

## **K&C Phase 4 – Status report**

# **Mapping Wetlands, Surface Structural Attributes, and Boreal Freeze/Thaw at Regional Scales with JAXA SAR Datasets**

*Kyle C McDonald*

*Department of Earth and Atmospheric Sciences  
City College of New York  
City University of New York  
New York, NY, USA*

# Mapping Wetlands, Surface Structural Attributes, and Boreal Freeze/Thaw at Regional Scales with JAXA SAR Datasets

## ***Collaborators:***

**Ana Carnaval, Maria Tzortziou (CCNY)**

**Laura Chavez (Michigan Tech)**

**Joel Cracraft (American Museum of Natural History)**

**Bruce Chapman, Erika Podest (NASA JPL)**

**Mahta Moghaddam (USC)**

**Lisa Rebelo (IWMI), Thiago Silva (UNESP)**

## ***CCNY Contributors:***

***Nicholas Steiner, Marzi Azarderakhsh***

***Kat Jensen, Brian Lamb, Aaron Davit,***

***Jessica Rosenqvist, Jessika Alvarez, Rehnuma Islam***

## Project Objectives

### Wetlands

Extension of on-going wetlands work

- 30-year record: JERS, PALSAR, PALSAR2

Alaska, Canada

South America

Great Lakes

Africa

New Wetlands Regions

Chesapeake Bay – Estuarine carbon, Land-ocean exchange

New England (Maine) – Land-ocean carbon exchange

River Deltas – Carbon & Conservation

Boreal peatlands - Canada

## Project Objectives

### Biodiversity and Structure

Amazon and Brazilian Atlantic Coastal Forest

- Biodiversity
- Biome classification, landcover structure

### Freeze/Thaw

Alaska, Canada, Northern Europe

- Thermokarst studies
- Permafrost carbon cycle

**Objective: Development of a data set to facilitate global and regional studies of the role of inundated wetlands in studies of climate biogeochemistry, hydrology, and biodiversity.**

### **I. Regional inundated wetlands data sets from Synthetic Aperture Radar (SAR)**

- Spatial coverage: Major global wetland regions, 100m resolution
  - Temporal coverage: 1-2 year time series at 17-to-46 day intervals during 2006-2009 † ‡
  - Retrospective 1990's-era from archived JERS data covering Alaska, Canada, Amazon
1. Wetland extent (maximum inundatable area, including water bodies).
  2. Wetland vegetation type (Non-vegetated, Herbaceous, Shrub, Woodland, Forest).
  3. Inundation state (Flooded, Non-flooded; 17-46 day intervals) ‡
  4. Annual inundation duration

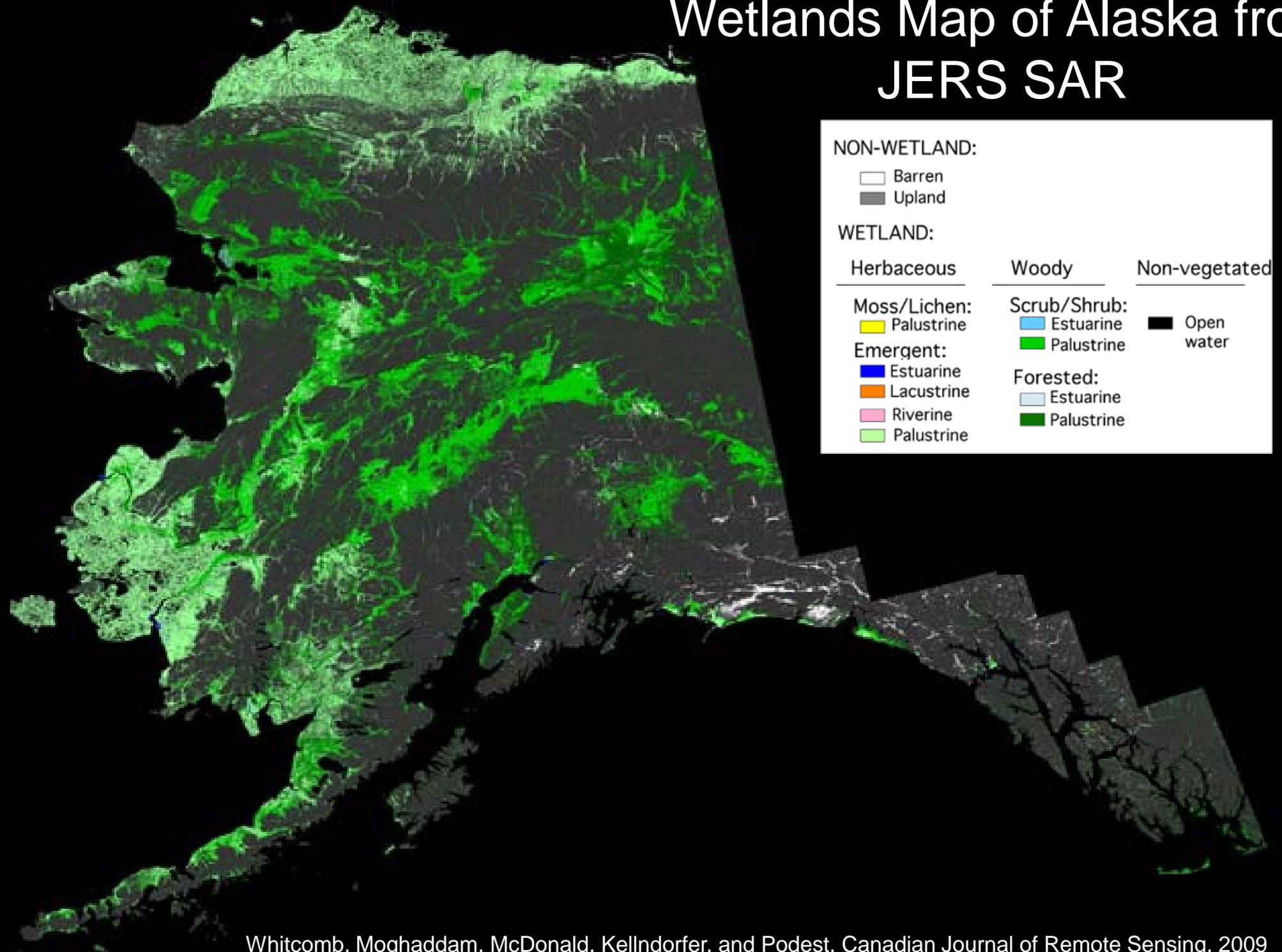
### **II. Global monthly inundation data sets derived from multiple satellite data sources**

- Spatial coverage: Global, 25 km resolution
  - Temporal coverage: Monthly monitoring with annual summaries, 1992-2013 †
1. Globally gridded monthly inundated area fraction
  2. Globally gridded annual inundation duration

† The domain of the 25-km and 100-m data sets excludes permanently frozen regions and seasonally frozen landscapes during the frozen season, although data from frozen seasons is used to improve classification accuracy.

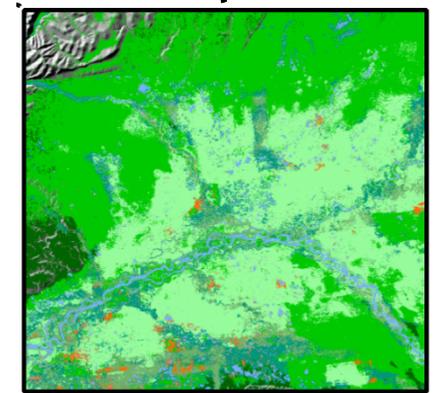
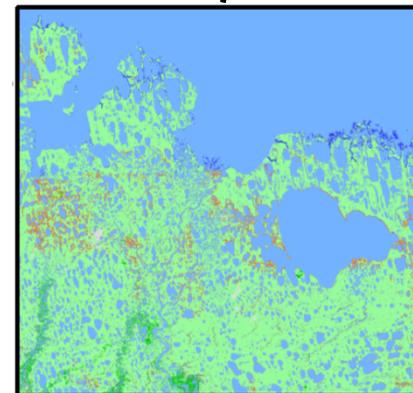
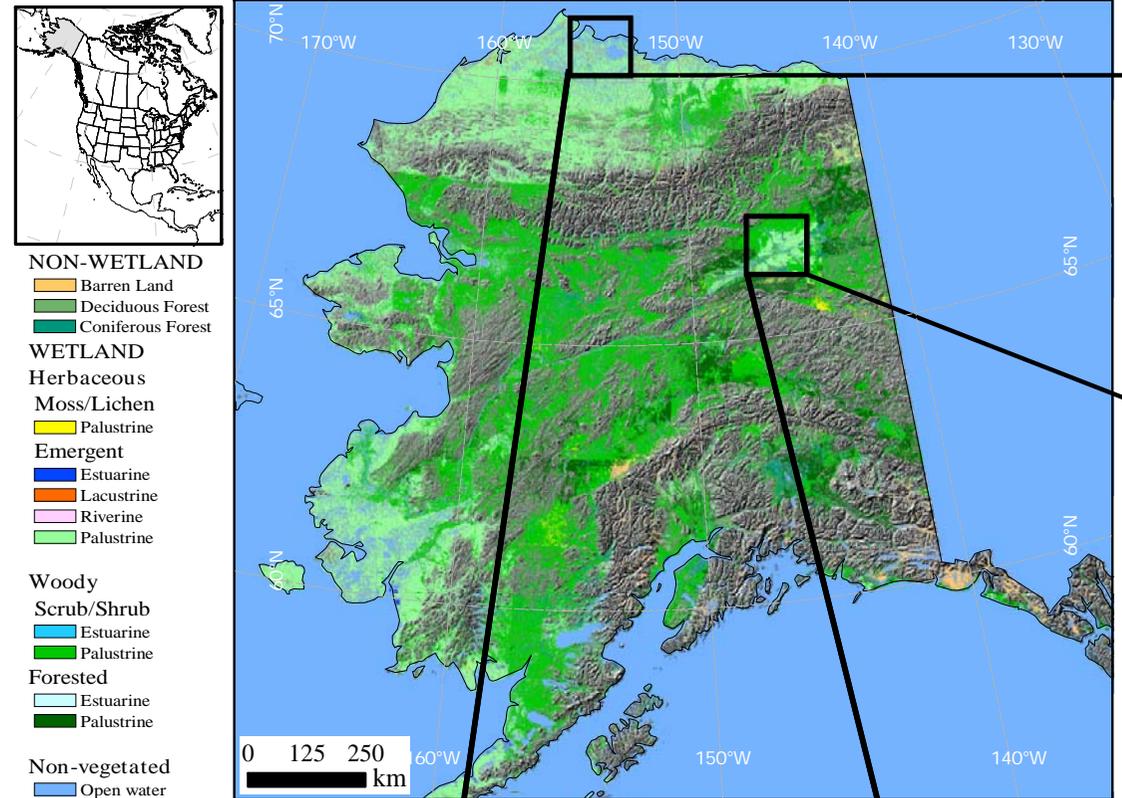
‡ PALSAR ScanSAR mode has 46-day exact repeat orbit with 17-day sub-cycles.

# Wetlands Map of Alaska from JERS SAR



## An updated map of wetlands in Alaska from PALSAR data

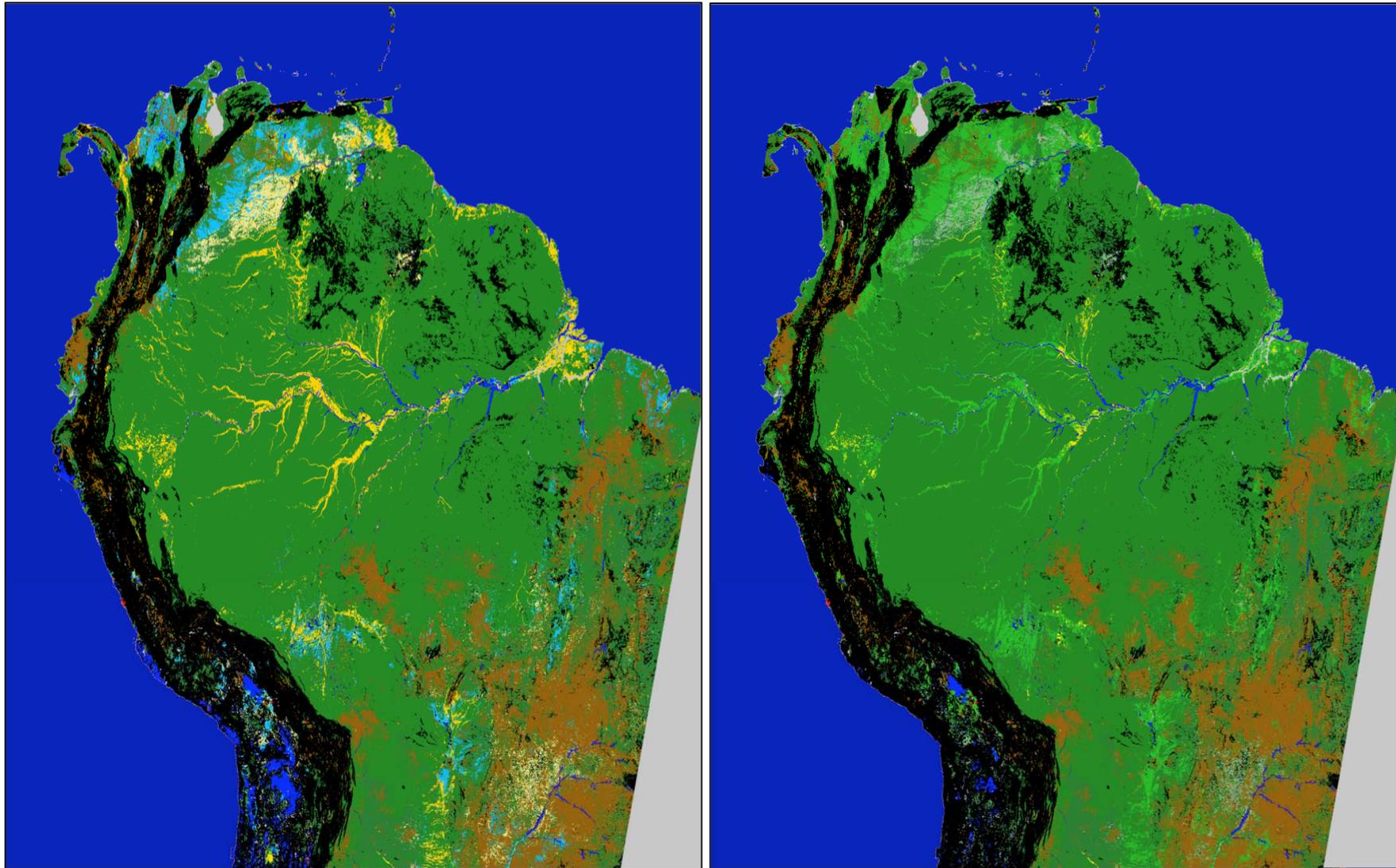
- 50 m spatial resolution map, using dual-polarization 2007 data
- Trained using National Wetland Inventory (NWI) data.
- Accuracy of geomorphology-vegetation classes 84 %
- Accuracy of wetland / non-wetland discrimination 94 %
- 0.59 million km<sup>2</sup> of wetlands mapped



ALOS

K&C Initiative  
An international science collaboration led by JAXA

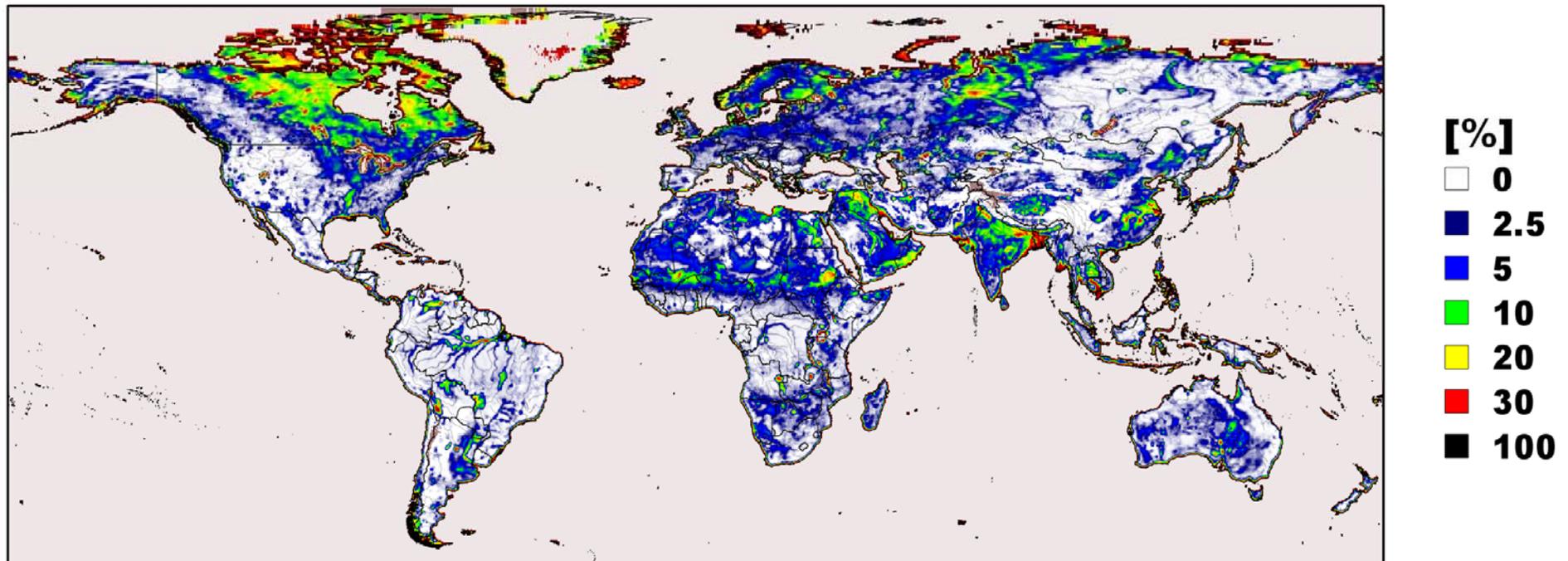
## Seasonal Maximum/Minimum Inundation from ALOS ScanSAR



Chapman et al, 2015

# Surface Water Microwave Product Series (SWAMPS)

**20+ Years of Daily Global Mappings of Inundation Extent  
at ~25 km Resolution for the Period 1992-2013**



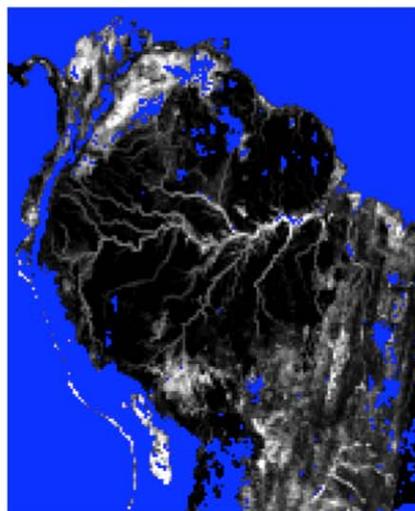
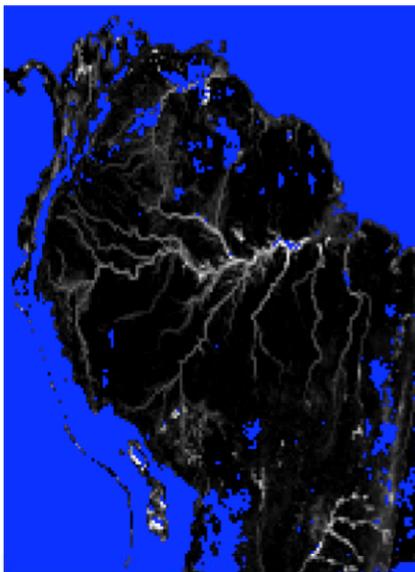
# Cross-Product Verification

Minimum

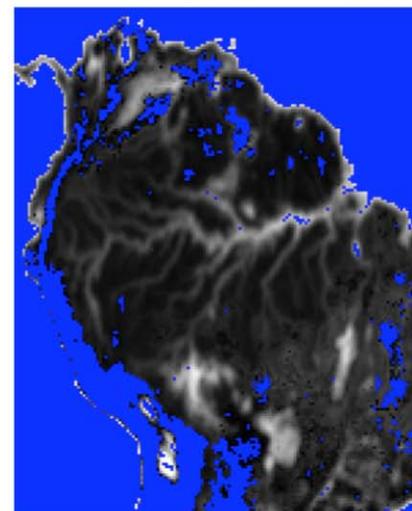
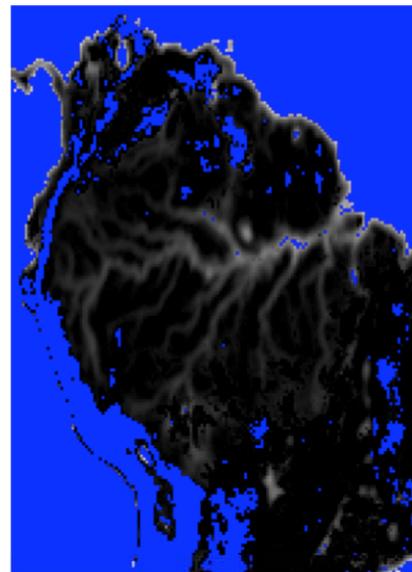
**Comparison of fractional  
open surface water area  
from coarse resolution  
sensors with ALOS  
SCANSAR open surface  
water fraction and  
inundated vegetation  
fraction**

Maximum

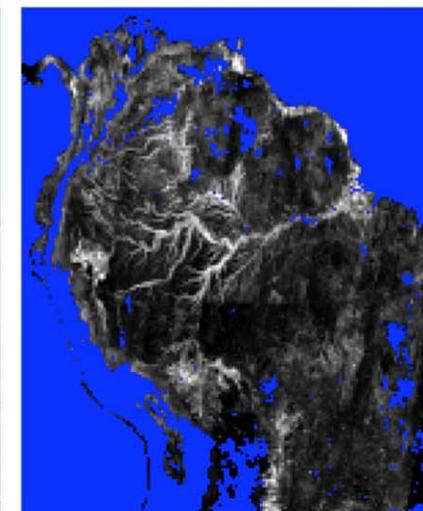
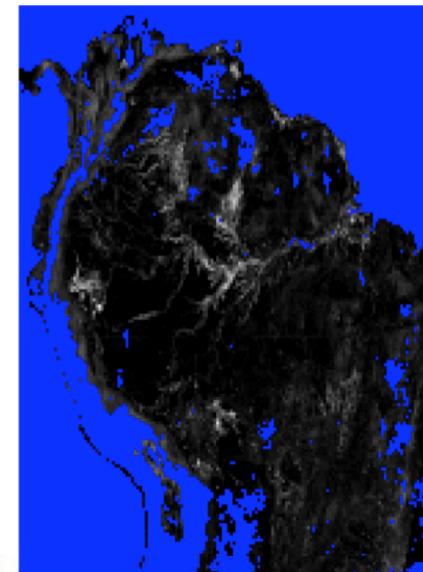
Fractional open surface  
water area

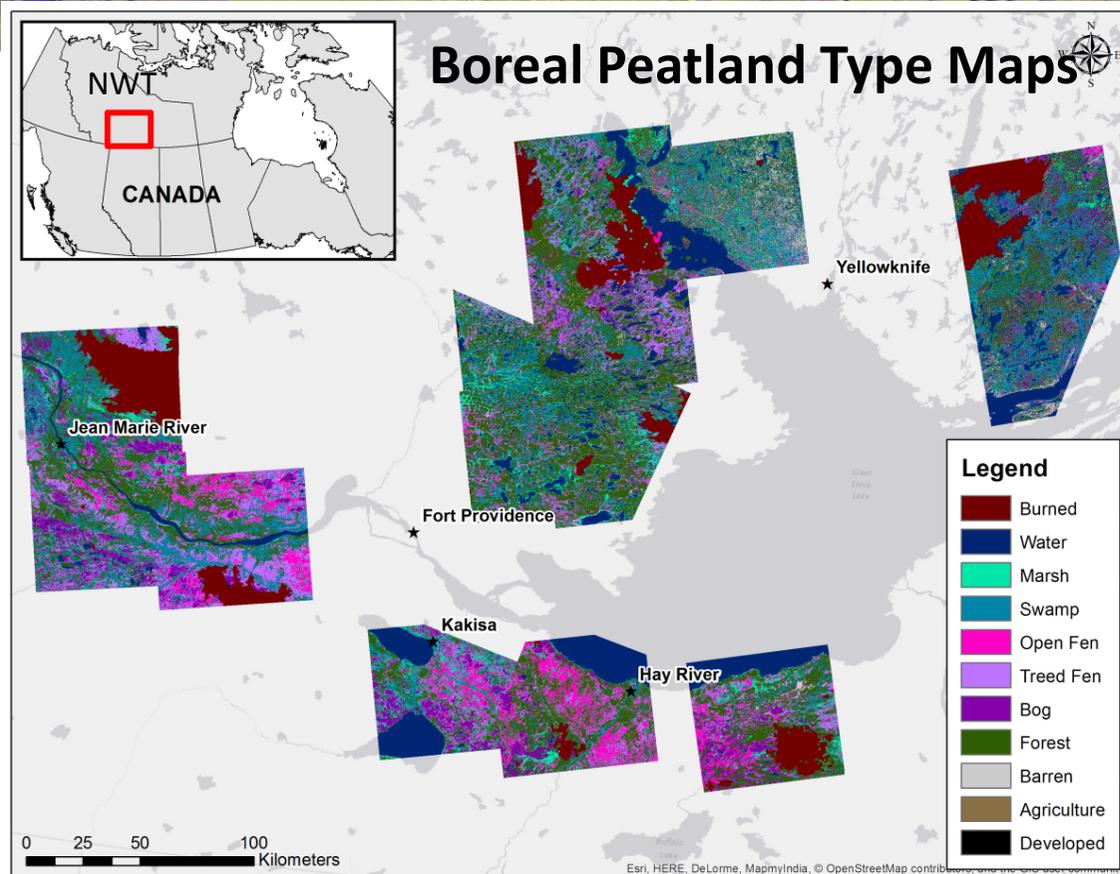


Open water from ALOS  
ScanSAR classification



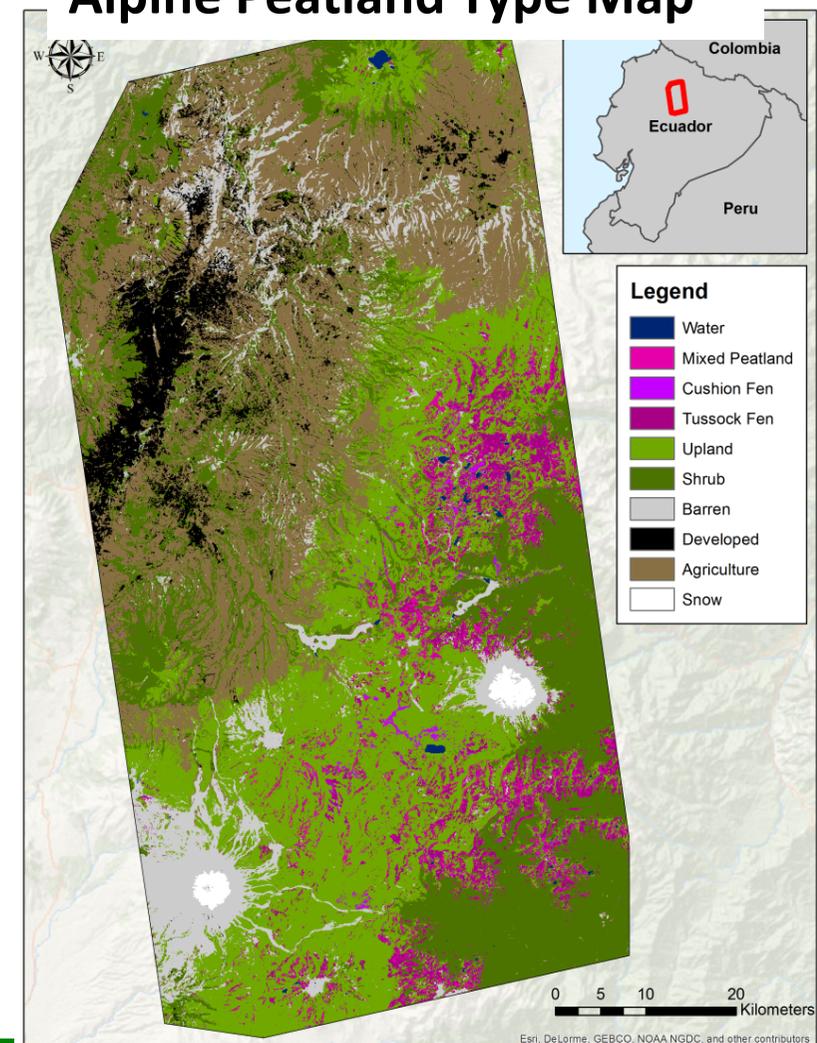
Inundated vegetation  
from ALOS ScanSAR  
classification





## Tropical and Boreal Peatland Type Mapping from PALSAR-Landsat multi-season Data

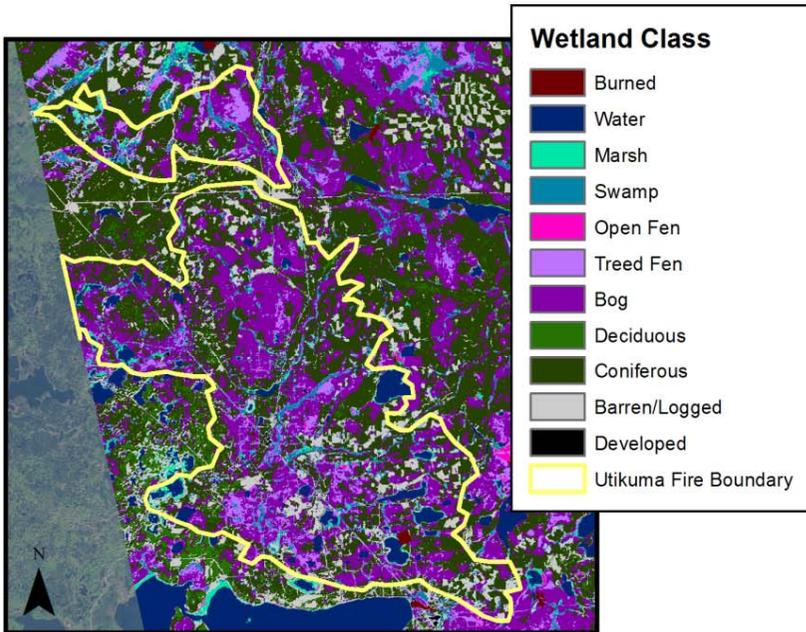
### Alpine Peatland Type Map



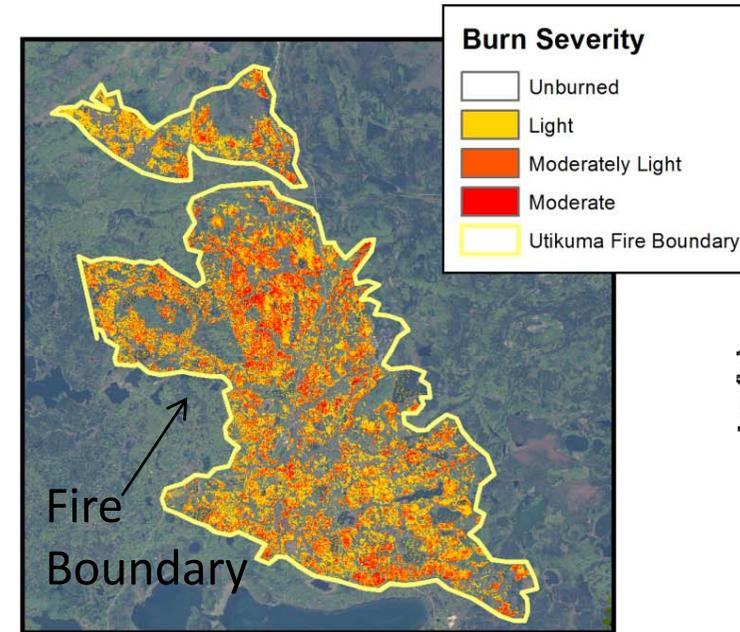
- Peatlands store large amounts of carbon belowground
- Important to know where and what type of peatland for accurate carbon estimates
- Peatland type maps created using multi-date PALSAR and Landsat fusion
- NWT Canada experienced 3.4 million hectares of broadscale wildfire in 2014, consuming many peatlands. Peatland maps are under development

## Carbon Emissions Estimates from Boreal Peat Wildfires are Improved with PALSAR-Landsat Peatland Type Maps

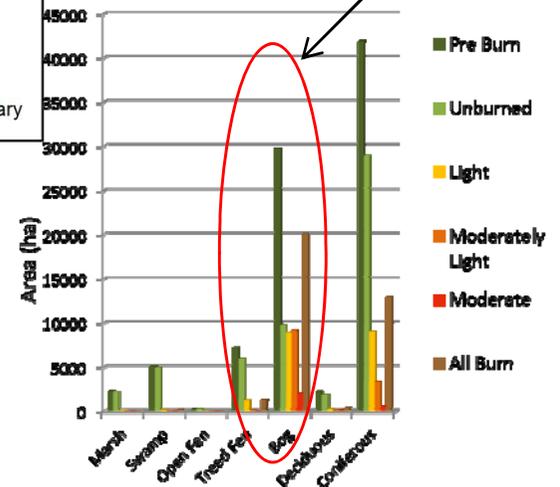
Peatland Type Map



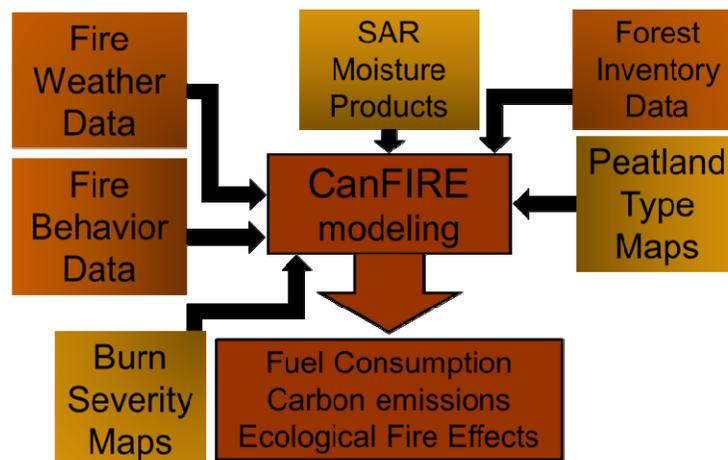
Peat Burn Severity Map



Peat Bogs burned more area of this wildfire



0 5 10 20 Kilometers



Integrated remote sensing products, field data and modeling answer regional scale research questions such as “do peatlands burn as frequently as uplands?” and “how does seasonality affect fuel consumption in peatlands?” “What are the emissions from Peatlands vs. Uplands?”

## Assessment with NASA UAVSAR Datasets

Classify and analyze high resolution imaging radar from the NASA UAVSAR mission, identifying wetlands components



Polarimetric Radar Backscatter  
[R: HH, G: HV, B: VV]

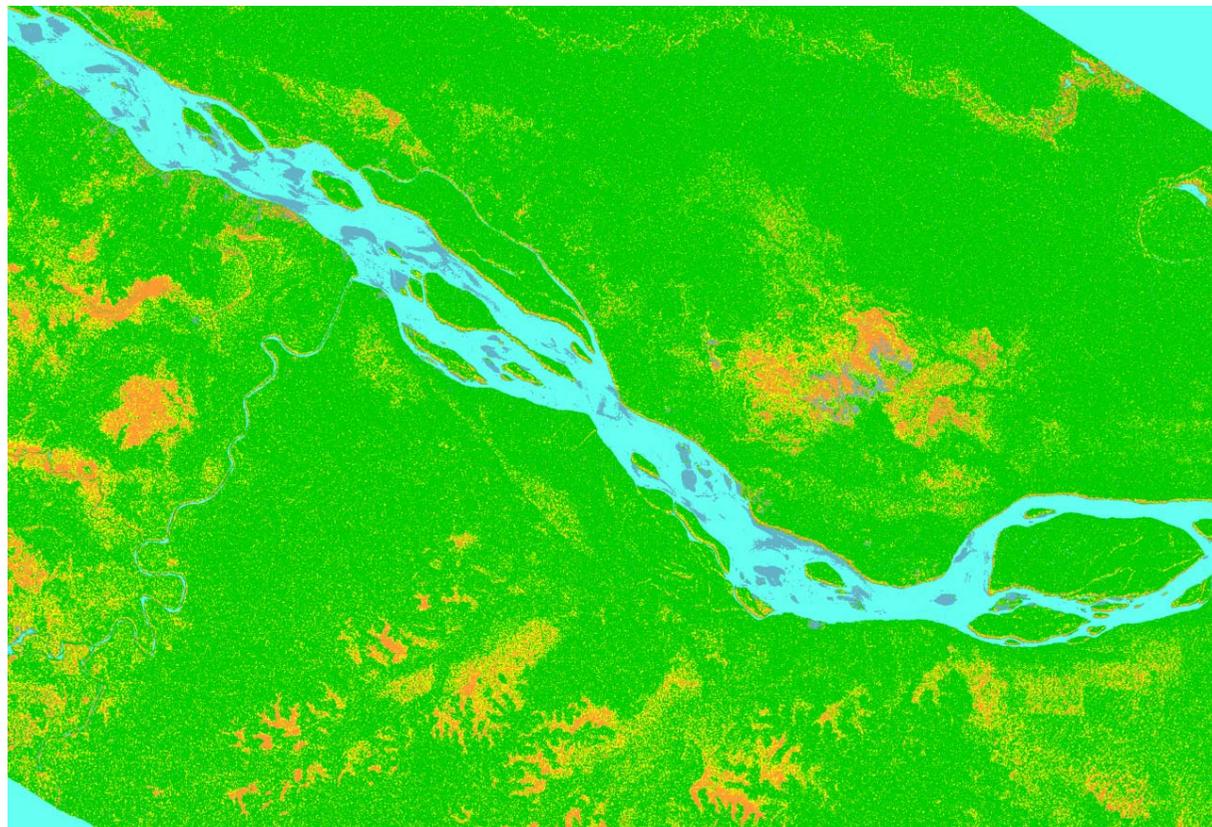
## Classification of UAVSAR data over Napo River in Peru (2013)

Vegetation inundation state determined along two transects within 2 days of UAVSAR data acquisition

Classification derived from Van Zyl decomposition of L-band Quad Pol data UAVSAR

Classification verified along field transects

This image spans incidence angles from  $20^\circ$  to over  $60^\circ$  (as determined from the SRTM DEM)



UAVSAR Inundation Classification : Green: not inundated  
Yellow and Orange: inundated vegetation  
Light and Dark Blue: open water/bare ground

UAVSAR data acquired  
March 31, 2013

Chapman et al, "Validation of forested inundation extent revealed by L-band polarimetric and interferometric SAR data", IGARSS 2014, Quebec, Canada July 2014

# Pacaya-Samiria National Reserve, Peru



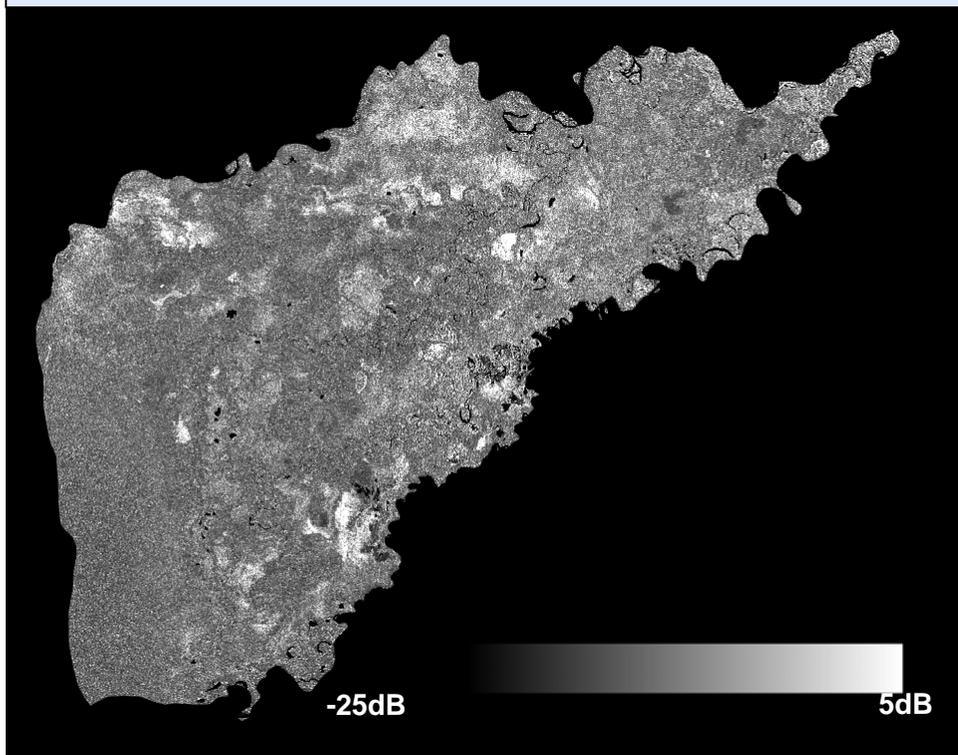
- Most extensive tropical flooded forest in the Peruvian Amazon
- Spans area of more than 20,000 km<sup>2</sup>
- Hosts rich biodiversity!



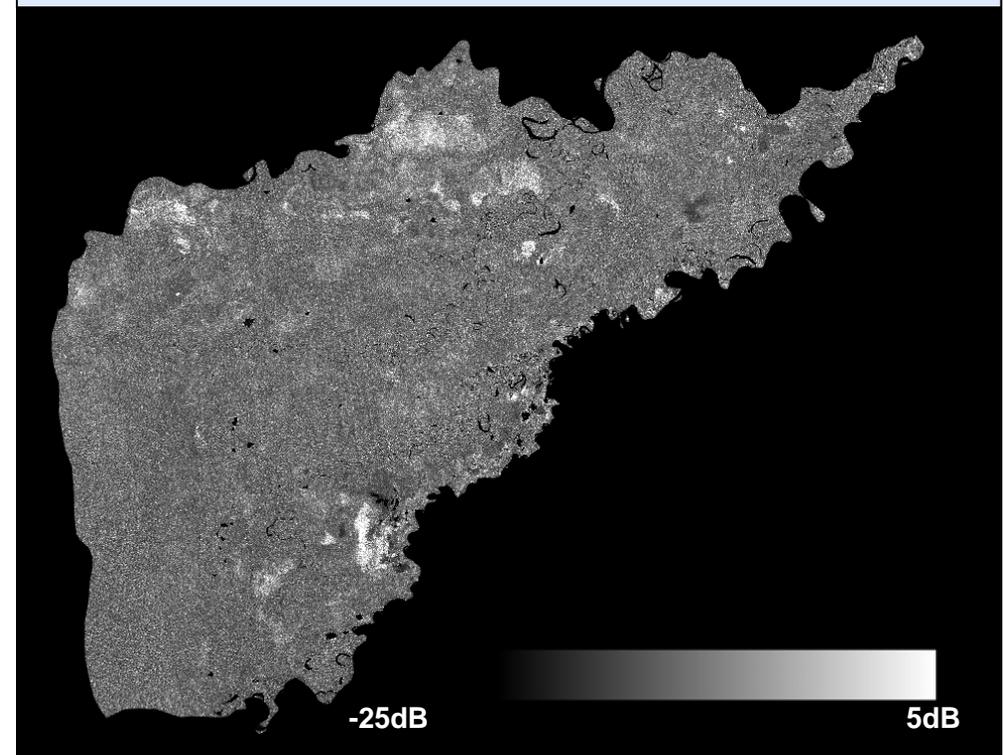
## Mapping of wetlands in Pacaya Samiria Reserve, Peru

### Seasonality of inundation patterns

Wet Season [Mar-Apr 2009]  
HH-polarized ALOS PALSAR Fine-Beam



Dry Season [Jul-Aug 2010]  
HH-polarized ALOS PALSAR Fine-Beam

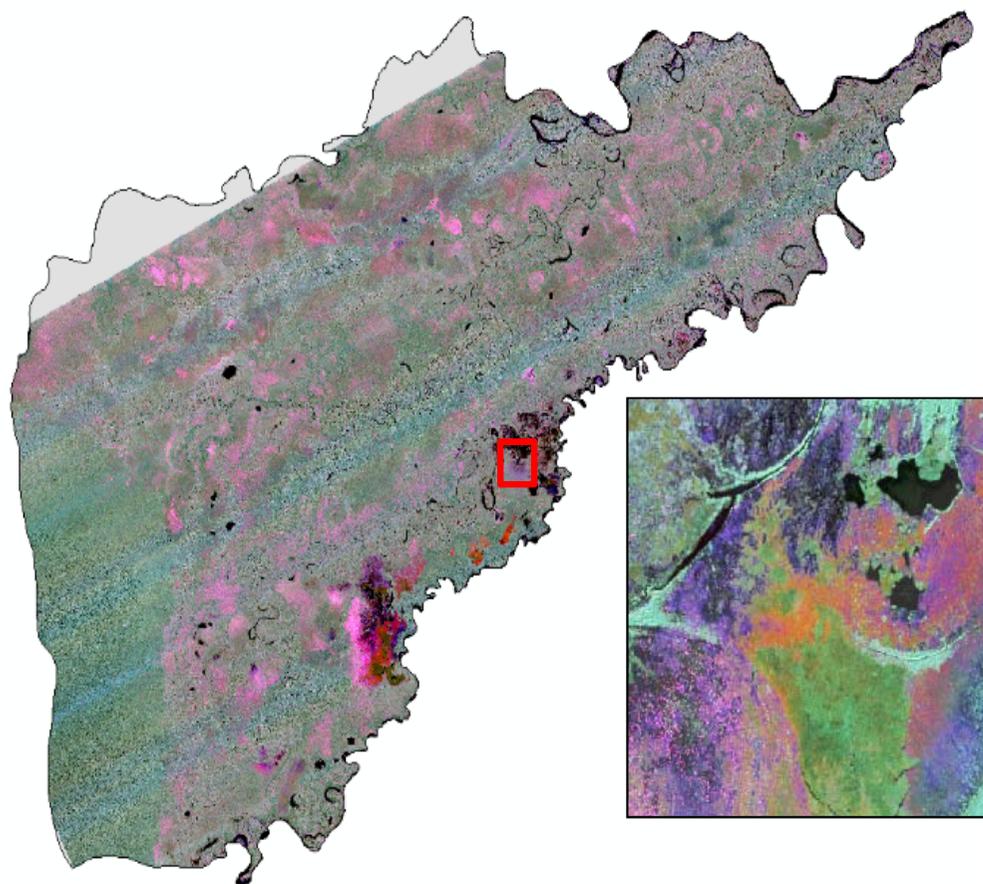


## **UAVSAR False Color Mosaic, normalized for incidence angle**

March 17 2013

R: HH, G: HV, B: VV

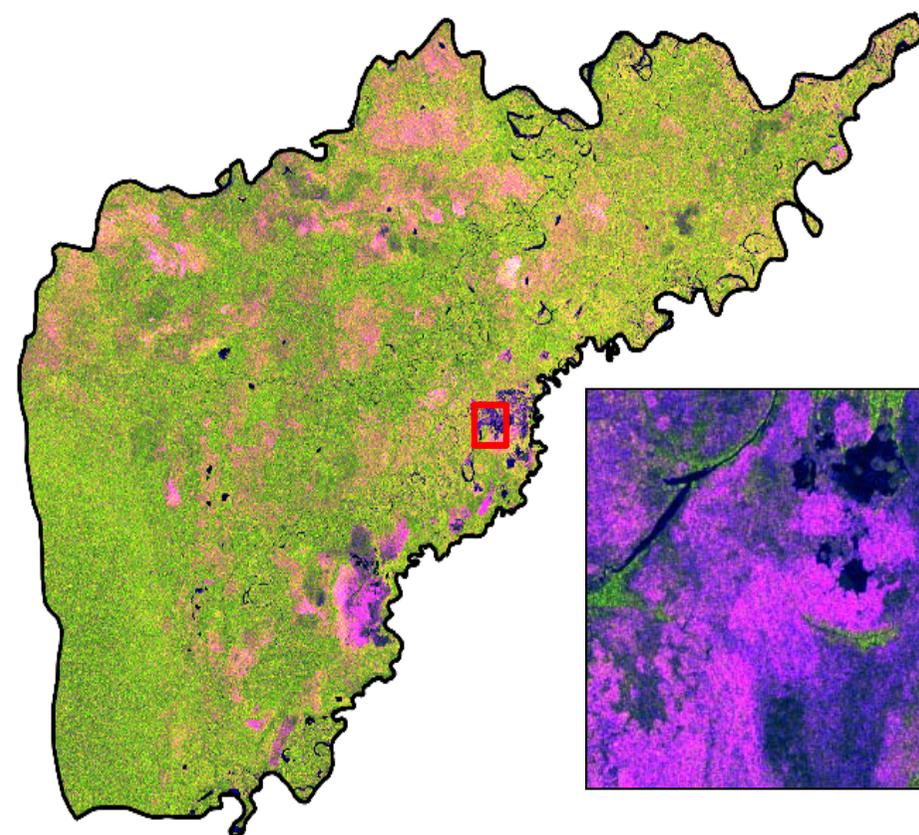
(grey indicates no data collected)



## **ALOS PALSAR False Color Fine Beam Mosaic**

March - April 2009

R: HH, G: HV, B: HH - HV



## Inundation monitoring in East Africa



Study Area:  
UTM Zone  
36P



- Land cover is highly heterogeneous:
  - Sahel desert in the north
  - Prominent agriculture between the Blue and White Nile Rivers
  - Sudd wetlands in the south west
- 55 PALSAR ScanSAR acquisitions between Dec 2008 – Nov 2010 [HH polarization only]
- Applied various classification algorithms on all 55 images:
  - K-means clustering
  - Decision tree
  - Maximum likelihood
  - Random forest

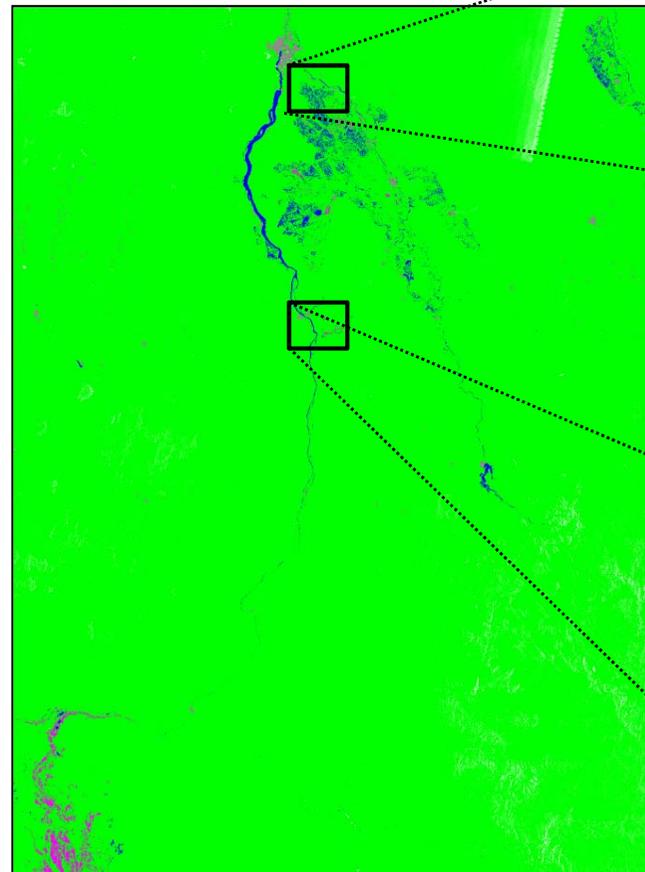
## Classification of ALOS PALSAR ScanSAR

## Inundation monitoring in East Africa

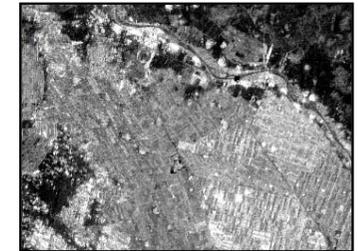
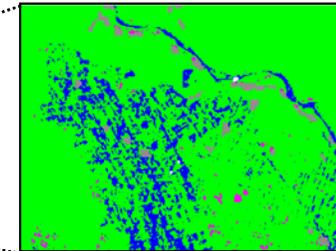
Google Earth



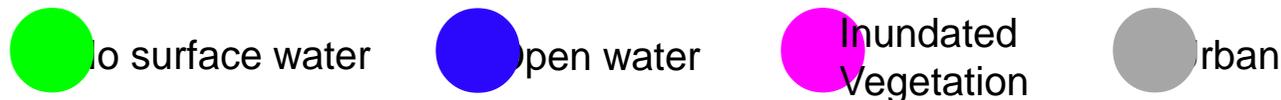
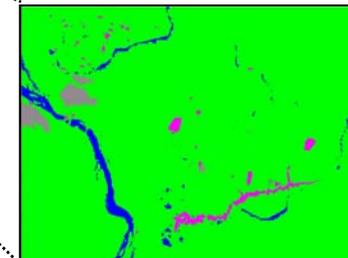
Decision-Tree Result



Rice paddies



Cotton fields



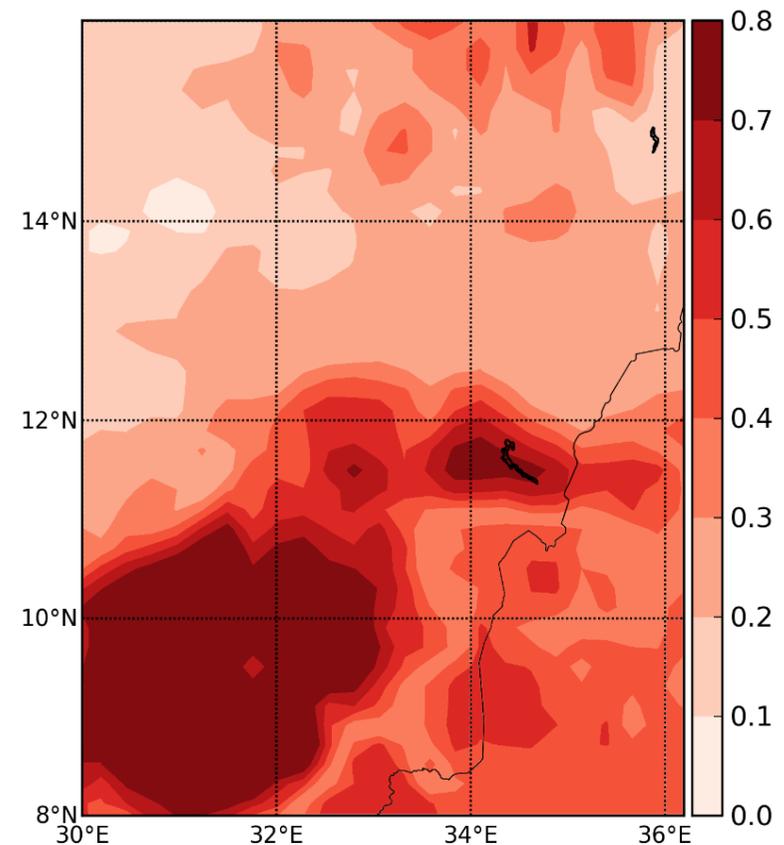
## Classification of ALOS PALSAR ScanSAR

## Comparison of mean SWAMPS and resampled classified SAR water fraction

The classified data are resampled to the 25km EASE grid, counting the number of pixels classified as Open Water and Inundated Vegetation and estimating the surface water fraction.

- 63% of the test area observed statistically significant correlation between inundated fractions derived from SWAMPS and those from classified PALSAR images
- Low  $R^2$  values were observed predominantly in highly arid regions

## Inundation monitoring in East Africa



Pixel-wise  $R^2$  coefficients between SWAMPS and resampled classified PALSAR images.

## Inundation monitoring in East Africa

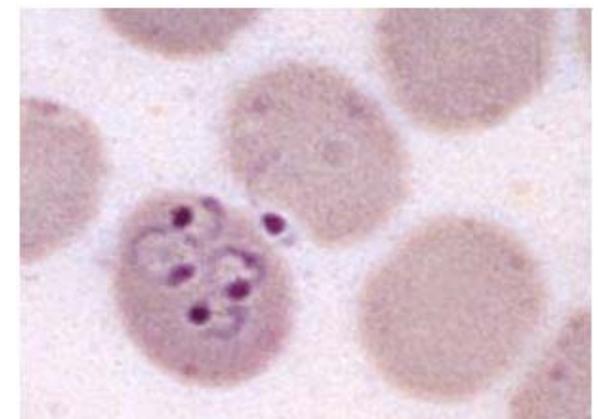
- Public health is a major concern in Africa where malaria epidemics are a burden on population and economies
- Vector-borne diseases are driven by environmental conditions that harbor specific vector habitat conditions such as:
  - 1) specific air temperature
  - 2) humidity for adult mosquito survival
  - 3) surface water for reproduction
- Natural disasters such as floods can also jeopardize the control systems and infrastructures developed in countries to fight infectious diseases
- **Project goal:** Deliver new and improved spatio-temporal characterization of surface water products to support the Ministries of Health in detecting risk areas of vector-borne disease transmission



Source: International Federation of Red Cross/Red Crescent



Source: CDC

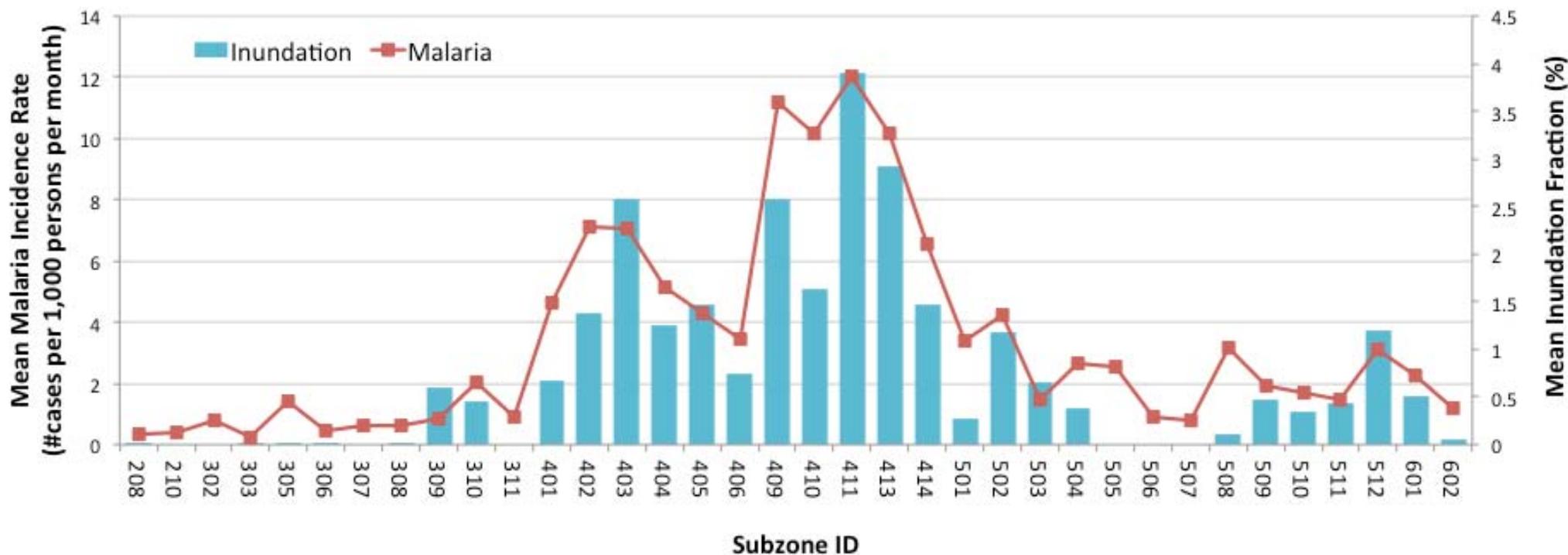


Source: CDC



## Inundation monitoring in East Africa

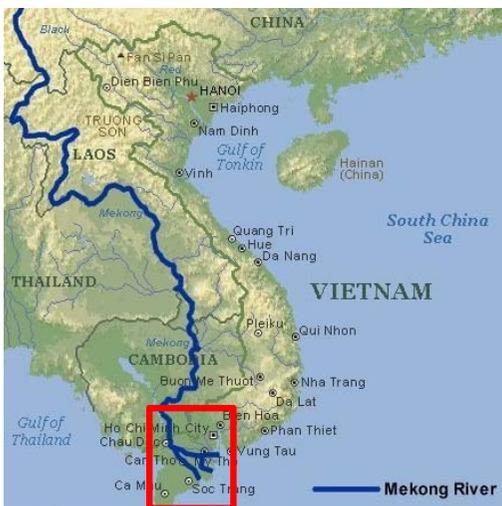
Correlative analysis of historical monthly malaria incidence rates (#cases per 1,000 persons) for subzones in the eastern African country Eritrea for 1996-2003



Spatial distribution of mean surface water fraction [SWAMPS] and mean malaria incidence rate across subzones of Eritrea (excluding coastal zones), 1999-2003.

## Monitoring change and threat to River Deltas Systems: Mekong Delta

- The Mekong River delta is the world's 3<sup>rd</sup> largest delta
  - ↓ densely populated
  - ↓ considered as Southeast Asia's "most important food basket"
  - ↓ rich in biodiversity at the world scale
- Increasingly affected by human activities and exposed to subsidence and coastal erosion
- Several dams have been constructed upstream of the delta and many more are now planned
- **GOAL:** Can we link landcover/land-use changes with inundation patterns using remote-sensing?



Mekong Delta in red

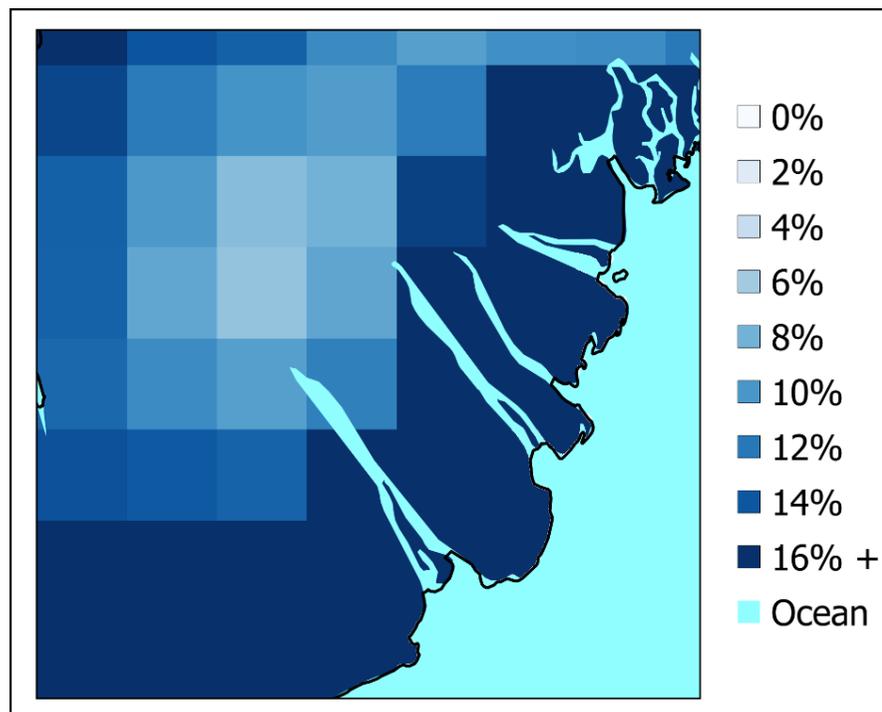


Source: National Geographic

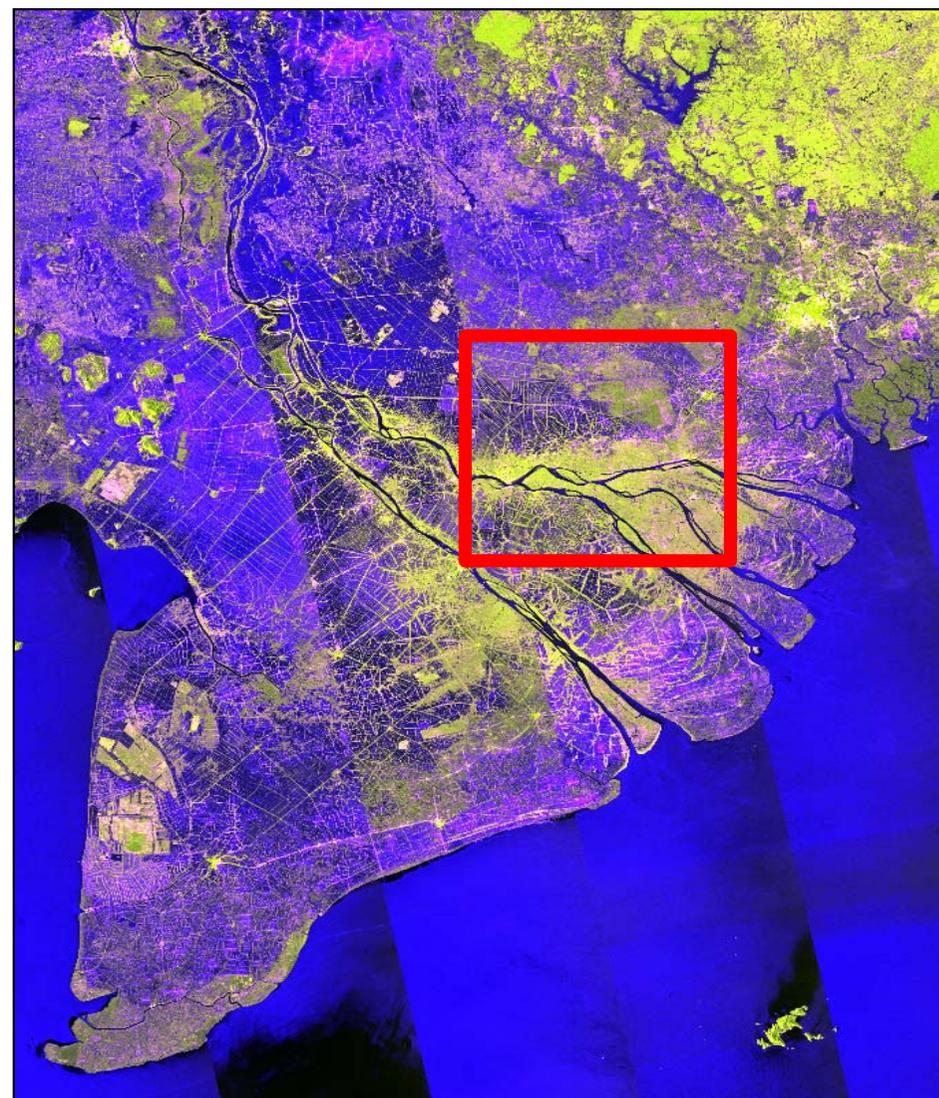


Source: National Geographic

## Mekong River Delta, Indochina



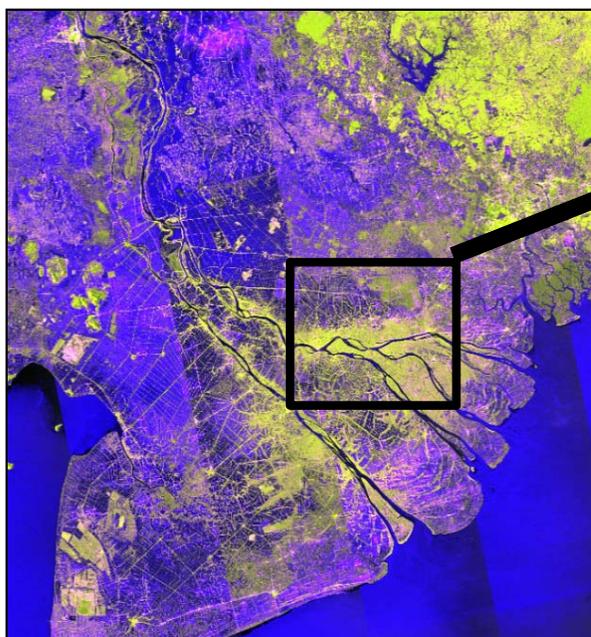
### SWAMPS 25km $f_w$ , Mekong River Delta, Jul 1-8 2009



ALOS PALSAR RGB backscatter mosaic, Jun – Aug 2009.  
R: HH, G: HV, B: HH-HV. 30m spatial resolution. Red box  
indicates region used in this study.

## Re-scaling of coarse resolution inundation area fraction

- Static physical information (e.g. topography) alone is not enough for accurate downscaling
- Use of ALOS PALSAR observations as ancillary input:
  - ↓ Identify biases in SWAMPS data product
  - ↓ Combine with machine learning algorithms to “learn” from SAR observations

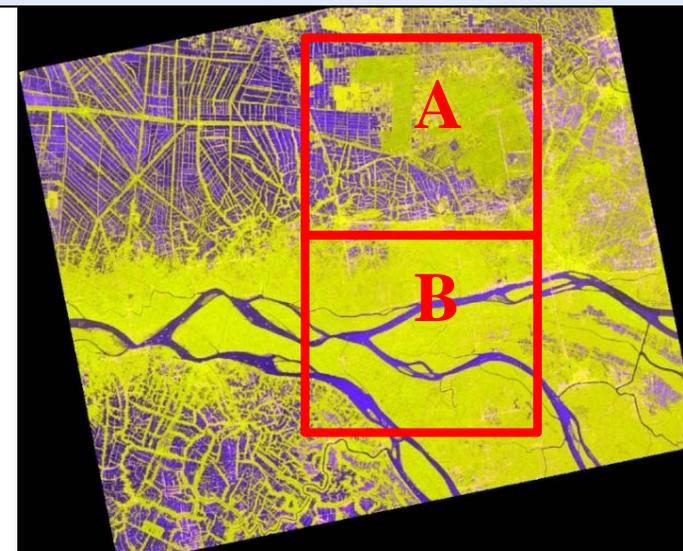


PALSAR FBD mosaic, Mar-May 2009  
[R: HH, G: HV, B: HH - HV]  
30m spatial resolution

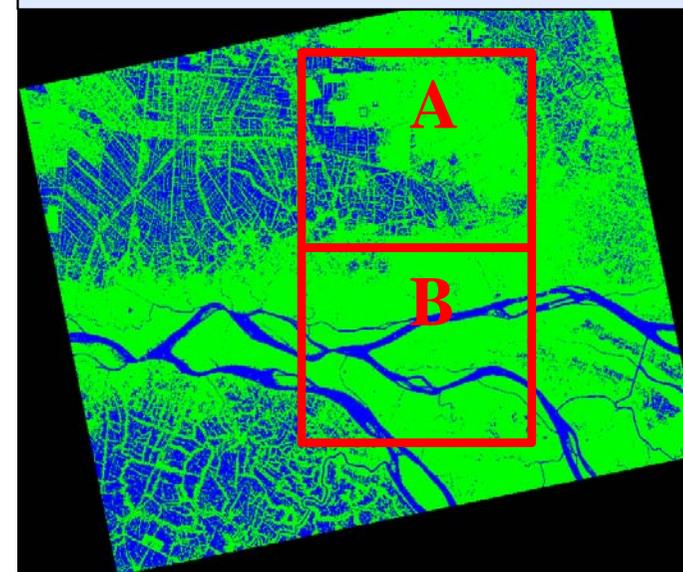
Classified open water in 24 reoccurring ALOS PALSAR fine-beam images (Feb 2007 – Jan 2011)

Aggregated 30m values of % classified as open water within two ease-grid cells (A & B) that are completely contained in image

PALSAR FBD, 11/18/10 [R: HH, G: HV, B: HH - HV]



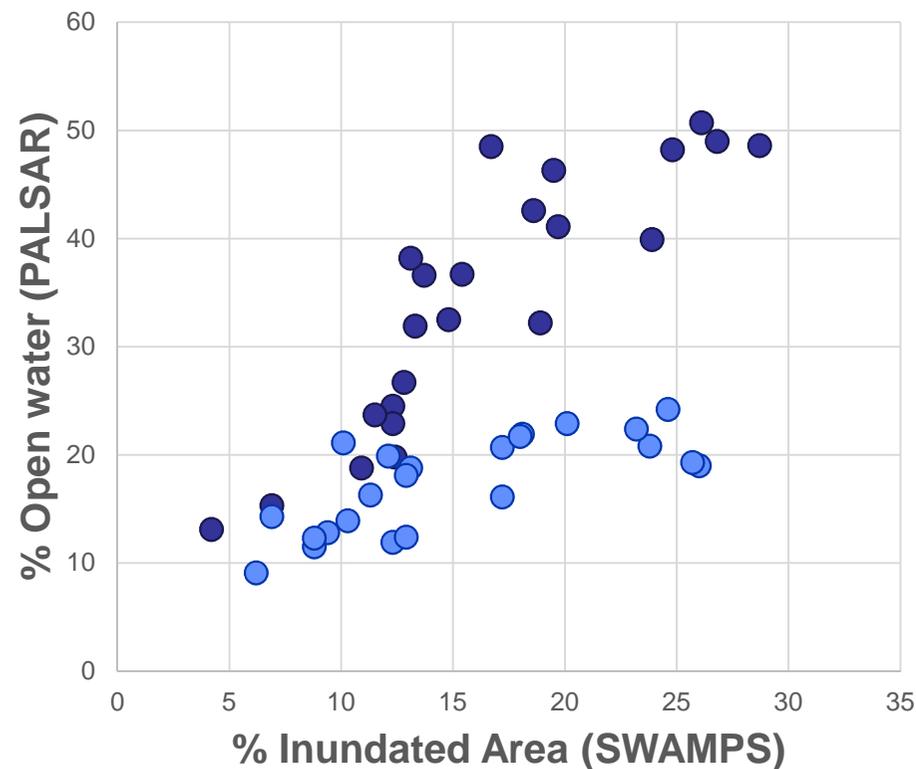
Decision Tree Classification  
Blue = Open Water, Green = Non-Inundated



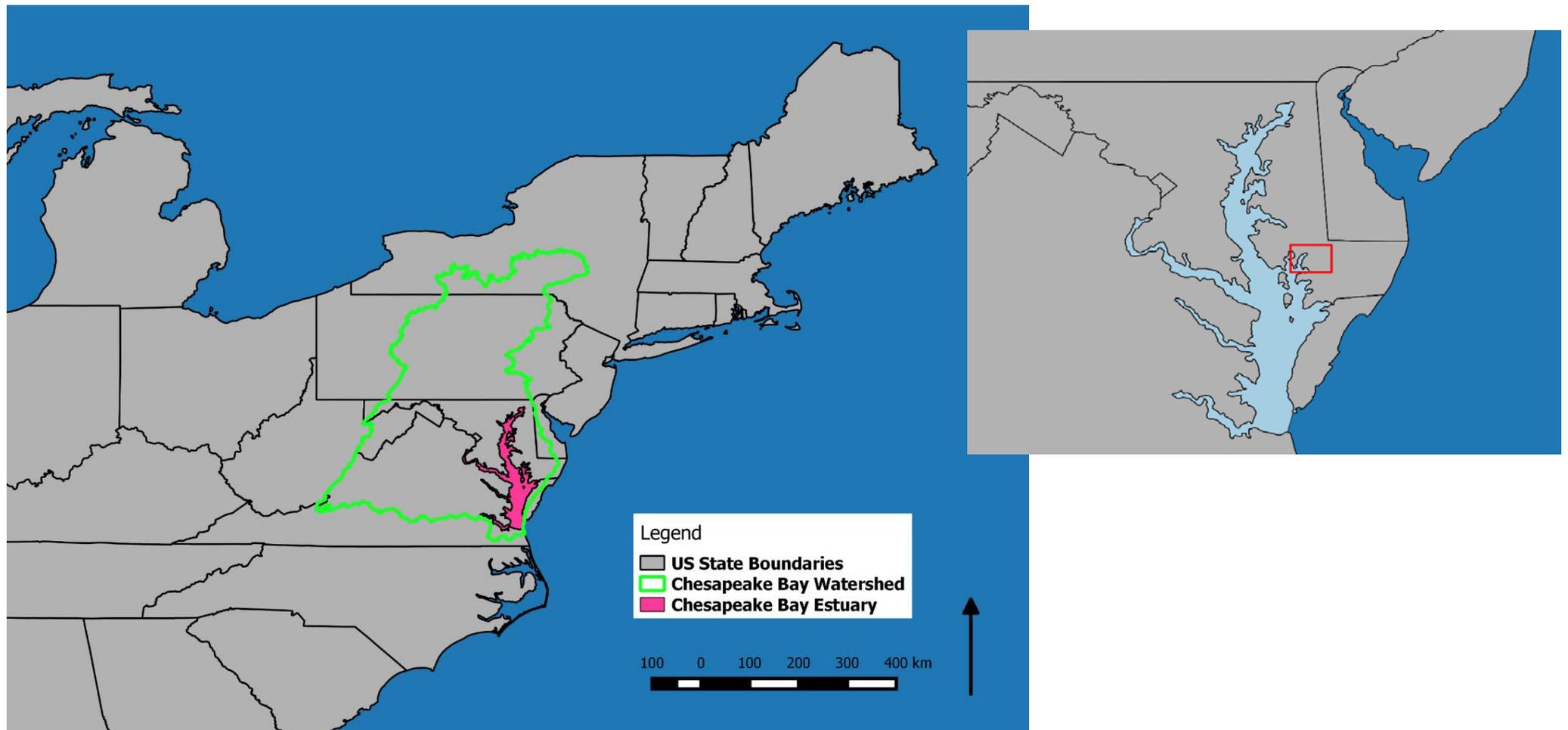
## Re-scaling of coarse resolution inundation area fraction

% Inundated Area Fraction [SWAMPS]  
vs.  
% Classified as Open Water [PALSAR]

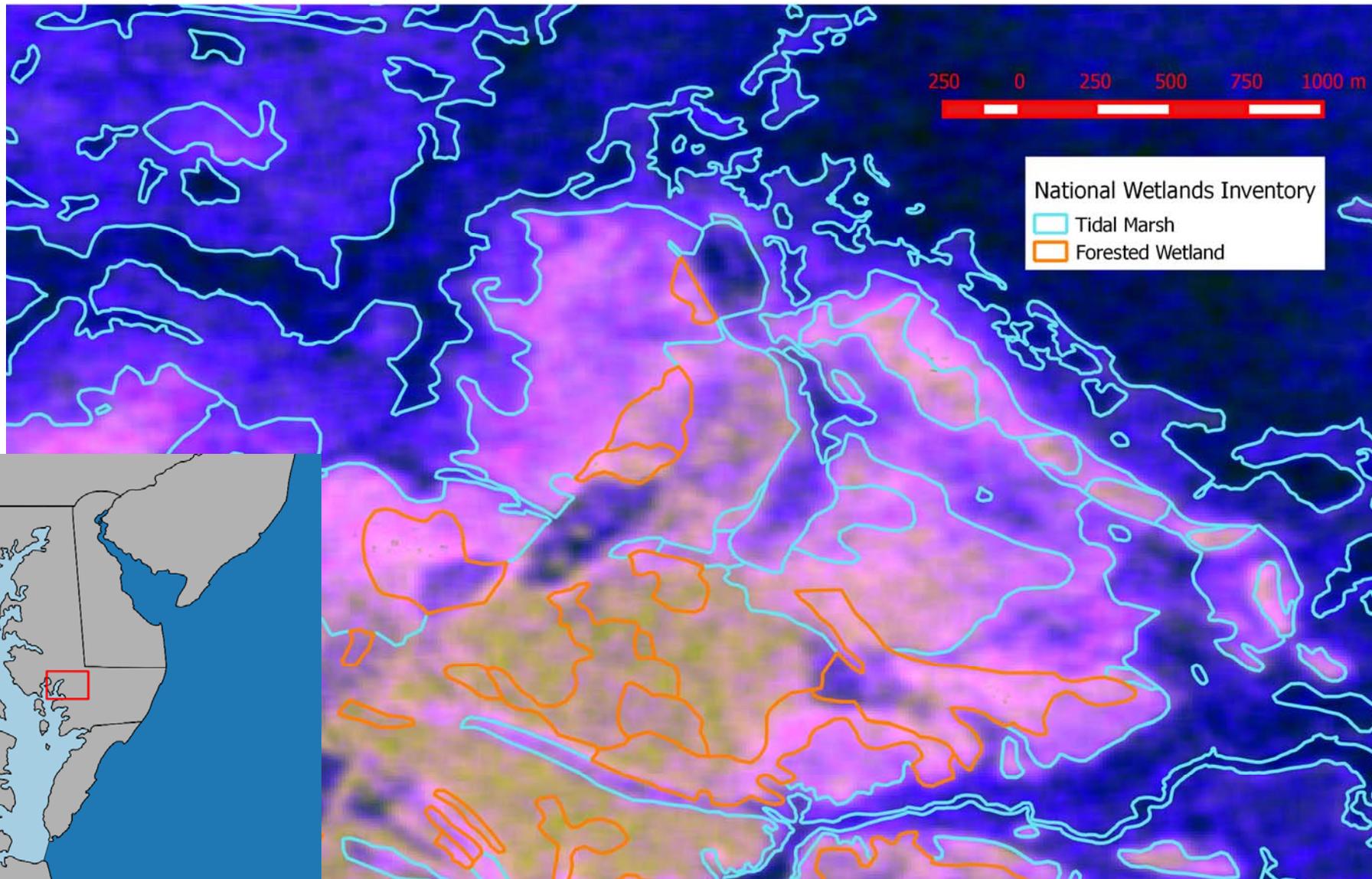
● A ● B



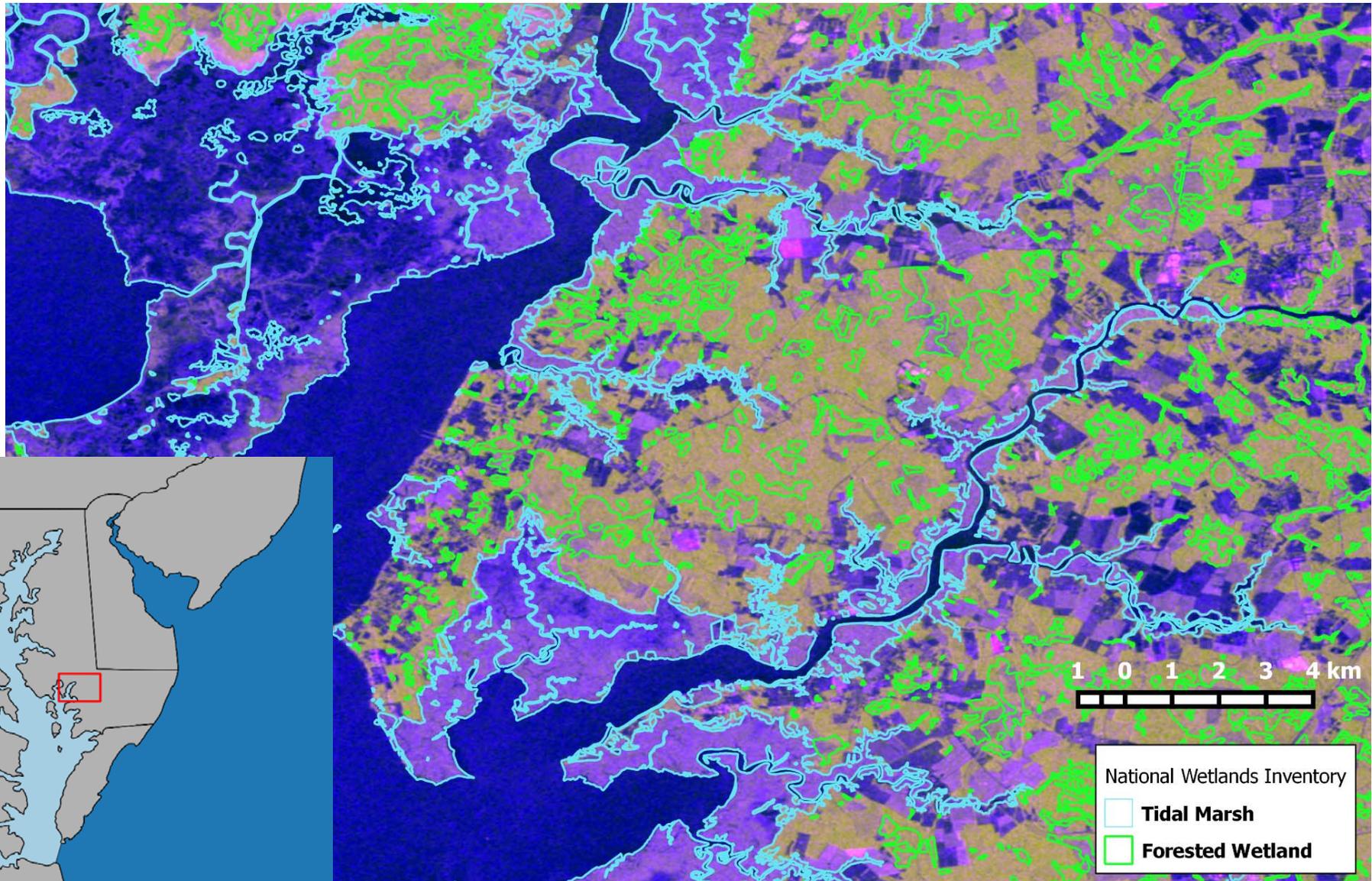
# Carbon exchange in tidal estuary systems: Chesapeake Bay



## PALSAR-1 False Color RGB (HH, HV, HH-HV): Blackwater National Wildlife Refuge

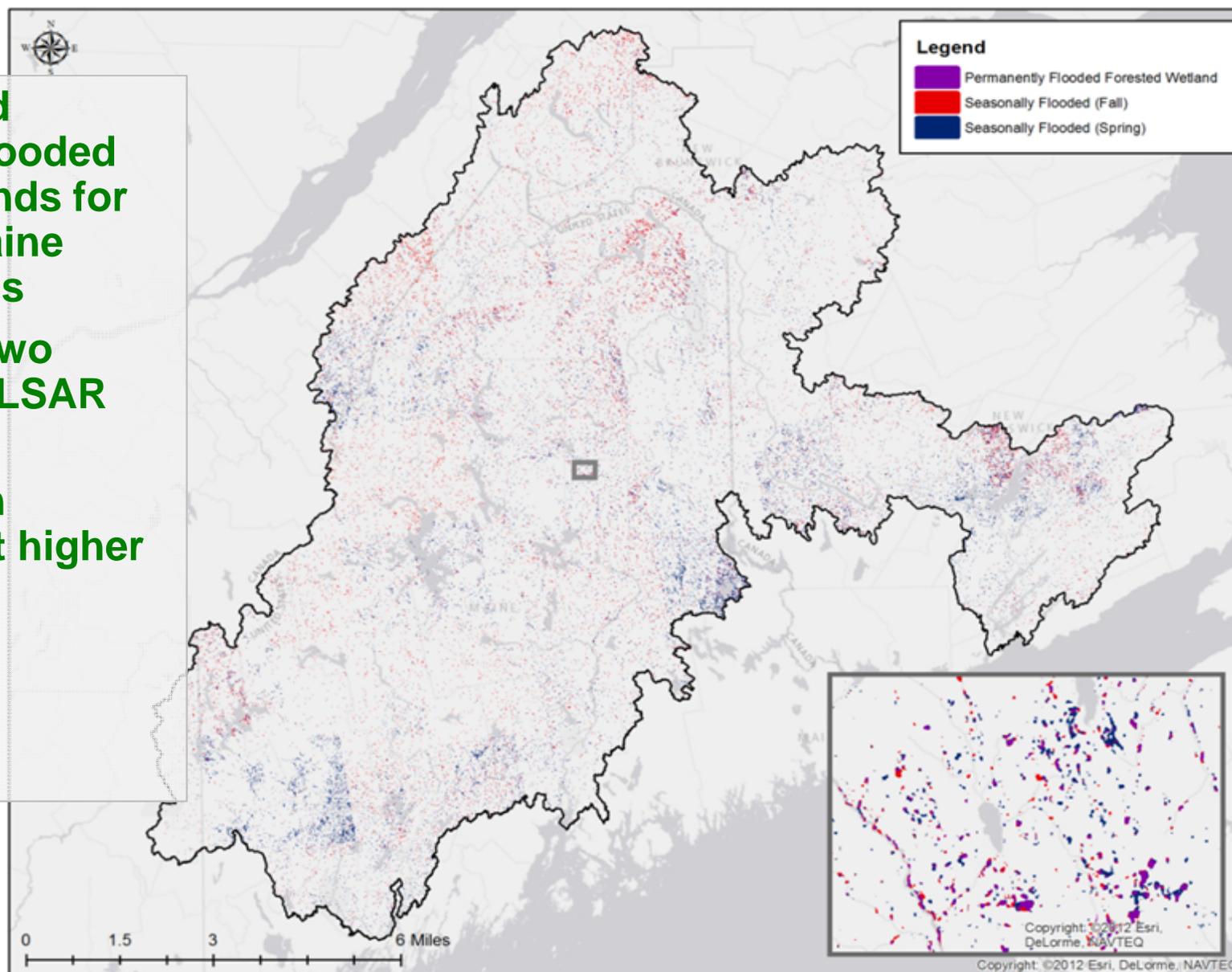


PALSAR-1 False Color RGB (HH, HV, HH-HV). July 2009. Maryland, Eastern Shore



## PALSAR-mapped forested wetland flood condition

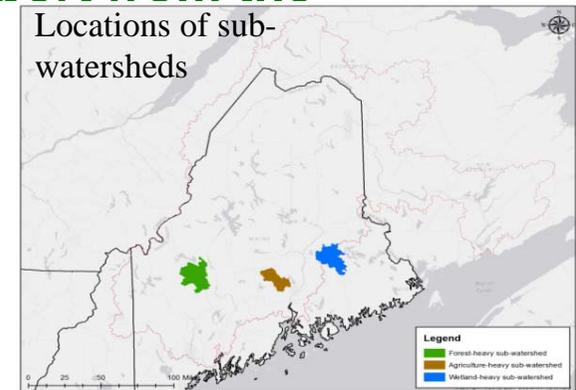
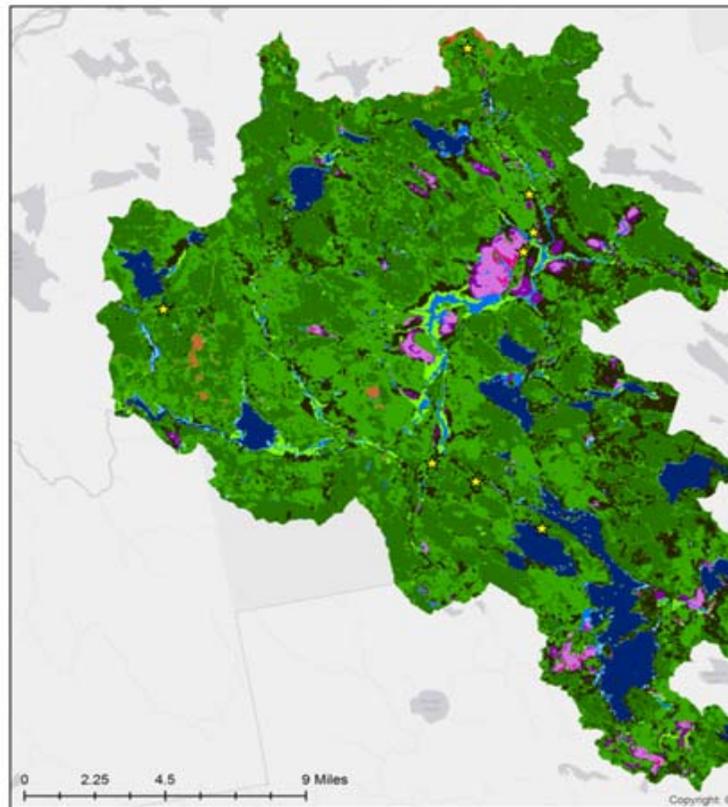
- Seasonally and permanently flooded forested wetlands for four Gulf of Maine drainage basins
- Mapped from two seasons of PALSAR images.
- Inset shows an example site at higher resolution.



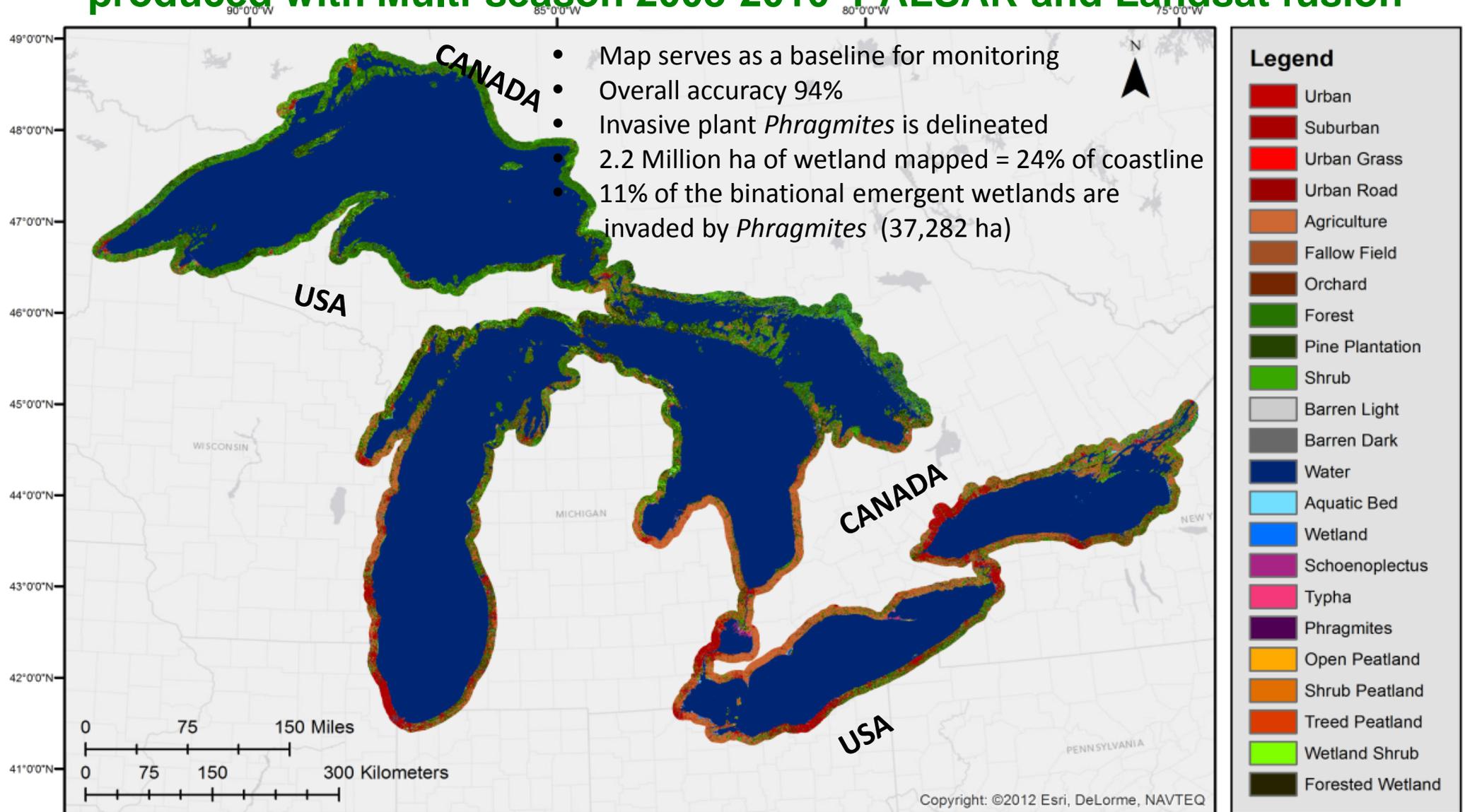
## Carbon Export to the Gulf of Maine

- Mapping the location of wetlands is important for carbon cycling because land cover and land use can impact the amount of carbon exported through streams and rivers. C models require accurate maps of wetlands and in particular forested wetlands to accurately estimate C export to the Atlantic Ocean.
- In Maine, forested wetlands were not well mapped with optical remote sensing methods, and they are important in defining the carbon export from the landscape.

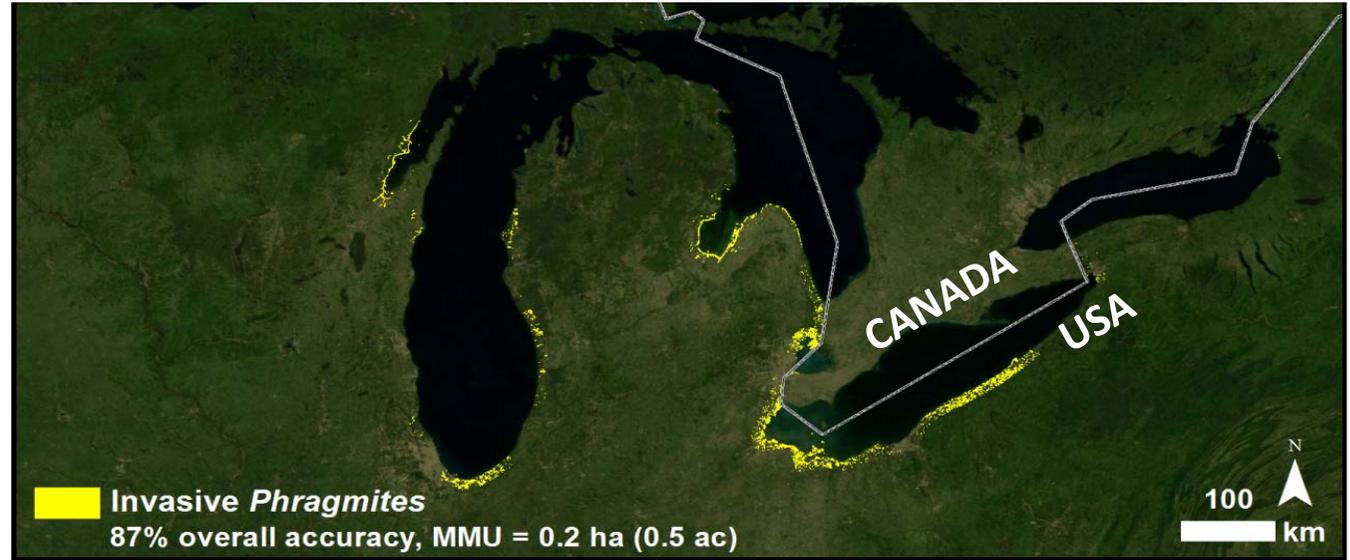
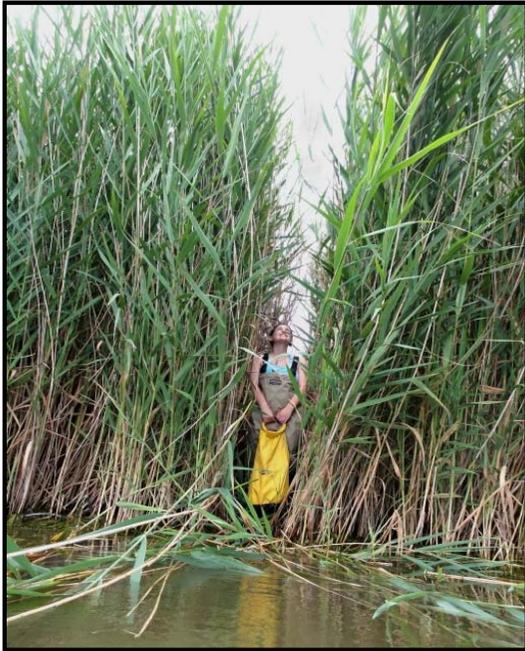
- Two seasons of PALSAR-Landsat fusion were used to identify wetlands, including cryptic forested wetlands.
- This map included delineation of flooded forested wetlands (i.e. swamps) vs. soggy peatlands and open herbaceous wetlands



## First Comprehensive Wetland Map of the Bi-National Coastal Great Lakes produced with Multi-season 2008-2010 PALSAR and Landsat fusion



# U.S.A. PALSAR-Derived Map of Distribution of Invasive *Phragmites australis* within 10 km of Coastline

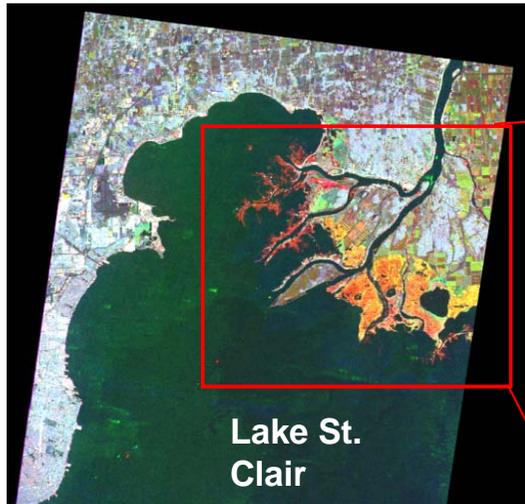
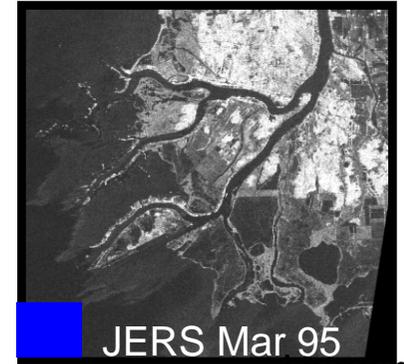
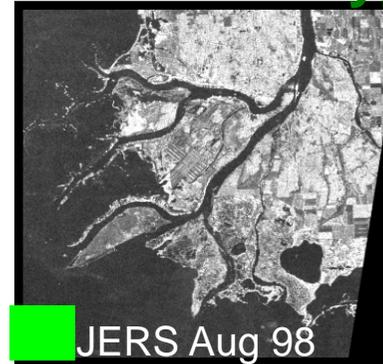
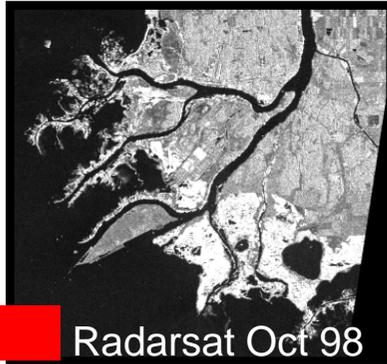


- Invasive wetland plant, *Phragmites australis*, is a tall dense grass taking over coastal Great Lakes (and other regions nationwide) over the past 30 years
- This plant is an increased Carbon sink over native wetland plant species
- Mapping was based on multi-season PALSAR for the USA Great Lakes Coastline
- PALSAR detects the high biomass and inundation

Great Lake	Total Mapping Area within 10 km of coastline	Hectares of <i>Phragmites</i> Mapped
Erie	96,862	8233
Michigan	578,320	6002
Ontario	102,056	13
Superior	<i>No Phragmites mapped</i>	<i>No Phragmites</i>
Huron	75,402	10,395
<b>Total</b>	<b>852,640</b>	<b>24,643</b>



# Multi-SAR Sensor Composite Distinguishes Wetland Plant Types



Dickinson Island  
Harsen's Island

Upland Forest

Cattail/*Scirpus* beds

Wet meadow-sedges

*Phragmites* dominant

Cattail dominant

22.7 km

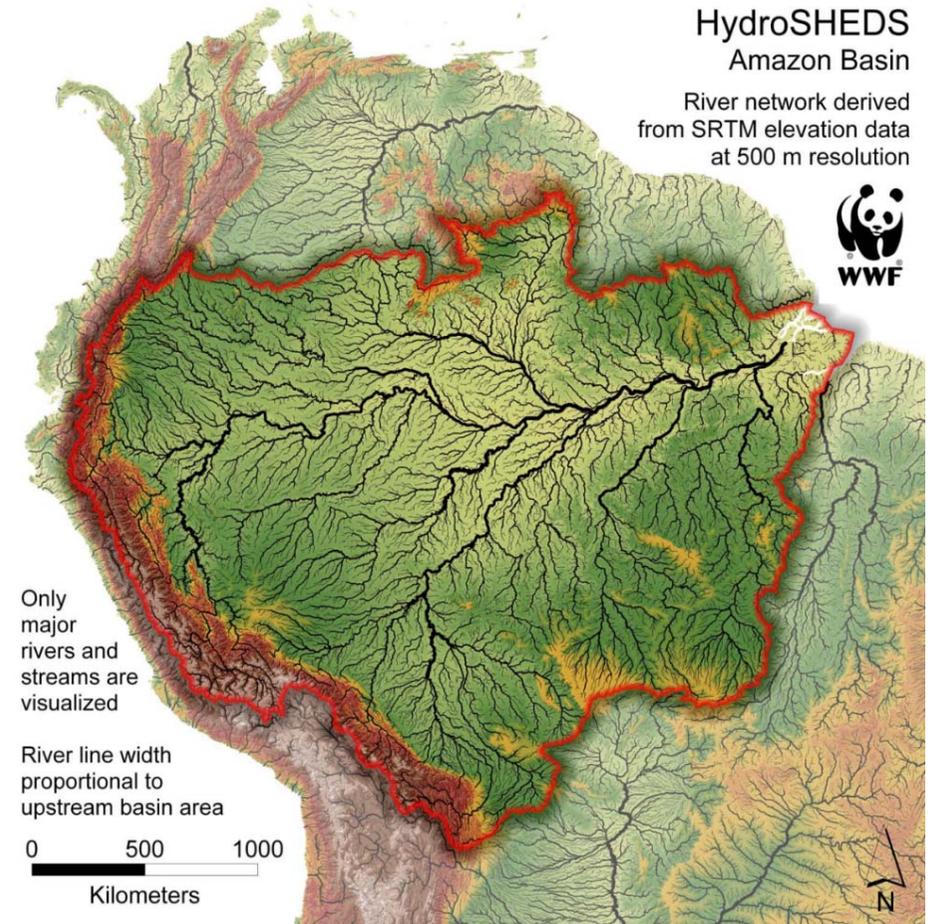
23.6 km

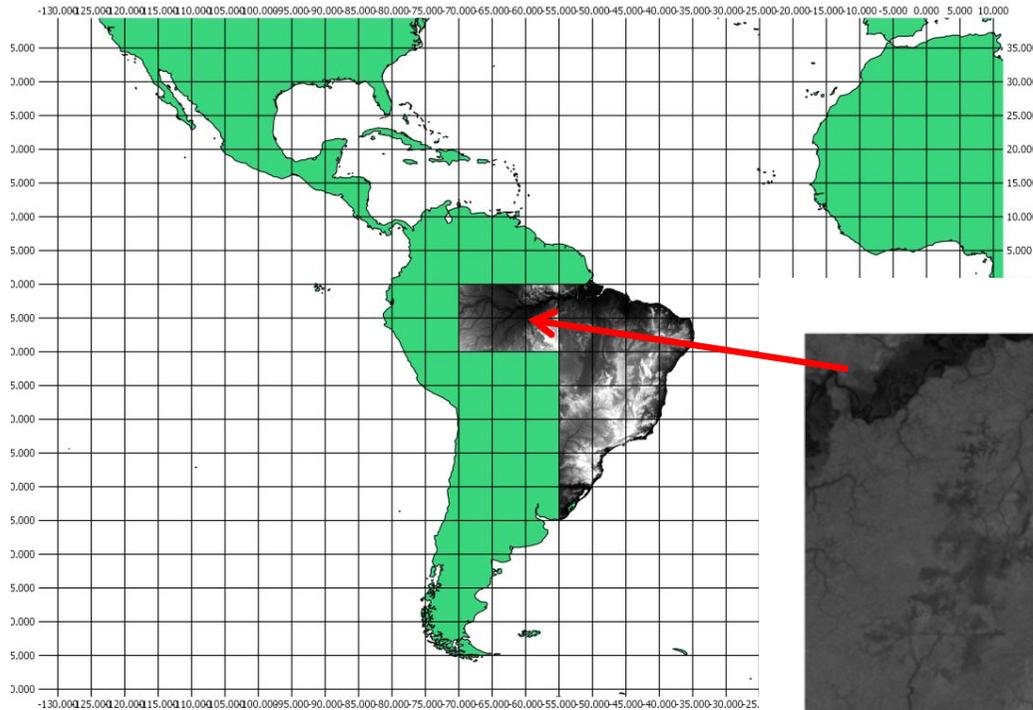
Multi-sensor L- and C-band radar composite depicts the biomass and flooding differences between the various emergent wetlands in this delta (factors relevant to Carbon stores and methane exchange).

JERS distinguishes the invasive plant *Phragmites australis* (medium green areas) which can be more than 5 m tall, and forms dense stands

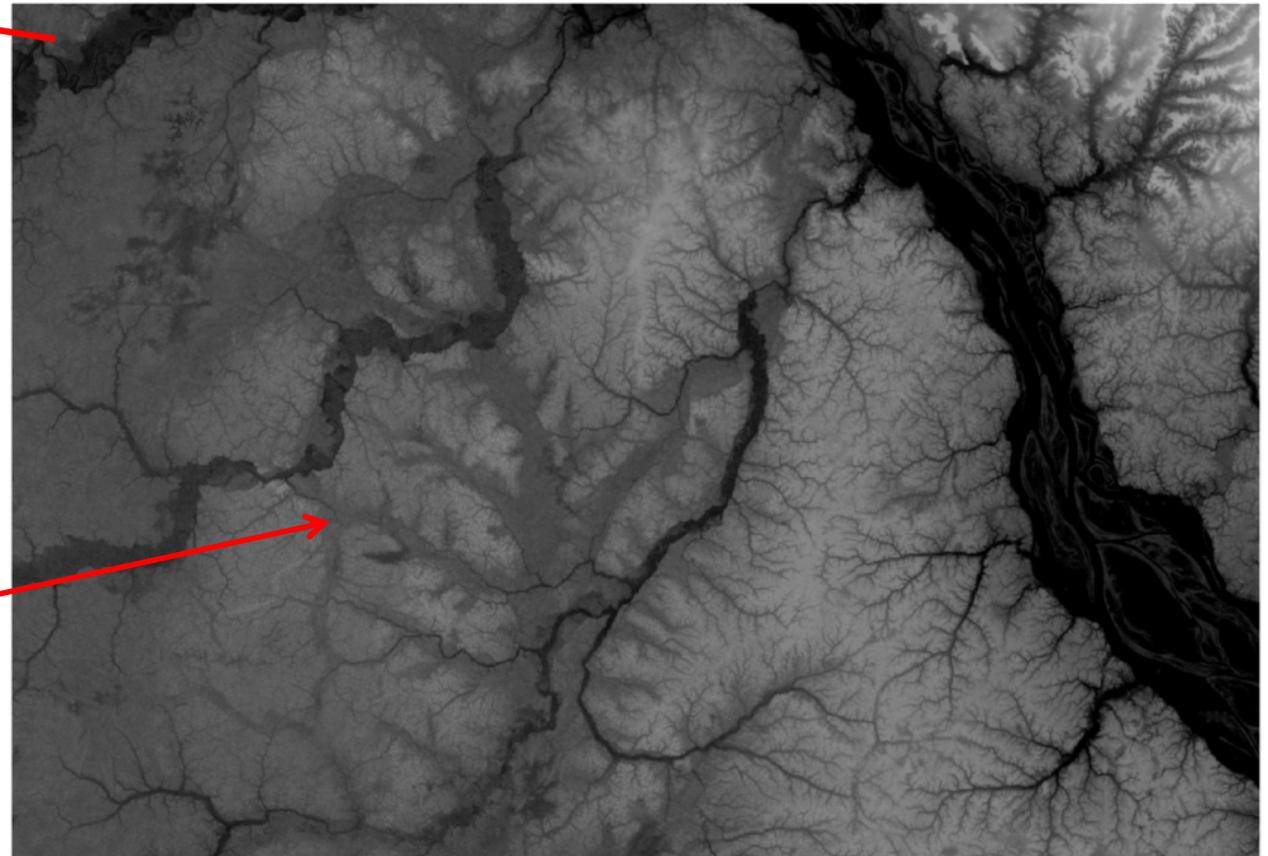
# Characterization of the geological and environmental history of Amazonia

- What has been the history of environmental change across Amazonia from the late Neogene to present?
- When did the Amazonian river drainage form?
- What was the Amazonian landscape like before the Amazon river formed and how did it change after the river formation?





## Integration using SRTM DEM: Amazon and the Atlantic Forest

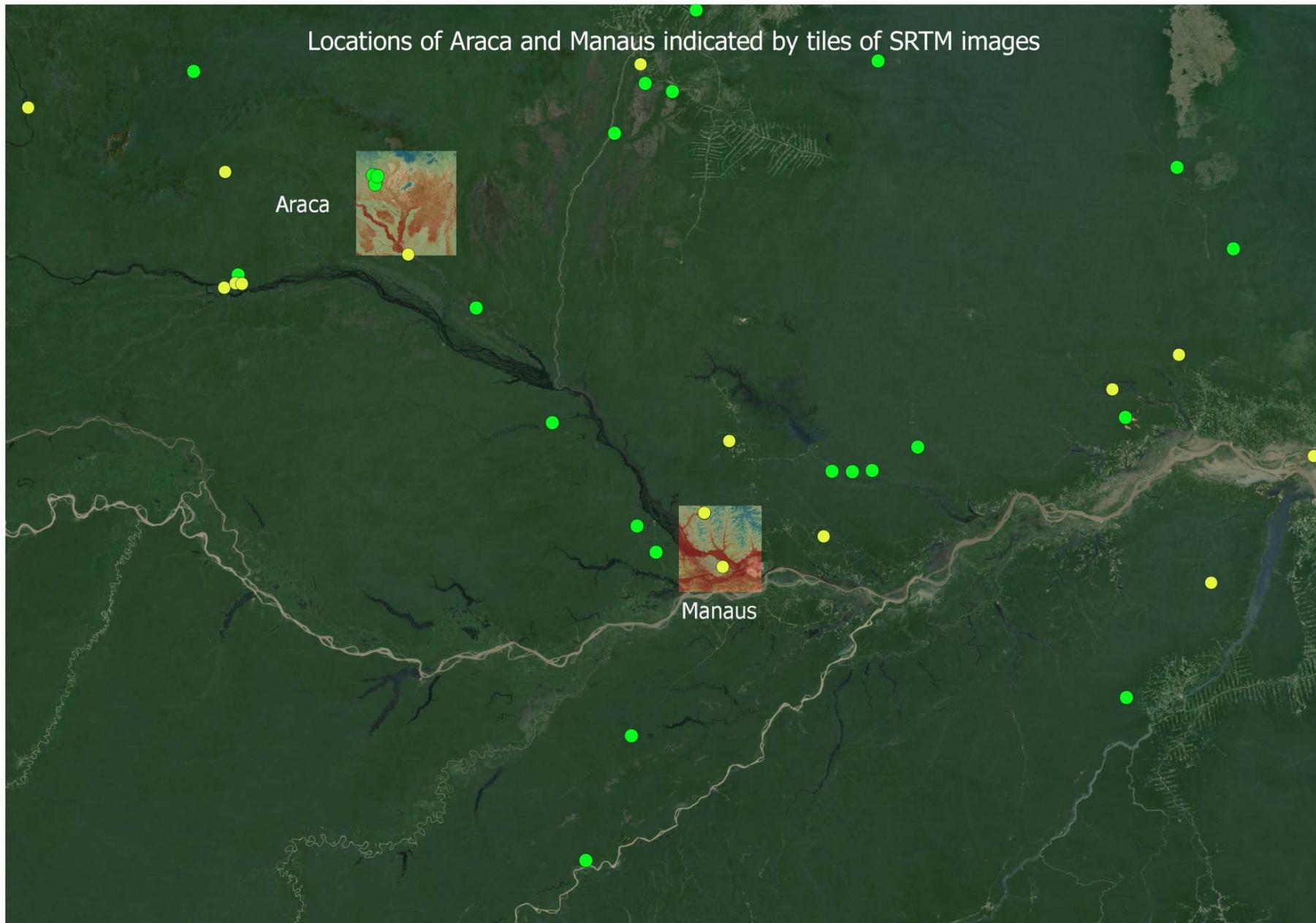


**Ancient  
river system**



Xenopipo (bird species endemic to white-sands) localities correspond to white-sand vegetation

Locations of Araca and Manaus indicated by tiles of SRTM images



# ALOS

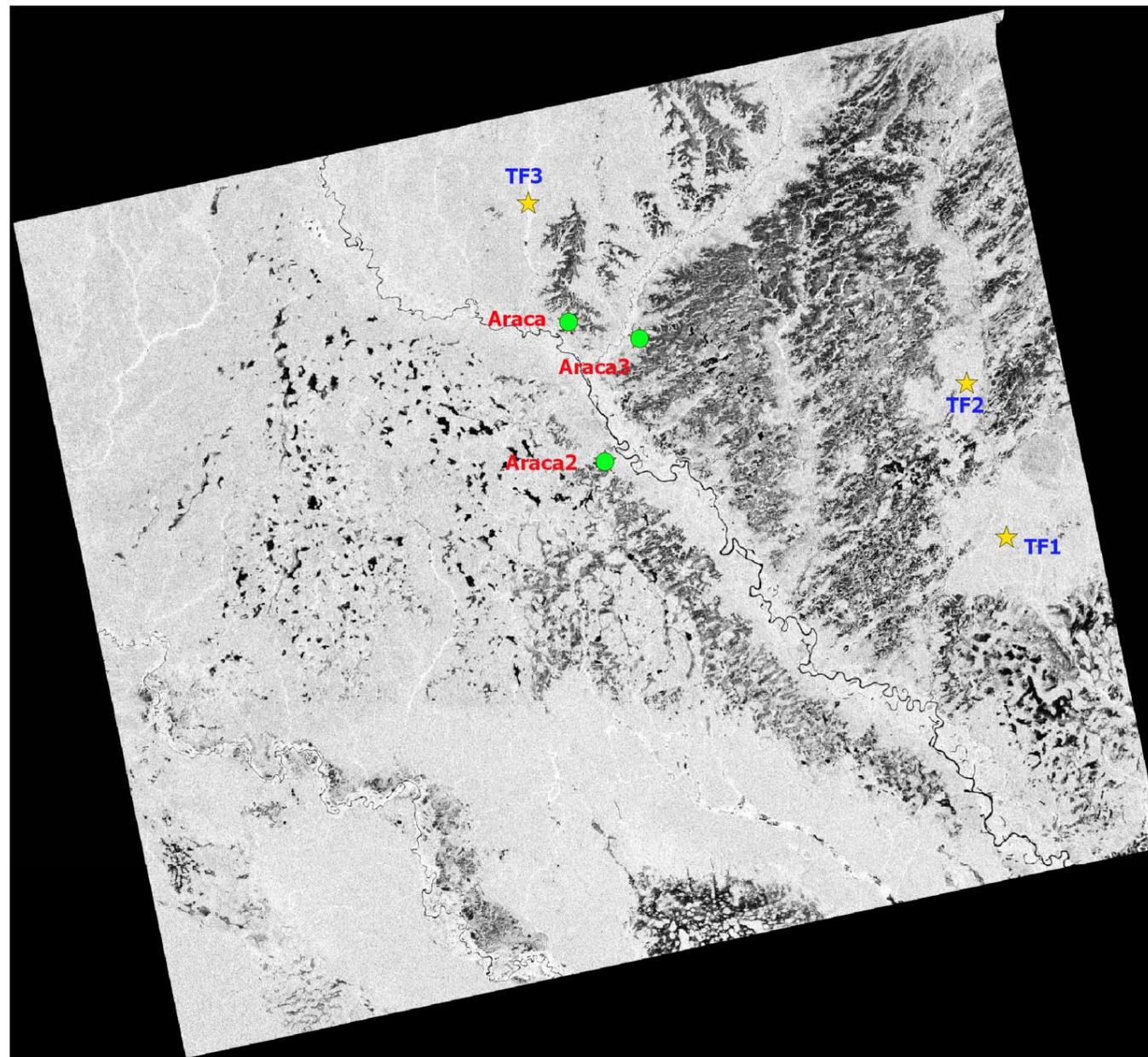
*K&C Initiative*  
*An international science collaboration led by JAXA*



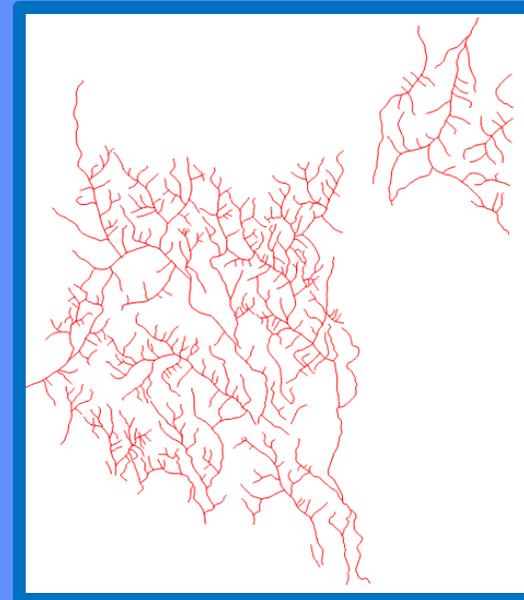
❖ **Photo credits: Eduardo Prata  
(INPA, Brazil)**

ALOS PALSAR  
HH-polarisation  
High-res terrain  
corrected FBD  
Acquisition date:  
Oct-16 2010

- White sand Vegetation
  - ★ Terra Firme
- 7190\_25177\_HH (dB)
- -30.603
  - -0.261416

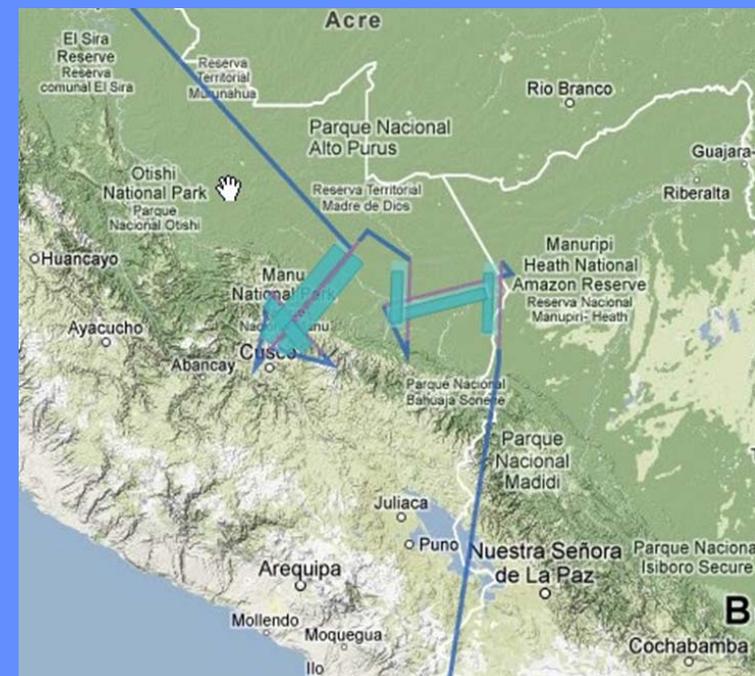


# Defining river branching structure: Planalto Peru



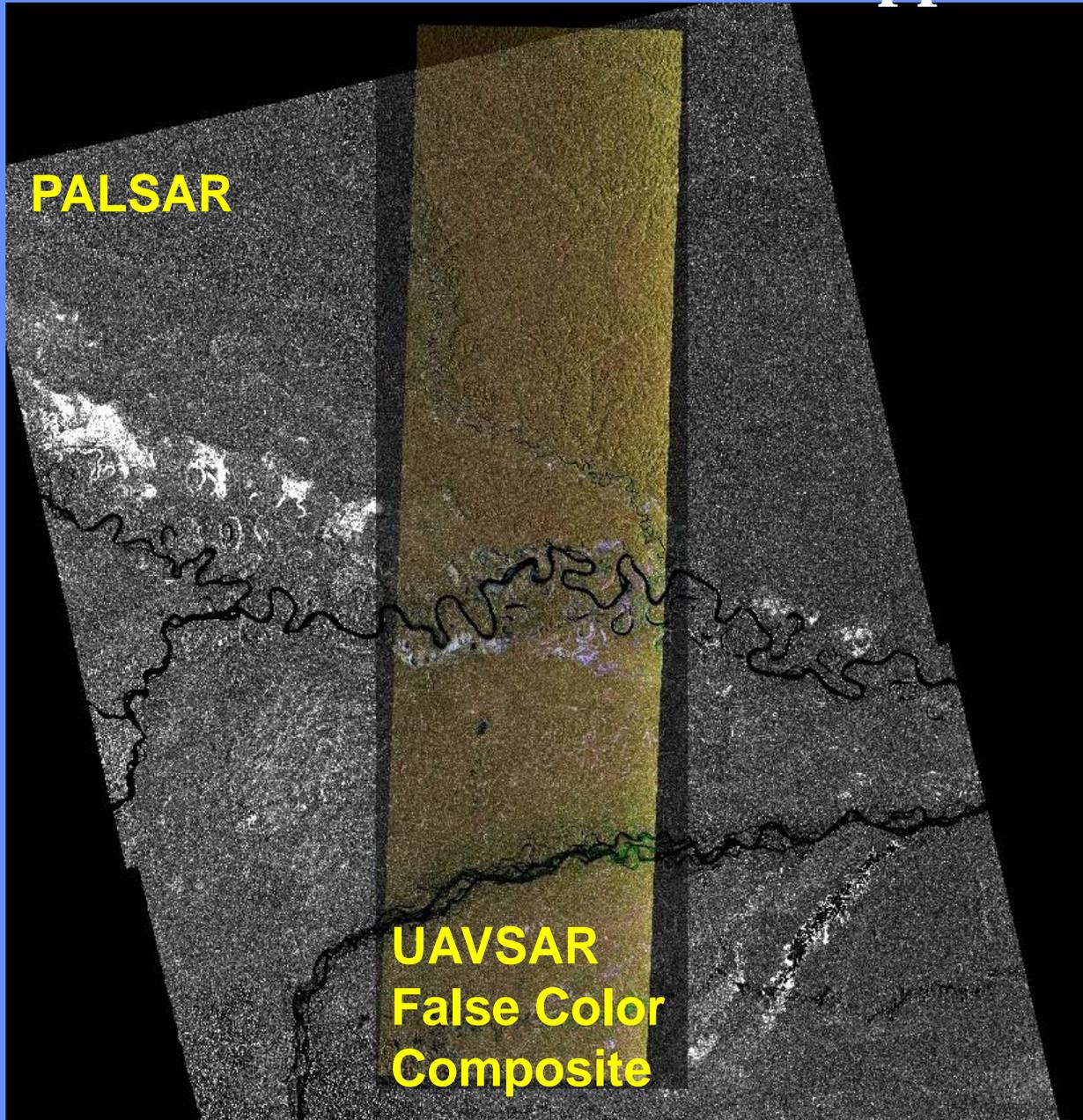
Edge detection algorithm applied to UAVSAR images

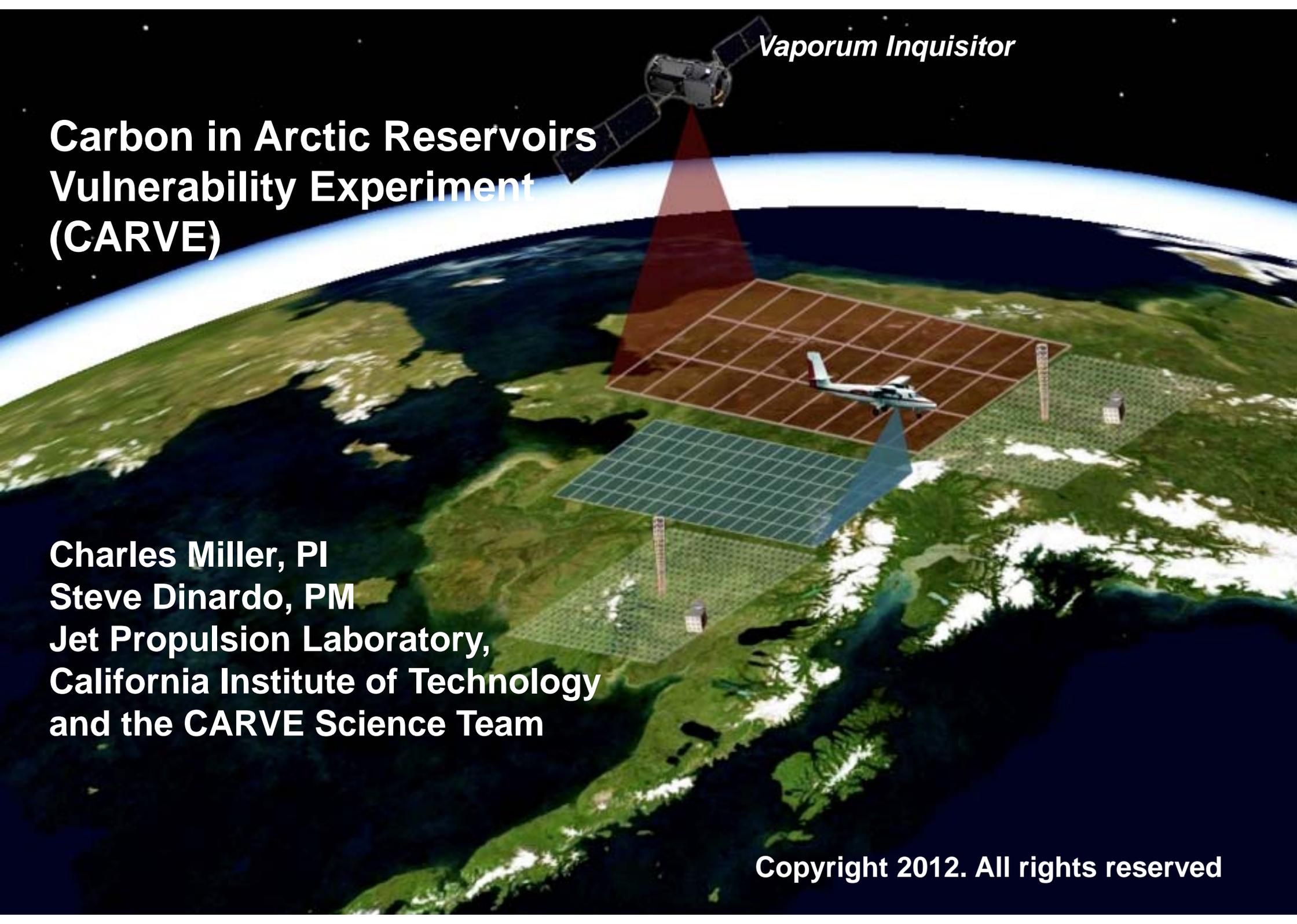
Manual clean-up applied to define structure





# UAVSAR-PALSAR Fusion Supports Basin-scale Analyses





*Vaporum Inquisitor*

# Carbon in Arctic Reservoirs Vulnerability Experiment (CARVE)

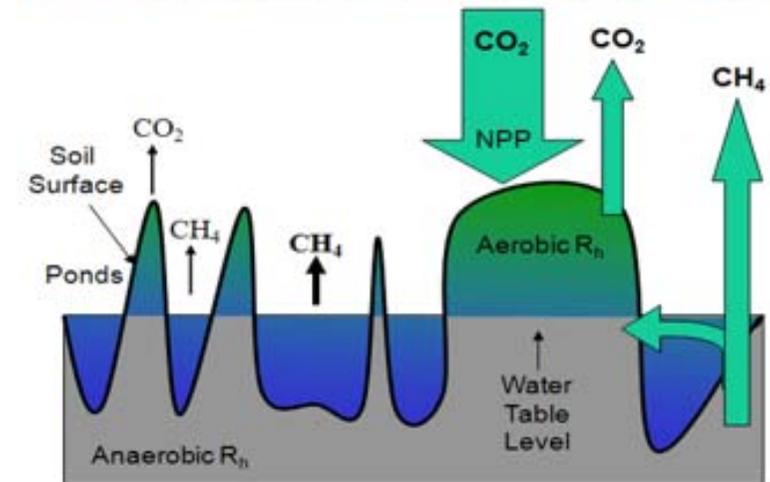
Charles Miller, PI  
Steve Dinardo, PM  
Jet Propulsion Laboratory,  
California Institute of Technology  
and the CARVE Science Team

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## CARVE Addresses

### Key Unanswered Science Questions

- 1) What are the sensitivities of the Alaskan Arctic carbon cycle and ecosystems to climate change?
- 2) How does interannual variability in surface controls (e.g., soil moisture) affect landscape-scale atmospheric concentrations and surface-atmosphere fluxes of  $\text{CO}_2$  and  $\text{CH}_4$  in the Alaskan Arctic?
- 3) What are the impacts of fire and thawing permafrost on the Alaskan Arctic carbon cycle and ecosystems?

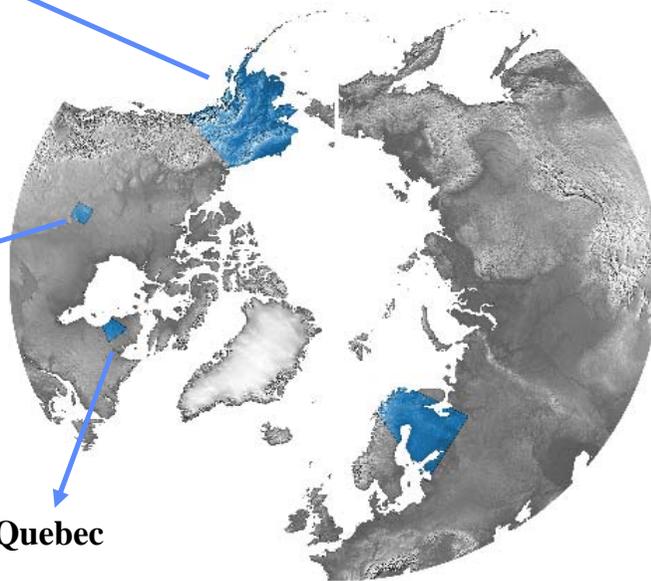


A mosaic of wet and dry areas is common for regions in the Arctic. Microtopography dictates the partitioning of soil respiration into aerobic processes ( $\text{CO}_2$  release) and anaerobic processes ( $\text{CH}_4$  release). The partitioning of carbon fluxes from Arctic ecosystems is not known accurately.

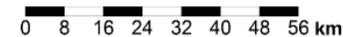
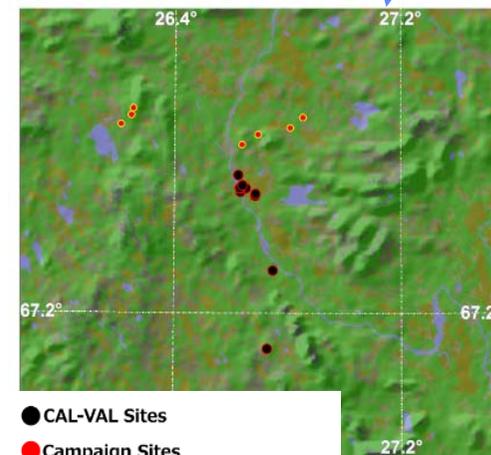
## Boreal/Arctic Freeze-Thaw



SMAP F/T Domain



Sodankylä, Finland



BERMS (Saskatchewan)



Old jack pine (OJP)  
at (53.92 N, 104.69 W)

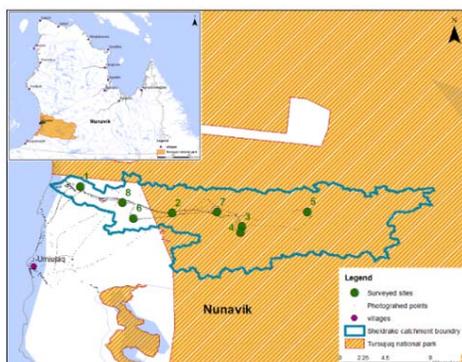
Young jack pine (YJP)  
also called H02  
at (53.95 N, 104.65 W)

Old black spruce (OBS)  
at (53.99 N, 105.12 W)

Fen (Fen)  
at (53.78 N, 104.62 W)

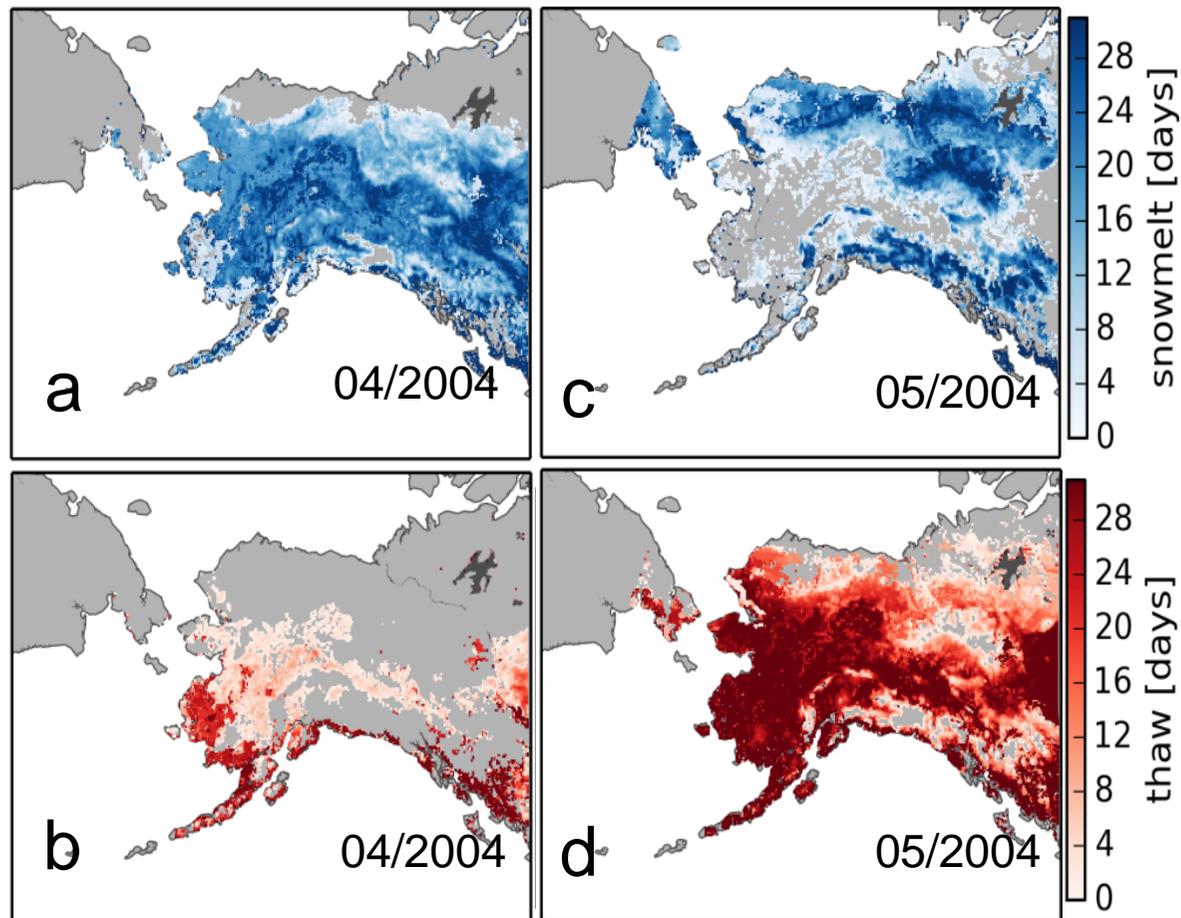
Mixed forest (Temp7)  
at (53.90 N, 104.88 W)

Umiujaq, Quebec

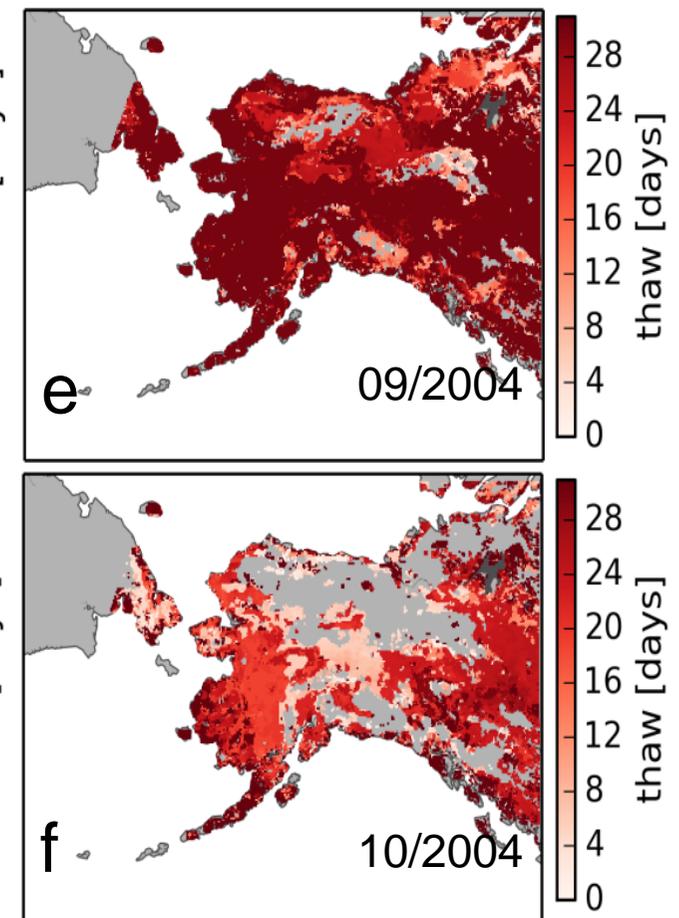


## Multi-frequency Freeze/Thaw algorithm development

Thaw Progression



Refreeze Progression



## Project Deliverables

- Updated wetlands maps from SAR datasets, providing 30-year record (Alaska, South America, Africa, Great Lakes)
- Global 25km resolution inundated area maps over 1992-2013 and updated annually, validated and assessed using PALSAR, PALSAR 2 and JERS SAR
- Wetlands map of the Chesapeake Bay estuary region.
- Biome maps of the Amazon basin and Brazilian Atlantic Coastal Forest.
- Prototype multi-sensor freeze/thaw state algorithm and maps for boreal study areas
- Delineation of contemporary land cover (vegetation) related to carbon and climate zonation for focused regions of the Amazon basin and the Brazilian coastal rain forest based on PALSAR and PALSAR 2.