Seasonal forecasts

presented by:



Seasonal Forecast

Worx

Latest Update: 9 October 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 <u>WALandman1981@gmail.com</u>



 $2002\,2004\,2006\,2008\,2010$

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 S forecasts over the same 9-years for below- (low yields), near- (about average) and above-normal (high yields). For example, in <u>2008</u> the forecast and observed index values are high and positive (figure on the left), and the ^O 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 (<u>http://www.cpc.ncep.noaa.gov/products/NMME/</u>; Kirtman et al. 2014). NMME real-time seasonal forecast and hindcast (re-forecast) data are obtained from the data library (<u>http://iridl.ldeo.columbia.edu/</u>) of the International Research Institute for Climate and Society (IRI; <u>http://iri.columbia.edu/</u>).
- 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 (<u>http://www.up.ac.za/en/geography-geoinformatics-and-meteorology/</u>). Statistical post-processing is performed with the CPT software (<u>http://iri.columbia.edu/our-expertise/climate/tools/cpt/</u>).
- 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



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







SST anomalies

1.5

0.5

-0.5

-1.5

2.5

1.5

0.5

-0.5

-1.5

2.5

-9.5

-9.5



Round-up: ENSO

- The UP model predicts weak La Niña conditions for spring and mid-summer, but ENSO-neutral conditions during the second half of summer
- Most forecast models continue to be for colder SST, with just a few warmer than the UP model (CS-IRI-MM)

CPC CONSOL vnamical Mode DYN AVG - NASA GMAD STAT AVG 2.5 - NCEP CF5v2 **IRI/CPC** - IMA BCC_CSM11 2.0 - SAUDI-KAU - LDEO - AUS/ACCES Anomaly (°C) UKMO - KMA SNU IOCAS ICM 1.0 - COLA COSM MetFRANCE SINTEX-F 0.5 CORDERED - GEDI CM2 - CMC CANSIF SST Nino -1.0 istical Models NTU CODA O- BCC BZDM -1.5 CPC MRKOV - CPC CA -O- CSU CLIPR -2.0 -O- JAP-NN -O- FSU REGI OBSERVED FORECAST - UCLA-TCD -2.5 SON OND ASO NDI DIF **IFM** FMA MAM

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

ROC Area (Above-Normal): OND Rainfall

15°E

20°E

25°E

30°E

35°E

40°E



OND 2020 Rainfall; ICs: Oct

ROC Area (Above-Normal): NDJ Rainfall



ROC Area (Above-Normal): DJF Rainfall

15°E

20°E

25°E

30°E

35°E

40°E



DJF 2020-21 Rainfall; ICs: Oct

ROC Area (Above-Normal): JFM Rainfall



Round-up: SADC Rainfall

 Favourable rainfall outcomes are expected over the larger part of the forecast region during spring and summer, except over parts of Namibia and Mozambique

Observed SADC Rainfall

Rainfall (% of normal): July-August-September 2020 Relative to July-August-September 1981-2010 rainfall



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



Maps prepared by Dr. Christien Engelbrecht

ROC Area (Above-Normal): OND Max Temp

Legend

ROC > 0.8 0.8 - 0.7 0.7 - 0.6 0.6 - 0.5 <= 0.5

Legend

> 0.8 0.8 - 0.7 0.7 - 0.6

0.6 - 0.5

ROC

40°E

20°

35°S

15°E

20°E

25°E

30°E

35°E

40°E



OND 2020 Max Temp; ICs: Oct

ROC Area (Above-Normal): NDJ Max Temp

Legend

> 0.8 0.8 - 0.7 0.7 - 0.6 0.6 - 0.5 <= 0.5

Legend

> 0.8

0.6 - 0.5

0.7 - 0.6

ROC

ROC

20°5

35°S

15°E

20°E

25°E

30°E

35°E

40°E



NDJ 2020-21 Max Temp; ICs: Oct

ROC Area (Above-Normal): DJF Max Temp

Legend

ROC > 0.8 0.8 - 0.7 0.7 - 0.6 0.6 - 0.5 <= 0.5

Legend

> 0.8

0.7 - 0.6

0.6 - 0.5

ROC

20°5

35°S

15°E

20°E

25°E

30°E

35°E

40°E



DJF 2020-21 Max Temp; ICs: Oct

ROC Area (Above-Normal): JFM Max Temp



20°S

25°S

30°S

35°S

15°E

20°E

25°E

30°E

35°E

40°E



>75

70-75 65-70 60-65

45-50

40-45 35-40 35-40 40-45

45-50

60-65

65-70

70-75 >75

ROC Area (Below-Normal): JFM Max Temp



JFM 2021 Max Temp; ICs: Oct

Round-up: SADC Max Temp

 Cooler maximum temperatures are likely in association with the increased likelihood of a wet summer season over parts of the region

Tailored Forecasts

- 1. Bapsfontein <u>end-of-season-yield</u> three-category probabilistic forecast for 2021
- 2. Probabilistic three-category <u>rainfall</u> forecast for the farm of Robbie Kingsley for Dec-Jan-Feb 2020/21
- 3. Probabilistic three-category malaria forecast for Limpopo for Dec-Jan-Feb 2020/21
- 4. Probability of exceedance Dec-Jan-Feb 2020/21 inflow forecast for Lake Kariba, Zambia/Zimbabwe
- 5. Probabilistic <u>rainfall</u> forecast for Jan-Feb-Mar 2021 for the farm Buschbrunnen near Grootfontein, Namibia

Crop-yield data and forecasts for a farm near Bapsfontein, South Africa Landman et al. (2019)



Bapsfontein



Oct Forecasts for DJF Rainfall vs Observed: Smithfield



Dec-Jan-Feb 2020/21 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)



Retrospective

REAL-TIME

Inflow forecast for Lake Kariba: onset season of DJF Muchuru et al. (2016)



JFM rainfall forecast for the farm Buschbrunnen near Grootfontein, Namibia Landman et al. (2016)



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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