The Effect of the Mean Bias Removal on MJO Forecast Skill

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Acknowledgement : CPC Monsoon Desk Program

Motivated by using the operational MJO forecasts



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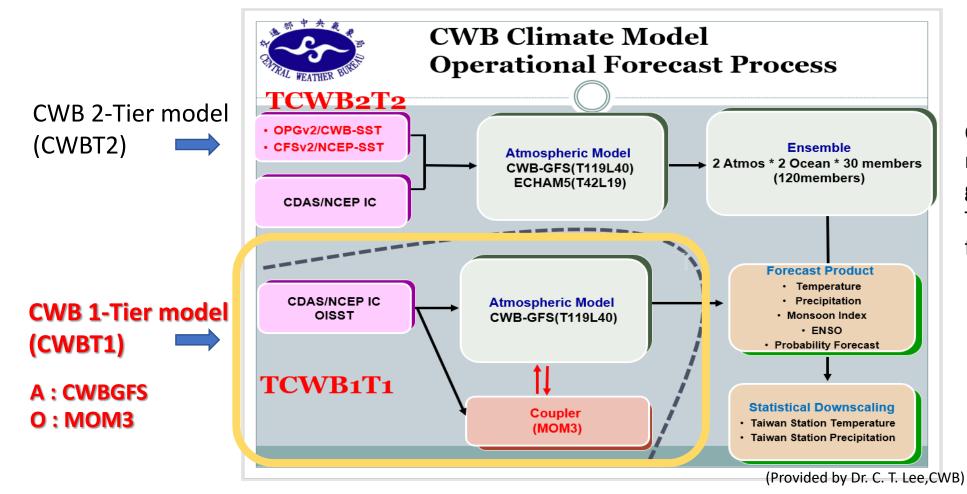
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Dynamical Model MJO Forecasts Supported by: U.S. CLIVAR International CLIVAR WCRP - WWRP/THORPEX YOTC MJO Task Force Phase Plots of MJO Index Forecasts NCPE **UKMA** NCPB ICPC NCES UKME CMET ECMF ECMN CPTC JMAN TCWB

BOMM

Taiwan Weather Bureau has coupled models for S2S forecasts

This work used CWB 1-tier model to discuss the effect of the bias removal on MJO forecasts.



CWB short-term climate models have provided guides for supporting Taiwan month/seasonal forecasts services for years.

More about CWBT1 : Poster #60 : Wu et al. Taiwan CWB 1-Tier model : Hindcast Analysis and Forecast Verification

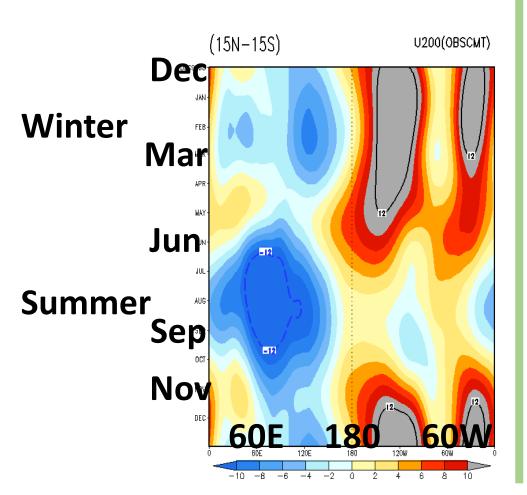
Take U200(15N-15S) for example :

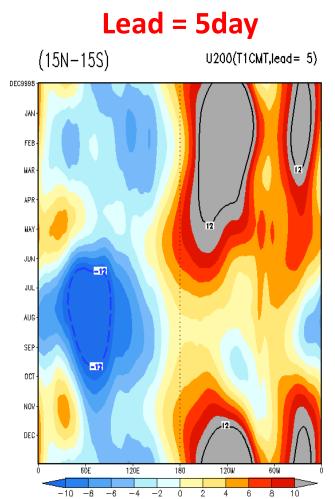
A look at CWB1T model climatology

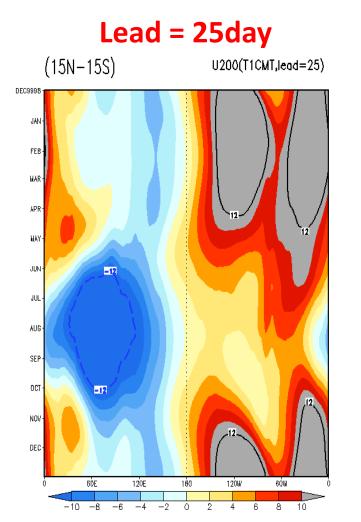
M_climatology (t, lead)

Based on 2001-2017

O_climatology(t)

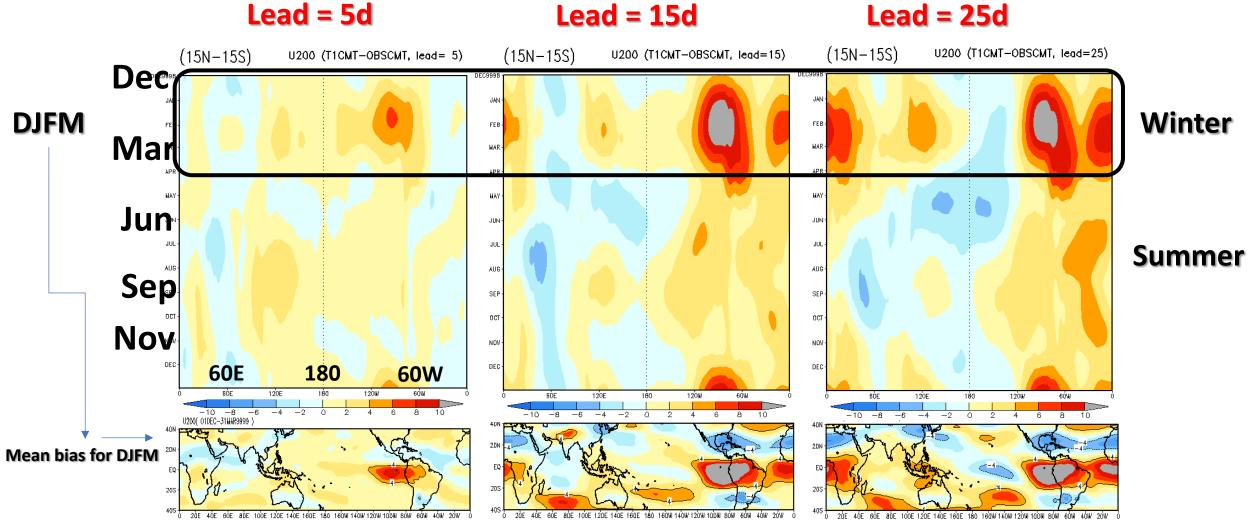






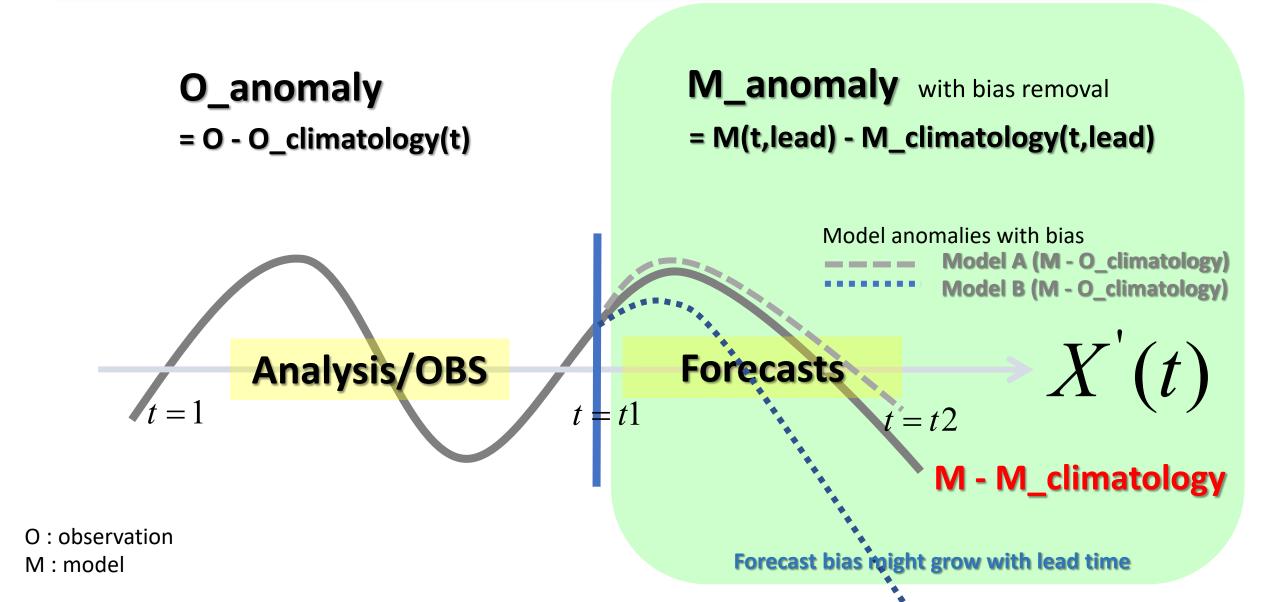
Model : CWB 1-Tier

Model bias(t, lead) = M_climatology(t,lead) – O_climatology(t)



Model mean bias could change with season and grow with the lead time

It's about the anomalies when comes to model forecasts.



Bias removal needs model climatology.

But sometime we just have the forecasts but without the model climatology......

http://www.cpc.ncep.noaa.gov/products/precip/CWlink/MJO/CLIVAR/clivar_wh.shtml

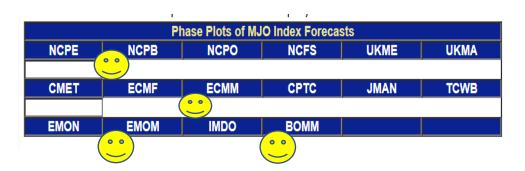


National Weather Service

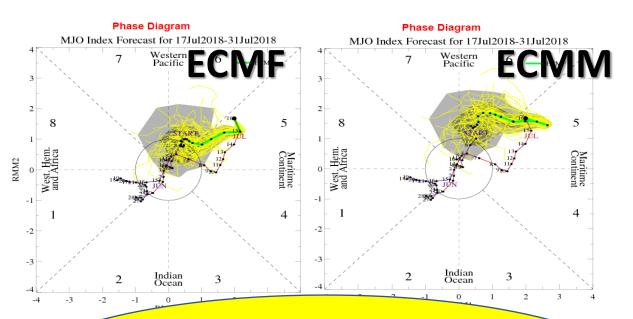
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Dynamical Model MJO Forecasts

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M_Anomaly = M – O_climatology (with model bias) M_Anomaly = M – M_climatology



How to interpret properly the biased and unbiased MJO forecasts? How different they are?

Q1 : How is the effect of the model mean bias on RMM forecast performance ? Why we need bias removal ?

No bias removal M_Anomaly = M(iy, t, lead) – O_climatology(t) With bias removal M_Anomaly = M(iy, t, lead) – M_climatology(t, lead)

Q2 : Model climatology (or mean bias) is usually got by a long period of hindcast data(eq. 30yrs or 20yrs). What can we do if we only have few years data ?

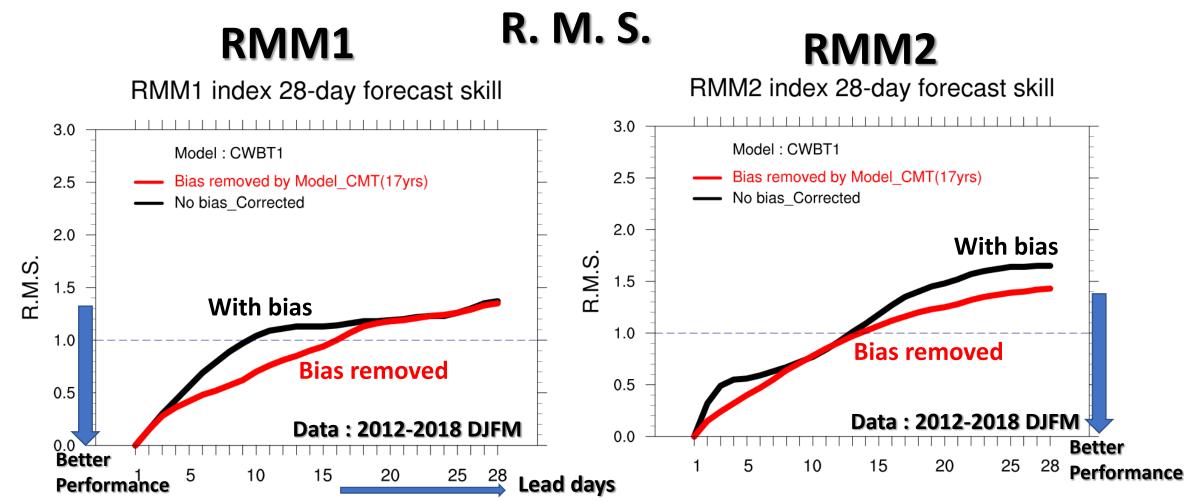
Can we work from O_climatology and mean bias?

M_Anomaly = M(iy,t,lead) – (O_climatology(t) + Mean bias(t,lead))

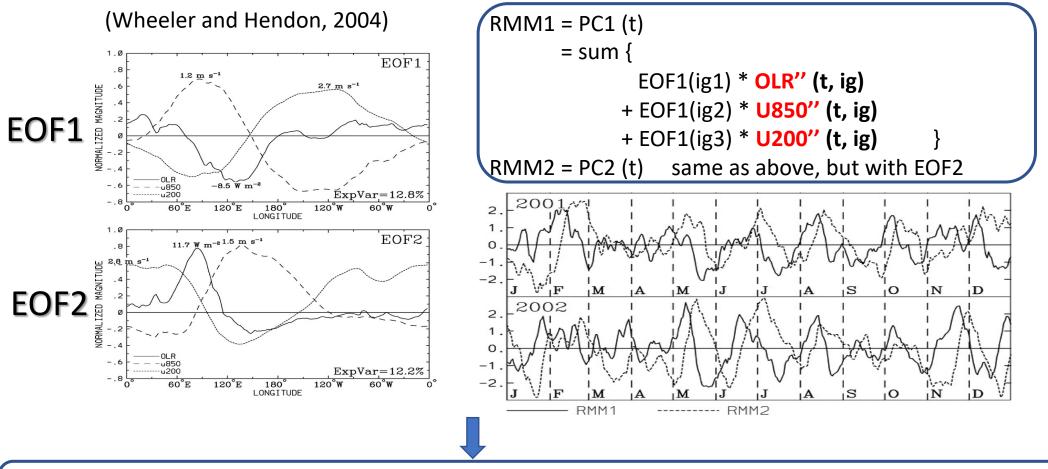
Can we work out a reasonable approach to estimate the model mean bias?

Q1 : How is the effect of the model mean bias on RMM forecast performance ?

Model : CWB 1-Tier



RMM index : the 2 leading PCs of the combined EOF from 3 variables(U200/U850/OLR)



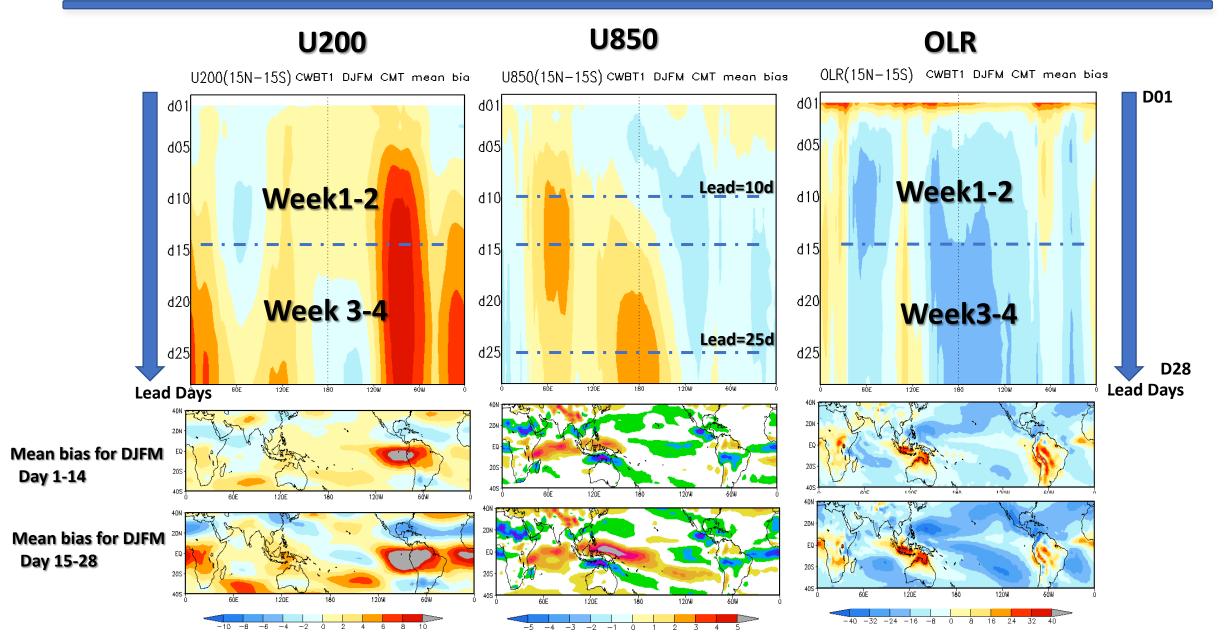
The contribution of mean bias on RMM can be estimated by :

RMM_BiasEffect = sum { EOF*OLR_bias + EOF*U850_bias + EOF*U200_bias } EOF : Fixed weighting function

➔ If model bias behavior changes with lead time, the bias effect would also do.

A Look at CWBT1 model mean bias

Season : DJFM

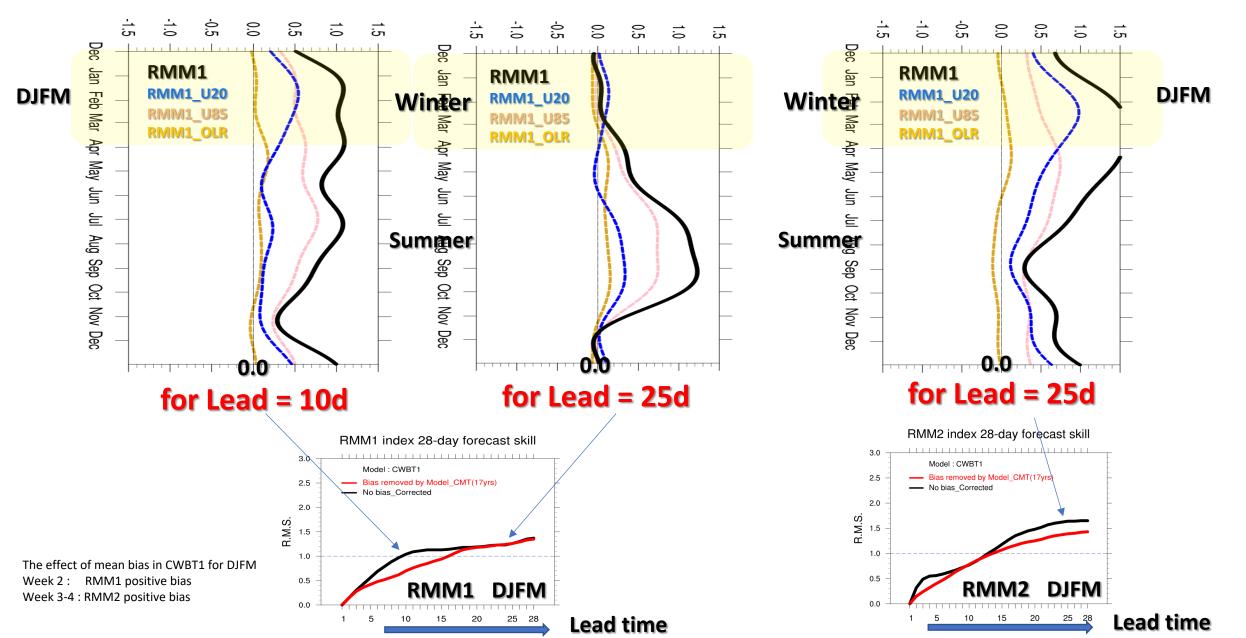


Mean bias contribution on RMM1

sum { EOF1*OLR_bias + EOF1*U850_bias + EOF1*U200_bias }

Mean bias contribution on RMM2

sum { EOF2*OLR_bias + EOF2*U850_bias + EOF2*U200_bias }

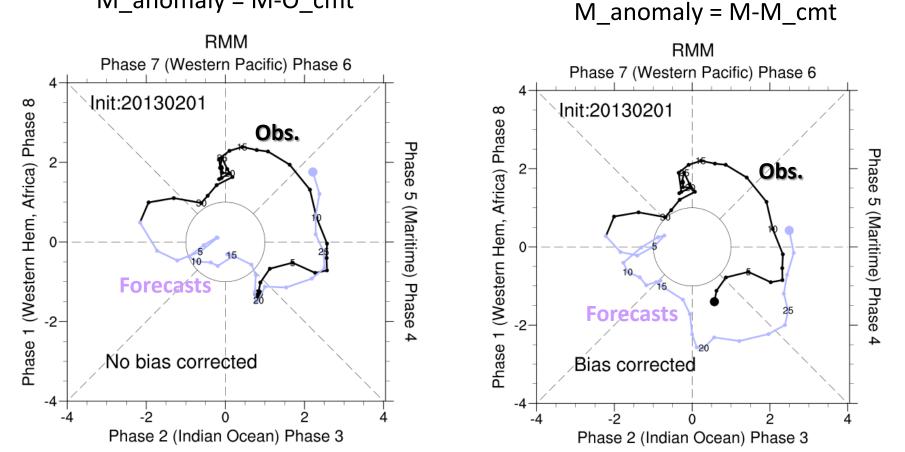


Model : CWB 1-Tier

An example :

No bias correction

M_anomaly = M-O_cmt

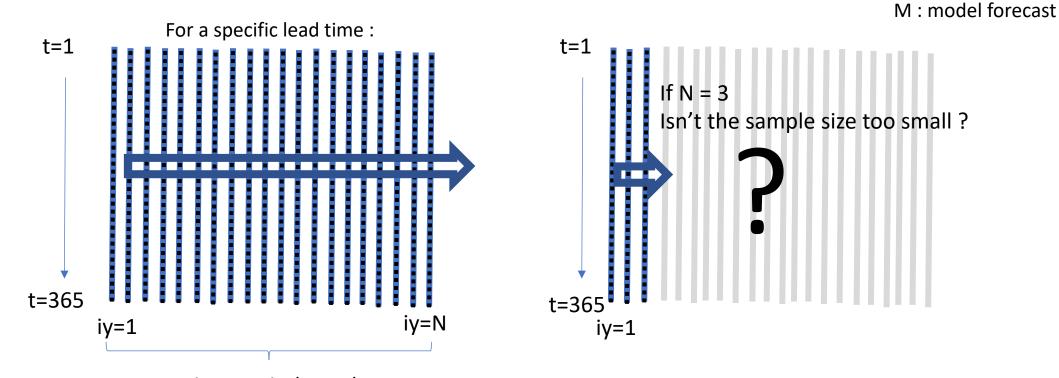


Bias corrected

Without the model bias information, we hardly can properly interpret the forecasts. It's better always try to pursuit bias corrected forecasts.

Q2 : Model climatology (or mean bias) is usually got by a long period of hindcast data(eq. 30yrs or 20yrs). What can we do if we only have few years data?

M_climatology (t, lead) = mean(M (iy,t,lead))



Muti-year Hindcast data

Alternative approach for bias corrected M_Anomaly

(A) Standard approach :

M_Anomaly = M(iy,t,lead) – M_climatology(t,lead)

(B) The Proposed alternative approach (when lack of hindcast):

M_Anomaly = M(iy,t,lead) – (O_climatology(t) + Mean bias(t,lead))

For a specific lead time :

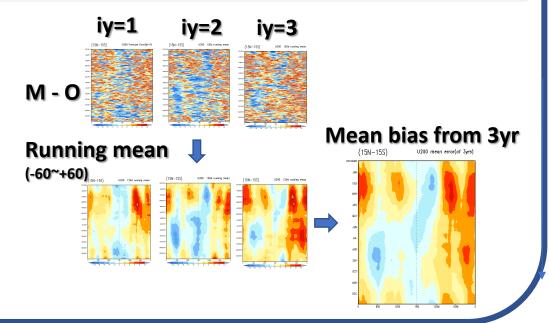
t=1

t=365

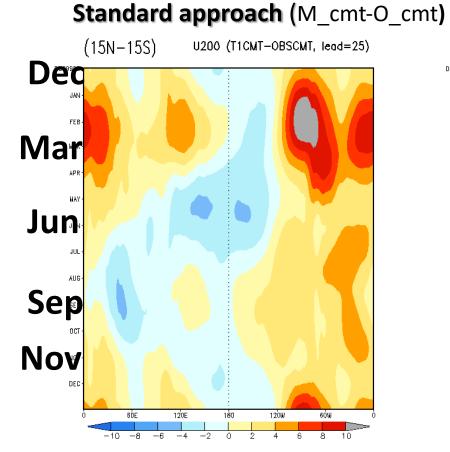
Get low frequency errors for each year : M(iy, t, lead) – O(iy, t) = Error(iy, t, lead) MovingAvg[(Error)] = Error_lowfreq(iy, t,lead)

Estimate the mean bias by averaging

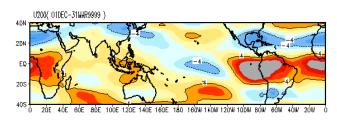
mean(Error_lowfreq) = Mean bias(t, lead)

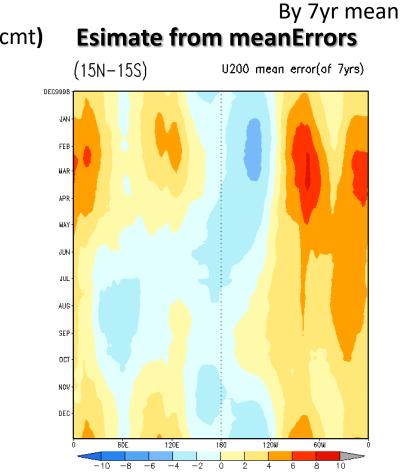


The CWBT1 mean bias (take lead 25d for example)

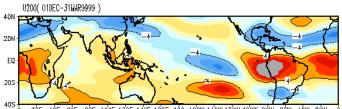


Mean bias for DJFM



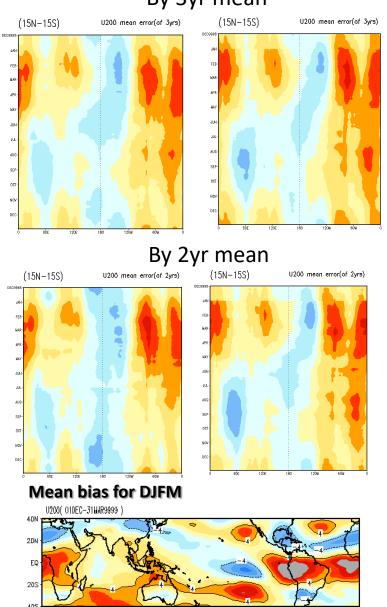


Mean bias for DJFM

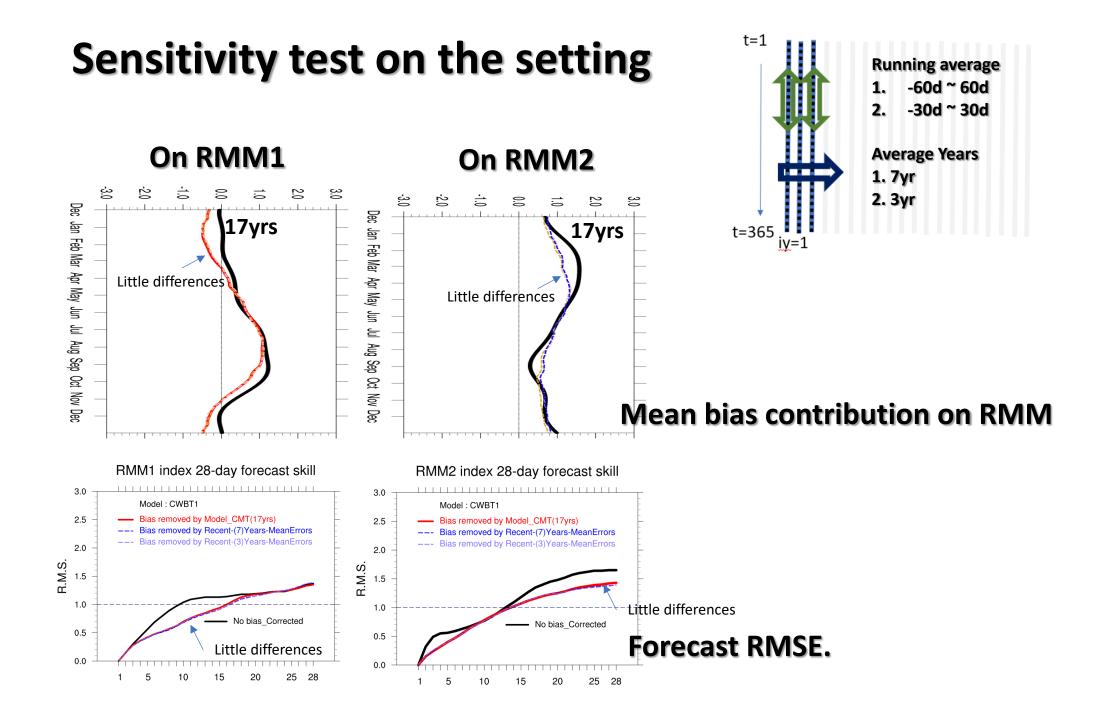


i 2ḋE 4ḋE 6ḋE 8ḃE 1ḋOE 12ḋE 14ḋE 16ḋE 18ḋ 16ḋW 14ḋW 12ḋW 1ḋDW 8ḋW 6ḋW 4ḋW 2ḋW ở

Esimate from mean Errors By 3yr mean



5 2016 4016 6016 8016 10106 12106 14016 16016 180 16011 14011 12011 101011 80111 60111 40111 20111 0





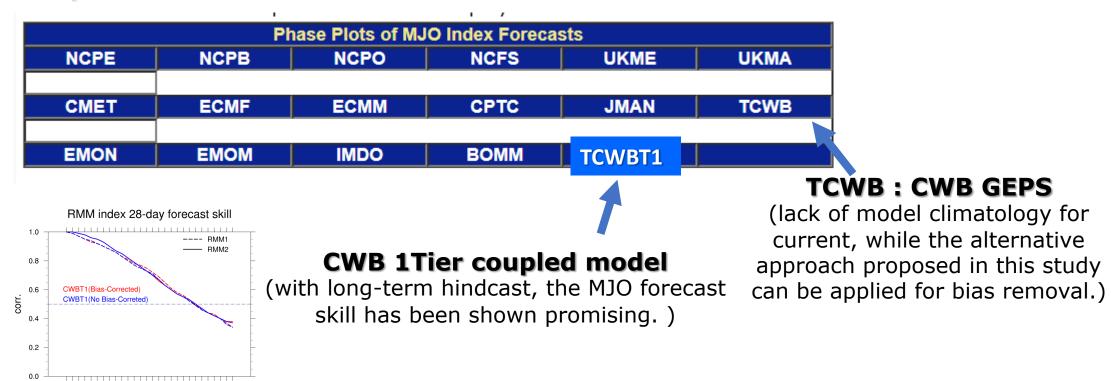
1 5 10 15 20

25 28

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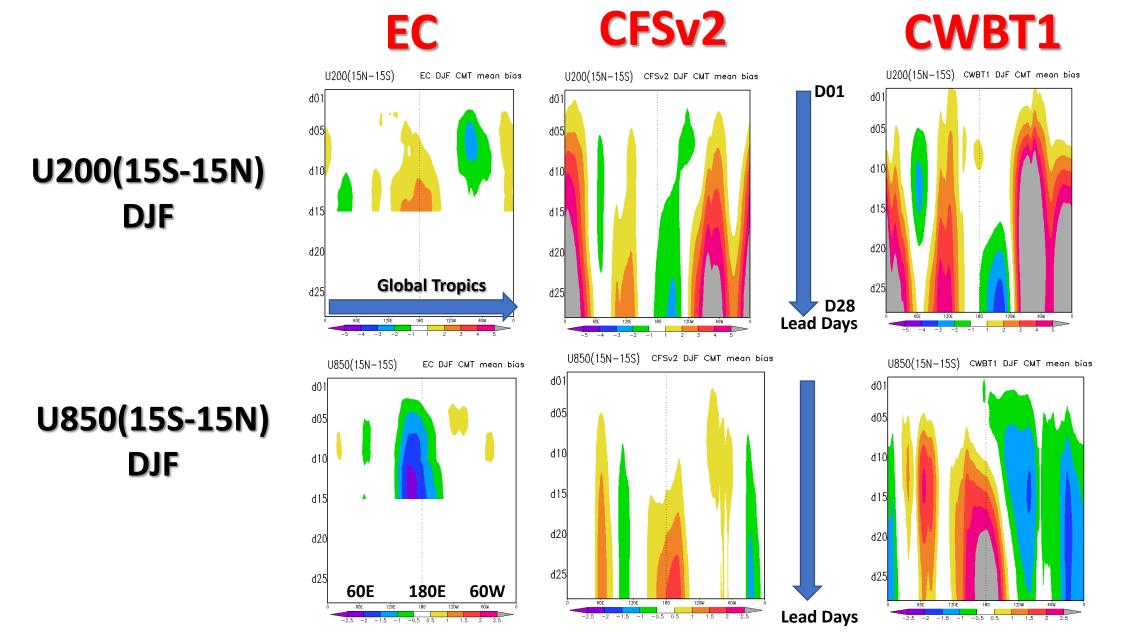
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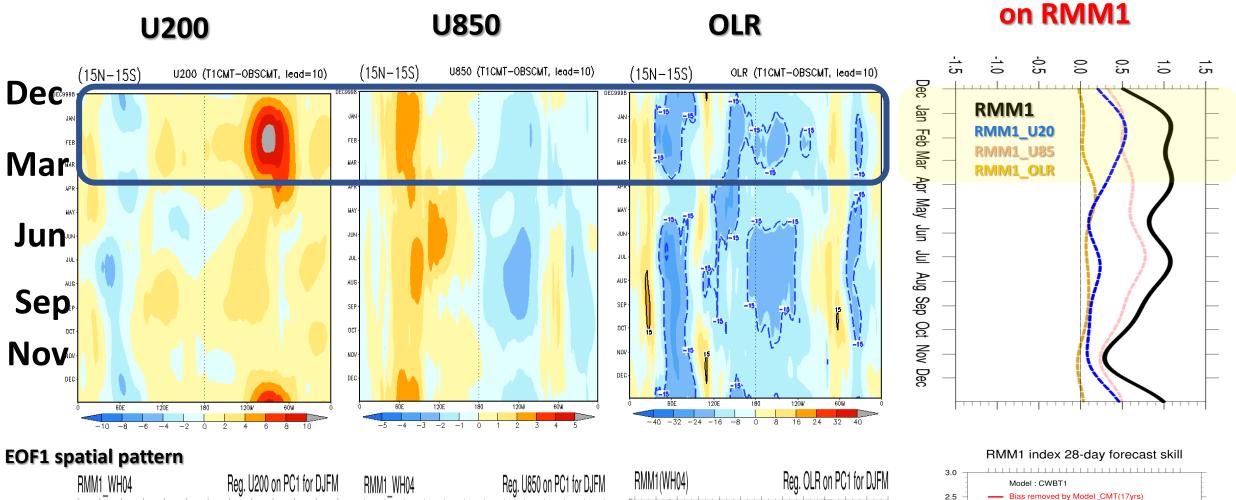


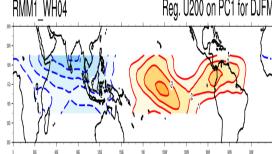
Summary

- This work shows the model bias removal effect on the MJO forecasts and argues a forecast guidance without bias removal would be hardly properly interpreted.
- Model mean bias in the forecast is usually removed based on the hindcast data. For forecast systems with no hindcast data, this study propose an alternative approach to estimate the model mean bias by using recent years forecast data.
- The model mean bias estimated from the proposed alternative approach resembles the results from the standard approach by using the hindcast data. The RMM forecast skill from the proposed alternative approach to removing the model bias is also comparable to the standard approach.

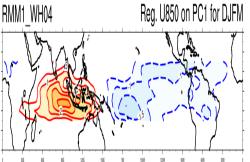
Each model has its own bias structure

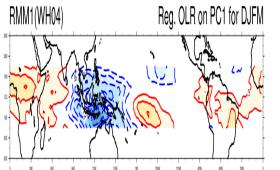


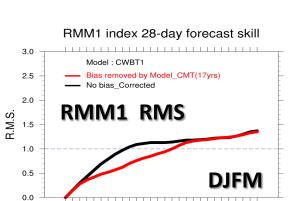




Mean bias for Lead = 10d



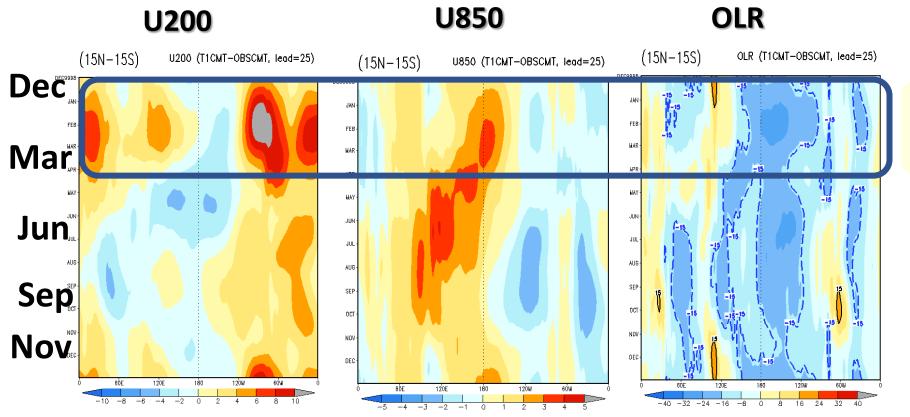




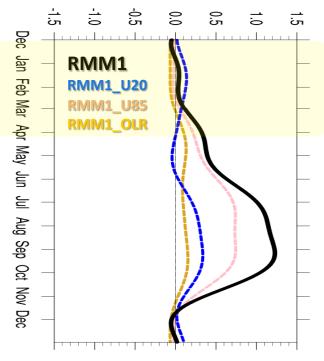
25 28

Mean bias contribution

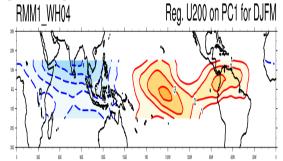
Mean bias for Lead = 25d

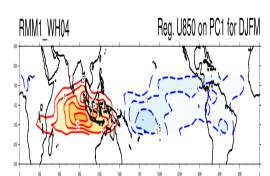


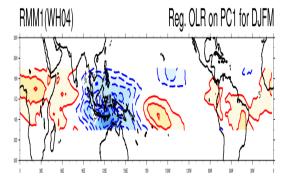
Mean bias contribution on RMM1

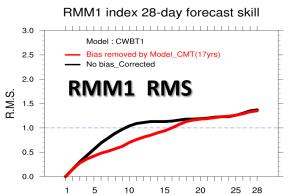


EOF1 spatial pattern

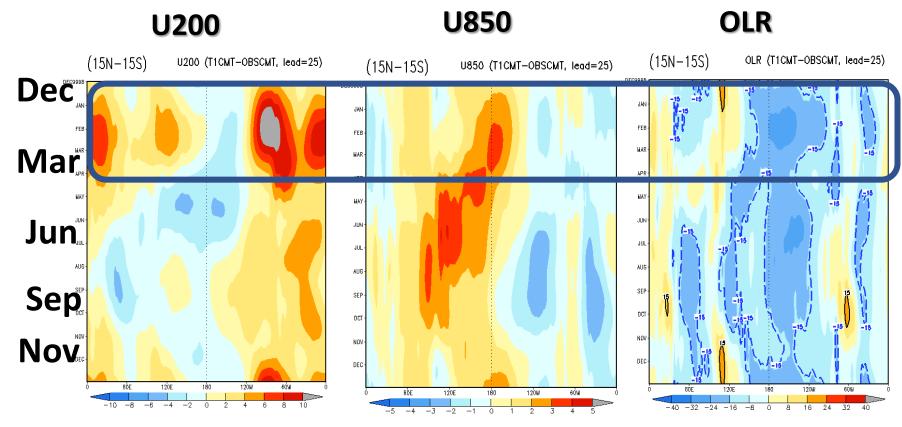




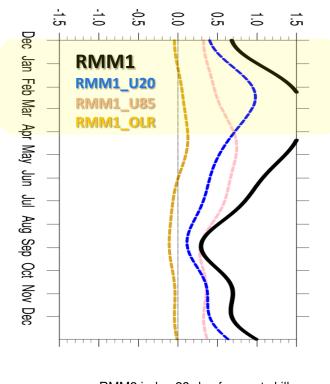




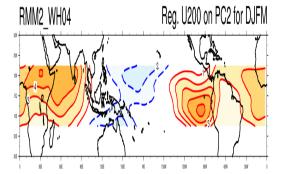
Mean bias contribution



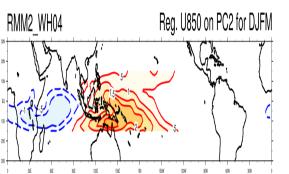
on RMM2

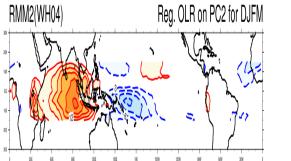


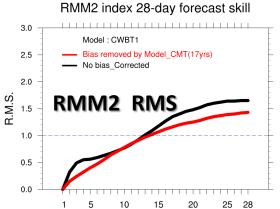
EOF2 spatial pattern



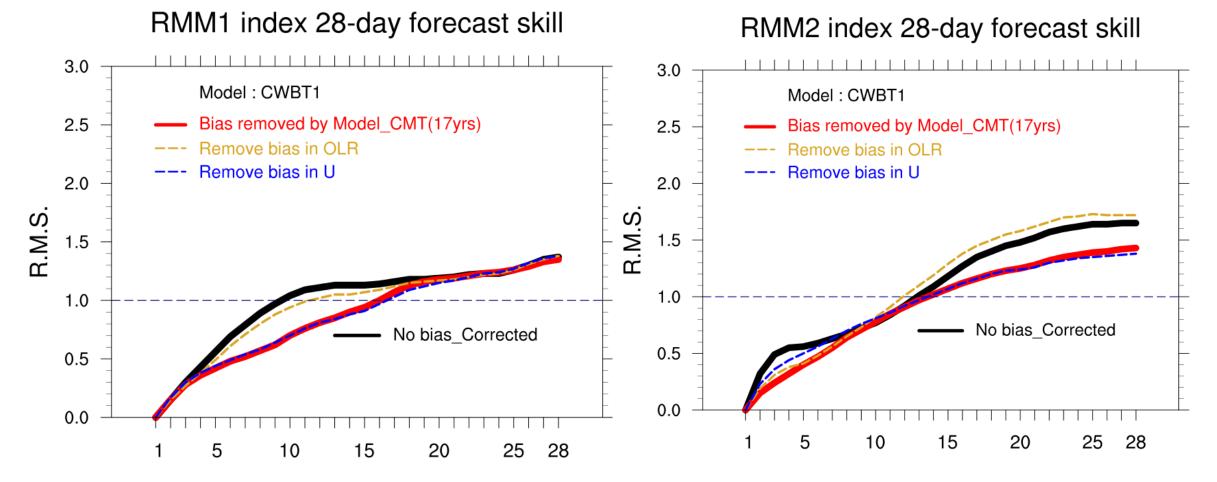
Mean bias for Lead = 25d





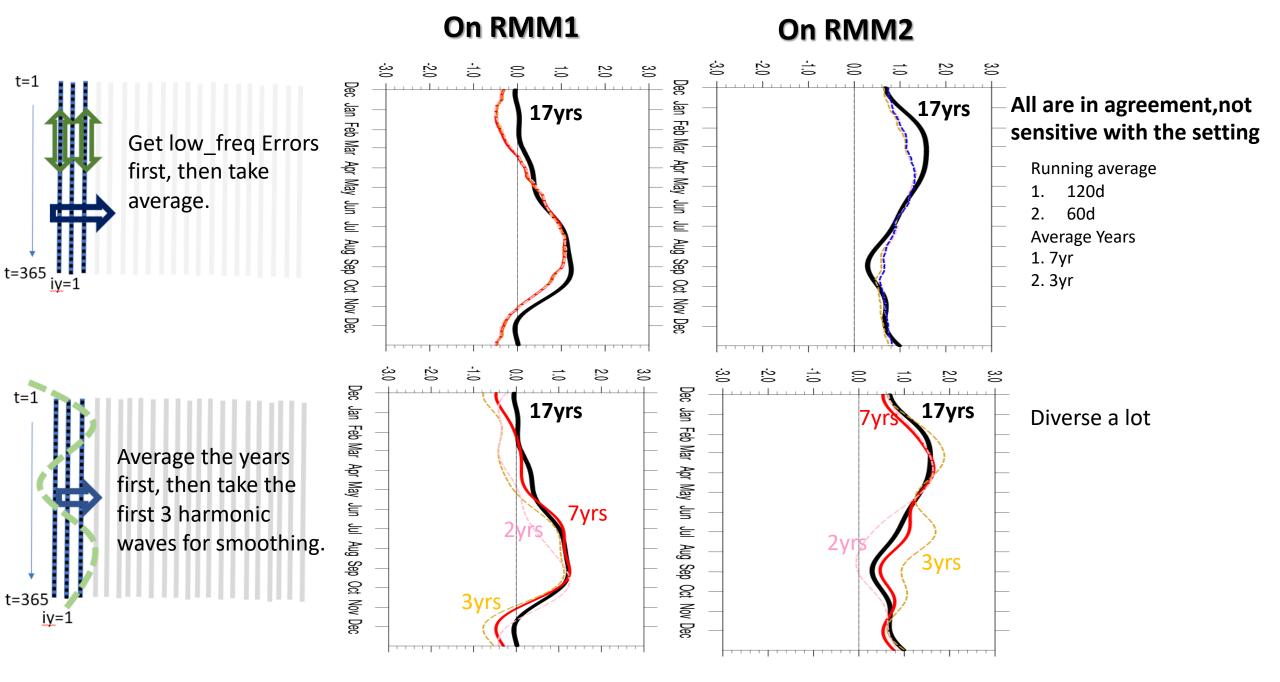


Model : CWB 1-Tier RMM1 RMM2



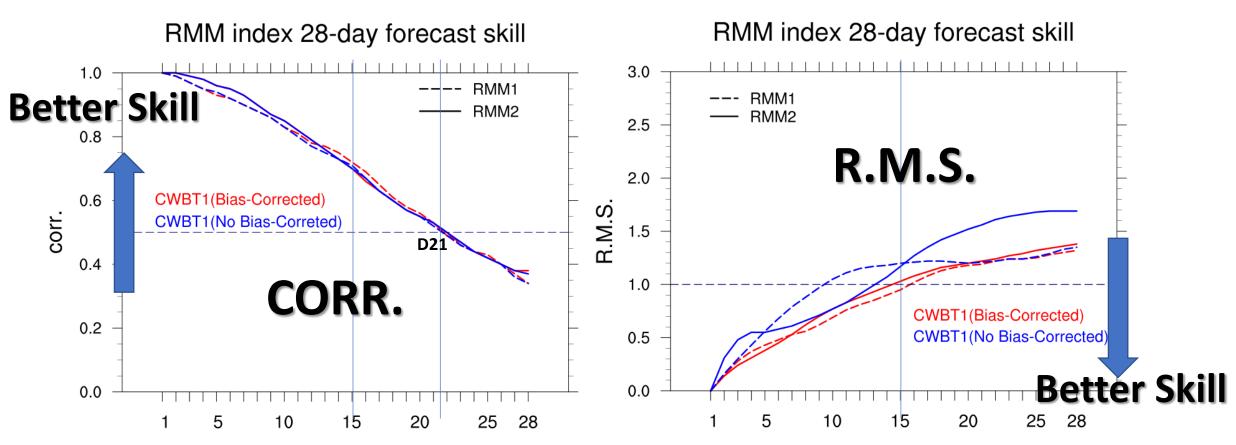
The bias removal effect is mainly from U field, the contribution from OLR is less.

Mean bias contribution



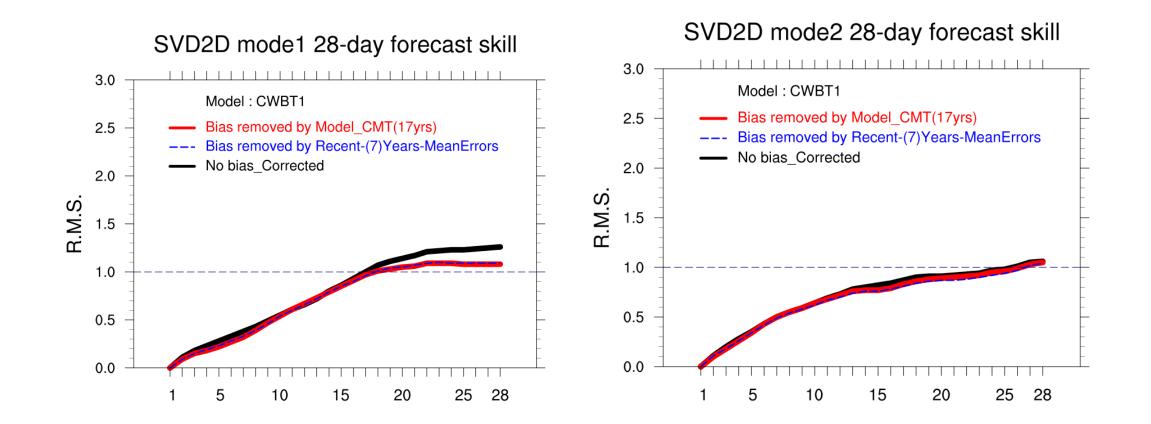
Study 1.1 : Effect of bias removal on RMM forecast skill score Model : CWB 1-Tier seasonal forecast model (CWBT1)

Data : 2012-2018 DJFM



For RMS, bias corrected forecast(Red) have better skill, while comparable skill was seen from correlation measurement. This might imply no bias corrected forecasts also can catch the MJO variation tendency.

SVD



Simmons et al. (1983), Ferranti et al. (1990)

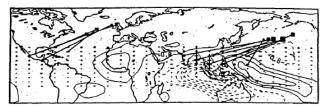
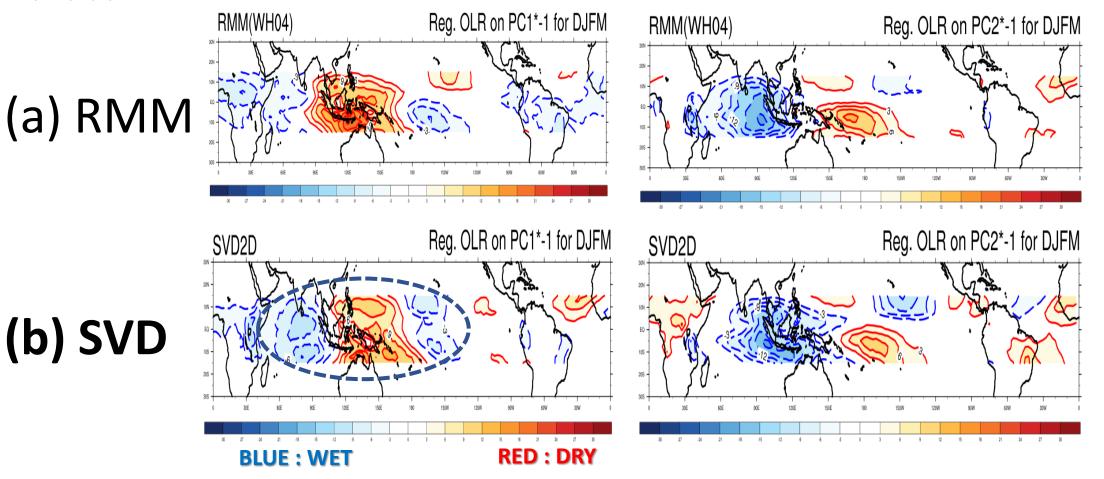


FIG. 8. Schematic illustration of the relationship between the location of the forcing region and the maximum midlatitude response excited from it. (From Simmons et al. 1983). Contours of the first EOF of OLR from the present study are superimposed.

A dipole heating structure over IO-WPC was seen from the SVD index, which is an important key feature in the tropical-extratropical interaction.

Reg (OLR , PC1)

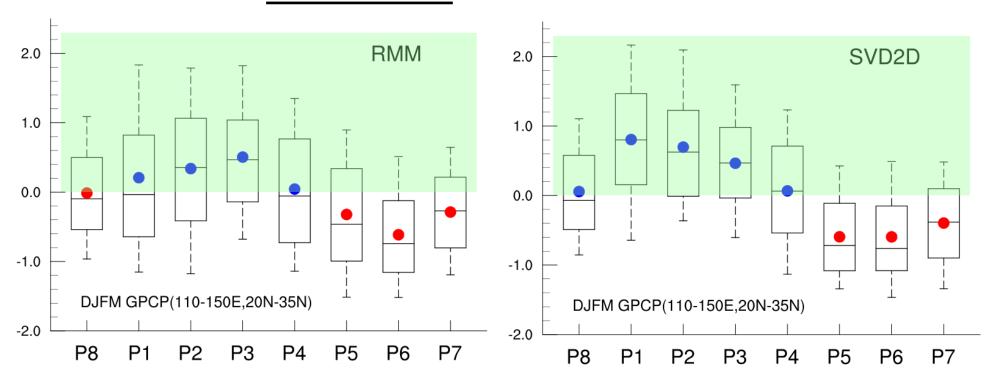
Reg (OLR , PC2)



The spatial pattern of the two leading modes of OLR for (a) RMM(WH04), (b) SVD.

The rainfall anomalies distribution in terms of the 8 MJO phases : RMM & SVD

SVD composite for the complete cycle of evolution of the EA rainfall anomaly through the 8 MJO phases is more conspicuous, while the RMM composite is more ambiguous.



East Asia DJFM rainfall GPCP(110-150E,20N-35N)

PDF of rainfall over EA jet entrance region(110-150E,20N-35N) in 8 MJO phases defined by: (a) RMM(WH04), (b) SVD. Solid dots denote the mean values.

Effect of bias removal on RMM forecast skill score Model : CWB 1-Tier seasonal forecast model

Data : 2012-2018 DJFM

RMM1

RMM2

