



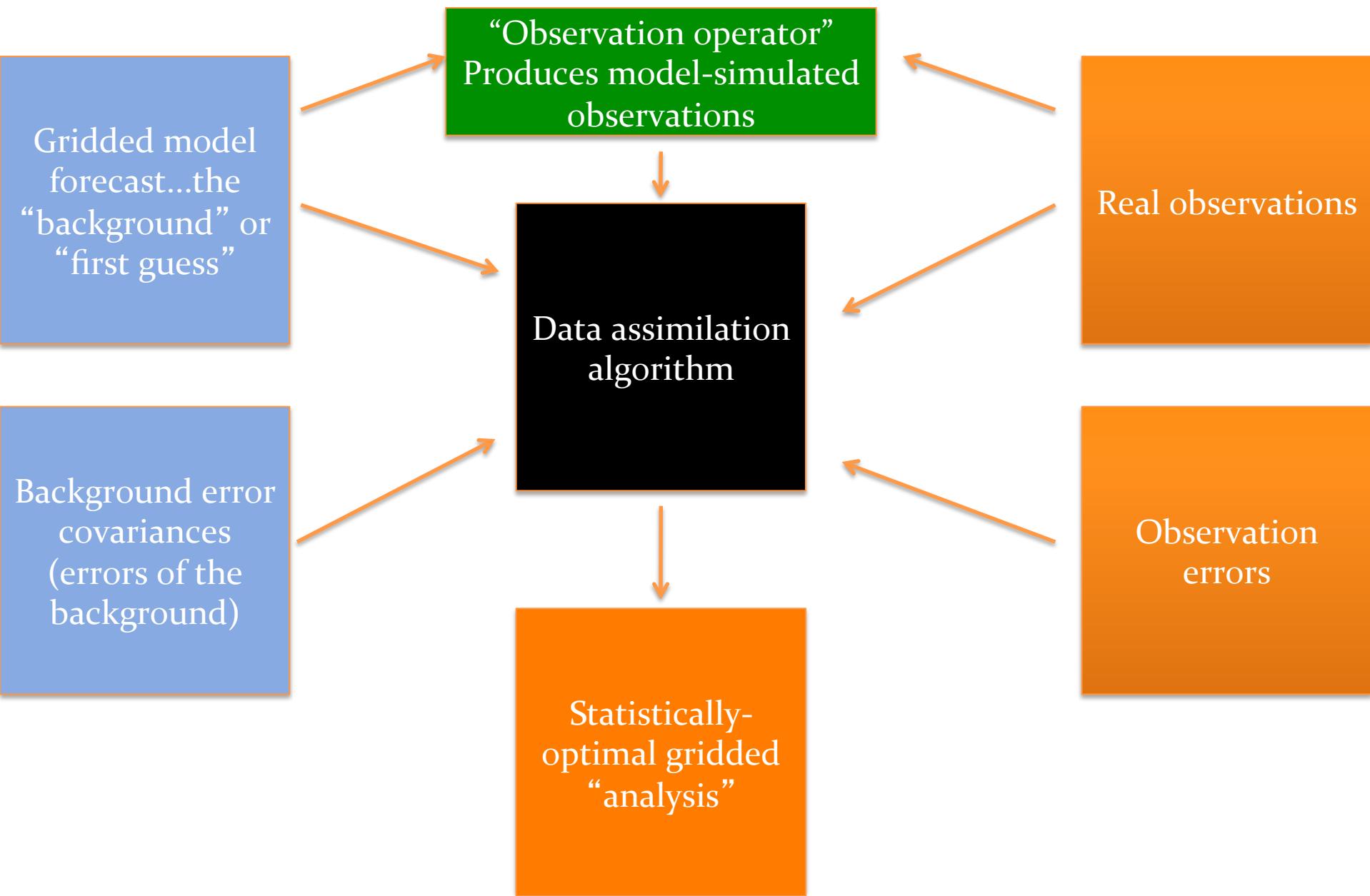
# Practical applications of hybrid variational-ensemble data assimilation approaches

Craig Schwartz  
[\(schwartz@ucar.edu\)](mailto:schwartz@ucar.edu)

NCAR/MMM

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# What is data assimilation?



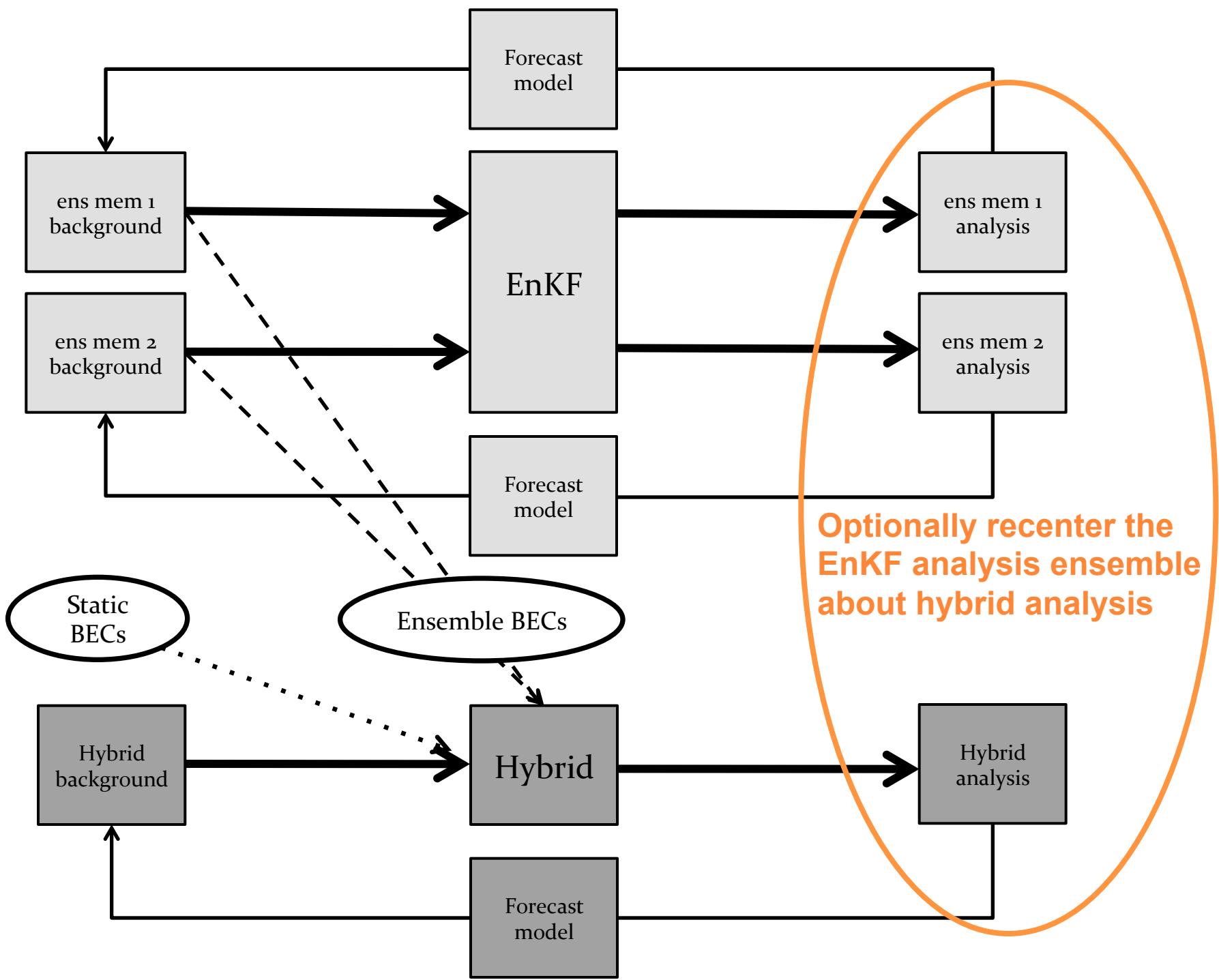
# A few data assimilation approaches

- Three-dimensional variational (3DVAR)
  - Background error covariances (BECs) typically fixed/time-invariant
  - May yield poor results when actual flow differs from that encapsulated within the fixed “climatology”
  - Produces a deterministic analysis
- Ensemble Kalman filter (EnKF)
  - Time-evolving, “**flow-dependent**” BECs estimated from short-term ensemble forecasts
  - Produces an ensemble analysis

# A few data assimilation approaches

- “Hybrid” variational/ensemble
  - Incorporates ensemble background errors within a variational (e.g.,  $\beta$ DVAR) framework
  - Combination of fixed and time-evolving BECs
  - Traditionally produces a deterministic analysis like ( $\beta$ DVAR)





**Dual-resolution hybrid DA with a focus  
on limited-area applications**

# Dual-resolution hybrid DA

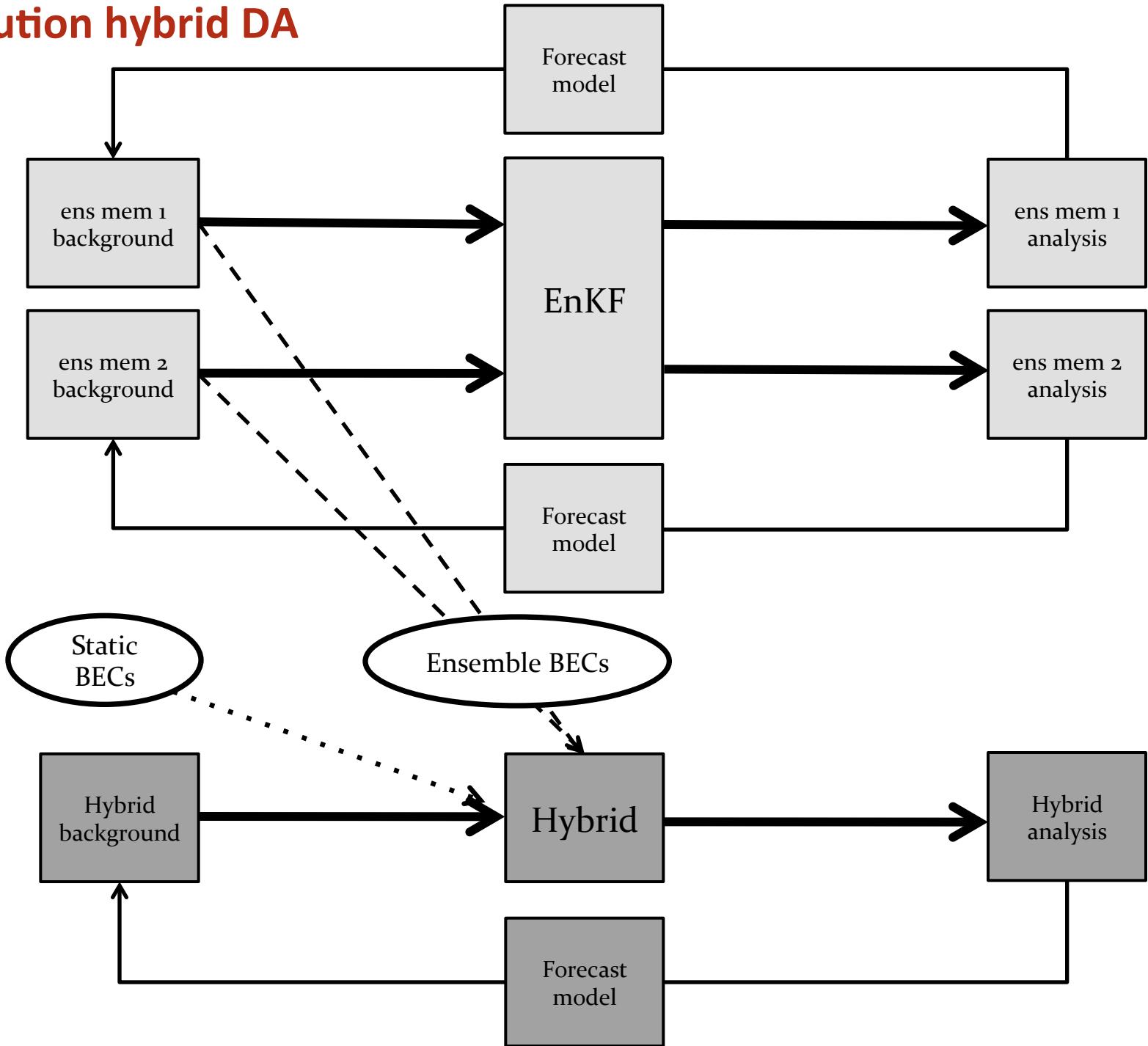
- The hybrid method has repeatedly been shown to initialize better forecasts than 3DVAR
  - Also usually comparable to or better than pure EnKFs
- Compared to 3DVAR, the primary expense of the hybrid is the **ensemble of forecasts**
  - Similar expense as pure EnKF

# Goal of dual-resolution hybrid DA

- We wish to perform high-resolution (HR) hybrid analyses using ensemble perturbations from a low-resolution (LR) ensemble
  - The ensemble *perturbations* provide flow-dependent BECs
- This approach considerably reduces computational expense since producing HR ensemble forecasts is unnecessary
  - The analyses are also faster
- Called “dual-resolution” (DR) hybrid DA

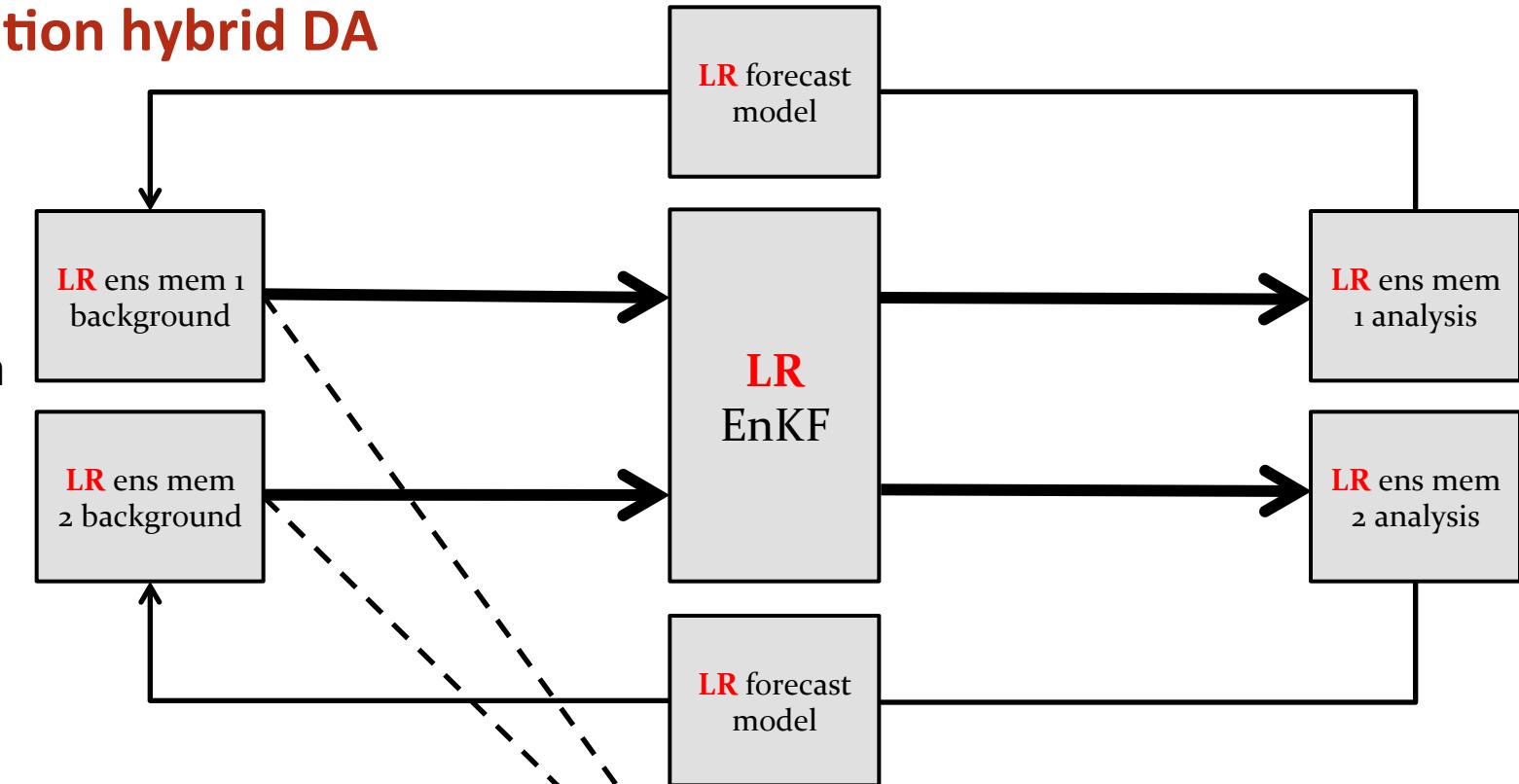
# Single-resolution hybrid DA

Both ensemble and deterministic “circuits” have identical resolutions

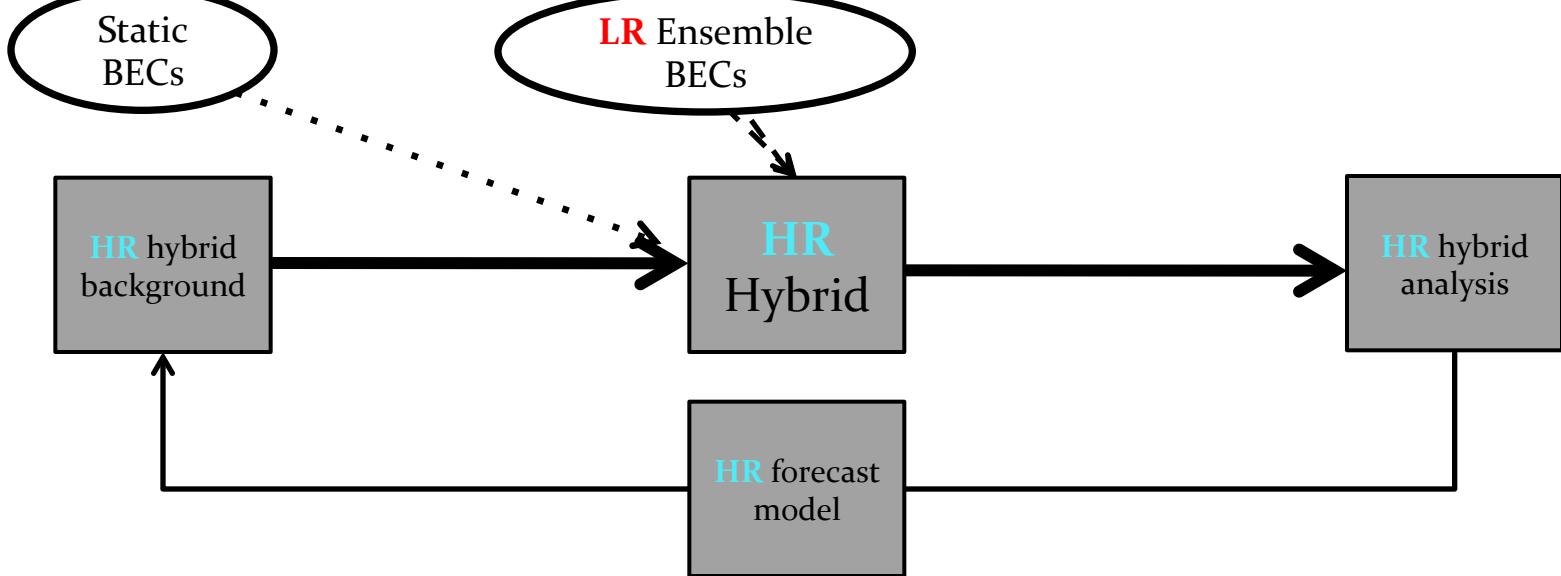


# Dual-resolution hybrid DA

Low-resolution  
ensemble  
“circuit”



High-  
resolution  
deterministic  
“circuit”



# Dual-resolution hybrid DA

- Practical aspect: Saving computational resources
- Scientific question: Does resolution of the ensemble perturbations matter?
- To examine these questions, several DA experiments were performed (Schwartz et al. 2015; MWR)

# Experiments assessing utility of DR hybrid

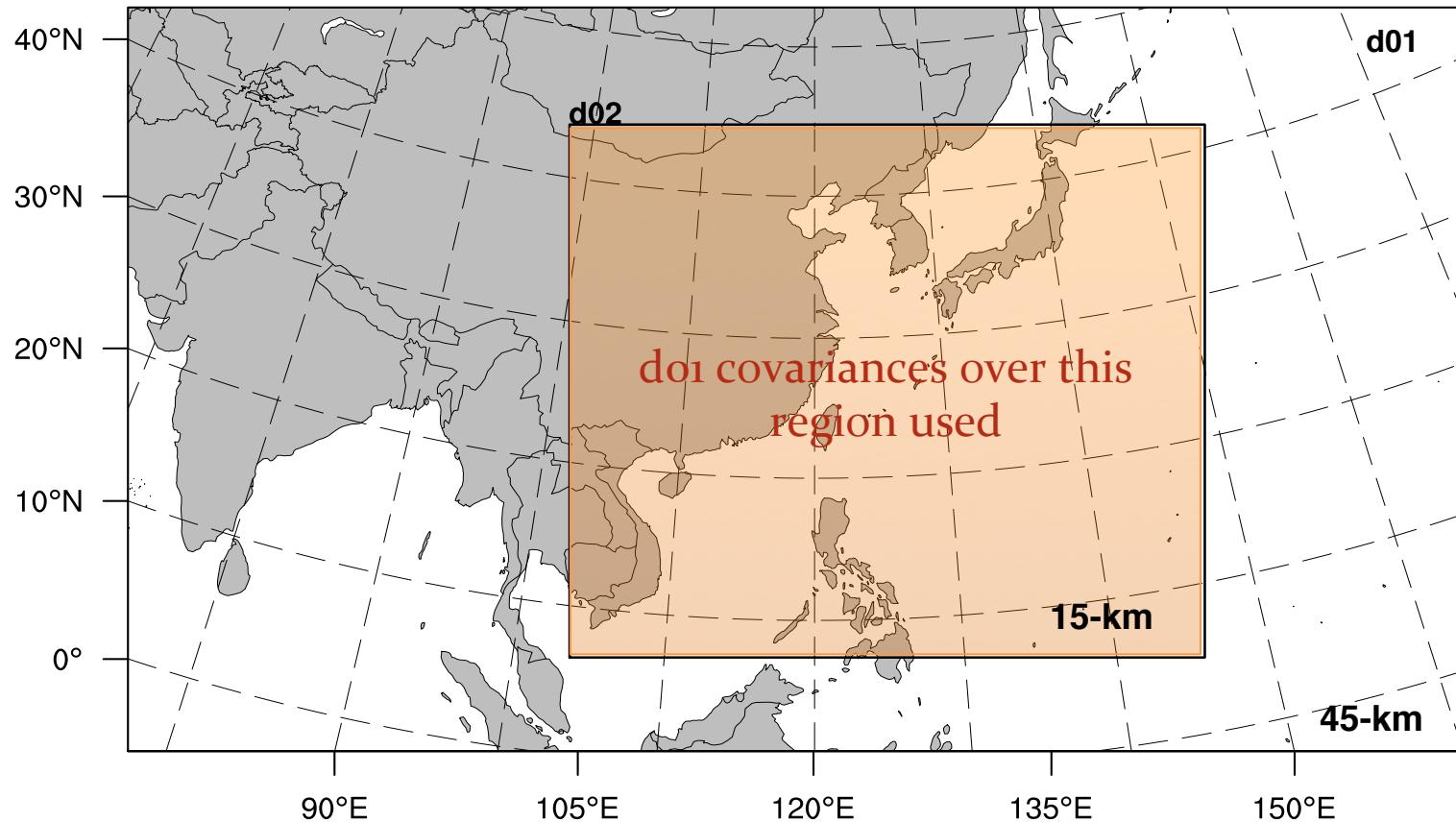
- Produced 45- and 15-km analyses, but **focus is on 15-km**
- Four continuously-cycling experiments:
  - 1) Single-resolution (SR) hybrid analyses at **15-km** grid length:  
a *15-km* ensemble was required
  - 2) Dual-resolution (DR) hybrid analyses at **15-km** grid length:  
a *45-km* ensemble was required
  - 3) Same as #2, but with EnKF re-centering
  - 4) Pure 3DVAR analyses at **15-km** grid length (control run)

# Some model configurations

- Version 3.3.1 of the WRF-ARW model
- 45 vertical levels, 30 hPa top
- Lateral boundary conditions from GFS forecasts
- Full suite of physical parameterizations

# Nested WRF model configuration

- For DR analyses, produce an analysis on d02 (15-km) using ensemble BECs from d01 (45-km)



# More settings

- An EnKF updated a 32-member WRF ensemble
  - Ran both 45- and 15-km cycling EnKF systems
  - 15-km ensembles provided input to 15-km *single-resolution* hybrid analyses
  - 45-km ensembles provided input to 15-km *dual-resolution* hybrid analyses
- Cyclic data assimilation (6-hr period) from 4-28 September 2008
- 72-hr nested WRF model forecasts initialized from hybrid and 3DVAR analyses beginning 1800 UTC 8 September—*focus on the 15-km forecasts*

# EnKF assimilation parameters

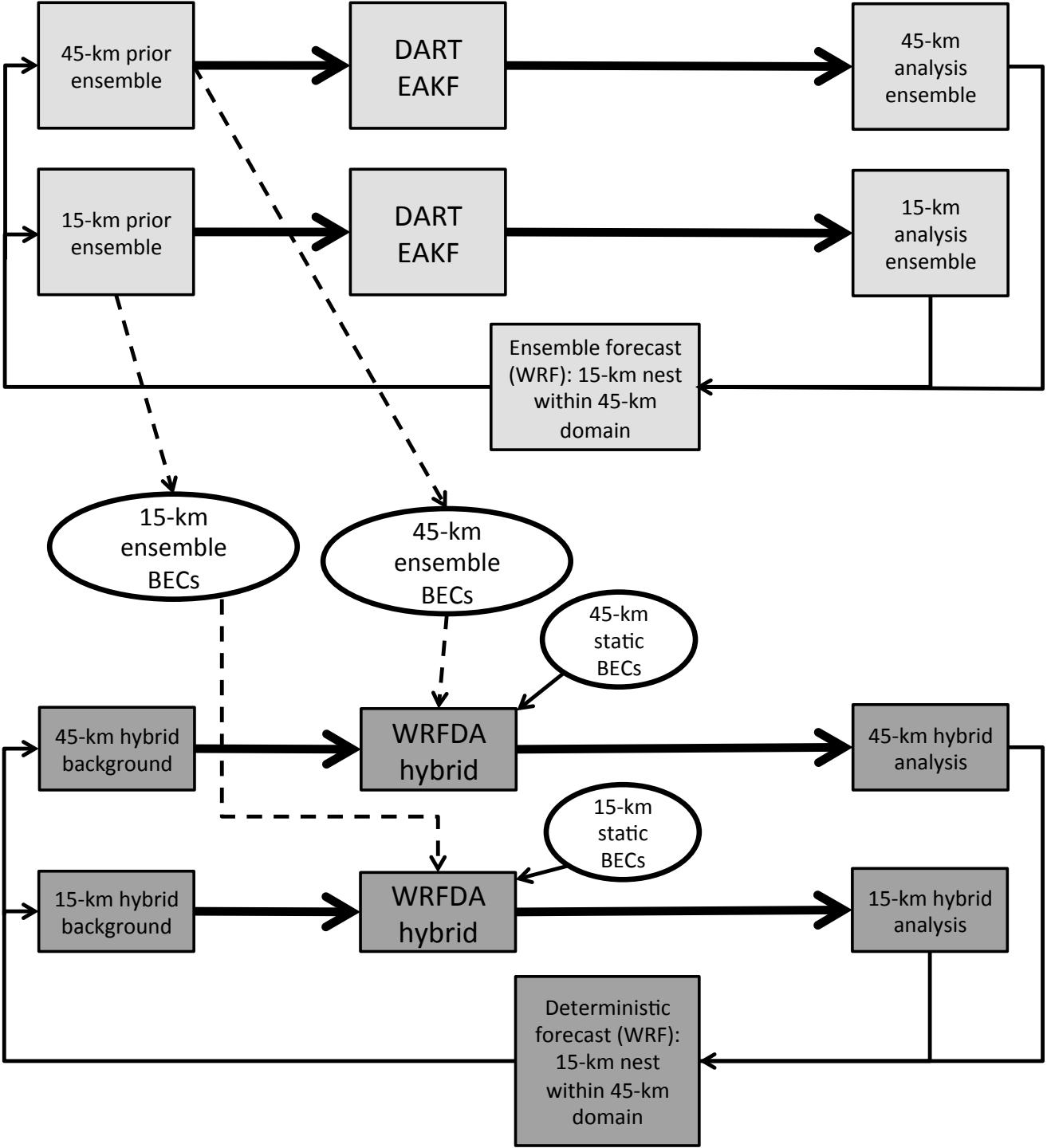
- The Data Assimilation Research Testbed (DART)
  - Ensemble Adjustment Kalman Filter (EAKF)
  - 32 ensemble members
  - Horizontal localization cutoff: ~1280-km from observations
  - Vertical localization cutoff: ~10-km from observations
  - Prior adaptive inflation
- Stochastic kinetic-energy backscatter scheme within WRF was used during model advances
  - Helped with ensemble spread
- Perturbed lateral boundary conditions for ensemble of model advances

# Hybrid assimilation parameters

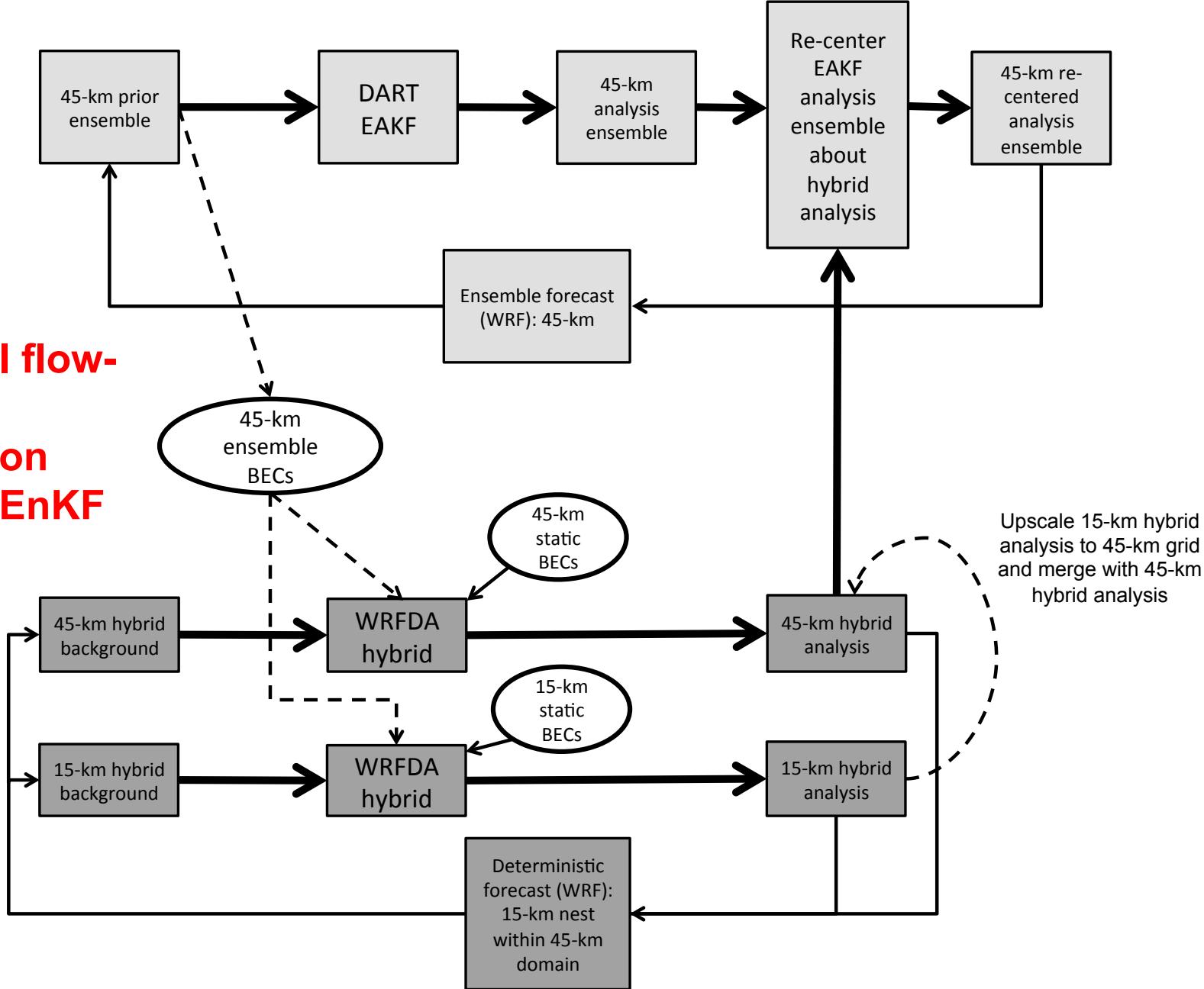
- Used the WRFDA hybrid system
- 75% ensemble, 25% climatological contributions to total BECs
- Horizontal localization: Similar as EnKF
- Vertical localization: Length-scale increased with height
- Same inflation as in DART applied to the prior perturbations before they were ingested into the hybrid



## Experimental flow-chart for the single-resolution system

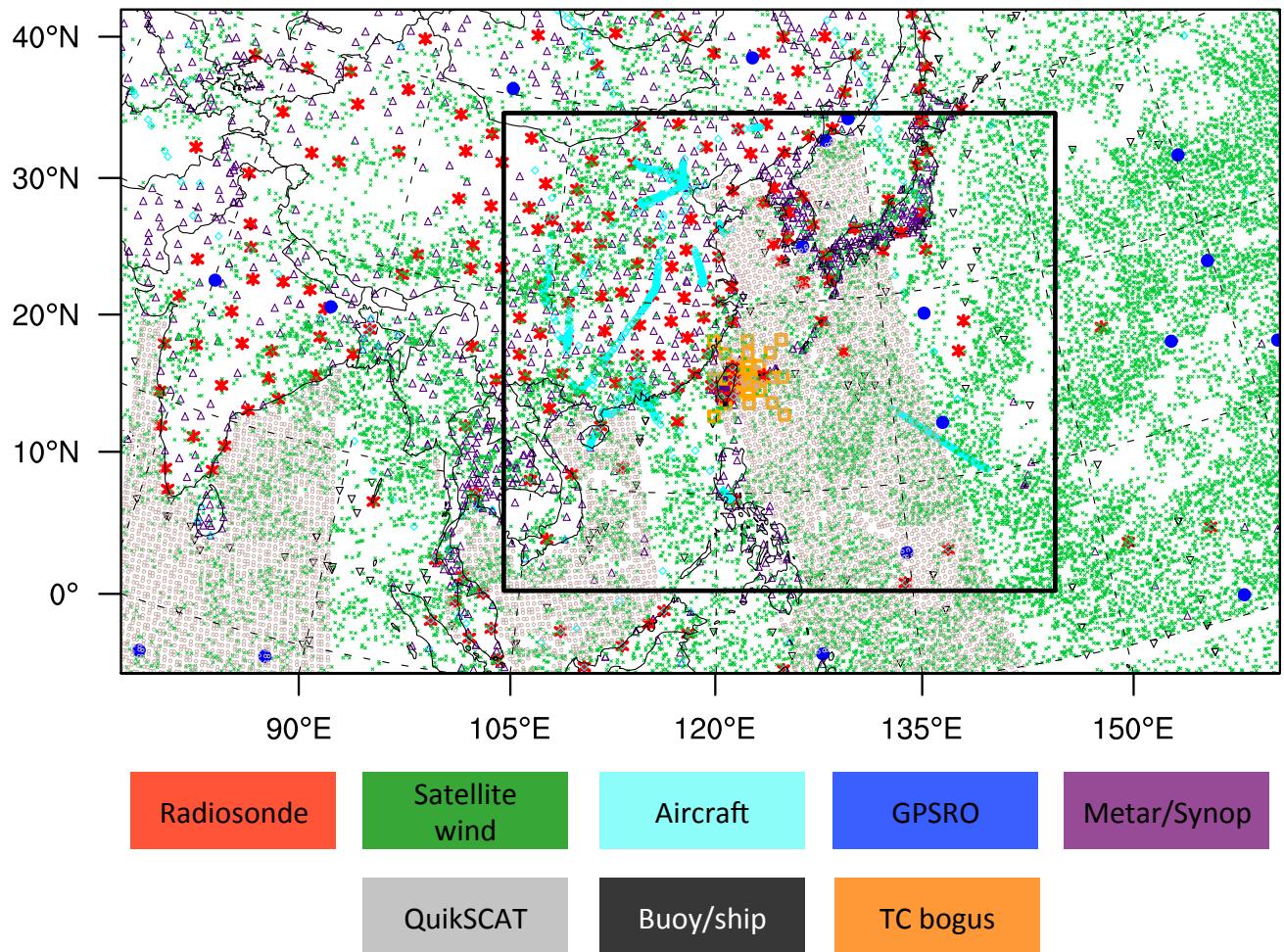


## Experimental flow-chart for the dual-resolution system with EnKF re-centering



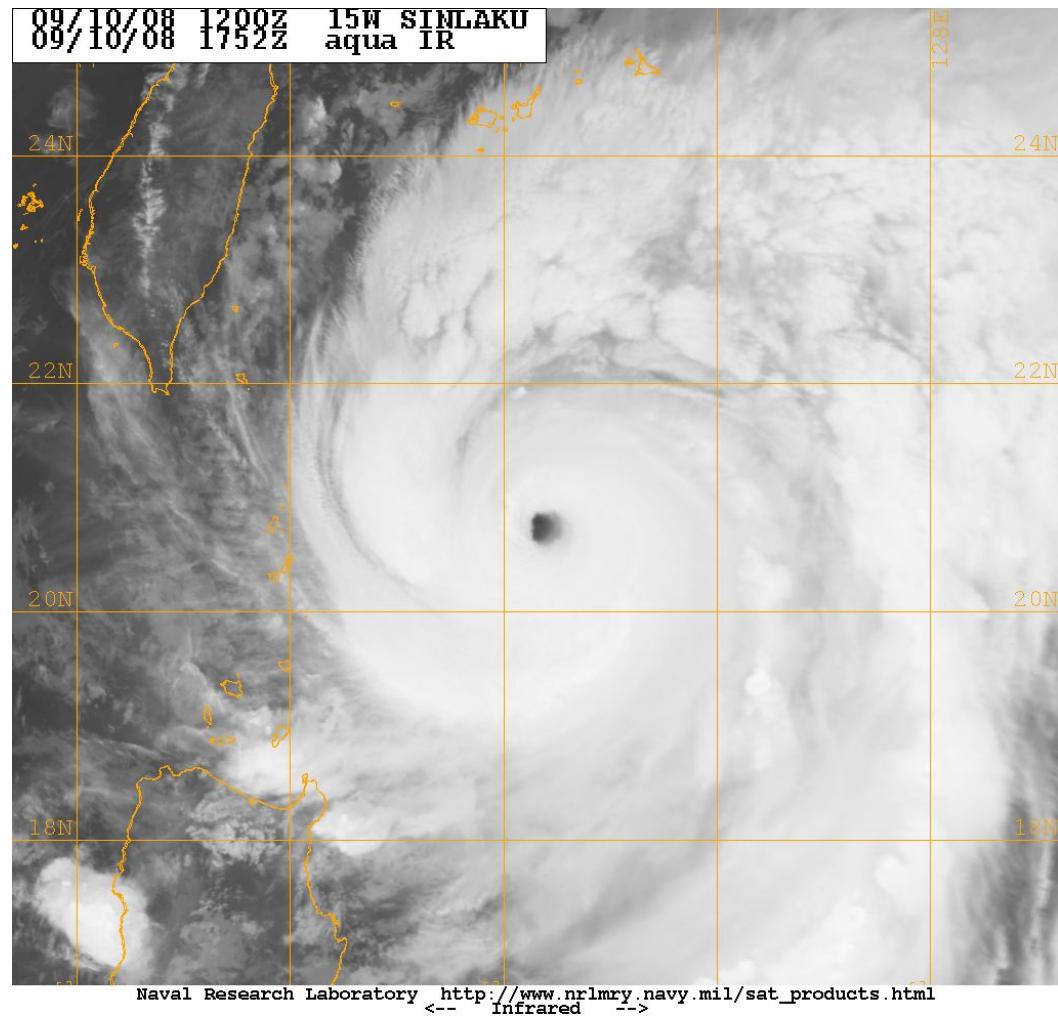
# Observation coverage

- Snapshot of available observations



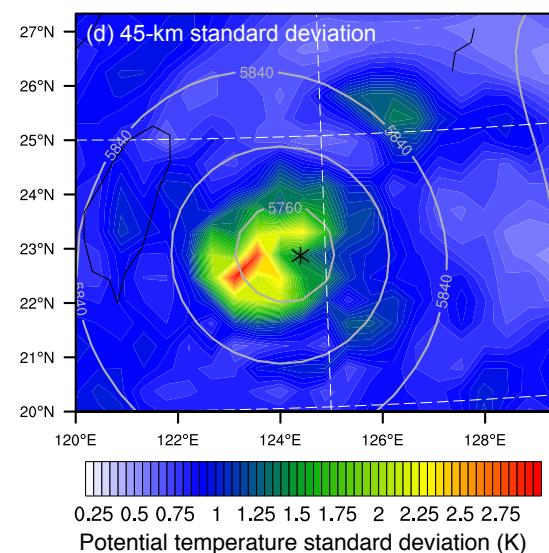
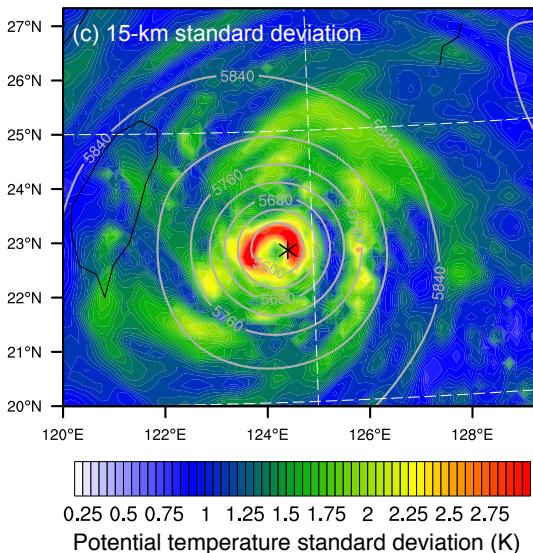
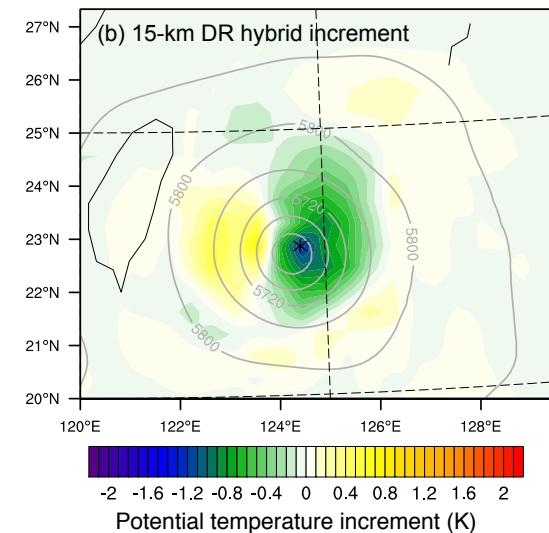
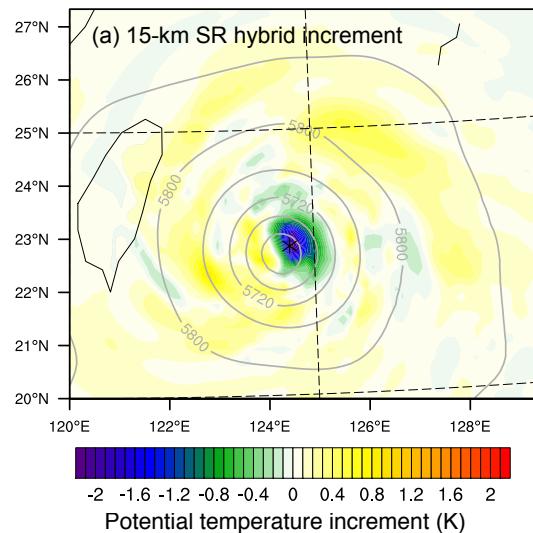
# Single observation experiment

- Temperature observation (innovation -2K, ob. error 1K) placed in center of typhoon Sinlaku



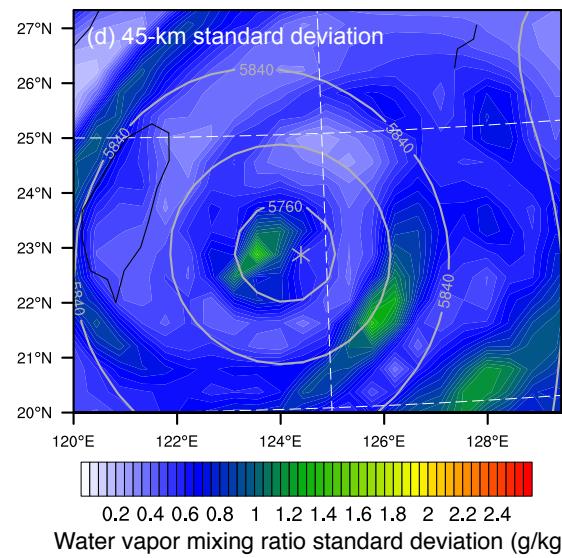
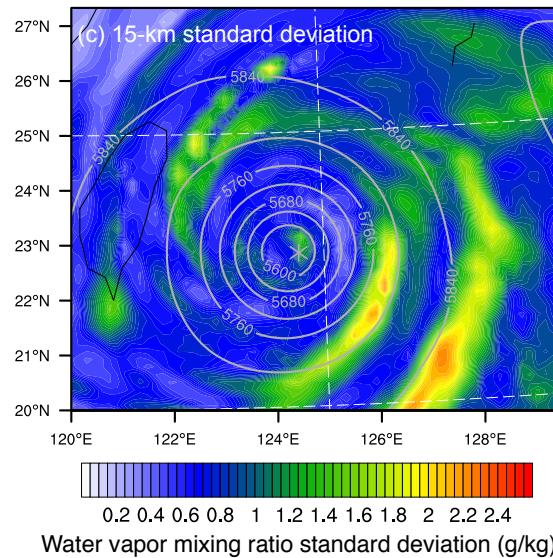
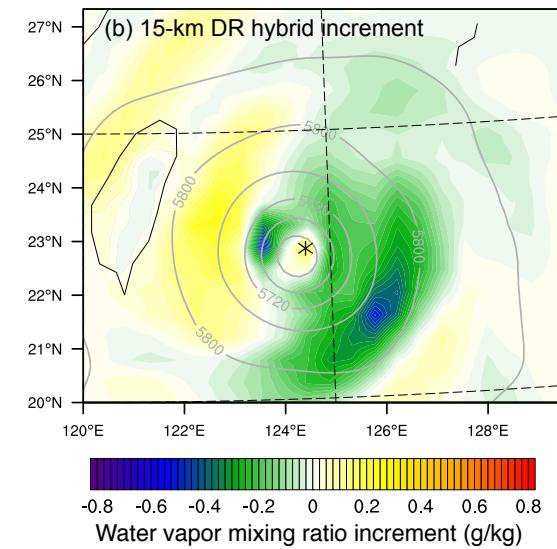
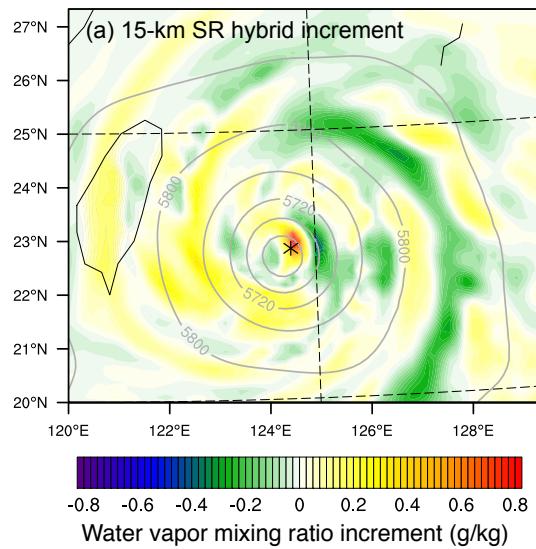
# Single observation experiment

- Temperature observation (innovation -2K, ob. error 1K) placed in center of typhoon
- **Potential temperature** increments and spread



# Single observation experiment

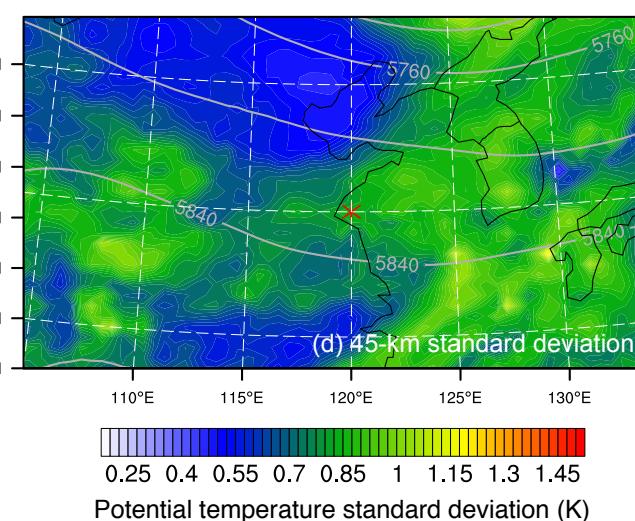
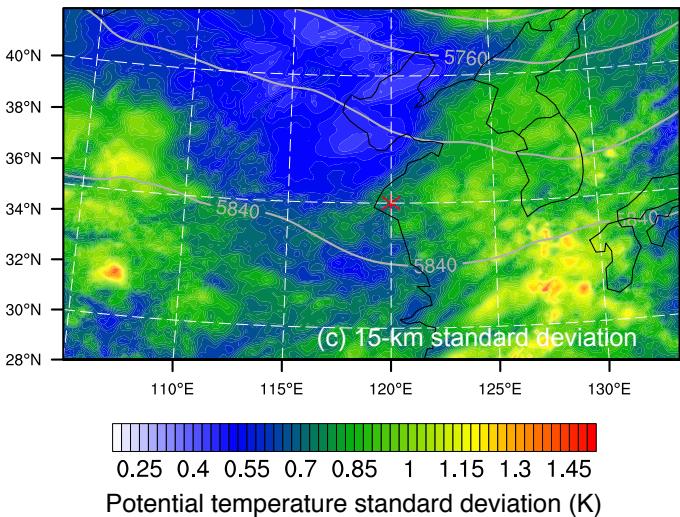
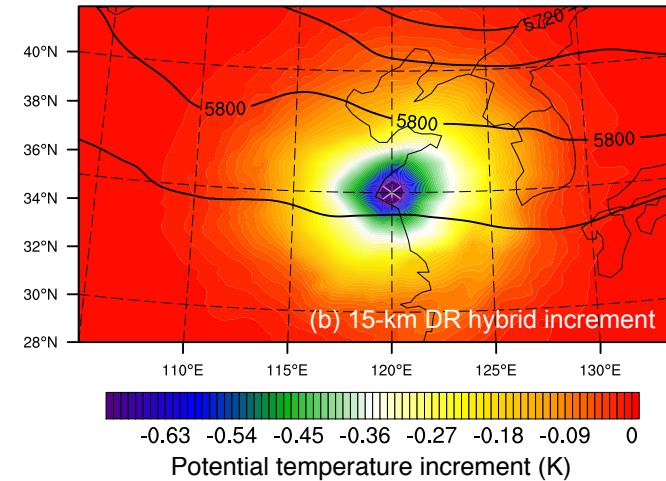
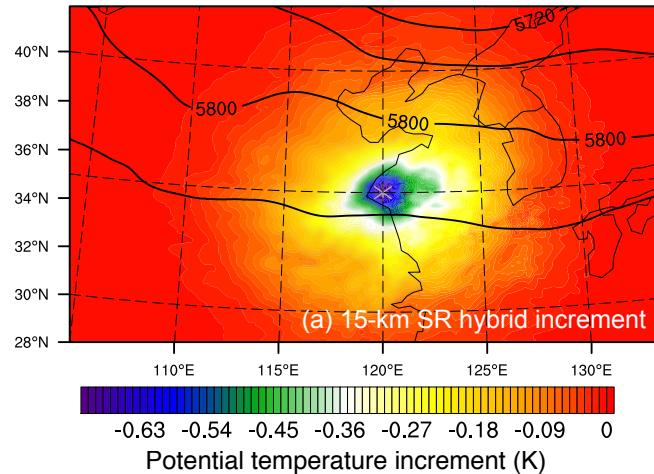
- Temperature observation (innovation -2K, ob. error 1K) placed in center of typhoon
- **Mixing ratio** increments and spread



# Single observation experiment

- Temperature observation (innovation -2K, ob. error 1K) placed in westerly flow

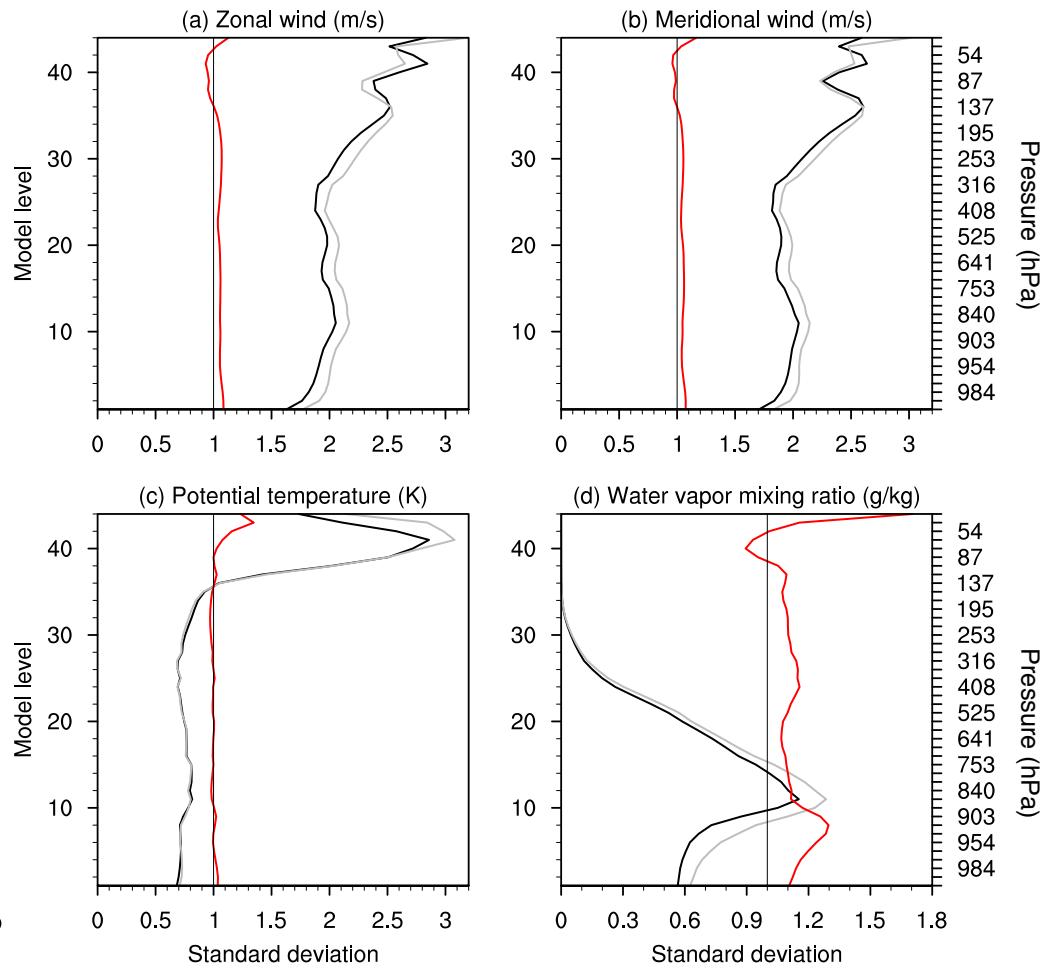
- Potential temperature increments and spread



# Ensemble spreads

- Average prior ensemble spreads (over 15-km domain)
- Averaged between 1800 UTC 8 September and 0000 UTC 28 September
- After prior inflation

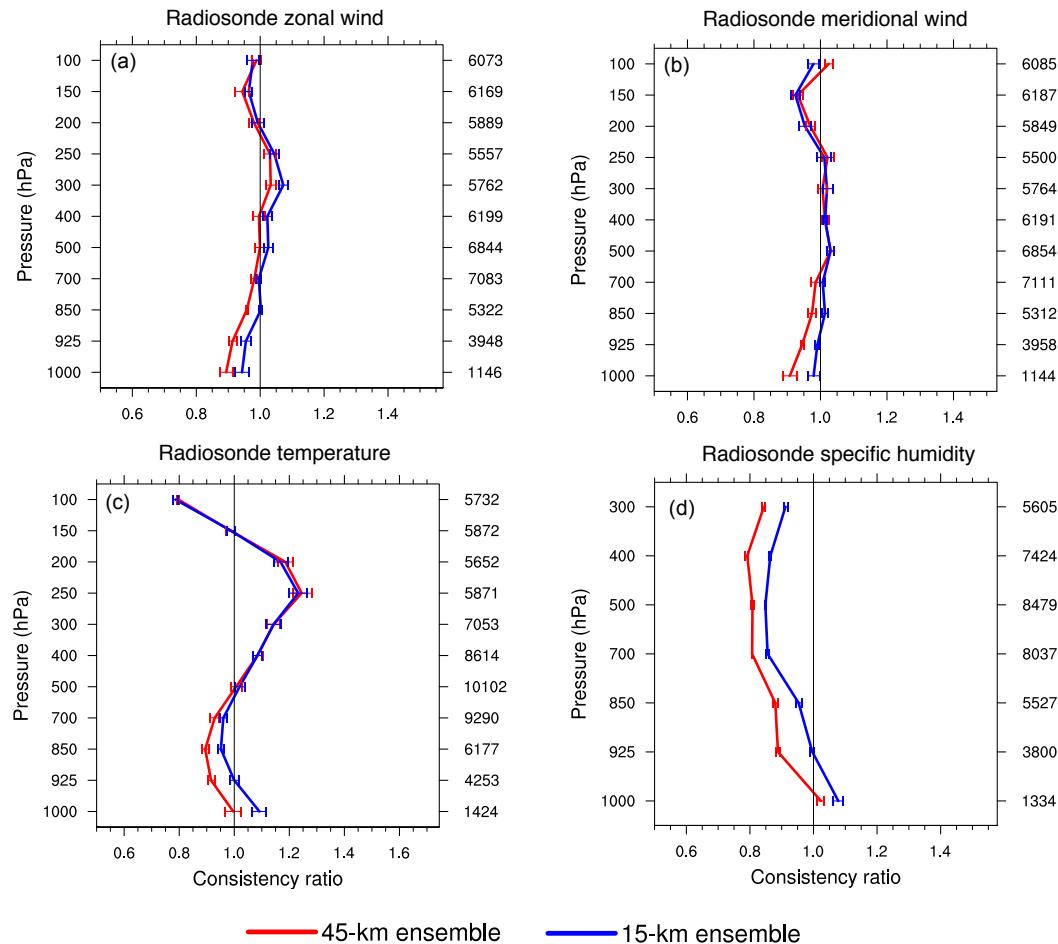
— 15-km ensemble  
— 45-km ensemble  
— Ratio of 15-km ensemble standard deviation to  
45-km ensemble standard deviation



# Ensemble spread/skill

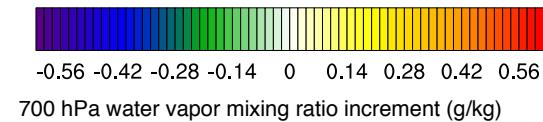
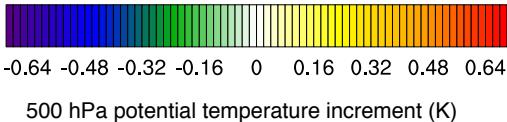
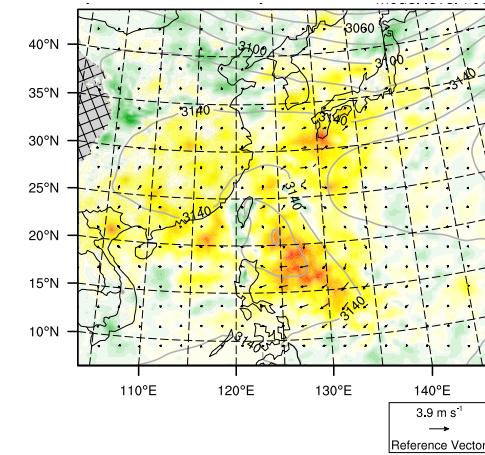
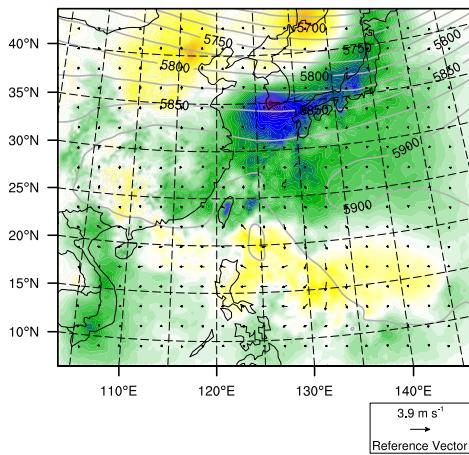
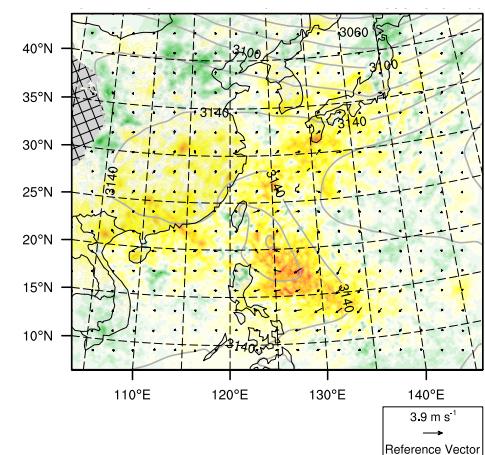
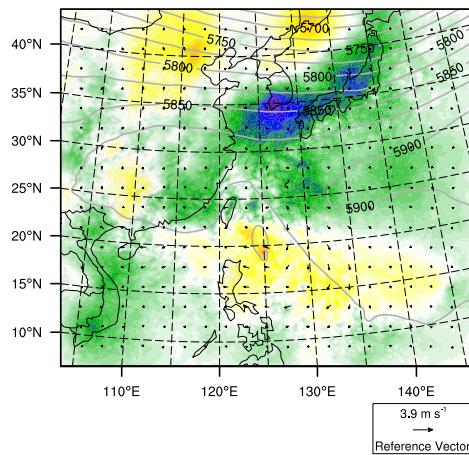
- How did the EAKF perform?
- Consistency ratios (CRs) averaged between 1800 UTC 8 September and 0000 UTC 28 September
- After prior inflation

$$CR = \frac{\text{Total spread}}{RMSE}$$



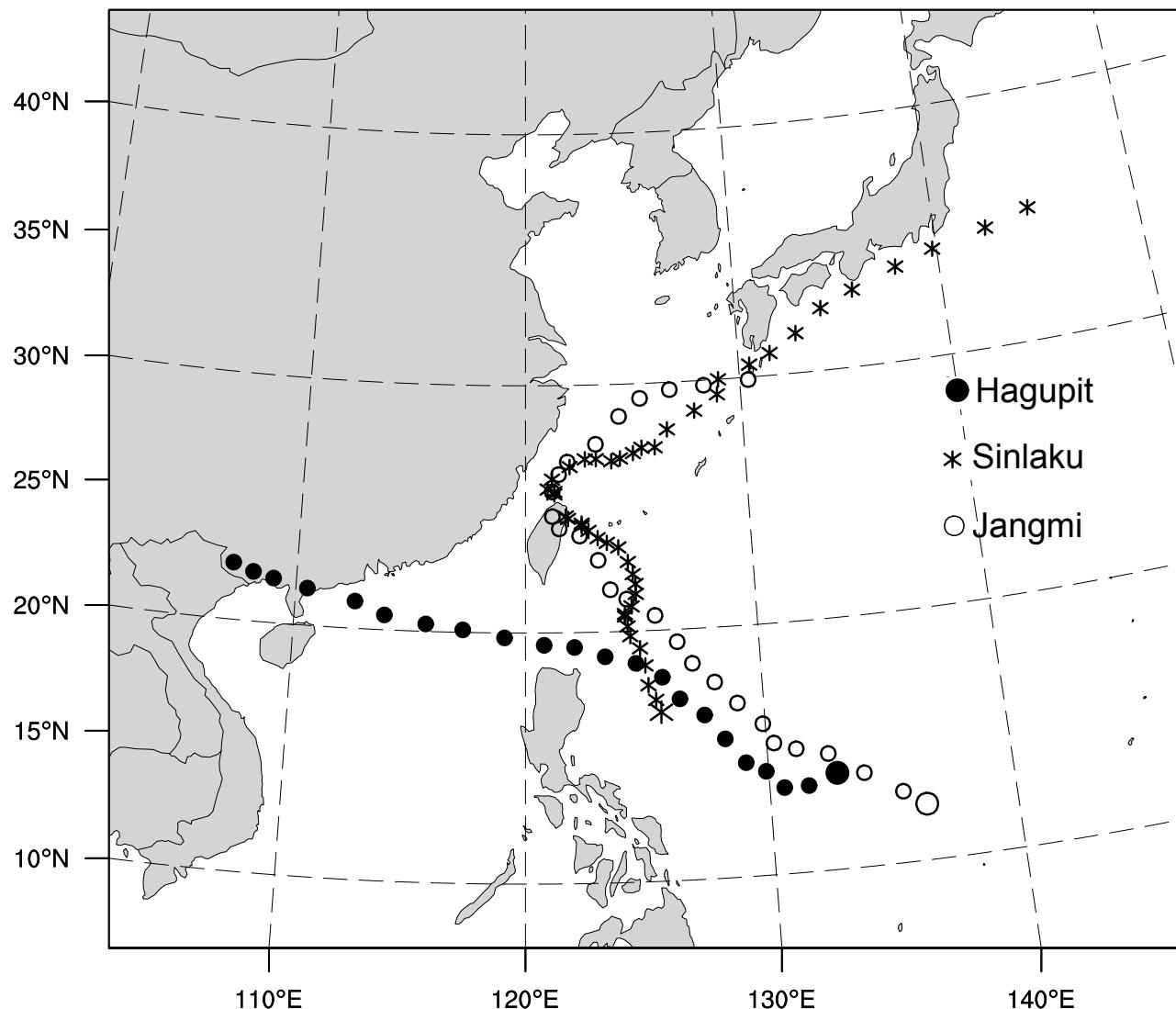
# Mean analysis increments

- Increments averaged between 1800 UTC 8 and 0000 UTC 28 September
- Which row is DR and which one is SR?
  - Can you tell the difference?



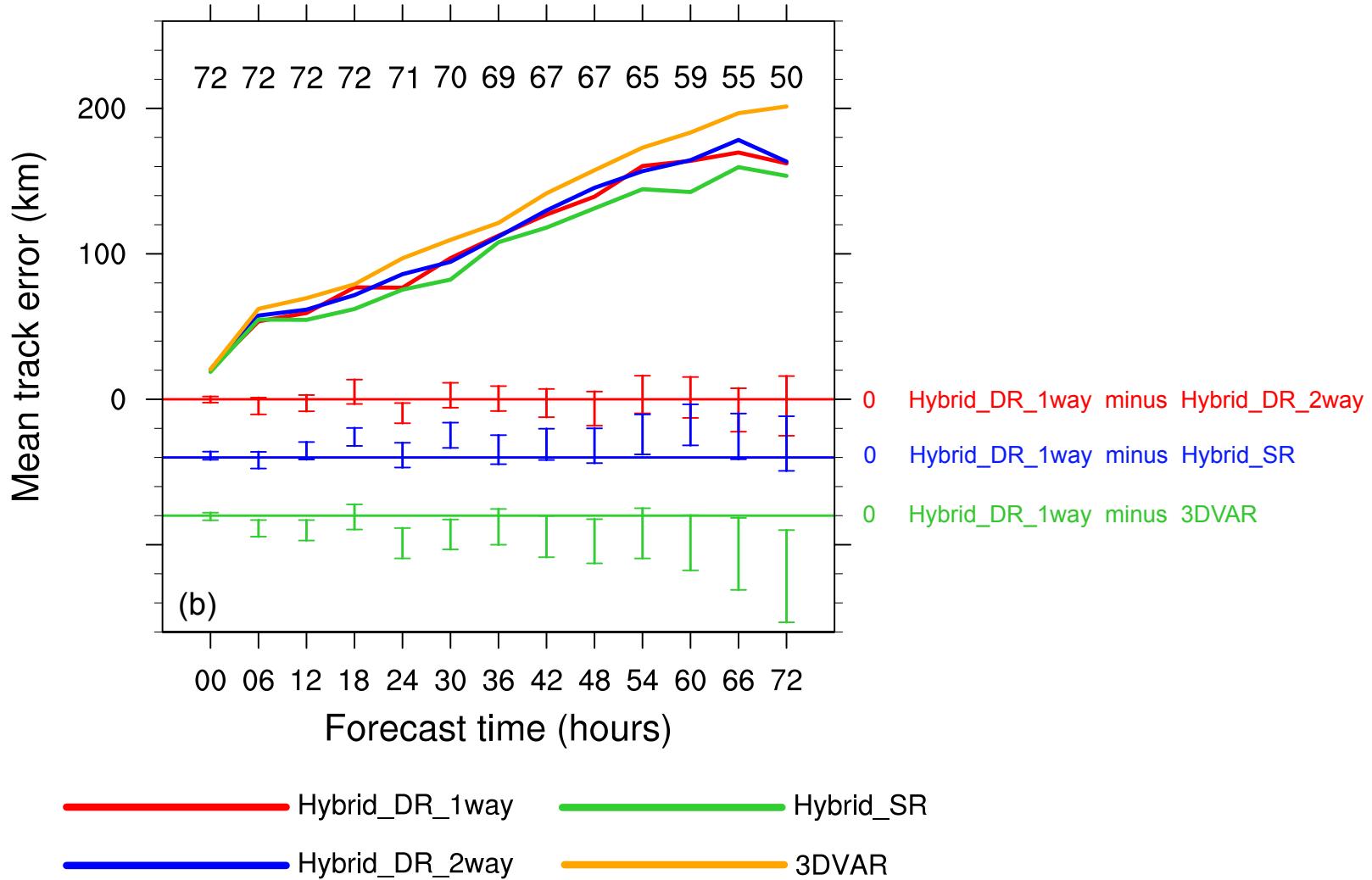
# Typhoon track forecasts

- Verified 15-km forecasts against tracks of 3 typhoons



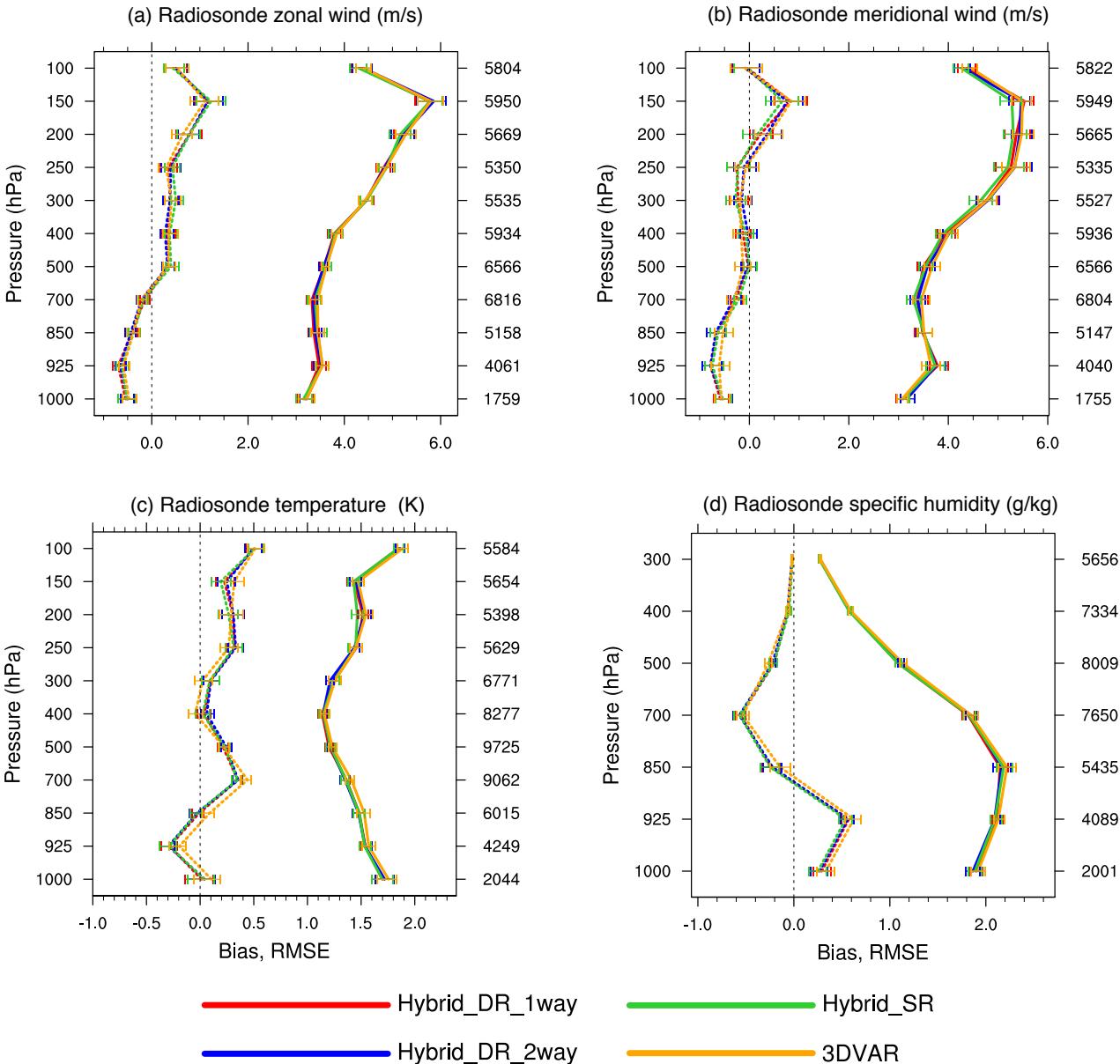
# Typhoon track forecasts

- Verified 15-km forecasts against tracks of 3 typhoons



# Verification against radiosondes

- 24-hr forecasts
- Aggregated over 78 forecasts initialized between 1800 UTC 8 and 0000 UTC 28 September



# Toward higher-resolution analyses

# Background

- Convection-allowing forecasts have traditionally been initialized from convection-parameterizing analyses
  - These forecasts have been good, but there are spin-up issues
- This work produces **convection-allowing *analyses*** over a large (e.g.,  $3000 \text{ km}^2$ ) domain
  - Many case studies have examined convection-allowing data assimilation over small domains

# The challenge

- Ultimately, we want to produce analyses at convection-permitting resolution using flow-dependent BECs derived from ensembles
- But, for DA purposes, computational constraints currently do not permit high-resolution *ensembles* over domains large enough to resolve synoptic scale features
  - This will change fairly soon

# Practical approach

- However, we can combine high-resolution *deterministic* backgrounds with low-resolution ensembles in a DR hybrid approach
  - Does not solve the issue of high-resolution ensembles
  - Permits high-resolution within ensemble DA systems
- Not clear how mixing a convection-allowing background with a convection-parameterizing ensemble will work

# DA experiments

- Full-cycling (6-hr period) between May 4 – June 30, 2013
- Five DA experiments (analyses every 6-hrs):
  - Pure EnKF (“EnKF”) : 20-km cycling
  - Pure 3DVAR (“3DVAR 20-km”) : 20-km cycling
  - Pure 3DVAR (“3DVAR 4-km”) : 4-km cycling
  - SR Hybrid (“Hybrid 20-km”) : 20-km cycling
  - DR Hybrid (“Hybrid 4-km”) : 4-km cycling
- Hybrid runs coupled to a 20-km, 50-member EnKF
- All assimilated identical conventional observations

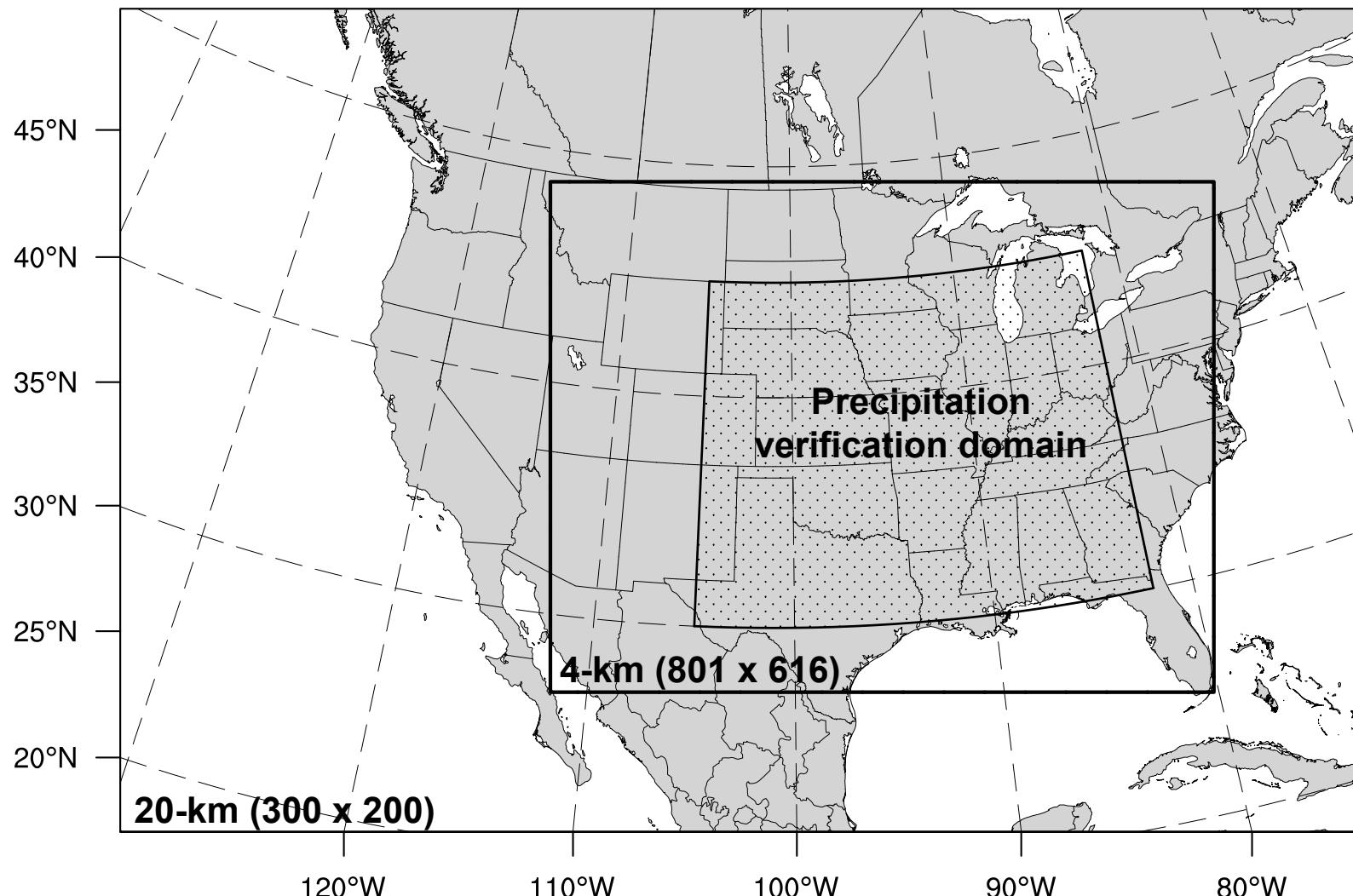
# Selected data assimilation settings

- 50 ensemble members in hybrid/EnKF
- Hybrid: 75% of background errors from ensemble, 25% from the static contribution
- Used posterior inflation for EnKF and localization in EnKF and hybrid

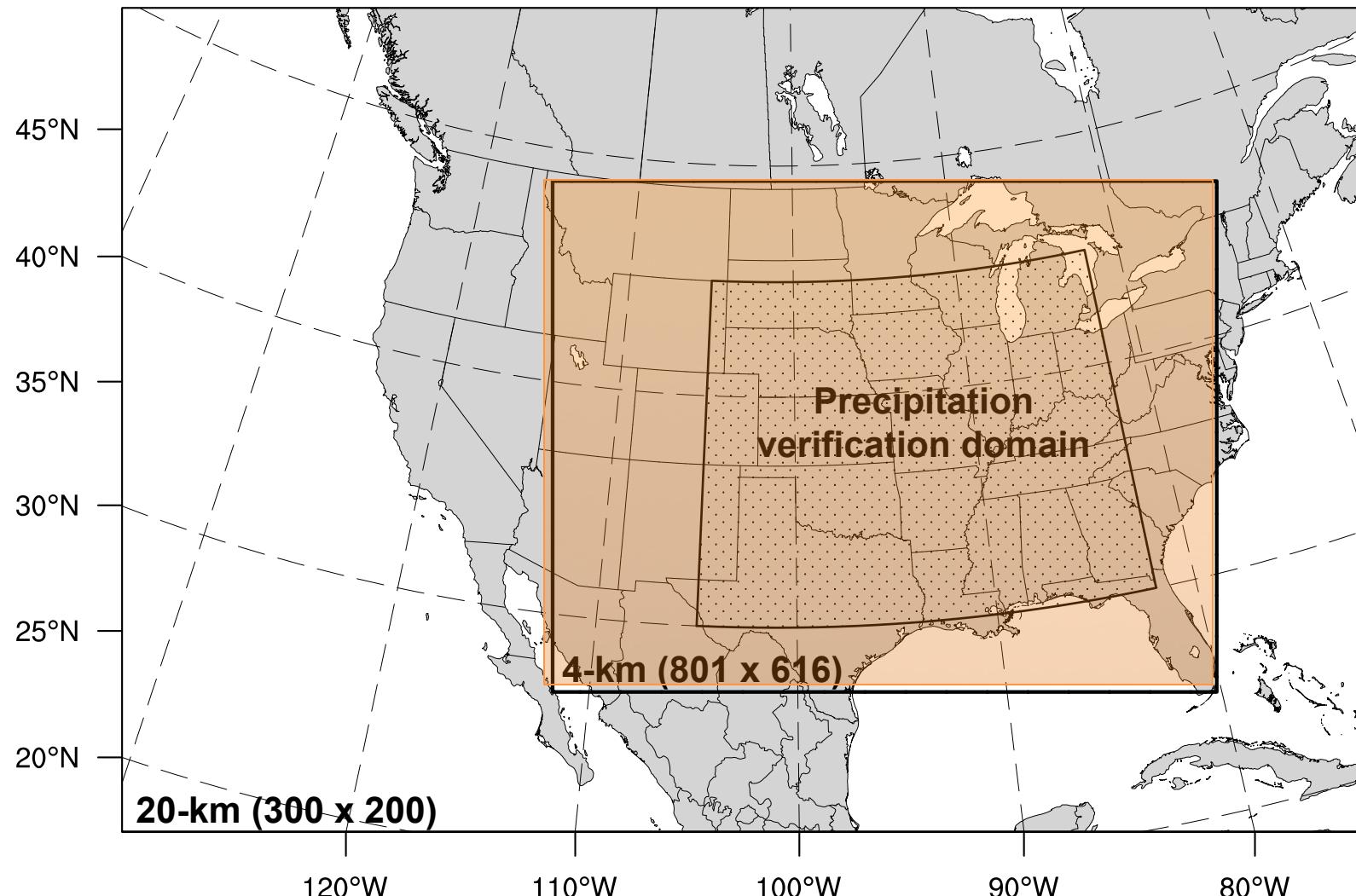
A very ill-behaved chihuahua/dachshund hybrid



# Computational domain

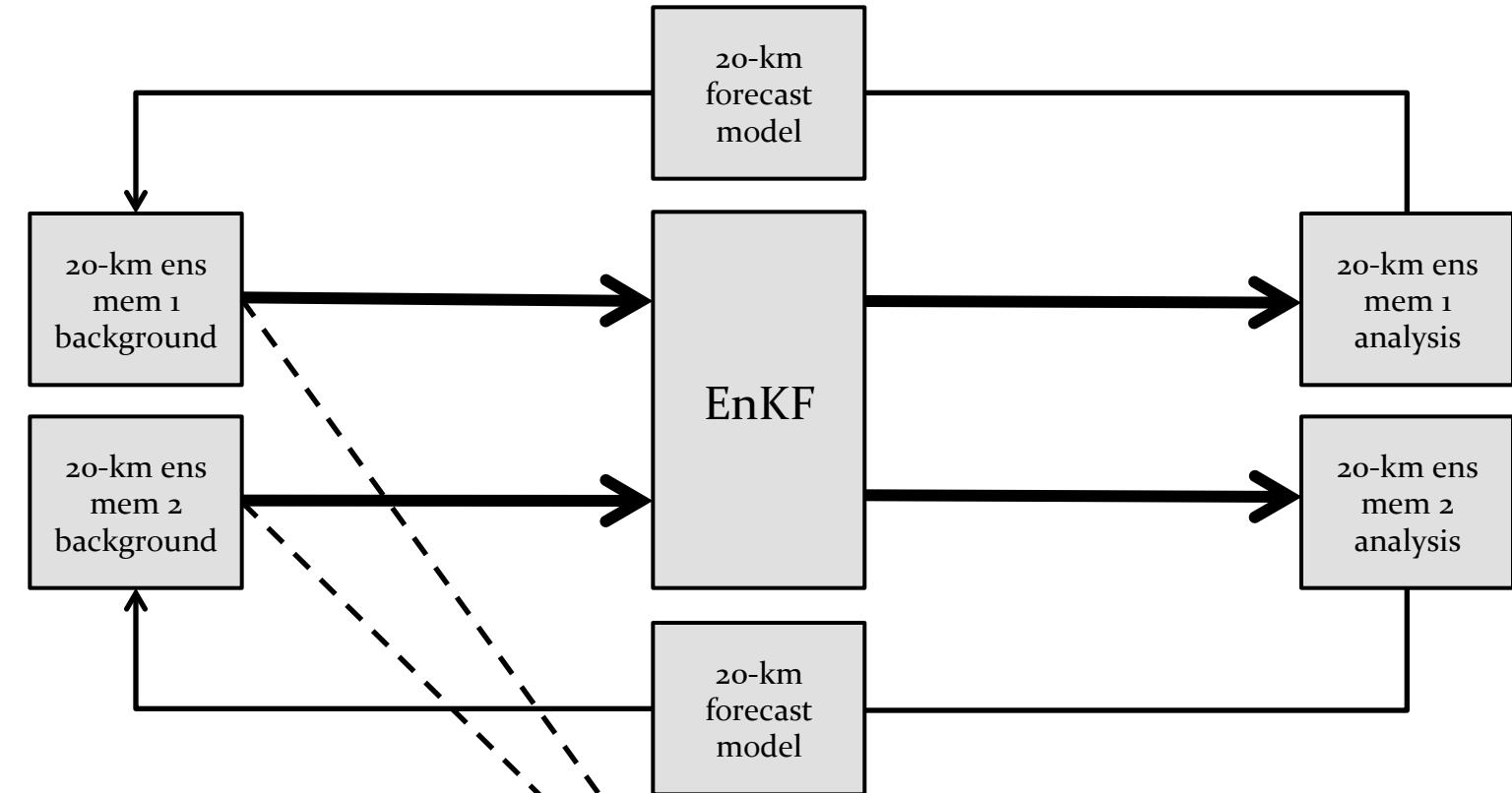


# Computational domain

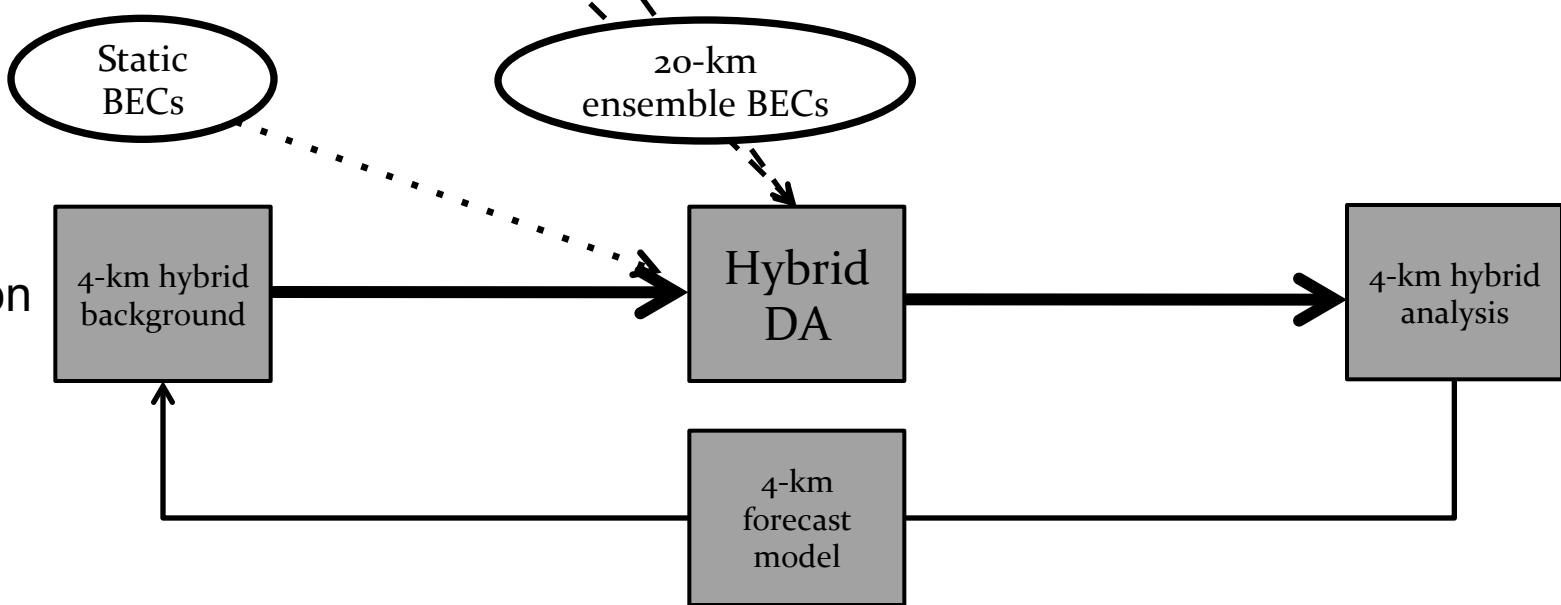


20-km ensemble covariances over orange region used for 4-km DR analyses

Low-resolution  
(20-km)



High-resolution  
(4-km)



# Forecast initialization

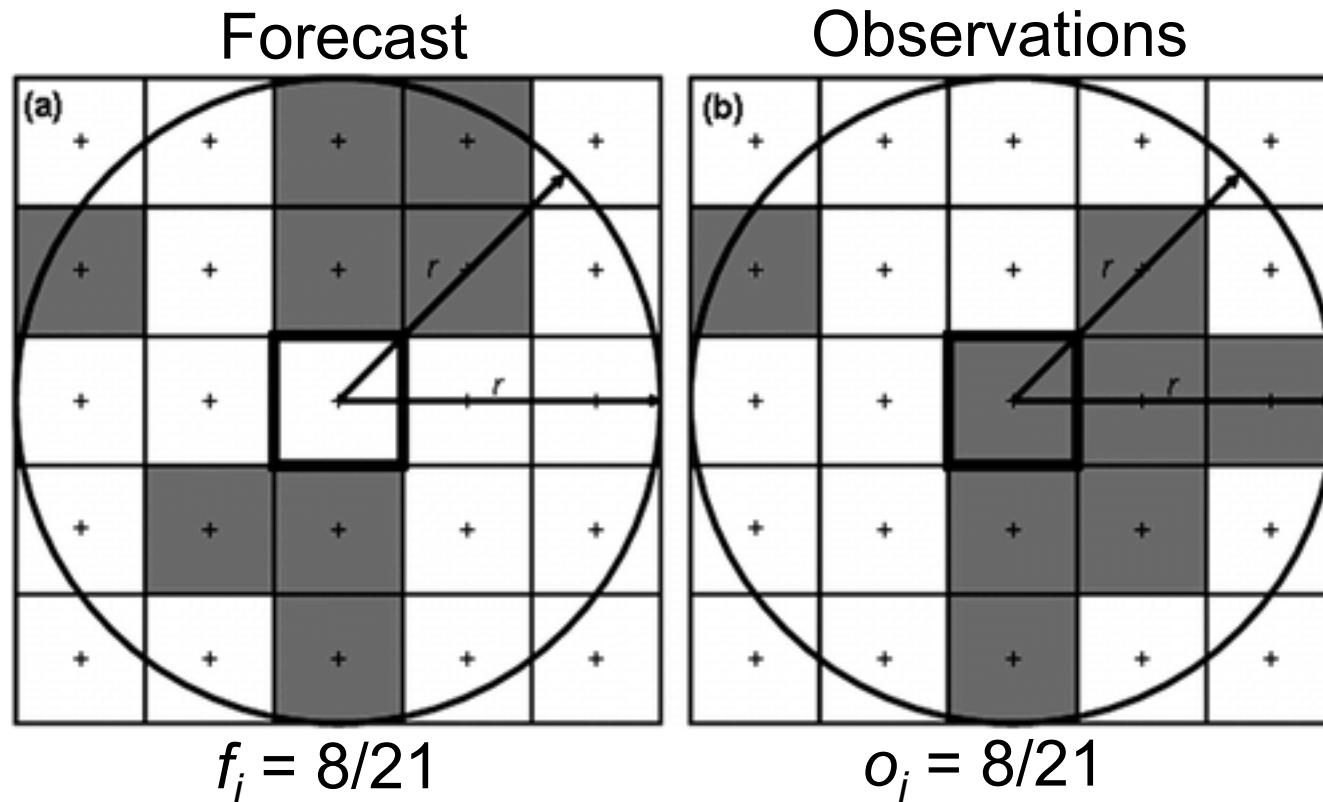
- 0000 UTC analyses initialized 36-hr 4-km WRF forecasts
- 4-km initial conditions were *downscaled 20-km analyses* in the 20-km 3DVAR, EnKF, and SR hybrid experiments
- *True 4-km analyses* initialized 4-km forecasts in the DR hybrid and 4-km 3DVAR experiments
- Forecast differences between SR and DR hybrid experiments due to analysis resolution
- Control: Interpolate 0000 UTC GFS analyses directly onto the domain and run forecasts

# Precipitation verification

- Focus on 4-km hourly precipitation forecasts
- NCEP Stage IV observations as “truth”
- All precipitation statistics aggregated over **55 4-km forecasts**
- Fractions skill score (FSS) quantifies displacement errors

# Fractions skill score (FSS)

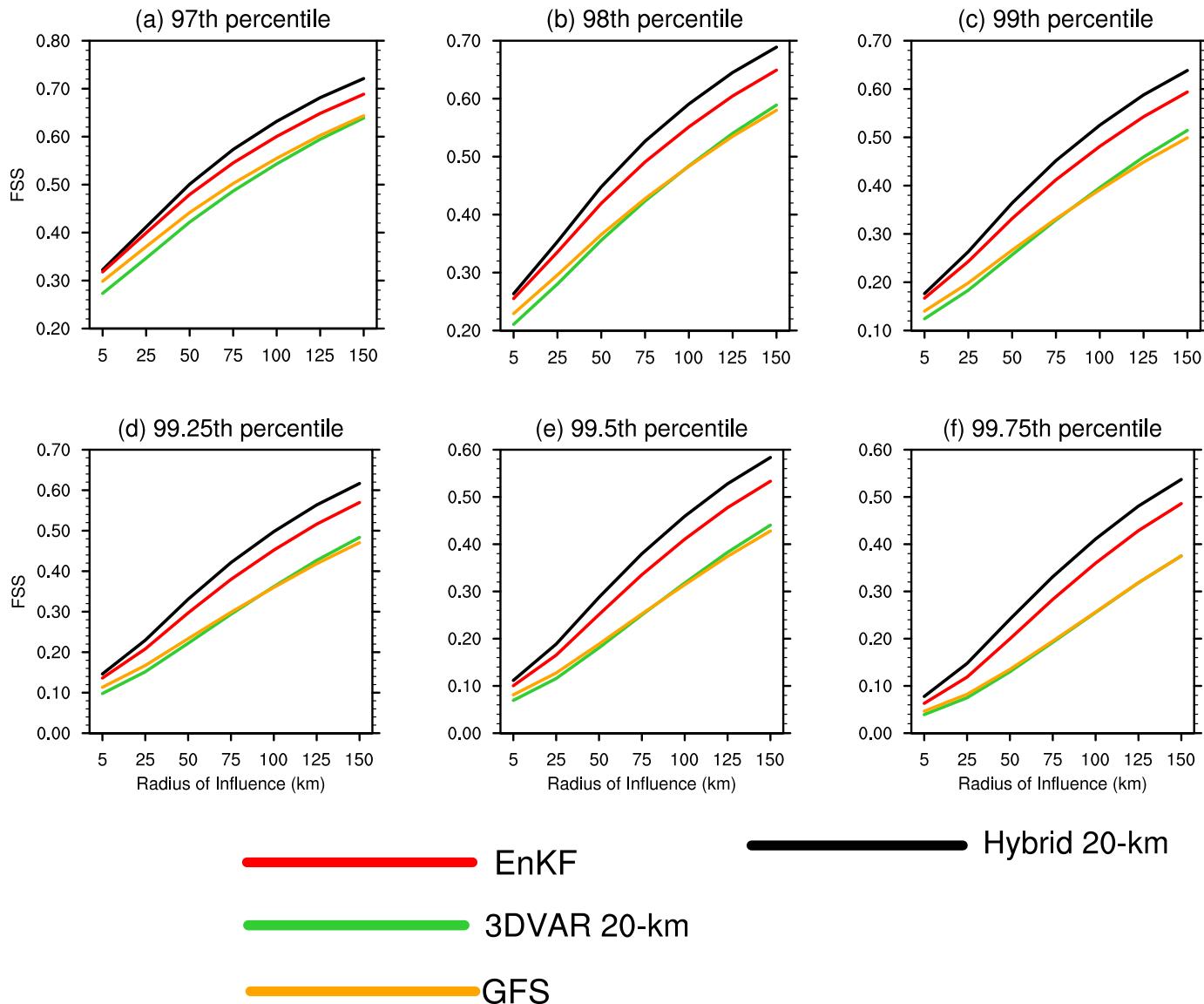
- A neighborhood approach to verification, which is needed to verify high-resolution forecasts



- Directly compare observed and forecast fractions to compute the FSS

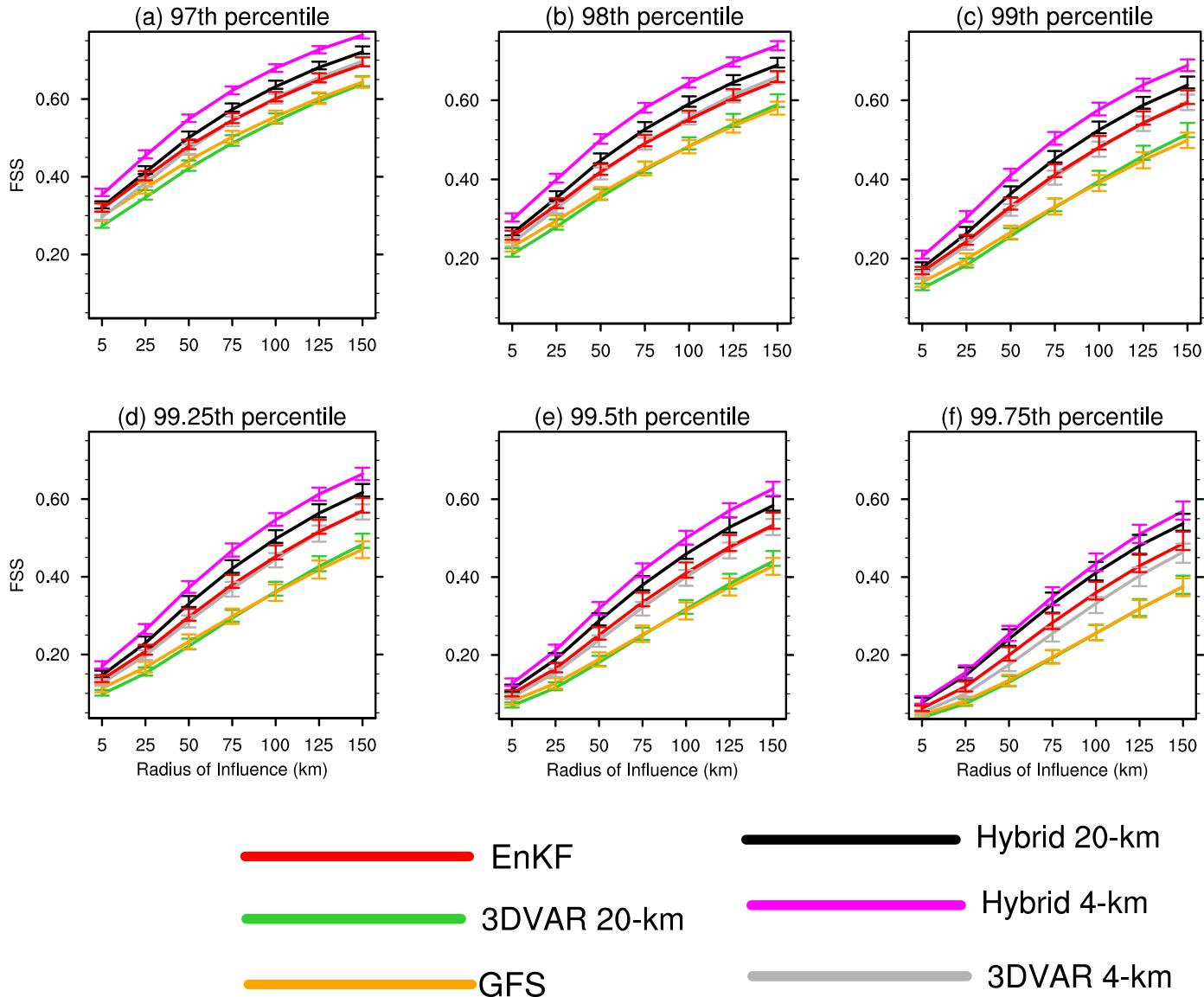
# Precipitation verification: the first 12-hrs

- Fractions skill score (FSS) aggregated over the **first 12 forecast hours** and 55 4-km forecasts



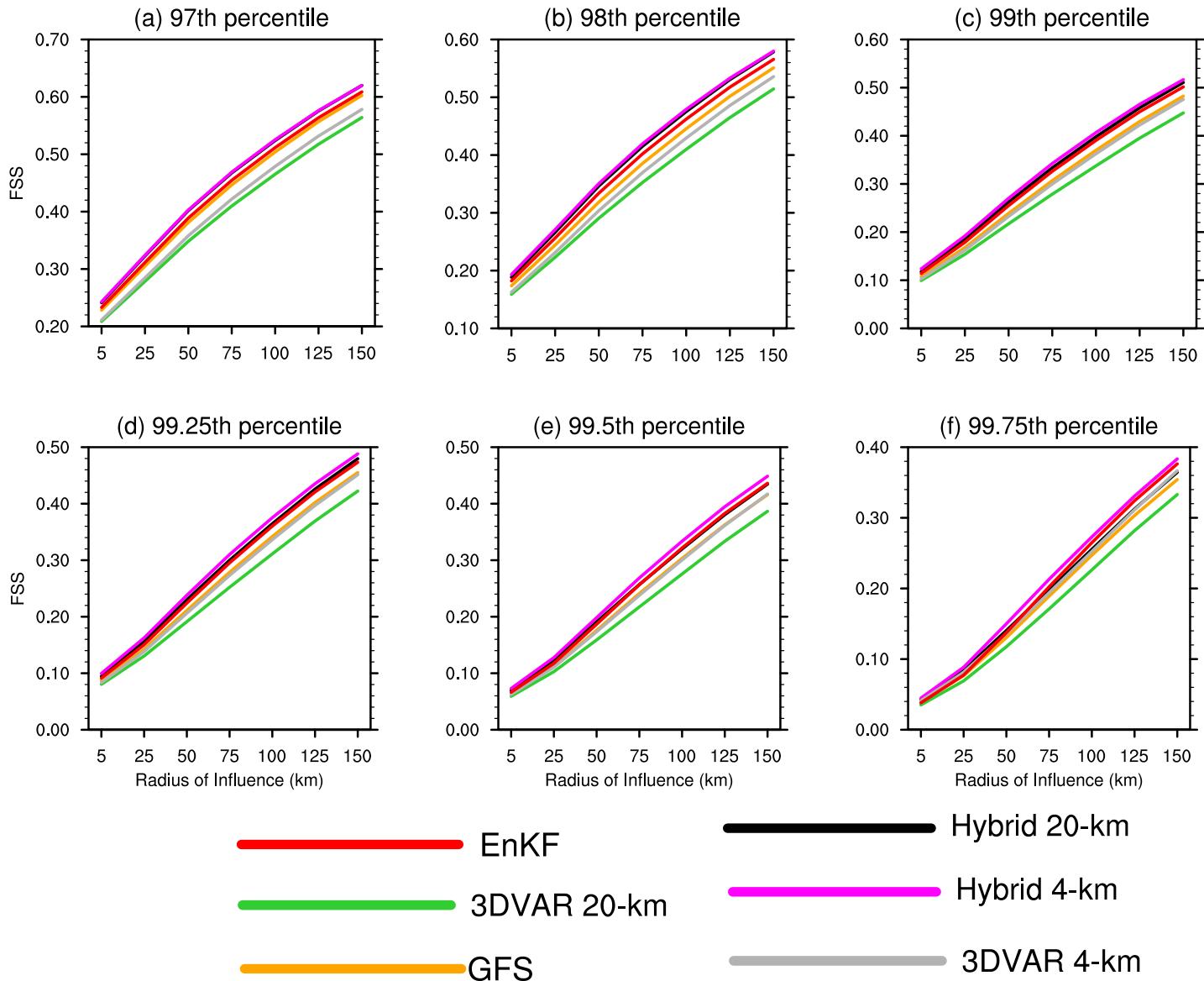
# Precipitation verification: the first 12-hrs

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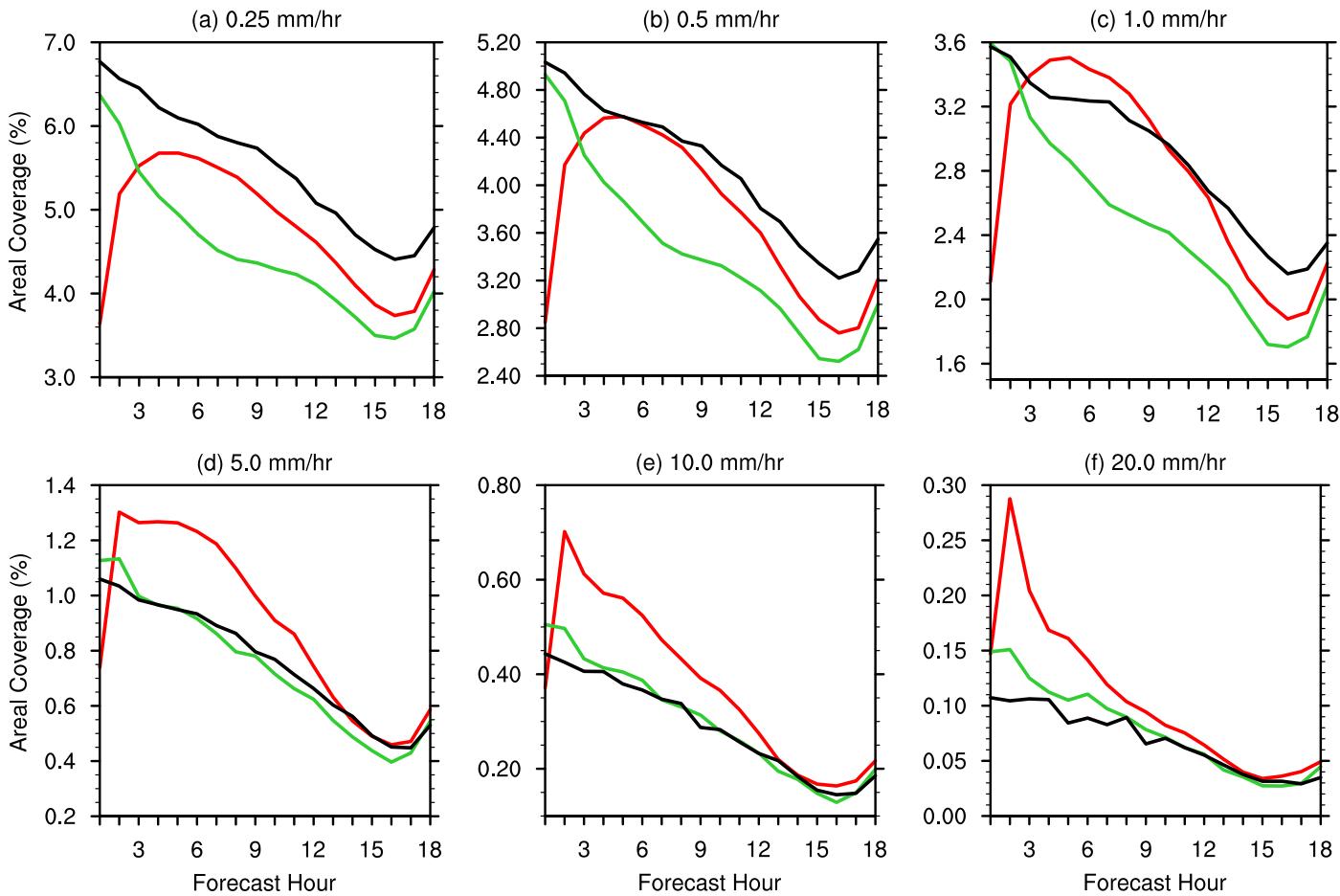
# Precipitation verification: 18-36-hrs

- Fractions skill score (FSS) aggregated over **forecast hours 18-36** and 55 4-km forecasts



# Areal coverages of precipitation

- Aggregate fractional coverage of precipitation exceeding certain thresholds aggregated over all 55 forecasts



— Hybrid 4-km  
— Hybrid 20-km  
— Stage IV

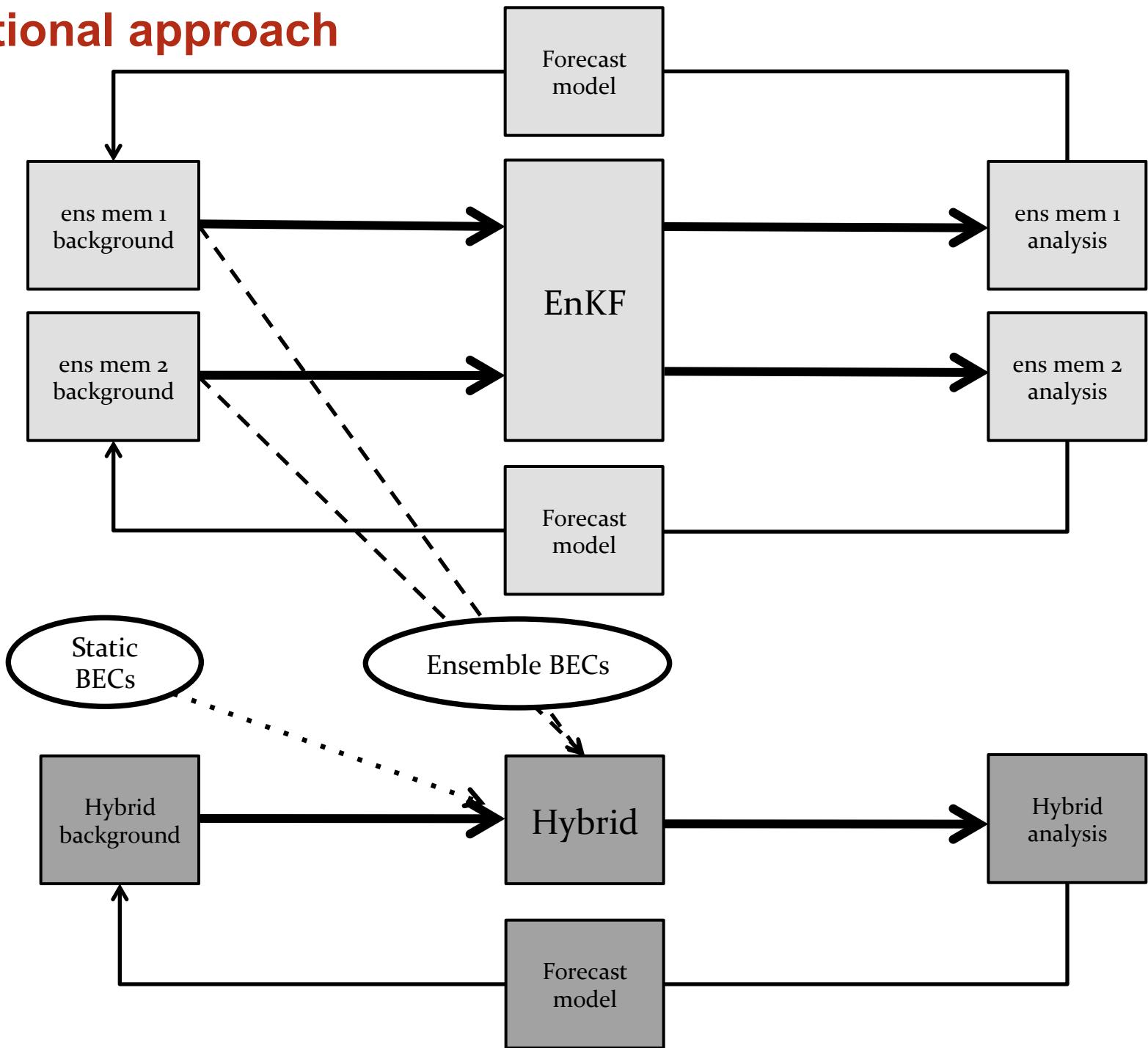
# **Introduction to integrated variational/ ensemble DA approaches**

# EVIL DA

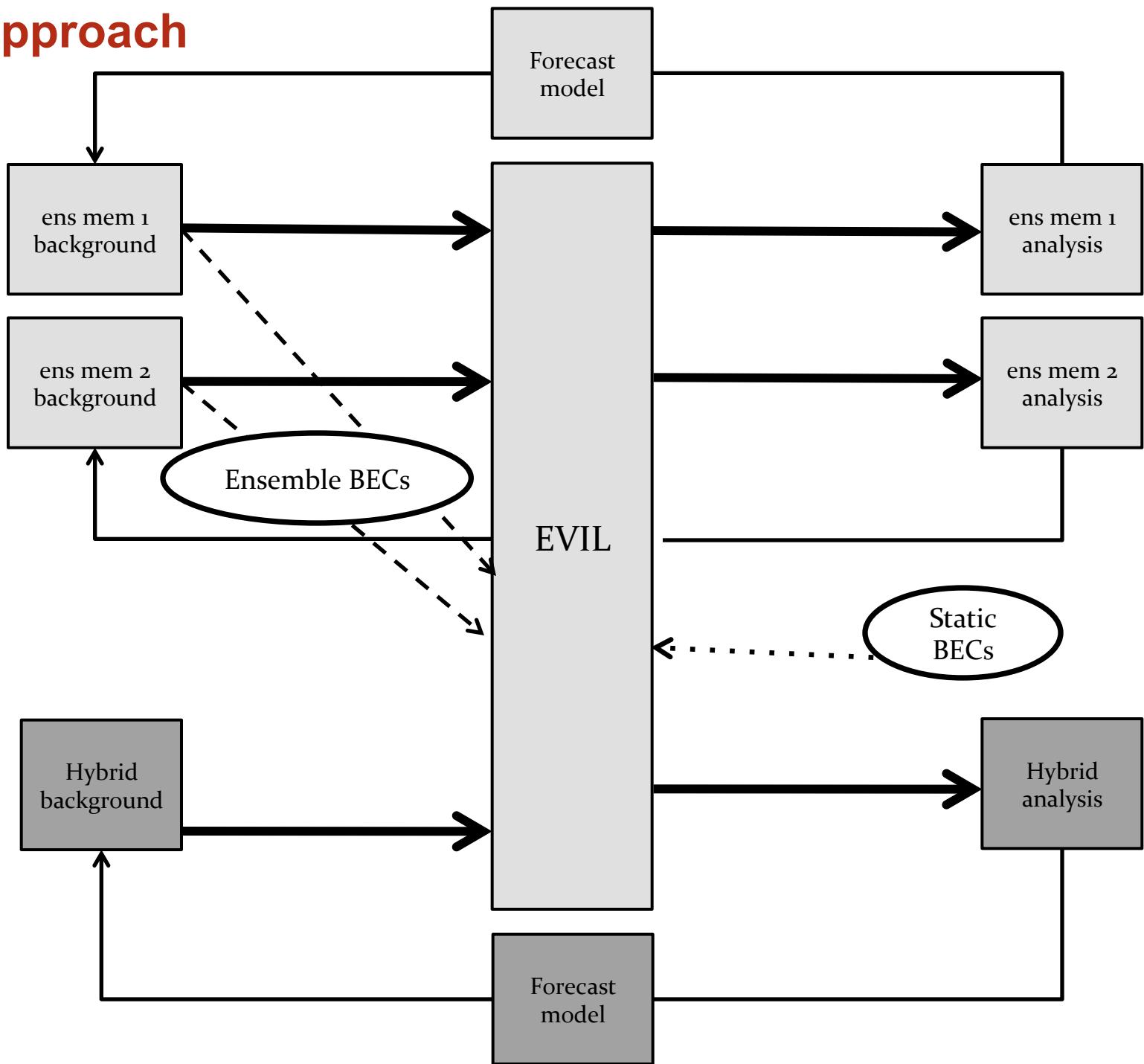
- EVIL stands for “Ensemble variational integrated localized”
- Developed by Tom Auligne (NCAR)
- Simultaneously updates both an ensemble and deterministic background
- Uses properties of the variational minimization to produce the analysis ensemble



# Traditional approach



# EVIL approach



# Summary

- The hybrid method is a practical way of incorporating ensemble BECs into DA systems
- Dual-resolution method is also a highly practical tool
  - Still not completely clear how much is lost through use of coarse-resolution ensemble
- Integrated variational/ensemble methods (e.g., EVIL) are “hot” research topics