
DEVELOPMENT AND PRODUCTION OF A MODERATE RESOLUTION FOREST TYPE MAP OF THE UNITED STATES

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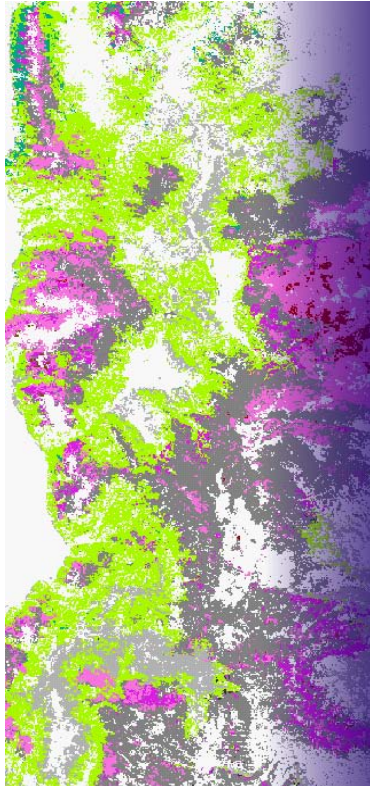
Zhu and Evans (1994) produced a forest group type map covering the entire United States (US) and Puerto Rico. This map was one of the key components for the development of the forest risk maps produced by the US Department of Agriculture Forest Service (USFS) Forest Health Monitoring (FHM) program. (Lewis 2002). The forest risk maps, which were actually produced in 2000, were non site-specific and only identified broad areas that had potential high risk of forest mortality or growth/volume loss from insects and diseases. The intent of these forest risk maps was to provide national information to policy makers to help determine national priorities. The forest risk mapping effort found 43 million hectares of forest land (14 percent of total forest lands) in danger of high tree mortality and growth/volume loss. According to Forest Health Protection (2004), tree mortality totaled 800,000 hectares in 2000. In 2003 tree mortality totaled over 4 million hectares. It is unknown how forest risk increased during this time period, but it is conceivable that forest risk also showed dramatic increases due to increases in climatic and anthropomorphic stresses and increases in globalization. Thus, there is a need to update and improve the forest risk maps. The improvements will allow the products to be used more specifically and at smaller scales. The forest group type map used for the forest risk mapping was produced in 1992 using AVHRR (Advanced Very High Resolution Radiometer) imagery with a spatial resolution of 1 km and, thus, the forest risk maps have a spatial resolution of 1 km as well (Zhu and Evans 1994). In 1999 the Moderate-resolution Imaging Spectroradiometer (MODIS) sensor with a spatial resolution of 250 meters became available. Additionally, other national remote sensing images and geographical information system (GIS) layers have become available since 1992.

These factors prompted Forest Health Monitoring (FHM), the Remote Sensing Applications Center (RSAC), and Forest Inventory and Analysis (FIA) to initiate the forest type and forest type group mapping effort described below.

Consistent national scale forest type information is important for modeling forest areas at risk of increased mortality due to insects and diseases. We used classification-trees to model forest type groups and forest types for the continuous United States (US) and Alaska. The modeling procedure used a geo-spatial dataset developed by the US Department of Agriculture Forest Service (USFS) for the predictor data and plot data from USFS Forest Inventory and Analysis (FIA) for the response data. The geo-spatial predictor dataset consisted of 269 remote sensing images and geographical information system (GIS) layers covering the continental US and 19 remote sensing images and GIS layers covering Alaska. The FIA plot data consisted of 78,127 forested plots covering the continuous US and 5,392 forested plots covering Alaska. Ten percent of the response data were separated from the modeling procedure and were used for accuracy assessment purposes. The overall accuracies for the continuous US for the forest type group and forest type were 69 percent and 50 percent, respectively. The overall accuracies for Alaska for the forest type group and forest type were 78 percent and 67 percent, respectively. To further verify the results, state summaries were derived from the final modeled products and compared to FIA state summary tables. For the continuous US, 95 percent of the forest type groups were within + 400,000 hectares of the FIA state summaries and 95 percent of the forest types were within + 290,000 hectares of the FIA state summaries. For Alaska, the forest type groups were within + 48,000,000 hectares of the FIA state summaries and the forest types were within + 27,000,000 hectares of the FIA state summaries; the FIA state summaries were only for southeast coastal Alaska whereas the state summaries derived from the modeled products included the whole state of Alaska.

Following the presentation of the background and reasons for development as well as the mapping methods and results the questions/issues below were discussed:

- Appropriate uses ...
 - Scale
 - Applications
- Use of map and plots together
- Map unit descriptors
- Forest types and type groups
 - “Right” classifications for risk mapping?
 - Necessary but not sufficient?
 - What structural characteristics are needed?




Development and Production of a Moderate Resolution Forest Type Map of the United States

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FIA Remote Sensing Band
All Regional Units

Presented at:
Evaluation of Quantitative Techniques
for deriving National scale Data
for Assessing and Mapping Risk Workshop
DoubleTree Hotel, Denver North
Westminster, Colorado

July 26-28, 2005



National Forest Type (NFT) Mapping Background

Impetus and Sponsorship

- ca. 2005 Forest Risk Map Revision (FHM)
- RSAC / FIA RSB Experience with National Mapping
- 12 Years since last map

What has changed in the last 10 years?

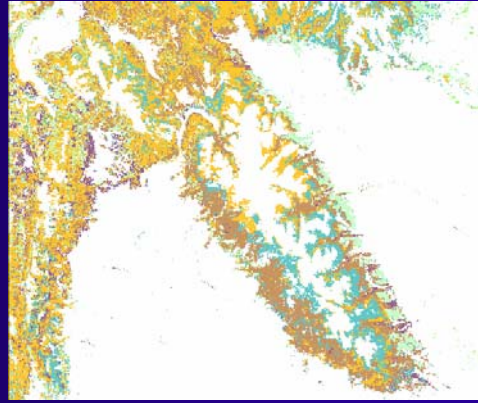
- Then ...
 - 1993 Forest Type Group Map (Zhu and Evans) covered the entire U.S. including Alaska, Hawaii, and Puerto Rico
 - 1993 Forest Type Group Map - 1-km spatial resolution (FIA Plots, AVHRR, elevation data were primary data sources)
- Now ...
 - Wide Availability of Nationwide Continuous Geospatial Data
 - New Satellite Imagery (MODIS, 250-m, 7-band)
 - New Modeling Techniques
 - Computing Environment Improvements

Why is it better than existing data?

Example: 1992 – 1 kilometer



Pilot Area Example: 2004 – 250 meter



Hypothesis: Increased Thematic Resolution may be possible with ...

- Improved data sources (remote sensing and otherwise)
- Higher spatial resolution data sources – smaller pixels increases homogeneity of pixels
- Improved modeling methods

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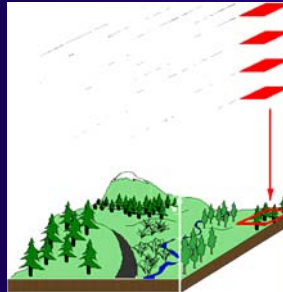
Presentation Overview

- Methods and Data
 - ◆ Predictor and Response Variables
 - ◆ Mapping zones and logic
 - ◆ Modeling method and Custom Tools
- Presentation of Results
 - ◆ Graphical
 - ◆ Accuracy vs. Utility Assessment
- Remaining Questions

General Methodology – Spatial Modeling

-- Geospatial Predictor Layers --

MODIS imagery.....
 MODIS Veg indices
 Elevation
 Aspect
 Slope
 STATSGO.....
 Etc.....



-- Response Variables –
 (FIA Plot Data)

**Predictive models developed for Forest Type
 using data from FIA plots and Geospatial Predictor Layers**

Figure Courtesy of Interior West FIA

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National Geospatial Predictors Database

- Elevation, slope, and aspect
- Ecological region layers
 - > USGS Mapping Zones, Bailey's Ecoregions, and Unified Ecoregions for Alaska
- STATSGO data layers (Lower 48)
 - > Available water capacity, soil bulk density, soil permeability, soil ph, soil porosity, soil plasticity, depth to bedrock, rock volume, soil types, and soil texture
- MODIS Vegetation Indices such as EVI, NDVI
- MODIS Vegetation Continuous Fields
 - > Percent tree cover, percent herbaceous cover, and percent bare ground
- MODIS fire points developed from the MODIS Active Fire Maps
- MODIS 8-day composites
 - > Multiple dates from the spring, summer, and fall
- All MODIS 32-day composite imagery that are cloud-free between the years 2001-2003
- USGS NLCD layers
- PRISM temperature and precipitation
 - > Minimums, maximums, and averages

Over 200 national predictor layers

Response Variables – FIA Plot Data

- 80,948 forested* FIA plots** for continental United States
- 5,392 forested FIA plots for Alaska
- Data collected between 1978 - 2004 – majority of the data collected between 2000 - 2004
- FIA variables – Forest Type and Forest Type Group

* The response variables included all FIA plots that were at least 50% forested. A forested plot is defined as being greater than 0.4 ha (1 acre) in size, greater than 120 feet in width, with greater than 10% stocking, and an undisturbed understory.

** FIA plots occur throughout the U.S. at an intensity of approximately one per 6,000 acres

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Response Variables – 141 FIA Forest Types

Jack Pine	Engelmann Spruce	Post Oak/Blackjack Oak	Black Cherry
Red Pine	Engelmann Spruce/Subalpine Fir	Chestnut Oak	Cherry/Ash/Yellow-poplar
Eastern White Pine	Grand Fir	Red Oak/White Oak/Hickory	Hard Maple/Basswood
Eastern White Pine/Eastern Hemlock	Subalpine Fir	White Oak	Elm/Ash/Locust
Hemlock	Blue Spruce	Northern Red Oak	Red Maple/Upland
Eastern Hemlock	Mountain Hemlock	Yellow-popular/White Oak/Northern Red Oak	Aspen
Balsam Fir	Alaska-Yellow-Cedar	Sassafras/Persimmon	Paper Birch
White Spruce	Lodgepole Pine	Sweetgum/Yellow-poplar	Balsam Poplar
Red Spruce	Western Hemlock	Bur Oak	Red Alder
Red Spruce/Balsam Fir	Western Redcedar	Scarlet Oak	Bigleaf Maple
Black Spruce	Sitka Spruce	Yellow Poplar	Gray Pine
Tamarack	Western Larch	Black Walnut	California Black Oak
Northern White-cedar	Redwood	Black Locust	Oregon White Oak
Longleaf Pine	Giant Sequoia	Southern Scrub Oak	Blue Oak
Slash Pine	Knobcone Pine	Chestnut Oak/Black Oak/Scarlet Oak	Deciduous Oak Woodland
Loblolly Pine	Southwest White Pine	Red Maple/Oak	Coast Live Oak
Shortleaf Pine	Bishop Pine	Mixed Upland Hardwoods	Canyon Live Oak/Interior Live Oak
Virginia Pine	Monterey Pine	Swamp Chestnut Oak/Cherrybark Oak	Tanoak
Sand Pine	Foxtail Pine/Bristlecone Pine	Overcup Oak/Water Hickory	California Laurel
Table Mountain Pine	Limber Pine	Atlantic White-cedar	Giant Chinkapin
Pond Pine	Whitebark Pine	Baldcypress/Water Tupelo	Pacific Madrone
Pitch Pine	Misc. Western Softwoods	Sweetbay/Swamp Tupelo/Red Maple	Mesquite Woodland
Spruce Pine	California Mixed Conifer	River Birch/Sycamore	Cercocarpus Woodland
Eastern Redcedar	Scotch Pine	Cottonwood	Intermountain Maple Woodland
Rocky Mountain Juniper	Australian Pine	Willow	Misc. Western Hardwood Woodlands
Western Juniper	Other Exotic Softwoods	Sycamore/Pecan/American Elm	Stable Palm
Juniper Woodland	Norway Spruce	Sugarberry/Hackberry/Elm/Green Ash	Mangrove
Pinyon Juniper Woodland	Introduced Larch	Red Maple/Lowland Cottonwood/Willow	Other Tropical
Douglas-fir	Eastern White Pine/Northern Red Oak/White Ash	Oregon Ash	Paulownia
Port-Orford-Cedar	Eastern Redcedar/Hardwood Longleaf Pine/Oak	Sugar Maple/Beech/Yellow Birch	Melaluca
Ponderosa pine	Shortleaf Pine/Oak Virginia Pine/Southern Red Oak		Eucalyptus
Incense Cedar	Loblolly Pine/Hardwood		
Jeffrey Pine/Coulter Pine/Bigcone Douglas-fir	Slash Pine/Hardwood		
Sugar Pine	Other Pine/Hardwood		
Western White Pine			
White Fir			
Red Fir			
Noble Fir			
Pacific Silver Fir			

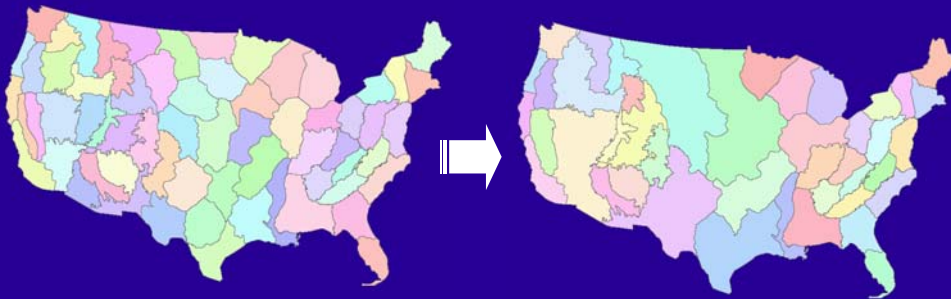
Response Variables – 28 FIA Forest Type Groups

White/Red/Jack Pine Group	California Mixed Conifer Group
Spruce/Fir Group	Exotic Softwoods Group
Longleaf/Slash Pine Group	Oak/Pine Group
Loblolly/Shortleaf Pine Group	Oak/Hickory Group
Pinyon/Juniper Group	Oak/Gum/Cypress Group
Douglas-fir Group	Elm/Ash/Cottonwood Group
Ponderosa Pine Group	Maple/Beech/Birch Group
Western White Pine Group	Aspen/Birch Group
Fir/Spruce/Mountain Hemlock Group	Alder/Maple Group
Lodgepole Pine Group	Western Oak Group
Hemlock/Sitka Spruce Group	Tanoak/Laurel Group
Western Larch Group	Other Western Hardwoods Group
Redwood Group	Tropical Hardwoods Group
Other Western Softwoods Group	Exotic Hardwoods Group

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Modeling Performed on Mapping Zones

For production purposes and to prevent anomalies, such as loblolly pine occurring in California, the continental U.S. was divided into the 65 USGS mapping zones. Alaska was not divided into mapping zones.



Because of the paucity of forest land in some mapping zones, several mapping zones were combined together creating a total of 43 mapping zones.

Modeling Using Cubist / See5

What is Cubist / See5? (www.rulequest.com)

- Powerful tool for generating rule-based models
 - Software helps determine which variables are important for modeling
- No assumptions about the data distribution (non-parametric models)
- Handles numeric and categorical data
- Handles large databases
- Able to incorporate both Remote Sensing and GIS data layers
- Easy to understand and efficient to run
- Easy to generate predictions based upon the models

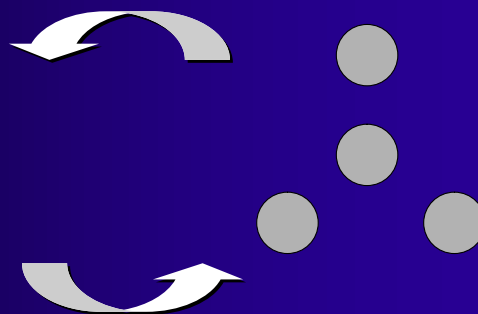
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Modeling / Mapping Tools

Remote Sensing/GIS Data



FIA Plot Data



Challenge:

How to combine Remote Sensing & GIS data with FIA plot data to produce a forest type geospatial product?

Loosely Coupled Imagine & Cubist / See5

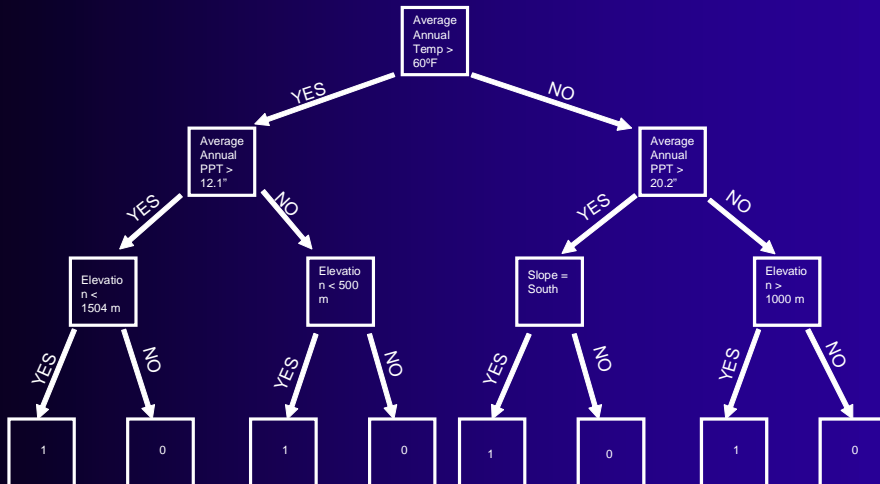
ERDAS Imagine GUI developed at RSAC that prepares FIA data (or any other plot data) for use with Cubist / See5 modeling application

The screenshot shows the 'Prepare FIA Data For Cubist' dialog box in ERDAS Imagine. It includes fields for input and output files, coordinate headers, and subset definitions. Several callout boxes provide additional context:

- Stacks and subsets imagery:** Points to the 'Text file containing list of images to stack' and 'Input Stacked Image' fields.
- Extracts the geospatial image information using plot coordinates:** Points to the 'Input file containing the coordinates' and 'Output file containing the coordinates' fields.
- Creates the .data and .names files for Cubist/See5:** Points to the 'Output Cubist Database' field.
- Combines the geospatial image information with the plot data:** Points to the 'File containing image information for the points' field.
- Randomly selects data to set aside for an accuracy assessment:** Points to the 'Percent' field set to 10.

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Classification Trees



Cubist / See5 Differences and Output

Cubist Output - Rule Sets

```

Rule 1 [50 cases, mean 17.4, range 0 to 132, est err 19.7]
if
  conus_mean_elevation_img <= 2770.062
  conus_modis32_2002225_albers_img_band6 > 1090
  conus_ndvi_2001097_img <= 32
then
  A_STAGE = -122.9 - 0.000104 X Coordinate + 0.076 us_tavg209_albers_img
            - 0.0083 conus_reflectance_2002097_img_band1
            - 0.039 ustmax09_albers_img - 0.037 ustmin07_albers_img
            + 0.0026 us_ppt14_annual_img + 0.0018 us_ppt12_dec_img
            + 0.0021 conus_ndvi_2002321_img - 0.006 us_tavg201_albers_img
Rule 2 [27 cases, mean 22.4, range 0 to 99, est err 25.5]
if
  conus_mean_elevation_img <= 2770.062
  conus_modis32_2001097_albers_img_band1 <= 1009
  conus_modis32_2002129_albers_img_band1 <= 1221
  conus_ndvi_2002225_img <= 3719
  conus_nlcd_percent_woodforest_img <= 71
  148_trasp_img in (1, 4, 5, 7, 8)
  percent_250m_img_band6 <= 5.36
then
  A_STAGE = 13.1 - 0.0562 conus_reflectance_2002097_img_band4
            + 0.0522 conus_reflectance_2002097_img_band1
            - 0.0043 us_ppt14_annual_img
            - 0.014 conus_reflectance_2002097_img_band7
            + 0.0051 conus_ndvi_2002225_img
            + 0.0041 conus_reflectance_2002097_img_band2
            + 0.006 conus_reflectance_2002097_img_band6
            + 0.0024 conus_ndvi_2002321_img
            - 0.005 conus_reflectance_2002321_img_band6
            + 0.011 conus_modis32_2001097_albers_img_band1
            + 0.005 conus_reflectance_2002321_img_band7
            - 0.012 conus_modis32_2001097_albers_img_band4
            + 0.0084 us_ppt12_dec_img
            - 0.004 conus_modis32_2001097_albers_img_band6
            + 0.002 conus_ndvi_2002097_img
            + 0.003 conus_modis32_2001097_albers_img_band7
            - 2 * percent_250m_img_band6 - 0.005 ustmin06_albers_img
            + 0.004 conus_modis32_2001097_albers_img_band2
    
```

Discrete Data

See5 Output - Decision Trees

```

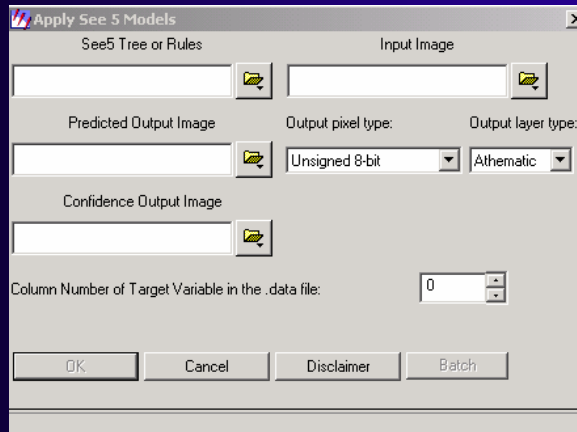
Decision tree
conus_reflectance_2002097_img_band3 > 1496:
... conus_mean_elevation_img <= 1622.195:
... conus_reflectance_2002097_img_band5 <= 5497: 2 (39/3)
... conus_reflectance_2002097_img_band5 > 5497: 4 (2)
... conus_mean_elevation_img > 1622.195:
... us_ppt08_avg_img <= 4328:
... us_ppt08_avg_img <= 2800:
... conus_mean_elevation_img <= 2333.972: 2 (29)
... conus_mean_elevation_img <= 2333.972:
... us_ppt14_annual_img <= 54807: 2 (37/8)
... us_ppt14_annual_img > 54807: 1 (4)
... us_ppt08_avg_img > 2800:
... conus_mean_elevation_img <= 2302.032:
... conus_nlcd_percent_mixedforest_img > 12: 2 (9/1)
... conus_mean_elevation_img <= 2302.032:
... conus_reflectance_2002225_img_band7 > 1788:
... conus_nlcd_percent_shrubland_img <= 46:
... conus_percent_slope_img > 3.220994: 2 (20)
... conus_percent_slope_img <= 3.220994:
... conus_percent_slope_img > 2.390486: 4 (2)
... conus_percent_slope_img <= 2.390486:
... conus_ndvi_2002097_img <= 801: 1 (2)
... conus_ndvi_2002097_img > 801: 2 (4)
... conus_nlcd_percent_shrubland_img > 46: [S1]
... conus_reflectance_2002225_img_band7 <= 1788:
... 140_trasp_img <= 6:
... 148_trasp_img > 4: 2 (5/1)
... 140_trasp_img <= 4: [S2]
... 148_trasp_img > 4:
... us_ppt10_oct_img > 4049:
... 148_trasp_img > 91: 1 (9/2)
... 140_trasp_img <= 91: [S3]
... us_ppt10_oct_img <= 4849: [S4]
... conus_mean_elevation_img > 2382.833:
... us_ppt02_feb_img <= 3751:
... conus_reflectance_2002321_img_band5 > 1484: 2 (70/13)
... conus_reflectance_2002321_img_band5 <= 1484:
... conus_ndvi_2002097_img <= 1576: 1 (2)
... conus_ndvi_2002097_img > 1576: 4 (3)
... us_ppt02_feb_img > 3751:
    
```

Continuous Data

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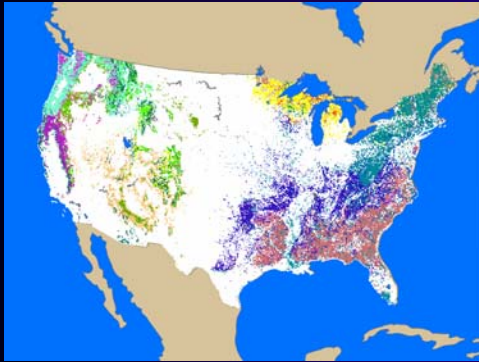
Spatial Representation of Model

Apply See5 Models

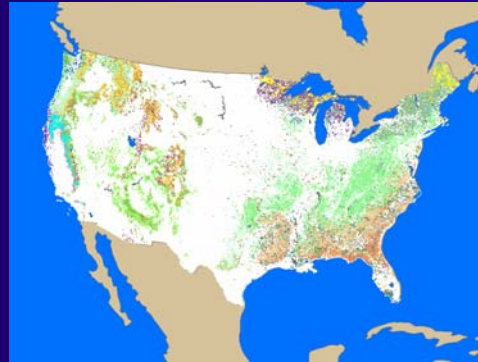


This ERDAS Imagine program applies the See5 models creating two image products: 1) a predicted output image with values representing the variables being modeled, and 2) a confidence output image.

Spatial Products – Continental U.S.



Forest Types (141)
50 % Accuracy



Forest Type Groups (28)
69 % Accuracy

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Spatial Products – Alaska



Forest Type
67% Accuracy



Forest Type Groups
78% Accuracy

Accuracy Assessment

Producer's	User's	Forest Type Group Class Name	# of AA Plots
65%	52%	White/Red/Jack Pine Group	161
73%	69%	Spruce/Fir Group	284
73%	73%	Longleaf/Slash Pine Group	122
69%	74%	Loblolly/Shortleaf Pine Group	663
88%	91%	Pinyon/Juniper Group	697
66%	74%	Douglas-fir Group	795
65%	65%	Ponderosa Pine Group	450
100%	25%	Western White Pine Group	4
62%	64%	Fir/Spruce/Mountain Hemlock Group	518
58%	53%	Lodgepole Pine Group	253
69%	52%	Hemlock/Sitka Spruce Group	128
13%	3%	Western Larch Group	30
50%	60%	Redwood Group	5
53%	27%	Other Western Softwood Group	64
59%	69%	California Mixed Conifer Group	194
0%	0%	Exotic Softwoods Group	6
61%	38%	Oak/Pine Group	367
75%	82%	Oak/Hickory Group	1471
51%	59%	Oak/Gum/Cypress Group	222
60%	40%	Elm/Ash/Cottonwood Group	200
65%	71%	Maple/Beech/Birch Group	663
71%	67%	Aspen/Birch Group	496
56%	41%	Alder/Maple Group	54
72%	61%	Western Oak Group	226
21%	21%	Tanoak/Laurel Group	19
55%	43%	Other Western Hardwoods Group	51
0%	0%	Tropical Hardwoods Group	2
0%	0%	Exotic Hardwoods Group	1

Discussion Items : Accuracy vs. Utility

- Appropriate uses ...
 - ◆ Scale
 - ◆ Applications
- Use of map and plots together
- Map unit descriptors
- Forest types and type groups
 - ◆ “Right” classifications for risk mapping?
 - ◆ Necessary but not sufficient?
 - ◆ What structural characteristics are needed?

