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“Characterization of Extremes”

Collaborators and co-authors on this work

Stanford Research Group

Daniel Swain

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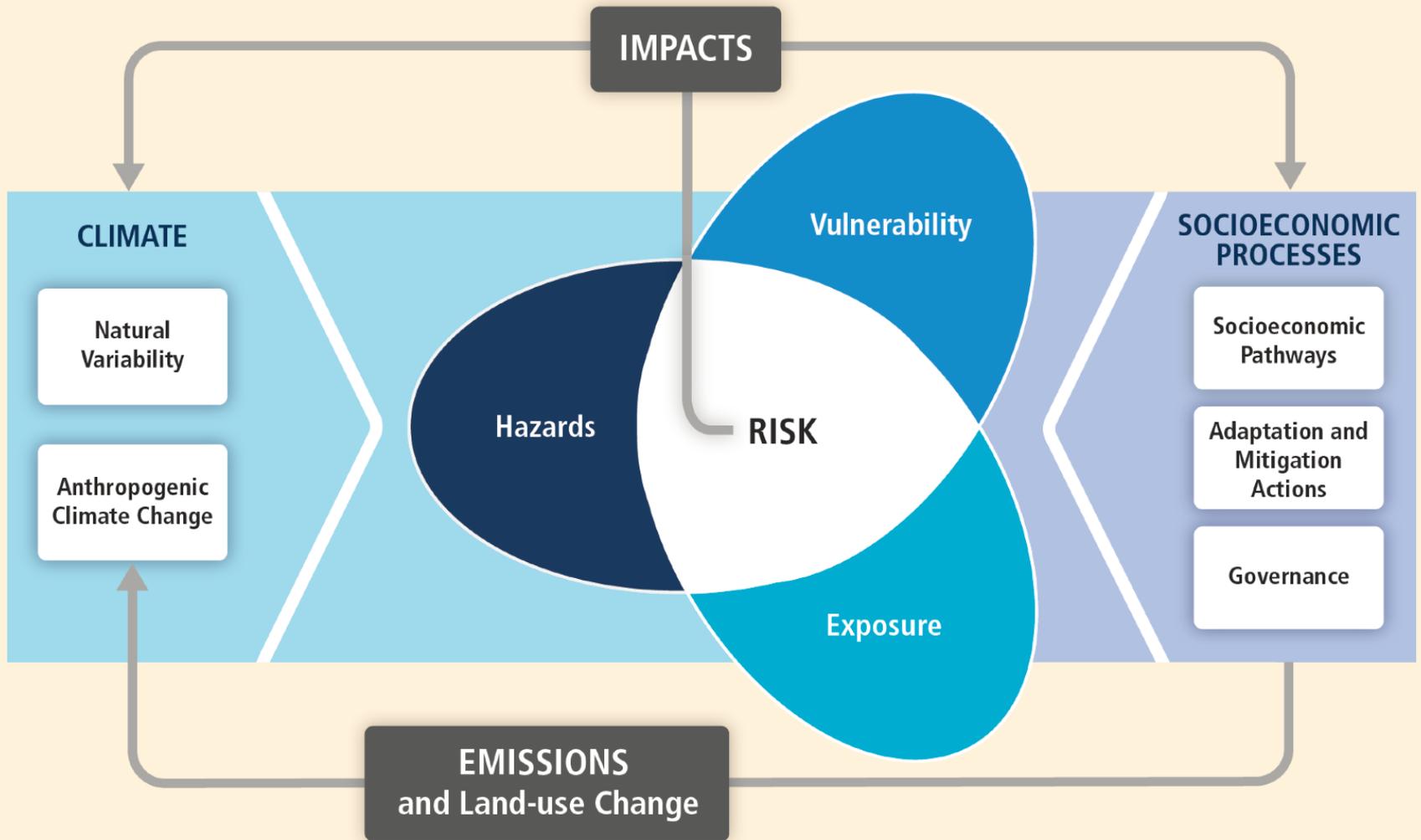
Michael Tsiang

Matz Haugen

Other Collaborators

Nat Johnson

IPCC: “Climate Risk Management”



A “Climate Dynamics” Risk Framing: Has the probability of the conditions that create extreme events changed?

$$\text{Risk} = f(\text{hazard, exposure, vulnerability})$$

What is probability of physical hazard in the current climate?

⇒ Scientifically tractable

⇒ Practically actionable

A “Climate Dynamics” Risk Framing: Has the probability of the conditions that create extreme events changed?

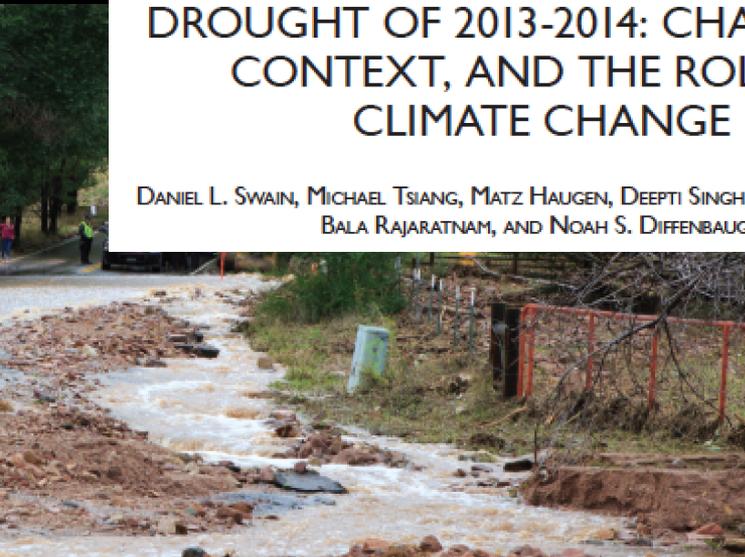
=> Have observed changes in surface extreme events been influenced by changes in the probability of atmospheric conditions?

- Trends in atmospheric patterns contributing to trends in extremes?
- Trends in atmospheric patterns contributing to individual extremes?
- Unprecedented atmospheric patterns contributing to individual extremes?

Unprecedented atmospheric patterns contributing to individual extremes?

2. THE EXTRAORDINARY CALIFORNIA DROUGHT OF 2013-2014: CHARACTER, CONTEXT, AND THE ROLE OF CLIMATE CHANGE

DANIEL L. SWAIN, MICHAEL TSIANG, MATZ HAUGEN, DEEPTI SINGH, ALLISON CHARLAND, BALA RAJARATNAM, AND NOAH S. DIFFENBAUGH



EXPLAINING EXTREME EVENTS OF 2013

From A Climate Perspective

Special Supplement to the
Bulletin of the American Meteorological Society
Vol. 95, No. 9, September 2014

Science Advances (April 1, 2016)

RESEARCH ARTICLE

CLIMATOLOGY

Trends in atmospheric patterns conducive to seasonal precipitation and temperature extremes in California

Daniel L. Swain,^{1*} Daniel E. Horton,^{1,2} Deepti Singh,^{1,3} Noah S. Diffenbaugh^{1,4}

Ongoing CA drought: One of many, many examples.....



Source: Huffington Post

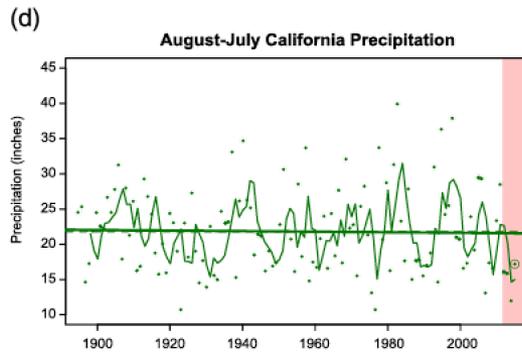
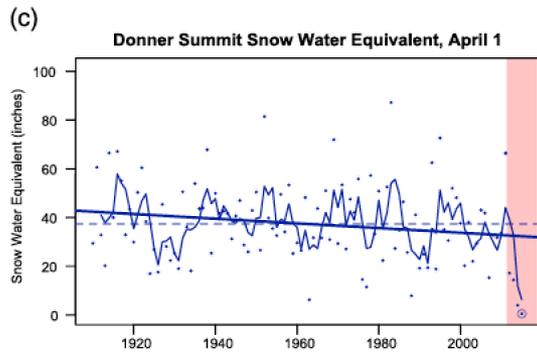
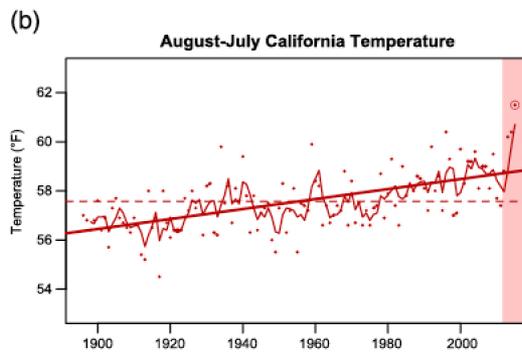
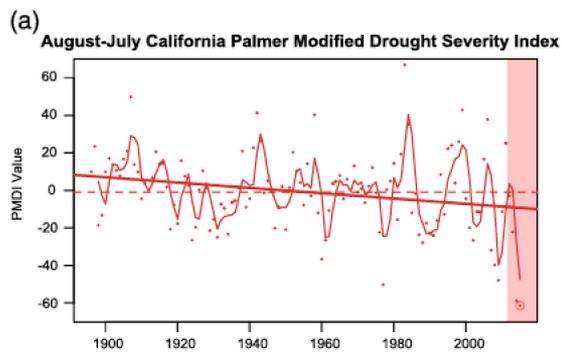


ASSOCIATED PRESS

UC Davis (Aug 2015):
\$2.7 billion damage, 21,000 jobs

Table ES-1. Summary of impacts of the 2015 California drought

Description	Impact	Base year levels	Percent change
Surface water shortage (million acre-ft)	8.7	18.0	-48%
Groundwater replacement (million acre-ft)	6.0	8.4	72%
Net water shortage (million acre-ft)	2.7	26.4	-10%
Drought-related idle land (acres)	540,000	1.2 million*	45%
Crop revenue losses (\$)	\$900 million	\$35 billion	2.6%
Dairy and livestock revenue losses (\$)	\$350 million	\$12.4 billion	2.8%
Costs of additional pumping (\$)	\$90 million	\$780 million	75.5%
Direct costs (\$)	million	NA	NA
Total economic impact (\$)	\$2.7 billion	NA	NA
Direct job losses (farm seasonal)	10,100	200,000#	5.1%
Total job losses	21,000	NA	NA



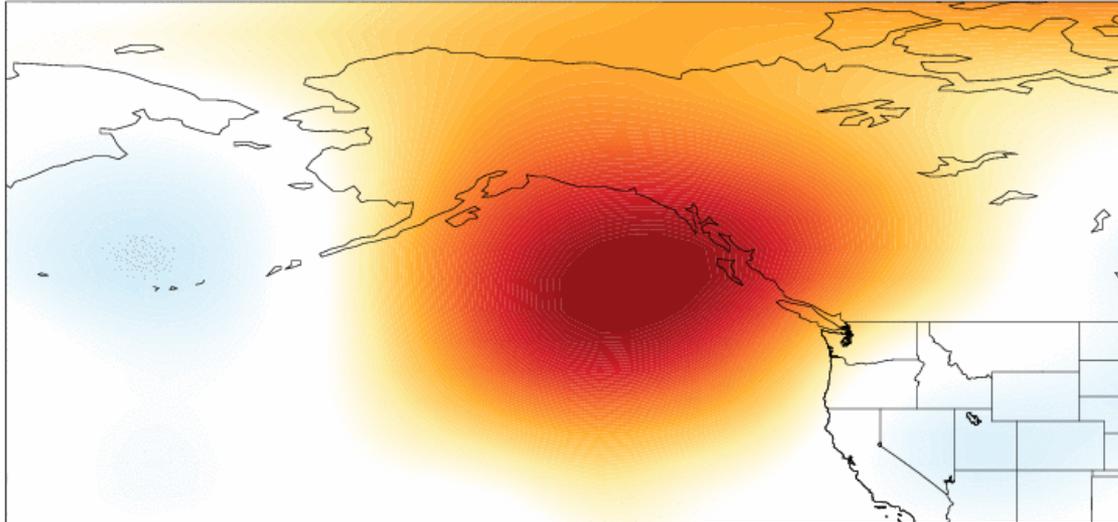
The current drought has produced many statewide records

Swain, GRL, 2015

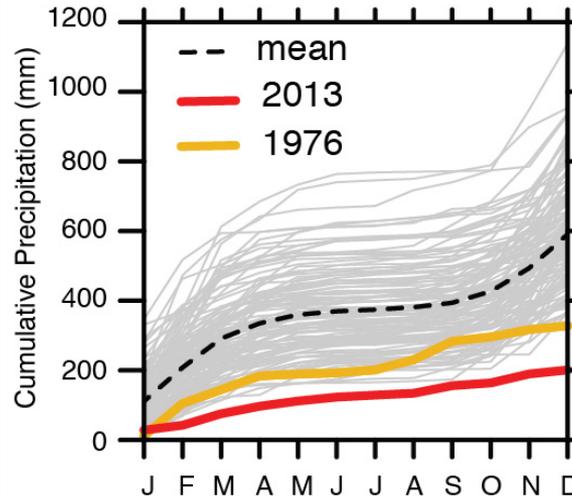
- most severe NOAA NCDC drought indicator values
- lowest calendar-year precipitation
- lowest 12-month precipitation
- lowest April 1st snowpack
- warmest calendar-year temperatures
- warmest winter temperatures
- warmest/driest 3-year period

The (Ongoing) California Drought (2012-2016)

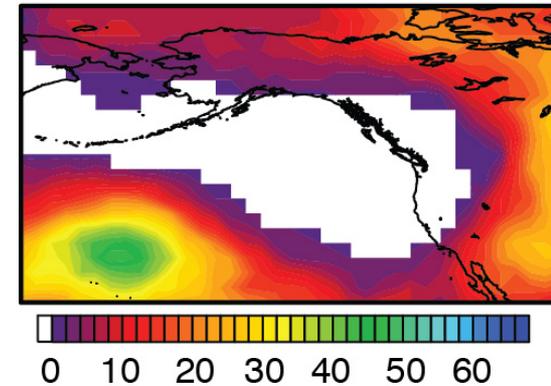
The Ridiculously Resilient Ridge 10/15/2013



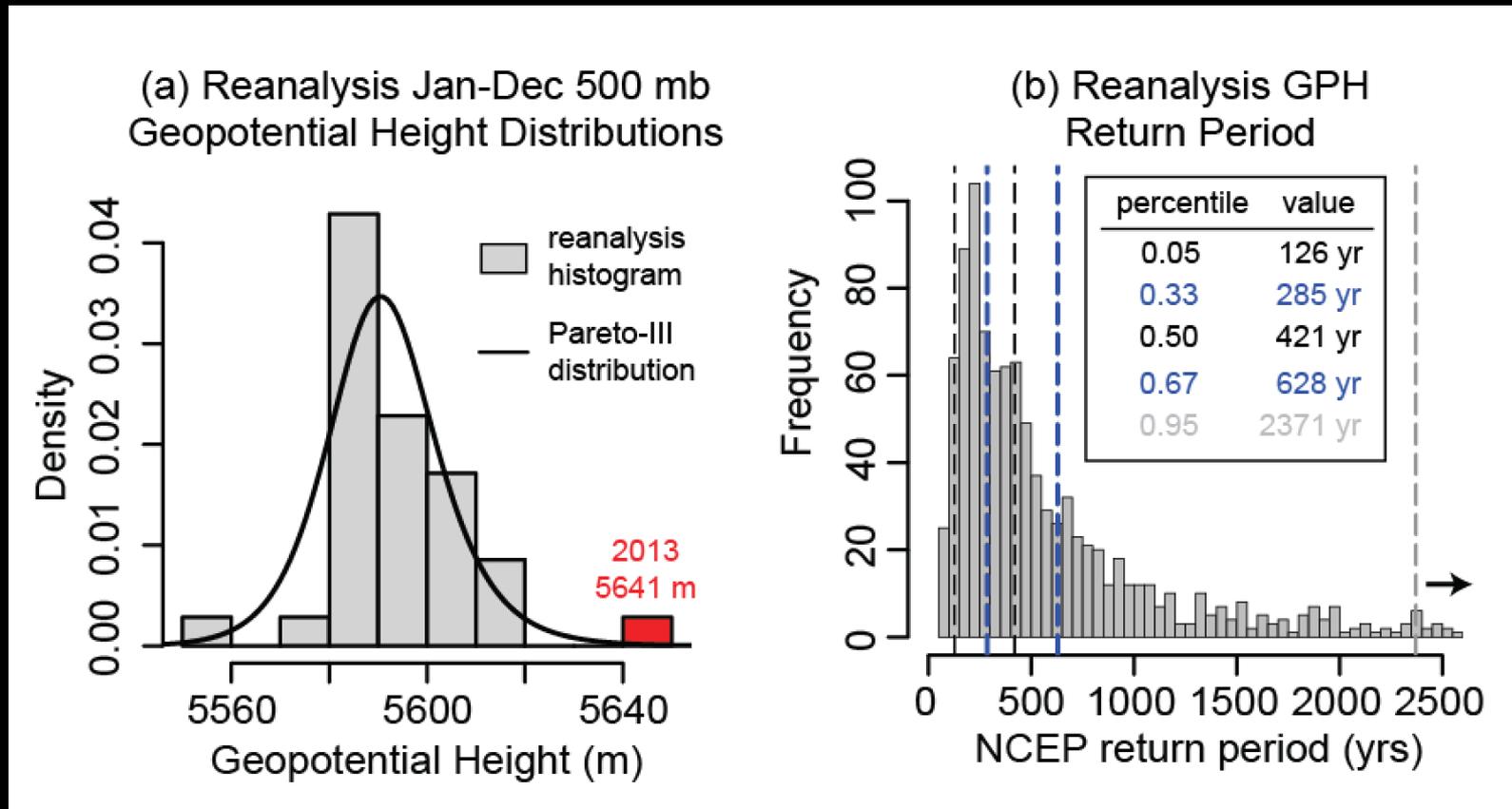
(a) Jan-Dec Cumulative Precipitation (mm)



(b) Exceedance of Jan-Dec Mean 500mb Geopotential Height (yr)



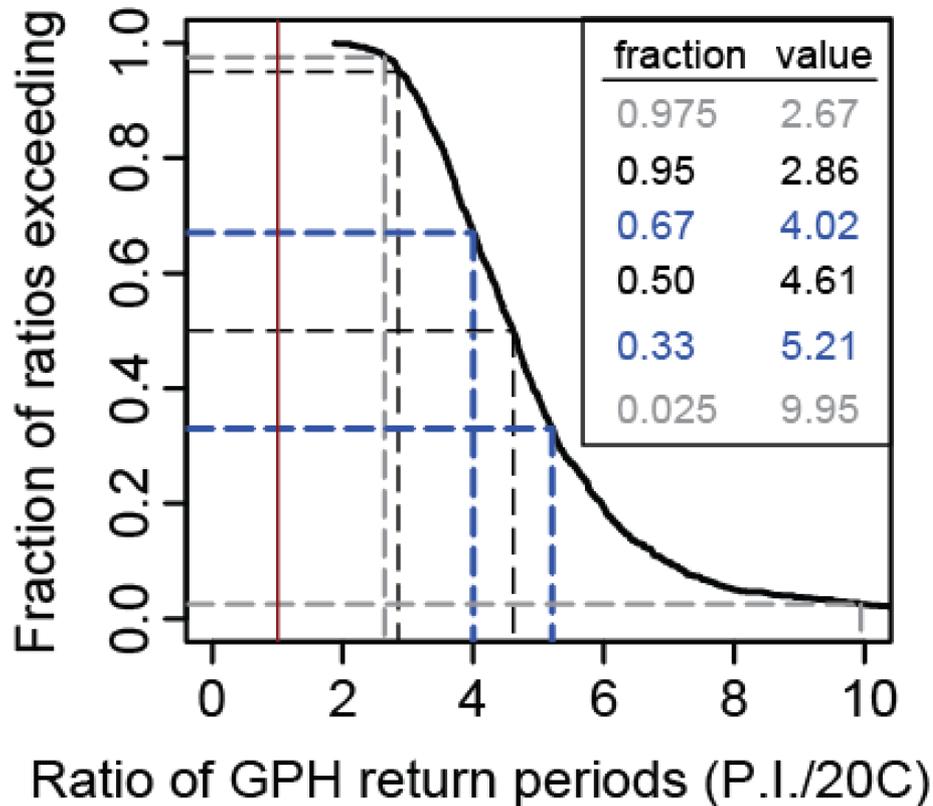
Return period of 2013 reanalysis 500mb GPH



- Observed persistent atmospheric pressure was at least a century-scale event, and 50% probability >420 years

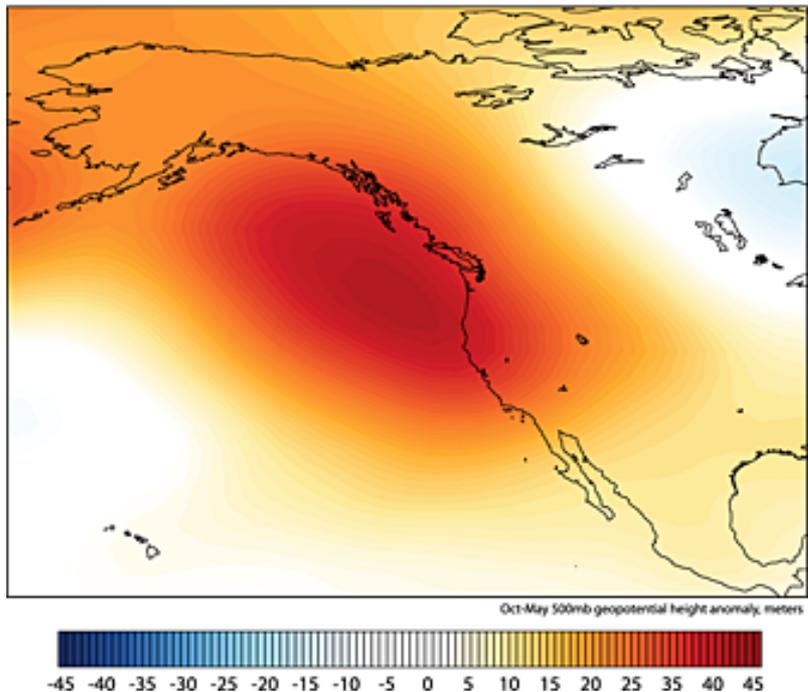
Probability of persistently high geopotential heights over the northeast Pacific

Ratio between CMIP5 Pre-Industrial (“PI”) and Historical (“20C”) Simulations



- “Very likely” (IPCC: 90%)
=> tripling
- “Likely” (IPCC: 66%)
=> quadroupling

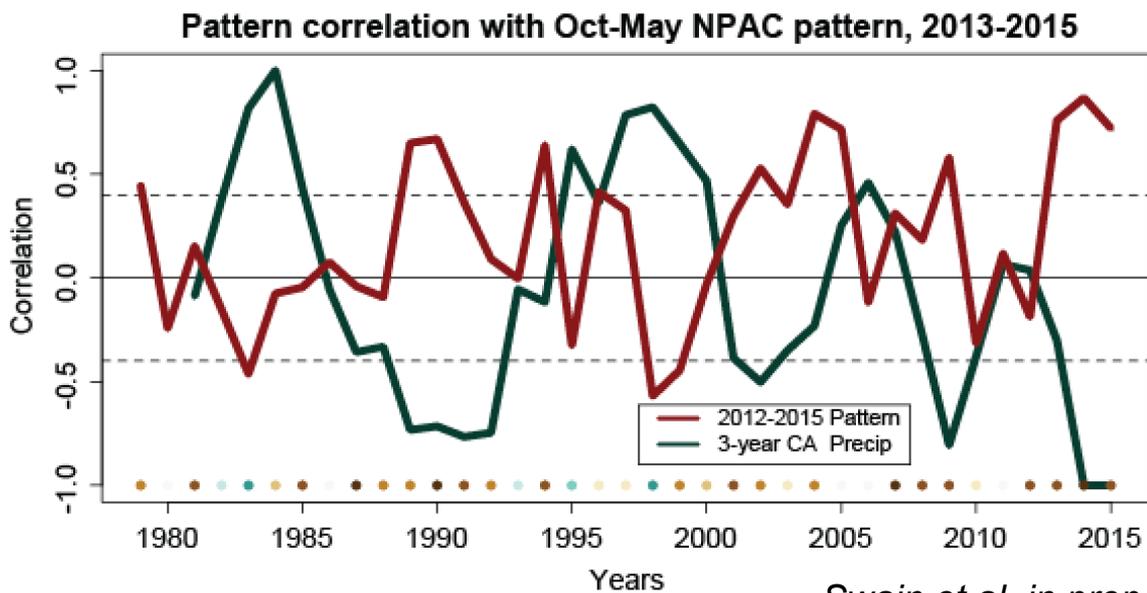
The Ridiculously Resilient Ridge, 2012-2015



BUT:

Has the probability of the *gradient* changed?

Swain, GRL, 2015



Swain et al. in prep



Trends in atmospheric patterns contributing to individual extremes?

RESEARCH ARTICLE

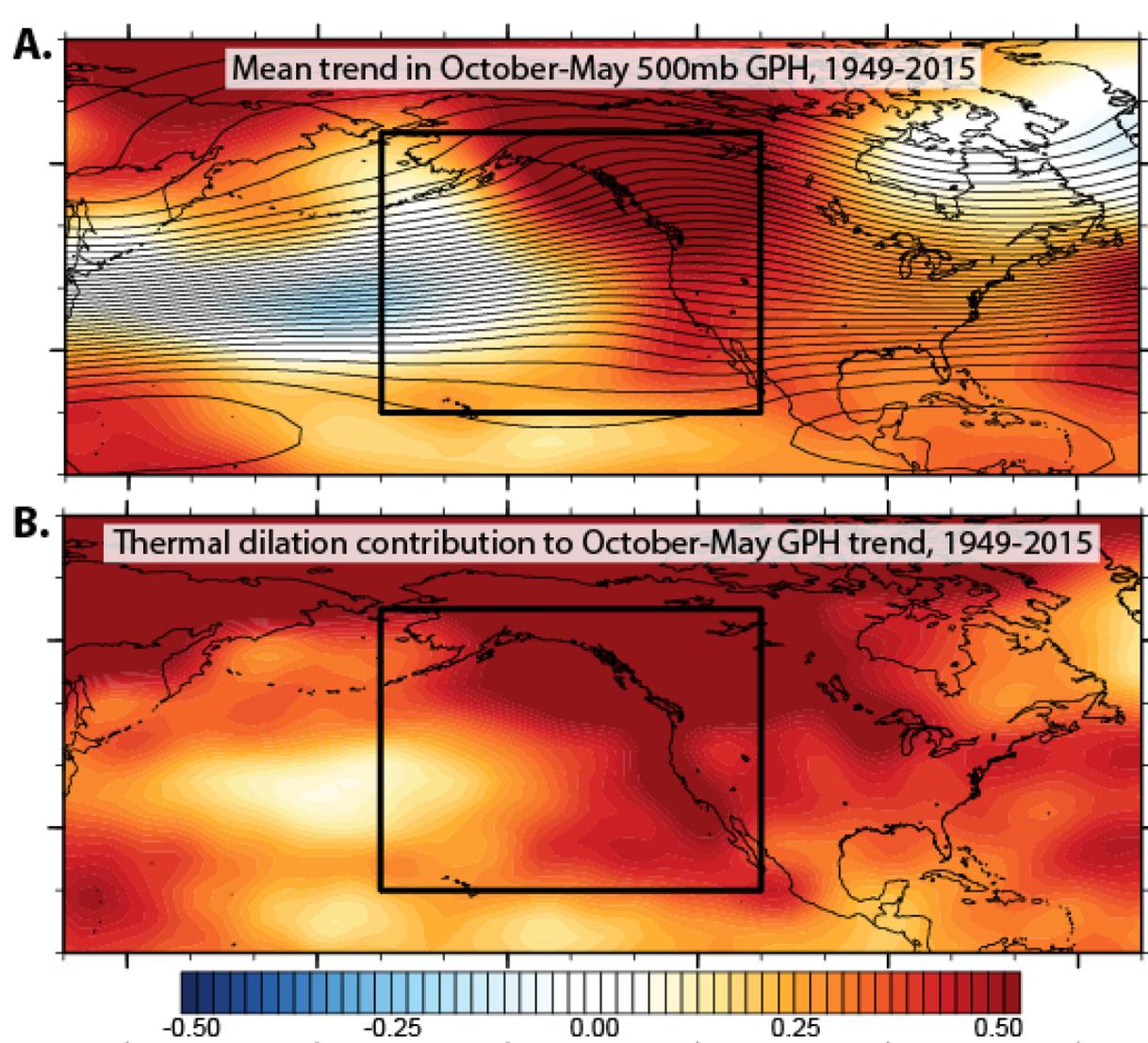
CLIMATOLOGY

Trends in atmospheric patterns conducive to seasonal precipitation and temperature extremes in California

Daniel L. Swain,^{1*} Daniel E. Horton,^{1,2} Deepti Singh,^{1,3} Noah S. Diffenbaugh^{1,4}

Swain et al., Sci. Adv., 2016

Trend in mean seasonal 500mb geopotential heights

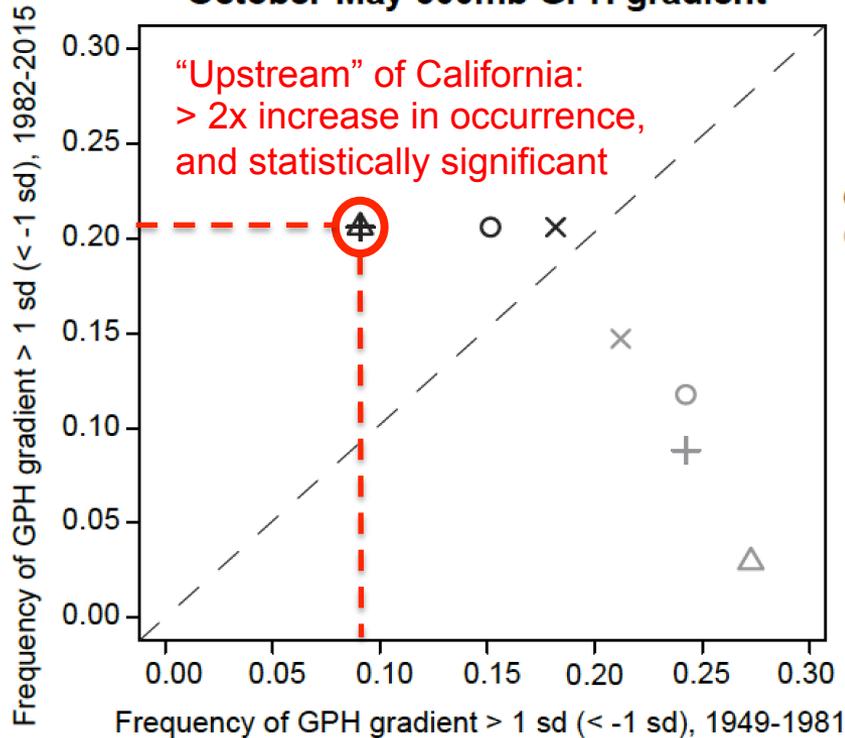


- Observed trend is highly non-uniform
- “Thermal dilation” has caused mean to become more like “Ridiculously Resilient Ridge”

Frequency of extreme 500mb geopotential height gradients

c.

Change in frequency of extreme
October-May 500mb GPH gradient



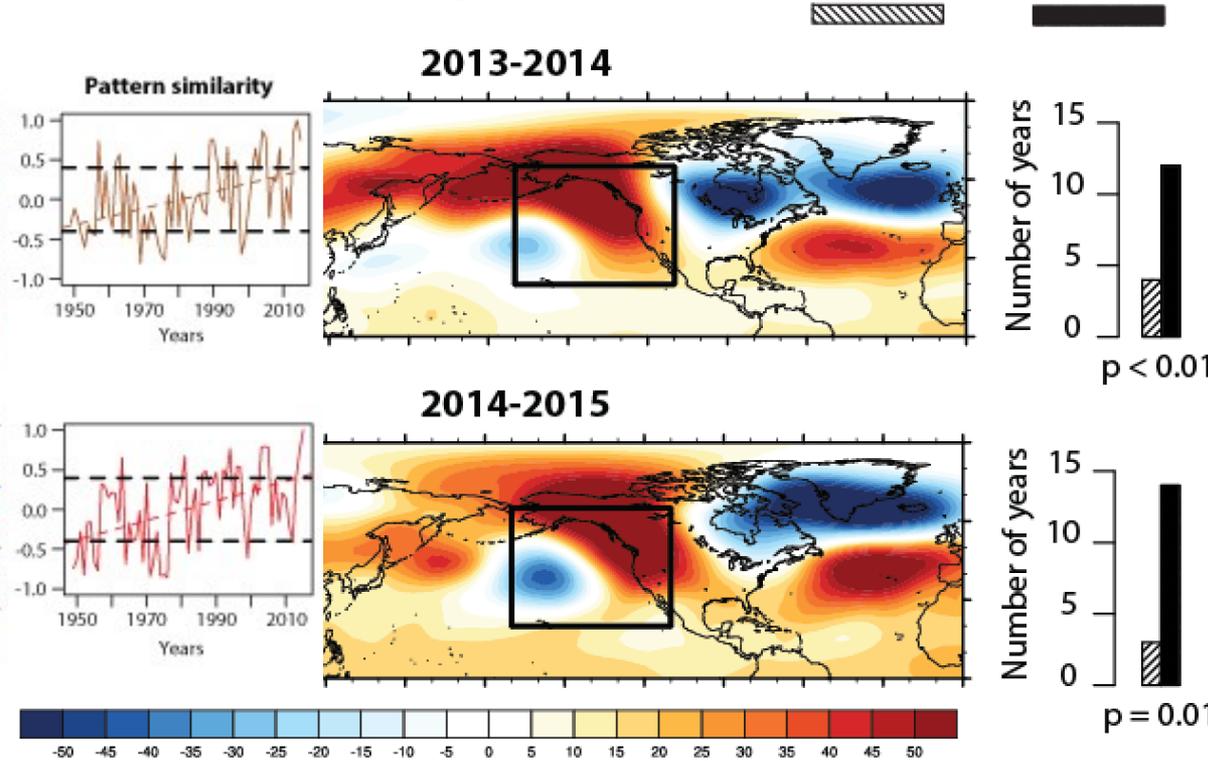
1949-1981

- Statistically significant trend in the gradient
- Statistically significant change in the probability of extreme gradient years

Probability of years similar to the “Ridiculously Resilient Ridge”



Frequency of moderate to high correlation (>0.4), 1949-1981 vs. 1982-2015



Historical Observations:

- Statistically significant increase in the occurrence of patterns similar to the “Ridiculously Resilient Ridge”, driven by pattern of surface warming
- But no decrease in the occurrence of patterns associated with extremely wet years

Trends in atmospheric patterns contributing to trends in extremes?

LETTER

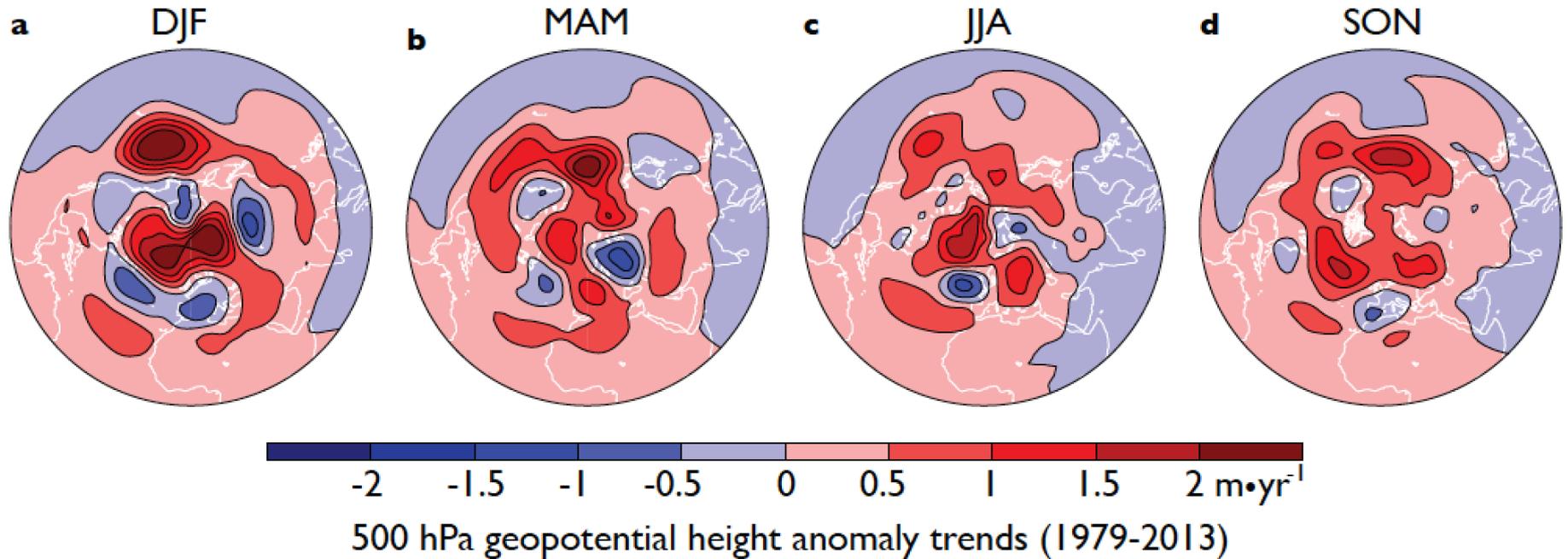
doi:10.1038/nature14550

Contribution of changes in atmospheric circulation patterns to extreme temperature trends

Daniel E. Horton^{1,2}, Nathaniel C. Johnson^{3,4,5}, Deepti Singh¹, Daniel L. Swain¹, Bala Rajaratnam^{1,2,6} & Noah S. Diffenbaugh^{1,2}

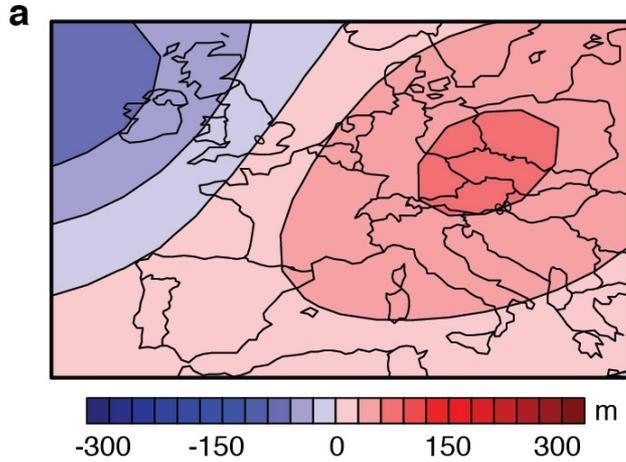
Horton et al., Nature, 2015

Reanalysis trends in 500mb geopotential heights

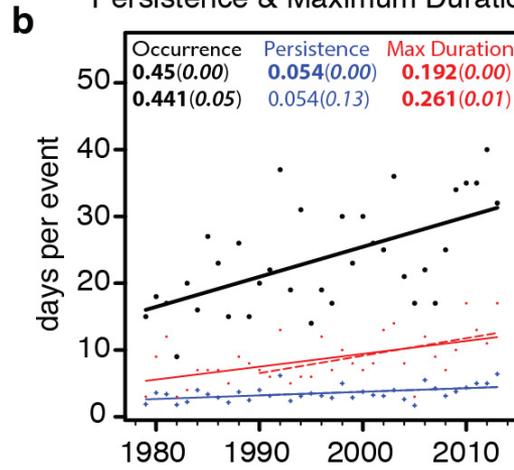


- Quite robust to numerous approaches to removal of systematic “thermal dilation”

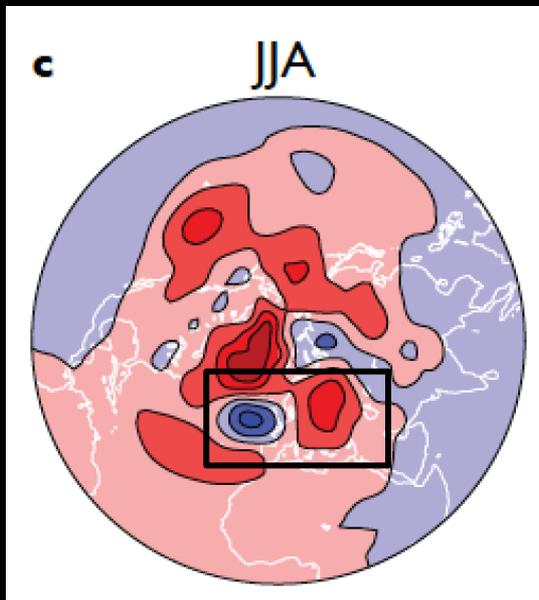
500 hPa Circulation Pattern



Trends in Pattern Occurrence, Persistence & Maximum Duration

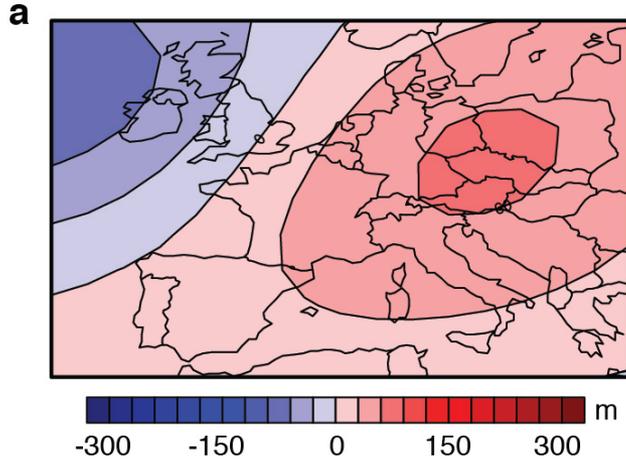


seasonal-mean GPH trend

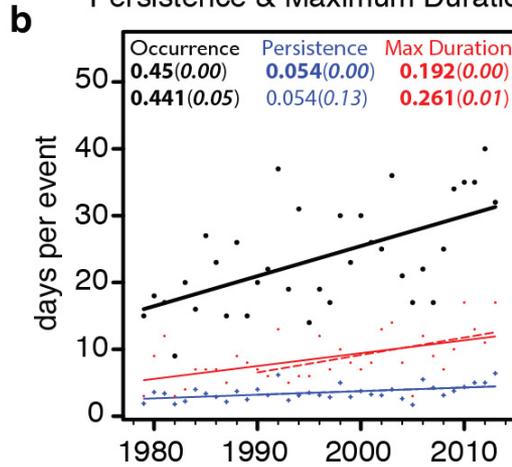


- Many mid-latitude regions show robust, statistically-significant changes in atmospheric pattern occurrence

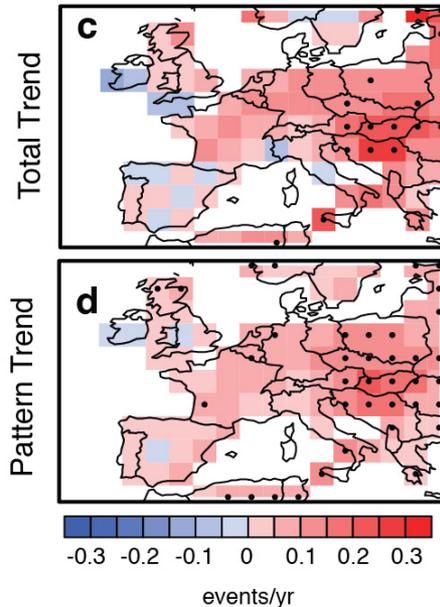
500 hPa Circulation Pattern



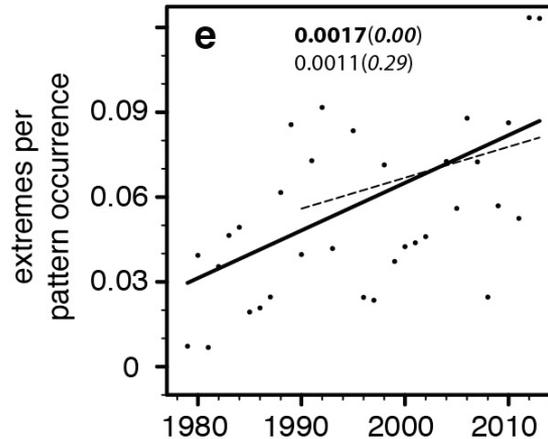
Trends in Pattern Occurrence, Persistence & Maximum Duration



Trend in Summer Hot Extremes (1979-2013)



Trends in Summer Hot Extremes per Pattern Occurrence



For those regions, changes in extreme temperature occurrence are driven ~20% (+/- 10%) by changes in atmospheric pattern occurrence

=> Are these changes "forced" or "noise"???

Example 3: Truly extreme 5-day precipitation in Northern India in June 2013

17. SEVERE PRECIPITATION IN NORTHERN INDIA IN JUNE 2013: CAUSES, HISTORICAL CONTEXT, AND CHANGES IN PROBABILITY

DEEPTI SINGH, DANIEL E. HORTON, MICHAEL TSIANG, MATZ HAUGEN, MOETASIM ASHFAQ, RUI MEI, DEEKSHA RASTOGI, NATHANIEL C. JOHNSON, ALLISON CHARLAND, BALAJI RAJARATNAM, AND NOAH S. DIFFENBAUGH



EXPLAINING EXTREME EVENTS OF 2013

From A Climate Perspective

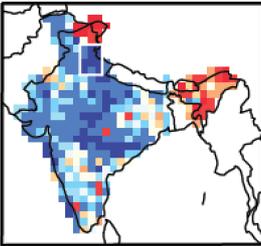
Special Supplement to the
Bulletin of the American Meteorological Society
Vol. 95, No. 9, September 2014



Source: Mashable

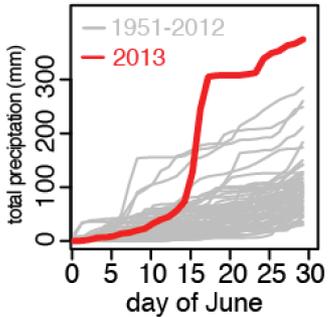
We can quantify the historical occurrence of simultaneous “ingredients”

(a) June Precipitation Percentile



0 30 60 90
percentile

(b) Cumulative Precipitation

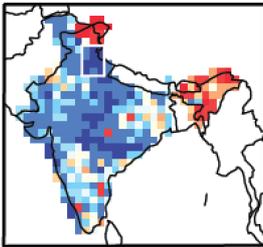


June 2013

- Wettest regional precipitation on record

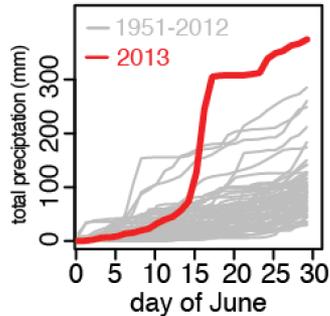
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(a) June Precipitation Percentile



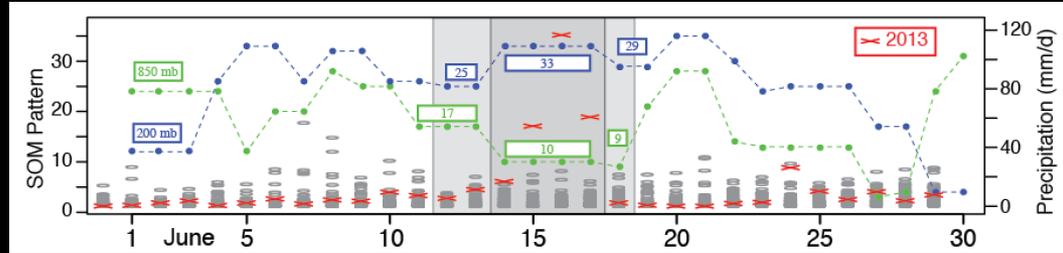
0 30 60 90
percentile

(b) Cumulative Precipitation



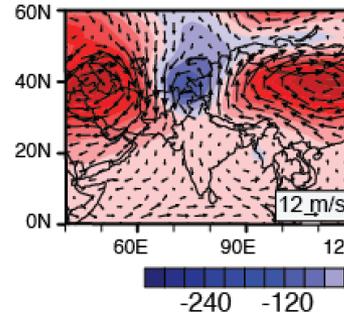
June 2013

- Wettest regional precipitation on record
- Only June co-occurrence of upper and lower-level patterns
- Climate models don't capture daily-scale variability

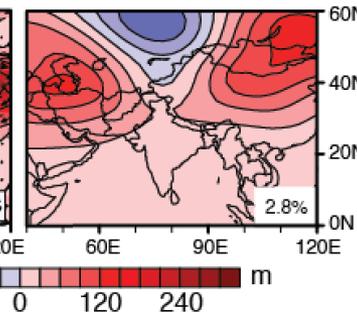


June 14-17 2013 Geopotential Height Anomalies

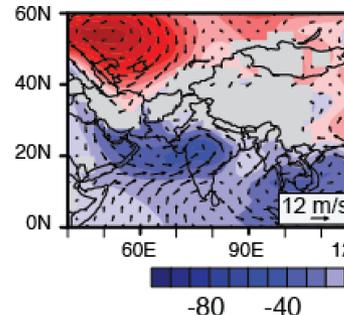
(e) 200mb Composite



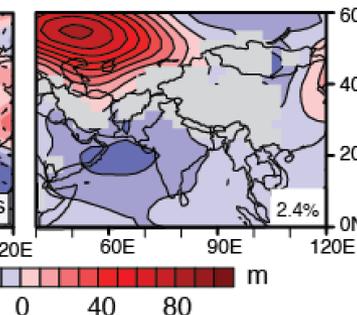
(g) SOM Pattern Match



(f) 850mb Composite

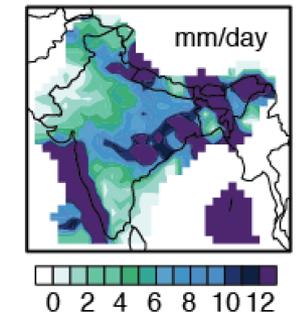


(h) SOM Pattern Match

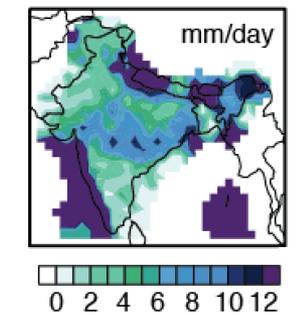


1951-2013 Precipitation

(i) 200mb SOM Pattern



(j) 850mb SOM Pattern



Some conclusions....

- A “risk” framing that focuses on the likelihood of events is both scientifically tractable and practically actionable
- Observations, climate models, computational resources, and analysis methods have developed to the point that we can test whether the likelihood of extreme events has changed
- This question is also possible for some unprecedented events, although observational records are relatively short, and climate models are limited
- For complex extreme events with many interacting components, how do we comprehensively analyze causality? And is absence of evidence = evidence of absence?
- Are there ways to better test hypotheses given relatively short observed record and the limitations of climate models?