

Experimental S2S Forecasts for California: *Wet/AR and Dry/Ridging Conditions*

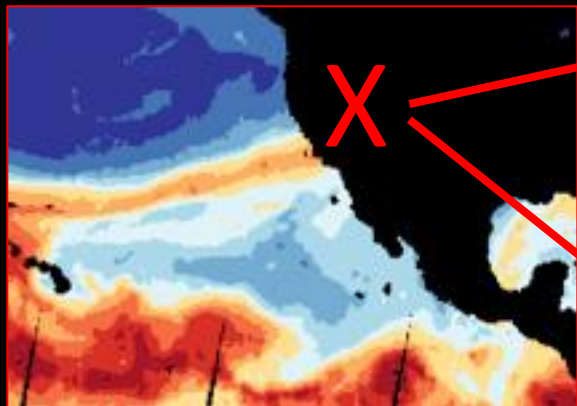
Duane Waliser

Mike DeFlorio, Peter Gibson, Bin Guan,
Elias Massoud, Huikyo Lee, Alex Goodman
Jet Propulsion Laboratory/Caltech
Pasadena, CA

In collaboration with
Center for Western Weather and Water Extremes (M. Ralph, A. Subramanian, etc.)
With Support from
CA Department of Water Resources (J. Jones)
NASA Energy and Water Cycle Research Program (J. Entin)

Western States Water Council
May 14-16, 2018

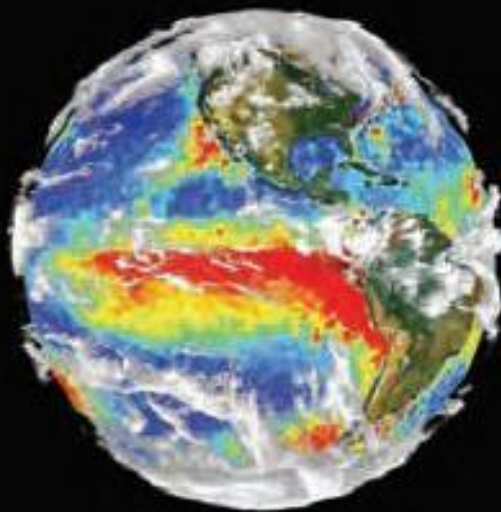
Western U.S. : Wet or Dry?



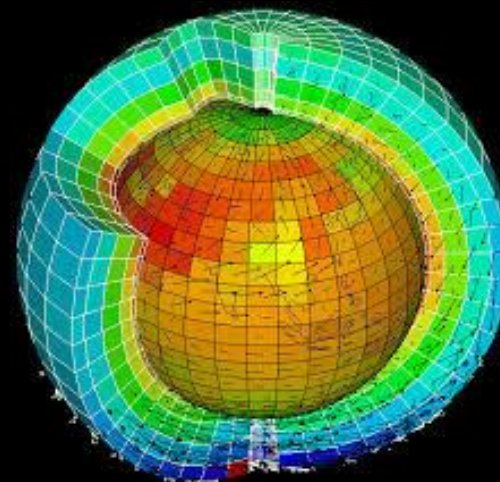
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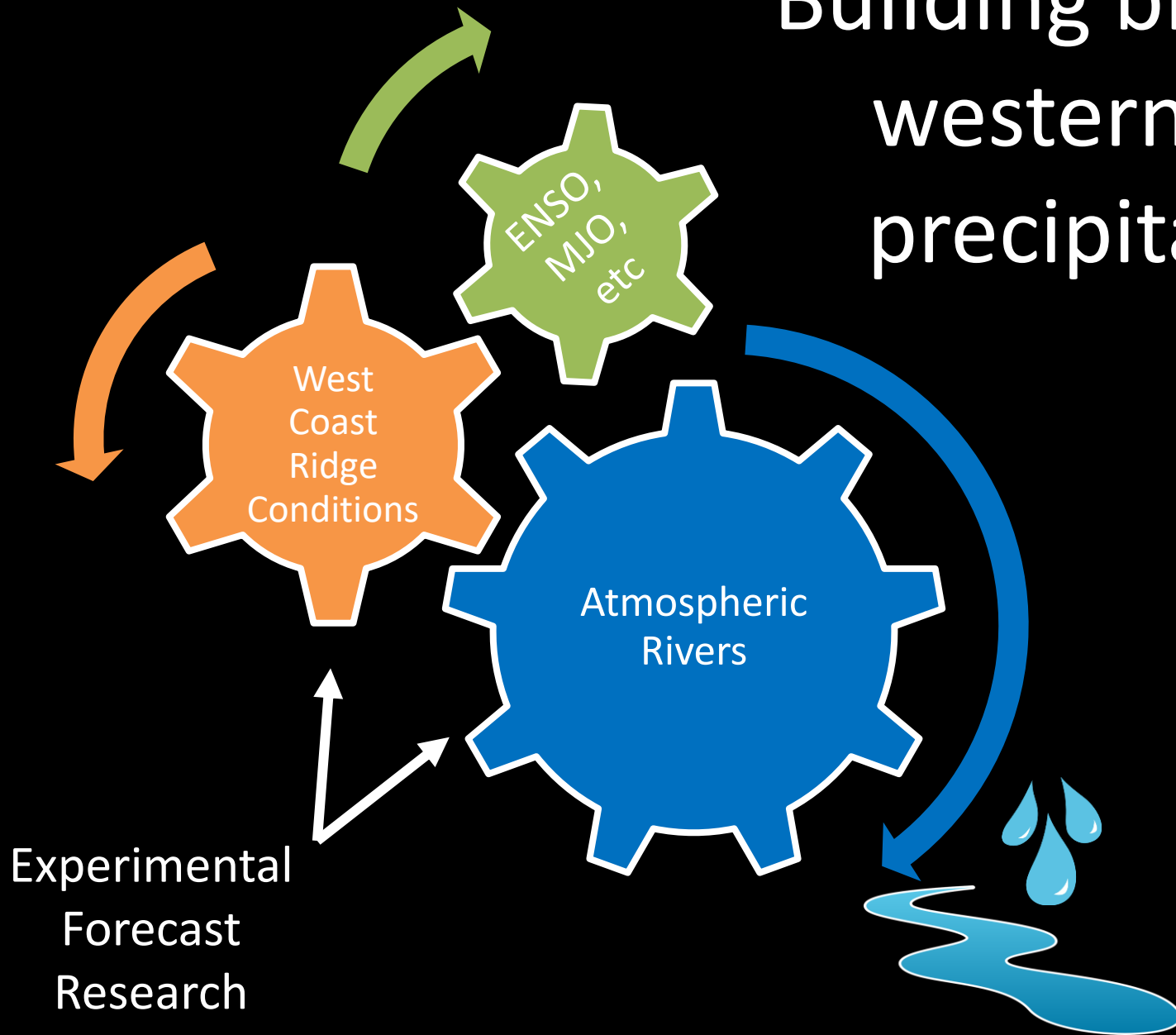
Objective:
Predict Western US
Precipitation Conditions



Research Ingredients:
Global Perspective
AR Detection Algorithm
S2S Project Database
(*e.g. NCEP, ECMWF, EC*)



Building blocks of western U.S. precipitation



Outline

I. Tools and Resources

- *Global AR Detection Algorithm*
- *WCRP/WWRP S2S Project & Forecast/Hindcast Database*

II. AR Predictions

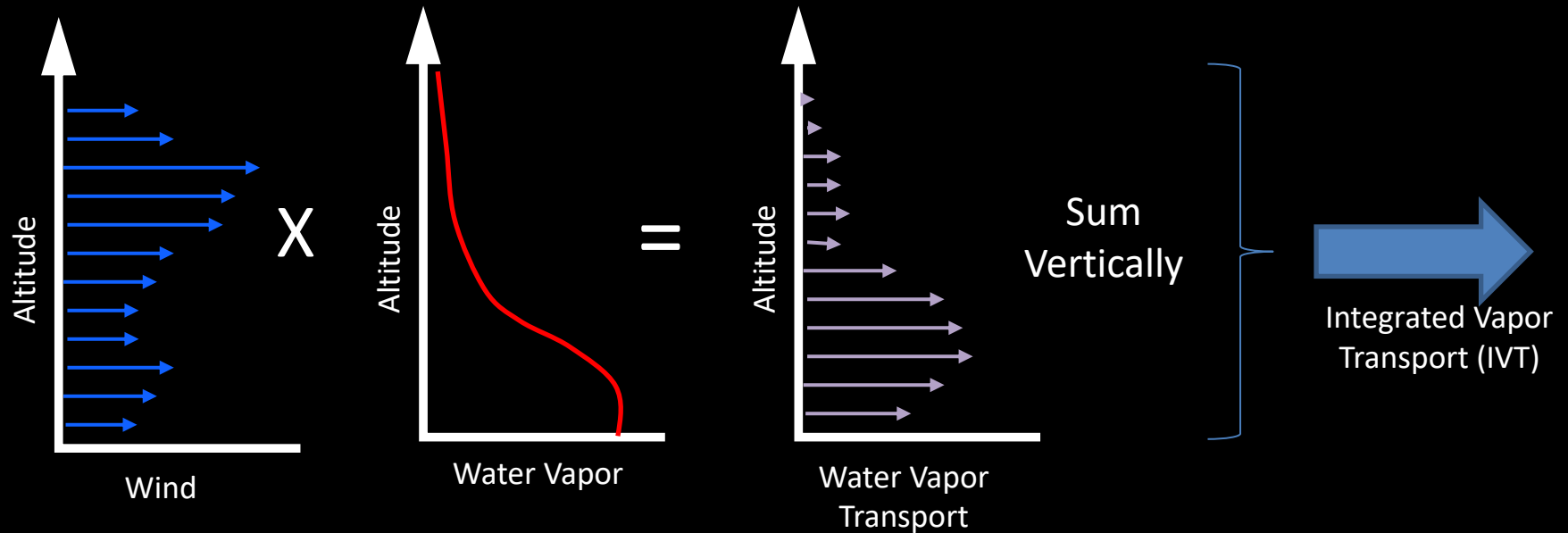
- *Weather predictions of Individual ARs (e.g. 0-15 days)*
- *Subseasonal predictions (e.g. 2-4 weeks)*
- *Experimental S2S Predictions*

III. Ridging: Modulating Droughts and ARs

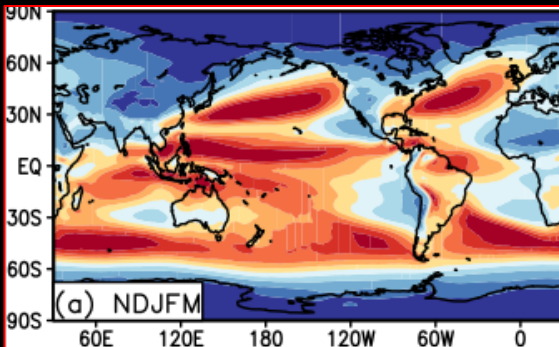
- *Early Considerations*

Global AR Detection

I. Compute IVT

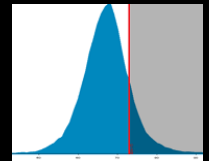


II. Map IVT globally



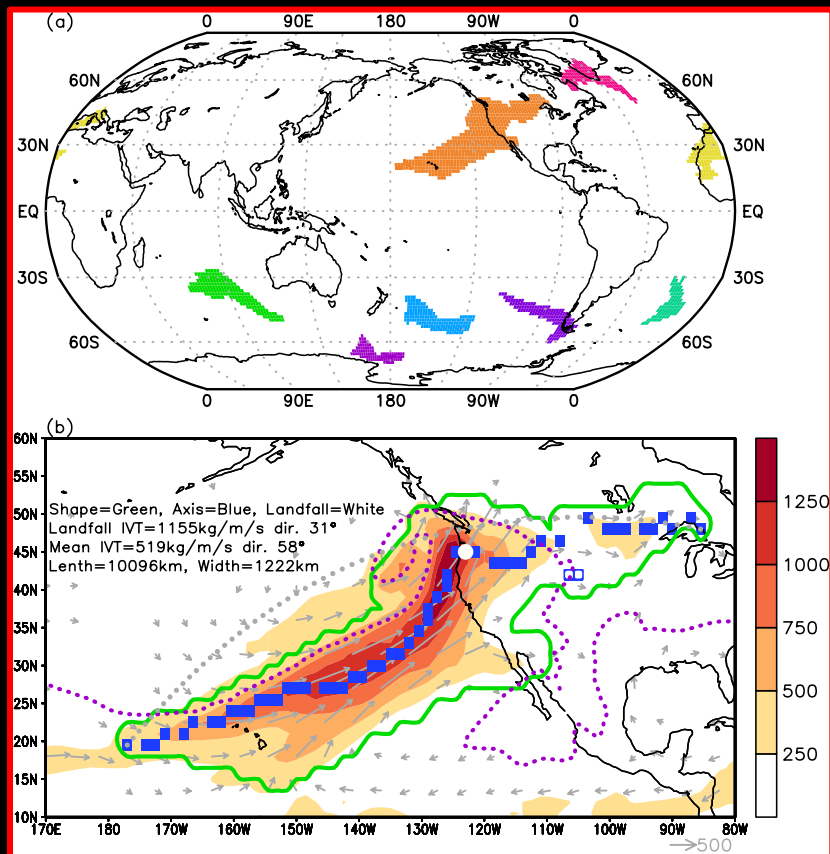
III. Apply AR Criteria

- IVT > 85th percentile
- Look for contiguous areas
- Length > 2000 km
- Length/Width > 2



Gives Long, Narrow Extreme Moisture Transports i.e. Rivers

Global AR Detection Algorithm

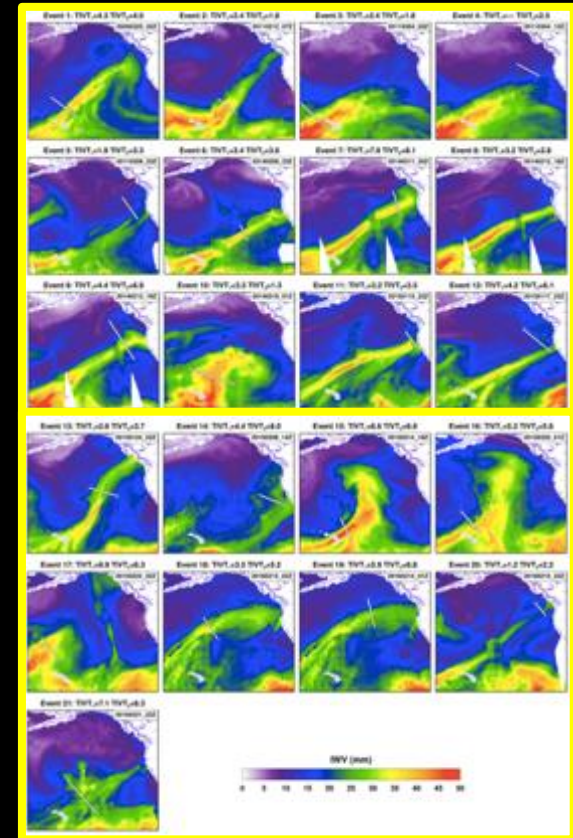
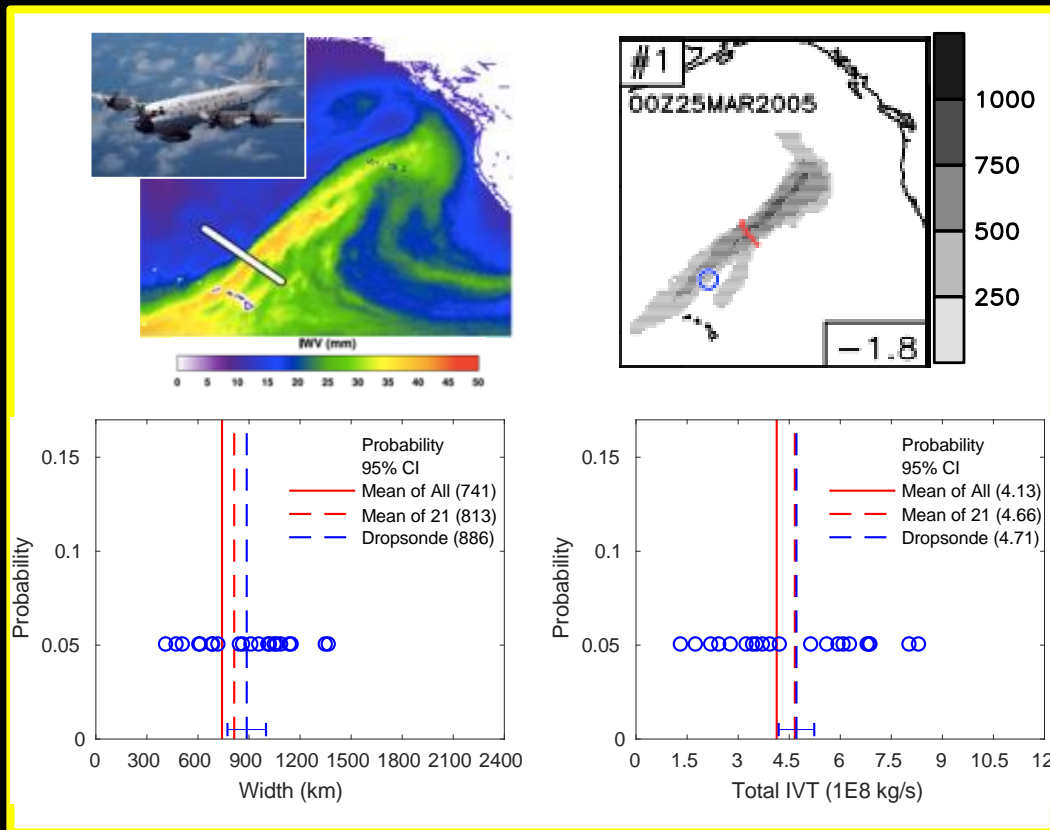


Guan and Waliser (2015)

- Based on Integrated Vapor Transport (IVT) fields and a number of common AR criteria (e.g. Ralph et al. 2004).
- Developed for global studies and for observations/reanalysis and models.
- Applied to:
 - ERA-I, MERRA-2, CFSR, NCEP/NCAR
- Code and databases available at:
 - <https://ucla.box.com/ARcatalog>
- Databases include AR Date, $IVT_{x,y}$, Shape, Axis, Landfall Location, etc.

Algorithm Validation Support from CalWater

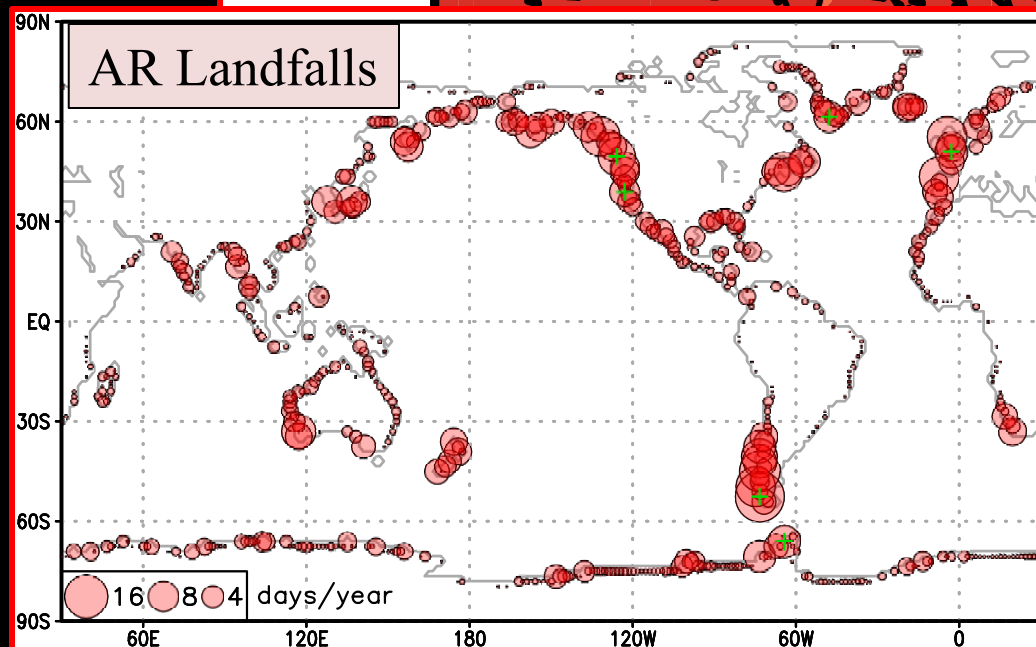
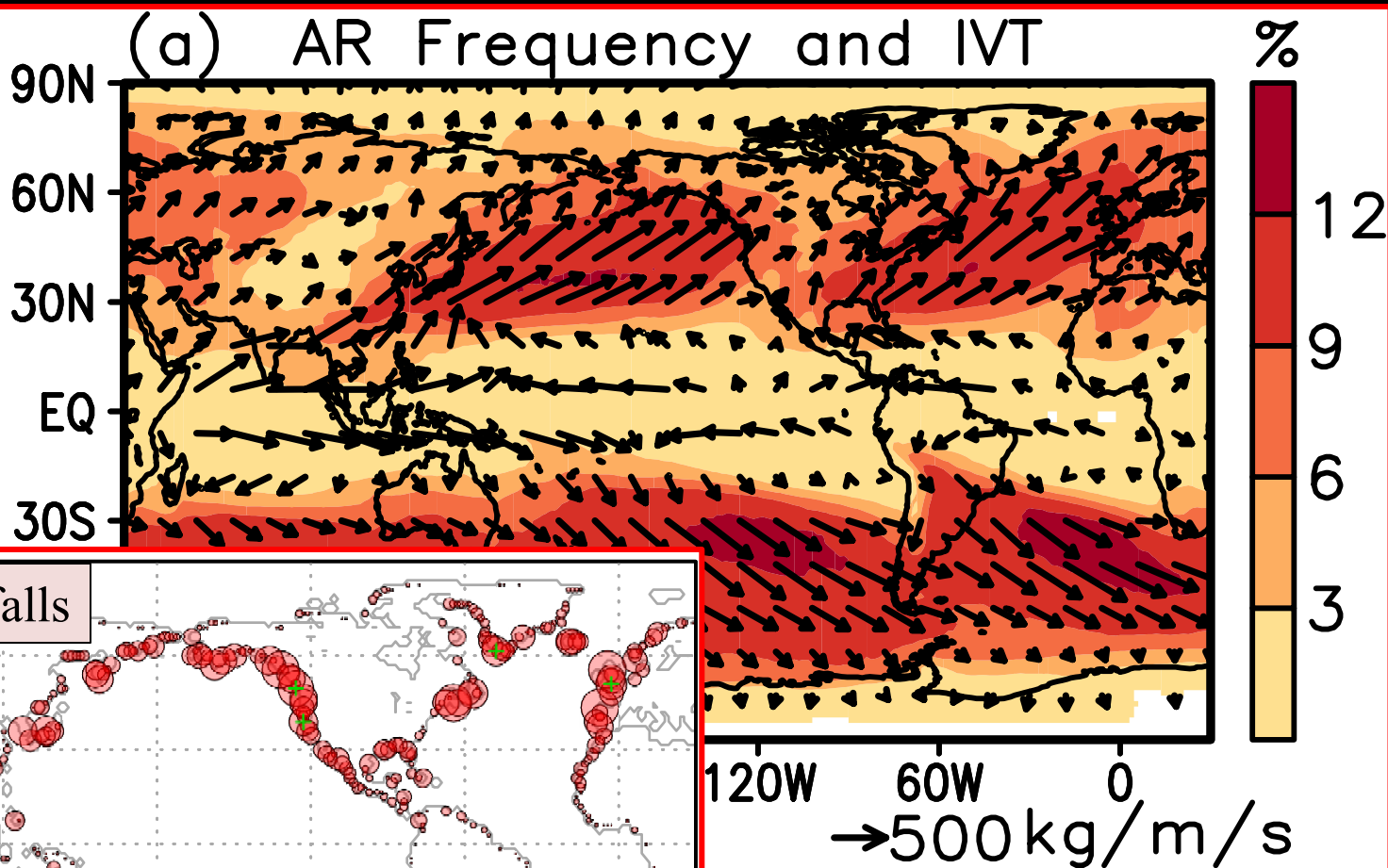
Guan, Waliser and Ralph (2018)



IVT Histograms Based On
 5636 NE Pacific ARs from ERA-I
 125-163W, 23-46N
 Jan 15-Mar 25 1979-2016

Ralph et al. (2017)
 21 AR Event Transects
 4.7 +/- 1.9kg/s
 Min 1.3; Max 8.3

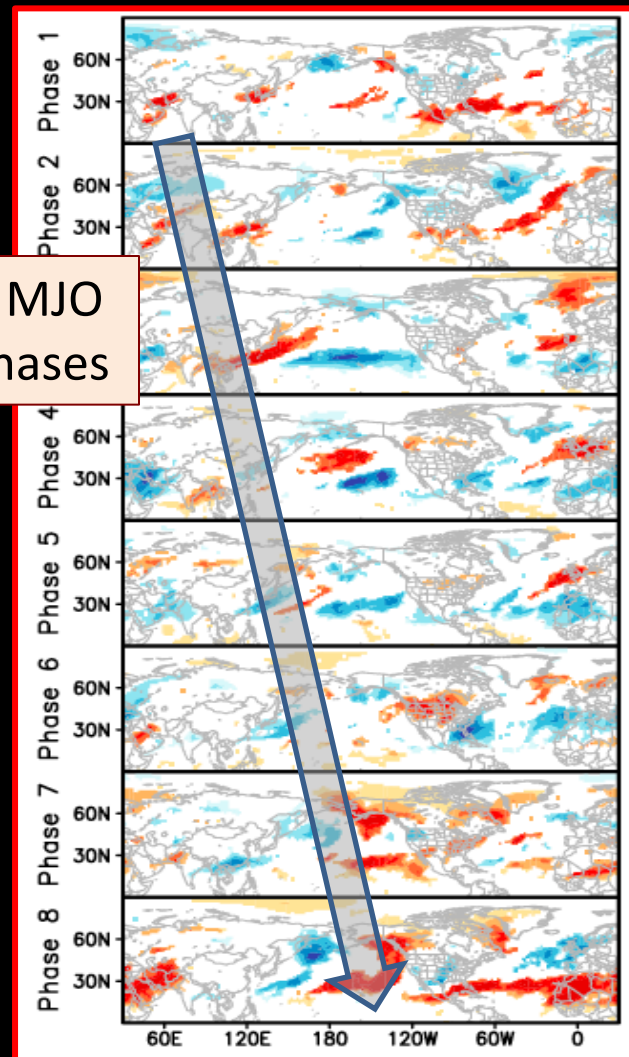
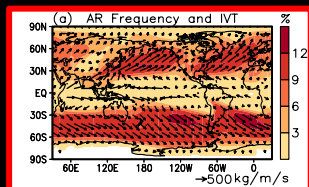
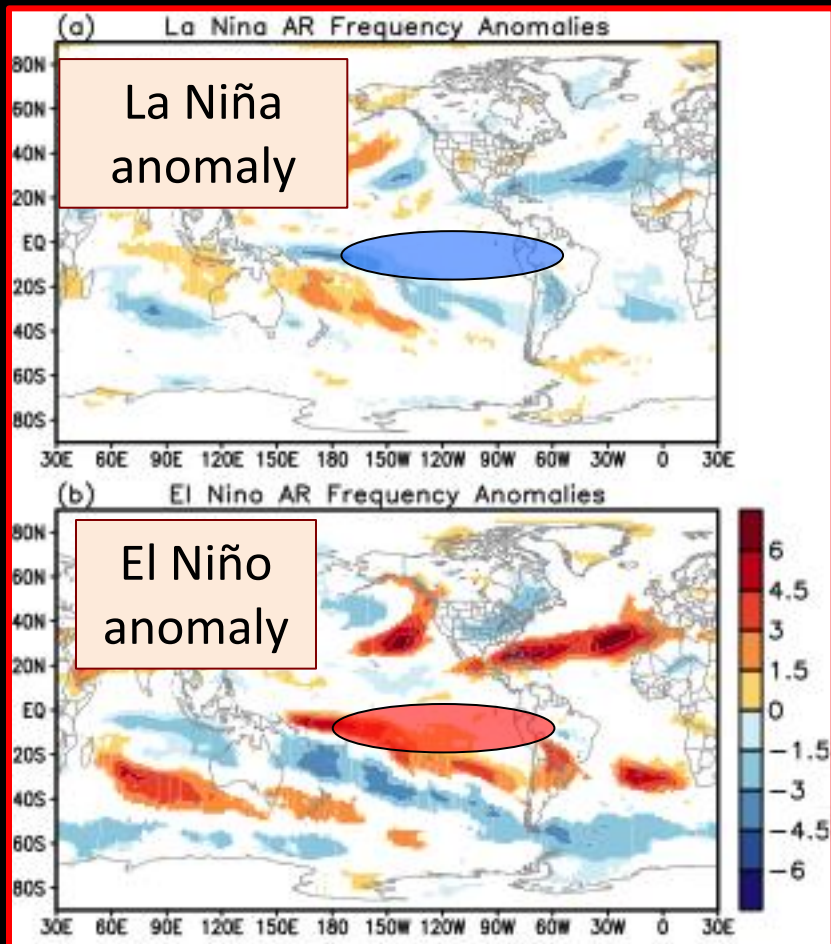
Global AR Characteristics



Climate Patterns and ARs

El Nino Southern Oscillation (ENSO)

Madden-Julian Oscillation (MJO)



Guan and Waliser (2015)

Subseasonal Forecast Database

WCRP-WWRP S2S Project

S2S Database

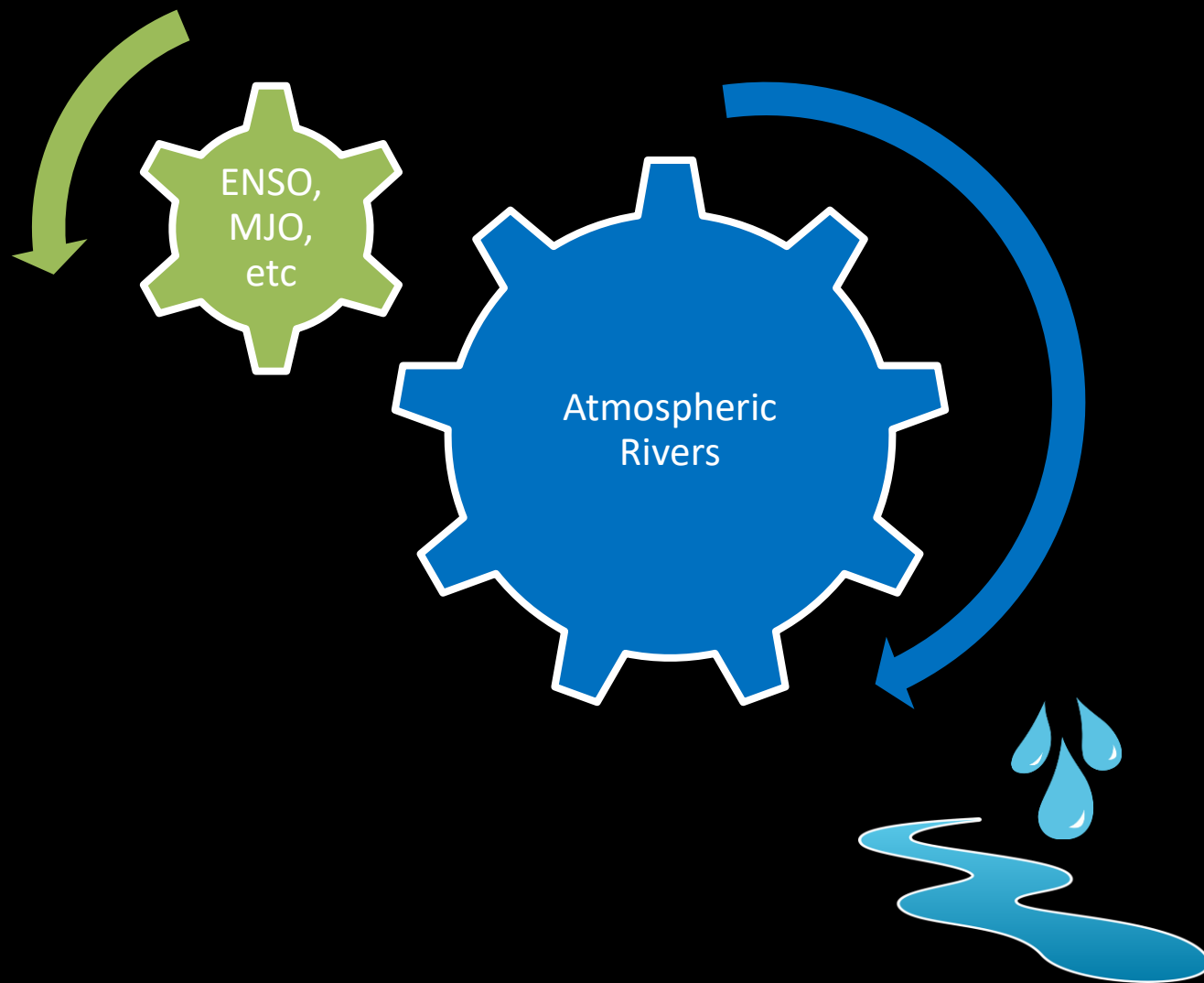


- Hindcasts for Research
- Forecasts for Experimental Products

	Time-range	Resol.	Ens. Size	Freq.	Hcsts	Hcst length	Hcst Freq	Hcst Size
ECMWF	D 0-46	T639/319L91	51	2/week	On the fly	Past 20y	2/weekly	11
UKMO	D 0-60	N216L85	4	daily	On the fly	1996-2009	4/month	3
NCEP	D 0-44	N126L64	4	4/daily	Fix	1999-2010	4/daily	1
EC	D 0-32	0.6x0.6L40	21	weekly	On the fly	1995-2014	weekly	4
CAWCR	D 0-60	T47L17	33	weekly	Fix	1981-2013	6/month	33
JMA	D 0-34	T319L60	25	2/weekly	Fix	1981-2010	3/month	5
KMA	D 0-60	N216L85	4	daily	On the fly	1996-2009	4/month	3
CMA	D 0-45	T106L40	4	daily	Fix	1886-2014	daily	4
CNRM	D 0-32	T255L91	51	Weekly	Fix	1993-2014	2/monthly	15
CNR-ISAC	D 0-32	0.75x0.56 L54	40	weekly	Fix	1981-2010	6/month	1
HMCR	D 0-63	1.1x1.4 L28	20	weekly	Fix	1981-2010	weekly	10

- Downloaded U,V,Q fields from all models' multi-decade subseasonal hindcasts.
- Computed Integrated Vapor Transport (IVT) and Applied AR Detection.
- Examined global AR event weather and AR activity subseasonal forecast skill

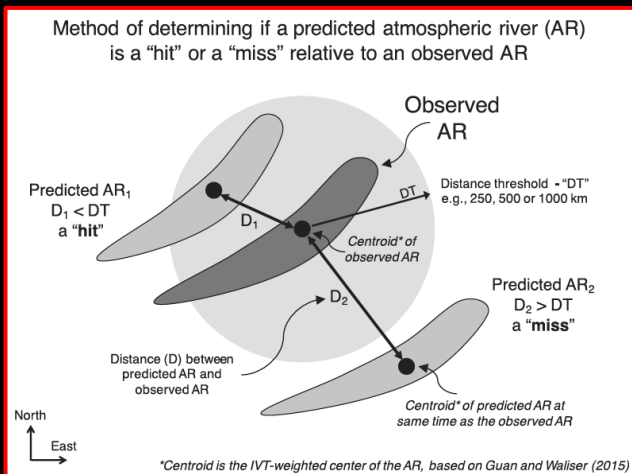
Start with AR Forecasting



Predicting AR Events

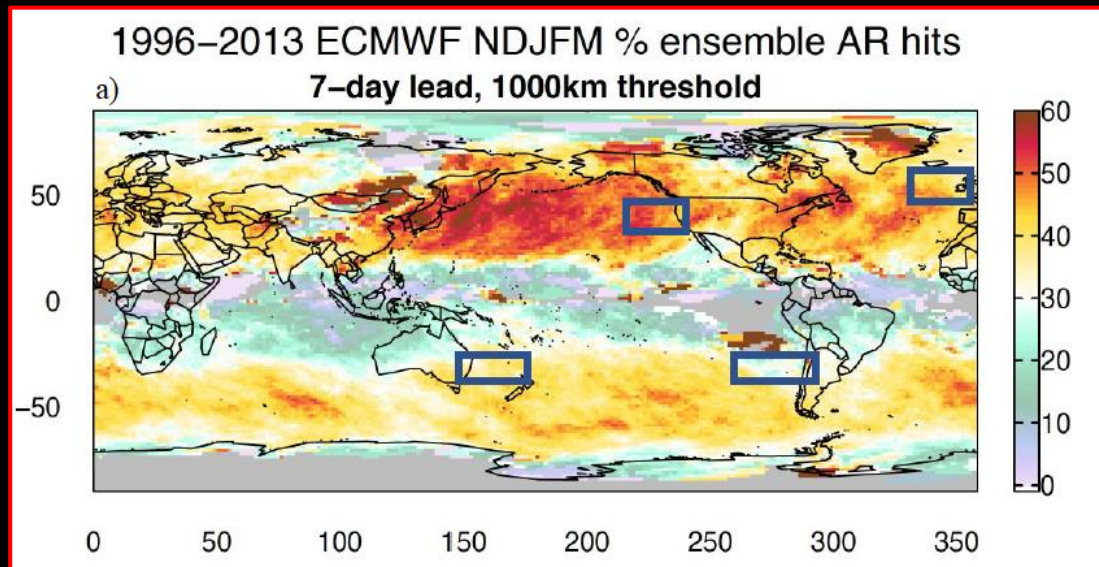
Considering “Weather” Lead Times

How well do our global NWP models – ECMWF in this case - predict AR occurrence & position?



ECMWF Subseasonal to Seasonal (S2S) hindcasts include twice-per-week, 11 member ensembles, from 1996-2013.

Courtesy WCRP/WWRP
S2S Project

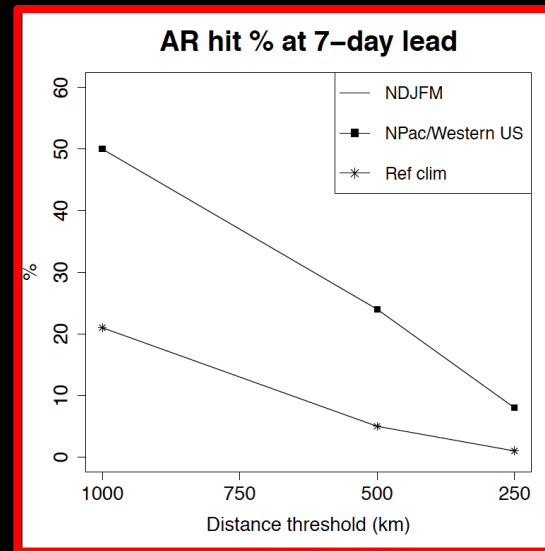
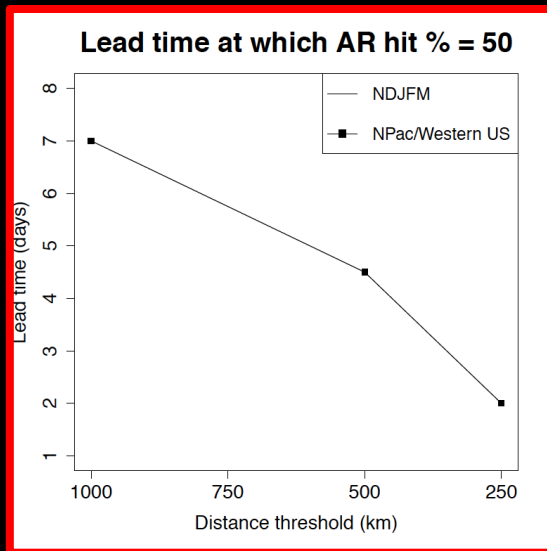
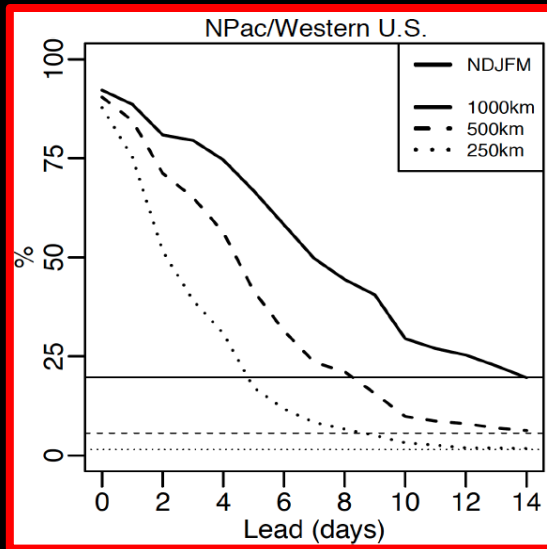


DeFlorio, Waliser, Guan, Lavers, Ralph, Vitart (2018a)

Predicting AR Events

Considering “Weather” Lead Times

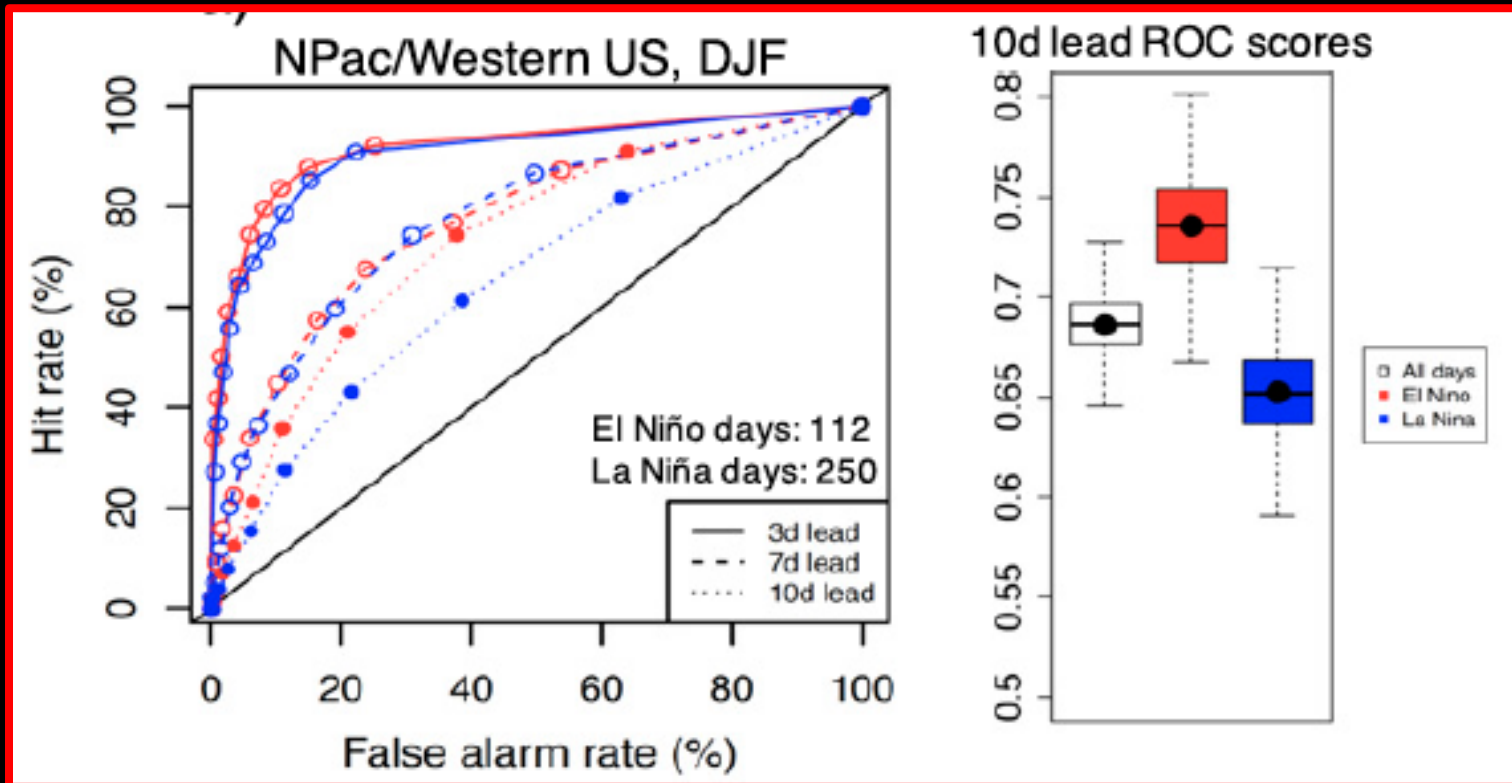
Decision Support Tradeoffs



Predicting AR Events

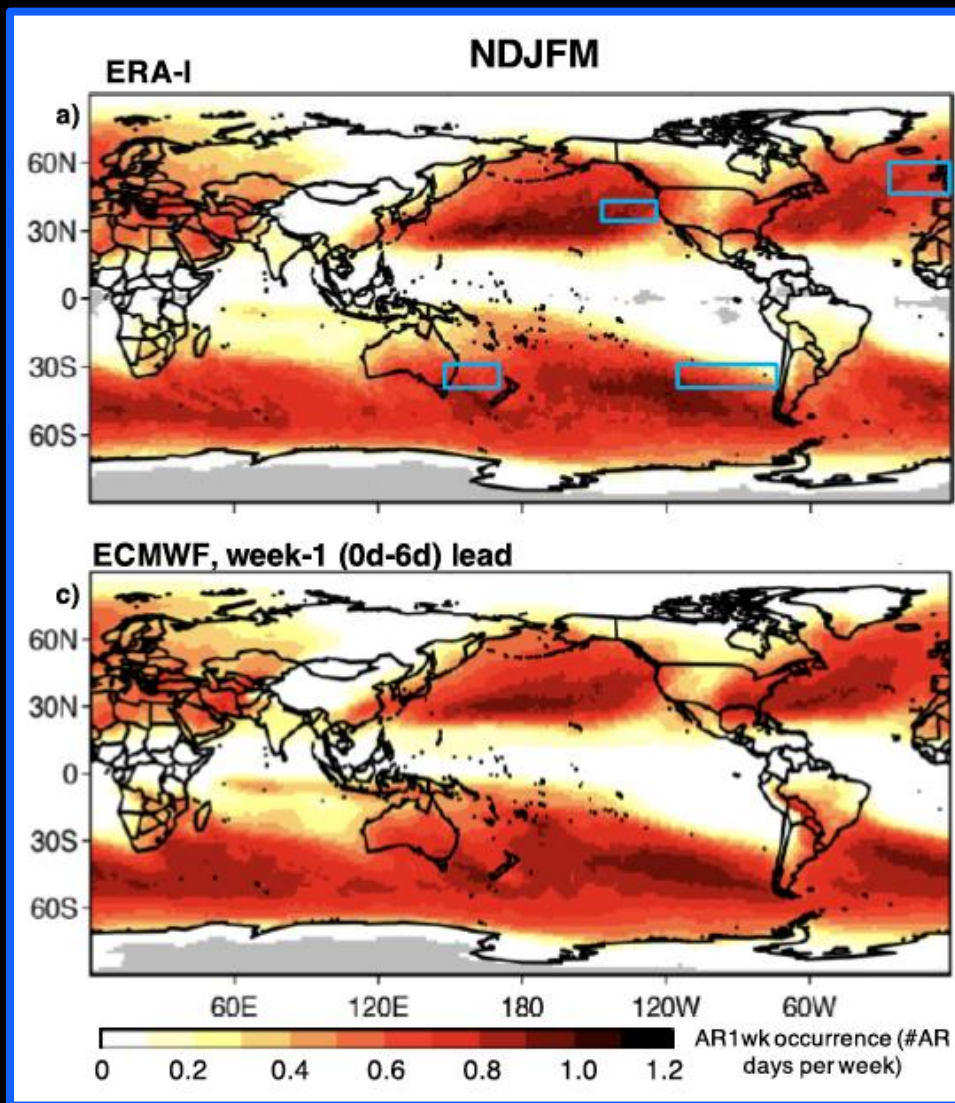
Considering “Weather” Lead Times

ENSO Modulation: Better Skill During El Nino



Predicting AR Activity

Considering S2S Lead Times



“Observations”

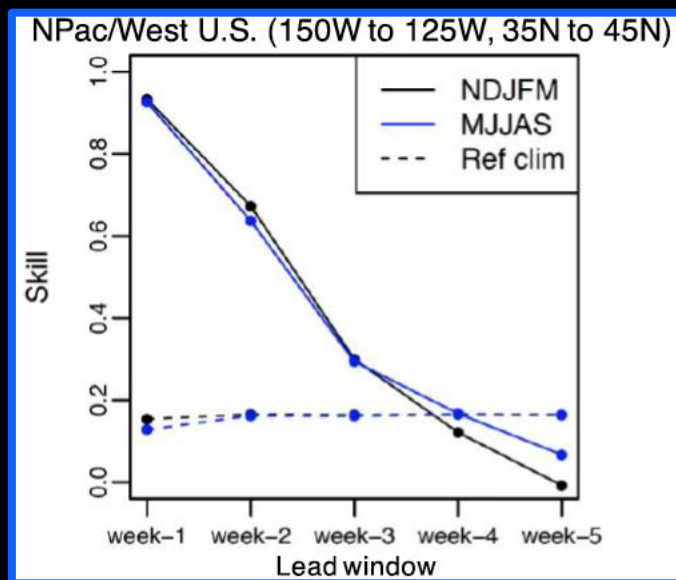
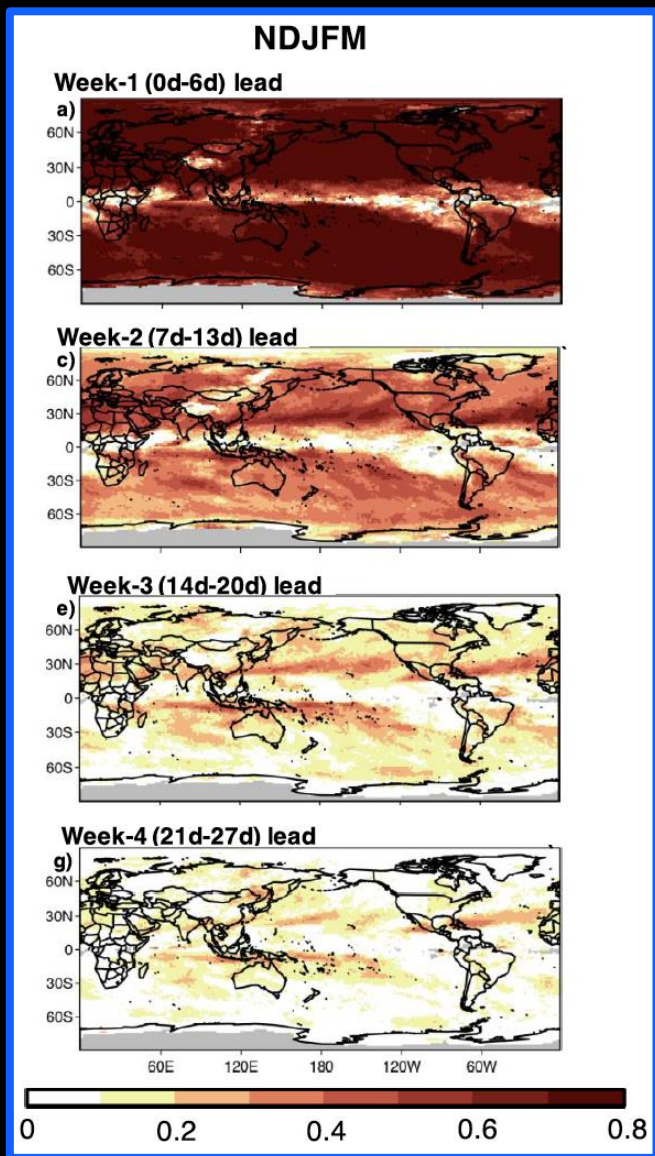
**For lead times > 2 weeks
Consider the
of ARs in a week**

Model - 1 Week Forecasts

Predicting AR Activity

Considering S2S Lead Times

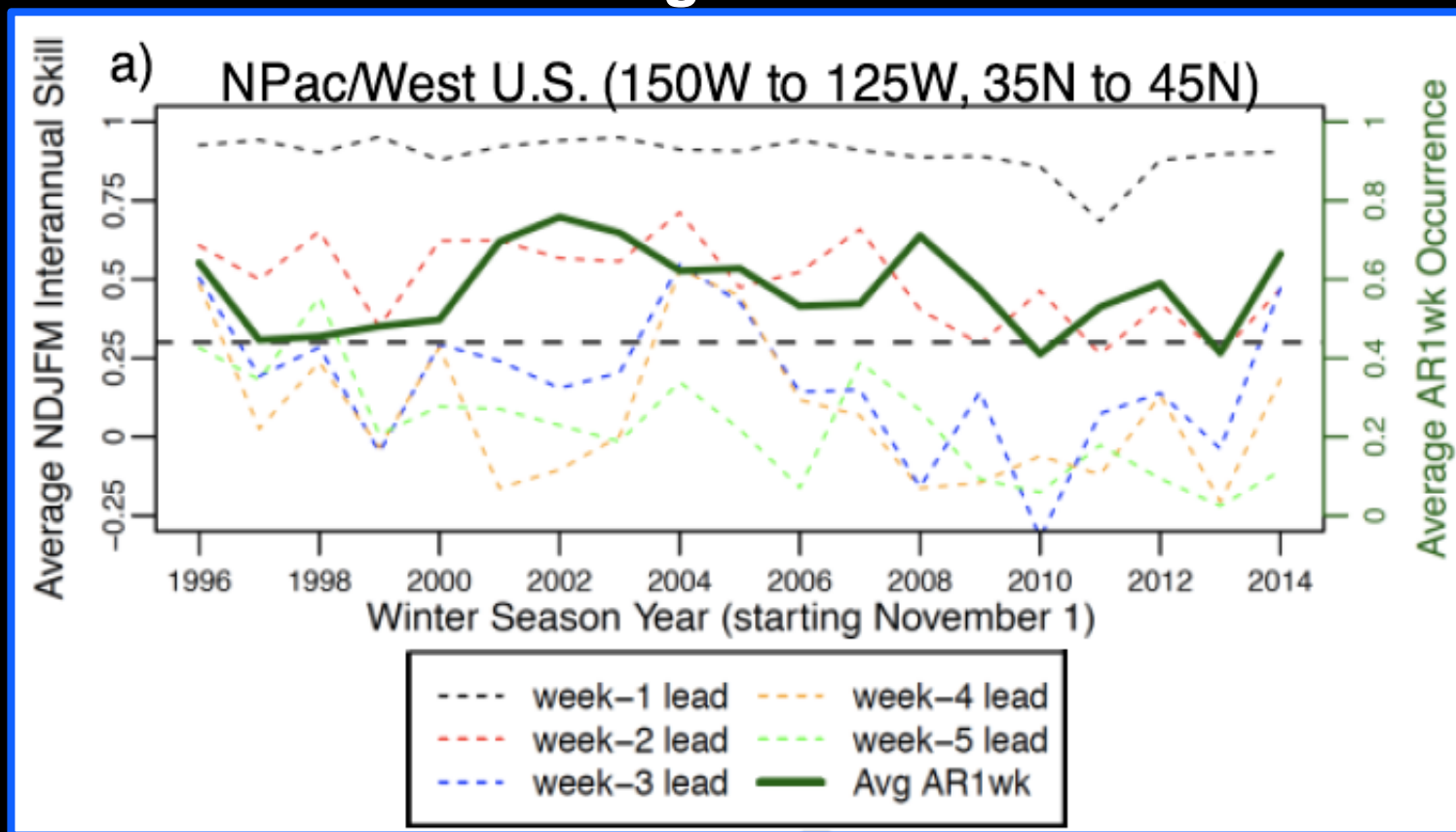
Some Model Prediction Skill over Western U.S. for Weeks 1, 2 and 3.



Predicting AR Activity

Considering S2S Lead Times

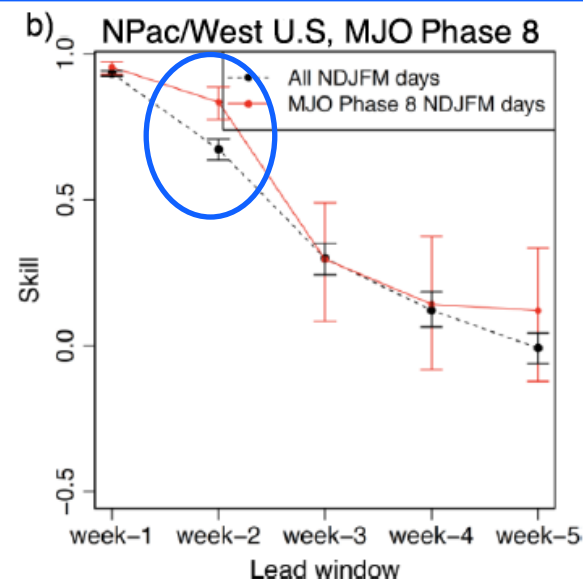
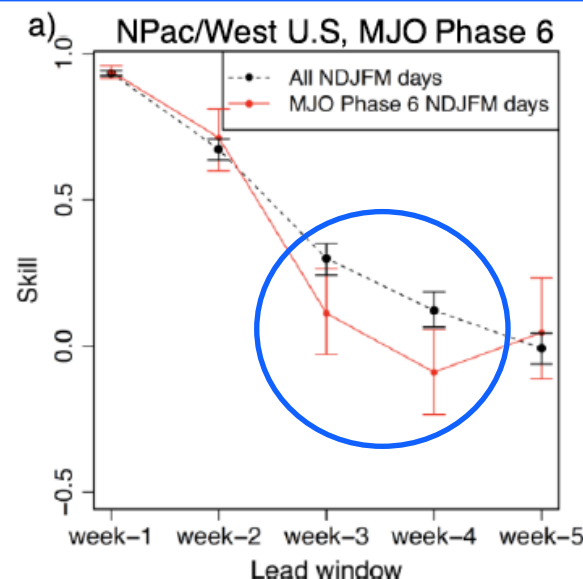
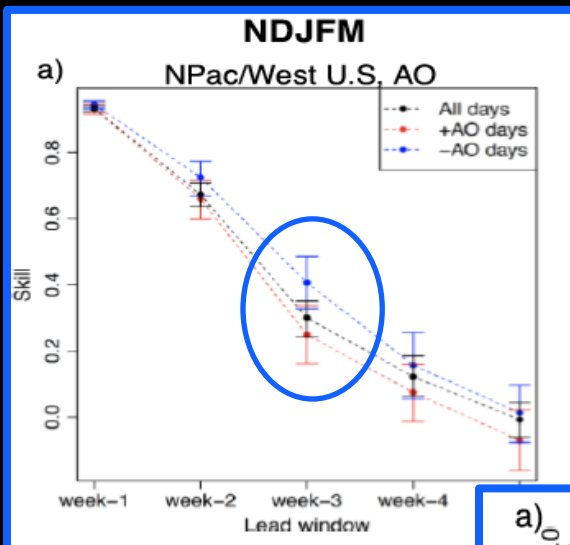
Intermittent Long-Lead Forecast Skill



Predicting AR Activity

Considering S2S Lead Times

Arctic Oscillation (AO) and Madden Julian Oscillation (MJO) Modulates S2S AR Forecast Skill



DeFlorio, Waliser, Guan,
Lavers, Ralph, Vitart
(2018b)

Experimental Synoptic and Subseasonal AR Forecasting for Winter 2017-18 and 2018-19



Duane Waliser
Mike DeFlorio
Alex Goodman



Bin Guan



Marty Ralph
Aneesh Subramanian
Sasha Gershunov



Frédéric Vitart



Jay
Cordeira



Jeanine Jones

Example of Experimental Atmospheric River Event Forecast*

Issued on Thursday, April 5, 2018

Contents:

Week 1 and 2: “Weather” - Typical presentation of US west coast weather/precipitation forecast over lead times of 1 to 14 days considering only the likelihood of an atmospheric river (AR) occurring on a given forecast day. *Novelty – a weather forecast presented only in terms of AR likelihood.*

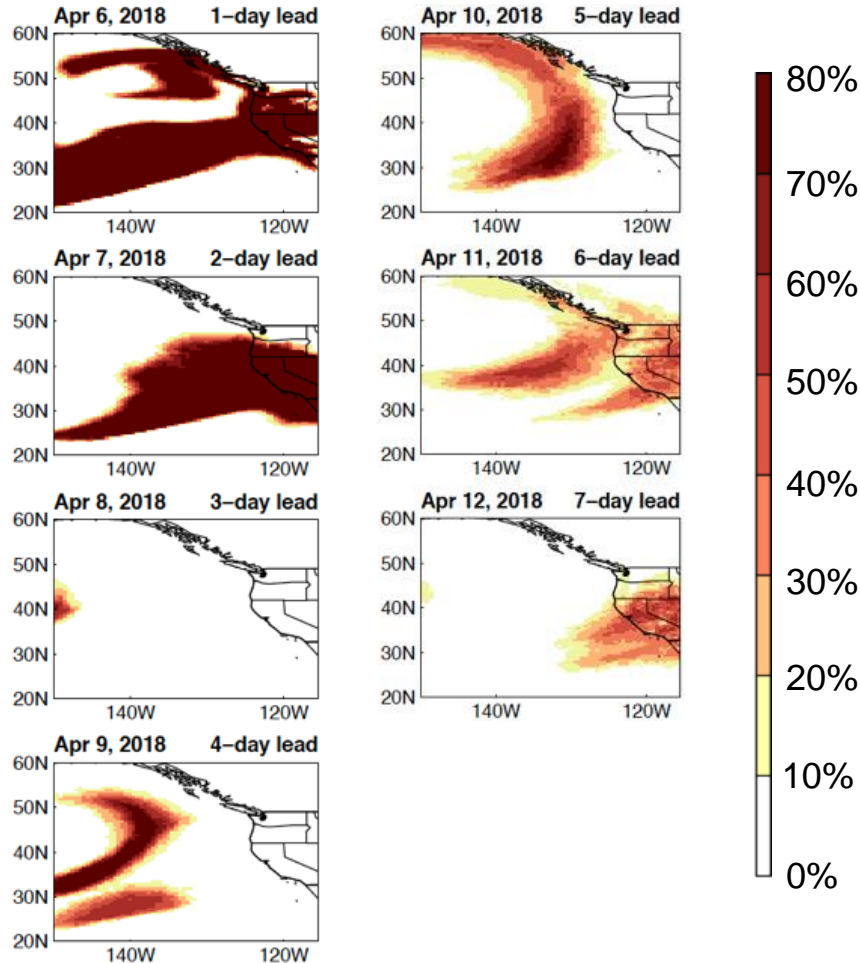
**This is an experimental activity for the 2017-18 and 2018-19 winters. Methodologies and hindcast skill are documented in DeFlorio et al. (2018a,b). Further validation of the real-time forecast results is required and underway. This phase of the research includes gathering stakeholder input on the presentation of information – feedback is welcome.*

POC: Michael J. DeFlorio (michael.deflorio@jpl.nasa.gov)



EXPERIMENTAL AR FORECAST

April 5, 2018 forecast: probability of AR occurrence during week-1



Week-1: ECMWF (1-day to 7-day lead)

Experimental AR forecast issued on Thursday, April 5, 2018 by M. DeFlorio, A. Goodman, D. Waliser, B. Guan, A. Subramanian, and M. Ralph using 51-member real-time **ECMWF** data for an Experimental AR Forecasting Research Activity sponsored by California DWR



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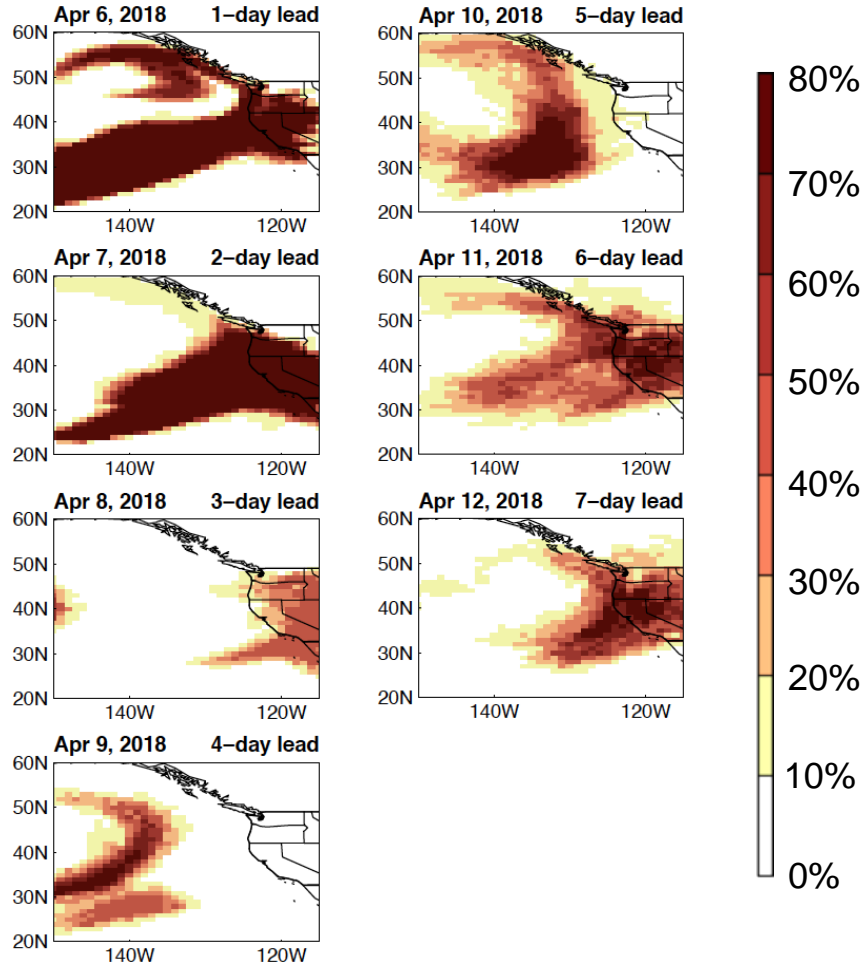


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Contact: M. DeFlorio
(michael.deflorio@jpl.nasa.gov)

EXPERIMENTAL AR FORECAST

April 5, 2018 forecast: probability of AR occurrence during week-1



Week-1: ECCC (1-day to 7-day lead)

Experimental AR forecast issued on Thursday, April 5, 2018 by M. DeFlorio, A. Goodman, D. Waliser, B. Guan, A. Subramanian, and M. Ralph using 21-member real-time ECCC data for an **Experimental AR Forecasting Research Activity** sponsored by **California DWR**



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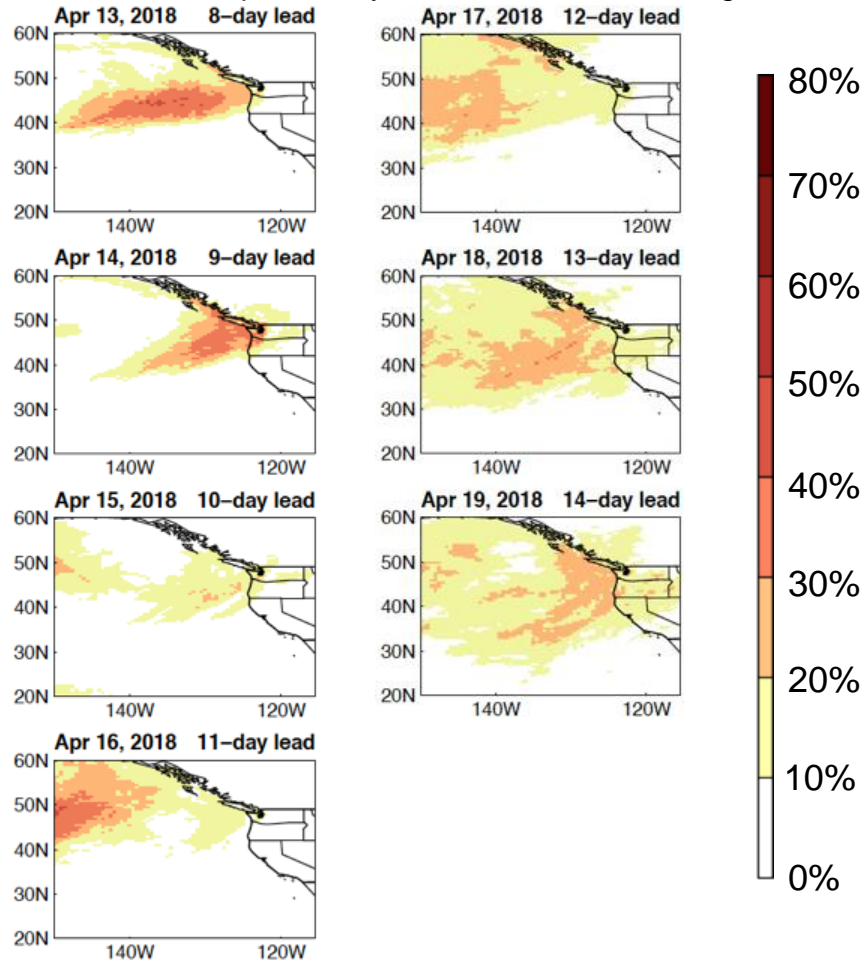


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EXPERIMENTAL AR FORECAST

April 5, 2018 forecast: probability of AR occurrence during week-1



Week-2: ECMWF (8-day to 14-day lead)

Experimental AR forecast issued on Thursday, April 5, 2018 by M. DeFlorio, A. Goodman, D. Waliser, B. Guan, A. Subramanian, and M. Ralph using 51-member real-time **ECMWF** data for an Experimental AR Forecasting Research Activity sponsored by California DWR



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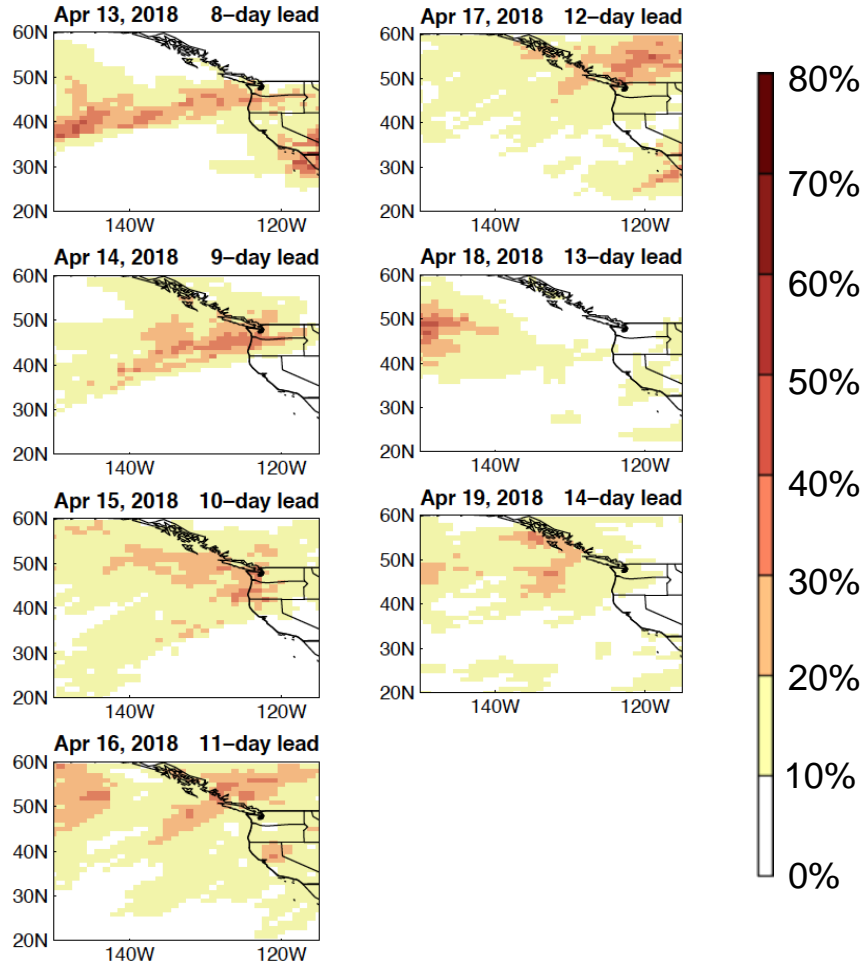


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EXPERIMENTAL AR FORECAST

April 5, 2018 forecast: probability of AR occurrence during week-1



Week-2: ECCC (8-day to 14-day lead)

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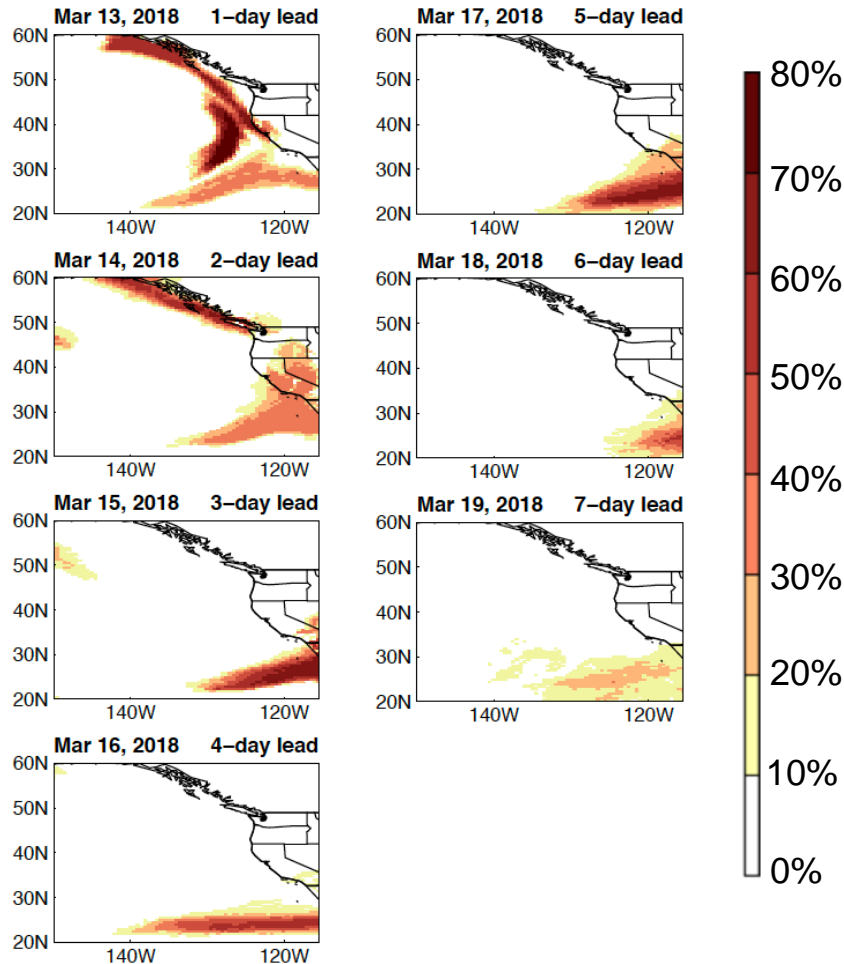


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EXPERIMENTAL AR FORECAST

March 12, 2018 forecast: probability of AR occurrence during week-1



Week-1: ECMWF (1-day to 7-day lead)

Experimental AR forecast issued on Monday, March 12, 2018 by M. DeFlorio, A. Goodman, D. Waliser, B. Guan, A. Subramanian, and M. Ralph using 51-member real-time ECMWF data for an Experimental AR Forecasting Research Activity sponsored by California DWR



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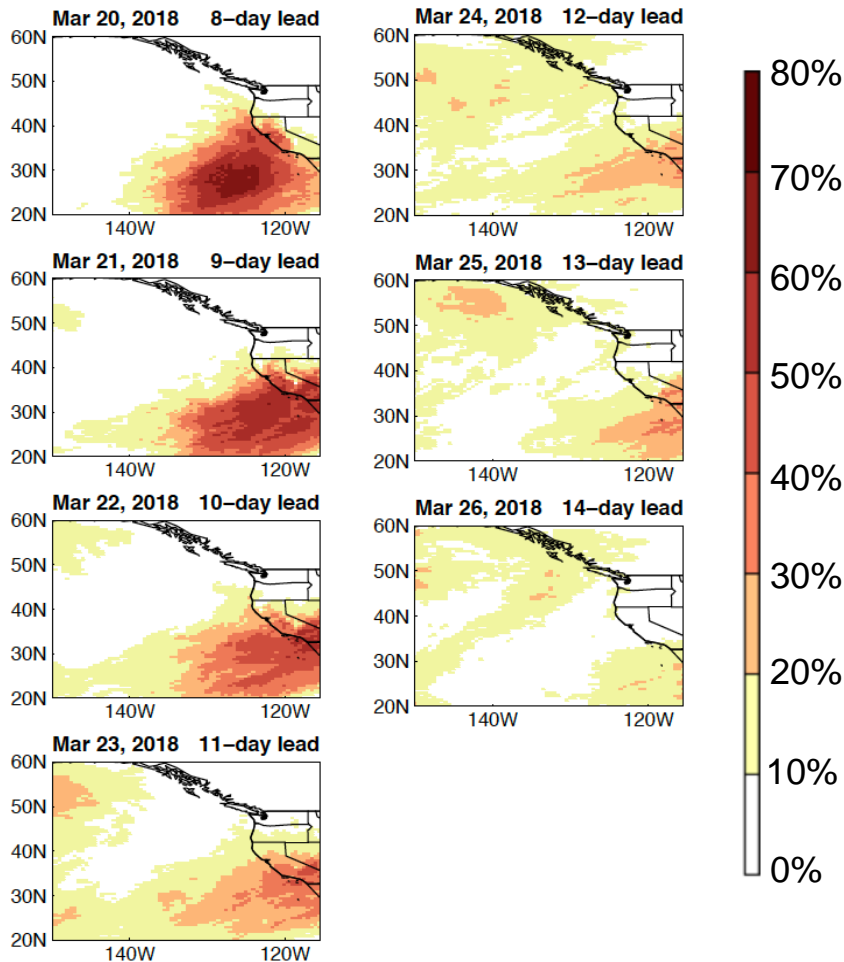


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EXPERIMENTAL AR FORECAST

March 12, 2018 forecast: probability of AR occurrence during week-2



Week-2: ECMWF (8-day to 14-day lead)

Experimental AR forecast issued on Monday, March 12, 2018 by M. DeFlorio, A. Goodman, D. Waliser, B. Guan, A. Subramanian, and M. Ralph using 51-member real-time ECMWF data for an Experimental AR Forecasting Research Activity sponsored by California DWR



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Example of Experimental Multi-model S2S Atmospheric River Activity Forecast*

Issued on Thursday, May 10, 2018

Contents:

Week 3: “Subseasonal” - US west coast weather/precipitation forecast for week 3 considering the number of atmospheric river days predicted to occur in the given forecast week.

Novelty – an S2S forecast presented only in terms of AR likelihood - specifically for week 3, an extended/long-range or “subseasonal” prediction

NCEP (National Centers for Environmental Systems) forecast system

ECMWF (European Centre for Medium-Range Weather Forecasts) forecast system

ECCC (Environment and Climate Change Canada) forecast system



**This is an experimental activity for the 2017-18 and 2018-19 winters. Methodologies and hindcast skill are documented in DeFlorio et al. (2018a,b). Further validation of the real-time forecast results is required and underway. This phase of the research includes gathering stakeholder input on the presentation of information – feedback is welcome.*

POC: Michael J. DeFlorio (michael.deflorio@jpl.nasa.gov)

EXPERIMENTAL S2S AR FORECAST

May 10, 2018 forecast: number of AR days during week-3

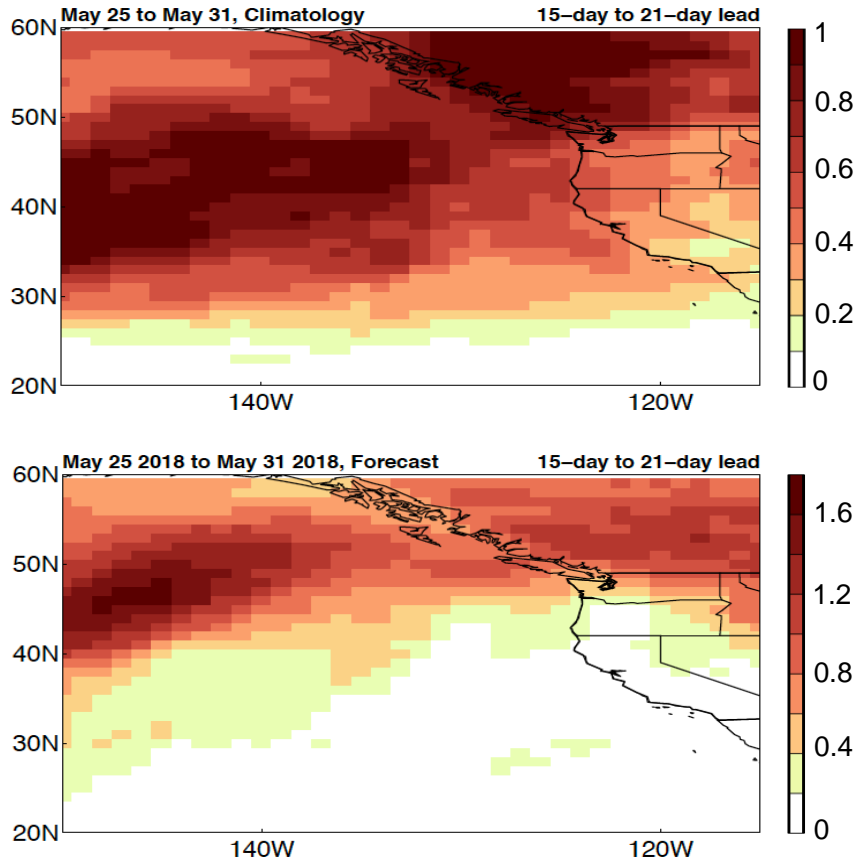


Week-3

(Combined 15-day to 21-day lead)

Top row: **hindcast climatology** (NCEP 1999-2010 data)
Bottom row: **real-time forecast** (NCEP 16-member ensemble)

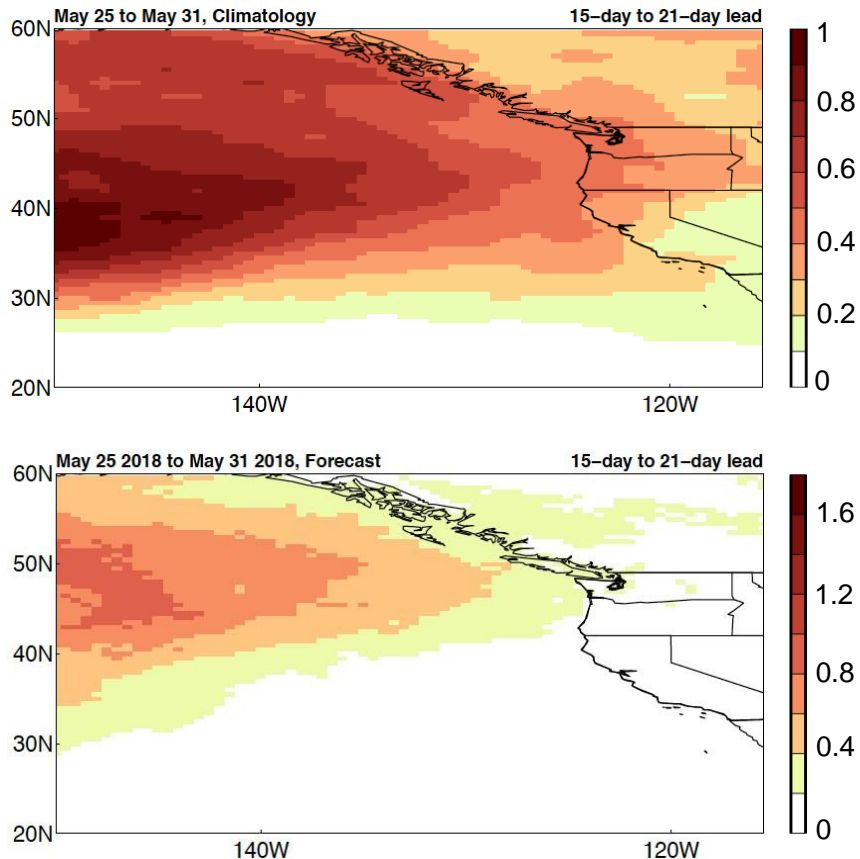
Experimental AR forecast issued on Thursday, May 10, 2018 by M. DeFlorio, A. Goodman, D. Waliser, B. Guan, A. Subramanian, and M. Ralph using 16-member real-time NCEP data for an **Experimental AR Forecasting Research Activity** sponsored by **California DWR**



Contact: M. DeFlorio
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EXPERIMENTAL S2S AR FORECAST

May 10, 2018 forecast: number of AR days during week-3



ECMWF



Week-3

(Combined 15-day to 21-day lead)

Top row: **hindcast climatology** (ECMWF 1996-2015 data)
Bottom row: **real-time forecast** (ECMWF 51-member ensemble)

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EXPERIMENTAL S2S AR FORECAST

May 10, 2018 forecast: number of AR days during week-3

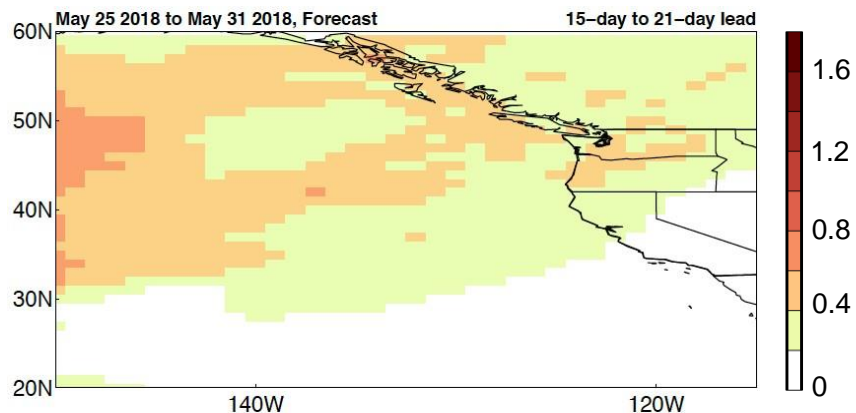
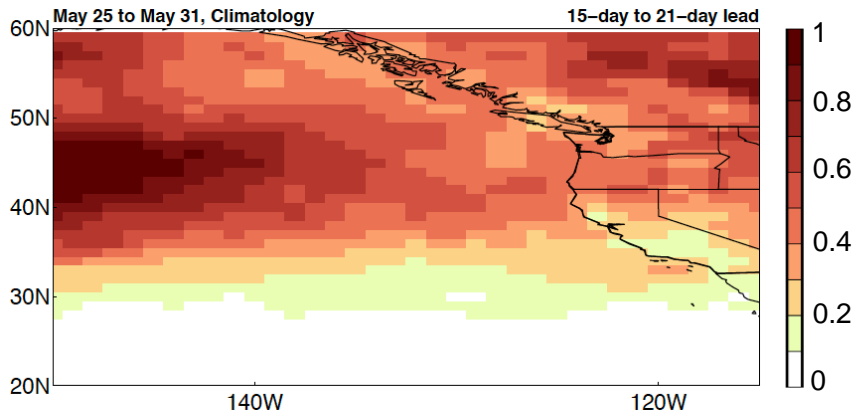


Week-3

(Combined 15-day to 21-day lead)

Top row: **hindcast climatology** (ECCC 1995-2014 data)
Bottom row: **real-time forecast** (ECCC 21-member ensemble)

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EXPERIMENTAL S2S AR FORECAST

May 10, 2018 forecast: number of AR days during week-3

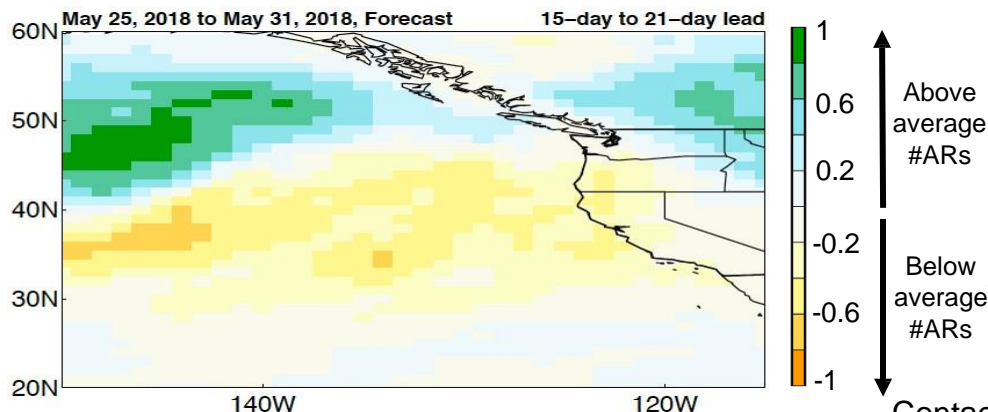
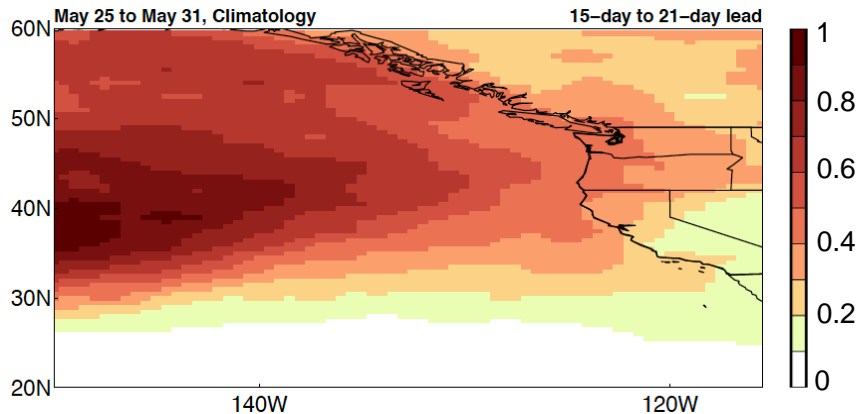


Week-3

(Combined 15-day to 21-day lead)

Top row: **hindcast climatology** (NCEP 1999-2010 data)

Bottom row: **real-time forecast** (NCEP 16-member ensemble)



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EXPERIMENTAL S2S AR FORECAST

May 10, 2018 forecast: number of AR days during week-3

ECMWF



Week-3

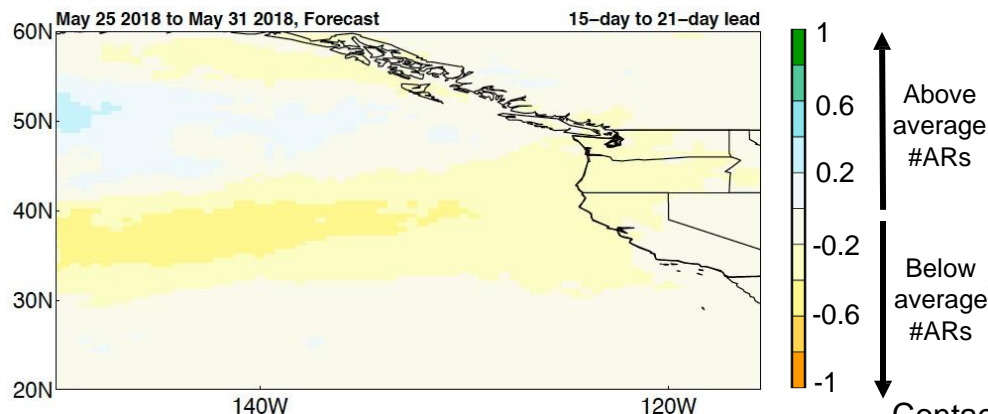
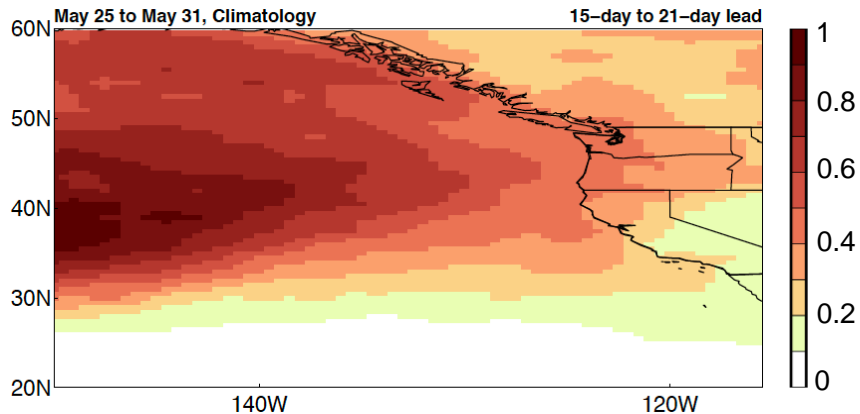
(Combined 15-day to 21-day lead)

Top row: **hindcast climatology** (ECMWF 1996-2015 data)

Bottom row: **real-time forecast minus climatology** (ECMWF 51-member ensemble)

Experimental AR forecast issued on Thursday, May 10, 2018 by M. DeFlorio, A. Goodman, D. Waliser, B. Guan, A. Subramanian, and M. Ralph using 51-member real-time ECMWF data for an

Experimental AR Forecasting Research Activity
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Above
average
#ARs

Below
average
#ARs



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EXPERIMENTAL S2S AR FORECAST

May 10, 2018 forecast: number of AR days during week-3

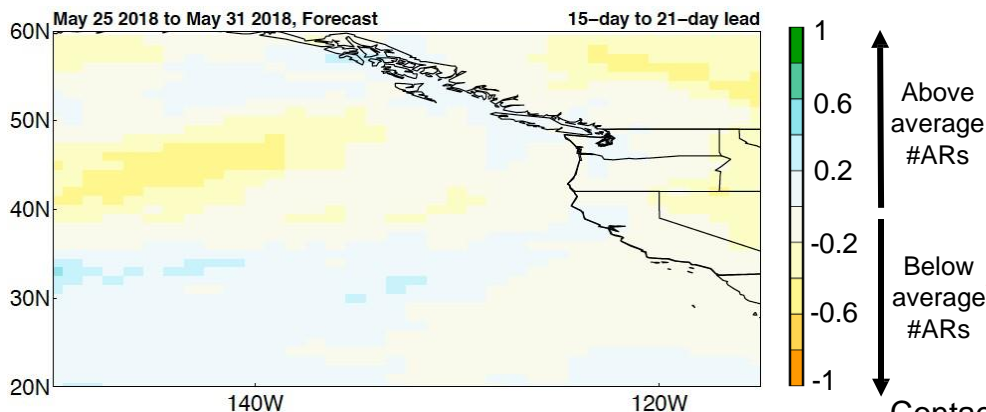
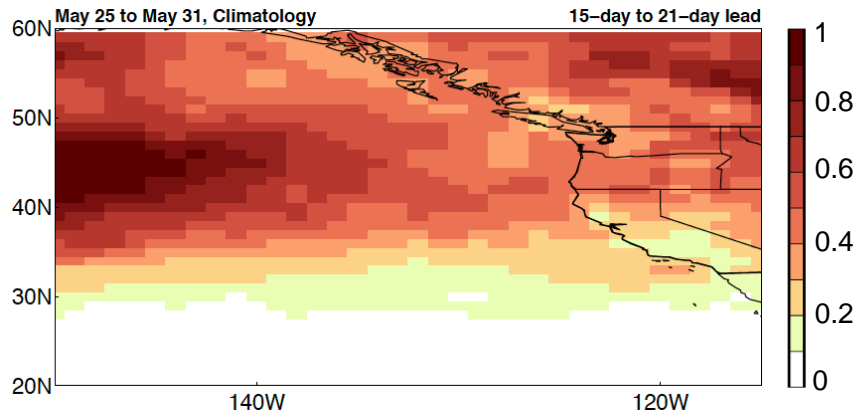


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Bottom row: **real-time forecast** (ECCC 21-member ensemble)



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Above
average
#ARs

Below
average
#ARs



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California Department of Water Resources is sponsoring research to explore subseasonal forecasts of AR activity.

The research is a collaboration between JPL and the University of California.

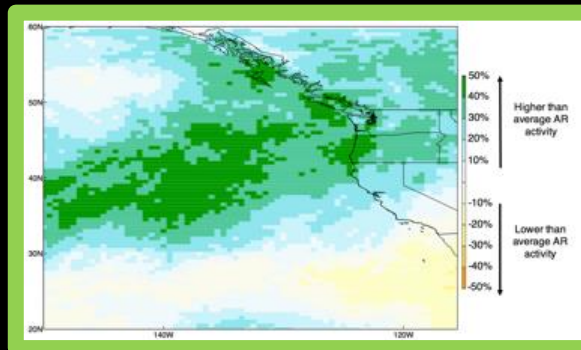
Near-term focus is on week-3 forecasts for AR activity for the winters of 2017-18 and 2018-19.



Flood image credit : NOAA

Forecast

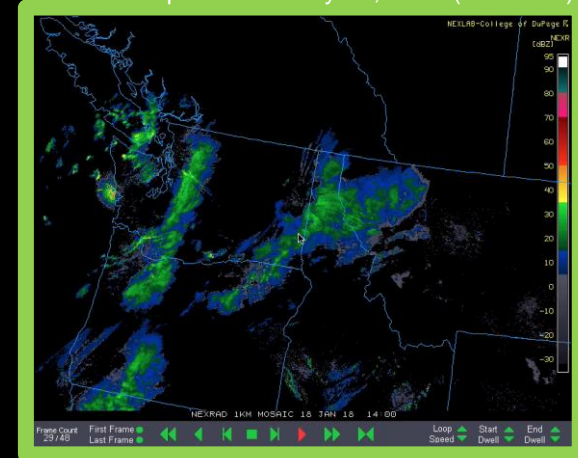
January 1, 2018 forecast for January 16-22



Experimental forecast suggests above average AR activity during 3rd week after forecast date

Observations

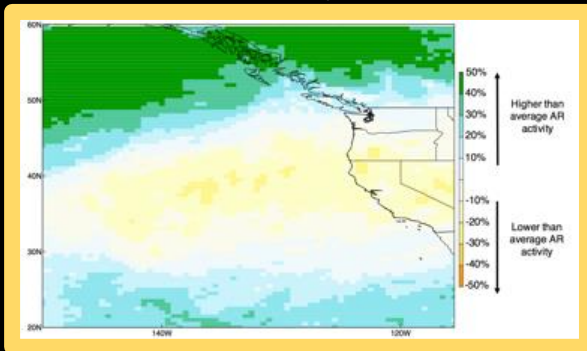
Radar loop from January 18, 2018 (AR event)



City	Cumulative Precipitation (inches)
Seattle	1.37
Portland	0.34
Medford	0.53
Crescent City	1.61

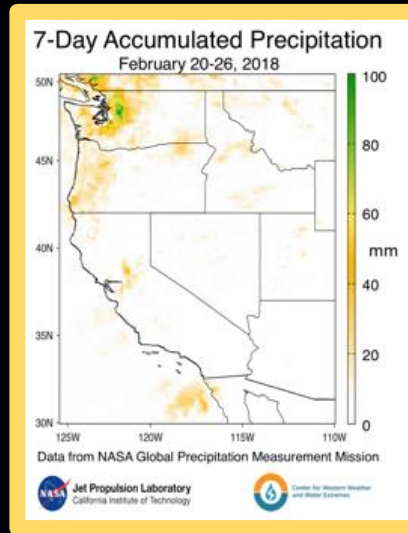
Forecast

February 5, 2018 forecast
for February 20-26



Experimental forecast
suggests below average AR
activity during 3rd week after
forecast date

Observations



City	Cumulative Precipitation (inches)
Oroville Dam	0.01
San Francisco	0.01
Los Angeles	Trace
San Diego	0.03



Experimental results to date show some
promise for intermittently being able to
deliver skillful longer-lead (e.g. 3-week)
forecast information for AR activity along
the west coast of the US.

Such forecasts will be valuable for flood,
hazard and water decision support.



Drought image credit : LA Times

Review of Winter 2017-2018 Activities & Winter 2018-2019 Goals

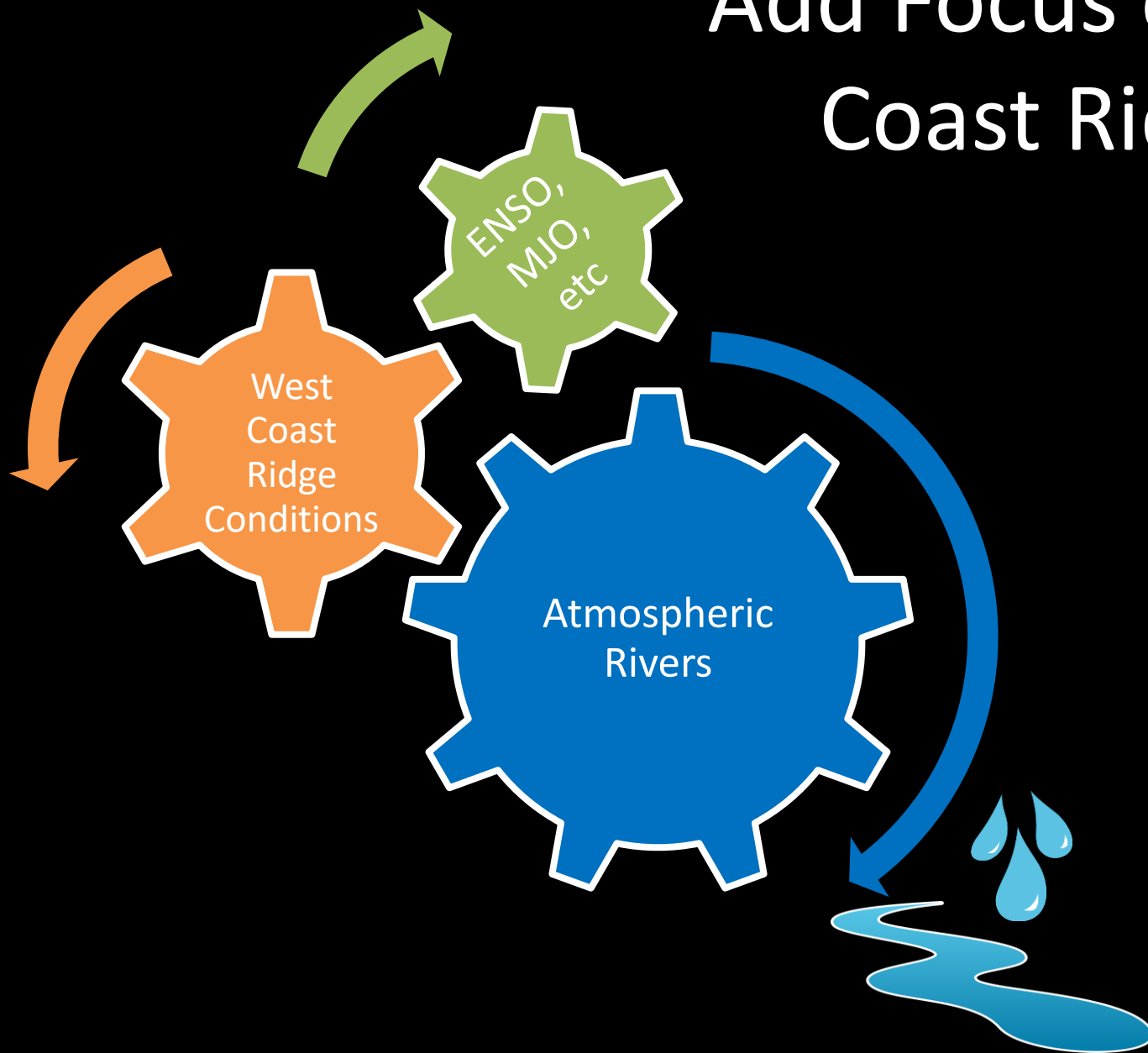
Winter 2017-2018: what we did

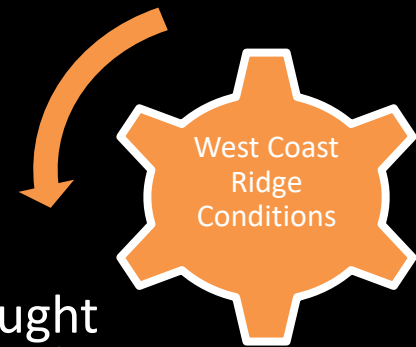
- Create an automated pipeline to:
 - Calculate IVT fields from ECMWF, ECCC, and NCEP forecast systems
 - Twice-a-week for ECMWF
 - Weekly or bi-monthly for ECCC, NCEP
 - Apply Guan and Waliser 2015 AR detection algorithm to IVT fields
 - Calculate AR1wk forecasts for week-1, week-2, and week-3 lead windows
- Developing strategy for calculating verification statistics - compare to hindcast skill benchmarks [DeFlorio et al. 2018b]
- Disseminate experimental forecasts and solicit feedback from sponsor, CW3E S2S weekly teleconferences, meetings, etc.

Winter 2018-2019: what we'll do

- Continue to produce near-real time experimental week-3 AR1wk occurrence forecasts for ECMWF, ECCC, and NCEP forecast systems, stratified by mean AR intensity (>250 kg/ms, >500 kg/ms)
- Calculate verification statistics for week-3 forecasts in real-time during winter 2018-2019
- Refine forecast products/metrics with stakeholders
- Turn useful parts of product pipeline into CW3E web products

Add Focus on West Coast Ridging





Project Goal

Improve our predictive capabilities of U.S. west coast drought conditions on S2S for improved water resource management.

Project Objectives

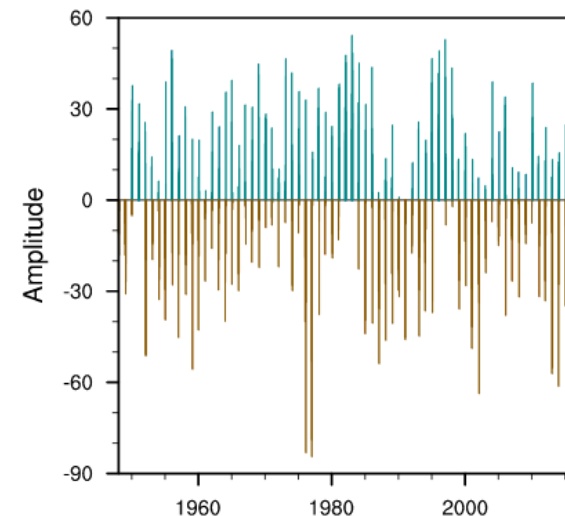
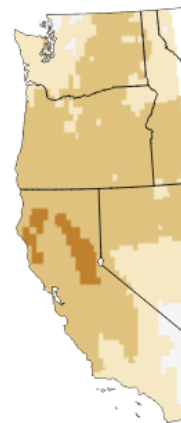
- Develop a **US West Coast Drought Ridging Index (WC-DRI)** that is closely tied to U.S. west coast drought
- Quantify subseasonal to seasonal (S2S) **prediction skill** of atmospheric ridging conditions relevant to western U.S. water resources in hindcasts of current generation weather/climate models.
- Identify sources of S2S **predictability** of ridging conditions relevant to U.S. west coast water resources and quantify their predictability characteristics.
- Develop and implement real-time **operational S2S monitoring and prediction** products related to ridging relevant to U.S. west coast water resources.

Standard Precipitation Index SPI over Western US

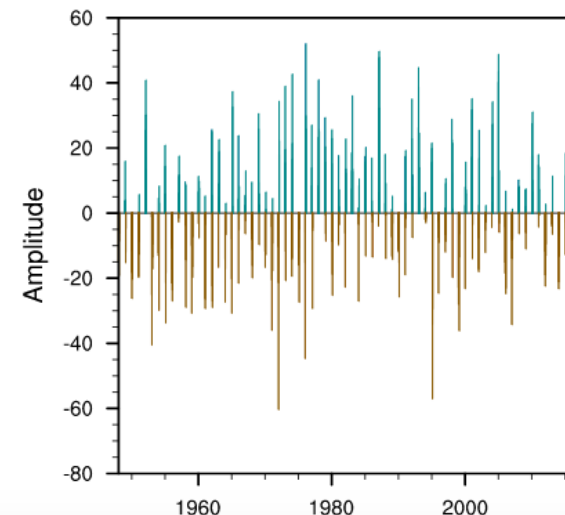
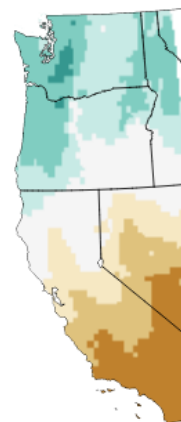
First 2 EOFs of SPI-3 in ONDJFM

- CPC-Unified gauge-based precipitation (0.25 deg.) 1950-2016
- *Drought definition may be modified in terms of **region** (e.g. Colorado vs. Western USA vs. CA only) to improve immediate relevance for stakeholder*

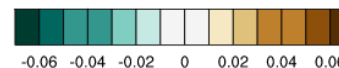
EOF1 39.3%



EOF2 20.6%

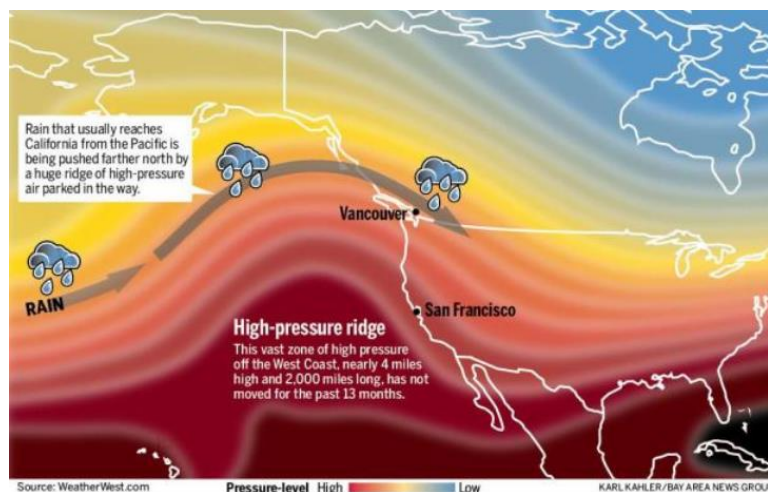


EOF3 5.2%

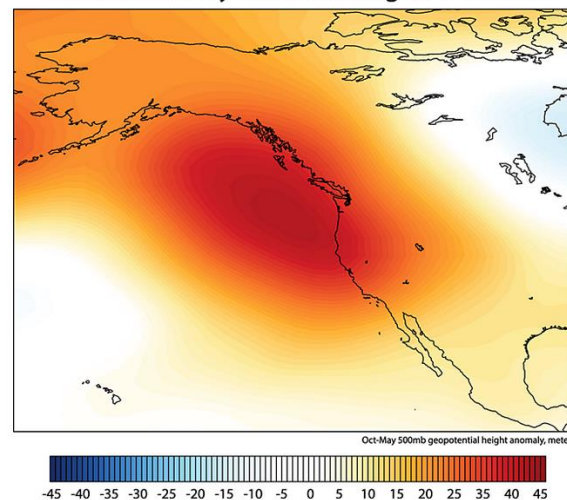


Defining a west coast ridge index:

‘Ridging’ is a general term describing the pattern of relatively high atmospheric pressure that influences the storm track.



500 hPa geopotential height anomaly
The Ridiculously Resilient Ridge, 2012-2015

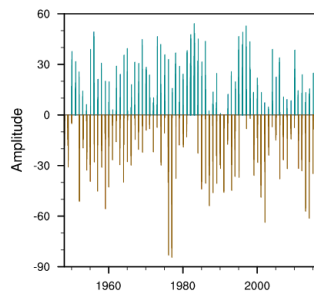


Swain (2015, GRL)

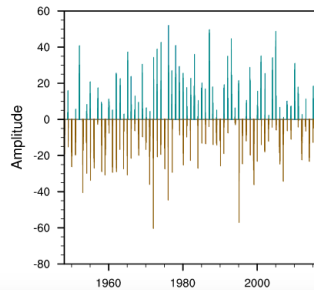
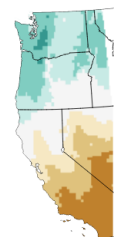
Candidates for a West Coast (drought) Ridge Index (WC-DRI) relevant to U.S. west coast Drought

3 month gpt. Height anomaly (500hPa)

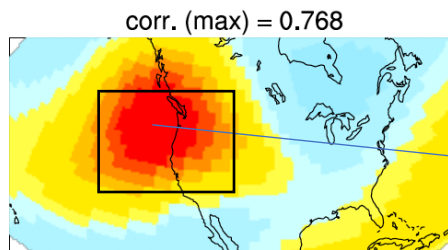
EOF1 39.3%



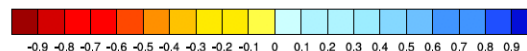
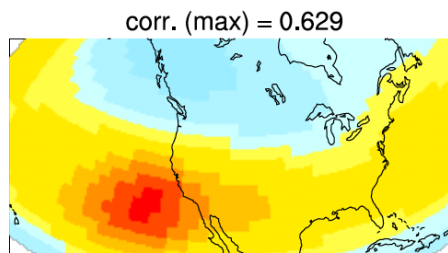
EOF2 20.6%



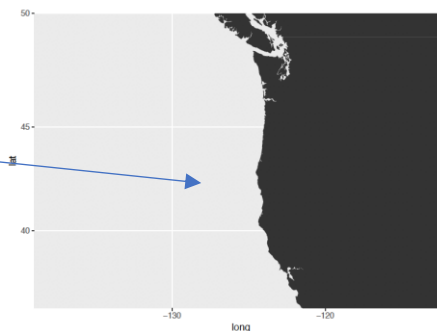
EOF1 SPI-3



EOF2 SPI-3



Pearson



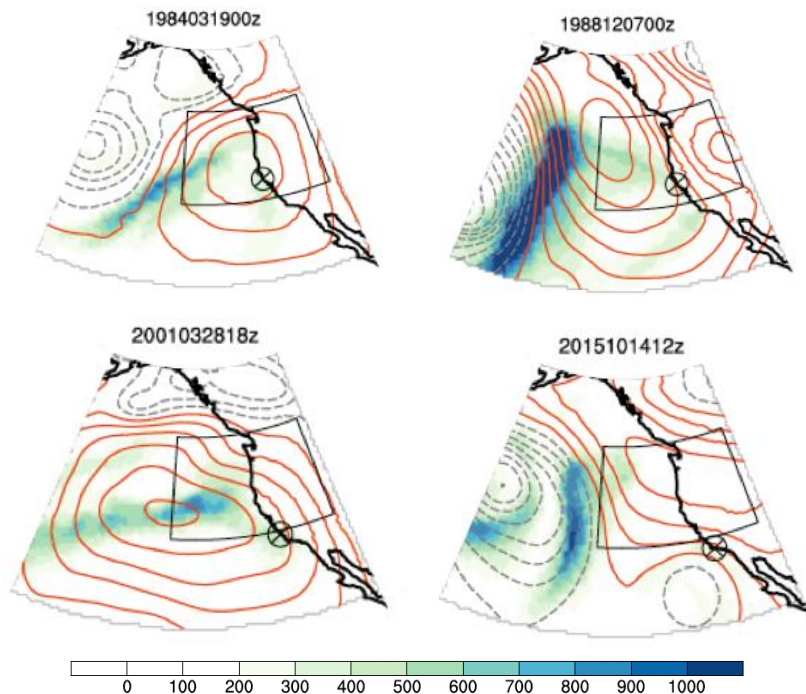
Area to compute ridge index over (WC-DRI)

- Area-weighted 500 hPa height anomaly
- % of > 0 anomalies for size/consistency

Based on MERRA-2 6-hourly data

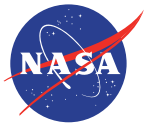
WC-DRI as an AR blocker

Red contours = positive geopotential hgt.; Shading = IVT; Cross = AR landfall location



When the WC-DRI >100m very few ARs make landfall over CA (16 in total, or <1% of all ARs between 1980-2016)

Those that do are very weak IVT at land fall (and even weaker precip.) and the AR has still been strongly affected by ridge



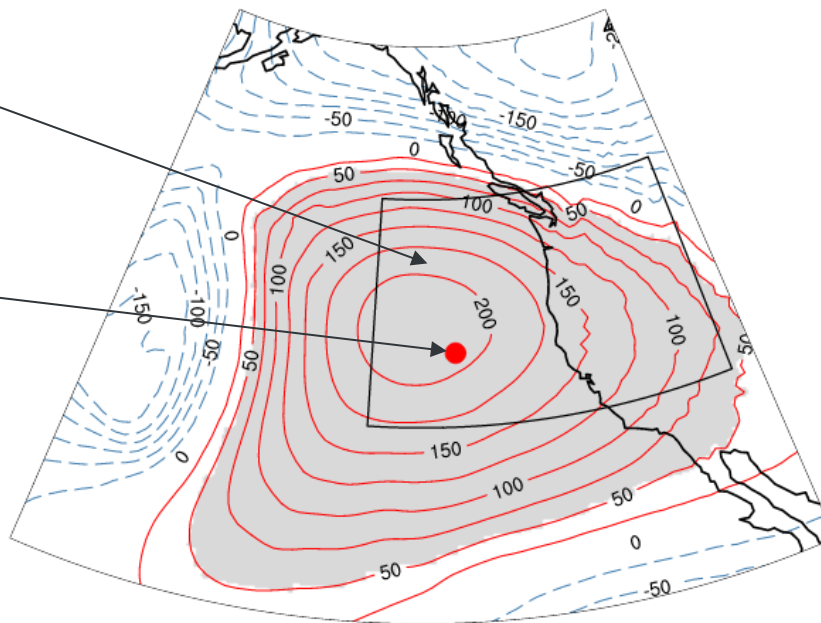
Tracking & Quantifying Ridges

Case 1: Big ridge in position

Gray shading
indicates ridge
object ($> +50m$
contour)

Ridge centroid
(red dot) lies in
critical boxed
region

MERRA-2 6hrly ridge index tracker: 1983011012

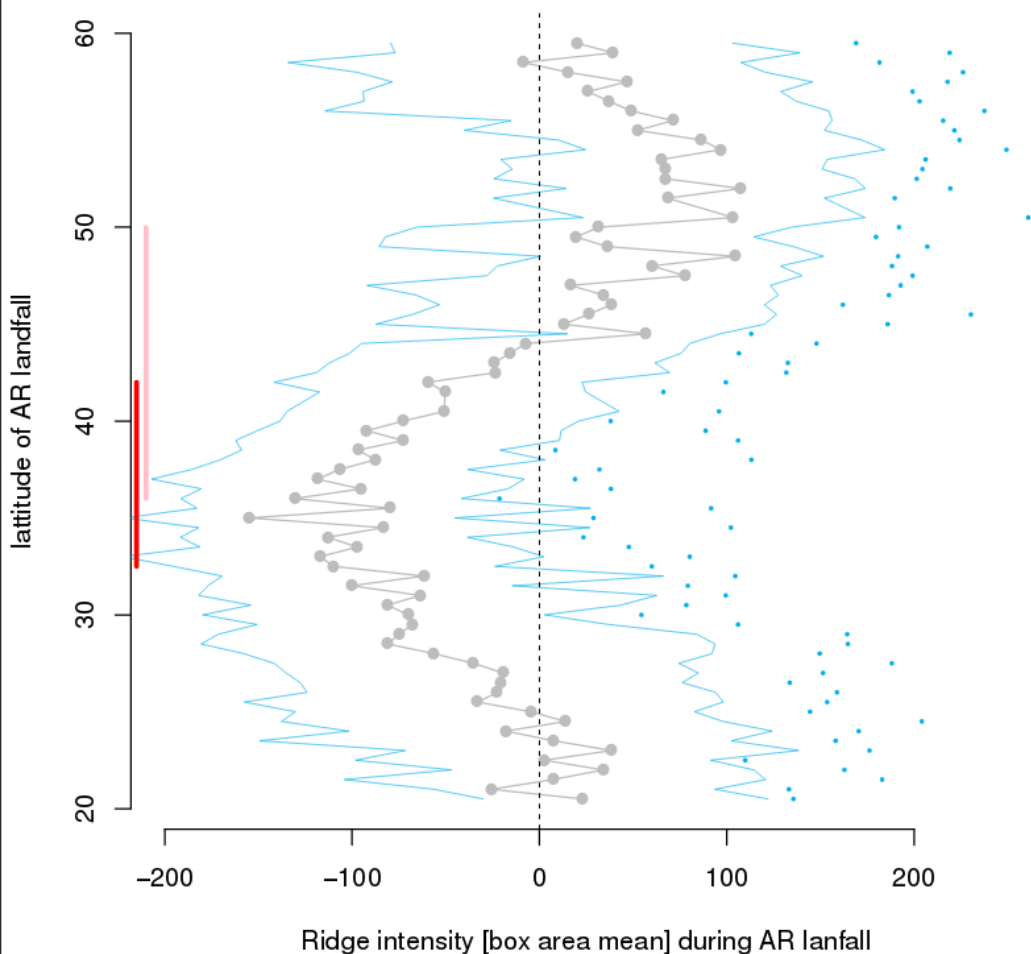


WCRI = 136.53m
WCRI (% positive) = 89.95%

Ridge position critical? YES
Persistence (n. 6hr steps) = 4
n. of ridge objects = 1
Area (km^2) = $7.36705e+06$
max lat. span (km) = 2650.47
max lon. span (km) = 3187.8

Stats
returned
from
tracker

all 6hrly LF AR occurrences between 20–60 deg.N
1980–2015 in MERRA2 | n= 13078



Ridge Effects on ARs

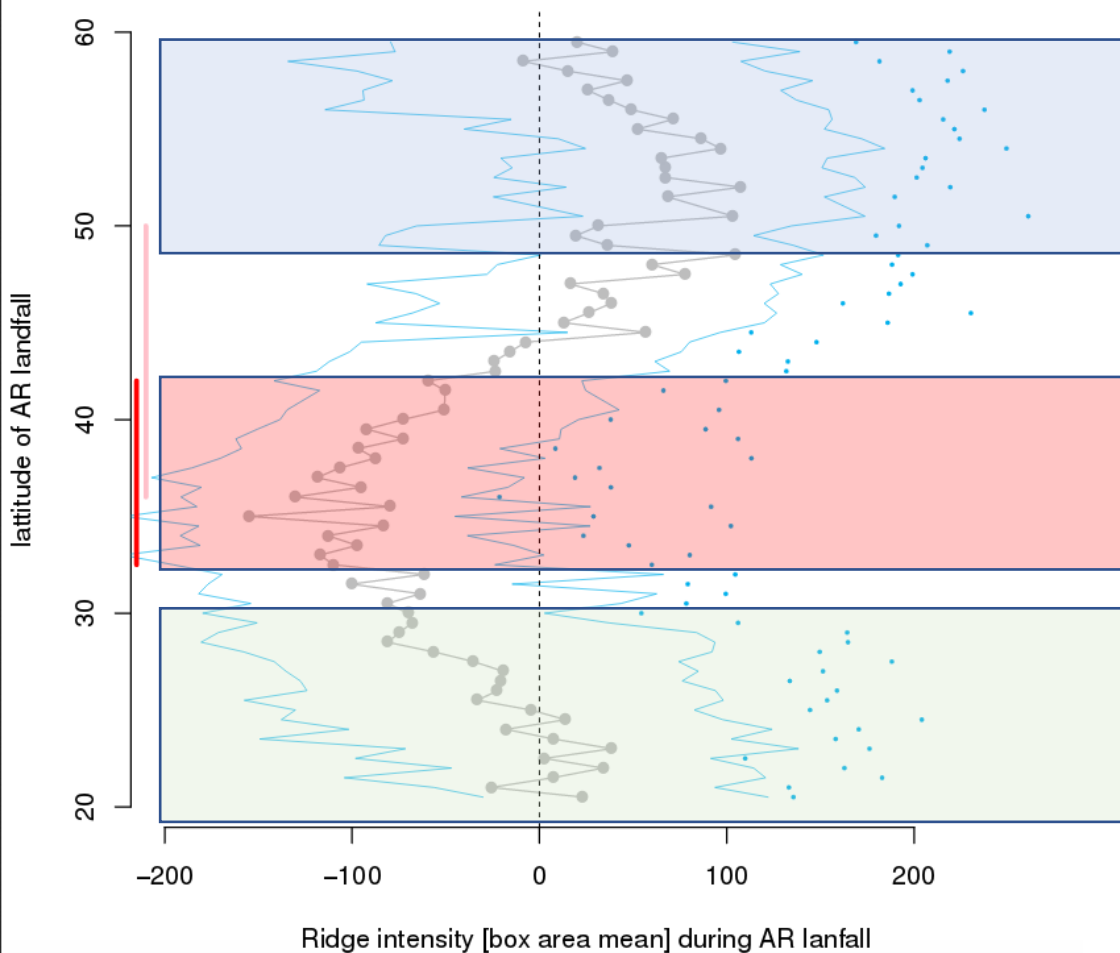
- grey line = median
- blue lines = 10/90 percentile range
- blue dots = 99th percentile

Over 90% of ARs that make landfall over CA (red region) are associated with negative Ridge intensities

Only a tiny fraction of ARs (<1%) make landfall over CA associated with Ridge intensity $\sim +100\text{m}$

Ridge Effects on ARs

all 6hrly LF AR occurrences between 20–60 deg.N
1980–2015 in MERRA2 | n= 13078



British Columbia =
more ARs under
positive ridge

CA= less ARs
under positive
ridge

Mexico =
unaffected by
positive
ridge?

Summary

- We're leveraging global AR detection algorithm and databases, S2S Predictions (e.g. NCEP, EC, ECMWF), and climate variations to explore/quantify AR prediction capabilities.
- We're quantifying skill in the present-day S2S forecasts for:
 - *Individual AR Events*
 - *Subseasonal AR Activity (# ARs in 1 week)*
- We're exploring prediction and predictability characteristics of U.S. west coast ridging conditions and impacts on ARs
- We're leveraging this research to help develop experimental forecast products relevant to U.S. west coast precipitation & water resource management.