



# Validating uncertainty estimates

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# So your retrieval has an uncertainty estimate...

- How do we know if the uncertainty estimate is useful or realistic?



Statistics!

# Reminder: uncertainty and error are different

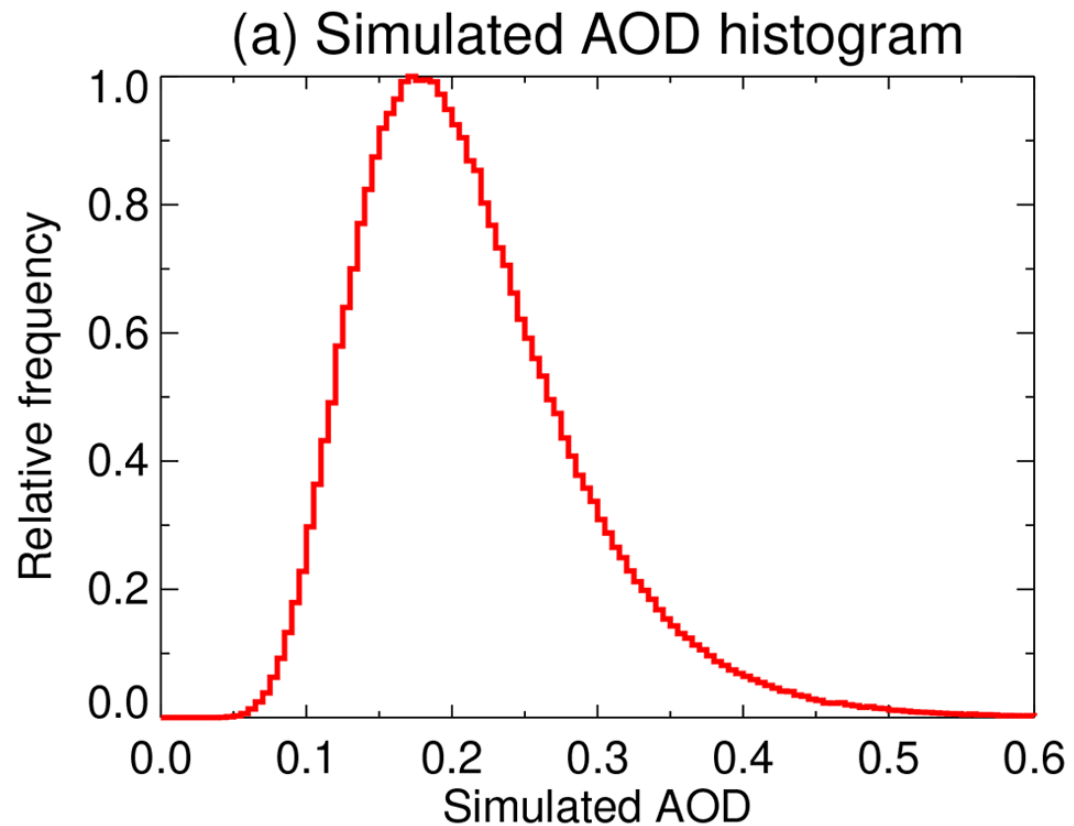
- Uncertainty is an *expected level of dispersion*
  - Typically one standard deviation ( $1\sigma$ , ~68% of the time as good as X)
- Error is an observed quantity
  - *A draw from the distribution* of possible errors given the uncertainty level



You *expect* 3.5  
You *get* 1, 2, 3, 4, 5, or 6

# Uncertainty distributions and error distributions

- Suppose AOD uncertainty is  $\pm(0.05+15\%)$ . Then:



Sayer *et al.*, AMT (2020)

# Solution: compare distributions of observed errors to expected discrepancy (ED)

- Observed error = (truth – retrieval)
- Expected discrepancy contains contributions from:
  - Retrieval uncertainty  $\sigma_{\text{sat}}$
  - Reference uncertainty  $\sigma_{\text{ref}}$
  - Spatiotemporal variability inherent in the comparison  $\sigma_{\text{var}}$
- If assumed independent, then:

**Propagated or simulated**

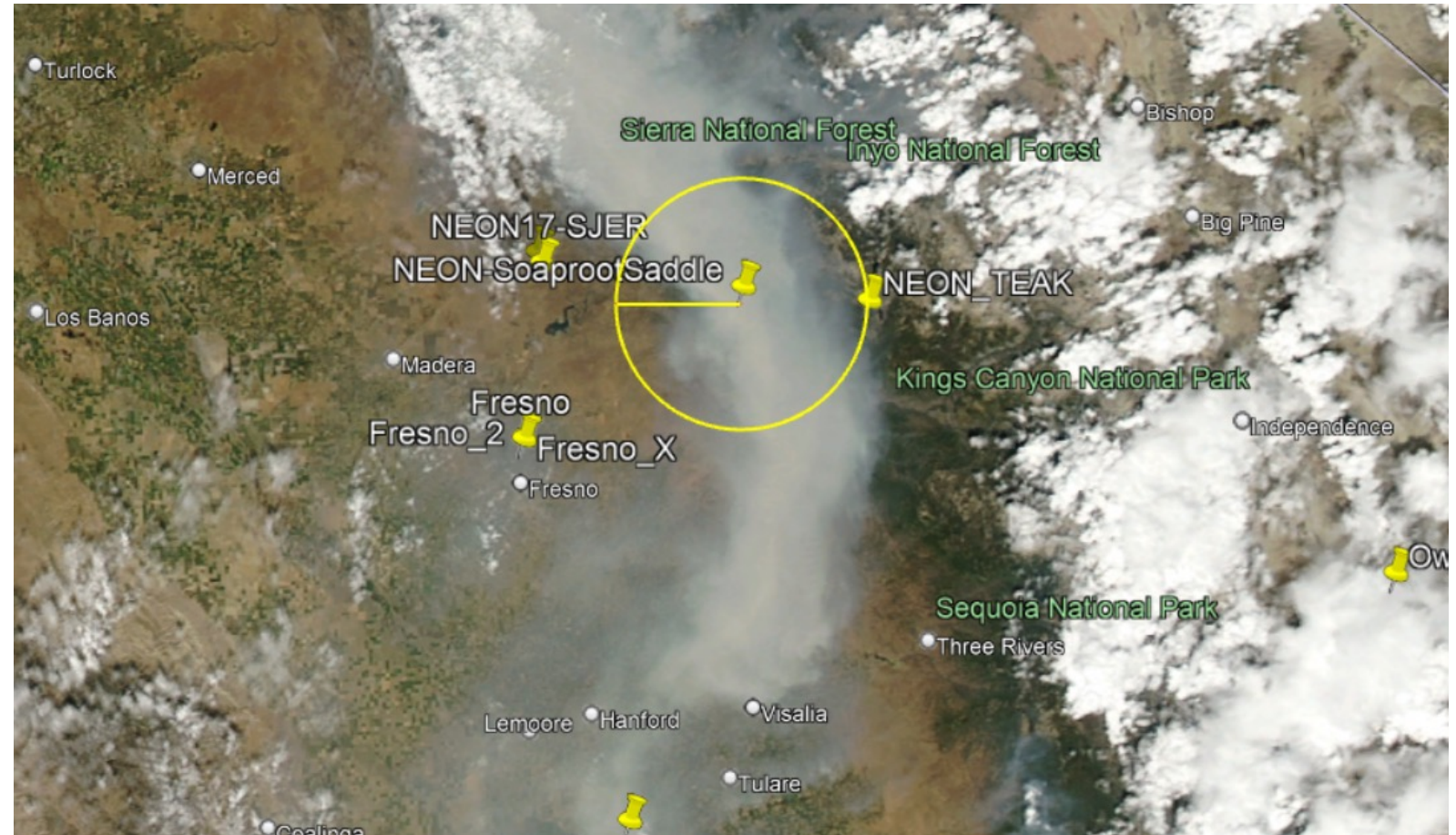
**Documented and small (hopefully)**

**Can be a real pain to generalise**

$$ED = \sqrt{(\sigma_{\text{sat}}^2 + \sigma_{\text{ref}}^2 + \sigma_{\text{var}}^2)}$$

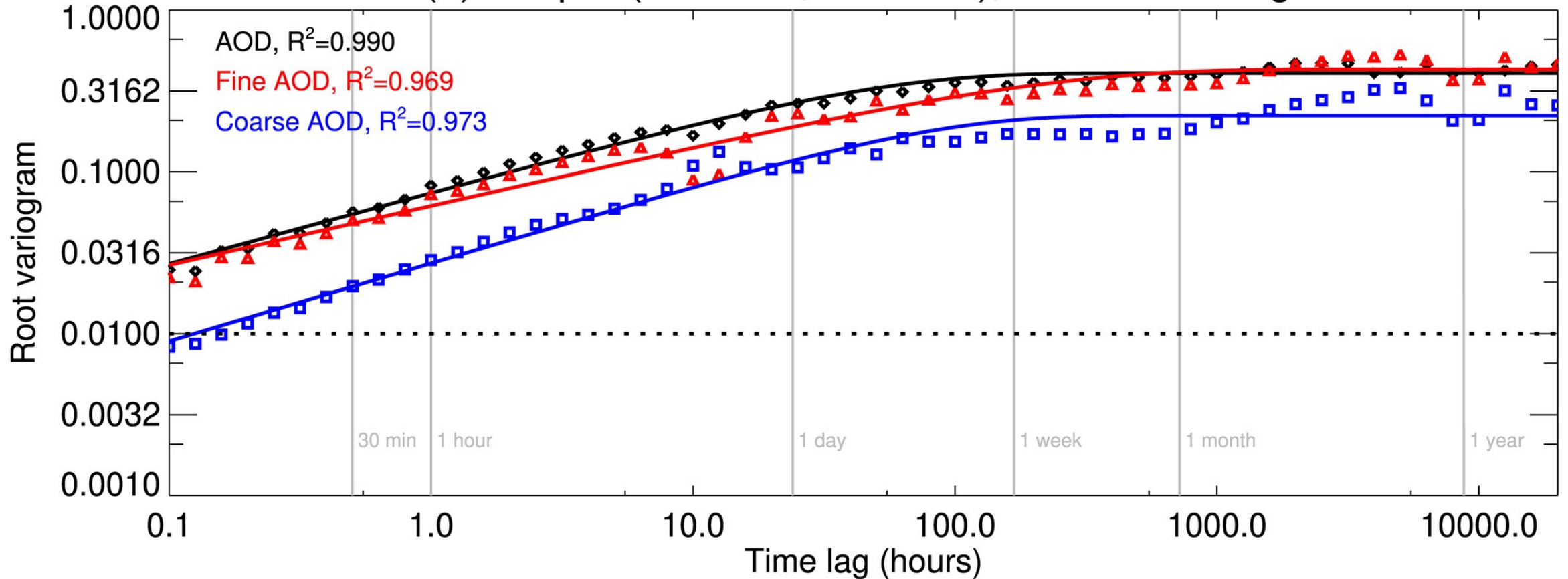
# Spatial and temporal variations can both be important

- Look at *spatial statistics* of satellite data in the area, *temporal statistics* of the ground data
- *Generalise* based on a model
- Use *robust statistics* (e.g., median not mean)
- *Shrink* area aggregated or comparison window

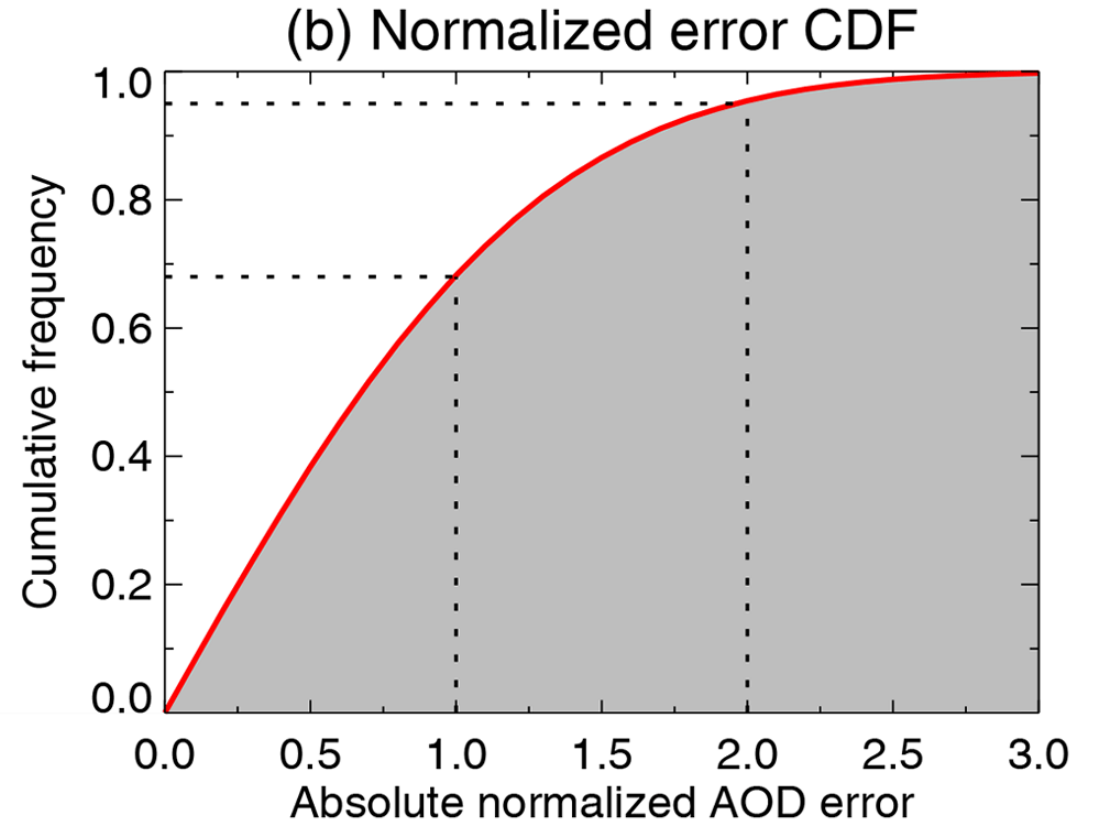
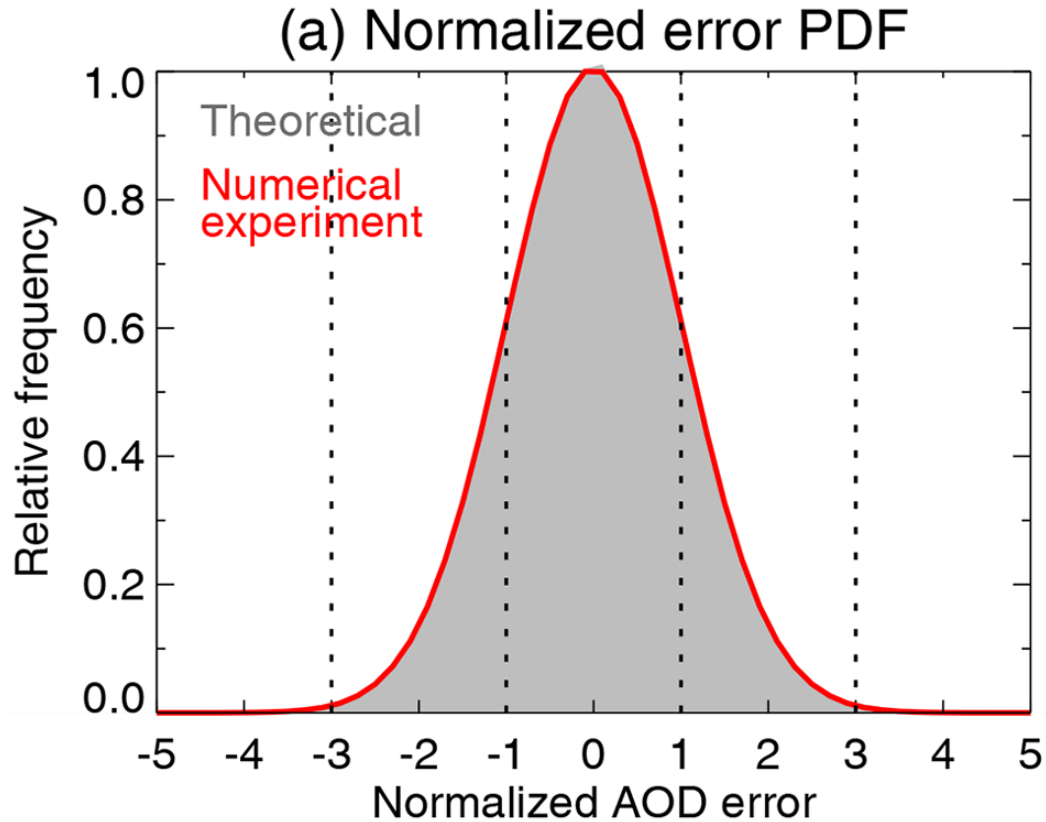


# Spatial and temporal variations can both be important

(a) Kanpur (26.51° N, 80.23° E), aerosol loading



# Look at distributions of normalised error

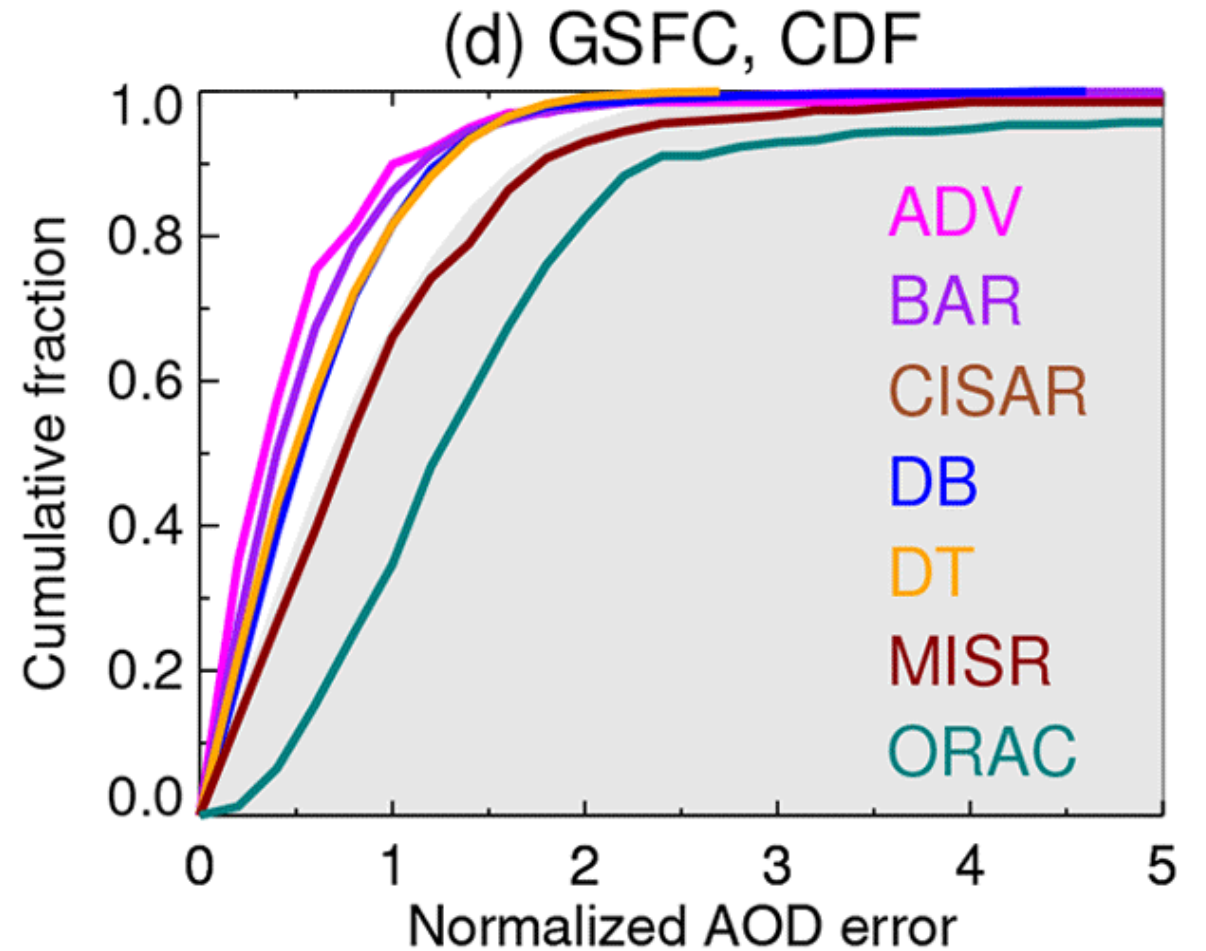
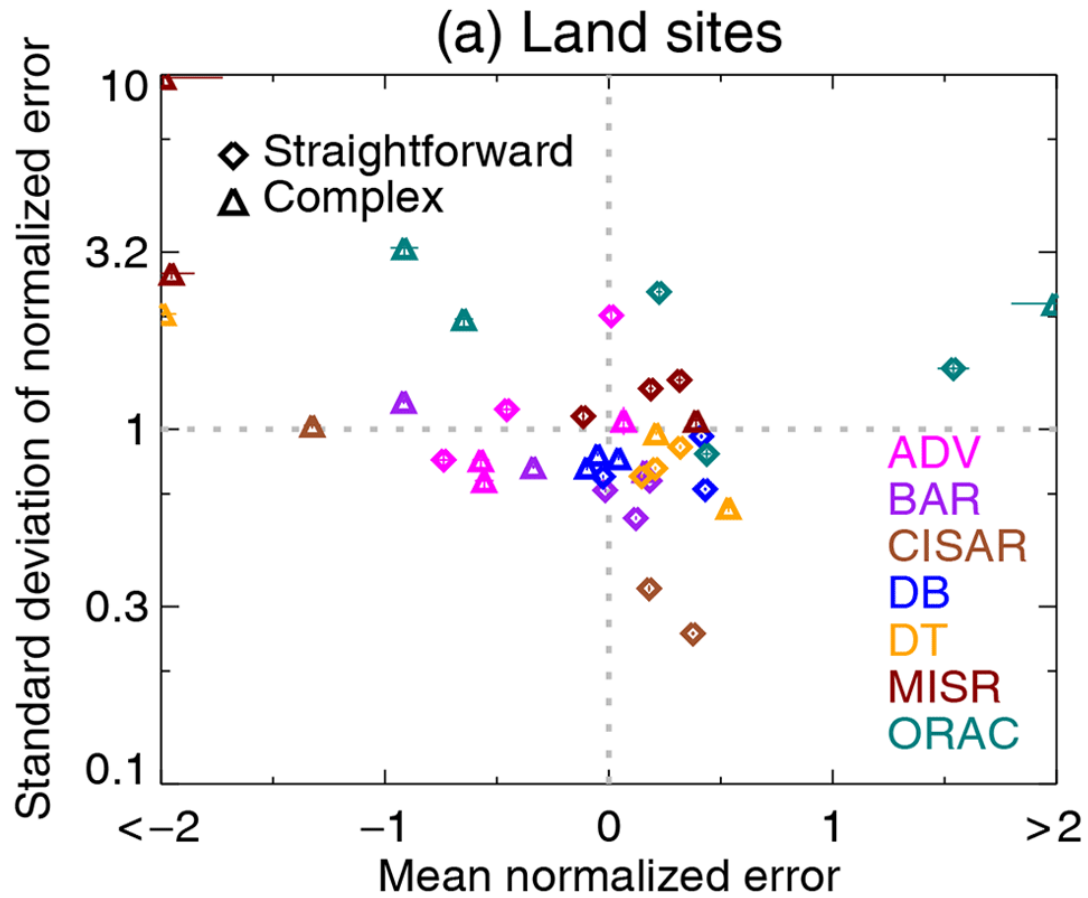


- Normalise as error/ED
- If unbiased and appropriate magnitude, distributions should follow  $N(0,1)$

Sayer *et al.*, AMT (2020)

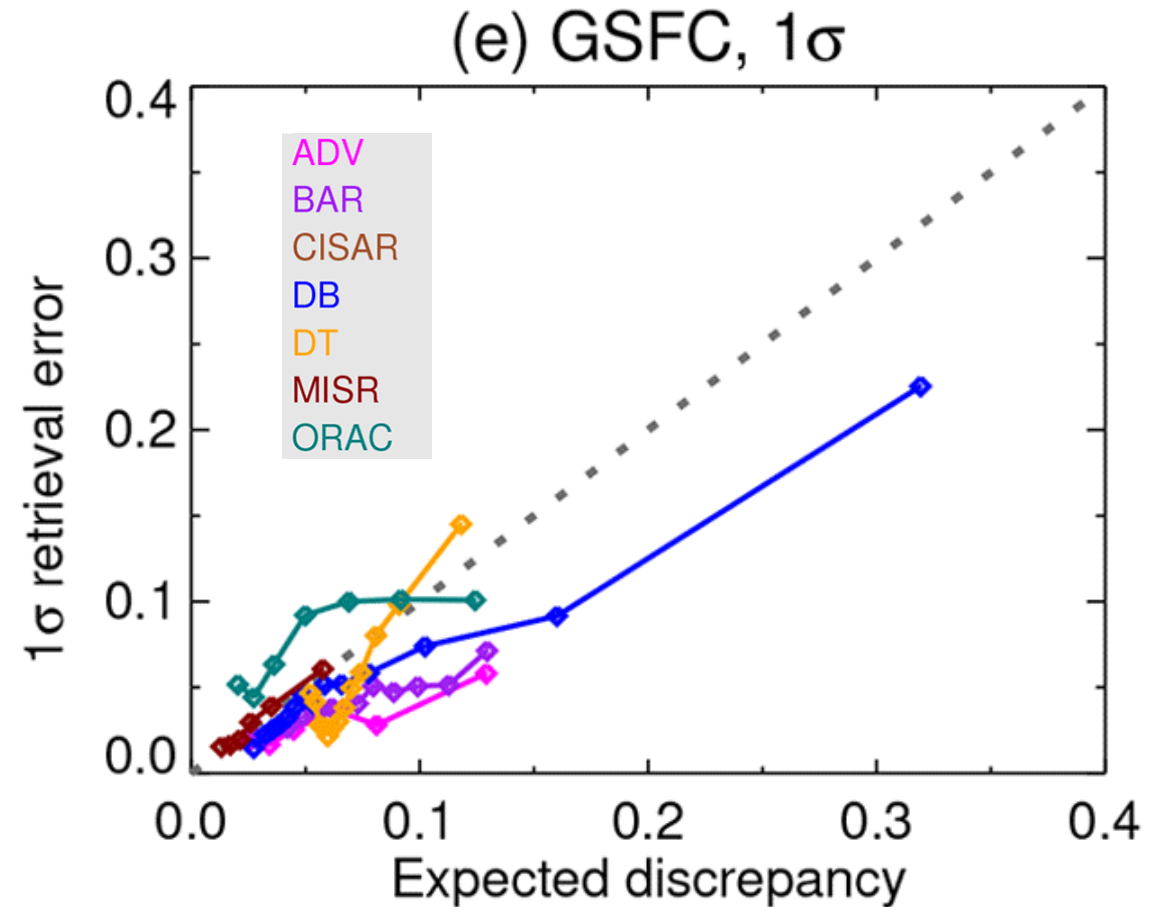


# Examples on real aerosol data



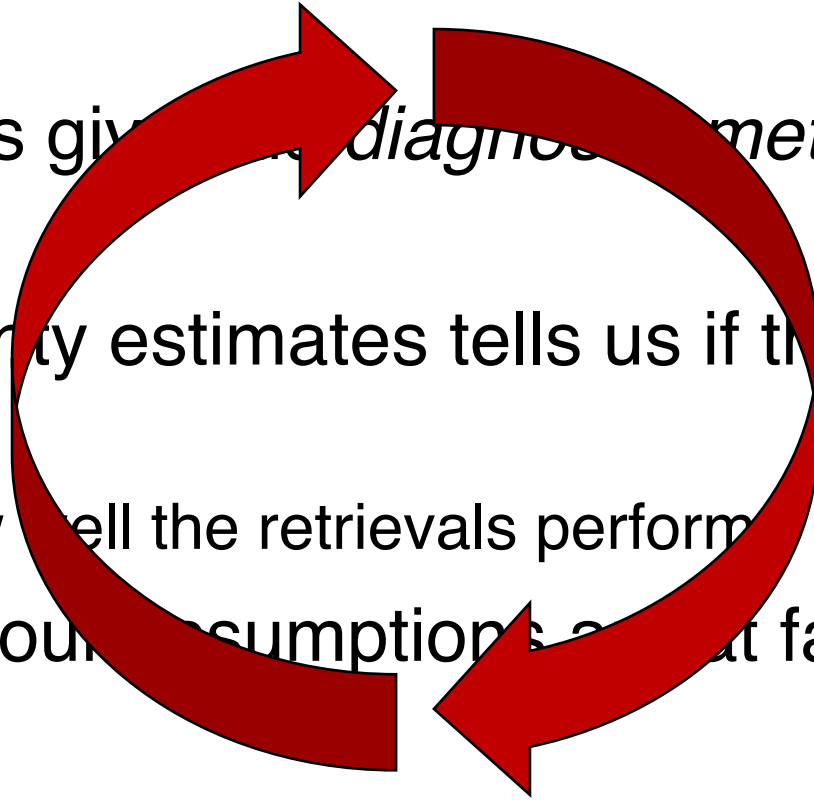
# Can we tell low-uncertainty situations from high ones?

- Stratify by ED
- Look at  $1\sigma$  point (68<sup>th</sup> percentile) of absolute error
  - Additional percentiles can tell you more about *how Gaussian* the error distribution is for a given ED
- Similar to *forecast calibration, skill scores, etc*



# Closing the loop

- Validating retrievals gives us *diagnostic metrics of quality* to report uncertainty
- Validating uncertainty estimates tells us if things are *working as we expect*
  - Regardless of how well the retrievals perform
- Figure out where your assumptions are at fault, then *refine* them!



# References and resources

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- Sayer, A. M. (2020), How long is too long? Variogram analysis of AERONET data to aid aerosol validation and intercomparison studies, *Earth and Space Science*, 7, e2020EA001290. <https://doi.org/10.1029/2020EA001290>
- Sayer, A. M., Govaerts, Y., Kolmonen, P., Lipponen, A., Luffarelli, M., Mielonen, T., Patadia, F., Popp, T., Povey, A. C., Stebel, K., and Witek, M. L. (2020), A review and framework for the evaluation of pixel-level uncertainty estimates in satellite aerosol remote sensing, *Atmos. Meas. Tech.*, 13, 373–404, <https://doi.org/10.5194/amt-13-373-2020>
- McKinna, L. I. W., Cetinić, I., & Werdell, P. J. (2021), Development and validation of an empirical ocean color algorithm with uncertainties: A case study with the particulate backscattering coefficient, *Journal of Geophysical Research: Oceans*, 126, e2021JC017231, <https://doi.org/10.1029/2021JC017231>