

Interpretation of CCA Loading Patterns and the Associated Time Series

Multiple Regression, Principal Component Regression, and CCA

Prediction of any element **Y** individually using **multiple regression**

Predictors can be elements of **X**

$$y_1 = b_1x_1 + b_2x_2 + b_3x_3 + \dots$$

$$y_2 = b_1x_1 + b_2x_2 + b_3x_3 + \dots$$

$$y_3 = \dots$$

OR

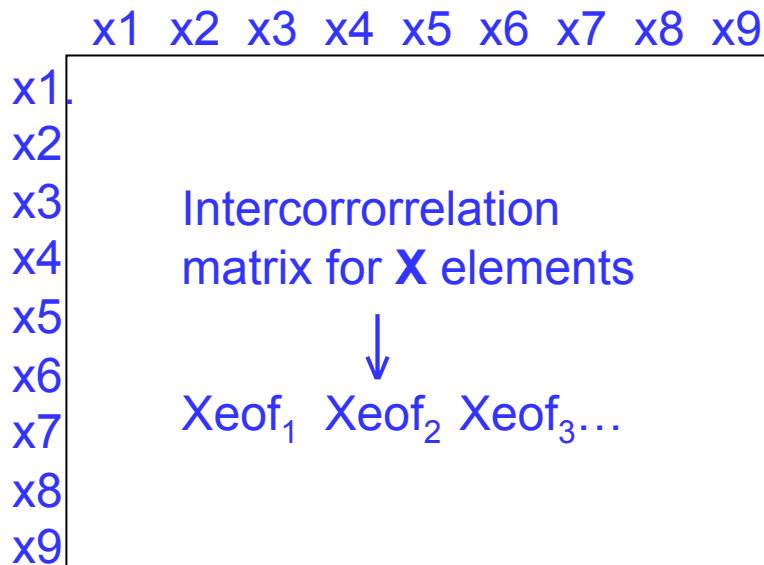
Predictors can be EOFs of **X**

(principal component regression)

$$y_1 = b_1XEOF_1 + b_2XEOF_2 + \dots$$

$$y_2 = b_1XEOF_1 + b_2XEOF_2 + \dots$$

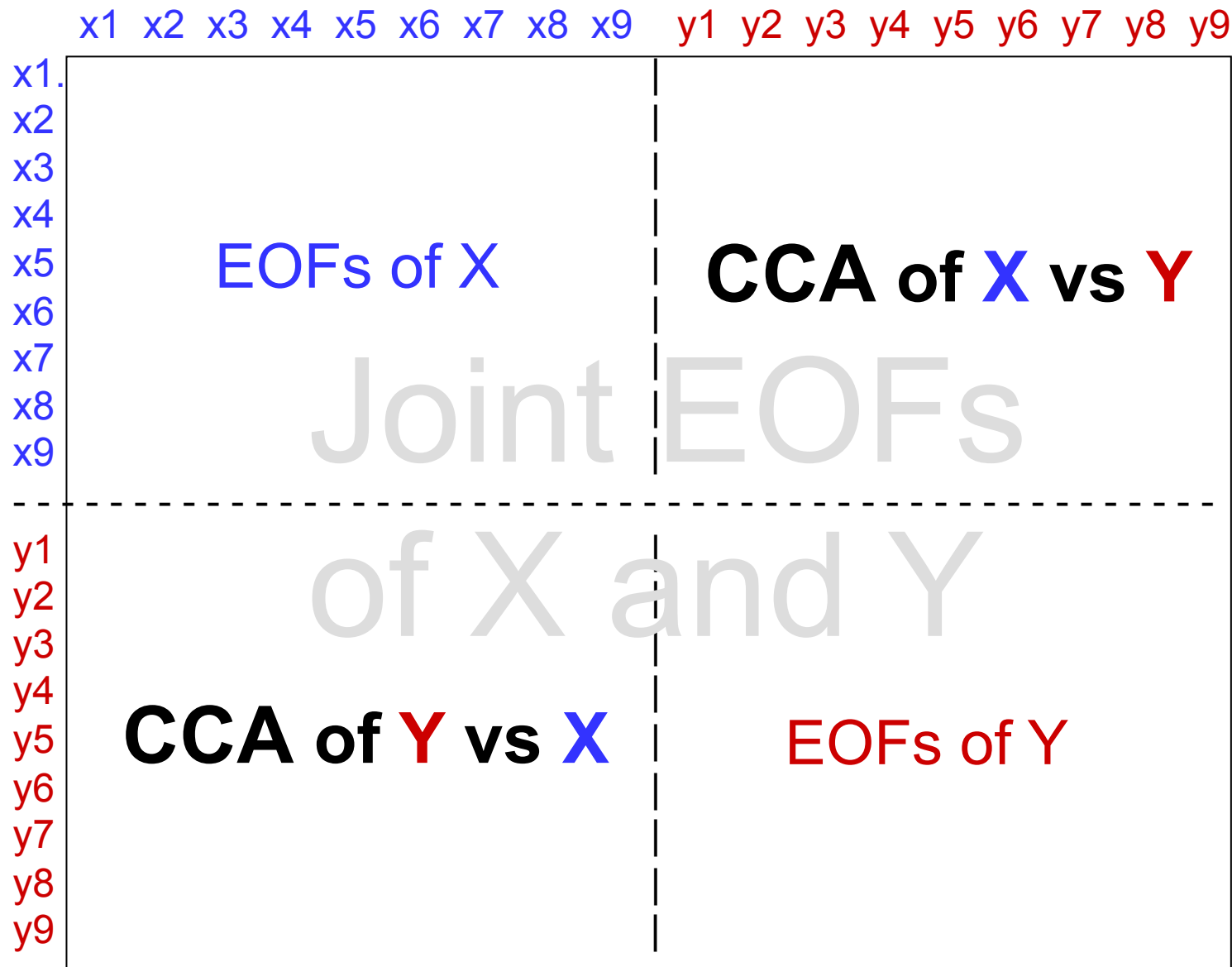
$$y_3 = \dots$$



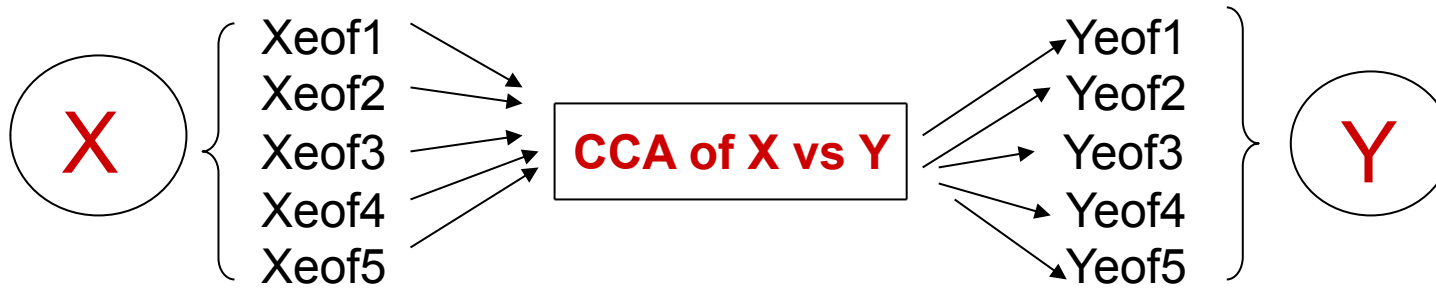
Introduction to CCA

CCA is like EOF analysis, except that there are **TWO data sets** (**X** and **Y**, or predictor and predictand), and the input matrix (correlation or covariance) contains **cross-dataset coefficients only** (an **X** element with a **Y** element; no **X-to-X** or **Y-to-Y**). Both **X** and **Y** can be time-extended or contain multiple fields.

Analyses of correlation matrix of **X** and **Y** fields: 9 elements of **X** and **Y**

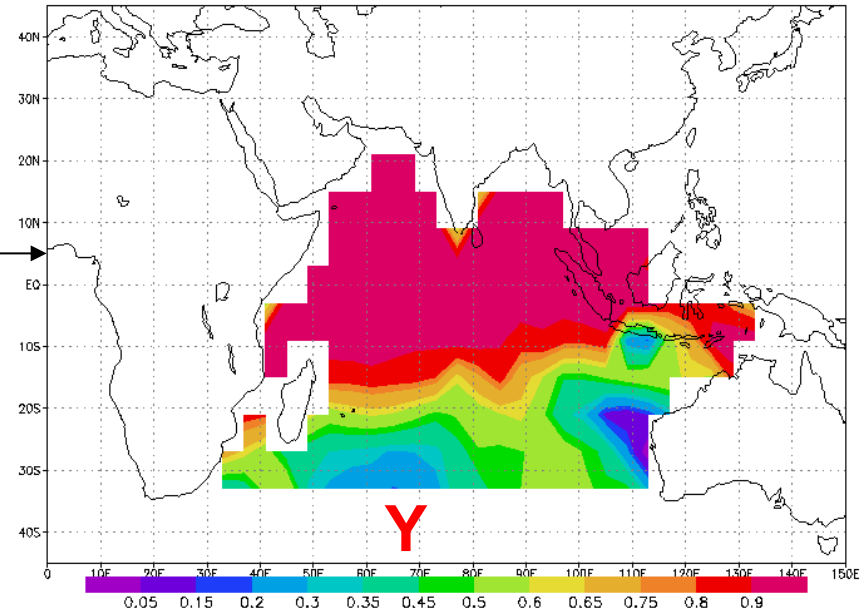
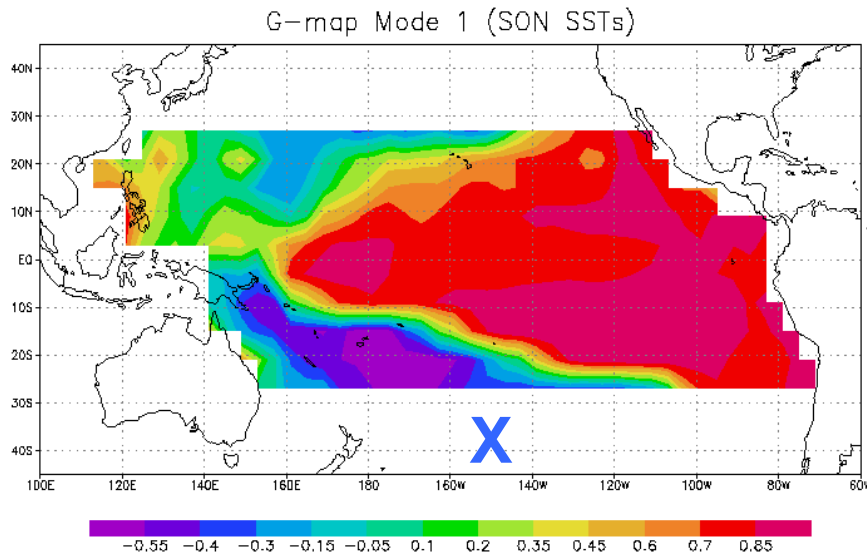


CCA can be done using just certain EOF time series of X and of Y:



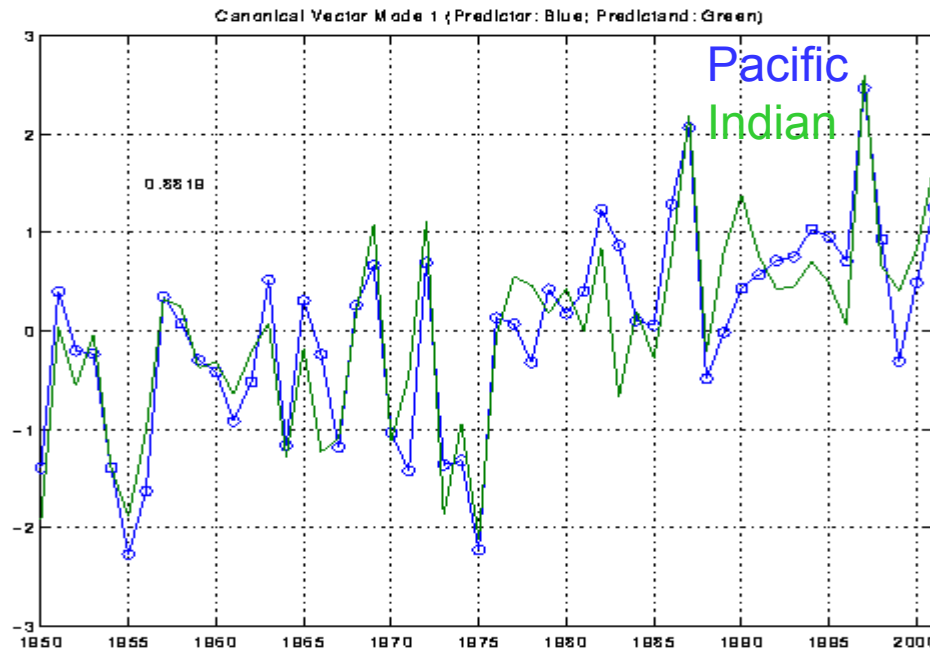
CCA mode 1

H-map Mode 1 (DJF SSTs)



X: tropical Pacific
SST, SON season

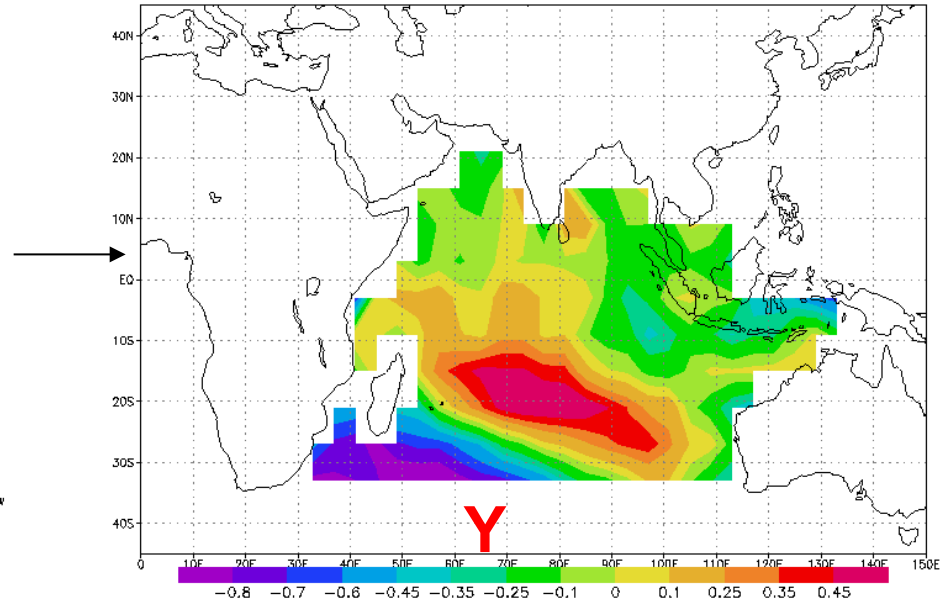
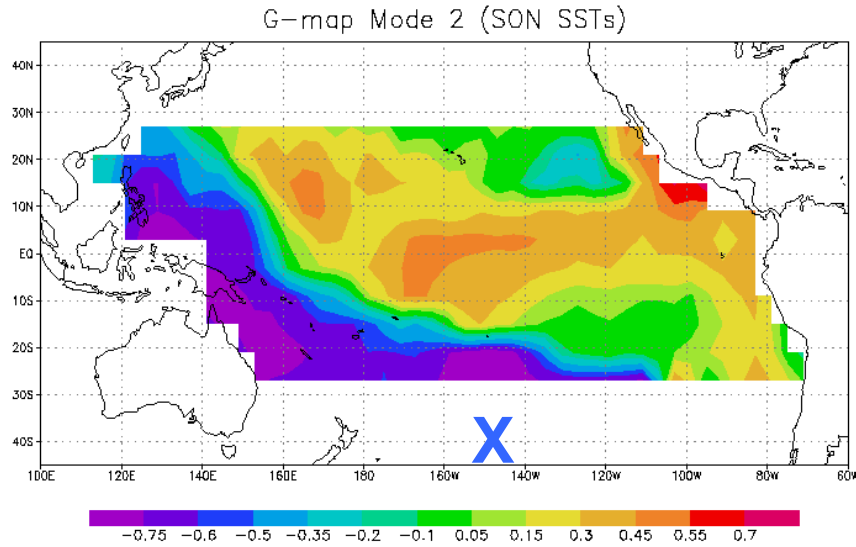
Y: Indian Ocean
SST, DJF season



canon
correl
=.882

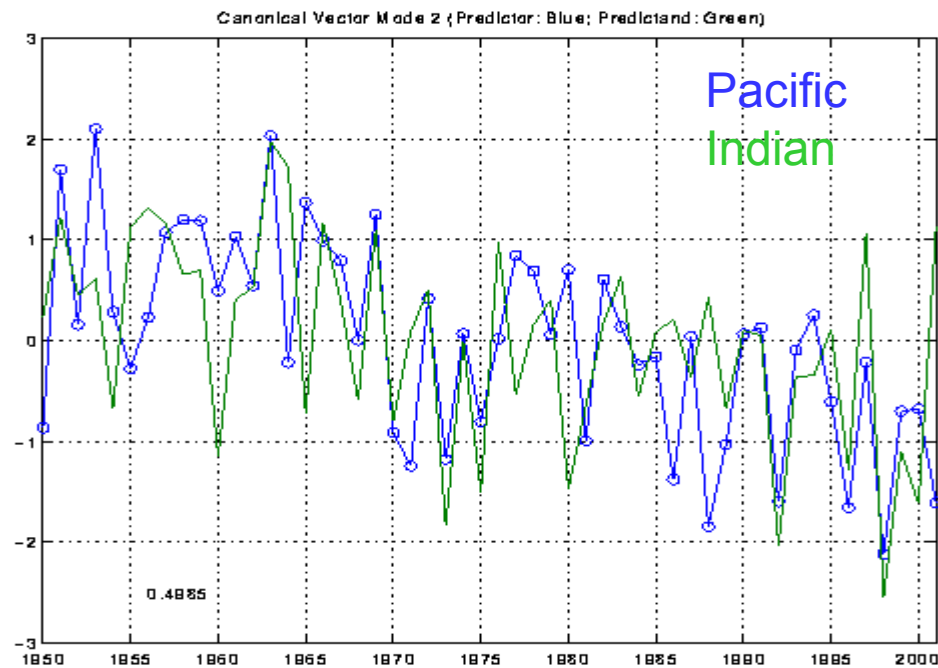
CCA mode 2

H-map Mode 2 (DJF SSTs)



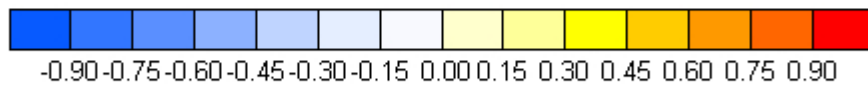
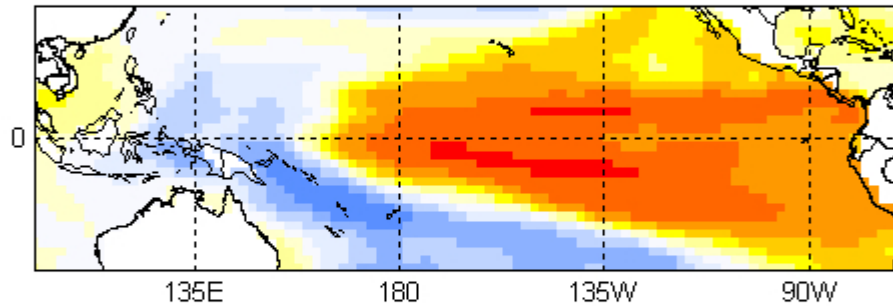
X: tropical Pacific
SST, SON season

Y: Indian Ocean
SST, DJF season

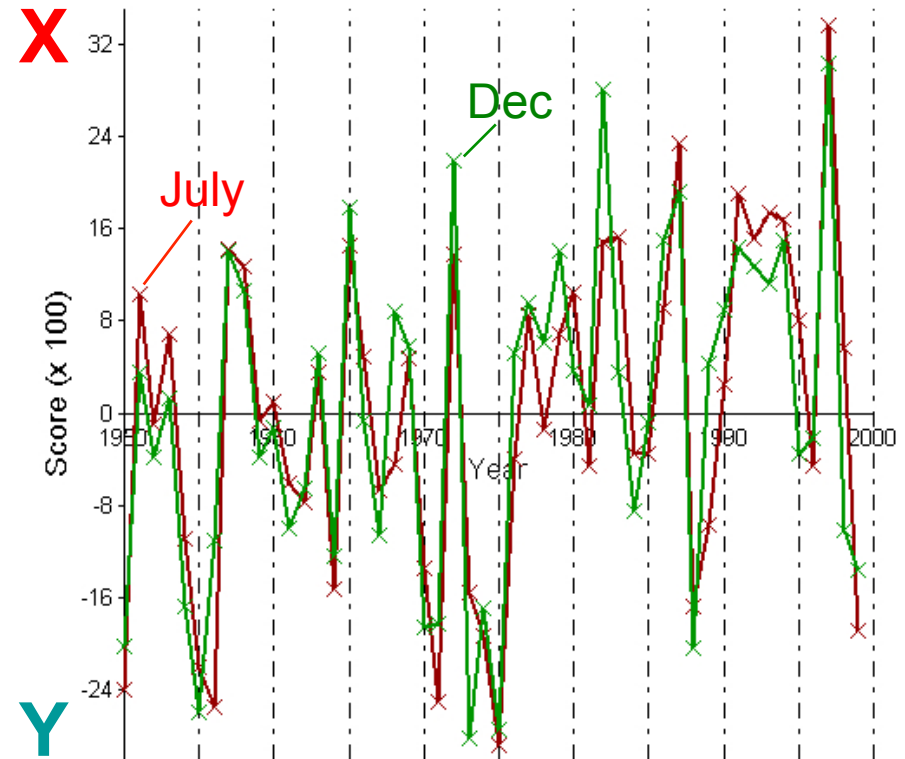
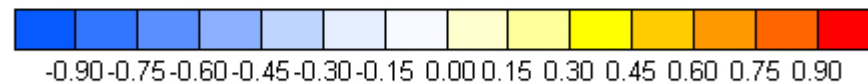
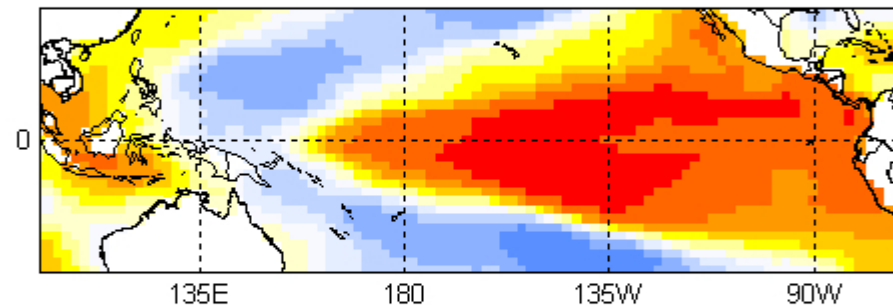


canon
correl
=.498

Another CCA example, using tropical Pacific SST for two separated months: July (X) and December (Y)

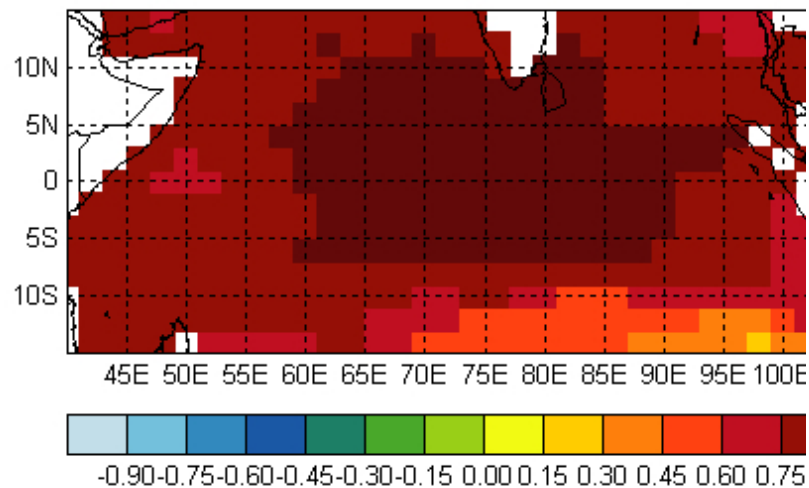


July vs. December sea-surface temperature, 1950-1999

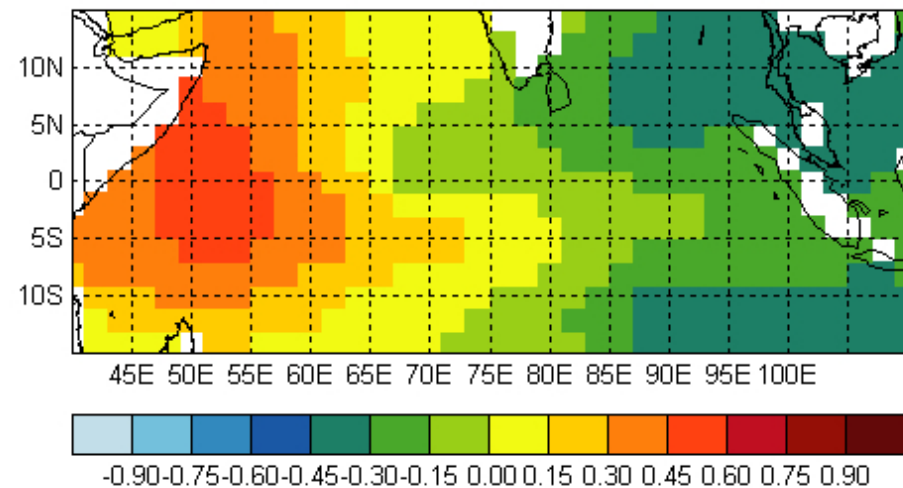


Buell patterns in EOFs:
Are the following Indian Ocean EOF modes real,
or are they caused mainly by the Buell pattern phenomenon?

X Spatial Loadings (EOF1)

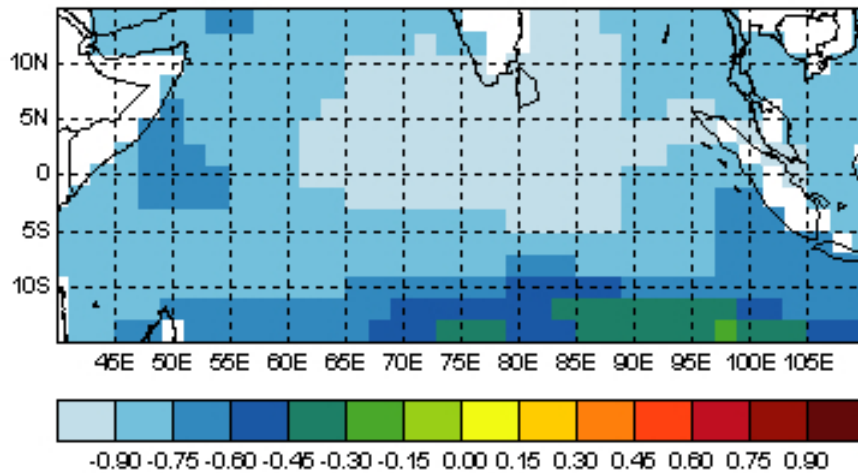


X Spatial Loadings (EOF2)

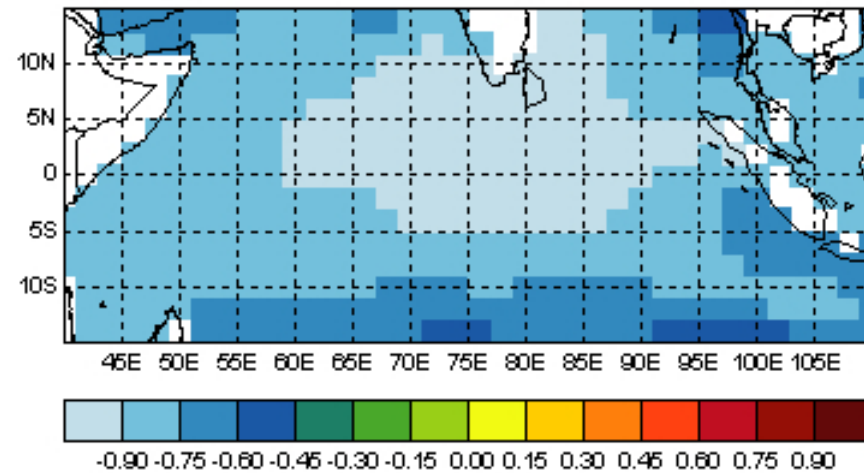


Buell patterns can sometimes also affect CCA

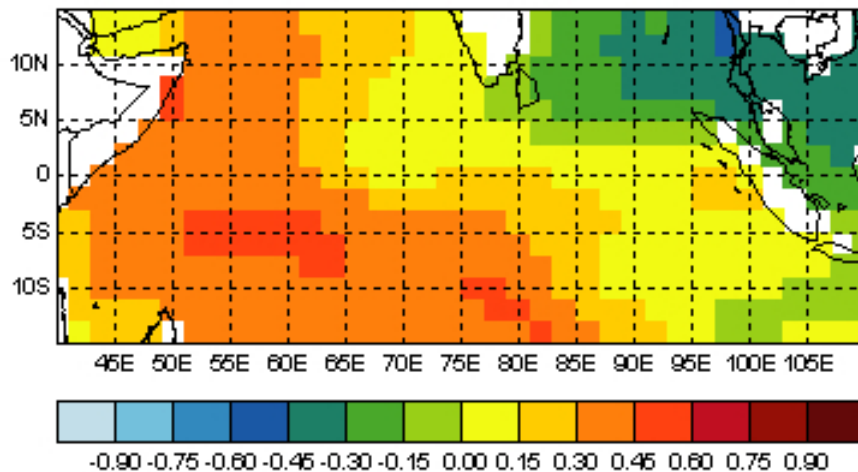
X Spatial Loadings (Mode1)



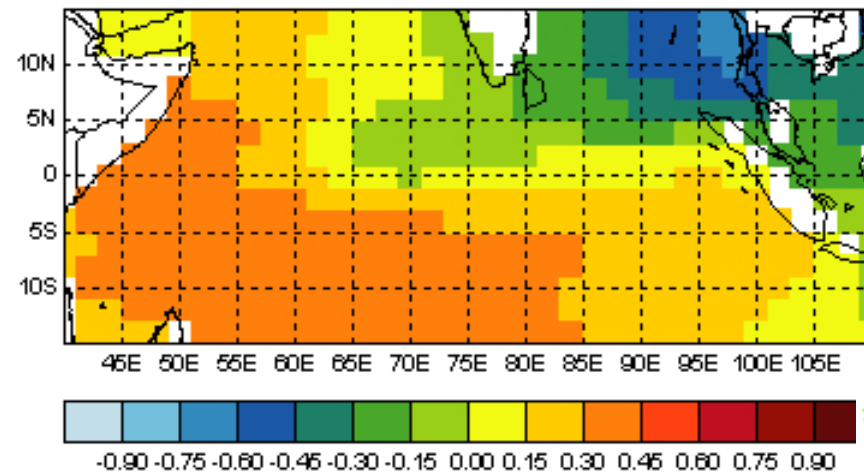
Y Spatial Loadings (Mode1)



X Spatial Loadings (Mode2)



Y Spatial Loadings (Mode2)



Two possible predictor designs in CCA

1. Observational predictor design

X is observed earlier
predictors, such as the
field of governing SST

2. Model MOS design

X is dynamical model
prediction of rainfall pattern
around a region of interest

Y is rainfall pattern
prediction for a
region of interest

1. is a purely statistical forecast system
2. Is a dynamical forecast corrected by a statistical adjustment

An Example of a CCA with observational predictor design

X data set: Tropical SST for Oct-Nov-Dec season

Y data set: Precipitation for 50N-50S for subsequent Jan-Feb-Mar

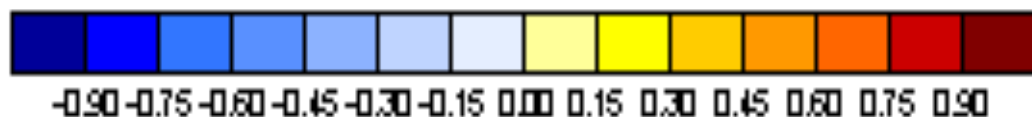
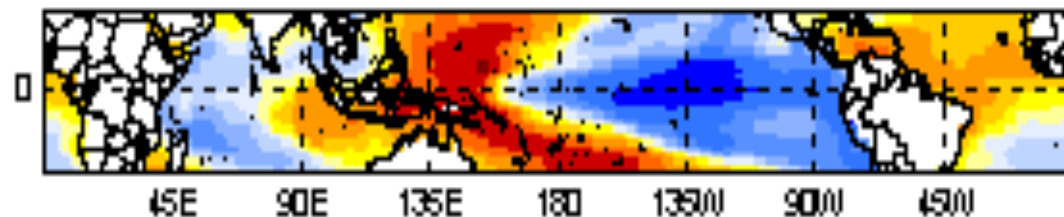
Physical basis for expecting a predictive pattern: During ENSO, the global SST anomalies are often strongest during Oct-Nov-Dec. The influence on global precipitation is seen most clearly during the northern winter season.

EOFs are first taken of X and of Y, and their time series are input to CCA.

CCA Mode 1 – a “non-pure” ENSO scenario (a general trend is also slightly mixed in)

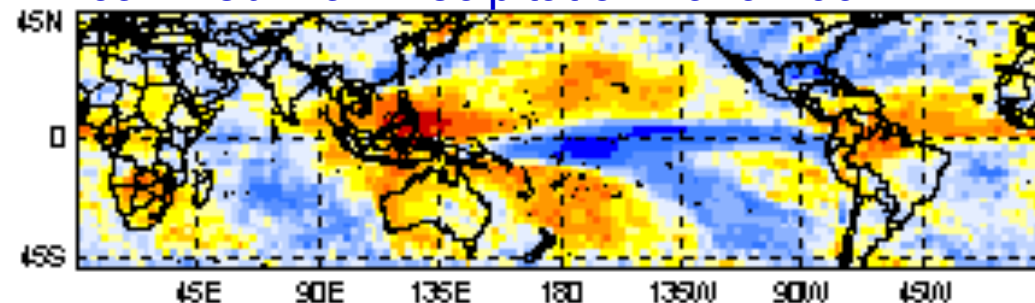
X Spatial Loadings (Mode1)

Oct-Nov-Dec SST 1978-2006

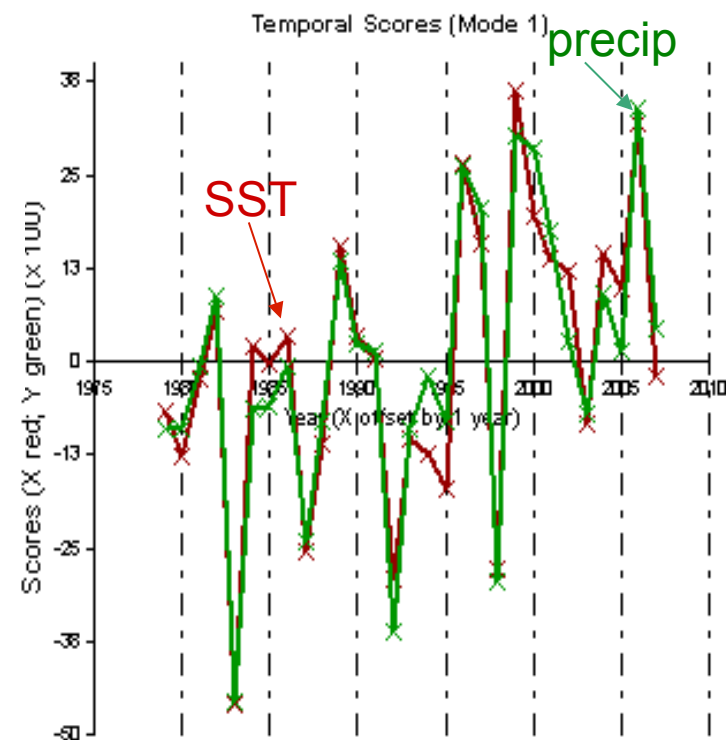


Y Spatial Loadings (Mode1)

Jan-Feb-Mar Precipitation 1979-2007

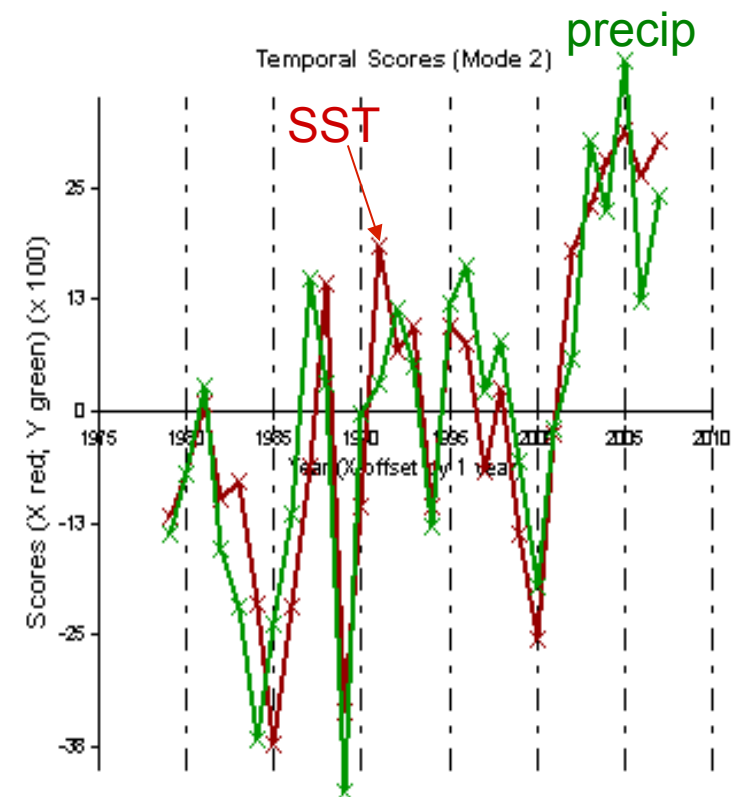
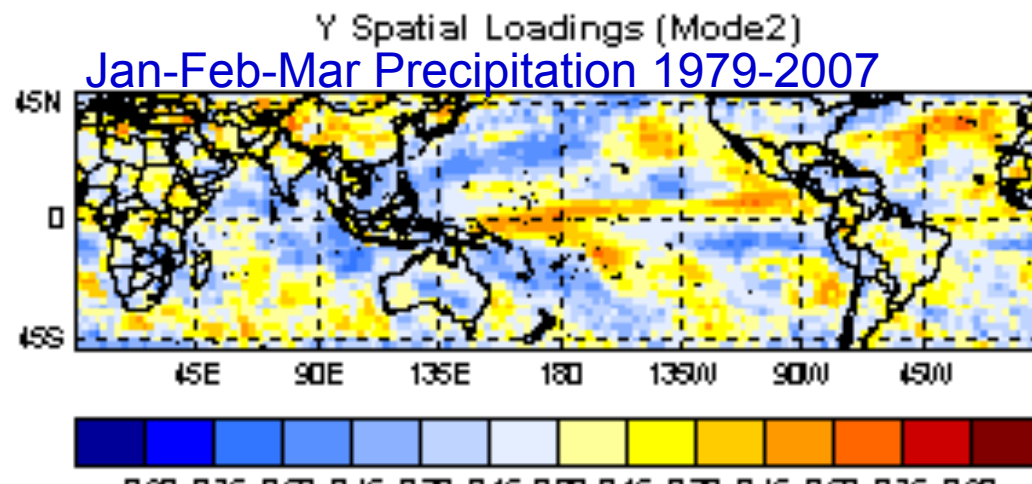
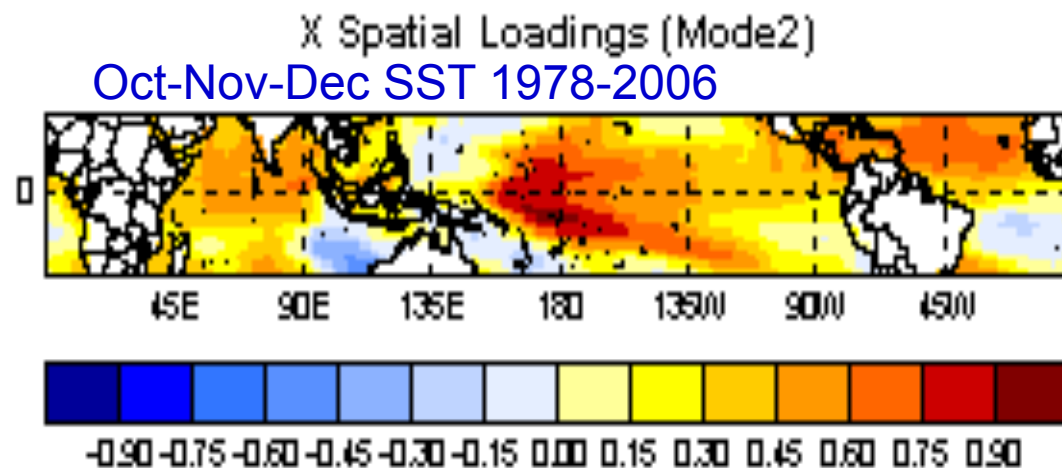


Temporal Scores (Mode 1)



X plotted in year of Y

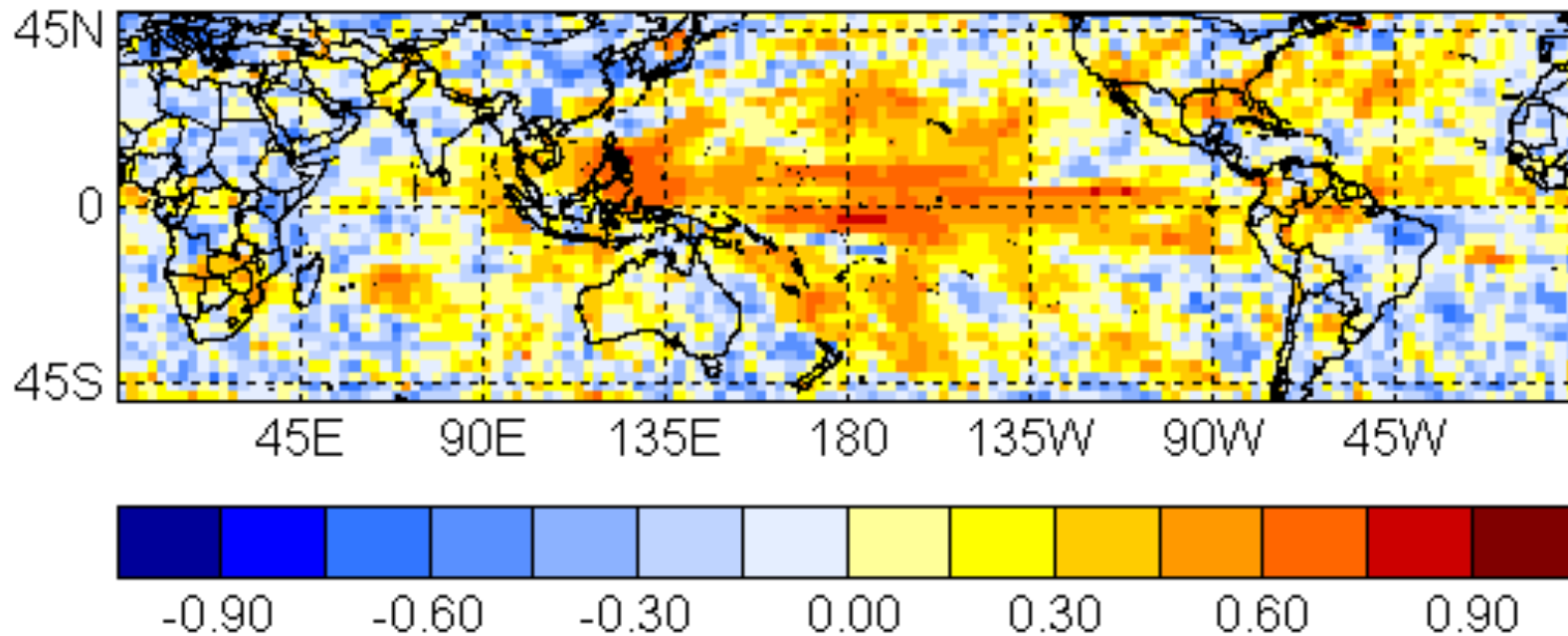
CCA Mode 2 – global climate change, plus a small residual of ENSO



X plotted in year of Y

Skill of CCA forecast of Jan-Feb-Mar precipitation, using Oct-Nov-Dec SST anomaly as predictor, 1979-2007

Pearson's Correlation



Another Example of a CCA with observational predictor design

X data set: Tropical SST for Oct-Nov-Dec season

Y data set: Air Temperature for 55N-55S for subsequent Jan-Feb-Mar

Physical basis for expecting a predictive pattern: During ENSO, the global SST anomalies are often strongest during Oct-Nov-Dec. The influence on global temperature is often strongest from the northern winter season through the northern spring season (there is some phase delay).

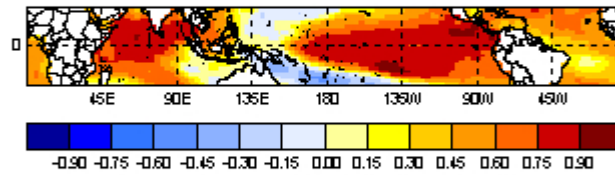
EOFs are first taken of X and of Y, and their time series are input to CCA.

OND
SST

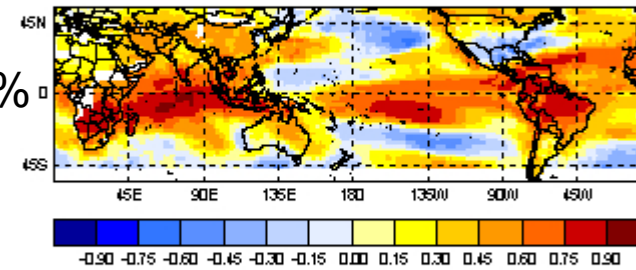
EOFs: Mode 1

JFM
Air Temperature
Y Spatial Loadings (EOF1)

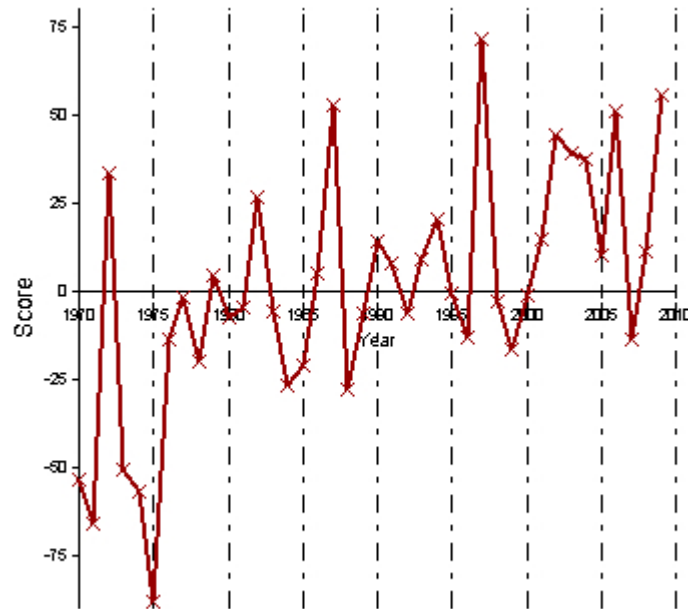
X Spatial Loadings (EOF1)



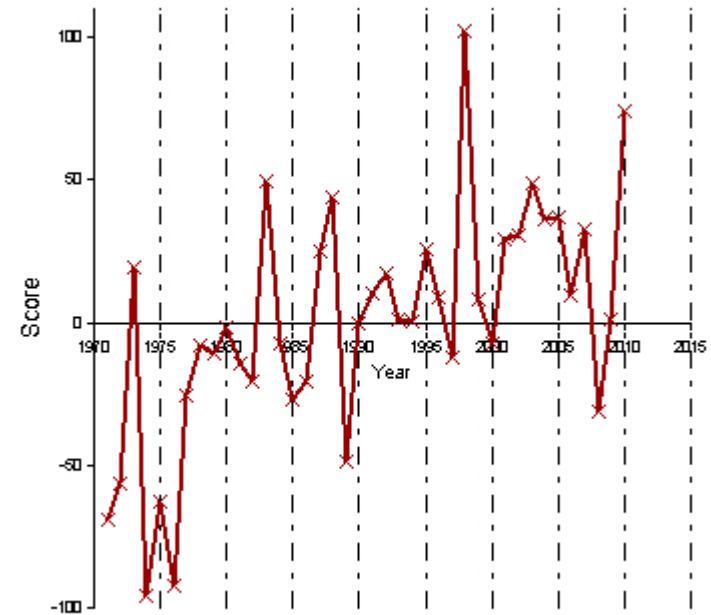
35% 22%



X Temporal Scores (EOF1)



Y Temporal Scores (EOF1)

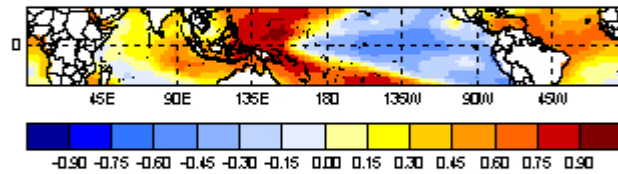


OND
SST

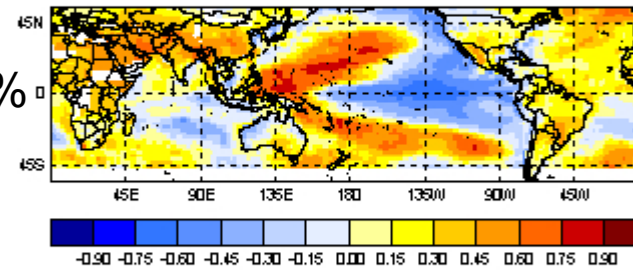
EOFs: Mode 2

JFM
Air Temperature
Y Spatial Loadings (EOF2)

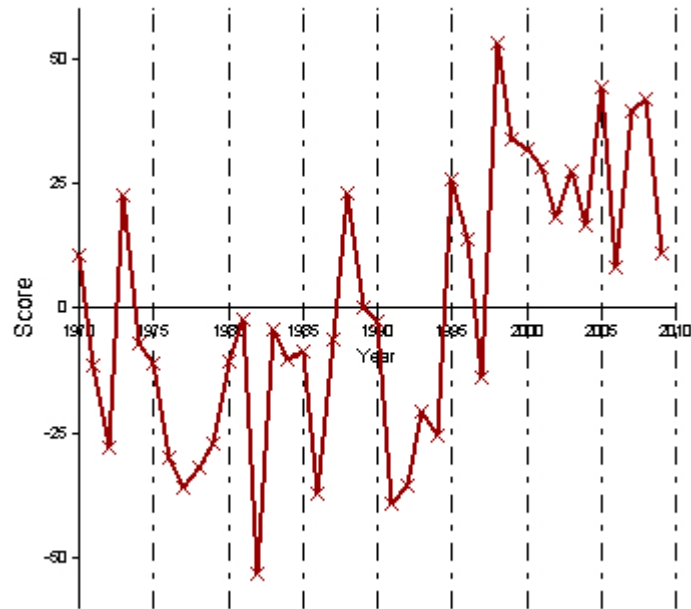
X Spatial Loadings (EOF2)



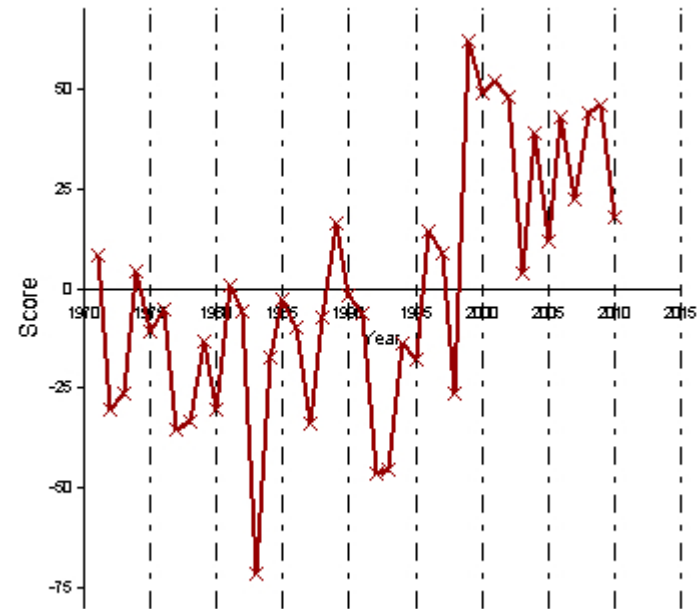
22% 12%



X Temporal Scores (EOF2)



Y Temporal Scores (EOF2)

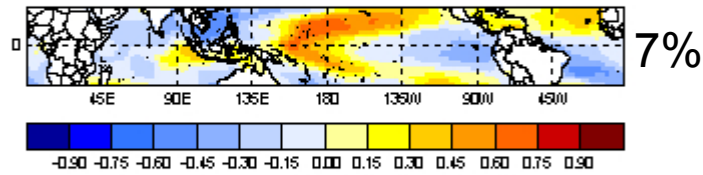


OND
SST

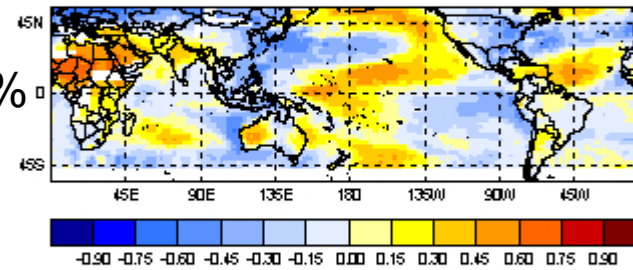
EOFs: Mode 3

JFM
Air Temperature
Y Spatial Loadings (EOF3)

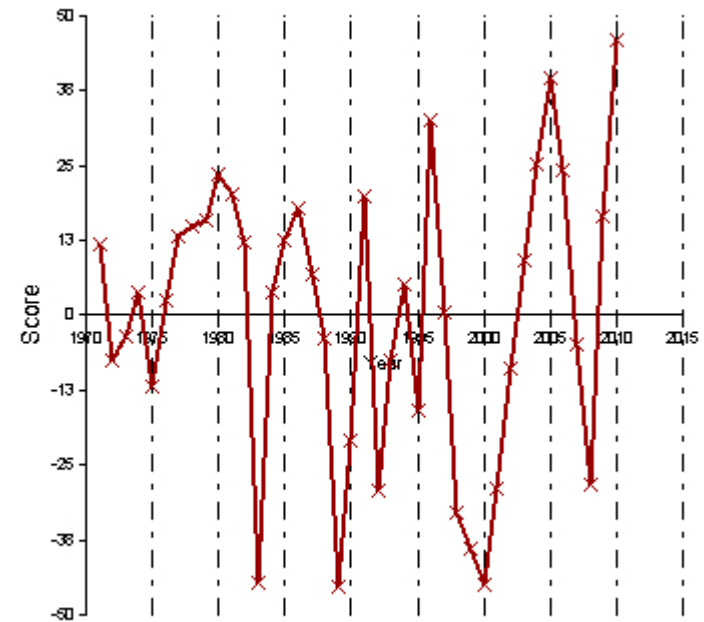
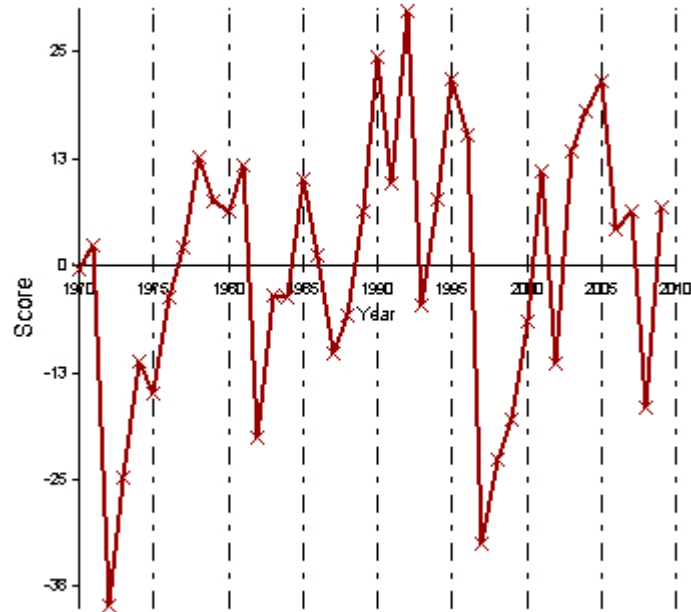
X Spatial Loadings (EOF3)



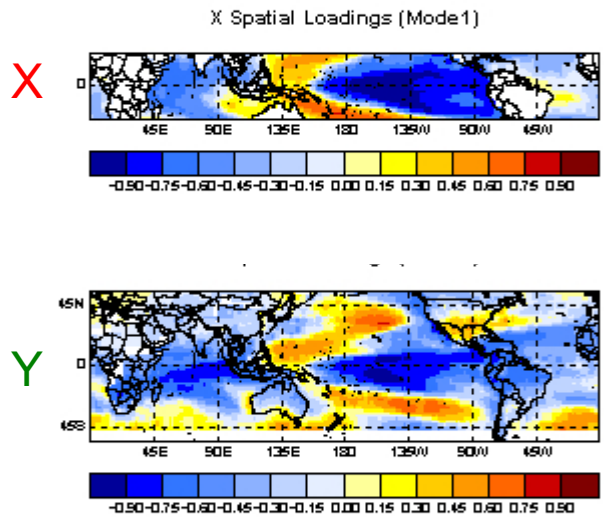
7%



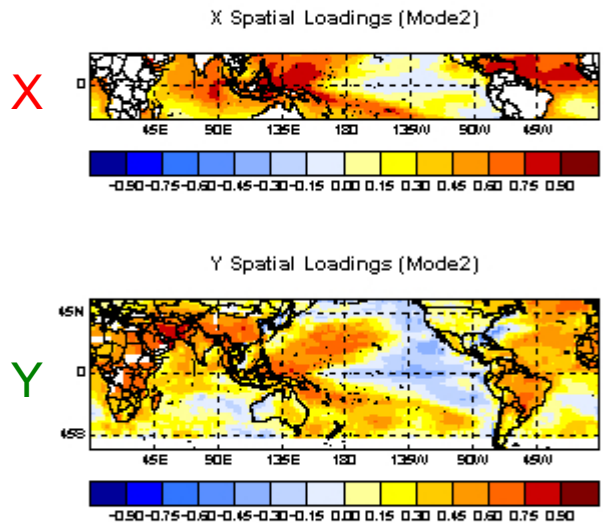
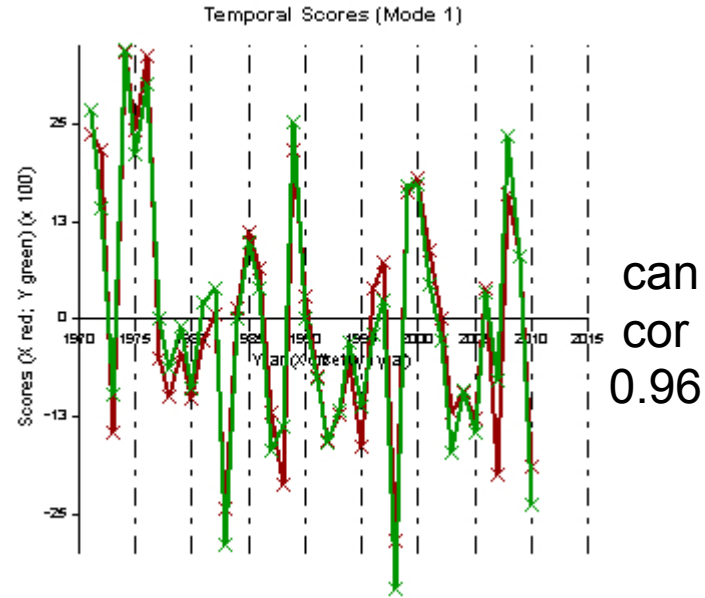
X Temporal Scores (EOF3)



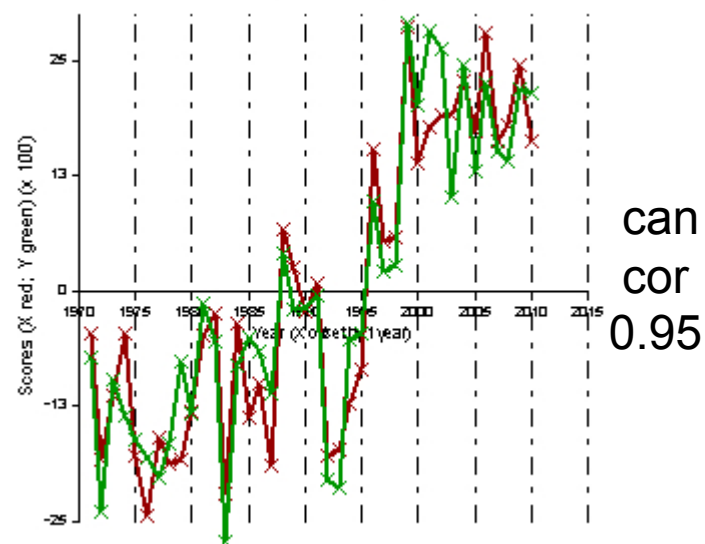
Leading CCA modes for OND SST versus JFM Air Temperature



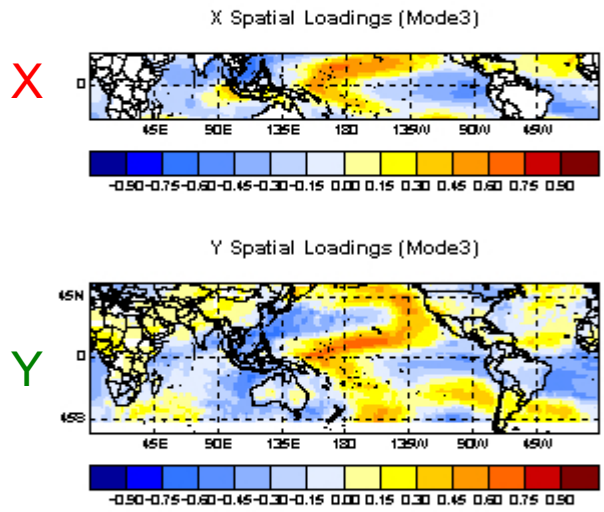
CCA
Mode
1



CCA
Mode
2



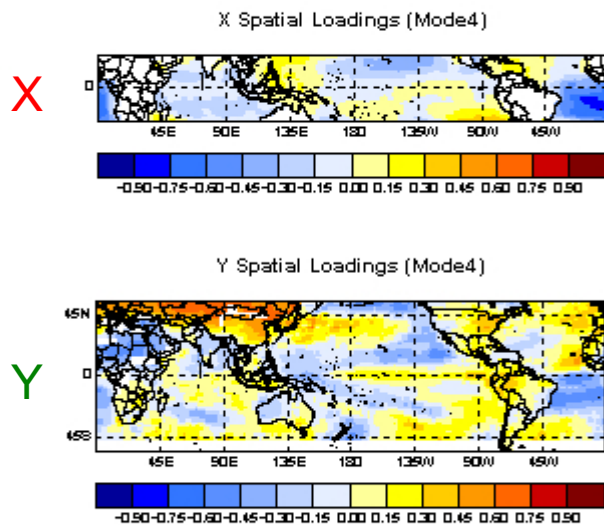
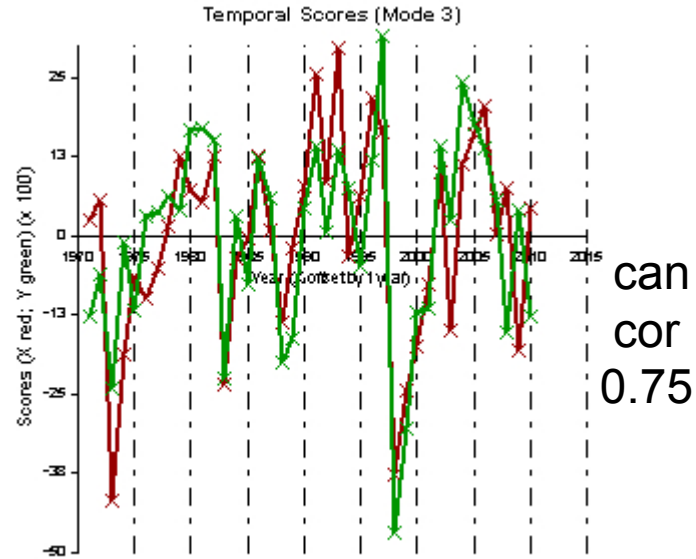
Leading CCA modes for OND SST versus JFM Air Temperature



X

Y

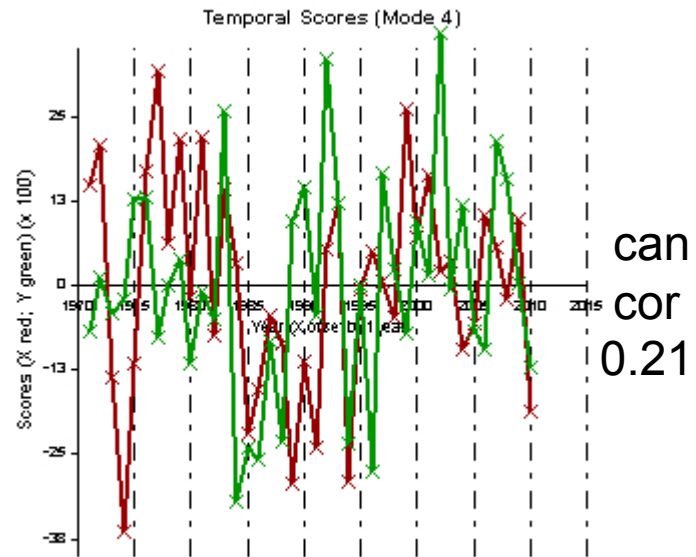
CCA
Mode
3



X

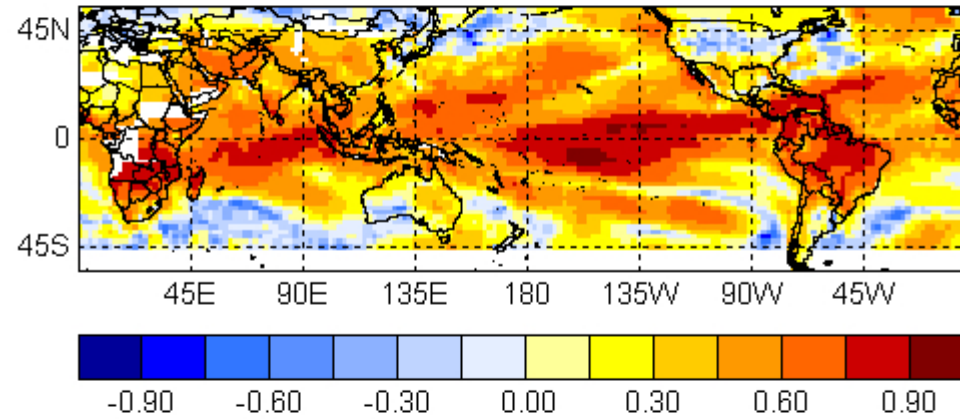
Y

CCA
Mode
4

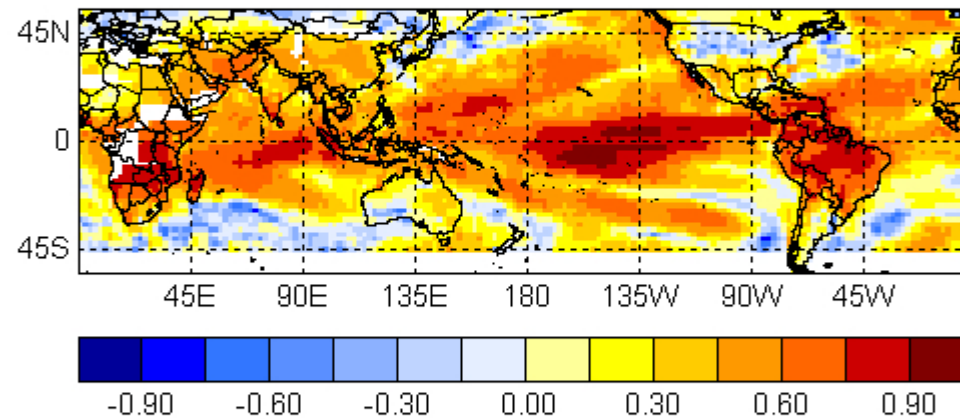


Skill of OND SST versus JFM Air Temperature

Pearson's Correlation



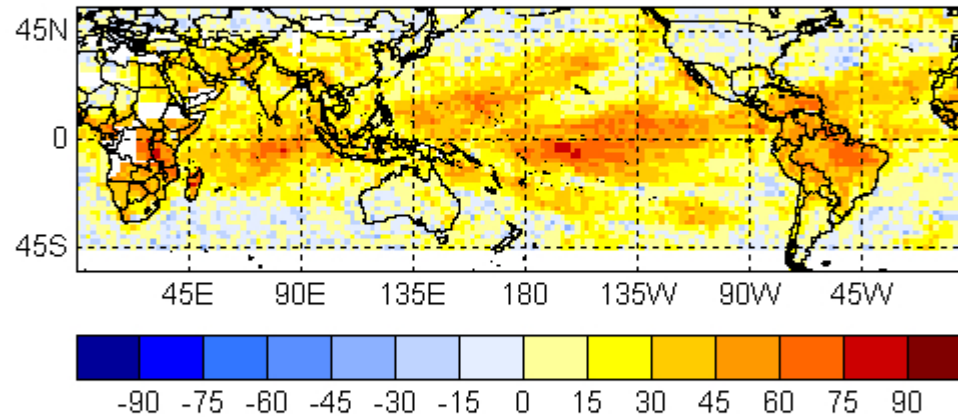
Spearman's Correlation



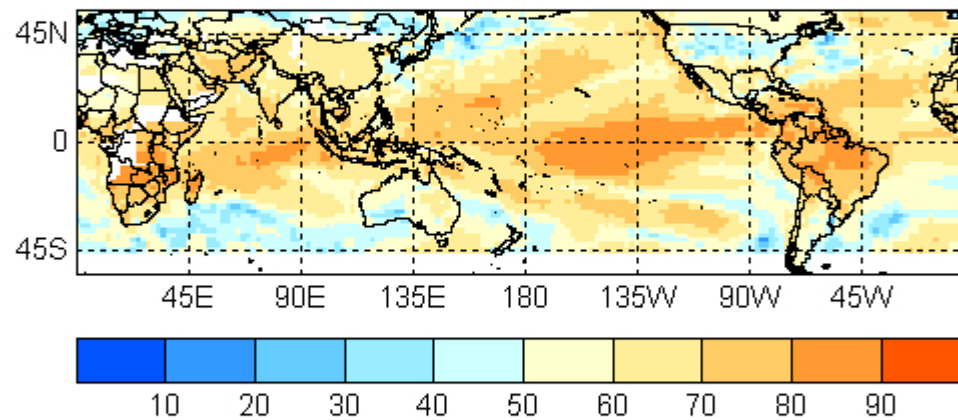
Skill of OND SST versus JFM Air Temperature

Heidke Skill Score

Hit Skill Score

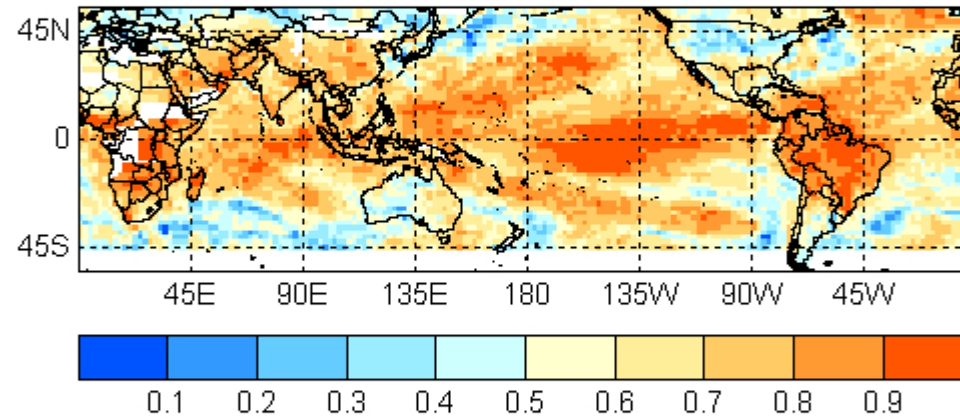


2AFC Score

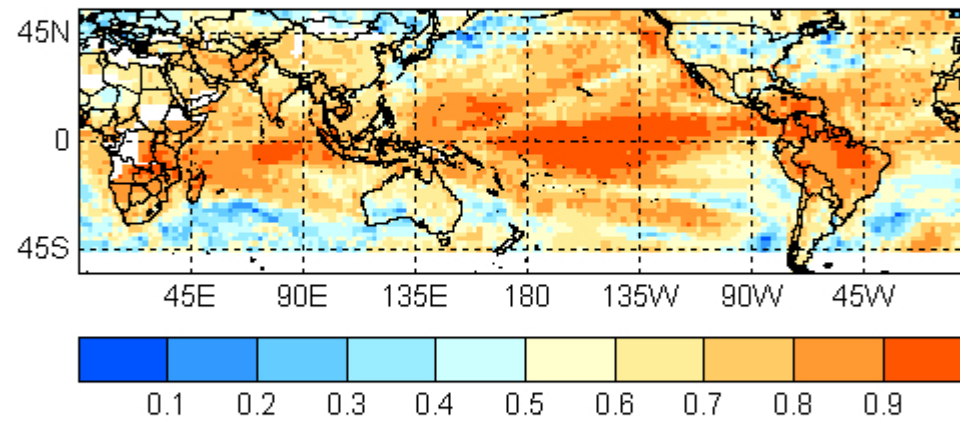


Skill of OND SST versus JFM Air Temperature

ROC Area (Above-Normal)



ROC Area (Below-Normal)

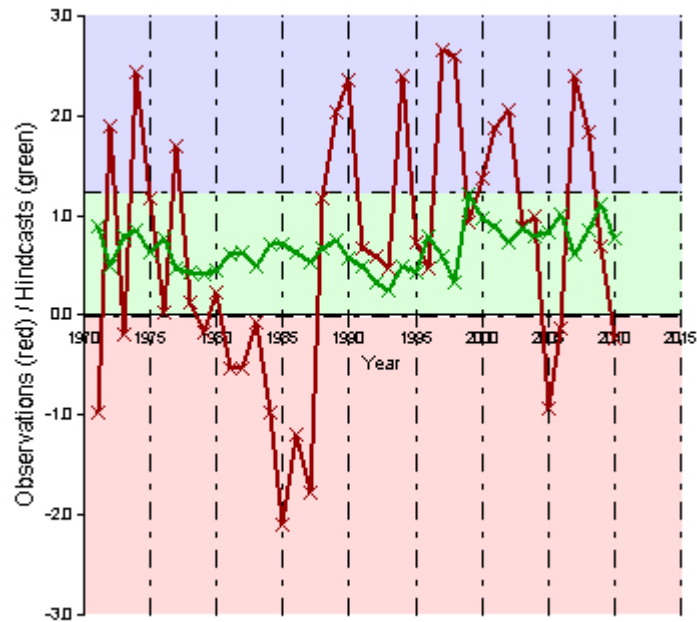


Hindcasts for a well-predicted and a poorly-predicted location

northern Italy

45N, 11E

Observations and Cross-Validated Hindcasts

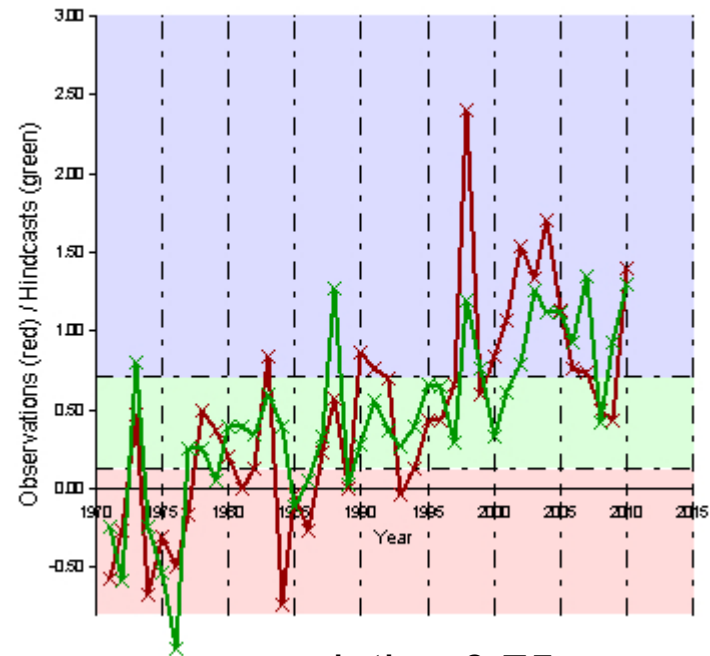


correlation -0.05
variance ratio 0.03

western Malaysia

5N, 103E

Observations and Cross-Validated Hindcasts



correlation 0.75
variance ratio 0.65

Ethiopian Seasonal Rainfall Prediction Using CCA

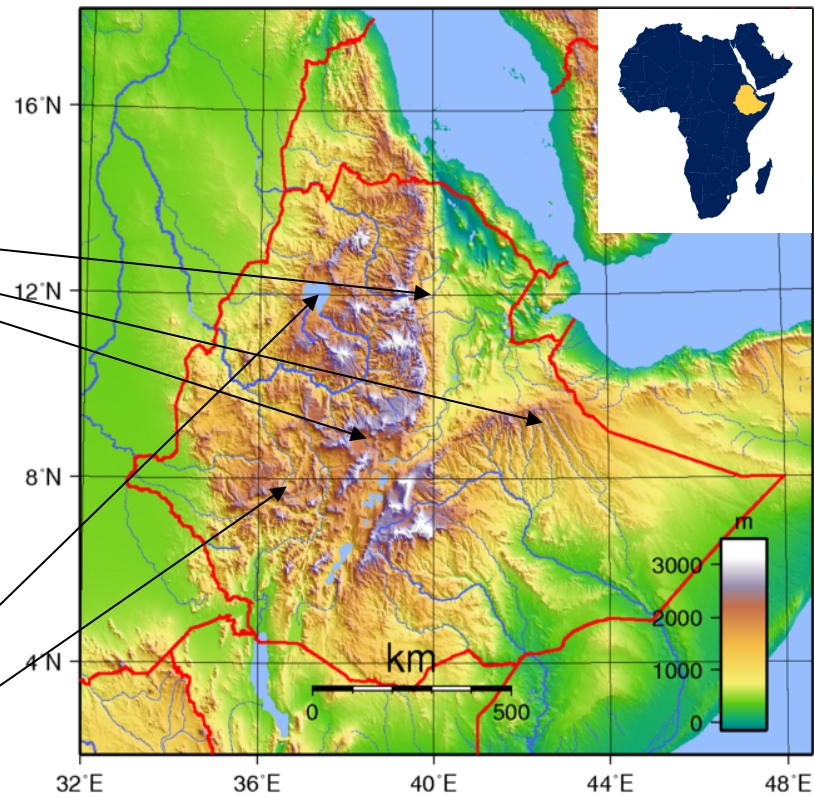
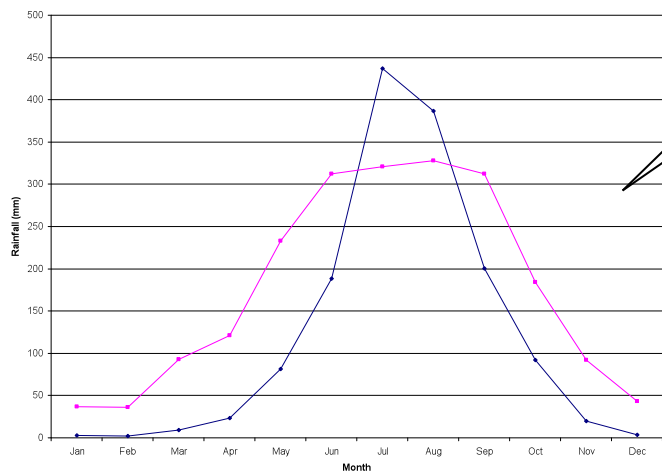
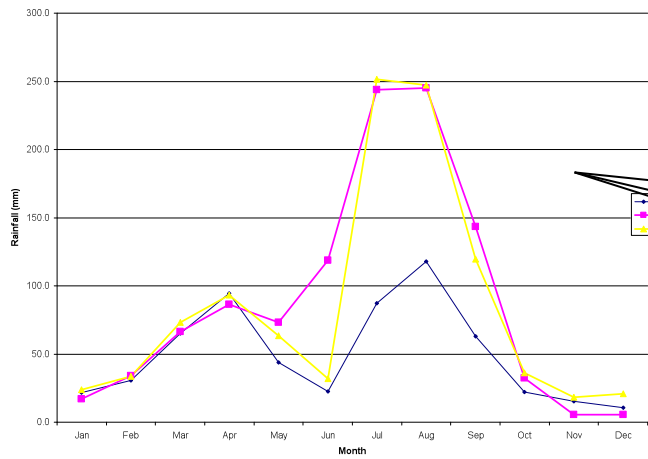
A. Empirical predictor design, using SST

B. Dynamical model correction design:

(1) Model rainfall → rainfall

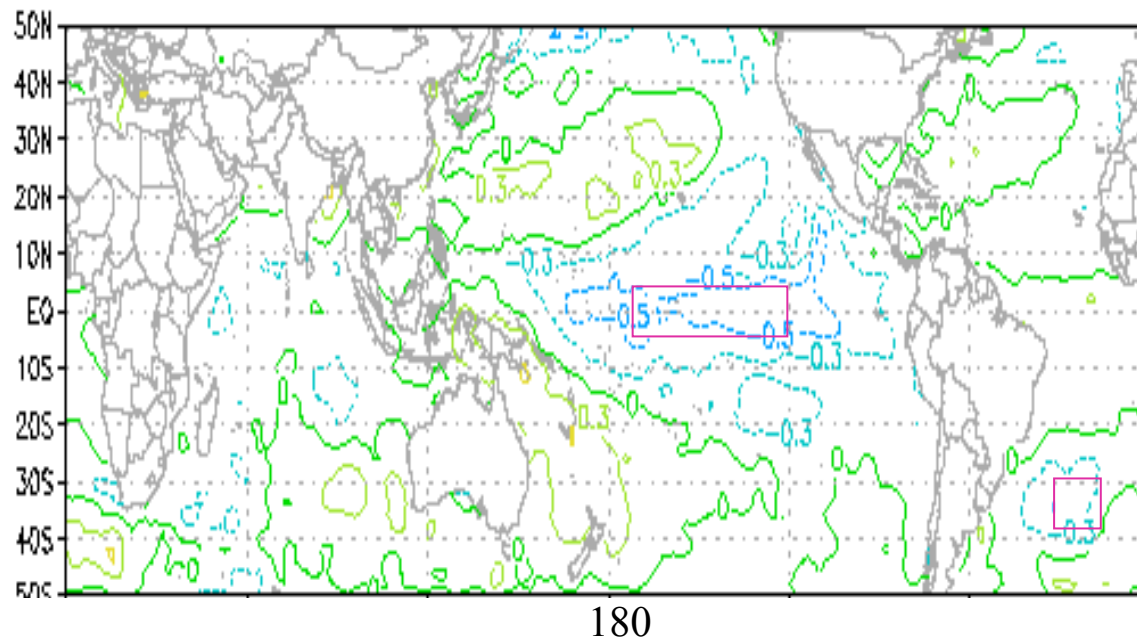
(2) Model v-wind → rainfall

Seasonal cycle of rainfall at various locations in Ethiopia

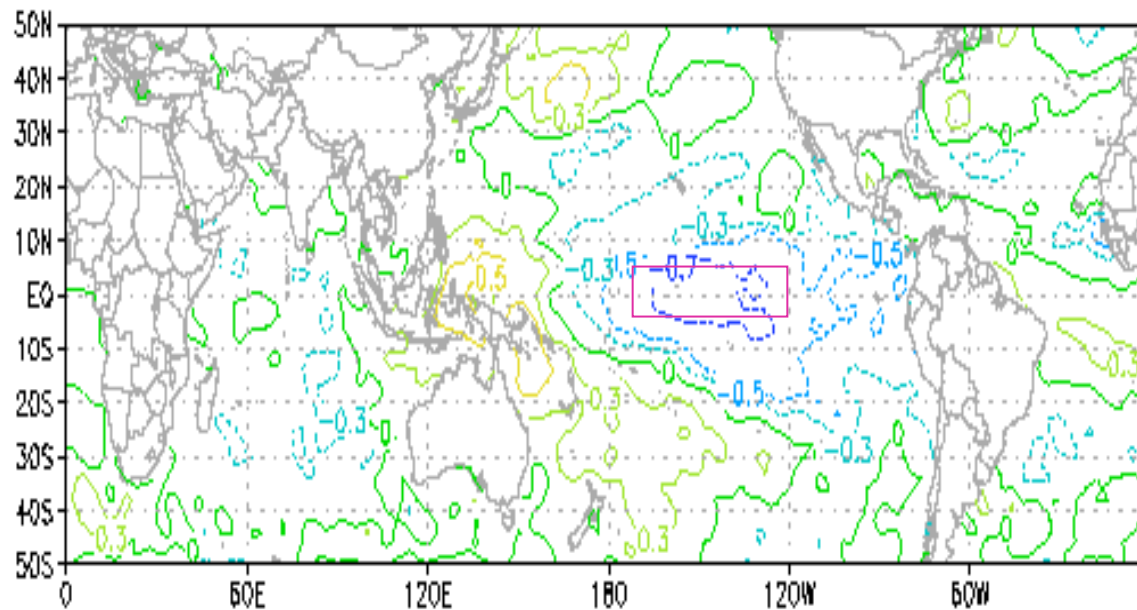


Except for southern and eastern Ethiopia, main rainy season is from June to September

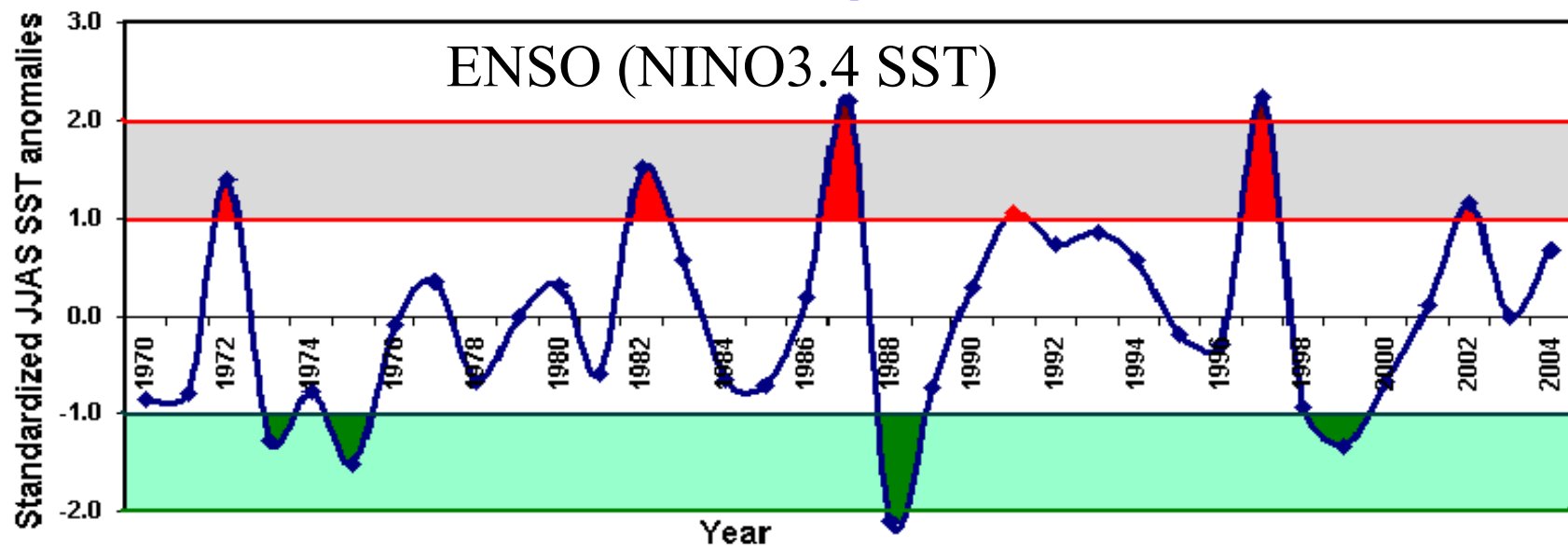
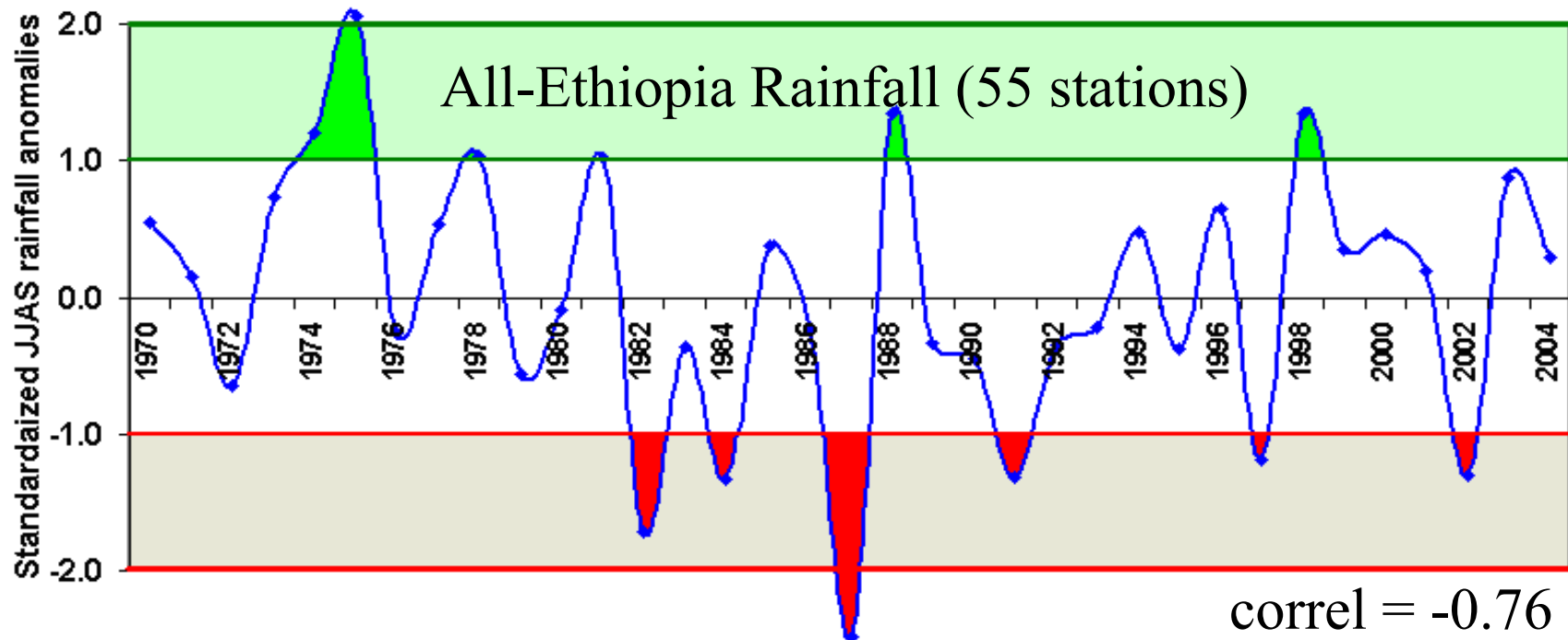
1970-2004 correlation: JJAS All-Ethiopia rainfall vs....



Correl:
May
SST
vs. all-
Ethiop
rainfall

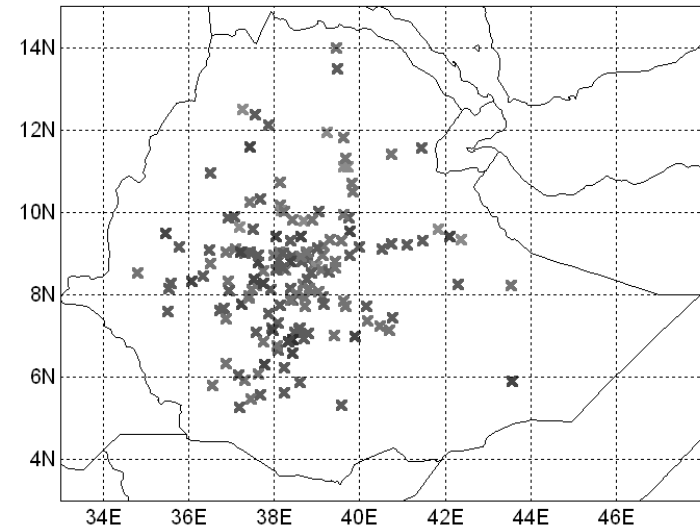


Correl:
Aug
SST
vs. all-
Ethiop
rainfall

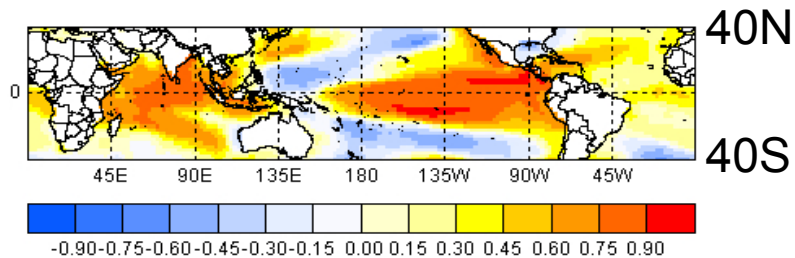


Data:

- 162 gauge data from National Meteorological Agency (NMA) of Ethiopia
- WCRP GPCP precipitation gridded data set
- ECHAM4.5 Precipitation and U & V wind components
- SST seasonal average for 40°N – 40°S: 2° x 2° resolution, from (NOAA/NCDC) version 2—i.e., the ERSSTv2 data set



Gauge stations network used

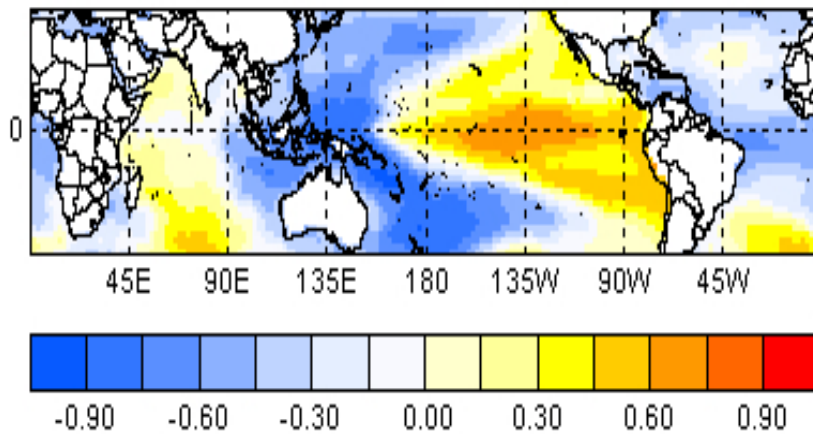


Empirical Prediction Design:

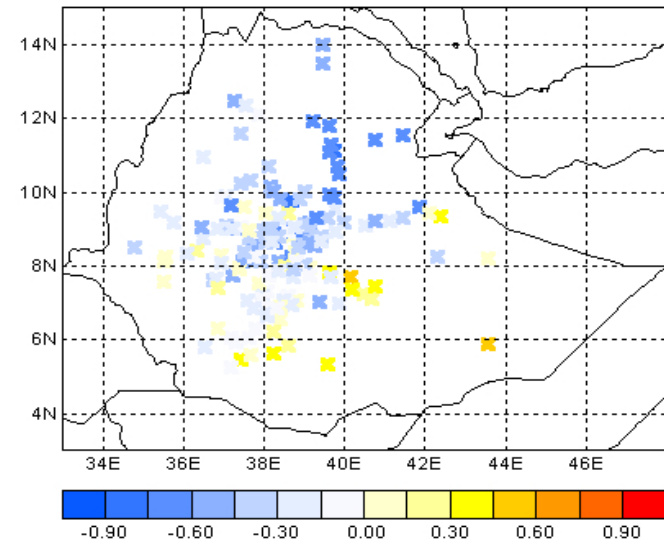
Mode 1 for JJAS *station rainfall*, using May SST as predictor:

Warm ENSO phase is associated with deficient rainfall in much of central and northern Ethiopia

SST in Elnino Scenario



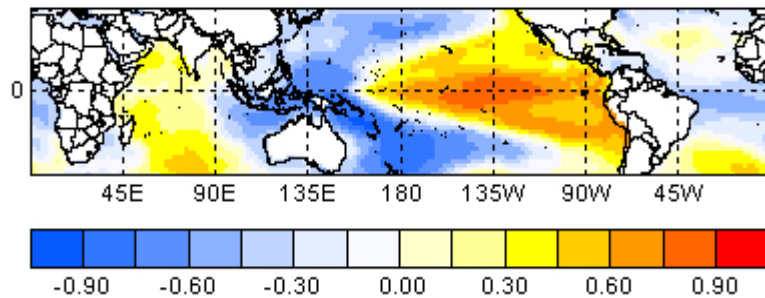
Ethiopian JJAS RF Spatial in Elnino Scenario



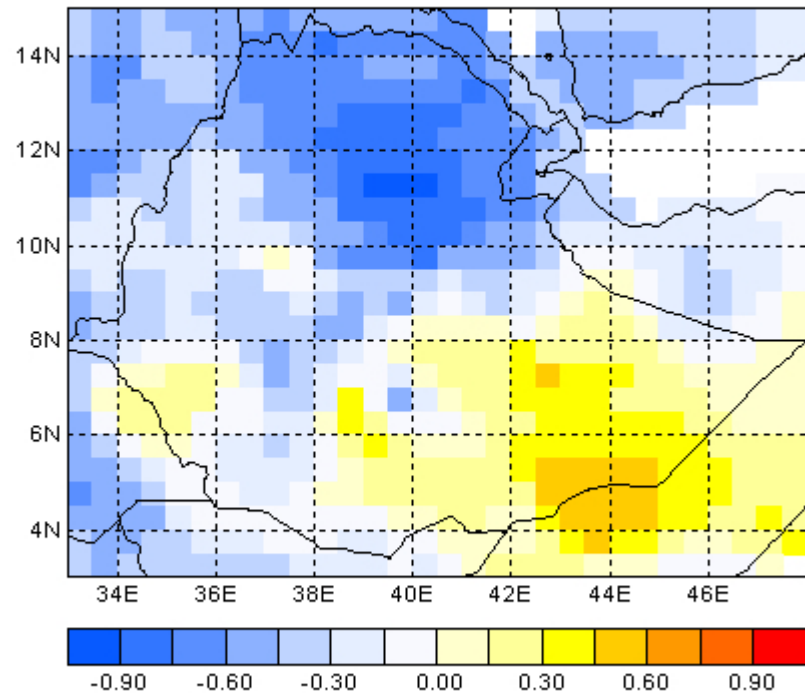
Empirical Prediction Design:

Mode 1 for JJAS WCRP (GPCC) *gridded rainfall*, using May SST as predictor:
Again, warm ENSO phase is associated with deficient rainfall in much of central and northern Ethiopia. Station and gridded rainfall data results agree.

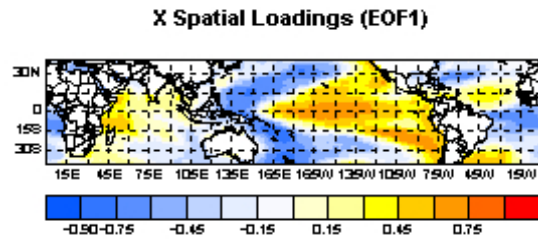
X Spatial Loadings (Mode1)



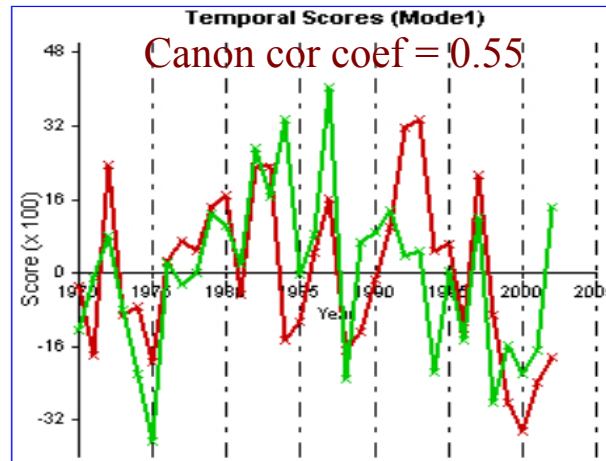
Y Spatial Loadings (Mode1)



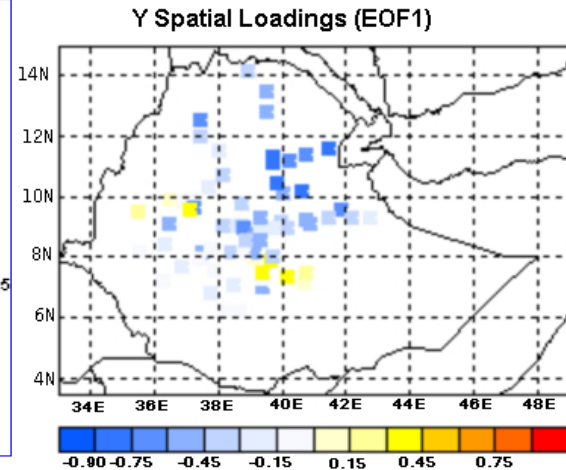
CCA using Climate Predictability Tool: 1 CCA mode only, using just 78 best stations



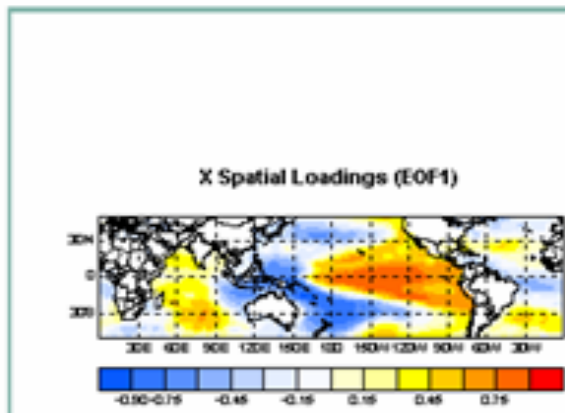
SST loadings,
May



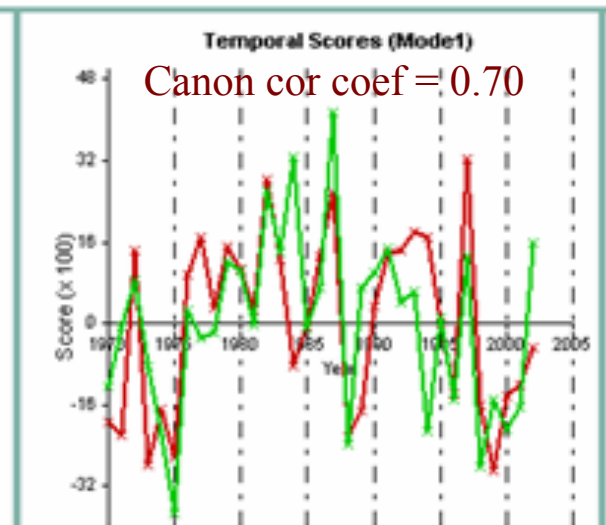
amplitude: **SST**, **rain**



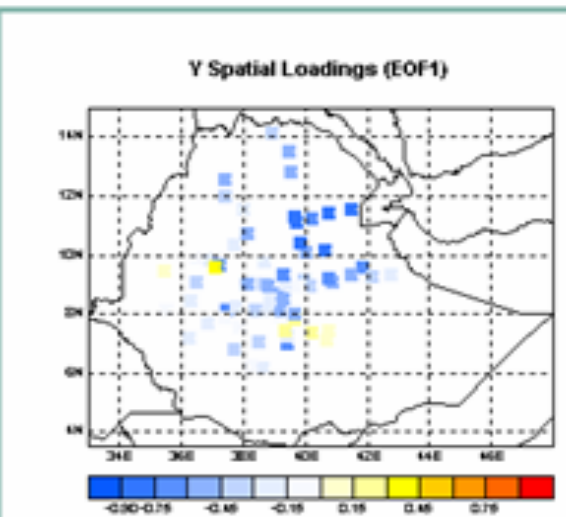
JJAS rainfall loadings



SST loadings,
JAS



amplitude: **SST**, **rain**



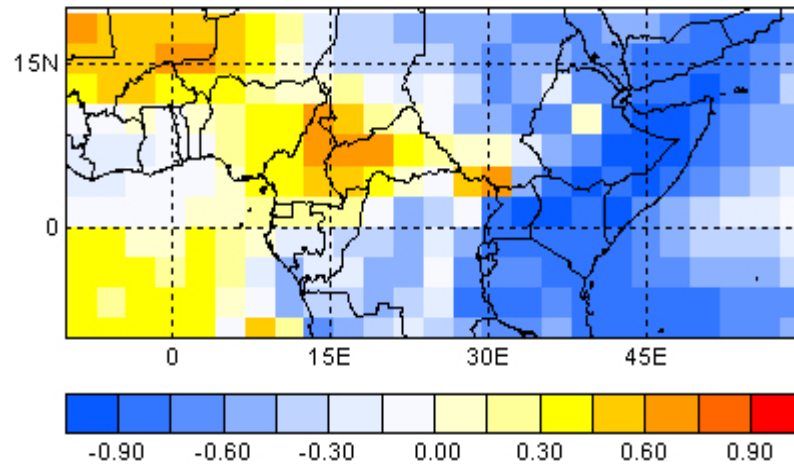
JJAS rainfall loadings

Dynamical model correction design:

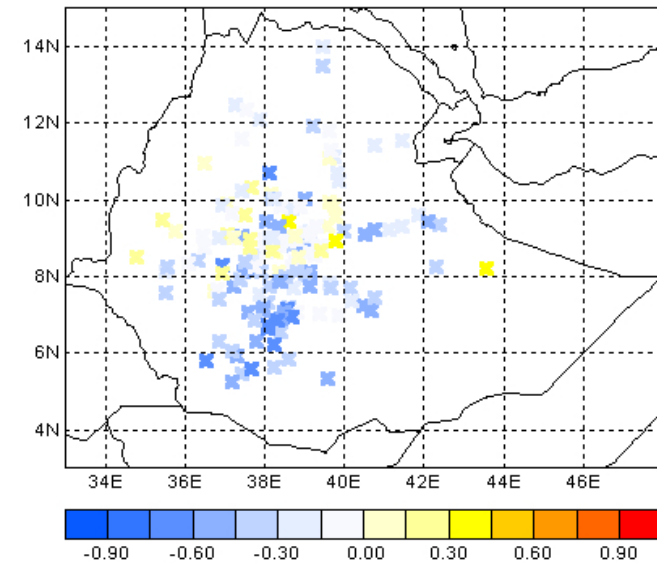
Mode 1 for JJAS station rainfall, using ECHAM4.5 AGCM rainfall forecast as predictor:

Pattern correction is slight, but local calibrations are noted.

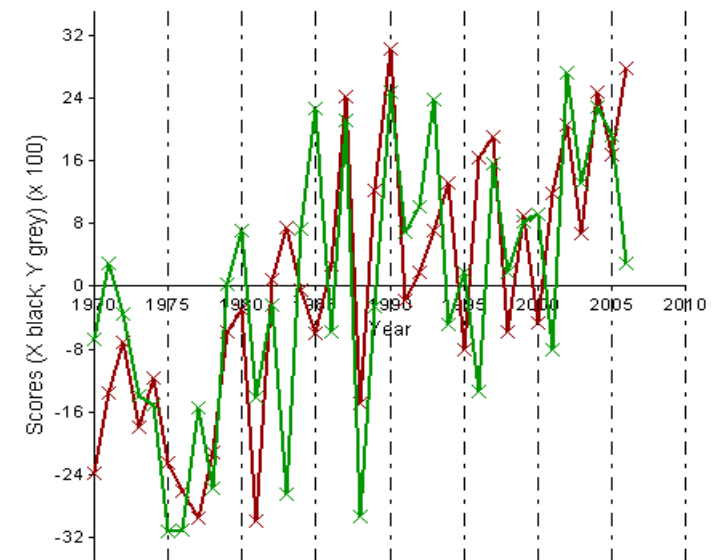
X Spatial Loadings (Mode1)



Y Spatial Loadings (Mode1)

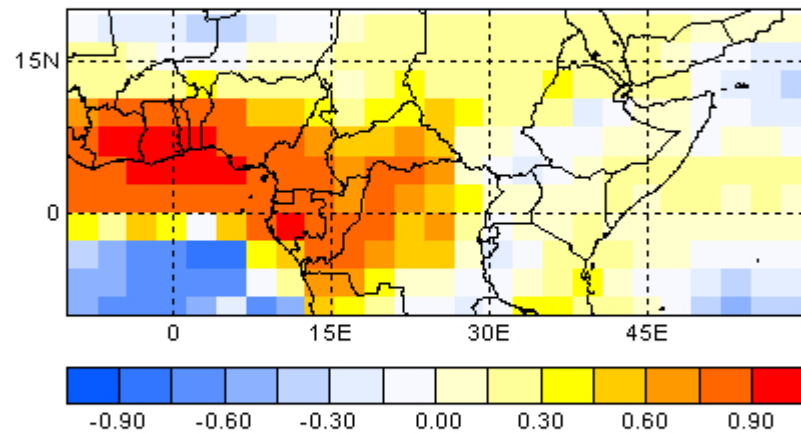


Temporal Scores (Mode1)

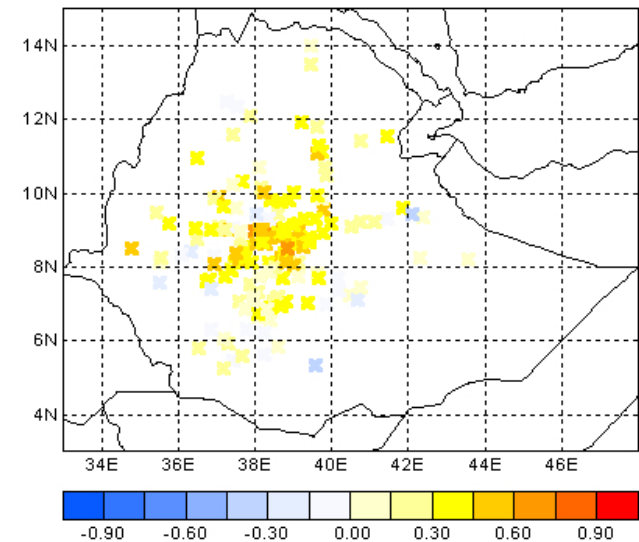


Dynamical model correction design:
Mode 2 for JJAS station rainfall, using
ECHAM4.5 AGCM rainfall forecast as predictor:
Pattern correction is substantial, and local
calibrations are also noted.

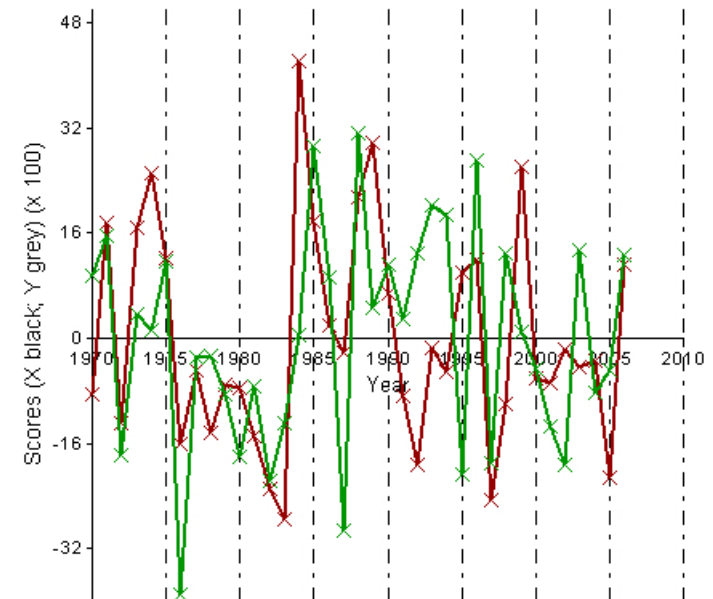
X Spatial Loadings (Mode2)



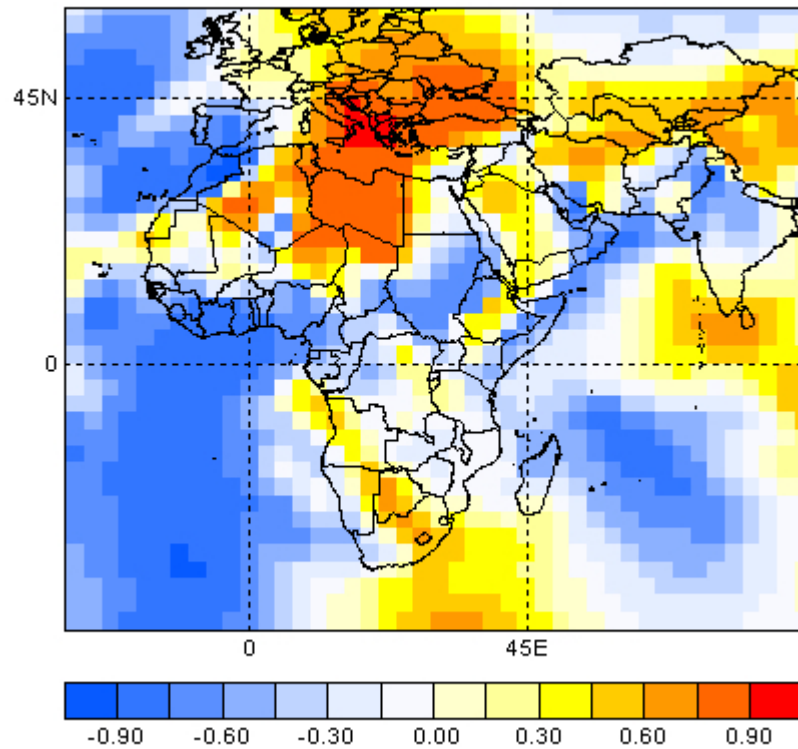
Y Spatial Loadings (Mode2)



Temporal Scores (Mode2)



X Spatial Loadings (Mode1)

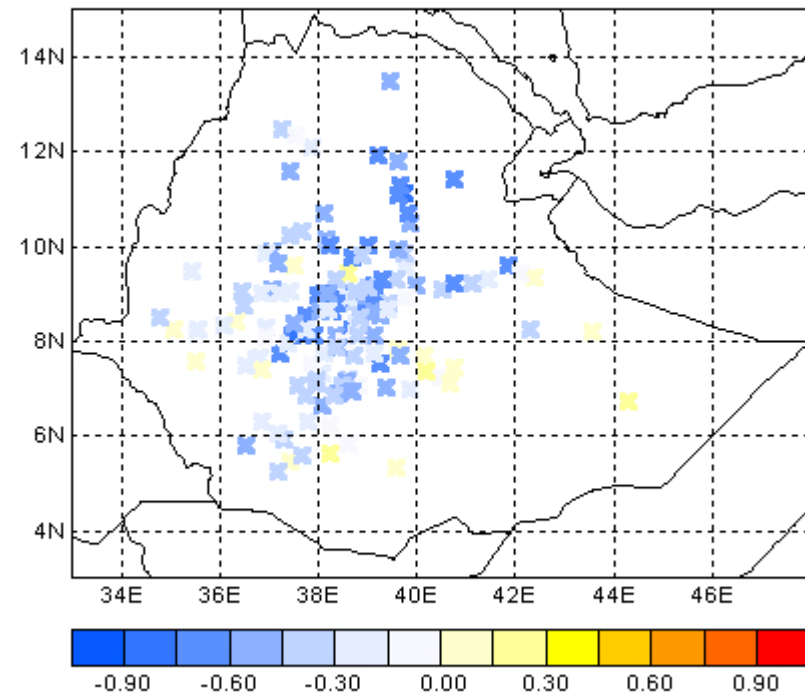


Dynamical model correction design:

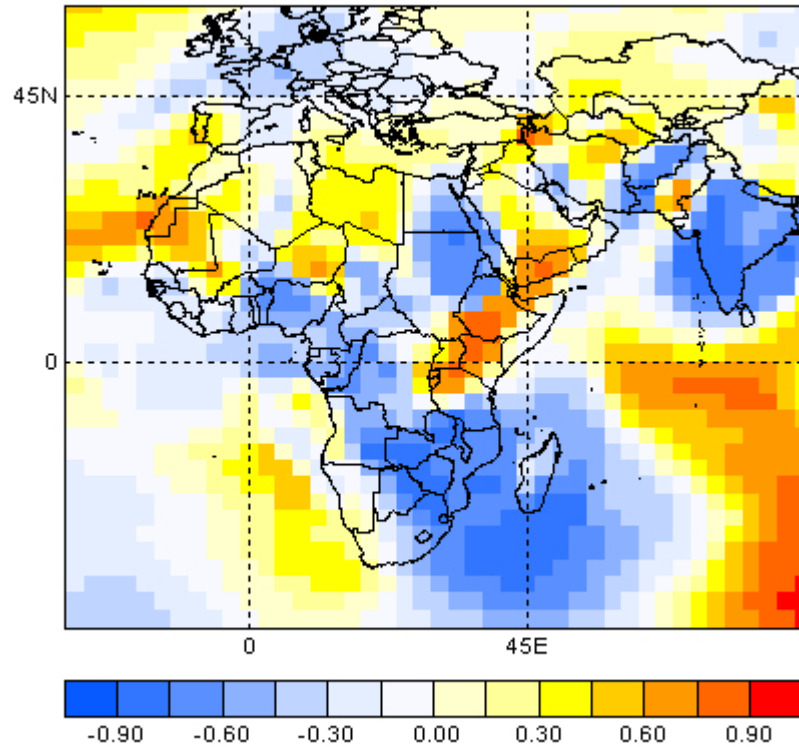
Mode 1 for JJAS v-wind component, using ECHAM4.5 AGCM v-wind forecast as predictor:

Pattern correction is substantial, and local calibrations are also noted.

Y Spatial Loadings (Mode1)



X Spatial Loadings (Mode2)

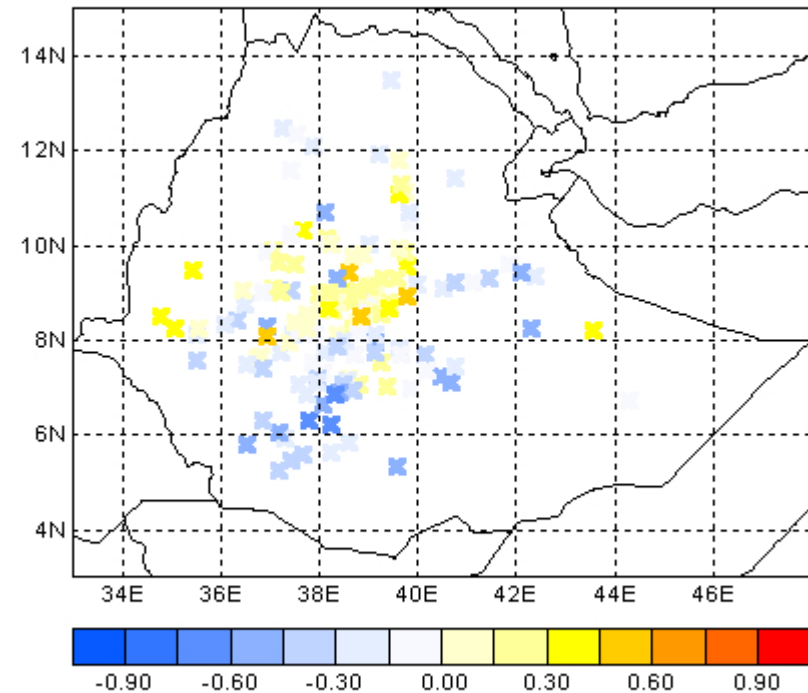


Dynamical model correction design:

Mode 2 for JJAS v-wind component, using ECHAM4.5 AGCM v-wind forecast as predictor:

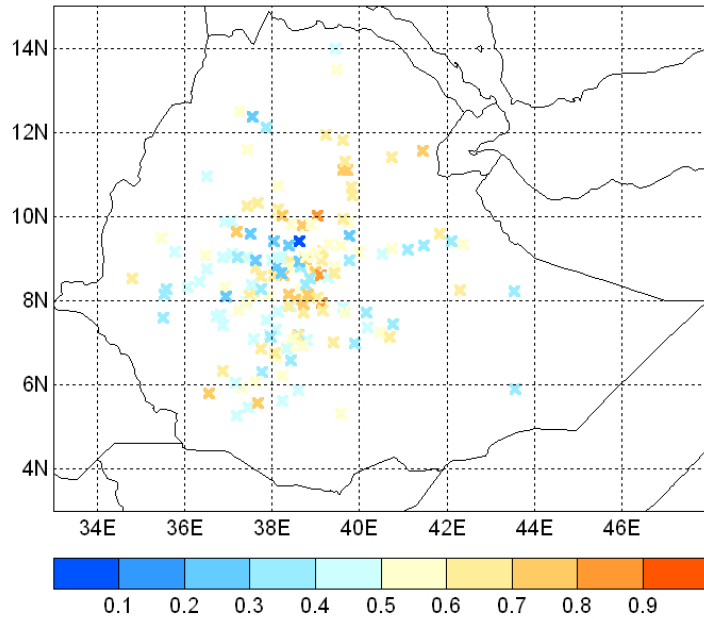
Pattern correction is substantial, and local calibrations are also noted.

Y Spatial Loadings (Mode2)

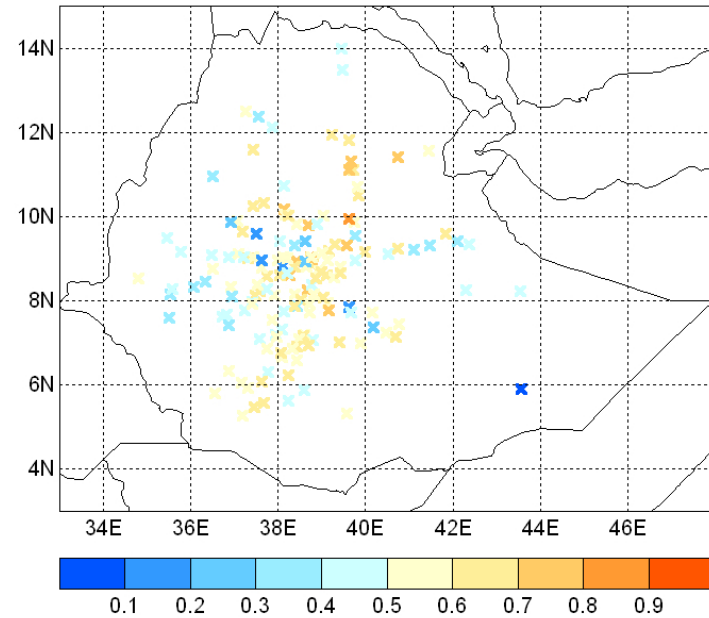


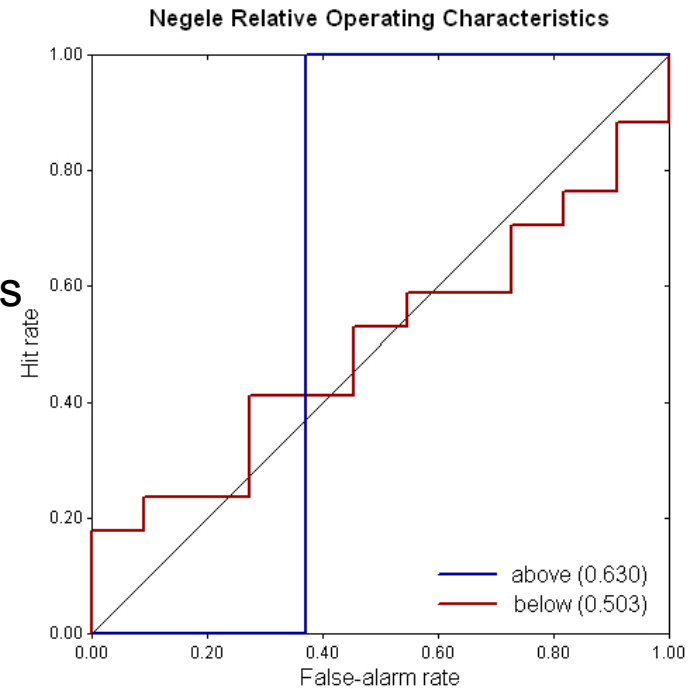
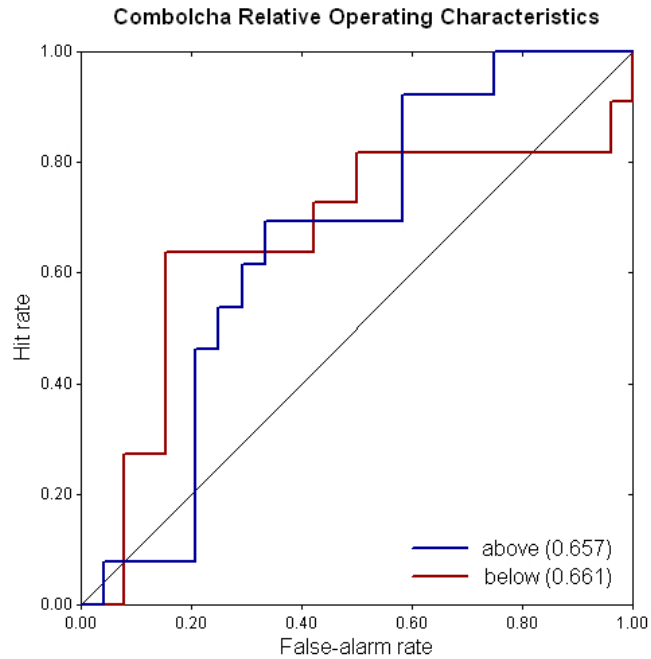
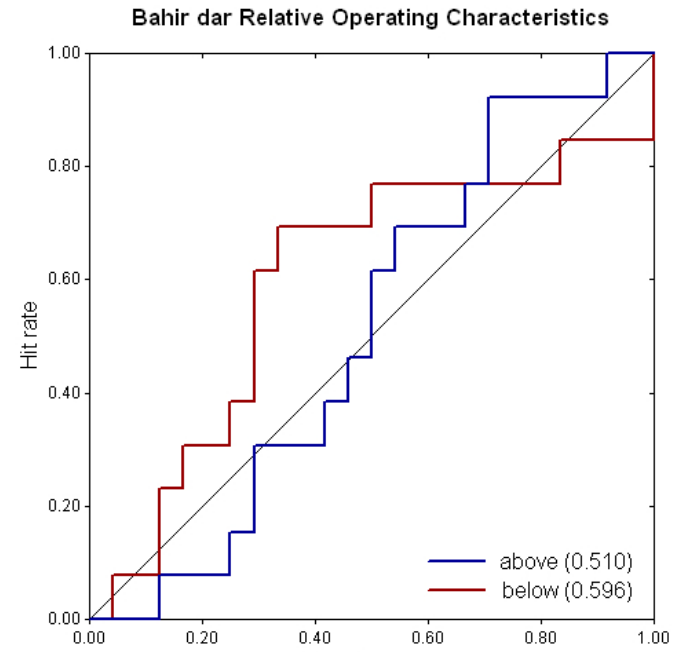
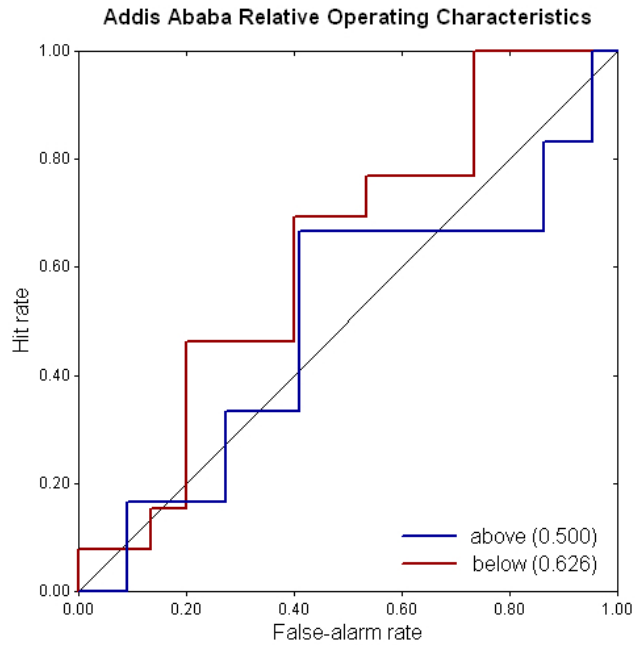
Skill map for JJAS rainfall using empirical (SST) design:
Area under Relative Operating Characteristics curve

ROC Area (Above-Normal)



ROC Area (Below-Normal)





ROC curves and ROC areas for four Ethiopia stations. Skill is modest but useful at some stations for certain tercile categories