

That Cost Me What!?

Demonstrating the Need and Utility of Catastrophe Models in
Quantifying Severe Thunderstorm Risk

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What is a Severe Thunderstorm?

AMS Glossary:

In general, any destructive storm, but usually applied to severe local storms in particular, that is, intense thunderstorms, hailstorms, and tornadoes.

Primary Sub-perils



Straight-line Wind (> 50 knots)



Hail (>=1" in diameter)



Tornado (EF0-EF5)

What is a Severe Thunderstorm?

Supercells



- Rotating, isolated
- Typical sub-perils
 - Tornado
 - Hail
 - Wind/downbursts
- Typical dimensions
 - Duration: hours
 - Spatial: ~100-1,000s of km²

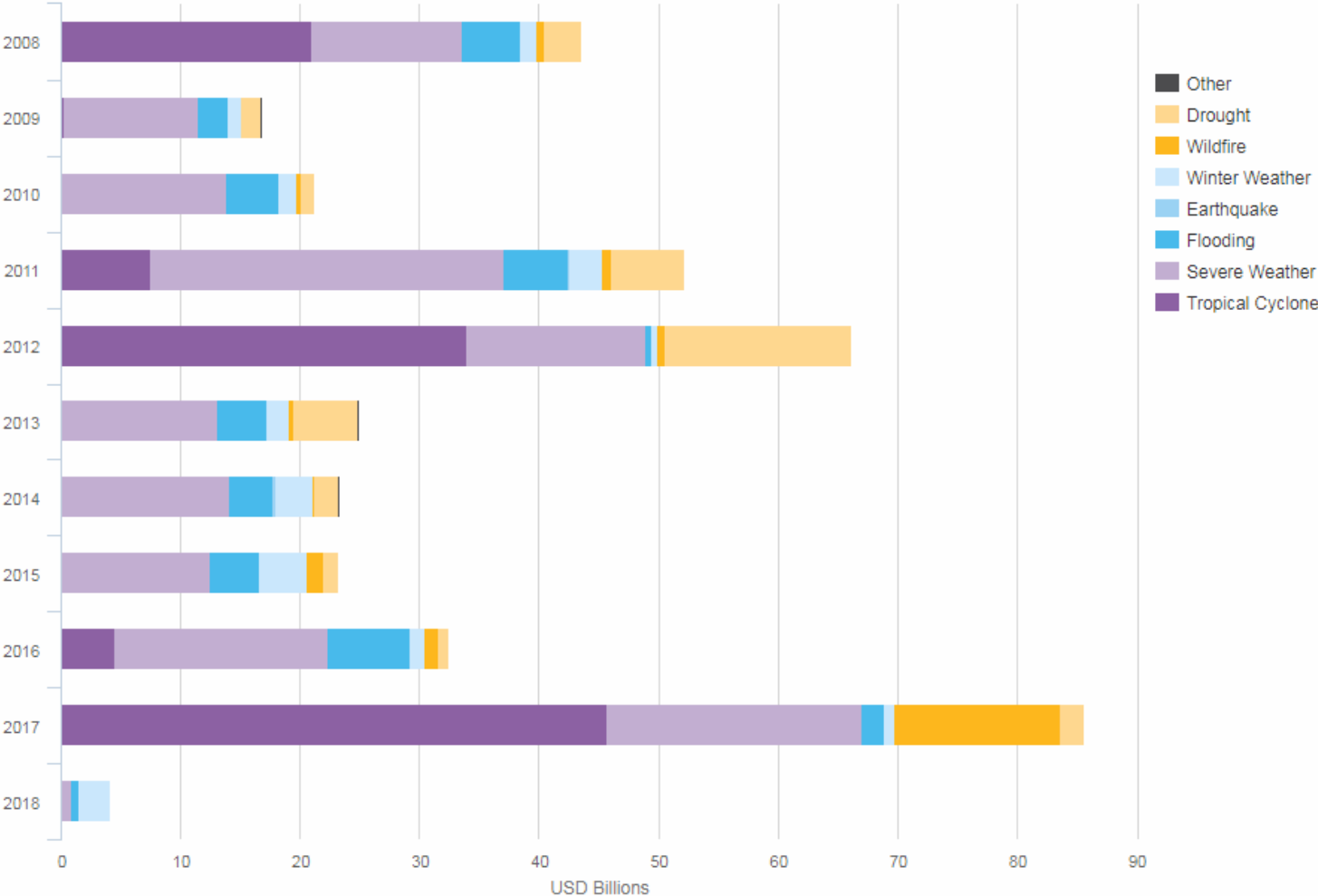
Multicells (Squall Lines, Derechos)



- Long-lasting
- Typical sub-perils
 - Wind/downbursts
 - Hail
 - Tornado
- Typical dimensions
 - Duration: days
 - Spatial: ~10,000-100,000s of km²

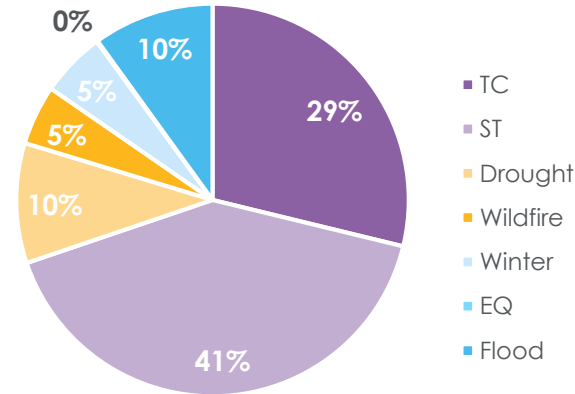
Severe Thunderstorms Losses can Rival Those Risks of More “Traditional” Concerns...

US Insured Losses



- 7/10 Years ST tops US Annual Insured Losses
- 4/10 with 15+ Billion in Loss

US Catastrophe Losses (Bln USD) 2008-2018



Source: AON Benfield Catastrophe Insight



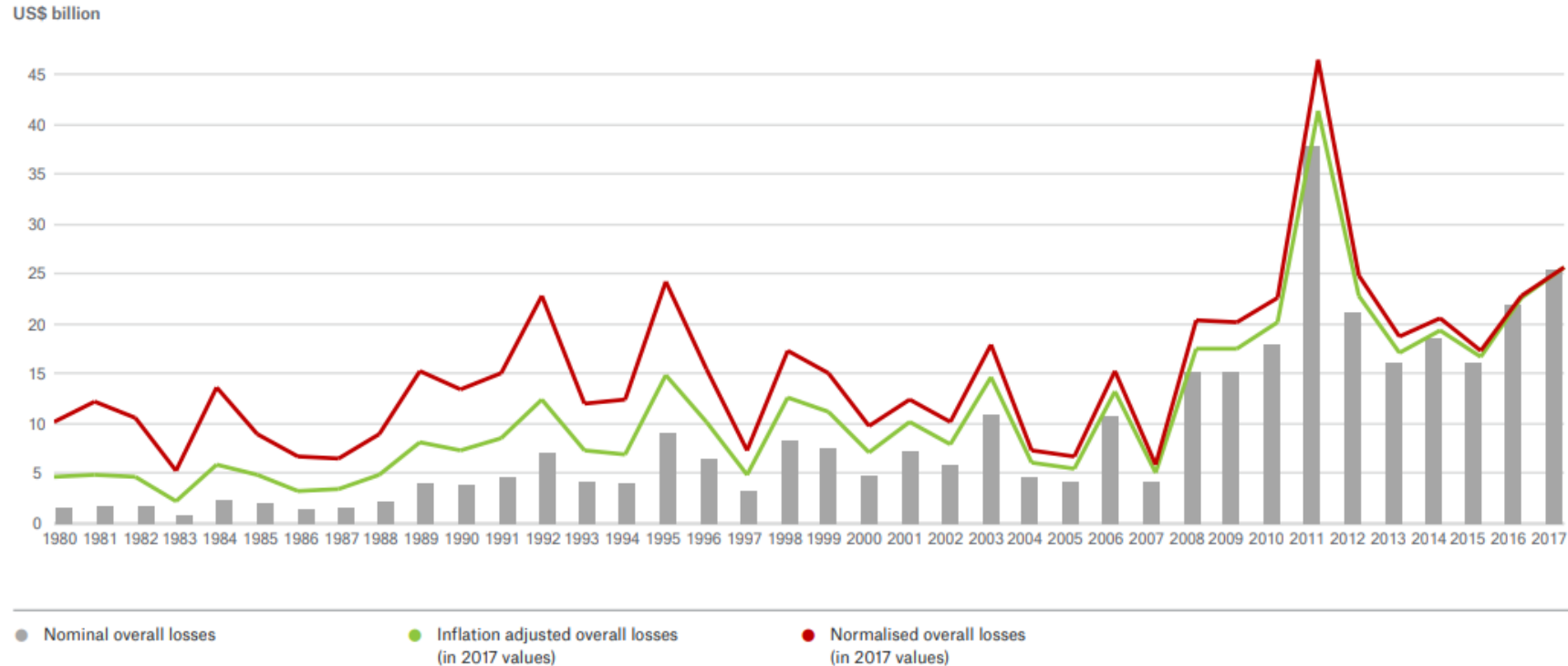
... And these Losses have Been Steadily Increasing Over the Last 30 Years

NatCatSERVICE



Overall losses in US\$: nominal, inflation adjusted, and normalised

Relevant convective storm events
in the United States 1980 - 2017

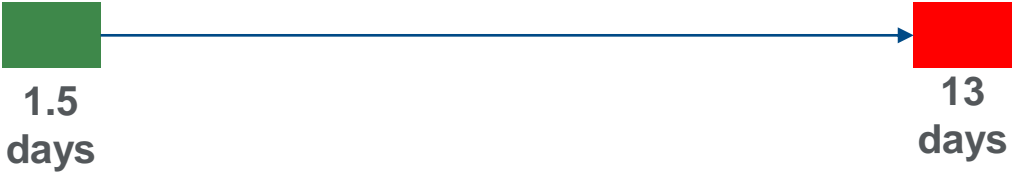
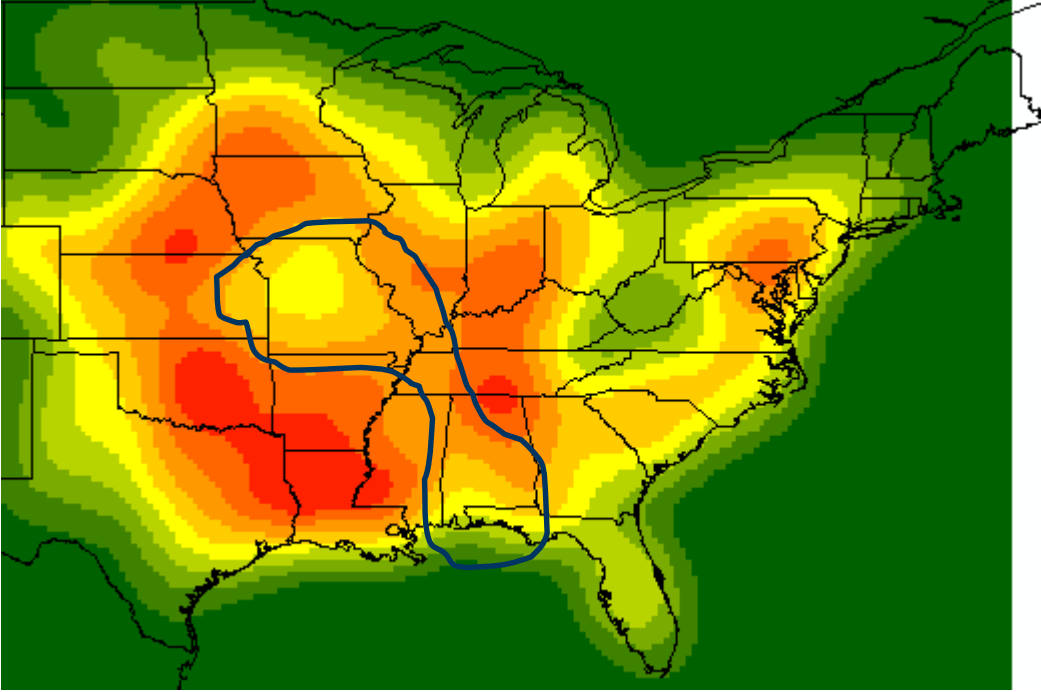


Inflation adjusted via country-specific consumer price index and consideration of exchange rate fluctuations between local currency and US\$.
Normalization via local GDP developments measured in US\$.

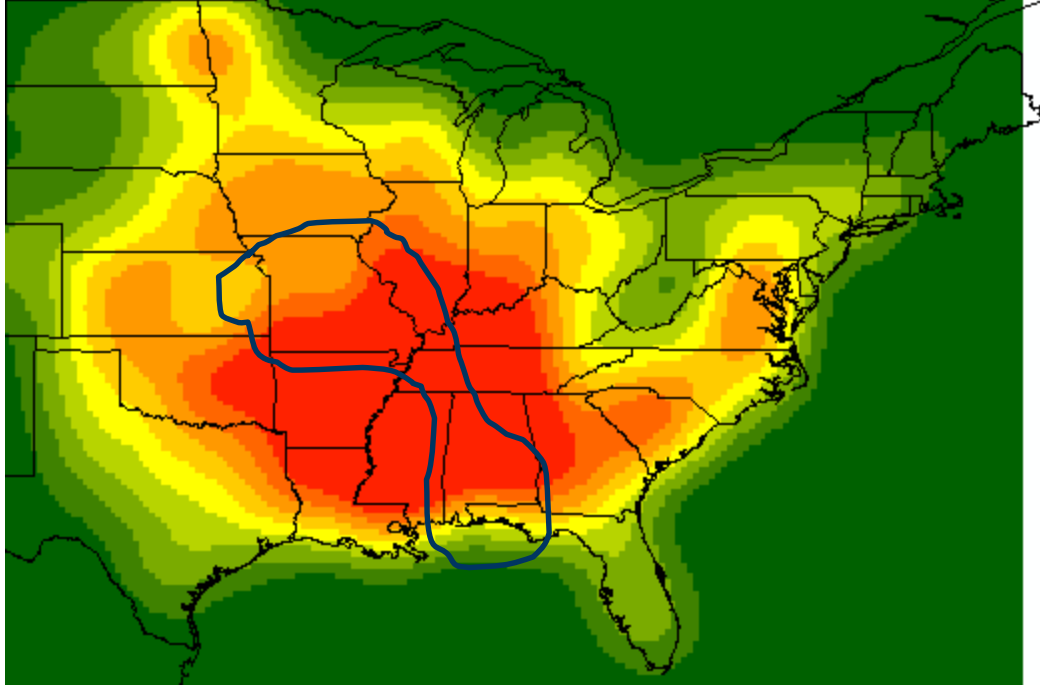


Thunderstorms Also Exhibit High Degrees of Year-to-Year Variability...

Average Annual Tornado Days
1990-2001

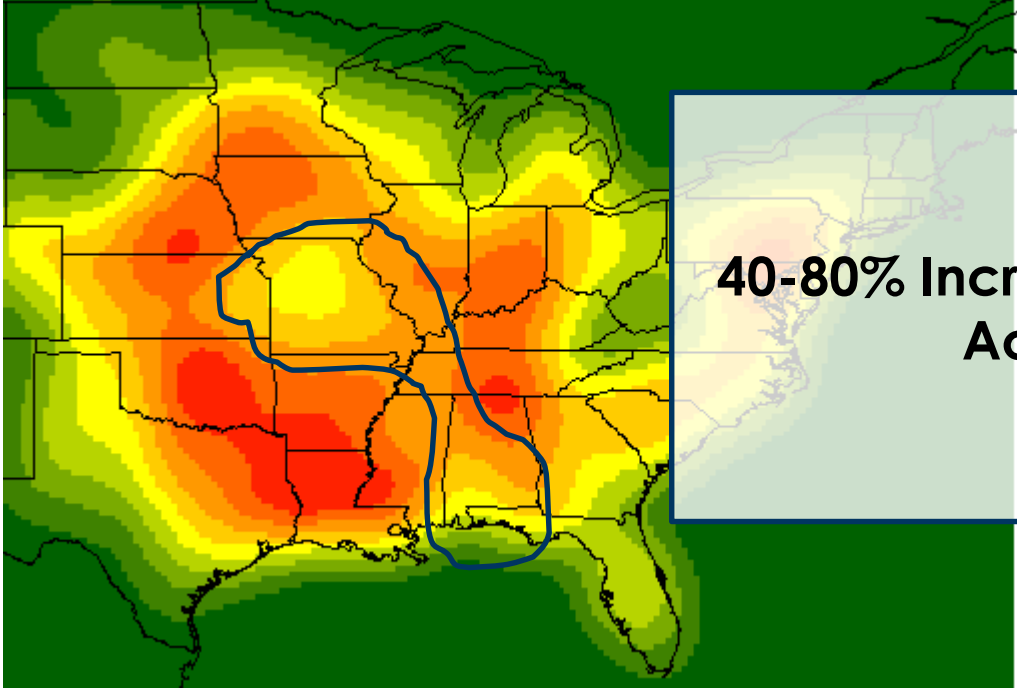


Average Annual Tornado Days
2000-2011

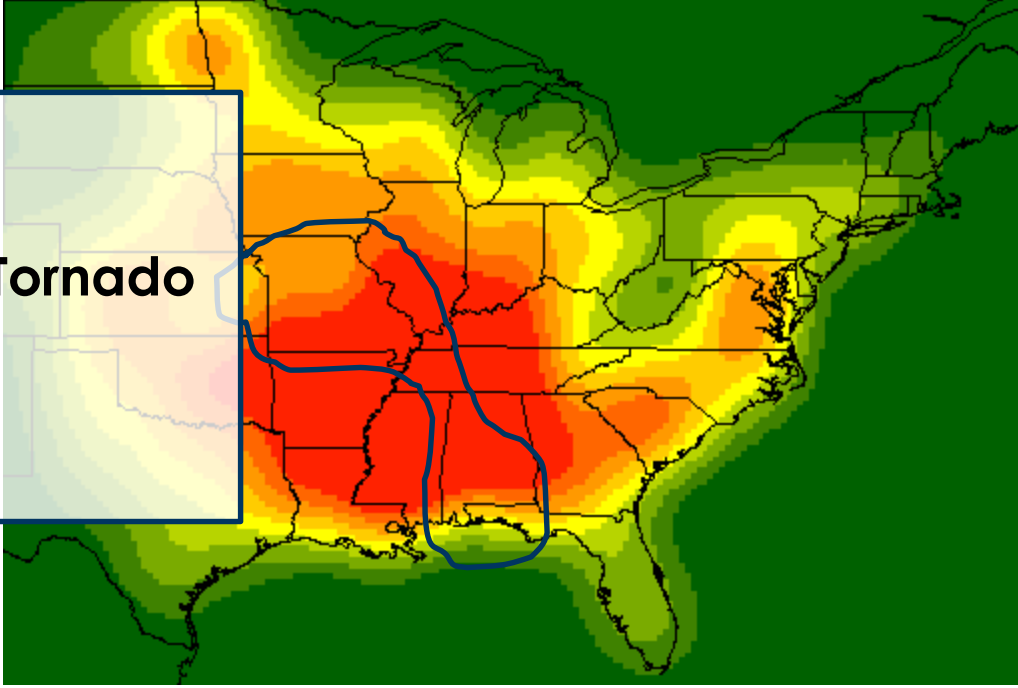


Thunderstorms Also Exhibit High Degrees of Year-to-Year Variability...

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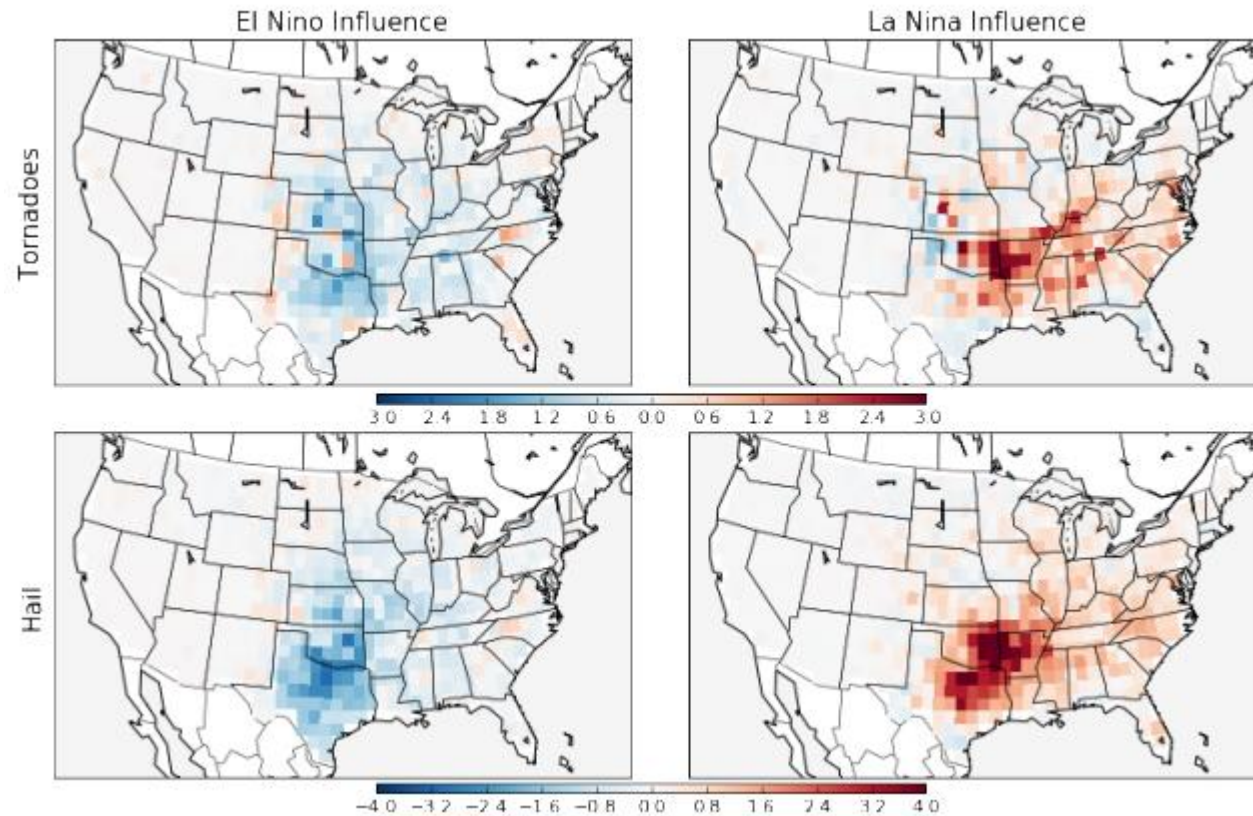


40-80% Increase in Tornado Activity!!



... That is Potentially Affected by Changing/Cyclical Climate Conditions

ENSO's Effect on Severe Thunderstorm Activity



But it's more than just ENSO!!

Pacific Decadal Oscillation (PDO)

North Atlantic Oscillation (NAO)

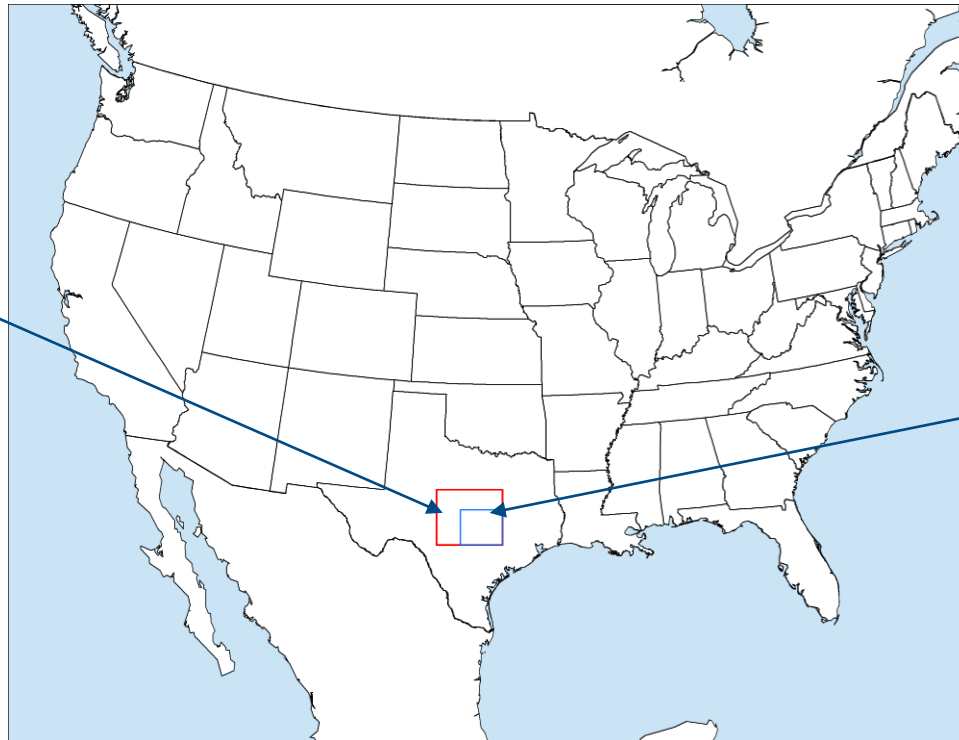
Pacific-North American Pattern (PNA)

Atlantic Multi-Decadal Oscillation (AMO)

...And Covers Relatively Small Areas

If I took EVERY tornado recorded for the last 70 years,
and placed them on the US without overlap:

1.3% of the
Eastern 2/3 of
CONUS

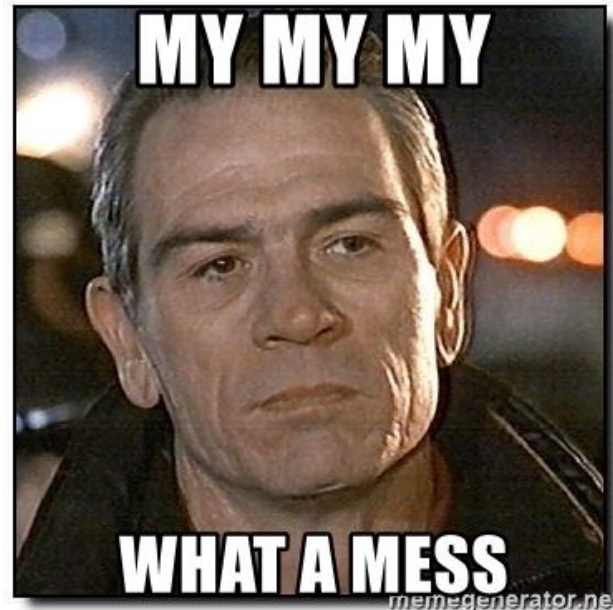


Limiting to the
last 20-yr's and
EF1+

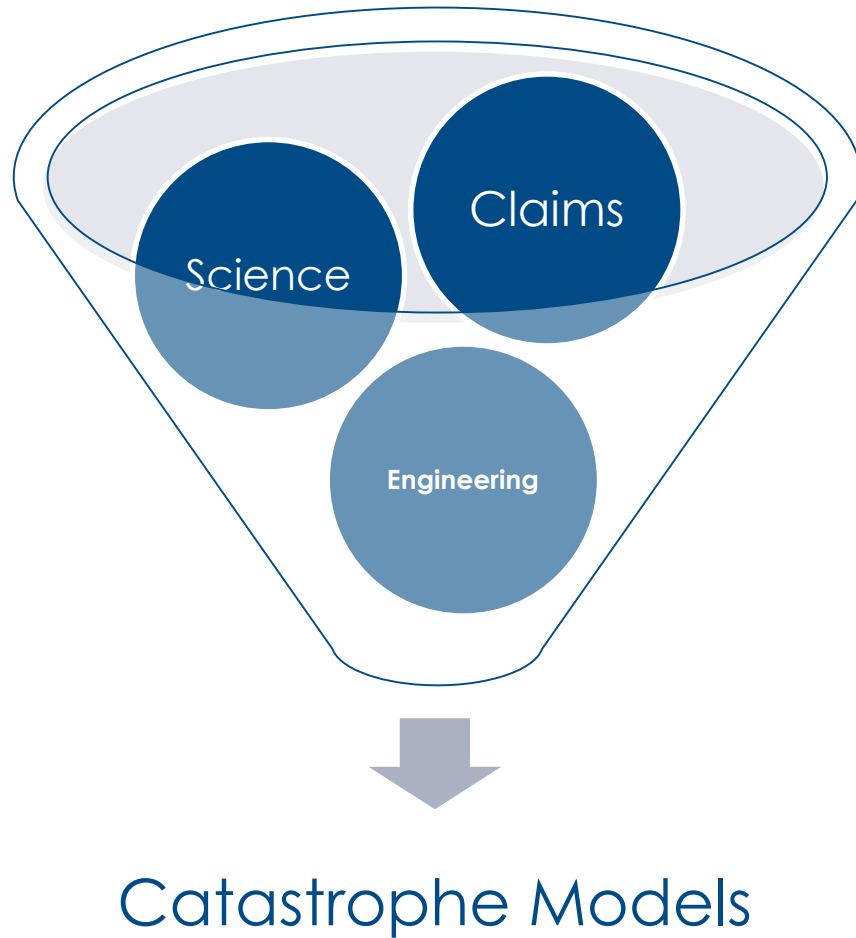
.5% of Eastern 2/3 of
CONUS

So What Does 2016-2017 (and Other Recent Years) Say About ST Risk???

- A Non-stationary, highly variable, potentially cyclical, spatially correlated, ill-observed phenomenon that can cause 10s of billions of dollars in Insured Losses per year
- In other words...

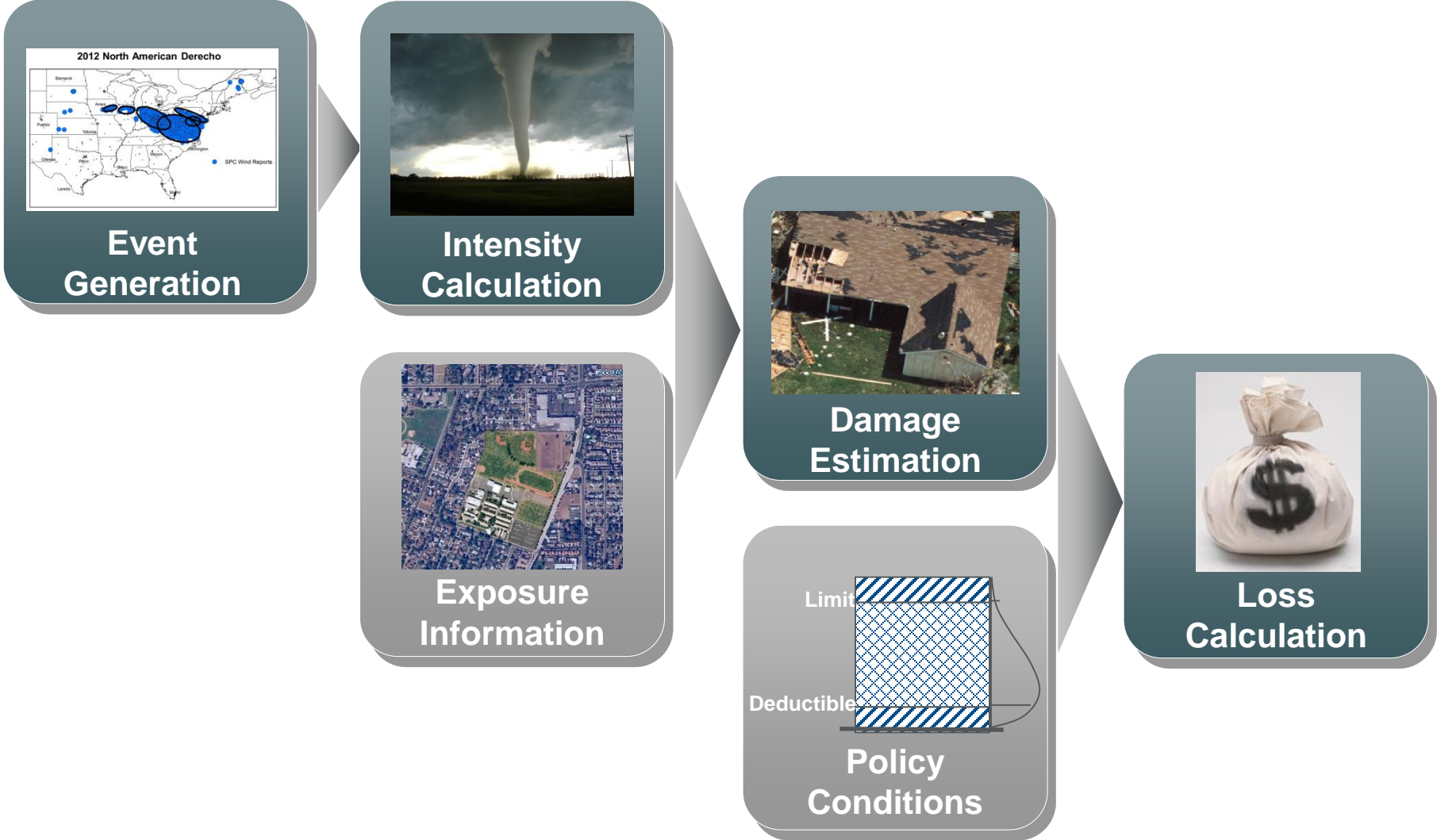


Don't Worry! The Models Can Help!

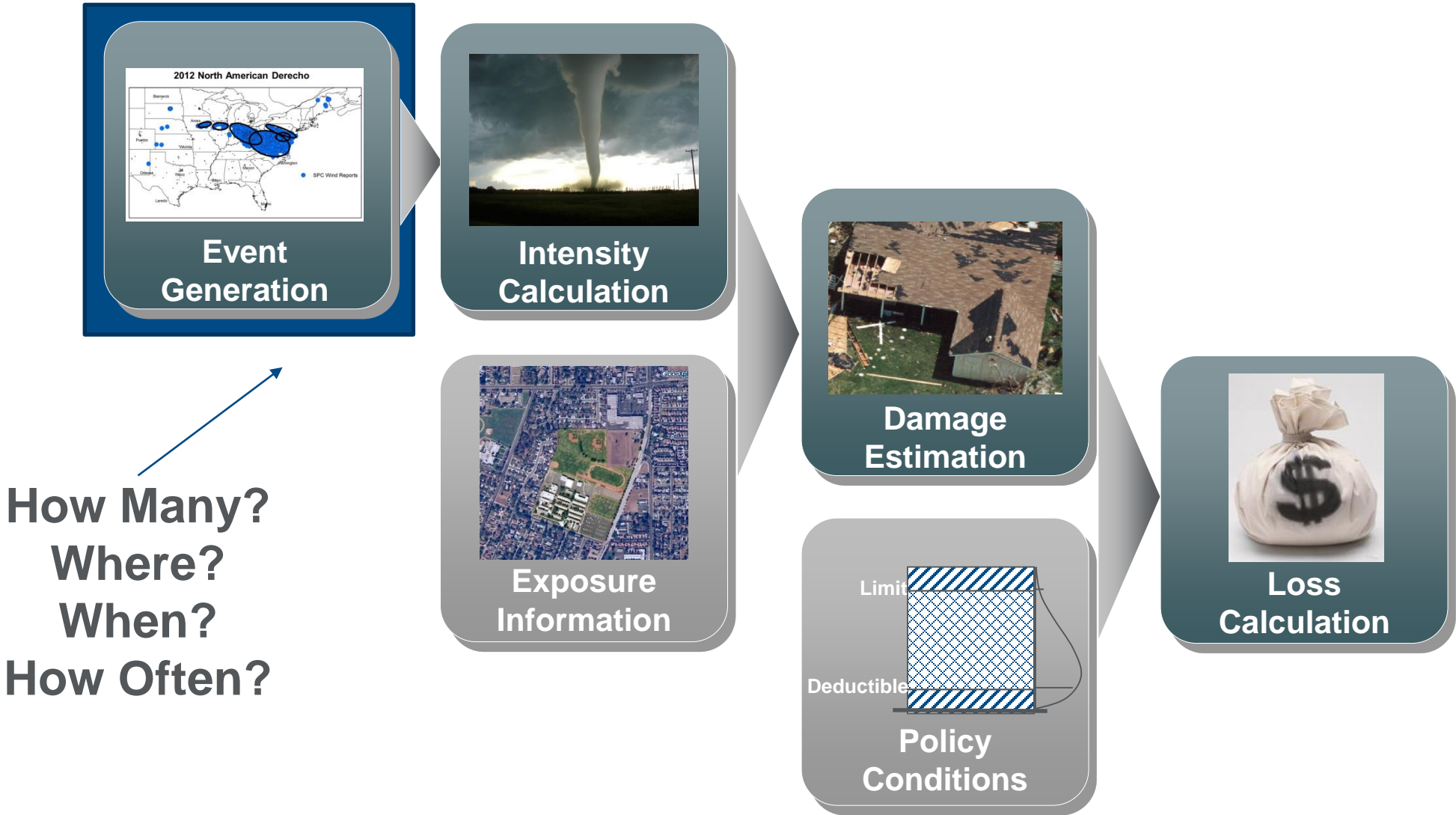


- Leverage multiple datasets to extend our “observational” dataset
 - Helps reduce variability and uncertainty
- Leverage engineering and science to differentiate risks in a robust way
- Test sensitivities to various parameters (e.g. missing data)

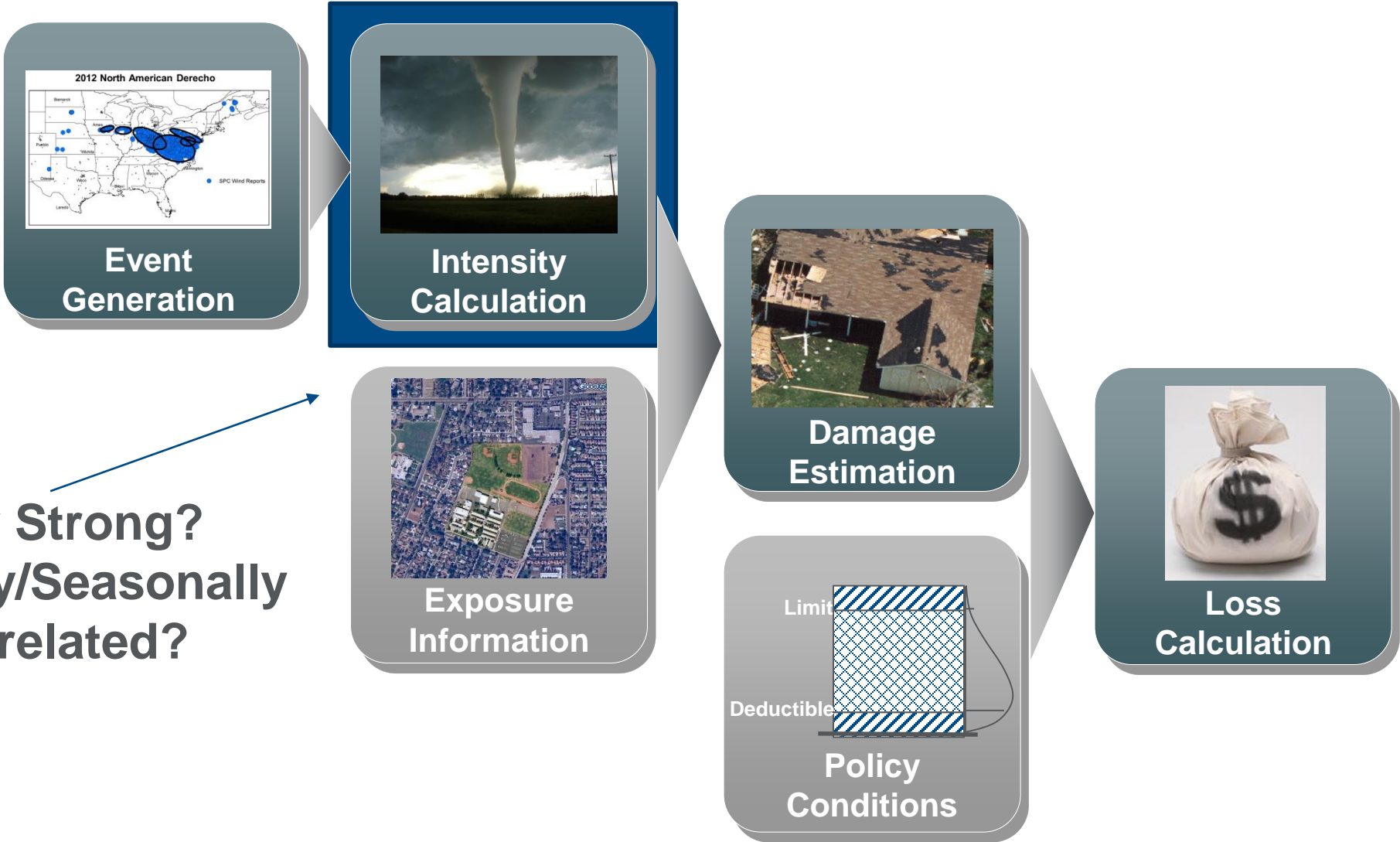
The Basic CAT Modeling Framework



The Basic CAT Modeling Framework

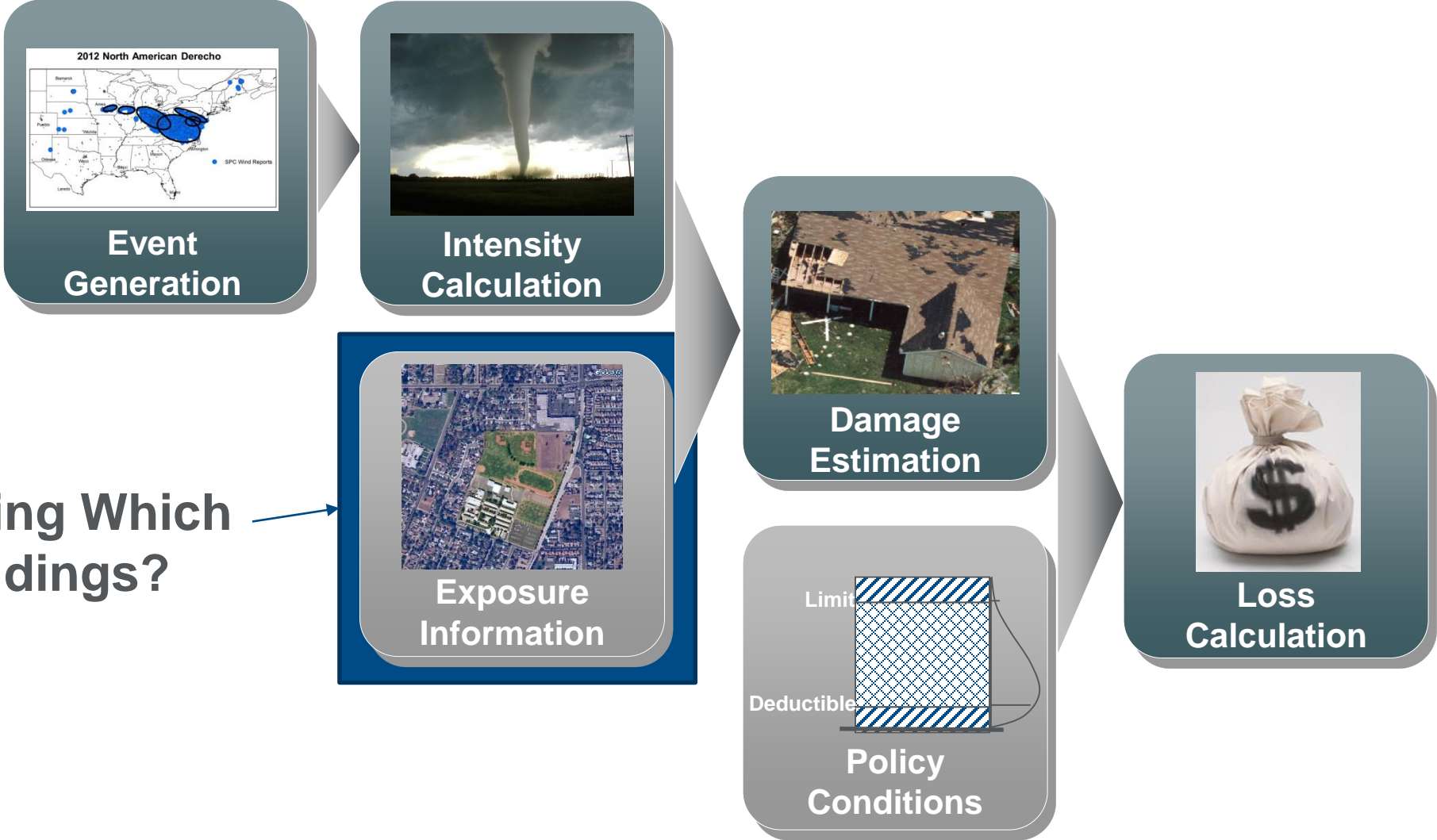


The Basic CAT Modeling Framework

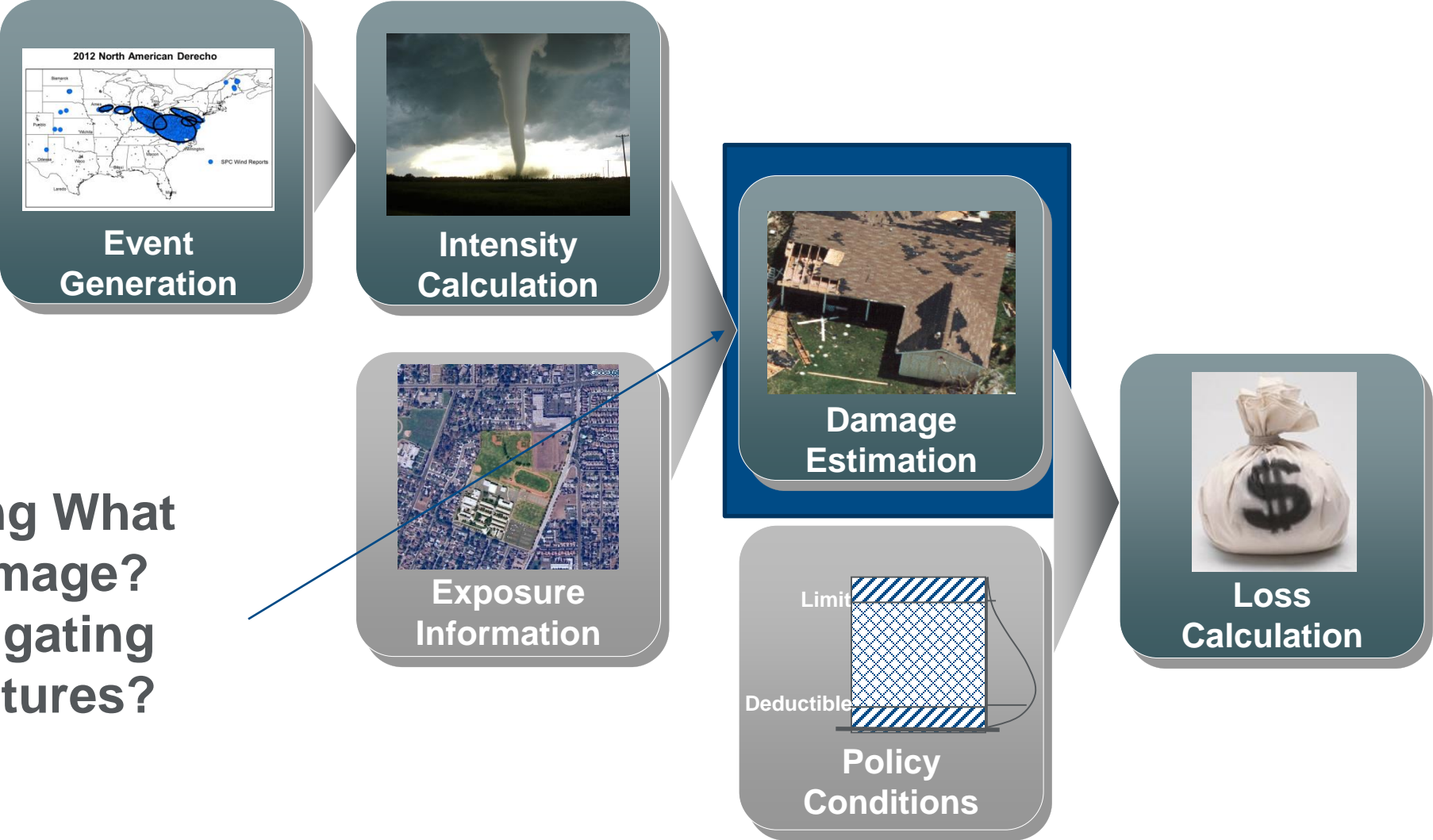


How Strong?
Spatially/Seasonally
Correlated?

The Basic CAT Modeling Framework

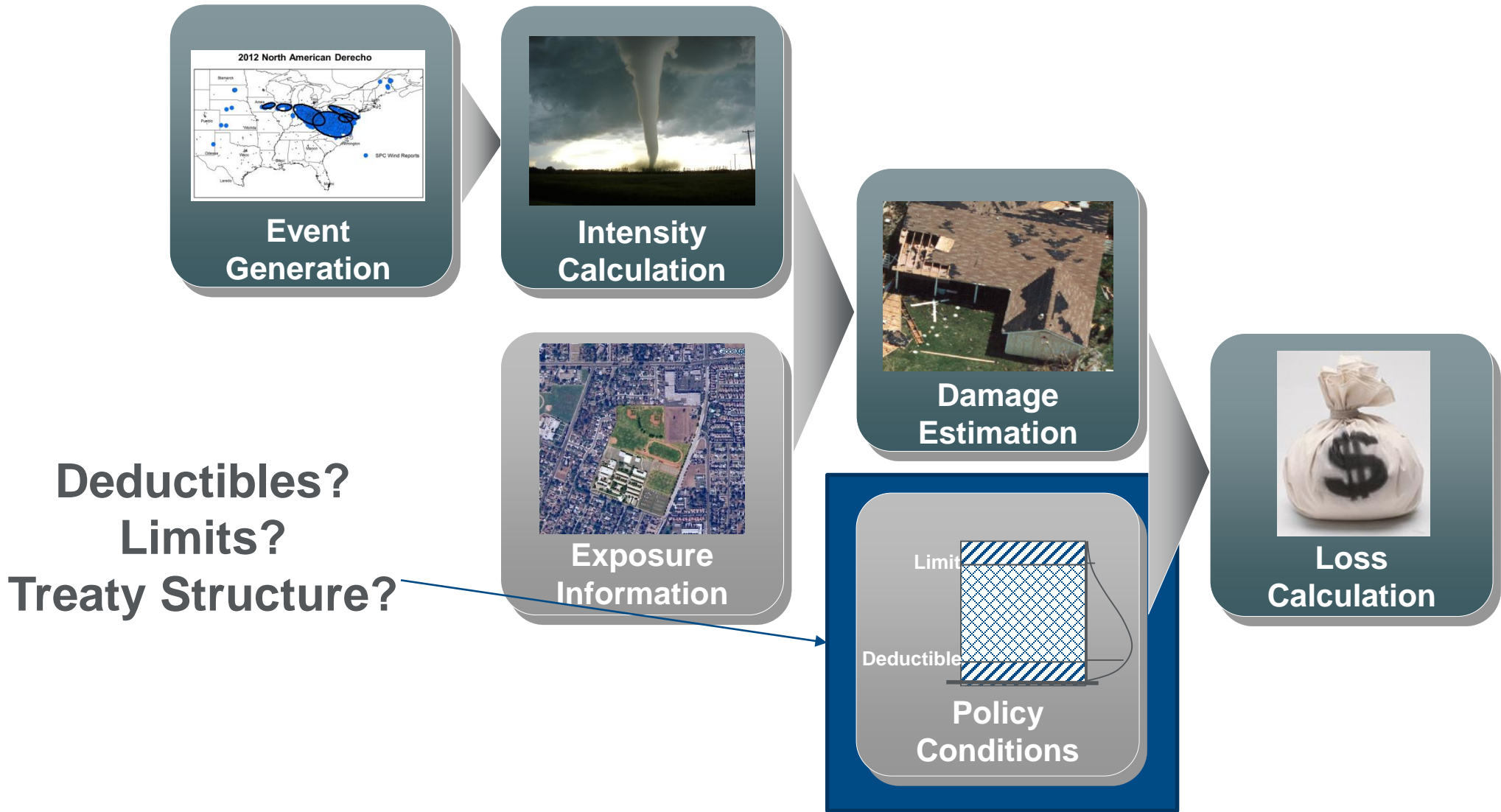


The Basic CAT Modeling Framework



Doing What
Damage?
Mitigating
Features?

The Basic CAT Modeling Framework



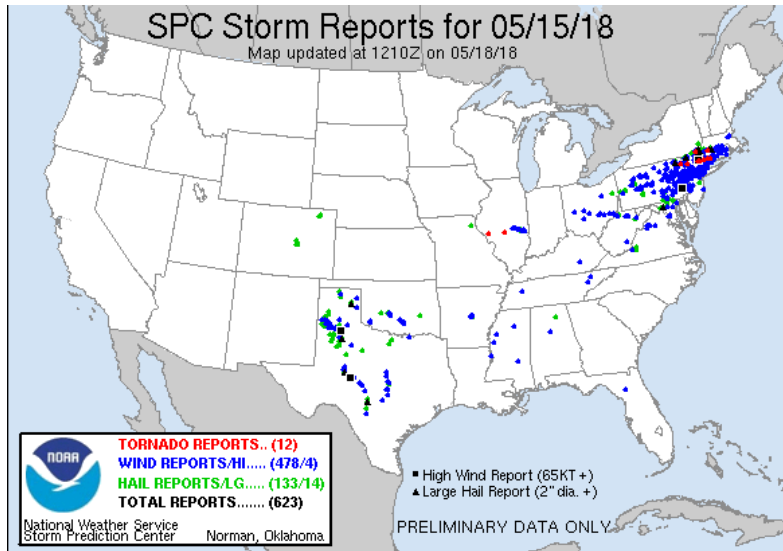
The Basic CAT Modeling Framework



Severe Thunderstorm Risk

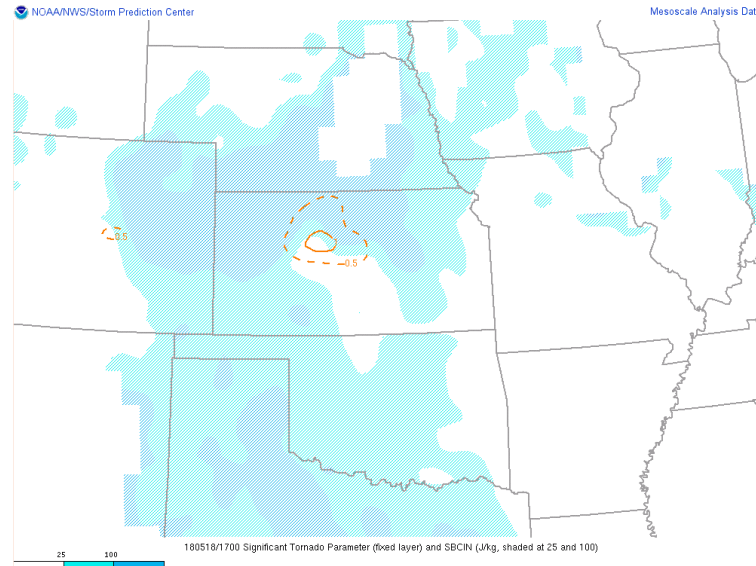
Event Generation: What Data Sources Do We Have to Quantify the Risk?

Eye-witness Reports



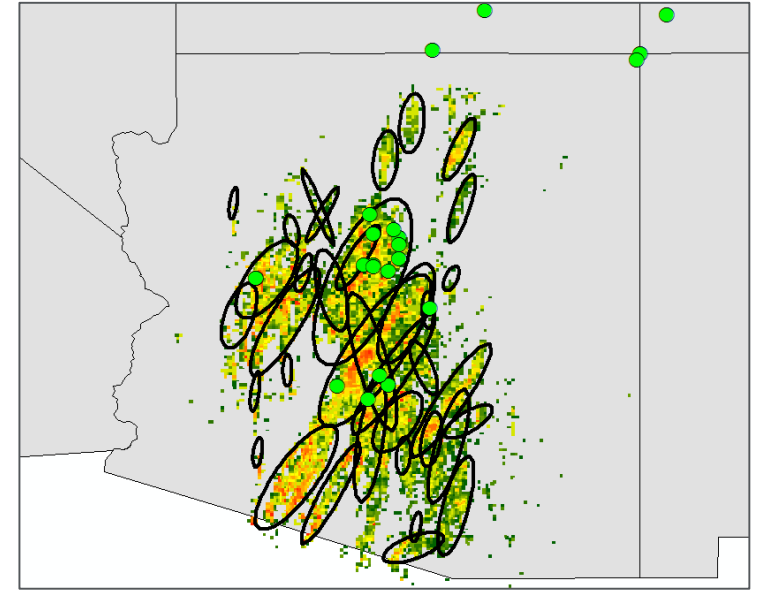
Ground "Truth" But Biased by Population Changes

Weather Data



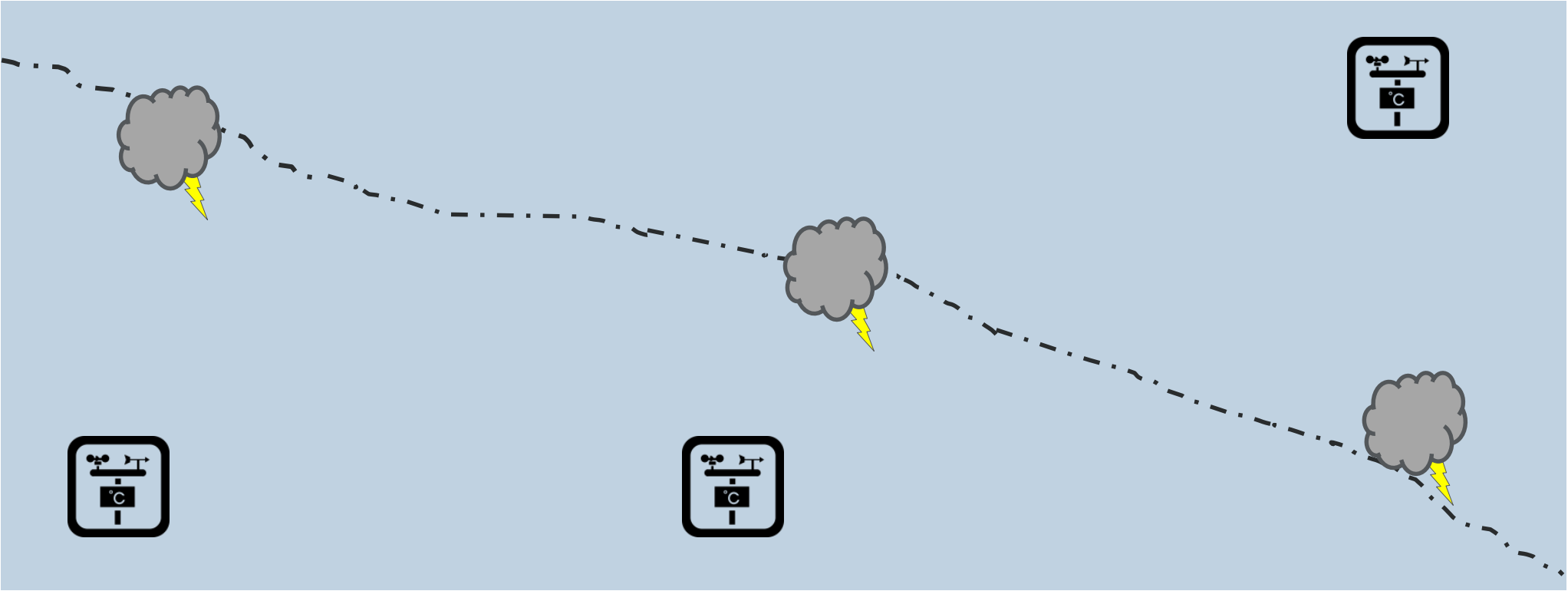
Temperature, Humidity, Wind Speed
BUT TYPICALLY not specific to
wind/hail/tornado

Other Datasets (e.g. Radar)



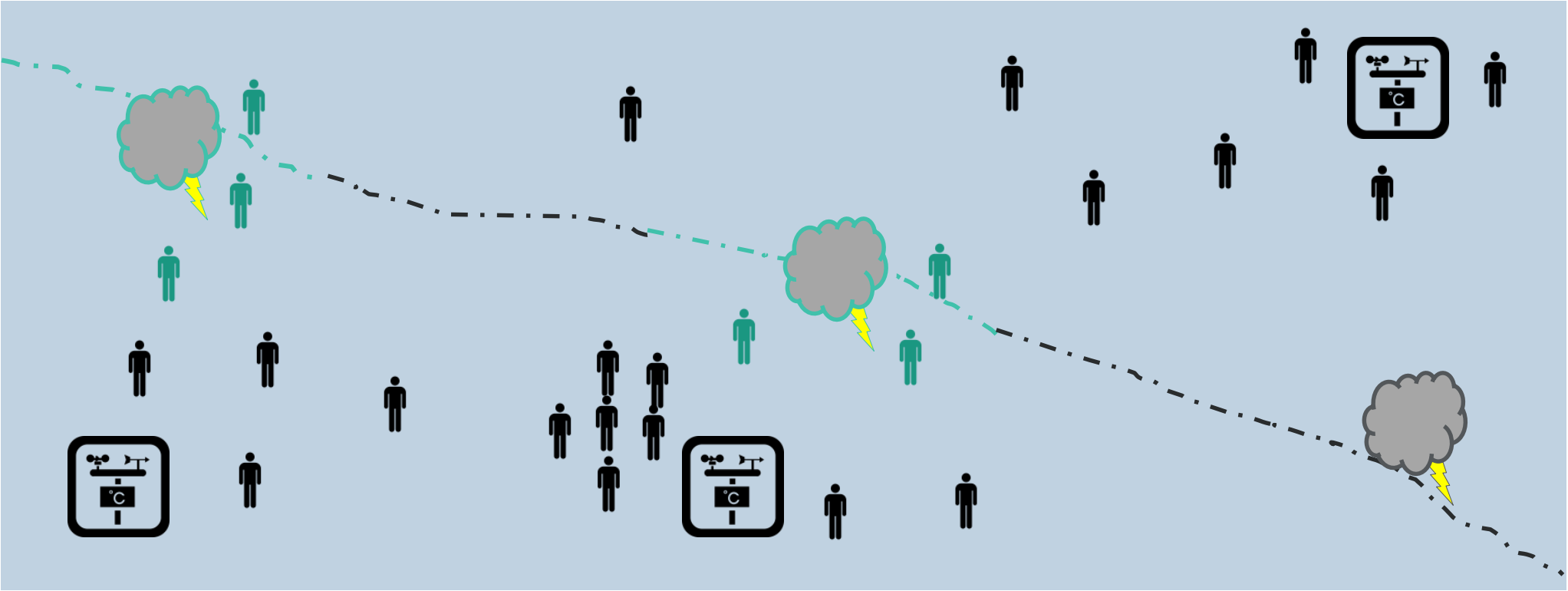
Depends on time and space coverage.
Uncertainties vary by data type

Scale of Storms Makes Them Difficult to Observe



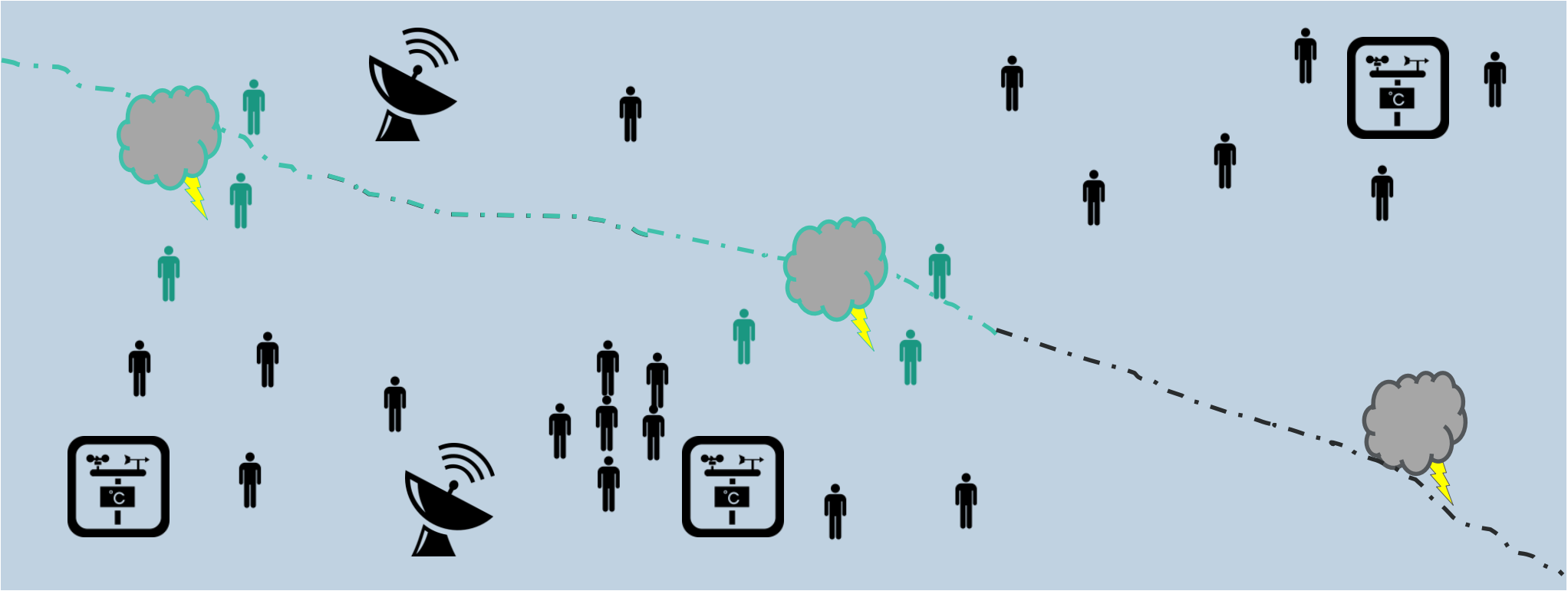
 Thunderstorm  Weather station

Scale of Storms Makes Them Difficult to Observe



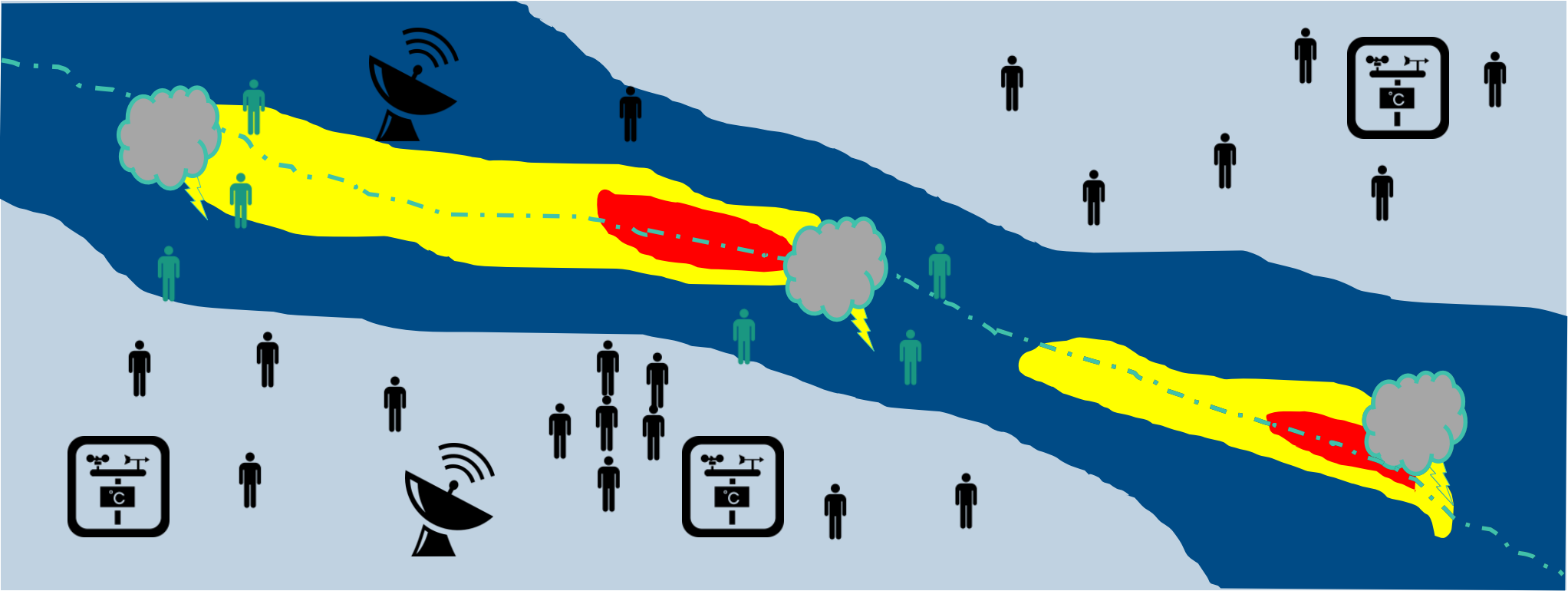
 Thunderstorm  Weather station  Observer

Scale of Storms Makes Them Difficult to Observe



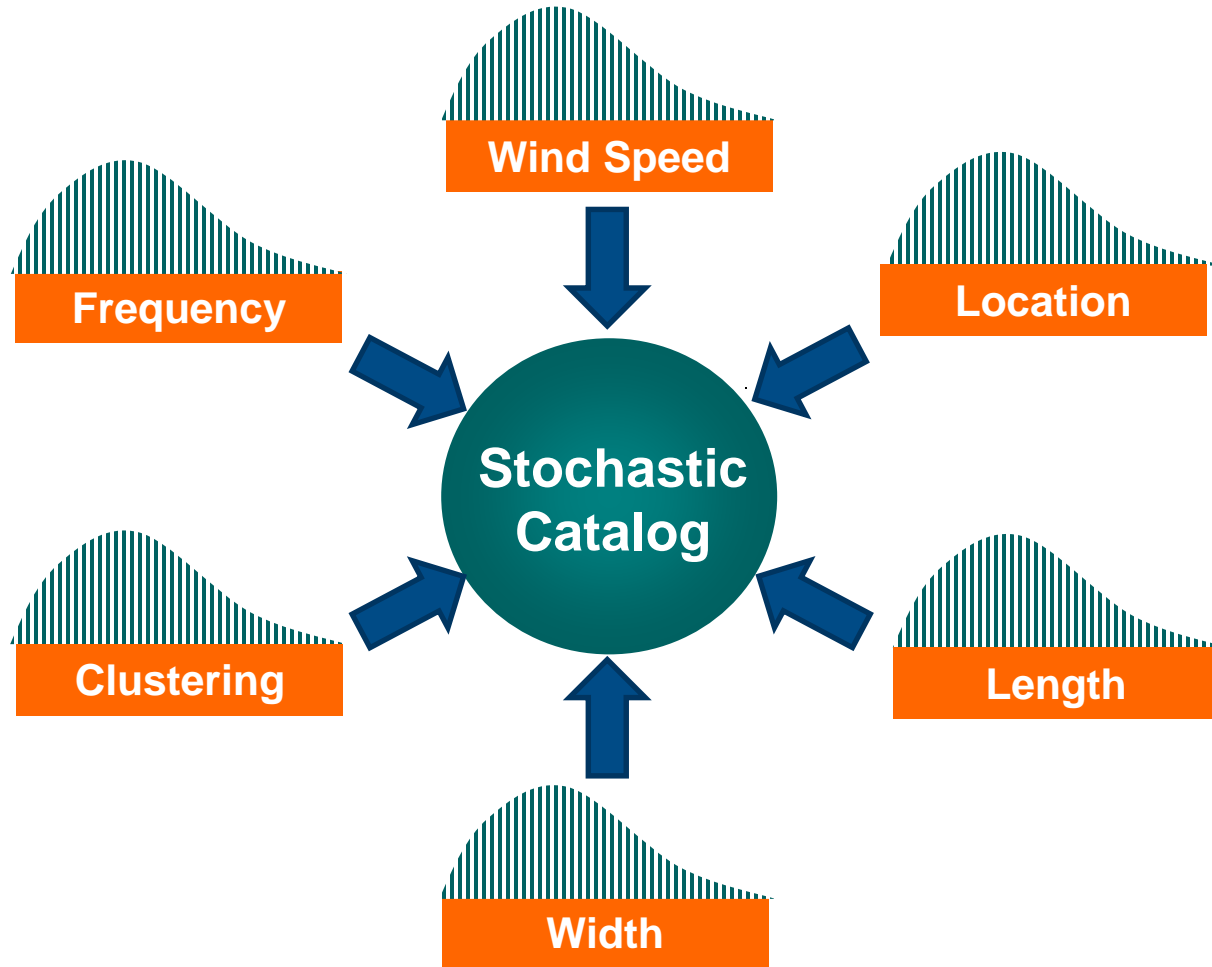
 Radar  Thunderstorm  Weather station  Observer

Scale of Storms Makes Them Difficult to Observe

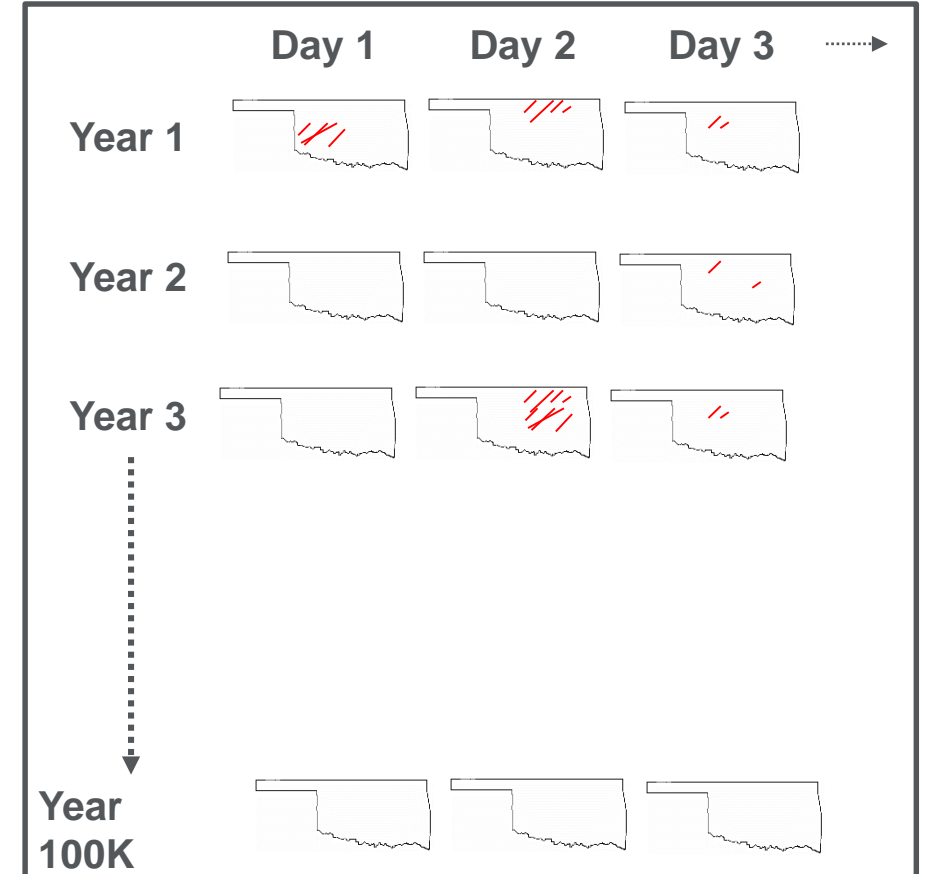


-  Radar
-  Thunderstorm
-  Weather station
-  Observer
-  Reanalysis

These Datasets Combine to Provide a Catalog of Plausible, Yet Perhaps Yet Unrealized Events



A 100K Year Stochastic Catalog



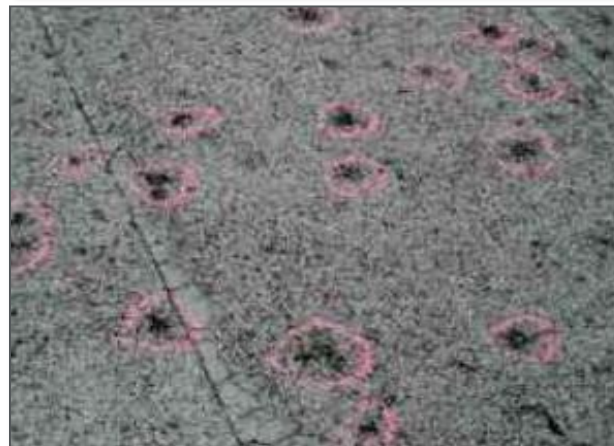
Now that We Have a Catalog... How Do We Calculate Damage?

Experiments



Source: Marshall et al., 2002

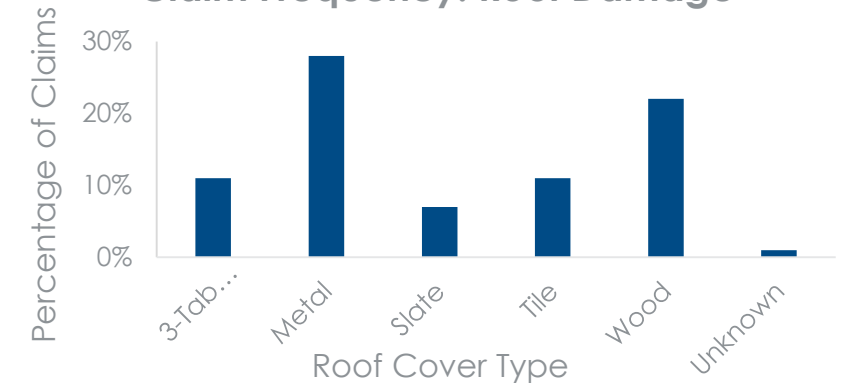
Damage Surveys



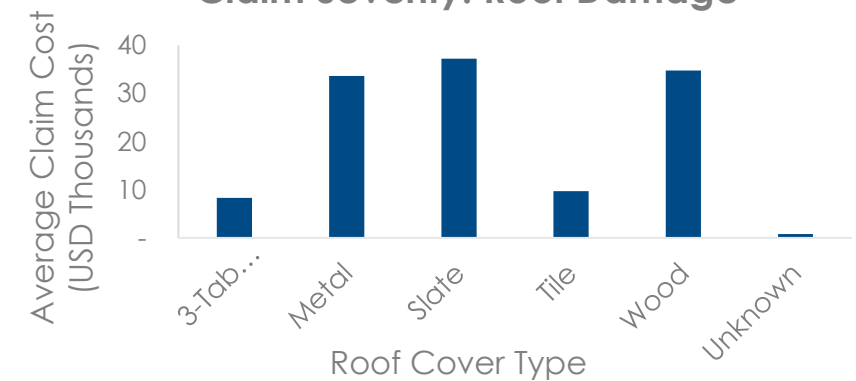
Source: RICOWI & AIR Damage Surveys

Claims Studies

Claim Frequency: Roof Damage



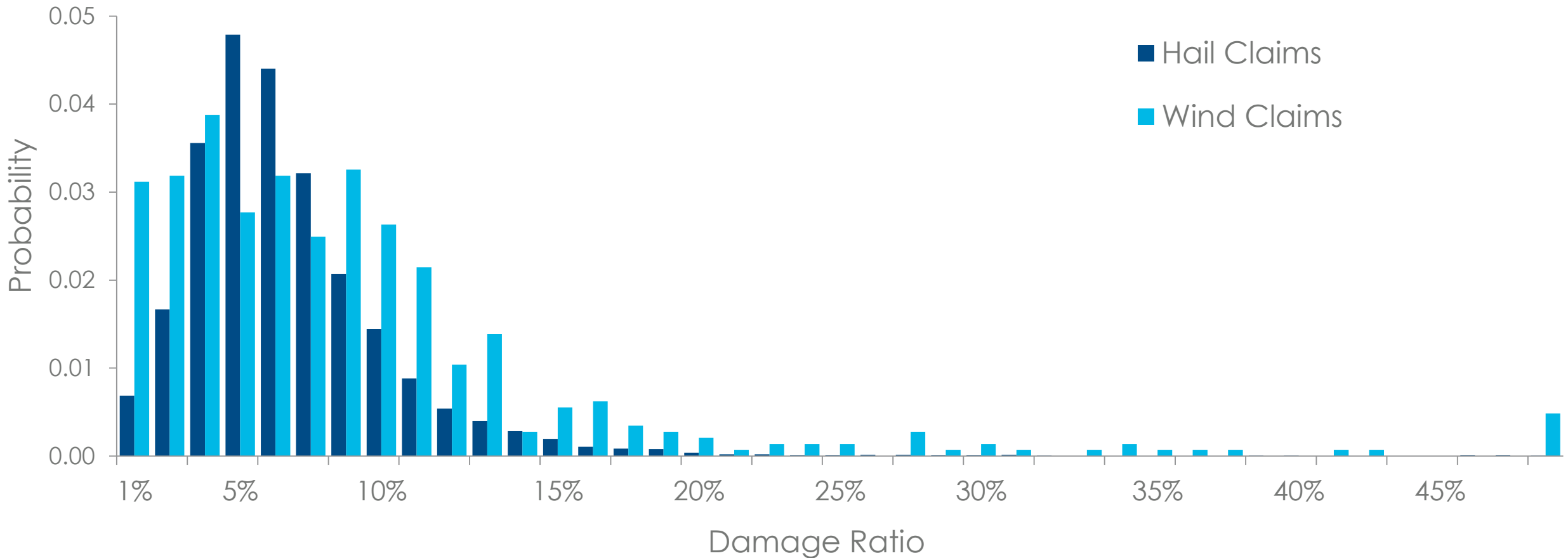
Claim Severity: Roof Damage



Source: IBHS, 2013

We Can Also Understand Uncertainty Directly Using Claims Data

Sample Damage Distributions from Hail and Wind Claims Data



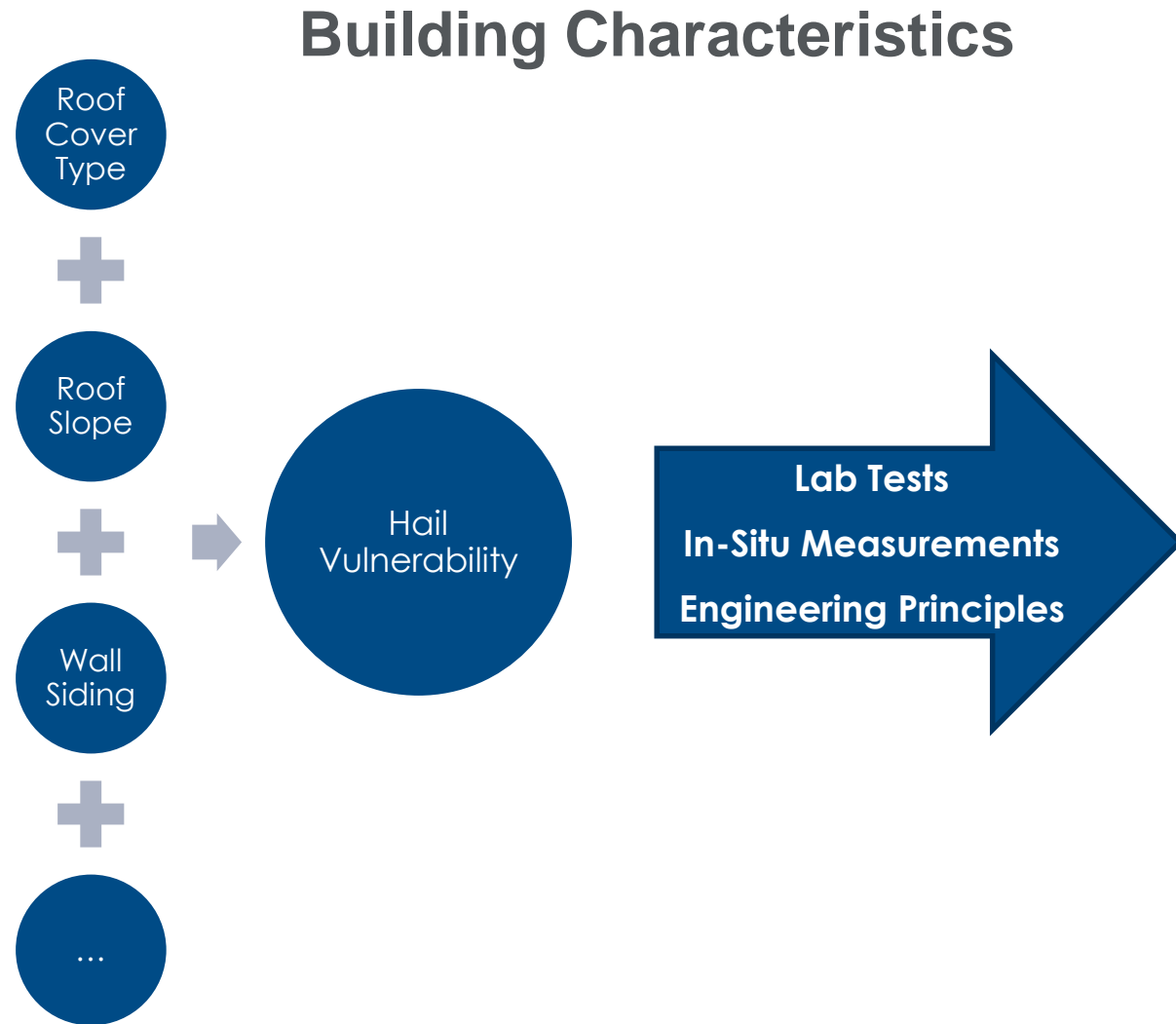
In Fact These Data Sources Allow Us to Consider Many Different Mitigation Factors?

Supported Features in the Severe Thunderstorm Model		
• Seal of Approval	• Roof Anchorage	
• Floor of Interest	• Year Roof Built	
• Building Condition	• Wall Type	
• Tree Exposure	• Wall Siding	
• Small Debris Source	• Glass Type	
• Large Missile Source	• Glass Percent	
• Terrain Roughness	• Window Protection	
• Adjacent Building Height	• Exterior Doors	
• Roof Geometry	• Building-Foundation Connection	
• Roof Pitch	• Internal Partition Walls	
• Roof Covering	• Wall Attached Structures	
• Roof Deck	• Appurtenant Structures	
• Roof Covering Attachment	• Roof Attached Structures	
• Roof Deck Attachment		
Newly Added Features ONLY for Hail	Hail Impact Resistance Roof Coverings:	
	✓ Class A	Least resistant ↓ Most resistant
	✓ Class B	
	✓ Class C	
	✓ Class D	

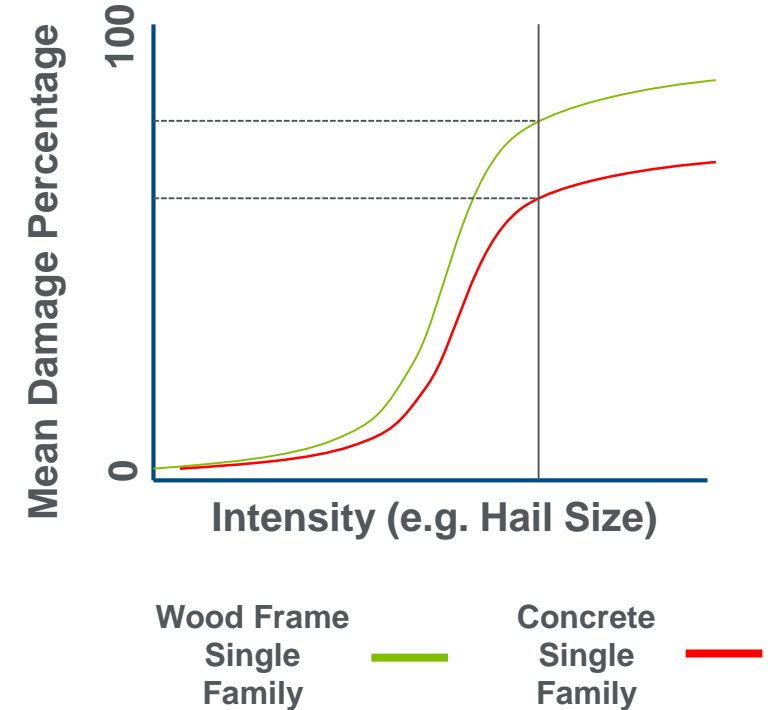
Note: Secondary features highlighted in green are supported for the hail sub-peril.



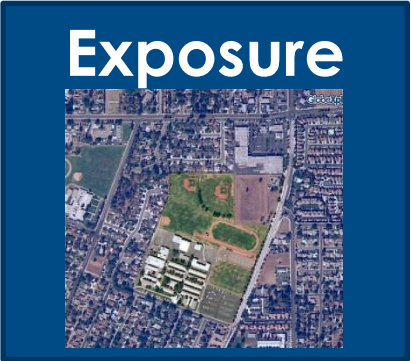
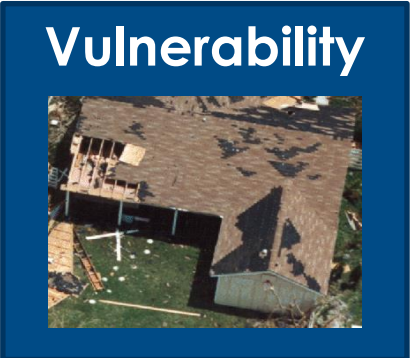
Damage Estimation: Translate Exposure, Building Characteristics, and Hazard into Damage



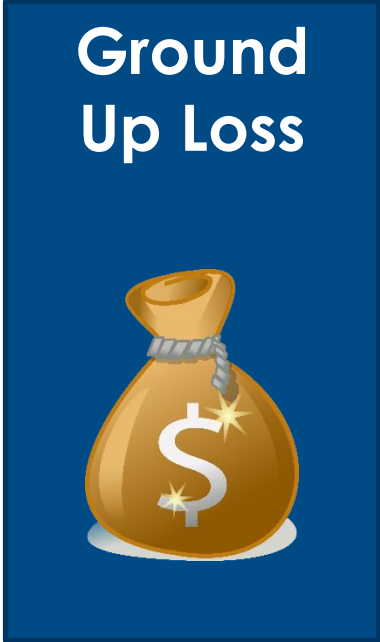
Damage Functions



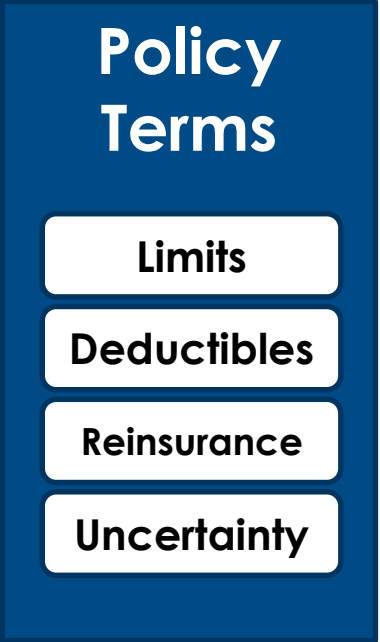
Loss Estimation: Apply Policy Terms and Uncertainty to Get Total Gross Insured Loss



=



+



Gross Loss

So How Can We Use This Tool to Manage the “Mess”?

The model alleviates many of the previous issues we encountered:

- **“Highly variable”**
More “years” of data allow for decreased variance and increased coverage
- **“Ill-observed”**
Application of meteorology and engineering allows for reasonable estimates in absence of claims data
- **“Non-stationary, spatially correlated, and potentially cyclical”**
Climate variability implicitly captured through use of historical atmospheric conditions

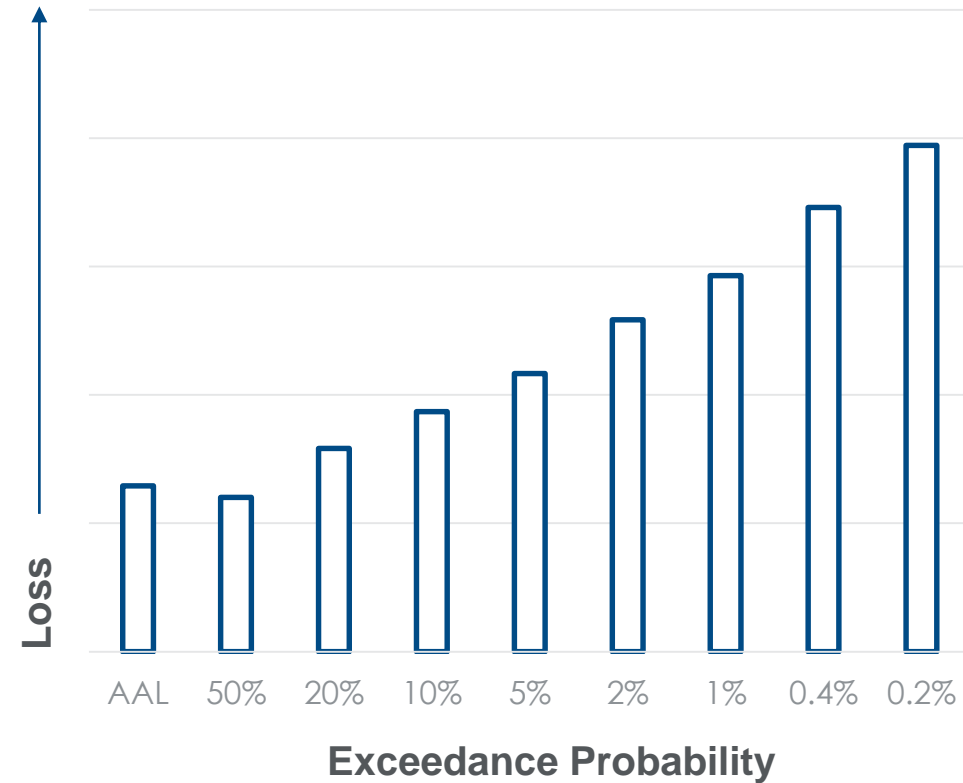
The model also allows us to answer key questions like...

“What was the Probability of that Happening?!”

Event Loss Table

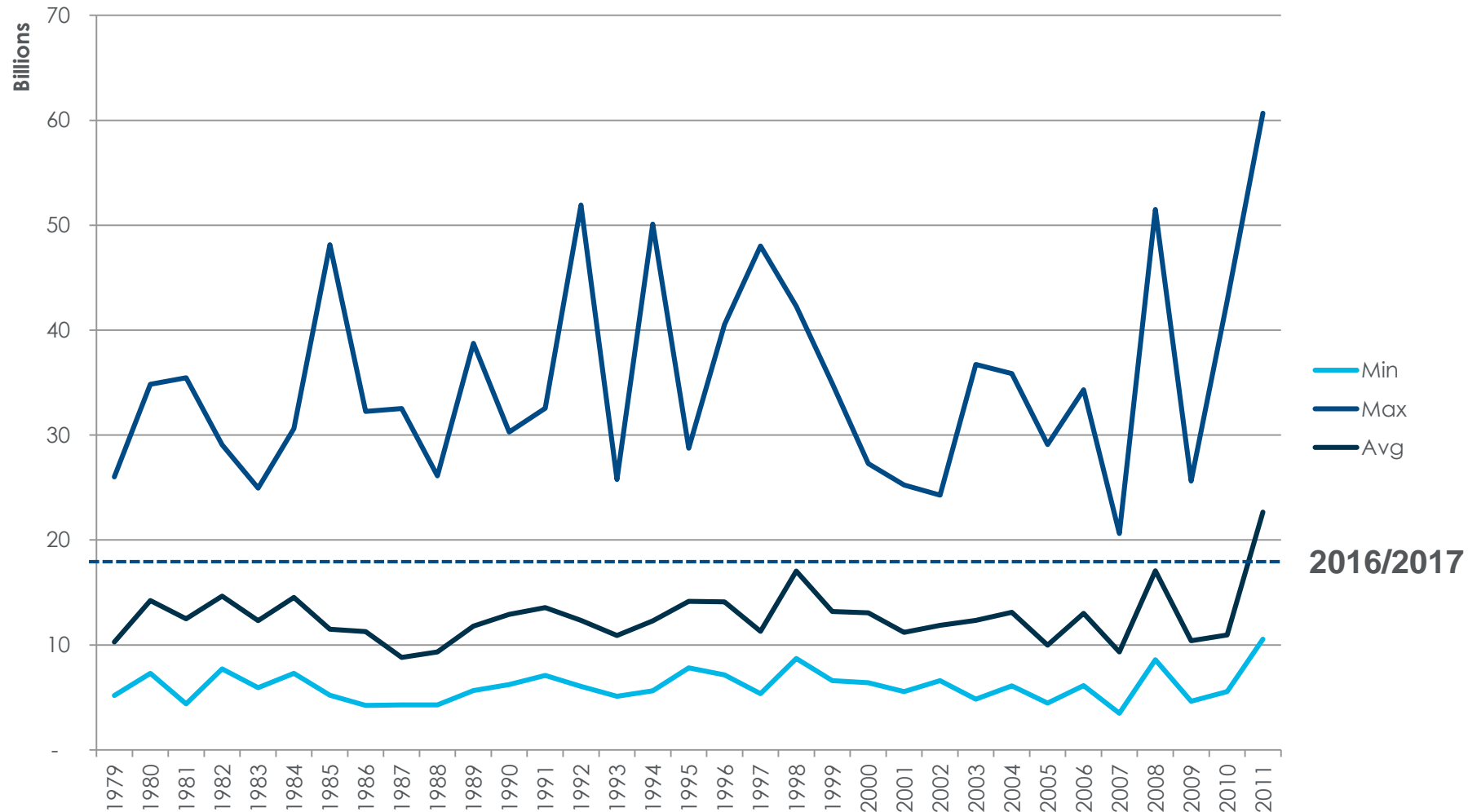
Event ID	Year	Month	Day	Loss
1	1	5	28	2,153,555
2	1	6	3	75,000,000
3	1	6	27	43,023,654
...
...
...
...
53229	10000	10	1	100,235,225
53230	10000	11	12	5,237,585
53231	10000	12	15	10,236,125

Exceedance Probability Curve



“How Much of an Outlier was 20XX?”

Model Simulated Losses By Basis Year



“What if it happened again!?”

1896 St. Louis Tornado



St. Louis c. 1875



St. Louis – Present Day

Method	Rate/yr	Loss 2014
Inflation Only	3.3%	\$ 543,653,610
GNP	4.9%	\$ 3,558,163,038
Tangible Wealth	6.3%	\$ 15,976,856,168
Modeled	N/A	\$ 7,256,136,150

The Case for CAT Models...



Severe thunderstorms present a complex yet serious risk to the insurance market

EP TVAR
PML

CAT models help “tame the mess”



Flexibility and robustness help view the risk from different perspectives