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STATE OF NEW MEXICO COUNTY OF LEA FIFTH JUDICIAL DISTRICT

REPUBLICAN PARTY OF NEW MEXICO, DAVID GALLEGOS, TIMOTHY JENNINGS, DINAH VARGAS, MANUEL GONZALES, JR., BOBBY and DEE ANN KIMBRO, and PEARL GARCIA,

Plaintiffs,

v.

Cause No. D-506-CV-2022-00041

MAGGIE TOLOUSE OLIVER, in her official capacity as New Mexico Secretary of State, MICHELLE LUJAN GRISHAM, in her official capacity as Governor of New Mexico, HOWIE MORALES, in his official capacity as New Mexico Lieutenant Governor and President of the New Mexico Senate, MIMI STEWART, in her official capacity as President Pro Tempore of the New Mexico Senate, and JAVIER MARTINEZ, in his official capacity as Speaker of the New Mexico House of Representatives,

Defendants.

<u>PLAINTIFFS' OPPOSITION TO LEGISLATIVE DEFENDANTS'</u> MOTION TO EXCLUDE THE EXPERT TESTIMONY OF SEAN P. TRENDE

Plaintiffs the Republican Party of New Mexico and a bipartisan group of New Mexico voters (collectively, "Plaintiffs") hereby file this Opposition To Legislative Defendants' Motion To Exclude The Expert Testimony of Sean P. Trende. The expert report submitted by Mr. Trende¹ powerfully shows that SB1 is an egregious partisan gerrymander through *both* independent methods of proof that Justice Kagan endorsed in *Rucho v. Common Cause*, 139 S. Ct. 2484 (2019). *See Grisham v. Van*

 $^{^1}$ Yesterday, Mr. Trende successfully defended his dissertation for his doctoral program at The Ohio State University. Accordingly, Mr. Trende will obtain his doctoral degree and the title "Doctor" on December 17, 2023. Supplemental Declaration of Sean P. Trende, ¶ 27 (Sept. 26, 2023) ("Trende Suppl. Decl.")

Soelen, No.S-1-SC-39481 (N.M. Sept. 22, 2023) (hereinafter "Opinion"). First, Mr. Trende's report meticulously demonstrates that Senate Bill 1 ("SB1") is an extremely partisan, near-perfect gerrymander, through a variety of qualitative analyses. Second, using a sophisticated simulation analysis—the same one Mr. Trende used successfully as the lead expert in cases invalidating the gerrymandered maps in New York and Maryland this redistricting cycle—Mr. Trende generated 2,040,000 simulated, partisan-neutral maps, showing that SB1 is an extreme outlier.

Legislative Defendants have no serious answer to Mr. Trende's independently sufficient qualitative analyses, and while they file their Motion to try to exclude Mr. Trende's simulation analysis, their arguments fall flat. Legislative Defendants claim that Mr. Trende's simulations-based conclusions are inadmissible because he did not save the 2,040,000 maps that he generated for purposes of his expert report. But Mr. Trende's approach of reporting the overall partisan distribution of the simulations, and not analyzing specific maps within that extremely large sample, is just what those trained in his method recommend—including Dr. Kosuke Imai, who pioneered the simulation approach that Mr. Trende uses. In any event, Legislative Defendants' claimed indignation that they wanted the 2,040,000 maps that Mr. Trende used in assessing the overall partisan distribution of the simulations in his report is now irrelevant because when Mr. Trende re-ran his algorithm, this provided Legislative Defendants with a full set of 2,040,000 maps, and those maps' partisan distribution leads to the exact same conclusions as Mr. Trende reported from

his first 2,040,000 maps run. So even if Legislative Defendants for some reason wanted to analyze individual maps within the set—contrary to what Dr. Imai recommends would be appropriate—they now have 2,040,000 maps to look at, which maps lead to the exact same conclusion as Mr. Trende articulated in his expert report.

STATEMENT

A. Plaintiffs submitted the expert report of Mr. Trende to support their conclusion that the Legislature acted with partisan intent and effect in adopting SB1. See Expert Report Of Sean P. Trende (Aug. 11, 2013) ("Trende Rep."); Opinion at 48. Mr. Trende is a renowned redistricting expert, Trende Rep.1–4 & Ex.1, who was also appointed by the Virginia Supreme Court to serve as a special master for its redistricting process, id. at 3–4. Most recently, and most relevant, Mr. Trende has used his simulation analysis in partisan-gerrymandering cases in both New York and Maryland, id. at 8, and that simulation analysis served as key evidence supporting invalidation of those maps as partisan gerrymanders, Harkenrider v. Hochul, 197 N.E.3d 437, 443, 453 (N.Y. 2022); Szeliga v. Lamone, No.C-02-CV-21-001816, 2022 WL 2132194, at *1, *29–33, *46 (Anne Arundel Cnty. Md. Cir. Ct. Mar. 25, 2022).

In his report here, Mr. Trende prepared *both* a qualitative-evidence analysis and a sophisticated-social-science analysis, which both independently demonstrate that SB1 is an extreme partisan gerrymander. See Trende Rep.31–74.

The qualitative-evidence analysis in Mr. Trende's report shows that SB1 has impermissible partisan effects, just like the qualitative data discussed by Justice Kagan with respect to the challenged Maryland map in *Benisek v. Lamone*, 348 F. Supp. 3d 493, 497–507 (D. Md. 2018), vacated and remanded sub nom. Rucho, 139

S. Ct. 2484, the companion case to *Rucho*, 139 S. Ct. at 2518–19 (Kagan, J., dissenting). That is because, with SB1, the Legislature made substantial, partisan shifts of voters between districts, ultimately balancing the Democratic-Party composition in each of the State's three congressional districts to maximize the Democratic Party's chance of winning all three districts. Trende.Rep.42 (addressing all three districts, using two different metrics); *id.* at 33–35, 42–43. That is a nearperfect gerrymander because a partisan mapdrawer needs to "rob Peter to pay Paul" to make any one district more Democratic, *id.* at 41; *see also id.* at 14–15, so "the best-case scenario for a [Democratic] gerrymanderer" in New Mexico looking to sweep all congressional races "would be drawing three districts" with a Democratic-party composition of "54.29%," *id.* at 14 (relying upon 2020 presidential election vote data). Finally, Mr. Trende found that the voter-registration data, Opinion at 46–47, leads to the same conclusion, as SB1 shifted District 2 from being roughly even registration between Republicans and Democrats, to a 13% registration advantage for Democrats, Trende Rep.38.

The sophisticated social-science analysis in Mr. Trende's report independently confirms that SB1 is an egregious partisan gerrymander. Trende Rep.43–75. Mr. Trende randomly generated one million maps that "incorporate [New Mexico's] physical and political geography and meet its declared districting criteria, except for partisan gain." *Rucho*, 139 S. Ct. at 2518 (Kagan, J., dissenting); *see* Trende.Rep.43–44. Mr. Trende instructed the simulation to "respect county subdivisions," "keep districts modestly compact," and "keep populations equal." Trende Rep.44. These

simulations had an average "Gerrymandering Index" of roughly 1.3%. *Id.* at 46. SB1, on the other hand, had a Gerrymandering Index of 6.4%, over four standard deviations from the mean, thereby demonstrating that SB1 is an extreme gerrymander. *Id.* Mr. Trende then prepared an additional million simulated maps that only moved the precincts that the SB1 drafters also moved between districts. *Id.* at 54–60. These simulated maps had an average Gerrymandering Index of 0.62%, whereas SB1 had a Gerrymandering Index of 2.95%, over seven standard deviations from the mean. *Id.* at 54. Finally, Mr. Trende ran three sets of additional simulations of 10,000 maps to confirm his results in various respects. *Id.* at 61–77.

Following Plaintiffs' production of Mr. Trende's expert report, Mr. Trende also provided Legislative Defendants with the code he used to produce his simulated maps. Because Mr. Trende, per his "usual practice," did not save the individual simulated maps, Plaintiffs' counsel requested, at Legislative Defendants' counsel's insistence, that Mr. Trende re-run his simulations so that Plaintiffs could provide Legislative Defendants with the individual maps. Trende Suppl. Decl. ¶ 4. Mr. Trende did so, thus producing an additional 2,040,000 maps for Legislative Defendants. *Id.* Legislative Defendants then scheduled a follow-up deposition, at which Mr. Trende learned that the code that he ran did not replicate the first batch of maps, but rather created a new set of 2,040,000 simulations. *Id.* ¶ 6. Although Mr. Trende inserted a command in his code known as "setting a seed" that normally would ensure that "anyone running the simulations would produce the exact same maps," *id.* ¶ 4, that "setting a seed did not work for this particular application if a

computer utilized more than 1 processing core," id. ¶ 6. So, in re-running the code, Mr. Trende actually produced to Legislative Defendants a second set of 2,040,000 maps. This second set of maps, unsurprisingly, generate the same partisan distribution as the first set, thereby further confirming Mr. Trende's conclusions because now 4,080,000 confirm that SB1 is an extreme partisan outlier. Id. ¶ 7. This second set of maps also rendered immaterial the fact that Mr. Trende did not save his original set of maps, as the "second set of 2,040,000 simulations also demonstrates that the [challenged map] is an extreme outlier" and "only strengthens the case against the [challenged map]." Id. (emphasis omitted), see also id. ¶¶ 9–25. Indeed, as Mr. Trende explains, even "had [he] only considered the second set of simulations, none of [his] conclusions in this matter would have changed." Id. ¶ 7.

B. Legislative Defendants have now moved to exclude one portion of Mr. Trende's expert report, namely, Mr. Trende's sophisticated social-science analysis, based largely upon the fact that Mr. Trende did not save the first set of 2,040,000 maps. Legislative Defs.' Opposed Mot. To Exclude The Unreliable Simulation-Based Expert Test. Of Sean P. Trende (Sept. 20, 2023) ("Leg.Mot.").

LEGAL STANDARD

Rule 11-702 of the New Mexico Rules of Evidence provides that "[a] witness who is qualified as an expert by knowledge, skill, experience, training, or education may testify in the form of an opinion or otherwise if the expert's scientific, technical, or other specialized knowledge will help the trier of fact to understand the evidence or to determine a fact in issue." Rule 11-702 NMRA. Rule 11-702 establishes "three requirements" for expert testimony to be admissible: "(1) that the expert be qualified;

(2) that the testimony be of assistance to the trier of fact; and (3) that the expert's testimony be about scientific, technical, or other specialized knowledge with a reliable basis." *Acosta v. Shell W. Expl. & Prod., Inc.*, 2016-NMSC-012, ¶ 22, 370 P.3d 761 (quoting *State v. Downey*, 2008-NMSC-061, ¶ 25, 145 N.M. 232, 195 P.3d 1244). The second and third element are at issue here.

Rule 11-702's second prong, "that the testimony be of assistance to the trier of fact," id. (citation omitted), "goes primarily to relevance' as '[e]xpert testimony which does not relate to any issue in the case is not relevant and, ergo, non-helpful," id. ¶ 23 (brackets in original) (quoting $Daubert\ v$. $Merrell\ Dow\ Pharm.$, Inc., 509 U.S. 579, 591 (1993)). To be relevant, the expert testimony must be "sufficiently tied to the facts of the case," such that it "will aid the jury in resolving a factual dispute." Downey, 2008-NMSC-061, ¶ 30 (citation omitted). That is, the methodology underlying the expert's opinions must "fit[]" the "facts of the case" and any assumptions grounded in the evidentiary record, "thereby prov[ing] what it purports to prove." Id.

Rule 11-702's third element, "that the expert's testimony be about scientific, technical, or other specialized knowledge with a reliable basis," addresses the reliability of expert testimony, which is also key for the testimony to be helpful to the trier of fact. *Acosta*, 2016-NMSC-012, ¶ 22 (citation omitted); *see also Downey*, 2008-NMSC-061, ¶ 25. A number of factors may be "pertinent to the trial court's determination of whether [] scientific evidence is reliable," including: "(1) whether a theory or technique can be (and has been) tested; (2) whether the theory or technique

has been subjected to peer review and publication; (3) the known potential rate of error in using a particular scientific technique and the existence and maintenance of standards controlling the technique's operation; . . . (4) whether the theory or technique has been generally accepted in the particular scientific field"; and (5) "whether the scientific technique is based upon well-recognized scientific principle and whether it is capable of supporting opinions based upon reasonable probability rather than conjecture." *State v. Anderson*, 1994-NMSC-089, ¶ 15, 118 N.M. 284, 881 P.2d 29 (citation omitted).

ARGUMENT

I. Mr. Trende's Method Of Looking At The Overall Partisanship Distribution Of The Simulated Maps Is Reliable, And Legislative Defendants' Feigned Desire to Look At Individual Maps Within The Simulation Set Is Contrary To The State Of The Art And Irrelevant Because Mr. Trende Gave Them An Additional Set of 2,040,00 Maps That Generates The Exact Same Conclusions

A. Mr. Trende's sophisticated social-science analysis satisfies each of the three requirements for admissibility under Rule 11-702. First, Mr. Trende is an eminently "qualified" redistricting expert, Acosta, 2016-NMSC-012, ¶ 22 (citations omitted), including with respect to simulation analysis that comprises his sophisticated social-science analysis here, as he has presented such analysis in multiple prior partisangerrymandering cases, supra p.3. Second, Mr. Trende's simulation analysis provides "assistance to" this Court as "the trier of fact," Acosta, 2016-NMSC-012, ¶ 22 (citations omitted), as it gives an objective measure of the extreme partisanship of SB1, using the "extreme outlier approach" endorsed by Justice Kagan in her Rucho dissent, Rucho, 139 S. Ct. at 2518 (Kagan, J., dissenting), without inserting partisan

considerations within his simulations, unlike the simulations of Legislative Defendants' simulation expert, see Pls.' Opposed Mot. To Exclude Expert Report And Expert Test. Of Dr. Jowei Chen at 8–14 (Sept. 22, 2023). Finally, Mr. Trende's simulation analysis rests on "scientific, technical, or other specialized knowledge with a reliable basis," Acosta, 2016-NMSC-012, ¶ 22 (citations omitted), using a "broadly accepted 'package' in [the program] R called 'redist,' which generates a representative sample of districts," Trende Rep.17 (citing, among another authorities, Cory McCartan & Kosuke Imai, Sequential Monte Carlo for Sampling Balanced and Compact Redistricting Plans, Annals of Applied Stat., (forthcoming 2023)).

B. Legislative Defendants argue that Mr. Trende's expert testimony as to his simulation analysis fails the third element of Rule 11-702 because Mr. Trende did not produce the 2,040,000 individual simulated maps underlying his expert report and the code capable of fully replicating those maps.² Legislative Defendants' arguments both misunderstand the scientifically appropriate method for analyzing a large set of maps in a simulation analysis and are irrelevant, in any event, because Mr. Trende has produced yet a second set of 2,040,000 maps, which second set leads to the exact same conclusion as the set of maps that Mr. Trende analyzed in his expert report.

Legislative Defendants' objection to Mr. Trende's simulation-based opinions is contrary to the state-of-the-art redistricting simulation methodology that Mr. Trende

² While Legislative Defendants suggest that "there is ample evidence that Mr. Trende is not qualified to render opinions regarding simulation analysis," Leg.Mot.6, their purported bases for this criticism are wrong for the reasons discussed below, *see infra* Part II, and, in any event, they expressly confine their Motion to "the reliability of Mr. Trende's expert opinions because of his decision to destroy the facts and data underlying his opinions," Leg.Mot.6.

employed in preparing his sophisticated social-science analysis. As Mr. Trende explains in his Supplemental Declaration, analysts "who use [the] simulation approach on which [Mr. Trende] relied" do not ordinarily "examine individual maps when performing the analysis," as the relevant data point is the "overall distribution." Trende Suppl. Decl. ¶¶ 2–3. The creator of Mr. Trende's simulation approach, Dr. Imai, has explained this very point, noting that, "[i]n order to use the simulation for evaluation," "one should never look at a single or a particular map[]," but rather must "look at the distribution of plans." Id. ¶ 3 (citation omitted) (emphasis added).³ Mr. Trende, consistent with this state of the art and thus his "usual practice," "did not save the individual maps." Trende Suppl. Decl. ¶ 2. Accordingly, Legislative Defendants' unusual request to see the individual maps that Mr. Trende generated is wholly unnecessary (and, indeed, nonsensical) for the type of analysis that Mr. Trende performed and, moreover, Legislative Defendants do not even try to explain what they would have done with the individual 2,040,000 maps that would have been permissible under Mr. Trende's state-of-the-art simulation method. See generally Leg.Mot.3-5, 6-7.

In any event, even if this Court were to conclude that it was error for Mr. Trende not to save and then produce his first set of 2,040,000 maps, Mr. Trende fully redressed Legislative Defendants' manufactured concerns by producing a second set of 2,040,000 maps to Legislative Defendants. Trende Suppl. Decl. ¶ 4. This

³ Legislative Defendants themselves rely favorably upon Dr. Imai's scholarship in their Response To Plaintiffs' Motion To Exclude The Expert Report And Expert Testimony Of Dr. Jowei Chen (Sept. 25, 2023).

second set of 2,040,000 maps has a partisanship distribution that is substantially the same to the simulated maps underlying Mr. Trende's simulation-based opinions. *Id.* ¶¶ 7, 9-25. For example, for the one-million simulated maps that used the presidential vote share to measure partisanship, Mr. Trende's expert report noted that 0.11% of the simulated maps had more extreme gerrymandering than the challenged map. Id. ¶ 10. That number was the same in the new production. Id. For the one-million simulated maps that looked only at those precincts that were swapped between the challenged map and the 2012-2020 plan, Mr. Trende's expert report noted that the gerrymandering index was "over seven" standard deviations from the mean. Id. ¶ 11. That number was 7.170 for the new production. Id.; see also id. ¶¶ 15–24. A review of the gerrymandering index plot figures produced by the original simulated maps as compared to the new batch similarly shows no material difference in the data generated by these two sets of maps and "leads to precisely the same conclusions as [Mr. Trende] la[id] out in [his expert] report," id. ¶¶ 15–25, namely, that the challenged map is an "outlier" and was likely drawn in "heav[y]" reliance on political considerations, id. ¶ 1; see id. ¶ 7 ("[H]ad I only considered the second set of simulations, none of my conclusions in this matter would have changed[.]"). The second set of maps, if such a set were necessary, thus provides a more than sufficient basis for holding that the simulations underlying Mr. Trende's expert report were "reliable." Acosta, 2016-NMSC-012, ¶ 22 (citation omitted); see also Downey, 2008-NMSC-061, ¶ 25; Rule 11-702 NMRA.

C. For the same reasons, Legislative Defendants' request that this Court exclude Mr. Trende's simulation-based opinions as a punishment for Mr. Trende's decision not to save his initial run of 2,040,000 maps is a nonstarter.

As a threshold matter, Legislative Defendants do not claim that Mr. Trende failed to save the individual maps in order to intentionally keep them from Legislative Defendants; rather, their complaint is merely with Mr. Trende's "usual practice." Trende Suppl. Decl. ¶ 2; see State v. Chouinard, 1981-NMSC-096, ¶ 16, 96 N.M. 658, 634 P.2d 680; Rest. Mgmt. Co. v. Kidde-Fenwal, Inc., 1999-NMCA-101, ¶ 13, 127 N.M. 708, 986 P.2d 504. Again, as Mr. Trende has explained, it is his "usual practice" not to save individual maps because "neither [he] nor others who use" the simulation approach "examine individual maps when performing the analysis." Trende Suppl. Decl. ¶ 2. Thus, while Legislative Defendants recognize that evidence exclusion is generally appropriate only where the opposing party has acted "deliberate[ly] or in bad faith," Leg.Mot.8, they do not contend that Mr. Trende acted deliberately here, Chouinard, 1981-NMSC-096, ¶¶ 14–16. Legislative Defendants' requested exclusion remedy is thus inappropriate. See id.; Rest. Mgmt. Co., 1999-NMCA-101, ¶ 14.

In any event, Legislative Defendants cannot credibly claim that the original batch of simulated maps is "material" to any arguments that they would want to make in this case, including because Mr. Trende produced a second set of maps that does just what his first set of maps do and shows the same results. *See Chouinard*, 1981-NMSC-096, ¶ 16. As Mr. Trende has explained, the individual simulated maps

are not themselves relevant to the simulation analysis under the state-of-art simulation methodology; rather, what matters for purposes of this analysis is the "overall distribution." Trende Suppl. Decl. ¶ 3. But to the extent the individual maps were material to whether Mr. Trende properly performed the analysis in his expert report, Mr. Trende has since provided Legislative Defendants with a set of 2,040,000 simulated maps that lead to all of the same conclusions put forth in Mr. Trende's expert report. *Id.* ¶ 7. Legislative Defendants do not contend that Mr. Trende's opinions would be any different had he formed them on the basis of the new production rather than on the original batch of simulated maps—nor could they, given that Mr. Trende's Supplemental Declaration makes clear that the second set of maps "only strengthens the case against" Senate Bill 1. *Id.* (emphasis omitted).

For similar reasons, Legislative Defendants cannot claim that they have suffered any prejudice from Mr. Trende's standard practice of not saving the individual maps underlying his expert reports. Mr. Trende's standard practice has no "effect" on Legislative Defendants' "ability to defend against Plaintiffs' claims," Rest. Mgmt. Co., 1999-NMCA-101, ¶ 15, as, again, only the "overall distribution"—not the individuals maps—is relevant to assessing whether Senate Bill 1 is an outlier, Trende Suppl. Decl. ¶¶ 2–3, and Legislative Defendants have, in any event, a second set of substantially similar maps in their possession, id. ¶¶ 7, 9–14. Thus, while Legislative Defendants assert that Mr. Trende's usual practice is "profoundly prejudicial," Leg.Mot.9, they do not explain how they are prejudiced, Trende Suppl. Decl. ¶¶ 7–14. Further, and contrary to Legislative Defendants' claim, Mr. Trende

has not "admitted" that Legislative Defendants "would be prejudiced" absent access to the original batch of maps. Leg.Mot.9. Rather, Mr. Trende has confirmed that the new production of simulated maps produces substantially the same results, Trende Suppl. Decl. ¶¶ 9–24, and that his expert opinions would be the same even if he had "only considered the second set of simulations," id. ¶ 7; see Rest. Mgmt. Co., 1999-NMCA-101, ¶ 15; Choinard, 1981-NMSC-096, ¶ 16.

II. Legislative Defendants' Remaining Quibbles With Mr. Trende's Report Provide No Basis For Exclusion

Legislative Defendants also suggest that Mr. Trende's simulation-based opinions are unreliable based upon minor inconsistencies between his expert report and his deposition testimony, as well as a mix-up in the computer scripts that Plaintiffs produced to Legislative Defendants. Leg.Mot.3–4. This too is wrong.

Legislative Defendants complain that Mr. Trende's expert report states that he performed his simulations "at home on a Dell Alienware desktop with an i9 processor," whereas Mr. Trende later confirmed that he "performed his simulations on a 16-core AMD processor." *Id.* at 3 (citations omitted). Legislative Defendants do not explain how the type of computer that Mr. Trende used to create his simulations is relevant to whether those simulations are reliable. *See id.* at 3, 6–7. And as Mr. Trende himself explained to Legislative Defendants, this minor inconsistency was "probably a leftover from having done it on a laptop once and forgetting that [he] didn't get an Intel chip on this, [he] got an AMD chip," which is "functionally equivalent" to "the i9." Leg.Mot. Exh. C.15 (Deposition of Sean P. Trende, Vol.2, 153:21–24).

As to Mr. Trende's purported "inconsisten[cies] regarding the number of simulations [he] performed," Legislative Defendants point to only one such inconsistency in Mr. Trende's expert report, which is an obvious "typo." Leg.Mot.3 (citation omitted). Specifically, on page 47 of his expert report, Mr. Trende states that he created "50,000 simulated maps," despite having created one million such maps. Trende.Rep.47. This is a clear typo, and Legislative Defendants' decision to waste this Court's time by bringing it up in a motion is unseemly gamesmanship.

Finally, Legislative Defendants take issue with the fact that Mr. Trende initially produced computer code that was only capable of generating sets of 100,000 maps, rather than 1,000,000 maps. Leg.Mot.3–4. But as Mr. Trende has explained, this is a "trivial" issue "that any beginning coder could address in a matter of seconds," and he did in fact "produce[] code with the number of simulations set to 1,000,000 shortly after the deposition." Trende Suppl. Decl.2 n.2. In any event, this complaint is ultimately irrelevant as, again, Legislative Defendants now have access to over two-million additional maps produced with Mr. Trende's code.

CONCLUSION

This Court should deny Legislative Defendants' Motion To Exclude The Expert Testimony of Sean P. Trende.

Dated: September 26, 2023

MISHA TSEYTLIN*
MOLLY S. DIRAGO*
KEVIN M. LEROY*
TROUTMAN PEPPER
HAMILTON SANDERS LLP
227 W. Monroe Street
Suite 3900
Chicago, IL 60606
(608) 999-1240 (MT)
(312) 759-1926 (MD)
(312) 759-1938 (KL)
(312) 759-1939 (fax)
misha.tseytlin@troutman.com
molly.dirago@troutman.com
kevin.leroy@troutman.com

Attorneys for Plaintiffs Manuel Gonzales, Jr., Dinah Vargas, David Gallegos, and Timothy Jennings

Respectfully Submitted,

HARRISON & HART, LLC

/s/Carter B. Harrison, IV
CARTER B. HARRISON, IV
924 Park Avenue SW, Suite E
Albuquerque, New Mexico 87102
(505) 312-4245
(505) 341-9340 (fax)
carter@harrisonhartlaw.com

Attorneys for Plaintiffs Republican Party Of New Mexico, David Gallegos, Dinah Vargas, Bobby and Dee Ann Kimbro, and Pearl Garcia

^{*}Admitted Pro Hac Vice

CERTIFICATE OF SERVICE

I hereby certify that a true and complete copy of the foregoing will be served on all counsel via the e-filing system.

Dated: September 26, 2023

/s/Carter B. Harrison, IV CARTER B. HARRISON, IV 924 Park Avenue SW, Suite E Albuquerque, New Mexico 87102 (505) 312-4245 (505) 341-9340 (fax) carter@harrisonhartlaw.com STATE OF NEW MEXICO COUNTY OF LEA FIFTH JUDICIAL DISTRICT

REPUBLICAN PARTY OF NEW MEXICO, DAVID GALLEGOS, TIMOTHY JENNINGS, DINAH VARGAS, MANUEL GONZALES, JR., BOBBY and DEE ANN KIMBRO, and PEARL GARCIA,

Plaintiffs,

v.

Cause No. D-506-CV-2022-00041

MAGGIE TOLOUSE OLIVER, in her official capacity as New Mexico Secretary of State, MICHELLE LUJAN GRISHAM, in her official capacity as Governor of New Mexico, HOWIE MORALES, in his official capacity as New Mexico Lieutenant Governor and President of the New Mexico Senate, MIMI STEWART, in her official capacity as President Pro Tempore of the New Mexico Senate, and JAVIER MARTINEZ, in his official capacity as Speaker of the New Mexico House of Representatives,

Defendants.

SUPPLEMENTAL DECLARATION OF SEAN P. TRENDE

- 1. In the course of preparing my initial expert report in this matter, I utilized a desktop computer to generate millions of simulated maps. I was able to look at the overall partisanship of these maps and to compare them to the partisanship of the Enacted Map. On this basis, following the approach I and others have used in similar matters, I was able to determine that Enacted Map was an outlier that would be extremely unlikely to have been produced as the result of a drawing process that did not rely heavily on partisan considerations.
- 2. On August 11, 2023, I produced to Defendants my expert report and the computer code I used to generate the analyses presented in my report. As is my usual practice, I did not save the individual maps. The reason is that neither I nor others who use simulation approach on which I relied typically examine individual maps when performing the analysis. While I may run a small sample set early in the process to make sure that the maps are behaving as expected—that the

underlying shapefile doesn't have missing precinct data or something of that nature—I do not examine maps in the full sample. Doing so would make little sense in the context of how the technique is supposed to work.

- 3. In fact, Dr. Kosuke Imai, who developed the simulation approach on which I relied, has been emphatic that one should not examine maps individually, but rather should pay attention to the overall distribution. In his previous sworn trial testimony, he stated: "So one thing that's very important, and I think is incorrect in the Dr. Voss report, is that *one should never look at a single or a particular maps simulated plans* [sic], right? In order to use the simulation for evaluation, you *have to look at the distribution of plans*. So -- in not, like, a one specific plan, but all 10,000 of them." Trial Tr. 51:1-51:7, *Graham v. Adams*, No. 22-CI-47(Ky. Cir. Ct. Apr. 5, 2022) (testimony of Kosuke Imai) (emphasis added), attached as **Exhibit A**.
- 4. I was asked by counsel to re-run my simulations and make the results available to counsel for Defendants. I did so, producing precinct assignment files¹ for 2,040,000 maps, which I believed to be the same as the 2,040,000 maps created by the initial simulations. This is because, due to a command I insert in my code known as "setting a seed," I believed that anyone running the simulations would produce the exact same maps.² This is the typical way coders ensure that

¹ A precinct assignment file gives a number that matches every precinct in an area with a district on a map. It does this for every map that is generated. This is the same way that Dr. Chen produces his maps.

² In my deposition of September 6, 2023, counsel for Defendants also raised concerns that the code I produced only created 100,000 maps instead of 1,000,000. It does appear that, at some point after my initial analysis was completed but prior to production, the number of maps produced was reset to 100,000. While I can't recall exactly why I did this, I likely did it during my pre-production review. Before production I will typically restart the computer and then re-run the code to ensure it will still run cleanly after everything is wiped from the computer's memory (e.g., in the environment in which a different analyst would run it). I likely reduced the number of simulations so that this process would be completed faster and forgot to change it back. Regardless, this is a trivial matter that any beginning coder could address in a matter of seconds. I also produced code with the number of simulations set to 1,000,000 shortly after the deposition.

outcomes are fully reproduceable, and I have never encountered an application where setting a seed does not result in fully reproduceable outcomes before.

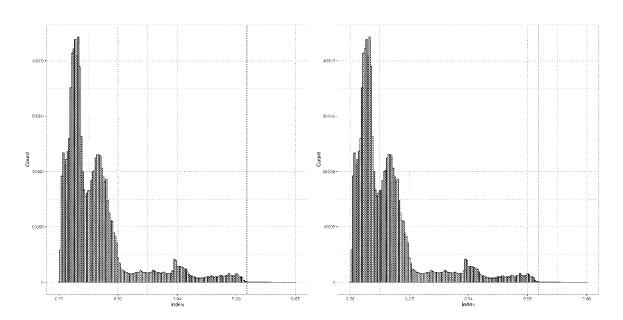
- 5. Counsel for Defendants requested an additional deposition, to take place within 48 hours. I complied and was prepared to answer to the best of my abilities any questions about the process used to generate the maps, the output from that process, or any other questions counsel may have had.
- 6. Over the course of this deposition, counsel produced documentation that suggested that setting a seed did not work for this particular application if a computer utilized more than 1 processing core. The documentation he showed me suggested that the output might not be "fully" reproduceable, but the documentation did not explain what "fully" means in that context. Deposition of Sean P. Trende, Vol.2 at 158:23, 162:22 (discussing deposition exhibits 25 and 26), attached as **Exhibit B**.
- 7. Examining the output from the produced maps (that is, the second set of 2 million maps produced to counsel) leads to three important conclusions.
 - a. The simulation outcomes are similar, and in the case of larger simulations, nearly identical, to the ones from the initial report. This is unsurprising. The entire purpose of the simulation exercise is to explore the set of maps that would be produced by neutral mapmakers under a given set of constraints. Just as public opinion polls become more precise as the number of individuals selected increases, so too does the estimate of the distribution of politics-neutral plans become more precise as the number of individuals selected increases. Thus, a reproduced set of 1,000,000 plans should vary very little from run to run.

- b. The conclusions I would draw from the set of simulations produced to counsel are the same as those found in the Trende Report. Because the second set of 2,040,000 simulations also demonstrates that the Enacted Map is an extreme outlier, examining these maps in the way that Dr. Imai described above would not lead an expert to different conclusions about the nature of the Enacted Map than those the Trende Report suggests. In fact, had I only considered the second set of simulations, none of my conclusions in this matter would have changed.
- c. This only strengthens the case against the Enacted Map. The fact that a second run of 2,040,000 maps leads to the same conclusions as the first run of 2,040,000 maps only demonstrates how robust the initial findings are.
- 8. One way to demonstrate this is to examine the results for the simulation set analyzed in the report and compare them to the results from the simulation set produced to Defendants' counsel.
- 9. The following table reports the results of simulations that were reported in the initial expert report. It also reports the results from the simulations that were produced to counsel.

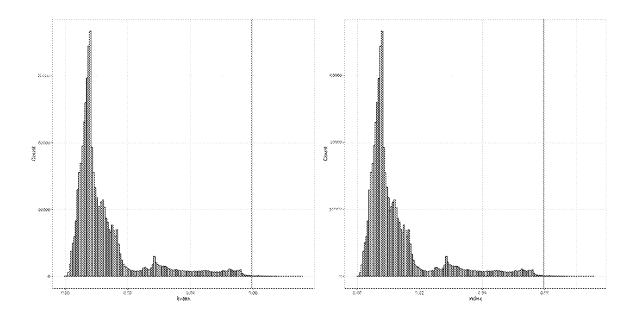
	St. Dev. of GI From Mean Percent of Plans in Eusemble More Extreme				
Scenario	Report	Production	Report	Production	
1 million simulations POTUS	over 4	4,107	0.11%	0.11%	
1 million simulations Truncated	over 7	7.170	0.00%	0.00%	
10,000 simulations Registration	over 3	3.227	1.92%	2.15%	
0,000 similations Registration Truncated	Not Reported	4.160	1.20%	0.00%	
10,000 simulations Citizens' Map	6.67	6.600	0.00%	0.00%	

- 10. As you can see, there is very little difference, if any, between the larger sets, and only slight differences between the smaller sets. For the 1 million simulations using presidential vote share as the measure of partisanship, the report states that the gerrymandering index produced was "over four" standard deviations from the mean. In the production set, the gerrymandering index was 4.107 standard deviations from the mean. In the report set, 0.11% of the maps in the ensemble had more extreme gerrymandering indices than the Enacted Plan; in the production set, the number is the same.
- Likewise, for the 1 million simulations that examined only those precincts that were swapped between the Enacted Plan and the 2012-2020 plan, the report states that the gerrymandering index produced was "over seven" standard deviations from the mean. In the production set, the gerrymandering index was 7.170 standard deviations from the mean. In the report set, none of the maps in the ensemble had more extreme gerrymandering indices than the Enacted Plan; in the production set, the number is the same.
- 12. The sets of 10,000 simulations show slightly more variation; that is to be expected. (That is why a set of say, 1,000 maps, would not be as reliable as a set of 1,000,000 maps.) But none of the bottom lines change. For the 10,000 simulations examining party registration, the initial report states that the gerrymandering index for the Enacted Map was "over 3" standard deviations from the mean of the ensembles; in the production set, the gerrymandering index of the Enacted Map is 3.227 standard deviations from the mean. The report further concludes that 1.92% of the maps were more extreme than the Enacted Map; the production set concludes that 2.15% of the maps were more extreme than the Enacted Map.

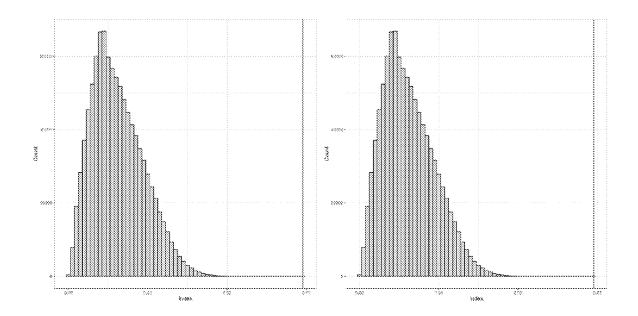
- 13. Likewise, in the ensemble of truncated maps that were run using registration as a measure of partisan identification, 1.2% of the maps had a gerrymandering index that was more extreme than the Enacted Plan in the report, while none were more extreme in the produced maps.
- 14. Finally, looking only at the precincts that were swapped between Citizen's Plan H and the Enacted Plan, the Enacted Plan had a gerrymandering index that was 6.67 standard deviations from the mean in the report and 6.6 standard deviations from the mean in the produced maps. Neither set had any maps with gerrymandering indices more extreme than the Enacted Map.
- 15. We can also compare the figures. For simplicity's sake, I will only provide the gerrymandering index plots, since they are effectively summary figures for the dotplots and boxplots.
- 16. We start with the gerrymandering index plots for the report set and the production set. The report set is on the left, while the production set is on the right.



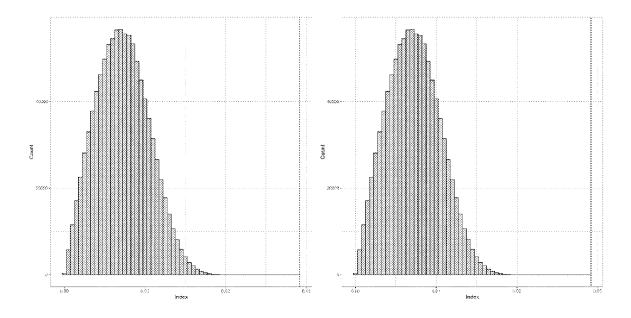
17. Second, we look at the sets using the partisan index as the measure of partisanship.



- 18. In both instances, there are no differences that would be relevant to the research question I was undertaking.
- 19. Next, I look at the truncated maps—that is, the maps that examined only the precincts that changed between the 2012-2020 map and the Enacted Map. Once again, the image from the report is on the left, while the image from the produced simulations is on the right. There may be slight differences, but they are hard to detect.

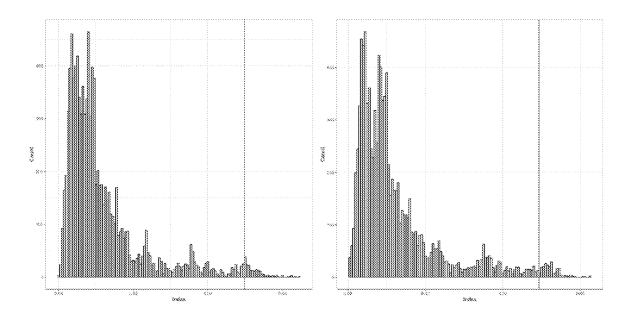


20. These are the images using the partisan index to measure partisanship.

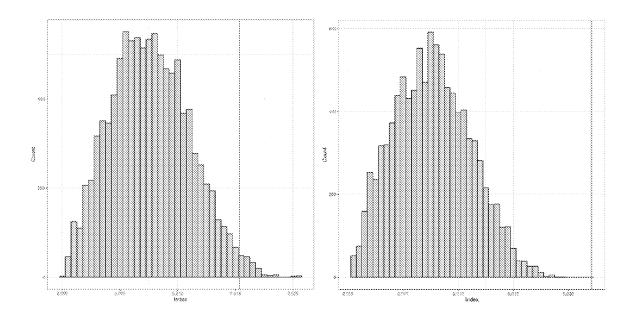


21. The remaining simulation sets involved just 10,000 simulations. Unsurprisingly, the differences, while still modest, are more pronounced. Here, we compare the set of 10,000 simulations run on the full map, using registration as the metric for partisanship. The image from

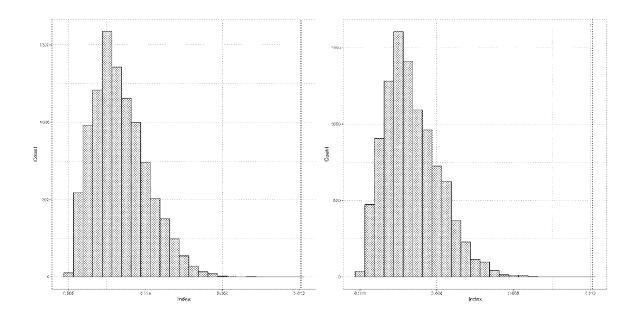
the report is on the left while the image from the production set is on the right. You can see the same "peaks and valleys" beginning to emerge between the sets, demonstrating the stability of the findings between the sets. If left to run for 1,000,000 simulations, those peaks and valleys would become even more stable.



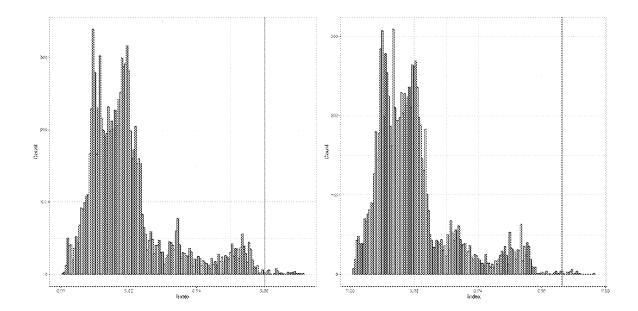
22. Here, we examine the precincts that were exchanged between the 2012-2020 map and the Enacted Map, using registration as the measure of partisanship.



23. Third, we examine the precincts that were swapped between Plan H and the Enacted Plan.



24. Finally, we examine the simulations that were run keeping Indian Reservations intact.



- 25. Critically, but not surprisingly, an analysis of the production set that Defendants' counsel received leads to precisely the same conclusions as I lay out in my report. In the simulations with an exceptionally large number of simulations, any potential differences between the production set and the set utilized for the report are insignificant.
- As stated earlier, the fact that a second run of the simulations produces substantially similar outputs is testimony to the robustness of the simulations and the reliability of my conclusion. Over the course of 4,040,000 maps, with multiple sets of constraints applied, *only a handful* are more extreme than the Enacted Plan.
- 27. Finally, on September 25, 2023, I successfully defended my dissertation for my doctoral program at The Ohio State University. Accordingly, I will now obtain my doctoral degree and the title "Doctor" on December 17, 2023.

I declare under penalty of perjury under the laws of the State of New Mexico that the foregoing is true and correct. See N.M. R. Civ. P. Dist. Ct.1-011(B).

Dated: September 26, 2023

/s/ Sean P. Trende SEAN P. TRENDE

CERTIFICATE OF SERVICE

I hereby certify that a true and complete copy of the foregoing will be served on all counsel via the e-filing system.

Dated: September 26, 2023

/s/Carter B. Harrison, IV CARTER B. HARRISON, IV 924 Park Avenue SW, Suite E Albuquerque, New Mexico 87102 (505) 312-4245 (505) 341-9340 (fax) carter@harrisonhartlaw.com

EXHIBIT A



CIVIL ACTION NO. 22-CI-47 GRAHAM, ET AL.

V.

ADAMS, ET AL.,

TRIAL DAY 1

DATE:

April 05, 2022



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1	FRANKLIN CIRCUIT COURT
2	CIVIL ACTION NO.: 22-CI-47
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5	GRAHAM, ET AL.,
6	Plaintiffs
7	
8	V.
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10	ADAMS, ET AL.,
11	Defendants
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16	TRIAL DAY 1
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1 **PROCEEDINGS** JUDGE WINGATE: -- every time you turn around. 2 3 I don't remember doing that when I was a youngster. But, anyway, they like to get up really early. All 4 5 right. Now, I thought that we would probably -- you 6 don't need to do openings or anything. I thought you'd just go right into your witnesses, and we'd go 7 from there. Is that all right, Michael? 8 (Inaudible), Your Honor. 9 MR. ABATE: JUDGE WINGATE: Okay. I'll probably call you 10 11 all your first names because I know you. Okay? All 12 right. You can call your first witness. MR. MADDOX: Your Honor? 13 14 JUDGE WINGATE: Yes. MR. MADDOX: Just two quick housekeeping 15 16 measures. 17 JUDGE WINGATE: Sure. MR. MADDOX: So, the first one is, yesterday, 18 19 we filed a motion -- or response to the motion to 20 dismiss. We don't intend to argue that today, but I 2.1 wanted you to know that it is in the record. 22 JUDGE WINGATE: Okay. MR. MADDOX: At the end of the proceeding this 23 24 week, we would expect to ask for judgment on our 25 cross claim and counter claim. The second one is,



1	we have prepared a written statement of the joint
2	stipulation that was reached at the last oral
3	hearing.
4	JUDGE WINGATE: Okay.
5	MR. MADDOX: So, we would, you know, tender
6	that to the Court. And we have included in a
7	binder, for everyone's use, certain printed
8	materials that we think would be helpful that come
9	from either the LRC website or the Secretary of
10	State's website, both of which have been stipulated
11	as admissible by all parties. I think that's right.
12	Casey?
13	MS. HINKLE: No objection, right.
14	JUDGE WINGATE: What's your name again?
15	MS. HINKLE: Casey Hinkle, Your Honor.
16	JUDGE WINGATE: I'm sorry?
17	MS. HINKLE: Casey Hinkle.
18	JUDGE WINGATE: Casey. Okay. So, you had
19	these stipulations written out?
20	MS. HINKLE: I believe
21	MR. MADDOX: Yes, Your Honor.
22	JUDGE WINGATE: Okay. Let's see what they look
23	like.
24	MR. MADDOX: And Alex is going to give it to
25	the Court. We've not yet filed it. We're tendering



1	it here in open court. He's also giving you a
2	notebook that we've already provided to the
3	plaintiff's Counsel, that includes relevant
4	materials from the websites.
5	JUDGE WINGATE: Okay. Can I just do the joint
6	stipulation as Exhibit 1? Would that be all right
7	for you-all?
8	MR. MADDOX: That would be that be great.
9	JUDGE WINGATE: Okay. Unless you've got your
10	
11	MS. HINKLE: I did pre-mark a couple things,
12	but that's okay.
13	JUDGE WINGATE: You did? Okay. Well, listen.
14	If you've pre-marked stuff, let's go with your
15	pre-marked stuff, and then we'll do this at break
16	MS. HINKLE: Okay.
17	JUDGE WINGATE: in between your case, if
18	that works. All right. All right. You may begin.
19	MS. HINKLE: Your Honor, similarly, I wanted to
20	bring your attention. There's a couple of binders
21	that we put over there as well.
22	JUDGE WINGATE: Is that your binder?
23	MS. HINKLE: It's our binders. Yes. And we
24	there's an empty binder for the witness's use, to
25	keep things organized as he may receive a lot of



1 paper. JUDGE WINGATE: Well, just go back and forth 2 and take it to them. How's that? 3 That's fine. MS. HINKLE: 4 5 JUDGE WINGATE: Or you can put your witness 6 binders on up the witness stand, if you want to go ahead and do that. 7 MS. HINKLE: Okay. You'd like the witness to 8 be seated here? 9 10 JUDGE WINGATE: Yeah. That's where they're 11 going to be. 12 MS. HINKLE: Okay. Great. Great. JUDGE WINGATE: All right? So, however 13 14 you-all want to do it. MS. HINKLE: All right. Well, the --15 16 JUDGE WINGATE: I designed this courtroom, so if you all hate it, it was designed by me. Okay? 17 18 MR. MADDOX: We love it, Your Honor. 19 JUDGE WINGATE: And if you know it , well, it's 20 sort of like if you're in a jury trial where you 2.1 say, "Voir dire, "Voir dare," you know? And I said, 22 how do you say that, to one lawyer, one time. 23 he said, however you say it is correct, Judge. 24 anyway, it's designed just like the historical 25 court, and that's -- you know, that's why. All



1	right. You ready to go?
2	MS. HINKLE: Yes, Your Honor.
3	JUDGE WINGATE: All right.
4	MS. HINKLE: The plaintiffs call Dr. Kosuke
5	Imai.
6	JUDGE WINGATE: Okay.
7	MS. HINKLE: Yes.
8	JUDGE WINGATE: Okay. Please raise your right
9	hand. Okay? Do you swear or affirm the testimony
10	you're about to give in this court today is the
11	truth and nothing but the truth?
12	THE WITNESS: Yes.
13	JUDGE WINGATE: Okay. Now, how do you say your
14	name again?
15	THE WITNESS: Kosuke Imai.
16	JUDGE WINGATE: Okay. Thank you.
17	THE WITNESS: Thank you.
18	MS. HINKLE: And Your Honor, would you prefer
19	the witness sit, so that you can see his face, or he
20	can face the audience? Okay.
21	JUDGE WINGATE: Nope. I'm seeing him on I'm
22	watching him on my monitor.
23	MS. HINKLE: I see. Thank you.
24	DIRECT EXAMINATION
25	BY MS. HINKLE:



1	Q	Good morning, Dr. Imai.
2	A	Good morning.
3	Q	Would you please state your name for the
4	record?	
5	A	Kosuke Imai.
6	Q	And where do you live, Dr. Imai?
7	A	I live in Newton, Massachusetts.
8	Q	Okay. And what is your current occupation?
9	A	I'm a professor in the department of
10	governmen	t and also in the department of statistics, at
11	Harvard U	Iniversity.
12		MR. MADDOX: Your Honor, may I interrupt? I'm
13	havin	g a hard time hearing him. Would you object if
14	I mov	red over into the jury box so that I could
15		JUDGE WINGATE: Nope. Any of you-all need to
16	move	over to the jury box? That's fine.
17		MS. HINKLE: And will you let us know if the
18	court	room microphone's not picking him up clearly?
19		JUDGE WINGATE: Yeah. I don't he you
20	just	have to sort of speak in the microphone.
21		THE WITNESS: Oh, okay.
22		JUDGE WINGATE: The microphone is on, even
23	thoug	h the lights are not there.
24		MS. HINKLE: Okay. Thanks.
25		THE WITNESS: Okay.

1	JUDGE WINGATE: All right.
2	MS. HINKLE: Yeah.
3	JUDGE WINGATE: Who's 18? Which one is 18,
4	L-E-X 18? She's telling me that you're blocking the
5	camera. There we go. Is that okay?
6	CLERK: Can Dr. Imai speak really fast?
7	MS. HINKLE: Just to test the microphone.
8	JUDGE WINGATE: Can you say something, so we
9	can see if the cameras are okay?
10	THE WITNESS: I'm Kosuke Imai.
11	JUDGE WINGATE: That's perfect.
12	COURT REPORTER: Yes.
13	JUDGE WINGATE: All right. Thank you. That's
14	all. That's all you need to do.
15	MS. HINKLE: Okay.
16	THE WITNESS: Okay.
17	BY MS. HINKLE:
18	Q So, Dr. Imai, you flew in from Newton,
19	Massachusetts, you said. And you explained that you're
20	a professor at Harvard; is that right?
21	A Uh-huh. That's right.
22	Q I'm going to ask you a little bit more about
23	your academic background and qualifications, because
24	you've been retained as an expert witness for the
25	plaintiffs in this matter, right?



1	A That's correct.
2	MS. HINKLE: So, Your Honor, if I could
3	approach the witness?
4	JUDGE WINGATE: Sure.
5	MS. HINKLE: I have Dr. Imai's CV, which he may
6	want to reference during his testimony. And
7	JUDGE WINGATE: You got one for me?
8	MS. HINKLE: Your Honor, this is a copy for
9	you.
10	JUDGE WINGATE: Okay. Thank you.
11	MR. MADDOX: Alex.
12	MS. HINKLE: And Morgan.
13	MR. MADDOX: The plaintiff's the plaintiff's
14	book.
15	MS. HINKLE: You have to use the binder.
16	THE WITNESS: Oh, okay.
17	MS. HINKLE: You can use this binder, if you'd
18	like, to keep things organized that way.
19	THE WITNESS: All right. Okay. Sure. Yeah.
20	BY MS. HINKLE:
21	Q And we've marked Dr. Imai's CV as Exhibit 1
22	for identification at this point. Dr. Imai, is this an
23	accurate and up-to-date CV that you prepared?
24	A Yes. I believe so.
25	Q And does this reflect your academic training



1	and certain other of your experience?
2	A Yes. I do. It does.
3	MS. HINKLE: Okay. We would move to introduce
4	this as Exhibit 1.
5	THE WITNESS: Okay.
6	JUDGE WINGATE: You have any objection to his
7	CV?
8	MR. MADDOX: No objection, Your Honor.
9	JUDGE WINGATE: Okay. So, ordered.
10	(PLAINTIFF'S EXHIBIT 1 ADMITTED INTO
11	EVIDENCE)
12	BY MS. HINKLE:
13	Q And Dr. Imai, can you start by telling us
14	where you did your undergraduate studies?
15	A I did my undergraduate studies at the
16	University of Tokyo.
17	Q And what did you study there? What subjects?
18	A Major is, you know, liberal arts, which
19	combines variety of subjects from mathematics to social
20	sciences of your choice, basically.
21	Q Okay. And did you continue your studies after
22	that degree?
23	A That's correct.
24	Q Where did you study next?
25	A I did the graduate degree at Harvard.



1	Q And what was the subject matter of that
2	degree?
3	A So I did study both statistics and political
4	science. In statistics I received master's degree in
5	statistics and then PhD, subsequently, in political
6	science.
7	Q And did you have a concentration within those
8	fields of study?
9	A Yes. I mean, you know, statistics in general
10	and application of statistics to social science
11	problems, questions. Sometimes they call political
12	methodology. It's a statistical methods for political
13	science.
14	Q And what drew your interest in those topics?
15	Why did you choose that as your concentration?
16	A Oh, yeah. That's a good question. I was
17	always interested in mathematics, computer science, you
18	know, from young age, and but I was also interested
19	in social problems, politics, economics, sociology. So
20	this is a way to combine my interest in mathematics and
21	data with the substantive interest in societal problems.
22	Q Okay. Your CV lists various honors and
23	awards, I think, on pages 2 and 3. One is a recognition
24	by Clarivate Analytics as a highly-cited researcher. Can

25

you explain to us what that means?

A Yeah. So this organization is a premier
organization that keeps track citation counts of
academic journals. And I was named for you know, one
of the few people who had, I think, produced multiple
papers of high citation impact. So that's I've
received that honor for last four years, since such
honor existed.

- Q Okay. And you received a PhD degree from Harvard, right?
 - A That's correct.

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- Q And that was in 2002?
- A Uh-huh -- 2003. Yes.
- Q Okay. What did you do, after you received that degree?
- A Yeah. So I started teaching at Princeton
 University. First as an instructor, and then assistant
 professor, associate professor, and then eventually
 promoted to professor -- full professor.

Q And what classes did you teach?

A Yes. So I teach, you know, statistics from undergraduate level to graduate level, mostly targeting students who are majoring in political science, public policy, you know, some engineering students who are interested in social problems as well. So those are the subjects that I teach.

1	Q Okay. I saw a reference on your CV to
2	Princeton's program in, "Statistics and Machine
3	Learning. " Can you describe for us what that is?
4	A Yeah. Sure. As you know, like last ten
5	years, many universities have invested data science
6	programs. So, Princeton was also, you know, no
7	exception. They wanted to build the program that
8	combines a variety of disciplines from social sciences,
9	to engineering, and even humanities. So there was an
10	interdisciplinary program they are building, and I was
11	program director, trying to coordinate, you know,
12	variety of educational and other efforts in in the
13	area of data science.
14	Q Thank you. And the position that you had with
15	Princeton, was that a tenure track position?
16	A Yes. So I started as a tenure track assistant
17	you know, instructor and an assistant professor, and
18	then promoted to associate professor is tenured, and
19	then full professor is tenured.
20	Q So, you've received tenure at Princeton?
21	A Yes.
22	Q And you, at some point, became a professor at
23	Harvard University?
24	A Right. That's right. So I was recruited by
25	Harvard in 2018.



1	Q And so you moved there in 2018. What is your
2	position at Harvard?
3	A So I hold the position the tenured full
4	professor position in both government department, which
5	is the political science department at Harvard, and
6	statistics department. And this is actually the first
7	such joint appointment in the history of Harvard.
8	Q And is this a tenured position?
9	A Yes.
L0	Q I saw a reference on your CV to Harvard's,
l1	"Institute for Quantitative Social Science." Can you
L2	explain to us what that is?
L3	A Yeah. So Institute of Quantitative Social
L4	Science is interdisciplinary institute at Harvard, which
L5	basically brings all the people who studies statistics,
L6	machine learning, computer science, and focusing on
L7	social science problems. And so, I'm part of that
L8	institute.
L9	Q Okay. What is your role with the institute?
20	A You know, I'm just a member of the institute.
21	I actively participate and organize workshops, you know,
22	advise graduate students, and yeah. I play a variety
23	of roles there.
24	Q And I assume you do research as an academic?

Α

25

Yes. I do.

Q What are your main areas of research?

A Yeah. So my main areas of research -- there are two of them. One is what we call causal inference. This is studying cause and effects. And in my case, I really focus on the cause and effects of public policy, different programs, government programs, non-government organizational (phonetic) activities. The second area of interest, which is perhaps more relevant for this case, is computational social science. So this is the area where you develop computational algorithms, to address and study social problems such as redistricting.

Q And have you published any books in your academic career?

A Yes. I have published book with the Princeton University Press in 2017, I think. And this is a textbook for quantitative social science. So this is introductory textbook for undergraduate students and beginning graduate students, who are interested in studying statistics and machine learning for social science programs that's been widely, widely used across major universities, in their teaching curriculum.

- Q And in addition to the textbook, you've also written various articles, Right?
 - A Yes. I have.
 - Q And are those listed in your CV?



2.1

1	A Yes. they're all listed in my CV.
2	Q Okay. And are these articles have they
3	been published in journals that are peer reviewed?
4	A Yes. So, I have, I think, more than 30
5	(phonetic) peer review journal publications.
6	Q Okay. And those, I think, are listed on pages
7	4 through 9 of your CV; is that right?
8	A Yeah. Yeah. I think so, if you say. Yeah.
9	That's right.
10	Q Okay. Does Harvard have a society for
11	excuse me. Are you familiar with an organization called
12	the, "Society for Political Methodology"?
13	A Yes. It is the Society for Political
14	Methodology is our international organization. It's a
15	premier academic society that basically, the main
16	for the scholars to study using statistics and, you
17	know, machine learning to study political science,
18	basically.
19	Q And are you a member of that society?
20	A Yes. I'm a member of the society. And I also
21	served as the president from 2017 to 2019 of that
22	society.
23	Q And how did you become president of that
24	society? Was there an election or something?
25	A Yeah. I was elected as the president.



1	Q And who are the members of that society?
2	A So the members of the society, there are more
3	than 1,000 academic scholars, basically. Many of them
4	are based in the United States, but there are many
5	others who are based in Europe and Asia. So it's an
6	international organization.
7	Q Thank you. And Dr. Imai, you're here today to
8	testify about redistricting. I'd like to start just by
9	generally asking, what type of analysis you used to
10	analyze the redistricting questions that are presented
11	by this case?
12	A So I specialize in simulation algorithms. I
13	have developed several such algorithms in the past. So
14	I use those algorithms to evaluate redistricting plans.
15	And that's the type of analysis I have expertise in, and
16	I - I conducted for this case.
17	Q Okay. And I'd like to ask you to explain, in
18	a general sense, if you can, how the simulation analysis
19	works. Do you start with certain inputs?
20	A Yeah. So usually, the the goals of
21	simulation analysis is to evaluate certain
22	characteristics of the proposed or enacted plan. And to
23	do that, what the simulation algorithm does is that you
24	specify a set of inputs. So the inputs include the

data. So data is often come from the census -- the

population data. And then, also a set of criteria. So you might be interested in, you know, a set of legal criteria. For example, you want the districts to be continuous or districts to have equal population, you know, or maybe that you want districts to be compact. So you will input the data as well as a set of these criteria. So, that's the choice of analysts. And then, what the algorithm does is it will generate a representative set of -- of the plans, the redistricting plan. So alternative redistricting plans that are consistent with those criteria you specified, based on the data you input. So that's basically what the simulation algorithm does.

Q Okay. And can you talk a little bit more about the criteria or constraints that you feed into the algorithm? Are there certain hard constraints? You know, can you assign weight to them? If you could explain that to the Court, please.

A Yeah. So that's a good -- a very good question. So there are two types of constraints that you can basically put in. Okay. So the first type is what I -- what I might call hard constraints. So these are the constraints that ensures that every single simulated plan will satisfy. So for example, in my algorithm, I'll put, like, continuity as a hard



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1	constraint. That is, every plan the algorithm simulates
2	has a continuous district. There's no plan that will
3	have discontinuous, you know, simulated plan. The other
4	set of constraints, you can think of it as soft
5	constraints. So these are the constraints that often
6	satisfied by the by the various degree. So you can
7	think of like a good example of this is, like,
8	compactness. So compactness is a measure of continuum.
9	It's not a dichotomy of whether a district is at
10	least, mathematically a district is compact or not
11	compact. There's more compact or less compact. So in
12	these soft constraints, you basically provide the
13	different degree of weights. So how much compactness
14	you want to, you know, impose, relative to some other
15	constraints.

Q Okay. And can you describe a little more for us what the output of the simulation algorithm is?

A Yeah. Simulation algorithms is -- literally, the output is many maps. And what's very important about the characteristics of these maps is that they are representative of the alternative plans that are consistent with the set of criteria specified. So think of this as, you know, like a simulated survey sampling, right? There are many, many districts you could draw under a set of constraints. It's impossible, actually.

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Astronomical number. So it's impossible to enumerate
every single possible map. So instead of trying to do
that, because that's computationally impossible, what we
try to do is we'll try to obtain the representative
sample of of that set. And that way, we can
characterize what that set will look like by just using
this this sample that we obtain. And this is very
similar to surveys. Like, instead of interviewing 200
million American voters, you sample, say, 1,000 people
or 2,000 people. And the reason why do that is because
that sample is representative of the population of
American voters. So we can understand the opinion, for
example, by just analyzing the survey sample.

Q And what are the applications for the simulation algorithm? What can it be used to do?

A Yeah. So the main application of the simulation algorithm -- redistricting algorithms, is to evaluate, you know, the characteristic, whether it's a partisanship or some other -- a partisan bias or some other characteristic ratio or dimension of the enacted plan. What's -- yeah. So that's -- that's sort of -- the evaluation is, you know, is the main goal of the simulation algorithm.

Q Can the simulation algorithm be used to create a map that might be enacted into law?



0	Okav.	And I t	hink yo	u men	tione	d th	at t	he r	nar
the policy	y makers	•							
take it a	nd then	enact i	t as a 1	map.	That	's a	role	e fo	or
the map tl	hat can	then	somebo	dy car	n	some	sta	te d	can
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A	So, no.	So my	opinio	n is t	that	the	simu	lati	ion

Q Okay. And I think you mentioned that the map can be used to evaluate -- or excuse me -- the algorithm can be used to evaluate an enacted plan. By what measures? In other words, what could the algorithm be used to test for?

A Right. So you can basically -- once you've obtained the simulated plan, that's representative of the plans that are consistent with the constraints you placed, then you can compare that with enacted plan. And then see whether the enacted plan, you know, for example, favors a particular party in comparison to the simulated plan. Right. So, if the enacted plan is favoring particular party way more than the simulated plan, you think that there is something beyond the set of factors you specified that read through that bias.

Q And how long have simulation algorithms been used to evaluate redistricting plans?

A Yeah. That's a very good question. I think, in the court -- in academic literature, I've been



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studying the simulation algorithms for ten years. I was one of the first academic researchers who really started developing the Monte Carlo methods, which has these representativeness (phonetic) characteristics -- mathematical characteristics. But I think that, in the court, my understanding is that, over the last five, six years, the simulation algorithms have been used in -- in a variety of courts across the country.

Q If you know, how were redistricting plans evaluated prior to the innovation of the simulation algorithm approach?

A Yeah. So that's the -- I think the biggest advantage of the simulation algorithm over traditional sort of way of evaluating redistricting plans -- by traditional way, I mean that, usually what researchers have done in the past is to compute some bias metrics for the enacted plan, for example. And then you say, okay, compared to this bias metric -- like, let's compare this, you know, metric with bias metrics of some other plans. So those plans may come from

Massachusetts, or New York, or Ohio, or somewhere else.

And compared to those plans, this plan that we are trying to evaluate is biased. But as you can -- you know, all probably can tell, such a comparison is problematic because, well, Kentucky is very different



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from Massachusetts. I think everybody agrees on that.
And so that, you know, you're not comparing apples and
apples, right? You're really comparing you don't
know why some whether a particular plan is biased if
you're just comparing that with other state plans
from other states. You can do the same thing within the
same state. Like, you can compare it with the previous
plan. But that could also be problematic because rules
can change, or the population could change. So things
could change. And so, you don't you're not really
comparing the same thing. So what the major
advantage of the simulation algorithm is basically you
use this data same data, right in my case, 2020
census data and same set of rules. right? Same set
of rules that Kentucky requires. And then be able to
generate alternate plans that are consistent with those
data and and rules. Instead of comparing with some
other states, some other different rules, or the
previous, you know, plan. So that's the major
advantage. And I think that's why, at least in the
academic circles, this became this has become the
dominant method to evaluate the redistricting plans.

Q And are there different types of algorithms that are used?

A Yes. The different types of algorithm that



are used to do this, they all belong to something called 1 Monte Carlo methods. So, it's a big family of methods. 2 It's called Monte Carlo methods. Monte Carlo methods 3 basically quarantees that there's a mathematical 4 5 quarantee for the representativeness of the plans that 6 you obtain. As I said, it's impossible to enumerate all plans. So you -- you obtain a, you know, random sample, 7 a representative plan. They're -- within the Monte 8 Carlo family, there are two types of algorithms. One is 9 called Markov chain Monte Carlo. So, Markov chain Monte 10 11 Carlo is you start with a particular map, and then we call this merge and split. So, you randomly pick two 12 districts that are adjacent to each other and then 13 14 split. And then you randomly pick two districts adjacent to each other, merge them, and split. That's 15 16 why it's -- we call merge split. And we repeat this many, many times to obtain different maps. But it's 17 done in a way that the resulting -- resulting simulated 18 19 plans are actually representative of the population of 20 the plans you're interested in. The second one is a 2.1 Sequential Monte Carlo, or some people prefer SMC. starts from the blank state -- okay. And then creates 22 one district at a time -- so you create one district. 23 24 You randomly create one district, and then you create 25 another district, and you create another district until



you create all the necessary districts. Okay. So, instead of starting -- okay. MCMC, you start from the -- the particular map and then start changing it. The SMC, you start from the blank state, and then you start creating the districts. But both are designed to sample from -- you know, obtain representative sample from the population of plans that you're interested in. So they serve the same purpose. It's just the different techniques to achieve that goal.

Q With respect to the MCMC method, the Markov chain Monte Carlo method that starts from an existing map, if that's the starting point for the algorithm, won't that starting point map always look like an outlier, in any analysis that you do?

A No. No. So that's -- well, that's incorrect in couple ways. So, first, Markov chain Monte Carlo has a mathematical guarantee that, you know, the -- the resulting plans are representative. And typically, what we do is we worry about -- you know, starting with the, say, enacted plan. The next plan will be different from enacted plan, but it might be actually very similar because we're just sort of merging the districts, and then spreading them in different ways. So what we do is something called burn-in. We just discard the initial set of plans -- certain number of plans, so that there

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is less impact of the initial plan on the resulting
you know, resulting simulated plans. So this is a
this type of practice is already established. It's not
like Markov chain Monte Carlo has been around for
many decades, and there is an established practice to
make sure that initial plan does not have impact on the
on the resulting plans that you obtain.

Q And so, discarding those initial plans that are created is called, "Burn-in"?

A Yeah. It's called burn-in. And we do that -- I do that in my report, the analysis is in my report, as well.

Q Okay. And are the two different types of algorithms that you've described, Sequential Monte Carlo and then the Markov chain Monte Carlo, are they designed to do different things?

A In theory, they're designed to do the same thing. Now, in practice, you know, redistricting can --you know, redistricting case can be quite different from state to state, like some states are larger. State House district, we have 100 districts instead of six districts in the congressional case. So, you know, some states, there are population centers. And some states impose complex rules. So depending on the situation, you want to be able to use different algorithm. And,

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you know, I can get into more detail of that, but, you know, based on my experience and expertise, I decide, in which case -- which algorithm is more appropriate, given a particular setting that I'm analyzing.

Q And did you use both types of algorithms, for the analysis you did in this case?

A That's right. So for the House districts, I used the merge split algorithm, which is MCMC algorithm. And then, for the congressional district, I used Sequential Monte Carlo.

Q And can you just tell us why you chose to use the MCMC approach for the House map in this situation?

A Yeah. So the House map has 100 districts, as I mentioned. And then, also, as a part of analysis, there is the sort of somewhat complicated restrictions on how the county splits should be -- should be done. And so these type of, you know, large number of districts with somewhat complex constraints, the merge split is -- is a better way of sampling the simulated plans. For the congressional district, that there's only six districts. There have to be a small number of districts. And there are sort of fewer rules that I needed to impose. And so, for those cases, the Sequential Monte Carlo is very effective because, unlike merge split, which sort of sequentially alters the

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district districting plan, you know, SMC spits out a
simulated plan one at a time. So it's they're
independent. Like, each one is separately generated. So
it's a more efficient way of obtaining a sample.

Q Have you had any role in developing the methods you've just described for us, the two types of algorithms, as used to evaluate redistricting plans?

A Yeah. So I have, you know, published, you know, a few articles that develop both type of methods, MCMC, as well as SMC.

Q And do you use a particular or type of software to effectuate the -- or run the algorithm?

A Yes. So I use the software package called, "Redis" (phonetic). It's the -- based on the R programming language, which is one of the popular statistical programming languages. And this is the software my collaborators and I have developed over a few years.

Q And is the Redis software package something that anyone can use?

A Yes. So, one of the things I wanted to do -and this is part of my academic principle, is to make
the methods available to everyone for free. So the
reason is that it allows other researchers to duplicate
and reproduce my results, which is important for

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scientific transparency and, you know, improvement. But
also, it allows other policy makers to use this. And
it's all free and open source. So open source means
that the code is available. So anyone can look at the
code that underlies the pack algorithm. And, you
know, if there's a mistake, they can point that out. Or
if there is improvement that can be made, they can do
that as well. And so that's unlike commercial
software, where the source code is not available. These
are open source, free for download by anyone.

Q And do you do anything to track how much the Redis software package has been used by others?

A All right. So I don't track download counts, but somebody else does. And so there is a web page that keeps track of download counts, you know, from several repository where this software is Housed. And according to their accounts, there are more than 30,000 times been -- that software has been downloaded.

Q Are you aware of any other academics or professionals that study redistricting, using these same methodologies that you've innovated and described for the court today?

A Yeah. So I've seen, you know, papers that -- by some other researchers who use this package, as well as other expert witness in other cases, who use this



1	package.
2	Q And there's something in your CV called the
3	"Algorithm-Assisted Redistricting Methodology Project,"
4	which is a mouthful.
5	A Yeah.
6	Q Is that what is that organization or
7	project? And can you describe for us what it does?
8	A Yeah. It's a project that I had, you know, at
9	Harvard, which basically has a group of graduate
10	students, undergraduate students who are interested in
11	using simulation methods to, you know, evaluating
12	redistricting plans, not only in the United States, but
13	also in other countries as well. So it's a research
14	group that that I lead.
15	Q And Dr. Imai, do you have any prior experience
16	serving as an expert witness in litigation matters?
17	A Yes. I served on several cases.
18	Q And are those matters listed at the end of
19	your CVA? I believe on pages 25 and 26.
20	A Yes. I believe so. Yes.
21	Q And it looks like there's seven cases listed
22	there, including this one as number seven?
23	A Yep. That's right.
24	Q Did all of those cases involve redistricting
25	proposals?



1	A That's right.
2	Q And what type of analysis, in general, did you
3	do in those other cases? Was it is it similar to the
4	simulation analysis you did here?
5	A Yes. I only do simulation analysis. I'm the
6	simulation guy.
7	Q Okay. So that's the only subject matter or
8	expertise
9	A Yeah.
10	Q that you've served in?
11	A Yeah. That's my expertise. I feel
12	comfortable with saying (phonetic) that.
13	Q To your knowledge, have you ever had your
14	expert report, or your opinions excluded by a court, in
15	one of these cases?
16	A I'm not aware of that.
17	Q Are you aware of any challenge to your expert
18	qualifications in any of those cases?
19	
	A I'm not aware of that,
20	A I'm not aware of that, MS. HINKLE: Your Honor, we would offer
20	MS. HINKLE: Your Honor, we would offer
20 21	MS. HINKLE: Your Honor, we would offer Dr. Imai as an expert witness in computational
20 21 22	MS. HINKLE: Your Honor, we would offer Dr. Imai as an expert witness in computational science, and in particular, so simulation analysis
20 21 22 23	MS. HINKLE: Your Honor, we would offer Dr. Imai as an expert witness in computational science, and in particular, so simulation analysis used to evaluate legislative redistricting



1	MS. BECKER: No objection, Judge.
2	JUDGE WINGATE: Okay. So, ordered.
3	BY MS. HINKLE:
4	Q So, Dr. Imai, you've been retained as an
5	expert witness by the plaintiffs in this case, right?
6	A That's correct.
7	Q And are you being paid for your services?
8	A Yes.
9	Q Is the fee that you're charging for your
10	services in this case, a standard fee that you charge?
11	A That's correct.
12	Q Does the compensation that you receive in this
13	case depend in any way on the opinions that you reach?
14	A No.
15	Q I also wanted to ask you, are you registered
16	to vote in the United States?
17	A No.
18	Q And have you often in these disputes,
19	there's sort of a Democratic Party side and a Republican
20	Party side in redistricting litigation. Were you
21	engaged by the Democratic side, in all the cases that
22	are listed in your CV?
23	A Yeah. Democratic side. Yes. But not
24	necessarily Democratic Party for all the cases.
25	Q Would you be willing to work for the



1	Republican side in one of these disputes?
2	A Sure.
3	Q Have you ever been asked to do so?
4	A Yeah. I've been reached out once by by a
5	lawyer representing I'm not sure if the Republican
6	Party or Republican side, but at that time,
7	unfortunately, I was already engaged by the other side,
8	so
9	Q In the same case?
10	A Same case, I had to decline.
11	Q Have you ever turned down an expert engagement
12	due to the political affiliation of the party requesting
13	your services?
14	A No.
15	Q Have you ever turned down an expert engagement
16	at all?
17	A Yes. I have.
18	Q And can you describe for us the circumstances
19	in that situation?
20	A Right. So I have, you know, declined one case
21	I just mentioned, I was already engaged by the other
22	side. I had to decline. I also declined the engagement
23	offer from the Democratic side, in cases where I felt
24	that the case they were trying to make didn't exist.
25	Q In other words, where there wasn't good



1	evidence of
2	A Right. So if I feel that empirical evidence
3	is not strong enough to support the case they're trying
4	to make, I don't feel comfortable presenting my, you
5	know, analysis, so
6	Q And that's happened before?
7	A That happened before. Yes.
8	Q So when you were retained in this case, what
9	were you asked to do?
10	A So I was asked to basically, analyze, and
11	evaluate, enact a plan both House and congressional
12	plan, using simulation operations.
13	Q And did you produce a written report that
14	reflects your opinions?
15	A Yes. I did.
16	MS. HINKLE: Your Honor, may I approach?
17	JUDGE WINGATE: Yes.
18	Q Dr. Imai, if you could look at this document,
19	and let me know if that is a accurate copy of your
20	expert report in this case?
21	A Yes, yes. That's the report I authored.
22	Q And do you adopt the opinions, that are
23	reflected in that report
24	A Yes.
25	Q as your opinions in this case?



1	MS. HINKLE: I would move to introduce
2	Dr. Imai's expert report as Exhibit 2.
3	JUDGE WINGATE: Okay. Any objection?
4	MR. MADDOX: No objection, Your Honor.
5	JUDGE WINGATE: Okay. Thank you.
6	(PLAINTIFF'S EXHIBIT 2 ADMITTED INTO
7	EVIDENCE)
8	BY MS. HINKLE:
9	Q So, Dr. Imai, I'd like to ask you in a little
L0	more detail, what you did to evaluate the plans for this
11	case. Did you have data, regarding Kentucky's
L2	population that you used for the simulation analysis?
L3	A Yes.
L 4	Q And where did you get that data?
L5	A I obtained that from Census Bureau.
L6	Q And did you get did you use any data,
L7	regarding prior elections for purposes of your analysis?
L8	
	A Yes. I did. So I used 2016 and 2019
L9	A Yes. I did. So I used 2016 and 2019 statewide elections data.
L9 20	
	statewide elections data.
20	statewide elections data. Q And where did you get the elections data that
20 21	statewide elections data. Q And where did you get the elections data that you used?
20 21 22	statewide elections data. Q And where did you get the elections data that you used? A So this is called VEST, Voting And Election



1	University of Florida and other universities.
2	Q Did you say it's the, "Voting and Election
3	Science Team"? Is that the source?
4	A Yeah. Voting it's called VEST Voting and
5	Election Science Team. I think that's the full name.
6	Q And to your knowledge, is that a widely used
7	source of election status?
8	A Yes. So this is sort of the go-to source for
9	academic researchers, and it's available publicly
10	available, and anyone can download that data as well.
11	Just like the census data that I used.
12	Q Was the election data that you used available
13	at the precinct level?
14	A Yes.
15	Q And with respect to the population data that
16	you obtained from the census, how granular was that
17	data?
18	A So, census data, you know, the most granular
19	level is available at the block level. However, because
20	election data is the, you know, smallest unit for which
21	election data available is precincts. So normally what



precinct level, and then analyze the precinct level data

academic researchers do and what I followed, is to

aggregate the census data population data to the

sets.

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Q So you mentioned that you used election results from 2016 and 2019 in Kentucky, which races did you use?

A Okay. So that's a great question. So, 2016, it was US presidential election and senate election. In 2019, there were six statewide elections I used governor, lieutenant governor attorney general, secretary of state, auditor, treasurer, and agricultural commissioner.

Q And why did you choose those elections?

Α So these elections are all statewide elections for which the election data available at the precinct level, to the best of my knowledge. And the reason why the academic researchers typically use statewide elections is because when you do a simulation, you're trying to generate lots of districts that are obviously different boundaries from the, you know, district boundaries that were in previous plan, under which the -- those elections were held. So if you look at, for example, like congressional election -- for example, like congressional election returns or the State House returns, those are based on the actual, you know, district boundaries of the previous plans. And what we want to know is, like, what the partisanship would look like under different redistricting plan. So to do that,

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we use the statewide elections where the district
boundaries within the state doesn't exist. So we can
more accurately measure the partisanship of the
districts that are under the simulated plans. So that's
you know, that's the general practice.

Q Does the algorithm that you use do anything to predict future election results or voting choices that voters may make?

A No. I think of it as measuring the partisanship -- partisan lean of each district under, you know, an active plan, as well as under a simulated plan. I don't think of this as a forecasting model or exercise. In fact, you know, those -- those models would require different type of inputs and statistical methods to do that. So for me, the -- these past elections are a way to measure the partisanship and partisan lean of different districting plans.

Q So does the algorithm just assume that voters will vote the same way they have in these past elections?

A Well, it's more like algorithm will take the previous election results as a way to measure what the partisanship of the resulting district will look like. You know, actually the algorithm itself doesn't use the partisanship, right? So, it's -- algorithm uses the



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population. Like, obviously my analysis only based on	
population data, cause you don't want to bias the	
results in, you know, towards one party or another. And	
so the algorithm, itself doesn't use any partisanship	
information. But when you evaluate the enacted plan	
relative to the simulated plan, we going to measure	
partisanship using the past election data. And that's -	
- you know, that's what typically is done in this type	
of analysis.	

- Q Okay. So, one of the redistricting plans you evaluated was for the Kentucky State House of Representatives, what type of algorithm did you use to evaluate the House map?
- A Yeah. So this is the House map I used, the Markov chain Monte Carlo. So that's the one that I used, the merge split algorithm.
- Q And you explained to us why you made that choice. Is that choice something that other academics have also made? In other words, is there agreement in your field, that the MCMC type of algorithm, is best suited for something like the House plan?
- A That's a good question. I hate to sort of characterize as a general agreement just because, you know -- you know, America -- US is a federalism, and each state has such a unique set of rules and political



geography. So I think it really, you know, depends on the set of circumstances you're in and trying to, you know -- also the goal -- goals of analysis could be different depending on the case, right? So it could be House (phonetic) in a gerrymandering case, or it could be racial gerrymandering case. And different cases bring different analyses, which may -- based on different algorithms.

Q Okay. So can you describe for the Court, what criteria you fed into the algorithm for your State House analysis?

A Sure. So what's nice about the simulation algorithms is that it's very transparent, in terms of inputs. So, you know, you reach the inputs, and they go in, and the plans come out. So, the set of inputs I used is basically I told the simulation algorithm to generate a total of 100 contiguous districts and -- for the House. And we -- I also set the population deviation to be plus or minus 5 percent for the House. So at most 5 percent deviation from the equal population criteria. And I made sure that the districts are also reached as compact, as the enacted plan. On average, based on this sort of set of measures that academics use to measure compactness, we try to minimize the number of counties that are being split by the districts. I also

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made sure that the simulator House plan have fewer county boundaries that split, in comparison to the enacted plan on average. They also made sure that the simulator House plan have fewer districts with more than two counties, right? So the districts comprised with more than two counties, in comparison to the enacted So there are fewer of those. And I also made plan. sure that the simulated plans have fewer counties with more than two districts. So some -- another way to think of this is exactly (phonetic) as a county. And if there's more than two districts as part of that county, I try to minimize -- you know, reduce that number of such counties. And importantly, that I did not use partisanship or racial information in the -- in the -in the simulation algorithm.

Q Why didn't you feed the algorithm any partisan criterion?

A Right. So the purpose of this analysis is to evaluate the parts and bias of the enacted plan. So what you want to do then is to compare that enacted plan with the alternate plans that are consistent with all the rules, but, you know, you don't want to partisanship to generate the biased plans. So I don't use partisanship when analyzing the partisan bias of the enacted plan.



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1	Q And why did you not feed the algorithm any
2	information about the race of voters in Kentucky?
3	A Right. So that also depends on the purpose of
4	the analysis, but use of race can be, you know, biased,
5	in terms of like racial gerrymandering. And also, that
6	for some analysis, like in you know, depending on the
7	case, you may consider like, you know, certain type of
8	majority minority districts, for example, to be created
9	in certain parts of the state. But in those cases,
10	you'd have to, you know typically the there has to
11	be a VRA claim, and there has to be some minorities that
12	establishes, you know, where and, you know, the majority
13	minority districts where that should be located, and
14	what should be the percentage. So, in this case, I
15	focused on analysis of partisan bias (Inaudible).
16	Q And you made a reference to the VRA, is that
17	the Voting Rights Act?
18	A That's right, Voting Rights Act.
19	Q Okay. And how many simulated plans did your
20	algorithm generate for the House map?
21	A So I basically generated a total of, you know,
22	10,000 simulated plans for me to analyze.
23	Q And you mentioned the technique earlier called
24	burn-in.

Α

Right.

Q	Is	that	something	that	you	used	here?
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A That's right. So, you know, in practice, what you do is you generate more than what you want. So in this case, I want to generate 10,000. So you generate more than that. So in my case, I generated 72,000 and burn-in basically discouraged the initial simulated plans. I discarded, I think, 1,000 of them for each --each chain, like the parallel chains that goes in. And then there's also a technique called thinning, to make sure that each simulated plan are not too dependent. So I use that technique too, and this is a very sort of standard general technique in the MCMC literature, to obtain the final 10,000 simulated plans.

Q Okay. Why not use as your set of simulations, the original 72,000? Is there some reason that you need to thin down to 10,000?

A Right. So -- yeah. So this is, again, like, a standard practice in MCMC literature, but, you know, the burn-IN is designed to reduce the impact of the initial plan. So initial 1,000 plan has been distorted. And thinning is a way to reduce the dependency of -- of the plan. And so that's what I did.

Q And I'd like to get into what your analysis showed, with respect to the Kentucky House plan.

A Sure.



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1	Q And we're going to use some terminology today,
2	and I want to make sure everyone understands what we
3	mean. You mentioned before that you applied a criteria
4	to require the number of county splits to be minimized.
5	We might call those, "Split counties." Does that mean
6	that all of your simulated plans have 23 split counties?
7	A Right. So in the for the House plan, the
8	enacted plan actually has 23 counties that are being
9	split. We could call this split counties and the
10	simulated plan, all of them also have 23 exact. So in
11	that sense, you know, simulated plan are equal
12	Q And is that the minimum number of counties
13	that need to be split?
14	A I think so. Yeah.
15	Q And you did some further analysis of the 23
16	counties that were split, right?
17	A That's correct.
18	Q And your report references something called,
19	"Multi-split counties." Can you just explain how you're
20	using that term?
21	A Right. So 23 counties that are being split is
22	the total number of counties that are split in some way,
23	but you can imagine the county can be split in many
24	different ways. Like county can split into say two
25	districts, or county can split into three or four

districts. Okay. So instead of just counting how many -- you know, counting how many counties are being split you in some way, I looked out father as to exactly how those counties are being split, so -- yeah.

Q And does figure 1 in your expert report, which is on page 9, does that reflect your analysis of multisplits?

A Right. So that this is one analysis idea, which based basically counting the number of counties -- like computing the number of counties that contain more than two districts. So instead of having one county split, you know, into two districts, it might split into three districts or four districts. So lots of splits within the county. So, figure 1 presents that.

Q And can you describe for us what this figure shows?

A Sure. So figure 1 is -- first, I think you can look at red line. So this is enacted plan. So enacted plan have 18 counties, that has more than two districts. Okay. So under enacted plan, there are 18 counties that are not just splitting into two districts, but three or four or more. Okay. Under simulated plans, on average, there are 15 counties. So on average, three counties or less of such -- such counties. And, you know, it ranges from 13 to 17.

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1	Q Okay. And is this the analysis that informed
2	your opinion that the House plan unnecessarily splits a
3	greater number of counties into more than two districts?
4	A That's right. So this figure shows that it's
5	possible to generate many, many maps and reduce things
6	(phonetic), so I can even generate more that have a
7	fewer number of counties that has more than two
8	districts than you know, compared to the enacted
9	plan.
10	Q Okay. And you also counted the number of
11	House districts that include all part of more than two
12	counties.
13	A That's right. Yeah. So that's another way of
14	thinking about how the counties are being split. You
15	know, previous one is that look at the county, and then
16	count how many districts are in the county. Another way
17	of thinking about it, like, look at the district and
18	then count how many counties are in the district. And
19	so, you can count the number of districts that has more
20	than two two counties. And yeah
21	Q Is that analysis shown in figure 2 of your
22	report, which is on page 10?
23	A Yes. That's correct.
24	Q And can you show us what this analysis
25	reflects?



1	A Right. So again, this is similar to figure 1.
2	So I forgot to mention that these gray bars are
3	histograms. So it tells you relative (phonetic)
4	frequency under simulated plants, how often different
5	number of districts with more than two counties
6	happened. And so under the enacted plan, there's 31
7	Districts that has more than two counties, whereas on
8	the simulated plans, you know, on average, there are 24
9	such districts, which is basically seven districts fewer
10	than the enacted plan on average. And, you know, it
11	ranges from 21 to 30.
12	Q And is that difference statistically
13	significant in your view?
14	A Yes. In fact, now my 10,000 simulated plans
15	has as many districts or more, with more than two
16	districts. So in that sense, it's outlier. The enacted
17	plan is an outlier, has way more districts with more
18	than two counties than any of the 10,000 simulated
19	plans.
20	Q And I think you also looked at the total

- Q And I think you also looked at the total number of county splits in the enacted plan compared to your simulated plans.
 - A That's correct.
- Q And I think that's figure 9 of your report, right?



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some

1	A Yes.
2	Q Which is on page 22. And can you just for us
3	what this is quantifying or showing us?
4	A Right. So you know, this is just accounting
5	for each again, it's for each county, you count the
6	number of districts, and then you add that number up
7	across all counties in the state. And under enacted
8	plan, there are 80 such splits. additional splits, more
9	than necessary. And then this video shows that, on
10	average, the simulated plan has a fewer additional
11	county splits.
12	Q Okay. Dr. Imai, are you aware that the
13	defendants in this case have retained and disclosed some
14	expert opinions?
15	A Yes.
16	Q And have you had an opportunity to review the
17	defendant's disclosed experts, which are Sean P. Trende
18	and Dr. Stephen Voss and his report?
19	A Yes. I have to had a chance to look at their
20	reports.
21	Q I'm going to ask you now about one of the
22	critiques of your analysis in Dr. Voss' report. He



claimed that, avoiding multi-splits in your algorithm

that caused urban counties to be carved up, such that

urban centers are represented by more districts.

23

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1	MR. MADDOX: Your Honor, we have a question
2	about this procedure. Whatever he's about to
3	testify to has not been disclosed in any prior
4	report. So this is outside the scope of what has
5	been disclosed by Dr. Imai, and we think it's
6	inappropriate.
7	MS. HINKLE: And Your Honor, we would respond
8	that it's within the scope of his initial opinions.
9	He's just explaining why critiques that have been
LO	lobbed at his analysis are incorrect, so
11	JUDGE WINGATE: I agree with you. I think it's
L2	a overruled. I think that your Dr. Voss can
L3	critique Imai, and Dr. Imai can critique Dr. Voss.
L4	I think that's fair enough. Okay? Thank you.
L5	BY MS. HINKLE:
L6	Q Okay. So Dr. Voss asserted that in one of his
L7	maps that he assessed is best for Democrats, that only
L8	had 13 splits, that there were certain cities, Bowling
L9	Green, Owensboro, and Hopkinsville that are split up
20	excessively, I think, is the assertion. And he actually
21	said that he saw that across your set of simulated
22	plans. Did you do anything to evaluate that critique by
23	Dr. Voss?
24	A Yes.



And can you describe for us what you did?

Q

A Yean. So one thing that's very important, and
I think is incorrect in the Dr. Voss report, is that one
should never look at a single or a particular maps
simulated plans, right? In order to use the simulation
for evaluation, you have to look at the distribution of
plans. So in not, like, a one specific plan, but all
10,000 of them. And to see, you know, in this case, how
often if you look at across all 10,000, how often
does a particular city will be being split among them?
So this is very similar to survey something, again,
like, just an example of that. Where, you know, if you
are interested in opinion of American voters and you
sample 1,000 voters, you don't want to just look at one
person who you happen to interview to infer what the
Americans think of a whole. And so it's always
important to look at the, you know, all 10,000 plans and
then see if there's a tendency that imposing these
county split constraints will have some impact on a
particular aspect of the plans you're interested in. And
when I look at Bowling Green, Owensboro

Q Hopkinsville --

A Hopkinsville, the -- basically these three constraints have no material impact on how often these cities are being split. So there's no empirical evidence that shows that these additional constraints



have impact on these -- these cities that -- as he asserts.

Q I just want to make clear what you did, make sure I understood what you said. So you removed the multi-split constraints that you fed into your algorithm, and then looked at those three cities, Bowling Green, Owensboro, and Hopkinsville. And you observed what? That there was no material change in the number of splits in those urban centers?

And it -- actually, that's exactly what Yeah. Dr. Voss did. So Dr. Voss basically took my code and then removed that particular constraint, and then actually simulated plans, which I duplicated, I got the exact same simulated plans. However, he did look at how often these plans split these cities. He just picked one map, and then said, look at this map, this map splits this city many times, this city many times, this city many times. What I did is took the output of what he did, and then actually look at how often these 10,000 maps split this city, and this city, this city. And when I compared that with my initial simulation, which had a multi-split, you know, constraint, there's no statistically significant difference. So what that suggests is that this particular constraint that -- that -- he removed has no material impact on -- on those



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1	splits of those cities. And this is you know,
2	another advantage of simulation algorithm is that
3	actually you can add a constraint or remove it, and then
4	figure out whether that has any systematic impact. You
5	know, the maps you're going to get is going to be
6	different from the maps you get if you had you know,
7	it's a random algorithm, so it will generate different
8	maps. But you need to look at, you know, whether these
9	two sets of maps have different characteristics and
10	tendencies (phonetic).
11	Q So the algorithm allows you to isolate
12	A Yeah.
13	Q the impact of particular criteria?
14	A Yeah. And Dr. Voss could have done that.
15	Q Did you do any partisan bias analysis of the
16	enacted House plan?
17	A Yes. I did.
18	Q And what did you do to evaluate that?
19	A Yeah. So I looked at the you know, the
20	for each district, I look at the parts and view
21	(phonetic) of that district based on those six elections
22	that I mentioned. Two elections from 2016 and, you
23	know, six elections statewide elections from the
24	2019.
25	Q And I'd like to ask you a little bit about the



1	analysis that's reflected in figure 3 in your report.
2	A Sure.
3	Q Which is on page 11 of the report. And this
4	is one that we have enlarged, in hopes that the Court,
5	and everyone else will be able to see it as Dr. Imai
6	explains for us. What's being shown.
7	MR. ABATE: Right here?
8	MS. HINKLE: Yeah. Thanks.
9	JUDGE WINGATE: Oh, we can see you. You're
10	good.
11	MS. HINKLE: Okay.
12	BY MS. HINKLE:
13	Q And Dr. Imai, if you'd like to come closer.
14	A Yes.
15	Q So to point things out
16	JUDGE WINGATE: You can get up here, point to
17	what you're it'll pick you up with her
18	microphone.
19	THE WITNESS: Yeah. Thank you.
20	MS. HINKLE: And Your Honor, if you could let
21	us know if you're having any trouble seeing this, we
22	may move it closer, so that you
23	JUDGE WINGATE: I can see it.
24	MS. HINKLE: Okay.
25	BY MS. HINKLE:



Q Dr. Imai, can you walk us through what's being depicted here on figure 3?

A Okay. So this is a somewhat complicated figure, but I'll try to explain. So what I did is, for each plan, I ordered districts by Democratic vote share. So from the least Democratic district to the most Democratic district -- you know, 100 districts of them. And first I wanted to sort of ignore --

JUDGE WINGATE: Let me ask you this, are you doing registration? Is that how you get this?

THE WITNESS: Yeah. That's a very good question. So it's based on the vote share -- average vote share across the past elections, so --

JUDGE WINGATE: Okay. So you're not just doing Democratic registration, you're doing voting patterns?

THE WITNESS: Voting patterns. Yes. Exactly, right at the precinct level. So because it's measured at the precinct level, I can aggregate to the district level to know whether a particular district has, you know, 40 percent of Democratic vote share versus, you know, 60 percent Republican vote share, on average, across the past elections for which I have data. Okay. So first I wanted to know what these funny boxes and then focus on these

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red dots. So these red dots are basically enacted (phonetic) plan. In the House, there are 100 districts, so five (phonetic) to 100 of their dots, nobody can't see anything. So, I focus on competitive districts that are closest to 50/50, prime (phonetic), which is the dotted line. So, all this --

Q Dr. Imai, sorry to interrupt. I just want to clarify one thing. There's reference to district numbers along the bottom of this graph. Are those the districts of the State House Representative districts?

A Yes. Yeah. So these numbers are not the particular district number. It's a 73rd and most Democratic, you know, the -- so the higher -- higher, the number is more Democratic it is, and lower the number is, the least Democratic. So, D1 would be the least Democratic district D100 would be the most Democratic district. And I'm focusing on from 73 to the 84 that are closest to, you know, its most competitive district based on the past election results. So what you see first is these red dots. And then on the y-axis, you see the Democratic vote share. So anything below the 50 percent is Republican leaning and anything above the 50 percent is Democratic leaning. And what you see for these -- these dots are enacted plan. So,



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for example, you know, the 77th District, based on the 1 order by Democratic vote share on the enacted plan, the 2 vote share is about -- Democratic vote share is about 3 47.5 percent -- point. And so one thing I wanted to 4 5 notice is the pattern. Okay. So below 50 percent, you 6 see these red dots -- you know, sort of flat -flattened here, and then there's a big gap, about 2.6 7 percentage point, going from this particular 79th 8 District, the older district, to the 80th District. 9 Which now closest like Democratic -- you know, these 10 11 districts are Democratic-leaning, because it's about 50 12 percent. Okay. So what this -- what this shows is that for the Democratic-leaning districts, these four 13 14 districts remains particularly -- relatively competitive, close to 50 percent. And yet, the 15 16 Republican-leaning districts tend to be far away from the 50 percent. Okay. So tends to be more safe. 17 18 fact, this district that's the closest to 50 percent is 19 -- it's right in the middle, 48 percent. So this would be considered as relatively safe Republican district. 20 Okay. So, --2.1 JUDGE WINGATE: Casey, could he -- just, this 22 23 district, was that D 79? 24 THE WITNESS: Yes. D79. That's right. 25 MS. HINKLE: That's right. Yeah.



JUDGE WINGATE: Yeah.

BY MS. HINKLE:

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And this jump, going from the D79 to D80, this Α big jump is in the academic literature considered -called this type of pattern as signature of gerrymandering, because basically the close Republicanleaning districts in this case are made -- made safer, whereas the close Democratic-leaning districts are made to be competitive. Now, what I did then, is to compare this with a simulated pattern (phonetic). Okay. not only just sort of seeing this pattern, which is -you know, it's often called the signature of gerrymandering in the literature, I want to know whether this is unusual. Like, I want to know whether a simulated plan also have this pattern. Okay. Well, simulated pattern, because there's 10,000 plans, I have this box spot. So box spot basically shows that this box contains 50 percent of simulated plan, so 5,000 of them out of 10,000. And then these lines, which called whiskers, are called typical range -- typical range of simulated plan. Okay. And this is actually median. And what you see is that simulated plan has no gap, right? It's very smoothly shift -- changing, in terms of vote (phonetic) share, from, you know, 73 to 84. And there's no, you know, jump anywhere. In fact, these two



districts or maybe even these three districts, 78,
79, 80, they tend to be Democratic-leaning on average,
whereas in under enacted plan, these are safe Republican
seats. So what this comparison shows and then if you
look at the Democratic-leaning districts that are very
close to 50 percent, you see that these, under enacted
plan, this is much closer to the to the 50 percent
line compared to the simulated plan. So, what this
shows is, you know, under enacted plan, Democratic-
leaning districts are being made competitive more
competitive than the simulation would show otherwise
would show. And the Republican-leaning districts are
being made safer, relative to the simulated plan.

Q Do you draw any conclusions from the data that's reflected on figure 3?

A Right. So this figure shows the evidence of partisan gerrymandering. Favoring the Republican Party, by making Republican-leaning districts safer, and making the Democratic-leaning districts more competitive than - compared to simulated plans.

Q And Dr. Imai, can you comment on the strength of that conclusion or opinion?

A Right. So in, you know, if you think of the statistical outliers -- like these are statistical outliers, right, beyond these -- you know, typical range



1	that you might you know, typical simulated plan
2	range. And what you see is that not only just the one
3	district, but the pattern of several districts that are
4	being made safer than the the simulated plan would
5	indicate. And again here again, it's all of these
6	four districts are being made more competitive in
7	comparison to the simulated plan. So this pattern as a
8	whole so I, you know, as a statistician, I don't want
9	to just put all my basket all my eggs in one
10	basket, but if you look at the multiple districts, you
11	see the pattern of partisan gerrymandering.
12	MR. MADDOX: I'm sorry, can you if you look
13	at which districts? Marginal did you say
14	marginal?
15	THE WITNESS: If you look at these districts
16	that are, you know, closer to the 50 percent.
17	MR. MADDOX: I really I just didn't
18	understand the word.
19	JUDGE WINGATE: Multiple. Multiple.
20	THE WITNESS: Oh, multiple.
21	MR. MADDOX: Multiple. Thank you.
22	THE WITNESS: Yeah.
23	BY MS. HINKLE:
24	Q And Dr. Imai, is the opinion that you just
25	described, is that dependent on your observations about

the enacted plan having more counties with multi-splits
than the simulated plans, or are those separate
opinions?
A Not yeah. They're separate opinions. You
know, they're obviously related, but they're separate
opinions.
Q And this analysis is reflects your
evaluation of Kentucky as a whole, right? All 100
districts?
A Right. Focusing on you know, relatively
competitive districts where you know that
redistricting could make a difference.
Q Did you do any local analysis of partisan bias
in the House map?
A Yes. I did.
MS. HINKLE: Wonder if I might want to take
this down?
JUDGE WINGATE: Yes. So let me ask one more
question. These are 12 are these 12 specific
districts that are the closest to being competitive
without the
THE WITNESS: That's correct. You know, under
the enacted plan, those are specific districts that
are close to, you know, competitive districts, as
you you said. Under simulated plans, they're not



1	necessarily the same districts, because it's a							
2	different plans, so the most competitive district							
3	may not be in the same location. It could be at							
4	different parts of the state, but it tells you how							
5	the competitive district fare, in terms of							
6	partisanship in comparison between the enacted plan							
7	and the simulated plan.							
8	JUDGE WINGATE: Good.							
9	THE WITNESS: And yeah, that's the analysis.							
10	And this is under the standard analysis the							
11	academic researchers do when evaluating the partisan							
12	bias of the enacted plan.							
13	BY MS. HINKLE:							
14	Q You mentioned a term in your testimony about							
15	figure 4 called, "The signature of gerrymandering," is							
16	that an accepted term in academic literature?							
17	A It's a term that has been published, not by							
18	myself, other researchers in in the article and the							
19	(Inaudible) journal.							
20	Q And it's commonly understood to refer to what							
21	you've							
22	MR. MADDOX: Objection, Your Honor. She's							
23	leading the witness.							
24	JUDGE WINGATE: He's an expert witness. It's							
25	all right. I'll let you lead too, Vic.							

1	MR. MADDOX: Thank you. I'll take you up on								
2	it.								
3	JUDGE WINGATE: Okay.								
4	BY MS. HINKLE:								
5	Q All right. So, I'd like to move on now to								
6	talk about your local								
7	A Okay.								
8	Q analysis that you did. I think you looked								
9	at Jefferson County and Fayette County, right?								
10	A That's right.								
11	Q And can you describe generally for us, what								
12	you observed when you looked at those two localities?								
13	A Yeah. What you observe is a pattern of								
14	combining basically, the Democratic voters in the urban								
15	area with the Republicans Republican voters in the								
16	rural area, to create the more Republican-leaning								
17	districts.								
18	Q And is your analysis of Jefferson County								
19	reflected on figure 4 of your expert report, that's on								
20	page 13? And I have that one in the large size as well.								
21	MS. HINKEL: Here's your copy.								
22	JUDGE WINGATE: All right.								
23	MR. MADDOX: Sorry.								
24	MS. HINKLE: Does that work?								
25	MR. MADDOX: Okay.								



MS. HINKLE: Okay.

BY MS. HINKLE:

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So this map shows Jefferson County and the Α surrounding area. And on the left, the maps are colored by -- again, the Democratic voter share based on the past elections under enacted plan. So on the enacted plan, you can see, like, districts, you know, 42, 43, and 30 are very, very Democratic, 34, 41, 40, 44 reasonably Democratic. And then, you know, in the rural area, much more Republican -- Republican districts. And the gray line are county lines -- county boundary line -- boundaries. And solid black line are the district boundary lines on the enacted plan. So, one thing you notice is -- let's look at, like, District 48, for example. So, District 48 takes the sort of urban area of voters and then combines with part of Oldham. sort of crosses the county border and spills into this rural county, and that has very strong Republican-leaning tendency. If you look at District 33, that's another example where you take the urban districts -- urban precincts, and then combine it with the rural districts. In this case, this particular district cross into two other counties. So it's part of Jefferson, but it's also a part of Oldham and part of Shelby. So these are the two districts who --



basically, I see this pattern in other places where the urban -- urban precincts are combined with the strong Republican voter base of the rural -- rural counties. And as a result, the district becomes more Republican. So, if you look at the 48 and 33, that color is a little bit pink, which means that now these two districts, despite the fact that there are -- Democratic voters live here, becomes Republican-leaning. Now, we don't want to just look at this, and we want to compare this with the simulated plans. The question is, is this unusual, or is this -- does this have to happen because of all this population constraints, and continuity (phonetic), and so on. So on the right, you see the same exact map, except now coloring is based on the average simulated plan. So -- at the precinct level. So what you can see this -- is that for each precinct, wherever you look at it, you can ask yourself, okay, on average, how -- how Democratic that district is -precinct to belong to, under the simulated plan. So what you see -- so let's look at 48 and 33. So if you look at 48, you see that slightly blue area here, which means that these voters tend to belong to the district that's slightly Democratic-leaning -- yeah, under the simulated plan. Even though these voters on the enacted plan is actually a part of District 48, which is



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Republican-leaning. Now, going towards closer to more rural area, these voters tend to belong to a more competitive district. So white means Democrats and Republicans are very close. And yet, under the enacted plan, they're a part of 48, which is Republican-leaning. And this area, which is a part of Oldham, part of --District 48. These borders actually tend to belong to more Republican district. That makes sense, because these areas are heavily Republican, so typically these voters could be a part of the district that is within Oldham. However, because these voters combined with the urban voters, 48 becomes essentially, Republican-leaning district. The same pattern appears in District 33. if you look at the District 33 in the urban area, these voters mostly are Democratic. They tend to belong to the much more competitive district under simulated plan. But when they're combined with the Republican voters in the Oldham County and the Shelby County, then the District 33 as a whole under the enacted plan, becomes -- becomes Republican-leaning. So, this pattern of, you know, combining basically the urban Democratic voters with often the rural county, by crossing the county border and creating a district, leads to, you know, Democratic voters belonging to more Republican-leaning districts, in comparison to the simulated plan. And you



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1	see some of these patterns, you know, even like District								
2	29, for example, these voters belong to more competitive								
3	districts. And yet, under the simulated plan, they								
4	would be part of the Republican district. So that's								
5	the Jefferson.								
6	Q Do you have an opinion of what this pattern								
7	shows?								
8	A Right. So this pattern basically shows the								
9	strategy of combining the Democratic urban voters with								
10	the Republican rural voters, to create a								
11	Republican-leaning district.								
12	Q Okay. And you did a similar analysis for								
13	Fayette County, where Lexington is, right?								
14	A That's that's correct.								
15	Q And your analysis of Fayette County is shown								
16	in figure 5, which is on page 15 of your report.								
17	MS. HINKLE: Just going to try to hold this up a								
18	little straighter.								
19	Q So, Dr. Imai, can you walk us through your								
20	analysis of Fayette County, as shown in figure 5?								
21	A Sure. So this is the same sort of set of								
22	figures that I just showed you for Jefferson. So on the								
23	left, you have enacted plan, and on the right, you have								
24	average simulation plan. Under the enacted plan, the								
25	District 77, which is I think it's the most								



1	Democratic-leaning district in the state. And, you
2	know, so that in a sort of in the very urban areas,
3	there's, like, a group of Democratic districts that's
4	created, but I wanted to focus on, like, District 88 and
5	District 45. So 88 takes the, sort of, surrounding
6	environs (phonetic) of this county and then spills over
7	into Scott. Then this this is the heavy Republican
8	area. So by combining some of the Democratic voters who
9	live in these areas with the large number of Republican
10	voters in Scott County, this 88 becomes Republican-
11	leaning. Similarly, if you look at the 45, 45 takes
12	some of the Democratic voters who live here and then
13	combine it with a large number of Republican voters who
14	live in the Jessamine County, again, by crossing the
15	county line. And this creates a Republican-leaning
16	district, even though there are many Democratic voters
17	live there. Now, compare this with the simulated plan.
18	So, under the simulated plan, the voters who live in
19	this area which under the enacted plan called it, the
20	District 88 they are more likely to belong to
21	Democratic-leaning district. So I'm, you know
22	most in many cases, these voters who live around here
23	is most likely to belong to the Democratic-leaning
24	district. However, under enacted plan, because it's
25	combined with this large area of Scott County, the 88



becomes Republican-leaning. Similarly, the voters who								
live in 45 District under the enacted plan, these voters								
are more likely to belong to the competitive districts.								
That's why it's white under the simulated plan. And								
yet, because of under the enacted plan, District 45								
combines these voters with a large number of Republican								
voters in Jessamine County. The 45 becomes a								
Republican-leaning district. So this is again, the same								
pattern as Jefferson, where the urban Democratic voters								
are combined with rural Republican voters, to create								
additional Republican-leaning district. And this is a								
achieved by packing Democratic voters in the center								
city. And you can see that these blue lines blue								
color is much darker than the blue colors under the								
simulated plan. So these voters in the center city								
generally belong to Democratic district, because that's								
where they live. However, under enacted plan, they are								
carved in a way that packs the Democratic voters but								
which then reduces the Democratic vote share or lean of								
the surrounding county, which helps create additional								
Republican-leaning districts, so								

- Q Thank you. Dr. Imai --
 - MS. HINKLE: Yeah. Thanks.
- Q Did you -- I noted it in the rebuttal reports, there was a suggestion that maybe the multi-split



constraint that you fed into your algorithm impacted your observations, with respect to partisan bias. Did you do anything to investigate that critique?

A Right. So I saw that critique that multi-spread constraint that I imposed may have a partisan implication. And as I said in the previous criticism, it's very important to look at all the simulated plans instead of just one or two simulated plans that were chosen in the rebuttal report. So what I did is just take the simulated plan, the -- you know, the --

Dr. Voss or Trende, I can't remember which, but they generated, and then look at that -- basically, he created these figures, right -- same set of figures. And I see no material difference, no statistical difference.

Q Okay. One of the rebuttal experts,
Mr. Trende, stated in his report that he calculated
something called, "Efficiency gap," on all of the maps
in your simulated set of 10,000 alternative House maps,
and asserted that the efficiency gap looks within normal
range on the enacted plan, under the analysis that he
did. Did you do anything to analyze Mr. Trende's
opinions in that regard?

A Yes. I did.

Q And can you describe for us what you did?



A So first efficiency gap is a measure of partisan bias. It's a measure that's used quite often in academic literature, as well as in many court cases. It's not the only measure, but it's one way to measure parts and bias of a particular -- particular plan. Should I explain what that is?

Q Sure.

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So the idea is that, you know, if -- if packing -- so it's trying to capture packing and So packing means, that you try to pack, you know, opposing party voters, supporters into one district, so that they have -- they have less -- you know, fewer chance of getting other districts -- the candidates elected in other districts. So they look at the wasted votes, so how many votes are cast that's beyond 50 percent? So that's unnecessary wasted votes. The other part of this efficiency gap is that cracking, which basically tries to crack the stronghold of the opposing party supporters, so that you -- you know, divide the supporters of a particular party into two districts. So in those cases, you might lose election by, say, close margin, but not quite enough. those votes get wasted. So they look at the -- how the wasted votes differ between Democrats and Republicans. So that's a measure that's -- you know, one measure of



partisan bias. It's not the only measure. There are other measures as well. But what Mr. Trende did in his rebuttal report -- and he calculated efficiency gap under enacted plan, and then compared that with the simulated plans. You calculate the efficiency gap for each simulated plan, and then look at the distribution of simulated plan -- efficiency gap on the simulated plan, and then compare that with the efficiency gap of the enacted plan. So that's a -- that's a good thing, in the sense that it's comparing the enacted plan with the simulated plan. Like, not just the one plan -- one simulated plan, but looking at the 10,000 simulated plans. So -- so I -- that's a good thing. However, what he did is to choose one particular election to compute this efficiency gap, and he chose 2016 presidential election. Okay. When I look at the other elections -- so I can basically repeat the same exercise, but usually in the academic literature, you don't want to rely on the single election, because single election -- as you know, has many different factors going in. Some candidates may be extremely popular or less popular. There may be some other events that happen during the campaign that could influence it. So most of the academic literature, when investigating the partisan bias of a particular plan, you look at wide



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range of races, and then average them out. So when you average across different races, many of these factors may cancel. And you get the general pattern of partisanship, instead of relying on a particular election. And so when I did that -- in fact, if I just take 2016 -- not just the US presidential election, but also Senate race, right? So those are two statewide races for the -- for which data is available for 2016, then his results go away. In fact, the analysis shows that enacted plan is an outlier favoring the Republican Party, as measured as using efficiency gap. If I use 2019 election, there are six of them. I get the same The enacted plan is actually an outlier results. favoring the Republican Party, based on the efficiency If I take all the elections, 2016, 2019 gap measure. together, and then compute the efficiency gap, I get the same exact results. The enacted plan is an outlier favoring the Republican Party over Democratic Party. So what Mr. Trende did was to choose this particular election, and was able to show well, in that case, you know, the enacted plan is within the simulated range. But as soon as you take more elections and combine them -- which is the right way to do because you don't want to rely on again, a particular election, then -- like his analysis -- you know, his result is -- it goes away.



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1	Q I'd like to move on to your evaluation of the								
2	enacted congressional map. Did you analyze Kentucky's								
3	enacted congressional map								
4	A Yes. I did.								
5	Q as part of your expert engagement. And								
6	what type of algorithm did you use to evaluate this map?								
7	A So for this the congressional analysis, I								
8	did I used the SMC, that's the Sequential Monte Carlo								
9	algorithm.								
10	Q And that's the approach that starts with a								
11	blank slate, right?								
12	A That's right. That's a start yeah that								
13	one start with a blank slate and then start building the								
14	district one at a time.								
15	Q And what criteria did you feed into your								
16	algorithm, when you were evaluating the congressional								
17	map?								
18	A So I made sure that the algorithm creates a								
19	total of six continuous districts. That's the number of								
20	congressional districts. And I used the overall								
21	population deviation of plus, minus 0.1 percent. So								
22	that's the at most, the simulated plan have the								
23	maximum deviation of plus, minus 0.1 percent.								
24	Q Do you know, in terms of real people, what								

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plus or minus 0.1 percent is?

A Yeah.	That's a	good	question.	Yeah.	So the				
choice of this	is based c	n the	fact that	: I'm wor	king				
with the precinct-level data. So precinct-level data									
is, you know, on average, I think maybe 2,000									
population, or something along those lines. And plus,									
minus 1 [sic] percent is usually between 700 to 800									
people in Kentucky.									

- Q And it's plus or minus 0.1 percent, right?
- A Plus, minus 0.1 percent. Yes.

Q And why not require your algorithm to require absolute equality among the districts?

A Right. So in the -- you know, when the states -- many states enact their congressional plan, they often go down to one person difference. So the population based on the census is different from the -- another district, at most one or two people, right? However, for simulation analysis, which is designed to evaluate the characteristics. It's not designed to generate the plan that someone can pick and enact -- because we are based working on the precinct-level data, we don't have ability to go down to one person, right? So one person would require census block level data for which election results are not available. So the fact that we use, as in many partisan analysis of -- in academic literature, we use precinct-level data. And



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Q Did you also include a compactness criteria in the algorithm, for purposes of congressional map?

A Yes. Yes. Generally, these algorithm are designed to generate compact districts, because if you think about, you know, all possible districts, then you'd have many snake-looking districts that we -- we would not care. So, we focus on -- these algorithm are designed to generate relatively compact districts.

Q And Mr. Trende's rebuttal report indicates, that you used a compactness parameter of one; is that right?

A That's correct.

Q And he thought that maybe a map drawer would choose 0.5 or two, as opposed to one, as a compactness measure. Do you have any reaction to that?

A So map drawers should not be using the algorithm to generate the enacted plan, so they should never choose the parameter. But if the point is to say, more realistic choice is the compactness parameter, it should be 0.5 or two, that's inaccurate. Because I've analyzed many others states as part of my academic research -- and of, you know, part of expert witness

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1	work, but you never choose those extreme values. That
2	would be really pushing the algorithm too far to, you
3	know, keep these theoretical guarantees that I
4	described, that are very important part of the
5	algorithm. So, typically we would may change like
6	1.1, 1.05, 0.97, 0.95 to make it a little bit more
7	compact, a little bit less compact. But never the range
8	of 0.5 or two that's suggested by Mr. Trende.
9	Q Okay. And did you feed any criteria relating
LO	to county splits into the algorithm, for purposes of
11	evaluating the congressional map?
L2	A Yes. I did.
L3	Q Can you describe those for us?
L4	A Yeah. So I made sure that the simulated plans
L5	have fewer than the number of counties that are being
L6	spread under the enacted plan.
L7	Q And again, did you use any partisan criteria,
L8	as part of the criteria for the algorithm?
L9	A No. And I should also note that each county
20	is spread, you know, at most once, because that's the
21	important part of the criteria. So the simulation I
22	instructed the simulation algorithm to just do that.
23	Q And did you use any racial criteria, as part



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of the algorithm?

No.

1	Q How many simulated plans did you generate, for
2	purposes of evaluating the congressional map?
3	A 10,000. And that choice is just motivated by
4	statistical efficiency. Like, if you have 10,000,
5	that's that's actually way more sufficient to yield
6	accurate conclusions.
7	Q Okay.
8	A And, you know, obviously, I could generate
9	more, but that's generally pointless, at that point.
L0	Q When you're generating the simulated plans, is
L1	it possible to freeze a particular district? In other
L2	words, to lock in one district and then simulate the
L3	remainder?
L4	A Yeah. That's possible.
L5	Q And is that something that you would recommend
L6	doing and evaluating a map using the simulation
L7	algorithms?
L8	A Depends on the context. So for example, you
L9	know, in some cases where a particular district boundary
20	is at dispute so, you know, if you have say two
21	districts, and the boundary between those two districts
22	is in dispute, then you could freeze the rest of the
23	state, and then generate those two districts to see how
24	unusual those boundaries are.
25	Q If you were trying to measure the compactness
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of an enacted plan, what impact would freezing a particular district have?

A Right. So that you have to be careful, because freezing one district, will basically freeze that district boundary surrounding it. So that has an impact on compactness of the surrounding district. And so the conclusion has to be, you know -- you have to be very careful, right, because it has that -- freezing that one district will have an impact on compactness of other districts, that are neighboring with the -- with the frozen district.

Q Did you consider Kentucky's historical congressional maps, in developing your algorithm?

A No.

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Q And why not?

A So typically, when I evaluate the partisan bias of the enacted plan, I do not bring in the previous maps. The reason is that, we don't know what went to the previous maps, what factors were considered to create the previous maps. And so one of the important aspect of simulation algorithm is transparency. So if - you know, you specify a set of criteria, and under that criteria, the algorithm will generate the plans. So when you -- if you input the previous plan, whatever the consideration that was used to generate that plan, would

1	affect the results and may bias my conclusion in one way
2	or another. So when if I evaluate the partisan, you
3	know, bias of an enacted plan, I don't use the previous
4	maps.
5	Q And Kentucky currently has one Democratic
6	representative and five Republican representatives in
7	the US Congress. Why not add a criteria to your
8	algorithm that would ensure at least one Democratic
9	representative from Kentucky?
10	A Yeah. Because that would bias my conclusion.
11	That would be sort of encouraging partisan
12	gerrymandering. So I, you know in order to evaluate
13	the partisan bias, you don't use the partisan
14	information, right? That would be a bad idea.
15	Q Okay. And you focused your analysis of the
16	congressional map on Franklin County, right?
17	A Yes.
18	Q Why did you do that?
19	A Franklin County is notable because it's part
20	of this District 1 that travels from the west side of
21	the state, all the way to the center of the state. And
22	Franklin county is a, you know, important part of that
23	district and dispute.
24	MS. HINKLE: And I don't seek to introduce this

map through this witness. But I want the Court to

be aware that we have included maps of the districts 1 in the front of your binders --2 3 JUDGE WINGATE: Yes. MS. HINKLE: -- if you'd like to look at that. 4 5 And if it's all right, Your Honor, I'd like to give 6 the witness one of those as well, to reference as needed. 7 JUDGE WINGATE: That's all right. 8 BY MS. HINKLE: 9 10 So that is -- is that the enacted 11 congressional map? 12 Α That's correct. So, what was the first step in your analysis 13 14 of looking at the enacted congressional map? All right. So, the first step of analysis was 15 16 to evaluate the compactness of this district. And how did you do that? 17 Α Well -- so, you know, one could look at it and 18 19 then see it's not compact. But because I'm a simulation 20 expert, what I do is, I'm going to compare the 2.1 compactness of District 1 under the enacted plan with simulated plans, the compactness of the district that 22 contains the Franklin County under the simulated plan. 23 24 And you never know that this shape may be necessary in 25 order to comply with, you know, population and other



criteria. So you always want to be able to you know,
you want to you want to be able to compare this with
a simulated plan that comply with all this other set of
requirement, and then see if this is an outlier.

Q And did you use the full set of 10,000 simulated plans to do this analysis?

A No. I subset it to the 93 percent of the simulated plan, so most of it, but -- which did not split the Franklin County.

Q And why did you make that choice?

A Because the enacted plan that's not spread to Franklin County, and I wanted to make sure that -- you know, that I'm comparing apples and orange -- apples instead of comparing two different -- completely different districts.

- Q And you compared the compactness of the enacted first district with those in your simulated plan, right?
 - A That's right.

Q What compactness measure did you use?

A So I used the measure called Polsby-Popper compactness score, which is one of the very standard metric of compactness measure. I also used the Reock measure, which is a related measure that's again, used in academic literature.



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1	Q And is your analysis of the compactness of the
2	first district reflected in figure 6 of your report on
3	page 17?
4	A That's correct.
5	Q And can you describe for us what this shows?
6	A Right. So this figure shows, again, the
7	compactness of enacted plan, which is the red line, and
8	compactness of the district that contains Franklin
9	County as a whole, which is shown as a histogram, the
10	gray bars. And the Polsby-Popper compactness score is
11	the larger the value is, the more compact it is. So, if
12	the value is smaller, that means less compact. And as
13	you can see, almost all the simulated plans generate the
14	district that contains the Franklin County as a whole,
15	that is much more compact than the District 1 of the
16	enacted plan. In fact, more than 99 percent of the
17	simulated plans can generate the district the
18	corresponding district that is more compact than the
19	District 1 in the enacted plan. Which led me to
20	conclude that District 1 is outlier, in terms of the
21	lack of compactness of that of that shape.
22	Q So you're measuring the compactness, looking
23	just at District 1, right?
24	A That's right.



Does the compactness of one district affect

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other districts?

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A Yes.

Q In other words, is there necessarily a tradeoff in the compactness, if you change the compactness level of one district versus another?

Α I see. So compactness of one district affects the compactness of the other districts because if you change the district boundaries of one district, you know, the district boundaries of the surrounding district have to change, and then that will also lead to the change in other districts. So in a sense, they're all related. However, there is no general tradeoff between -- like, if you make one district more compact, you have to make another district less compact or vice In fact, you can create a map that all the districts are non-compact. You can imagine, just like make a lot of snakes, and that will lead to the map that has many, many districts that are non-compact. So they are related, but there's no general tradeoff, right? fact, what the simulation shows, it is possible to create -- because this simulation actually makes sure that the average compactness level is the same as the enacted plan. So my simulated plan, on average -average across districts, have the same compactness level as the enacted plan, right? But the enacted plan



creates this District 1 that it's highly non-compact.
What the simulation idea shows, is that even if you keep
the overall compactness average compactness the same,
I don't need to create this highly non-compact district.
I can create the district that are much more compact
across across the board on, you know on average,
basically.

Q So if you made the first district in

Kentucky's map more compact, does that necessarily mean
that the other districts become less compact?

A No. And that's exactly what the simulation shows, right? So it's possible to make the District 1 more compact, without changing the overall level of compactness of the map.

Q Okay. Dr. Voss' rebuttal report suggests that your relaxed -- as he describes it, a relaxed standard for population equality caused your simulation to produce more compact maps. You've described for us why you used the population measure that you did, but could you -- could you tell us what you did, if anything, to investigate Dr. Voss' critique, in this regard?

A Yes. So the Dr. Voss critique on my choice of population deviation -- which is 0.1 percent, about 700 to 800 people difference across -- from the ideal target population was positive to me because in his report, he

1	says I you know, this choice is too big. So the 0.1
2	percent is too large. But as I explained at if
3	you're using the precinct-level data, which is the
4	analysis that I'm conducting, then the 0.1 percent, 700
5	to 800 people, is appropriate choice because the
6	precinct is not as small as the census blocks
7	(phonetic). Okay?. And he says that in the report, he
8	chose you know, he pinched he reduced that
9	population deviation, but then he cites number 0.001,
10	which is 0.1 percent, which is exactly what I did,
11	right? So that was puzzling to me because he's saying
12	that, well, I picked a too too big a number and he
13	says I set it you know, he set it to 0.001, but
14	that's exactly the same number I chose. So I was a
15	little confused. But then when I looked at his code, he
16	actually set it to 0.00001, which is basically 0.001
17	percent, or seven or eight people. Okay? So instead of
18	choosing 700, 800 people, which I did because of the
19	size of the precinct, he chose in the return of the
20	analysis he conducted in his report, he chose 0.001
21	percent which is seven to eight person people. Okay?
22	Q And can you describe for the impact that

Q And can you describe for the impact that choice has on the algorithm?

A Right. So first of all, that's not appropriate choice because precincts are much bigger.



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Precincts are not the size of seven, eight, six, five 1 people. On average, I think, you need a couple thousand 2 people. And so, if you set the tolerance (phonetic) --3 population tolerance to that low, there are so few 4 5 precincts that you'll be able to move to generate the 6 plans. And so when I rerun his algorithm -- the software that I wrote, generates lots of warning. 7 Basically, it says this is not a good choice, and it has 8 some potential impact on the theoretical properties of 9 the algorithm because you're choosing too tight a 10 11 population threshold, given the dataset that you're 12 analyzing, and so that's one concern. So that any results that might come out from such a tight population 13 14 threshold, when the pop -- data itself is precinct-level There may not be mathematical quarantee that --15 16 that make these algorithms so attractive and powerful. How -- in addition, when I actually run this, right, 17 even though it gives lots of warnings, you still get --18 19 generate 10,000 plans. And I look -- I recreated the 20 figures that are in my report, using those output. 2.1 There's no material difference. So this is, again -okay. What's important is to look at the distribution 22 23 of the plans as a whole. So you cannot just choose one 24 particular plan of 10,000 and draw some general 25 conclusions. In order to general -- draw general



1	conclusions, you need to look at the entire simulated
2	output. And when you do that, the population deviation,
3	at least of his choice, has no material impact on the
4	conclusions that I drew.
5	Q And did you do anything to analyze the
6	partisan bias in Kentucky's congressional map?
7	A Yes. I did.
8	Q And can you describe for us what you did to do
9	that?
10	A Right. So for this analysis, I looked at the
11	Democratic vote share of the districts that contain
12	Franklin County. So, that's basically for for the
13	enacted plan, that's District 1.
14	Q And is your analysis, in this regard,
15	reflected in figure 7 on page 18 of your report?
16	A That's right.
17	Q And can you please describe for us what this
18	figure 7 shows?
19	A Right. So the figure 7, just like previous
20	figures, focus on the districts that contain Franklin
21	County as a whole. So for under the enacted plan,
22	this would be District 1. And District 1, under the
23	enacted plan, has, you know, Democratic share of votes
24	around 35 percent. And the gray histogram on basically
25	shows what would be the Democratic vote share of this

1	corresponding district under simulated plan. And what
2	you see here is that under simulated plan, the Franklin
3	County will belong to the district that is much more
4	Democratic. Okay? So compared to the enacted plan.
5	So enacted plan is basically making Franklin County a
6	part of much more Republican-leaning district, in
7	comparison to the simulated plans.
8	Q And is your observation shown in figure 7, is
9	that statistically significant?
10	A Yes. Again, it's more than 99 percent of the
11	plans have higher Democratic vote share for the
12	corresponding district than the enacted plan. So I say
13	this is statistical. All right.
14	Q And is your opinion, with regard to the
15	partisan impact of the enacted plan dependent on your
16	observations with respect to compactness of the enacted
17	congressional plan? Are these separate?
18	A They're separate conclusions. You know,
19	obviously, they're related because the way that District
20	1 is constructed is this, you know combining the
21	highly Republican-leading counties with
22	Democratic-leading county, to make additional Republican
23	district.
24	Q Are you giving an opinion today, that Franklin
25	county should be in a in a district that's



1	represented by a Democratic representative?
2	A No.
3	Q When you were developing the criteria for your
4	algorithm, were you attempting to create a map that
5	might elect two Democratic representatives from
6	Kentucky?
7	A No. Because my goal is to evaluate the
8	partisan bias of the plan. So I did not use partisan
9	information as input to my algorithm.
10	MS. HINKLE: Can you give me a minute?
11	JUDGE WINGATE: Are we at a good breaking point
12	for lunch?
13	MS. HINKLE: I think we are.
14	JUDGE WINGATE: Huh?
15	MS. HINKLE: We are, Your Honor. Thank you.
16	JUDGE WINGATE: Okay.
17	MR. MADDOX: Your Honor, I may I ask is if
18	Mr. Imai's examination has concluded or do you have
19	further questions?
20	JUDGE WINGATE: No. She's got further
21	questions. I'm just want to go to lunch.
22	MR. MADDOX: Yeah. I understand. I thought
23	maybe she was done.
24	MS. HINKLE: Well, if you would give me a
25	minute to confer, I can do that.



1	JUDGE WINGATE: Yean. Why don't we just see if
2	you're yeah. If you're if you've got a couple
3	more questions, that's all right.
4	MS. HINKLE: You know, I'd like to reserve the
5	chance to ask a few more questions after lunch, if I
6	may?
7	JUDGE WINGATE: Yes, ma'am. You-all can talk
8	over lunch, and it looks like you're getting pretty
9	close to the end.
10	MS. HINKLE: I am certainly very close.
11	JUDGE WINGATE: Okay. All right. Usually, I
12	give an hour and 15 for lunch. So we will return at
13	1:30. Okay? 1:30. Thank you all.
14	MR. ABATE: Your Honor, is it
15	JUDGE WINGATE: You all need to talk to me
16	about anything? You can come up here.
17	MR. ABATE: Yeah. We were just wondering
18	if it's possible to do a slightly shorter break? I
19	don't know how long cross examination will last.
20	Dr. Imai does have his a plan to return to the
21	airport tonight
22	JUDGE WINGATE: Why don't you just go down to
23	Buffalo Trace and do a tour or something? Yeah.
24	What time does his plane leave?
25	(OFF THE RECORD)



1	JUDGE WINGATE: We're on the record. Okay.
2	You're still under oath, Doctor, okay? All right.
3	You may continue.
4	MS. HINKLE: We have no further questions on
5	direct exam.
6	JUDGE WINGATE: Okay. No further questions.
7	Very good. All right. Very good.
8	CROSS EXAMINATION
9	BY MS. BECKER:
10	Q Good afternoon, Doctor. I'm Heather Becker. I
11	represent the Commonwealth. I want to understand two
12	points from your testimony. Using your ensemble and the
13	vote share that you calculated, it's true that 76 of
14	Kentucky's House districts lean in favor of Republicans,
15	right?
16	A I don't recall the exact number.
17	MS. BECKER: Okay. Can I use your box bar
18	(phonetic)?
19	MS. HINKLE: Sure.
20	MS. BECKER: This is going to be okay for the
21	cameras?
22	CLERK: Yeah.
23	BY MS. BECKER:
24	Q So, again, looking at your chart, 76 of
25	Kentucky's House districts lean in favor of Republican -



1	- of the Republican Party, right?
2	A That's a different you mean that 76 out of
3	100?
4	Q Uh-huh.
5	A That's not necessarily the case because
6	should I explain the reason or?
7	Q Sure.
8	A So each of these (Inaudible) is a distribution
9	of, you know, the district for like like older
10	(phonetic) at 76. So, it doesn't mean like every single
11	point. You cannot really compare across districts. So
12	it's possible that in for a particular simulated
13	plan, you know, 78 or 74 of the districts are leaning
14	towards one party or another. So you'd want to
15	calculate the actual number, expect the number of
16	than five (phonetic) districts under the simulated plan.
17	So, that would be a different prop (phonetic). This
18	prop wouldn't necessarily tell you that.
19	Q So, right here
20	A Right.
21	Q the average of your 76th ordered district,
22	falls the median, falls below the 50 percent line.
23	A Right.
24	Q You would say that leans Republican?
25	A So, the average, 76 I guess, my question



maybe I'm not understanding your question. But average
Democratic listen, Democratic, you know, vote share
of the average 76 simulated plan is yes for what 49,
you know point, whatever there.
Q And everything before it?
A Well, that's everything before what? I
just want to be careful about what I'm trying to
being asked.
Q This is your ensemble?
A Right.
Q The average of your ensemble districts would
order at 76 leading Republican?
A Right. Among the all 76, you know ordered
districts among the simulated, the average vote share
for that district is below 50 percent. Yes. That's
right.
Q Okay. And taking the average vote share of
the district that contains Franklin County in your
congressional simulation, the average Democratic vote
share was 43 percent, right?
A I don't memorize what is numbers. So I don't
so it's this is congressional, not the House?
Q Yes.
A Okay. What was the question again? Sorry.
Q When you look at the average vote share of the



1	district that contains Franklin County in your
2	congressional ensemble
3	A Oh, okay.
4	Q it's 43 percent?
5	JUDGE WINGATE: You're talking about Republican
6	votes, right?
7	MS. BECKER: He does it ordered by a Democratic
8	vote share. So it would be a 43 percent democratic
9	vote share.
10	A 43 yeah. Yeah. Okay. Right. So 43.6
11	percent. You know, among the simulated plans that
12	contain for the for the district that contains
13	Franklin County, has a 43.6 percent on average
14	Democratic voter share. That's right.
15	BY MS. BECKER:
16	Q Okay. So, I think that's the bottom line on
17	your report. What I want to do now is talk about how
18	you got there. So I want to make sure I understand.
19	Your ensemble for the House analysis contained 10,000
20	maps, right?
21	A That's correct.
22	Q And you generated a like number for your
23	congressional analysis?
24	A I generated 10,000 simulated plans for
25	congressional analysis as well.



1	Q	And your algorithms could have made many
2	different	sets of 10,000, right?
3	A	Yes.
4	Q	Is 10,000 the universe of all the maps that
5	could have	e been created?
6	А	No.
7	Q	And you didn't look at any of the simulations
8	in your e	nsemble, did you?
9	А	What do you mean, "Look at"?
10	Q	You didn't look at you didn't generate maps
11	from your	simulations? You didn't look at what they
12	looked lil	ke in real life?
13	А	I did I did look at some of them.
14	Q	Before you received our reports?
15	А	Right. I mean, not all of them, but some of
16	them. Yes	5.
17	Q	All right. For your work in this case, you
18	used your	Redis software, correct?
19	А	That's correct.
20	Q	And does you call it, "R"?
21	А	Yeah. R is the statistical programming
22	language t	that the Redis based off.
23	Q	Does R contain both your SMC and MCMC
24	algorithms	s?
25	А	S what do you mean, "Contain"?
	1	



1	Q Are they written into the R program?
2	A It's part of the R package. Some parts are
3	written in, you know, C program just because it's
4	faster.
5	Q I don't want to go through it in great detail,
6	but I would like to go through some of your code with
7	you. Okay?
8	A Okay.
9	MS. BECKER: Can you go
10	BY MS. BECKER:
11	Q Are you familiar with this code?
12	A This is the congress. This is the code for
13	the congressional simulation?
14	Q So, this is the code you ran an R for your
15	congressional simulation analysis?
16	A Uh-huh. Uh-huh.
17	Q For right now, I'd like to mark it for
18	identification as Exhibit 1. Can you locate, for me, in
19	this document, where the algorithm you used to generate
20	your analysis is?
21	A So the algorithm is in the package. So, this
22	is the code that caused (phonetic) the algorithm.
23	Q So your algorithm's not in here?
24	A Yeah. Algorithm is in the package Redis. So
25	Redis has a set of code that's, you know, contain that



in that package. And this code caused the Redis. 1 Okay. I'd like to look at the same document 2 3 for your House analysis. Α Okay. 4 5 Q So is this the House analysis -- the House 6 code analysis that you used? Uh-huh Α 8 I'd like to mark this for identification as 9 Exhibit 2. Can you locate, in this document, where your 10 code is that you wrote for your analysis? 11 Α I'm not sure I'm understanding your question. 12 The code that you used to generate your 0 simulation ensemble, where is it in this document? 13 14 Α Simulate? Yeah. So, this -- it's 03 -- or 03 simulate SHDMS. 15 16 And it goes on for 44 pages? Well, it depends on what you mean by, with --17 so -- anyway. It has all -- all the prepping the data, 18 19 and setting of constraints, and all that is, you know, prior to actually scoring (phonetic) the simulation 20



1	Q Is that pre-code located in the R package?
2	A So this code is not a part of the package.
3	This is the a code that caused the R package
4	function, which has the algorithm program, if that makes
5	sense.
6	Q I would also need to know what your R package
7	said, to know how to interpret this?
8	A What do you mean by, "Interpret"?
9	Q To be able to use this, I would also need to
10	know your R code, right?
11	A You need to be able to install the package and
12	to run this. That's right.
13	Q Well, let's look at your R code.
14	JUDGE WINGATE: Heather, is the first one going
15	to be Exhibit 1 and then 2, then this 3?
16	MS. BECKER: Well, I guess, as a matter of
17	housekeeping, Judge, we do want to make sure that
18	our binder is numbered as 1, what we did at the very
19	beginning.
20	JUDGE WINGATE: Okay. Let me see here.
21	MS. BECKER: So, we could do that as 1 and then
22	I'll do the Congress code is 2. House code is 3.
23	And what Alex is handing you, the R code is 4.
24	JUDGE WINGATE: Well, what in your binder,
25	where are they listed?



MS. BECKER: Those are not in our binder. These 1 are --2 3 JUDGE WINGATE: Got you. MS. BECKER: Those are just the stipulated 4 5 documents. 6 JUDGE WINGATE: Got you. Got you. Got you. 7 Got you. So, the first one you have is 1? CLERK: No. 8 9 JUDGE WINGATE: No? 10 MS. BECKER: It would be 2. 11 JUDGE WINGATE: 2. CLERK: District binder is 1, the stipulated 12 facts. 13 14 JUDGE WINGATE: The stipulated facts is 1? Right. 15 MS. BECKER: 16 JUDGE WINGATE: Now, I'm understanding. and then 4 --17 18 MS. BECKER: 2 would be the one that says, "Run 19 Congress." 3 would be the one that says, "Run 20 House." And then 4 will be the one that starts with 2.1 the "@RD name." 22 JUDGE WINGATE: Got you. This one right here. 23 MS. BECKER: All right. So --24 JUDGE WINGATE: I got them now. All right. I'm 25 going put these all in here.



BY MS. BECKER:

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- Q So, the -- I'm going to -- let me just mark that one as number 4. So the document we're on is number 4. This is your -- this is your R code, correct? The R package software?
- A I -- I think so. I mean -- I assume you print it out from the Redis file. I mean, I don't memorize it with line. So, you know, assuming that this was printed out from the actual package. Yes.
- Q So once I have all three of these sets of code, I'm ready to start your simulation process, yes?
- A You have to first, you know, install the package. The package is a set of programming files. So you have to, you know, download that and install. And then once that's done, then yes, the other R files can be used to generate the simulated plans.
- Q And so that's roughly 11 files, and 13 libraries, and roughly 90 pages of code that I would need to have under my belt, before I could start what you did?
- A Yes. But -- yeah. that's -- I mean, you have to have them. Yes. Without them, it's -- you wouldn't be able to run. That's correct. But you know, R, itself, has many, many files. So if you -- you know, if you -- if your definition is you have to have all these



programs, then you have to have all R all the code
that's a backbone of R has to be printed out as well.
That would be, you know, hundreds of files.
Q So, you say that you use R, so that anyone can
recreate your work?
A That's correct.
Q I would need an expert to tell me how to do
what we just walked through. I might need you.
A Yeah. But others can not just me, but
others can also use it as well. So yeah. You may
need some expertise to use R and associate packages, but
yes, that's correct. But you don't need me, per se.
Like you could have somebody else who is familiar with R
and R packages.
Q All right. You've never been appointed to
draw a redistricting plan, have you?
A No.
Q I have a couple questions about some of the
new analysis you unveiled today in your direct
testimony.
A Sure.
Q You reviewed Professor Voss' report. You
reviewed Mr. Trende's report. When did you form the
opinions that you shared today?



This weekend, I think, after I received and

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reviewed the you know, the report and the and the
code associated code and data.
Q And did you disclose your opinions to your
Counsel?
A What do you mean by, "Disclose"?
MS. HINKLE: I'm just going to object, to the
extent your question is trying to invade our
communications and work products.
MS. BECKER: Judge, I'm entitled to
JUDGE WINGATE: No. It's not really that. It's
just when did you disclose, that's a typical
question. Yeah. You can answer that question,
Dr. Imai.
A Okay. Yeah, I show the results I shared
the results of the analysis with Counsel.
BY MS. BECKER:
Q When?
A When? Sunday.
Q Did you provide Counsel with any of the
underlying data for your conclusions?
A I used the data I received from you know,
for the Dr. Voss and Dr. Trende's Mr. Trende's
analysis. So they the Counsel had those.
Q I'd like to look at your CV really quickly.
A Okay.



1	Q Counsel, on direct, asked you about your
2	publications. How many of your publications relate to
3	the work and analysis that you're doing here today?
4	A Okay. Yes. Three of them relate very closely
5	to what I'm doing today. But there are there are
6	other there are other publications that are about
7	simulation algorithms and, you know, general area of
8	research. But three are specifically about
9	redistricting simulation algorithms.
10	Q I'd like to talk about some of your some of
11	your report now.
12	A Okay.
13	Q So looking at your House analysis, the you
14	would agree that the input criteria that you choose are
15	important to the outcome?
16	A Yes. I do agree.
17	Q So they have to be chosen carefully?
18	A That's that's correct.
19	Q And if you use additional or other criteria
20	that could change your conclusions?
21	A That could. Yes.
22	Q So I want to look at page 7 of your report.
23	Down here in paragraph 16, you have several bulleted
24	points. These criteria these are the constraints
25	that you imposed in your simulations, right?



1	A Right. I mean, in the, you know, actual	
2	constraint itself is mathematical but this described.	
3	Q Okay. So these are the criteria and then you	
4	assign constraint levels to the criteria?	
5	A Right. So this is my attempt of, you know,	
6	describing the constraints that I used.	
7	Q I think what you're talking about, you	
8	described a little bit better on page 22. Can you turn	
9	to the appendix of your report?	
10	A Right. That's the details.	
11	Q So you say that you set a county split	
12	constraint at a level of ten. And you set a county	
13	multi-split avoidance at a constraint of seven, and a	
14	custom constraint at a level of ten.	
15	A This is paragraph 11, on page 22, is that	
16	Q I'm sorry?	
17	A Is it this is paragraph 11 on page 22?	
18	Q 10 and 11. Yes.	
19	A 10 and 11. Yes.	
20	Q What's the significance of a constraint of	
21	seven?	
22	A Do you mean statistical I'm just trying to	
23	understand your question.	
24	Q Sure. You chose a constraint of seven.	
25	A Uh-huh.	



If you -- no one told you to set it at seven, 1 0 you picked seven. 2 Oh, okay. How did I -- why did I choose 3 Α seven? 4 5 Q Well, so I guess, two questions not to ask 6 compound. You picked seven, and what would the difference have been if you picked one, or two, or five? 7 Oh, okay. I don't believe I tried those one 8 to five specific numbers, but the general principle to 9 choose this constraint is to -- at least in this case, 10 11 that trying to minimize the number of spreads, whatever 12 the constraints trying to, you know, reduce, to the extent that algorithm is actually capable of doing that. 13 14 So algorithm has multiple diagnostics that basically tells you whether -- you know, because if you make the 15 constraint too strong, obviously there wouldn't be any 16 primes -- or a very small number of primes that would be 17 18 able to satisfy that. So, you know, you reduce it to 19 the point where -- like, the algorithm's still 20 performing well. And the other thing is that there's 2.1 multiple constraints. So you have to, you know, reduce each one of them to the extent that still the algorithm 22 is performing well, based on the general diagnostics 23

Q So you wanted your algorithm to discourage



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that's available.

1	multi-splits and you felt that a constraint of seven
2	would accomplish that?
3	A Right. And in pushing to below that, I felt
4	that we would start impacting the efficiency of the
5	algorithm. So, that's the you know, that's the level
6	I chose.
7	Q And you didn't have any reason to believe that
8	the Kentucky General Assembly was drawing its plan with
9	a constraint to avoid multi-splits, at a level of seven,
10	did you?
11	A No. So my yeah no.
12	Q You said you didn't run it with a different
13	constraint level?
14	A I did run it with different values. I didn't,
15	you know, record every single one of them, but I settled
16	on these values and because I found that these values
17	are still maintaining the efficiency of algorithm, while
18	reducing these con splits, as much as possible.
19	Q And did you include that criterion because
20	Plaintiff's Counsel told you to?
21	A Which one?
22	Q The multi-split constraint?
23	A Yeah. So the interpretation of the section
24	33, I did (Inaudible) on Counsel.

Do you have an independent understanding of

Q

Т	what Kentucky law requires?
2	A No. I'm not a lawyer.
3	Q Have you ever read the case Jensen v. State
4	Board of Elections?
5	A No.
6	Q You'd said that you tried different constraint
7	levels and that the algorithm was running efficiently at
8	seven. What does the efficiency of the algorithm mean?
9	A Right. So algorithm can stuck if you increase
10	the strengths of the constraint too much. Because then
11	algorithm won't be able to find another plan that will
12	satisfy that constraint. So in this Markov chain Monte
13	Carlo and Sequential Monte Carlo literature, there are
14	set of diagnostics techniques that one can use to make
15	sure that algorithm are, you know, running efficiently.
16	Q So I know what you told your algorithm to
17	consider. You didn't instruct your algorithm to
18	consider race?
19	A No.
20	Q You didn't instruct your algorithm to consider
21	communities of interest?
22	A No.
23	Q You didn't instruct your algorithm to consider
24	where schools are?
25	A No.



Q	You didn't instruct your algorithm to consider	
where churches are?		
А	No.	
Q	You didn't instruct your algorithm to consider	
where ne	ighborhoods are?	
A	No. No, to the extent	
Q	You didn't	
A	Sorry. So no to the yeah. I didn't	
incorpora	ate those factors directly, but that doesn't	
necessar	ily mean that those, you know, say for example,	
neighborl	hoods (phonetic) aren't kept together because	
to, you l	know, to the extent the counties, for example,	
correspon	nds to neighborhoods. And to that extent, the	
simulate	d plans may have those characteristics. But I	
didn't d	irectly tell algorithm, keep this particular	
neighborl	hood together or, you know, churches, or schools	
in the co	ertain districts. No.	
Q	And you didn't instruct your algorithm to	
consider	the location of county seats?	
А	No.	
Q	You didn't instruct your algorithm to consider	
the major transportation corridors in this state?		
A	No.	
Q	And you didn't instruct your algorithm to	
consider	where natural boundaries are, like rivers or	

1	mountains?
2	A No. But to the extent that they might
3	coincide with, you know, county boundaries.
4	Q The county boundaries?
5	A Yeah. That's right.
6	Q And you didn't instruct your algorithm to
7	consider where incumbents or candidates live?
8	A No.
9	Q And you didn't instruct your algorithm to
10	consider or try to prevent double bunking?
11	A No.
12	Q And you didn't instruct your algorithm to
13	consider maintaining the continuity of representation?
14	A No.
15	Q And you didn't instruct your algorithm to
16	consider core retention of districts?
17	A No.
18	Q So not a single one of the simulations in your
19	ensemble considers any of the things we just talked
20	about?
21	A Not directly considers that.
22	Q Wouldn't you agree though, that those are all
23	well-established, traditional redistricting criteria?
24	A What do you mean by, "Traditional
25	redistricting criteria"?



Q That those are things courts have told us over
time, are reasonable for redistricters to consider when
enacting a plan?
A I don't want to say these are the set of
traditional redistricting criteria. I think in the
academic literature reached, you know, things like
population, quality, compactness are considered a
traditional redistricting criteria. Other things that
you've listed may or may not. I don't really wish to
express opinion on exactly what counts as traditional
redistricting criteria.
Q You can say, I don't know.
A Oh, okay. Okay. Well, I know about them, but
I don't want express opinions on whether they count as
traditional redistricting criteria.
Q But at the end of your simulation analysis
or at least the first part of it, you conclude that
House bill two makes three additional splits to counties
more than the average necessary in your ensemble. I'm
looking at the chart on page 9 of report.
A Right. Okay. You mean figure 1?
Q Yes.
A Okay. Right. So on average, a simulated plan
has, you know, 15 and enacted plan is 18. So the

difference is (Inaudible) --

Q We'll move on to the next step of your
analysis. You then went on to calculate the partisan
vote share. And you said, you used six statewide races
from 2019 in Kentucky, and two 2016, federal statewide
race. How did you weight those races?
A Equally.
Q So each of the six constitutional office races
are given the same weight so, is it a one-to-one or
did you
A One-to-one.
Q The presidential and US Senate race, those are
both statewide races. Are state legislative races,
statewide races?
A No.
Q Are presidential races and US Senate races
good predictors for legislative races state
legislative races?
A I haven't done analysis of Kentucky, you know,
election forecasting, so I don't know.
Q So it's not your expert opinion that those
races are good predictions, because you couldn't form
that opinion?
A I used them as a major of partisan you
know, partisanship at the precinct-level, as standard
(phonetic) in the academic literature.



Q So by selecting those races, you're assuming
that voting patterns and voting history don't change,
right?
A No.
Q So if someone who votes one way in the
presidential race, you assume votes the same in a Senate
race, the same in all six constitutional office races?
A No. I don't make that assumption. I'm using
them as a measure of partnership at the precinct-level.
It has nothing to do with the voting you know,
prediction of voting behavior.
Q But you would agree that voter preferences do
change?
A Yes yeah. A little. Yeah. They could
change.
Q And that the candidate quality could really
impact turnout or support for a particular candidate?
A Sure.
Q But you didn't consider candidate quality when
you were selecting your races?
A So I used all the statewide elections for
which I had the precinct-level results. So I did not
consider candidate characteristics.
Q And so you didn't consider the pertinent
races, when you were picking those particular returns to



1	look at?
2	A No.
3	Q And specifically with your the selected
4	state races that you chose, you didn't do anything to
5	account for the clear outlier of the gubernatorial race,
6	did you?
7	A No. That's the point of combining multiple
8	races. You don't want to rely on a particular race. And
9	so, by averaging all the different races, you tried to
10	get a good measure of partnership.
11	Q But you certainly noticed that for all the
12	other five state constitutional offices, Republican
13	candidates, won handily?
14	A I actually didn't even consult who won. I
15	took those election results, and took the average, and
16	this is standard practice.
17	Q You. So the races you chose didn't
18	contemplate at all that Matt Bevin ran a terrible
19	campaign?
20	A Nope. I didn't do that.
21	JUDGE WINGATE: Is that what you-all stipulated
22	for?
23	MR. MADDOX: We'll stipulate to that.
24	BY MS. BECKER:
25	Q So, but when you included Andy Beshear's vote



1	share in your calculation, you didn't consider that Matt
2	Bevin said, that teachers kill kids?
3	A No. I didn't even know about that, so
4	Q And you didn't know that he called teachers
5	thugs?
6	A No.
7	Q You didn't know that he threatened the
8	northern Kentucky population with a toll bridge?
9	A Oh, no.
LO	Q And you didn't know that he removed expanded
L1	public assistance to the Commonwealth?
L2	A No.
L3	Q You didn't know that large populations of the
L4	Republican Party disliked Matt Bevin?
L5	A No. I didn't know that.
L6	Q That. So you didn't account for any of that
L7	when you included Andy Beshear's high Democratic vote
L8	share in your calculation?
L9	A No. So the taking the, you know, average to -
20	- so that you try to get general measure of
21	partisanship, not specifically any candidate or any
22	race.
23	Q I want to look back at we could look up
24	here if you like, but this is on page 11 of your report.
25	When you were characterizing where the Democratic-lean



versus the Republican-lean breaks, you used the flat 50
line as the line of demarcation for that, right?
A Yeah. That's right.
Q But you don't have any reason to believe that
the statewide average vote share of Democrats at 50
percent is when Democrats and legislative races actually
start winning races?
A Right. So, this is it's you know, it's
just the average vote share across multiple elections
that I looked at. So this is not a prediction of what
might happen in the next election. This is just measure
of, you know, possibly for one way or another.
Q But I think what you said earlier, was that
you highlighted these particular elections because they
were the competitive ones. That suggests that the 50
percent line is important.
A Right. I mean, in order to identify you
know, it is a measure of partnership. So when the
measure is close to 50/50, those are districts that tend
to be competitive in the next elections as well.
Q But you have no reason to believe that's
actually true in Kentucky?
A Well, you know, general tendency in many
states is that these type of averaging past election

results tend to correlate with the, you know, future

117 election. Just to the extent that past election is correlated with the future election. I'm talking about Kentucky. Even Kentucky I -- but, you know, I haven't Α done analysis in that sense. Right. Yeah. 0 So, no --I don't have a specific analysis to show you Α that. And if the threshold that is appropriate is somewhere closer to 51, 52 or 53 percent, would that change your analysis? Well, it may change the -- well, it doesn't Α really change the analysis. The fact that those D76 to the D79 is an outlier. That fact is not changed. doesn't matter how dotted line moves up and down. the fact that those D76 to D79 red dots are below the simulated prime Democratic portion (phonetic), that fact won't change. In fact, the simulate -- you know, that box (Inaudible) and dots won't change. It just what's going to change is just the dotted line going up and down. So -- but D76 is not an outlier and D77 is not an outlier?



the visual inspection here. But -- yeah. Anyway, the -

Okay. Well -- yeah. Sorry. I'm just doing

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- most the D76 and D77, you know, whether you call
this outlier or not, vast majority of (Inaudible) primes
have higher than prior vote share (phonetic) for those
districts. So that factor won't change.

- Q But it also wouldn't change that you're predicting -- or guessing based off your vote share, that the ordered District 76 and below are clearly Republican districts?
- A Yeah. Each election may have some swings, right, as -- as you all know. You know, some elections, Democrats do better. In other elections, Republicans do better. But what's important is the relative difference between the red dots and the box bar (phonetic). And that won't change, even if there's a uniform swing.
- Q But if the dotted line moves to 51 percent median, District 77 and 78 are below that relevant line?
- A Right. If the dotted line moves to, you know, 52 percent -- and yes, those red line -- red dots become below the dotted line. But what I'm saying is that the fact that the enacted plan systematically deviates from the simulated plan, that fact won't change. Because remember simulation doesn't use election results at all. So it's the -- you know, when you evaluate. That's when the election results come in,
 - Q Do you know how many seats Republicans



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1	currently hold?
2	A In the House?
3	Q Yes.
4	A I don't remember exactly.
5	Q Do you know how many votes are needed to
6	override a veto?
7	A I don't remember exactly.
8	Q Do you know how many votes are needed to pass
9	a bill?
10	A Don't want to I don't like to, you know
11	yeah. I don't remember exactly, so
12	MS. BECKER: Need to do a little bit of setup.
13	JUDGE WINGATE: Guess we're going to get some
14	color copies. I'm glad to see that you-all have got
15	color copies. Because, you know, I told you, we
16	don't' have color copiers in the judiciary.
17	MR. MADDOX: You still don't have that copier,
18	Judge?
19	JUDGE WINGATE: Uh-uh. I even asked for one, I
20	said, you know, I need a color copier, I'm doing
21	this big case, you know. And they said they
22	laughed, and they said, they'll give you the color
23	copies.
24	MR. ABATE: Which one is this?
25	MS. BECKER: This is old congressional the



1	old Judge, if you got a color printer, Staples
2	won't be making a mint anymore.
3	JUDGE WINGATE: Okay. Now do I need both these
4	or is one for
5	MS. BECKER: Well, there's one from Morgan.
6	JUDGE WINGATE: Here's one from Morgan. Okay.
7	I think I've got two here. I've got two here. Do I
8	need two?
9	CLERK: There should be a
10	JUDGE WINGATE: Do I got the same ones?
11	CLERK: You've got the same ones.
12	JUDGE WINGATE: Thanks.
13	MS. BECKER: She said that, where this is,
14	might block the camera.
15	JUDGE WINGATE: Oh, got you.
16	MS. BECKER: Can you just say something, so she
17	can check the camera?
18	THE WITNESS: Hello.
19	MS. BECKER: We're not worried about the news.
20	We're worried about the record.
21	THE WITNESS: Oh, okay.
22	MS. BECKER: You're good on the record.
23	THE WITNESS: Okay.
24	BY MS. BECKER:
25	Q So, in your opinion summary, you say that



1	there are districts in Jefferson and Fayette County that
2	improperly adjoin Republican precincts to make seats
3	safer, right? You focus in on districts 33, 48, 88, and
4	45. I'm on page 13 of your report the beginning on
5	13.
6	A Yes. I focus on yeah 33 and 48 in
7	Jefferson. And mention a couple other districts as
8	well.
9	JUDGE WINGATE: It looks like the 48 needs to
10	go into Oldham also. Am I reading that right?
11	MS. BECKER: That so that is my point.
12	JUDGE WINGATE: And the 33 looks like it used
13	to go into Oldham, just not as much or more 33
14	is more into Oldham, right? Now, under the new
15	plan?
16	MR. MADDOX: I'll just leave them here.
17	BY MS. BECKER:
18	Q So, the Judge has beat me to the chase here,
19	but I want to look at District 48. So, this is the old
20	map. This is the map that was drawn in 2013. And so,
21	you can see District 48 here and District 48 over there.
22	They make the same cut into Oldham county. Are you
23	aware that on the new map, the only change here is one
24	precinct?
25	A No. I'm not aware.



1	Q And Judge also noted that District 33 has
2	always gone into Oldham County. Can you see that here,
3	as well as over there?
4	A I see that.
5	Q And you understand that District 36 gained
6	population. So this portion had to be taken up
7	somewhere. And you see that was done with District 33.
8	And for the first time you understand Shelby County
9	exceeded the population of an ideal district, and it had
10	to shed population.
11	MS. HINKLE: Are you asking him if he knows
12	that or asking him to accept it?
13	MS. BECKER: I'm asking him if he knows that.
14	A No.
15	BY MS. BECKER:
16	Q And you see that part of Shelby County was
17	attached to 33?
18	A Yeah. I see that. Yeah.
19	MS. BECKER: Can I use your Fayette County
20	insert?
21	MS. HINKLE: Sure.
22	BY MS. BECKER:
23	Q And so over here you said that District 88 has
24	been made more Republican by adding Scott County. Do
25	you see how white shaded that portion of Scott County



т	is? It's not a large portion of Republican voters.
2	A Oh, you mean on the right map?
3	Q Yes.
4	A Yeah. So that is showing that on average,
5	across simulated plan, those white areas would have
6	belonged to a more competitive district. It's not
7	showing that both Democrats and Republicans live there
8	necessarily. It's showing that particular area would
9	have belonged to more competitive districts, under the
10	simulated plan.
11	Q Okay. I want to look back at your CV.
12	A Okay.
13	Q Prior to 2012 [sic], you had not offered
14	expert testimony in any litigation.
15	A Prior to 20 what year did you?
16	Q 2021.
17	A Oh, yes. Correct.
18	Q And that includes partisan gerrymandering
19	litigation?
20	A Right. That's correct.
21	Q Earlier, Counsel asked you if you had ever
22	declined a job, and you'd indicated that you had. What
23	jobs had you declined, beyond the one where you were
24	already retained by the other side?
25	A I was yeah. I was asked by the lawyers



of the counsel representing New York Democrats for the
New York redistricting case, recently.
Q And you'd indicated that you declined that job
because you didn't think they'd be able to prove their
case?
A I didn't feel comfortable based on the
analysis I've done myself. I don't feel comfortable
proceeding with that case providing expert witness
case in that.
Q Had you declined any other jobs?
A Trying to remember. I don't think so. Oh,
but I don't want to yeah. I feel like I may have,
and I may not have, so because these, you know, these
are short conversations that happened, and I sort of
don't remember after that. So I may have, but not,
like, often.
Q Did you decline any work in the Maryland
redistricting litigation?
A I wasn't asked by approached by anyone in
the Maryland case.
Q okay. I want to talk about your algorithms
now. You introduced your MCMC algorithm in 2020, right?
A You mean are you talking about the specific
publication or?

Is that the first publication where you

Q

Yes.

introduced that algorithm?

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A So in academia there's often a huge lag between when you have a paper and then, you know, there's (Inaudible). But the Journal of Computational and Graphical Statistics paper, I think that's what you're referring to. Is that what you're referring to? I'm just trying to make sure it's 2020. It's not some other years. Yeah. That's right. So, the Journal of Computational and Graphical Statistics. Yeah. 2020 is the publication year. Yes.

Q Are you looking at number --

JUDGE WINGATE: Do what?

CLERK: I think something's wrong with the system. Hold on.

MS. BECKER: Oh, yeah. It's going red.

JUDGE WINGATE: What did you just say?

CLERK: Give me a second.

JUDGE WINGATE: We've never had that happen.

CLERK: I don't know if it's, like,

overheating. It's making a loud noise. I can call

Amy if you want, or you can just pray it's

recording.

JUDGE WINGATE: Why don't you -- we'll just

keep it because it's lights are still on. Would you

just step out and talk to Amy, and see if she knows



1	what it is?
2	CLERK: Yeah.
3	JUDGE WINGATE: I've never had that beep happen
4	in 23 years. Okay. You can continue whenever you
5	want.
6	MS. BECKER: I'm sorry, Judge. Did you say
7	that we could go, or no?
8	JUDGE WINGATE: Well, let's just wait. Let's
9	do this. Let's stop and just wait. Might be better
10	to wait and talk to Amy.
11	(OFF THE RECORD)
12	JUDGE WINGATE: not having yet. Okay. Okay.
13	You may continue.
14	MS. BECKER: Does it matter that it's still
15	red?
16	JUDGE WINGATE: What?
17	CLERK: We want it to be red.
18	MS. BECKER: Oh, you want it to be red? Okay.
19	CLERK: If it's yellow, it's bad. Or flashing.
20	BY MS. BECKER:
21	Q I think when we stopped, we were talking about
22	on your CV, page 4, the article you have cited as number
23	12. We were talking about that as the first article,
24	where you introduced your MCMC algorithm in 2020?
25	A That's right.



Q And that wasn't the only article you wrote
about the MCMC algorithm you were working on. You wrote
another one, right?
A Yes.
Q I'd like to hand you a copy of that.
MS BECKER: And Judge, I think we're on
Commonwealth Exhibit 5.
JUDGE WINGATE: Uh-huh, that's the one you're
on.
BY MR. BECKER:
Q Professor, are you familiar with this article?
A Yes.
Q Did you write this article?
A Yes. With collaborators.
Q I'm sorry?
A With collaborators. Yes.
Q And it was published in the Journal of
Computational and Graphical Statistics?
A That's correct.
Q And it was published in 2020, sometime early
in that year?
A Yeah. I don't know exactly when, you know, it
was yeah. I don't know exactly when it was published
back during that year.
MS. BECKER: I'd like to admit this as an



1	exhibit, please.
2	JUDGE WINGATE: Okay. Let's how about we go
3	ahead and admit 1, 2, 3, 4 and 5 right now. Okay.
4	Is there any in objections, Michael, to any of
5	those, or Casey?
6	MS. HINKLE: No, Your Honor.
7	JUDGE WINGATE: Okay.
8	(COMMONWEALTH'S EXHIBIT 1 ADMITTED INTO
9	EVIDENCE)
10	(COMMONWEALTH'S EXHIBIT 2 ADMITTED INTO
11	EVIDENCE)
12	(COMMONWEALTH'S EXHIBIT 3 ADMITTED INTO
13	EVIDENCE)
14	(COMMONWEALTH'S EXHIBIT 4 ADMITTED INTO
15	EVIDENCE)
16	(COMMONWEALTH'S EXHIBIT 5 ADMITTED INTO
17	EVIDENCE)
18	BY MS. BECKER:
19	Q And then so you continue to work on this
20	algorithm, you authored a second article, correct?
21	A Yes. I've written multiple papers.
22	Q But on this particular topic, the next article
23	you wrote was what you have at number 13, another 2020
24	article?
25	A Yes. That's that's correct.



1	Q Are you familiar with this article?
2	A Yes.
3	Q You wrote this article?
4	A Yes.
5	Q And it was published in the Journal of
6	Statistics and Public Policy?
7	A That's correct.
8	Q In 2020?
9	A That's correct.
10	MS. BECKER: Move to admit this as
11	Commonwealth's Exhibit 6.
12	JUDGE WINGATE: Okay. Any objection?
13	MS. HINKLE: No, Your Honor.
14	JUDGE WINGATE: Okay. It's admitted.
15	(COMMONWEALTH'S EXHIBIT 6 ADMITTED INTO
16	EVIDENCE)
17	BY MS. BECKER:
18	Q I notice that you wrote this article, as well
19	as the one before with a gentleman by the name of
20	Benjamin Fifield.
21	A Yes.
22	Q What does he do now?
23	A He's a data analytic analyst for ACLU.
24	Q These papers were peer-reviewed?
25	A That's correct.



And they were approved for publication? 1 Q That's correct. 2 Α Did you introduce your SMC algorithm in 2021? 3 Q Α 2020. 4 2020 was the first draft? 5 Q 6 Α Well, yes. So, these papers, you know, takes time to be appearing in print. So the publication date 7 does not necessarily correspond to when the method was 8 developed. 9 10 I want to make sure you heard my question. Q 11 asked about your SMC algorithm. 12 Α Right. SMC, I think the first draft was 2020. 13 Okay. I'd like to hand you a copy of your 0 14 working paper --Α 15 Okay. 16 0 -- for that algorithm. 17 Α Okay. 18 So this document, it says -- are you familiar Q 19 with this document? 20 Α Yes. 21 Did you write this? Q 22 Α Yes. And your first draft was in 2020? 23 Q 24 Α That's correct. 25 And it says, "This draft, August 10, 2021"? Q



1	A That's correct.
2	Q Was this article published in a journal?
3	A It's in the review process.
4	Q So it's in the peer-review process?
5	A That's correct.
6	Q So this is a working paper?
7	A That's correct.
8	MS BECKER: I'd like to move to admit this as a
9	Commonwealth's Exhibit 7.
10	JUDGE WINGATE: Okay. Do you have any
11	objection?
12	MS. HINKLE: No.
13	JUDGE WINGATE: Okay.
14	(COMMONWEALTH'S EXHIBIT 7 ADMITTED INTO
15	EVIDENCE)
16	BY MS. BECKER:
17	Q I want to talk about some of your statements
18	in this working paper. I'm looking on the first page in
19	the abstract.
20	A Okay.
21	Q Where it says, "For successful application."
22	I'm going to read it to you. It says, "For successful
23	application, sampling methods must scale to large maps
24	with many districts, incorporate realistic legal
25	constraints, and accurately and efficiently sample from



1	a selected target distribution. Unfortunately, most
2	existing methods struggle in at least one of these
3	areas." What, "Existing methods," were you talking
4	about?
5	A This is a general statement. So it's not
6	specific particular, you know, algorithm, per se.
7	Q You wrote this paper to address concerns with
8	the MCMC algorithms that were prevailing at the time,
9	right?
10	A That's correct.
11	Q And so you're saying here, that your MCMC
12	algorithm cannot in actuality sample from a specific
13	target distribution?
14	A I didn't say that.
15	Q You say, it suffers from one of these
16	weaknesses, correct?
17	A Yes. But that's different from saying to not
18	sample. So it's in the context of academic research,
19	we always try to improve the existing algorithms and,
20	you know, that's that's the context. So we always
21	want to, you know, include what's out there. That's why
22	we do research.
23	Q I want to read not the next sentence, but
24	the one after it. You write, "Because it samples
25	directly, the SMC algorithm can efficiently explore the



relevant space of redistricting plans better than the existing Markov chain Monte Carlo, MCMC algorithms, that yield dependent samples," is that a true statement? Statement is true, but it's not -- it's all Α relative, right? We trying to improve the performance of the existing algorithms. Will turn with me to the second page. 0 I'm looking at the fifth whole paragraph. Page 1, or page 2? Α It's labeled as page 1. Q Α Oh, page 1. But it's the second page of that document. Okay. Okay. Α You say, "MCMC algorithms can, in theory, sample from a specific target distribution and incorporate constraints through the use of an energy In practice however, existing algorithms struggle to mix and traverse through a highly complex space, making scalability difficult and accuracy hard to Some of these algorithms make proposals by prove. flipping precincts at the boundary of existing districts, and rendering it difficult to transition between points in the state space, especially as more constraints are imposed." Did I read that accurately?



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I think you did.

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1	Q Now, the articles that you cite
2	parenthetically here, do you cite your own article in
3	both of those statements?
4	A Yeah. I'm trying to prove myself, too, so
5	Q Can you turn with me to numbered page 3?
6	JUDGE WINGATE: Did you say page 30?
7	MS. BECKER: Yes.
8	Q About midway through this page, you refer to
9	another expert in your field, Wendy Tam Cho, and her
10	criticisms of the existing MCMC algorithms. And in
11	response, you write the third full paragraph. I'm going
12	to read just a portion of it. "First, the distributions
13	that some of these algorithms sample from are not made
14	explicit, leaving open the possibility that the
15	generated ensemble is systematically different from the
16	true set of all valid plans. Second, even when the
17	distribution is known, the MCMC algorithms used to
18	sample from it may be prohibitively slow to mix, and
19	cannot yield a representative sample." Did I read that
20	correctly?
21	A Yes. You did.
22	Q Would you turn with me to page 13 of this
23	article. I'm not going to read all of this page, but
24	what I would like to ask you is a question, this

summarizes what's going on here. So what I read on this

page, is that at least at two separate points, your MCMC algorithm failed to yield reliable results, correct?

- A What do you mean by, "Your MCMC algorithm"?
- Q So throughout this article where you're citing back to yourself, you call the comparison, "The state of the art MCMC algorithm." And in the second paragraph on page 13, you say, "The upper panel of figure 4A shows the resulting density estimates. While the target distribution is highly multimodal, there's a good agreement between the SMC sample and the reference distribution. In contrast, the MCMC samples fail to accurately capture the left tail of the distribution, and over sample certain values of the right tail," does it say that?
- A Yeah, so this MCMC algorithm that I used in this article is not the same as the one I developed. So it's something that's different by a different author. But, you know, these are comparisons of -- in the academic article, you know, validation exercises to see how challenging problems (phonetic) can be addressed, you know, efficiently by one method over another, so...
- Q The last sentence on this page reads, "In comparison, the MCMC algorithm was not able to sample accurately from this target distribution in 20,000 iterations." Did I read that correctly?



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A Yes. But this is not a general statement. So
it's in this particular example that are actually
somewhat contrived because here in these these
examples, you can actually enumerate the all possible
ways. So you can actually, you know know exact truth
true distribution is, and it's a very challenging
setup that we, you know, set these things up, so that
to see how these different algorithms perform in these
specific applications. So and I don't want to take
this sentence out of context and, you know, make it into
a general statement.

Q I want to look on page 14. It says in the first paragraph, four lines down, you write, "Since the merge split MCMC algorithm is not specifically designed to enforce this hard constraint, we do not present its results." In this paragraph where you're not presenting the results because the algorithm is not designed to enforce the hard constraint, isn't the hard constraint, a hard multi-split (phonetic) constraint?

A I'm sorry. Should I answer now?

Q Yes.

A So the multi-split constraint is not a hard constraint that I used in the Kentucky case. It's not a hard constraint. So this hard constraint I'm talking about in this article is just the total number of



counties being spread. So in SMC, you can actually turn 1 that into hard constraints, so that no simulated plans 2 3 have, you know, more than certain number of counties that are being split. But multi-split constraint is 4 5 actually a soft constraint that I used. 6 JUDGE WINGATE: Jill, I used to fine people \$50, and give it to your domestic violence group. 7 Remember that? 8 9 MS. ROBINSON: Yeah. I do. 10 JUDGE WINGATE: Sorry. MS. HINKLE: It's okay. 11 BY MS. BECKER: 12 One final question. Can you turn with me to 13 0 14 page 15? 15 Α Okay. 16 I'm looking at the last sentence of the section that's on this page. 17 IJh-huh. 18 Α 19 Q Says, "This implies that SMC is several times 20 more effective than the state of the art MCMC algorithm, 21 in terms of run time per effective sample. Although 22 additional study is warranted, our results suggest that the proposed algorithm may be substantially more 23 24 effective when applied to real world redistricting 25 problems." Did I read that accurately?



A Yes. You did.

Q And you stand by all the statements that you made in this working paper?

A I do.

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Q But you used your MCMC algorithm for your State House analysis?

Right. So, again, the -- which algorithm you Α use depends on, you know, what you are studying. So the reason why I used the MCMC algorithm for the state case is that it better handles certain types of constraints when the number of districts is large. And so the SMC -- because SMC builds the one district at a time, as opposed to MCMC algorithm, where you start with the redistricting. So the SMC is unable to see, you know, certain type of constraints that requires you to know the entire redistricting plan itself. So, you know, again, these are sort of statements that's applicable to the particular applications I have in -- in this specific paper. But, you know, again, I don't want to -- I don't want anyone to generalize this statement to any case out there in the world. You know, it has to be considered for both -- both types of algorithms.

Q I appreciate your explanation. But I asked you if you used your MCMC algorithm to do your State House analysis; yes or no?



1	A I did use MCMC algorithm for the State House.
2	Yes.
3	Q Okay. I want to go back really quickly and
4	talk about specifically districts 48 and 33 on the new
5	map. You would agree with me, Professor, that District
6	48 on that map is more compact than District 48 on this
7	map, right?
8	A Depends on how you measure compactness.
9	Q Just give it the eyeball test. You don't have
10	your computer.
11	A As a statistician, I don't do the eyeball
12	test.
13	Q I want to talk about we talked earlier
14	about the population growth in District 36 and in
15	District 58, and how that impacted District 33. Are you
16	familiar with the community of interest that this area
17	covers?
18	A I did not use community of interest in my
19	simulation algorithm.
20	Q That's not my question. I'm asking you, if
21	you're familiar with the community of interest in this
22	area?
23	A No.
24	Q So you're not aware that there's a fire
25	station district in Peewee Valley that crosses the



1	county line?
2	A No.
3	Q And you're not aware that this area is called
4	Peewee Valley, and there's a women's prison that's
5	actually located in Shelby County?
6	A No.
7	Q And that this area is intricately intertwined?
8	A No.
9	Q I want to go back to our discussion of Matt
10	Bevin.
11	A Okay.
12	Q So this is the 2019 gubernatorial race. Now
13	we talked about all the things that you didn't know
14	about him. Did you know that election was decided by
15	less than 5,000 votes?
16	A No.
17	Q And did you know that the other five elections
18	were decided by at least 100,000 votes difference?
19	A No.
20	Q Did you know that our Secretary of State ran
21	against a former Miss America? Well, she was the
22	winner, right? She won.
23	A I have no idea.
24	JUDGE WINGATE: Did you know that Matt Bevin
25	attacked a very well-known judge here in Franklin



1	County? During the break I saw Phillip (phonetic),
2	and he said to mention that.
3	MS. BECKER: I left that out of the list.
4	BY MS. BECKER:
5	Q You would agree that an election that's won by
6	5,000 votes is an outlier, compared to five other
7	elections, at the same time, won by over 100,000 votes?
8	A I mean, compared to those other elections, but
9	you know, many elections are close, just in general.
10	Q Okay. I want to switch topics and talk about
11	your congressional analysis.
12	A Okay.
13	Q Your report only mentions Franklin County,
14	right?
15	A That's correct.
16	Q But there are other counties that switched
17	districts in the new congressional plan, right?
18	A Right. But as I mentioned, I didn't consider
19	the previous map. So that's not part of the criteria.
20	Q So you were told to focus on Franklin County?
21	A What do you mean by, "Told"?
22	Q You didn't look at the old map, but you
23	acknowledged that other counties switched. Why focus in
24	on Franklin County?
25	A Oh, I see. Well, the Franklin County is



you know, if you look at the district one, it's the edge
of this lengthy district that you know, that's
comprising that District 1. So it's, you know, one
place upon instruction of Counsel, that's the part that
I focused upon.
Q So you didn't focus on Franklin County because
that's where the plaintiffs wanted to file suit?
A Oh, I wasn't aware of who filed suit about,
you know, where they are located or anything like that.
Q Do you know the compactness measure for the
other five districts in Senate Bill 3?
A I did look at it at some point. I don't have
it on top of my head.
Q Do you know the compactness measure for the
whole map?
A Oh, yeah. I looked at that as well, as I
think that's in the report in the appendix. Yeah. So
it's it's figure 10. So that's an overall
compactness score average of the plans. So the red line
is the enacted plan, and the histogram is the simulated
simulated plans. That's overall, not (Inaudible).
Q So you're saying figure 10 is the analysis of
the whole plan?
A Yeah. So this isn't (phonetic) entire
overall. Yeah. So the Polsby-Popper is an average



across districts. And the other measure the other
compactness measure is a plan-wide measure. So there's
no specific district level measures.
Q So what I see here, is that the enacted plan
falls right within the average range of your simulation,
for compactness of the plan as a whole?
A Right. So exactly. That's the point. So
on average, I made sure that compactness of the
simulated plan is similar to the enacted plan. So
that's that's by design. But what I showed is that
even if you keep the overall level of compactness the
same, District 1 is highly non-compact.
Q But you don't know the compactness measures of
the other five districts?
A I did look at it at some point. I didn't
include it in the figure, but I did look at the
compactness of other districts as well. And if you take
the average, it will be the red line. So some are more
compact to offset the un-compactness of the District 1.
Q It's true isn't it, that every other district
Q It's true isn't it, that every other district is more compact, over the last map?
is more compact, over the last map?



Polsby-Popper method, right?

A Yes. For this. But I also did the Reock,
which is actually more computationally intensive. That's
figure 13. That's only I did this only for District
13 sorry, District 1 in figure 13.
Q Is the Polsby-Popper measure built into your
SMC algorithm?
A No. And neither Reock, in this other measure
that's in figure 13. They're not part of the algorithm.
Q Have you ever heard of let me make sure I
get the names right. Have you ever heard of Nicholas
Stephanopoulos and Eric McGhee?
A Yes. Yes.
Q Are they well-respected experts in your field?
A Yes. Nick is my colleague at Harvard Law
School.
Q So you don't want to say anything bad about
him, is that what you're saying?
A He's a great scholar.
Q Have you ever read their work, "The Measure of
a Metric"?
A Yes. I am aware of that (Inaudible) paper.
Q So are you familiar with their statement in
that article, where they say, "Scholars have not
selected a gold standard among the metrics," he's
talking about the Measures of Compactness, "But rather



have managed to use them productively in research, by	
combining multiple measures and adjusting weights for	
each specific purpose"?	
A Okay. Well, I don't memorize what he wrote.	
If he says that in the article, that must be what he -	_
he meant.	
Q But you didn't use any other metric beyond	
primarily the Polsby-Popper and then Reock as a	
crosscheck. You didn't use any of the other standard	
available methods?	
A What other measures are you talking about?	
Q So, there's the Inverse Convex Hull.	
A Okay.	
Q The Schwartzberg method.	
A Okay.	
Q You didn't use either of those?	
A Yeah. No. But you know, the Polsby-Popper,	
to be to be fair is the most commonly used method.	
You know, obviously, compactness can be measured in	
different ways.	
Q But the Polsby-Popper method does not like	
sharp curves?	
A Right. So different measurements try to	
capture different aspects of compactness. That's	
correct.	



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Q So in a state where we have a lot of rivers that are winding, and mountains that don't respect straight lines, and bounded by a river, the Polsby-Popper might not be the most favorable method to use?

Α Well, one could debate the properness of different compactness measures. But one advantage of the simulation method is that -- you know, because you are comparing -- so it's difficult to say, well, because Polsby-Popper is 0.2, that's too low or too high. But one advantage that simulation method offers is that you're actually comparing with other alternative plans using the same exact measure. So you're holding the measurement constant, and then doing a comparison accounting for all the geographical features, and rules, and other things. So, you know, I do feel comfortable using a simulation method and doing a comparison based on Polsby-Popper or some other measures, whereas, you know, interpreting these numbers as you pointed out, as it is, might not be appropriate, depending on the state.

Q The vast majority, if not all of your congressional analysis, is premised on compactness, yes?

A Well, first half is compactness, and the second half is, you know, partisanship.

Q Can you tell me the balance of weight you gave



to your population equality constraint versus the weight
you gave to your compactness?
A Oh, okay. In these algorithms, population
constraint is a hard constraint. So when you specify
it, the algorithm will generate the simulated plans that
always satisfy the population constraint. So we'll
never exceed that threshold, whereas compactness is a
relative (phonetic) measure. It's not a dichotomy.
Q So when your algorithm is creating districts
in each of your simulations, it is forced to follow your
population constraint
A First. Yeah.
Q but in doing so, is guided by your
compactness measure?
A Yeah. That's one way to think about it.
Another way to think about it is to consider a set of
simulated plans that, you know, simulate consider a
set of alternative plans that satisfy population
constraint. And then among those consider compactness,
you know, try to say select give more weight to the
more compact districts, for example.
Q I want to talk about your population
constraint.
A Okay.



You're aware that some congressional plans try

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1	to observe a strict one person deviation, right?
2	A Yes. I am. I'm aware.
3	Q But you didn't adopt that requirement in this
4	case?
5	A That's correct.
6	Q Are you aware that the enacted plan Senate
7	Bill 3 does just that, it's within one person equal
8	population?
9	A I think I've seen that at one point.
10	Q But your constraint was a plus or minus 0.1
11	percent, which I think I heard you say earlier, is a
12	spread of about 700 to 800 people?
13	A That's right. That's maximum. So some plans
14	are much lower than that but that's the maximum allowed
15	deviation.
16	Q You'd agree, that when the law requires as
17	mere as practicable, that one person is objectively
18	better than 800?
19	A I'm not a lawyer, but one person is smaller
20	than 700 to 800.
21	Q I want to go back to where we started.
22	A Okay.
23	Q You'd agree, that if the vast majority of
24	plans that are generated by your simulation method,
25	using what you call neutral redistricting criteria would



1	produce the same seat share in the enacted plan, then
2	the conclusion that there's a partisan bias that is not
3	supported?
4	MS. HINKLE: Objection to form. I'm sorry. I
5	just didn't follow the question.
6	MS. BECKER: Sure.
7	BY MS. BECKER:
8	Q So if the vast majority of the plans in your
9	ensemble have the same basic seat share as the enacted
10	plans, it's not right to assume that there's been a
11	partisan bias?
12	A Well, it depends on what you mean by,
13	"Partisan bias," I suppose. Right. It's that's I
14	guess, the whole difficult thing. But yeah.
15	Q Now when we talked about your House
16	conclusions, I think we decided we agreed that your
17	simulations suggest that 76 districts should lean in
18	favor with the Republican Party?
19	MS. HINKLE: Objection. It's inconsistent with
20	prior testimony.
21	JUDGE WINGATE: Okay. What's your objection,
22	again? I didn't hear you. I don't hear very good.
23	MS. HINKLE: I'm sorry. I think that's
24	inconsistent with Dr. Imai's prior testimony. Of
25	course, he can explain if it is or isn't, but I



1	JUDGE WINGATE: Yeah. It's overruled. I think
2	your question was, the 76 districts that are
3	Republican?
4	MS. BECKER: Right.
5	JUDGE WINGATE: And then
6	MS. BECKER: Which I think we've established
7	that.
8	JUDGE WINGATE: Yeah. I think it's been
9	established, when you were pointing to the one map.
10	So you can ask your question to him, again.
11	MS. BECKER:
12	BY MS. BECKER:
13	Q Isn't it true that under your simulation
14	analysis, the Republican Party in Kentucky should expect
15	for 76 districts to lean in its favor?
16	A Oh, I see what you're trying to ask. Okay. So
17	this is average vote share of Democratic average vote
18	share. And if you are you thinking about the seat
19	share? Like, how many seats the Republicans would win,
20	given any direction?
21	Q Now, I know that you're not capable of
22	rendering that opinion. So I'm just asking if your
23	simulations suggest that 76 districts lean in favor of
24	the Republican Party?
	the Republican Party.



according to my simulation, you know, on average lean 1 towards Republican, if you used these, you know, average 2 vote share from the past directions that I used. JUDGE WINGATE: Well, the follow-up question to that, is it 76 districts in the new plan? Or are there more that lean Republican? Is that what your 7 follow-up here? I'd just like to establish, that MS. BECKER: using his own simulations and his data, that we all 9 agree that 76 districts lean Republican. 10 11 JUDGE WINGATE: Well, their question is does 12 81, or 83, or after the plan, is there any way that he can -- that he has a prediction for that? Does 13 14 that make any sense? So I think when you're asking is, 15 MS. BECKER: 16 is he capable of predicting whether the districts 77, 78, 79 or 80 go Republican in an actual 17 election? 18 19 JUDGE WINGATE: Or lean Republican in this new analysis. Where's the cutoff? Where's the 50/50? 20 I was trying to figure that. 2.1 THE WITNESS: Right. So that really depends on 22 23 each election, right? There's just always, you 24 know, swing from one direction to another, based on 25 a variety of factors, including candidate popularity



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1	and other factors. So what I was establishing in
2	the figure 3 is that the 76 district and the enacted
3	plan is much more Republican leaning than the vast
4	majority of simulated 76 districts.
5	JUDGE WINGATE: Okay.
6	BY MS. BECKER:
7	Q Question about your congressional analysis.
8	A Okay.
9	Q So, on page 18, you say, "Under these
10	simulated congressional plans, the Democratic vote share
11	for the district that contains Franklin County is 43.6
12	on average, correct?
13	A That's correct.
14	Q And if we're using a strict 50 line, that
15	district is not likely to lean Democratic?
16	A Again, this is a, you know, measure of
17	partisanship, based on the literally, average of past
18	direction vote share. So, you know, I don't want this
19	to interpreted as like a prediction of the future
20	election or anything like that. It's just a measure. In
21	the past elections, 43 percent of voters voted for
22	Democrat on average. But yeah. It's less than 50
23	percent of voters, if that's what you're saying
24	Q So using your simulation data, we should
25	reasonably expect in the Commonwealth to have a



1	congressional delegation of five Republicans and one
2	Democrat?
3	A Again, I'm not expressing any opinion on
4	likely, you know, number of Democrats or Republican
5	seats in the future actions. That's I didn't do
6	that. No. It's just
7	MS. BECKER: Judge, can I have a minute?
8	JUDGE WINGATE: Yeah.
9	MS. BECKER: Judge, we'll pass the witness
10	back.
11	JUDGE WINGATE: Okay. You got any follow-up?
12	MS. HINKLE: It's very brief, Your Honor.
13	JUDGE WINGATE: Okay.
14	REDIRECT EXAMINATION
15	BY MS. HINKLE:
16	Q Dr. Imai, you were asked some questions about
17	the use of the NCM type of algorithm for purposes of
18	analyzing the Kentucky State House map. Are you
19	confident that that was the right type of algorithm to
20	use for the task, to which you put the algorithm in this
21	instance?
22	A Yes. Otherwise, I wouldn't put it in my
23	report.
24	Q And have you used the MCMC type of algorithm
25	in any of your prior expert engagements?



A Yes. A ton before.
Q And had produced reports on the basis of the
MCMC algorithm or expressed opinions in court, based on
that type of algorithm?
A Yes.
Q And those opinions, to your knowledge, have
been accepted by the courts?
A Yes.
MS. HINKLE: Thank, Your Honor. Nothing
further.
JUDGE WINGATE: Okay. Do you got any follow-
up?
MS. BECKER: No, Judge.
JUDGE WINGATE: Okay. Can we release this
witness? You okay to release him? Okay. You can
try to make your flight or you can go to Buffalo
Trace, and stand in line with everybody else. Thank
you, Doctor. I appreciate you being here.
THE WITNESS: Thank you very much. Thank you.
JUDGE WINGATE: All right. Let's have the
lawyers up here to talk about where we're at. And
also, Heather, do you want to make these part of
your exhibit? Are they in your book?
MS. BECKER: Well, they're in the book. Yeah.
I just thought it'd be easier for you-all to look at



those. 1 2 JUDGE WINGATE: Okay. Good enough. Yeah. Good 3 enough. MS. BECKER: Sorry. Did I miss a logistical 4 5 question? JUDGE WINGATE: Yeah. We're just -- where are 6 7 we at, as far as witnesses? What do you want to do? MS. HINKLE: I think we would prefer to call 8 another witness today, just to be mindful of the 9 10 court's time tomorrow as well. We do have a number 11 of witnesses to get through tomorrow. 12 JUDGE WINGATE: Okay. Good enough. MS. BECKER: So, even if we don't finish the 13 14 witnesses, I think it might be appropriate to get 15 started. 16 JUDGE WINGATE: Okay. Who do you want to do today else? 17 18 MR. ABATE: We're going to give it (phonetic) 19 to Trey Hieneman --20 JUDGE WINGATE: Okay. MR. ABATE: -- with the Kentucky Democratic 2.1 22 Party. Sounds good. How long 23 JUDGE WINGATE: Uh-huh. 24 do you expect him to be? 25 MR. ABATE: Certainly not as long as Dr. Imai.



1	JUDGE WINGATE: Okay.
2	MR. ABATE: I would imagine we would definitely
3	finish his direct today
4	JUDGE WINGATE: Okay.
5	MR. ABATE: Depending on how long the Court
6	wants to go, but
7	JUDGE WINGATE: Yeah. Okay. All right. Let's
8	start him then. Okay. Thank you-all.
9	MR. ABATE: Thank you.
10	MS. HINKLE: We want to take down everything,
11	right?
12	MR. ABATE: Yes. I may end up wanting that in
13	Jefferson County
14	MS. BECKER: Okay.
15	MR. ABATE: now, perhaps.
16	CLERK: Hey Casey, when you move that again,
17	will you angle it more this way? Because it blocks
18	the witness, if Judge asks him questions on the
19	MS. HINKLE: Oh, sure.
20	MS. BECKER: board.
21	MS. HINKLE: Yeah. If you could just take them
22	down for a minute
23	MR. ABATE: You can put them down.
24	MS. HINKLE: Yeah.
25	MR. ABATE: Just leave Jefferson County.



1	MS. HINKLE: Feel free to boss us around. When	
2	we put it up when we're not always mindful of	
3	those angles.	
4	JUDGE WINGATE: You ready to go?	
5	MR. ABATE: Yes, sir. If the Court is ready.	
6	Plaintiffs will call Trey Hieneman as witness.	
7	JUDGE WINGATE: Okay. Mr. Hieneman, please	
8	raise your right hand. Do you swear or affirm the	
9	testimony you're about to give in this Court today	
10	is the truth and nothing but the truth?	
11	THE WITNESS: I do.	
12	JUDGE WINGATE: All right. You may be seated.	
13	DIRECT EXAMINATION	
14	BY MR. ABATE:	
15	Q Thank you. Could you please state your name	
16	for the record, please?	
17	A Trey Hieneman.	
18	Q Great. And Mr. Hieneman, what is your current	
19	position?	
20	A I am the political director for the Kentucky	
21	Democratic Party.	
22	Q How long have you held that job?	
23	A I began that job in March of 2019, so just	
24	over three years.	
25	Q Okay. What are your duties as political	



director for the KDP?

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A I have a number of roles. I maintain relationships with interested groups, and organizations, county parties. I am the campaign strategist in-house for several campaigns. I work closely with our legislative caucuses on issues like recruitment. And then obviously, with any legislative talking points and things like that they may need.

Q Can you describe for the Court, your educational background?

A Sure. I have a bachelor of arts in political science from the University of Kentucky. And a masters of arts in political management from The George Washington University.

Q Okay. Thank you. Can you tell us a little bit about your work history, kind of going backwards? Tell us how you got to your role you're in now?

A Sure. From 2009 to 2013, I was communications director in the Office of the House Majority Caucus Chair. Then from there, I went to work for an outside organization. Then in 2019, returned to the party to work for Governor Beshear's election in 2019.

Q Okay. And you have been since that time -- since you left that campaign, you've been employed with the KDP?



1	А	Correct.
2	Q	And have you been in the same role the whole
3	time?	
4	А	Yes.
5	Q	Okay. Mr. Hieneman, do you have any
6	experienc	e in putting together legislative maps, like
7	the ones	we've been looking at here?
8	А	In my experience, during my tenure with the
9	Legislati	ve Research Commission and in the Office of the
10	Majority	Caucus Chair, I advised legislators on
11	developin	g the redistricting maps in the previous cycle.
12	Q	And did you have any involvement in preparing
13	a proposal in this legislative cycle for a redistricting	
14	map?	
15	А	Yes. I worked with our legislative leadership
16	and advis	ed them on drafting of House Bill 191.
17	Q	So, House Bill 191 was the Democratic
18	proposal?	
19	А	Correct.
20	Q	And was that introduced in the legislature?
21	А	Yes. It was.
22	Q	But it obviously did not become law?
23	А	Correct.
24	Q	Okay. Tell us about your role in helping to
25	craft Hou	se Bill 191?



1	A Sure. I advised the members of the House
2	Democratic Caucus on what they needed to emphasize when
3	drawing a map. Obviously, there are constitutional
4	requirements, federal requirements that they needed to
5	emphasize, and I advised them on those.
6	Q Can you tell us what those criteria were that
7	you used in drawing the maps or advising the members on
8	how to draw maps?
9	A Certainly. So the first criteria is the
10	constitutional one person, one vote requirement of being
11	plus or minus 5 percent about the mean for all
12	districts. There's also the section for
13	Q I'm sorry.
14	A Oh, yes.
15	Q I just want to stop you there. 5 percent of
16	population?
17	A Of the average population size.
18	Q Okay.
19	A So, 4,500,000 roughly Kentuckians divided by
20	145,000, and then 5 percent plus or minus about that
21	mean, one way or another. There are also the Kentucky
22	constitutional requirements that I advised them on,
23	obviously, the minimum number of split counties. But
24	then we also emphasized the exact verbiage underneath
25	Section 33 of not pairing more than two counties

together, and no piece of a county to form a district.
Q Okay. And just so I understand, you were
looking at the language of Section 33?
A That's correct. Those were the paramount
criteria that we used.
Q Okay. We'll get into a little more detail in
a second, but let's sort of walk through some of these.
On the population variants, did House Bill 191 satisfy
the plus or minus 5 percent standard you referred to?
A It did. No district exceeded 5 percent or
105 percent of the mean, and no district was under 95
percent of the mean.
Q Okay. Do you recall how many counties House
Bill 191 split?
A House Bill 191 split 23 counties.
Q Okay. And we've heard some testimony that's
the minimum is
A That's correct.
Q Can you explain why that is? Because that's
not mathematically the minimum, correct?
A That's correct. The mathematical minimum is
actually 21 counties. There are 21 counties that have a
population over 105 percent of the mean, so they have to
be divided. However, because of geography in the
Jackson Purchase region, particularly around Calloway,



1	Marshall, and Trigg Counties, one of those has to be
2	split. Because there's no county that can be either
3	split beforehand or can be paired with any of those
4	counties to make a district. And the similar the
5	same is true in southeastern Kentucky around Bell,
6	Harlan, Perry, Letcher, that region as well. So two
7	additional splits on top of the 21, gets you to 23. And
8	that is the actual practical minimum.
9	Q So if you added two of those counties
10	together, it still wouldn't be big enough to get within
11	the five percent, is it?
12	A It would either be too small or too large.
13	Q Or too large. Okay. Okay. So you had to
14	split 23. So you also talked about I believe you
15	said, the total number of times counties were divided
16	A Correct.
17	Q was a metric that you looked up; is that
18	correct?
19	A That's correct. Minimizing that number or
20	that piece of a county that then could be added to
21	another time. So minimizing the number of times that a
22	county is actually divided.
23	Q Okay. And then the other factor you
24	considered, was how many times three or more counties
25	were aggregated together?



1	A That's correct.
2	Q Okay. Have you reviewed House Bill 2, which
3	became the enacted
4	A I did. Yes.
5	Q maps? And let's talk a little bit about
6	your analysis of HB 2. I just want to start with the
7	basics. What did you what data did you look at to
8	analyze HB 2?
9	A Sure. So I worked off of the files that were
10	produced by the Legislative Research Commission. So the
11	actual map itself, as well as the Shapefiles that the
12	GIS staff with the Legislative Research Commission
13	produced, to to make my analysis.
14	Q What is a, "Shapefile"?
15	A So a Shapefile is basically a computer file
16	that's generally used in redistricting, that
17	encapsulates a number of different things, the block
18	level, the polygon shapes that that lay out the
19	districts. That you can then import and export into
20	different programs to generate the maps.
21	Q Okay.
22	A And generate maps that have already been
23	generated by the same program.
24	Q And did you look at all the same factors of
25	HB- 2 that we just discussed for 191, things like



1	population variants, county splits, multi-split
2	counties, and districts with three or more counties?
3	A Yes. I did.
4	Q Did you prepare an affidavit in this case,
5	that documents your comparisons?
6	A Yes. I did.
7	MR. ABATE: Your Honor, I'm not offering this
8	as evidence, but I would like to show the witness
9	the affidavit
10	JUDGE WINGATE: That's fine.
11	MR. ABATE: which is also in the binder we
12	gave you. This is marked number well, that was
13	Exhibit 32 of our preliminary injunction motion
14	that exhibit sticker predated.
15	BY MR. ABATE:
16	Q Is this a copy of the affidavit that you
17	prepared, Mr. Hieneman?
18	A Yes. It is.
19	Q Okay. And if you look down towards the last
20	page, that's your signature?
21	A Yes. It is.
22	Q Okay. I want to walk through some of the
23	details of the comparison
24	MR. MADDOX: Your Honor, I object to the
25	procedure. I don't believe it's proper for a fact



1	witness to have a cheat sheet in front of him, as
2	he's testifying.
3	JUDGE WINGATE: I've had all kinds of fact
4	witnesses with cheat sheets.
5	MR. MADDOX: All right.
6	JUDGE WINGATE: So I'll allow it. But I'll
7	note your objection for the record.
8	MR. MADDOX: Thank you, Your Honor.
9	MR.ABATE: Your Honor, we'll move forward.
10	BY MR. ABATE:
11	Q So, Mr. Hieneman, I'd like to look and I'd
12	like to look at your comparison of HB 2 and HB 191 on
13	each of these metrics. Let's start with the one that
14	talks about the number of times counties are divided.
15	Can you remind us again, why it is that you counted the
16	total number of times the counties were split?
17	A So I counted those because by counting those
18	numbers of splits, you can also calculate the number of
19	times that a piece of a county is being used to add onto
20	another district.
21	Q And how did you actually determine the number
22	of you told us you counted the number of counties
23	that were split, and it was 23. But how did you
24	determine the number of total county splits? Excuse me.
25	A Sure. By analyzing those 23 counties, you can



1	see each time that that county make or has a piece of
2	a district. You know, using McCracken County, for
3	example, in
4	MR. ABATE: May I approach, Your Honor?
5	JUDGE WINGATE: Yes.
6	A in this this McCracken County, for
7	example, has parts of
8	MR. MADDOX: What are you showing him?
9	THE WITNESS: It's this map.
10	MR. ABATE: Your sorry.
11	THE WITNESS: I'm sorry, the House Bill 2.
12	MR. MADDOX: Thank you.
13	MR. ABATE: That's the map of House Bill 2 you
14	all provided to the prior witness.
15	A Using McCracken County, for example, it has
16	parts of four House districts, so you would say that
17	that is split three times.
18	BY MR. ABATE:
19	Q Okay. And so, you I mean, visually, you're
20	looking at the map?
21	A Correct.
22	Q Okay. And how many times did you determine
23	that HB 2 has split counties in this method?
24	A Under my calculation and by reviewing it
25	visually, House Bill 2 split the 23 counties, 80 times.



1	Q Did you do the same calculation for 191?
2	A I did.
3	Q And how many times did you determine that
4	House Bill 191 split counties?
5	A House Bill 191 split the same 23 counties, 60
6	times.
7	Q Okay. And both of those bills complied with
8	the plus or minus 5 percent population?
9	A That's correct.
L0	Q Okay. So the 80 total splits was not
L1	necessary to achieve population equality?
L2	A No. It was not.
L3	MR. MADDOX: Objection, Your Honor. I really
L4	don't want him to be (Inaudible) on this. That was
L5	a leading question, and he needs to ask him
L6	questions that the witness can answer.
L7	JUDGE WINGATE: Well, I think he can answer
L8	that. That's overruled. Okay?
L9	MR. ABATE: Understood, your Honor.
20	BY MR. ABATE:
21	Q I want to look at another metric that you
22	consider you've testified that you considered in
23	taking account in drafting 191. And that was the number
24	of times HB 2 took a portion of one county and joined it
25	to a neighboring county to form a district.



1	A Correct.
2	Q How many times did HB 2 do that?
3	A Under my count, that happened 45 times.
4	MR. ABATE: And Your Honor, I guess since the
5	affidavit is not coming in
6	Q could you briefly tell us the numbers of
7	the districts in which this happened?
8	A Sure. It is districts 1, 2, 3, 5, 6, 8, 10,
9	14, 16, 18, 19, 22, 26, 27, 33, 37, 39, 45, 48, 52, 55,
10	56, 61, 63, 69, 71, 73, 78, 80, 82, 83, 85, 86, 87, 88,
11	89, 90, 91, 92, 94, 95, 96, 97, 98, and 100.
12	Q Thank you for humoring me with that, Mr.
13	Hieneman. I won't make you list them all out, but did
14	you count the number of times that HB 191 did the same
15	thing, take a portion of one county and joined it to a
16	neighboring county?
17	A Yes.
18	Q And how many times was that?
19	A So under my count for House Bill 191, that
20	actually occurred 31 times.
21	Q Okay. And then finally, that affidavit that
22	you identified talked about the number of times that HB
23	2 created districts containing three or more counties.
24	How many times did HB 2 do that?
25	A House district or House Bill 2 did that 31



1	times.
2	Q Okay. How many times did House Bill 191 do
3	that?
4	A House Bill 191 did that 23 times.
5	Q Okay. Thank you. Mr. Hieneman, did you take
6	a look at specific cities within Kentucky to determine
7	how HB 2 and HB 191 treated them?
8	A I did.
9	MR. ABATE: Great. Your Honor, I would like to
10	show the witness an exhibit here with certain
11	MS. HINKLE: 191?
12	MR. ABATE: No. The side-by-sides. Sorry,
13	it'd be the last tab. And for opposing Counsel,
14	this was the last tab in the binder that we handed
15	you this morning and for the Court.
16	JUDGE WINGATE: Okay.
17	BY MR. ABATE:
18	Q Mr. Hieneman, I'm going to show you what we
19	what exhibit number are we up to here?
20	MS. HINKLE: 3.
21	Q 3. I'm going to mark this as Plaintiff's
22	Exhibit 3. Mr. Hieneman, have you seen this document
23	before, these images?
24	A Yes.
25	Q Can you tell us what they are?



A These are maps detailing side-by-side the
district layout across various cities, across the
Commonwealth. Between the 2013 map that was enacted by
the General Assembly House Bill 2, and then House Bill
191.
Q Did you create these images?
A I did create these.
Q Can you tell us how you did that?
A Sure. After uploading the Shapefiles into an
online program called Dave's Redistricting app, I was
able to isolate these individual cities, while
overlaying the district maps over top of them.
Q And were these the same images that you
provided to Councel for use with the legal pleadings
provided to Counsel for use with the legal pleadings
A Correct.
A Correct.
A Correct. Q prepared in this case?
A Correct. Q prepared in this case? MR. ABATE: Your Honor, I'd like to move the
A Correct. Q prepared in this case? MR. ABATE: Your Honor, I'd like to move the admission as Exhibit 3.
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1	Q Mr. Hieneman, I'd like to kind of walk through
2	some of these districts that you've singled out here.
3	And I'll just start with the first one on the first
4	page. And this is a map of Bowling Green. Can you tell
5	us what the three images on this page show? First
6	describe what they are, and then we'll talk about it.
7	A Sure. First off, this is the City of Bowling
8	Green overlayed three different times, first with the
9	district layouts for the 2013 map House Bill 2 and then
10	House Bill 191.
11	Q Okay. In the existing map for this is
12	district the City of Bowling Green, how is the city
13	divided up for districting under the 2013 map?
14	A So you can see there, by and large, District
15	20 encompasses downtown Bowling Green. And then
16	outlying districts that come in from other counties
17	because Warren County is one of those that has to be
18	divided, take out parts of the outskirts of the county -
19	- or of the city, I'm sorry.
20	Q Now, how does House Bill 2, the enacted map,
21	treat Bowling Green?
22	A Basically, it cracks the City of Bowling Green
23	right down the middle between District 17 and District
24	20, while leaving District 19 pretty much taking up that
25	that same portion.

1	Q And does that affect the political leanings of
2	the district?
3	A It does. Under the calculations that were
4	done through Dave's Redistricting app using a a I
5	would call it, a composite score of election results,
6	over different cycles. I believe, it's the 2012
7	presidential, the 2016 presidential, the 2016 US Senate,
8	the 2019 gubernatorial and the 2019 attorney general. It
9	aggregates the different partisanships performance of
10	these. And what it showed is that under House Bill 2,
11	District 20 would go from being a Democratic performing
12	district as it is under 2013, to being almost 10 percent
13	more Republican leaning, and moving outside of the City
14	of Bowling Green. Whereas, it was overpopulated
15	following the 2020 Census, so it really needed to
16	condense. But instead of condensing, it was shifted.
17	Q So you said currently, there was a who
18	represents currently the district?
19	A State Representative Patti Minter.
20	Q You said there was population growth in
21	Bowling Green?
22	A Correct.
23	Q And what is likely to happen to the
24	representation there, under HB 2?
25	A You will have, with the three districts that -



- that are encompassing, this city, you'll have three
Republican representatives.
Q What would how would HB 191 have treated
Bowling Green?
A It kept District 20 almost wholly, again,
within the City of Bowling Green, condensing it, again,
because it exceeded the population that's allowable. And
keeping it wholly intact within Bowling Green, as it has
been for decades.
Q Let's move onto the next page if we could. And
I want to ask you about Covington. How does the
existing map, under which the current legislature was
elected treat the City of Covington?
A So, District 65 encompasses most of downtown
Covington. But as you can see, it kind of has an
annexed tail that comes down with it. And really House
District 65 makes up most of the City of Covington, with
District 64 coming in on the bottom.
Q And who currently represents that district
District 65?
A Democratic Representative Buddy Wheatley.
Q What does HB 2 do to the City of Covington?
A As you can see on the map there, it basically
splits the City of Covington into three different
pieces, and pushes what you don't see on this map,



unfortunately, is that it pushes District 65 outside of
the City of Covington, and deep into parts of Kenton
County that are not a similar community to downtown
Covington.
Q And as result of that, what do you expect will
happen?
A So a district that that is typically about
10 percent more Democratic would actually become about
10 percent more Republican.
Q So, unlikely to be any Democratic
representative
A Correct.
Q representation?
A You would have three Republican
representatives for Covington.
Q How would HB 191 have treated Covington?
A As you can see, it still splits the city, but
it keeps the downtown portion, particularly there in the
concentrated part along the river, wholly within
District 65, as it has been for decades.
Q And based on your analysis, do you know what
the political leaning of the District 65 would've been
under 191?
A I don't right off hand. I believe, we kept it
significantly more Democratic, 53 percent.

1	Q It more closely resembles the existing
2	district?
3	A Correct.
4	Q Can we talk about Erlanger nearby, staying in
5	Northern Kentucky?
6	A Sure.
7	Q How has that district historically been
8	well, how is it, under the current map, treated?
9	A You can see under the current map, it is
10	divided into three different representative districts,
11	the bulk of it being in District 69, but the tail end
12	there on the south part being in District 63 and 64.
13	Q How does HB 2 treat the City of Erlanger?
14	A It actually cracks it even further by pushing
15	63 more into the northern part, and continuing to split
16	it three ways.
17	Q Is that a significant change?
18	MR. MADDOX: Objection, your Honor. It's
19	purely a matter of speculation and opinion.
20	JUDGE WINGATE: Well, I think he can answer it,
21	if he knows. Or he can say, no, it's not
22	significant, or I think it's significant, or
23	MR. MADDOX: Maybe we don't need the answer.
24	JUDGE WINGATE: I think it's all right.
25	A So, the 69th House District has been



	110 111112, Gallett on 1-p2 2 35, 2022
1	represented by former Erlanger city councilman and has -
2	- this basically, is going to push that district less
3	out or more out of Erlanger, as it has been pretty
4	much the heart of Erlanger, as you can see in the 2013
5	map.
6	BY MR. ABATE:
7	Q And how would HB 191 have treated
8	A It kept the City of Erlanger almost entirely
9	whole. There's a very small fraction, because precinct
10	lines.
11	Q What about the City of Florence? How does the
12	current map the 2013 map sorry, when I say,
13	"Current," I'm referring to the map that the elected the
14	current legislature, as opposed to the one just enacted.
15	A You can see there again the bulk of it is
16	within one district, but it would certain pieces on
17	the outskirts coming in there, to take up the rest of
18	the population. But by and large, it is within one

whole district.

And how does HB 2 treat the City of Florence? Q

It splits it into pretty much two different Α districts, with a third taking on a third part in the northern part of the district -- or of the city, I'm sorry.

And the resulting districts under HB 2, what



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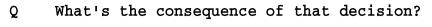
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	±,
1	did you determine about the
2	A They all become they all become
3	significantly more Republican performing.
4	Q Significantly more Republican? Well, what
5	about HB 191, how would that have treated the City of
6	Florence?
7	A You can see, it basically keeps into two
8	districts only, with the bulk of it being in District
9	85.
L0	Q I would ask you to turn to the next page,
L1	which includes maps of Georgetown. How does the 2013
L2	map treat Georgetown?
L3	A So the City of Georgetown, by and large, is
L4	within the 62nd House District. It is split, as you can
L5	see, between the 61st coming in from the north, and then
L6	the 78th, which actually comes around and gets a little
L7	bit there on the west side.
L8	Q How does HB 2 treat that?
L9	A Well, what you can see here, is that basically
20	House Bill 2 drives a spike right through the City of
21	Georgetown to where it's split. The northern part is
22	actually separated from the southern part by District
23	88.



A District 88, which has up to this point been



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wholly encased within Fayette County, becomes 1 significantly more Republican performing. And how would HB 191 have treated Georgetown? It has the City of Georgetown split into two pieces as well, but trying to maintain that continuity by putting most of it into District 56 and then the outlying portions into District 62. 8 Based on your calculations, how does that compare, in terms of political competitiveness? So you would have two --10 Α 11 Q To HB 2, excuse me. 12 You would have two Republican leaning Α Yeah. districts in House Bill 2. I would call House District 13 14 56 under House Bill 191 competitive, and District 62 being Republican. 15 16 If I could turn your attention to the next page, Mr. Hieneman, which is Hopkinsville. 17 18 (phonetic) Hopkinsville treated under the current 2013 19 maps? 20 So, for the most part, it is contained within Α two districts. There are a couple of small precincts 2.1 there that are tied into District 4 as it comes in, but 22 for the most part it is contained mostly within 8, and 23



And how did HB 2 address the City of

Kentuckiana Reporters

P.O. Box 3983

Louisville, KY 40201

Q

then the rest is in 9.

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Hopkinsville?

A So, as you can see, it splits the City of Hopkinsville basically right down the middle. But what's most egregious about this is there are two precincts that are actually numbered, Walnut Street 1 and Walnut Street 2. They are the most Democratic performing precincts in Christian County. They're also overwhelmingly African American, and they are divided between House District 8 and House District 9.

Q How would HB 191 have treated Hopkinsville?

A House Bill 191 kept the City of Hopkinsville mostly within House District 8, while trying to maximize African American representation in that district, and making it almost 40 percent African American voting age population.

Q How were the Walnut Street precincts treated in --

A They were both wholly contained within House District 8.

Q Finally, I'm going to ask you to look at Richmond, which is the last page of this exhibit. How does the 2013 map treat the City of Richmond?

A The City of Richmond has been almost wholly contained within the city -- or the 81st House District for generations.



1	Q Is that a competitive district?
2	A It is. Aside from the 2018 or the 2020
3	elections, which had some extenuating circumstances, the
4	2018 and the 2016 elections were both decided by less
5	than 1 percent.
6	Q What does HB 2 do to the City of Richmond?
7	A So as you can see from House Bill 2 map there,
8	it actually splits the city into three different pieces,
9	particularly with House Districts 91 and 89 taking
10	pieces of the City of Richmond, and then tacking them
11	onto counties outside of Madison County.
12	Q What is the consequence of that, according to
13	your calculations?
13	<pre>your calculations? A This would go from probably if not the most</pre>
14	A This would go from probably if not the most
14 15	A This would go from probably if not the most competitive district in the state, historically over the
14 15 16	A This would go from probably if not the most competitive district in the state, historically over the past couple of elections, to having three solidly
14 15 16 17	A This would go from probably if not the most competitive district in the state, historically over the past couple of elections, to having three solidly Republican districts.
14 15 16 17 18	A This would go from probably if not the most competitive district in the state, historically over the past couple of elections, to having three solidly Republican districts. Q And how would House Bill 191 have treated
14 15 16 17 18	A This would go from probably if not the most competitive district in the state, historically over the past couple of elections, to having three solidly Republican districts. Q And how would House Bill 191 have treated Richmond?
14 15 16 17 18 19	A This would go from probably if not the most competitive district in the state, historically over the past couple of elections, to having three solidly Republican districts. Q And how would House Bill 191 have treated Richmond? A House Bill 191 kept the City of Richmond
14 15 16 17 18 19 20	A This would go from probably if not the most competitive district in the state, historically over the past couple of elections, to having three solidly Republican districts. Q And how would House Bill 191 have treated Richmond? A House Bill 191 kept the City of Richmond entirely within the 81st House District.



about a different subject. You mentioned -- well, how

did you strike that. Let me start over. When			
drawing 191, did you consider any factors of race in			
crafting the map?			
A We did.			
Q How so?			
A So obviously, the constitutional paramounts of			
plus or minus 5 percent about the mean minimizing the			
county splits, minimizing the number of splits sort of			
limiting that number of pieces of a county, and then			
limiting the number of three or more counties are			
articulated in Section 33. While we don't have any			
actual Voting Rights Act districts, we did make every			
effort to make sure that minority population and			
minority representation was maintained within this map.			
Q And I want to unpack that a little bit. Can			
you explain for the court what you mean when you say,			
"We don't have any Voting Rights Act districts"?			
A Sure. We don't have in Kentucky's State			
House, there are no districts that are protected under			
Section 2 of the Voting Rights Act for minority			
representation.			
Q Okay. But you considered these factors when			
drawing 191?			
A That's correct.			

Did you compare the results of the districts

Q

1	drawn und	er 191 to House Bill 2
2	А	Yes.
3	Q	on these metrics? And can we walk through
4	the diffe	rent kinds of metrics you looked at? What
5	kinds of	districts did you consider with racial
6	А	Sure. There were four. The first being a
7	considera	tion of majority, minority populations. So,
8	that	
9	Q	What does that mean?
10	А	So that is 50 percent or above 50 percent
11	of the vo	ting age population being non-White.
12	Q	Okay. So that doesn't could that be
13	multiple	races combining?
14	А	Correct.
15	Q	Okay. What other you said there were four.
16	А	There were.
17	Q	Can you name the second kind?
18	А	So yeah. The others so subsets of that,
19	the first	is a plurality Black district, because those
20	are distr	icts where the Black voting age population is
21	the plura	lity of the voting age population. It may not
22	necessari	ly constitute a majority. In fact, it didn't
23	in any of	these these districts, but it does
24	constitut	e a plurality. The third are what are called,
25	"Coalitio	n districts." That is where two minority



1	groups combine to exceed 50 percent of the voting age
2	population. And then the last one are what are called,
3	"Influence districts," and that is where a a minority
4	population in almost every instance actually, in
5	every instance, it's the Black voting age population
6	exceeds 20 percent, giving them an opportunity to
7	influence the election of a representative of their
8	choice.
9	Q How did House Bill 2 and House Bill 191
10	compare, in terms of majority minority districts?
11	A Both maps contained six majority minority
12	districts.
13	Q Okay. How did House Bill 2 and House Bill 191
14	compare, in terms of plurality Black districts?
15	A So if I recall, I believe, that House Bill 2,
16	contained five plurality Black districts, whereas House
17	Bill 191 contained four plurality Black districts.
18	JUDGE WINGATE: How many? I'm sorry.
19	MR. MADDOX: Sorry. The door was shut. Yeah.
20	Q Can you repeat that last answer, sir?
21	A Sure. House Bill 2, as I recall, contained
22	five plurality Black districts, and House Bill 191 had
23	four plurality Black districts.
24	JUDGE WINGATE: Okay.
25	Q How about how did the two maps compare, in

terms of coalition districts?

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A So, House Bill 2, as I'm recalling, had no coalition districts, that is no two combined minority groups formed a coalition to exceed 50 percent voting age population. Under House Bill 191, District 77 was a coalition between Hispanic and Black population exceeding 50 percent voting age population.

Q And on the final metric, which you talked about influence districts, how did the two maps compare?

A So, again, as I recall, I believe House Bill 2 contained three influence districts, whereas House Bill 191 contained five influence districts.

Q Okay. Thank you very much for those facts. I appreciate it. So I'm going to shift gears again, and I'm going to ask you a little bit about the impacts of House Bill 2 on the Kentucky Democratic Party, based on your work in the party and your role. First of all, were you in involved in candidate recruitment for the 2022 election cycle?

A Yes.

Q Tell us about your involvement in that process, please?

A Sure. I worked closely with the legislative leadership and their anointed member of their caucus who heads up recruitment to find leads, vet leads, make



contact with individuals to gauge interest, connect them to the legislators to gauge their level of interest in running for state representative.

- Q Did the passage of HB 2 impact candidate recruitment efforts for the party for the 2022 elections?
 - A Absolutely.
 - Q How so?

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A Coupling the uncertainty of what the map, you know, rolling out on a state holiday right at the end of the year, and the uncertainty that created, the biggest hurdle that it created was several districts where candidates had been recruited, and then were drawn out of their districts, to where now we have no candidate.

Q Can you name some specific examples?

A Sure. You know, House District 24 -- or House District 21, I should say. He's now running in 24. John Pennington was our recruited candidate in House District 21. He was drawn out into now District 24. District 81, Martina Jackson was our recruited candidate. She was drawn out in the cracking of the City of Richmond into District 91. District 36, Derek Penwell was our recruited candidate in that district, and he was drawn out to where we have no candidate District 36. In District 29, Suzanne Kugler was our recruited candidate.

1	She was drawn out and now we have no candidate in
2	District 29.
3	Q Do you know how many candidates overall are
4	running for office, in the State House, as Democrats?
5	A So, I believe of contested races, that number
6	is 56 57, actually.
7	Q How does that compare to past cycles?
8	A It's down almost 25 percent from 2020. I
9	believe, we had 77 contested races in 2020 to now we
10	have 57 potentially contested races in 2022.
11	Q How does HB 2 affect candidate recruitment?
12	A By changing the lines. Even within split
13	counties, it draws potential recruits out, who who
14	were qualified candidates that we had identified, and
15	potentially persuaded to run. You know, it by
16	changing significantly the performance of those
17	districts, it dissuades candidates from wanting to to
18	run in a district, where the results are predetermined.
19	Q So, is it harder to recruit candidates in a
20	more polarized manner?
21	A Yes. Yes.
22	Q How did HB 2 treat Democratic incumbent
23	legislators? Or let me ask more specifically, did it
24	pair any existing sets of incumbent legislators?



Democratic legislators were paired in

Α

Jefferson County. There were two sets in the 41st
District. State Representatives Mary Lou Marzian and
Josie Raymond were paired together, creating a brand new
-- with no incumbent still within Jefferson County. And
similarly, Representative McKenzie Cantrell and
Representative Lisa Willner were paired together. Again,
leaving an open district in Jefferson County, beside the
district where they drew the those lines.

Q What does it mean when a new district is open?

A There's no incumbent that can seek reelection there. So it means, that you have to significantly work to hold a district like that because incumbency provides a lot of leverage in -- in an election.

Q And what does it mean when two incumbents are paired against one another?

A It basically means -- yeah -- that either they run against each other, or one steps aside. And that happened in both of these instances.

Q Can you explain that, what has happened?

A Sure. Representative Mary Lou Marzian, after being in the legislature for, you know, decades, has decided to step down and not seek reelection. And then Representative Cantrell having only served, I believe, three terms, is now running for Court of Appeals because she didn't want to be in a race against a fellow

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Democratic incumbent.

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- Q Did HB 2 pair any Republicans against one another?
 - A It did, but those --
- Q I was going to say, are those equivalent to the way they pair Democrats?
- A They're not. Those parents come because of population losses, specifically Representative Lynn Bechler is paired into the 12th District now, and that comes because Hopkins County, having previously been split under the 2013 map, has to come back whole. And so by doing that, there's no incumbent within Hopkins County. So, Crittenden County, which cannot be split and where he is -- currently resides has to be paired into another district. Similarly, population losses in eastern Kentucky -- in southeastern Kentucky in particular, meant pairing incumbent Representatives Bobby MCCool and Norma McCormick. And that district -- Norma McCormick's district, District 93, was relocated to Fayette County in House Bill 2.
- Q As a result of these changes, are there areas of the state where it'll be harder to recruit candidates than in previous years?
- A It is. Again, because of changing of district lines, and the way that some of these cities in



particular have been changed, you know, we have areas of the state where having a voice is becoming significantly harder.

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A You know, we currently have no Democratic representative candidates in the Jackson Purchase region, and there's six districts there.

Q Okay. What does it mean for the Kentucky Democratic Party, if you have no candidates or no elected leaders in a particular region?

A Yeah. You have no one to carry your message and carry your banner, even though the party represents over a million Kentuckians across Kentucky.

Q What other impacts do you foresee for the party, if there are regions of the state with no elected representatives?

A It makes it harder to convince donors to support those candidates that do decide to run in those regions, because they feel that those are lost causes, and they don't feel that they are able to allocate their resources as effectively. And so, it means a lot more sweat equity, and hard work to A, convince those candidates to run, and B for those candidates to actually make a competitive stand.

Q Will it affect your volunteer base in any way?



1	A Definitely. By having the or not having
2	the ability to compete and you know, by taking a
3	district in Warren County that has traditionally been 6
4	or 7 percent more democratic than Republican and turning
5	it plus-10 Republican, that diminishes volunteer
6	enthusiasm, and the ability to generate support for
7	local candidates.
8	Q Why are local volunteers important to the
9	party?
10	A They are the grassroots. They are the people
11	who help do the day-to-day operations of a campaign, you
12	know, knocking on doors, stuffing envelopes, making
13	phone calls, sending text messages.
14	Q Will these you mentioned, like, a
15	fundraising, lack of local volunteers. Will that affect
16	the party's ability to run statewide races?
17	A It could. You know, we we saw in years
18	past that, you know, not having resources can diminish
19	the ability of down ticket candidates to compete in a
20	statewide election.
21	Q So, HB 191 would've given the Republicans a
22	super majority?
23	A Correct.
24	Q So why help us understand why the
25	difference between HB 2 and HB 191 matters to the party?

A Well, the first is the Constitution matters.
The exact verbiage in Section 33 matters. And by
excessively exceeding it, these districts are created
outside of
MR. MADDOX: Objection, Your Honor. This calls
for a legal conclusion. He's not a lawyer.
JUDGE WINGATE: I think he can answer. But I
can read Section 33. I've read it backwards and
forwards a 1,000 times.
BY MR. ABATE:
Q In addition to any legal reasons, why else
does the difference between 191 and HB 2 matter to the
party, from an operational perspective?
A Yeah. It you know, by there are shifts
that are occurring in this state. And by being able to
dilute the the Democratic votes in certain areas, it
makes it so that the next election is guaranteed. But
you're also guaranteeing elections to come, as opposed
to trying to make those competitive in the future. This
isn't House Bill 191, isn't about 2022. It's about
·
2030, and 2040, and beyond.
2030, and 2040, and beyond. O Are there a different number of competitive
Q Are there a different number of competitive



produced by the Dave's Redistricting app, I believe,

1	that the number actually is cut in half between House
2	Bill 2 and House Bill 191. It goes from 17 in House
3	Bill 191 to, I believe, nine in House Bill 2. And by
4	competitive, I define that as within 10 percent one way
5	or another. So 55 to 45, one way or another.
6	Q And why does it matter if there are fewer
7	competitive races across the state?
8	A It basically allows the most extreme
9	legislation to come out. There's no bipartisanship.
10	There's no compromise. It's we we have the votes,
11	we're going to do it our way, and our majority is safe
12	no matter what.
13	Q You mentioned earlier that HB 2 was released
14	on a holiday, I think. Did you or did the KDP have any
15	advanced knowledge of what the maps in HB 2 would be?
16	A No. No.
17	Q When did you first see HB 2?
18	A I saw House Bill 2 for the first time as a
19	screenshot on a Twitter account, on a map that was
20	placed on a wall.
21	Q Can you unpack that a little bit; when was
22	that?
23	A On December 30th, when the House majority
24	leadership unveiled House Bill 2 as their proposed plan.
25	They held a press conference and I believe, members of



1	the media were there, and they shared that on social
2	media. And that was the first time that I actually saw
3	the proposal from there. I requested the Shapefiles
4	after it had been filed, to begin an analysis of it.
5	Q Do you recall when you received the
6	Shapefiles?
7	A That would've been first day of session, so
8	what, January 3rd, 4th, somewhere around there.
9	Q And do you remember when the bill was passed?
10	A I believe session started on the 4th, so it
11	would've been the 8th, 9th. I can't it was the
12	Saturday of that week, I believe.
13	Q Was that a sufficient amount of time to
14	analyze HB 2?
15	A Not from a public perspective. I mean, you
16	know, the the bill was put up on the first day,
17	immediately moved through committee that week, and then
18	out of the chamber within the first couple of days.
19	Q Did the timing of that announcement, impact
20	your ability to recruit candidates in new districts?
21	A Yes. I mean, although, the filing deadline
22	was postponed, there was still a lot of uncertainty over
23	when when this bill would be enacted. Or, you know,
24	there was a bill filed that was enacted on, to continue

to move the primary -- to move the filing deadline. That

1	created a lot of uncertainty as well.
2	MR. ABATE: Give me one second to confer with
3	co-Counsel here, if I may.
4	THE WITNESS: Sure.
5	MR. ABATE: Your Honor, I don't have any more
6	direct questions. I guess, my one question would be
7	a procedural one, which is I had asked Mr. Hieneman
8	to look at this map. I don't know if we ever
9	introduced these as exhibits. Did you move that as
10	an exhibit with Dr. Imai?
11	MS. BECKER: Not the big ones, but the small
12	maps are in our binder.
13	MR. ABATE: Okay.
14	JUDGE WINGATE: Small maps are in there, too.
15	MR. ABATE: Okay. Great. Well, then I think
16	we can skip introducing the big map as a separate
17	exhibit. And for now, I will pass the witness.
18	JUDGE WINGATE: Why don't we take a little
19	break? How long do you think your cross will be?
20	MR. MADDOX: I could have an hour, Your Honor.
21	JUDGE WINGATE: I'd like to finish this guy.
22	MR. ABATE: I would, too.
23	JUDGE WINGATE: I'd like to finish him, so he -
24	- you know. Let's go ahead and take a ten-minute
25	break and then we'll go to about 5:30 today. Okay?



1	All right. Thank you all.
2	(OFF THE RECORD)
3	JUDGE WINGATE: There you go. Take your seat
4	again. You're still under oath.
5	MR. ABATE: Your Honor?
6	JUDGE WINGATE: Yes.
7	MR. ABATE: While he's sitting down, I have one
8	housekeeping item. We refer to the map for HB 191,
9	which is a tab in the binder that we gave you this
10	morning, but it was not introduced as an exhibit
11	yet. So we would just like to move to introduce that
12	as Plaintiff's Exhibit 4.
13	JUDGE WINGATE: Okay. That'll be HB 191. Okay.
14	We've got it in. If there's any objection
15	there's no objection.
16	MR. MADDOX: No objection.
17	(PLAINTIFF'S EXHIBIT 4 ADMITTED INTO
18	EVIDENCE)
19	JUDGE WINGATE: Okay. You may begin.
20	CROSS EXAMINATION
21	BY MR. MADDOX:
22	Q Good afternoon, Mr. Hieneman. My name is
23	Victor Maddox. We've never met before, have we?
24	A No.
25	Q So I represent the Commonwealth, along with



Ms. Becker and I'd like to ask you some questions about
your testimony today. First of all, you were involved
in the preparation of HB 1 in the 2012 regular session,
right? The law that was passed and signed by Governor
Beshear, but ruled unconstitutional by the Supreme
Court, right?
A I offered advice and opinion to legislators
who introduced and voted on the bill. Yes.
Q Right. You said that you, in your job for the
House at that time, you were part of your
responsibilities was working with the Shapefiles and
A At that time, no.
Q No?
A No.
Q Okay. What was your involvement?
A Again, I did work with the Maptitude software,
and helped legislators craft what districts that they
wanted.
Q Okay. And the legislatures, at that time,
there was what a 59 Democrat majority in the House?
A 58, 59, I believe.
Q And as I remember, the Republicans in the
Senate had the majority. And so the way HB 1 in 2012
came together, the respective Houses agreed to pass each
other's maps, right?



1	A I believe, that's correct.
2	Q Okay. Now you said a moment ago that one of
3	the reasons HB 191 matters is because the Constitution
4	matters, right?
5	A Yes.
6	Q I want to show you a close up of a portion of
7	the map, that was enacted into law by the Kentucky
8	General Assembly with Speaker Stumbo and Governor
9	Beshear, both members of the Kentucky Democratic Party,
LO	right?
L1	A Yes.
L2	Q Okay. And what you see here is a close up of
L3	what I called in the 2012 case, Fischer Four, "The
L4	Pulaski strip." You're familiar with that, aren't you,
L5	sir?
L6	A No. But I see what you're referencing. Yes.
L7	Q Okay. So District 80 on the right is
L8	Rockcastle County, and it connects with this narrow
L9	strip through Pulaski County that's sort of the
20	taupe-y color, to Casey County; you see that?
21	A Yes.
22	Q And do you see there where the Lincoln County,
23	Rockcastle County, Pulaski County lines sort of come
24	together?
25	A Yes.



1	Q That's like a little spec of land, where the
2	District 85 connecting Rockcastle County and Casey
3	County, actually connects Pulaski to Rockcastle; isn't
4	that right?
5	A Judging by the picture. Yes.
6	Q Okay. Do you
7	A I've never visited there
8	Q Right. But the Constitution in Section 33
9	requires counties to be contiguous, doesn't it?
LO	A Yes.
L1	Q Okay. Now, as a political operative, you're a
L2	director of a party, and you've worked with
L3	redistricting maps, and you're familiar with the
L4	Constitution, I believe, even though you're not a
L5	lawyer. Does that strike you as respecting the
L6	Constitution's continuity requirement?
L7	A I would say they're contiguous. I would say
L8	it's also more than two counties paired together.
L9	Q So you think that's unconstitutional?
20	A Yes.
21	Q Even though the Democrat Party in 2012 thought
22	it was perfectly fine, right?
23	A Yes.
24	Q Okay. When did the Democrat Party decide that
25	it was unconstitutional to connect more than two



counties?

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A Well, I have -- I'm not, nor have I ever been, a member of the Kentucky General Assembly. I didn't get to vote on this bill. I got to talk to members about what they wanted, and then help them figure out what they wanted to do on the software. I didn't get to actually draw the map for the individuals. I didn't get to make the strategic decisions.

Q And I appreciate that. But my question was, when did the Democrat Party of Kentucky decide that it was unconstitutional to join more than two counties in creating a map? Because it clearly wasn't their view in 2012, right?

A I would say that the members of the General Assembly decided on this map.

Q Okay. So the Kentucky Democrat Party didn't have a view on whether it was --

A I don't think they took an official position on redistricting at that time. Members of the party may have.

Q Okay. So my question again is, when did the Kentucky Democrat Party decide that it was unconstitutional to join more than two counties?

A I don't think the party has ever taken that official position.

1	Q That's their position in this lawsuit, isn't
2	it?
3	A It is the position that Section 33 requires
4	that, and that House Bill 191 requires that, and the
5	our challenge is that House Bill 2 did not do that.
6	MR. MADDOX: Okay. Now, Your Honor, I would
7	like to offer the map I've just shown as
8	Commonwealth Exhibit Number 8, I believe.
9	JUDGE WINGATE: Yes. It's number 8.
10	MR. MADDOX: And offer it into evidence.
11	JUDGE WINGATE: You got any objection? Anybody
12	going to object?
13	MR. ABATE: No.
14	JUDGE WINGATE: Okay. It comes in.
15	(COMMONWEALTH'S EXHIBIT 8 ADMITTED INTO
16	EVIDENCE)
17	BY MR. MADDOX:
18	Q So you talked about there being 23 county
19	splits that are required, right?
20	A Yes.
21	Q And into the current population numbers, that
22	23 counties have to be split. Now you remember I
23	believe, you would remember that in 2012, that number
24	was 24 counties, right?
25	A I believe, that's correct. Hopkins and



1	Henderson were required to be divided, and Shelby was
2	not. So, yes.
3	Q Right. So do you remember that on
4	January 12, 2012, Speaker Stumbo in the floor of the
5	House said the following, when told that it would be
6	unconstitutional to divide 28 counties, because that's
7	what HB 1 in 2012 did, right? It divided 28 counties,
8	right?
9	A I honestly don't remember, but I'll take your
10	word for it.
11	Q Okay. I think the record will show that.
12	Mr. Stumbo said, when told that the law required and the
13	Constitution required that the minimum number of
14	counties be split and that was 24. He said
15	MR. MADDOX: and this is available, Your
16	Honor, at the KET website. We can introduce that
17	for the record, or you can take judicial notice of
18	it.
19	BY MR. MADDOX:
20	Q At 81 minutes I believe it's at the 81.13
21	mark on the tape, he says, "I would submit to you, that
22	the word possible means what you can get past, and what
23	you can get done in light of all the circumstances." And
24	a few seconds later, he said, "My interpretation of that



is, possible means what you can pass in light of the

1	spirit of the document. What you can pass that makes	
2	sense in the modern world. What you can pass through	
3	this body and the Senate and get signed into law." So,	
4	Speaker Stumbo, at least, high ranking Republican or	
5	Democrat who was involved in passing HB 1 in 2012	
6	essentially said, that the minimum number of counties	
7	that needed to be divided, Section 33 notwithstanding,	
8	was a political proposition, right?	
9	A I don't	
10	Q The minimum number of counties that you can	
11	get passed	
12	A If that was his opinion, then that was his	
13	opinion.	
14	Q And that's what the majority Democrat Party	
15	did in the House, right? They passed that law.	
16	A They passed it. Yes.	
17	Q Okay. I asked you earlier	
18	JUDGE WINGATE: Vic, regarding that tape,	
19	sometimes you just put it on a disc, and put a	
20	exhibit sticker on it, and we'll enter it in the	
21	next couple days.	
22	MR. MADDOX: Yeah. Your Honor, what I might	



Supreme Court in the 2012 case -- the Fischer court

also do is, I read from that -- that was in the

brief that Representative Fischer filed in the

23

24

1 case. JUDGE WINGATE: Okay. Well, then that would be 2 3 good enough. MR. MADDOX: I can offer that as an exhibit 4 5 here. I only have one copy, but I don't mind 6 introducing that into the record. JUDGE WINGATE: Yeah. That's --MR. MADDOX: We do have copies. I'm sorry. 8 9 We got copies of it, don't we? JUDGE WINGATE: 10 MR. MADDOX: We have extra copies here. 11 JUDGE WINGATE: Okay. Yeah. I would 12 appreciate that. MR. MADDOX: I would offer that as Exhibit 9. 13 14 JUDGE WINGATE: Hey, I'm doing this for the Supreme Court. I'm making a record, you-all. You 15 16 all understand that don't you? MR. MADDOX: So, I would offer that as 17 Commonwealth Exhibit Number 9. 18 19 JUDGE WINGATE: It'd be 9. 20 MR. MADDOX: Your Honor, the brief filed by 2.1 Representative Fischer in Legislative Research Commission v. Joseph M. Fischer Supreme, Court case 22 number 2012-SC-0091. 23 24 JUDGE WINGATE: Okay. Do you-all got any 25 opposition to that?



1	MR. ABATE: We have no objections.
2	JUDGE WINGATE: Okay.
3	(COMMONWEALTH'S EXHIBIT 9 ADMITTED INTO
4	EVIDENCE)
5	MR. MADDOX: Your Honor, I read from page 12,
6	quoting Speaker Stumbo.
7	BY MR. MADDOX:
8	Q The other thing that I would offer and ask
9	Mr. Hieneman, since we have an Exhibit 9 in the record
10	now, is are you aware that in the spec of land there in
11	Pulaski County that connected it to Rockcastle, there
12	were five voters?
13	A No.
14	MR. MADDOX: Okay. Your Honor, I would submit
15	that Exhibit Number 9 will demonstrate that, on page
16	15 as well.
17	JUDGE WINGATE: Okay. Thank you.
18	BY MR. MADDOX:
19	Q Now, I was going to ask you, Mr. Hieneman, if
20	you're familiar with the case that the Supreme Court did
21	in the Jensen matter?
22	A I'm familiar with the name of the case.
23	Q Okay. Well, the Democrat Party is asking this
24	Court to rule as a matter of law, that you can't split a
25	county more than twice, and you can't or you have to



split the minimum number of counties multiple times, and
you can only join two counties to form a district; is
that right?
A As well as the plus and minus 5 percent about
the mean, and the (Inaudible) number of split counties.
Q Right. So, in the Fischer Four case, the
Supreme Court sort of addressed the question of how many
counties had to be divided. And it rejected the LRC's
position that they could divide 28 counties or as many
as were politically possible. But in footnote 17, they
said this, and I wondered if you and the Democrat Party
were aware of it because it's part of
MR. ABATE: Your Honor, I'd like to object. It
seems like Mr. Maddox is trying to re-argue the case
from the preliminary injunction, and he already
objected to Mr. Hieneman
JUDGE WINGATE: Well, I think he can ask him
these questions. And if he doesn't know, he doesn't
know.
MR. ABATE: All right.
MR. MADDOX: Thank you, Your Honor.
BY MR. MADDOX:
Q I'm asking you, Mr. Hieneman, because you're
representing the Kentucky Democrat Party, if the party

was aware of this statement by our Supreme Court, in

support of the argument that more counties than the minimum should be divided, that the legislature should have more discretion. The Court said, "The LRC notes that by dividing more counties than the mathematical minimum, larger portions of more populous counties would remain intact. We decline to address the LRC's assertion because this is essentially the same argument made and rejected in Jensen. The appellant there asked the Court to require division of the minimum number of counties, only after each county large enough had obtained a district. The Court rejected this argument, upholding a requirement articulated in Fischer Two, to divide the fewest counties mathematically possible."

A No.

Q So essentially, LRC there was suggesting the Supreme Court that the rule they wanted would allow fewer counties to be multiply split, right? That you wouldn't have to have so many smaller fragments. You could keep larger portions of counties together, when forming districts, right?

A Sure.

Q And the Supreme court said they wouldn't even consider that, (Inaudible) Jensen rejected it, right?

A I'm not an attorney. I haven't read the



1	Jensen case.
2	Q Okay. Let me go now to some of your other
3	testimony. Before I do that, as one of your
4	responsibilities, you said that you are responsible for
5	candidate recruitment for the Kentucky Democrat Party,
6	right?
7	A I work with the legislative leadership to
8	recruit candidates to help them recruit candidates.
9	Q Okay. So you work on recruitment, but you're
10	not in charge of recruitment; is that it?
11	A No.
12	Q I'm sorry. "No," means I'm wrong right or
13	wrong?
14	A No. I am not in charge of the recruitment of
15	state legislative candidates.
16	Q Thank you. But you are involved?
17	A Yes.
18	Q Okay. So, do you know generally, if the
19	Democrat Party's approach toward candidate recruitment
20	is to meet with people one on one in the communities and
21	in question, and go to the churches, or the unions, or
22	the schools, or wherever it might be where potential
23	candidates could be located, and try to actively
24	identify such candidates?
25	A However the legislative leaders choose to



recruit their candidates is their prerogative.

Q Okay. Are you aware of any program by the Kentucky Democrat Party, whereby you engaged a California consultant to send text messages to random people asking them if they wanted to run for the Kentucky House?

A The program was not random. But, yes. I am familiar with the program.

Q Can you explain it for the Court?

A It's a -- a camp -- or it's a program from a company that texts to registered Democrats with -- to cell phone numbers that have been appended to the voter file, that the Kentucky Democratic Party purchases. And texts those folks to engage them and ask their level of interest. There are errors that occur because cell phone numbers are not as easy to pair as landlines and addresses.

Q So you get phone numbers, and you have a California company send text messages, and then if whoever receives that text message expresses interest, the company follows up and somehow tries to entice this person at the other end of the phone to become a candidate; is that it?

A They share their information with us. They offer to communicate with them, and then pass that



1	information to us. And then the legislative leaders and	
2	the county party leadership can make follow-ups on that.	
3	Q How long has that program been in place; do	
4	you know?	
5	A Couple months. it started at the end of last	
6	year.	
7	Q So you never tested it in a previous election	
8	cycle?	
9	A No.	
10	Q And have you had any success with it in this	
11	one?	
12	A Yes.	
13	Q Okay. So how many people have you recruited	
14	that way?	
15	A I honestly don't know.	
16	Q But it's been a successful program?	
17	A We have identified successful recruits from	
18	that program.	
19	Q Okay. And that's not withstanding HB 2,	
20	correct?	
21	A Most of those came from counties that were not	
22	divided yet.	
23	Q Okay. So is it your concern that	
24	A They also came from local or they are also	
25	running for local and county offices as well.	



Q I see. Is it is the Kentucky Democrat Party's		
position, that it's the counties that have been divided		
that makes it more difficult for the party to recruit		
candidates? Is that your basic position?		
A Not necessarily. It's the realignment of		
counties with different counties to make districts that		
also compose problems.		
Q Okay. Do you agree that the candidate		
recruitment in an off-year election like this, between		
presidential elections, is more difficult for the party		
that holds the White House?		
A Electorally?		
Q Recruitment of candidates and election of		
those candidates to office?		
A Electorally, yes. I don't have evidence of		
recruitment, one way or another.		
Q Well, if it's more difficult to win office,		
does it make sense that it would be more difficult to		
get people to change what they're doing and become a		
candidate, if the likelihood of them winning is less?		
A I don't know.		
MR. ABATE: Objection to form.		
JUDGE WINGATE: You got to explain all these		
objection to forms.		
MR. ABATE: He		

1	JUDGE WINGATE: I've been doing it for 22	
2	years. I had a Chicago lawyer say it every single	
3	time there was a question asked, so	
4	MR. ABATE: Well, thankfully, Your Honor, we're	
5	not doing that.	
6	JUDGE WINGATE: Okay.	
7	MR. ABATE: Mr. Maddox just said it's more	
8	difficult to win elections in State House in an	
9	off-year, and there's no evidence of that in the	
LO	record.	
11	JUDGE WINGATE: The I think he was asking	
L2	him if he knows there was any evidence of that.	
L3	MR. MADDOX: I was.	
L4	JUDGE WINGATE: So you can ask him again if he	
L5	knows there's any yeah.	
L6	BY MR. MADDOX:	
L7	Q Do you know, sir?	
L8	A What's the question? I'm sorry. I	
L9	Q Do you know if it's more difficult for	
20	candidates to win office in off-year elections, when the	
21	party in the White House in an off-year election,	
22	when that party holds the White House?	
23	A For State House, I don't know.	
24	Q Okay. Do you agree, though, that in election	
25	years that are presidential election years, the quality	



of the candidate and the appeal of the candidate at the		
top of the ticket matters right on down the line, both		
federal and state offices, would you agree with that?		
A That the quality of candidate matters in a		
particular race?		
Q Yes, sir.		
A Yes.		
Q And that, if the presidential candidate is		
unappealing in a state like Kentucky, that it makes it		
more difficult for candidates of that same party to do		
well, doesn't it?		
A It can.		
Q Okay. So now, in 2016, Hillary Clinton was		
the Democrat's candidate in Kentucky, right?		
A Correct.		
Q And she basically called just about everybody		
in eastern Kentucky a deplorable, didn't she?		
MR. ABATE: Objection.		
A I disagree.		
Q She called		
A I'm a native to eastern Kentucky, and I didn't		
consider myself deplorable.		
Q Well, no. Because you weren't a Trump		
supporter, right? She called the Trump supporters a		
basket of deplorables, didn't she?		



1	A	But your question implies that there are only
2	Trump sup	porters in eastern Kentucky, and that's not
3	true.	
4	Q	No. I said, about half.
5	А	No. You said, the population of eastern
6	Kentucky.	
7	Q	I think the record's going to show that I said
8	she called about half	
9	А	Did she call them deplorable? Yes.
10	Q	Okay. And she said they were racist, didn't
11	she?	
12	А	That I don't know.
13	Q	She said they were xenophobic, they were
14	homophobi	c, they were Islamophobic. And it's really
15	hard to c	all somebody something worse than a racist in
16	21st Cent	ury America, isn't it?
17	А	I don't know.
18	Q	Okay. So does that seem like an appealing way
19	for a can	didate to sort of help Democrats win election
20	in Kentuc	ky in 2016?
21	А	By calling them deplorable?
22	Q	Yes.
23	A	No. It doesn't help.
24	Q	Okay. What happened to the Democrat Party in
25	the House	races in 2016?



1	A They lost the majority.
2	Q They lost 17 seats, didn't they?
3	A Correct.
4	Q So they didn't just lose the majority. They
5	went from 53 in the majority to, I think, 36 in the
6	minority?
7	A Correct.
8	Q So they went from a majority to being a super
9	minority, right?
10	A Correct.
11	Q And that was the first time in what, 95
12	years the Democrat Party had not controlled the State
13	House?
14	A I believe, that's correct.
15	Q Okay. Now, in 2018, I think it was kind of a
16	wash, right?
17	A I believe Democrats gained a couple seats.
18	But, yeah.
19	Q Okay. A couple of special elections involved,
20	weren't there?
21	A No. There were there were a couple of
22	Republican held seats that I think it went from 36 by
23	the start of the 2019 General Assembly, it was like 39.
24	Q And that's an interesting point, right?
25	Because 2018 was an off year, and the Democrat the



1	Republica	ns held the White House, right? And the
2	Republica	n Party lost a couple of seats in that off-year
3	election,	right?
4	А	Yes.
5	Q	Okay. So it suggests that I was right, when I
6	said earl	ier that it's more difficult for can
7	А	One election doesn't create a trend.
8	Q	Right. So the third election in the last
9	three cyc	les, what happened?
10	А	Democrats lost seat in the State House.
11	Q	Lost how many?
12	A	They lost seats.
13	Q	They lost 11 seats, didn't they?
14	A	I believe, that's correct.
15	Q	So, they went to 75/25 in the minority, right?
16	A	I believe, that's correct.
17	Q	Okay. So now we've got three elections in a
18	row. And	if you draw a line from the 2016 to the 2018
19	to the 20	20, does that demonstrate a trend?
20	A	I what's the trend?
21	Q	Well, the trend is, the Democrat Party lost
22	over 52 p	ercent of its House memberships, in that
23	three-yea	r period, right?
24	A	In two elections.
25	Q	Three elections.
	l	



Well, but there's a blip up in '18. 1 Α Right. And so when you draw a line through 2 Q three points, and there's a blip up, but then the third 3 4 point is below where the first point is, that line is down, right? 5 6 Α Sure. And so all of that happened to the 7 Okay. 8 Democrat Party in Kentucky, without any regard to any 9 sort of mapmaking process, right? Α 10 Sure. 11 Q HB 2 didn't have a thing in the world to do 12 with it, did it? 13 Α No. 14 Okay. Now, you complained about the timing of 15 HB 2 in your testimony, right? 16 Α Yes. You said it was released on a holiday. 17 December 30th is not --18 19 Α It's a state holiday. 20 Very important difference, right? You didn't Q 21 And I said, that's a state holiday. 22 Α 23 Okay. S, for the most part of people in the 24 state, it's not a holiday at all, is it? 25 Α No.

1 Q Okay. Were you working on the 30th? I was. 2 Α 3 Q Okay. So it wasn't a holiday for you either? 4 Α No. 5 Q Okay. Now, you understand, right, that the 6 General Assembly couldn't do any redistricting law for -- that led to HB 2, right, until the United States 7 8 Census Bureau released the census data, right? 9 Α Correct. 10 And typically, that happens early in the year. 11 So the census is in the aughts (phonetic), the '10s, the '20s, and typically it's early in the spring, right? 12 13 That data comes out by April, I think, right? 14 Α Correct. But last year was a COVID year, right? 15 0 16 Α Yes. So, when did the data come out? 17 Q Α I don't honestly remember the date that it 18 19 came out. 20 Q Wasn't it in September? I -- again, I don't remember. 2.1 Α 22 So the General Assembly had between the Q Okay. 23 date that data came out -- let's call it September 1st, 24 just to be general. And the end of the year to do a 25 bill, right?

1	A Bills can be filed, I believe, until the
2	middle of March.
3	Q Right. But they couldn't do a redistricting
4	bill until they knew what the population density was,
5	right? So, they only had that four-month window,
6	correct?
7	A But again, they could have filed a bill
8	significantly later, and shown it to the public.
9	Q Significantly later than what?
10	Q Then the first day of session, when they
11	introduced the bill.
12	Q Okay. So they could have filed it sooner, but
13	they only had a four-month window to actually come up
14	with a bill, and get it vetted internally, and do
15	whatever else they wanted to do, and then get it
16	introduced, right?
17	A Sure.
18	Q Are you suggesting to the court that the
19	Democrat Party when or the Democrats, when they
20	controlled the House, and in that 95-year period
21	beforehand, that they got their bills introduced much
22	sooner?
23	A No. No.
24	Q Okay. So, what happened with HB 2 is what's
25	happened traditionally in Kentucky, when redistricting



1	bills are introduced, right?
2	A I'll defer to you on that. I've only been
3	involved in one working for the LRC.
4	Q Okay. So here's another question. Your
5	the your party's position, the Democrat Party and the
6	plaintiff in this case position is, that the legislature
7	damaged the party because it held the bill, and didn't
8	introduce it until the first day of the session, right?
9	A I would say that that damaged the public.
10	Q But the public isn't a plaintiff here today.
11	The party is, right?
12	A And our complaint is about the
13	constitutionality of the bill.
14	Q Right. But you were complaining to Judge
15	Wingate about the timing of its release and
16	introduction. And you seem to say be saying that,
17	that timing made it more difficult for the Democrat
18	Party to recruit candidates?
19	A It did.
20	Q Right?
21	A Yes.
22	Q Okay. So now, Andy Beshear is a Democrat,
23	isn't he?
24	A Yes.
25	Q And he's the only man in the state who has the



power to call a special session to deal with
redistricting, right?
A Yes.
Q So if the Democrat Party did the Democrat
Party ask Governor Beshear to call a special session, so
that redistricting could be done before the regular
session, and the Democrats would then be better able to
recruit candidates?
A No.
Q Why didn't it call why didn't it ask the
governor to do that?
A I don't know.
Q Okay. Fair enough. Are you familiar with the
concept of political geography?
A As I need a little bit more information on
that.
Q Well, people tend to be partisans one way or
the other. They vote for one party or another. They're
registered as Democrats, or Republicans, or
Independents. They live in cities, or they live in
rural areas, or suburbs, that sort of thing. Are you
familiar with how that affects sort of electoral
success, and success recruiting candidates?
A Somewhat.
O Okay. Are you familiar with the proposition,



that I think is commonly accepted in political science,
that Democrats tend to concentrate in urban areas of
larger cities, and Republicans tend to locate in
suburban areas or in rural areas in small towns?
A I'm not aware of that as a political science
concept. No.
Q Okay. So, you haven't ever looked into the
concept of political geography and the sort of the
disadvantage in the redistricting process that the
Democrats often face, simply because they are
concentrated in urban districts?
A No. I haven't looked at that.
Q Okay. Do you have any information you can
share with the Court, on the extent to which the
migration within Kentucky in the last decade has tended
to further concentrate Democrats in urban areas and
disperse Republicans in other areas?
A I don't know about moving Democrats and
Republicans, but I am aware of demographic changes and
population movement, regardless of partisanship.
Q All right. Do you agree that where people
choose to live is an important factor in the partisan
makeup of House districts across Kentucky?
A I'm sorry. Repeat that.

Yeah. I asked you if you agree that the --

Q

where people choose to live is an important factor in
the partisan makeup of House districts across Kentucky?
A No.
Q So let me ask you about the House District 40,
for instance. Are you familiar with House District 40?
A Yes.
Q Where is that?
A Shively, roughly. It's in Jefferson County.
Q Yeah. I think it's more like Shawnee Park,
right? And you know where Shawnee Park is, don't you?
A I'm not overly familiar with too many of the
smaller cities within Jefferson County.
Q Okay. Well, let's see if we can take a quick
look at it. So I think you're right, Mr. Hieneman.
It is Shively. So I apologize. You've got a set of
maps there in front of you in that notebook.
MR. MADDOX: And Your Honor, this is the
stipulated notebook that we called Exhibit 1, I
believe.
Q If you look at tab number one, which is a map
of this is HB 2, I believe; is that right? HB2. You
see that?
A Yes.
Q So if you look at exhibit or at the
Jefferson County detail up in the top left corner of



that, you'll see that District 43 is there in the
that bend in the Ohio River, right on the northwest
corner of Jefferson County, right?
A Yes.
Q Okay. So do you know what the black
population of that district is?
A The population or the voting age population?
Q Voting age population.
A I'm not aware. I know that it is a
plurality-black district.
Q So 35 to 45 percent?
A I would say over 45 percent.
Q 45 percent black voting age population. And
that suggests, to a reasonable objective observer, that
the Democrat vote percent in that district is going to
be extremely high, doesn't it?
A Yes.
Q Okay. And if you look at that the
geography of that district, you really can't go west and
gain any population that might sort of change that,
right? Because there's a river there, and then there's
Indiana.
A Correct.
Q And you can't go north because there's a river
there, and then there's Indiana, right?



1 Α Correct. And if you go south, you're in the 42nd 2 Q 3 District. 42nd District is just like the 43rd District, 4 right? 5 Α Yes. 6 0 About 45 percent black, very high Democrat population, a partisan split. So you can't go to that 7 8 district and get anybody who might sort of reduce the 9 partisan makeup of that district, to make it less 10 Democrat, right? 11 Α Correct. So those two districts, and there's others 12 0 just like it, 40, 44 are no different. 13 Those are 14 basically districts where -- because Democrats chose to 15 live in the urban center. And the geography of the 16 state, with the river, and the boundaries, and the like, and the relationship of one Democrat district to 17 another, means it's very, very difficult for a map maker 18 19 to make that district one that's going to be less than 20 highly Democrat, right? 2.1 Α Yes. 22 Okay. And so to the extent that Kentucky has 23 a given statewide vote percentage for Democrats, a 24 partisan split, and to the extent that a big number of 25 those people are concentrated in just a few districts,



that suggests that the rest of the state is going to be
a whole lot more Republican, doesn't it?
A Yes.
Q And that's not because of HB 2 or the
mapmakers. That's because of the political geography of
Kentucky; isn't that right, sir?
A With due respect to those, it's also because
of redlining.
Q Okay. So redlining is a you're talking
about a practice whereby somehow people were forced to
live in these areas, right?
A It it's a historical practice of yeah,
limiting the expansion of minority communities.
Q Okay. And that you're saying that took
place in the west end of Jefferson County, right?
A I'm saying that has historically happened. Not
necessarily I don't know because I'm a historian and
I'm definitely not a Jefferson County native resident,
but that has typically been what has concentrated some
communities.
Q You don't believe the Republican Party has
been in
A No.
Q control of Jefferson County
A NO



1	Q in the last 50 years, do you?
2	A No.
3	Q So you're not suggesting that redlining has
4	got anything to do with the makeup of HB 2, do you?
5	A No.
6	Q Okay. You know, that's another interesting
7	point. You know, when I was a kid, I lived in the west
8	end of Louisville. I think I lived in what would now be
9	the 40th District. It was right off of Algonquin
10	Parkway. You know where that is?
11	A No.
12	Q Okay. Well, it's in the west end. And it was
13	a long time, I confess, but my parents moved us to a
14	different part of the state. When I came back to
15	Kentucky after law school, I lived in the 34th District.
16	And that's Mary Lou Marzian's district, right?
17	A She was elected in that, but no. She
18	wouldn't she doesn't live in there, under House Bill
19	2.
20	Q The 34th House district right now is Mary Lou
21	Marzian's district, isn't it?
22	A She was elected to that under the 2020 in
23	2020.
24	Q And she's been there for 30 years, right?
25	A Roughly, yes.



Q Well, about 20 of those years, she represented
me. And you know, she never once voted the way I wanted
her to. Not once. And so, you know what I did, I got
sick of it. And so, I moved to another district, right?
I moved to a district that Jerry Miller represents, in
the 36th District. And Jerry Miller is in a district
that's so Republican, that at least one year in the last
three cycles, he didn't even have an opponent.
MR. ABATE: Your Honor, is this testimony or is
this a question for the witness?
MR. MADDOX: So, it's leading to a question.
JUDGE WINGATE: I think he could answer his
questions.
MR. MADDOX: It's leading to a question.
MR. ABATE: I never heard the question.
MR. MADDOX: Thank you for encouraging me to
move on, though, Michael.
BY MR. MADDOX:
Q So the question is, sir: People have a choice
about where they want to live, don't they?
A Some do. Some don't have the economic means
to move.
Q Okay. And you're suggesting that some of the
people in the 40th, and the 42nd, and those districts
don't have a choice about where to move? And that's

A I'm not saying about them in particular. The
same can be said in eastern Kentucky as well.
Q Okay. But certainly, people in the 34th
District do, right? That's a well-off district, isn't
it?
A Again, they're it's socioeconomics I'm not
familiar with.
Q Have you ever been to the Highlands, the
Cherokee Triangle area?
A Is that where that is? I don't know.
Q Okay. So people can make decisions about
where they want to move to, right? And that affects the
political geography. People can self-segregate into
more partisan for their purposes or less partisan for
their purposes districts, right?
A Sure.
Q Okay. And when that happens, over a period of
years or decades, the legislature is compelled to deal
with the people they have in the counties they find
them, right? And they have to draw lines, and sometimes
those lines lead to an electoral disadvantage for one
party or another; isn't that right?
A It can. Yes.
Q Okay. So while I've got you, up here on the
big map



1	MR. MADDOX: and this is tab 11, Your Honor,
2	in Exhibit 1.
3	Q We have a copy of HB I think this is the
4	2013 map, right? So, this is the existing the
5	districts that were recently repealed, and where Mary
6	Lou Marzian was actually elected in, right? And you can
7	see 34th District is right here. That looks like a
8	reasonably compact district, doesn't it?
9	A Yes.
L0	Q Okay. And do you know if the shape of that
L1	district has basically changed at all, in the last 30
L2	years?
L3	A I don't know.
L4	Q Okay. But over here on 43, and 41, and 42,
L5	those districts looks like bacon strips running east to
L6	west, don't they?
L7	A They are long. Yes.
L8	Q Those are not compact districts, are they?
L9	A I don't know the compact scores from
20	Q Right.
21	A from a statisticians
22	Q But using the eyeball test, you can tell me,
23	can't you, that those districts are long and narrow
24	compared to the same districts in HB 2?
25	A 41 is completely relocated. So, yes. And



1	then 40 I would say 43 and 42 are more compact. Yes.
2	Q Okay. And do you know what the partisan
3	makeup of these districts is, in their current
4	configuration?
5	A I not right off hand.
6	Q Okay. Is it important for mapmakers to try to
7	make districts more compact?
8	A If at all possible.
9	Q Okay. And it's certainly possible in western
10	Jefferson County, because we have a map in front of you
11	with more compact districts, don't we?
12	A They are compact.
13	Q Okay. Let's move on. So your job, with
14	respect to HB 1, was basically to extract some data
15	with respect to this case, excuse me, was to extract
16	data about HB 191 and HB 2, and then sort of present
17	that so that the plaintiffs could incorporate that into
18	their complaint, right?
19	A Yes.
20	Q And you did that with the Dave's Redistricting
21	website, basically, right?
22	A Correct.
23	Q Okay. When you went to Dave's Redistricting
24	website, did you notice that it tells you a lot of
25	information about the various districts and the metrics

1	of the plan itself; did you notice that?
2	A Some. Yes.
3	Q Okay.
4	A I'm not familiar with and I don't I'm
5	not a statistician, so I don't know a lot of those
6	details.
7	MR. MADDOX: Yeah. I want show you can you
8	give the witness
9	MR. MAGERA: Absolutely.
10	BY MR. MADDOX:
11	Q I want to Mr. Magera is going to hand you a
12	printout of a page that deals with HB 191. And this
13	comes from the Dave's Redistricting website. You
14	recognize this as coming from Dave's, don't you?
15	A Yes.
16	MR. MADDOX: Okay. Your Honor, I would offer
17	this as Commonwealth's Exhibit 10.
18	JUDGE WINGATE: Okay. Is there any objection?
19	MR. ABATE: I've never seen this before.
20	MR. MADDOX: Well, let me go through it and
21	JUDGE WINGATE: Well, how about this? Why
22	don't you ask him about?
23	MR. MADDOX: I'll go through it.
24	JUDGE WINGATE: And then, at the end of it, you
25	can



1	MR. MADDOX: And then we'll offer it.
2	JUDGE WINGATE: Okay.
3	BY MR. MADDOX:
4	Q So, what I've marked as the Commonwealth's
5	Exhibit 10 for identification, is a printout of a
6	screenshot from Dave's Redistricting. You can see there
7	it says "HB 191" in the middle, right? Sort of
8	A Yes.
9	Q top. And over to the left-hand side, it
10	says, "Kentucky 46,018." It has a number of things you
11	can do. And then over on the right-hand side, it has,
12	"District details 47," and it has a lot of different
13	information about voting age, and population, and
14	partisan issues, right; you see that?
15	A Yes.
16	Q So you've seen these kinds of pages before in
17	your work, right?
18	A Yes.
19	Q Okay. And this relates to District 47 in
20	particular. You can see where it says on the bottom
21	there on the under, "Composite 2012, 2019 Democrat,
22	35," I think that's 0.7 or 0.9 percent. And,
23	"Republican, 60.9 percent"; you see that?
24	A Yes.
25	Q So it looks like Dave's has said, if we look



at some elections over the last eight or nine seven
years, this district tends to vote 35 to 40 percent
35 to 36 percent Democrat and about 64 percent
Republican, right?
A Correct.
Q Okay. Now, if you go look to the next
page, do you see where it says, "Efficiency gap," there?
On the left-hand side, under, "Metrics," extreme right -
- extreme left?
MR. ABATE: I only have one page.
JUDGE WINGATE: Yeah. I've only got one.
MR. MADDOX: Yeah. Sorry, Your Honor. Mine
are combined. Mr. Magera's going to hand out
JUDGE WINGATE: Got you.
MR. MADDOX: the second page. Alex, are you
handing out just one or the other two?
MR. MAGERA: Just one.
MR. MADDOX: Okay. We'll do it page by page.
And Your Honor, I would offer this as for
identification as well, can I just make it all
one exhibit? Is that
JUDGE WINGATE: Yeah. That's what I was going
to say. I'm going to staple mine. How's that?
MR. MADDOX: That's what we'll do. It's going
to be a three-page exhibit.

1	JUDGE WINGATE: Is it going to be a three-
2	page?
3	MR. MADDOX: Three page.
4	JUDGE WINGATE: Okay.
5	BY MR. MADDOX:
6	Q So, Mr. Hieneman, you can see there on the
7	second page of Exhibit 10 for identification, the
8	efficiency gap number. Do you see that?
9	A Yes.
LO	Q Okay. And it says, "Efficiency gap, 9.26." Do
L1	you know what that means?
L2	A I don't. I mean, other than the description
L3	that's written there. I'm not a statistician, I don't -
L4	- I'm not familiar with it.
L5	Q Okay. So we've seen in the record, in
L6	connection with earlier filings, that the efficiency gap
L7	number that number would suggest that the map is a
L8	pro-Republican map; do you agree? And if that's the
L9	case let me ask you to assume that for a moment. Can
20	you tell me why the Democrat Party sponsored a map that
21	became HB 191, that has a pro-Republican leaning and an
22	efficiency gap of 9.26?
23	A Again, I don't know what efficiency gap
24	percent means. What I advised the members of the
25	Democratic Caucus of the General Assembly was to draft a



map, that complied with the written text of Section 33.	
Q How many seats do you think would be elected	
by Republic be elected for Republicans under HB 191?	
A In '22 or in 2030?	
Q In the coming election?	
A It's I believe the estimates from this	
website, was 76, something like that. I do I	
honestly don't remember right off hand.	
Q So Dave's Redistricting, sort of an objective	
website, one that the Republican the Democrat	
Party, excuse me, used to help it put together its own	
map 191, says that map's going to result in 76	
Republicans, right?	
A After the 2022 election, and again, not	
assuming candidate quality or anything like that. Just	
based off statistical estimation.	
Q If you and the Democrat Party were able to	
draft a map that would've resulted in more Democrats	
being elected, would you have tried to do that?	
A What I advised the members of the legislative	
cauc of the Democratic Legislative Caucus to do was	
draft the map that was compliant with the written text	
of Section 33.	
Q Okay. And when you say, "The written text,"	
did you include the authoritative interpretation of that	

1	written text, that the Supreme Court has handed down?
2	A I'm not an attorney and did not have
3	interpretation of that. I could simply read the text of
4	the mandates in it.
5	Q But the Constitution is a legal document,
6	right?
7	A Yes.
8	Q And so to interpret it, if you you have to
9	rely on either some training or some authoritative
10	source, don't you?
11	A The authoritative source was the written words
12	of Section 33.
13	Q Okay. So you formed your own layperson's
14	opinion about the meaning of Section 33 when you advised
15	the Democrat Party on how to construct HB 191; is that
16	your testimony?
17	A I advise members of the General Assembly to do
18	so.
19	Q Even better. Okay. Did anybody in the
20	General Assembly ask you if there was a legal opinion
21	supporting that?
22	A No.
23	Q Who did you talk to specifically, in the
24	General Assembly?
25	A The legislative leadership.



1	Q And that was who?
2	A Representative Jenkins Joni Jenkins,
3	Representative Derrick Graham, Representative Angie
4	Hatton.
5	Q And not one of them asked you if your
6	layperson's interpretation was supported by a legal
7	opinion, from anybody with any legal training?
8	A They did not ask me that.
9	MR. MADDOX: Third page. Your Honor, I would
10	like to now turn to the third page of what I've
11	called Commonwealth's 10 for identification.
12	JUDGE WINGATE: Again, I'm making this one
13	exhibit, if that's all right?
14	MR. MADDOX: Yes. Thank you, Your Honor.
15	BY MR. MADDOX:
16	Q And so, Mr. Hieneman, the this document -
17	MR. MADDOX: Where's the precinct's one? With
18	the Okay. All right.
19	Q So the last line in this document, sir,
20	addresses the splitting of precincts. Do you see that?
21	A Yes.
22	Q Okay. And it says, to achieve almost exact
23	district population, 99 precincts may also have to be
24	split in but 24 are split. Do you understand that is
25	what HB 191 does?



1	A Yes.
2	Q Okay. Do you know how many precincts HB 2
3	splits?
4	A No.
5	Q Have you ever done anything to inform yourself
6	or the Democratic Party about whether HB 2 splits even
7	one precinct?
8	A I have not.
9	Q Do you think the legislature was entitled to
10	adopt a rule that says, we're not going to split any
11	precincts, when we redistrict the state?
12	A Are they entitled to that? If that is their
13	mapmaking principle, that's they're certainly
14	entitled to that.
15	Q Do you understand that there are good reasons
16	why you don't want to split precincts, when you're
17	creating legislative districts? For one, it saves the
18	county clerks a whole lot of time and money. Would you
19	agree with that?
20	A I've not worked in the clerk's office. I
21	don't know.
22	Q Okay. In any event, HB 191 splits at least 24
23	precincts, right?
24	A According to this. Yes.
25	Q And you don't have any reason to doubt it,



1	right?
2	A No.
3	Q Okay. So when you go to Dave's Redistricting
4	and that's what you did, right?
5	A Yes.
6	MR. MADDOX: Your Honor, I would like to offer
7	Exhibit 10, at this point 11 10 for
8	identification
9	
LO	JUDGE WINGATE: It's 10. Do you-all have any
L1	objections?
L2	MR. MADDOX: I mean, into evidence.
L3	MR. ABATE: No.
L4	JUDGE WINGATE: Okay. Go ahead then.
L5	(COMMONWEALTH'S EXHIBIT 10 ADMITTED INTO
L6	EVIDENCE)
L7	BY MR. MADDOX:
L8	Q All right. So when you go to Dave's
L9	Redistricting, Mr. Hieneman, it allows you to upload the
20	maps and the files that go into creating the maps that
21	are HB 2 and HB 191, right?
22	A Correct.
23	Q And you so you went to the LRC website, and
24	you can go to the link, and it says, "Here are the
25	Shapefiles files, right? And you got those Shapefiles,



1	and then you went to Dave's, and you uploaded the LRC
2	files to the Dave's website, right?
3	A I was given the files from LRC staff.
4	Q Okay. Even better. But all you had to do
5	then was take the LRC shape files, and put them into
6	Dave's, right? And then when you did that, what did
7	Dave's ask you and what did it do for you?
8	A I uploaded the map, and I guess, ran analysis.
9	Q Okay. And you don't have any training in
10	statistics, or quantitative analysis, or computational
11	science?
12	A No.
13	Q And you don't have any training in election
14	history or political science in the country or in
15	Kentucky, right?
16	A I have a degree in political science.
17	Q Oh, you do?
18	A Yes.
19	Q Okay. Is that the one you got at George
20	Washington?
21	A No. I have a master's in political management
22	from George Washington.
23	Q Okay. So, you have an undergraduate degree in
24	political science. You got that when?
25	A 2004.



1	Q Do you consider yourself an expert in
2	redistricting techniques?
3	A No.
4	Q Do you consider yourself an expert in computer
5	programming?
6	A No.
7	Q Okay. So after you uploaded the maps, you
8	asked Dave's to analyze it for you, and that's what it
9	did?
10	A It generates those automatically.
11	Q Okay. And then and that's where you got
12	the files, the maps for the different cities we looked
13	at earlier?
14	A Yes. You can isolate by city.
15	Q And it gave us it would've given you if
16	you'd asked for it and wanted to keep it all the data
17	that we called Exhibit 11, right?
18	A Yes.
19	Q Okay. 10, I'm sorry. Heather's keeping me
20	honest. And if I wanted to take those shame shape
21	files, or if Judge Wingate wanted to take those files,
22	he could do the same thing you did to generate
23	JUDGE WINGATE: Probably not. I can barely run
24	that computer.
25	MR. MADDOX: Well, I'm not far behind you,



Judge. 1 2 BY MR. MADDOX: But anybody generally can do this, right? 3 4 whole point is, it's set up so that anybody can do it, 5 right? 6 Α Yes. Okay. Now you'll notice on page 2 of Exhibit 7 Q 8 10, at the very bottom, it says, "Use plan score to 9 further assess the degree to which a map is 10 gerrymandered." Do you see that? 11 Α Yes. 12 And then it has a little tab you can click on, 13 and that'll take you to plan score, right? 14 Α I believe, I --You didn't do that? 15 0 16 Α No. Okay. So you don't know what the plan score 17 data for HB 191 is, right? 18 19 Α I don't. 20 Okay. Mr. Hieneman, I want to ask you about Q 21 some of the districts you talked about in your testimony 22 earlier. So, the first one was Bowling Green, right? 23 That's the first map on Plaintiff's Exhibit Number 3, 24 right? 25 Α Yes.



1	Q Do you see the purple area under, "2013 Map,"
2	the purple area that has a 17 in it?
3	A Yes.
4	Q That is part of District 17 from the map that
5	was enacted in 2013, right?
6	A Yes.
7	Q So if you look at tab 11 tab 10 in our
8	Exhibit number 1 for the Commonwealth, that's another
9	map of the 2013 plan. Do you see District 17 there is
10	mostly Butler County. Do you see that?
11	A Geographically, yes.
12	Q Yes. And it actually kind of looks like a
13	jellyfish, doesn't it, with like the tentacles, the
14	tendrils hanging down below sort of the base of the
15	jellyfish. And that purple that blue section on tab
16	10, but the purple section in your map that goes down
17	all the way from the northwestern Warren County border
18	and encircles District 20. Isn't that how you would
19	describe that?
20	A I wouldn't say it encircles, but it comes
21	underneath. Yes.
22	Q It comes underneath, and it comes back up on
23	the eastern side, right? So it comes down on the west,
24	goes all the way across the south, comes back up halfway

on the east, right?

1	A Yes.
2	Q Okay. Does that look like a gerrymander
3	district to you?
4	A I don't know the details that come with that,
5	in terms of the racial or voting population to know one
6	way or another.
7	Q But you're complaining in the party the
8	Democrat Party is complaining about gerrymandering in
9	this case, right?
10	A As it relates to yes, the manipulation of
11	district lines to favorite one party over another.
12	Q Okay. So, when HB 1 in the 2012 map created
13	the Pulaski strip and the Pulaski spec to join Casey and
14	Rockcastle, was there a partisan advantage to doing
15	that?
16	A I don't know.
17	Q Okay. Is it your let me ask you to take
18	another look at Plaintiff's Exhibit 3. Your view is
19	that this changed from what HB 2 does from the 2013
20	map, changed the district number 20 from a district that
21	was Democrat to one that's likely to be Republican,
22	right?
23	A Yes.
24	Q Okay. And you said there was population
25	growth in Bowling Green. Do you have any figures for



the Court, on how the population of the city of Bowling
Green grew?
A I I don't have those.
Q Okay. So, as you sit here today, even though
you told the Judge that the population in Bowling Green
grew, and that's part of the reason why the district had
changed, you don't know what the numbers are, right?
A The population of District 20 exceeded the
maximum population of the district.
Q In fact, every district in Warren County is
more than the ideal population in the current map,
right?
A That's the only district wholly contained
within Warren County. I don't recall from the other
districts where they are.
Q Okay.
A I believe, 17 is definitely over the others. I
can't say one or another.
Q Okay. Do you know what the compactness scores
are for HB 2's District 20 versus the 2013 map?
A I do not.
Q Okay. Do you when you look at tab 10 in
Exhibit 1, do you see where part of District 19 in 2013
was in Edmonson and Warren, but then District 23 in

Barren County took a little part of Warren County. Do

1	you see that?
2	A Yes.
3	Q Now in the current map, HB 2, there's no
4	Barren County population in Warren County in the
5	District 19, right?
6	A Correct.
7	Q It's because Barren County population grew as
8	well, right?
9	A I believe so. Yes.
L0	Q Right. And so the map makers were faced with
l1	a choice of how to change the districts in that
L2	geographical part of the state, where both Warren and
L3	Barren County had grown in population, right?
L4	A Yes.
L5	Q Okay. Now looking again at HB the 2013
L6	map, can you tell me how many districts are in Warren
L7	County? It looks like there's 19.
L8	A Five, I believe; is that correct? Five
L9	Q Well, let's count them. We got 19.
20	A 17, 19, 20, 16, and 23, 22. Did I say 22?
21	Q You said 23, but you meant 22?
22	A Well, no. 23 is in there as well. So, 16,
23	17, 19 20, 22, and 23. So, six.
24	Q So, there were six different districts in that
25	one county, that the Democrat Party thought was just



1	fine, right?
2	A The members of the General Assembly thought
3	that.
4	Q Okay. Now how many districts are in Warren
5	County, under HB 2?
6	A Four.
7	Q So, that seems like a big improvement, doesn't
8	it?
9	A Yes.
10	Q Okay. Let me ask you about Erlanger
11	Erlanger. So in the first map on your exhibit this
12	is the second page of Plaintiff's Exhibit 3, the
13	democratic map in 2013 split Erlanger into three
14	different districts, right?
15	A Correct.
16	Q The HB 2 in the current bill splits Erlanger
17	into three different districts, right?
18	A Correct.
19	Q And when did the Democrats decide that
20	Earlenger really should be just one district? Because
21	that's what HB 191 does, right?
22	A It preserves the boundaries of the City of
23	Erlanger in the district.
24	Q But that wasn't important just nine years ago,
25	right?



A You'll have to ask the members of the General
Assembly from 2012.
Q Whom you advised, correct?
A Yes.
Q Did you advise them, that it was important to
keep Erlanger as a single district?
A That was not what I one of the places I
advised on.
Q So you didn't deal with that part of the
state?
A No.
Q Who did?
A I honestly I don't remember.
Q Okay. Let's look at Florence. It's the next
page of Plaintiff's Exhibit 3. There's, in 2013, one,
two, three, four districts in Florence, right? That's
generally a Republican area, right?
A Yes.
Q And it looks like the Democrats were trying to
crack Florence in 2013; wouldn't you say?
A Again, you'd have to ask the members of the
General Assembly.
Q Well, they divided in into four different
counties or four different districts, right? Now it's
only in three, right, HB 2?



A Yes.
Q Okay. Do you know you I think your cheat
sheet told us what the partisan makeup was; can you
remember?
A Yeah. There were three Republican districts
under House Bill 2.
Q Okay. And is there let's see. So HB 2
creates three Republican districts, 62 percent, 60
percent, and 63 percent. Your bill would keep it intact
60 percent. So where's the partisan disadvantage to the
Democrat Party, by Erlanger being either three districts
or one district, with 60 percent or more Republicans?
A Looking at this isolated, I don't think can
give it a whole perspective.
Q Well, you isolated it on Plaintiff's Exhibit
3, not me, sir.
A And that's true. But that is to demonstrate
what happened to the city, not necessarily what happened
at large.
Q Okay. But as you sit here today, you can't
tell the Judge what the partisan disadvantage is, right?
A No.
Q Okay. Let's look at the next page.
Georgetown, 2013 Democratic Party in charge.
Georgetown's one, two, three districts, right?

1	A Yes.
2	Q Okay. HB 2, two districts. So that's an
3	improvement, right?
4	A Yes. Okay.
5	Q Current map, HB 2 improvement over existing
6	map, sponsored by the Democrat Party nine years ago. And
7	you said that it puts a spike through the middle of
8	Georgetown, I believe, right?
9	A Dividing the city. Yes.
10	Q Okay. And if we look at your cheat sheet, it
11	says, Georgetown has been divided into two
12	unrepresentative (phonetic) majority Republican
13	districts, 52 percent and 58 percent. Whereas in
14	contrast, HB 1 keeps it intact in a competitive
15	district, right? So HB 191 you say, has 53 percent and
16	that's competitive, right? But HB 2 creates two
17	different districts, one of which only has 52 percent,
18	right? So you've got to concede that's even more
19	competitive, don't you?
20	A Yes.
21	Q Where's the partisan disadvantage for the
22	Democrat Party?
23	A Again, it's not only one isolated incident.
24	It's the map at large.
25	Q Right. And you understand that when you do a



map at large, you've got to start somewhere, and you've 1 got to move across the state. And the geography makes 2 3 it very difficult sometimes to keep things just the way 4 you might like them, right? 5 Α Yes. 6 0 Okay. Let me ask you about the -- I think 7 it's the last page. No. It's the next page on the 8 Plaintiffs 3, and that's Hopkinsville. 2013 map, the 9 city had four districts, right? I believe, it's three. 10 Α 11 Q I'm sorry, three districts. Excuse me. District 4, District 8, District 9. HB 2, it only has 12 13 two districts, right? 14 Α Correct. 15 So that's an improvement, right? 0 16 Α Correct. 17 Q Okay. Now HB 191 has two districts, right? 18 Yes. Α 19 Q And your only real concern about Hopkinsville 20 is that it split two Black precincts. It moved one 21 Black precinct into one district, and the other into 22 another district, right? It did dilute the Black population -- or 23 24 voting age population, relative to House Bill 191. 25 Q Now were -- how would you characterize the



Black population district in the existing bill 2013
bill? Was that a majority minority? Was that a
plurality? Was that a coalition or an influence?
A I don't recall.
Q Okay. So after HB 2, did it change from one
of those categories to another, or can't you say?
A I can't say that it changed one way or
another, because I don't know what District 8 or 9
looked like before that.
Q Yeah. That reminds me. You testified that
When you did HB 191, you explicitly considered race in
deciding where districts should be, right?
A Not necessarily. It was not paramount. It
was a secondary factor, in an effort to help maximize
those racial districts. The primary factors were from
Section 33.
Q So here's the thing. When you consider race,
you either consider race or you don't, did you consider
race in 191?
A In certain areas, where it was able to draw
one of these districts. Yes.
Q And is that because the Democrat Party had
made an analysis and reached a conclusion, that that was
required under Supreme Court precedent that race be

specifically considered in these districts?

1 Α No. In fact, you understand don't you -- and the 2 Q 3 parties certainly should doesn't it, that it is illegal 4 to consider race, unless the Voting Rights Act requires 5 it, right? 6 Α No. You don't know that? 7 Q Α No. 8 9 Let me finally take you to the last page of 10 Plaintiff's Exhibit 3, and that's Richmond. So there 11 was one district in the previous map that's three now. 12 And now, it's one in HB 191. So going by my standard, 13 HB 2 makes Richmond a little worse, right? Worse than 14 three districts -- one district to three districts, 15 right? 16 Correct. Okay. And the partisan disadvantage there is 17 Q 18 -- sorry, I got to get a cheat sheet -- is, again, the 19 difference between more Republican districts. 20 in one case heavily Republican 72 percent. And then a 21 competitive district, right? Again, not one that at 51 22 percent, the Democrat Party would count on winning, but at least it would be competitive, right? 23 24 Α Yes.



MR. MADDOX: Okay. Your honor, I know we're

1	running out of time, and I'm trying to be as
2	conscious as I can. I think I'm getting close to
3	done.
4	JUDGE WINGATE: Okay.
5	BY MR. MADDOX:
6	Q You talked about fundraising well, a few
7	other things. Let me before I go to fundraising, let
8	me ask you about the congressional map real quick.
9	MR. MADDOX: Where do we have that? Is that in
L O	our tabs? In our stipulations?
L1	Q If you would look to table of contents SB
L2	3. So, tabs 11 and 12, I think. Yes. So tab 11,
L3	Mr. Hieneman, is SB 3, that's the congressional map.
L4	And tab 13 is the congressional map from 2012, again,
L5	signed in the law by Democrat governor and passed by a
L6	Democrat House, right?
L7	A Correct.
L8	Q Okay. So One of the principle concerns that
L9	the Democrat Party has announced in this complaint which
20	it's filed, is that Franklin County under SB 3 is not
21	kept in the district that's centered on the Bluegrass,
22	right?
23	A Yes.
24	Q And it's not kept in the central Kentucky
25	region where it naturally belongs, right?



1 Α Yes. So, take a look at tab 13, in Exhibit 1. 2 3 that'll show you the congressional map from the 2013 --4 2012 bill, I'm sorry -- that was passed, as I said, with 5 a lot of Democrat support. You see how part of 6 Jessamine County that goes right up to the Fayette County line, is in the 2nd District with Owensboro? 7 Α Yes. 8 9 So, Which is more central to the Bluegrass and 10 the central Kentucky, and Lexington region, northwestern 11 Franklin County or Jessamine County on the Fayette 12 County line? I can't say that one is valued over the other. 13 Α 14 Jessamine County is at least as central to the 15 Bluegrass and central Kentucky, as Franklin County; 16 wouldn't you say? 17 Α Sure. 18 Okay. Did the Democrat Party have any concern 0 19 at all for Jessamine County being carved up in two -- I 20 mean, first of all, it was a split County. And then 21 second of all, it was moved into a district that sort of 22 centered on Bowling Green and Owensboro. Was there any objection to that? 23 24 Α Not at the time it was passed.



Okay. Was there an objection that developed

Q

Т	later?
2	A Not that I'm aware of.
3	Q Okay. This reminds me. I'm sorry. I want to
4	go back to one other district map. And I want to take
5	you to so it's the 2013 map, but in Exhibit 1, tab
6	number 4, we have a blow up of wait a minute. What
7	is this?
8	JUDGE WINGATE: Where are you at?
9	MR. MADDOX: I'm trying to understand what my -
10	- tab 4 is HB 2. So do we have the 2013 there for
11	these? We don't, do we?
12	MS. BECKER: No. We don't (Inaudible)
13	MR. MADDOX: Okay. All right. Well, I'm just
14	going to have to use I'm sorry, Judge. We can
15	put that aside.
16	JUDGE WINGATE: It's all right.
17	BY MR. MADDOX:
18	Q I want to ask you about the large map. So,
19	this is the District 44 and 28 split. Do you see that?
20	A Yes.
21	JUDGE WINGATE: Which one are you on?
22	MR. MADDOX: And Judge, this is this is
23	JUDGE WINGATE: 2013 or the
24	MR. MADDOX: Yes, sir Your Honor, 2013. And
25	so, if you go back to tab 10, the insert at the top



1	for Jefferson County.
2	JUDGE WINGATE: Yes.
3	MR. MADDOX: You'll see that.
4	BY MR. MADDOX:
5	Q And Mr. Hieneman, do you see where District 44
6	and District 28 come down into that narrow section of
7	Jefferson County, that terminates at the Hardin County
8	line? And so, there's like a tri-county area, Bullitt
9	County, Jefferson County, and Hardin County?
10	A Yes.
11	Q Are you familiar with the geography in that
12	part of the Jefferson County?
13	A I'm not.
14	Q Do you know what Dixie Highway is?
15	A I have heard the name. Yes.
16	Q Okay. So, looking at the geography, I'm not a
17	cartographer, but it looks like the split between the
18	Ohio River and the Bullitt County line is maybe a mile
19	or two, maybe less. And my understanding is and we
20	can try to firm this up for Your Honor later, is that
21	Dixie Highway runs right down the middle of that. And
22	that the district line runs right down the middle of
23	Dixie Highway. Do you have any information on that?
24	A I didn't work on Jefferson County in 2012. And
25	I'm not familiar with the geography.



Q Okay. If that were the case and this little					
tail of Jefferson County consists of communities called					
Valley Station, and Kosmosdale, and that sort of thing.					
Do you understand those to be communities of interest?					
A I assume they're cities in Jefferson County.					
Q Right.					
A I'm not familiar with them but I understand if					
you say they're cities. Like I said, I'm not familiar.					
Q Do you know of any good reason why District 44					
under the 2013 map, ran from the Hardin County line all					
the way up to Shively, if it wasn't to provide a					
partisan advantage to the Democrat Party?					
A I have no idea. Again, I didn't work in Jeff					
Jefferson County in 2012.					
Q Okay. So let me just ask you about funding					
and I think we'll be done. You've complained about the					
impact that HB 2 either has had, or might have on the					
Democrat Party's funding, right?					
A Could. Yes.					
Q Could have, is that what you're saying?					
A I believe, that's what I said. Yeah.					
Q Okay. So your testimony to the court is that					
it might damage your fundraising efforts, right?					
A It makes it harder for us to compete, and that					
could jeopardize fundraising.					



Q Okay.	It's a fact, isn't it, that throughout
2021, the Democr	rat Party out-raised the Republican Party
in Kentucky; isr	n't that right?
A Correc	ct (phonetic).
Q By a s	substantial amount, right?
A Correc	ct.
Q Do you	n have the numbers?
A I don'	t.
Q Okay.	
A I don'	t work in fundraising.
Q Okay.	
A And fi	inance.
Q So, ev	ven though the party had a super minority
in the House had	l lost 52 percent of its members in the
House in just th	aree election cycles, the party still
substantially ou	it-raised the Republican Party in
Kentucky, right?	
A In 202	21. Yes.
Q Okay.	Now you don't know what's going to
happen in 2022 h	pecause there haven't been any
fundraising repo	orts. There's been no, like donation
records and stuf	ff made public, right?
A Not th	nat I'm aware of.
Q Okay.	I mean, we're just at the end of the
first quarter, l	last week, right?
	2021, the Democration Kentucky; israely as a second of the Correct Q By a second Q Do you A I don't Q Okay. A I don't Q Okay. A And fin Q So, exim the House had the House in just the substantially out Kentucky, right? A In 202 Q Okay. happen in 2022 he fundraising reports and stuff Q Okay.



A And I don't work in fundraising and finance.

Q Okay. Finally, let me ask you, you were complaining about candidates who were drawn out of their districts. They had announced that they were going to run in the district, and they then learned that they didn't live in the district they planned to run in, right?

A Yes.

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And I think you listed three or four of those districts, right? Are you aware that Kentucky revised statutes provide a mechanism for the party to nominate a replacement candidate, in those circumstances where a candidate who is unopposed in the primary drops out? I think this is the way it works. And the Secretary of State will correct me if I'm wrong. But basically, if a vacancy occurs, this is KRS 18105, sub 3. "If a vacancy occurs in the nomination of an unopposed candidate, or a nomination made by the primary or the certification of candidates for the regular election made under 118215 because of, among other things withdrawal, then " -let's see. "The governing authority of the party may provide for filling the vacancy. But only after the certification is made that the statute's been satisfied." So has the Democrat Party filled those vacancies or tried to?

A I've only learned of that in the past week.
Q Oh, okay. So when you filed the lawsuit, you
didn't know about that?
A No.
Q Okay. Where did you learn of that?
A After the Secretary of State's Office provided
us with a nomination form.
Q Shouldn't a political director know that sort
of thing?
A I'm not an attorney. I don't interpret
revised statute.
Q But you interpret Constitution?
A Not an interpretation, just a strict reading.
MR. MADDOX: Okay. Thank you, Mr. Hieneman. No
other questions, Your Honor.
JUDGE WINGATE: Do you have any follow-up?
MR. ABATE: Give us one second to confer.
JUDGE WINGATE: Okay.
MR. ABATE: It won't be much, if we do.
JUDGE WINGATE: Ray (phonetic), you should have
asked him when he was talking about moving to the
Highlands, and maybe he was a rich lawyer, and made
a little bit of money. What are you thinking?
THE WITNESS: I had to get out, Judge.
JUDGE WINGATE: Everybody wants to live in the



Highlands of Louisville. But every time I drive 1 down through there, I'm like, there's more cars 2 3 parked on the street than any place in America, I think. 4 5 MR. MADDOX: So, judge, I'll tell you my lot 6 was 60 feet wide by 120 feet deep. JUDGE WINGATE: Okay. Yeah. MR. MADDOX: Tells you all you need to know, 8 right? 9 10 JUDGE WINGATE: Did you have off-street 11 parking? 12 MR. MADDOX: I did have a garage. Thank you. 13 Okay. JUDGE WINGATE: There you go. 14 MR. MADDOX: But like most Highland garages, it was falling down. 15 16 JUDGE WINGATE: It was falling down. My best friend and my best man in my wedding lived on 17 18 Crescent Avenue. 19 MR. MADDOX: Yeah. 20 JUDGE WINGATE: And they used to live on 2.1 Bailey. 22 MR. MADDOX: Yep. JUDGE WINGATE: And when -- you know, I really 23 24 didn't know how to take the LSAT. And I went up 25 here and took this UofL professor of political

1	science professor tour and stayed with them and we
2	would go out and drink. I'll probably go off
3	record.
4	(OFF THE RECORD)
5	JUDGE WINGATE: Do you have any questions?
6	MR. ABATE: Your Honor, we're not going to ask
7	any redirect at this time. There a few issues we'll
8	probably deal with Professor Caughey.
9	JUDGE WINGATE: Okay. And if you need to if
10	you need to recall him, you know, he's going to be
11	hanging, you're going to be hanging around, right?
12	Well, good enough. All right. Let's talk about
13	tomorrow morning. We can go off the record about
14	this.
15	(TRIAL ADJOURNED AT 5:54 P.M.)
16	
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CERTIFICATE OF REPORTER 1 2 COMMONWEALTH OF KENTUCKY AT LARGE 3 I do hereby certify that the said matter was reduced to 4 type written form under my direction, and constitutes a 5 true record of the recording as taken, all to the best 6 of my skill and ability. I certify that I am not a 7 8 relative or employee of either counsel, and that I am in 9 no way interested financially, directly or indirectly, in this action. 10 11 12 13 14 15 16 17 18 Brooke Andrew 19 20 2.1 22 BROOKE ANDREW, 23 COURT REPORTER / NOTARY 24 COMMISSION EXPIRES ON: 11/27/2025



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