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WILDFIRE ELECTRIC GRID SECURITY

Mitigate Electric Grid Ignitions and Major Wildfire Consequences

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Sandia National Laboratories is a multimission laboratory managed and operated by National Technology and Engineering Solutions of Sandia LLC, a wholly owned subsidiary of Honeywell International Inc. for the U.S. Department of Energy's National Nuclear Security Administration under contract DE-NA0003525.



Goal 1: Mitigate grid caused wildfires

Goal 2: Mitigate consequence from wildfire to grid
and critical infrastructure



**WILDFIRE AND THE
ELECTRIC GRID NEXUS**

WILDFIRE IMPACTS

- Impacts to society, property, and *loss of life*
- Damage to critical infrastructure
- Business impacts (utility business impacts)
- Supply chain impacts for replacement infrastructure
- Insurance
- Lawsuits
- and much more

Lawsuits: hundreds of millions for wildfire impacts. For example:

- 2017 Thomas Fire,
- 2018 Camp Fire,
- 2020 Archie Creek Fire,
- 2020 Zogg Fire,
- 2021 Dixie Fire.

Oregon

- **Buzzard Complex in 2014:** Nearly 400,000 acres burned, costing \$11 million, caused by lightning (14 July to 11 Sep)

California

- **Dixie Fire, 2021:** Second-largest, nearly 0.97 million acres burned, \$637 million cost, caused by PG&E power line (13 July to 23 Oct)
- **August Complex, 2020:** The largest wildfire, nearly 1.03 million acres burned, \$116 million cost, 1 death, caused by lightning (17 Aug to 11 Nov)
- **Camp Fire, 2018:** Deadliest and most destructive, 154,000 acres burned, \$120 million cost, 85 deaths, caused by PG&E power line (8–25 Nov)
- **Mendocino Complex, 2018:** Largest in state history until the Dixie Fire in 2021, 459,000 acres burned, \$220 million cost, 1 death, caused by a hammer spark and under investigation (27 July to 18 Sep)
- **Thomas Fire, 2017:** Seventh-most destructive, 270,000 acres burned, \$124 million cost, 23 deaths, caused by downed power lines from Southern California Edison (4 Dec 2017 to 12 Jan 2018)

Hawaii

- **Hawaii fires, 2023:** Deadliest US wildfire in over a century, nearly 17,000 acres burned, 101 deaths, \$5.5 million cost, possibly caused by downed power lines, exacerbated by dry conditions and high winds (14 June to 6 Aug)

Washington

- **Okanogan Complex, 2015:** Largest wildfire on record, 304,000 acres burned, 3 deaths, 120 homes destroyed, \$44.5 million cost, caused by lightning (15 Aug to 19 Sep)
- **Carlton Complex, 2014:** Nearly 260,000 acres burned, 2 deaths, \$68 million cost, caused by lightning (14 July to 28 Sep)

Nevada

- **Martin Fire, 2018:** Largest in Nevada's history, 436,000 acres burned, \$10 million cost, caused by human activity (5–21 July)

South Dakota

- **Legion Lake Fire, 2017:** Third-largest in state history, nearly 54,000 acres burned, \$2.2 million cost, 2 deaths, caused by a tree falling on a Black Hills Energy power line 11–13 Dec)

Colorado

- **Marshall Fire, 2021:** Most destructive in terms of buildings destroyed, about 6,000 acres burned, 2 deaths, \$2 million cost, cause unknown (30 Dec 2021 to 1 Jan 2022)
- **Cameron Peak Fire, 2020:** Largest wildfire in state history, about 209,000 acres burned, \$133 million cost, cause unknown (13 Aug to 4 Dec)
- **Spring Creek Fire, 2018:** Third-largest in state history, about 108,000 acres burned, \$35 million cost, caused by human activity (27 June to 6 Dec)

Oklahoma and Kansas

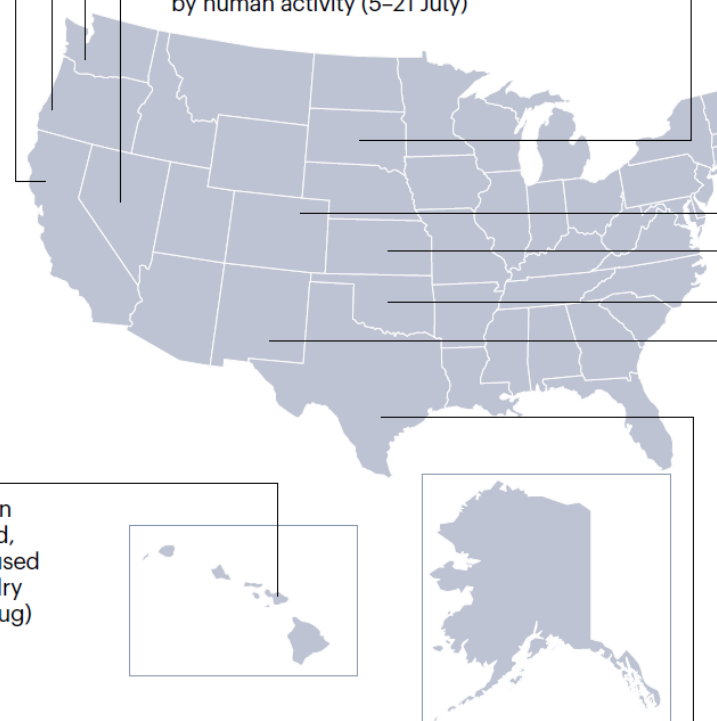
- **Northwest Oklahoma Complex, 2017:** Impacted parts of Kansas and Oklahoma, largest wildfire in Kansas history, nearly 800,000 acres burned, 6 deaths, thousands of cattle killed, \$3.2 million cost, cause unknown (7 March to 24 April)
- **Anderson Creek Fire, 2016:** Impacted parts of Kansas and Oklahoma, the second-largest wildfire in Kansas, nearly 370,000 acres burned, \$1.75 million cost, sparked by a vehicle (23 March to 4 April)

New Mexico

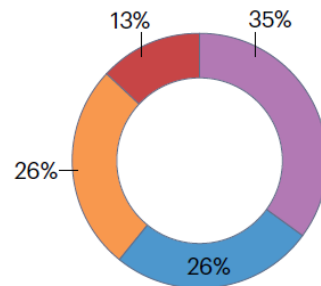
- **Hermits Peak Fire, 2022:** Largest and most destructive wildfire in state history, nearly 342,000 acres burned, \$330 million cost, caused by human activity (7 April to 20 Oct)
- **Black Fire, 2022:** Second-largest fire, nearly 325,000 acres burned, \$60,000 cost, caused by human activity (13 May 10 Nov)

Texas

- **Smokehouse Creek Fire, 2024:** Largest wildfire in Texas history, nearly 1 million acres burned, \$4.6 million cost, 2 deaths, caused by downed power lines (26 Feb to 14 March)



Percentage of wildfires by cause



Alaska

- **Lime Complex Fire, 2022:** Nearly 865,000 acres burned, \$12 million cost, caused by lightning (15 June to 26 July)
- **Old Grouch Top Fire, 2019:** Nearly 307,000 acres burned, \$61,000 cost, caused by lightning (5 June to 1 Aug)
- **Ruby Area Fires, 2015:** Nearly 422,000 acres burned, \$2 million cost, caused by lightning (2 June to 4 Aug)
- **Tanana Area Fires, 2015:** Nearly 500,000 acres burned, \$14 million cost, caused by lightning (14 June to 6 Aug)

Soroush Vahedi, Junbo Zhao, Brian J. Pierre. Key Wildfire Events in the U.S. (2014-2024). Data collected from the Annual National Climate Report (2014-2023) by the National Centers for Environmental Information and the Wildland Fire Summary and Statistics Annual Report (2014-2023) by the National Interagency Coordination Center.

*this slide may not have the most up to date information.

WILDFIRE ELECTRIC GRID SECURITY PROGRAM



Mitigating Electric Grid Initiated Wildfires and Protecting our Critical Infrastructure from Wildfires



Wildfire Electric Grid Security: Planning / Monitoring



Sandia National Labs

ELECTRIC GRID IGNITED WILDFIRES



- Conditions and drivers that led to the ignition event.

	Smokehouse-Creek, TX	Maui Fires, HI	Hermit's Peak, NM	Marshall Fire, CO	Caldor Fire, CA	Dixie Fire, CA	Camp Fire, CA	Woolsey, CA	Carr Fire, CA	Tubbs Fire, CA	Atlas Fire, CA
Ignition Cause											
Ignition Date	Feb '24	Aug '23	Apr '22	Dec '21	Aug '21	Jul '21	Nov '18	Nov '18	Jul '18	Oct '17	Oct '17
High Winds											
Low Humidity	40%	40%	7%	12%	16%	15%	11%	5%	21%	4%	12%
% of Avg. Snowpack	na	na						na		na	na
Live Fuel Moisture	70%	unknown	73%	96%	108%	75%	95%	51%	114%	56%	56%
Dead Fuel Moisture	4%	6%	2%	8%	3%	3%	5%	7%	3%	4%	4%
Dry Grasses											
Tree Mortality											
Drought											
Unusual Heat											
Acreage		•		•							
Cost to Utility	TBD	TBD	NA	TBD	NA	\$45m	\$13.5B	\$2.2B	NA	\$415m	\$415m
Suppression Cost	TBD	TBD	\$968 per acre	\$333 per acre	\$1220 per acre	\$661 per acre	\$667 per acre	\$588 per acre	\$691 per acre	\$2703 per acre	\$1156 per acre
Total Cost	\$0.5B	\$5.5B	\$4.0B	\$2.0B	\$1.2B	\$1.2B	\$16.5B	\$6.0B	\$0.8B	\$9.6B	\$3.3B

Drivers

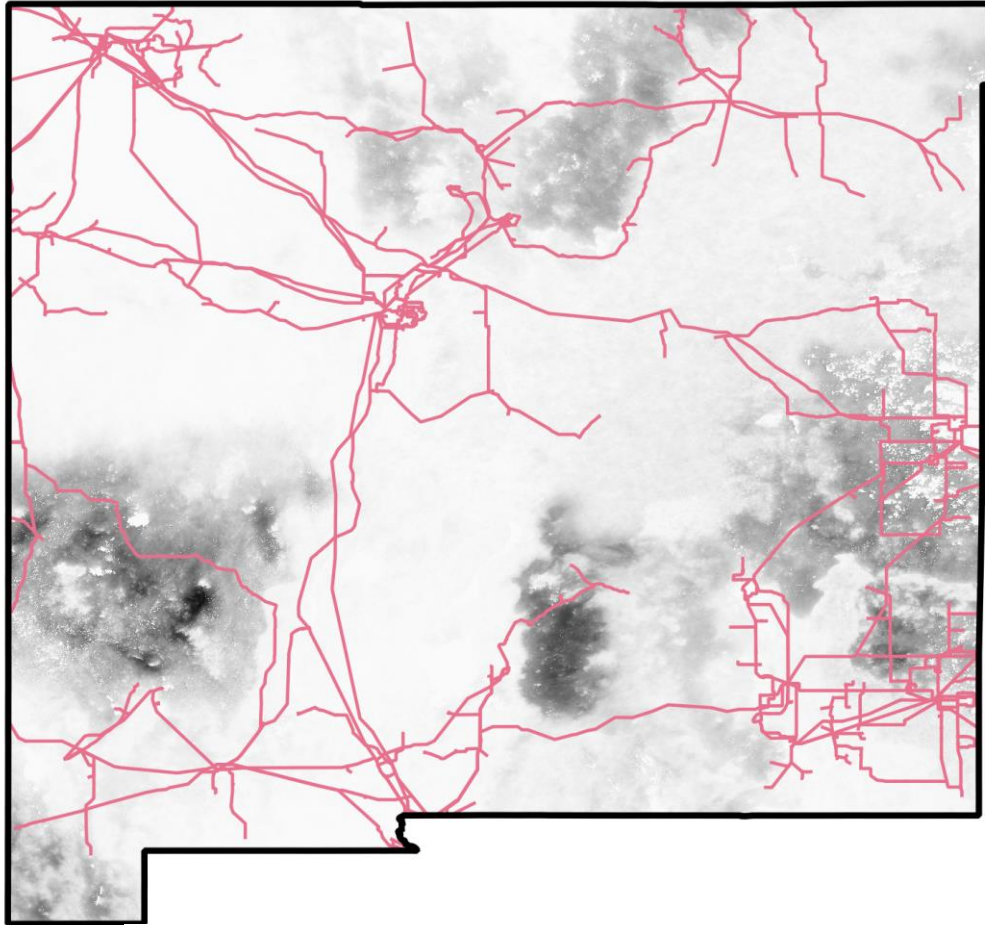
Live Fuel Moisture Content Scale

- 30-50% : Treat as dead fuel
- 50-80%: Yellowing / curing
- 80-100%: Green color pales
- 100-200%: Mature foliage
- +200%: Fresh foliage, growing

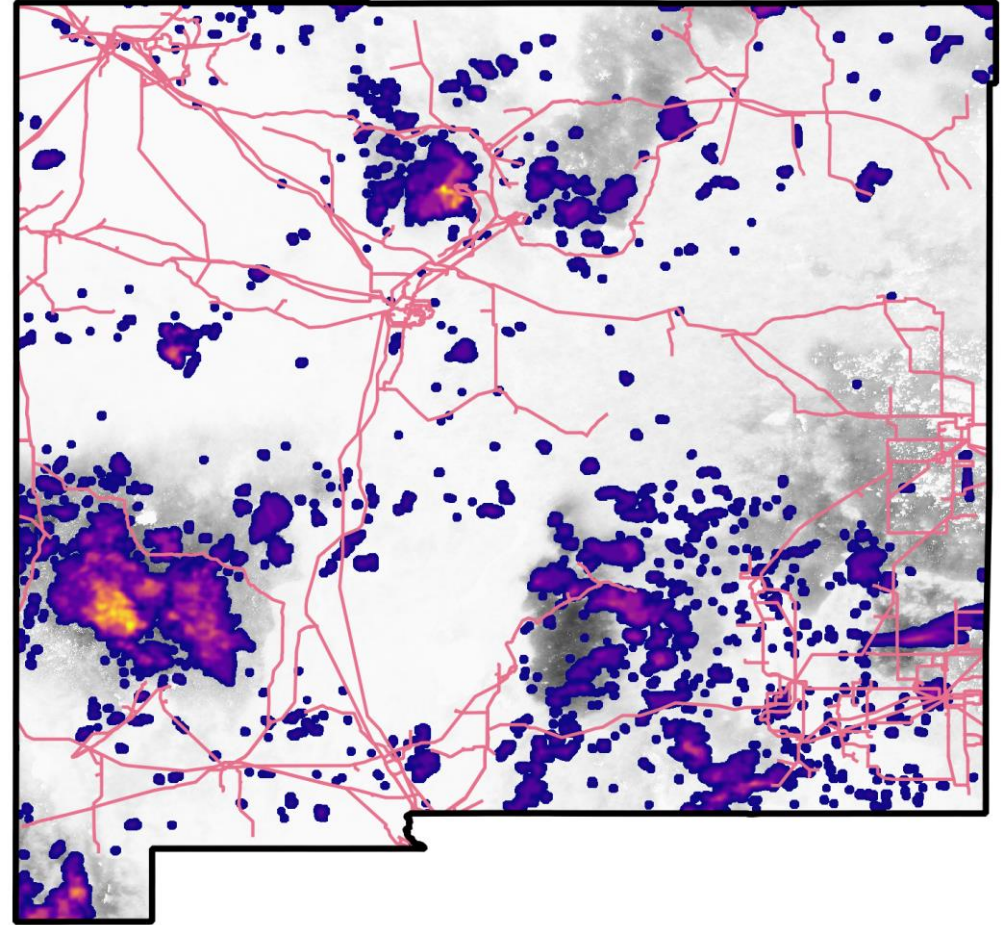
US Drought Monitor Scale

- Abnormally Dry (D0)
- Moderate Drought (D1)
- Severe Drought (D2)
- Extreme Drought (D3)
- Exceptional Drought (D4)
- = 10k Acres

WILDFIRE RISK: NEW MEXICO



— Transmission Lines
Burn Probability
Low High

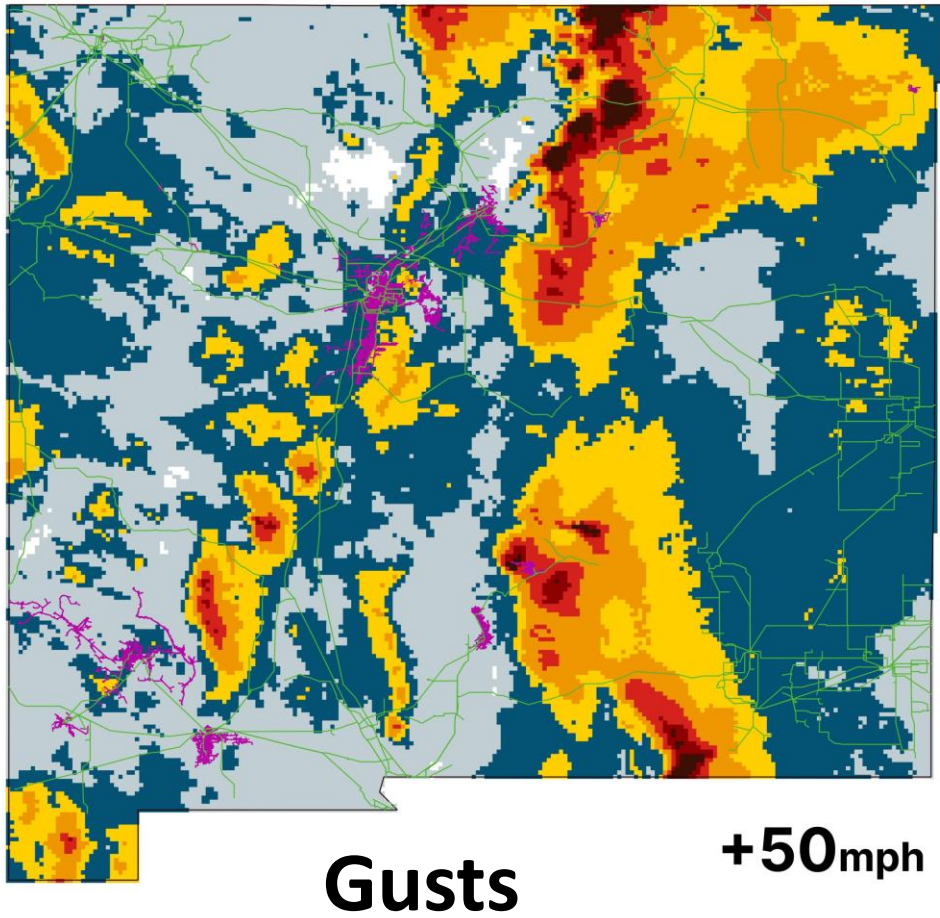


Historic Fire Frequency 1990 - 2023
Low High

GUSTS AND SUSTAINED WINDSPEED | NEW MEXICO 2023 DATA









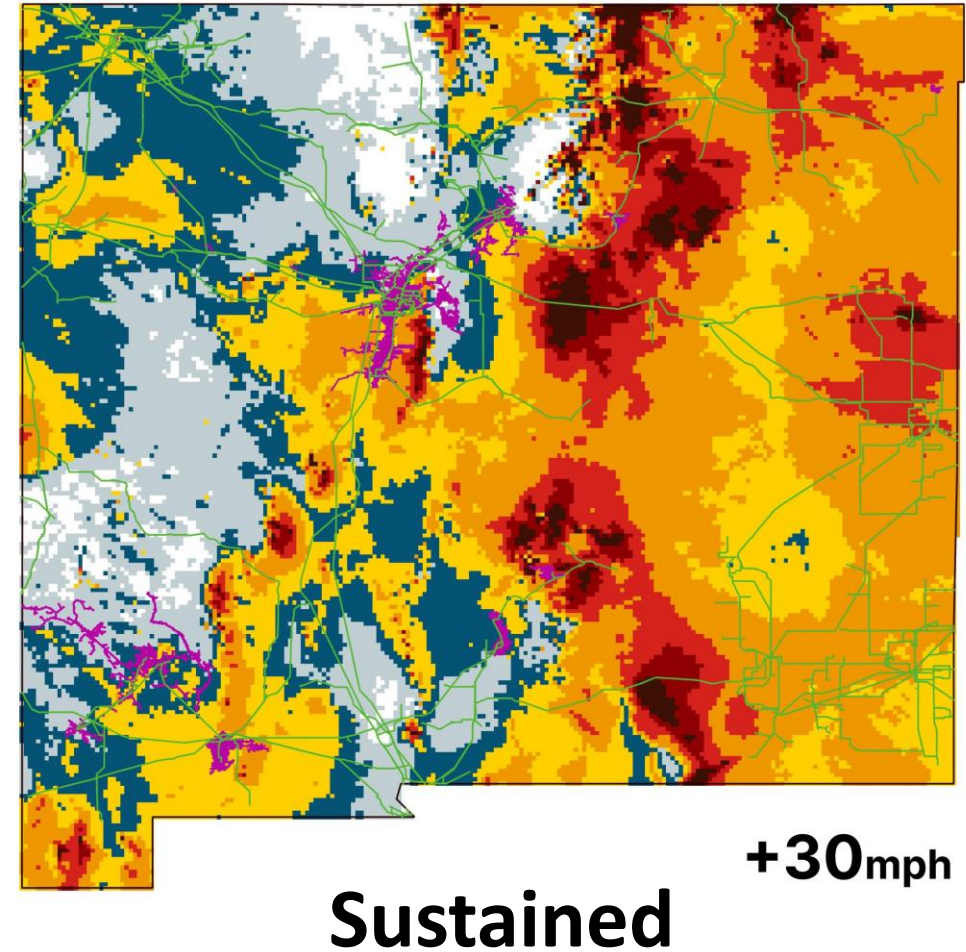
Wind obviously has a major impact on wildfire risk, especially spread rate, which can often lead to the largest wildfires



 **PNM Distribution**
 **Transmission Lines**

Hours during the Year

-  < 10 hours
-  10 - 25 hours
-  25 - 50 hours
-  50 - 100 hours
-  100 - 150 hours
-  + 200 hours



Vegetation Classification for Fire Spread Modeling



Problem:

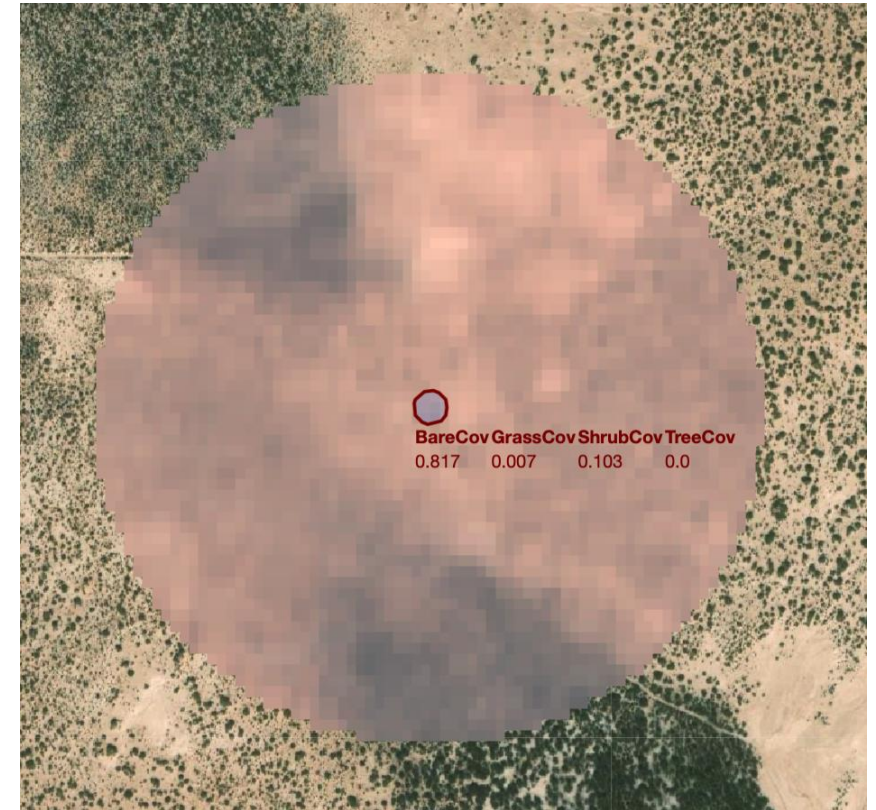
- Utilities require accurate up-to-date fire spread models to safely and efficiently distribute mitigation resources.
- Accurate modeling of wildfire spread requires detailed and up-to-date input data on the vegetation conditions such as amount of biomass, vegetation cover type and flammability characteristics.

Approach:

- Leverage machine learning (ML) with frequent revisit multiband satellite imagery to predict vegetation coverage for accurate up-to-date fuel model generation.
- ML model down-selection and optimization.
- Run fire spread simulations against historic wildfires to validate fuel model generation.

Impact:

- Increase the accuracy and reliability of fire spread models with On-the-fly fuel model generation from up-to-date satellite data.



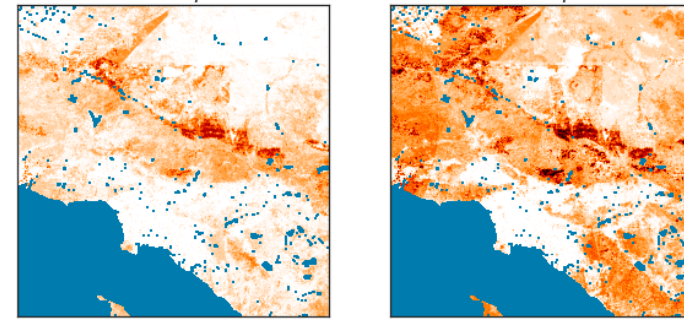
Predicted Vegetation Coverage *Pixel* (inner circle) from ML model trained on BLM labelled data. Outer circle is an RGB stacked 10m satellite image representing a fraction of the model input.

WILDFIRE MODELING



1. Data-driven Dynamic Wildfire Risk Maps

- Satellite imagery, dynamic accurate ML-derived vegetation characterization, weather station data
- Wildfire spread modeling

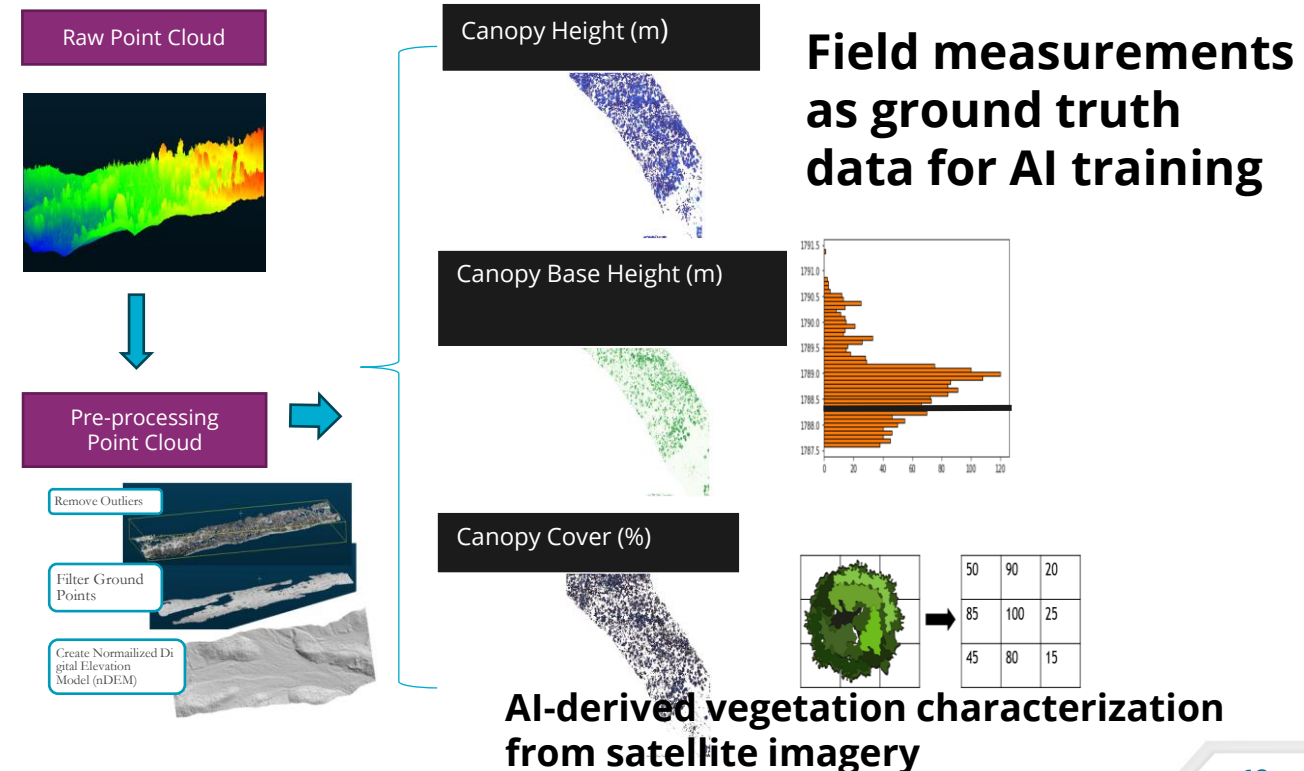
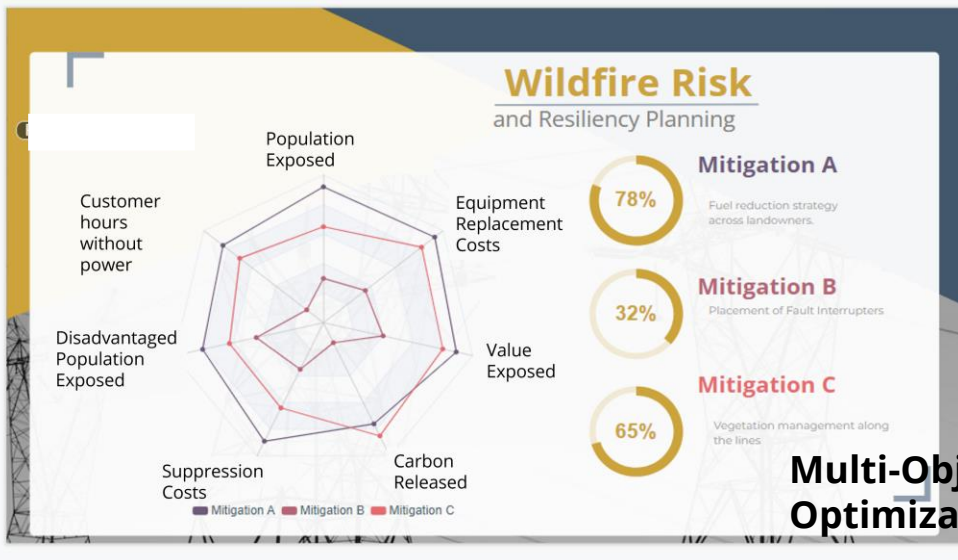


Dynamic Wildfire Risk

2. Dynamic Monitoring for Grid Vulnerability to Fire

- Wildfire risk – electric grid impacts, possible cascading failure.

3. Optimized Vegetation and Resiliency Treatments to Reduce Wildfire Threat



Future Burn Probability: Next Fire Season Impacts to Electric Grid



Problem:

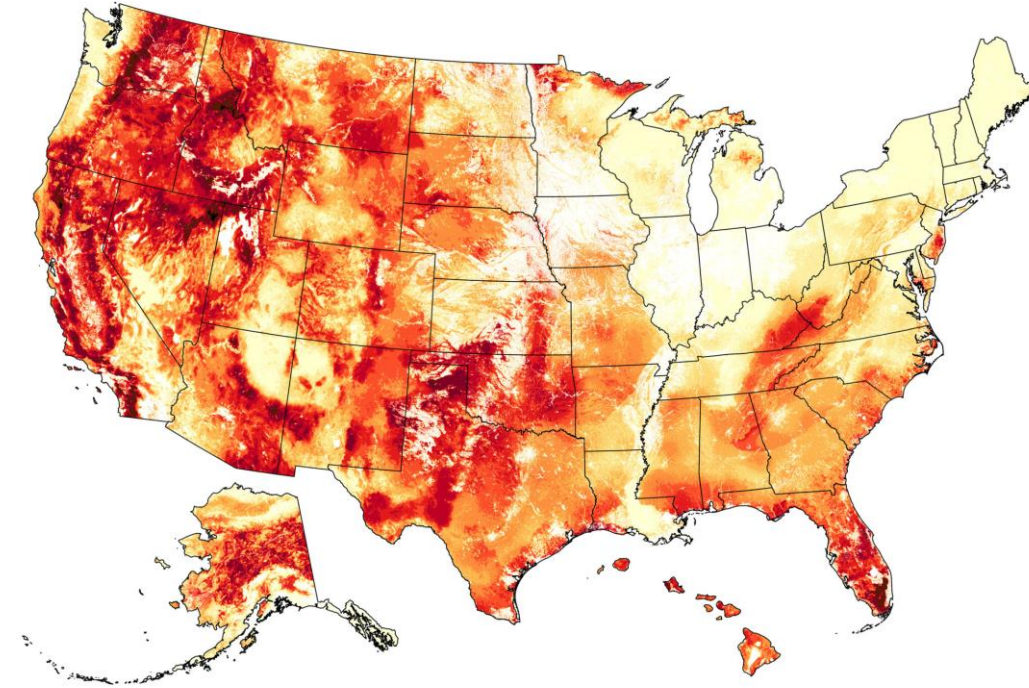
- Utilities need to plan months in advance
- Fire weather and burn probability forecasts are often short term (next 7-days), there is a need for mid-term (months ahead) forecasts

Approach:

- Use **machine learning** methodology with satellite imagery and field data to **predict vegetation growth** 6 months ahead
- Downscale **seasonal forecasts**
- Run WRF-SFIRE fire simulations with **long-term weather forecast** data and forecasted vegetation data
- Ensemble runs and overlay with GIS data of electric grid assets
- Back-test on historic fire seasons

Impact:

- Forecasting wildfire risk for the upcoming fire season increasing preparation and enables fast response in the event of a wildfire
- Partnering with utilities, USFS, and companies to beta-test the tool and align with utility operator needs



Burn Probability: Mapped burn probability for the United States mapped in 2022 by the USFS.

PUBLIC SAFETY POWER SHUTOFF (PSPS) OPTIMIZATION FOR WILDFIRE RESILIENT GRID OPERATION



Investigating the best strategies for PSPS operations by combining optimization models and data-driven PSPS forecasts

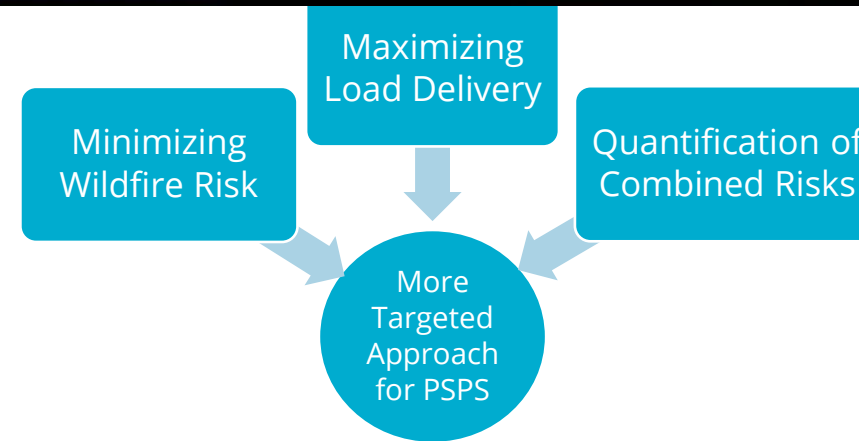
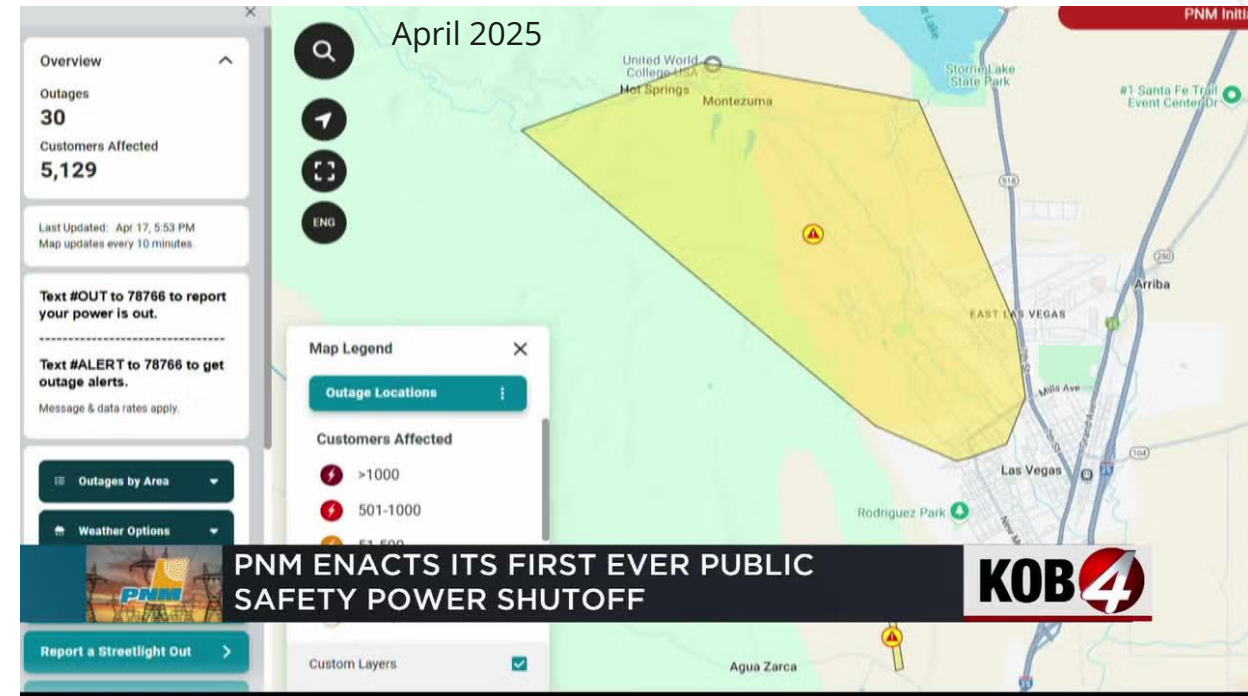
Problem: It is challenging to achieve both safety from wildfire and reduce the number of customers impacted by PSPS. PSPS action may severely impact critical loads and leave the grid network vulnerable to additional contingencies in post-PSPS periods.

Approach:

- Development of optimization models to minimize load-shedding and impact on customers without compromising wildfire risk
- Ensuring more critical loads getting served in PSPS events
- Leveraging the PSPS forecasts to complement the optimization model and help utilities select the best PSPS strategy.

Expected outcome and impact:

- Assist utilities in identifying optimal deployment of PSPS, to minimize wildfire risk, maximizing service to customers especially critical loads, and minimizing the impact to the bulk electric grid security (e.g. a “weakened” state possibly leading to cascading outages).



Thank you KOB4 for reporting

PyroKit: Wildfire Risk and Mitigation Planning Tool



Problem:

- Assessing current conditions and risk is critical to avoid electric-equipment started fires and to be prepared for encroaching wildfire.

Approach:

- Leveraging previous R&D investments at Sandia National Labs, we will use machine learning algorithms trained on satellite imagery and weather station data to determine wildfire risk and mitigation effectiveness.
- Data models include:
 - Dynamic utility wildfire risk maps
 - Burn Probability
 - Red Flag Warning
 - Customizable Public Safety Power Shutoff (PSPS) calculator
 - Contingency Analysis
 - Mitigation Scenarios / Solutions

Impact:

- Provide decision makers with an interactive map which shows fire risk and then allows the user to run mitigation scenarios based on user input management objectives.

The screenshot displays the PyroKit web application interface. The central map shows a wildfire risk area in orange/red, with a tooltip for FID: 132 showing a treatment area of 101 acres and a shape area of 408059.656. The interface includes several panels:

- Mitigation Scenarios:** A list of six scenarios: 1. Number of Population, 2. Number of Structures, 3. Number of Vulnerable Communities, 4. Wildfire Suppression Costs, 5. Equipment Replacement Costs, and 6. Customer hours without power. A note indicates users should rank these by importance (1-6) and click "Run Mitigation Scenario".
- Treatment Information:** A table showing details for Treatment ID: 1, including Treatment Area (137), Land Owner (USFS), Treatment Type (Physical and Prescribed Burn), Physical treatments (Canopies of trees reduced by 50%, Canopy base height increased 1m, Removal of ladder fuels under trees), Prescribed Burn treatments (Surface fuel reduced from high load to low load, Example: Very high dry climate shrub density to low dry climate shrub density, Removal of high load of litter, Reduction of dense grass biomass), Treatment History (Commercial Thinning 2019), and Land Owner (USFS).
- Treatment Objectives:** A radar chart comparing Impact and Fire Behavior across six metrics: Number of Vulnerable Communities, Number of Structures, Population Exposed, Customer Hours, Equipment Replacement Costs, and Wildfire Suppression Costs. A legend indicates "Data Driven" results.
- Impact of Data Driven Mitigation Plan:** A table showing the impact of the plan on six metrics:

Metric	Value
1. Structures Exposed	12,407
2. Population Exposed	7,699
3. Disadvantaged Population Exposed	3,369
4. Wildfire Suppression Costs	\$85,165,056
5. Equipment Replacement Costs	\$6,496,352
6. Customer Hours Without Power	135

WILDFIRE TOOLS

- Many tools to be aware of to assist in wildfire modeling, characterizing wildfire risk, spread modeling, forecasting, wildfire behavior analysis.
- Examples of other lab risk tools:
 - RADR-Fire (PNNL)
 - AHA (INL)
 - WildfireGPT (ANL)
 - Wildfire Tool Inventory and Evaluation (EPRI)

Table 1 | Wildfire model comparisons

Wildfire model	Developer	Primary application	Key features	Benefits	Limitation	Used in PSRA
FireSim	Technosylva	Deterministic and probabilistic modelling, real-time fire behaviour prediction	Physics-based wildfire models, initial attack assessment, impact analysis, urban encroachment algorithms, real-time data calibration	Quickly determines fire path and impacts, all-in-one platform: wildfire risk forecasting, spread predictions, risk mitigation and fire behaviour analysis	May not capture all complexities of fire behaviour, not free for use	Used by PG&D, SCE, San Diego Gas & Electric, Xcel Energy, Bear Valley Electric Service, Liberty for wildfire mitigation plan
IFTDSS	US Forest Service	Fuel treatment planning and wildfire risk assessment	Web-based application, integrates multiple models (FlamMap, FARSITE, BehavePlus)	User-friendly interface, comprehensive US data, step-by-step fuels treatment testing, supports decision-making, free access, generates maps, graphs and tables	Requires detailed input data	Using IFTDSS, they provided a wildfire characterization package enabling proactive decision-making for the wildfire mitigation plan
Solid Fire Model	NA	1D/2D flame model for deterministic/probabilistic wildfire risk modelling, fire management, firefighting	Physics-based approach, detailed fire behaviour simulation, computes radiative heat flux transfer	An easy-to-use tool for evaluating wildfire risk, aiding fire management decisions and integrating into power system risk assessments	Does not account for crown fires and spotting, represents the flame only as a radiant surface (solid-flame assumption) and may lack accuracy	The developed resilience assessment quantifies how wildfire characteristics such as ignition probability, intensity, spread rate, temperature and severity affect the failure likelihood of power system components
ELMFIRE	Chris Lautenberger	Real-time and historical fire spread forecasting	Physics-based model that considers fuel, topography, weather and fire suppression; Monte Carlo analysis	Real-time forecasting quantifies fire risk exposure		
FlamMap	US Forest Service	Deterministic fire behaviour prediction and landscape analysis under constant conditions.	Physics-based model, produces raster maps, integrates multiple fire models, provides environmental condition data	Detailed fire behaviour maps, comprehensive analysis		
FARSITE	US Forest Service	2D deterministic fire growth simulation	Huygens, combines models for surface, spot, crown fires, wave dissemination models	Combines multiple fire models, generates fire propagation maps, essential for forest fire extinction decision-making		
MTT	US Forest Service	Underlying model for FlamMap and FSIm	Physics-based prediction for fire perimeter expansion, calculates MTT across a 2D network of landscape nodes	Approximates complex fire behaviour models at low computational cost (makes it well suited for running many wildfire simulations), predicts fire behaviour and perimeter expansion effectively	Not designed to predict final fire extent — final perimeters depend on simulation duration, requires detailed input data	A study evaluated wildfire risk-mitigation measures by PG&D, using MTT for detailed ignition risk predictions based on data from over 25,000 miles of high-risk lines ⁵⁴
Burn-P3	Natural Resources Canada (NRCan)	Landscape-scale wildfire simulation	Physics-based model that uses Prometheus model, evaluates fire characteristics and produces burn probability maps	Detailed predictions, supports planning, open source	Extensive input data, computationally intensive	No

Wildfire Electric Grid Security: Ignition / During Wildfire

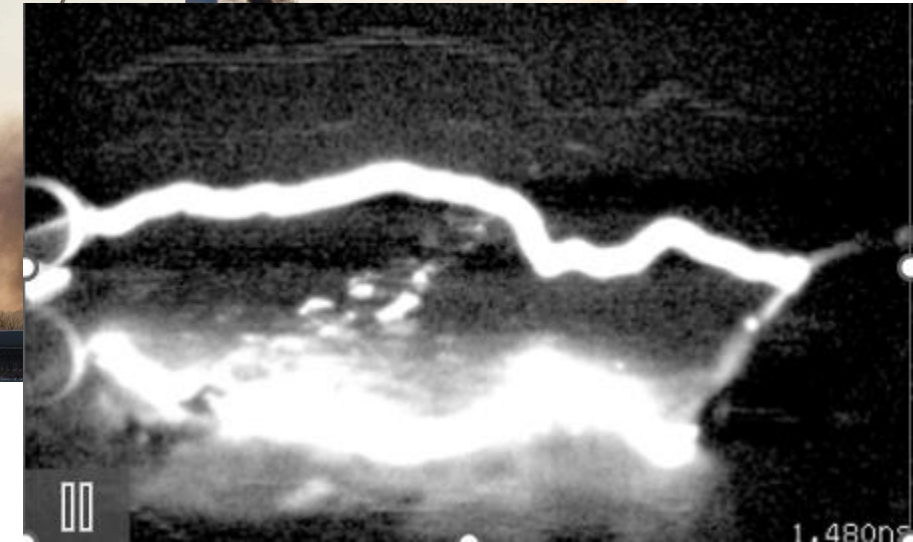
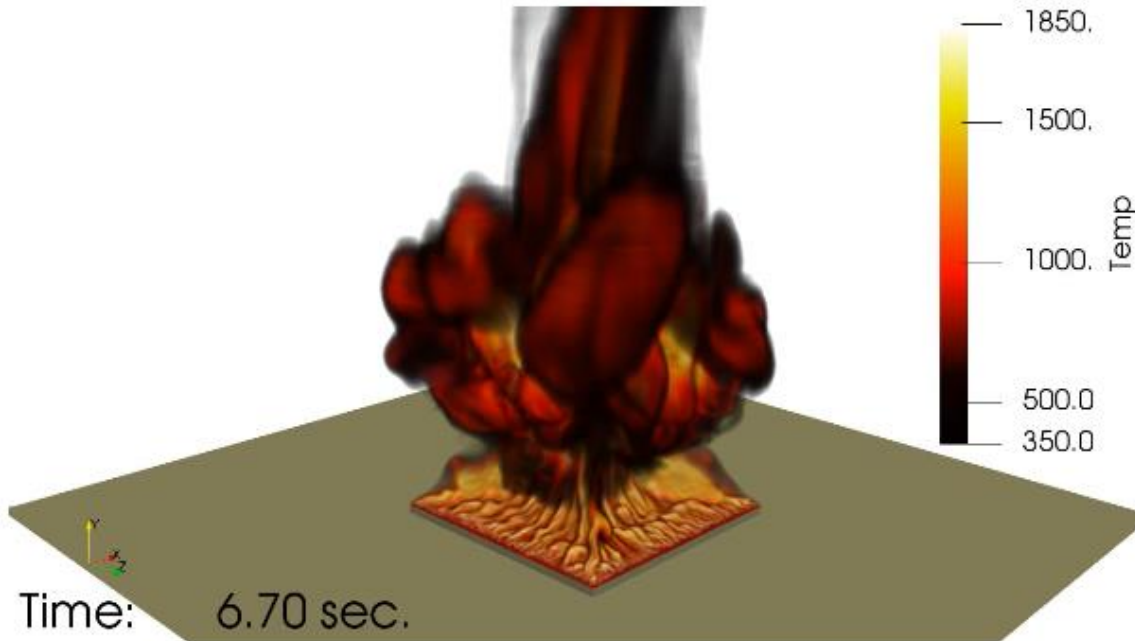


Sandia National Labs

LIGHTNING MODELING

1. First principles of wildfire ignition by lightning

- Better understand probability of lightning ignition given different situations to inform wildfire response.
- Create an experimentally-driven computational model to understand the predictability of lightning-ignited fire using Sandia's lightning strike lab capabilities for testing.



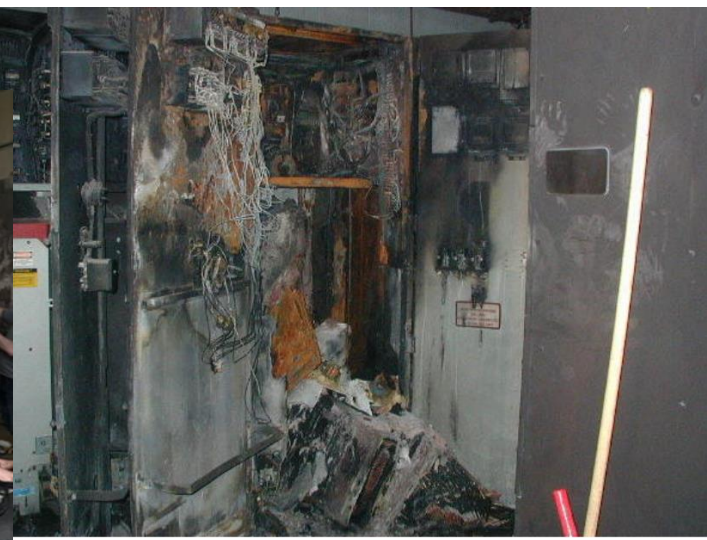
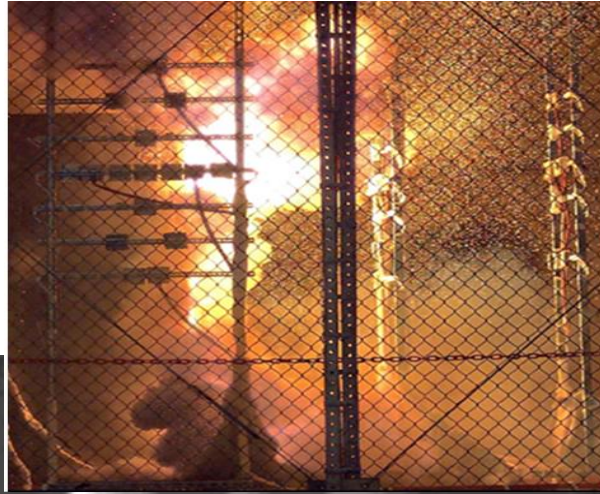
2. Novel lightning monitoring of critical assets for wildfire risk assessment

- Develop a novel lightning monitoring system that provides total lightning current (and energy), to predict ignition by lightning.



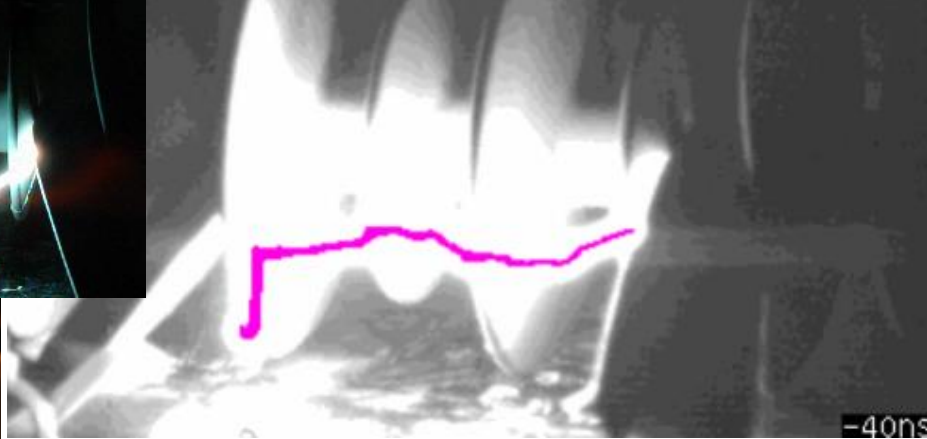
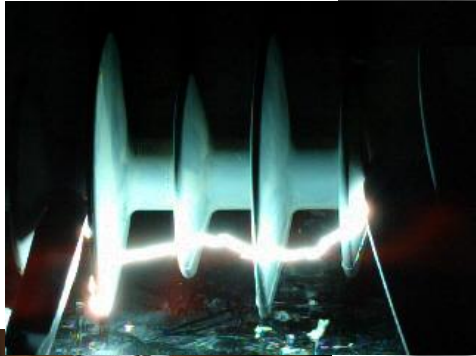
ARC MODELING

- Testing arc probabilities
- Testing arc ignition probabilities



Nuclear Energy

“There was a failure of the main contacts of a 25 year old 4.16 kV breaker to close fully, causing a HEAF event... the fire persisted for three hours until water was applied.” ~San Onofre Nuclear Generating Station



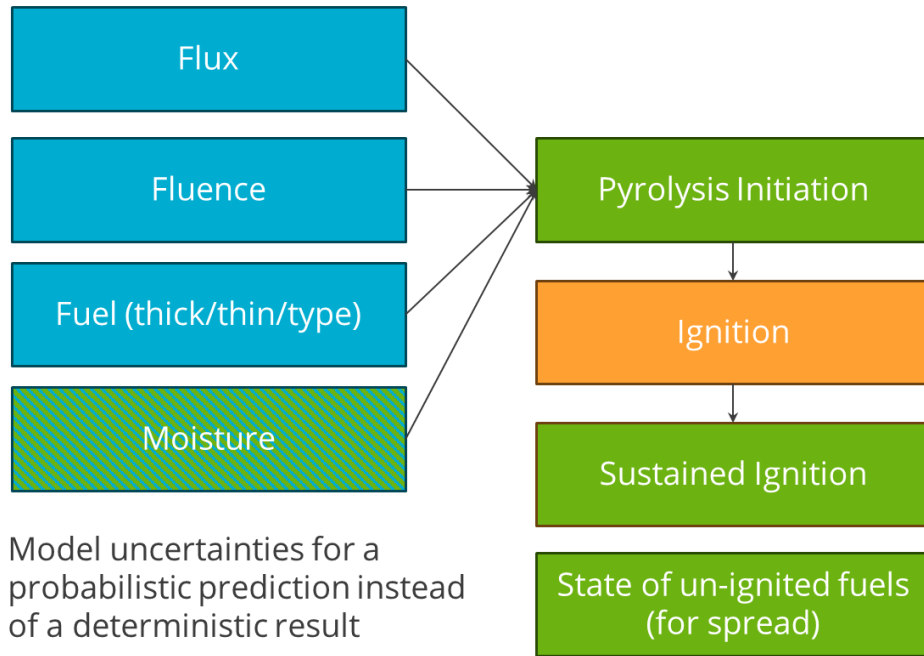
Insulator Flashover from Wildfire Contaminants

- High voltage breakdown testing of clean vs. contaminated/aged insulators
- Define risk metrics for contamination and aging at which point risk of failure / faults increase
- Develop failure thresholds and tracking criteria (working with wildfire propagation projects at Sandia) that can be used for grid health predictions in response to wildfires.

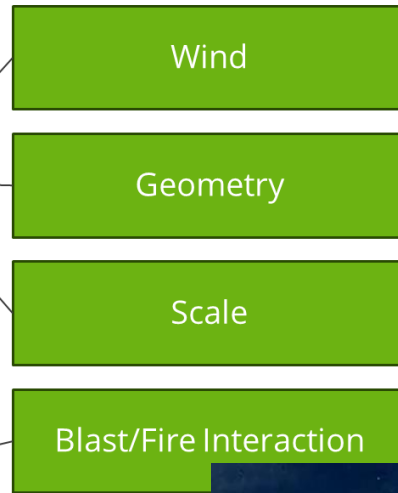
LEVERAGE IGNITION MODELING FROM OTHER SOURCES



Existing Models:



Potential New Phenomenology:



5-year testing campaign, conducted high-flux exposures at Sandia National Thermal Test Facility.

Leveraging ignition models from other sources.

- Identified missing physics in most existing models:
 - Wind effects on ignition
 - Geometry effects on ignition
 - Effect of scale on ignition



Protective Relaying To Reduce Wildfire Risks and Possibly Reduce Public Safety Power Shutoff (PSPS) Events



- New power system protection methods and new technologies can reduce wildfire grid ignitions significantly.
- Sandia protective relaying R&D efforts to significantly reduce future wildfire risks

Protection Planning for Faults

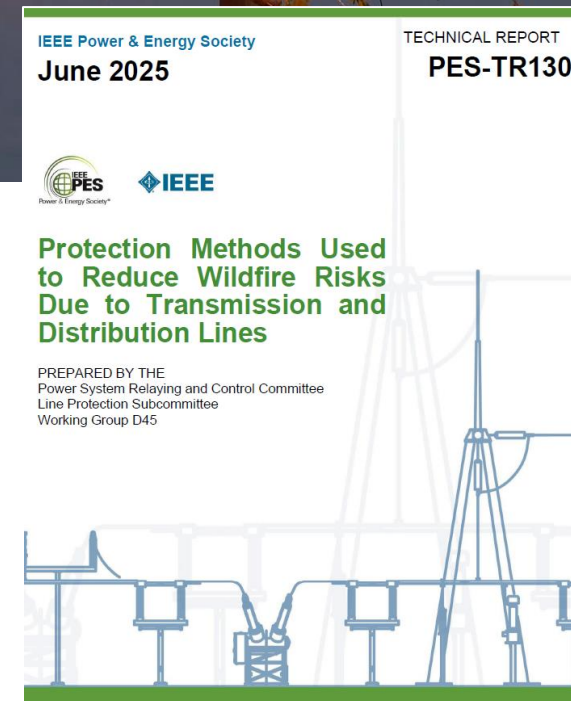
1. Rapid Earth Current Fire Limiter (RECFL)

Preemptive Protection Response

2. Adaptive Settings – Fast Sensitive Trip, No Reclose
3. Incipient Failure Detection

Fast Fault Clearing

4. AI-Based Traveling Wave Relay
5. Communication-Assisted Fast Protection



AI-BASED TRAVELING WAVE PROTECTION SCHEME FOR DISTRIBUTION SYSTEMS



Use fast fault location to quickly detect and isolate an incident before it leads to a wildfire

Problem: The potential for fire ignition is proportional to the duration of the arc, and current protection schemes **generally take around 100 milliseconds** to a second to operate

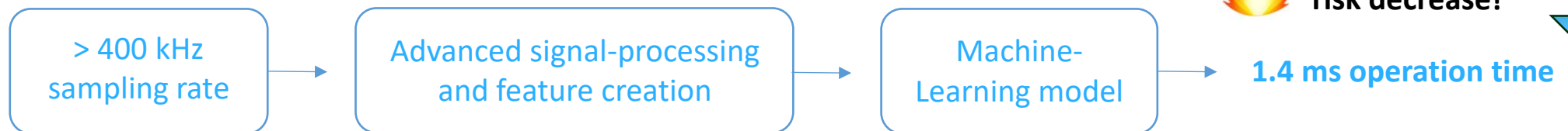
Solution:

- Develop fast, local, bi-directional, data-driven fault detection and location schemes for distribution systems, including DER high-penetrations, that operate in **less than 2 milliseconds**
- Use high-frequency (1 MHz) **traveling wave** methods combined with physics-informed **Artificial Intelligence** can learn correlations to determine the fault location – Ability to detect fault location in the distribution system within 100 meters

Conventional protection



Fast, AI data-driven protection

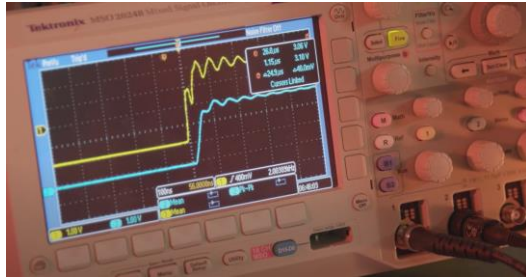


Trained on past faults and simulated faults

AI-BASED PROTECTION – TECHNOLOGY VALIDATION EFFORTS AND DEMONSTRATIONS



Kirtland Air Force Base, Albuquerque, NM (2023)

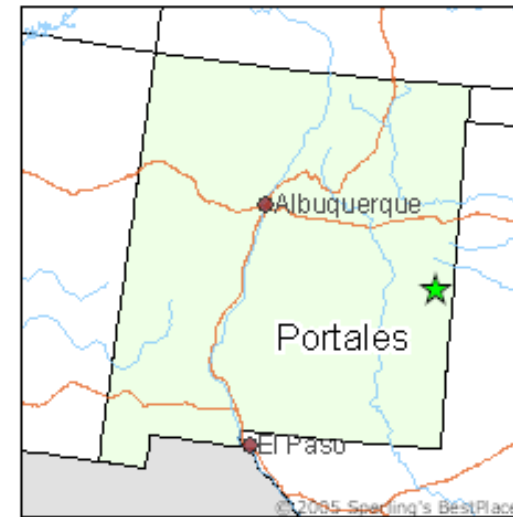


[Link to YouTube demonstration video](#)



Portales, NM (2024-2025)

New Mexico

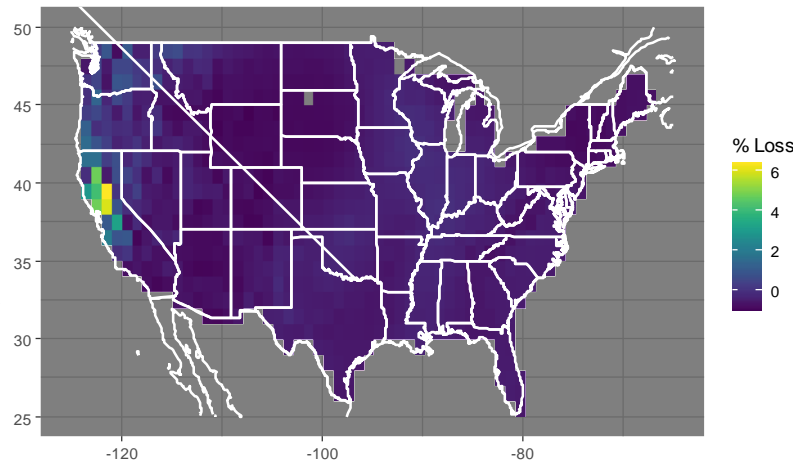


Risks/challenges to investigate:

- Integration with MV commercial sensors
- Impacts of system model accuracy

SMOKE MODELING

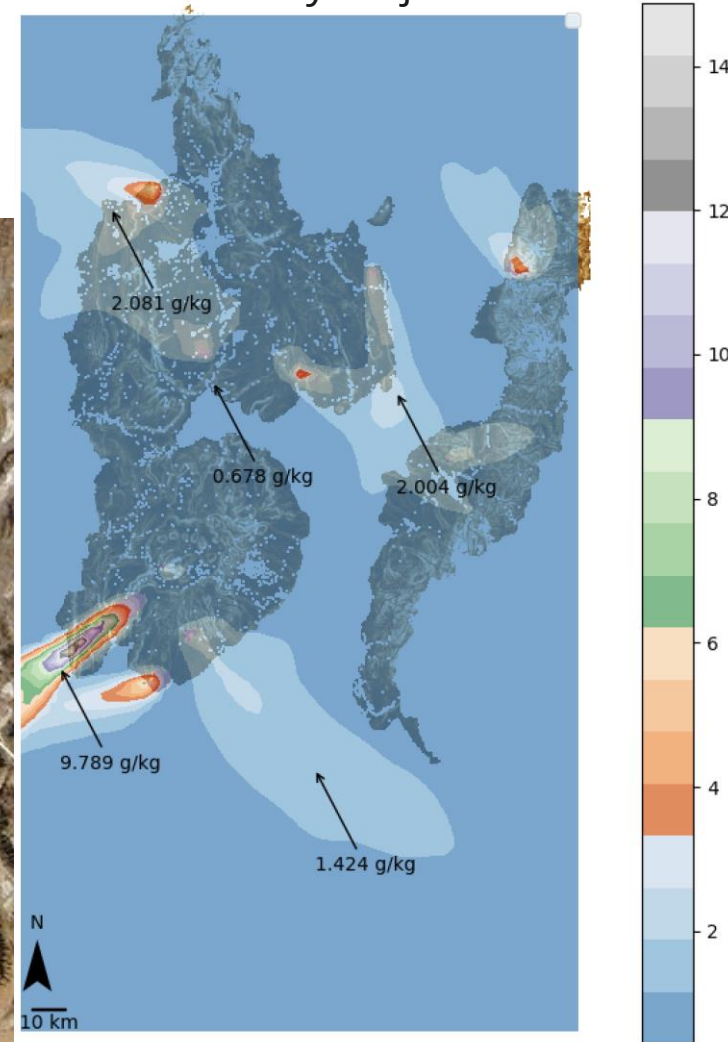
Energy Production Ratio - Difference (Annual - Predicted)
Camp Fire: 11-14-2018



Custom WRF-Fire code and LANDIS-II running on Sandia HPC environments

Smoke plumes from many major fires simulated

- Developed a coupled landscape, weather, wildfire, smoke modeling platform.
- Wide-scale future smoke impact across the U.S. (plum modeling).
- Help forecast future smoke impact to energy systems, primarily solar power.
- Smoke impact to military installation mission assurance.



VISUALIZING UNCERTAINTY: PSPS AND EVACUATION



Decision making for Public Safety Power Shutoffs and Evacuations.

Design choices, visualizations, impact decision making. How should data be presented to a grid operator to make the optimal decisions.

State uncertainty is uncertainty about the current or future state of some phenomenon

- Very common in weather forecasts, hazard maps, AI/ML outputs

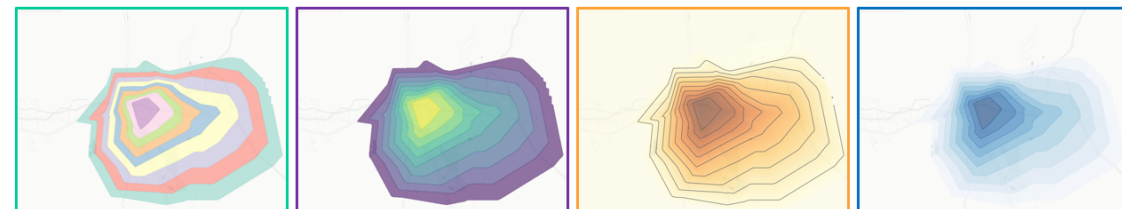
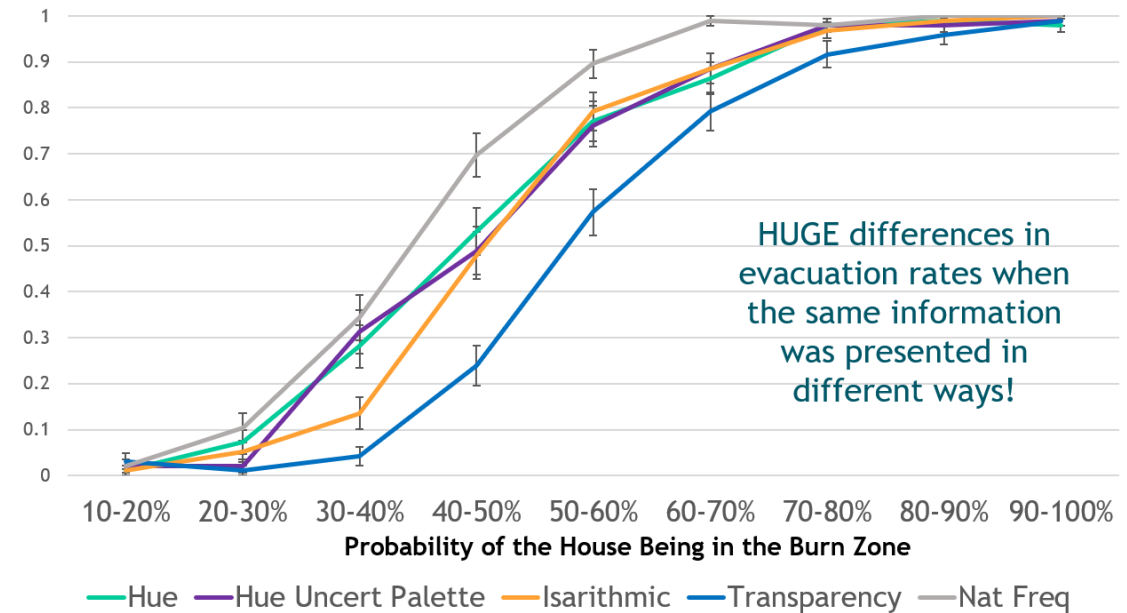
Humans are notoriously bad at understanding state uncertainty and probability

- Prior research suggests that **different representations of uncertainty can lead to different patterns of decisions**, but we don't yet know when and why this happens.

Grid operations (and modeling) involve complex, heterogeneous, uncertain information

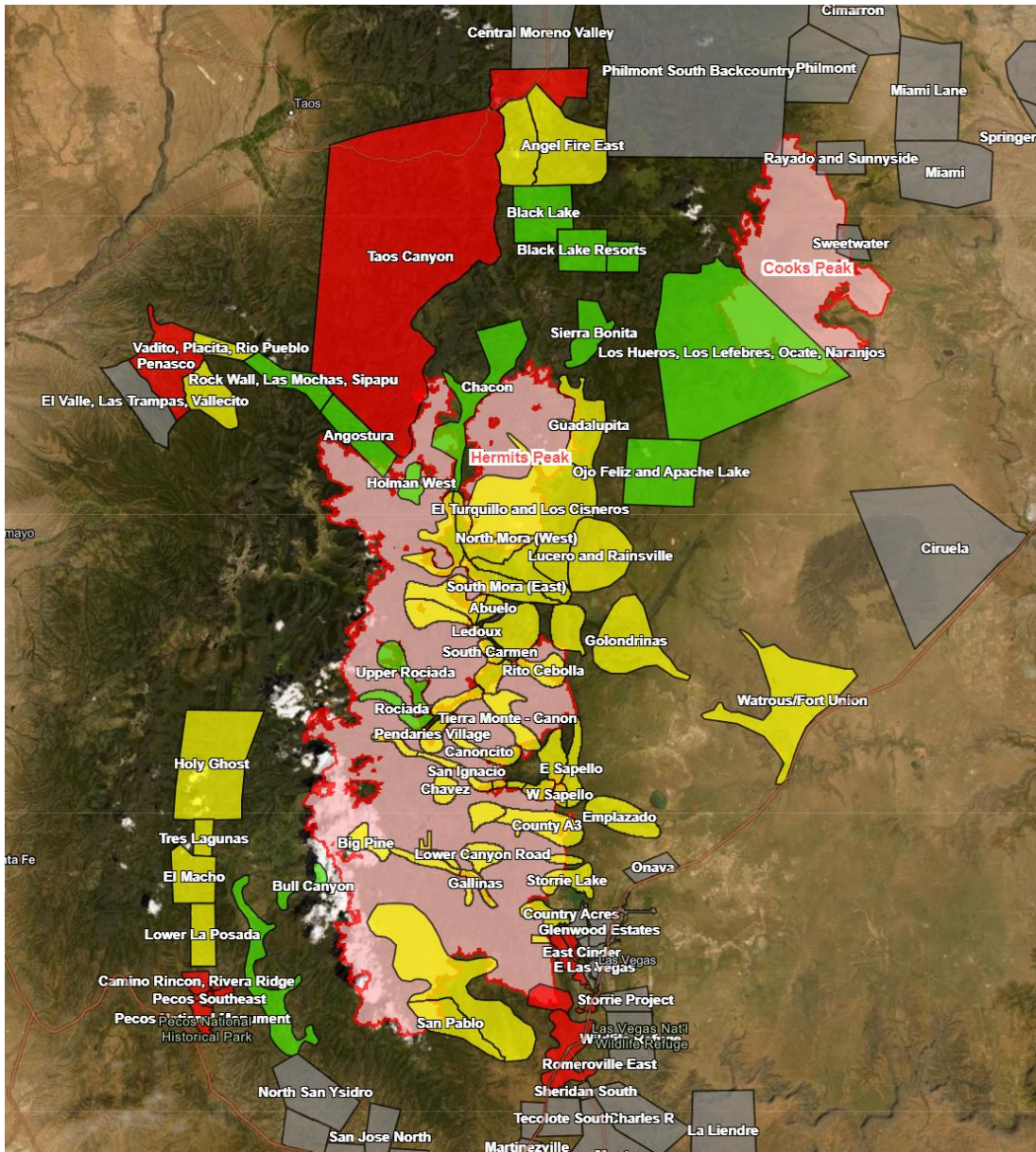
- **How should that data be presented to support optimal decision making?**

Wildfire evacuation decision dependent on how the information is presented



Your house is located in the **40 to 50%** burn likelihood zone.

THESE CHOICES MATTER



New Mexico wildfire evacuation material from long ago.

NEW MEXICO



Wildfire Electric Grid Security: Recovery / Post Wildfire



Sandia National Labs

ACCELERATED GRID RECOVERY POST WILDFIRE



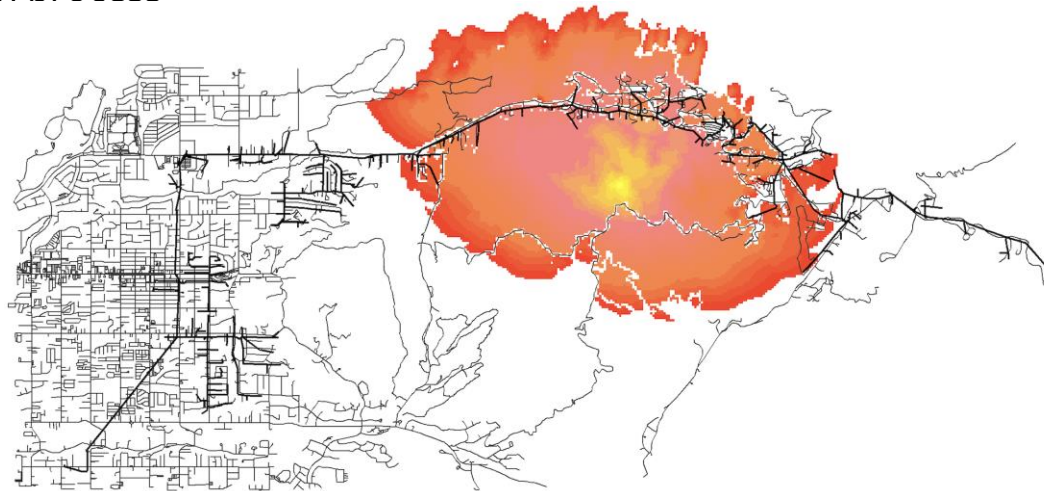
Problem: Electrical outage time due to wildfires needs to be minimized.

Approach: Today, linemen are frequently stationed in areas affected by wildfires. They await safety approval to access damaged areas and then they proceed to assess and repair the affect grid infrastructure.

Expected outcome and impact: In partnership with PNM/SCE, this approach will manage key information that can accelerate grid recovery

- Document existing capabilities
- Created an optimization method enabling an accelerated recovery decision process

FlamMap fire progression coupled with grid and transportation infrastructure



Load Restoration	System Recovery
Residential	Transmission
Commercial	Distribution
Industrial	Adequacy/Reliability

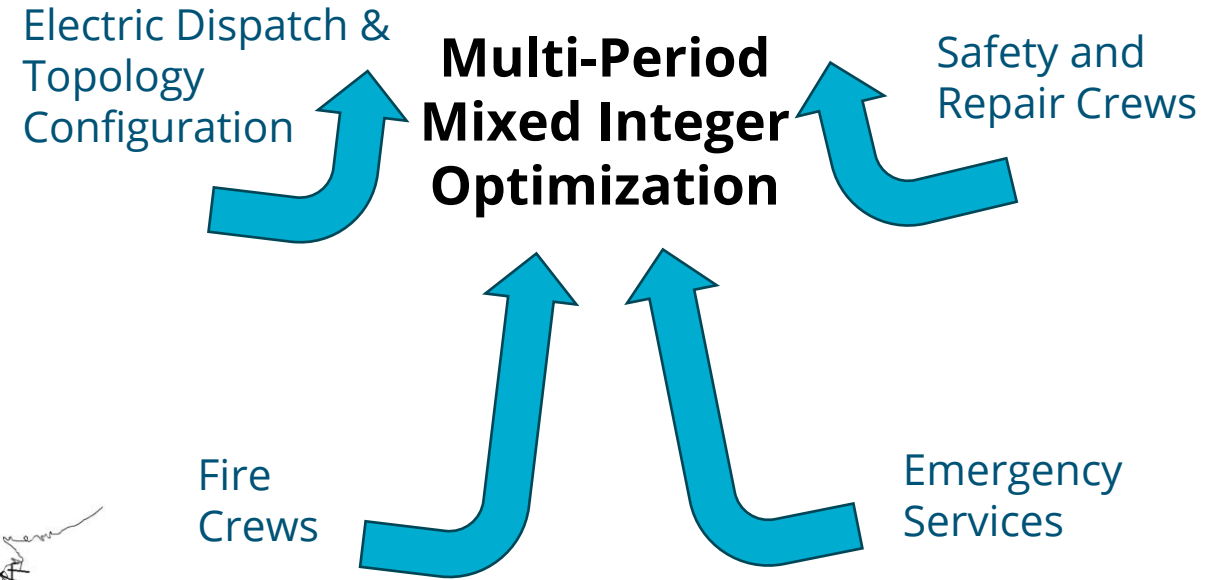


Figure: Optimal Recovery of Wildfires for Electric Utilities



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DOE CESER Natural Hazards Program

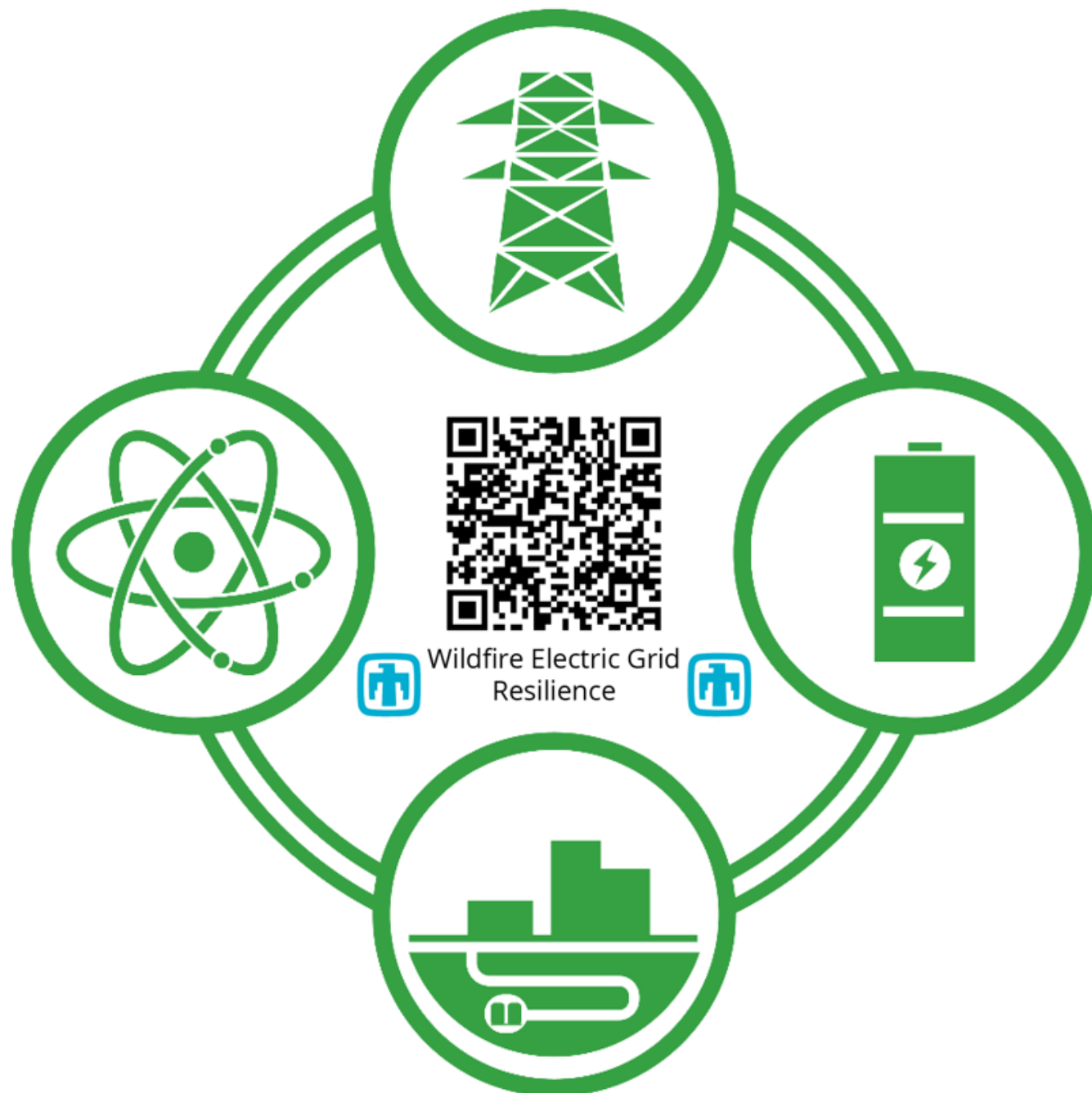
Thank you to the dedicated staff that make these projects successful and impactful

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<https://energy.sandia.gov/programs/electric-grid/wildfire-electric-grid-resilience/>