

NCAR

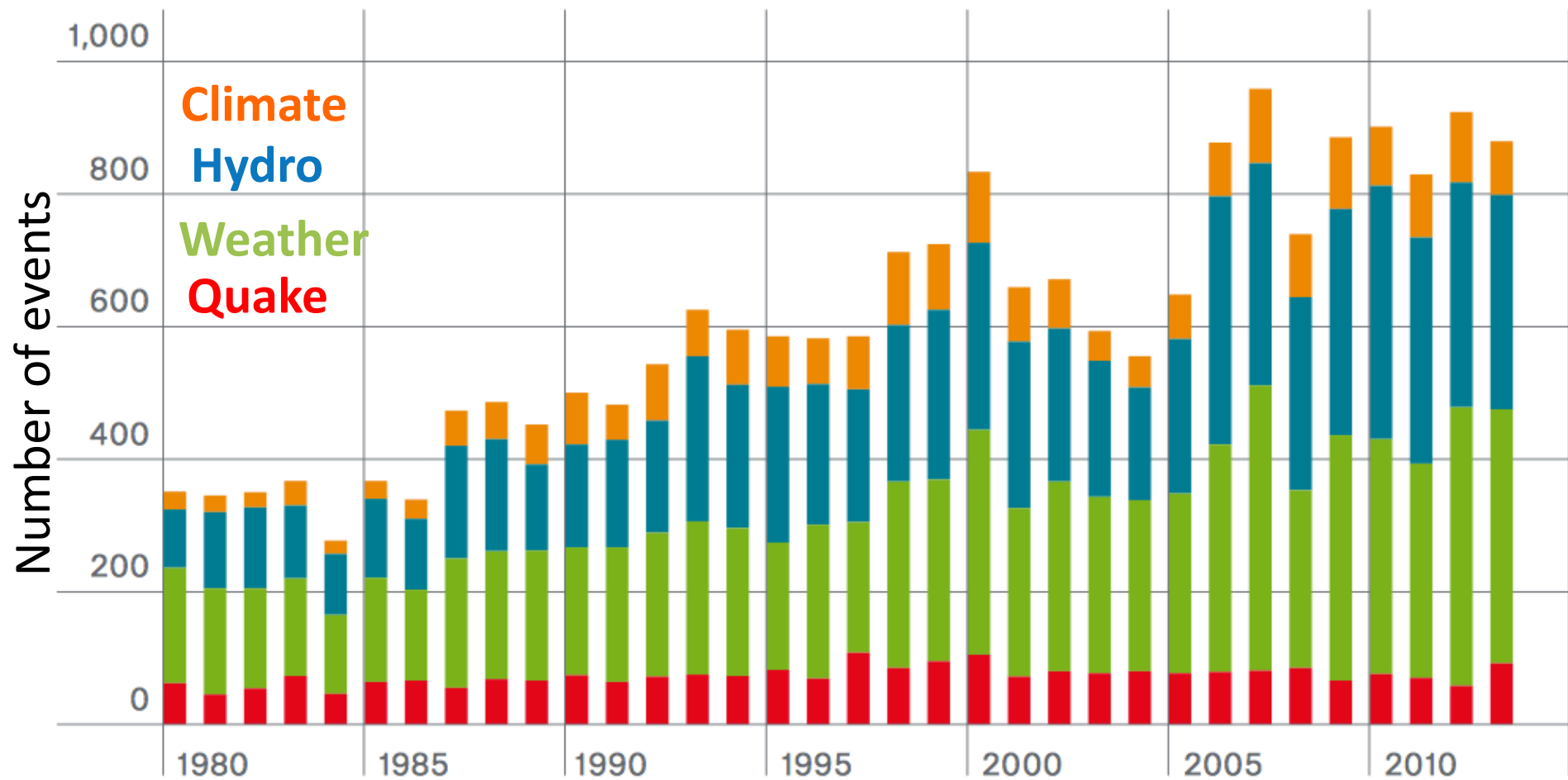
NATIONAL CENTER FOR ATMOSPHERIC RESEARCH



Pathways Connecting Physics and Climate Resilience

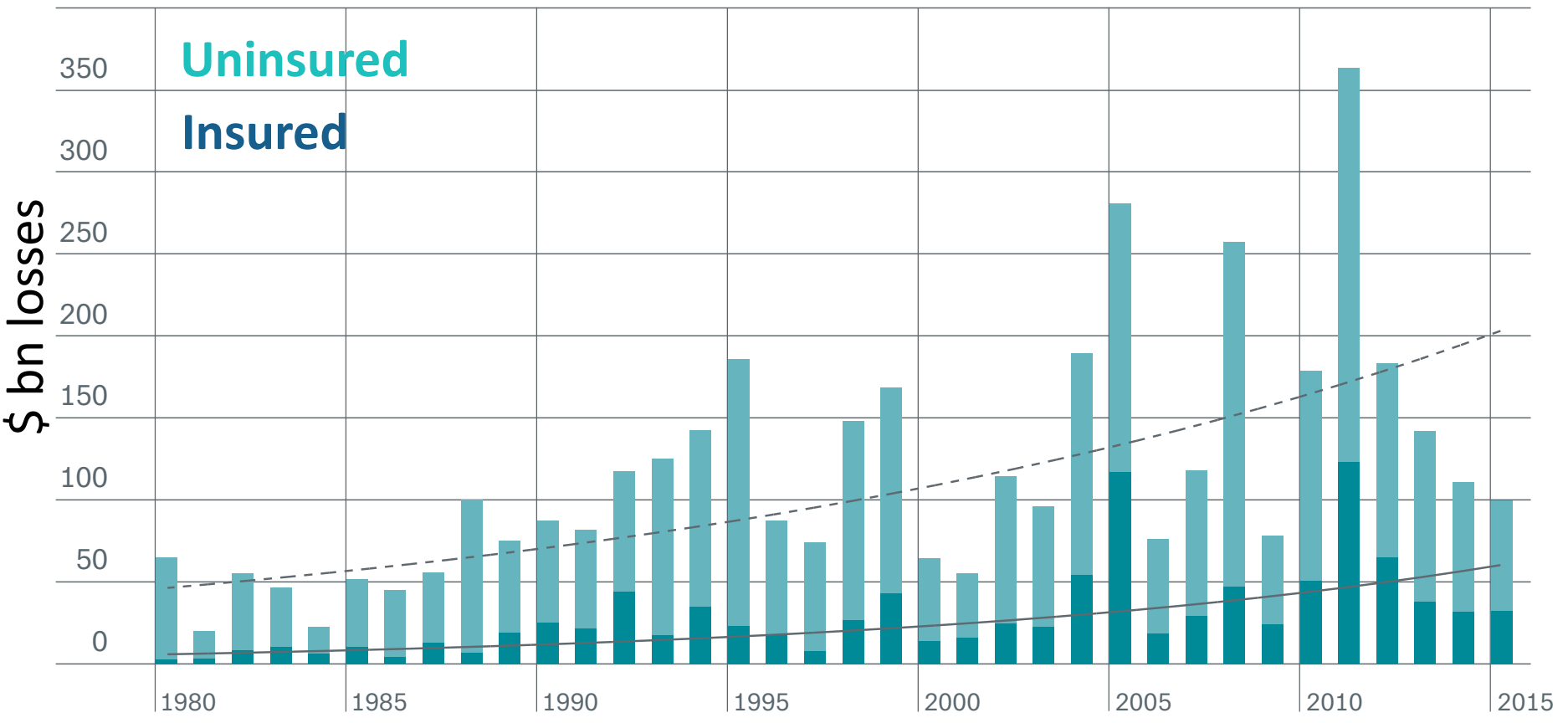
James Done

Increasing Number of Events



Munich Re 2014

Increasing Losses

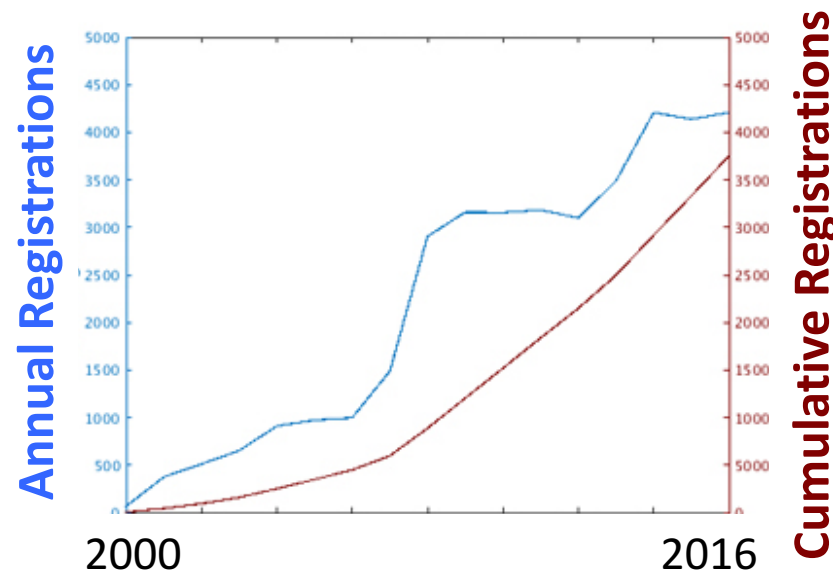


Munich Re 2014

- Highest quality science

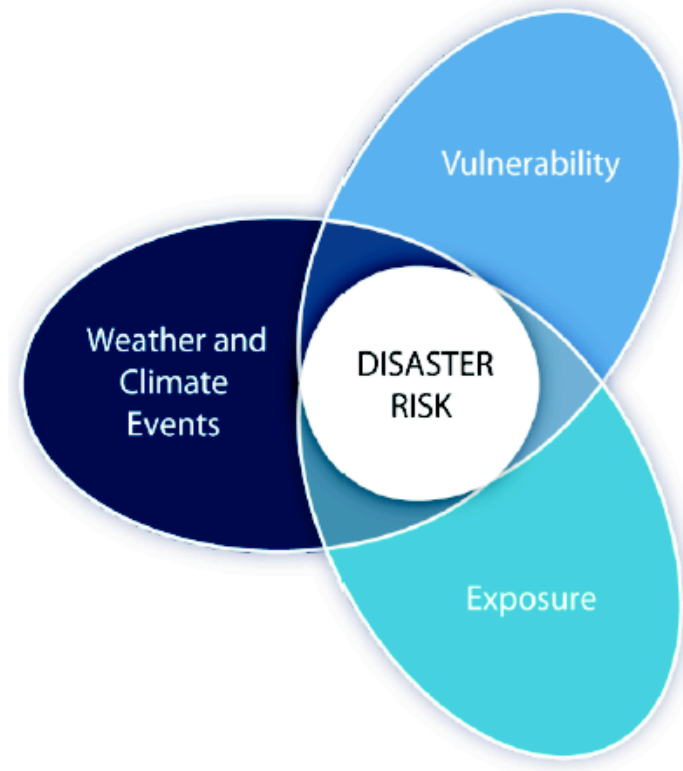
Rank	Institution	C/N = citations per paper	N = number of papers	C = number of citations
1	NCAR (NATL CTR ATMOSPHERIC RES)	34.92	4,718	164,732
2	HARVARD UNIV	29.81	1,908	56,878
3	PRINCETON UNIV	28.13	1,701	47,857
4	UNIV WASHINGTON	25.16	3,967	99,813
5	UNIV COLORADO BOULDER	25.04	5,233	131,049

- Most widely-used community models



- Next-generation instruments, and datasets

Physics is one of many factors

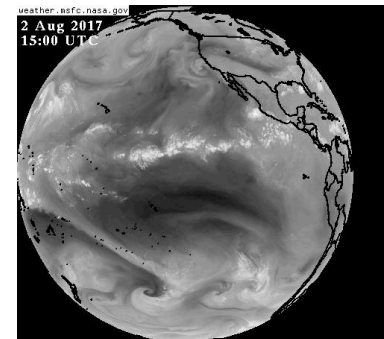


Adapted from IPCC SREX, 2012

- Incorporate physics into weather and climate risk assessment.
- Risk management practice informs the physically-based approaches.

The Mother Lode:

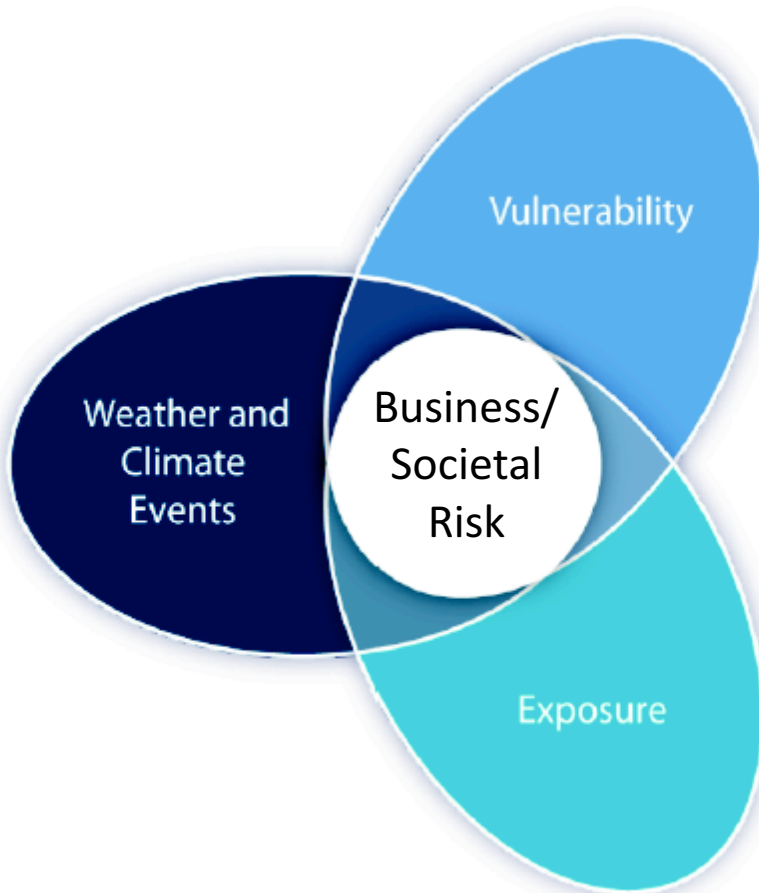
Capacity to understand mesoscale phenomena at the global scale.



- Stakeholder:** The reinsurance industry
- Need:** Understand losses
- Current practice:** Gradient wind
- Hypothesis:** Terrain effects drive TC wind losses
- Physics:** Terrain effects
- Results:** New view of footprints and wind climate
- Resilience action:** Optimize reinsurance portfolios

Thanks to Ming Ge (NCAR), Yuqing Wang (U. Hawaii),
Geoff Saville and Ioana Dima-West (Willis Towers Watson)

Minimizing Risk



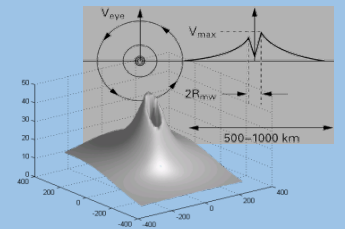
Adapted from IPCC SREX, 2012

- Understand inland wind decay.
- Understand historical losses.
- Quantify wind risk in regions of sparse data
- Validate catastrophe models.



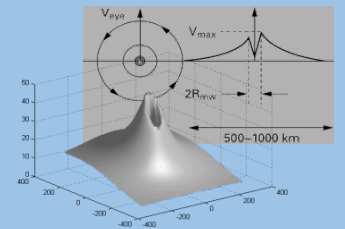
Parametric radial wind profiles:

- fast, but smooth fields, surface wind factors.



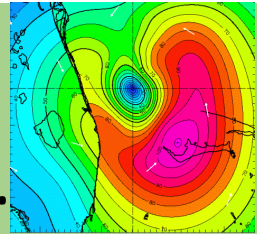
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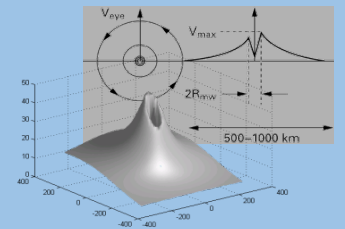
Spatial analysis of observations:

- asymmetries, but few storms, globally inconsistent.



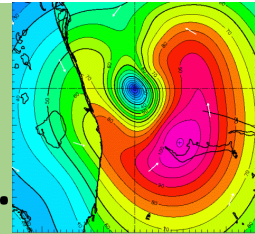
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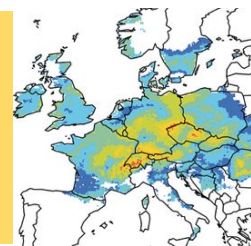
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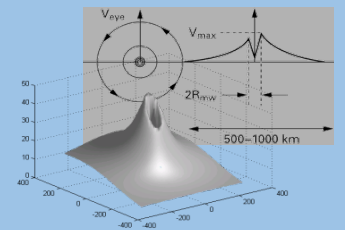
Geostatistical spatial modeling:

- fast, only applied to European windstorm so far.



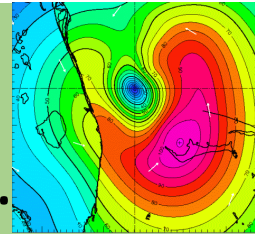
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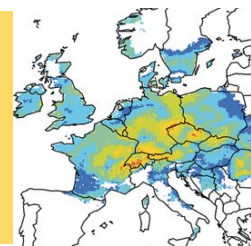
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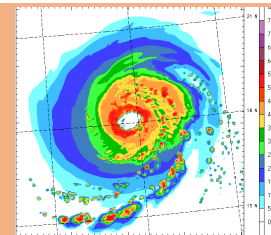
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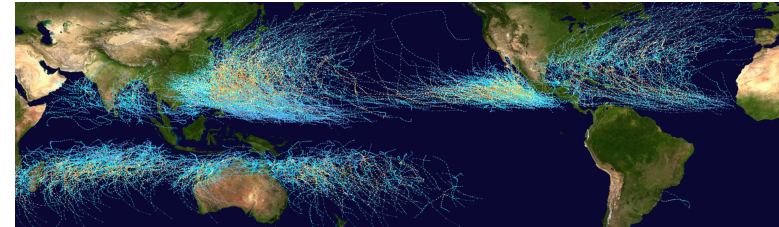
Numerical modeling:

- many physical processes, but slow, track error.

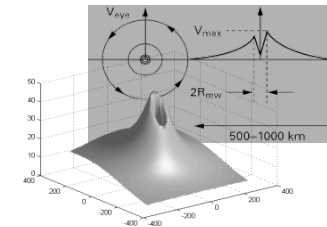


Adding Physical Processes

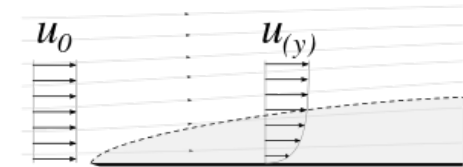
Historical track data
(EBTrACS and JTWC)



Holland et al. (2010)
parametric pressure profile



KePERT and Wang (2001) numerical
boundary layer model.



Fast, some topographic and roughness effects, no track error, but missing processes (e.g., strong thermal effects).

Diagnoses boundary-layer flow using dry equations of motion for a given pressure field.

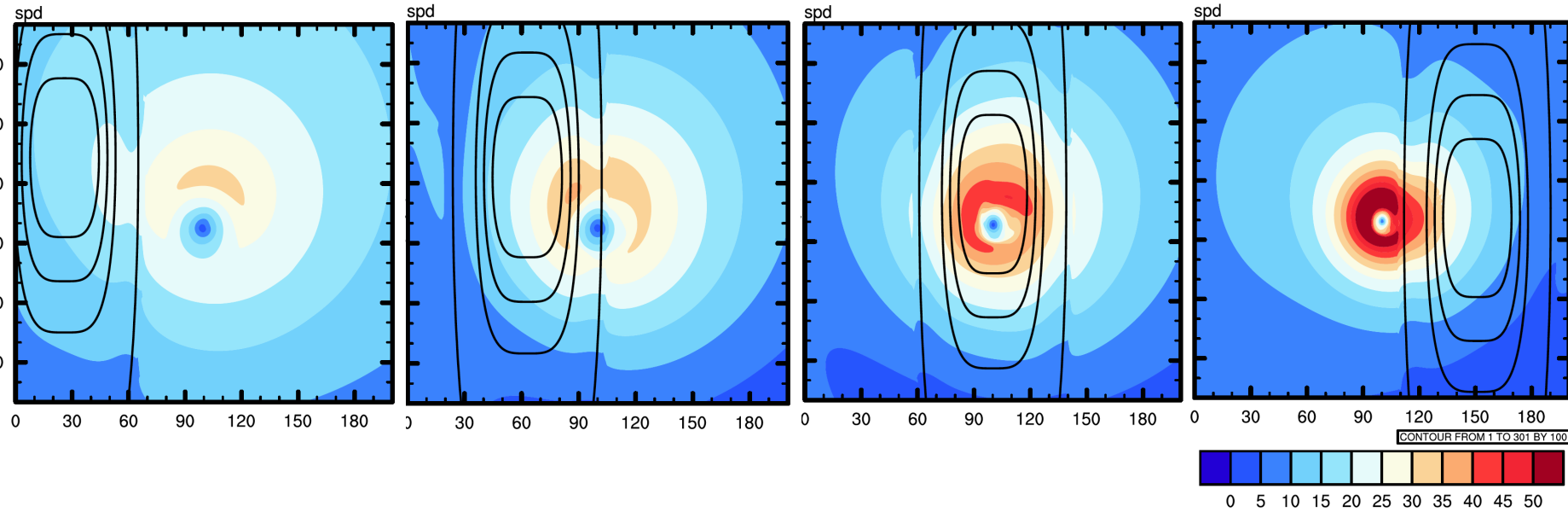
- High-order turbulence scheme
 - prognostic TKE, turbulence dissipation.
 - diagnostic length scale (<80m).
- Ignores strong thermal effects.
- Rapidly achieves steady state.

1. Allow storms to move:
 - add environment pressure gradient to TC forcing,
 - add storm translation velocity to horizontal advection.

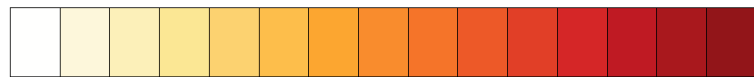
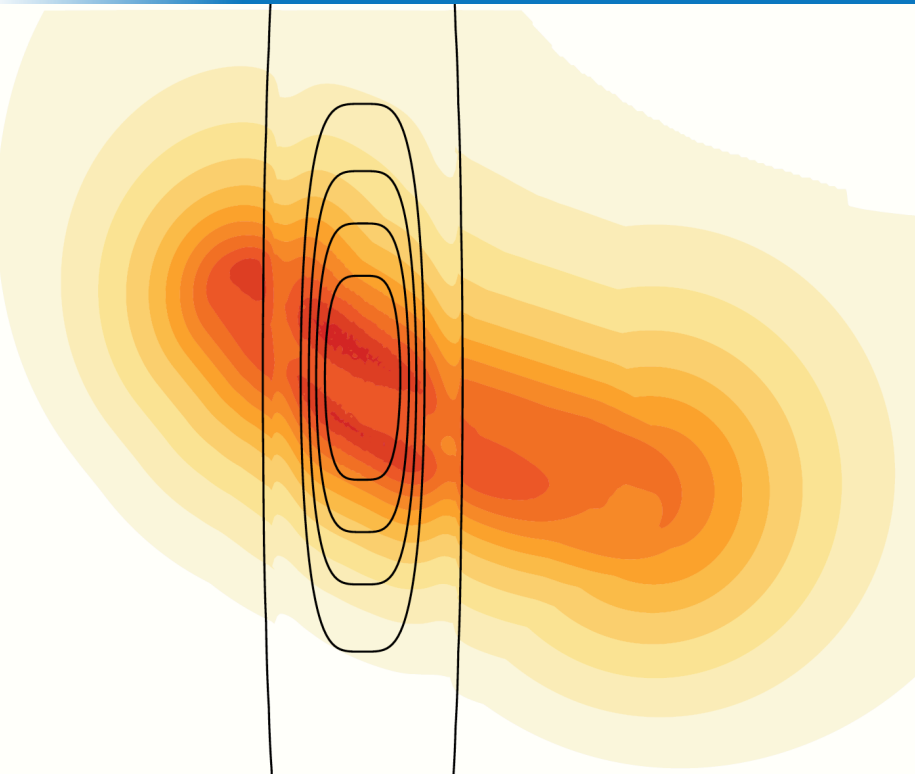
1. Allow storms to move:
 - add environment pressure gradient to TC forcing,
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2. Allow storms to change intensity and size:
 - update pressure gradient and allow winds to respond,
 - force gradient winds at model top.

1. Allow storms to move:
 - add environment pressure gradient to TC forcing,
 - add storm translation velocity to horizontal advection.
2. Allow storms to change intensity and size:
 - update pressure gradient and allow winds to respond,
 - force gradient winds at model top.
3. Include some topographic effects:
 - included in model equations.
4. Include variable surface roughness effects:
 - drag coefficient = $f(\text{terrain height})$.

Passing Over an Idealized Hill



Ike Encounters a Surprise Island

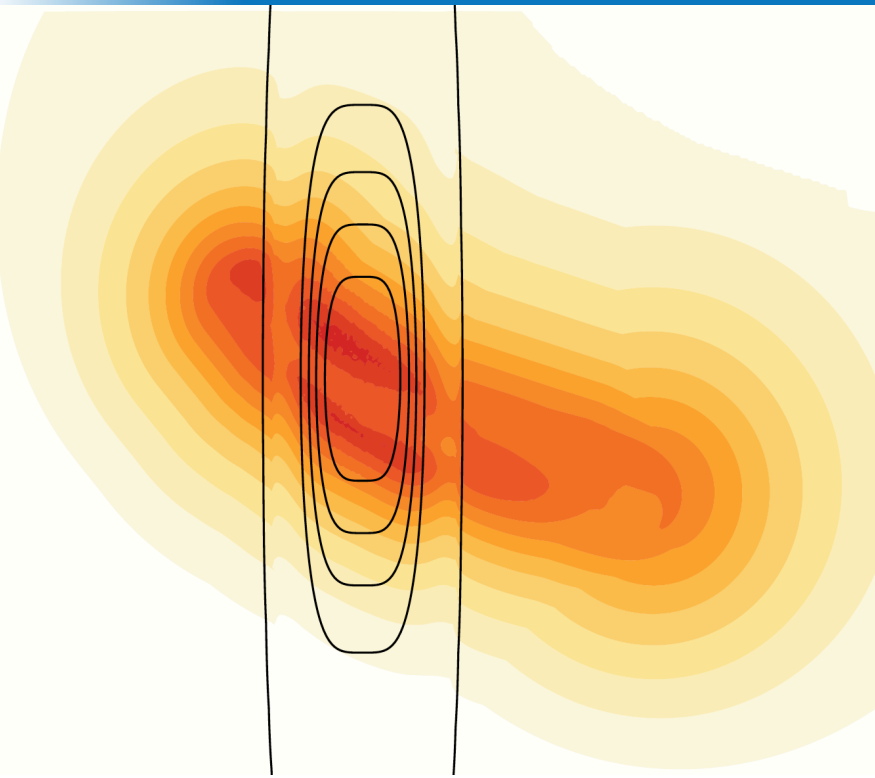


5 10 15 20 25 30 35 40 45 50 55 60 65 70

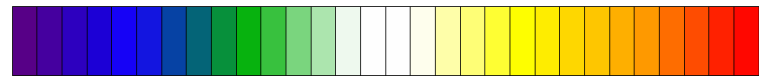
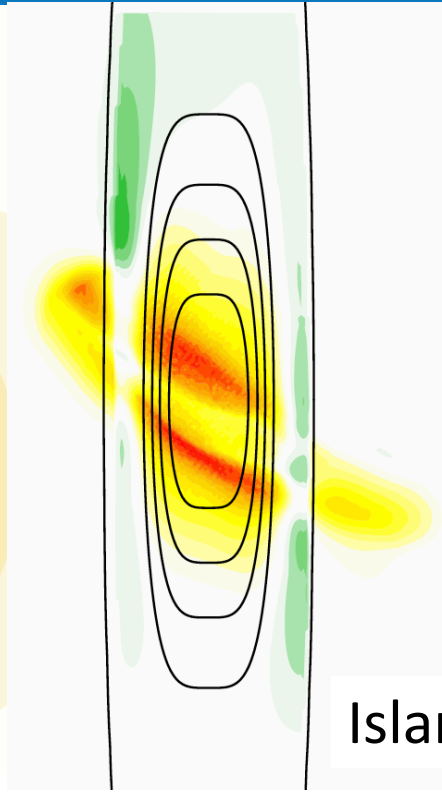
Max wind speed (ms^{-1})



Ike Encounters a Surprise Island

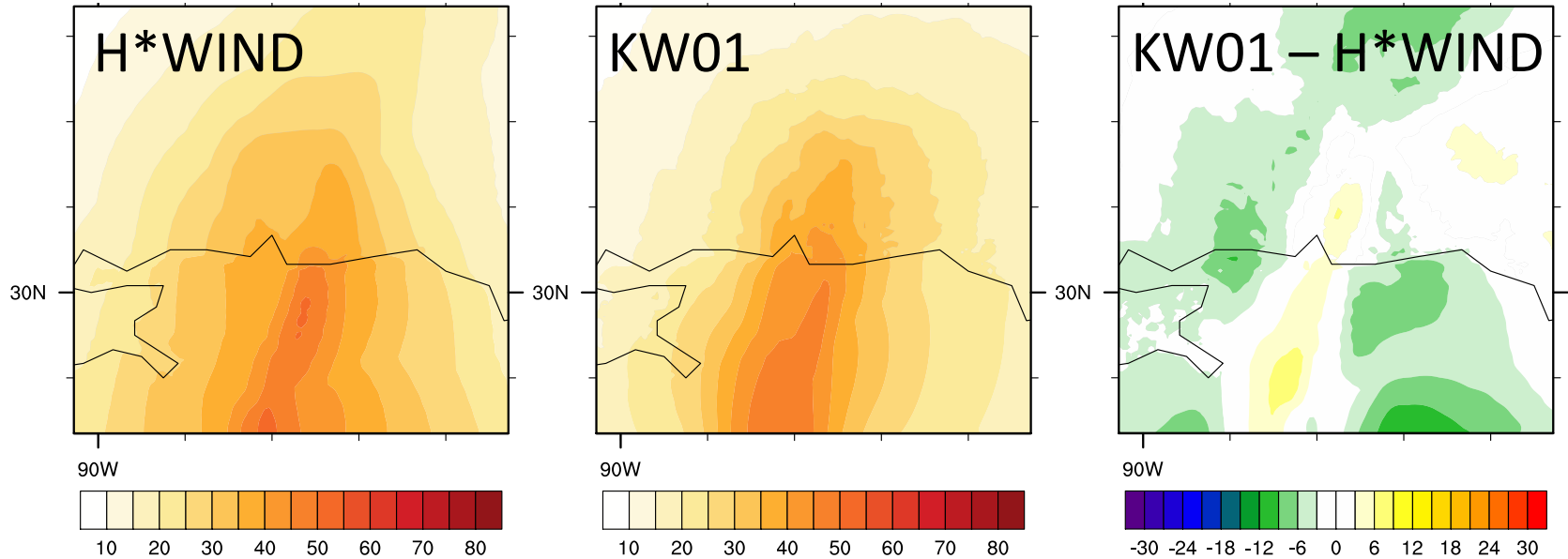


5 10 15 20 25 30 35 40 45 50 55 60 65 70
Max wind speed (ms^{-1})

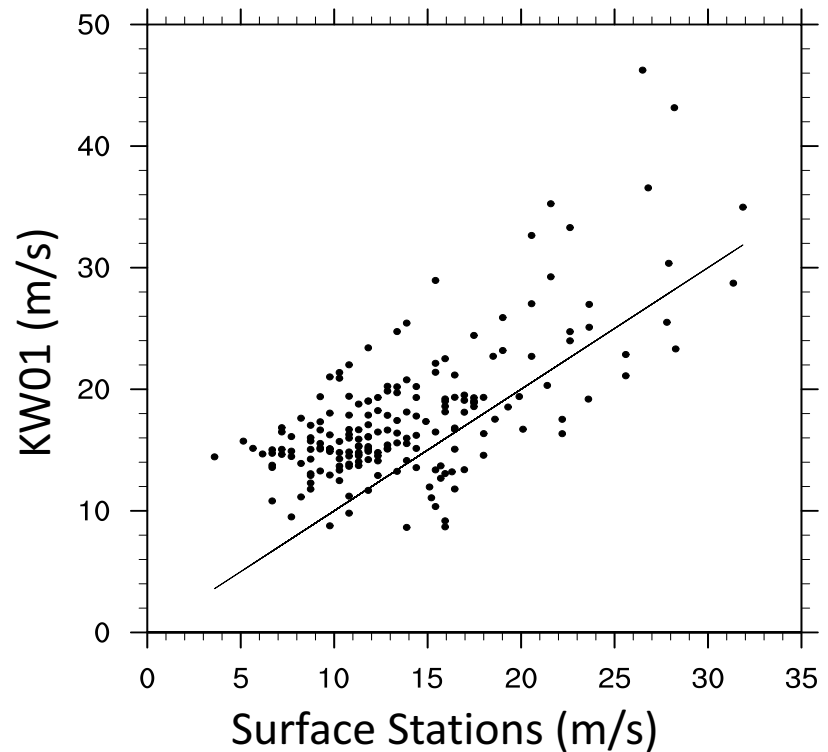
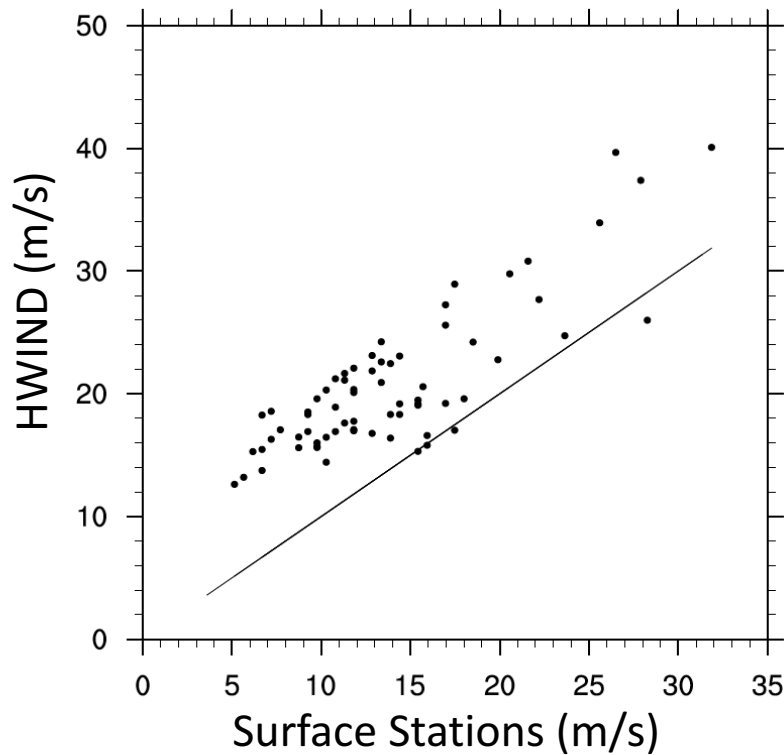


-12 -9 -6 -3 0 3 6 9 12
Max wind speed (ms^{-1})

Island – No Island



- HWIND has greater asymmetry.
- HWIND adjusts land data for open terrain.
- KW01 includes smaller scales.

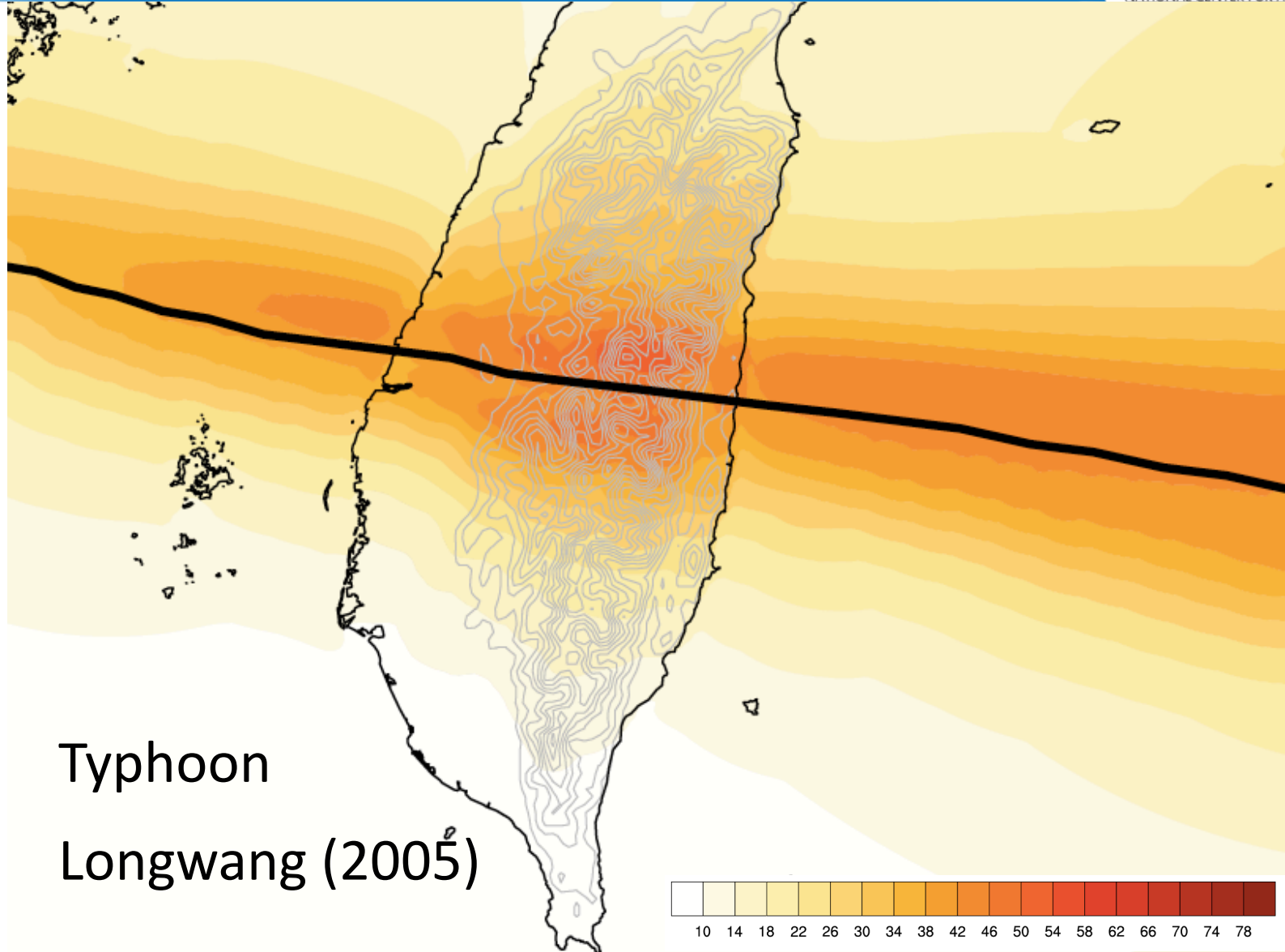


- HWIND and KW01 have high bias.
- KW01 comparable to HWIND.
- KW01 has potential to outperform analyses in complex terrain.

Results: 250 Footprints



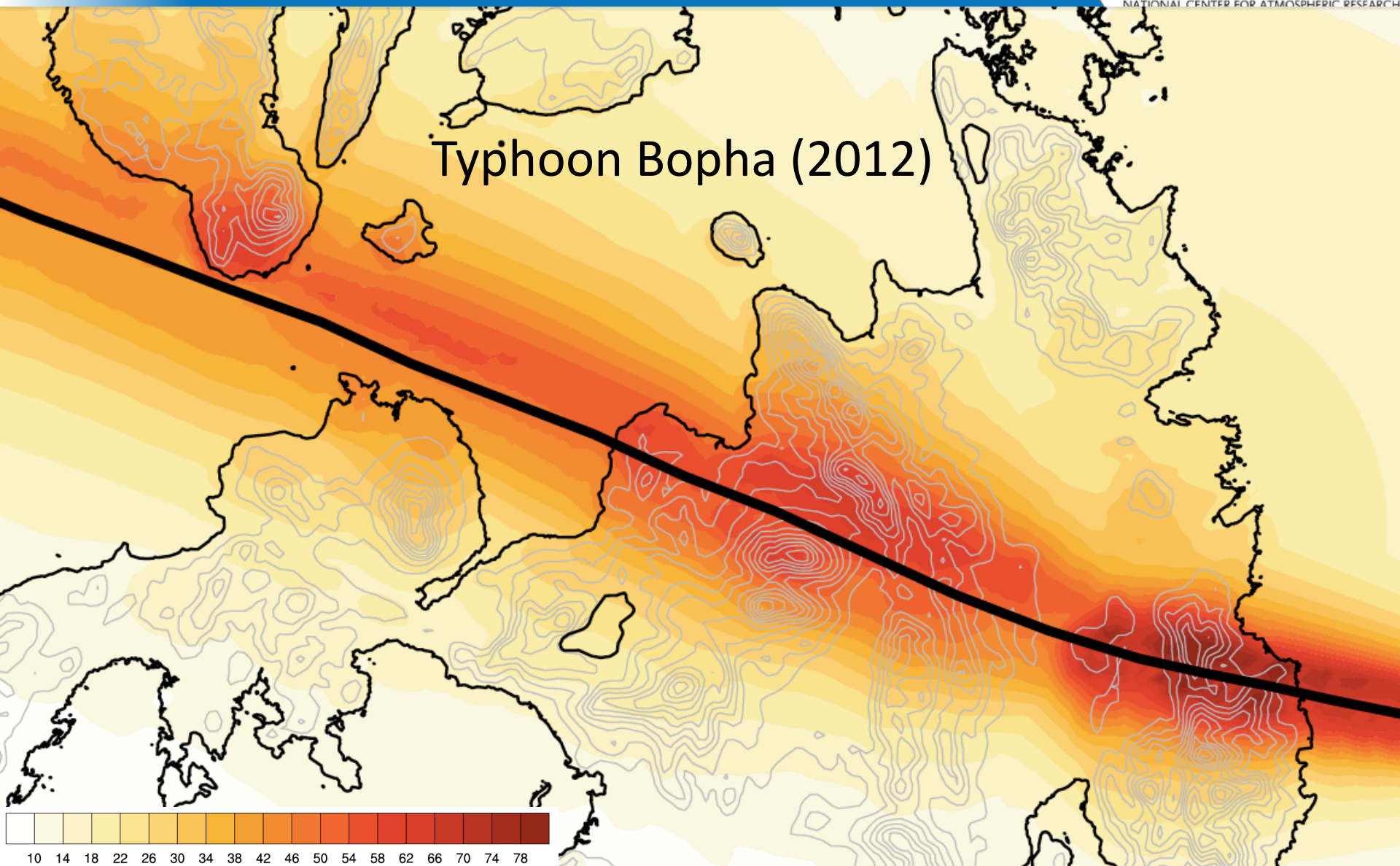
Results: High Terrain



Typhoon
Longwang (2005)

Results: Complex Topography

Typhoon Bopha (2012)

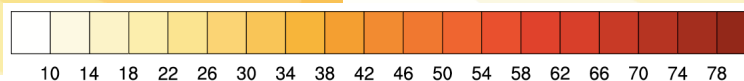


New views of wind risk aloft

Typhoon Ellen (1983)

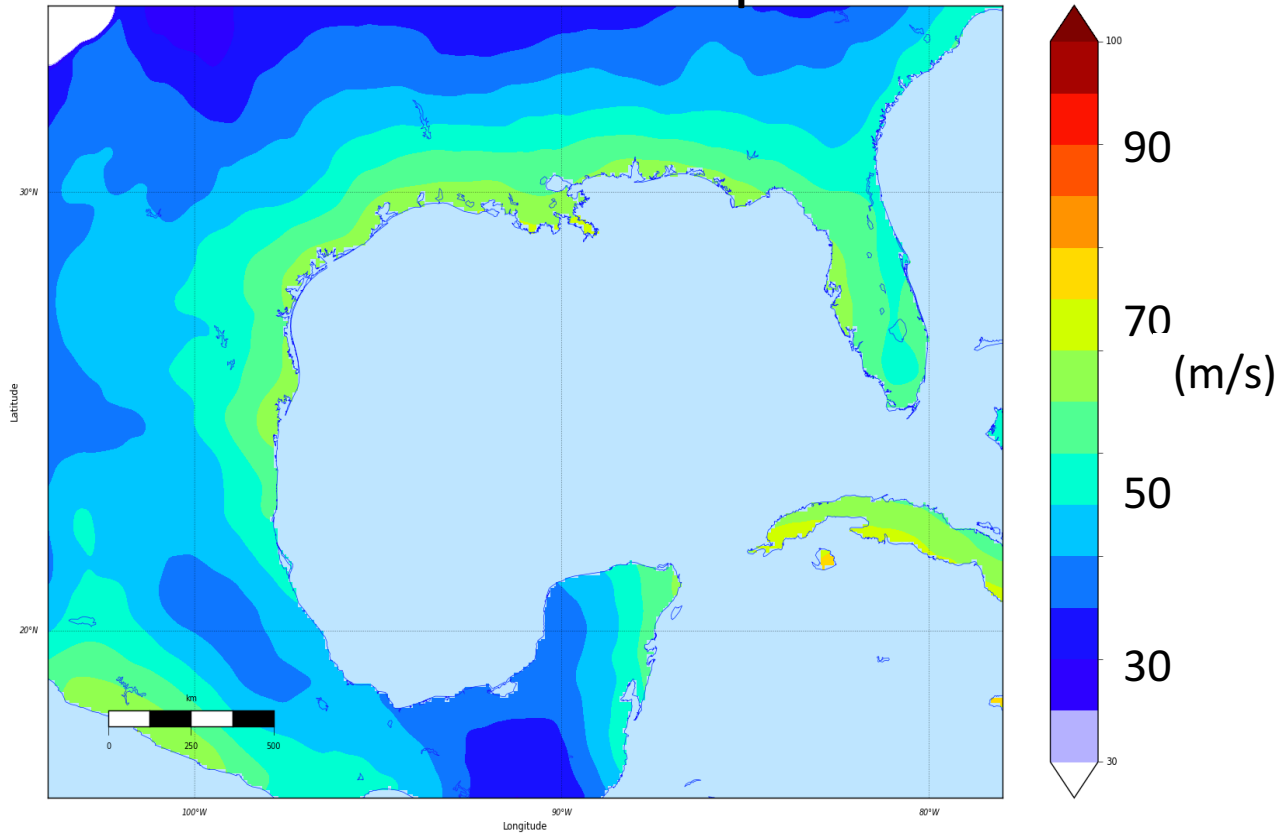
Max wind at 10m

Max wind at ~80m



(ms^{-1})

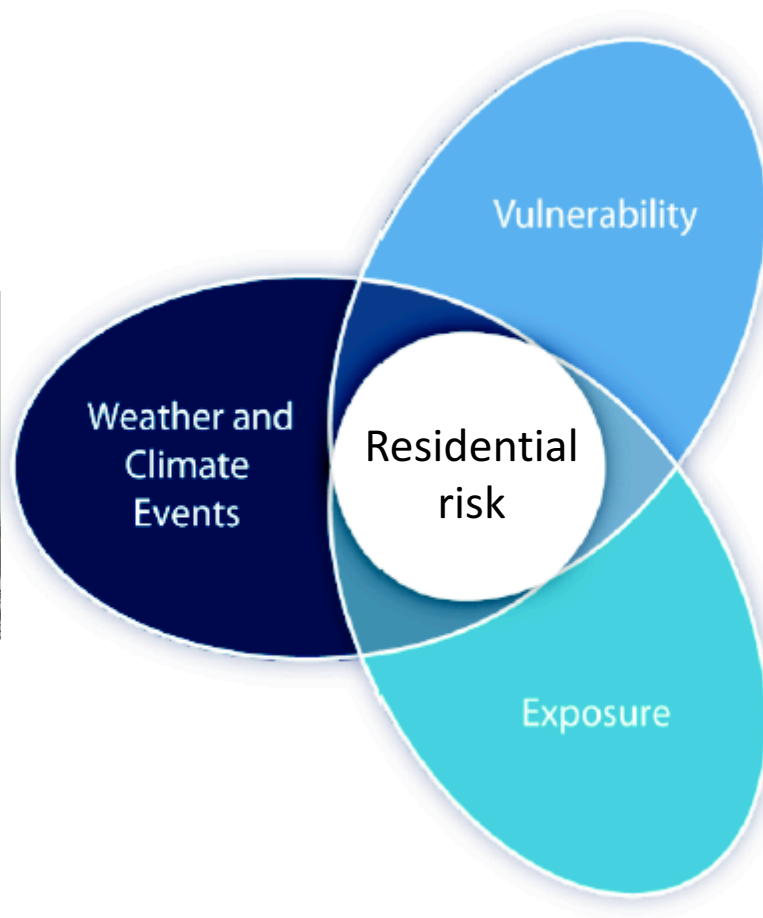
1% annual wind speed



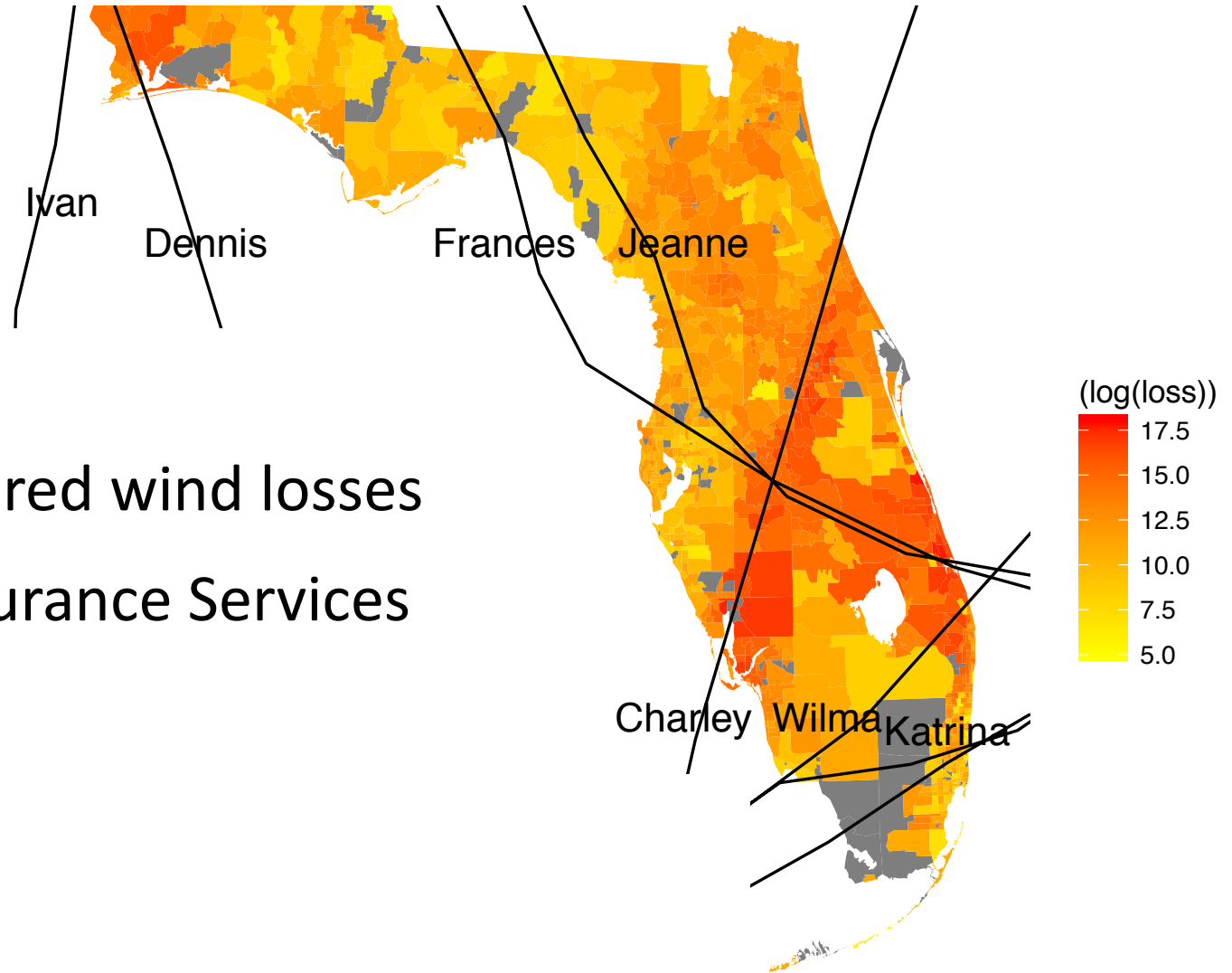
- New view of wind climate
- Optimized global exposure for business/societal resilience

- Stakeholder:** FL division of emergency management
- Need:** Effectiveness of the building code
- Current practice:** Code based on wind speed
- Hypothesis:** Losses also driven by other wind effects
- Physics:** Multiple wind field parameters
- Results:** Quantified loss reductions
- Resilience Action:** Informed building code updates, policy

Thanks to Jeff Czajkowski (U. Penn), Kevin Simmons (Austin College)

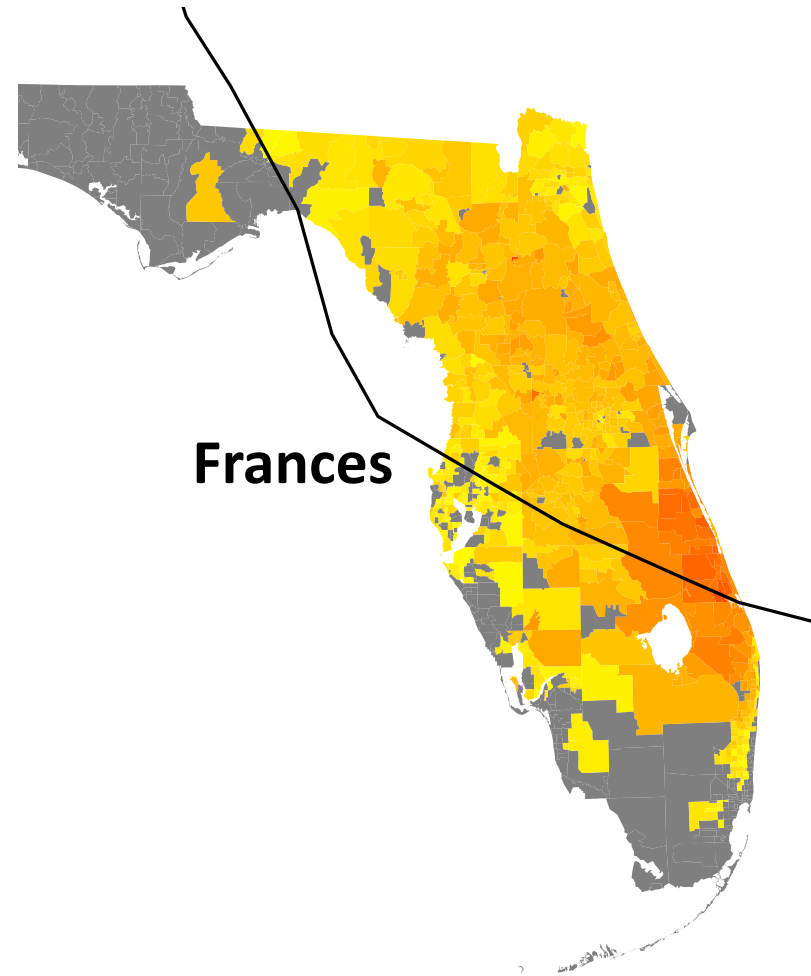
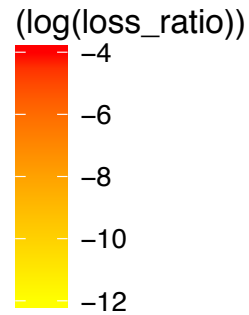
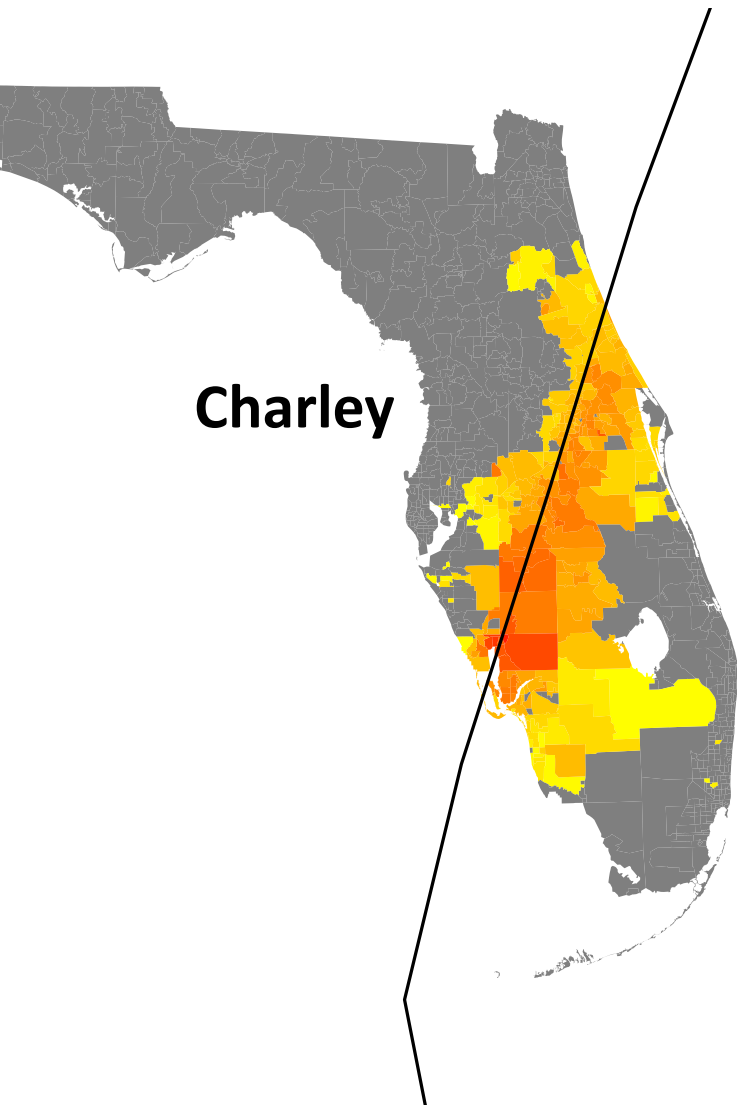


Adapted from IPCC SREX, 2012

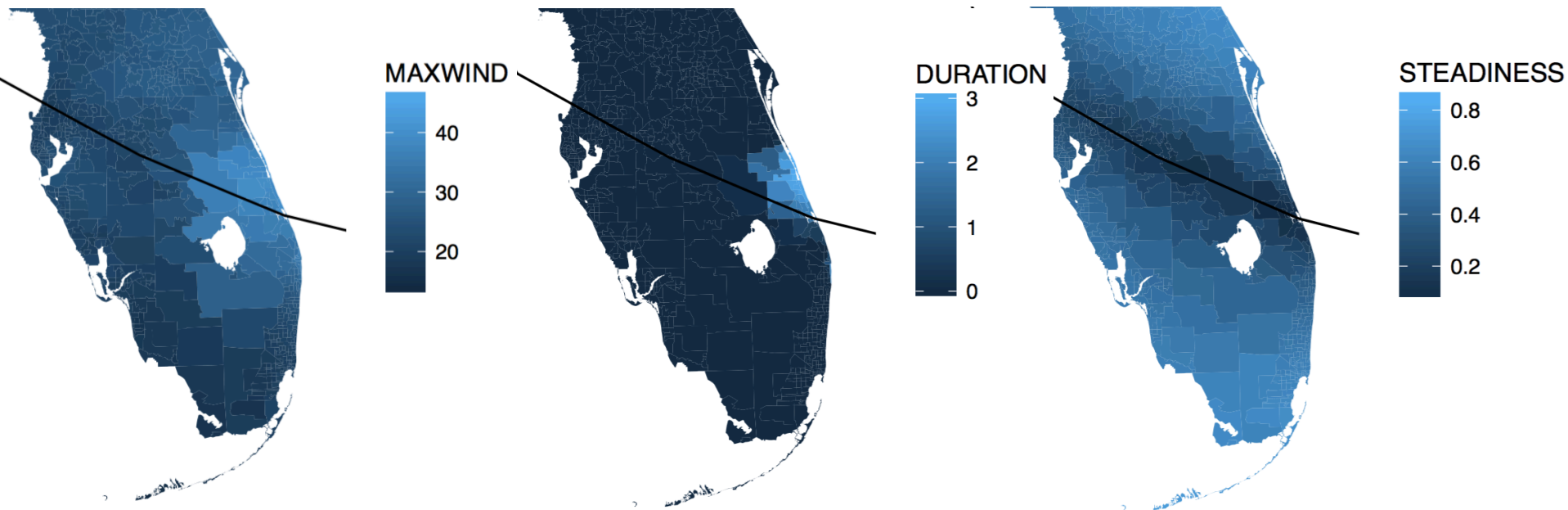


- Florida insured wind losses
- Source: Insurance Services Office

Losses by hurricane

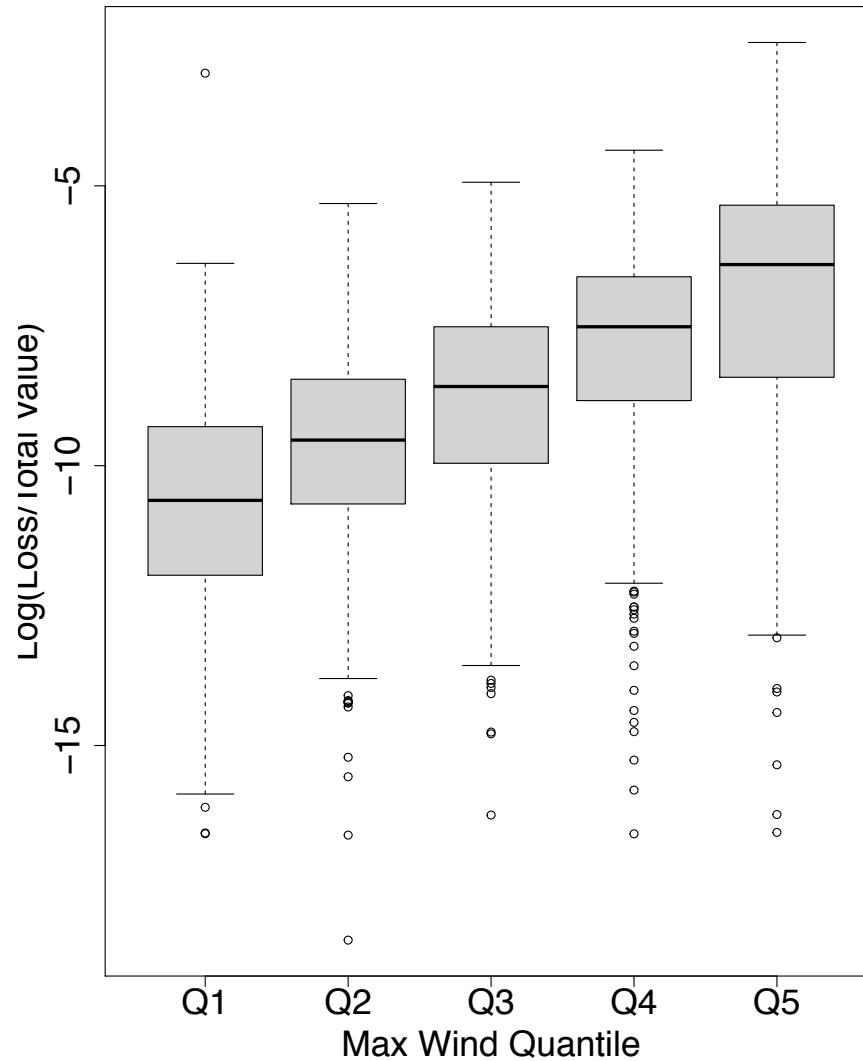


Hurricane Frances (2004)

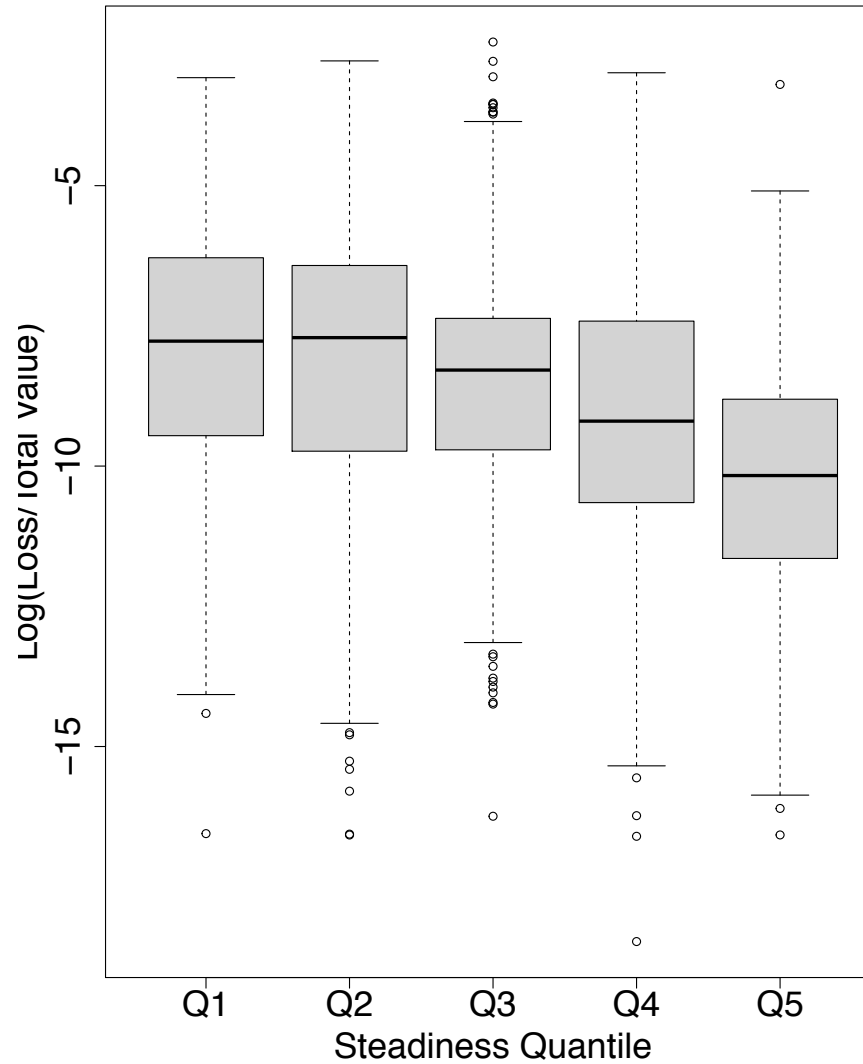


Data source: NOAA H*WIND

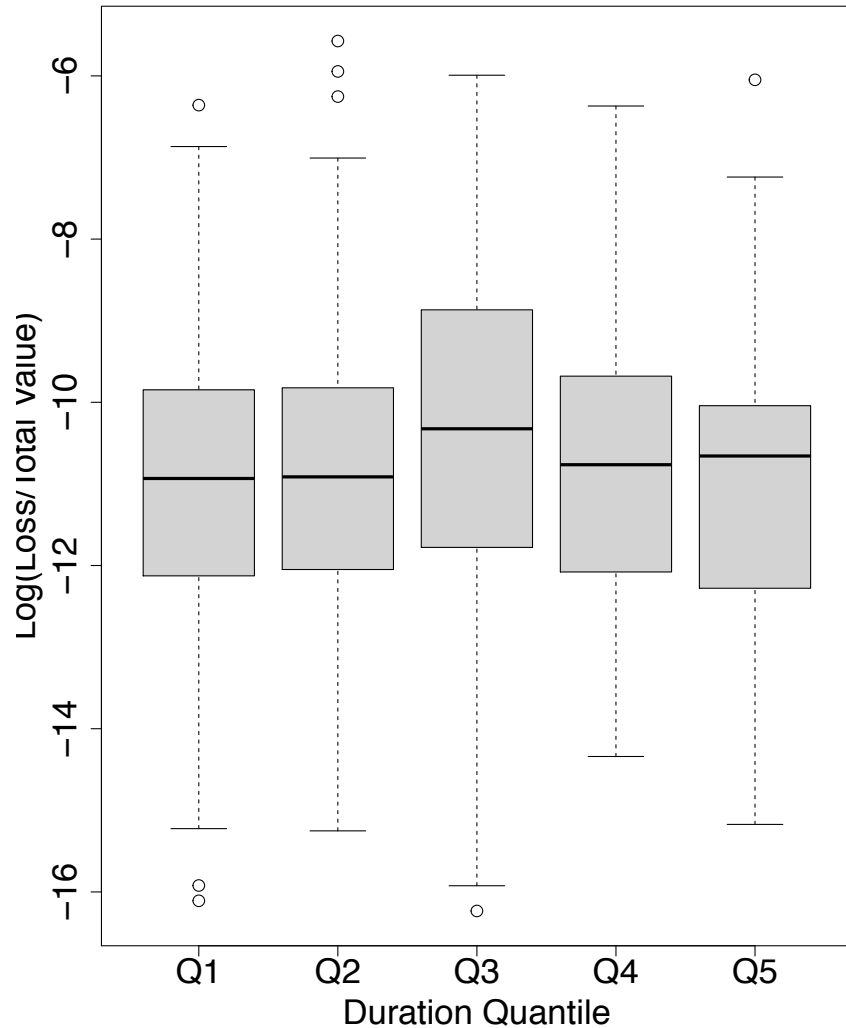
Loss increases with wind speed



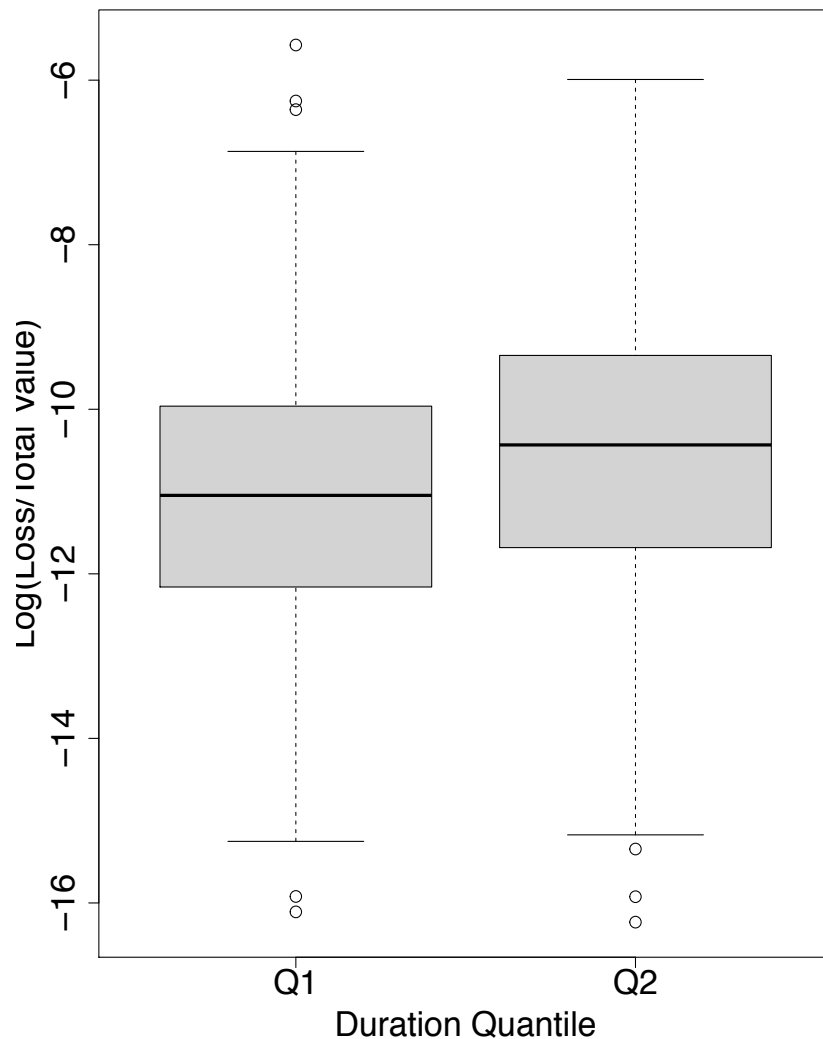
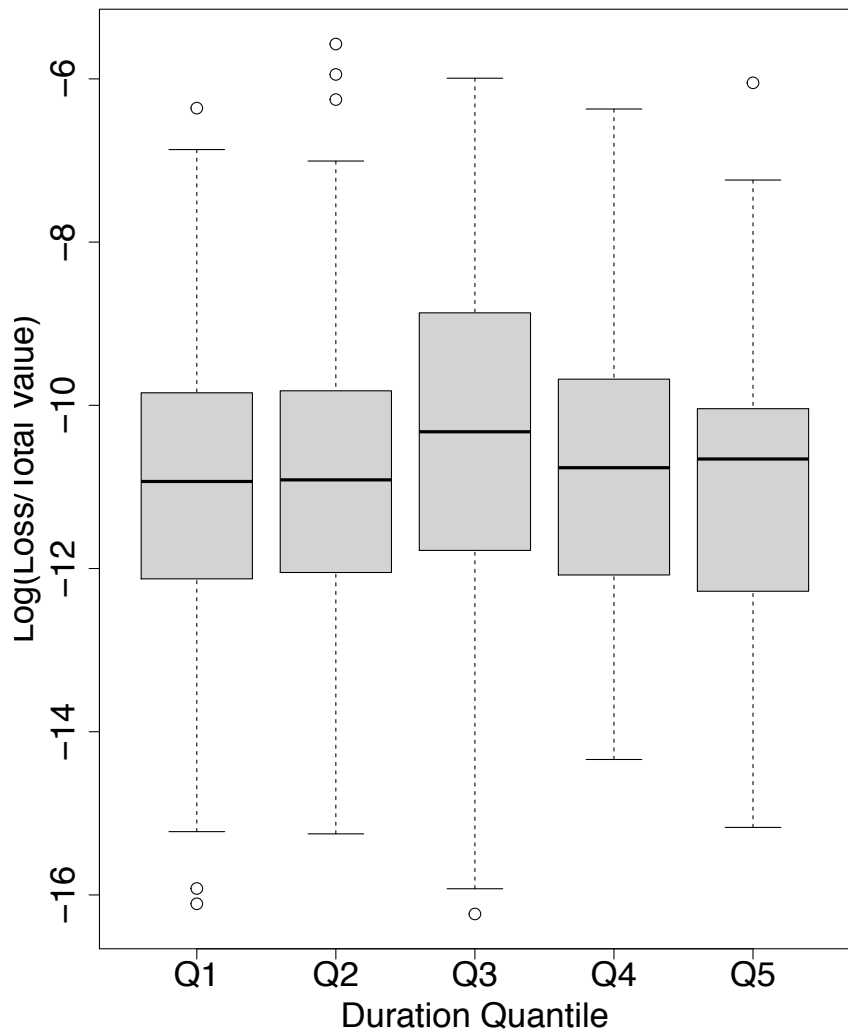
Loss decreases with steadiness



Duration vs. Loss



Duration important at low speeds



major_hurricane
minor_hurricane
high_duration
high_steadiness

$$\text{Ln}(\text{losses}) = f(\text{categorical wind factors} + \text{exposure and vulnerability factors} + \text{interaction effects} + \text{time and space fixed effects})$$

built_2000s = 1 if homes built in the 2000s

Multiple wind parameters drive loss

Parameter	Coefficient	Significance
	Estimate	Pr > t
major_hurricane	2.49	<.0001
minor_hurricane	1.76	<.0001
high_duration	0.50	<.0001
high_steadiness	-0.78	<.0001
built_2000s	-1.13	<.0001
# obs	10564	
r^2	0.34	

- Loss sensitive to wind speed, then steadiness, then duration.
- Homes built to code drive down losses by 68% compared to homes not built to code.

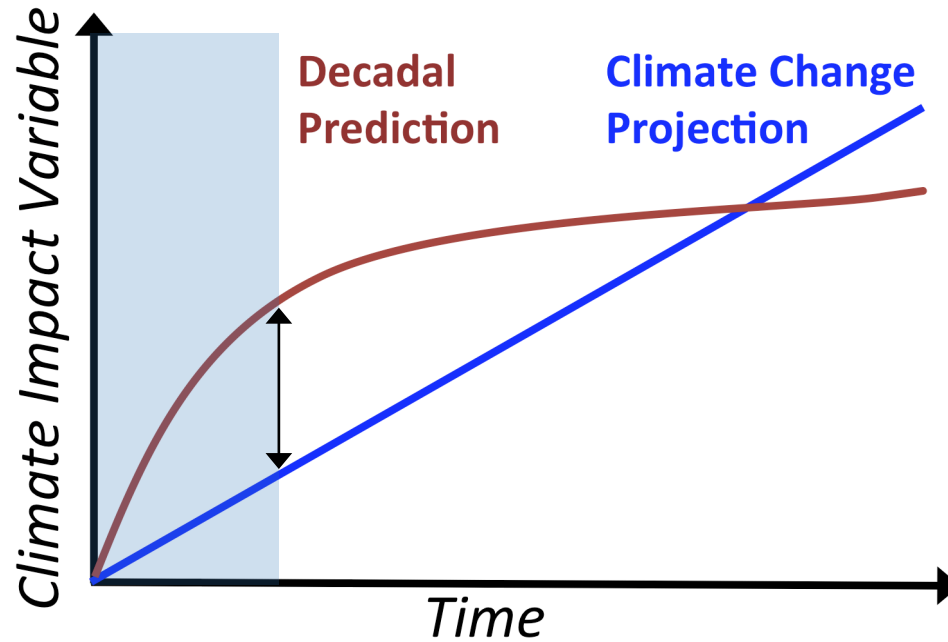
Done et al. (2017)

Understanding Decision Climate Interactions on Decadal Scales

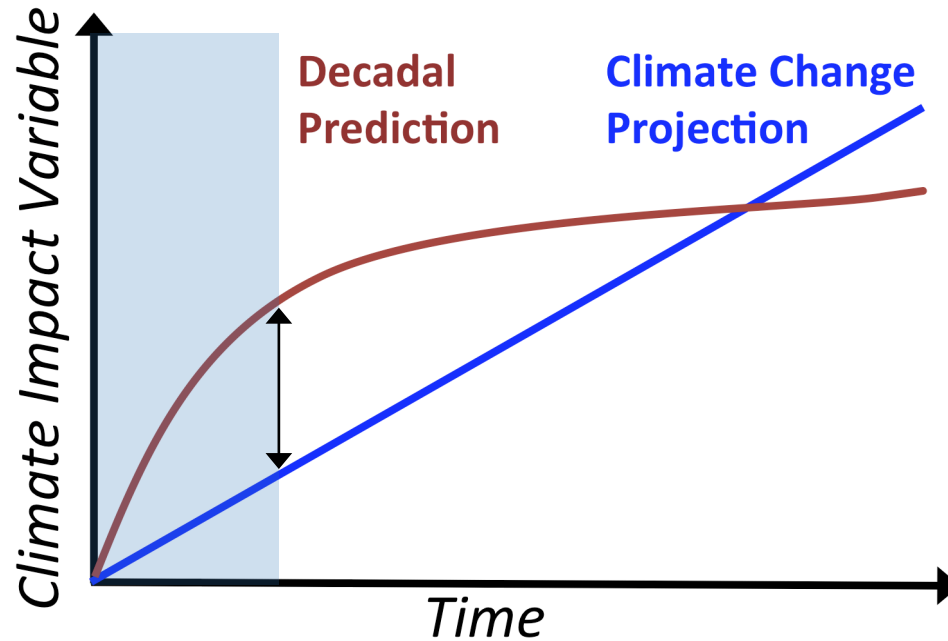
Stakeholders:	Water resource and flood control managers
Need:	Operations, modest infrastructure
Current practice:	Daily, seasonal forecasts and climate change
Hypothesis:	Decadal prediction is useful
Physics:	Remote controlled local, decadal climate
Results:	Intersection of need and decadal prediction
Resilience Action:	Informed operations, medium-term planning

Thanks to the UDECODE team

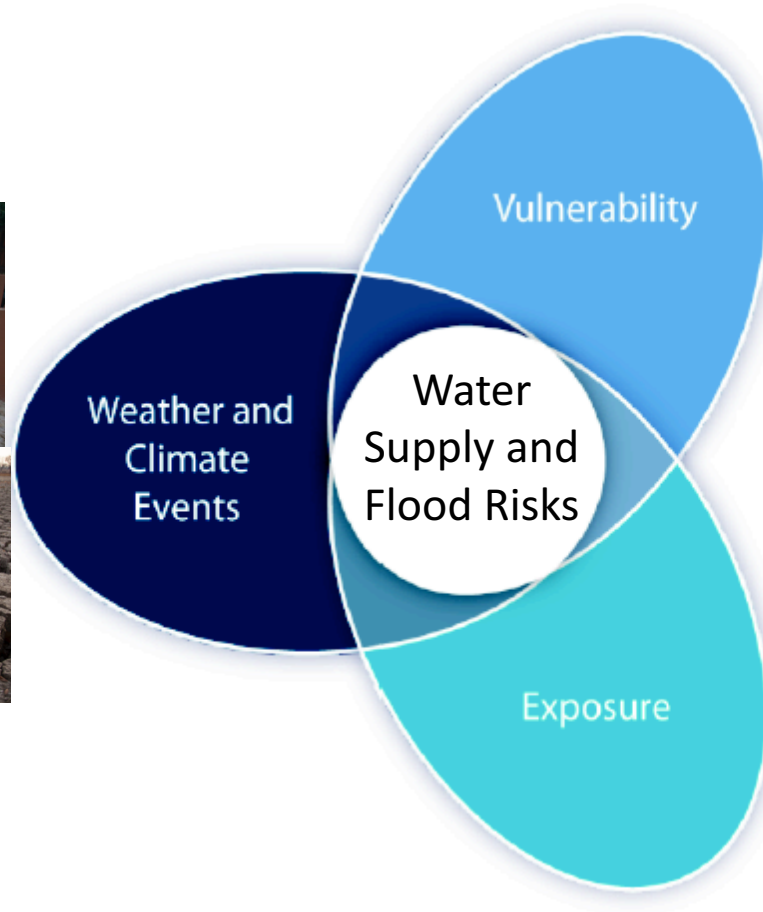
Why Decadal Prediction?



Why Decadal Prediction?

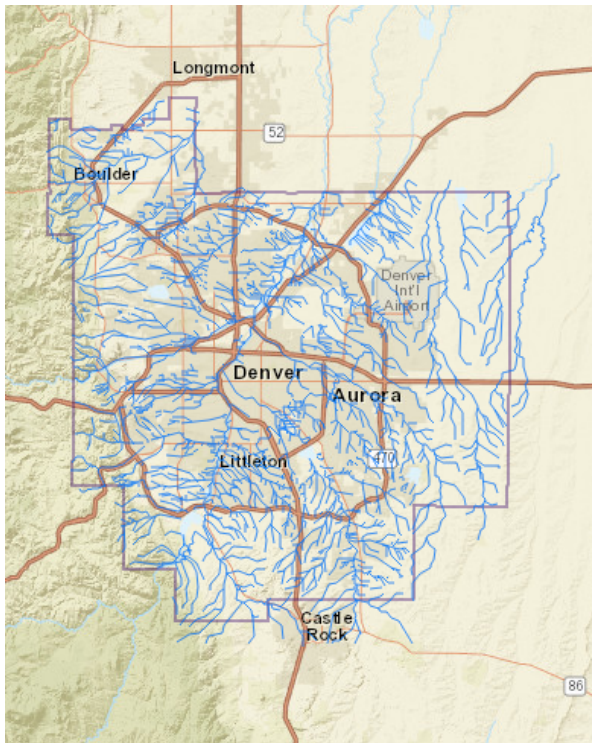


Minimizing supply risk, and flood risk



*Adapted from
IPCC SREX, 2012*

Peak flow and sustained high flow likelihoods, for the design and construction of natural channels.



Heather Lazrus

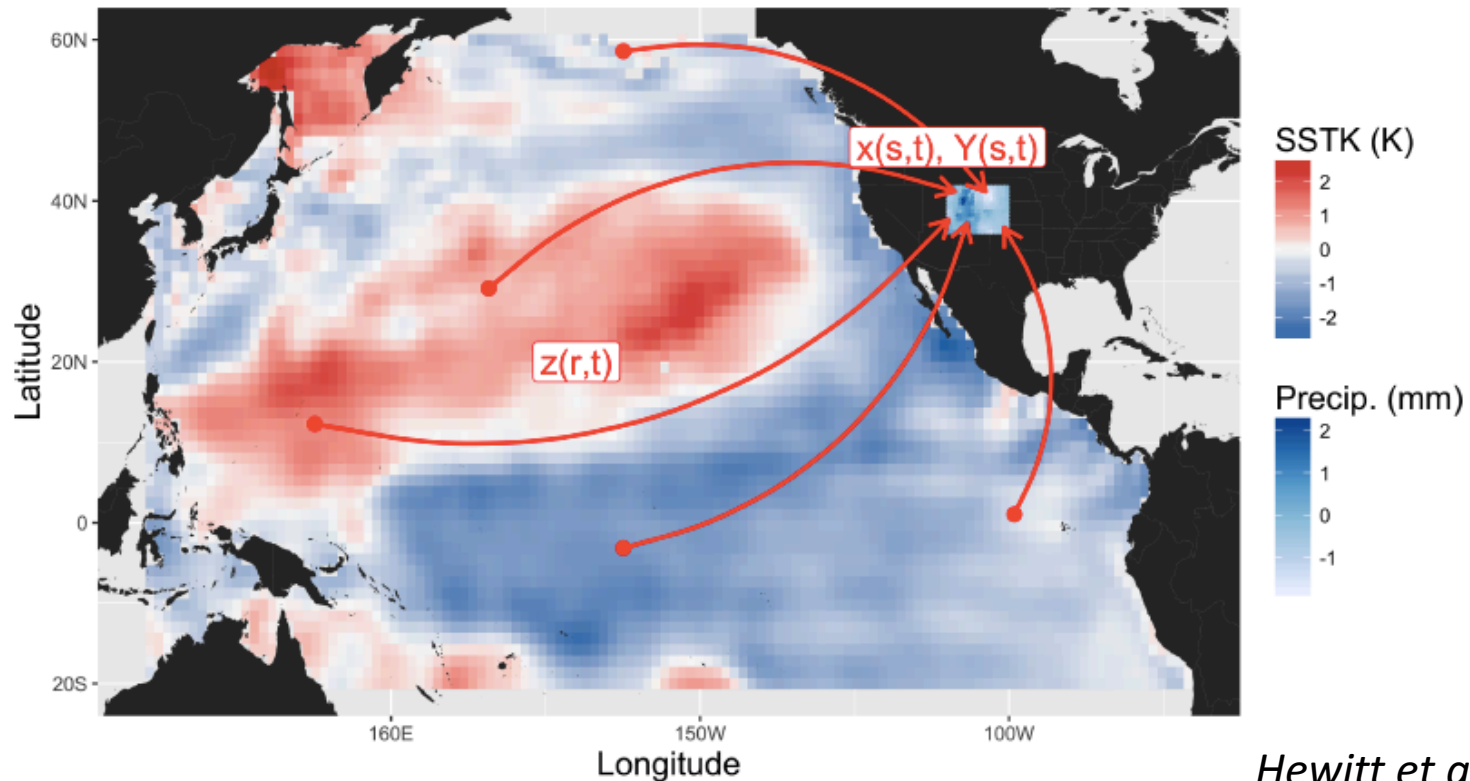
Number and characteristics of big precipitation events, for drought relief and reservoir management.

Heather Lazrus



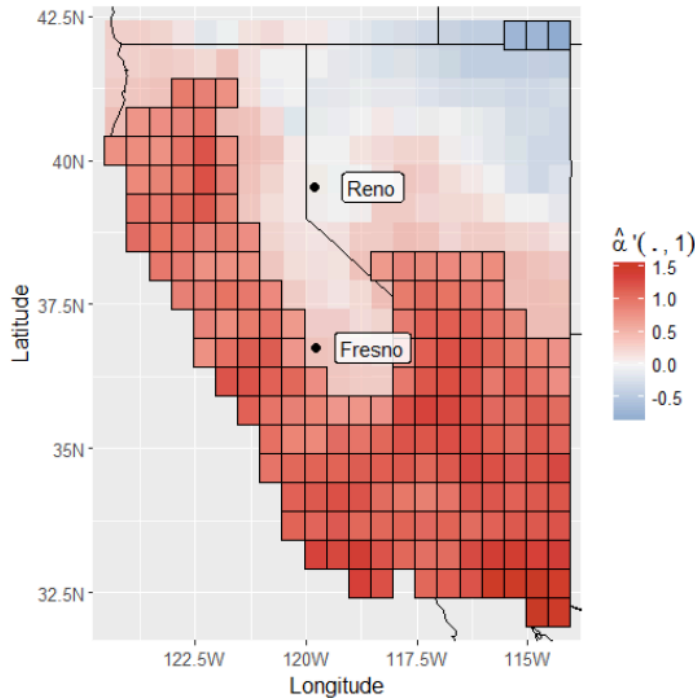
Can SSTs Predict Precipitation?

$$\underbrace{Y(s, t)}_{\text{Std. Precip. anomaly}} = \underbrace{x(s, t)^T \beta}_{\text{Local effects}} + \underbrace{w(s, t) + \varepsilon(s, t)}_{\text{Spatial + Independent error}} + \underbrace{\int_{D_Z} z(r, t) \alpha(s, r) dr}_{\text{Teleconnection effects}}$$

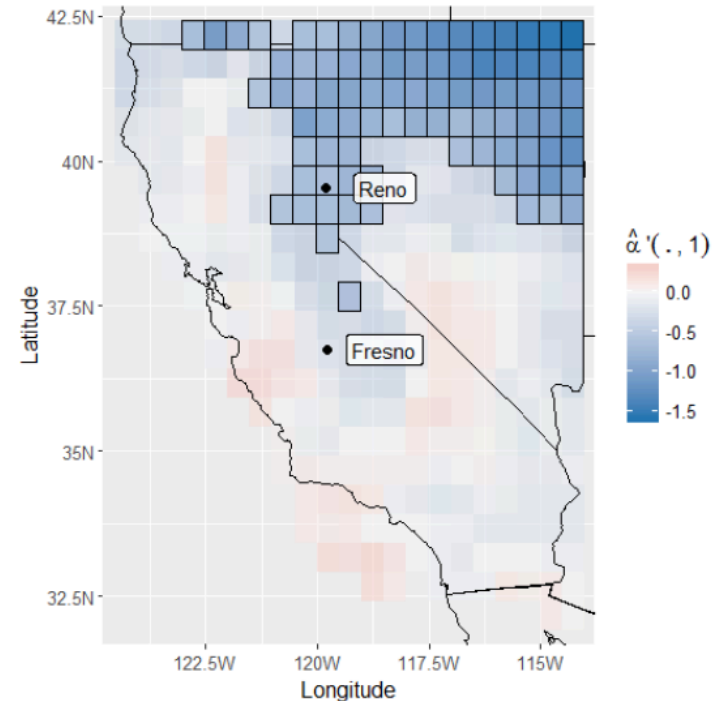


Hewitt et al. (2017)

Model with remote effects only.



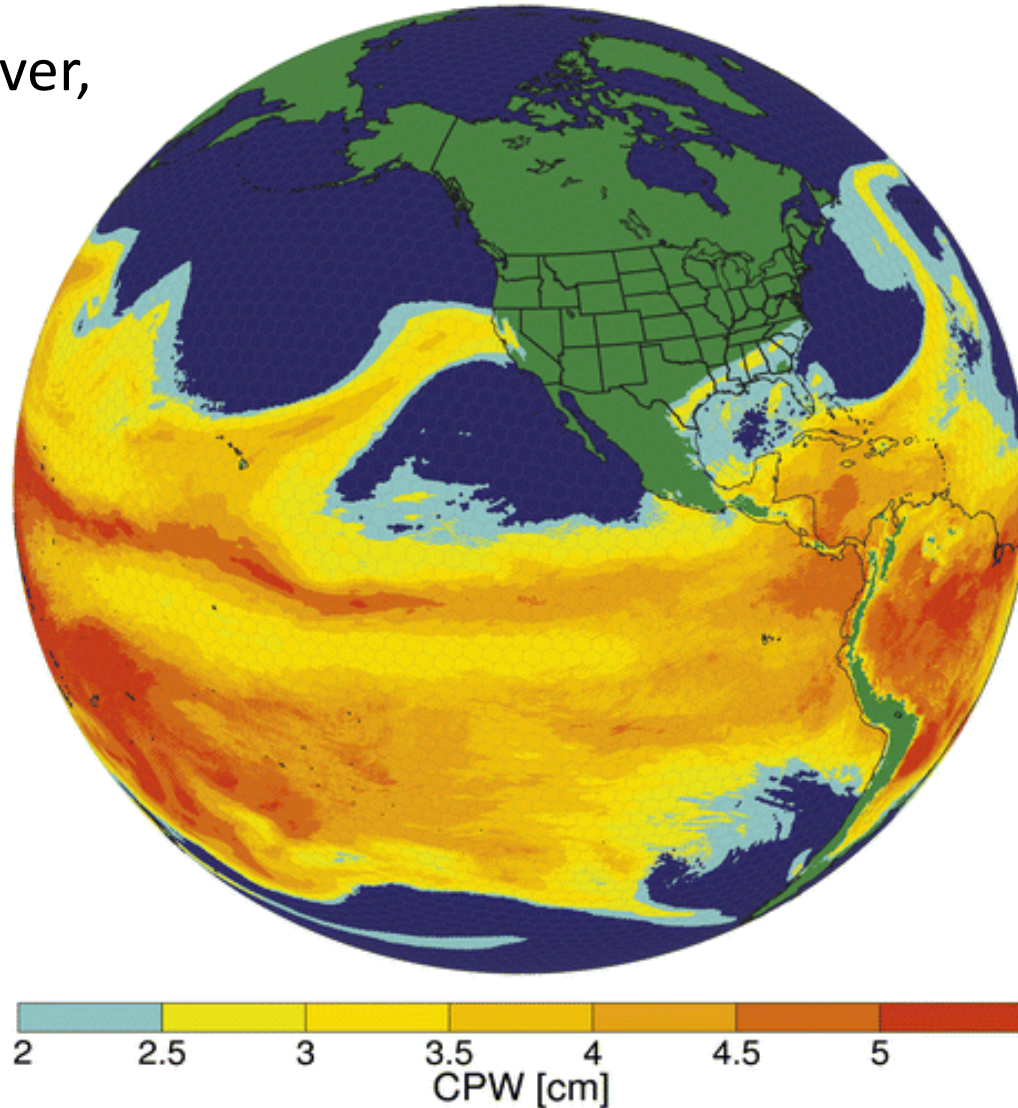
Model with remote + local effects



Casey Shafer, Josh Hewitt, Jennifer Hoeting (CSU)

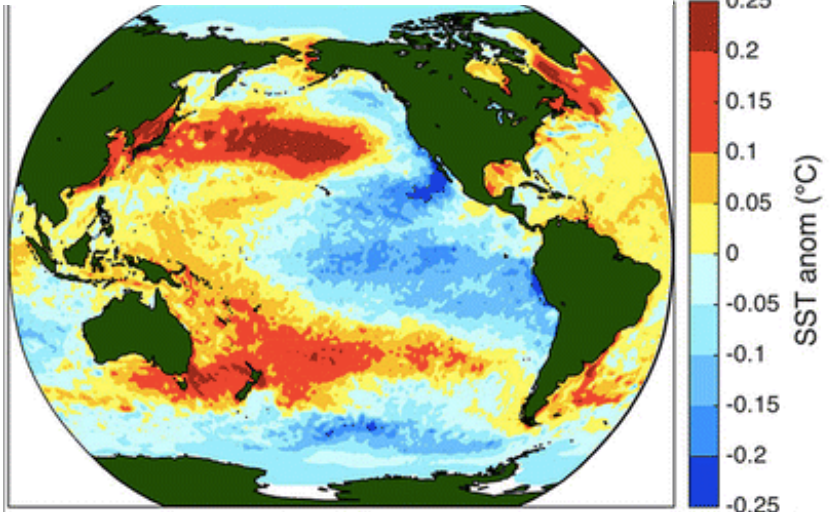
Planned MPAS Experiment

Atmospheric River,
MPAS

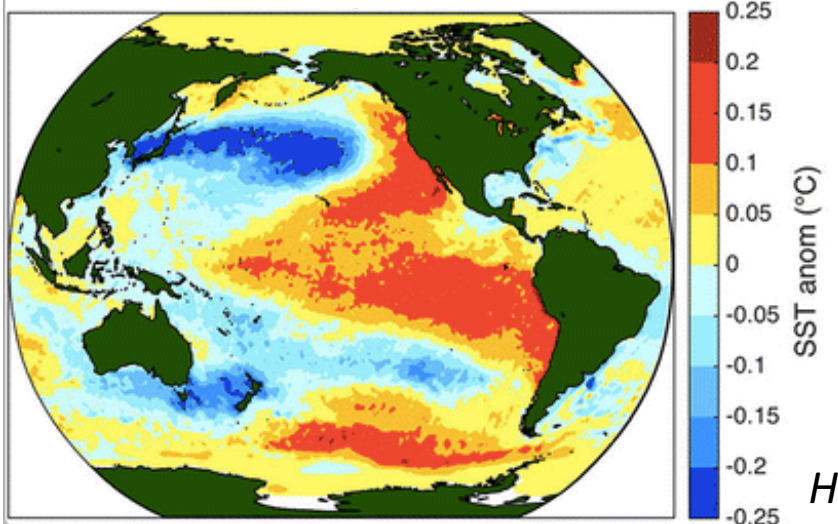


Hagos et al (2015)

Negative Phase PDO



Positive Phase PDO



AR characteristics:

- timeseries analysis
- rain/snow ratio, freezing level.

Henley et al (2015)

- Physical science informed by needs
- Compatible with management practice
- Two-way
- A key component of a broader effort.



Significant advances expected through understanding mesoscale phenomena at global scales.

Collaborate through



NCAR is sponsored by the National Science Foundation