# Assessment of conditional forecast skill for Brazilian precipitation

Amulya Chevuturi | Liang Guo | Matthew Young | Nick Klingaman Steve Woolnough | Chris Holloway | Emily Black | Pier Luigi Vidale

a.chevuturi@reading.ac.uk

Workshop on Predictability, dynamics and applications research using the TIGGE and S2S ensembles (2-5 April 2019)







## Objectives

Diagnosing and Understanding Brazilian Subseasonal Tropical and Extratropical Processes (DUBSTEP)

**Aim:** Assessing **prediction skill for sub-seasonal rainfall** variability in Brazil, including **conditional skill** based on large-scale atmospheric circulations.

- S2S prediction skill for Brazilian precipitation
- Conditional forecast skill for Brazilian precipitation; teleconnections from large-scale climate variability

#### Motivation





#### S2S Data

Model	Native Resolution	Years available	Initialisation frequency	Ensemble members	Reference
UKMO (GloSea5-GC2)	0.8° x 0.5°	1993–2015	1 <sup>st</sup> , 9 <sup>th</sup> , 17 <sup>th</sup> , 25 <sup>th</sup> of each month	7	MacLachlan et al., (2015)
NCEP (CFSv2)	~100 km	1999–2010	Daily	4	Saha et al., (2014)
ECMWF (IFS)	0.7°	1998–2017	Twice per week	11	Vitart et al., (2004)

Evaluation of S2S skill for weekly averaged precipitation in all 4 seasons (DJF, MAM, JJA, SON)

- Lead times of 1-5 weeks
- Weekly averaging increases skill by reducing noise from daily rainfall variability.
- Observations: GPCP (satellite + gauge)
- Analysis Horizontal Resolution: 1.5° × 1.5°
- Challenge: Comparing models with different initialisation dates and ensemble sizes
  - Lagged ensembles for NCEP (last 7 days) and ECMWF (last 3 forecasts)
- Common time period of analysis: 1999-2010
- Evaluation of effects of the ENSO and MJO on the precipitation over Brazil in S2S for DJF
  - Oceanic Nino Index (ONI) is used for ENSO
  - Wheeler–Hendon real-time multivariate (RMM) index for MJO

### S2S Rainfall Forecast Skill

Bias Week 1









Anomaly correlation coefficient (ACC) Week 1



÷

## S2S Rainfall Forecast Skill

Bias Week 5



UKMO

NCEP

ECMWF







Anomaly correlation coefficient (ACC) Week 5



4

#### Forecast Skill over Brazil (40°S-20°N, 90-20°W)

- In Bias, ECMWF has dry bias during wet seasons but wet bias during the dry seasons.
- In terms of RMSE, UKMO's skill deteriorates quicker than the other two systems.
- All models show **lowest ACCs in JJA**, which suggests issues in predicting rainfall during the dry season.
- Skill is generally **highest in ECMWF** and UKMO and **lowest in NCEP**
- Models lose useful skill (ACC > 0.5) after Week 2.



# Brier Skill Score (BSS)



• Higher forecast skill for above and below normal precipitation categories in week 1 than normal precipitation category.

## **DJF Precipitation Anomaly Bias - ENSO**



#### Forecast Skill over Brazil (40°S-20°N, 90-20°W)

- Models underestimate:
  - dry anomalies during El-Niño
  - wet anomalies during La-Niña
- RMSE are lowest in ECMWF and the highest in UKMO.
- Skill is lower during El-Niño than in Neutral or La-Niña phases in all models.



## MJO

- MJO in Phases 8, 1
  and 2 leads to
  precipitation
  anomalies that are:
  - wet over SACZdry over SESA
- The precipitation anomalies are reversed during Phases 3, 4 and 5.
- Precipitation
  suppressed in Phase 0 (inactive MJO).



#### GPCP Precipitation Anomalies (DJF)





4.0

- 3.5

- 3.0

- 2.5

- 2.0

- 1.5

- 1.0

- 0.5

- 0.0

-0.5

-1.0

-1.5

-2.0

-2.5

- -3.0

- -3.5

-4.0

# Precipitation Anomaly Bias – Week 1

UKMO



Phase 0







Phase 3 Phase 4 Phase 5





Phase 6



Phase 7



Phase 8



- Strongest errors are observed during MJO **Phase 1**: dry bias over NWSA & SESA and wet bias over eastern Brazil (except ECMWF).
- Phase 8 has a wet bias over central Brazil.
- Phases 4 and 5 show a wet bias over northern SA and somewhat dry bias over the central Brazil.

#### Forecast Skill over Brazil (40°S-20°N, 90-20°W)



- Forecast skill decreases with lead time in all MJO phases.
- Forecast skill highest in Week 1 for the observed MJO in Phase 1.
- Forecast skill decreases with lead time, and models usually have no skill beyond Week 2.

### Conclusions

Models exhibit large precipitation biases

• Errors might be associated with tropical convection, orography and regional SSTs

Models show lowest skill in JJA season

• Indicating issues in predicting rainfall during the dry season

Models underestimate the strength of the MJO and ENSO teleconnections

• S2S rainfall predictions are better during Neutral ENSO conditions

Precipitation errors are well established within the first week of the forecasts

• Skill declines with lead time and models mostly lose useful skill after Week 2

Future work includes

- Forecast skill analysis for Brazilian operational model
- Analysing model skill using CHIRPS dataset over different sub-regions
- MJO teleconnections analysed using model MJO indices