

# Seasonal forecasts

presented by:

Tel: +27(82)644-5304  
Willem.Landman@up.ac.za



<https://tinyurl.com/ybrb3a72>



UNIVERSITEIT VAN PRETORIA  
UNIVERSITY OF PRETORIA  
YUNIBESITHI YA PRETORIA



# Seasonal Forecast Worx

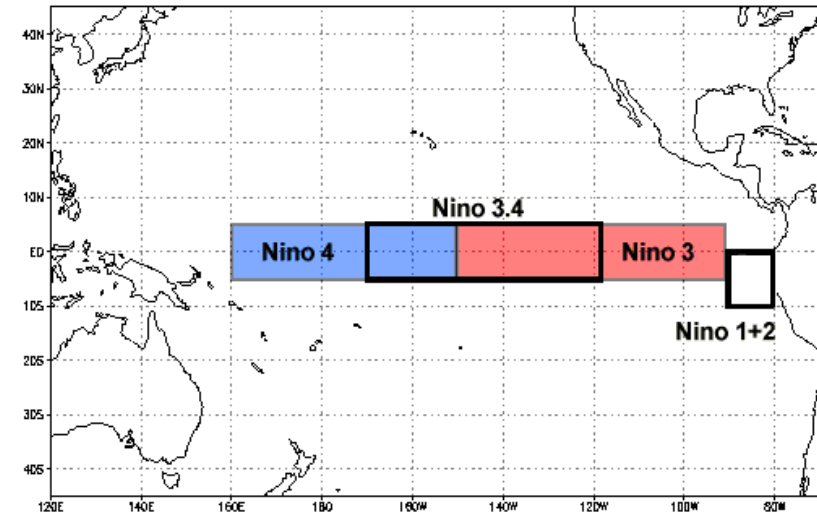
Latest Update: 9 February 2019

- The seasonal forecasts presented here by **Seasonal Forecast Worx** are based on forecast output of the coupled ocean-atmosphere models administered through the North American Multi-Model Ensemble (NMME) prediction experiment (<http://www.cpc.ncep.noaa.gov/products/NMME/>; Kirtman et al. 2014). NMME real-time seasonal forecast and hindcast (re-forecast) data are obtained from the data library (<http://iridl.ldeo.columbia.edu/>) of the International Research Institute for Climate and Society (IRI; <http://iri.columbia.edu/>).
- NMME forecasts are routinely produced and are statistically improved and tailored for southern Africa and for global sea-surface temperatures by employees and post-graduate students in the Department of Geography, Geoinformatics and Meteorology at the University of Pretoria (<http://www.up.ac.za/en/geography-geoinformatics-and-meteorology/>). Statistical post-processing is performed with the CPT software (<http://iri.columbia.edu/our-expertise/climate/tools/cpt/>).
- Why do we apply statistical methods to climate model forecasts?  
“...**statistical correction methods treating individual locations (e.g. multiple regression or principal component regression) may be recommended for today’s coupled climate model forecasts**”. (Barnston and Tippett, 2017).
- Why do we not use just a single model in our forecasts for southern Africa?  
“...**multi-model forecasts outperform the single model forecasts...**” (Landman and Beraki, 2012).
- For the official seasonal forecast for South Africa, visit the South African Weather Service website at <http://www.weathersa.co.za/home/seasonal>

# ENSO and Global SST Forecasts

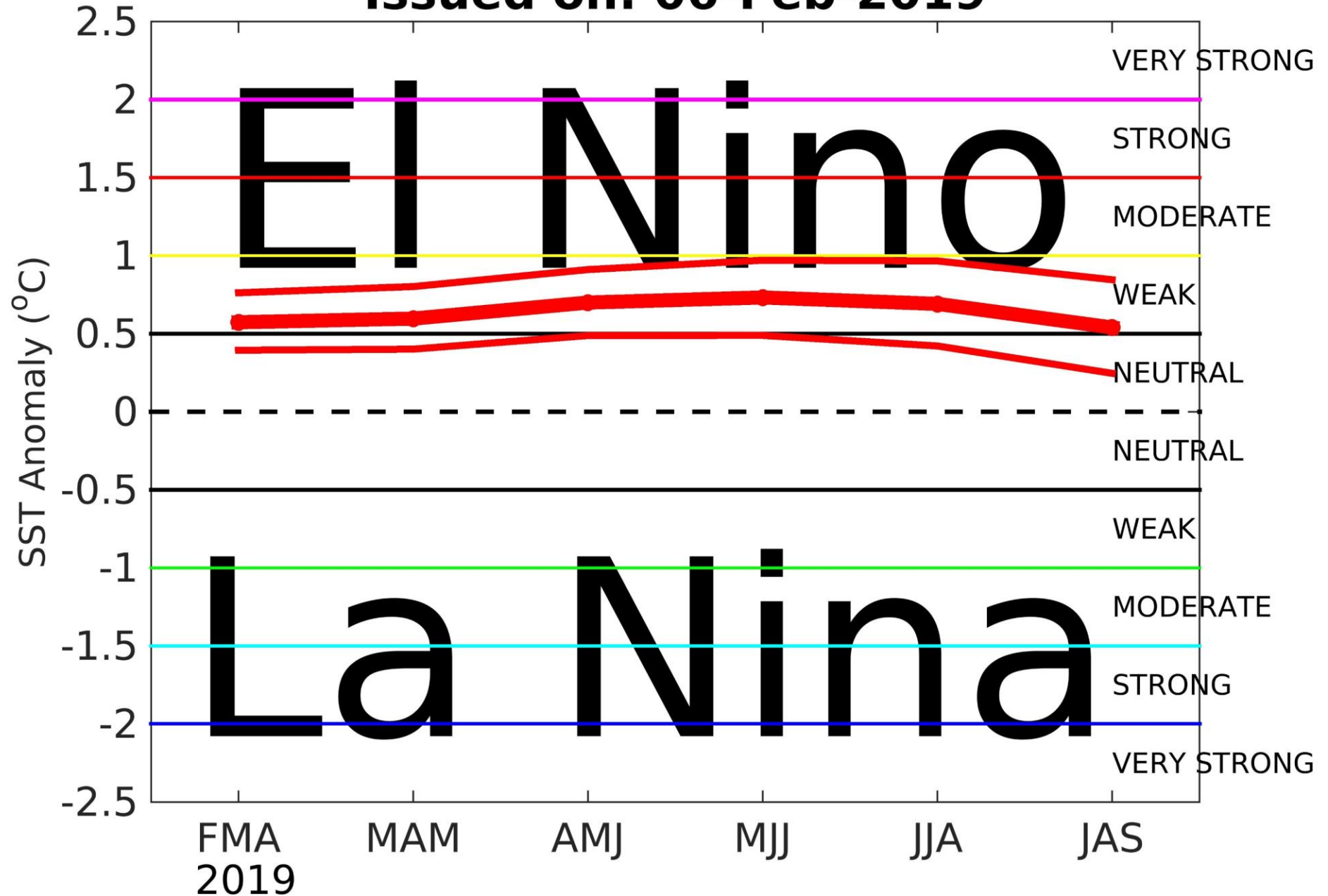
# Prediction Method

- Forecasts for global sea-surface temperature (SST) fields are obtained through a combination of NMME models and a linear statistical model that uses antecedent SST as predictor (Landman et al. 2011). Forecasts for the Niño3.4 area (see insert) are derived from the global forecasts.
- Three-month Niño3.4 SST forecasts are produced for three categories:
  - **El Niño:** SST above the 75th percentile
  - **La Niña:** SST below the 25th percentile
  - **Neutral:** Neither El Niño nor La Niña



# CSiriMM Nino3.4 SST Forecast

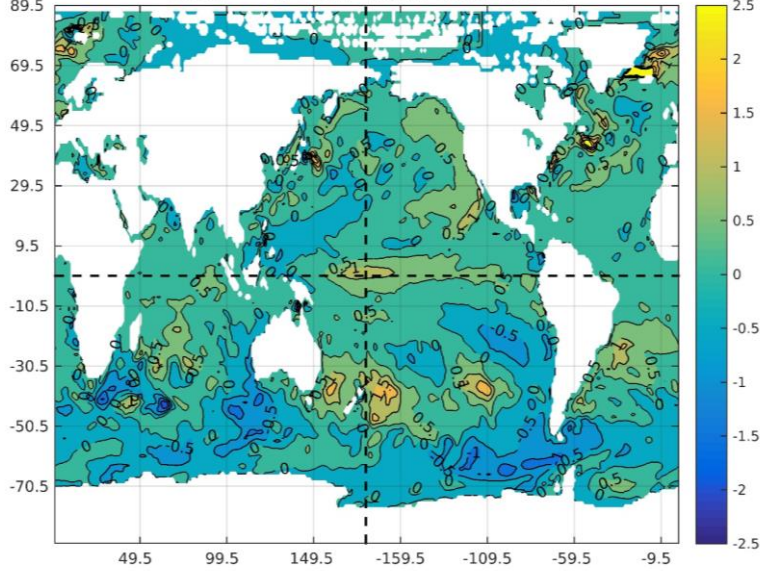
## Issued on: 06-Feb-2019



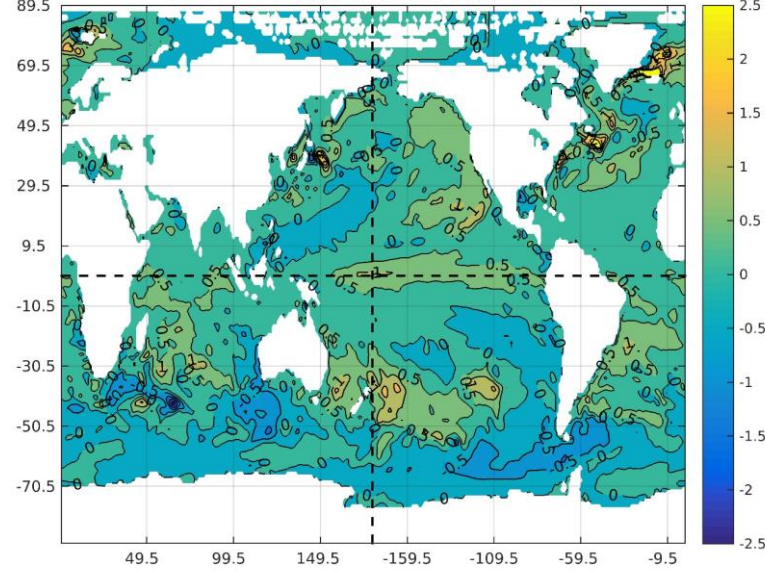


# SST anomalies

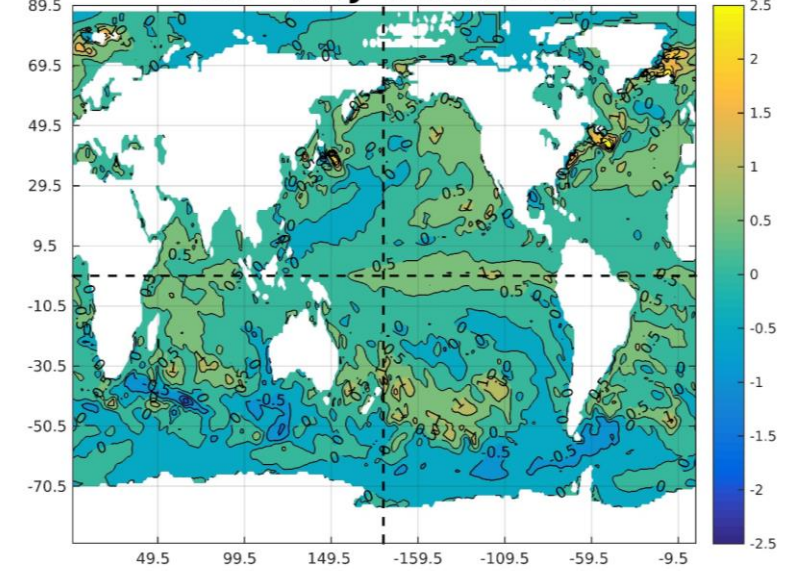
**SST FMA IC:Feb2019**



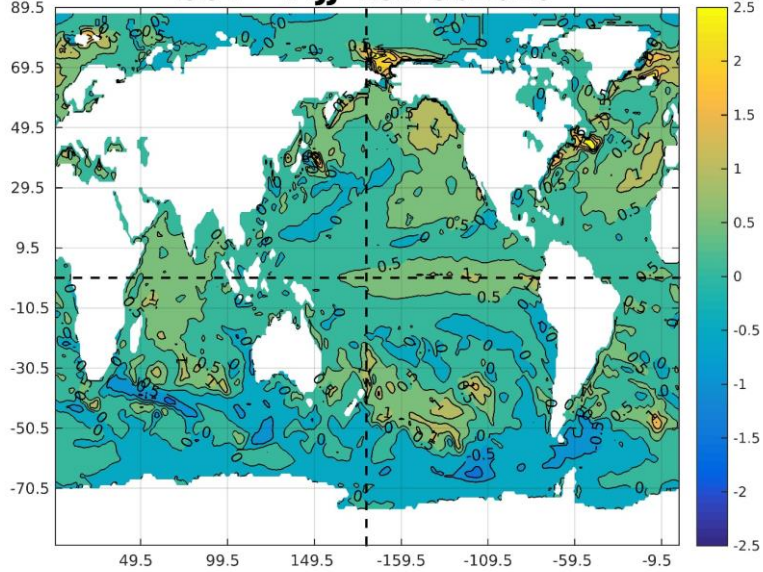
**SST MAM IC:Feb2019**



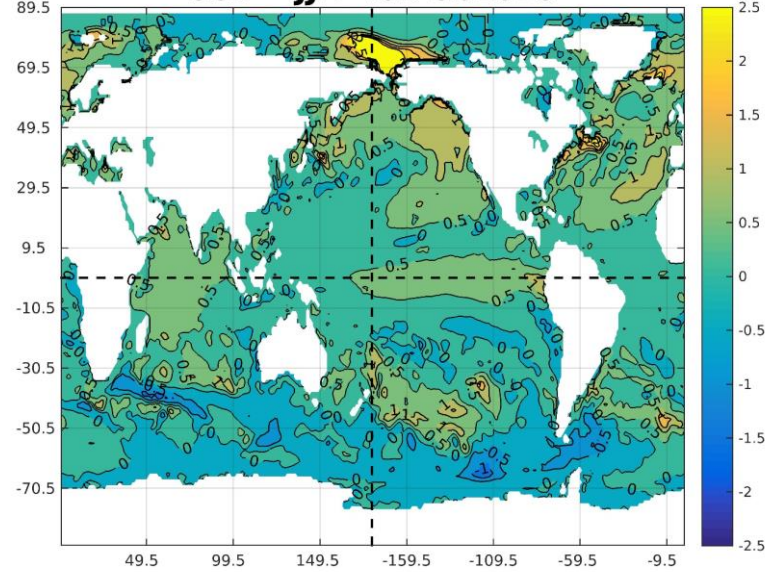
**SST AMJ IC:Feb2019**



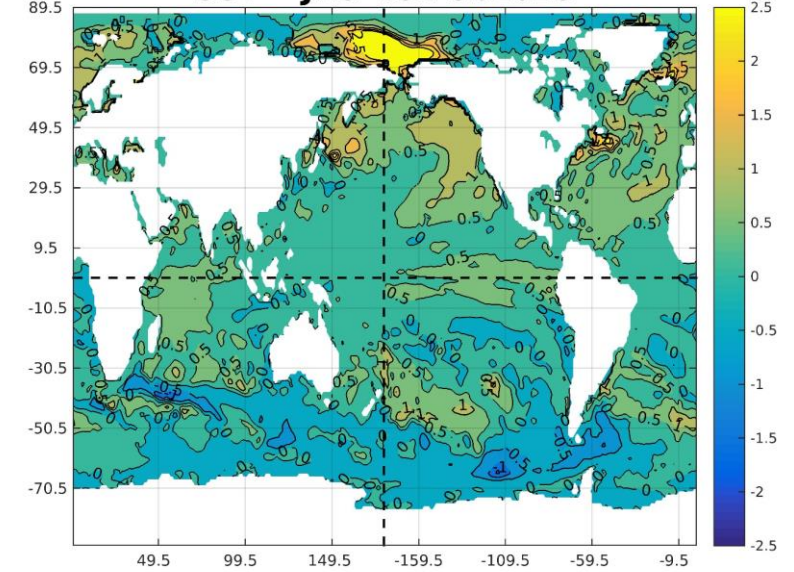
**SST MJJ IC:Feb2019**



**SST JJA IC:Feb2019**



**SST JAS IC:Feb2019**



# Round-up: ENSO

- Weak El Niño into early spring predicted.
- From the CPC/IRI El Niño Watch: *Despite the above-average ocean temperatures across the equatorial Pacific Ocean, the overall coupled ocean-atmosphere system continued to reflect ENSO-neutral.*

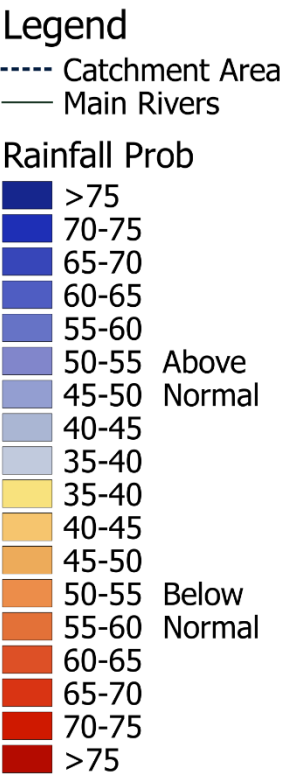
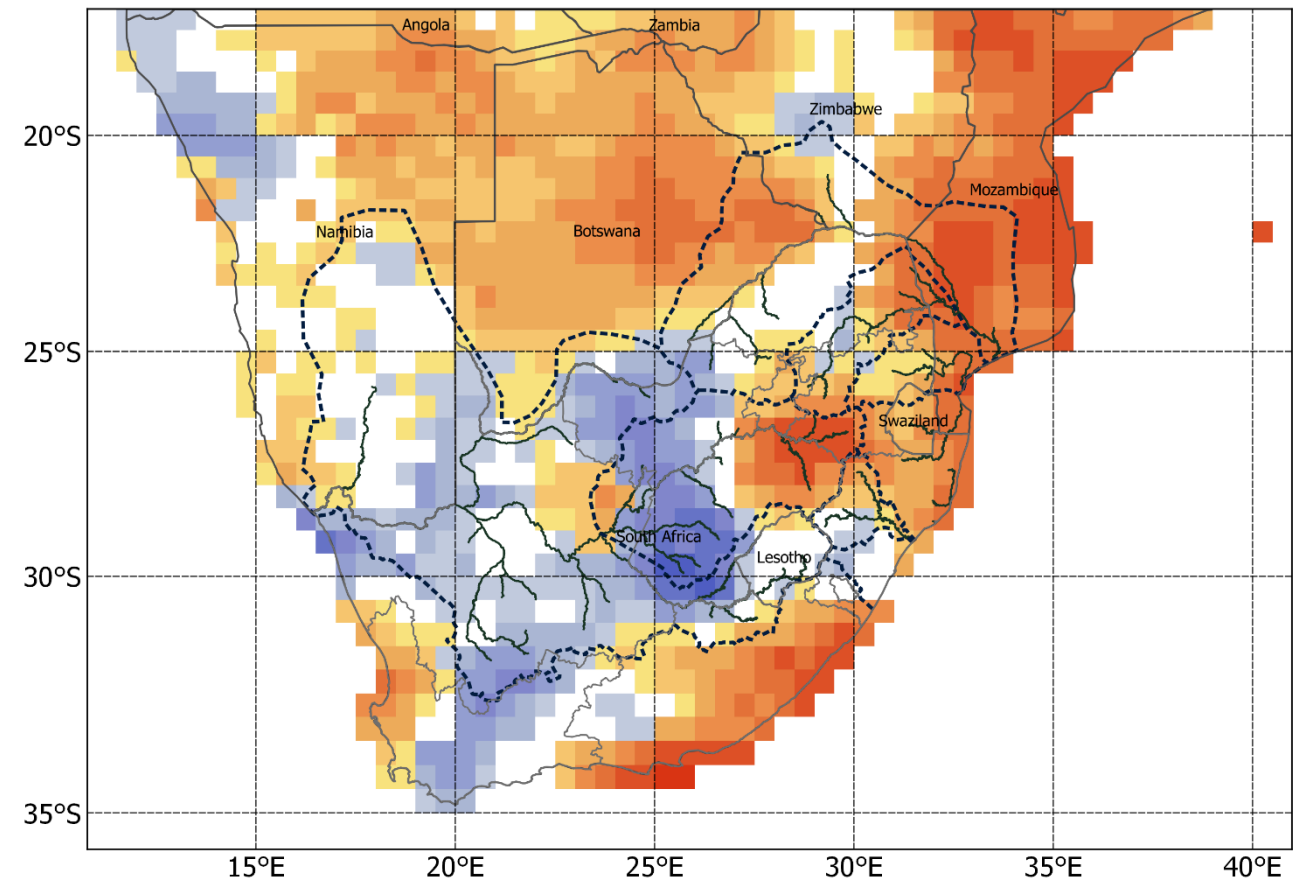
# Southern Africa Forecasts



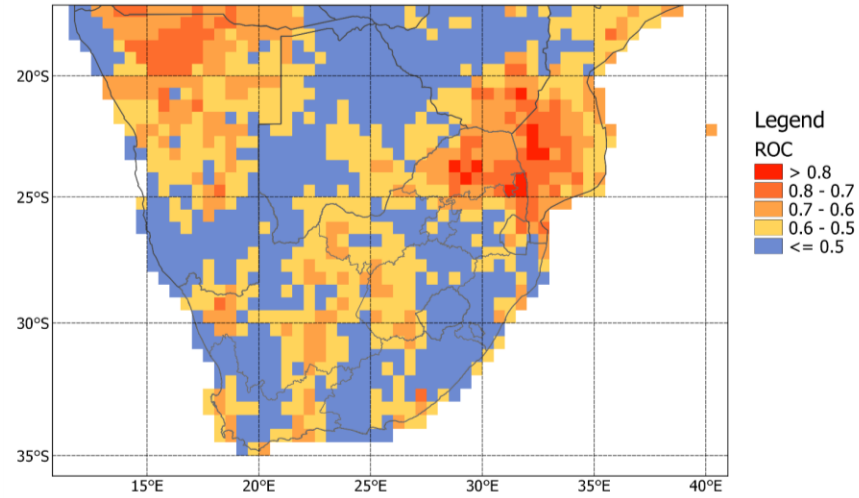
# Prediction Method

- Three-month seasons for seasonal rainfall totals and average maximum temperatures of NMME ensemble mean forecasts are interpolated to Climatic Research Unit (CRU; Harris et al. 2014) grids ( $0.5^{\circ} \times 0.5^{\circ}$ ) by correcting the mean and variance biases of the NMME forecasts. Probabilistic forecasts are subsequently produced from the error variance obtained from a 5-year-out cross-validation process (Troccoli et al. 2008). Forecasts cover a 6-month period.
- Forecasts are produced for three categories:
  - **Above:** Above-normal (“wet” / “hot”, rainfall totals / maximum temperatures higher than the 75th percentile of the climatological record)
  - **Below:** Below-normal (“dry” / “cool”, rainfall totals / maximum temperatures lower than the 25th percentile of the climatological record)
  - **Normal:** Near-normal (“average” season)
- Verification:
  - ROC Area (Below-Normal) – The forecast system’s ability to discriminate dry or cool seasons from the rest of the seasons over a 32-year test period. ROC values should be higher than 0.5 for a forecast system to be considered skilful.
  - ROC Area (Above-Normal) – The forecast system’s ability to discriminate wet or hot seasons from the rest of the seasons over a 32-year test period. ROC values should be higher than 0.5 for a forecast system to be considered skilful.

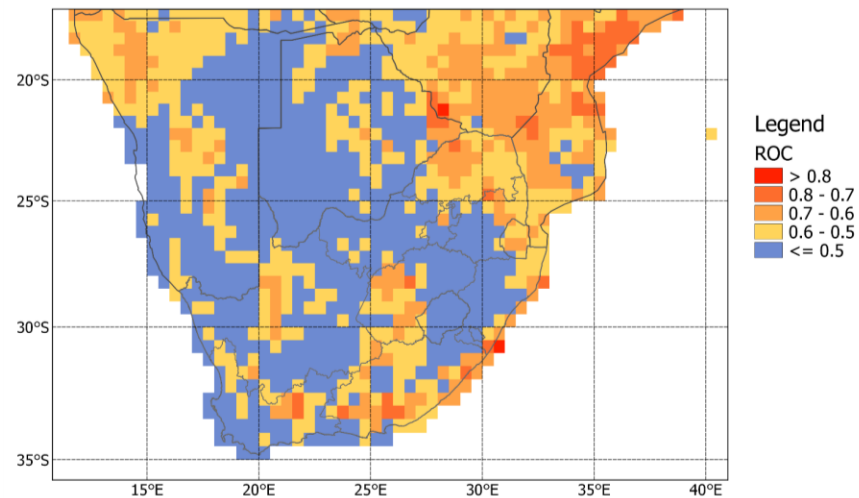
# FMA 2019 Rainfall; ICs: Feb



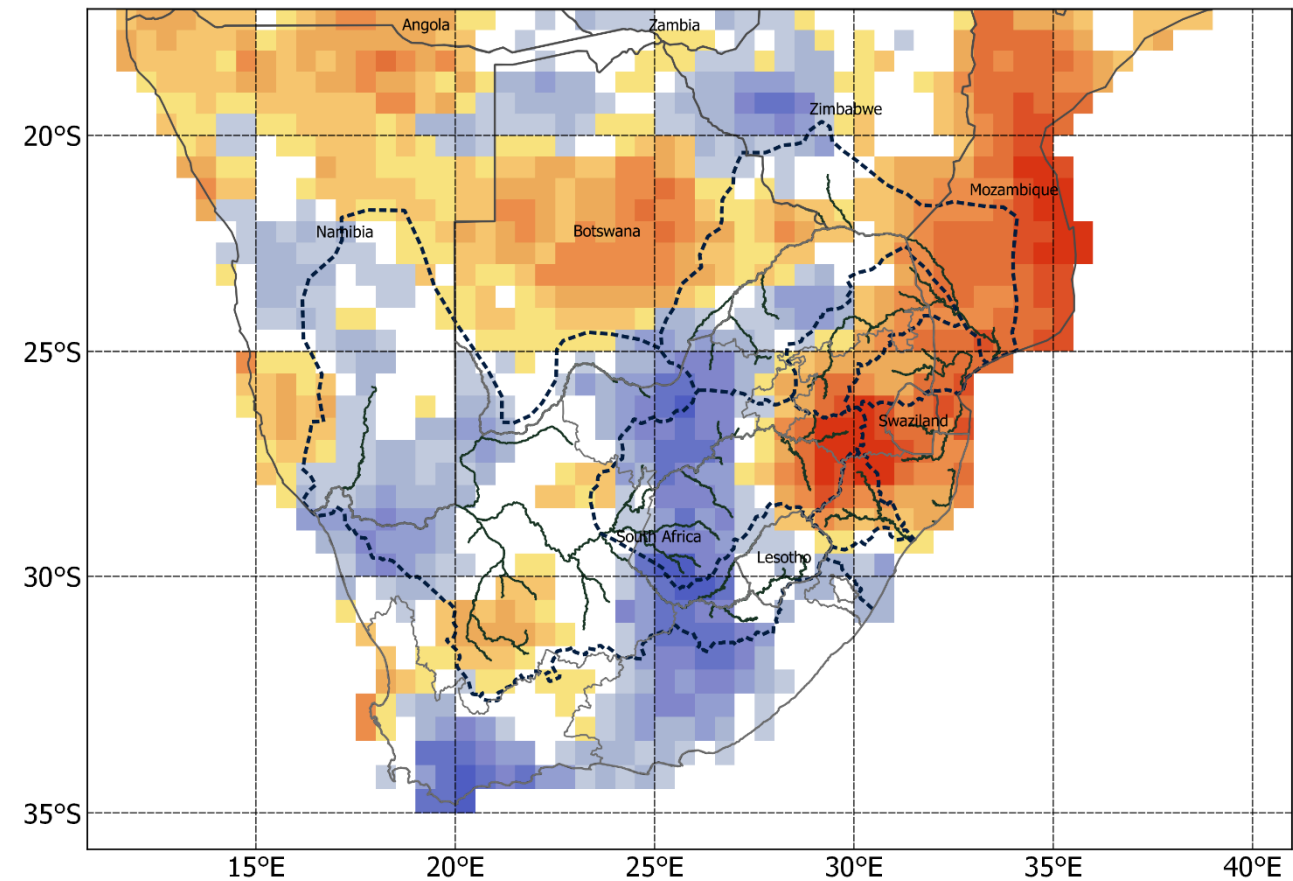
## ROC Area (Above-Normal): FMA Rainfall



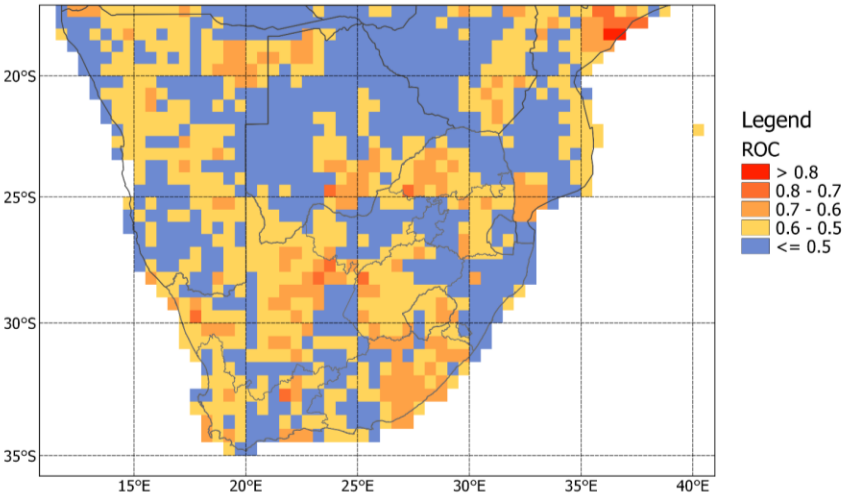
## ROC Area (Below-Normal): FMA Rainfall



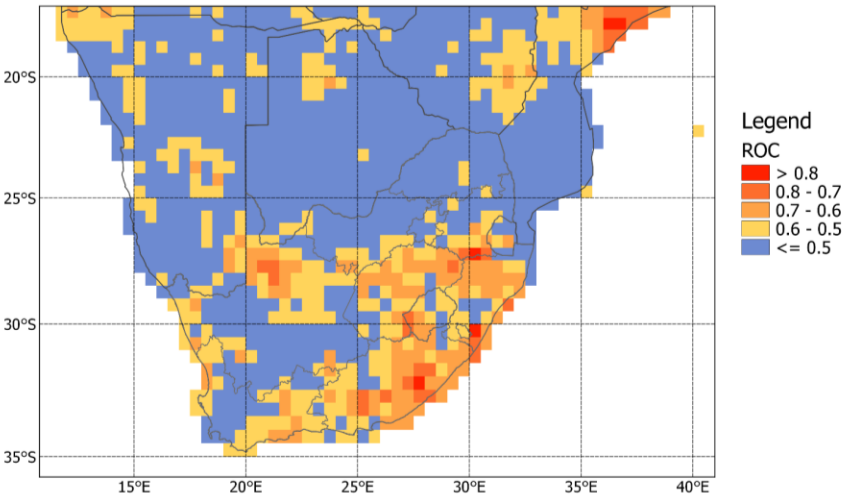
# MAM 2019 Rainfall; ICs: Feb



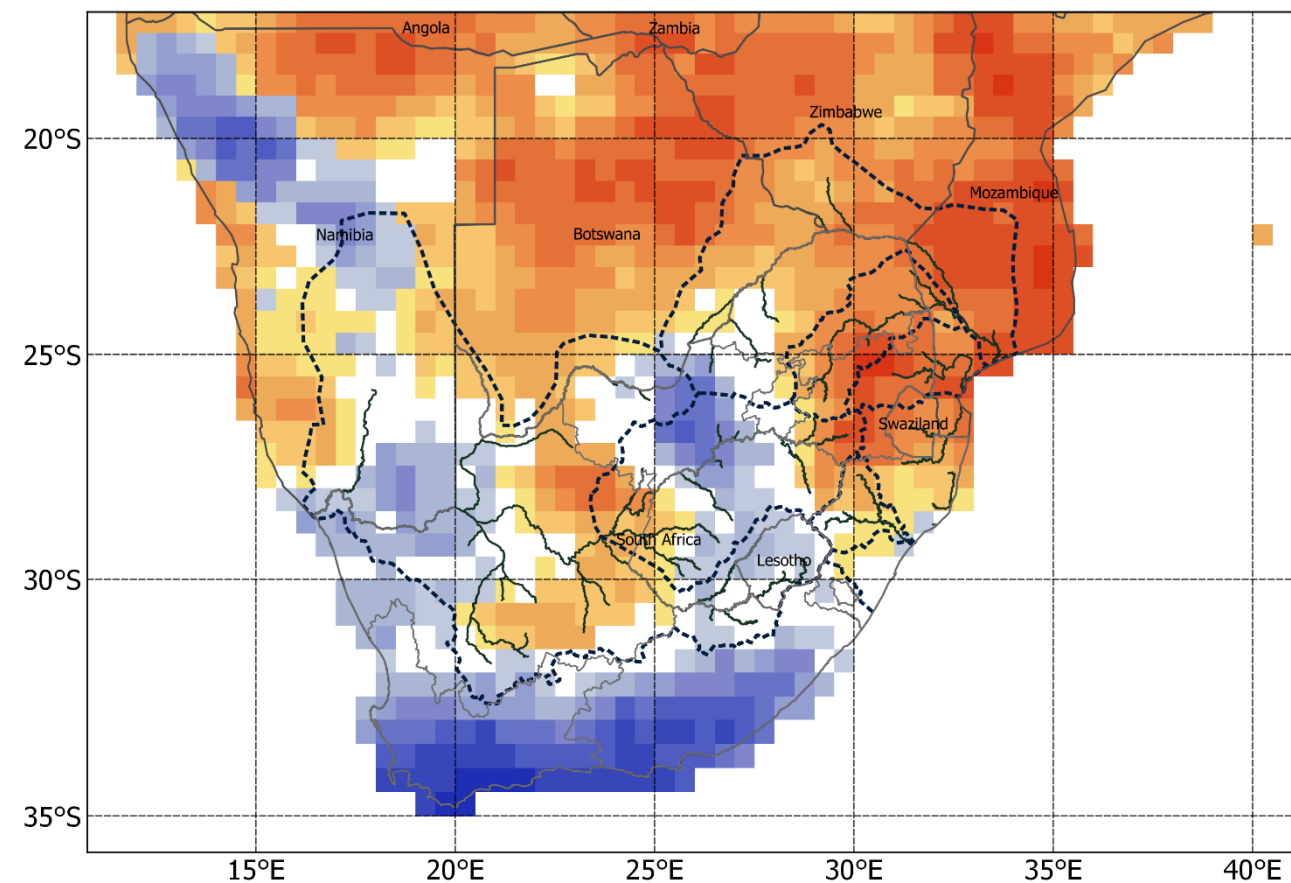
## ROC Area (Above-Normal): MAM Rainfall



## ROC Area (Below-Normal): MAM Rainfall



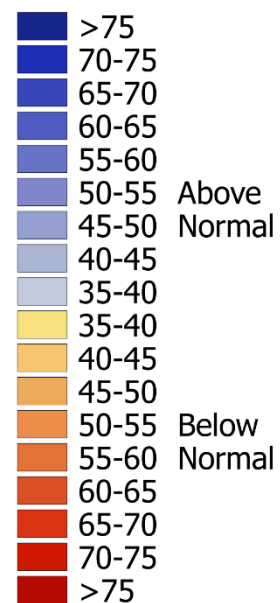
# AMJ 2019 Rainfall; ICs: Feb



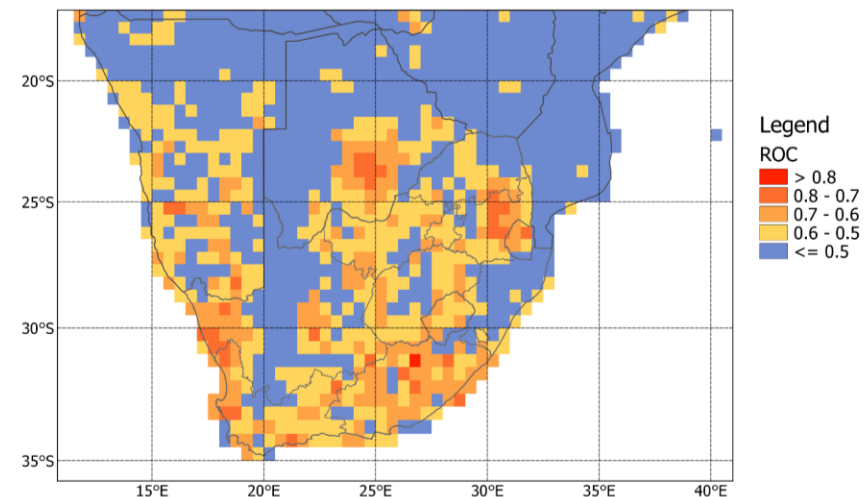
## Legend

--- Catchment Area  
— Main Rivers

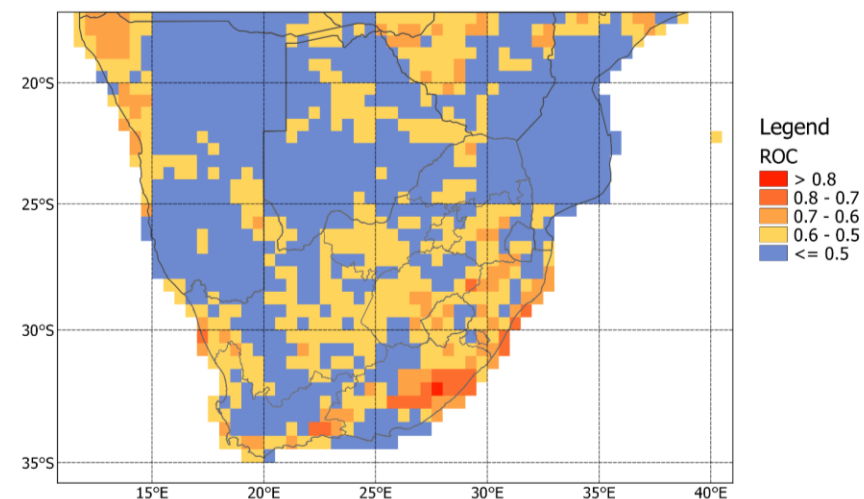
## Rainfall Prob



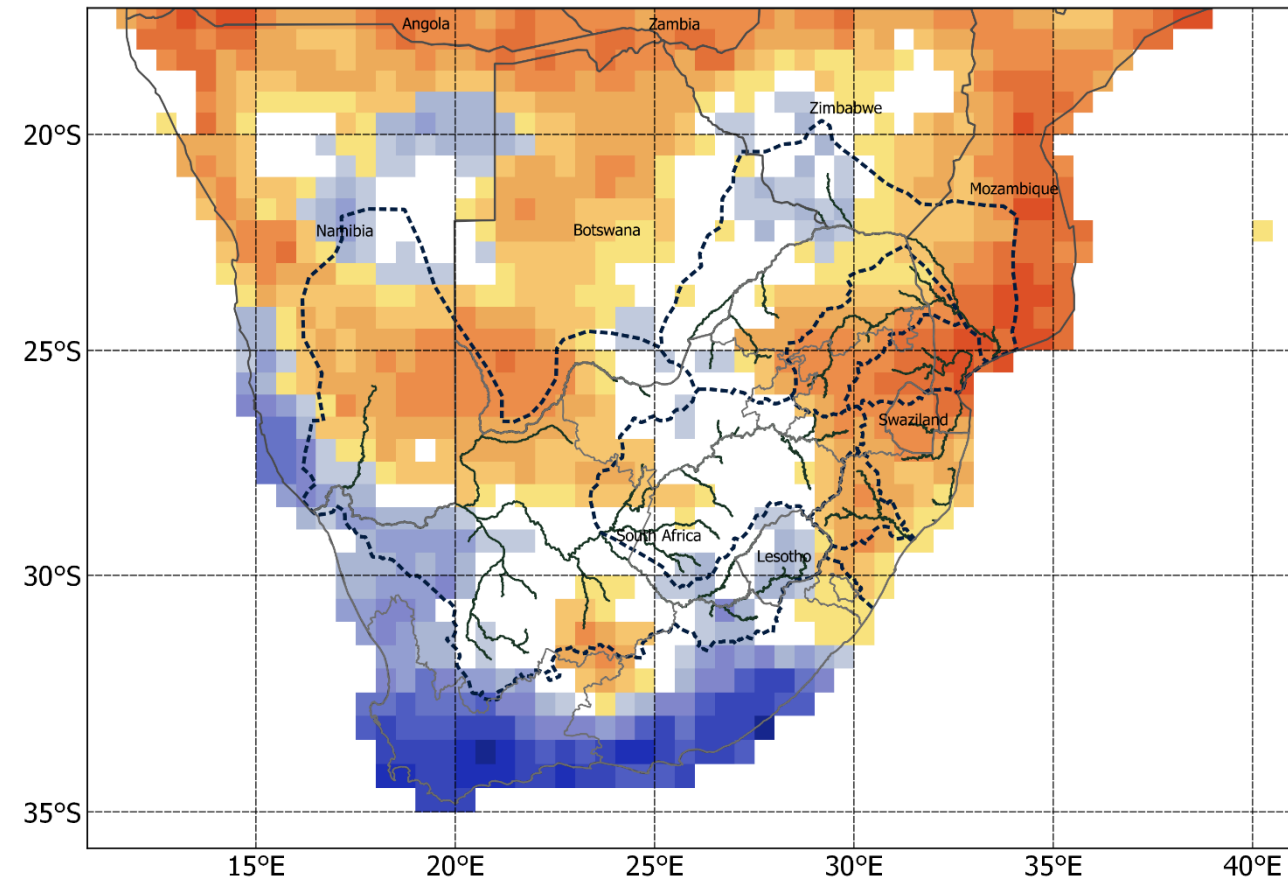
## ROC Area (Above-Normal): AMJ Rainfall



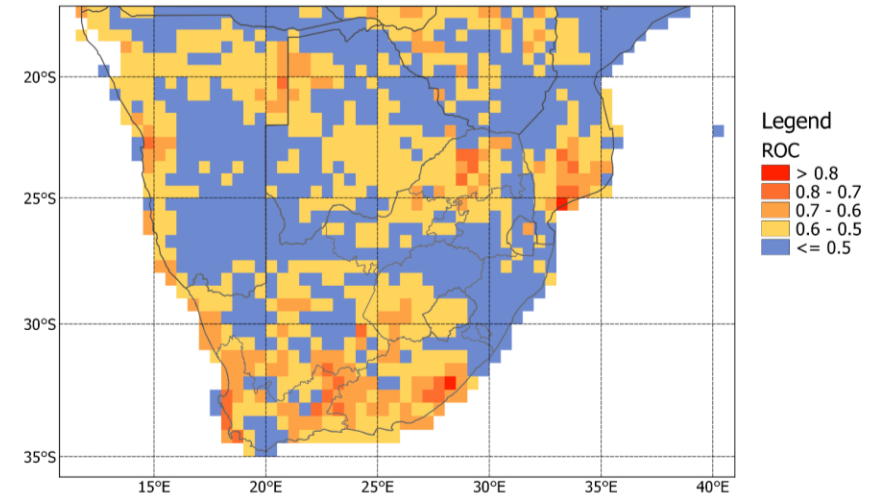
## ROC Area (Below-Normal): AMJ Rainfall



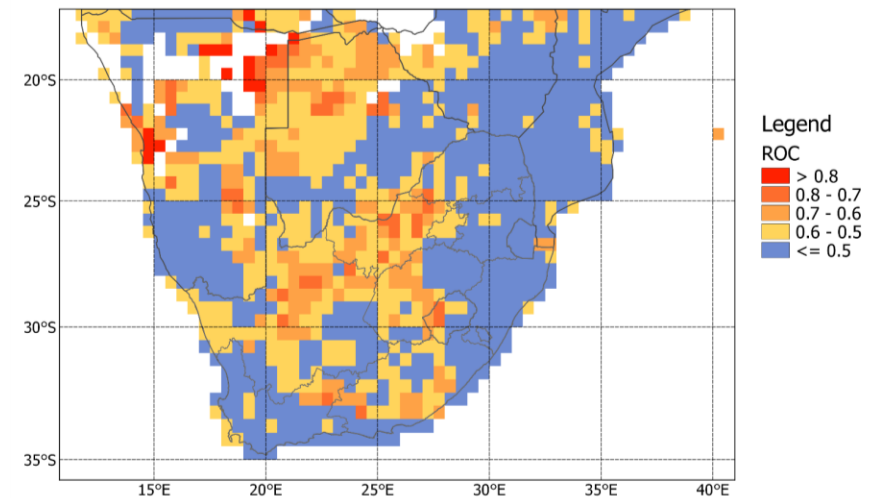
# MJJ 2019 Rainfall; ICs: Feb



## ROC Area (Above-Normal): MJJ Rainfall



## ROC Area (Below-Normal): MJJ Rainfall

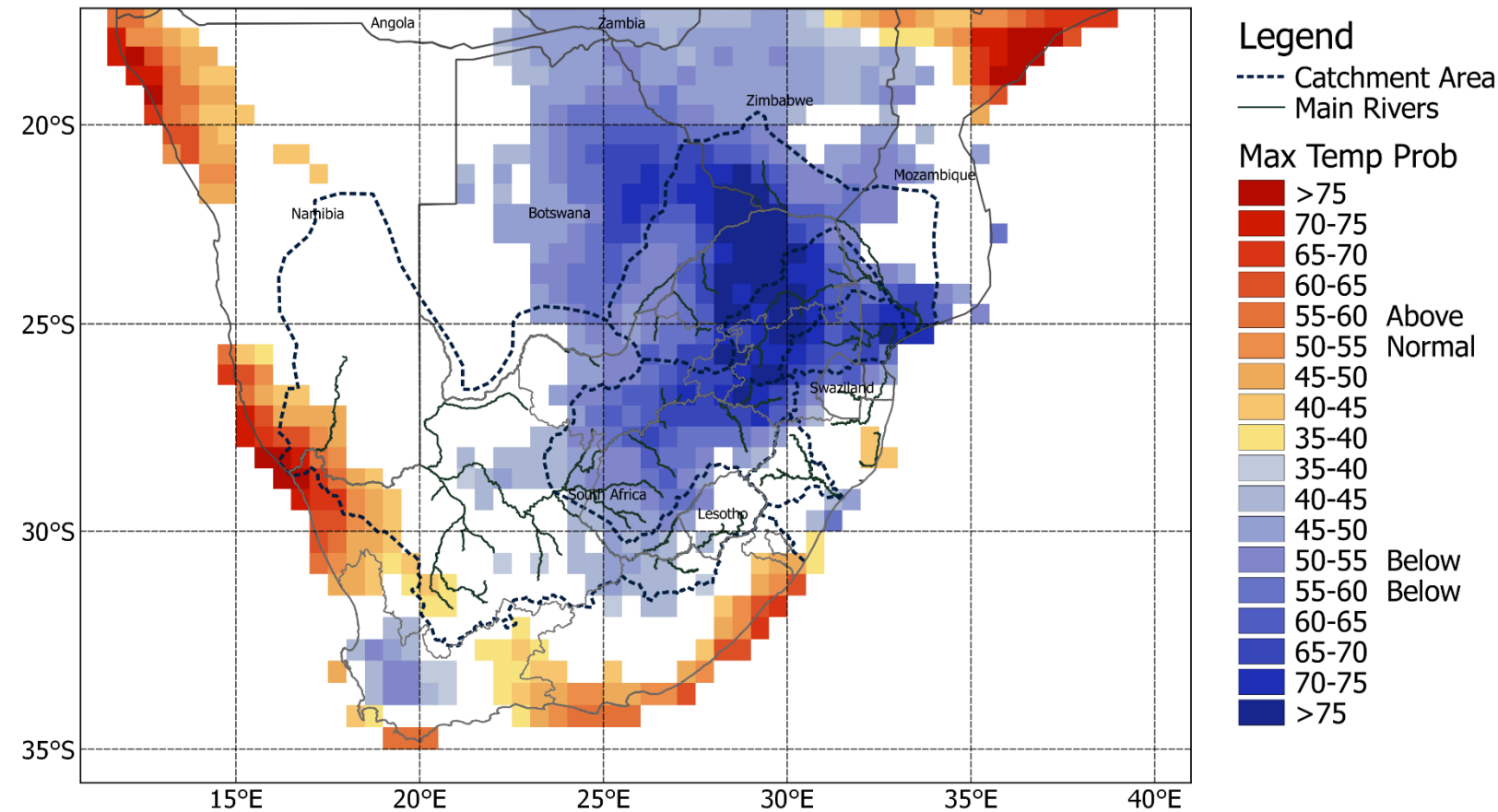




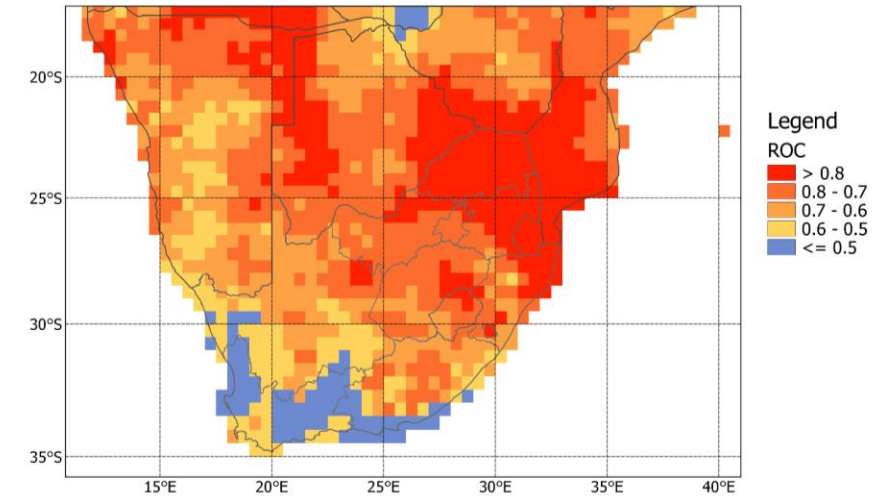
# Round-up: SADC Rainfall

- From the CPC/IRI El Niño Watch: *The atmospheric anomalies (across the equatorial Pacific Ocean) ... have not yet shown a clear coupling to the above-average ocean temperatures.*
- The forecasts for the remainder of summer do not show a typical El Niño-type rainfall pattern of mainly below-normal rainfall totals.
- Enhanced probabilities for favourable rainfall outcomes are predicted over the southern and southwestern Cape areas towards the end of the forecast period.

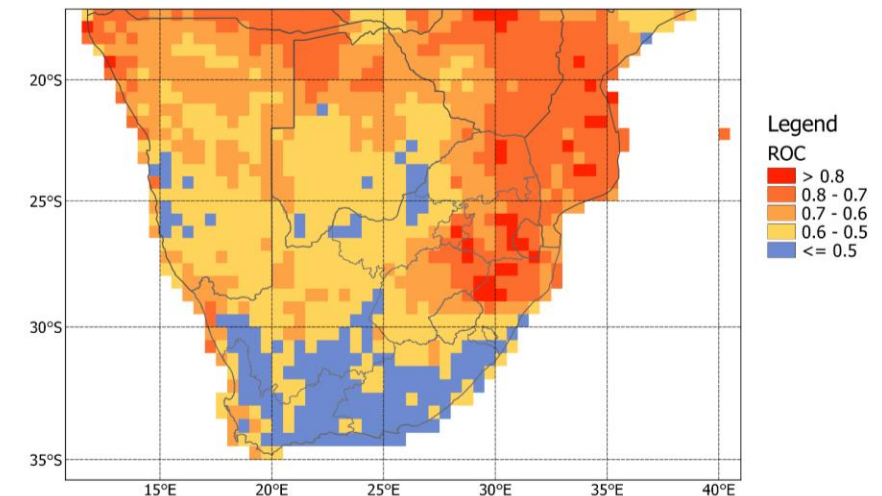
# FMA 2019 Max Temp; ICs: Feb



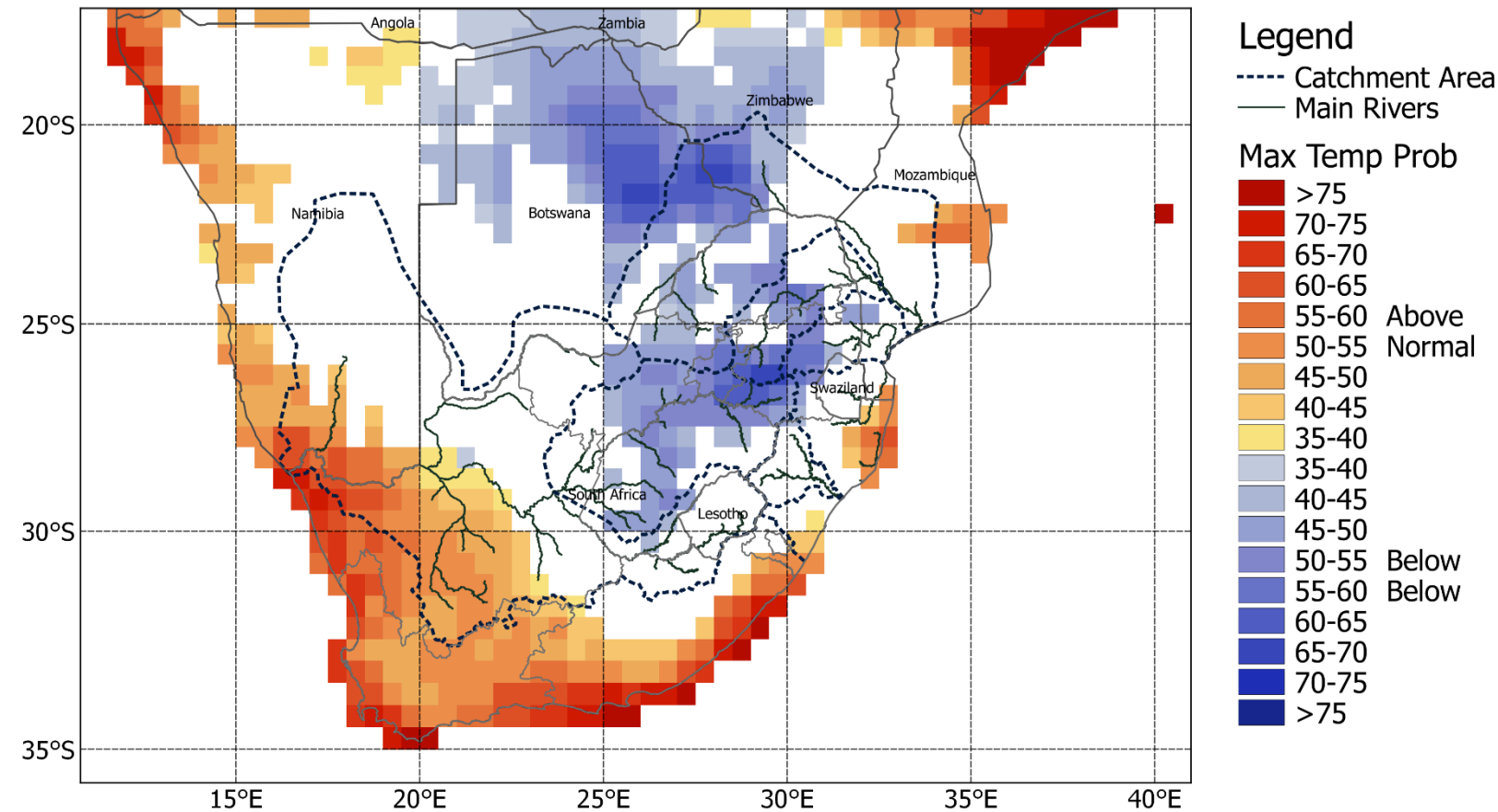
## ROC Area (Above-Normal): FMA Max Temp



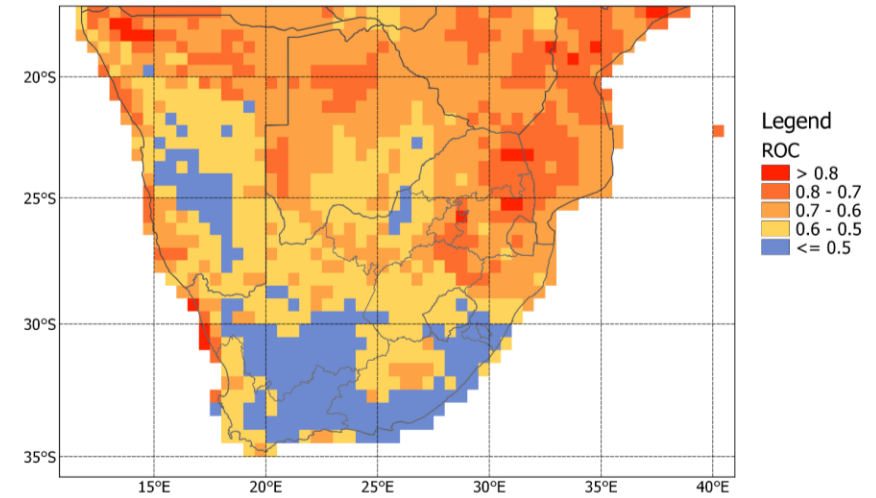
## ROC Area (Below-Normal): FMA Max Temp



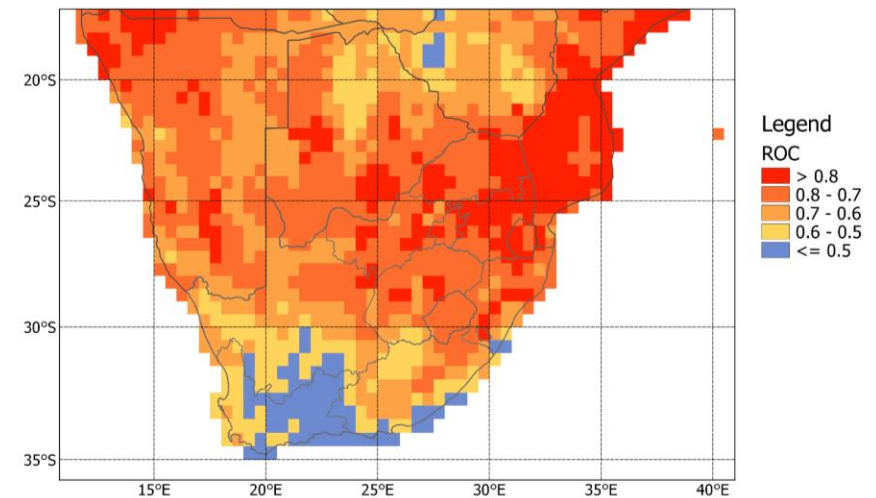
# MAM 2019 Max Temp; ICs: Feb



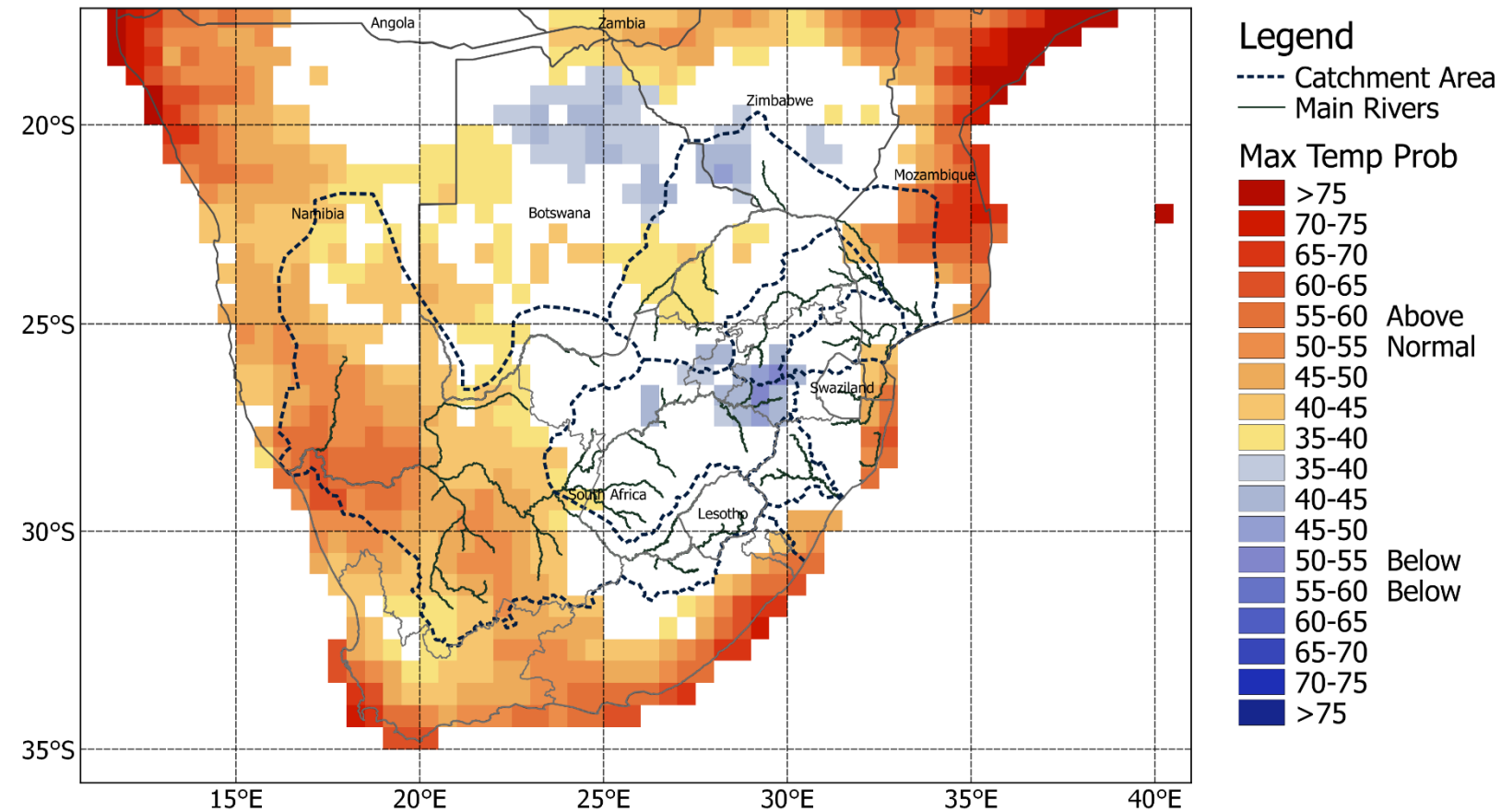
## ROC Area (Above-Normal): MAM Max Temp



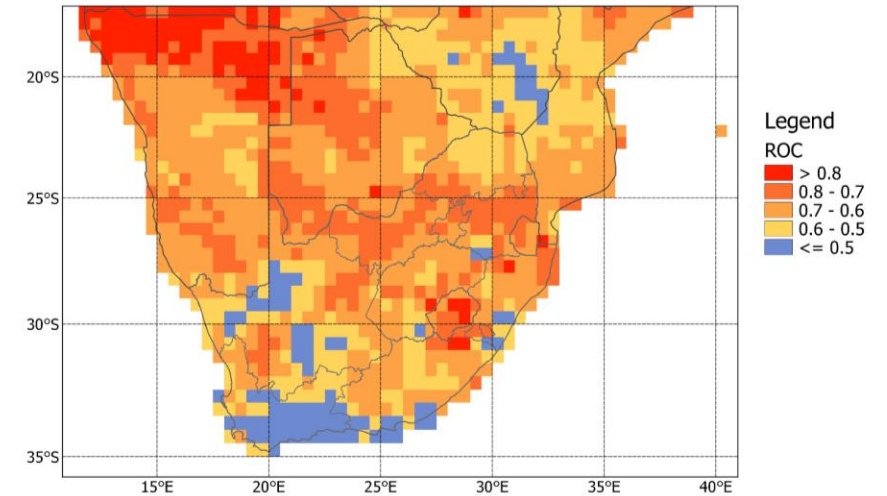
## ROC Area (Below-Normal): MAM Max Temp



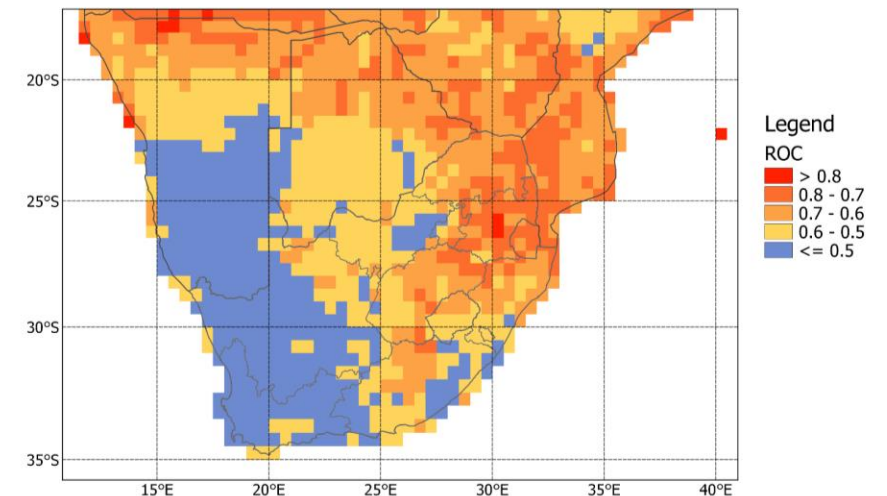
# AMJ 2019 Max Temp; ICs: Feb



## ROC Area (Above-Normal): AMJ Max Temp

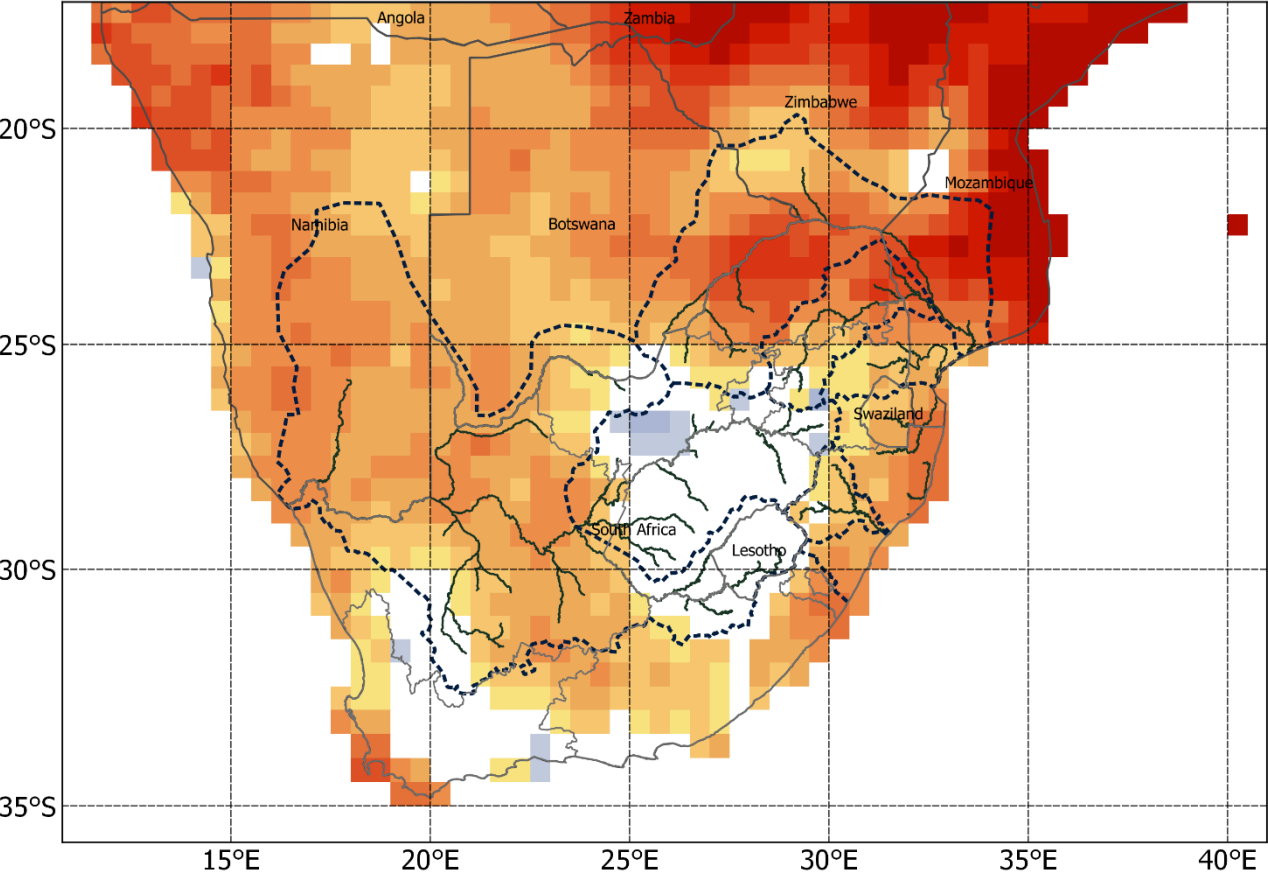


## ROC Area (Below-Normal): AMJ Max Temp

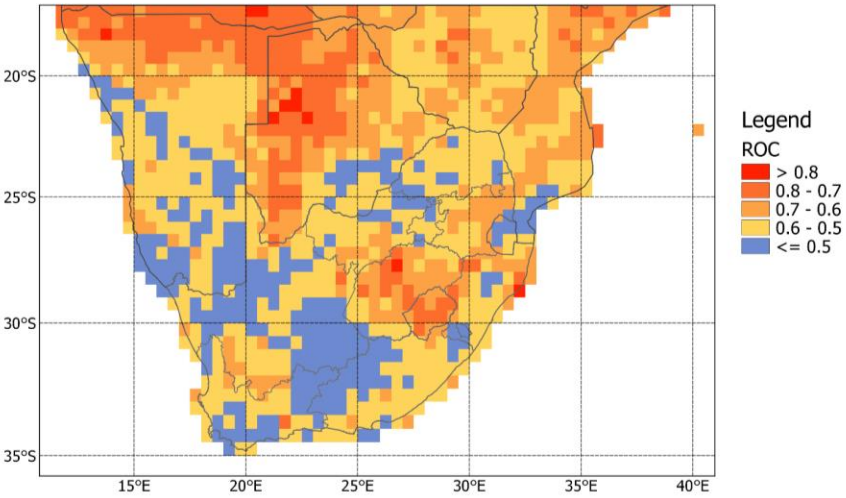




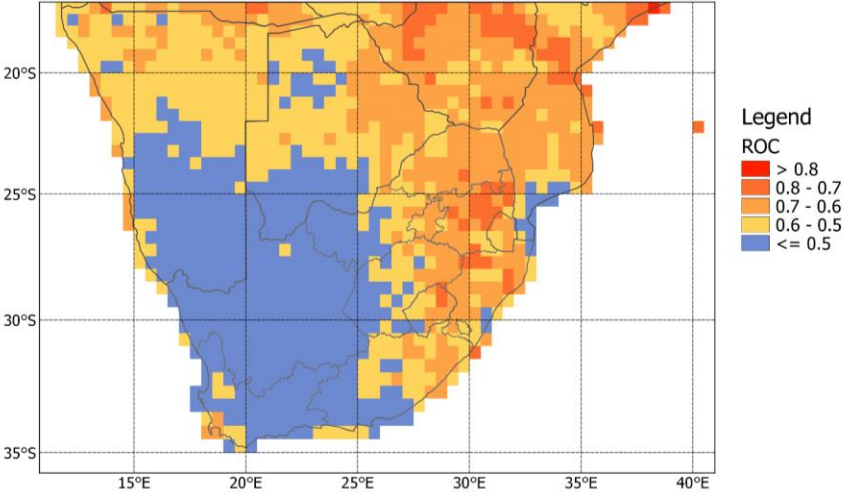
# MJJ 2019 Max Temp; ICs: Feb



ROC Area (Above-Normal): MJJ Max Temp



ROC Area (Below-Normal): MJJ Max Temp





# Round-up: SADC Max Temp

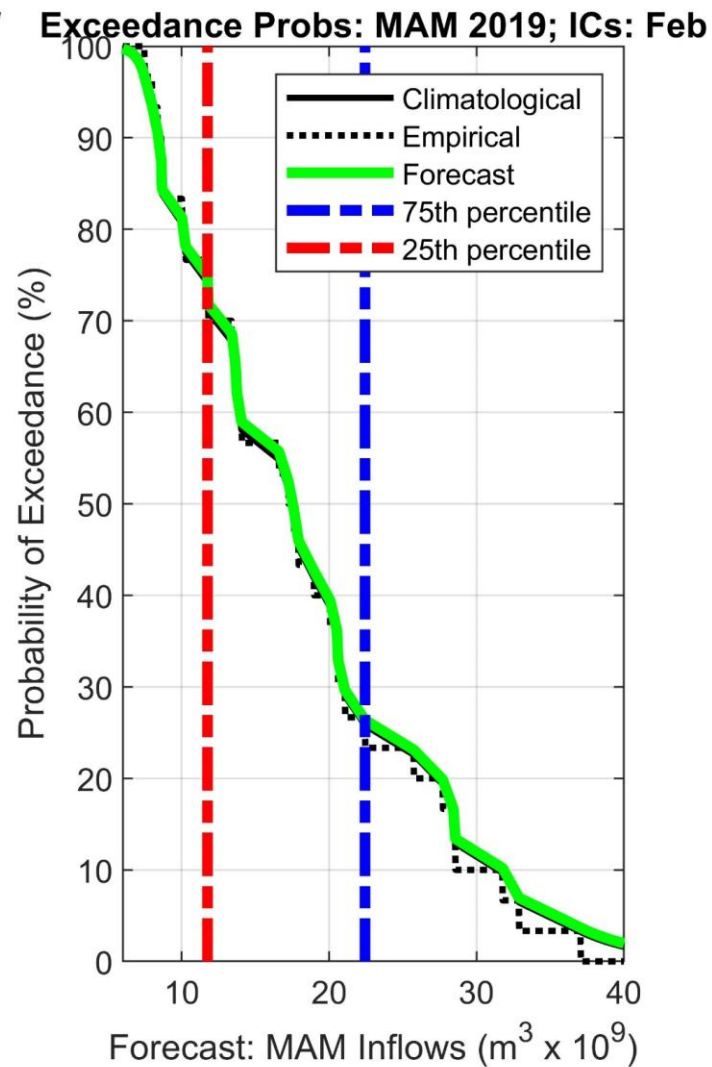
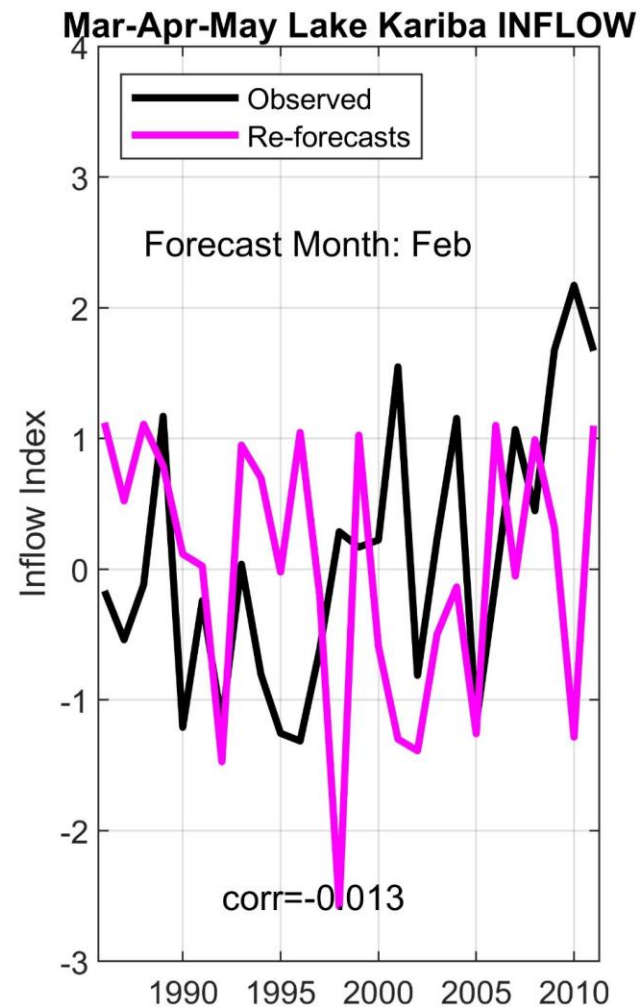
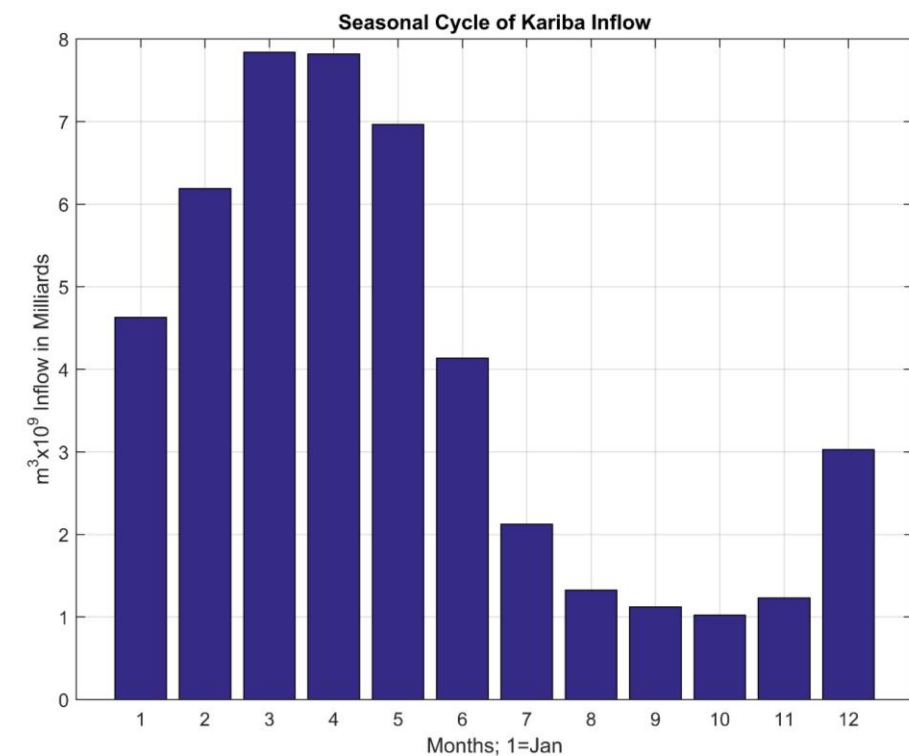
- Mainly cool maximum temperatures predicted over the larger part of the region are replaced by predominantly high maximum temperatures (for that time of the year) towards the end of the forecast period.

# Tailored Forecasts

1. Probability of exceedance Mar-Apr-May 2019 inflow forecast for Lake Kariba, Zambia/Zimbabwe

# Inflow forecast for Lake Kariba: onset season of MAM

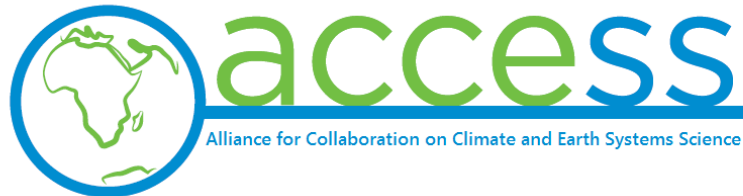
Muchuru et al. (2016)



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- The National Research Foundation through the Incentive Funding for Rated Researchers
- ACCESS (Alliance for Collaboration on Climate and Earth System Science) through the project “Investigating predictability of seasonal anomalies for societal benefit”





# Student participation in forecast system development



**Stephanie Hinze, BSc (Honours)(Meteorology):**

Statistical downscaling using large and high-resolution data sets, forecast displays for SADC rainfall and maximum temperatures, forecast verification



**Surprise Mhlongo, BSc (Honours)(Meteorology):**

Improving on SST forecast system through pattern correction, correlation vs covariance approaches, forecast output combination (multi-model approaches), mean and bias correction, and correct for skill