



# **Observed Changes in Extremes Affecting California: What and Why**

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Western Regional Climate Center



Intensity - Duration - Frequency (IDF) Curves \*

What is our expectation of

1) How much, 2) Over what time interval, 3) How often ?

\* Note: Often show accumulation rather than intensity (sums versus rates)

Much, much scrutiny Very widely used Critical for infrastructure design Scientific, legal, financial underpinnings

These and other climatic extremes built into building codes



Fri Aug 08 16:35:55 2008

#### **NOAA HDSC**

Duration			
5-min ——	120-m <del>-</del>	48-hr <del>-×-</del>	30-day ———
10-min 🔶	3-hr <del>-*</del> -	4-day <del>-</del>	45-day 🛶
15-min 🕂	6-hr 🔶	7-day 🔶	60-day 😽
30-min <del></del>	12-hr 🕂	10-day ——	
60-min <del>-×-</del>	24-hr <del>-8-</del>	20-daý <del>-</del> ⊟-	



7 miles

### **Extremes matter**

Societal physical and behavioral infrastructure are built around:

**Central tendencies - the many** 

High probability / low consequence events

**Distribution tails - the few** 

Low probability / high consequence events

Huge societal investments (\$B, \$\$B, \$\$B) to withstand rare events

Where do we obtain these probabilities? Pertinent decisions are about the future: hence, these are forecasts The commonsense assumption: Past is Prologue The past as a reliable guide to the future Past statistics = Future statistics The past (as established "fact") has built-in credibility as a forecast Automatic, relatively painless buy-in Climate stationarity is implicit in this assumption

### Two major audiences to satisfy:

Scientific What is intellectually defensible, accurate, correct?

#### **Societal**

Acceptance by the engineering profession
 Methodologies that are understandable, transparent, etc
 Practical and implementable
 Acceptance by the planning community
 Huge \$\$\$ at stake in building for extremes
 Acceptance by political process

For both audiences Stationary climate Reasonably constrained range of options Considerable experience base to work from Nonstationary climate Many more possible options to choose from Lots of ways to "be nonstationary" Experience base very limited

Not "just" a science problem, but a joint science-society problem

### **Stationarity**

### There does not seem to be a uniform meaning for this word

Often used in a statistical sense

Values that are i.i.d.

independent and identically distributed

- each new event independent of the last
- always drawn from the same statistical distribution(s)

In a physical sense

The causal mechanisms in the physical world (that generate the statistics) continue to occur, and to interact with each other, in approximately the same manner, over the time period for which stationarity is claimed.

However, in the physical world, the temporal characteristics of physical causation are constantly changing, all the time, on all time scales

Stationarity concept must contain embedded implicit time scales

#### The climate system as an operator

y = f(x), where x is input, operated on by f, with output y

s = H(ppt; parameters), where

ppt falls as input from sky (rain or snow) ...
... is operated on by Hydrologic system H ...
... with output s (some kind of hydrologically relevant quantity)

s could represent streamflow, lake level, groundwater status, etc

#### Η

is not linear in ppt

is sometimes simple, usually very complex, often extremely complex is dependent on many parameters representing state of system Some of these themselves depend on the climate system (feedbacks) Some of these are externally manipulated (e.g., human activities) could be natural (catchment, basin, major river system, lake, etc) need not be "natural" (roof, parking lot, tailings pond, alfalfa field, etc) acts as a kind of complex time filter

Variability in output s may result from variability in input ppt or in H or both That is, nonstationarity in s could result from driver or system nonstationarity

Rarity of input ppt need not produce same rarity in output s 25 year precipitation event need not produce 25 year hydro event ppt

Η

S









### American River @ Fair Oaks (Sacramento CA) Annual Maximum Three-Day Average Flow Reconstructed Natural Flow below Folsom Reservoir



### **National Research Council**

### January 1999





FIGURE 1.1 Main features of the American River watershed. SOURCE: Sacramento District, USACE, 1991.





Bulletin # 17B of the Hydrology Subcommittee

Revised September 1981 Editorial Corrections March 1982

INTERAGENCY ADVISORY COMMITTEE ON WATER DATA

U.S. Department of the Interior Geological Survey Office of Water Data Coordination Reston, Virginia 22092



Water Year Oct-Sep Precip

South Coastal California

1895/96 thru 2009-10







### Note:

**Typical distributions with exponential tails:** 

Binomial Normal Lognormal Gamma Gumbel Weibull Double (or Stretched or extended ) exponential

**Typical distributions with heavy tails:** 

Pareto Cauchy Frechet Stable laws

### **Pareto Distribution**

If X is a random variable with a Pareto distribution, then the probability that X is greater than some number  $\times$  is given by

$$\Pr(X > x) = \begin{cases} \left(\frac{x_{\rm m}}{x}\right)^{\alpha} & \text{for } x \ge x_{\rm m}, \\ 1 & \text{for } x < x_{\rm m}. \end{cases}$$

### **Applies to**

**Frequency distribution of wealth** 

**Sizes of cities** 

Sizes of internet files

Sizes of Bose-Einstein condensate clusters

Sizes of sand grains

**Sizes of meteorites** 

**Sizes of forest fires** 

Sizes of insurance losses

The expected value of a random variable following a Pareto distribution with α > 1 is

$$E(X) = \frac{\alpha x_{\rm m}}{\alpha - 1}$$

(if  $\alpha \leq 1$ , the expected value does not exist).

• The variance is

$$\operatorname{var}(X) = \left(\frac{x_{\mathrm{m}}}{\alpha - 1}\right)^2 \frac{\alpha}{\alpha - 2}.$$

(If  $\alpha \leq 2$ , the variance does not exist).

• The raw moments are

$$\mu_n' = \frac{\alpha x_{\rm m}^n}{\alpha - n},$$

but the *n*th moment exists only for  $n < \alpha$ .



A random variable that follows a heavy tailed distribution (such as Pareto) can be extremely large with non-negligible probability



"The biggest one is yet to come." - Jim Goodridge CA DWR



### Pareto alpha

### San Jose CA Monthly Precip 1908-2000

Alpha Parameter of Rainfall Distributions in San Jose



Figure 1 (from Panorska et al. 2007). The graph contains 6 boxplots of simulated distributions of L. The first five boxplots were done using 10,000 observations of L from Pareto samples of size 1,000 with a varying from 0.5 (first boxplot) to 5 (second to the last boxplot). The last boxplot corresponds to 10,000 observations of L from exponential samples of size 1,000. The inset blows up the last two boxplots.

Simulated PDFs of L

### **Redder = heavier tails**



**Figure 2 (from Panorska et al. 2006).** Log likelihood ratio (L) computed for daily excesses over local 75<sup>th</sup> percentile at each of the 560 stations. (a) Values close to zero,  $L \le 1$  (blue and green x's) represent approximately exponential tails, while yellow, red and black circles represent progressively heavier tails. (b) Level of confidence,  $(1 - \gamma)^*100$ , for rejecting the null hypothesis (H<sub>o</sub>) of exponential tails. Blue x's represent exponential tails, green x's represent stations at which the H<sub>o</sub> cannot be rejected with reasonable (90%) confidence. Yellow and progressively redder circles represent stations at which H<sub>o</sub> can be rejected with 90, 95, 98 and 99% confidence in favor of the Pareto alternative. For example, H<sub>o</sub> can be rejected at 81% of stations with 95% confidence.

Percentiles are for days with measurable precipitation. A.P. tried 50-95 percentile.



Same for the quality controlled larger station subset - 1124 stations.

This was done with a flexible choice of three thresholds (80<sup>th</sup>, 90<sup>th</sup> and 95<sup>th</sup>%-iles used in the peaks over threshold approach, see Panorska et al. 2007 for details) and six confidence levels ranging from 0.005 to 0.1.

- Small green circles represent heavy tails outcome for at least one threshold choice at lowest confidence – happens at 30% of the stations
- Larger blue circles represent heavy tails outcome for at least half of the possible threshold/confidence choices
- Big red circles represent heavy tails all around this happens at 10% of the stations



Same thing with reasonably-sized circles. Here, we can better see that heavy tails are observed only at a minority of stations, even in the northern Plains and the Mid-West, where much of the heavy-tailed behavior is observed.

On the whole, depending on threshold and significance level choice, snow accumulations at 10 - 30% of stations are heavy tailed. This is consistent with the result of Panorska et al. 2007 (figure 3a) where all precipitation (SWE in case of snow) at 37% of North American stations was estimated as heavy tailed in DJF using the 80<sup>th</sup> %-ile threshold with 95% confidence. Also, the spatial pattern of snow accumulation heavy tails generally agrees with the spatial pattern of wintertime heavy precipitation tails.

### Panorska, Gershunov, Redmond

Station	Log	P[p > 0]	$75^{\text{th}}$ %-ile( $p_{p>0}$ )	Max <sub>obs</sub> (p)	100-yr event	Pareto	
	likelihood				Exp and Pareto	$P[p > p_{exp}^{100}]$	
	ratio (L)	(%)	(mm)	(mm)	(mm)	(%)	
Sacramento	1.60	16	10.7	96	85 and 99	2.3	
Nashville	3.15	26	16	153	127 and 154	3.4	
St. Louis	4.93	30	11.2	142	114 and 144	4.1	
Houston	15.2	27	16.3	253	195 and 292	6.5	
Fargo	28.6	27	5.8	118	85 and 167	12.0	
Miami	41.8	36	13.7	377	181 and 346	9.8	

### PRECIPITATION STATISTICS AT SELECTED STATIONS

**Table 2.** Precipitation statistics at selected stations for the common observational period 1950 - 2001: L; probability of precipitation (i.e. % of days with recorded precipitation); 75<sup>th</sup> percentile of daily total on days with precipitation; maximum recorded daily total; the estimated 100-year event assuming exponential and Pareto tails; and the Pareto probability of exceeding the exponential 100-yr event. The last column can be interpreted as the factor by which the 100-yr event estimated assuming exponential tail is more likely to occur assuming Pareto tail. Alternatively, the Pareto return period for an exponential 100-yr event is 100 years divided by the value in the last column at a specific station. So, for example, in <u>Sacramento</u>, the 100-yr event estimated using the exponential assumption, can be expected, according to the Pareto assumption, to be exceeded in a 44 yr period (i.e. 100/2.3), while in Miami, it should be exceeded within 10 years (100/9.8).



### **For Helena MT ASOS Location**

## **Precipitation Frequency Data Output**

NOAA Atlas 2

Montana 46.6056°N 111.9636°W Site-specific Estimates

Мар	Precipitation (inches)	Precipitation Intensity (in/hr)
2-year б-hour	0.71	0.12
2-year 24-hour	1.25	0.05
100-year б-hour	1.69	0.28
100-year 24-hour	2.81	0.12

Hydrometeorological Design Studies Center - NOAA/National Weather Service 1325 East-West Highway - Silver Spring, MD 20910 - (301) 713-1669 Fri Jul 23 15:26:20 2010

> Using AZ example and ratios 100/2 = 2.3, so est 100yr = 2.75" 10/2 = 1.5, so est 100yr = 1.87"

### At WRCC: www.wrcc.dri.edu/pcpnfreq.html

### Western U.S. Precipitation Frequency Maps

Source: NOAA Atlas 2 published in 1973. (HDSC/NWS Office of Hydrology)

Note: To maintain image integrity and detail each image is almost 1 MB in size.

To obtain more information or the text material that accompanies these maps contact the Western Regional Climate Center at 775-674-7010

(wrcc@dri.edu)

	2 Yr, 6 Hr	5 Yr, 6 Hr	10 Yr, 6 Hr	25 Yr, 6 Hr	50 Yr, 6 Hr	100 Yr, 6 Hr	2 Yr, 24 Hr	5 Yr, 24 Hr	10 Yr, 24 Hr	25 Yr, 24 Hr	50 Yr, 24 Hr	100 Yr, 24 Hr
Arizona	$\checkmark$											$\checkmark$
Northern California												
Southern California												
Colorado	$\checkmark$								$\checkmark$			$\checkmark$
Idaho	$\checkmark$											
Montana	$\checkmark$					$\checkmark$			$\checkmark$	$\checkmark$	$\square$	$\checkmark$
Nevada	$\checkmark$					$\checkmark$			$\checkmark$	$\checkmark$		$\checkmark$
New Mexico	$\checkmark$		$\checkmark$			$\checkmark$			$\checkmark$			$\checkmark$
Oregon	$\checkmark$					$\checkmark$			$\checkmark$	$\checkmark$		$\checkmark$
Utah	$\checkmark$					$\checkmark$			$\checkmark$	$\checkmark$	$\checkmark$	$\checkmark$
Washington	$\checkmark$					$\checkmark$			$\checkmark$	$\checkmark$	$\square$	$\checkmark$
Wyoming	$\checkmark$											$\checkmark$

Western Regional Climate Center, wrcc@dri.edu



### **10-year 24-hour precipitation in tenths of inches**

#### RD 22 DHÁ 26 -30-24 -28 32 32 634 87 30 26 ELEN 24 3 EAK) 24 26-28 ARLOW 30 24 30 BIG TIMBER THREE WRKS GSTON (20 628 1618 18 18 18 18 14 20 22 20 18 2022 26 30 34 22201816 + 20 52000 2220 1 22

### More detail ... 10-year 24-hour precipitation in tenths of inches



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#### **NOAA HDSC**

Duration			
5-min ——	120-m <del>-</del>	48-hr <del>-×-</del>	30-day ———
10-min 🔶	3-hr <del>-*</del> -	4-day <del>-</del>	45-day 🛶
15-min 🕂	6-hr 🔶	7-day 🔶	60-day 😽
30-min <del></del>	12-hr 🕂	10-day ——	
60-min <del>-×-</del>	24-hr <del>-8-</del>	20-daý <del>-</del> ⊟-	



#### NOAA's National Weather Service NOAA Hydrometeorological Design Studies Center Home Site Map News Organization General Info Current NWS Precipitation Frequency (PF) Documents and Related Studies Homepage Current Projects **HDSC** 1. PF documents Precipitation 1.1 PF documents by state/territory and duration Precipitation Frequency (PF) 1.2 PF documents by title Frequency PF Data Server 2. PF related studies Status of PF Documents Viewing the documents requires the Adobe Acrobat Reader (click here to download). **Products** Probable Maximum Precipitation (PMP) As of PMP Documents 1. PF documents 2011 May 19 Record Precipitation 1.1 PF documents by state/territory and duration Duration (D) Contact Us State/Territory 1 hr ≤ (<) D ≤ 24 hr D > 24 hr D < (≦) 1 hr Inquiries Contiguous U.S. List-server Alabama Tech Memo HYDRO-35 (1977) Technical Paper 40 (1961) Technical Paper 49 (1964) Arizona NOAA Atlas 14, Vol 1 (2004) NOAA Atlas 14, Vol 1 (2004) NOAA Atlas 14, Vol 1 (2004) USA.gov Arkansas Tech Memo HYDRO-35 (1977) Technical Paper 40 (1961) Technical Paper 49 (1964) NOAA Atlas 14, Vol 6 (2011) -

NOAA Atlas 14, Vol 6 (2011)

NOAA Atlas 2, Vol 3 (1973)

Technical Paper 40 (1961)

NOAA Atlas 14, Vol 2 (2004)

Coming in May

Technical Paper 49 (1964)

Technical Paper 49 (1964)

NOAA Atlas 14, Vol 2 (2004)

NOAA Atlas 14, Vol 6 (2011)

Tech Memo HYDRO-35 (1977)

NOAA Atlas 14, Vol 2 (2004)

Arkell & Richards (1986)

FAQ

California

Colorado

Delaware

Connecticut

**Potential sources of heavy precipitation:** 

**Frontal passages** 

**Cyclonic storms** 

**Orographic uplift** 

**Tropical storms** 

**Convergence zones** 

**Embedded cumulus** 

**Atmospheric rivers** 

**Blocked atmospheric patterns** 

**ENSO** conditioning

**MJO activity** 

Human effects (?)

### World record precipitation locations.

### ... A variety of circumstances



**NOAA HDSC** 



To produce precipitation, and heavy precipitation:

It is not sufficient that there be abundant atmospheric moisture (humidity). The main issue is that there must be a mechanism to wring out moisture. (Examples: Sudan, Kuwait)

It is also not necessary that there be abundant atmospheric moisture. If dry, there must be advective mechanisms to replenish moisture. (Example: Heavy winter snows near Lethbridge/Banff at -20 F)

Heavy precipitation can result from

Abundant local moisture

But almost always also requires (especially at time scales > a few hours)

Replenishment and import as local supply is consumed


Redmond, K.T., Y. Enzel, P.K. House, and F. Biondi, 2002. Climate variability and flood frequency at decadal to millennial time scales. pp. 21-45, in *Principles and Applications of Paleoflood Hydrology*, editors: P.K. House, R.H. Webb, and V.R. Baker, American Geophysical Union, 385 pp.



Baker, American Geophysical Union, 385 pp.

"Stationarity is dead"

Stationarity was never really fully alive.

\*

"The history of climate is a nonstationary time series." \*

Corollaries: There are no true climatic "normals". We never know enough. We can never stop observing.

- \* P.C.D. Milly, Julio Betancourt, Malin Falkenmark, Robert M. Hirsch, Zbigniew W. Kundzewicz, Dennis P. Lettenmaier, Ronald J. Stouffer, 2008. Stationarity is dead: Whither water management?. Science, 319 (5863), 573-574, 1 Feb 2008.
- \* Reid A. Bryson, 1997. The Paradigm of Climatology: An Essay. Bulletin of the American Meteorological Society, 78(3), 449-455.

Stationarity, if even alive, is not feeling well ... "under the weather"

Climate change: The "present future" will slowly depart from its "prior future" Stationarity slowly but progressively becoming a less valid assumption

Evidence points this way ...... but is not completely unambiguous How much until this departure is "significant" ?

(not so much in statistical terms, but in practical terms) How do we adjust all the statistics of the past to reflect the expected future?

#### A thought experiment

Suppose we had a perfect (complete and accurate) observed time series of a climate element indefinitely far back into the past from a location of interest. e.g. 1-minute measurements for the last 100, thousand, 10000, 100000, million years

We wish to make a decision about some future time interval The expected lifetime of a culvert The expected lifetime of a railroad bridge The expected lifetime of a waste settlement lagoon The expected lifetime of a major dam In many cases, the interval may consist of the time until we next revisit the issue (because human infrastructure is constantly being reshaped for many purposes)

The big question :

#### What part of the past is relevant to what part of the future?

How many years would we go back, for a decision related to how many years in the future?

The usual answer:

There's never enough data. Beggars can't be choosers. More is always better.



FIG. 2. Time series of the percent contribution of the upper 10 percentile of daily precipitation events to the total annual precipitation area-averaged across the United States. Smooth curve is a nine-point binomial filter, and the trend is also depicted.



#### **Trends in Extreme Daily Precipitation, 1910-1996, by category.**

Thomas R. Karl and Richard W. Knight, 1998. Secular Trends of Precipitation Amount, Frequency, and Intensity in the United States. Bulletin of the American Meteorological Society, 79 (2), 231–241.



Fig. 5. The contribution to the trends in Fig. 1 attributed to trends in precipitation intensity. Trends are expressed as in Fig. 1.

### Karl and Knight, 1998.

Contribution to change in total precipitation from different intensity classes.

1910-1996.

-- The intensity of precipitation has increased for very heavy and extreme precipitation days only.

-- The proportion of total annual precipitation derived from heavy and extreme precipitation events has increased relative to more moderate precipitation.

# Trends in heavy (5 pct), very heavy (1 pct), and extreme (0.1 pct) daily precipitation. 1910-1999, USA.

TABLE 1. Trend characteristics in annual precipitation totals; in heavy (upper 5%), very heavy (upper 1%), and extreme (upper 0.1% of daily rain events) precipitation totals; and in the fraction of total precipitation occurring in heavy, very heavy, and extreme precipitation events over the contiguous United States, 1910–99. Asterisks (\*) indicate trends that are statistically significant at the 0.05 or higher level.

Precipitation		Annual precipitation		Contribution to annual totals			
		Linear t	rend	Fraction	Relative change		
	Mean value (mm)	Estimate [% (10 yr) <sup>-1</sup> ]	Variance (%)		Estimate [% (10 yr) <sup>-1</sup> ]	Variance (%)	
Total Heavy Very heavy Extreme	750 195 62 12	0.6 1.7 2.5 3.3	5* 12* 15* 11*	1.00 0.26 0.08 0.016	1.0 1.9 2.7	20* 17* 9*	

Pavel Ya. Groisman, Richard W. Knight, David R. Easterling, Thomas R. Karl, Gabriele C. Hegerl, and Vyacheslav N. Razuvaev, 2005. Trends in Intense Precipitation in the Climate Record *Journal of Climate*, 18 (9), 1326–1350.

## Trends in number of days with heavy (5 pct), very heavy (1 pct), and extreme (0.1 pct) daily precipitation. 1910-1999, USA.

TABLE 2a. Trend characteristics in the number of days with heavy and very he	eavy precipitation over the contiguous United States,
1910-99 (percentile definition). Asterisks (*) indicate trends that are sta	tistically significant at the 0.05 or higher level.

	Day	s with precipitati	on	Contri pree	ibution to total days with cipitation above 1 mm	
		Linear t	rend	Fraction	Relative change	
Events	Mean (days yr <sup>-1</sup> )	Estimate [% (10 yr) <sup>-1</sup> ]	Variance (%)		Estimate [% (10 yr) <sup>-1</sup> ]	Variance (%)
Total days with precipitation above 1 mm Heavy (upper 5% of precipitation events) Very heavy (upper 1% of precipitation events)	75 4.4 0.88	0.5 1.5 2.2	6* 12* 14*	1 0.06 0.012	1.0 1.7	11* 13*

Pavel Ya. Groisman, Richard W. Knight, David R. Easterling, Thomas R. Karl, Gabriele C. Hegerl, and Vyacheslav N. Razuvaev, 2005. Trends in Intense Precipitation in the Climate Record *Journal of Climate*, 18 (9), 1326–1350.



tions of duration and return period. The time series have

been smoothed with a 7-yr moving average filter. Return

periods of 1 year (red), 5 years (blue), and 20 years (orange)

are plotted on each graph.



5-Day

10-Day

**30-Day** 



Figure 1. Location of stations with less than 10% missing daily precipitation data for 1895–2000. The symbol 'o' (in blue) indicates that long-term data were available prior to CDMP while the symbol 'x' (in red) indicates newly available long-term stations.

Extreme Precipitation Index
United States
1895-2000.
Selected durations
And
Return periods (1, 5, 20 yrs)
(Station density effects removed)

Ken E. Kunkel, Dave R. Easterling, Kelly T. Redmond, and Ken G. Hubbard, 2003.

Temporal variations of extreme precipitation events in the United States: 1895-2000.

Geophysical Research Letters, 30:1717.

#### West Coast Precipitation. Annual Series of 2-Day Maximum.

#### Number of events per decade.



#### Trends 1950-2009.



Figure 3: Sixty-year trends for the top 60 events for all stations (a) and for averages of the stations over 2° latitude bands (b). The trends are given in percent change relative to the mean over the 60-year period. The vertical dotted lines indicate the state boundaries and the brackets indicate the 95% interval derived from using formula (1).

Figure 4: Sixty-year trends for the top 20 events for all stations (a) and for averages of the stations over 2° latitude bands (b). The trends are given in percent change over the 60-yr period. The vertical dotted lines indicate the state boundaries and the brackets indicate the 95% interval derived from using formula (1).



Trends 1950-2009 Annual Max 2-Day Precipitation

Units in mm/day rather than percentages

Figure 5: Sixty-year trends (mm per year) for the maximum annual two-day precipitation for coastal locations from southern California to British Columbia. The vertical dotted lines indicate the state boundaries and the brackets indicate the 95% confidence interval derived from using formula (1).

#### West Coast Streamflow. Annual Series of 1-Day Maximum Discharges.



Number of events per decade.

Trends 1950-2009.

Figure 7: Decadal variation of the top 20 and 60 average daily discharges for 1950-2009 for unregulated rivers north and south of 45°N latitude.



Young, Zieger, Babanin, 2011. Global Trends in Wind Speed and Wave Height. Science, 332 (6028), 451-455.



#### **Satellite Winds and Waves**

7 satellites since 1985, analysis thru 2008 GEOSAT difficulties with wind, so record starts 1991, thru 2008 Separate distributions each month, thus, seasonal cycle to remove 1 billion altimeter observations, 90,000 per 2x2 deg grid, 300 per month Very well correlated with buoy data of wind and waves RMS error of satellite-derived waves less than 0.2 meter RMS error of satellite-derived winds less than 1.5 m/s for 10-m wind

Rayleigh distribution, Hs=10 m: .1 > 10.7 m, .01>15.1 m, .001>18.6 m



#### Statistical Wave Distribution



Figure S6: Colour contour plot of mean monthly wind speed trend (% per annum) from the NCEP/NCAR reanalysis data (25). Points which are statistically significant according to the SK test are shown with dots. This figure can be directly compared with Fig. 1.

#### 90<sup>th</sup> Pctl Wind Speed Trend <sup>90th percentile wind speed (1991–2008)</sup> 1991-2008.



90th percentile significant wave height (1985-2008)





#### 99<sup>th</sup> Pctl Wind Speed Trend<sub>99th percentile wind speed (1991-2008)</sub> 1991-2008.

#### Table 1

Comparison of trend estimates for buoy and altimeter data. The top panel shows wind speed and the bottom panel shows wave height, with the locations grouped by geographic region. Bold values are statistically significant at the 95% level (bold and underscored) and at the 90% level (bold) where two significance tests were passed (the normal distribution and the homogeneity test) (SOM).

				Buoy trend (cm/s/year)			Altimeter trend (cm/s/year)		
				Mean	90th	99th	Mean	90th	99th
Gulf of Mexico	42001	25.9	89.7	1.79	3.00	4.53	0.57	4.50	10.11
	42002	25.8	93.7	1.88	3.07	<u>6.29</u>	1.13	0.00	0.69
North Atlantic	44004	38.5	70.4	<u>4.01</u>	4.42	7.34	0.57	2.41	10.94
	44011	41.1	66.6	0.48	2.46	4.63	0.51	2.16	<u>13.03</u>
	41002	32.4	75.4	3.66	7.50	12.99	-0.47	2.21	10.73
North Pacific	46001	53.3	148.0	2.90	<u>4.93</u>	7.56	<u>5.33</u>	7.46	10.54
	46002	42.6	130.5	<u>1.99</u>	2.14	2.83	<u>3.24</u>	5.25	10.42
	46005	46.1	131.0	4.02	<u>6.43</u>	<u>8.89</u>	4.26	<u>5.50</u>	13.76
	46006	40.9	137.5	3.52	<u>4.15</u>	12.70	2.45	3.33	10.04
	46035	57.1	177.8	<u>5.62</u>	10.00	<u>9.08</u>	1.06	-0.61	0.02
Hawaii	51001	23.5	162.3	2.86	<u>3.59</u>	4.42	<u>3.99</u>	2.77	4.96
	51002	17.1	157.8	2.12	1.40	0.92	2.90	3.63	<u>6.43</u>
Gulf of Mexico	42001	25.9	89.7	0.24	0.00	1.42	-0.41	0.43	<u>2.41</u>
	42002	25.8	93.7	0.55	0.50	1.00	-0.44	0.24	1.46
North Atlantic	44004	38.5	70.4	0.14	0.40	1.27	-0.54	0.51	2.74
	44011	41.1	66.6	0.42	<u>1.11</u>	1.47	0.34	1.64	<u>5.20</u>
	41002	32.4	75.4	-0.05	0.00	0.54	-0.41	-0.02	2.82
North Pacific	46001	53.3	148.0	-0.45	0.00	0.50	0.08	1.24	<u>3.03</u>
-	46002	42.6	130.5	0.06	0.00	-0.06	0.01	0.58	2.59
	46005	46.1	131.0	0.36	0.00	1.84	0.42	<u>1.67</u>	<u>4.50</u>
	46006	40.9	137.5	0.98	1.25	1.61	-0.21	0.24	2.64
	46035	57.1	177.8	-0.31	-0.95	-2.54	-0.36	0.84	2.59
Hawaii	51001	23.5	162.3	-0.71	-0.71	-0.65	-0.88	<u>-0.95</u>	-0.06
	51002	17.1	157.8	0.02	0.00	-0.51	-0.16	0.27	0.66

Young, Zieger, Babanin, 2011.



Soda Springs Store March 27, 2011

#### Tom Knudson Sacramento Bee







## South Central Sierra Snow Lab

East

Photo: Dave Simeral



## A Key HMT Finding:

 atmospheric rivers are a key to extreme precipitation and flooding, as well as water supply and stream flow on the U.S. West Coast

Examples of AR events that produced extreme precipitation on the US West Coast, and exhibited spatial continuity with the tropical water vapor reservoir as seen in SSM/I satellite observations of IWV.

#### **Thanks to Marty Ralph**

#### November 6-7, 2006. One storm, three state daily precipitation records !!! Never in U.S. history have two state records been set on the same day.

## Preliminary State-Record Maximum 24-Hour Precipitation

New Record: Former Record: June Lake, Washington, elevation 3,340 feet Mt. Mitchell, Washington, elevation 3,600 feet

New Record: Former Record: Lee's Camp, Oregon, elevation 660 feet Port Orford, Oregon, elevation 150 feet

New Record: Bear Mountain, Idaho, elevation 5,400 feet Former Record: Rattlesnake Creek, Idaho, elevation 4,000 feet



15.20 inches on November 6-7, 2006 14.26 inches on November 23-24, 1986

14.30 inches on November 6-7, 2006 11.65 inches on November 19, 1996

9.40 inches on November 6-7, 2006 7.17 inches on November 23, 1909

Old records broken by

- 23 % Oregon
- 06 % Washington
- 31 % Idaho

## Square:

Many Glacier Lodge





**Fig. 3.** Trends in satellite-derived P - E for the period July 1987 through August 2006. The largest change was over the warm pool in the western Pacific: a wet area that became wetter.

**Table 1.** Statistics on the variation of global evaporation, global precipitation, and over-ocean water vapor for the period July 1987 through August 2006. The error bars on the trends are given at the 95% confidence level. The values in parentheses are in terms of percentage change, rather than absolute change.

Parameter	Mean	Standard deviation	Trend
Evaporation	961 mm year <sup>-1</sup>	10.1 mm year <sup>-1</sup> (1.1%)	$12.6 \pm 4.8 \text{ mm year}^{-1} \text{ decade}^{-1}$ (1.3 ± 0.5% decade <sup>-1</sup> )
Precipitation	950 mm year <sup>-1</sup>	12.7 mm year <sup>-1</sup> (1.3%)	$13.2 \pm 4.8 \text{ mm year}^{-1} \text{ decade}^{-1}$ (1.4 ± 0.5% decade <sup>-1</sup> )
Total water	28.5 mm	0.292 mm (1.0%)	$0.354 \pm 0.114 \text{ mm decade}^{-1}$ (1.2 ± 0.4% decade <sup>-1</sup> )

Frank J. Wentz, Lucrezia Ricciardulli, Kyle Hilburn, Carl Mears, 2007. How much more rain will global warming bring? Science, 317, 233-235.

## The World's Warm Oceans



**Thru February 2011** 

**Typically "El Nino"** 



Typically "La Nina"

Courtesy Klaus Wolter & Mike Timlin, NOAA Climate Diagnostics Center

#### **Global Sea Surface Temperature Departure from Average, 2011 March 6-12**



**NOAA Physical Science Division, ESRL** 







Temporal Variability of

Orographic Effect on Precipitation

Sacramento (10') Versus Tahoe City (6230')

July thru June Oct-March Percent of Annual:

83% at Tahoe










-200 -100

0 100 200

P(N) 970205

**Thru February 2011** 

**Typically "El Nino"** 



Typically "La Nina"

Courtesy Klaus Wolter & Mike Timlin, NOAA Climate Diagnostics Center



#### PDO index values for 2009



Percent change in annual precipitation and in amount on wettest day of the year, when both models agree on sign. 2 x CO2. Canadian Model 1 & Hadley Model 3.

## Annual amount

#### Wettest day



FIG. 13. Consensus estimates of changes in mean annual precipitation in the  $2 \times CO_2$  experiments from CGCM1 and HadCM3 GCMs over North America. The red end of the scale depicts decreases and the blue increases. The pattern shows the average precipitation change between the models, it is only shown where the simulations with each model are consistent with the respective other model at the gridpoint level.

Pavel Ya. Groisman, Richard W. Knight, David R. Easterling, Thomas R. Karl, Gabriele C. Hegerl, and Vyacheslav N. Razuvaev, 2005. Trends in Intense Precipitation in the Climate Record *Journal of Climate*, 18 (9), 1326–1350.

### **GFDL 2.1 A2 Emissions Scenario**

Ratio of Maximum Daily Precipitable Water (SRESA2 2071 to 2100 / 20C3M 1961 to 1990) Note: Precipitable waters derived from model specific humidity, 20C3M = mean of ensembles 2,4,5



## **GFDL 2.1 A2 Emissions Scenario**

Ratio of Maximum Daily Precipitable Water (SRESA2 2071 to 2100 / 20C3M 1961 to 1990) Note: Precipitable waters derived from model specific humidity, 20C3M = mean of ensembles 2,4,5



# Urbana Illinois IDF Curves and Probable Maximum Precipitation Present and Future Estimate from Clausius-Clapeyron Considerations





Some issues for consideration -1

How adequate are our observing systems to record the extremes of interest?

How adequate is our tracking of extremes? Ability to notice changes with relatively little delay Helps contribute to awareness of "changing extremes" possibilities

Extremes in hydrologic drivers versus extremes in hydrologic response Hydrologic systems as "filters" Land use change as an example Agriculture Urban and impermeable surfaces Same atmospheric sequence may not produce same hydro consequences

How well do / can models represent the properties of extremes? Ability to represent the entire probability distribution Ability to represent return intervals correctly Can models produce heavy-tailed distributions

Wet hydrologic extremes on short time scales exhibit heavy tails We may be underestimating the size and frequency of rare events

**Orographic ratio connections to large scale climate** 

Some issues for consideration - 2

Atmospheric rivers as a weather - climate connection

Better understanding of MJO (Madden-Julian Oscillation) relationships to ENSO and other oscillations West Coast heavy precipitation probabilities Windows of opportunity that may or not be realized (as heavy precip)

Multi-day duration episodes Slow-moving cutoff systems Sequences of multiple rapidly translating systems

The importance of a few singular events Increases from north to south Lack of a few events sets the stage for drought

Effect of non-precipitation hydrologic elements Changes in hydrologic events resulting from changes in temperature

Decadal-scale climate variability of extreme event likelihood

Social science investigations relating to risk perception Improved societal appreciation/understanding of probabilistic concepts "The improbable is bound to happen one day."

- Emil Gumbel

"The biggest one is yet to come."

- Jim Goodridge CA DWR

