

# Seasonal forecasts

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# Seasonal Forecast

# Worx



<https://tinyurl.com/ForecastProf>

UNIVERSITEIT VAN PRETORIA  
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**Seasonal Climate Forecasts**

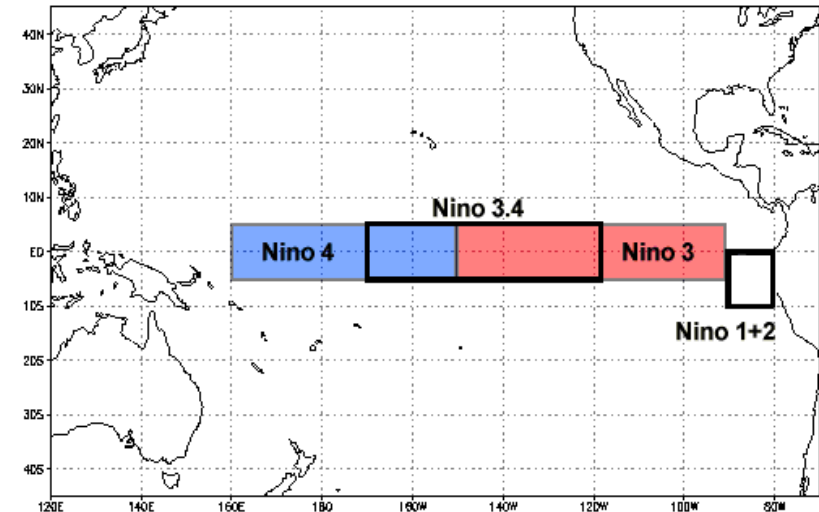
Latest Update: 14 September 2023

- 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?  
“...**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/images/data/longrange/gfcsa/scw.pdf>

# ENSO and Global SST Forecasts

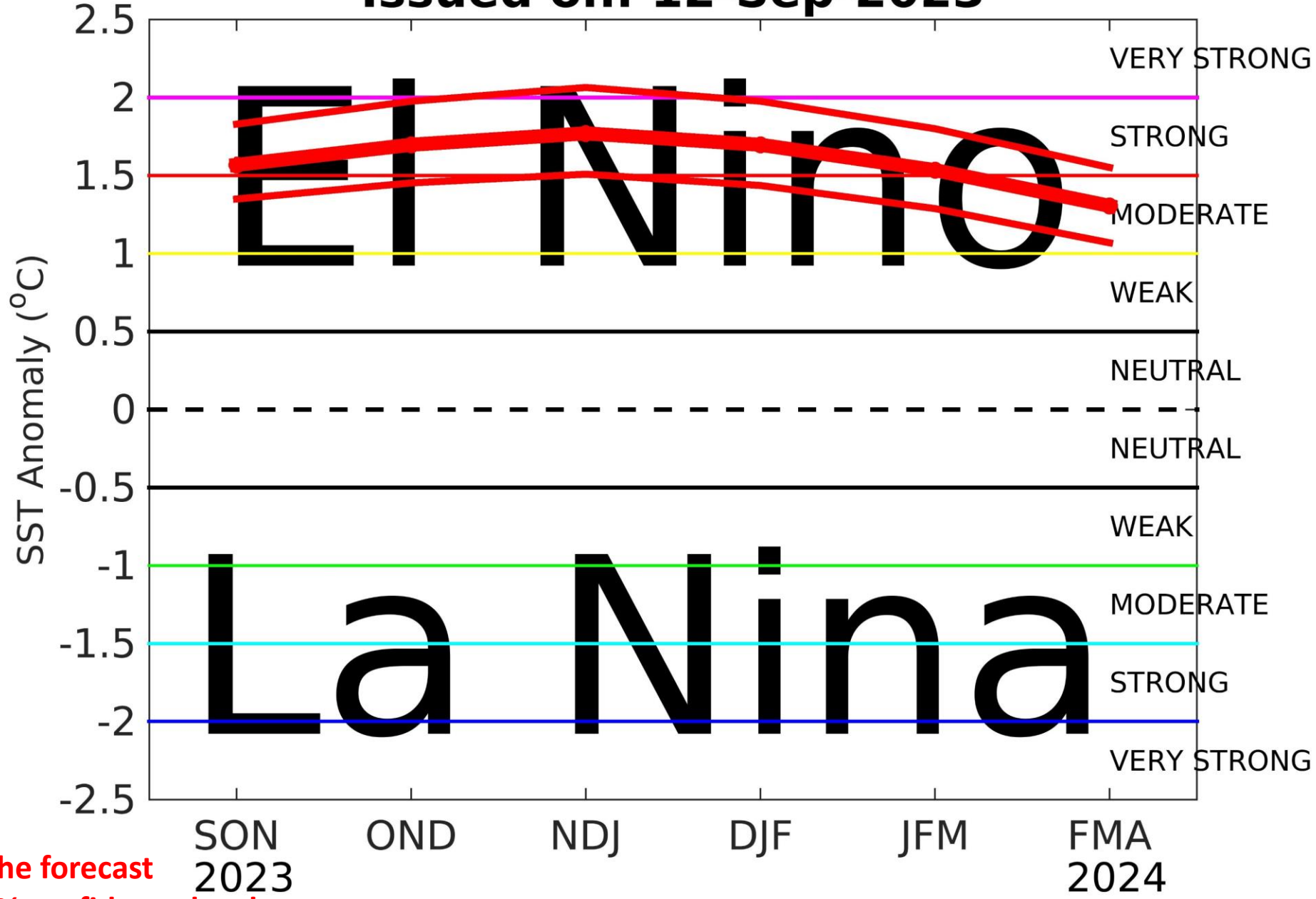
# Prediction Method

- 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 a predictor (Landman et al. 2011). Forecasts for the Niño3.4 area (see insert) are derived from the global forecasts.
- SST forecasts from the NMME models are variance and bias corrected.
- 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

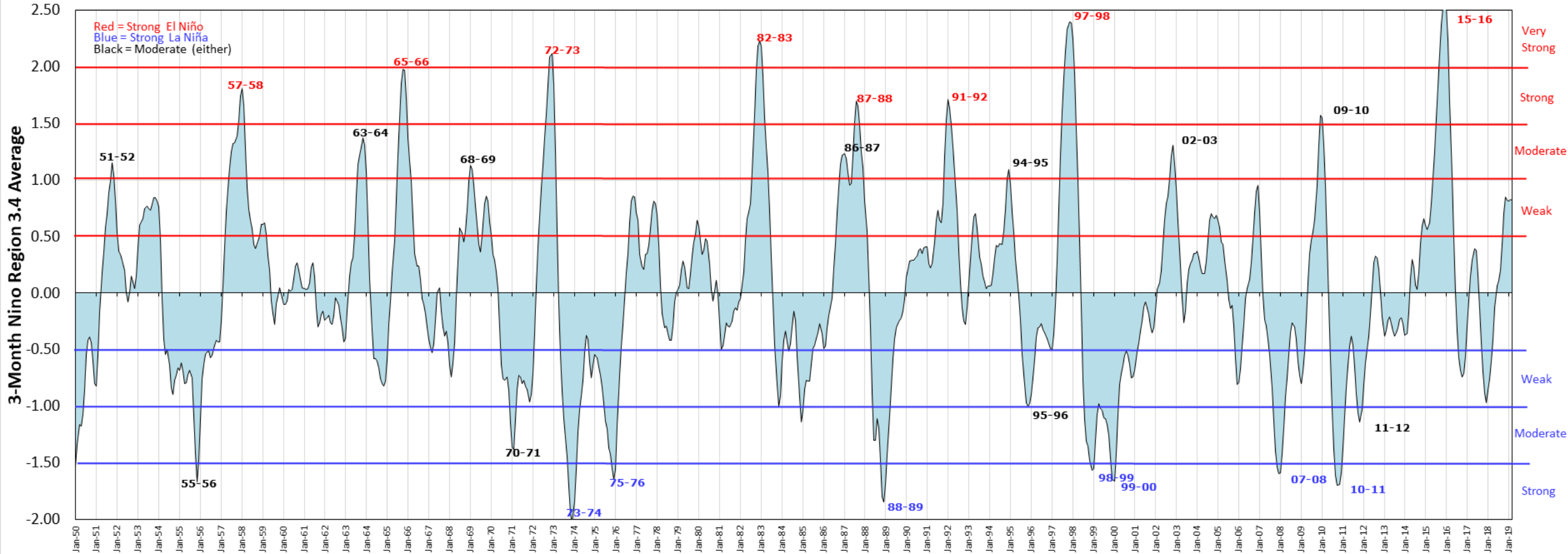
## Issued on: 12-Sep-2023



**Middle red line: the forecast**  
**Thin red lines: 25% confidence levels**

# Oceanic Niño Index (ONI)

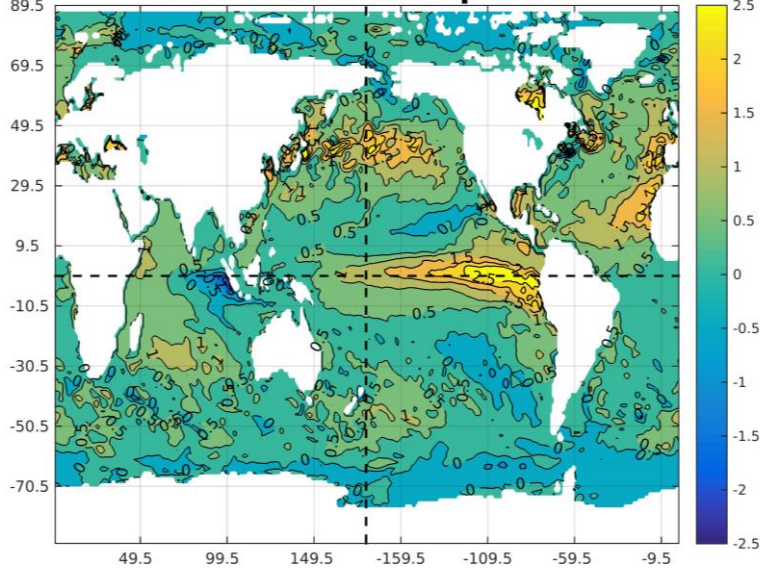
[http://www.cpc.ncep.noaa.gov/products/analysis\\_monitoring/ensostuff/ensoyears.shtml](http://www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ensoyears.shtml)



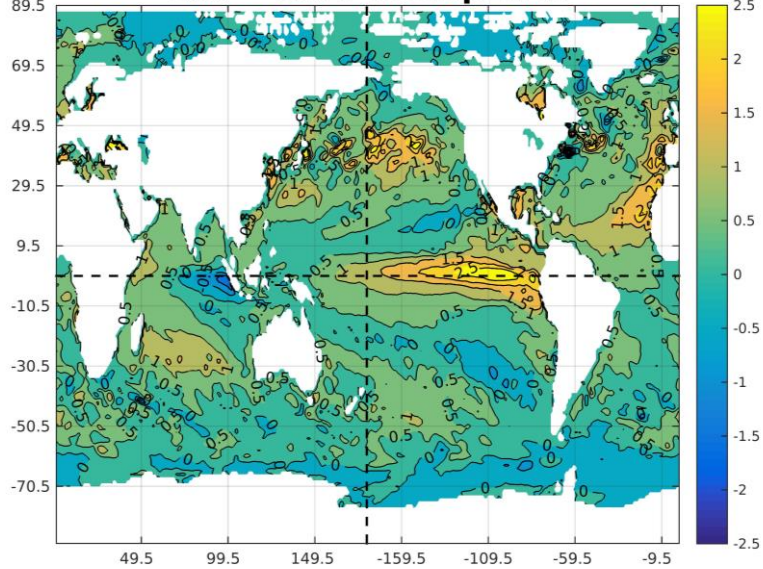
IC: the month in which the forecast was made

# SST anomalies (in °C, where blue is cooler and orange is warmer)

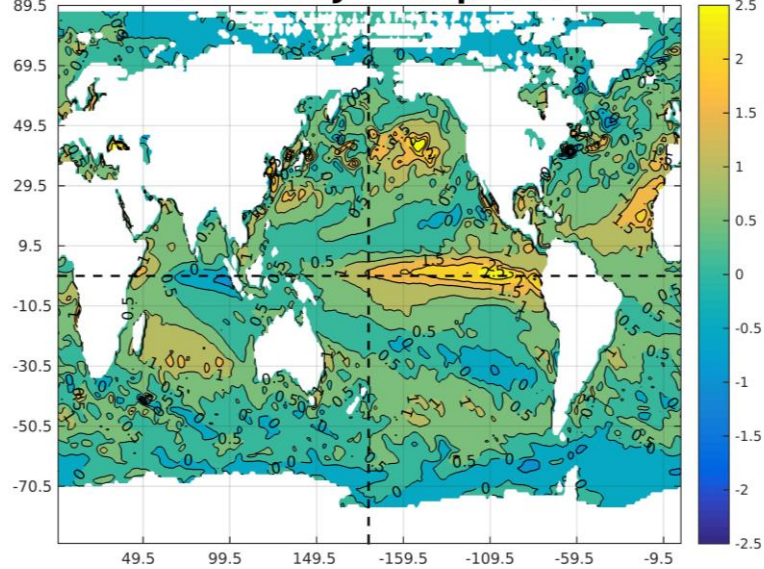
SST SON IC:Sep2023



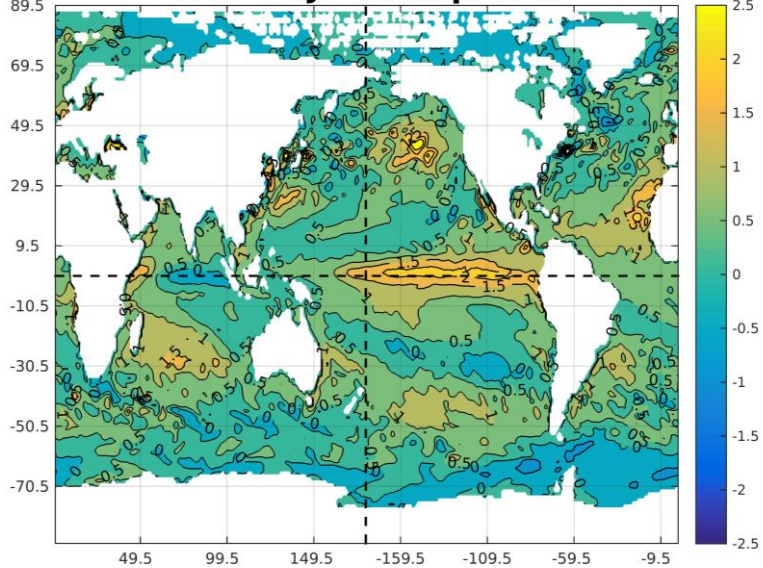
SST OND IC:Sep2023



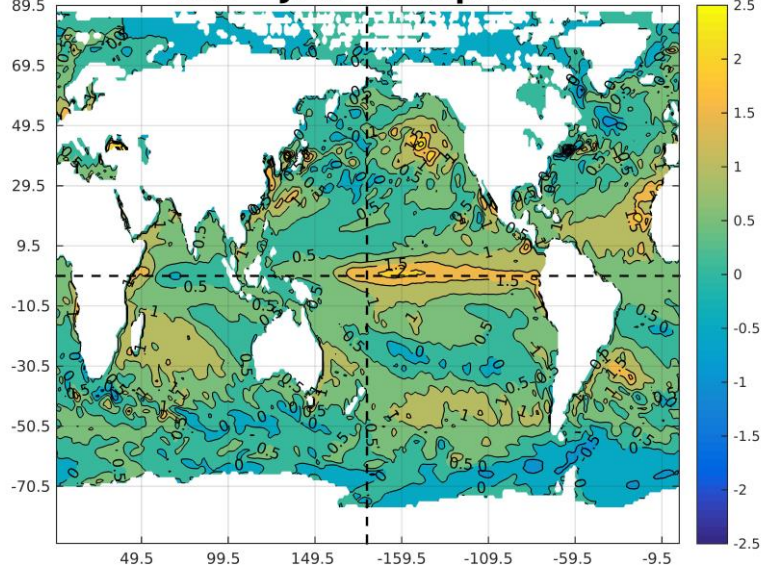
SST NDJ IC:Sep2023



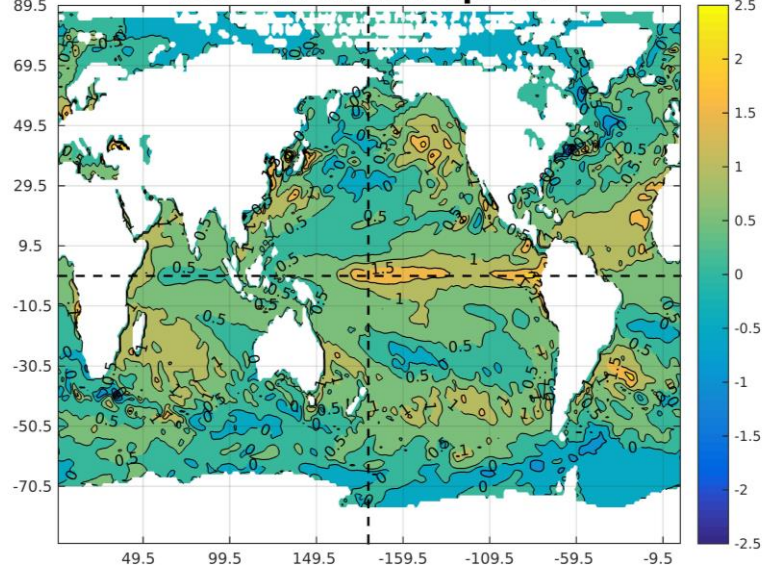
SST DJF IC:Sep2023



SST JFM IC:Sep2023

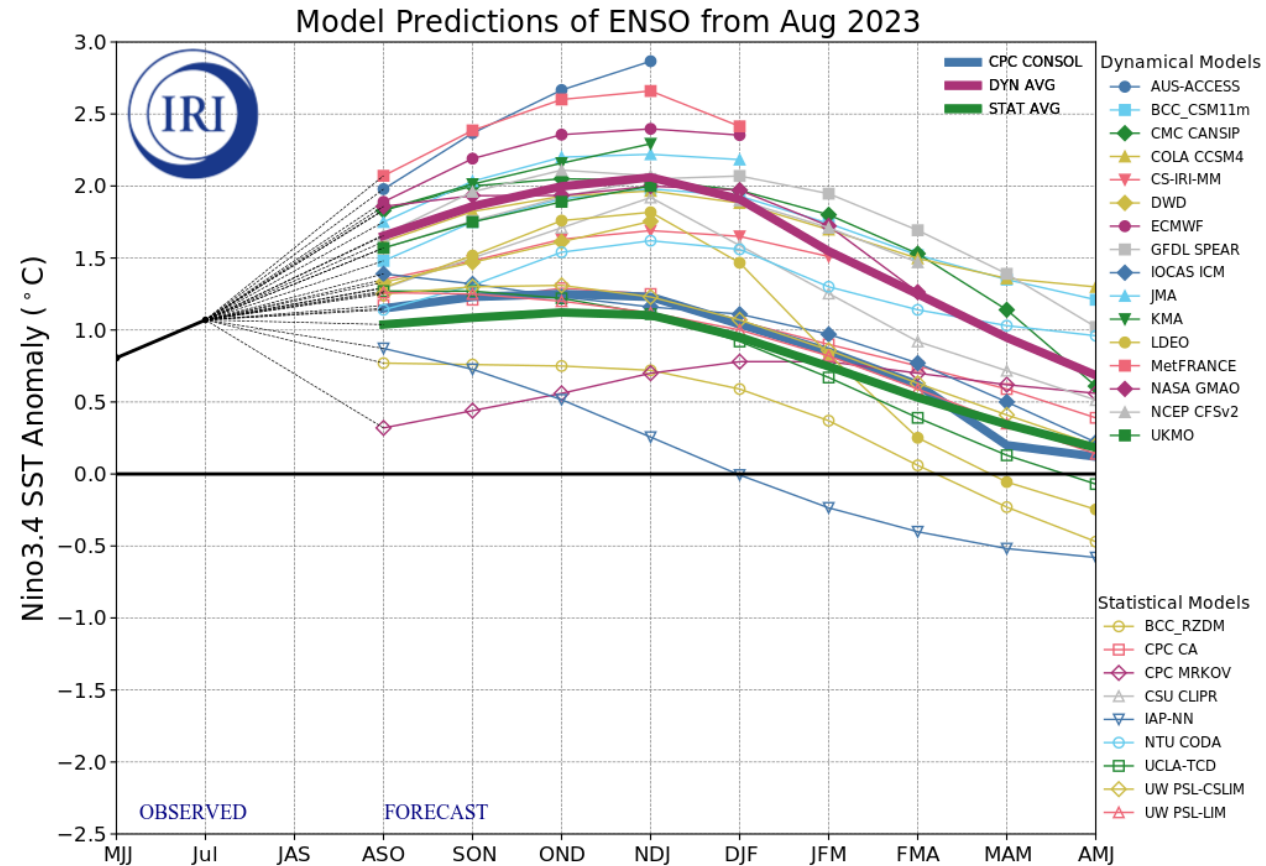
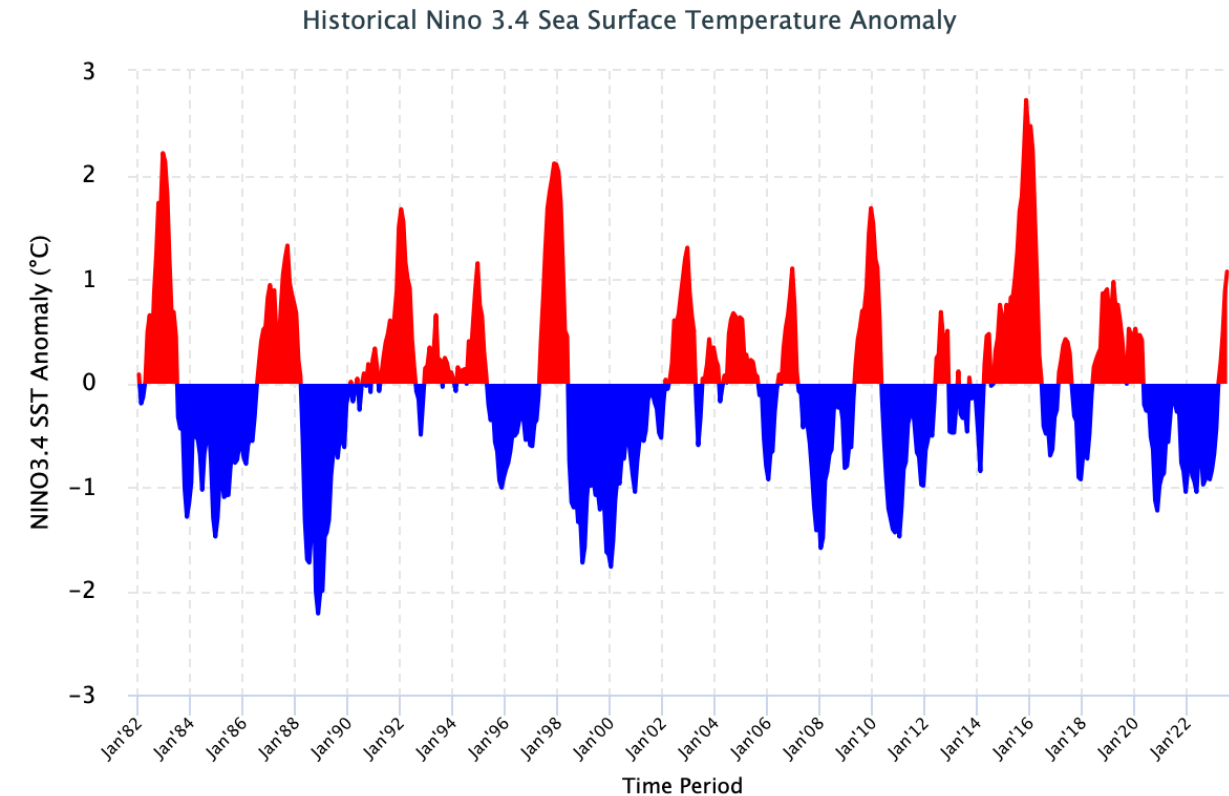


SST FMA IC:Sep2023



# Round-up: ENSO

- The UP model is predicting a strong El Niño event during the Austral mid-summer period





Africa forecasts, south of 20°N

# Prediction Method

- Three-month seasons for seasonal rainfall totals and average maximum temperatures of NMME ensemble mean forecasts are recalibrated to the Climatic Research Unit (CRU; Harris et al. 2014) grids (0.5°x0.5°). 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” rainfall totals / “hot” maximum temperatures higher than the 75th percentile of the climatological record)
  - **Below:** Below-normal (“dry” rainfall totals / “cool” maximum temperatures lower than the 25th percentile of the climatological record)
  - **Normal:** Near-normal (“average” season)

## **NEW!!!**

- Verification of forecast performance:
  - ROC Area (Below-Normal) – The forecast system’s ability to discriminate dry or cool seasons from the rest of the seasons over a 23-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 23-year test period. ROC values should be higher than 0.5 for a forecast system to be considered skilful.
  - The white areas on the forecast maps
    - Forecasts for the near-normal category do not have skill and are therefore not shown
    - Forecasts associated with ROC values less than or equal to 0.5 (no skill) are also not shown

# Forecasts are probabilistic

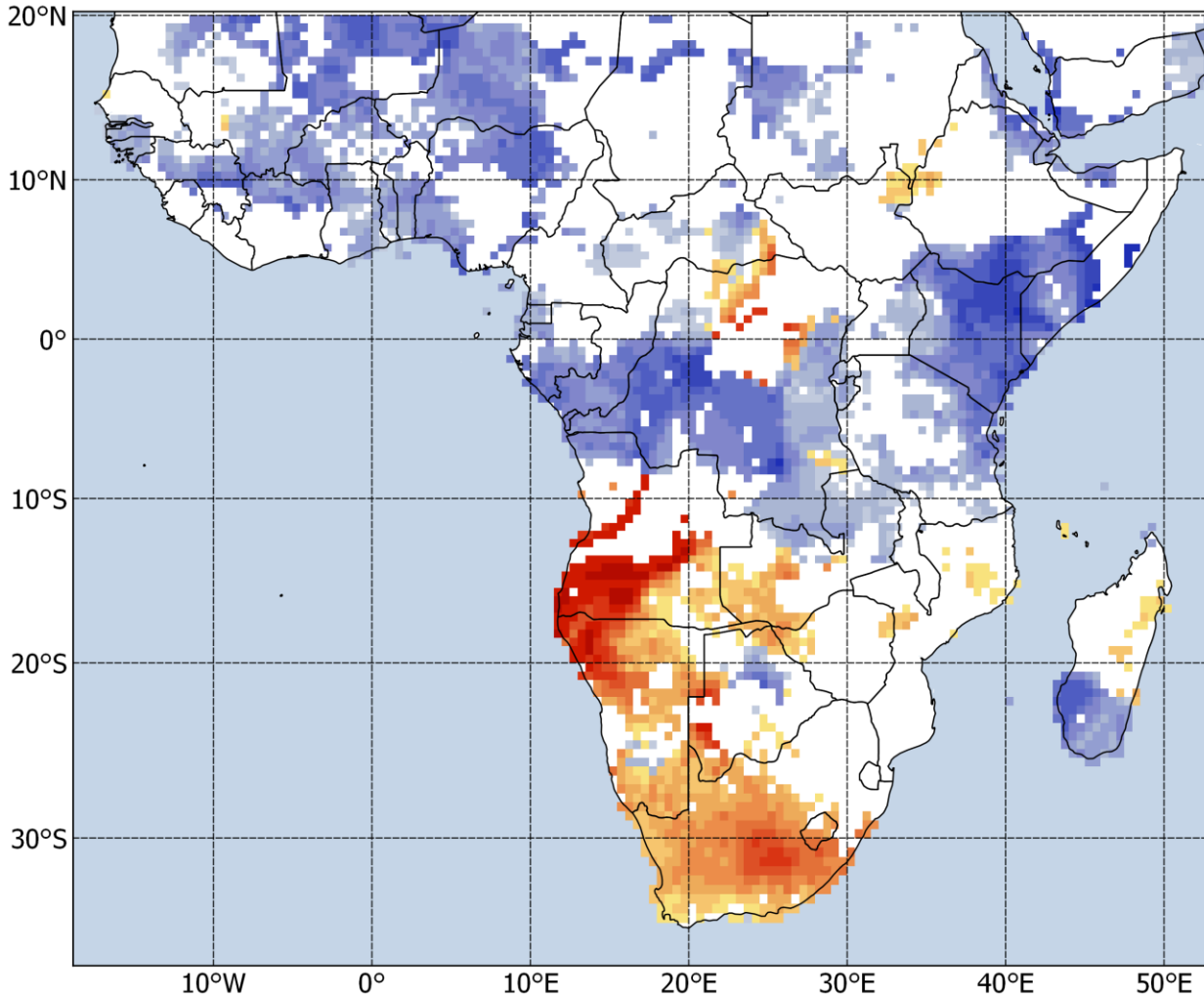
Probabilistic forecasts can help users understand risks and opportunities (forewarned is forearmed) in order to make more informed decisions.

The seasonal rainfall and maximum temperature forecasts to follow are expressed in probabilities, shown as the % chance of the most likely outcome of 3 categories. The colour of the scale reflects the most likely category and the % shows the probability of that outcome. Only ONE of the ROC area skill assessment maps should be consulted, depending on the category shown on the forecast map (Above- or Below-Normal), and the higher the ROC value, the more skilful the forecast for that pixel is. The probabilities shown are always less than 100% - so there is no absolute certainty that the less favoured outcome will not occur. For example, if the forecast claims a 75% chance of below-normal rainfall totals for a season (i.e. drought), it means that 1 out of 4 times it will **not** develop into a drought.

The nature of a probabilistic forecast implies that the less likely outcomes are always possible. In fact, for the probabilistic forecasts to be considered reliable, the less likely outcomes will and must occasionally occur.

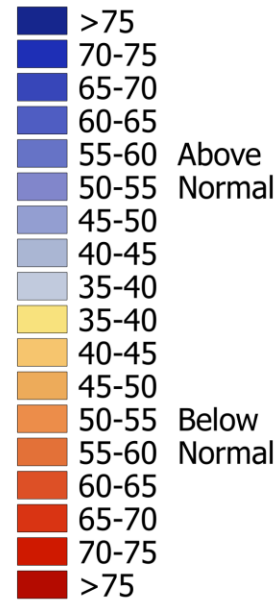
Note: Probabilistic forecasts are considered reliable when the forecast probability is an accurate estimation of the relative frequency of the predicted outcome. In other words, forecasts are reliable if the observation falls within the category (Below-, Near- or Above-Normal) as frequently as the forecast implies

# SON 2023 Rainfall; ICs: Sep



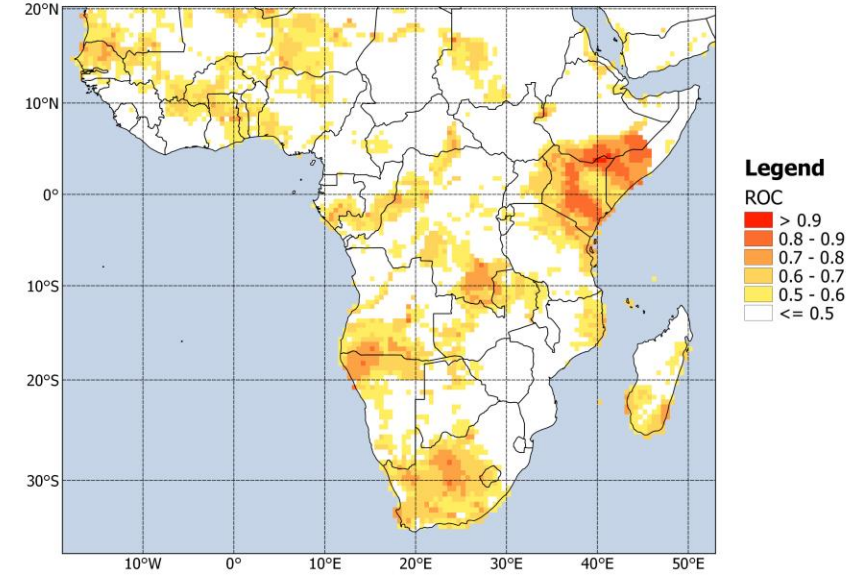
## Legend

### Rainfall Prob

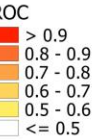


□ No forecast

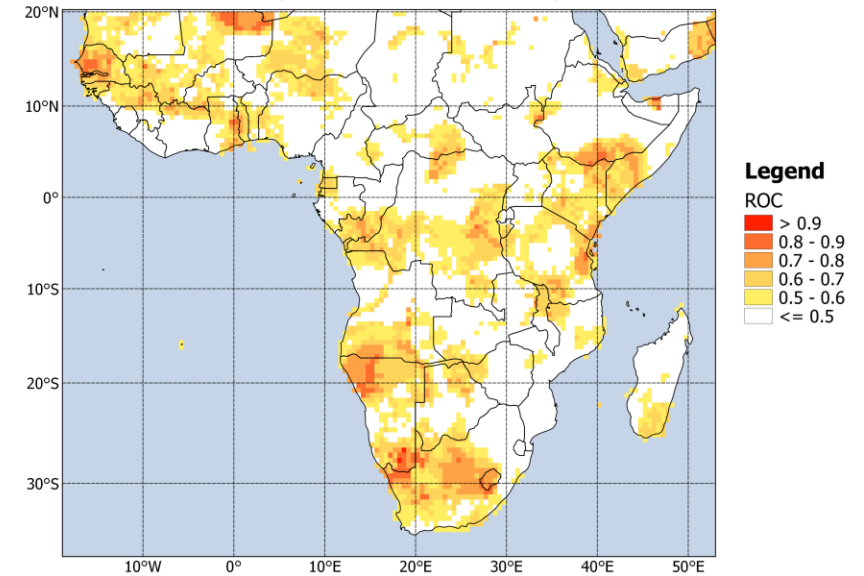
## ROC Area (Above-Normal): SON Rainfall



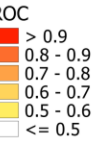
## Legend



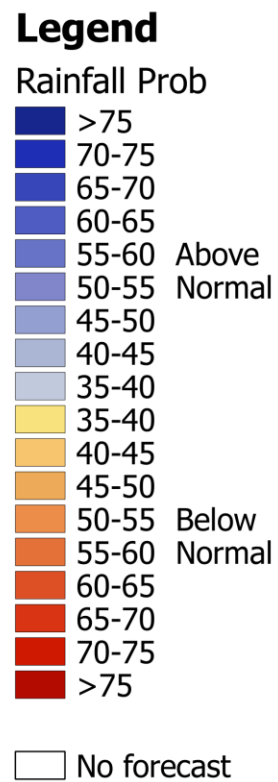
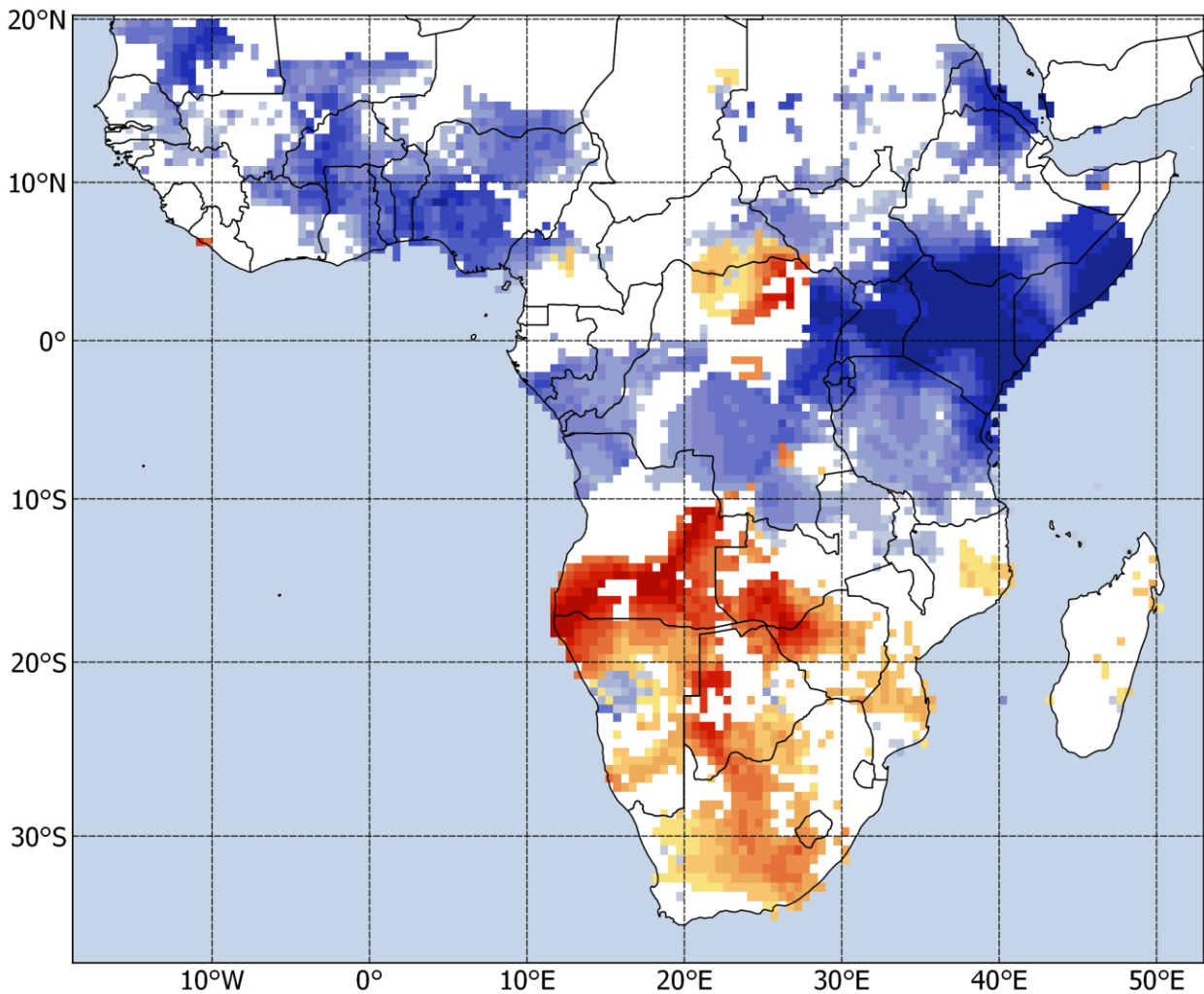
## ROC Area (Below-Normal): SON Rainfall



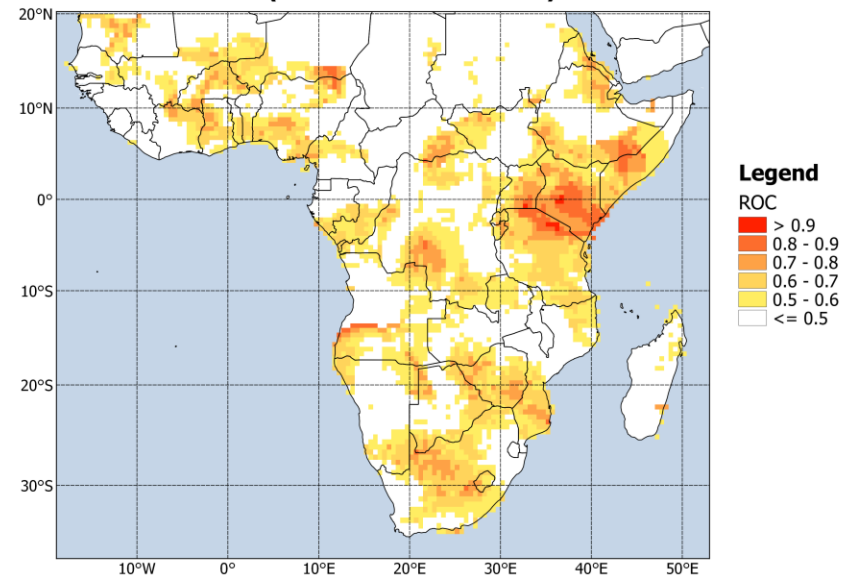
## Legend



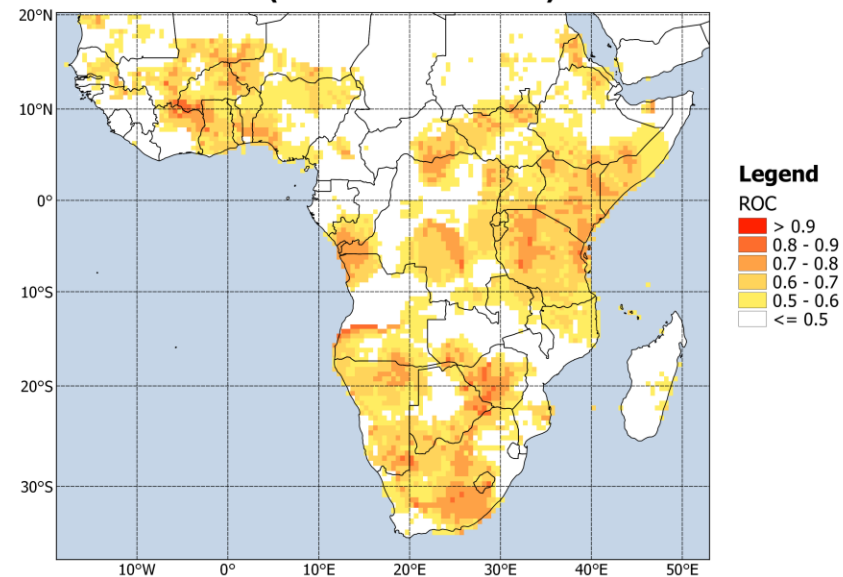
# OND 2023 Rainfall; ICs: Sep



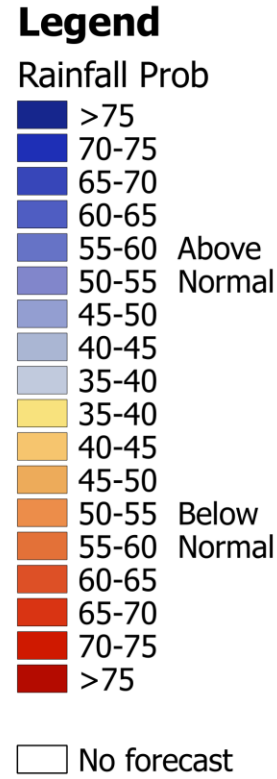
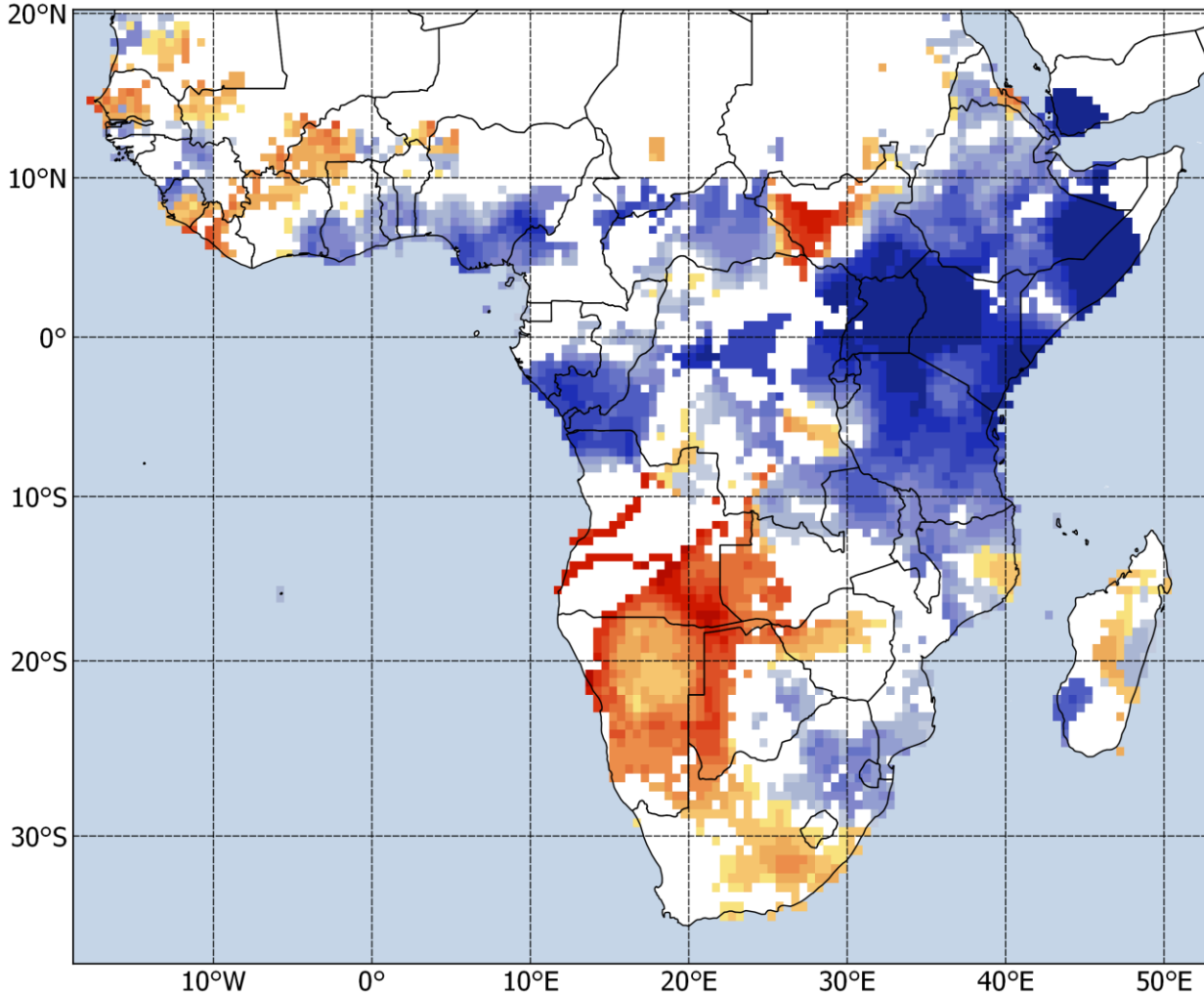
## ROC Area (Above-Normal): OND Rainfall



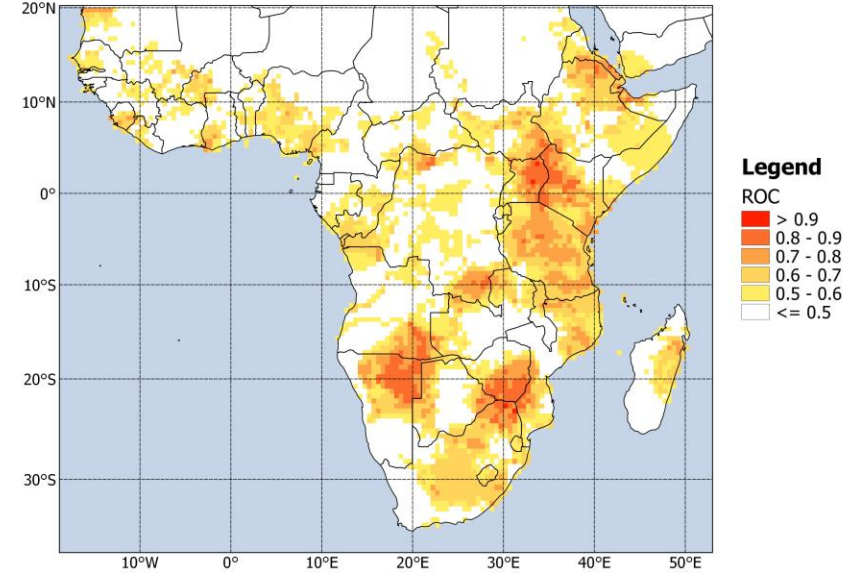
## ROC Area (Below-Normal): OND Rainfall



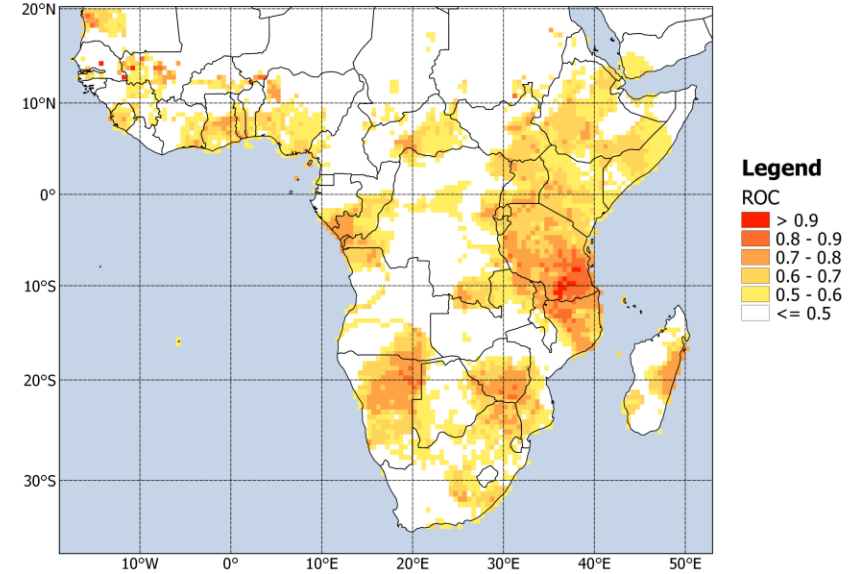
# NDJ 2023/24 Rainfall; ICs: Sep



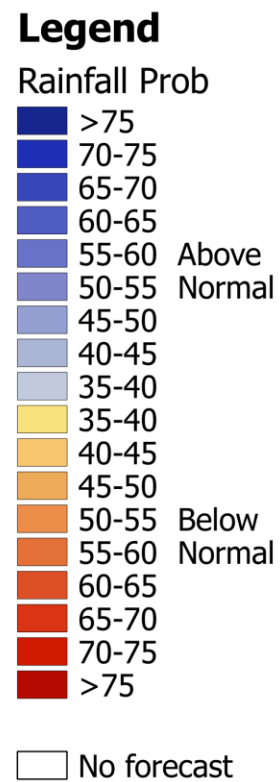
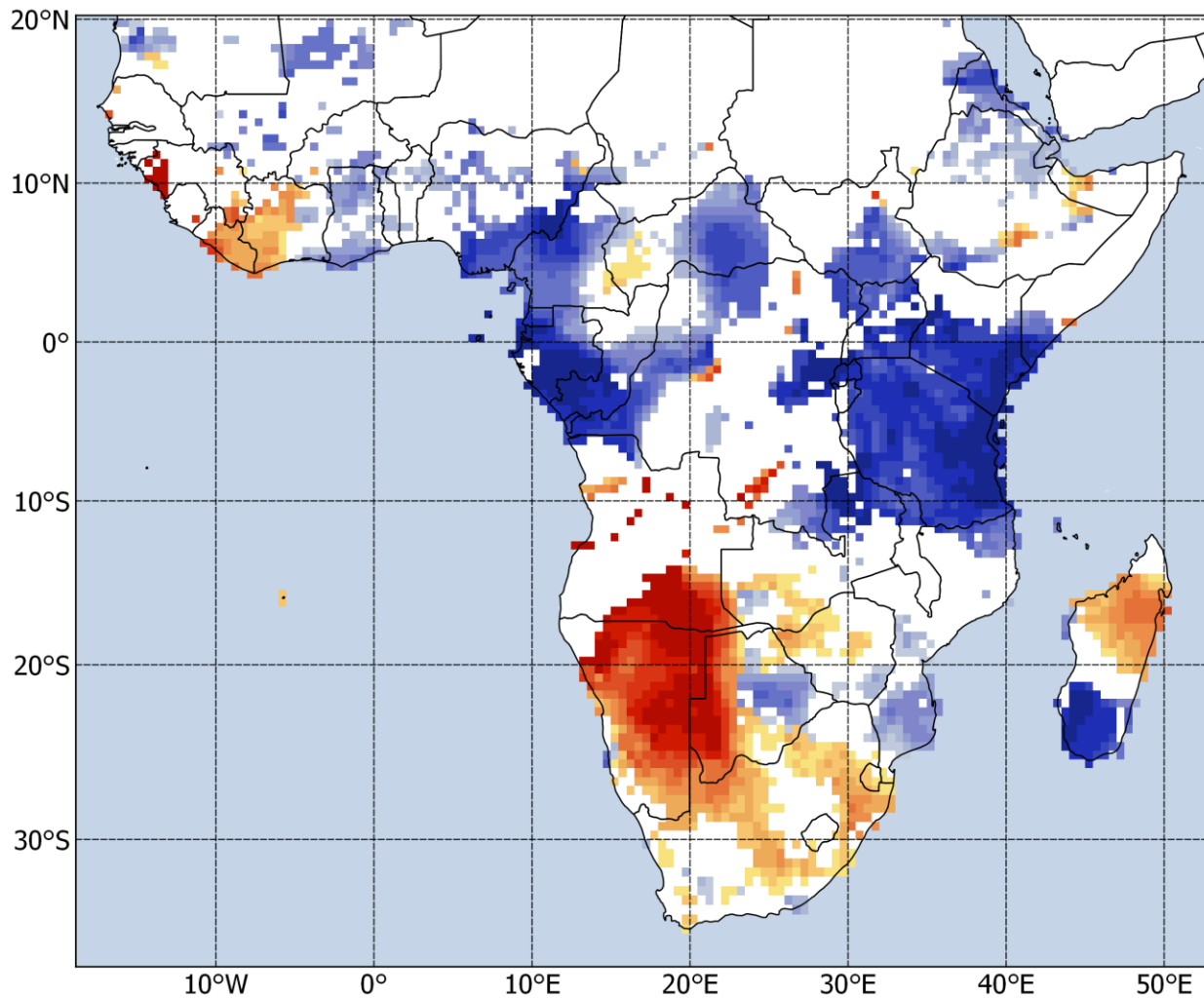
## ROC Area (Above-Normal): NDJ Rainfall



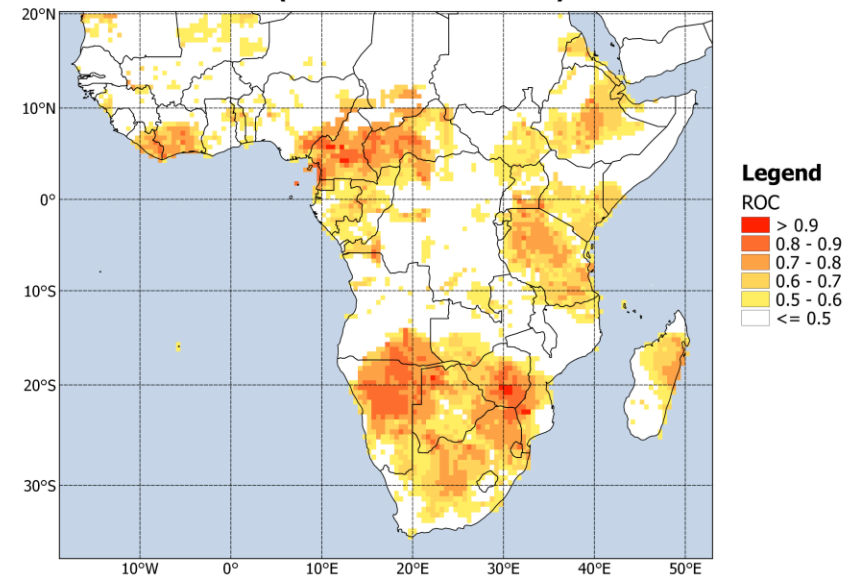
## ROC Area (Below-Normal): NDJ Rainfall



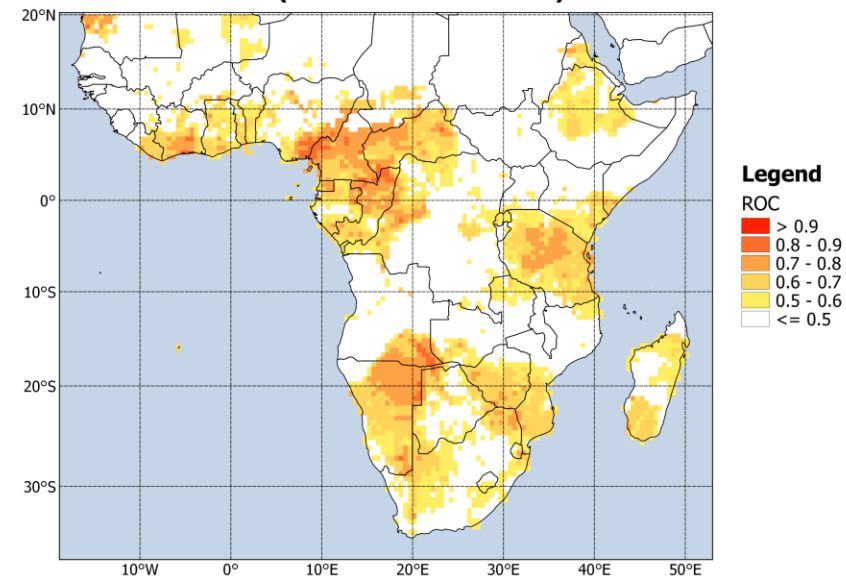
# DJF 2023/24 Rainfall; ICs: Sep



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



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



# Round-up: Rainfall south of 15°S

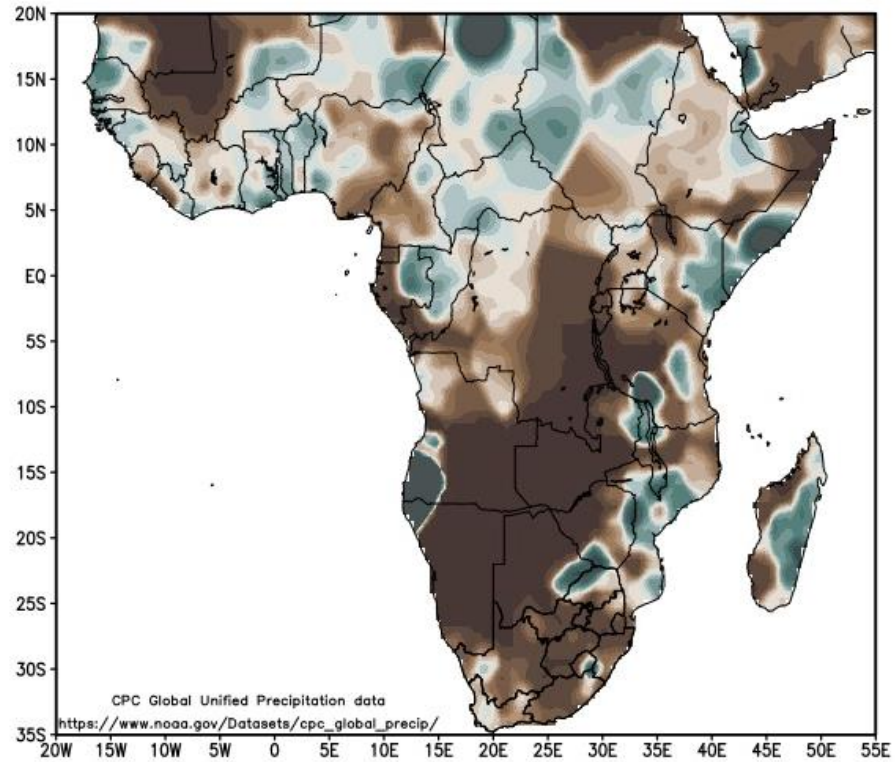
El Nino events (*as expected this summer*) are most often associated with below-normal rainfall totals over the summer rainfall region

- At the present time, the forecasts show that the central and western parts of the region are the most strongly linked with expected below-normal rainfall conditions
- Above-normal rainfall totals are predicted over the northeastern parts when entering the New Year, which is an outcome not usually associated with El Nino seasons
- Owing to the uncertainty of the current forecasts for the larger area, it is strongly advised to closely follow forecast updates

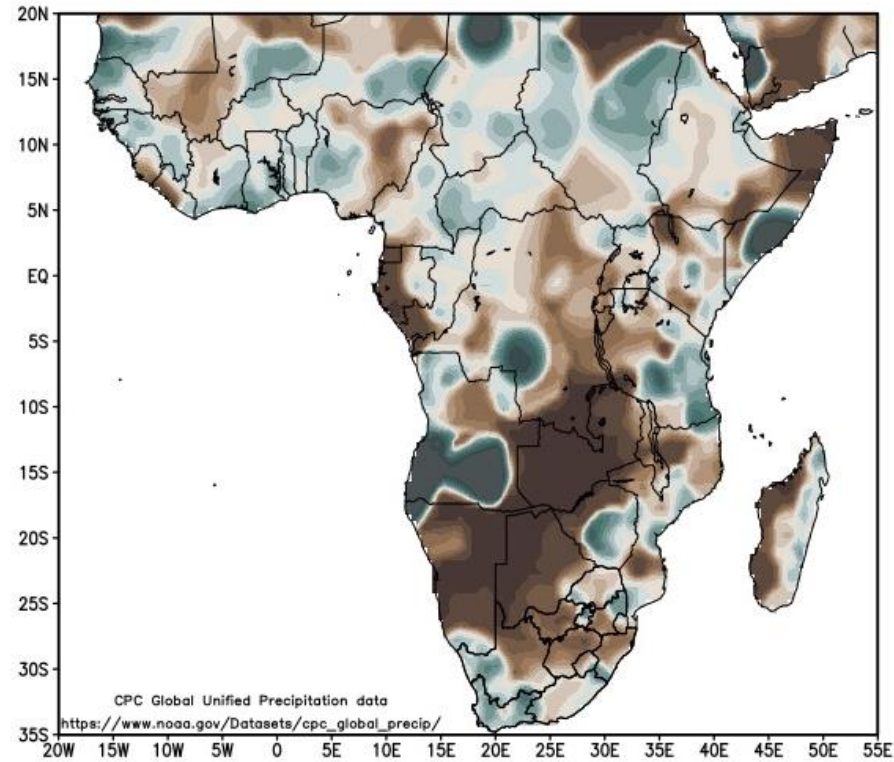


# Observed Rainfall

Rainfall (% of normal): August 2023  
August long-term mean: 1981–2010



Rainfall (% of normal): JJA 2023  
JJA long-term mean: 1981–2010

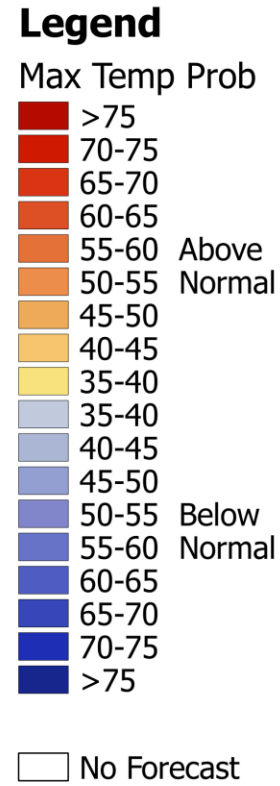
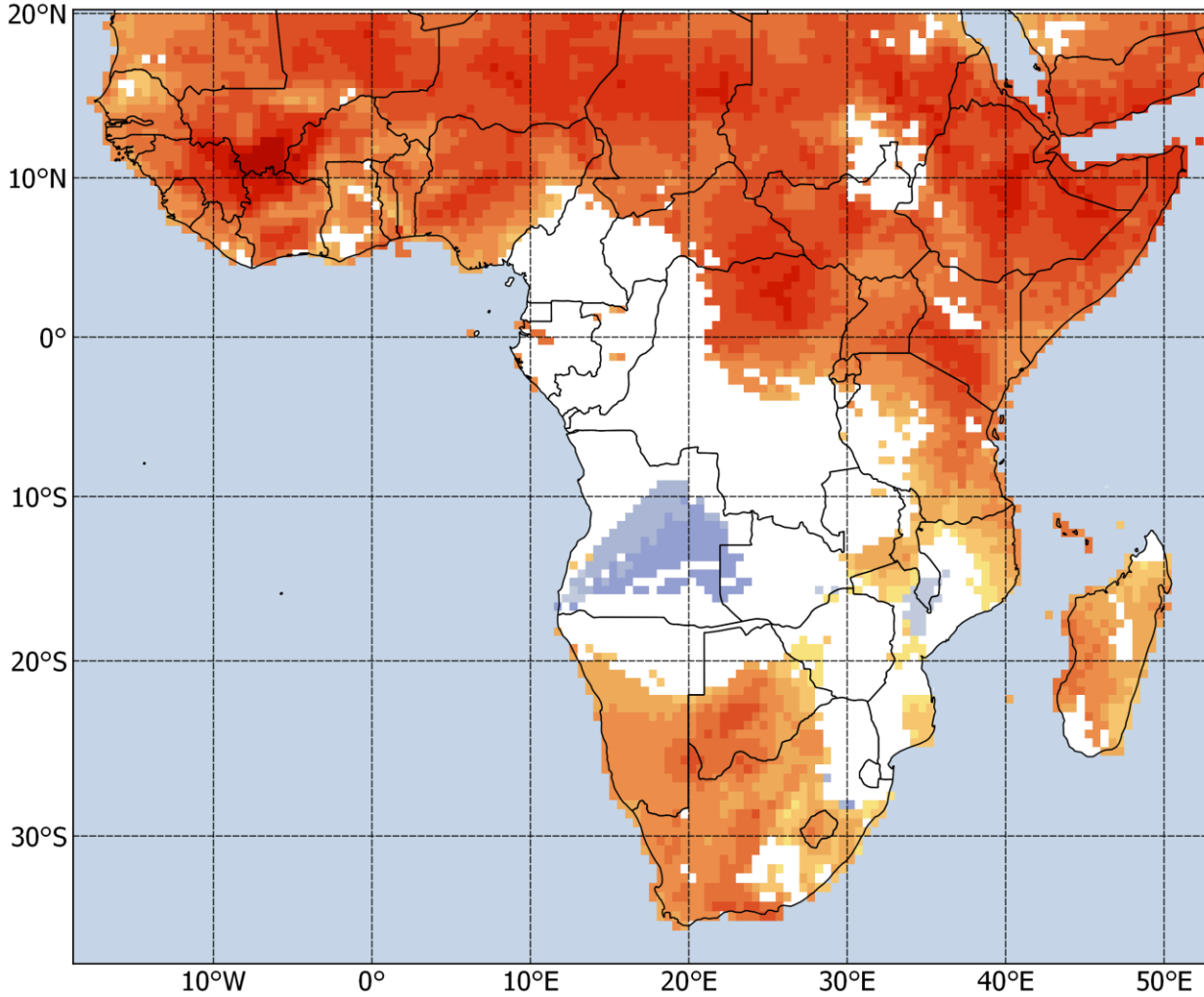


Recorded rainfall for August and the June-July-August season show below-normal rainfall over the brown areas and above-normal rainfall over the green areas

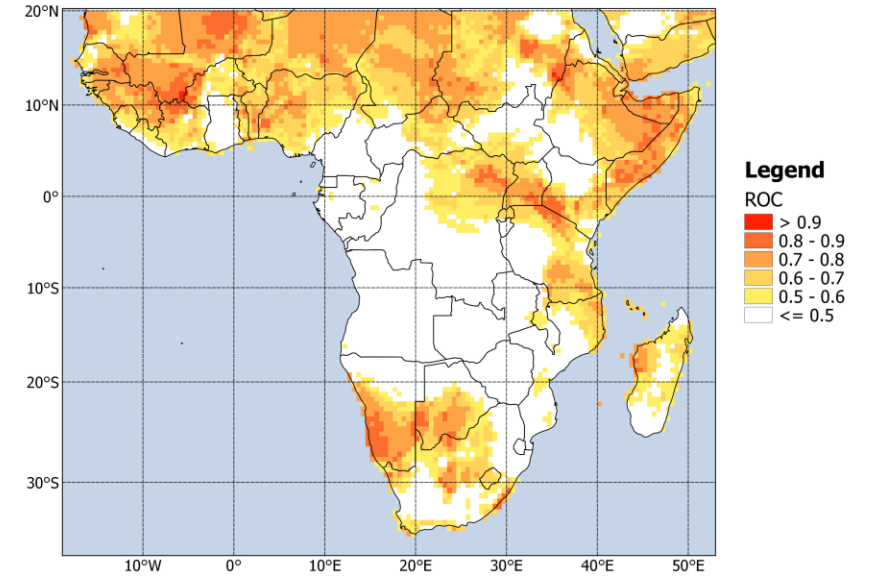
0 10 20 30 40 50 60 80 100 120 140 160 200 250 300 400 500 (%)

0 10 20 30 40 50 60 80 100 120 140 160 200 250 300 400 500 (%)

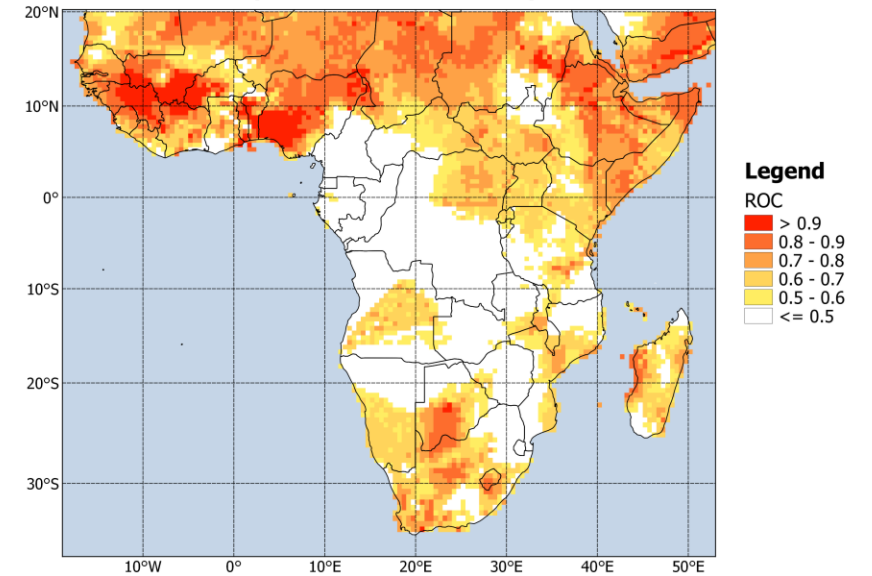
# SON 2023 Max Temp; ICs: Sep



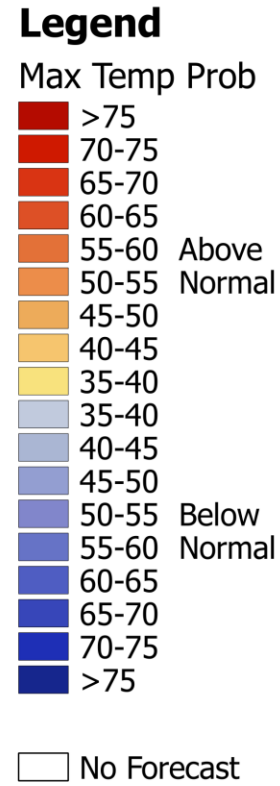
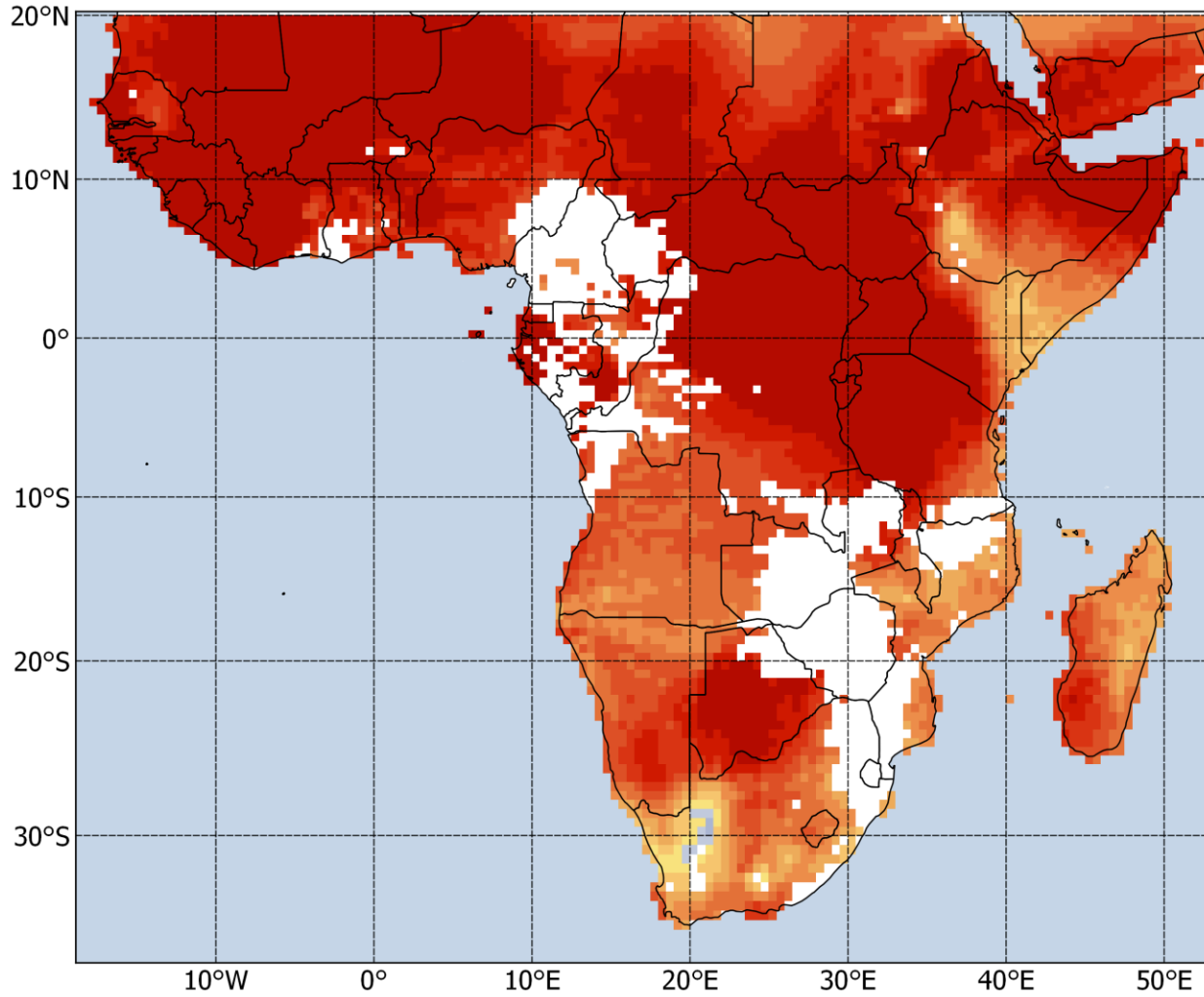
## ROC Area (Above-Normal): SON Max Temp



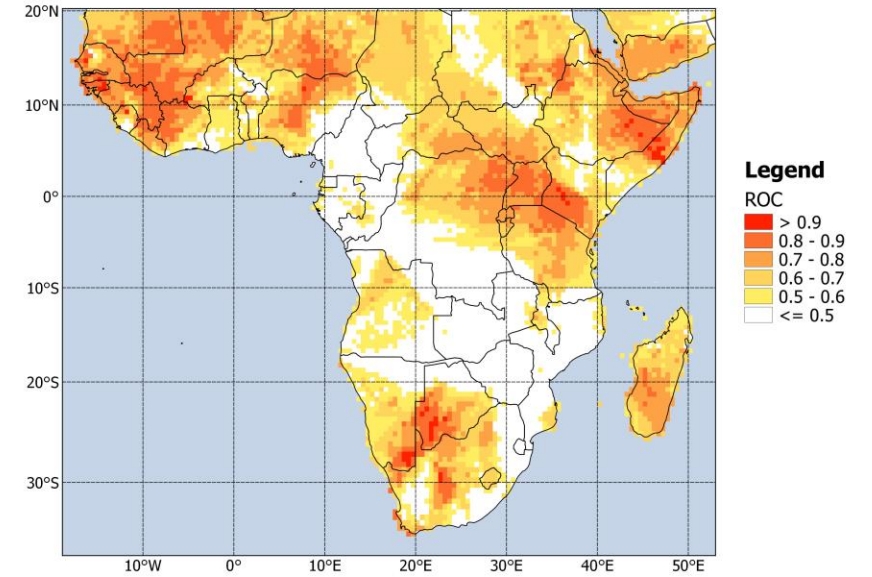
## ROC Area (Below-Normal): SON Max Temp



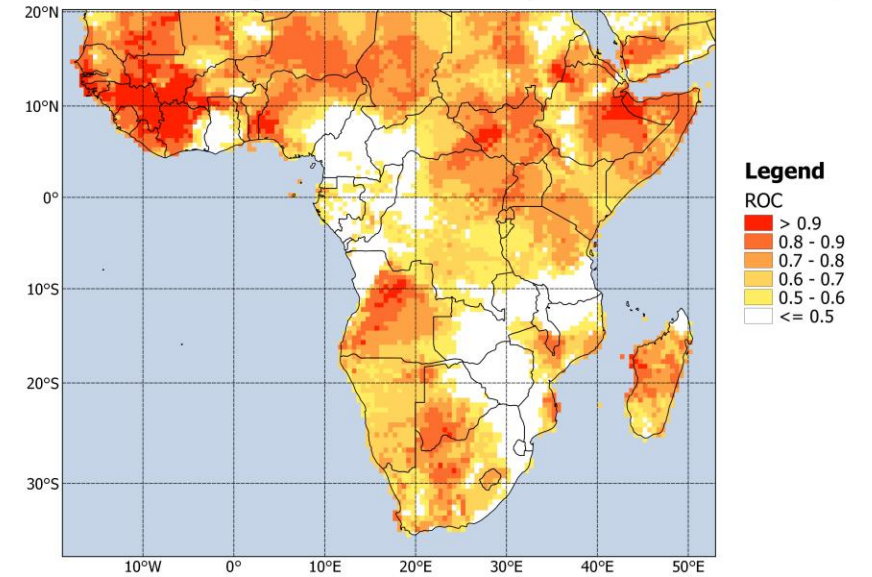
# OND 2023 Max Temp; ICs: Sep



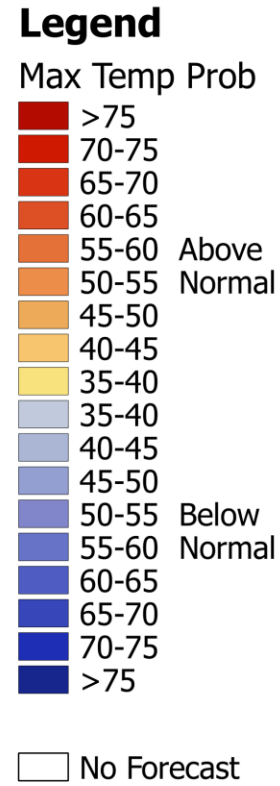
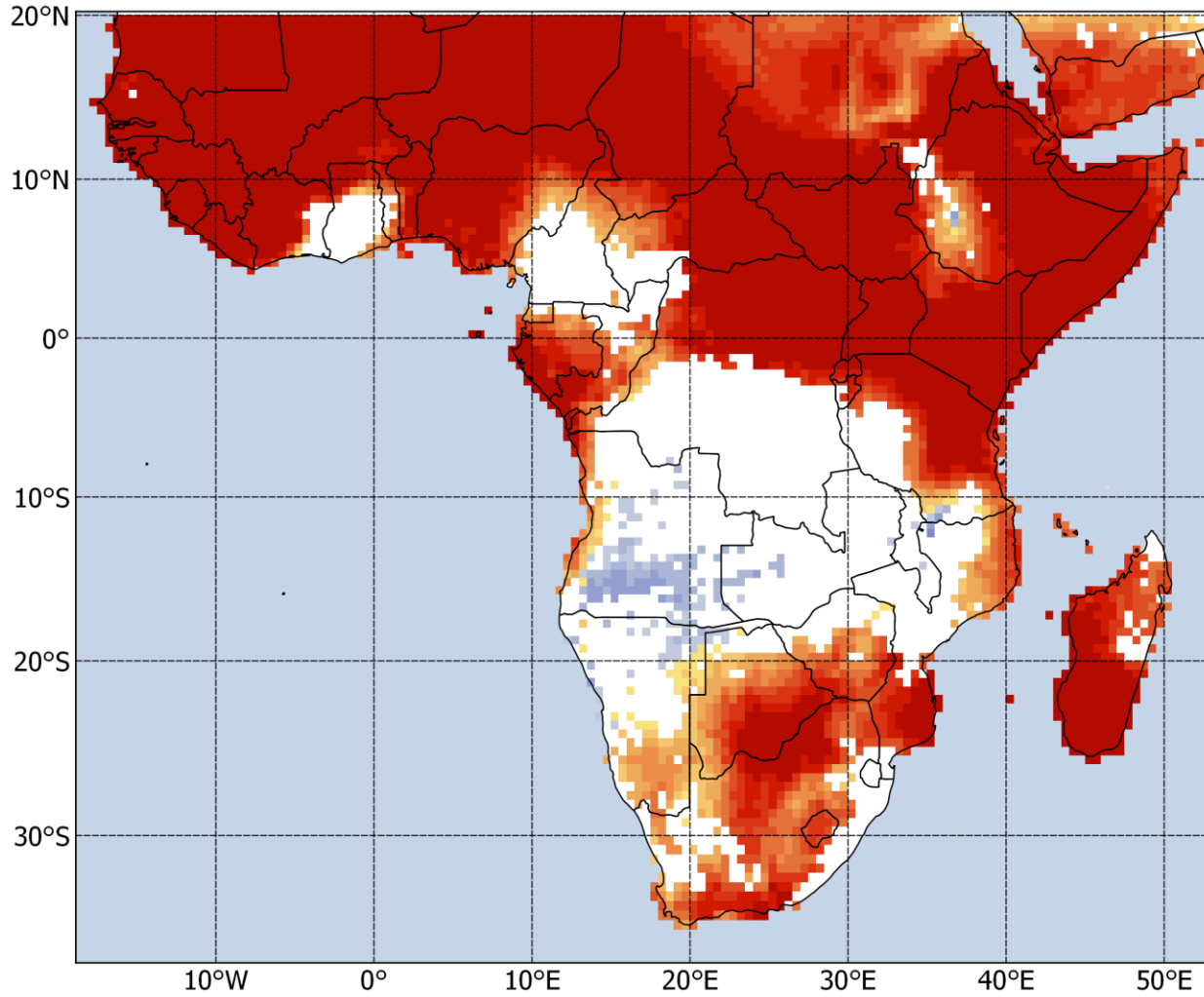
## ROC Area (Above-Normal): OND Max Temp



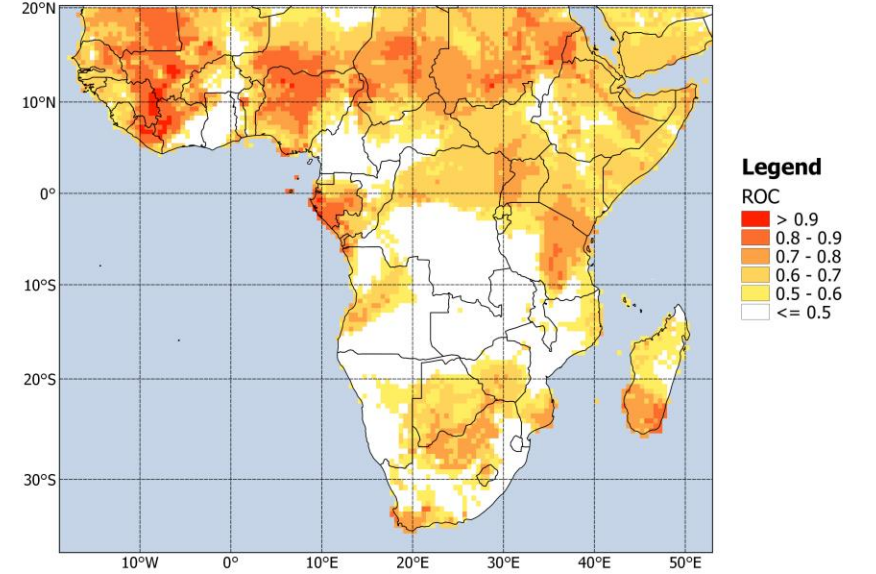
## ROC Area (Below-Normal): OND Max Temp



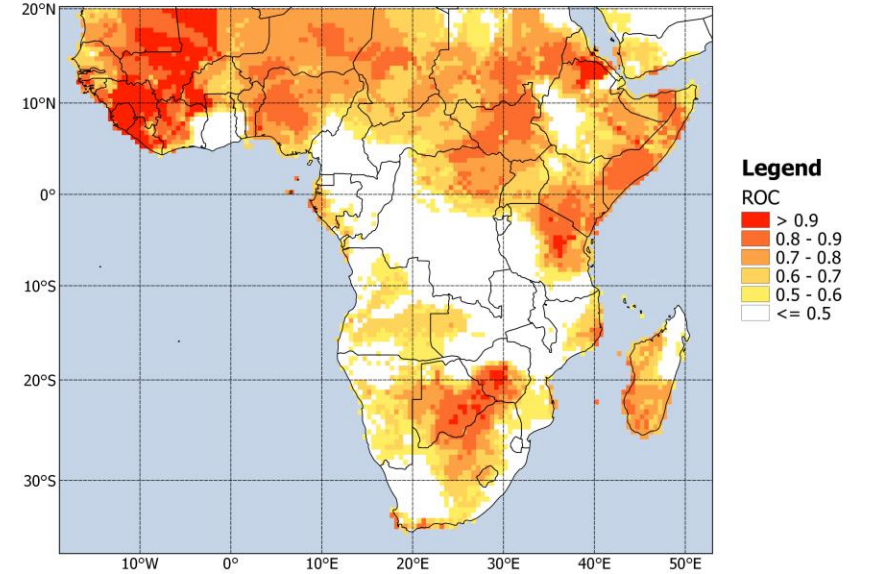
# NDJ 2023/24 Max Temp; ICs: Sep



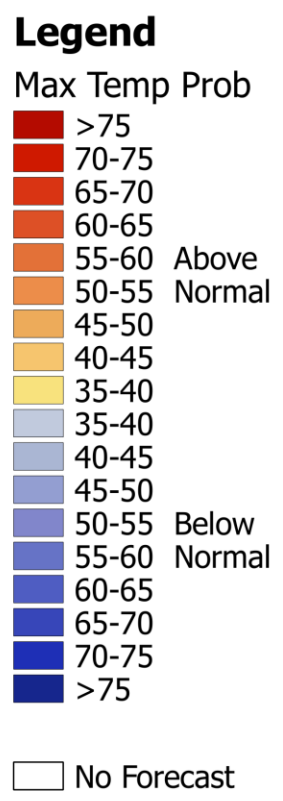
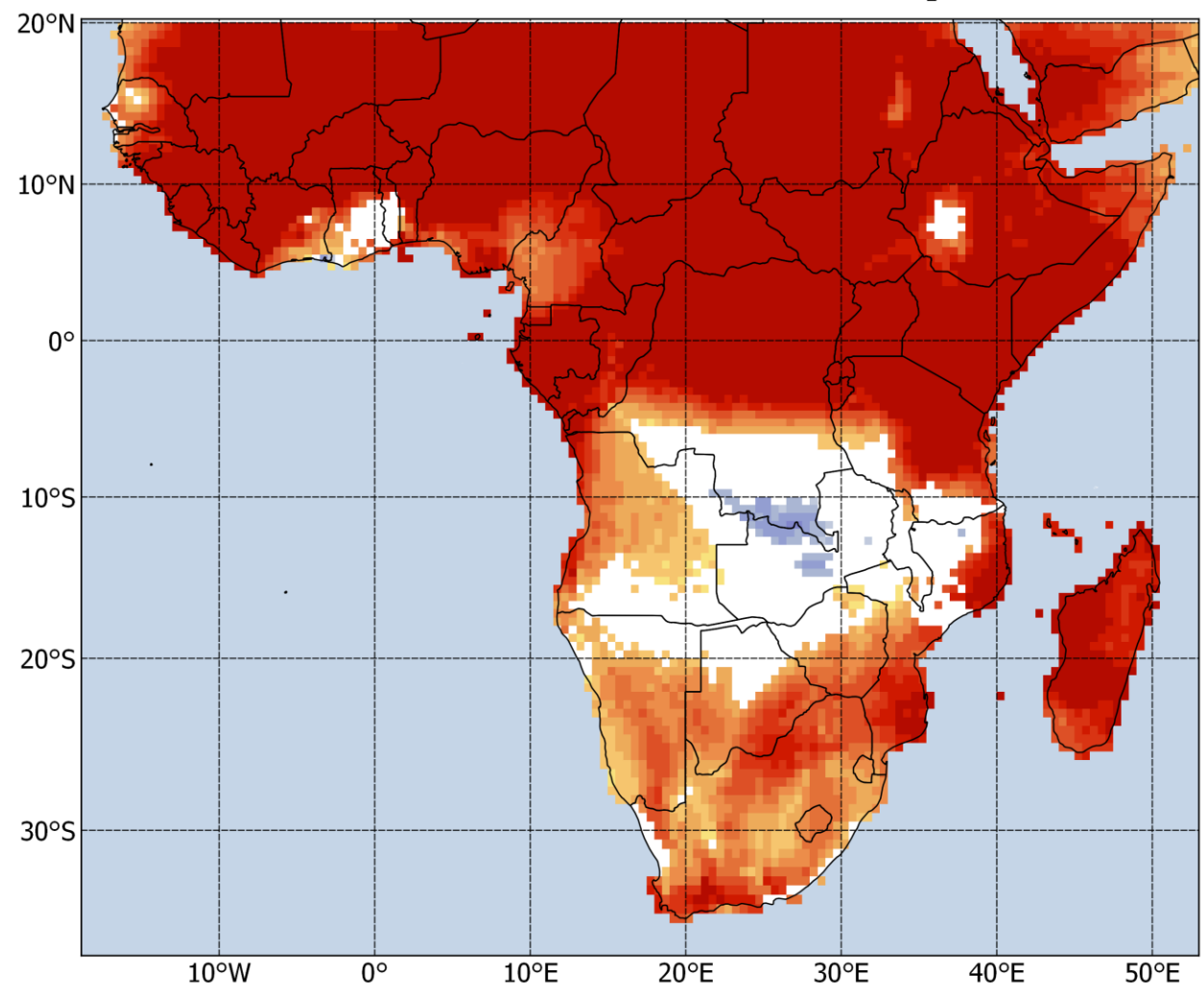
## ROC Area (Above-Normal): NDJ Max Temp



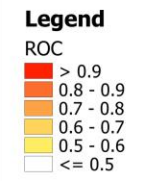
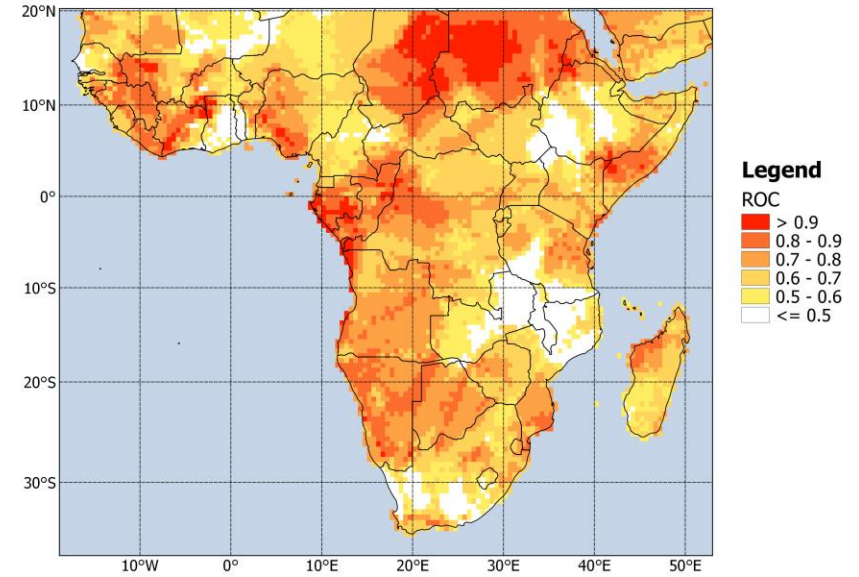
## ROC Area (Below-Normal): NDJ Max Temp



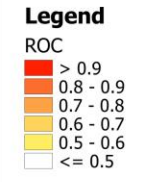
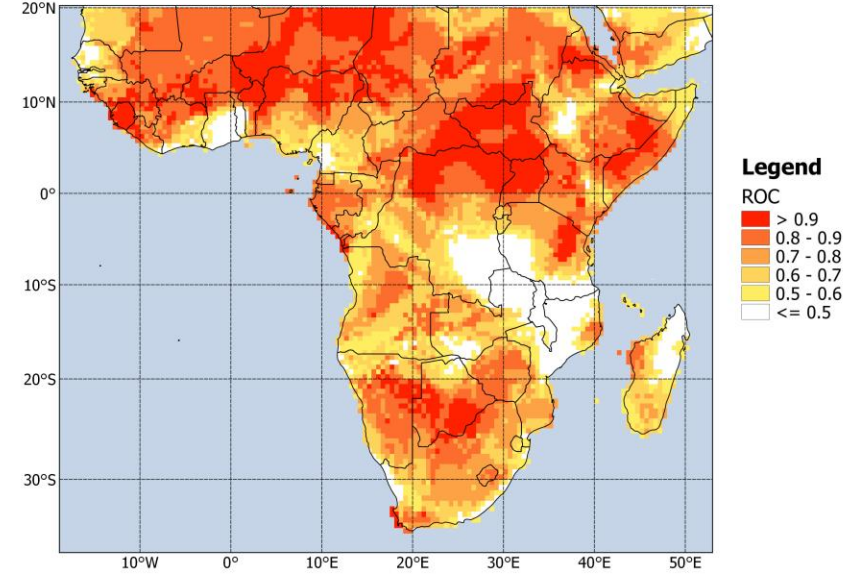
# DJF 2023/24 Max Temp; ICs: Sep



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



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



# Round-up: South of 15°S Max Temp

- Above-normal maximum temperatures over almost the entire region are highly likely during the forecast period

# Tailored Forecasts

Translating forecasts into indices on a range of relevant space and time scales that can inform regional decision-making. The following forecasts are shown to indicate the potential of seasonal forecasting for real-life applications

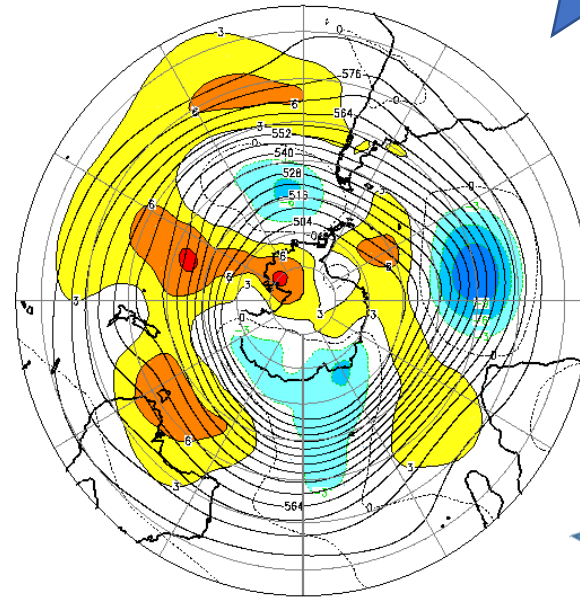
1. Bapsfontein end-of-season-yield three-category probabilistic forecast for 2024
2. Probability of exceedance Dec-Jan-Feb 2023/24 inflow forecast for Lake Kariba, Zambia/Zimbabwe

# The prediction scheme

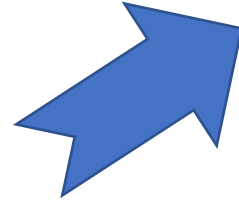
1. Phenomena to be predicted should contain a climate signal (e.g. ENSO) in the data;
2. Observed and model time series must be over sufficiently long enough periods so that robust statistical relationship can be developed;
3. and some form of quality control of the observed data had taken place.



Climate models of the NMME



Seasonal forecast of the climate models, statistically linked to observed values



Riverflow



Malaria

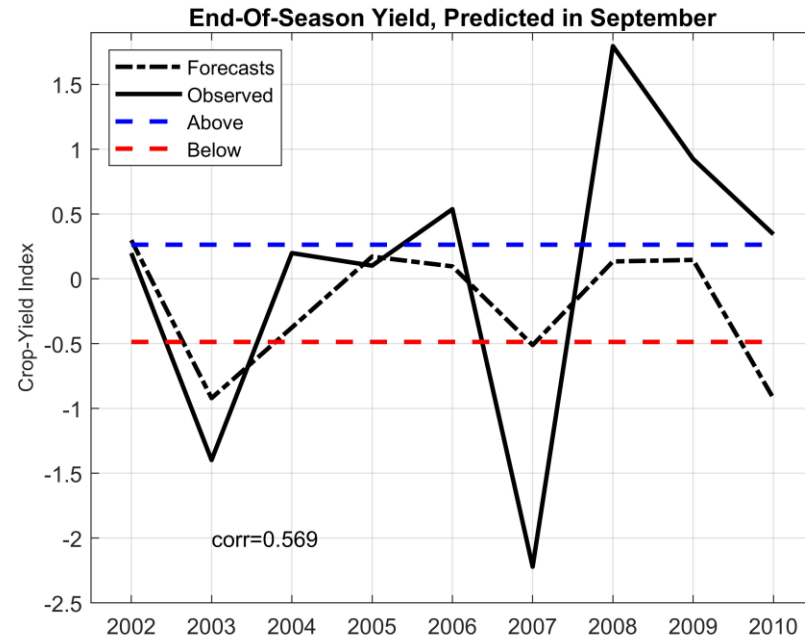


Dry-land crop yield



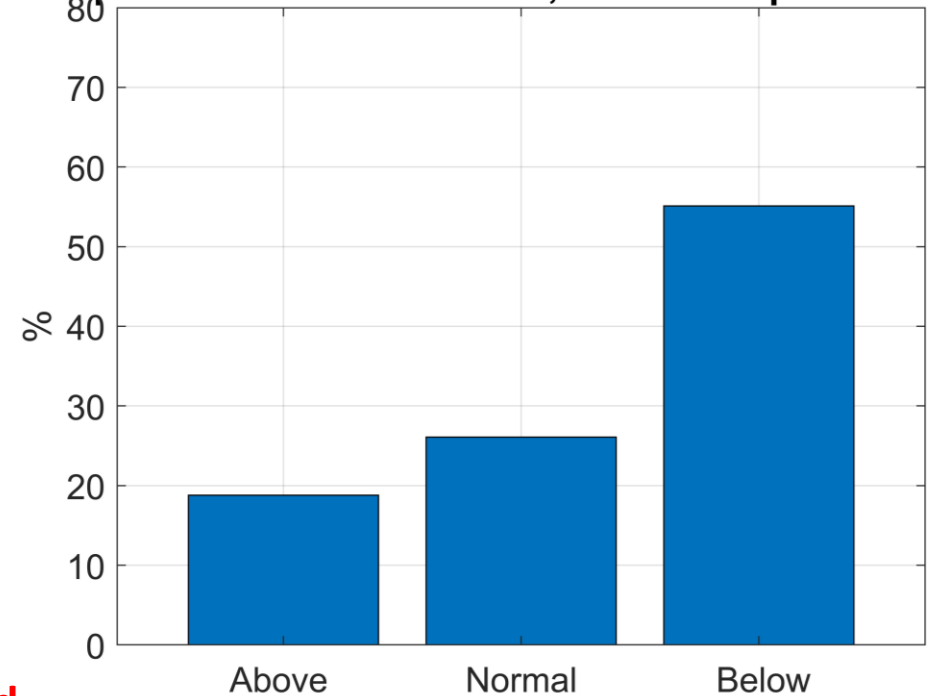
# Dry-land crop-yield data and forecasts for a farm near Bapsfontein, South Africa

Landman et al. (2019)



**Re-forecasts for end-of-season crop yield**

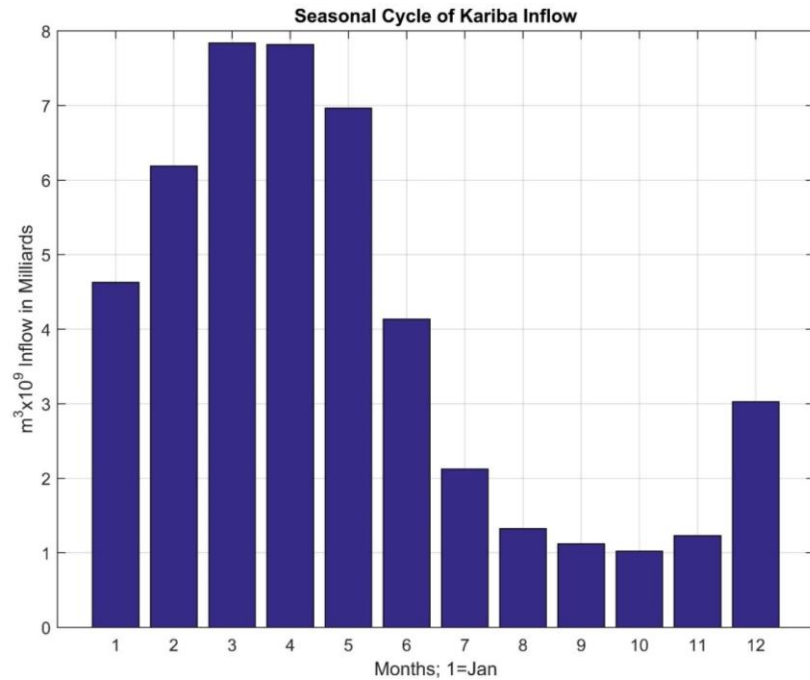
**Crop-Yield Forecast for 2024, Made in September 2023**



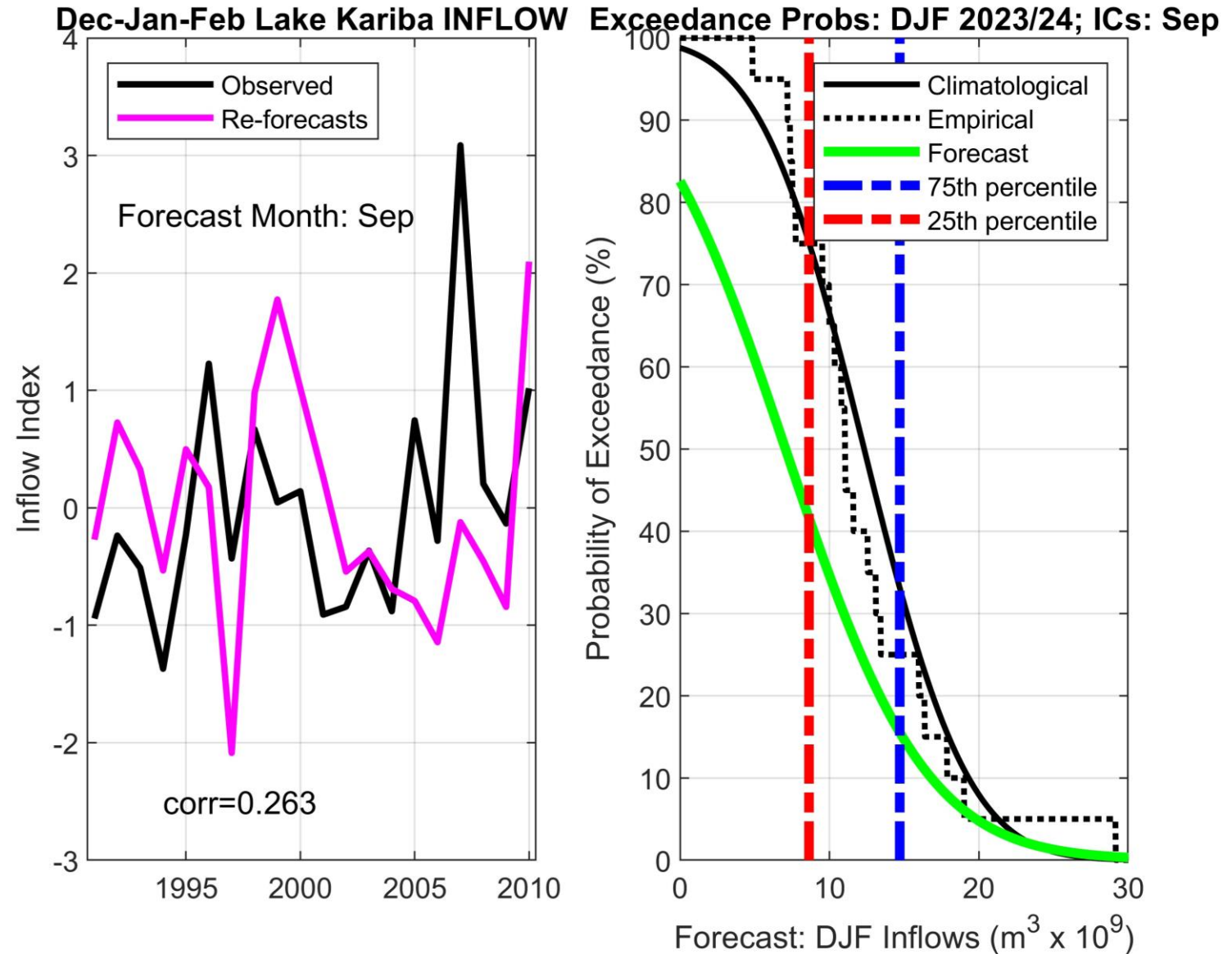
% is the probability of the respective categories to occur

# Inflow forecast for Lake Kariba: onset season of DJF

Muchuru et al. (2016)



For the forecast on the far right: The black curve represents the climatological probability distribution of DJF inflows, and the green curve represents the predicted probability distribution for the coming DJF season. The vertical dashed lines represent the category thresholds. The easiest way to interpret the green forecast curve would be to consider a curve above (below) the thick black curve to be probabilistic forecasts for anomalously high (low) DJF inflows.



# Round-up: Tailored products

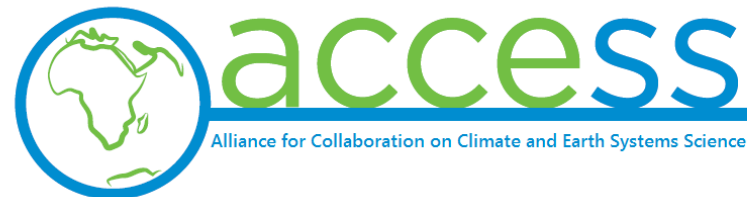
- The yield and inflow forecasts shown are in agreement with the expected outcome during El Nino seasons – low yield and low flows

# References

- Barnston, A.G. and Tippett, M.K., 2017: Do statistical pattern corrections improve seasonal climate predictions in the North American Multimodel Ensemble models? *Journal of Climate*, 30: 8335-8355. doi: 10.1175/JCLI-D-17-0054.1
- Harris, I., Jones, P. D., Osborn, T. J., and Lister, D. H., 2014: Updated high-resolution grids of monthly climatic observations - the CRU TS3.10 Dataset. *International Journal of Climatology*, 34: 623-642. doi: 10.1002/joc.3711
- Kirtman, B. P. and Co-authors 2014: The North American Multimodel Ensemble: Phase-1 seasonal-to-interannual prediction; Phase-2 toward developing intraseasonal prediction. *Bulletin of the American Meteorological Society*. 95, 585–601. doi: <http://dx.doi.org/10.1175/BAMS-D-12-00050.1>
- Landman, W.A., and Beraki, A., 2012: Multi-model forecast skill for midsummer rainfall over southern Africa. *International Journal of Climatology*, 32: 303-314. doi: 10.1002/joc.2273.
- Landman, W.A., Archer, E. and Tadross, M., 2016: Decision-relevant information on seasonal time scales – the case of a farm in northern Namibia. *Conference Proceedings of the 32nd Annual Conference of the South African Society for Atmospheric Science*, Cape Town, 31 October to 1 November 2016, pp 69-72. ISBN 978-0-620-72974-1.
- Landman, W.A., Archer, E. and Tadross, M. (2019): How costly are poor seasonal forecasts? Peer reviewed abstracts, 35th Annual conference of the South African Society for Atmospheric Science, Vanderbijlpark, 8 to 9 October 2019, pp 60-63. ISBN 978-0-6398442-0-6.
- Landman, W.A., Archer, E.R.M and Tadross, M.A (2020a). Citizen science for the prediction of climate extremes in South Africa and Namibia. *Frontiers in Climate*, 2:5, doi: 10.3389/fclim.2020.00005
- Landman, W.A., DeWitt, D., and Lee, D.-E., 2011: The high-resolution global SST forecast set of the CSIR. *Conference Proceedings of the 27th Annual Conference of South African Society for Atmospheric Sciences*, 22-23 September 2011, Hartbeespoort, North-West Province, South Africa. ISBN 978-0-620-50849-0
- Landman, W.A., Sweijd, N., Masedi, N. Minakawa, N. (2020b). The development and prudent application of climate-based forecasts of seasonal malaria in the Limpopo province in South Africa. *Environmental Development*, 35, 100522, doi: 10.1016/j.envdev.2020.100522.
- Landman, W.A., Tadross, M., Archer, E., Johnston, P. (2023). Probabilistic vs deterministic forecasts - interpreting skill statistics for the benefit of users. *Water SA*, 49(3), 192–198. <https://doi.org/10.17159/wsa/2023.v49.i3.4058>
- Landman, W.A., DeWitt, D. Lee, D.-E., Beraki, A. and Lötter, D., 2012: Seasonal rainfall prediction skill over South Africa: 1- vs. 2-tiered forecasting systems. *Weather and Forecasting*, 27: 489-501. DOI: 10.1175/WAF-D-11-00078.1
- Muchuru, S., Landman, W.A. and DeWitt, D., 2016: Prediction of inflows into Lake Kariba using a combination of physical and empirical models. *International Journal of Climatology*, 36: 2570–2581, DOI: 10.1002/joc.4513.
- Troccoli, A., Harrison, M., Anderson, D.L.T. and Mason, S.J., 2008: *Seasonal Climate: Forecasting and Managing Risk*. NATO Science Series on Earth and Environmental Sciences, Vol. 82, Springer, 467 pp.

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The forecast is produced by Prof Willem Landman of the University of Pretoria, South Africa, and issued on or around the 15th of each month. Please feel free to contact me at [WALandman1981@gmail.com](mailto:WALandman1981@gmail.com)

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Disclaimer: The author has compiled this forecast guidance as a service to users for application in appropriate sectors, but cannot be held responsible for inaccuracies contained therein

# 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



**Shepherd Muchuru, PhD (Meteorology):**

Statistical modelling to relate large-scale features to seasonal inflows into Lake Kariba in southern Africa. Two predictions systems: 1) using antecedent seasonal rainfall totals over the upper Zambezi catchment as predictor in a baseline model, and 2) using predicted low-level atmospheric circulation of a coupled ocean–atmosphere general circulation model as predictor.



**Pearl Gosiambe, BSc (Honours)(Meteorology):**

Development of hydro-climate predictions models for dam levels and downstream flows of the Vaal Dam. Predictors considered include historical rainfall over the catchment, SST and output from global climate models.





**Idani Mandiwana, BSc (Honours) (Meteorology):**

Seasonal rainfall forecast verification of real-time forecasts produced by SFW over the 5-year period from 2018 to 2022. Area is SADC south of 17° South.