

# Forecast Sensitivity to Observation Impacts and the U.S. Integrated Ocean Observing System

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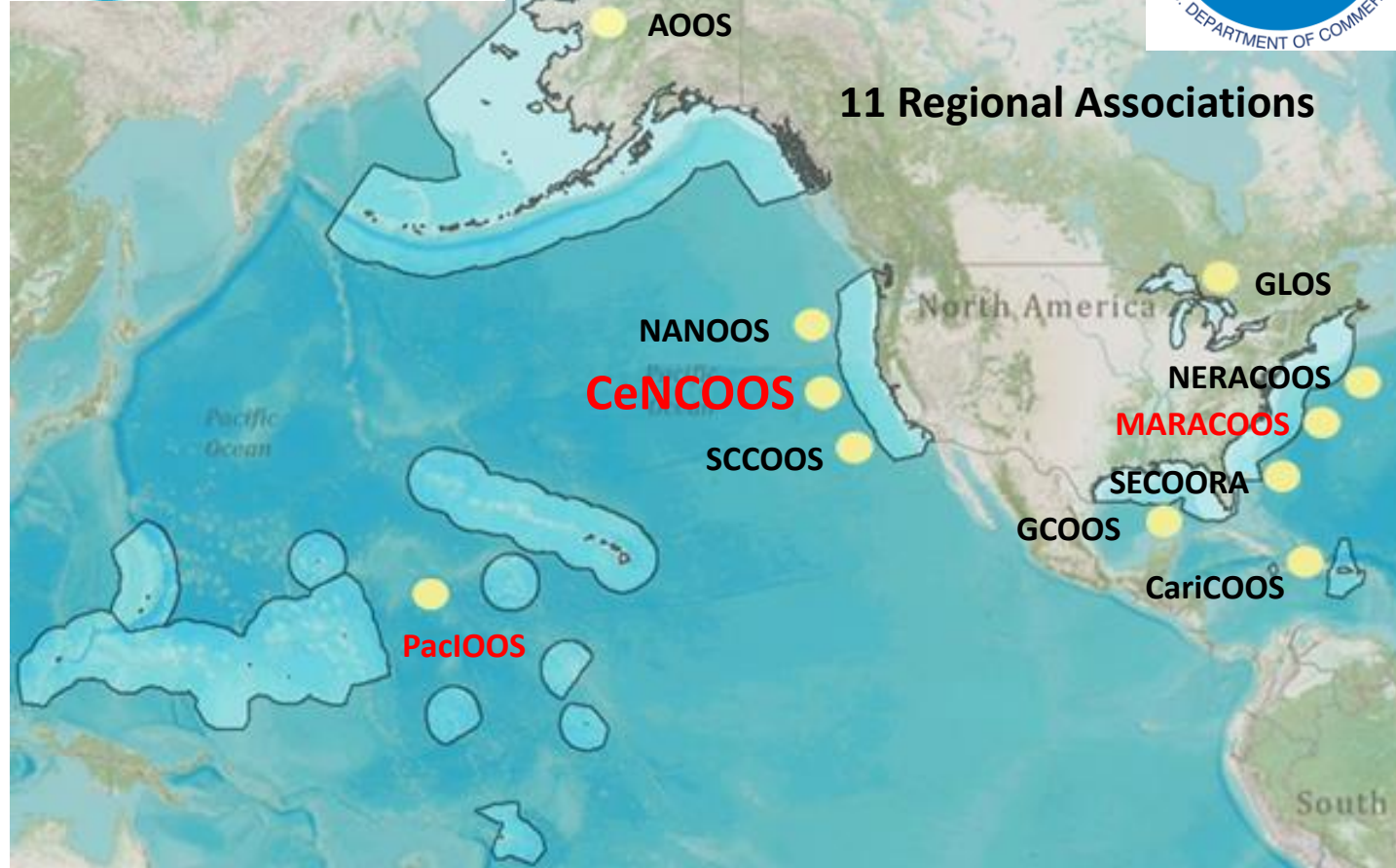
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**CenCOOS, MARACOOS & PacIOOS** near real-time analysis-forecast systems are all based on Regional Ocean Modeling System (ROMS) 4-dimensional variational (4D-Var) data assimilation.

$$\mathbf{x}^a = \mathbf{x}^b + \mathbf{B}\mathbf{H}^T(\mathbf{H}\mathbf{B}\mathbf{H}^T + \mathbf{R})^{-1}(\mathbf{y} - \mathbf{H}(\mathbf{x}^b))$$

$\mathbf{x}^a$  = analysis state-vector

$\mathbf{x}^b$  = background state-vector

$\mathbf{y}$  = observation vector

$\mathbf{H}$  = observation operator (NLROMS)

$\mathbf{H}$  = TL observation operator (TLROMS)

$\mathbf{H}^T$  = AD observation operator (ALROMS)

$\mathbf{B}$  = background error covariance

$\mathbf{R}$  = observation error covariance

Incremental, dual formulation, Lanczos formulation of restricted  $\mathbf{B}$ -preconditioned CG to minimize:

$$J = \delta\mathbf{x}^T \mathbf{B}^{-1} \delta\mathbf{x} + (\mathbf{d} - \mathbf{H}\delta\mathbf{x})^T \mathbf{R}^{-1} (\mathbf{d} - \mathbf{H}\delta\mathbf{x})$$

$\delta\mathbf{x} = \mathbf{x}^a - \mathbf{x}^b$  = increment

$\mathbf{d} = (\mathbf{y} - \mathbf{H}(\mathbf{x}^b))$  = innovation

**CeNCOOS** = Central and Northern California Ocean Observing System

**MARACOOS** = Mid-Atlantic Regional Association Ocean Observing System

**PacIOOS** = Pacific Island Ocean Observing System





Photo Dan Costa

Taggers



Argo floats

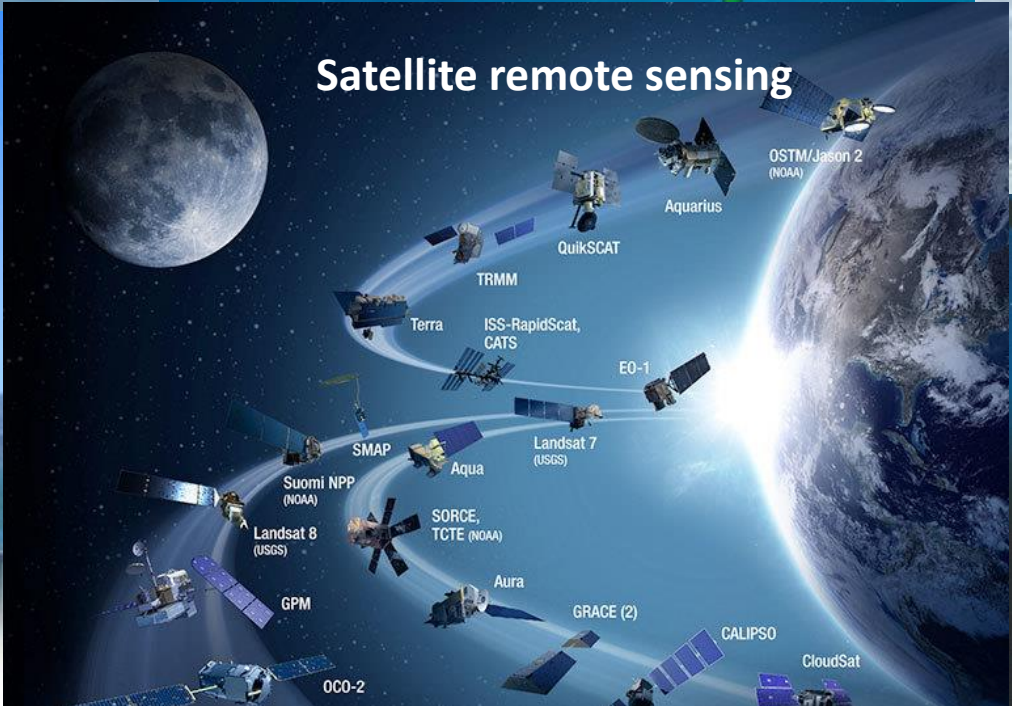


Buoys

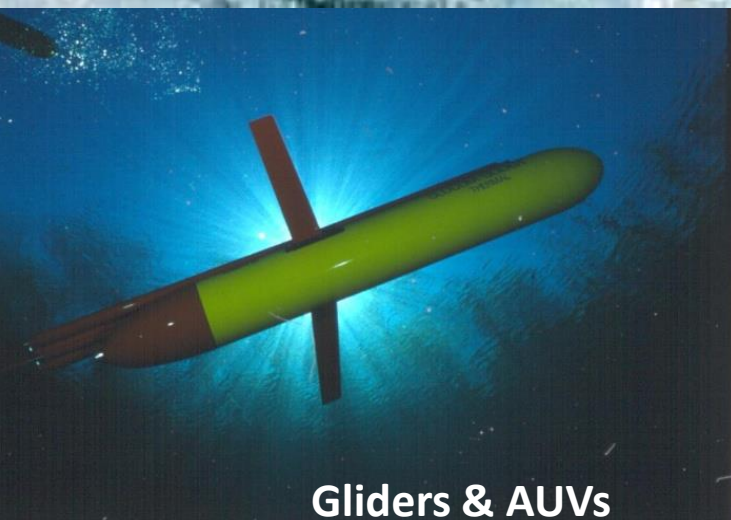


High frequency radars

San Francisco Bay, California



Satellite remote sensing

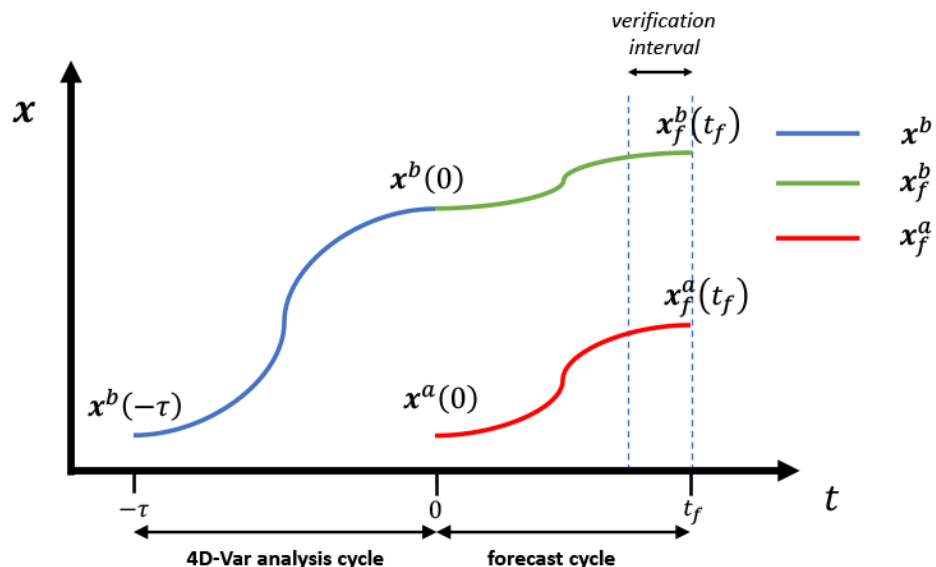


Gliders & AUVs

**What impact does each component of the observing system have on forecast skill?**

# Forecast Sensitivity to Observations

Baker & Daley (2000); Langland & Baker (2004); Errico (2007); Tremolet (2008); Zhu & Gelaro (2008)



Forecast skill metric:

$$e(t_f) = \left( x_f(t_f) - x^t(t_f) \right)^T \mathbf{C} \left( x_f(t_f) - x^t(t_f) \right)$$

In practice  $x^t(t_f) \rightarrow x^a(t_f)$ , a “*verifying analysis*.”

Alternatively, we can use independent observations  $y_f$ :

$$e(t_f) = \left( y_f(t_f) - y(t_f) \right)^T \mathbf{C} \left( y_f(t_f) - y(t_f) \right)$$

The change in forecast skill due to assimilating observations:

$$\delta e = e^a(t_f) - e^b(t_f)$$

To 3<sup>rd</sup>-order:

$$\delta e \approx \mathbf{d}^T (\mathbf{H}\mathbf{B}\mathbf{H}^T + \mathbf{R})^{-1} \mathbf{H}\mathbf{B} \frac{1}{2} \left( \partial e^a / \partial \mathbf{x}^a |_{t=-\tau} + \partial e^b / \partial \mathbf{x}^b |_{t=-\tau} \right)$$

Observation impacts:

$$\delta e \approx \mathbf{d}^T \mathbf{g} = \sum_{i=1}^N d_i g_i$$

$$\delta e < 0$$

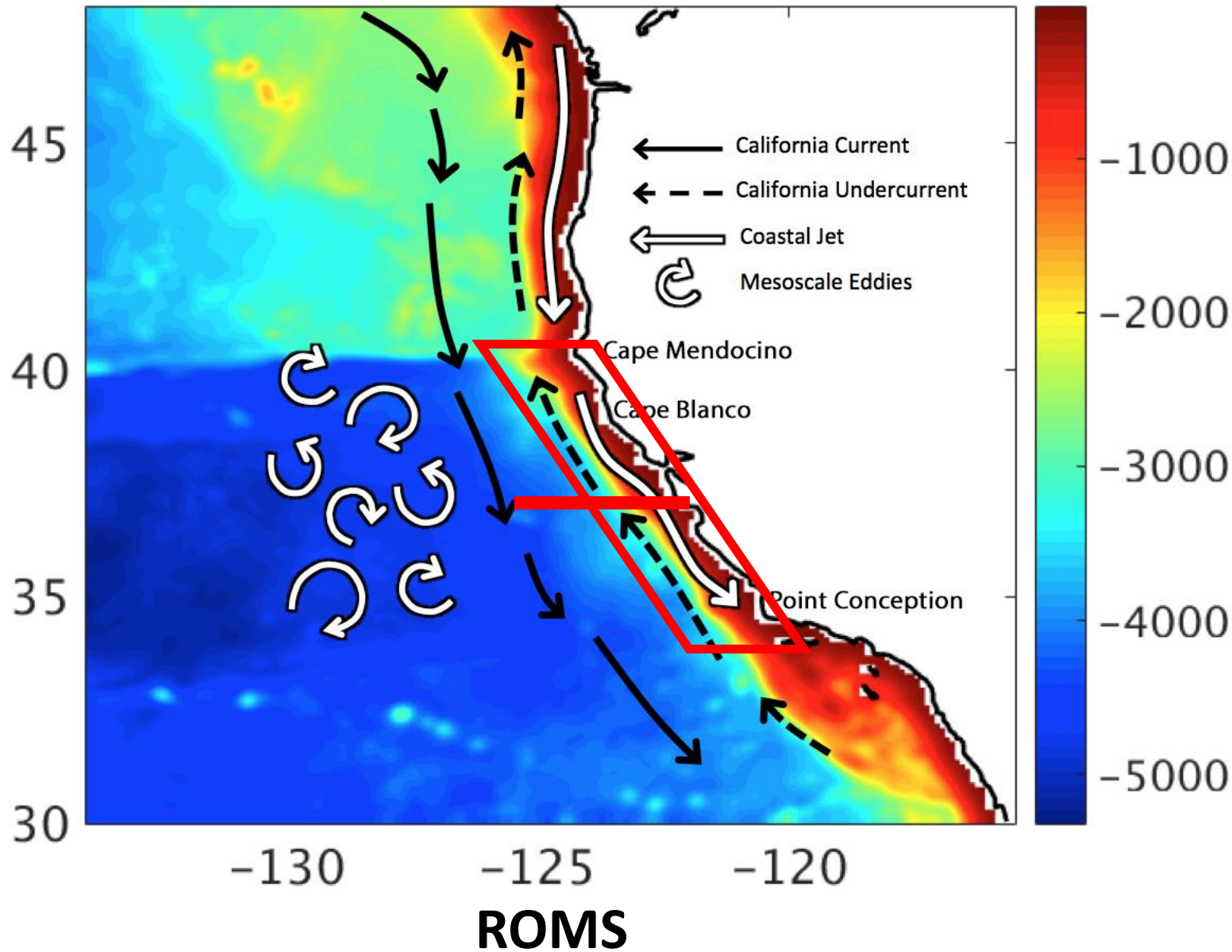
Obs improve forecast skill

$$\delta e > 0$$

Obs degrade forecast skill



# The California Current System



- 1/10<sup>th</sup> degree resolution, 42  $\sigma$ -levels
- COAMPS surface forcing
- Global HYCOM open boundary conditions
- Observations (2018 & 2019):
  - satellite SST
  - Aviso altimetry
  - Argo profiling floats
  - gliders
  - HF radar surface radial currents
- Background quality control of obs
- 4-day 4D-Var windows (1 outer-loop, 9 inner-loops)
- 4-day forecasts ("*hindcasts*")
- Forecast metrics:
  - central CA SST MSE  $e_{SST}$
  - central CA surface current MSE  $e_V$
  - central CA upwelling transport MSE  $e_W$
- 37N transport MSE  $e_{37N}$

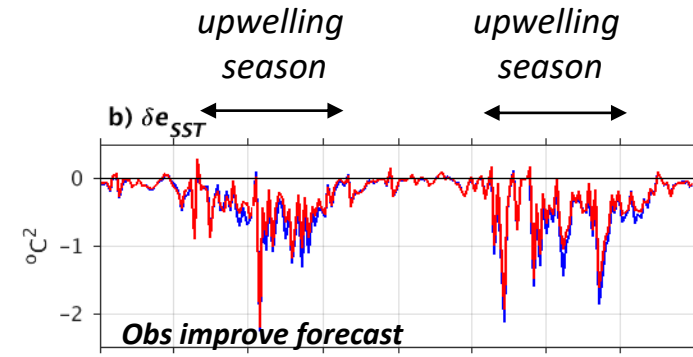
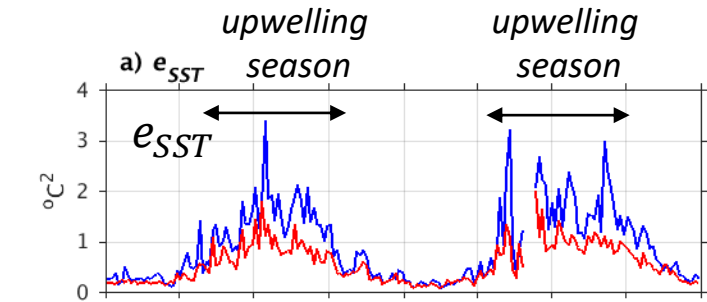
Directly observed

Indirectly observed

# 4-Day Forecast Errors

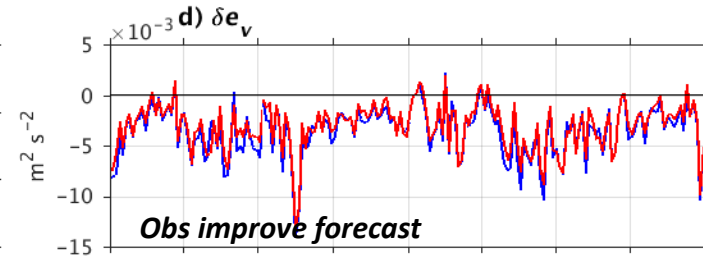
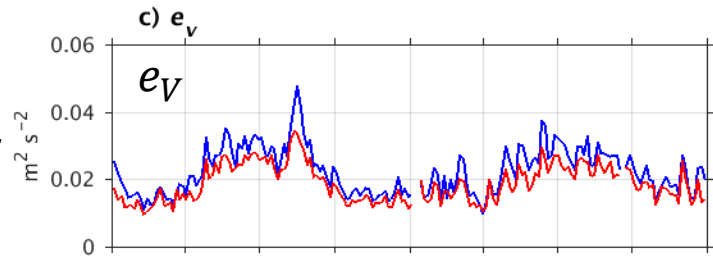
$\delta e < 0$   
**Obs improve forecast skill**

**Central CA SST MSE**



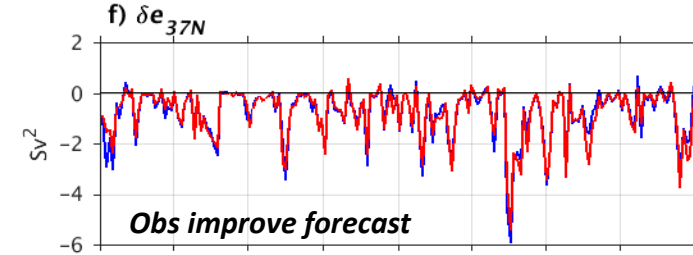
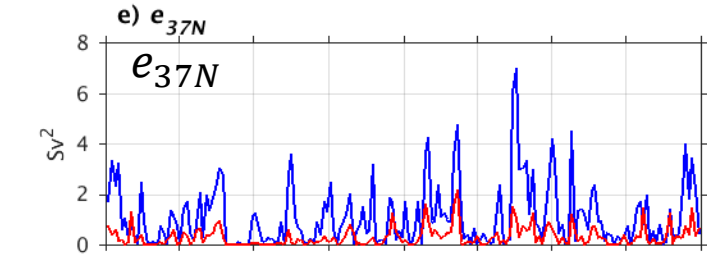
$\delta e_{SST}$

**Central CA currents MSE**



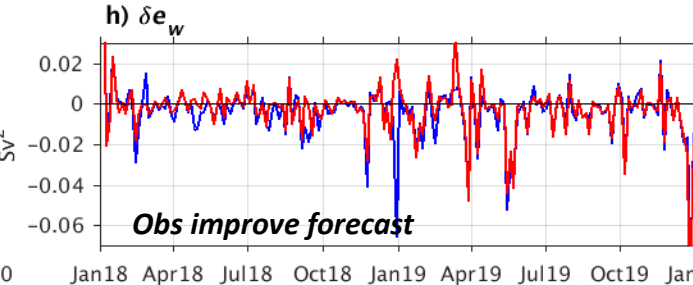
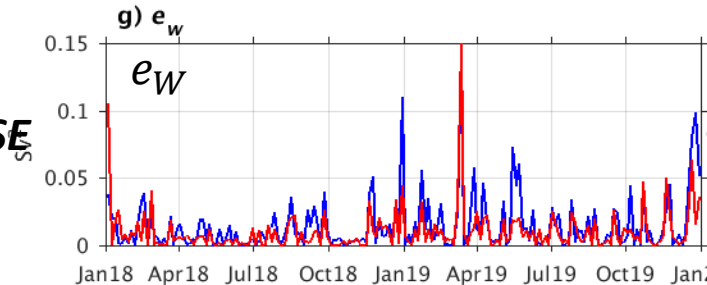
$\delta e_v$

**37N transport MSE**



$\delta e_{37N}$

**Central CA upwelling MSE**

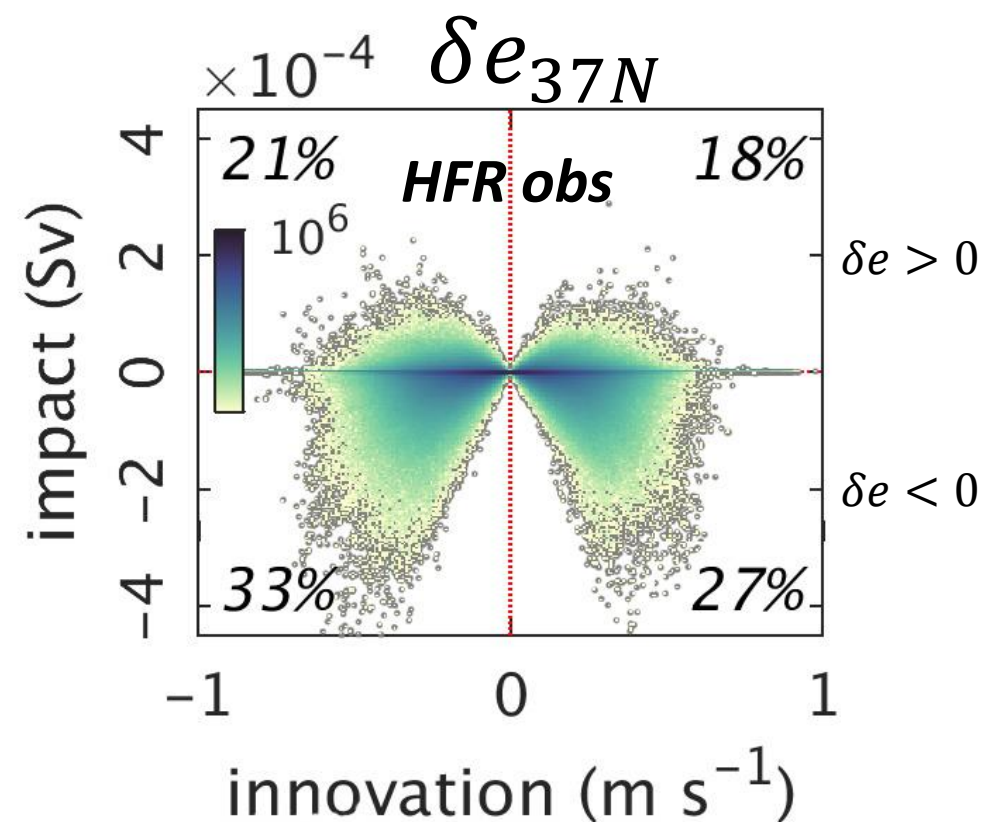
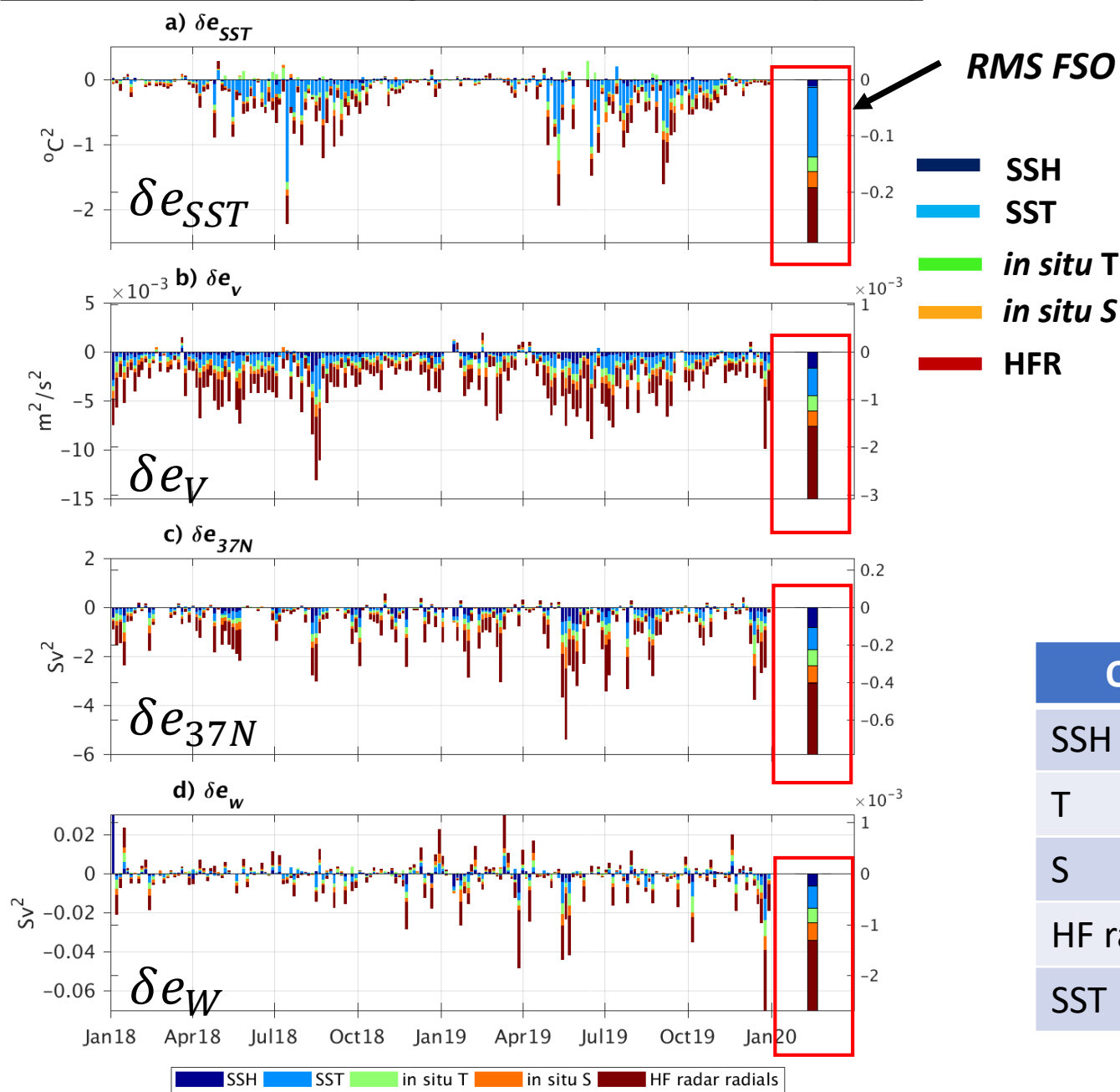


$\delta e_w$

— background  
 — analysis

—  $\delta e = e^a - e^b$   
 —  $\delta e$  3<sup>rd</sup>-order approx

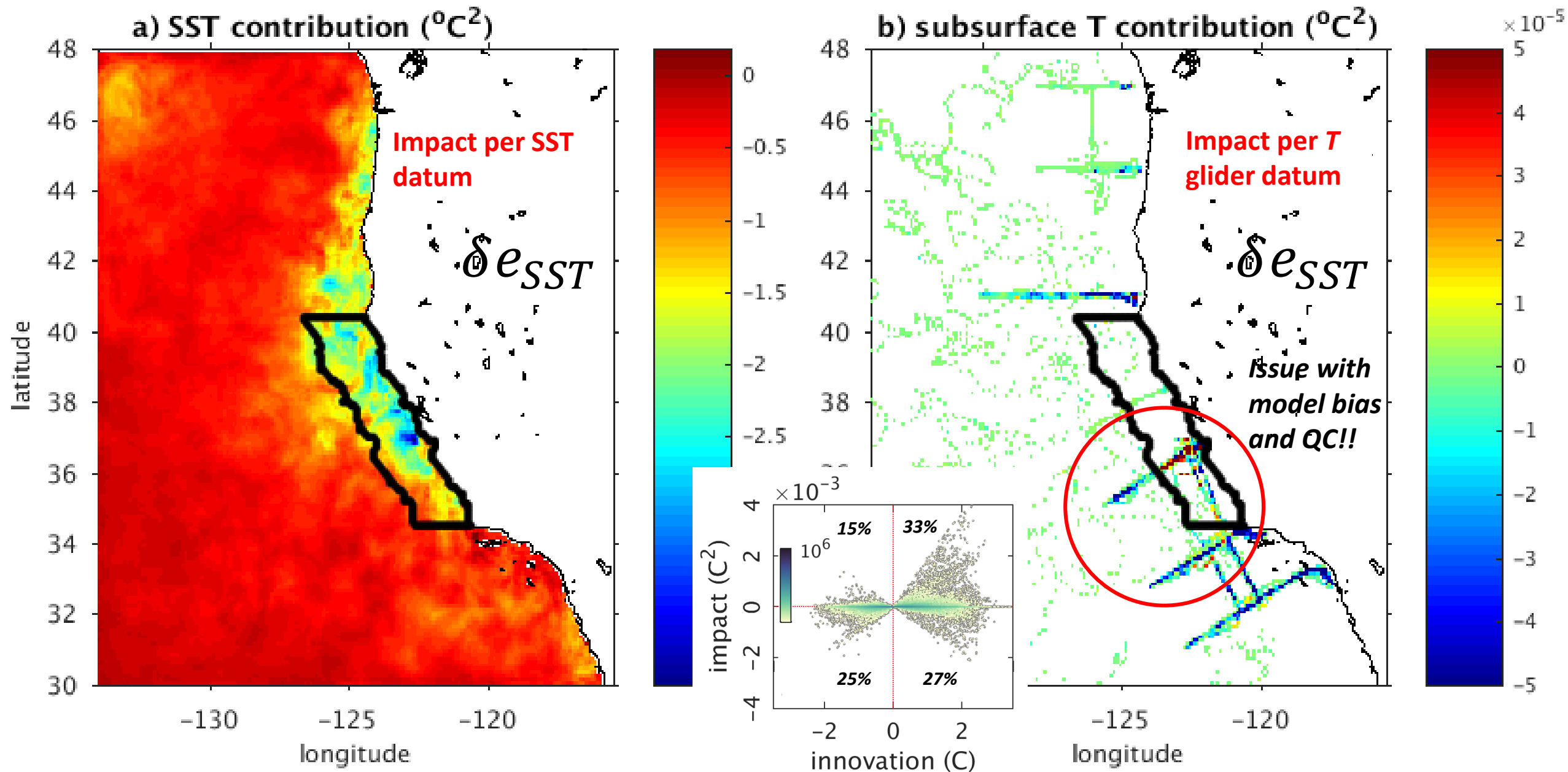
# Forecast Sensitivity to Observations (FSO)



Obs	$e_{SST}$	$e_v$	$e_{37N}$	$e_w$
SSH	56%	66%	62%	52%
T	38%	41%	40%	34%
S	38%	41%	41%	35%
HF radial	58%	65%	61%	51%
SST	63%	62%	58%	50%

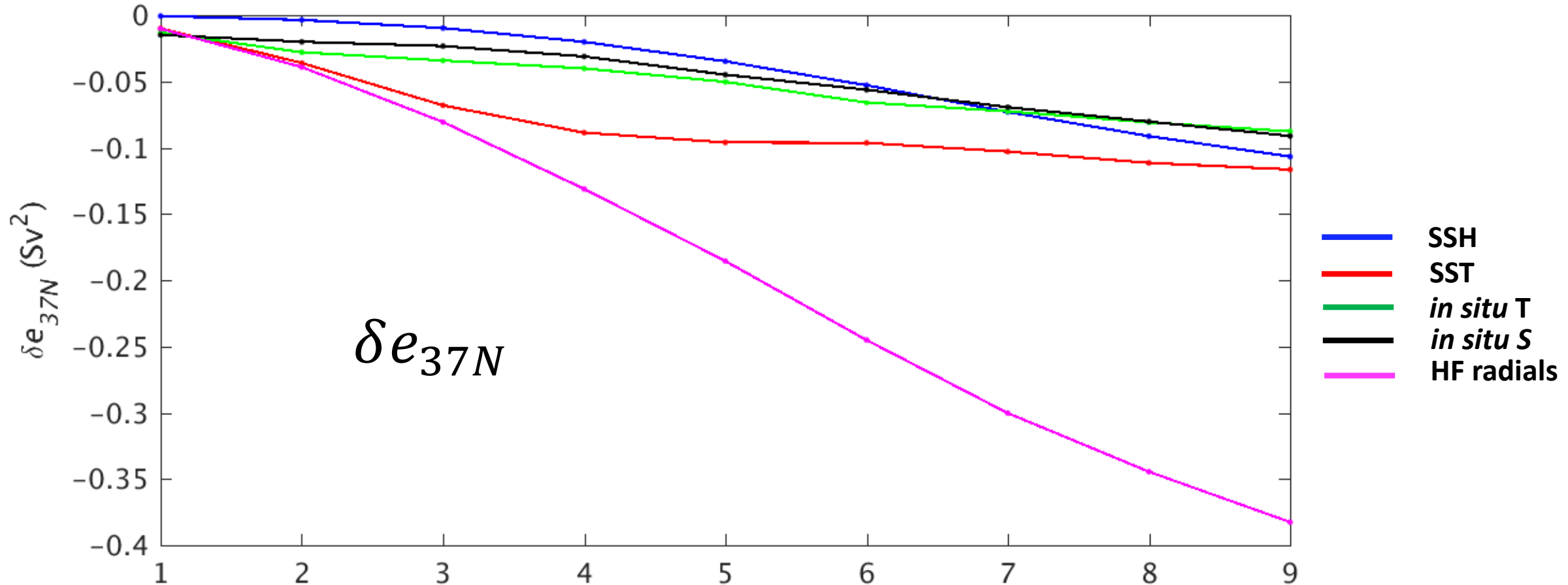
**Percentage obs that improve forecast skill**

# Observing System Monitoring





# System Performance



Mean impact per datum versus inner loop averaged over all analysis-forecast cycles

# Summary

- Only ~50% of obs improve forecast (agrees with experience in NWP)
- Should more observations should be assimilated (*cf* Gelaro *et al.*, 2010)?
- Can better use made of existing observations (*i.e.* can more info be extracted from some obs)?
- Data thinning required to reduce relative impact of high volume obs (*e.g.* HF radar, SST – *work in progress*)
- FSO is useful for monitoring observing system and performance of 4D-Var system
- FSO can be applied to *highly targeted* forecast metrics (*e.g.* drifter trajectories in support of search-and-rescue)