

Predictive Skill and Bias of Synoptic Variables during Extended Extreme Precipitation Events in S2S Project Models

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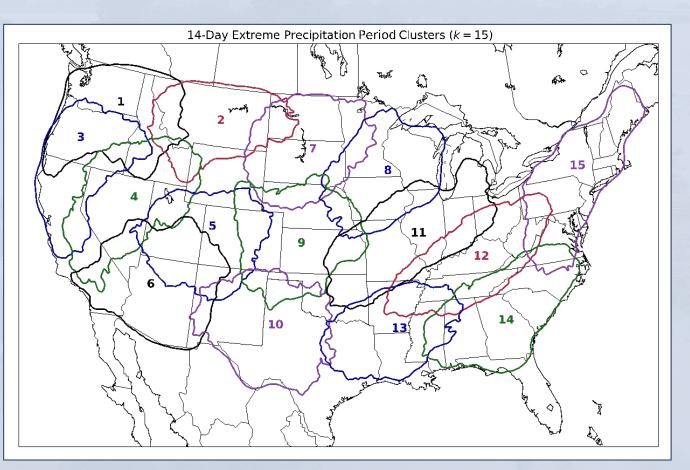
Trinity River May 25th 2015

14-day Extreme Precipitation Period Database

- 14-day extreme period database (Dickinson et al. 2021)
- January 1915 December 2018
- 851 total events
- 15 clusters

Thresholds:

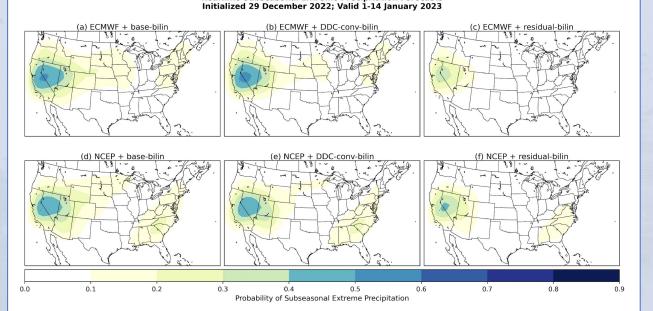
- ✓ Observed 14-day precip ≥ 14-day 99th percentile
- ✓ At least 7 days must receive precip ≥ long-term daily mean
- ✓ Area ≥ 200,000km²



Verification of Synoptic Variables

Why Synoptic Variables?

- Prediction skill of Precipitation in S2S models is low or near zero beyond Week 2 (Pan et al. 2019, Becker et al. 2020, McAfee et al. 2023)
- Geopotential Height and Specific Humidity together were shown to be a potentially skillful predictor (Schroers and Martin 2020)
- Statistically postprocessing model output has been shown to increase skill of precipitation process (Gagne et al. 2014, Hill and Schumacher 2021).



Realtime Subseasonal Extreme Precipitation Forecasts

Figure 6.20: Realtime forecasts initialized on 29 December 2022 from (a-c) ECMWF and (d-f) NCEP subseasonal models valid for 1-14 January 2023. The dynamical forecasts were input into the (a,c) base-bilin, (b,d) DDC-conv-bilin, and (c,e) residual-bilin U-Nets.

Dickinson 2023

Verification

Verifying the ensemble averages of the extreme periods 14-day averaged forecasted anomalies against ERA5 observations.

Models used:

• ECMWF, UKMO, CNRM

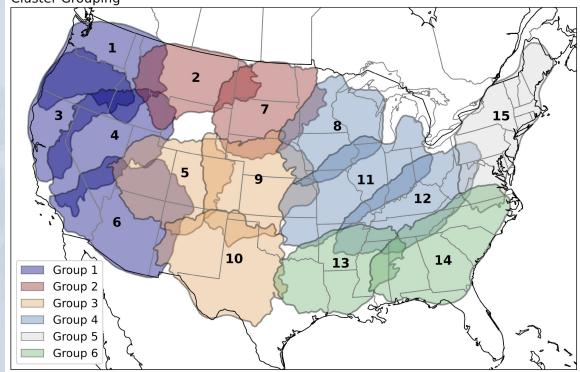
Common Period:

• 1996 – 2014

Variables verified:

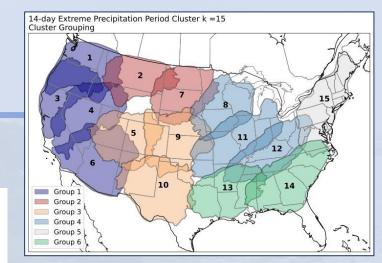
- 500 hPa Geopotential Height (CONUS domain)
- 700 hPa Specific Humidity (group domain)
- 850 hPa Specific Humidity (group domain)

14-day Extreme Precipitation Period Cluster k =15 Cluster Grouping



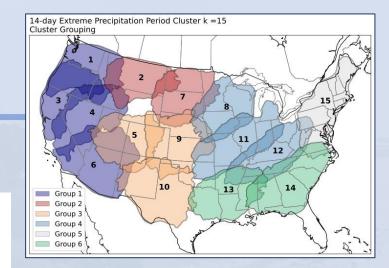
Verification – 500 hPa GPH

	Geopot ECMWF	ential He	ight at 5	500 hPa		икмо					CNRM				_	
1	- 0.83	0.65	0.47	0.43	_	0.81	0.53	0.38	0.27	-	0.75	0.5	0.39	0.33		ACC - 0.6
2	- 0.75	0.53	0.41	0.41	-	0.72	0.42	0.39	0.28	-	0.65	0.43	0.33	0.34		- 0.4
Clusters	- 0.75	0.52	0.35	0.33	-	0.69	0.44	0.21	0.15	-	0.63	0.45	0.29	0.28		- 0.2
Grouped	- 0.78	0.55	0.43	0.28	-	0.69	0.38	0.28	0.18	-	0.69	0.42	0.27	0 .21		- 0.0
5	- 0.83	0.61	0.41	0.24	-	0.79	0.42	0.27	0.33	-	0.75	0.46	0.39	0.26		0.4
€	- 0.79	0.6	0.52	0.44	-	0.66	0.51	0.36	0.27	_	0.69	0.46	0.35	0.32		0.6
	<u>1/2 2/3</u> 3/4 4/5 Week					1/2	2/3 We	3/4 ≘ek	4/5		1/2	2/3 We	3/4 eek	4/5		_



Verification – 850hPa Spechum

	Specific ECMWF	Humidit	ty at 850) hPa		икмо				(ONRM				_
1	- 0.76	0.49	0.47	0.4	-	0.49	0.32	0.22	0.19	-	0.71	0.44	0.37	0.38	ACC - 0.6
2	- 0.67	0.48	0.17	0.38	-	0.47	0.13	0.12	-0.11	-	0.57	0.12	0.19	0.37	- 0.4
Clusters	- 0.69	0.39	0.12	0.16	-	0.35	0.12	-0.01	0.01	-	0.39	0.2	0.08	0.05	- 0.2
Grouped	- 0.66	0.37	0.22	0.09	-	0.5	0.04	0.14	0.03	-	0.45	0.2	0.11	0.03	- 0.0
5	- 0.76	0.2	0.01	-0.09	-	0.5	-0.01	-0.27	-0.32	-	0.47	0.14	0.07	-0.19	0.4
6	- 0.53	0.32	0.27	0.32	-	0.26	0.06	0.18	0.12	-	0.28	0.09	-0.17	0.1	0.6
	1/2	2/3 We	3/4 eek	4/5		1/2	2/3 We	3/4 eek	4/5		1/2	2/3 We	3∕4 ≘ek	4/5	

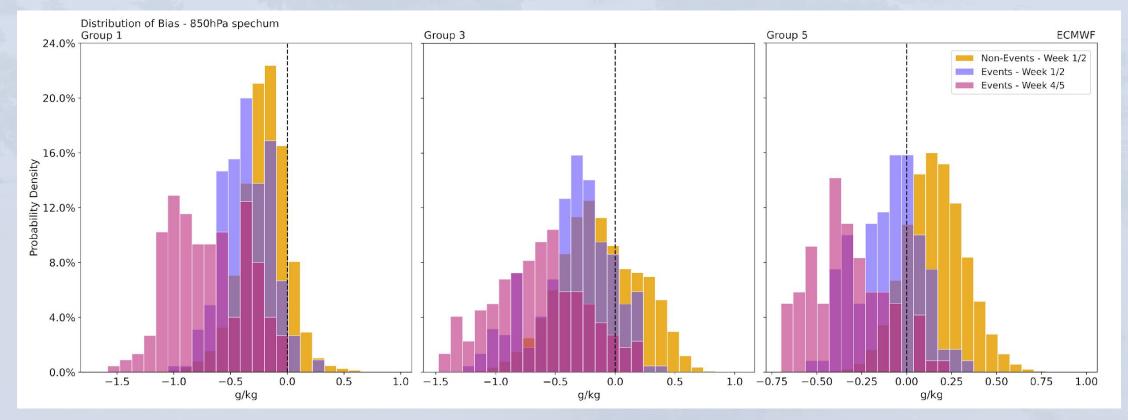


700hPa specific humidity has similar trends, with overall lower skill

Extreme Event Bias within Synoptic Variables

Bias of 850hPa Specific Humidity

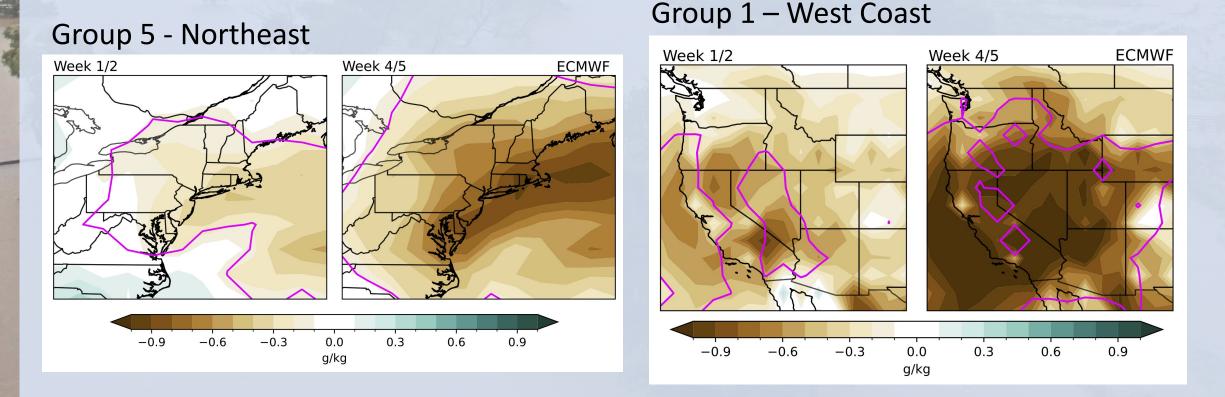
Comparing distributions of event and non-event bias



Bias of 850hPa Specific Humidity

Geospatial distributions of Event Bias

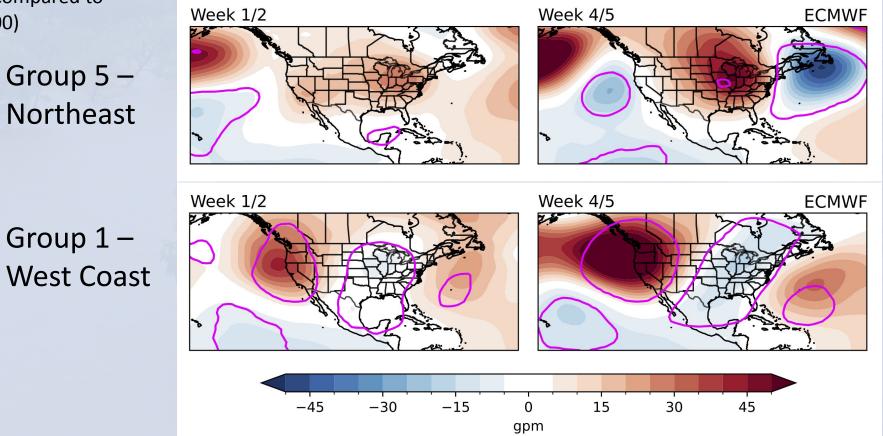
P-value: 90% confidence interval compared to bootstrapped non-events (n = 2500)



Bias of 500hPa Geopotential Height

Geospatial distributions of Bias P-value: 90% confidence interval compared to bootstrapped non-events (n = 2500)

Group 5 – Northeast



Conclusions

- Skill of synoptic variables is low past Week 2/3, with more skill seen on the West Coast. Similar to S2S model precipitation skill of these events.
- There is a dry bias in low level specific humidity that increases with lead time.
- Models have an inability to produce
 500hPa dipoles that are common to most extreme events cases past Week 1/2.
- There are cases where the synoptic variables are well forecasted with better precipitation forecasts. Primarily on the West Coast of the CONUS.

Next Steps:

What does the extra-tropical cyclone activity look like during these extreme events? Do the S2S models properly represent ETC activity during events?



Information regarding climatology, typical storm reports, and seasonality of the extreme precipitation events is now accessible to all at pres2ip.com.



SCAN ME

Contact: maschroers@ou.edu



About PRES²iP

<u>Prediction of Rainfall Extremes at Subseasonal to Seasonal Periods</u>

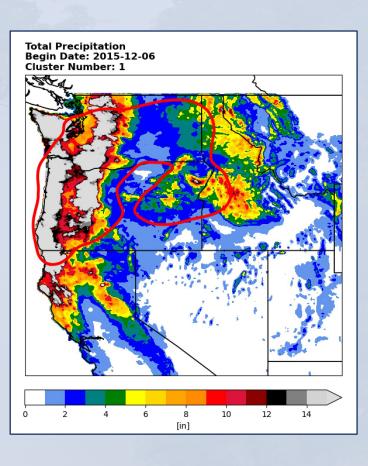
The goal of this project is to enhance the physical understanding of the large-scale dynamics and forcing of S2S extreme precipitation events, improve our capability to predict these events, and increase communication between research and stakeholder communities with regard to extreme precipitation.

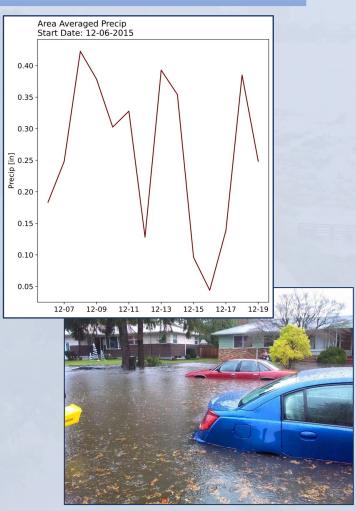
PRES²iP

What do these extreme periods look like?

West Coast Extreme Period in December 2015

- \$35 Million of Property Damage
- Occurred after regional drought.
- Typically made up of multiple systems passing over the same area.



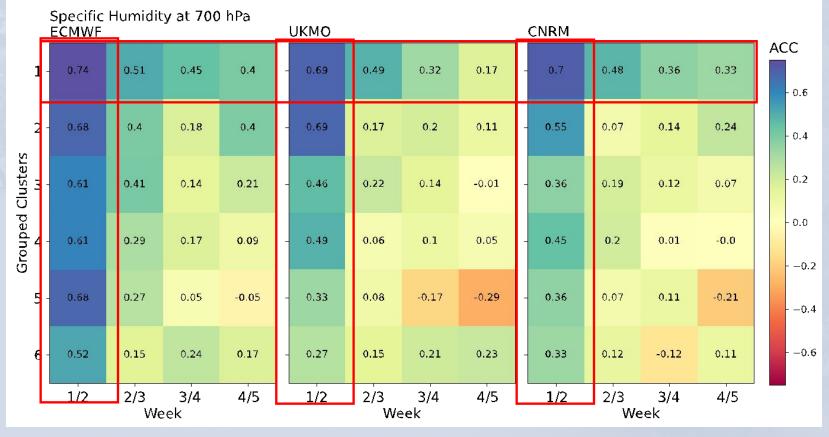


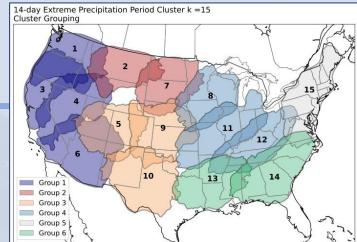
S2S Model Descriptions

Verifying the ensemble averages of the extreme periods 14-day averaged forecasted anomalies against ERA5 observations.

S2S Model	Time Range	Hindcast Freq	Hindcast Period	Ensemble Size	Ocean Coupled	Sea-Ice Coupled	Extreme Periods Covered
ECMWF	46	2/week	1996-201 4	11	Yes	No	171
UKMO	60	4/month	1996-201 4	2	Yes	Yes	171
Meteo-Fr ance	61	1/week	1996-201 4	15	Yes	Yes	171

Verification – 700hPa Spechum

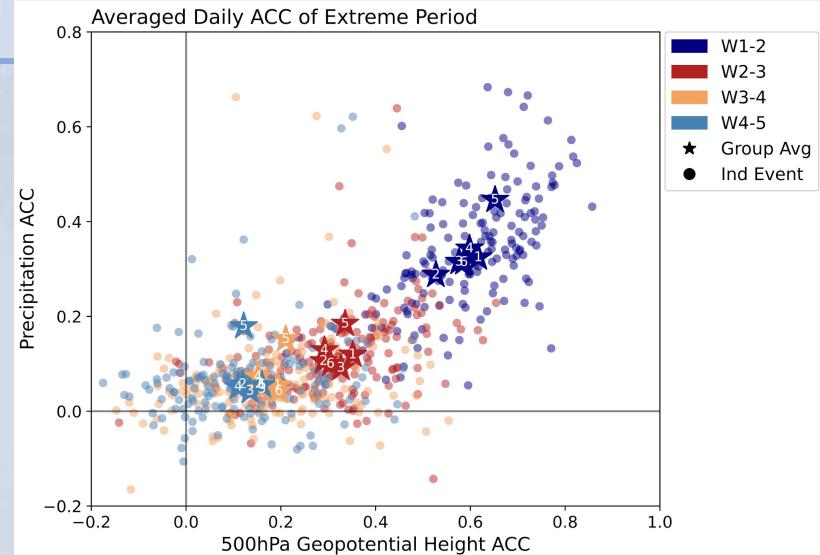




Comparison to Precipitation Skill

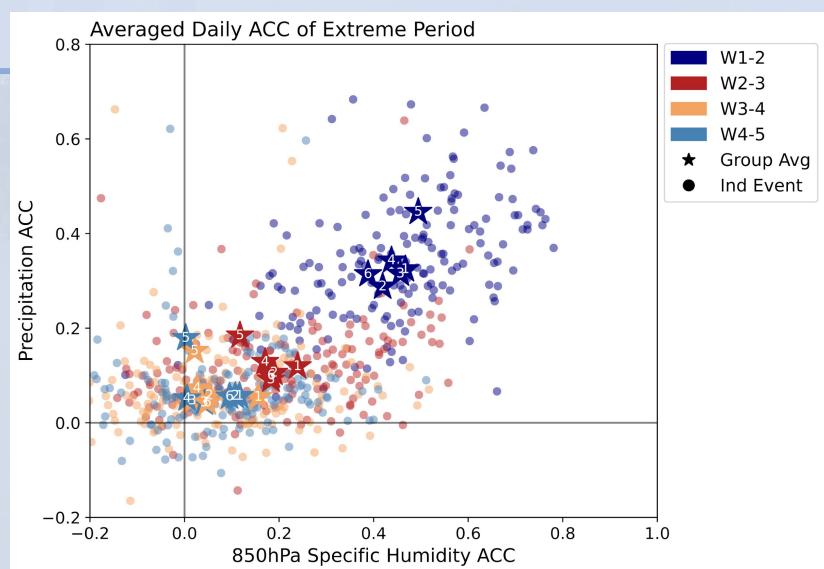
Precipitation ACC (y) vs. 500hPa geopotential ACC (x)

- Positive correlation at Week 1/2
- Less to no correlation seen past Week 1/2
- Group Averages are similar



Comparison to Precipitation Skill

Precipitation ACC (y) vs. 850hPa spec humidity ACC (x)



Bias of 850hPa Specific Humidity

Comparing distributions of event and non-event bias

