Al in Consequential Decisions: The Need for Transparency

Sonia Gipson Rankin, University of New Mexico School of Law

Al and Consequential Decisions

that have long-term effects on people's lives: **Employment** (automated hiring) Health care, education, social services, fraud detection **Criminal justice:** pretrial, sentencing, parole, predictive policing **Pros:** evidence-based, objective, accurate, avoids stereotypes

- All is being used in both the public and private sector to make decisions

 - Housing: credit, lending, tenant screening, public housing waiting lists
- Cons: based on historical data, treats people as statistics, black boxes
 - What do citizens and governments need to know about these systems?

Transparency vs. Black Boxes

- What data does an Al use about a defendant or applicant?
- Where does this data come from?
- What does the AI do with this data to make a decision, a score, or a recommendation?



- Do the people affected by an AI, and the decision makers advised by it, understand the logic behind its decisions?
- Do they know what its limitations are, and what kinds of errors it can make?
- Can we independently lacksquareassess Als for accuracy and fairness, or do we just have to take the vendor's word for it?







$INPUT \rightarrow$



Transparency, Due Process, and Equal Protection

Procedural Due Process U.S. CONST. amend. XIV, § 1 N.M. CONST. art. II, § 14

Substantive Due Process U.S. CONST. amend. IX, § 1 U.S. CONST. amend. XIV, § 1 N.M. CONST. art. II, § 18

Equal Protection U.S. CONST. amend. XIV, § 1 N.M. CONST. art. II, § 18

Example #1: Pretrial Supervision

- Public Safety Assessment: Simple point system, publicly known weights
- Based on criminal record: Past convictions, past failures to appear
- Uses age, but not race, gender, employment, education, or environment

Sonia Gipson Rankin, Technological Tethereds: Potential Impact of Untrustworthy Artificial Intelligence in Criminal Justice Risk Assessment Instruments, 78 WASH. & LEE L. REV. 647 (2021)

PUBLIC SAFETY ASSESSMENT RISK FACTORS

RISK FACTOR	WEIGHTS
FAILURE TO APPEAR maximum total weight = 7	points
Pending charge at the time of the offense	No = 0 Yes = 1
Prior conviction	No = 0 Yes = 1
Prior failure to appear pretrial in past 2 years	0 = 0 1 = 2 2 or more = 4
Prior failure to appear pretrial older than 2 years	No = 0 Yes = 1
NEW CRIMINAL ACTIVITY maximum total weigh	nt = 13 points
Age at current arrest	23 or older = 0 22 or younger = 2
Pending charge at the time of the offense	No = 0 Yes = 3
Prior misdemeanor conviction	No = 0 Yes = 1
Prior felony conviction	No = 0 Yes = 1
Prior violent conviction	0 = 0 1 or 2 = 1 3 or more = 2
Prior failure to appear pretrial in past 2 years	0 = 0 1 = 1 2 or more = 2
Prior sentence to incarceration	No = 0 Yes = 2
NEW VIOLENT CRIMINAL ACTIVITY maximum t	otal weight = 7 poin
Current violent offense	No = 0 Yes = 2
Current violent offense & 20 years old or younger	No = 0 Yes = 1
Pending charge at the time of the offense	No = 0 Yes = 1
Prior conviction	No = 0 Yes = 1
Prior violent conviction	0 = 0 1 or 2 = 1 3 or more = 2
Source: Laura and John Arnold Foundation	

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Example #2: Child Welfare and Protective Services



Oregon Department of Human Services to End Its Use of Child Abuse Risk Algorithm

Allegheny County, PA (Pittsburgh)

- Uses prior allegations, publicly funded mental health and drug/alcohol services, jail bookings
- Predicts removal from home within 2 years, rereferral after initially being screened out, or injury



Example #3: Fraud Detection Government's Use of Algorithm Serves Up False Fraud Charges

Using a flawed automated system, Michigan falsely charged thousands with unemployment fraud and took millions from them.

- benefits at a rate of 400 percent plus interest
- 93% of the charges were erroneous
- Agency failed to repay millions of dollars for years

Agency charged 40,000 people, billing them at five times the original

Sonia Gipson Rankin The MiDAS Touch: Atuahene's "Stategraft" and the Implications of Unregulated Artificial Intelligence, 98 N.Y.U. L. REV. ONLINE 225 (2023)





Al can help inform consequential decisions if...

- People affected by them understand what data about them is used and what the AI does with this data
- Decision makers advised by them understand what they mean and what mistakes they can make
- Policymakers understand their strengths and weaknesses
- They are regularly and independently assessed for accuracy and fairness, rather than relying on vendor's claims

All this requires transparency!



Moses, M.E., Gipson Rankin, S.M., Medical artificial intelligence should do no harm. NAT REV ELECTR ENG (2024).



Types of Transparency

"Where constitutional rights are involved, transparency is paramount."

Simple notice: Alert consumers or applicants that an Al is being used **Applicant Challenges**: Allow applicants to see their data and correct it (e.g. FCRA) **Self-assessment**: Require AI developers to assess their own product for bias, and perform due diligence to avoid it (like an impact statement) **Local studies**: Require AI deployers to periodically test the AI for accuracy and bias on local data to make sure it works well for local populations **Independent assessments**: Independent third parties (e.g. ISR at UNM) Full transparency: Public disclosure of design and methods, sources of data, and how the AI uses that data to produce its output

— Computing Community Consortium











