The development and prudent application of climate-based forecasts of seasonal malaria in the Limpopo province in South Africa

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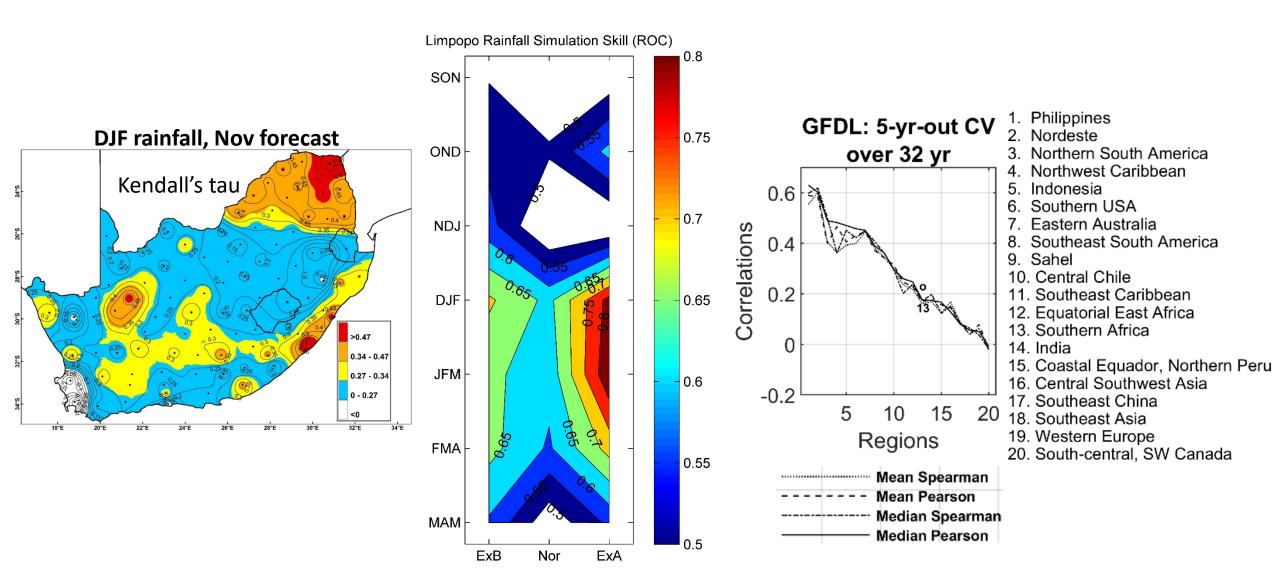
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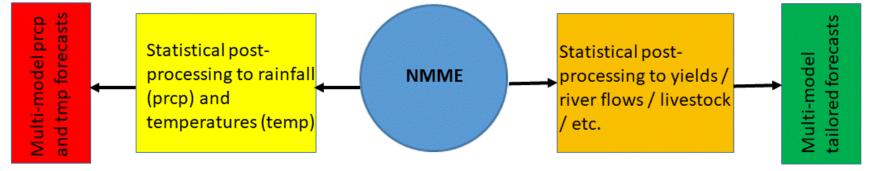
journal homepage: www.elsevier.com/locate/envdev

South African Society for Atmospheric Sciences 2020 Online Seminar, 20 November

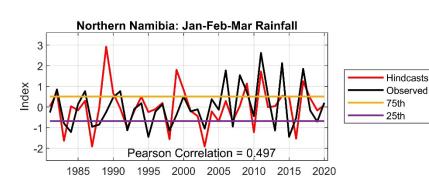


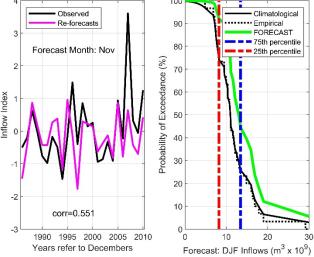
Limpopo seasonal rainfall predictable

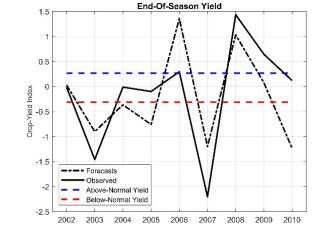
Skill for southern Africa modest, Limpopo skill elevated



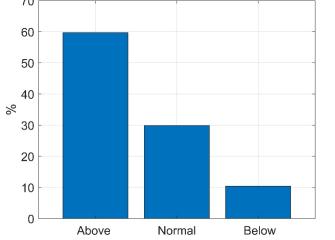
Dec-Jan-Feb Lake Kariba INFLOW Exceedance Probs: DJF 2020/21; ICs: Nov



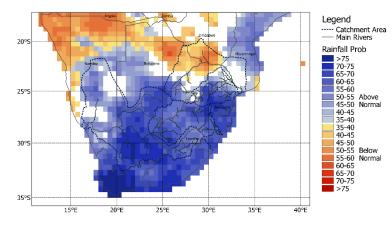


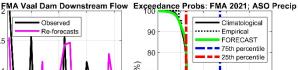


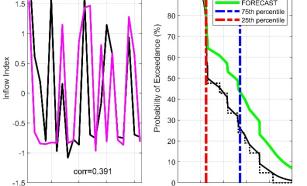
Crop-Yield Forecast for 2021, Made in November 2020



DJF 2020/21 Rainfall; ICs: Nov

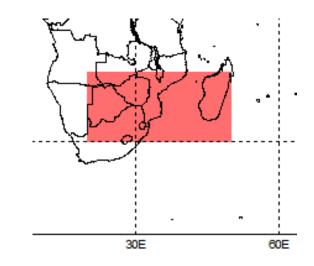




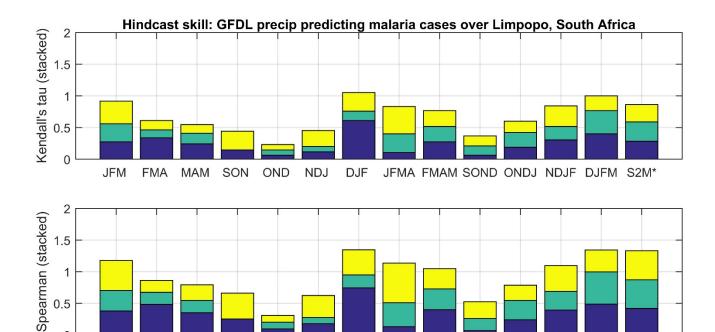


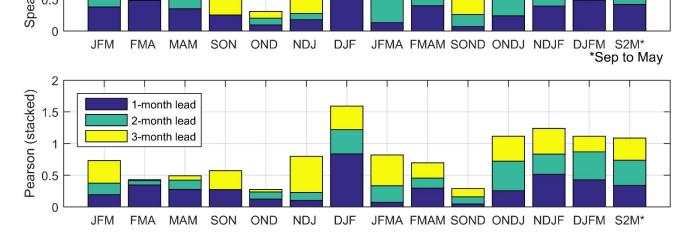
-2 -1 0 1 2

2000 2005 2010 2015

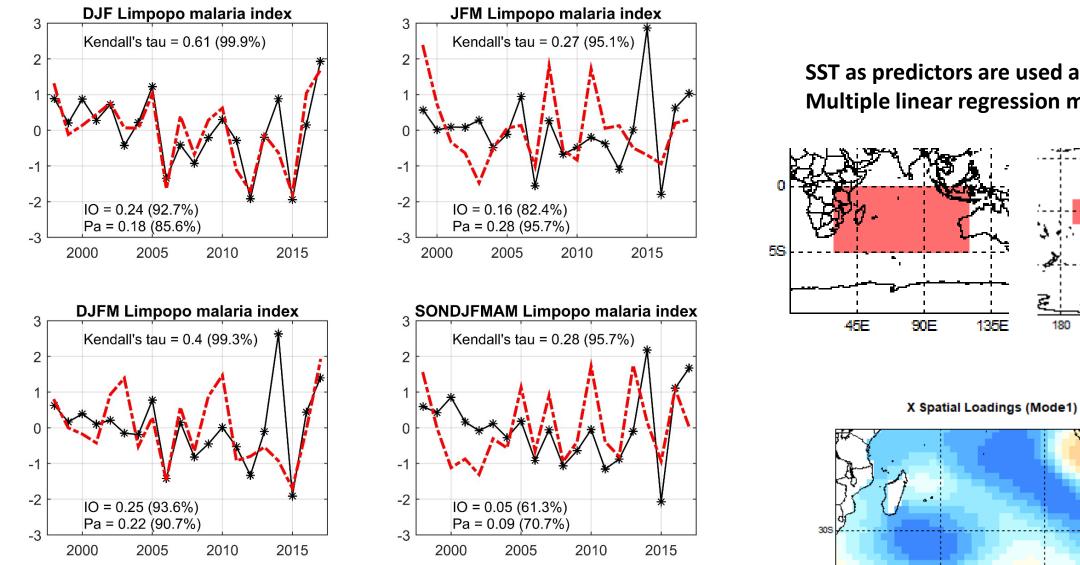


- Natural logs (In) of the malaria seasonal data resulted in data to become normally distributed (Lilliefors test)
- GFDL-CM2p5-FLOR-B01 seasonal rainfall forecasts are statistically downscaled to seasonal malaria values (canonical modes of forecasts used in a multiple linear regression model as predictors)



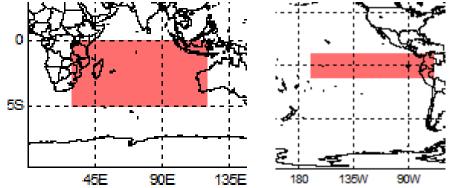


Correlations between hindcasts (re-forecast) and observed for the seasons indicated. Lead-times up to 3 months are presented.

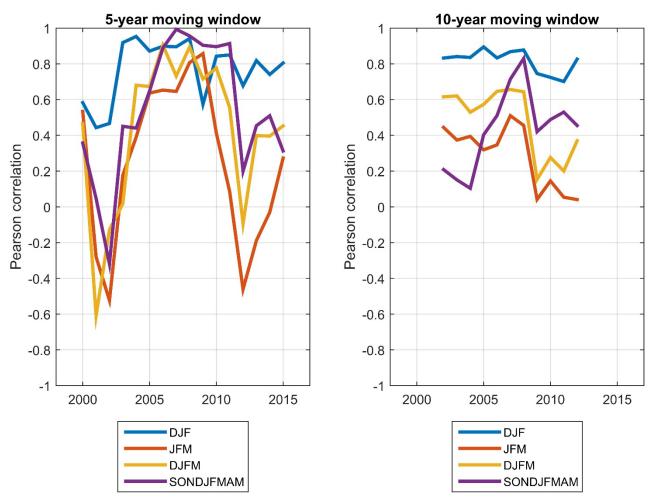


Cross-validated malaria cases hindcasts (1-month lead) for the four seasons indicated. Time series have been normalised. The top value on each panel is the Kendall's tau (Kt) correlation between hindcast (red dashed) and observed (black asterisked). IO: Kt when using Indian Ocean SST as predictor; Pa: Kt when using equatorial Pacific Ocean SST as predictor.

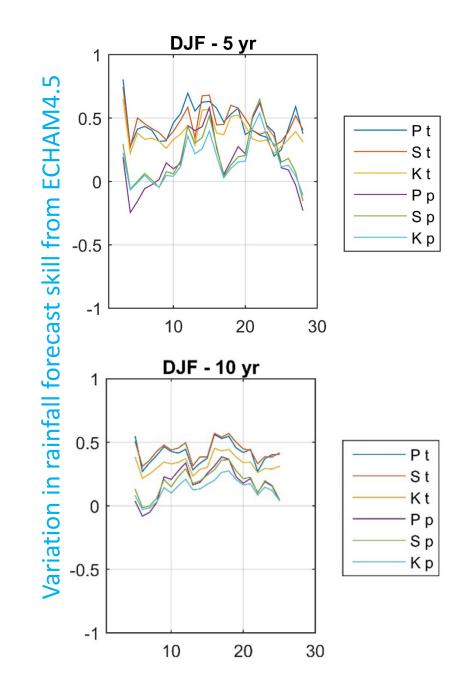
SST as predictors are used as baseline: Multiple linear regression models



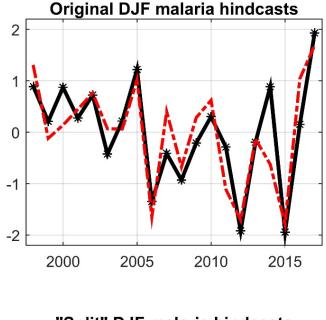
60E 90E 0 90 -0 75 -0 60 -0 45 -0 30 -0 15 0 00 0 15 0 30 0 45 0 60 0 75 0 90

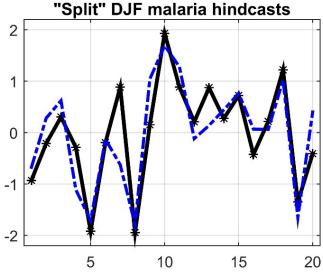


Respectively 5- and 10-year moving windows are used to calculate correlations between observed and hindcasts for the four seasons indicated. The DJF season is not only found to be associated with highest skill (correlations), but is also the season during which forecast skill remains the most consistent.

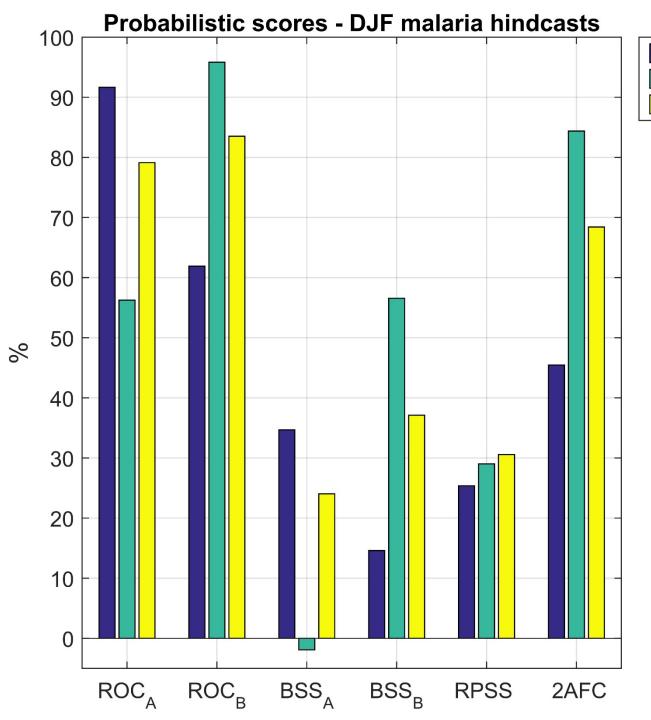


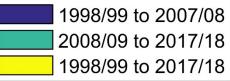
- Seasonal forecasts should be expressed probabilistically – and also need to be verified as such
- 2. **Probabilistic forecasts** are produced through the calculation of error variances of downscaled ensemble mean values
- 3. Typically, the first 10 years of the 20-year period are used to train the statistical model to produce probabilistic forecast for the second 10-year period
- 4. To obtain 20 years of probabilistic forecasts, we swapped the two 10-year periods around, with the result that the second 10year period can be used to also produce probabilistic forecasts for the first 10-year period. This process results in the production of 20 years of probabilistic forecasts





Top panel shows the DJF hindcasts from before. Bottom panel shows the cross-validated hindcasts by placing the second 10 years of the 20-year period In front of the first 10 years.





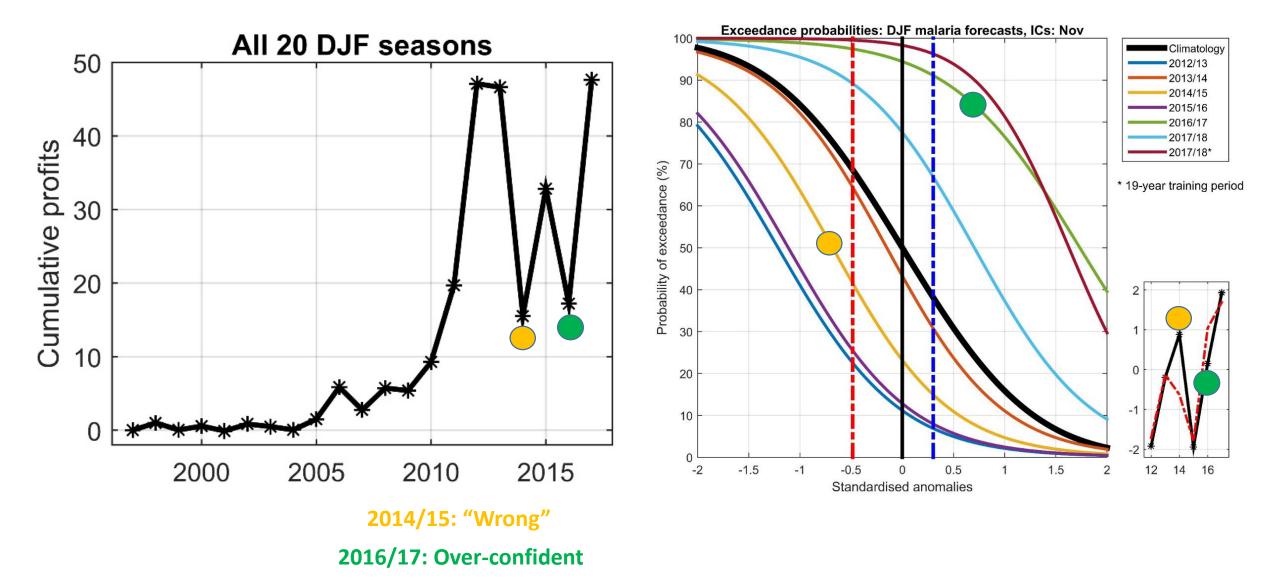
Resolution slopes:

	Above	Below
98/99-07/08	1.60	0.80
07/08-17/18	0.42	1.52
98/99-17/18	0.98	1.12



Note on resolution slopes: The <u>full period</u> shows slopes that are <u>close to perfect reliability</u>

- First 10 years more above-normal DJF cases high ROC-A and BSS-A
- 2. Second 10 years more below-normal DJF cases high ROC-B and BSS-B
- 3. THUS skill level is a function of verification period
- 4. **2AFC** (probability of a correct decision made by, for example, a **health practitioner**), shows values great than 50% for the whole and for second 10-year period



Summary:

- Limpopo is area of relatively high seasonal predictability
- Existing **experience** in statistical downscaling for **applications** modelling applied here for **malaria predictions**
- Justification presented to use physical global coupled models' output as input in malaria models (statistical, in this case)
- Issues to be cognizant of:
 - Verification **period** may strongly influence perceived forecast **skill**
 - Some predicted years may simply be "wrong"
 - Such years may have **serious implications** on future health operations

The final paragraph of the paper:

- We have developed and presented a robust and skillful seasonal forecast system for mid-summer malaria incidence...
- ...we have demonstrated how occasional false-positive and false-negative forecasts can result in material financial losses.
- Therefore we urge caution in relying on these forecast models exclusively for disease management....
- ...we therefore suggest... to make sure that good climate monitoring systems are in place to supplement forecasts from models...
- ...we propose that disease surveillance and control activities should not be replaced by forecasts...
- ...forecasts should only be used to supplement health practices that are currently going on in the region.