

Seasonal forecasts

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<https://tinyurl.com/ybrb3a72>



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Seasonal Forecast Worx

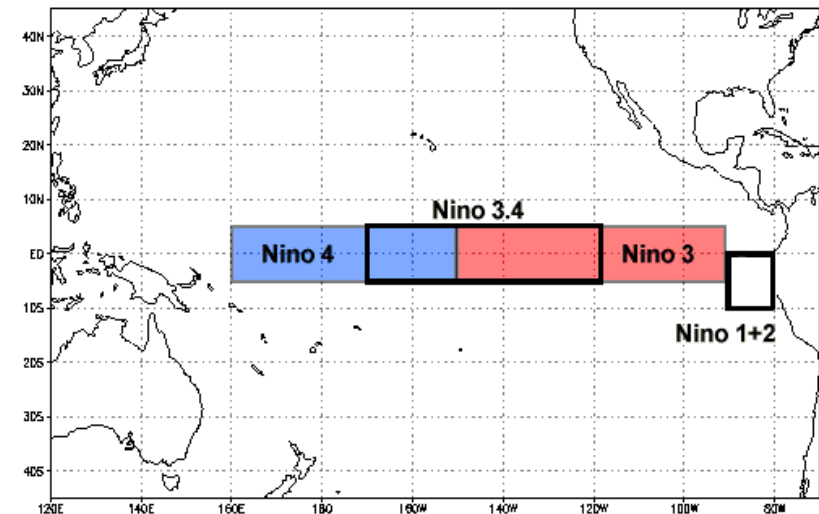
Latest Update: 15 November 2021

- 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

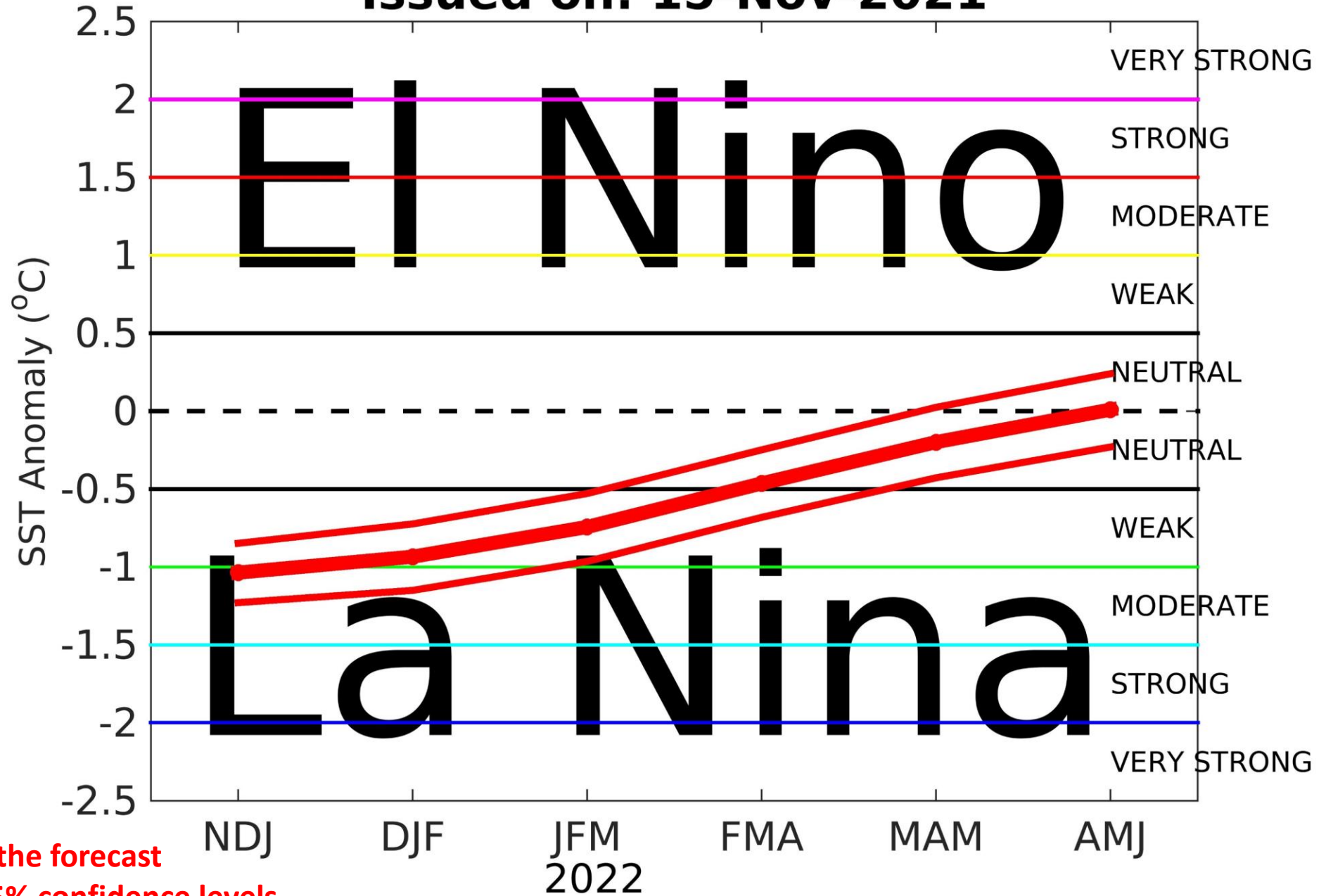
Prediction Method

- 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

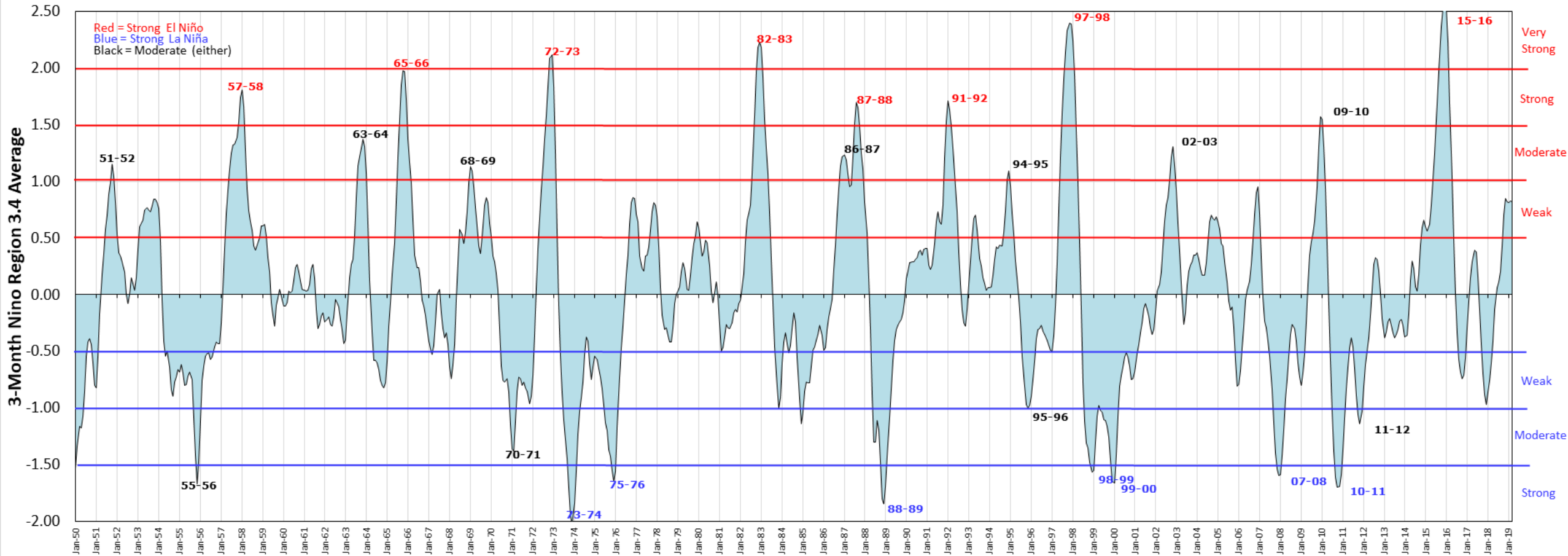
Issued on: 15-Nov-2021



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

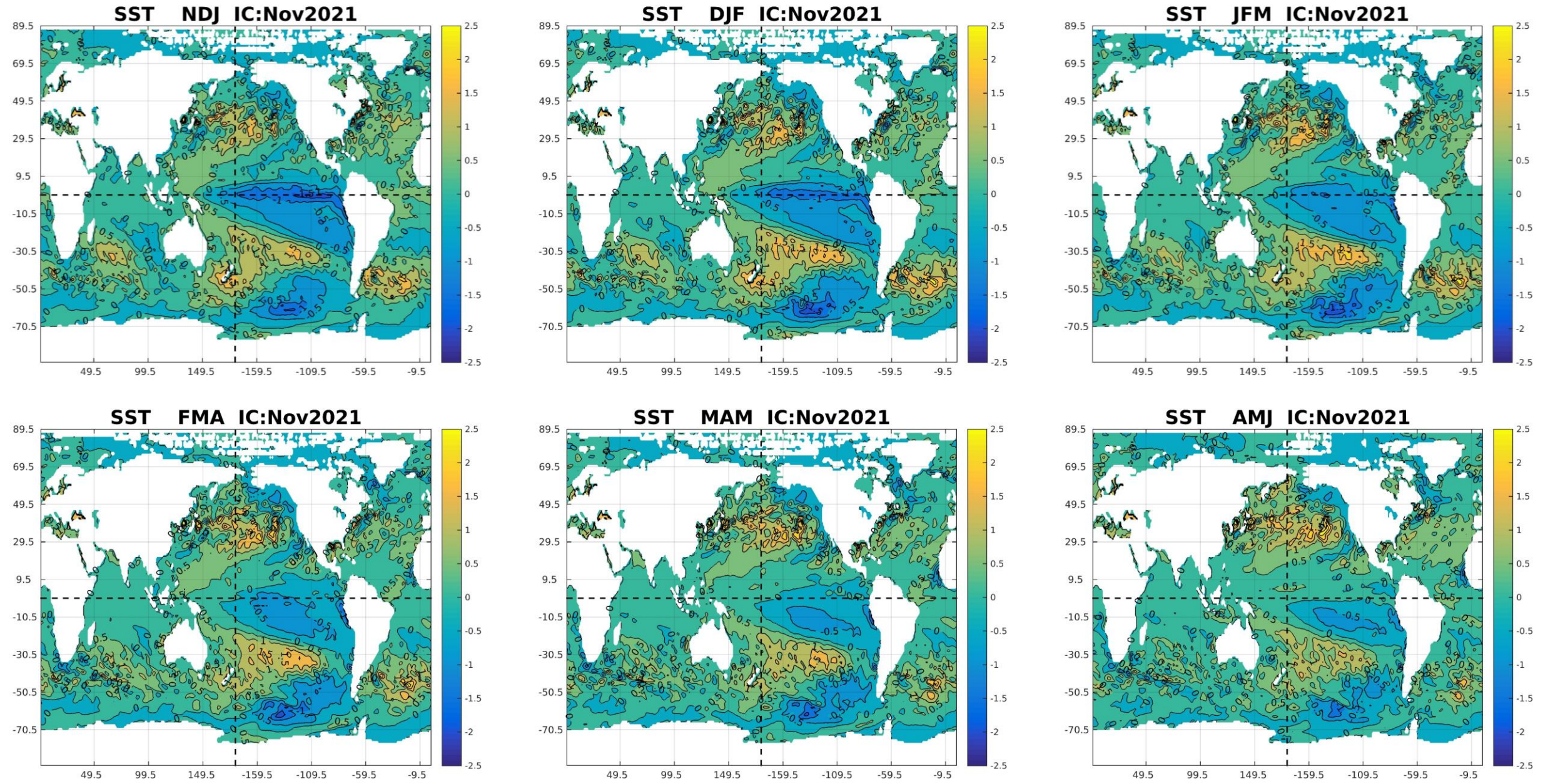
Oceanic Niño Index (ONI)

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



IC: the month in which the forecast was made

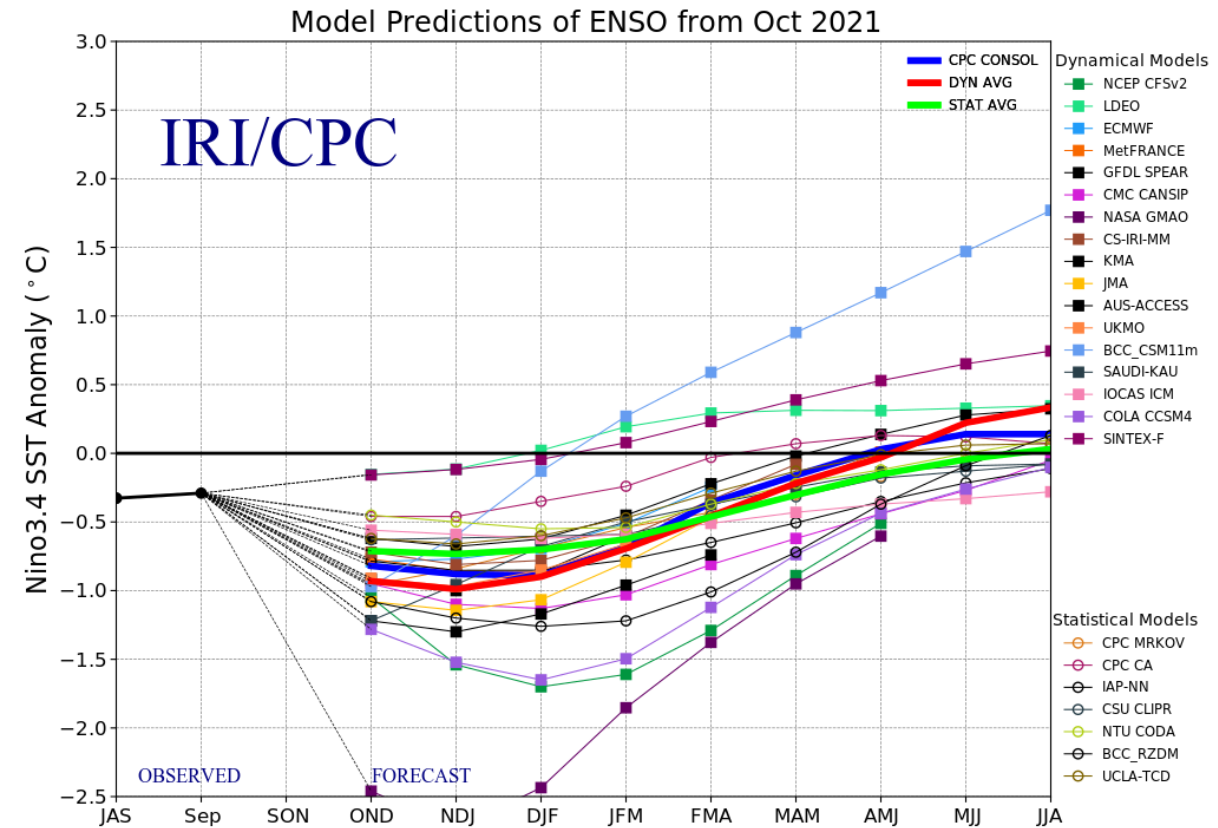
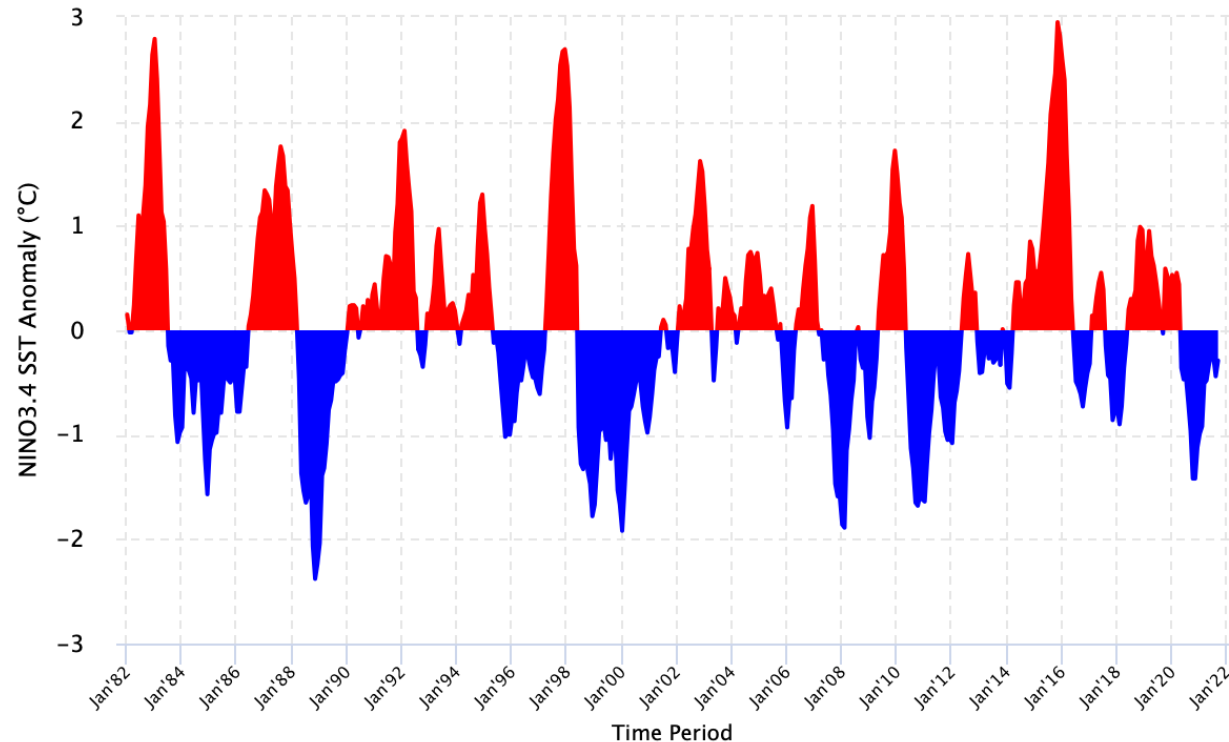
SST anomalies (in °C, where blue is cooler and orange is warmer)



Round-up: ENSO

- The UP model predicts moderate to weak La Niña conditions during mid-summer, supported by the plume of forecasts below

Historical Nino 3.4 Sea Surface Temperature Anomaly



Southern Africa Forecasts

Prediction Method

- Three-month seasons for seasonal rainfall totals and average maximum temperatures of NMME ensemble mean forecasts are interpolated to the 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” 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)
- Verification of forecast skill:
 - 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.

Forecasts are probabilistic

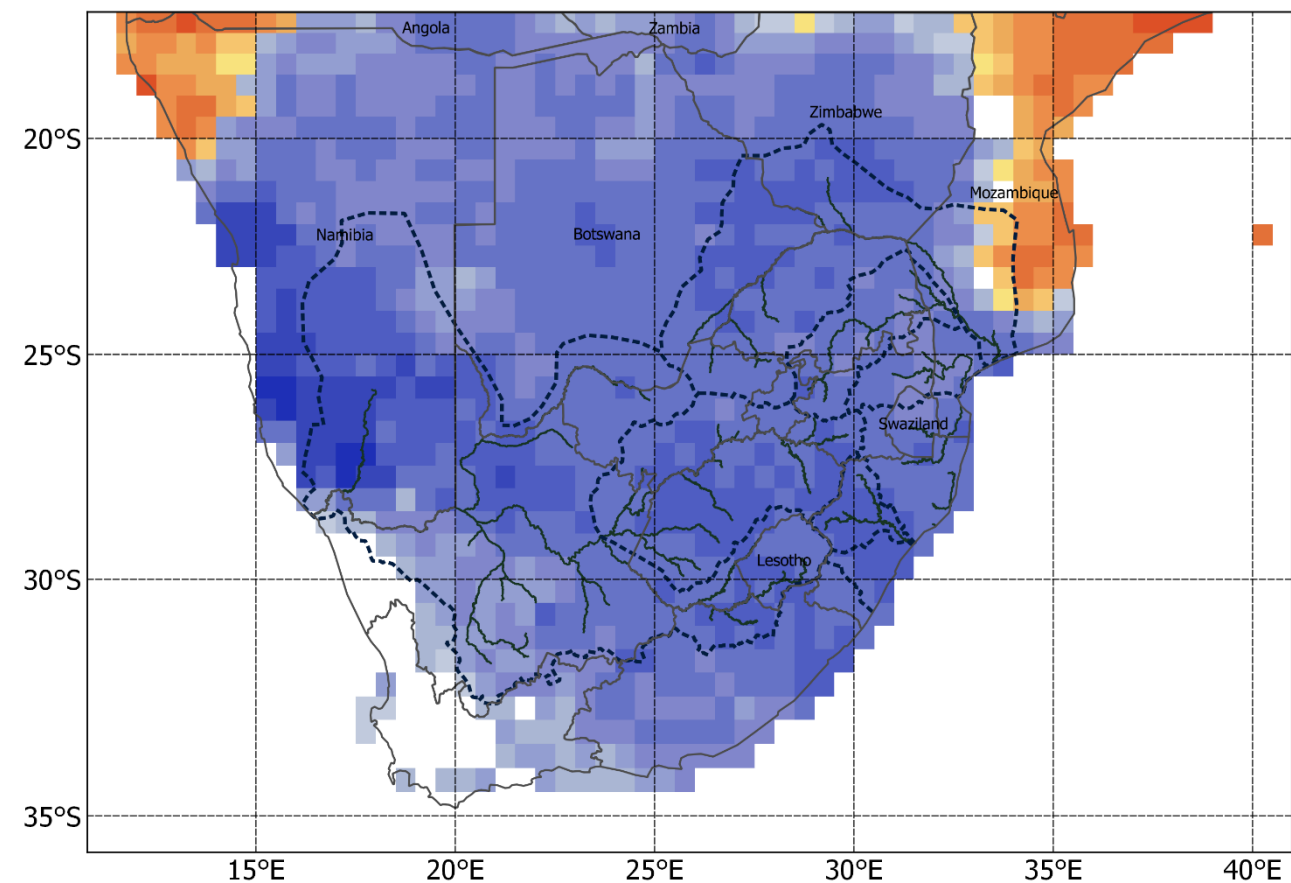
Probabilistic forecasts can help users understand risks and opportunities (forewarned is forearmed) in order to make more informed decisions.

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 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% 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.

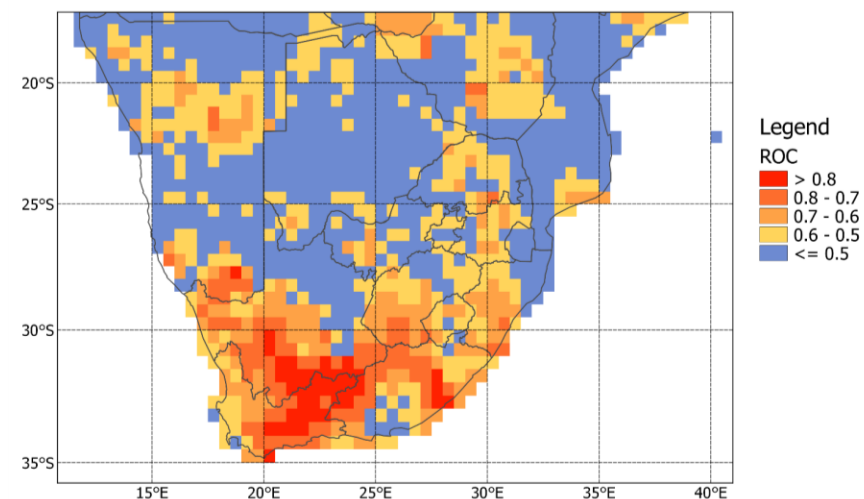
The nature of a probabilistic forecast implies that the less likely outcomes are always possible. In fact, for the probabilistic forecasts to be considered reliable, the less likely outcomes will and must occasionally occur.

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

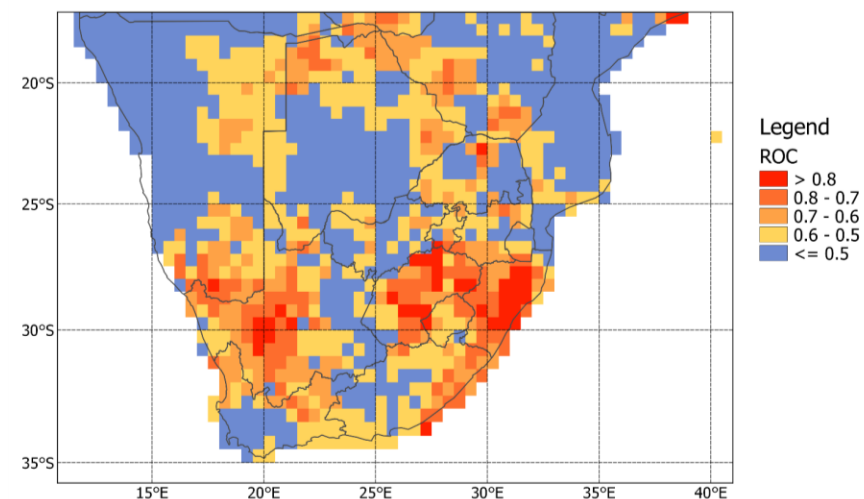
NDJ 2021/22 Rainfall; ICs: Nov



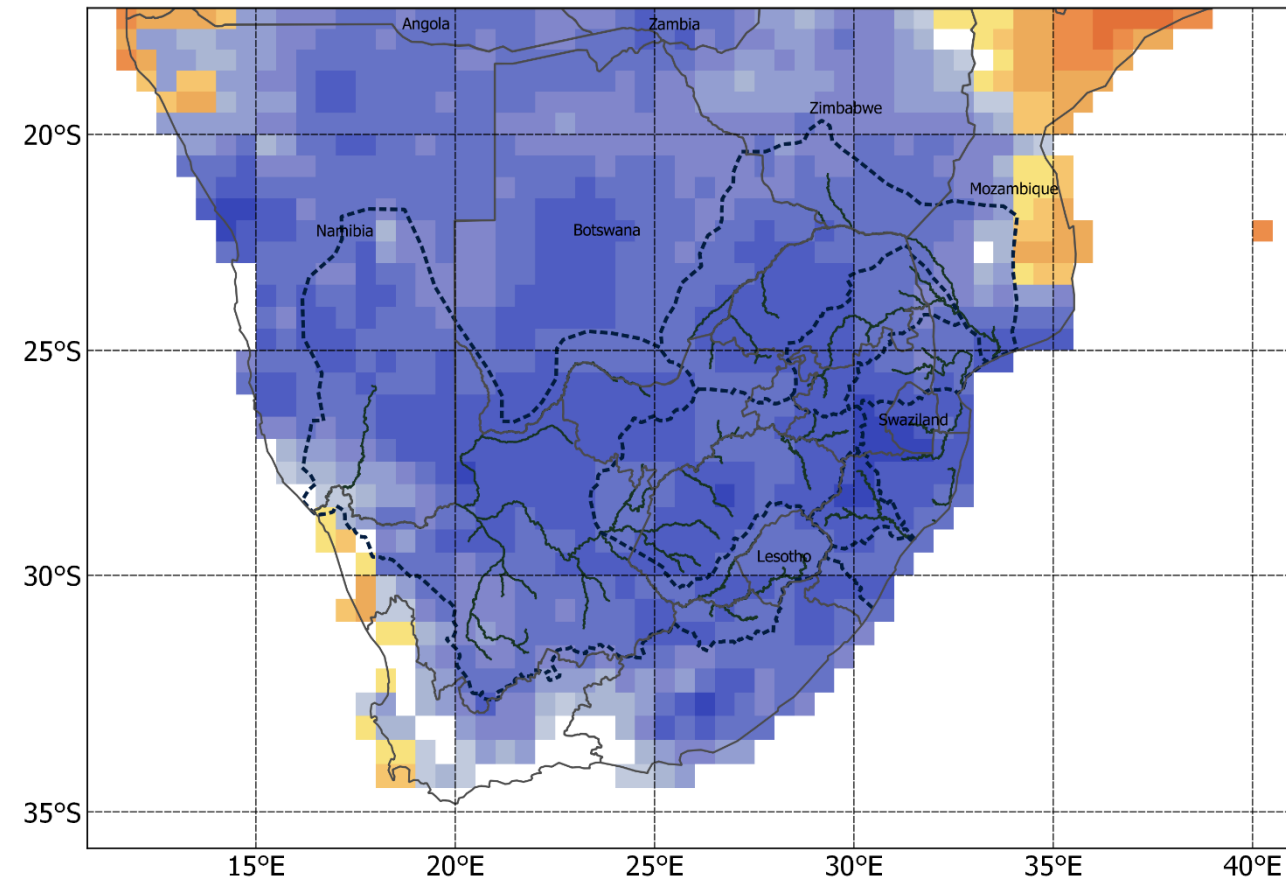
ROC Area (Above-Normal): NDJ Rainfall



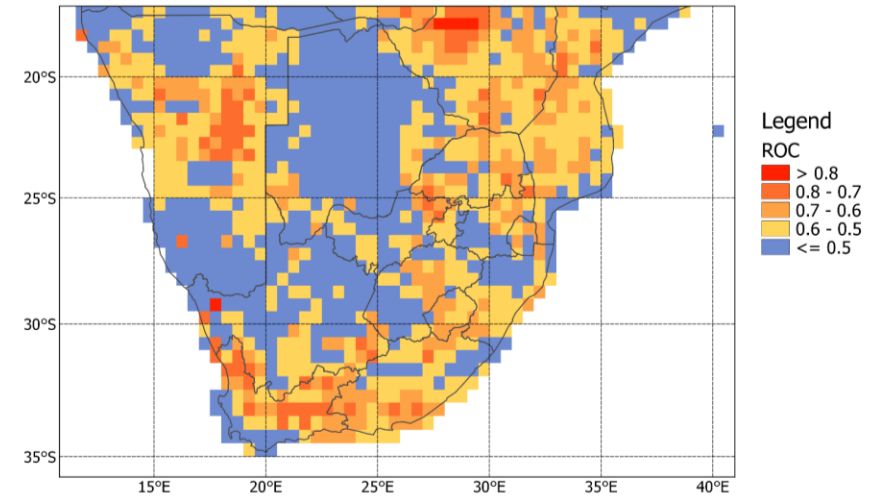
ROC Area (Below-Normal): NDJ Rainfall



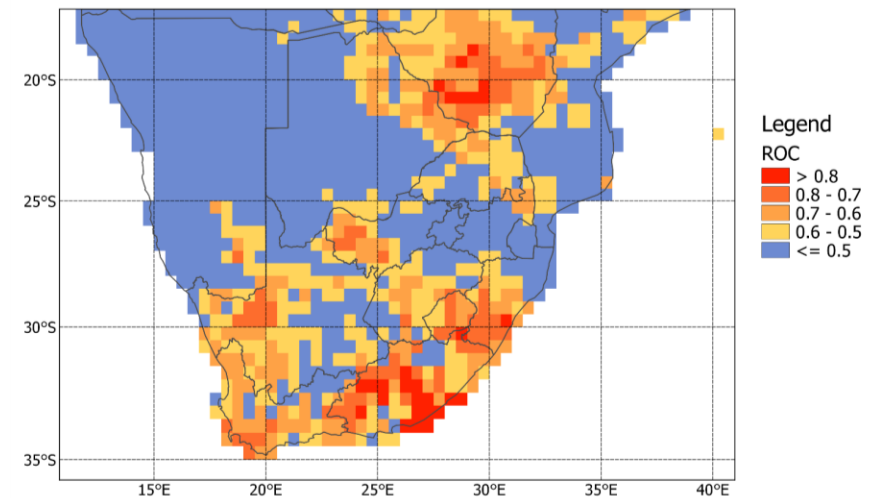
DJF 2021/22 Rainfall; ICs: Nov



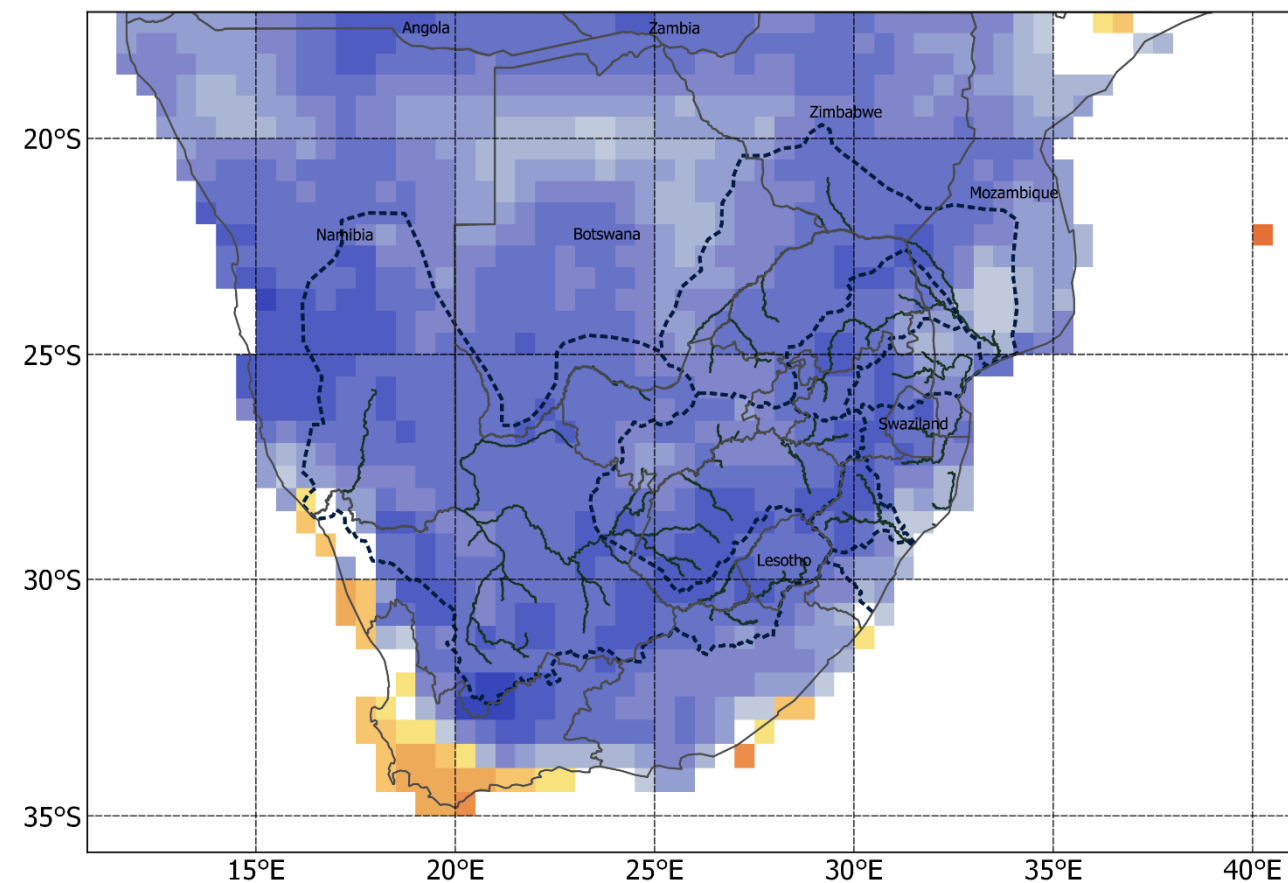
ROC Area (Above-Normal): DJF Rainfall



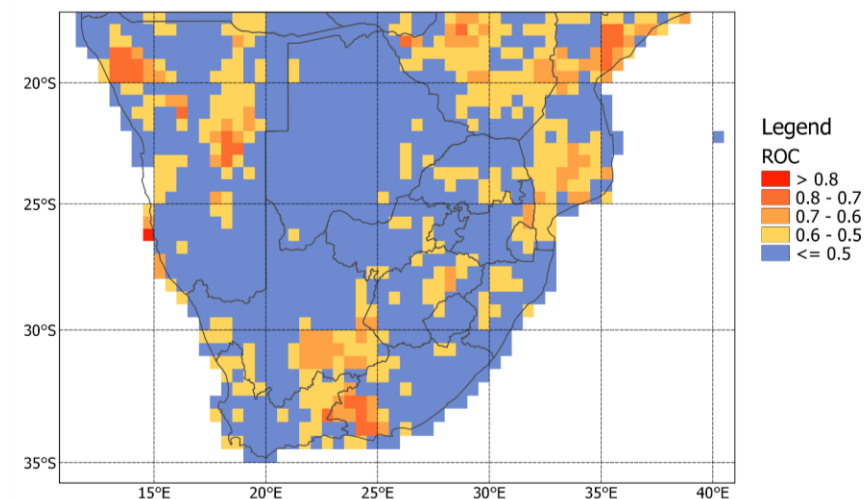
ROC Area (Below-Normal): DJF Rainfall



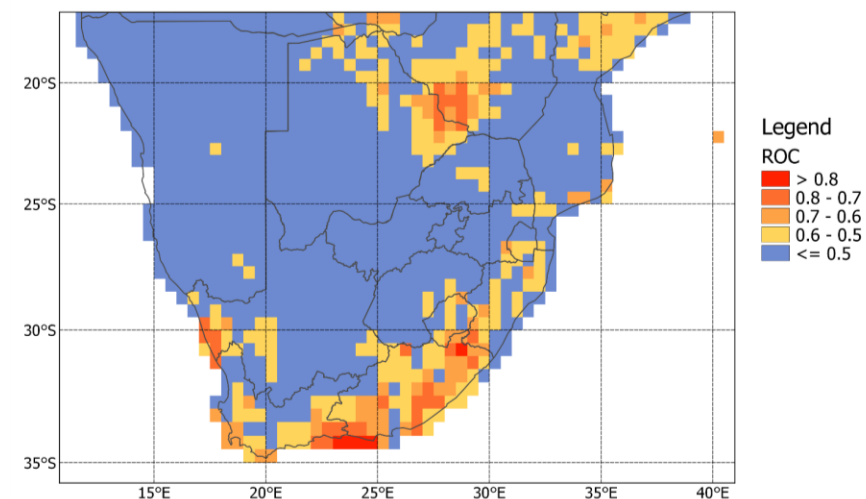
JFM 2022 Rainfall; ICs: Nov



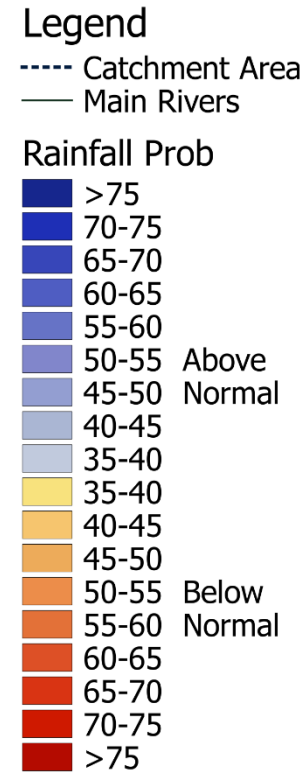
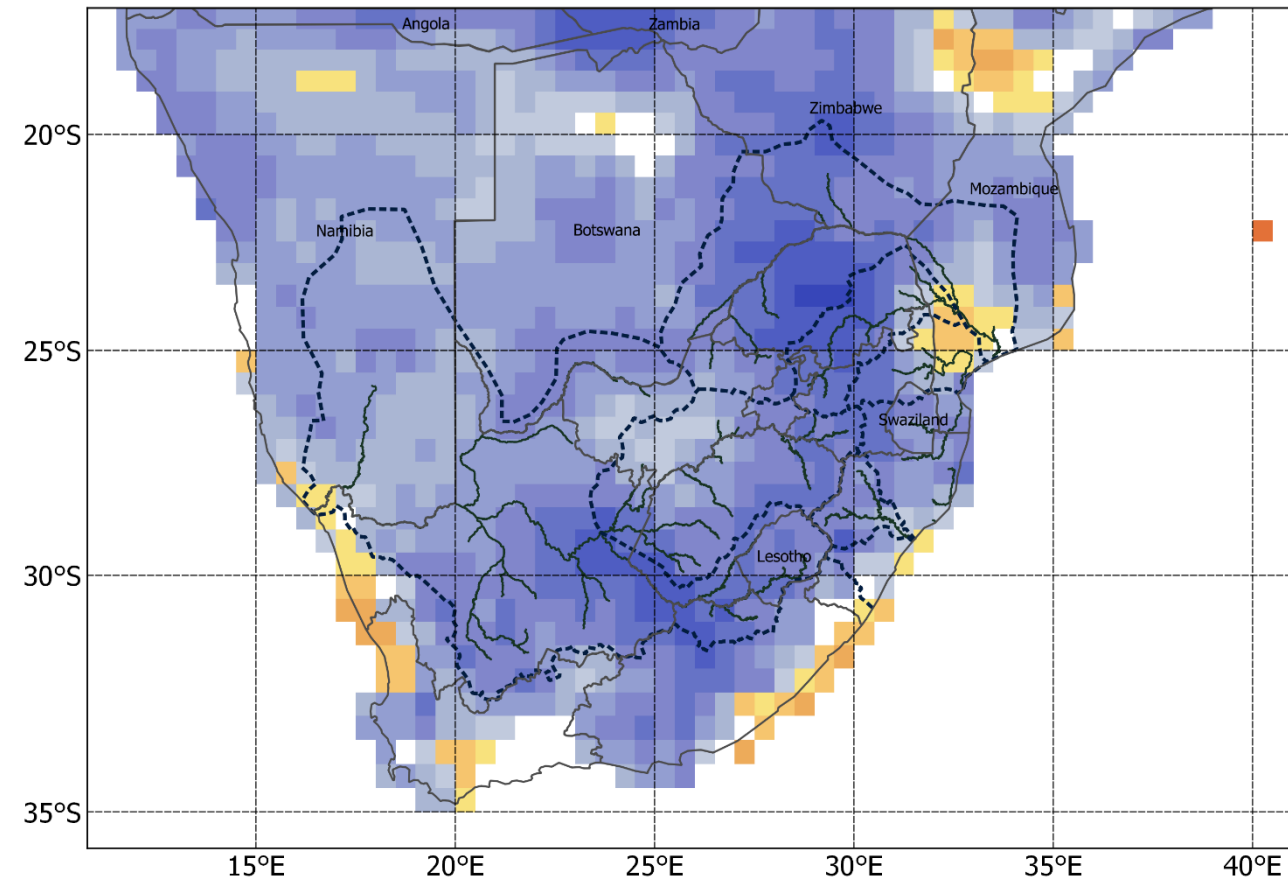
ROC Area (Above-Normal): JFM Rainfall



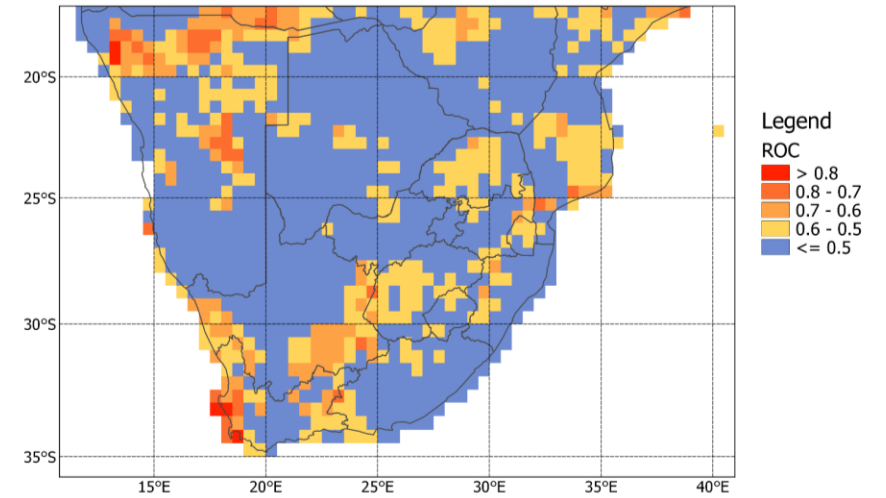
ROC Area (Below-Normal): JFM Rainfall



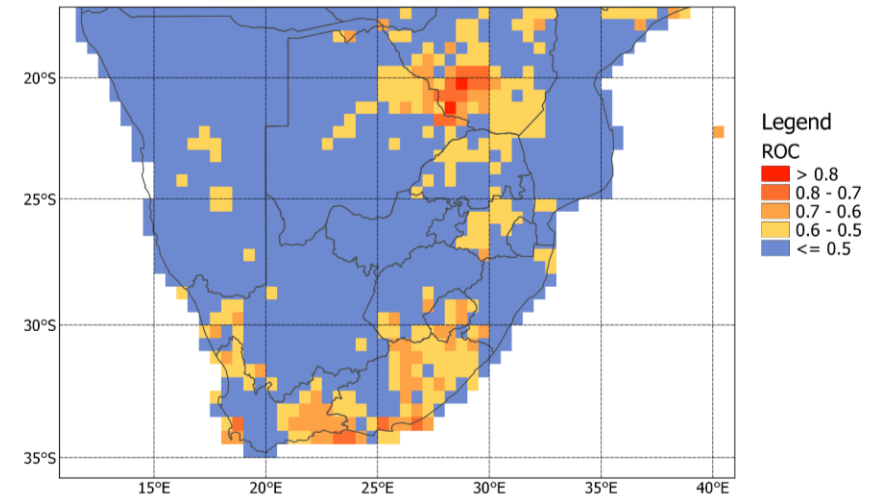
FMA 2022 Rainfall; ICs: Nov



ROC Area (Above-Normal): FMA Rainfall



ROC Area (Below-Normal): FMA Rainfall

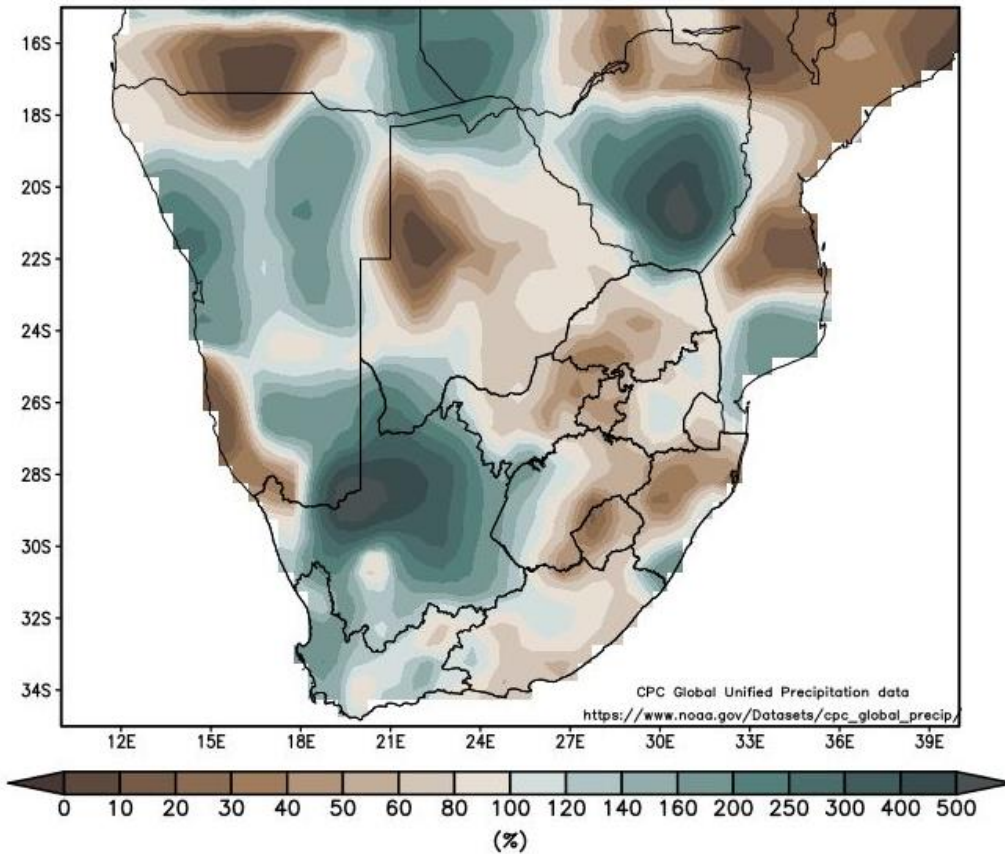


Round-up: SADC Rainfall

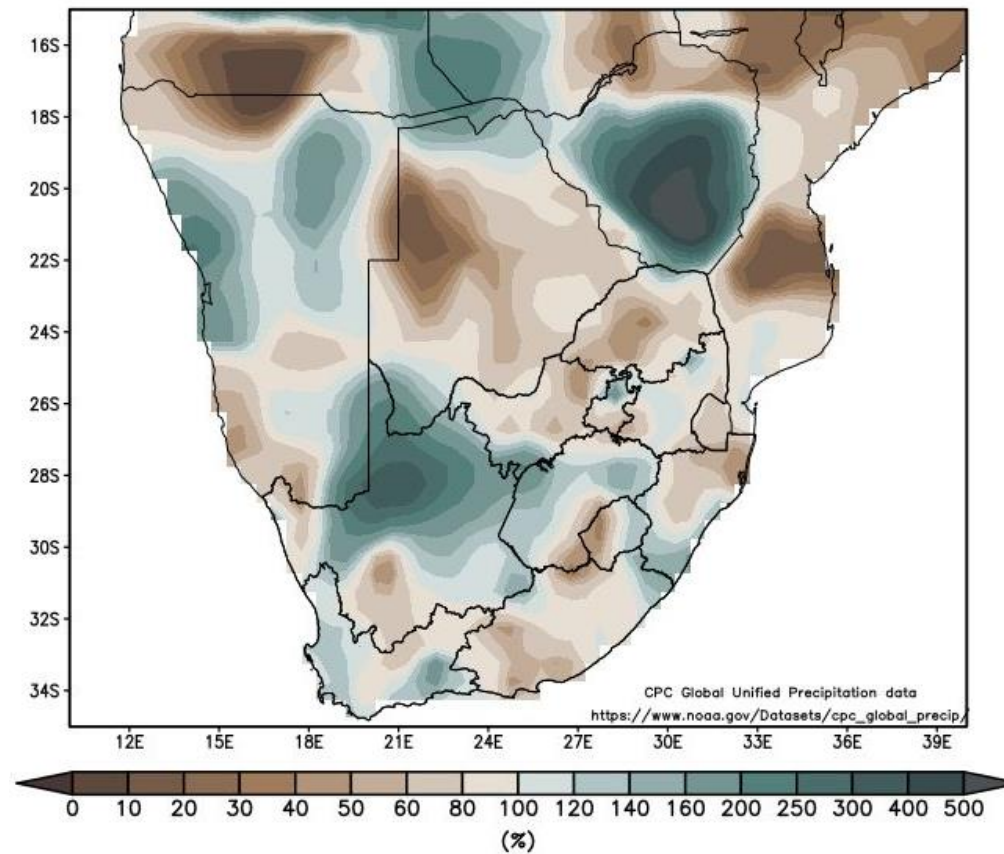
- Predominantly favourable rains are expected over the larger part of the forecast region, except for the Southwestern and Southern Cape, and the northwest and northeast of the SADC region at first
- Take note of the low forecast skill found over large areas during the second half of the forecast period

Observed SADC Rainfall

Rainfall (% of normal): October 2021
October long-term mean: 1981–2010



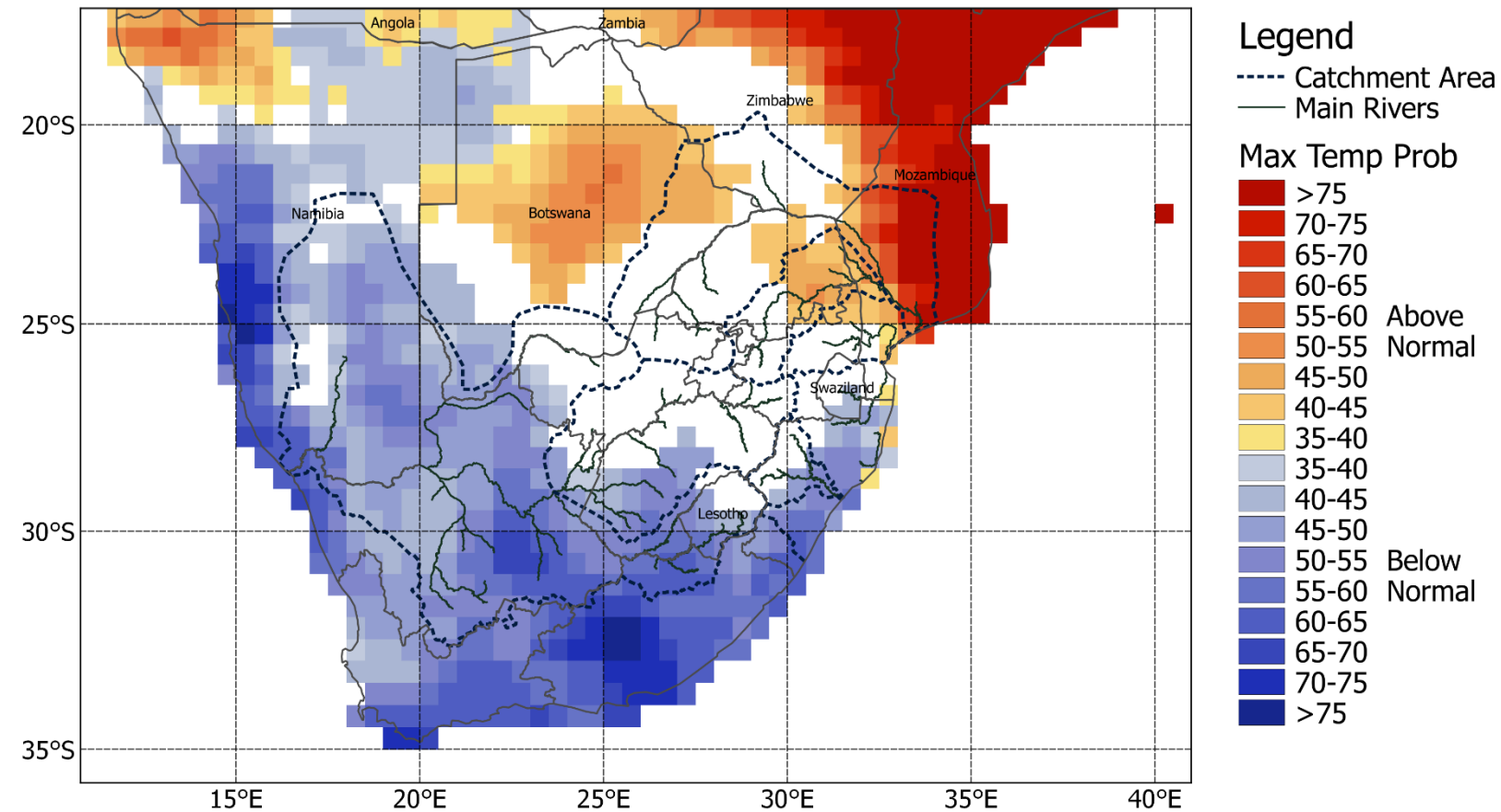
Rainfall (% of normal): ASO 2021
Relative to ASO 1981–2010 rainfall



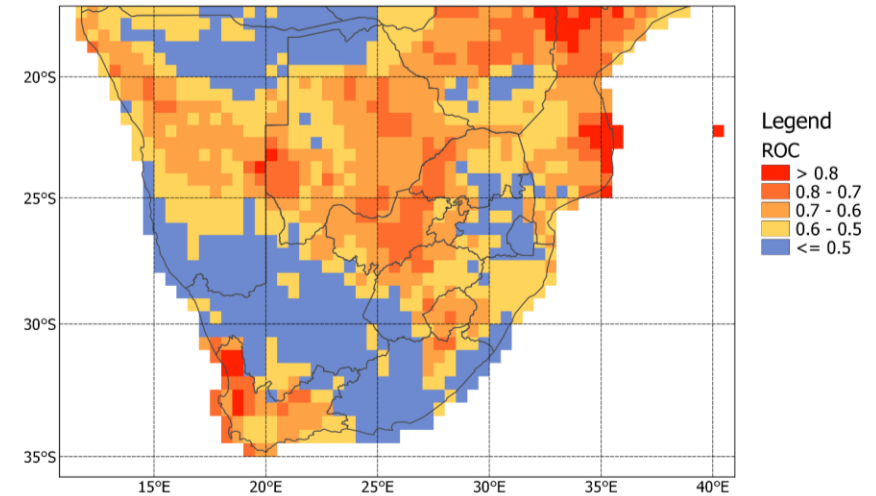
Recorded rainfall for October and the Aug-Sep-Oct season show below-normal rainfall over the brown areas and above-normal rainfall over the green areas

Maps prepared by Dr. Christien Engelbrecht

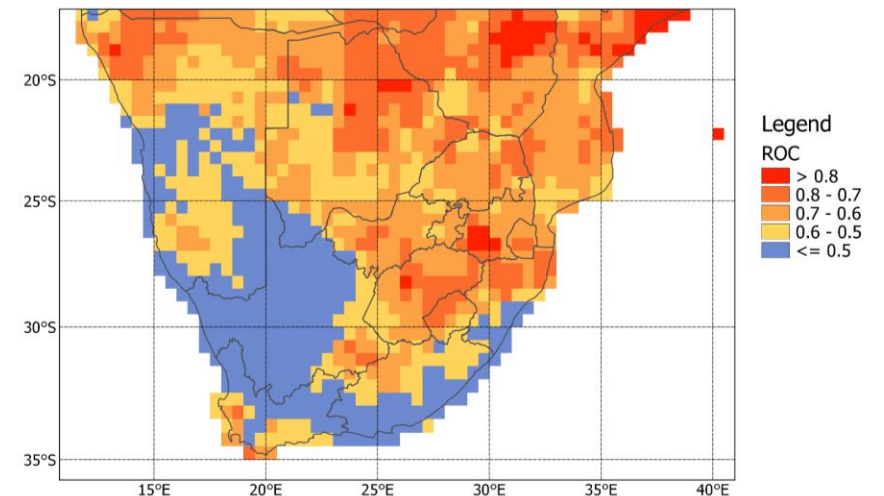
NDJ 2021/22 Max Temp; ICs: Nov



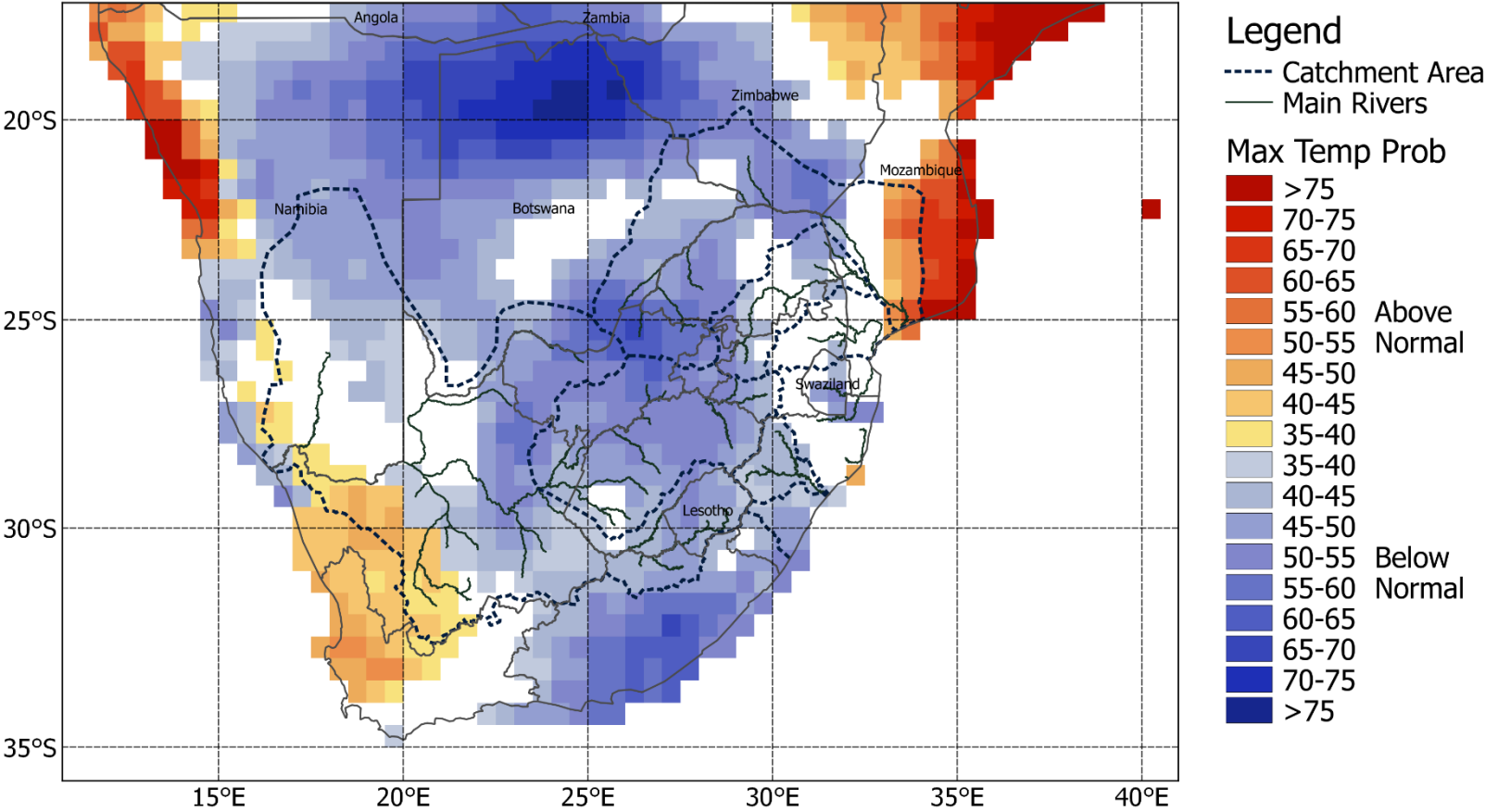
ROC Area (Above-Normal): NDJ Max Temp



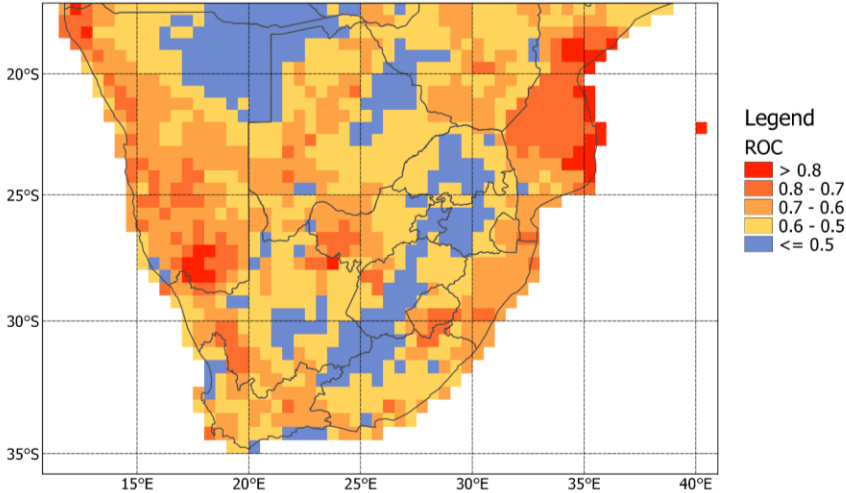
ROC Area (Below-Normal): NDJ Max Temp



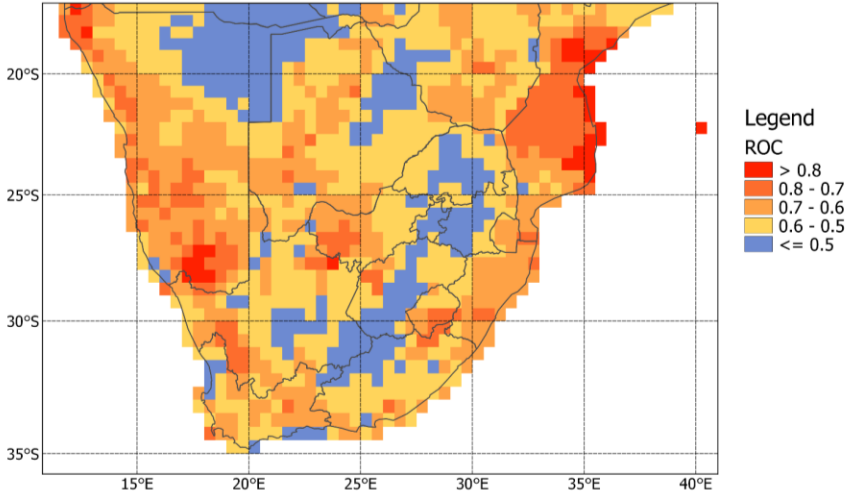
DJF 2021/22 Max Temp; ICs: Nov



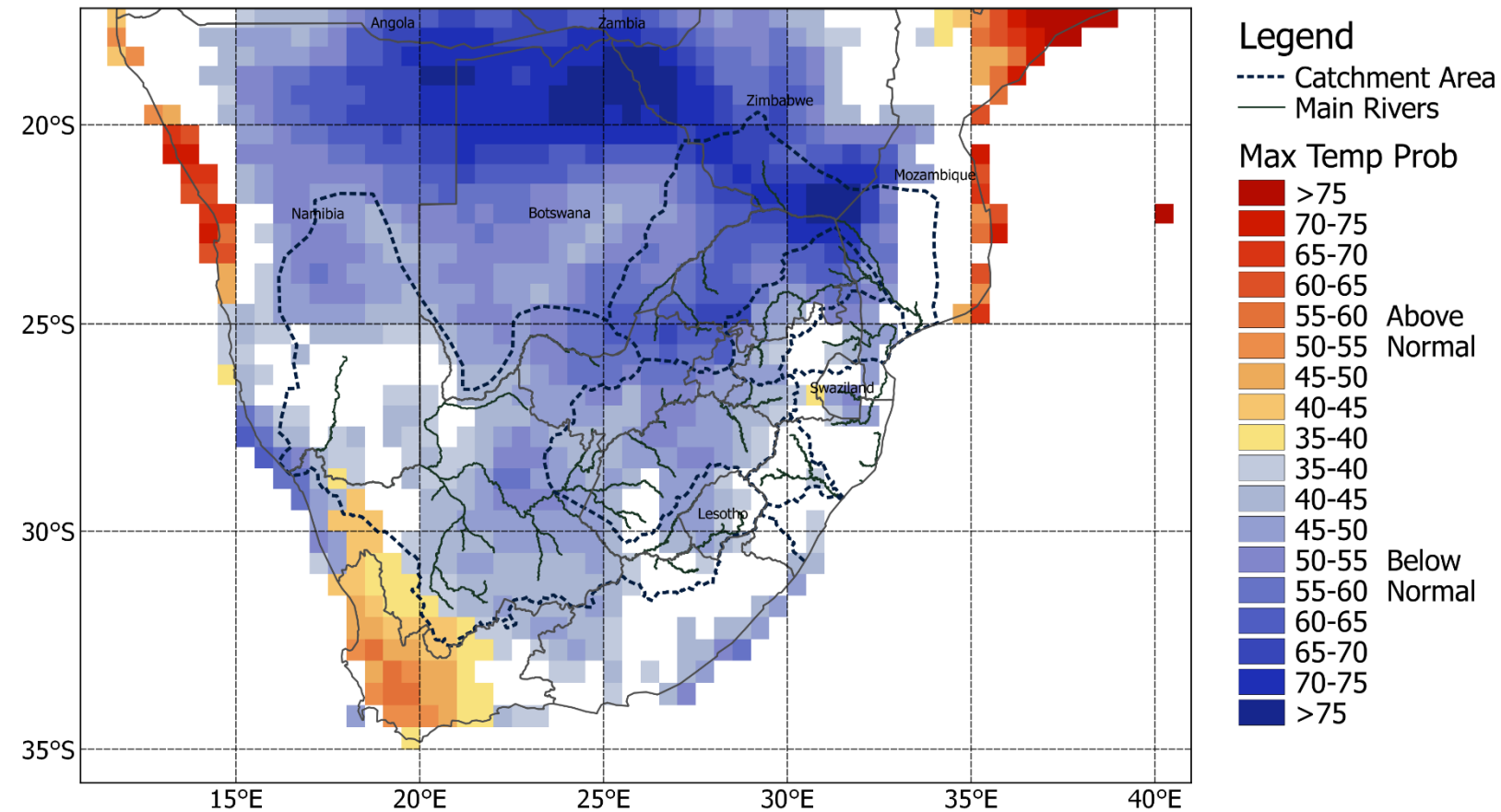
ROC Area (Above-Normal): DJF Max Temp



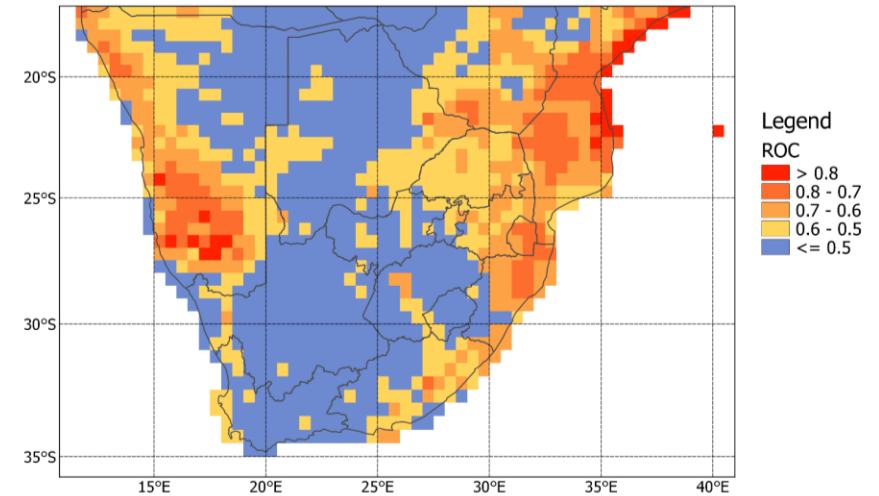
ROC Area (Above-Normal): DJF Max Temp



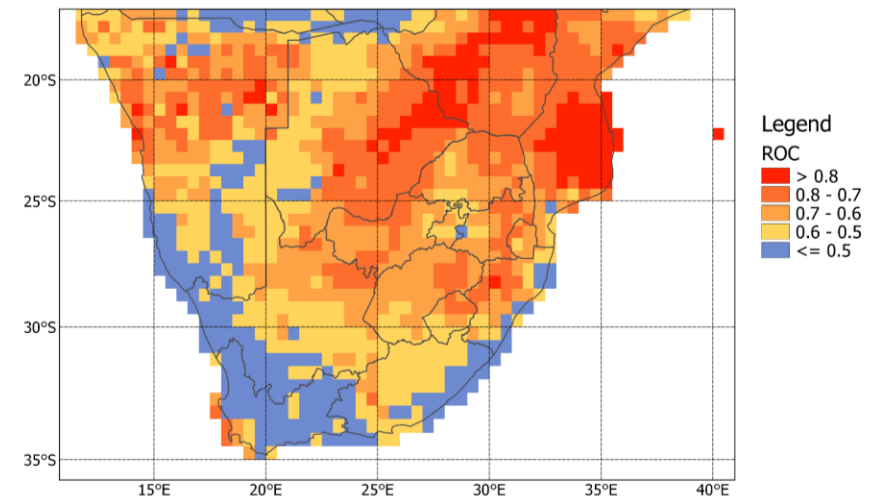
JFM 2022 Max Temp; ICs: Nov



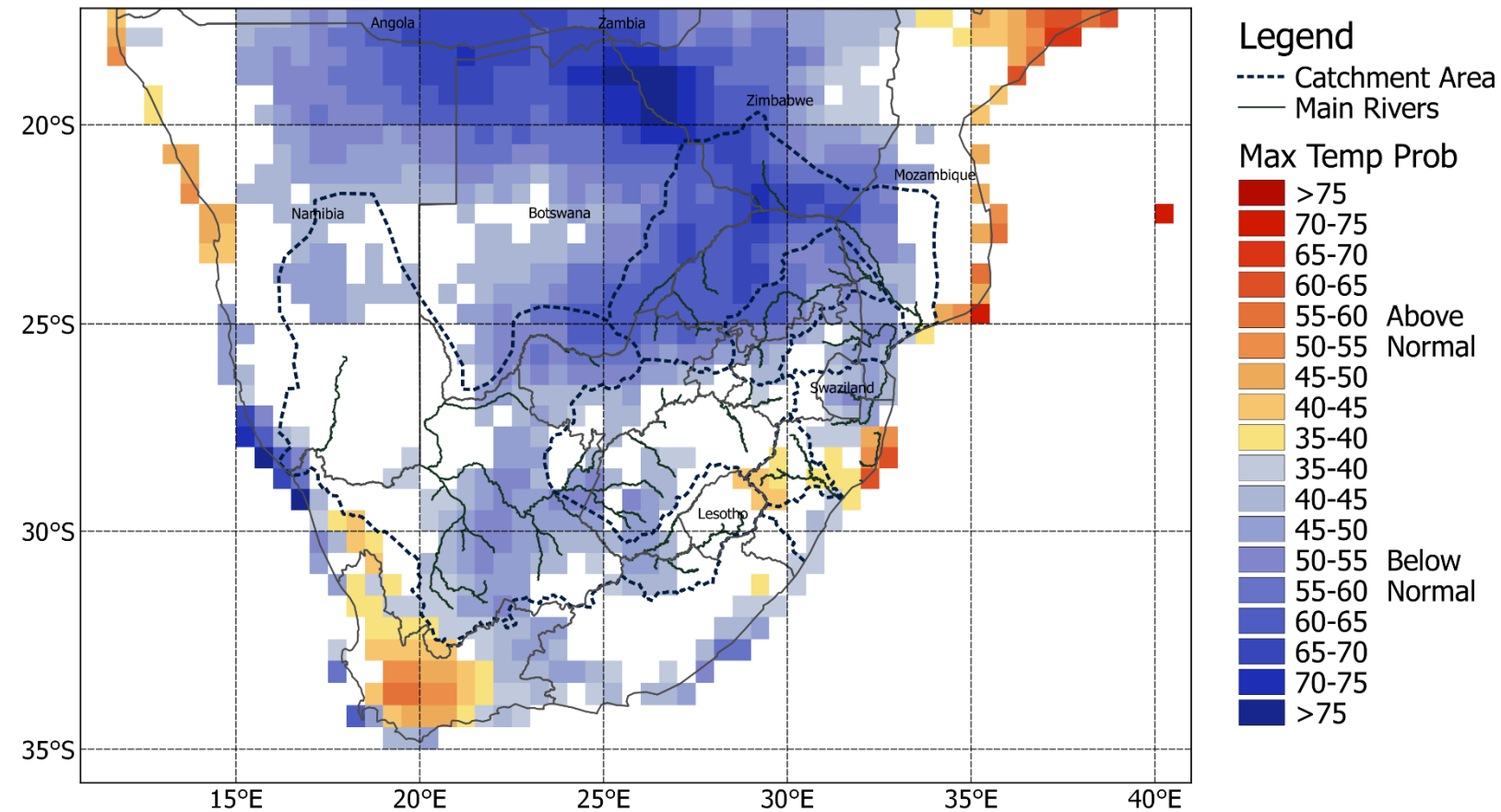
ROC Area (Above-Normal): JFM Max Temp



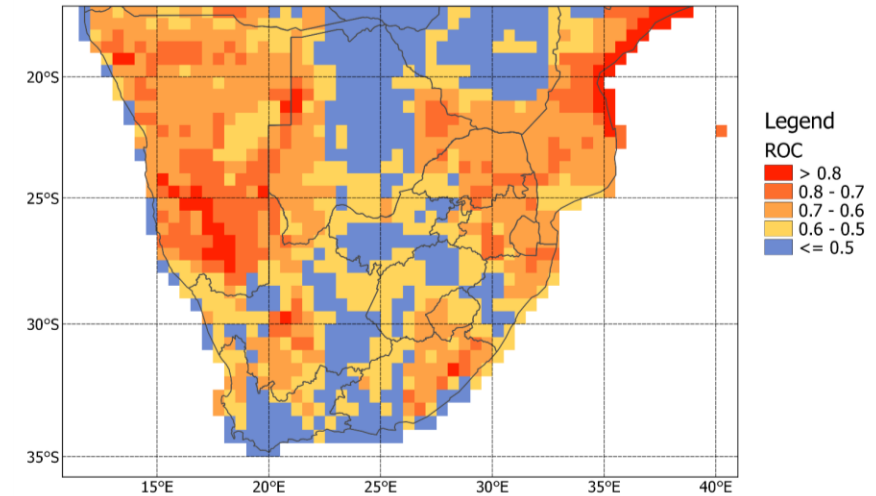
ROC Area (Below-Normal): JFM Max Temp



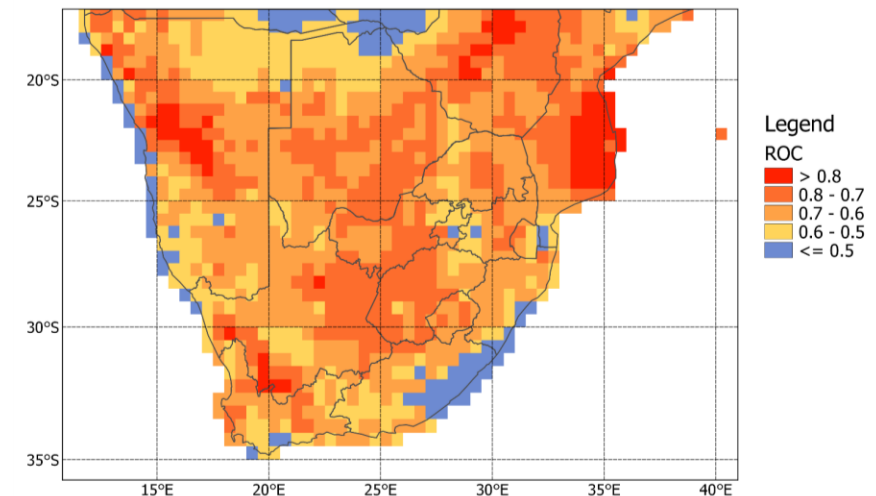
FMA 2022 Max Temp; ICs: Nov



ROC Area (Above-Normal): FMA Max Temp



ROC Area (Below-Normal): FMA Max Temp



Round-up: SADC Max Temp

- Cooler than normal maximum temperatures are expected from mid-summer to the end of the forecast period in the Central, Eastern and Northern parts
- Higher than normal maximum are expected over the western parts, including the SW Cape, and the far northeastern parts

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

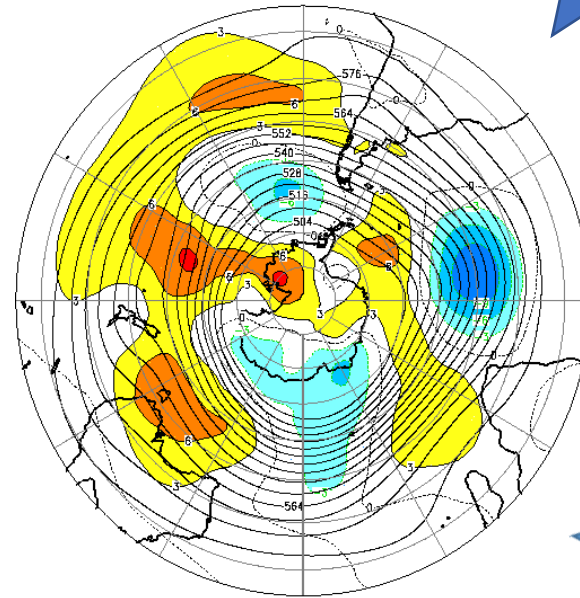
1. Bapsfontein end-of-season-yield three-category probabilistic forecast for 2022
2. Probabilistic three-category rainfall forecast for the farm of Robbie Kingsley for Dec-Jan-Feb 2021/22
3. Probabilistic three-category malaria forecast for Limpopo for Dec-Jan-Feb 2021/22
4. Probability of exceedance Dec-Jan-Feb 2021/22 inflow forecast for Lake Kariba, Zambia/Zimbabwe
5. Probabilistic rainfall forecast for Jan-Feb-Mar 2022 for the farm Buschbrunnen near Grootfontein, Namibia
6. Probability of exceedance Jan-Feb-Mar 2022 downstream flow forecasts for Vaal Dam

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



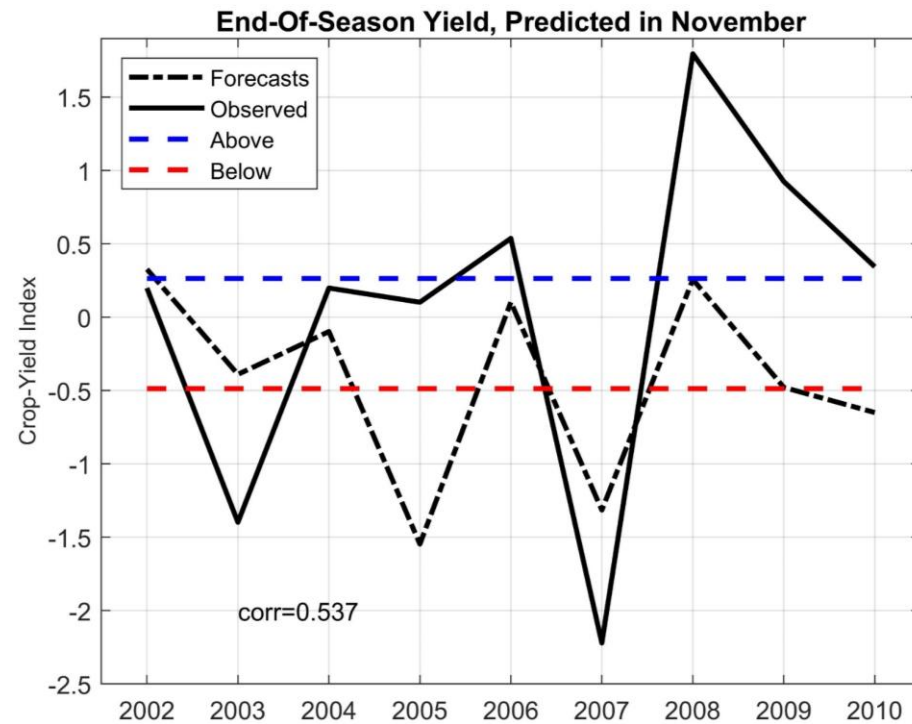
Malaria



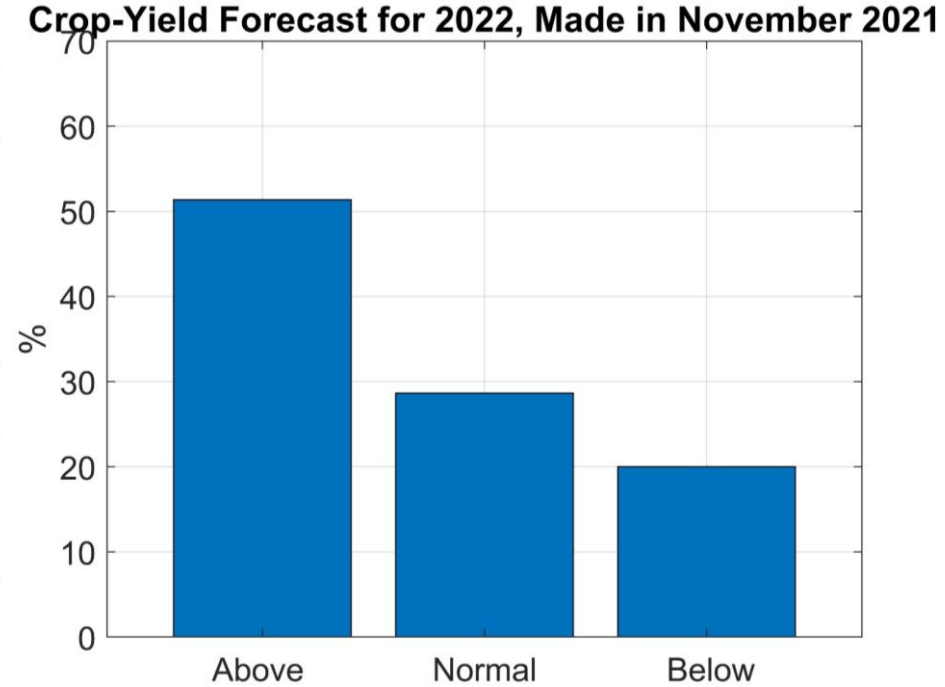
Dry-land crop yield

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

Landman et al. (2019)

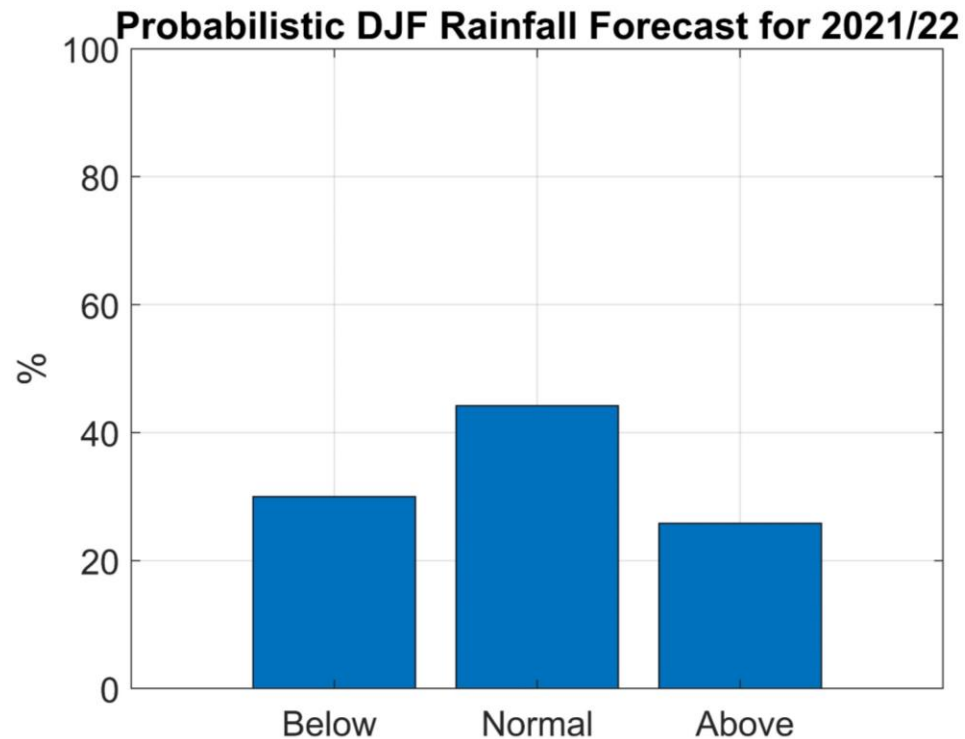
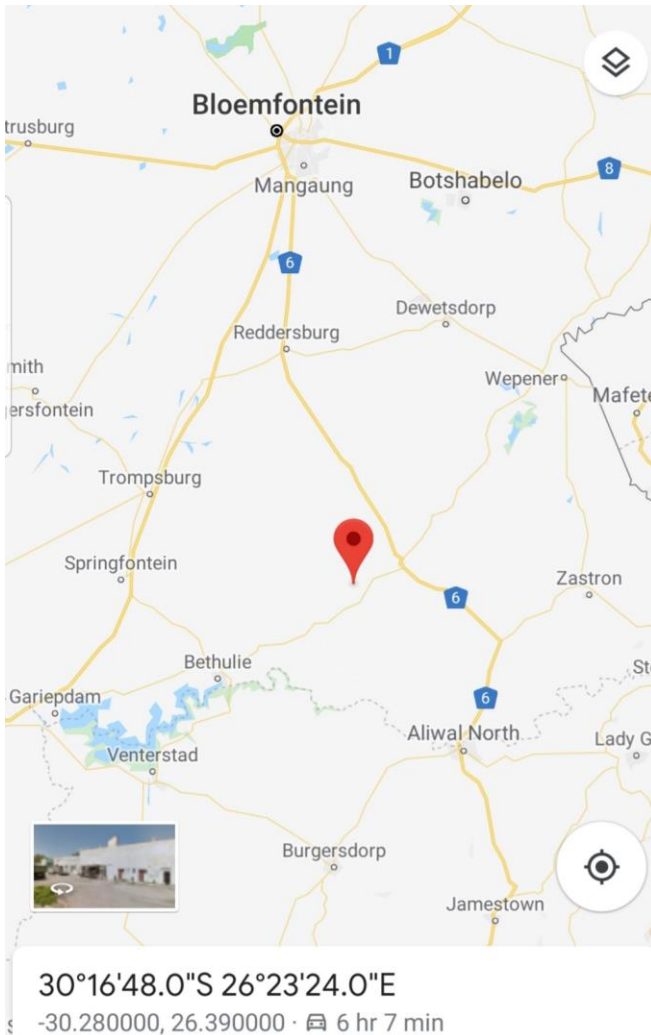
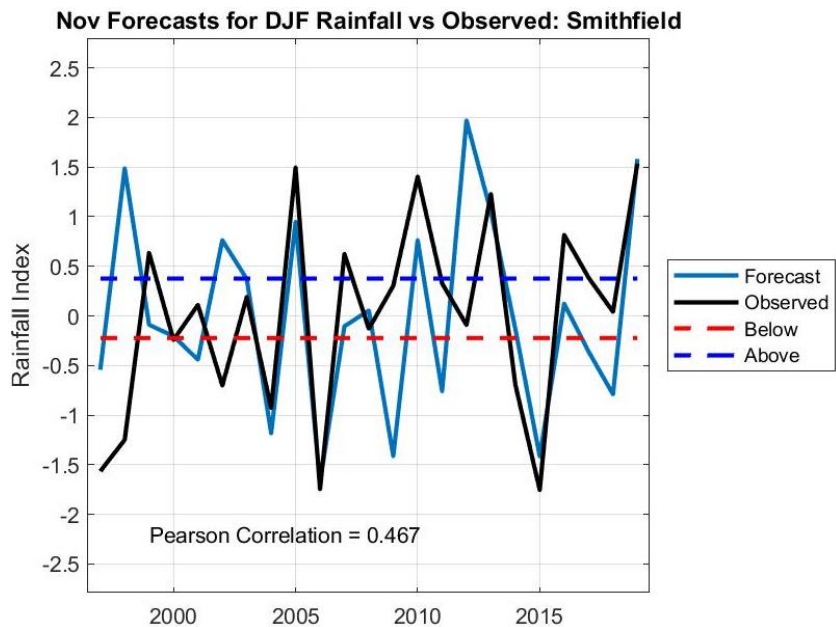
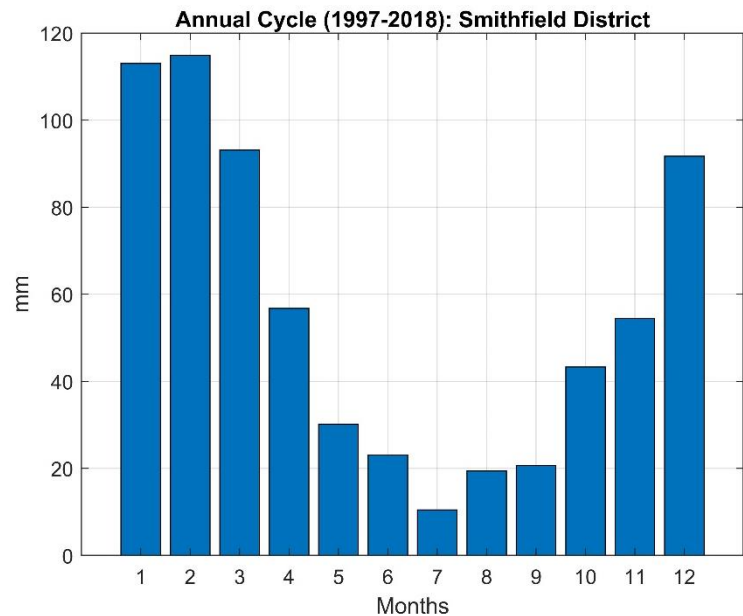


Re-forecasts for end-of-season crop yields



Dec-Jan-Feb 2021/22 rainfall forecast for farm in the Smithfield district (see map). Rainfall data provided by the farmer, Mr. Robbie Kingsley

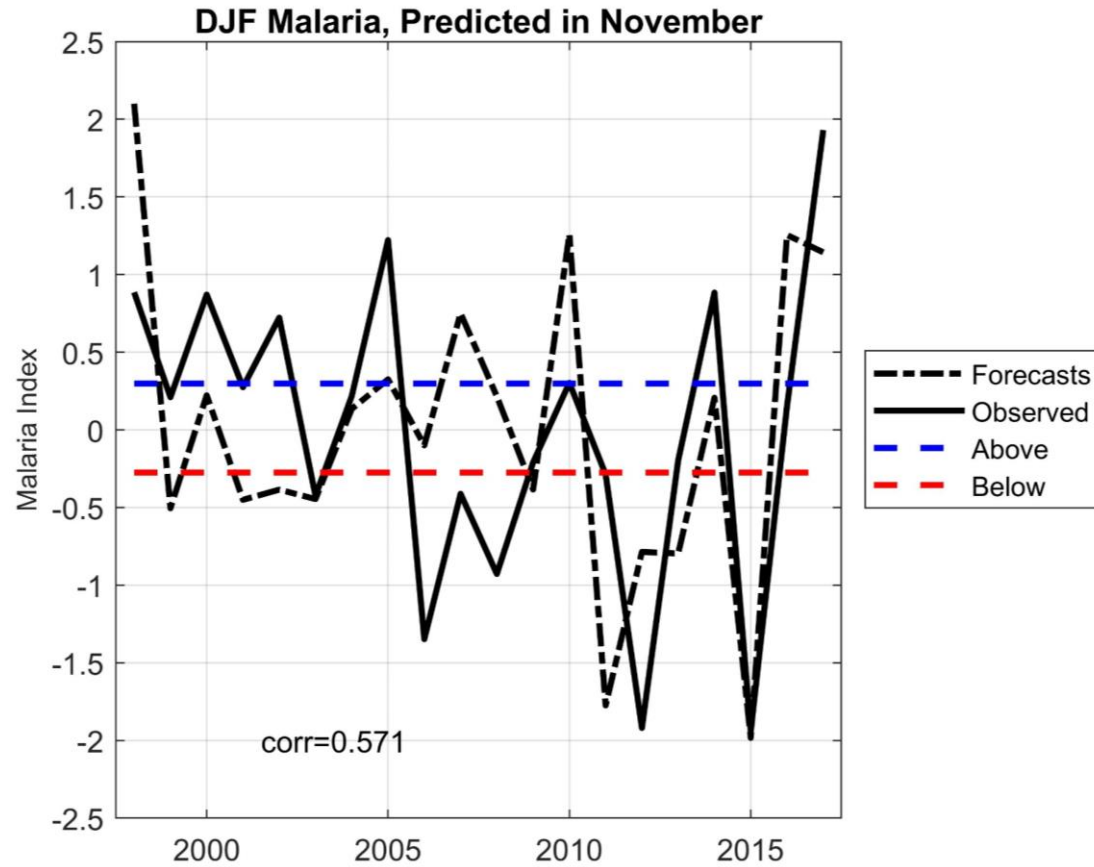
Landman et al. (2020a)



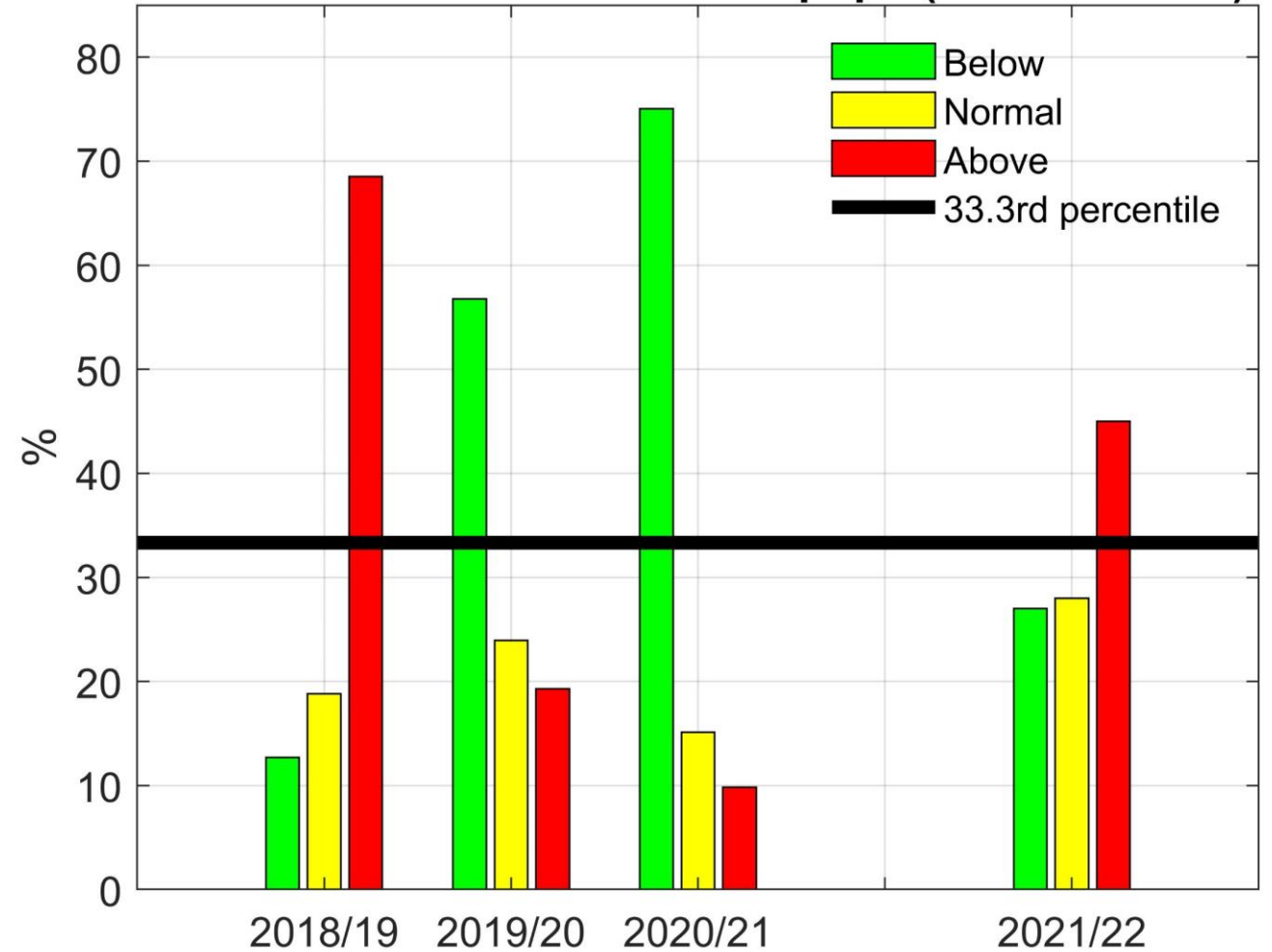
Malaria forecast

Landman et al. (2020b)

Hindcasts



DJF Malaria Forecasts for Limpopo (Issued in Nov)

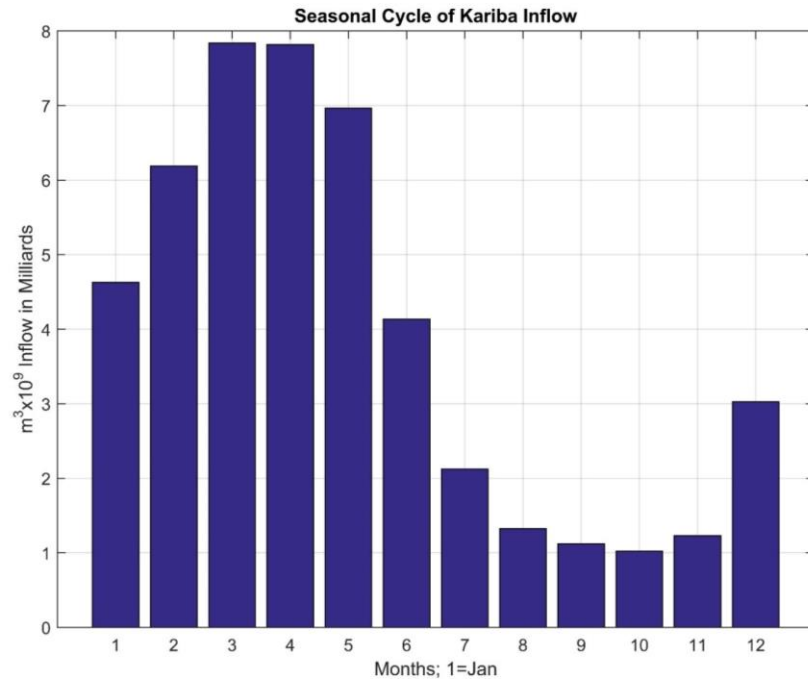


Retrospective

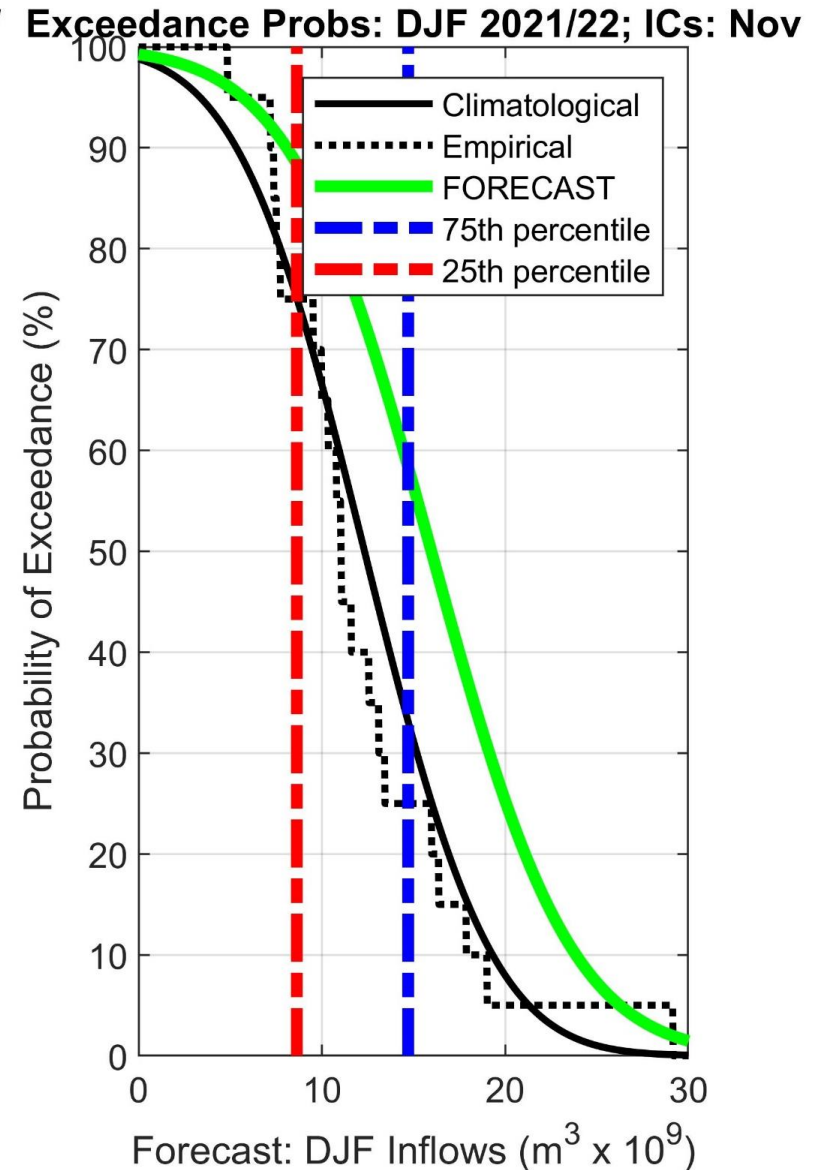
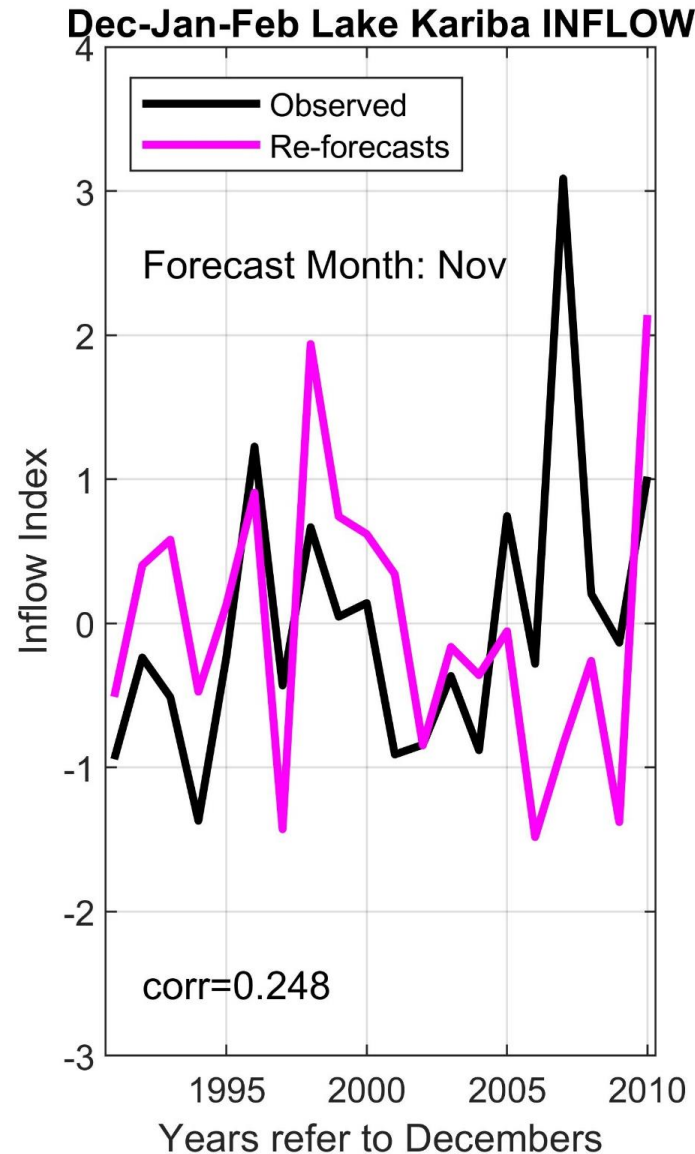
REAL-TIME

Inflow forecast for Lake Kariba: onset season of DJF

Muchuru et al. (2016)

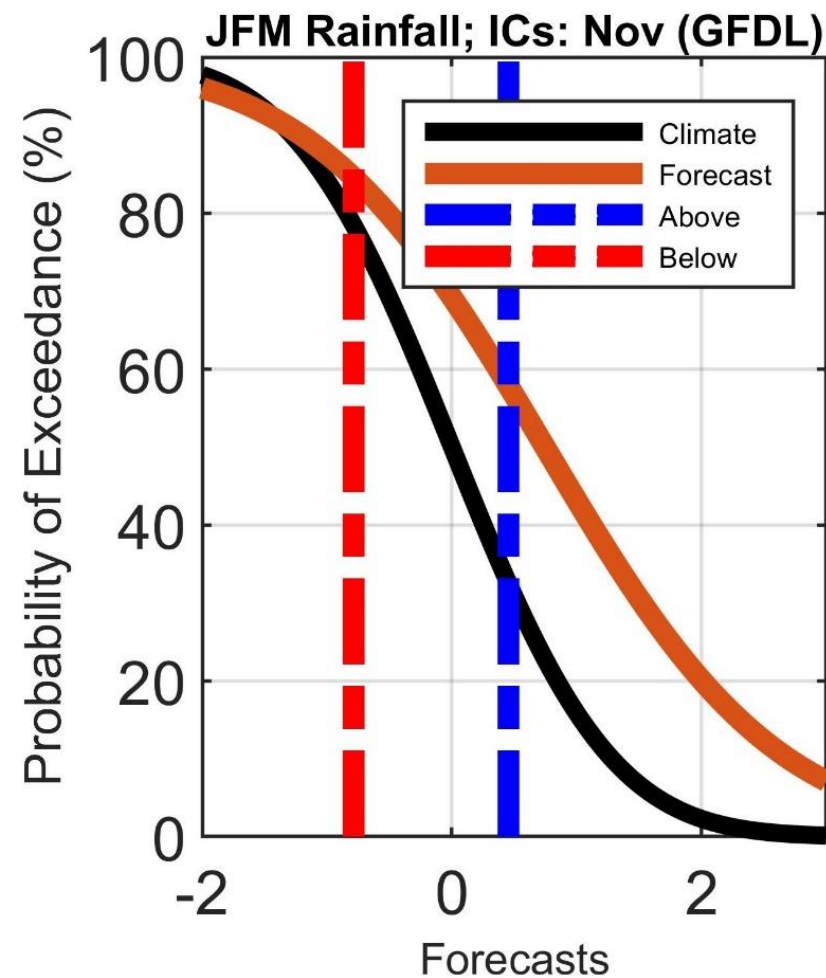
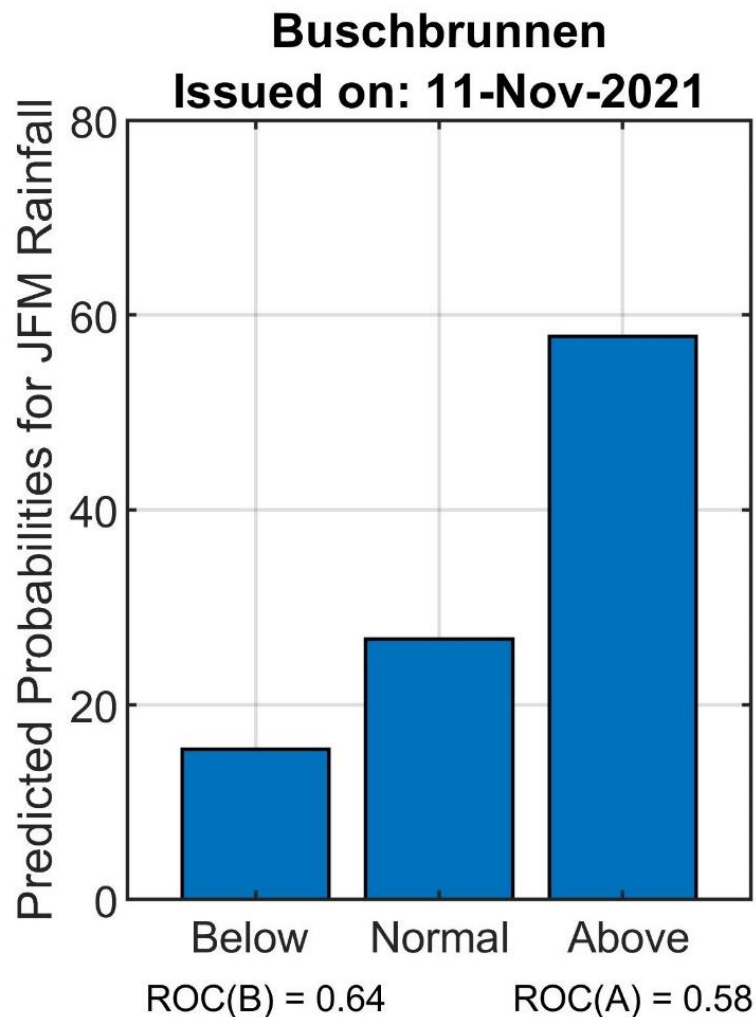
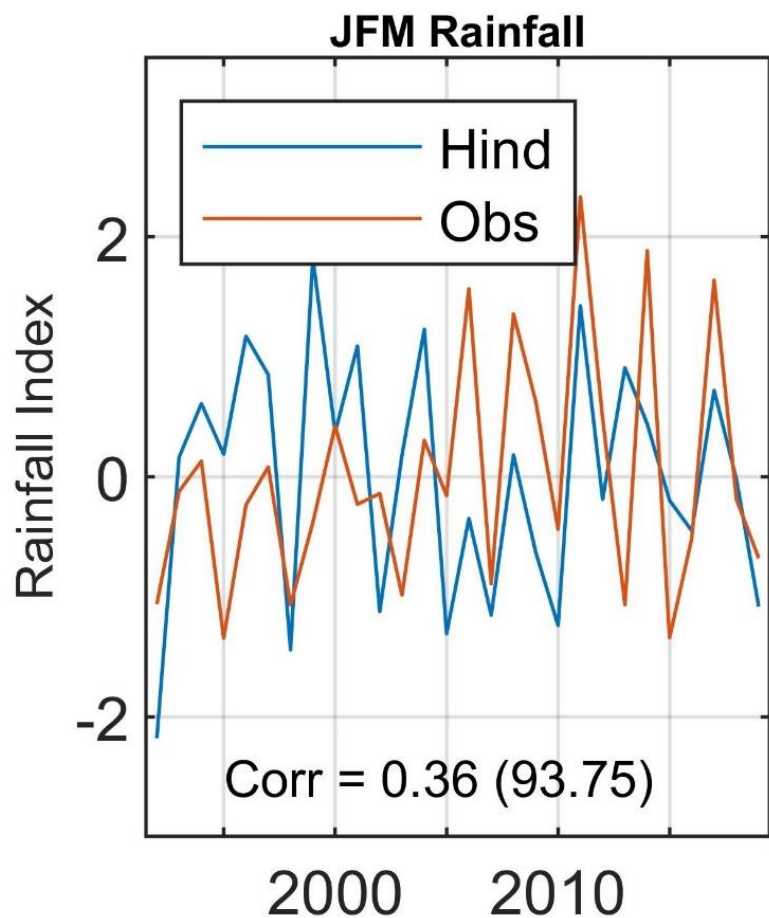


For the forecast on the far right: The black curve represents the climatological probability distribution of DJF inflows, and the green curve represents the predicted probability distribution for the coming DJF season. The vertical dashed lines represent the category thresholds. The easiest way to interpret 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.



JFM rainfall forecast for the farm Buschbrunnen near Grootfontein, Namibia

Landman et al. (2016)



Current Project, administered by the WRC: *RainSolutions*

Vaal Dam Downstream Flow Predictions

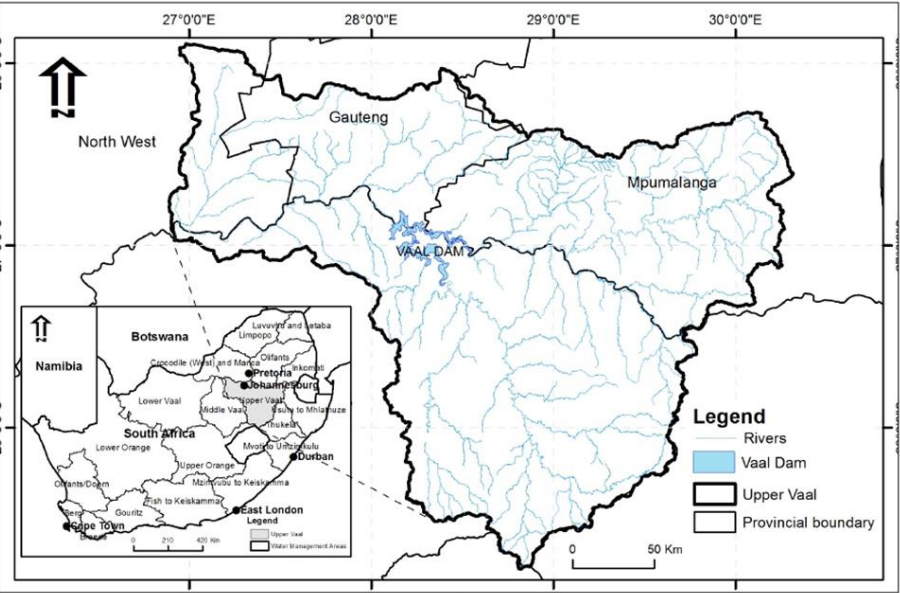
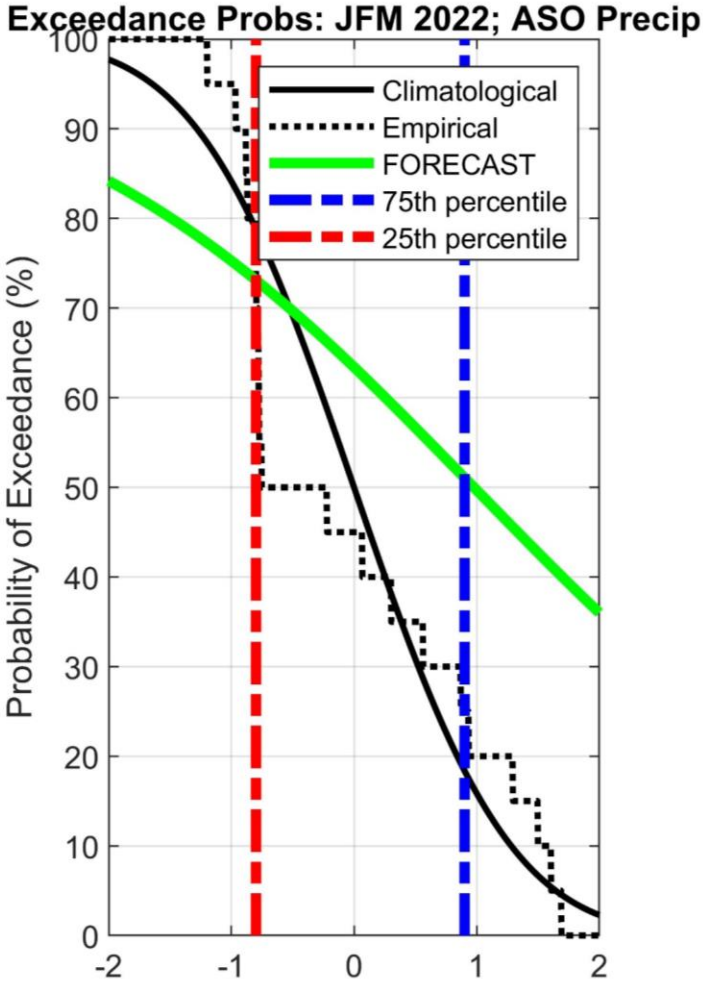
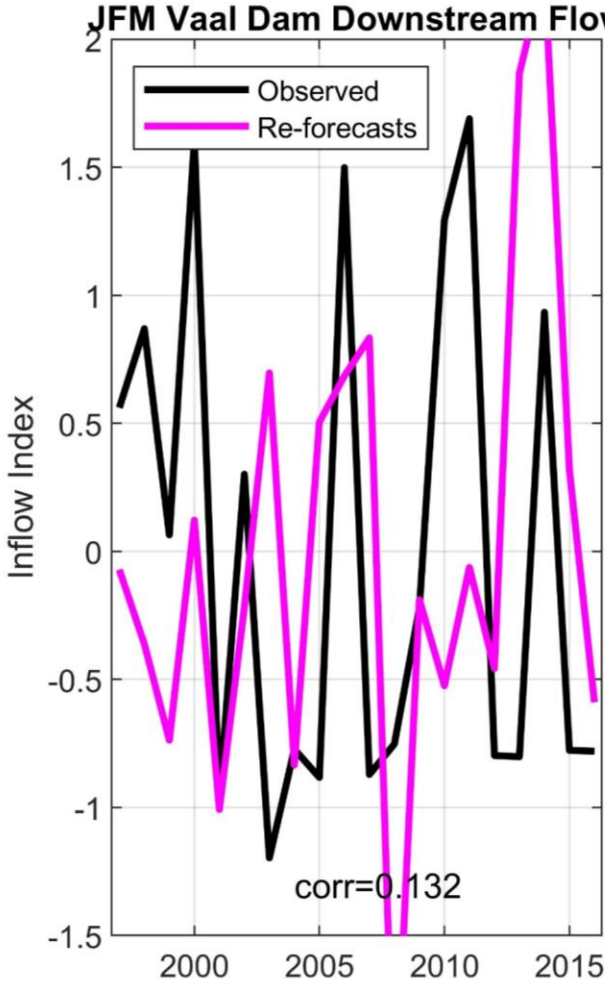


Figure 1
Location of Vaal Dam in Upper Vaal Water Management Area in South Africa (Modified from DWA, 2010)



Round-up: Tailored products

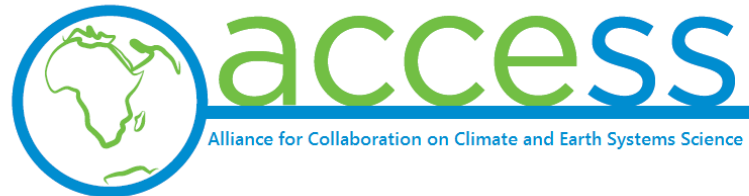
- The tailored forecasts shown are in agreement with the rainfall forecasts for Southern Africa
 - Increased farm rainfall and crop yield
 - Increased malaria occurrence over Limpopo
 - Increased inflow into Lake Kariba
 - Increased outflow from the Vaal Dam

References

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- The National Research Foundation through the Incentive Funding for Rated Researchers (since 2017)
- 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 2021)



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.