

Seasonal forecasts

presented by:

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Seasonal Forecast Worx



<https://tinyurl.com/ForecastProf>

UNIVERSITEIT VAN PRETORIA
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Seasonal Climate Forecasts

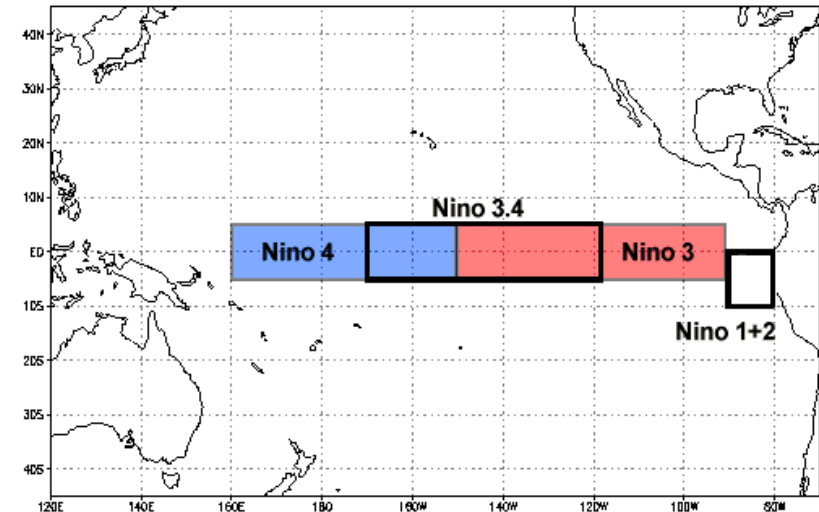
Latest Update: 22 January 2024

- 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>

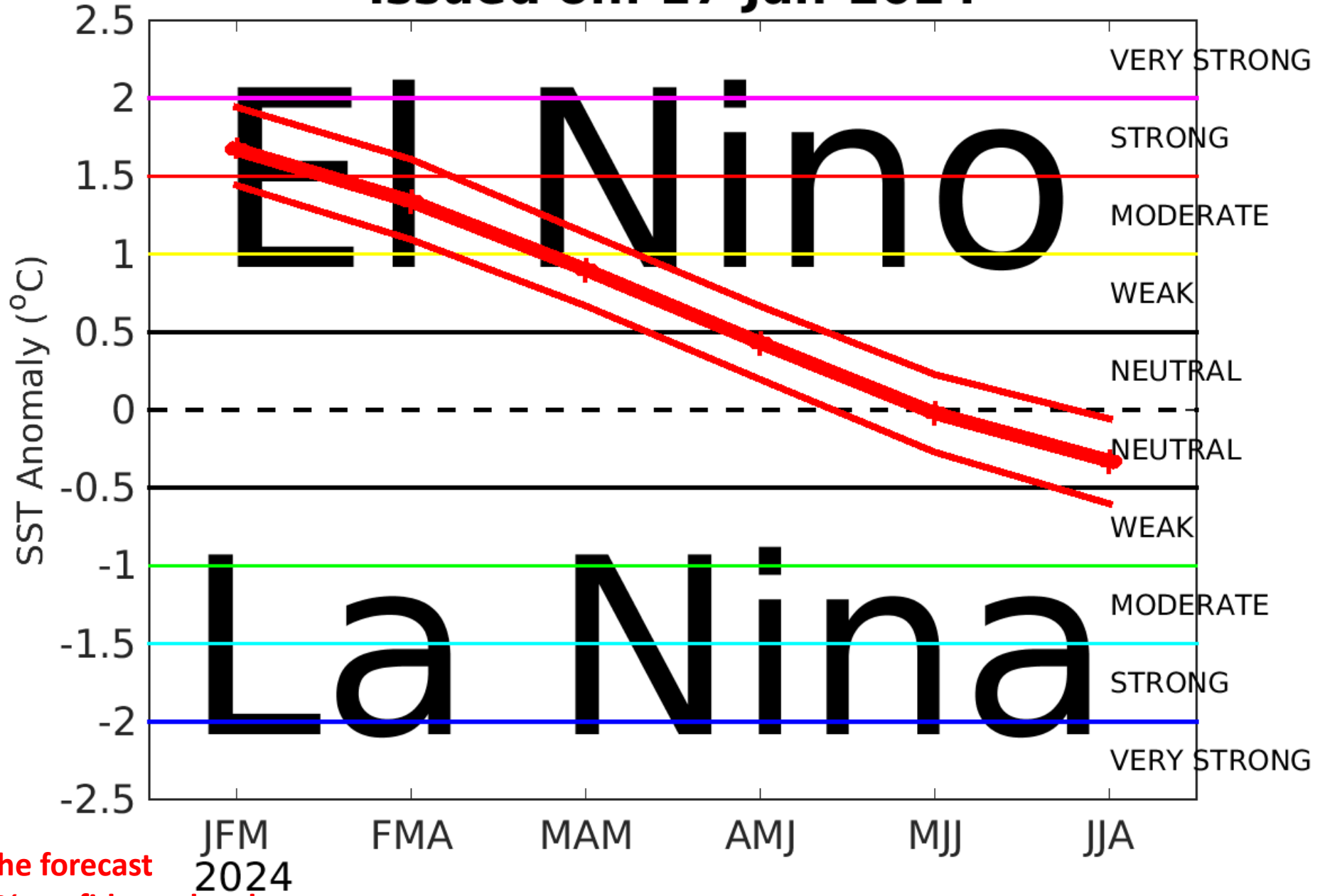
ENSO and Global SST Forecasts

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



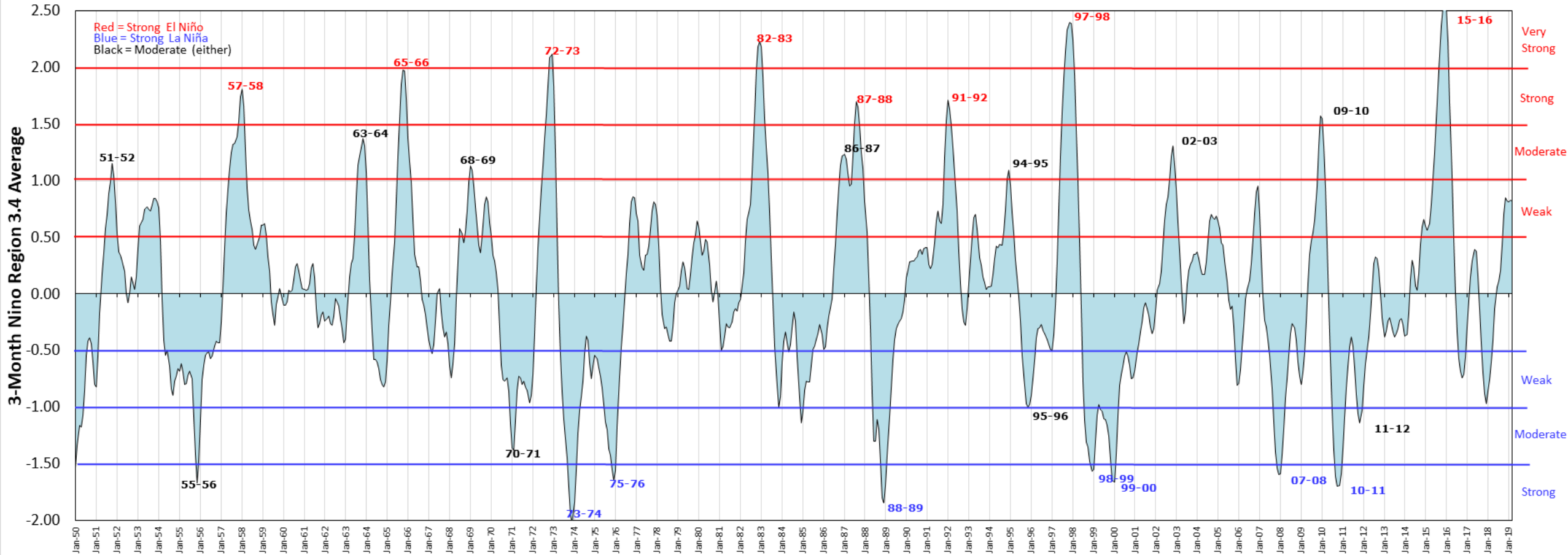
CSiriMM Nino3.4 SST Forecast Issued on: 17-Jan-2024



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

Oceanic Niño Index (ONI)

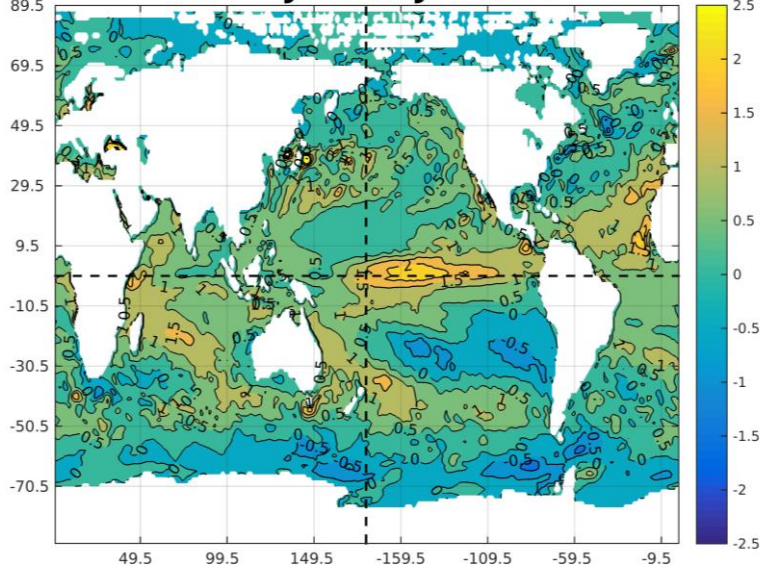
http://www.cpc.ncep.noaa.gov/products/analysis_monitoring/ensostuff/ensoyears.shtml



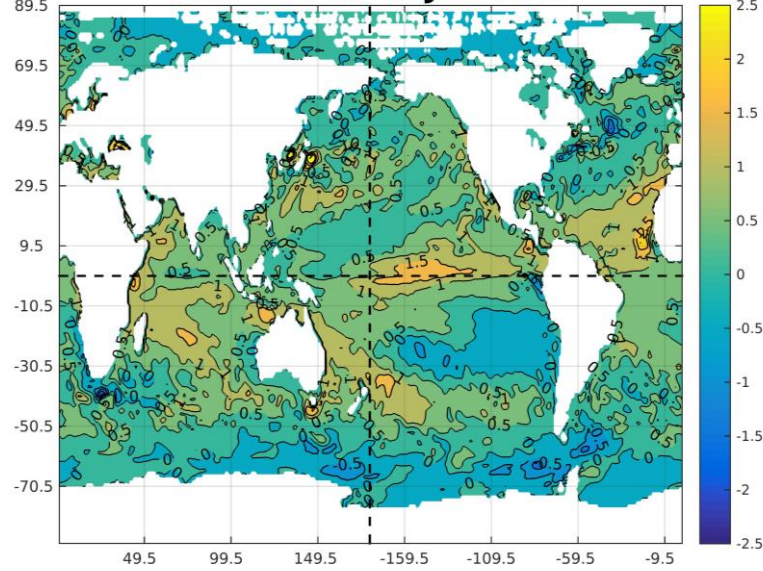
IC: the month in which the forecast was made

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

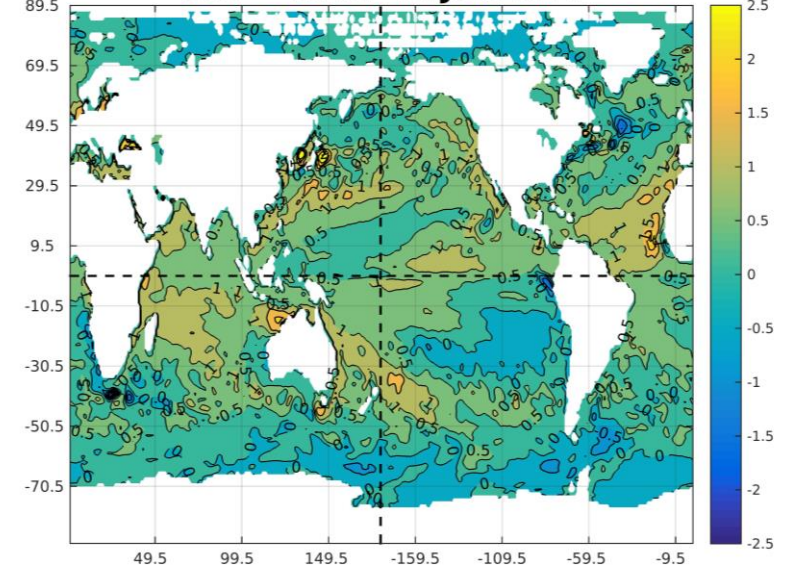
SST JFM IC:Jan2024



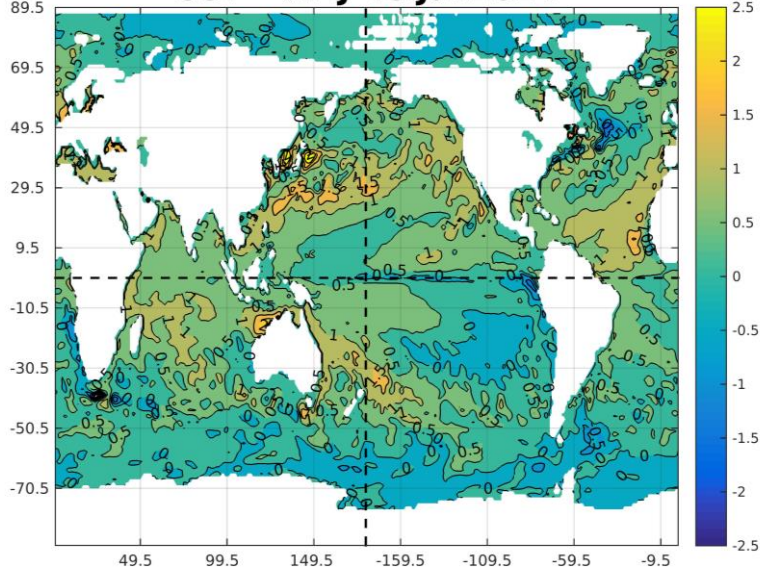
SST FMA IC:Jan2024



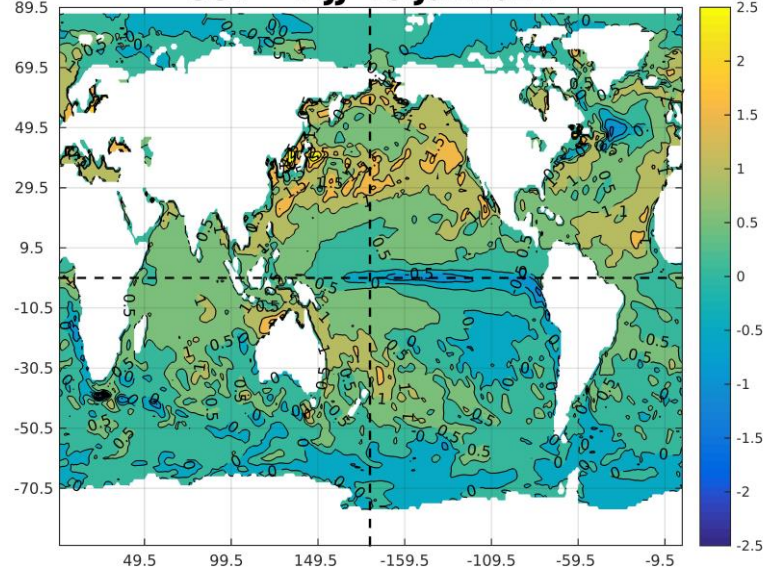
SST MAM IC:Jan2024



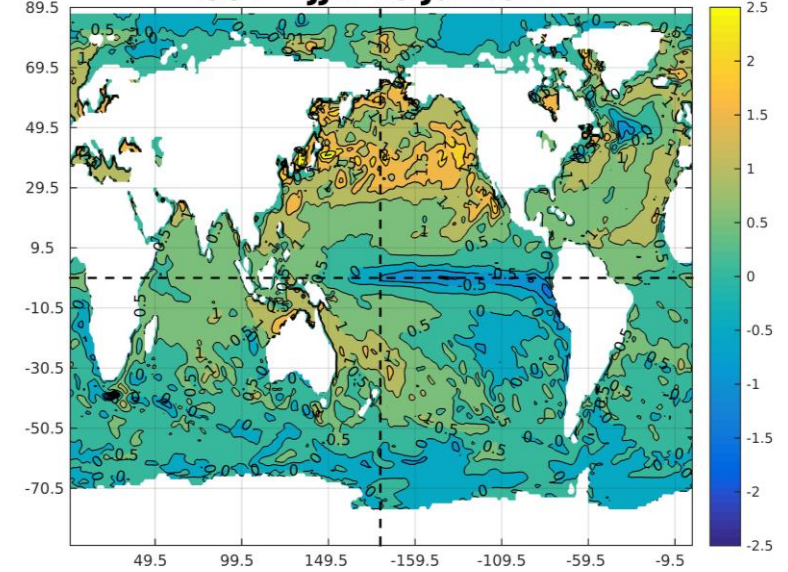
SST AMJ IC:Jan2024



SST MJJ IC:Jan2024

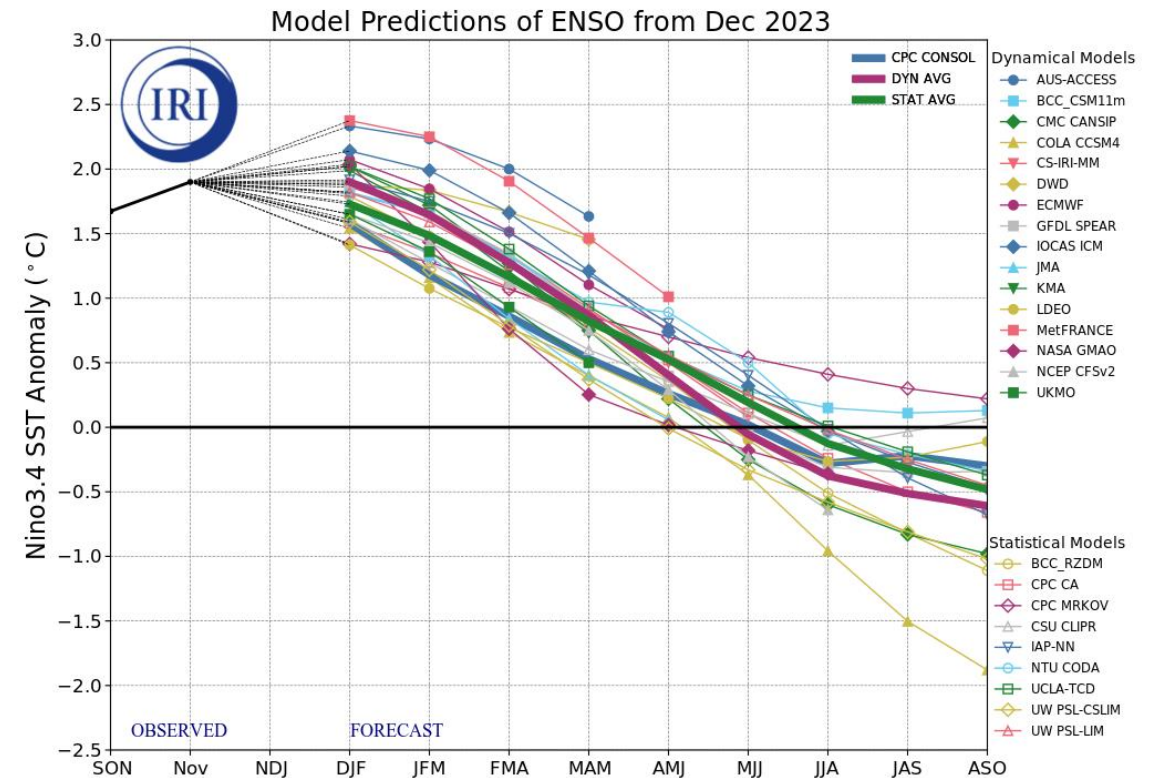
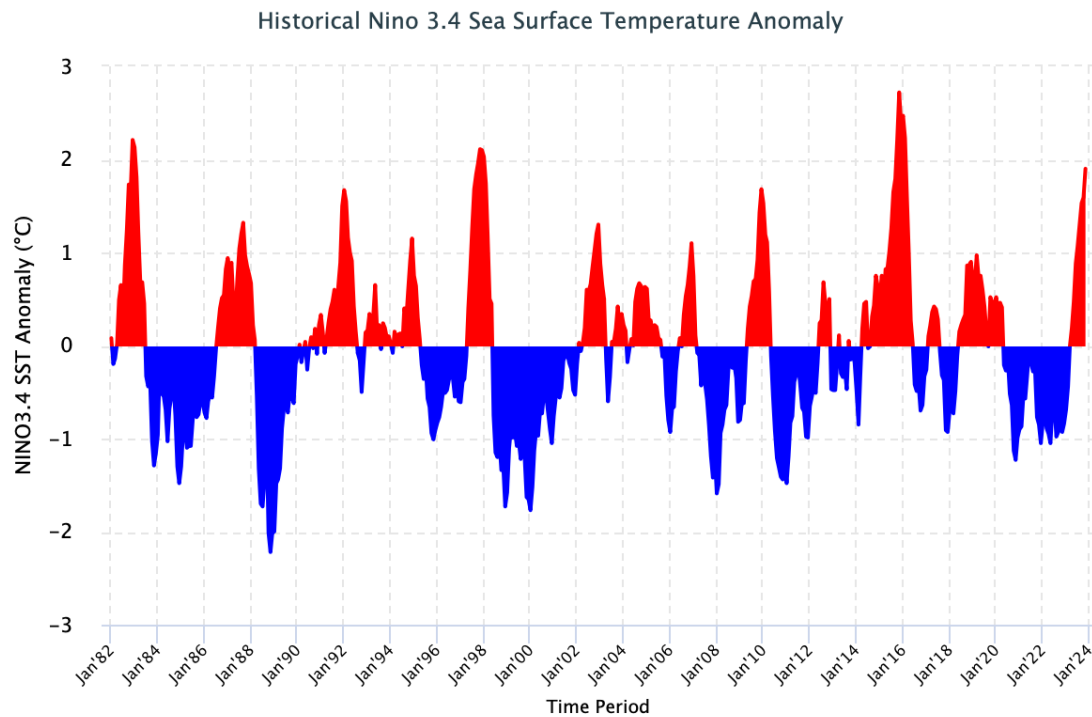


SST JJA IC:Jan2024



Round-up: ENSO

- The UP model (on previous pages) is predicting a medium to strong ($>1^{\circ}\text{C}$) El Niño event to continue during the Austral summer period, which is in agreement with some of forecasts produced internationally (right panel below)
- Heading towards ENSO-neutral by Winter



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

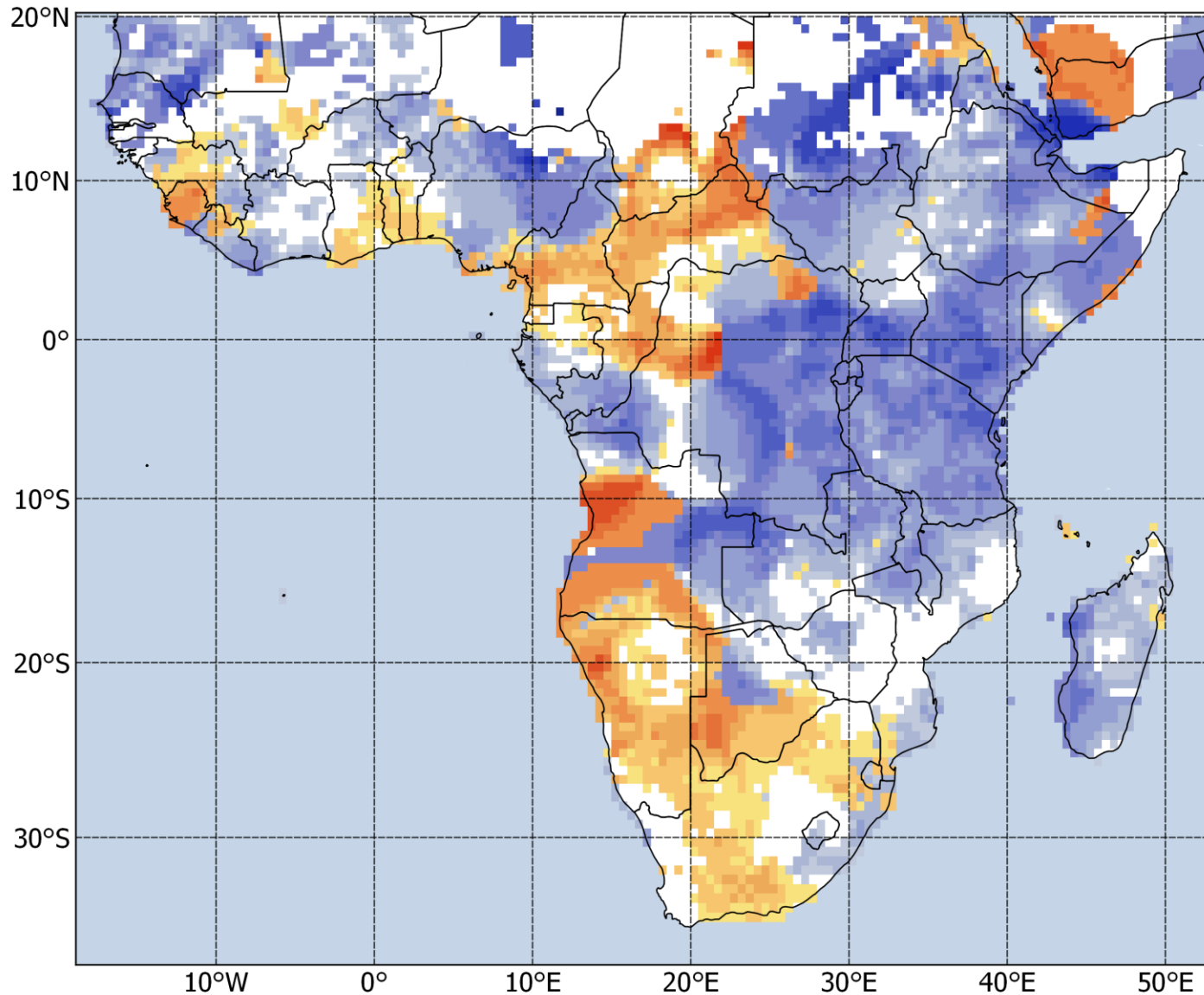
Probabilistic forecasts can help users understand risks and opportunities (forewarned is forearmed) in order to make more informed decisions.

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.

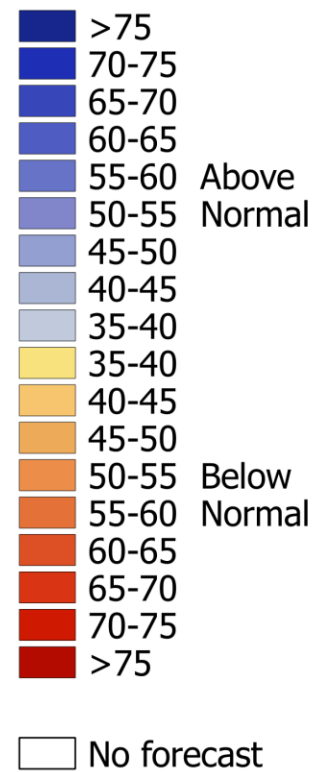
Note: Probabilistic forecasts are considered reliable when the forecast probability is an accurate estimation of the relative frequency of the predicted outcome. In other words, forecasts are reliable if the observation falls within the category (Below-, Near- or Above-Normal) as frequently as the forecast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

JFM 2024 Rainfall; ICs: Jan

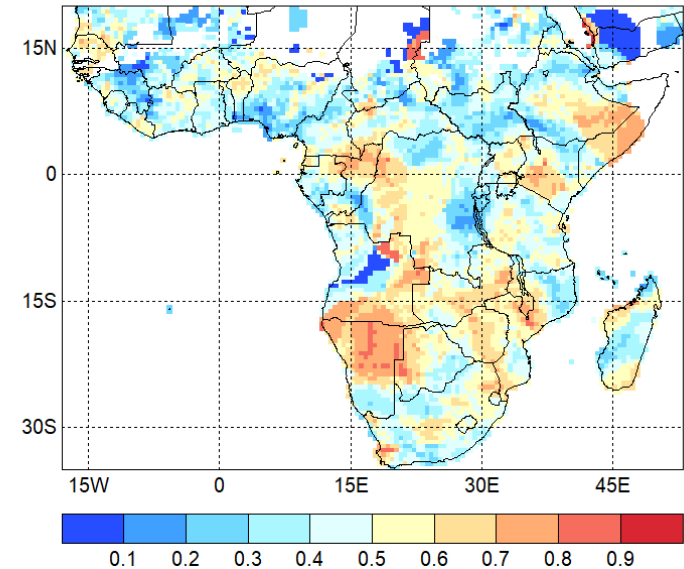


Legend

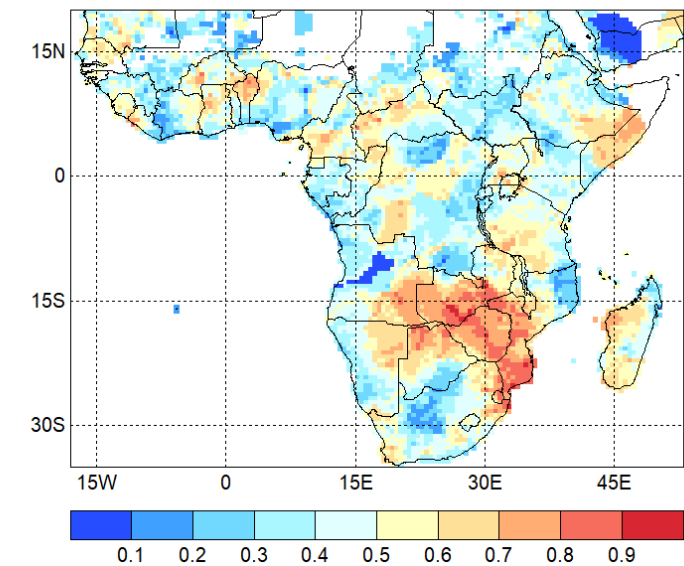
Rainfall Prob



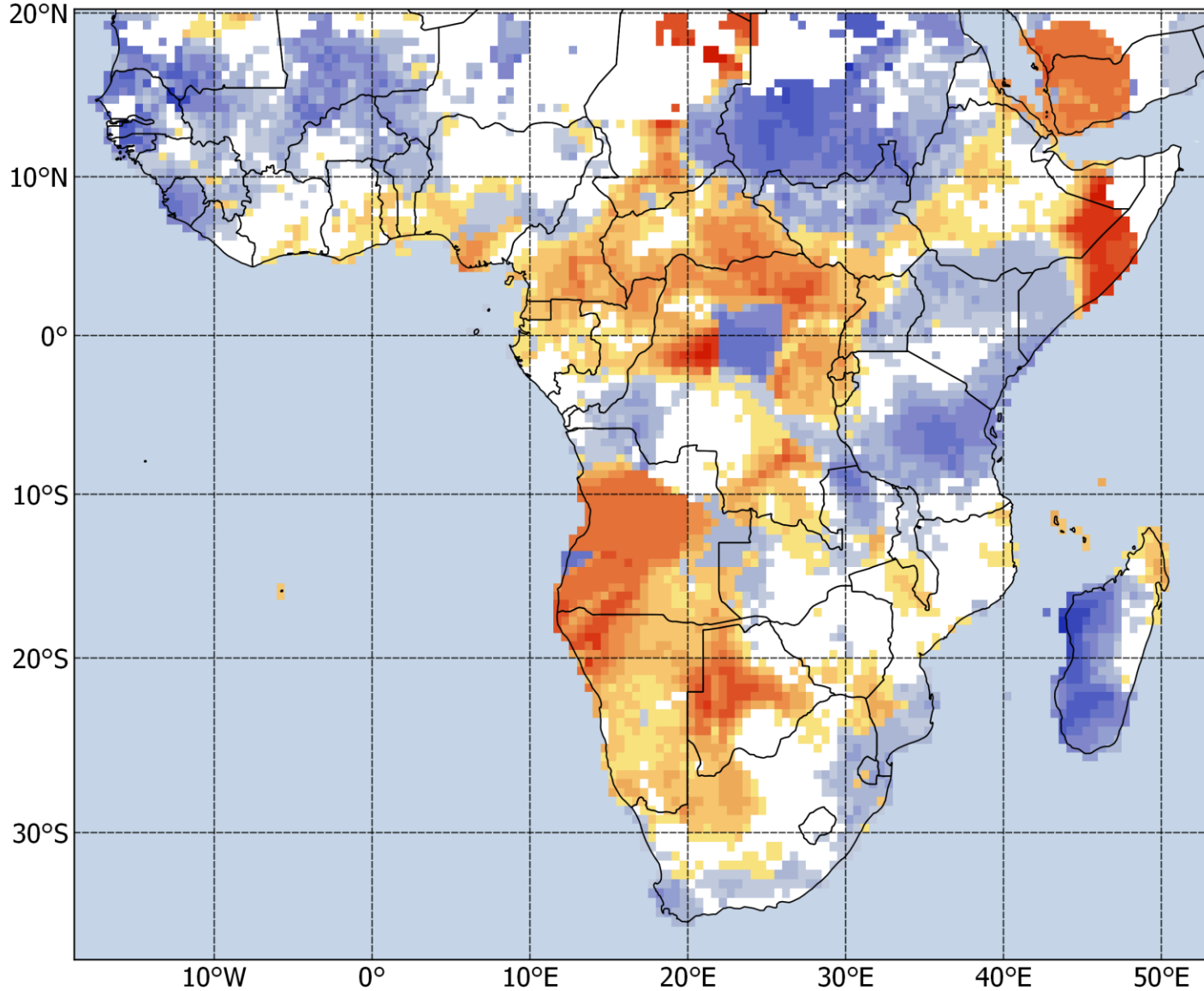
ROC area (above-normal): JFM Rainfall



ROC area (below-normal): JFM Rainfall

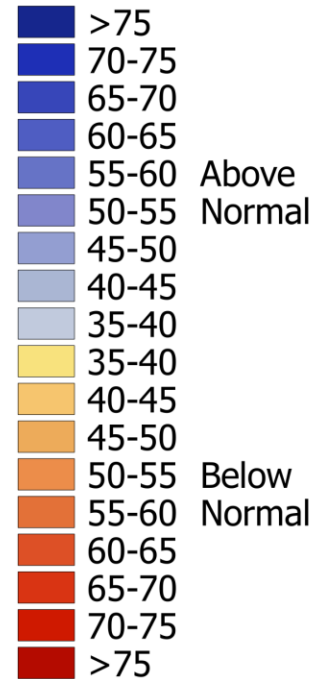


FMA 2024 Rainfall; ICs: Jan



Legend

Rainfall Prob

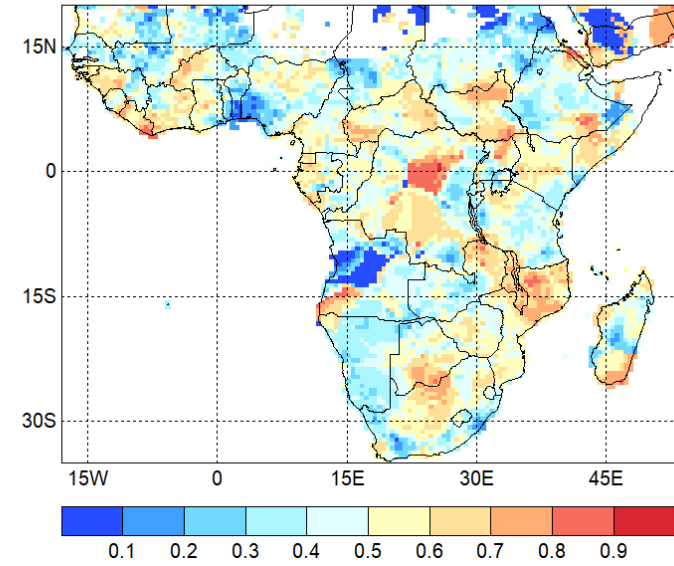


Above Normal

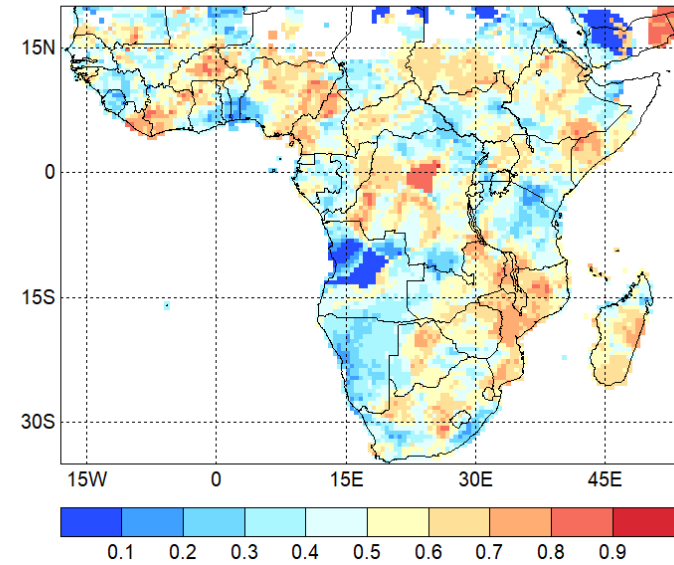
Below Normal

No forecast

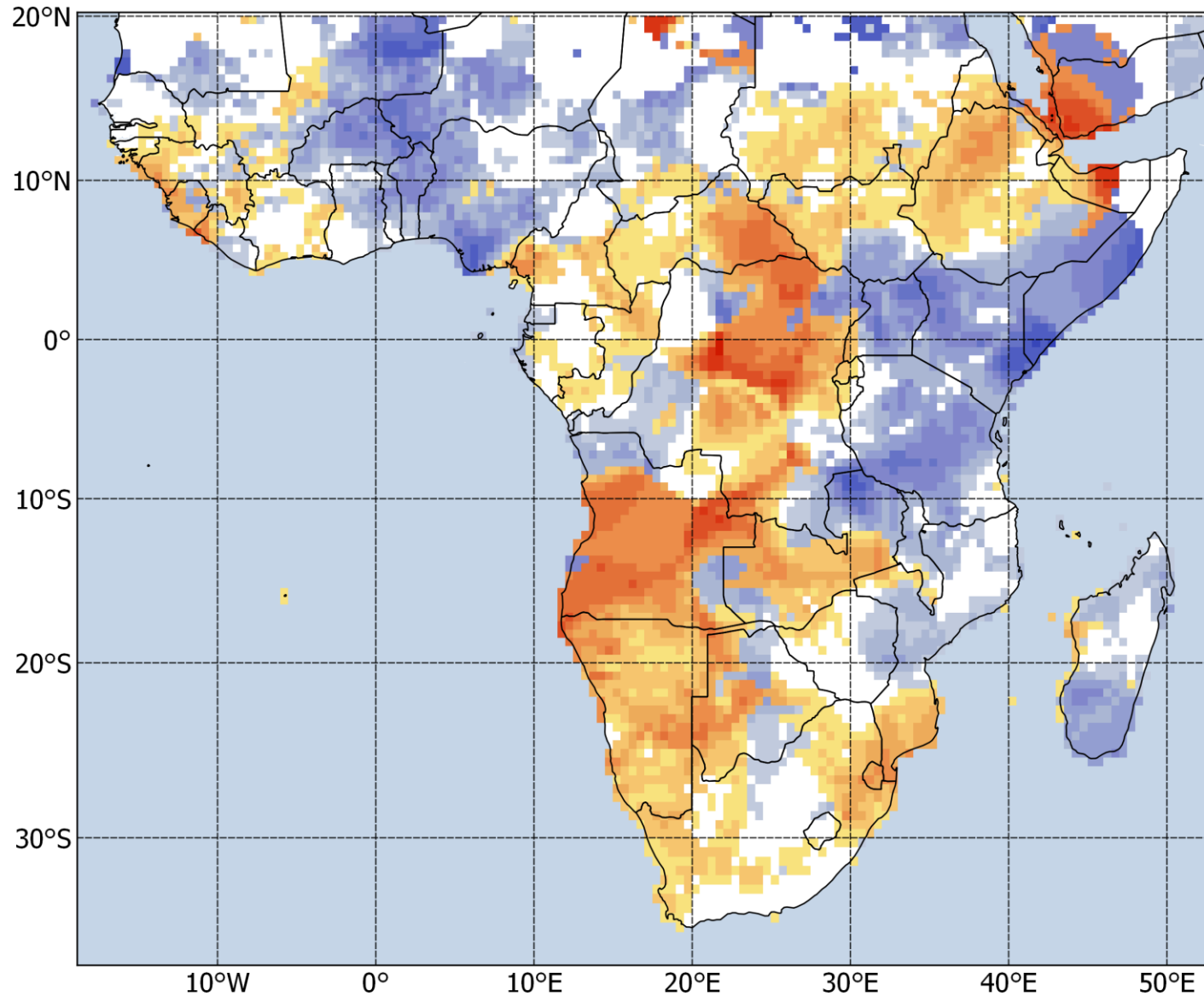
ROC area (above-normal): FMA Rainfall



ROC area (below-normal): FMA Rainfall

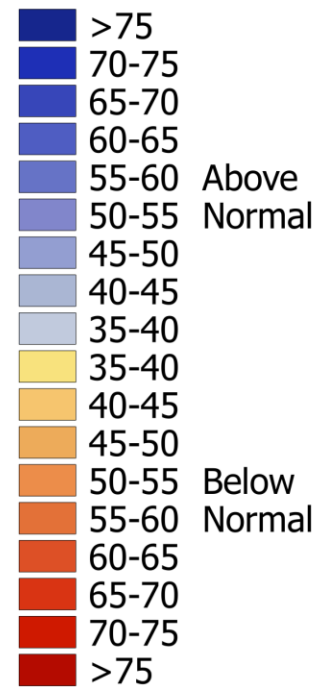


MAM 2024 Rainfall; ICs: Jan



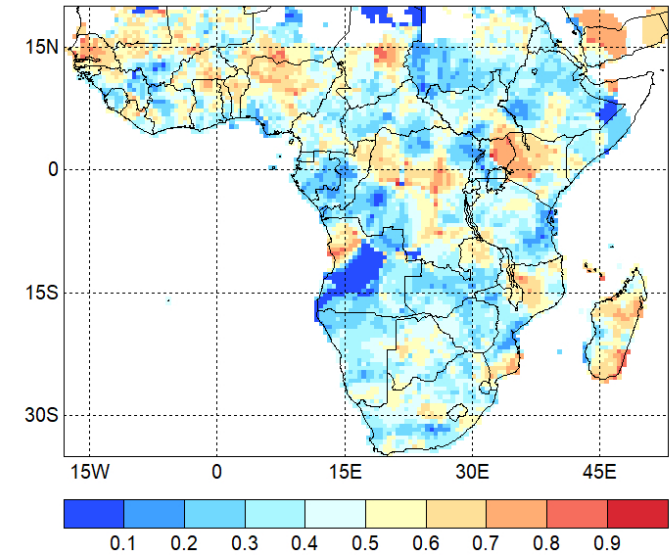
Legend

Rainfall Prob

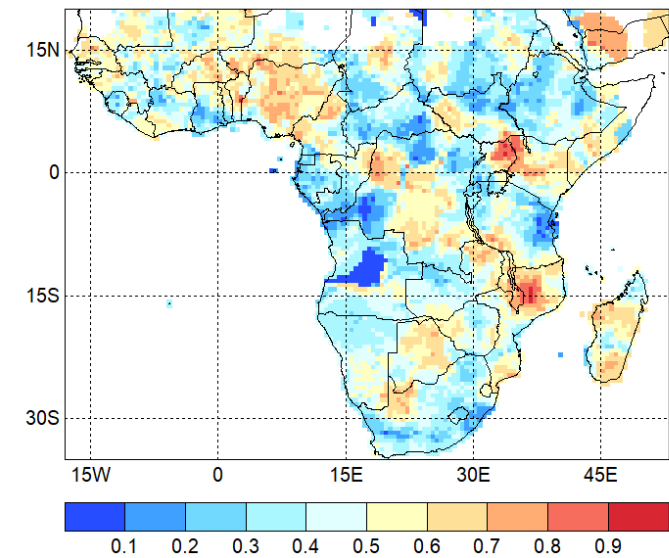


□ No forecast

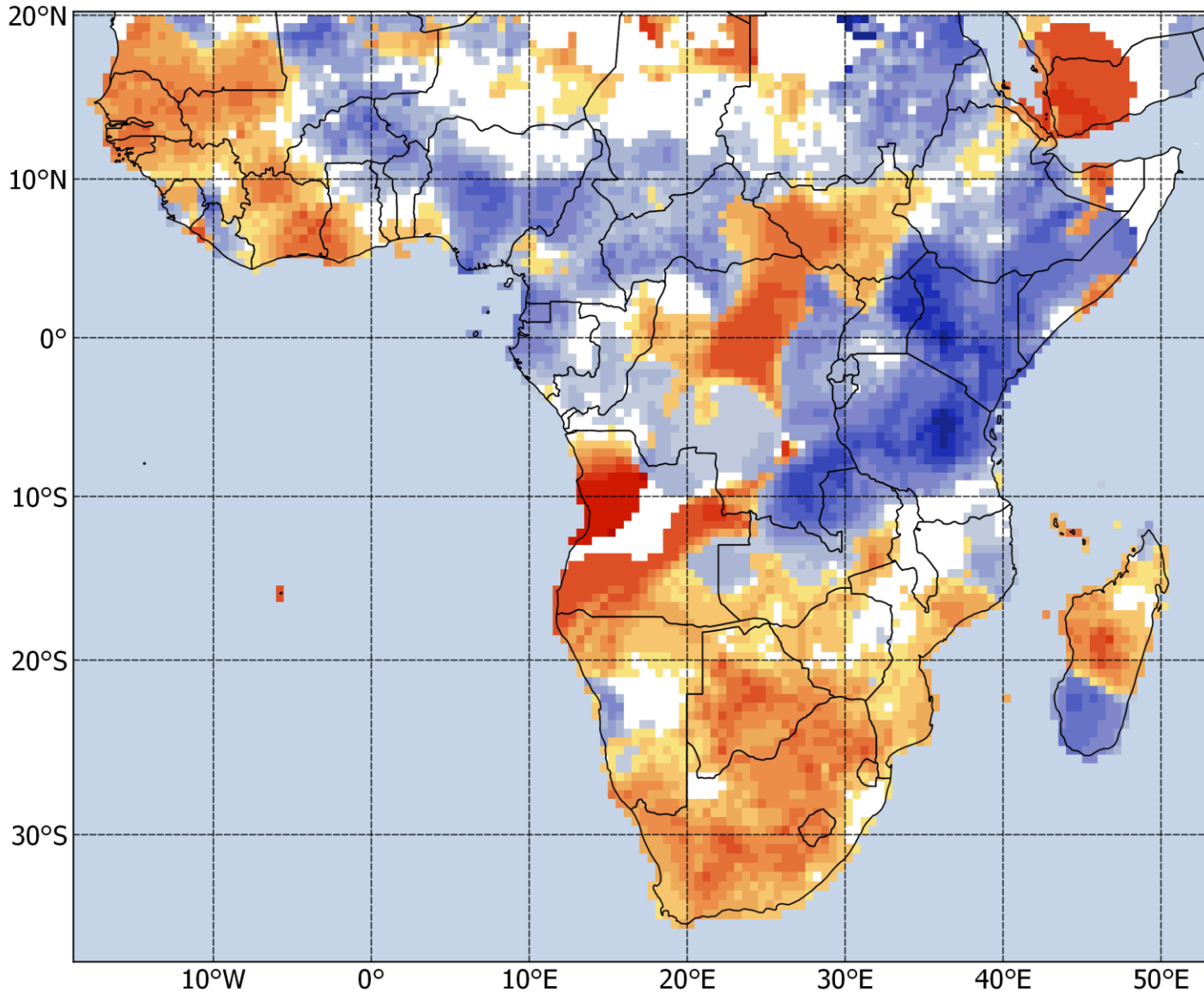
ROC area (above-normal): MAM Rainfall



ROC area (below-normal): MAM Rainfall

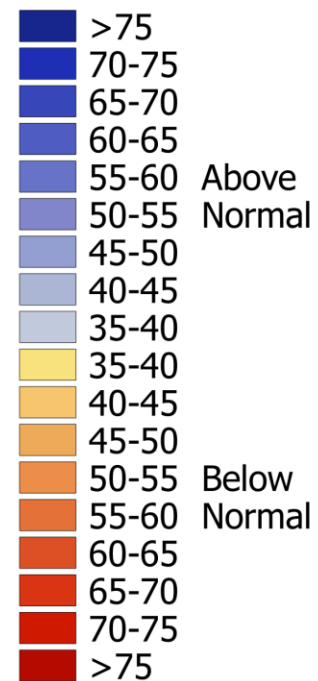


AMJ 2024 Rainfall; ICs: Jan



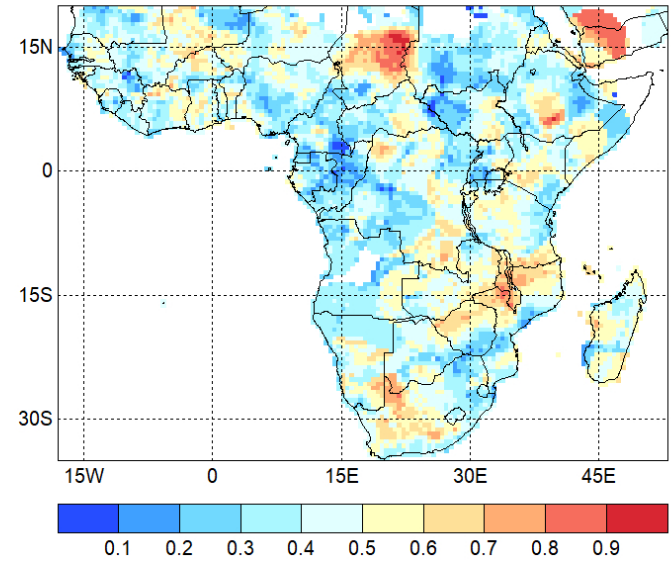
Legend

Rainfall Prob

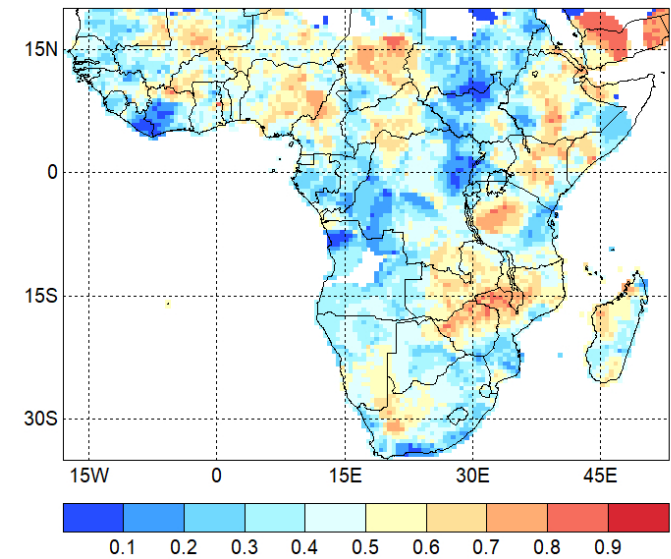


□ No forecast

ROC Area (Above-Normal): AMJ Rainfall



ROC Area (Below-Normal): AMJ Rainfall

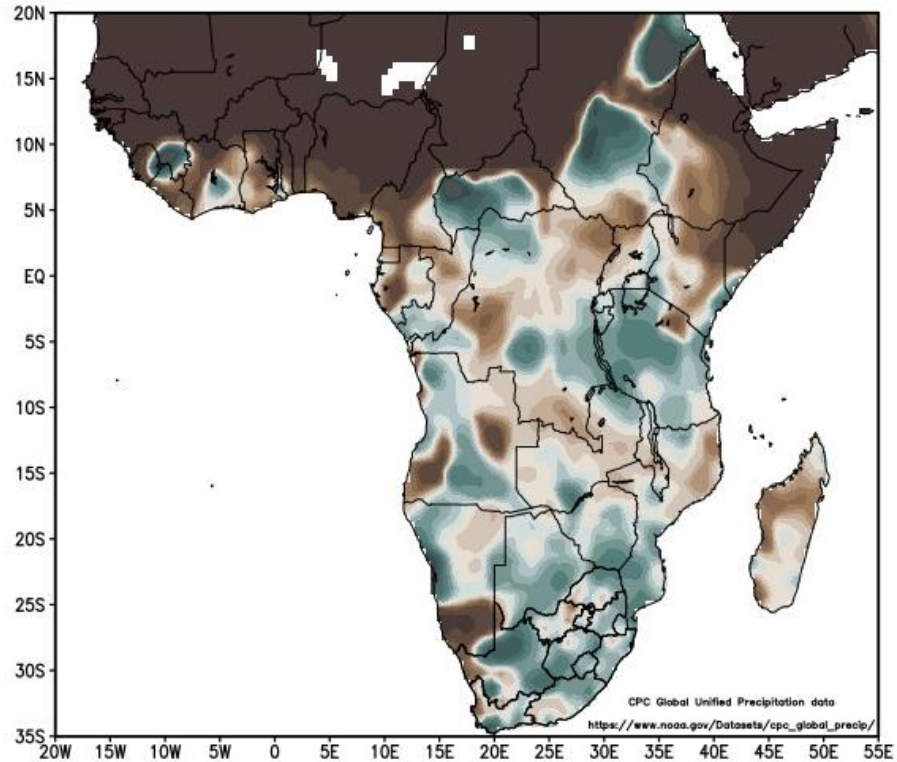


Round-up: Rainfall south of 15°S

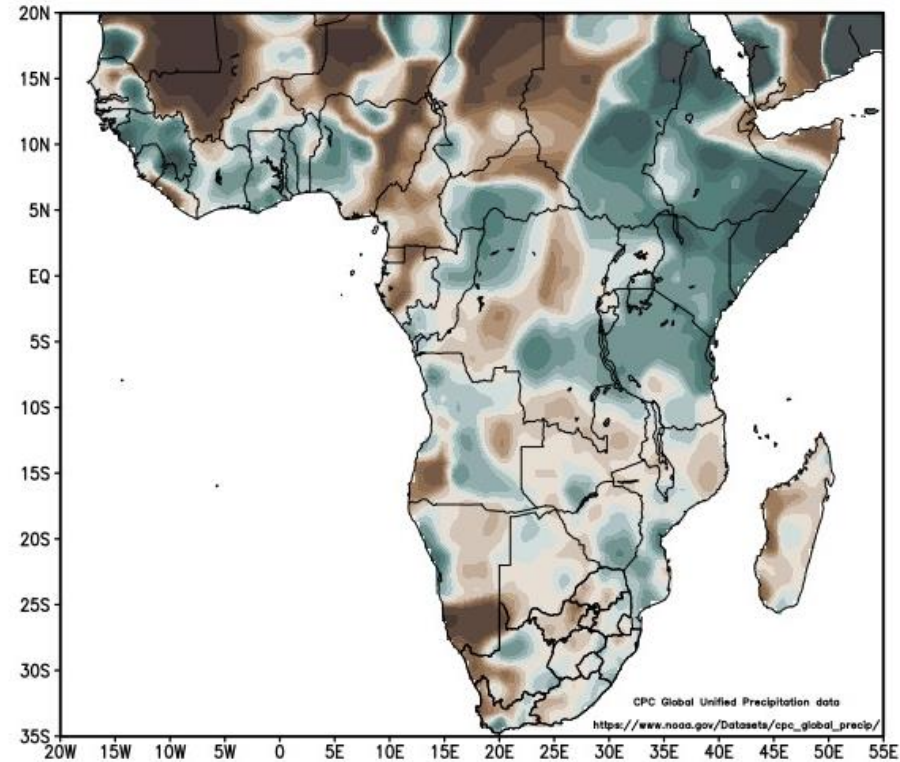
- JFM and FMA: Enhanced probabilities of below-normal rainfall totals, except over the east and south-eastern areas where above-normal totals are likely
- MAM and AMJ: Although the forecast is for below-normal rainfall totals in most areas, the forecast skill for these two seasons is low and this reduces confidence in the forecasts for these seasons

Observed Rainfall

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



Rainfall (% of normal): October–November–December 2023
October–November–December long-term mean: 1981–2010

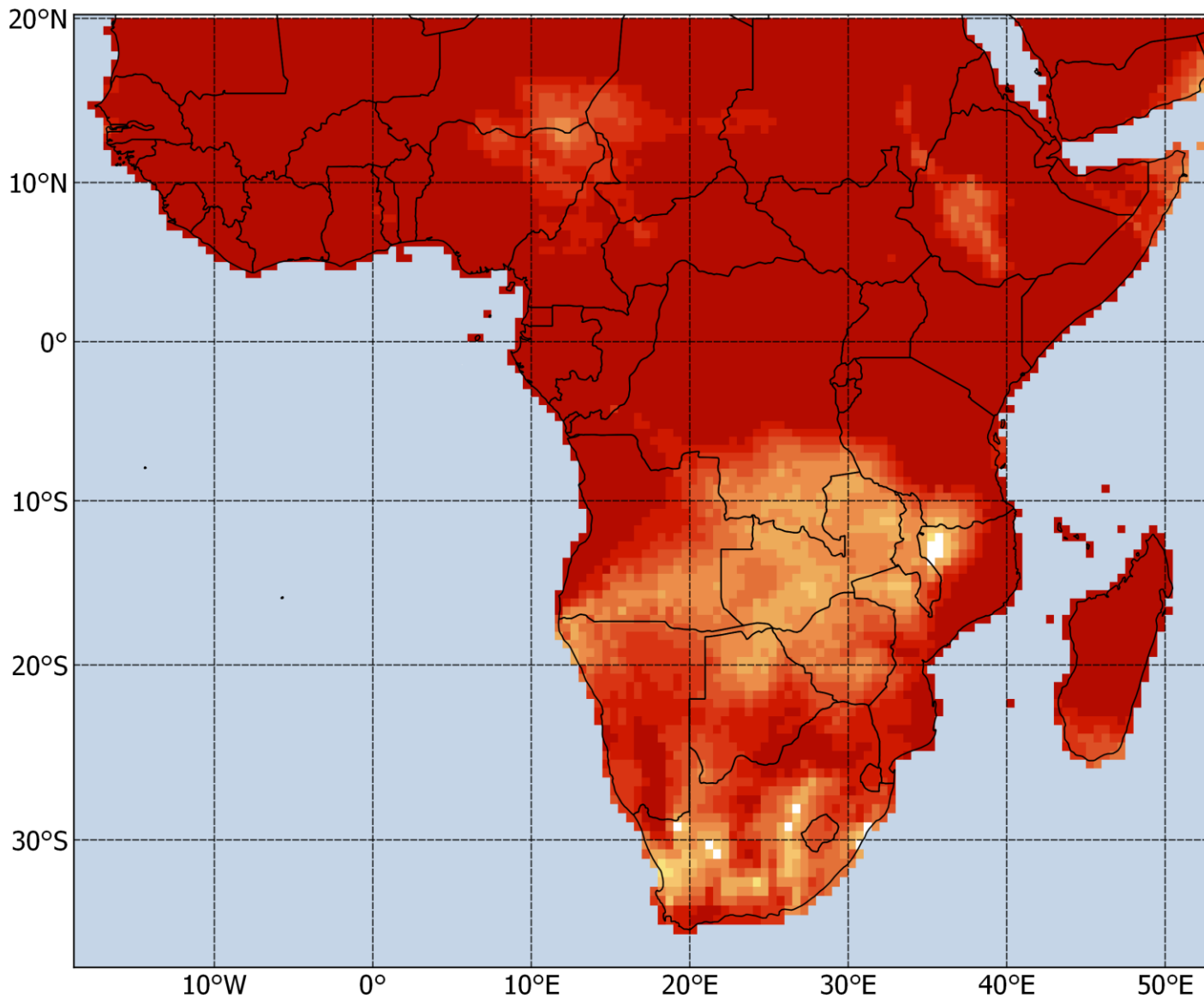


Recorded rainfall for December and the October–November–December season show below-normal rainfall over the brown areas and above-normal rainfall over the green areas

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

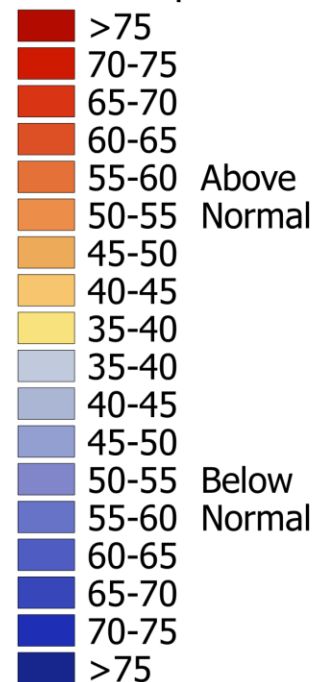
0 10 20 30 40 50 60 80 100 120 140 160 200 250 300 400 500 (%)

JFM 2024 Max Temp; ICs: Jan



Legend

Max Temp Prob

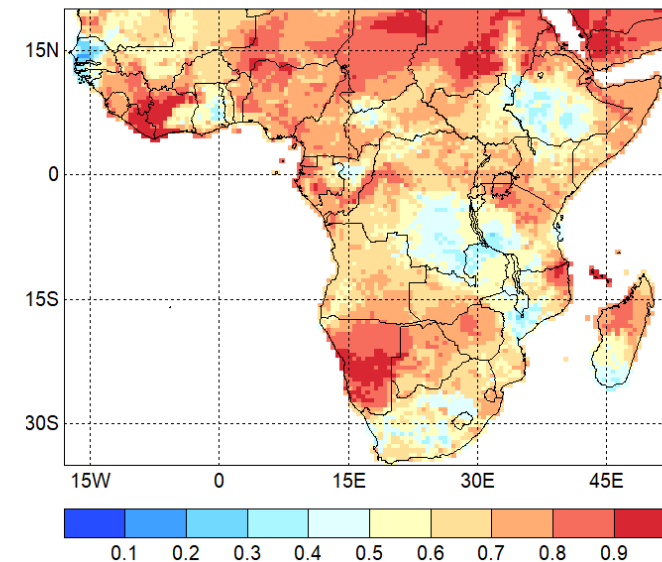


Above
Normal

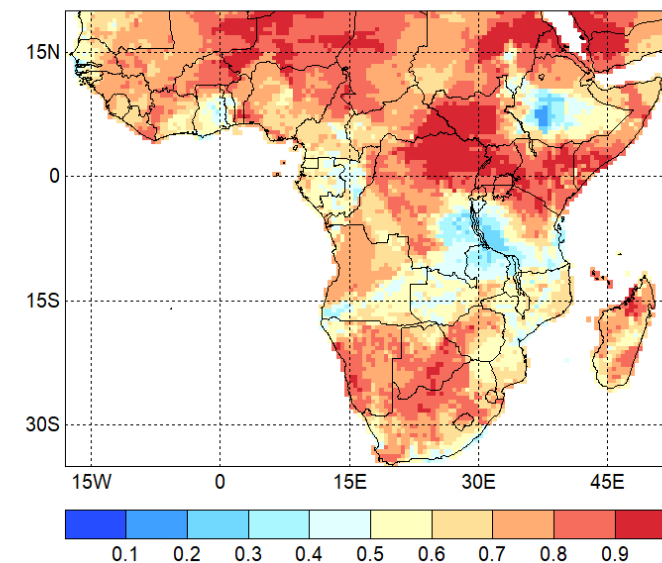
Below
Normal

No Forecast

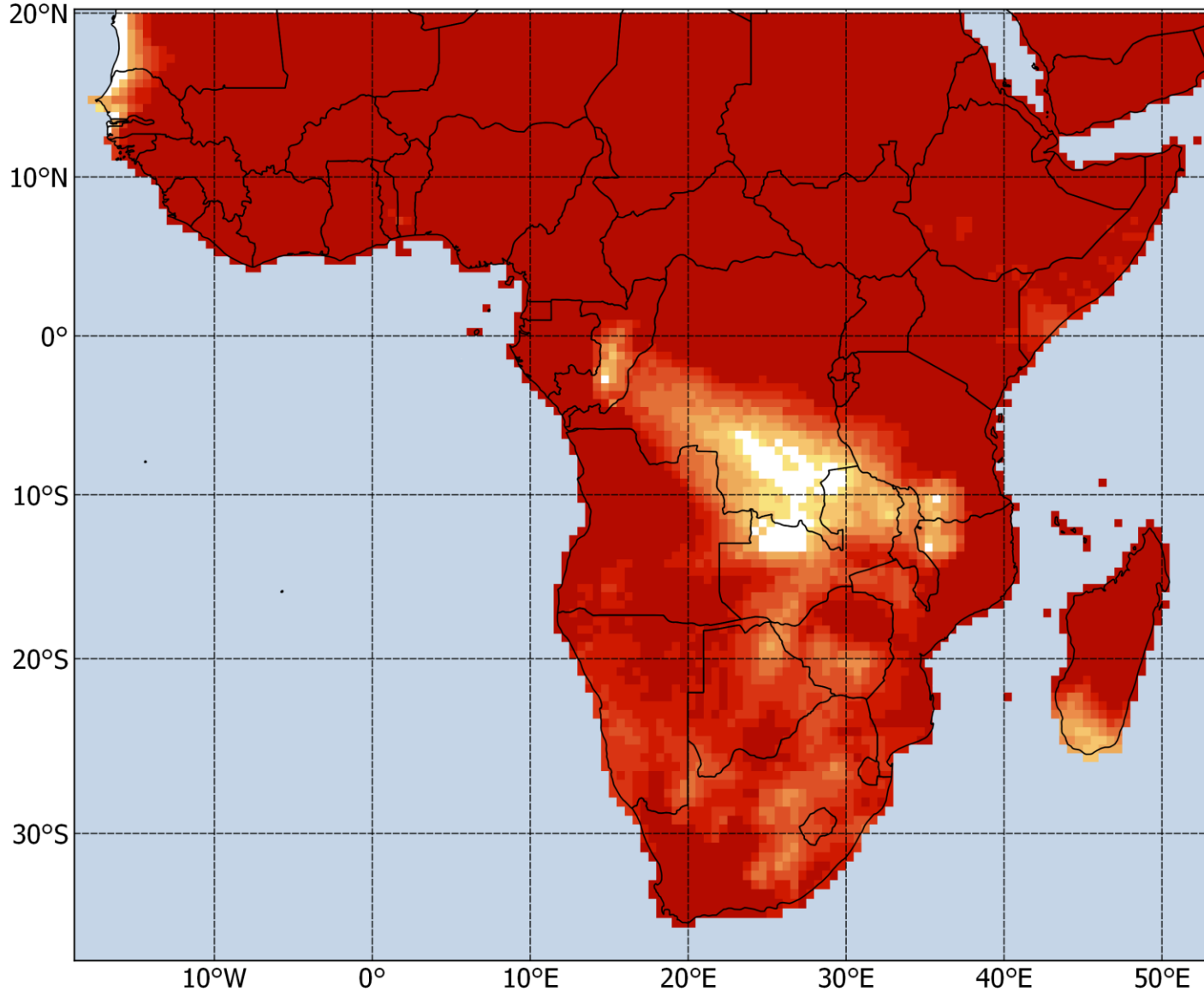
ROC Area (Above-Normal): JFM Max Temp



ROC Area (Below-Normal): JFM Max Temp

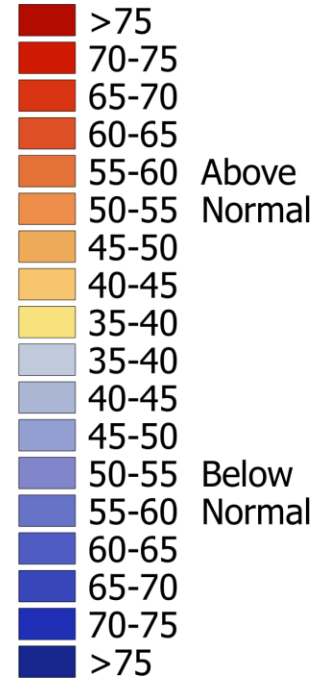


FMA 2024 Max Temp; ICs: Jan



Legend

Max Temp Prob

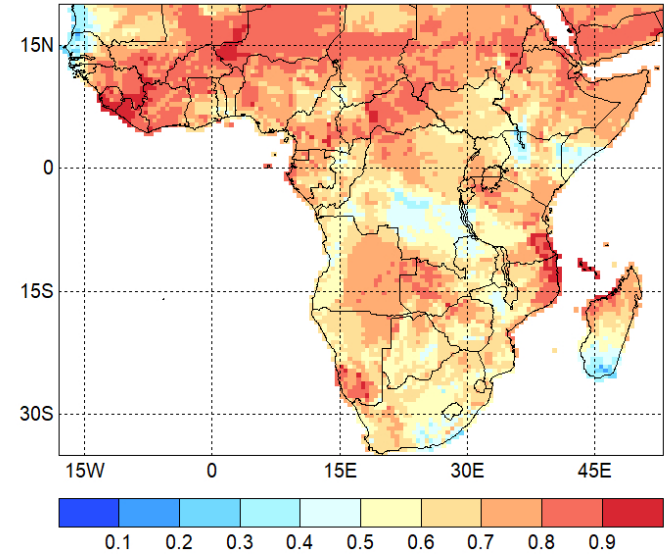


Above Normal

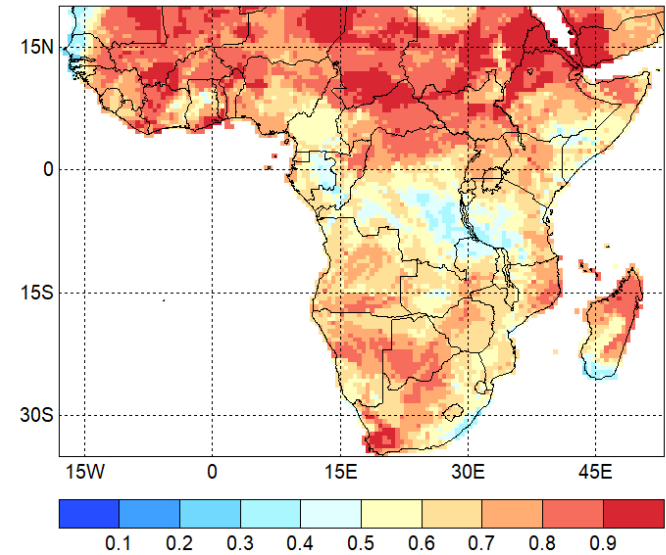
Below Normal

No Forecast

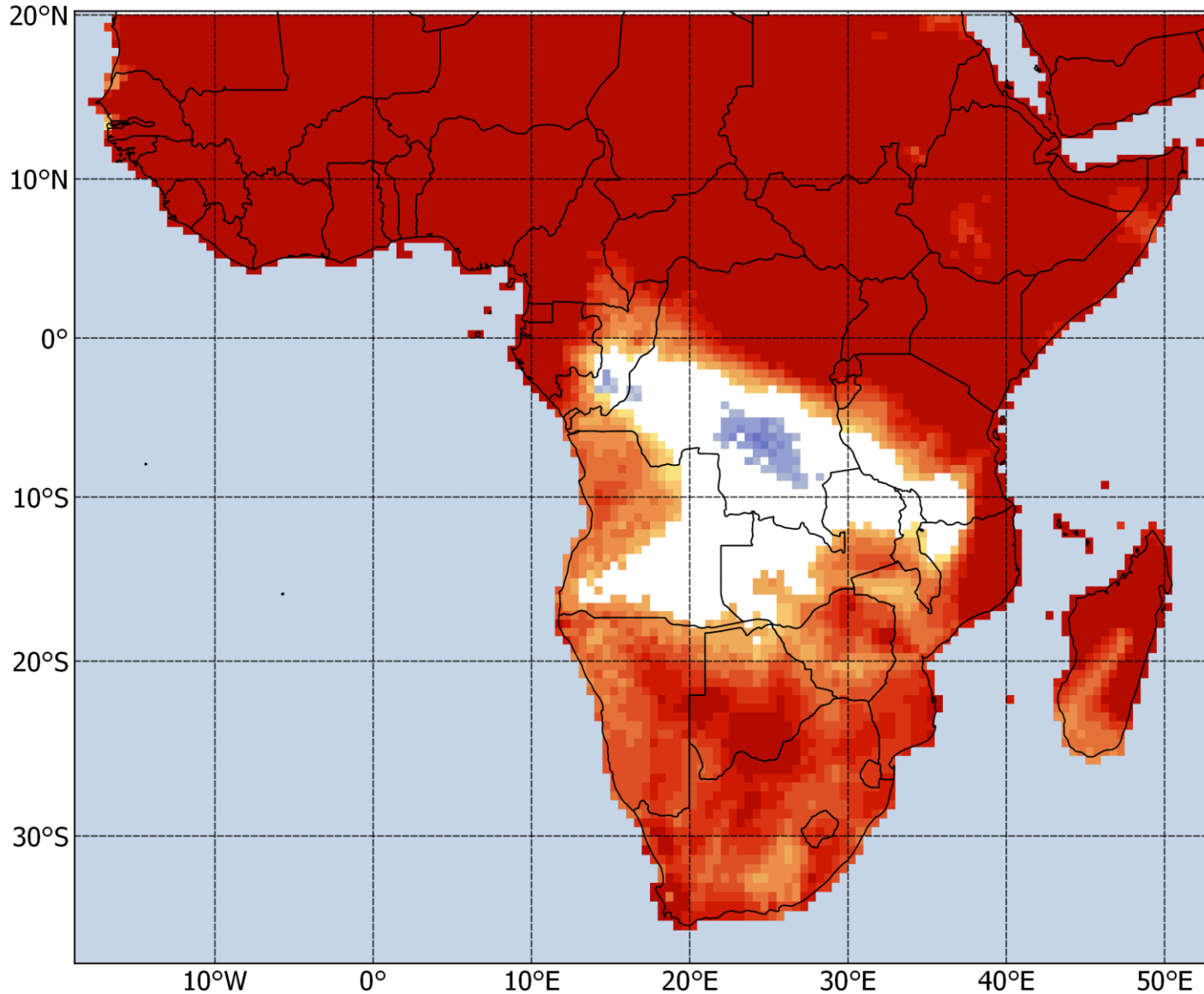
ROC Area (Above-Normal): FMA Max Temp



ROC Area (Below-Normal): FMA Max Temp

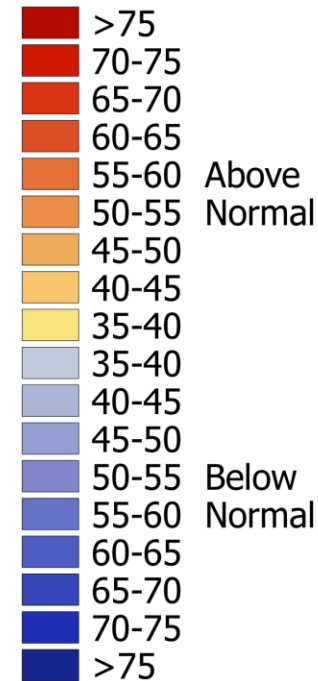


MAM 2024 Max Temp; ICs: Jan



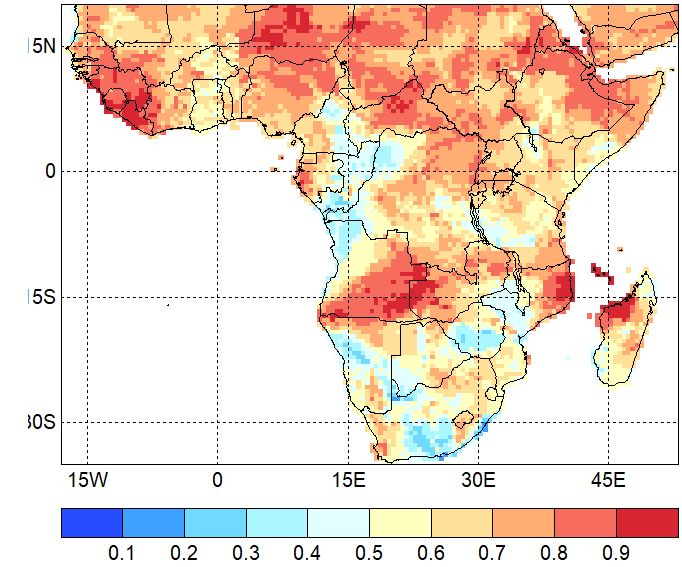
Legend

Max Temp Prob

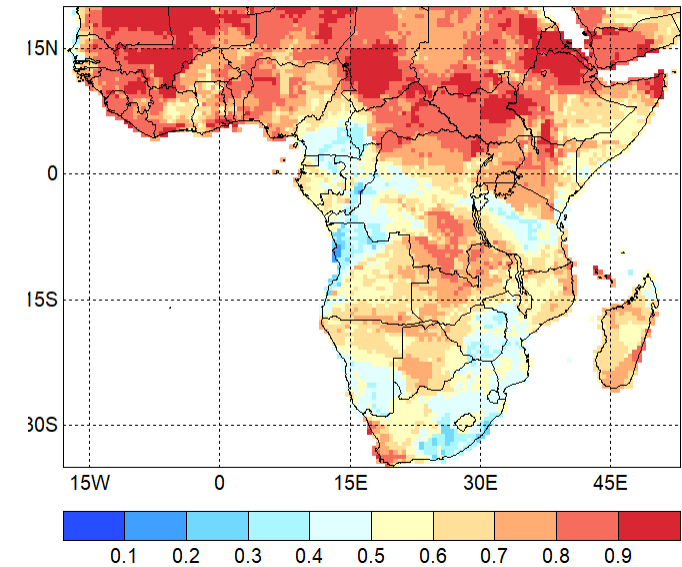


White box: No Forecast

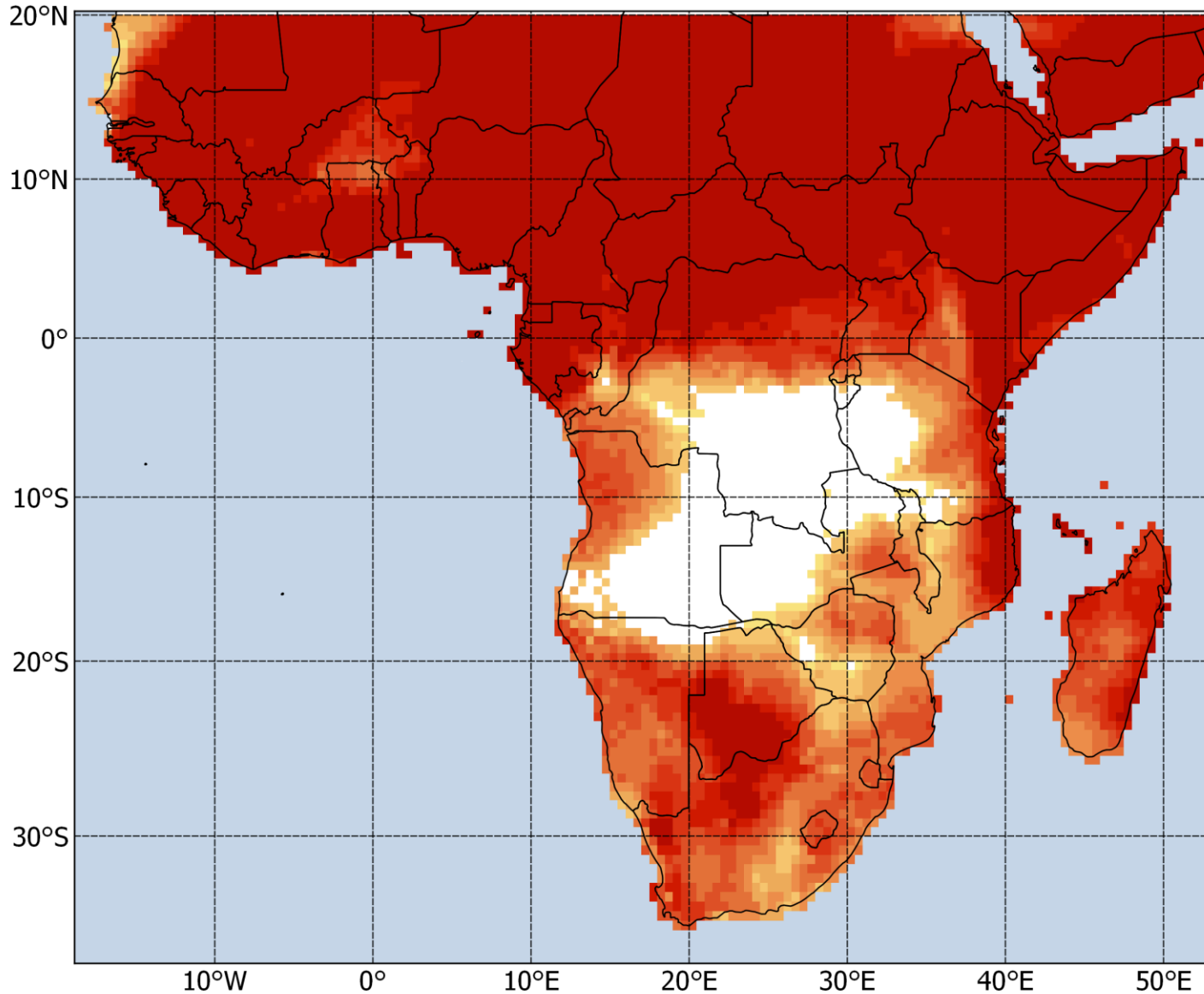
ROC Area (Above-Normal): MAM Max Temp



ROC Area (Below-Normal): MAM Max Temp

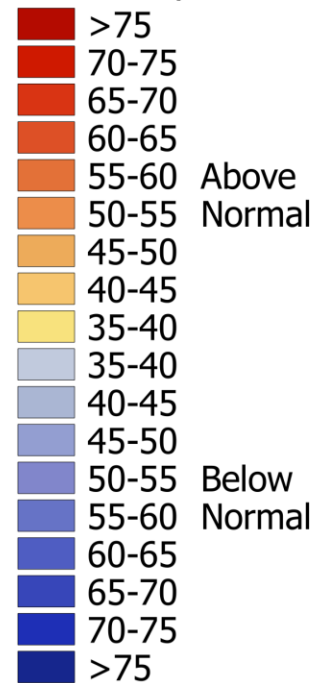


AMJ 2024 Max Temp; ICs: Jan



Legend

Max Temp Prob

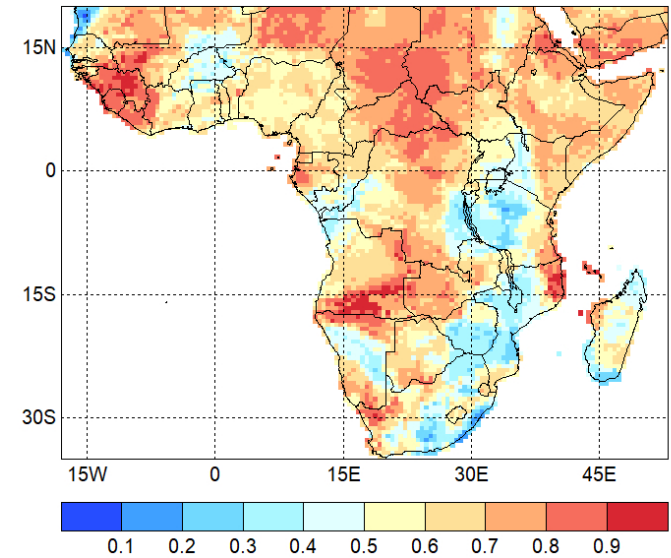


Above
Normal

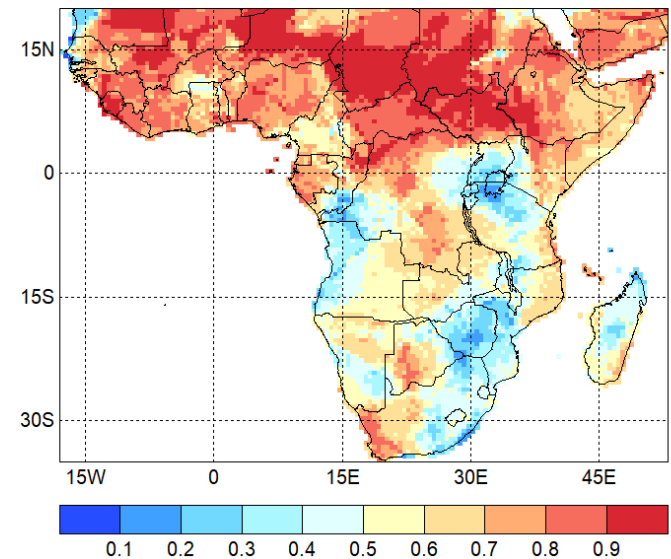
Below
Normal

White Box No Forecast

ROC Area (Above-Normal): AMJ Max Temp



ROC Area (Below-Normal): AMJ Max Temp



Round-up: South of 15°S Max Temp

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

References

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 - Project: Application of knowledge for the management of extreme climate events (APECX; 2022 to 2024)
- ACCESS (Alliance for Collaboration on Climate and Earth System Science) through the project “Investigating predictability of seasonal anomalies for societal benefit” (2018 to 2021)
- Water Research Commission through administering the international project “Research-based Assessment of Integrated approaches to Nature-based SOLUTIONS (RainSolutions)” (2020 to 2022)



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

Acknowledgments to Dr Peter Johnston of the University of Cape Town for professional comments and advice

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.