

# An evaluation of CFSv2 tropical convection

GEWEX Convection-Permitting Climate Modeling Workshop



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*September 8, 2016*

# Question: Is poor simulation of convection in the tropics undermining NWP on all timescales?

Convection in the tropics significantly impacts global weather on numerous timescales through atmospheric teleconnections.

Examples include ENSO and the MJO.

Thus, poorly simulated tropical convection could reduce NWP skill on many temporal/spatial scales.



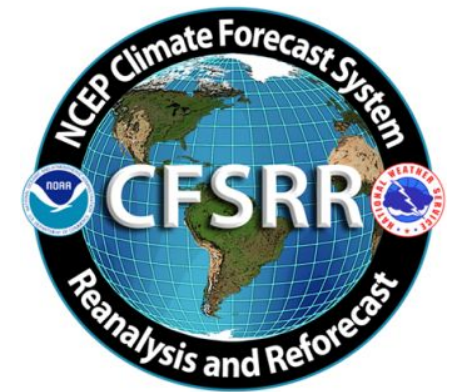
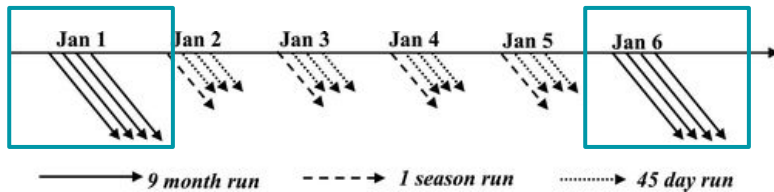
# Motivating questions

- How does global model skill degrade with lead time?
- How does the representation of tropical convection change with lead time?
- Can some of the skill decline be explained by poor simulation of tropical convection?

# Verification methodology

## Forecasts used:

- 4-member CFSv2 reforecast ensemble mean
- 9-month forecasts
- Initialized every 5 days during 1982-2008



## Verification dataset:

- GDAS analyses

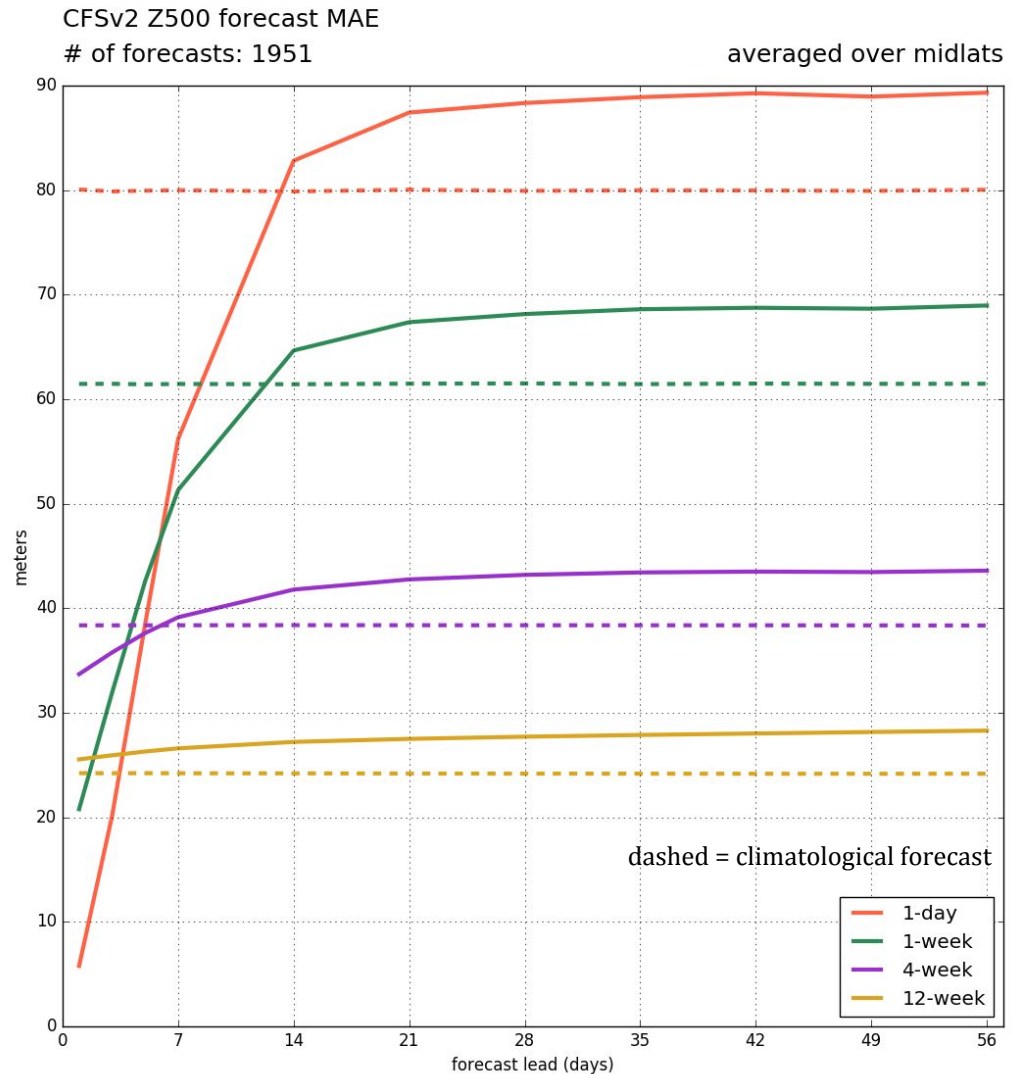
## Temporal timescales:

- Ranging from daily to seasonal
- Weekly-average forecasts primarily used here

# **CFSv2 skill vs lead time**

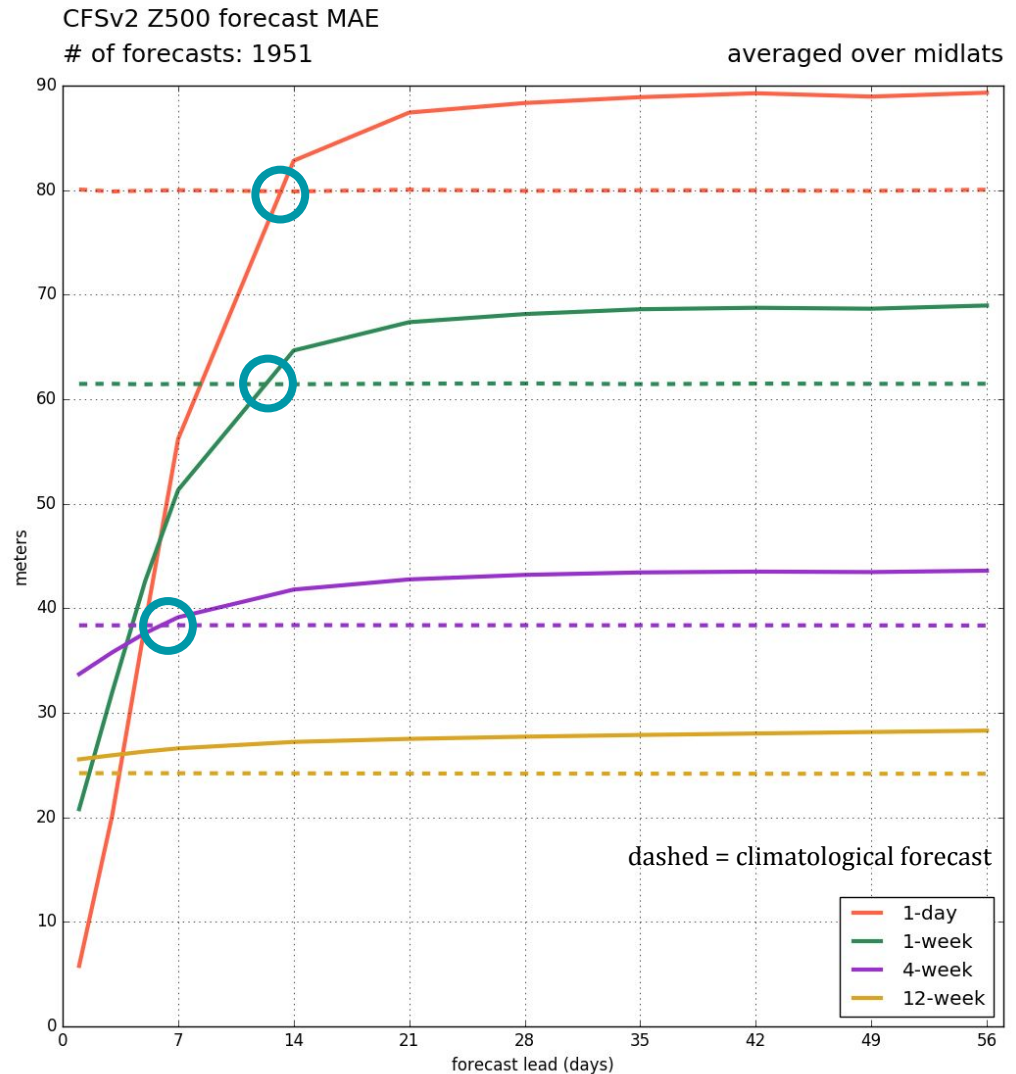
# Z500 MAE vs lead time

- MAE averaged from 30N to 70N
- Errors saturate by week 3 for all averaging periods



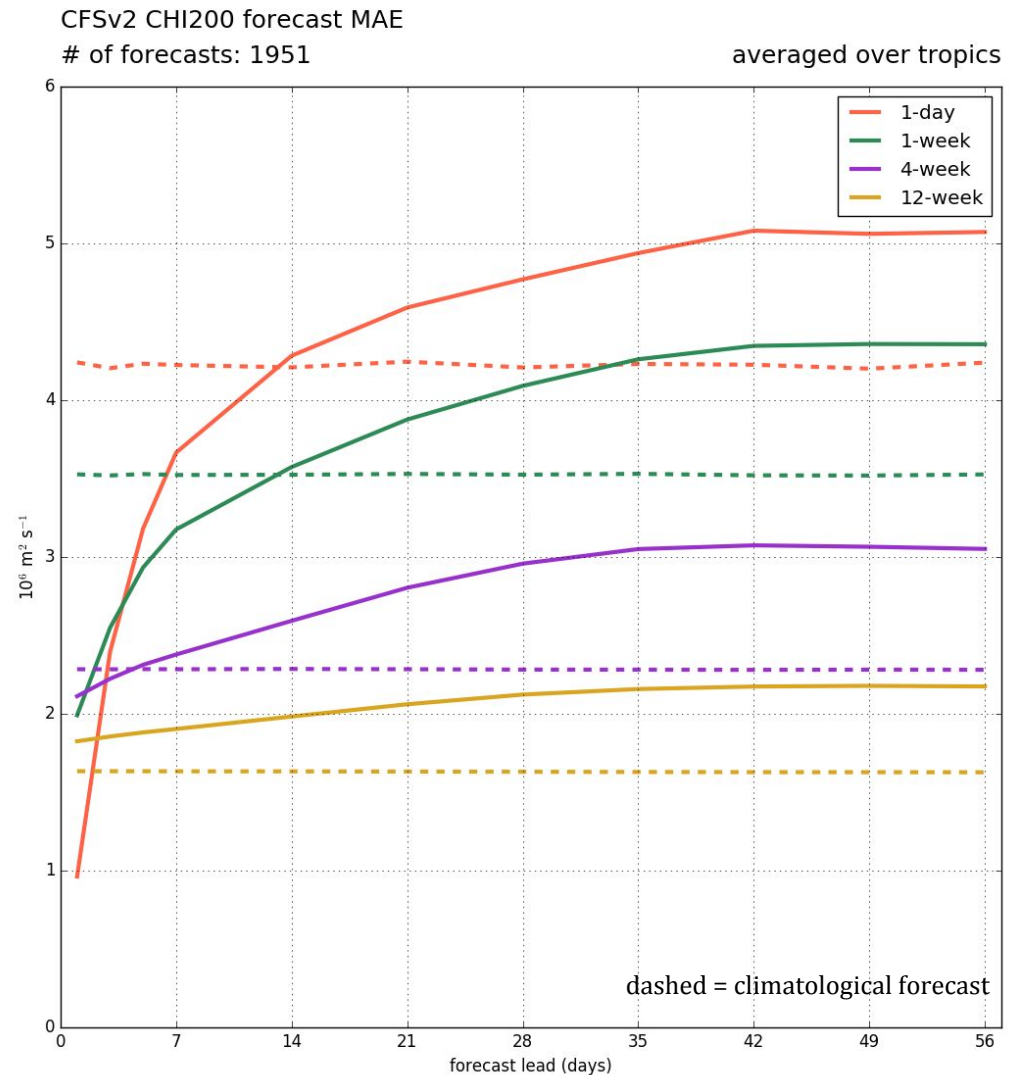
# Z500 MAE vs lead time

- MAE averaged from 30N to 70N
- Errors saturate by week 3 for all averaging periods
- Skill over climatology fades in the first 2 weeks of the forecast



# CHI200 MAE vs lead time

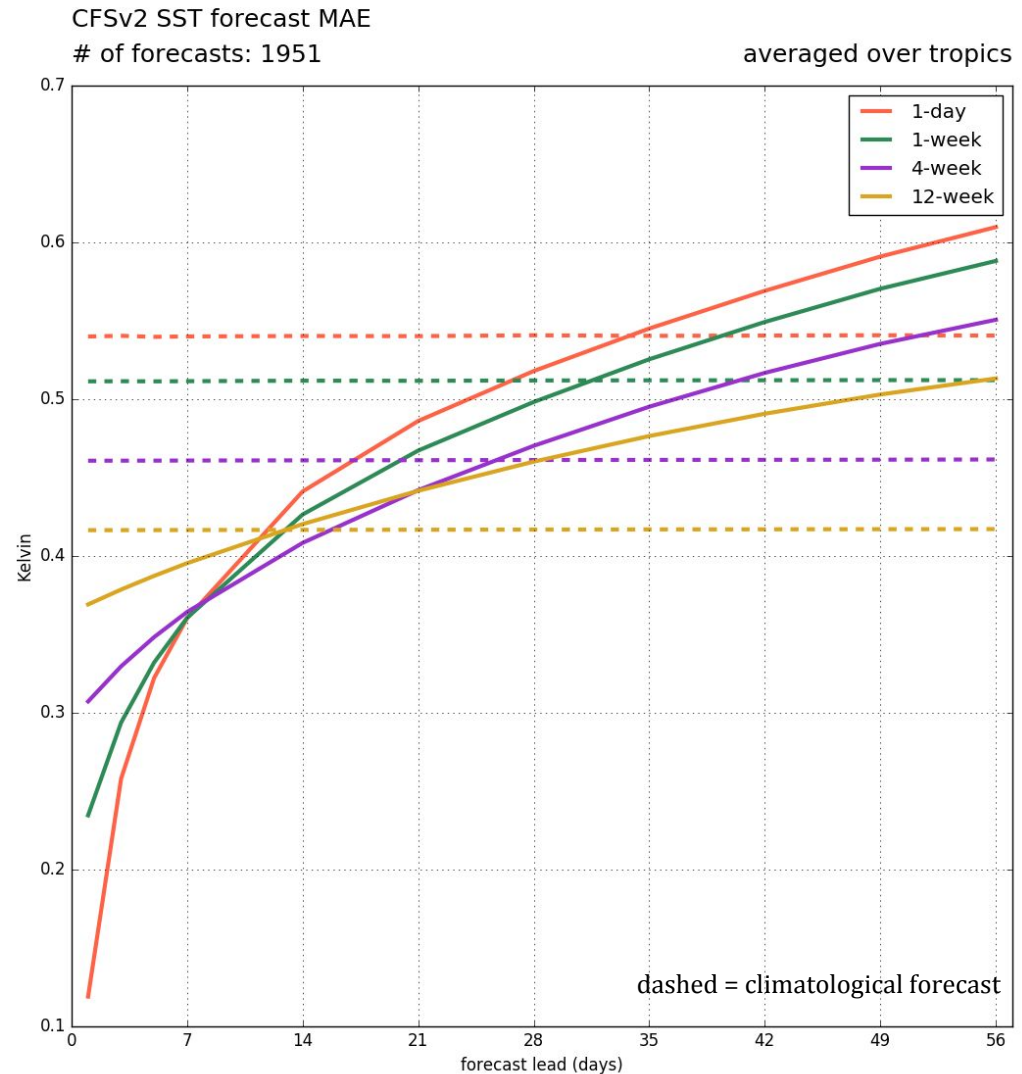
- CHI200 - analog for tropical convection
- MAE averaged from 10S to 10N
- Skill over climatology is similar to Z500, though errors saturate later in the forecast





# SST MAE vs lead time

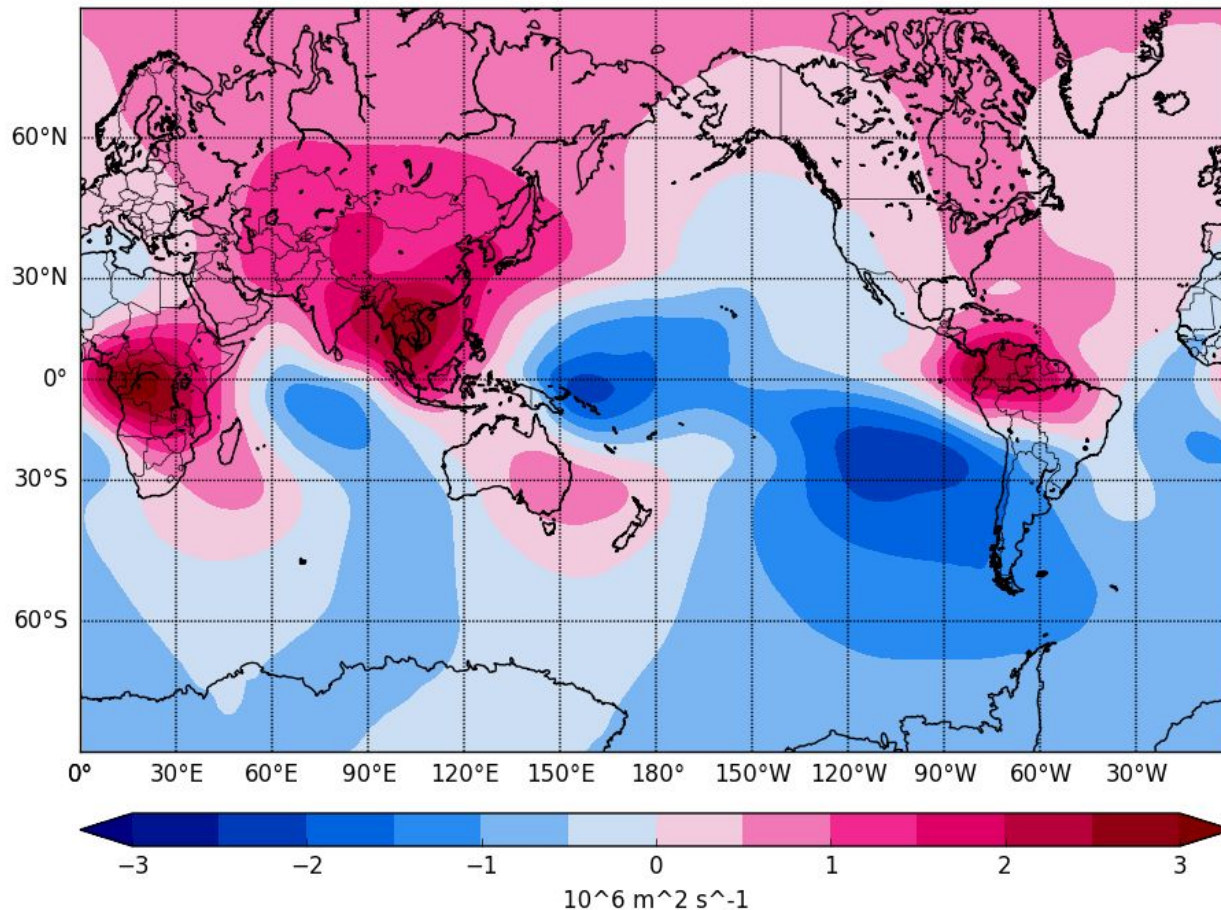
- MAE averaged from 10S to 10N
- Errors do not saturate in the first ~2 months
- Skill over climatology extends much further in time than with Z500 and CHI200



# CHI200 annual bias

CFSv2 1-week-ave CHI200 forecast Bias  
# of forecasts: 1951

Week 4

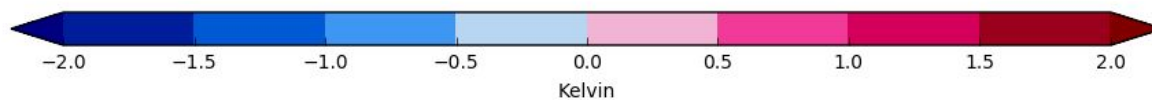
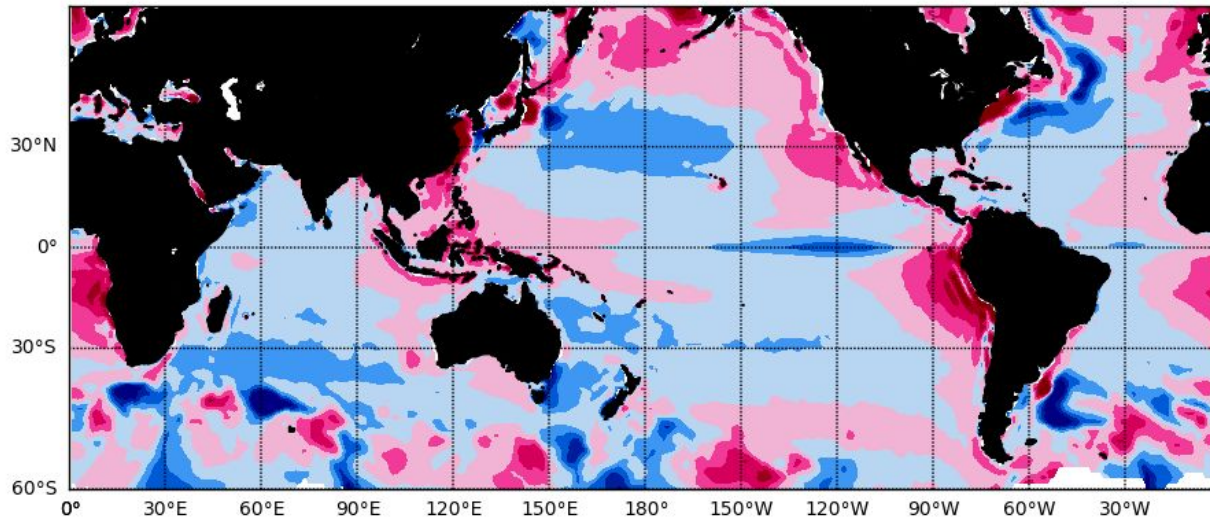


- Too little convection over tropical land
- Too much convection over Indian Ocean, equatorial West Pacific, and Southeast Pacific

# SST annual bias

CFSv2 1-week-ave SST forecast Bias  
# of forecasts: 1951

Week 8



- Warm bias in subtropical stratocumulus environments
- Misplaced boundary currents
- Too cold in equatorial East Pacific (El Niño region)

# Initial questions/observations

- Why does extended SST skill not project onto the atmosphere, e.g. CHI200 and Z500?
- How do the mean state biases impact simulated tropical convection?
- The mean state biases in SST and CHI200 do not seem to correlate with one another

**How tropical convection  
forecasts change with  
lead time**

**A) Hovmöller comparison:  
analyses vs a single CFSv2  
forecast**



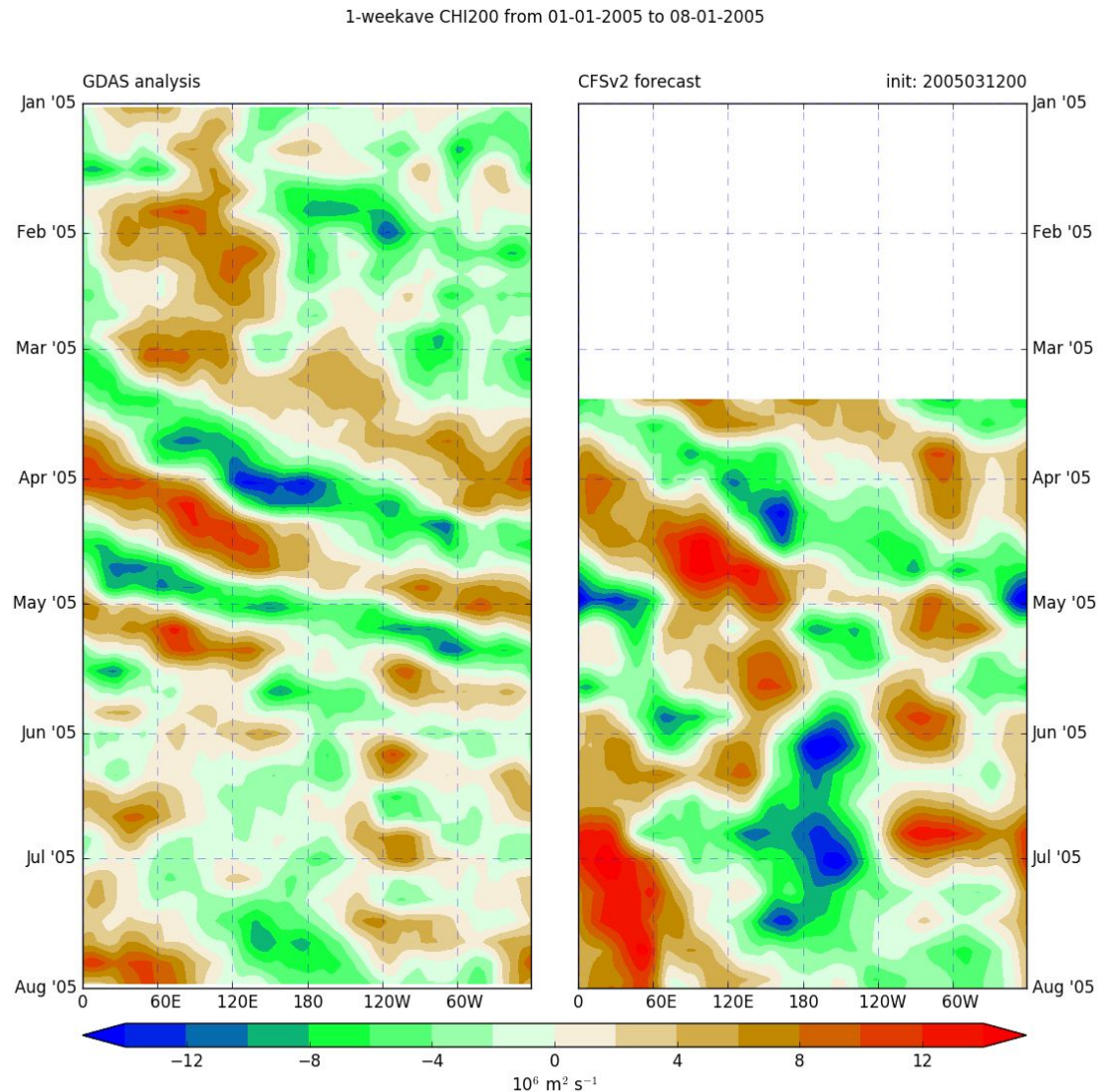
# CHI200 Hovmöller: analyses vs single forecast

Time period:

Spring/summer 2005

Forecast initialization:

March 12, 2005



# CHI200 Hovmöller: analyses vs single forecast

Time period:

Spring/summer 2005

Forecast initialization:

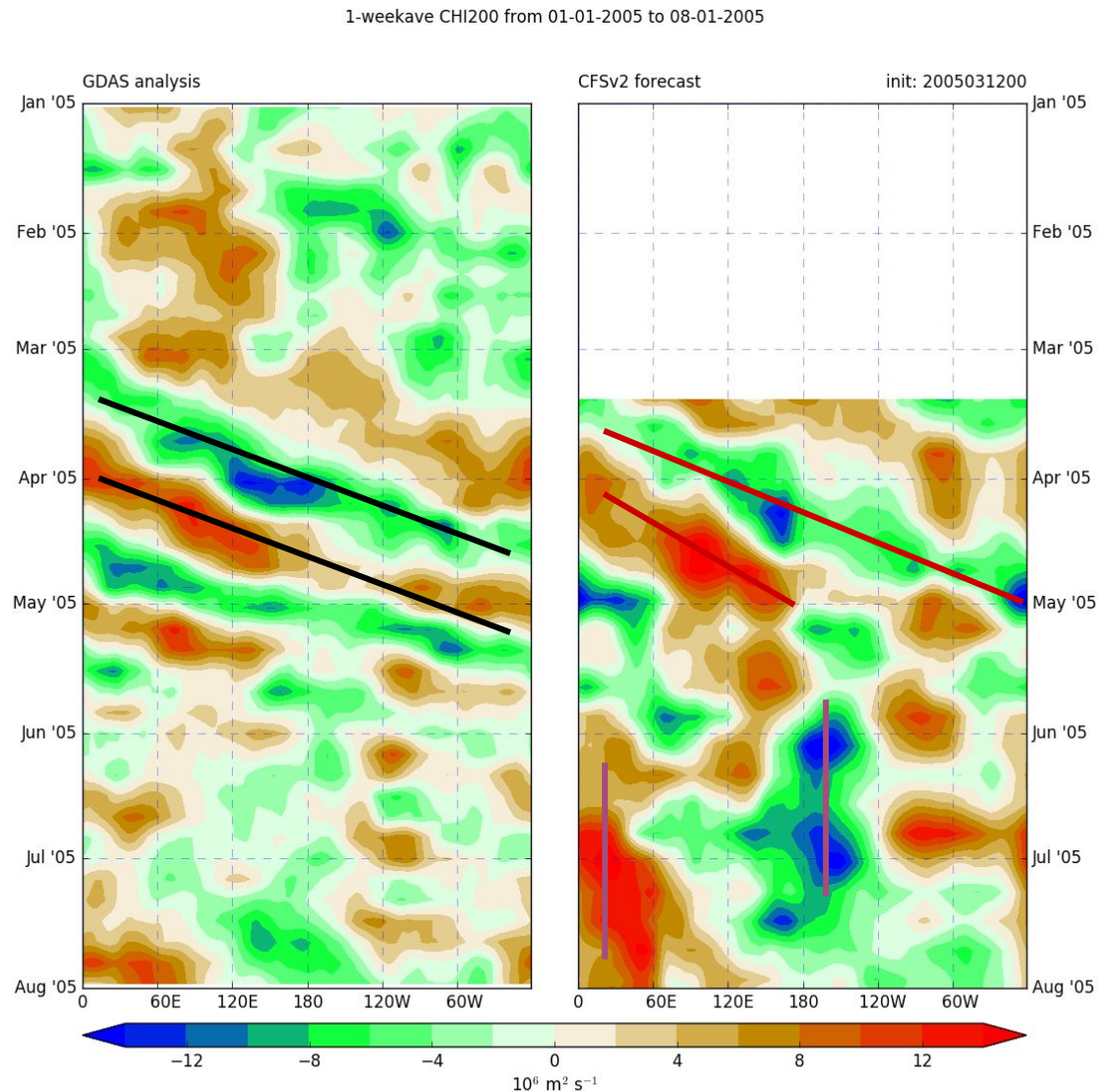
March 12, 2005

Short leads:

Clear propagating features  
(albeit with slower phase  
speed)

Long leads:

Non-propagating anomalies





# CHI200 Hovmöller: analyses vs single forecast

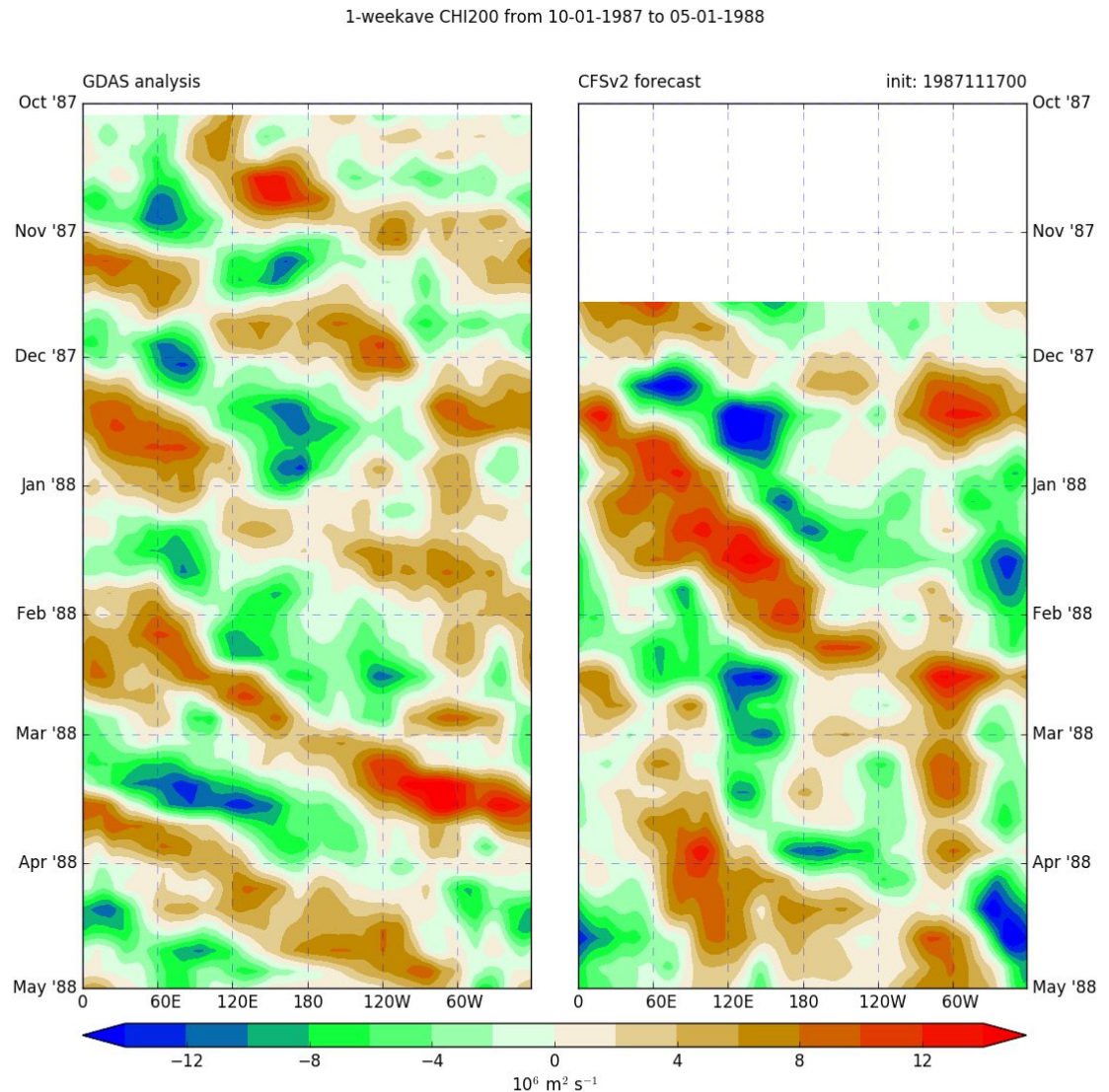
## MORE EXAMPLES

Time period:

Winter/spring 1987-1988

Forecast initialization:

November 17, 1987



# CHI200 Hovmöller: analyses vs single forecast

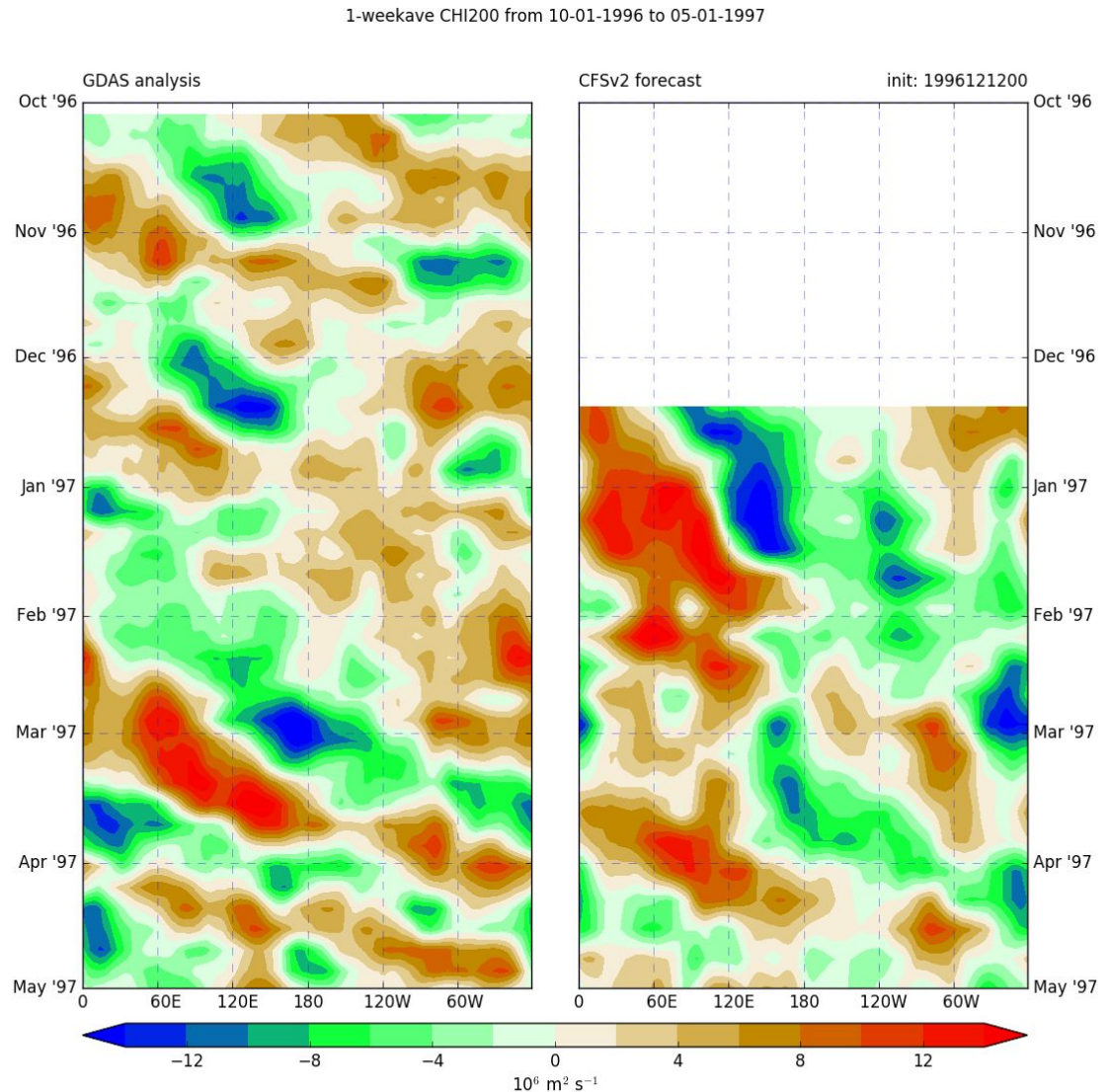
## MORE EXAMPLES

Time period:

Winter/spring 1996-1997

Forecast initialization:

December 12, 1996

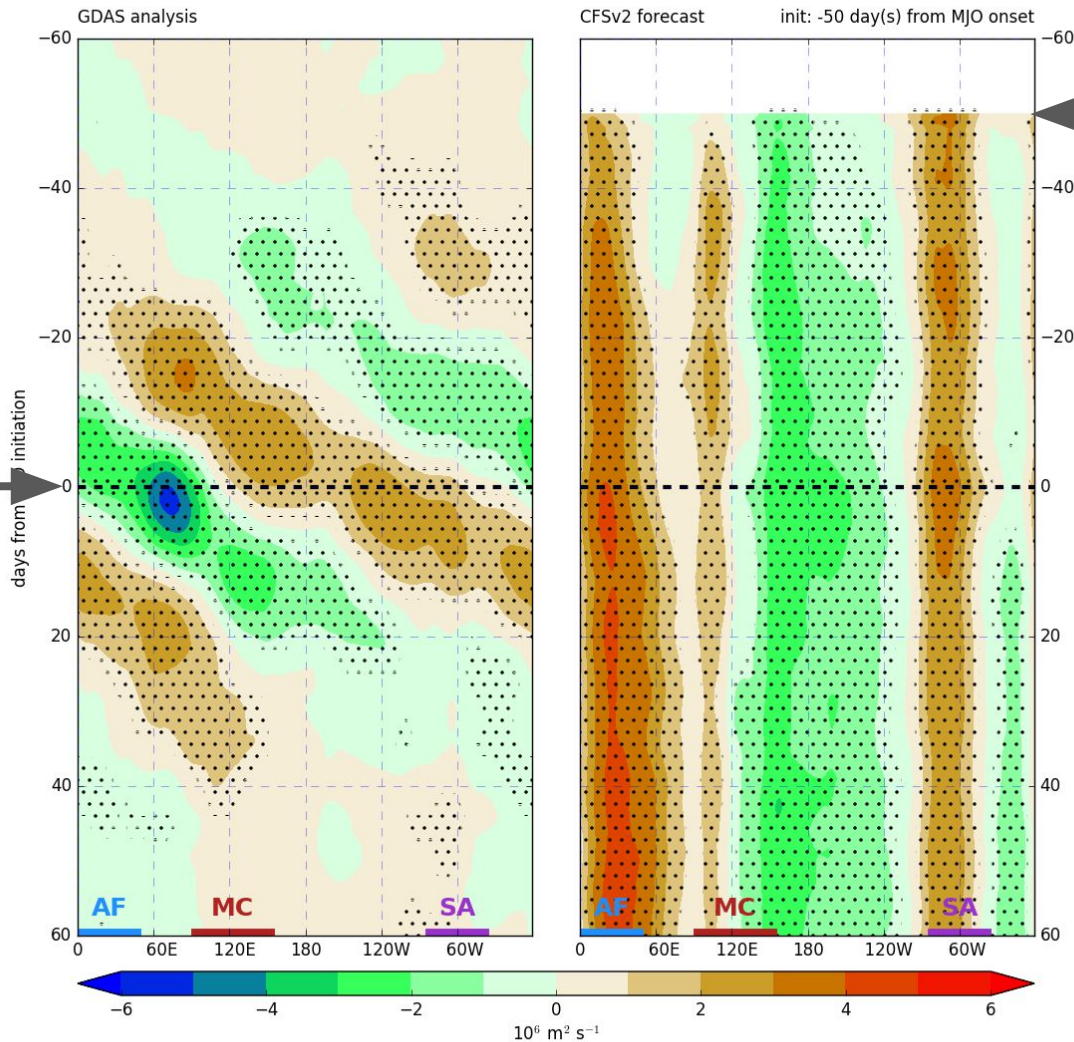


## **B) MJO Hovmöller composites**

# CHI200 Hovmöller composites

1-weekave CHI200 "all" MJO composite: 163 events  
 $r=0.05$

MJO initiation  
(as determined  
by an Indian  
Ocean  
convection  
index)



Forecasts  
initialized

Stippling =  
95% sig.

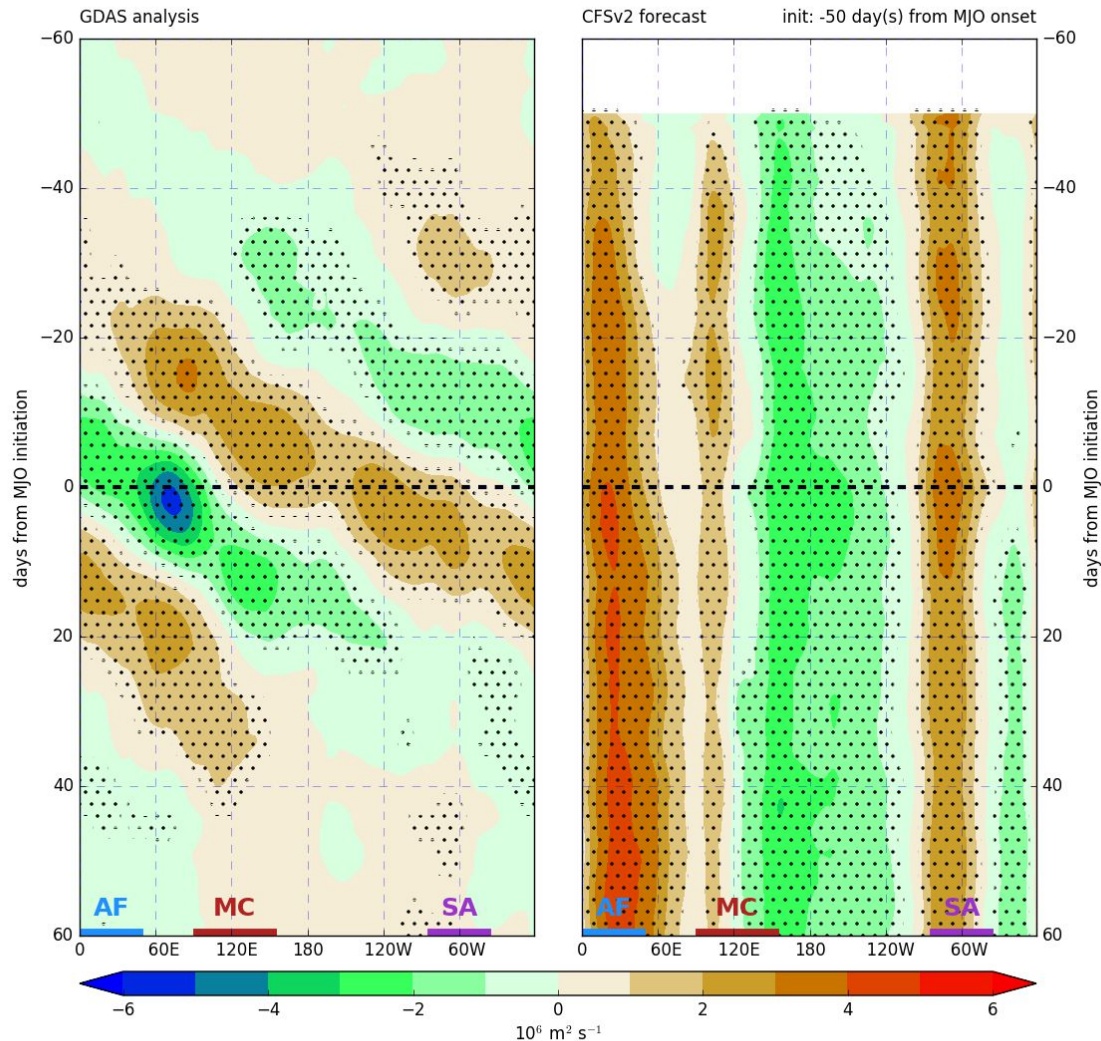


# CHI200 Hovmöller composites

1-weekave CHI200 "all" MJO composite: 163 events  
 $r=0.05$

Long leads:  
No MJO signal.

Model biases dominate, e.g.  
dry equatorial  
land and wet  
West Pacific

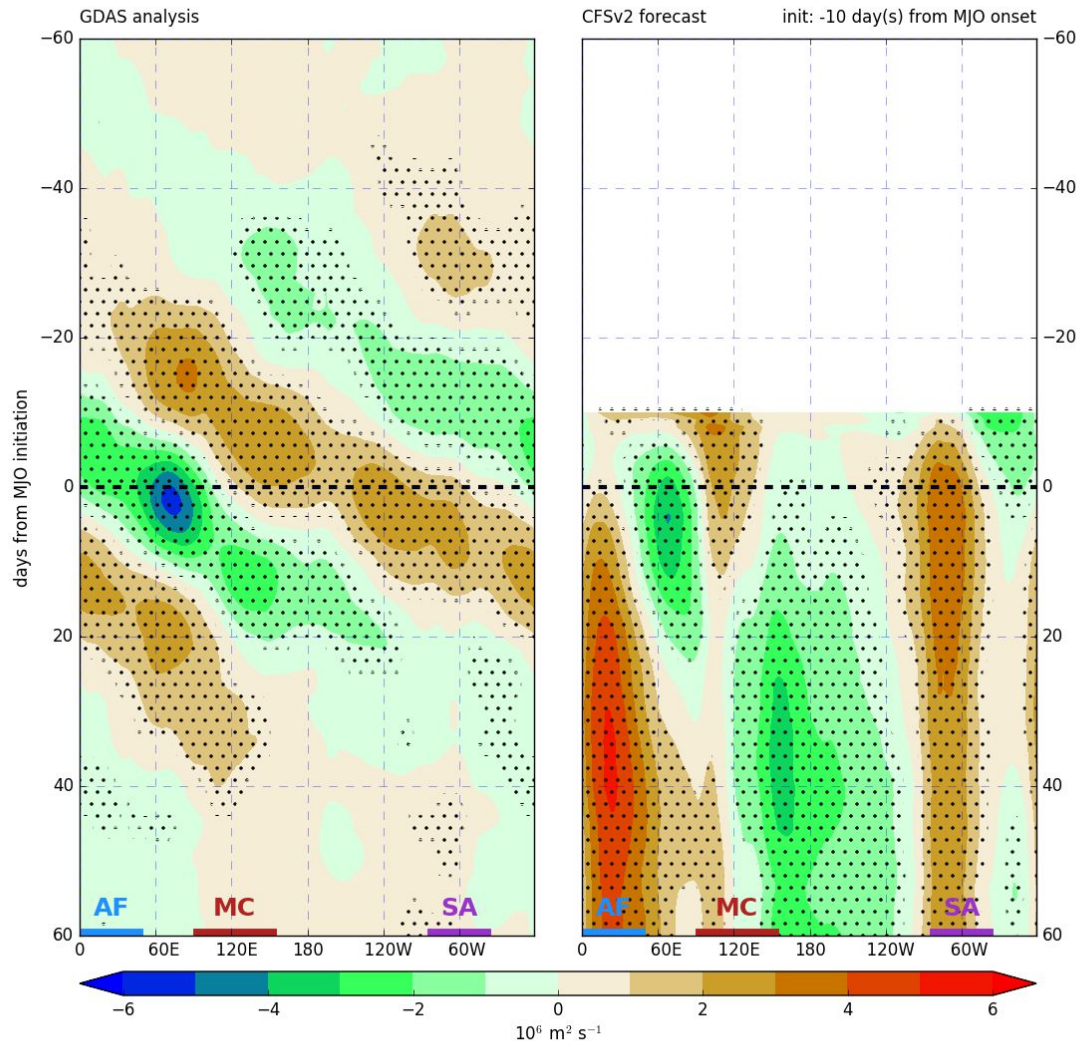


# CHI200 Hovmöller composites

1-weekave CHI200 "all" MJO composite: 163 events  
 $r=0.23$

Long leads:  
No MJO signal.

Model biases dominate, e.g.  
dry equatorial  
land and wet  
West Pacific



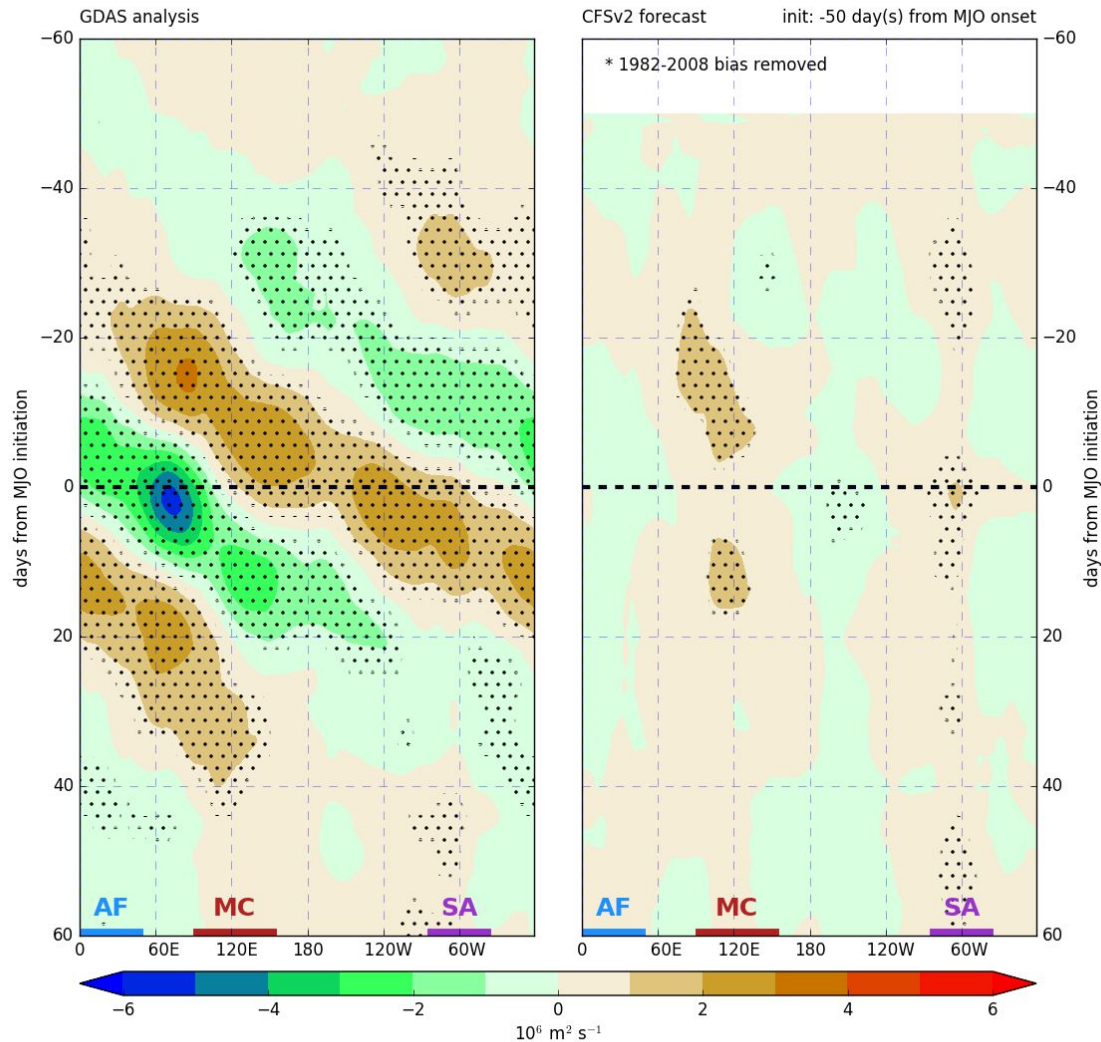
Shorter leads:  
Weak MJO  
signal.

Model biases  
take over ~2  
weeks into the  
forecast

# CHI200 Hovmöller composites

\* Lead-  
dependent  
1982-2008  
bias removed  
from forecasts

1-weekave CHI200 "all" MJO composite: 163 events  
 $r=0.24$

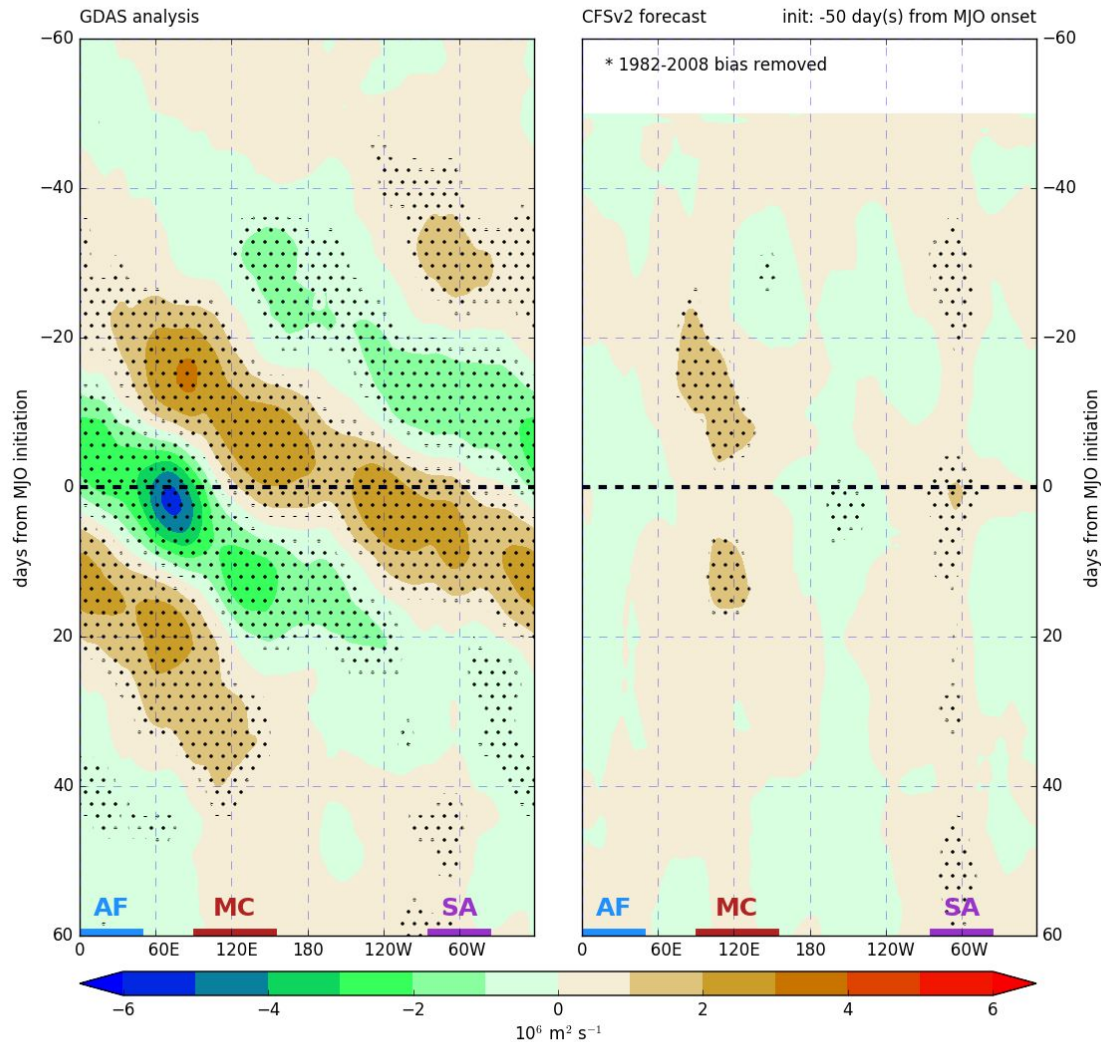




# CHI200 Hovmöller composites

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1-weekave CHI200 "all" MJO composite: 163 events  
 $r=0.24$



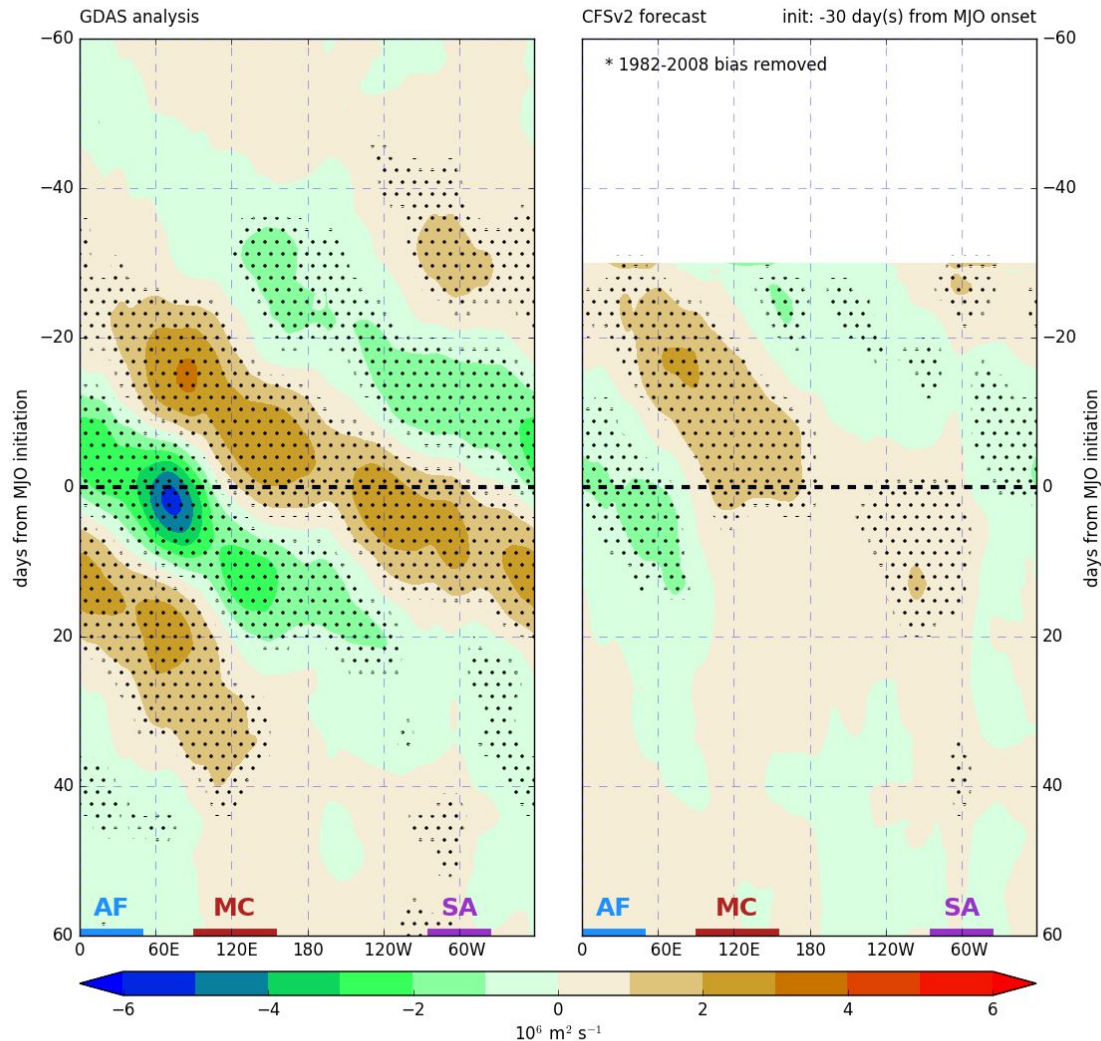
Long leads:  
No MJO signal



# CHI200 Hovmöller composites

\* Lead-dependent  
1982-2008  
bias removed  
from forecasts

1-weekave CHI200 "all" MJO composite: 163 events  
 $r=0.53$



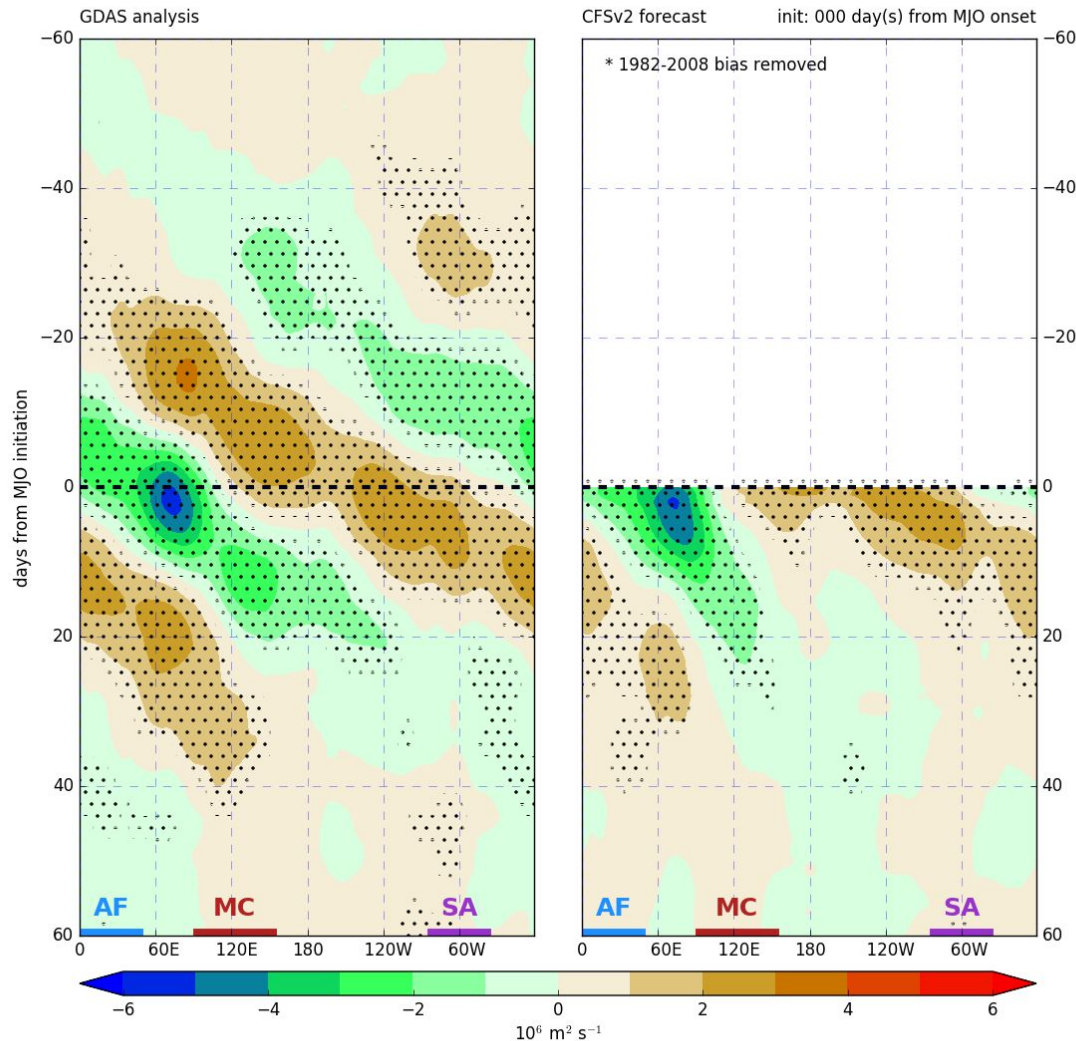
Long leads:  
No MJO signal

~30-day lead:  
Weak/slow MJO

# CHI200 Hovmöller composites

\* Lead-dependent  
1982-2008  
bias removed  
from forecasts

1-weekave CHI200 "all" MJO composite: 163 events  
 $r=0.76$



Long leads:  
No MJO signal

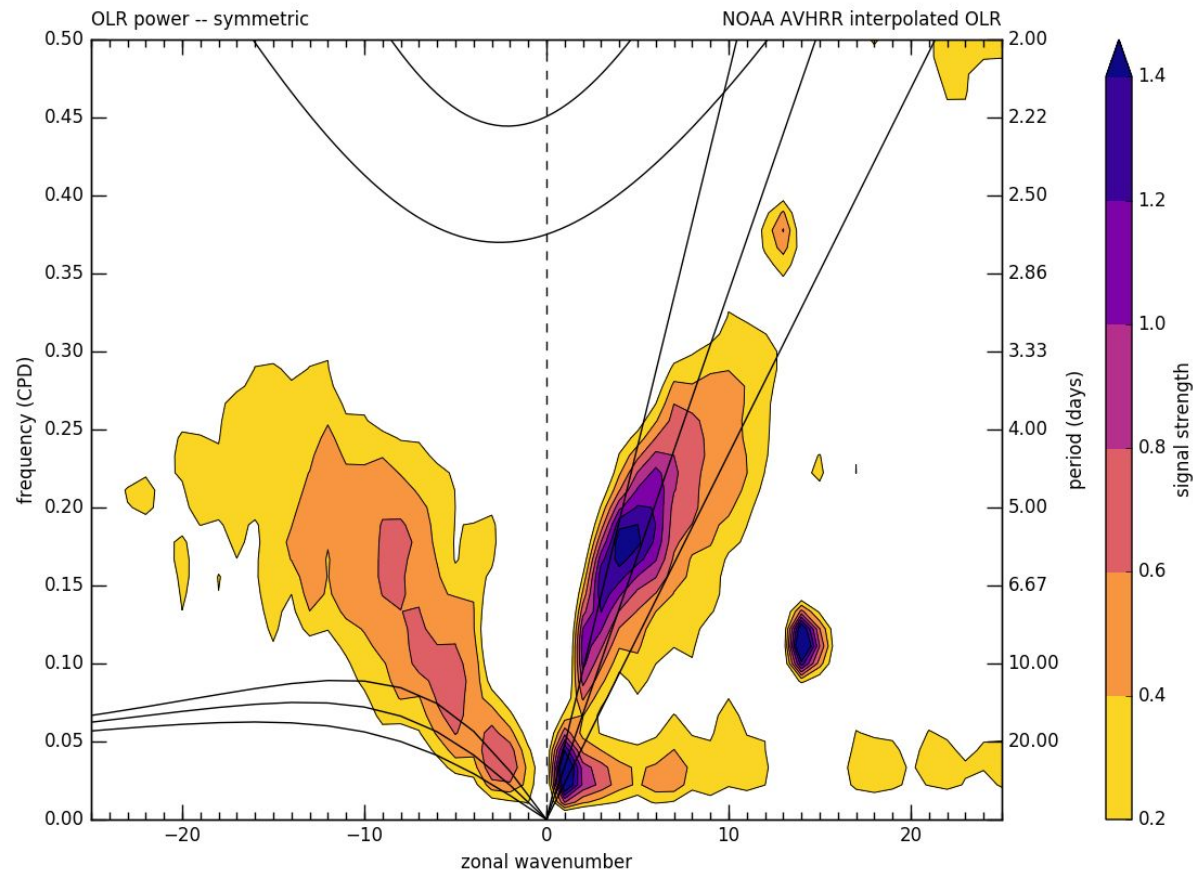
~30-day lead:  
Weak/slow MJO

Shorter leads:  
MJO intensity improves,  
Phase speed still too slow,  
M.C. Barrier

# **C) Wavenumber- frequency power spectra**

# Space-time power spectrum - satellite obs

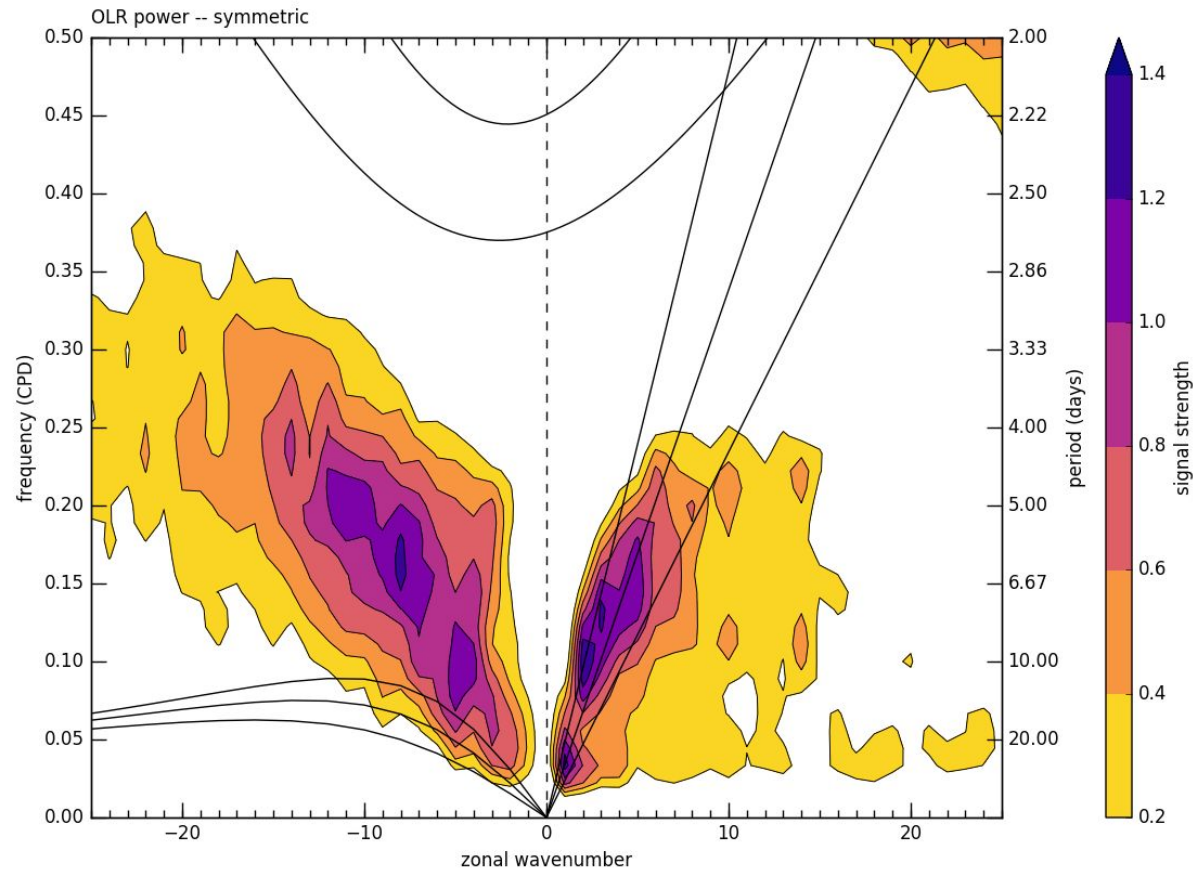
- Computed with daily, latitude-averaged (15S to 15N) data (here, AVHRR OLR)
- Amount of variance at different  $k$ ,  $\omega$
- Shown as ratio between the power and the background “red” spectrum
- Black lines are the dispersion curves for Kelvin, Rossby, and inertia-gravity waves



See Wheeler and Kiladis (1999) for more details

# Space-time power spectrum - analyses

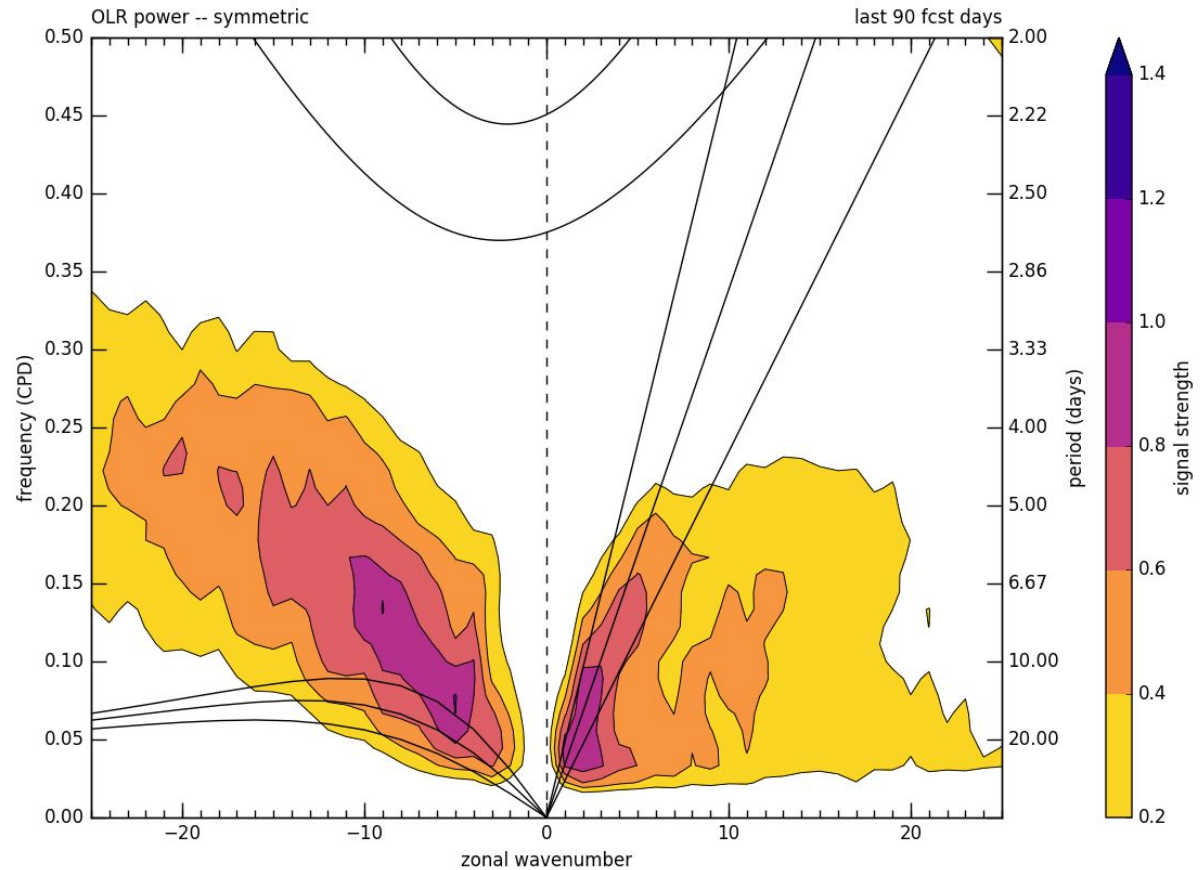
- MJO and Kelvin waves are the prominent eastward-propagating features
- More westward-propagating power than in the observations





# Space-time power spectrum - CFSv2 forecasts

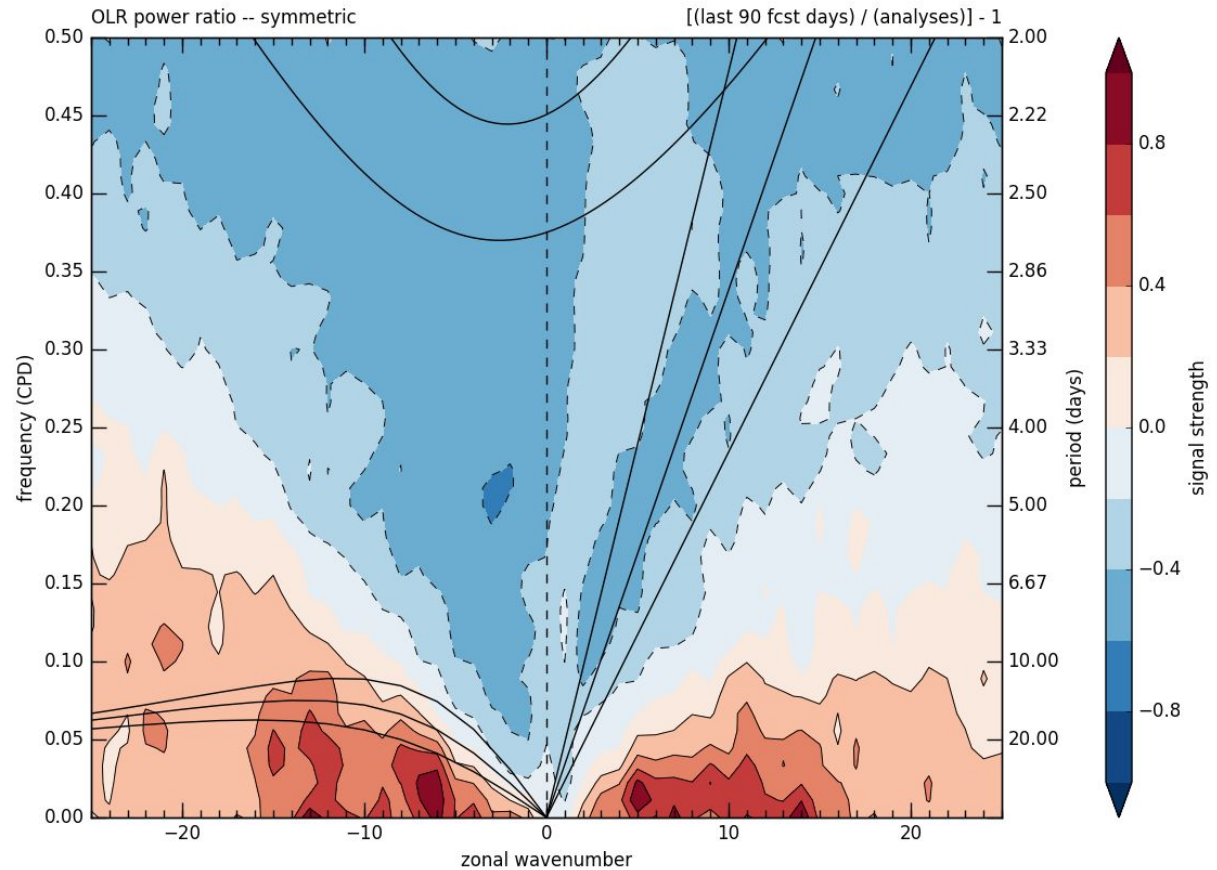
- *Last 90 days* of CFSv2 9-month forecasts
- MJO and Kelvin waves are less distinguished from each other and from the background spectrum
- Power over a wider range of wavenumbers and frequencies



# Raw power ratio: CFSv2 vs analyses

$$\frac{(\text{raw power: forecasts})}{(\text{raw power: analyses})} - 1$$

- Long-lead forecasts tend to have:
  - Higher power at lower frequencies
  - Lower power at high frequencies



# Conclusions so far

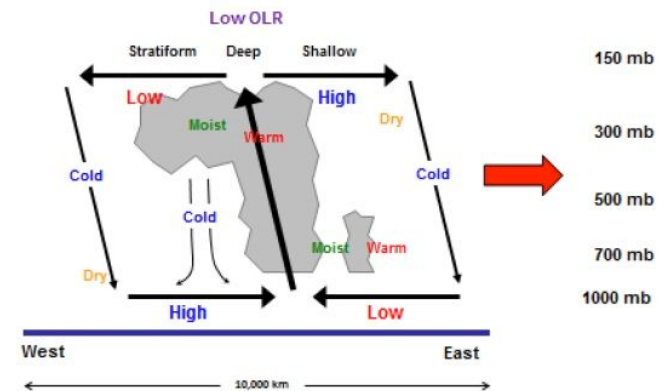
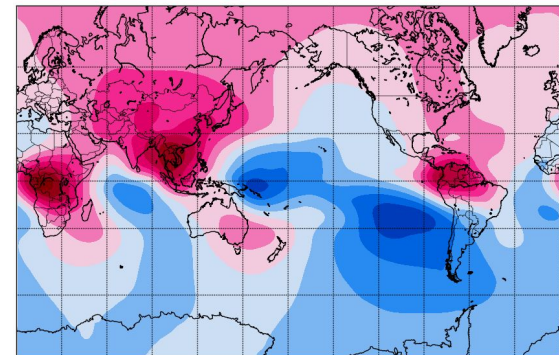
- CFSv2 tropical convective skill does not permeate far into the subseasonal time range (only the first few weeks of a forecast)
- Robust biases in the model's tropical convection and moisture fields may inhibit the formation of realistically propagating convective features like the MJO
  - The wet bias in the west Pacific likely acts to inhibit MJO propagation (Kim et al. 2014)
- The lack of coherence between the SST biases and CHI200 biases, *and* the extended SST forecast skill, suggests that the issue lies in the atmospheric model, rather than the ocean



# Some thoughts on parameterized convection

- Convective parameterizations can affect propagating convection in two ways:
  1. Indirect effect: Changing the mean state in the model, which has a huge impact on features like the MJO
  2. Direct effect: Altering the simulated convectively-coupled waves themselves by poorly representing the interactions between clouds and dynamics
- We should **test** to see how propagating convection might improve with explicit convection in the tropics

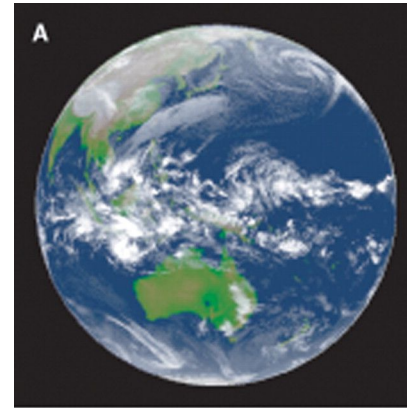
CHI200 bias



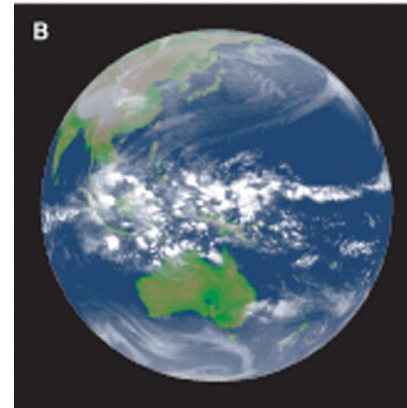
Convection-permitting models are able to...

# Convection-permitting models are able to...

- Realistically simulate MJO events (e.g., Miura et al. 2007)



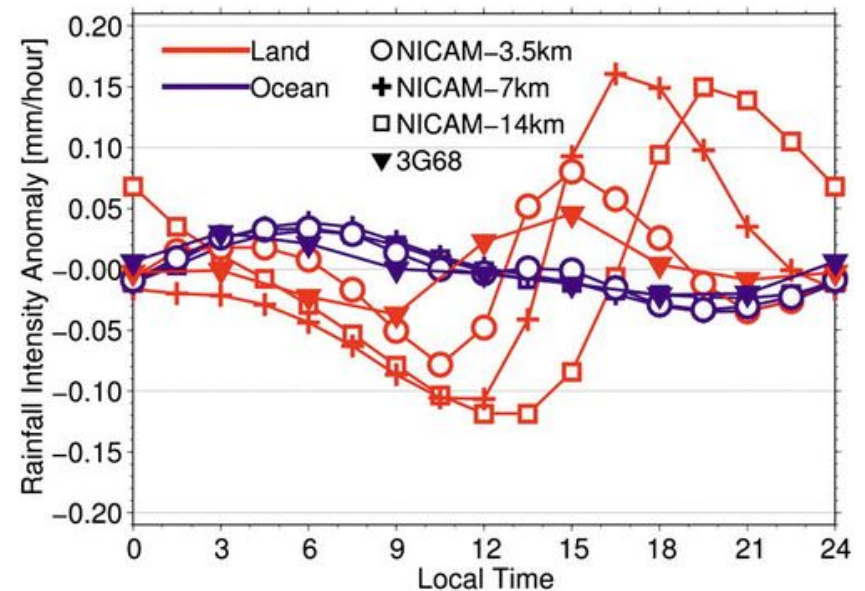
MTSAT-1R



NICAM 3.5km

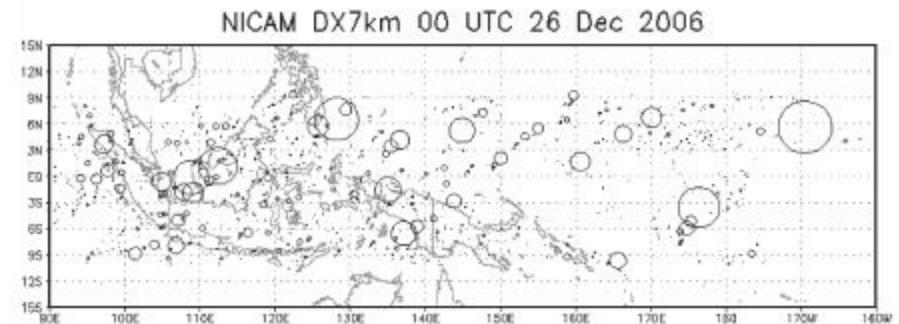
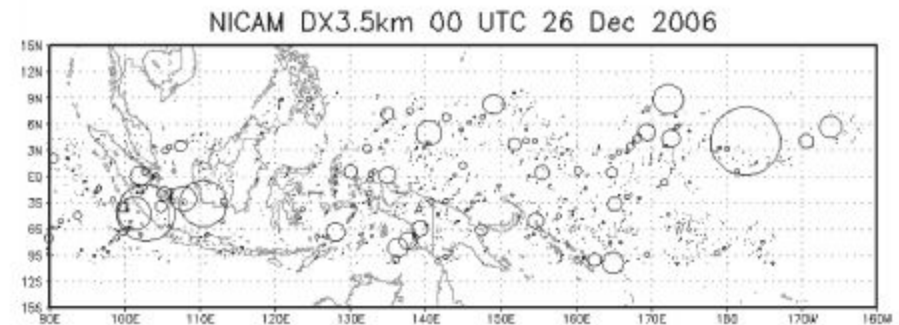
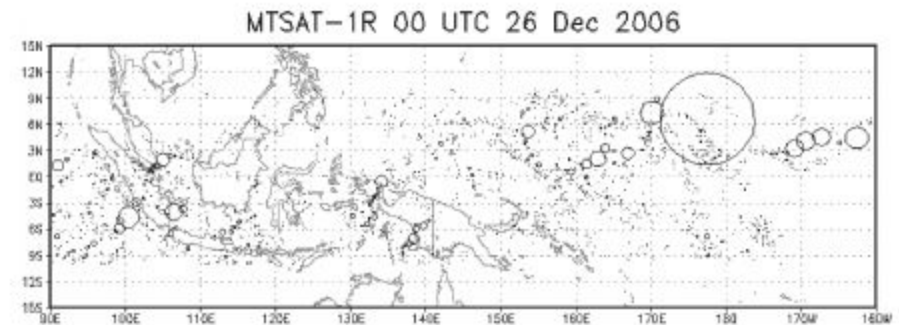
# Convection-permitting models are able to...

- Realistically simulate MJO events (e.g., Miura et al. 2007)
- More accurately reproduce the rainfall diurnal cycle (Sato et al. 2009)



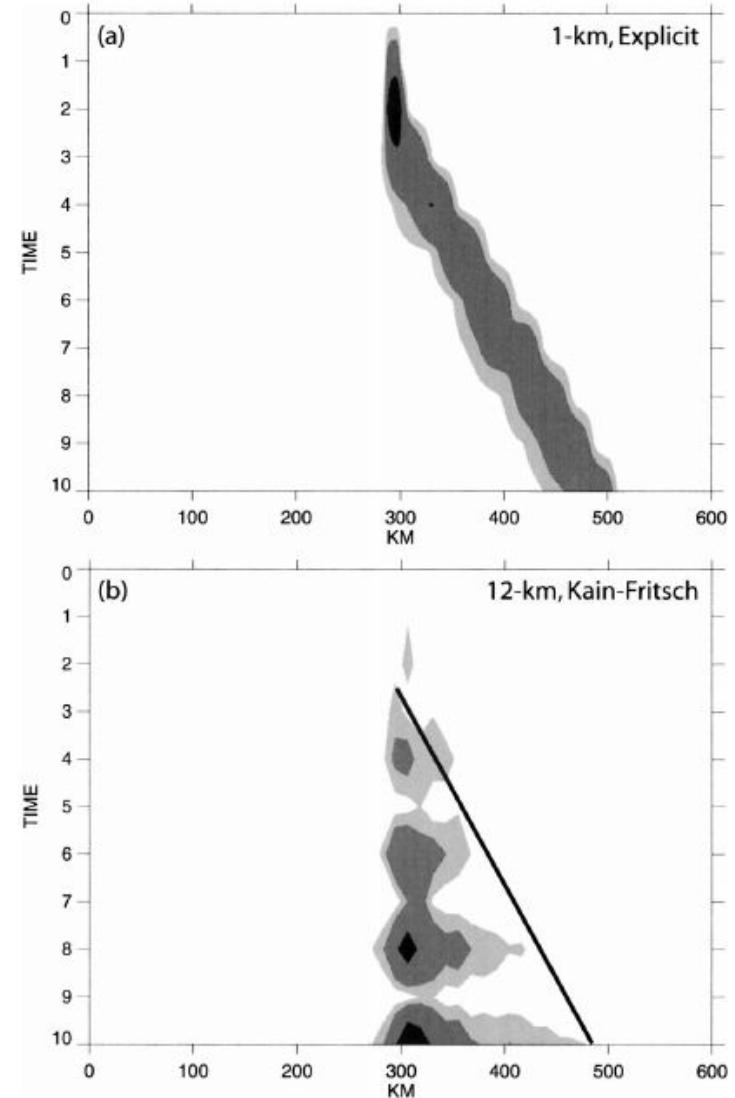
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- Improve the distribution and characteristics of deep convection in the tropics (Inoue et al. 2008)



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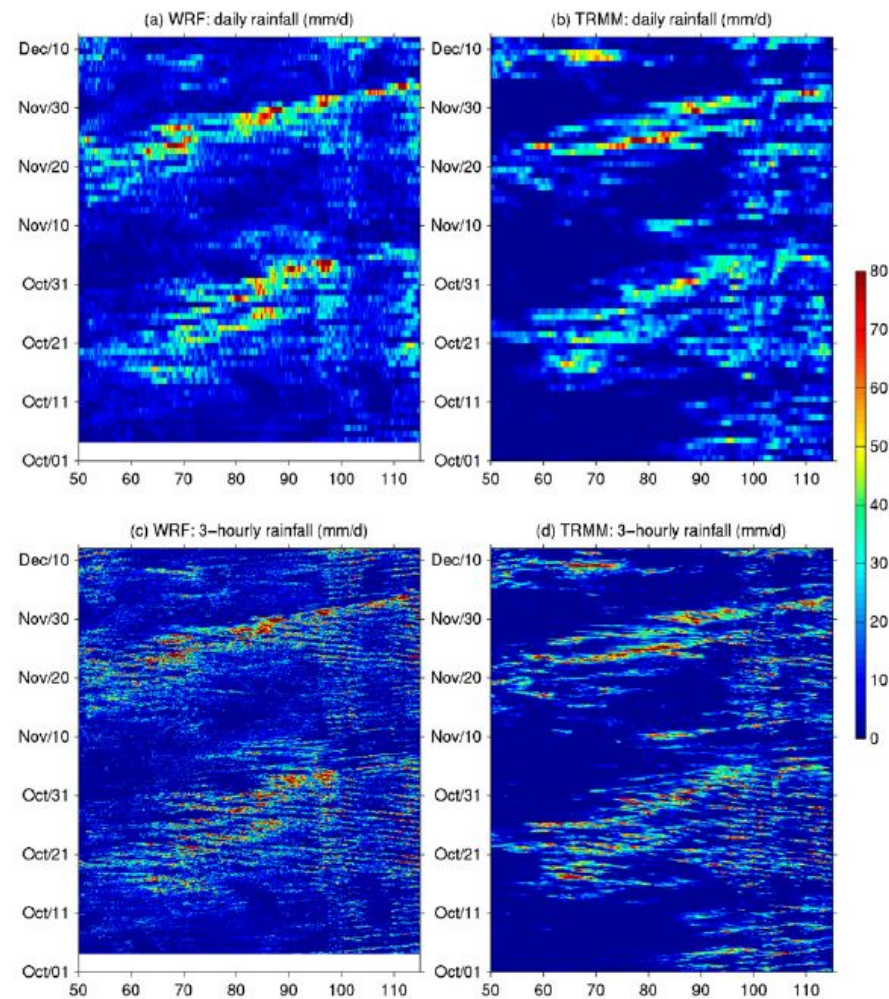
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- Improve the propagation characteristics of convection
  - In the midlatitudes (Davis et al. 2003)





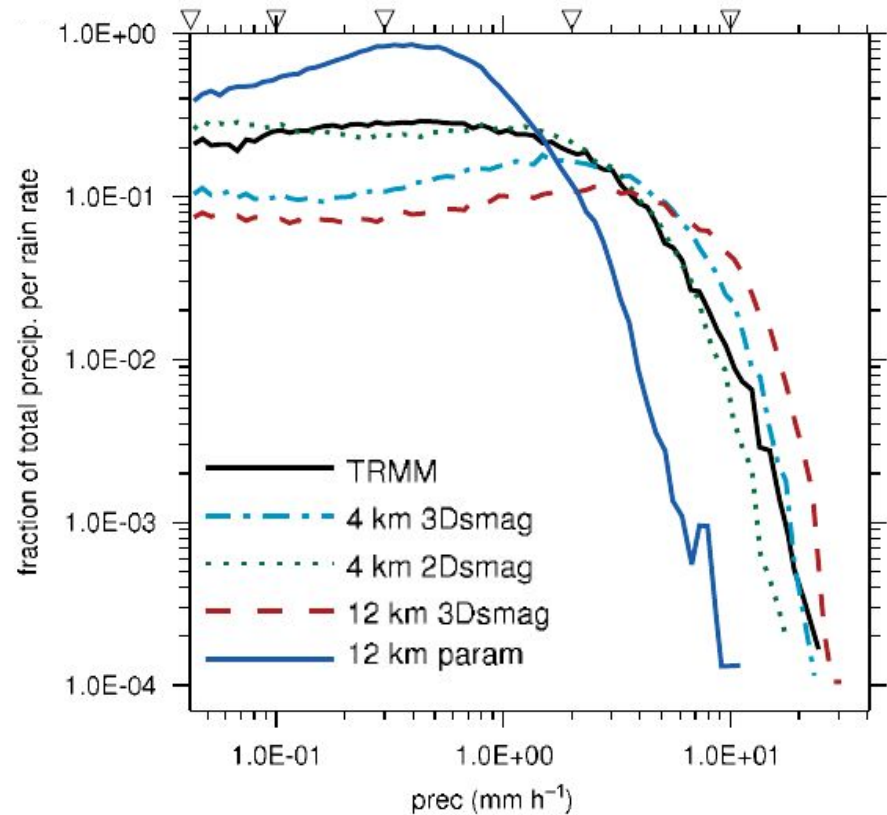
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- Realistically simulate MJO events (e.g., Miura et al. 2007)
- More accurately reproduce the rainfall diurnal cycle (Sato et al. 2009)
- Improve the distribution and characteristics of deep convection in the tropics (Inoue et al. 2008)
- Improve the propagation characteristics of convection
  - In the midlatitudes (Davis et al. 2003)
  - *And* in the tropics (Wang et al. 2015)



# Convection-permitting models are able to...

- Realistically simulate MJO events (e.g., Miura et al. 2007)
- More accurately reproduce the rainfall diurnal cycle (Sato et al. 2009)
- Improve the distribution and characteristics of deep convection in the tropics (Inoue et al. 2008)
- Improve the propagation characteristics of convection
  - In the midlatitudes (Davis et al. 2003)
  - *And* in the tropics (Wang et al. 2015)
- Mitigate the issue of “too much light rain” in many models (Holloway et al. 2012)





# Moving forward: a grand experiment

**The plan:** Test a convection-allowing subseasonal forecast framework

- a. Use a variable-resolution model, such as MPAS
- b. 1-2 month integrations
- c. High-resolution mesh (2-4 km) over the global tropics to allow explicit convection



## Questions that will be addressed:

1. How will explicit convection in the tropics change the mean state (i.e., the biases) in the model?
2. Does explicit convection produce more realistic propagating convection, like the MJO?
3. Will the improved tropical convection, and the associated midlatitude teleconnections, improve the fidelity of global subseasonal prediction?



**Thank you**

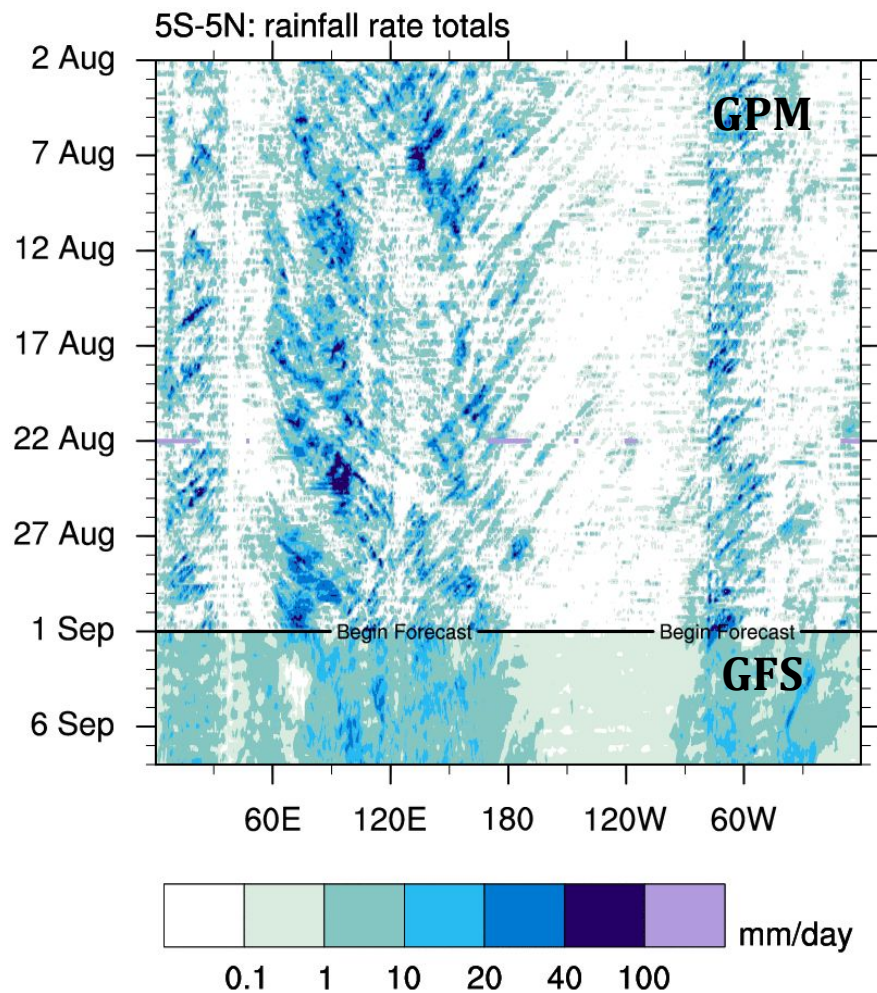
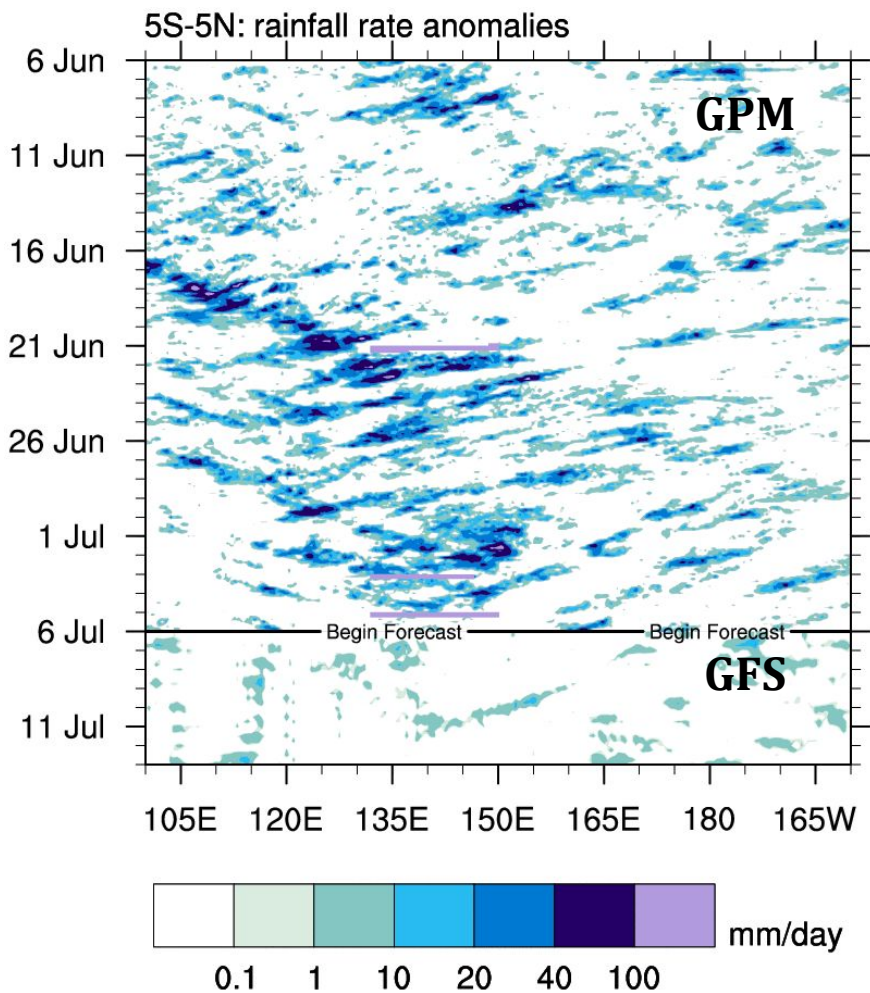
# References

- Davis, C. A., K. W. Manning, R. E. Carbone, S. B. Trier, and J. D. Tuttle, 2003. Coherence of warm-season continental rainfall in numerical weather prediction models. *Mon. Wea. Rev.*, **131**, 2667-2679.
- Holloway, C. E., S. J. Woolnough, and G. M. S. Lister, 2012. Precipitation distributions for explicit versus parametrized convection in a large-domain high-resolution tropical case study. *Q. J. R. Meteorol. Soc.*, **138**, 1692-1708.
- Inoue, T., M. Satoh, H. Miura, and B. Mapes, 2008. Characteristics of cloud size of deep convection simulated by a global cloud resolving model over the western tropical Pacific. *J. Meteor. Soc. Japan*, **86A**, 1-15.
- Kim, D., J.-S. Kug, and A. H. Sobel, 2014. Propagating versus nonpropagating Madden-Julian Oscillation events. *J. Climate*, **27**, 111-125.
- Miura, H., M. Satoh, T. Nasuno, A. Noda, and K. Oouchi, 2007. A Madden-Julian Oscillation event realistically simulated by a global cloud-resolving model. *Science*, **318**, 1763-1765.
- Saha, S and Coauthors, 2014. The NCEP Climate Forecast System version 2. *J. Climate*, **27**, 2185-2208.
- Sato, T., H. Miura, M. Satoh, Y. N. Takayabu, and Y. Wang, 2009. Diurnal cycle of precipitation in the tropics simulated in a global cloud-resolving model. *J. Climate*, **22**, 4809-4826.
- Wang, S., and Coauthors, 2015. Regional simulation of the October and November MJO events observed during the CINDY/DYNAMO field campaign at gray zone resolution. *J. Climate*, **28**, 2097-2119.
- Wheeler, M. and G. N. Kiladis, 1999. Convectively coupled equatorial waves: Analysis of clouds and temperature in the wavenumber-frequency domain. *J. Atmos. Sci.*, **56**, 374-399.

**Extra slide(s)**



# GFS tropical convection

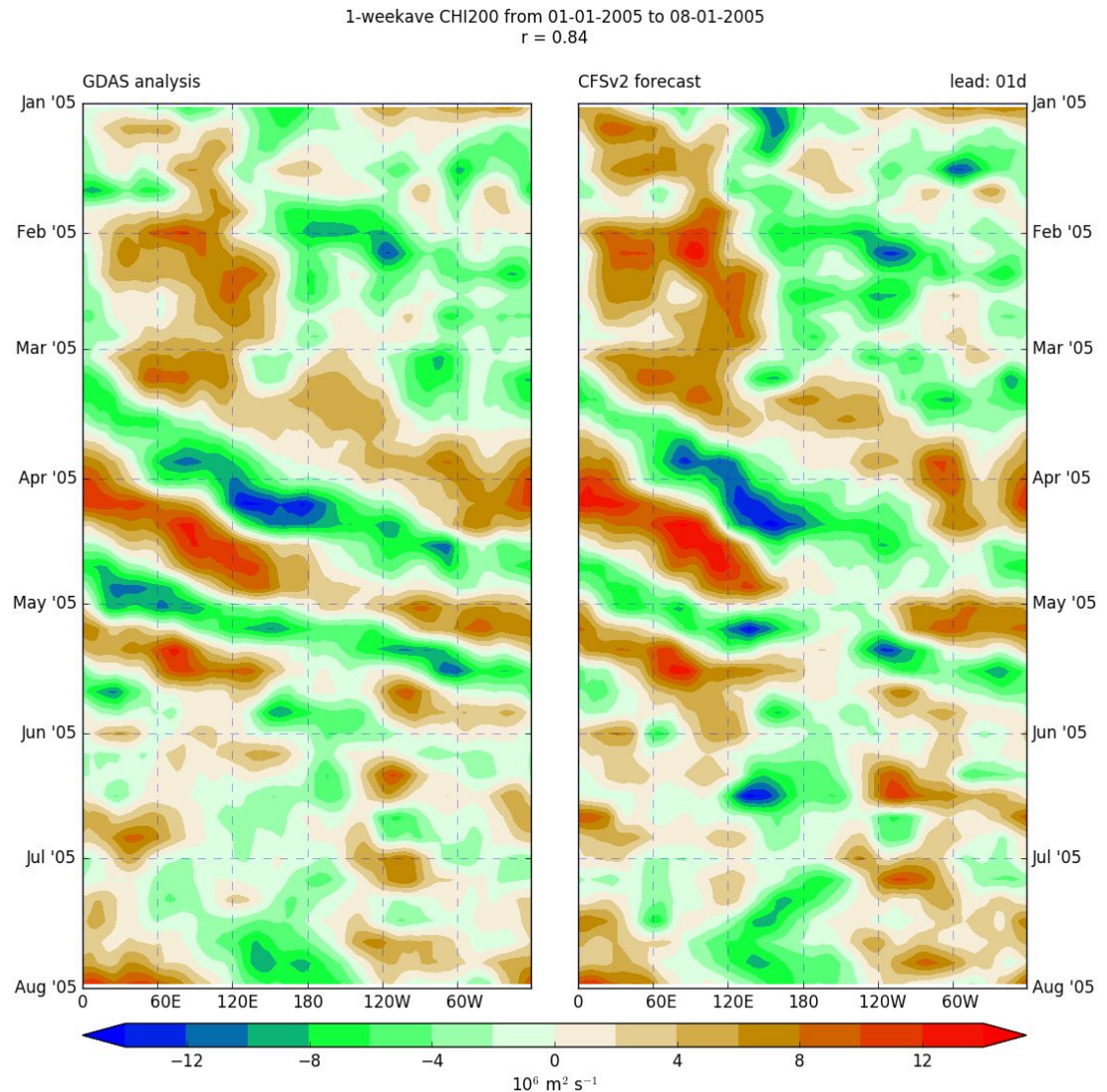


# CHI200 Hovmöller: analyses vs week-1 forecasts

Time period:

Spring/summer 2005

\* CHI200 averaged from 5S to 5N





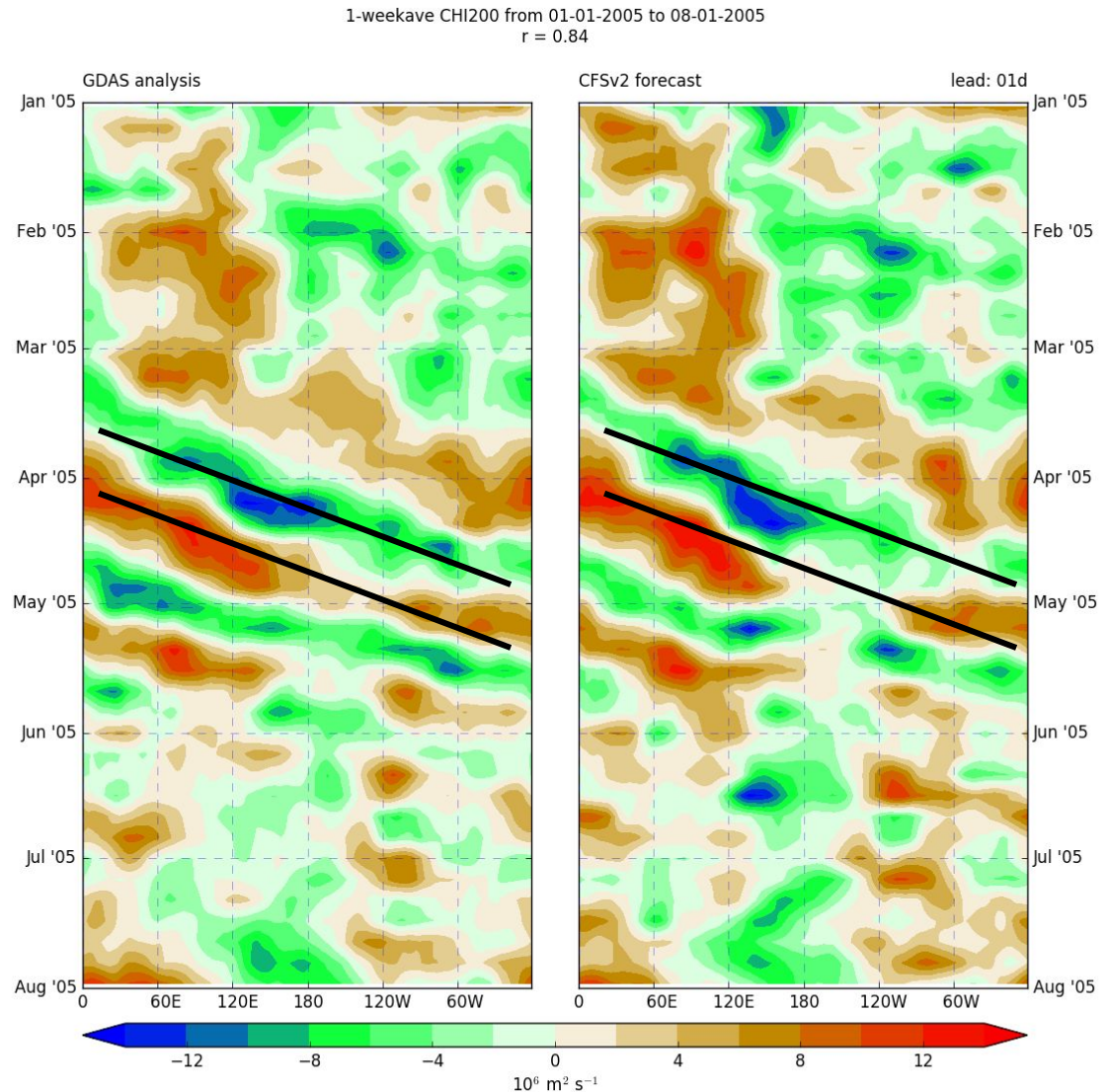
# CHI200 Hovmöller: analyses vs week-1 forecasts

Time period:

Spring/summer 2005

Short leads:

Clear propagating features in both the analyses and forecasts



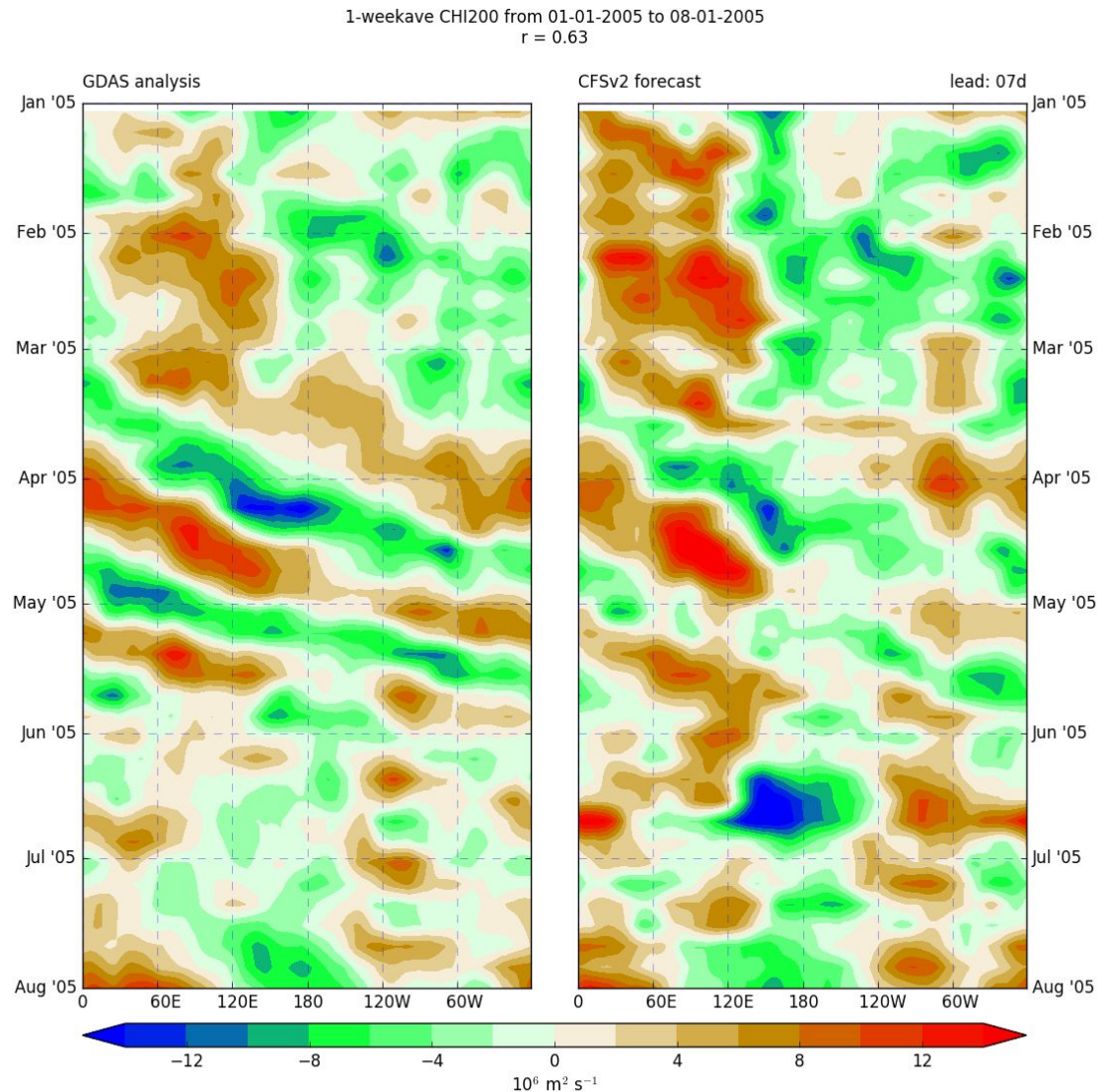
# CHI200 Hovmöller: analyses vs week-2 forecasts

Time period:

Spring/summer 2005

Short leads:

Clear propagating features in both the analyses and forecasts





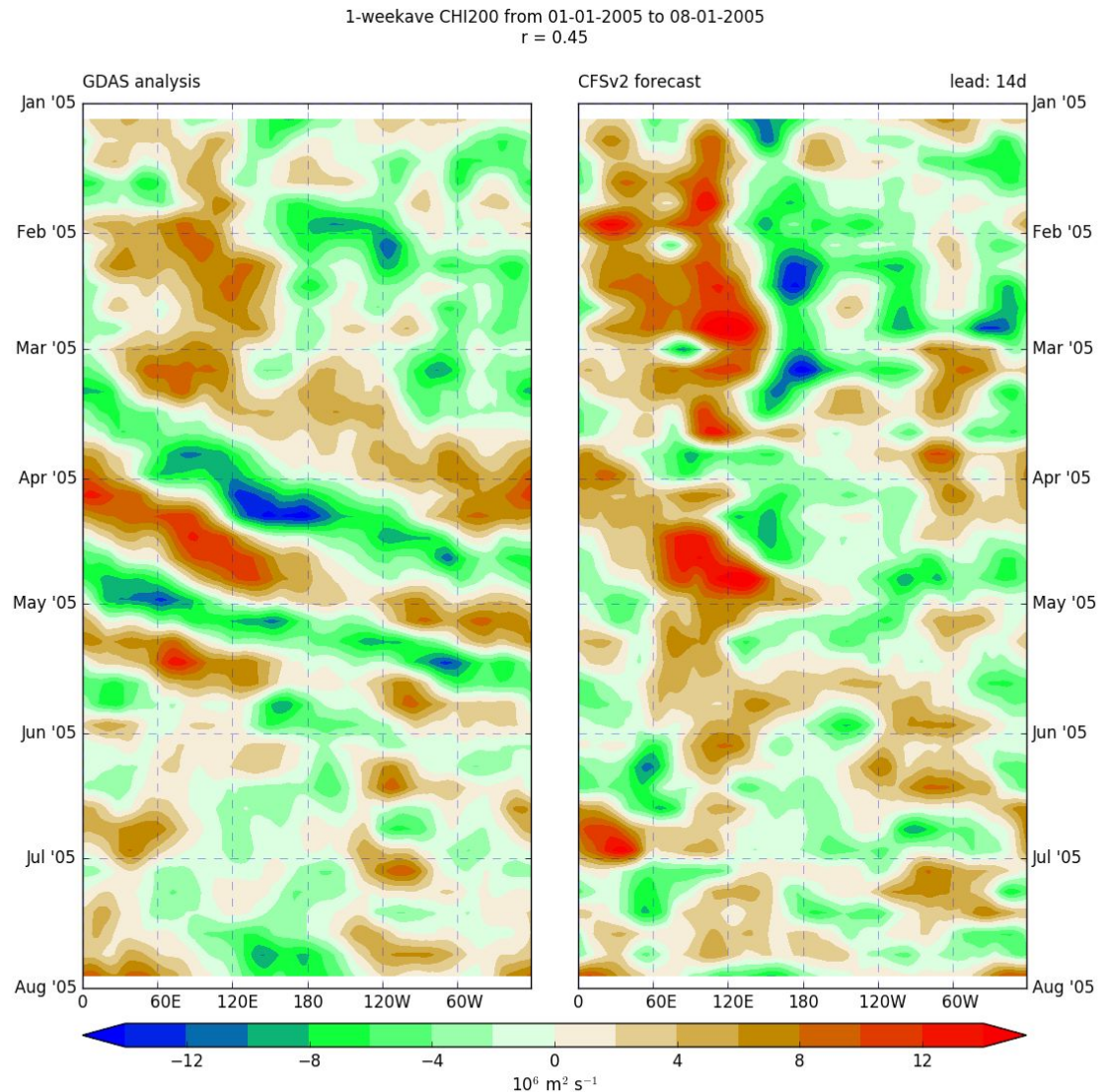
# CHI200 Hovmöller: analyses vs week-3 forecasts

Time period:

Spring/summer 2005

Short leads:

Clear propagating features in both the analyses and forecasts





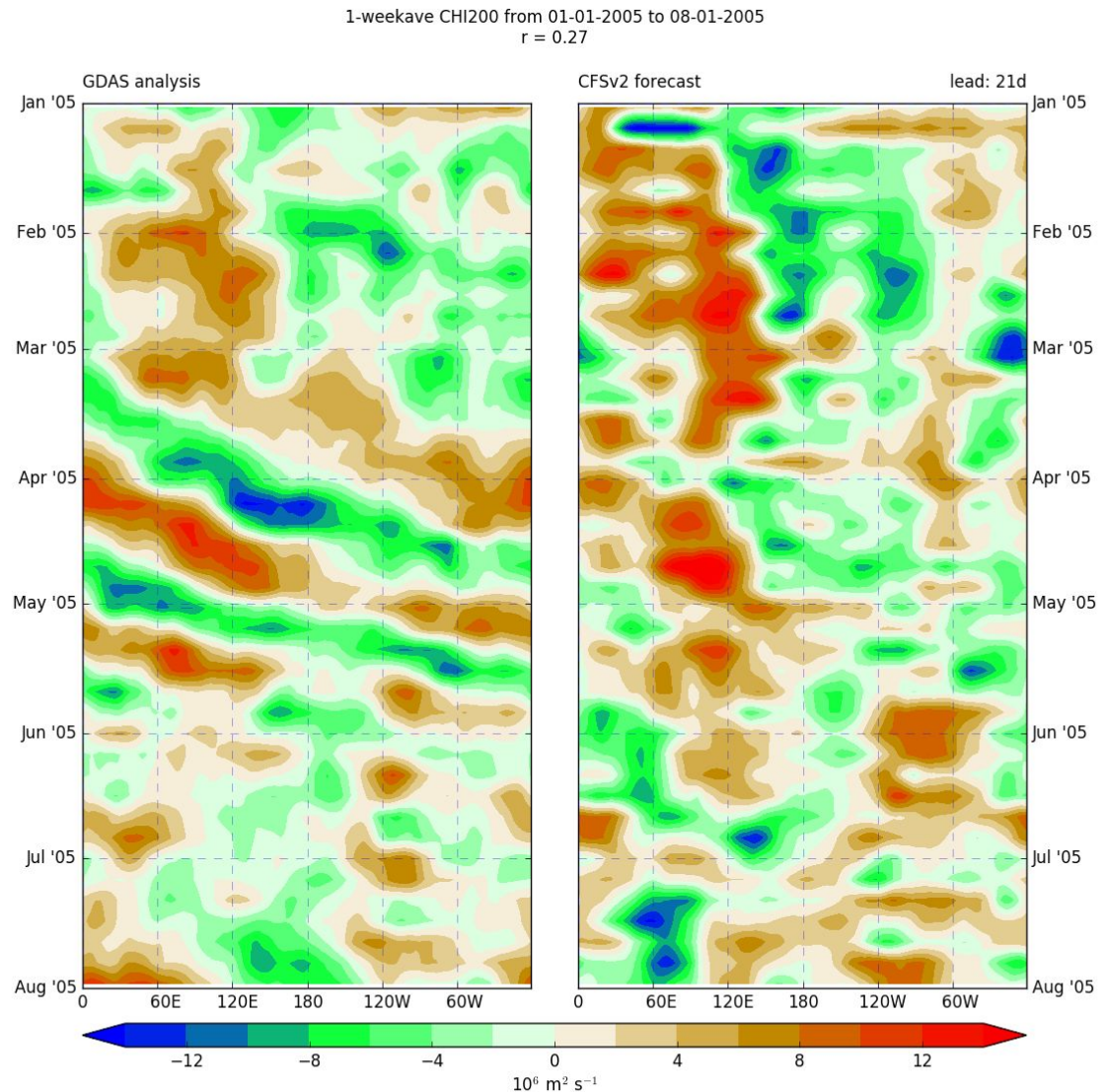
# CHI200 Hovmöller: analyses vs week-4 forecasts

Time period:

Spring/summer 2005

Short leads:

Clear propagating features in both the analyses and forecasts



# CHI200 Hovmöller: analyses vs week-5 forecasts

## Time period:

Spring/summer 2005

## Short leads:

Clear propagating features in both the analyses and forecasts

## Long leads:

Coherent propagating structures are lost, while more stationary patterns develop

