

Atmospheric River and S2S Prediction Research

Colorado State University

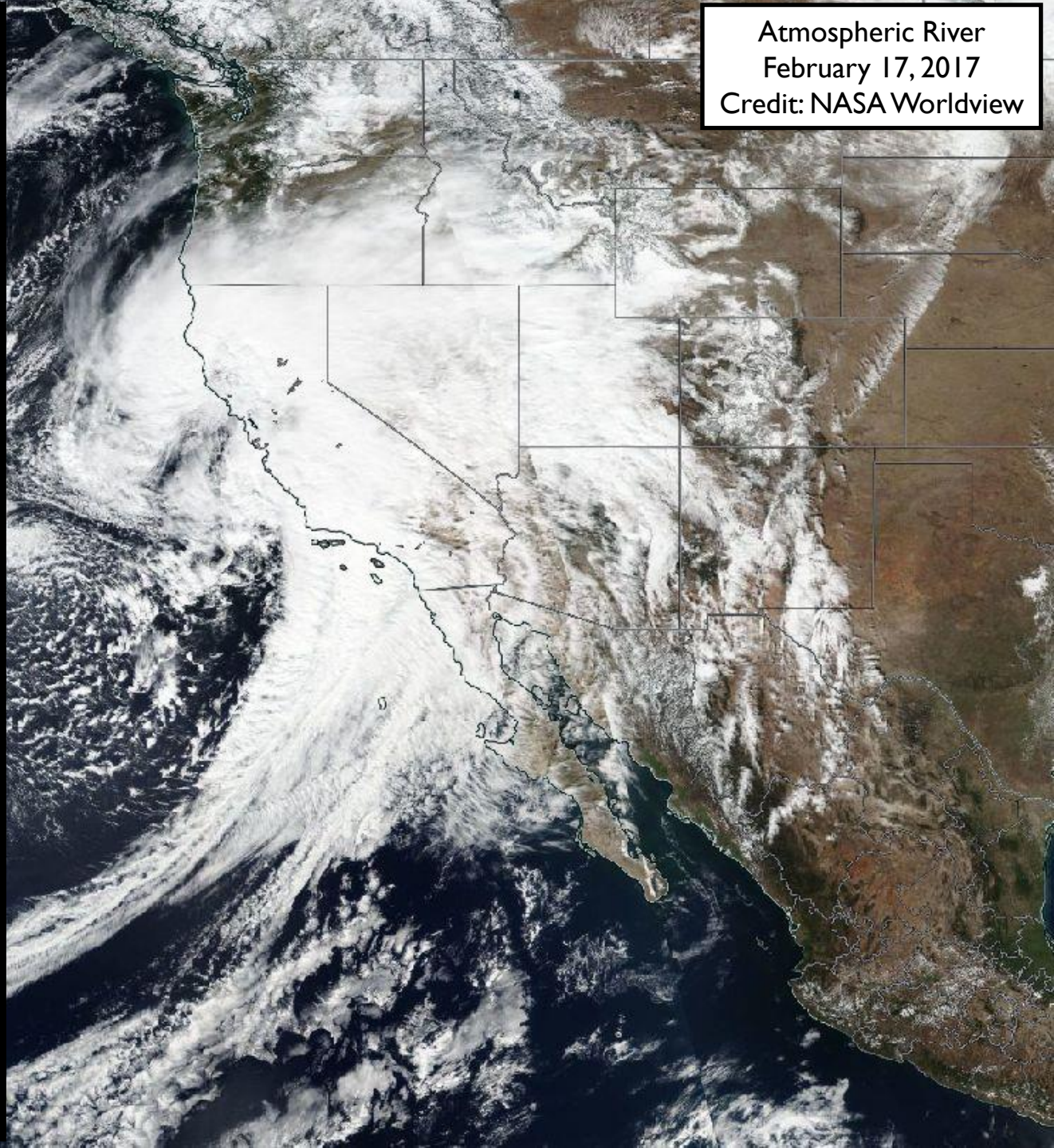
Cory Baggett
(cbaggett@rams.colostate.edu)

Western States Water Council
Workshop

May 15, 2018



Atmospheric River
February 17, 2017
Credit: NASA Worldview



The Researchers



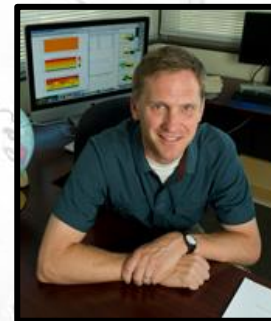
Cory F. Baggett
Research Scientist at CSU



Elizabeth A. Barnes
Professor at CSU



Bryan D. Mundhenk
USAF Lt Co at Asheville, NC



Eric D. Maloney
Professor at CSU



Kai-Chih Tseng
Graduate Student at CSU



Kyle M. Nardi
Graduate Student at CSU

Atmospheric Rivers (ARs)

Atmospheric River
February 17, 2017
Credit: NASA Worldview

Atmospheric rivers are "long, narrow, and transient corridors of strong horizontal water vapor transport that frequently lead to heavy precipitation. They are the largest 'rivers' of fresh water on Earth, transporting on average more than double the flow of the Amazon River."

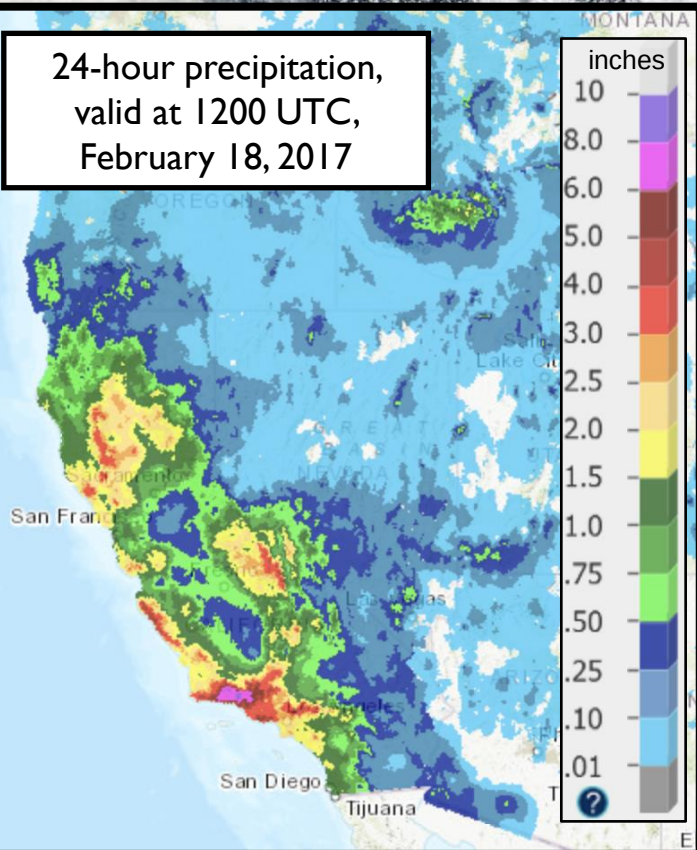
- AMS Glossary



Atmospheric Rivers (ARs)

Atmospheric River
February 17, 2017
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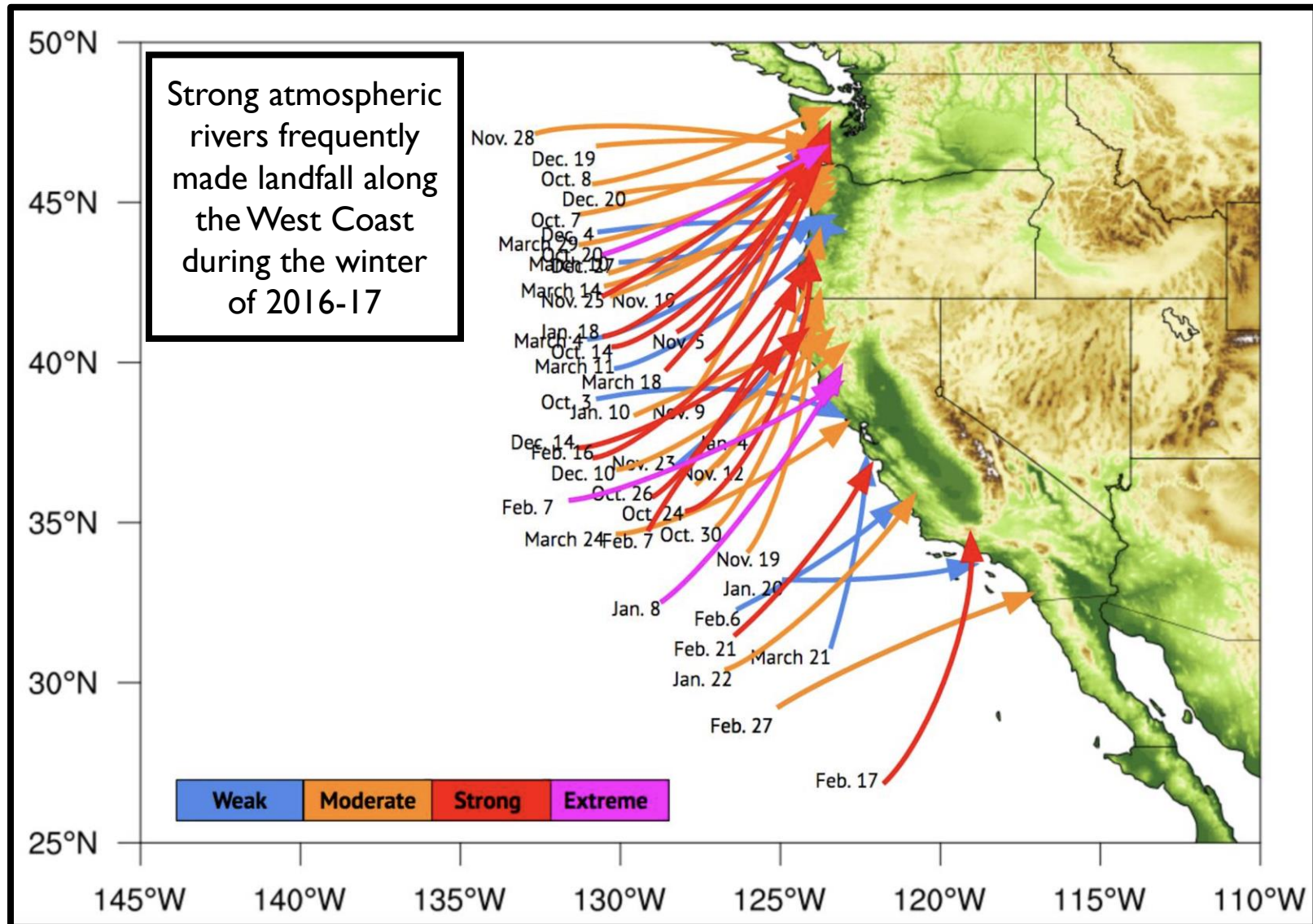
24-hour precipitation,
valid at 1200 UTC,
February 18, 2017



Atmospheric rivers are crucial to water supply along the U.S. West Coast:

- Account for 20-50% of California's precipitation and streamflow (Dettinger et al. 2011)
- Break ~40% of California's droughts (Dettinger 2013)

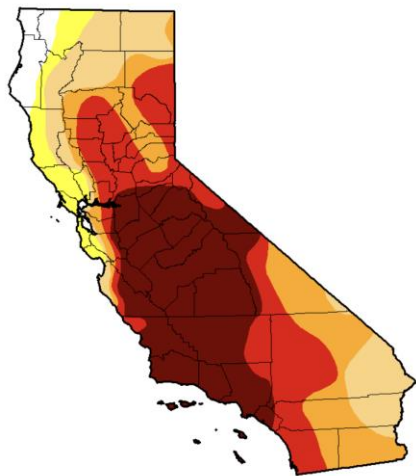
Atmospheric River Landfalls 2016-17



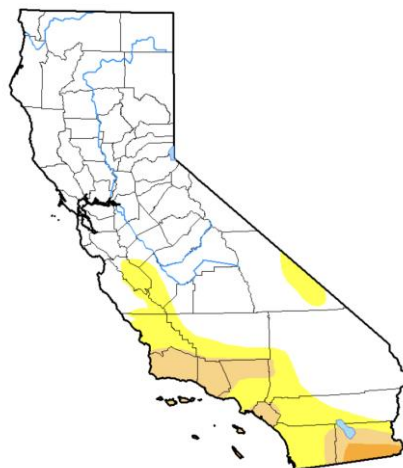
Ebb and Flow of California Droughts



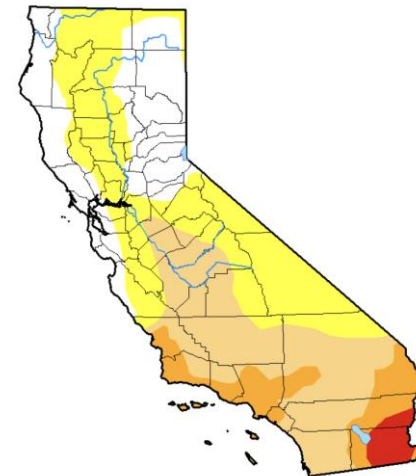
April 5, 2016



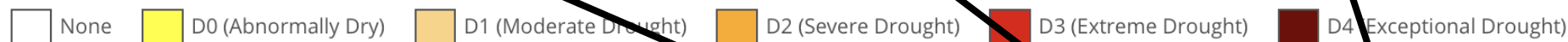
April 4, 2017



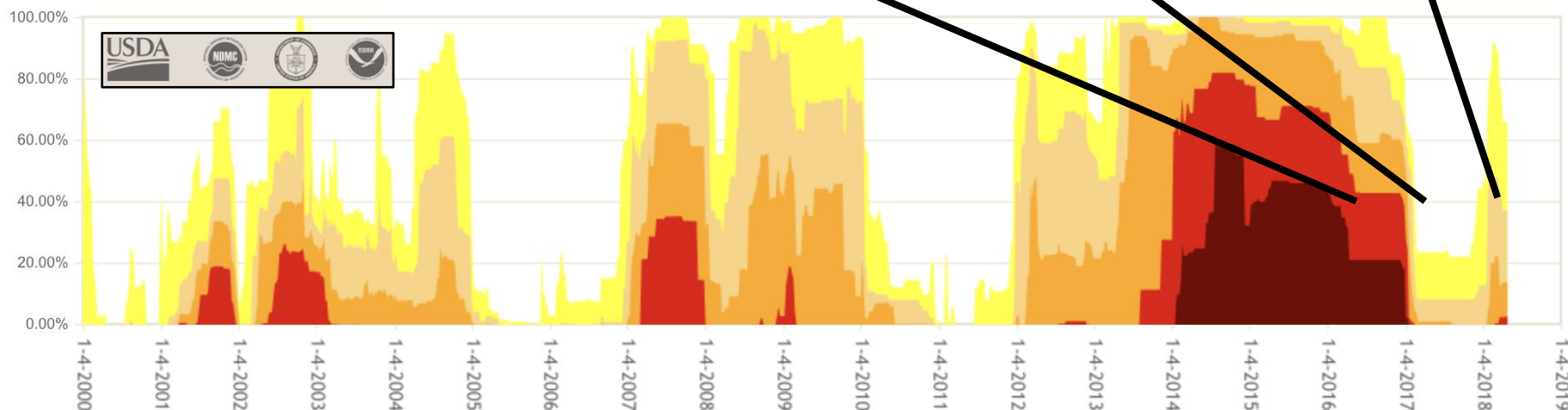
April 3, 2018



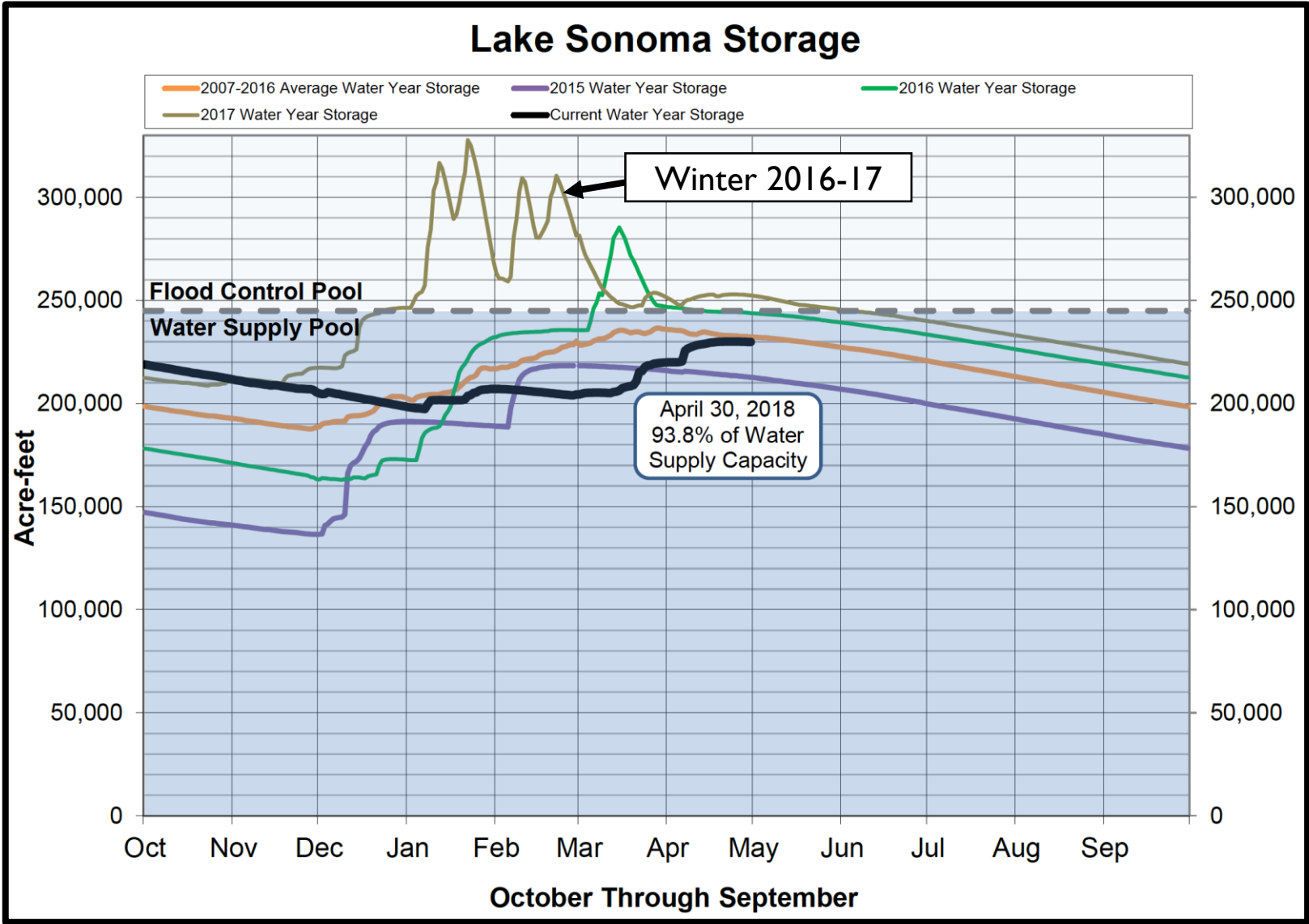
Drought Classification



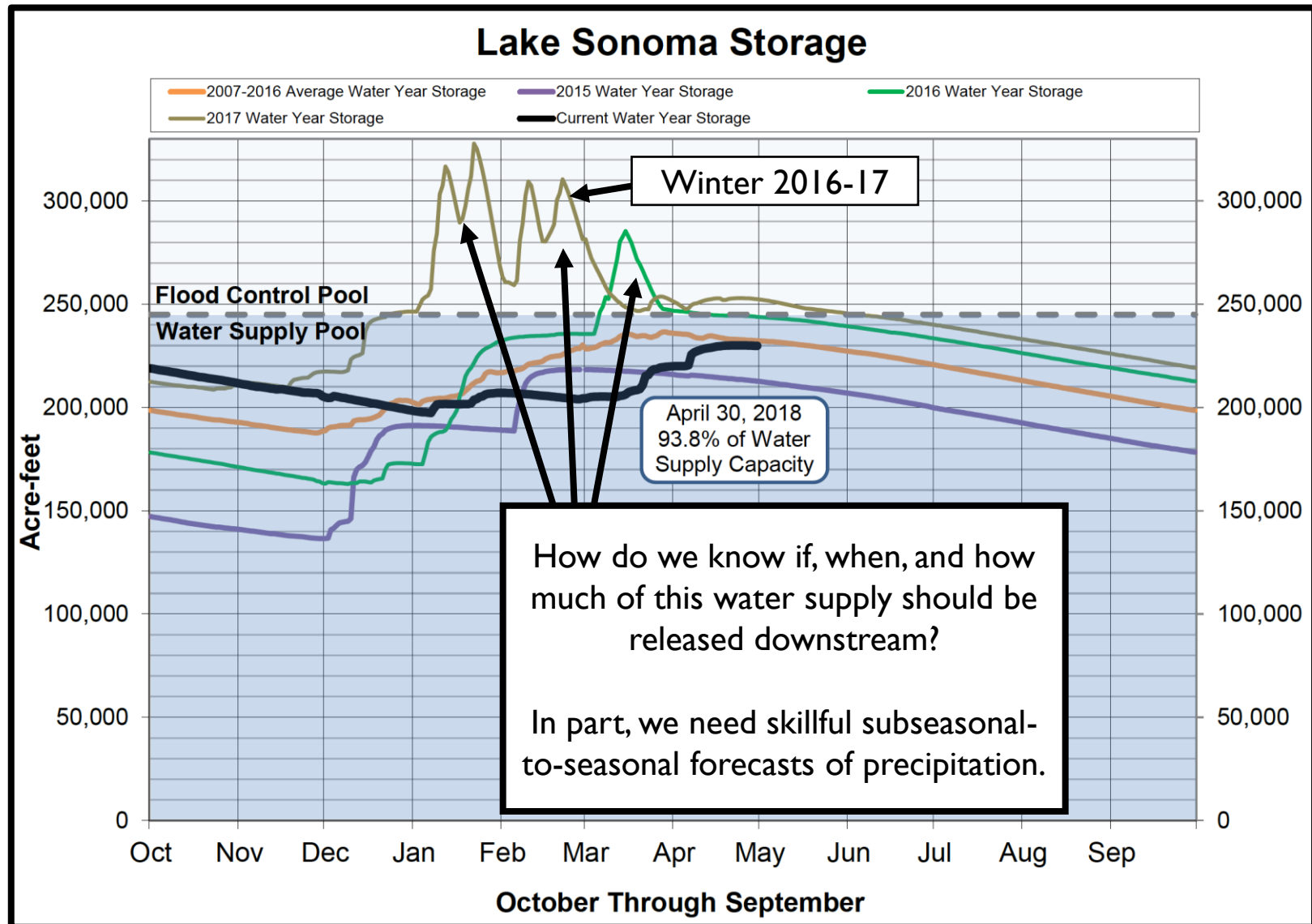
California Drought Severity by % Area 2000-2018



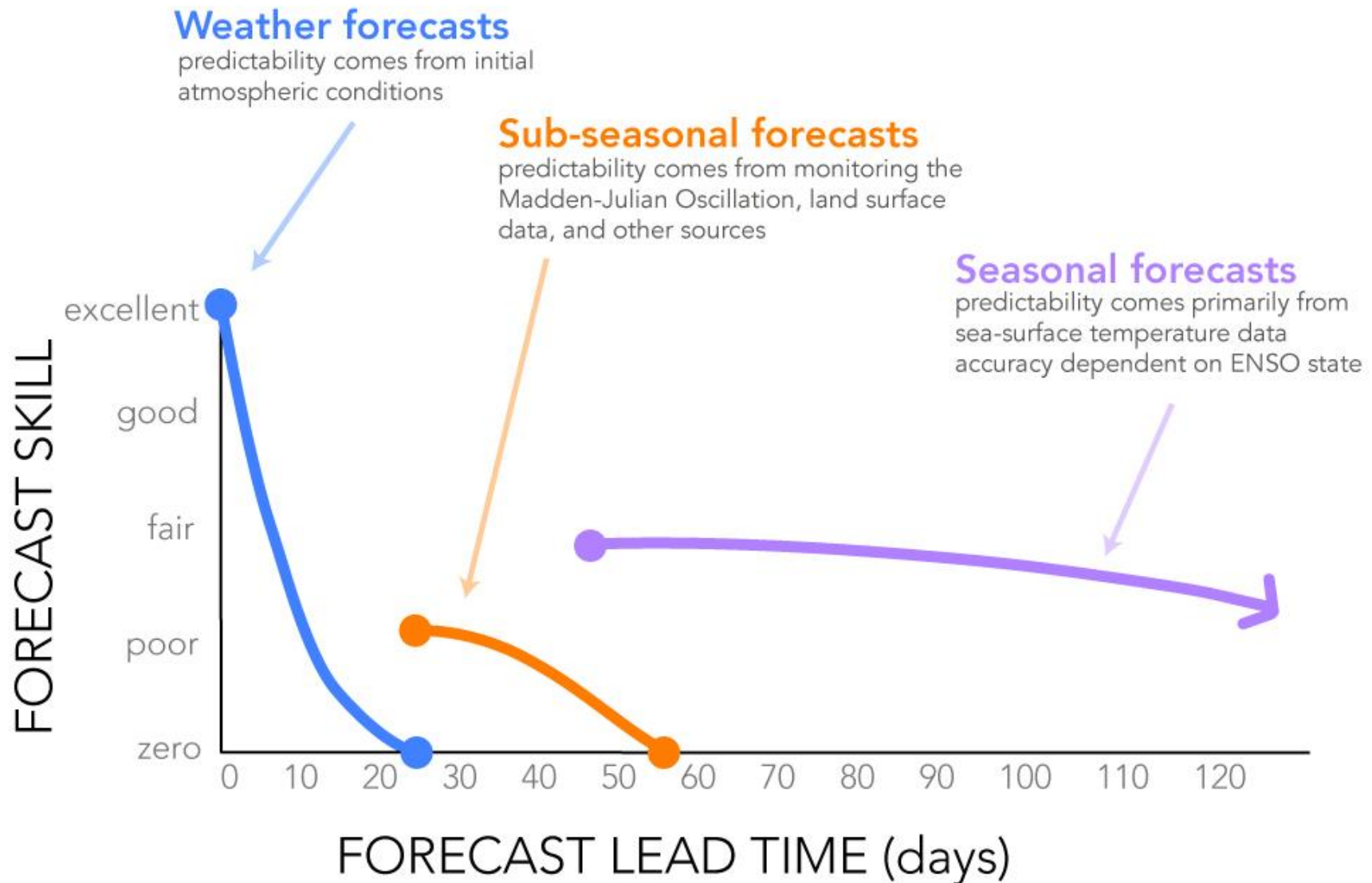
Lake Sonoma Reservoir Conditions



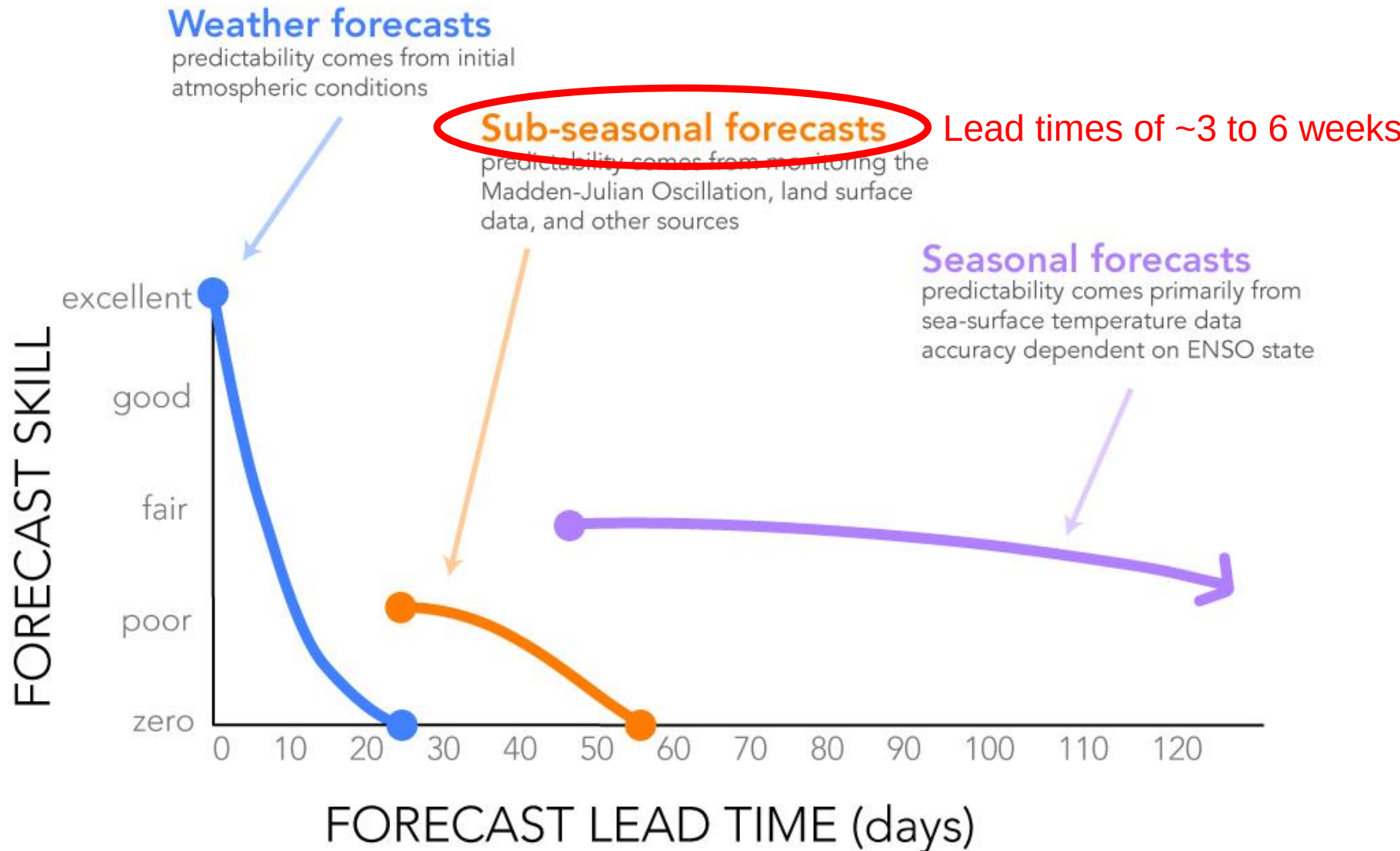
Lake Sonoma Reservoir Conditions



The Subseasonal-to-Seasonal (S2S) Forecast Gap



The Subseasonal-to-Seasonal (S2S) Forecast Gap

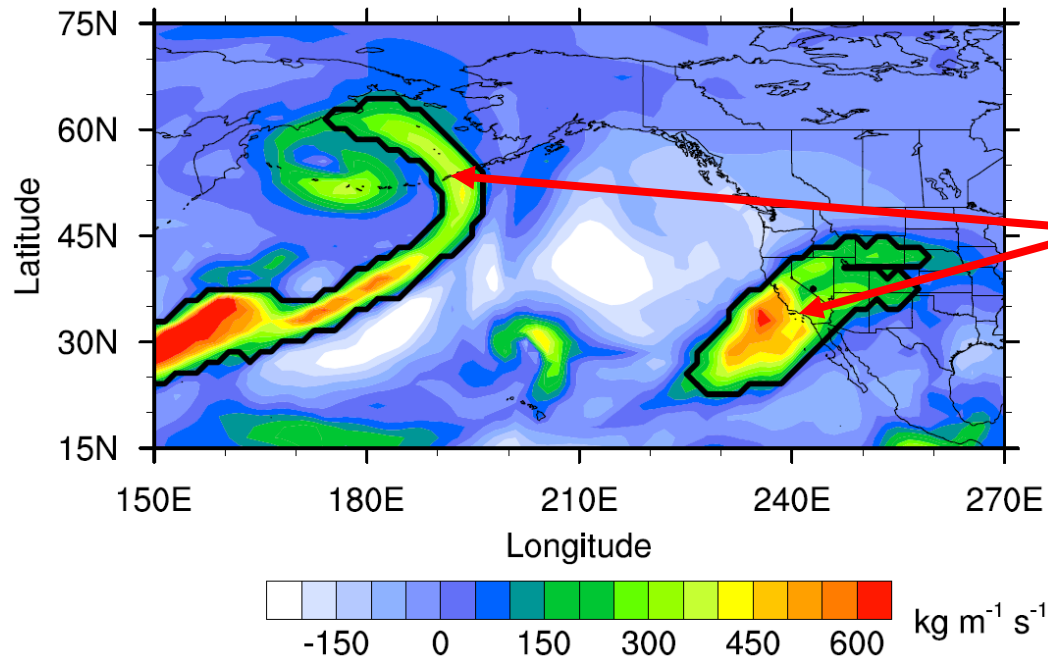




Can state-of-the-art numerical weather models help us in forecasting atmospheric river activity at subseasonal-to-seasonal lead times (~3 to 6 weeks in advance)?

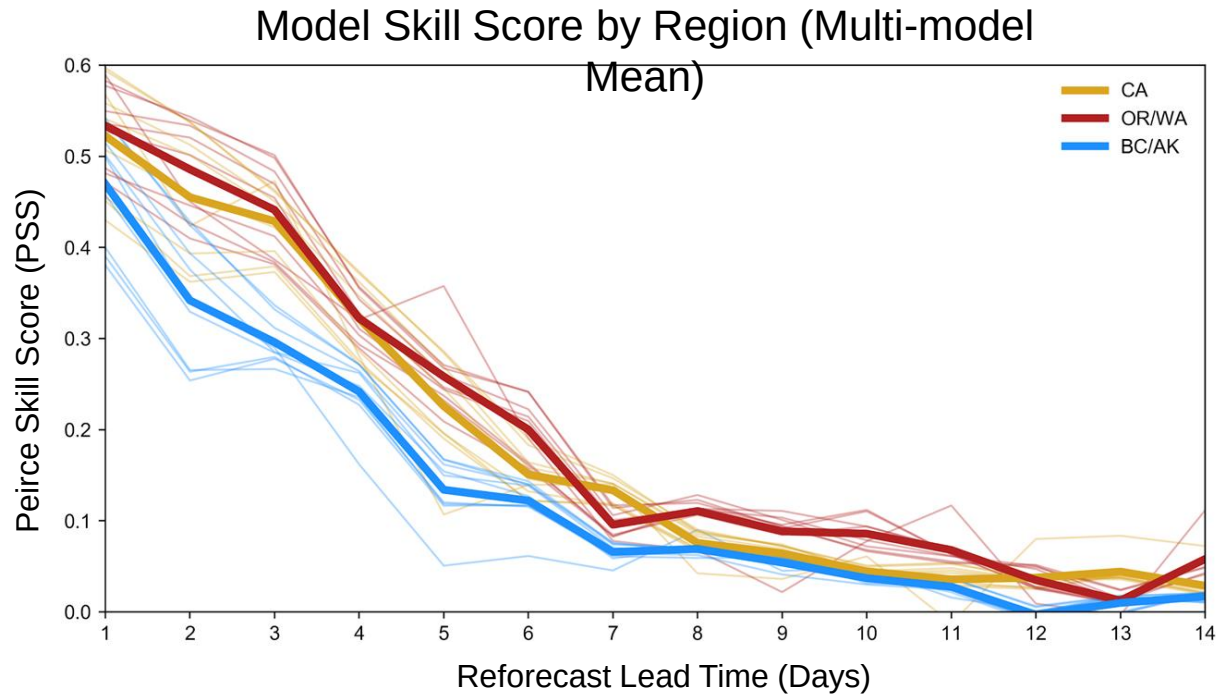
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Snapshot of integrated vapor transport



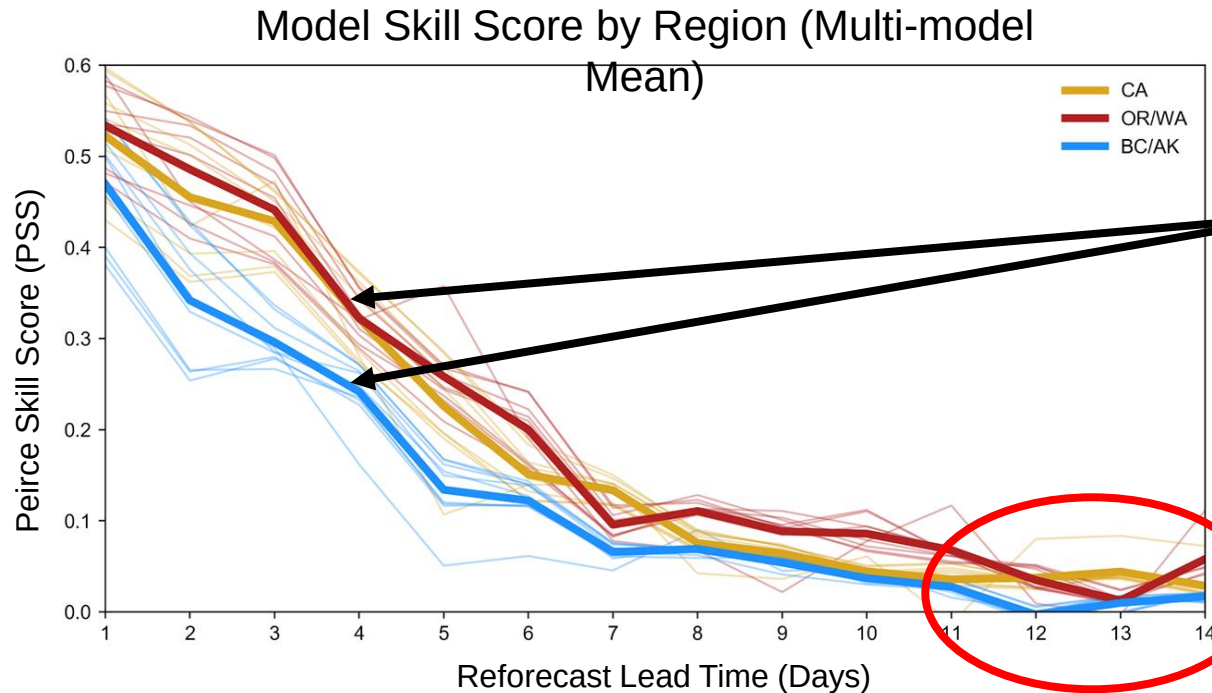
We identify atmospheric rivers in weather models and observations (ERA-Interim reanalysis data) using fields of water vapor transport (IVT) (Mundhenk et al. 2016).

Atmospheric River Activity - Model Skill Scores



Reforecasts from: BOM, CMA, CNRM,
ECCC, ECMWF, HMCR, JMA, NCEP,
UKMO

Atmospheric River Activity - Model Skill Scores



Model skill varies by region, with lower skill in BC/AK compared to CA and OR/WA

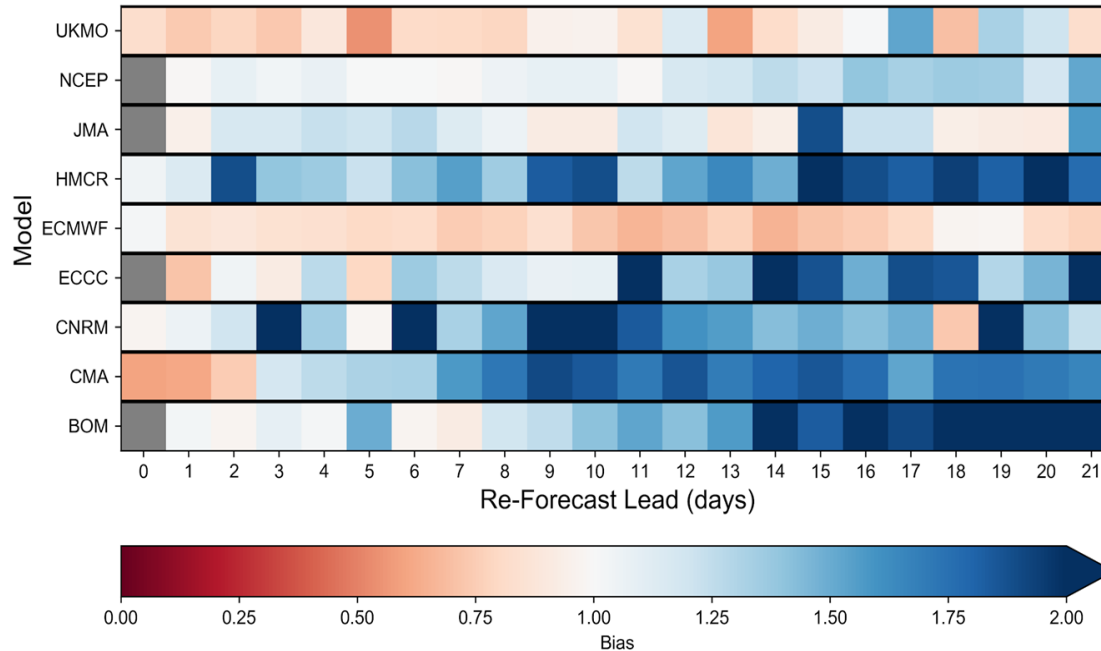
At lead times approaching 14 days, models have no additional skill compared to a climatological forecast

Reforecasts from: BOM, CMA, CNRM, ECCO, ECMWF, HMCR, JMA, NCEP, UKMO

Atmospheric River Activity - Model Bias



AR Bias for the West Coast of North America

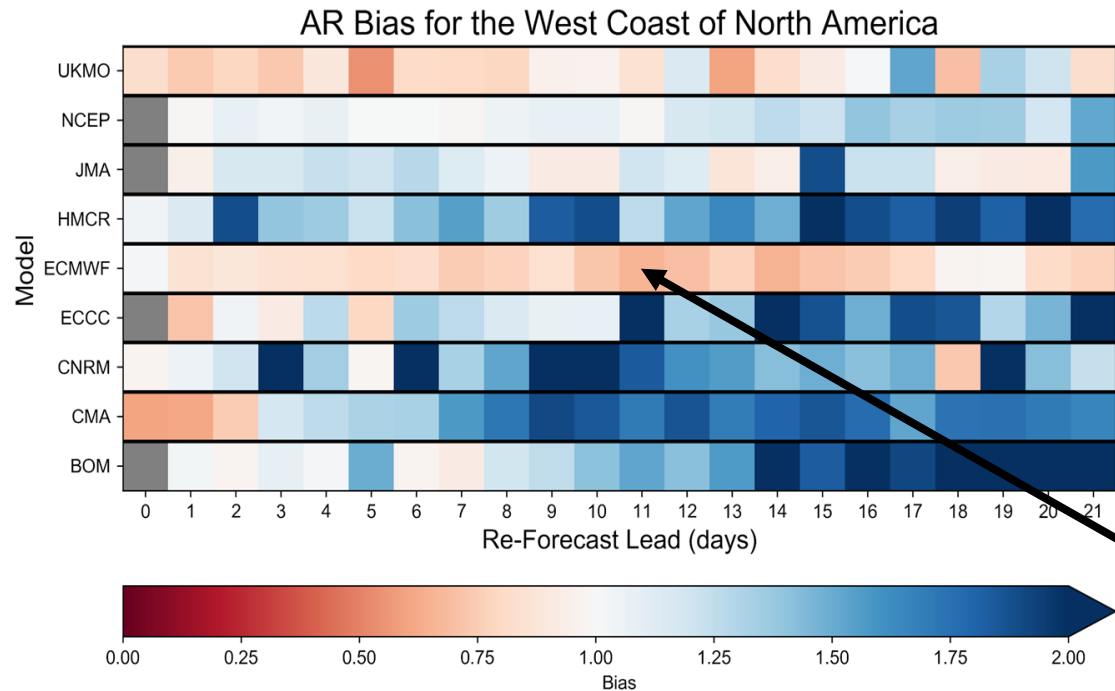


Blue indicates more
ARs forecasted than
observed

Red indicates less
ARs forecasted than
observed

$$\text{Model Bias} = \frac{\# \text{ of Forecasted ARs}}{\# \text{ of Observed ARs}}$$

Atmospheric River Activity - Model Bias



Blue indicates more ARs forecasted than observed

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ECMWF has a “quiet” bias

$$\text{Model Bias} = \frac{\# \text{ of Forecasted ARs}}{\# \text{ of Observed ARs}}$$



Can state-of-the-art numerical weather models help us in forecasting atmospheric river activity at subseasonal-to-seasonal lead times (~3 to 6 weeks in advance)?

Apparently no, the models have skillful lead times only extending to ~14 days, in agreement with prior studies (Wick et al. 2013; Nayak et al. 2014; DeFlorio et al. 2018).



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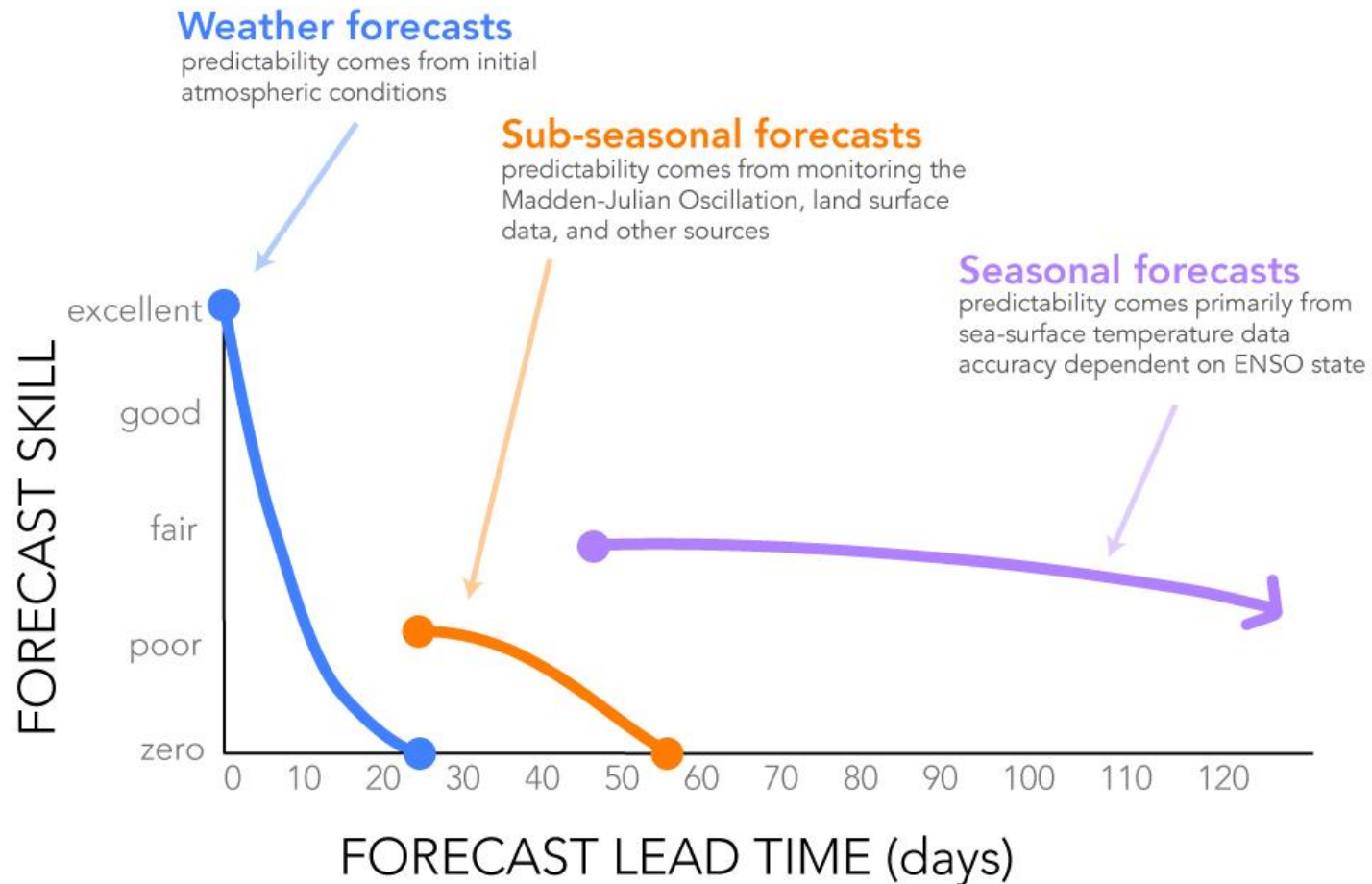
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Question



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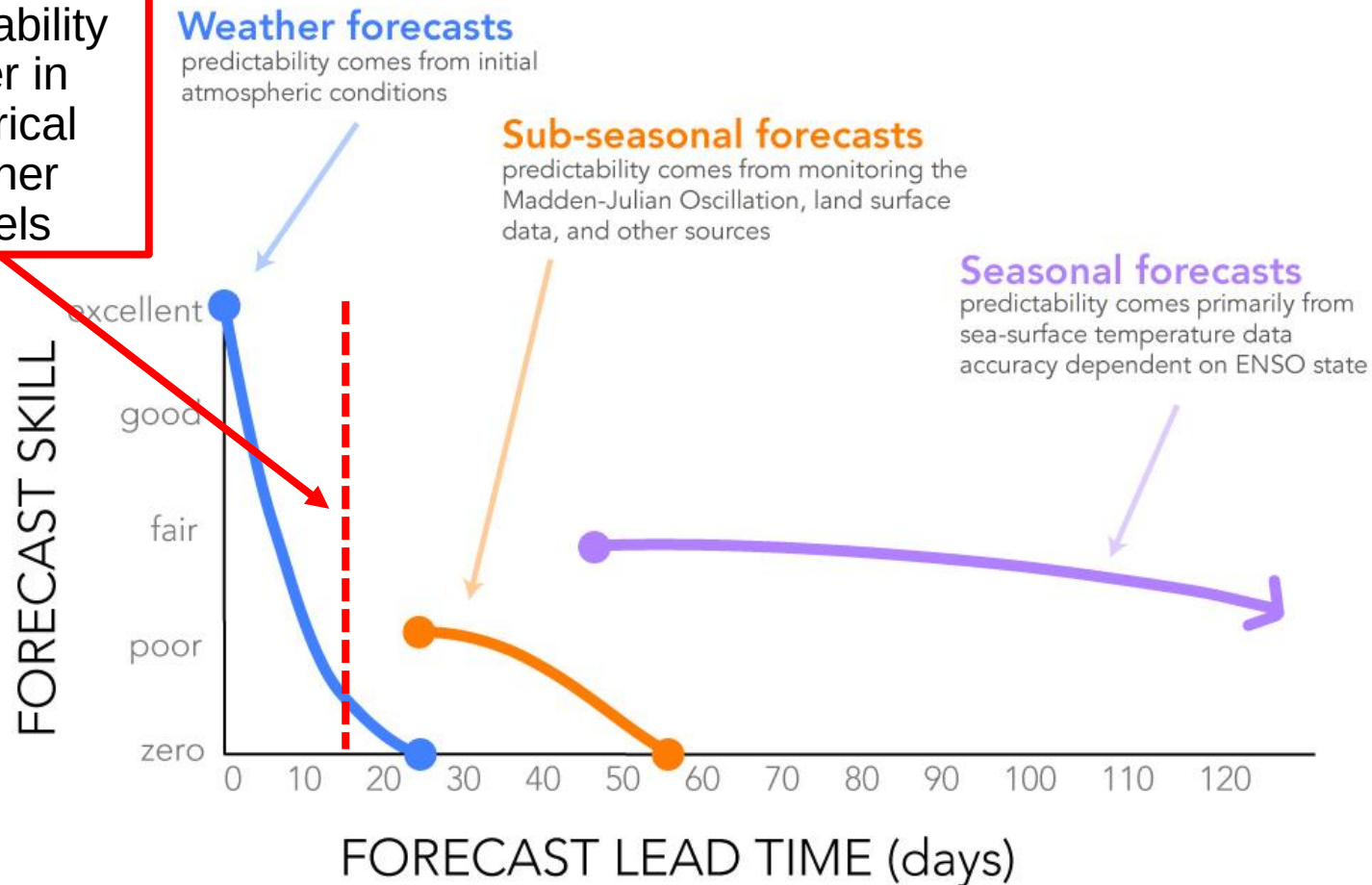


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Predictability
barrier in
numerical
weather
models



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Predictability barrier in numerical weather models

Weather forecasts

predictability comes from initial atmospheric conditions

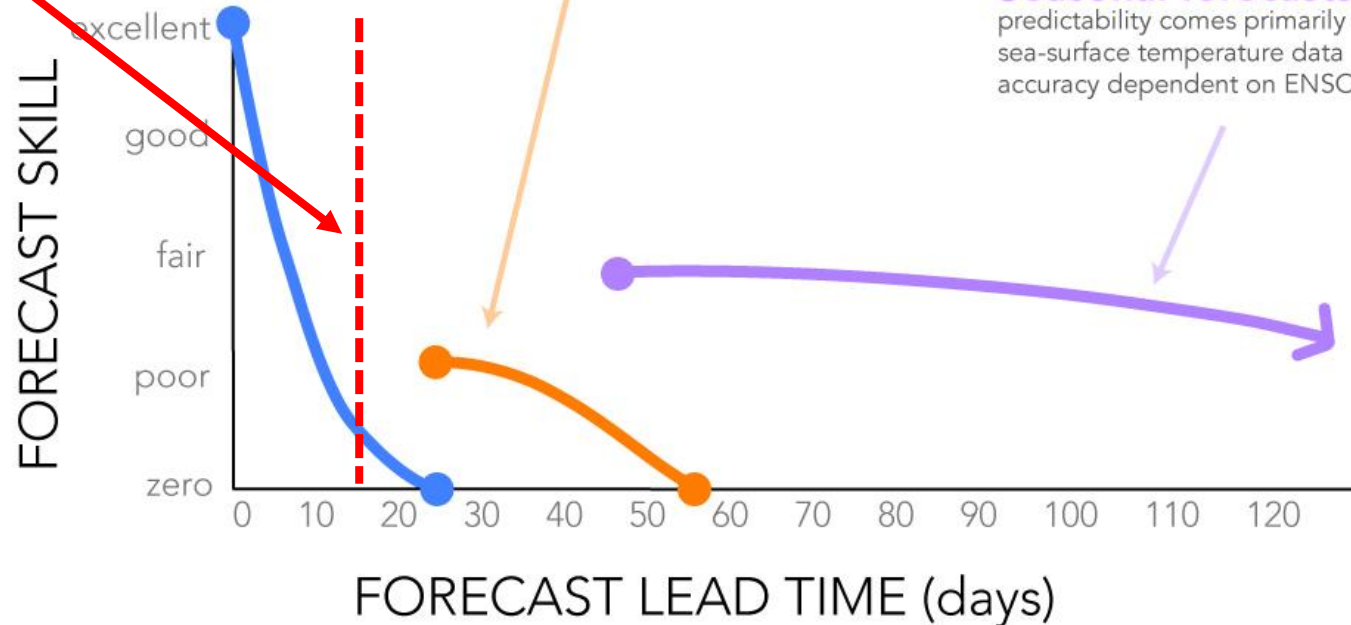
Sub-seasonal forecasts

predictability comes from monitoring the Madden-Julian Oscillation, land surface data, and other sources

The Madden-Julian Oscillation (MJO) ?

Seasonal forecasts

predictability comes primarily from sea-surface temperature data accuracy dependent on ENSO state



Question



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Predictability barrier in numerical weather models

Weather forecasts

predictability comes from initial atmospheric conditions

Sub-seasonal forecasts

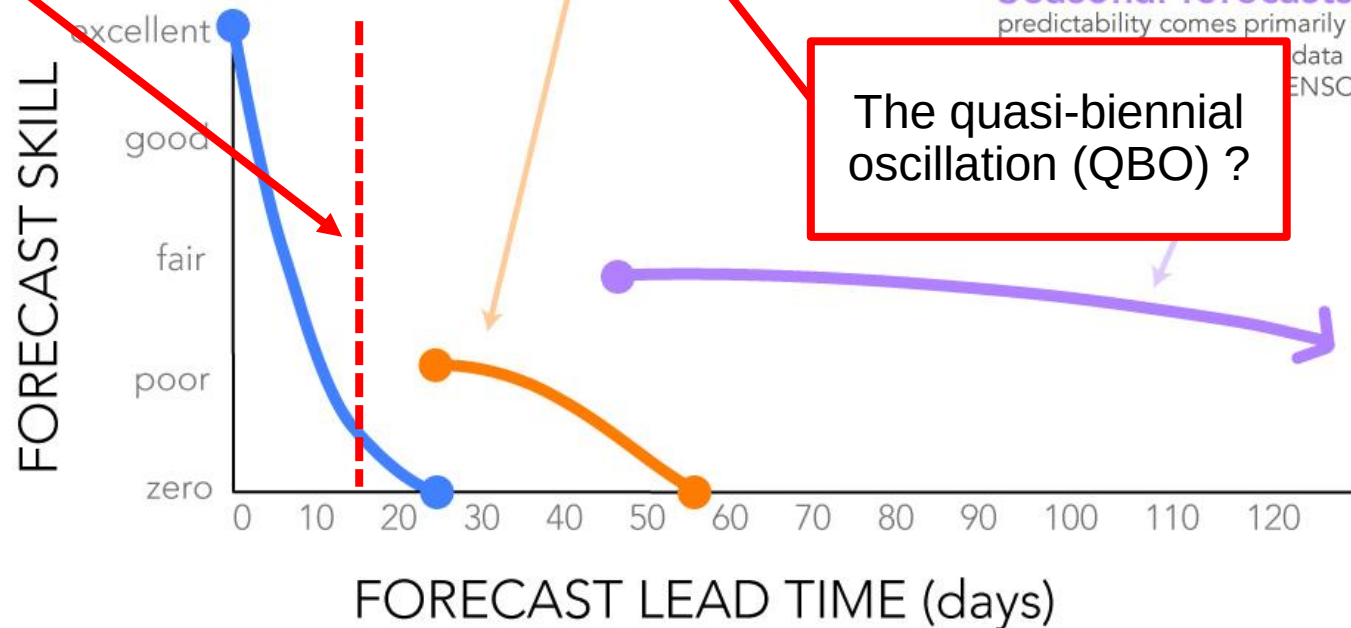
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The Madden-Julian Oscillation (MJO) ?

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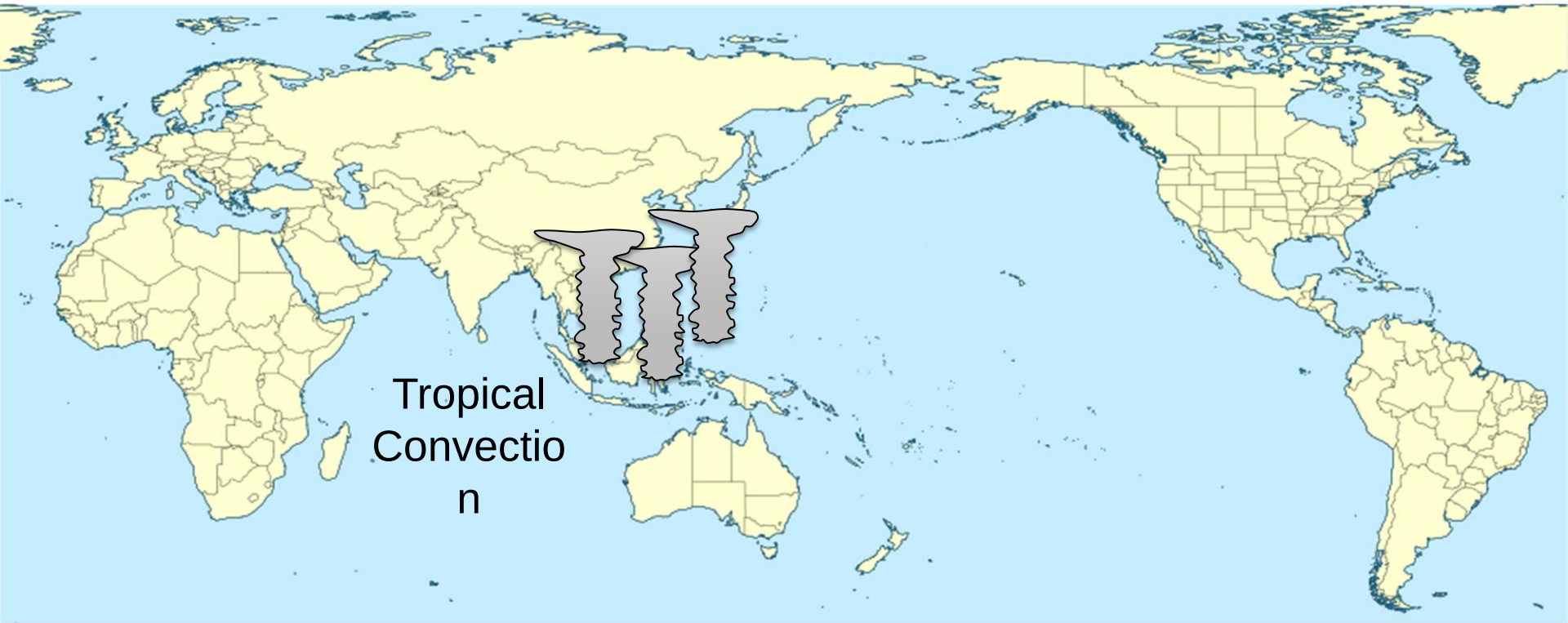
predictability comes primarily from data and ENSO state

The quasi-biennial oscillation (QBO) ?



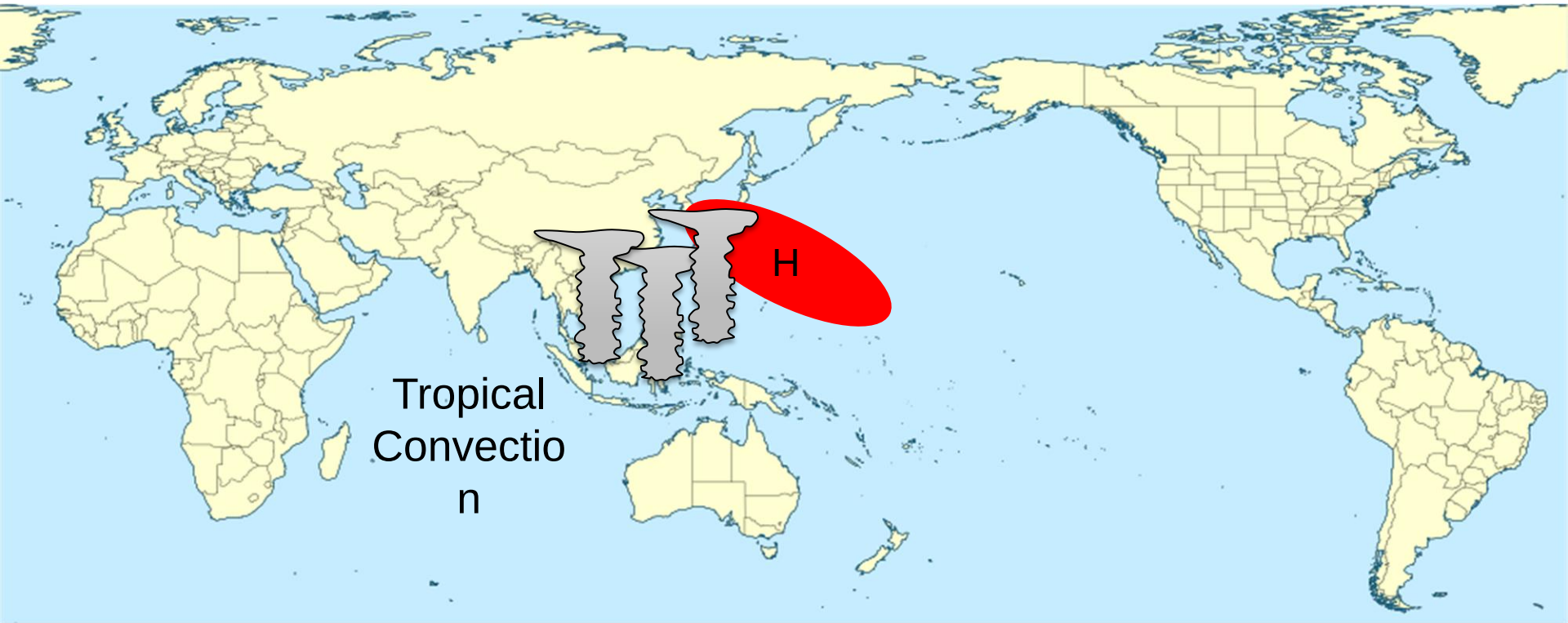
The Madden-Julian Oscillation – a source of predictability at S2S lead times

Since the 1980s, we have known...



The Madden-Julian Oscillation – a source of predictability at S2S lead times

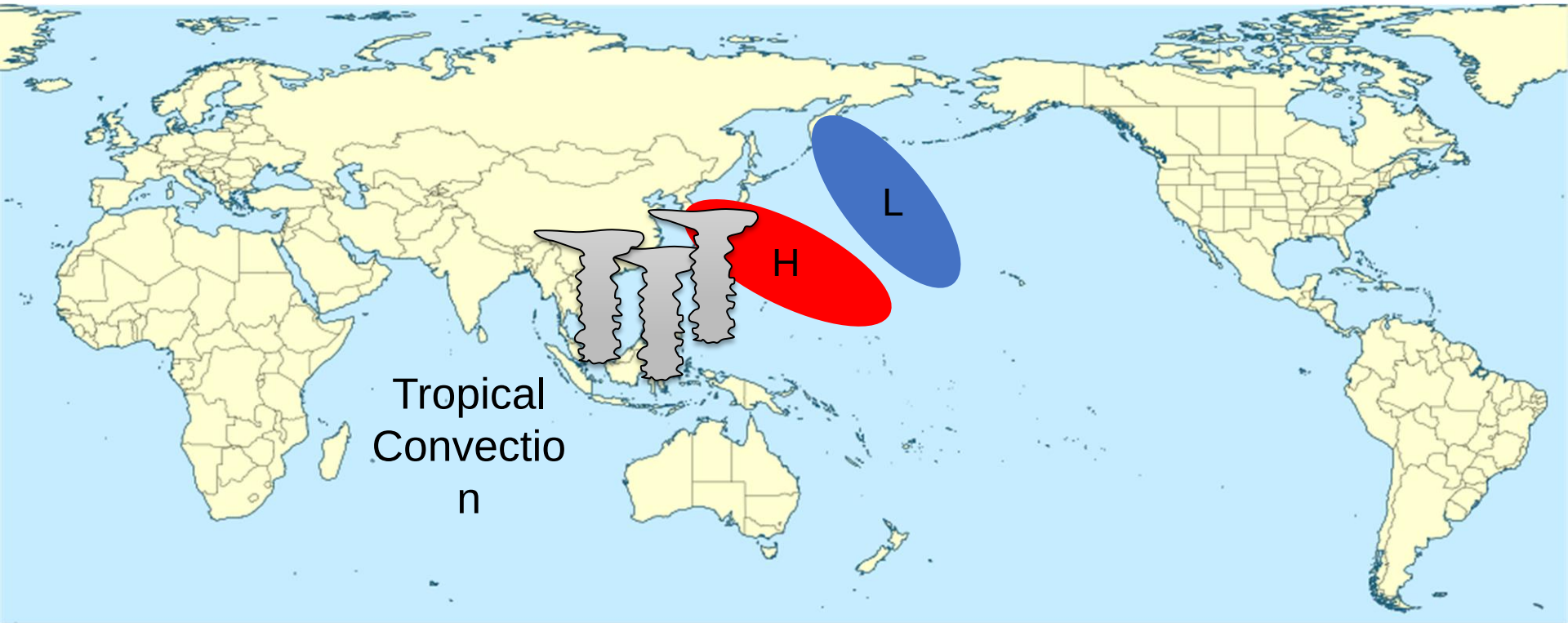
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Like creating ripples in a pond, tropical convection can produce atmospheric waves that reach North America in ~7 to 10 days

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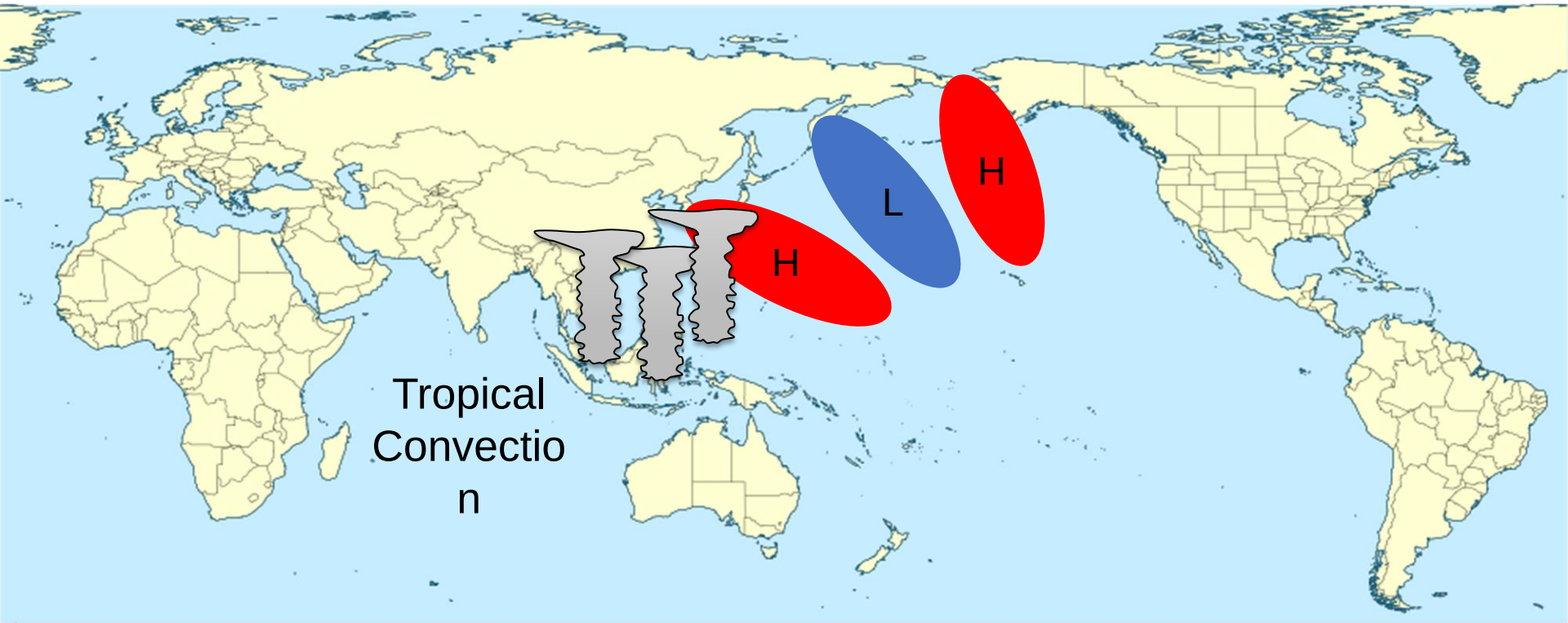
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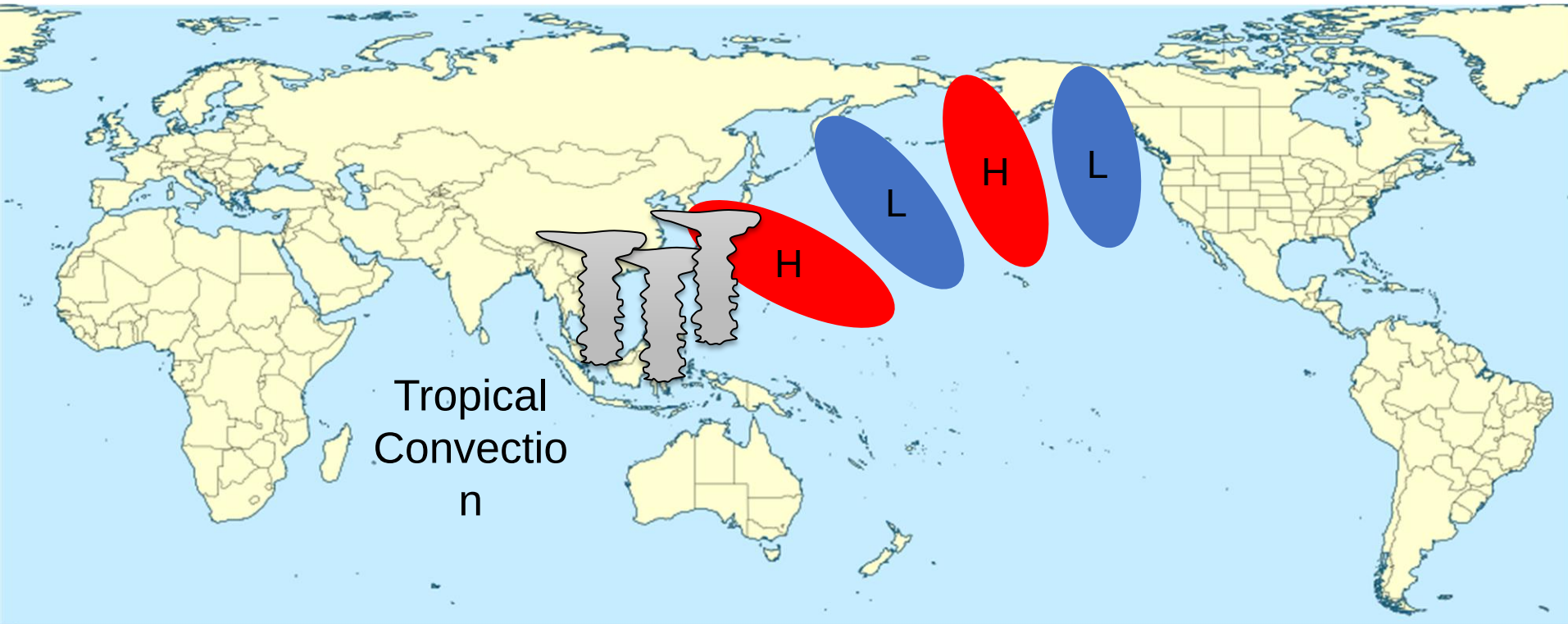
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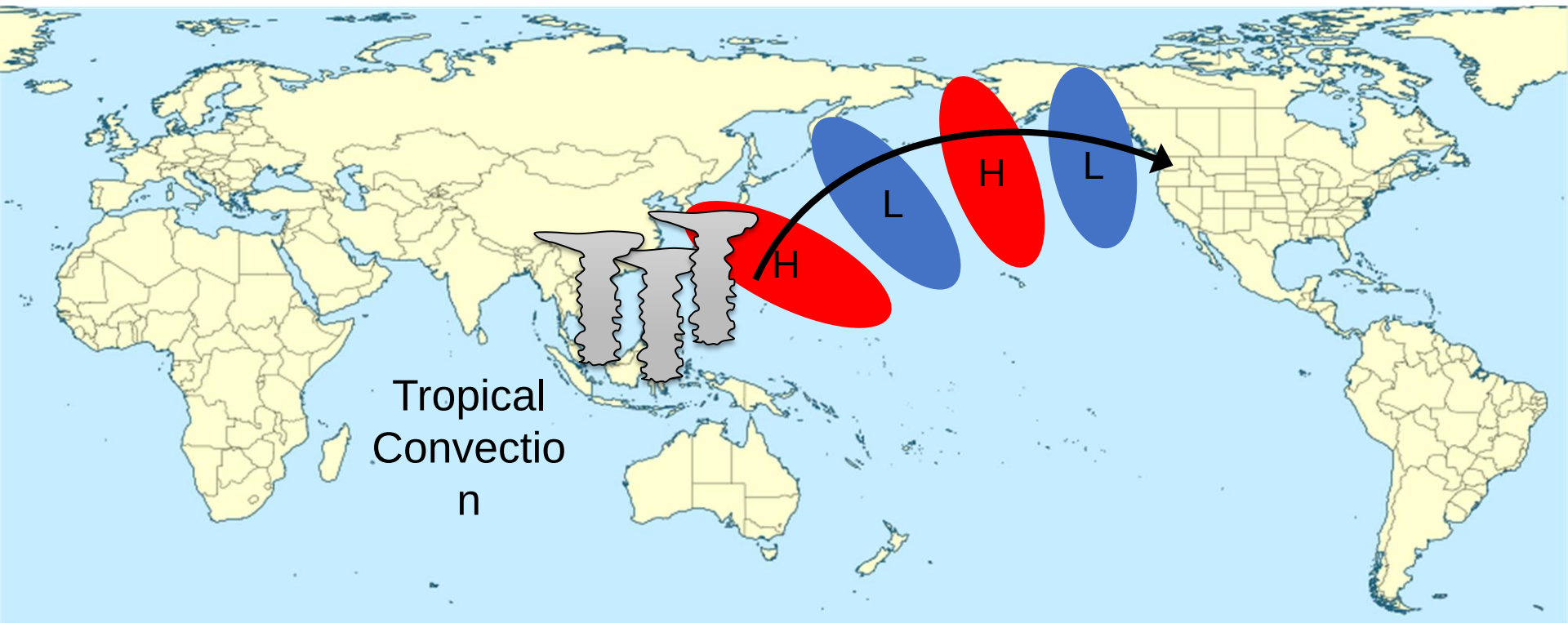
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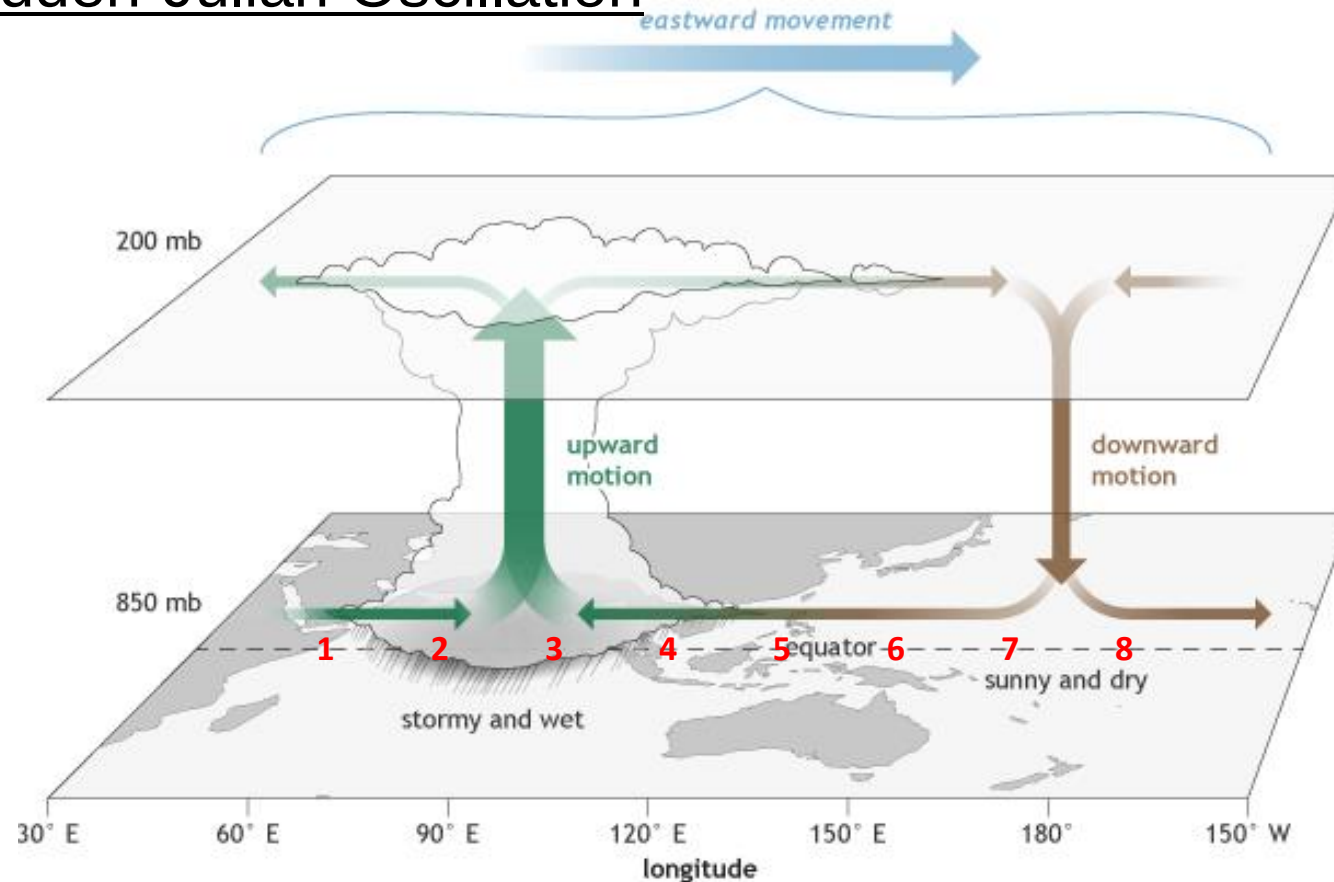


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One form of tropical convection is the...

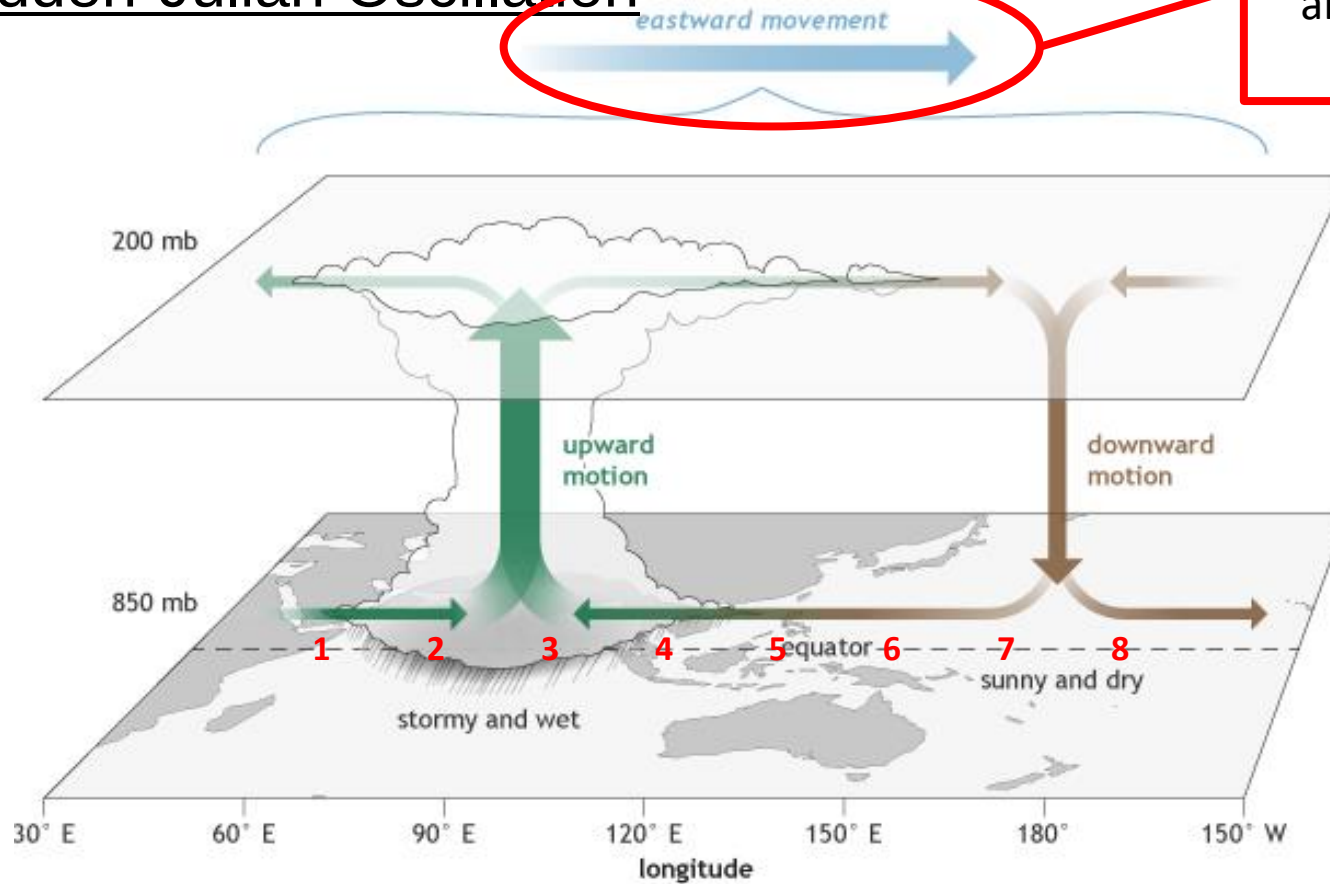
Madden-Julian Oscillation



The Quasi-biennial Oscillation – modulates the MJO?

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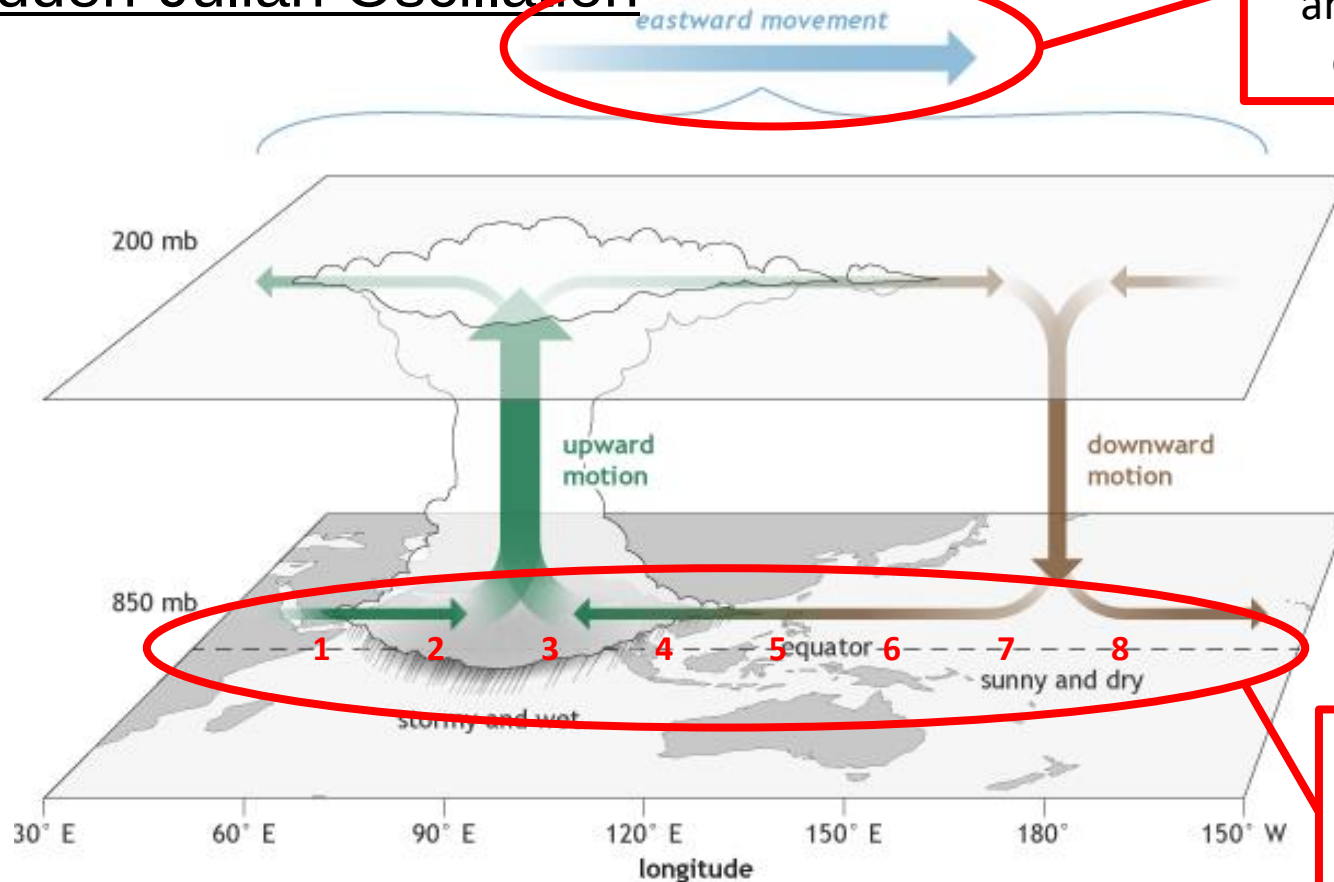


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Madden-Julian Oscillation



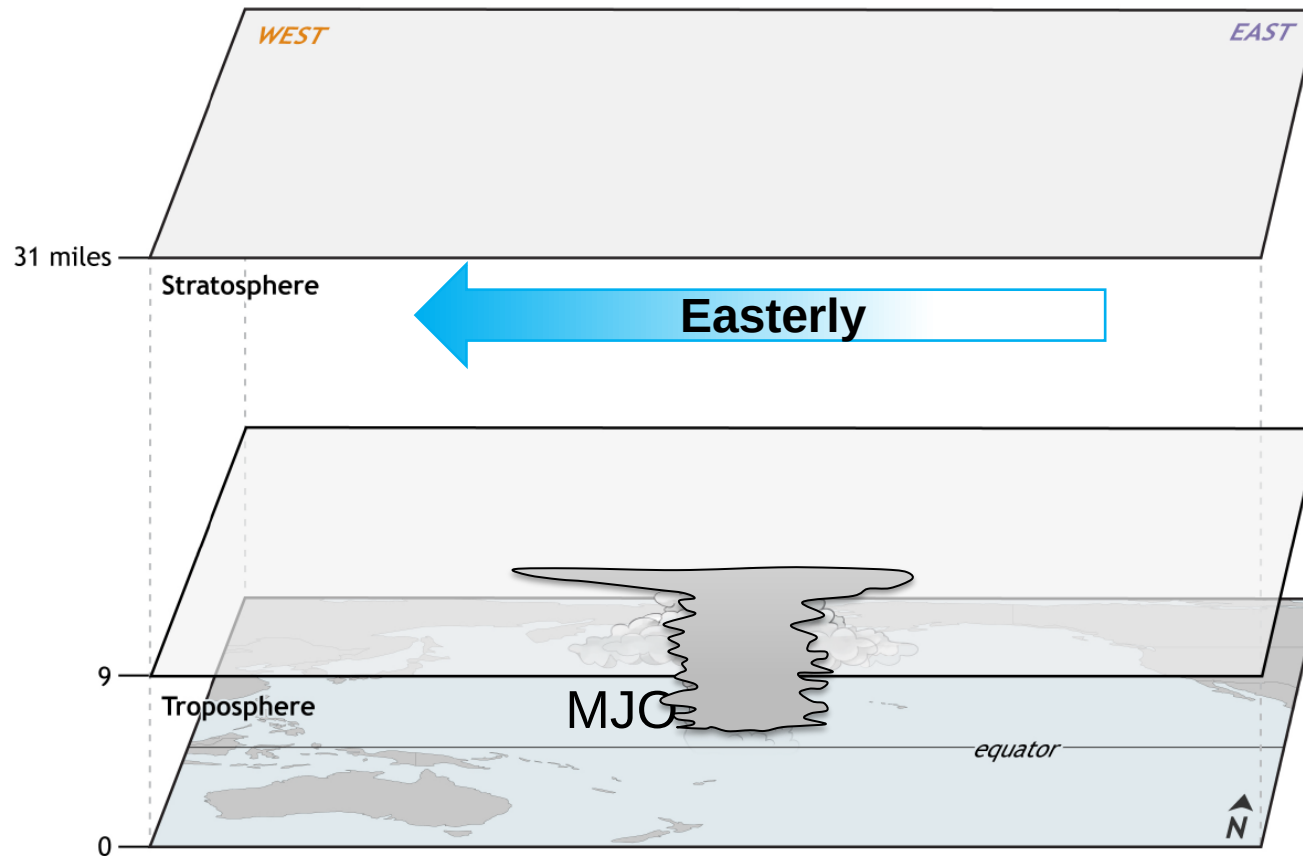
~30 to 90 days
to travel
around the
equator

Location of
MJO along the
equator is
divided into 8
phases

The Quasi-biennial Oscillation – modulates the MJO?



The Quasi-biennial Oscillation

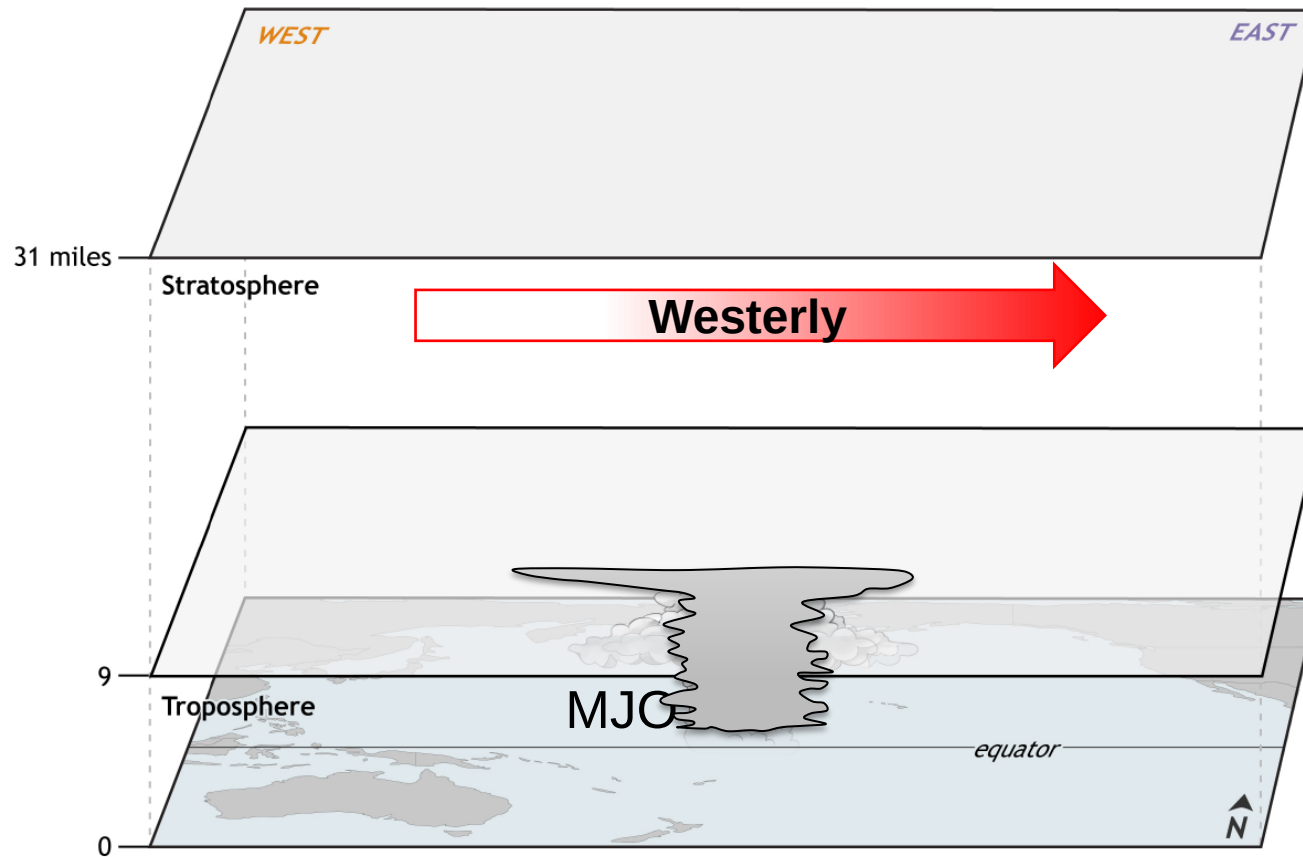


The QBO is characterized by stratospheric zonal winds that cycle between easterly and westerly phases every 2 to 3 years

The Quasi-biennial Oscillation – modulates the MJO?



The Quasi-biennial Oscillation

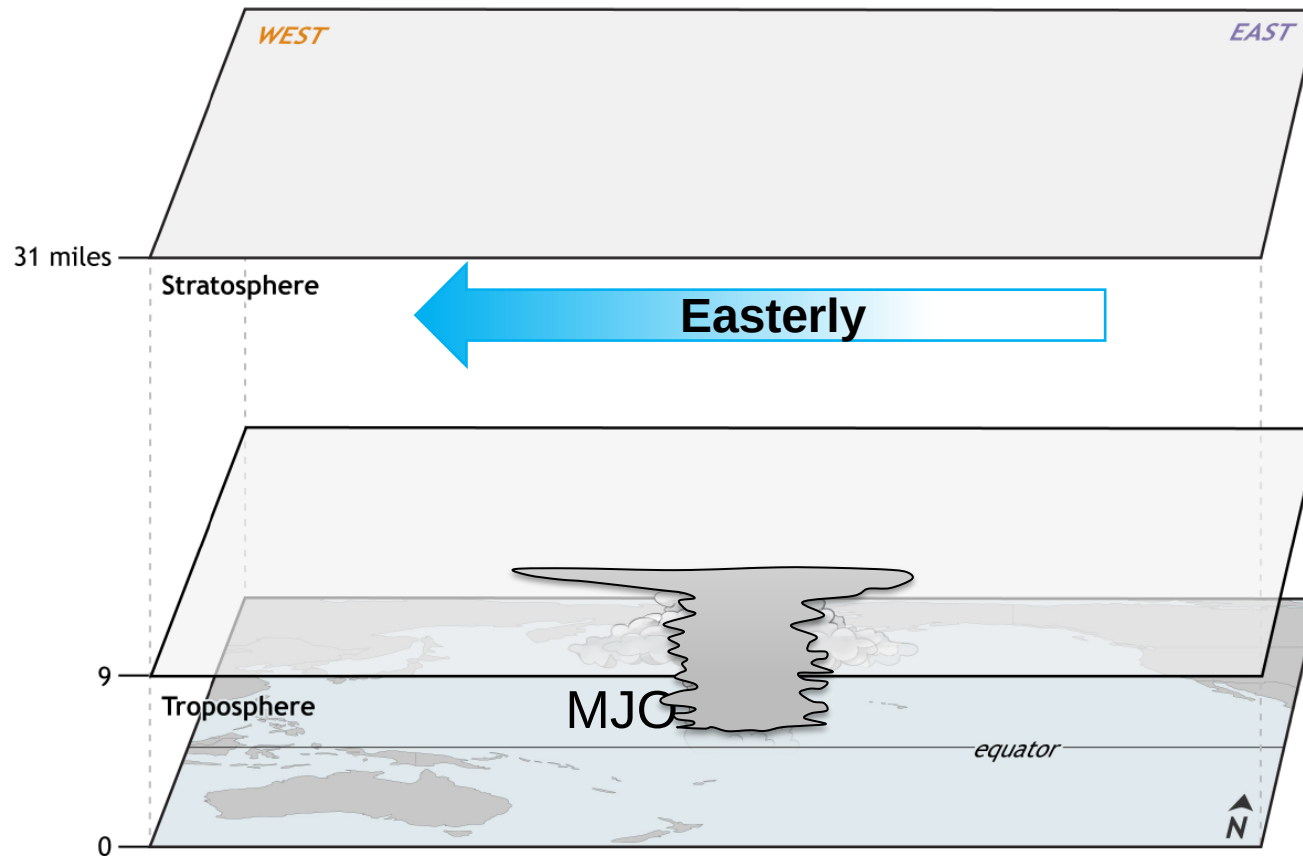


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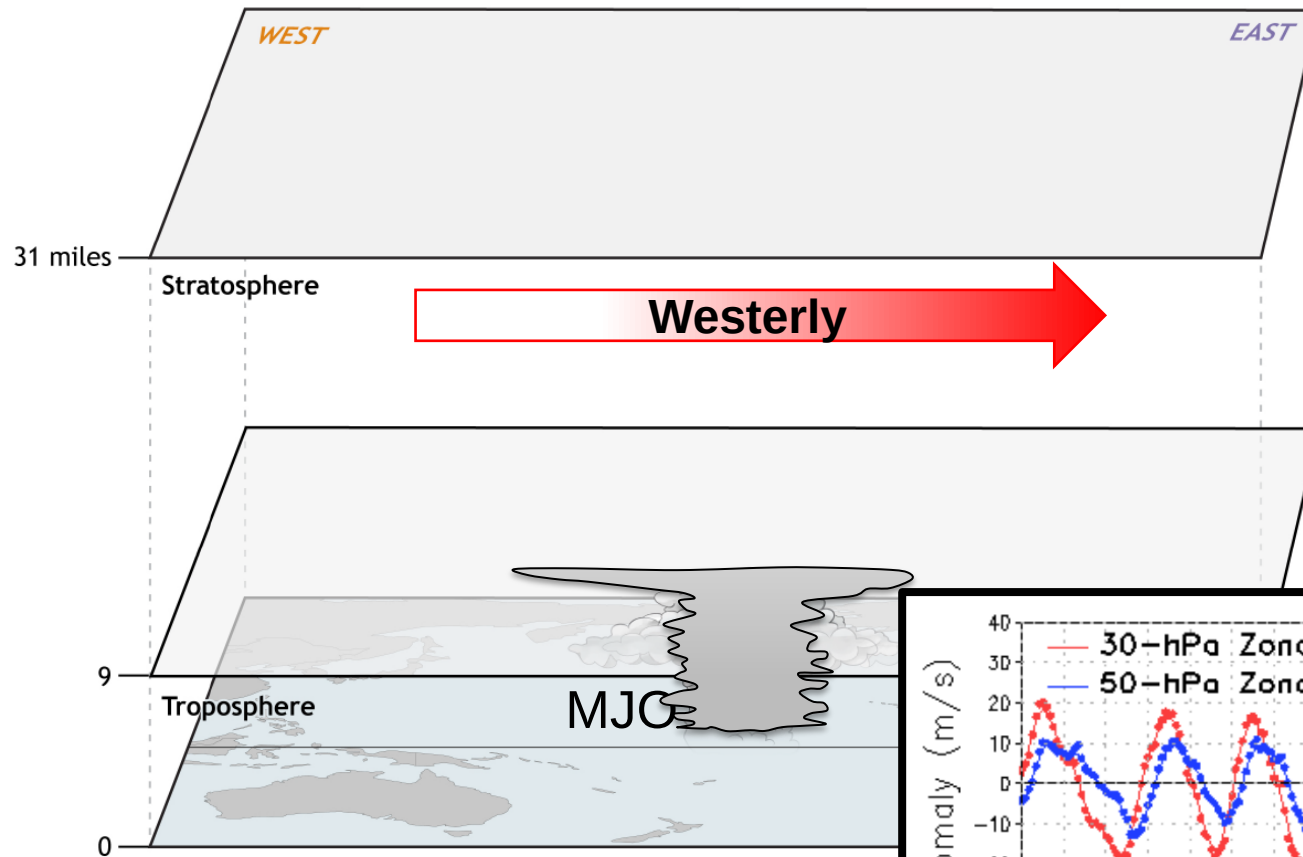


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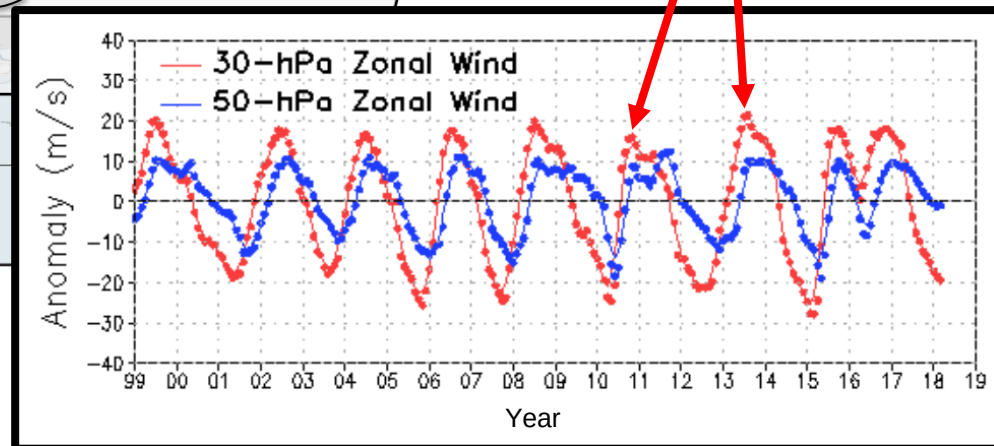
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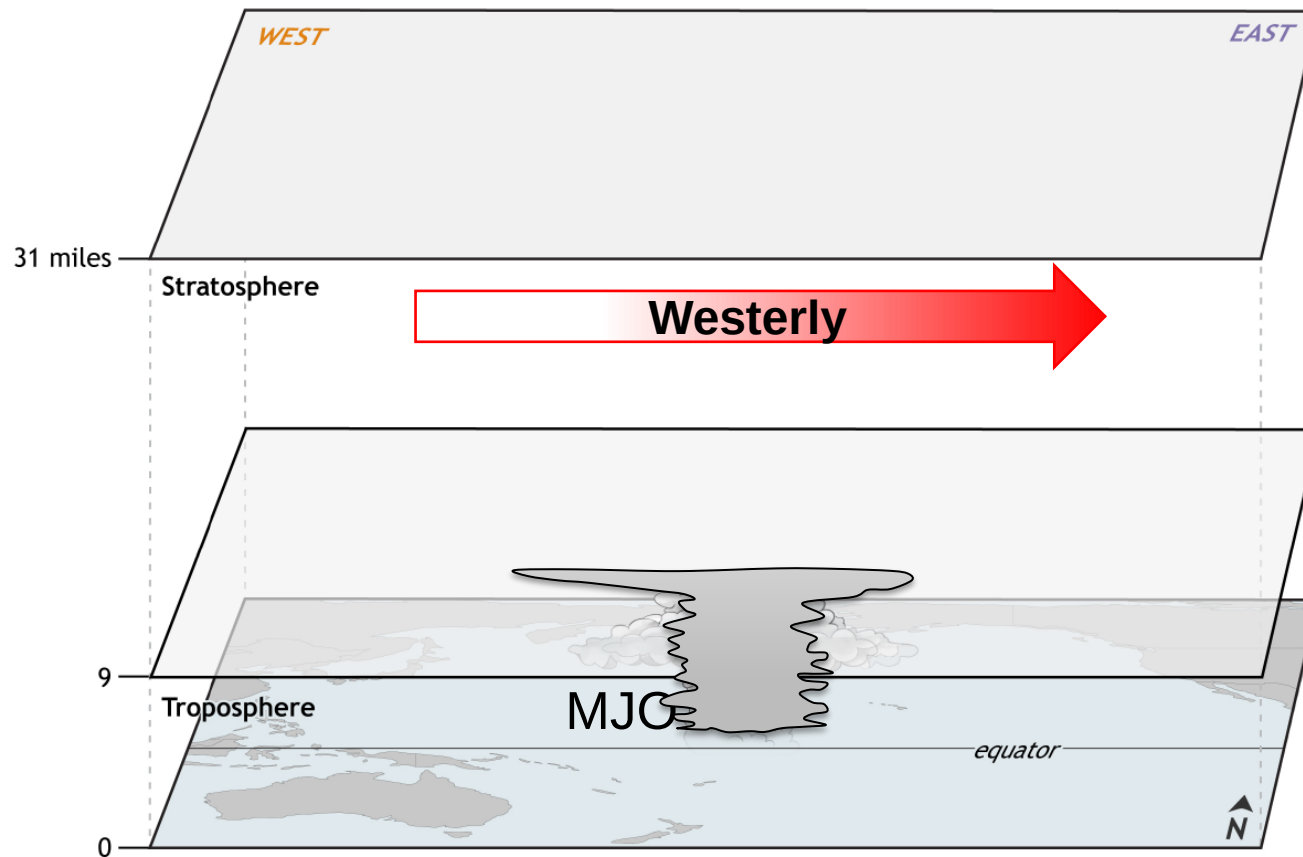
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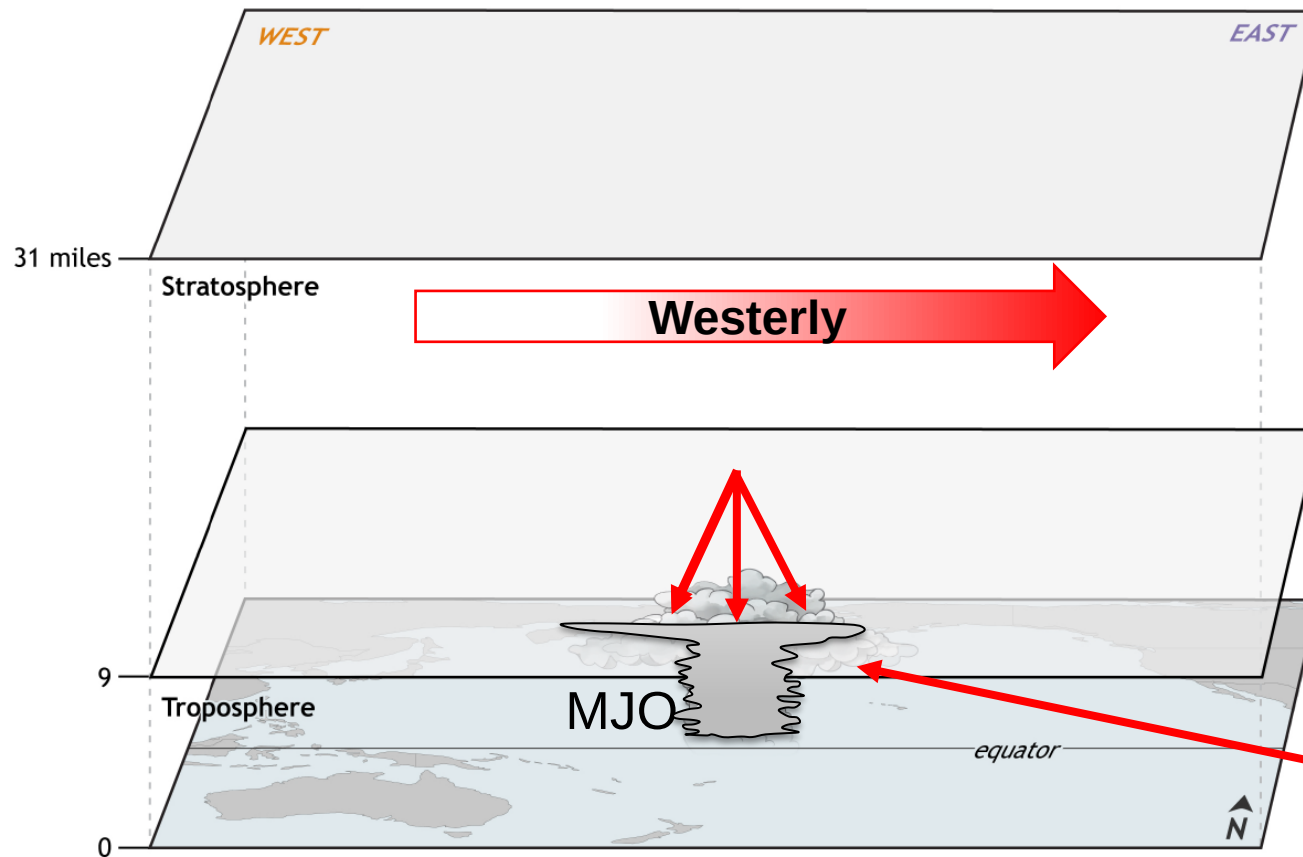


Recent work has shown that the QBO has an impact on the MJO and its teleconnections, but the mechanism is still a subject of ongoing research (Yoo and Son 2016; Son et al. 2017; Zhang and Zhang 2018)

The Quasi-biennial Oscillation – modulates the MJO?



The Quasi-biennial Oscillation

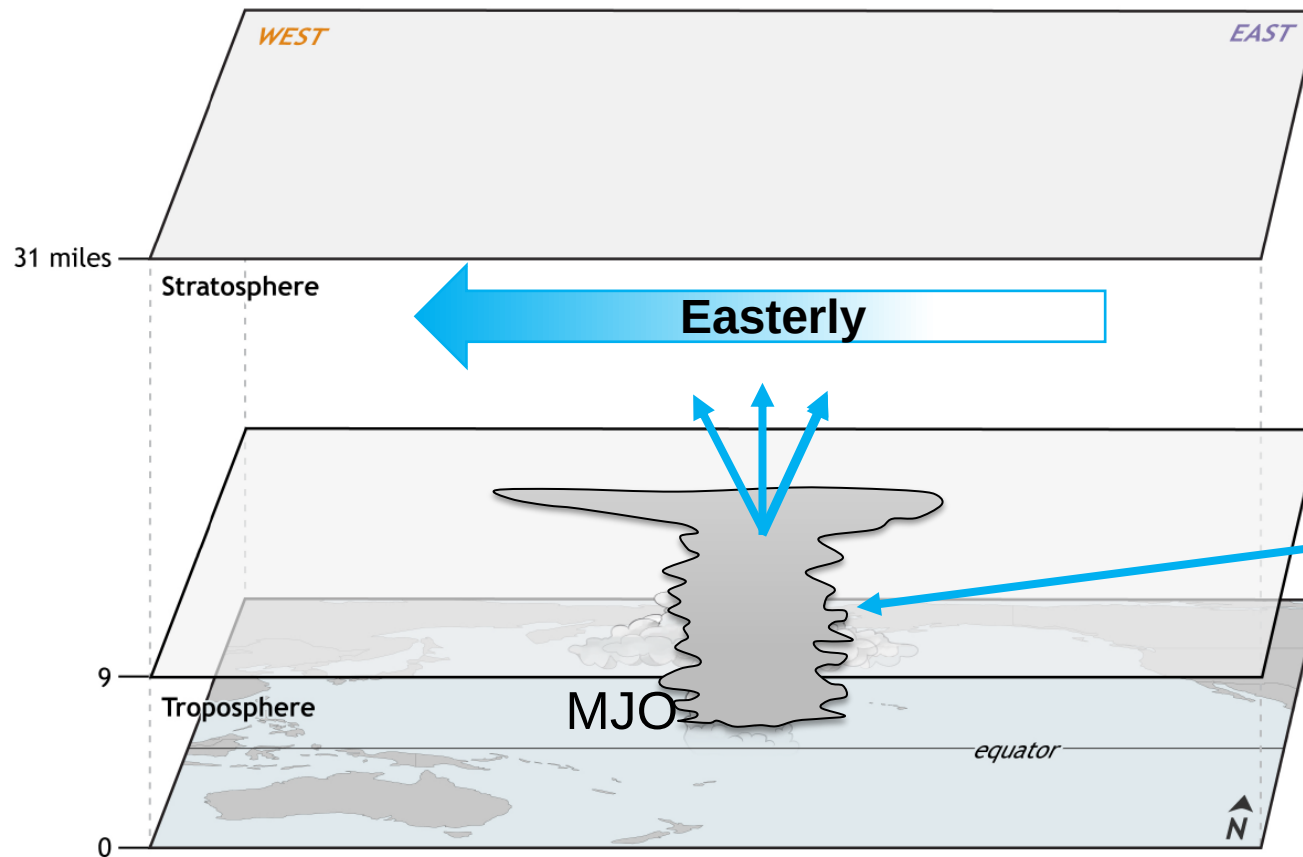


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MJO is *suppressed* during the westerly phase of the QBO

The Quasi-biennial Oscillation – modulates the MJO?

The Quasi-biennial Oscillation

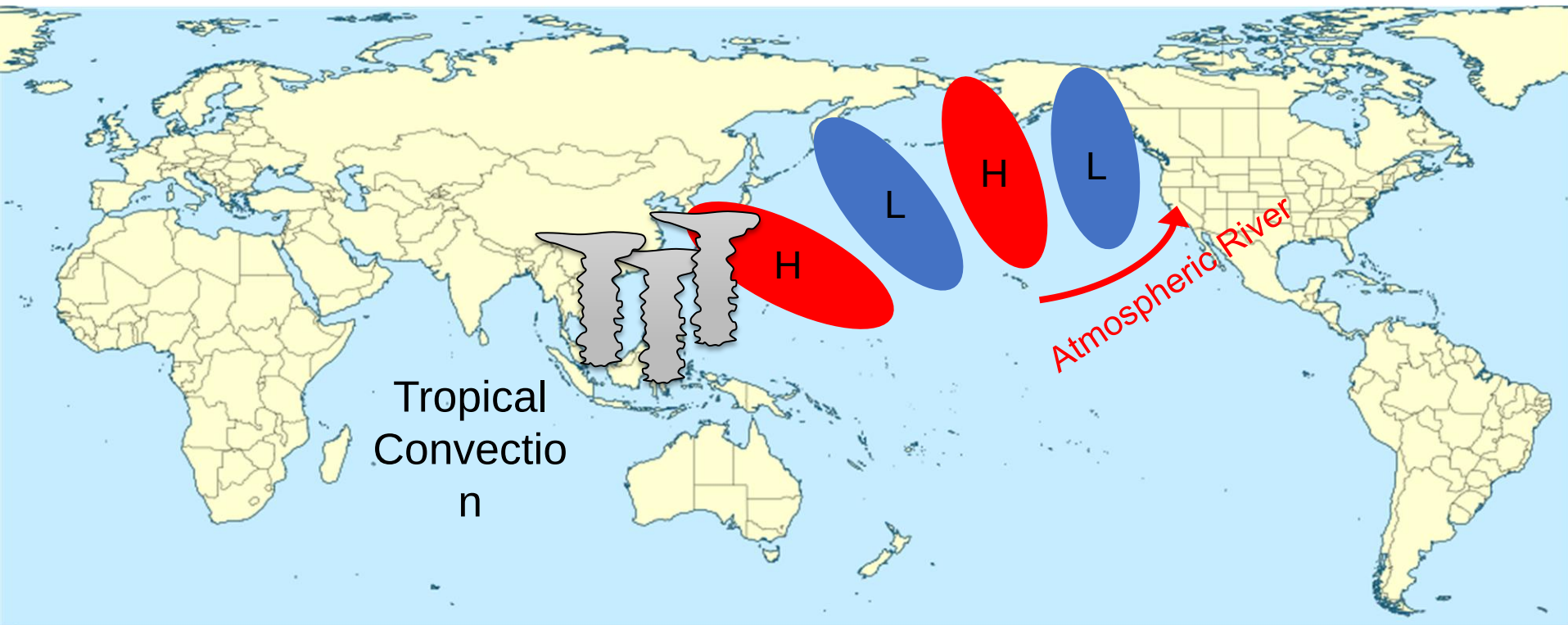


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MJO is *enhanced* during the easterly phase of the QBO

The Madden-Julian Oscillation – a source of predictability at S2S lead times

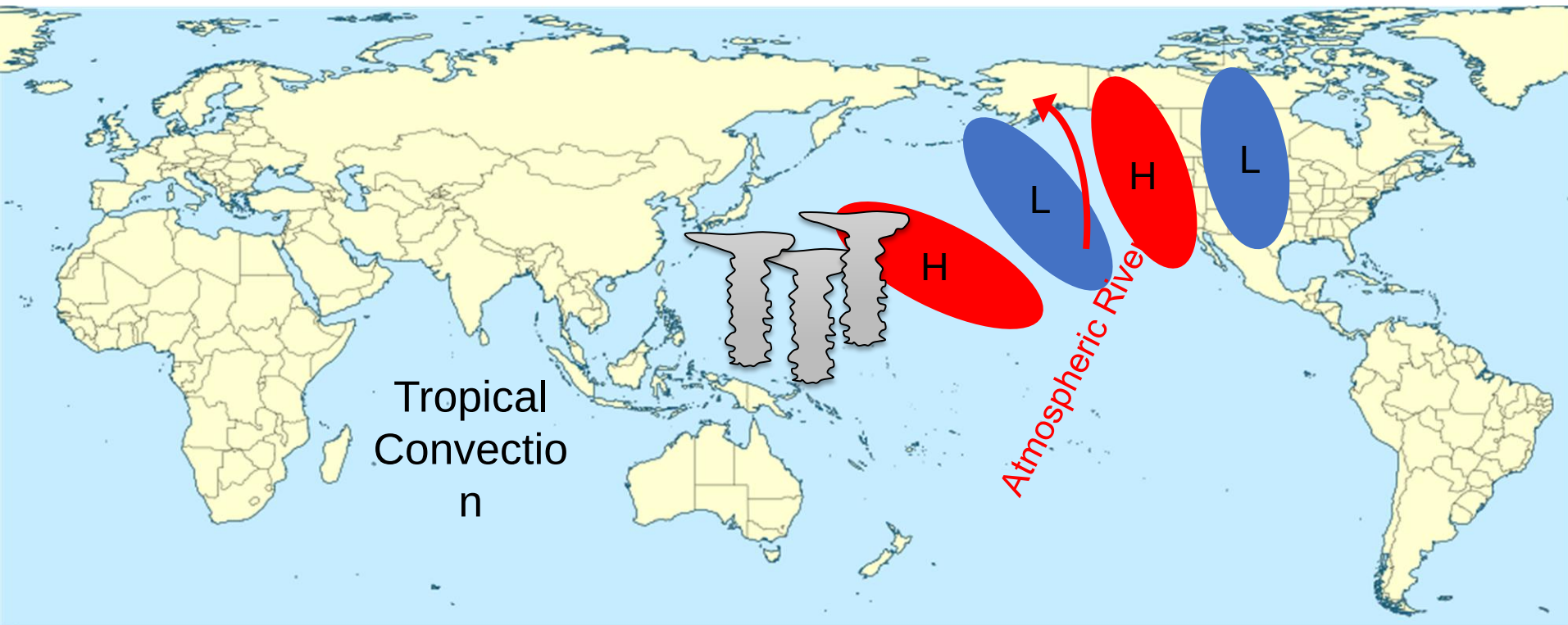
So, why do the MJO and QBO matter?



Because the pressure and wind patterns they produce over the North Pacific impact if and where atmospheric rivers make landfall along the U.S. West Coast.

The Madden-Julian Oscillation – a source of predictability at S2S lead times

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The MJO and the QBO are potential sources of predictability.



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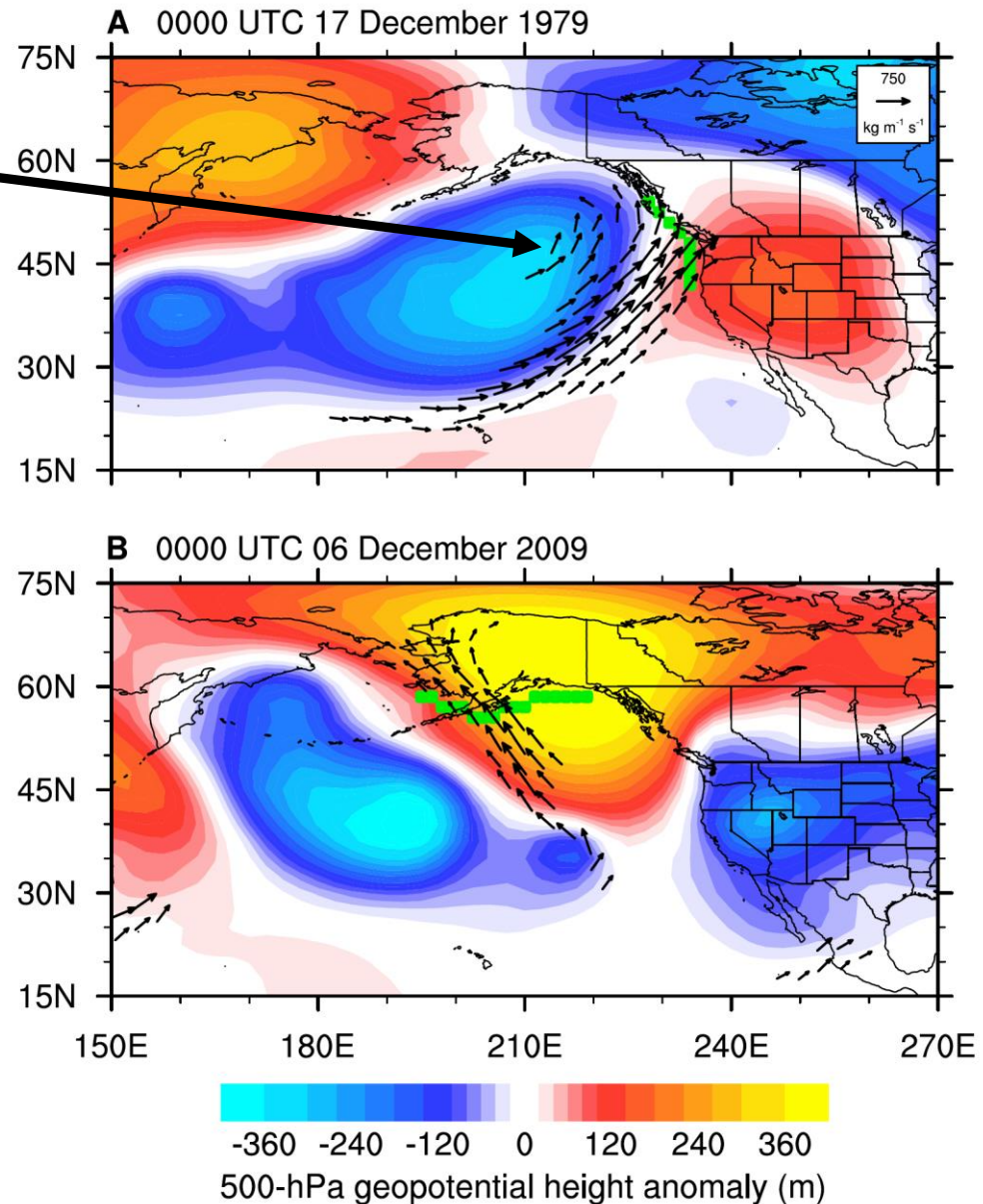
The MJO and the QBO are potential sources of predictability.

Does the *current state* of the MJO and QBO offer information about *future* atmospheric river activity along the U.S. West Coast?

If yes, can we use the *current state* of the MJO and QBO to skillfully forecast atmospheric river activity at subseasonal-to-seasonal lead times (~3 to 6 weeks in advance)?

Observational Work - AR Landfall Events

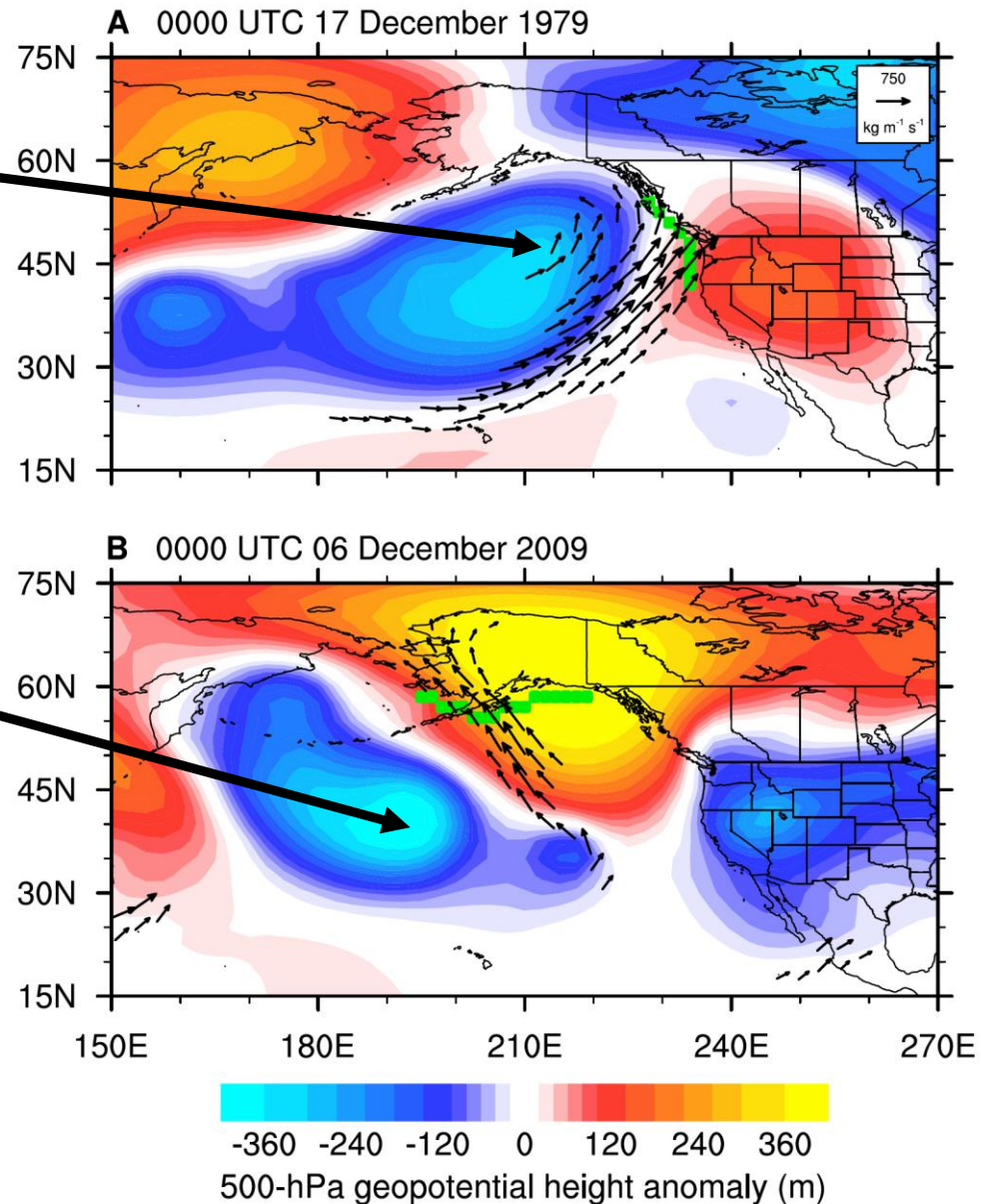
Black arrows outline the location of an atmospheric river



Observational Work - AR Landfall Events

Black arrows outline the location of an atmospheric river

Shaded regions represent 500-mb geopotential height anomalies



Observational Work - AR Landfall Events

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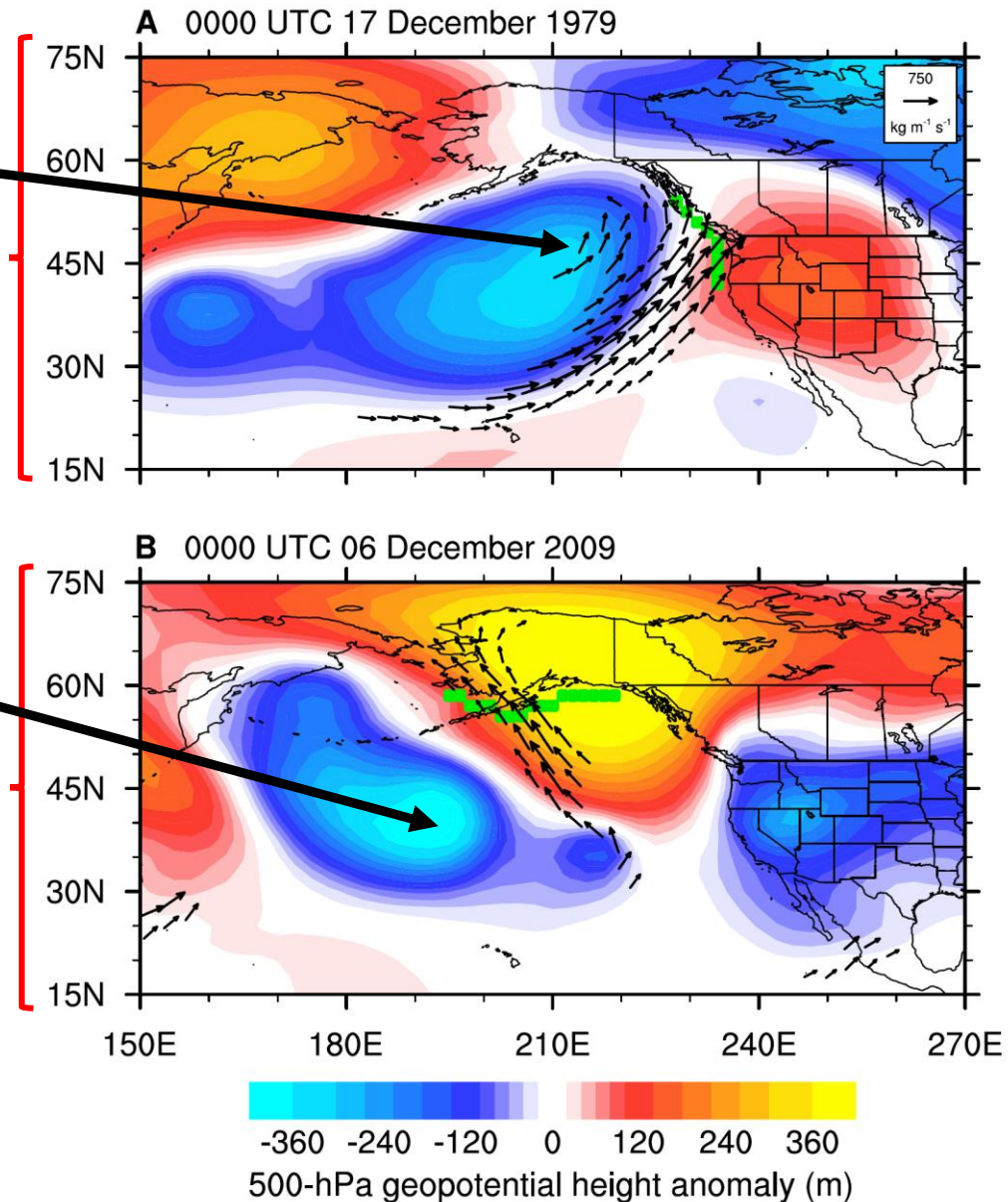
Easterly QBO

Shaded regions represent 500-mb geopotential height anomalies

versus

Both events occurred in the third week following MJO phase 5, but during opposite QBO phases

Westerly QBO

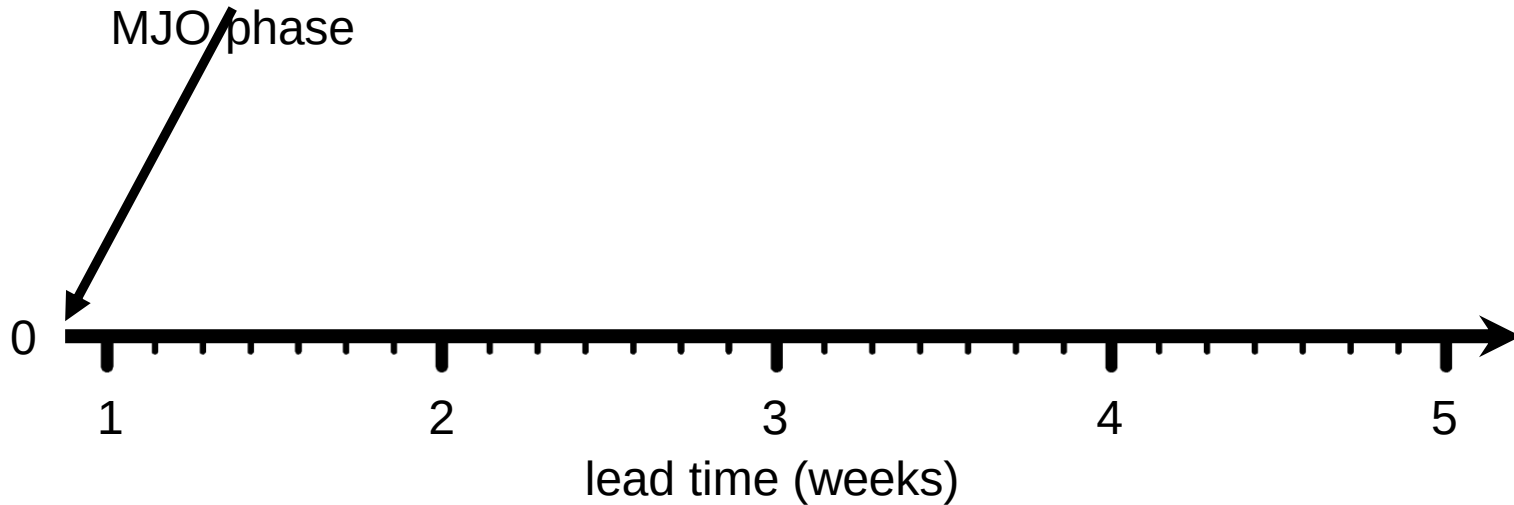


Weekly Composites of Anomalous Atmospheric River Activity

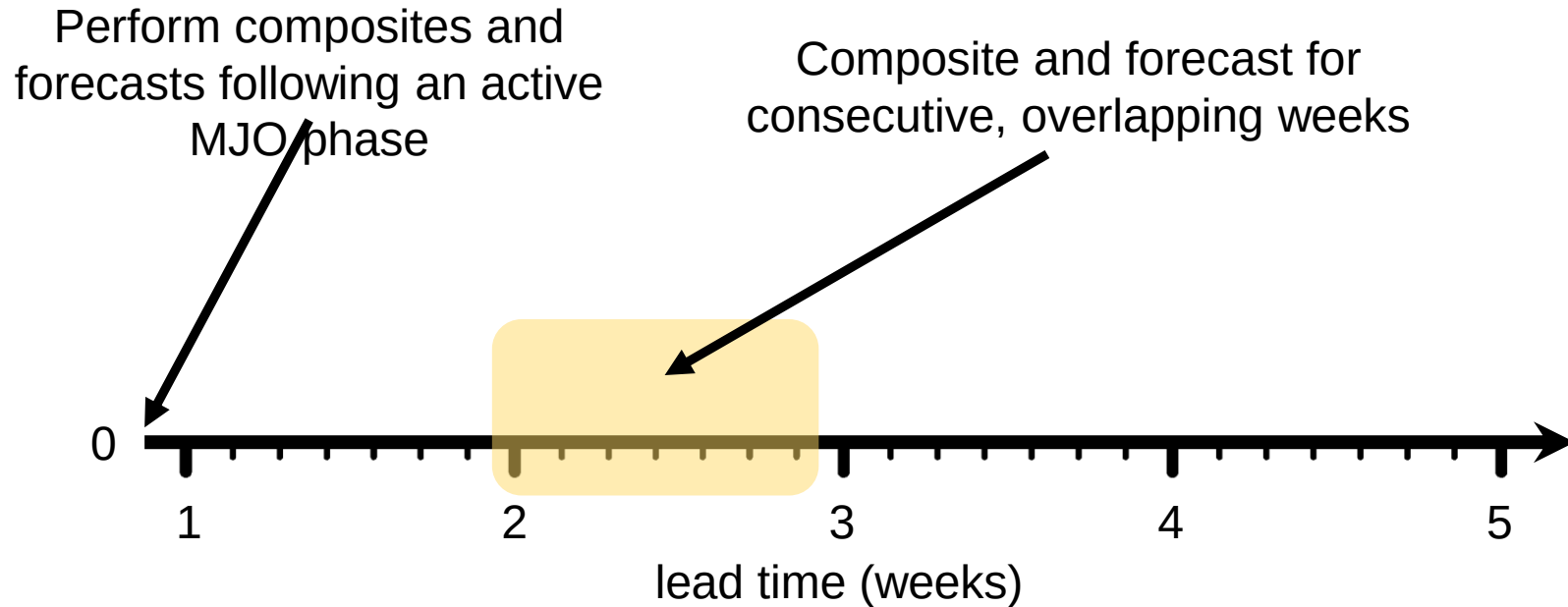


Perform composites and
forecasts following an active

MJO phase



Weekly Composites of Anomalous Atmospheric River Activity



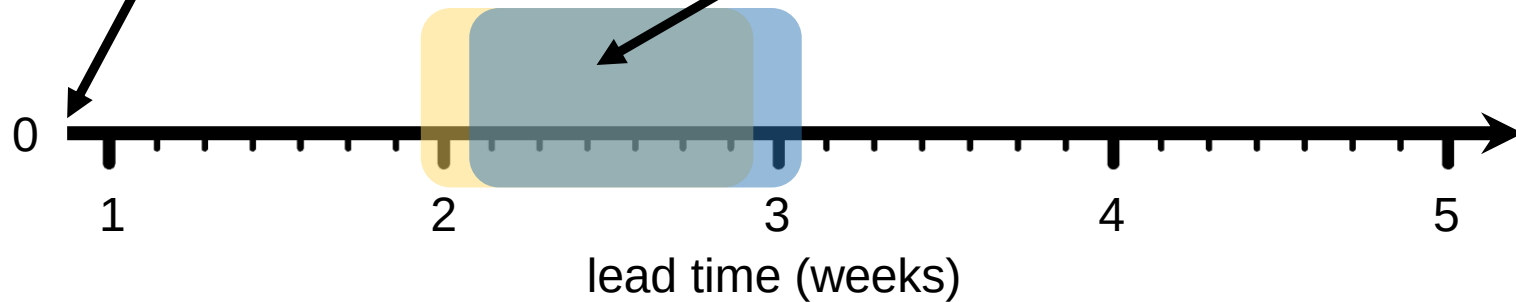
Days 8 to 14

Weekly Composites of Anomalous Atmospheric River Activity



Perform composites and forecasts following an active MJO phase

Composite and forecast for consecutive, overlapping weeks



Days 8 to 14

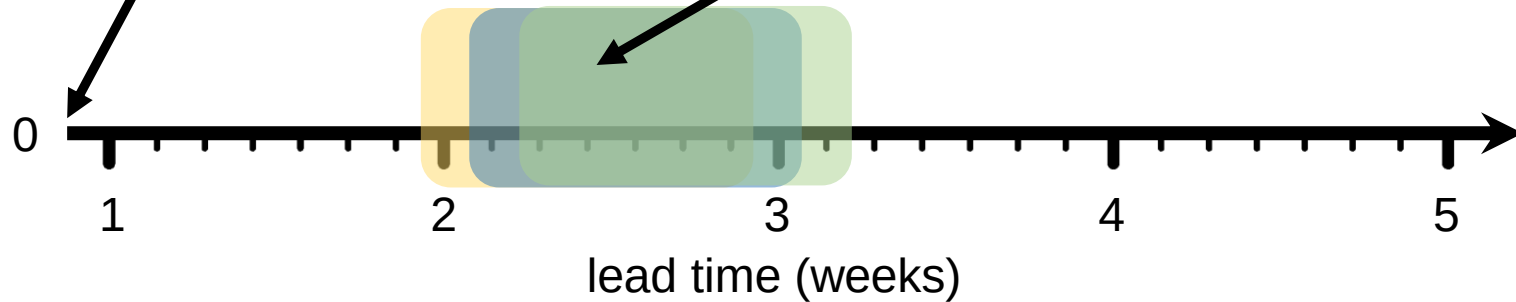
Days 9 to 15

Weekly Composites of Anomalous Atmospheric River Activity



Perform composites and forecasts following an active MJO phase

Composite and forecast for consecutive, overlapping weeks



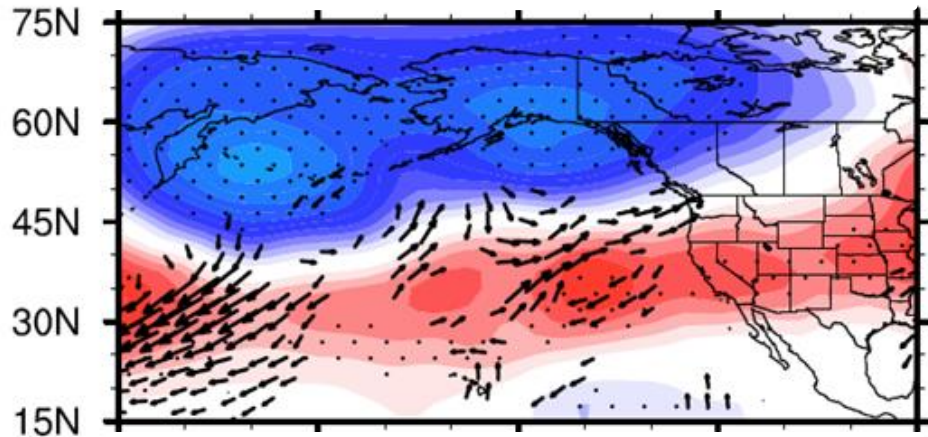
Days 8 to 14

Days 9 to 15

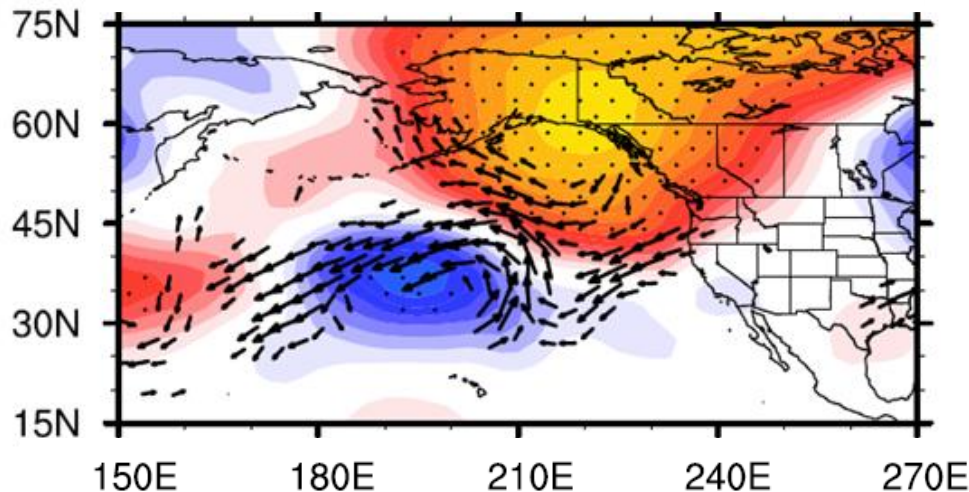
Days 10 to 16

Observational Work - Composites

Easterly QBO Composite



Westerly QBO Composite



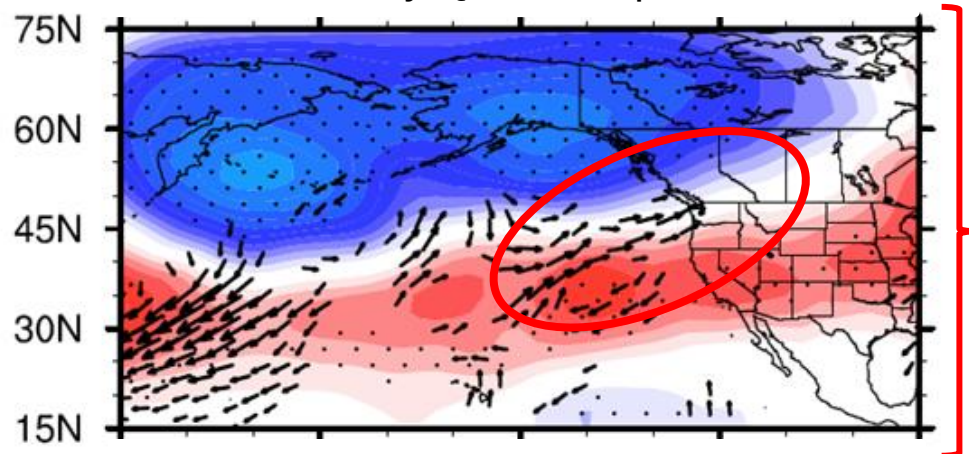
500-hPa geopotential height anomaly (m)

Averaging the third week following each MJO phase 5 event, we obtain vastly different composites when the phase of the QBO is considered.

Observational Work - Composites

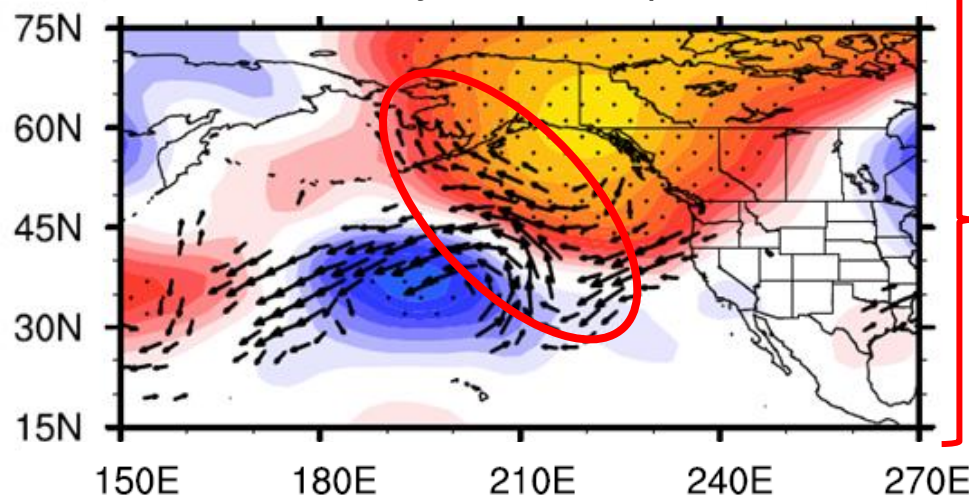


Easterly QBO Composite



Easterly
QBO: favors
OR/WA AR
landfalls

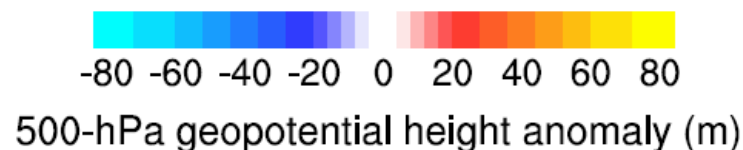
Westerly QBO Composite



Westerly
QBO: favors
AK AR
landfalls

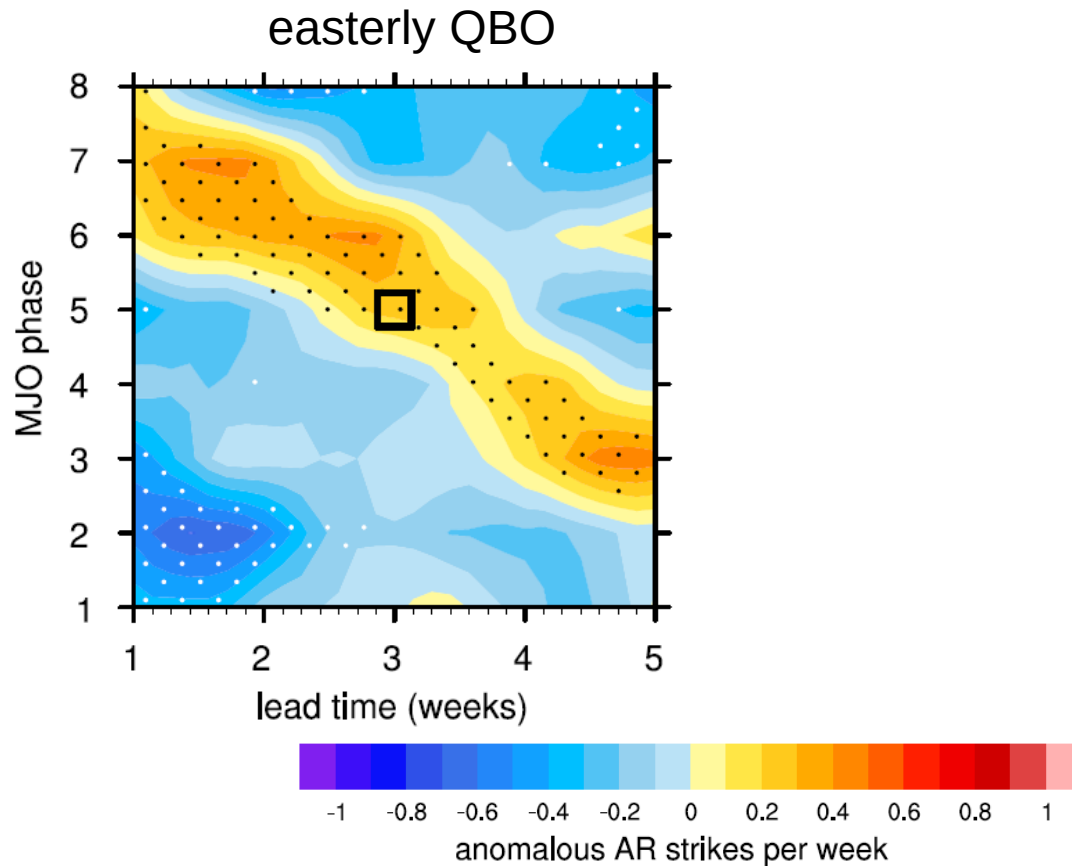
versus

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Observational Work – Pacific Northwest Composites

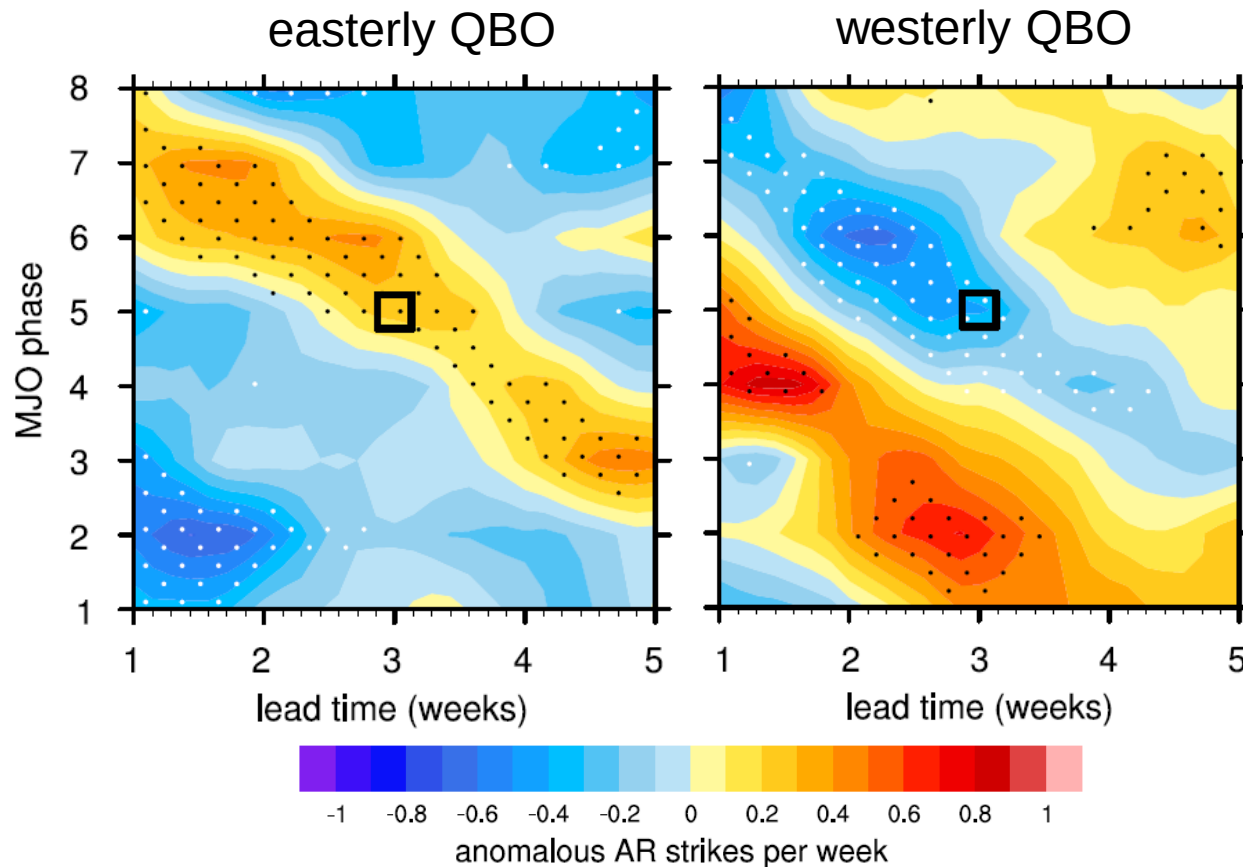
Pacific Northwest



Observational Work – Pacific Northwest Composites



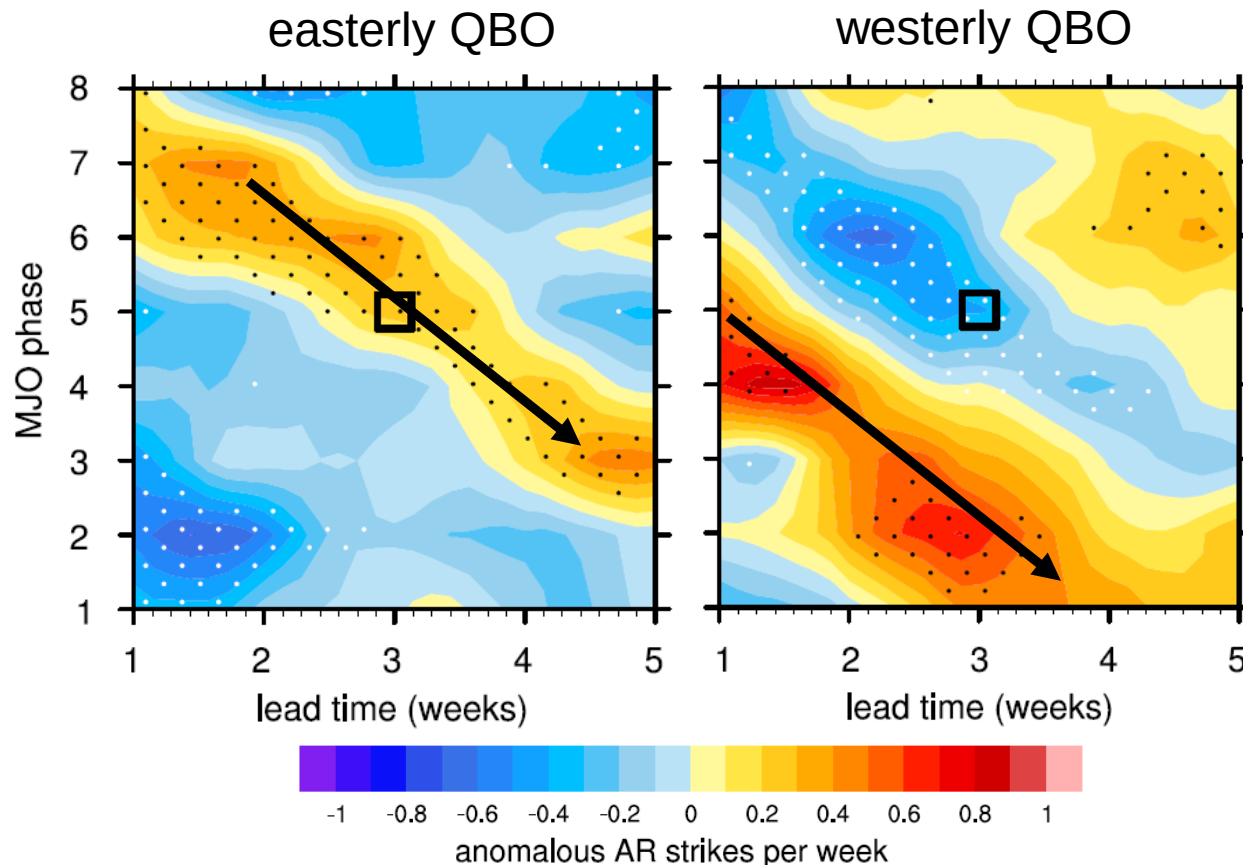
Pacific Northwest



Observational Work – Pacific Northwest Composites



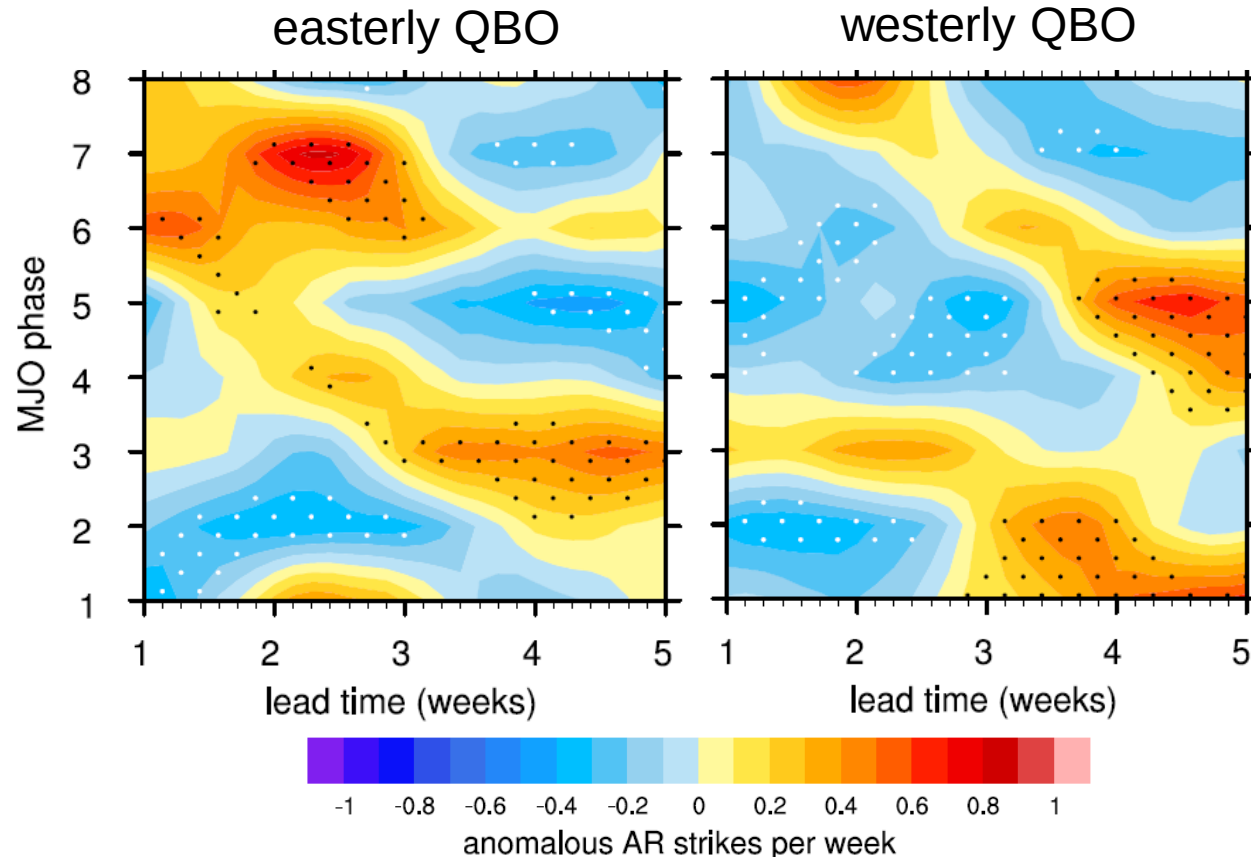
Pacific Northwest



The signals propagate across lead time and MJO phase but vary strongly according to the phase of the QBO.



California

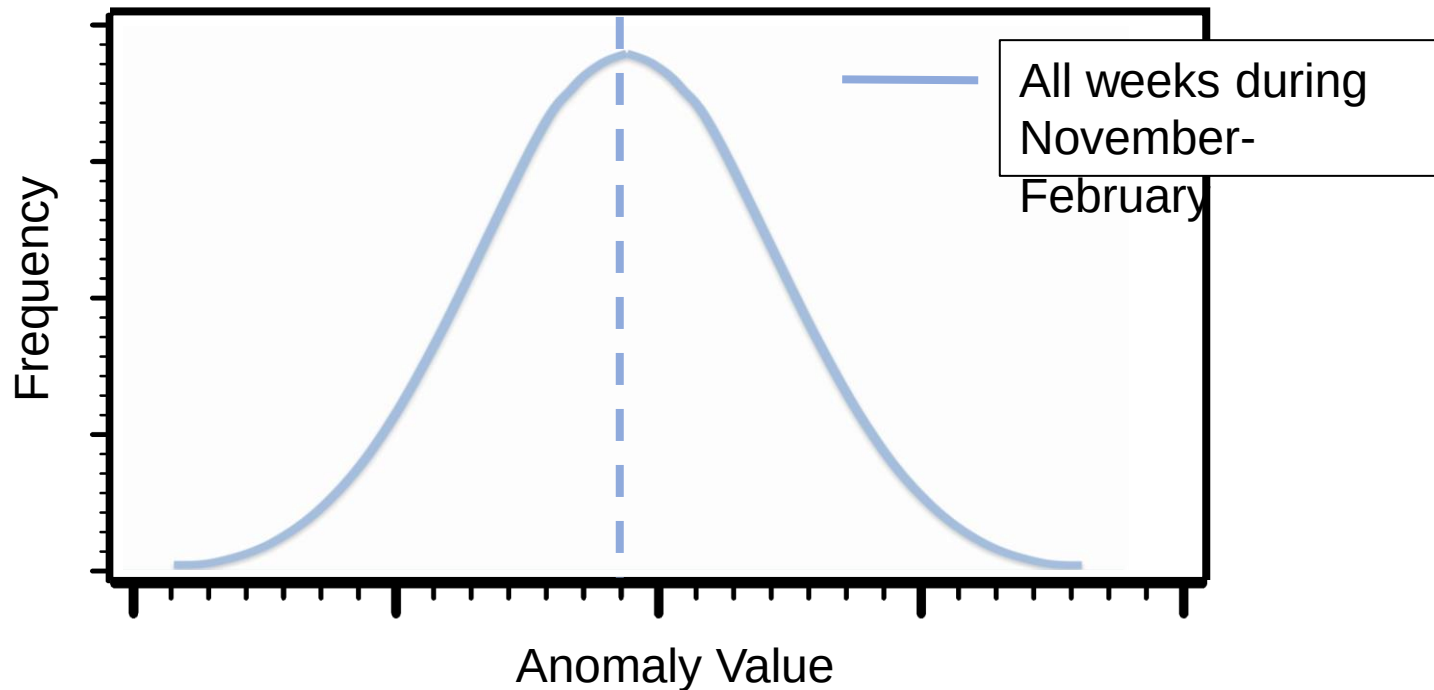


The signals are not as clear in California as in the Pacific Northwest, but modulation by QBO phase is still evident.

Empirical Forecast Work - Description of the Forecast Method



How do we construct an atmospheric river activity forecast using the current state of the MJO and QBO, e.g., in California?

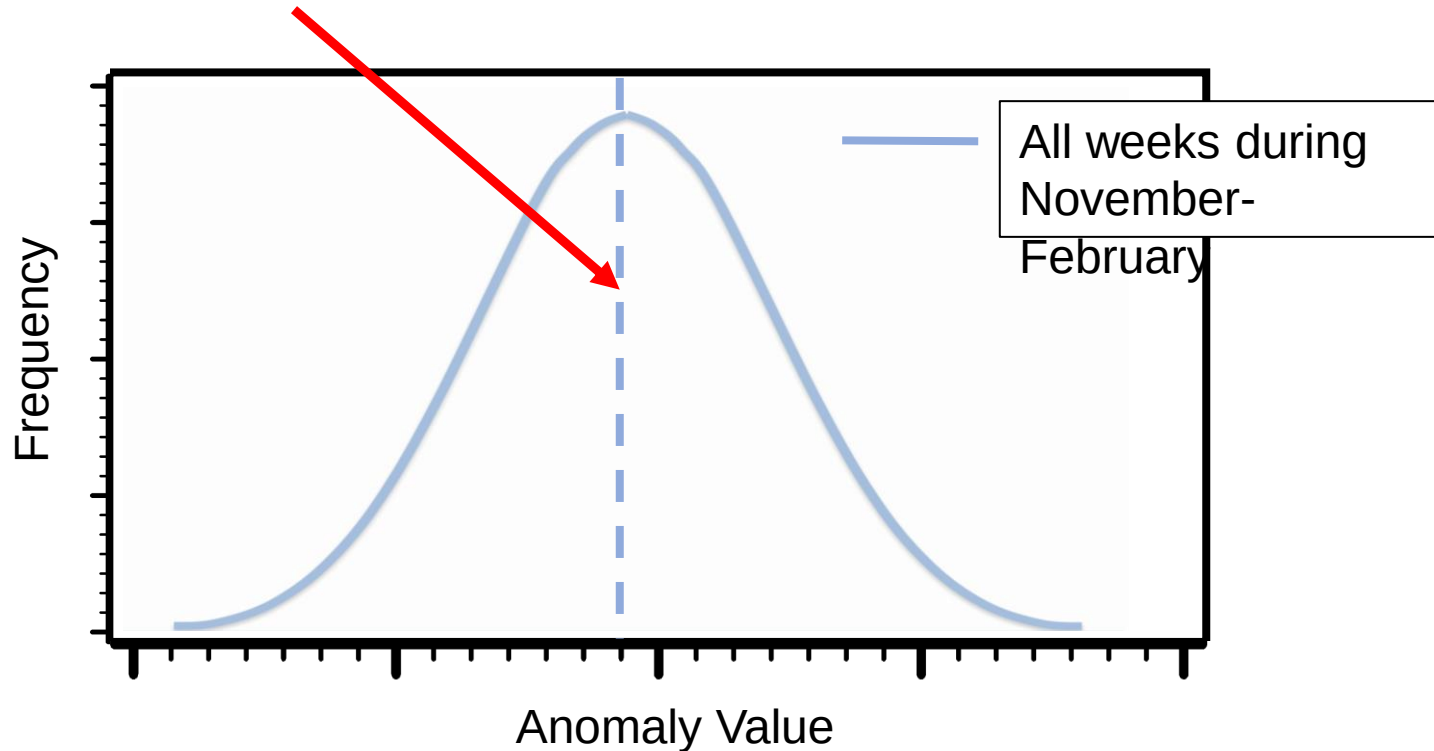


Empirical Forecast Work - Description of the Forecast Method



How do we construct an atmospheric river activity forecast using the current state of the MJO and QBO, e.g., in California?

Normal AR activity during any given week

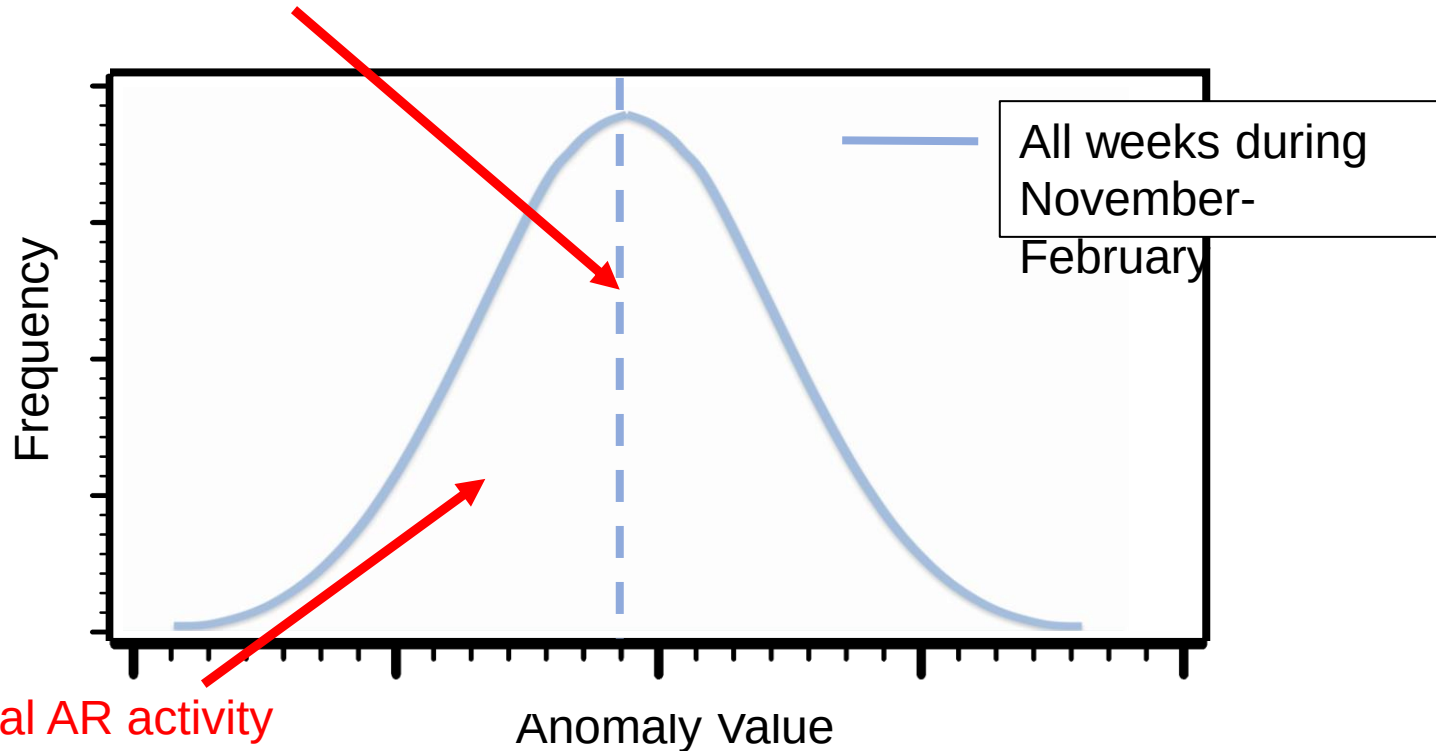


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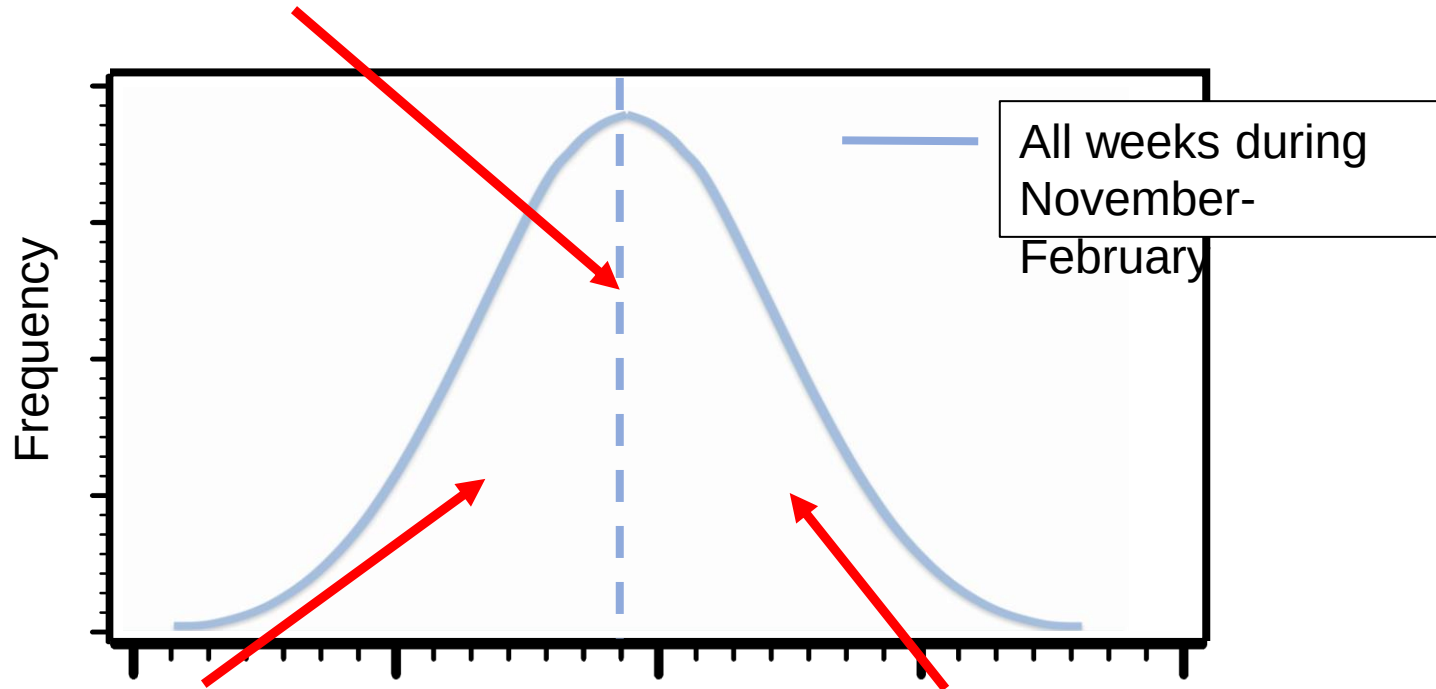
Below normal AR activity
during any given week

Empirical Forecast Work - Description of the Forecast Method



How do we construct an atmospheric river activity forecast using the current state of the MJO and QBO, e.g., in California?

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Below normal AR activity
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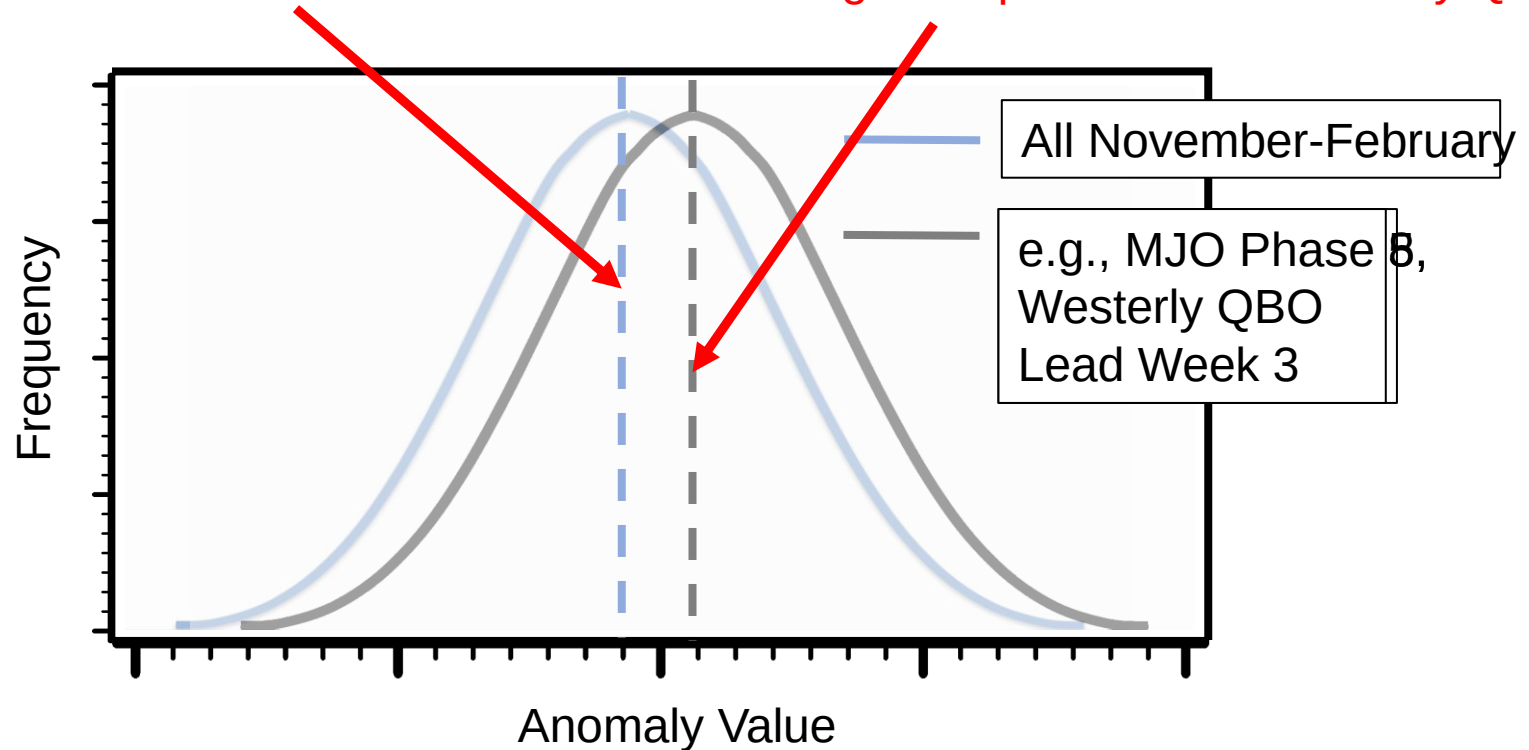
Above normal AR activity
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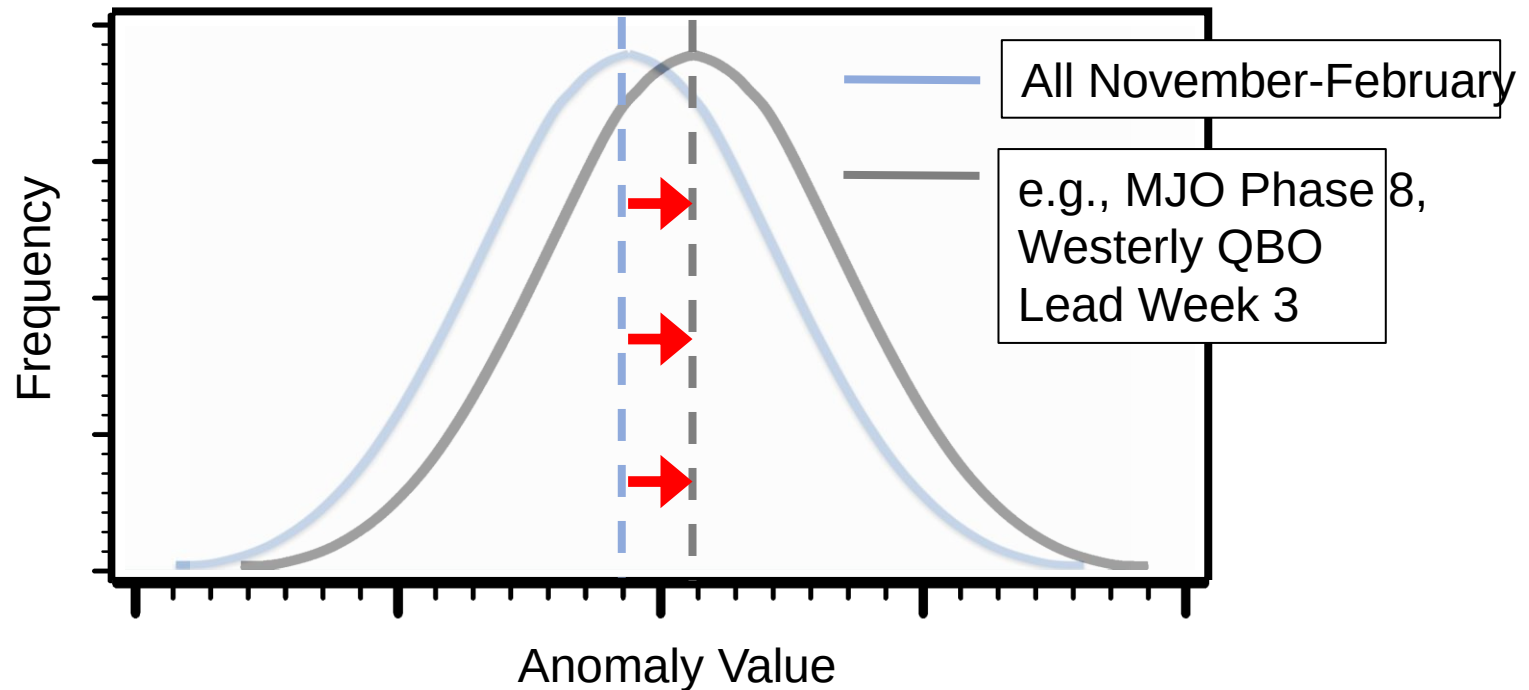
Normal AR activity during the third week following MJO phase 8 in a Westerly QBO



Empirical Forecast Work - Description of the Forecast Method



Normal shifts to the right, so we forecast above normal atmospheric river activity during the third week following MJO phase 8 in a Westerly QBO.

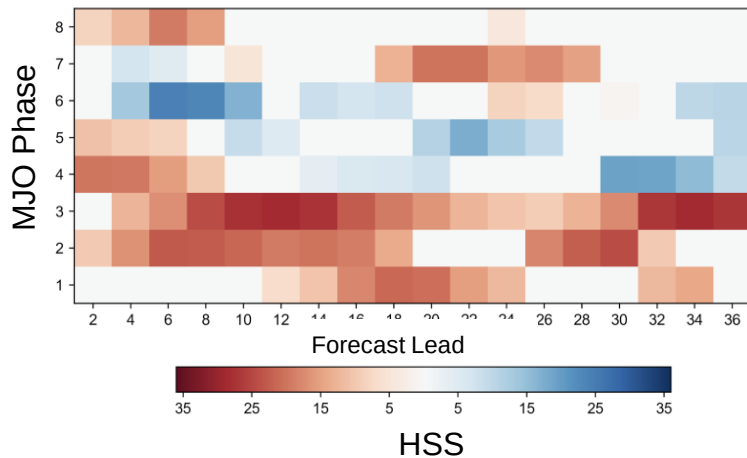


Empirical Forecast Work - Heidke Skill Score for California Forecasts of AR Activity

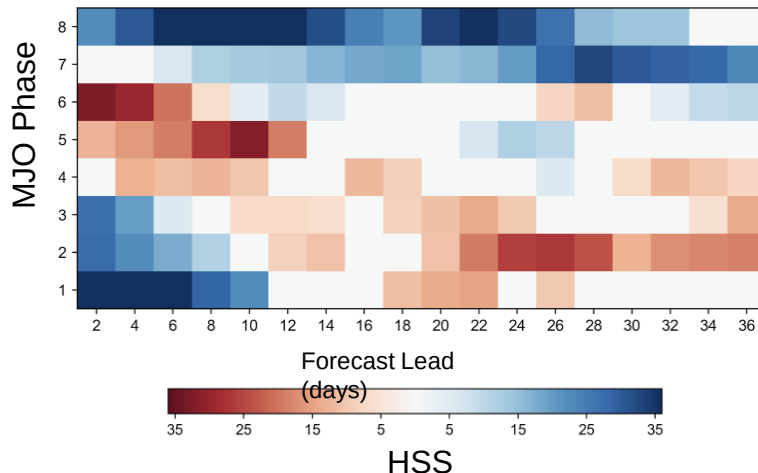


AR Activity Forecasts

HSS for AR activity in CA, during the easterly QBO



HSS for AR activity in CA, during the westerly QBO



We generate forecasts for all combinations of MJO phase, QBO phase, and lead time.

Heidke Skill Score (HSS)

$$HSS = \frac{(H - E)}{(T - E)} \times 100,$$

of correct forecasts

of total forecasts

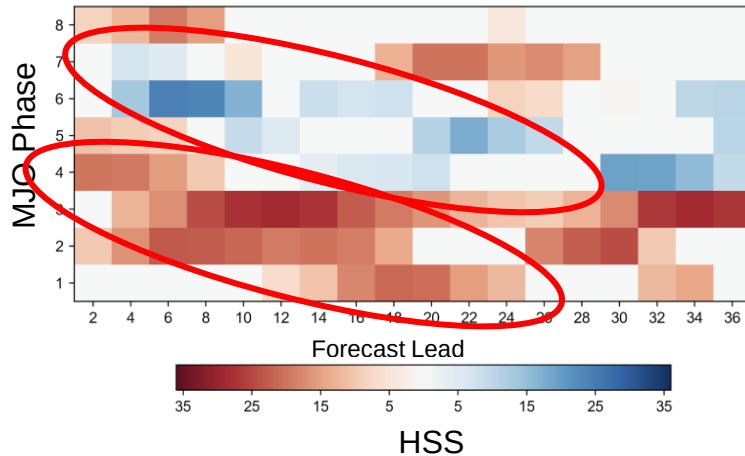
of correct by chance = $T/2$

Empirical Forecast Work - Heidke Skill Score for California Forecasts of AR Activity

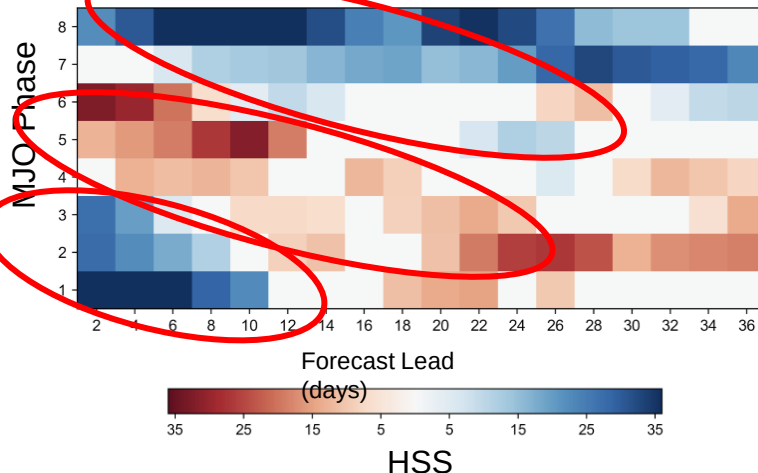


AR Activity Forecasts

HSS for AR activity in CA, during the easterly QBO



HSS for AR activity in CA, during the westerly QBO



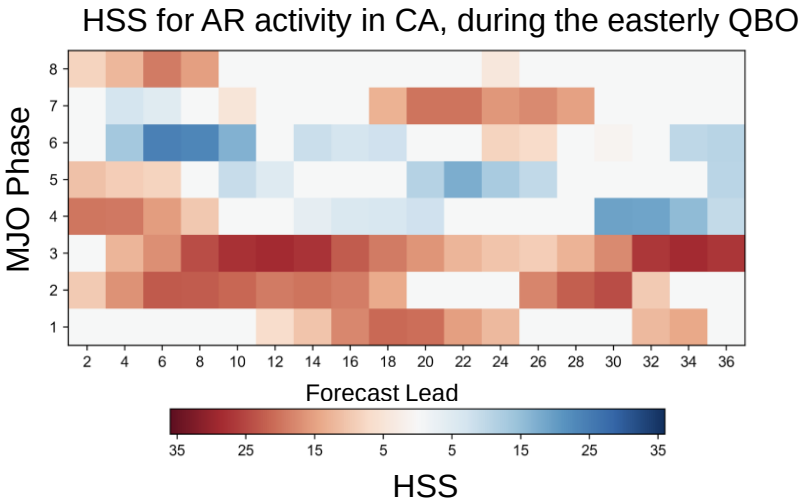
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Heidke Skill Score (HSS)

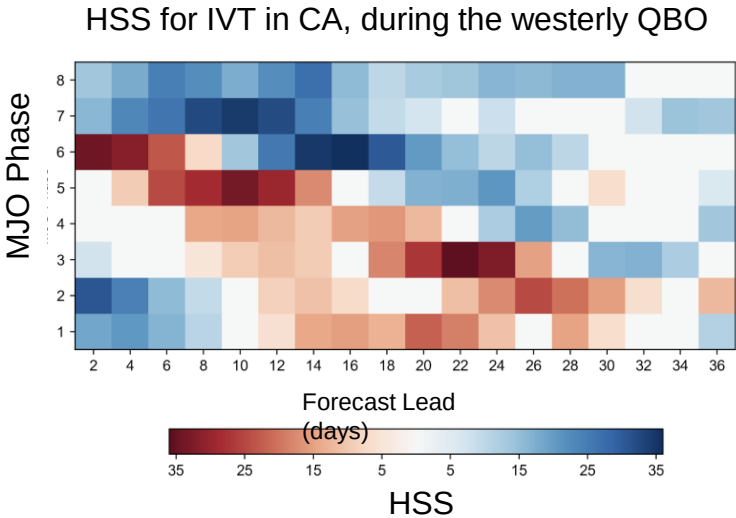
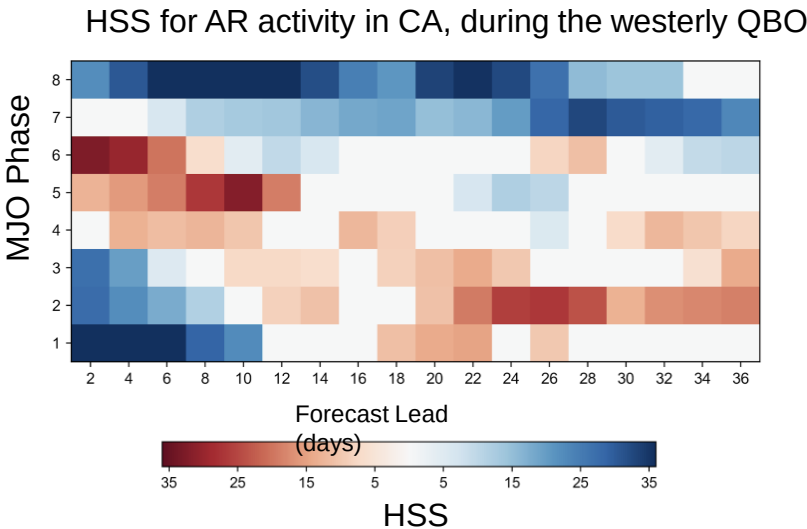
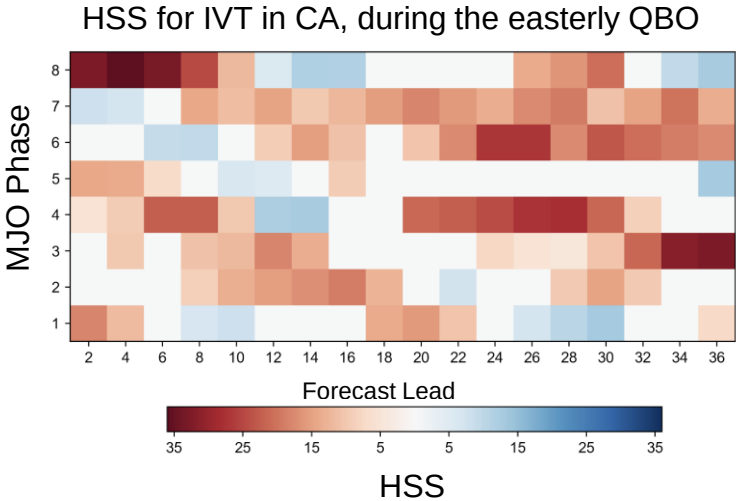
$$HSS = \frac{(\overset{\text{\# of correct forecasts}}{H} - E)}{(\underset{\text{\# of total forecasts}}{T} - \underset{\text{\# of correct by chance} = T/2}{E})} \times 100,$$

- Skillful forecasts out to 5+ weeks
- Skills propagate across lead time and MJO phase, consistent with the composites

AR Activity Forecasts

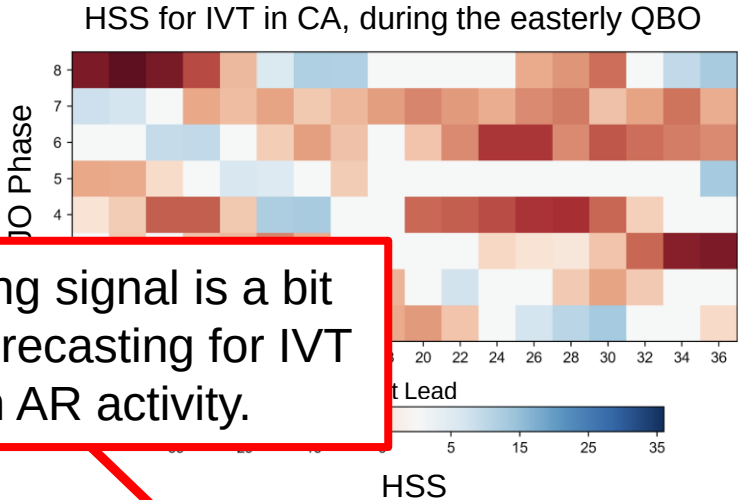
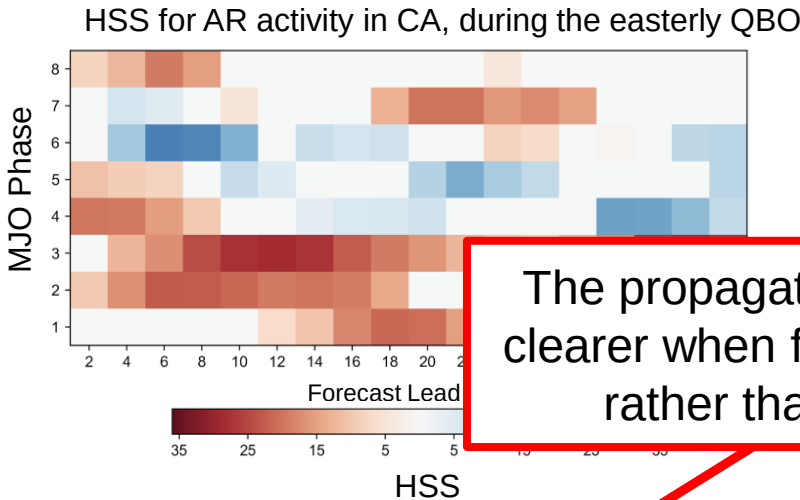


IVT Forecasts

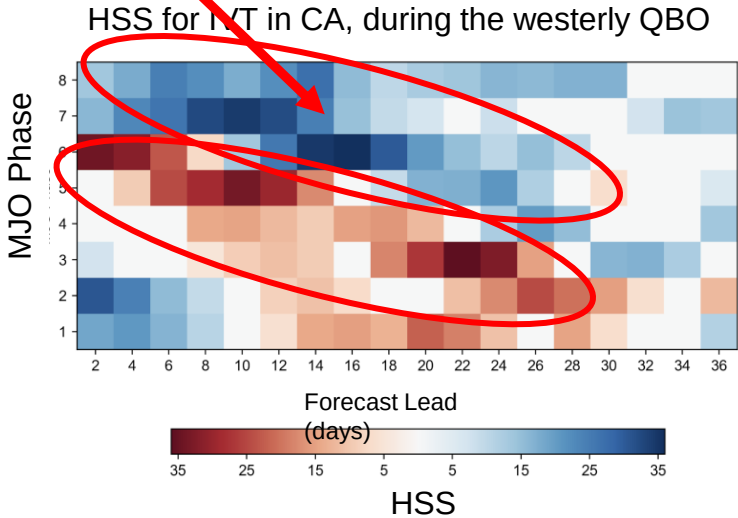
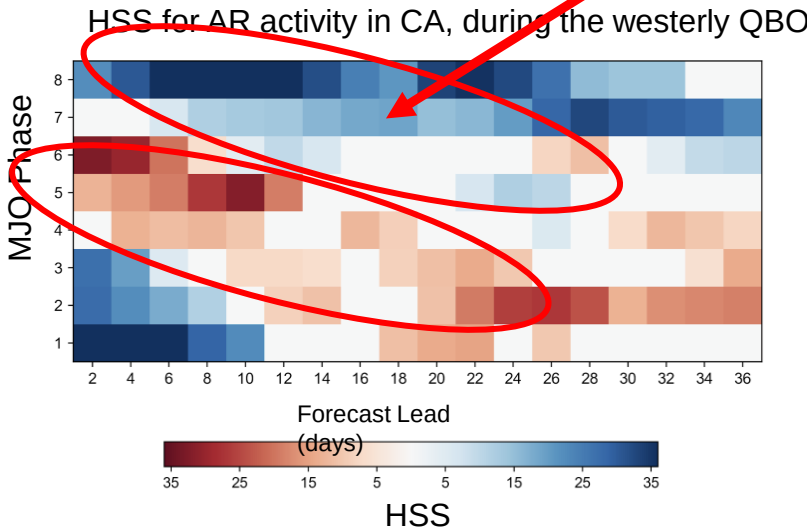


AR Activity Forecasts

IVT Forecasts



The propagating signal is a bit clearer when forecasting for IVT rather than AR activity.





Does the *current state* of the MJO and QBO offer information about *future* atmospheric river activity along the U.S. West Coast?

Yes, 3 to 6 weeks in advance.

If yes, can we use the *current state* of the MJO and QBO to skillfully forecast atmospheric river activity at subseasonal-to-seasonal lead times (~3 to 6 weeks in advance)?

Yes, skillful forecasts of opportunity exist 3 to 6 weeks in advance.



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TEASER of work in progress:

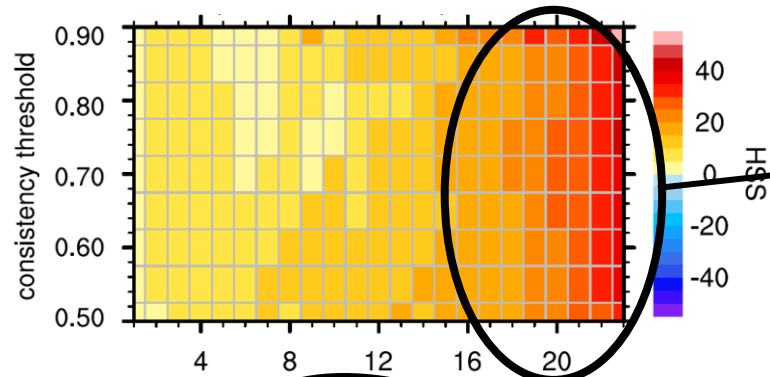
Back to the numerical models. Are they really hopeless beyond 14 days? What if we remove the bias? Or use a lagged ensemble approach to forecast based on model run to model run consistency?

TEASER Slide

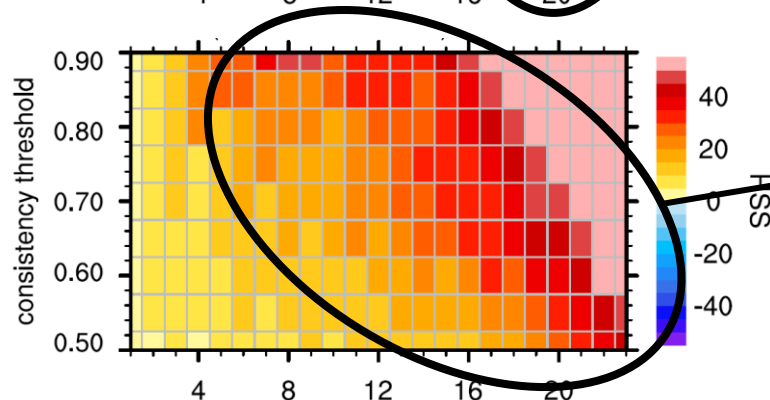
ECMWF - Bias Correction and Consistency Forecasting



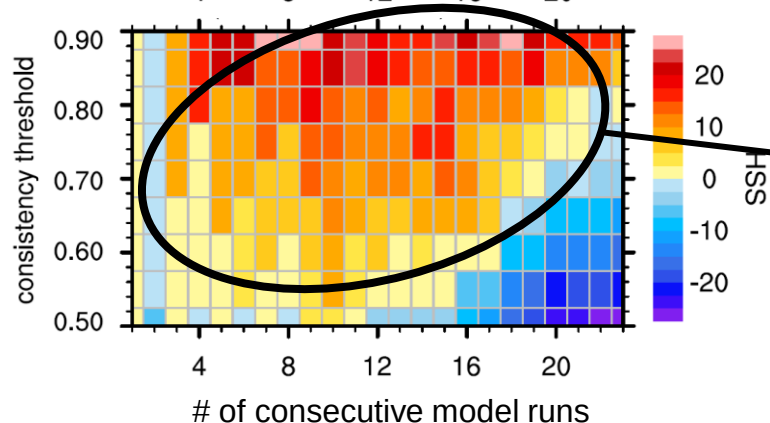
ECMWF ensemble mean AR forecasts for CA



Bias not removed forecast
Consistency forecasting is generally most skillful with lead times of less than 3 weeks.



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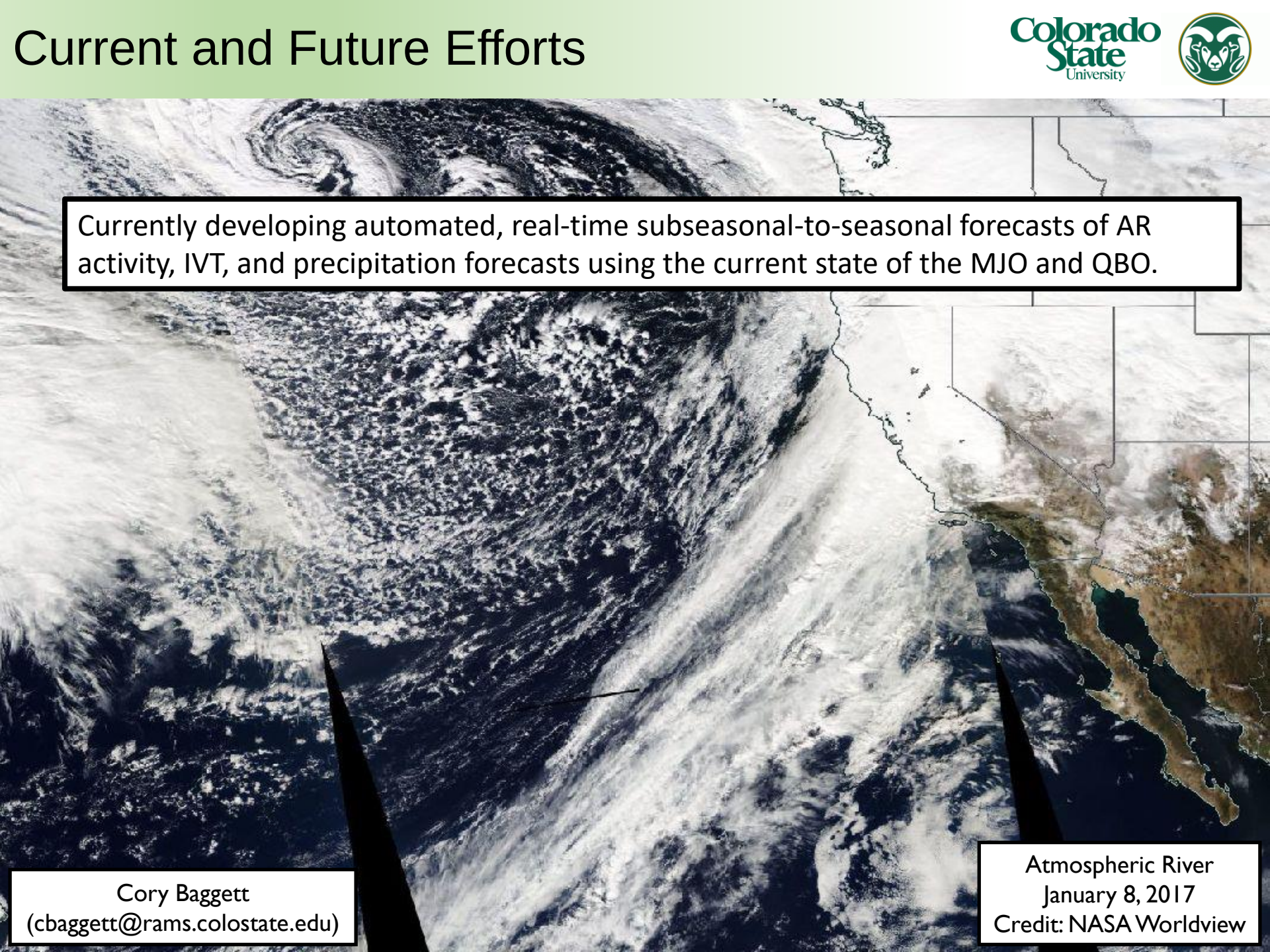


But how does the bias removed forecast compare to simply using the most recent model run?

There is a period of time when using the most recent model run is better, but at longer leads consistency forecasting has higher skill, especially at higher

consistency thresholds

Current and Future Efforts



Currently developing automated, real-time subseasonal-to-seasonal forecasts of AR activity, IVT, and precipitation forecasts using the current state of the MJO and QBO.

Cory Baggett
(cbaggett@rams.colostate.edu)

Atmospheric River
January 8, 2017
Credit: NASA Worldview

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Thank you!
Questions?

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Extra Slides

Regions of interest



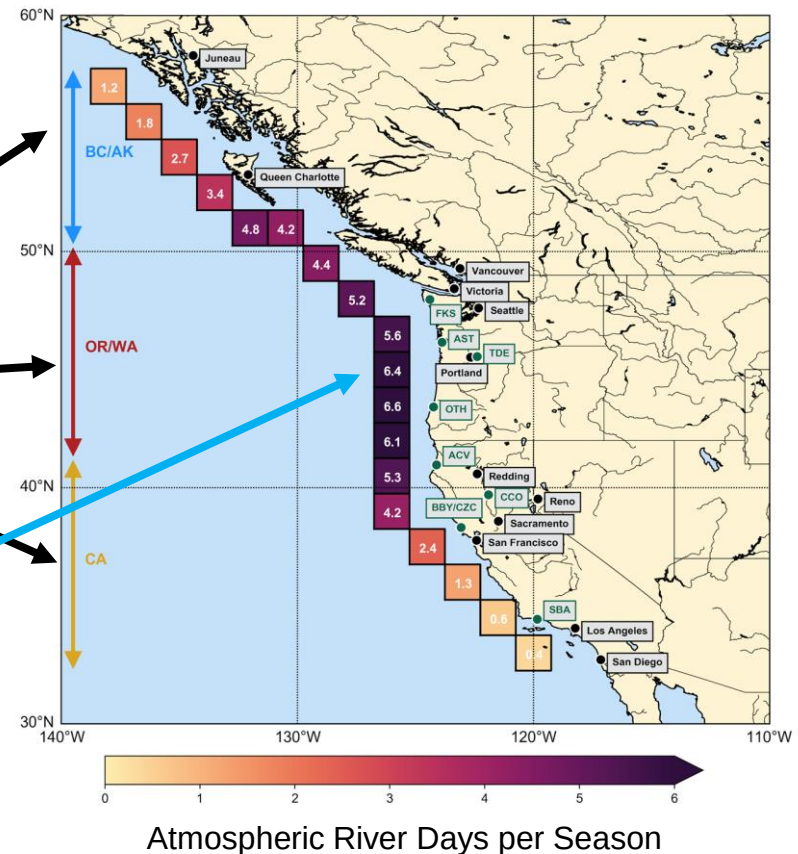
ERA-Interim Reanalysis data is the “truth” by which the models are judged.

The detection region is divided into 3 sub-regions with different atmospheric river climatologies:

- British Columbia / Alaska
- Oregon / Washington
- California

Atmospheric river frequencies are highest in the Pacific Northwest

Atmospheric River November-February Frequencies





2 x 2 Contingency Table (Wilks 2016)

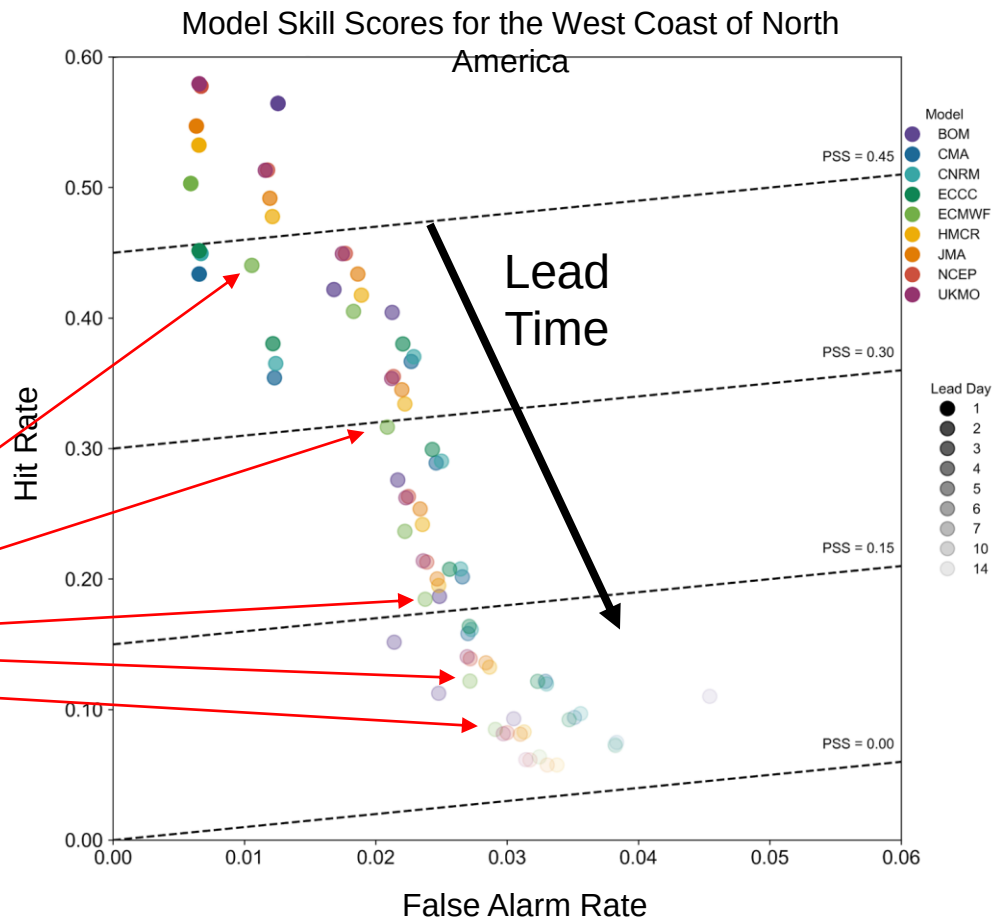
		<u>Observed?</u>	
		Yes	No
<u>Forecast?</u>	Yes	Hits	False Alarms
	No	Misses	Correct Rejections

Allows for the calculation of

- Bias,
- Hit Rate,
- False Alarm Rate
- Peirce Skill Score (PSS).

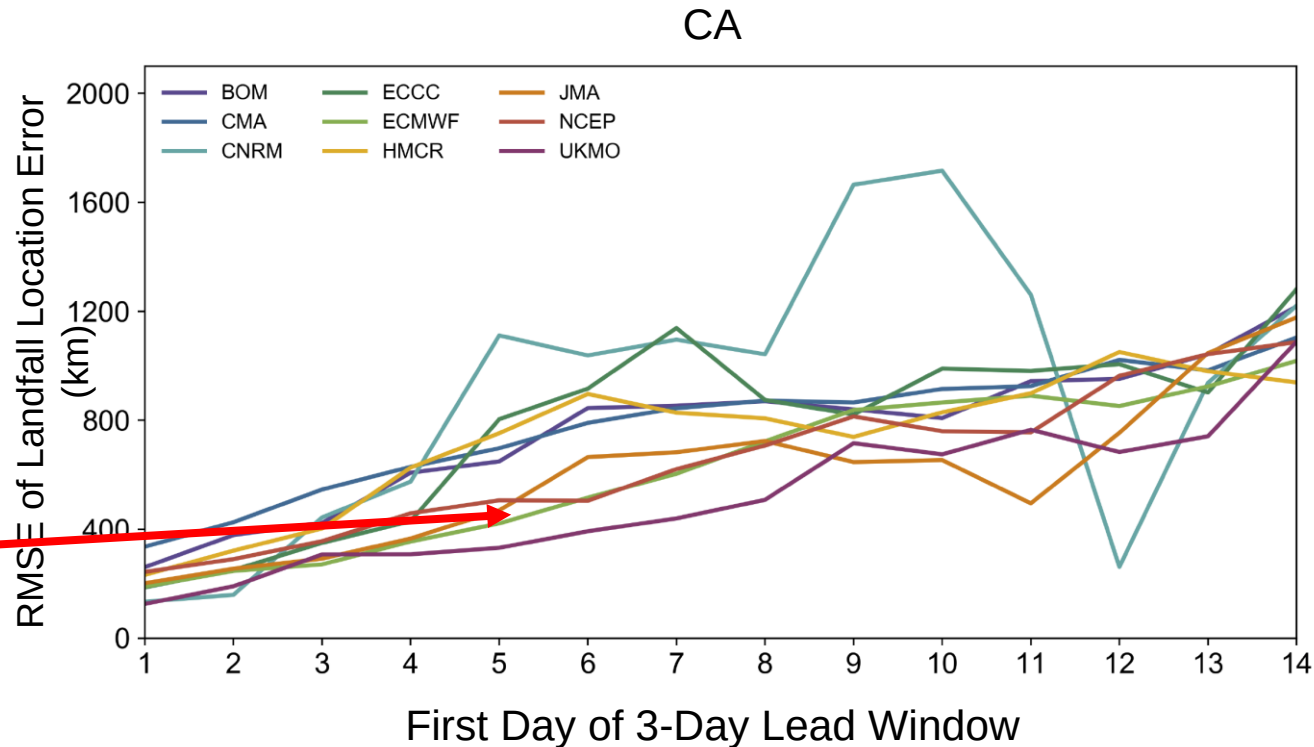
Hit Rate decreases and False Alarm Rate increases as lead time increases.

ECMWF consistently has the lowest False Alarm Rate, consistent with its “quiet” bias

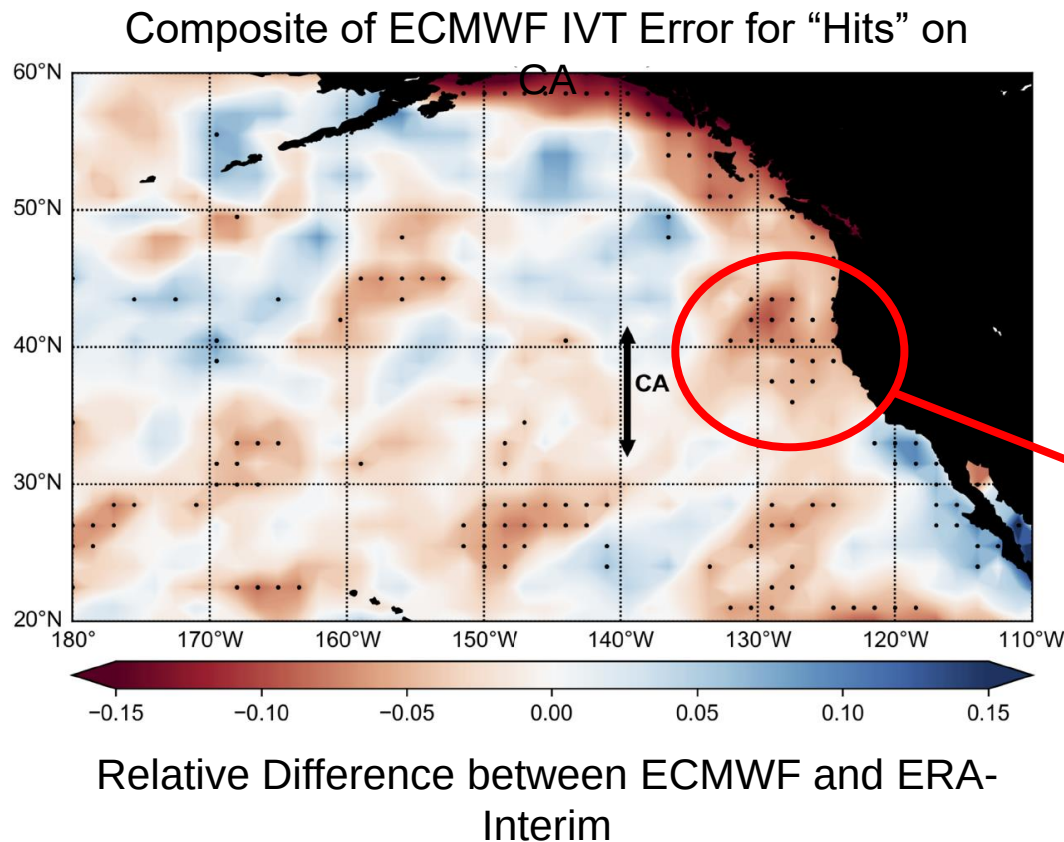


Colorado State University

ECMWF is no exception.



Atmospheric River IVT Errors



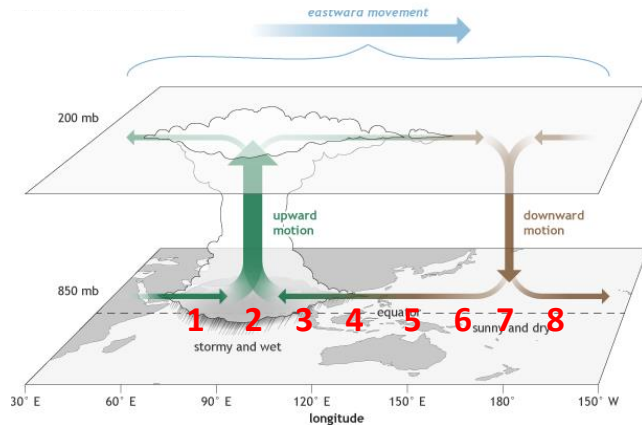
Red colors indicate that the model, on average, re-forecasts lower IVT values compared to reanalysis.

ECMWF shows, on average, offshore low-IVT errors.

The Madden-Julian Oscillation – a source of predictability at S2S lead times

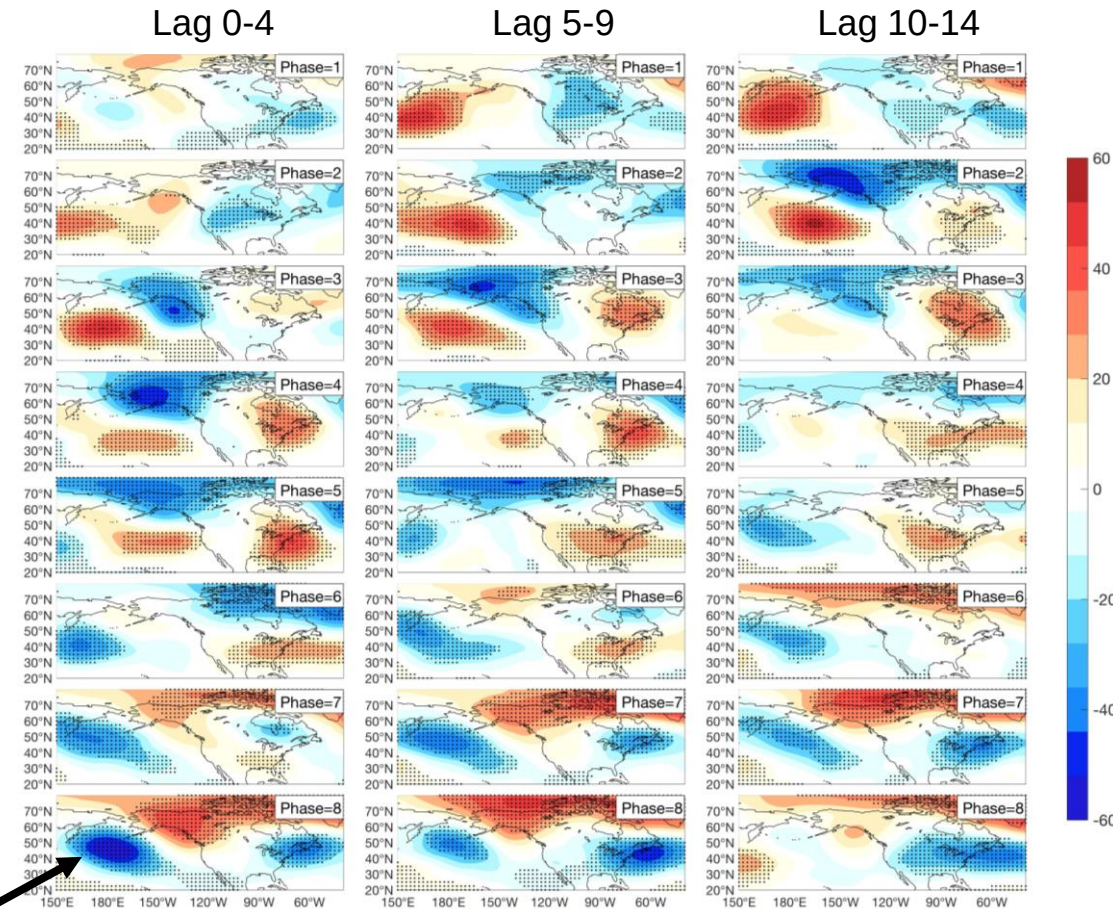


Madden-Julian Oscillation



Shaded regions represent 500-mb geopotential height anomalies, according to MJO phase

Teleconnection

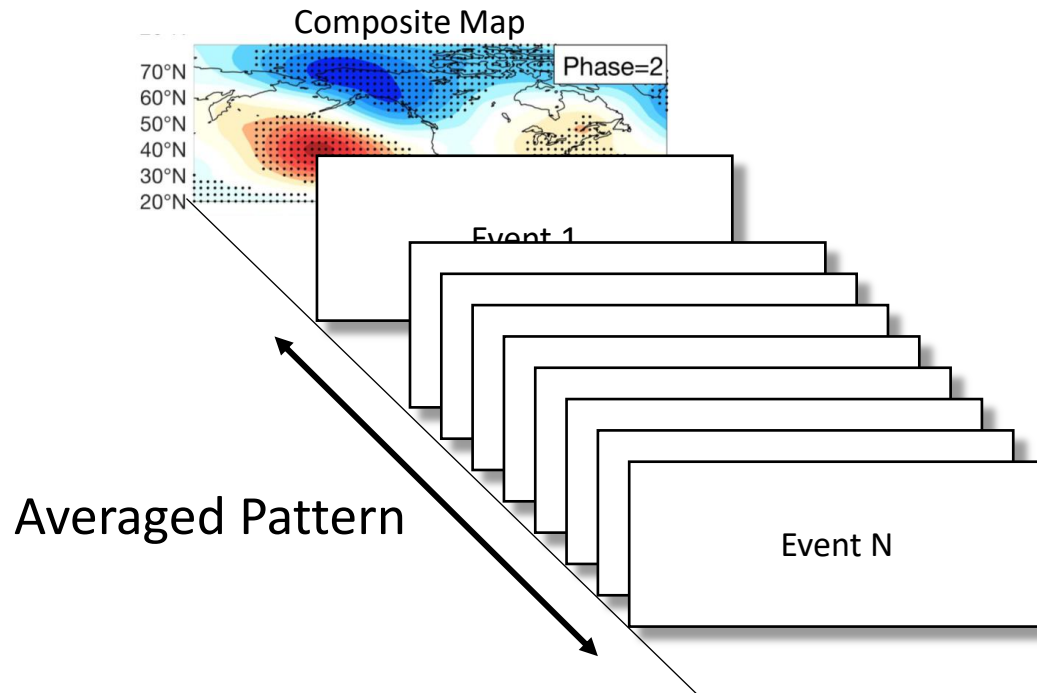


Can we use the MJO to forecast atmospheric river activity along the U.S. West Coast at S2S lead times of ~3 to 6 weeks?

Pattern Consistency

At least two reasons can result in a strong MJO teleconnection signal:

- A few strong events dominate the signal
- Or similar teleconnections continually occur



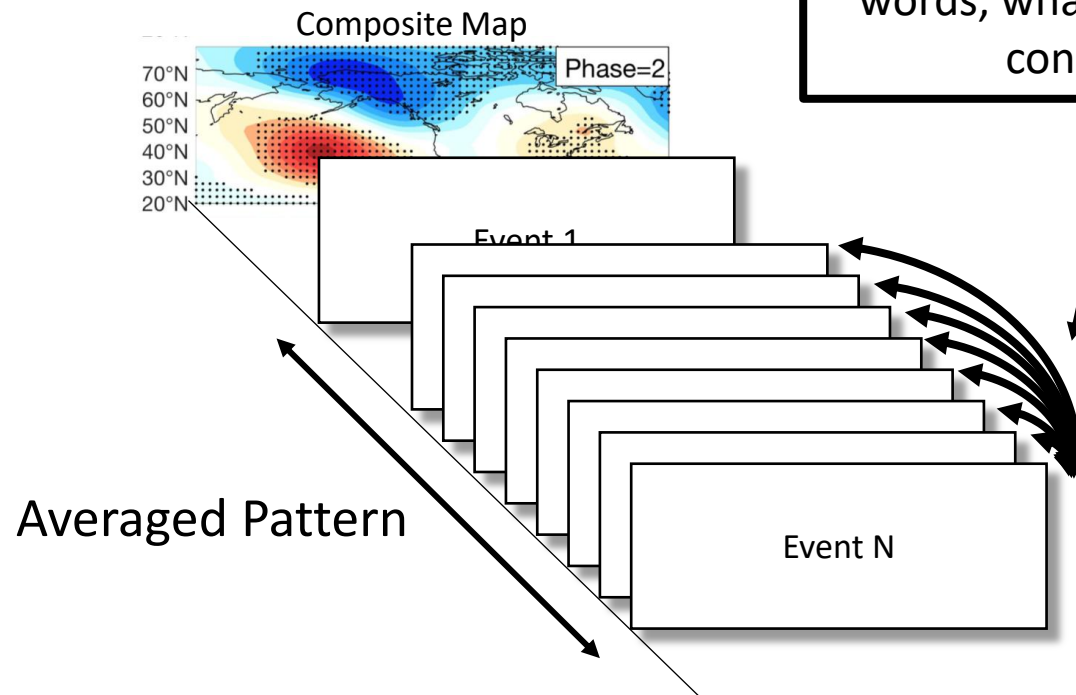
Pattern Consistency of Events per MJO phase



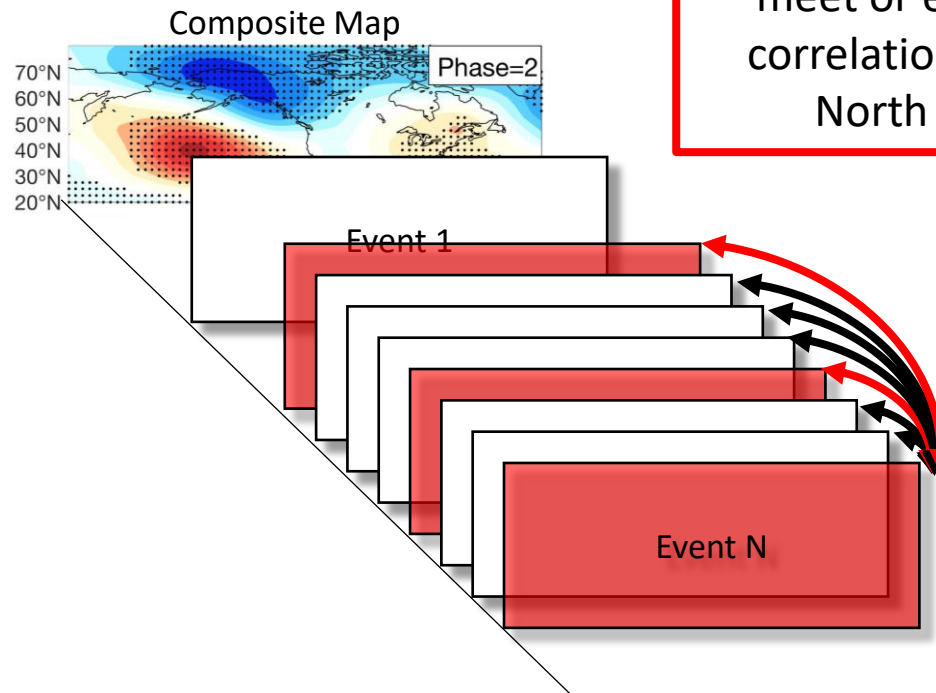
At least two reasons can result in a strong MJO teleconnection signal:

- ~~• A few strong events dominate the signal~~
- Or similar teleconnections continually occur

We will test the second idea by asking, how similar are individual events to each other per MJO phase? In other words, what is their pattern consistency?

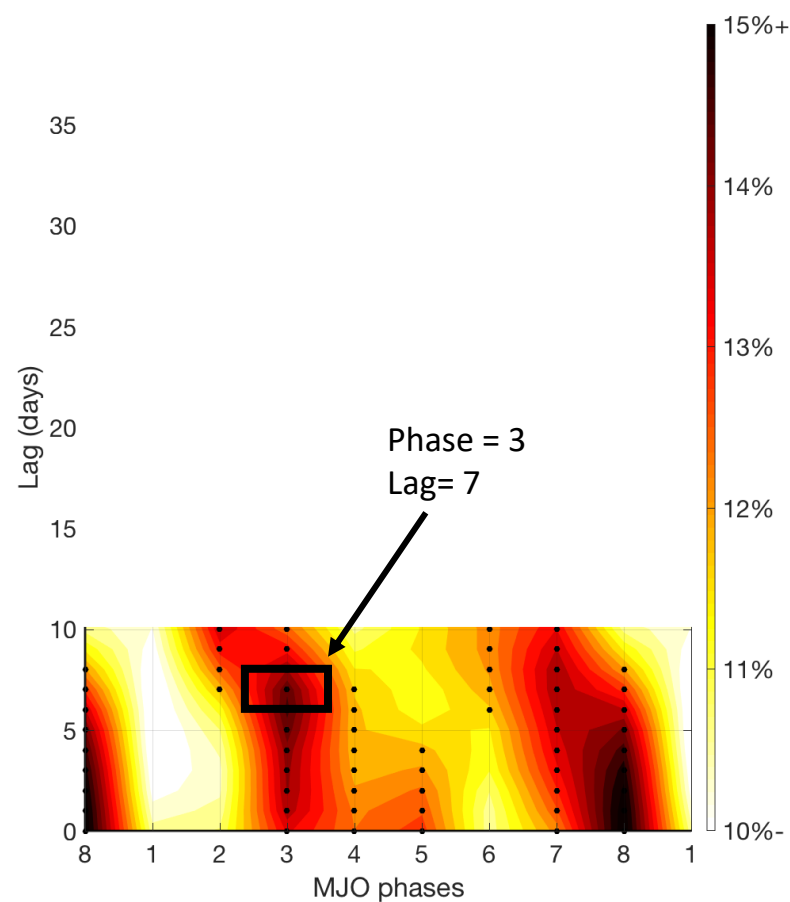


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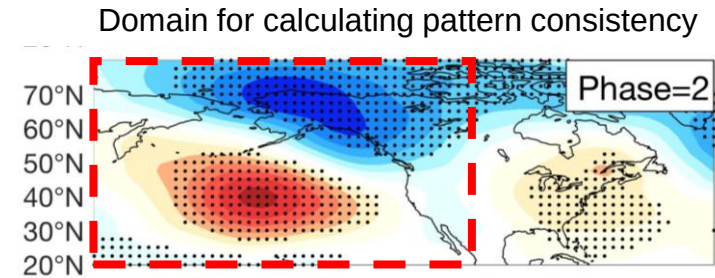


We quantify pattern consistency by counting the number of pairs of events that meet or exceed a pattern correlation of 0.5 over the North Pacific basin.

Pattern Consistency of Events per MJO phase

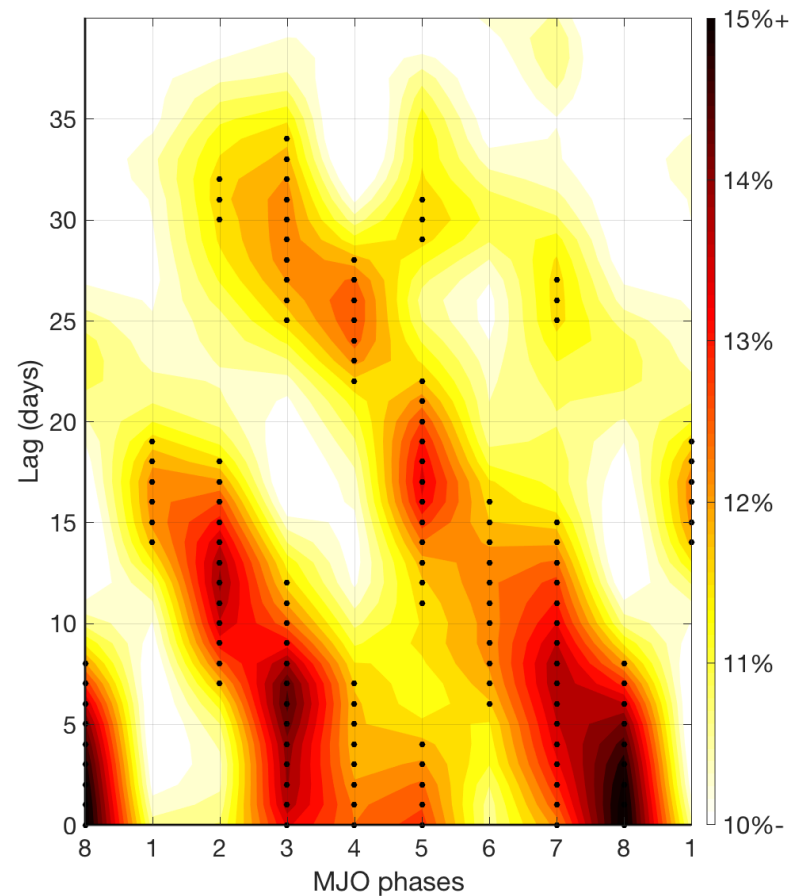


High Pattern Consistency

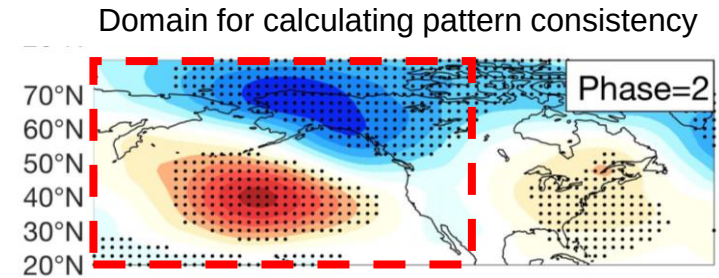


Low Pattern Consistency

Pattern Consistency of Events per MJO phase

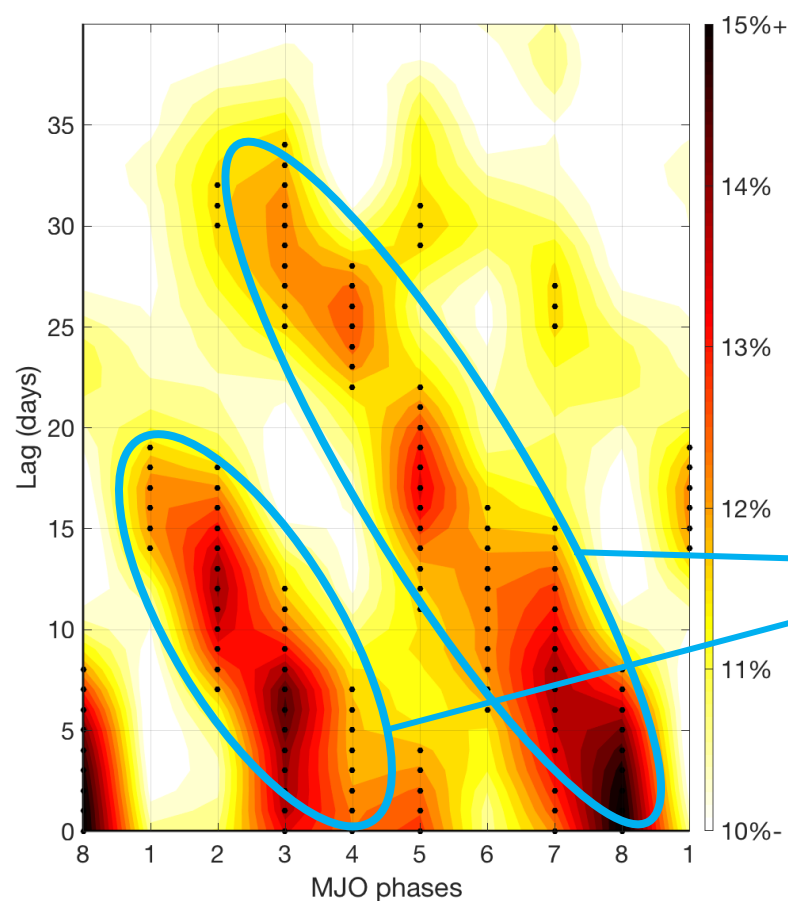


High Pattern Consistency

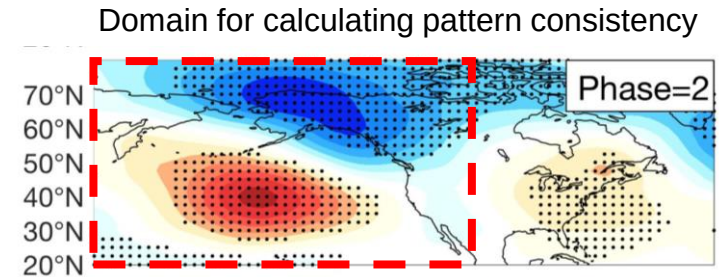


Low Pattern Consistency

Pattern Consistency of Events per MJO phase



High Pattern Consistency

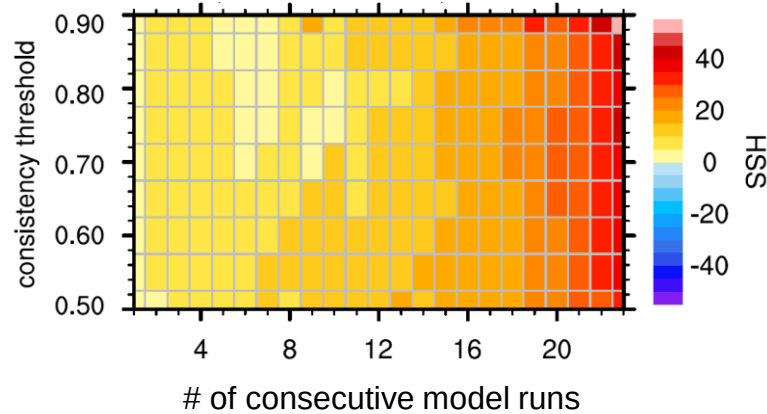


The pattern consistency is a function of MJO phase and time lag.

Low Pattern Consistency



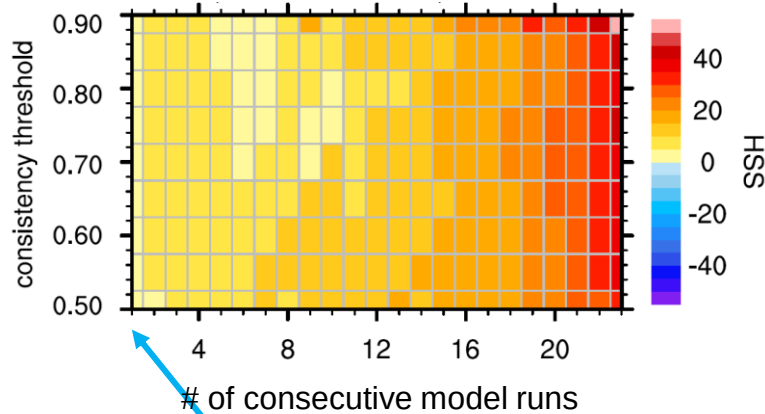
ECMWF ensemble mean AR forecasts for CA



Skill scores of consistency forecasts are calculated across the parameter space of 1) number of consecutive model runs used and 2) consistency threshold



ECMWF ensemble mean AR forecasts for CA

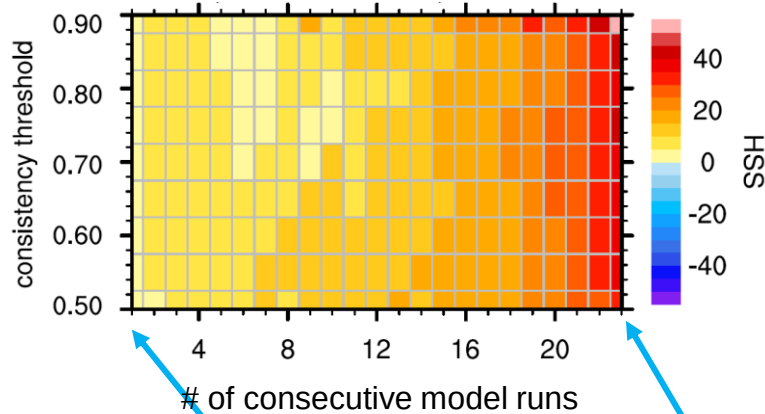


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Longest lead time forecast (~6 weeks). Only one model run has seen the forecast valid date.



ECMWF ensemble mean AR forecasts for CA



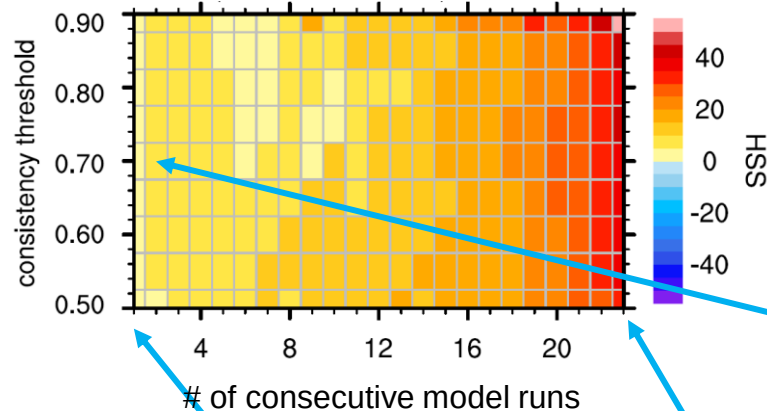
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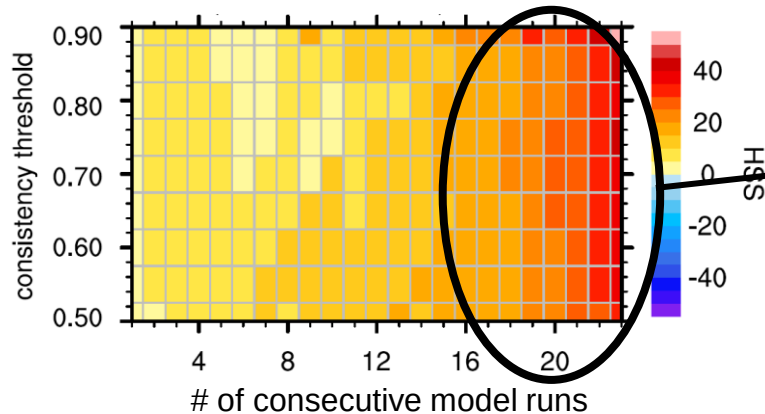
Threshold = the fraction of consecutive model runs that agree on the sign of the AR anomaly. E.g., 0.70 implies 70% agree on the sign of the anomaly.

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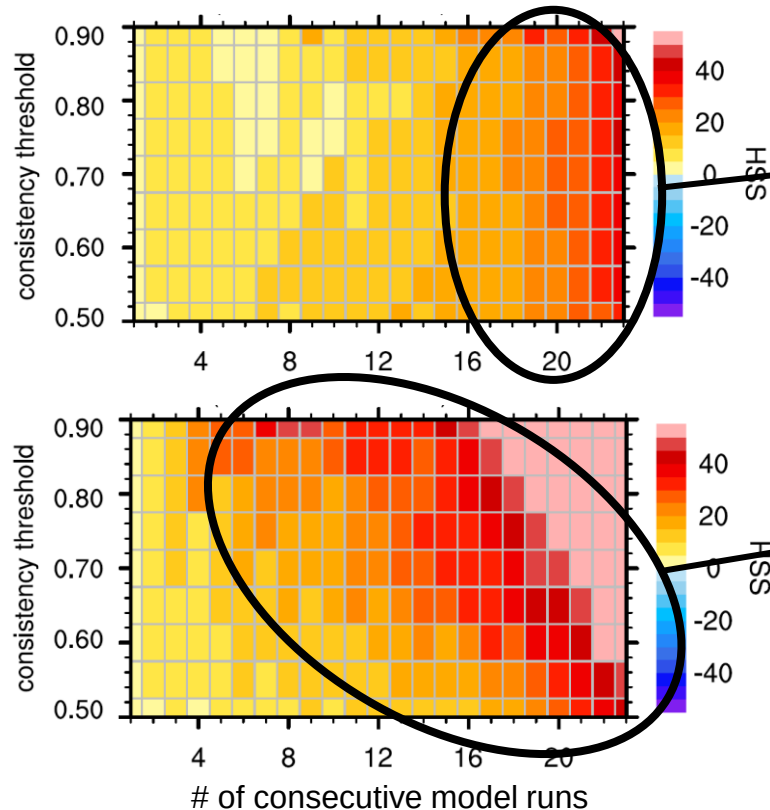
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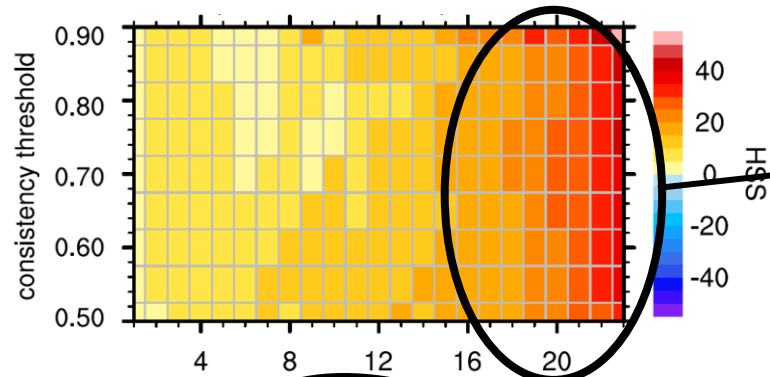
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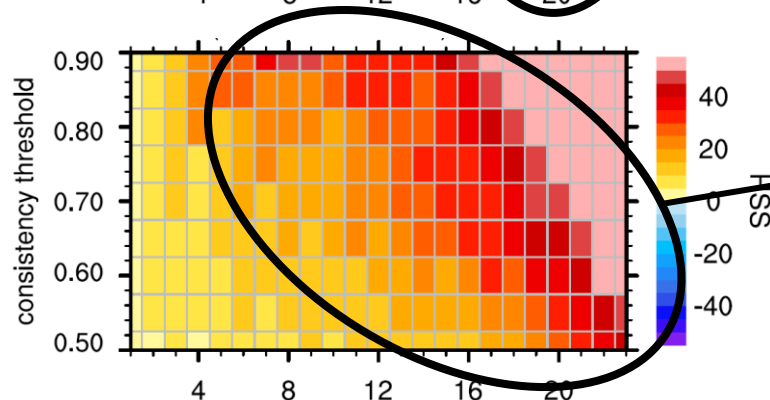
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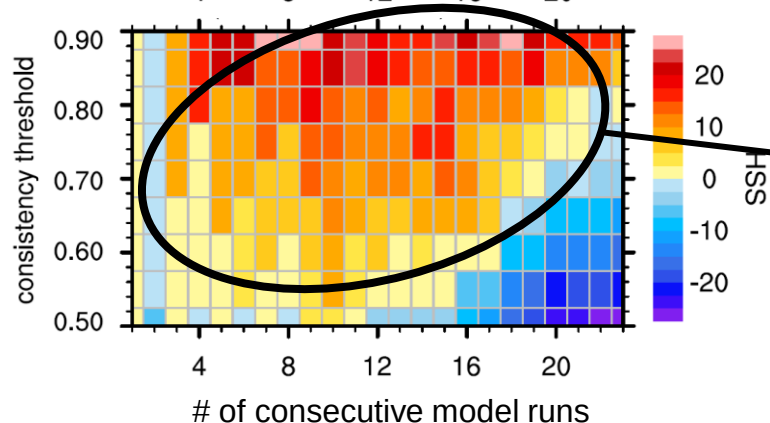
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