

# Earth Science Data Records (ESDR) of Global Forest Cover Change

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# Project Background

- Funded primarily through NASA's MEASURES program
  - Significant LCLUC funding

- Multi-institute collaboration

- University of Maryland



- John Townshend (PI)

- Chengquan Huang, Saurabh Channan, Joe Sexton, Raghuram Narasimhan, Min Feng, Kuan Song, Xiaopeng Song, DoHyung Kim, Danxia Song, Paul Davis, Sam Goward, et al.

- NASA/GSFC and contractors

- Jeff Masek, Robert Wolfe, Eric Vermote,
- Feng Gao, Bin Tan, Greg Ederer, Jim Tucker, et al.



- South Dakota State University

- Matt Hansen, Peter Potapov



# Primary Goals

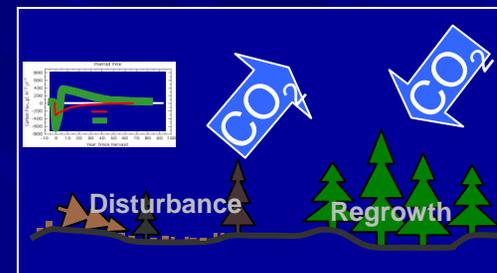
– Develop global FCC products to support

■ Modeling

- Climate
- Carbon: REDD
- Hydrology

■ Biodiversity and conservation

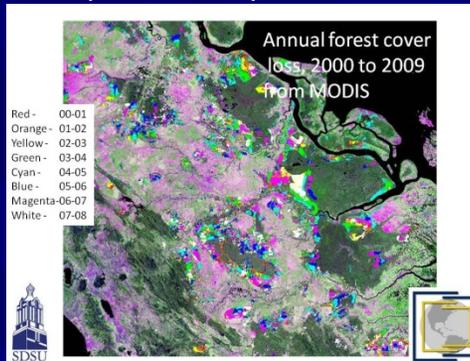
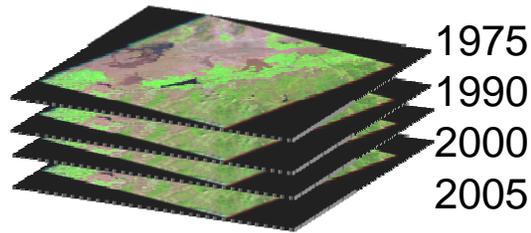
– Demonstrate routine global FCC monitoring capability



# Deliverables

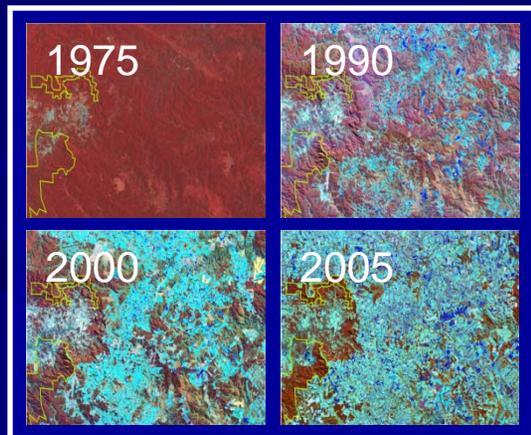
Forest cover change  
(MODIS) ESDR

“Global” Land Survey (GLS)

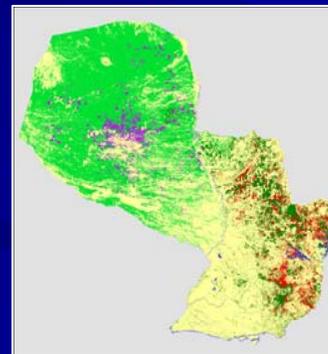


MODIS VCF  
(2000-2005)

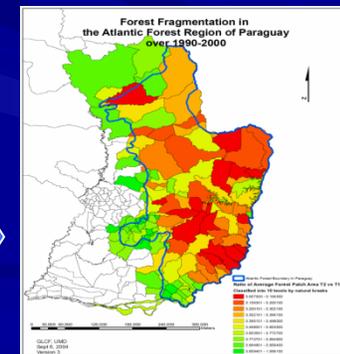
Landsat-MODIS  
consistency



Surface reflectance ESDR



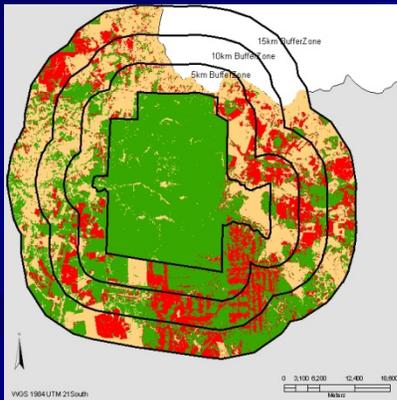
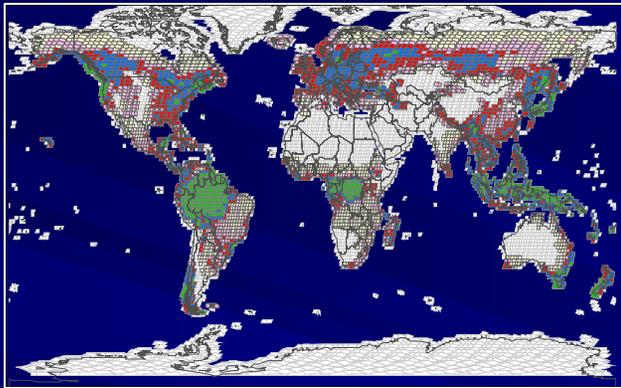
Forest cover change  
(Landsat) ESDR



Fragmentation & change  
ESDR

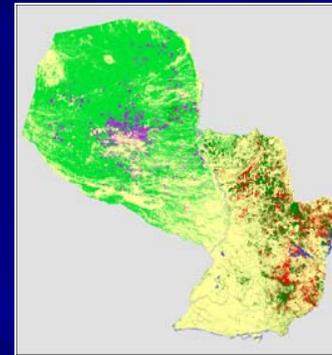
# Subsets and Aggregation

Global Products

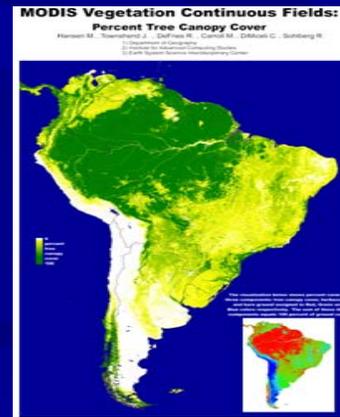


Complete subset for all WDPA protected areas

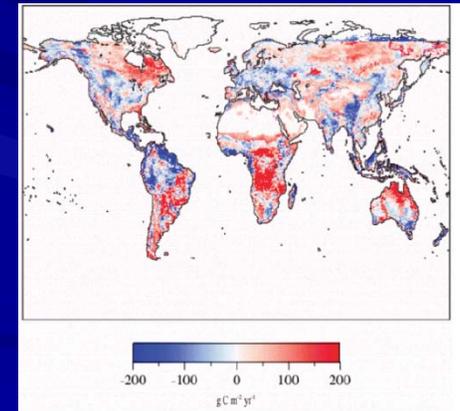
Landsat/MODIS resolution



Aggregation

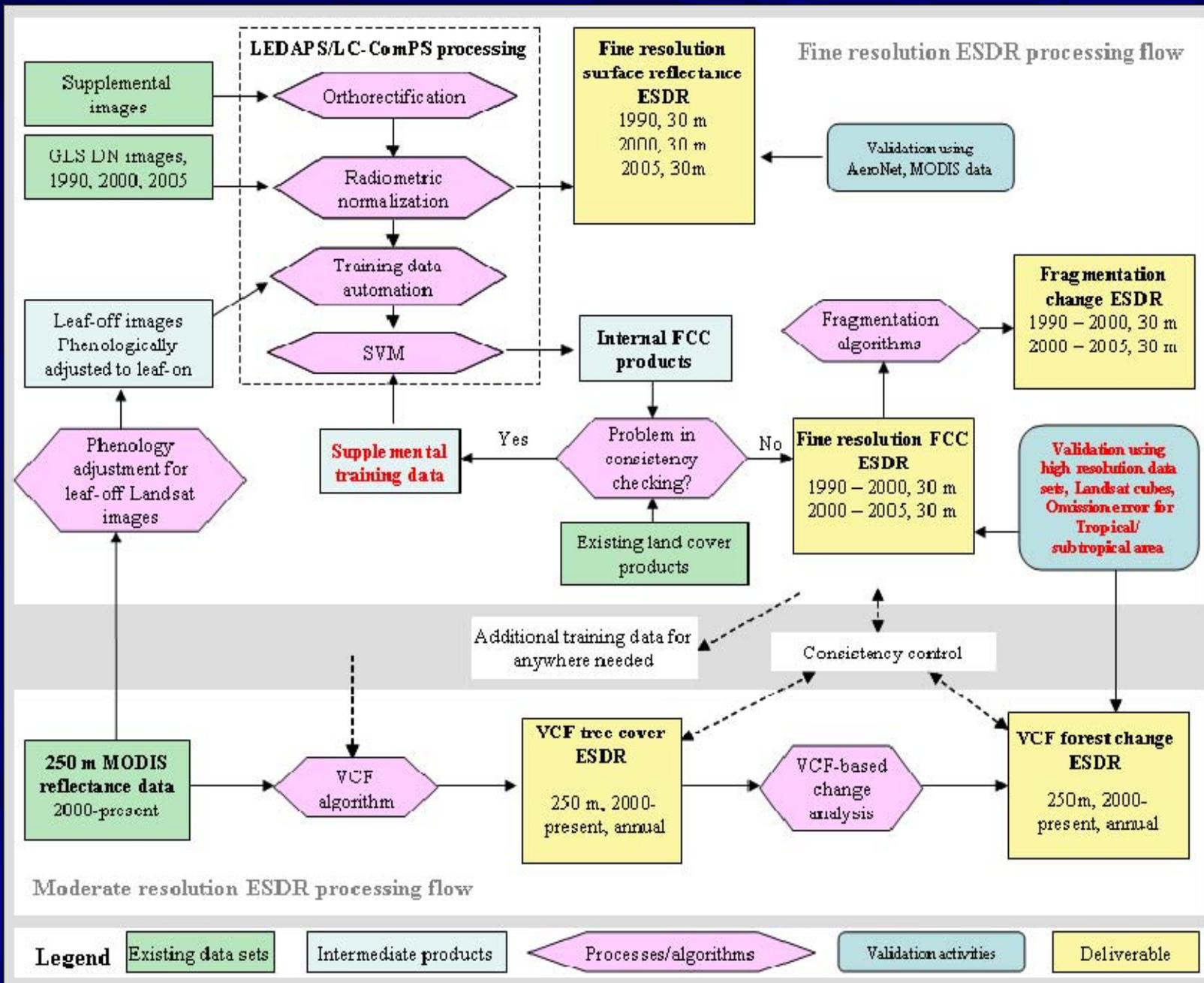


250 m – 5 km  
(MODIS-AVHRR)



5 km – 1 degree  
(model grids)

# GFCC-ESDR Processing Flow

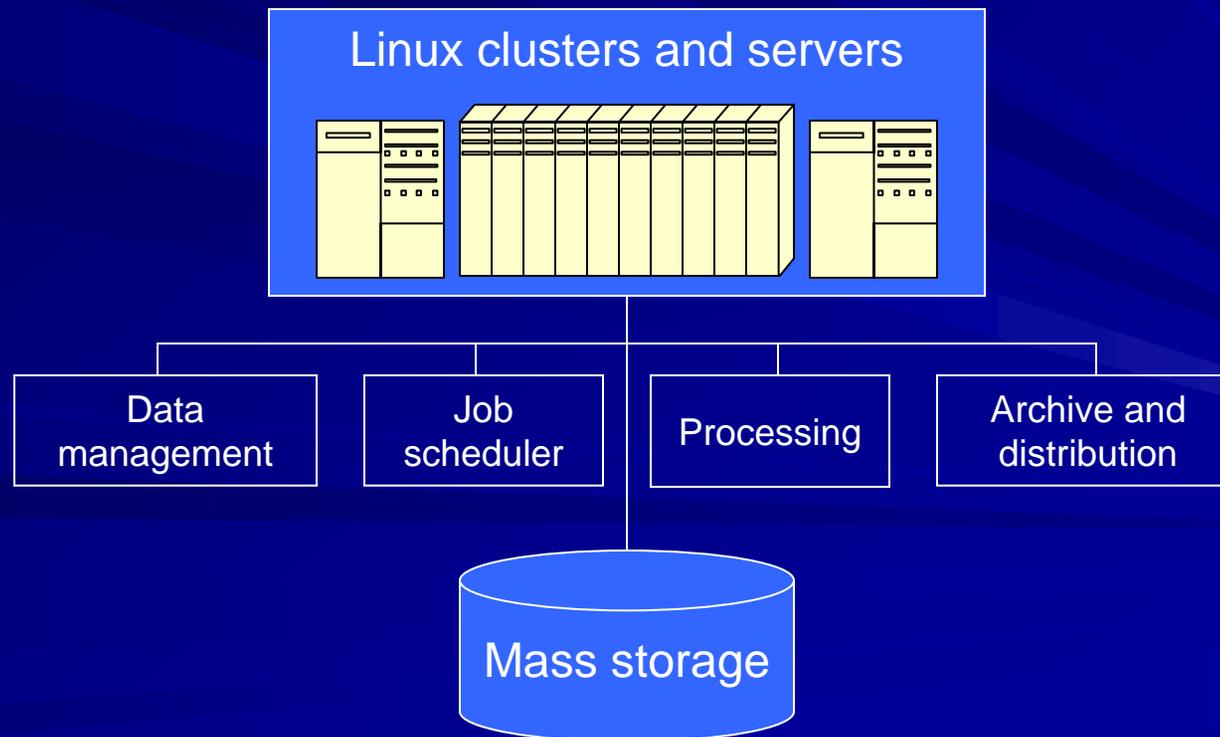


# Key Requirements of the Approach

- Mass processing capability
  - ~9000 images x 4 epochs
- Radiometric adjustment/atmospheric correction
  - Geometry mostly satisfactory
- Automated change mapping
- Consistency and quality checking

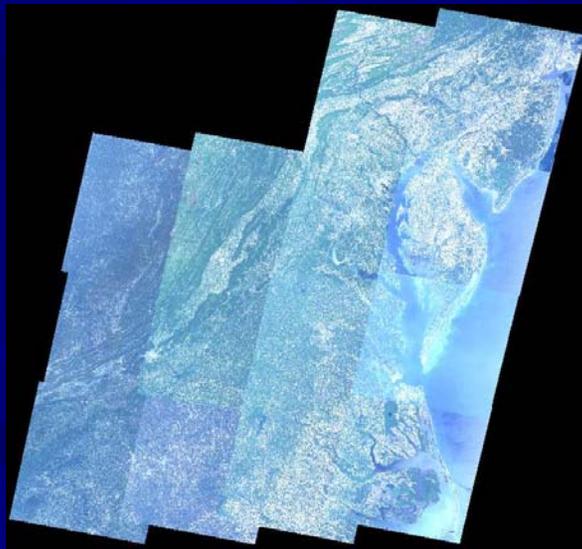
# Mass Processing Capability

- LEDAPS (Landsat Ecosystem Disturbance Adaptive Processing System)
  - An adaption of MODAPS by NASA/GSFC (Masek)
  - Have been used to process a few thousands Landsat images
  - UMD enhancement
    - **~9,000 Landsat images x 4 epochs (10 times more than can be handled by current LEDAPS)**



# Atmospheric Adjustment

- Based on the MODIS 6S approach (Masek et al. 2006)
  - Fully automated
  - Validated at many locations



Before adjustment



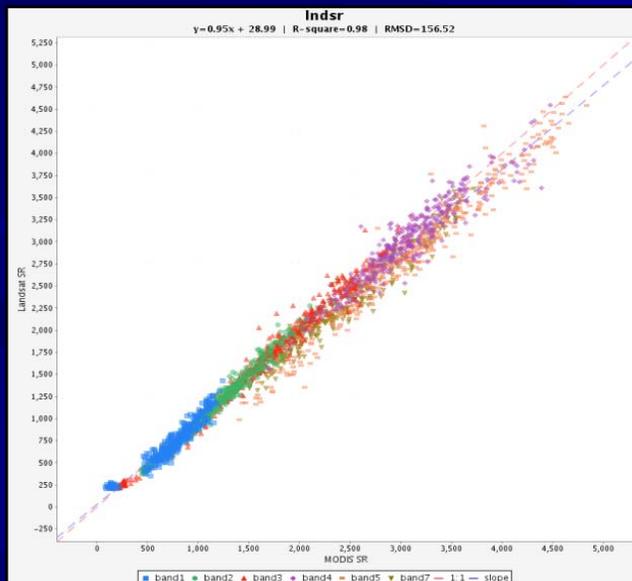
After adjustment

# Landsat SR Mosaic for North America

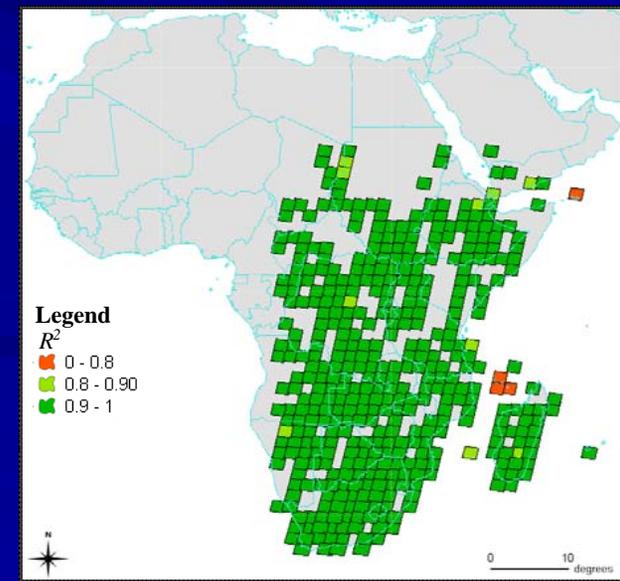


# Landsat-MODIS SR Comparison

- Every Landsat SR checked against MODIS SR
  - MODIS SR validated extensively
  - Landsat 7 SR  $\approx$  MODIS SR (daily)
    - Same acquisition date,  $\sim$ 30 minutes apart (Landsat 7 images)
    - Aggregate Landsat to MODIS resolution
    - Use samples from homogeneous areas
  - Fully automated, will be applied to all Landsat 7 images
- Landsat 5 SR  $\approx$  MODIS SR (NBAR)

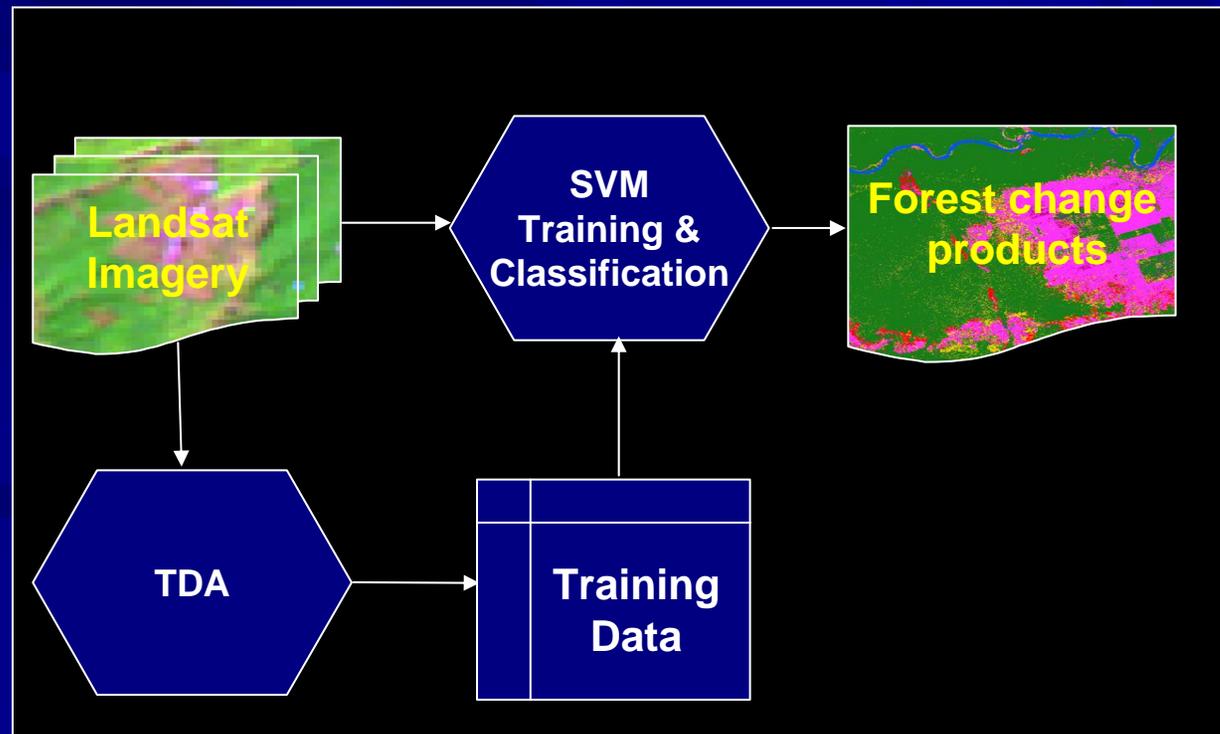


R square  
RMSD



# Automated Forest Change Mapping Method

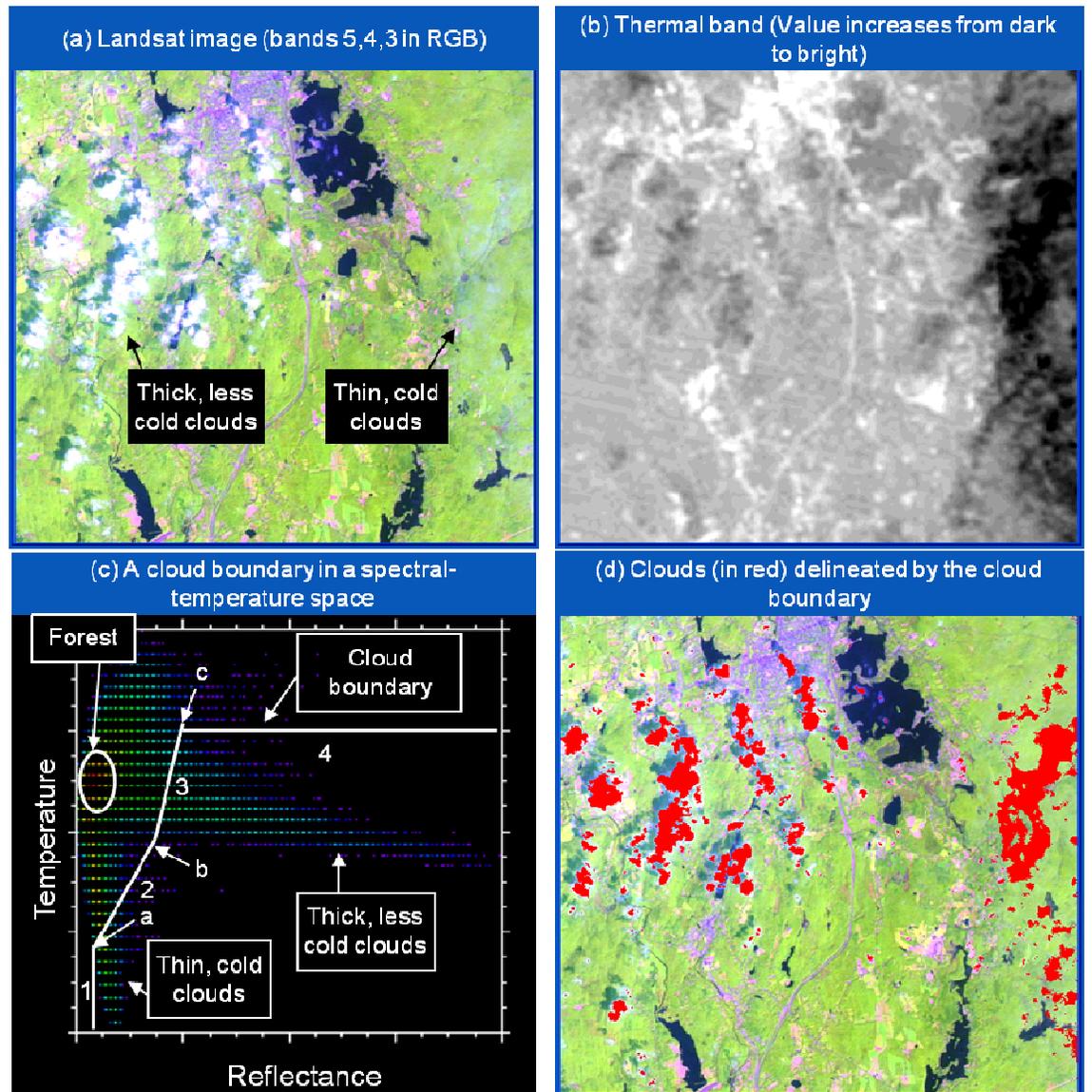
- SVM (support vector machines) more accurate than other classifiers (Huang et al., 2002; Pal & Mather, 2005, and others)
- Automation possible by TDA (training data automation) (Huang et al., 2008)



# Cloud and Shadow Masking

Fig. 1. Use of a cloud boundary in a spectral-temperature space to delineate cloud pixels. In the spectral-temperature space shown in (c), the color of each dot indicates the number of pixels plotted at that point, which increases from purple to blue, green, yellow, and red. The cloud boundary consists three segments, which are labeled from 1 to 3, and are defined and joined by points a, and b. See section 2.4 on discussions of the cloud boundary.

(Huang et al., in press)

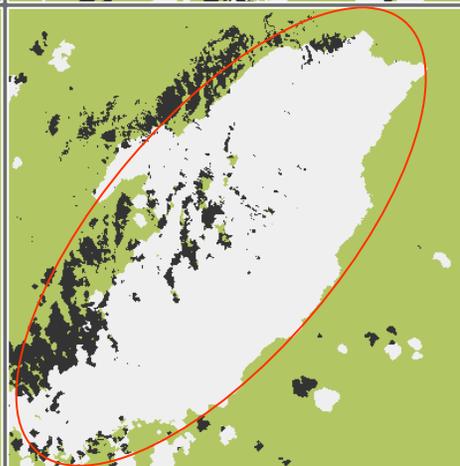
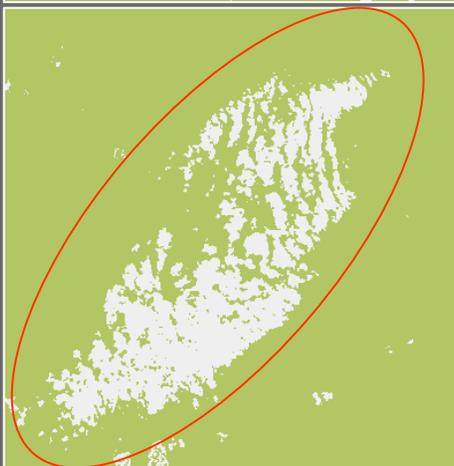
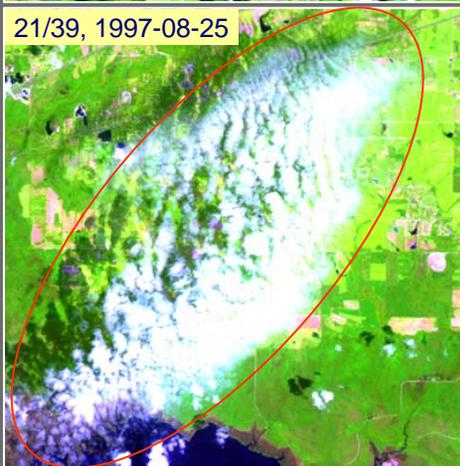
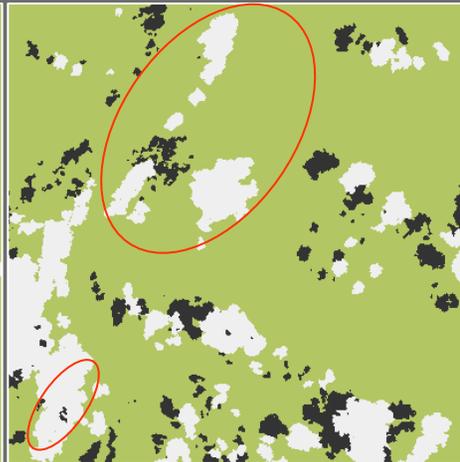
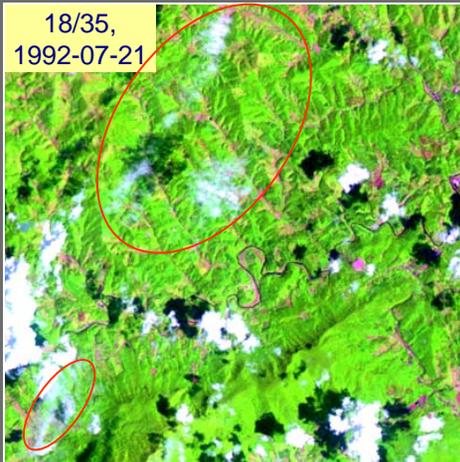


# Cloud and Shadow Masking

Landsat image, R, G, B =  
bands 5,4,3

cloud mask by the Luo et  
al. (2008) method

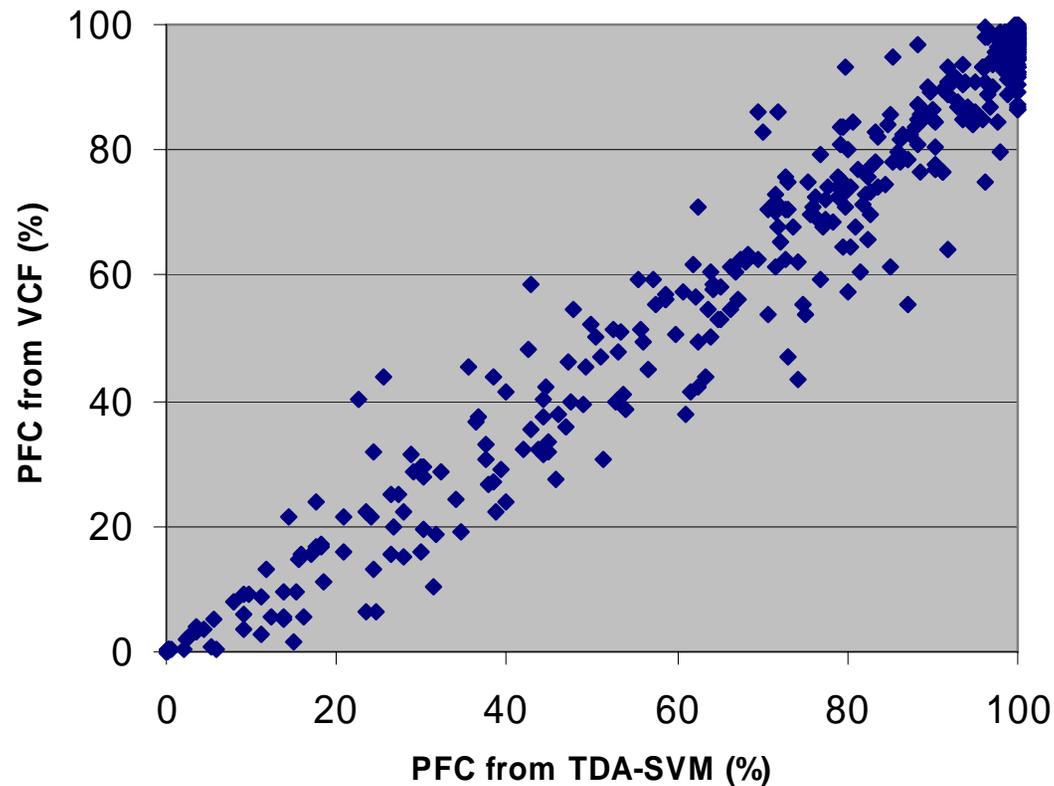
Cloud mask by this study  
(Huang et al., in press)



# FCC Products Assessment

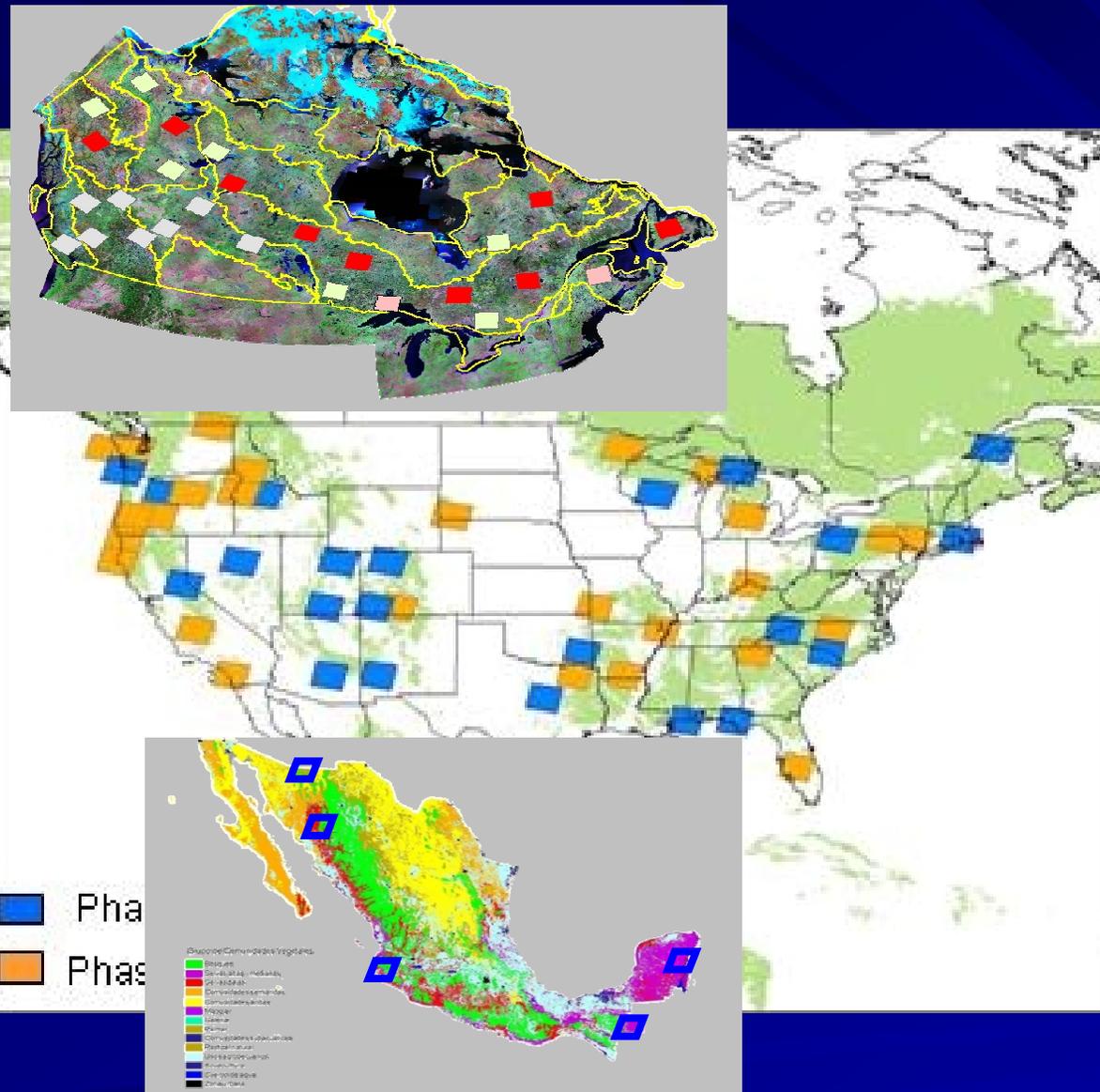
- Comparison with existing land cover products
  - Global & regional products
    - Rule out gross errors
  - NAFD dense time series change products
    - Highly reliable
    - Accuracy estimates
- Design-based accuracy assessment

# Use Existing LC Products to Rule Out Gross Errors



*Figure 5. Percent forest cover (PFC) within 10 km grid cells for the 2000 epoch calculated using a TDA-SVM derived FCC product and the MODIS VCF tree cover product. The near 1:1 relationship between the two sets of PFC values is an indicator of the high quality of the TDA-SVM derived FCC product.*

# Dense Time Series FCC Products



State level:

Utah, Mississippi, Alabama, Maryland, North Carolina

LVIS acquisition sites

# Accuracy Assessment

Random Sample (known inclusion probability)



Labeling Tool - Mozilla Firefox

File Edit View History Bookmarks Tools Help

http://159.226.111.3:57080/validation/~view.do?collect=5

Most Visited Customize Links Free Hotmail Windows Marketplace Windows Media Windows

Labeling Tool

Global Land Cover Facility  
Earth Science Data Interface

Login Register

Labeling Tool

Reset Extent

Satellite Street

Ready Sample: 116.375E, 40.020N 116.376E, 40.018N

1975	1990	2000	2005
09/19/1978(04,02,01) 09/19/19	09/06/1992(40,30,20) 09/06/19	06/30/1999(40,30,20) 06/30/19	09/05/2006(40,30,20) 09/05/20
p133r32_3m19780920 [04, 02, 01]	p123r032_5dx19920907 [40, 30, 20]	p123r032_7dx19990701 [40, 30, 20]	L7112B032_03220060906 [40, 30, 20]
31,19,21	83,26,29	118,75,87	63,108,91
landcover unknown	landcover forest	landcover forest	landcover urban

Land-cover for all epochs: forest crop grass urban water others mixed badimage unknown

Submit mfeng (Tuesday, December 15, 2009 5:52:21 PM) 237 / 651 Previous Next Go to... Shapefile

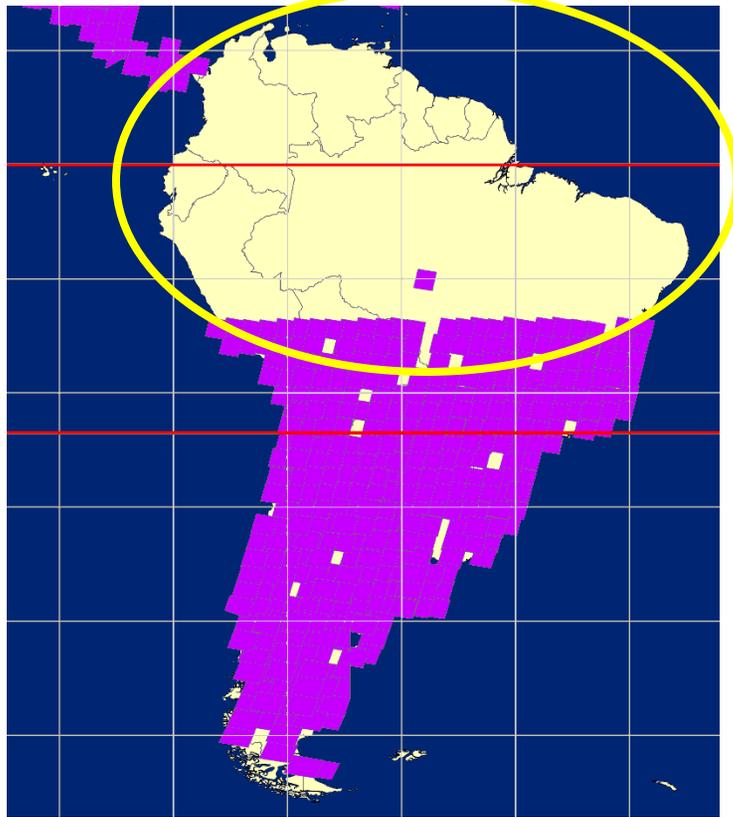
Done

# Summary

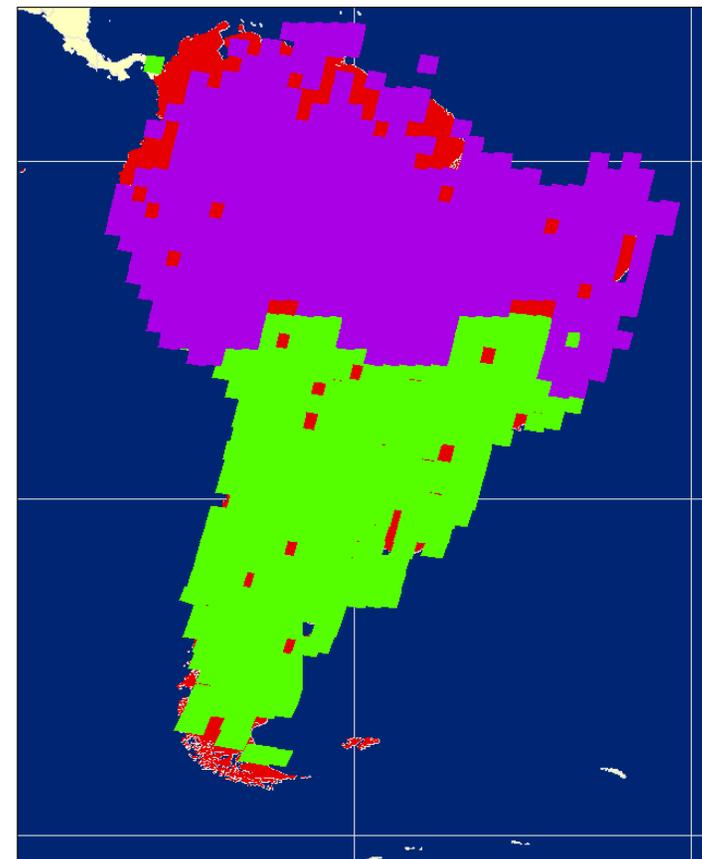
- Producing global forest cover change products will be achieved through
  - Fully automated algorithms
  - Systematic quality checking/verification
  - Mass processing capability
- Issues with the GLS Data Sets
  - Major data holes
  - Difficulty to calibrate the 1990 data set

# A Major Data Hole in the GLS 1975

No data for the north part of SA



INPE has the data to fill the gap



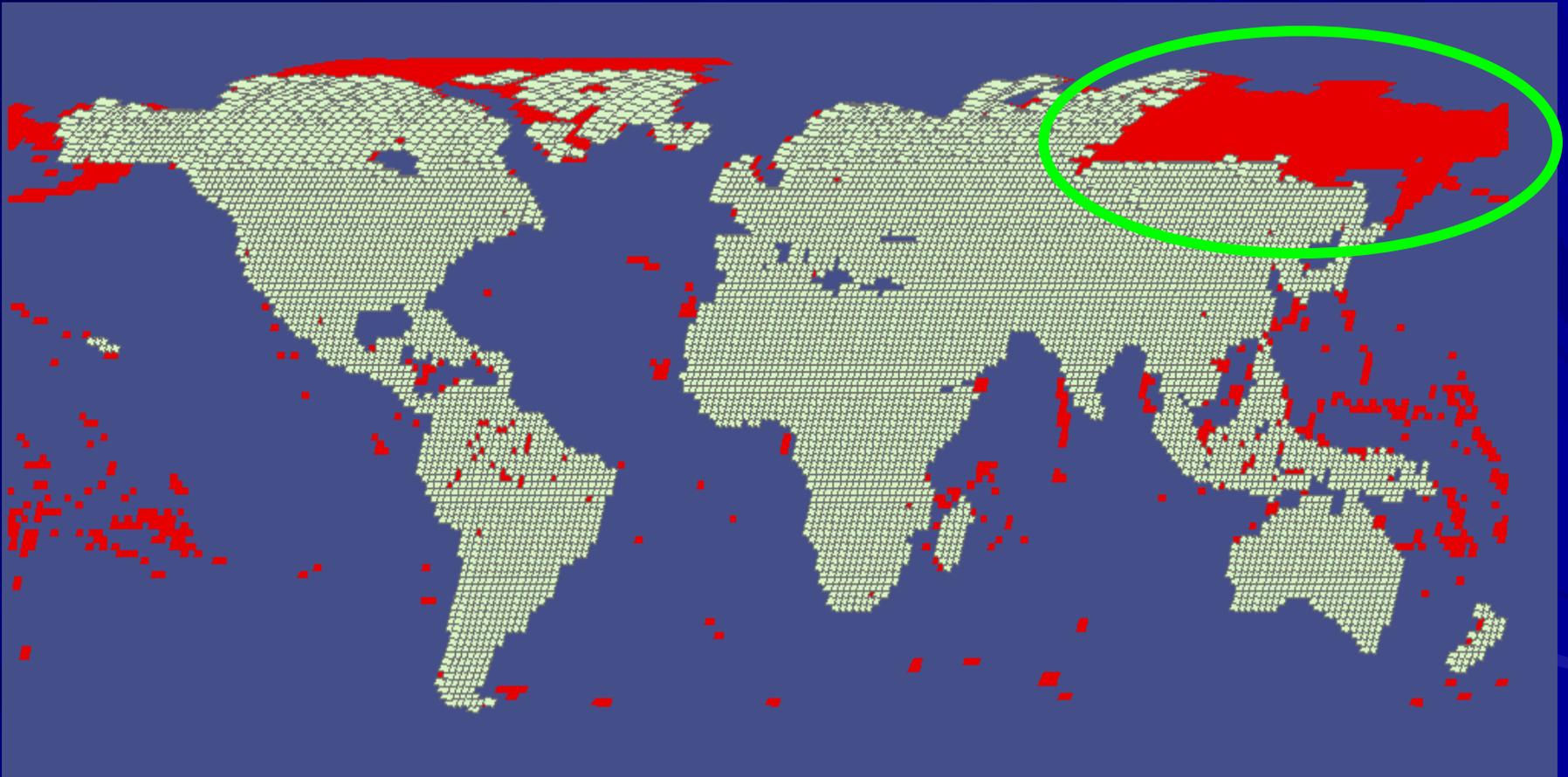
- Geocover MSS 1975
- Holes that can be filled by INPE
- Hole that cannot be filled by INPE

0 250 500 1,000 1,500 2,000 Miles

The GLCF: [www.landcover.org](http://www.landcover.org)

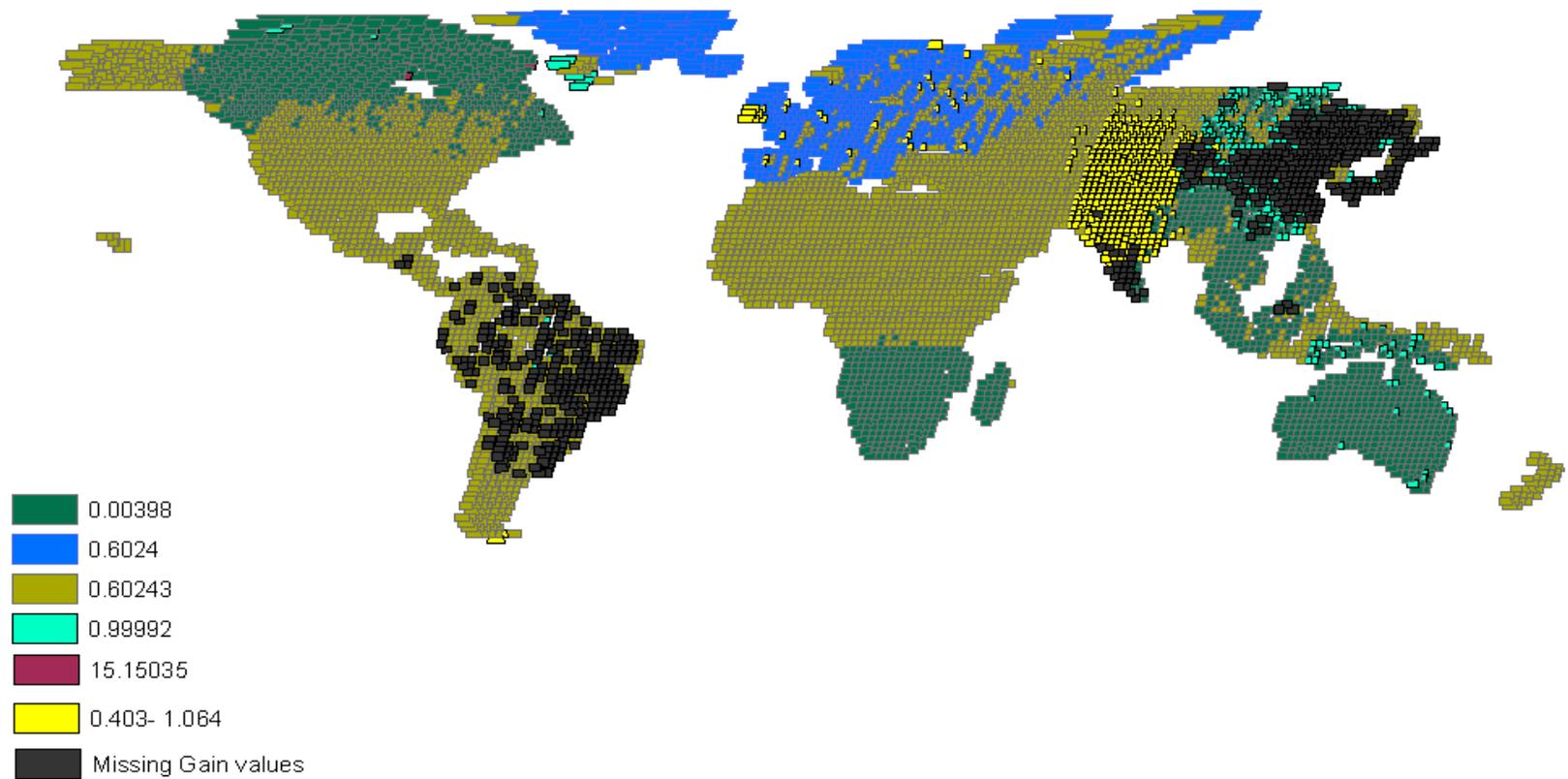
Source: ESRI 2000, GLCF 2009

# Eastern Siberia Not Covered in 1990



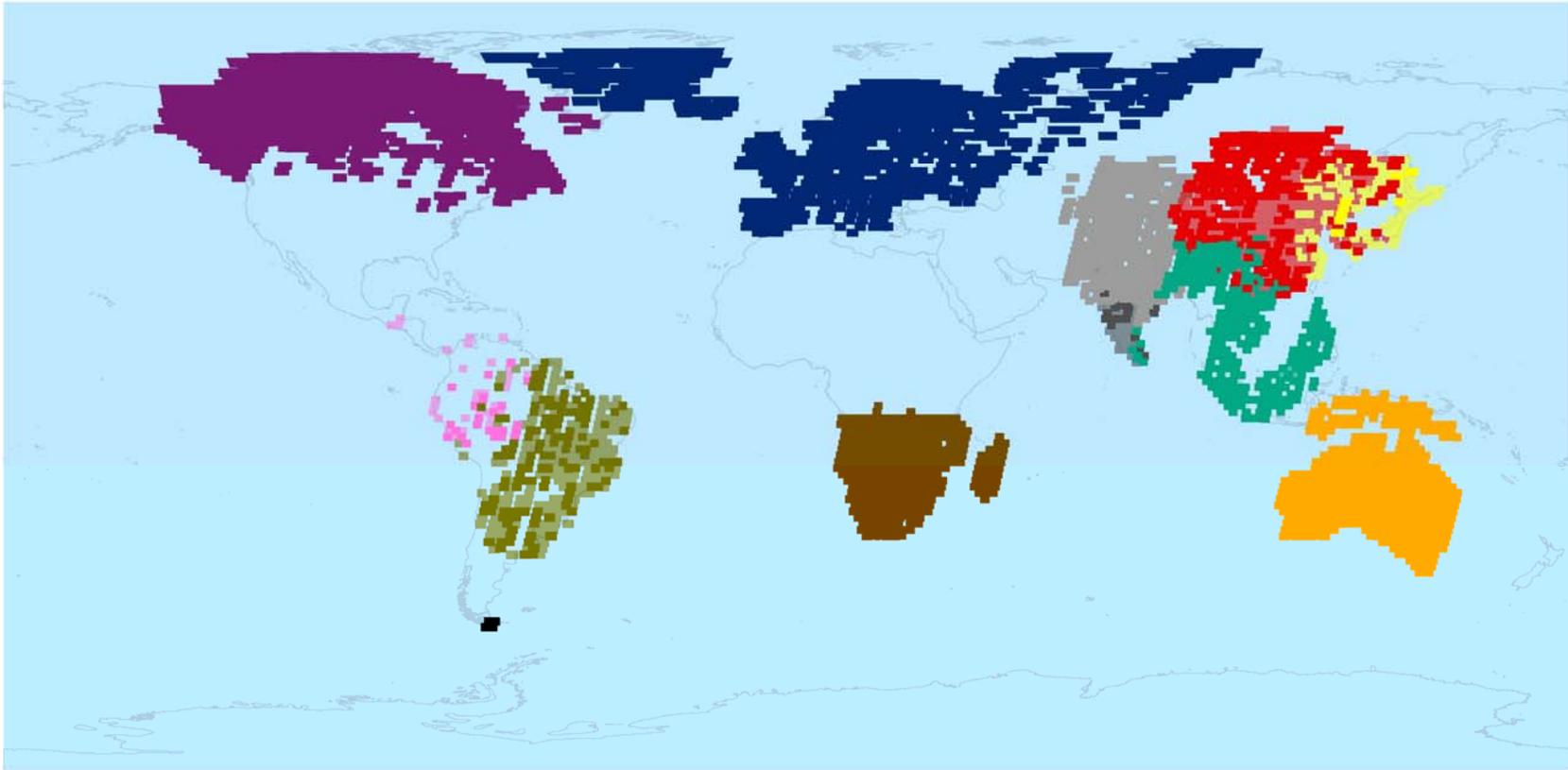
# GLS 1990 Calibration Challenge

GLS 1990 Band 1 Gain Values



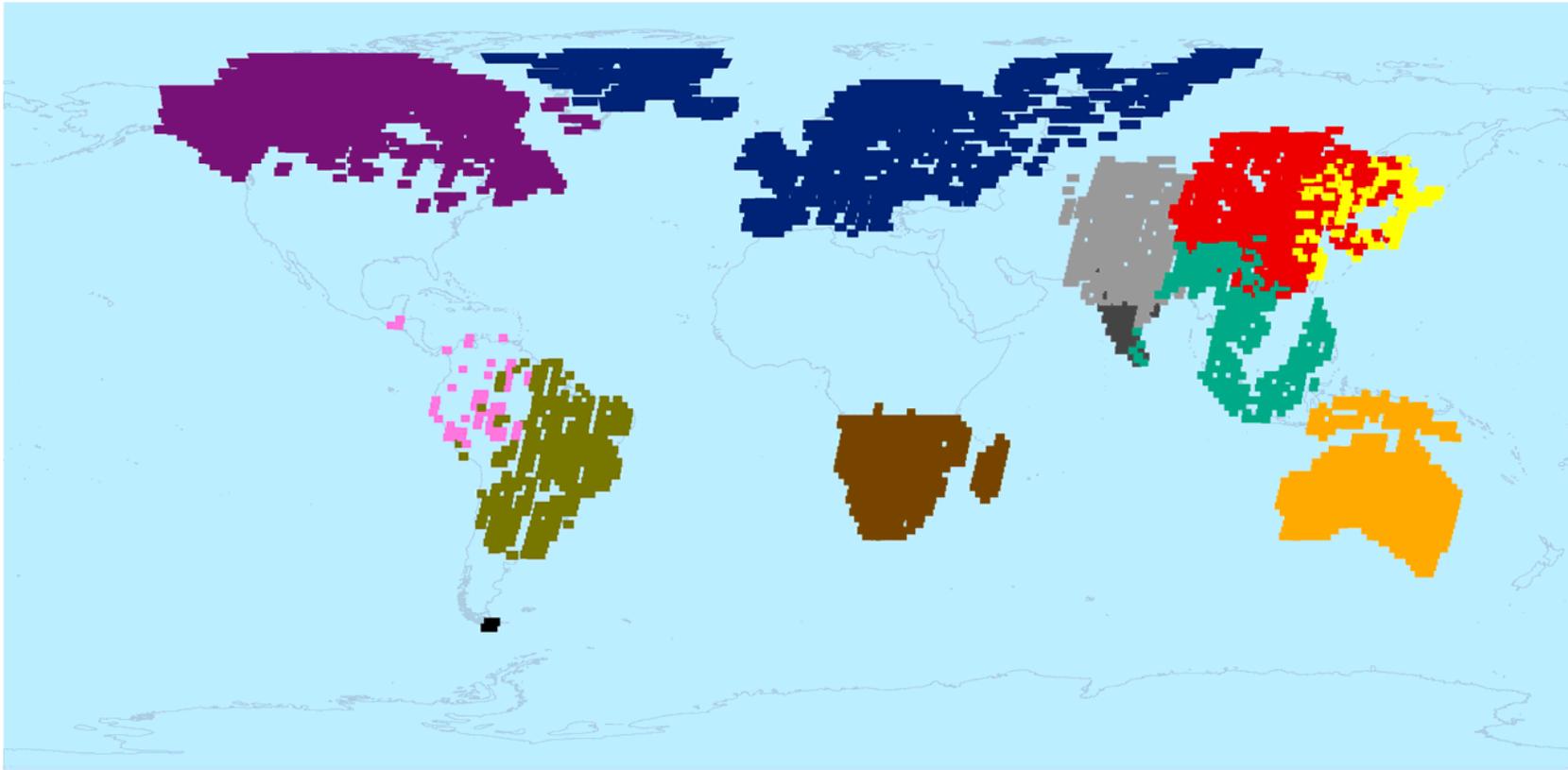
## GeoCover 1990 International GRS & GLS1990 Band1 Gain Values

*Similar spatial pattern for Bands 2,3,4,5 and 7*



 Australia (ACRES): 0.00398, 0.60243, 0.99992	 Ecuador (CLIRSEN): -9999, 0.60243	 Japan (RESTEC): -9999, 0.60243
 Brazil (INPE): -9999, 0.60243, 0.99992	 EOSAT: 1.05473, 1.05509	 Pakistan: 0.00398, 0.40258-1.06365
 Canada: 0.00398, 0.60243, 0.99992, 15.15035	 Europe (ESA): 0.6024, 0.60243, 0.7314	 South Africa (SAC): 0.00398, 0.60243
 China: -9999, 0.00398, 0.60243, 0.99992, 1.06365	 India (NRSA): -9999, 0.00398, 0.48733 -1.06365	 Thailand: -9999, 0.00398, 0.60243, 1.06365

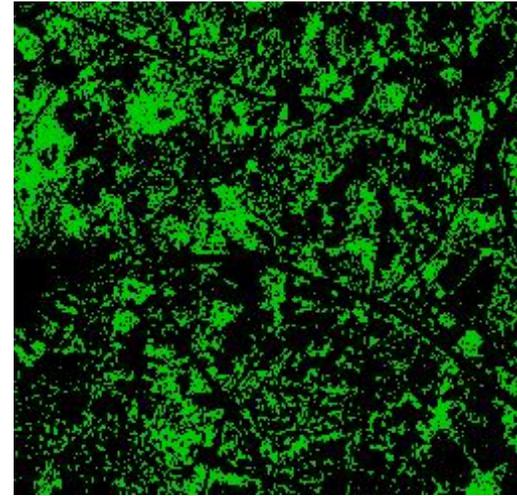
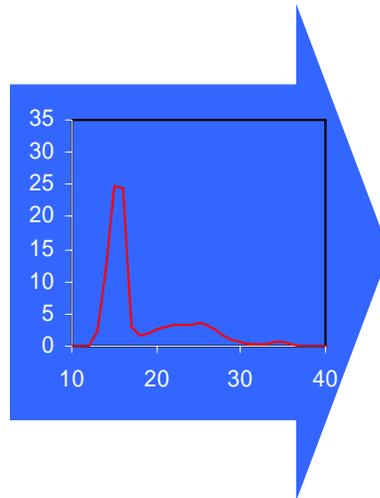
## GeoCover 1990 International GRS & GLS1990 Band6 Gain Values



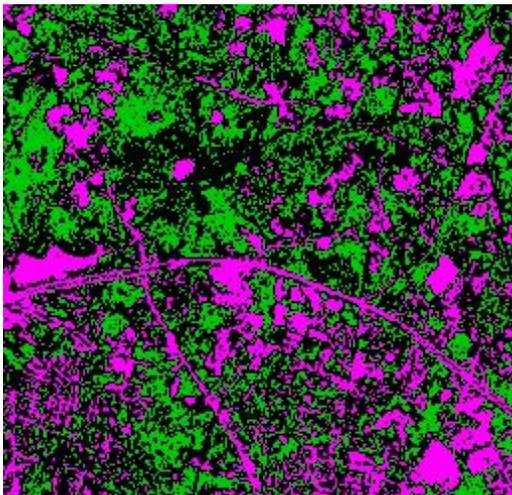
 Australia (ACRES): [-9999, 0.05515, 3.20107	 Ecuador (CLIRSEN): [-9999, 0.05515	 Japan (RESTEC): [-9999, 0.05515
 Brazil (INPE): [-9999, 0.05515, 3.20107	 EOSAT: 1.52431	 Pakistan: [-9999, 0.05515, 0.66433 - 1.53031
 Canada: [-9999, 0.05515, 1.52432, 3.20107	 Europe (ESA): [-9999, 0.05515, 0.0552	 South Africa (SAC): [-9999, 0.05515
 China: [-9999, 0.05515, 0.89048 - 1.35381, 3.20107	 India (NRSA): [-9999, 0.89048, 0.92357- 1.10559	 Thailand: [-9999, 0.05515, 1.03389

# Backup Slides

# Training Data Automation



**Forest  
Index**



**Threshold FI**



(Huang et al. 2008)

# SVM More Accurate

## ■ Classify using optimal class boundary

Method	single-scene test	multiple-scene test
MLC	85%	60%-70%
Bayesian	86%	60%-70%
Contextual MLC	88%	60%-70%
DT	90%	80%
<b>SVM</b>	<b>92%</b>	<b>85%</b>

## ■ TDA-SVM tested in many places

- Huang, C., Song, K., Kim, S., Townshend, J., Davis, P., Masek, J. & Goward, S.N. (2008). Use of a Dark Object Concept and Support Vector Machines to Automate Forest Cover Change Analysis. *Remote Sensing of Environment*, 112, 970-985.

# Coordination with Related Projects

## ■ NASA LCLUC

- Chandra Giri, USGS EROS
  - Tropical Mangrove Forests: Global Distributions And Dynamics (1990-2005)
- Matthew Hansen, South Dakota State University
  - Producing Composite Imagery And Forest Cover And Change Characterizations For The Humid Tropics - A Contribution To The MDGLS Activity
- Mutlu Ozdogan, University of Wisconsin
  - Land-Use And Land-Cover Changes In Temperate Forests Of European Russia: The Past, The Current, And The Future
- David Skole, Michigan State University
  - Enhancing Global Observations And Information On Tropical Forest Change Using Landsat Global Data
- John Townshend, University of Maryland
  - Three Decades Of Forest Cover Change In The Americas Evaluated Using The Geocover And MDGLS Data Sets
- Xiangming Xiao, University of New Hampshire
  - Developing Land Cover Classification Products In Monsoon Asia Over The Period Of 2004-2007 Through Integration Of Landsat And ALOS/PALSAR Images

# Coordination with Related Projects

## ■ Other NASA Projects

- Jeff Masek, NASA/GSFC

- Recent North American Forest Dynamics via Integration of ASTER, MODIS, and Landsat Reflectance Data

- Jeff Masek, NASA/GSFC

- Landsat Ecosystem Disturbance Adaptive Processing System (LEDAPS)

# Stay tuned ...

- Five-year project started this summer
- Current progress
  - System setup
  - Data ingestion
  - Prototype studies