

When monitoring is not enough:

Predicting post-wildfire
trajectories from MODIS
time series



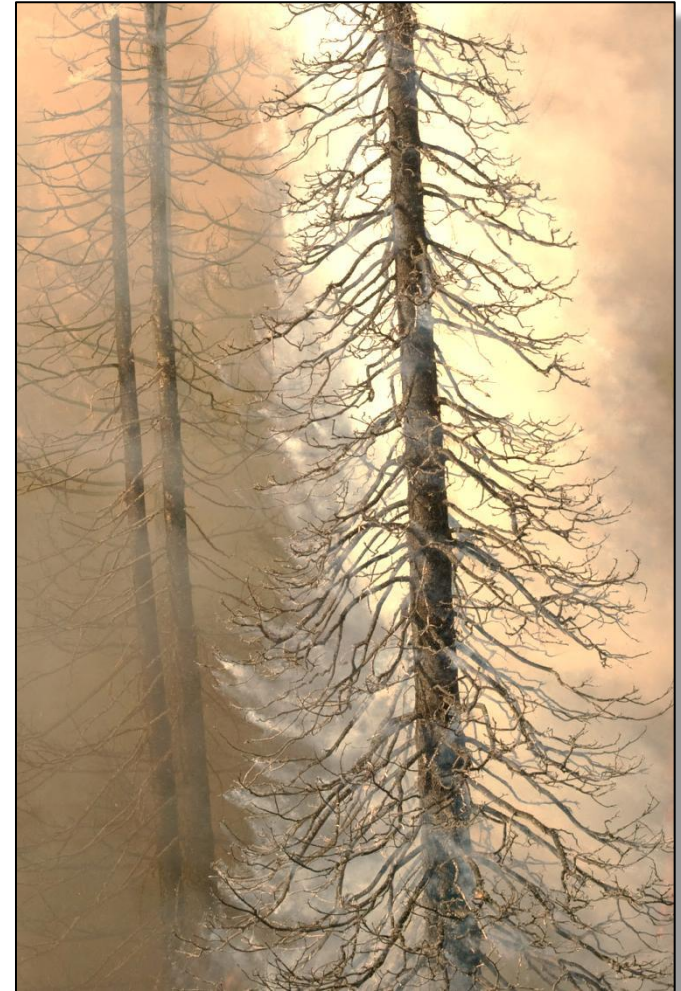
Steven P. Norman¹

William W. Hargrove¹

Danny C. Lee¹

Joseph P. Spruce²

William M. Christie¹



¹. USDA Forest Service Southern Research Station, Asheville NC
Eastern Forest Environmental Threat Assessment Center

². NASA-Stennis, Stennis Space Center MS



Outline

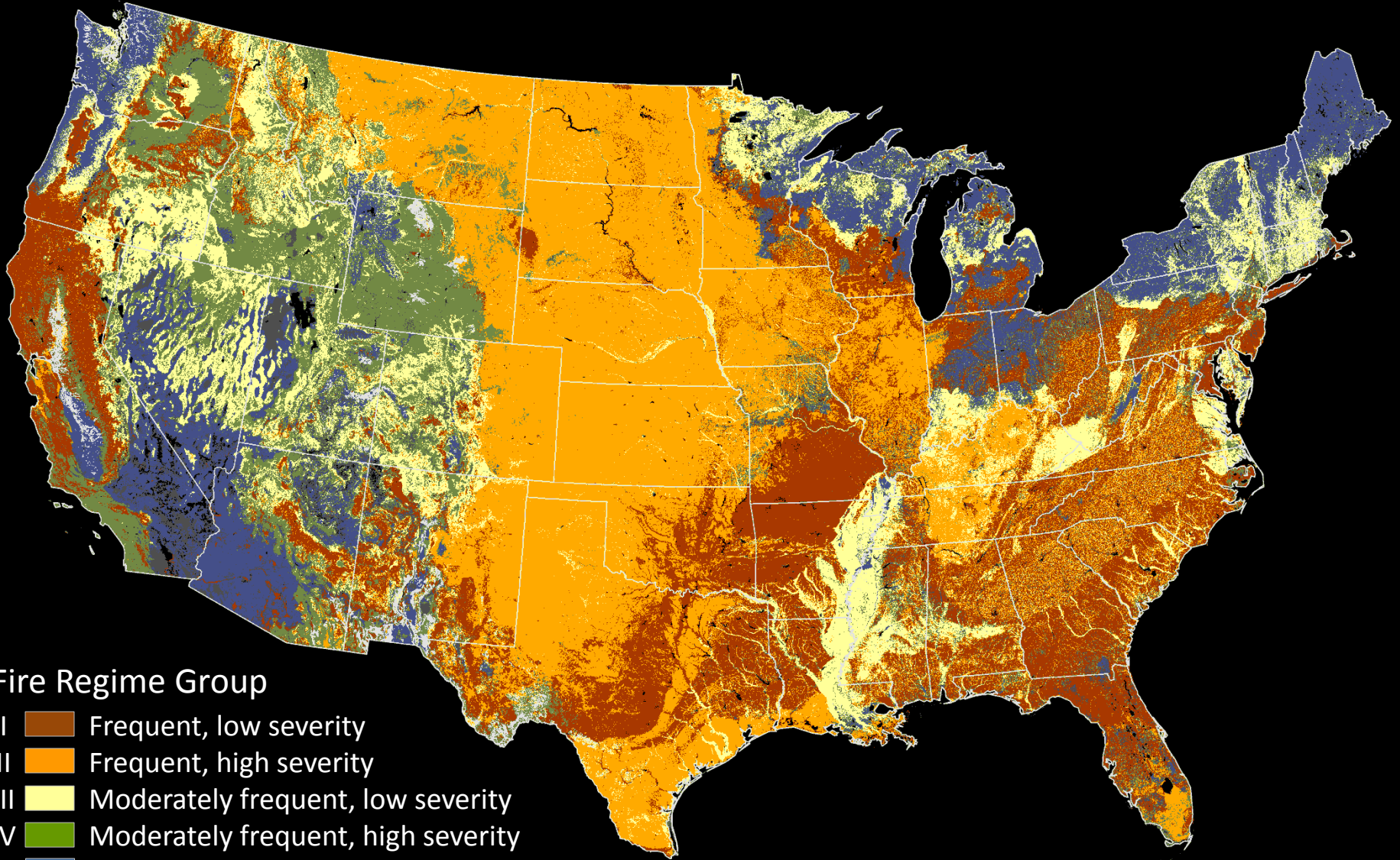
PART I—The wildland fire and management crisis of the Western US

PART II—Theoretical challenges that can impede management success

PART III—Efficient coarse-filter landscape monitoring

PART IV—Predicting long-term wildfire effects

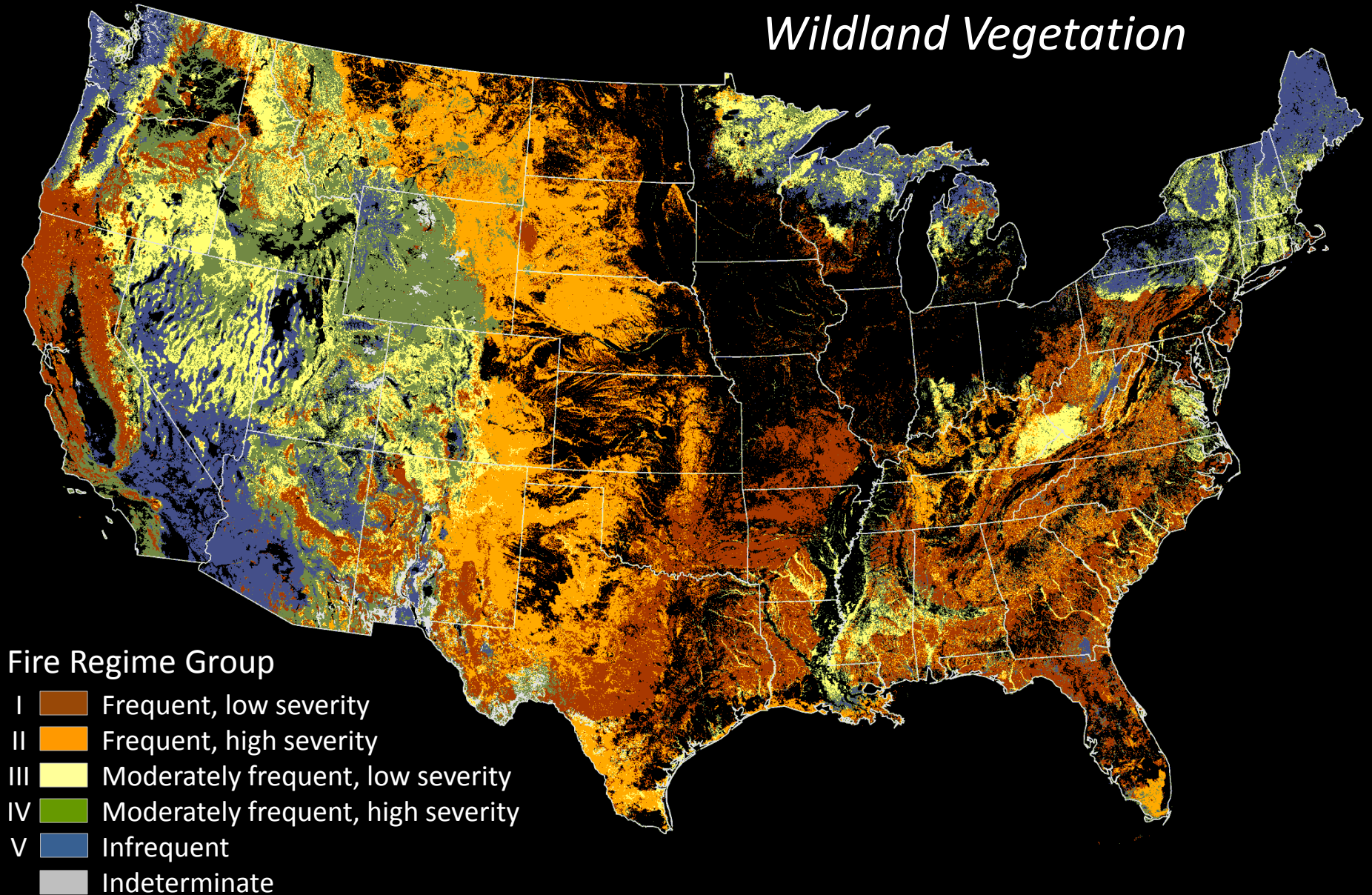
Historical Fire Regimes



Fire Regime Group

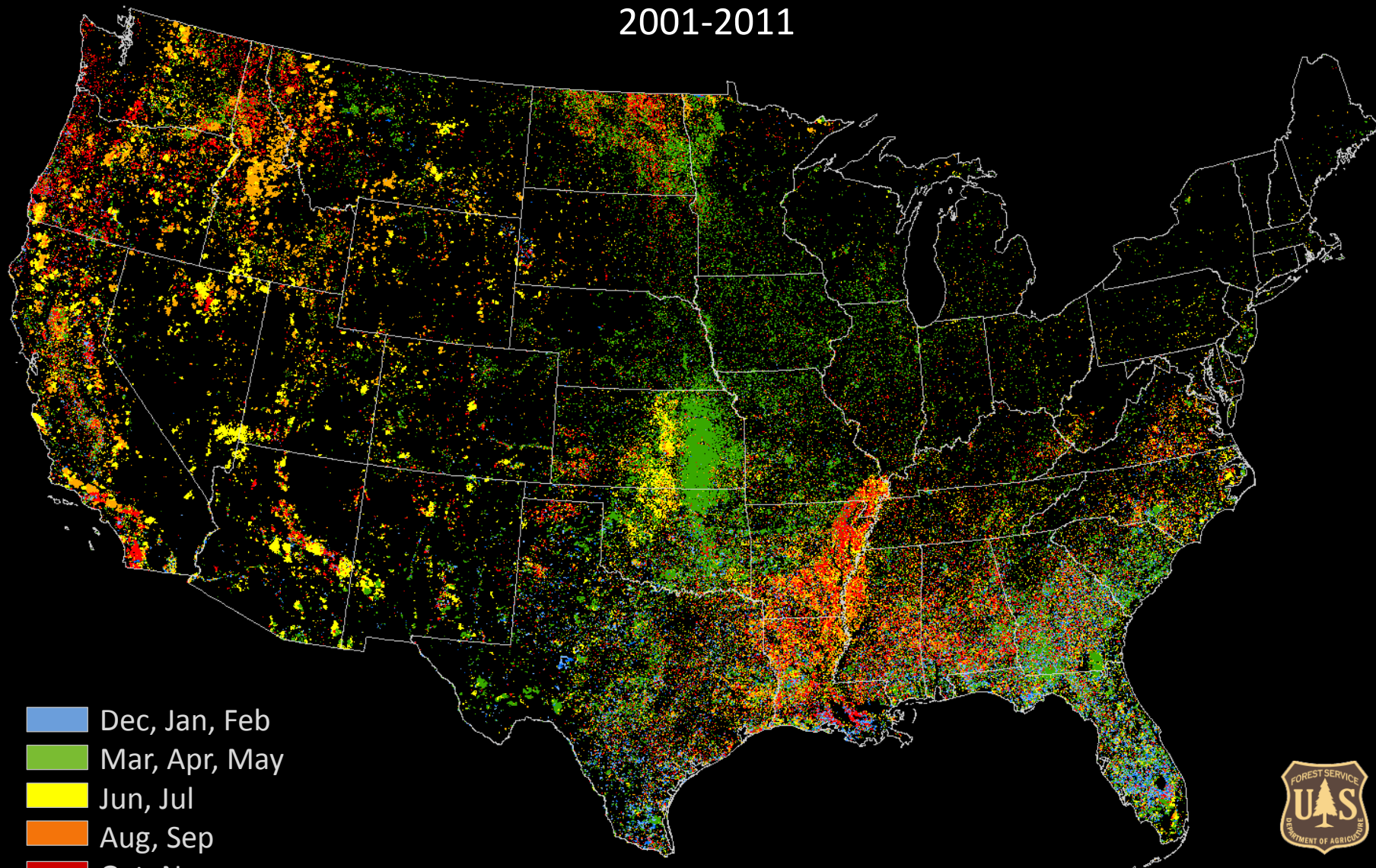
- I  Frequent, low severity
- II  Frequent, high severity
- III  Moderately frequent, low severity
- IV  Moderately frequent, high severity
- V  Infrequent
-  Indeterminate

Historical Fire Regimes of *Existing* Wildland Vegetation



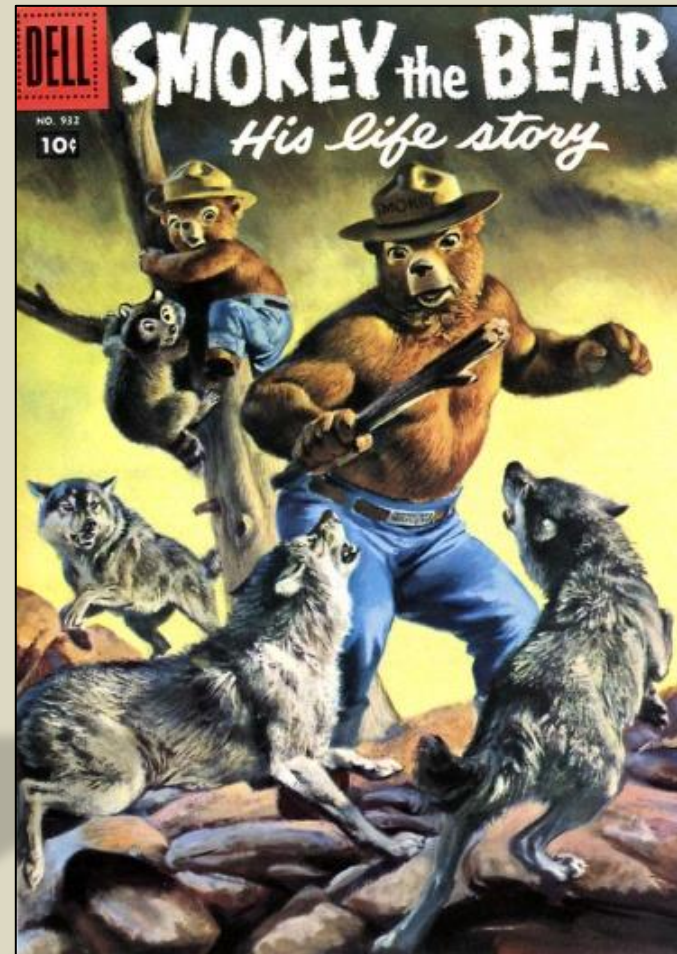
Seasonality of all fires

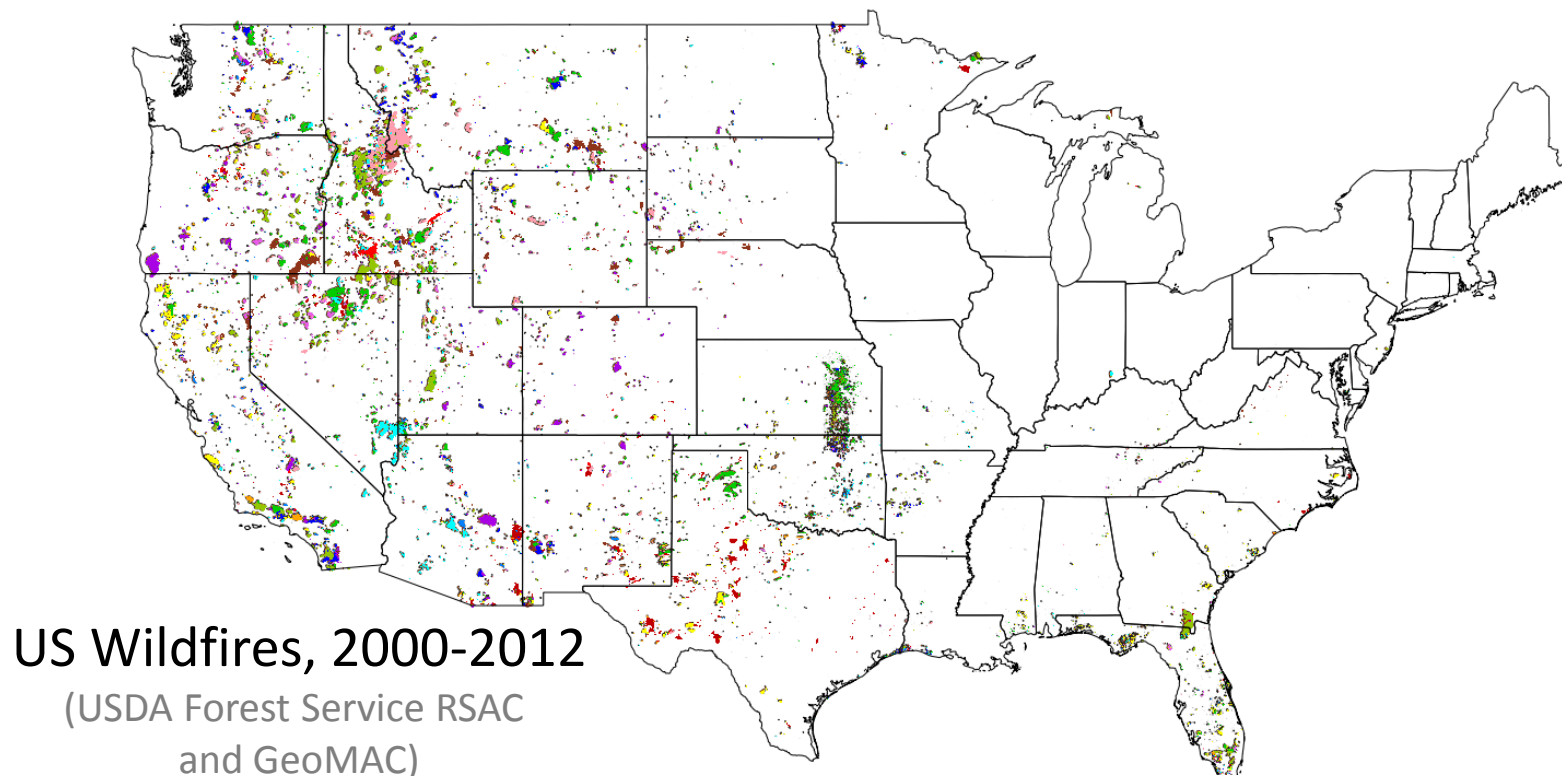
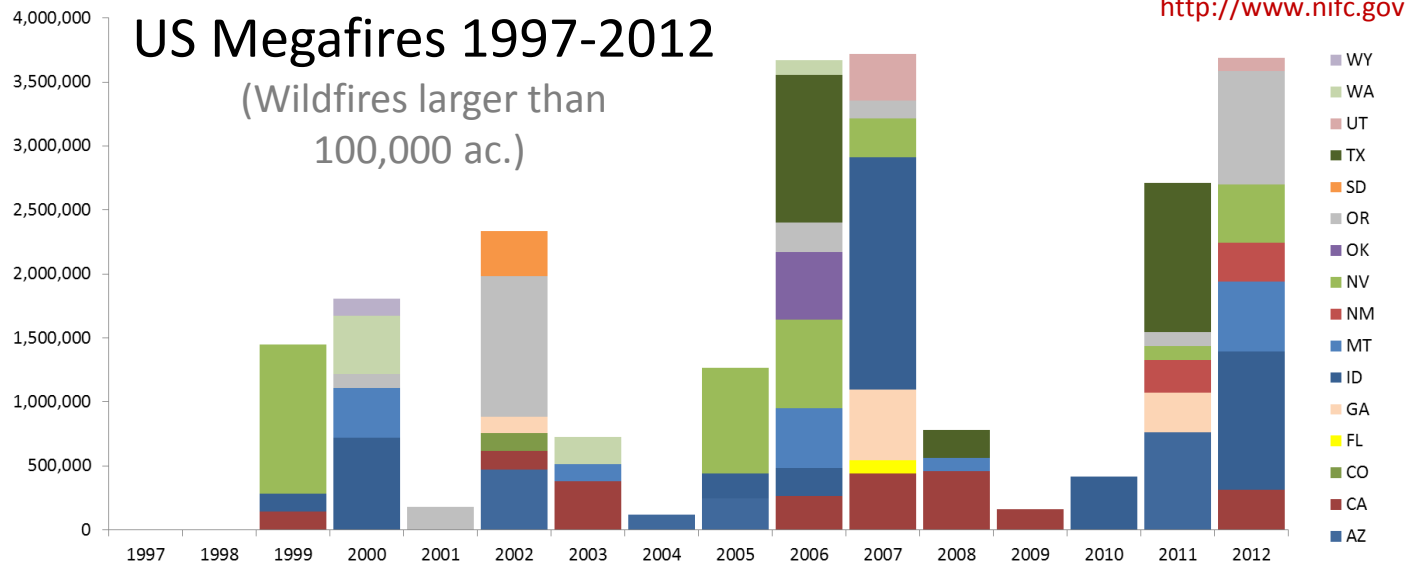
as inferred from MODIS hotspots
2001-2011



N=1,009,782
Steve Norman 5/2012

Does our wildland fire problem result from
our past management decisions?





Little Bear Fire, Lincoln National Forest NM, 2012



Trinity Ridge Fire, Boise National Forest ID, 2012



Waldo Canyon Fire and Colorado Springs CO, 2012







Outline

PART I—The wildland fire and management crisis of the Western US

PART II—Theoretical challenges that can impede management success

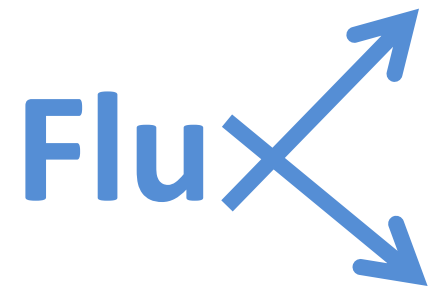
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PART IV—Predicting long-term wildfire effects

PART II—Theoretical challenges that can impede forest management success

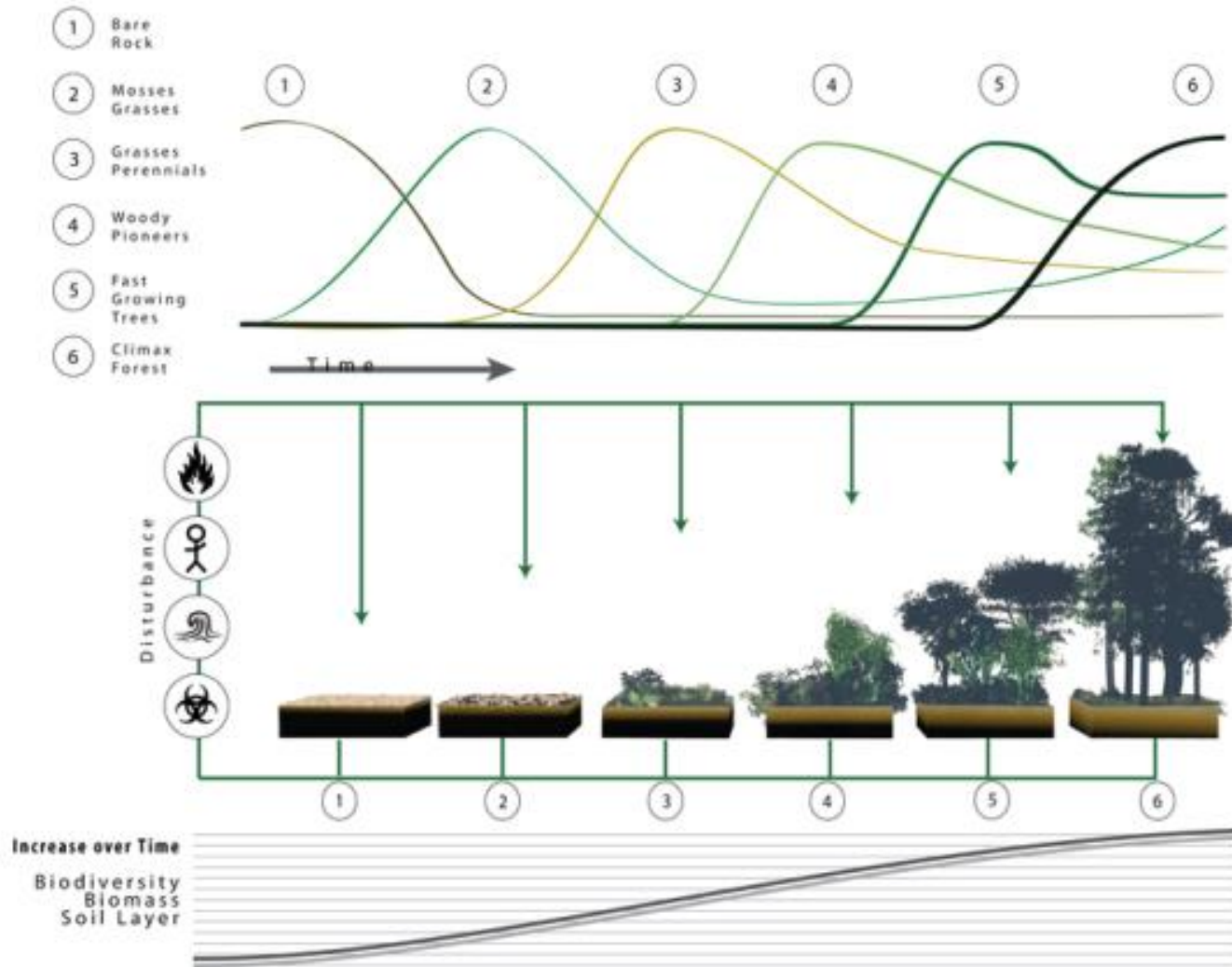


STABILITY



PART II—Theoretical challenges that can impede forest management success

A typical cyclical disturbance-succession model (at the stand-scale)



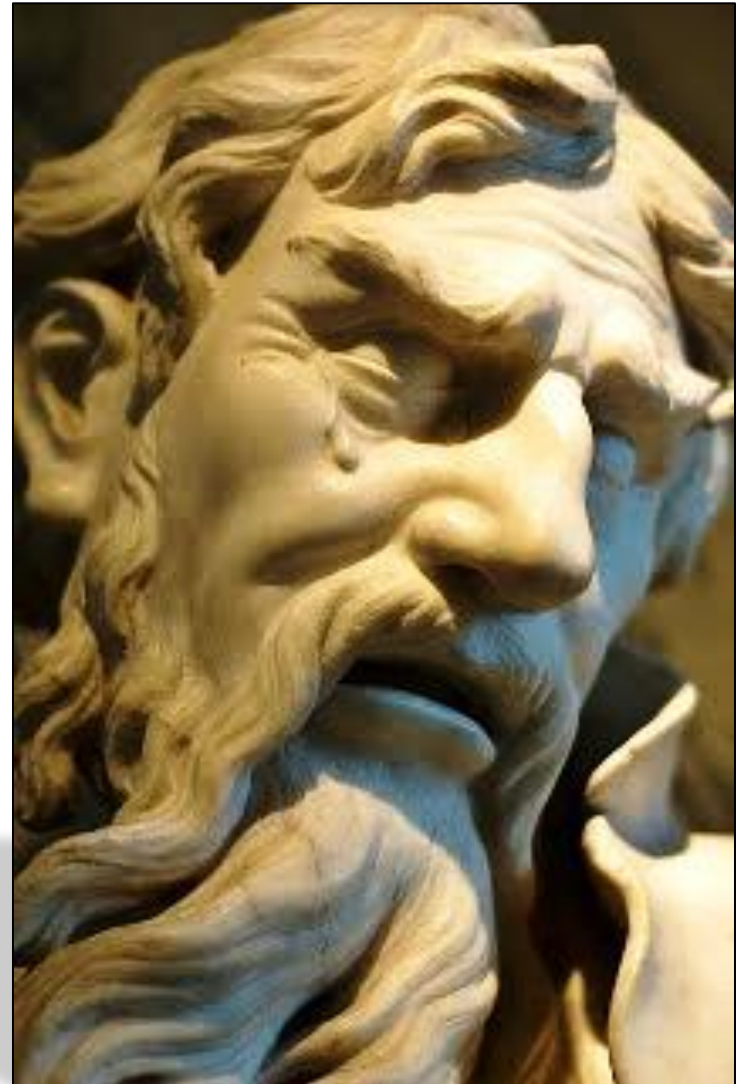
CYCLICITY
*at the stand scale
could translate to*
STABILITY
*at the landscape
scale*

ποταμῶι γαρ οὐτ' ἐστὶν ἐμβῆναι δις τῶι αὐτῶι.
σκίδνησι καὶ πάλιν συνάγει.....καὶ πρόσσεισι καὶ ἀπείσι

*“after all, one does not step
into the same river twice.
waters disperse and come
together again ... they keep
flowing on and flowing away”*

Heraclitus

c. 535 – c. 475 BCE



PART II—Theoretical challenges that can impede forest management success

DRIVERS of STABILITY in the fire environment

- Evolutionary-scale species responses to fire
- Evolutionary-scale competitive relationships among species
- Fuel accumulation and succession
- Consistency of climate (fire seasonality, microclimate)
- Topography
- Consistent management regimes

PART II—Theoretical challenges that can impede forest management success

DRIVERS of STABILITY in the fire environment

- Evolutionary-scale species responses to fire
- Evolutionary-scale competitive relationships among species
- Fuel accumulation and succession
- Consistency of climate (fire seasonality, microclimate)
- Topography
- Consistent management regimes

DRIVERS of FLUX in the fire environment

- Inconstant human influence:
 - changing ignition rates or patterns
 - fire management (e.g., fire suppression)
 - land management (e.g., grazing, logging)
- Introductions of non-native species
- Climate change
- Complex disturbance interactions



“Only I can prevent forest fires? What's fire?”

Buffelgrass



Cheatgrass



Cogongrass



Tamarix spp.



**Complex disturbance interactions (drainage, logging, hurricanes, repeated fires)
in the Okefenokee NWR, Georgia**



**Complex disturbance interactions (drainage, logging, hurricanes, repeated fires)
in the Great Dismal Swamp NWR, Virginia**



Complex disturbance interactions (beetle kill, drought, wildfire) in the interior West



Complex disturbance interactions (blowdown, wildfires) in the 2012 Wenatchee Fire, Washington



PART II—Theoretical challenges that can impede forest management success



STABILITY

Flux





Outline

PART I—The wildland fire and management crisis of the Western US

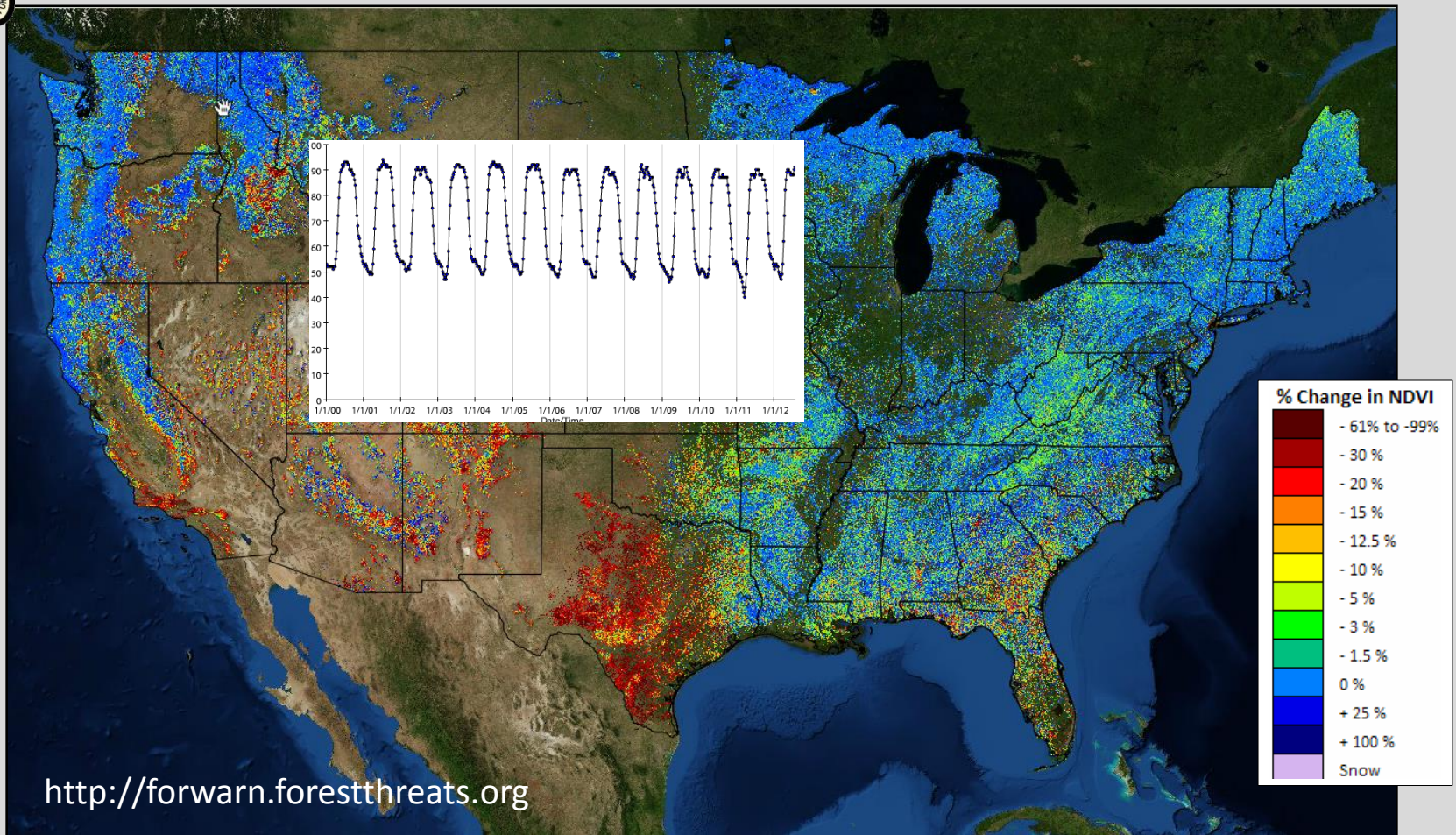
PART II—Theoretical challenges that can impede management success

PART III—Efficient coarse-filter landscape monitoring

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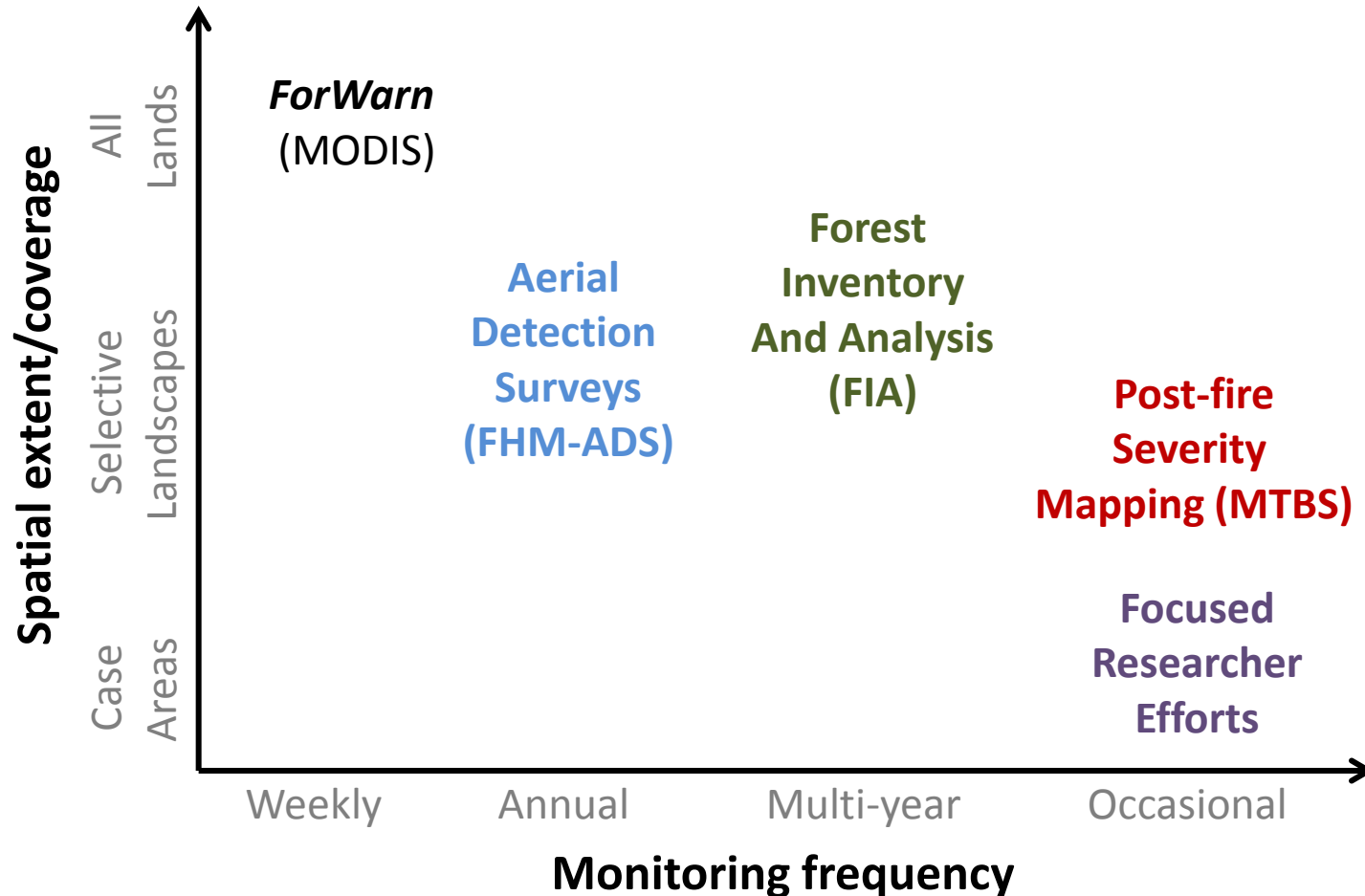


- Normalized Difference Vegetation Index (NDVI) from MODIS
- 46 periods per year (8-day intervals)
- 2000 to present
- 232 meter resolution
- Includes NDVI time series and change maps



PART III—Efficient coarse-filter landscape monitoring

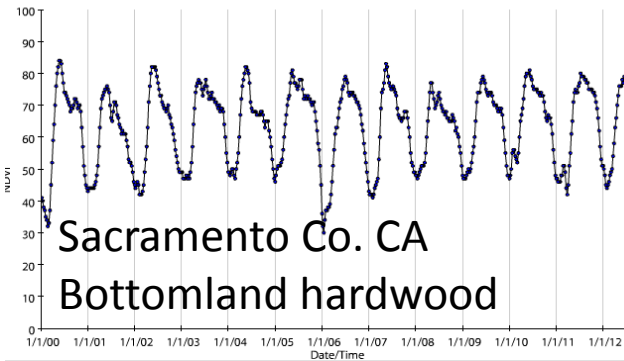
Available approaches and datasets



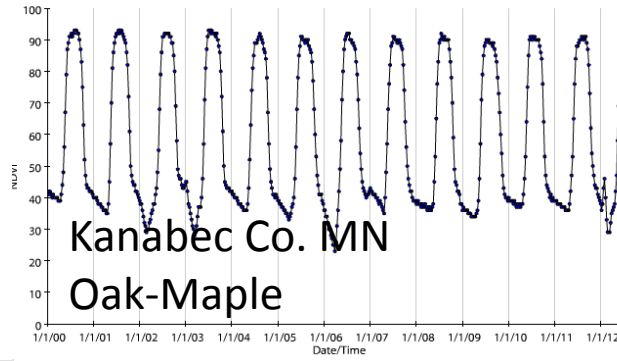
PART III—Efficient coarse-filter landscape monitoring

Phenological signatures of **deciduous forest** dominated pixels

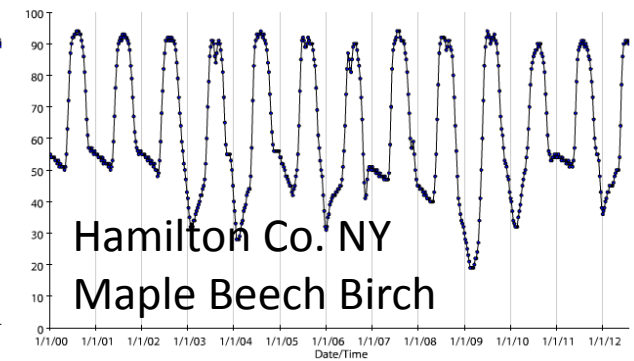
MODIS NDVI for Lat: 38.606167 Lon: -121.495572



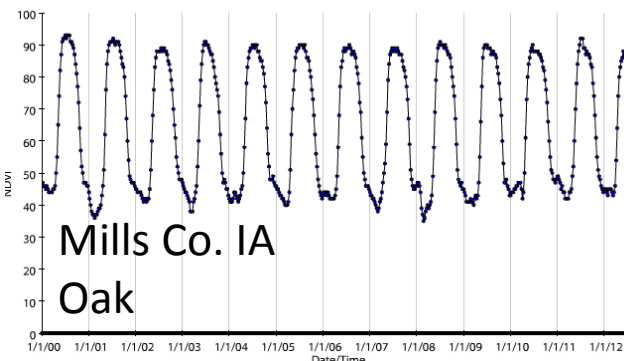
MODIS NDVI for Lat: 46.135957 Lon: -93.170101



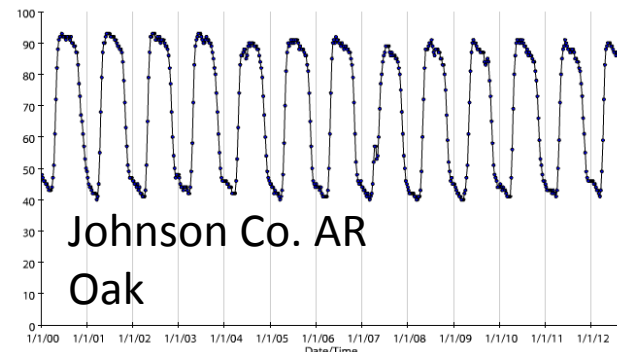
MODIS NDVI for Lat: 43.442462 Lon: -74.575822



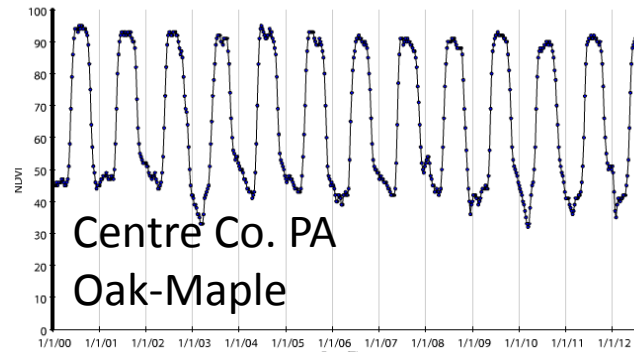
MODIS NDVI for Lat: 40.953757 Lon: -95.776403



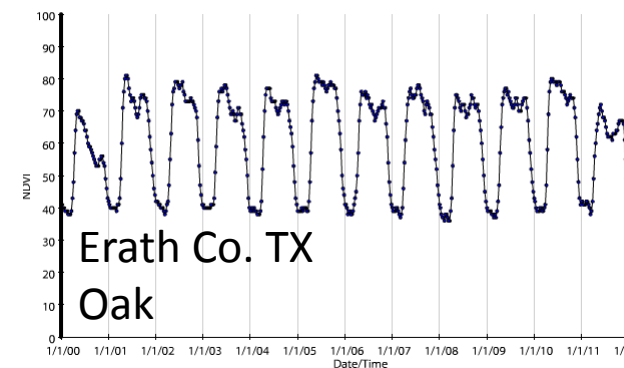
MODIS NDVI for Lat: 35.662042 Lon: -93.238221



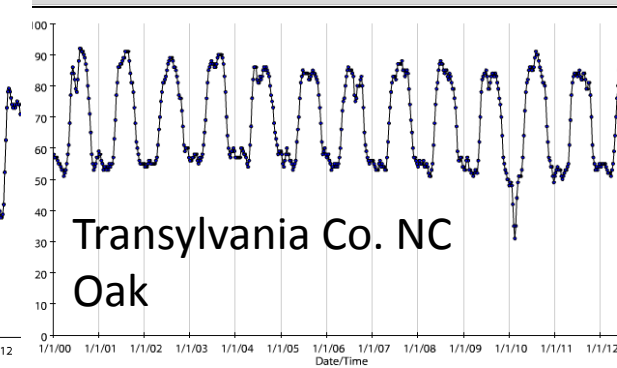
MODIS NDVI for Lat: 40.817019 Lon: -77.788463



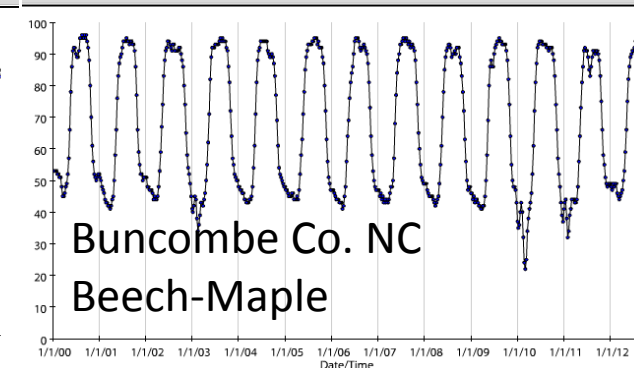
MODIS NDVI for Lat: 32.408265 Lon: -98.411352



MODIS NDVI for Lat: 35.299932 Lon: -82.786252



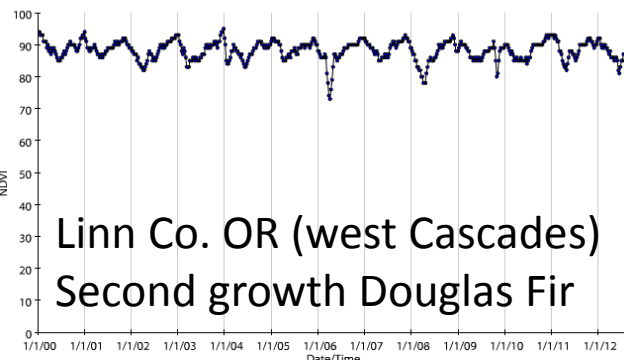
MODIS NDVI for Lat: 35.730570 Lon: -82.423199



PART III—Efficient coarse-filter landscape monitoring

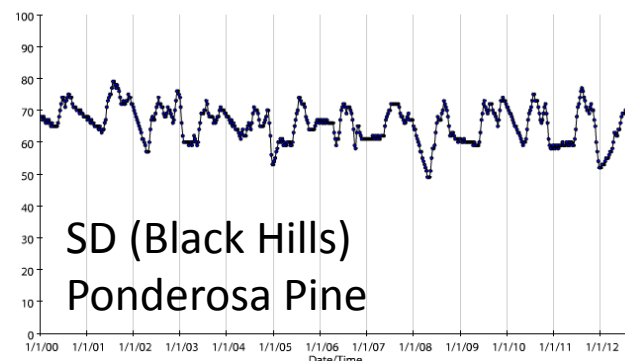
Phenological signatures of **conifer forest** dominated pixels

MODIS NDVI for Lat: 44.559993 Lon: -122.498349



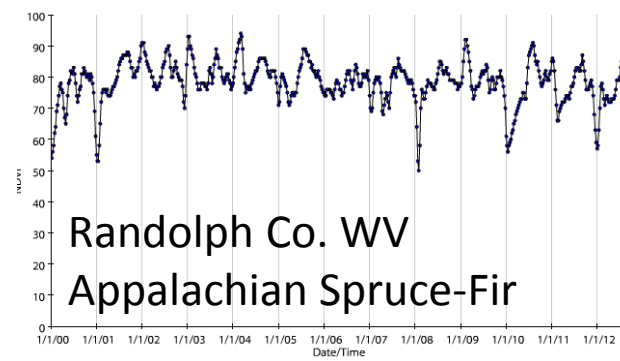
Linn Co. OR (west Cascades)
Second growth Douglas Fir

MODIS NDVI for Lat: 44.076127 Lon: -104.044490



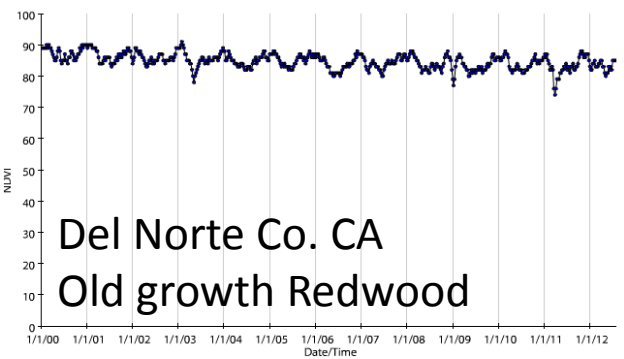
SD (Black Hills)
Ponderosa Pine

MODIS NDVI for Lat: 38.613930 Lon: -79.845828



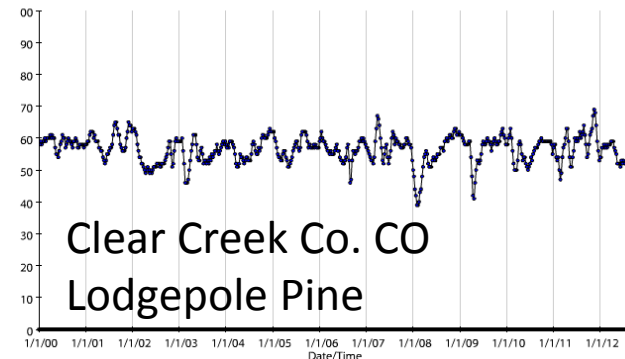
Randolph Co. WV
Appalachian Spruce-Fir

MODIS NDVI for Lat: 41.591584 Lon: -124.061824



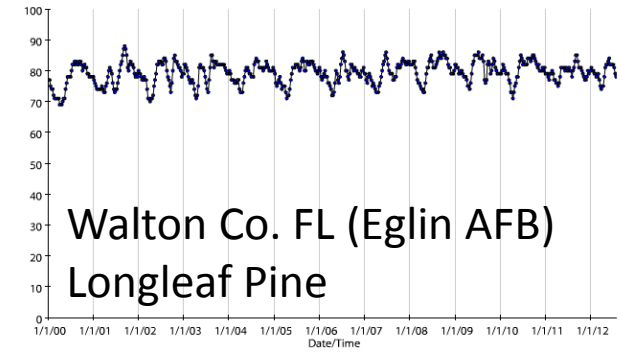
Del Norte Co. CA
Old growth Redwood

MODIS NDVI for Lat: 39.720617 Lon: -105.585688



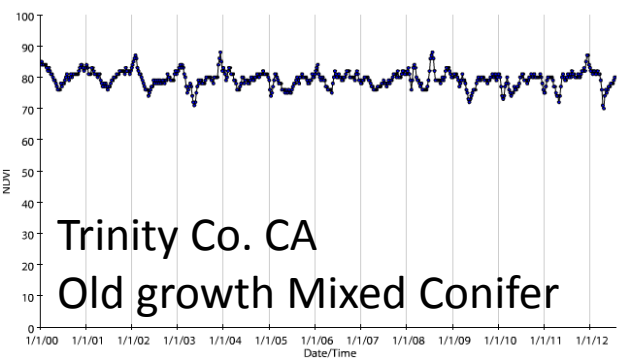
Clear Creek Co. CO
Lodgepole Pine

MODIS NDVI for Lat: 30.533331 Lon: -86.280592



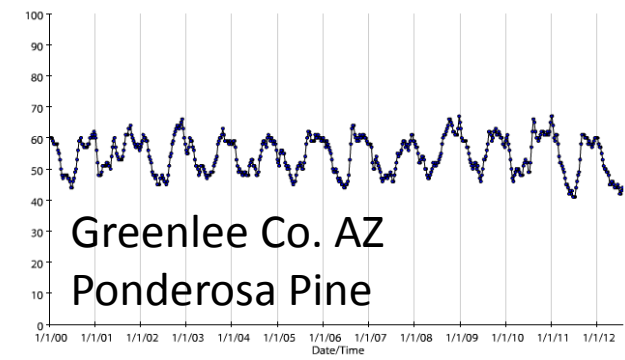
Walton Co. FL (Eglin AFB)
Longleaf Pine

MODIS NDVI for Lat: 40.577527 Lon: -123.268067



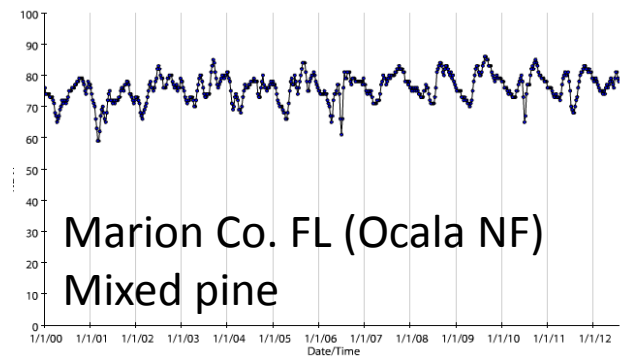
Trinity Co. CA
Old growth Mixed Conifer

MODIS NDVI for Lat: 33.641112 Lon: -109.138410



Greenlee Co. AZ
Ponderosa Pine

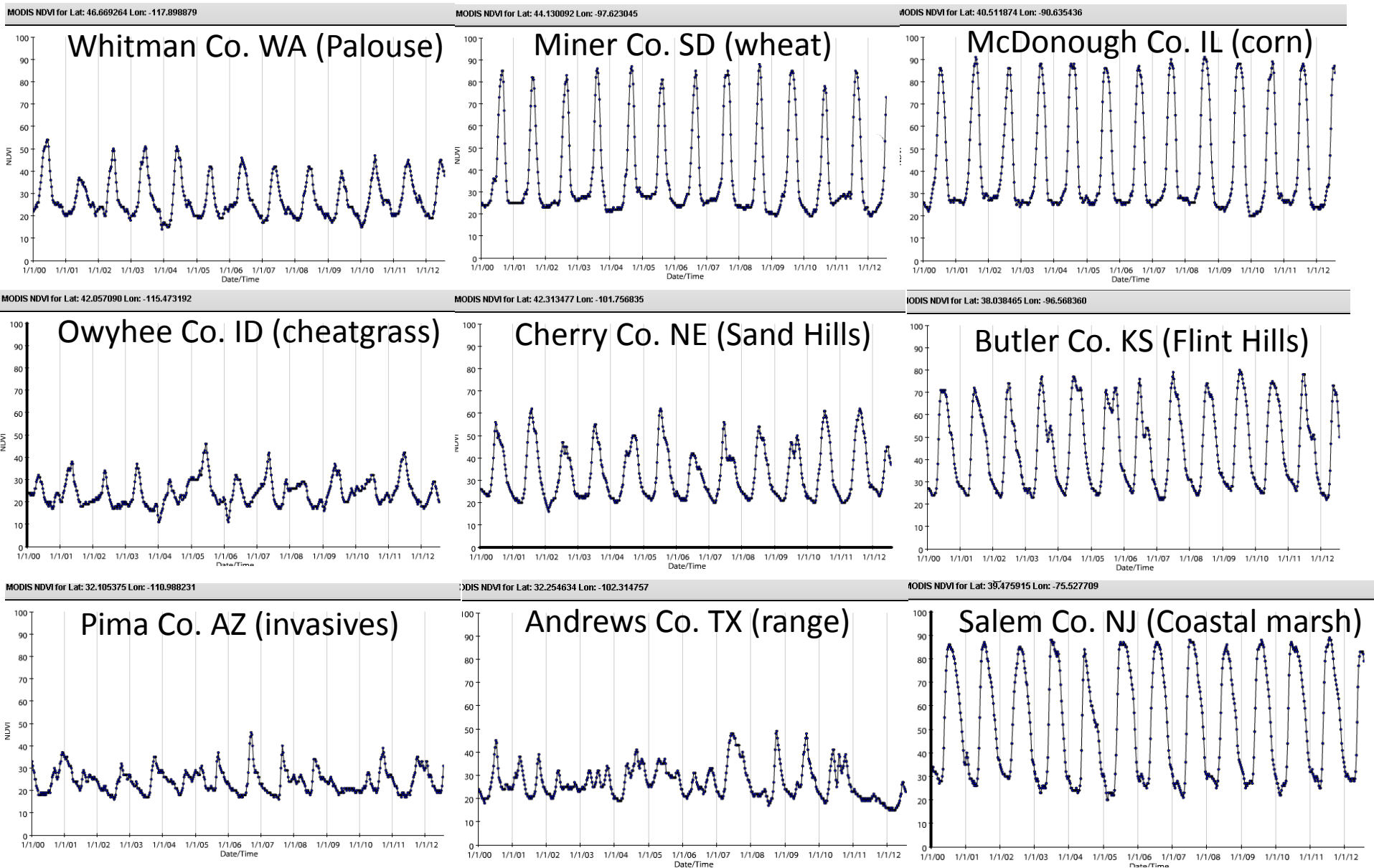
MODIS NDVI for Lat: 29.449665 Lon: -81.862985



Marion Co. FL (Ocala NF)
Mixed pine

PART III—Efficient coarse-filter landscape monitoring

Phenological signatures of **grass** dominated pixels



PART III—Efficient coarse-filter landscape monitoring

Potential measures of fire effects and desired vegetation

Maximum NDVI

Minimum NDVI

Mean NDVI

Median NDVI

Percentiles of the annual distribution

Amplitude of NDVI (of extremes)

NDVI Difference (between thresholds)

Duration above some threshold

Area under the growing season curve

Key Measures for Vegetation Change Associated With Wildland Fire:

(1) LIVING BIOMASS

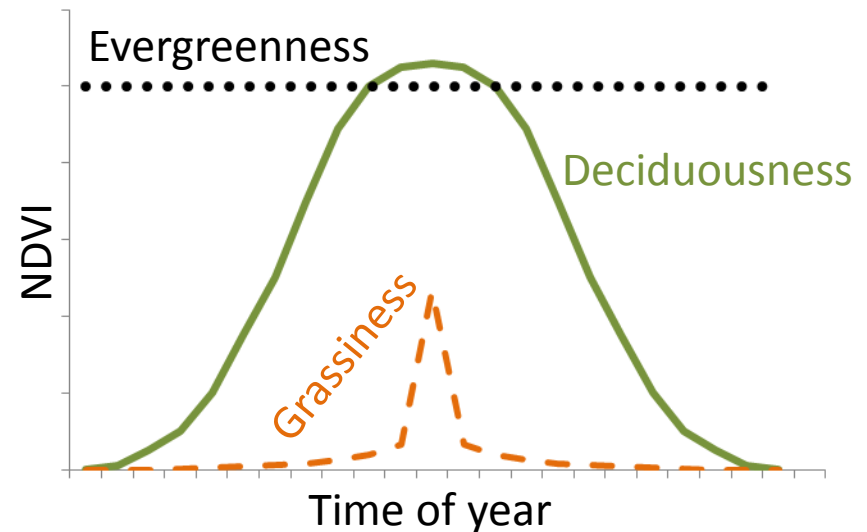
(median NDVI)

(2) CONIFER (EVERGREEN) FRACTION

(~25th %ile of annual distribution)

(3) GRASS FRACTION

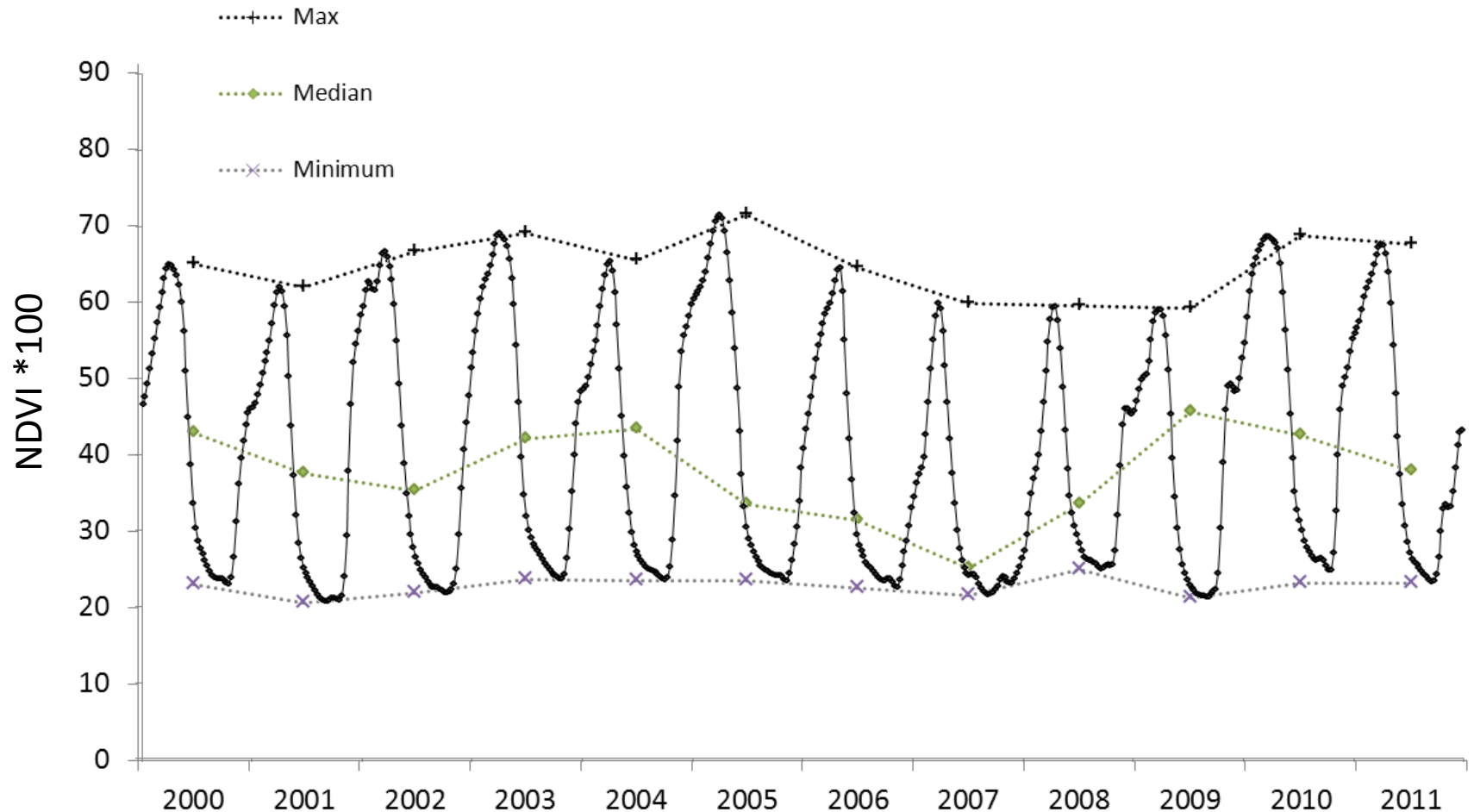
(peakedness of upper distribution)



PART III—Efficient coarse-filter landscape monitoring

The max., median and min. NDVI for Willows CA non-native grasslands

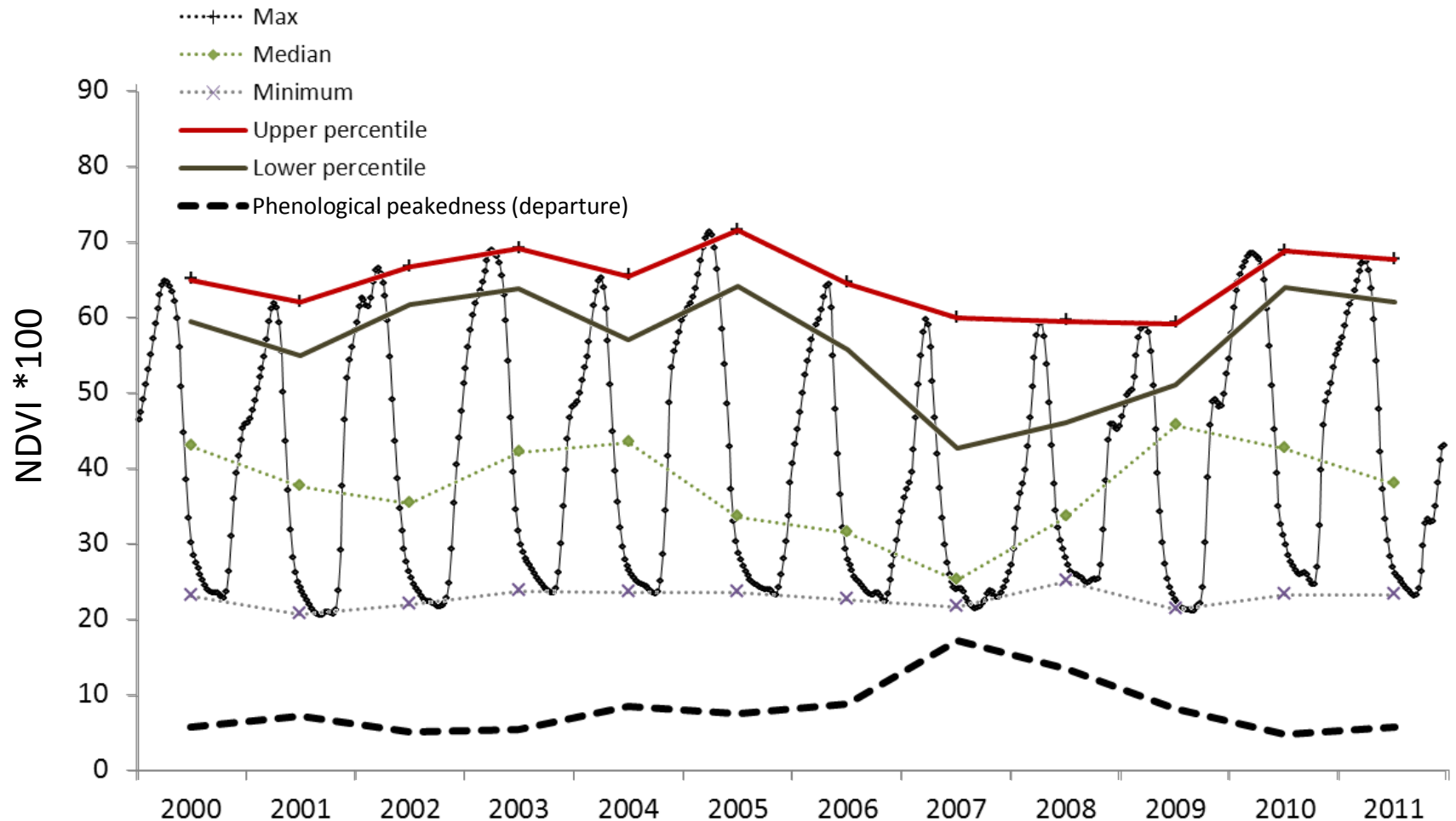
(The landscape mean of 9,345 MODIS pixels)



Note the inter-annual volatility of this measure typically caused by climate variation

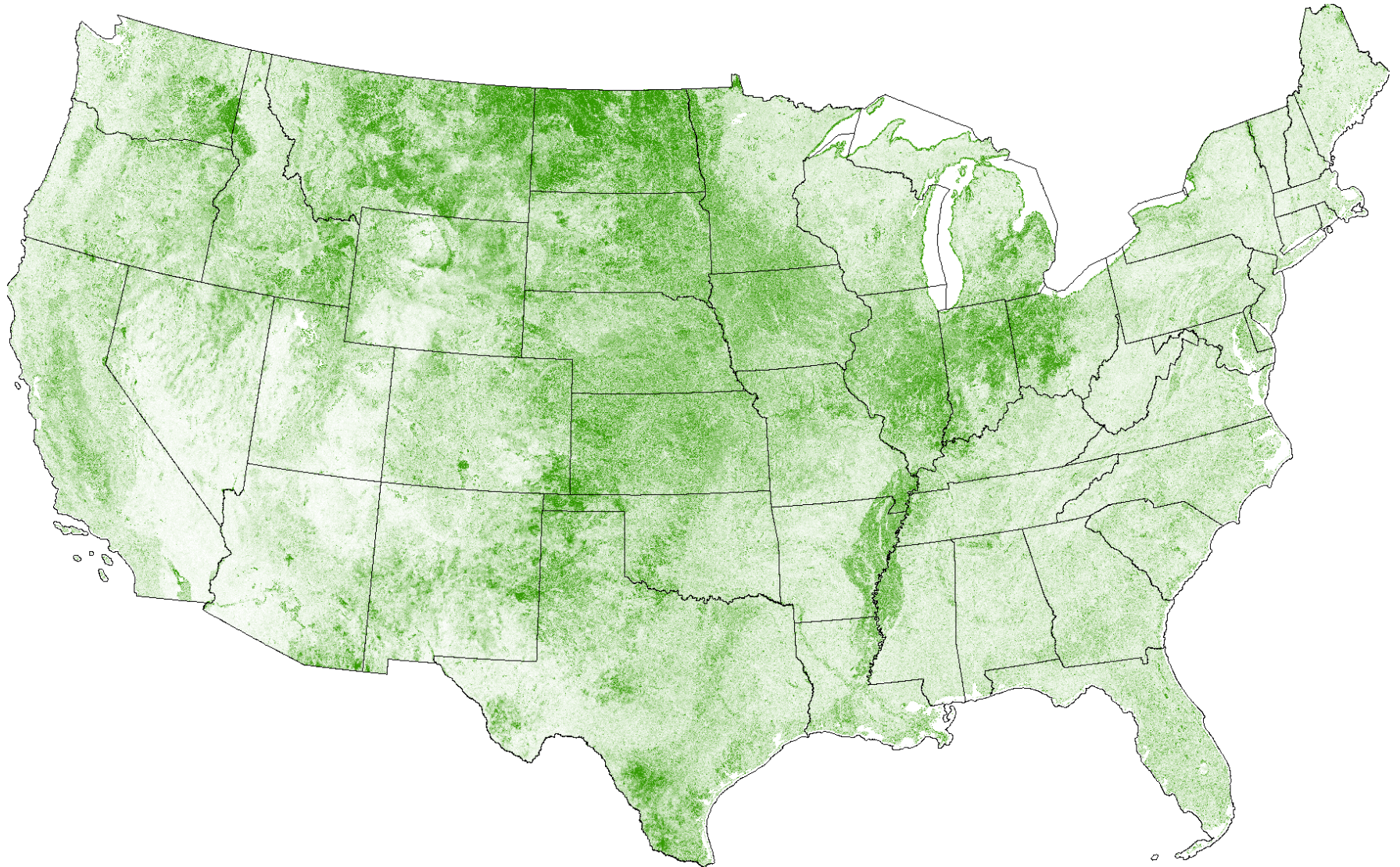
PART III—Efficient coarse-filter landscape monitoring

Phenological peakedness as the difference between the **max** and **80th** percentile of the calendar year distribution



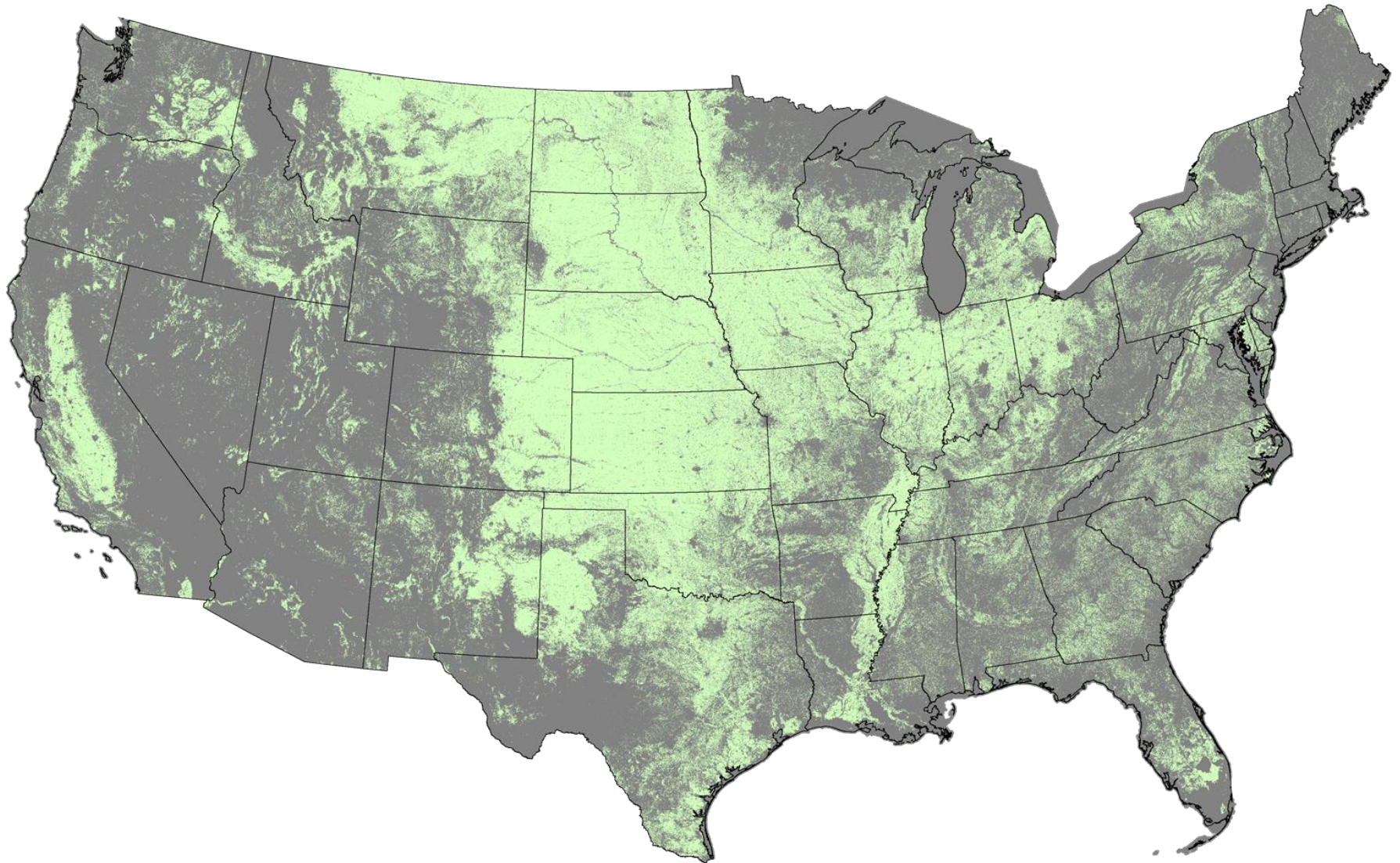
PART III—Efficient coarse-filter landscape monitoring

Phenological peakedness as the difference between the **max** and **85th** percentile of the **2002** fiscal year distribution



PART III—Efficient coarse-filter landscape monitoring

National Land Cover Dataset (NLCD 2006): grassland/herbaceous, pasture/hay, cultivated crops





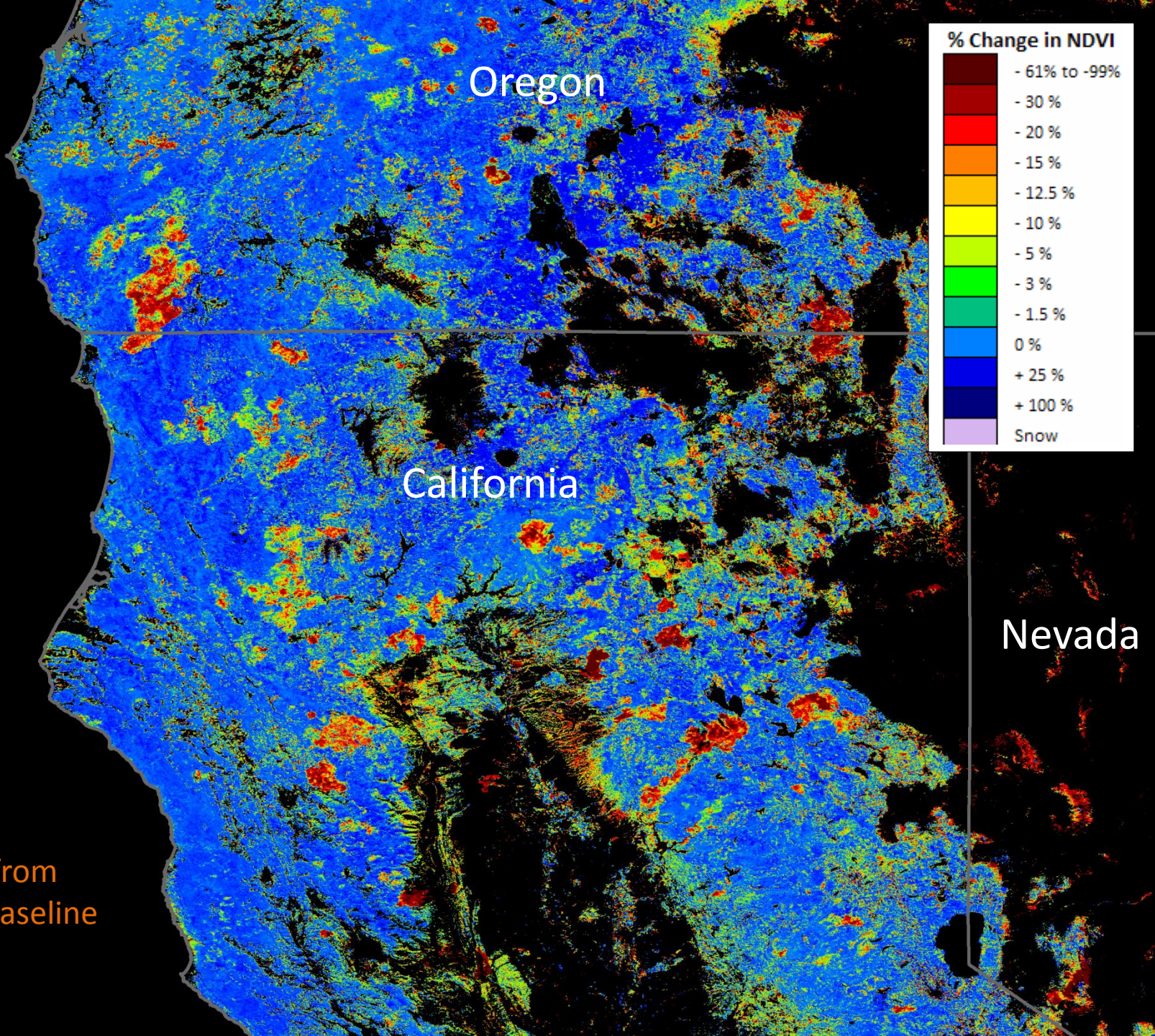
Outline

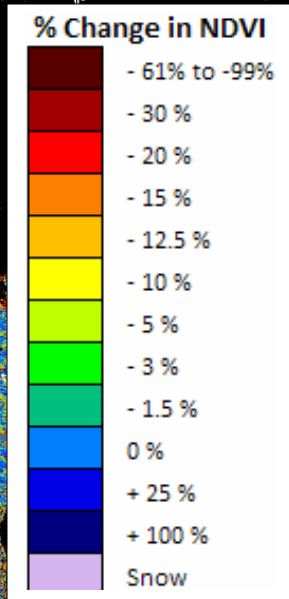
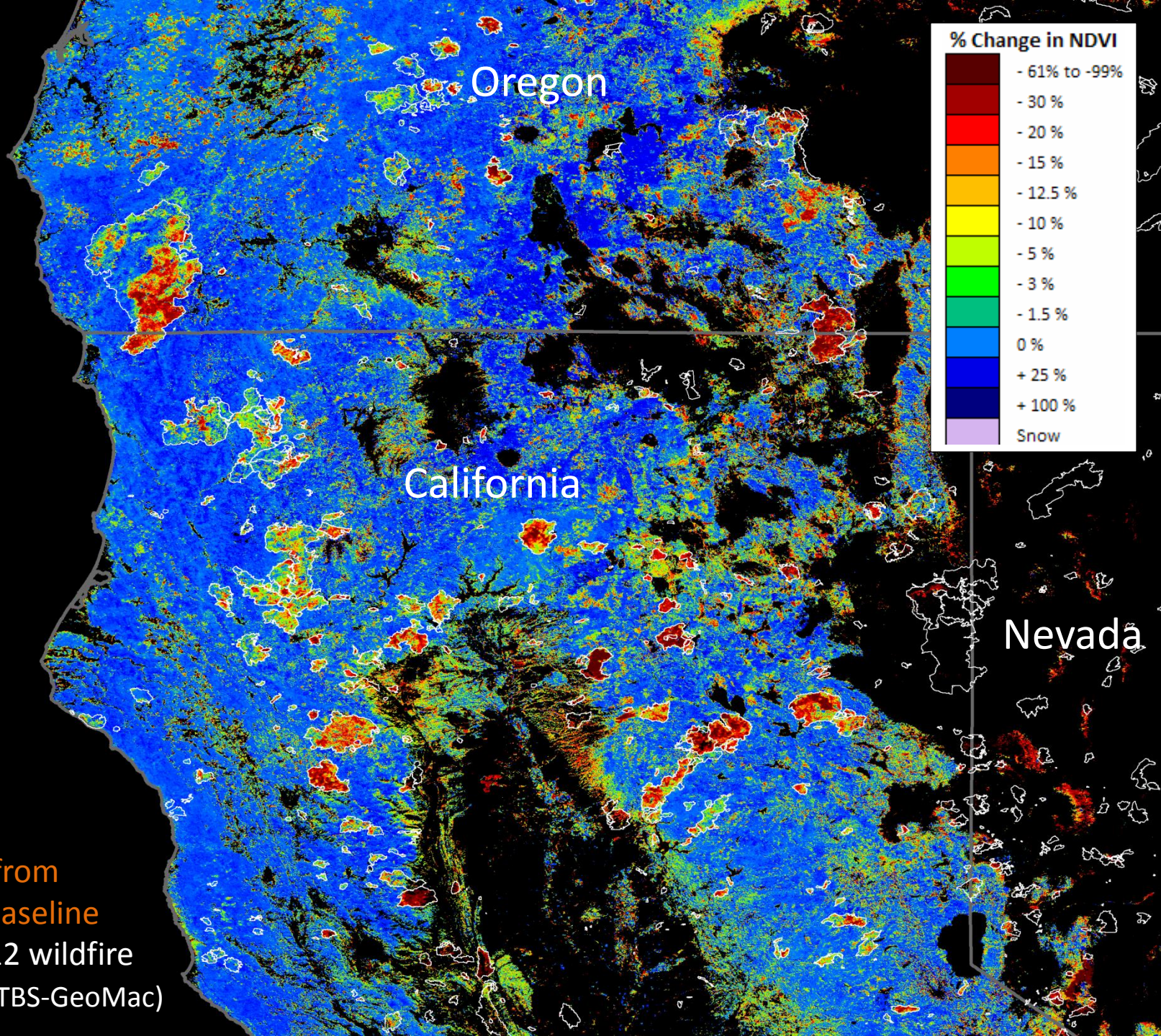
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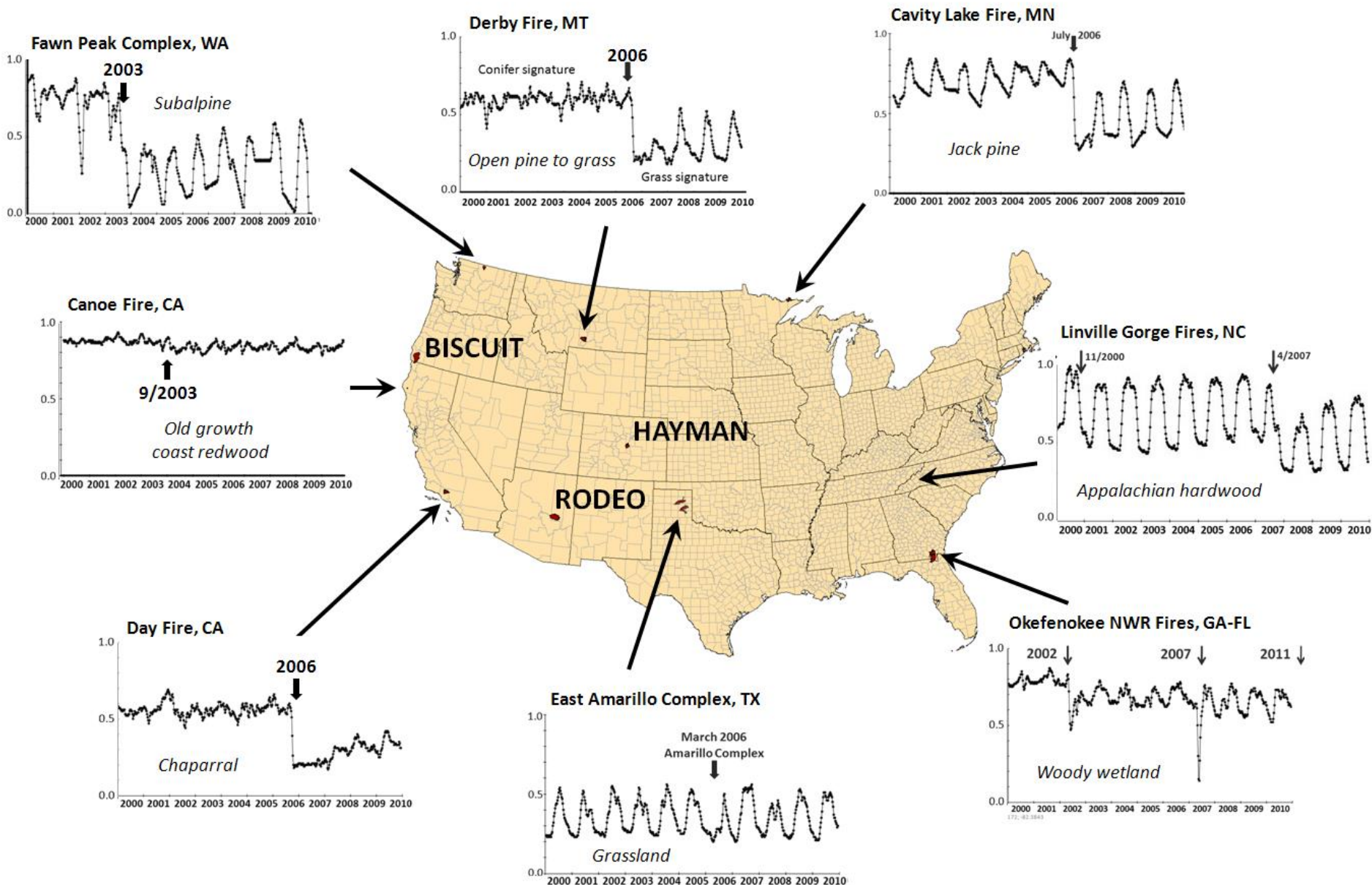
PART IV—Predicting long-term wildfire effects





ForWarn
Aug. 12, 2013
NDVI Change from
All-Year Max Baseline
with 2000-2012 wildfire
Perimeters (MTBS-GeoMac)

PART IV—Predicting long-term fire effects



Map Layers

Political Boundaries and Roads

Fires

Historical Aerial Disturbance Surveys (USFS FHM)

Forest Disturbance Detection Maps (MODIS NRT)

Forest NDVI Change (USFS TACs-NAS)

Change from 1-Year Baseline (recent)

current_jul20_aug12.tif

previous1_jul12_aug4.tif

previous2_jul4_jul27.tif

Change from All-Year Baseline (all data)

current_jul20_aug12.tif

previous1_jul12_aug4.tif

previous2_jul4_jul27.tif

Change from 3-Year Baseline (<3-Yr)

current_jul20_aug12.tif

eMODIS Cloud Assessment Products

Drought- US Drought Monitor

USGS/NDMC VegDRI Drought Assessment Products

Other Weather and Climate

Map Tools

Legend (click to clear layer)

TACs-NASA Products

% Change in NDVI

States

Print Map

Export Map to JPG

Graph NDVI

Share this Map

Biscuit Fire
2002

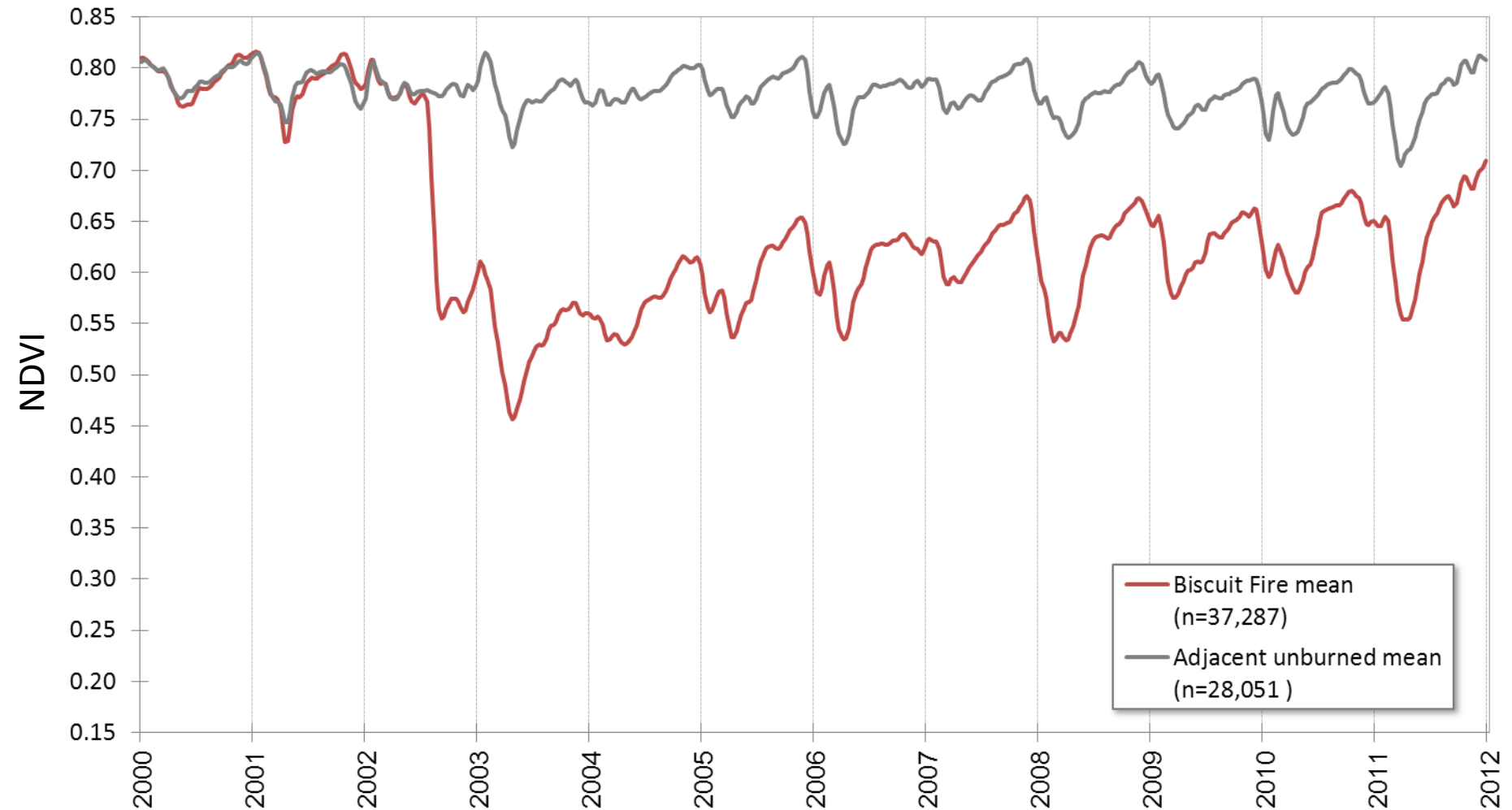
OR
CA

Date: August 12, 2013
Baseline: Max of all years



PART IV—Predicting long-term fire effects

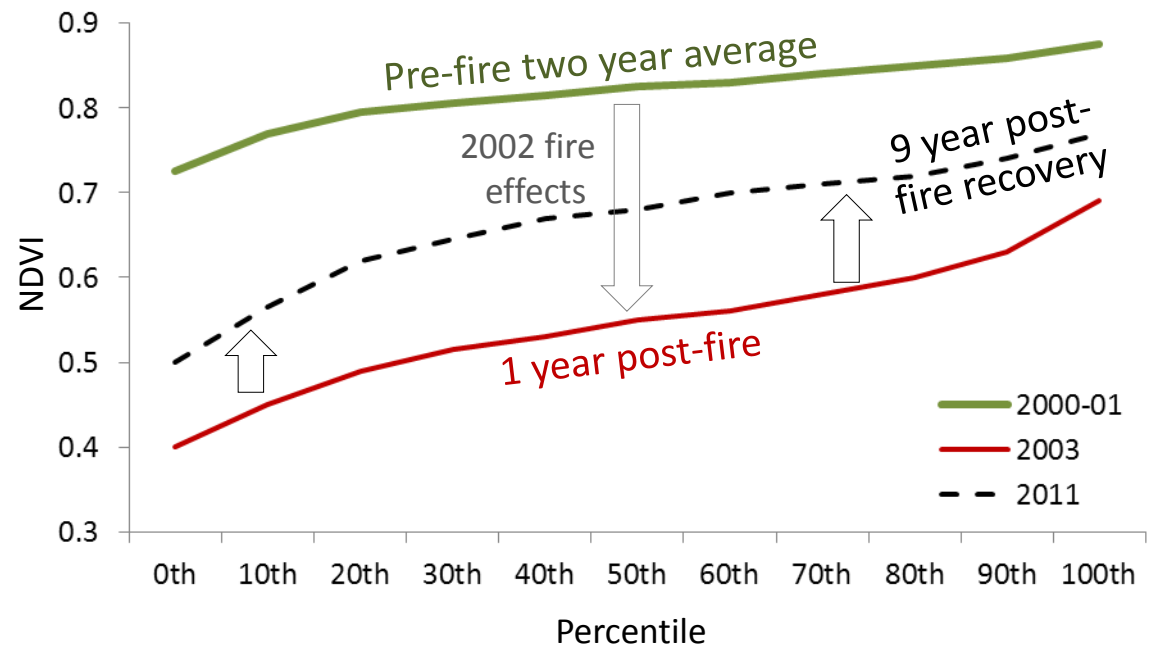
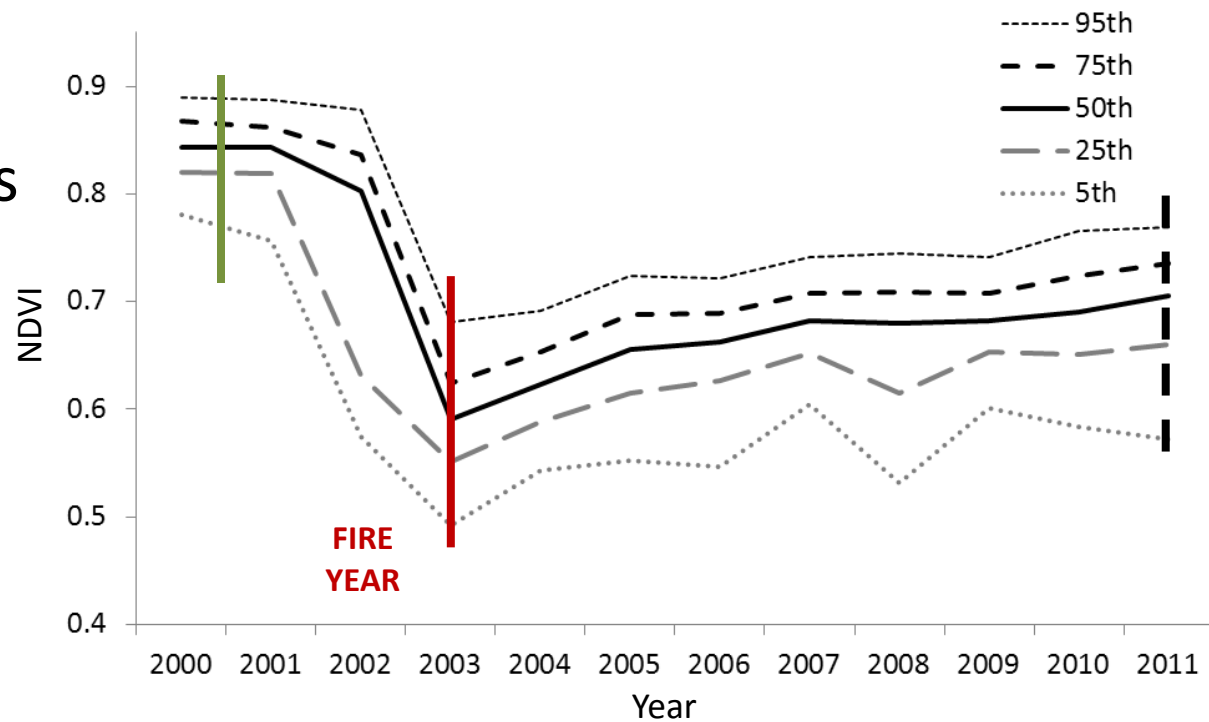
Mean observed MODIS NDVI for the 2002 **BISCUIT FIRE** and adjacent lands



In the red, 37,288 pixels are averaged (y) for each of 552 periods (x). This is 20,582,976 values.

PART IV—Predicting long-term fire effects

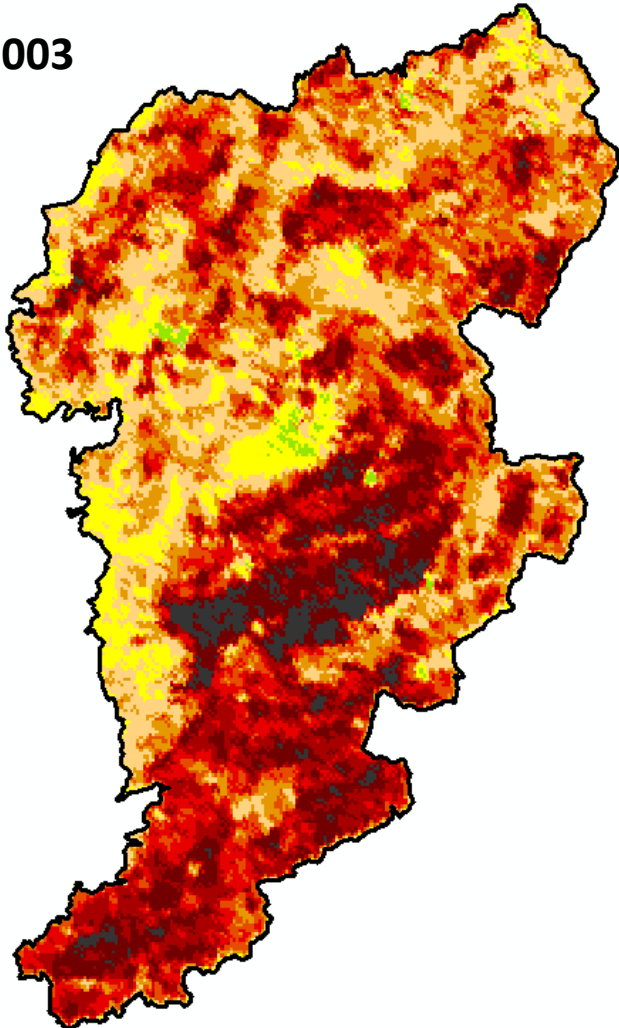
Change over time across percentiles of the annual NDVI distribution for the **BISCUIT FIRE**



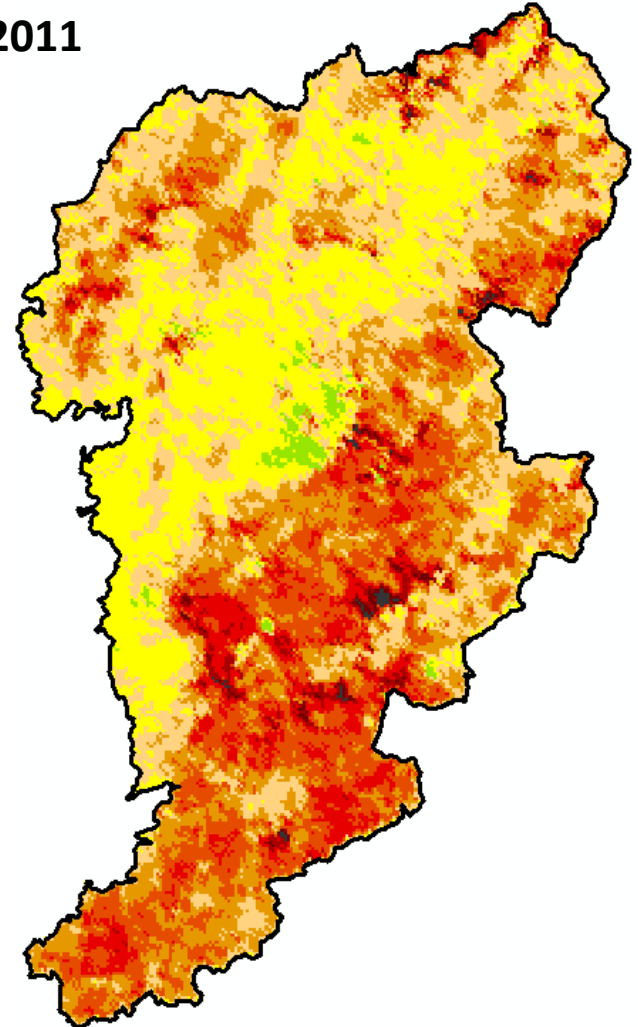
PART IV—Predicting long-term fire effects

Percent change in the Evergreen Fraction for the 2002 **BISCUIT FIRE**

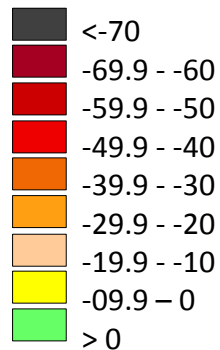
2003



2011



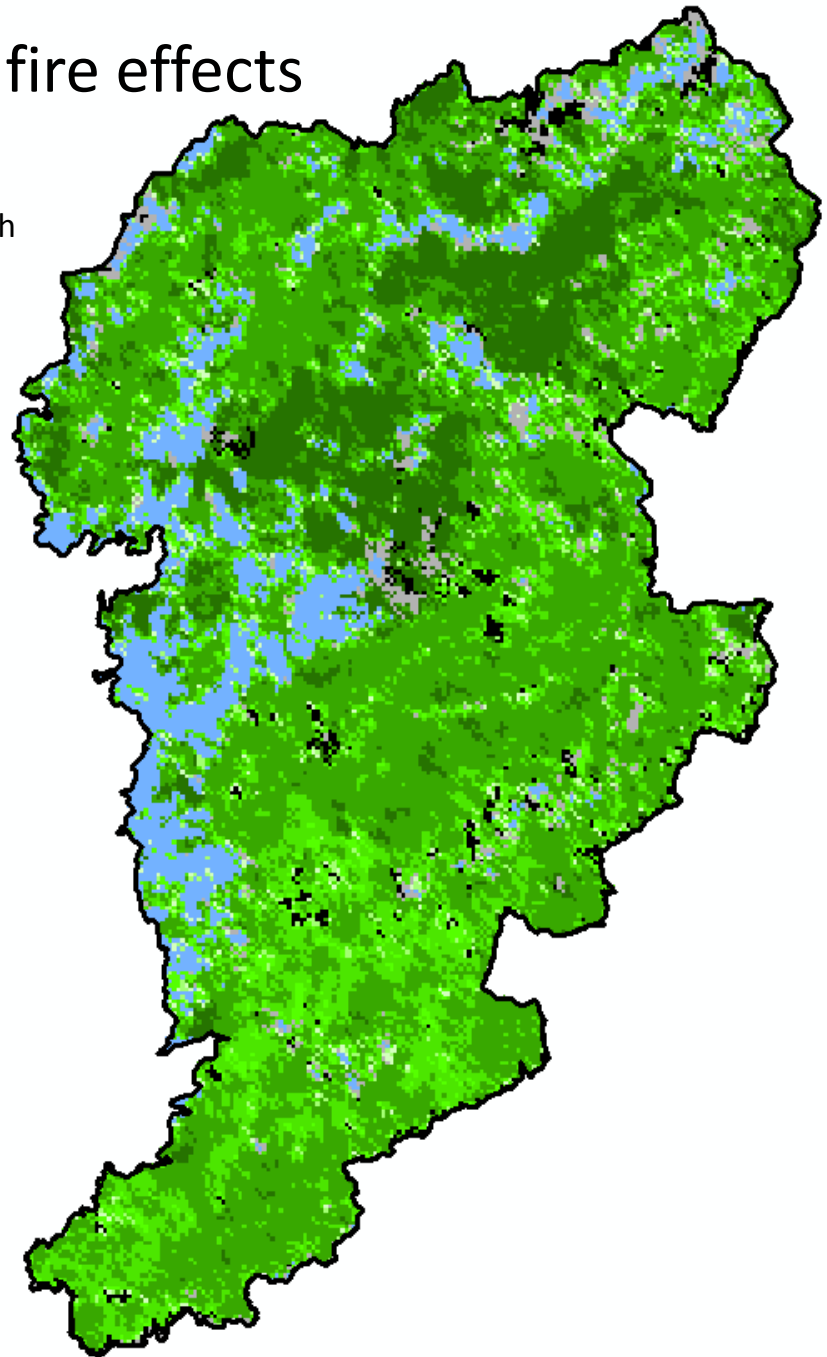
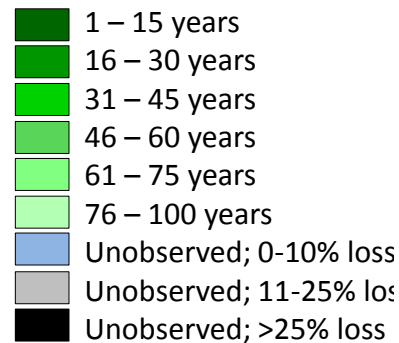
Percent change in
25th percentile
from 2000-01



PART IV—Predicting long-term fire effects

Predicted years to NDVI recovery for the **BISCUIT FIRE** based on change in the 50th percentile of calendar year distributions (annual medians), 2004-2011

Predicted Recovery



Rodeo-Chediski 2002

Wallow 2011

AZ NM

Whitewater-Baldy 2012

Date: August 12, 2013
Baseline: Max of all years

Map Layers

- Political Boundaries and Roads
- Fires
- Historical Aerial Disturbance Surveys (USFS FHM)
- Forest Disturbance Detection Maps (MODIS NRT)
- Forest NDVI Change (USFS TACs-NAS)**
 - Change from 1-Year Baseline (recent)**
 - current_jul20_aug12.tif
 - previous1_jul12_aug4.tif
 - previous2_jul4_jul27.tif
 - Change from All-Year Baseline (all data)**
 - current_jul20_aug12.tif
 - previous1_jul12_aug4.tif
- eMODIS Cloud Assessment Products
- Drought- US Drought Monitor
- USGS/NDMC VegDRI Drought Assessment Product
- Other Weather and Climate

Map Tools

Legend (click to clear layer)

TACs-NASA Products

% Change in NDVI

- 61% to -99%
- 30 %
- 20 %
- 15 %
- 12.5 %
- 10 %
- 5 %
- 3 %
- 1.5 %
- 0 %
- + 25 %
- + 100 %
- Snow

Counties

Print Map

Export Map to JPG

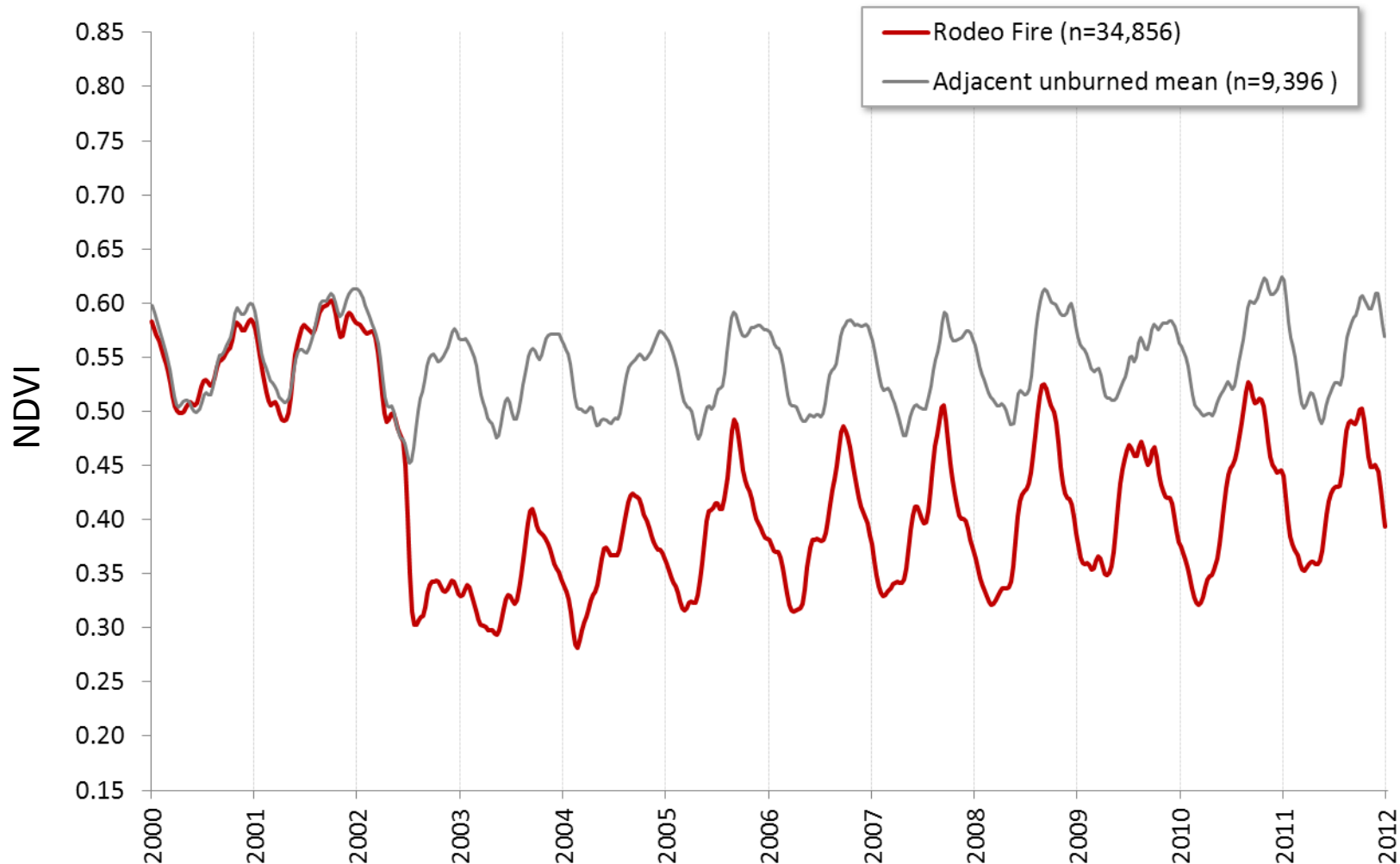
Graph NDVI

Share this Map



PART IV—Predicting long-term fire effects

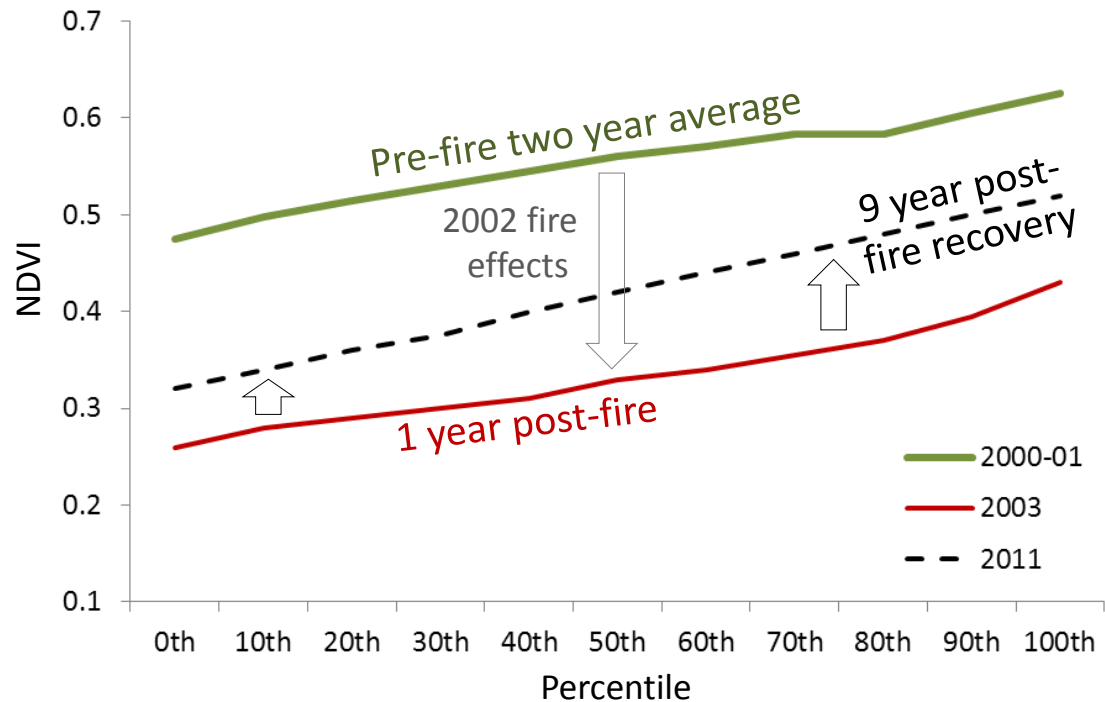
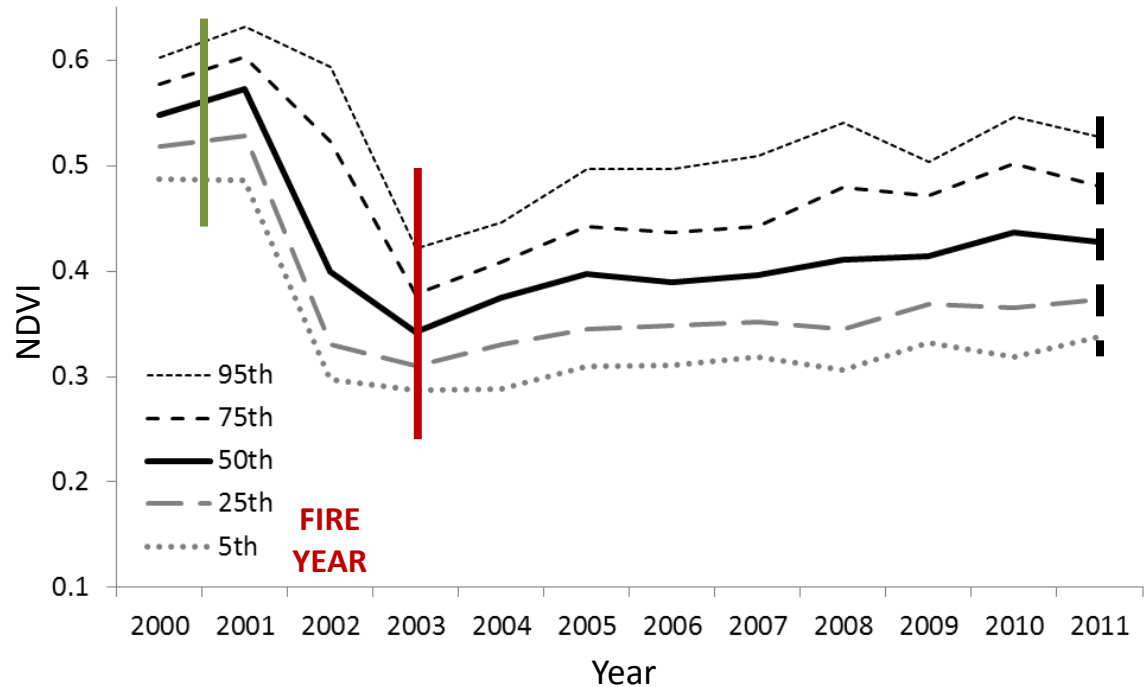
Mean observed MODIS NDVI for the **RODEO-CHEDISKI FIRE** and adjacent lands



Note: For only this one of the three fires, adjacent unburned values were not simply from the grid box surrounding the fire, but only those cells with mean 2000-01 values GTE 0.50 because areas within the box of this fire include low elevation grasslands, not mountain forests. Without this, the mean of the adjacent is close to 0.45 not 0.55 as it is with this selective adjustment.

PART IV—Predicting long-term fire effects

Change over time across percentiles of the annual NDVI distribution for the **RODEO-CHEDISKI FIRE**



PART IV—Predicting long-term fire effects

Percent change in the Evergreen Fraction for the 2002 **RODEO-CHEDISKI FIRE**

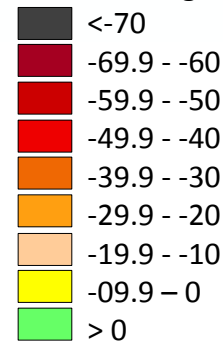
2011

Change in 25th Percentile
2000-01 to 2011

Change in 25th Percentile
2000-01 to 2003

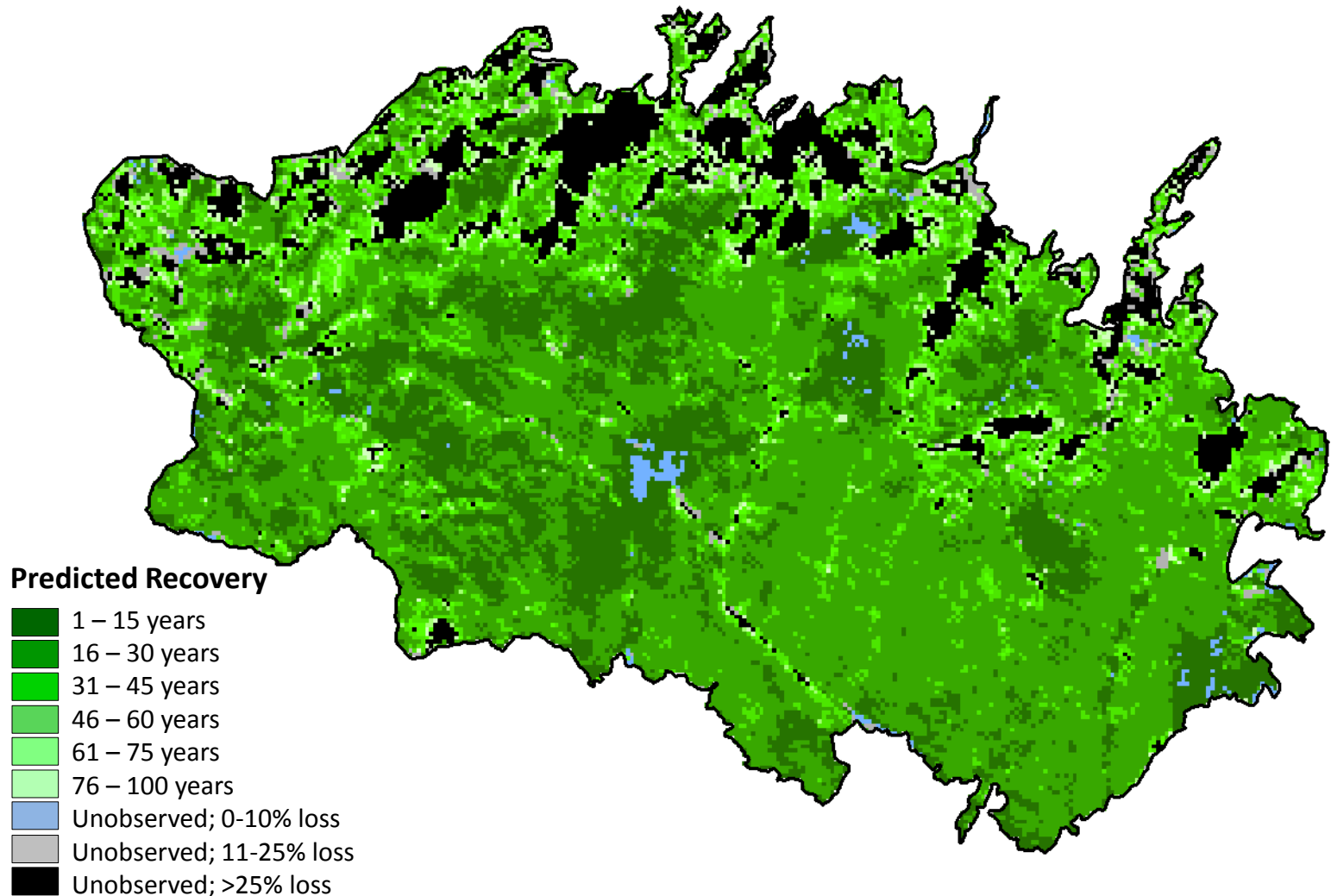
2003

% change



PART IV—Predicting long-term fire effects

Predicted years to NDVI recovery for the **RODEO-CHEDISKI FIRE** based on change in the 50th percentile of calendar year distributions (annual medians), 2004-2011





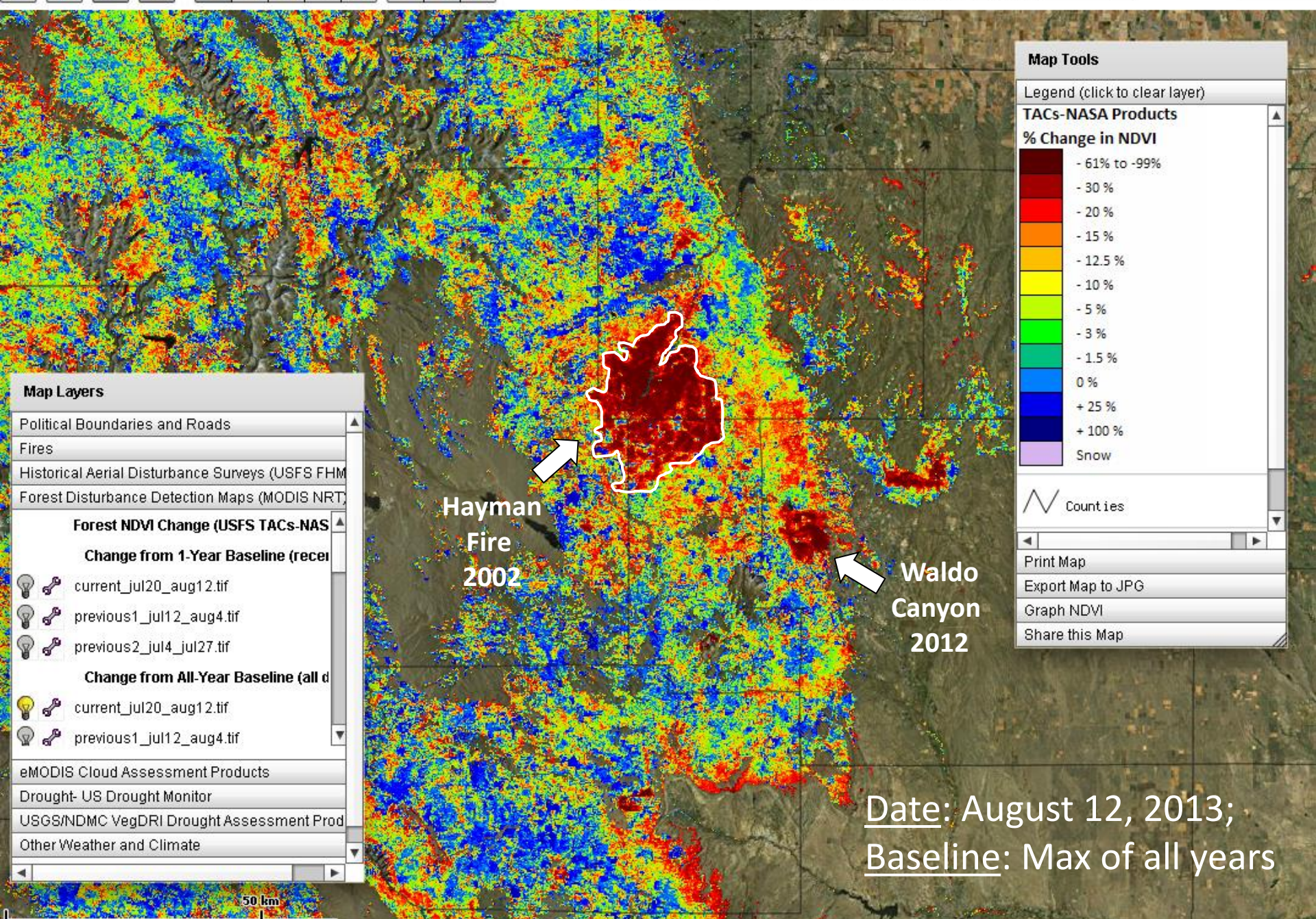
Basemap

Imagery

Theme

CONUS Vegetation Monitoring Tools

Find Area



Map Layers

Political Boundaries and Roads
Fires
Historical Aerial Disturbance Surveys (USFS FHM)
Forest Disturbance Detection Maps (MODIS NRT)

Forest NDVI Change (USFS TACs-NAS)

Change from 1-Year Baseline (recent)

current_jul20_aug12.tif
previous1_jul12_aug4.tif
previous2_jul4_jul27.tif

Change from All-Year Baseline (all data)

current_jul20_aug12.tif
previous1_jul12_aug4.tif

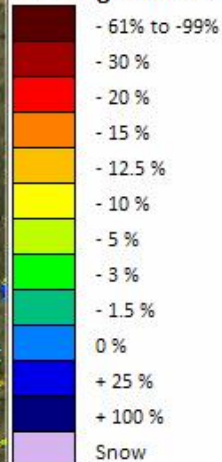
eMODIS Cloud Assessment Products
Drought- US Drought Monitor
USGS/NDMC VegDRI Drought Assessment Product
Other Weather and Climate

Map Tools

Legend (click to clear layer)

TACs-NASA Products

% Change in NDVI



Counties

Print Map

Export Map to JPG

Graph NDVI

Share this Map

Hayman
Fire
2002

Waldo
Canyon
2012

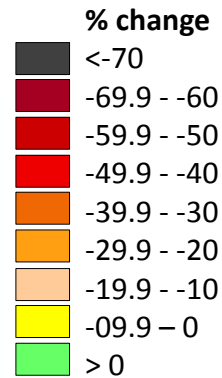
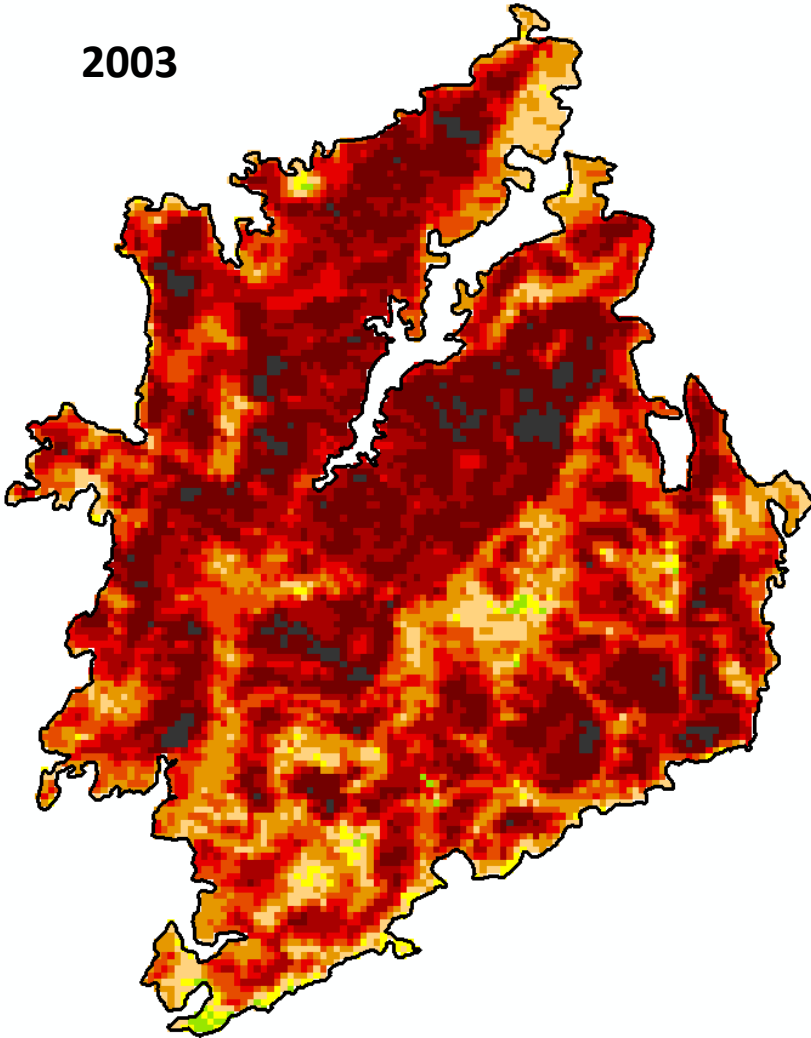
Date: August 12, 2013;
Baseline: Max of all years

PART IV—Predicting long-term fire effects

Percent change in the Evergreen Fraction for the 2002 **HAYMAN FIRE**

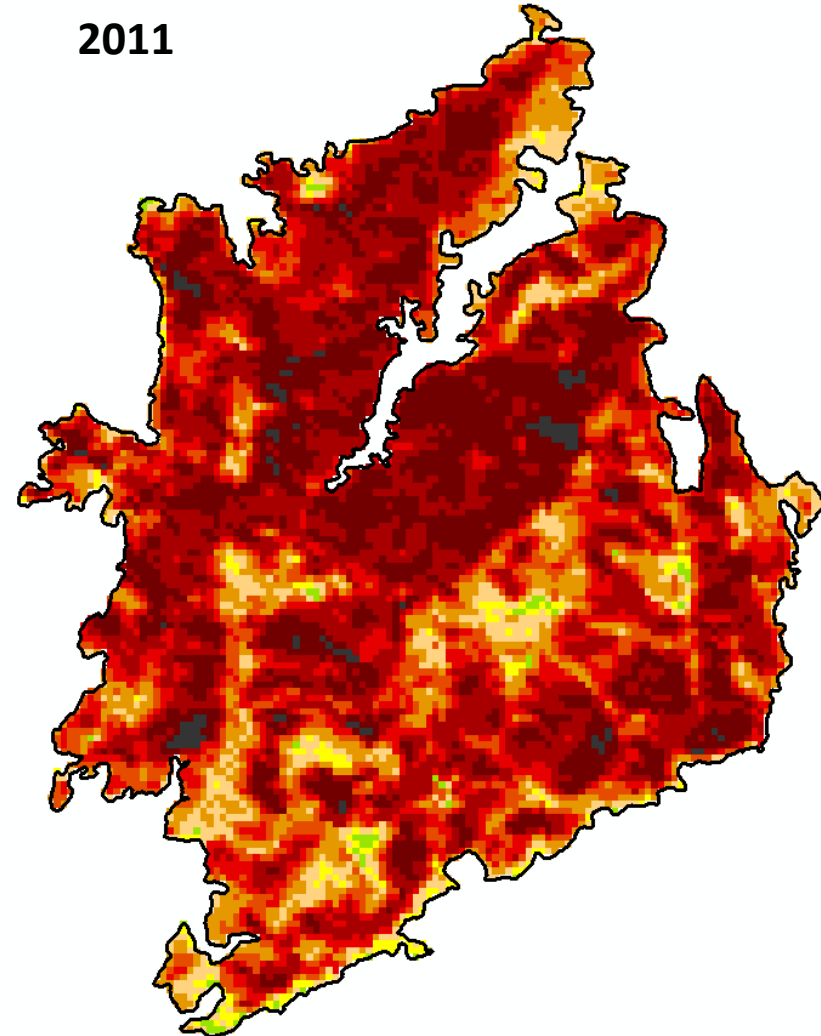
Change in 25th Percentile
2000-01 to 2003

2003



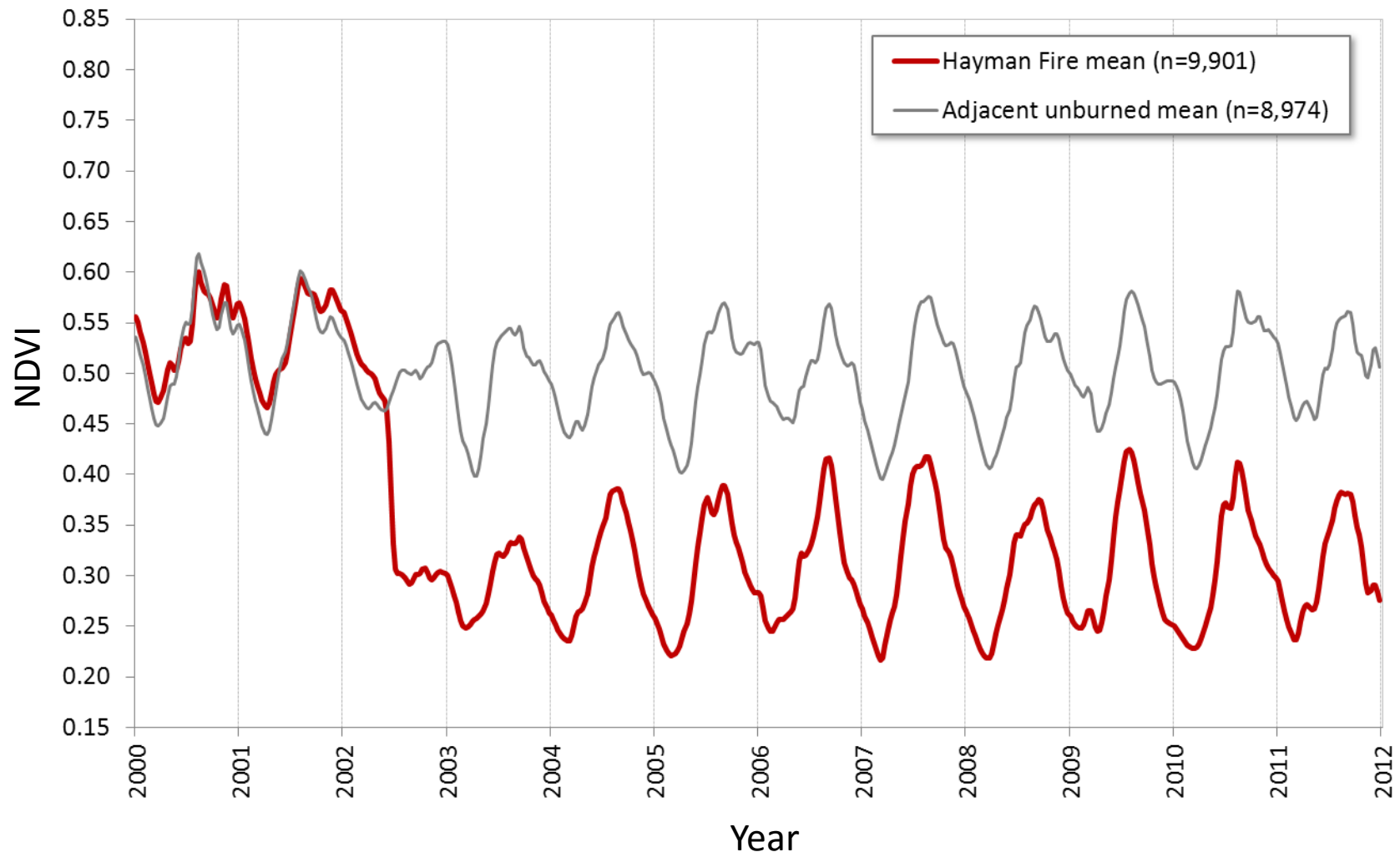
Change in 25th Percentile
2000-01 to 2011

2011



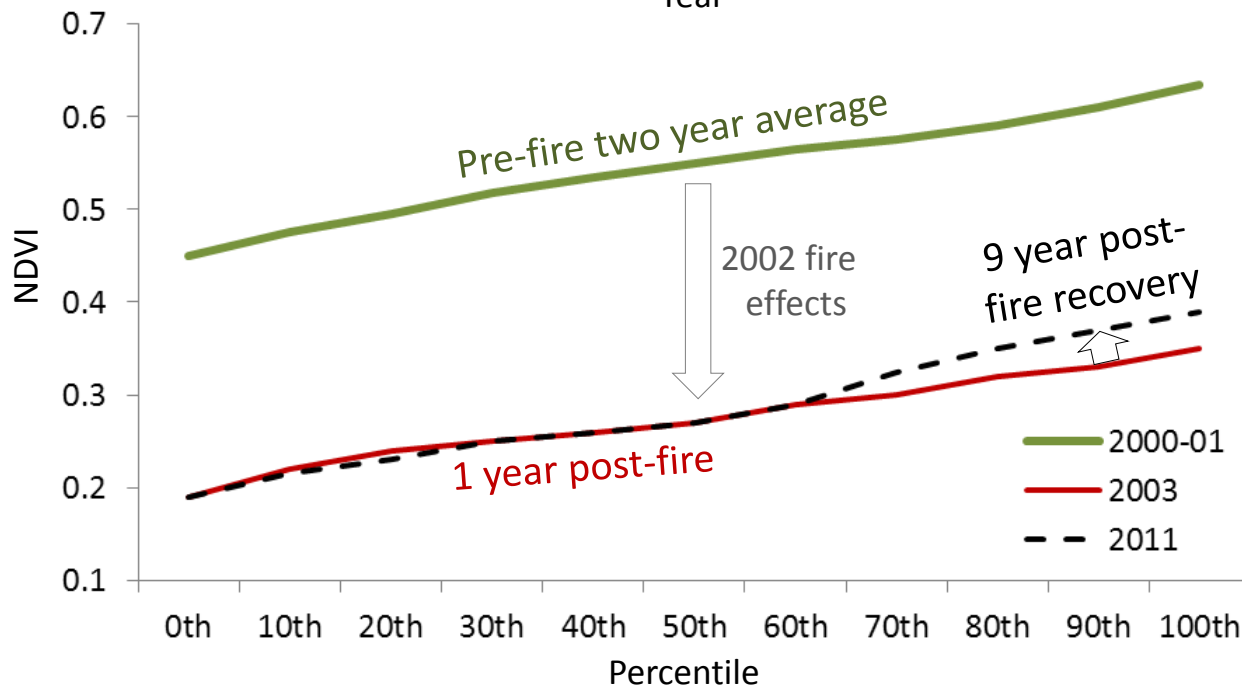
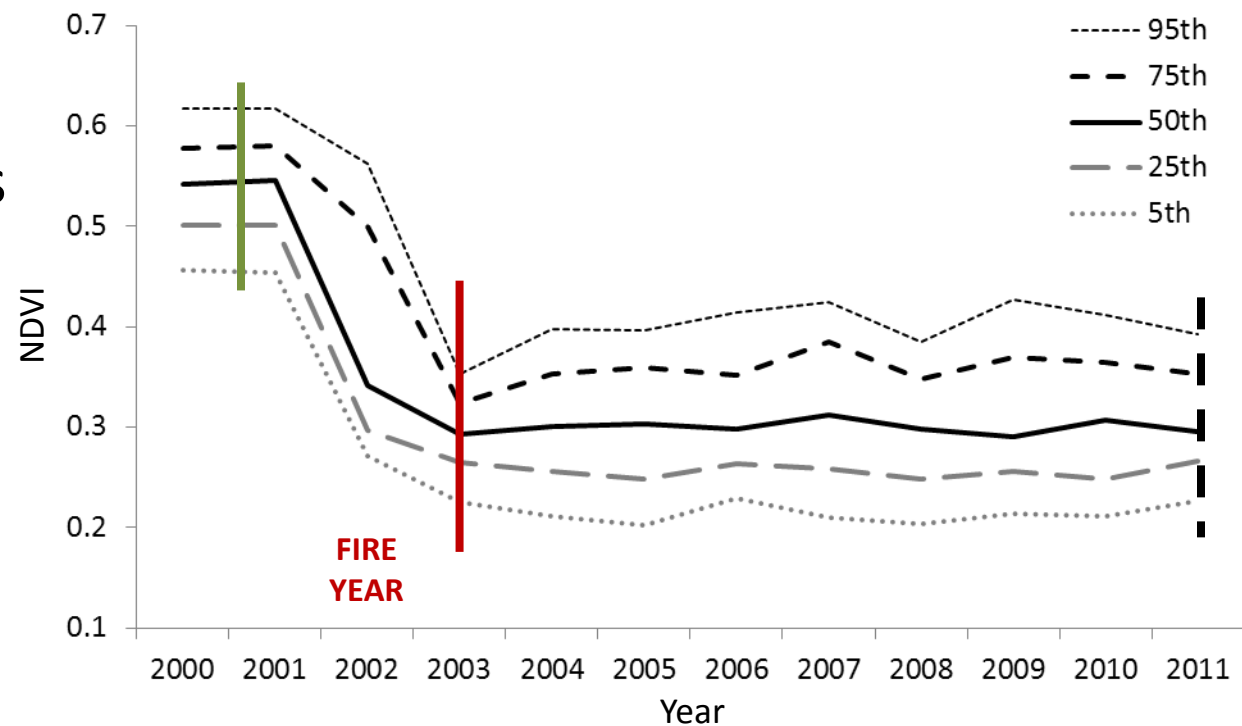
PART IV—Predicting long-term fire effects

Mean MODIS NDVI for the 2002 **HAYMAN FIRE** and adjacent lands



PART IV—Predicting long-term fire effects

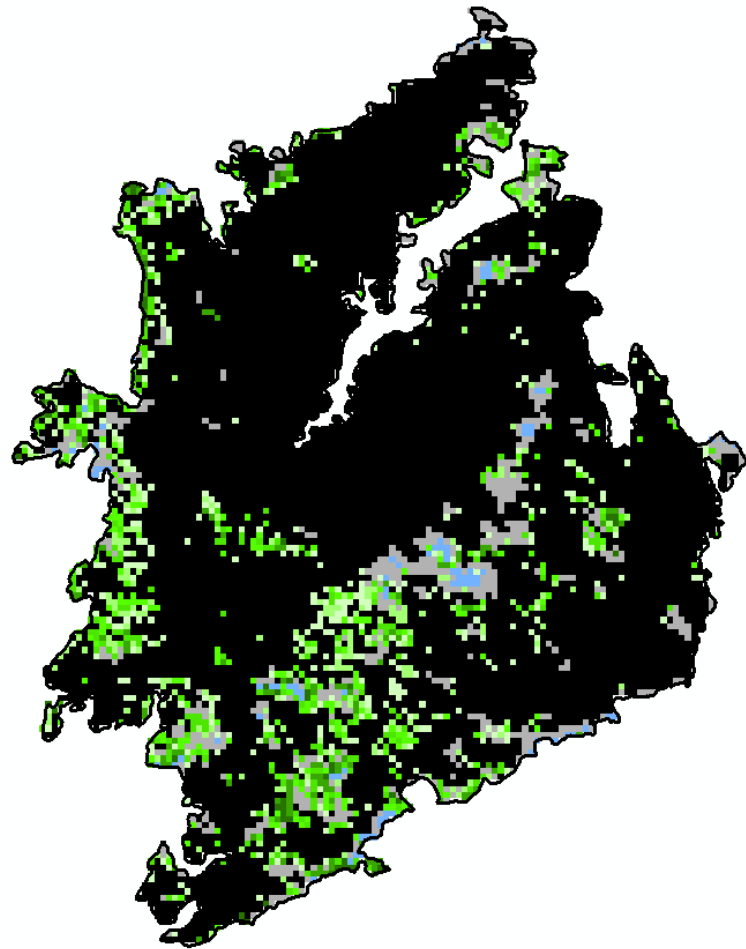
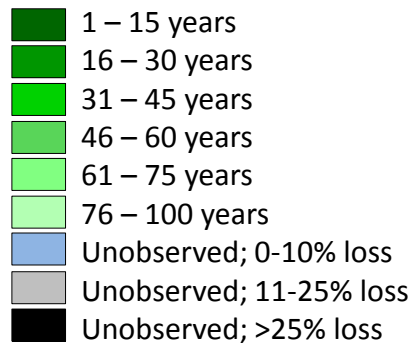
Change over time across percentiles of the annual NDVI distribution for the **HAYMAN FIRE**, Colorado.



PART IV—Predicting long-term fire effects

Predicted years to NDVI recovery for the **HAYMAN FIRE** based on change in the 50th percentile of calendar year distributions (annual medians), 2004-2011

Predicted Recovery





Summary

- (1) High frequency, moderate resolution MODIS NDVI provides insights into *short* and *long-term* fire effects.
- (2) Recovery to pre-fire or progress toward *desired conditions* can be *predicted*.
- (3) This approach also provides a uniform coarse filter mechanism for *ecological process* monitoring.
- (4) This functions for many other disturbances, and therefore for coarse aspects of disturbance interactions including cumulative effects from causes, both indigenous and novel.