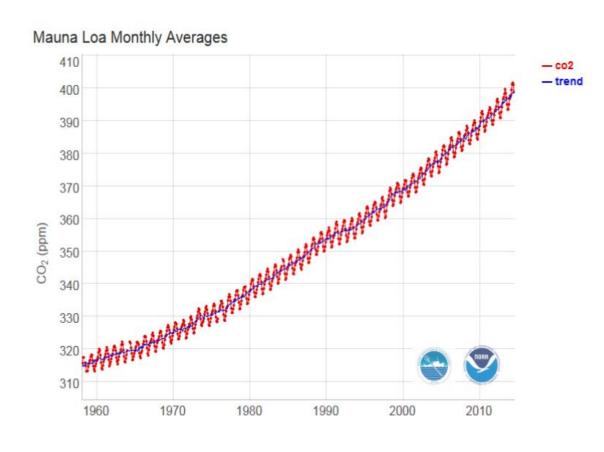
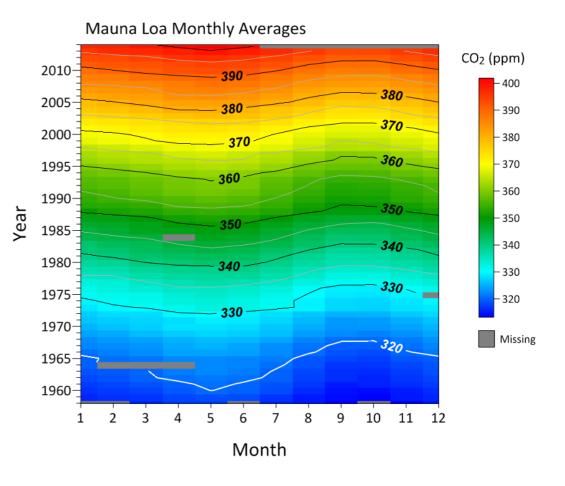


Innovative Ways to Visualize and Analyze Environmental Time-Series Data



Brown Bag webinar NOAA Central Library August 20, 2014



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Webinar outline

- Visualization criteria
- Composition vs configuration metrics
- Examples and applications
- Analysis techniques
- Users
- Questions



Criteria for time-series visualization

Show all data

- **Display large datasets**
- Handle wide data ranges
- Clearly show each data point
- Don't hide/cover/clump points

Identify patterns

- Allow for easy comparisons
- Spot outliers and artifacts
- Show both short and long-term patterns
- Identify natural vs artificial patterns



Data sources and challenges

Observations

Data collected for any number of environmental sciences Hydrology, engineering, oceanography, biology/ecology, meteorology, limnology, ...

Model output

HEC-RAS, HEC-GeoRAS, HEC-HMS, PHABSIM, SEFA, ELOHA, WaterFALL, WFET, SWAT, CrEAM, ABF, NEABF, SAM2, IHA,...

Big Data

Obstacles to understanding Identify temporal signatures Need different display techniques



Time-series metrics (for hydrology)

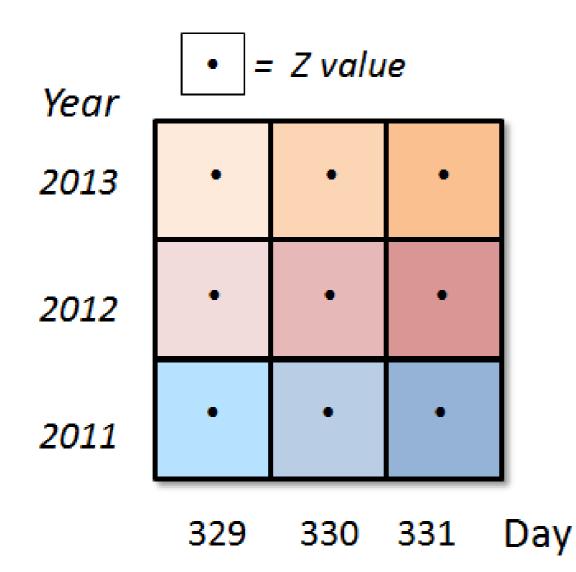
- Most indices are simple¹
- Need time-scale analysis to find patterns¹
- Hundreds of indices exist²

Magnitude	55%	Composition	
Frequency	8%		
Duration	26%		Temporal
Timing	6%	Configuration	component
Change	5%		

Visualization overlooked as analysis method

¹ Long, K. S. 1994 ² Olden, J.D., and N.L. Poff. 2003

Temporal map approach (raster graph)

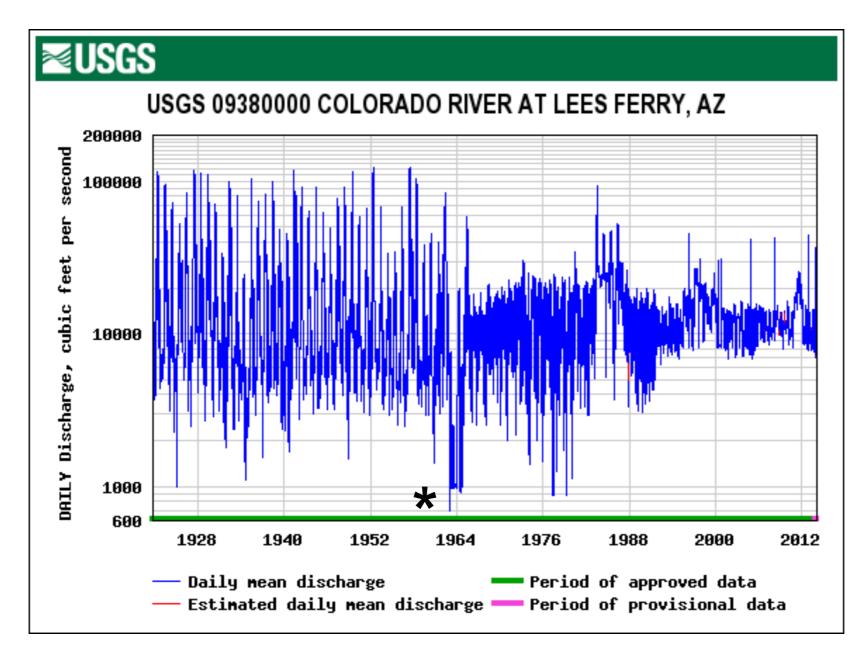


X = Short-term coordinate

Y = Long-term coordinate

Traditional hydrograph

92 years daily data (33,700 points)

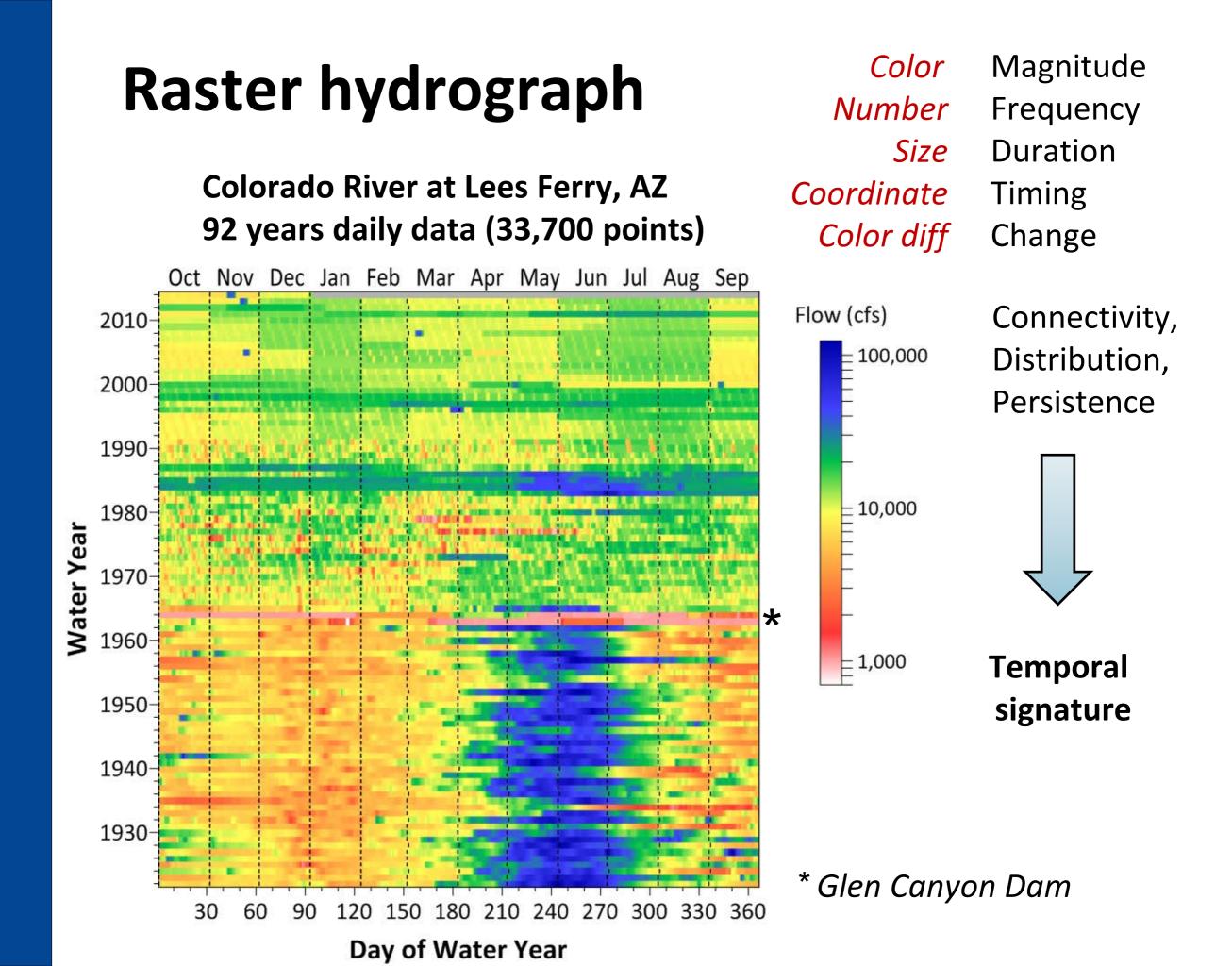




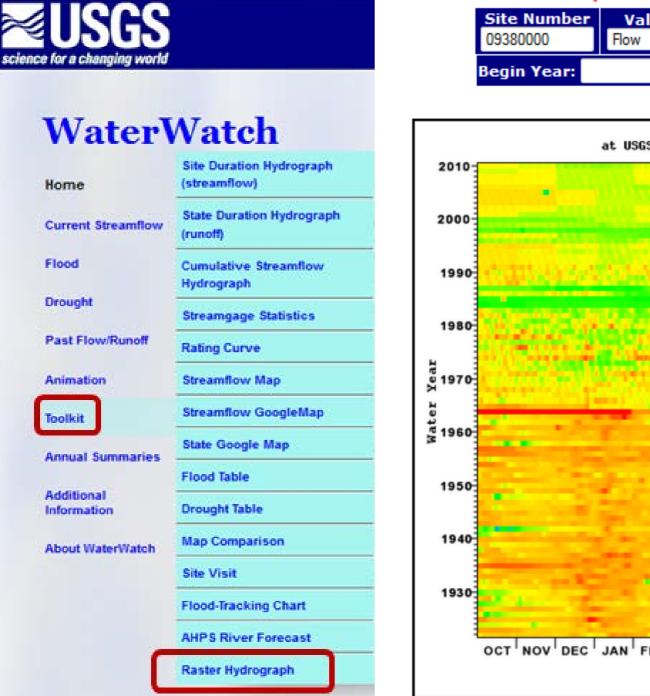
* Glen Canyon Dam

Raster hydrograph

Colorado River at Lees Ferry, AZ 92 years daily data (33,700 points) Temporal Oct Nov Dec Jan Feb Mar Apr May Jun Jul Aug Sep patterns Flow (cfs) 2010 9 Snowmelt runoff $\pm 100,000$ 1. 2000 Drought 2. 10Storm flow 3. 1990-Vegetation signal 4. Tunnels closed 5. $\pm 10,000$ 6. El Nino 1980 6 Water Year 7. Artificial flood Sundays 8. 1970 Christmas 9. * 10. Monthly change 1960 = 1,000 3 1950 1940-4 2 1930-* Glen Canyon Dam 60 90 120 150 180 210 240 270 300 330 360 30 **Day of Water Year**



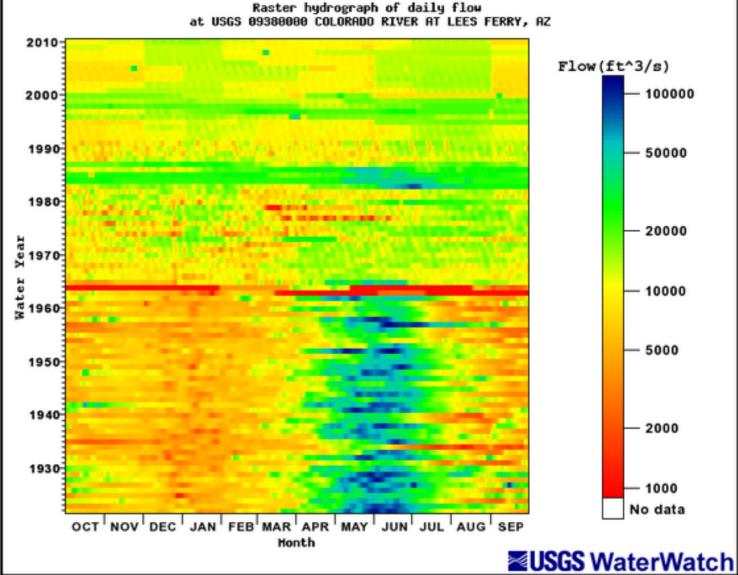
Adopted by the USGS



Streamflow Raster-Hydrograph Builder

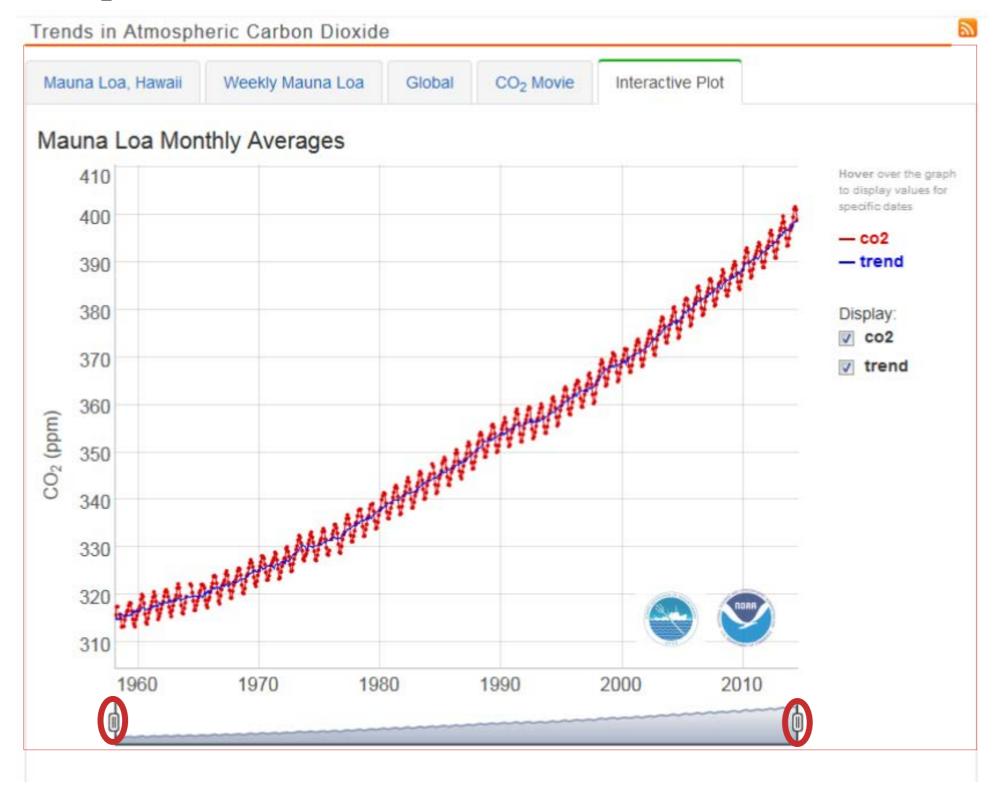
(Warning: It may take several minutes to process)

Site Number	Value type		Flow type		Year type	
09380000	Flow	-	Daily streamflow	/ •	Water year 🛛 👻	GO
Begin Year:		End Yea	r:	Leger	nd Unit: <mark>ft^3/s</mark>	•



Traditional display

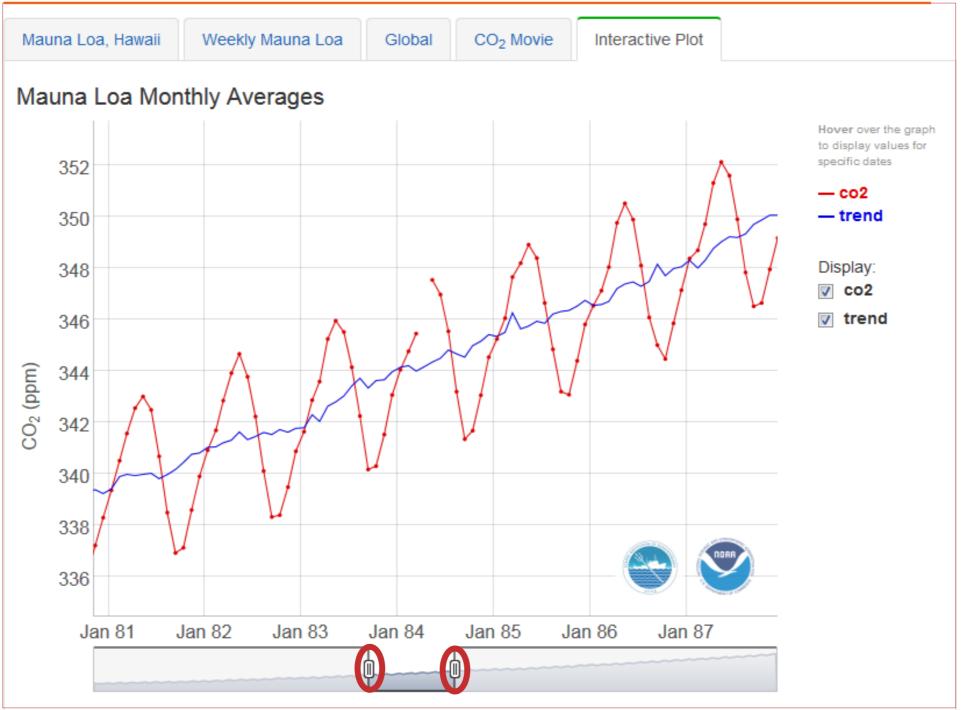
CO₂ dataset



Zoom feature

CO₂ dataset

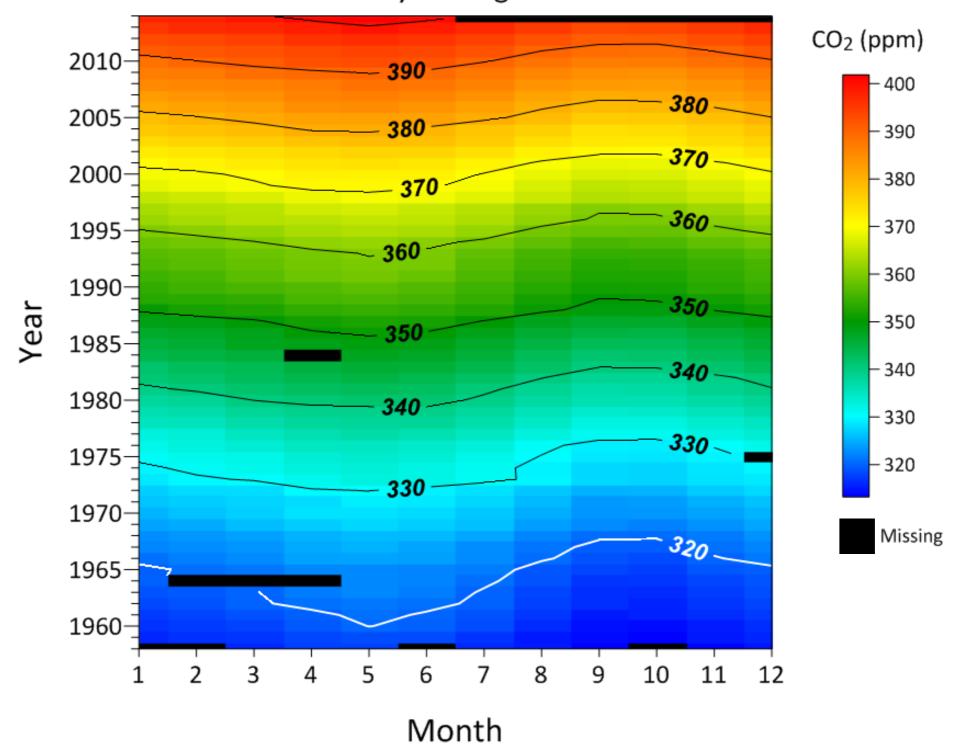
Trends in Atmospheric Carbon Dioxide

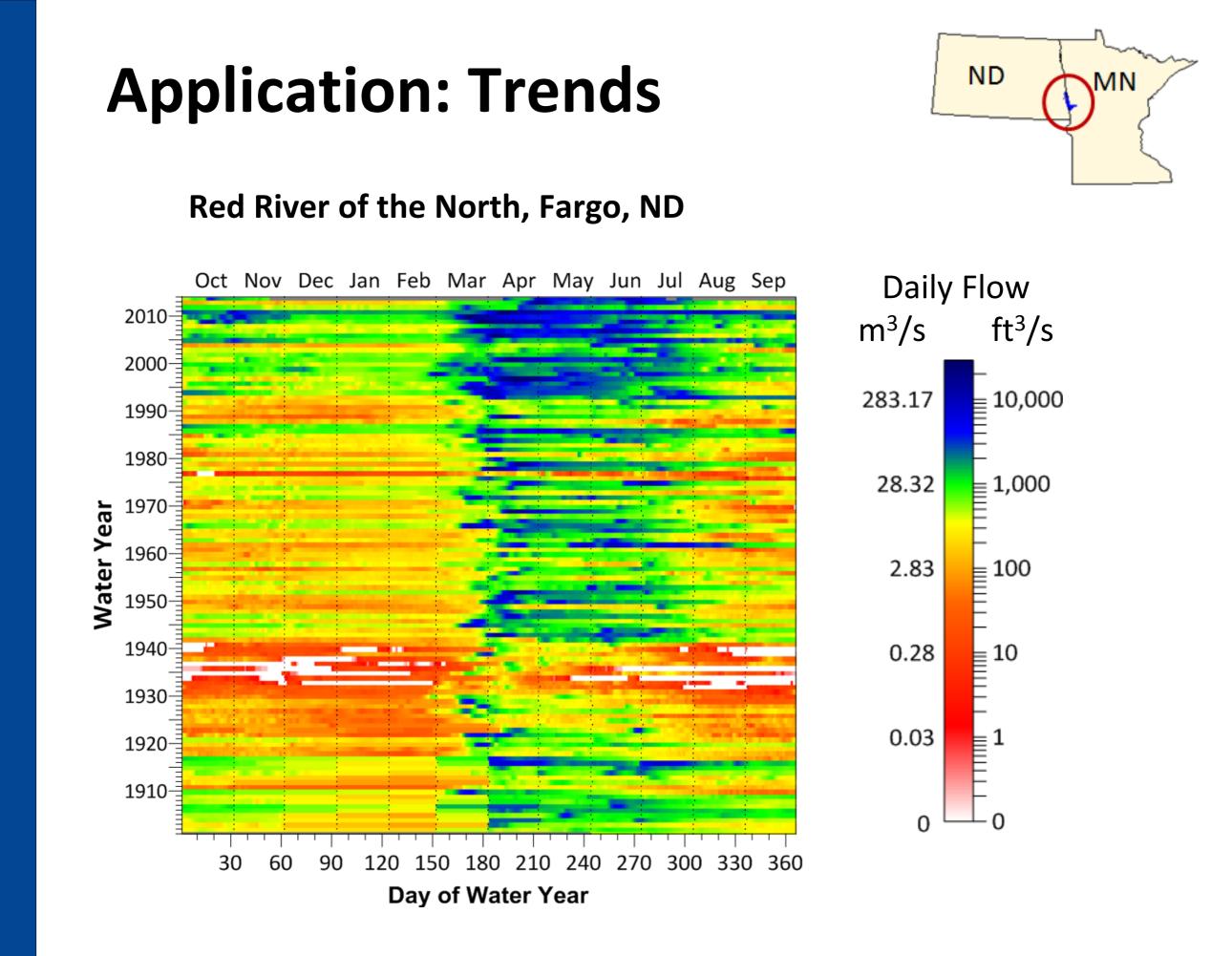


Application: Raster display

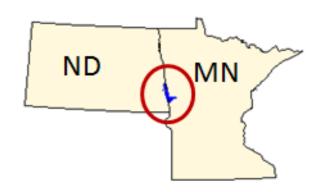
CO₂ dataset

Mauna Loa Monthly Averages

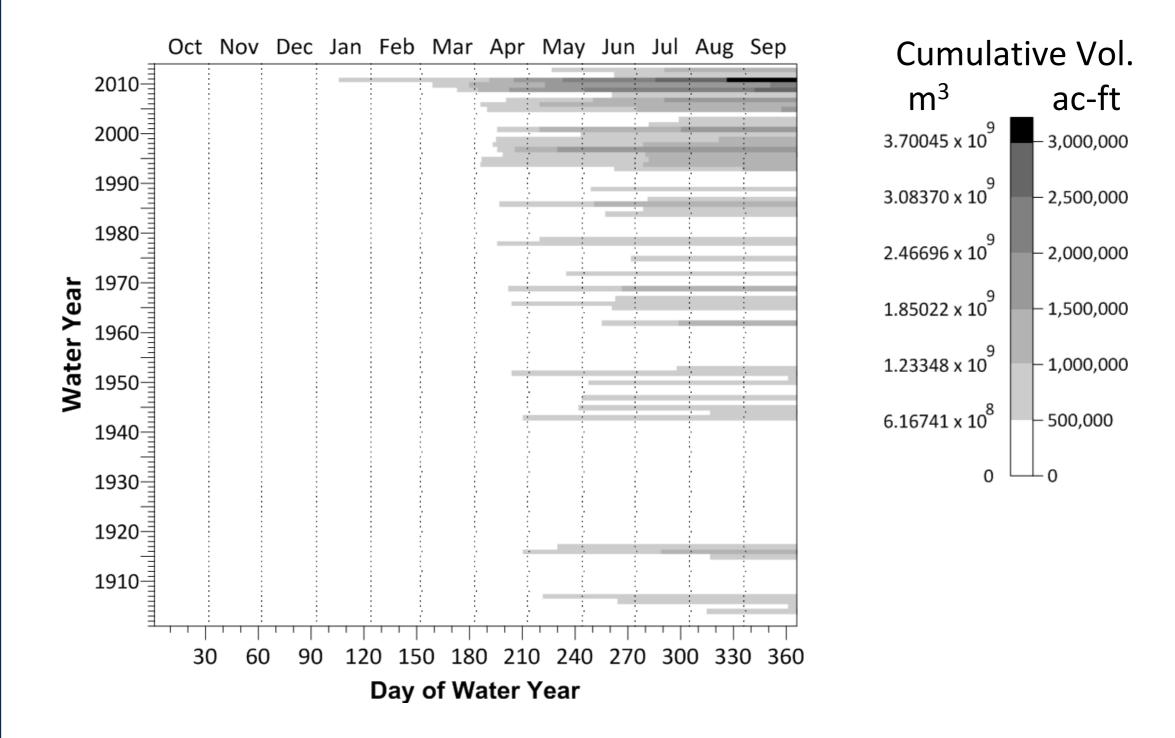




Application: Yearly volume

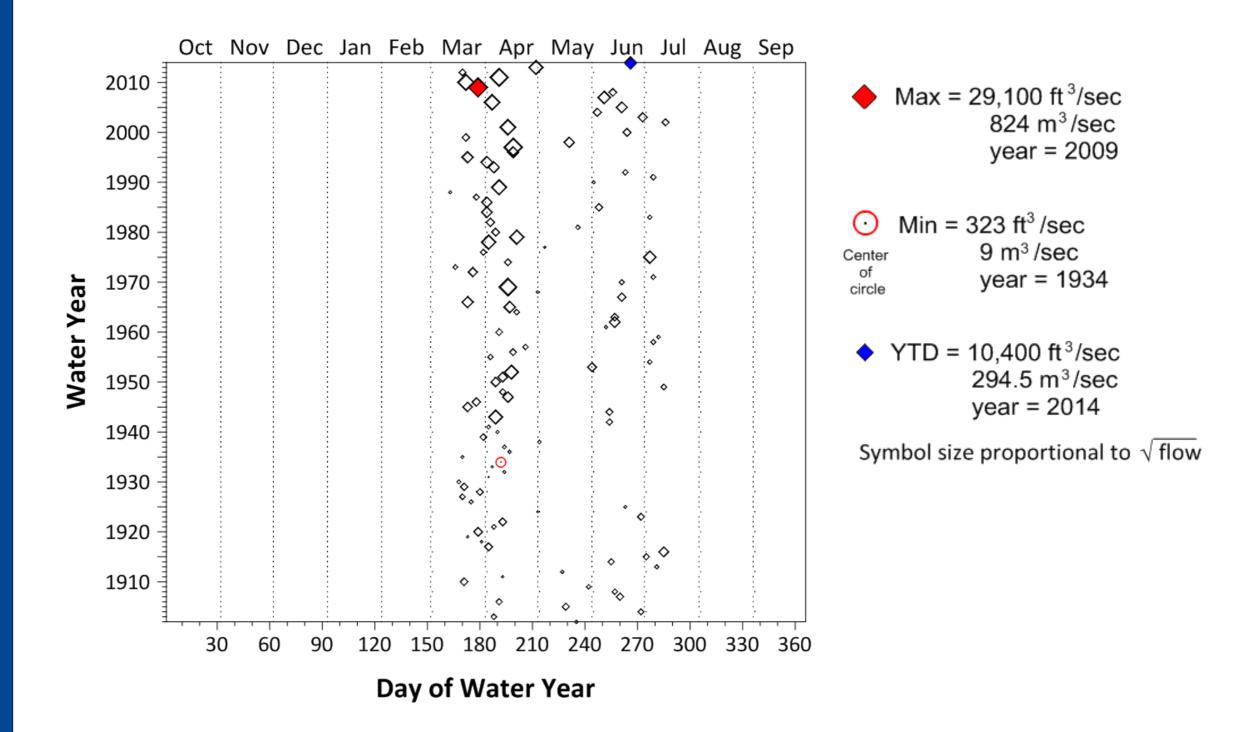


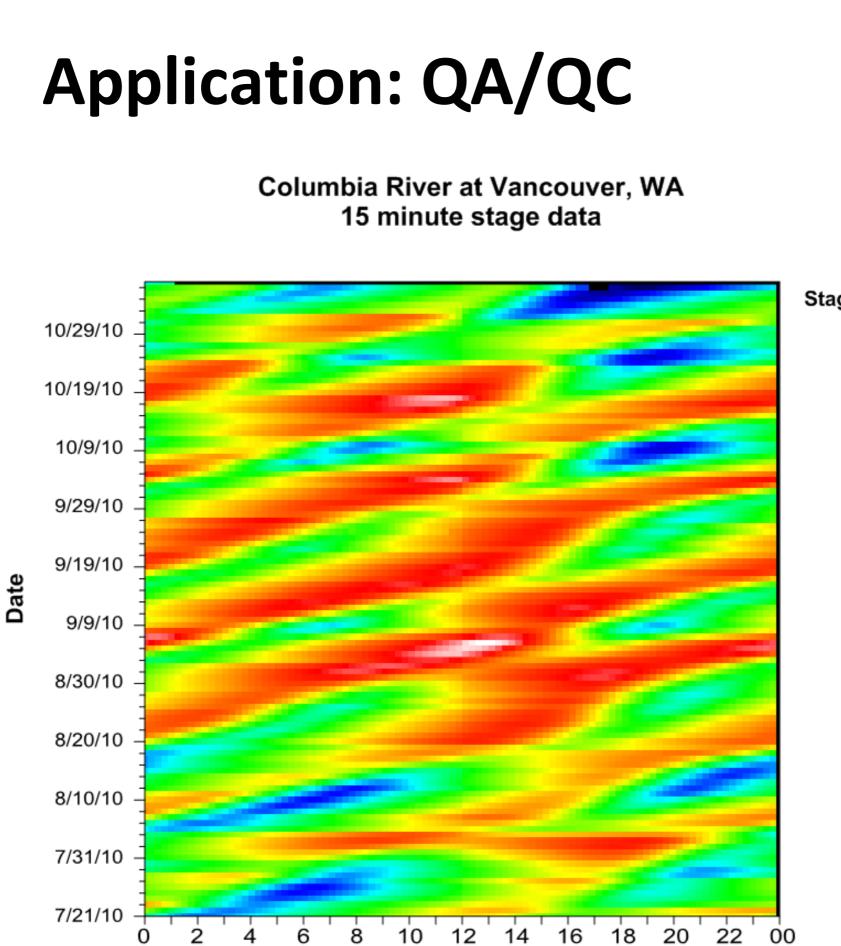
Red River of the North, Fargo, ND



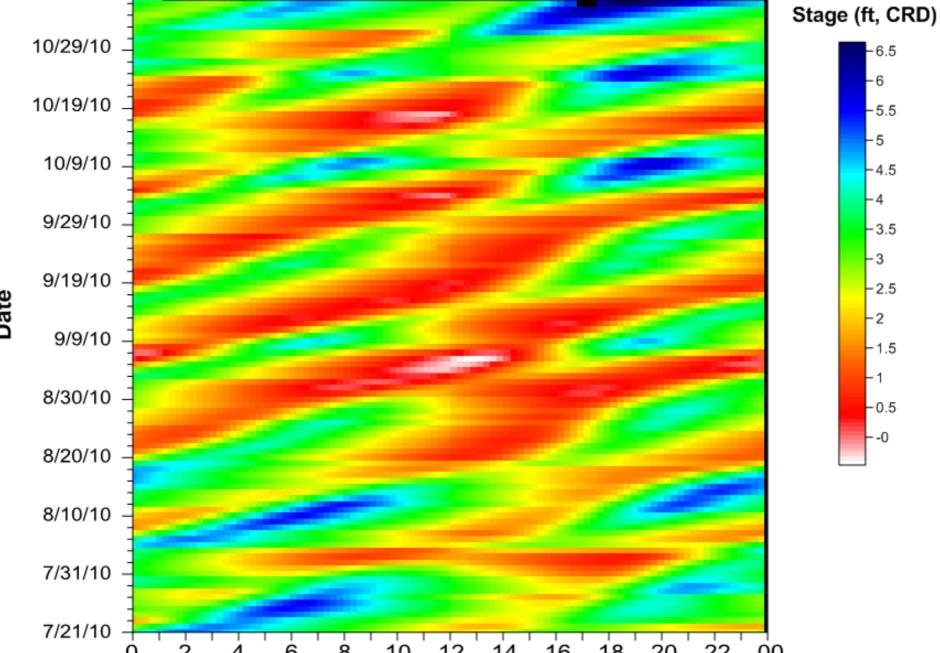
Application: Annual maximum flow

Red River of the North, Fargo, ND





Columbia River Snake River OR

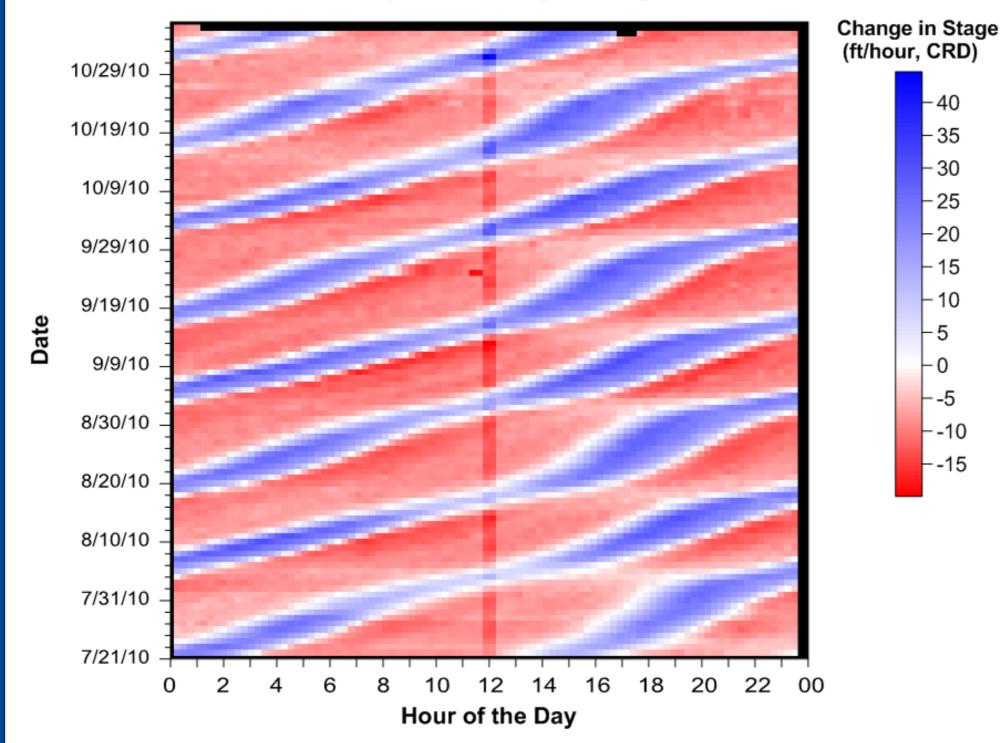


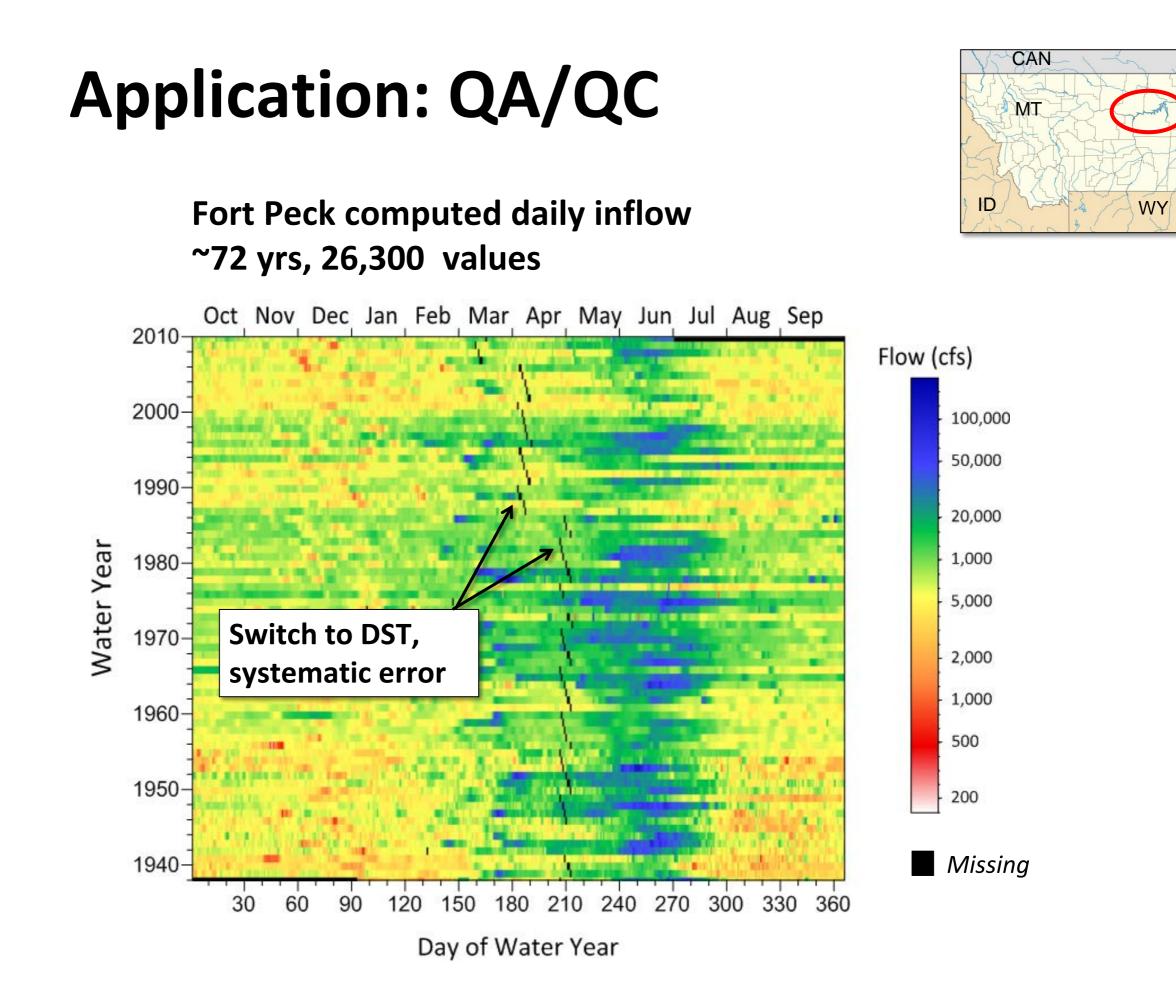
Hour of the Day

Change in stage



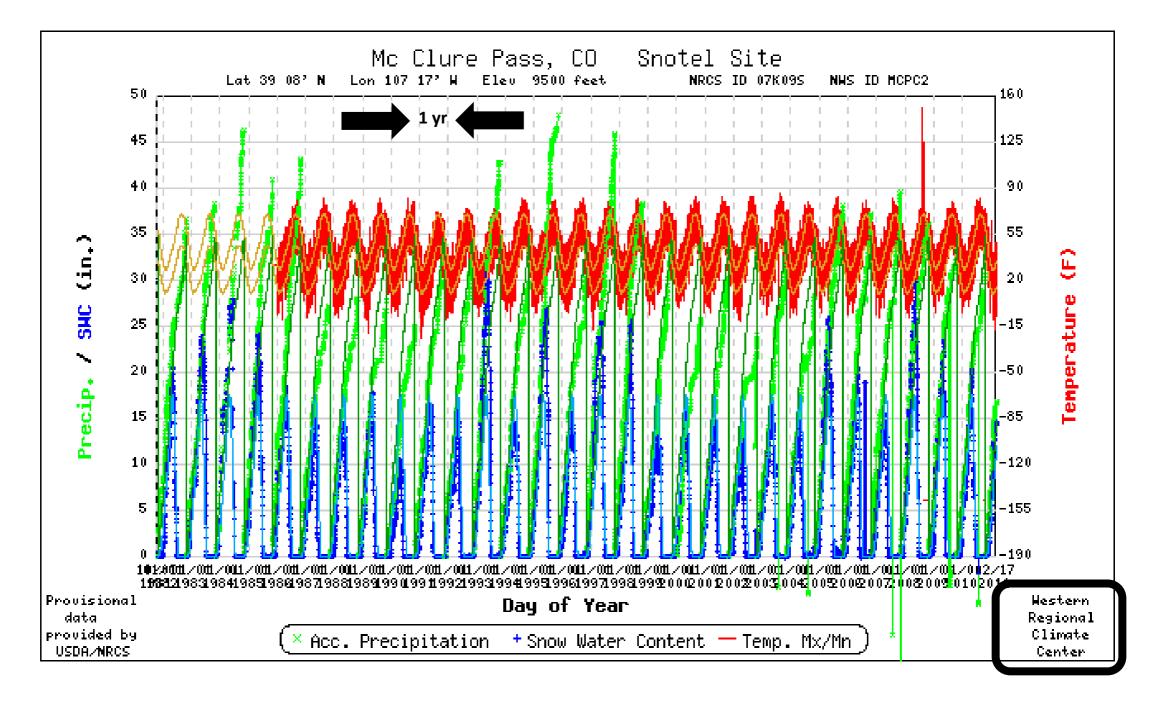
Columbia River at Vancouver, WA Hourly rate of stage change



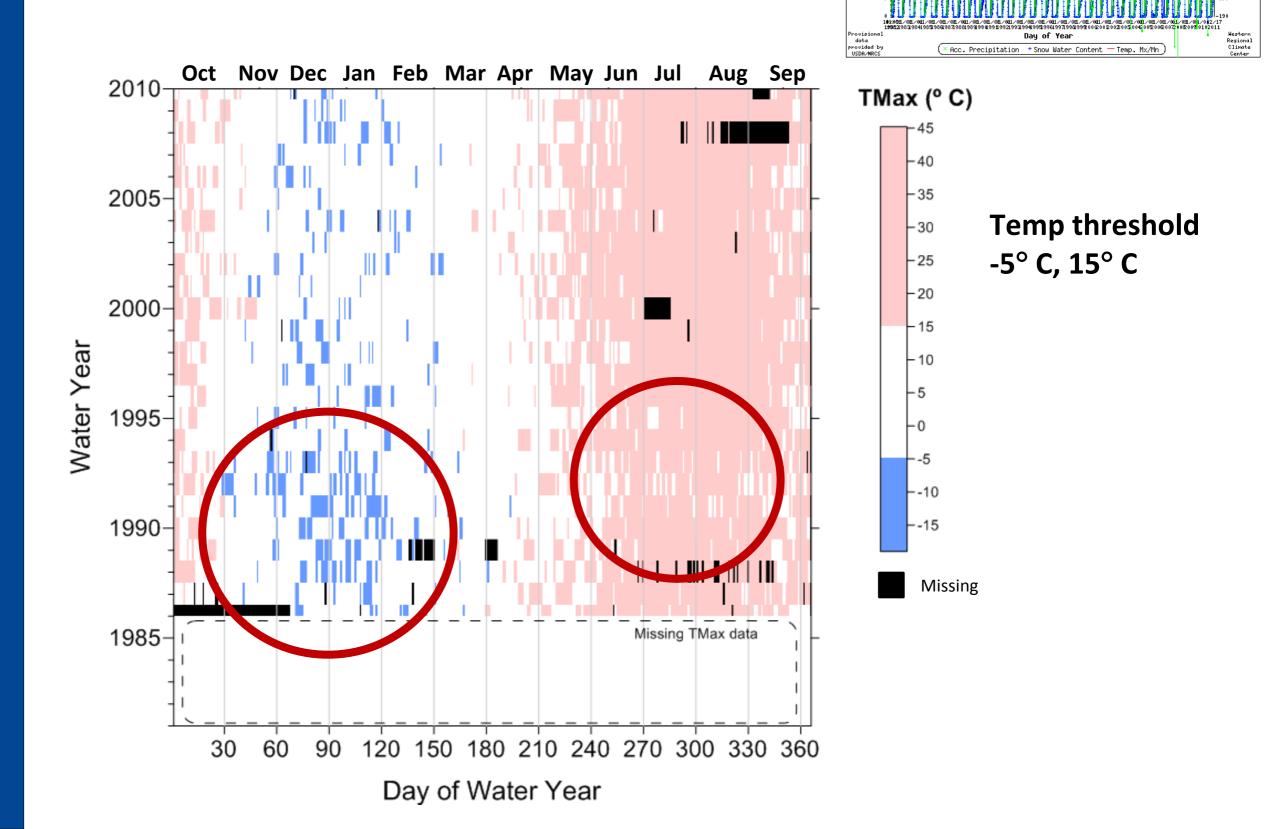


Application: pattern comparison

Temperature, Snow Water Equivalent, Precipitation



Temp

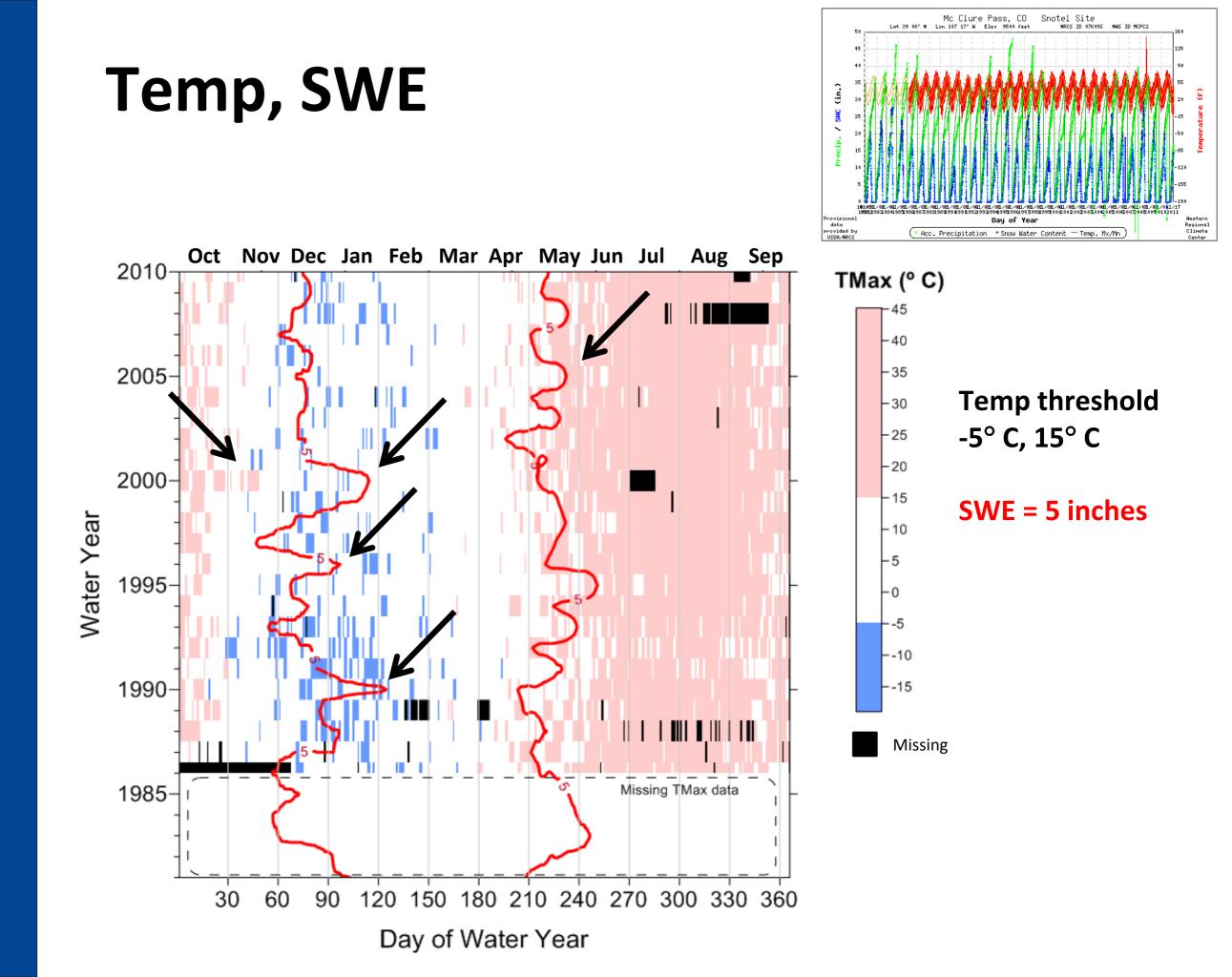


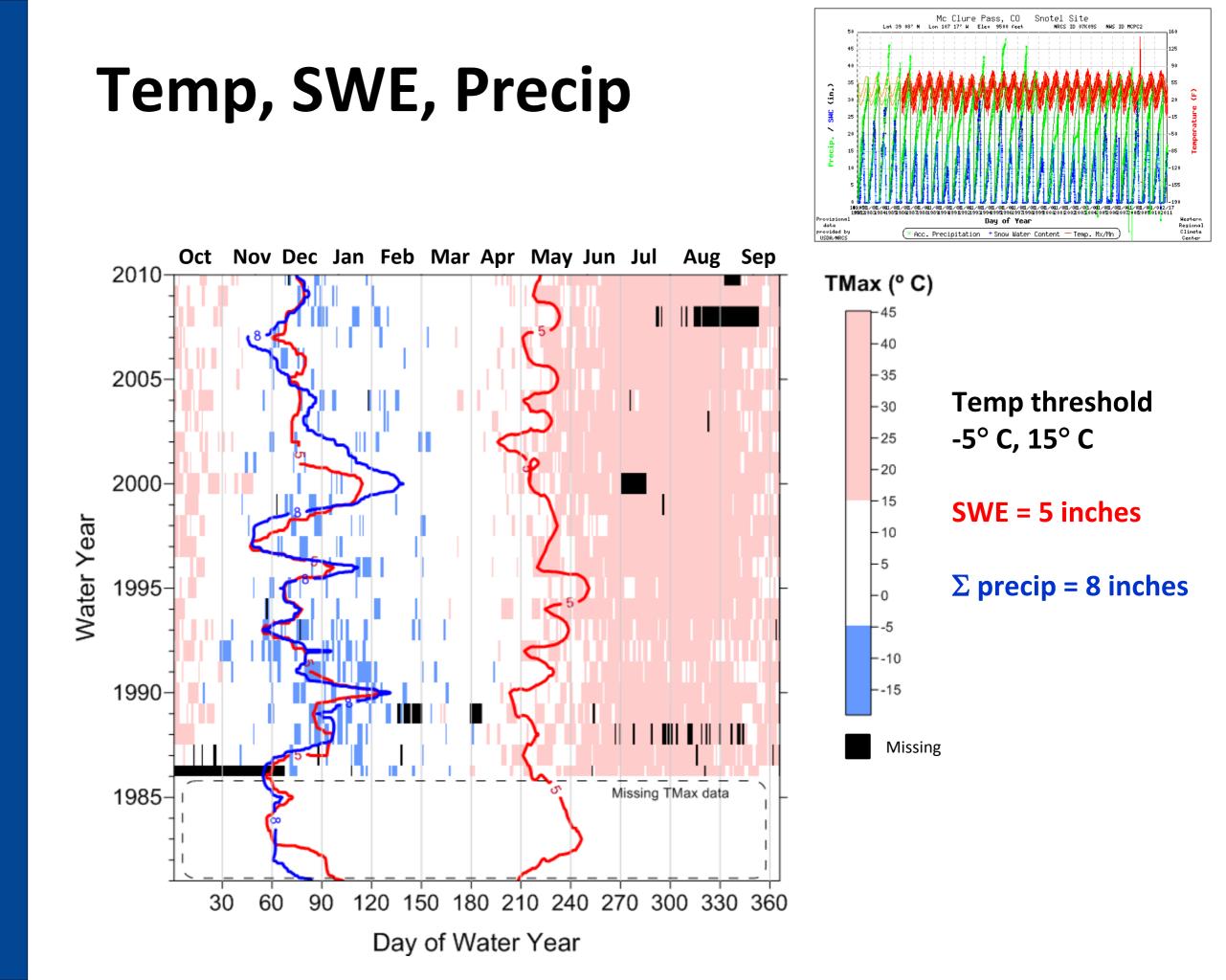
Mc Clure Pass, CO Lon 107 17' W Elev 9500 feet

(in.)

ġ

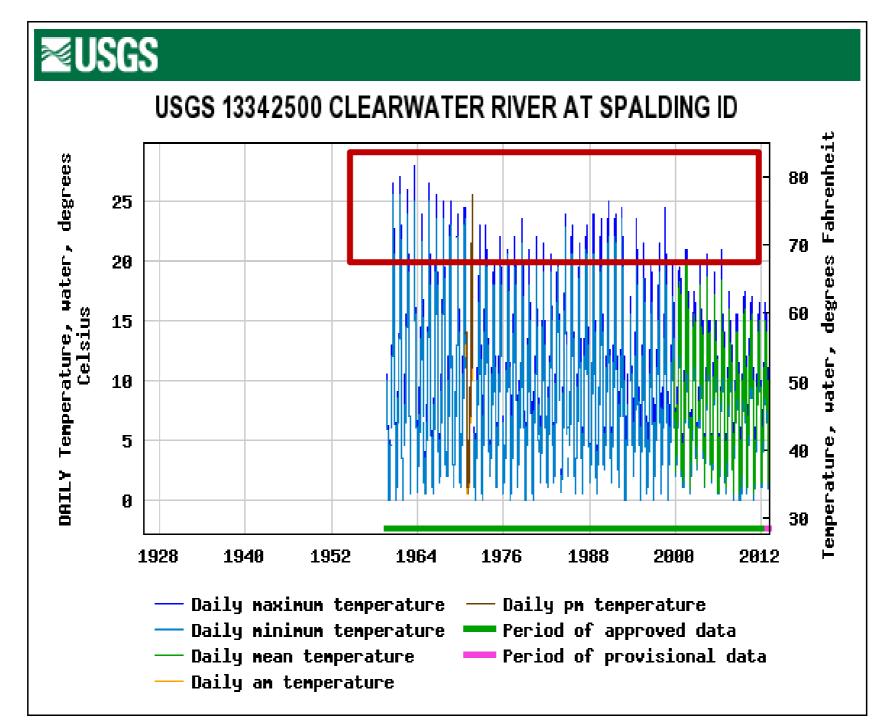
Snotel Site NRCS ID 07K09S





Application: Water temp

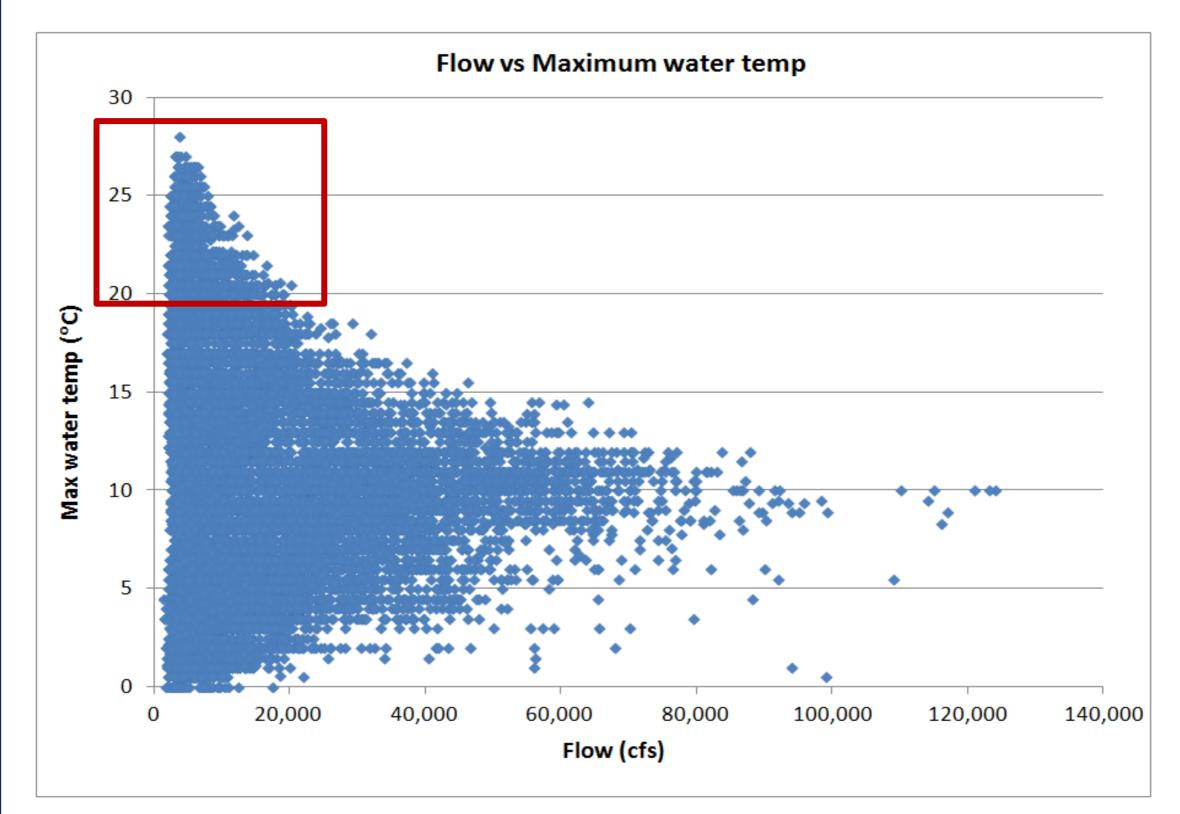
Indentify flows with maximum water temperature \geq 20° C (68° F)





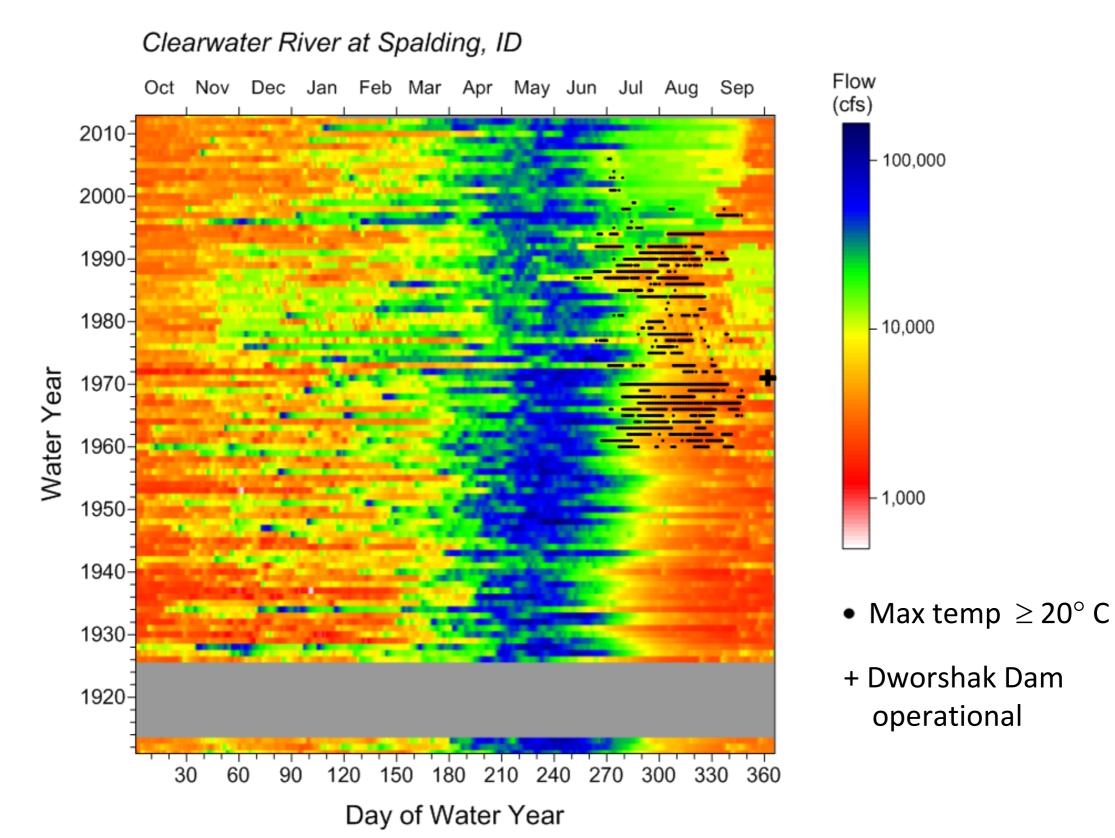
Max water temp vs flow

Indentify flows with maximum water temperature $\geq 20^{\circ}$ C (68° F)



Max water temp vs flow

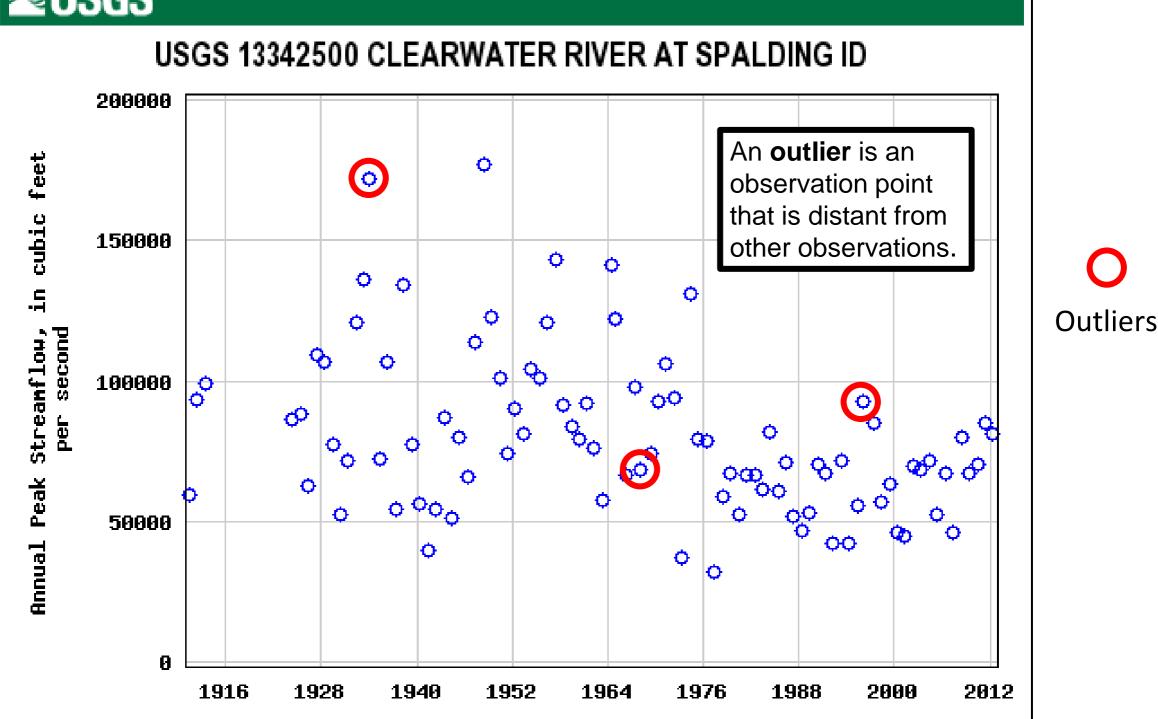
Flow, critical temperature and time



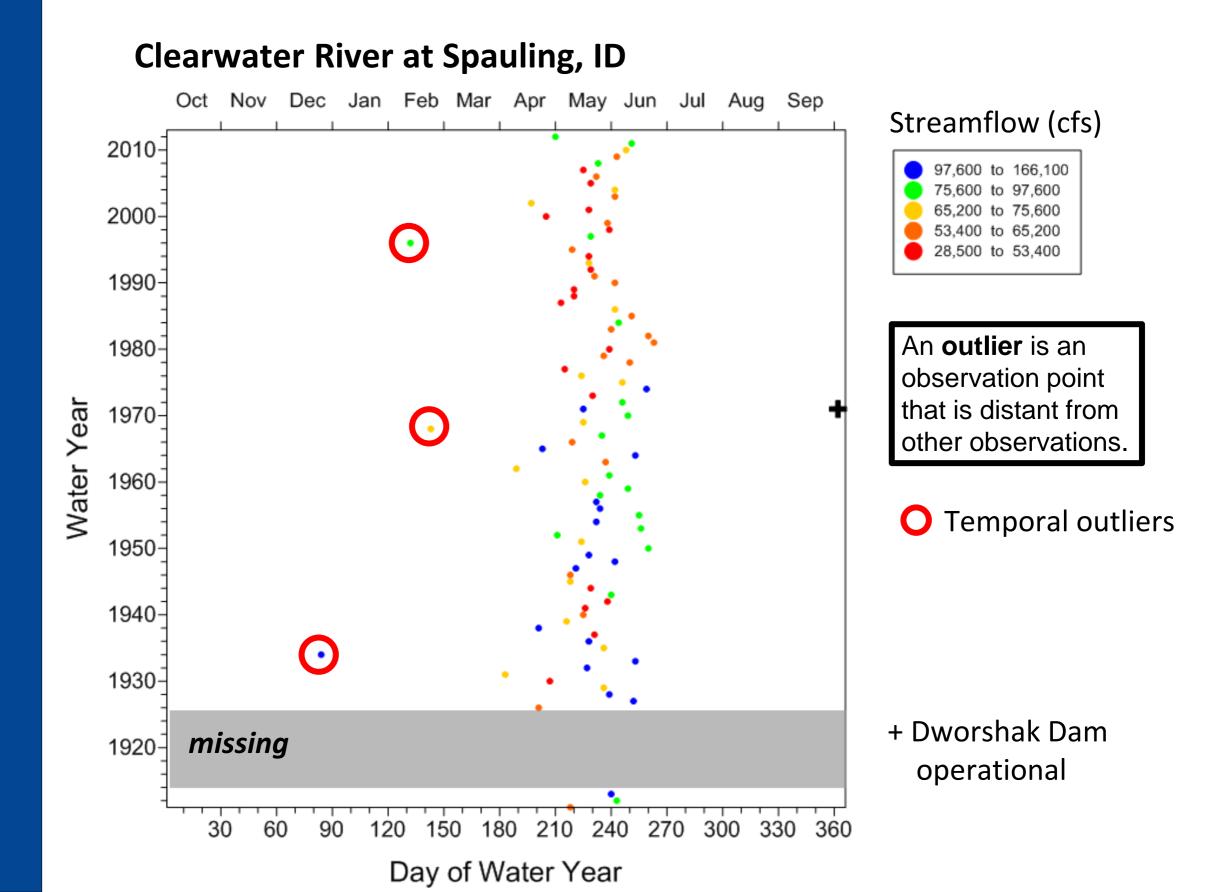
Application: Outlier detection

Annual maximum daily mean flow (cfs)

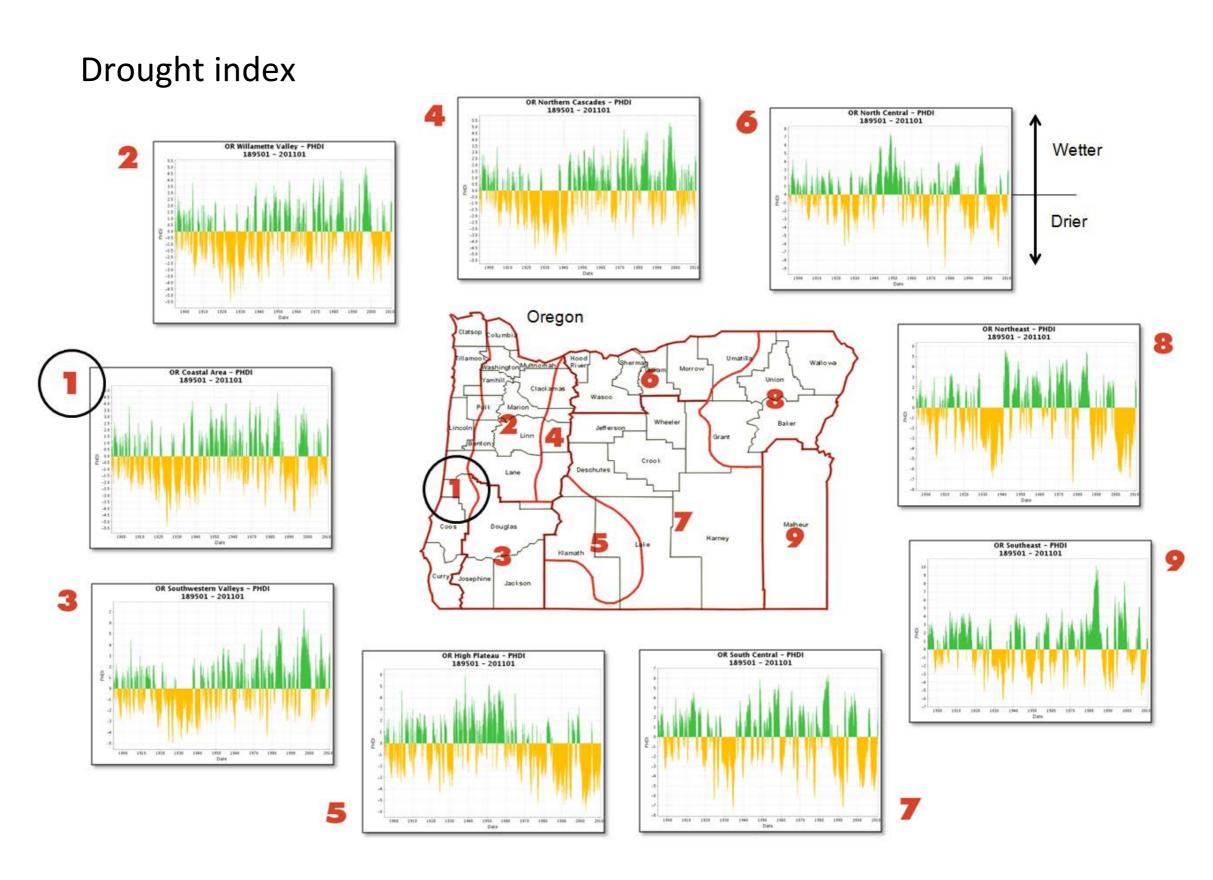
≊USGS

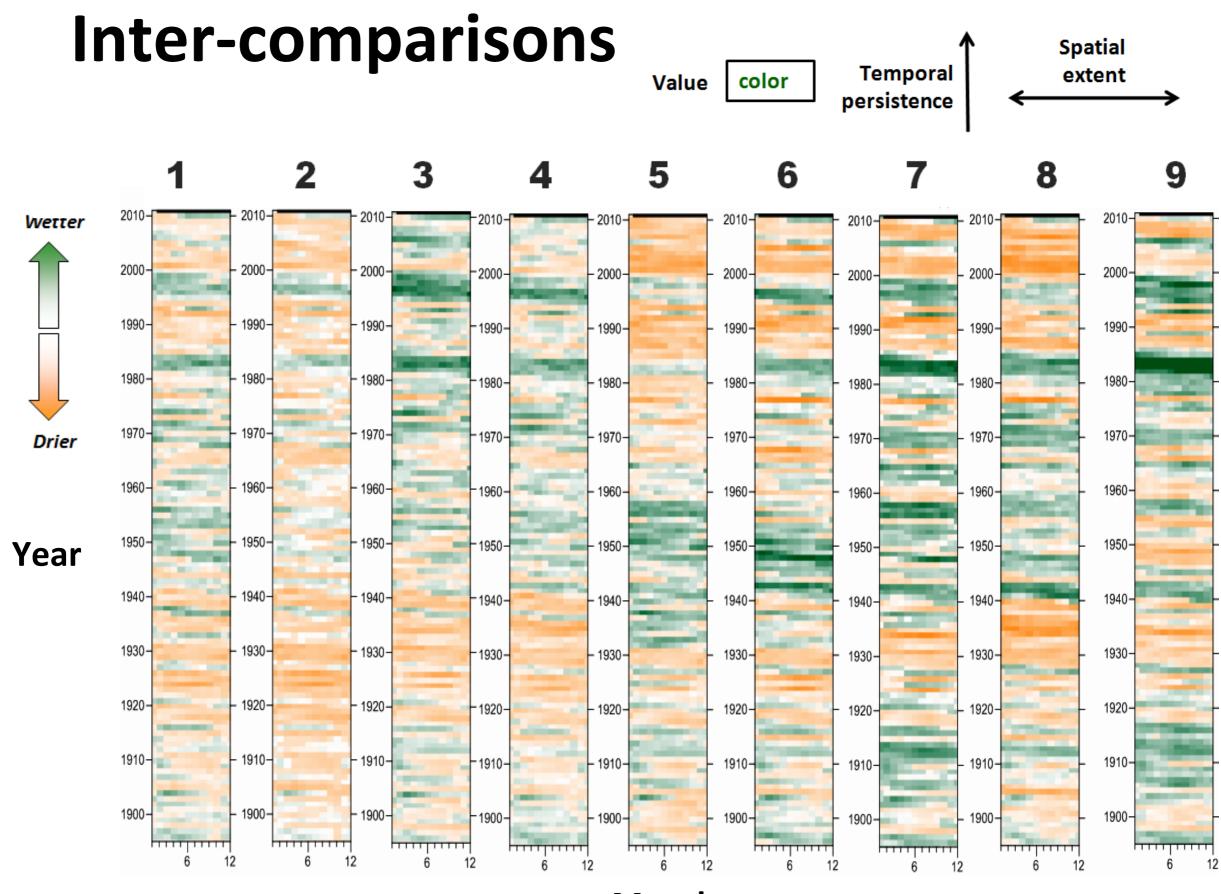


Outlier identification



Application: Multi-site comparison

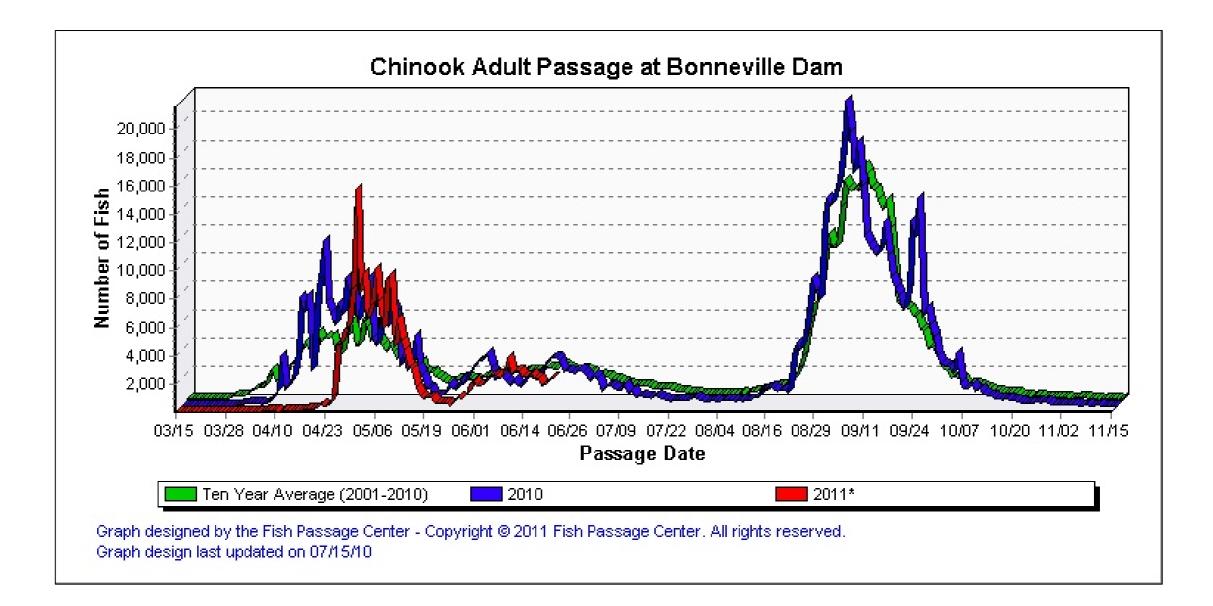




Month

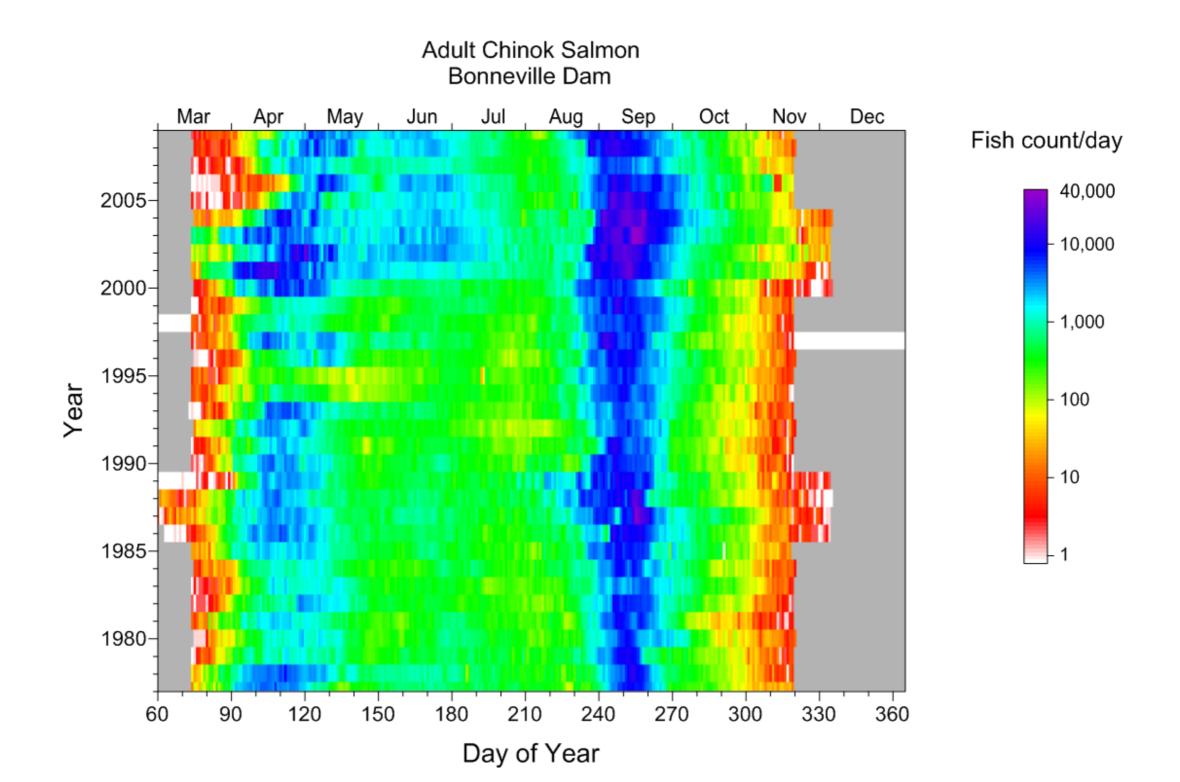
Application: salmon migration

Existing Bonneville Dam fish count plot



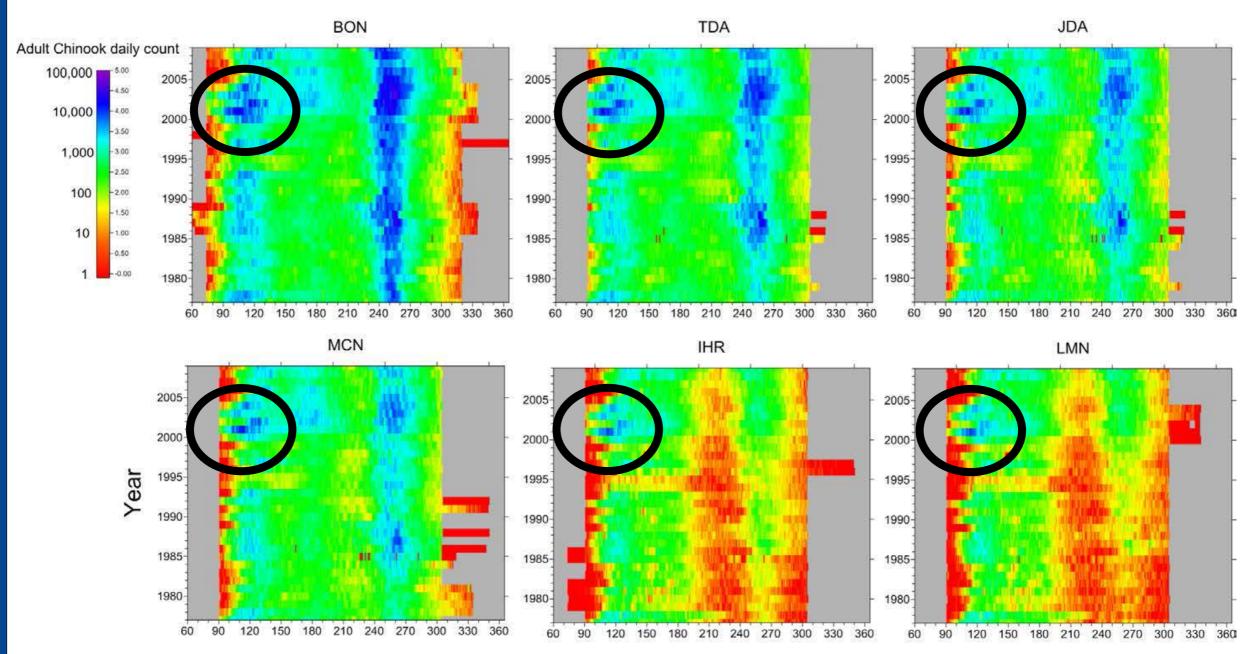
Site salmon migration

Bonneville Dam fish count – entire record



System salmon migration

Lower Columbia System fish count – entire record



Day of the Calendar Year

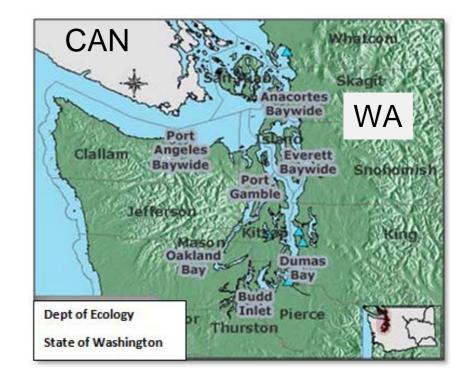
Case study

Paralytic shellfish toxins in Puget Sound*

How can multiple environmental time-series be integrated into a single summary plot?

<u>Topics</u>

Water quality Habitat monitoring Hydrometeorology Oceanography Public health Economics Climate change Decision support Ecological forecasting and trends Data visualization



Aquatic habitat

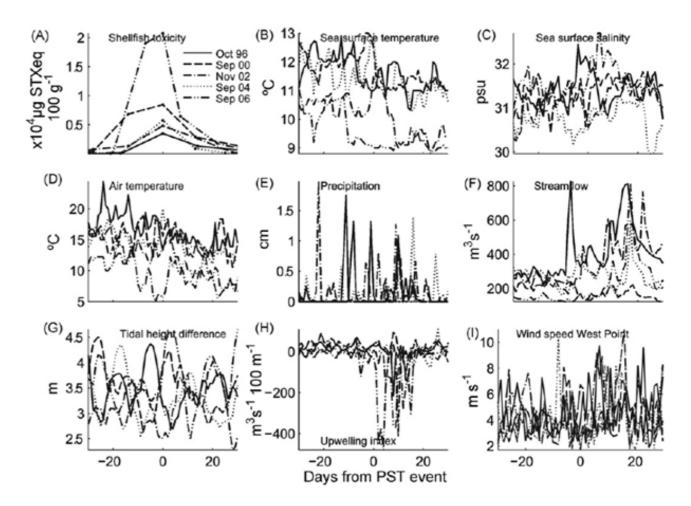
Background information

Time-series datasets:

Environmental factors

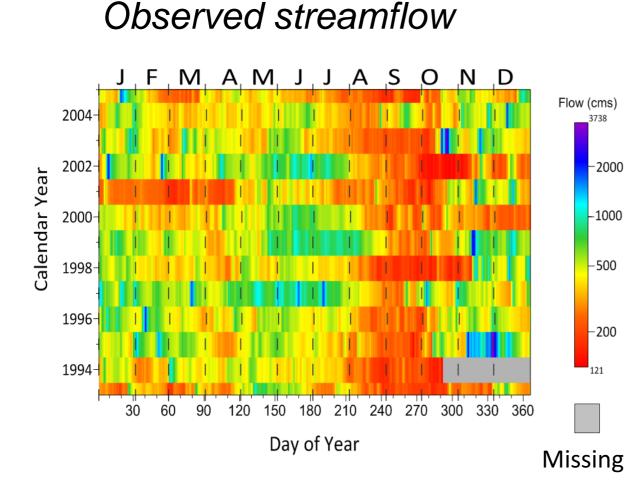
- 1. Sea surface temp ($^{\circ}C$)
- 2. Sea surface salinity (psu)
- 3. Air temp ($^{\circ}C$)
- 4. Precipitation (cm)
- 5. Streamflow (m^3s^{-1})
- 6. Tidal height difference (m)
- 7. Upwelling $(m^3 s^{-1} 100 m^{-1})$
- 8. Wind speed (ms^{-1})

Traditional plots:



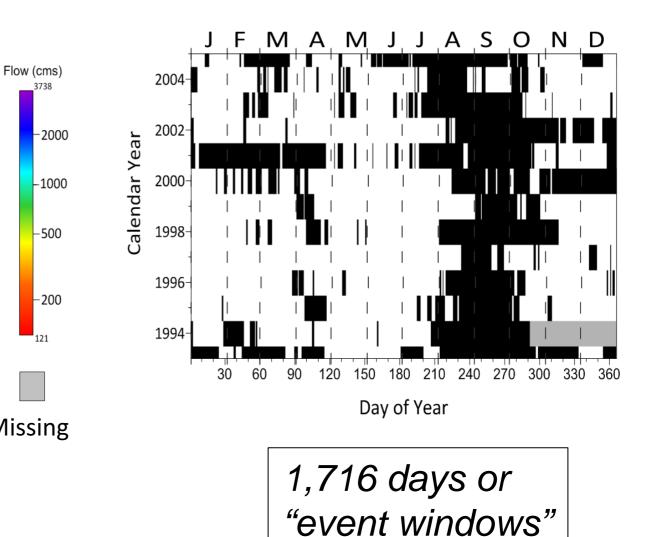
Criterion approach

Binary filter

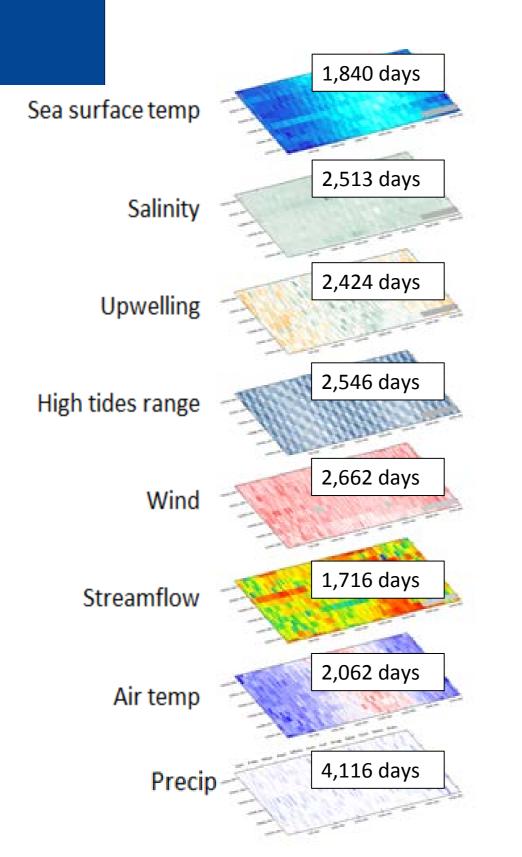


Flow $\leq 350 \text{ m}^{3}\text{s}^{-1}$

■ *Met* = 1, □ *Not met* = 0



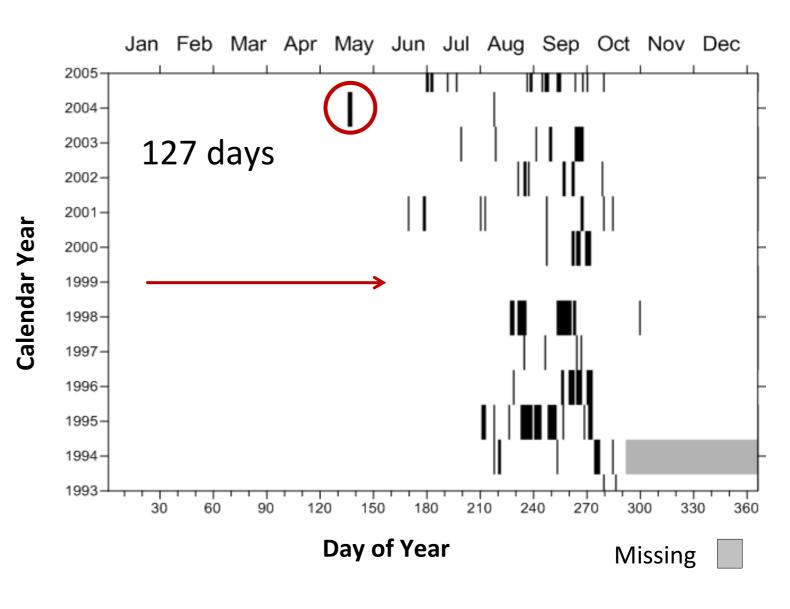
Analysis results



Apply specific criterion to specific layer

If Σ layers = 8 for specific day; Then event window present

Potential Event Windows



Problematic: 3-D plots

- •Graph display issues
- Axis labels
 - Axis direction

Limited analysis

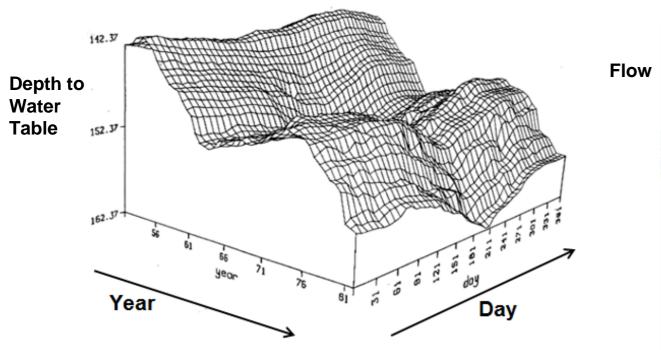
- Lighting angle

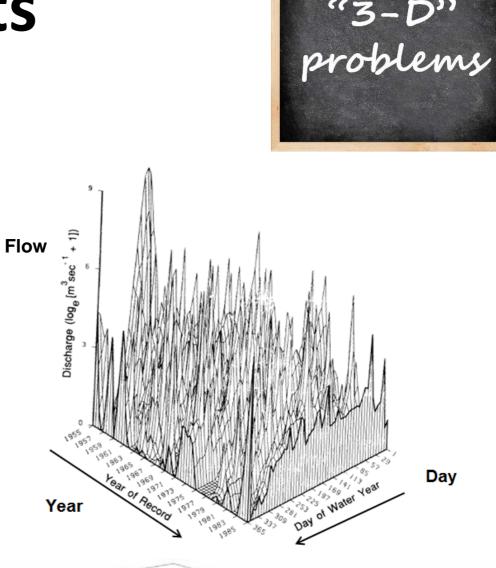
- Hidden data

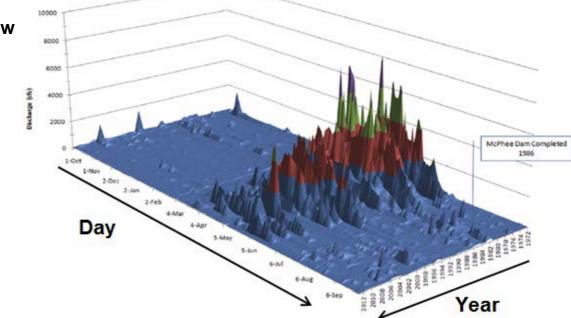
- View direction

Perspective

- Surface texture
- Temporal data ≠ Spatial data







Raster Users

Bureau of Land Management National Park Service NOAA US Army US Fish and Wildlife Service US Forest Service USDA – Agricultural Research Service US Geological Survey

Idaho – WRRI New Hampshire – Environmental Services Hood River, OR – Water Planning Group

> Brazil - National Water Agency Republic of Ireland - EPA

Arizona State University Fort Lewis College Juniata College Kent State University North Dakota State University Oregon State University Texas A &M University of Massachusetts - Amherst University of North Carolina - Charlotte University of Arizona University of Oregon Virginia Tech University

Nat. Center for Atmospheric Research The Nature Conservancy Normandeau and Associates, Inc

Brazilian National Water Agency (ANA)

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Pluviometric ID: precipitation characteristics at a glance

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Abstract

This article proposes the Pluviometric ID, a graphic representation of the main precipitation characteristics from a given dataset. The Pluviometric ID exceeds the traditional depiction of rainfall data as it allows a fast and simultaneous perception of parameters such as seasonal and interannual precipitation pattern changes, historic data series (length and integrity), cumulative and distributed (daily) long-term mean precipitation, among others. When plotted on a web map, the Precipitation ID quickly projects the regional precipitation patterns and trends towards a particular visually chosen direction Non-standard patterns may eventually arise and be easily spotted.

Received: 10 October 2013 Revised: 26 February 2014 Accepted: 13 March 2014

Keywords: precipitation data analysis; precipitation visualization; pluviometric ID

Summary

- Natural way to view large datasets
- Quickly review and interpret
- Develop new types of products and analyses
- Cost effective and time efficient method



References

de Pessôa, J. A. 2014. Pluviometric ID: precipitation characteristics at a glance. Atmosph. Sci. Lett.. doi: 10.1002/asl2.501

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Moore, S.K., et al., 2009. Recent trends in paralytic shellfish toxins in Puget Sound, relationships to climate, and capacity for prediction of toxic events. Harmful Algae 8, 463–477. doi:410.1016/ j.hal.2008.1010.1003.

Olden, J.D., and N.L. Poff. 2003. Redundancy and the choice of hydrologic indices for characterizing streamflow regimes. River Research and Applications 19, 101-121.



Thank you!

Questions?

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Extra slides

