Seasonal forecasts

presented by:

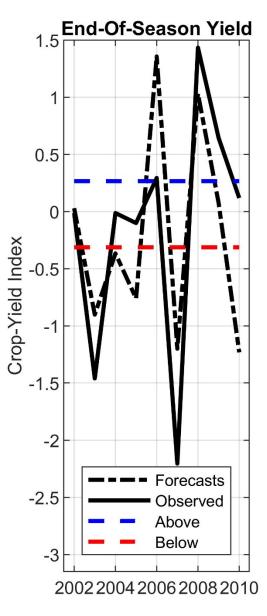


Seasonal Forecast

Worx

Latest Update: 9 May 2020

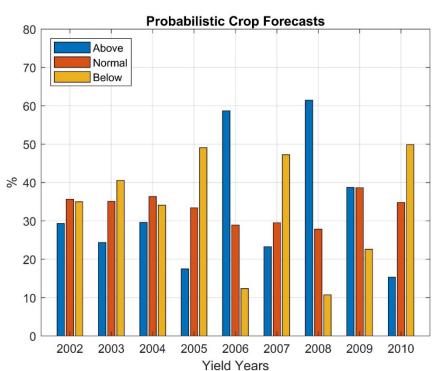
Are you a farmer who wants to make use of science-based seasonal predictions for your farm? If you are interested to be part of an initiative at the University of Pretoria that involves the development of seasonal forecast systems for farms, specifically tailored to farmers' needs, please send an email to WALandman1981@gmail.com

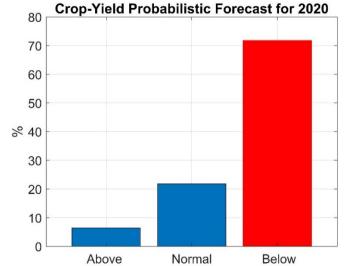


Have a look at this example of end-of-season crop yield forecasts for a farm near Bapsfontein.

The farmer provided several decades of crop-yield data and these data were subsequently used

to create a crop forecast model specific to the farm





Above is the crop-yield forecast for the coming season. The forecast is for enhanced probabilities of below-normal (low) crop yield for the farm. The farmer may be able with support to use this forecast information to plan for the coming season

On the left are time series of forecast and observed crop yields at the time of harvest for the years indicated. Next to the time series are probabilistic forecasts over the same 9-years for below- (low yields), near- (about average) and above-normal (high yields). For example, in 2008 the forecast and observed index values are high and positive (figure on the left), and the highest predicted probability is for above-normal yield (figure in the middle).

Share your data and become part of this initiative

- 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://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).
- The UP seasonal forecast is a research product. For the <u>official</u> seasonal forecast for South Africa, visit the South African Weather Service website at:

Weather Service

http://www.weathersa.co.za/images/data/longrange/gfcsa/scw.pdf

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°x0.5°), 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.

Forecasts are probabilistic

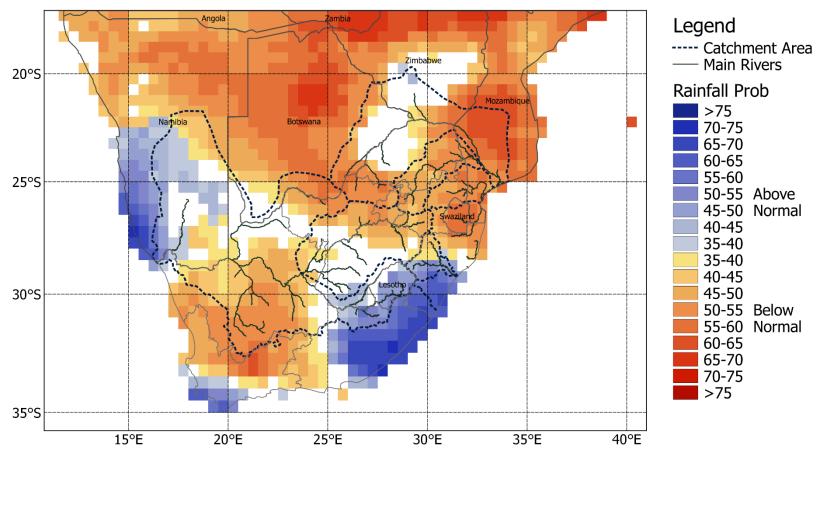
Probabilistic forecasts can help users understand risks and opportunities (forewarned is forearmed) in order to make more informed decisions.

The seasonal rainfall and maximum temperature forecast to follow are probabilities (% chance) of only the most likely outcome for below-, near-, or above-normal (B, N or A). 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.

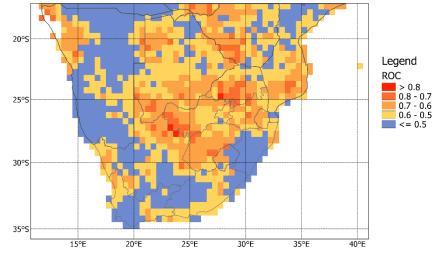
The nature of a probabilistic forecast implies that the less likely outcomes are always possible. In fact, for the probabilistic forecasts to be considered reliable, the less likely outcomes will and must occasionally occur.

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 (B, N or A) as frequently as the forecast implies

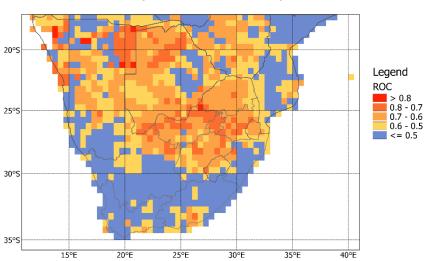
MJJ 2020 Rainfall; ICs: May



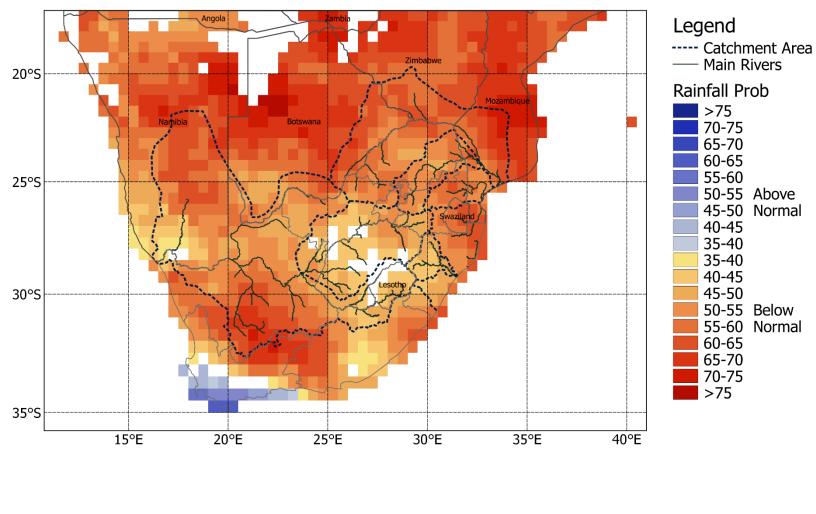
ROC Area (Above-Normal): MJJ Rainfall



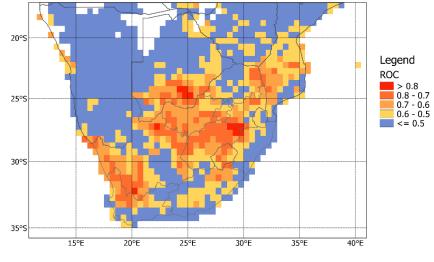
ROC Area (Below-Normal): MJJ Rainfall



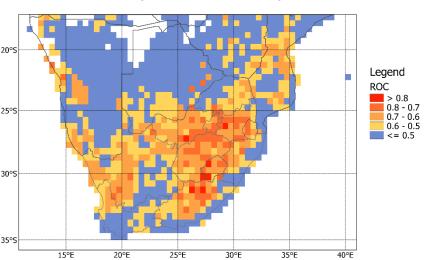
JJA 2020 Rainfall; ICs: May



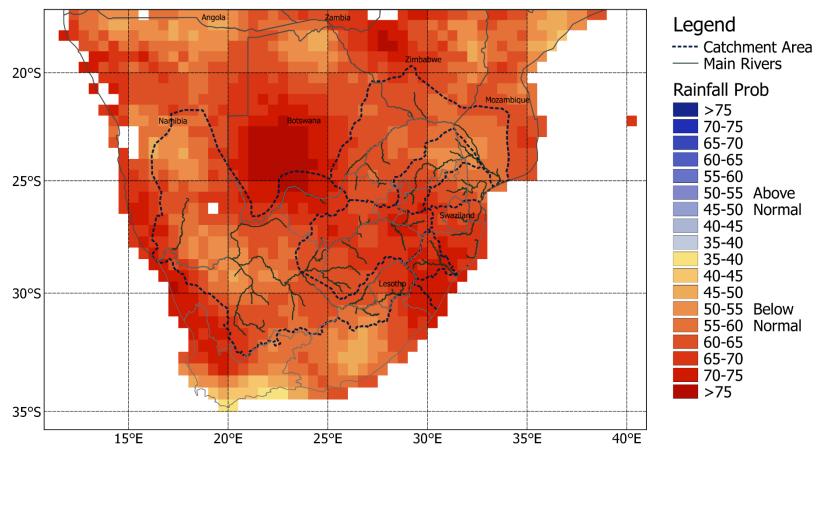
ROC Area (Above-Normal): JJA Rainfall



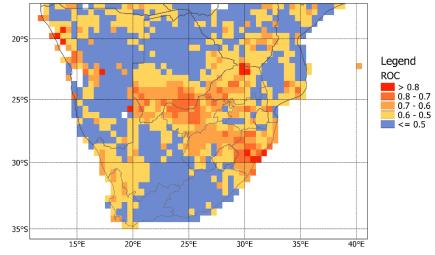
ROC Area (Below-Normal): JJA Rainfall



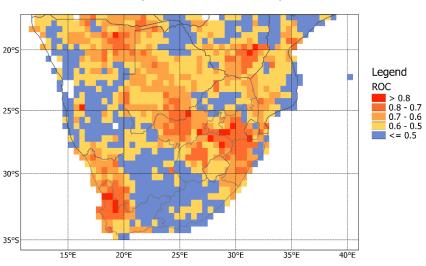
JAS 2020 Rainfall; ICs: May



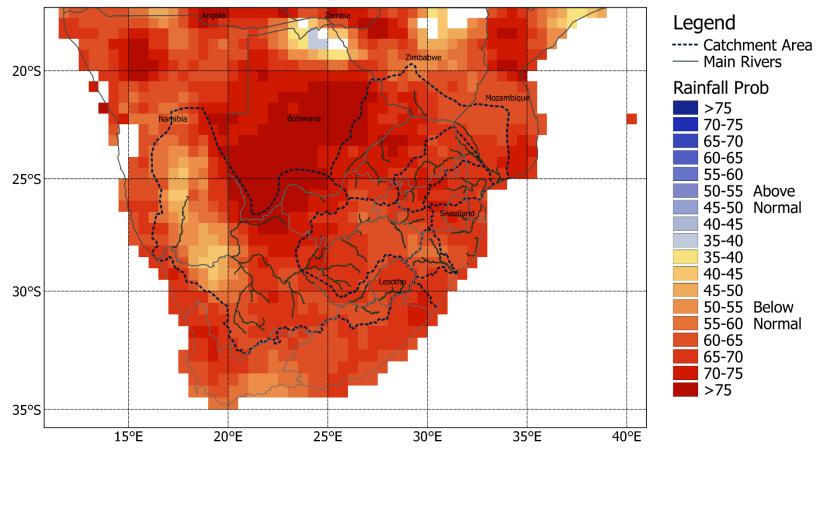
ROC Area (Above-Normal): JAS Rainfall



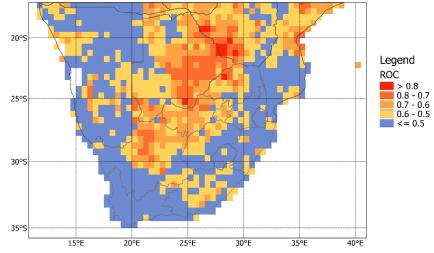
ROC Area (Below-Normal): JAS Rainfall



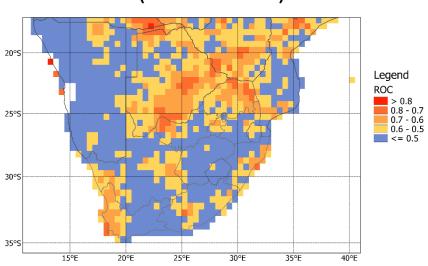
ASO 2020 Rainfall; ICs: May



ROC Area (Above-Normal): ASO Rainfall



ROC Area (Below-Normal): ASO Rainfall

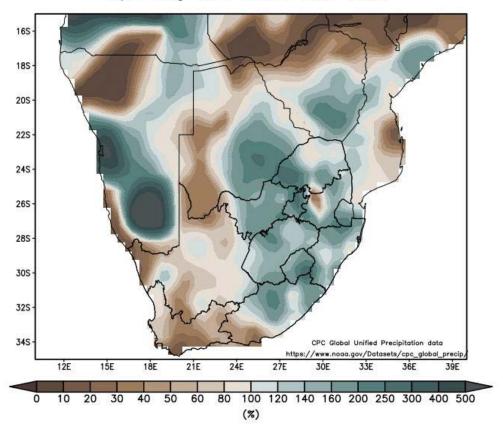


Round-up: SADC Rainfall

 Mostly dry, but with some indication of favourable rainfall outcomes over the SW Cape during mid-winter

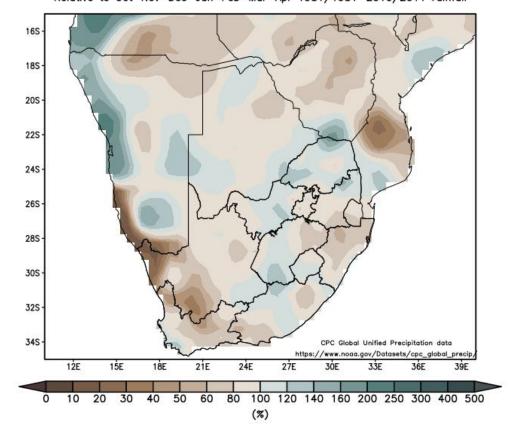
Observed SADC Rainfall

Rainfall (% of normal): April 2020 April long-term mean: 1981-2010

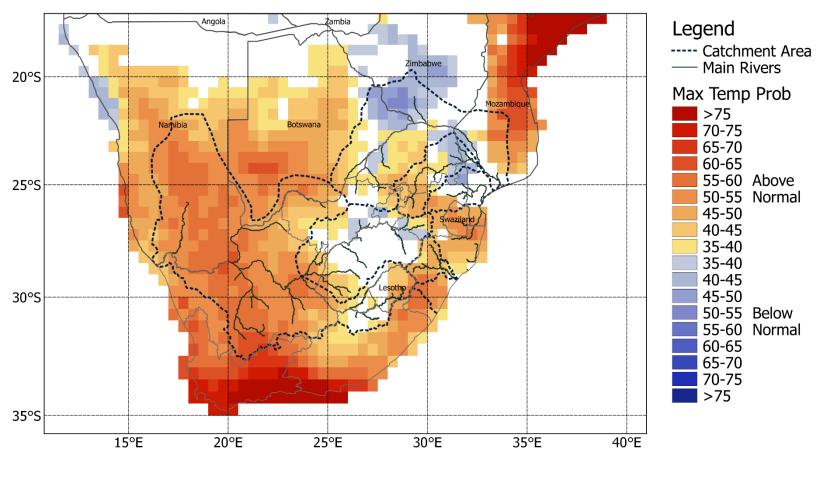


Rainfall (% of normal): Oct-Nov-Dec-Jan-Feb-Mar-Apr 2019/2020

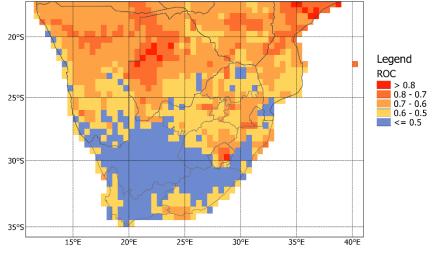
Relative to Oct-Nov-Dec-Jan-Feb-Mar-Apr 1981/1981-2010/2011 rainfall



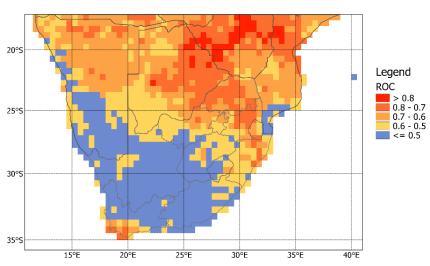
MJJ 2020 Max Temp; ICs: May



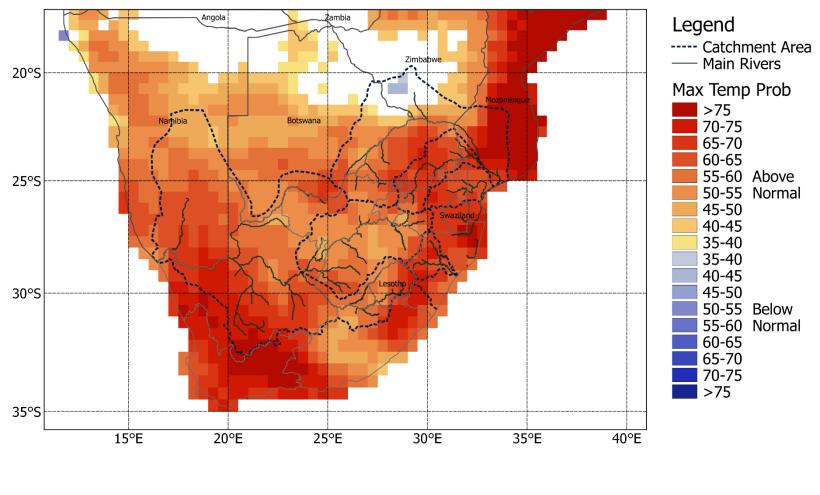
ROC Area (Above-Normal): MJJ Max Temp



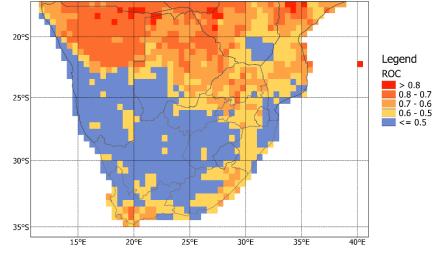
ROC Area (Below-Normal): MJJ Max Temp



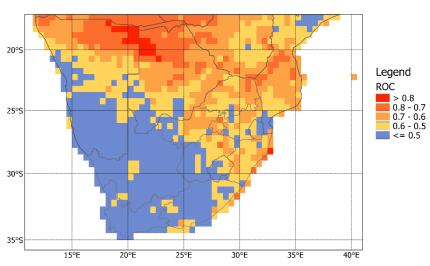
JJA 2020 Max Temp; ICs: May



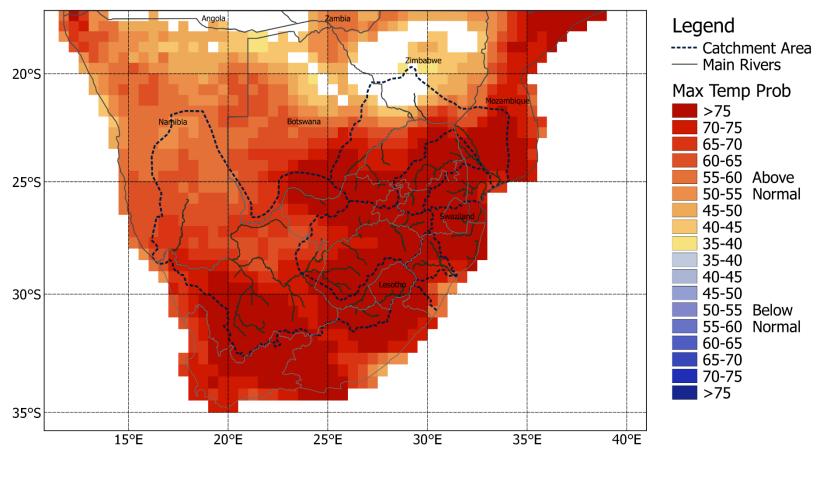
ROC Area (Above-Normal): JJA Max Temp



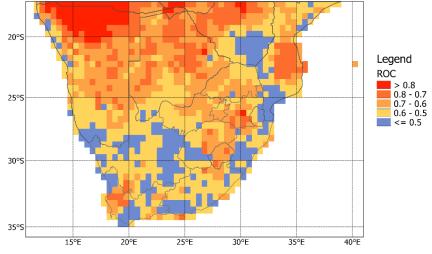
ROC Area (Below-Normal): JJA Max Temp



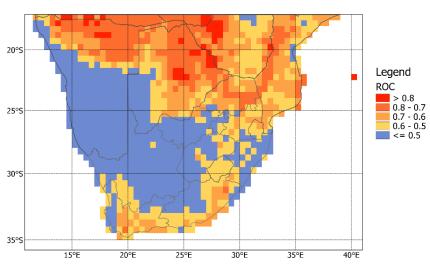
JAS 2020 Max Temp; ICs: May



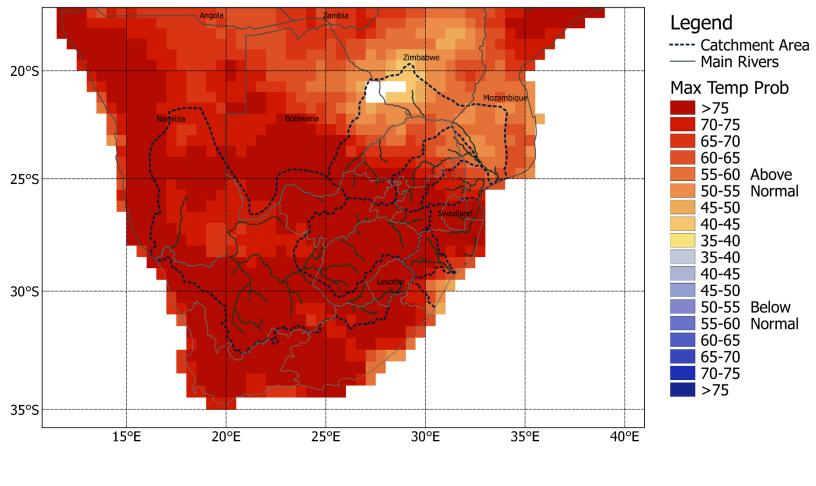
ROC Area (Above-Normal): JAS Max Temp



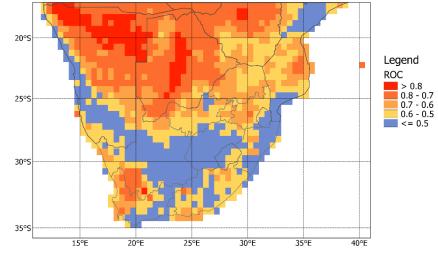
ROC Area (Below-Normal): JAS Max Temp



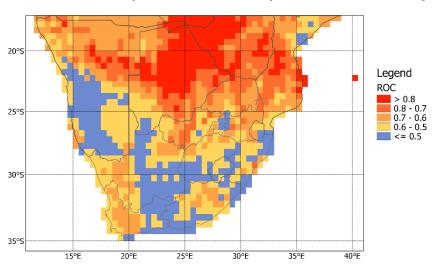
ASO 2020 Max Temp; ICs: May



ROC Area (Above-Normal): ASO Max Temp



ROC Area (Below-Normal): ASO Max Temp



Round-up: SADC Max Temp

- Cool early winter conditions over some parts
- Predominantly high maximum temperatures are likely during the forecast period

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 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