



Data Assimilation - 001. FSO & EFSO -

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with many thanks to Mrs. T. Saito and K. Fujimura



Observation Diagnosis

Observation Space Analysis

- ▶ Innovation statistics
- ▶ Degrees of freedom to signals

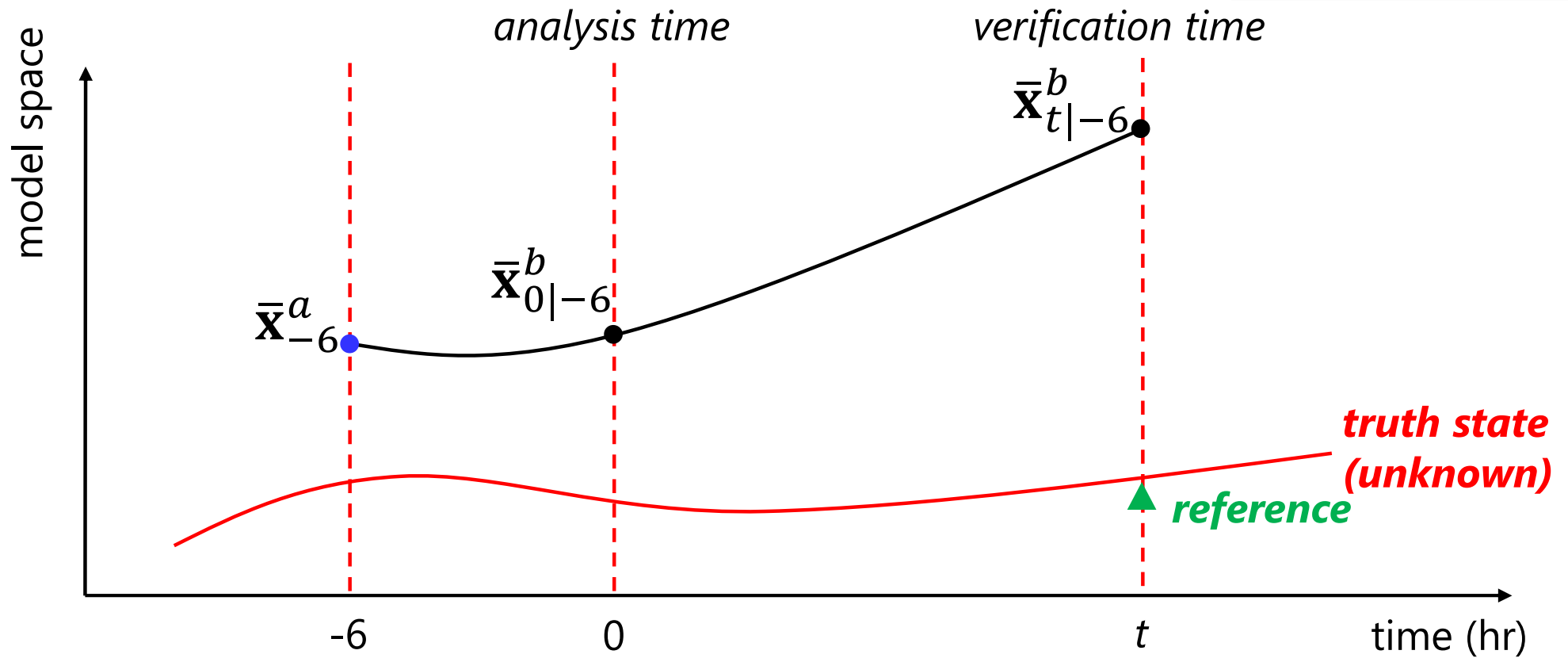
Forecast Sensitivity

		Forecast Sensitivity to Observation (FSO)	
		Adjoint	Ensemble
DA (iteration)	iterative	Langland and Baker (2004)	Buehner et al. (2018)
	deterministic	N/A	Kalnay et al. (2012)

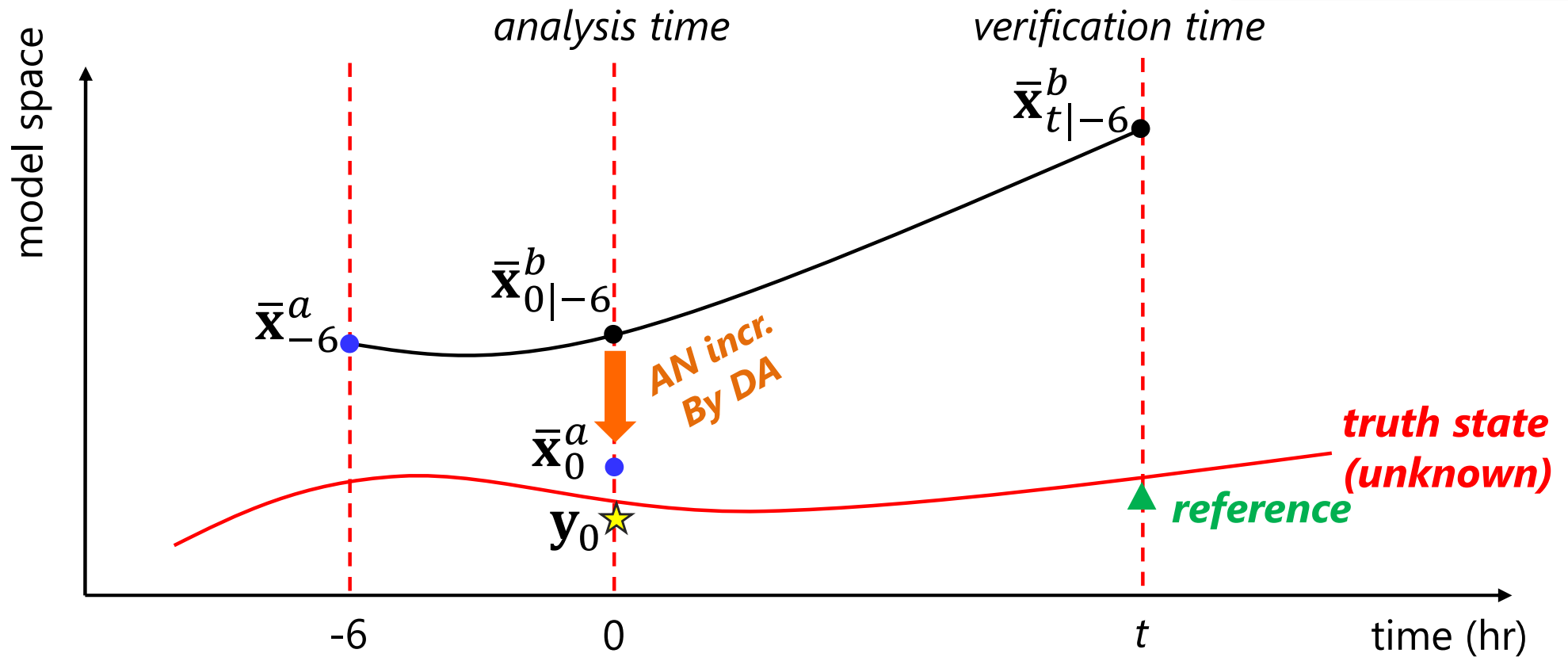
		Forecast Sensitivity to \mathbf{R} (FSR)	
		Adjoint	Ensemble
DA (iteration)	Adjoint	Daescu (2008)	N/A
	Ensemble	N/A	Hotta et al. (2017)

Forecast Sensitivity to Observation (FSO)

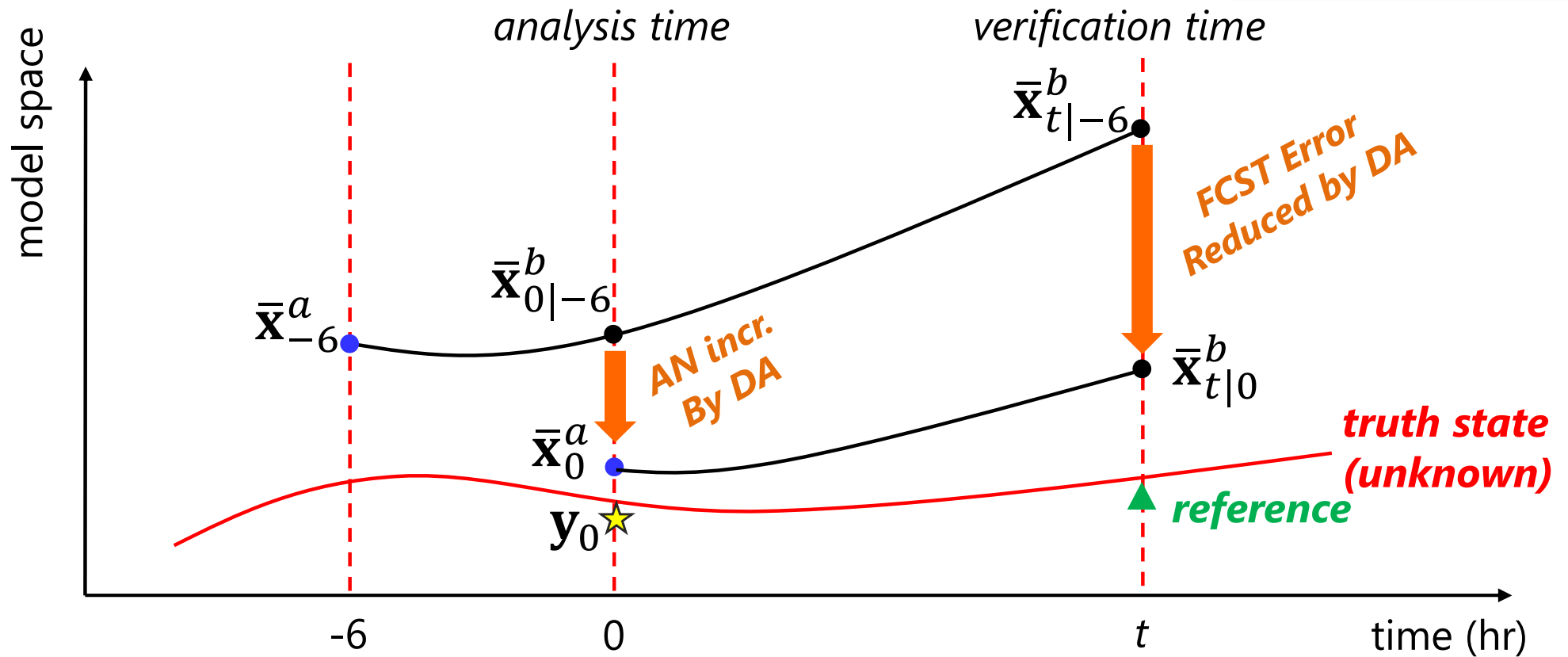
FSO: Forecast Sensitivity to Obs



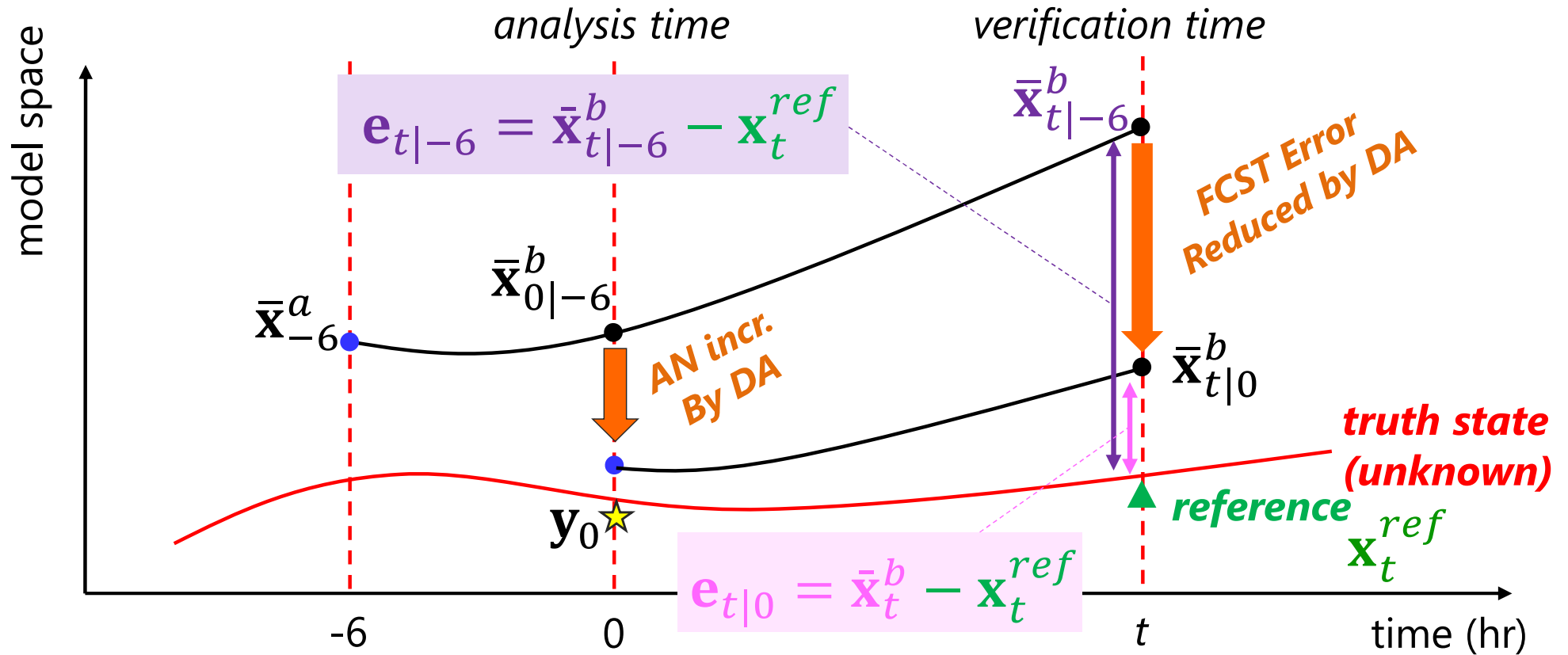
FSO: Forecast Sensitivity to Obs



FSO: Forecast Sensitivity to Obs



FSO: Forecast Sensitivity to Obs



The difference b/w $e_{t|0}$ and $e_{t|-6}$ depends only on the obs DA'ed at $t=0$.

FCST Error Reduction $J = \Delta e^2 = e_{t|0}^2 - e_{t|-6}^2 = \mathbf{e}_{t|0}^T C \mathbf{e}_{t|0} - \mathbf{e}_{t|-6}^T C \mathbf{e}_{t|-6}$

Error Reduction w.r.t. Obs $\frac{\partial J}{\partial \mathbf{y}} \in \mathbb{R}^p$

C : square norms
(e.g. L2 for L63/L96, dry/moist energy norm for NWP)

FCST Sensitivity to Obs (FSO)

$$\begin{aligned}
 J = \Delta e^2 &= \mathbf{e}_{t|0}^T \mathbf{C} \mathbf{e}_{t|0} - \mathbf{e}_{t|0}^T \mathbf{C} \mathbf{e}_{t|-6} \\
 &= (\mathbf{e}_{t|0} - \mathbf{e}_{t|-6})^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) \\
 &= (\bar{\mathbf{x}}_{t|0}^b - \bar{\mathbf{x}}_{t|-6}^b)^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) \\
 &\approx [\mathbf{M}(\bar{\mathbf{x}}_0^a - \bar{\mathbf{x}}_{0|-6}^b)]^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) \\
 &= (\mathbf{M}\mathbf{K}\delta\mathbf{y}_o)^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) \\
 &= \delta\mathbf{y}_o^T \mathbf{K}^T \mathbf{M}^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6})
 \end{aligned}$$

liner
error
growth
approx.

$$\mathbf{K} \in \mathbb{R}^{n \times p}, \mathbf{M} \in \mathbb{R}^{n \times n}$$

too large memory

needs adjoint (\mathbf{K}^T & \mathbf{M}^T)

$$\begin{aligned}
 \bar{\mathbf{x}}_0^a - \bar{\mathbf{x}}_{0|-6}^b &= \mathbf{K}(\mathbf{y} - H(\bar{\mathbf{x}}_{0|-6}^b)) \\
 \bar{\mathbf{x}}_0^a - \bar{\mathbf{x}}_{0|-6}^b &= \mathbf{K}\delta\mathbf{y}_o
 \end{aligned}$$

$$\delta\mathbf{y}_o \in \mathbb{R}^p$$

$$\mathbf{u} = \mathbf{K}^T \mathbf{M}^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) \in \mathbb{R}^p$$

$$\therefore \frac{\partial J}{\partial \mathbf{y}} \approx \mathbf{K}^T \mathbf{M}^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6})$$

$$J = \sum_{i=1}^p \delta y_{o,i} u_i$$

i.e., impact of i th obs is $\delta y_{o,i} u_i$

Ensemble FSO

$$\begin{aligned}
 J = \Delta e^2 &= \mathbf{e}_{t|0}^T \mathbf{C} \mathbf{e}_{t|0} - \mathbf{e}_{t|0}^T \mathbf{C} \mathbf{e}_{t|-6} \\
 &= (\mathbf{M} \mathbf{K} \delta \mathbf{y}_o)^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) \\
 &\approx \delta \mathbf{y}_o^T \mathbf{R}^{-1} \mathbf{Y}_0^a \mathbf{X}_{t|0}^{bT} \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) / (m - 1)
 \end{aligned}$$

$$\delta \mathbf{y}_o \in \mathbb{R}^p$$

$$\mathbf{u} = \mathbf{K}^T \mathbf{M}^T \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) \in \mathbb{R}^p$$

$$J = \sum_{i=1}^p \delta y_{o,i} u_i \quad \text{i.e., impact of } i\text{th obs is } \delta y_{o,i} u_i$$

$$\mathbf{M} \mathbf{K} \delta \mathbf{y}_o = \mathbf{M} \mathbf{A} \mathbf{H}^T \mathbf{R}^{-1} \delta \mathbf{y}_o$$

ensemble approx.

$$\mathbf{A} = \frac{1}{m - 1} \mathbf{X}_0^a (\mathbf{X}_0^a)^T$$

$$\approx \mathbf{M} \mathbf{X}_0^a (\mathbf{H} \mathbf{X}_0^a)^T \mathbf{R}^{-1} \delta \mathbf{y}_o / (m - 1)$$

$$= \mathbf{X}_{t|0}^b \mathbf{Y}_0^{aT} \mathbf{R}^{-1} \delta \mathbf{y}_o / (m - 1)$$

$$\delta \mathbf{Y}_0^a \equiv \mathbf{H} \delta \mathbf{X}_0^a$$

Experiments w/ Lorenz96

Kotsuki, S., Greybush, S., and Miyoshi, T. (2017):

Can we optimize the assimilation order in the serial ensemble Kalman filter?
A study with the Lorenz-96 model.

Mon. Wea. Rev., 145, 4977-4995. doi: 10.1175/MWR-D-17-0094.1

Experimental Setting

- ▶ Serial EnSRF (Whitaker and Hamill 2001)
 - ▶ Ensemble size : 8
 - ▶ # of observations : 40
 - ▶ Adaptive multiplicative inflation (Miyoshi 2011)

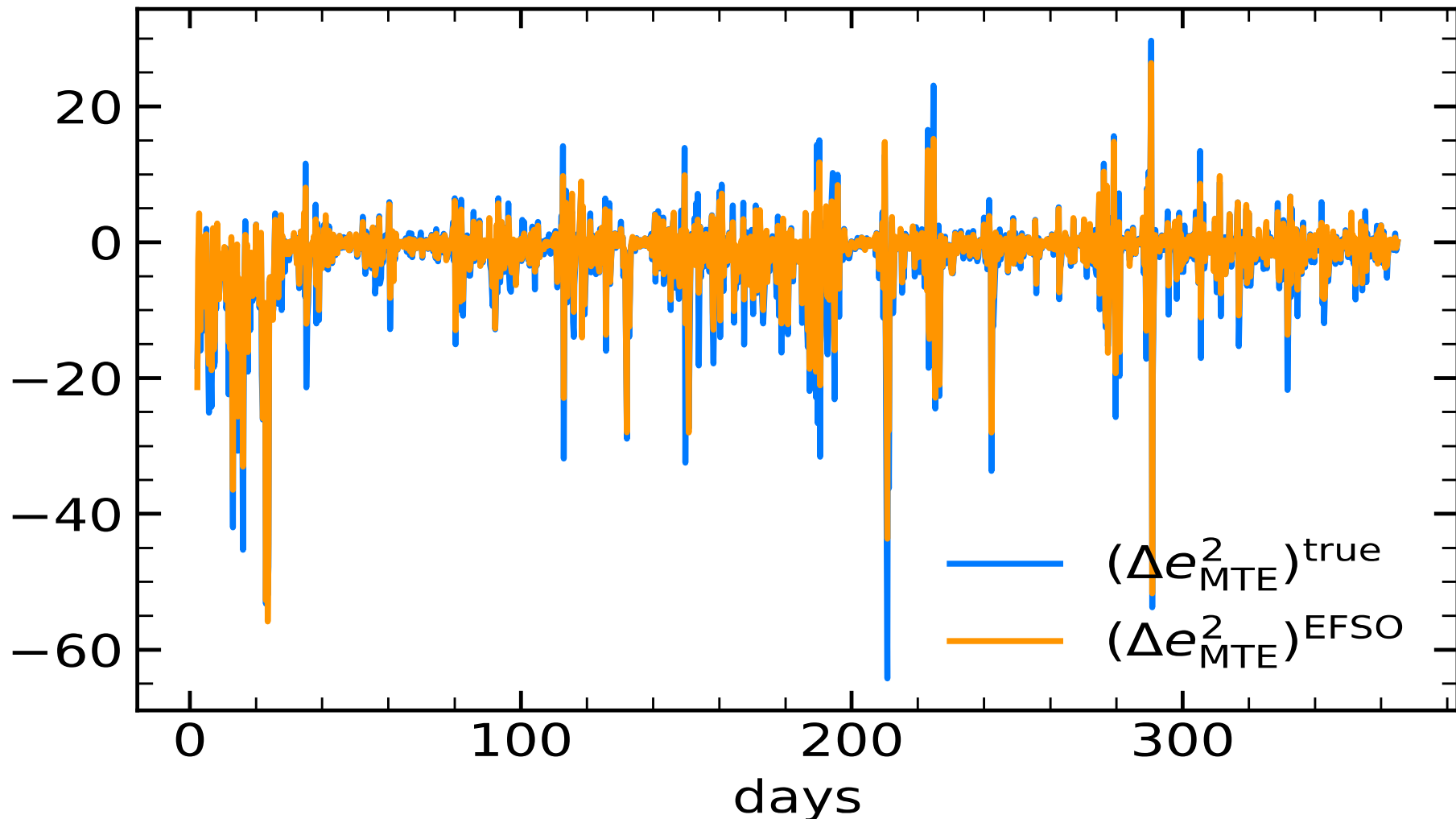
- ▶ Assimilation order
 1. Randomly-determined
 2. **From better to worse obs based on EFSO**

cf. Kotsuki et al. (2017, MWR)

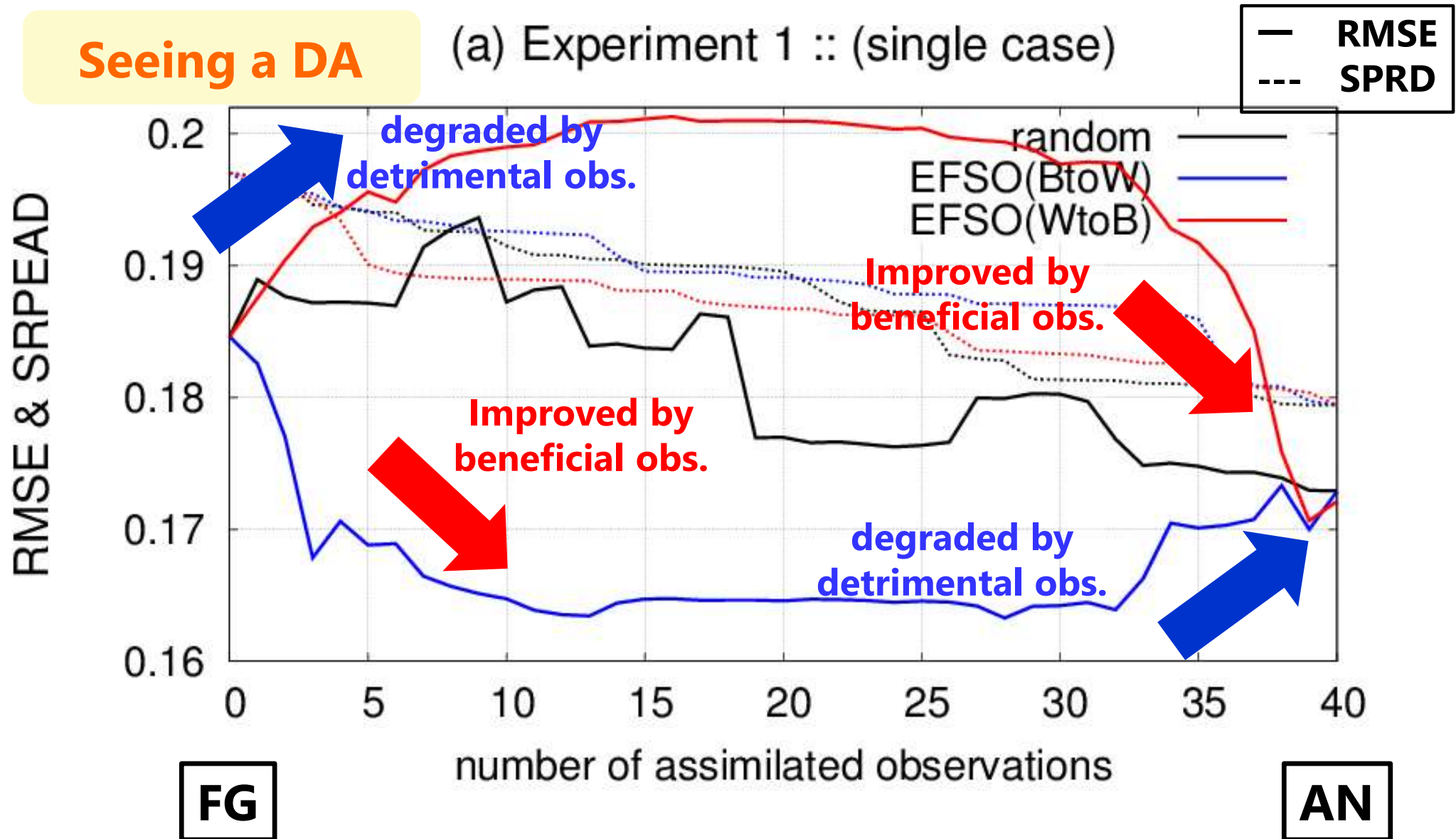
Debug

$$\Delta e_{true}^2 = \mathbf{e}_{t|0}^T \mathbf{C} \mathbf{e}_{t|0} - \mathbf{e}_{t|0}^T \mathbf{C} \mathbf{e}_{t|-6}$$

$$\Delta e_{EFSO}^2 = \delta \mathbf{y}_0^T \mathbf{R}^{-1} \mathbf{Y}_0^a \mathbf{X}_{t|0}^{bT} \mathbf{C} (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6}) / (m - 1)$$



A case w/ serial EnSRF

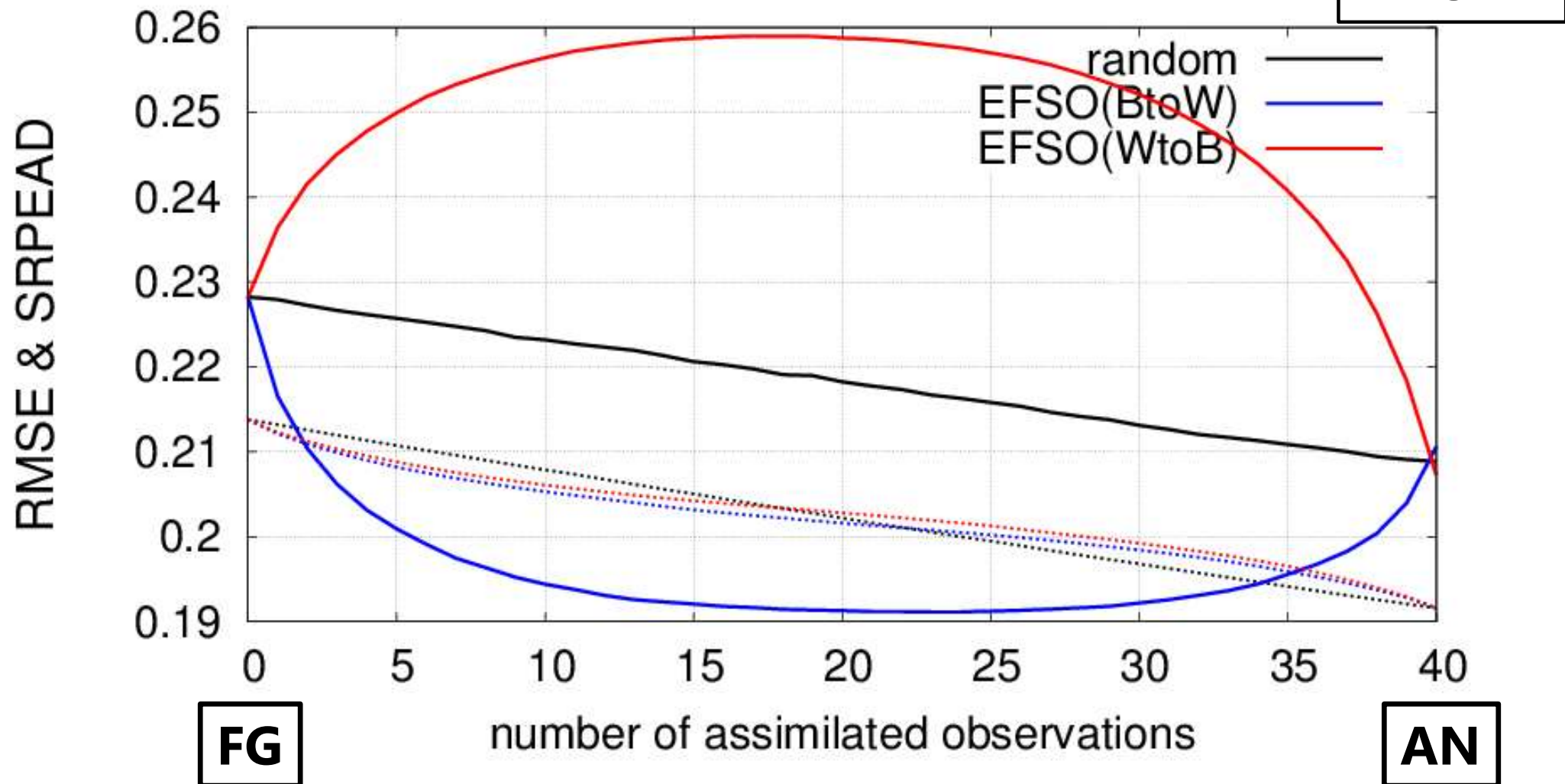


—: random order

—: EFSO better → worse obs.

Ave of 1460 cases

(c) Experiment 1 :: (ave. of 1460 cases)



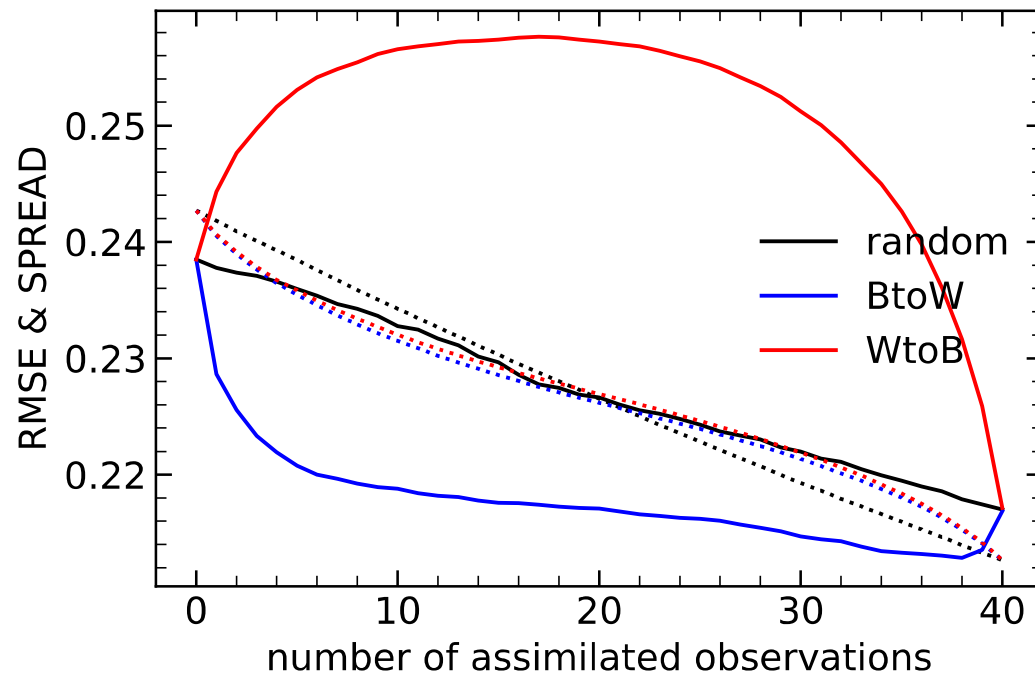
—: random order

—: EFSO better → worse obs.

—: EFSO worse → better obs.

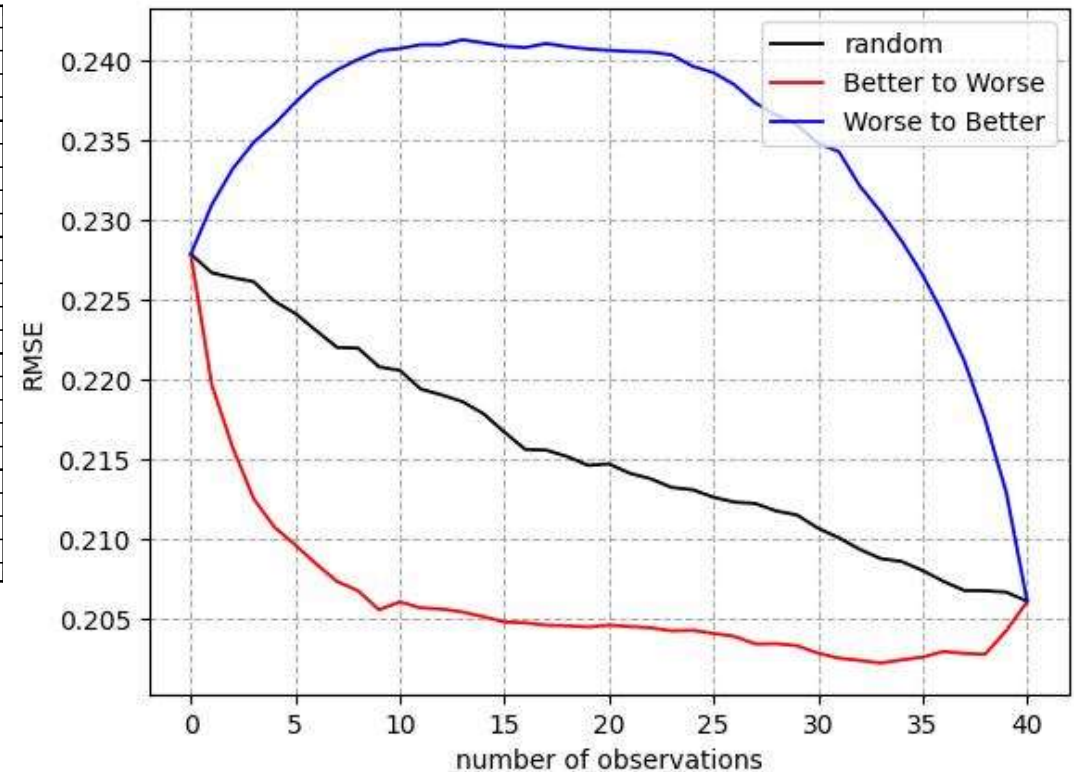
Replication by students

Fujimura



Saito

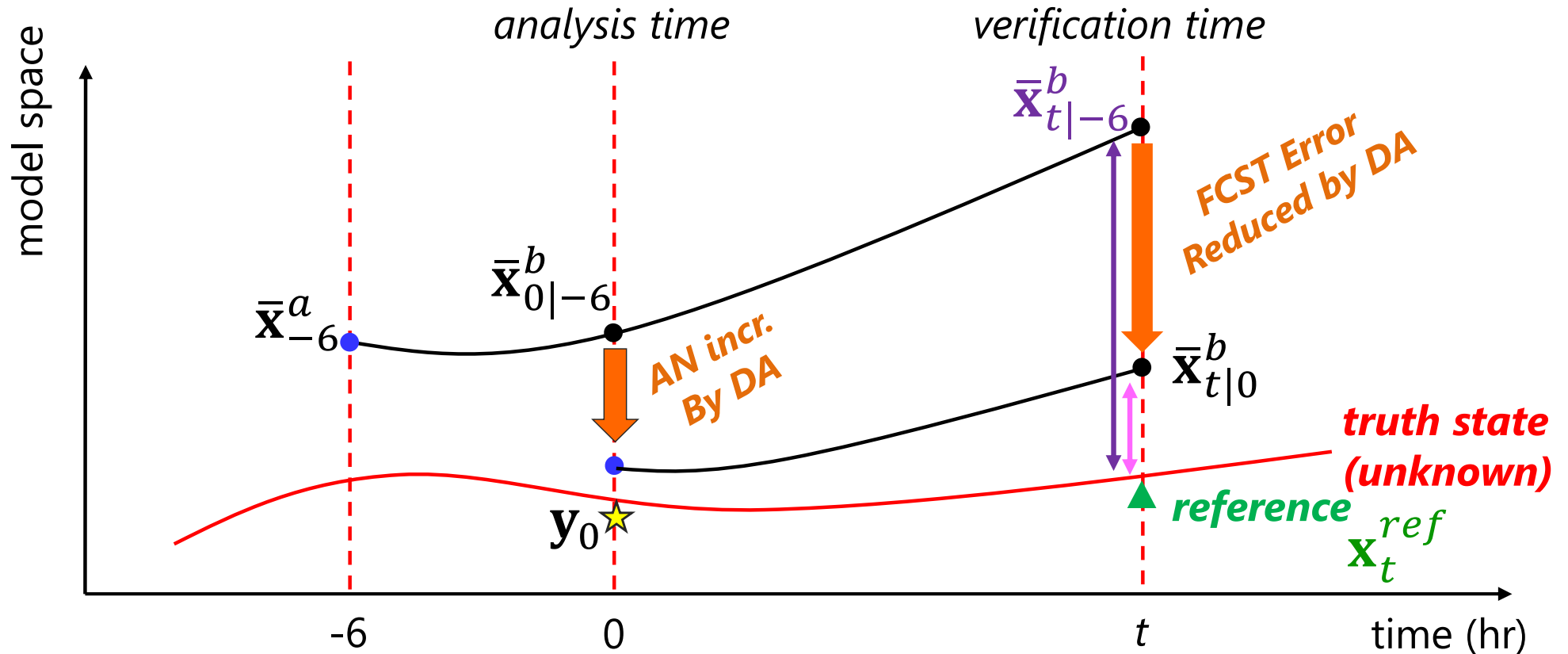
serial EnSRF



Experiments w/ NICAM-LETKF

Kotsuki, S.*, Kurosawa, K., and Miyoshi, T. (2019):
On the Properties of Ensemble Forecast Sensitivity to Observations.
Q. J. R. Meteorol. Soc., 145, 1897-1914. doi: 10.1002/qj.3534

Reference & Norms



Moist Total Energy for Error Norm

$$\Delta e_{MTE}^2 = (\mathbf{e}_{t|0}^T \mathbf{C} \mathbf{e}_{t|0} - \mathbf{e}_{t|-6}^T \mathbf{C} \mathbf{e}_{t|-6}) / 2$$

Verified against subsequent Anl.,
or Anl from other center

$$\mathbf{e}_t = \bar{\mathbf{x}}_t^b - \mathbf{x}_t^{ref}$$

Normalized Obs. Departure for Error Norm

$$\Delta e_{NOD}^2 = (\mathbf{d}_{t|0}^T \mathbf{R}_t^{-1} \mathbf{d}_{t|0} - \mathbf{d}_{t|-6}^T \mathbf{R}_t^{-1} \mathbf{d}_{t|-6}) / p$$

Verified against future obs

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

Ensemble FSO

Moist Total Energy for Error Norm (Ota et al. 2013)

$$\Delta e_{MTE}^2 = (\mathbf{e}_{t|0}^T C \mathbf{e}_{t|0} - \mathbf{e}_{t|-6}^T C \mathbf{e}_{t|-6}) / 2$$

$$\mathbf{e}_t = \bar{\mathbf{x}}_t^b - \mathbf{x}_t^{ref}$$

$$\approx \frac{1}{2} \frac{1}{m-1} \delta \mathbf{y}_0^T \mathbf{R}^{-1} \mathbf{Y}_0^a \mathbf{X}_{t|0}^{bT} C (\mathbf{e}_{t|0} + \mathbf{e}_{t|-6})$$

obs-minus-FG AN ptb FCST ptb
in obs space

Normalized Obs. Departure for Error Norm (Sommer and Weissmann 2016)

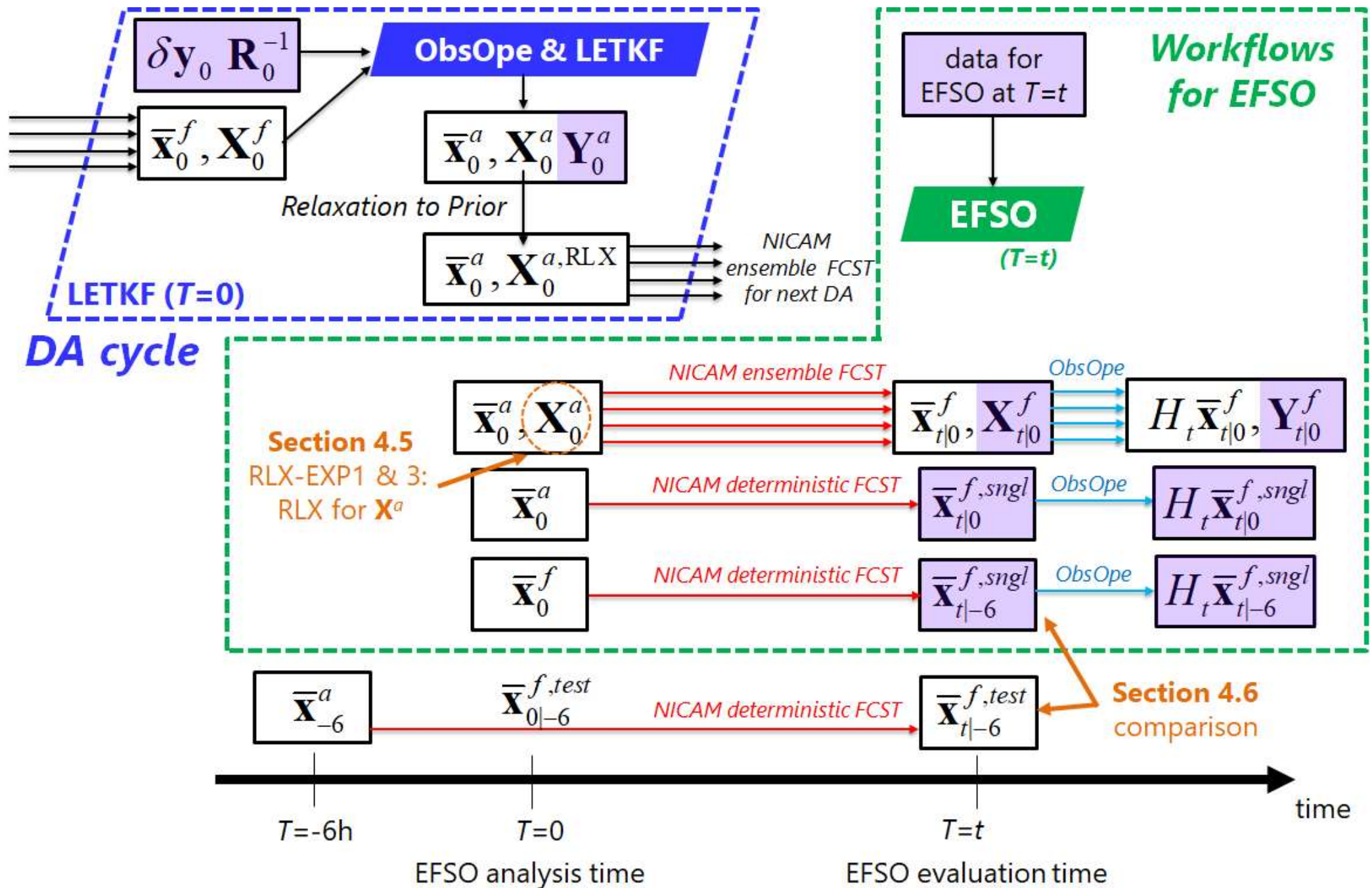
$$\Delta e_{NOD}^2 = (\mathbf{d}_{t|0}^T \mathbf{R}_t^{-1} \mathbf{d}_{t|0} - \mathbf{d}_{t|-6}^T \mathbf{R}_t^{-1} \mathbf{d}_{t|-6}) / p$$

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

$$\approx \frac{1}{p} \frac{1}{m-1} \delta \mathbf{y}_0^T \mathbf{R}^{-1} \mathbf{Y}_0^a \mathbf{Y}_{t|0}^{bT} \mathbf{R}_t^{-1} (\mathbf{d}_{t|0} + \mathbf{d}_{t|-6})$$

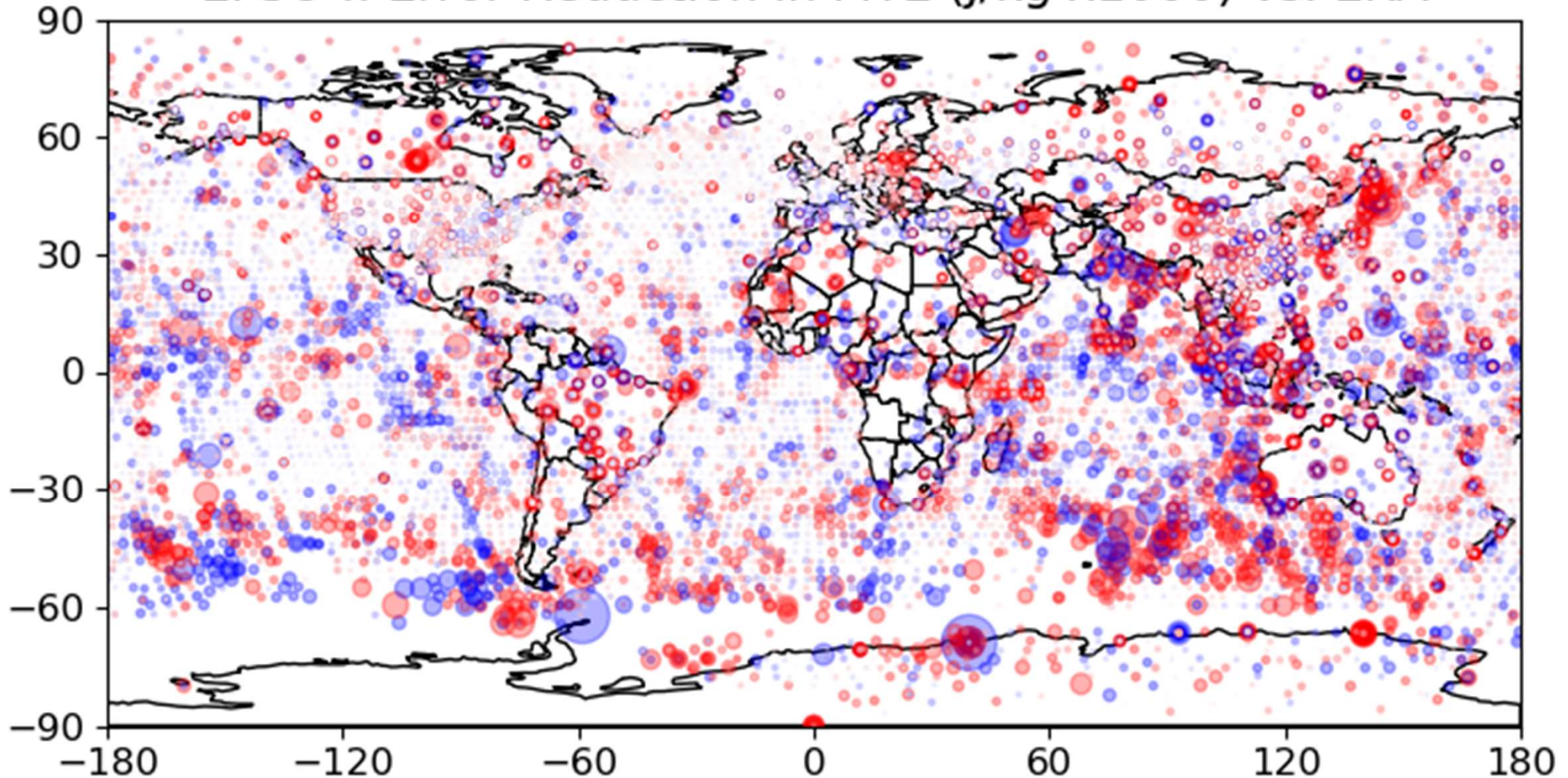
obs-minus-FG AN ptb FCST ptb
in obs space

EFSO with NICAM-LETKF



Estimated Impacts by EFSO

EFSO :: Error Reduction in MTE (J/kg x1000) vs. ERA



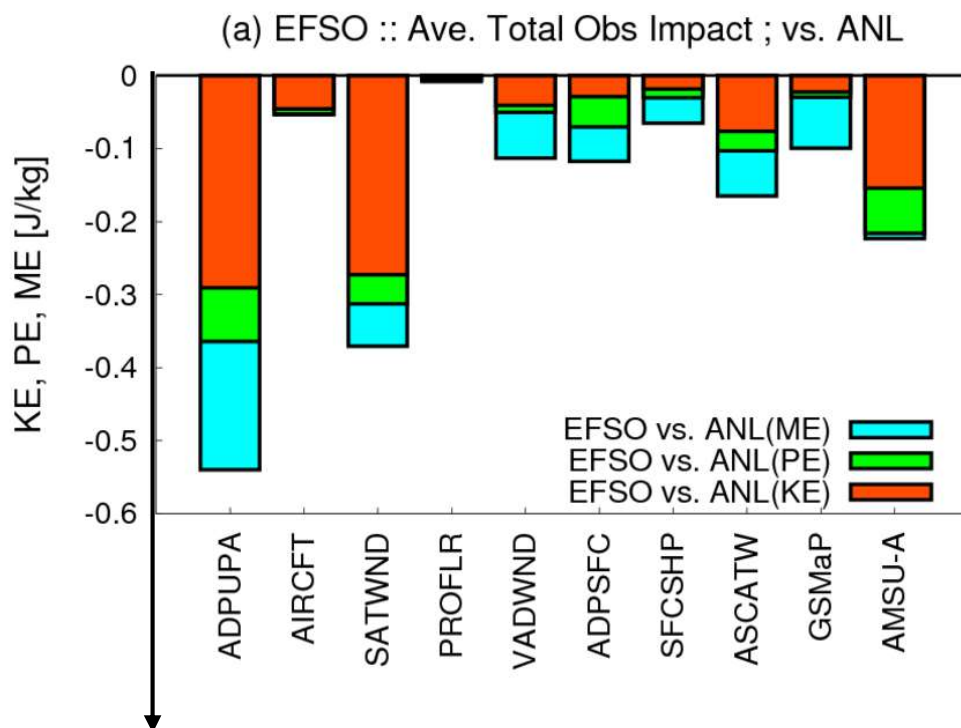
● ***Beneficial observation***

● ***Detrimental observation***

2014/07/11/00UTC; vs. ERA interim

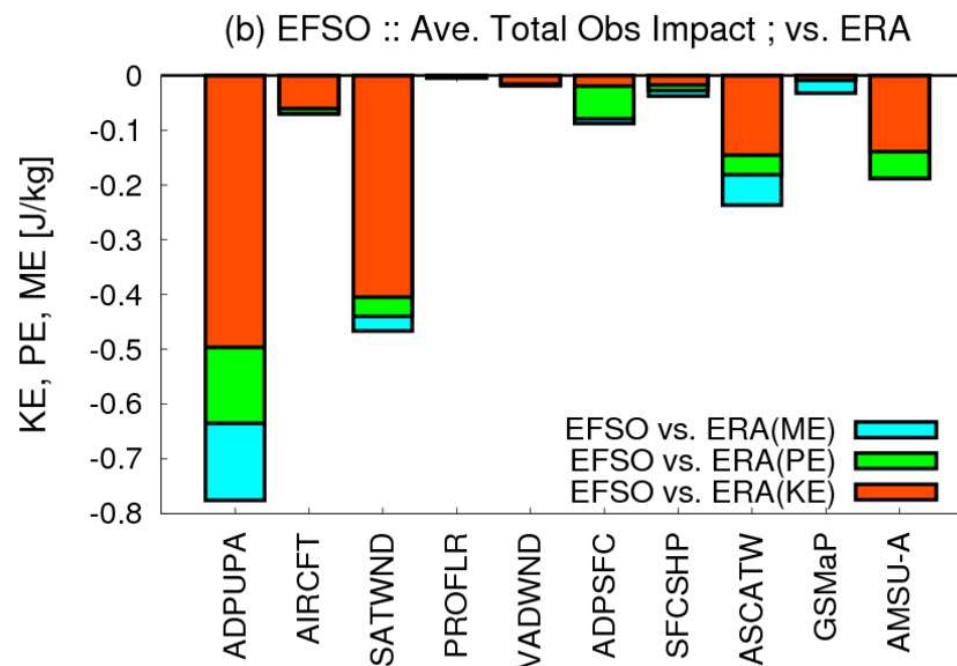
FCST **MTE** Error Reduction

vs. NICAM-LETKF



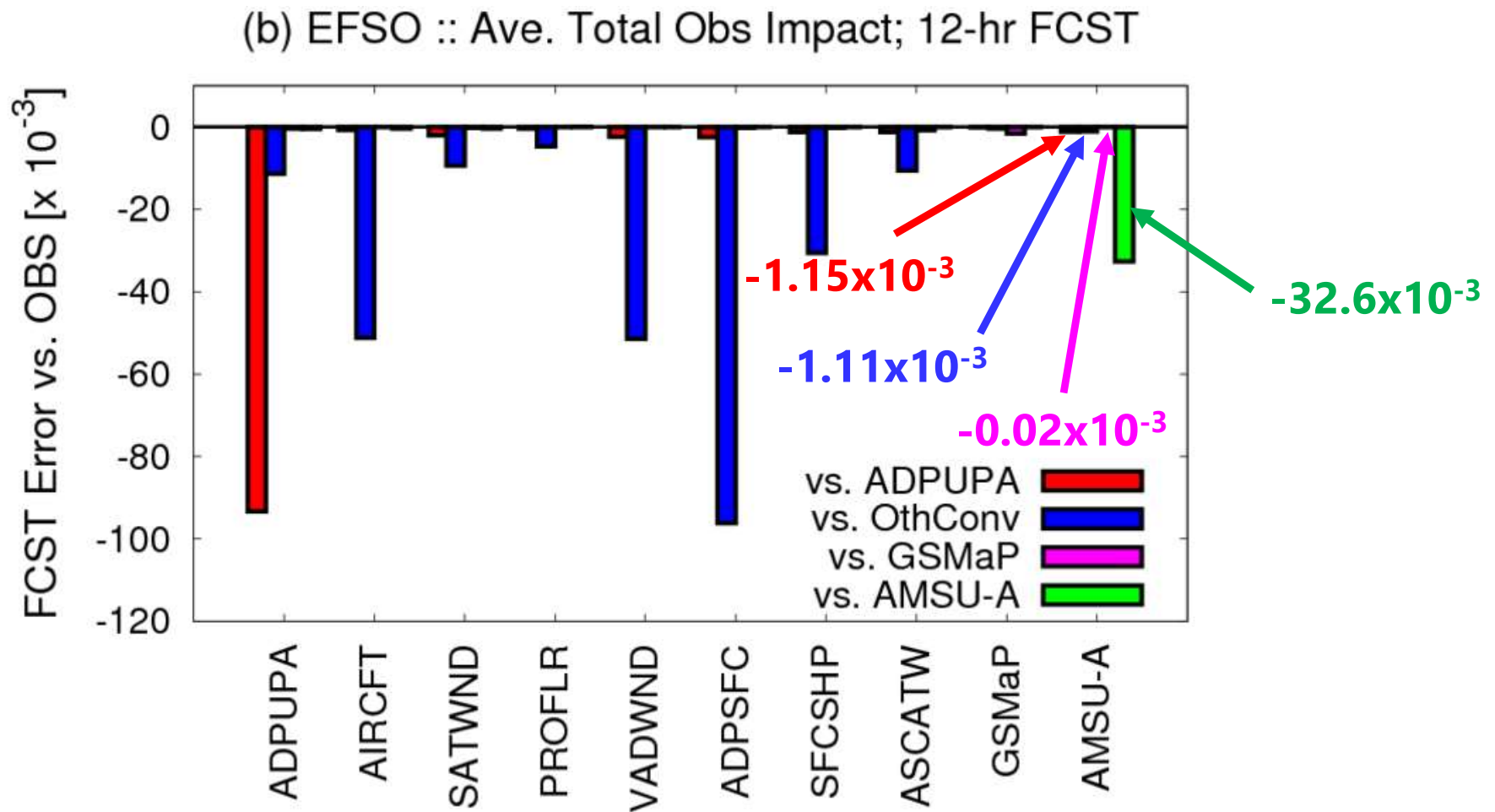
Positive Impact
(improving 6-hr FCST)

vs. ERA Interim



ME : Moist Energy
PE : Potential Energy
KE : Kinetic Energy

FCST NOD Error Reduction



Each type of observations mainly contributes to the improvement of the observed variable.

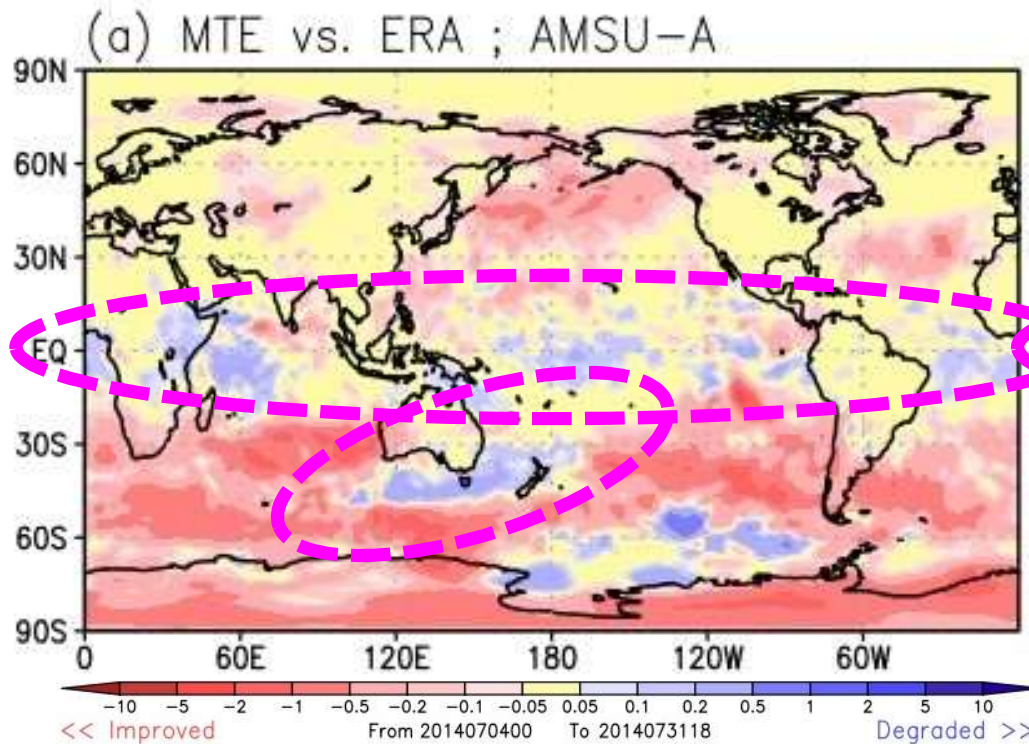
Monthly average in July 2014; FT 12hr

Detection of detrimental AMSU-A radiances

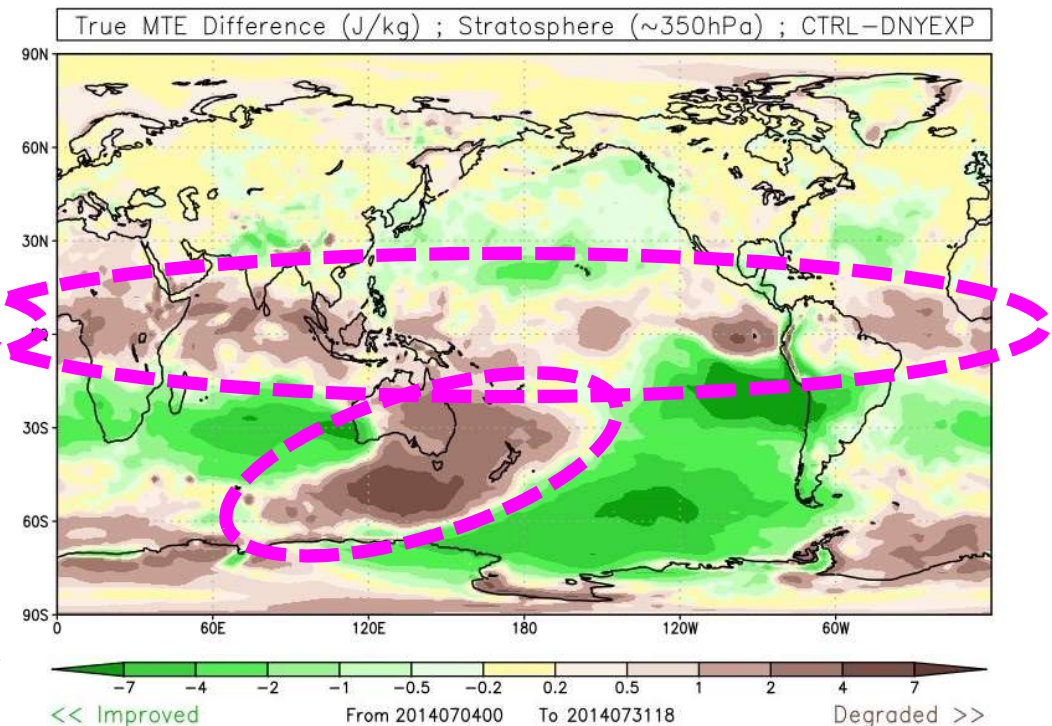
Impacts of AMSU-A on upper atmos. ($\sim 350\text{hPa}$, vs. ERA)

EFSO Estimates

w/wo AMSU-A



detrimental →



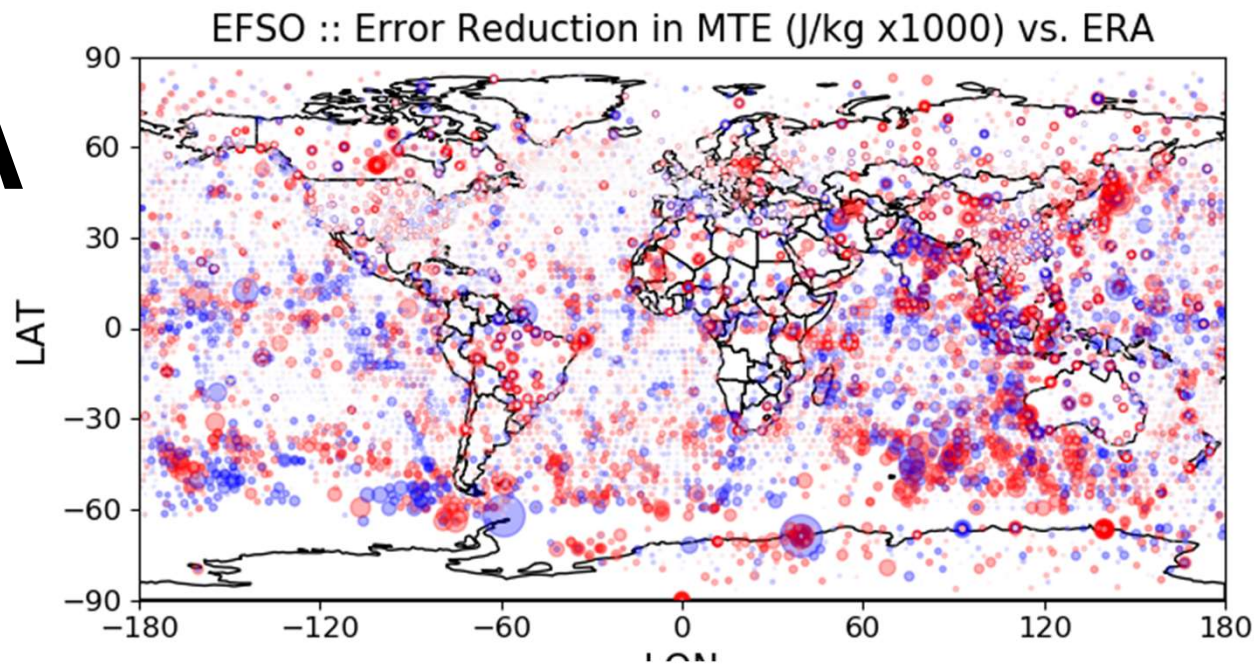
detrimental →

We can evaluate impacts of AMSU-A by EFSO without expensive w/wo OSE experiments

On Beneficial & Detrimental Obs

FCST Error Reduction (2014071100UTC)

vs. ERA

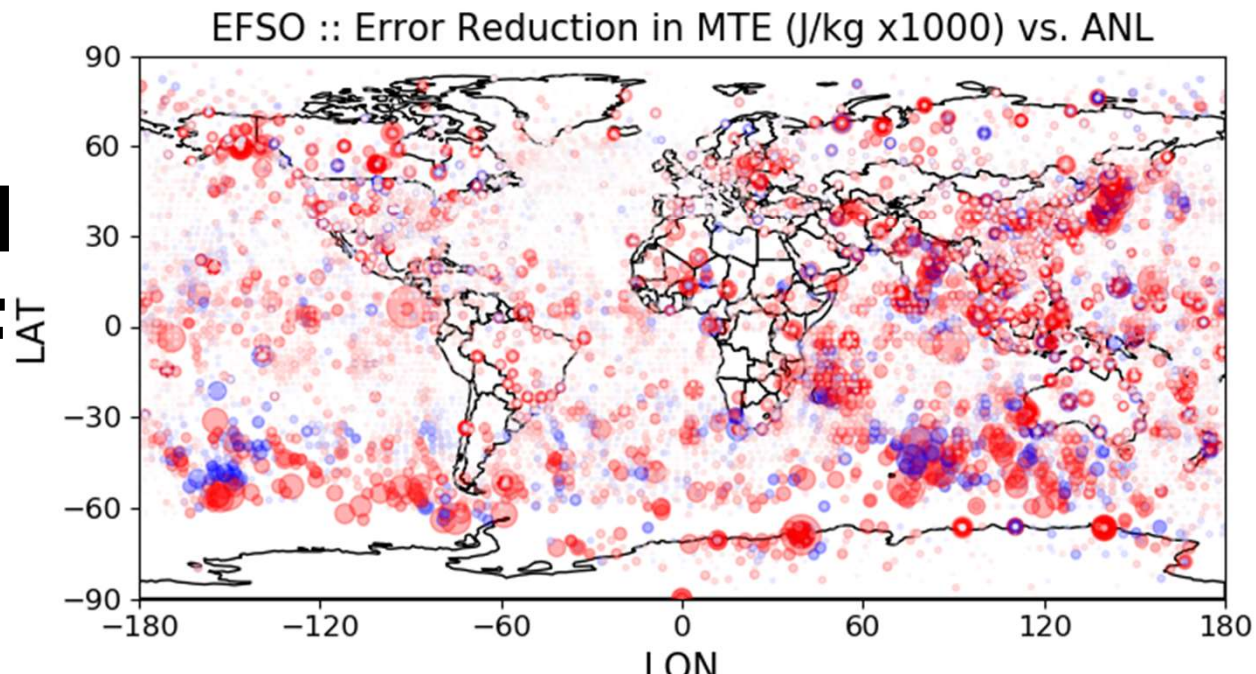


FCST degraded

FT 06hr

FCST improved

**vs.
NICAM
-LETKF**



FCST degraded

FT 06hr

FCST improved

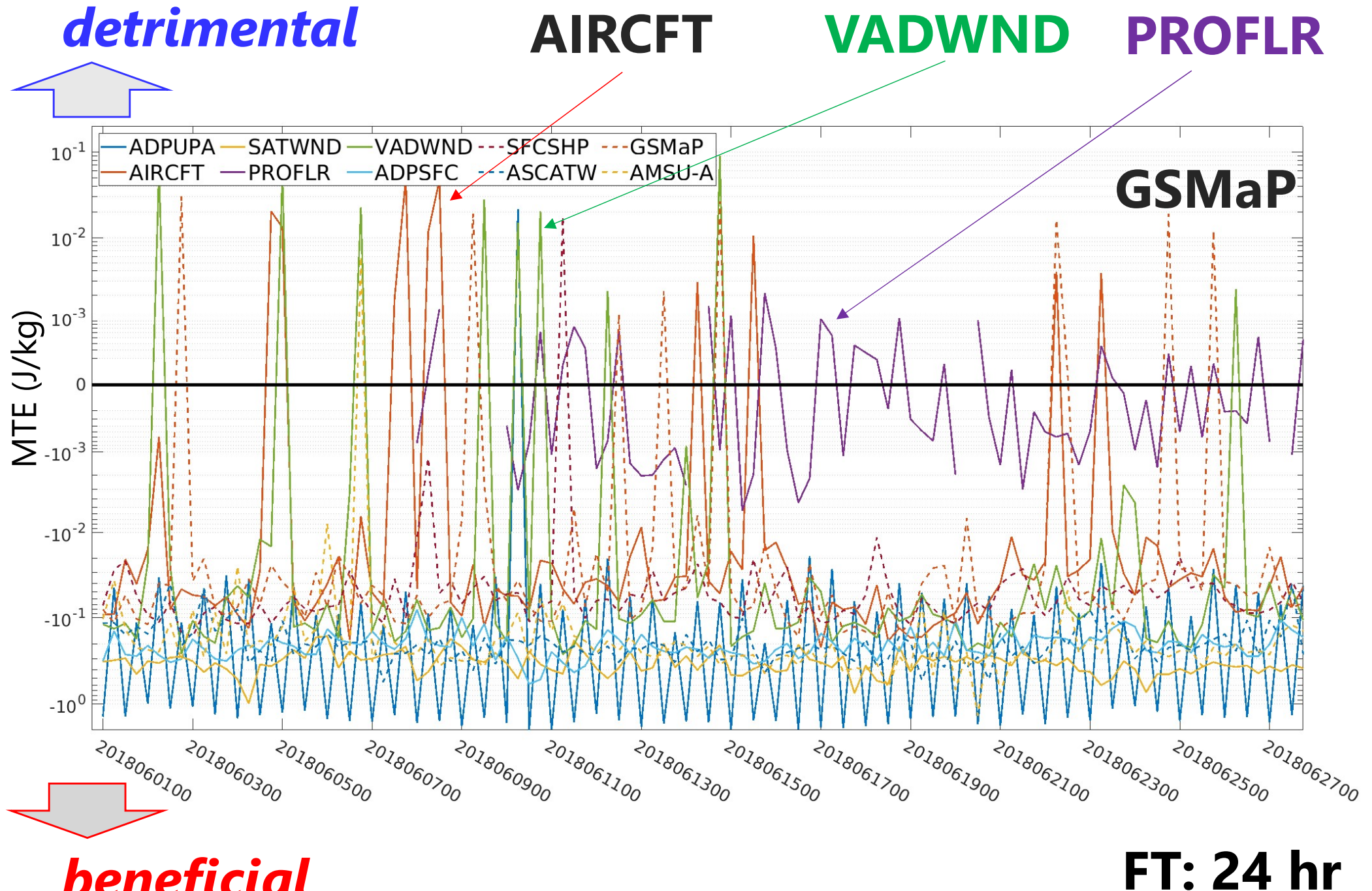
Fraction of Beneficial Obs

Ref.	Subseq. AN	ERA Interim	AMSU-A
FT 06hr	<u>58.8 %</u>	55.4 %	53.1 %
FT 12hr	<u>56.1 %</u>	54.2 %	53.2 %

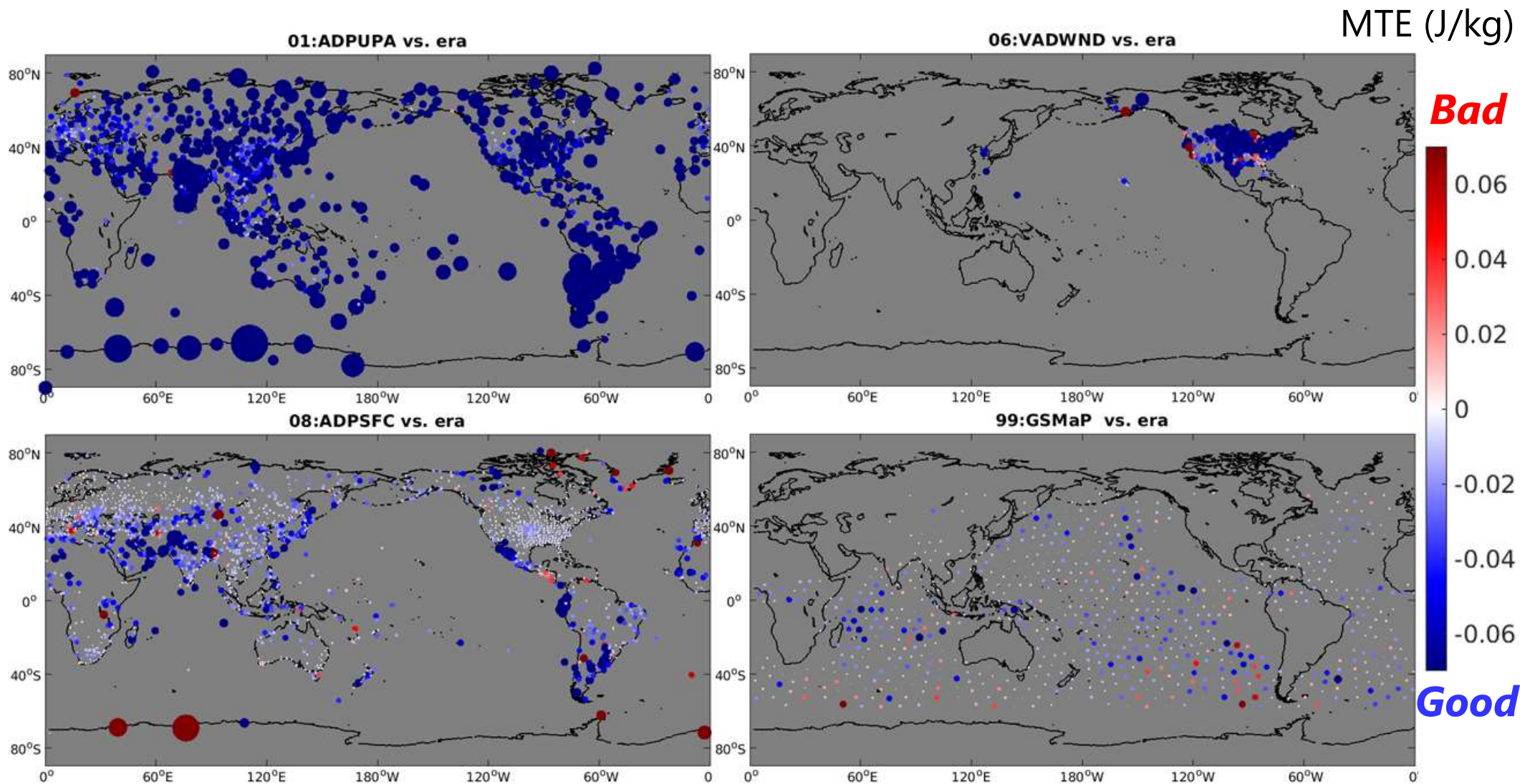
FSO may overestimate observational impact with subsequent analyses for verification reference.

→ It is good to use an independent anl for reference

Impacts in Time (vs. ERA Interim)



Impacts in Space (vs. ERA Interim)



統計的に予報を改悪する観測を除くと良くなるのか？（課題）

FT: 24hr

SAMPLE: 2018060100-2018063018

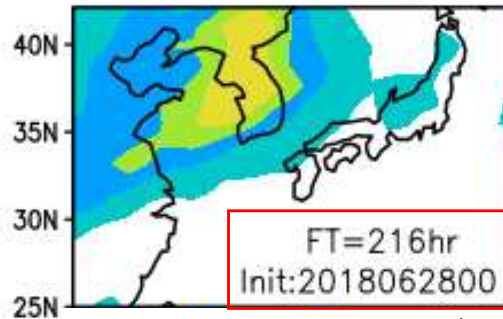
A Case of Heavy Rain in July 2018

Kotsuki, S.*, Terasaki, K., Kanemaru, K., Satoh, M., Kubota, T. and Miyoshi, T. (2019):
Predictability of Record-Breaking Rainfall in Japan in July 2018:
SOLA, 15A, 1-7. doi: 10.2151/sola.15A-001

NICAM-LETKF Ens. FCSTs

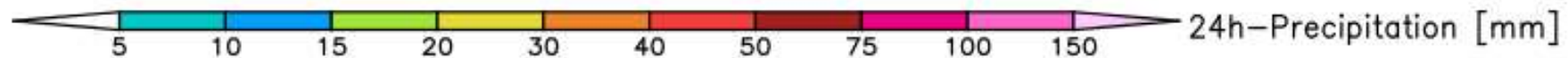
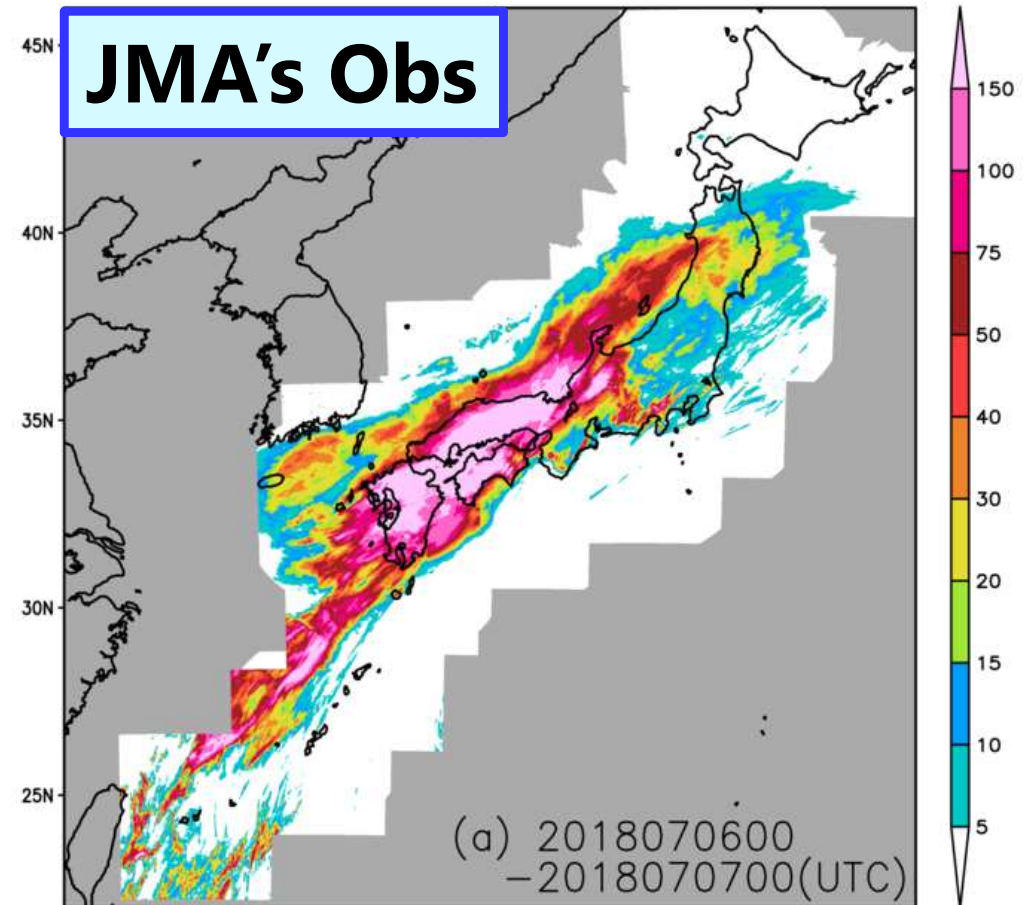
NEXRA Precip. FCSTs (100-Ensemble)

Period:2018070600-2018070700(UTC)



Forecast Lead Time (F⁻
& Initialized Date

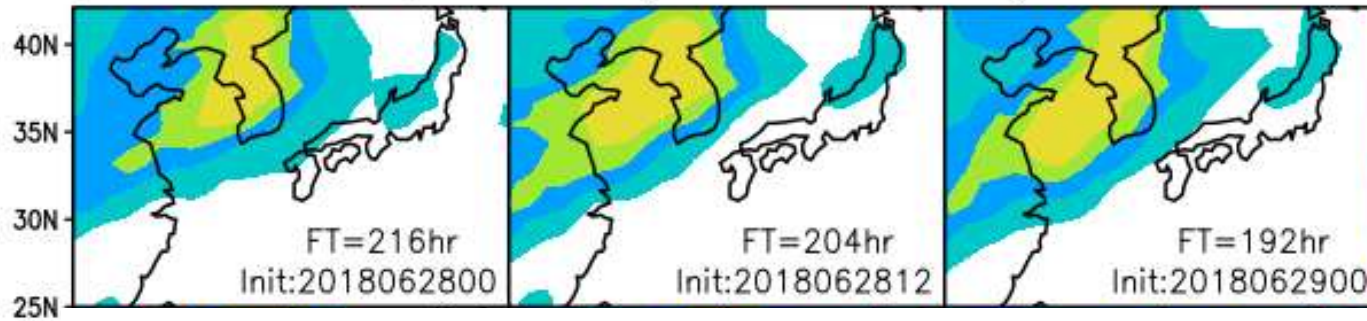
24-hr Accumulated Precip



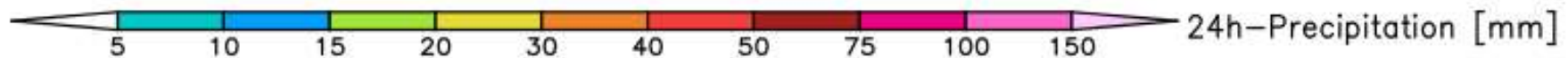
NICAM-LETKF Ens. FCSTs

NEXRA Precip. FCSTs (100-Ensemble)

Period:201870600-201870700(UTC)



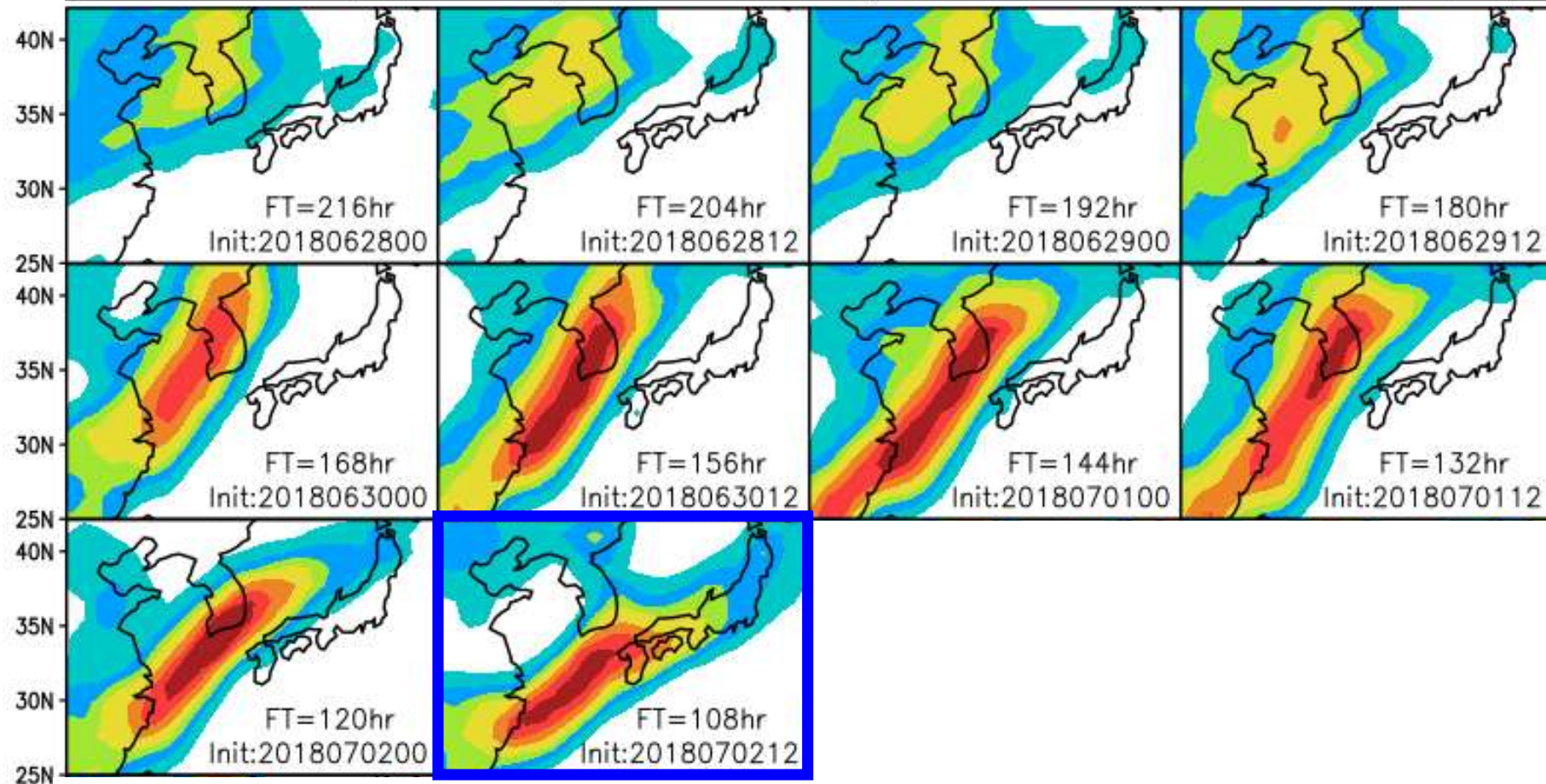
shifting initialized time



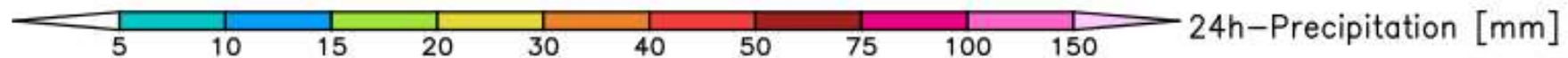
NICAM-LETKF Ens. FCSTs

NEXRA Precip. FCSTs (100-Ensemble)

Period:201870600-201870700(UTC)



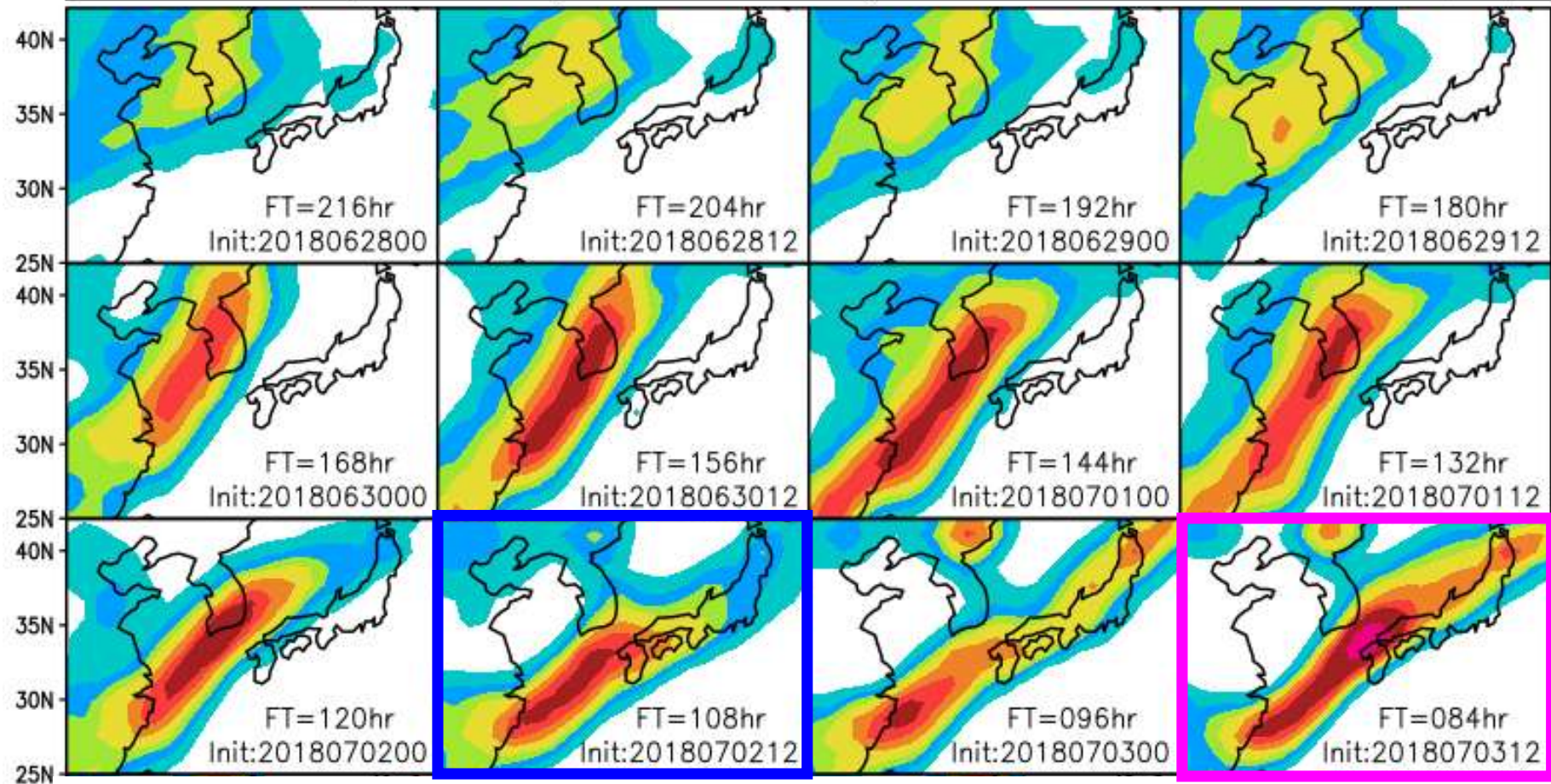
First Turning Point



NICAM-LETKF Ens. FCSTs

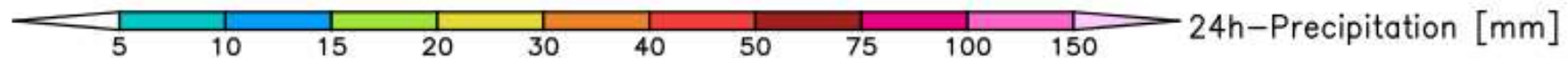
NEXRA Precip. FCSTs (100-Ensemble)

Period:201870600-201870700(UTC)



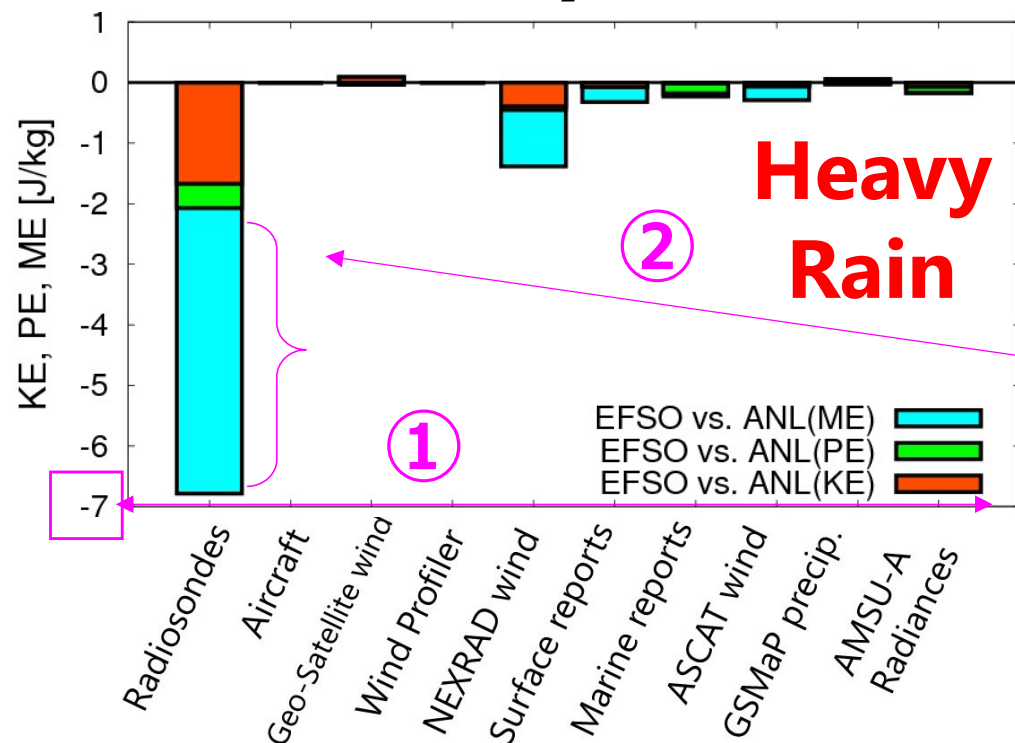
First Turning Point

Second Turning Point

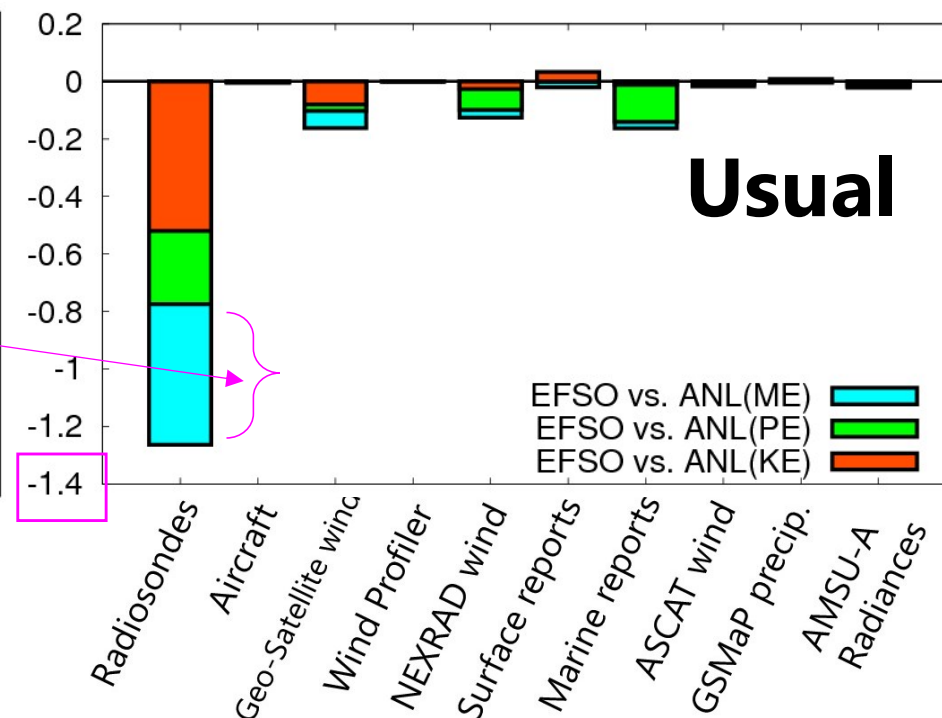


Obs Impacts

Obs at 07/03 12UTC



Ave 4/01~4/07



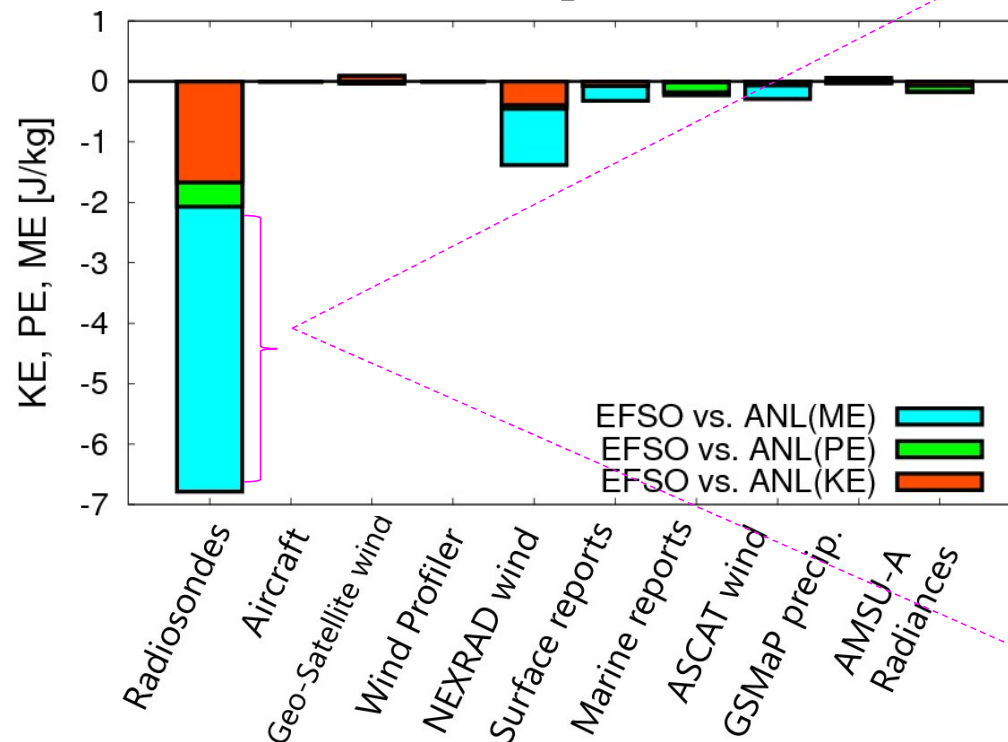
- ① Large impacts than usual → rapid error growth
② Impacts on moist field is bigger

■ ME : Moist Energy
■ PE : Potential Energy
■ KE : Kinetic Energy

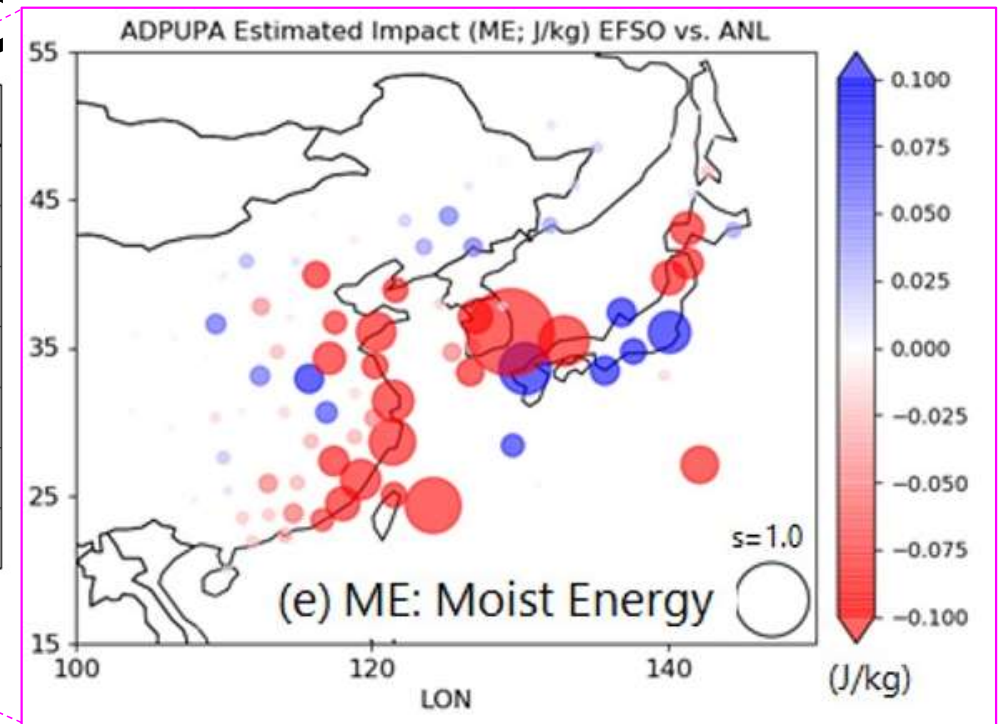
(水蒸気の改善)
(気温・気圧の改善)
(風向風速の改善)

Radiosondes' Impacts

Obs at 07/03 12UTC



●: beneficial radiosondes
●: detrimental radiosondes



■ ME : Moist Energy
■ PE : Potential Energy
■ KE : Kinetic Energy

(水蒸気の改善)
(気温・気圧の改善)
(風向風速の改善)

Thank you for your attention!

Presented by Shunji Kotsuki
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Further information is available at
<https://kotsuki-lab.com/>

