# Atmospheric Predictability

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### Initial-Condition Errors: Scale Sensitivities

Consider two *different* questions

Is upscale error growth important?
(even if it is not exactly a "spectral cascade")

 Given initial errors of *fixed absolute magnitude*, does their *horizontal scale* influence predictability?

#### Lorenz's 1969 Answer: Experiments A & B



"Evidently when the initial error is small enough, its spectrum has little effect upon the range of predictability."

Implications of Experiment B were largely overlooked

# Small *relative errors* in the large-scales can destroy predictability.



#### Influence of Scale: Lorenz Model

- Small relative errors in the large scales rapidly propagate down to the smallest resolved scale.
- Those small-scale errors subsequently propagate back upscale as if they had simply originated in the small scales.
  - Upscale growth is responsible for the finite limit to intrinsic predictability
- No easy way to diagnose the scale of the "original errors".

#### How relevant is the Lorenz model?

- It does not include
  - Baroclinic instability
  - Deep convection
  - Inhomogeneity and nonstationarity
- Nonlinear effects are incorporated only crudely.
- Incorrectly assumed k<sup>-5/3</sup> slope for the background KE spectrum at large-scales.
- Deep Convection?

# Systems

- Four cases: both weakly and strongly forced systems
  - 24-hr control simulations

- WRF model, 2.5 km horizontal grid spacing
- GFS analysis for initial conditions
- Six ensemble simulations branch off each control at hour 6
- Different background perturbations among ensemble members in the near-surface moisture field
  - Monochromatic square wave in horizontal, random phase
    - Small-scale ensemble: x & y wavelengths 20 km ( $\lambda = 14$  km)
    - Large-scale ensemble: x & y wavelengths 200 km ( $\lambda = 140$  km)
  - Perturbation amplitude of 1% of control moisture field
  - 1-km e-folding decay scale away from the surface

# Synoptic Overview

- Sea-level pressure
- 500 hPa heights
- 500 hPa vertical velocity (contours)



#### **Control Simulations**

- Simulated composite reflectivity
- 12 hours after initialization from GFS
- Hour 6 in the ensembles
- 2.5 km horiz. resolution



#### Pertubation KE Growth: April 2017 Case



#### Fractions Skill Score

- 1 mm/hr precip threshold
- 5, 20, 80 km verification radii
- Weak forcing: 14-km perturbations grow faster than 140-km perturbations



#### Influence of Scale – Convective Systems

- Equal amplitude 1% humidity errors at 14 and 140 km produce:
  - Similar losses in predictability in strongly forced cases
  - More rapid error growth in weakly forced cases
- Short-wavelength errors influence convective initiation
  - Important in weakly forced cases
- Long-wavelength errors influence convective organization
  - Important in strongly forced cases

# Implications for data assimilation on the mesoscale

- Characteristic velocities at wavelengths of 200-400 km are 5 times larger than those at 2-4 km.
- Equal improvements: (> 6-hr forecast)
   from reducing IC errors at 2-4 km below 50%
   200-400 km below 10%
   (equal absolute errors in KE')

NEXRAD Coverage Below 10,000 Feet AGL



### Predictability and Microphysics



Fine-scale rain gauge network across ridge

#### MM5 vs Rain Gauges

Black: observations

Gray: MM5 forecast



#### MM5 vs Rain Gauges WY 2005

Black: observations

Gray: MM5 forecast



#### Predictability and "Physics"

Don't test a family of physics parameterizations in simulations using single deterministic initial condition!

#### References

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## Another measure of predictability

#### Fractions skill score

(Roberts and Lean, MWR, 2008)



# Strong/Moderate Forcing

Synthetic radar reflectivity



### Weak Forcing

Synthetic radar reflectivity



#### Implications for data assimilation: I

Parseval's relation

$$\int_{S} u^2(x) \, dx = \int_{-\infty}^{\infty} \hat{u}(k) \hat{u}^*(k) \, dk$$

KE in wavenumber band  $(k_1, k_2)$ 

$$E(k_1, k_2) = \int_{k_1}^{k_2} \hat{u}(k)\hat{u}^*(k) + \hat{v}(k)\hat{v}^*(k) dk$$

#### Implications for data assimilation: II

•  $k^{-5/3}$  KE spectrum

$$\frac{E(k_1, k_2)}{E(k_3, k_4)} = \frac{\lambda_1^{2/3} - \lambda_2^{2/3}}{\lambda_3^{2/3} - \lambda_4^{2/3}}$$

- Ratio of velocities in 200-400-km band to those in 2-4km band is 0.21
- Which is the easier goal? Reduce errors at 200-400 km below 10% Reduce errors at 2-4 km below 50%

Error saturation (KE'/KE) in layer 10 < z < 12 km

- Similar errors at 12 hr in all cases
- Small-scale errors produce more saturation at 6 hr in the weakly forced cases
  - More variation in Cl

