



Applying Remote Sensing Metrics to Quantify Invasive Annual Plant Distributions and Map Potential Fire Hazard Across Mojave Landscapes

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Fire Science Management for the Mojave Ecoregion—Barstow, CA, June 11, 2014

Invasive Species Detection and Fire Hazard: Objectives

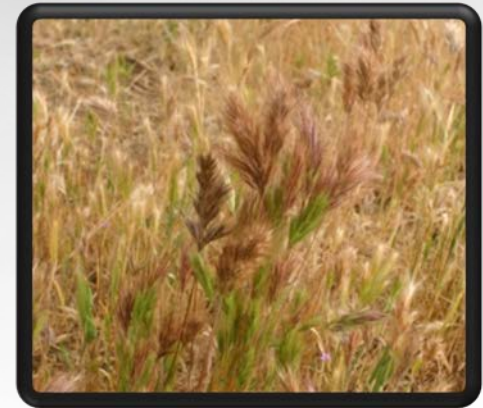


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- ▶ Develop and evaluate models to predict relative abundance of nonnative annuals using environmental attributes and remotely sensed data
- ▶ Construct relationship between measured cover and biomass that could be related to potential fire risk
- ▶ Provide model and tools to evaluate and compare relative fire hazard for Mojave

Bromus rubens &
B. tectorum

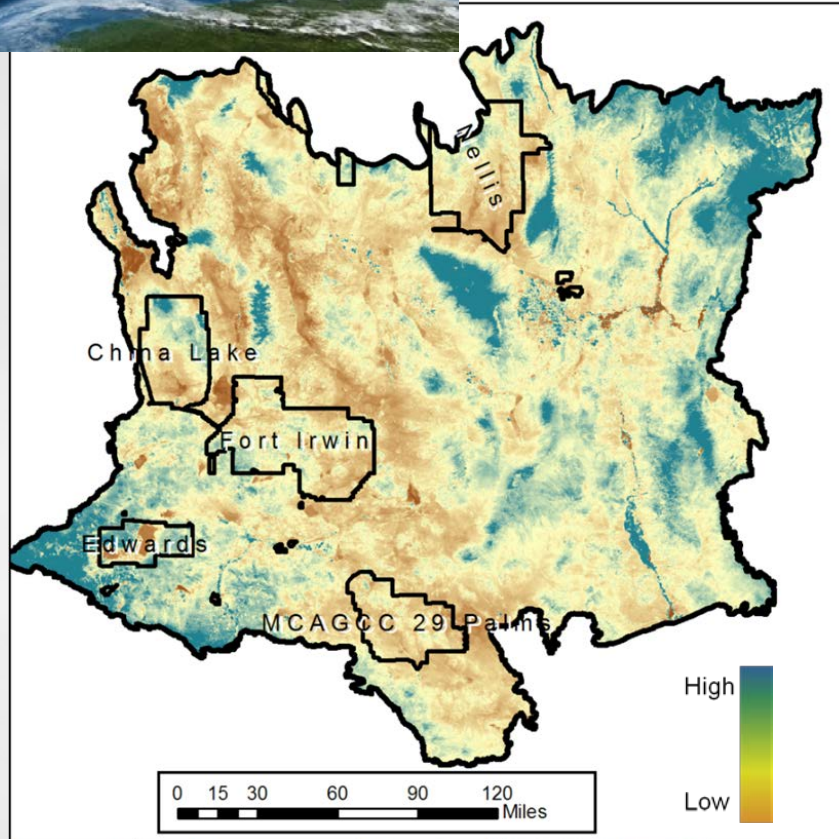


Schismus arabicus & *S. barbatus*

Erodium cicutarium



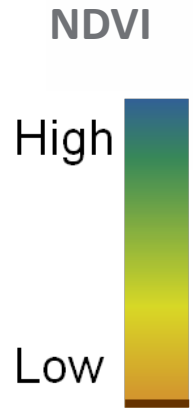
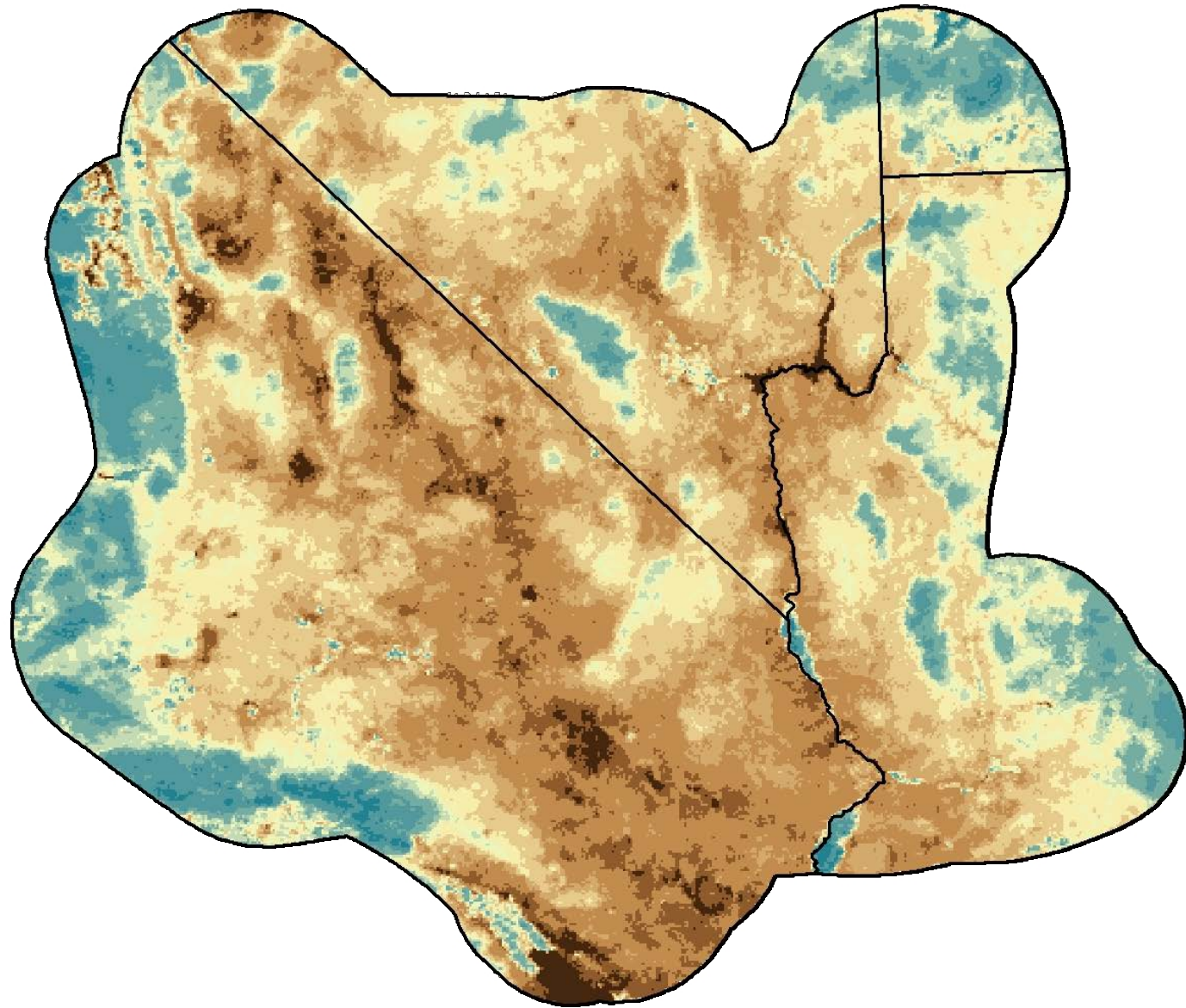
Invasive Species Detection and Fire Hazard Analysis



- ▶ MODIS weekly NDVI composite images
 - From 2000-2014
 - USGS eMODIS product
 - 250 m pixel
- ▶ Utilize weekly MODIS NDVI data to describe phenology and relative productivity; develop models for current and past distribution of nonnative annuals.
- ▶ Develop landscape models to assess relative annual fire hazard



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December

August





Data > Field Plot Measurements

- ▶ 600 plot locations
- ▶ Collected during 2 campaigns (2009 & 2011)
- ▶ 50-meter plot size
- ▶ Cover measured for all species in subplots
- ▶ Biomass collected in a subset of the 2011 plots



Plotcode	x	y	AlnannfbTotal	AlnAnnfb+erocic	broarv	brorub	brotec	hordsp	schara	schbar	SchismusTotal	vulbro	Alnanngr Total	erocic	Alnbielb Total	AlnperfbTotal	AlnpergrTotal	baresoil	litter
66SER11	667150	3902163	0.0	2.3	0.0	0.1	0.0	0.0	0.0	0.0	0.0	0.0	0.1	2.3	2.3	0.0	0.0	75.3	1.7

Phenological signatures or “Phenometrics”

Calculated for the hydrologic year between beginning of October and end of following September

Peak NDVI —

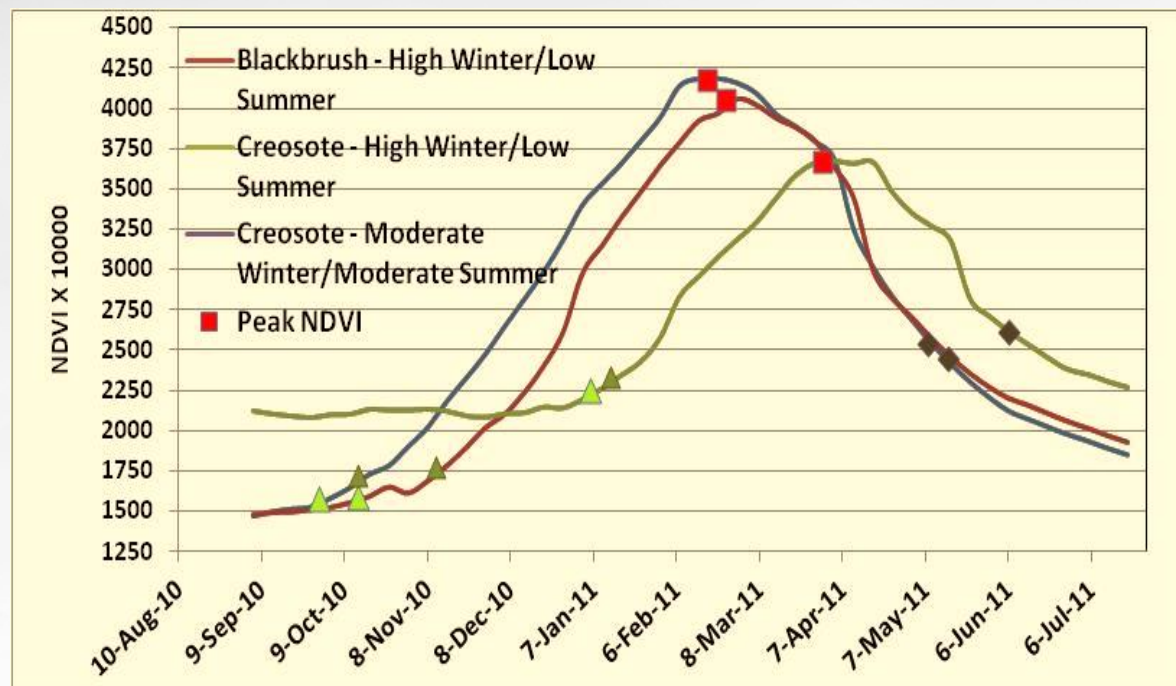
Maximum weekly NDVI during the growing season.

Start of Season (SOS) —

Weeks during which green-up is detected

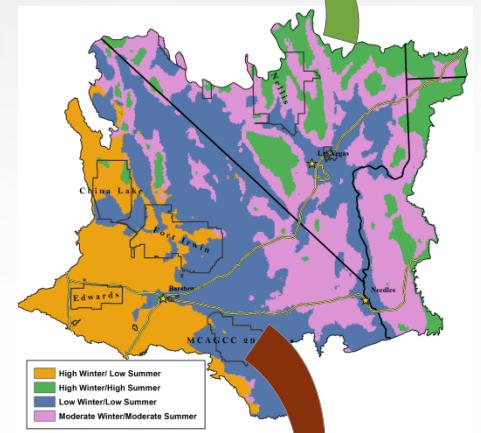
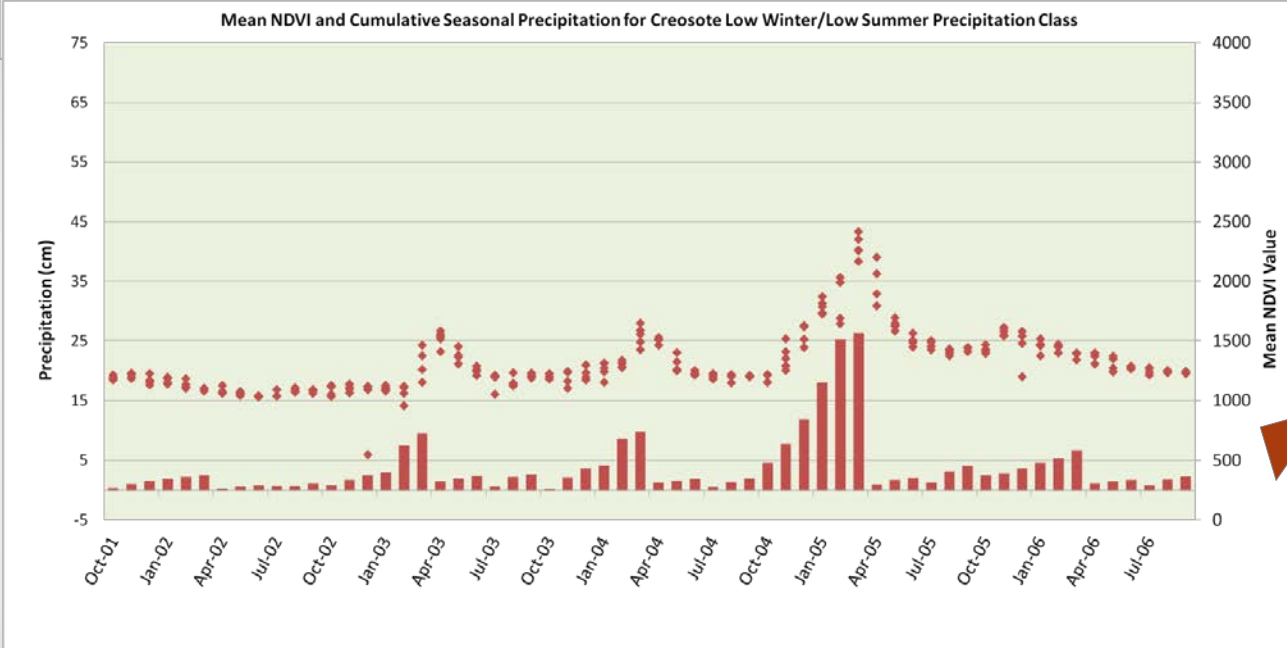
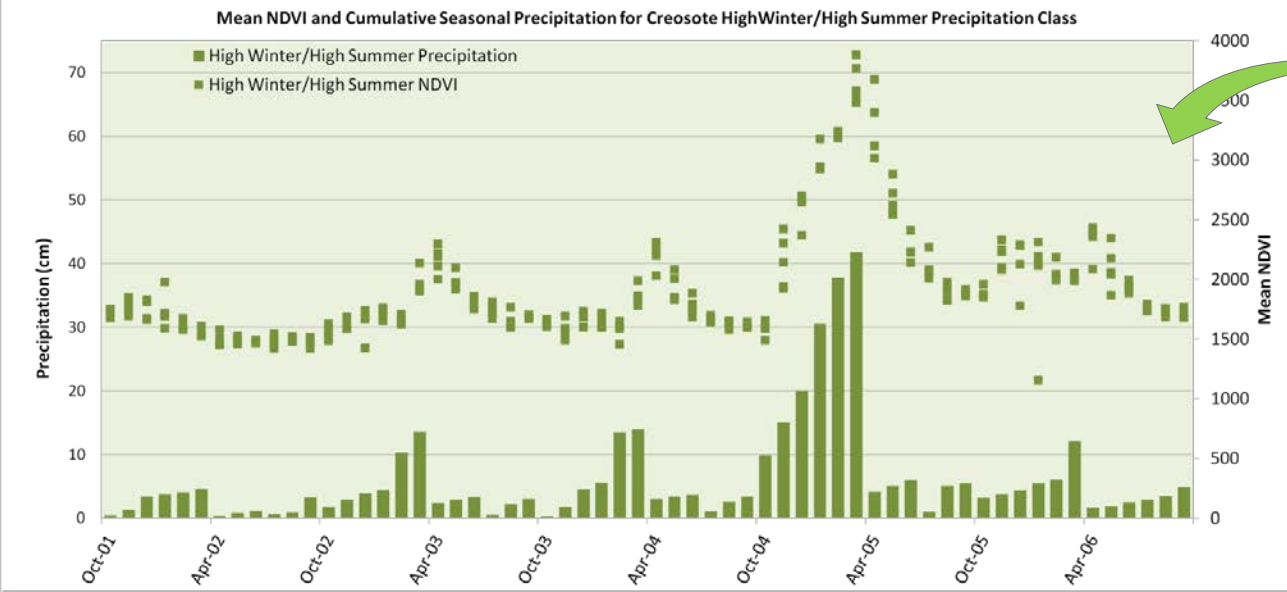
End of Season (EOS)—

Week or month where senescence is nearly complete





Vegetation response to rainfall differs by zone





Data > Field Plot Measurements

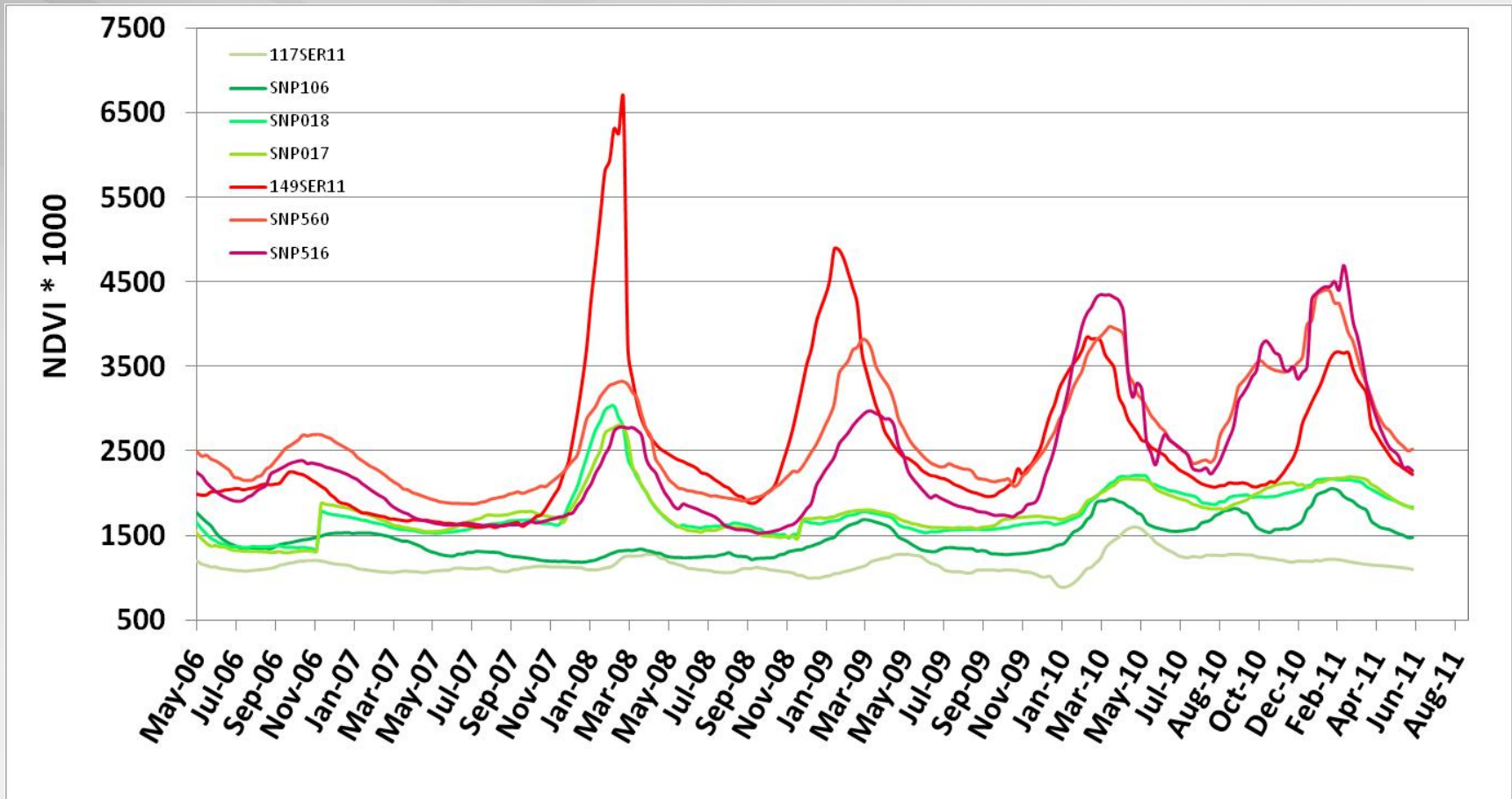
- ▶ More than 600 plot locations
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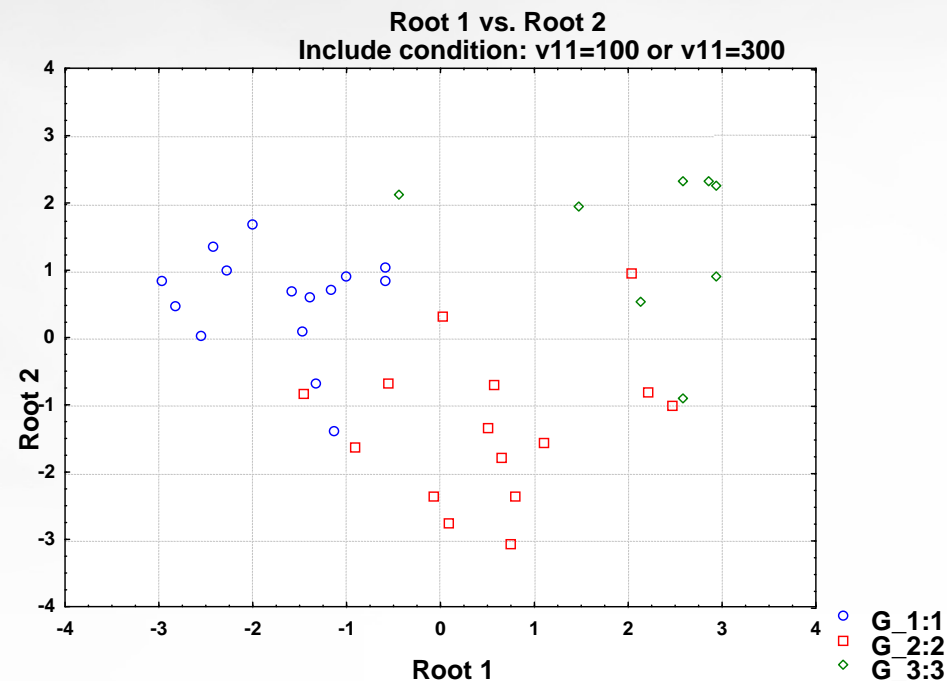
Data—NDVI Signatures For Field Locations with >25% Cover NIA and <5% Cover NIA



NDVI signatures for plots in Blackbrush in the
Moderate Winter/Moderate Summer Precipitation Class
Noisy data smoothed via Savitsky-Gorlay filtering

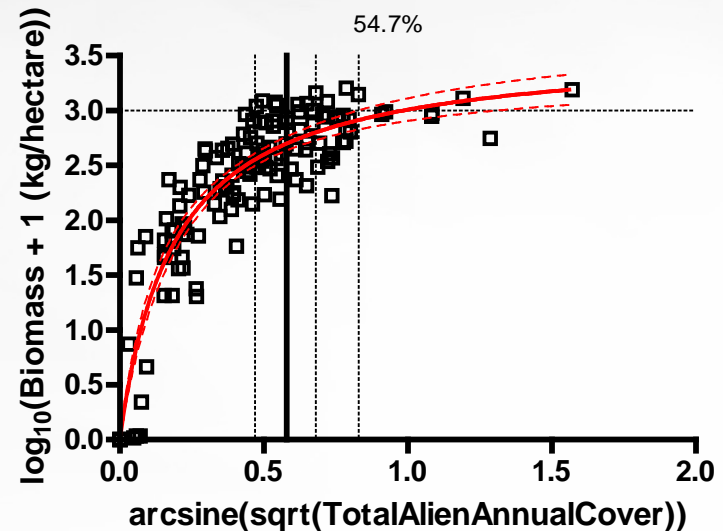
Detecting and Mapping Nonnative Annual Species

- ▶ Vegetation data from 515 field locations assigned categories relevant for predicting fire risk
 - Low nonnative annual cover ($\leq 10\%$)
 - Intermediate (10% to 40%)
 - High ($> 40\%$)
- ▶ 25% of dataset withheld for validation
- ▶ Discriminant models
 - standardized variables
 - by precipitation regimes and precipitation regime x vegetation association.
- ▶ Predictor variables:
 - Cumulative winter precipitation, monthly precipitation
 - Start of season (SOS) NDVI
 - Peak NDVI
 - Slope between SOS and Peak NDVI



Detecting and Mapping Nonnative Annual Species

- ▶ Relate measured nonnative annual cover to biomass measures
- ▶ N=128, $R^2 = 0.82$



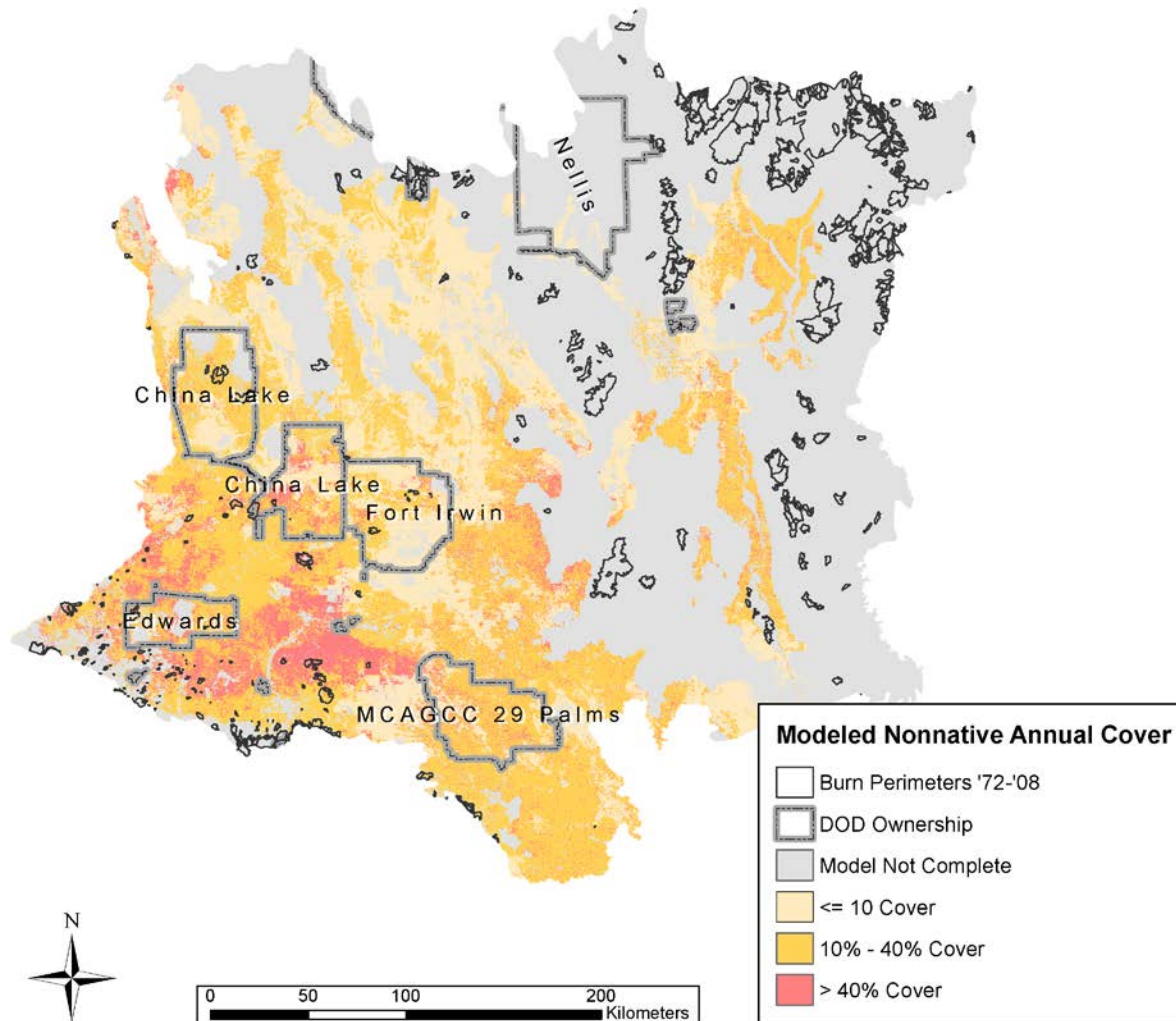
$$y = (b_{\max} \cdot X) / (K_d + X)$$

Fine Fuel Thresholds ^(a)	Log10 (biomass+1)	Cover = 0.0474*biomass
200 kg/h	2.30103	9.48
1000 kg/h	3	47.4

(a) based on values from Rao et al. (2010), Brooks et al. (2007), Brooks (1999), and Brooks (2002).

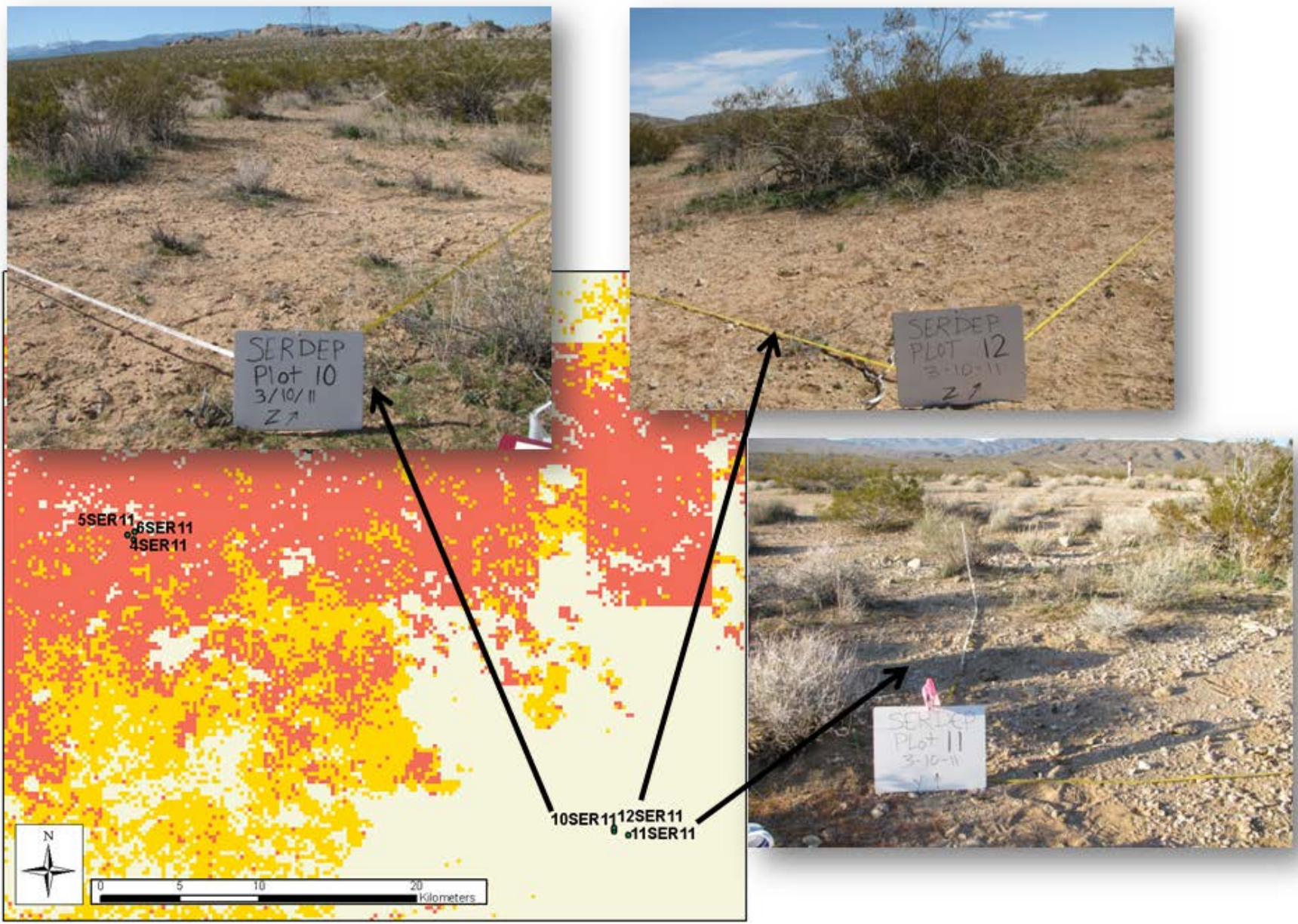
Model Testing and Validation

Model	Veg+Precip Classes	Modeled Sample Size	Validation Sample Size	Modeled Correct Classification	Modeled Maximum Error Size	Modeled Percent errors > 5	Validation Maximum Error Size	Validation Percent Errors > 10	1-Percent Errors > 10
M1	100, 300, 412 < 1720 m	38 +10(412)	22	86.8 (50% for 412)	9.96 (13.8 for 412)	8% (30% for 412)	4.7	0%	100%
M2	202, 205, 208, 210, 211	25	11	96	7.4	4%	37.4	27%	73%
M3	204 < 896 m	39	17	76.9	25.2	8%	47.1	24%	76%
M4	204 >= 896 m	44	8	77.3	10.7	7%	7.1	0%	100%
M5	403, 404, 406, 407, 410, 411	36	12	75.0	14.8	17%	35.6	17%	83%
M6	401, 405	57	23	82.5	30.2	12%	11.3	9%	91%
M7	412 >= 1720 m	33	8	75.8	11.9	9%	8.1	0%	100%
M8	402 >= 1200 m	62	15	75.8	36.6	8%	32	13%	87%
M9	402 < 1200 m	40	11	82.5	22.5	13%	31.8	27%	73%



Example of Model Results for 2010-2011 Growing Season

- Represents two precipitation regions
 - High Winter/Low Summer
 - Low Winter/Low Summer
- **88% correct classification based on limited validation dataset**
- Represents **57% of the study area**
- **Weighted mean classification accuracy = 75% for initial discriminant models**

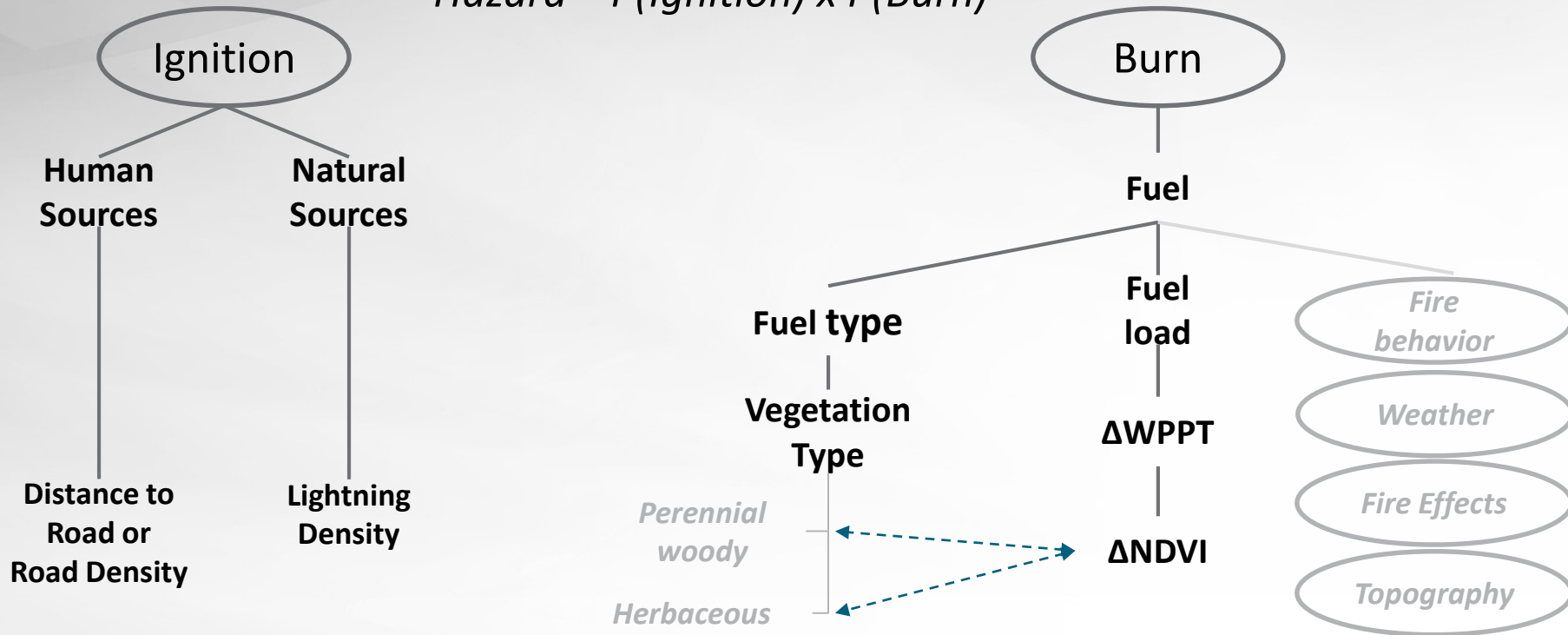


Model Development

- ▶ Model based primarily on 2009 and 2011 field data
- ▶ Further data collection occurred during 2012 and 2013 for development and validation
 - Dry years – little growth of invasive annuals
- ▶ Existing models not robust
- ▶ Step back to drawing board
 - More complex models?
 - More data?
 - Logistics for incorporating current data and complex models into accessible information in timely manner
- ▶ What other metrics or modeling approaches can we use to evaluate potential fire hazard

Conceptual Model for Preseason Assessment of Fire Hazard

$$\text{Hazard} = P(\text{Ignition}) \times P(\text{Burn})$$

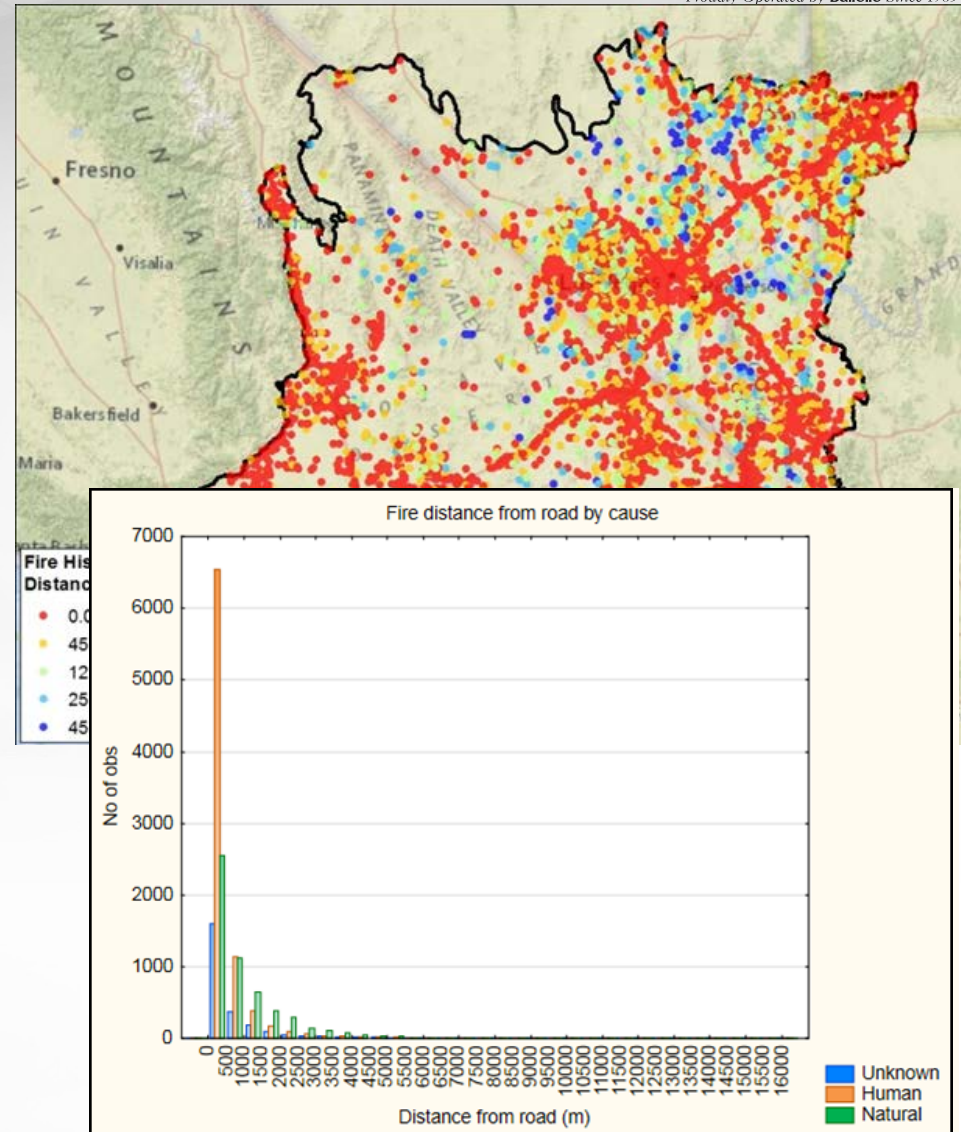


$\Delta\text{NDVI} = \text{peak NDVI in current year (i)} - \text{median of peak NDVI}_{2000-2010}$

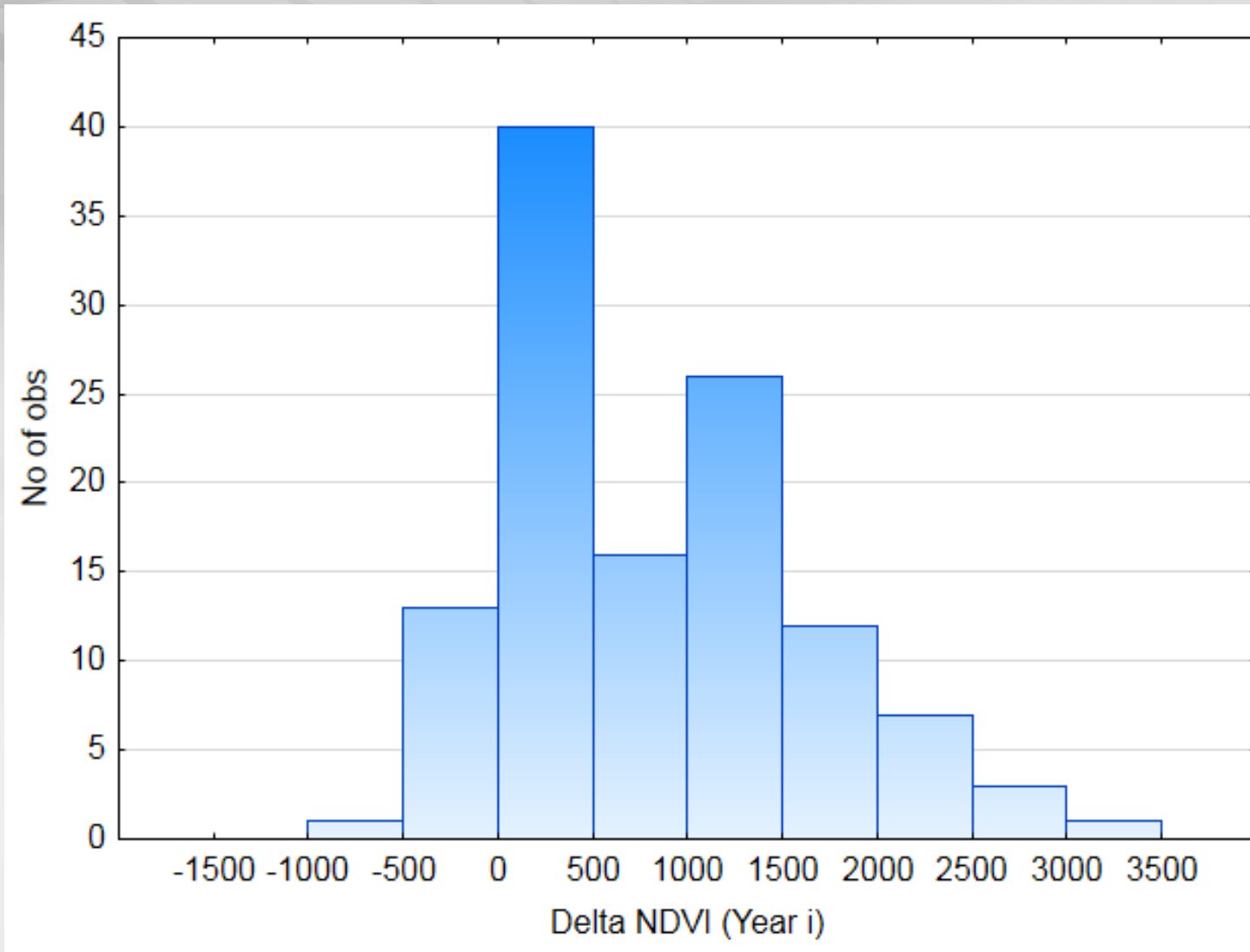
Mean $\Delta\text{WPPT} = \text{Winter precipitation for current year} - \text{median winter precipitation}_{1971-2010}$

Preseason Fire Hazard Assessment

- ▶ Use the database of fire starts for 1980 to 2012
- ▶ 81% of fire **starts** occur <1000 m from a road
- ▶ 52% human-caused, 33% natural, 15% unknown cause
- ▶ Inverse relationship between all fire types and distance from road is function of road density in Mojave.
- ▶ Relationship for human-caused fires.

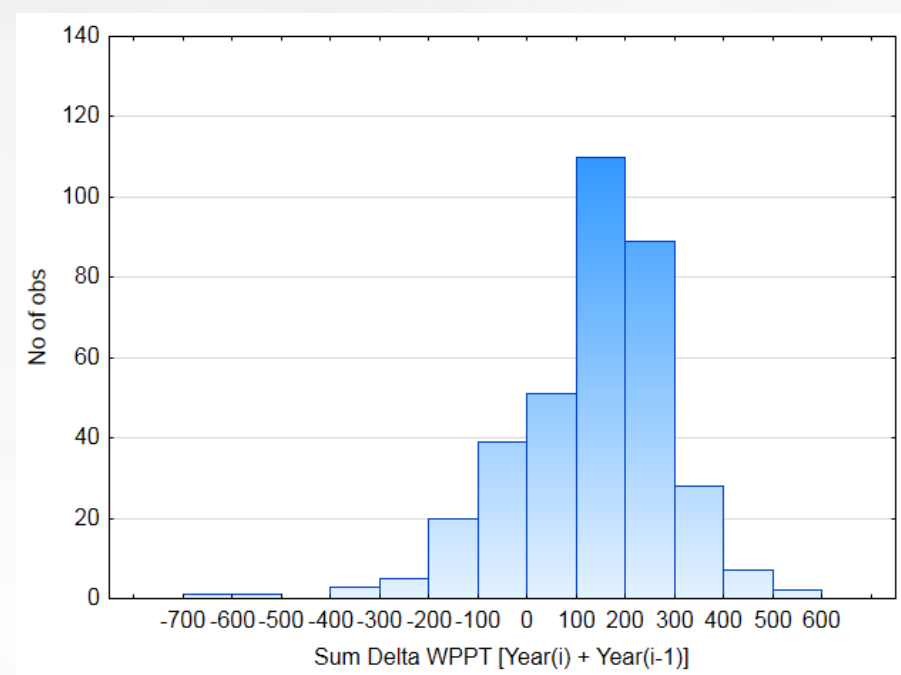
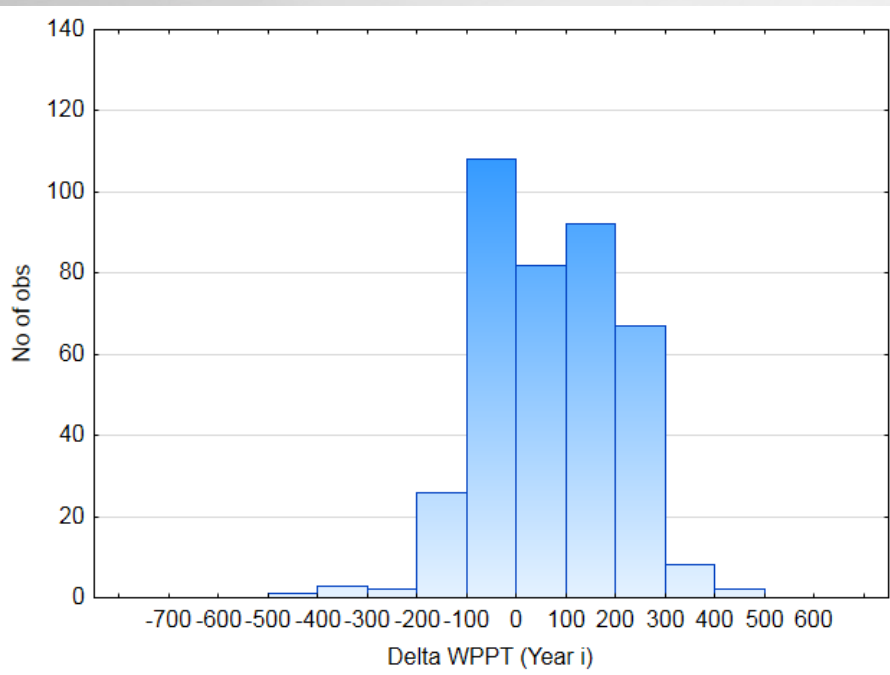


Mean Δ NDVI for Burned Areas 2000-2010



Δ NDVI = peak NDVI in current year (i) – median of peak NDVI₂₀₀₀₋₂₀₁₀

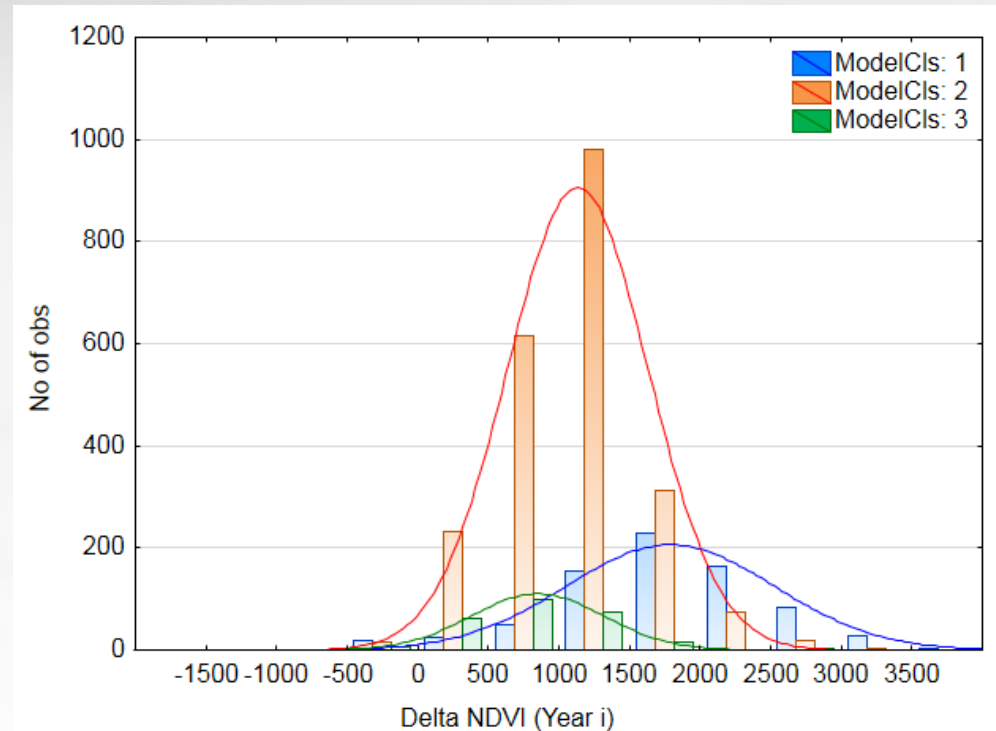
Mean Δ WPPT for Burned Areas (2000 – 2010)



Mean Δ WPPT = Winter precipitation for current year –
median winter precipitation₁₉₇₁₋₂₀₁₀

Subsampling by Vegetation Type within Burns

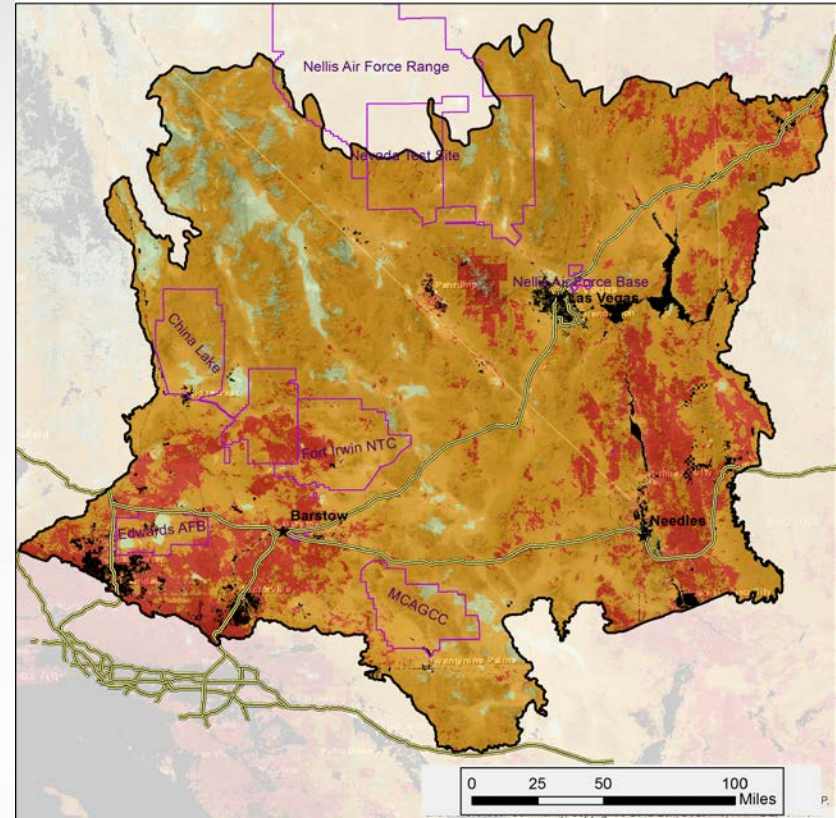
- ▶ Class 1 –low elevation
 - Creosote, Salt Desert Shrub and Mojave Scrub Shrub
- ▶ Class 2—mid elevation
 - Blackbrush, Sagebrush
- ▶ Class 3—higher elevation
 - Juniper, Pinyon Pine, Chapparal



65% of fires occurred in Blackbrush and Creosote vegetation types for the entire fire record

Preseason Fire Hazard

- ▶ Use precipitation zones as strata to divide region into sampling areas
- ▶ Using 10-year record of data:
 - Fire history (Burned/Unburned)
 - Δ NDVI
 - Δ WPPT
 - Ignition variables
 - Vegetation Type
- ▶ Apply logistic regression to derive parameter estimates to predict probability of fire hazard in that season



Preseason Fire Hazard: Logistic Regression Model

▶ Sampling Strategy

■ Stratified by

- Precipitation Zone
- Vegetation Type
- Year

- Sampled 1,000 fires and 1,000 non-fires from each Precipitation Zone – Vegetation Type combination

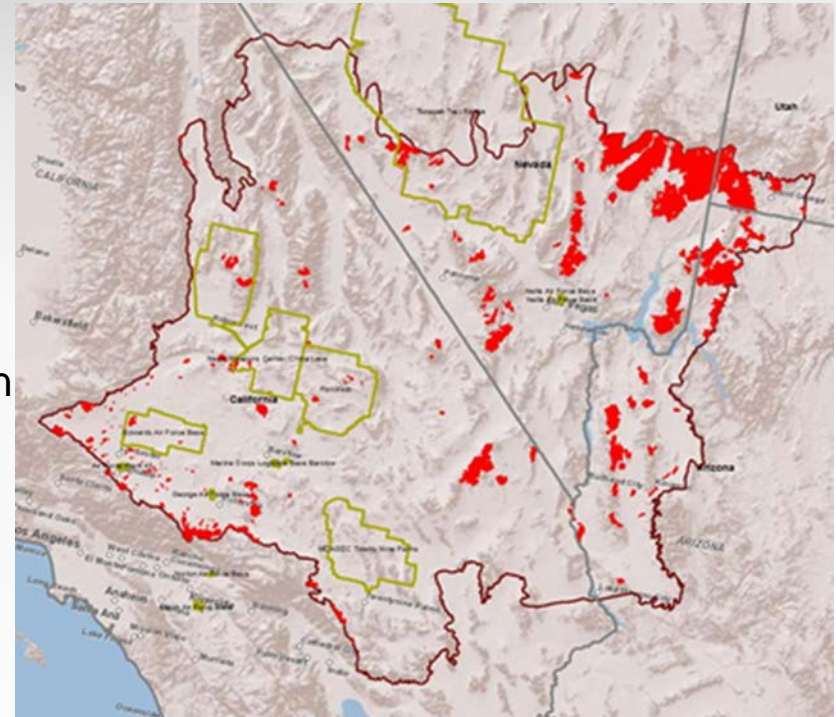
▶ Final Model

■ Logistic Regression

■ Variables included:

- Distance to Road
- Lightning Density
- Delta winter precipitation
- Delta NDVI
- Dwppt²
- Lightning Density²

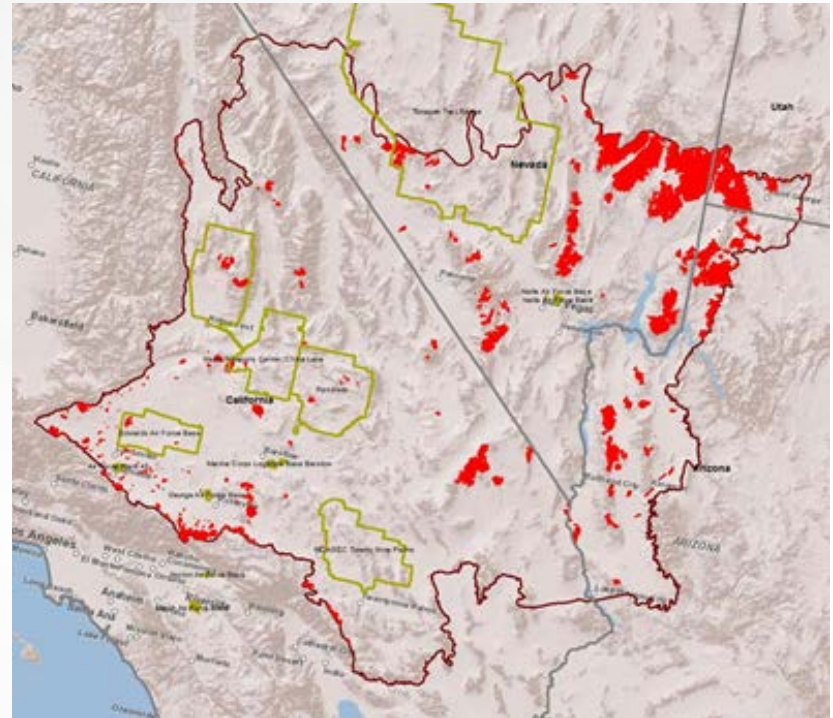
- Parameters estimated using 20X cross-validation.



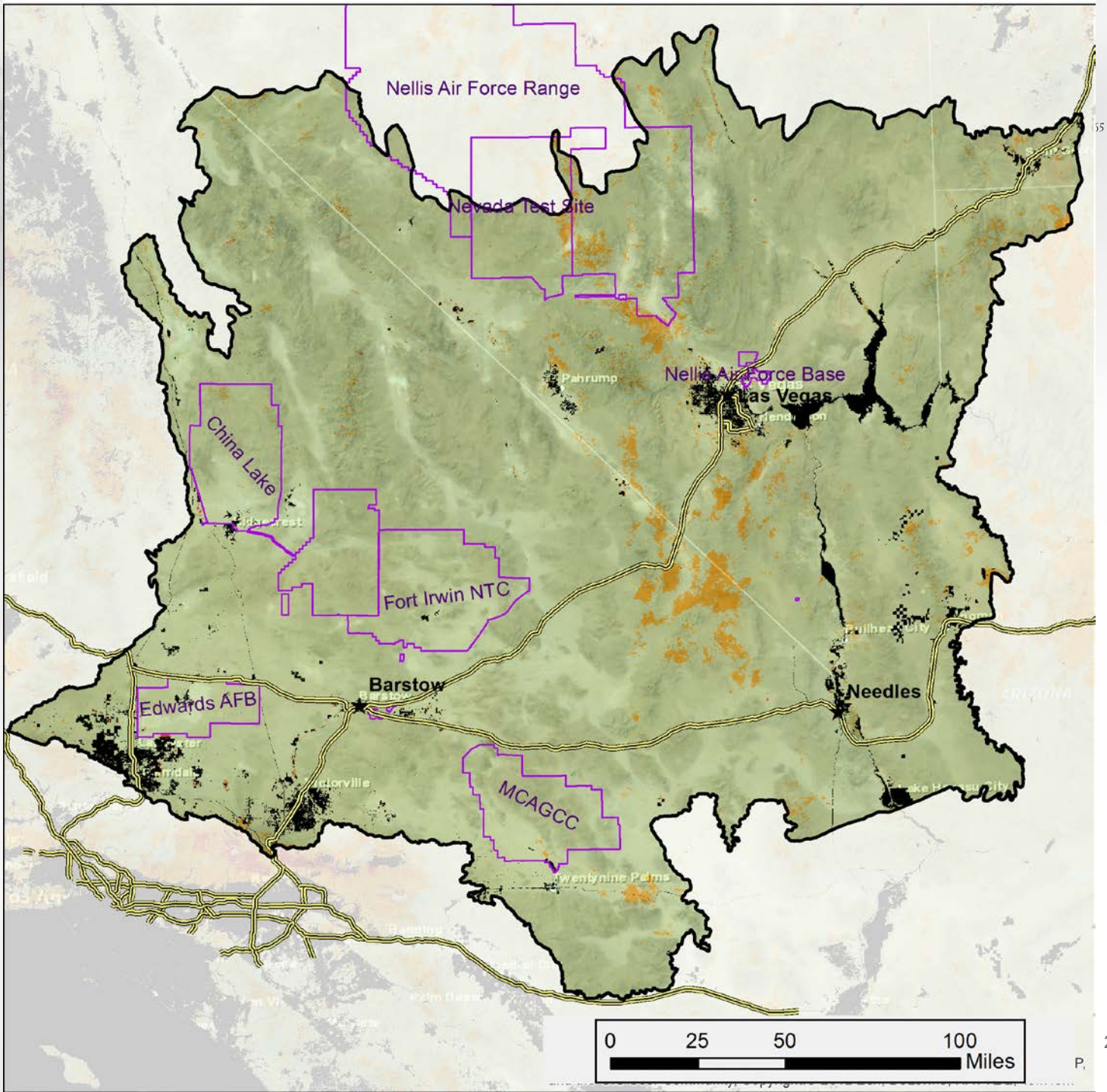


Preseason Fire Hazard: Logistic Regression Model

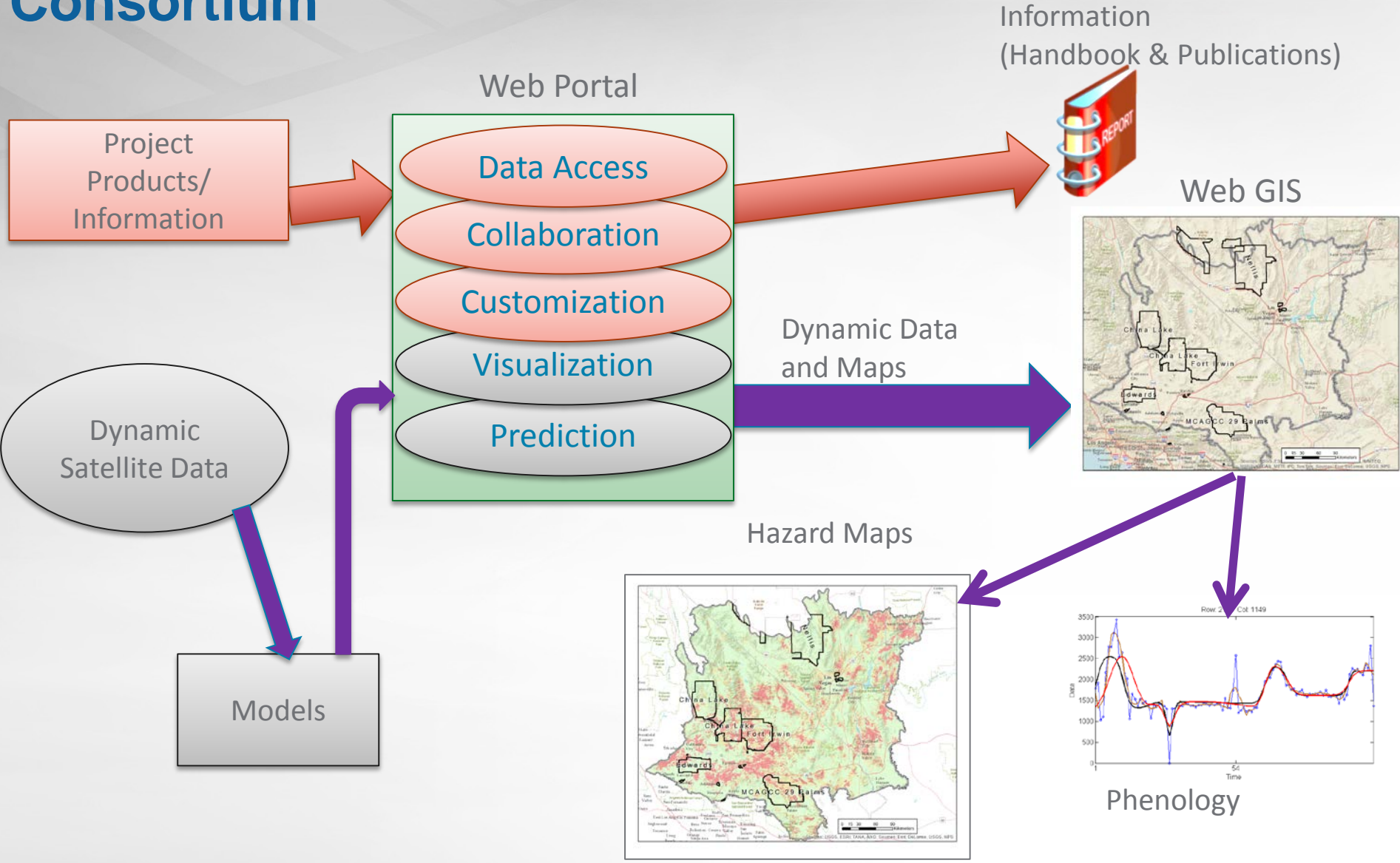
- ▶ Model generates predicted probability of fire for each pixel (for each year or future year)
- ▶ More complicated models considered with little to no improvement in classification accuracy
- ▶ Classification Accuracy
 - Accuracy = 77.8%
 - False Positives = 6.3%
 - False Negatives = 15.8%



2001
2002
2003
2004
2005
2006
2007
2008
2009
2010
2011
2012
2013



Mojave Fire Management Portal linked through California Fire Science Consortium





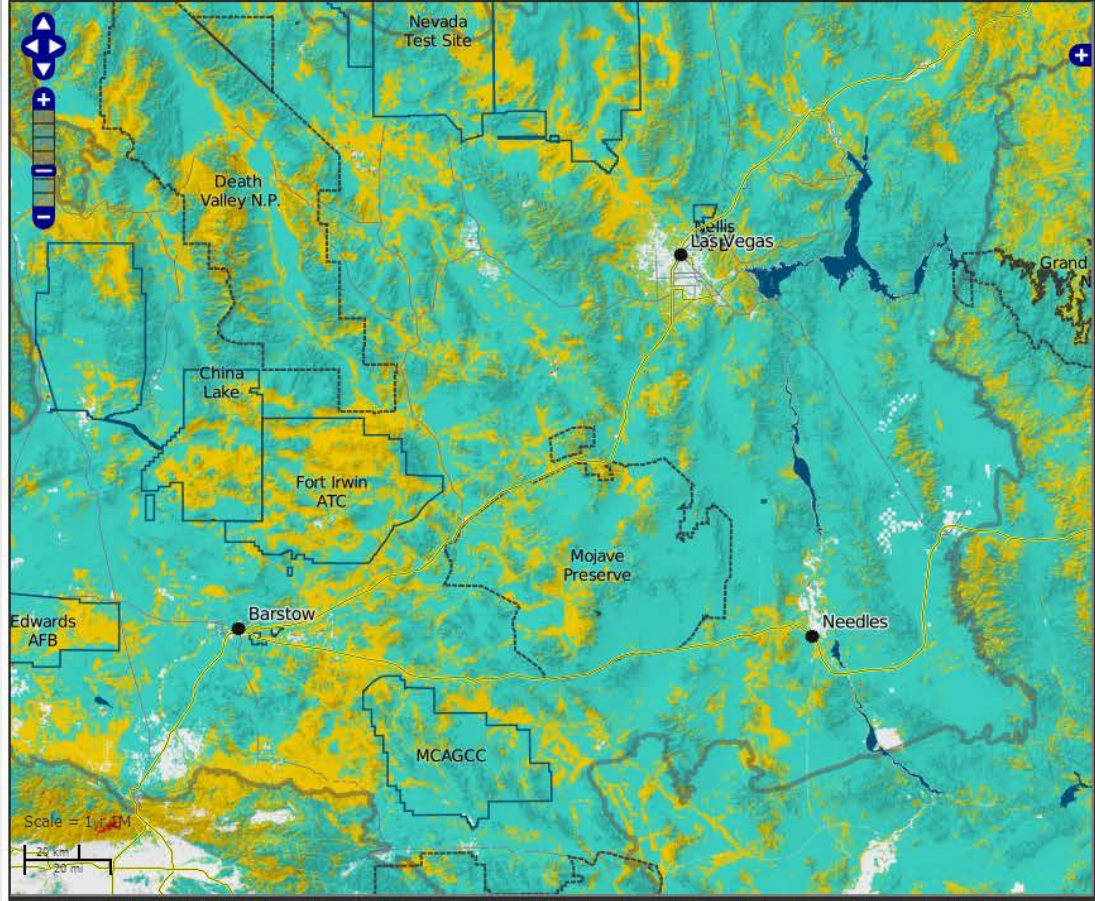
<http://gisx.pnl.gov/Mojave>

404 Not Found | PNNL WEB PORTAL | LANDFIRE-Vegetation Pro | profound effect - Google | Mojave Fire Risk

gisx.pnl.gov/Mojave/

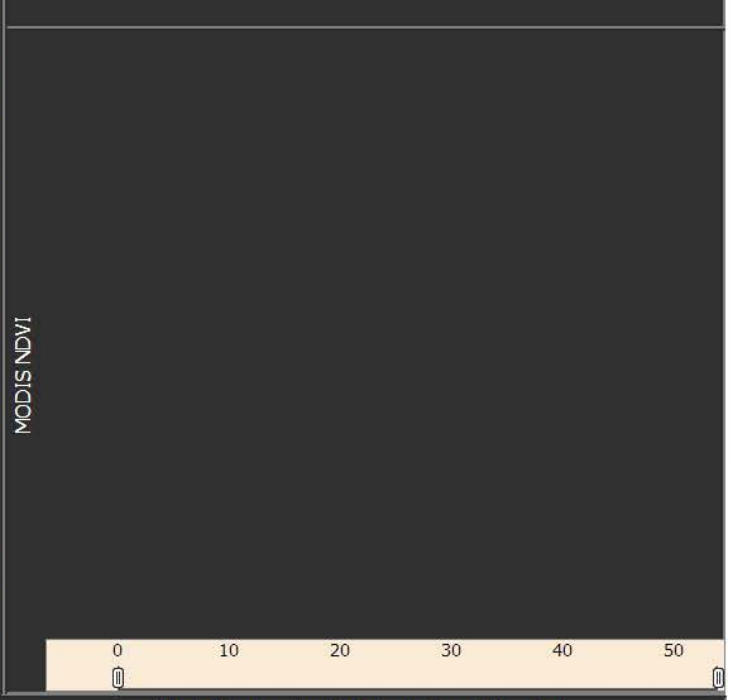
Predictive Tools for Managing Altered Fire Regimes

Caused by plant invasions in the Mojave Desert



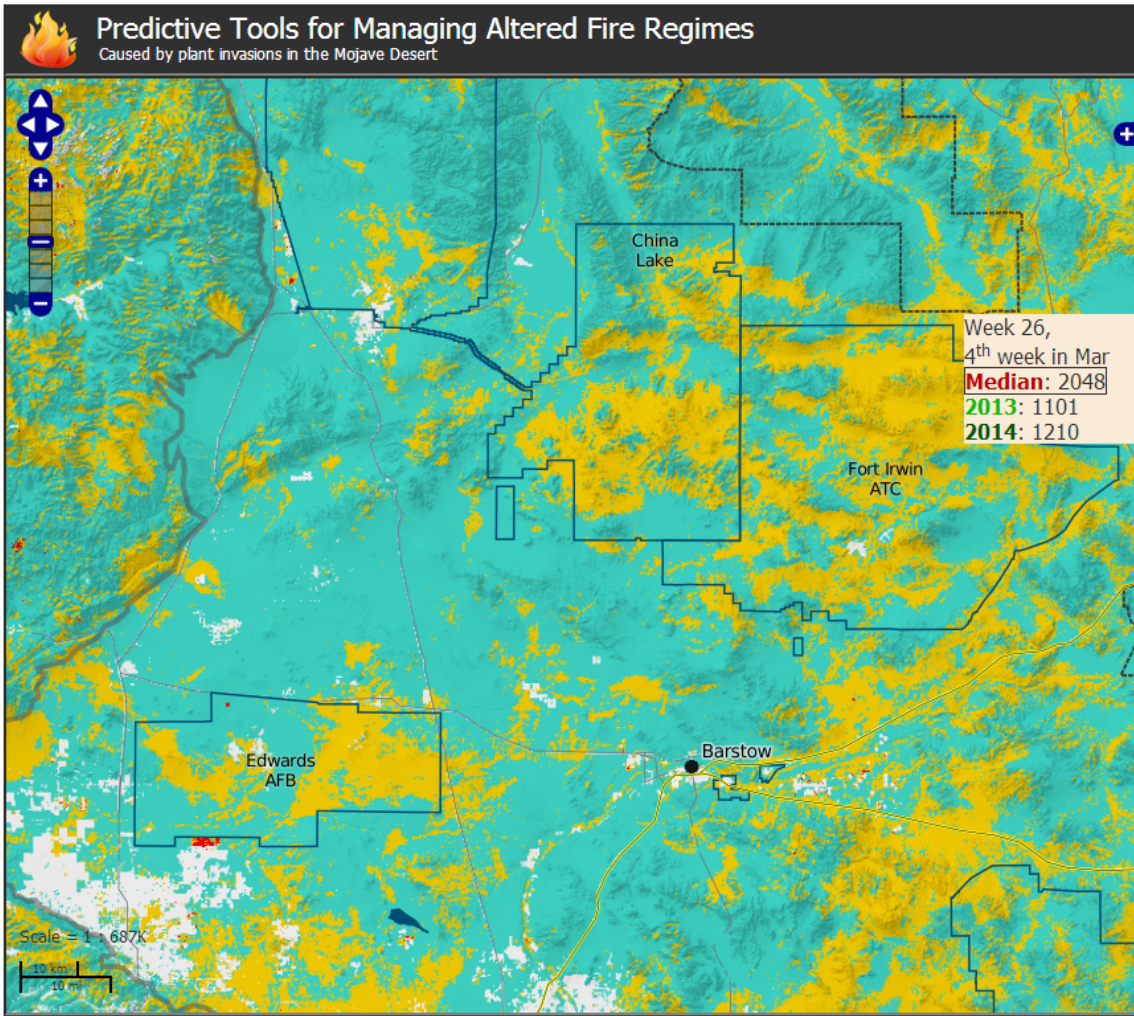
Easting (335095 - 831594)
Northing (3725726 - 4222225)

Bolded Lines are Displayed



Projection (WGS84 UTM 11n) x: 646998.94 y: 3799207.38 (meters) Model Index {row: 1712, col: 1445} Week Number for Hydrologic Year Oct - Sept

Windows taskbar with icons for Internet Explorer, File Explorer, Task Scheduler, Google Chrome, LibreOffice, and other applications. System tray shows date and time: 6:48 AM 6/11/2014.



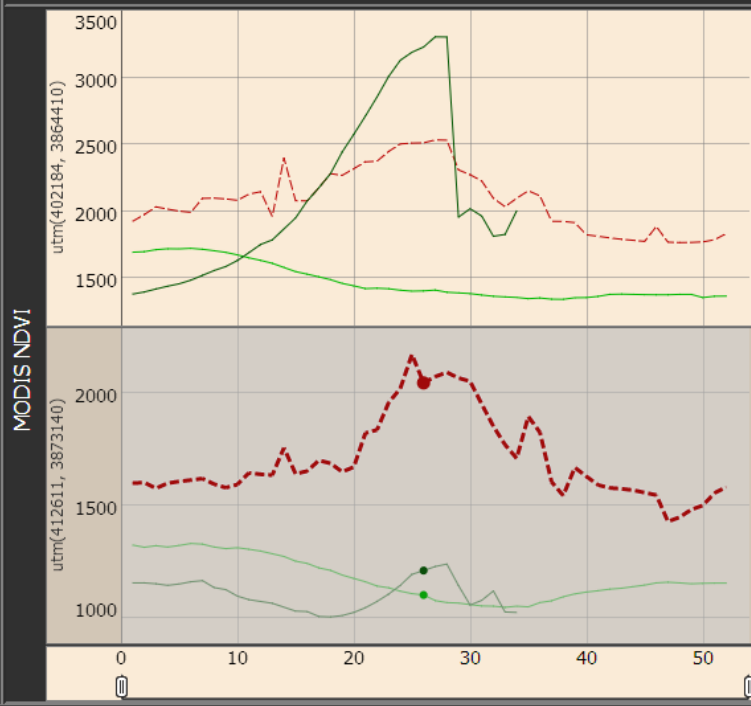
Week 26,
4th week in Mar
Median: 2048
2013: 1101
2014: 1210

Easting (335095 - 831594)

Northing (3725726 - 4222225)

utm(412611, 3873140)

Bolded Lines are Displayed



Projection (WGS84 UTM 11n) x: 411883.82 y: 3859559.56 (meters) Model Index {row: 1470, col: 505} Week Number for Hydrologic Year Oct – Sept