



WORLD BANK GROUP
Climate Change

Challenges and Opportunities in Monitoring of Emission Reductions in World Bank Land Use Carbon Finance Programs

Andres Espejo

Forest Carbon Partnership Facility

BioCarbon Fund

May 25, 2017

OUR OBJECTIVES



Promote and reward reduced greenhouse gas emissions and increased sequestration through better land management, climate-smart agriculture, and smarter land use planning and policies.

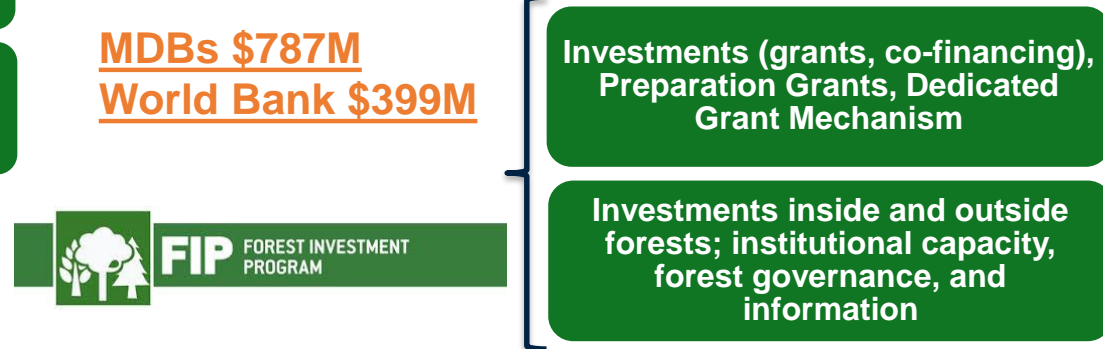
- Integrate sub-national development agenda with low-carbon pathways.
- Support forest countries to maintain and improve **livelihoods**, **conserve biodiversity** and leverage significant private and public sector finance to achieve transformational change.
- **Demonstrate approaches that can be applied nationally** i.e., national low-carbon strategies and global mechanisms of support such as REDD+.

WORLD BANK FOREST CLIMATE FUNDS (\$2.3 BILLION)

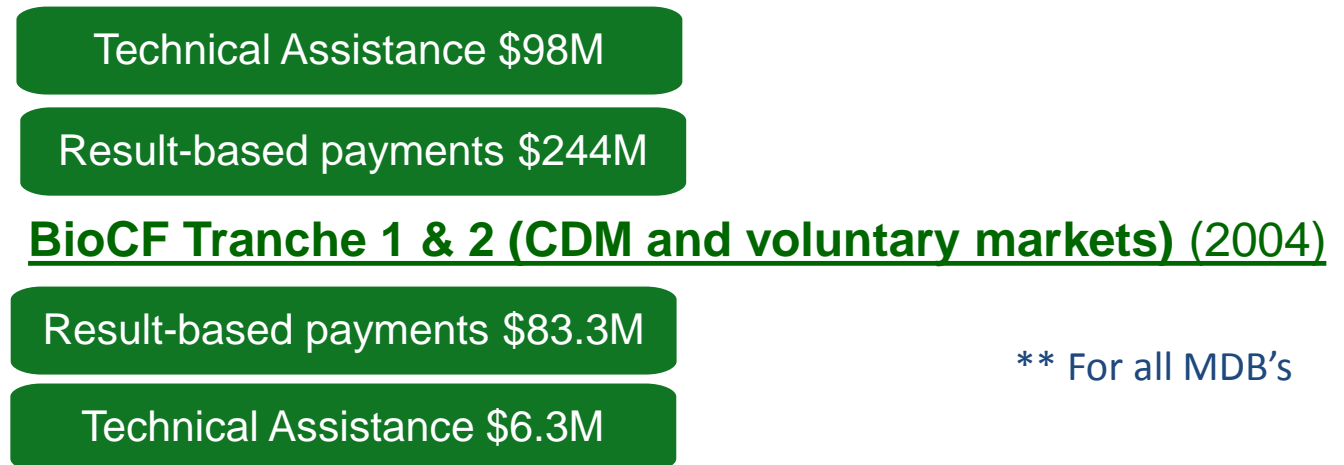
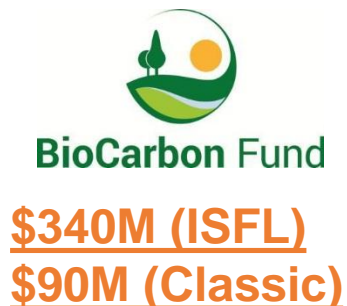
Forest Carbon Partnership Facility (2008)



Forest Investment Program (FIP) (2009)

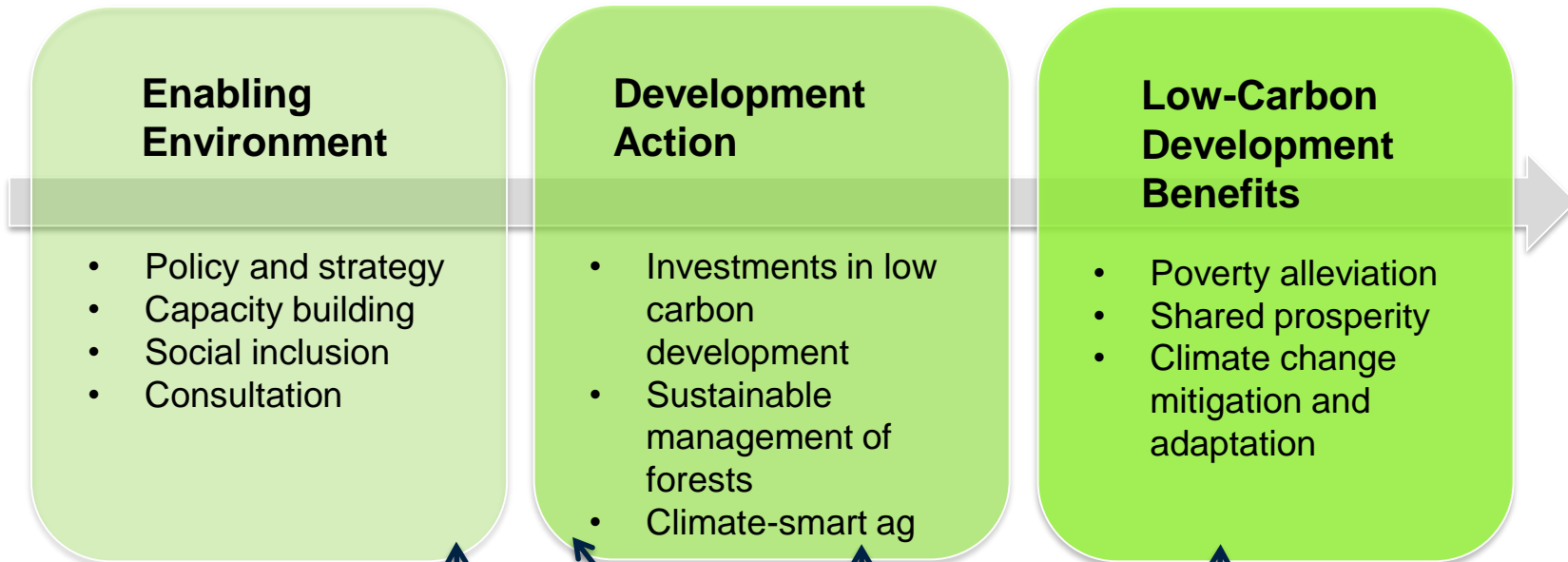


Initiative for Sustainable Forest Landscapes (ISFL)(2013)



** For all MDB's

BUSINESS MODEL



We provide:



**Grant Funding;
Technical Assistance**

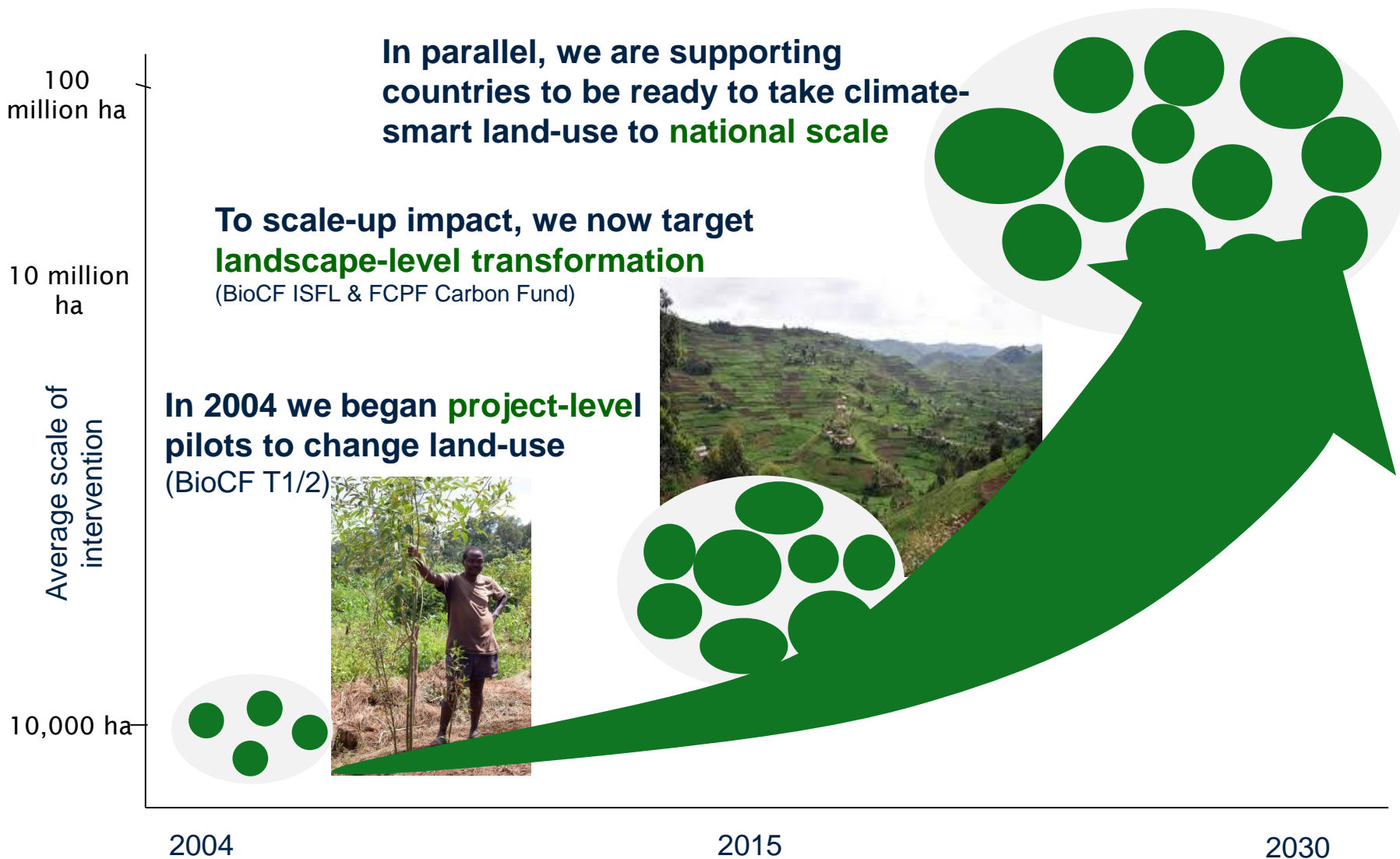
**Results-Based Finance for
Emission Reductions**

We “crowd-in”:

*Private and Public Finance,
including IDA, IBRD, GEF
financing*



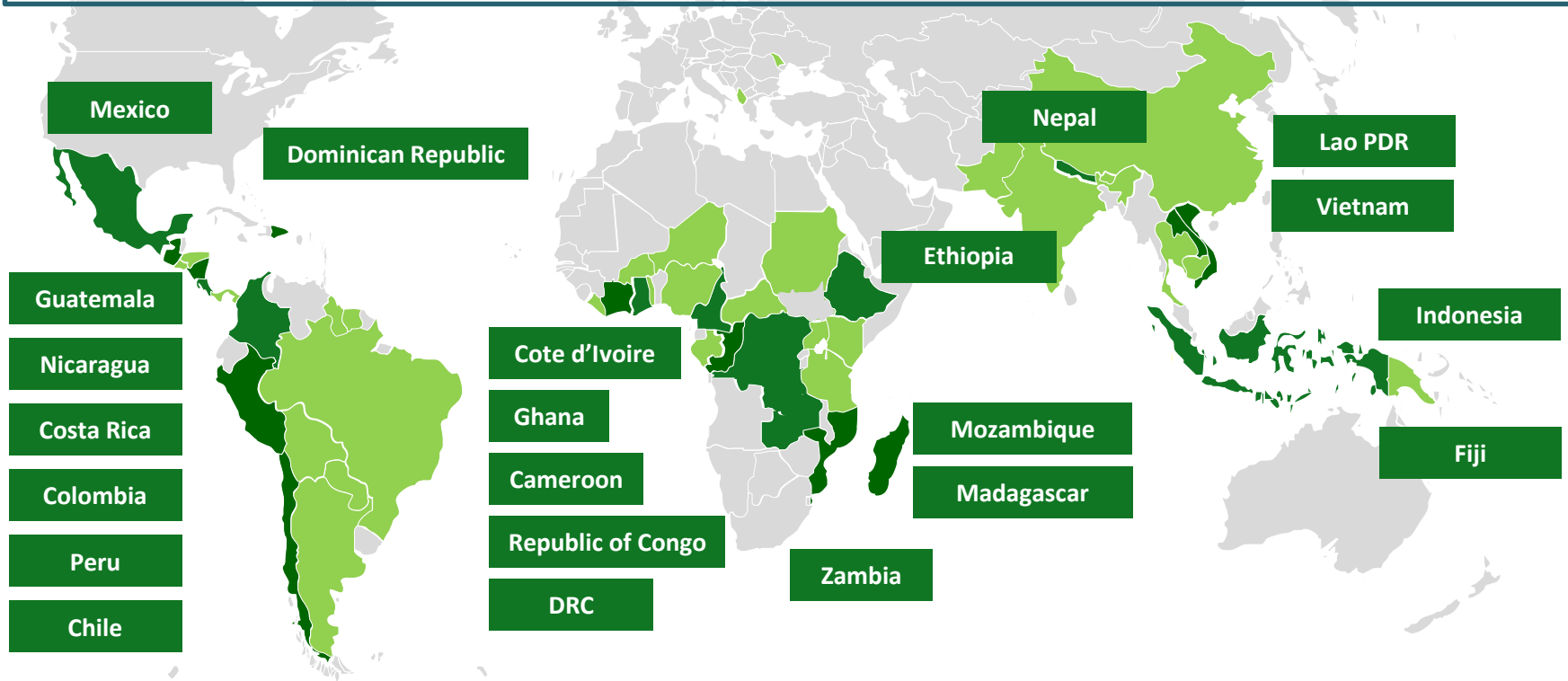
WHY WE ARE SCALING-UP OUR PROGRAMS



WHERE WE WORK

 **22 countries** with large scale climate-smart land-use programs

 **54 countries** with REDD+ readiness support or projects



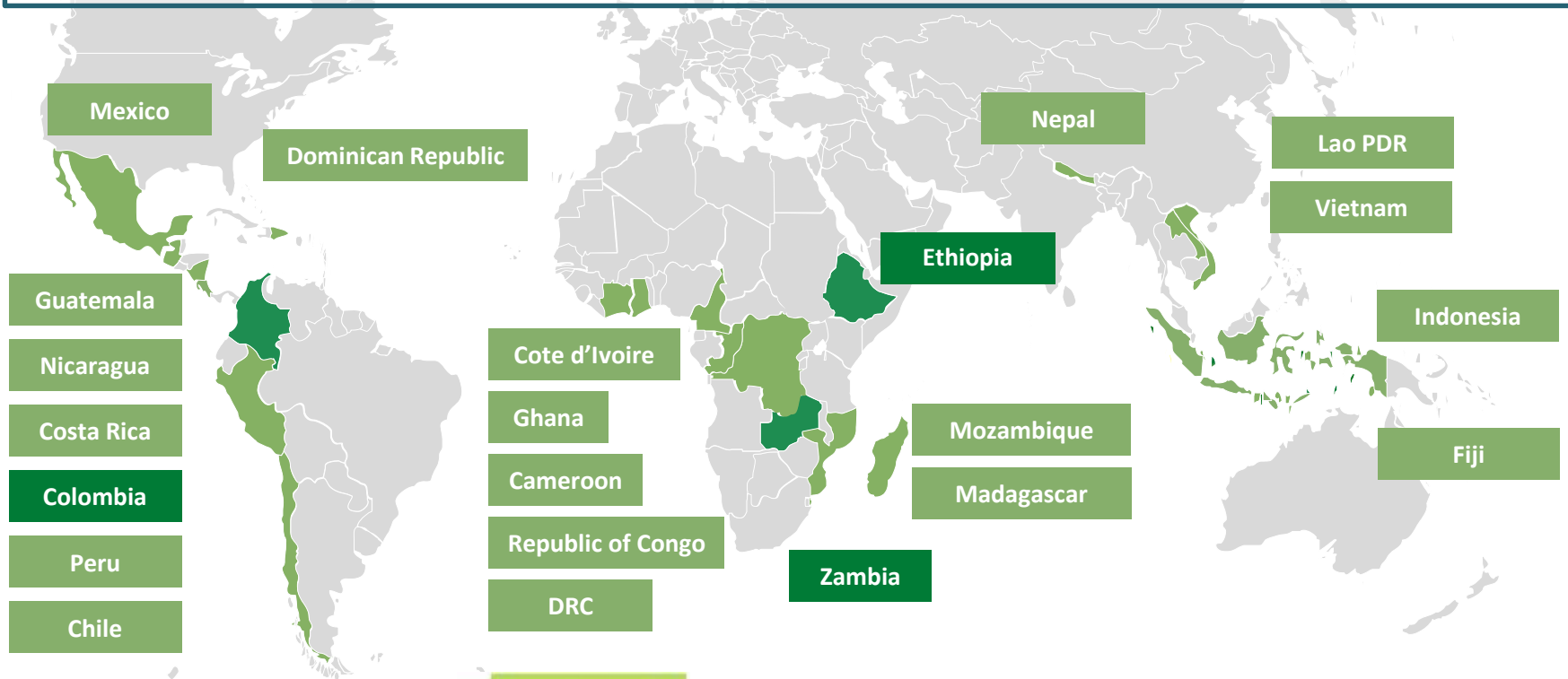
WHERE WE WORK – LAND USE PROGRAMS



19 ER programs under FCFP Carbon Fund



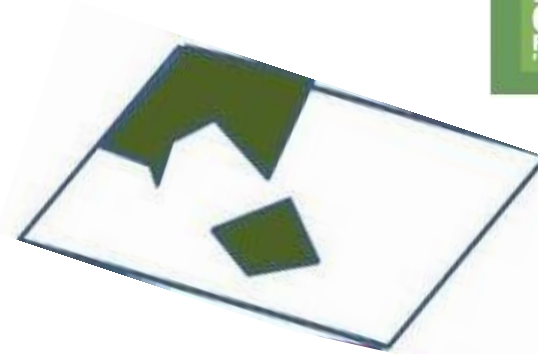
3 ER programs under Carbon Fund - ISFL



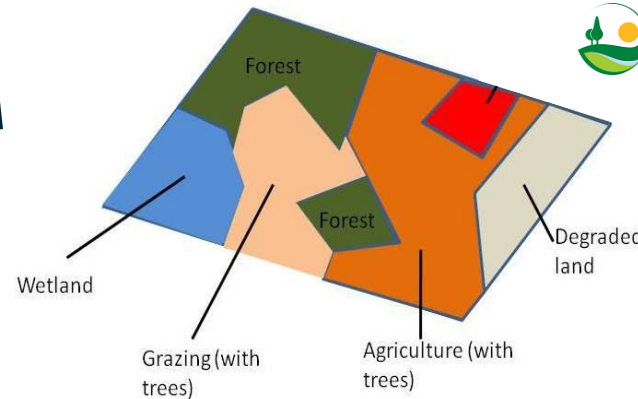
DIFFERENCE BETWEEN FCPF CF AND ISFL



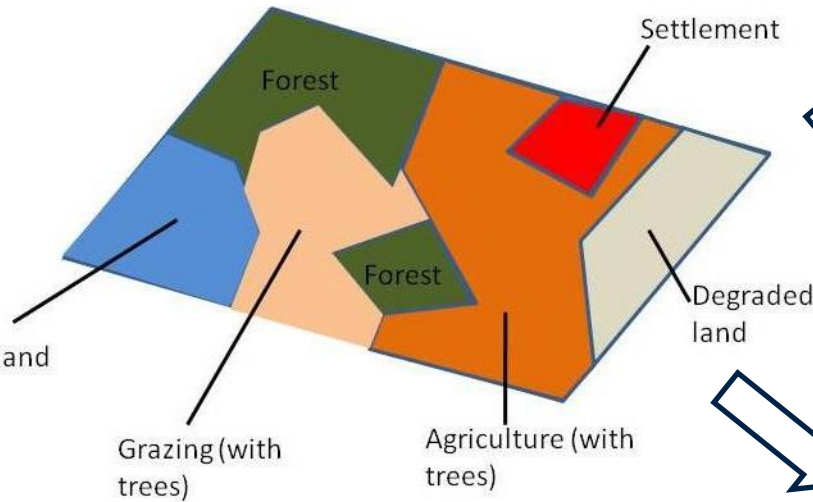
Carbon Fund



REDD+ \approx Forestry sector

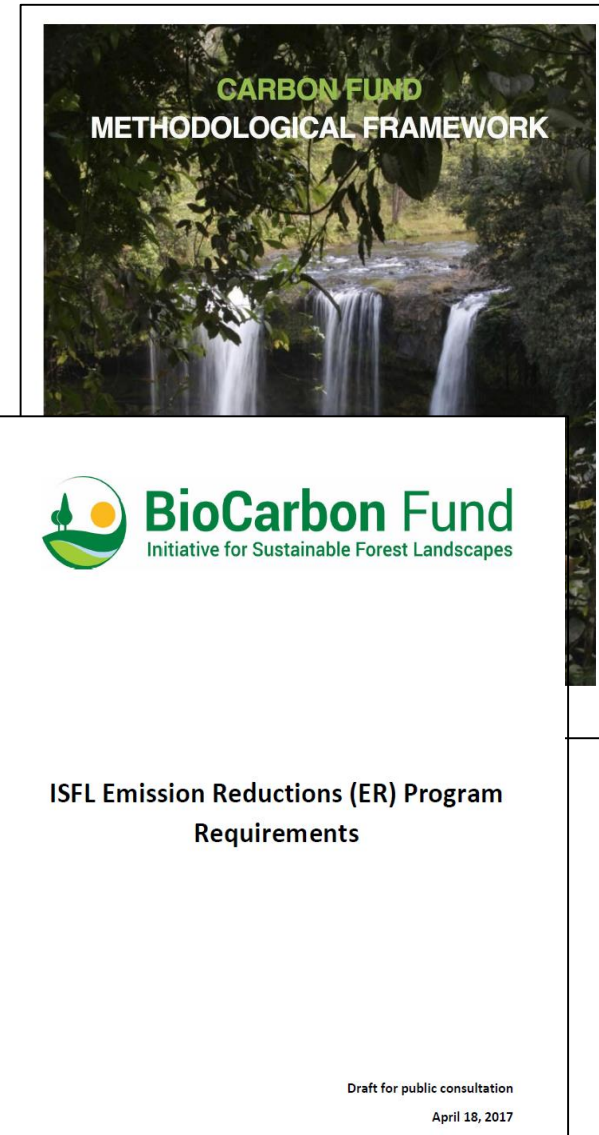


Landscape \approx AFOLU sector



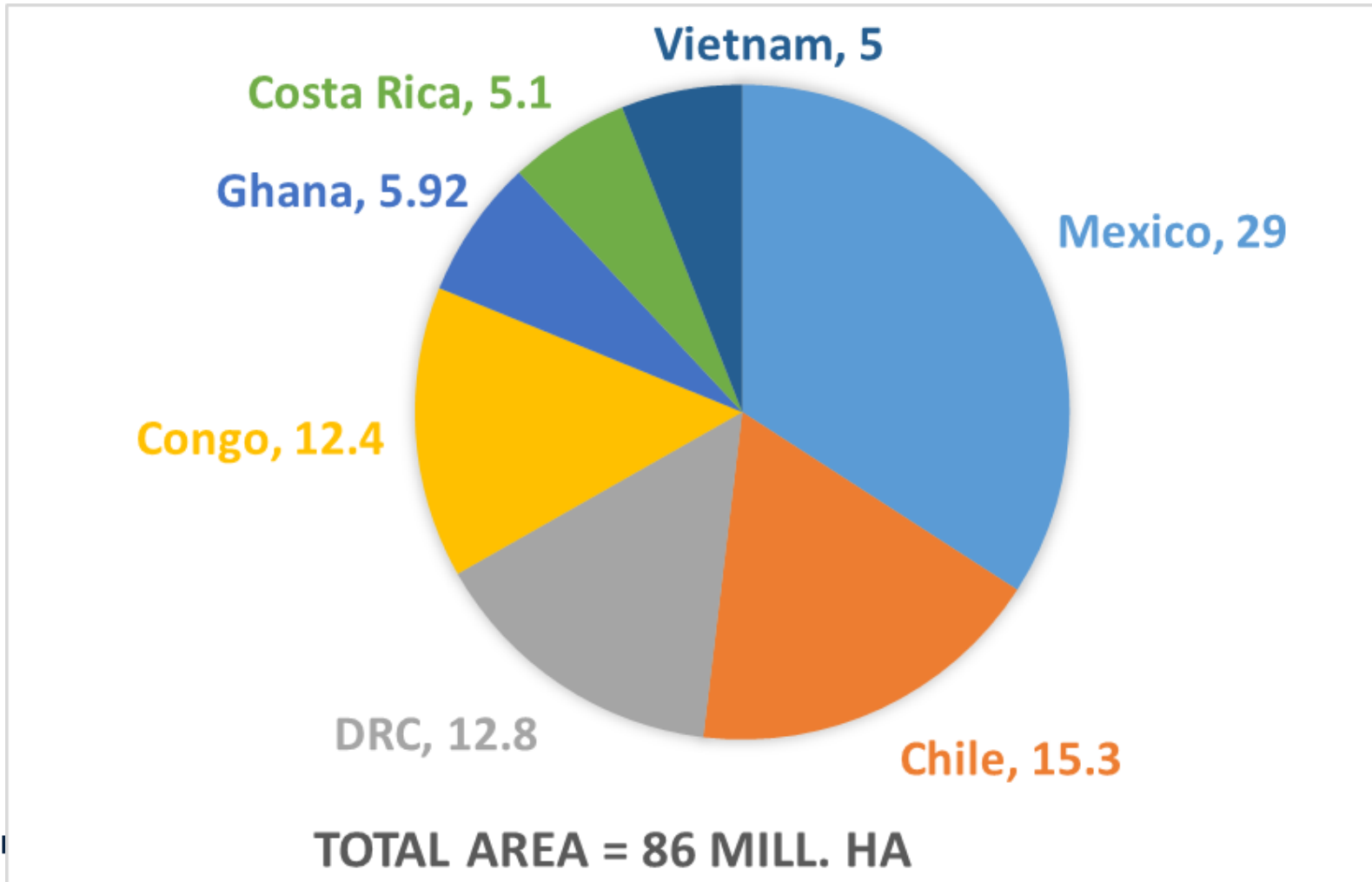
GHG ACCOUNTING REQUIREMENTS

- ER programs have to present RL and MRV system designs compliant with methodological requirements
- Some highlights:
 - ✓ GHG emissions from forest degradation or FL-FL must be accounted for
 - ✓ GHG emissions and removals have to be estimated with IPCC Tier 2, Tier 1 may be used exceptionally
 - ✓ Uncertainties estimated via Monte Carlo methods
 - ✓ Discounts are applied to ERs if HWCI >15% at 90% of confidence



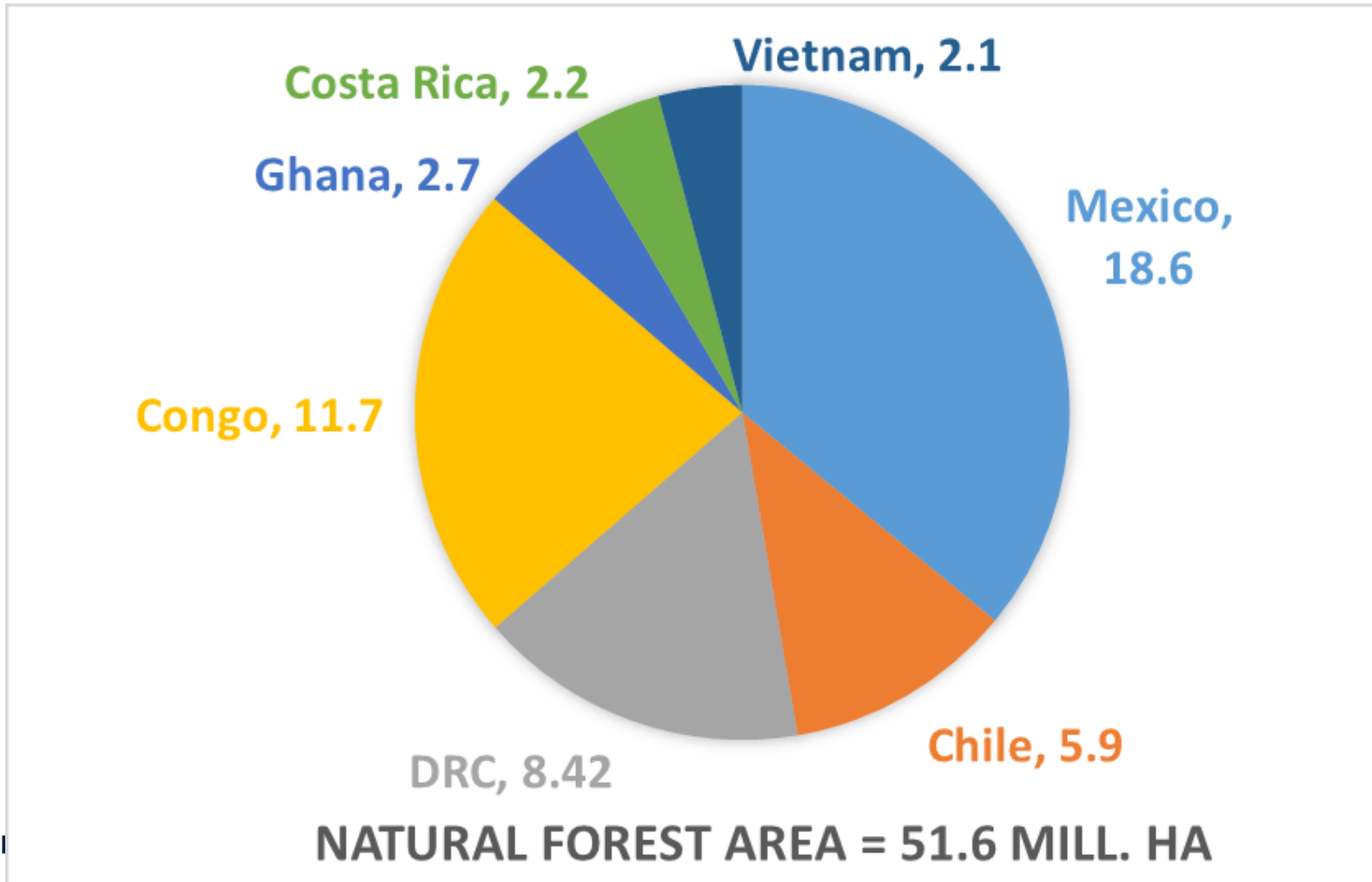
REFERENCE LEVELS OF ER PROGRAMS

- 7 ER programs have presented Reference Levels so far
- Some figures...



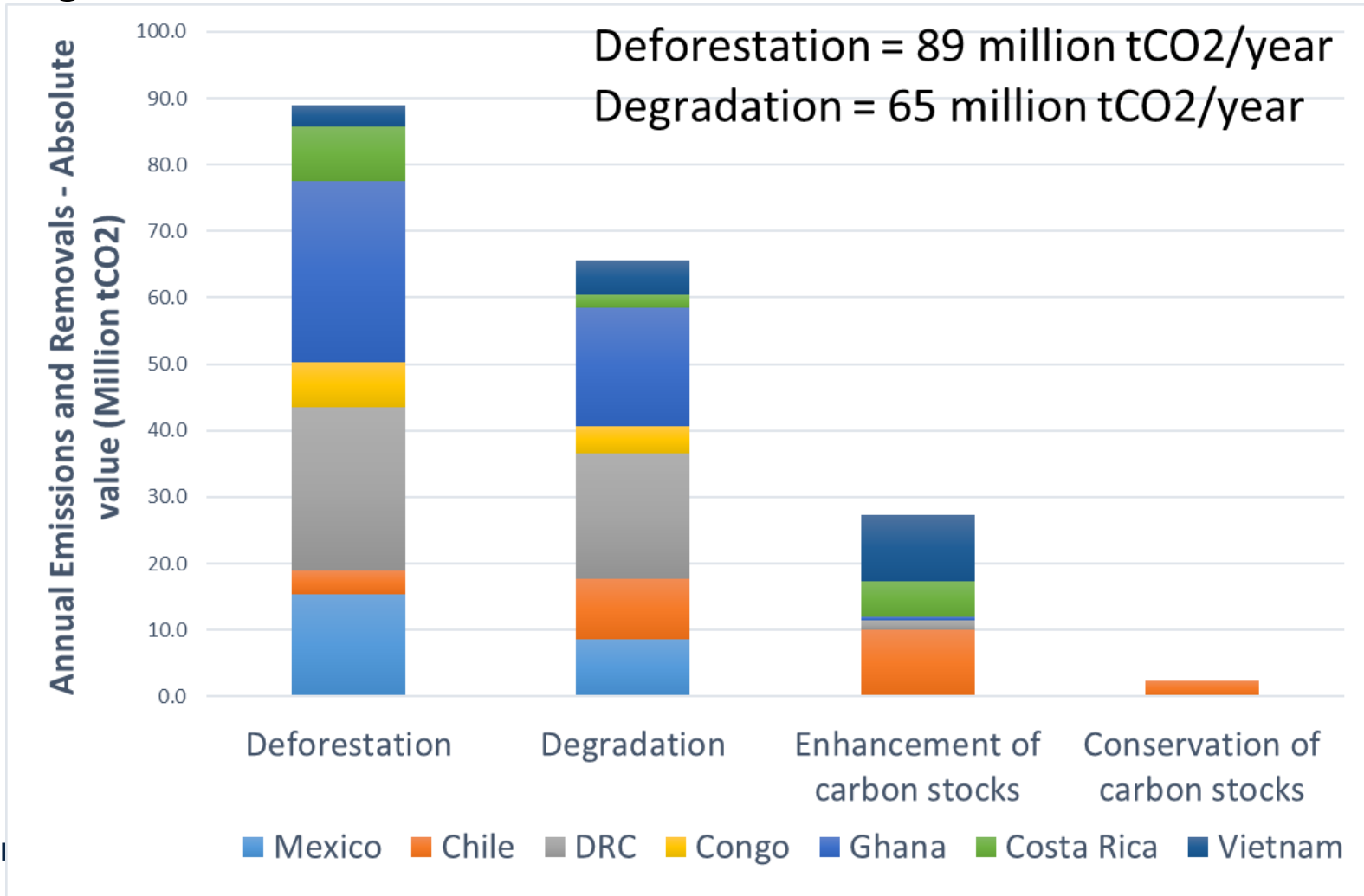
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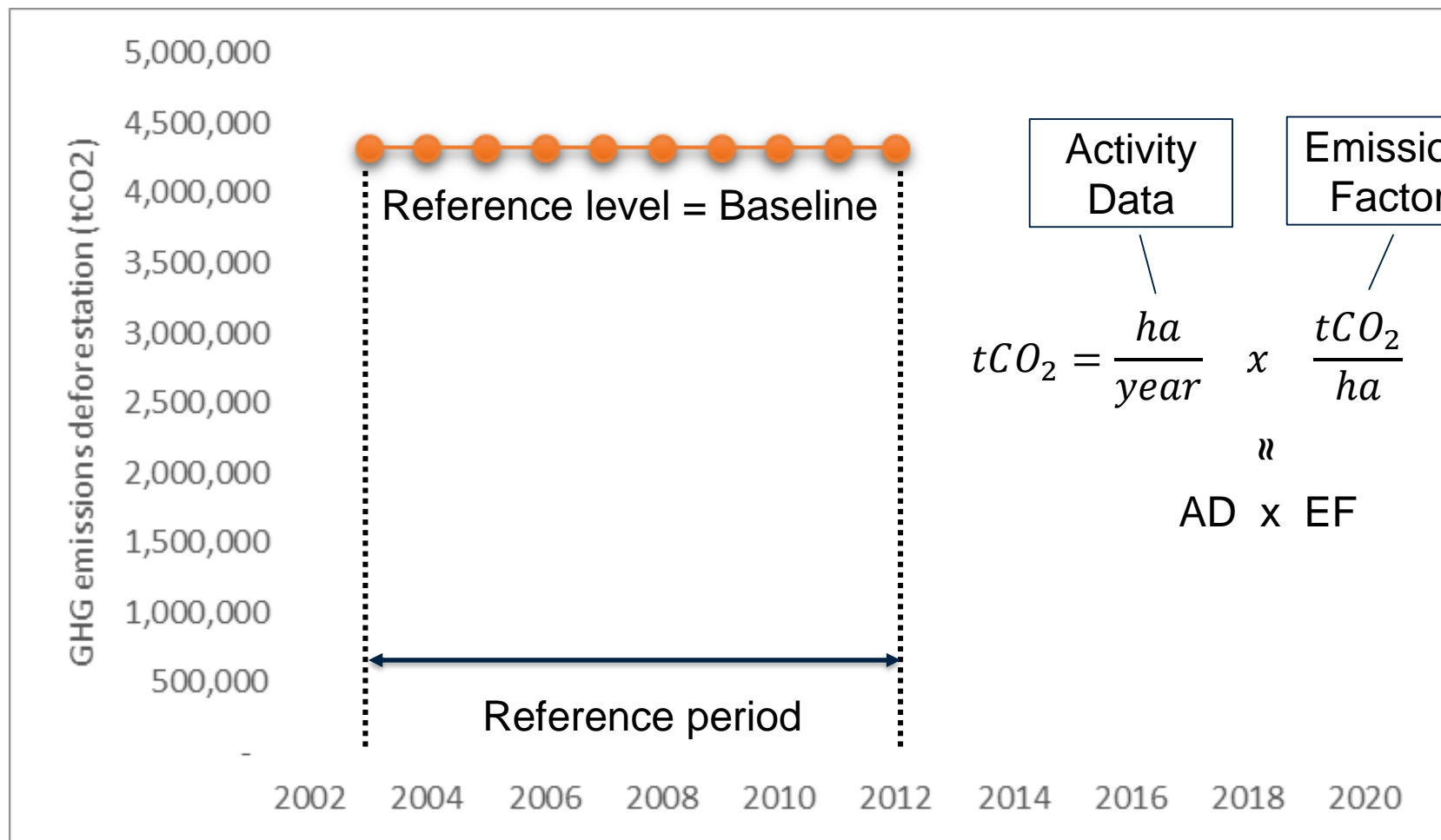
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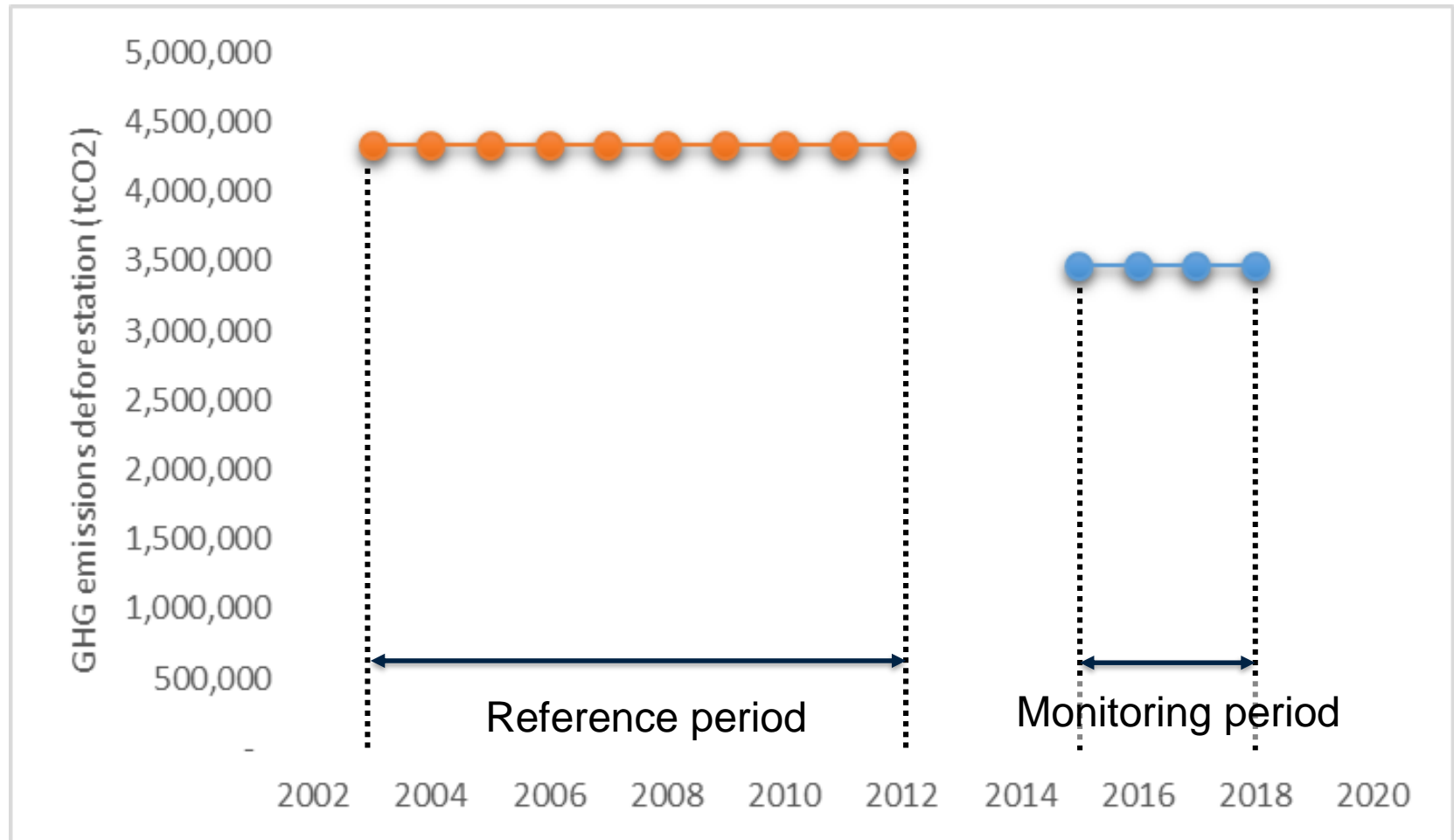
HOW TO ESTIMATE EMISSION REDUCTIONS

- A simple case...



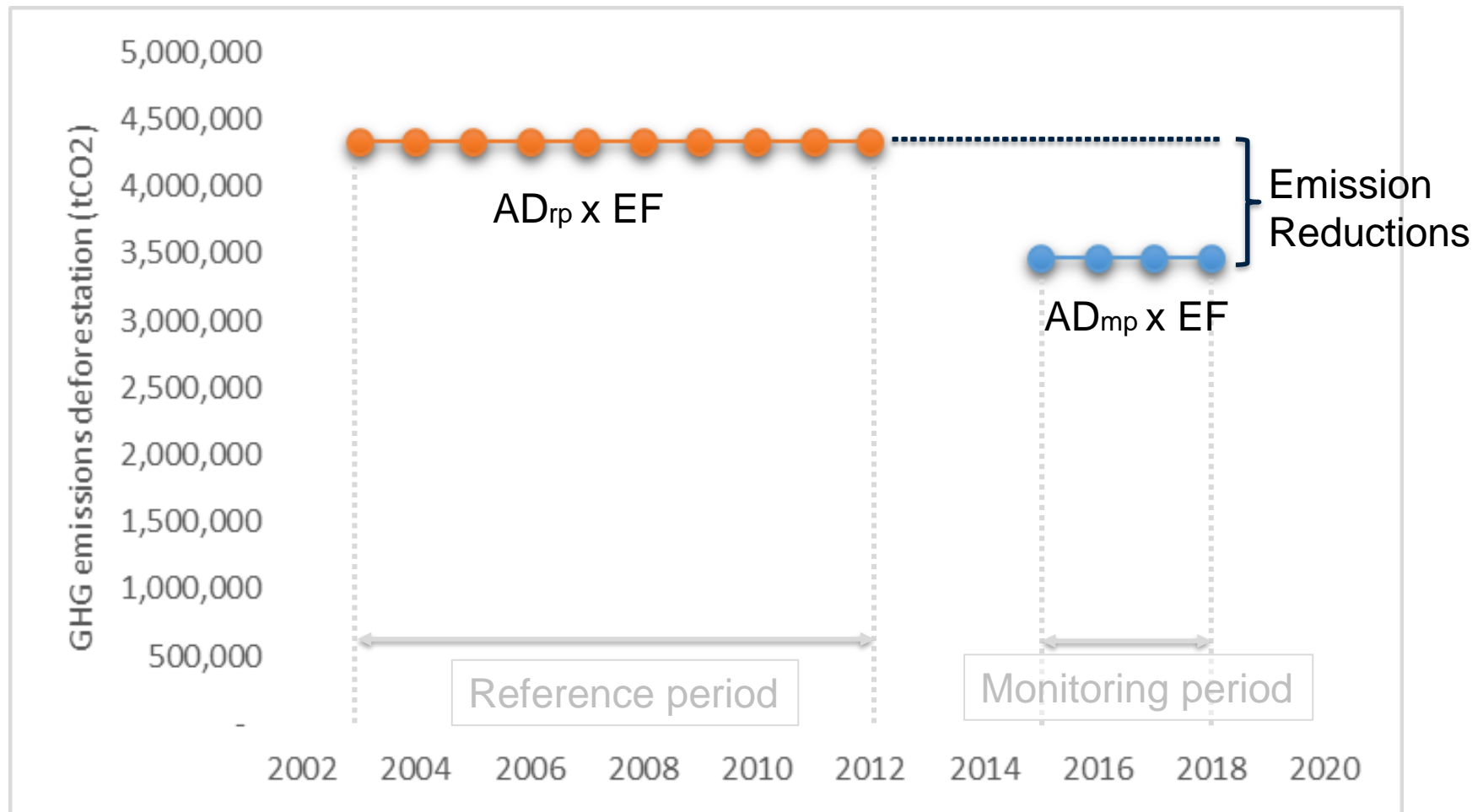
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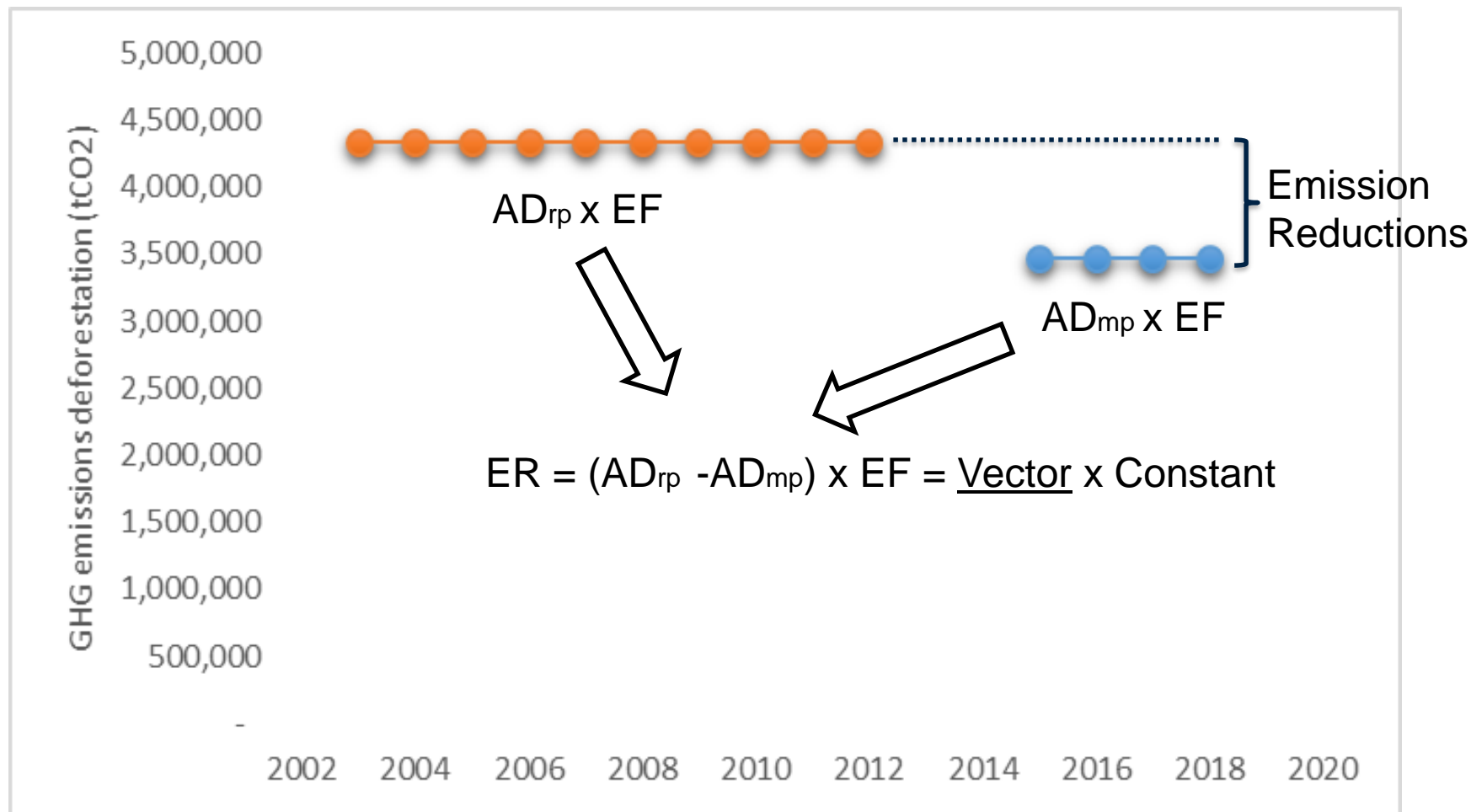
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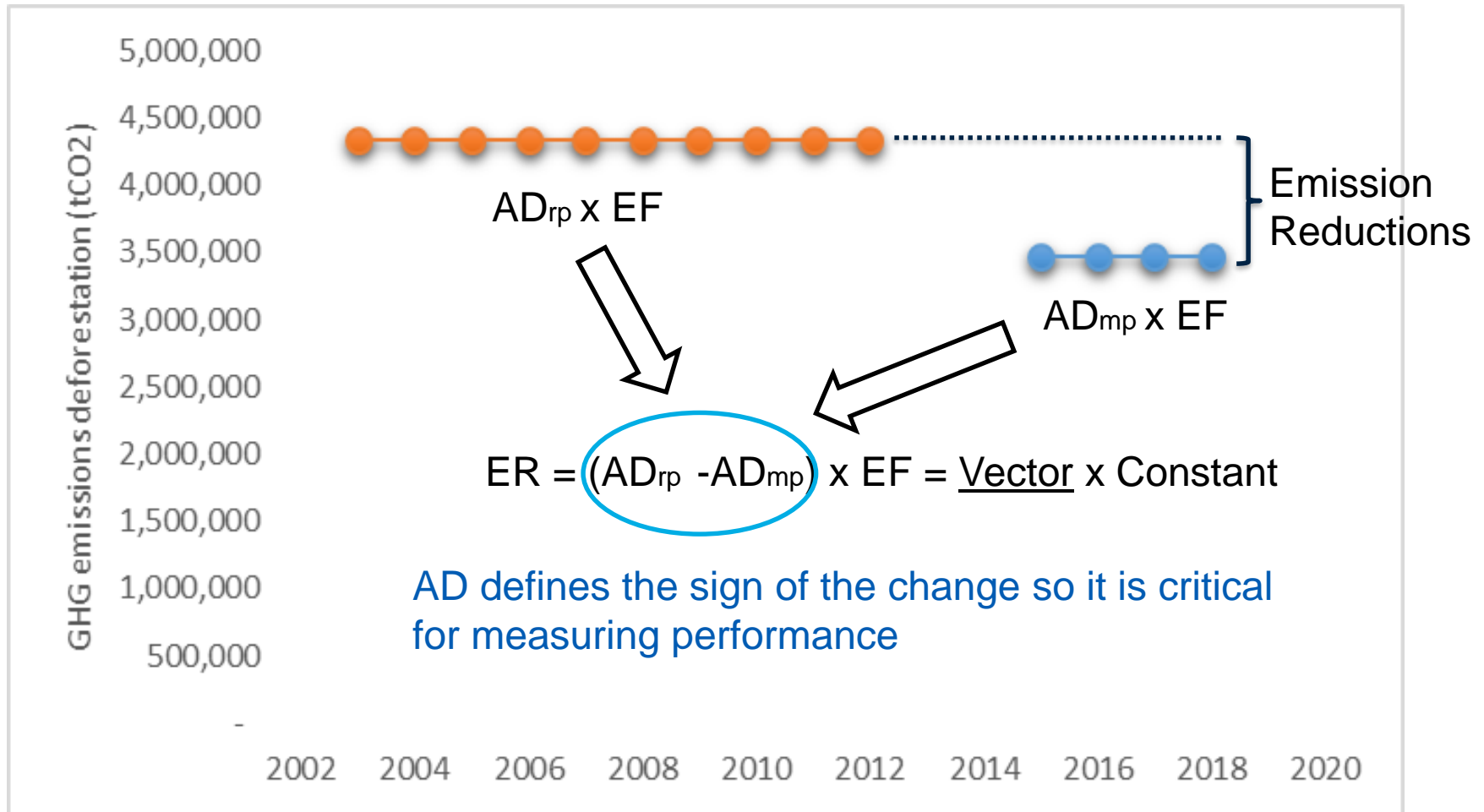
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HOW TO ESTIMATE EMISSION REDUCTIONS

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ACTIVITY DATA ESTIMATION – CHALLENGES

- Activity data has been usually estimated with EO data with two approaches:
 - Wall-to-wall approach, i.e. maps
 - Sampling approach
- However, the use of maps has some issues...

ACTIVITY DATA ESTIMATION – CHALLENGES

- Example of probabilities matrix of change map

| | | Reference | | | | Total | User's accuracy |
|---------------------|-------|-----------|------|------|------|-------|-----------------|
| | | DF | AF | F | NF | | |
| Map | Class | | | | | | |
| | DF | 0.03 | 0.00 | 0.01 | 0.01 | 0.04 | 0.62 ← |
| | AF | 0.00 | 0.03 | 0.00 | 0.01 | 0.04 | 0.75 |
| | F | 0.01 | 0.02 | 0.50 | 0.04 | 0.57 | 0.88 |
| | NF | 0.02 | 0.01 | 0.02 | 0.29 | 0.34 | 0.85 |
| | Total | 0.06 | 0.06 | 0.53 | 0.35 | 1.00 | |
| Producer's accuracy | | 0.49 | 0.50 | 0.94 | 0.84 | | |
| Overall accuracy | | | | | | 0.85 | |

62% of pixels classified as deforested are actually deforested

49% of area actually deforested has been mapped

*DF = Deforestation
AF = Afforestation

ACTIVITY DATA ESTIMATION – CHALLENGES

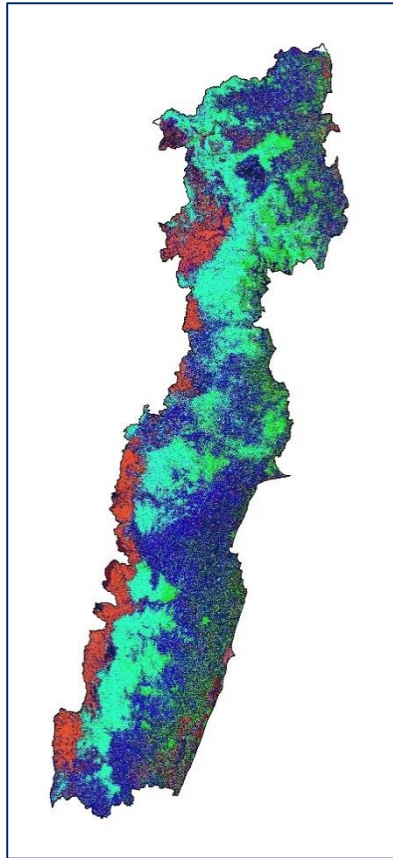
Good practices for estimating area and assessing accuracy of land change



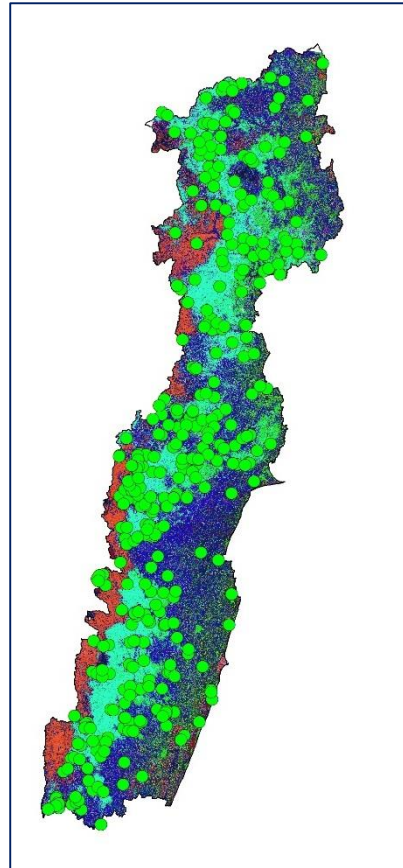
Pontus Olofsson ^{a,*}, Giles M. Foody ^b, Martin Herold ^c, Stephen V. Stehman ^d,
Curtis E. Woodcock ^a, Michael A. Wulder ^e

- Olofsson et al. (2014) is the first attempt to provide guidance in order to address the challenges of using maps to estimate Activity Data
- The approach is to use sample reference data and change maps for stratification, in order to obtain a stratified estimate (design-based inference)

ACTIVITY DATA ESTIMATION – CHALLENGES



Map = Strata



Sampling of
reference data

$$\hat{\mu}_{STR} = \sum_{h=1}^H w_h \hat{\mu}_h$$



| Change class | | StRS estimate | U% 90% |
|--------------|--------------------|---------------|--------|
| SF | Stable Forest | 0.407 | 3% |
| SNF | Stable Non-Forest | 0.439 | 3% |
| AF | Afforestation | 0.083 | 7% |
| DF | Deforestation | 0.025 | 19% |
| FE | Forest Enhancement | 0.018 | 7% |
| FD | Forest Degradation | 0.028 | 19% |

Inference

ACTIVITY DATA ESTIMATION – CHALLENGES

Good practices for estimating area and assessing accuracy of land change

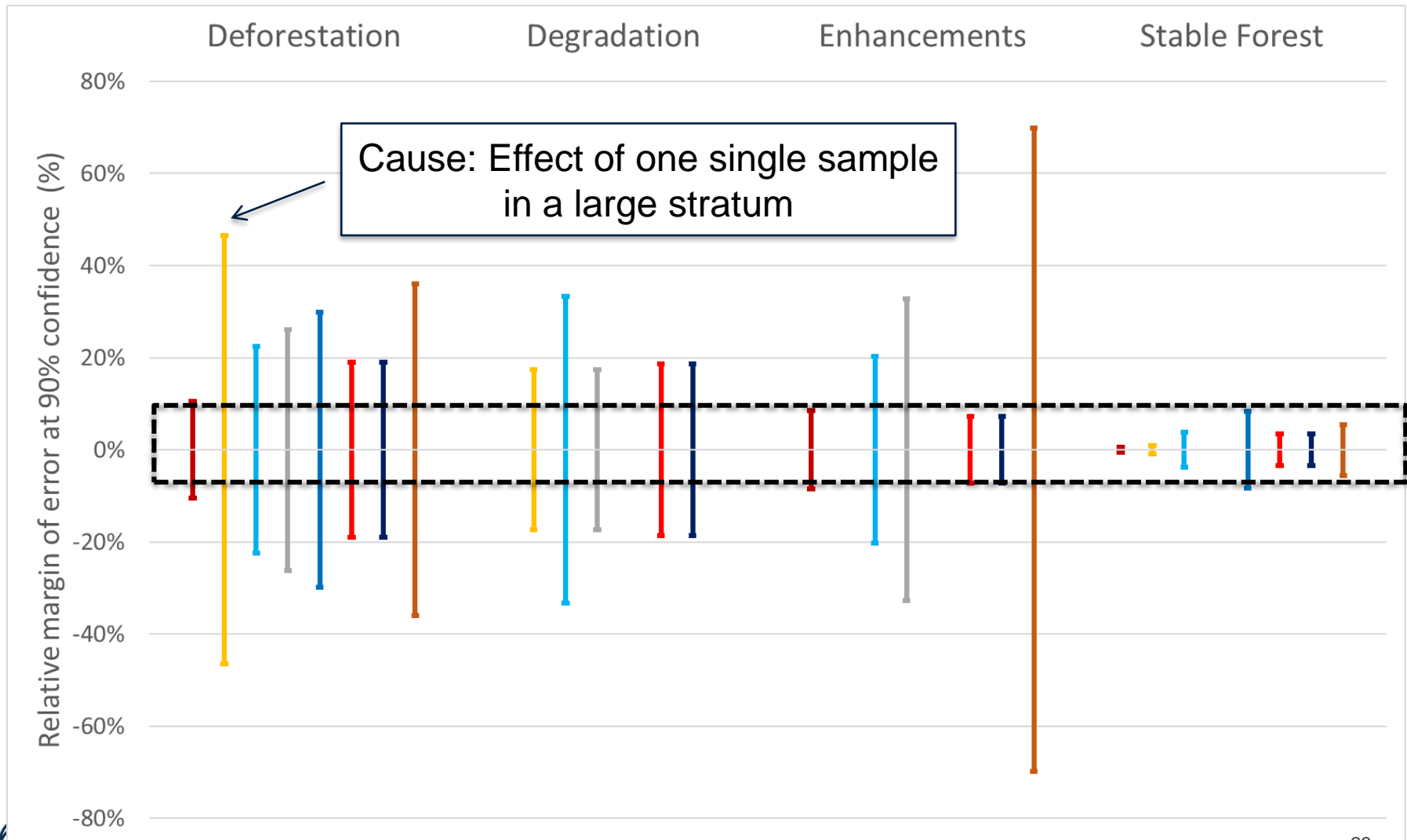


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- Five out of seven programs of the CF have applied this guidance
- Costa Rica has not applied it as it has applied complex Tier 3 integration frameworks for estimating the RL
- However, some challenges have been faced when establishing their Reference Levels...

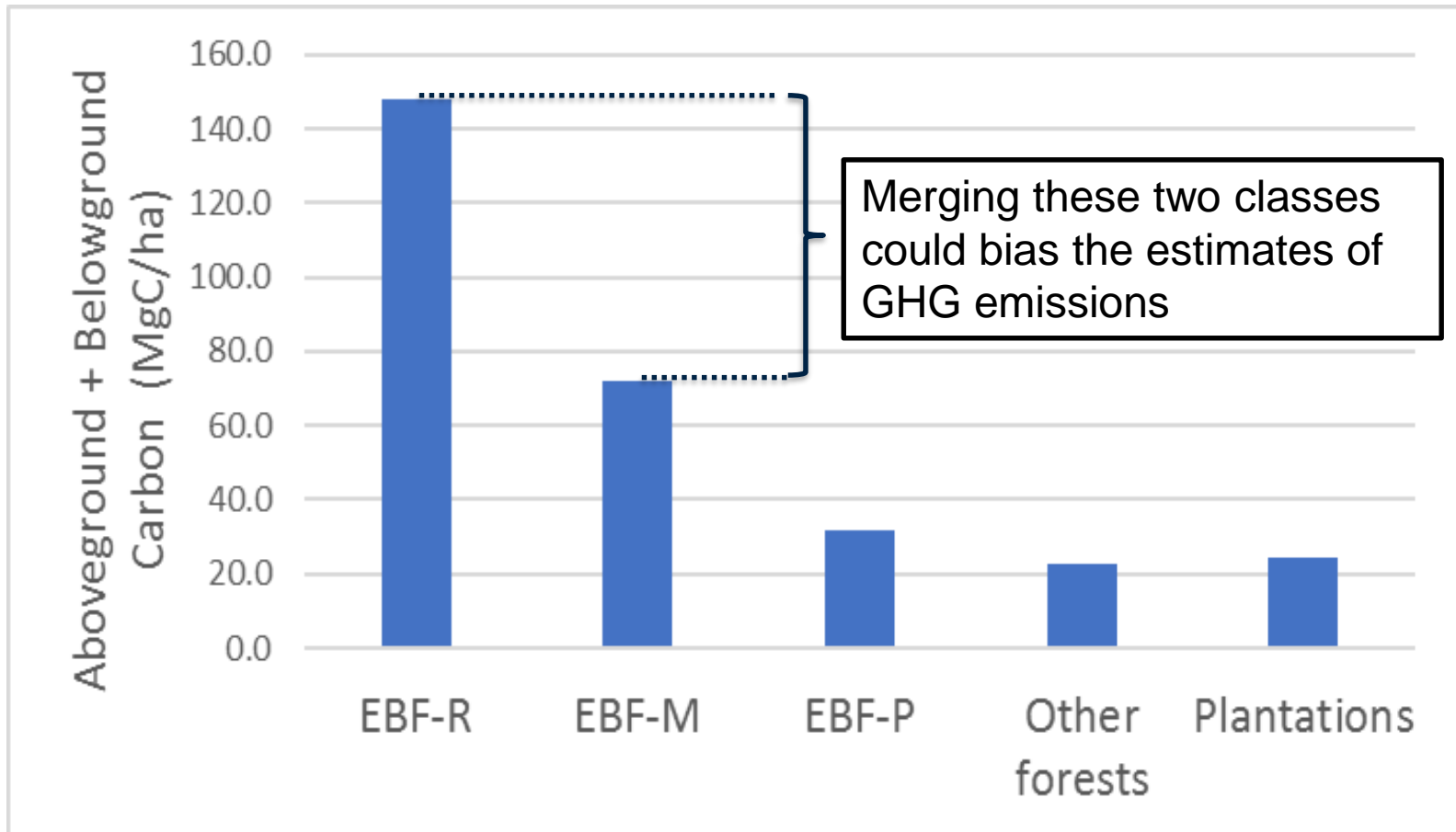
ACTIVITY DATA ESTIMATION – CHALLENGES

- Challenge 1: Too large statistical uncertainty



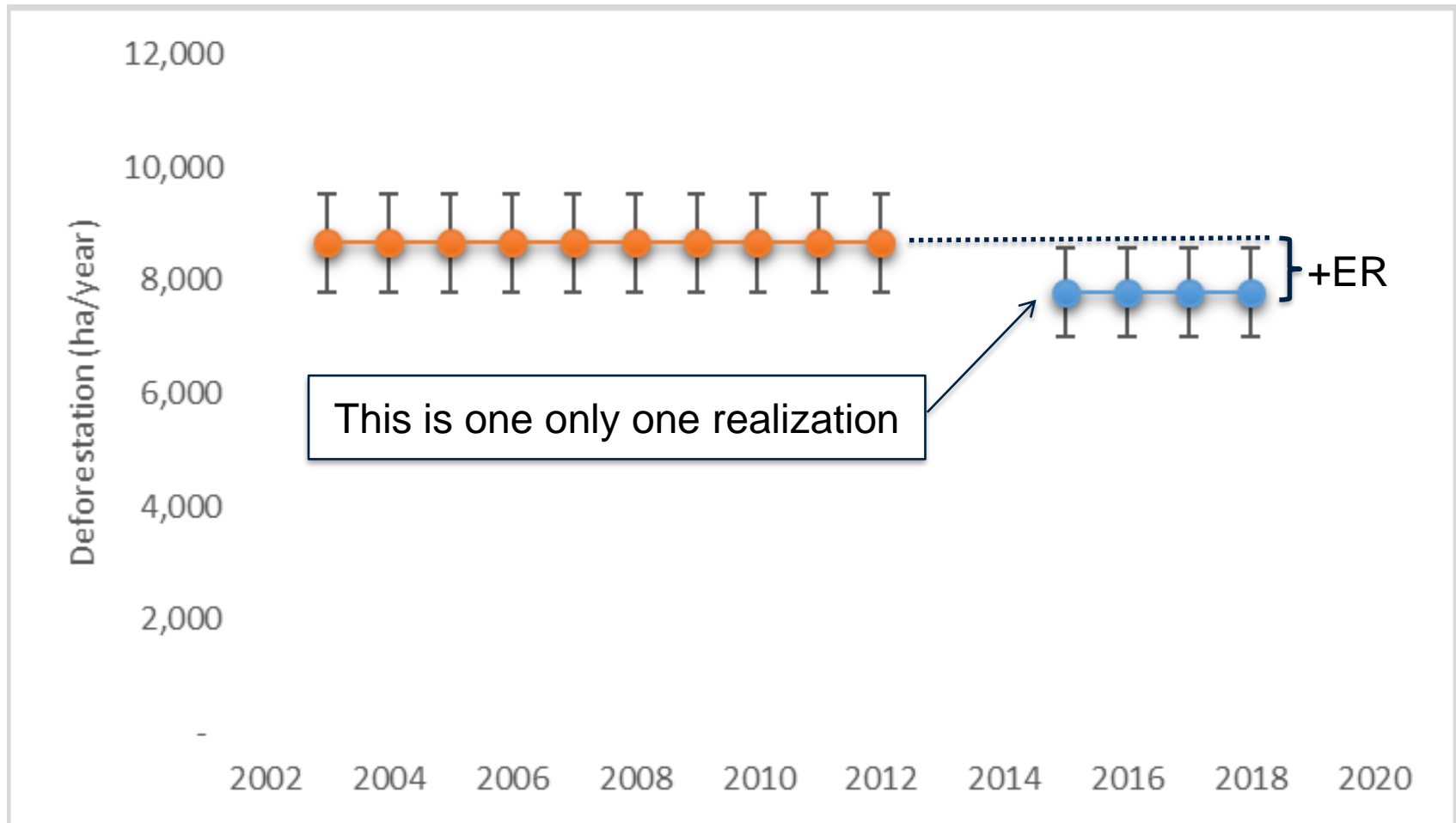
ACTIVITY DATA ESTIMATION – CHALLENGES

- Challenge 2: Difficulty in the application when large complex classes (e.g. complex integrated methods)



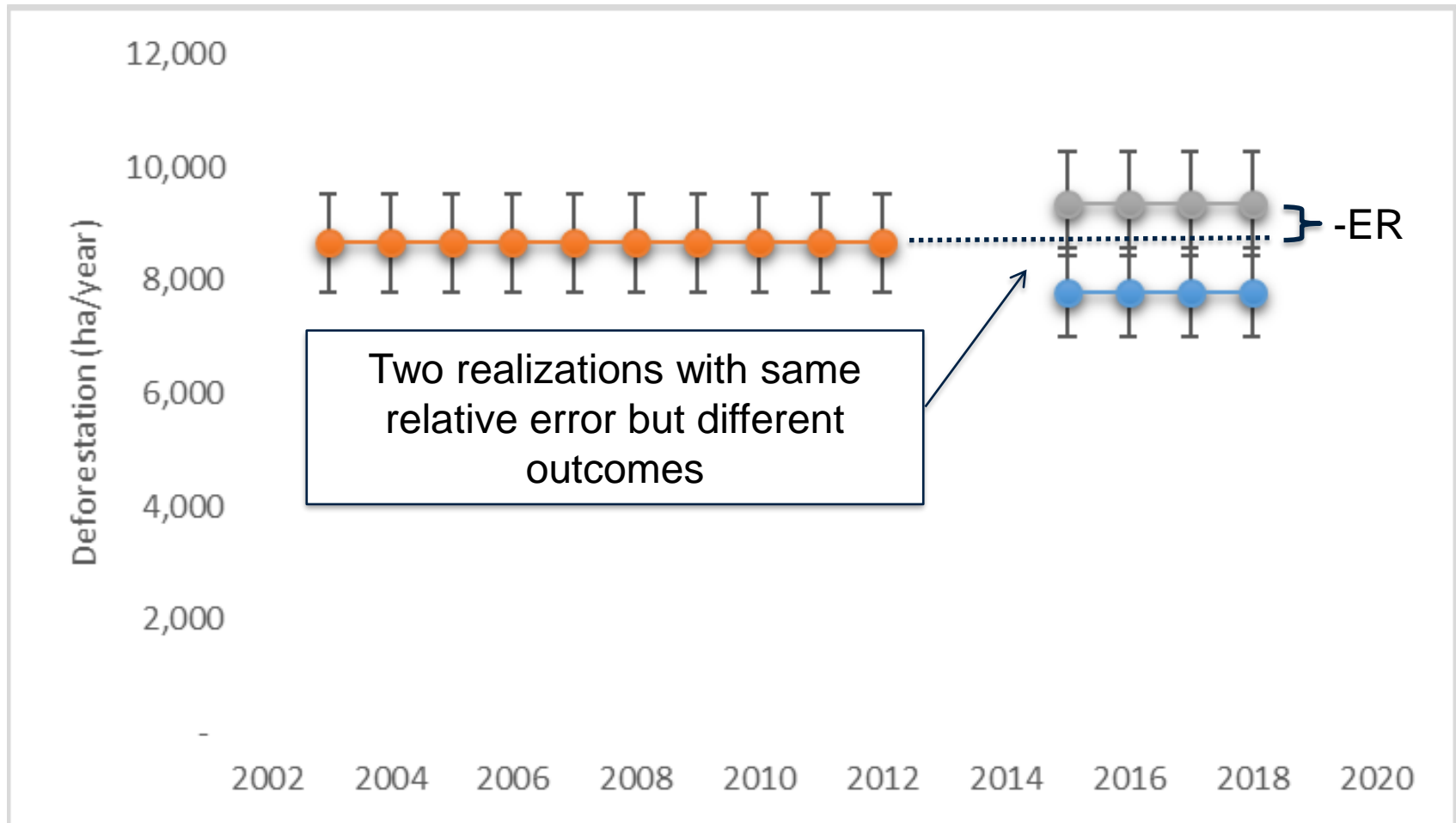
ACTIVITY DATA ESTIMATION – CHALLENGES

- Challenge 3: How to estimate ERs with precision?



ACTIVITY DATA ESTIMATION – CHALLENGES

- Challenge 3: How to estimate ERs with precision?



ACTIVITY DATA ESTIMATION – OPPORTUNITIES

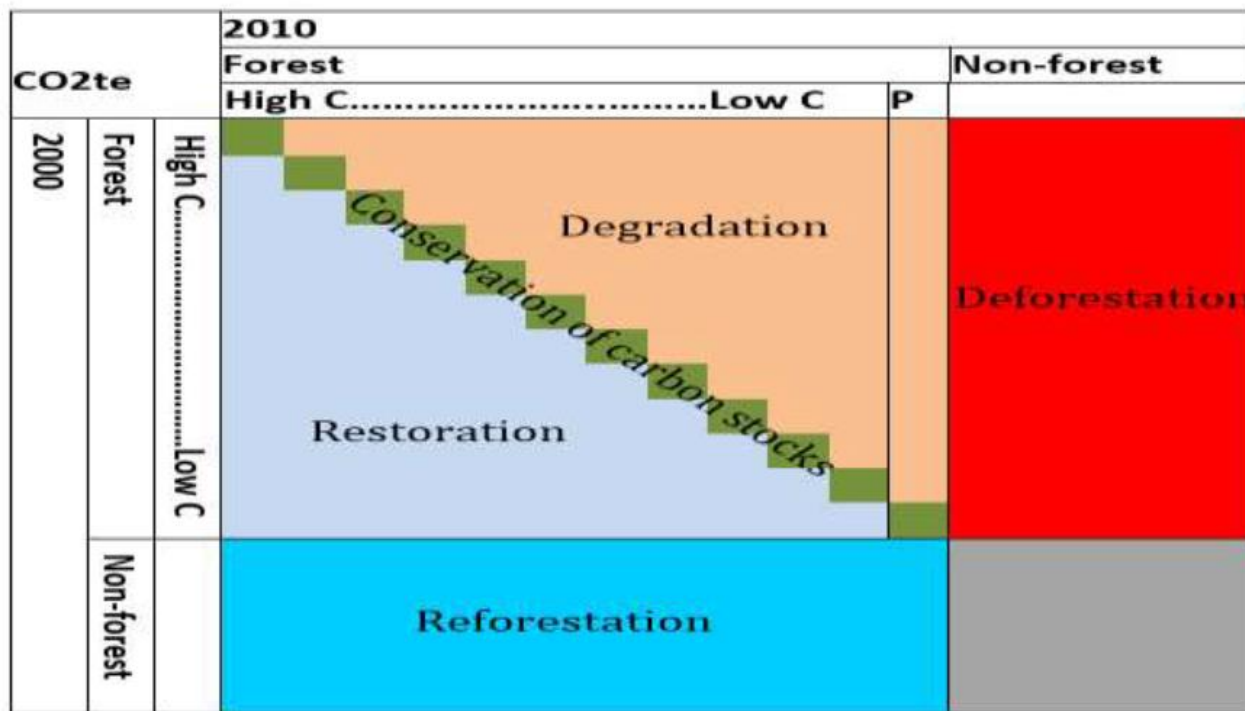
1. How to reduce uncertainty of AD in design-based inference?
2. How to estimate uncertainty in complex legends or high integration methods?
 1. Options in design-based inference
 2. Options in model-based inference
3. How to estimate the change of AD and its uncertainty?
 1. Options of sampling designs
 2. Montecarlo simulations

OTHER CHALLENGES AND OPPORTUNITIES - DEGRADATION

- The methodological requirements of the CF/ISFL require accounting of GHG emissions from degradation
- ER programs have piloted different methods to estimate degradation
- ER programs have successfully estimated GHG emissions from degradation...
- ...yet, still many uncertainties and limitations

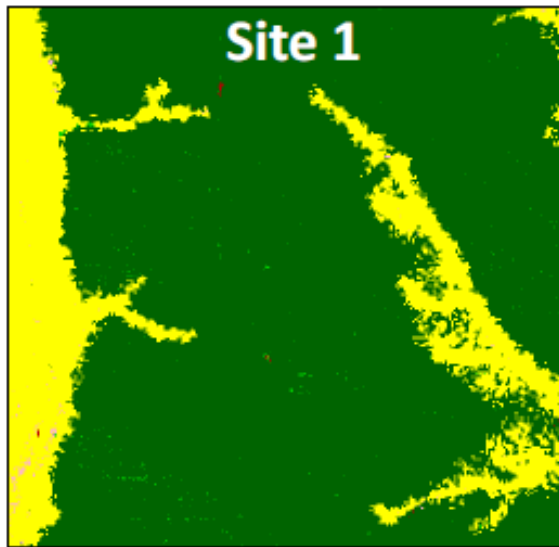
OTHER CHALLENGES AND OPPORTUNITIES - DEGRADATION

- **Mexico, Vietnam and DRC:** Degradation detected as transitions between forest types (e.g. primary to secondary forest)
- Some issues: only detects high disturbance degradation, high uncertainty in the classification

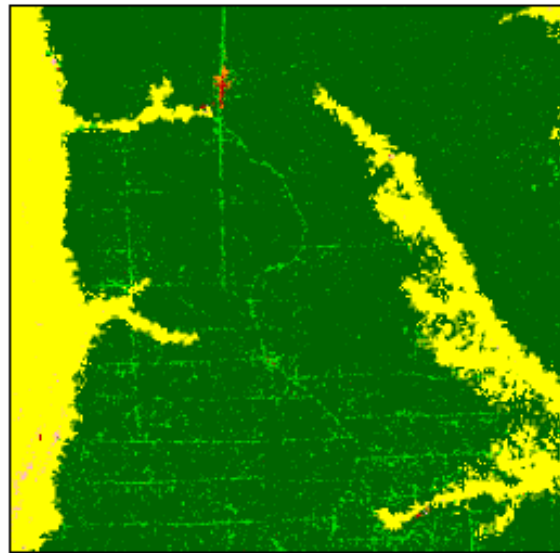


OTHER CHALLENGES AND OPPORTUNITIES - DEGRADATION

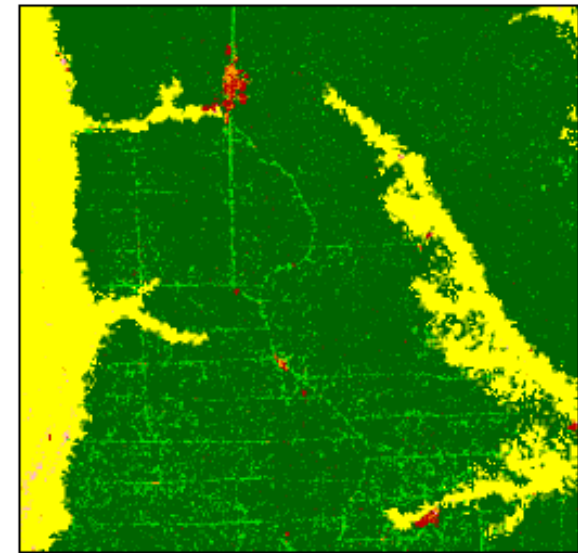
- **Congo, Madagascar:** Degradation is detected through changes in vegetation indices in a temporal series of medium resolution imagery
- Some issues: High commission errors, no VHR imagery available for validation,



2003



2012



2015

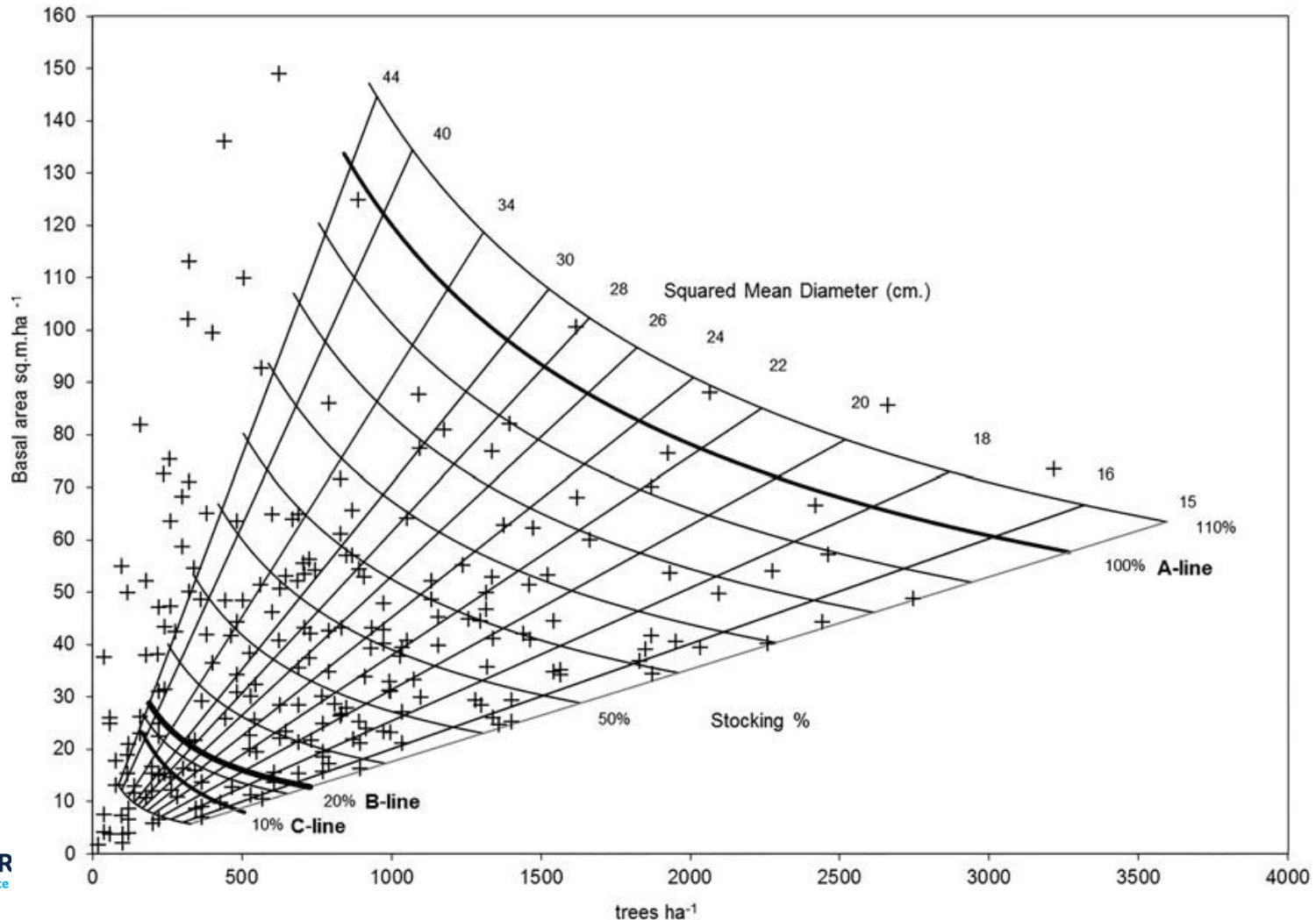
OTHER CHALLENGES AND OPPORTUNITIES - DEGRADATION

- **Costa Rica, Madagascar:** Degradation is detected through changes canopy cover observed in VHR imagery
- Some issues: coverage of VHR imagery, impossibility to detect small changes in canopy



OTHER CHALLENGES AND OPPORTUNITIES - DEGRADATION

- **Chile:** Using stocking tables built with NFI data, and they are applied to spatial explicit stocking models



OTHER CHALLENGES AND OPPORTUNITIES - DEGRADATION

- **Ghana, Congo:** Using extracted timber volumes as proxies of degradation by multiplying volumes to damage factors
- **Issues:** uncertain extracted volumes, volumes of illegal logging not available

| Factor | | Value (tCO ₂ /m ³) | Uncertainty |
|-------------------------------|-----|--|-------------|
| Emission from Extracted Log | ELE | 0.79 | 0.02 |
| Logging Damage Factor | LDF | 2.46 | 0.17 |
| Logging Infrastructure Factor | LIF | 0.50 | 0.13 |
| Total Emission Factor | TEF | 3.75 | 0.21 |

OTHER CHALLENGES AND OPPORTUNITIES - WETLANDS

Age, extent and carbon storage of the central Congo Basin peatland complex

Greta C. Dargie^{1,2*}, Simon L. Lewis^{1,2*}, Ian T. Lawson³, Edward T. A. Mitchard⁴, Susan E. Page⁵, Yannick E. Bocko⁶ & Suspense A. Ifo⁶

- Peatlands in the Congo basin store a quantity that is equivalent to 95% of the above-ground carbon stocks of the tropical forests of the entire Congo Basin
- These areas are not yet under threat
- However, research is needed in order to understand the carbon dynamics and estimate potential impact of future policies over these areas

OTHER CHALLENGES – NEW RESEARCH

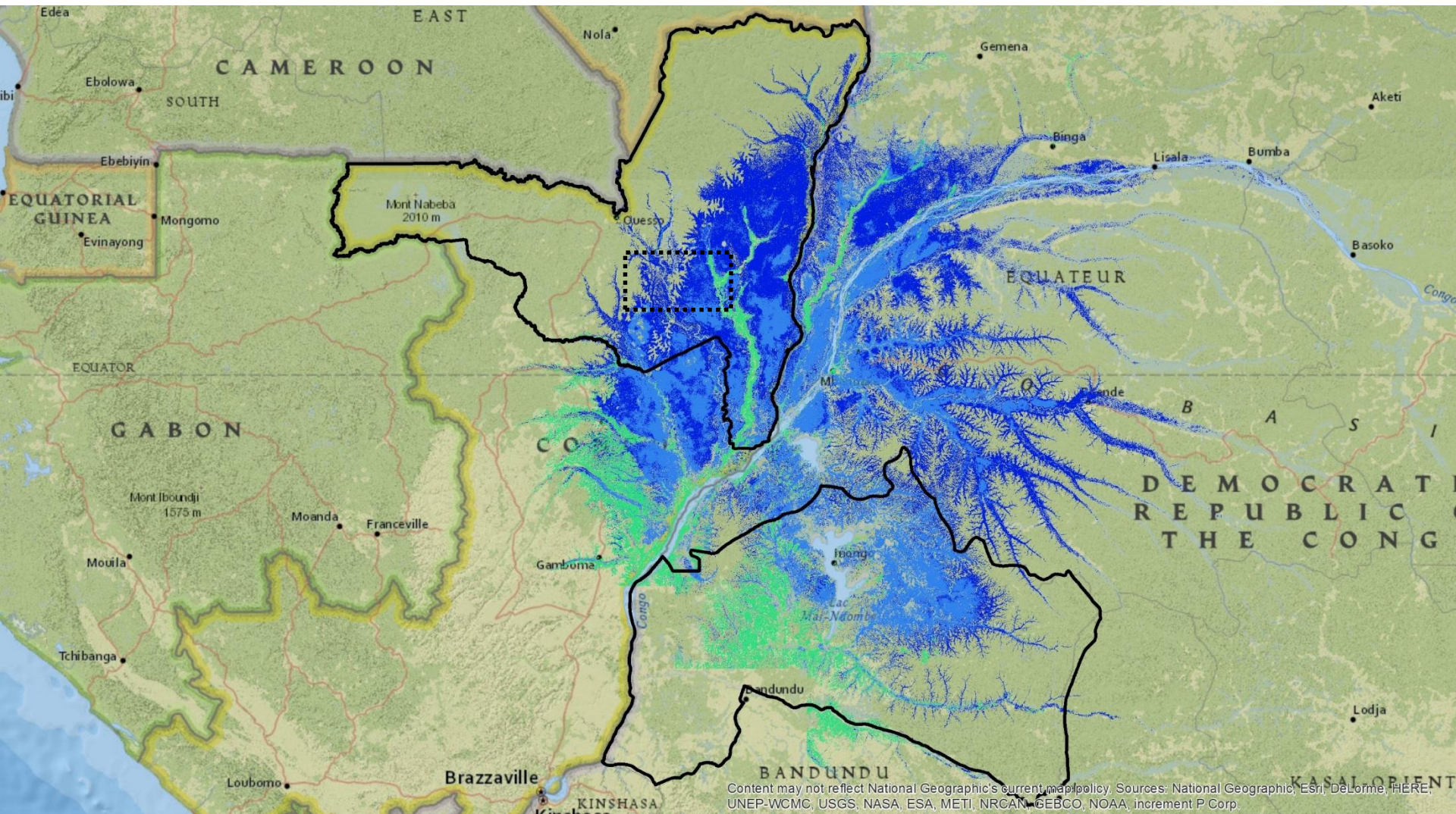


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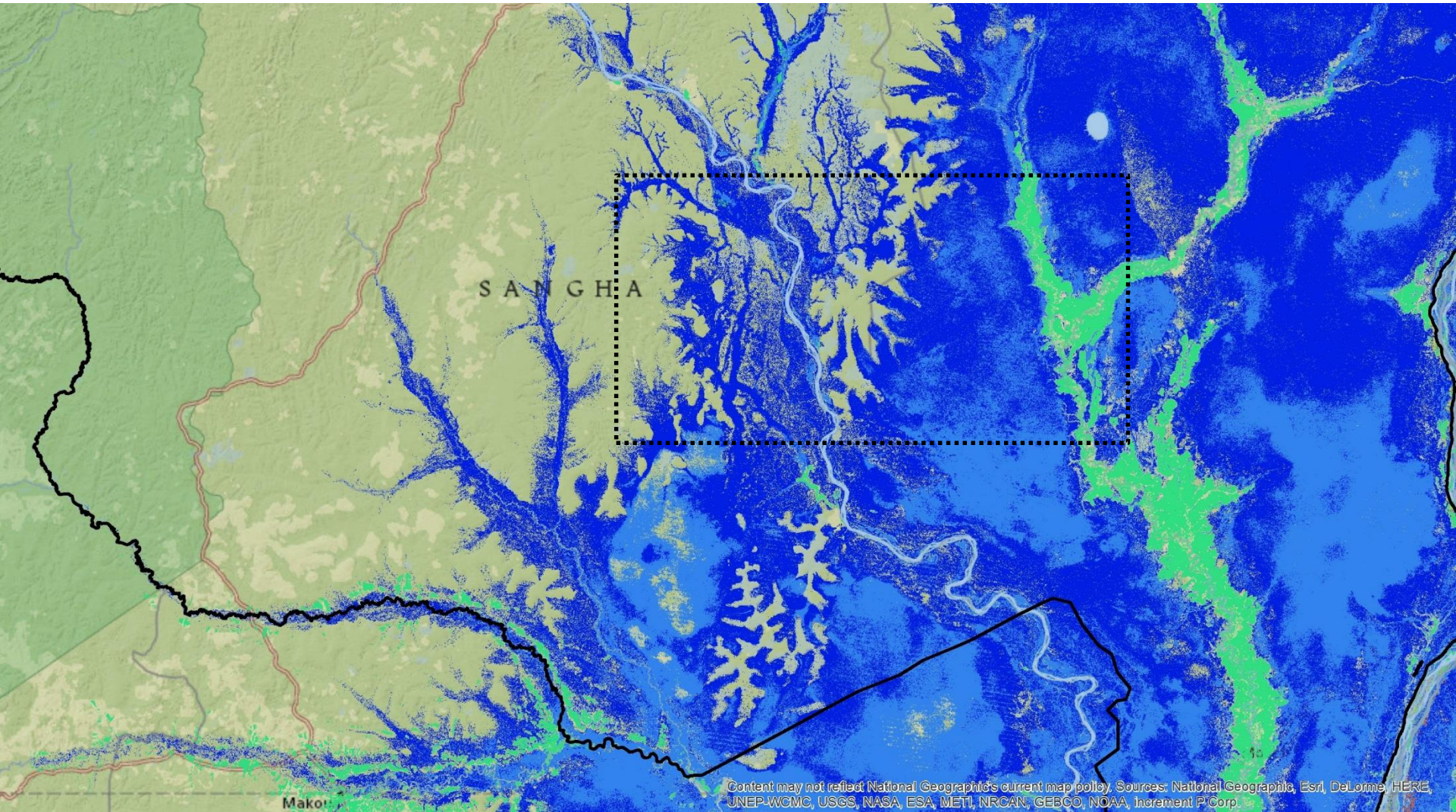


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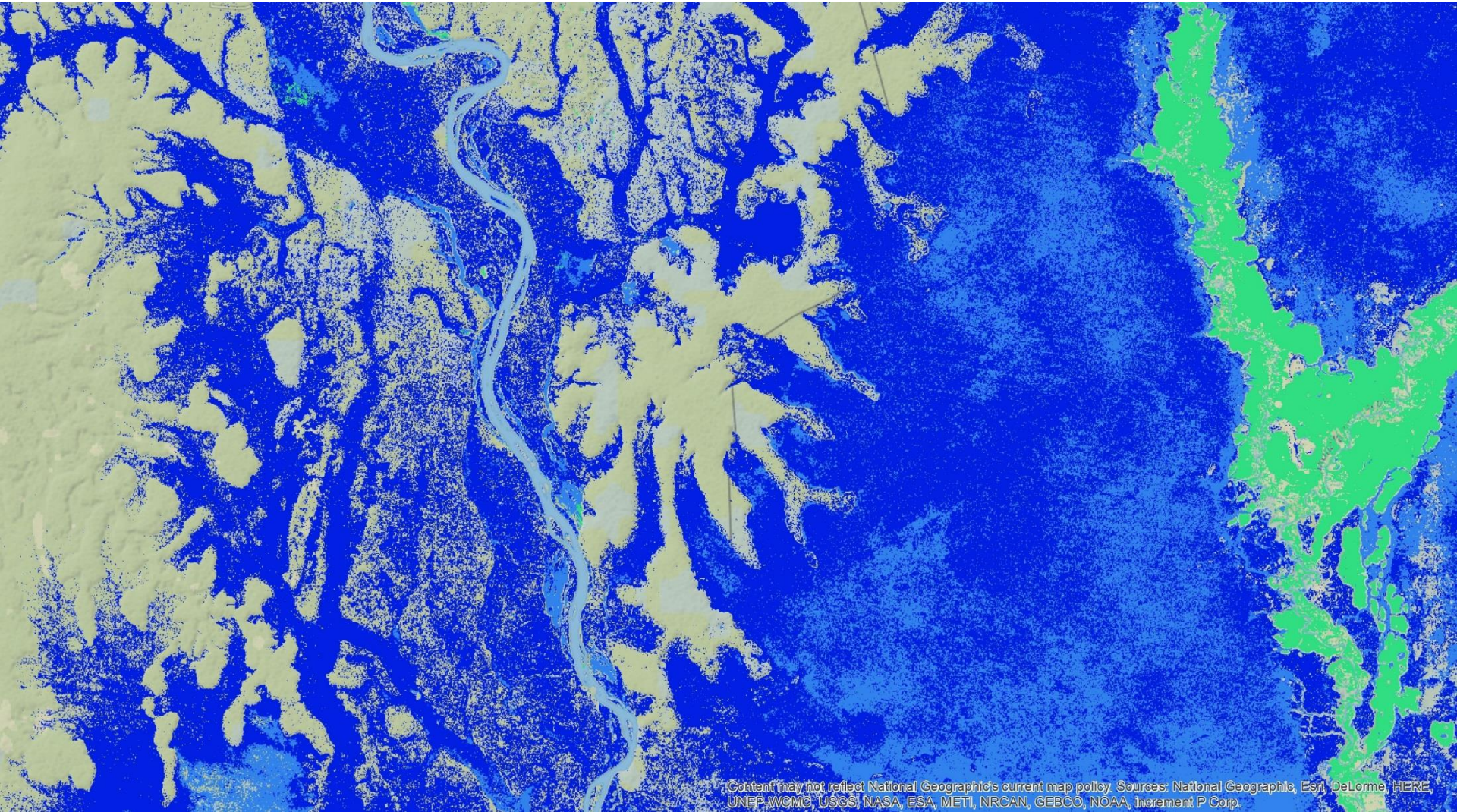
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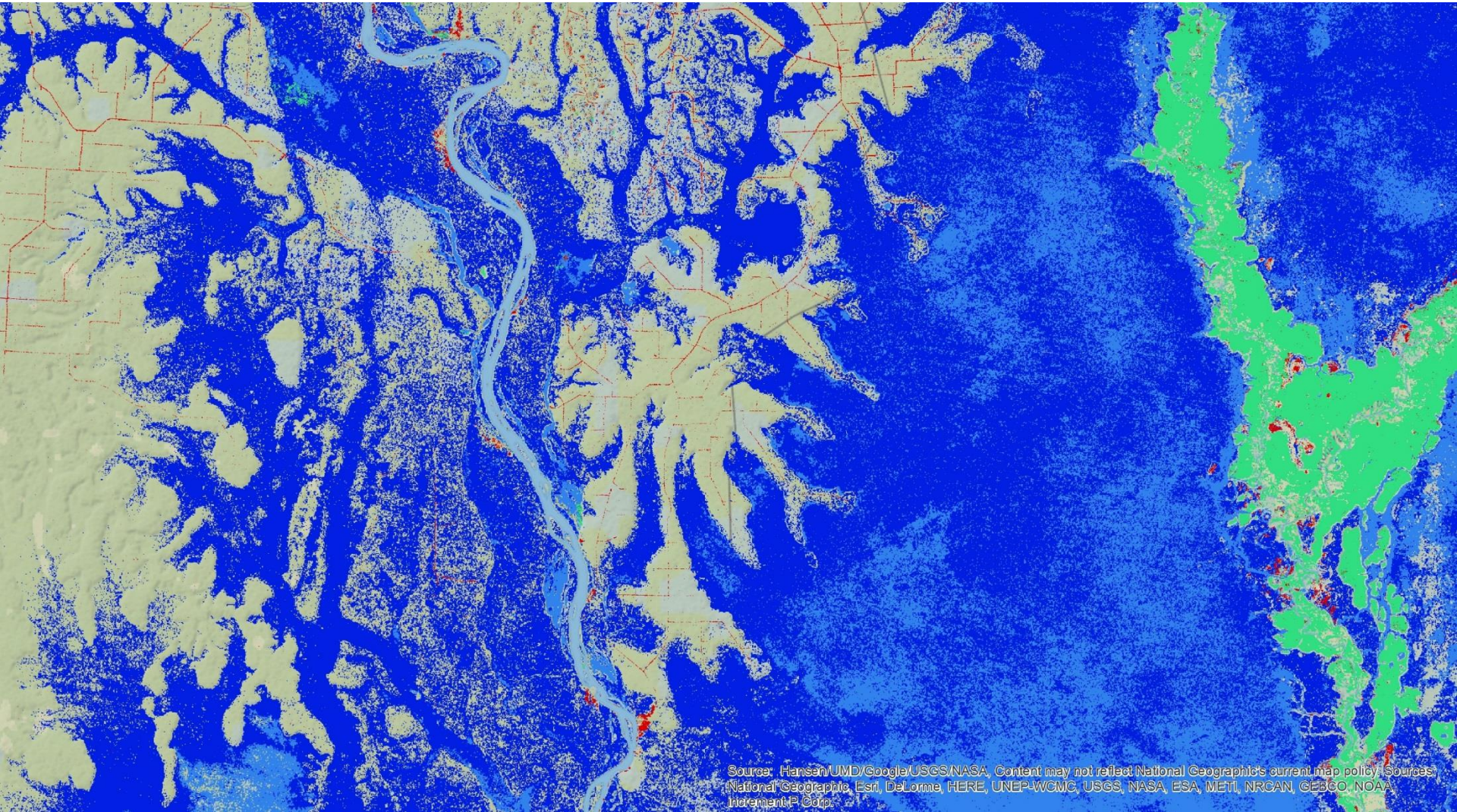


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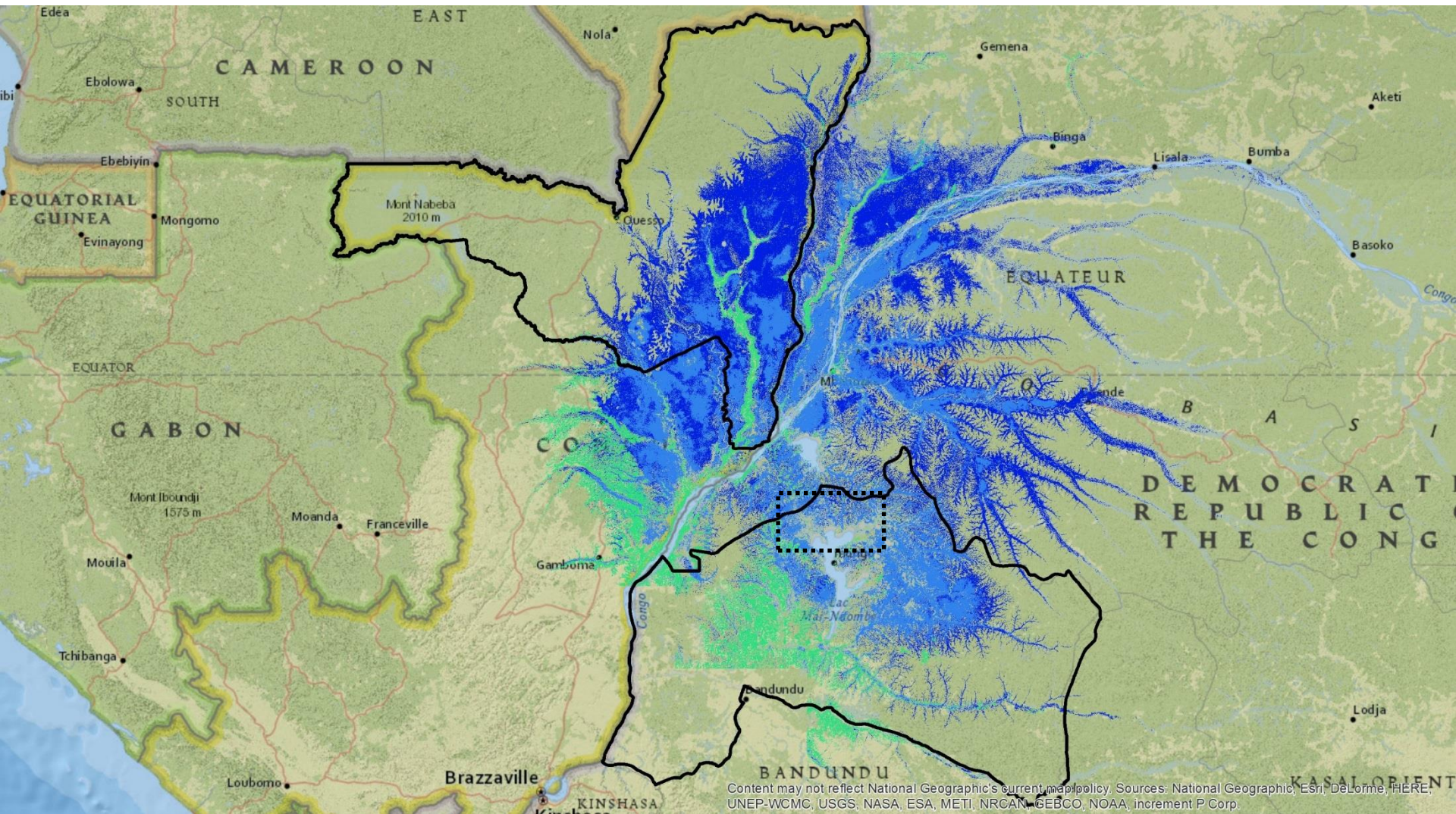


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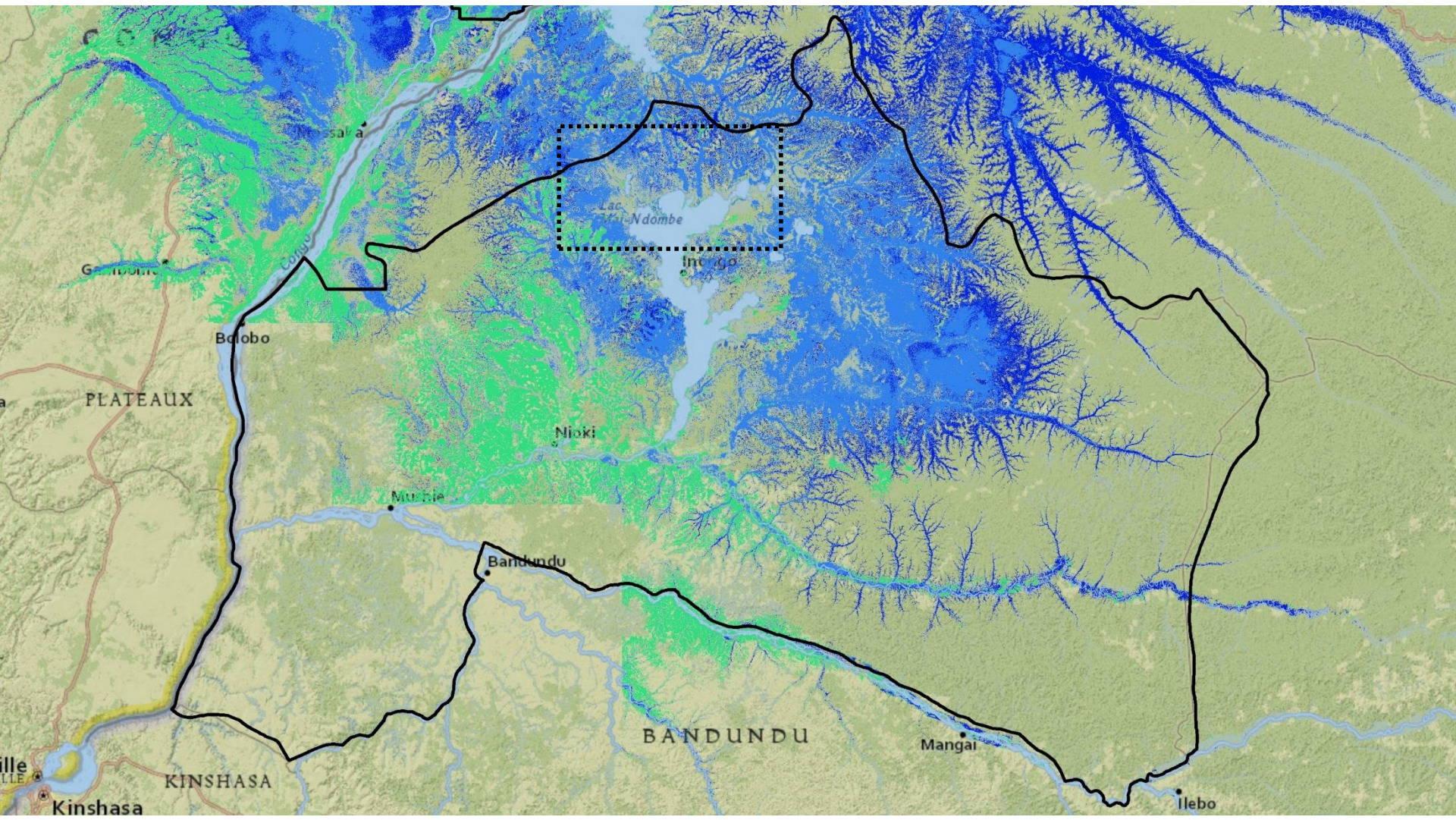
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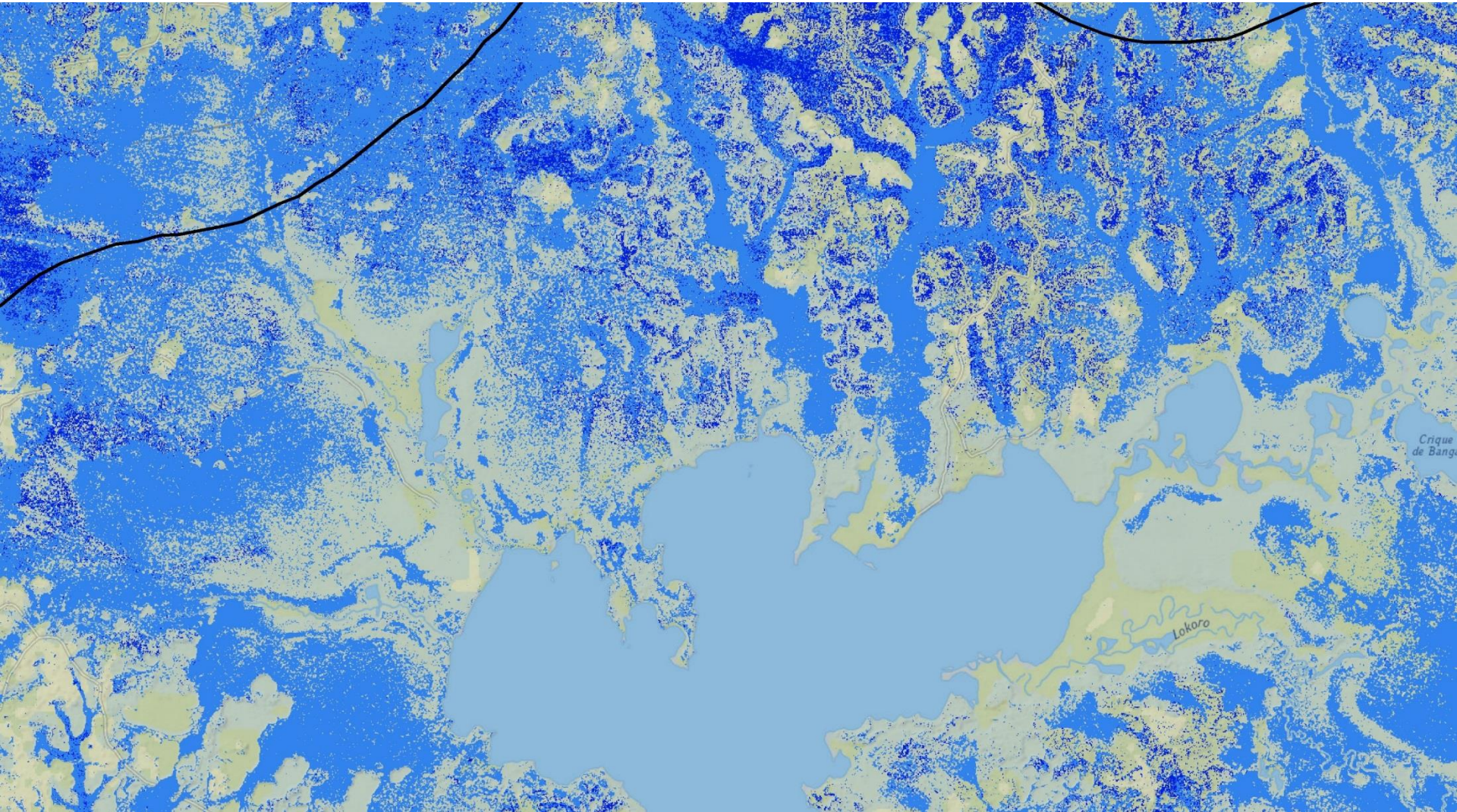
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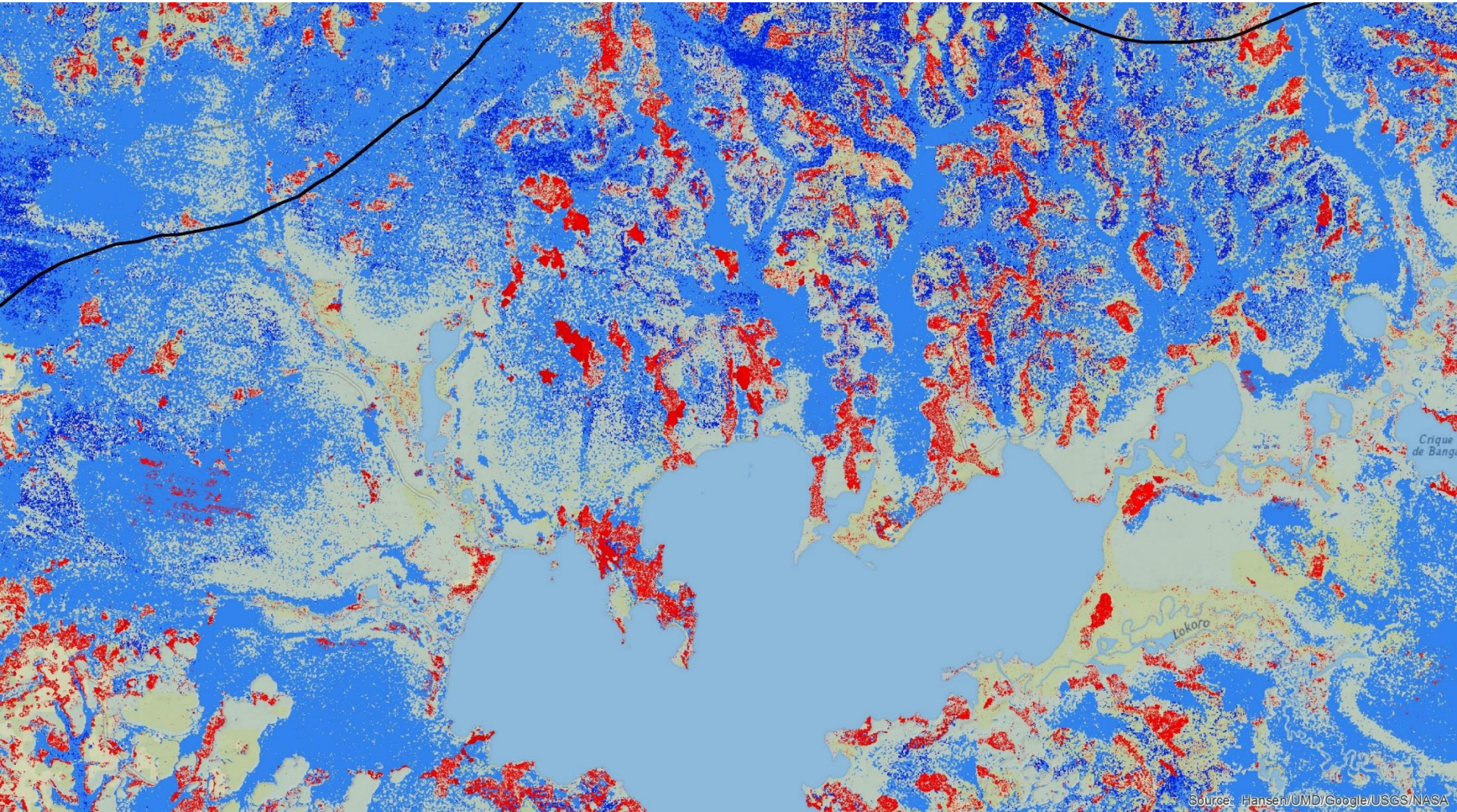
OTHER CHALLENGES – NEW RESEARCH



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Source: Hansen/UMD/Google/USGS/NASA



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THANK YOU

