

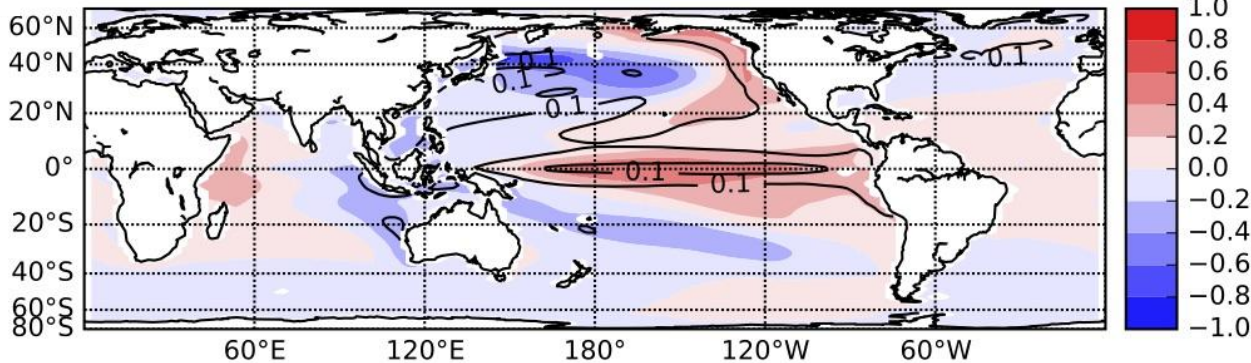
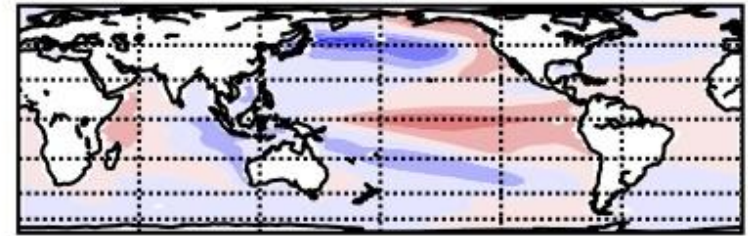
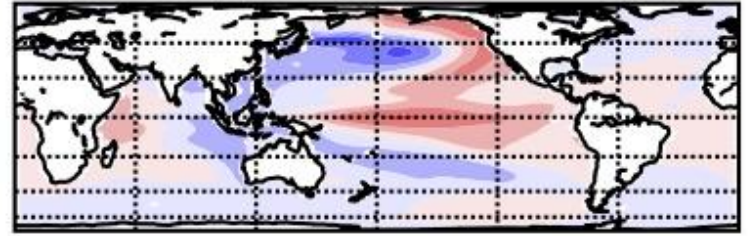
Using ensemble EOFs to identify ENSO precursors in the CESM

Naomi Goldenson
Andy Rhines
Cecilia Bitz
Malte Stuecker

CESM Workshop - June 2017

Motivation for Ensemble EOFs:

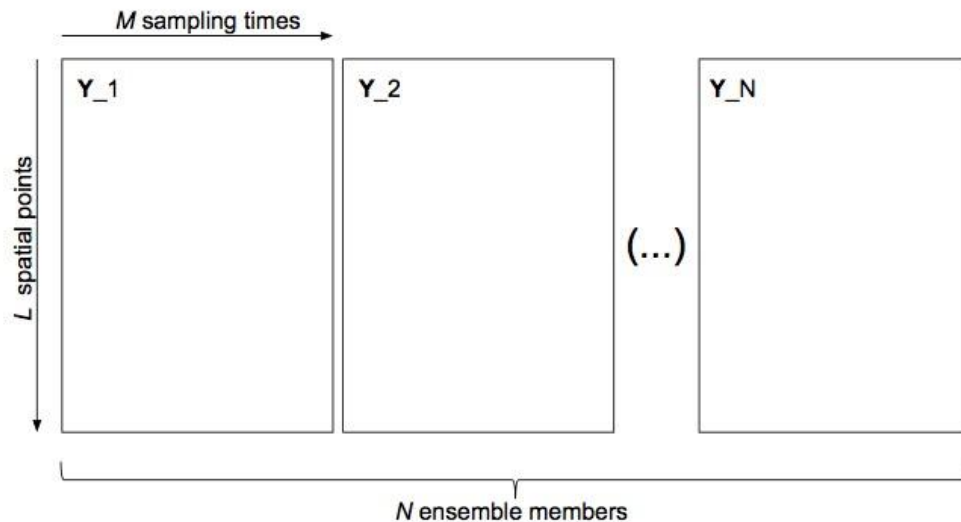
The PDO, calculated separately for each member of the large ensemble, can vary quite a bit.



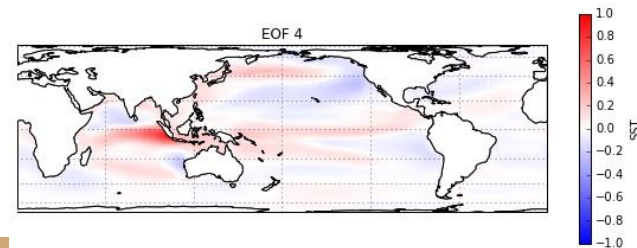
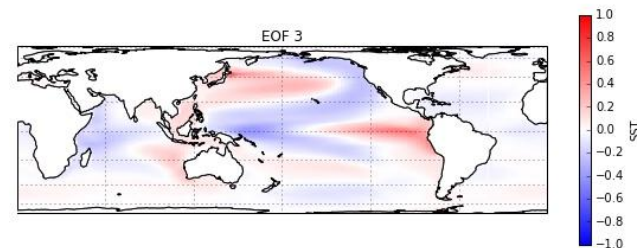
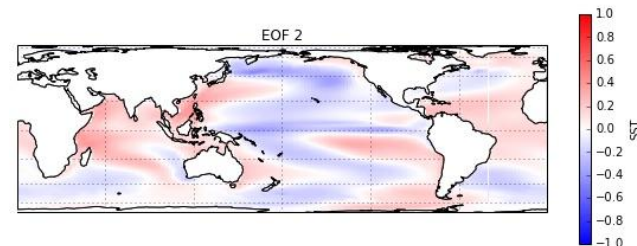
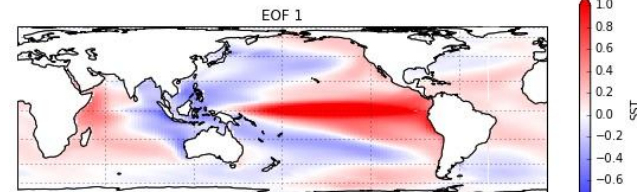
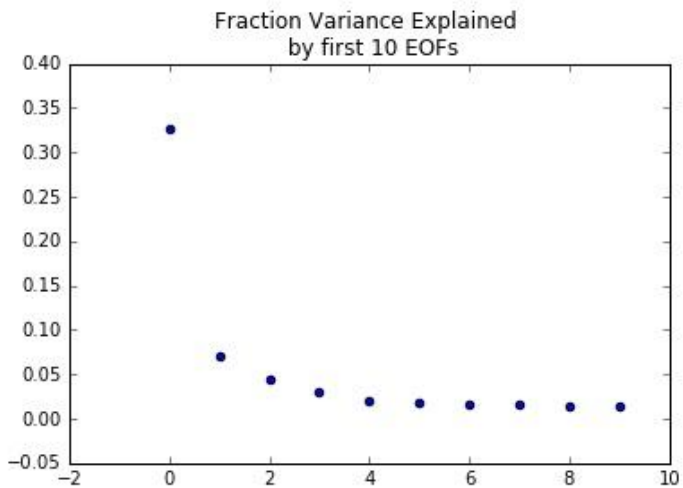
Ensemble EOFs

We stack 40 members of the CESM large ensemble to find one consistent set of “ensemble EOFs”.

- Global SSTs
- Ensemble mean trend removed
- Ensemble mean seasonal cycle removed from monthly data in blocks of 30 years.



First few patterns: ENSO and ... ?

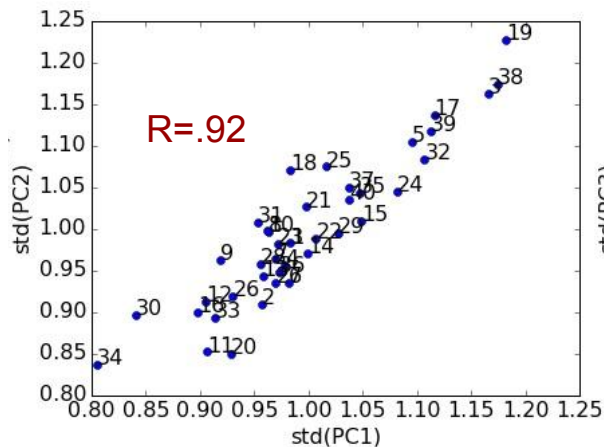


“partial eigenvalues” across ensemble members

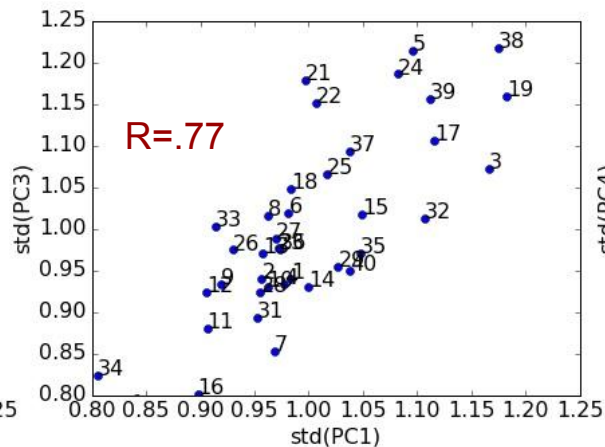
correlation between the magnitude of variance between...

	PC1	PC2	PC3	PC4
PC1	1	0.92	0.77	0.79
PC2		1	0.77	0.65
PC3			1	0.47

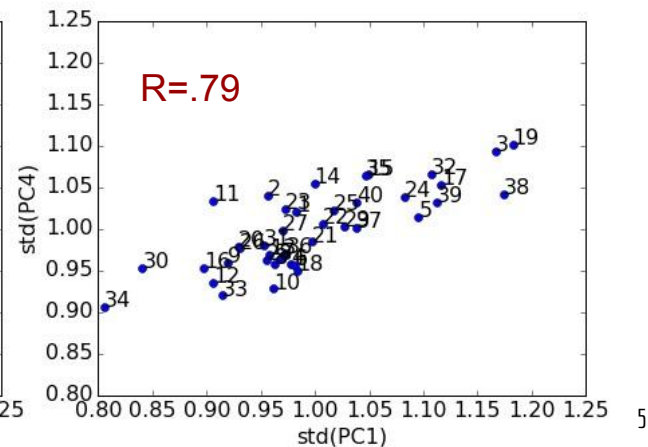
the 1st and 2nd PCs



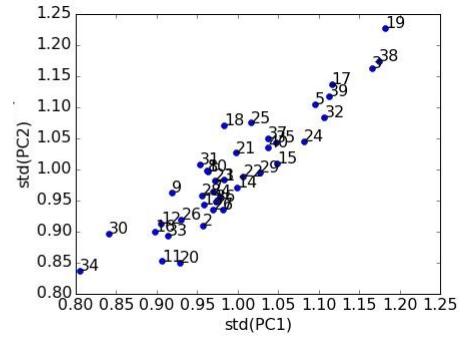
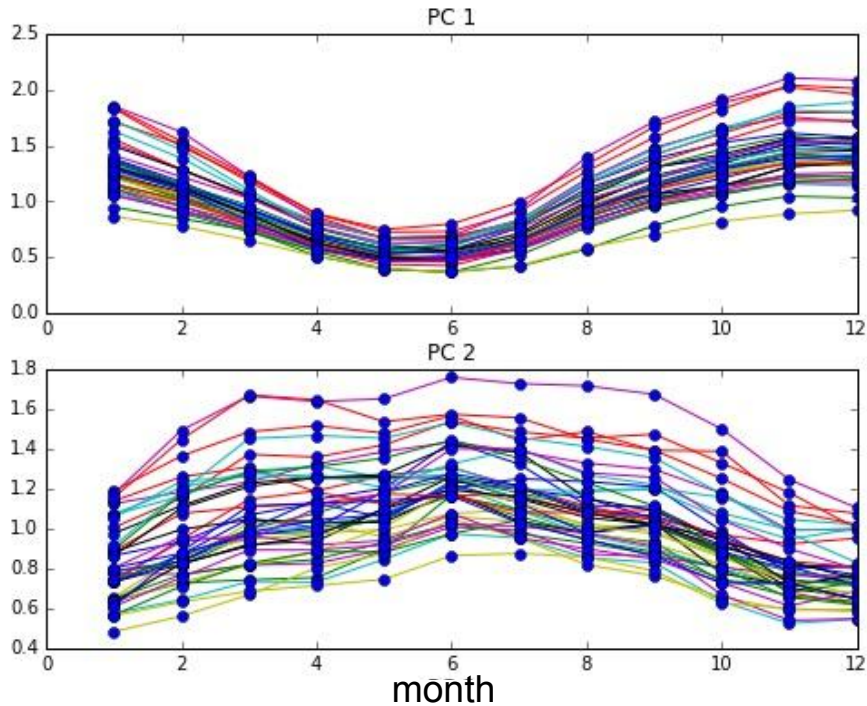
the 1st and 3rd PCs



the 1st and 4th PCs



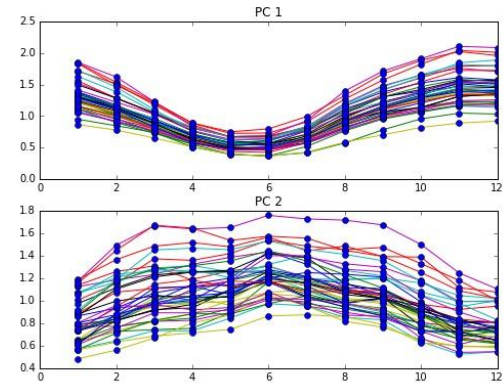
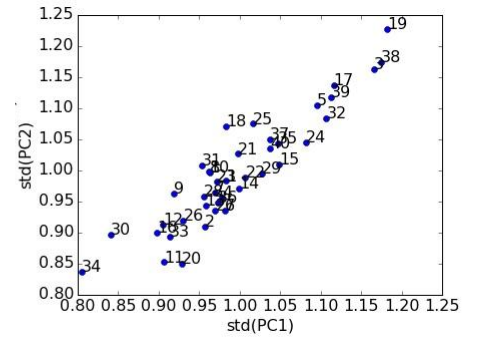
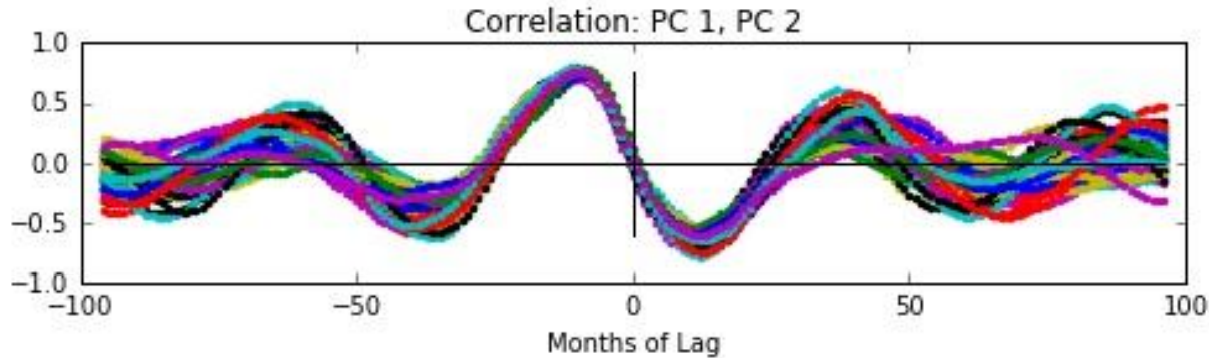
What is going on here?



PC2 has more variance in
Northern summer

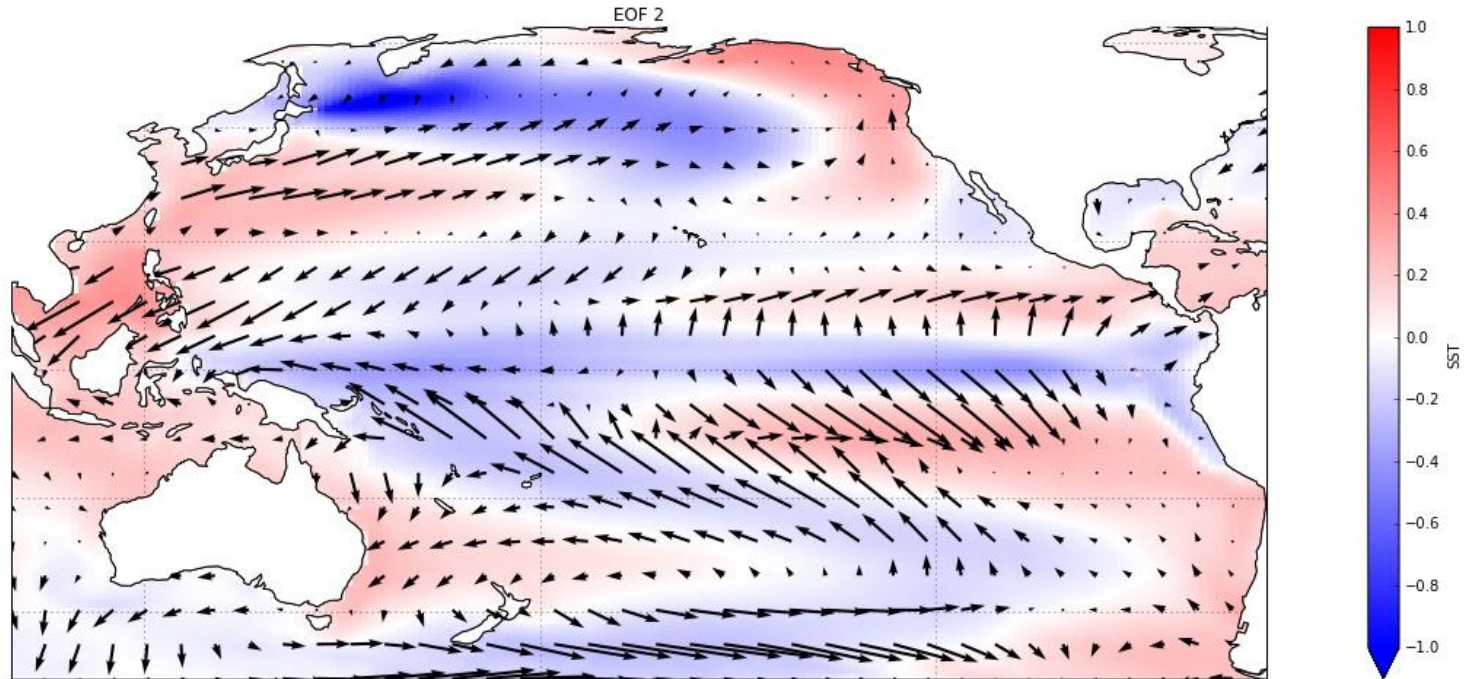
What is going on here?

PC2 and PC1 are strongly cross-correlated:

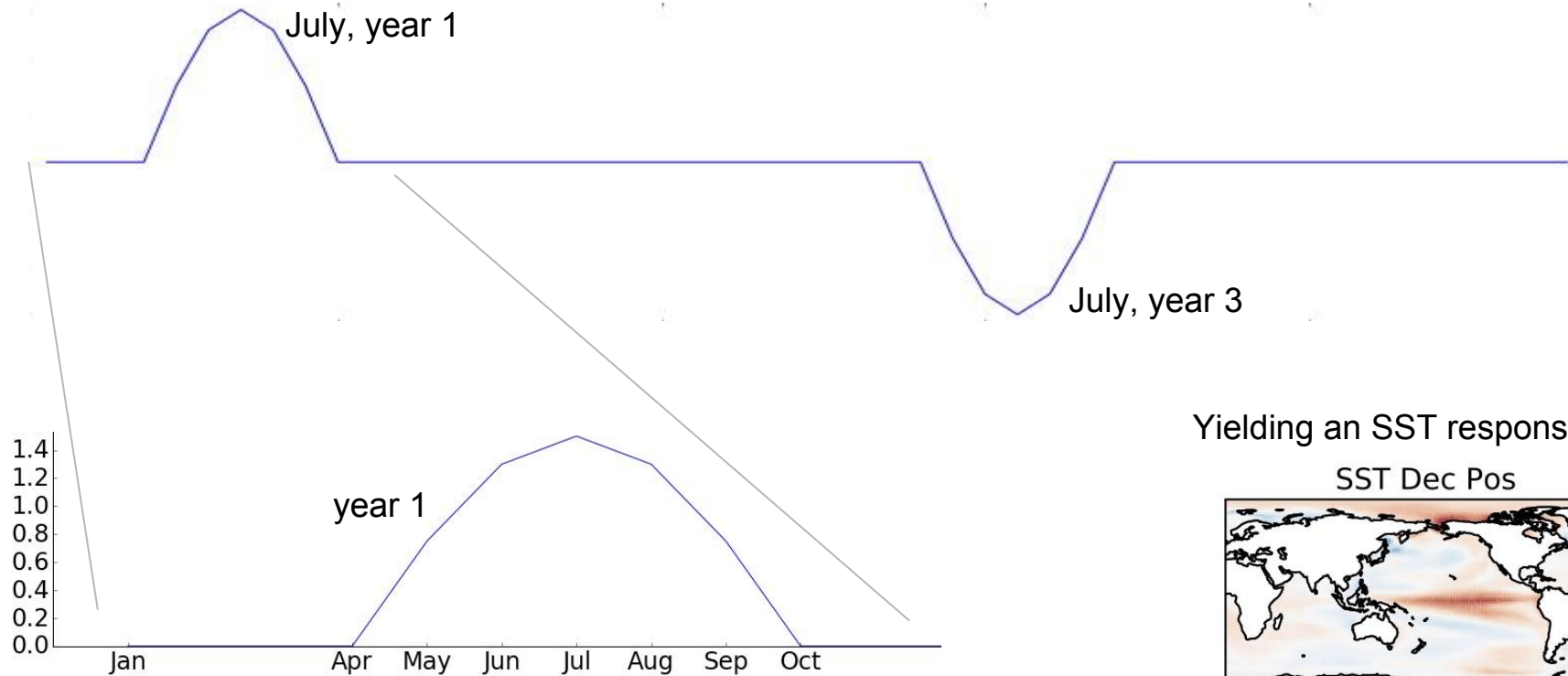


Can we induce an ENSO-like response?

We regress wind stress on PC2...

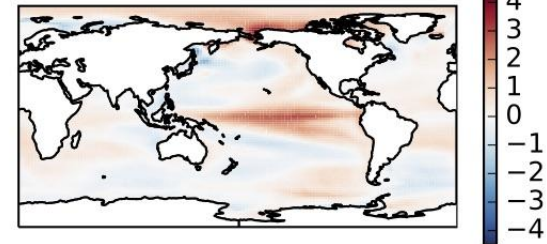


...and drive a cyclic anomaly from climatology

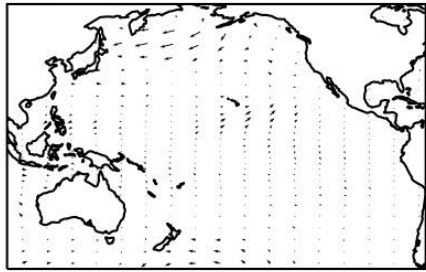


Yielding an SST response:

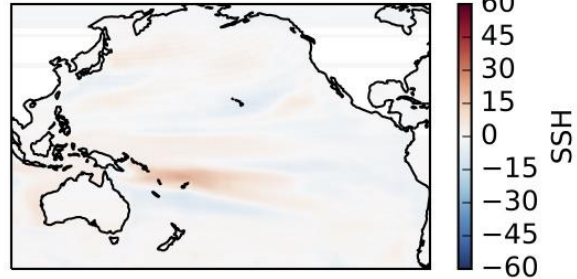
SST Dec Pos



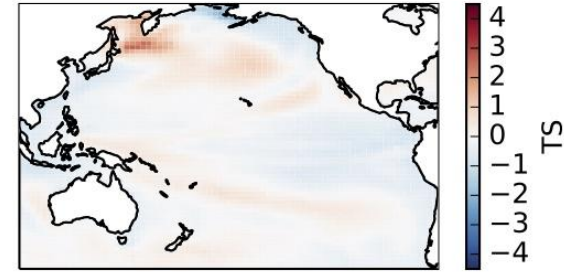
Wind stress anom



SSH anom



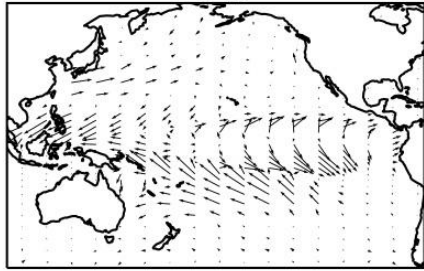
SST anom



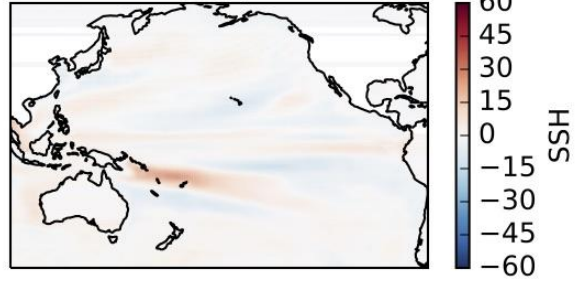
Composite anomalies, April, year 1 of 4 year cycle.

April

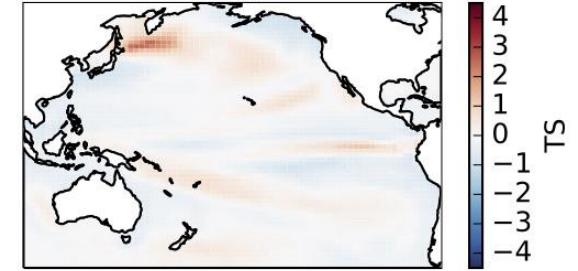
Wind stress anom



SSH anom

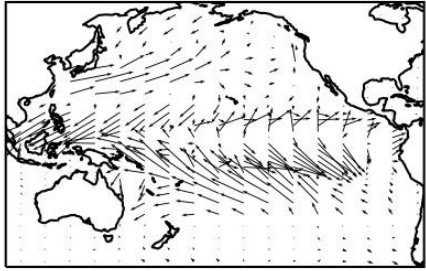


SST anom

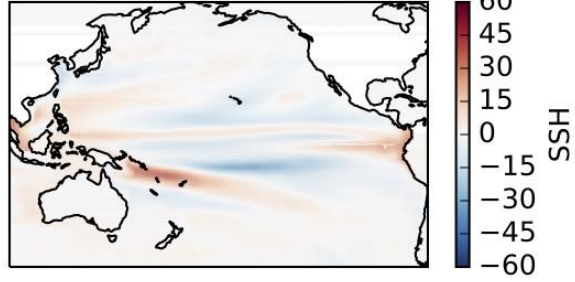


May

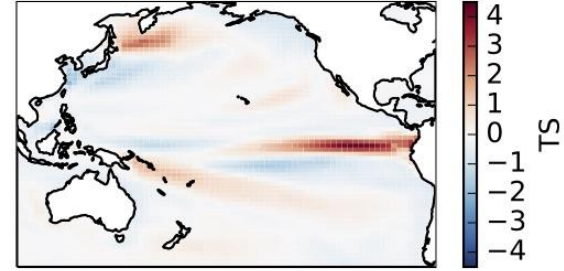
Wind stress anom



SSH anom

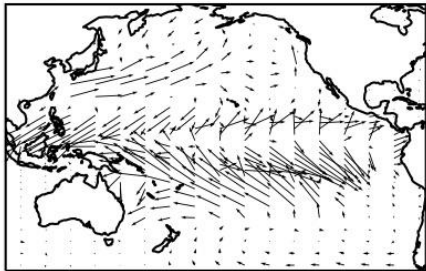


SST anom

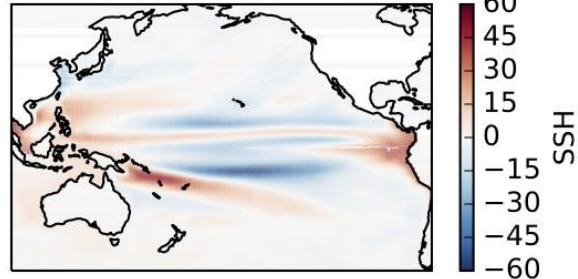


June

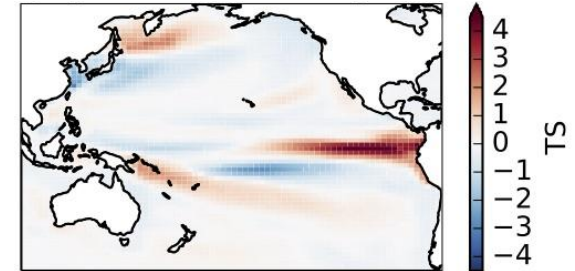
Wind stress anom



SSH anom



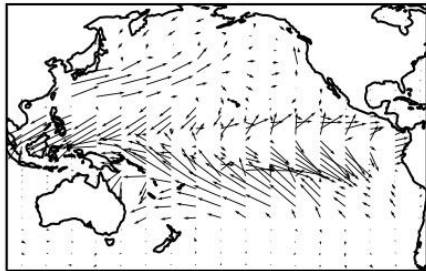
SST anom



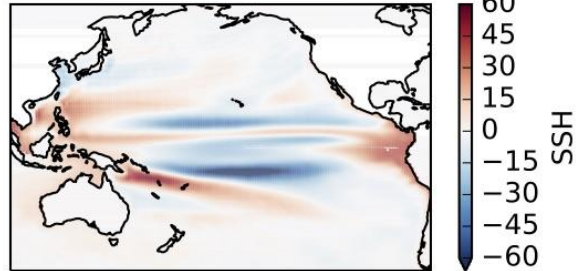
forcing is maximum

July

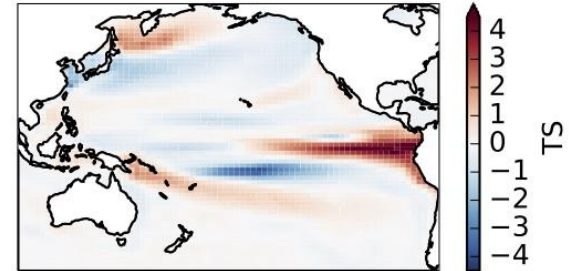
Wind stress anom



SSH anom

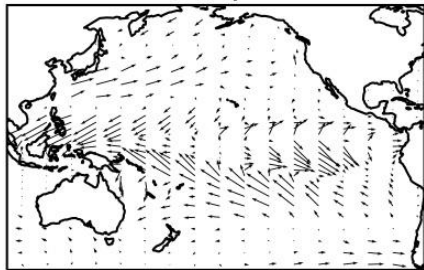


SST anom

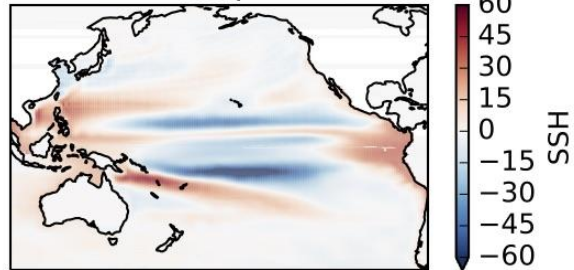


August

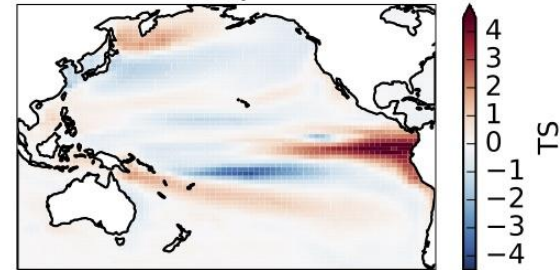
Wind stress anom



SSH anom

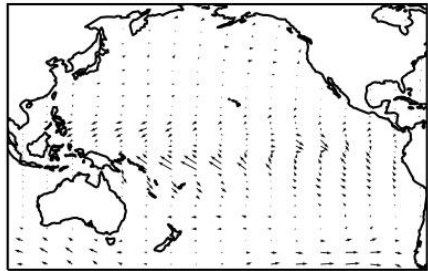


SST anom

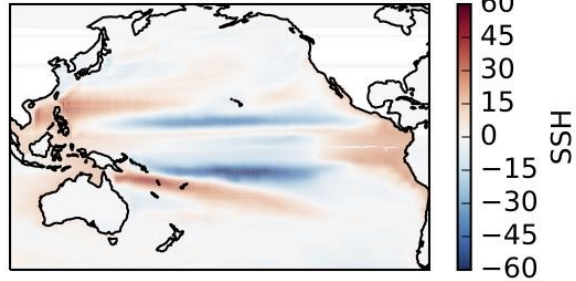


September

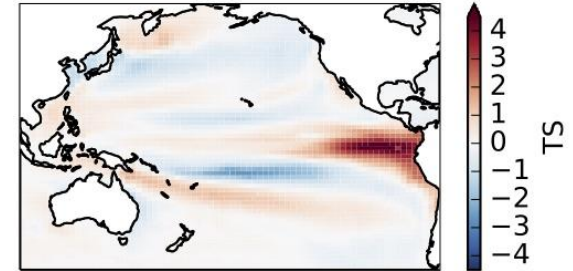
Wind stress anom



SSH anom

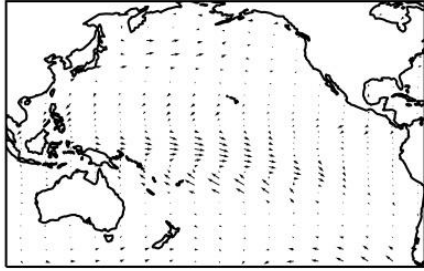


SST anom

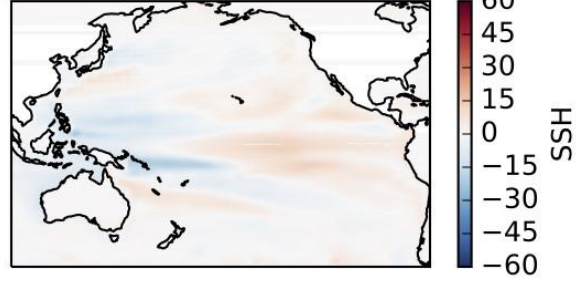


October

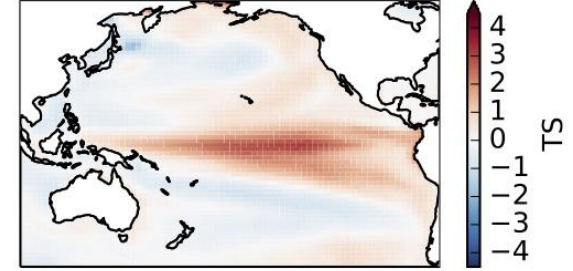
Wind stress anom



SSH anom

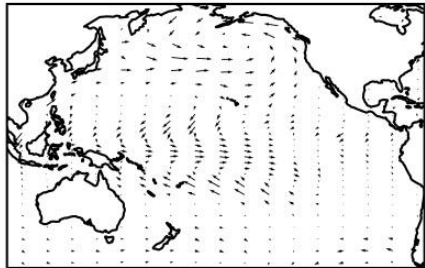


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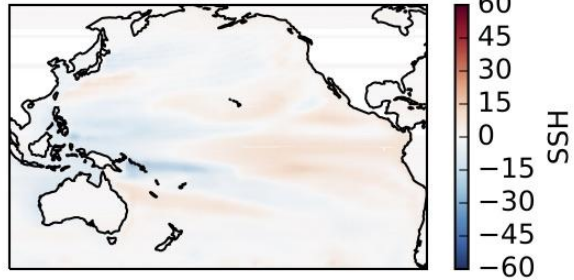


November

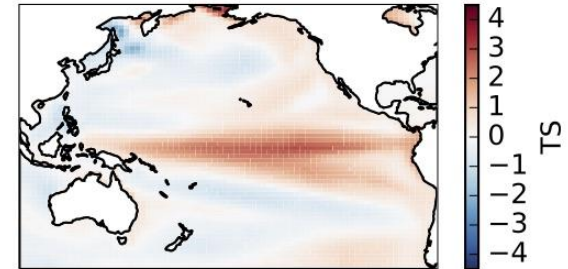
Wind stress anom



SSH anom

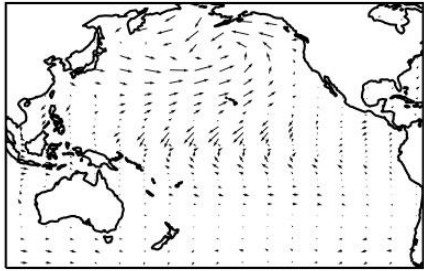


SST anom

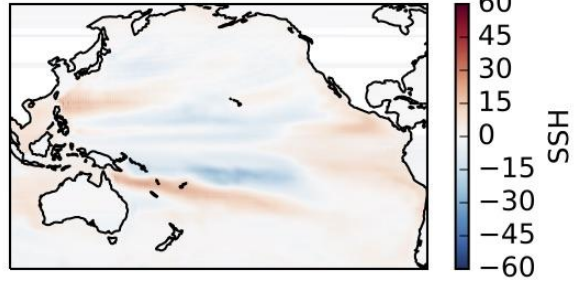


December

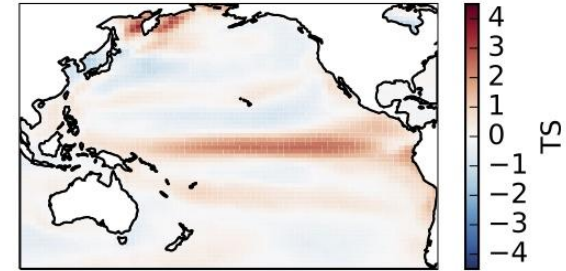
Wind stress anom



SSH anom



SST anom

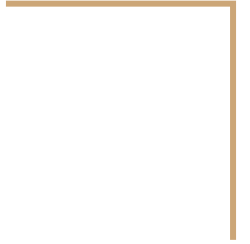


January

Summary

- Ensemble EOFs let us compare variance across ensemble members.
- Variance associated with precursor and decay of El Niño appears in EOFs after the 1st.
- These are not necessarily synonymous with physical patterns of SSTs that lead and lag El Niño.
- Wind stress associated with PC2, which leads PC1, can be used to generate an El Niño-like response.

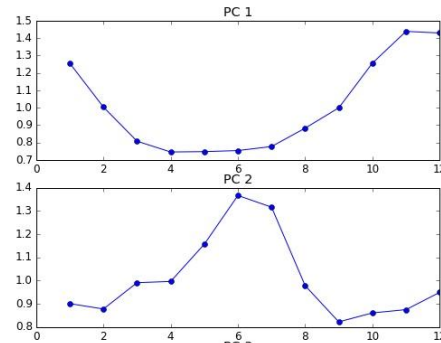
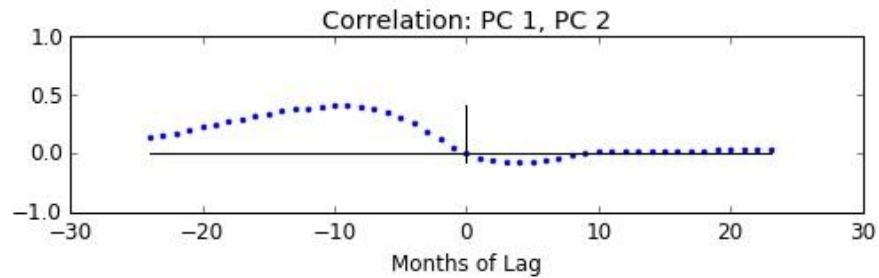
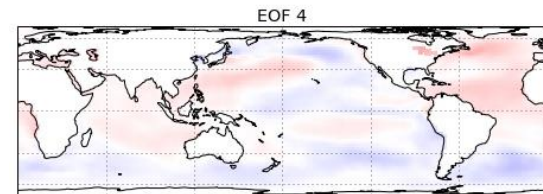
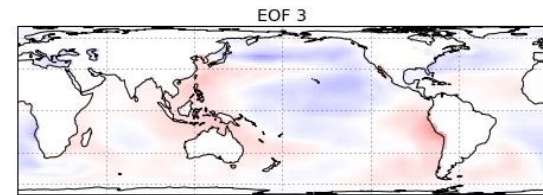
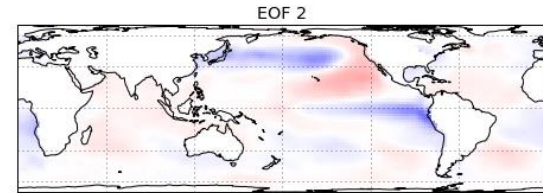
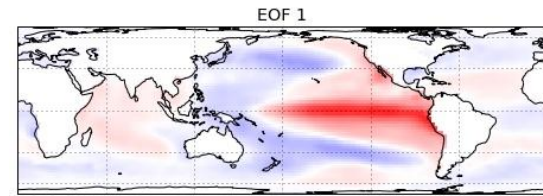
Extra Slides



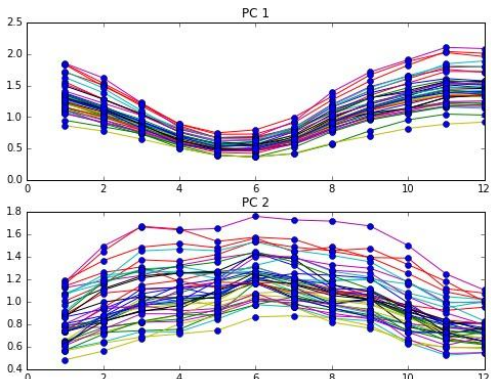
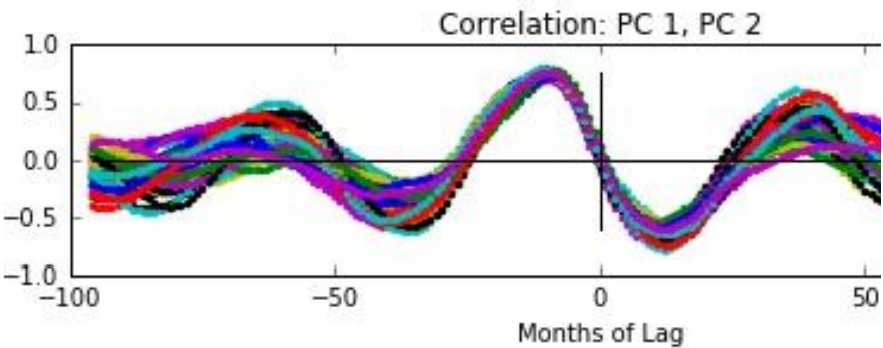
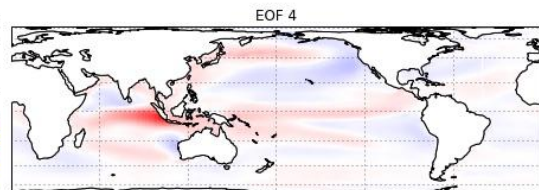
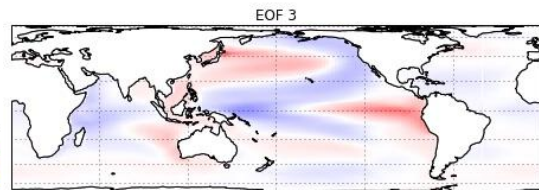
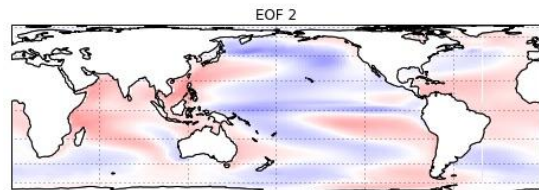
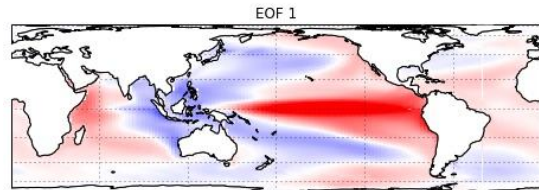
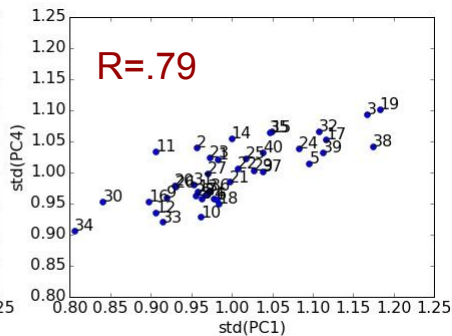
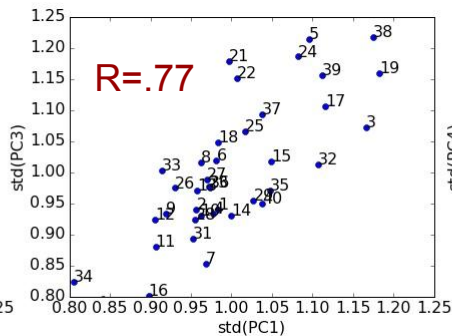
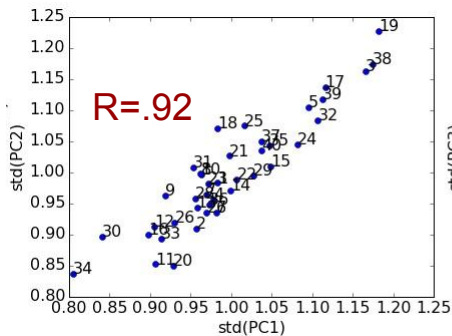
To calculate EOFs on a very large matrix...

- where “very large” means 31200 x 20755 (it occupies 9.5 GB in a .csv file)
- Incremental PCA saves the day ([Ross et al.](#))
- I use an implementation in the Python library: [SciKit-Learn](#)

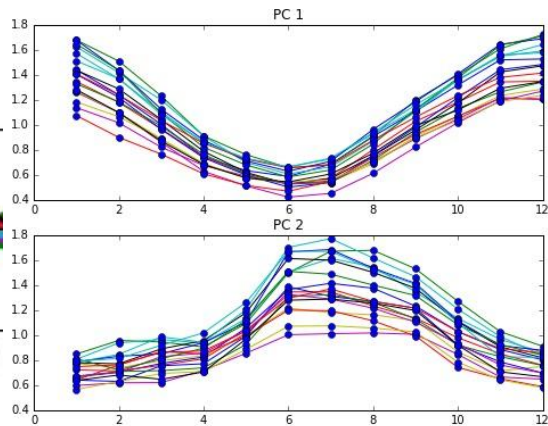
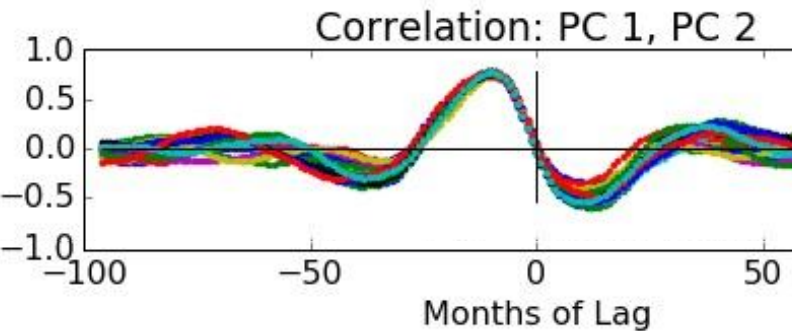
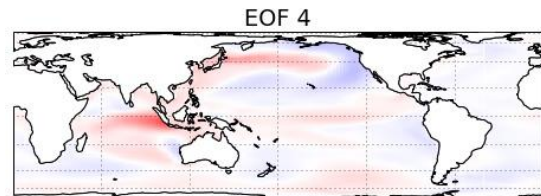
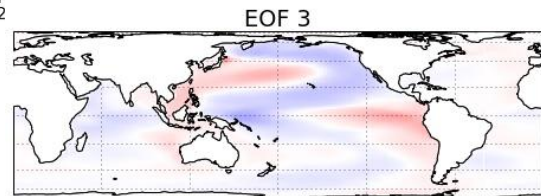
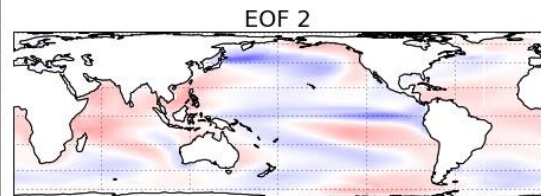
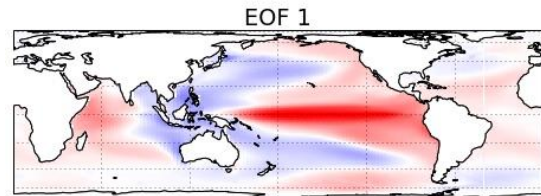
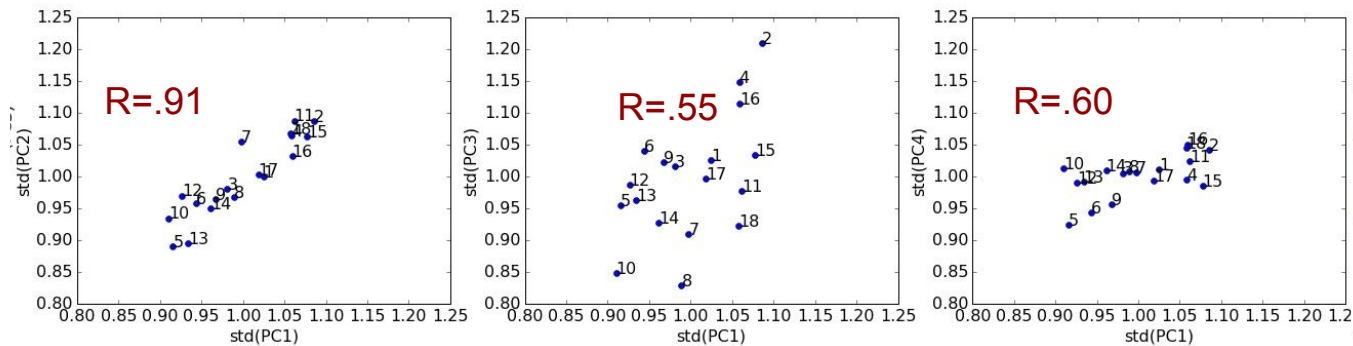
Observations (NOAA's ERSST)



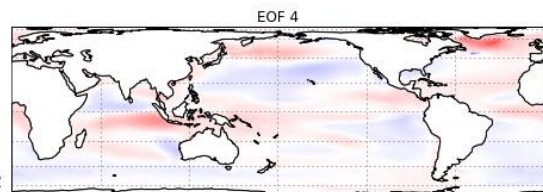
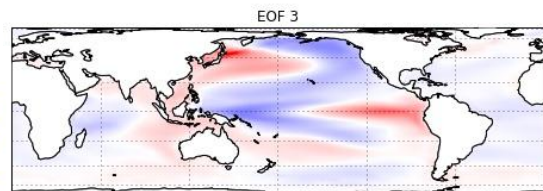
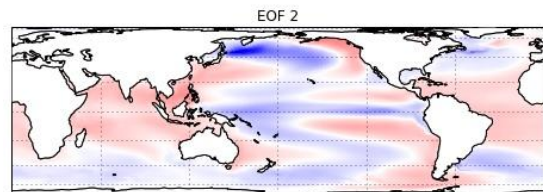
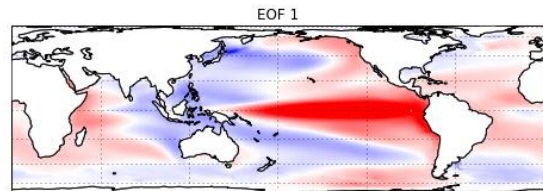
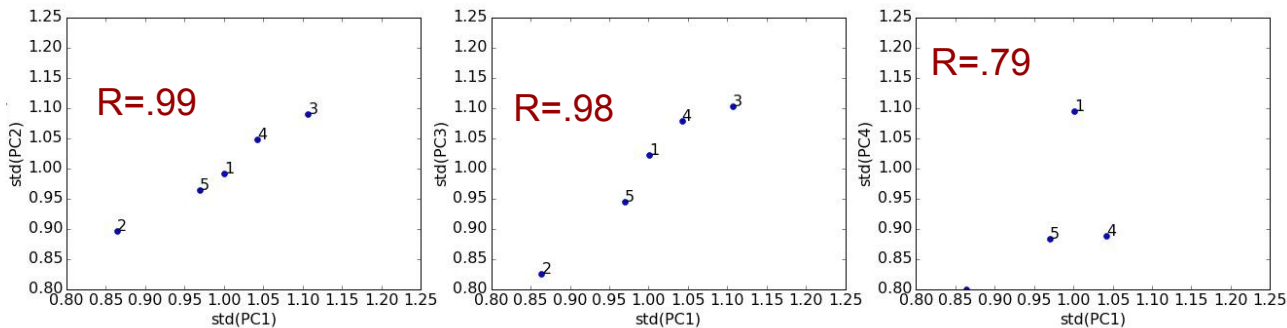
CESM LE summary slide



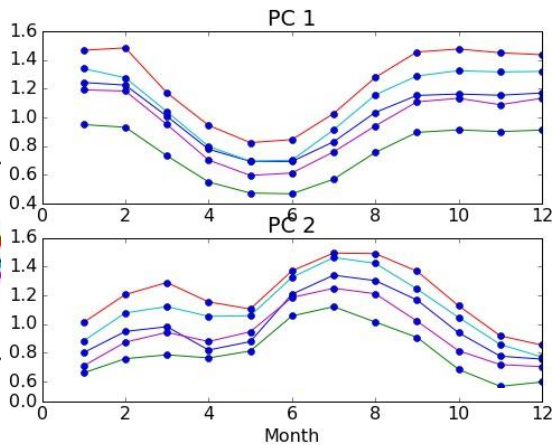
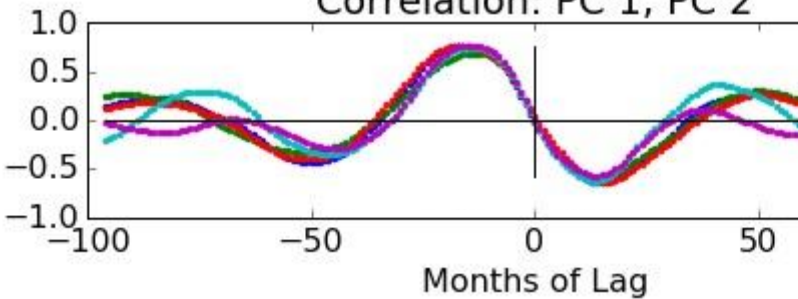
The long control run of CESM.



The control run of CCSM4 2deg.

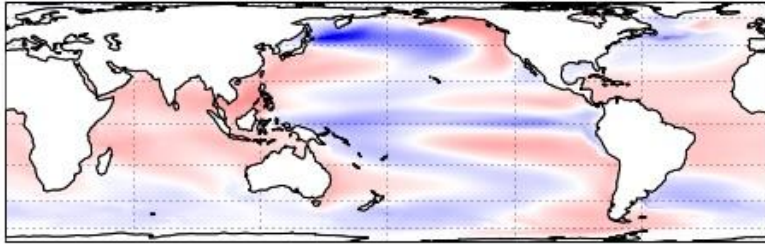


Correlation: PC 1, PC 2

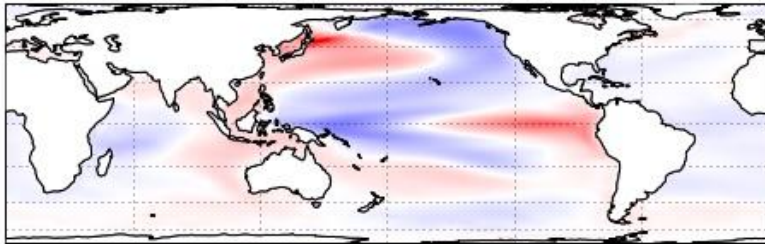


Finding more physical patterns?

EOF 2

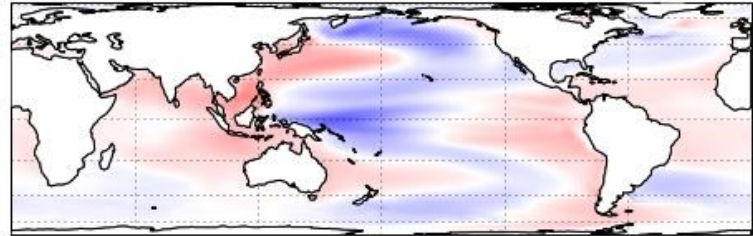


EOF 3



45° rotations

Rot PC2 and 3 +



Rot PC2 and 3 -

