

Measurement of GHGs CO₂, N₂O and CH₄

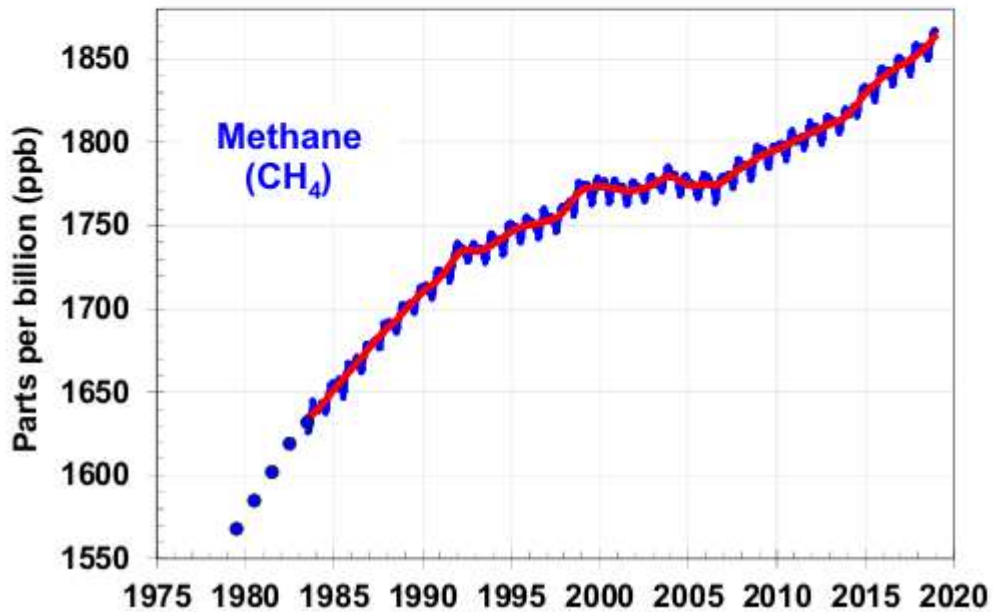
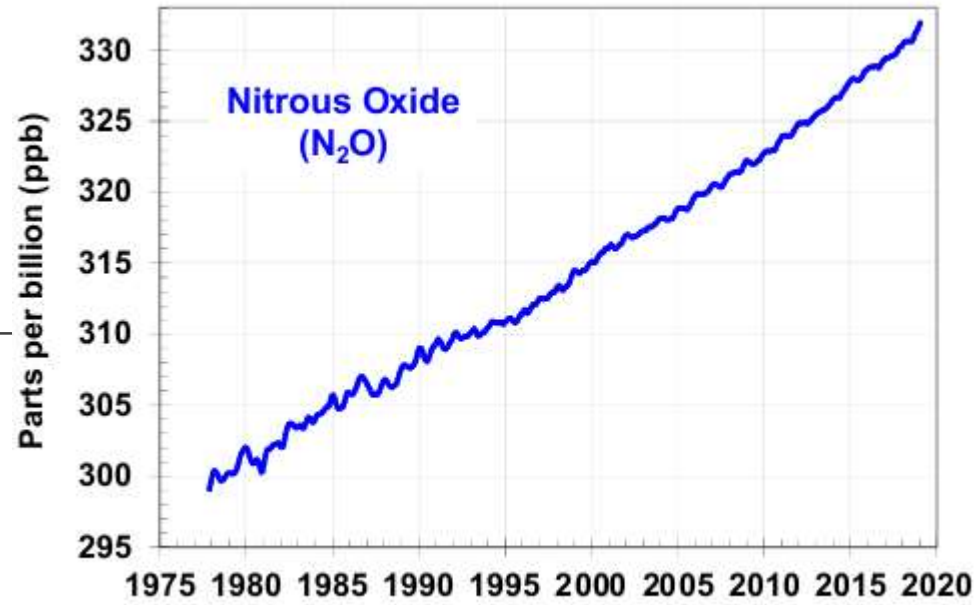
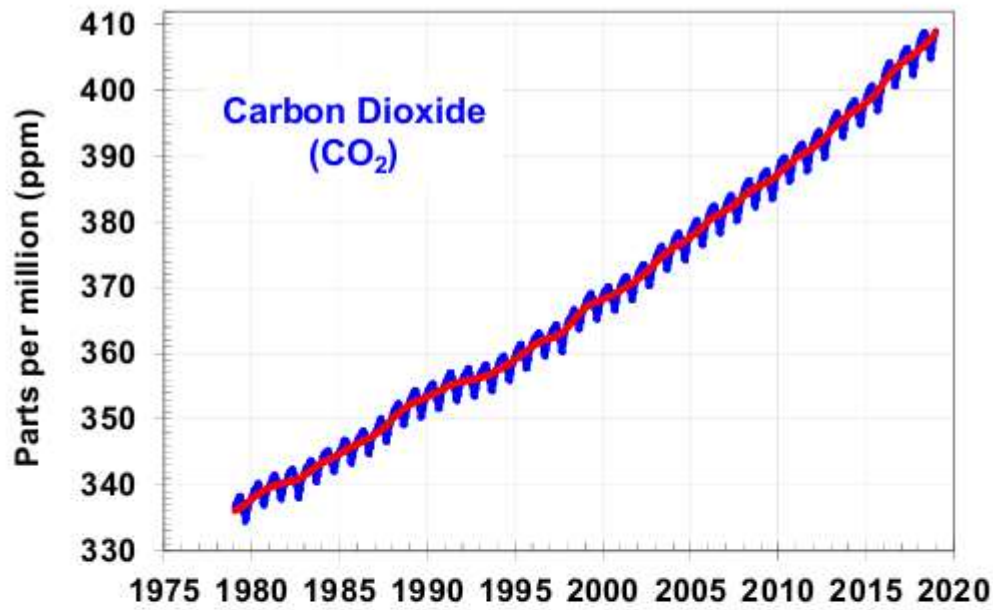
Suman George

MU Murdoch
University



AEGIS (Action on Emission of GHGs for Integrated Sustainability) Research Consortium





26 October 2022 (WMO) - In yet another ominous climate change warning, atmospheric levels of the three main greenhouse gases - carbon dioxide, methane and nitrous oxide all reached new record highs in 2021,



CO₂

CARBON DIOXIDE

The primary greenhouse gas, responsible for about three-quarters of emissions

CH₄

METHANE

Accounts for about 16% of all greenhouse gas emissions

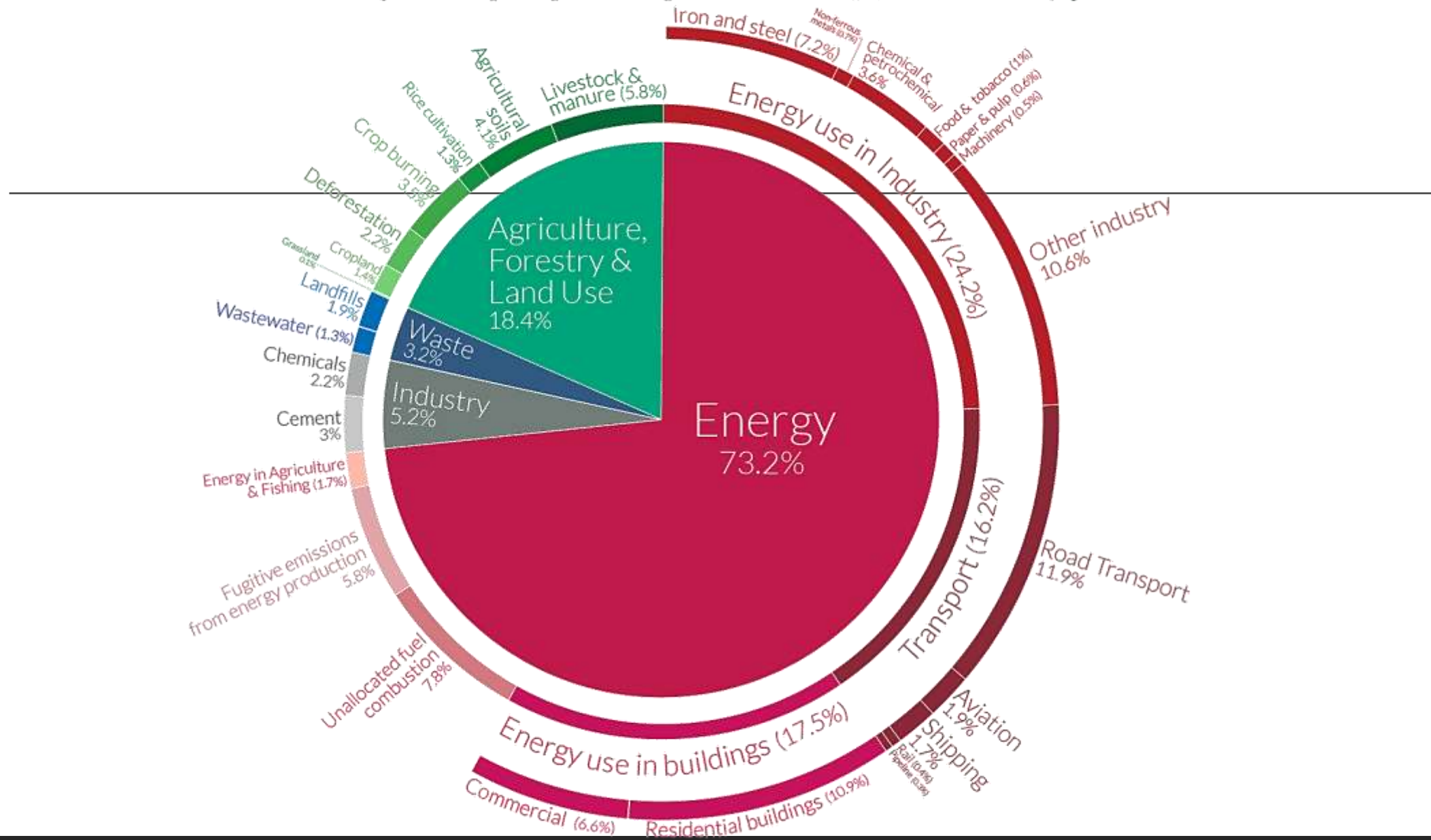
N₂O

NITROUS OXIDE

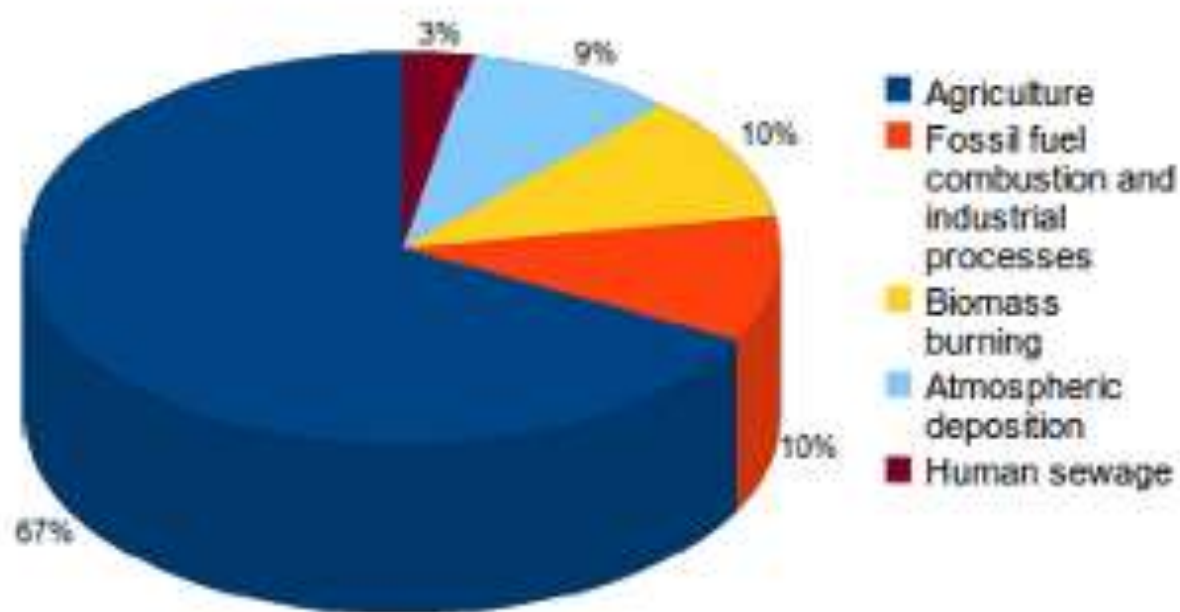
Accounts for only 6% of all greenhouse gas emissions but it is 264 times more powerful than carbon dioxide

Global greenhouse gas emissions by sector

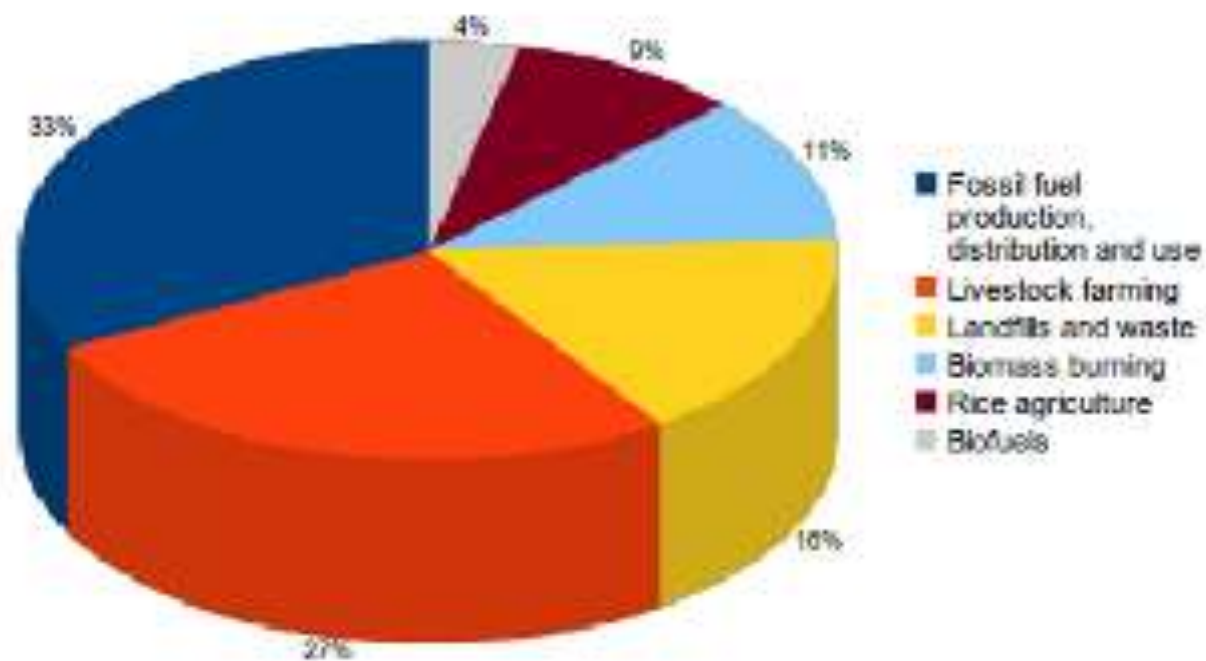
This is shown for the year 2016 – global greenhouse gas emissions were 49.4 billion tonnes CO₂eq.



Human sources of nitrous oxide



Human sources of methane



UNFCCC RESOURCE GUIDE

MODULE 3: NATIONAL GREENHOUSE GAS INVENTORIES FOR PREPARING THE NATIONAL COMMUNICATIONS OF NON-ANNEX I PARTIES

gas-by-gas basis and in units
of mass, estimates of anthropogenic emissions of carbon
dioxide (CO₂), methane (CH₄) and nitrous oxide (N₂O)

N₂O emissions (direct & indirect)

- ❑ N₂O is produced naturally in soils through microbial processes of nitrification, denitrification
- ❑ Main controlling factor → N availability in the soil (depends on N inputs, including N released from mineralization of SOM)
- ❑ Direct & indirect emissions of N₂O from managed soils occur
- ❑ N inputs include: Synthetic and organic fertilizer & N mineralisation associated with land use and/or management change
- ❑ Direct N₂O emissions from mineral soils are estimated when SOM is lost through oxidation, due to land-use or land management changes and this loss is accompanied by a mineralisation of N (F_{SOM})
- ❑ Indirect N₂O emissions occur through 2 pathways: volatilisation & leaching/runoff. Under tier 1, only indirect N₂O emissions from N leached resulting from mineralization of SOM associated with land use/management changes

N₂O emissions (direct & indirect)

$$N_2O - N_{emissions} = F_{SOM} \cdot EF_1 \quad \text{Equation 11.1}$$

Activity Data
Emission Factor

| Emission factor | Default value | Uncertainty range |
|---|---------------|-------------------|
| EF ₁ for N additions from mineral fertilisers, organic amendments and crop residues, and N mineralised from mineral soil as a result of loss of soil carbon [kg N ₂ O-N/kg N ₂] | 0.01 | 0.003 - 0.03 |
| EF _{11R} for flooded rice fields [kg N ₂ O-N/(kg N ₂)] | 0.003 | 0.000 - 0.006 |

$$F_{SOM} = \sum_{LU} \left[\left(\Delta C_{Mineral,LU} \cdot \frac{1}{R} \right) \cdot 1000 \right] \quad \text{Equation 11.8}$$

F_{SOM} : The net annual amount of N mineralised in mineral soils as a result of loss of SOC associated with change in land use and/or management system of practices, kg N.
 $\Delta C_{Mineral,LU}$: SOM oxidised in mineral soils as a consequence of land use and/or management change. This term is calculated by applying the methodology described in previous slides for estimating SOC changes, t C.
 R: The C:N ratio of the soil organic matter.

The IPCC default value is **15** for forest land/grassland conversion to cropland & **10** for management changes in cropland

- To convert kg of N₂O-N emissions into tonnes of N₂O emissions, the result of equation 11.1 needs to be multiplied by 44/28 and by 10⁻³

N₂O emissions (direct & indirect)

$$N_2O - N = F_{SOM} \cdot Frac_{LEACH(N)} \cdot EF_5$$

Equation 11.10

- Input data needed are AD, leaching fraction and EF
- F_{SOM} is the same calculated for direct N₂O emissions

Activity Data
Emission Factor

$N_2O_{(L)} - N$: Annual amount of N₂O-N produced from leaching and runoff of N released from SOM mineralized, as consequence of land use and/or management change, in regions where leaching/runoff occurs, kg N₂O-N yr⁻¹.
 $Frac_{LEACH(N)}$: Fraction of all N mineralised from SOC losses in mineral soils, associated with changes of land use and/or management change, that is leached and runoff, kg N (kg of N additions)⁻¹ (Table 11.3).
 EF_5 : emission factor for N₂O emissions from N leaching and runoff, kg N₂O-N(kg N leached and runoff)⁻¹ (Table 11.3).

| Leaching fraction kg N (kg N additions or deposition by grazing animals) ⁻¹ | Used for | Value |
|--|---|-------|
| $Frac_{LEACH(N)}$ | N losses by leaching / runoff for regions where soil water-holding capacity is exceeded | 0.30 |

| Factor | Default value | Uncertainty range |
|--|---------------|-------------------|
| EF_5 [leaching/runoff], kg N ₂ O-N (kg N leaching/runoff) ^{-1, 23} | 0.0075 | 0.0005 - 0.025 |

CH₄ emissions

- ❑ CH₄ emissions from mineral soils occur on Inland Wetland Mineral Soils (IWMS) that are rewetted (e.g., for cultivation of crops)
- ❑ Management activities that alter the water table on lands containing IWMS can impact CH₄ emissions*
- ❑ IWMS are aquic soils (USDA) or gleysols (World Reference Base), having restricted drainage, leading to periodic flooding and anaerobic conditions
- ❑ Only 2013 IPCC Wetlands Supplement provides default methodology for estimating CH₄ emissions from IWMS
- ❑ Recall that CH₄ emissions from rice cultivations are reported under the agriculture sector
- ❑ IWMS might occur in any of the six land-use categories

CH₄ emissions

$$CH_{4-IWMS} = \sum_c (A_{IWMS} \times EF_{CH_4-IWMS})_c$$

2013 IPCC Supplement on Wetlands, chapter 5, Equation 5.1

Activity Data Emission Factor

CH_{4-IWMS}: Annual CH₄ emissions from managed lands on IWMS where management activities have raised the water table level to or above the land surface, kg CH₄ yr⁻¹.

A_{IWMS}: Total area of managed lands with mineral soil where the water table level has been raised, ha.

EF_{CH₄-IWMS}: Emission factor from managed lands with mineral soil where water table level has been raised, kg CH₄ ha⁻¹ yr⁻¹ (Table 5.4 of 2013 IPCC Supplement on Wetlands).

c: Climate region.

Land representation

- The area of managed lands with IWMS or dry mineral soil, where water table level has been raised, should be stratified by climate region

TABLE 5.4
DEFAULT EMISSION FACTORS FOR CH₄ FROM MANAGED LANDS WITH IWMS WHERE WATER TABLE LEVEL HAS BEEN RAISED

| Climate Region | EF _{CH₄-IWMS} (kg CH ₄ ha ⁻¹ yr ⁻¹) | 95% Confidence Interval ^A | Number of Studies |
|----------------|--|--------------------------------------|-------------------|
| Boreal | 76 | ±76 ^B | 1 ^C |
| Temperate | 235 | ±108 | 21 |
| Tropical | 900 | ±456 | 18 |

^AThe 95% confidence interval is calculated from the mean, standard deviation, and the critical values of the t distribution, according to the degrees of freedom. These are not expressed as a percentage of the mean.

^BBridgham *et al.* (2006).

^CThis study (Bridgham *et al.*, 2006) is a synthesis of numerous studies; see publication for details.

IPCC OUTLINES A TIERED APPROACH FOR ESTIMATING GHG EMISSIONS

TIER 1

MULTIPLYING
ACTIVITY DATA (AD)
BY AN
INTERNATIONAL
DEFAULT FACTOR
REPRESENTING
EMISSIONS PER UNIT
OF ACTIVITY

TIER 3

MORE DETAILED -
FURTHER
STRATIFICATION OF THE
AD, AND EMISSION
ESTIMATES, DIRECT
MEASUREMENT, OR
OTHER EQUIVALENT
REGION-SPECIFIC
APPROACHES.

TIER 2

APPROACHES
GENERALLY INVOLVE
THE APPLICATION OF
A COUNTRY-SPECIFIC
EMISSION FACTOR TO
NATIONAL- OR
REGIONAL-LEVEL AD;

How to ground truth the modelled information

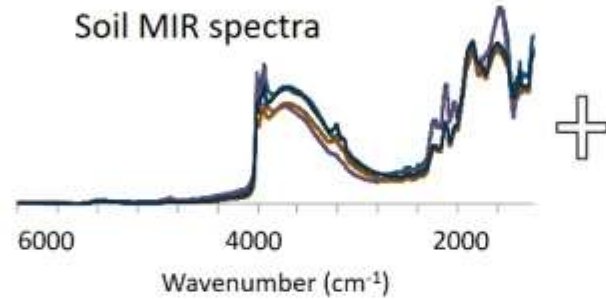
Tier one - Multiplying activity data by an international default factor representing emissions per unit of activity.

Use a Tier 3 level tool to ground truth will be obviously very different

With Tier 1 – we are trying to generate a base map when there is none with intention to progress towards higher Tiers

Even then we need to develop the capability to ground truth for carbon offsetting/carbon trading, helping in the transition to higher tiers, this is also an opportunity to improve the model.

Technologies: Soil spectroscopy



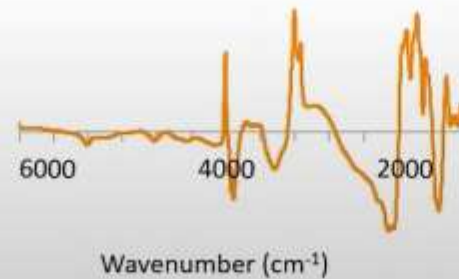
Analytical Data

| | <u>OC (%)</u> |
|--------|---------------|
| Soil 1 | 20.5 |
| Soil 2 | 14.8 |
| Soil 3 | 12.7 |
| Soil 4 | 10.0 |
| ⋮ | ⋮ |

↓
Multivariate Analysis



Calibration Model



Regression Coefficients

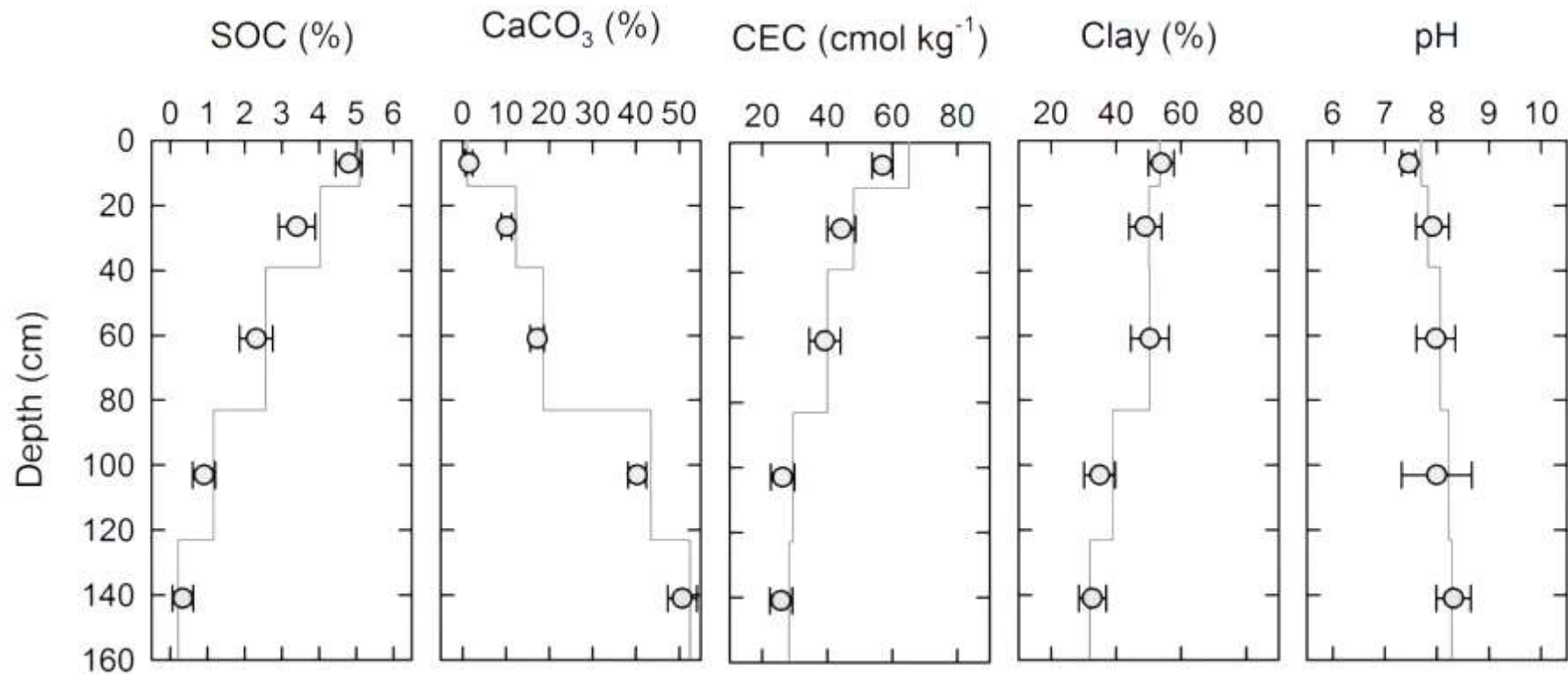
Soil spectra
from samples
with no
analytical data

→

Predicted Data

| | <u>OC (%)</u> |
|---------|---------------|
| Soil 21 | 13.1 |
| Soil 22 | 17.6 |
| Soil 23 | 14.5 |
| Soil 24 | 25.8 |
| ⋮ | ⋮ |

Technologies: Soil spectroscopy



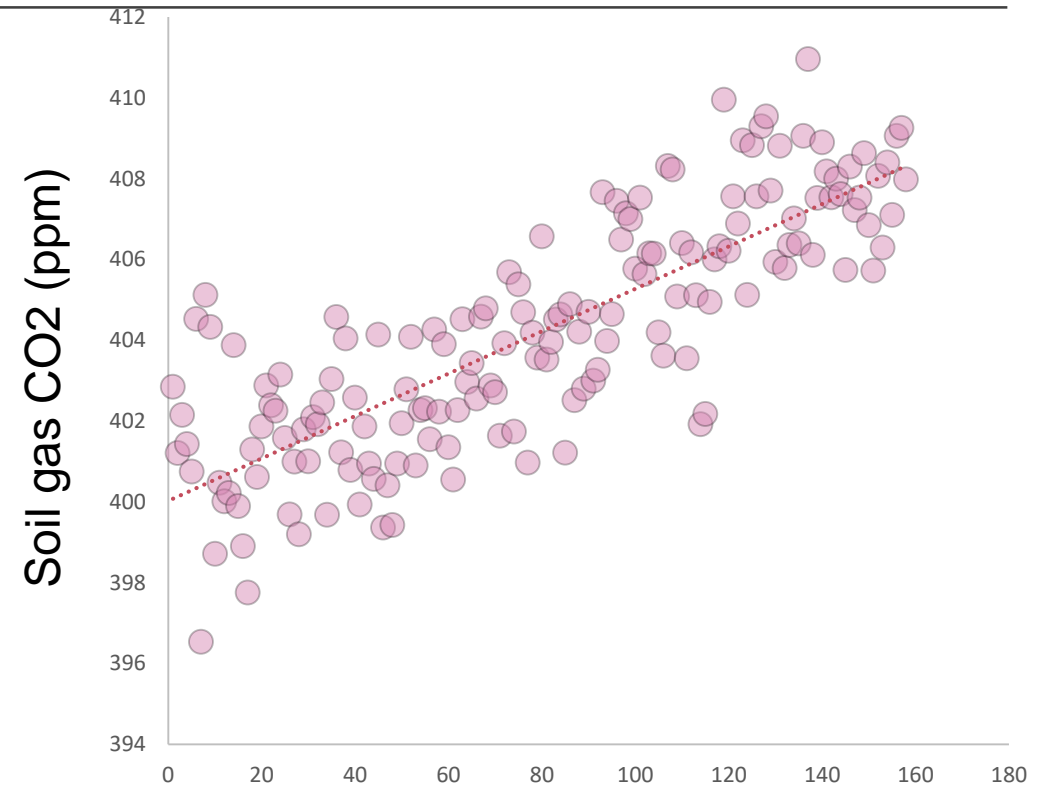
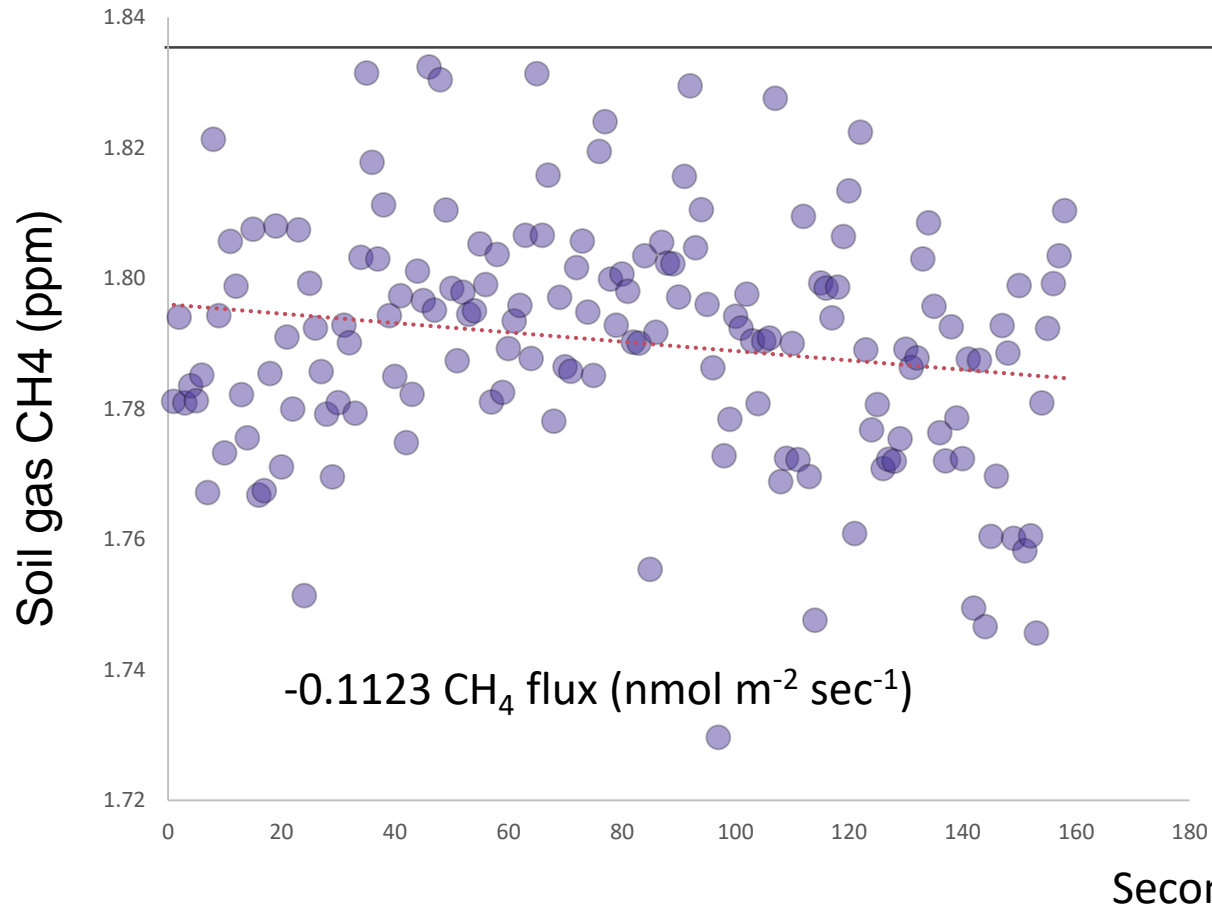
Vertisol, Kerr County, Texas

⊙ Spec-based estimate — Reference analytical data

Source: Sanderman et al. 2020 SSSAJ

Why Measure CO_2 , N_2O and CH_4

Control sites



NEED OF THE HOUR:

**TANDEM GHG MRV WITH IMPLEMENTATION
OF CLIMATE-SMART MITIGATION PROJECTS
IN AGRICULTURE**

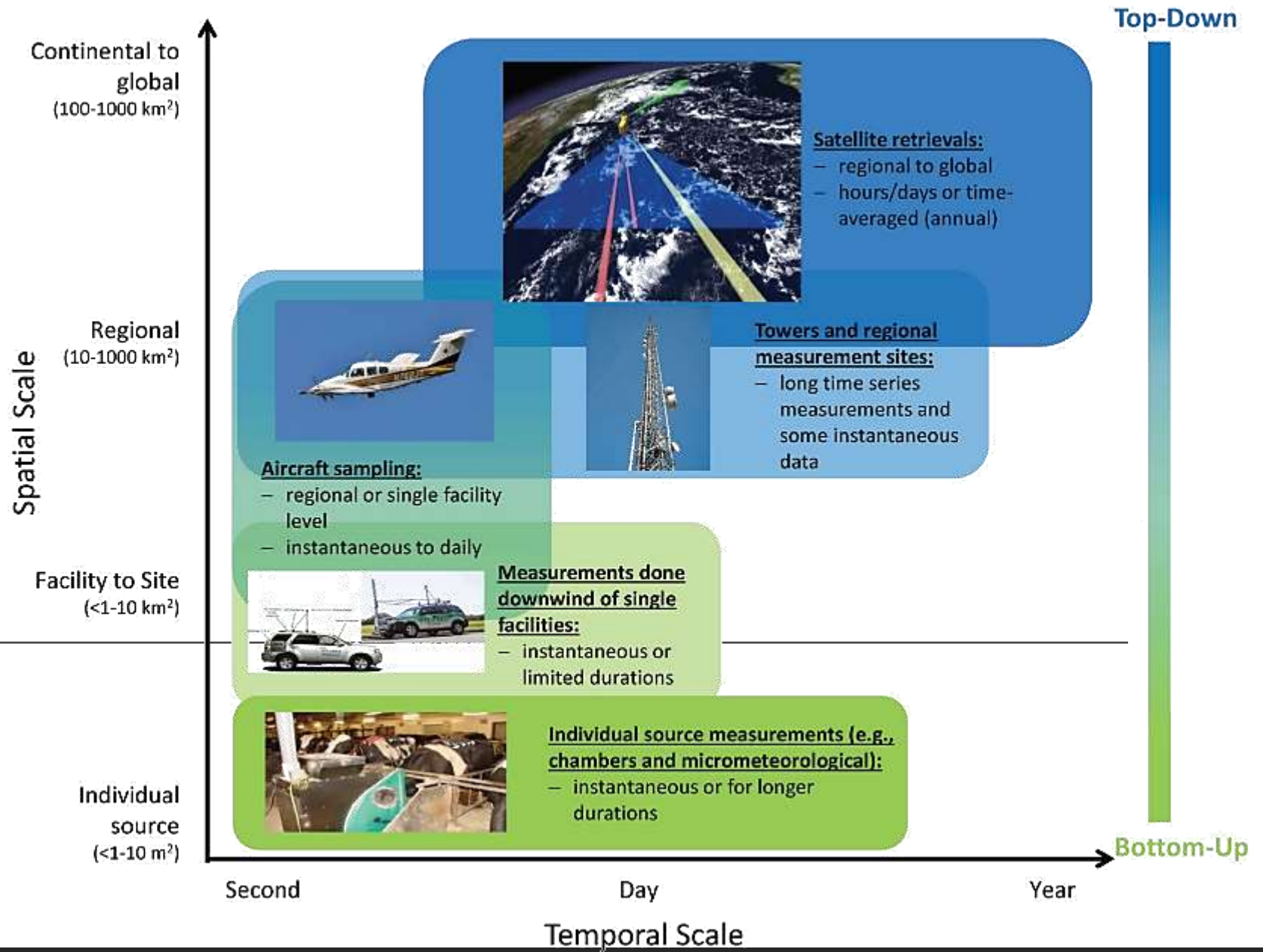


**MONITORING CLIMATE
MITIGATION/ADAPTATION
TARGETS THROUGH THE LIFE
OF A PROJECT**

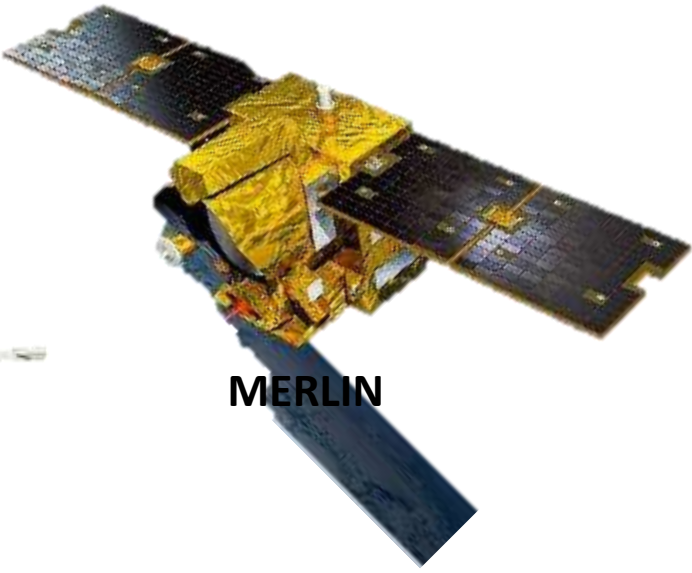
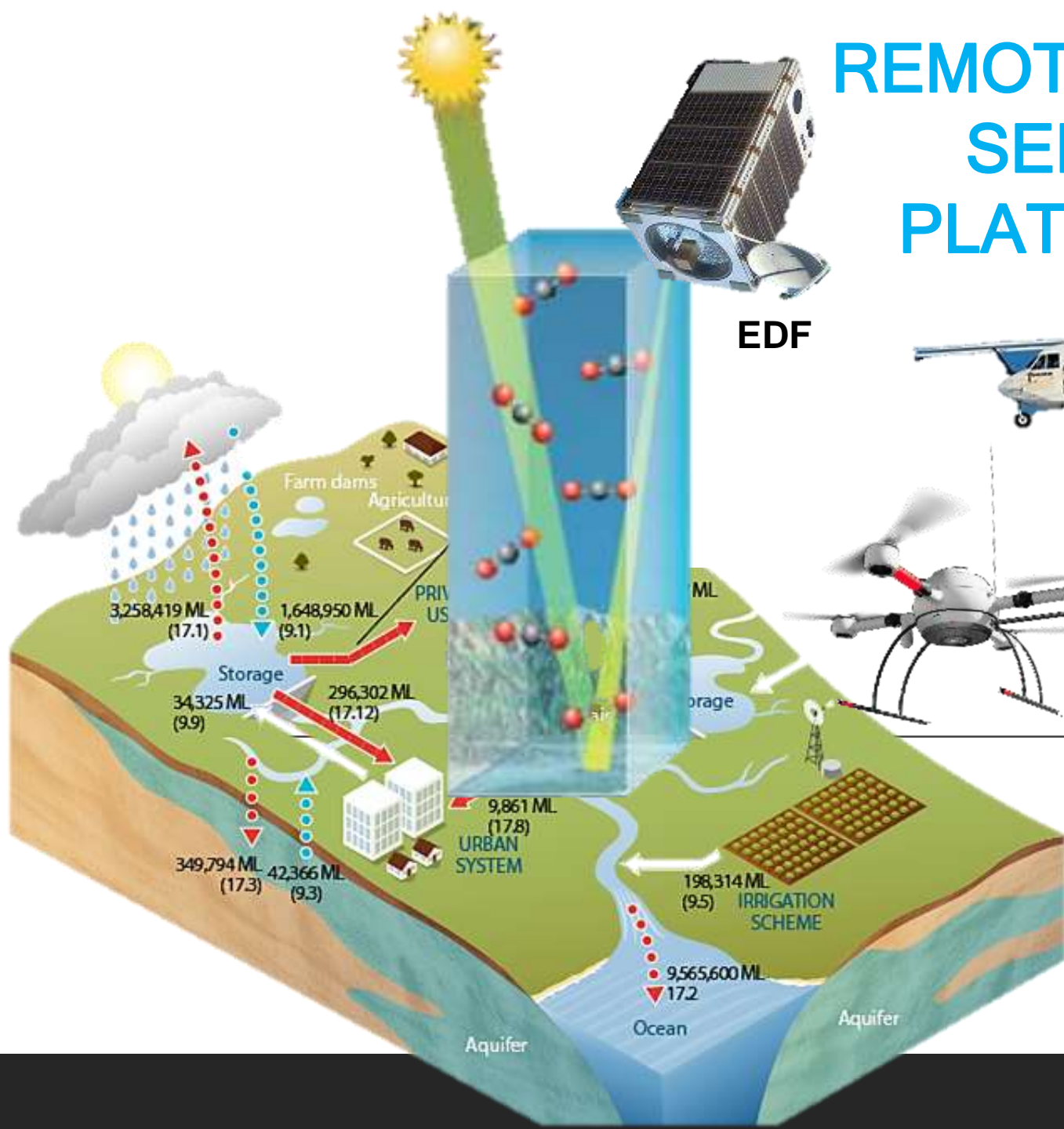
BIG PICTURE SCIENCE: LEARNING FROM THE FEEDBACK LOOP

**USUALLY DURING EXECUTION OF VARIOUS CLIMATE
ADAPTATION/MITIGATION PROJECTS OPPORTUNITY
FOR CONTINUOUS LEARNING AND R&D
DEVELOPMENT IS MISSED.**

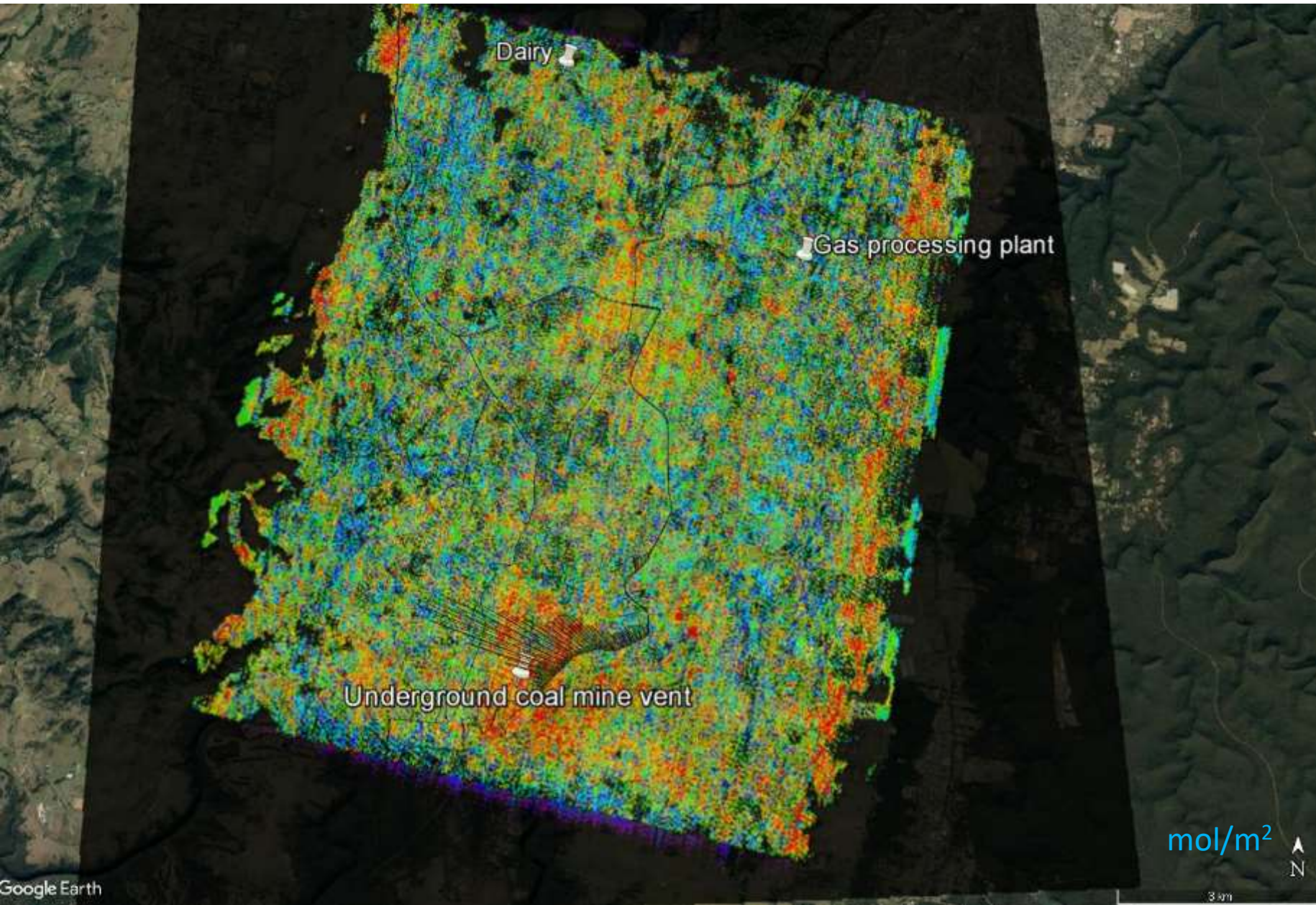
MEASURING GHG IN A 4D SPACE



REMOTE/AERIAL SENSED PLATFORMS



- SCIAMACHY
- SENTINEL 5
- GOSAT
- GeoCARB
- METHANESAT
- IASAI
- GHGsats



CH₄ from GHGSTAT

The measurement of methane from space is a highly demanding task because the sources are usually small, compounded by the large vertical column from space to earth (>500 km),

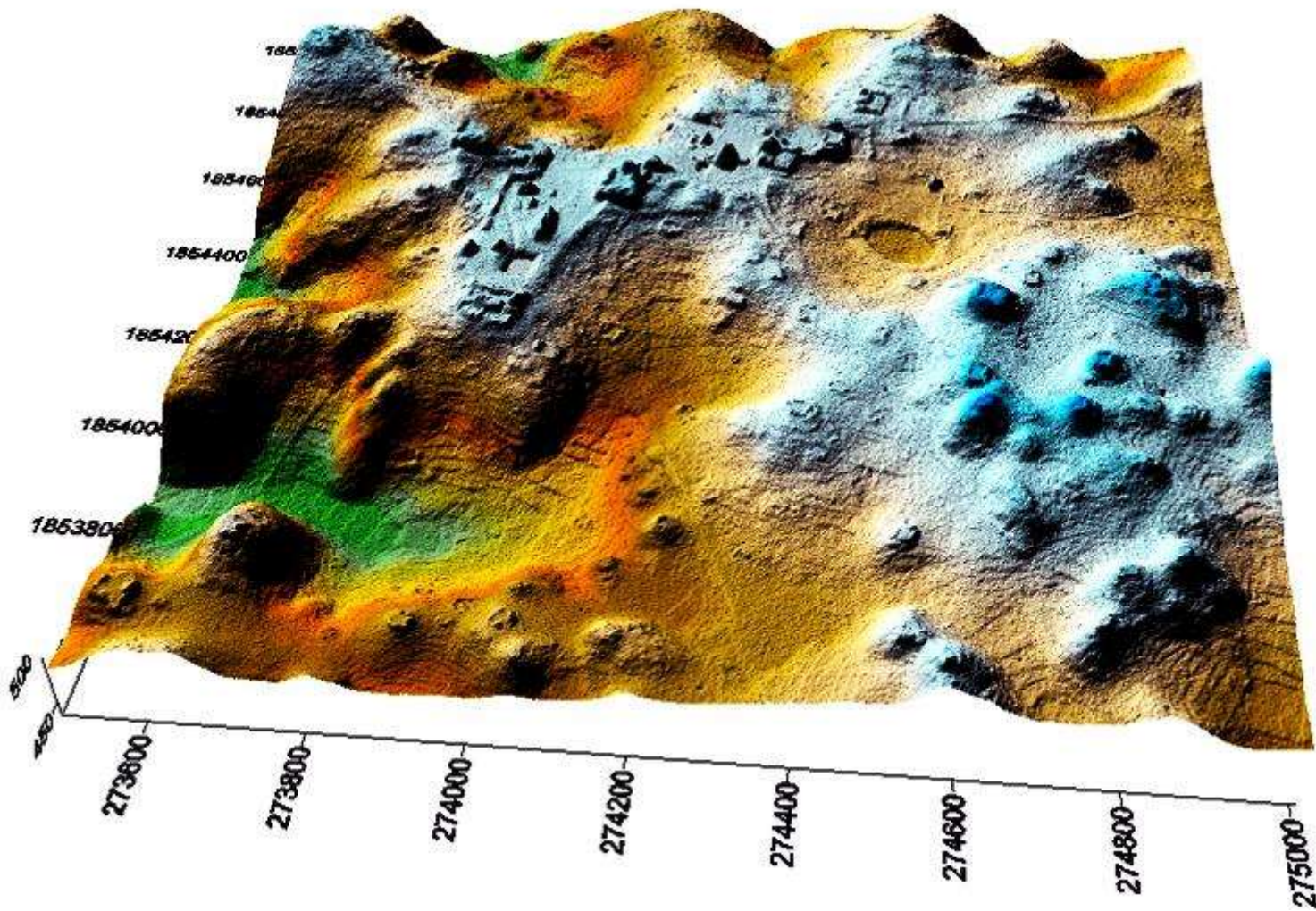
MONITORING SMARTER AND NOT HARDER

TIER 3 LEVEL LOWERING MONITORING UNCERTAINTY

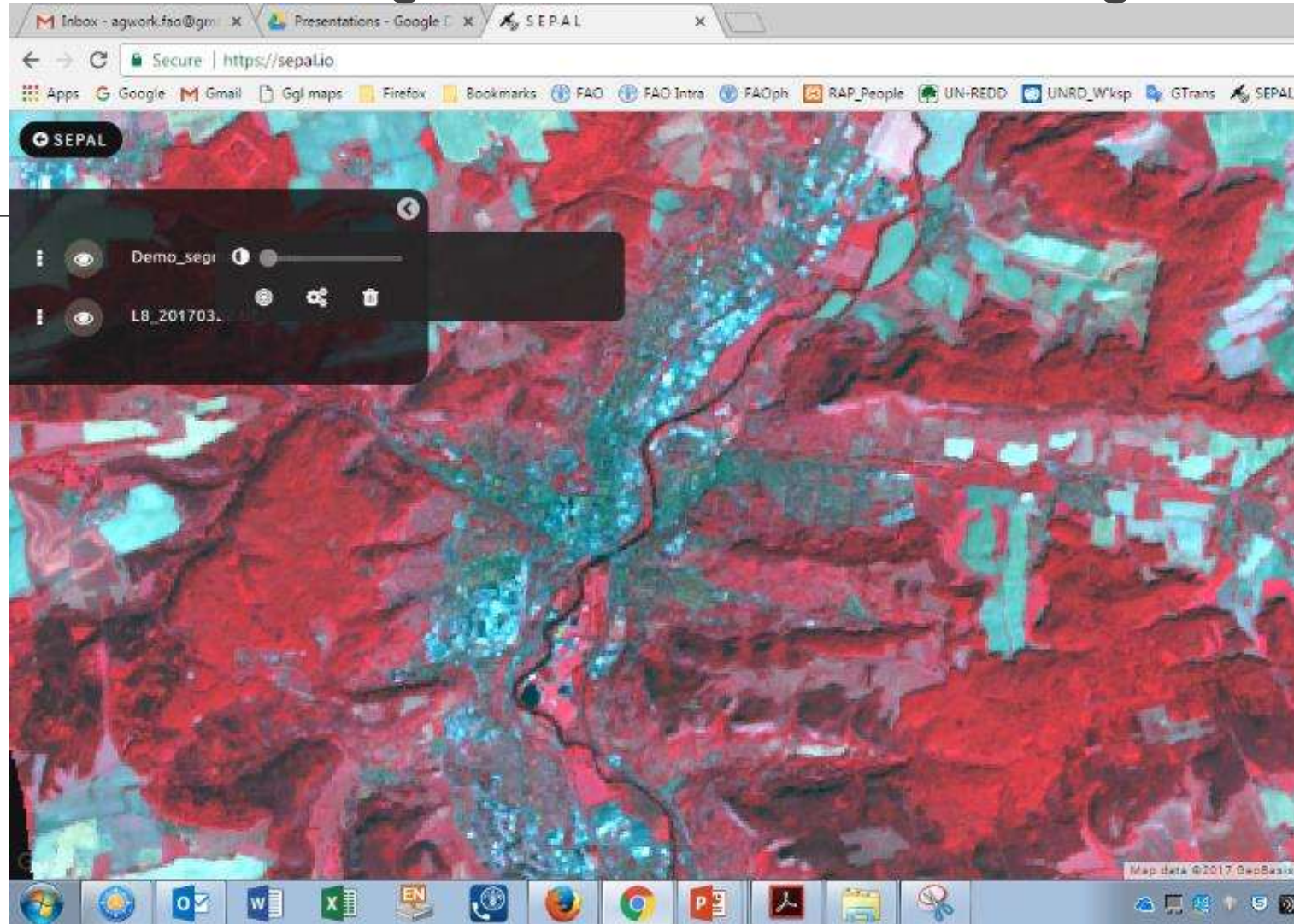
- SMART SAMPLING
- STRATEGIC GROUND-TRUTHING
- IMPROVING MODEL OUTPUT
- MICROBIAL COMPONENT
- REMOTE SENSED PARAMETERS
- SURROGATE VARIABLES
ACCURATE BUT
CHEAP

CITIZEN SCIENCE

A hyperspectral image multispectral remote sensing images to inform about sampling

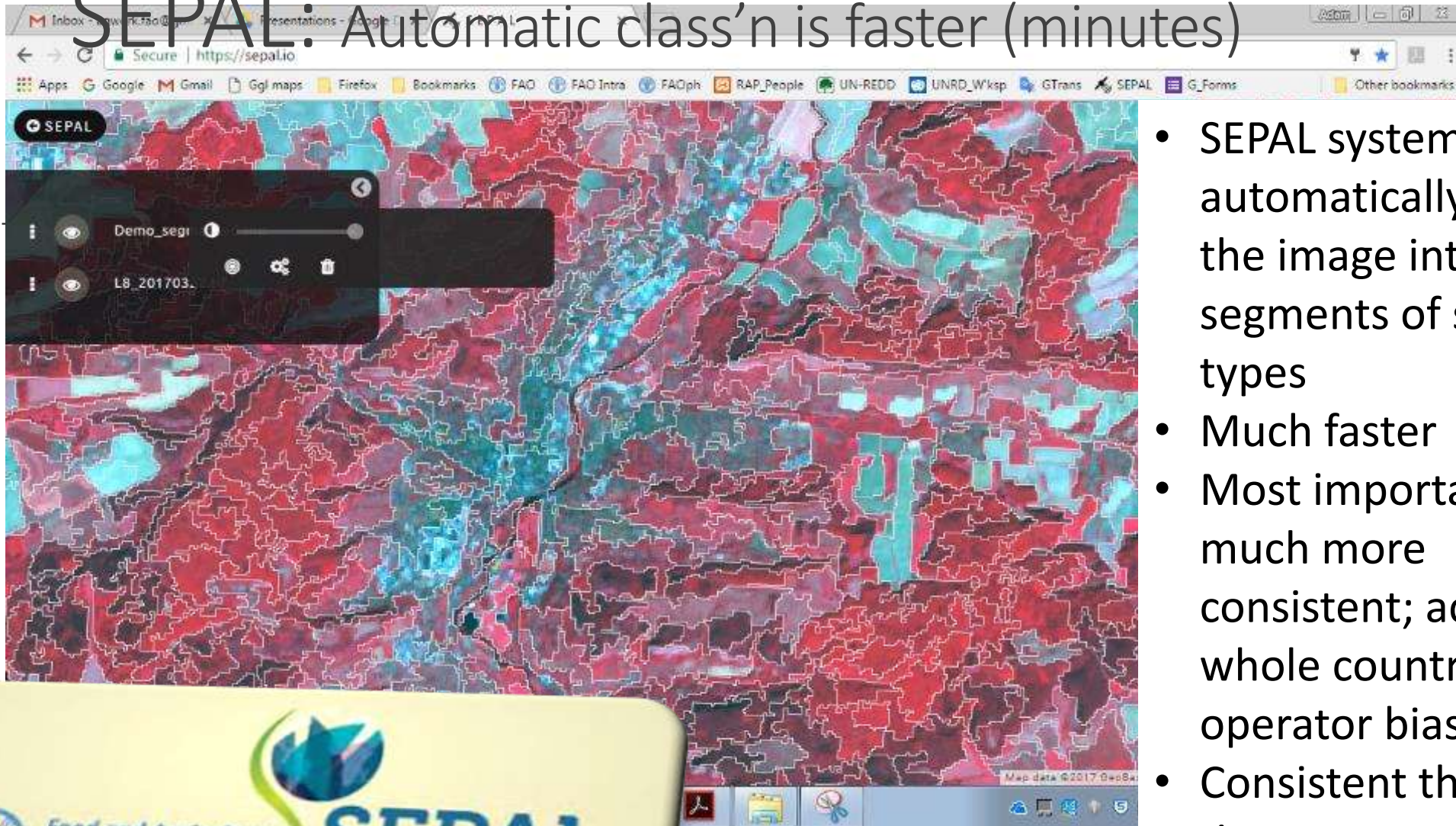


Current Satellite Image classification takes long time



- Typical satellite image in “false colour” where red areas are forest and vegetation
- In Indonesia, assessors manually classify these images visually to areas of similar land cover
- It is difficult, slow work; hard to be consistent over time and between operators

SEPAL: Automatic class'n is faster (minutes)



- SEPAL system can automatically classify the image into segments of similar types
- Much faster
- Most importantly – much more consistent; across the whole country (no operator bias)
- Consistent through time over years



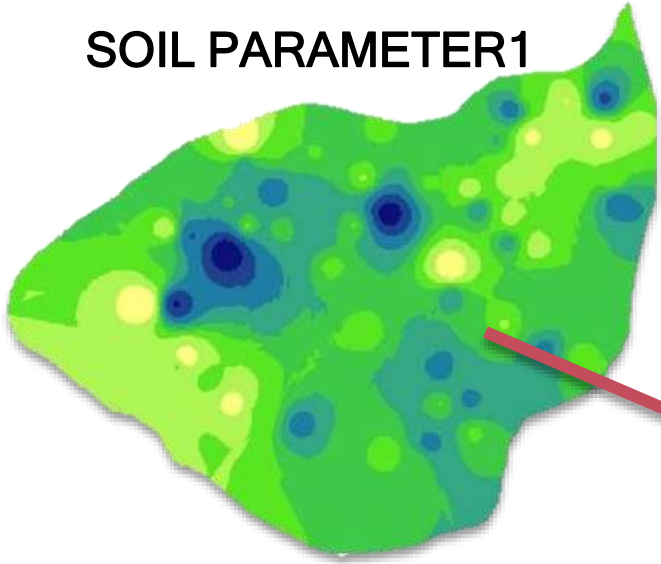
Food and Agriculture
Organization of the
United Nations



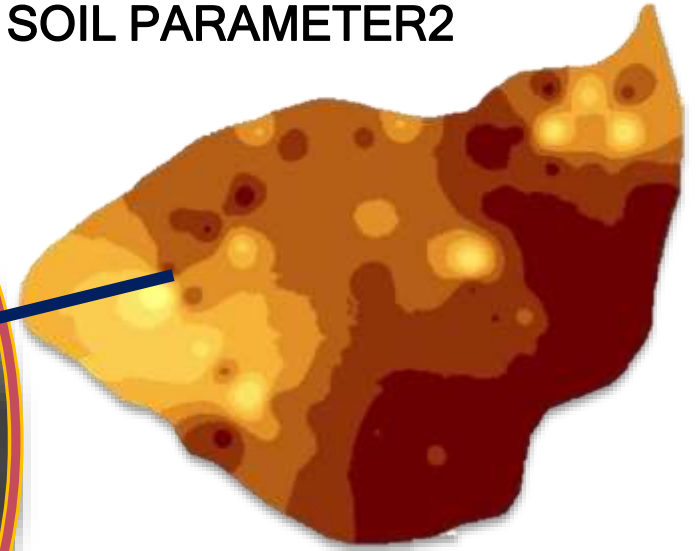
SEPAL

SYSTEM FOR EARTH OBSERVATION
DATA ACCESS, PROCESSING &
ANALYSIS FOR LAND MONITORING

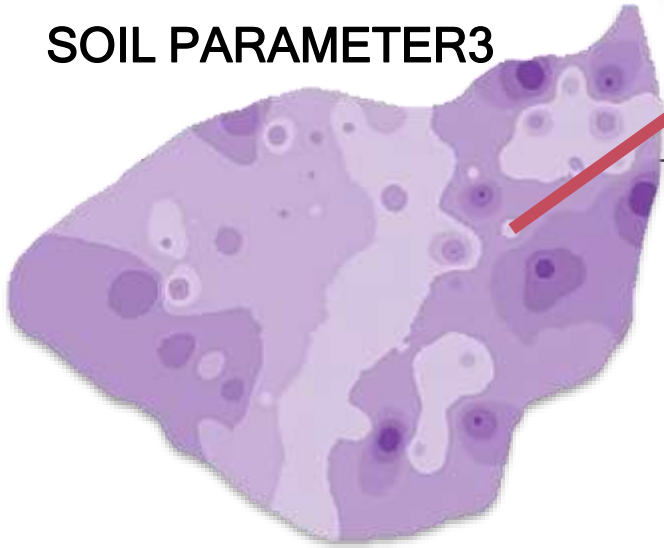
SOIL PARAMETER1



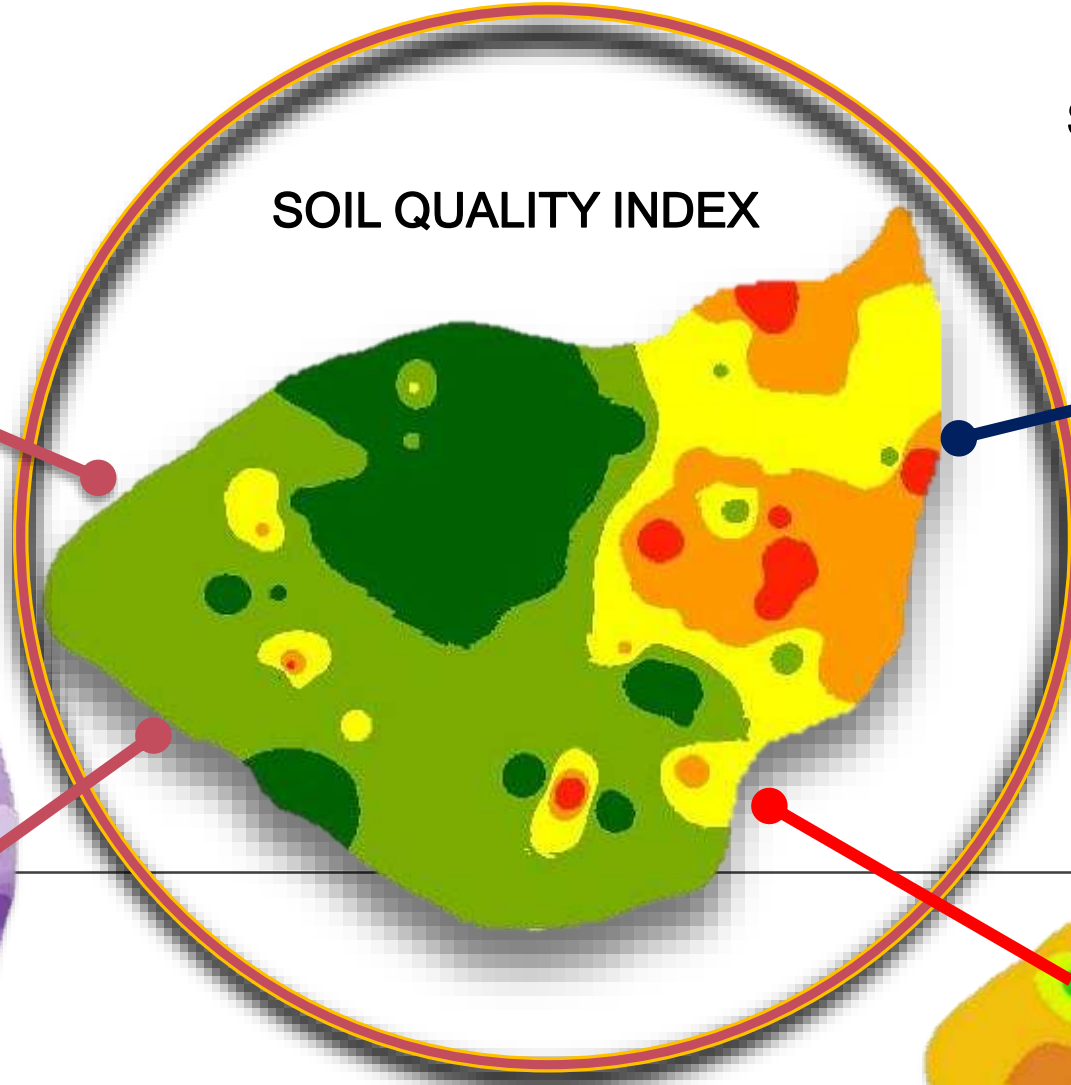
SOIL PARAMETER2



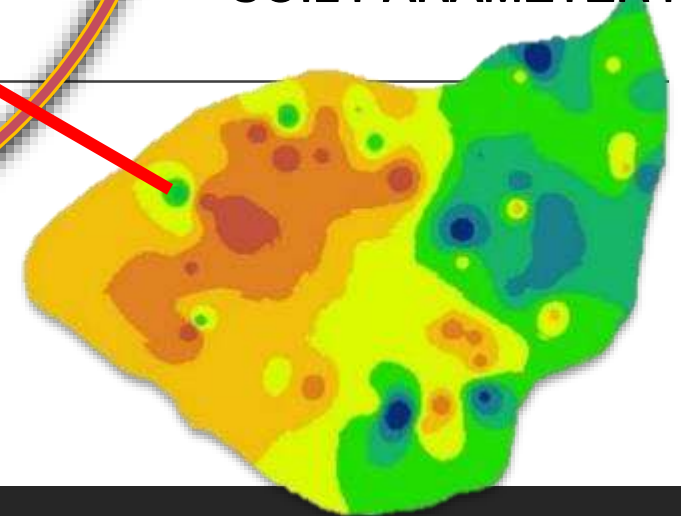
SOIL PARAMETER3



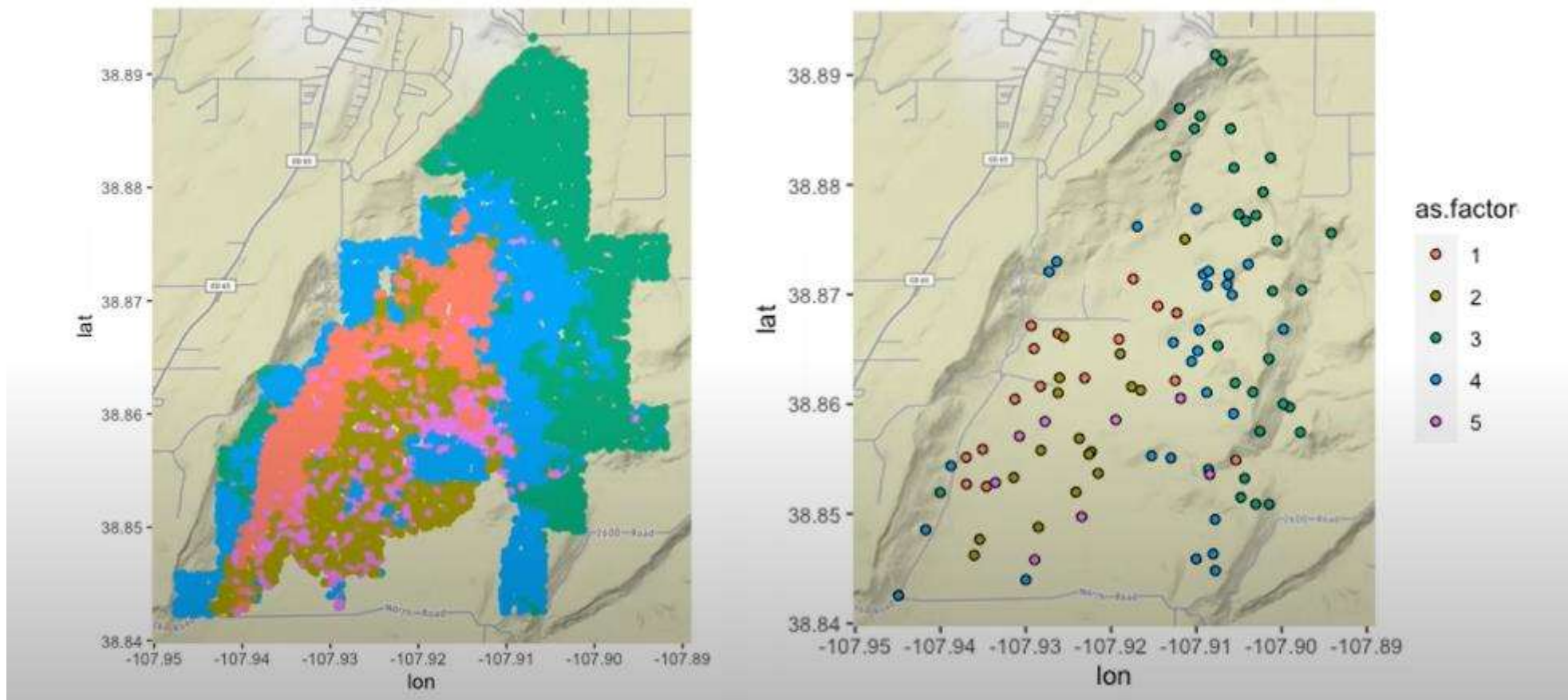
SOIL QUALITY INDEX



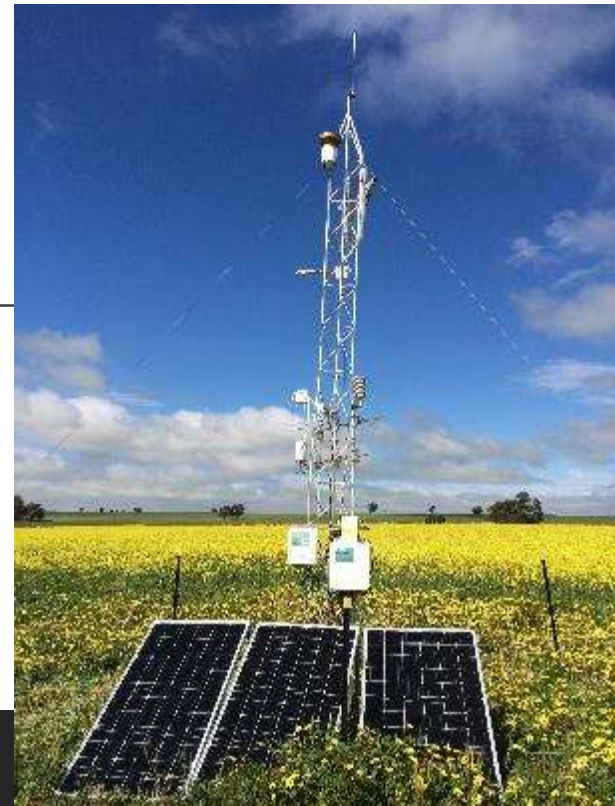
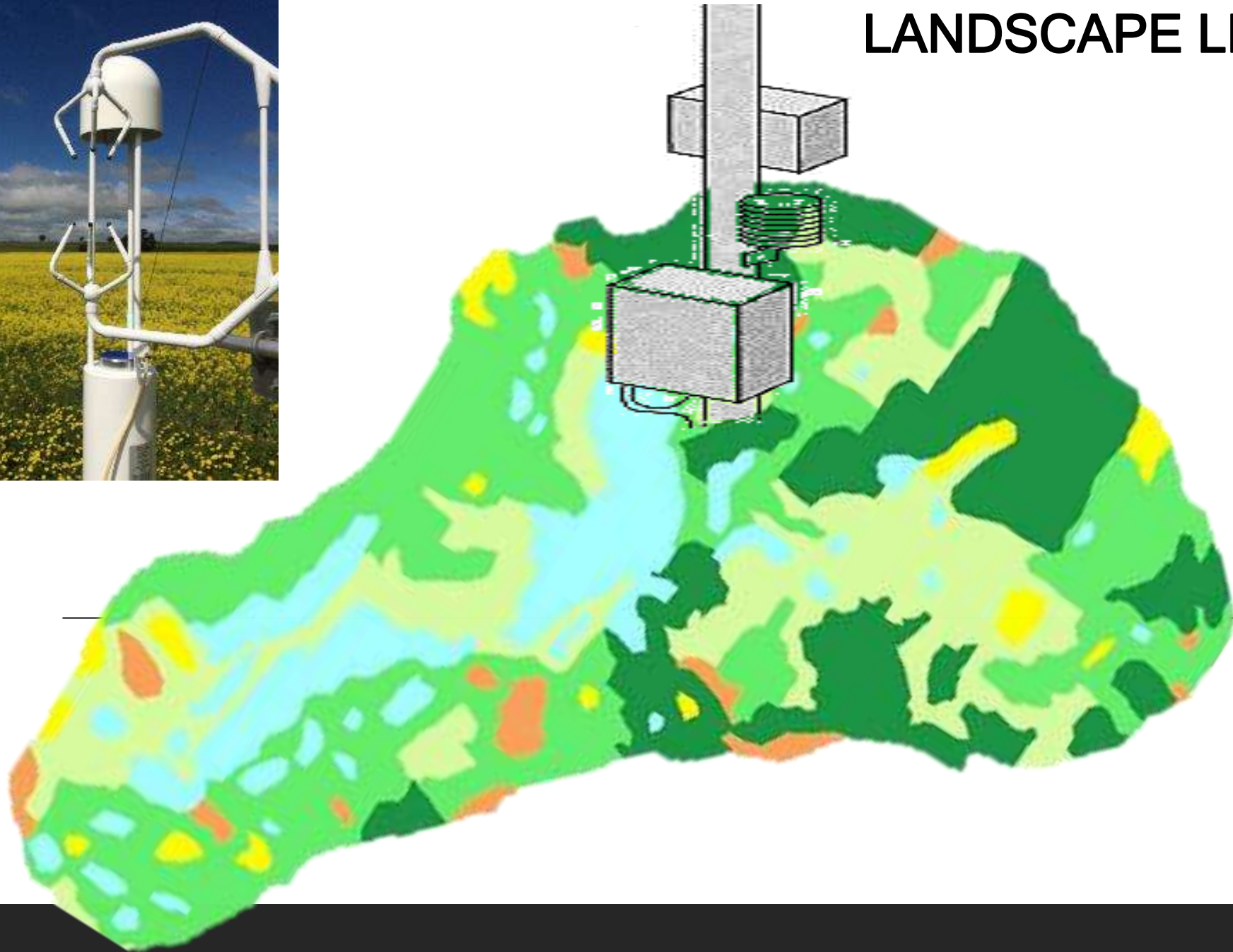
SOIL PARAMETER4

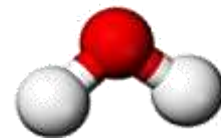
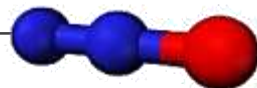


Stratification based upon prior information with data-driven sample selection

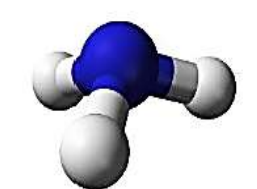
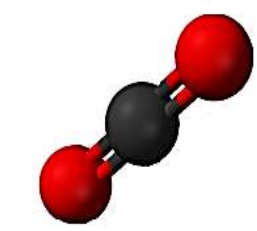
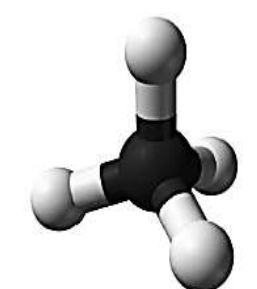
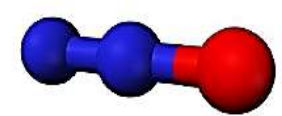


LANDSCAPE LEVEL GHG MRV





Picarro G2508 Analyzer – Portable instrument that can simultaneously measure nitrous oxide (N_2O), carbon dioxide (CO_2), ammonia (NH_3) and water vapor (H_2O) –



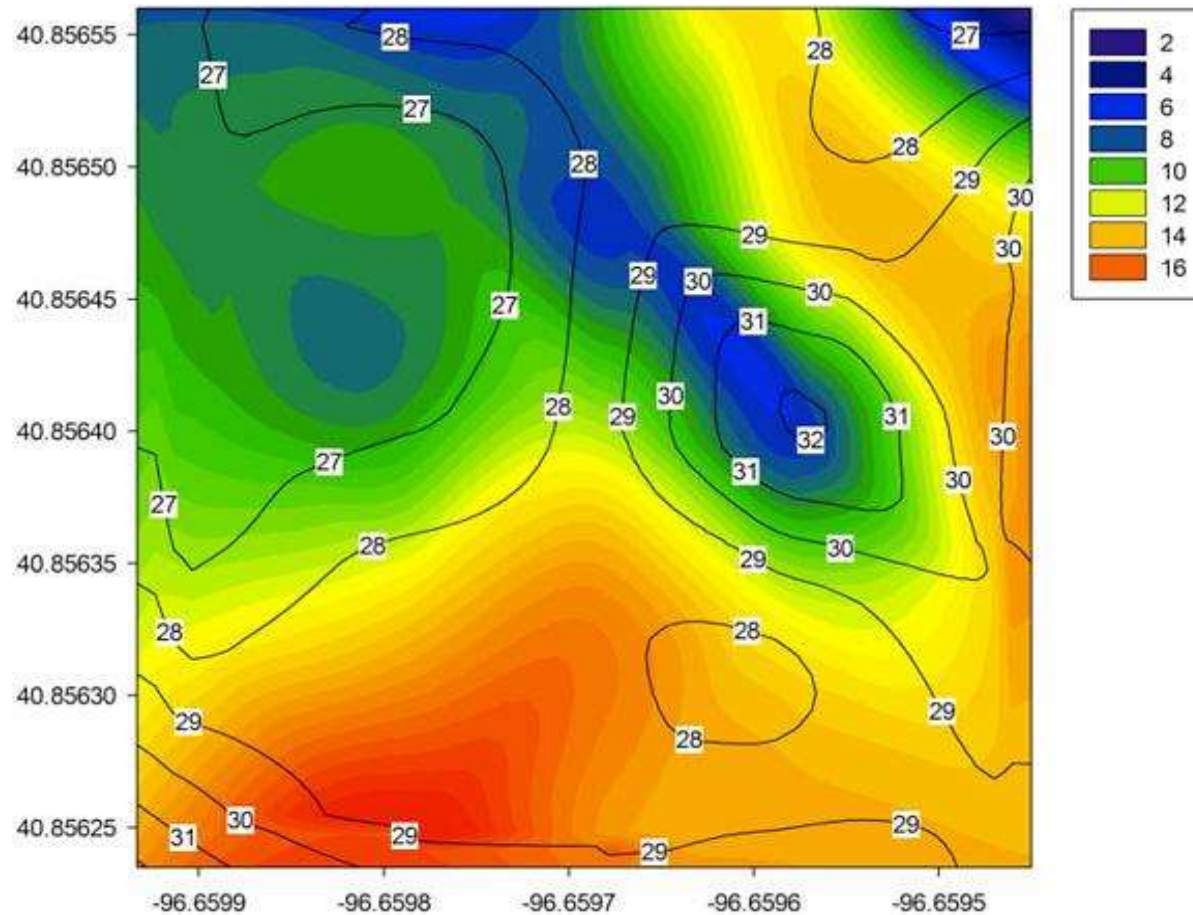
Measurement precision

N_2O : < 5 ppb precision in 5 m

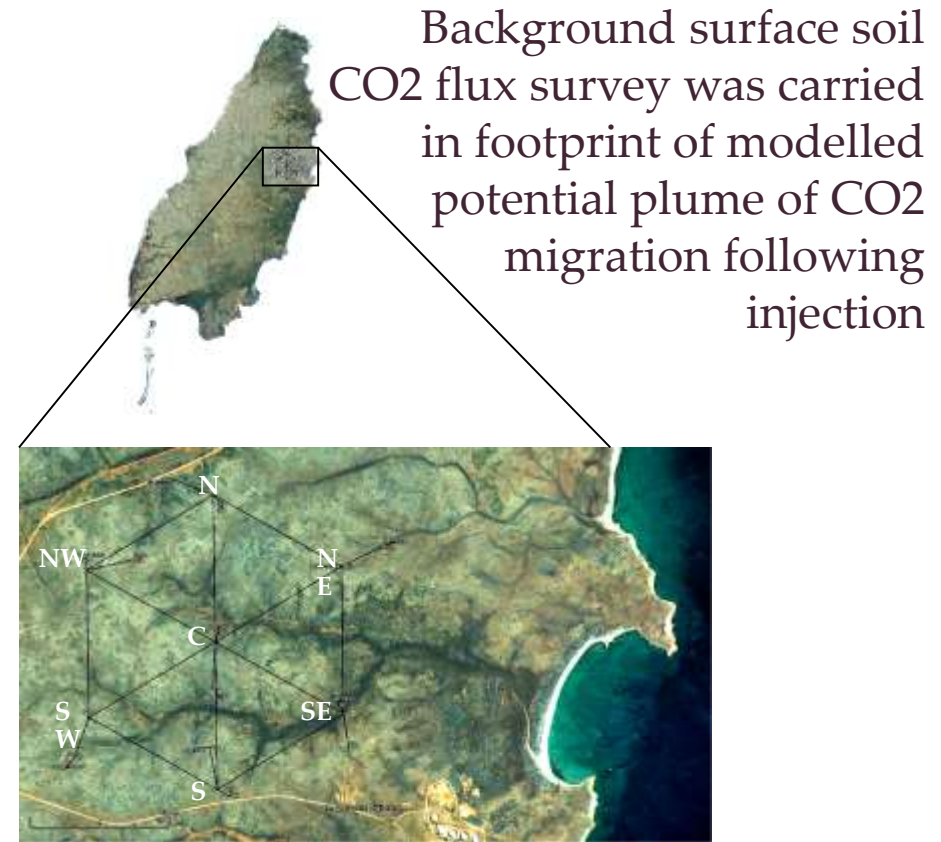
CH_4 : < 5 ppb precision in 5 m

H_2O : < 100 ppm precision in

Soil CO₂ flux /concentration mapping



Evaluating spatial variation in soil CO₂ flux across the Gorgon CO₂ plume area on barrow island, western Australia

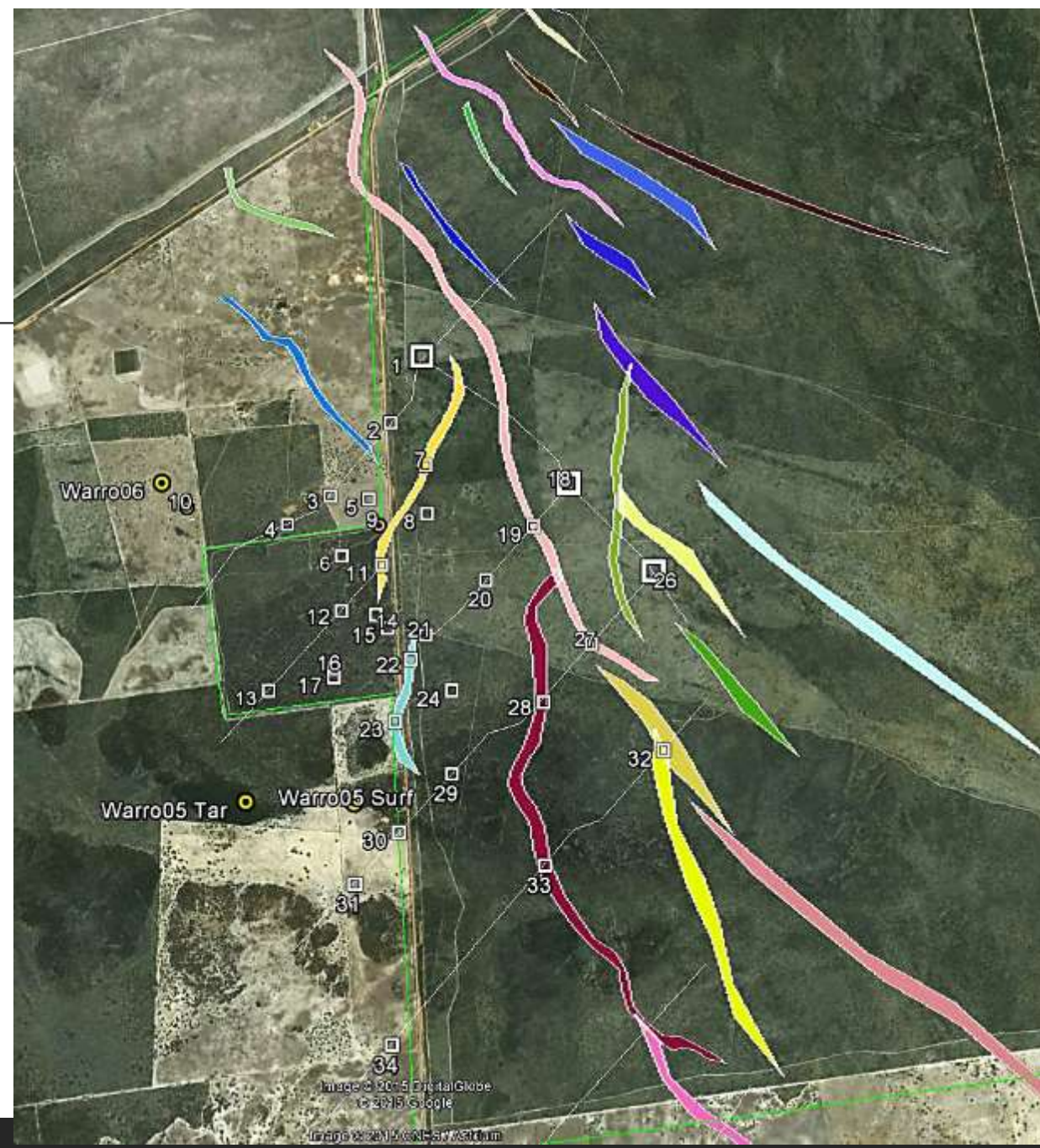


Surface CH₄ survey (Winter) Latent



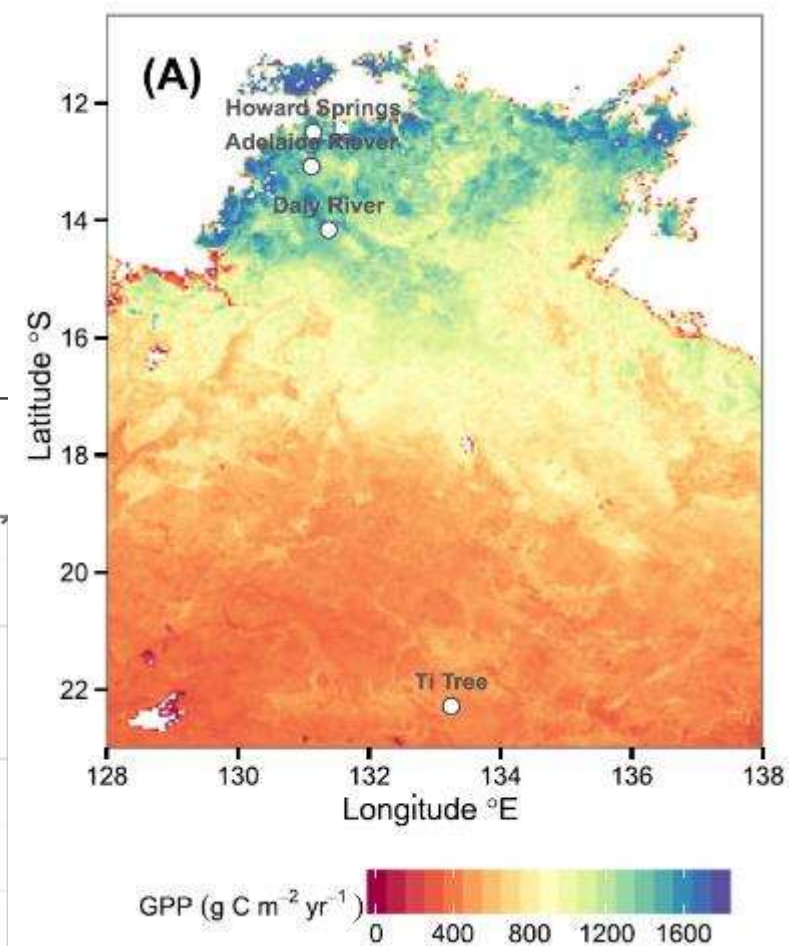
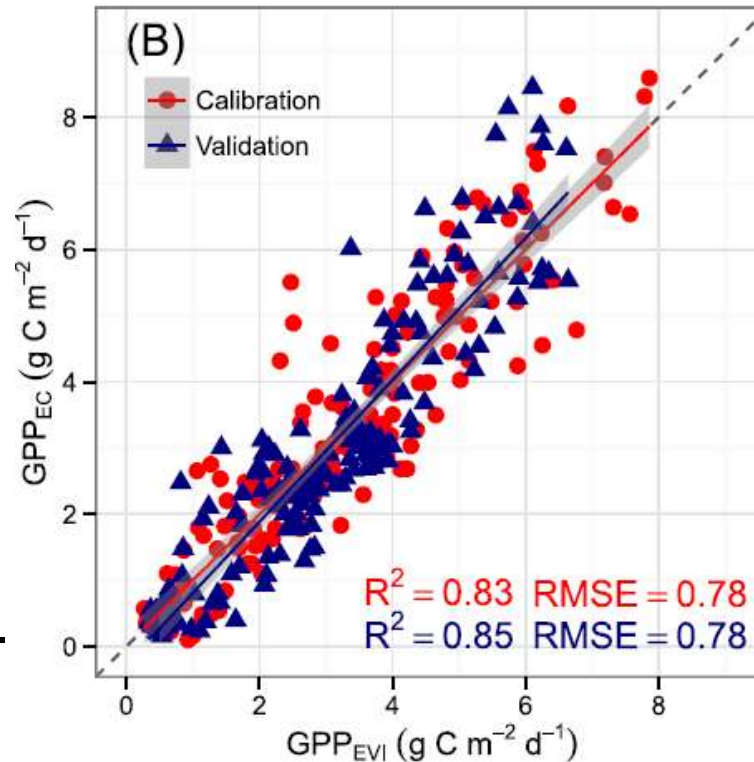
Soil sampled from each location





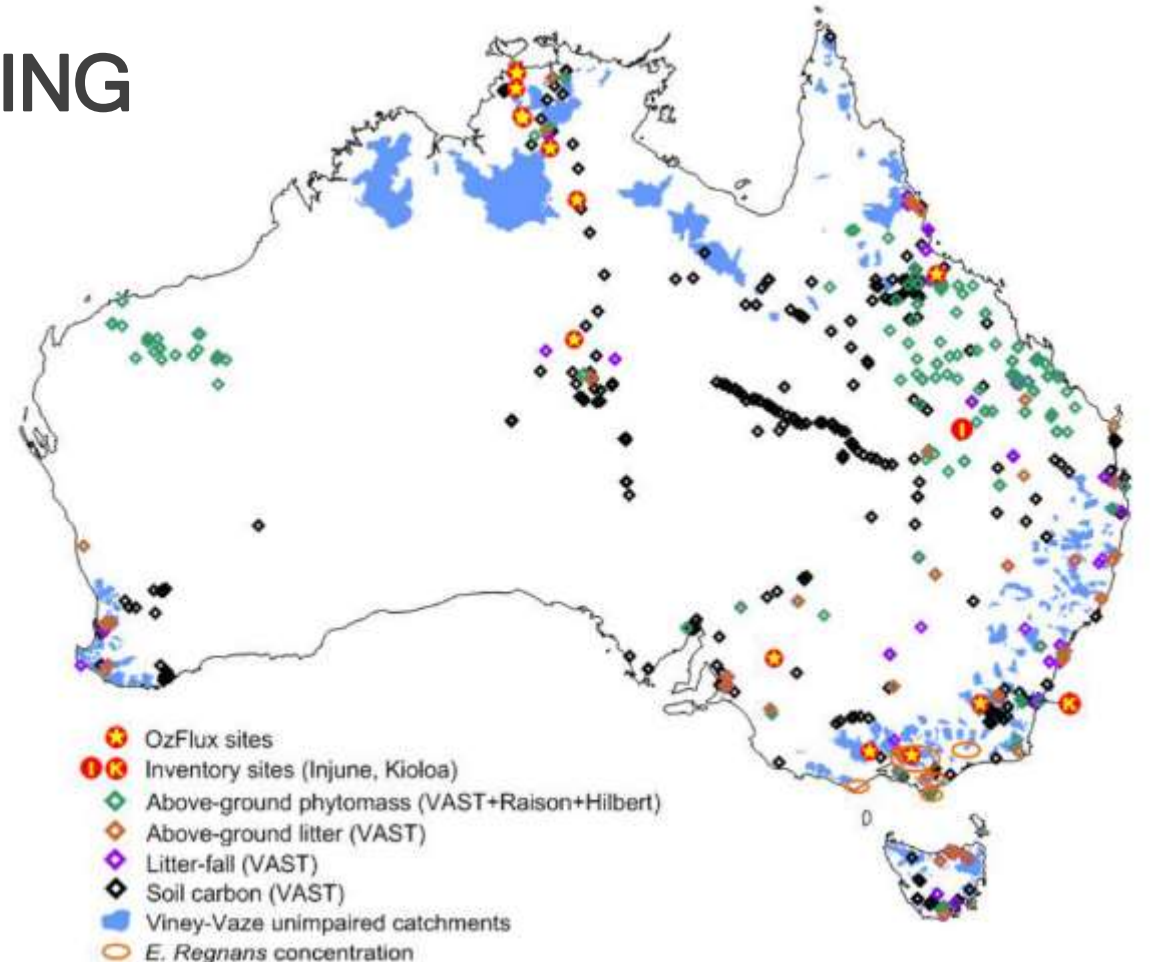
INTEGRATION WITH REMOTE SENSING

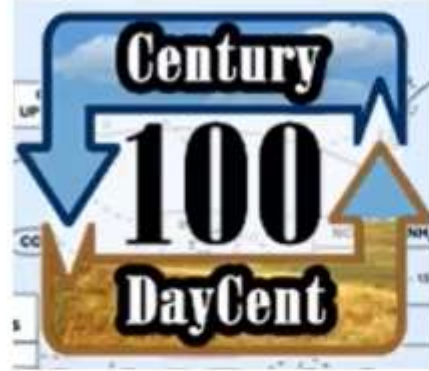
- Remote sensing does not measure quantity needed.
- Crucial role in development, calibration and validation.
- Applications from point to landscape to address research question
- Use of spectral indices of vegetation for productivity and ET.
- LiDAR for structure and biomass.
- Microwave for soil moisture, etc.



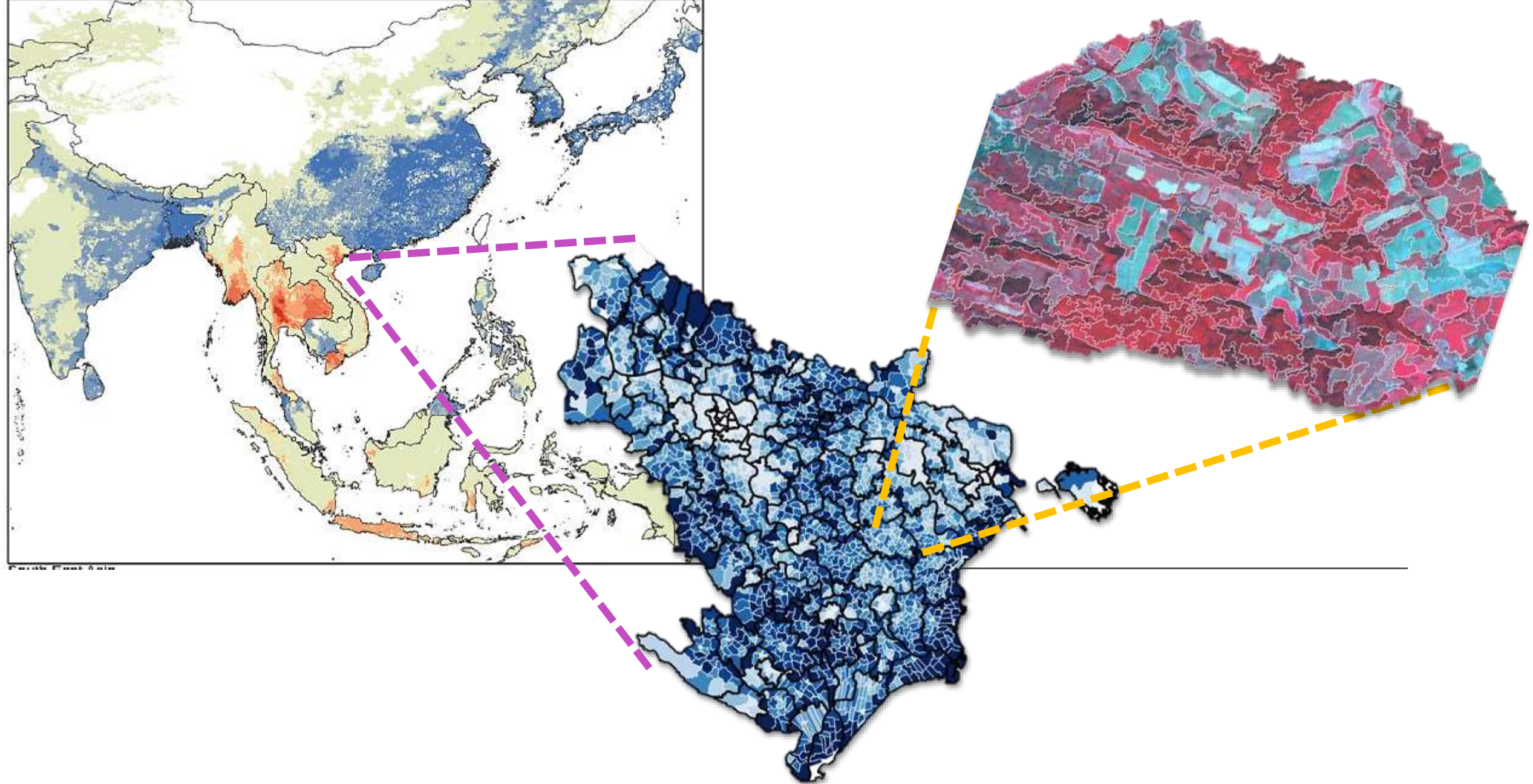
INTEGRATION WITH MODELLING

- Development, calibration and validation.
- DNDC + ORYZA
- Driven by gridded meteorology
- Strategically placing Model-data fusion to construct a consistent parameter set for model uncertainties.



The logo for DNDC (Dry Deposition and Diffusion Calculator) features the letters "DNDC" in a bold, black, sans-serif font. A blue circular arrow encircles the text, pointing clockwise.The logo for RothC is a simple black-outlined square containing the text "RothC" in a black, serif font.

- Markets & protocols are dominated by a handful of traditional models
- Next-gen models better represent modern C concepts
- Extremely high user skill requirements
- Debate over how to represent uncertainty
- Multi-model ensembles are still lacking
- Limited data for model training and validation



A main impetus for constituting the consortium is to bring the fragmented emissions research from institutional/country level to a regionally consistent, scientifically interrogated “package of practices” implemented at a regional scale



AEGIS aims to fill knowledge gaps for accurately quantifying GHG emissions when implementing some of the most promising mitigation options.



Action On Emissions of GHGs for Integrated Sustainability

China



Laos



Japan



Cambodia



Vietnam



Thailand



Indonesia

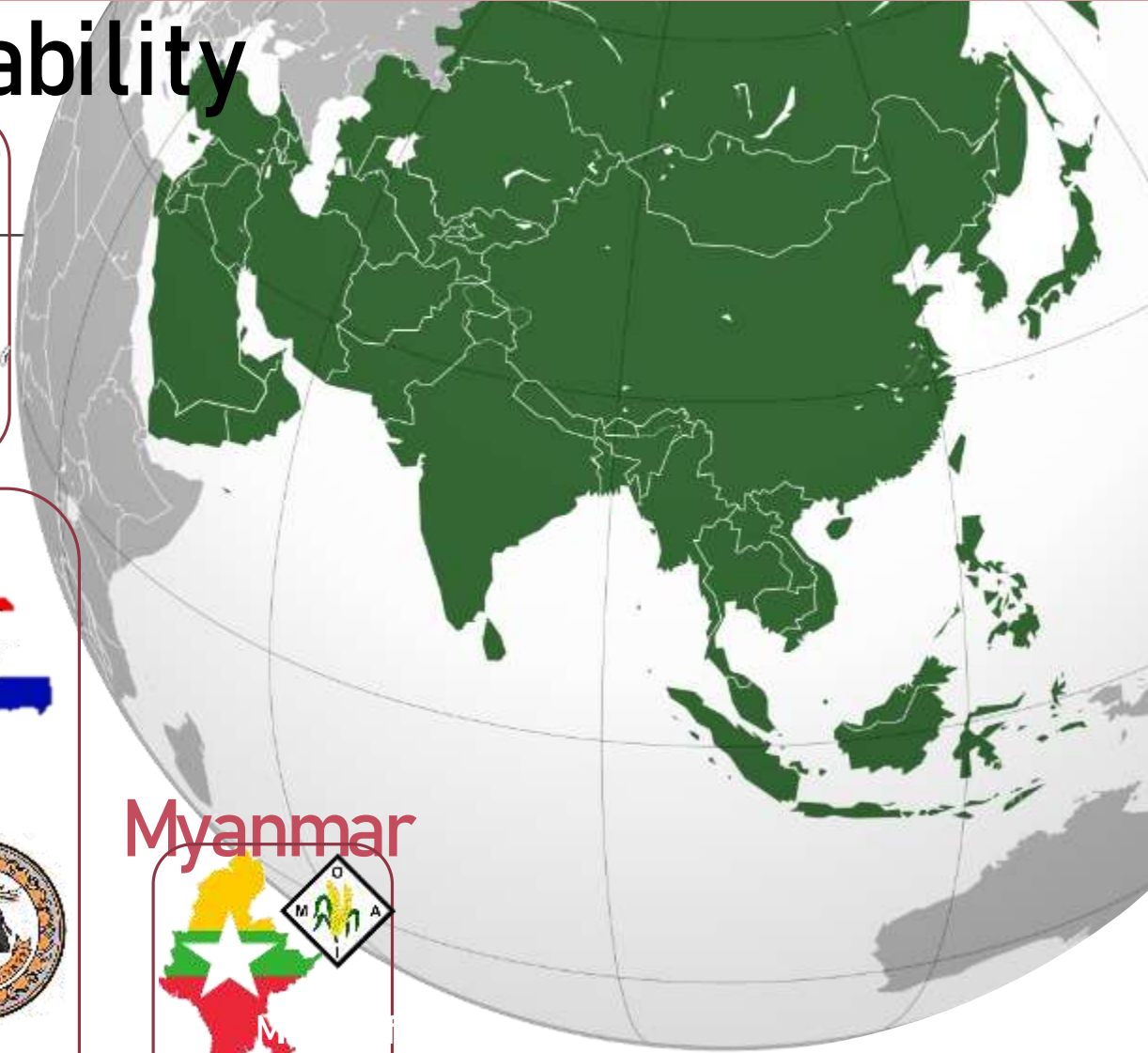


University

Myanmar



Irrigation





THE UNIVERSITY OF
**WESTERN
AUSTRALIA**



**QUEEN'S
UNIVERSITY
BELFAST**



Centre for
Ecology & Hydrology
NATURAL ENVIRONMENT RESEARCH COUNCIL



Food and Agriculture
Organization of the
United Nations

Bringing together key Strategic Research, Policy and End User Partners to define the barriers to progress in each area, integrate these policy/end user/science challenges into a coherent cross-discipline framework

