# <sup>1</sup> National CO<sub>2</sub> budgets (2015-2020) inferred from atmospheric CO<sub>2</sub> observations in support of the Global Stocktake

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## Goals

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### • Contribute to Global Stocktake (GST) Activities of the Paris Agreement

- GST to monitor Paris agreement implementation (e.g., emissions and removals of CO<sub>2</sub>)
- GST to evaluate the collective progress made in achieving goals.
- Goal of the pilot dataset: Start a conversation.
  - Provide a pilot product of emissions and removals of CO<sub>2</sub>
  - Illustrate the type of dataset we can provide.
  - Identify current limits of our approach and where research is needed.
  - Inform development of Monitoring and Verification System

### • Long term goal:

 Provide countries with precise and accurate carbon budgets to track AFOLU (Agriculture, Forestry and Other Land Use) and unmanaged lands. Complement bottom-up datasets.

### What is in our dataset?

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- Quantities provided:
  - Net carbon exchange (net surface-atmosphere CO<sub>2</sub> flux)
  - Change in terrestrial carbon stocks ( $\Delta C_{loss}$ ).
  - Fossil fuel emissions and lateral C fluxes
  - And their uncertainties!
- Spatiotemporal scale:
  - Annual net fluxes over (2015-2020)
  - Country totals and as 1° x 1° degree.

# <sup>1</sup> Methods - CO<sub>2</sub> flux inversions

### v10 OCO-2 Model Intercomparison Project (MIP)

12 flux inversion models follow protocol with common data assimilated and fossil fuel emission inventory.
Each group free to choose prior NBE and ocean fluxes

#### Includes four MIP experiments that use different datasets:

- In situ (IS)
- Land nadir + land glint (LNLG)
- Land nadir + land glint + in situ (LNLGIS)
- Land nadir + land glint + ocean glint + in situ (LNLGOGIS)

#### Data coverage over 2015-2020



## Methods - CO<sub>2</sub> flux inversions

Each modeling group estimates the Net Carbon Exchange (NCE) = Fossil Fuel + Net Biosphere Exchange

- Estimates provided on a 1º x 1º grid.
- We aggregate to country totals.
- Take model median as best estimate.
- Uncertainty is estimated as the standard deviation across model estimates.

NCE fluxes Aggregated to Country Totals



Net Carbon Exchange (NCE) for 2015–2020

- <sup>6</sup> Methods carbon stock loss ( $\Delta C_{loss}$ )
- Land carbon stock loss (ΔC<sub>loss</sub>) estimated by combining top-down NCE with other carbon flux datasets.
- Calculate:

 $\Delta C_{loss} = NCE - FF - F_{crop\,trade} - F_{wood\,trade} - F_{rivers\,export}$ 

**FF**: CO<sub>2</sub> emissions from fossil fuels and cement production. (ODIAC w/ fractional uncertainties of Andres et al. (2014))

**F**<sub>crop trade</sub>: lateral flux of carbon due to farming (Deng et al. 2022, assume std = 30%).

**F**wood trade: lateral flux of carbon due to wood harvesting. (Deng et al. 2022, assume std = 30%).

**F**<sub>rivers export</sub>: lateral flux of carbon due to rivers. (mean of Deng et al. 2022 and DLEM, Uncertainty = absolute difference)

Andres et al. (2014), Tellus B, <u>https://doi.org/10.3402/tellusb.v66.23616</u> Deng et al. (2022), ESSD, <u>https://doi.org/10.5194/essd-14-1639-2022</u>

#### Carbon fluxes for a given land region



### Results – carbon stock loss ( $\Delta C_{loss}$ )

#### **Example 2015–2020 Carbon Budgets for Four Countries**

- Recall:  $FF + F_{crop trade} + F_{wood trade} + F_{rivers export} + \Delta C_{loss} = NCE$
- Figure below shows how each component contributes to the NCE for a few specific countries, constrained by atmospheric CO<sub>2</sub> measurements.
- Increasing land carbon stocks decrease NCE relative to FF emissions for USA, but the opposite occurs for Indonesia.



### Results – carbon stock loss ( $\Delta C_{loss}$ )

**Example Carbon Budget Time Series for Four Countries** 

- Provide annual net fluxes for six years covering 2015 through 2020.
- Interannual variations in NCE are driven primarily ΔC<sub>loss</sub> due to climate variability and trends in FF.
- Droughts reduce carbon uptake by the ecosystem. Variability associated with El Niño in the tropics is a strong driver of variability in ΔC<sub>loss</sub>.



# Lessons learned and path forward

#### Lots of Obs in pipeline

- Data-dense GeoCarb, CO2M and GOSAT-GW
- Regional expansions of in situ measurements.

#### Keys to future success:



- Increased ground-based and aircraft-based CO<sub>2</sub> measurements in poorly sampled regions will identify retrieval biases and improve confidence. Some regions show substantial differences between OCO-2 and in situ inversions that are not well understood. Need more independent CO<sub>2</sub> data in tropics.
- Uncertainty quantification should incorporate Bayesian uncertainties. Spread between flux inversion ensemble members largely captures systematic errors (model transport, inversion set-up) but not Bayesian component.
- **Refine inversions systems.** Including adding missing processes (e.g., atmospheric CO<sub>2</sub> production).

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# Where to access the data:

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