

**Exec Summary:**  
**TLDR:**

1. Nudging is worse than the open-loop ensemble.  
2. Ensemble data assimilation improves forecasts but updating aquifer storage is tricky.

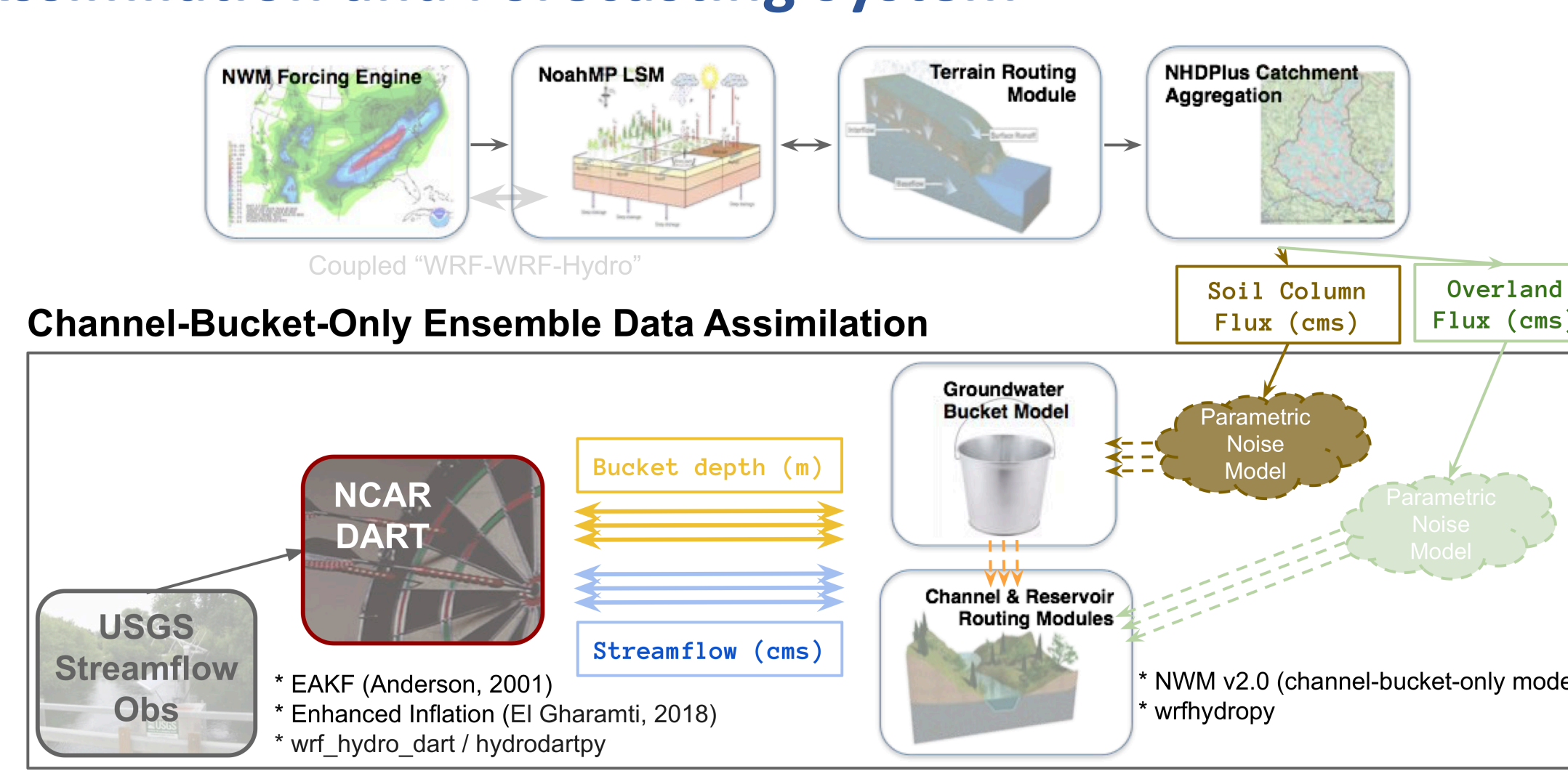
3. Forcing errors dominate forecasts: improvements from initial conditions fade after ~10 hours.  
4. Novel wavelet approach identifies “events” and their timing errors.

### Ensemble Streamflow Assimilation and Forecasting System

**US National Water Model:**

- NOAA Office of Water Prediction
- Operational on continental US
- Forced by NWP products from NOAA National Weather Service.
- 1km land model
- 250m overland and subsurface hydrologic routing
- Streams and lakes on vector network (~1.5 km)
- *Streamflow nudging data assimilation: USGS observations*
- Aquifer “bucket” model

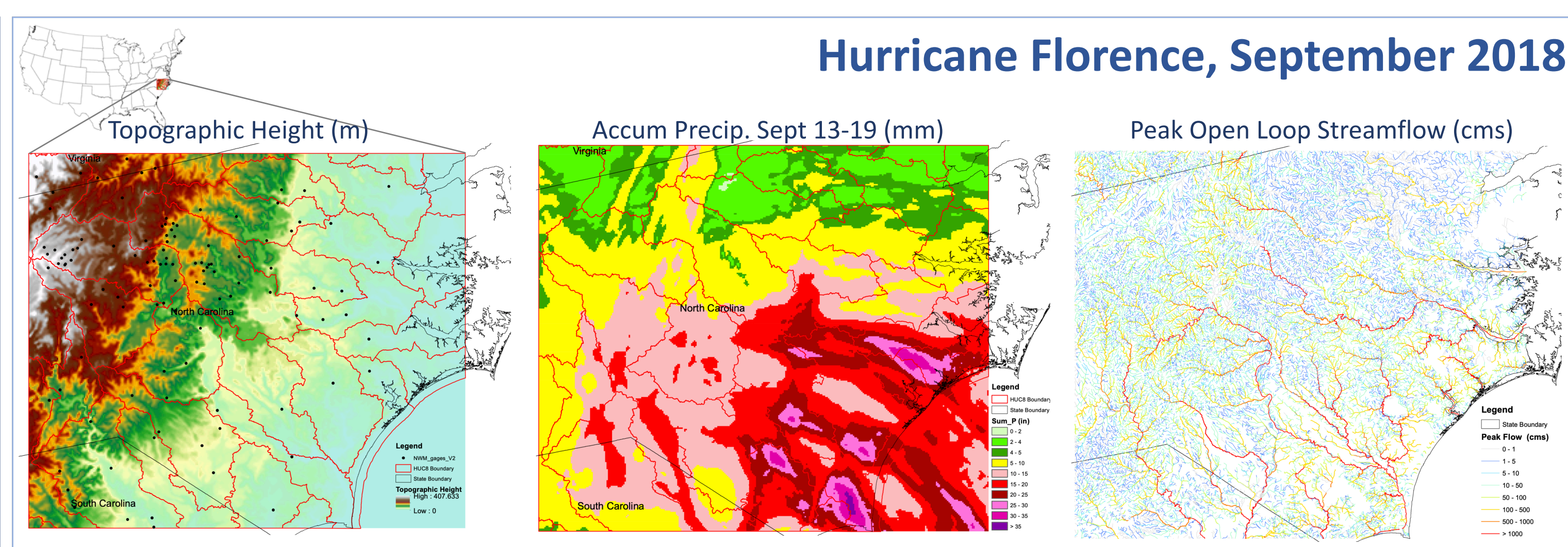
**Channel-Bucket-Only Ensemble Data Assimilation**



**NCAR's Data Assimilation Research Testbed (DART):**

- Ensemble assimilation is performed within NCAR's DART.
- Ensemble Adjustment (deterministic) Kalman Filter (Anderson, 2001)
- Spatially- and temporally- varying adaptive inflation (Gharamti, 2018)

### Hurricane Florence, September 2018



- Record-breaking rainfall totals (> 30 in = 76 cm)
- Storm surge
- 105 USGS streamflow gages in the domain
  - *Real-time flows not viable for assimilation experiments, using post-corrected flows*
- Widespread flooding
- Realtime NWM AnA forcings used for both analysis and forecasts (*hindcast experiments*)

### Methods and Design

**Uncertainty:**

1. Flux perturbations (depicted above)
2. Channel parameters (6 including geometry and mannings N)
3. Bucket parameters (2)

**State Updating:**

1. Streamflow
2. Streamflow & bucket depth

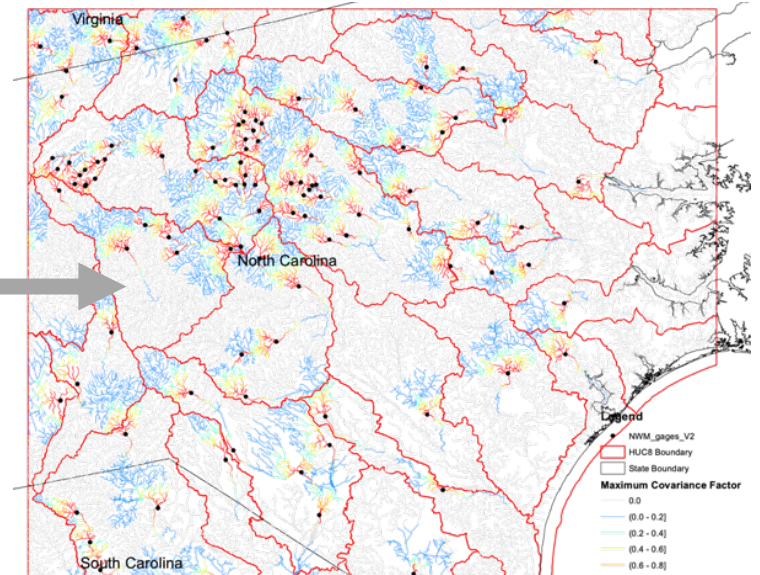
**Experiments:**

- **open-loop-multiphysics:** channel & bucket param uncertainty
- **nudging:** deterministic, flux perturb (all ensembles have perturbed fluxes)
- **nobucket-channel-multiphysics:** No bucket updating, channel param uncertainty
- **nobucket-multiphysics:** No bucket updating, channel & bucket param uncertainty
- **bucket-multiphysics:** Bucket updating, channel & bucket param uncertainty

**Observation Error Variance:** 20%

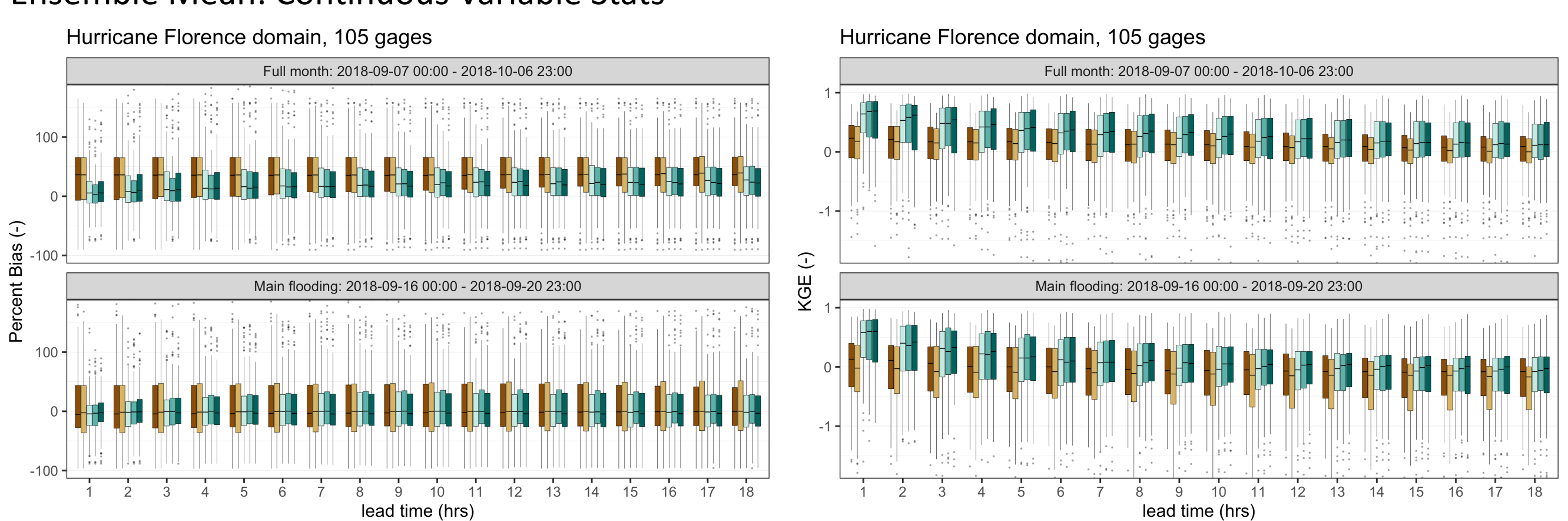
**Localization:** Along-stream metric (as-the-fish-swims) localization strategy

**Gaussian Anamorphosis:** technique for the non-gaussian updated states.... Still in development.

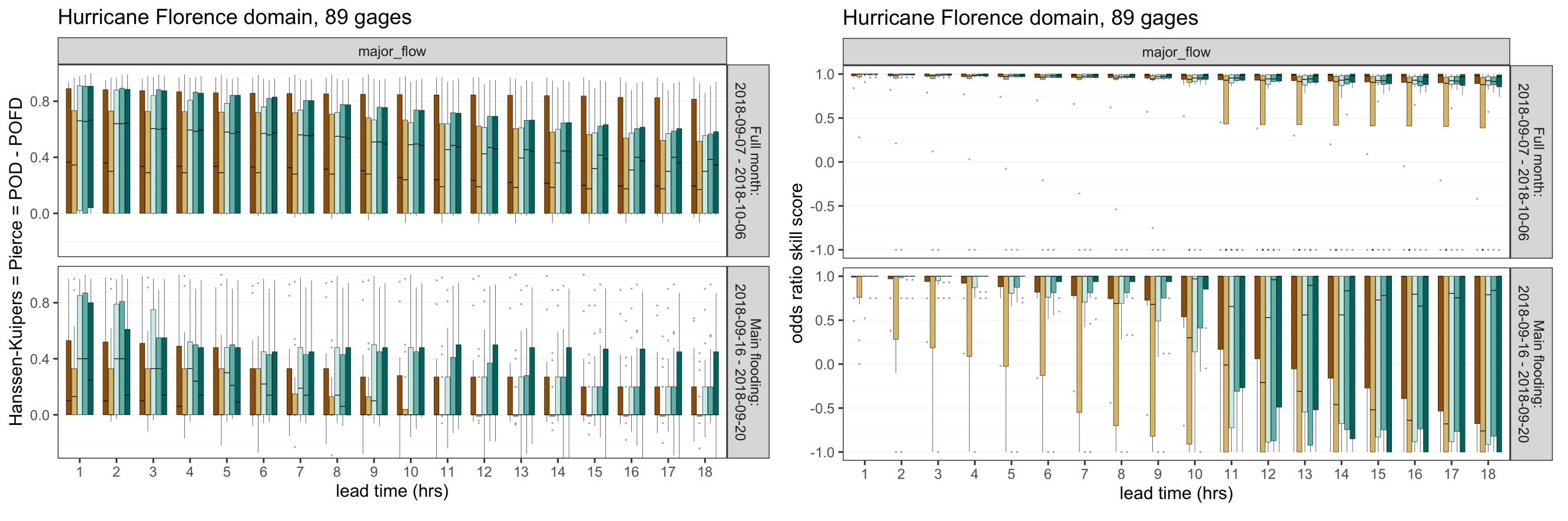


### Forecast Evaluation

**Ensemble Mean: Continuous Variable Stats**



**Ensemble Mean: Dichotomous Variable (Contingency) Stats**



**Nudging is worse than the open-loop ensemble:**

- Ensemble mean evaluation:
  - Full period: generally similar distributions to open-loop at all lead times
  - Flooding period: generally worse distributions than open-loop
    - Hansen-Kuipers = POD – POFD: IQR negative for large lead times
    - ORSS: Nudging struggles against random chance to discriminate major floods
- Ensemble distribution evaluation: Time-lagged ensembles
  - “Incomplete” until verifying: Evaluate as a function of min lead time
  - Time-lagged ensemble CRPSS: Much worse than open-loop

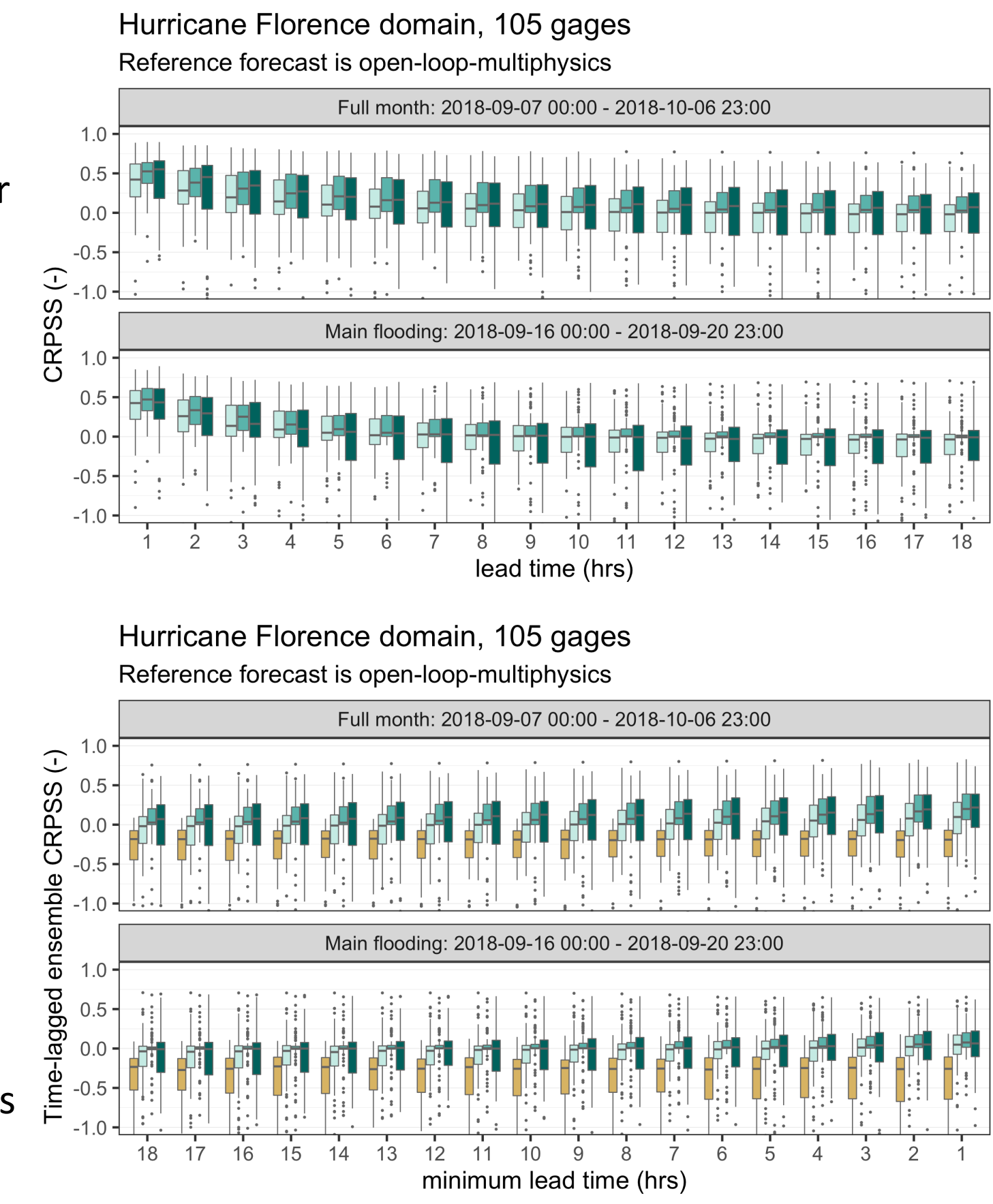
**Forcing errors dominate forecasts:**

- Ensemble mean evaluation:
  - For all stats, the distributions of the assimilation experiments converge towards the open-loop distribution within about ~10 hours
  - The presented stats are the exception to this rule: convergence is slower
- Ensemble distribution evaluation:
  - Two of the experiments converge to the reference (open-loop) around 9h
- **Coupled atmosphere – hydro data assimilation to fix “forcing” errors**

**Ensemble DA improves forecasts, updating bucket is tricky:**

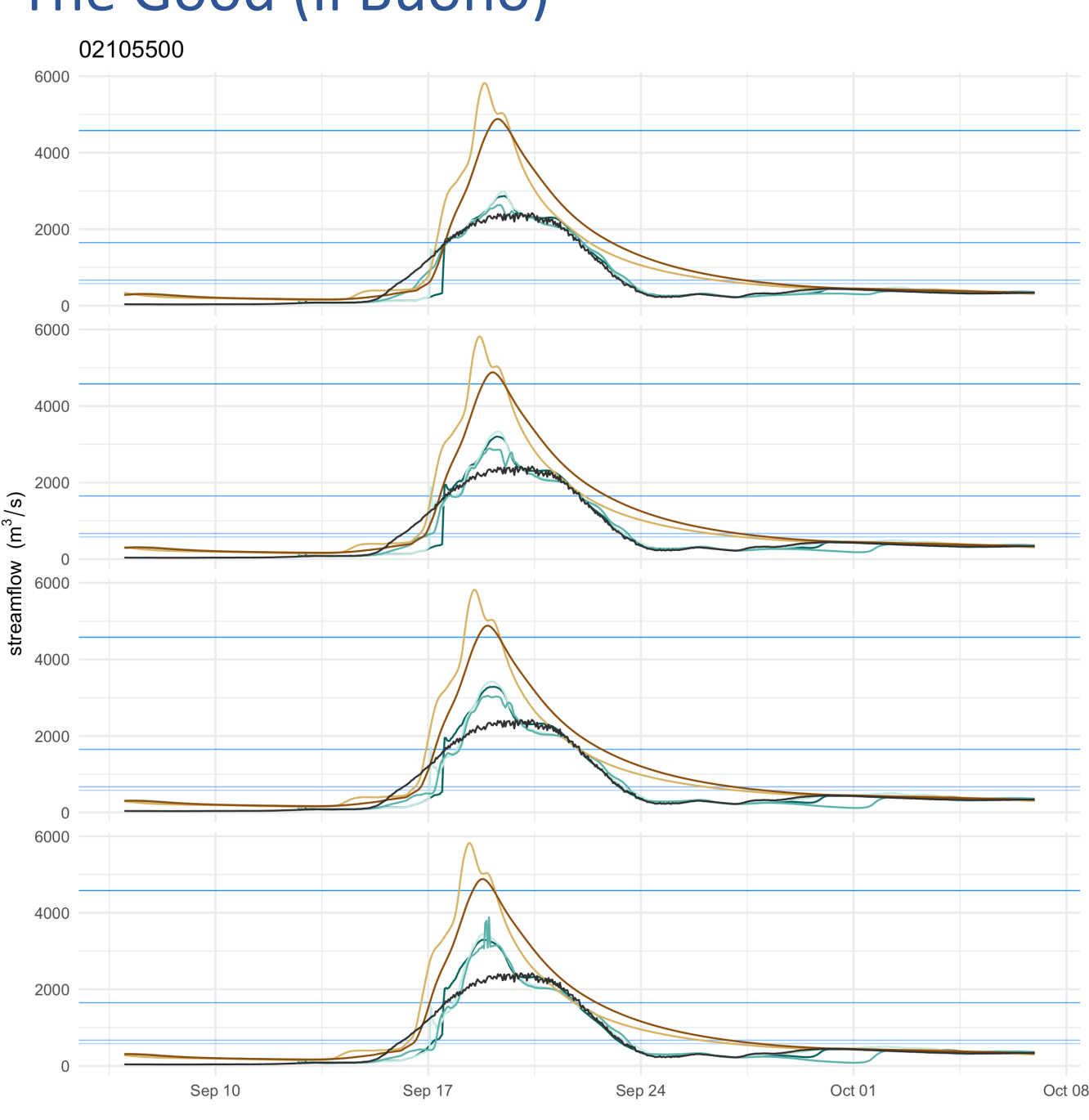
- Updating the bucket (aquifer) level has pros and cons
- Ensemble mean evaluation:
  - Generally bucket-multiphysics has similar or slightly better distributions as the other assimilation experiments
- Ensemble distribution evaluation:
  - Bucket-multiphysics has wider distributions with better and worse CRPSSs

**Ensemble Distribution: Probabilistic Variable Stats**

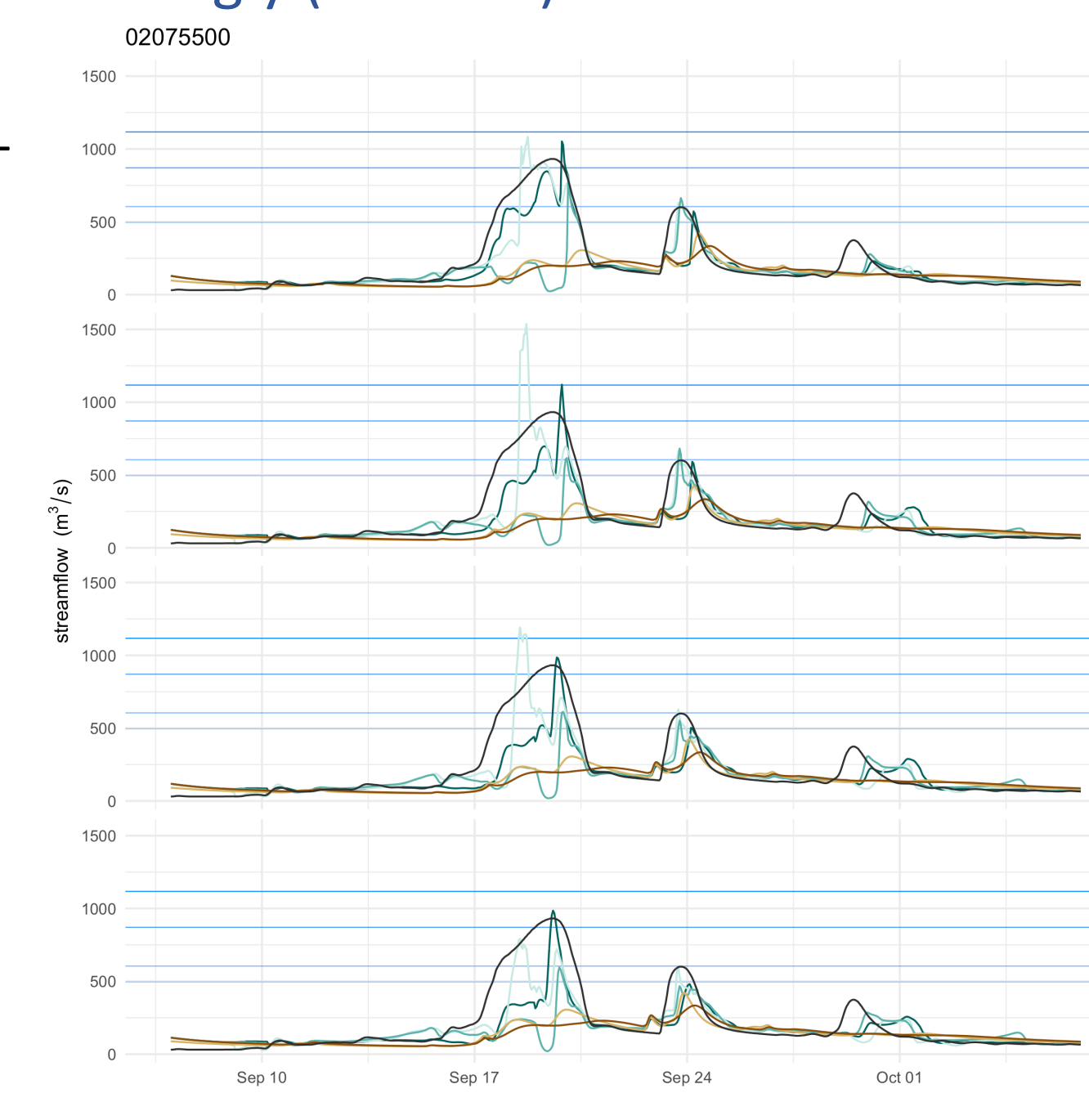


### Evaluation Illustrated: Example Ensemble Mean Forecast Hydrographs

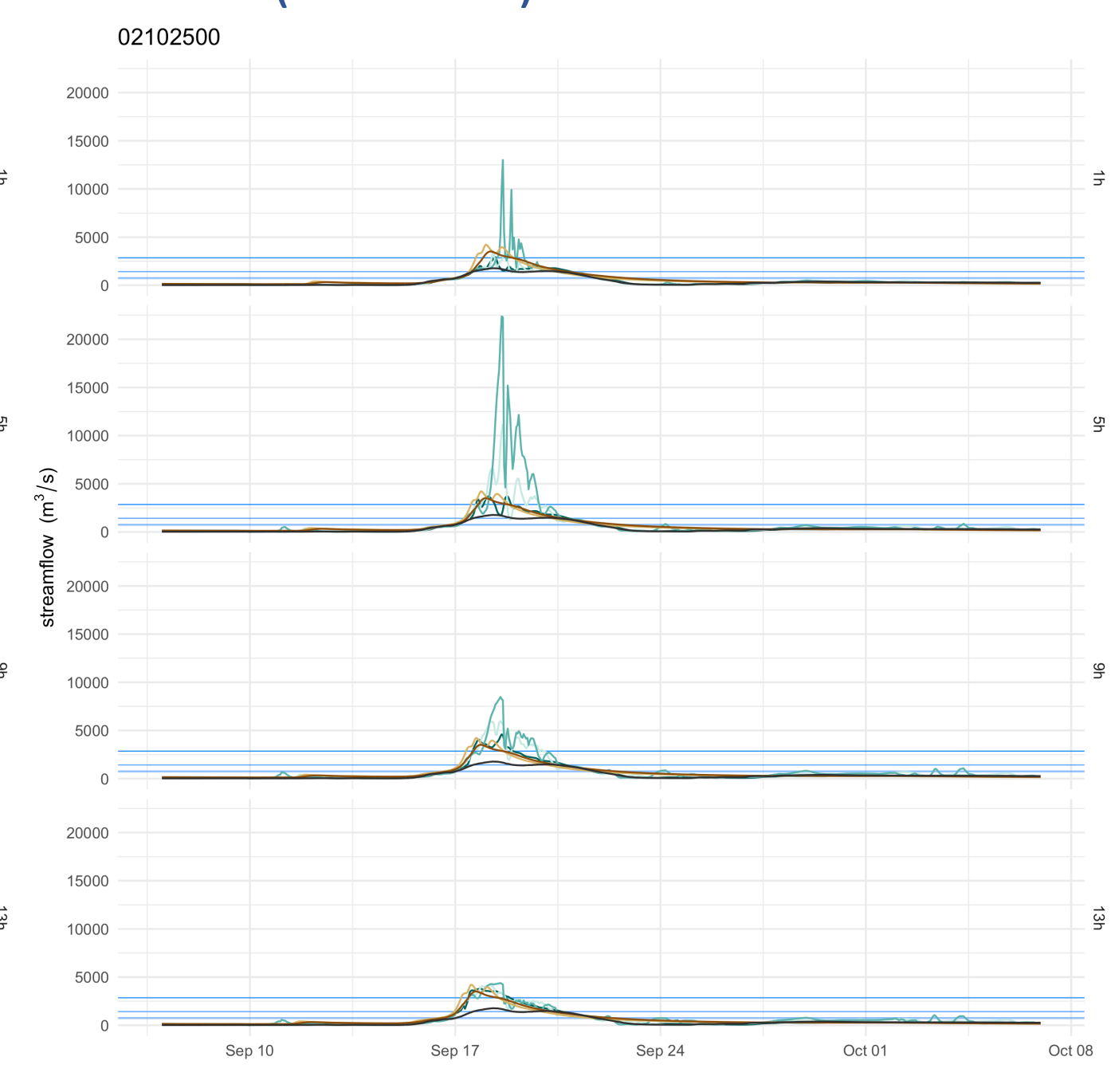
**The Good (Il Buono)**



**The Ugly (Il Brutto)**



**The Bad (Il Cattivo)**



**experiment**

- bucket-multiphysics
- nobucket-channel-multiphysics
- nobucket-multiphysics
- nudging
- open-loop-multiphysics
- observed

**The Good:**

- Strong bias correction (high bias reduction)
  - Even at long lead times, not converging to OL
  - Are upstream corrections to thank?
- Assimilation eliminates a “major” false alarm
- Descending limb is much more accurate
- General bias improvements outside flooding

**The Ugly:**

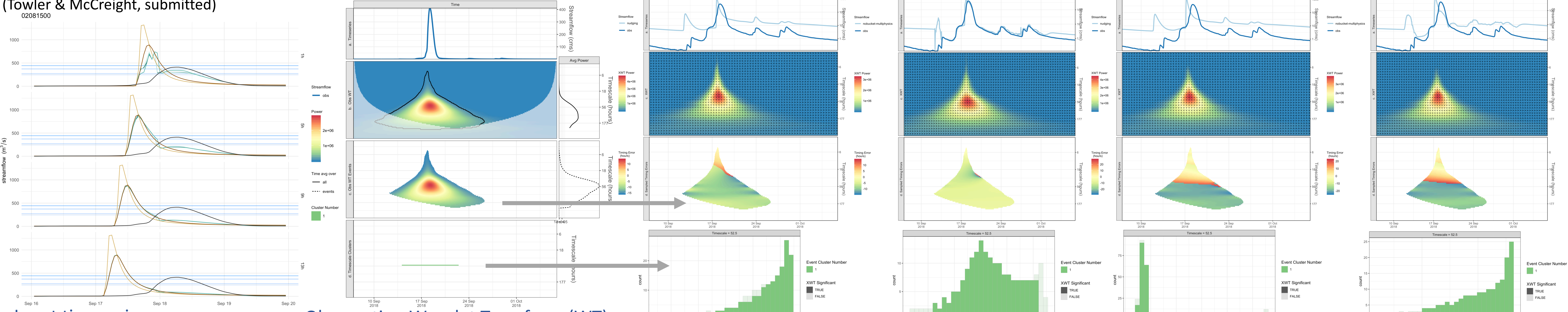
- Bias correction (low bias correction)
- Open loop and nudging register “miss”
- bucket-multiphysics: much better during first event, worse during second event
- nobucket-channel-multiphysics: False alarms for first event at medium lead times
- nobucket-multiphysics: unrealistic oscillatory behavior, worse misses than nudging and open-loop. **Smoother?**

**The Bad:**

- False alarms galore!
- Highly unrealistic spikes reflected in the mean are under investigation

### Novel Wavelet Event Detection and Timing Error Quantification

(Towler & McCreight, submitted)



**Input timeseries:**

- Simple event example
- Observed
- Modeled:
  - 1 and 13 hour forecasts
  - Nudging & nobucket-multiphysics

**Observation Wavelet Transform (WT):**

- Significance defines events (b -> c)
- Events have dimensions: time, period
- Peaks in event spectrum: “characteristic” periods of observed events = ~53hr
- Timescale event clusters (c -> d)

**Cross-Wavelet Transforms (XWT):**

- XWT phase (panel b) gives timing error on observed events (panel c)
- XWT significance (not shown) defines timing error significance
- Timing error stats (bottom) calculated on events’ characteristic periods defined by the observation WT and its event spectrum

**Timing Error Distributions:**

- Stats can be taken on the significant timing errors
- 1 hr: nudging is ~11hr early, nobucket-multiphysics is ~5hr early
- 13 hr: nudging is ~23hr early, nobucket-multiphysics is ~20.5hr early
- Timing errors are converging with lead time (as are forecasts).