

### **Mapping Industrial Forests in Tropical Monsoon Asia**

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### **Acknowledgement:**

A network of international collaborators for studying in tropical monsoon Asia

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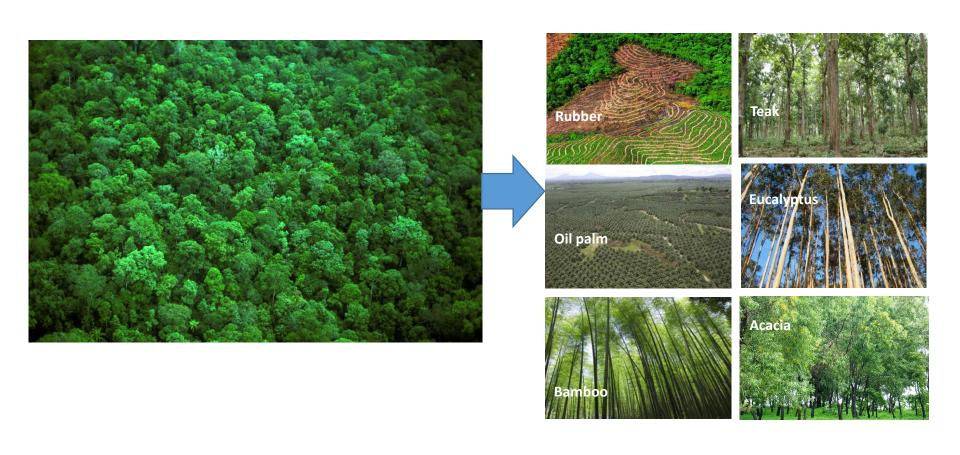
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### Why care about industrial forests?

Increasing tropical forests converted into industrial forests

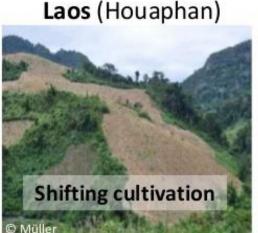


### Why care about industrial forests in monsoon Asia?

Most of the global forest conversion from natural to industrial forests happened in tropical monsoon Asia.

e.g., ~90% global palm oil ~97% global natural rubber production (Li and Fox, 2012)









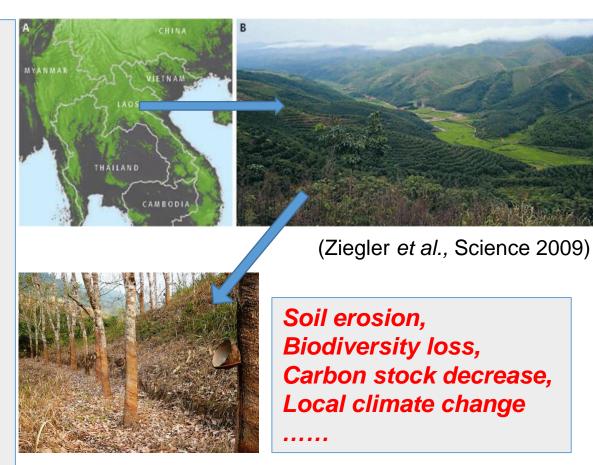
(Muller, 2012)

### Rubber plantation expansion in tropical monsoon Asia

#### Some facts

- >95% of global rubber plantations from Southeast Asia
- Rubber production increased by more than 1,500 % from 1961-2011
- By 2020, almost 640,000 hectares of protected areas will be converted to rubber trees

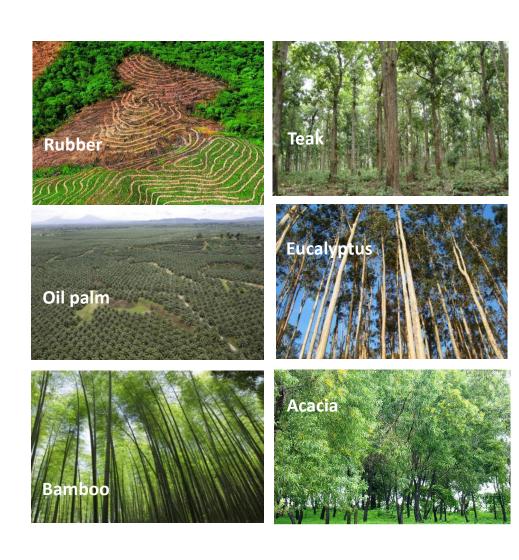
(Ahrends et al., GEC 2015)



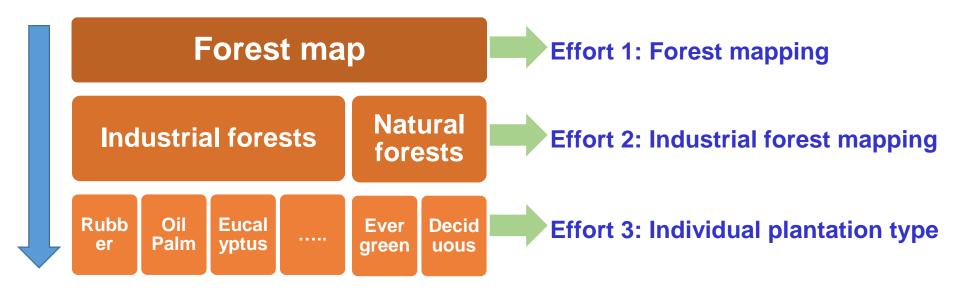
However, industrial forest maps are very limited

### Project objective

to map the major industrial forest plantations in tropical monsoon Asia and quantify their changes



### From forest mapping to industrial forest mapping



### How well existing forest maps support conservation policy?



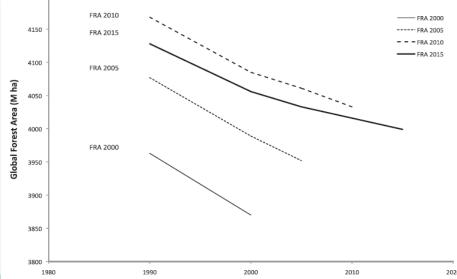


Fig. 2. Trends in global forest area 1990-2015 in Forest Resources Assessments (FRAs) 2000-2015.

FAO FRA 2000/2005/2010/2015

(Keenan et al., FEM, 2015)

Great discrepancy in existing global forest maps (Sexton *et al.,* Nature Clim Chang 2015)

estimates of global forest area range from 32.1×106 to 41.4×10<sup>6</sup> km<sup>2</sup>

Previous satellite-based

### How well existing forest maps support conservation policy?

 A continued sharp decrease of net forest area losses with -3.3 Mha/yr from ca. 2010 to ca. 2015, mainly in the tropics.

----FAO FRA 2015

Tree cover loss is rapidly accelerating in the tropics in 2014.

## Uncertainty in forest dynamics

## Baseline Map of Carbon Emissions from Deforestation in Tropical Regions

Nancy L. Harris, 1\* Sandra Brown, 1 Stephen C. Hagen, 2 Sassan S. Saatchi, 3,4 Silvia Petrova, 1 William Salas, 2 Matthew C. Hansen, 5 Peter V. Potapov, 5 Alexander Lotsch 6

30% of previously published estimates for an overlapping time period. The largest source of uncertainty in our analysis is the estimates of gross forest-cover loss across large regions. This

(Harris et al., Science 2012)

## Objective 1: Improve forest maps through integration of optical and microwave images

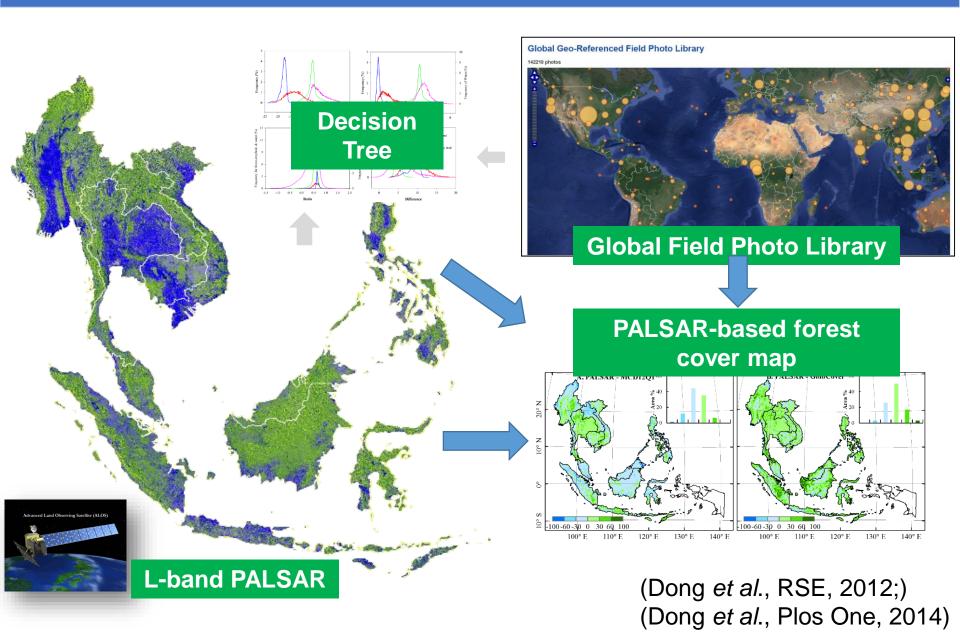
- □ Existing forest mapping efforts are generally based on optical RS.
- □ Advantages of L-band PALSAR has been recognized, but limitedly used due to data availability.

Hypothesis: Inclusion of both forest structure and canopy information will help to improve forest mapping

Optical RS: sensitive to vegetation/foliage cover L-band PALSAR: sensitive to structure (e.g., tree height, biomass) and vegetation/surface moisture conditions

### Cloud-free PALSAR based forest mapping

(Ver\_0)



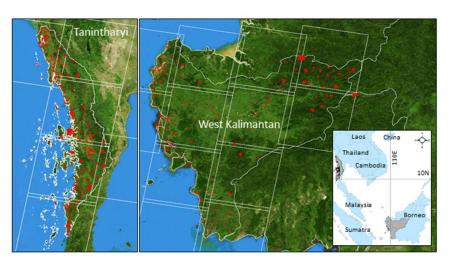
### Improved PALSAR+Optical RS based forest mapping (Ver\_1)

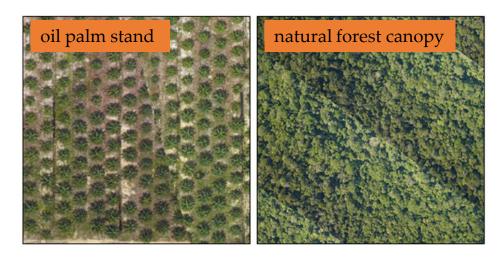
PALSAR-based Max NDVI-based **Forest Map Forest Layer Vegetated Layer** Fraction of forest cover Vegetation and bare land Vegetation and bare lands Non-Forest Vegetation cover Forest Country B. NDVI<sub>max</sub> C. Forest map A. PALSAR FNF

(Qin et al., Scientific Report, 2016)

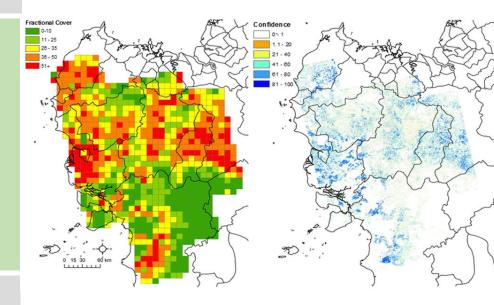
# Objective 2: Identify and map industrial forests through integration of optical and microwave images

Hypothesis: Industrial forests/monoculture plantations are featured with structured planting and smooth/uniform canopy, in comparison to complex canopy and multi-layered structure from natural forests





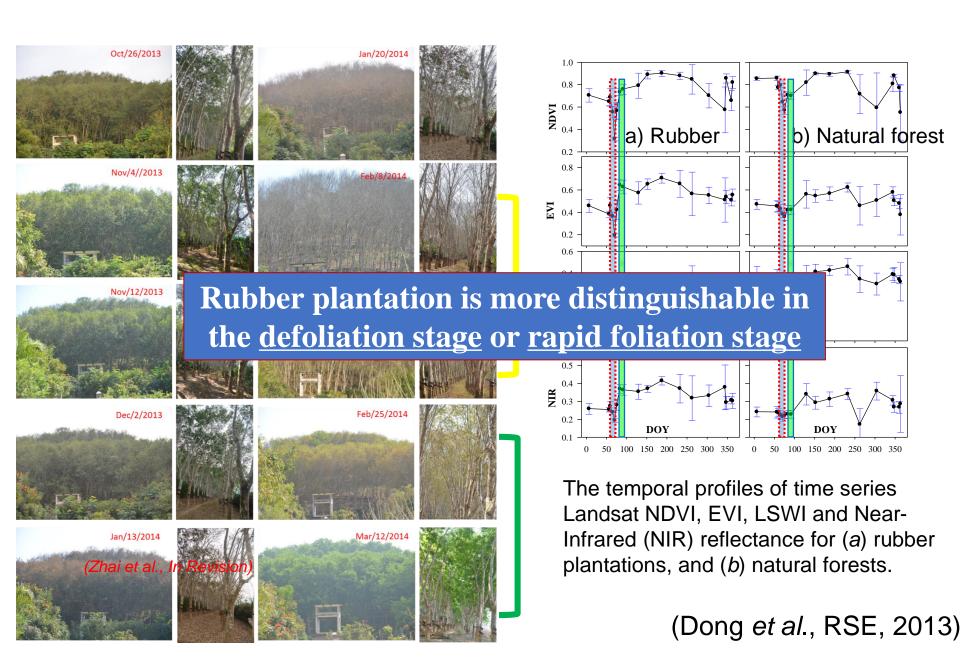
- Landsat 8, PALSAR 2, Sentinel-1
- Common ratios (HH/HV²)
- Vegetation indices (NDVI, LSWI, SATVI, and NDTI)
- Texture indices (gray-level cooccurrence matrix, GLCM), included mean, variance, homogeneity, contrast, dissimilarity, entropy, second moment, and correlation
- Classification and Regression Tree-Random Forest (CART-RF approach)



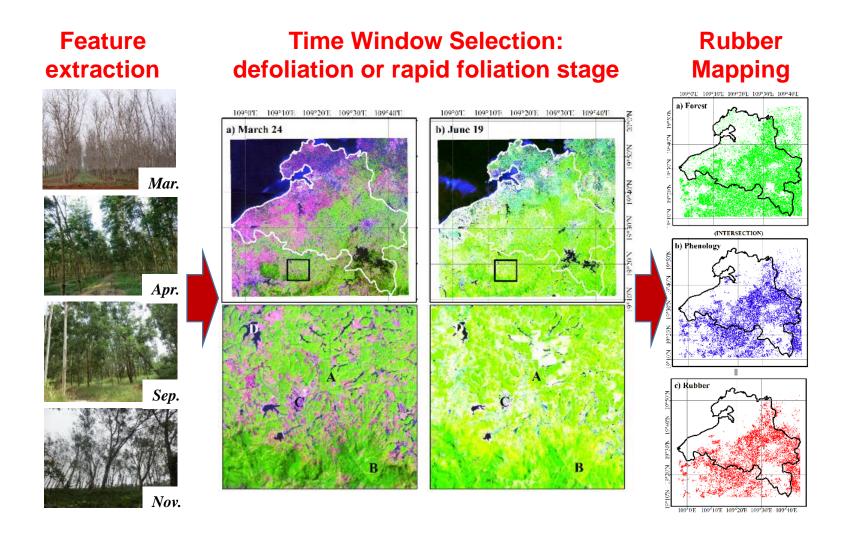
(Torbick et al., Remote Sensing, 2016)

# Objective 3: Identify and map individual industrial forest types through integration of optical and microwave images

**Hypothesis:** Deciduous rubber plantations have unique phenology feature in its canopy, which can be tracked by time series optical images.



### Rubber plantation mapping using phenology-based approach

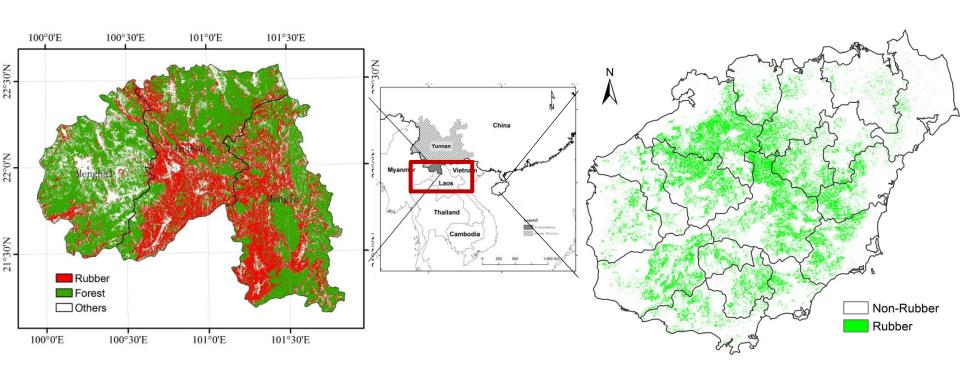


(Dong et al., RSE, 2013)

### Rubber plantation maps in two rubber production hotspots

### Xishuangbanna

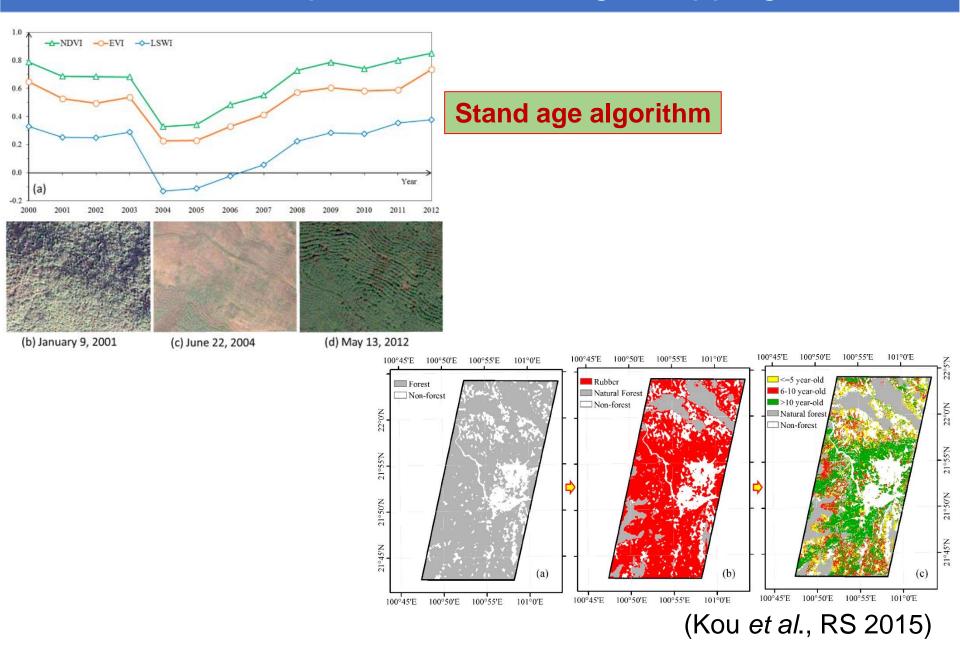
### Hainan Island



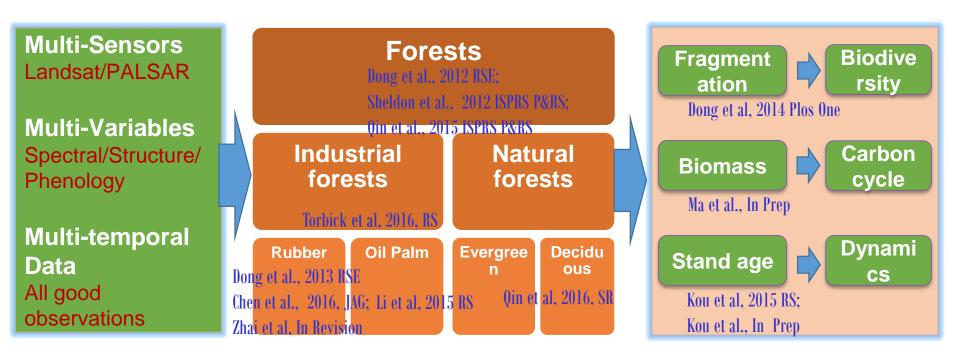
(Kou et al., In Review)

(Chen et al., Int J Appl Earth Obs, 2016)

### Rubber plantation stand age mapping



### Summary of our efforts on forest/plantation mapping

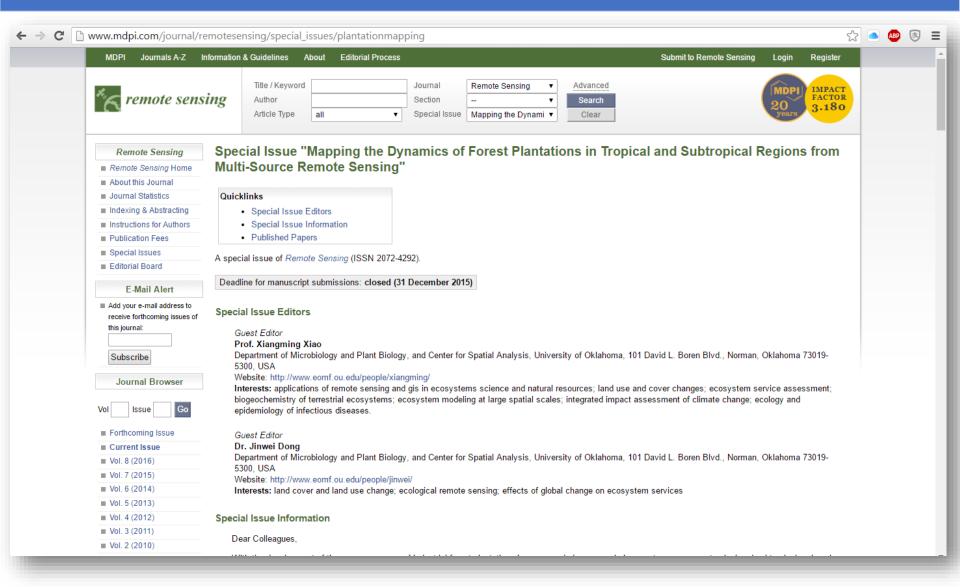


**Methodology** 

**Map Products** 

**Applications** 

### Special Issue about forest plantations on Remote Sensing



### Forest and Plantation RS Session in 2015 AGU Fall Meeting

### **@AGU FALL MEETING**

San Francisco | 14-18 December 2015

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BROWSE PROGRAM

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#### B51L

e Sensing of Forests and Plantations in the Tropical and Subtropical Zones I

Friday, 18 December 2015: 08:00-10:00 2006 (Moscone West)

Primary Conveners:

Xiangming Xiao, University of Oklahoma Norman Campus, Norman, OK, United States

Jinwei Dong, University of Oklahoma Norman Campus, Norman, OK, United States

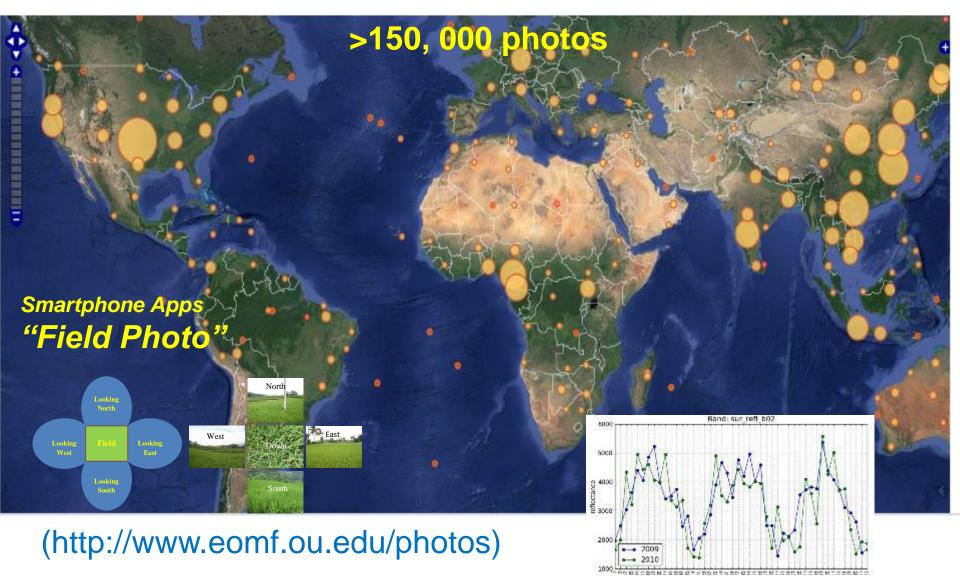
Chairs: Xiangming Xiao, Jinwei Dong and Yuanwei Qin, University of Oklahoma Norman Campus,

Norman, OK, United States

OSPA Liaisons: Xiangming Xiao, University of Oklahoma Norman Campus, Norman, OK, United States

08:00	B51L-01	Dynamics of Industrial Forests in Southeast United States Assessed using Satellite and Field Inventory Data (Invited) (81600) Chengquan Huang, University of Maryland, Department of Geographical Sciences, College Park, MD, United States and NAF-C team
08:15	B51L-02	Changes in rubber plantation in the cross-border area of mainland Southeast Asia through analysis of PALSAR and time series Landsat images (79275)  Jinwei Dong¹, Xiangming Xiao¹, Yuanwei Qin¹, Bangqian Chen², Weilī Kou³, Deli Zhai⁴, Geli Zhang⁵, Yao Zhang¹, Yuting Zhou¹ and Jie Wang¹, (1)University of Oklahoma Nomman Campus, Norman, OK, United States, (2)Organization Not Listed, Washington, DC, United States, (3)University of Oklahoma, Norman, OK, United States, (4)Kunming Institute of Botany, Chinae, 60 Academy of Sciences, Kunming, Chinae, (5)University of Oklahoma, Center for Spatial Analysis, Norman, OK, United States
08:30	B51L-03	Tropical-Forest Biomass Dynamics from X-Band, TanDEM-X DATA (Invited) (63944) Robert N Treuhaft, NASA Jet Propulsion Laboratory, Pasadena, CA, United States
08:45	B51L-04	Topographic Distributions of Emergent Trees in Tropical Forests of the Osa Peninsula, Costa Rica (Invited) (76240) Chris Balzotti <sup>1,2</sup> , Gregory Paul Asner <sup>1,2</sup> , Phillip Taylor <sup>3</sup> , Rebecca J Cole <sup>4</sup> , Brooke B Osborne <sup>5</sup> , Cory C. Cleveland <sup>9</sup> , Stephen Porder <sup>7</sup> and Alan R Townsend <sup>3</sup> , (1)Carnegie Institution for Science Washington, Washington, DC, United States, (2)Carnegie Institution for Science, Department of Global Ecology, Stanford, CA, United States, (3)University of Colorado at Boulder, Boulder, CO, United States, (4)University of Hawaii at Manoa, Honolulu, HI, United States, (5)Brown University, Department of Ecology and Evolutionary Biology, Providence, RI, United States, (6)University of Montana, Missoula, MT, United States, (7)Brown University, Providence, RI, United States
09:00	B51L-05	Reconstructing Land Use History from Landsat Time-Series. Case study of Swidden Agriculture Intensification in Brazil (69205) Loic Durrieux, Catarina C Jakovac, Latifah H Siti and Lammert Kooistra, Wageningen University, Wageningen, Netherlands
09:15	B51L-06	Tracing Forest Change through 40 Years on Two Continents with the BULC Algorithm and Google Earth Engine (85023)  Jeffrey A Cardille, McGill University, Natural Resource Sciences, Montreal, QC, Canada
09:30	B51L-07	Using CLASlite to Map Deforestation in Makira Natural Protected Area, Madagascar (79928) Alison Nicole Thieme <sup>1</sup> , Sean McCartney <sup>1</sup> , John Rogan <sup>1</sup> , Florencia Sangermano <sup>1</sup> and David Wilkie <sup>2</sup> , (1) Clark University, Worcester, MA, United States, (2) Wildlife Conservation Society, Bronx, NY, United States

### Cal/Val LC database: Global Geo-Referenced Field Photo Library



(Xiao *et al.,* EOS, 2011)

Individual photos are linked with time series MODIS data (2000-present)

### Summary

- 1. Cloud-free PALSAR data can effectively contribute to the existing forest mapping efforts. For example, the combined Landsat/PALSAR-based forest mapping approach is more robust.
- 2. Unique phenology signature makes rubber plantation identification possible in a simple way; however, its extensive application in other regions and other species should consider different factors.
- 3. Emerging data sources (e.g., Landsat 8, Sentinel-1, Sentinel-2, and PALSAR-1/2) provide unprecedented opportunities for industrial forest mapping; however, new algorithms and computing technologies (e.g., cloud computing) are needed to make full use of these big data.

### Work plan in 2016

- 1. Improved forest maps by integrating PALSAR and Landsat/Sentinel-2.
- 2. Extensive application of the rubber plantation approach on tropical monsoon Asia.
- 3. Development of new mapping algorithms for other industrial forest types.
- 4. Enhanced field photo and ground truth data portal (freely open to the public).

