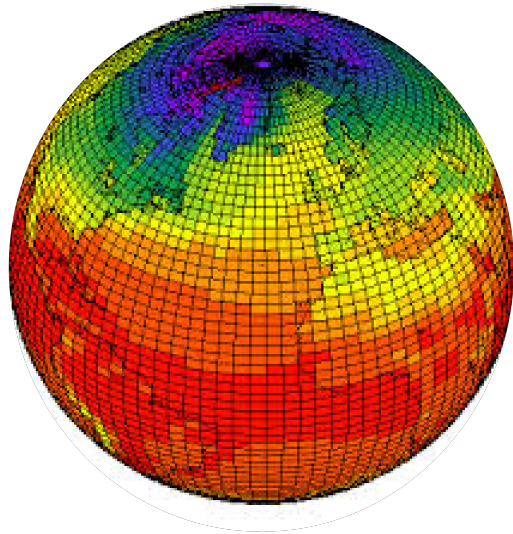


Model



Observations



Reality



Reality

Model A

Observation A



Observation B

Is model system B superior to model system A?

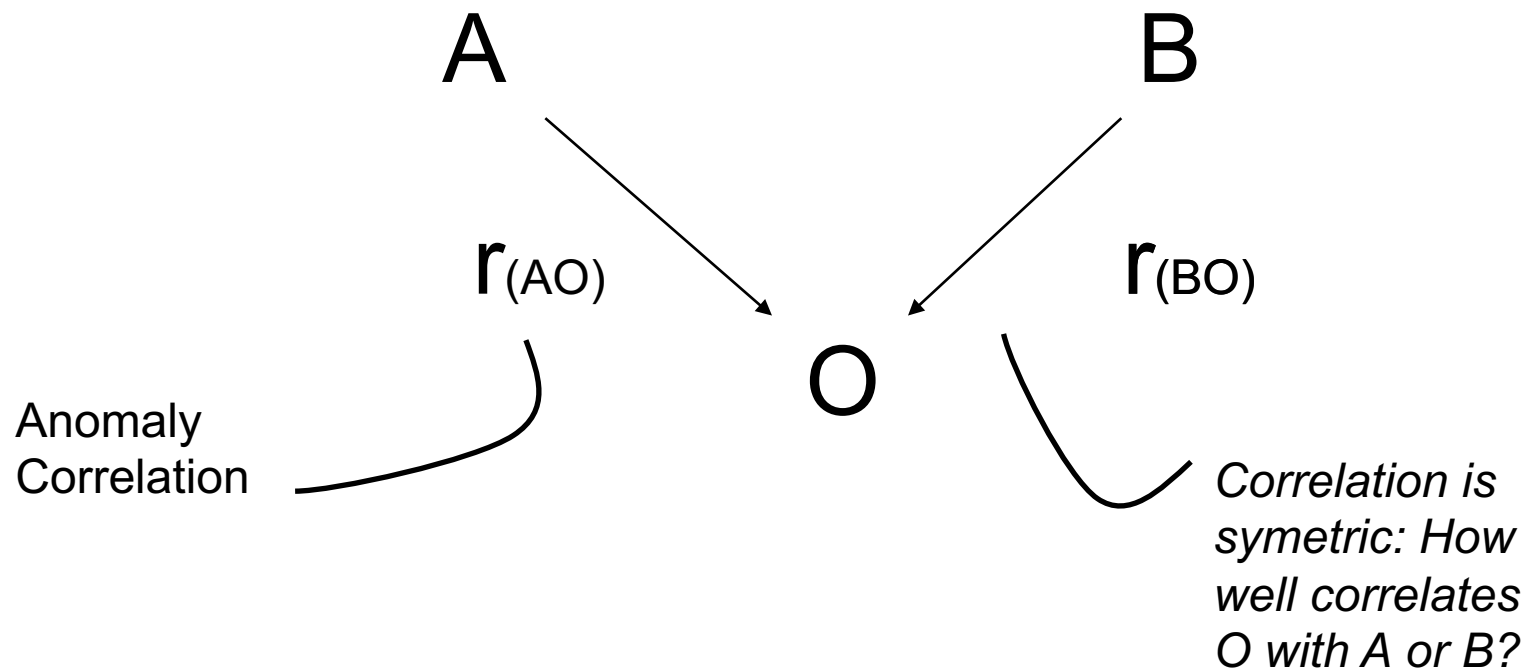
A

Low horizontal resolution

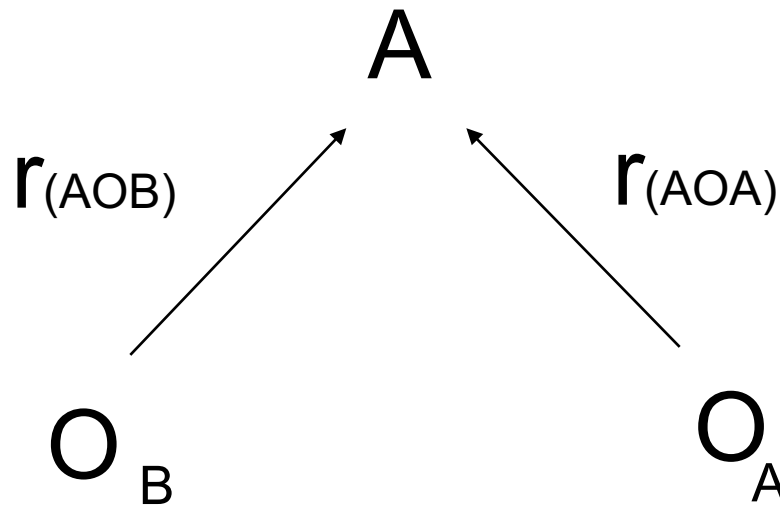
B

High horizontal resolution

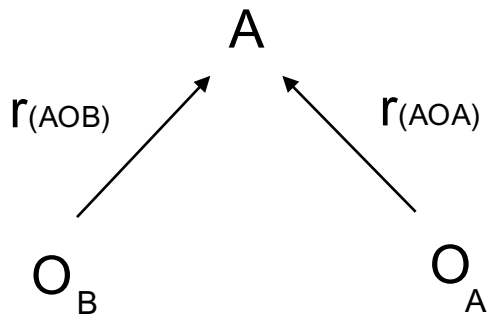
Compare hindcast skill with an observation



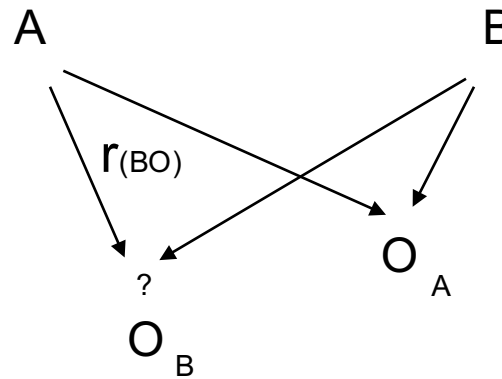
Which observation is better? A useful question?



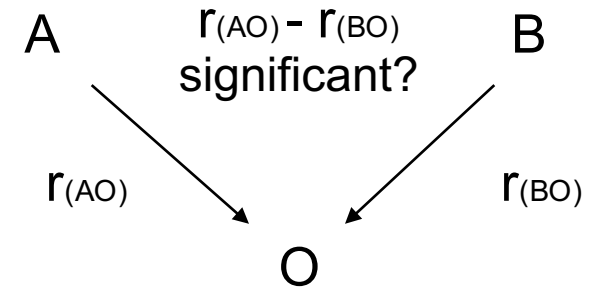
Overview this presentation



(a) Which observations has the smallest error

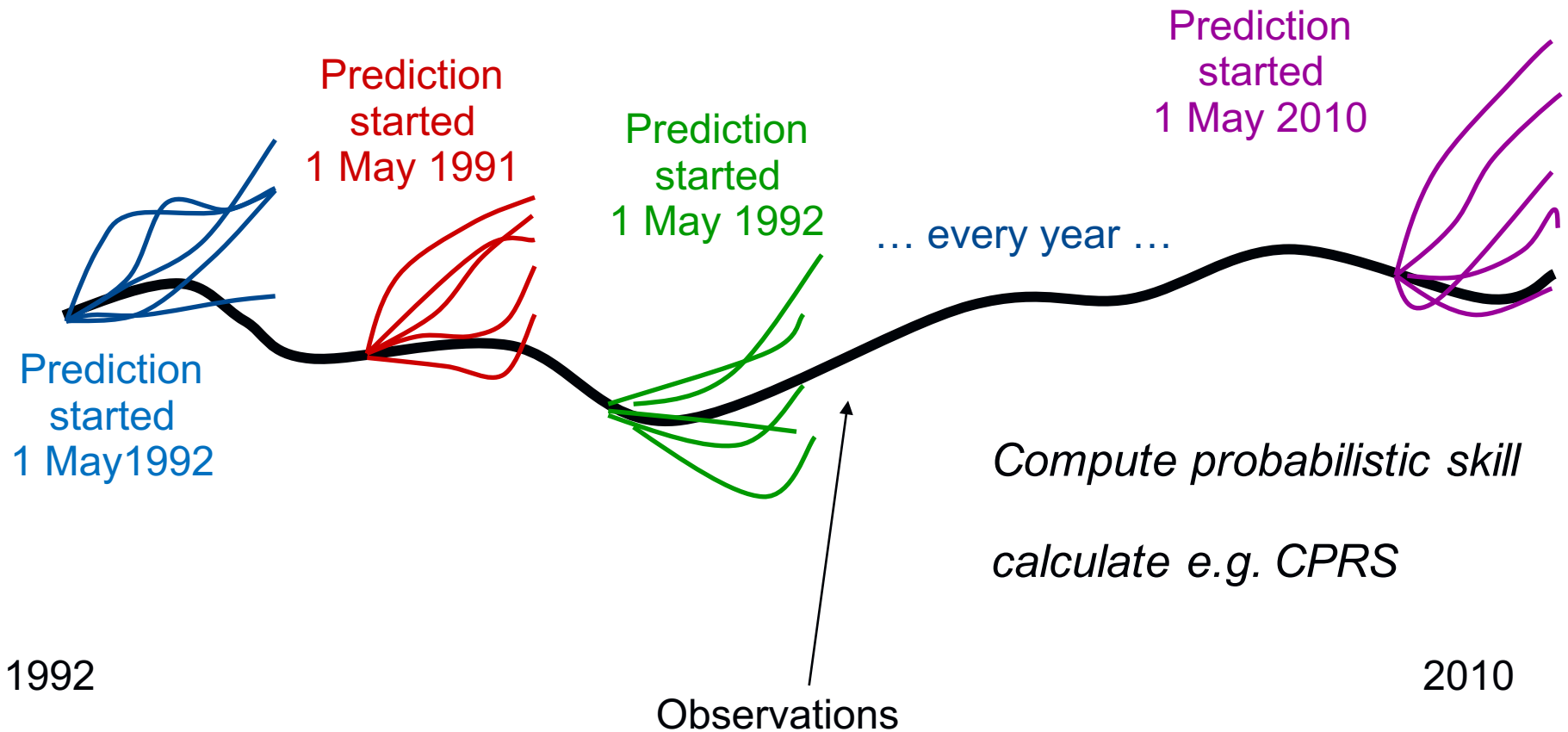


(b) How important is observational uncertainty

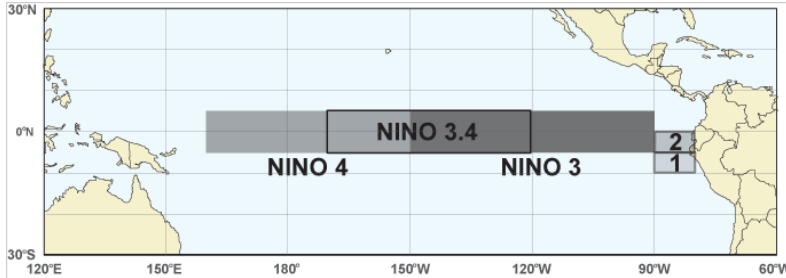


(c) How to detect improvements in models or observations

Seasonal forecast skill



Target: Global and Niño3.4 SST



Observations: 4 Sea-surface Temperature (SST) observations: ESA-CCI, HadISST, ERSST4, ERA-Interim

Prediction started
1 May 1991

Prediction started
1 May 1992

Prediction started
1 May 1992

... every year ...

Prediction started
1 May 2010

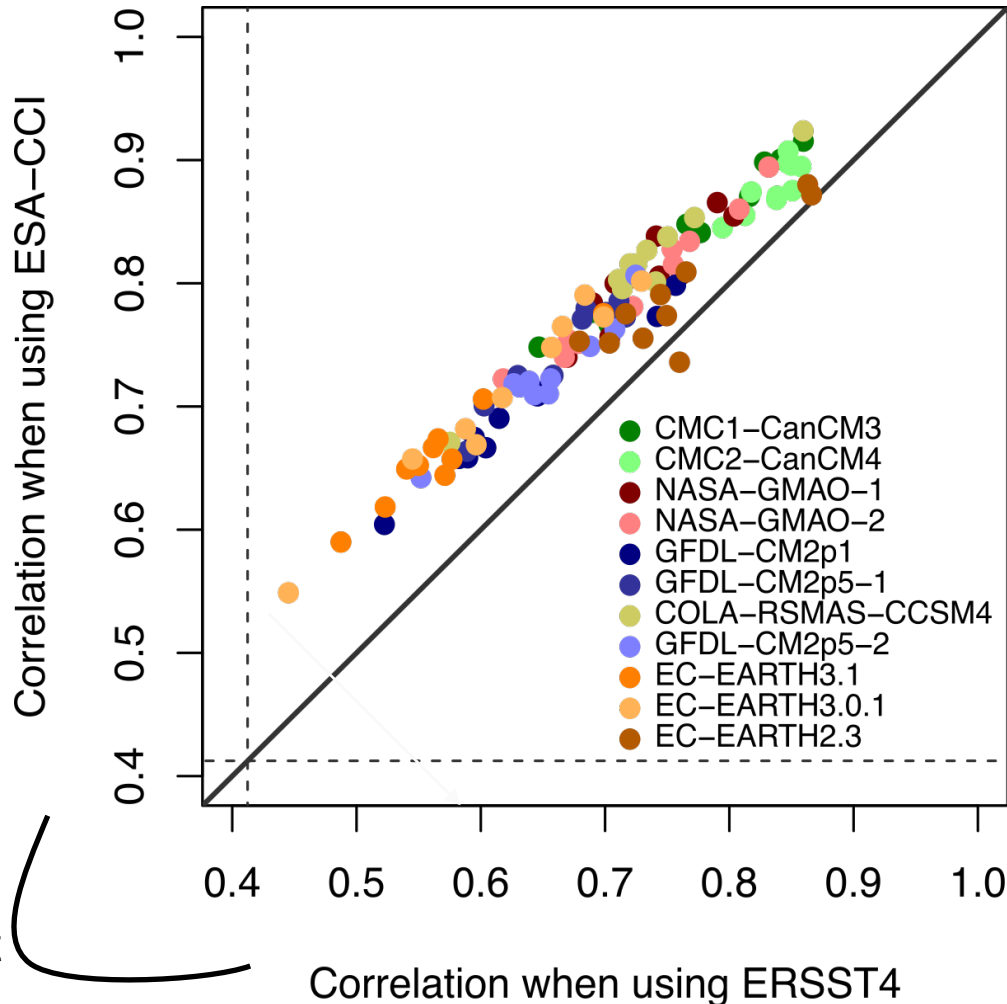
Models: EC-Earth (3 versions), ECMWF S4, North American Multi-Model Ensemble (NMME, 7 models) 10 – member forecast each

Compute ensemble-mean to distill climate signal

calculate e.g. anomaly correlation

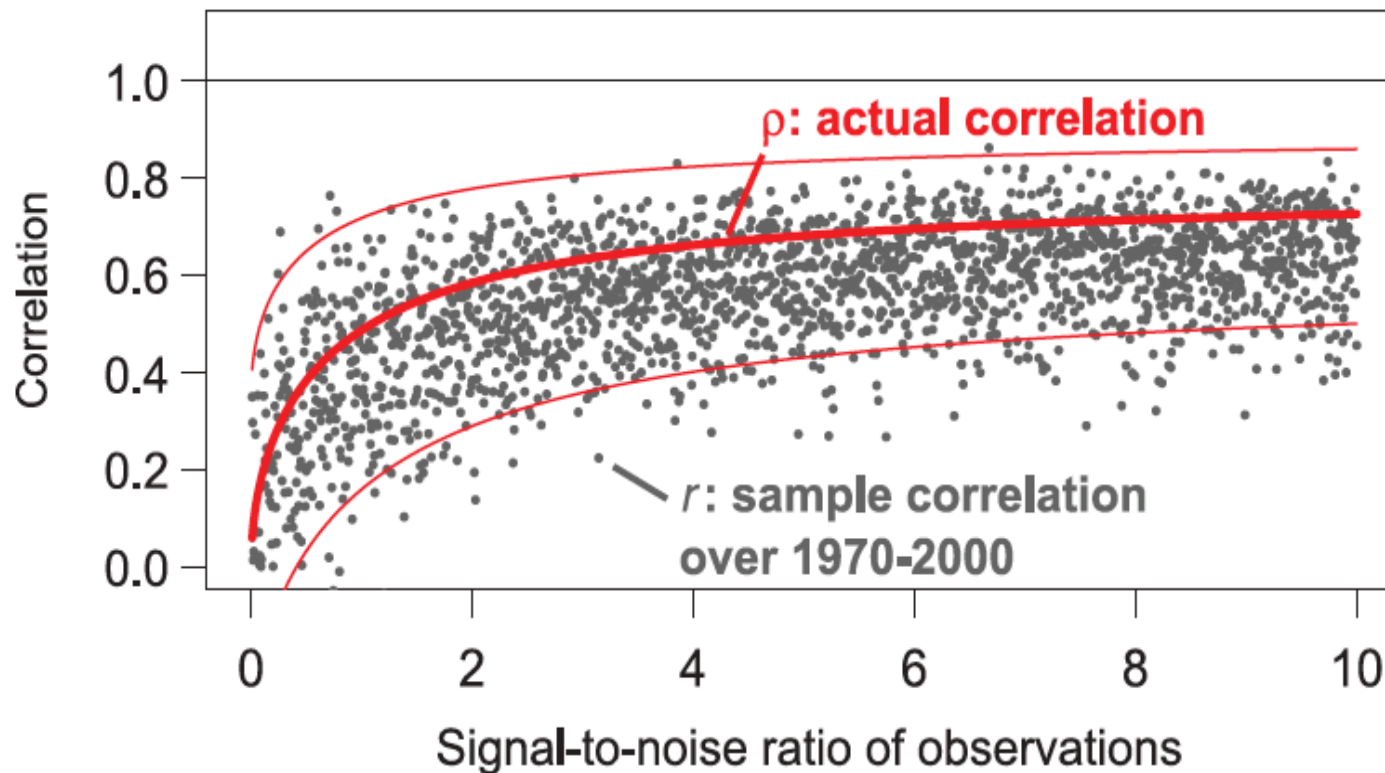
CCI SST yields systematic higher correlation skill across many models

Independent from the models

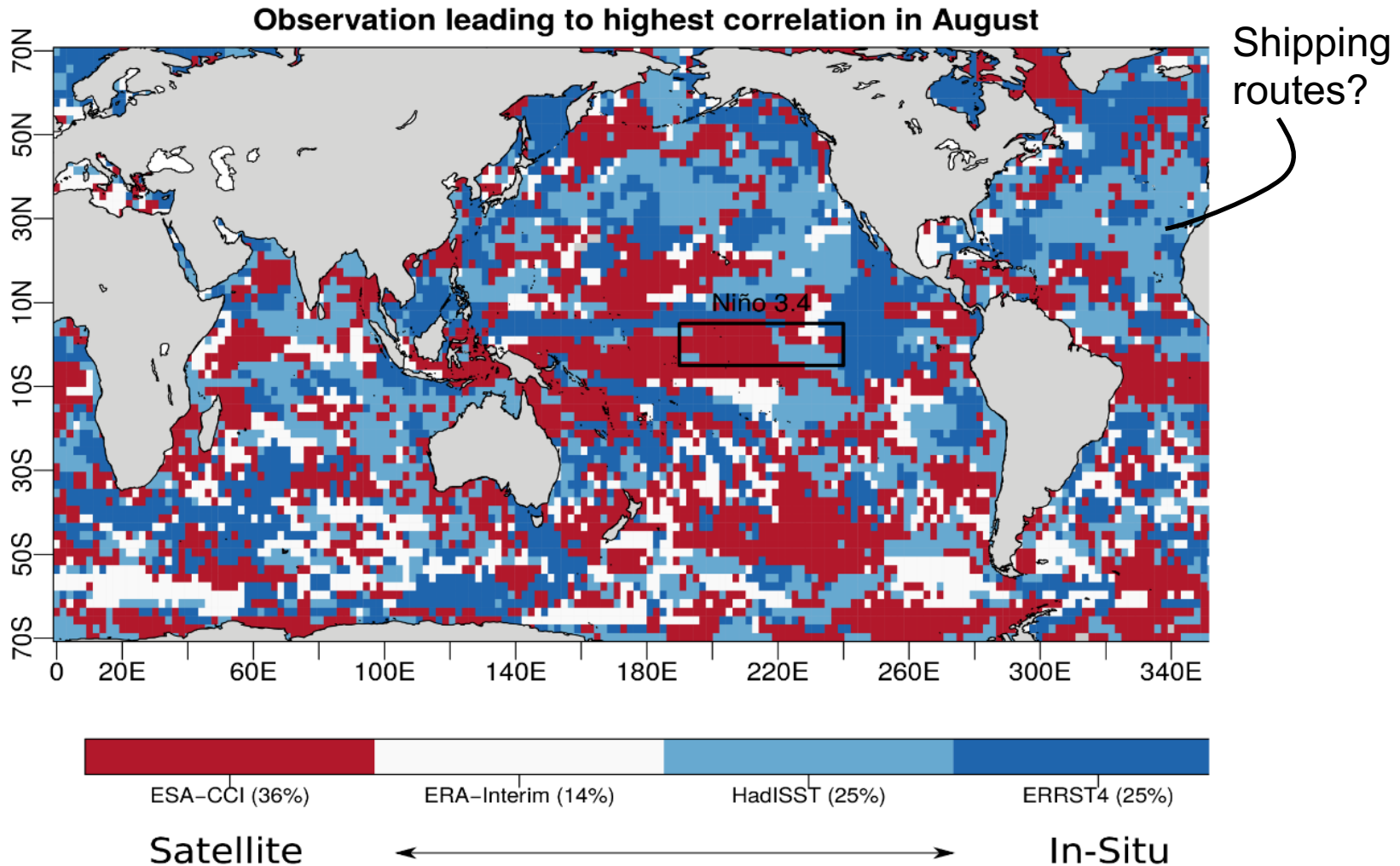


Fully independent datasets

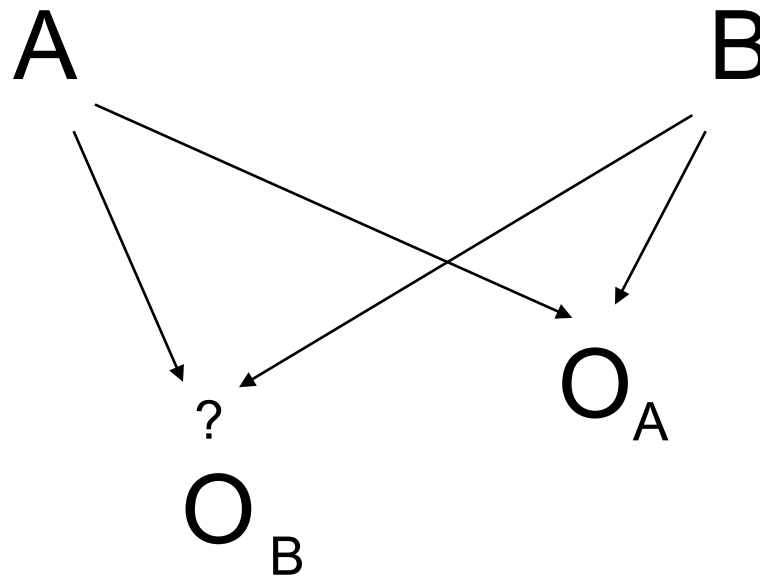
Correlation reduces with noise either co-variates: observational uncertainty reduces forecast skill



Choice of observation may differ on the location, overall CCI best

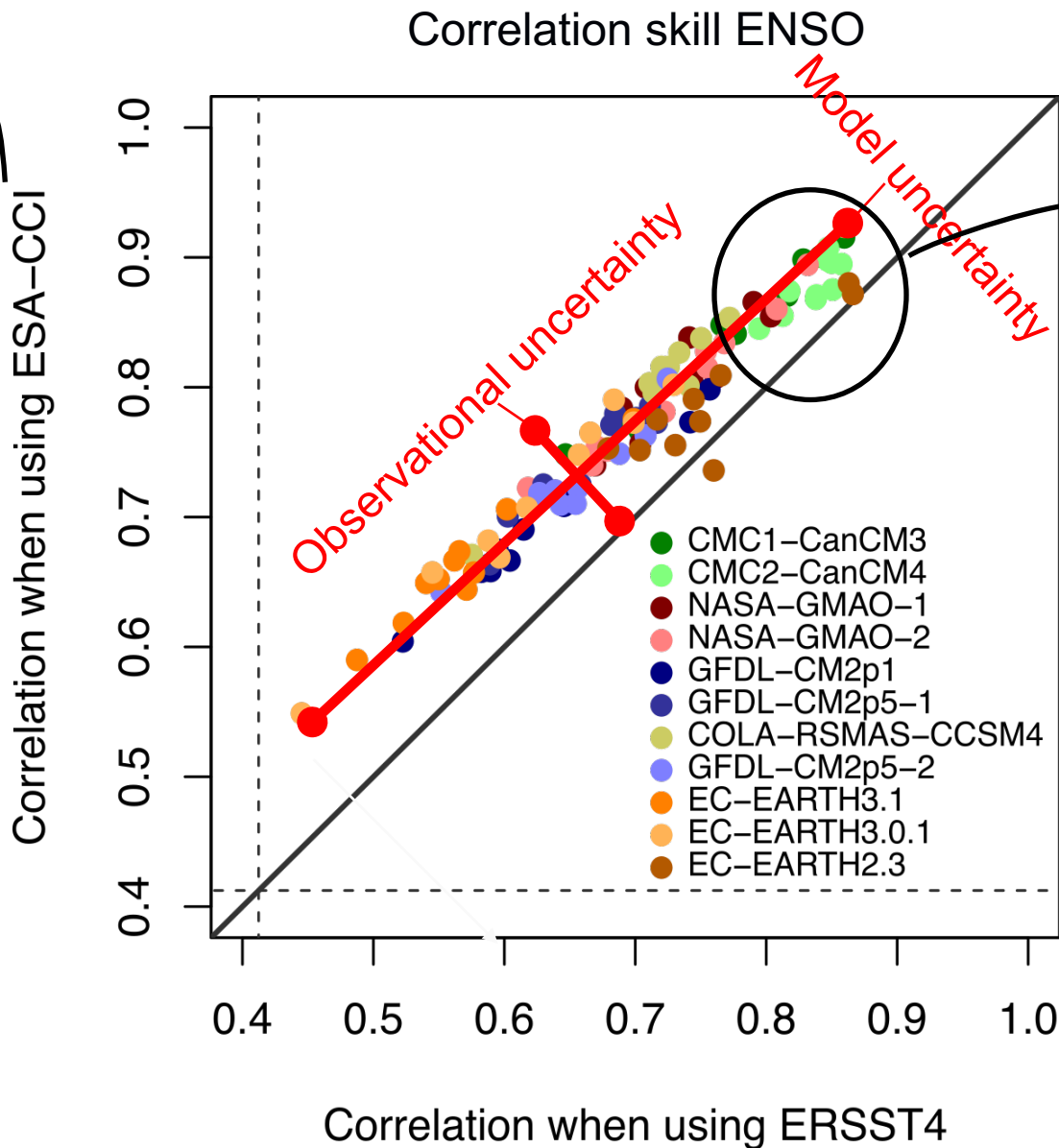


How important is the observational uncertainty?



Acknowledging joint uncertainty

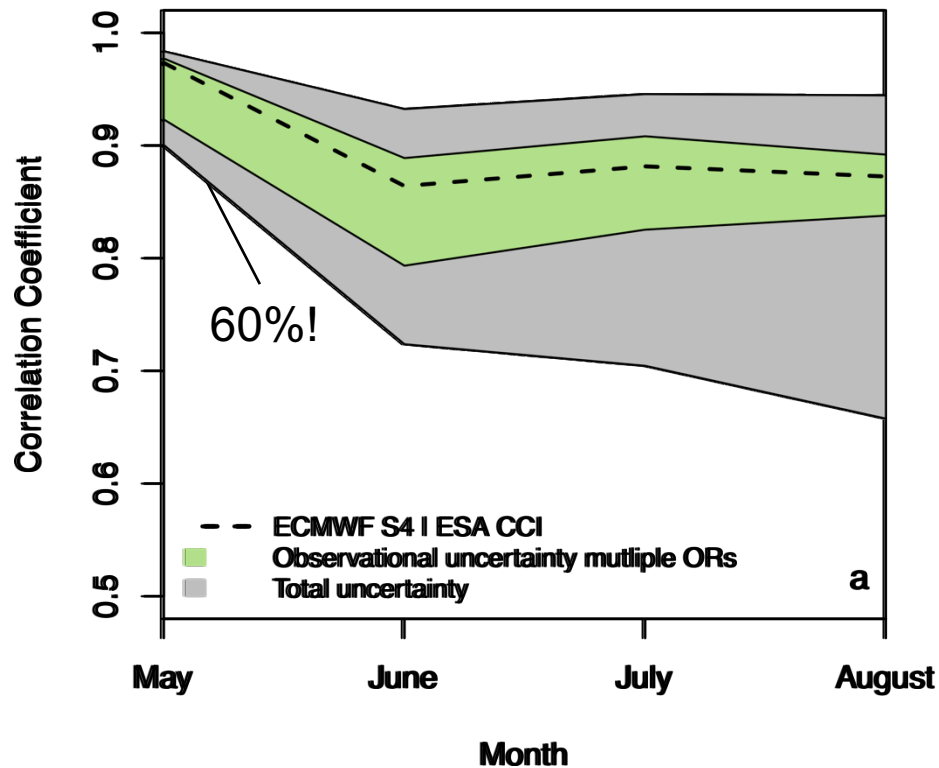
Independent from the models



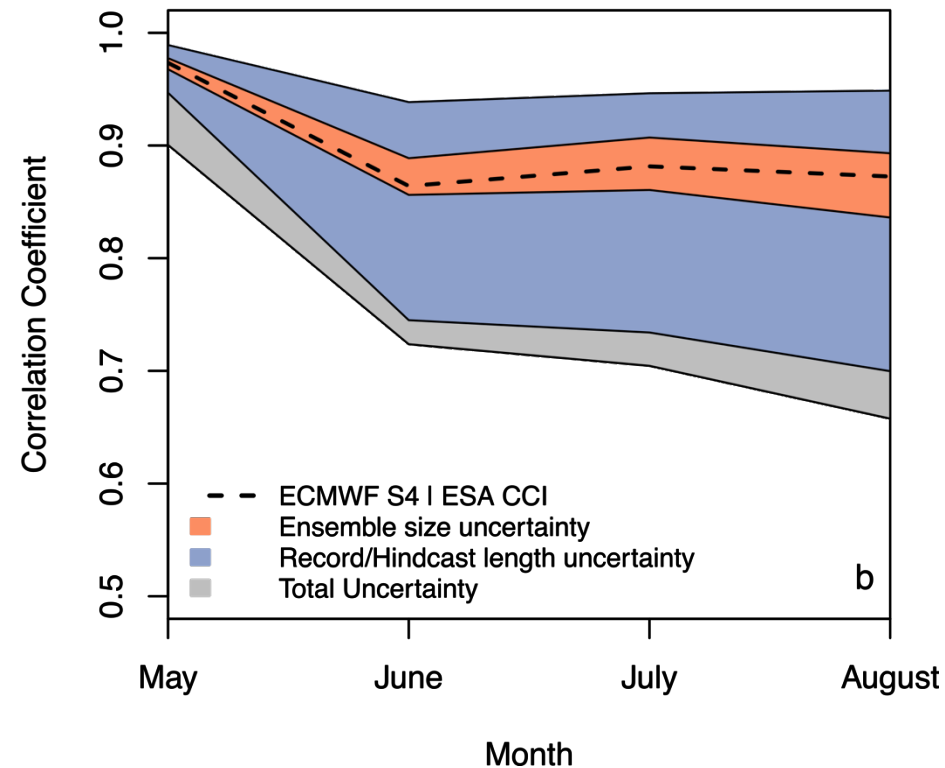
How important for individual system?

Comparison to sample uncertainties: observational uncertainty is an important source of verification uncertainty for ENSO

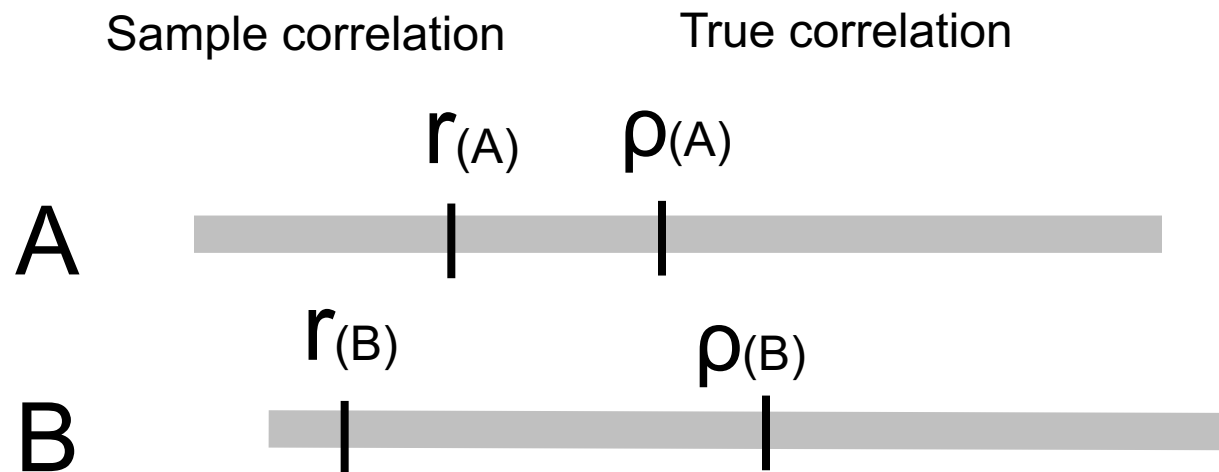
ENSO Prediction (Observational uncertainty)



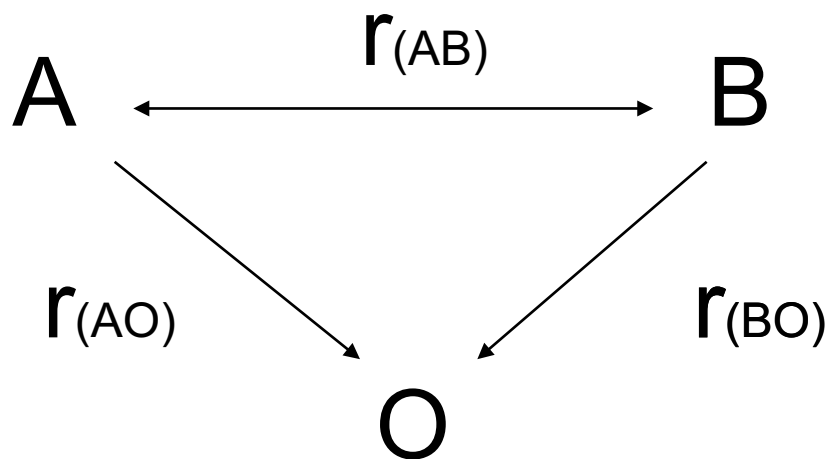
ENSO Prediction (Sampling Uncertainty)



Are the differences in performance of models or observations significant $r_{(CCI)} > r_{(ERSST)}$?



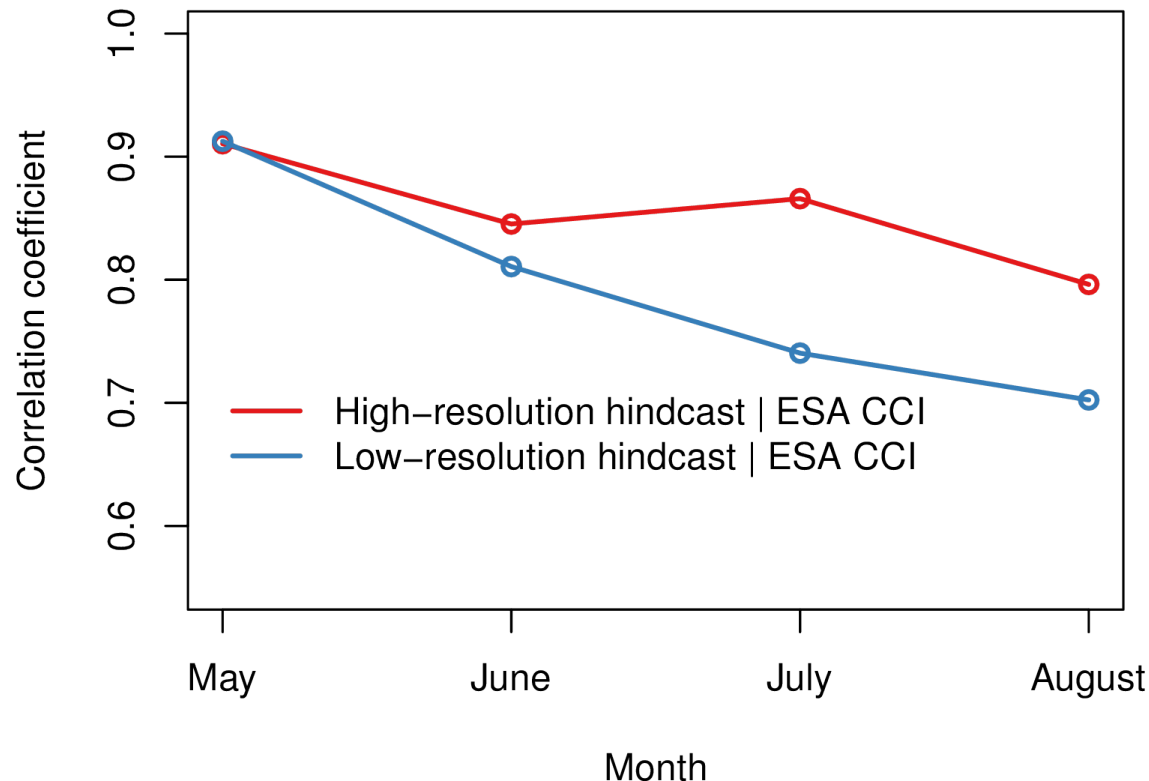
Models and observations are statistically dependent!



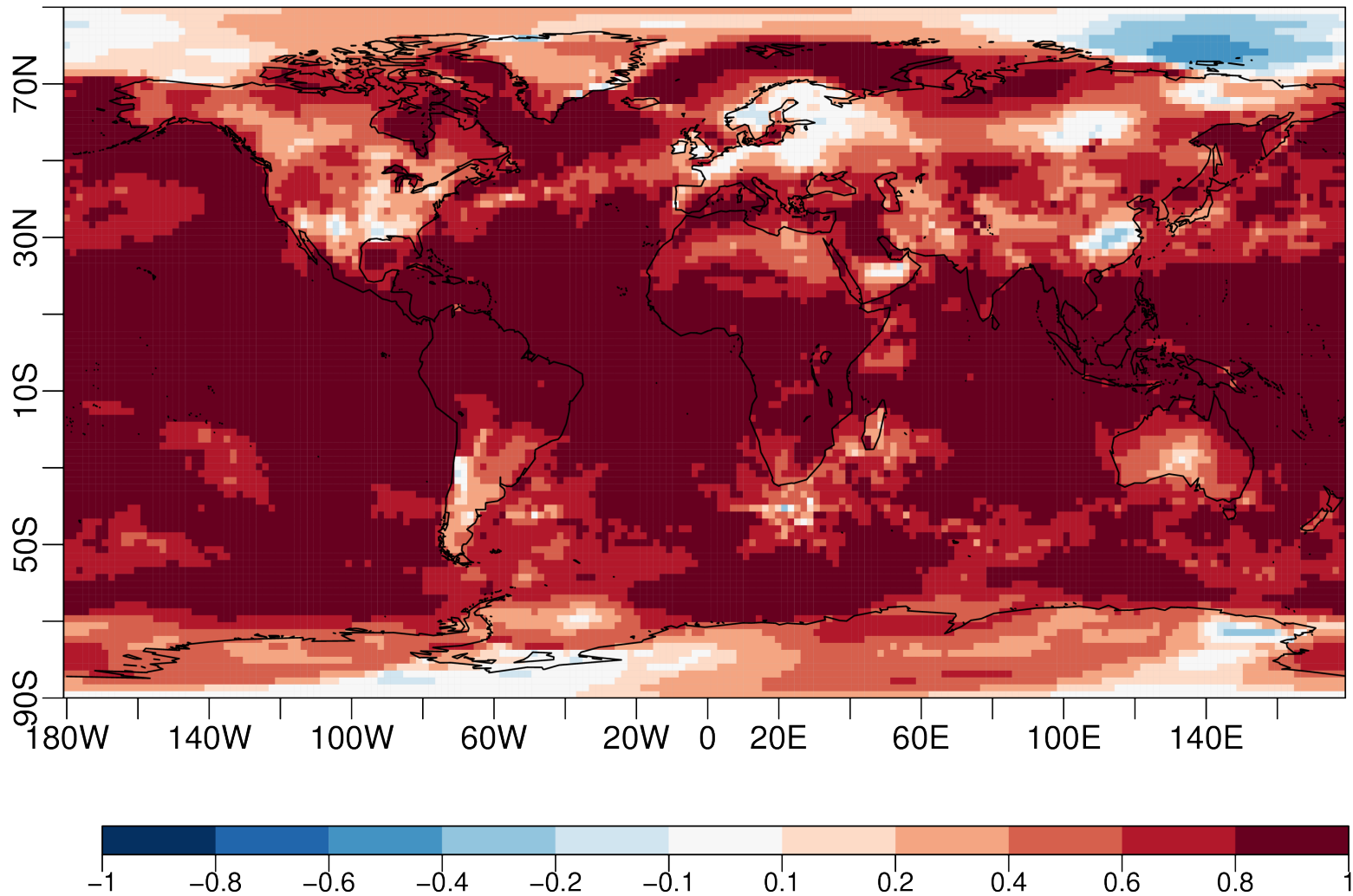
Fisher-test (common test in community) assumes independence while newer tests (Steiger, 1980, Zou, 2007) don't.

High-resolution hindcasts improves El Niño Southern Oscillation (ENSO) predictions, but change not significant at 5% (Fisher-test)

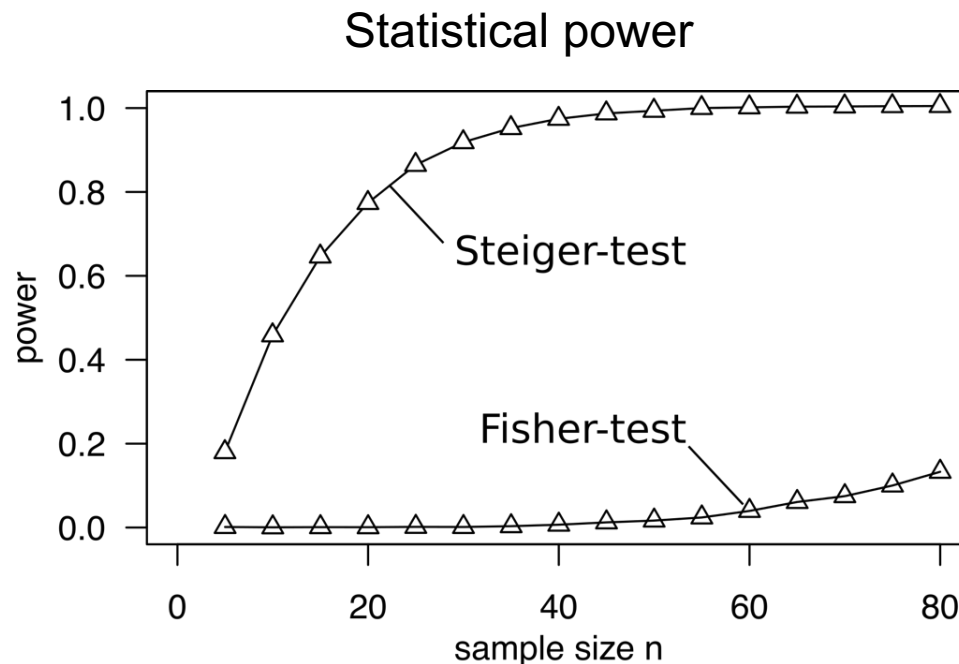
Prediction skill ENSO: Increase in resolution



Correlation of Low and High-resolution hindcast



**Power to detect a difference between increases dramatically.
Improvement now statistical significant at 1% level.**

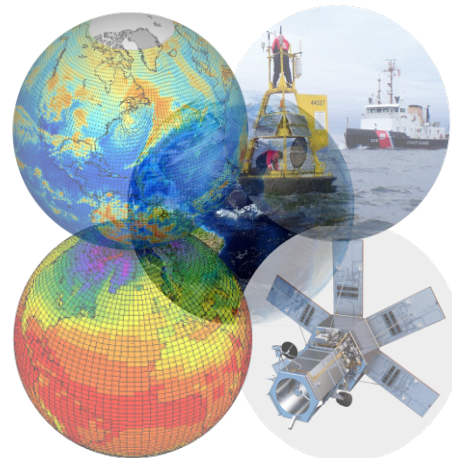


In medicinal science only studies with power > 80% are accepted, a guideline for forecasting?

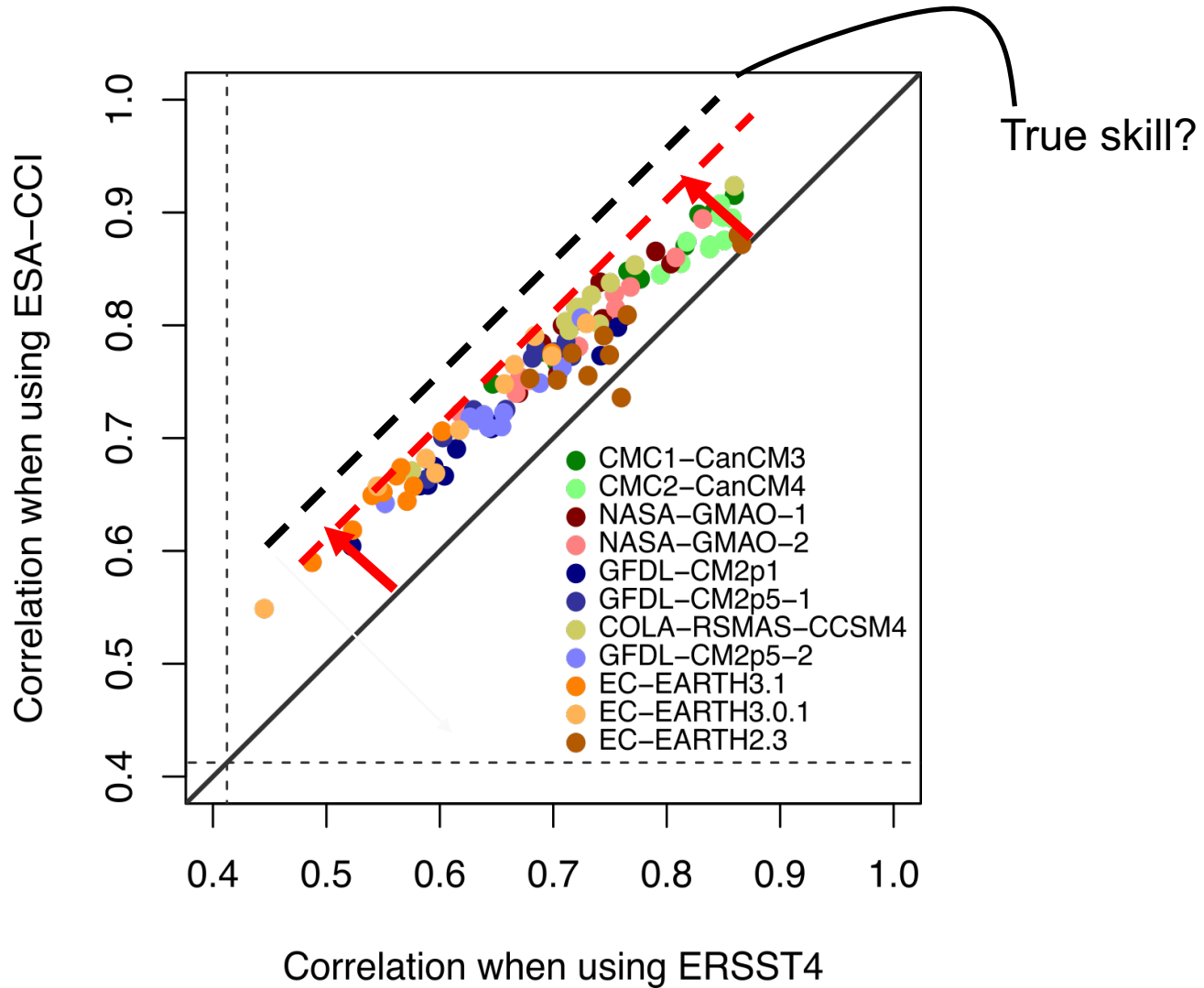
Models and observations are both approximations of the truth and uncertainty in both sources can be important.

Models can be valuable in assessing observational quality and thus guide a more objective dataset selection

Testing improvements in models and observations requires the consideration of dependence between all source of information

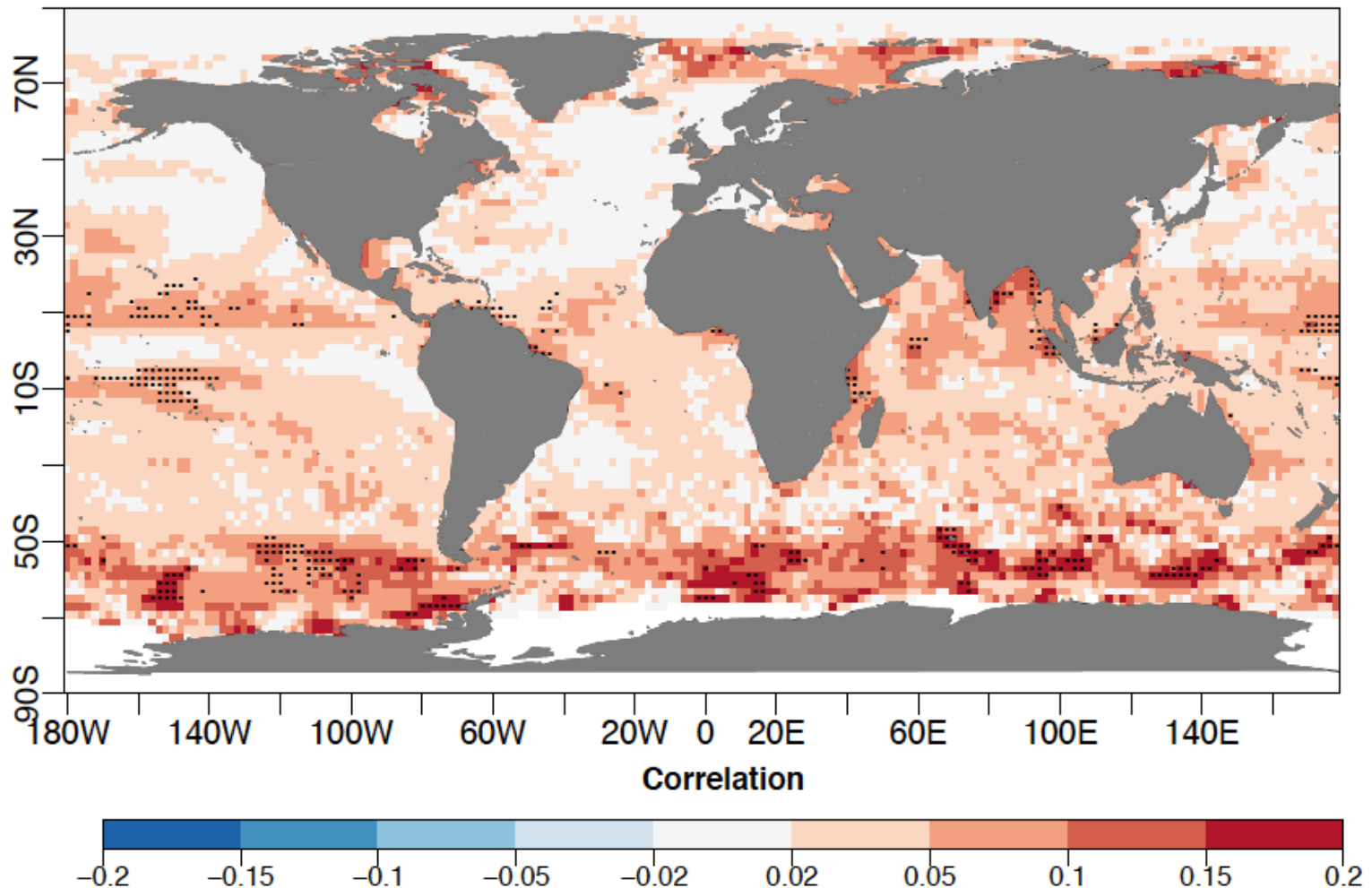


True climate predictions skill is systematically underestimated due to uncertainties in the observations

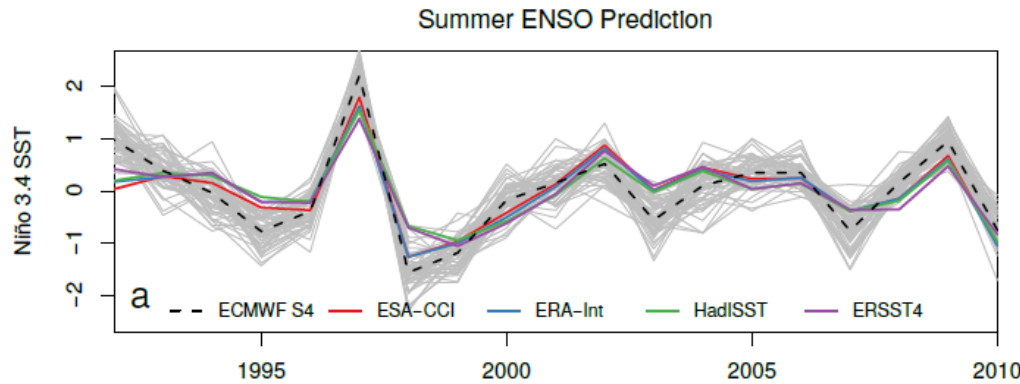


Seasonal SST forecast skill is underestimated up to 0.2 correlation

Lost skill due to observational uncertainty



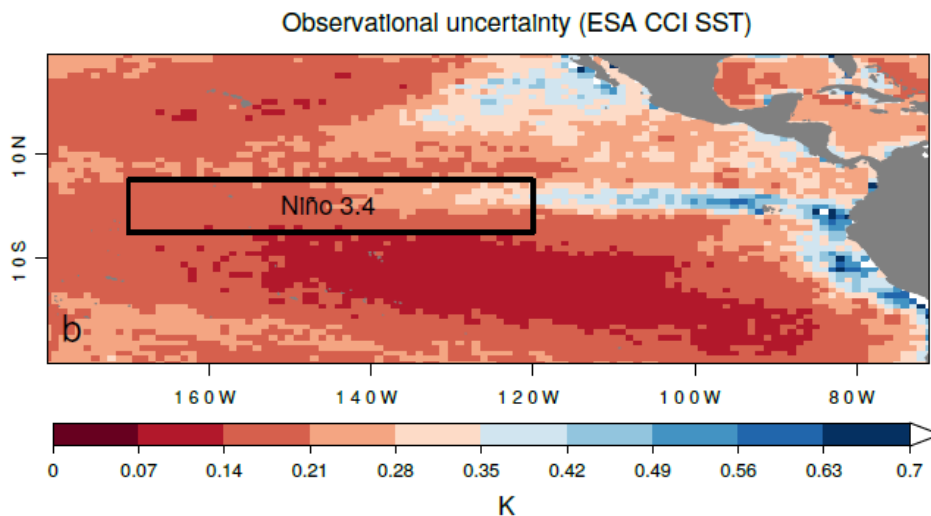
Model evaluation often requires spatial and temporal averaging,
requires the consideration of error correlation scales



Monthly
Niño3.4



Uncorrelated uncertainty
reduces by $1/\sqrt{N}$
but errors are not
uncorrelated!



Daily, 4 km

Quantifying observational uncertainty is a challenge and propagation scales represented to by the models is a big gap – A stronger interaction is required with the observational data community

ESA *Climate Modelling User Group (CMUG)* is going to explore observational uncertainty in model – observation inter-comparison strongly in the future

Metrics and new statistical tests are required that can make use of the observational uncertainty data that future data sets are going to provide

Thank you!



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EXCELENCIA
SEVERO
OCHOA

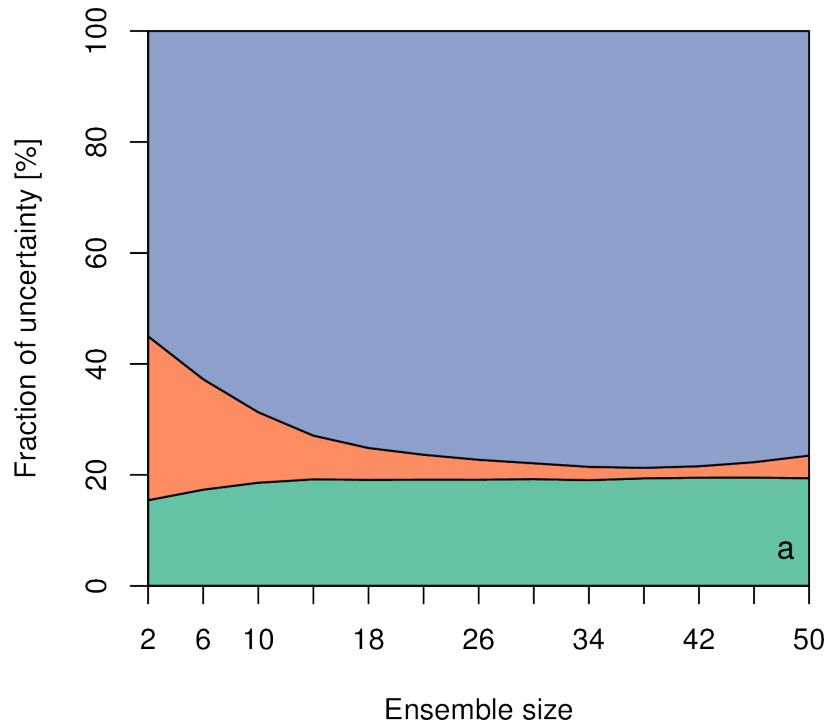
Massonnet, F., Bellprat, O., Guemas, V., Doblas-Reyes, F. J., (2016). Using climate models to estimate the quality of global observational data sets, *Science (AAAS)*

Bellprat, O., Massonnet, F., Siegert, S., Guemas, V., Doblas-Reyes, F. J. (2017). Exploring observational uncertainty in verification of climate model predictions, *Remote Sensing of the Environment (RSE)*, *in review*

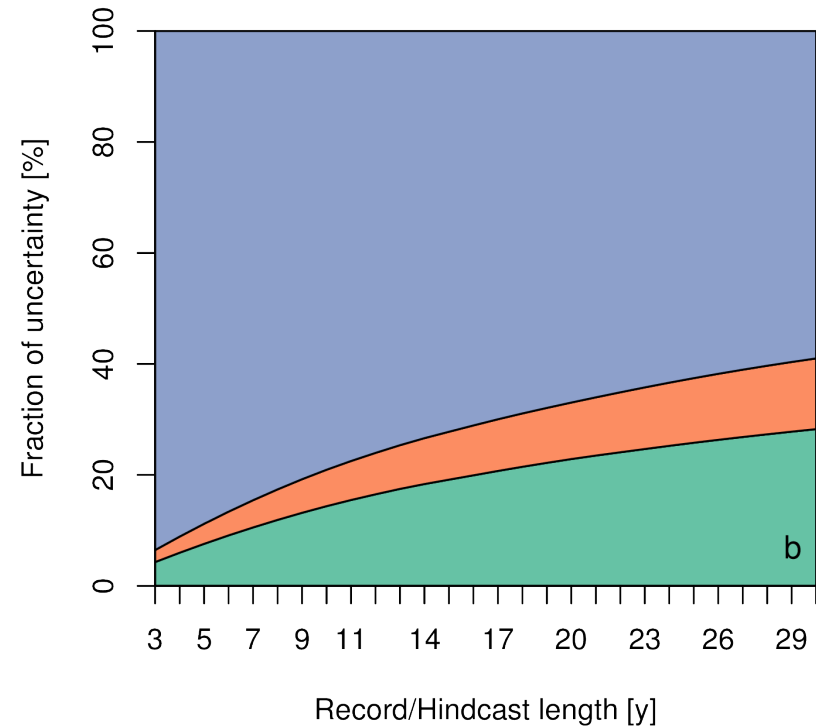
Siegert, S., Bellprat, O., Menegoz, M., Stephenson, D., Doblas-Reyes, F. (2016). Detecting improvements in forecast correlation skill: Statistical testing and power analysis. *Monthly Weather Review*

Extra Slides

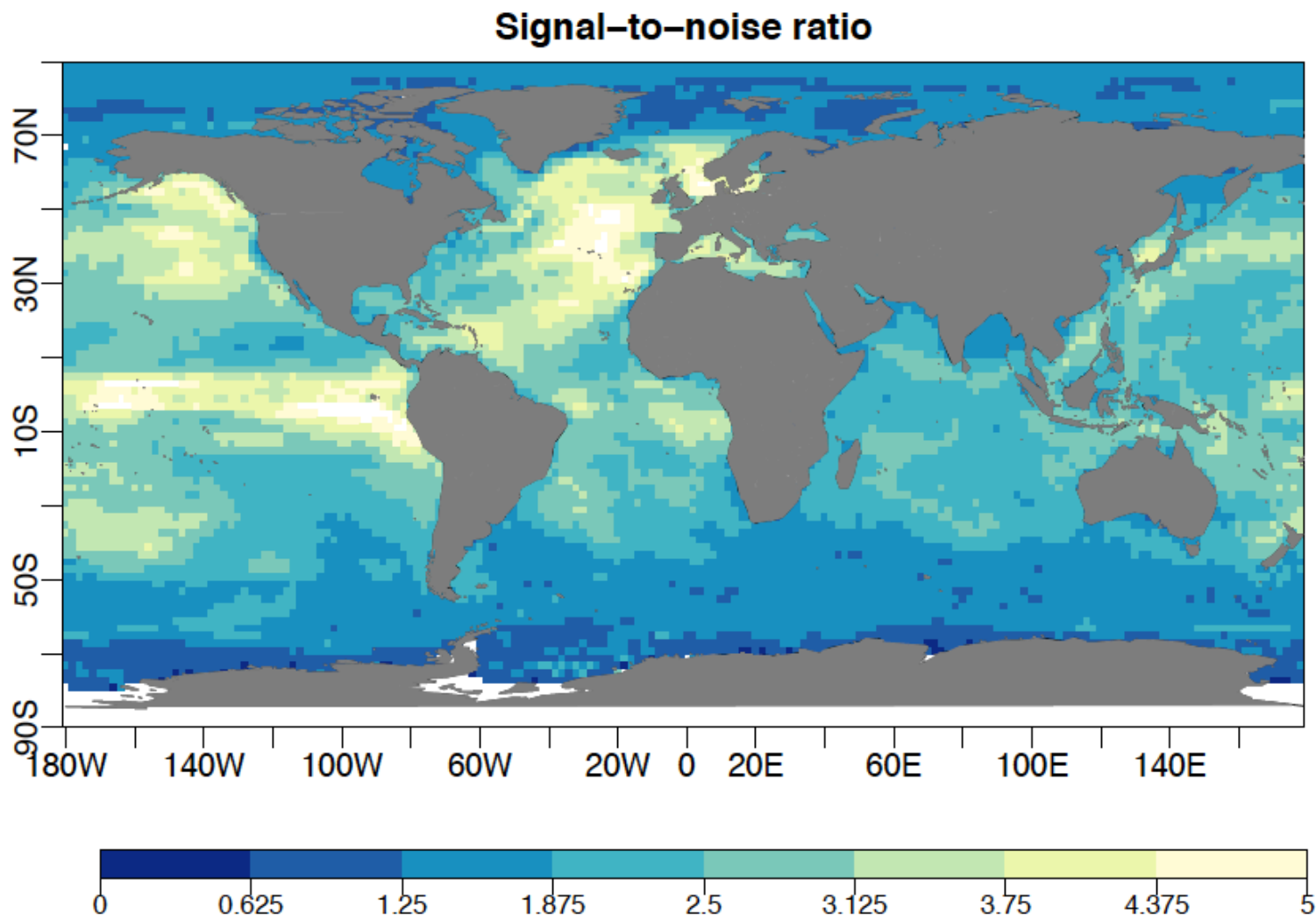
Correlation Uncertainty ENSO



Correlation Uncertainty ENSO

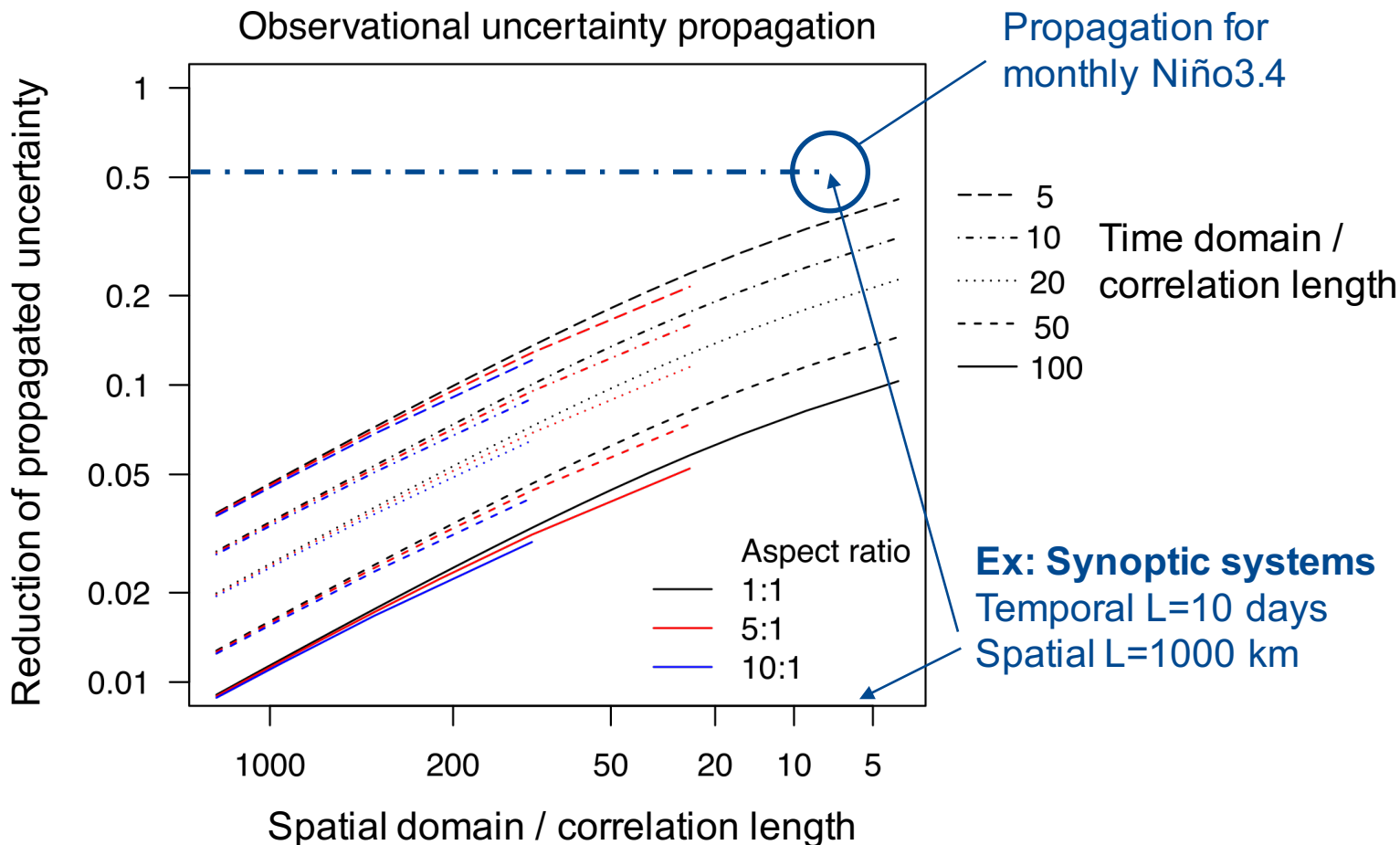


Signal (inter-annual variability) versus observational uncertainty (noise)



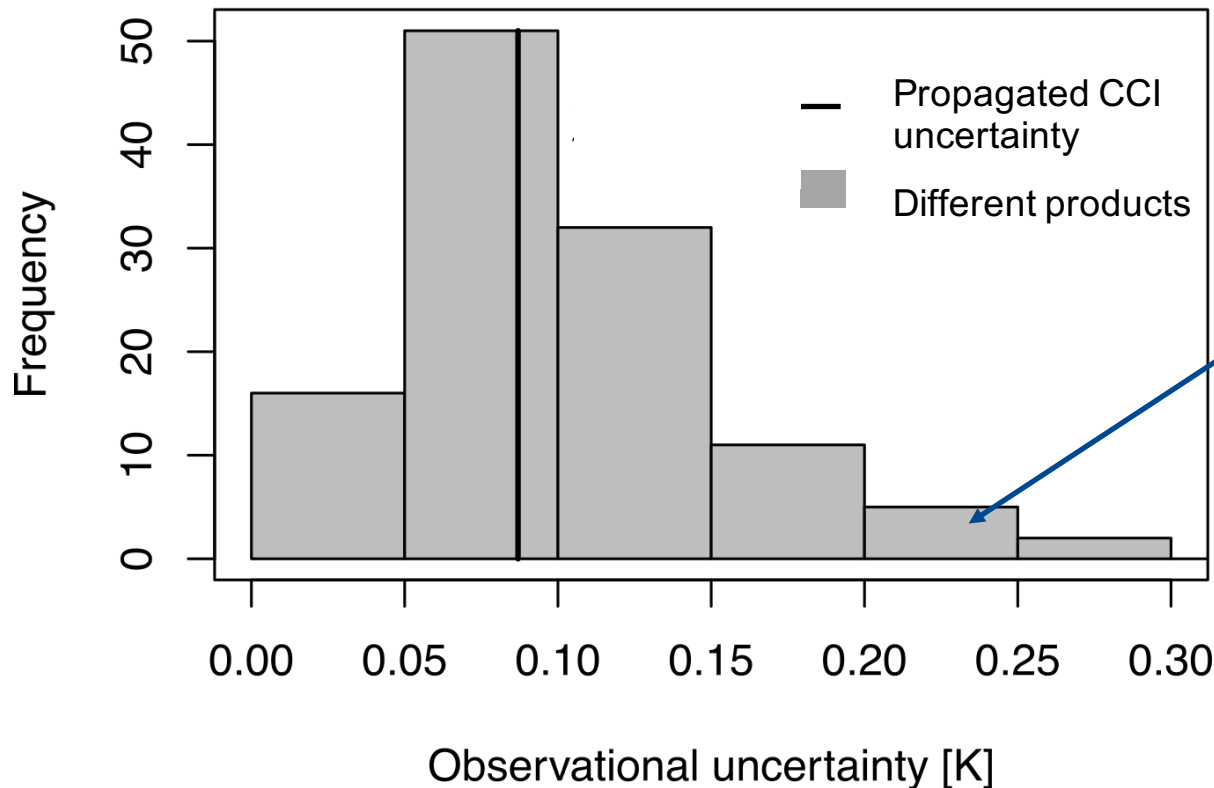
A “look-up” propagation figure

Use of error correlation scales: analytical solution that allows to look-up propagation factors



Propagation assuming synoptic scales (1000 km, 10 days) of weather systems agrees well with deviations between existing products

Observational uncertainty Niño3.4 SST

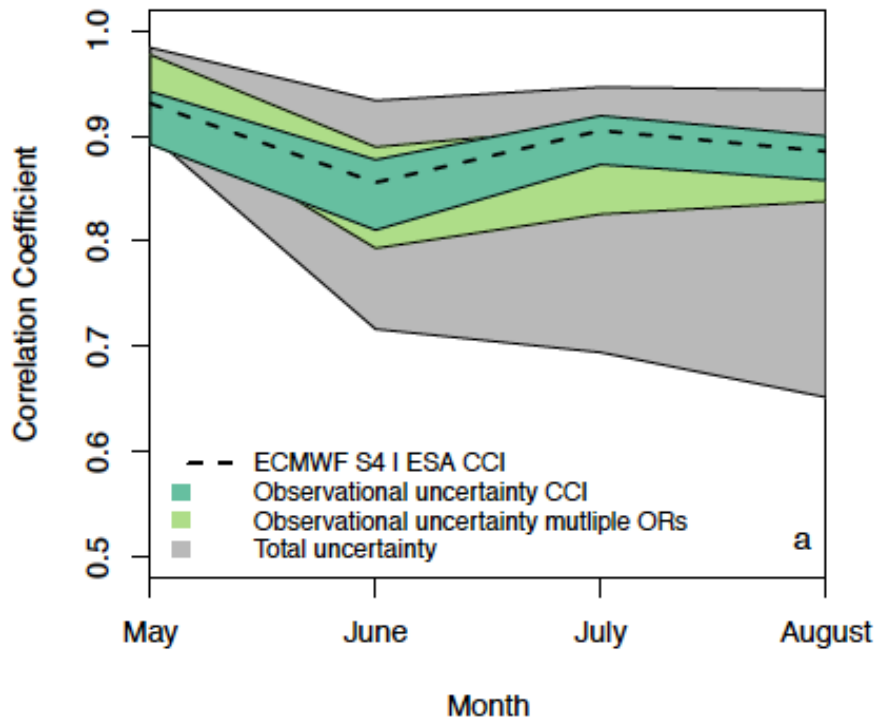


Assumes all observations are equally "likely"

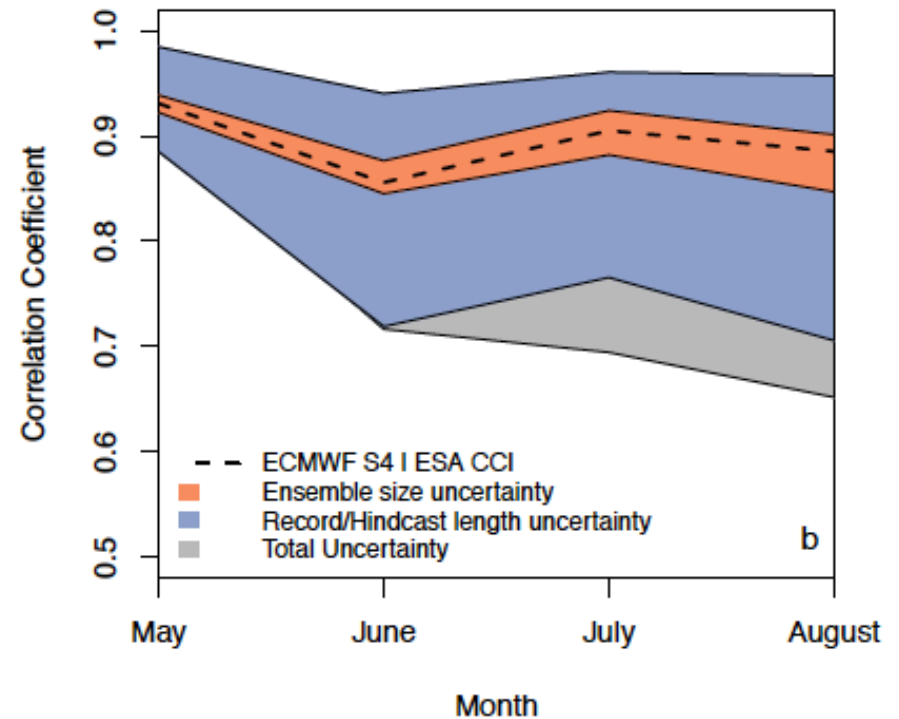
ESA CCI Uncertainty estimate



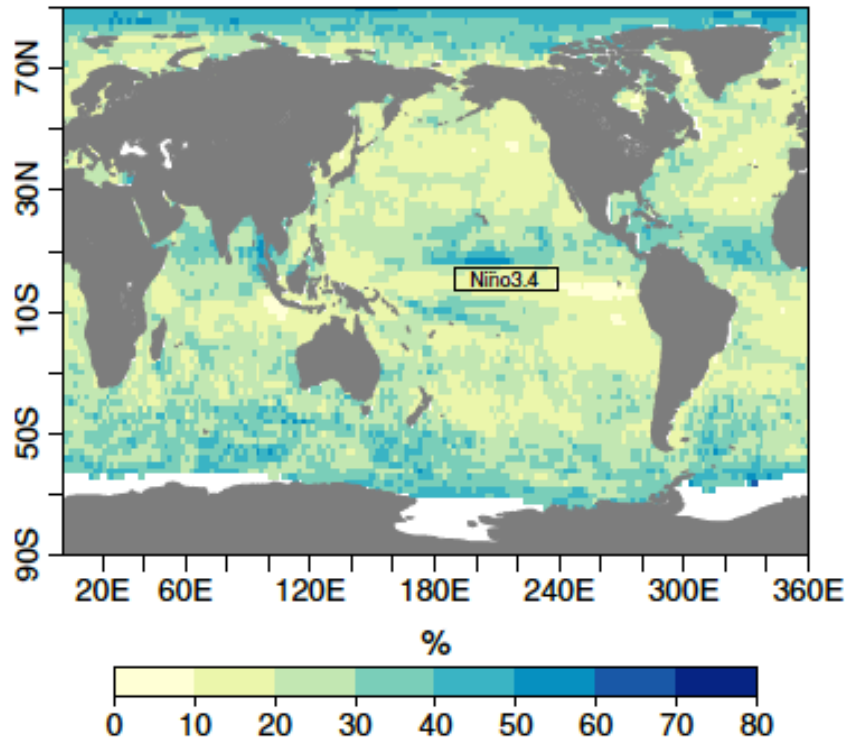
ENSO Prediction (Observational uncertainty)



ENSO Prediction (Sampling Uncertainty)



Observational uncertainty



Ensemble size uncertainty

