





Multi-resolution quantification and driver assessment of hot spots of global forest disturbance

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Using PlanetScope data to quantify forest loss area: Peru prototype study





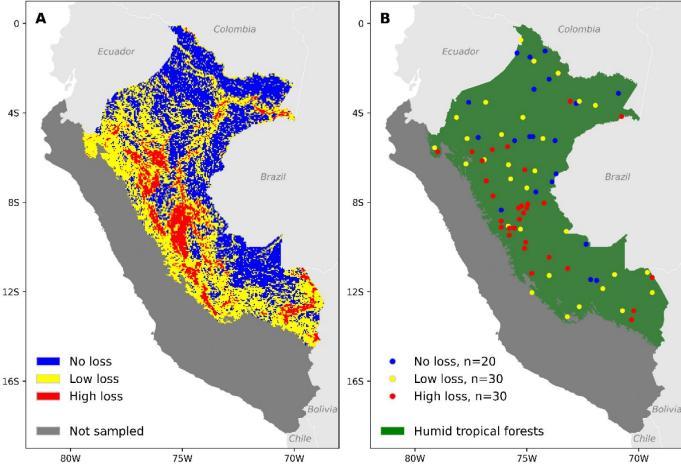


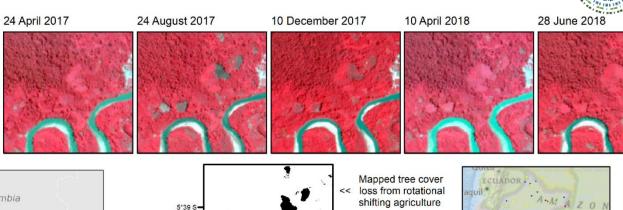
Pickering et al. (2021)

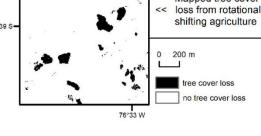
Communication

Using Multi-Resolution Satellite Data to Quantify Land Dynamics: Applications of PlanetScope Imagery for Cropland and Tree-Cover Loss Area Estimation

Jeffrey Pickering *, Alexandra Tyukavina ⁽⁰⁾, Ahmad Khan, Peter Potapov, Bernard Adusei, Matthew C. Hansen ⁽³⁾ and André Lima









- ~3 m reference data (PlanetScope)
- 80 5x5 km reference blocks
- Reference maps loss/no loss for 2017-2018
- Natural vs. anthropogenic forest loss estimates

Results

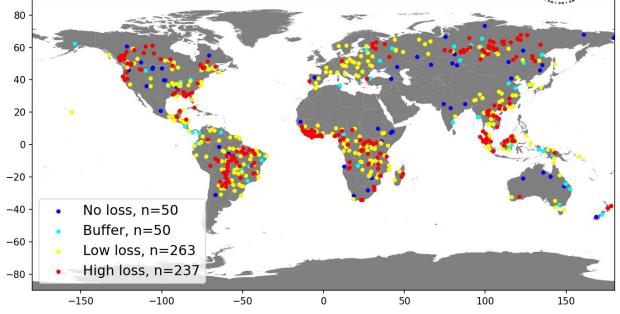
- 9% SE of total forest loss area estimate
- 8% SE anthropogenic loss area
- Regression estimate: 5% SE of total loss area

Current NASA LCLUC-funded project on hotspots of global forest loss

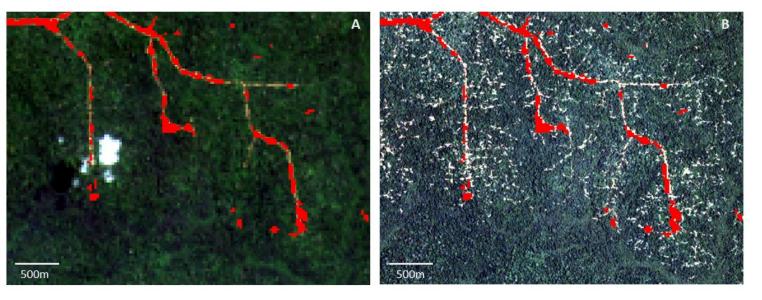




- Global sample: 600 5x5km blocks, stratified random sample;
- Reference satellite data: PlanetScope (~3m) and Sentinel-2 (10m);
- Time interval: 2018 loss + loss drivers based on 3 years of imagery after disturbance.



Locations of 5x5km sample blocks, colored by sampling stratum



Red – global forest loss map

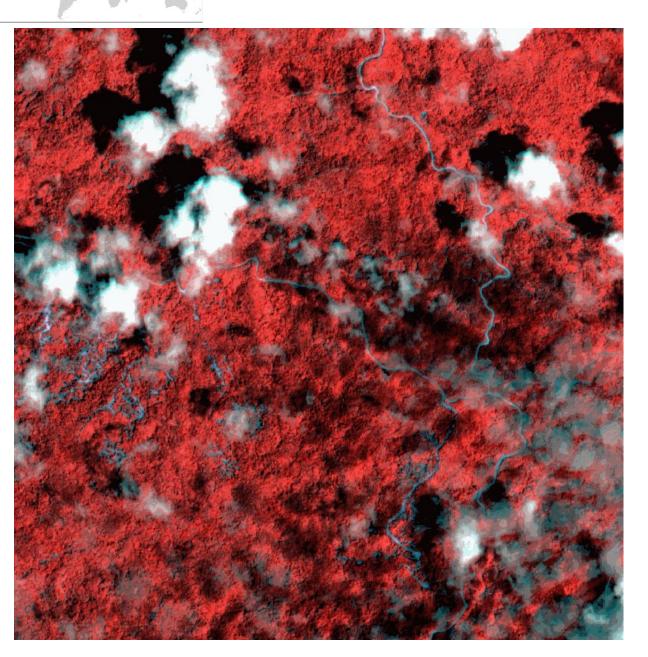
left (A) – over the best cloud-free post-disturbance Landsat observation

right (B) – over the PlanetScope image concurrent with disturbance









- One-two minimally cloudy PlanetScope images per month Dec. 2017 - Jan. 2019 + all available Sentinel-2 images for each 5x5 km block;
- Stacking all images (no cloud filtering);
- Supervised classification (decision trees) of each block separately: yes/no year 2018 forest loss;
- Iterative quality assessment and map improvement process;
- Point-based validation of resulting reference block maps (upon completion of all block maps);
- Visual interpretation of PlanetScope basemaps + Landsat + GoogleEarth to assign each reference loss pixel to pre-disturbance forest type and loss driver

Image stack example, training data and classification result for a cloudy block in Malaysian Borneo: Dec. 2017 – Jan. 2019, 21 PlanetScope images + (not pictured) 30 Sentinel-2 images

Drivers of forest loss: direct (proximate) drivers vs. underlying causes





"Proximate causes *<direct drivers>* are <u>human activities</u> or immediate actions at the local level ... that originate from intended land use and directly impact forest cover"

• Infrastructure extension



• Agricultural expansion



• Wood extraction



"Underlying driving forces <causes> are fundamental <u>social</u> <u>processes</u> ... that ... either operate at the local level or have an indirect impact from the national or global level"

• Demographic factors



Economic factors



Technological factors



• Policy & Institutional factors



Cultural factors



• Other factors: social trigger events, natural catastrophes, predisposing environmental conditions

Adapted from: Geist & Lambin (2002) Proximate Causes and Underlying Driving Forces of Tropical Deforestation, BioScience, Volume 52, Issue 2

Drivers of global forest loss: legend

Initial forest cover

- Natural forest
- Timber plantation
- Non-timber plantation
- Palm plantation



Large rubber plantation, palm plantations intermixed with patches of natural forest Côte d'Ivoire, block 372

Initial direct driver of loss

- Direct human clearing
- Mechanical (mechanized)
- Mechanical (manual)
- Flooding (dams)
- Fire

Natural disturbances:

- Floods (natural, e.g. river meandering)
- Insects
- Hurricanes/Windfalls
- Drought
- Earthquakes/Land slides

MARYLAND



Land cover / land use 3 years after disturbance

- Shifting cultivation
- Forestry/Clearcut
- Timber tree plantation
- Non-timber tree plantation
- Palm plantation
- Selective logging
- Cropland
- Pasture
- Settlement
- Commercial construction
- Road
- Mining
- Energy infrastructure
- Flooded (dams)
- Natural disturbances
- Human clearing with uncertain purpose



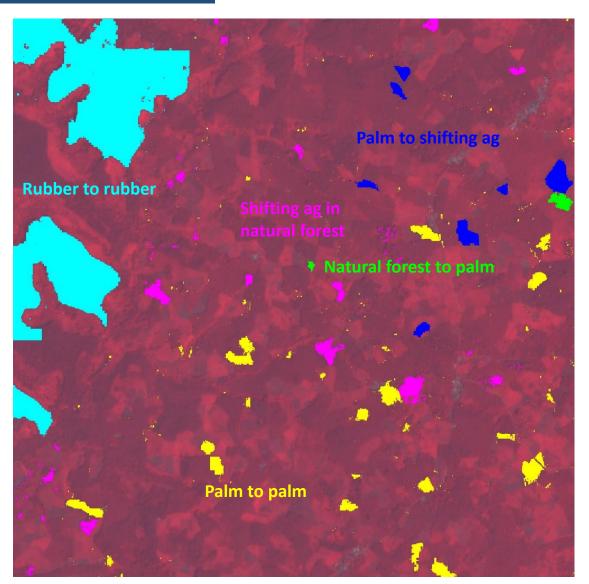
Multiple loss drivers per sample block, Côte d'Ivoire







Very high resolution image (Maxar) 03/05/2021



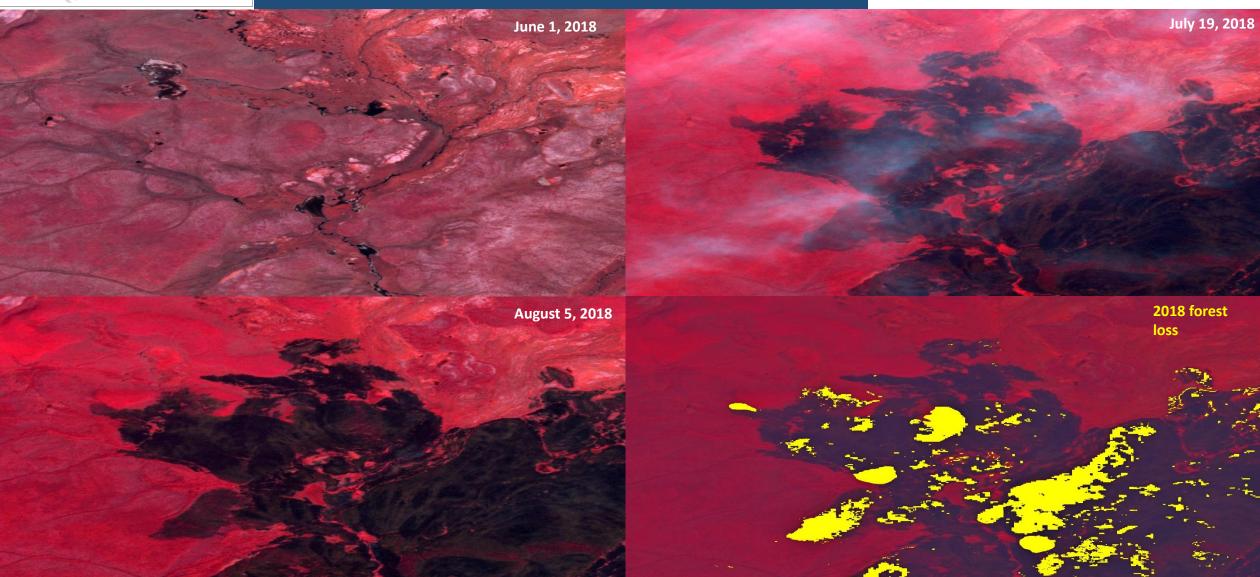
2018 forest loss map based on PlanetScope and Sentinel 2 data



Fire in Sakha Republic, Russia



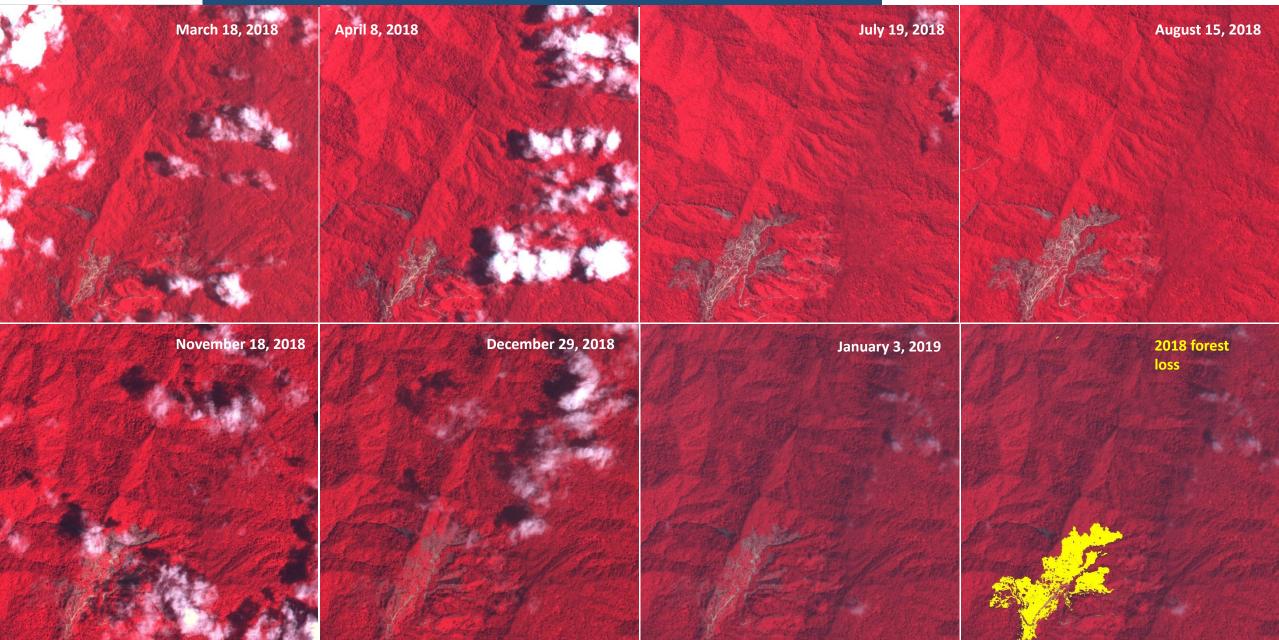




Industrial logging in Malaysia





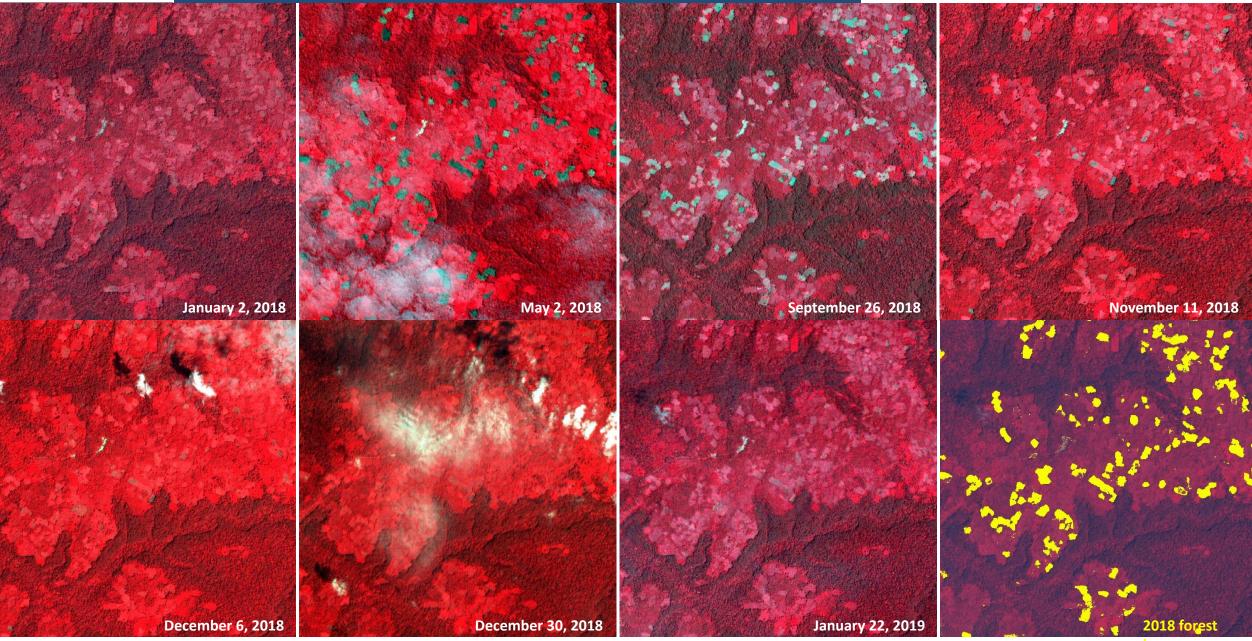




Shifting cultivation in the DRC



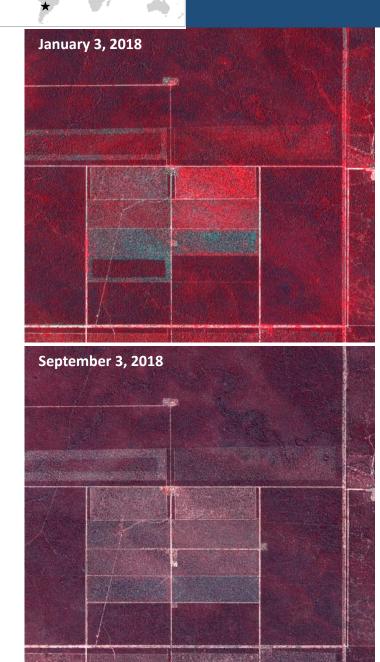


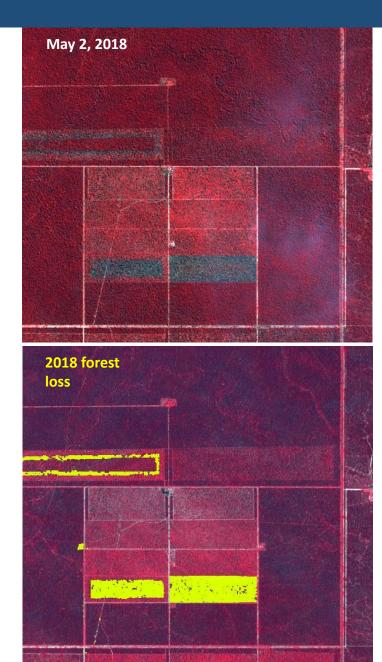


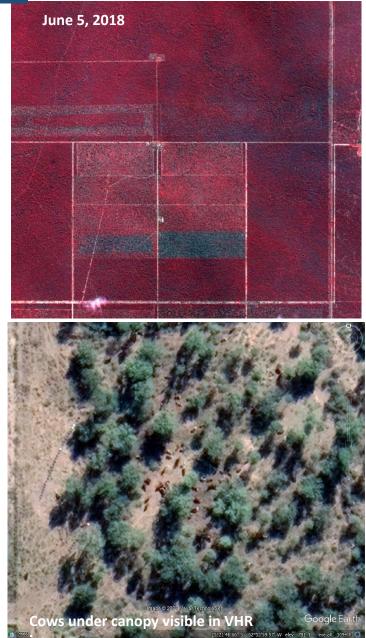
Clearing for pasture in Argentina







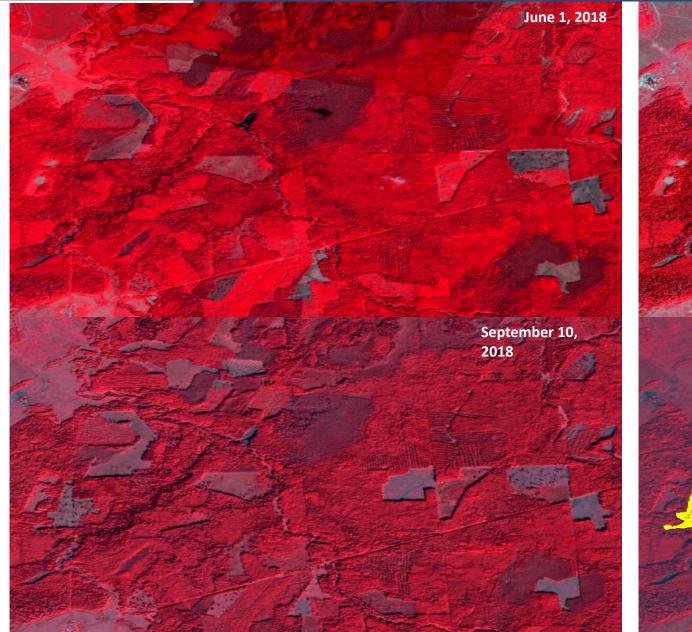


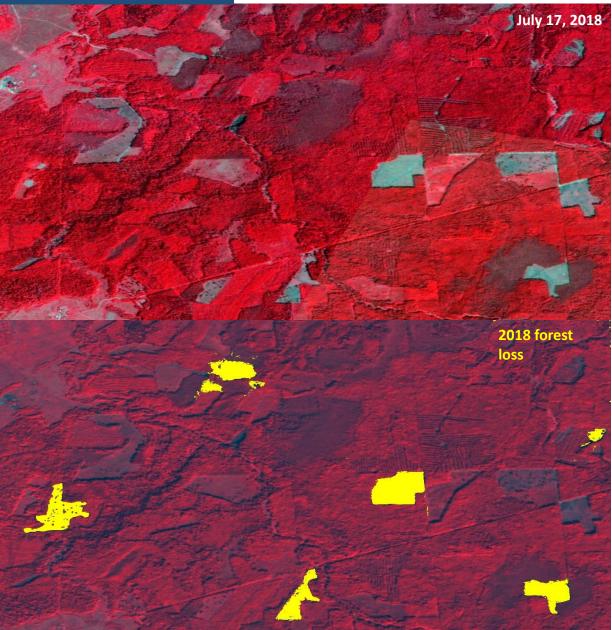


Forestry in Novgorod region, Russia













Timber plantation management (thinning) in Brazil

November 11, 2018







Initial forest disturbance type

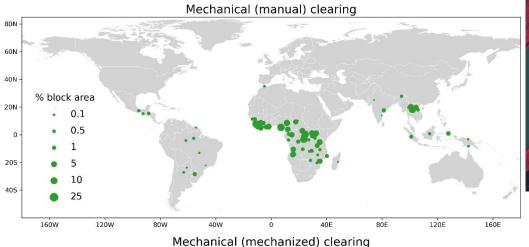
MARYLAND



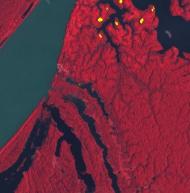
Natural disturbances, wildfires and flooding

Mechanical clearing: manual vs. mechanized

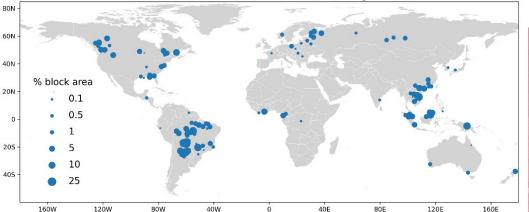
Criteria: clearing size, presence of access roads for machinery + auxiliary information on land use practices



Example of **manual** clearing



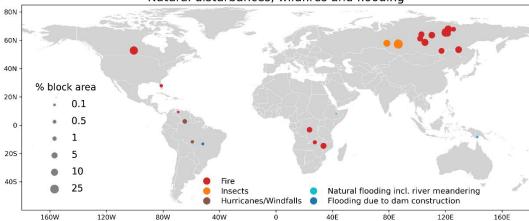
Shifting cultivation, State of Amazonas, Brazil



Example of mechanized clearing



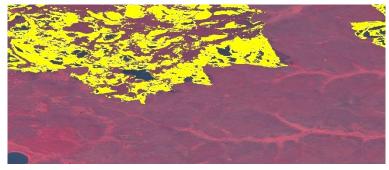
Clearing for pasture, Paraguay



Example of fire

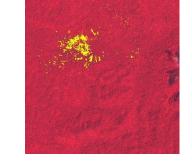
Example of insect damage

Tomsk region, Russia



Sakha Republic, Russia

Example of **windfalls/hurricanes**



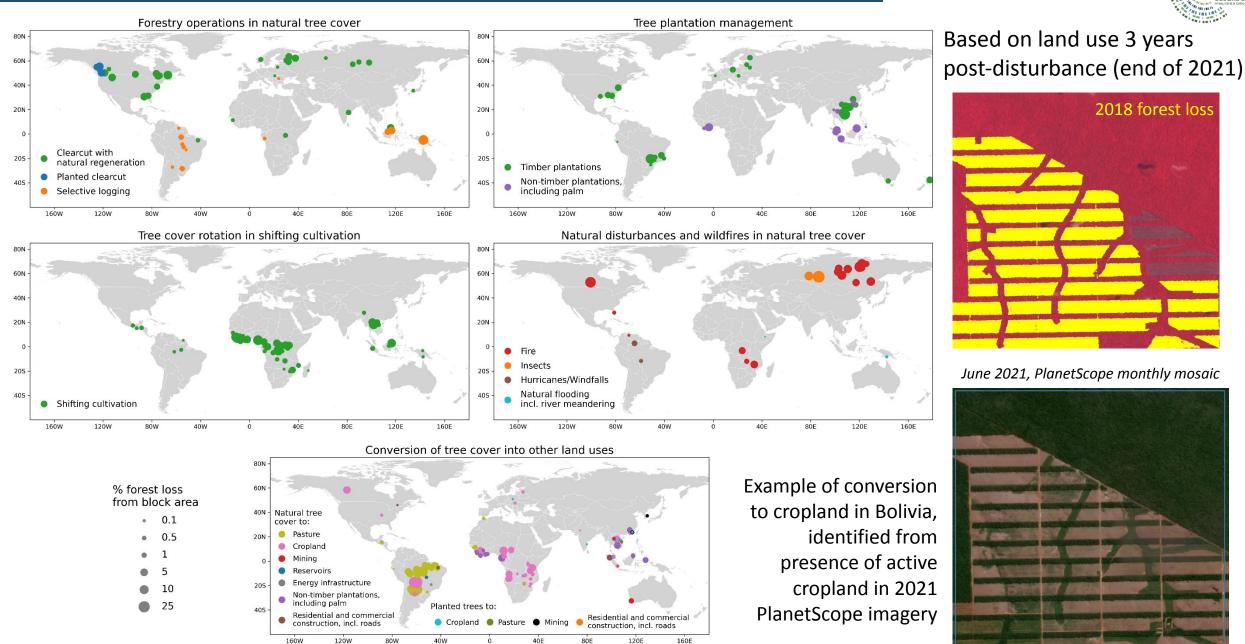
Amazonas state, Venezuela

Proximate cause (direct driver) of forest loss

UNIVERSITY OF MARYLAND

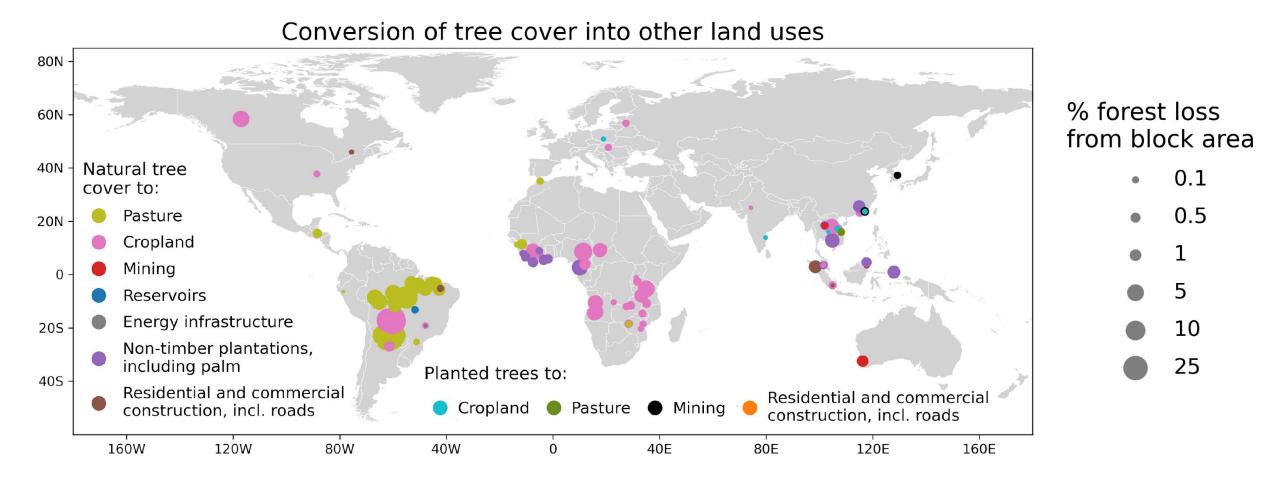


2018 forest loss









Project timeline





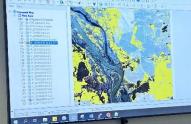
Project task	2021				2022				2023			2024	
	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1	Q2	Q3	Q4	Q1
1. Finalize stratification and sample selection for baseline loss year (2018)				6 20									
 Select, download and pre-process 2018 PlanetScope and Sentinel-2 data for all sampled blocks 													
3. Classify forest loss from high resolution reference data for each sampled block													
4. Attribute direct drivers of forest loss and pre- disturbance forest type to mapped loss pixels													
5. Perform accuracy assessment of reference forest loss maps													
6. Perform statistical analysis, finalize findings, share data													
7. Write manuscript and submit it for publication in a peer-reviewed journal													
Project reports, presentations at LCLUC Science Team meetings													
Travel to visit research collaborators							Ē					-	
Presentation of preliminary findings at the ESA Living Planet Symposium													
Presentation of final results at the AGU Fall Meeting													

Project task status: prange – completed

green – in process blue – not yet started

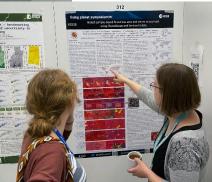
We are here





QA session – checking block mapping results

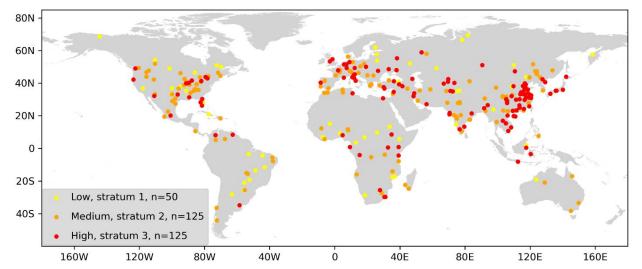
ESA Living Planet Symposium 2022



Current LCLUC project augmentation: CSDA BlackSky evaluation

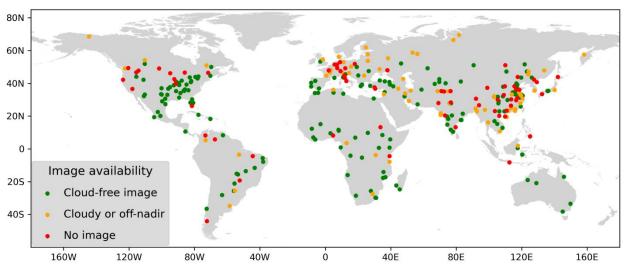






Distribution of selected blocks, n=300, 3 strata

Image availability, May 1st 2023



Purpose:

Employ BlackSky imagery to validate time-series of global change in **built-up area**, which is one of the drivers of global forest disturbance, and estimate the global built-up area in 2022-23.

Study Objectives:

- Estimate built-up land area from the sample of BlackSky and drivers of built-up area change;
- Estimate the accuracy of Landsat- and Planet-based maps.

Study Design:

Stratified random sample of 300 2.5 km blocks; stratification is based on existing Landsat-based built-up maps.

Note from a first-time PI:

It is exciting to task a satellite!





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