Seasonal forecasts

presented by:



Seasonal Forecast

https://tinyurl.com/ForecastProf

UNIVERSITEIT VAN PRETORIA
UNIVERSITY OF PRETORIA
YUNIBESITHI YA PRETORIA

Worx

Seasonal Climate Forecasts

Latest Update: 15 July 2023

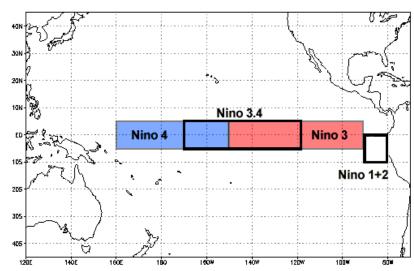
- The seasonal forecasts presented here by Seasonal Forecast Worx are based on forecast output of
 the coupled ocean-atmosphere models administered through the North American Multi-Model
 Ensemble (NMME) prediction experiment (http://www.cpc.ncep.noaa.gov/products/NMME/;
 Kirtman et al. 2014). NMME real-time 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 <u>official</u> seasonal forecast for South Africa, visit the South African Weather Service website at http://www.weathersa.co.za/images/data/longrange/gfcsa/scw.pdf

Weather Service

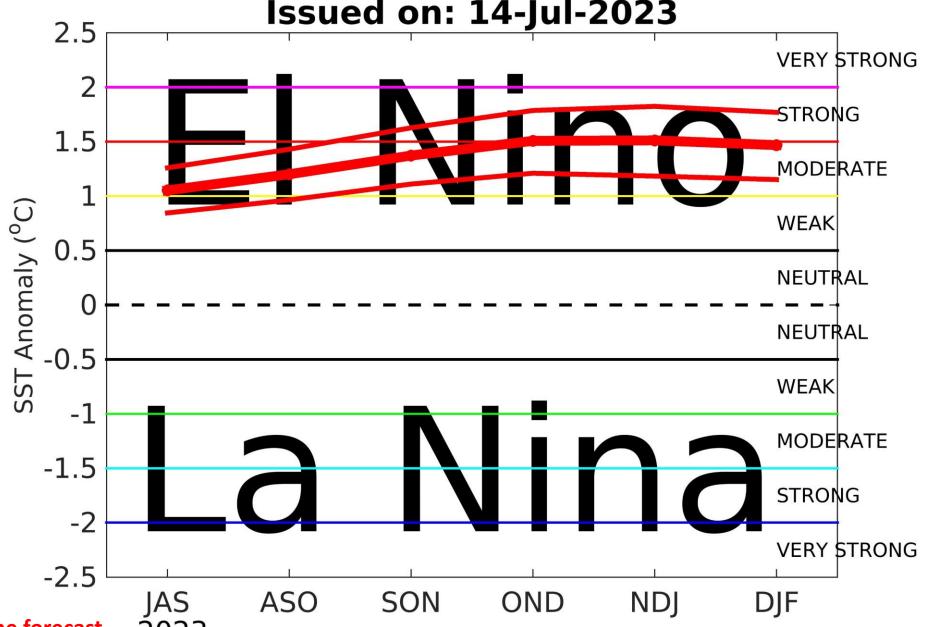
ENSO and Global SST Forecasts

Prediction Method

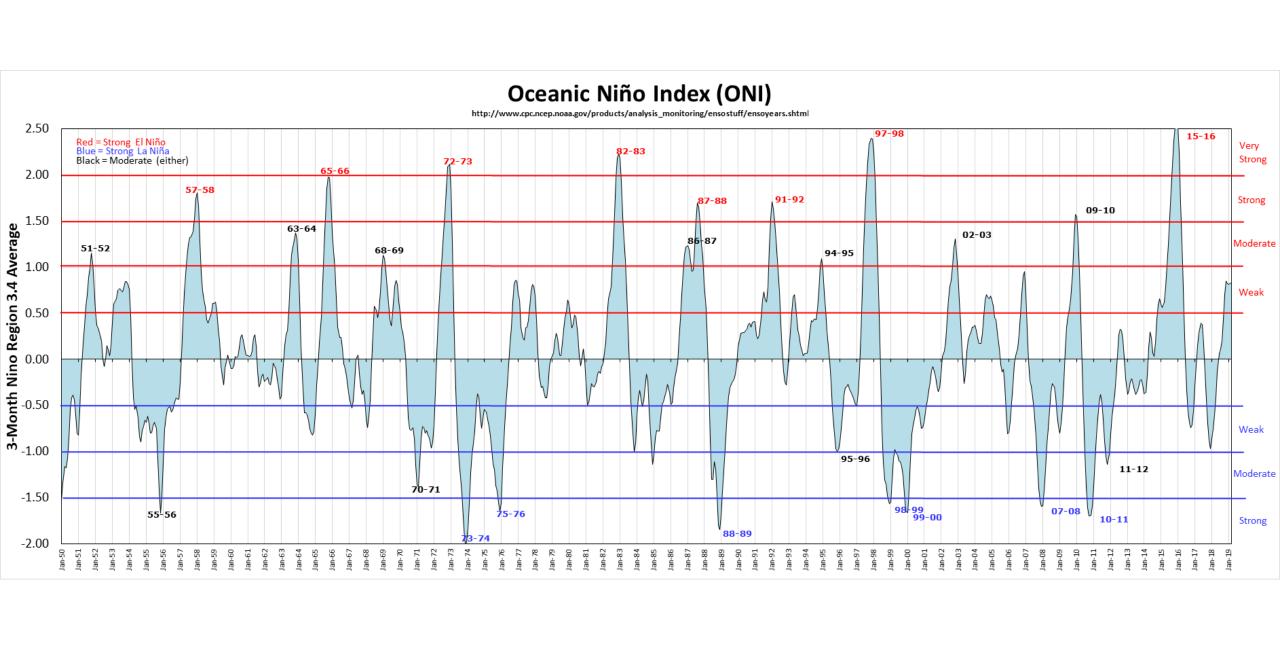
- Forecasts for global sea-surface temperature (SST) fields are obtained through a combination of NMME models and a linear statistical model, that uses antecedent SST as a predictor (Landman et al. 2011). Forecasts for the Niño3.4 area (see insert) are derived from the global forecasts.
- SST forecasts 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



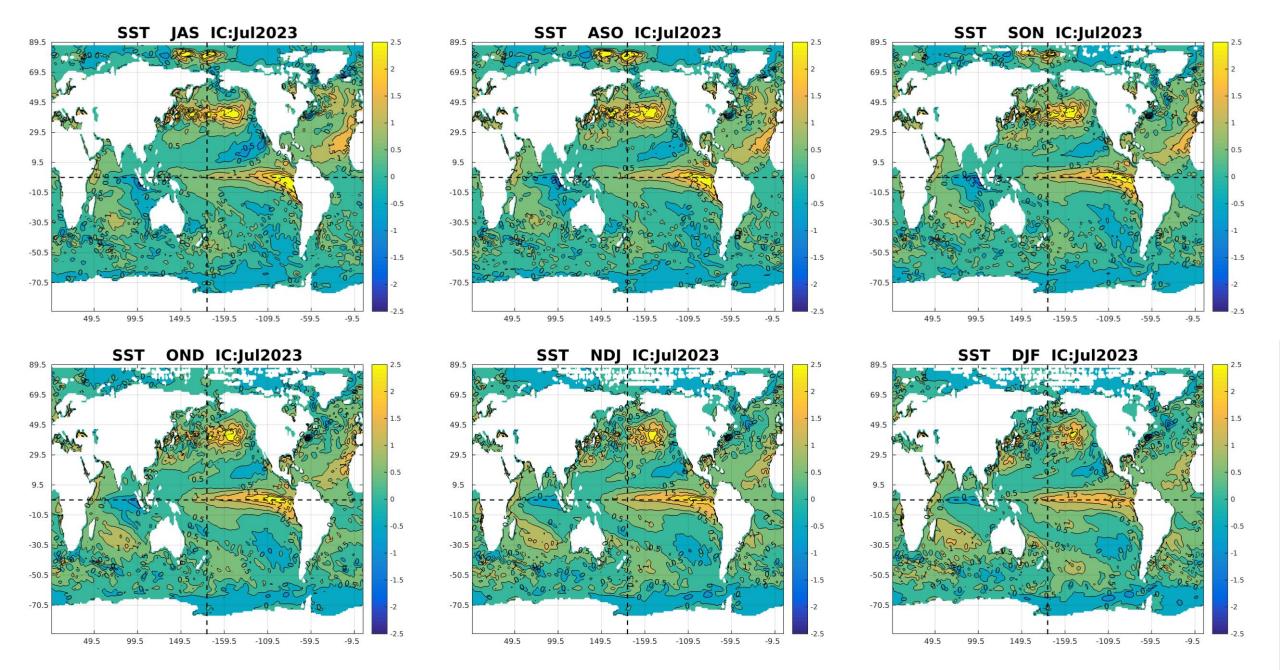
CSiriMM Nino3.4 SST Forecast Issued on: 14-Jul-2023



Middle red line: the forecast 2023
Thin red lines: 25% confidence levels

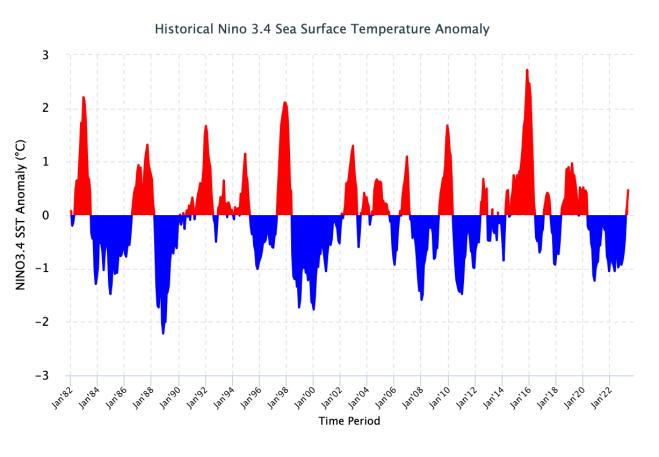


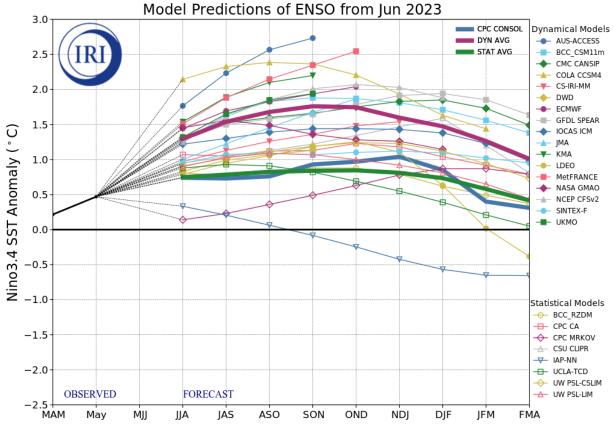
SST anomalies (in °C, where blue is cooler and orange is warmer)



Round-up: ENSO

 The UP model predicts a moderate to strong El Niño event during the Austral summer





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 interpolated to the 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" 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 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.

Forecasts are probabilistic

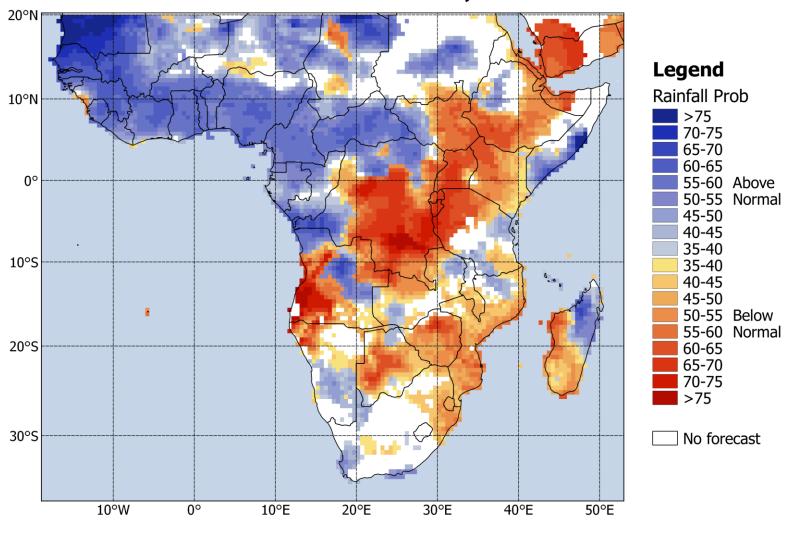
Probabilistic forecasts can help users understand risks and opportunities (forewarned is forearmed) in order to make more informed decisions.

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.

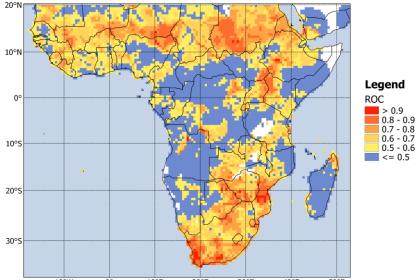
The nature of a probabilistic forecast implies that the less likely outcomes are always possible. In fact, for the probabilistic forecasts to be considered reliable, the less likely outcomes will and must occasionally occur.

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

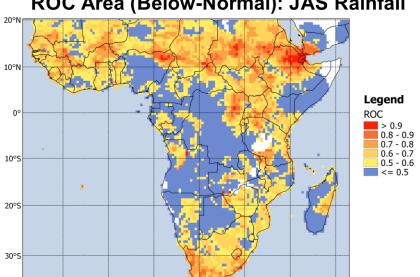
JAS 2023 Rainfall; ICs: Jul



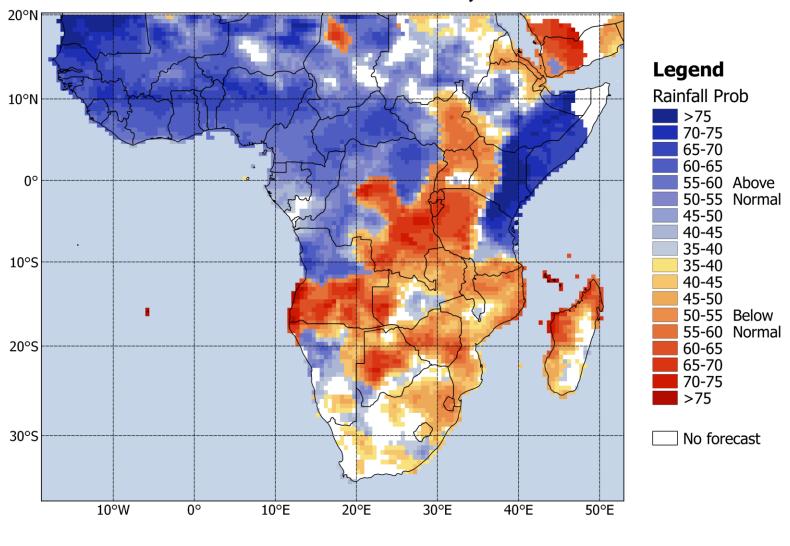
ROC Area (Above-Normal): JAS Rainfall



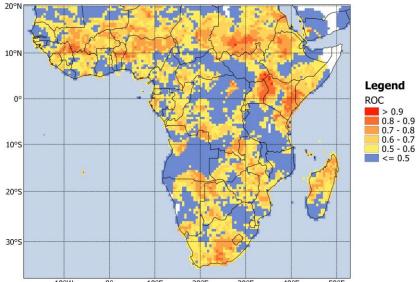
ROC Area (Below-Normal): JAS Rainfall



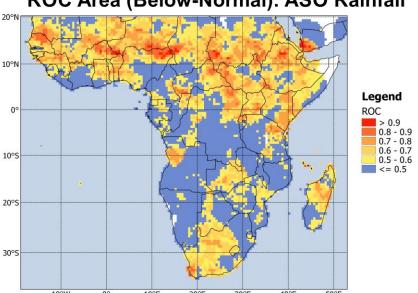
ASO 2023 Rainfall; ICs: Jul



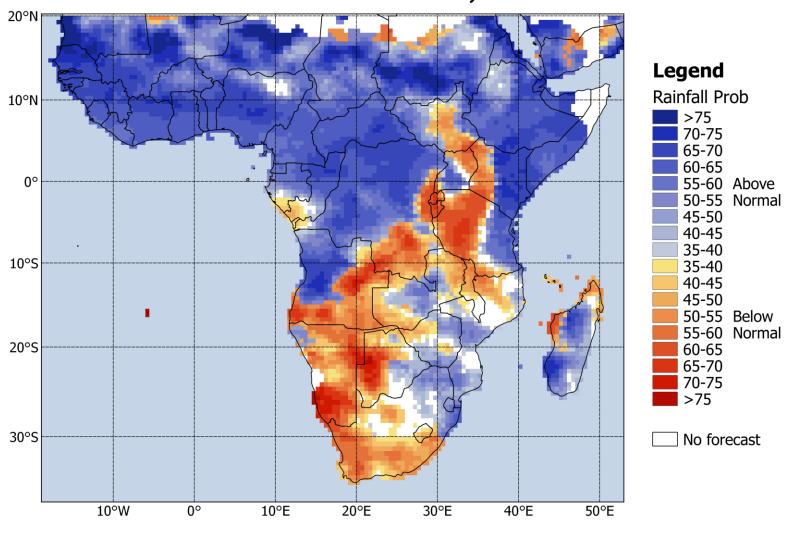
ROC Area (Above-Normal): ASO Rainfall



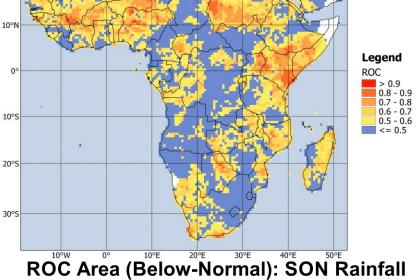
ROC Area (Below-Normal): ASO Rainfall

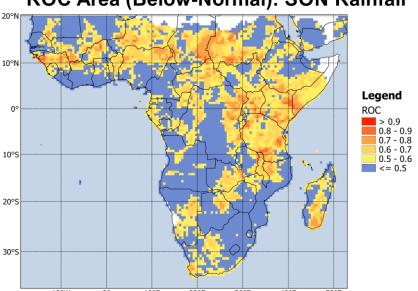


SON 2023 Rainfall; ICs: Jul

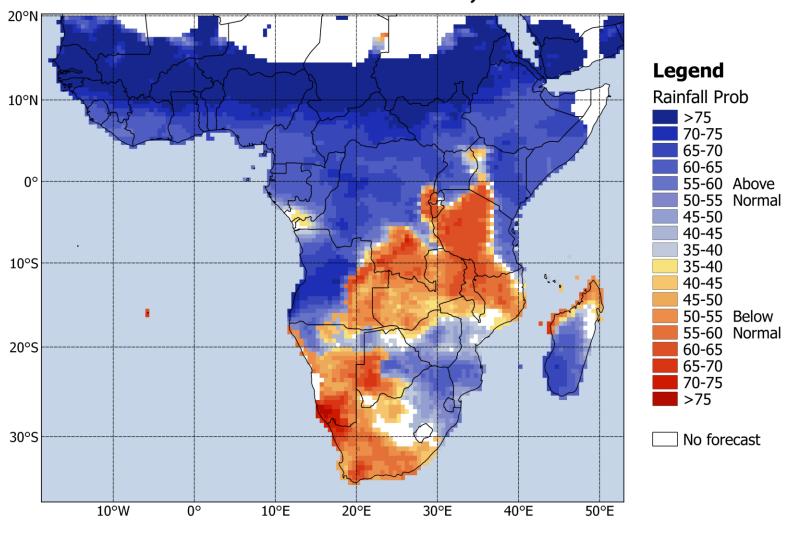


ROC Area (Above-Normal): SON Rainfall Legend

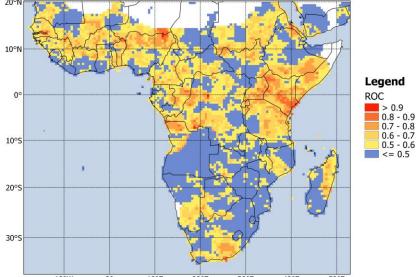




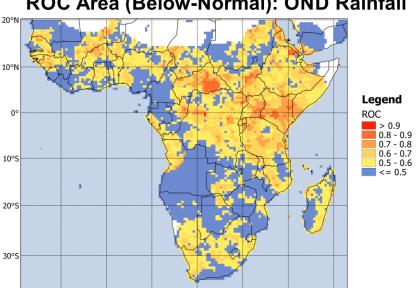
OND 2023 Rainfall; ICs: Jul



ROC Area (Above-Normal): OND Rainfall



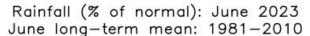
ROC Area (Below-Normal): OND Rainfall

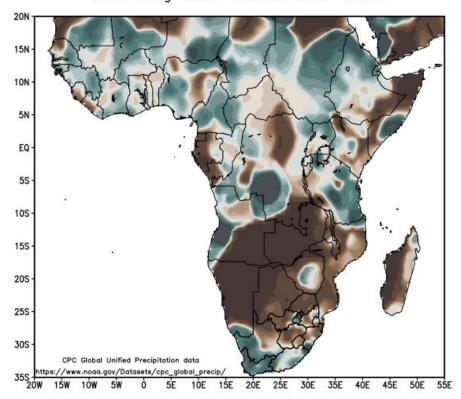


Round-up: Rainfall south of 15°S

- Above-normal rainfall is expected over the western winter rainfall region and southern coast until early spring, followed by below normal rainfall until the end of the forecast period
- Below-normal rainfall is expected over the forecast region during spring and early summer where there is skill (ROC>0.5)

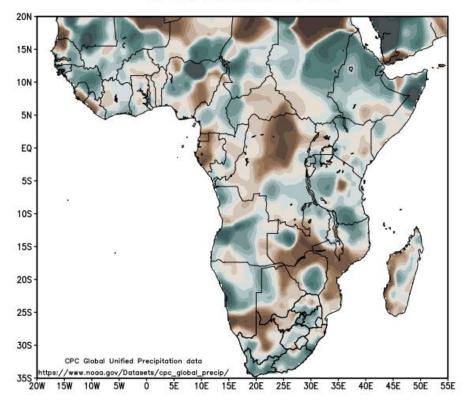
Observed Rainfall





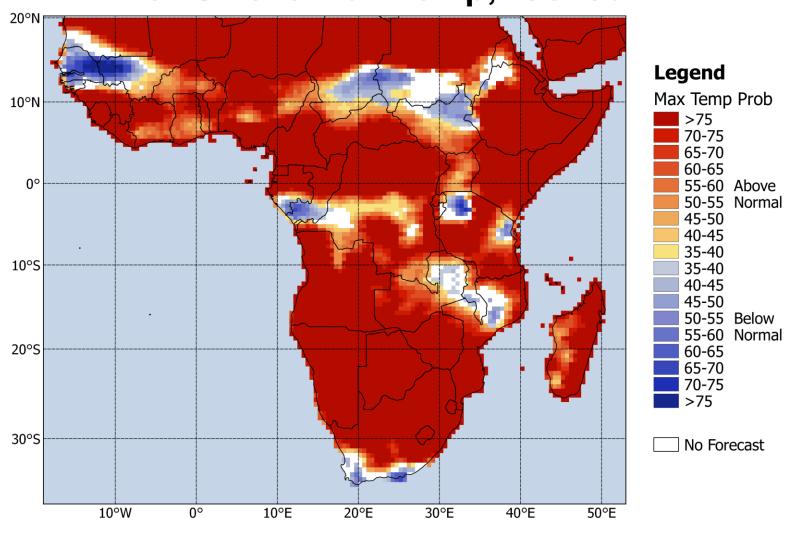
Rainfall (% of normal): AMJ 2023

AMJ long-term mean: 1981-2010

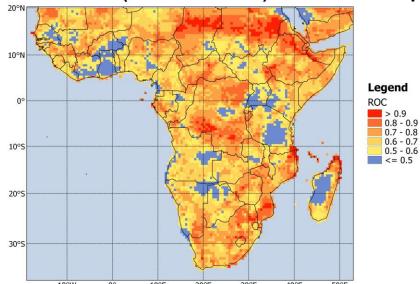


Recorded rainfall for June and the April-May-June season show below-normal rainfall over the brown areas and above-normal rainfall over the green areas

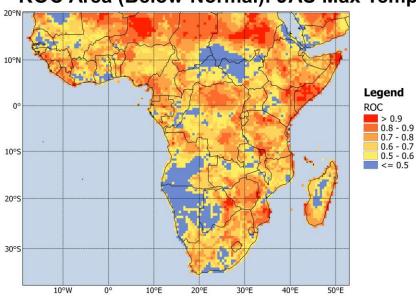
JAS 2023 Max Temp; ICs: Jul



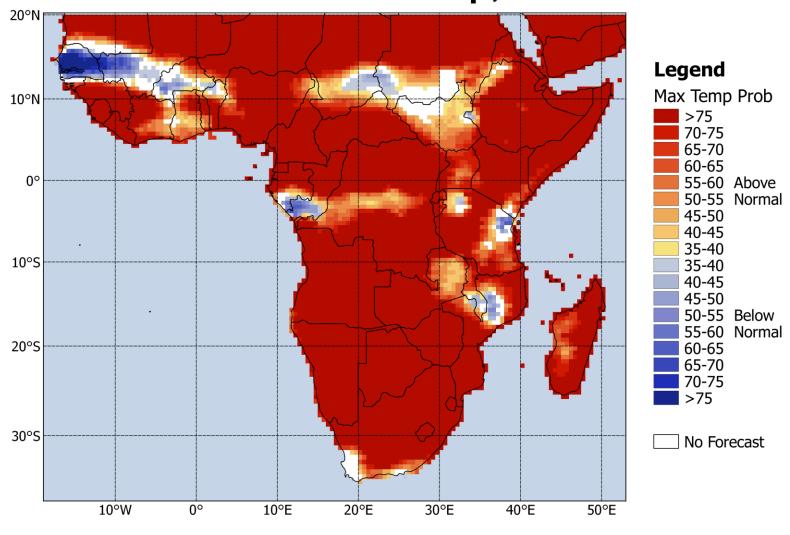
ROC Area (Above-Normal): JAS Max Temp



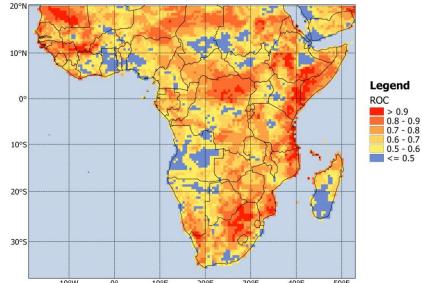
ROC Area (Below-Normal): JAS Max Temp



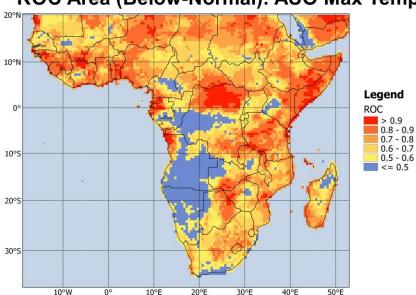
ASO 2023 Max Temp; ICs: Jul



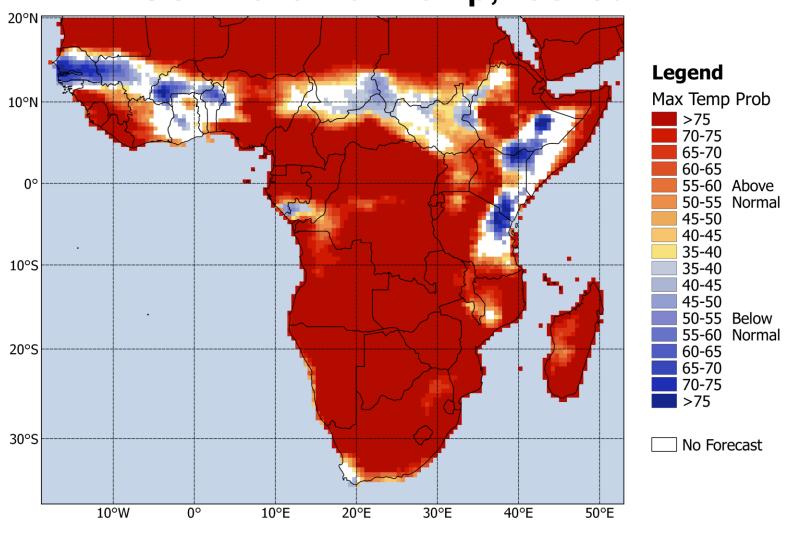
ROC Area (Above-Normal): ASO Max Temp



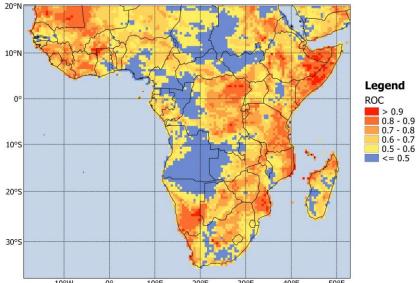
ROC Area (Below-Normal): ASO Max Temp



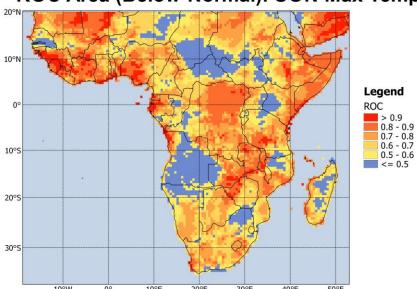
SON 2023 Max Temp; ICs: Jul



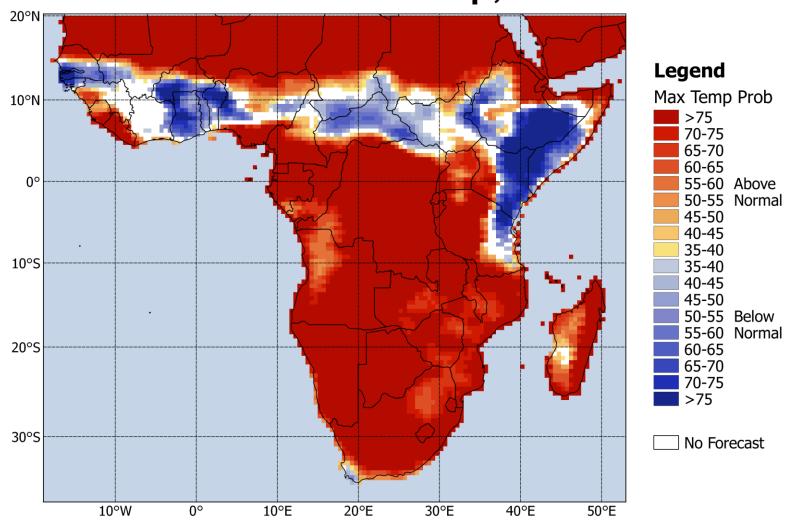
ROC Area (Above-Normal): SON Max Temp



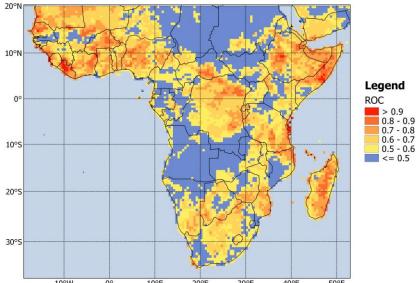
ROC Area (Below-Normal): SON Max Temp



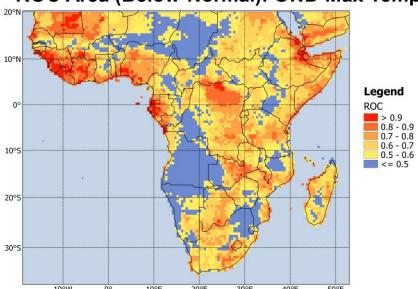
OND 2023 Max Temp; ICs: Jul



ROC Area (Above-Normal): OND Max Temp



ROC Area (Below-Normal): OND Max Temp



Round-up: South of 15°S Max Temp

- Enhanced likelihood of above-normal maximum temperatures over almost the entire region are expected during the forecast period
- Below-normal maximum temperatures only likely over the southern coastal and southwestern Cape during winter and early spring

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., 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.

Financial support from...

- The National Research Foundation
 - Incentive Funding for Rated Researchers (since 2017)
 - 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 set

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 Gosiame, 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.