

Risk-Based Strategies for Managing Wildfire Exposure in a Changing Landscape

Meet the Speakers



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**5 of the top 10
California wildfires
occurred between
2014 and 2017**



**More than 10 million
acres burned in 2017**



**USD 18 billion in North
American wildfire cat
losses since 2014**



- 1. Is wildfire risk increasing?**
- 2. What are my probabilities of loss?**
- 3. Do I understand my risk?**



An aerial photograph of a residential neighborhood. The top half of the image shows several intact, two-story houses with gabled roofs and driveways. The bottom half of the image shows the same area in a state of complete devastation, with only the foundations and scattered debris remaining. The text is overlaid in the center of the image.

No one should be surprised any more.

**The question is not,
“Will it happen?”**

**The question is,
“What are we going to do about it?”**

How Can the Insurance Industry Stay Effective in the Face of Change?

- Changes to market dynamic after 2017 losses
- Increased availability of data and analytics
- Regulatory environment changes



Today's Agenda

Wildfire Behavior and Exposure Growth

How Structure Vulnerability Impacts Risk

Catastrophe Modeling Process and Use Cases

Understanding Accumulated Risk

Putting It All Together

Wildfire Behavior and Exposure Growth

Why Were the 2017 Wildfires Catastrophic?

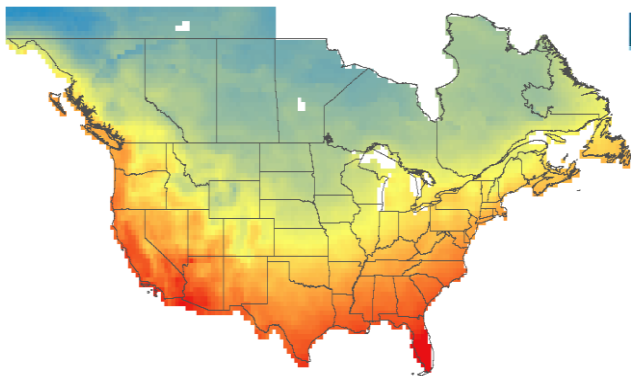


AIR Tubbs Fire Damage Survey 2017

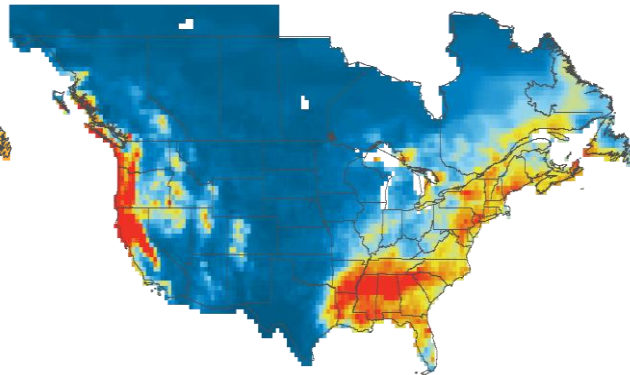
- Fuels resulting from antecedent weather conditions
- Conditions during event
- Wildland-Urban Interface (WUI)

Weather Patterns Drive Variability in Wildfire Activity

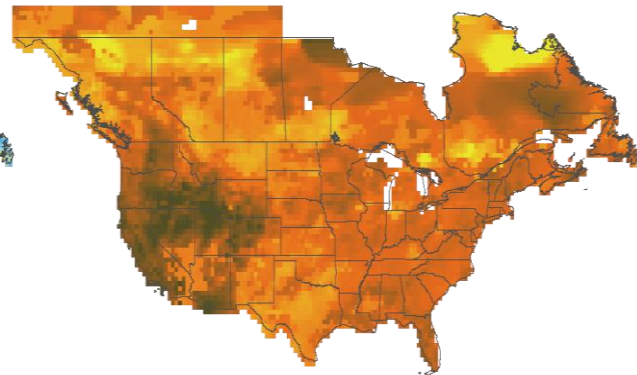
Temperature



Precipitation



Drought



Weather Patterns Drive Fuel Combustibility

Fine Fuels

Fire behavior is correlated to the drought conditions during the **growing** season

Wet Growing Season = **↑** Fire Activity

Dry Growing Season = **↓** Fire Activity



Coarse Fuels

Fire behavior is correlated to the drought conditions during the **fire** season

Wet Fire Season = **↓** Fire Activity

Dry Fire Season = **↑** Fire Activity



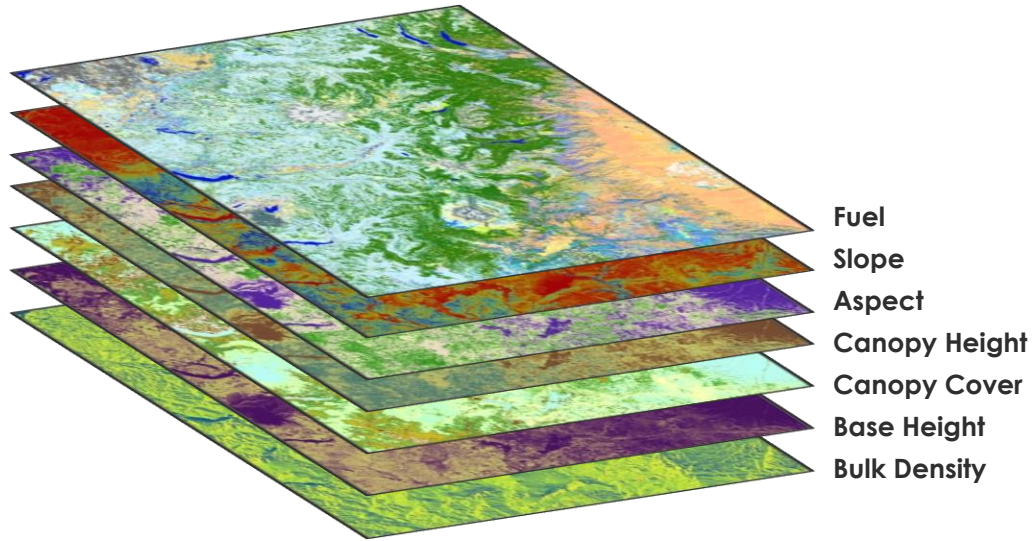
Fire Ignition Risk Patterns Follow Human Presence



Understanding proximity to roads is one piece of quantifying where ignitions are likely to occur

- Humans ignite the vast majority (84%) of wildfires in the U.S.
- Human-ignited fires account for half of annual area burned
- Human-caused ignitions are more prevalent in the WUI and close to the built environment

Wind Speed, Fuel, and Topography Contribute to Intensity



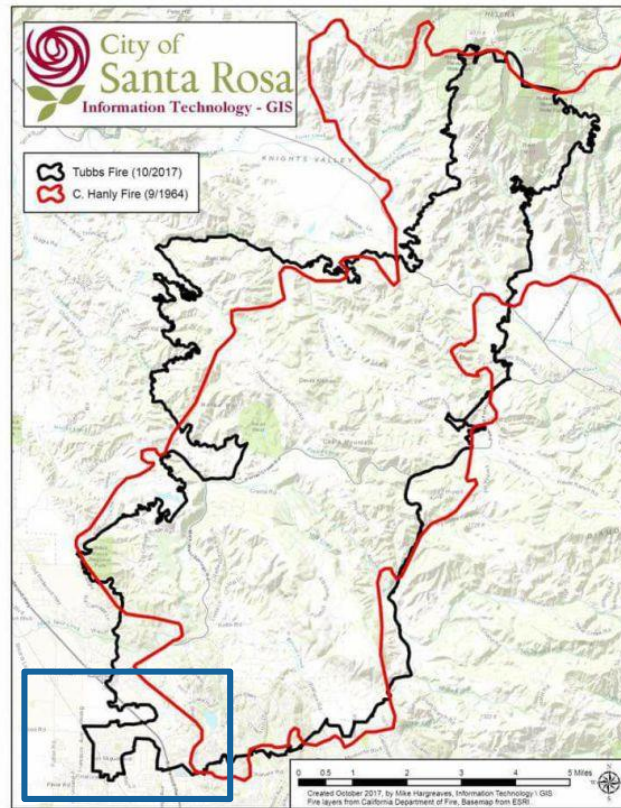
*Fires with high wind speeds
produce higher losses*

The WUI Is Critical to Understanding Wildfire Risk

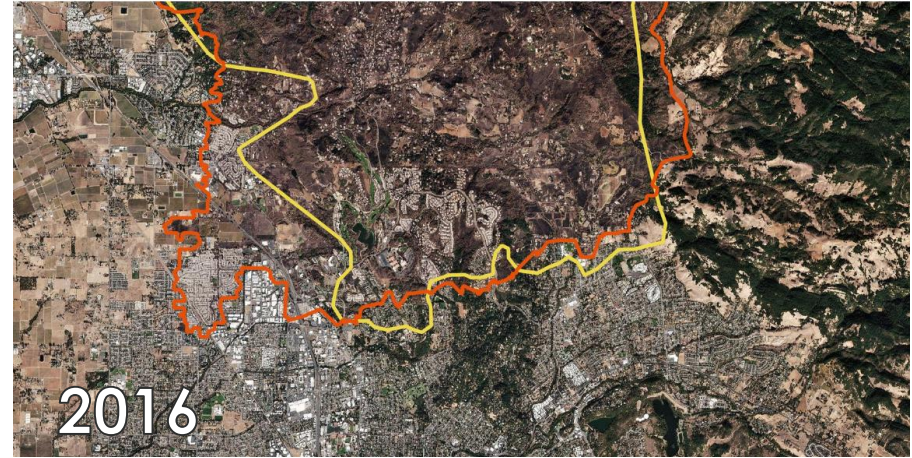
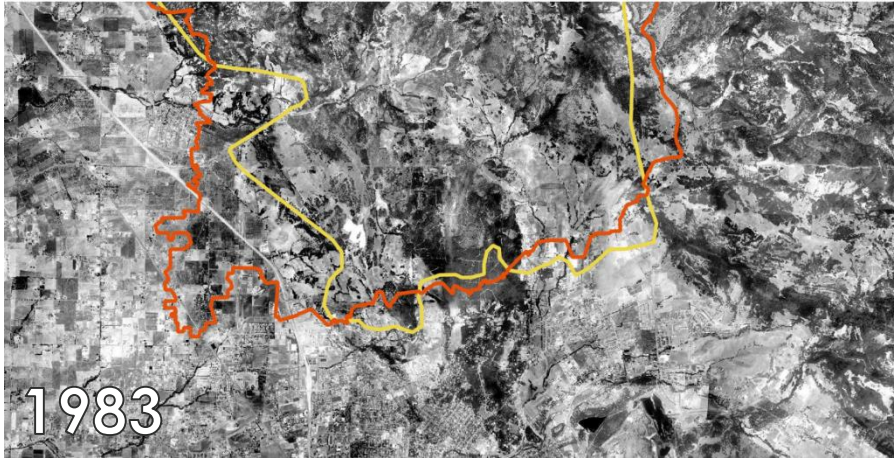


Fountain Grove, AIR Tubbs Fire Damage Survey 2017

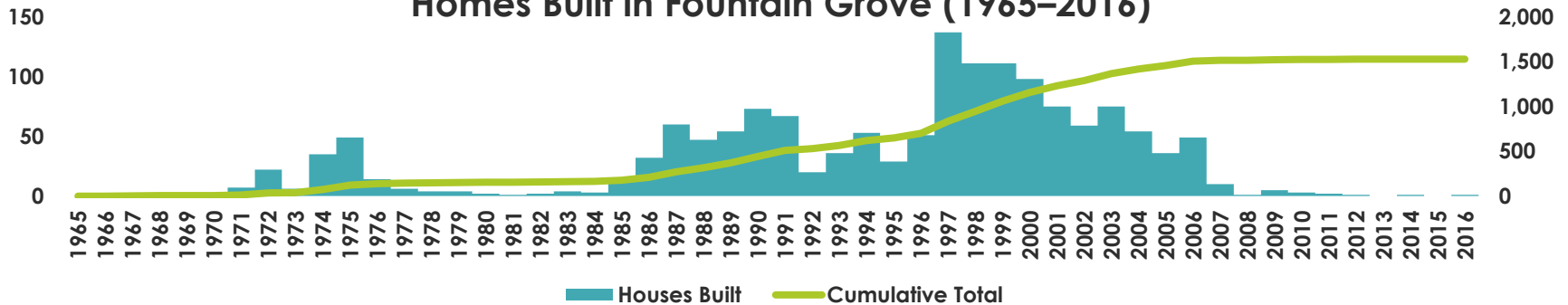
- Residential development intersects with undeveloped wildlands
- Largest economic losses from wildfire will occur in the WUI
 - One-third of U.S. households are located in the WUI
 - 4,000 acres of wildland are converted to WUI daily



Northern California Historically Had Hazard, but Not Risk



Homes Built in Fountain Grove (1965–2016)



Sagara, Eric and Kanik, Alexandra (2018). 'Built to burn.' revealnews.org.

Ember Creation Is the Primary Driver of Structure Ignition

- Material within the fire perimeter is thrown ahead of the main fire front
- Fires can cross roads, rivers, and other natural fire breaks
- Structures are ignited as embers collect on/around them



How Structure Vulnerability Impacts Risk

Drivers of Structure Loss in Wildfire

- Embers, not flames
- Insurance inspections have traditionally focused on fuels, not the structure itself
- Shift from just hazard identification to mitigation and risk reduction



Incorporating Certification

- Certification is effective when homeowner actions are tracked
- Assessment can be conducted by public safety professional, risk manager, or forester
- Feed back into platform to alter underwriting eligibility and portfolio-level risk

Proposed California Tiered Mitigation Standard

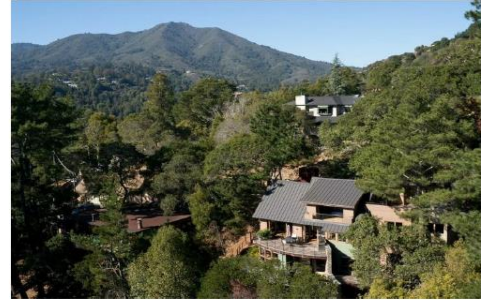
Tier 1: Non-combustible construction with defensible space

Tier 2: Hardened structure with defensible space

Tier 3: Defensible space

CA DOI Perspective

1. Single score to gate underwriting eligibility creates frustration due to inevitable variability in measuring severity at the property level
2. Advocating for required insurance quote or premium credit based on mitigation requirements
3. Open to new data-driven requirements for homeowners to ease availability of insurance in high risk areas

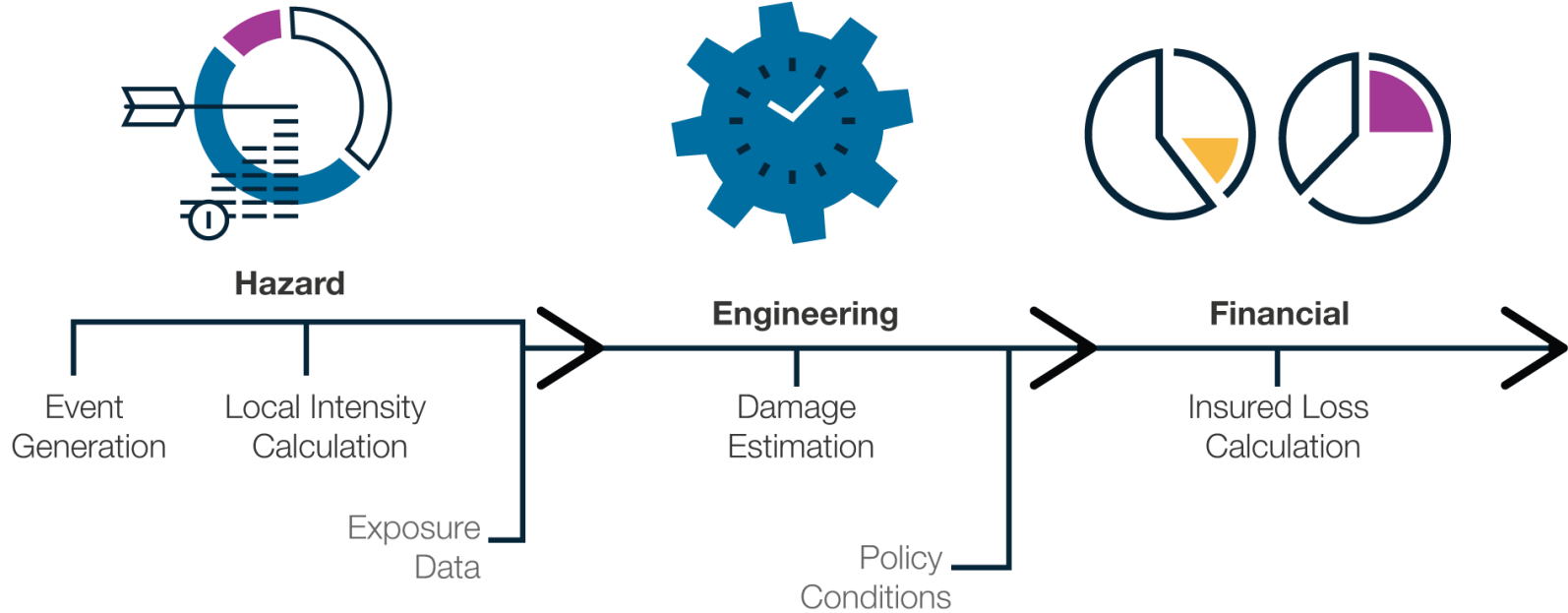


**The Availability and Affordability of Coverage
for Wildfire Loss in Residential Property
Insurance in the Wildland-Urban Interface
and Other High-Risk Areas of California:
CDI Summary and Proposed Solutions**

California Department of Insurance

Catastrophe Modeling Process and Use Cases

What Is Catastrophe Modeling?



Questions Catastrophe Models Are Designed to Answer

An aerial photograph of a large forest fire. Thick white and grey smoke billows upwards from the flames, which are visible as bright orange and red patches on the forest floor. The surrounding trees are mostly green, but some are charred and blackened. The sky is filled with smoke, and the overall scene is dramatic and intense.

What is the probability of a given level of loss in a wide range of catastrophe scenarios?

How intense are they likely to be?

Where are future events likely to occur?

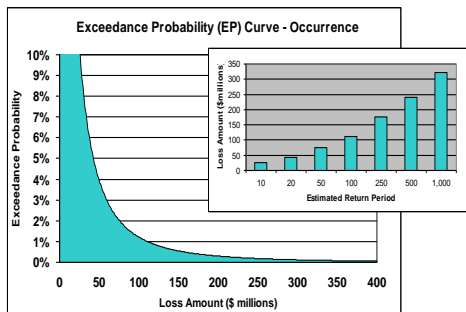
How frequently are they likely to occur?

For each potential event, what is the estimated range of damage and insured loss?

Catastrophe Models Provide a Wide Range of Outputs

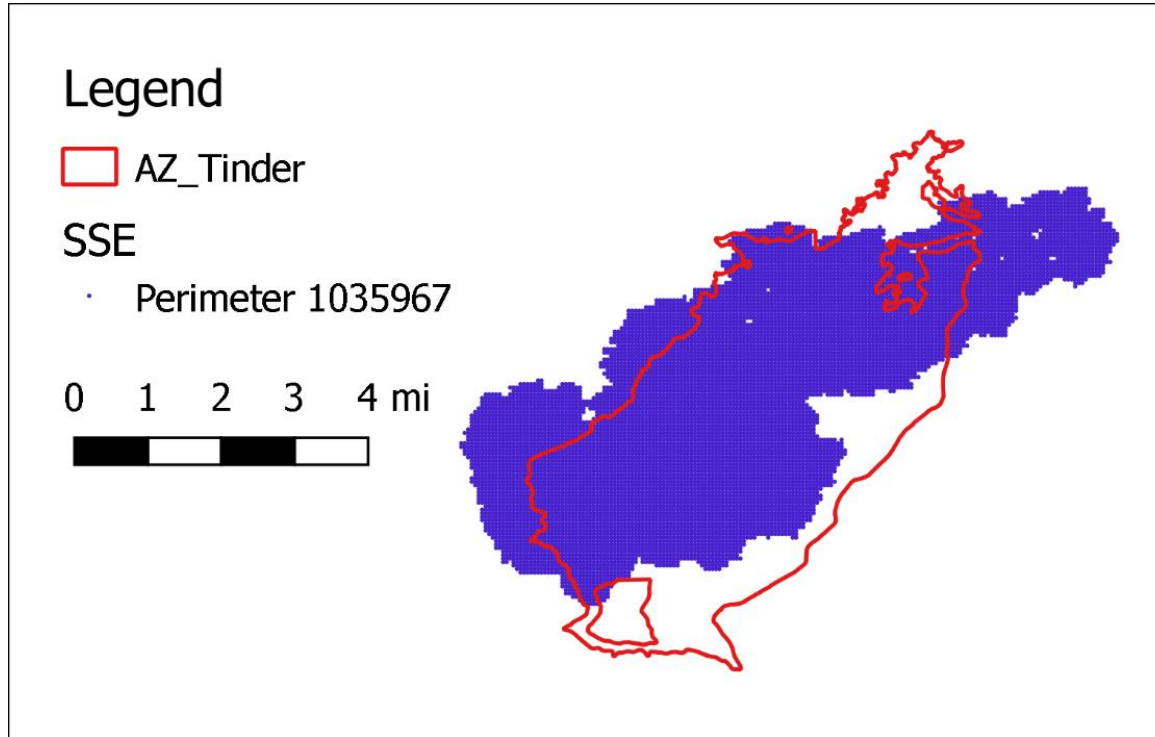
Summary EP Table									
Agg/Occ	Perspective	AAL(EV)	SD	20	50	100	250	500	1,000
AGG	Ground Up	6,630,536	24,753,884	30,886,184	65,842,762	95,130,374	167,601,978	225,760,876	371,719,930
	Retained	4,694,386	17,067,704	22,626,276	41,566,458	66,643,214	104,231,455	161,829,671	244,552,125
	Gross	1,936,150	9,400,331	7,081,307	22,815,924	40,152,431	68,563,509	116,955,551	143,440,838
	Net of Pre-Cat	1,936,150	9,400,331	7,081,307	22,815,924	40,152,431	68,563,509	116,955,551	143,440,838
OCC	Ground Up	6,015,196	23,605,984	27,770,160	60,079,085	89,660,760	162,588,072	217,879,127	344,790,310
OCC	Retained	4,229,041	16,363,640	19,909,887	37,191,259	59,281,633	99,172,459	156,816,491	244,536,739
OCC	Gross	1,824,107	8,987,294	6,675,370	22,202,705	39,498,091	68,519,883	111,110,691	140,808,555
OCC	Net of Pre-Cat	1,824,107	8,987,294	6,675,370	22,202,705	39,498,091	68,519,883	111,110,691	140,808,555

- Exceedance Probability (EP): The probability that a loss will exceed a certain amount in a simulated year
- Average Annual Loss (AAL): The average of all modeled events, weighted by their probability of annual occurrence

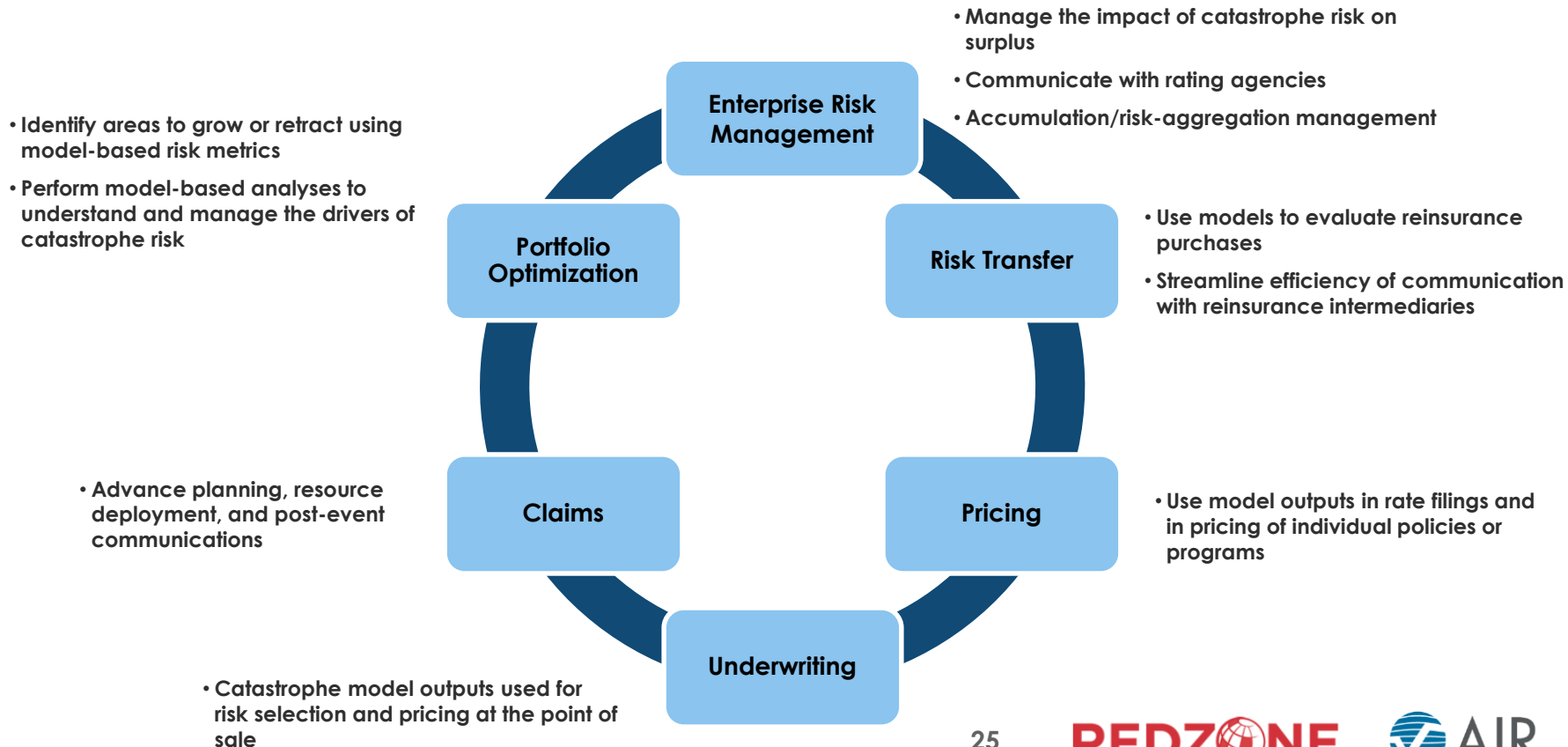


Year	Event ID	Peril	Ground Up	Gross
			Mean	Mean
7046	190252	TC	354,521,220,468	329,990,128,579
7726	208487	TC	352,861,090,276	329,408,630,882
9657	260515	TC	315,533,376,782	290,415,112,308
6636	179421	TC	311,391,560,655	286,924,607,423
5082	137043	TC	288,892,594,064	261,430,695,848
7695	207623	TC	277,575,451,393	257,873,322,806
5227	141054	TC	264,574,317,908	241,334,395,213
2859	76992	TC	254,471,024,991	234,848,485,701
373	9943	TC	251,145,478,732	227,156,239,856
8090	218346	TC	241,426,505,957	218,192,035,523
2956	79565	TC	242,405,569,550	216,834,993,072

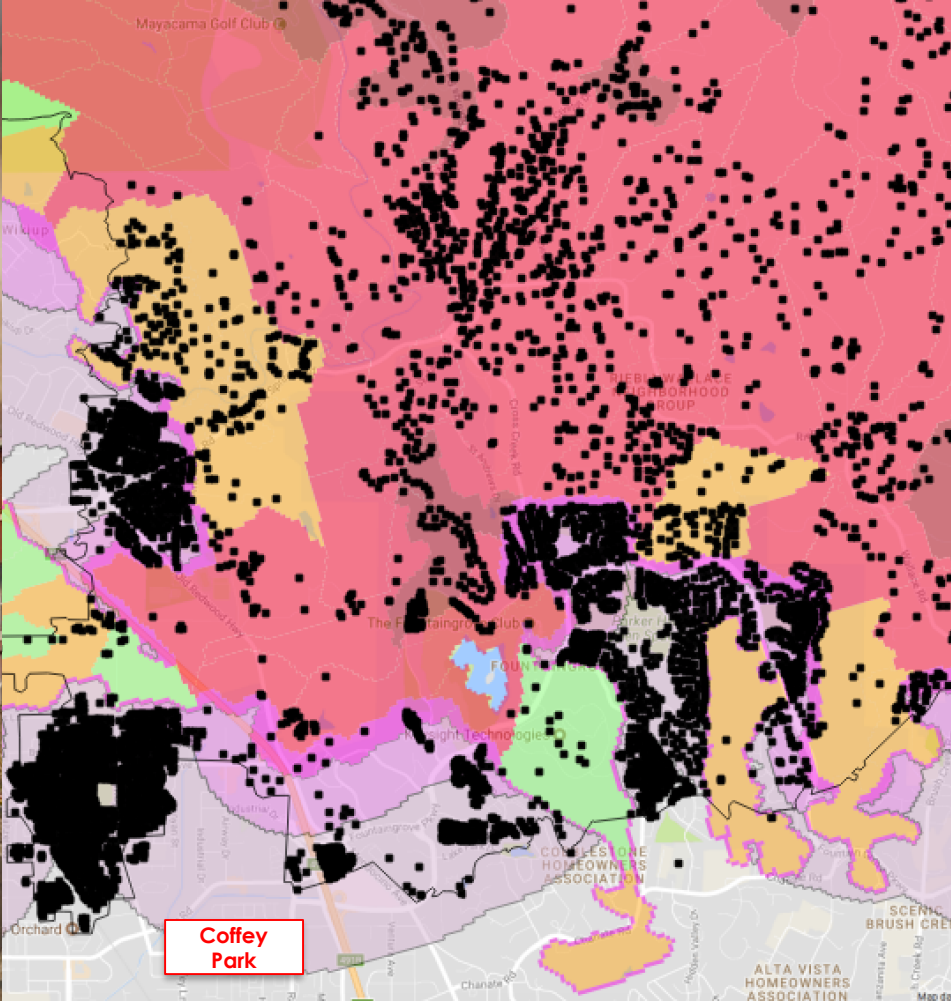
Similar Stochastic Events (SSEs) Support Understanding of Hazard Concentration and Real-Time Loss Estimates



How (Re)insurers Use Model Output

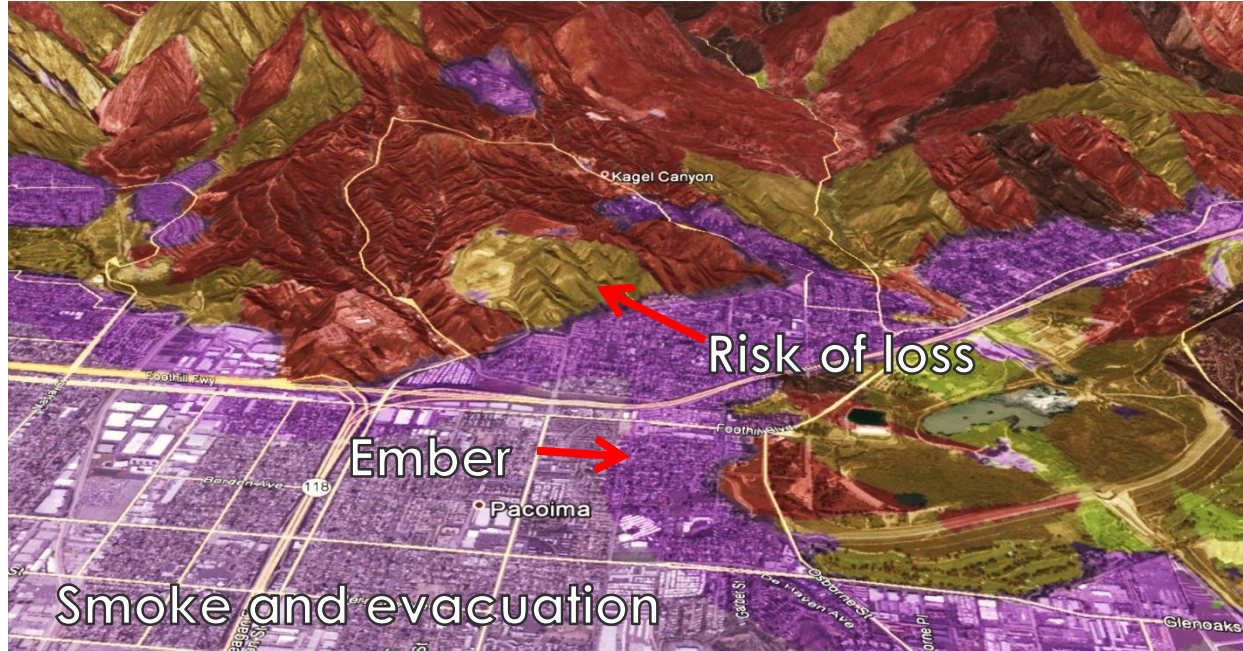


Understanding Accumulated Risk



- 50% of the structure losses in the Tubbs Fire were in urban areas
- Accurately classifying the WUI ember risk dramatically improved these estimates: 96% of the burned homes were classified as *at risk*

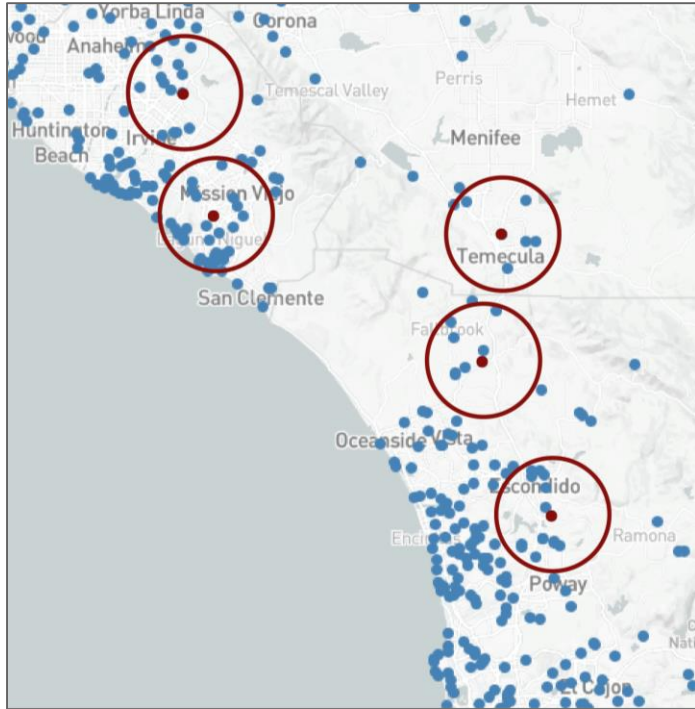
Understand the Comprehensive Risk



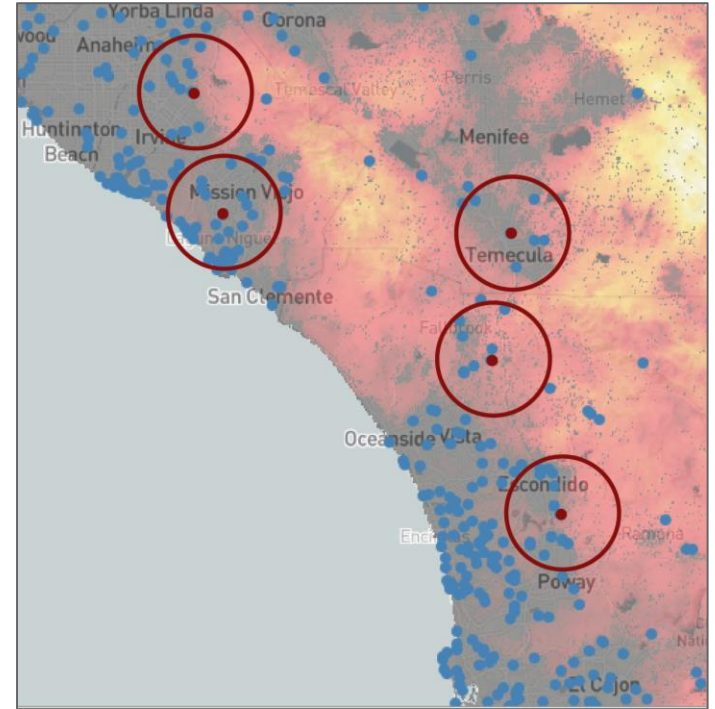
- Incorporate fire frequency and severity for more accurate assessments of loss potential
- Identify potential for ember showers and structure-to-structure ignition
- Provide an overall score and AALs to inform underwriting decisions

Accumulation by Proximity vs. Risk

Accumulated liability using policy proximity

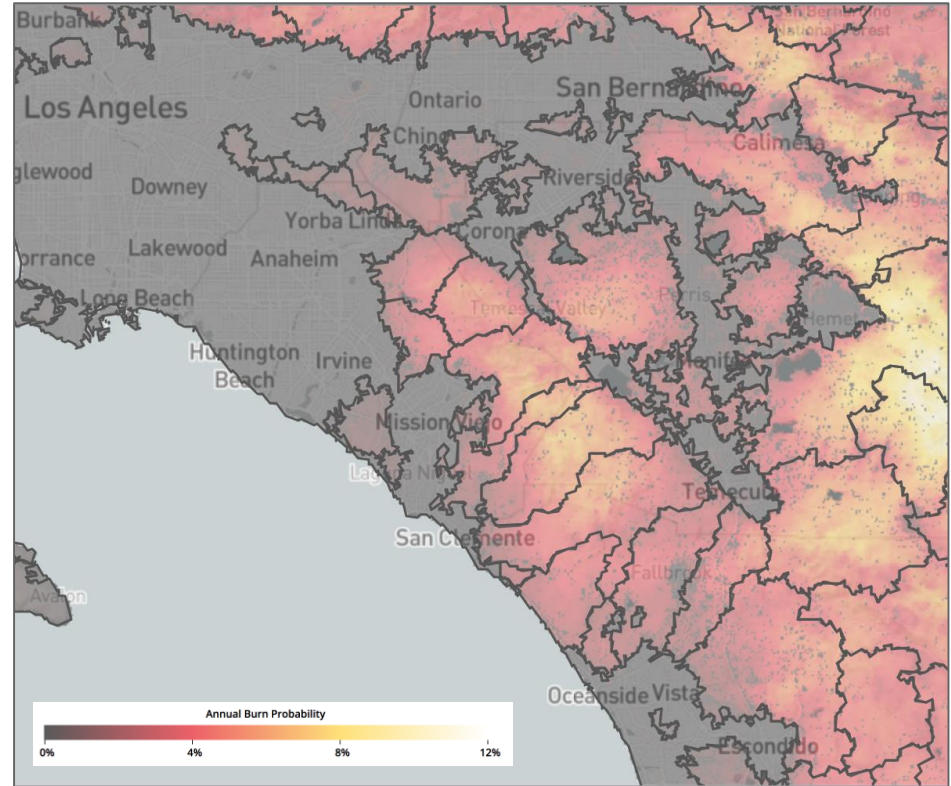


Proximity-based zones on wildfire risk map



Defining Correlated Risk Zones

- 10,000 years of wildfire simulations
- 25 years of historical perimeters
- Ember risk and hazard data for the WUI



Correlated Risk Zones: Context for “Dots on a Map”

Zone Analysis

274

Policy Count

\$345m

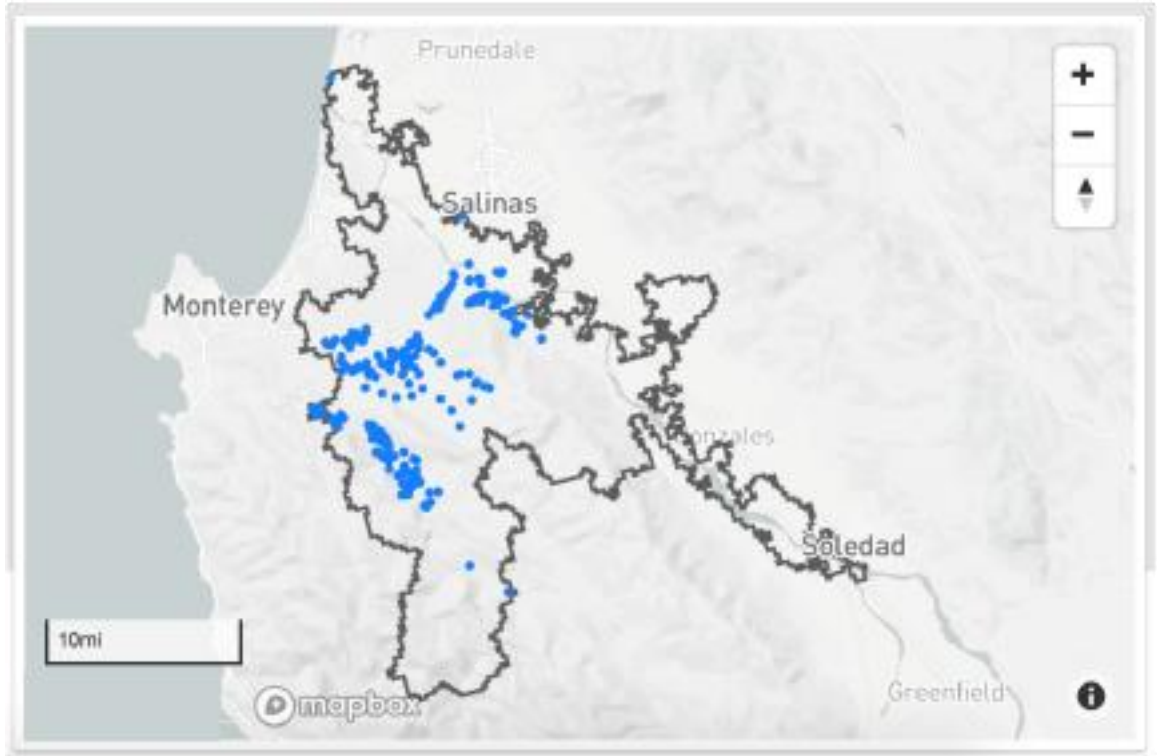
Total Insured Value

1.97%

Max Burn Probability

\$2.1m

Probable Maximum Loss



Putting It All Together

Applying New Tools to Wildfire Risk Management

- ✓ Physically-based catastrophe modeling
- ✓ Risk-based accumulation modeling
- ✓ Underwriting support
- ✓ Portfolio management and optimization
- ✓ Targeted decision support