

The image shows a natural landscape. In the foreground, there is a field of tall, dry, yellow grasses. Behind the grass, a dense forest of tall, thin trees, likely pines or firs, stretches across the background. The sky is a clear, bright blue. The text is overlaid on the image in a white, bold, sans-serif font.

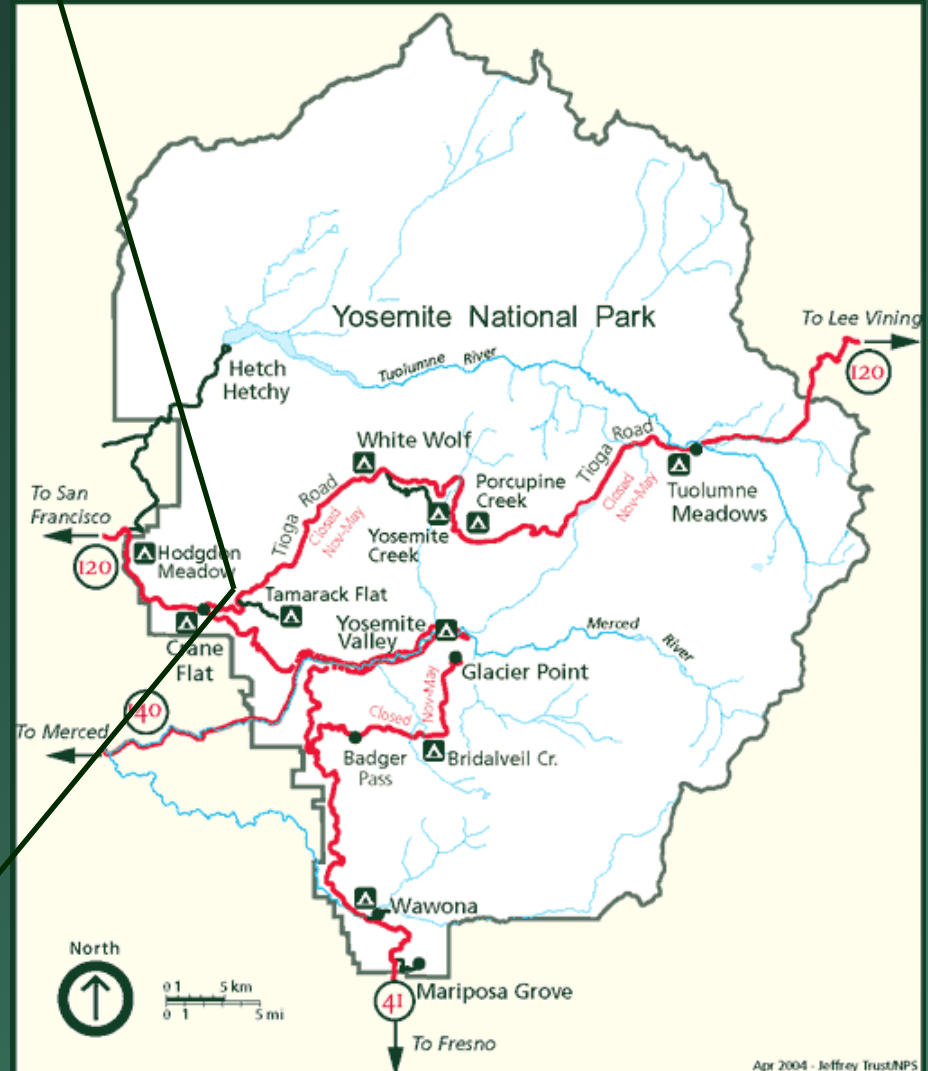
Development of a California-wide Soil Moisture Monitoring Strategy, and Application to the Hetch Hetchy Watershed

**Lorrie Flint, Jennifer Curtis, Michelle Stern, and Alan Flint
U.S. Geological Survey, Sacramento**



Gin Flat

Elevation: 7050' (2149m)



Dana Meadows Site Elevation: 9728' (2965m)



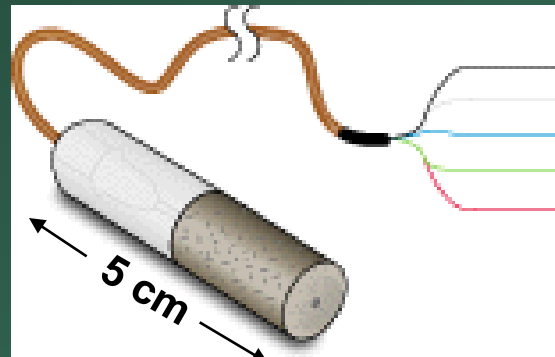
Picture Taken: 06/29/2011

Soil Moisture Monitoring Instrumentation



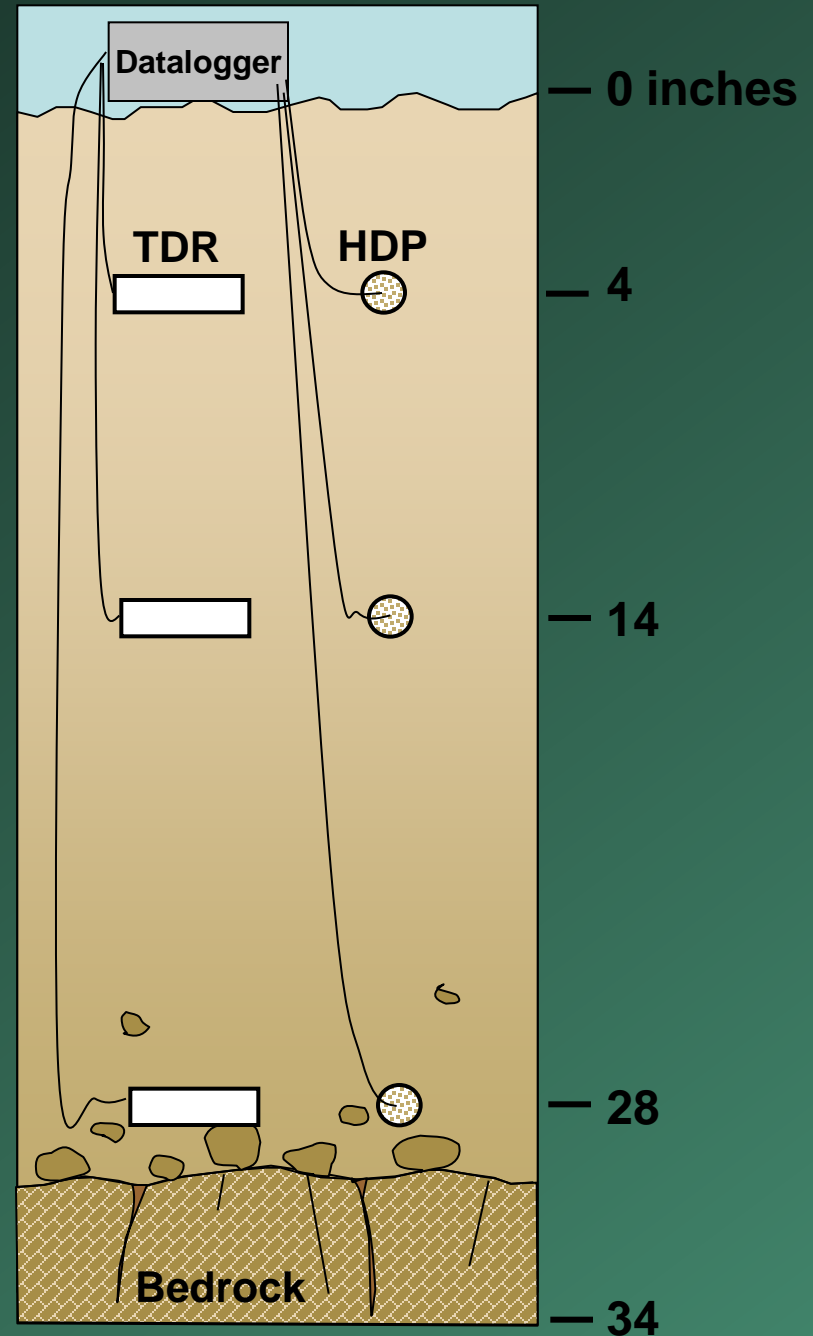
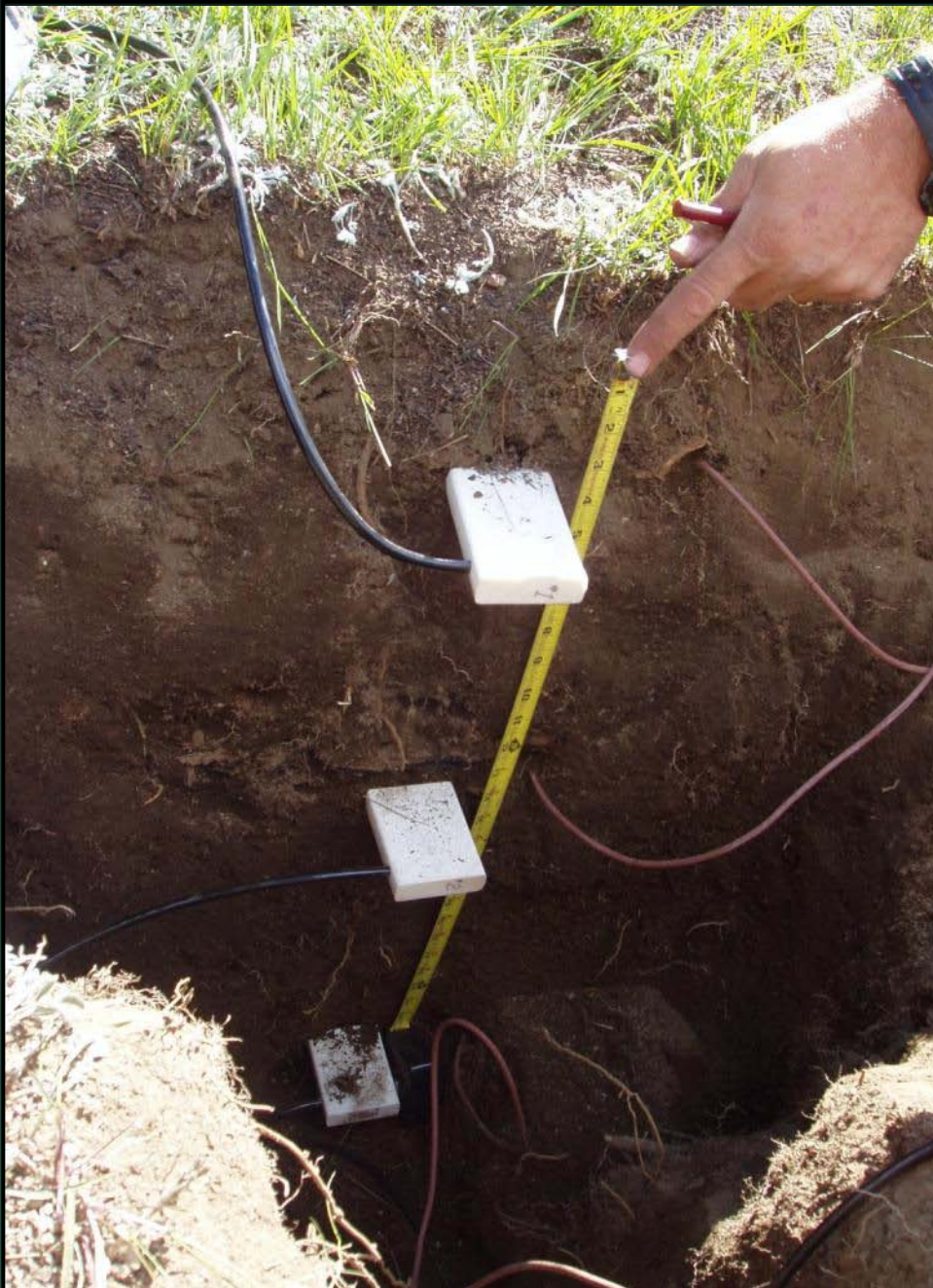
Water content:

**Time Domain
Reflectometry
(TDR)**

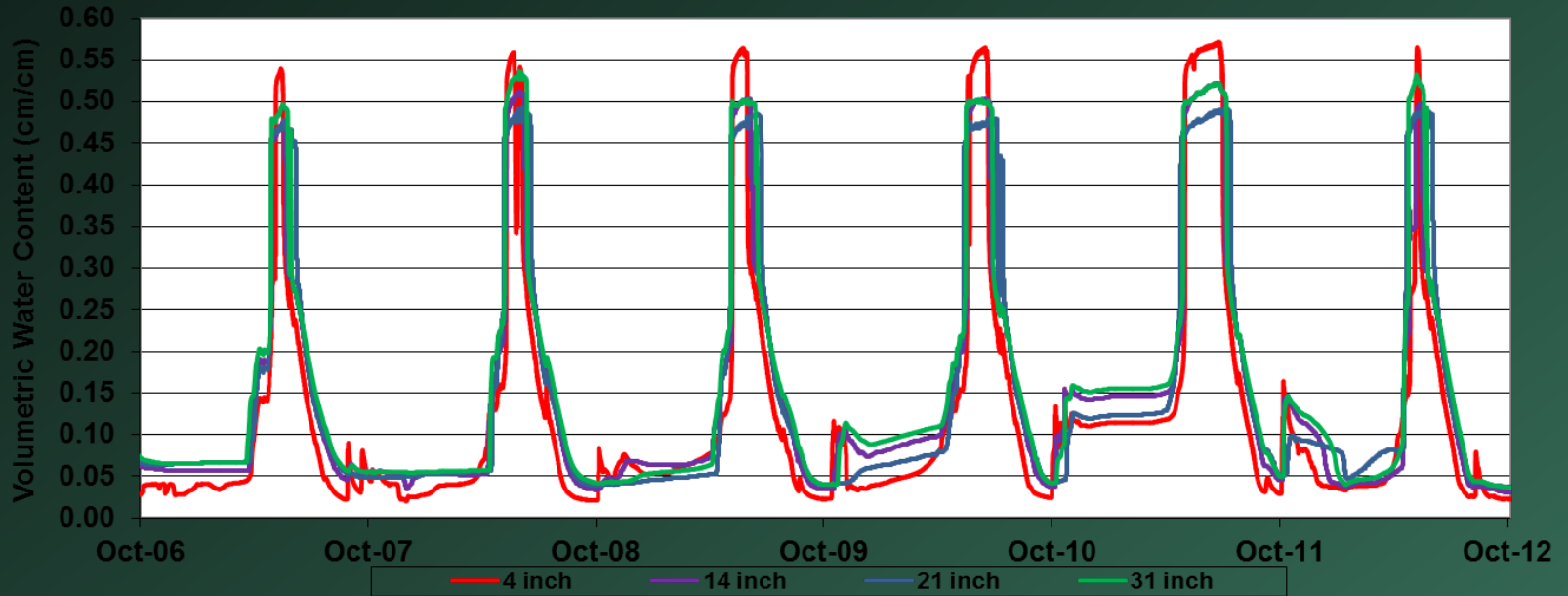


Water potential:

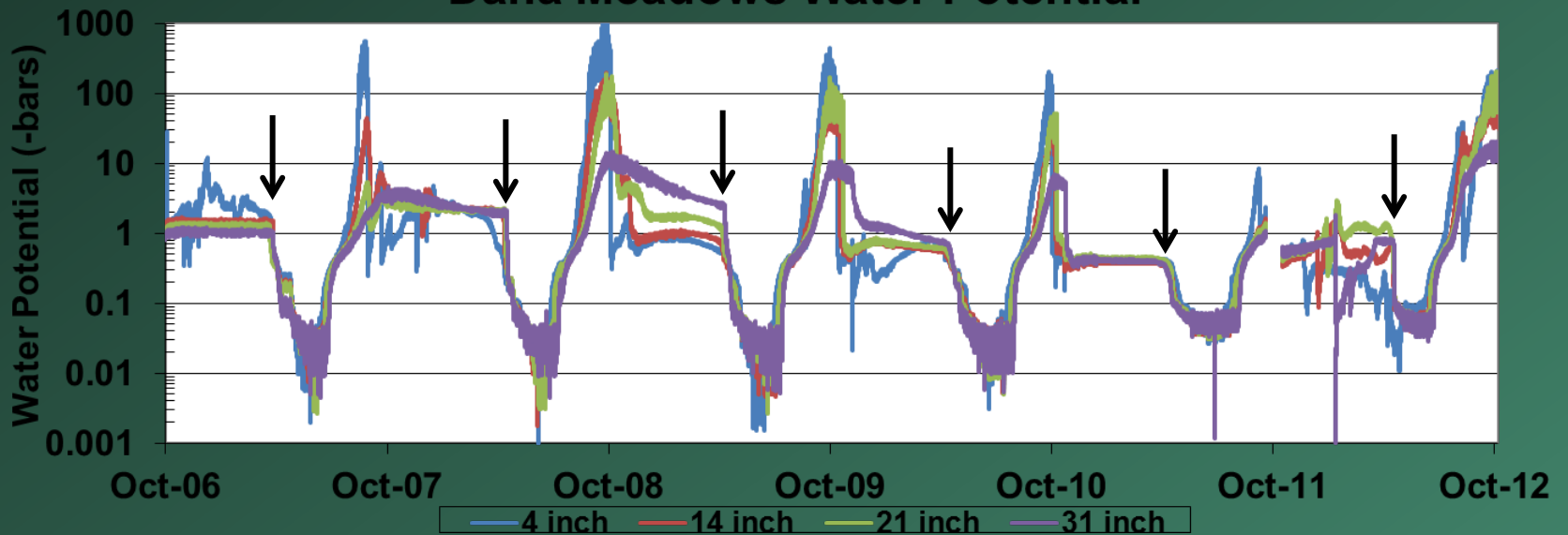
**Heat Dissipation
Probe
(HDP)**



Dana Meadows Water Content



Dana Meadows Water Potential





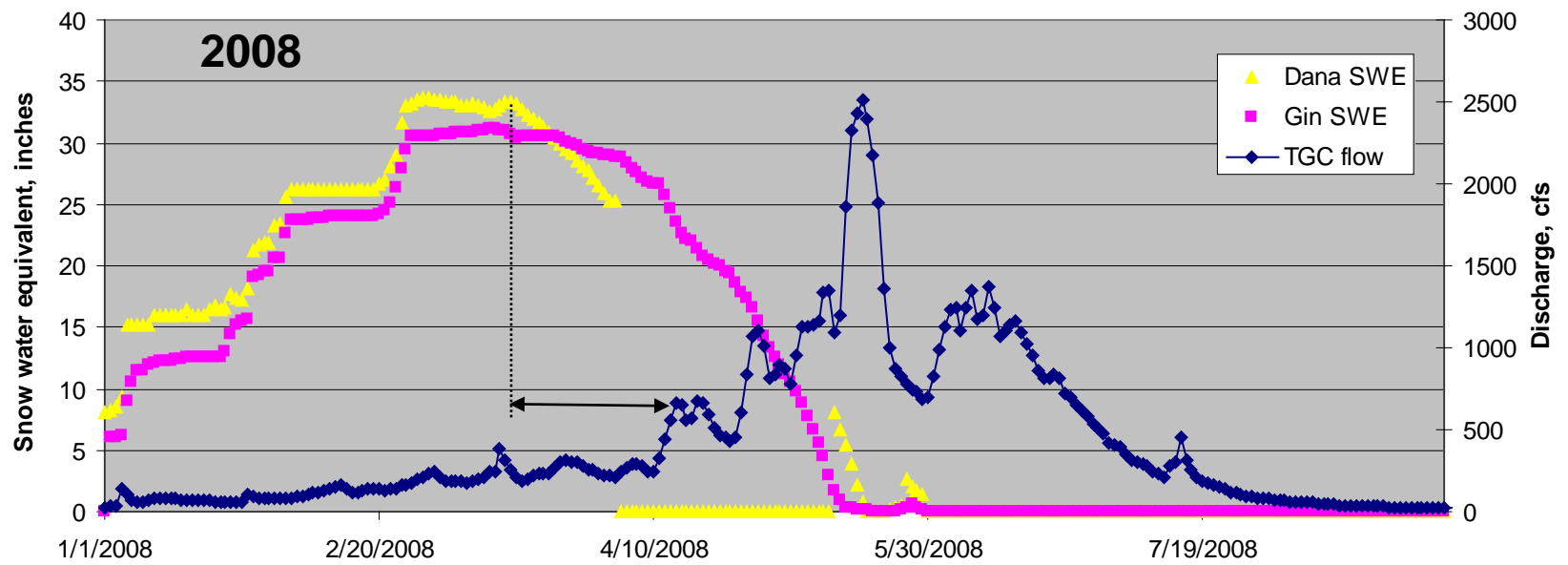
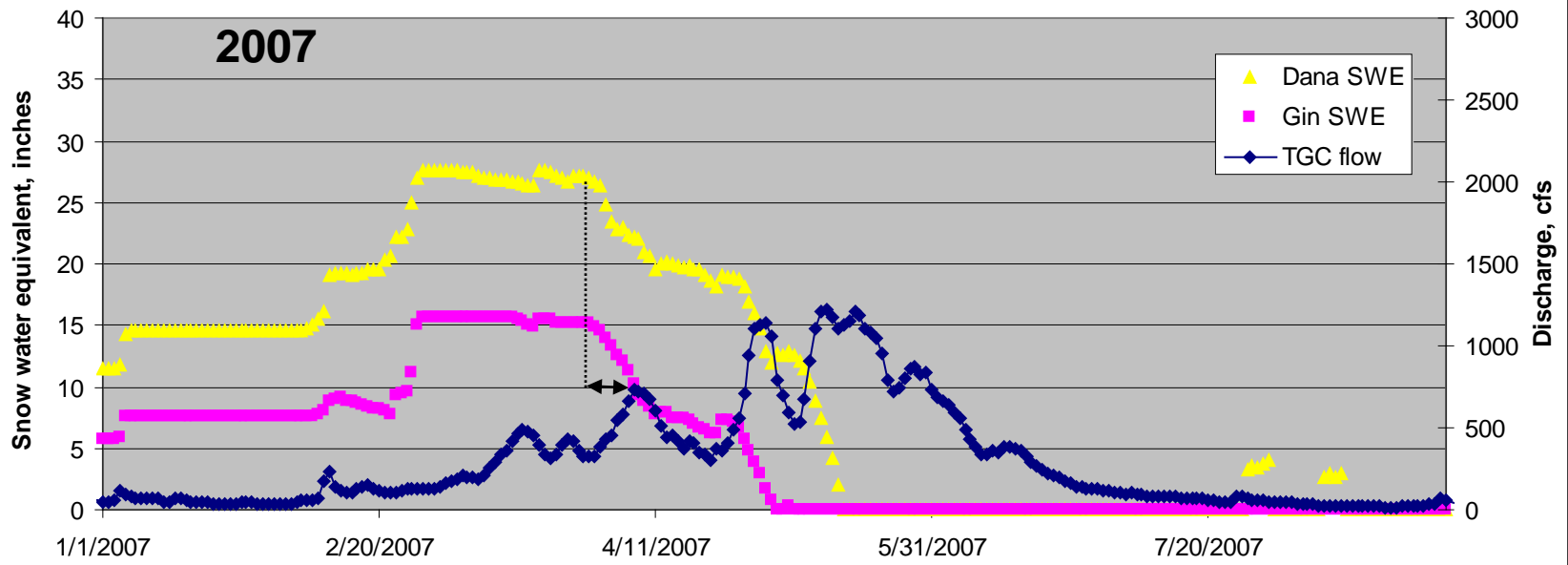
Applications

- Hydrologic model calibration/validation
- Calculations of volume of runoff lost to soil moisture replenishment under varying climatic conditions
- Real-time application for forecasting

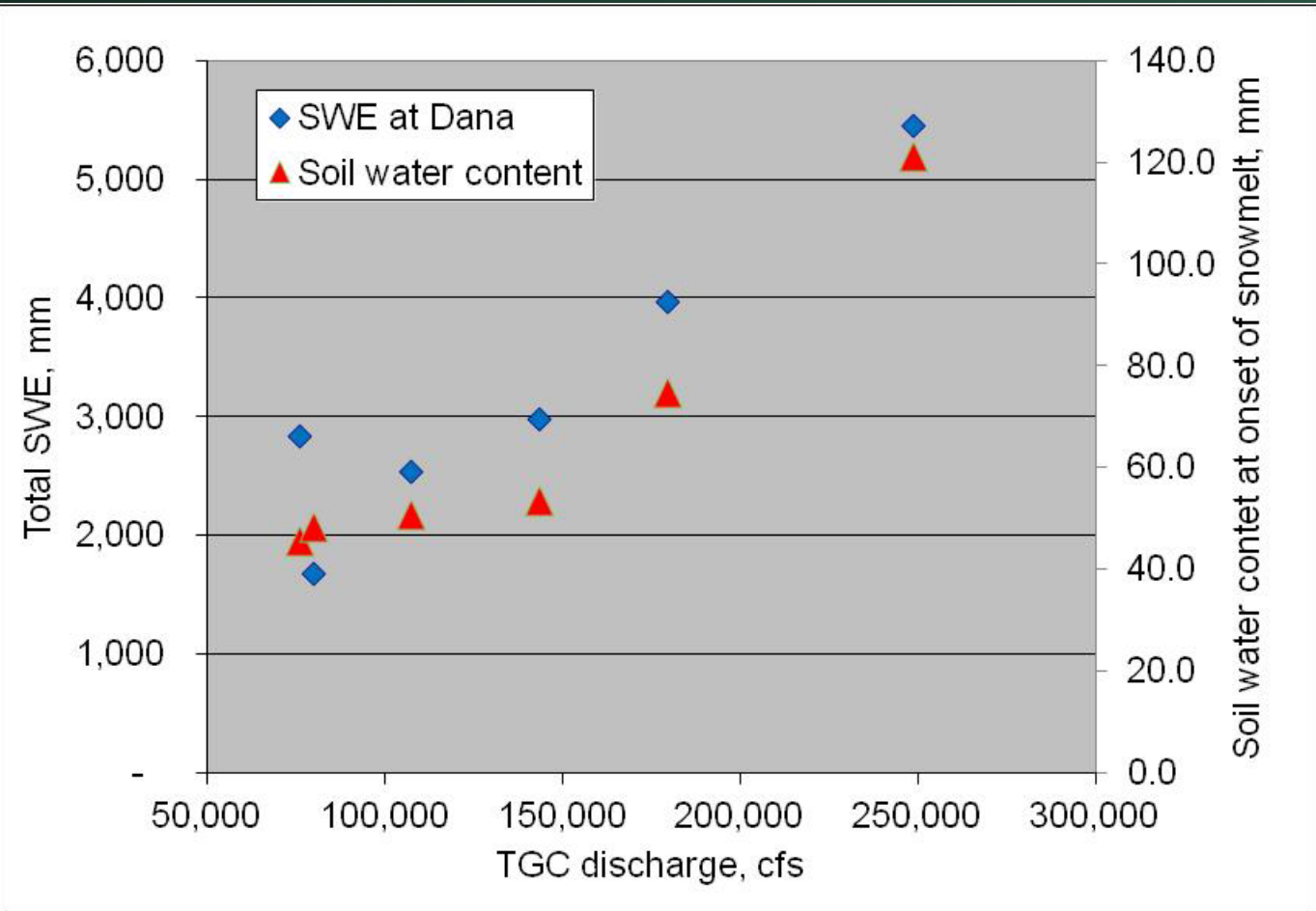
Tuolumne R. @ Grand CYN of Tuolumne

Elevation: 3830' (1167m)

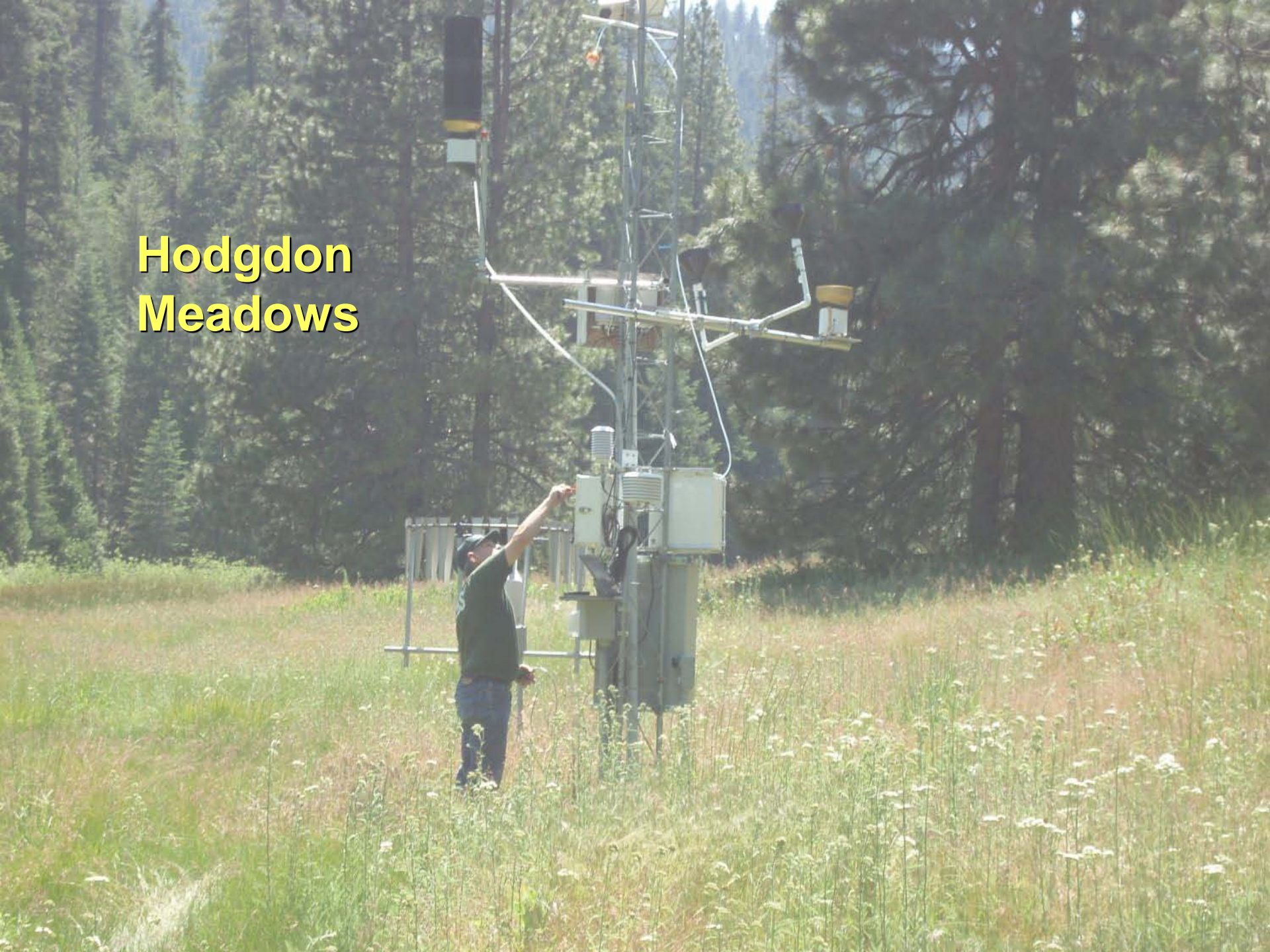




	2007	2008	2009
Onset of snowmelt	3/31	3/15	4/1
Lag time to rise in flow (days)	0	24	18
Driest potential at bottom of soil (bars)	-1	-4	-12
Soil water potential at onset (bars)	-0.8 - -1.2	-1.5 - -1.8	-0.1 - -2.7
Soil water content at onset (v/v)	0.1-0.15	0.04-0.06	0.05-0.07
# weeks to wet soil profile	0	5	6
# weeks to full saturation	7	11	8



Hodgdon Meadows



Pepperwood Preserve Sonoma County



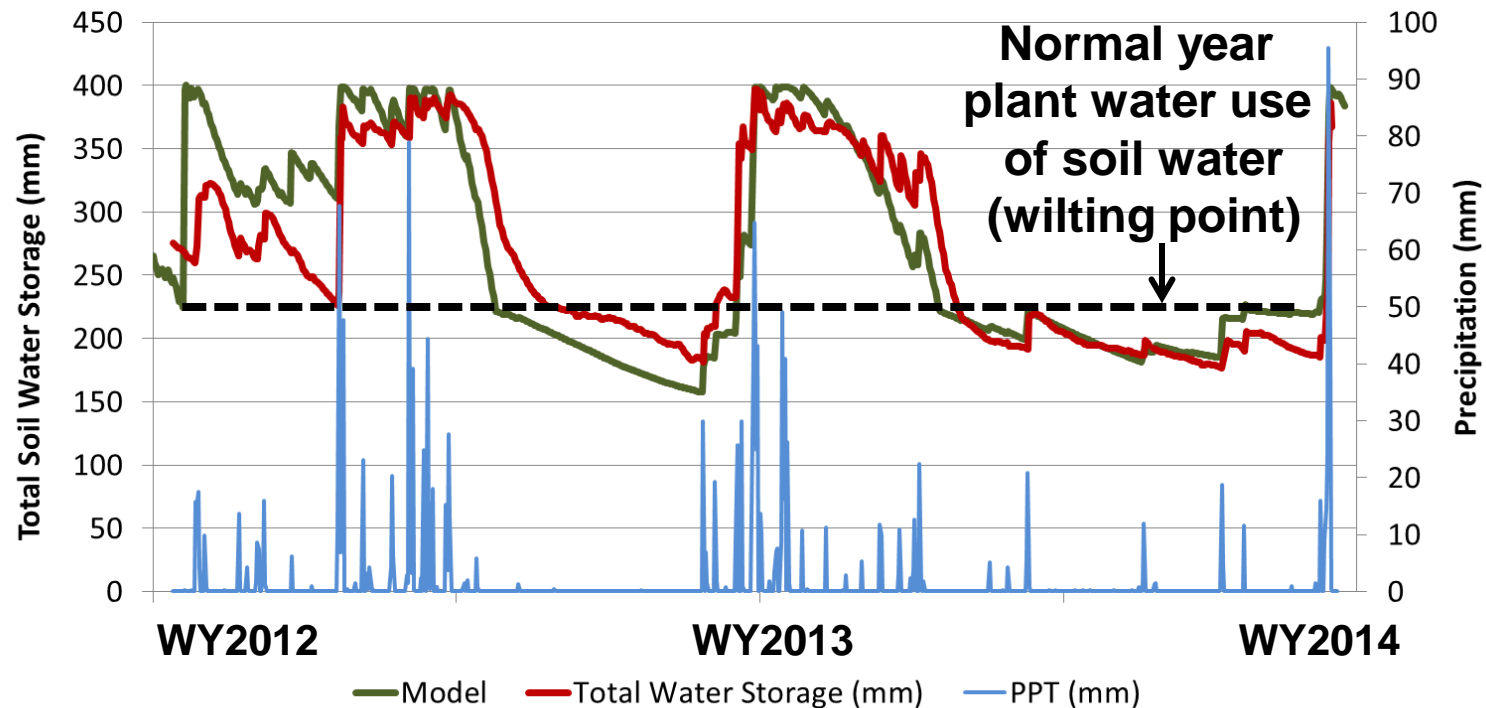
Why Water Potential?

- Helps you understand processes
- Helps identify hydraulic properties
- When plants shut down (wilting point) and drainage is occurring

Soil Moisture Monitoring

(headwaters of Mark West Creek)

Pepperwood Preserve Grassland Soil Moisture Monitoring

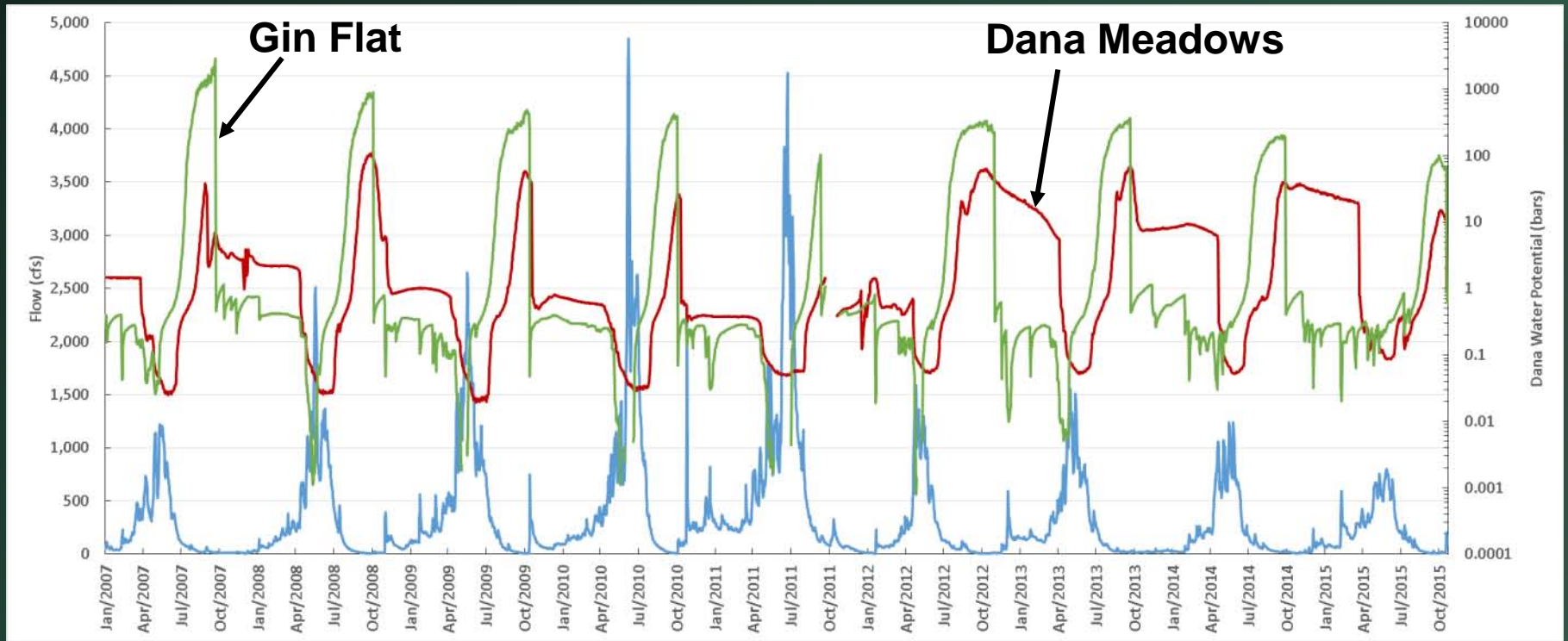


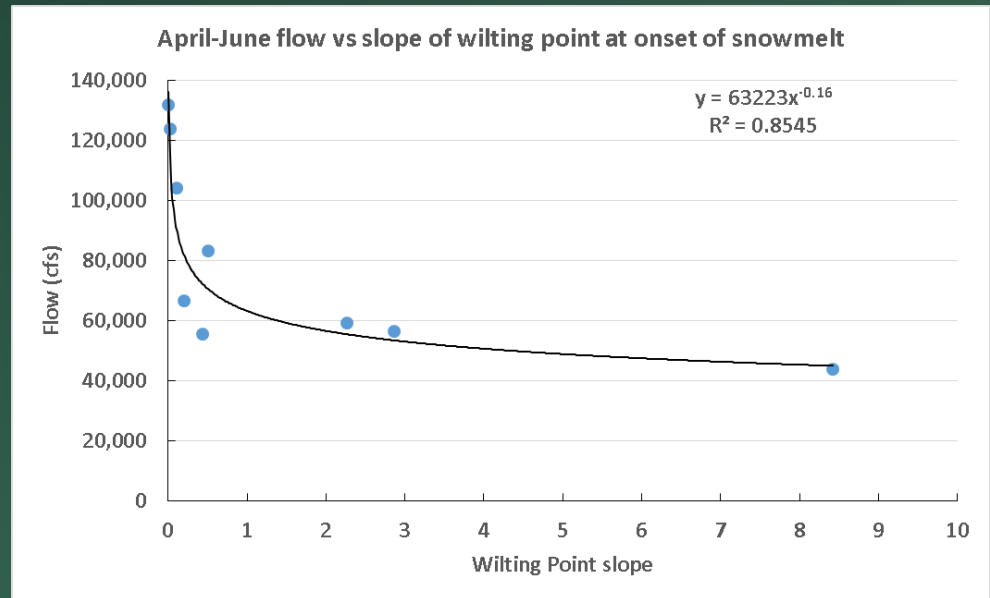
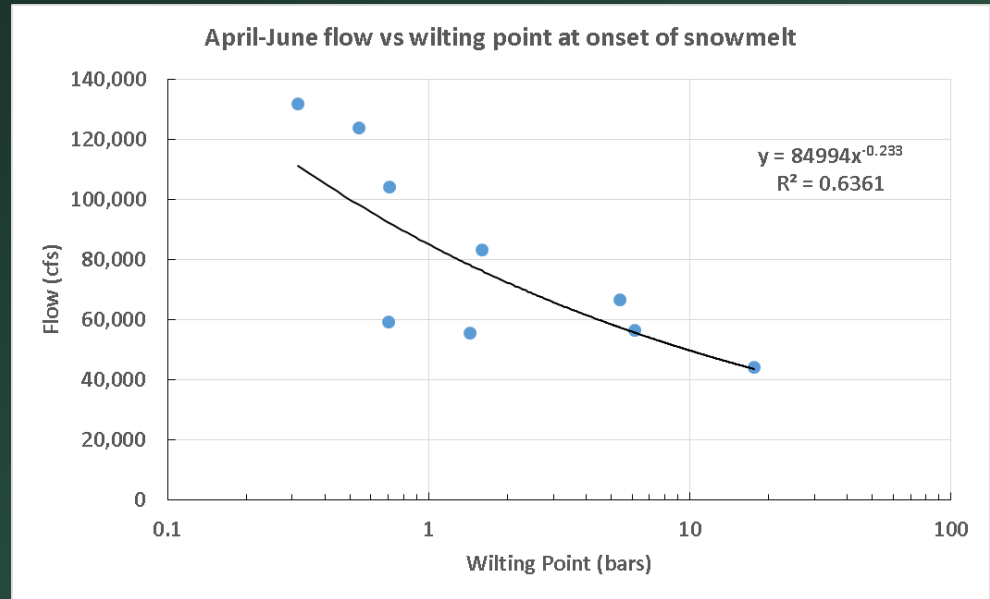
Why Multiple Depths?

- Helps you understand physical and physiological processes
- Rain gage, rooting zone, ponding zone



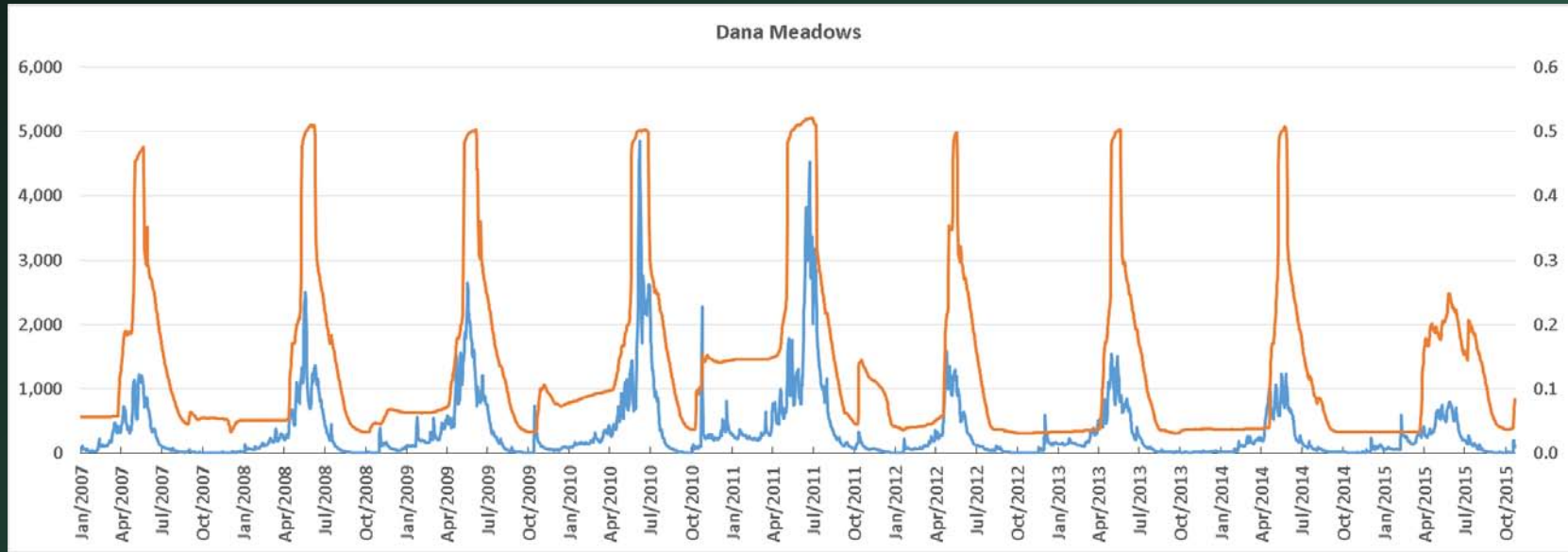
Can we use soils data for forecasting?





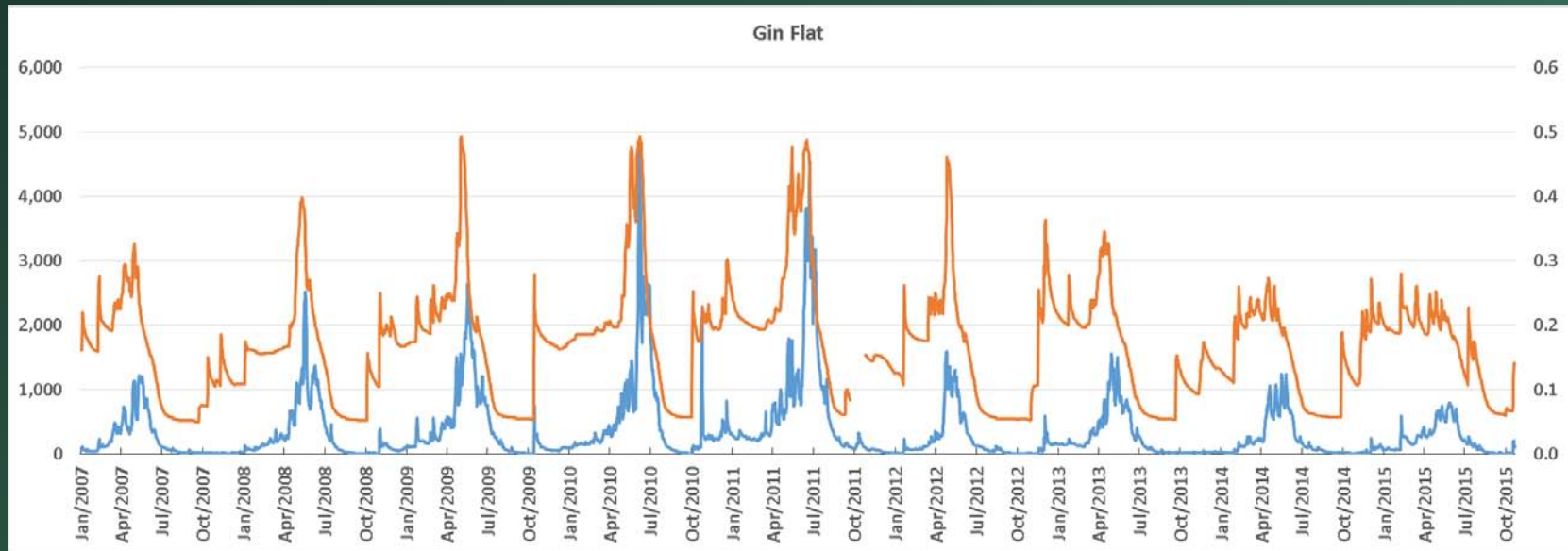
Can we forecast streamflow with soil moisture?

Streamflow, cfs

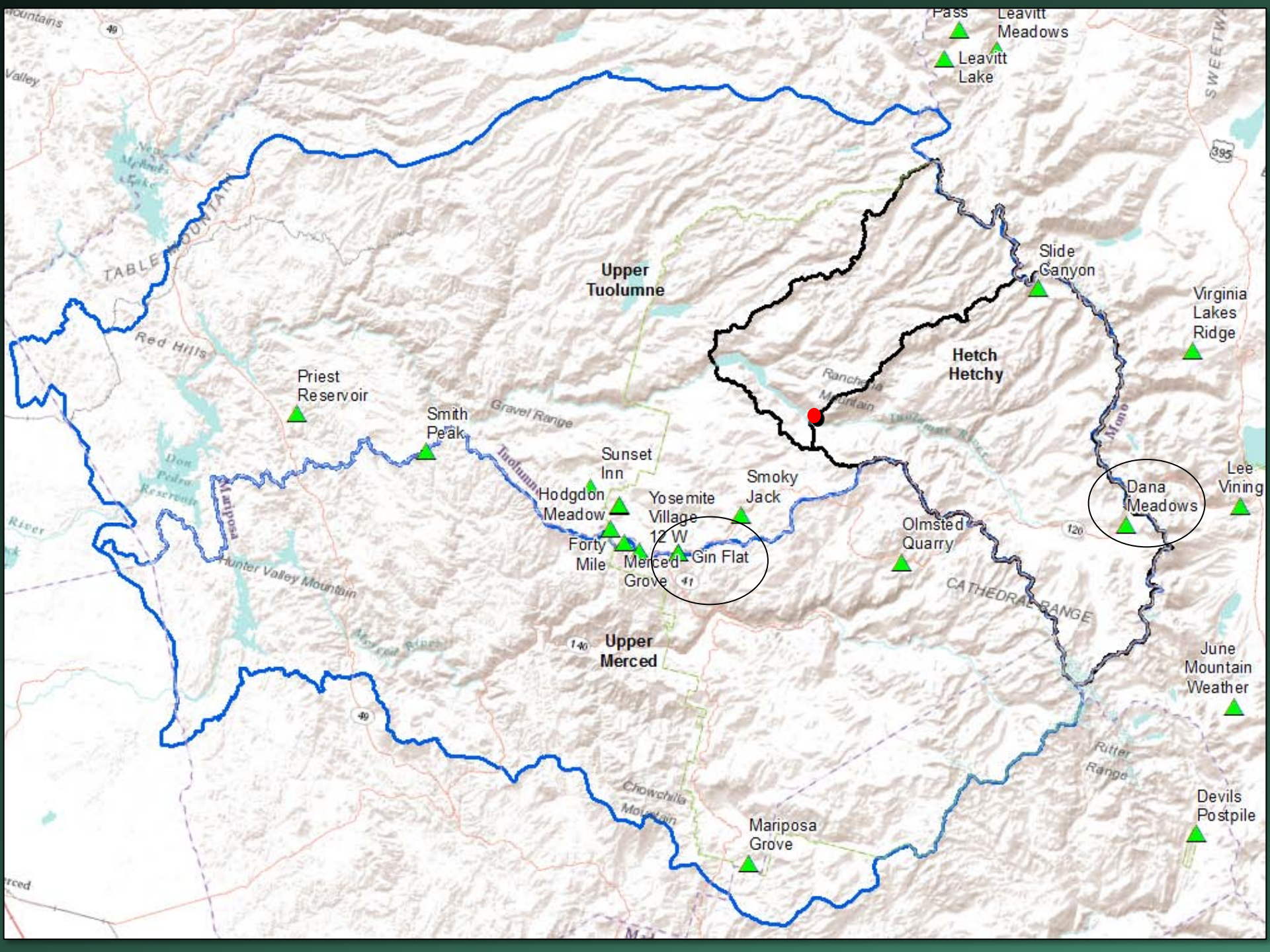


Saturation

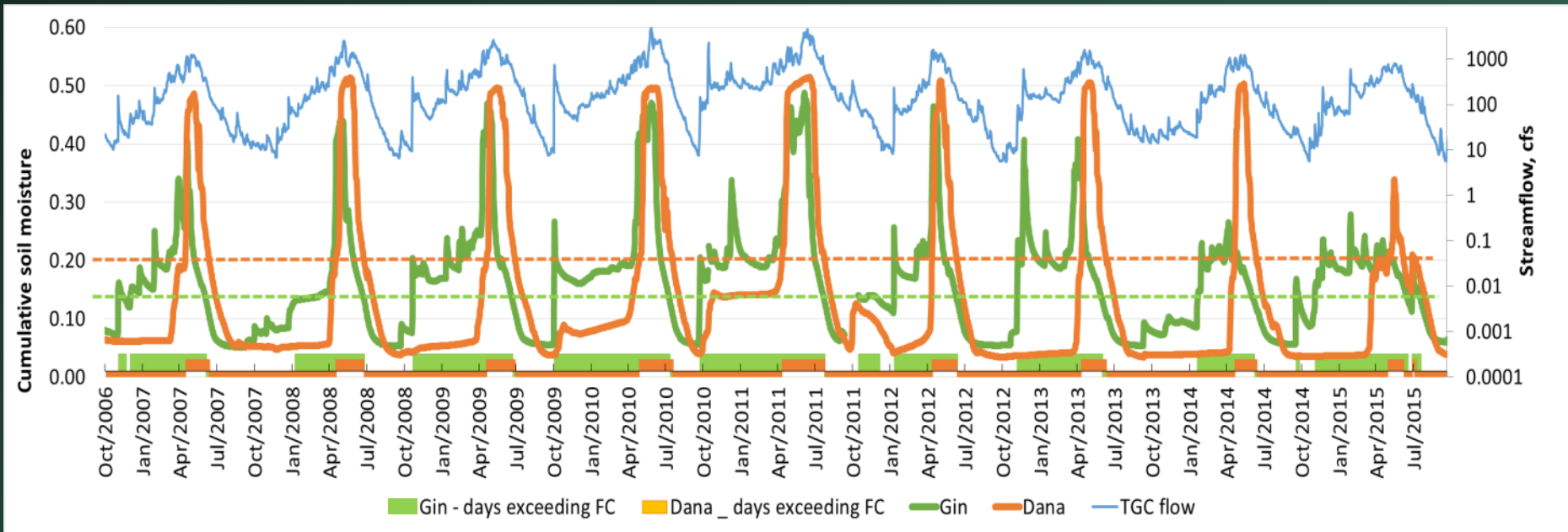
Streamflow, cfs



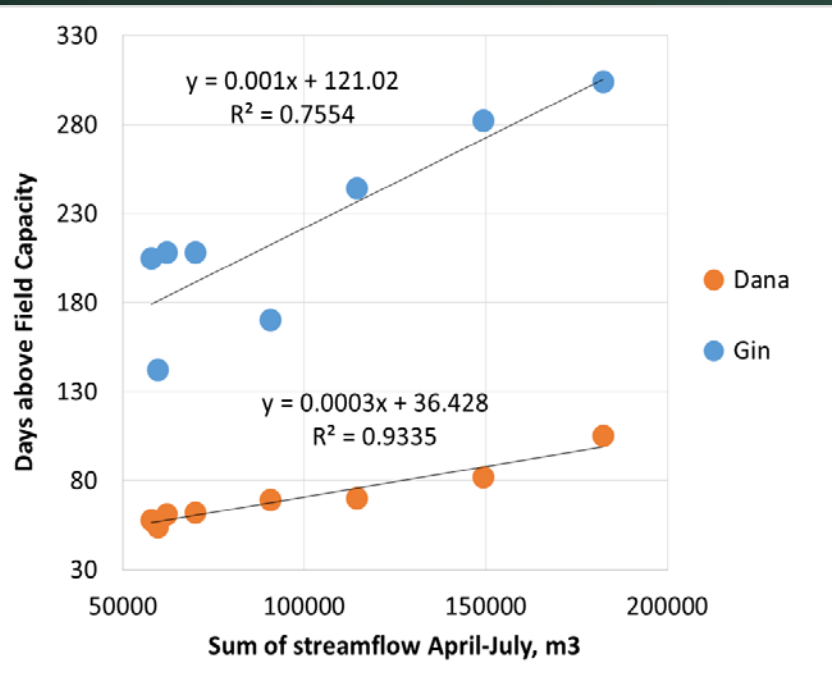
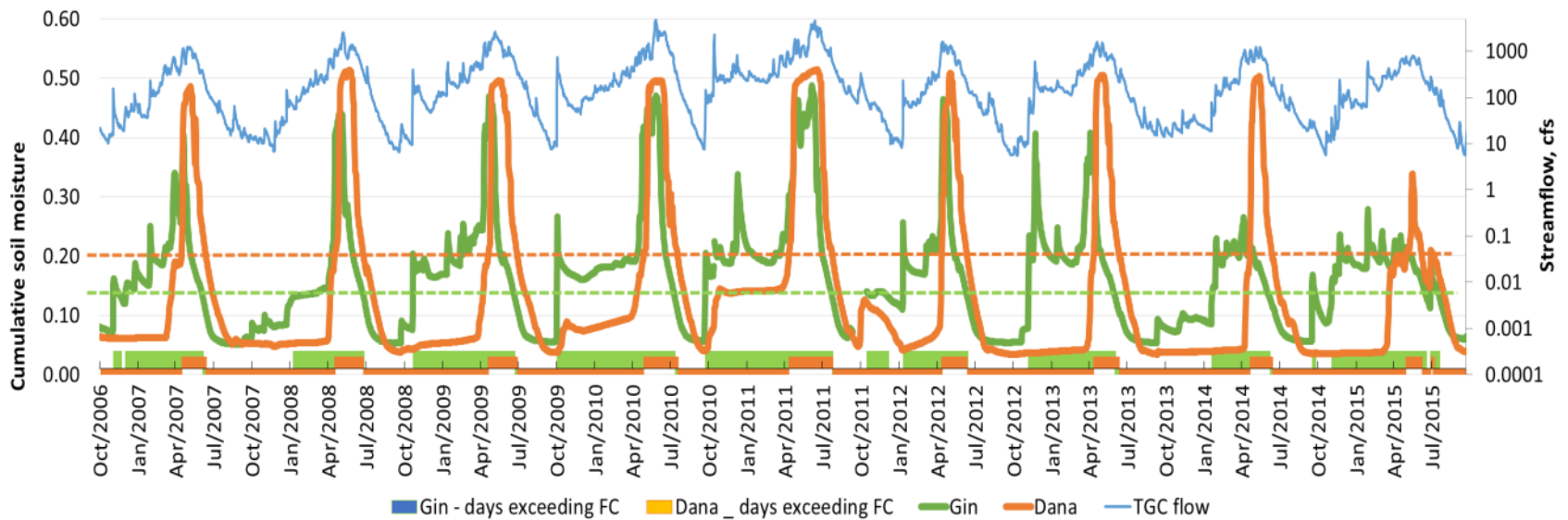
Saturation



Tuolumne River Basin/Hetch Hetchy Subbasin

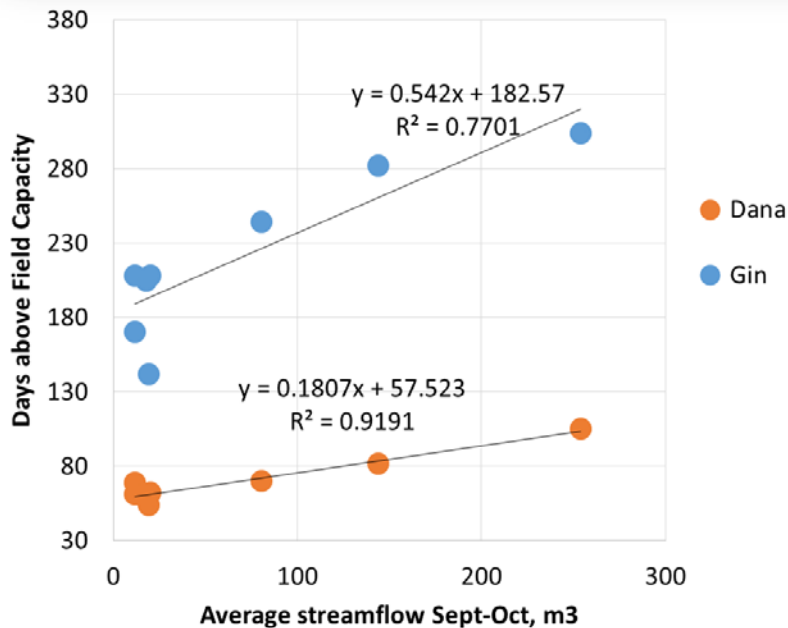
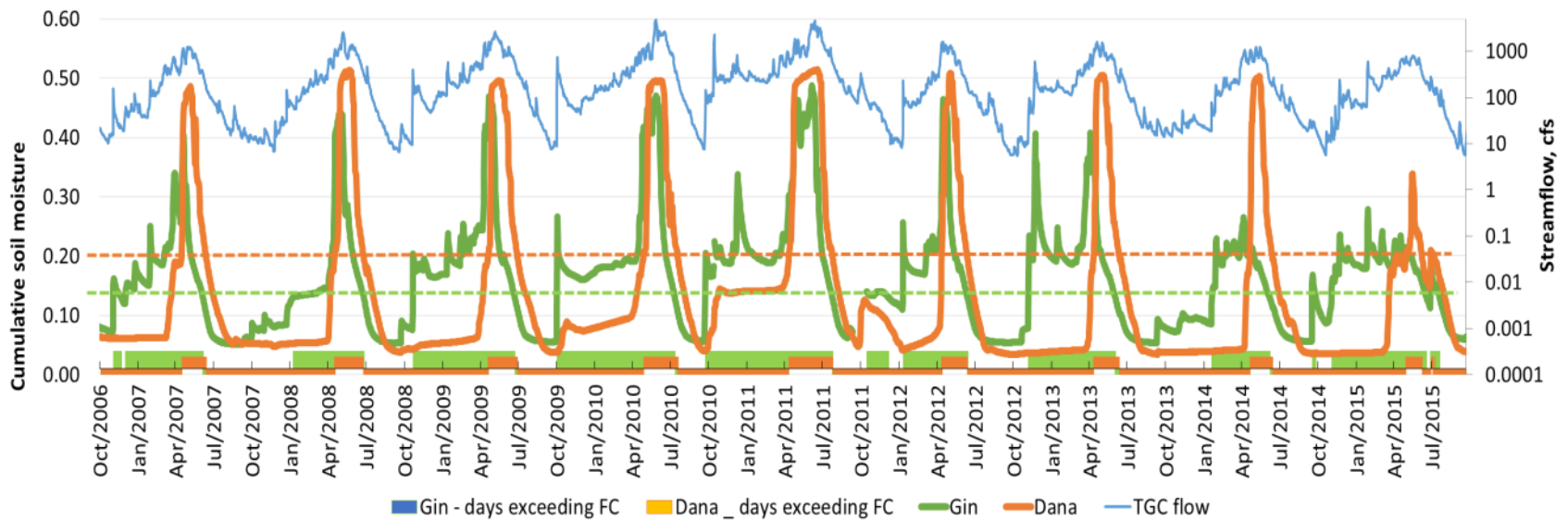


Threshold Field Capacity
Dana Meadows = 20% saturation
Gin Flat = 16% saturation



Forecasting Springtime Runoff

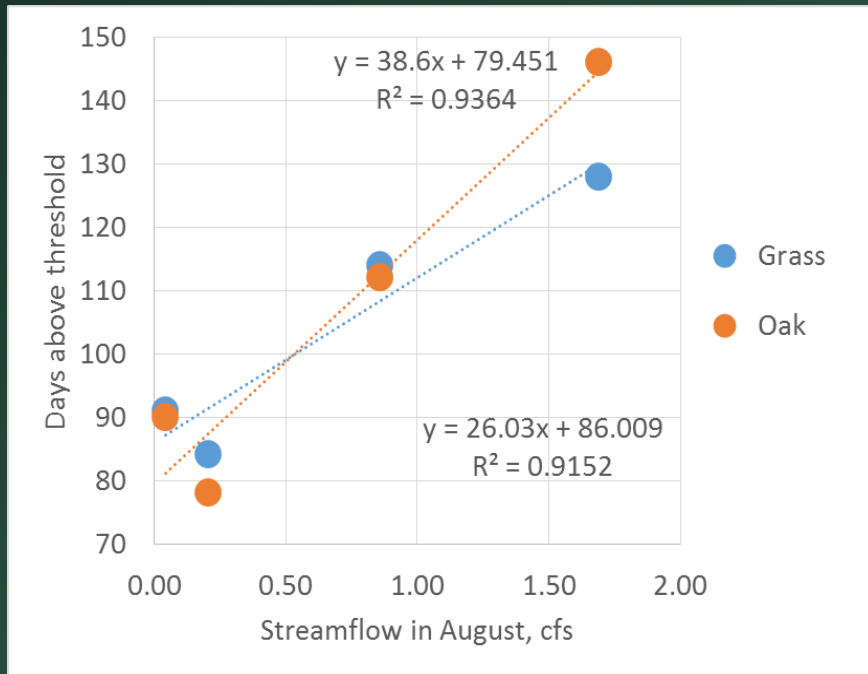
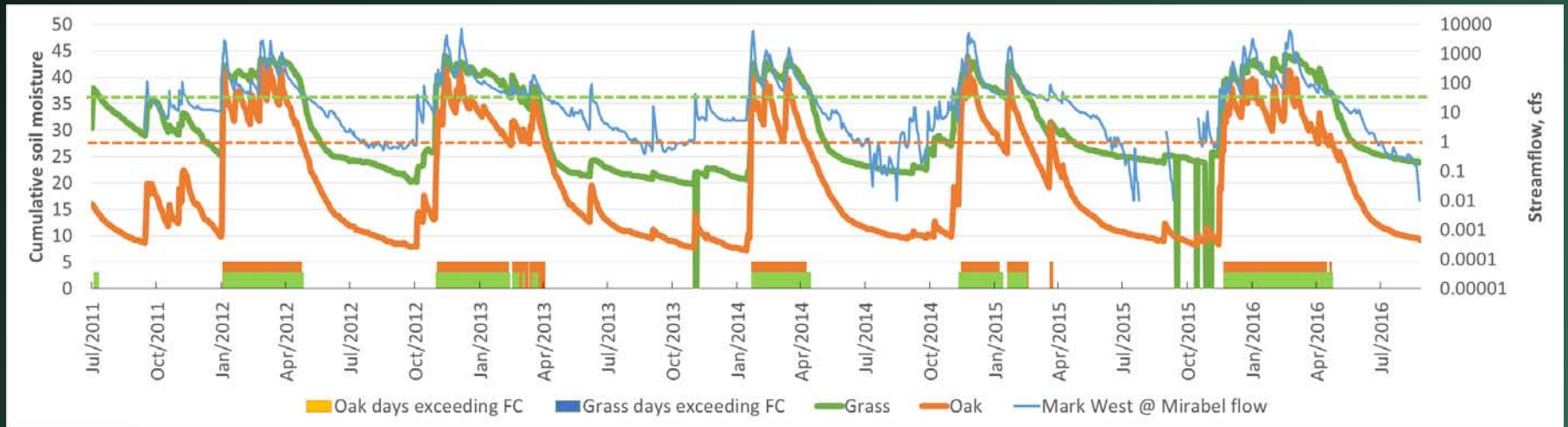
Threshold Field Capacity
 Dana Meadows = 20% saturation
 Gin Flat = 16% saturation



Forecasting Summer Low Flows

Threshold Field Capacity
 Dana Meadows = 20% saturation
 Gin Flat = 16% saturation

Sonoma County Pepperwood Preserve



Forecasting Summer Low Flows

Threshold Field Capacity
 Grass = 36.6% saturation
 Oak = 27.7% saturation

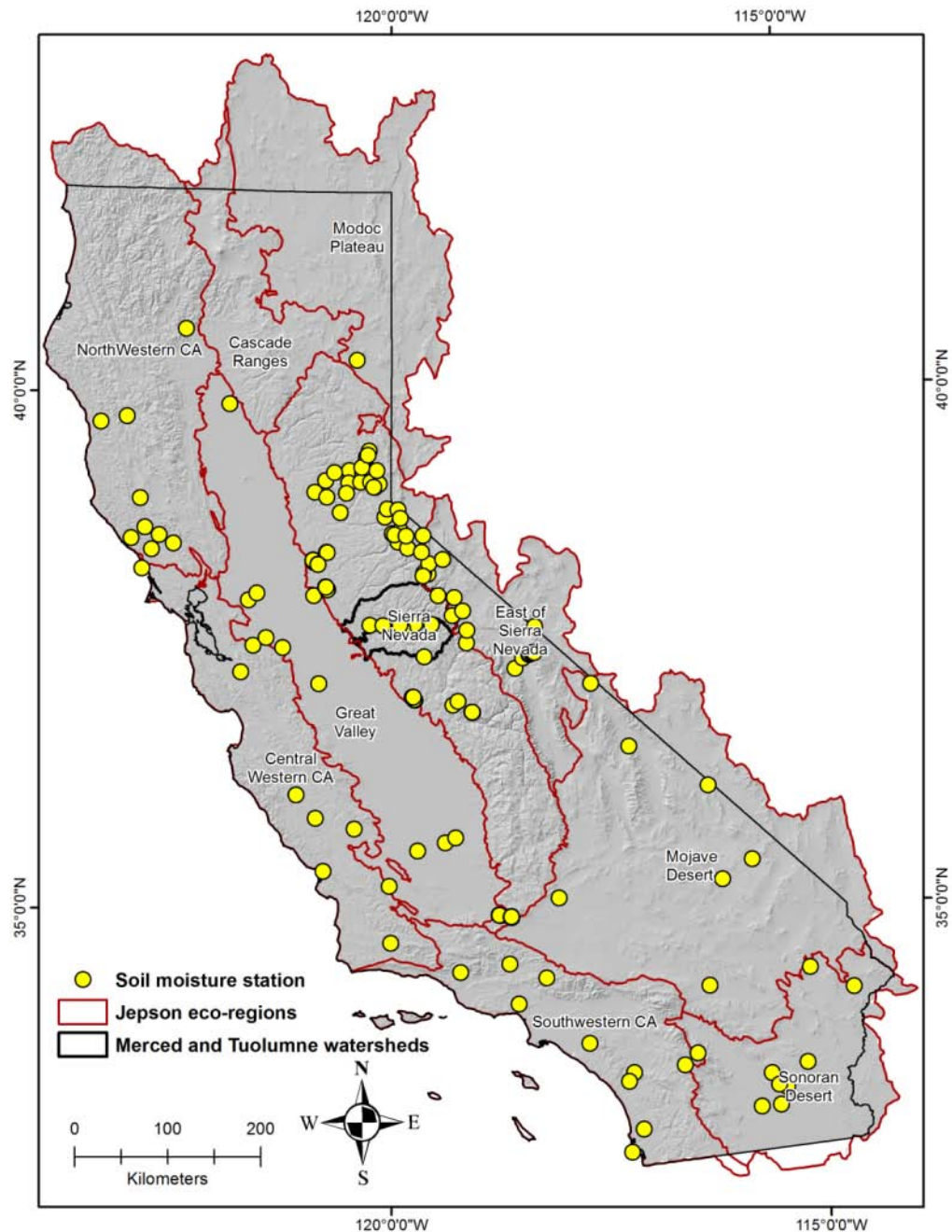


Soil moisture monitoring for forecasting

- **Snowpack is critical but not always the whole story**
- **Point measurements help us understand the state of the system at that location and may inform reservoir operations**
- **Soil moisture measurements can help to validate or constrain models, which can distribute and quantify processes for all watersheds**

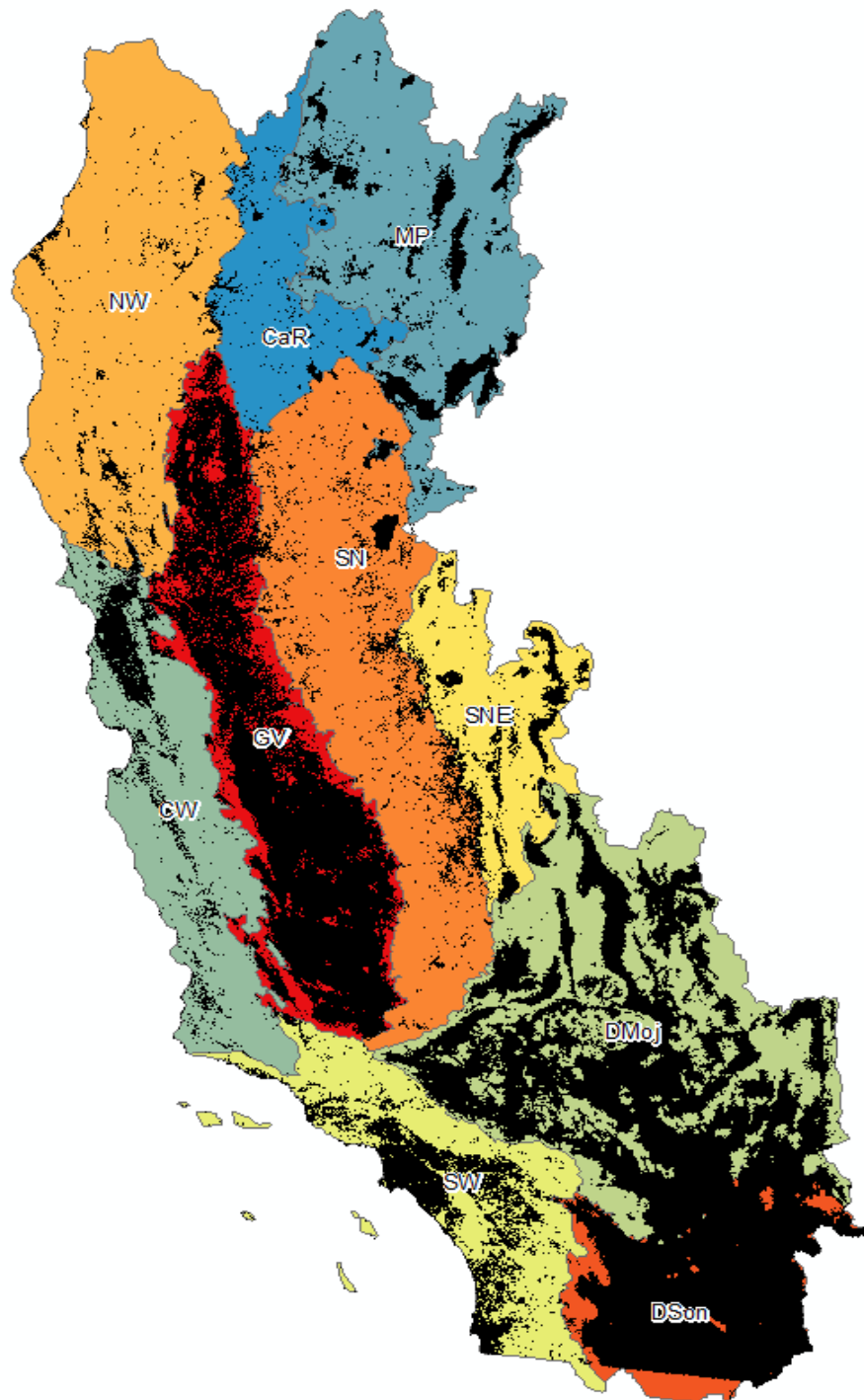
Monitoring Strategy

- **Statewide monitoring locations should**
 - **Mask out unlikely or non useful locations**
 - **Capture range of soil types**
 - **Capture range of climatic conditions**
 - **Capture range of soil responses to climate**
 - **Provide scaleable strategy for multiple applications from statewide to individual watersheds**



Jepson 10 boundaries

- NorthWestern CA
- Cascade Ranges
- Modoc Plateau
- Central Western CA
- Great Valley
- Sierra Nevada
- East of Sierra Nevada
- Southwestern CA
- Mojave Desert
- Sonoran Desert

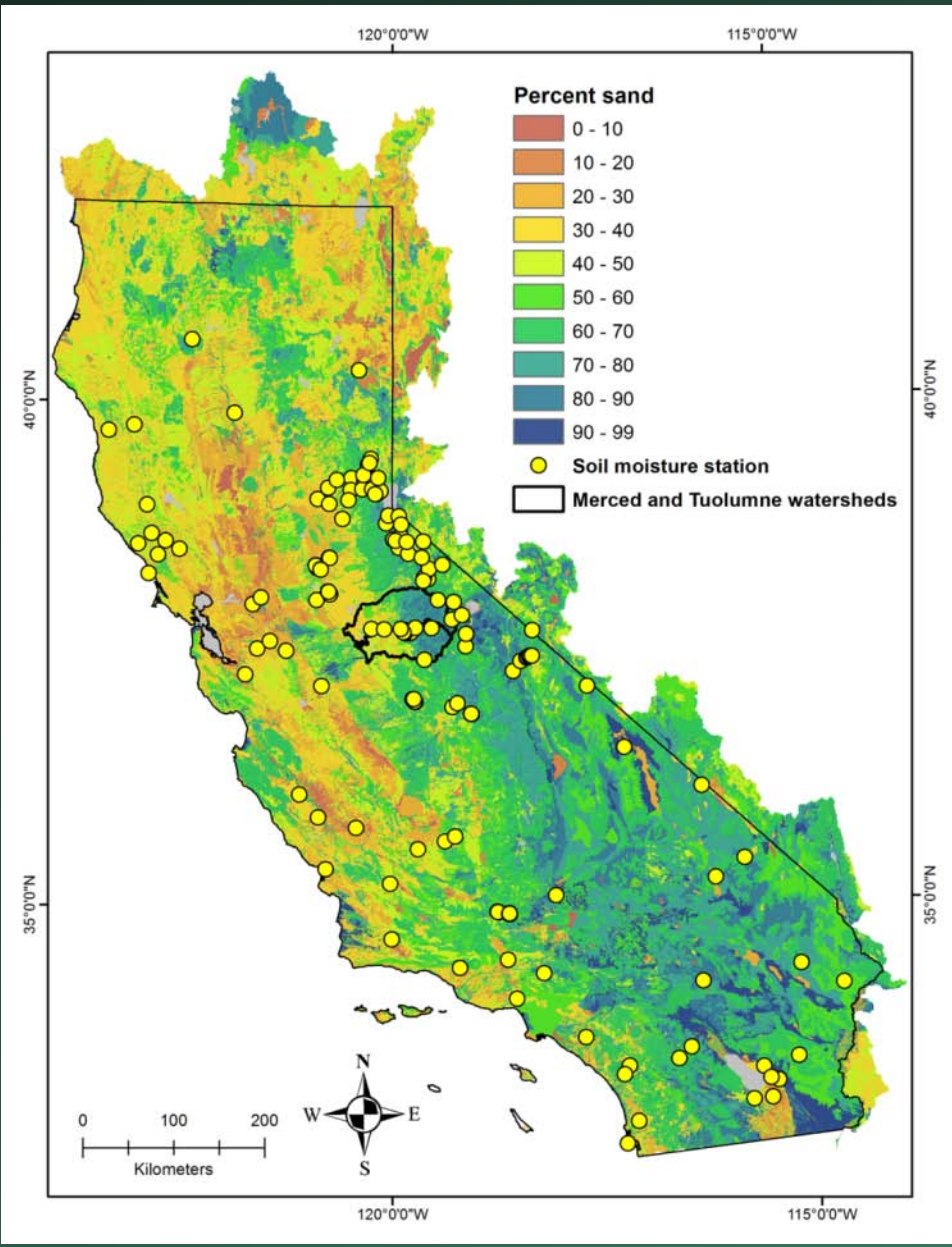


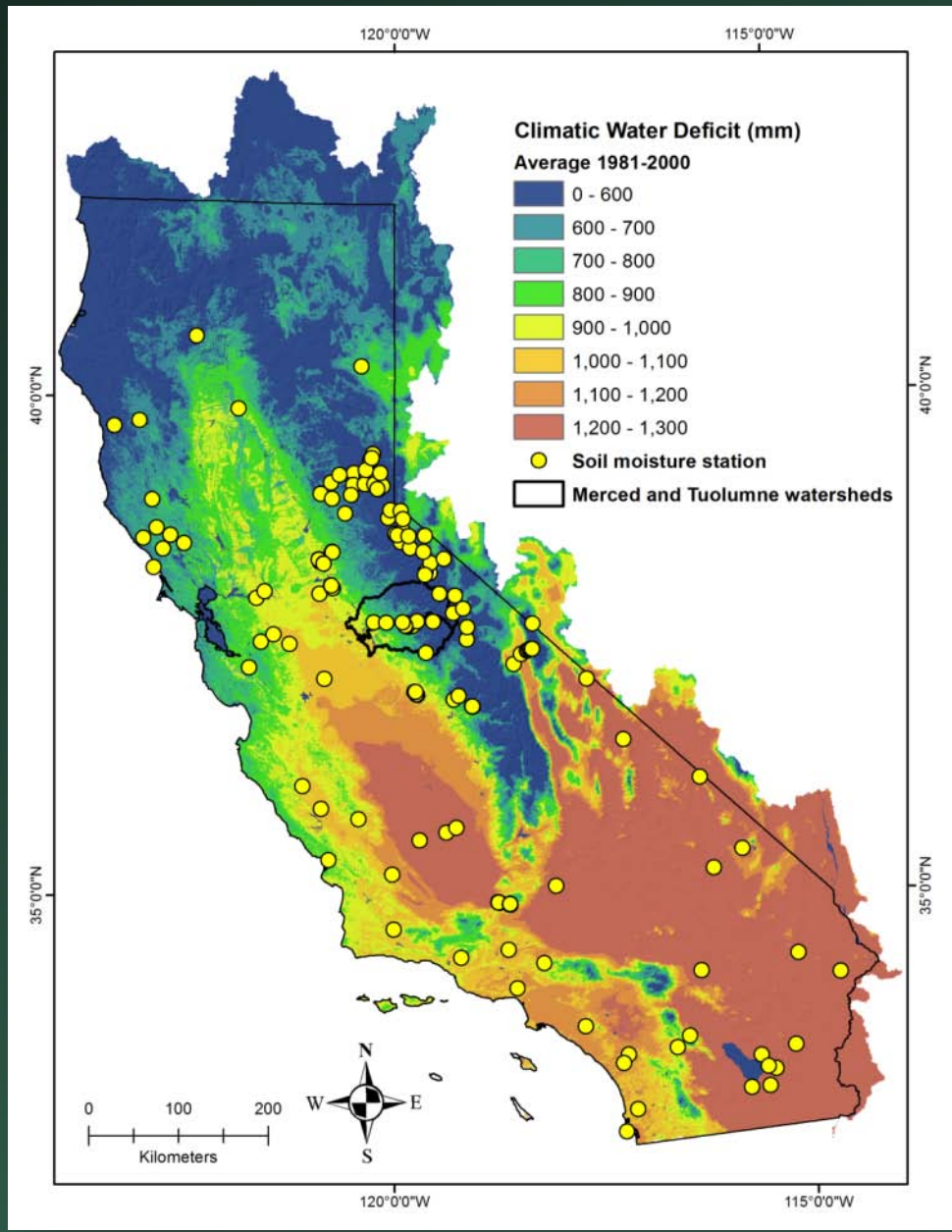
- Mask Layer Includes:
- WY98_rch+run <10mm
 - water
 - vineyard
 - urban
 - riverine
 - rice
 - orchVineyard
 - marsh
 - lacustrine
 - irrigRowFieldCrops
 - irrigHay
 - irrigGCrops
 - evergreenOrchard
 - estuarine
 - deciduousorchard
 - cropland
 - Barren

Soil moisture monitoring strategy

- Mask out unnecessary locations
- Select a range of variables that may describe soil moisture response units
- Scale variables 0-1
- Use principal components analysis to remove variables that are highly correlated
- Perform unsupervised cluster analysis
 - CA, ecoregions, watershed
- Topographic analysis for fine scale site selection







PCA California: Correlation ($\leq 50\%$): CWD and PET 93%
 Precip and Recharge 74%
 Precip and Runoff 81%
 Precip and CWD 65%
 Precip and PET 55%
 CWD and runoff 51%
 Porosity and AWC 49%

Run #1 - CA

Name	layer	variance	Eigen value	Percent of Eigen values	Percent Total
cwd	8	0.0097	0.0198	52.16	52.16
psand	5	0.0092	0.0086	22.61	74.77
precip	3	0.0050	0.0043	11.36	86.14
pet	6	0.0049	0.0021	5.56	91.69
rch	10	0.0032	0.0014	3.59	95.29
run	9	0.0024	0.0009	2.26	97.54
porosity	2	0.0013	0.0006	1.65	99.20
awc	4	0.0012	0.0003	0.75	99.95
snow	7	0.0011	0.00002	0.05	99.999
rch/run	1	0.000001	0.000001	0.0014	100

PCA California

CWD

Percent sand

Precipitation

Porosity

Snow (st. dev)

Run #2 - CA

Name	layer	variance	Eigen value	Percent of Eigen values	Percent Total
cwd	5	0.0097	0.0139	52.83	52.83
psand	3	0.0092	0.0083	31.47	84.30
precip	2	0.0050	0.0024	9.17	93.47
porosity	1	0.0013	0.0009	3.43	96.89
snow	4	0.0011	0.0008	3.11	100.00

PCA Sierra Nevada:

Correlation: CWD and PET 89%
Precip and Runoff 71%
CWD and Precip 64%
Precip and CWD 65%
Precip and Recharge 57%
CWD and runoff 52%
Precip and PET 49%

Run #1 – Sierra Nevada

Name	layer	variance	Eigen value	Percent of Eigen values	Percent Total
psand	5	0.0076	0.0112	42.82	42.82
cwd	8	0.0056	0.0076	29.17	71.99
pet	6	0.0027	0.0026	9.98	81.97
precip	3	0.0026	0.0018	7.05	89.02
rch	10	0.0024	0.0010	3.90	92.91
run	9	0.0019	0.0009	3.30	96.22
porosity	2	0.0013	0.0007	2.74	98.95
snow	7	0.0012	0.0003	0.98	99.93
awc	4	0.0009	0.00002	0.06	99.997
rch/run	1	0.000001	0.000001	0.0031	100

PCA Sierra Nevada:

CWD

Percent sand

Precipitation

Porosity

AWC

Snow (st.dev)

Run #2 – Sierra Nevada

Name	layer	variance	Eigen value	Percent of Eigen values	Percent Total
psand	4	0.0076	0.0086	44.84	44.84
cwd	6	0.0056	0.0069	35.81	80.65
precip	2	0.0026	0.0012	6.46	87.12
porosity	1	0.0013	0.0010	5.16	92.28
snow	5	0.0012	0.0009	4.45	96.73
awc	3	0.0009	0.0006	3.27	100.00

Understanding multivariate classification

- In an **unsupervised** classification, you do not know what features are actually at any specified location, but you want to aggregate each of the locations into one of a specified number of groups or clusters.
- Assignments to each class or cluster is dependent on the **multivariate statistics** that are calculated on the input rasters.
- Each location is characterized by a set or vector of values and can be visualized as a point in a multidimensional attribute space.
- A class or cluster is a grouping of points in this multidimensional attribute space. Two locations belong to the **same class or cluster if their attributes (raster values) are similar**.

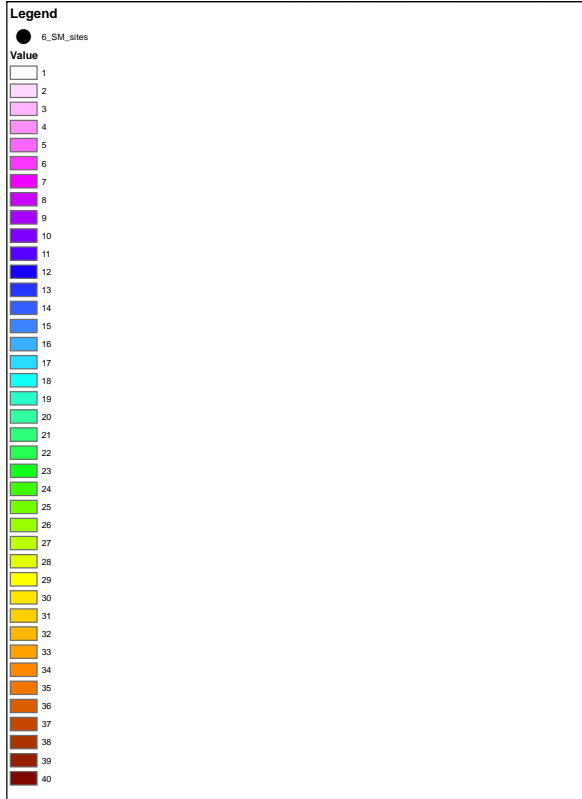
CA 30 class – cluster analysis

30

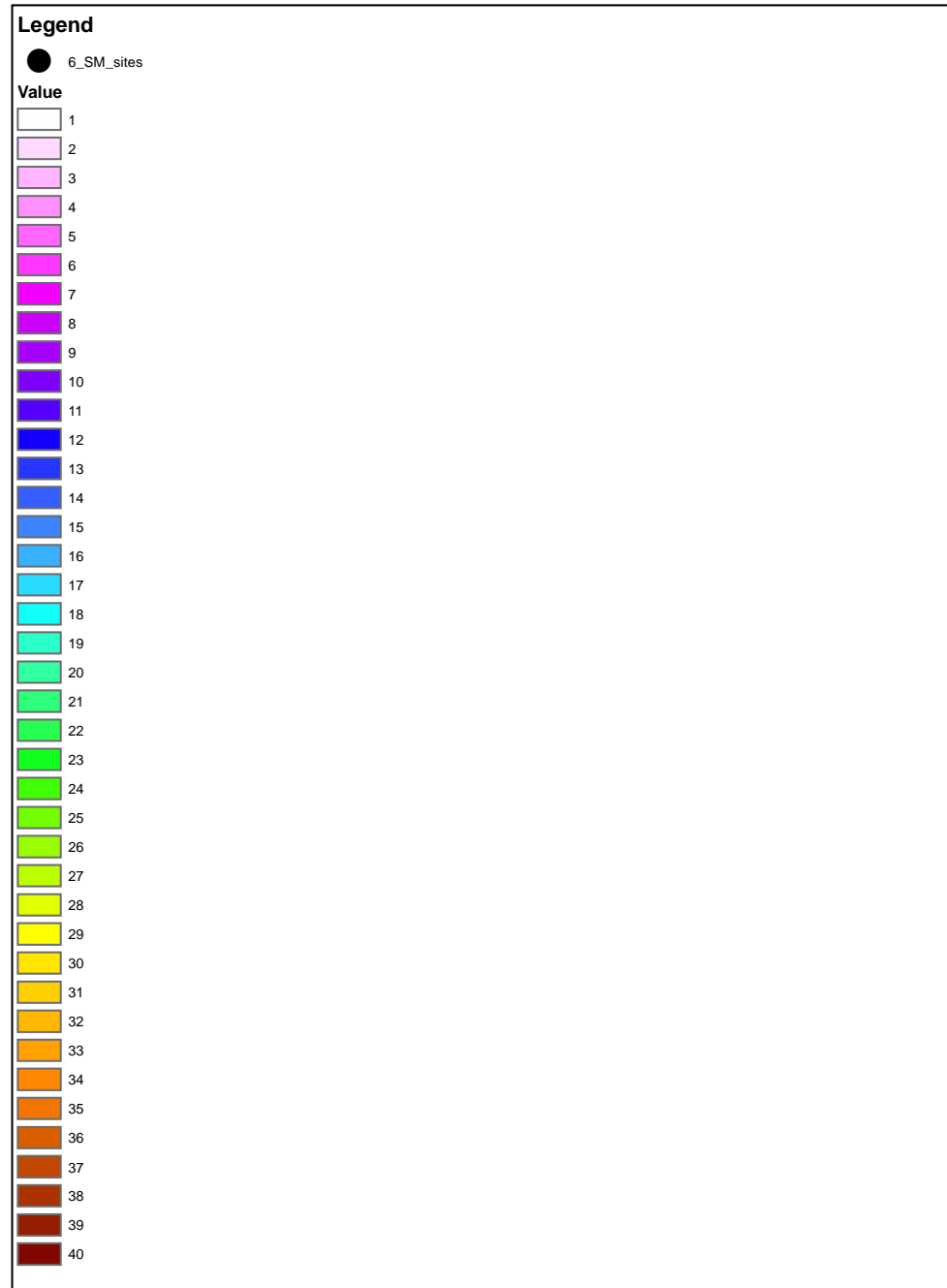
GinFlat, Forty Mile, Dana,
Smoky Jack, Olmstead = Class 17
Merced Grove = Class 14

30

CA 40 class - cluster analysis



Ginniat, Forty mile = Class 25
Dana, Smoky Jack, Olmstead = Class
Merced Grove = Class 23



SN 10 class – cluster analysis

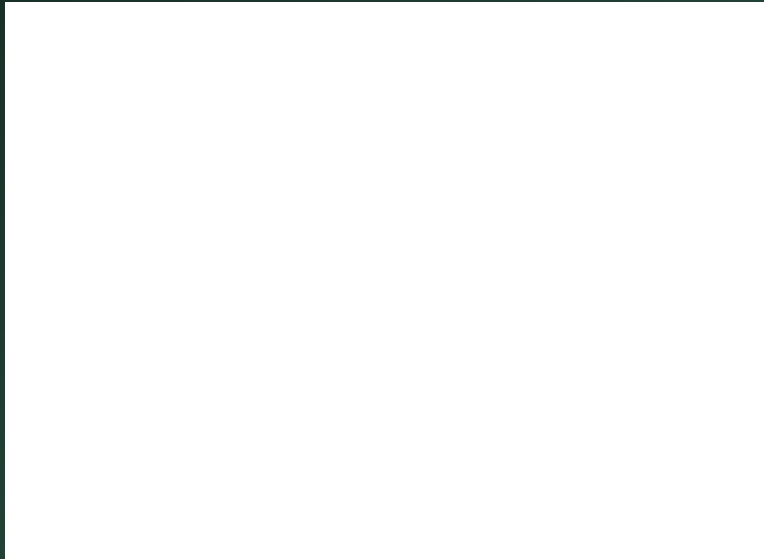
10

Dana = Class 3

Merced Grove = Class 7

GinFlat, Forty Mile, Smoky Jack, Olmstead = Class

Merced -Tuolumne 10 class – cluster analysis



Dana (2965m) = Class 7
Olmstead (2601m) = Class 8
GinFlat (2149m),
Smoky Jack (2180m) = Class 9
Merced Grove (1810m),
Forty Mile (1720m) = Class 10

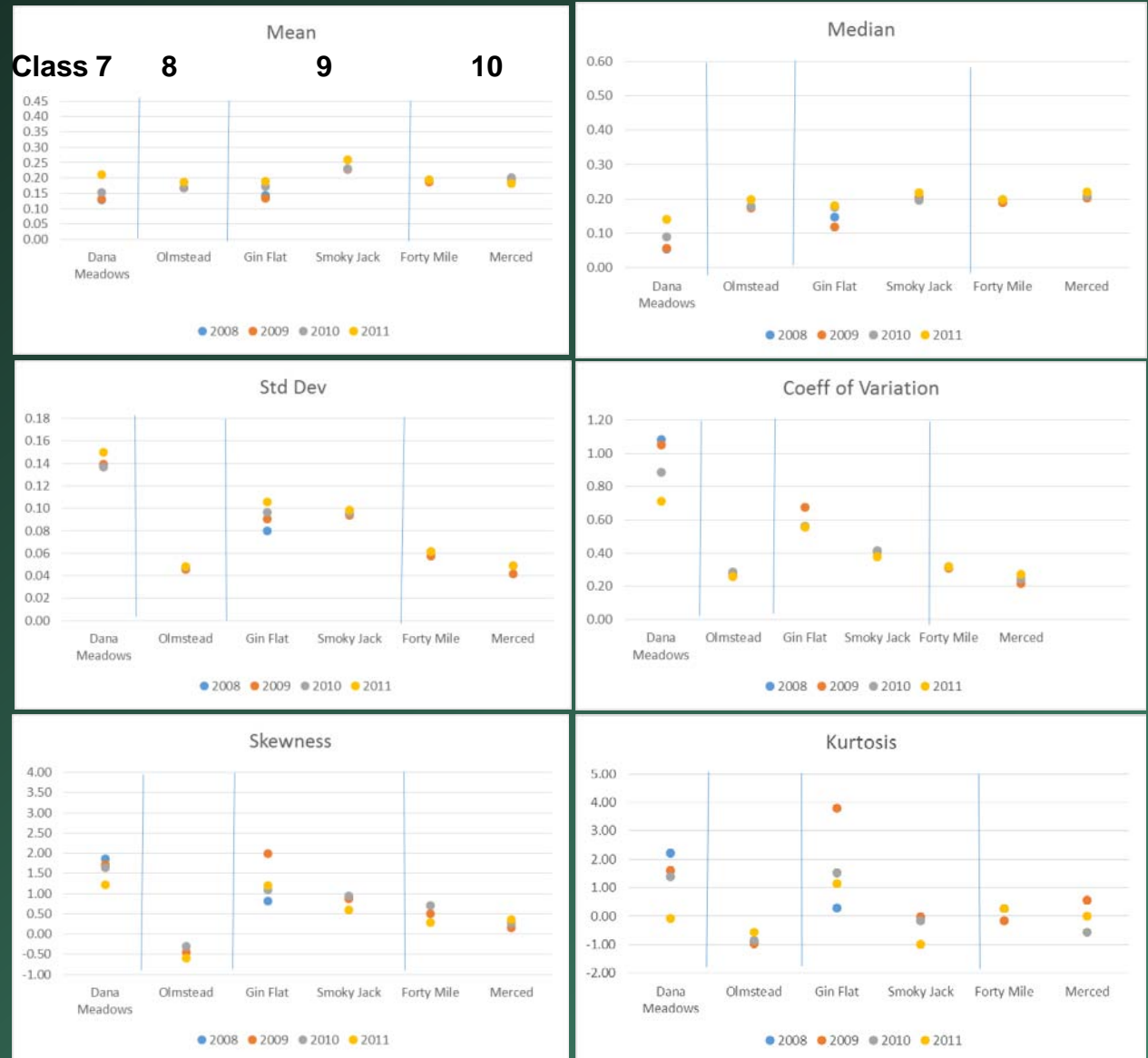
Merced -Tuolumne 6 class – cluster analysis



Merced Grove (1810m) = Class 3
Dana (2965m) = Class 4
Olmstead (2601m), Smoky Jack (2180m),
GinFlat (2149m),
Forty Mile (1720m) = Class 5

Verification annual WY data

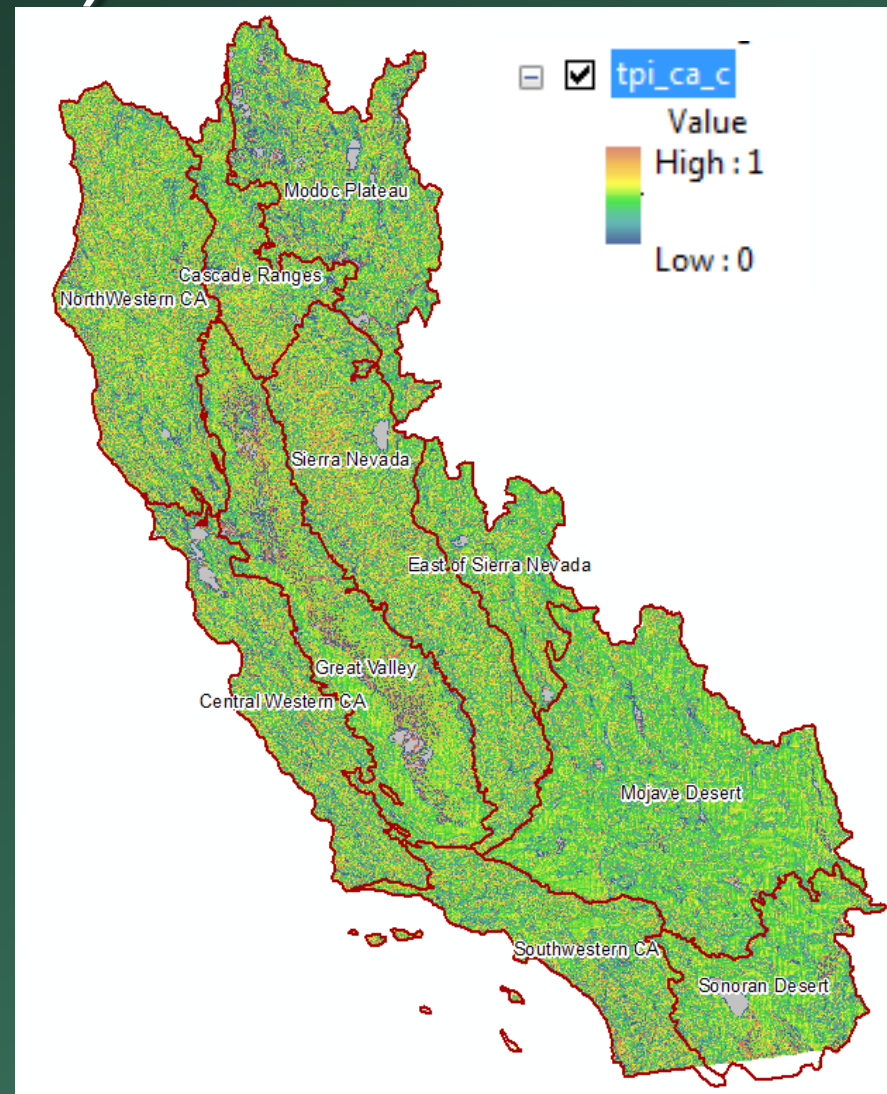
- ANNUAL WY DATA
- Looked at statistics using MT 10 classes
- Class 7 – lowest median values and highest CV values
- Class 8 – very little among year variability in any of the stats
- Class 9 – higher variability in SKEW and KURT, relatively high SD and CV but lower than class 7
- Class 10 – very little among year variability in Mean, Median, SD, and CV













Topographic Position Index (TPI)

- Topographic position index (*TPI*) is an algorithm increasingly used to measure topographic slope positions and to automate landform classifications.

$$TPI = (DEM - \text{focal min}) / (\text{focal max} - \text{focal min})$$



LC_ca_1000

-  canyons, deeply incised streams
-  midslope drainages, shallow valleys
-  upland drainages, headwaters
-  u-shaped valleys
-  plains
-  open slopes
-  upper slopes, mesas
-  local ridges, hills in valleys
-  midslope ridges, small hills in plains
-  mountain tops, high ridges



Merced -Tuolumne
10 class –
cluster analysis



Merced -Tuolumne
6 class –
cluster analysis



Soil Moisture Datasets: Yosemite National Park, Sierra Nevada and Pepperwood Preserve, Coastal Ranges, California

Prepared in cooperation with the Department of Water Resources

By Michelle Stern, Frank Anderson, Lorraine E. Flint and Alan L. Flint



Soil conditions before the onset of winter precipitation in California affect surface runoff, sub-surface water storage, and overall water supply management. (Photos taken by Frank Anderson)

USGS OFR describing soil moisture and water potential from 3 sites in YNP (2007-2016) and 2 sites in Sonoma County (2011-2016)