

IRI Forecast Systems update

A stylized background map of the Pacific Ocean. A prominent band of yellow and red colors stretches from the bottom left towards the top right, representing a forecast or data visualization. The rest of the map is a solid dark blue.

Ángel G. Muñoz

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Outline



1. Seasonal
 - A. New Forecast System (2017; so this is a refresher)
 - B. Skill Maps
2. Sub-seasonal
 - A. New Forecast System (2018)
 - B. Skill Maps

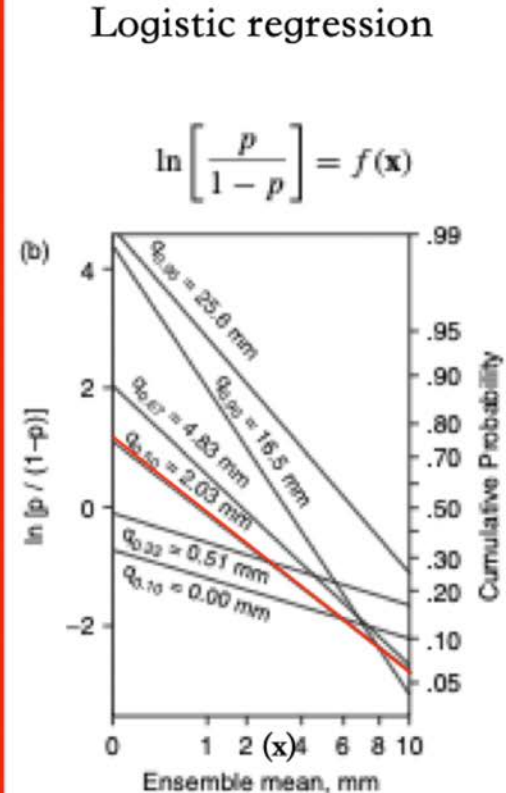
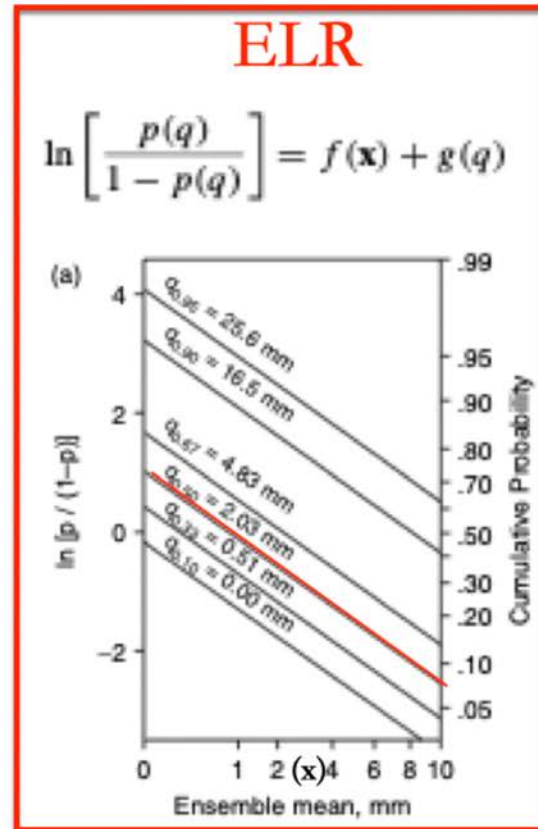
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New Seasonal Forecast System

Re-calibration using Extended Logistic Regression

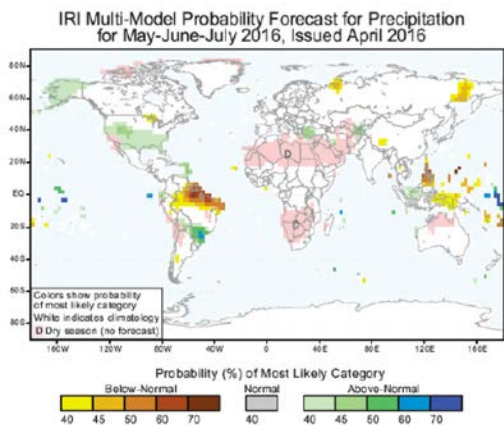
GFS Day 6 – 10
Precip Forecast for
Minneapolis
28 Nov – 2 Dec
2001
Wilks (2009)



Applied at each grid point, using forecast ensemble mean

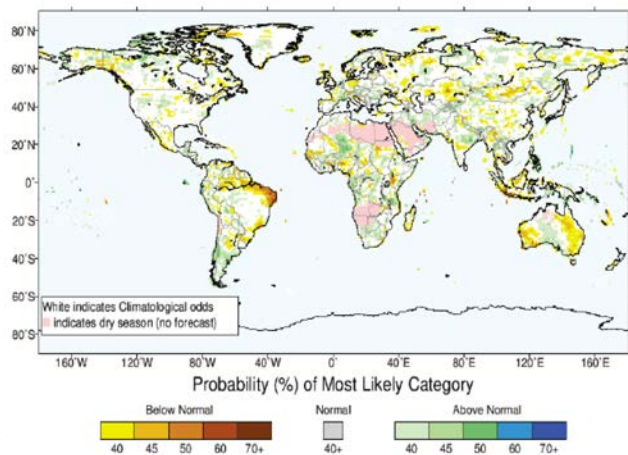
Old system

Seasonal Forecast Systems



New system

IRI Multi-Model Probability Forecast for Precipitation for May-June-July 2017, Issued April 2017



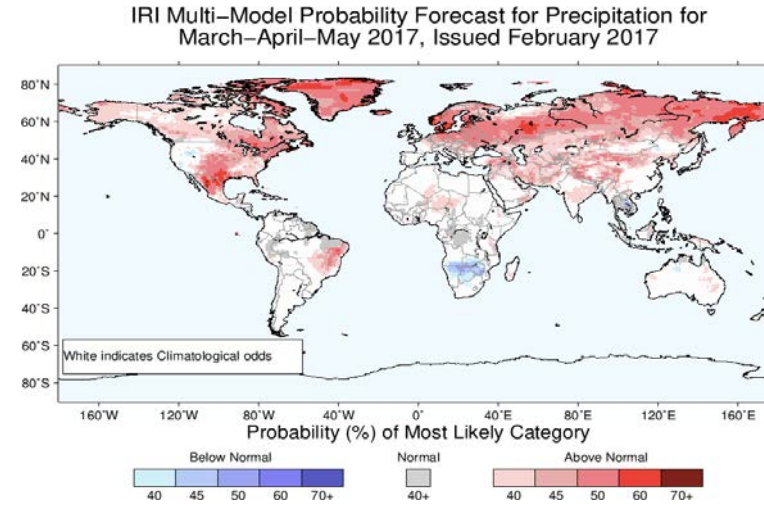
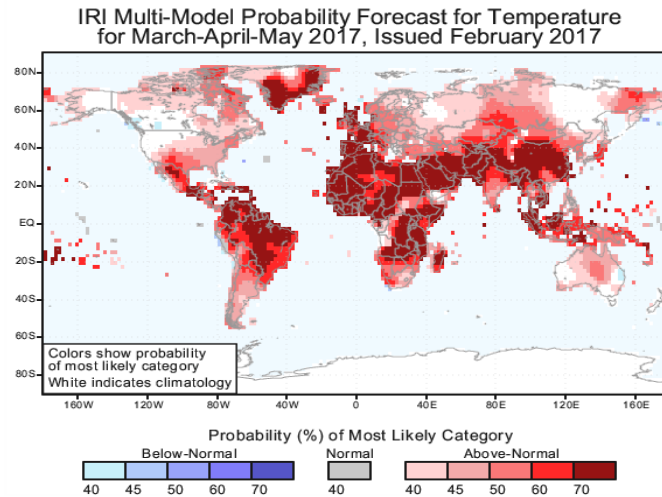
Courtesy of N. Acharya and the SCF team @ IRI

| | Existing IRI forecast | New IRI forecast |
|---------------------------------|----------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------------|---------------------------------------------------------------------------------------|
| GCM used (Predictors) | 2-tier (uncoupled) ECHAM 4.5, CCM3.6, COLA, GFDL,CFSv2 | 1-tier (coupled) NMME models |
| Observed data used (Predictand) | Precip: CMAP Temp: CAMS | Precip: CPC-CMAP Temp: GCHN updated |
| Forecast Resolution | 2.5 degree grid | 1 degree grid |
| Calibration method | <ul style="list-style-type: none">• Pattern-based correction of ensemble means<ul style="list-style-type: none">- PC Regression based on tropical precip EOFs- Spread estimate from historical forecasts with forecast SST• Equal weighting of corrected models• Parametric forecast probabilities (T - Gaussian, P - transformed Gaussian) | Extended Logistic Regression (Non-Gaussian) at grid point level. |
| Dry mask | Forecast are only produced when the climatology being more than 30 mm precipitation in any given season | Forecast are only produced when the at least 10% of the training sample are non-zero. |
| Making Flexible forecast | Used mean and SD of the forecast, then use parametric approach | Integrated part of the ELR method |

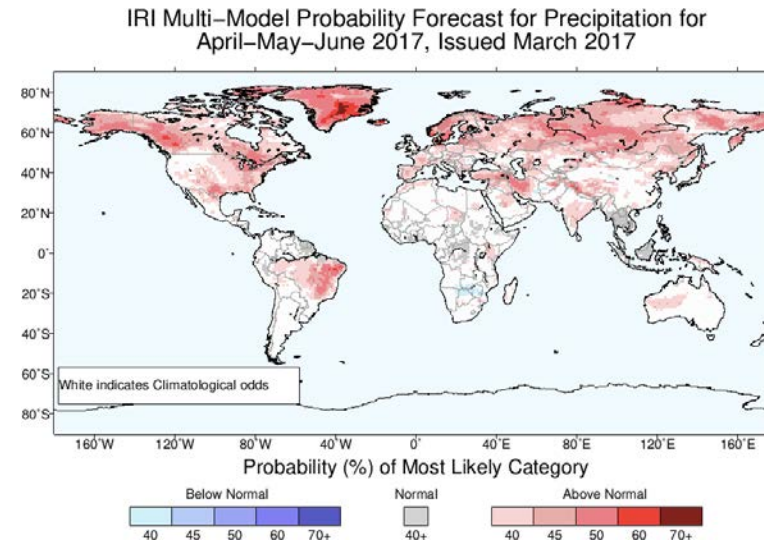
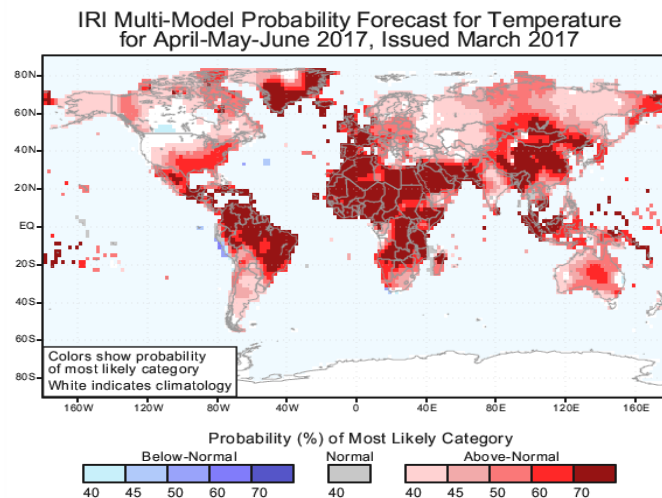
Old system

New system

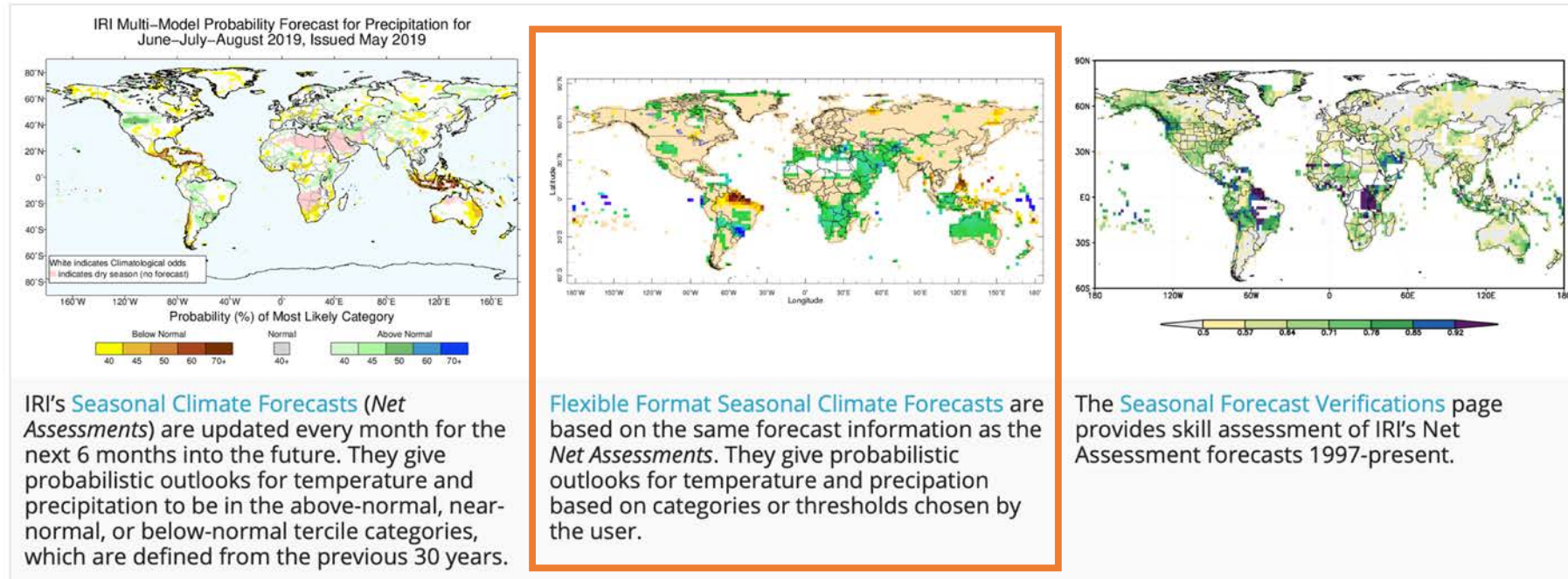
Feb
forecast,2017



Mar
forecast,2017



Seasonal Climate Forecasts



The IRI probabilistic seasonal climate forecast product is based on a re-calibration of model output from the [U.S. National Oceanographic and Atmospheric Administration \(NOAA's North American Multi-Model Ensemble Project \(NMME\)\)](#). This includes the ensemble seasonal prediction systems of [NOAA's National Centers for Environmental Prediction](#), Environment and Climate Change Canada, [NOAA/Geophysical Fluid Dynamics Laboratory](#), [NASA](#), [NCAR](#) and [COLA/University of Miami](#). The output from each NMME model is re-calibrated prior to multi-model ensembling to form reliable probability forecasts. The forecasts are now presented on a 1-degree latitude-longitude grid.

[Details of the forecast system, post-processing, and recommended references for citation can be found here.](#) Forecasts from the individual NMME models are shown on [NOAA CPC's website](#). Verifications of IRI's real-time forecasts issued since 1998 can be found on the [Seasonal Climate Verifications](#) pages.

https://iri.columbia.edu/our-expertise/climate/forecasts/#Seasonal_Climate_Forecasts



Precipitation Flexible Seasonal Forecast

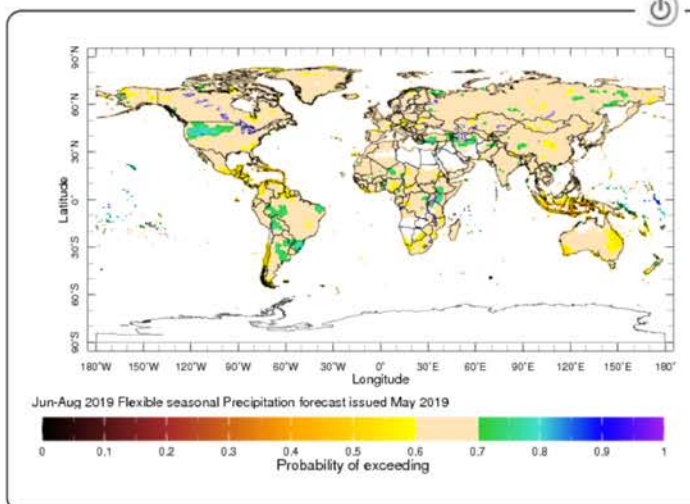
This seasonal forecasting system consists of probabilistic precipitation seasonal forecasts based on the full estimate of the probability distribution.

Please refer to our [licensing agreement](#) for permission to use any IRI forecast material.

Probabilistic seasonal forecasts from multi-model ensembles through the use of *statistical recalibration*, based on the historical performance of those models, provide reliable information to a wide range of climate risk and decision making communities, as well as the forecast community. The flexibility of the full probability distributions allows to deliver interactive maps and point-wise distributions that become relevant to user-determined needs.

The default map shows globally the seasonal precipitation forecast probability (colors between 0 and 1) of exceeding the 50th percentile of the distribution from historical 1982-2010 climatology. The forecast shown is the latest forecast made (e.g. Dec 2017) for the next season to come (e.g. Jan-Mar 2018). Four different seasons are forecasted and it is also possible to consult forecasts made previously. The forecasts are directly computed from the *extended logistic regression* model as probabilities of exceeding (or non-exceeding) of every 5th percentile of the climatological distribution. The specific quantile (in steps of 5 percentile points) can then be selected. The user can also specify a quantitative value in physical units (here seasonal total precipitation in mm) for probability of exceeding or non-exceeding. The final probability maps are smoothed spatially with a 9x9 point Gaussian smoother.

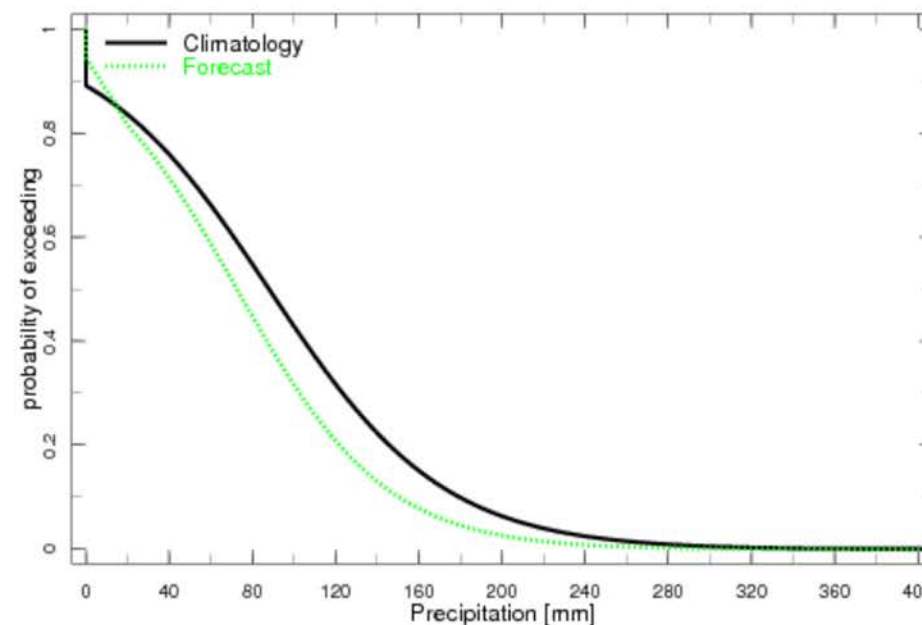
Clicking on a point on the map will show the local probability of exceeding and probability distribution of the forecast (green) together with the climatological distribution (black).



| Target Date | Issue Date | Lead Time |
|--------------|-----------------|-----------|
| Jun-Aug 2019 | 0000 1 May 2019 | 2.5 |

Forecast made for [148E-149E, 29.5S-28.5S]
located in or near **Goondiwindi (R), Queensland, Australia**

Probability of Exceeding



Jun-Aug 2019 Flexible seasonal Precipitation forecast issued May 2019



Precipitation Flexible Seasonal Forecast

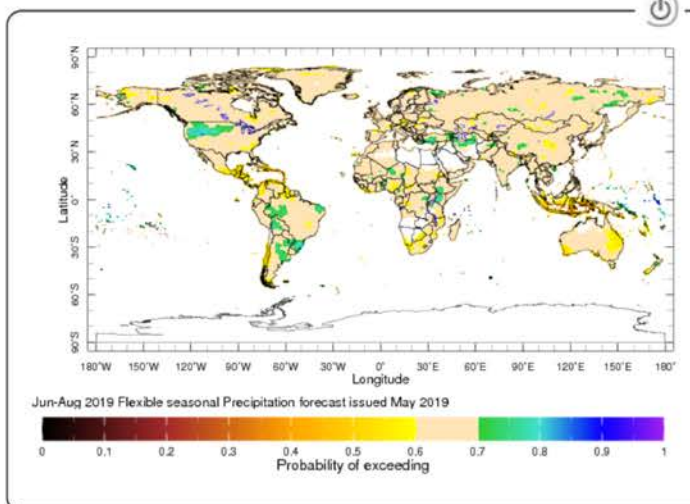
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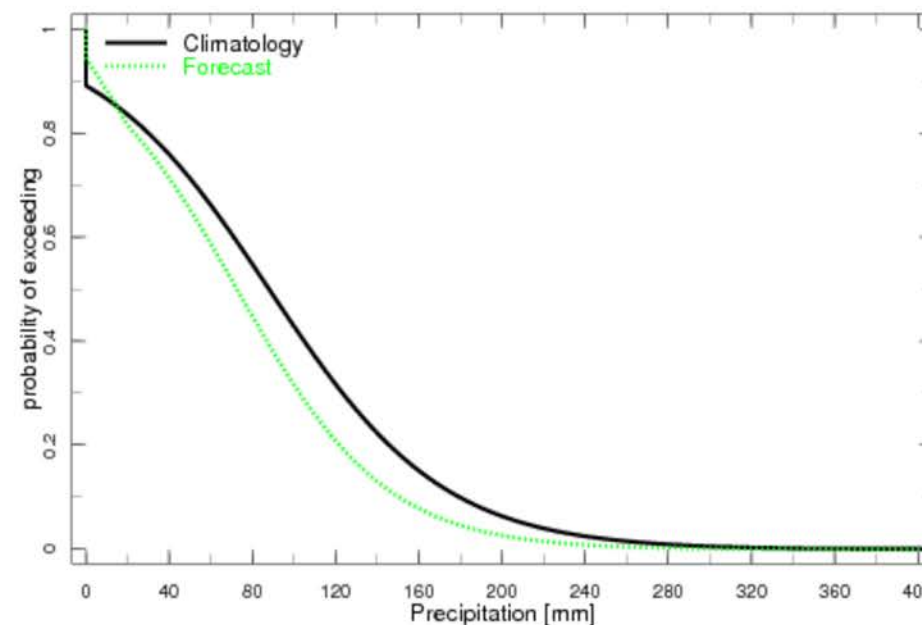
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Probability of Exceeding



Jun-Aug 2019 Flexible seasonal Precipitation forecast issued May 2019



Description

Expert Mode

Options

Instructions

home nachiketa .ELR_seasonal_RPSS_hindcast_1982_2010

.precip .rpss_precip_ELR.nc .rpss

latitude -85 85 RANGE

longitude -180 180 RANGEEDGES

startcolormap DATA

-0.1 0.2 RANGE

white DodgerBlue DodgerBlue

-0.05 bandmax

cyan cyan

-0.0125 bandmax

gray gray

0.0125 bandmax

yellow yellow

0.025 bandmax

gold gold

0.05 bandmax

orange orange

0.075 bandmax

DarkOrange DarkOrange

0.1 bandmax

tomato tomato 0.125

bandmax

OrangeRed OrangeRed 0.15

bandmax

red red 0.175

bandmax

magenta magenta 0.2

bandmax

magenta endcolormap

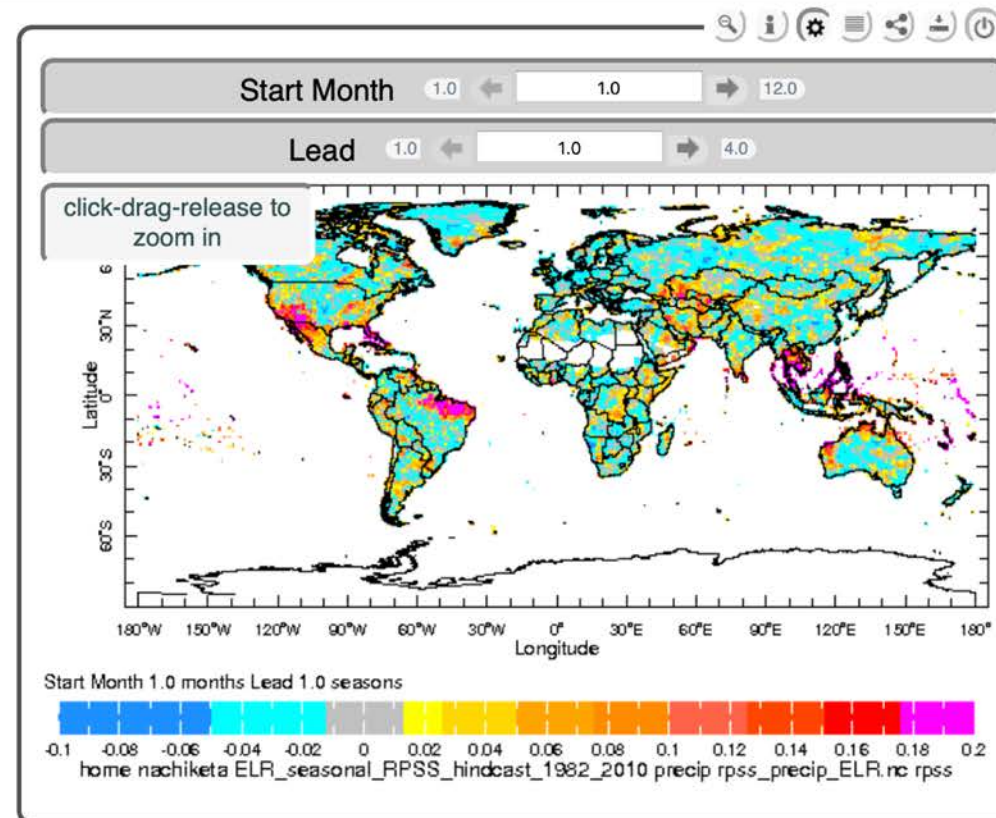
a- -a longitude latitude fig- colors coasts_gaz countries_gaz -
fig

OK

reset

- home[]
- home nachiketa ELR_seasonal_RPSS_hindcast_1982_2010 precip rpss_precip_ELR.nc
rpss[longitude latitude strtmonth l lead]
- home nachiketa ELR_seasonal_RPSS_hindcast_1982_2010 precip rpss_precip_ELR.nc rpss[longitude latitude strtmonth l lead]
- grid: /longitude (degree_east) ordered (180W) to (180) by 1.0 N= 361 pts :grid
- grid: /latitude (degree_north) ordered (85S) to (85N) by 1.0 N= 171 pts :grid
- fig: colors coasts_gaz countries_gaz :fig

EPSG:4326

<http://wiki.iri.columbia.edu/index.php?n=Climate.SeasForecastDev>

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Climate

Subseasonal Forecasts

SubX Forecasts

Precipitation Probability Forecast

Region

Global

Forecast Issued

0000 24 May 2019

Target Period

✓ 1-14 Jun 2019

8-21 Jun 2019

Description

Dataset Documentation

Instructions

Contact Us

Precipitation Probability Forecast

Calibrated Subseasonal Tercile categories precipitation real-time forecasts from the Subseasonal eXperiment ([SubX](#))

The default map shows the latest forecast for weeks 2-3 ahead (i.e. the 14-day Saturday-Friday target period, 9 to 22 days after the forecast is issued), as probability of the dominant tercile category. Previous forecasts can be viewed through the control bar menu. The weeks 3-4 forecast (i.e. the 14-day target period, 16 to 29 days after the forecast is issued) is also available. New forecasts are issued weekly on Fridays.

When navigating to a forecast of which Target Period is in the past, a smaller side map shows a verification of the forecast as the observed tercile values according to the 1999-2014 training period of the calibration of the forecast.

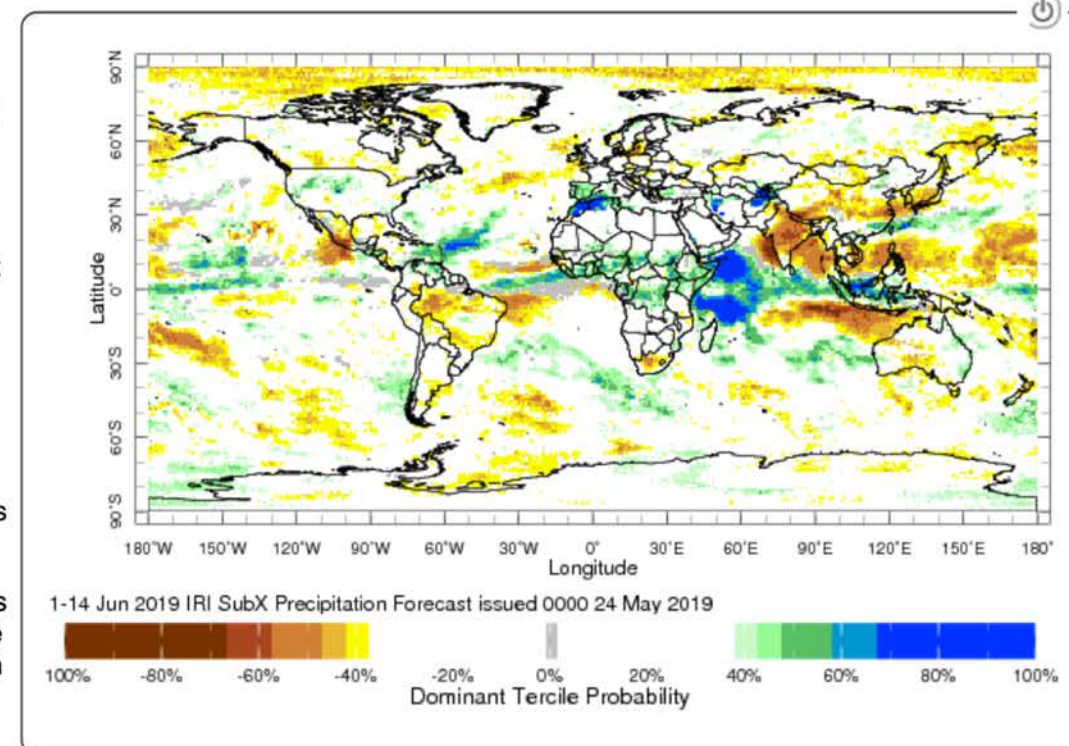
Clicking on the forecast map will show, for the clicked grid box, the probabilities for the 3 forecasts categories (Below-, Near- and Above- Normal).

The probabilistic forecasts shown here are obtained from the statistical calibration of three models (NCEP CFSv2, NCEP GEFS, and NOAA/ESRL FIM HYCOM, each run on Wednesdays) from the SubX database which are combined with equal weight to form multi-model ensemble precipitation tercile probabilities forecasts. Individual model forecasts are calibrated separately for each point, start and lead using Extended Logistic Regressions (ELR; Vigaud et al, 2017) based on the historical performance of each model, and thus provide reliable intra-seasonal climate information in regards to a wide range of climate risk of concerns to the decision making communities and for which subseasonal forecasts are particularly well suited.

Subseasonal-to-seasonal forecasting techniques are still under development. This Maproom shows the type of experimental forecast information currently being created at these time scales in real time.

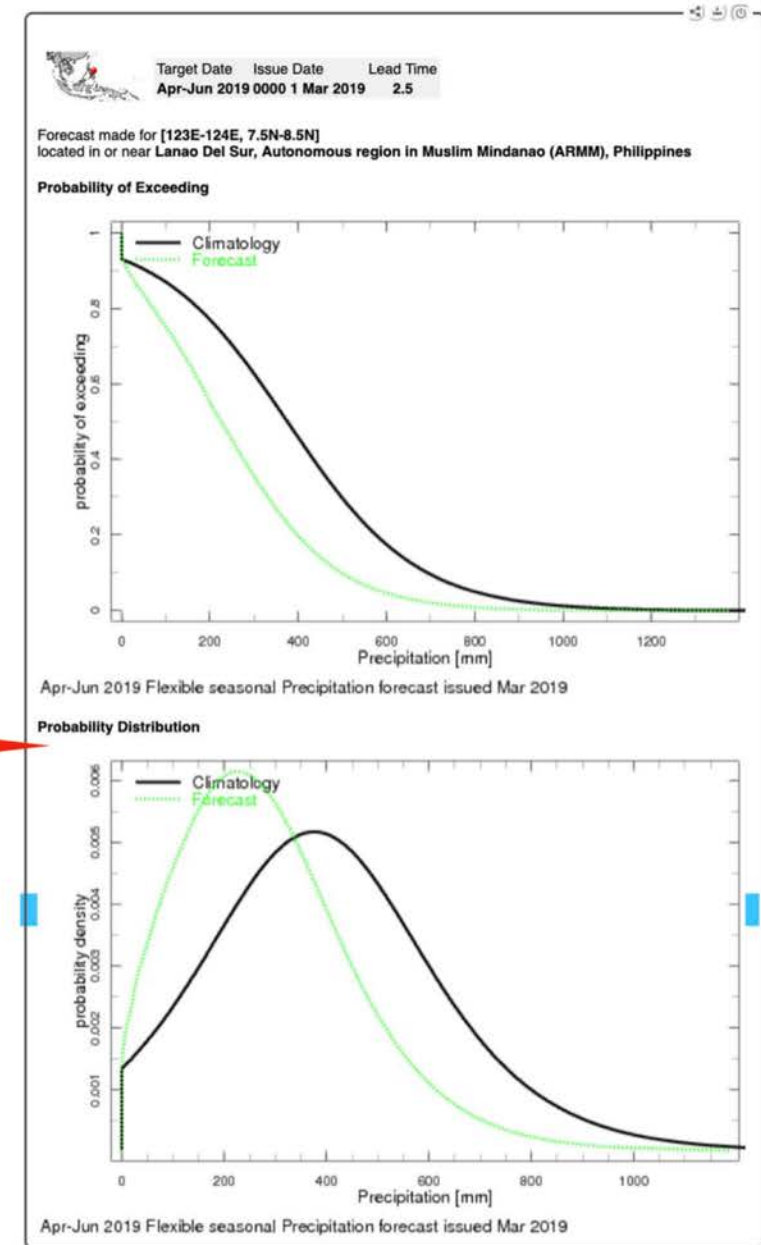
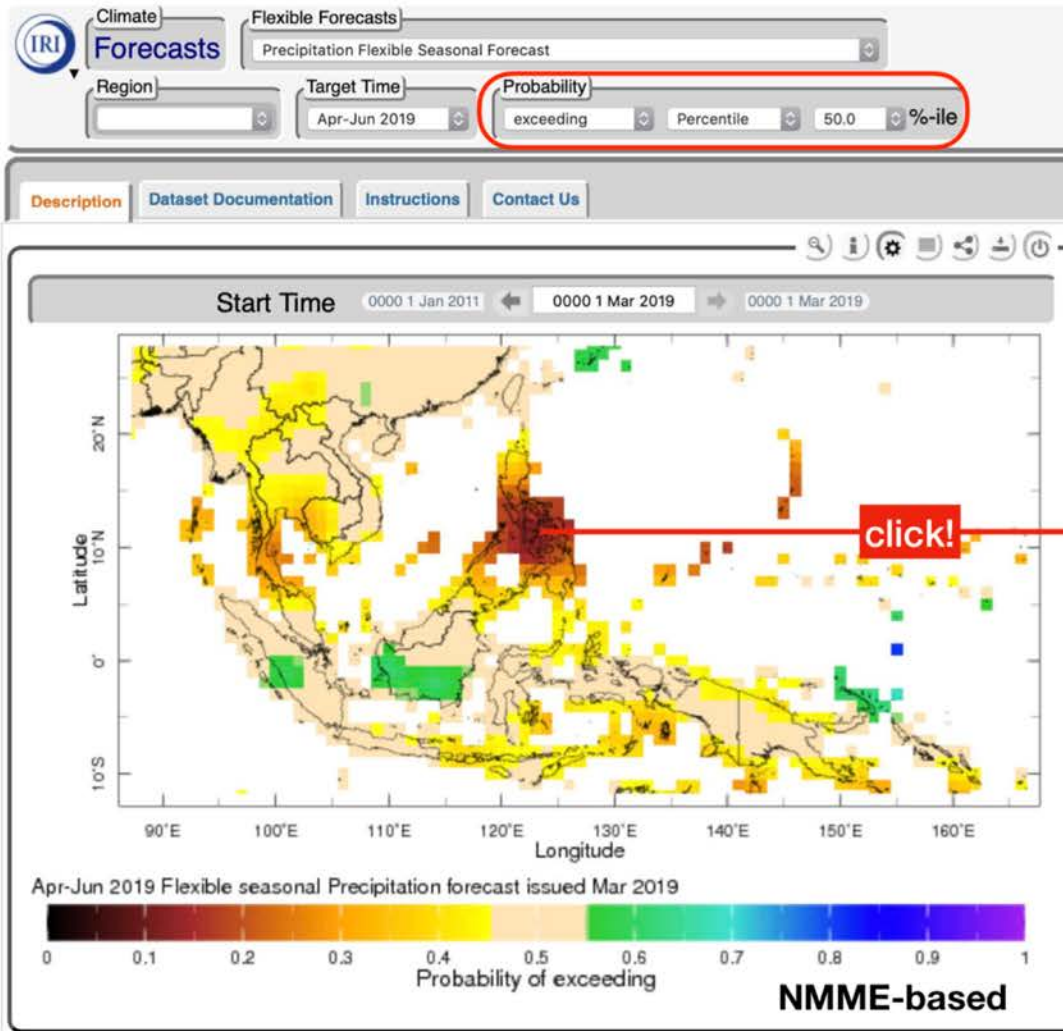
References:

- Pegion, K. et al. 2018: [The Subseasonal Experiment \(SubX\): A multi-model subseasonal prediction experiment](#) submitted to Bull. Amer. Meteor.
- Vigaud, N., A.W. Robertson, and M.K. Tippett, 2017: [Multimodel Ensembling of Subseasonal Precipitation Forecasts over North America](#). Mon. Wea. Rev., 145, 3913–3928



http://iridl.ldeo.columbia.edu/maproom/Global/ForecastsS2S/precip_subx.html

Flexible Format Seasonal Forecast Maprooms



Precipitation Probability Forecast

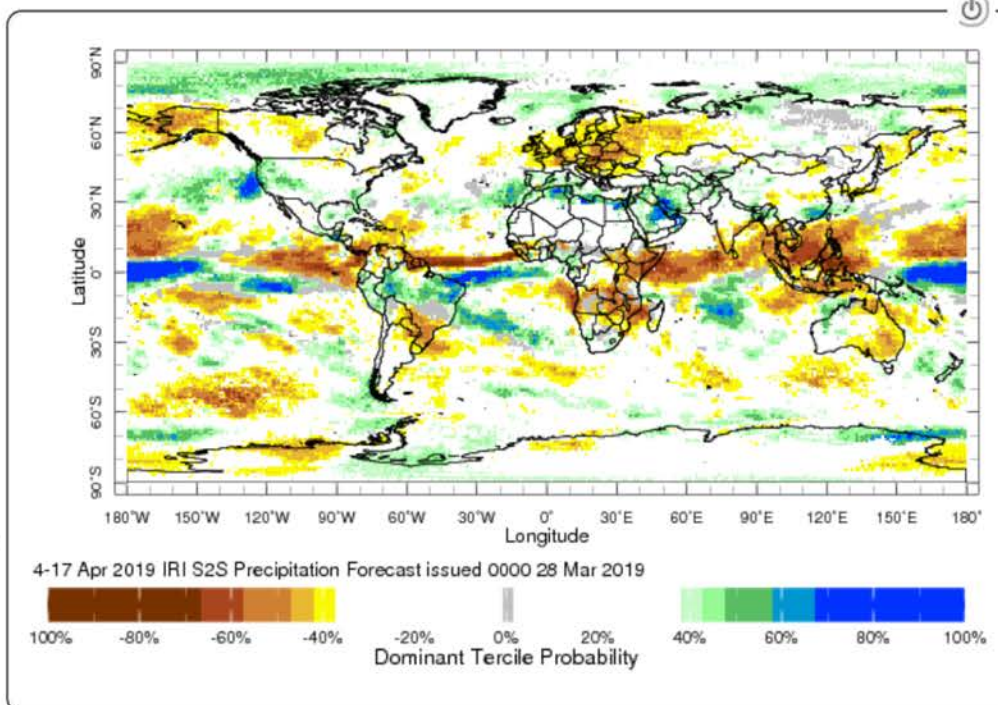
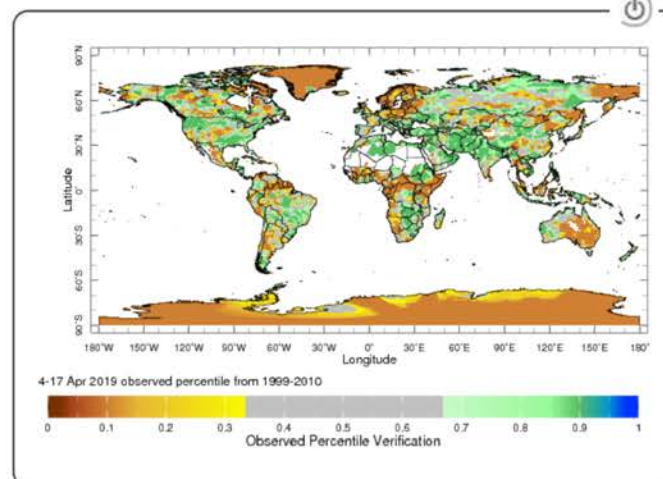
Calibrated Subseasonal Tercile categories precipitation forecasts lagged behind real time from the S2S database.

The default map shows the latest forecast for weeks 2-3 ahead (i.e. the 14-day Thursday-Wednesday target period, 8 to 21 days after the forecast is issued), as probability of the dominant tercile category. Previous forecasts can be viewed through the control bar menu. The weeks 3-4 forecast (i.e. the 14-day target period, 15 to 28 days after the forecast is issued) is also available. The smaller side map shows a verification of the forecast in current view as the observed tercile values according to the 1999-2010 training period of the calibration of the forecast. New forecasts are issued weekly on Thursdays but are released on a monthly basis. Please note that these forecasts are "lagged", i.e. they are not available in real time.

Clicking on the map will show, for the clicked grid box, the probabilities for the 3 forecast categories (Below-, Near- and Above- Normal).

The probabilistic forecasts shown here are obtained from the statistical calibration of three models (ECMWF, NCEP CFSv2 and CMA, each run on Thursdays) from the Subseasonal to Seasonal (S2S) Prediction Project database (Vitart et al, 2017) which are combined with equal weight to form multi-model ensemble precipitation tercile probabilities forecasts. Individual model forecasts are calibrated separately for each point, start and lead using Extended Logistic Regressions (ELR; Vigaud et al, 2017) based on the historical performance of each model, and thus provide reliable intra-seasonal climate information in regards to a wide range of climate risk of concerns to the decision making communities and for which subseasonal forecasts are particularly well suited.

As subseasonal-to-seasonal (S2S) forecasting techniques are being developed, and more and more models are made available in (near) real-time, this Maproom shows the type of forecast information that can be currently delivered at these time scales.



http://iridl.ldeo.columbia.edu/maproom/Global/ForecastsS2S/precip_s2s.html

Precipitation Hindcast Skill

Subseasonal skill score based on the historical performance of each model and their multi-model ensemble.

The different skill scores are mapped by calendar month. The forecasts lead times are combined over the weeks 2-3 and the weeks 3-4 from the forecast start time (i.e., 14-day long periods respectively 8 to 21 days and 15 to 28 days after the forecast is issued). Forecasts skill scores combine start times by calendar month and across years 1999 to 2010.

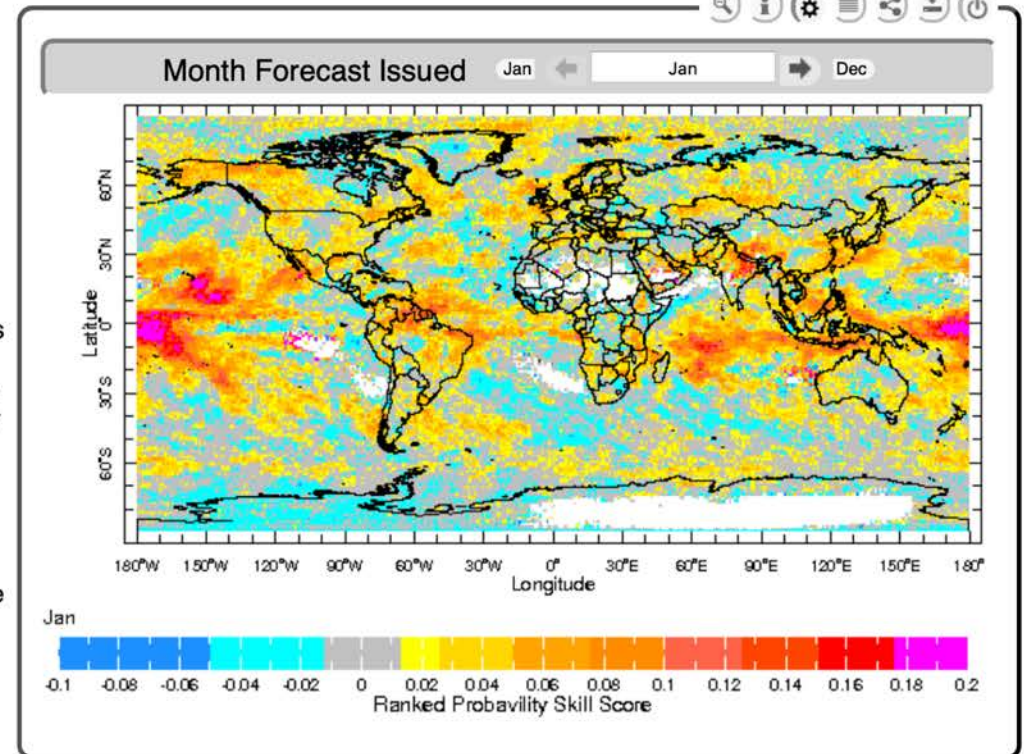
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These skill scores diagnostics maps give a sense of where and when (issued which months of the year and for which weekly lead times) subseasonal forecasts may have the potential to provide useful information.

The actual forecasts, of which these skill scores are measuring the historical performance, are to be found in the Experimental Precipitation Subseasonal Forecast Maproom.

Skill scores definitions:

- **RPSS:** Ranked Probability Skill Scores (RPSS; Epstein (1969); Murphy (1969, 1971); Weigel et al. (2007)) are used to quantify the extent to which the calibrated predictions are improved compared to climatological frequencies. RPSS values tend to be small, even for skillful forecasts. The approximate relationship between RPSS and correlation being such that a RPSS value of 0.1 corresponds to a correlation of about 0.44 (Tippett et al. 2010).
- **Spearman Ranked Correlation:** the Spearman Anomalies Correlation Coefficient corresponds to the ranked correlation between MME forecasts and observed anomalies, which is particularly appropriate to verify probabilistic forecast
- **ACC:** the Anomalies Correlation Coefficient is the correlation between MME forecasts and observed anomalies.



<http://iridl.ldeo.columbia.edu/maproom/Global/ForecastsS2S/s2sskill.html>

How do IRI's Subseasonal and Seasonal forecast skills compare?

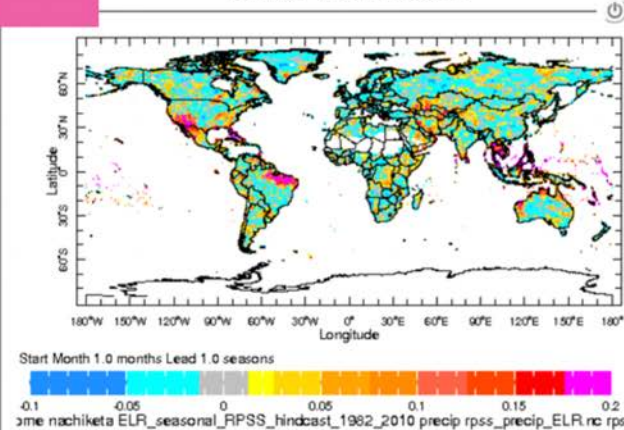
Precipitation
Forecasts issue in Jan
Hindcasts

**Seasonal
Skill**

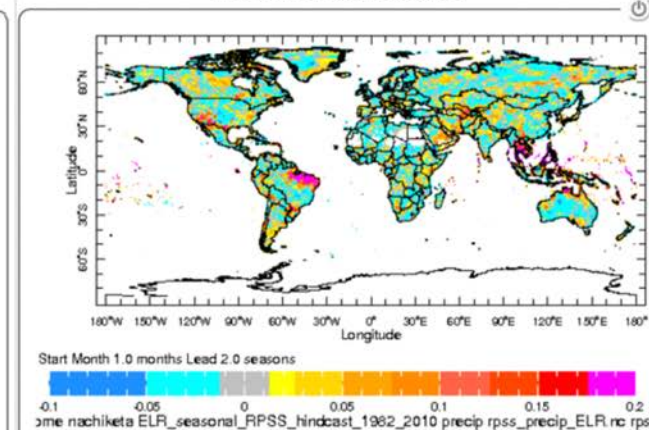
*Subseasonal
forecasts are
skillful over
broader areas
than seasonal
ones, but skill-
maxima are
lower*

**Subseasonal
Skill**

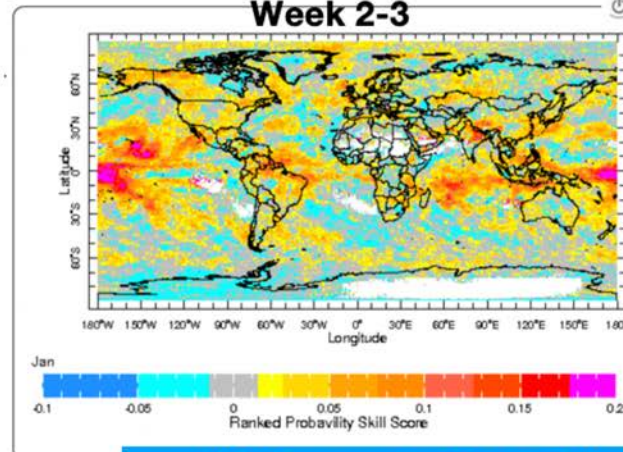
FMA from Jan



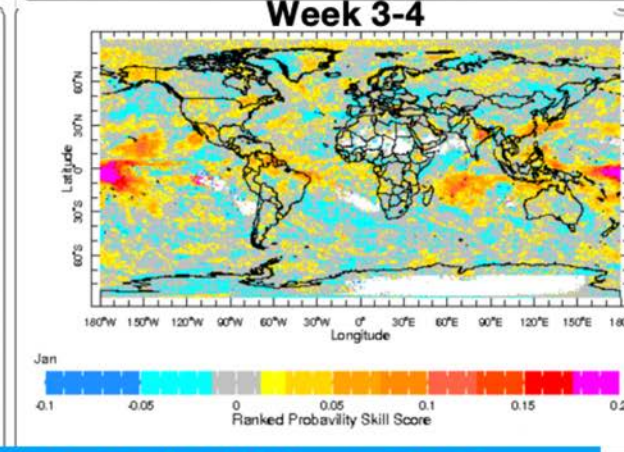
MAM from Jan



Week 2-3

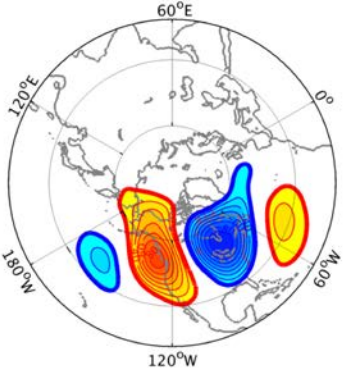


Week 3-4

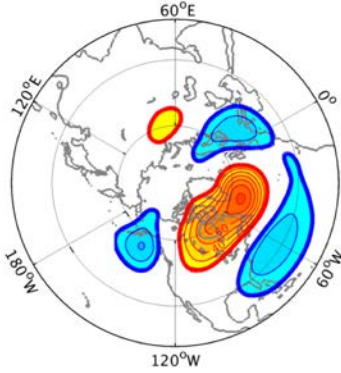


Skill is expressed here in Ranked Probability Skill Score which is a multi-category generalization of the Brier Skill Score

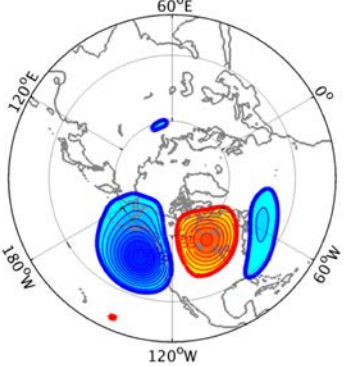
a) WR 1: West Coast Ridge



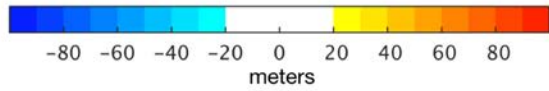
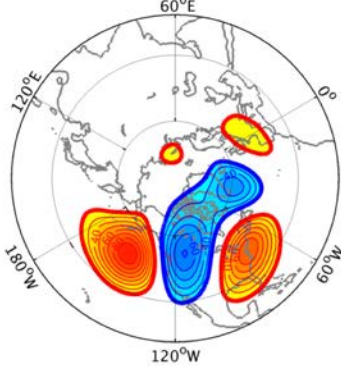
b) WR 2: Greenland High



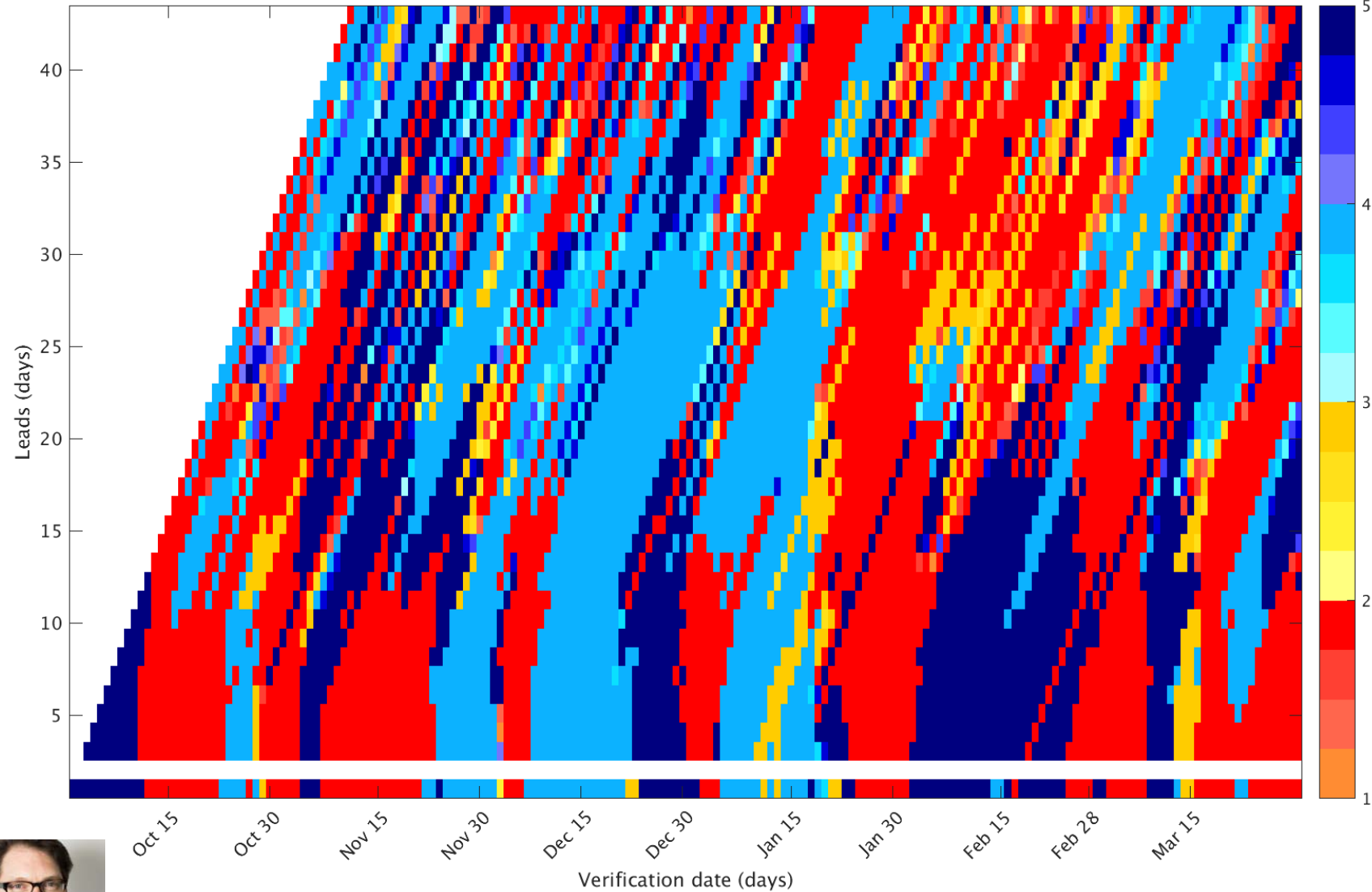
c) WR 3: Pacific Trough



d) WR 4: Pacific Ridge



CFSv2 daily winter WRs forecast from Oct 1 to Mar 31 2019



Vigaud, Robertson, Tippet (2018)
Special thanks to Jing Yuan!!

<http://wiki.iri.columbia.edu/index.php?n=Climate.S2S-WRs>



IRI Forecast Systems update

A stylized background map of the Pacific Ocean. A prominent band of yellow and red colors stretches from the bottom left towards the top right, likely representing a forecast of a climate event like an El Niño or La Niña. The colors transition from yellow to red, indicating varying intensities of the forecast.

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