



**CAFE**  
Climate Advanced Forecasting  
of sub-seasonal Extremes



# *Ensemble forecast of the Madden Julian Oscillation using a stochastic weather generator based on analogs of Z500*

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Machine Learning Workshop

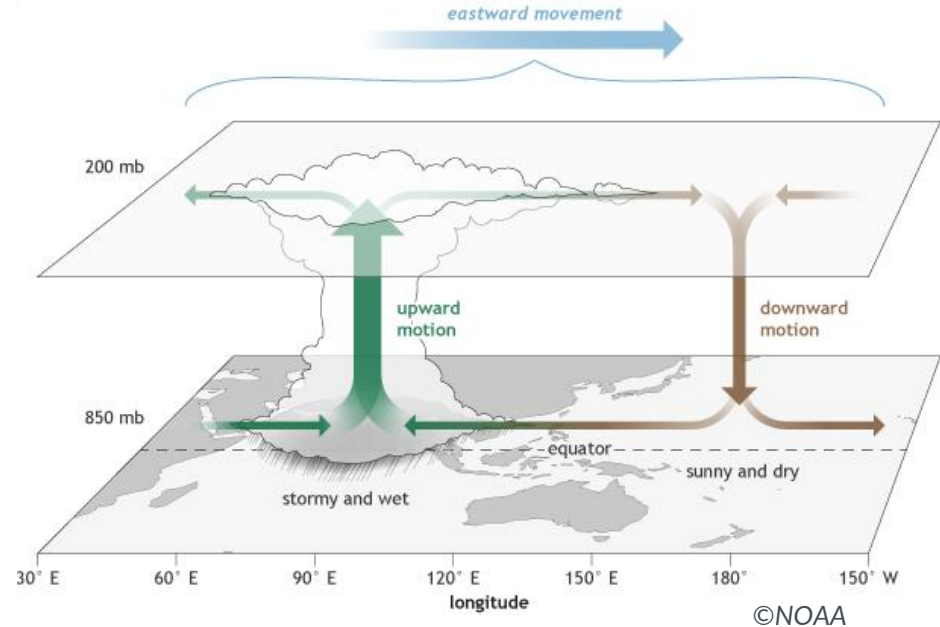
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# Motivation

## Madden Julian Oscillation

- Major sub-seasonal fluctuation in tropical weather,
- Source of intra-seasonal fluctuations in the Indian monsoon,
- Impacts the tropical & extra-tropical weather.



# Motivation

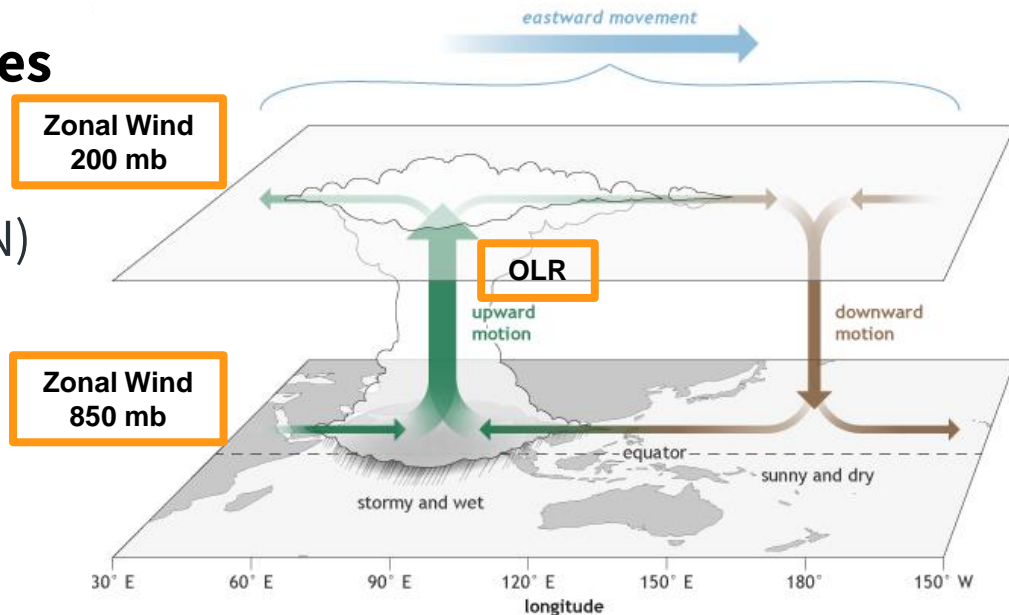
## Madden Julian Oscillation indices

RMM1 & RMM2 (between 15°S – 15°N)

EOF

- Zonal Wind at 850 mb
- Zonal Wind 200 mb
- OLR

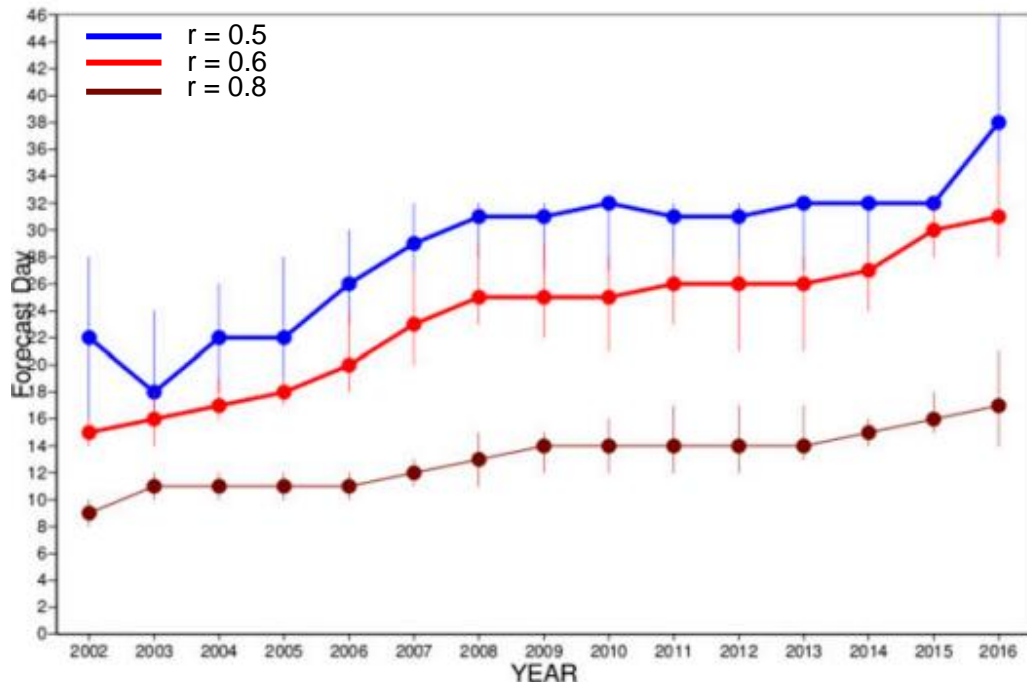
→ Amplitude & phase



©Wheeler and Hendon, 2004 (Monthly Weather Review)

# Current state of Art

## Evolution of the MJO skill scores



©Kim et al, 2018 (J. Climat) & Vitart 2014

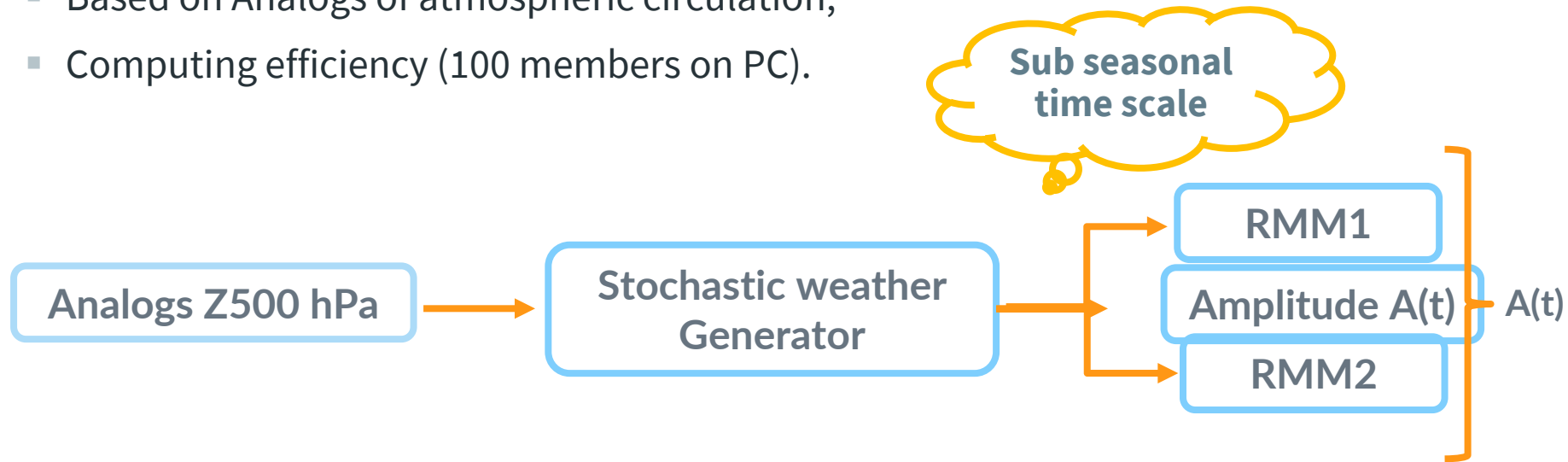
- MJO indices (RMM1 & RMM2)
- **Forecast of MJO  $\approx$  46 days**

## Scientific challenge

Would it be possible to forecast the MJO using a probabilistic approach? for a sub seasonal lead time?

# Our forecast approach

- Analog weather generator developed by P. Yiou (2013) ,
- Tested on **NAO** (Yiou and Déandreis, 2019), **precipitation** (Krouma et al.,2021)
- Based on Analogs of atmospheric circulation,
- Computing efficiency (100 members on PC).



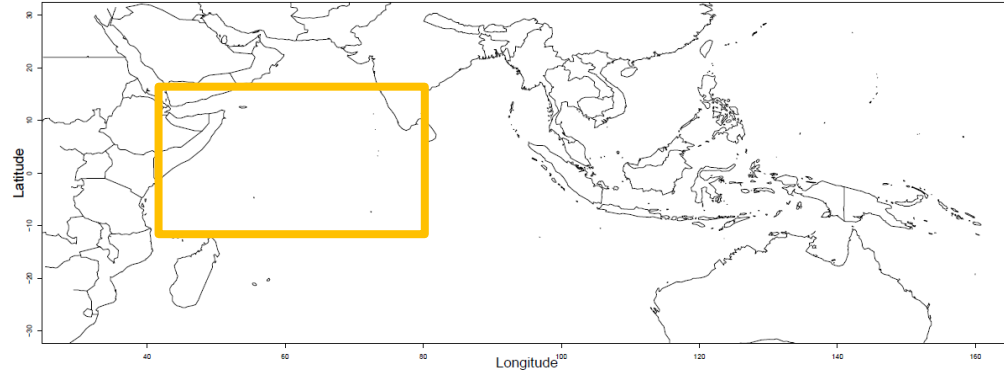
# Step 1. Compute Analogs of Geopotential Hight 500 mb

## Data.

- **Z 500** (NCEP) 1979 to 2020
- resolution  $2.5^\circ \times 2.5^\circ$

## Analog.

- Calendar distance between  $(d, d') \leq 30$
- $y_d \neq y_{d'}$
- $Min (D_{(d,d')}) = [\sum_X [Z500(x, d) - Z500(x, d')]^2]^{1/2}$



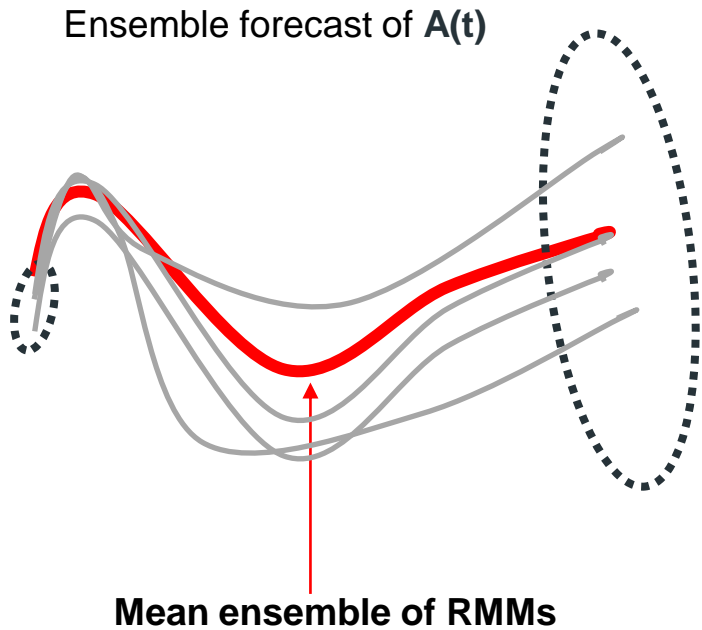
Amplitude  $A(t)$

RMM1 & RMM2

# Step 2. Stochastic weather generator (SWG)

- Generate  $n_{trajectories} = 100$  random trajectories,
- Simulate Amplitude from trajectories of Z500 for lead times **T= 1 to 90** days.

# Assess of the Forecast skill



## 1. Probabilistic score.

### Continuous Rank Probabilistic score

$$CRPS(P, x_a) = \frac{1}{n} \sum_{t=1}^n \int_{-\infty}^{+\infty} (P(x) - P_a(x))^2 dx$$

$$CRPSS = 1 - \frac{CRPS}{CRPS_{Ref}}$$

CRPSS = 1 → Perfect forecast

©Zamo and Naveau, 2018 (Mathematical Geosciences)

## 2. Scalar score.

### Bivariate Correlation & RMSE

$$COR(\tau) = \frac{\sum_{t=1}^N [a_1(t)b_1(t, \tau) + a_2(t)b_2(t, \tau)]}{\sqrt{\sum_{t=1}^N [a_1^2(t) + a_2^2(t)]} \sqrt{\sum_{t=1}^N [b_1^2(t) + b_2^2(t)]}}$$

$$RMSE(\tau) = \sqrt{\frac{\sum_{t=1}^N [ |a_1(t) - b_1(t, \tau)|^2 + |a_2(t) - b_2(t, \tau)|^2 ]}{N}}$$

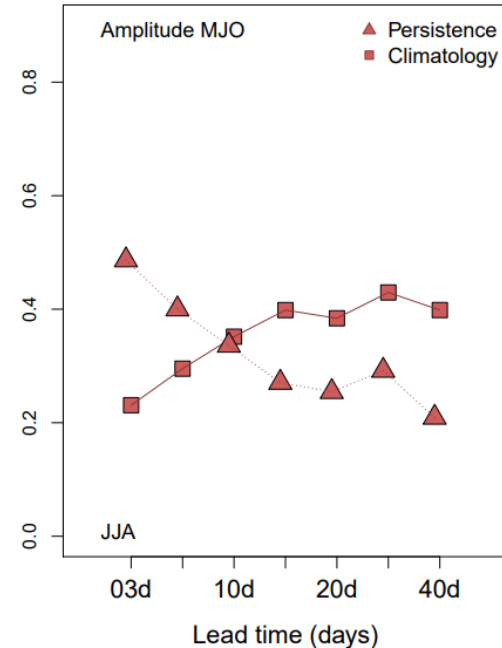
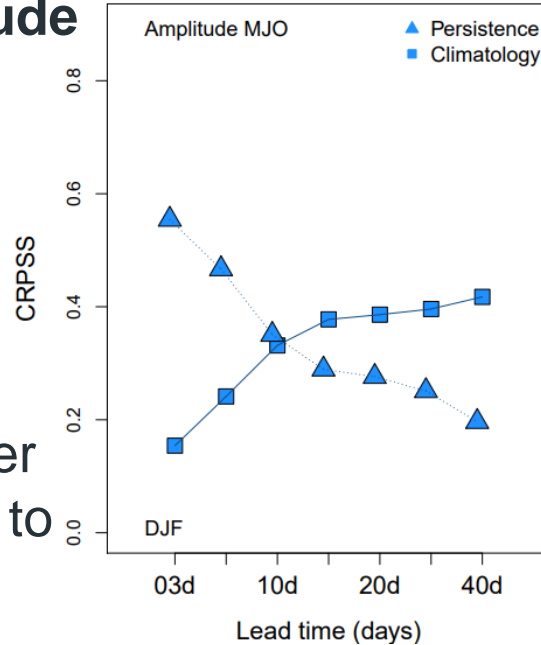
→ RMSE =  $\sqrt{2}$  & COR = 0.5

©Kim et al, 2018 (J. Climat) & Rashid et al., 2011 (Clim Dyn)

# Evaluation of the ensemble forecast

## Skill score for the MJO Amplitude

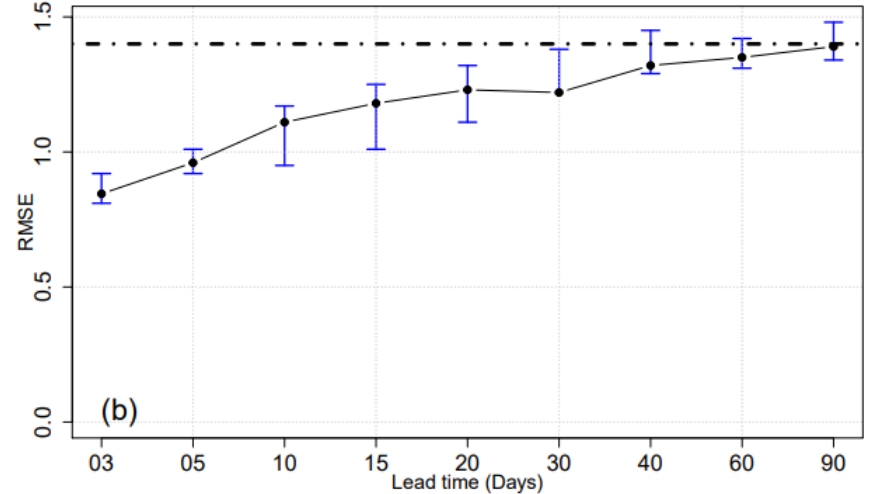
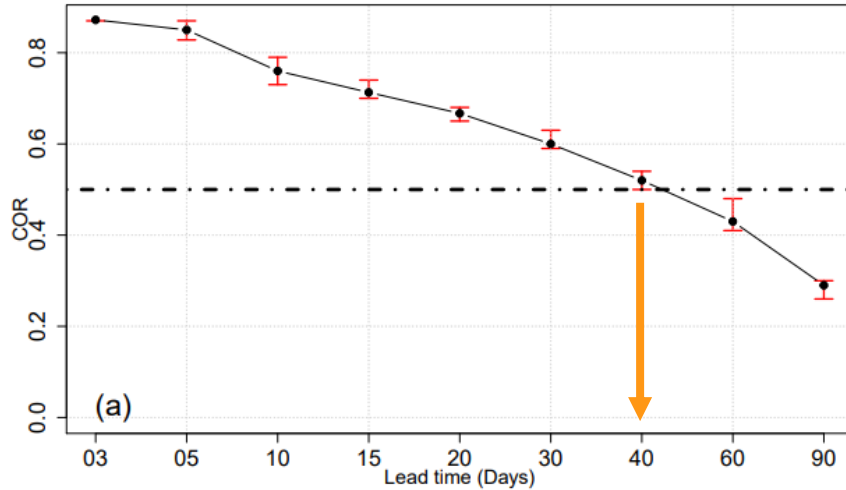
- CRPSS shows a positive improvement until 40 days for DJF and JJA.
- The correlations indicate the spread across the 100-member ensemble forecasts are equal to **0.62**.





# Evaluation of the ensemble mean forecast

## Skill score for the RMMs

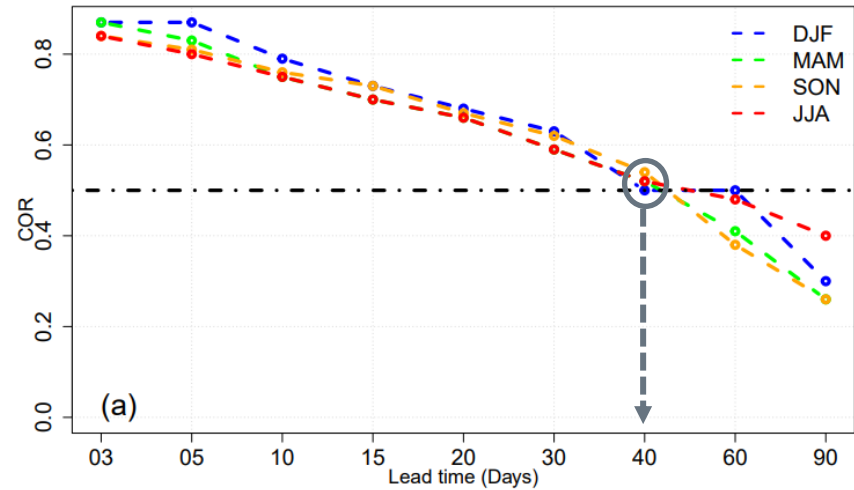
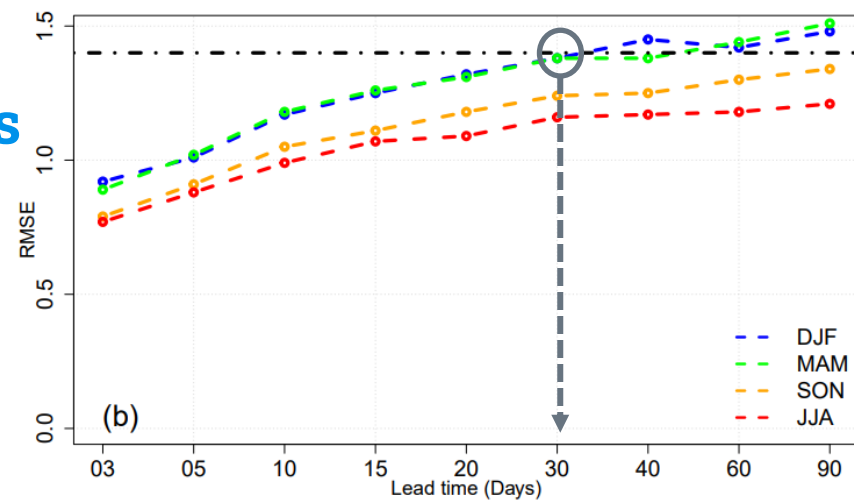


➔ The SWG forecast of the RMMs shows good skill until 40 days.



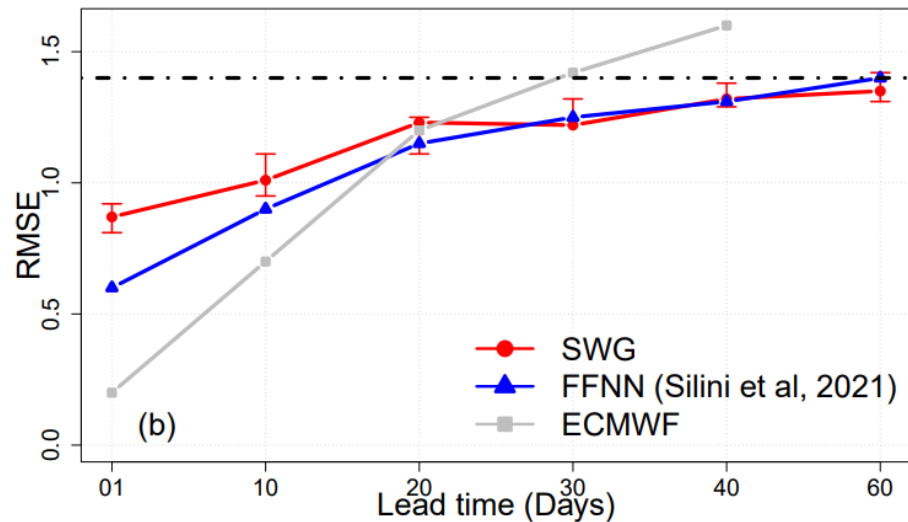
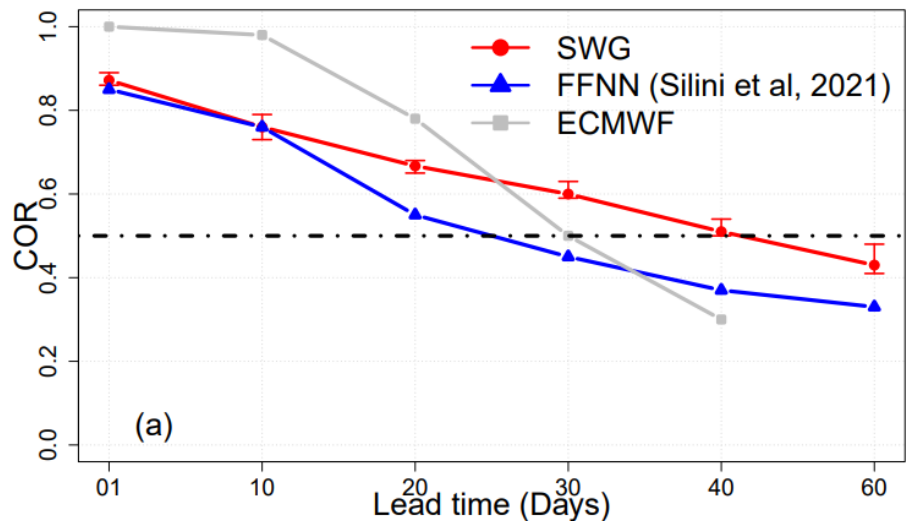
## Sensitivity of the forecast to seasons

- ◎ The forecast for DJF and MAM has a good skill until 30 days,
- ◎ The forecast for SON and JJA could reach 40 days.



## SWG forecast Vs other forecasts

- ⊙ ECMWF Forecast is more reliable for small lead time ( $\approx 10$  days).
- ⊙ Silini et al. (2021) forecast has the same skill until 10 days.
- ⊙ SWG forecast could reach 40 days.



# Conclusions

- ◎ The performance of analogs weather generator shows skill to forecast the MJO indices from analogs of Z 500 mb.
- ◎ Atmospheric circulation is useful for forecasting MJO.
- ◎ The comparison with ECMWF forecast confirmed the good performance of SWG quantitatively and qualitatively.

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[Ensemble forecast of the Madden Julian Oscillation using a stochastic weather generator based on circulation analogs](#)

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