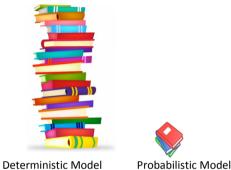
Evaluating a New Calibration Method for Seasonal Probabilistic Prediction for Indian Summer Monsoon

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Motivations

A single deterministic rainfall forecast is not sufficient for predicting seasonal Indian summer monsoon rainfall (ISMR) which is characterized by large variability. The user community should be given probabilistic forecasts that convey the inherent uncertainty within the prediction.

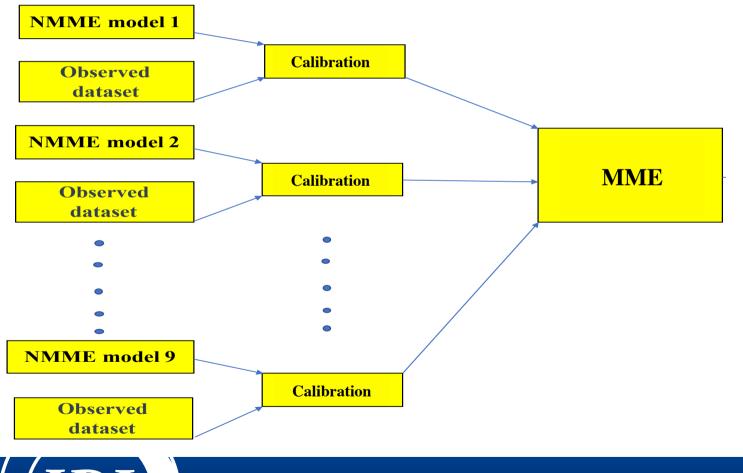


- Though a plethora of study exists to make a deterministic model for predicting ISMR, only a few studies have described the probabilistic prediction system.
- Common approach to make such probabilistic forecast
 - 1) Counting ensemble members of GCMs (Uncalibrated).
 - 2) Combine (average, multiple linear regression) multiple GCMs (Uncalibrated) and convert in to probabilistic space using Gaussian distribution.



Objective

Evaluation of a calibrated probabilistic forecast system using NMME models in a non-Gaussian framework for ISMR.





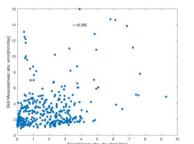
NMME datasets in IRI DL	Model lists
https://iridl.ldeo.columbia.edu/SOURCES/.Models/.NMME/	
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The lead-1 (using initial conditions of May) hindcast runs for mean rainfall of JJAS spanning over 1982-2010 is used.

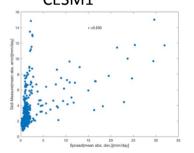


Representation of Uncertainty in GCMs

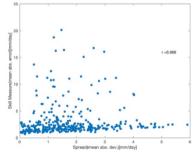
Spread-Error relationship CFSv2



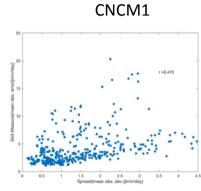
CESM1



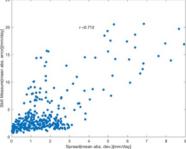
GFDL3

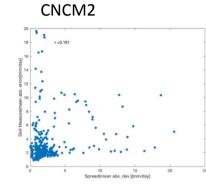




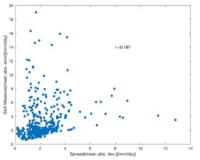


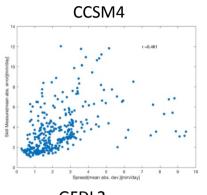
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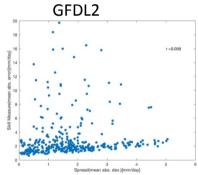




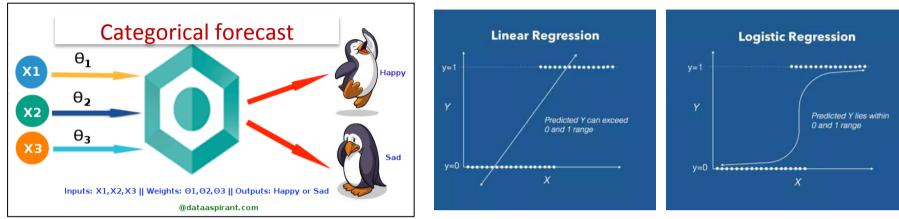
GFDL1







Calibration Method



Logistic Regression

Logistic regression is well famous method to make probability forecast

Where p is the (cumulative) probability of not exceeding the quantile q

Logistic regression (LR), a nonlinear regression method where probability itself can be considered as the predictand rather than a measurable physical quantity, is an alternative model for Gaussian approach.



Calibration Method

$$\ln\left[\frac{p}{1-p}\right] = f(x) + g(q) \quad \text{Where}$$

Limitations:

- * Probabilities of different categories estimated by fitting separate equations for selected predictand quantile thresholds (q), yielding a collection of threshold probabilities rather than full forecast probability distributions
- ✤ However, the most problematic consequence of separate equations for different predictand thresholds is that forecasts derived from the different equations are not constrained to be mutually consistent.

Extending Logistic Regression:

Extending LR (ELR) by including the predictand threshold as an additional predictor (link function g itself function of the quantile q), allows the derivation of full predictive distributions to avoid the problem of potentially incoherent forecast probabilities (Wilks, 2009).

Cumulative probability for a smaller predict d threshold cannot be larger than the probability for a larger threshold.

Mola. Wilks, D., 2009: Extending logistic regression to provide full-probability-distribution MOS fore- casts. (MPLOF. Spp:), 18, 381-101, 1979, USING 422. I logistic regression is por a 1988) (SMBSPVer8



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JKIII ASSESSITIETIL. Counting Ensemble vs

ELR

Rank Probability Skill Score (RPSS) **Counting Ensemble** ELR 40° N 30° N 30 20° M 10[°] N 10 70[°] E 80[°] E 90[°] E 100 1 70[°] E 80[°] E 90[°] E 100[°] E 0.85 0.8 0.75 0.7 RPSS 0.65 0.6 0.55 2009 982 003 005 900 00 ---Counting Ensemble ---ELR

Reliability Diagram

Counting Ensemble

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For a single GCM (GFDL3)



No Skill

No Resolution

0.8

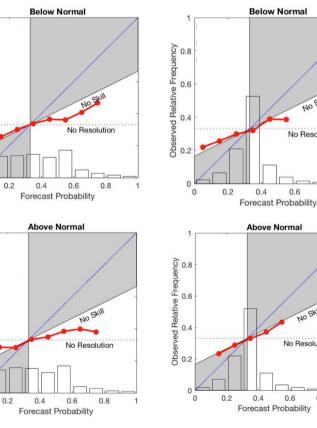
1

0.6

No Skill

No Resolution

0.8

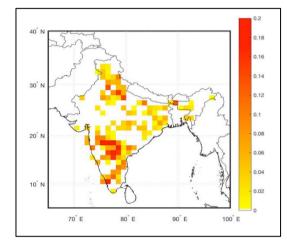


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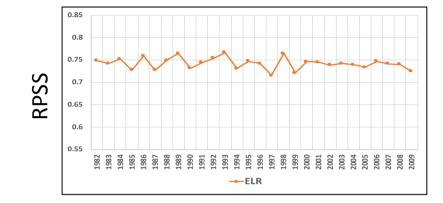
0.6

Skill Assessment: ELR based MME

Rank Probability Skill Score (RPSS)

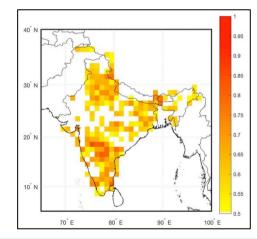






Above Normal **Below Normal** 2 0.8 2 0.8 0.6 0.6 e 0.4 0.4 No Resolution No Resolution q 0.2 Q 0. 0.2 0.4 0.6 0.8 0.4 0.6 0.8 0.2 1 0 Forecast Probability Forecast Probability

Generalized ROC Score (GROC)

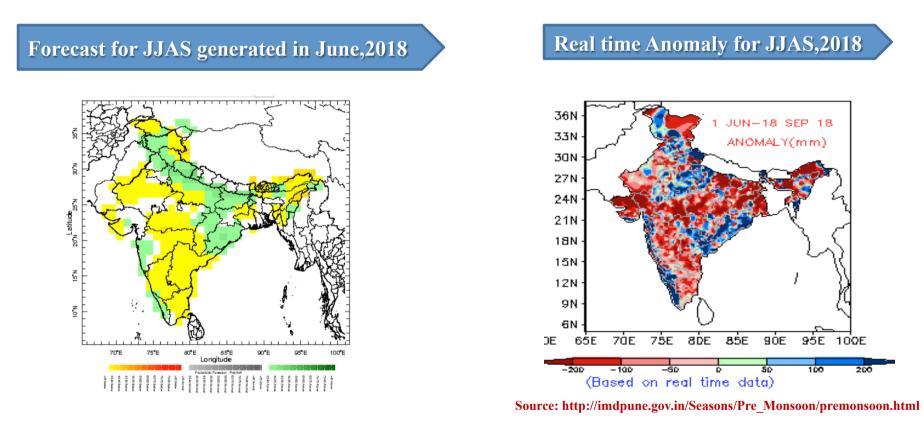


Courtesy: Simon Masson and Ángel Muñoz



Real time Forecast (tercile) and Verification

Real-time experimental probability forecast generated by above stated calibration method for monsoon 2018.



✤ The spatial pattern of deficit and excess rainfall well capture by Forecast.



Real time Forecast (full distribution) and Verification

Forecast for JJAS generated in June,2018

CDF

200 300 Precipitation [mm]

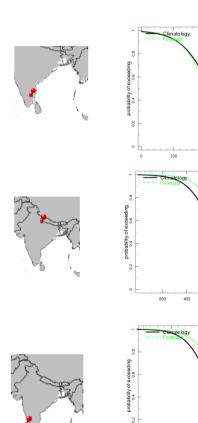
Precipitation [mm]

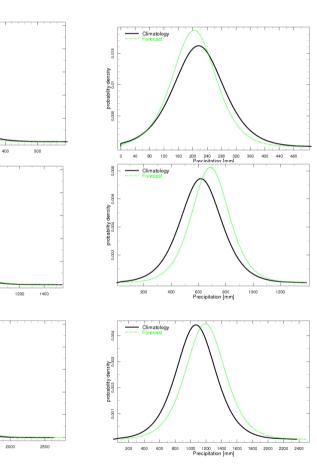
1000 1500 Precipitation [mm]

500

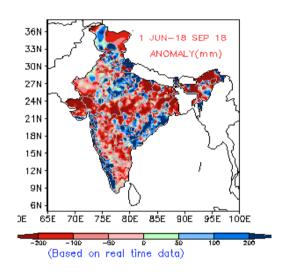
1000

Real time Anomaly for JJAS,2018





PDF





Remarks

ELR based forecast show some good hope for making reliable probability forecast.

The proposed calibration method also make a flexible forecast format (full distribution rather tercile) that allows users to glean information from those part of forecast distribution what matters most to them such as the probability of extremely dry/wet conditions.

Room for Improvement: ELR based forecast is less sharp.





Thanks! nachiketa@iri.columbia.edu

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