Prediction of the Indian Ocean Dipole with Canonical Correlation Analysis

^{1,2}Zewdu Segele, ^{1,2}Endalkachew Bekele, ²Wassila Thiaw (<u>zewdu.segele@noaa.gov</u>)

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Introduction

- •The climate in equatorial East Africa is extremely sensitive to SST variability in Indian Ocean; Indian Ocean Dipole is a critical input for predicting October-December over East Africa.
- •SST variations in the northern and southern Indian Ocean (IO) also affect summer rainfall in the northern parts of the Greater Horn of Africa. For example, model sensitivity studies showed that warmer Northern-IO and cooler Southern-IO enhance July-August rainfall over Ethiopia.
- Develop SST prediction model using Canonical Correlation Analysis (CCA) and assess the skill of the statistical predictions with that from NMME.

Methodology

•We assess the multivariate joint relationship between a predictand Y (n X q) and a set of X (n X p)predictors using Canonical Correlation Analysis. The Eigen structures

$$R_{xx}^{-1} R_{xy} R_{yy}^{-1} R_{yx}$$
 $R_{yy}^{-1} R_{yx} R_{xx}^{-1} R_{xy}$

are used to compute the canonical coefficients A and B. Using $x_i = X_i - \overline{X_i}$ and $y_i = Y_i - \overline{Y_i}$, the prediction equation is given as

$$n\widehat{Y_q} = \bigcap_n X_p P_p A_r P_r L_r P_q^{-1}$$
$$n\widehat{Y_q} = nX_p P_q A_r L_r P_q^{-1} - \overline{nX_p} P_q A_r P_r L_r P_q^{-1} + \overline{nY_q}$$

Predictands and Predictors

- •ERSSTv5: SST predictand over the Indian Ocean and lagged predictors over the Indian, Atlantic and Pacific Oceans; Press
- •NCEP/NCAR Reanalyses: MSLP, zonal and meridional wind over the Indian Ocean;
- Predictand and Predictors are available for 1981 to present;
- •NMME IOD forecasts for 7 models are used to compare results with CCA IOD predictions.



Prediction Strategy

•Retroactive prediction approach is used, in which the first 30 years (1981-2010) are used to identify and select predictors and develop the prediction model, and the remaining 13 years are used for validation of prediction.

•Real-time forecasts are developed every month by repeatedly constructing models from 1981-2010 base period and increasing model development period every year until all data from 1981 to present are used.

•Monthly forecasts are made for each observation month for CCA and initialization month for NMME









Two Predictors combinations (cross-validated correlation)



Correlation between Observed and CCA and NMME (Raw) Predicted IOD indices



Timeseries of observed and CCA (left) and NMME (right) predicted Indices for October



Real-Time forecast from July 2023 observed SST



Real-Time forecast from November 2023 observed SST

Real-Time forecast from December 2023 observed SST (Jan initialization)

Real-Time forecast from January 2024 observed SST (Feb initialization)

Real-Time forecast from February 2024 observed SST (Mar initialization)

- Pacific SST predictor gives the highest correlation of 0.92 over western Indian Ocean but yields strong negative correlations over southeastern Indian Ocean.
- Atlantic SST shows least skill except over the Bay of Bengal. However, it gives high correlation when used with MSLP.
- Using meridional wind predictor reduces the extent of negative correlation over southeastern Indian Ocean. However, the skill is not strong elsewhere.
- CCA and NMME perform poorly in Jan-Apr and best during Sept-December months; Prediction skill weakens around ENSO predictability barrier.
- CCA IOD skill is comparable with the raw NMME IOD predictions. The ensemble mean shows increased skill compared to individual models, especially in September.

Thank you for your attention!