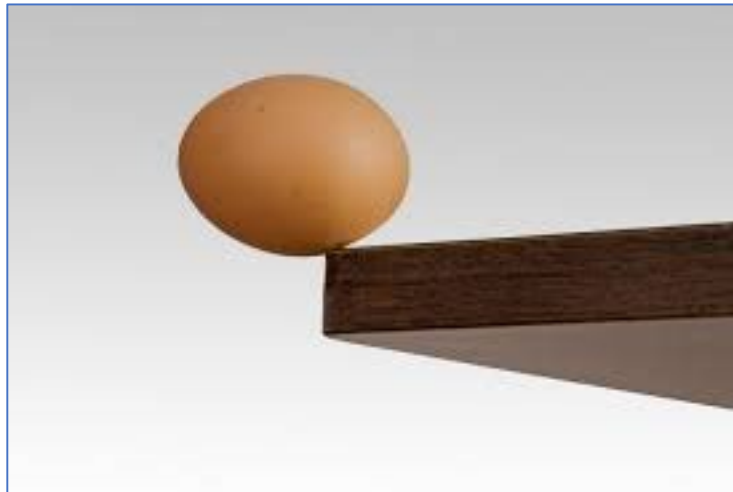


# Tipping Points in Weather Prediction

Extreme sensitivity of forecasts to the atmospheric state and what to do about it



# “Ordinary” Forecasts

Extended Forecast for  
Phoenix AZ

Today



Sunny

High: 89 °F

Tonight



Clear

Low: 64 °F

Saturday



Sunny

High: 91 °F

Saturday  
Night



Clear

Low: 65 °F

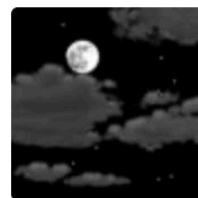
Sunday



Sunny

High: 92 °F

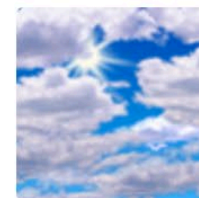
Sunday  
Night



Partly Cloudy

Low: 67 °F

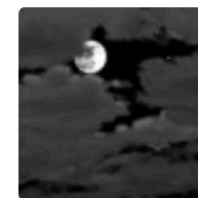
Monday



Partly Sunny

High: 89 °F

Monday  
Night



Mostly Cloudy

Low: 66 °F

Tuesday



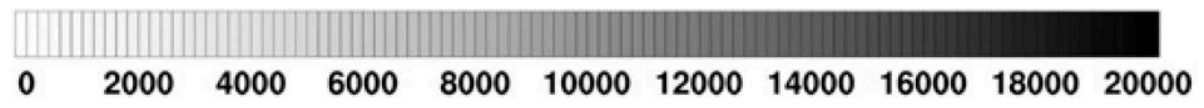
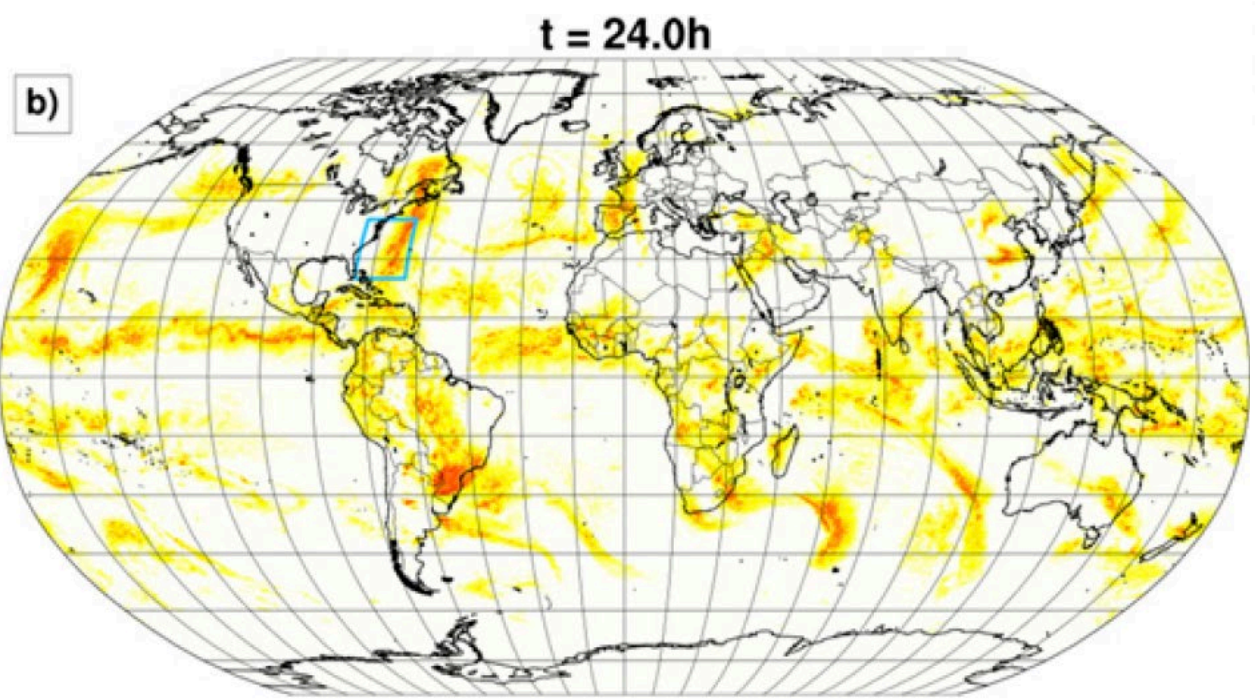
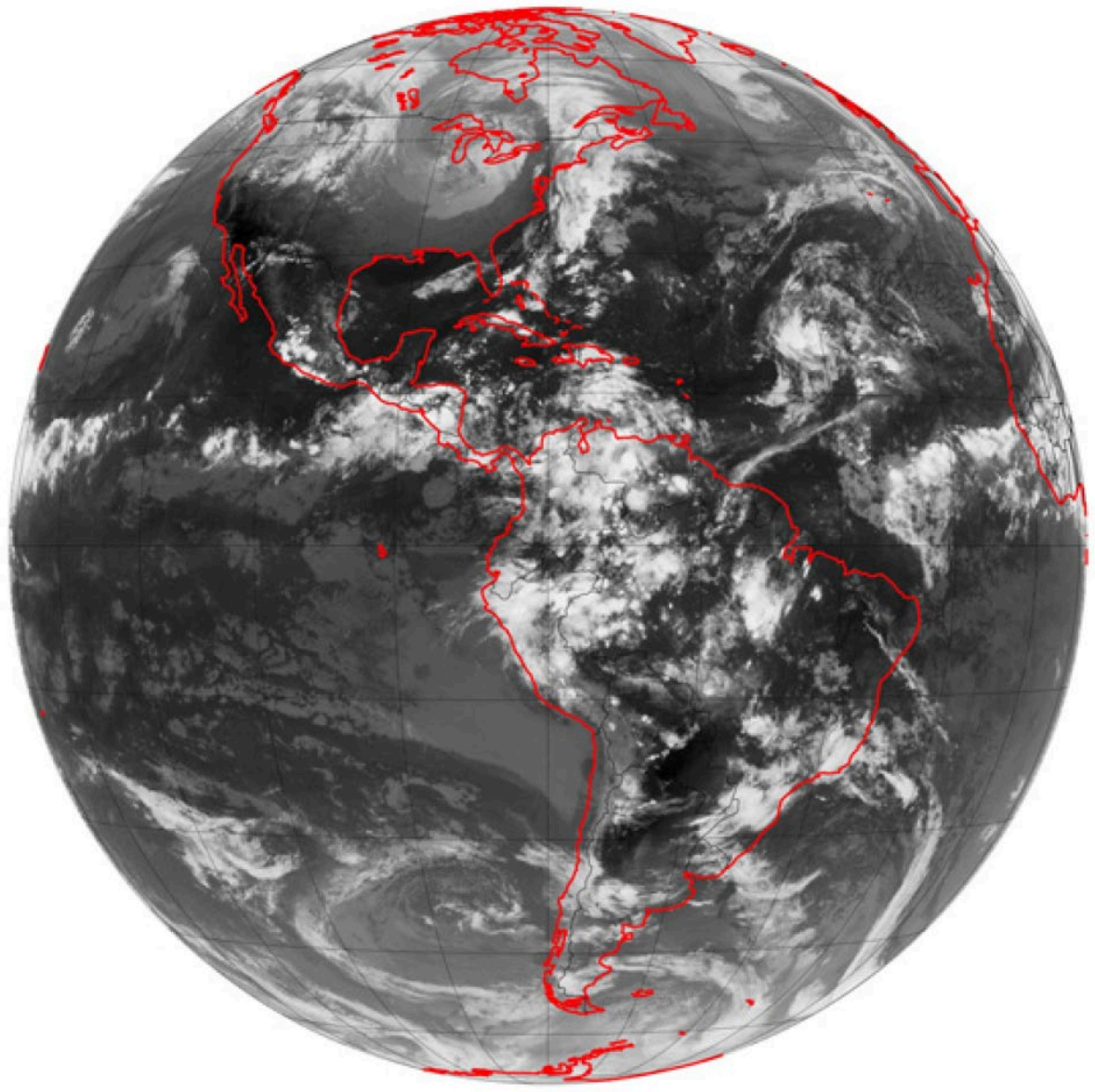
Mostly Sunny

High: 86 °F

**Not what we are talking about**

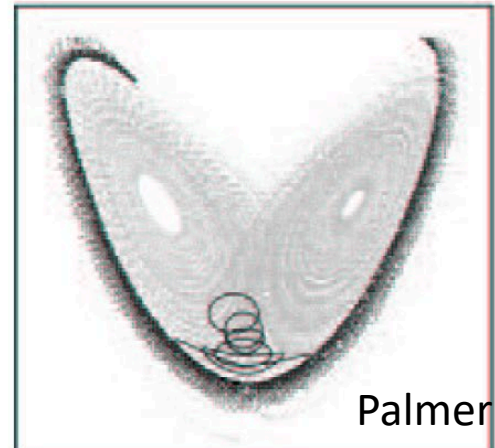
# Error growth is localized and “feature” based

(a) GOES-East IR Image (0000 UTC 20 October 2012)



# Extreme Sensitivity

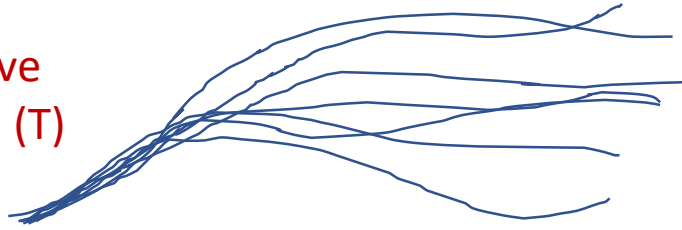
- Indicative of strong non-linearity and extreme error growth
- Often applied to the climate system, but is more general
- Deterministic prediction is difficult (impossible?) when system is in a state of extreme sensitivity





# Types of Thresholds

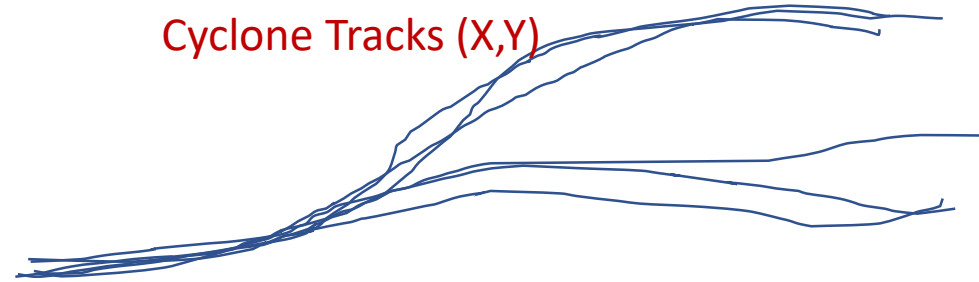
Convective  
initiation (T)



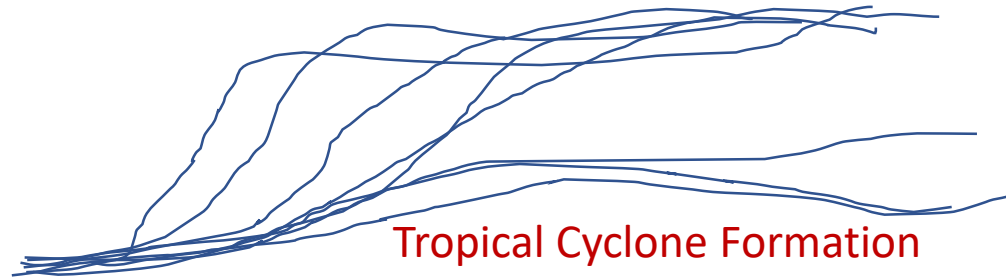
time →

Transition to very uncertain state

Cyclone Tracks (X,Y)



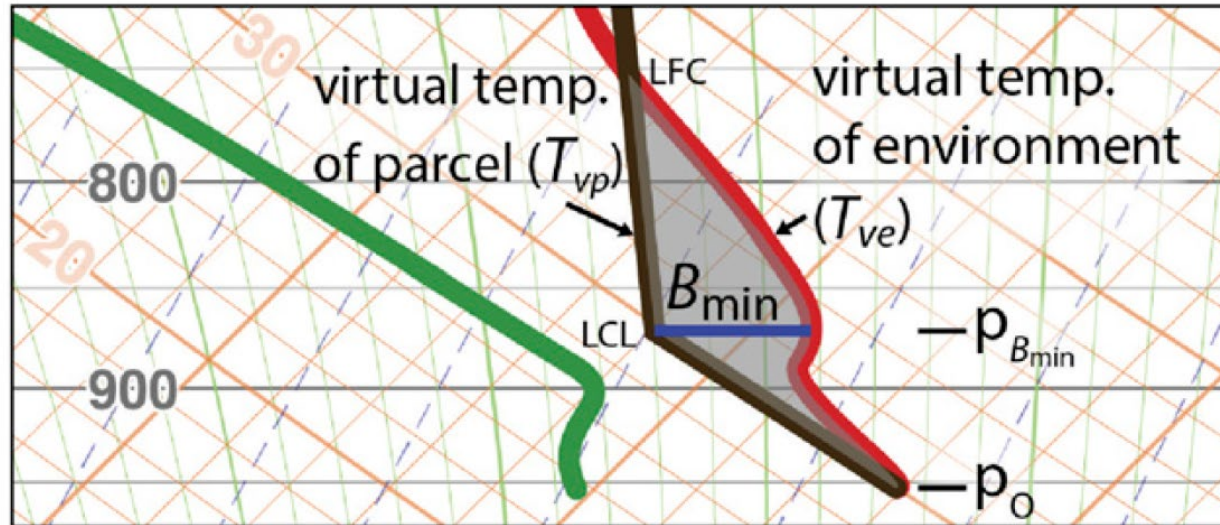
Transition to separated states (e.g. clusters)



Tropical Cyclone Formation  
and Intensification (V);  
Warm-front passage (T)

Uncertain transition

# Convection Initiation

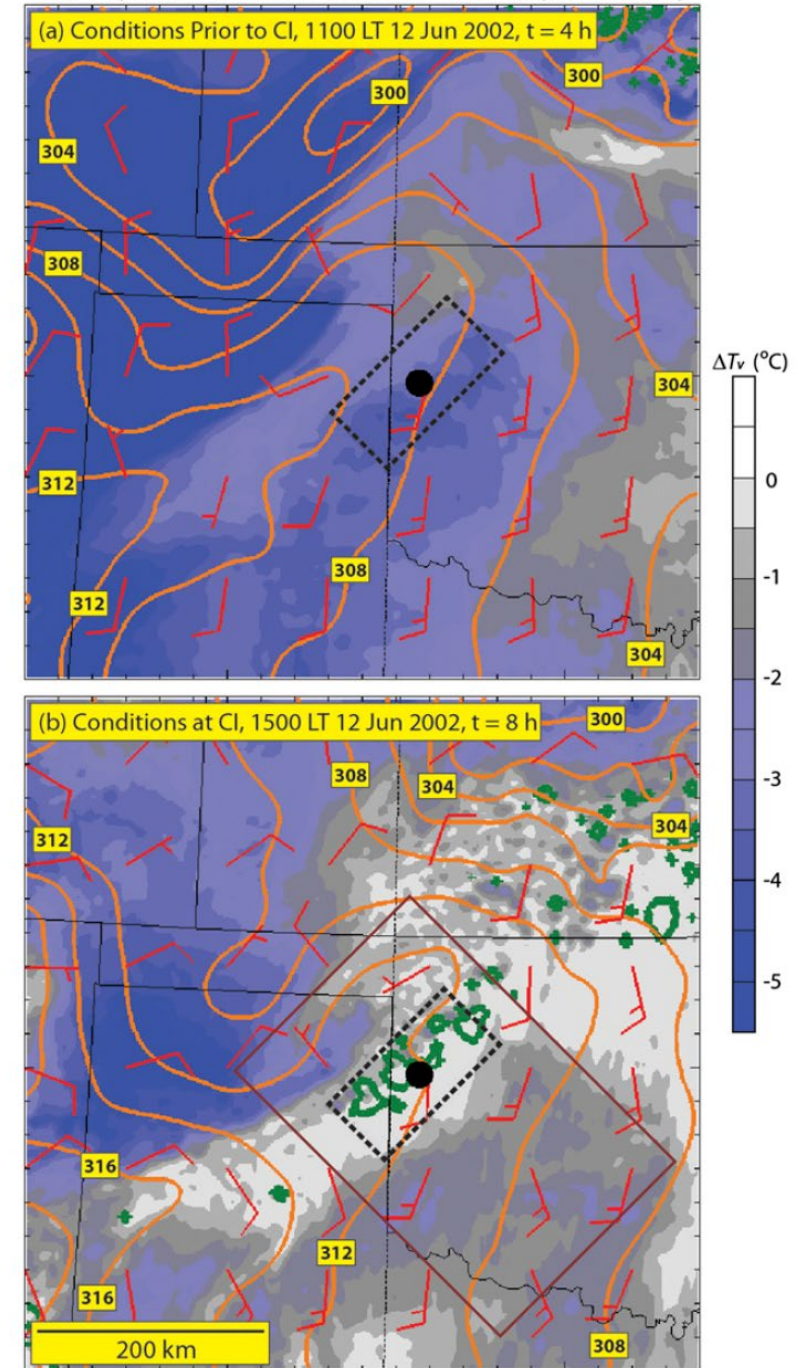


$B_{min} \sim 0$  is “tipping point”

$$\left. \frac{\partial B}{\partial t} \right)_{p_B} = \frac{\partial T_{vp}(p_B)}{\partial t} - \frac{\partial T_{ve}(p_B)}{\partial t}.$$

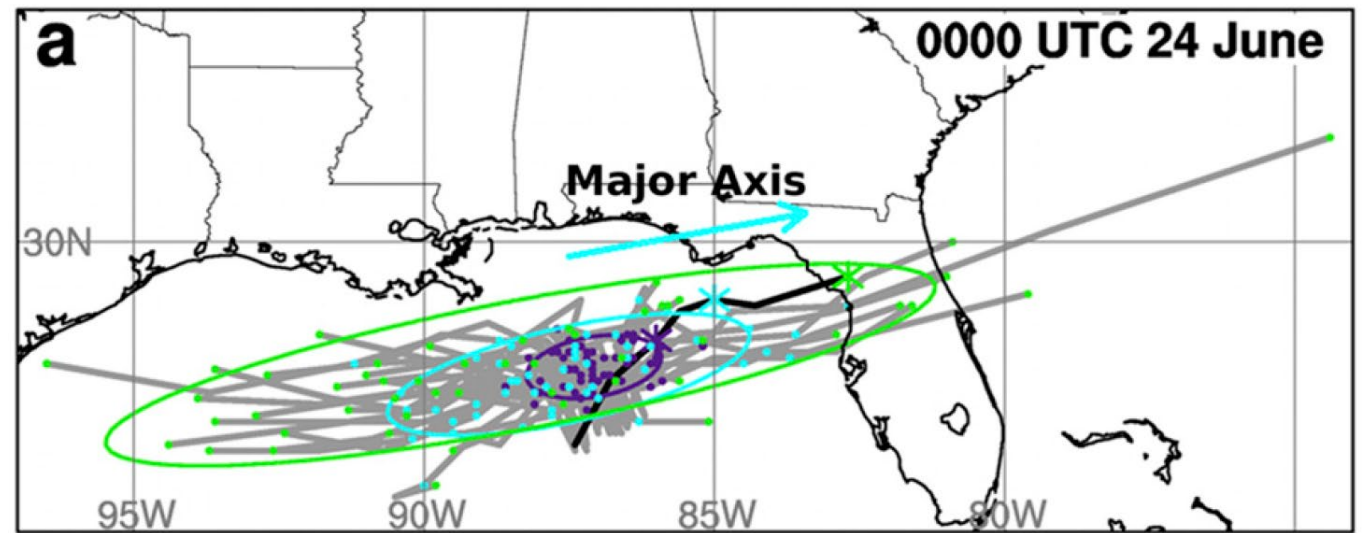
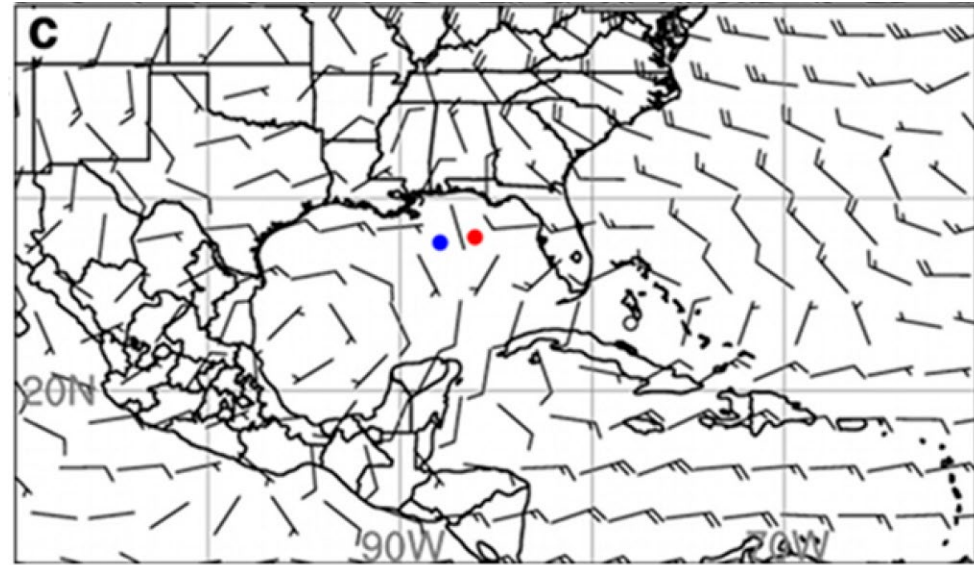
Large trend means even if  $B_{min} \sim 0$ , system may have some predictability

Case 1  $\theta$ , Winds and Bmin at Model Level 42 (0.2 km AGL)



# Deformation and Blocking

- tropical convection – mid-latitude interaction
- hurricane motion
- Split flows and blocking



Torn et al., 2018: MWR

# What can be done about extreme sensitivity?

- Relate ensemble spread to features to simplify interpretation
  - Already done informally in forecasting
  - Machine learning (e.g. Gagne et al. 2017, W&F)
- Focused observations for specific sensitivities
  - example: TC position relative to axis of contraction (in deformation)
- Predict the predictability: quantify forecast confidence  $C(t)$

**Hypothesis: Even with extreme sensitivity, the time of a marked change in confidence may be predicted, even if the outcome itself cannot be.**



# Very Uncertain Forecasts: Predicting Predictability

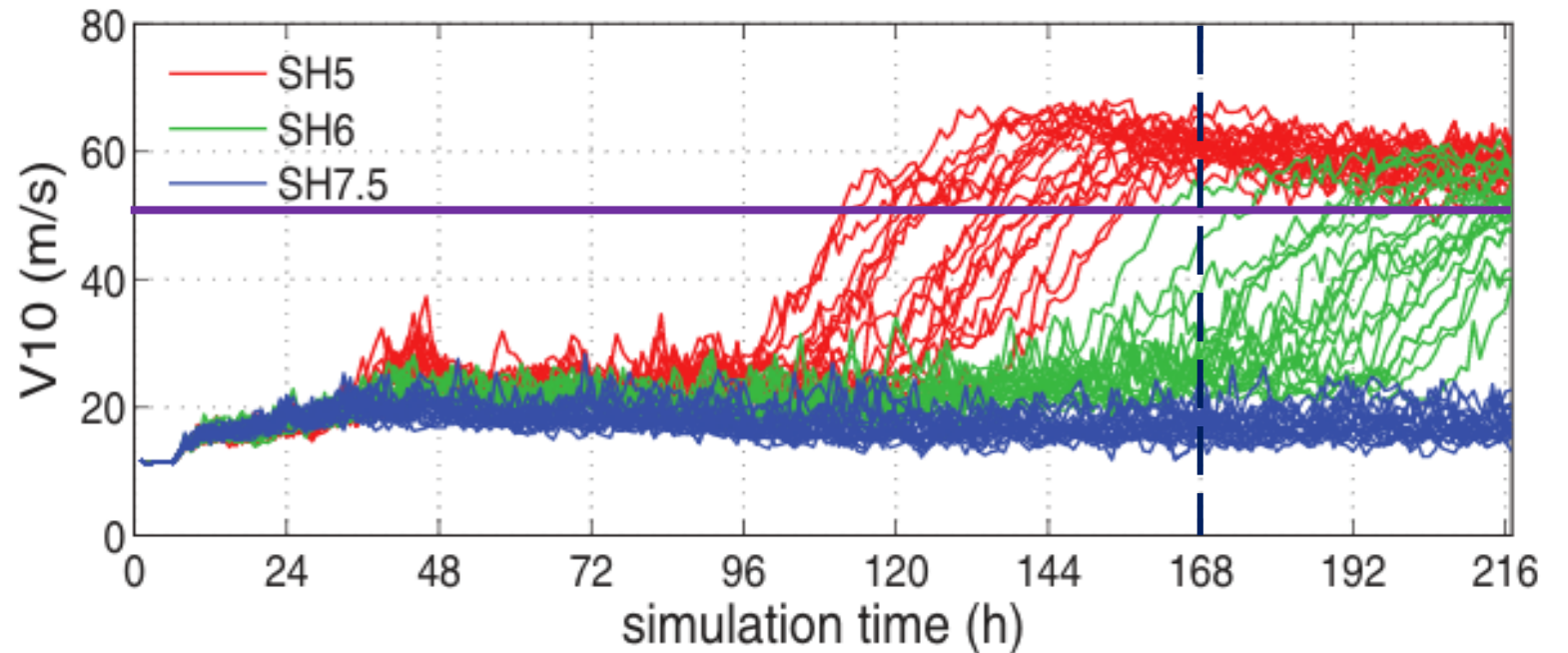
Emergency Manager (EM) wants forecast at day 7: Will there be a major hurricane (MH)?

$P(\text{MH}) \sim 0.35$

EM says that is not good enough to make a decision.

When will EM know with 80% confidence about a MH at day 7?

If you say 'Day 6', you are fired.

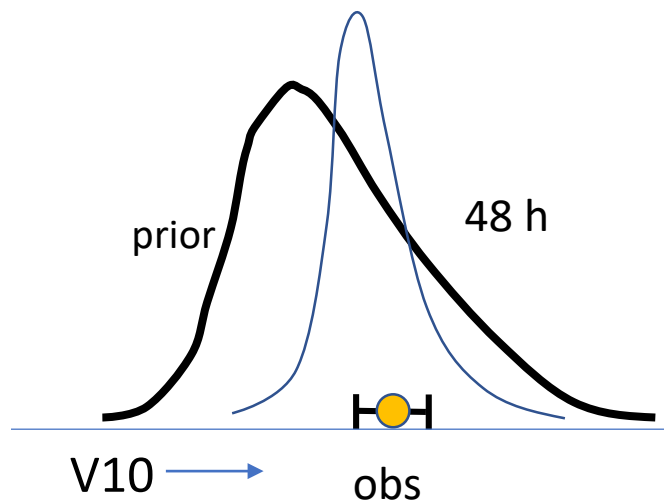
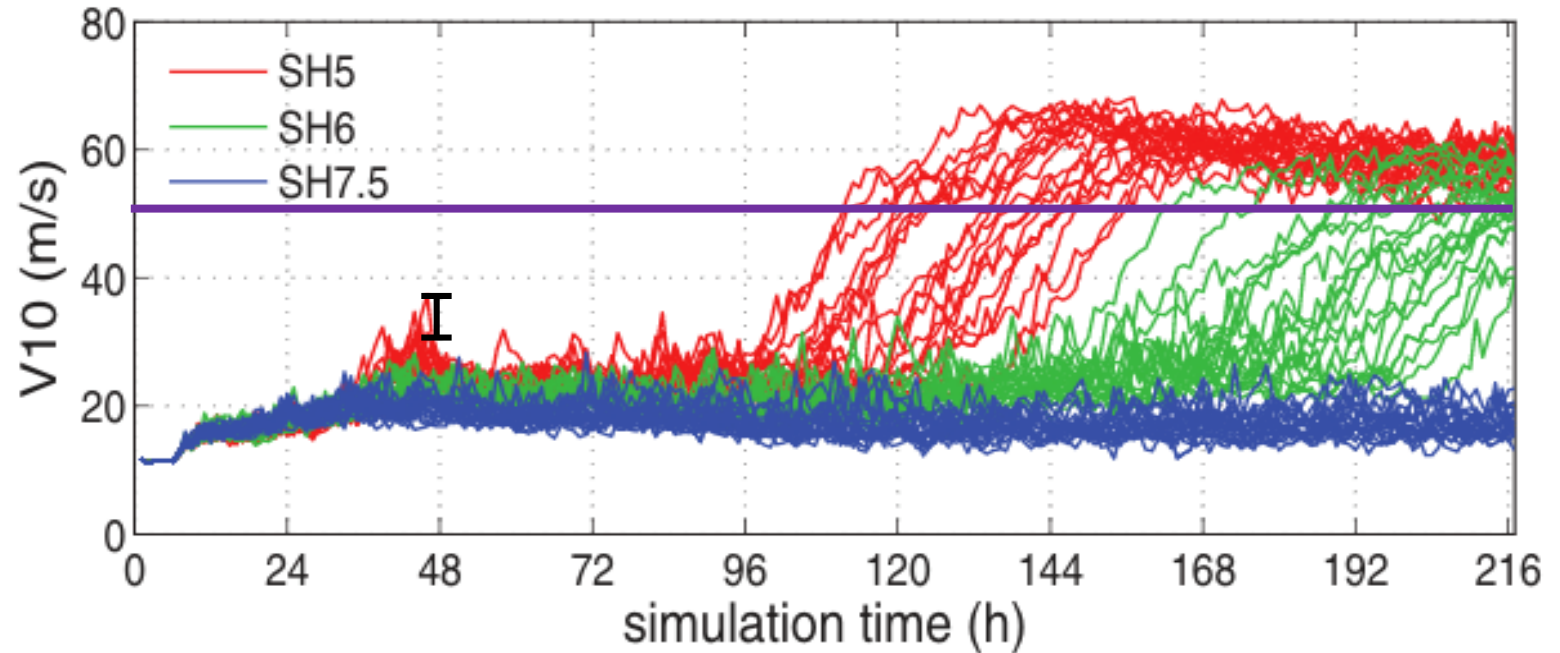


Tao and Zhang, 2015: JAMES

Two issues:

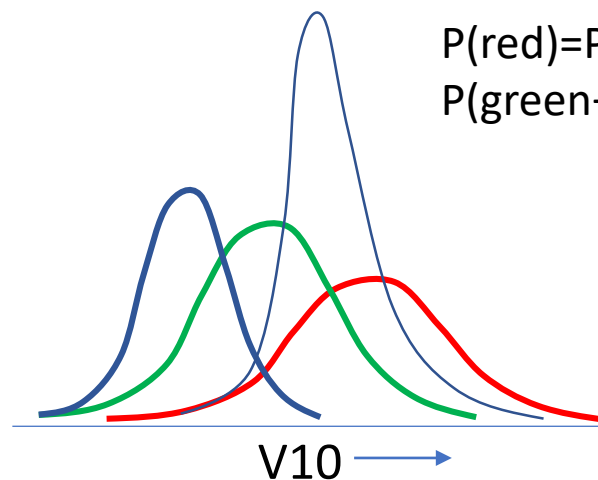
1. Uncertainty for a given environment
2. Uncertainty about the environment

$P(\text{MH}, t=0) \sim 0.35$  (at 168 h)



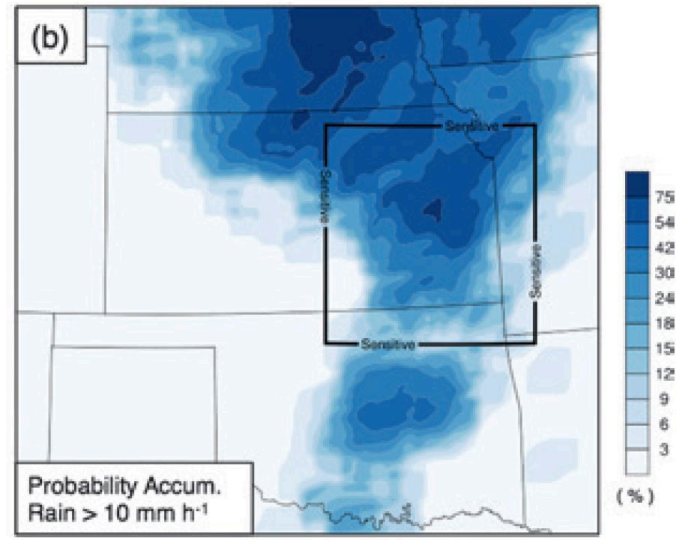
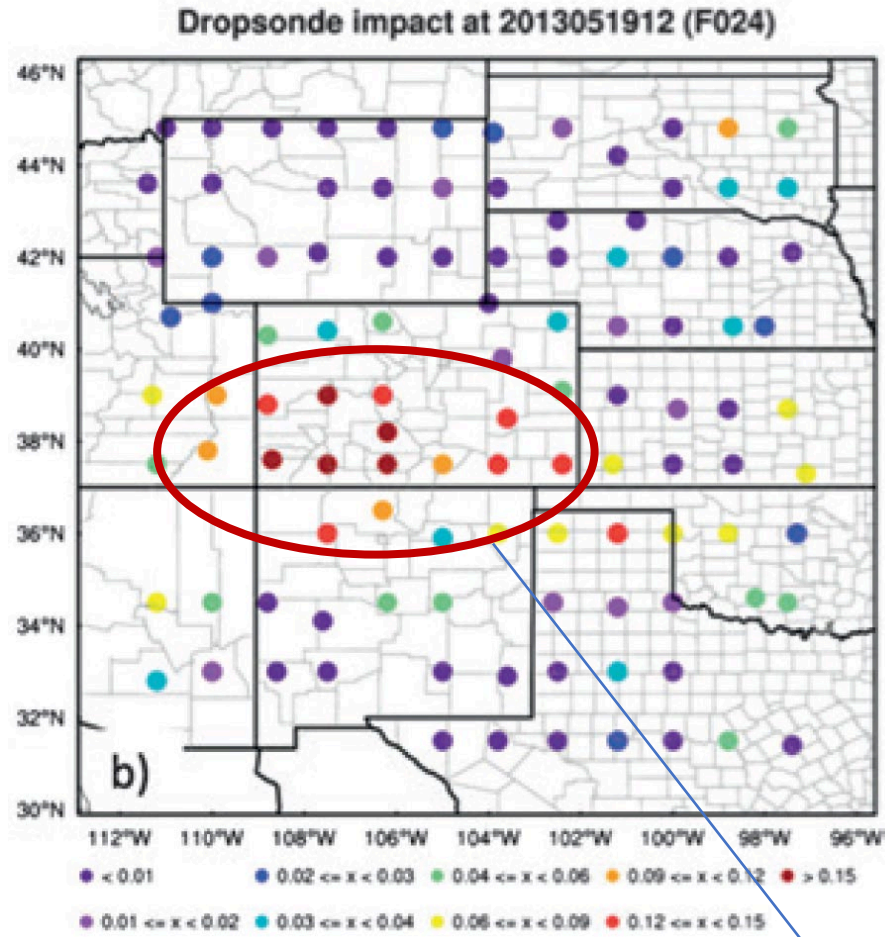
$P(\text{red})=P(\text{MH})\sim 0.7$

$P(\text{green+blue})=P(\text{no MH})\sim 0.3$



- Now, consider the full distribution of possible observations
- Then consider observations preceding this time

# Mesoscale Predictability Experiment



One could ask: Given the actual observing system, and its errors, when will I know more certainly the rainfall in the box?

Trapp et al., 2015: BAMS

Drops here will have the largest influence on rainfall 12 h later in box

# Summary

- Extreme sensitivity => extreme uncertainty
- Predicting “confidence” in a scenario
- Advances: Ensemble techniques, ensemble sensitivity (or adjoint sensitivity)
- Issues: Does this make any sense? More formalism
- Challenges
  - Relate ensemble variation to “features”
  - Requires clustering in ensemble outcomes
  - Challenges: coupling machine learning and data assimilation
- Requires reduced model error; focused observations may help

