

# Observing System Simulation Experiments

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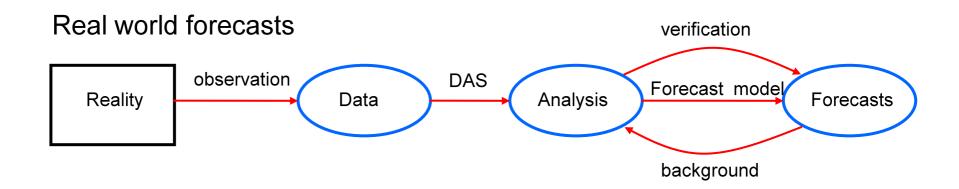
#### What is an OSSE?

An OSSE is a modeling experiment used to evaluate the impact of new observing systems on operational forecasts when actual observational data is not available.

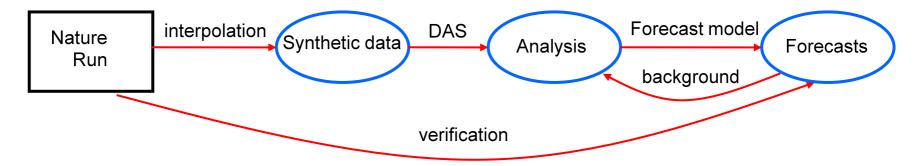
- A long free model run is used as the "truth" the Nature Run
- The Nature Run fields are used to back out "synthetic observations" from all current and new observing systems.
- Suitable errors are added to the synthetic observations
- The synthetic observations are assimilated into a different operational model
- Forecasts are made with the second model and compared with the Nature Run to quantify improvements due to the new observing system



#### OSSEs vs. the Real World



#### **OSSE** forecasts





## Why do an OSSE?

- 1. You want to find out if a new observing system will add value to NWP analyses and forecasts
- 2. You want to make design decisions for a new observing system
- 3. You want to investigate the behavior of data assimilation systems in an environment where the truth is known



#### When not to run an OSSE

- When you can't model the phenomena you are interested in
- When you can't simulate your new observations
- When you can't assimilate your new observations



#### Nature Runs

- Nature Runs act as the 'truth' in the OSSE, replacing the real atmosphere.
- Usually, a long free (non-cycling) forecast from the best available model is used as the NR
  - Model forecast has continuity of fields in time
  - Sometimes an analysis or reanalysis sequence is used, but the sequence of states of truth can never be replicated by a model
- Always a push for bigger, higher resolution NR

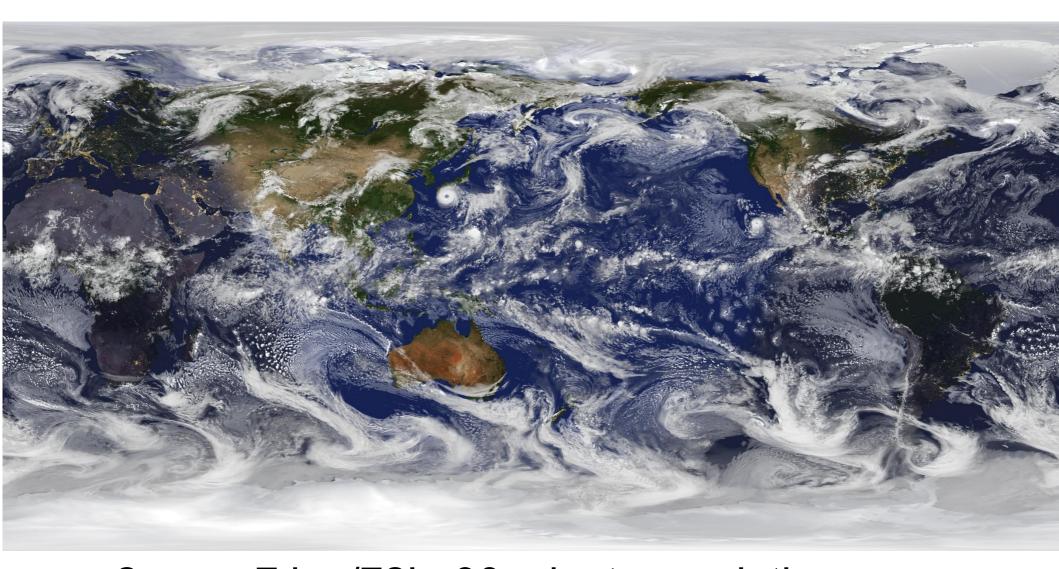


## Nature Run Requirements

- Must be able to realistically model phenomena of interest
  - Dynamics and physics should be realistic
  - Must produce fields needed for "observations"
  - Should be verified against real world
- Ideally is 'better' than the operational model to be used for experiments
- Preferably a different model base is used for the NR and the experimental forecast model to reduce incestuousness

#### **G5** Nature Run





2 year, 7 km/72L, 30 minute resolution 15 aerosols, ozone, CO, CO2





#### Common Problems with Nature Runs

- Nonexistence
- Identical or fraternal twins
- Outdated by the time you get to use them
- Gigantic output files and huge computational resource requirements
  - Output saved at full spatial resolution but 30 min + intervals

Full resolution (1.5 km, 10 min)

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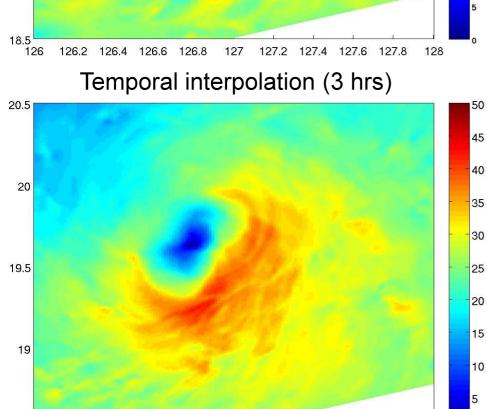
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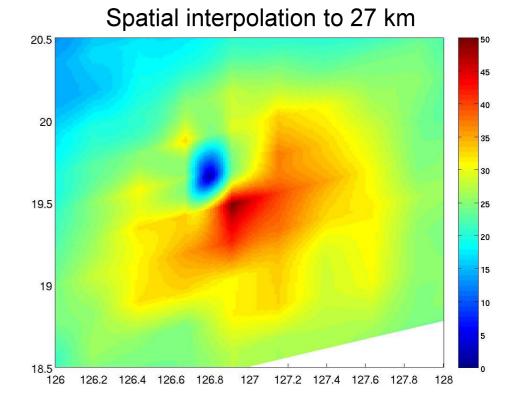


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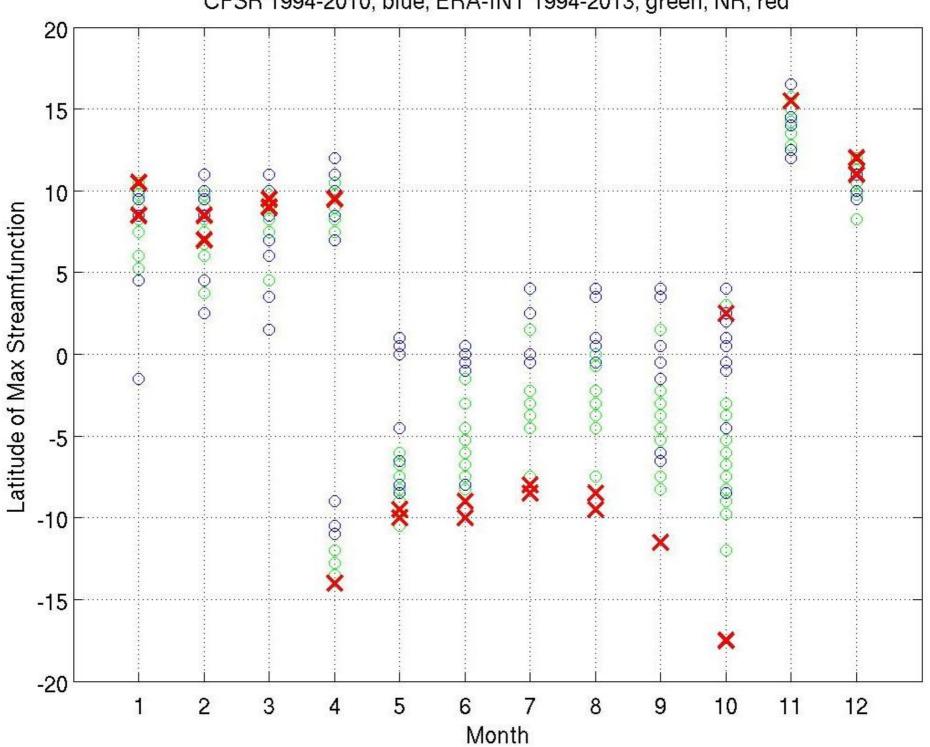
Comparison of temporal and spatial Interpolation errors compared to 1.5 km run for Typhoon Guchol (2012).



#### Nature Run Validation

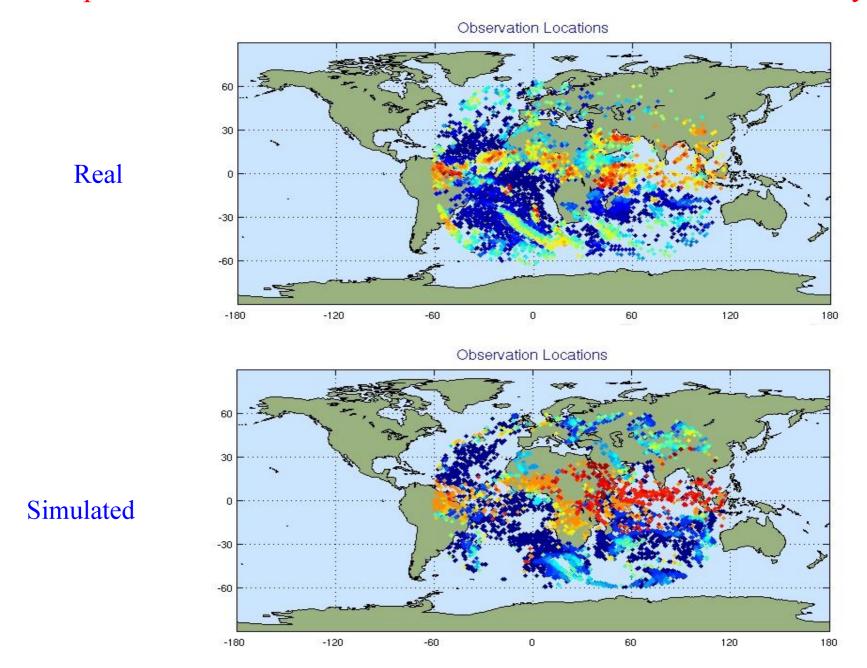
- Evaluate if NR is sufficiently realistic to yield meaningful results
- In addition to the phenomena of interest, the NR needs to realistically replicate fields needed to generate synthetic observations
- Can't validate everything; corollary don't expect a NR to come pre-validated for your needs

Latitude of maximum monthly mean zonal mean streamfunction CFSR 1994-2010, blue; ERA-INT 1994-2013, green; NR, red



## Synthetic Observations

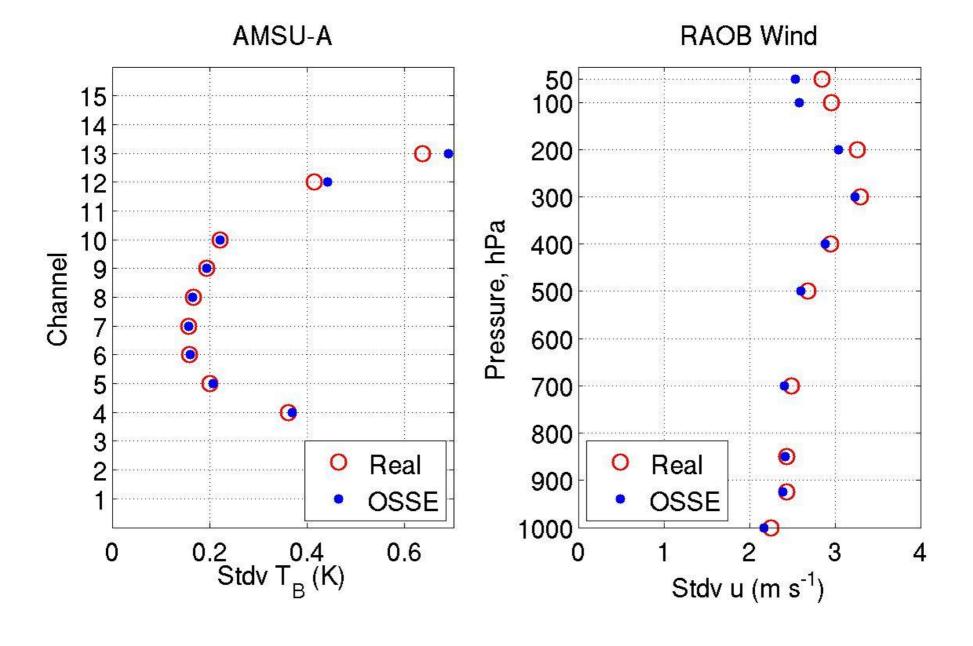
Example of METEOSAT AMV observations at 0000 UTC 10 July





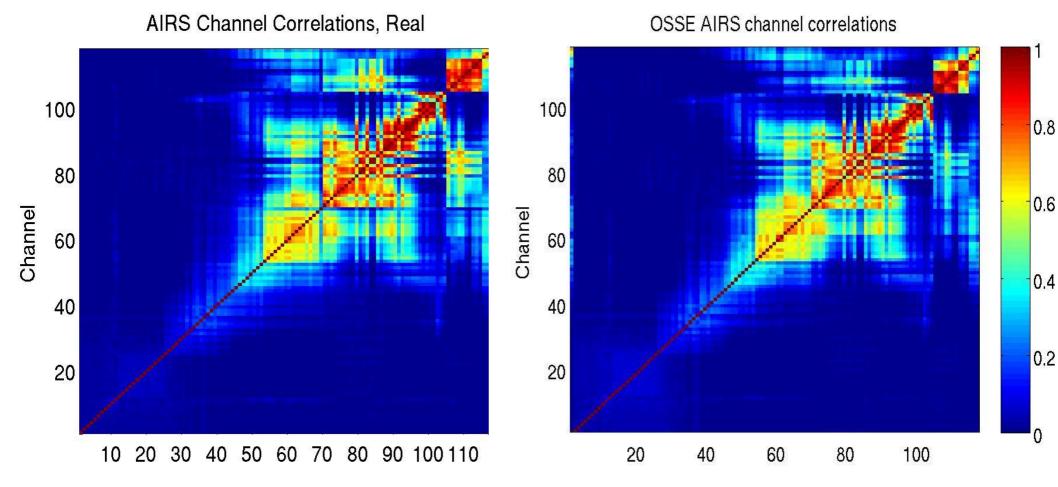
#### **Observation Errors**

- Synthetic observations contain some intrinsic interpolation/operator errors, but less than real observations (usually)
- Synthetic errors are created and added to the synthetic observations to compensate
- Error is complex and poorly understood
  - Error magnitude
  - Biases
  - Correlated errors

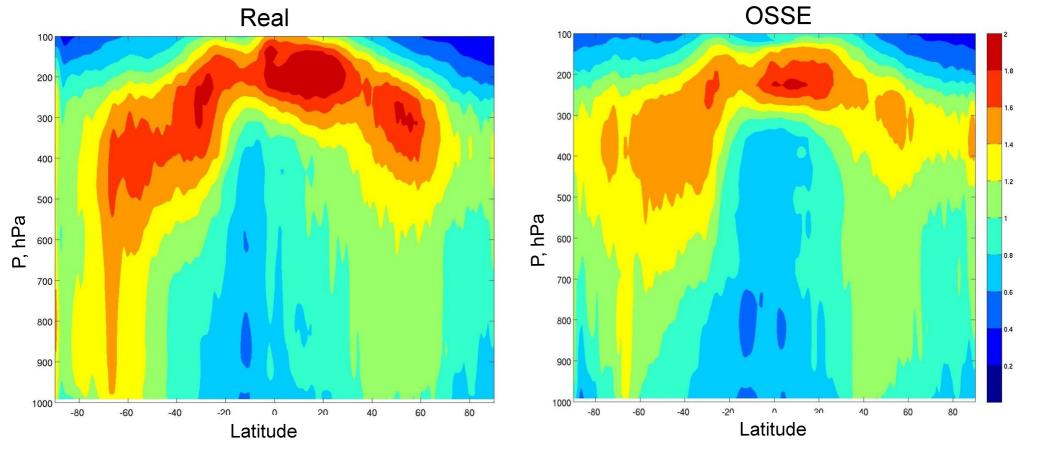


O-F is fairly easy to calibrate because you can manipulate O directly.





Some observation correlations are relatively easy to calibrate

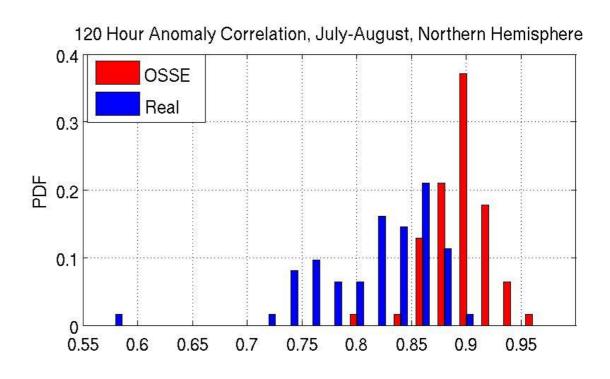


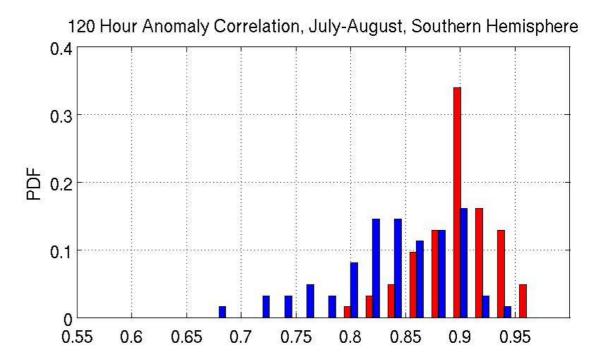
A-B (analysis increment) is a little harder to calibrate, as A and B are not directly controlled

Model error strongly influences forecast skill in the longer term forecast, so calibration is not possible (unless you want to mess with your model).

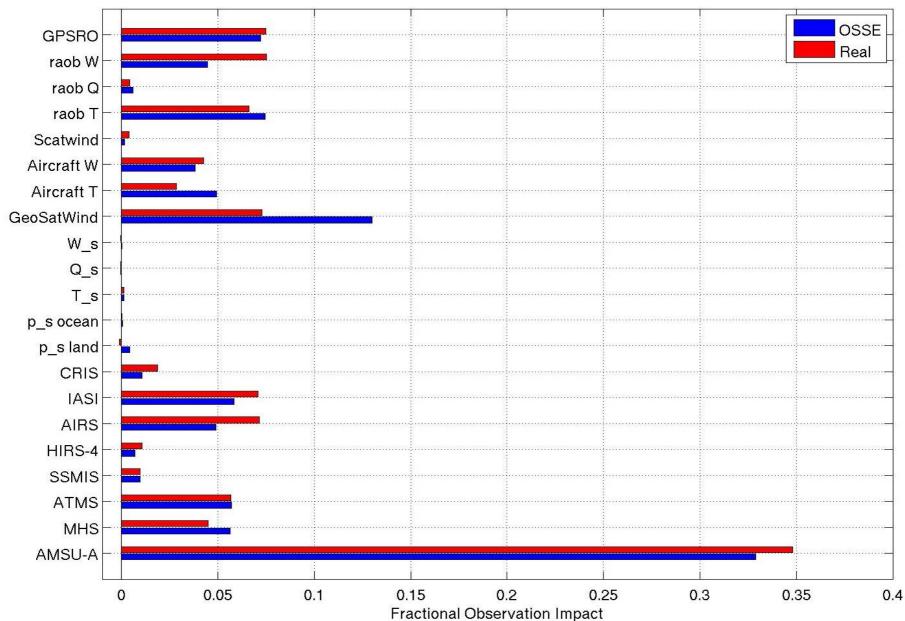
Red: OSSE Blue: Real

500 hPa anomaly correlations of geopotential height, G5NR





## Why believe OSSE results?



New observations can be put into context relative to existing observation impacts



### Criticisms of OSSEs

- Results only apply within the OSSE system no concrete connection to the real world
- Even the best OSSEs are far from perfect: incestuousness, difficulty in generating observations and errors, deficiencies of the Nature Run
- By the time the new instrument is deployed, both the global observing network and the forecast models/DAS will be different
- Examples of sloppy or unsuccessful OSSEs



## Takeaway

- OSSEs can provide useful information about new observational types and the workings of data assimilation systems
- Careful consideration of research goals should guide each step of the OSSE process
- Validate your OSSE!
- OSSEs are hard, good OSSEs are harder