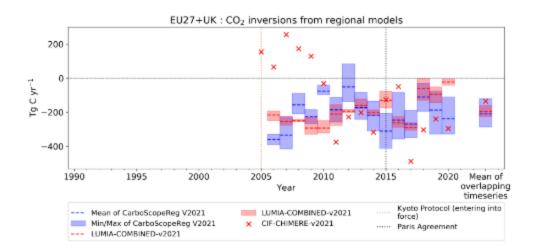


## T3.3 Other CO2 regional inversions!

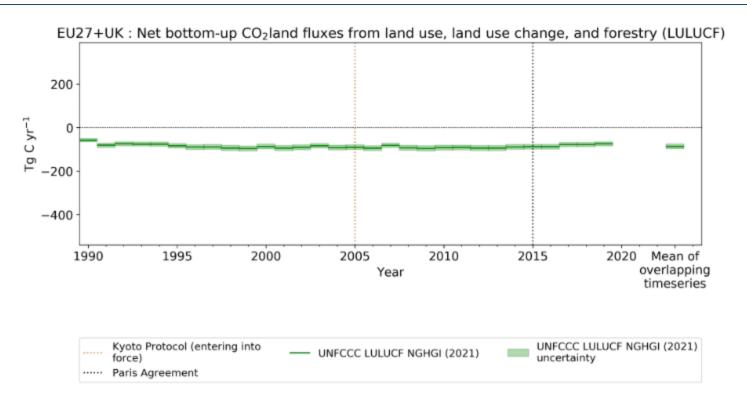
- Community Inversion Framework CO2 inter-comparison
  - CIF CHIMERE is working : first results obtained
  - CIF LMDZ (globa) completed
  - CIF STILT / Flexpart underway
- As well as LUMIA inversion system
  - ⇒ Differences that need to be resolved!
- Synthesis in WP5 still use the different estimates
  (See summary talk (P. Peylin) for a few comparisons)





## High uncertainty in LULUCF fluxes

#### **Inventories**

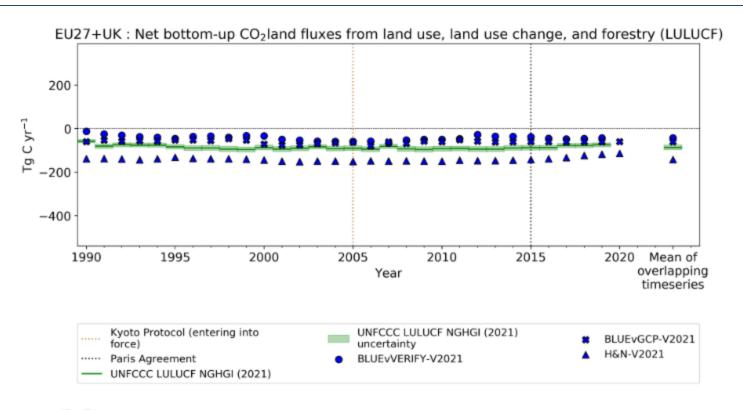




## High uncertainty in LULUCF fluxes

#### **Inventories**

#### Bookkeeping



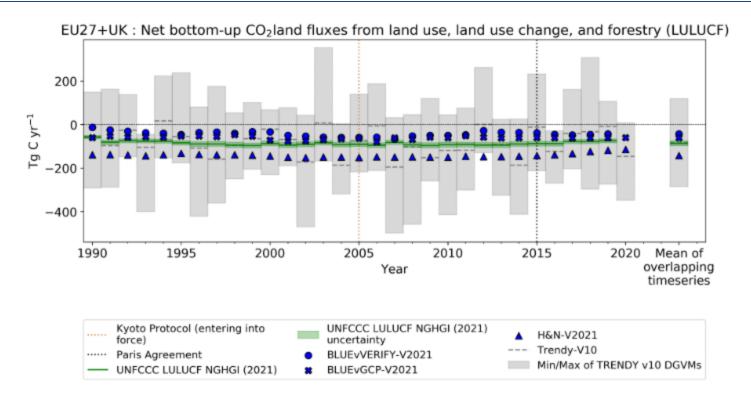


## High uncertainty in LULUCF fluxes

**Inventories** 

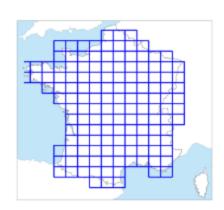
Bookkeeping

**DGVMs** 



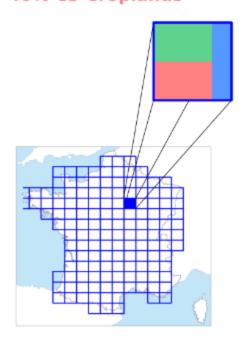


# Dynamic global vegetation models (TRENDY)





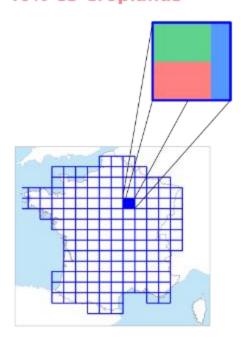
# Dynamic global vegetation models (TRENDY)

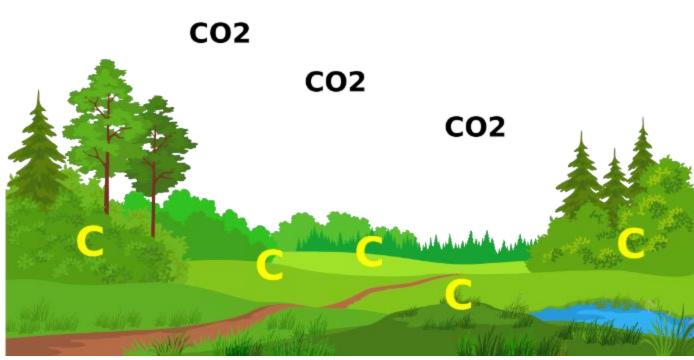




# Dynamic global vegetation models (TRENDY)

# Net biome production (NBP)



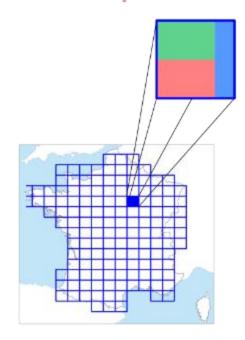


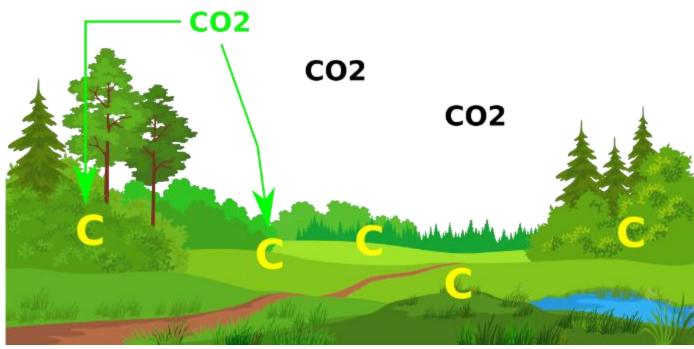


# Dynamic global vegetation models (TRENDY)

# Net biome production (NBP)

NBP = Photo



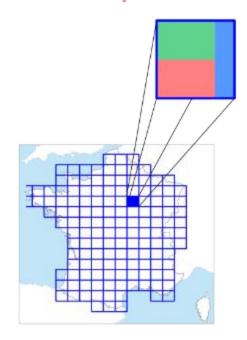


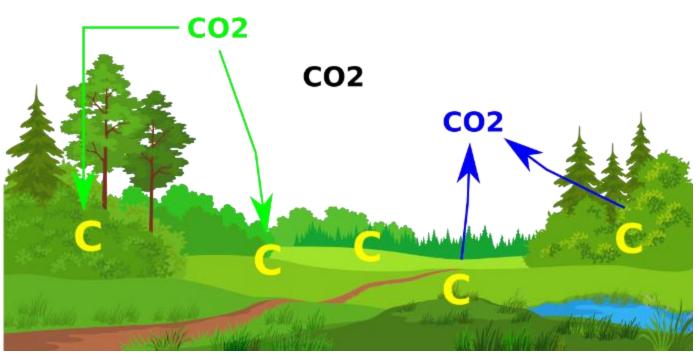


# Dynamic global vegetation models (TRENDY)

# Net biome production (NBP)

NBP = Photo - Resp



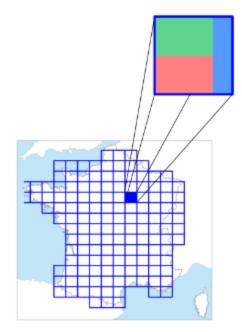


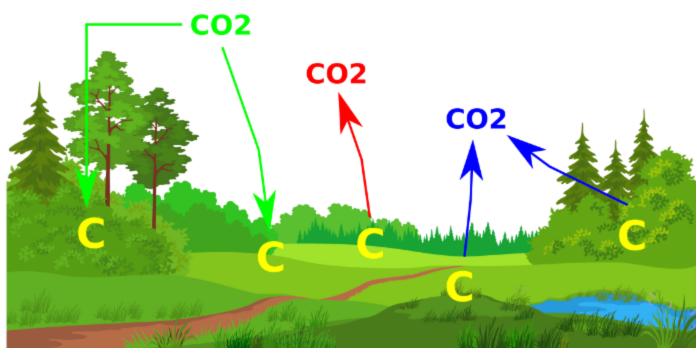


Dynamic global vegetation models (TRENDY)

# Net biome production (NBP)

NBP = Photo - Resp - Harv







# Dynamic global vegetation models (TRENDY)

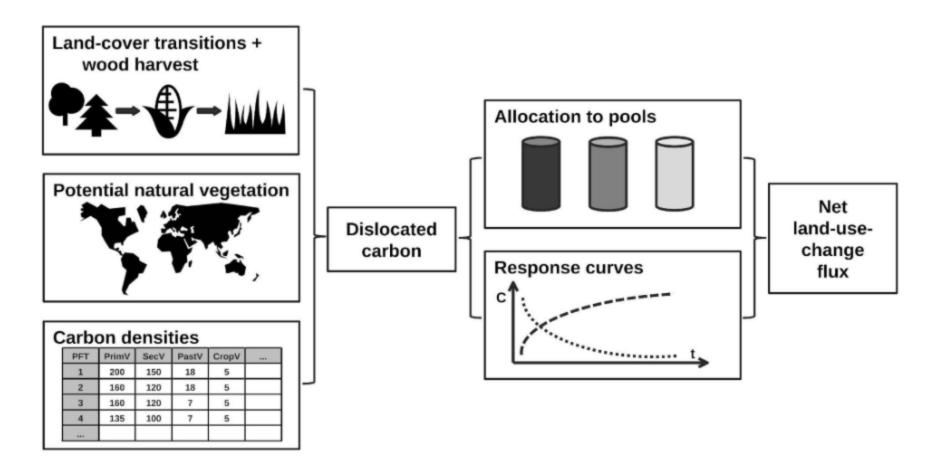
#### Factorial experiments

	S0	S1	S2	S3	S4 (if app.)
CO2	static	dynamic	dynamic	dynamic	dynamic
Climate	static	static	dynamic	dynamic	dynamic
Land cover/land use	static	static	static	dynamic	dynamic
Nitrogen deposition (if applicable)	static	static	static	dynamic	static

Difference between simulations gives the effect of that driver (e.g., S2-S1 gives the climate influence)

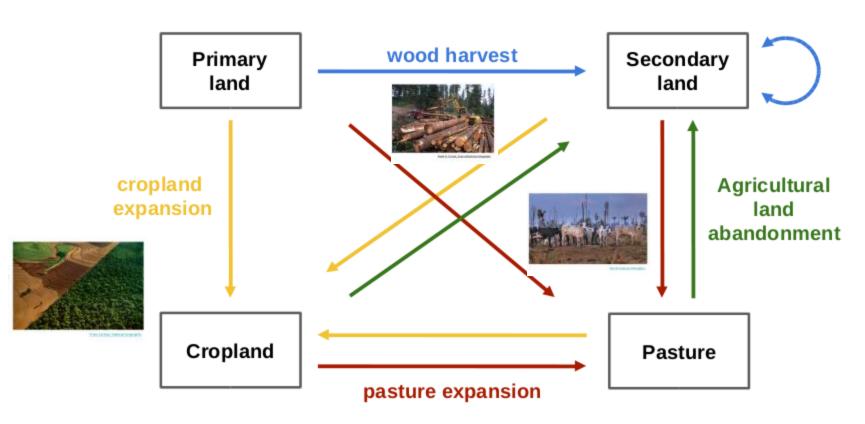


## Bookkeeping models (BLUE)





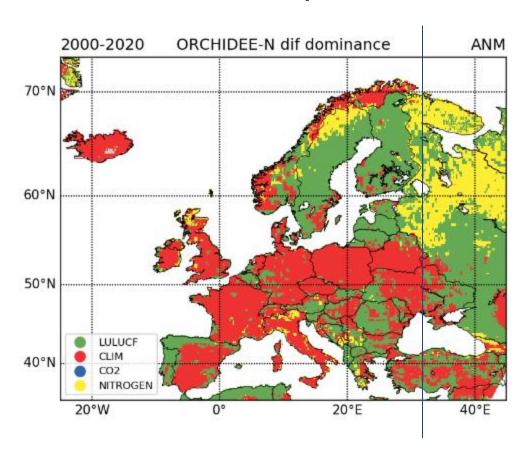
# Bookkeeping models (BLUE)



Adapted from Kerstin Hartung

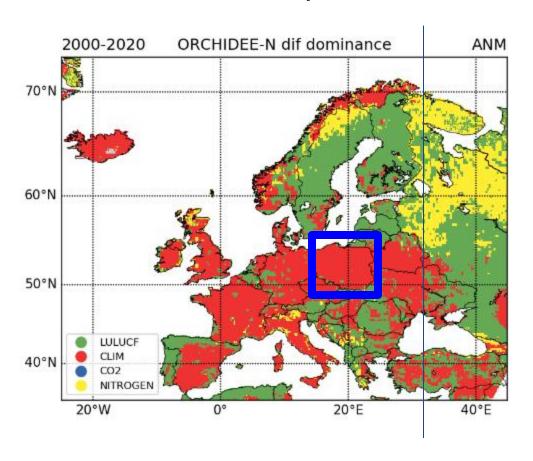


# Dynamic global vegetation models (TRENDY)





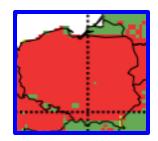
# Dynamic global vegetation models (TRENDY)

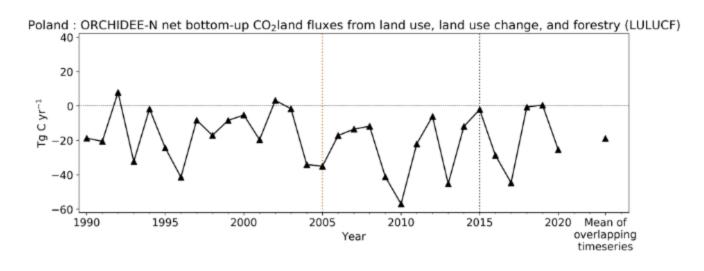




# Dynamic global vegetation models (TRENDY)

## Factorial experiments

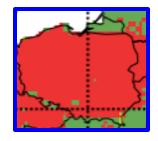


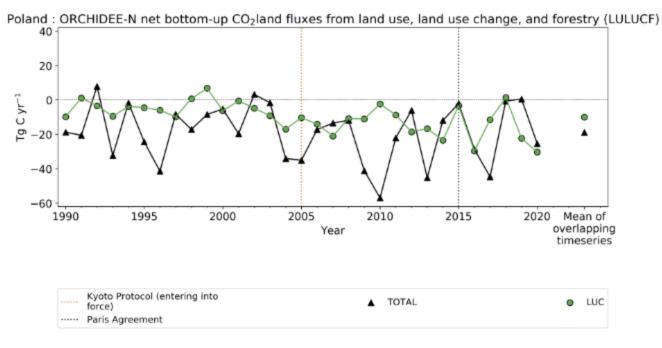






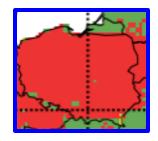
# Dynamic global vegetation models (TRENDY)

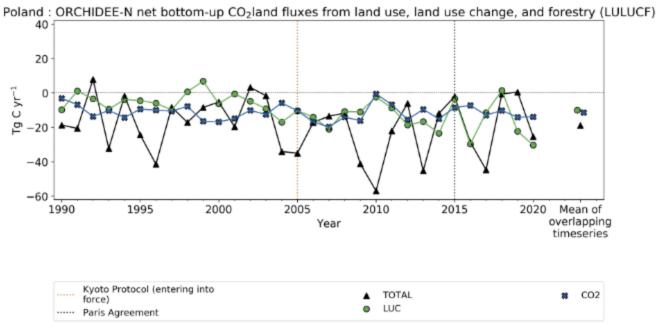






# Dynamic global vegetation models (TRENDY)

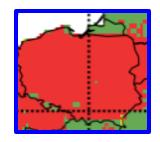


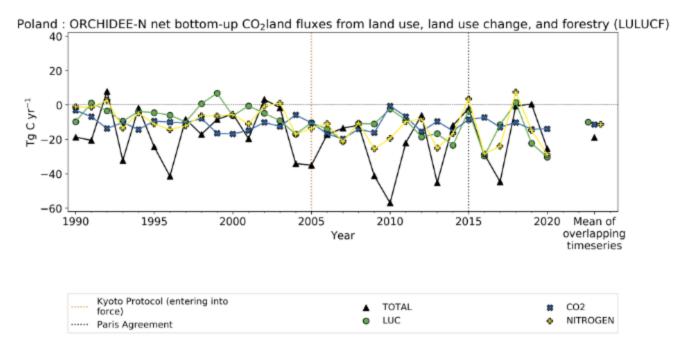






# Dynamic global vegetation models (TRENDY)



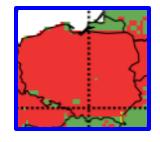


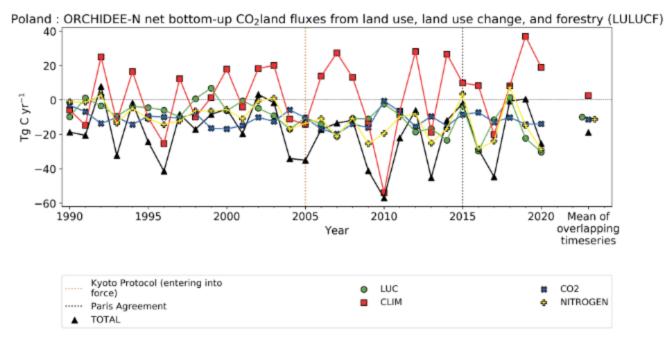




# Dynamic global vegetation models (TRENDY)

#### Factorial experiments



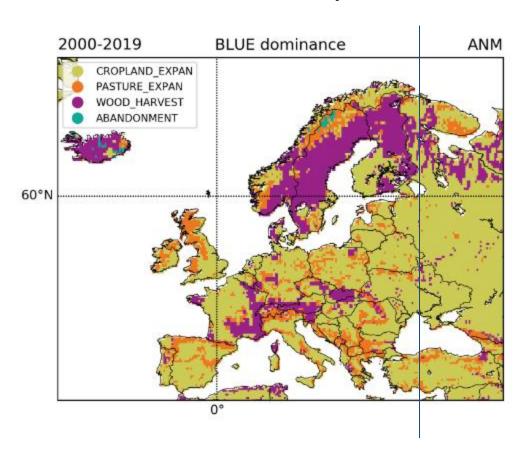


cc VERIFY Project



# Bookkeeping models (BLUE)

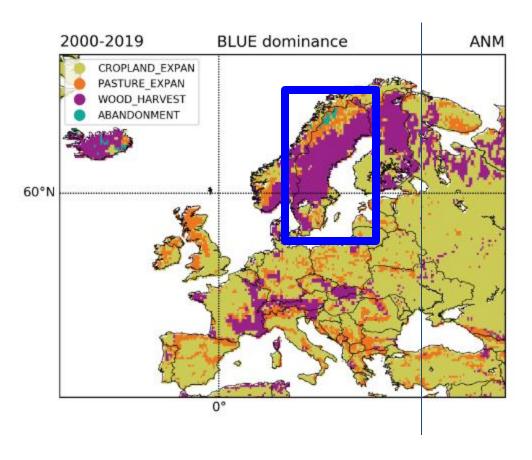
## Emission components





# Bookkeeping models (BLUE)

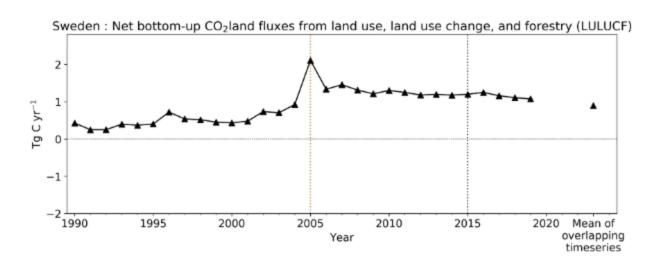
## Emission components

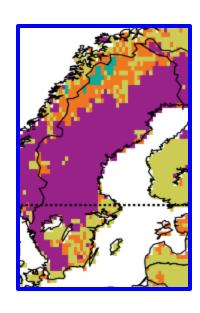




# Bookkeeping models (BLUE)

## Emission components



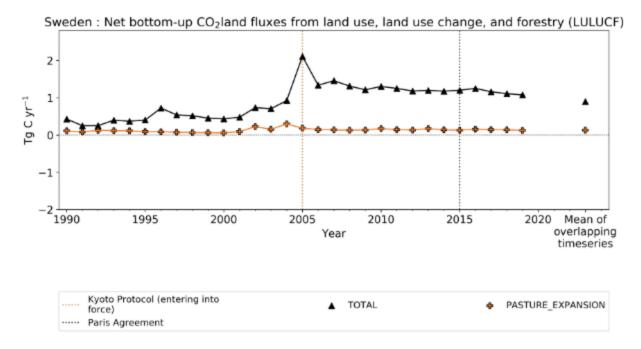


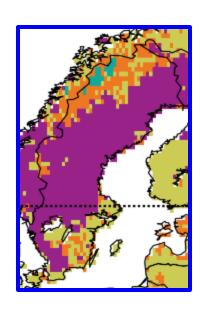
Kyoto Protocol (entering into ...... Paris Agreement ▲ TOTAL force)



# Bookkeeping models (BLUE)

## Emission components

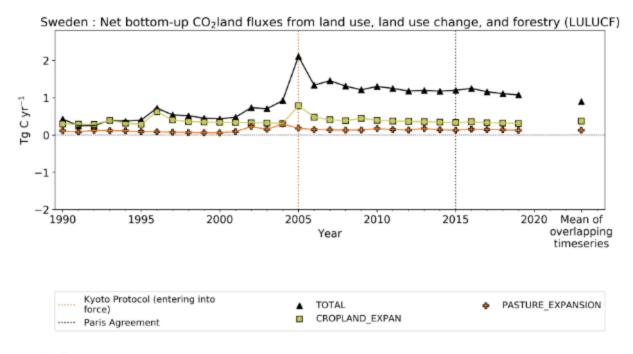


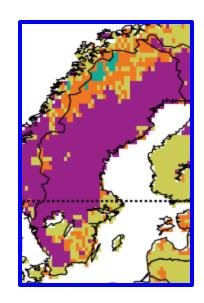




# Bookkeeping models (BLUE)

## Emission components

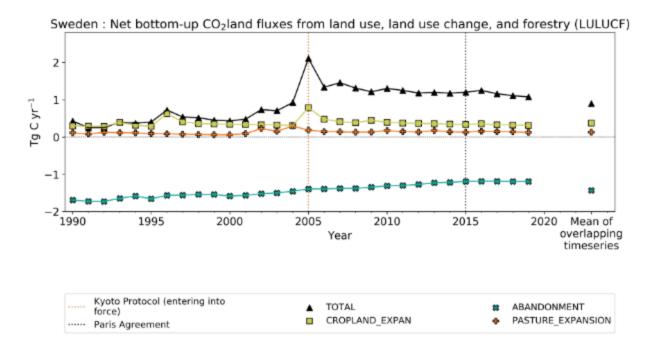


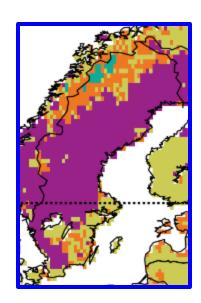




# Bookkeeping models (BLUE)

## Emission components

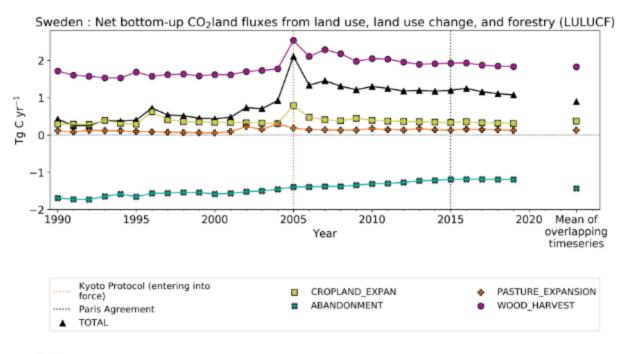


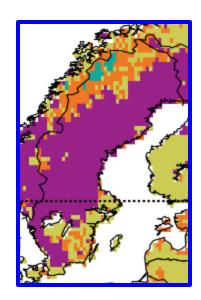




# Bookkeeping models (BLUE)

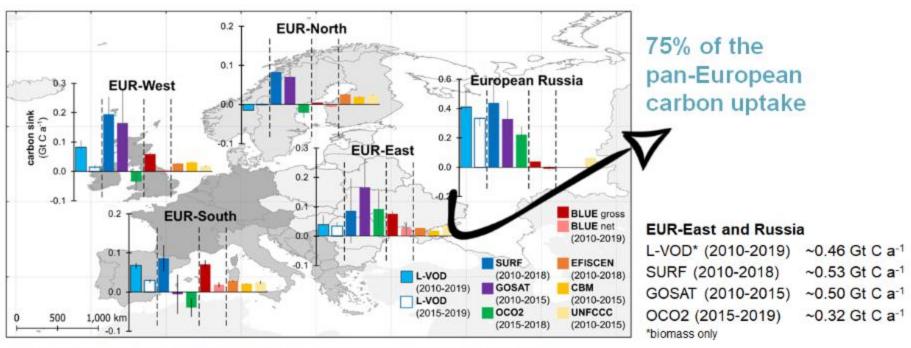
## **Emission components**







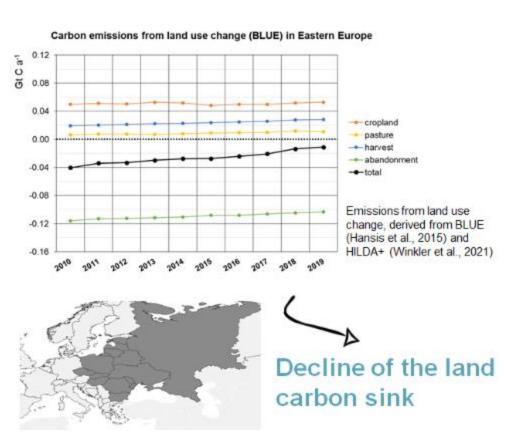
# 1. How large is the EE carbon sink?

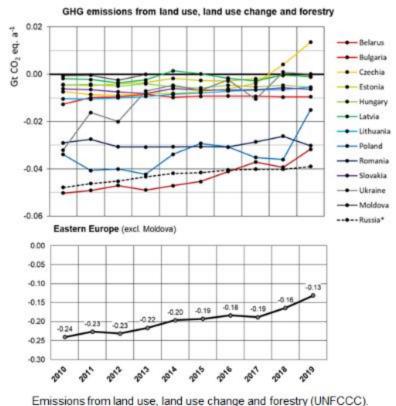


Carbon uptake from L-VOD AGB change, satellite inversions (SURF, GOSAT, OCO2), land use emission models and inventories (BLUE, EFISCEN, CBM, UNFCCC) across Europe between 2010 and 2019.



# 2. What are the temporal patterns?

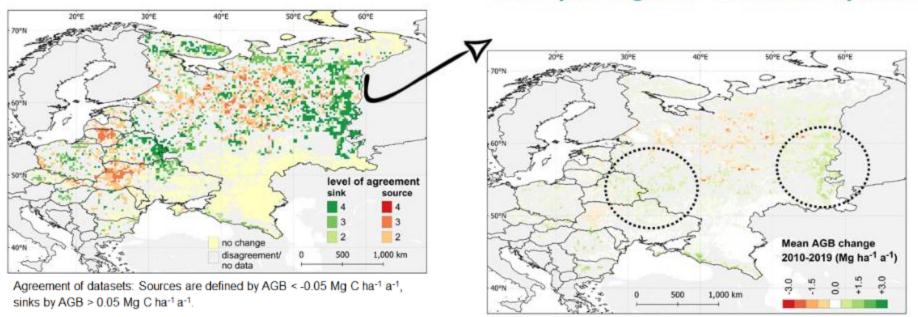






# 3. What are the spatial patterns?

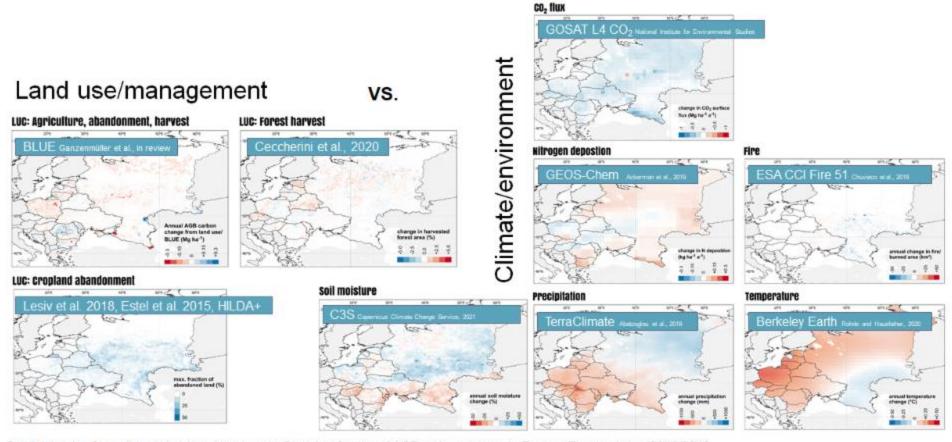
#### Hot spot regions of carbon uptake



Carbon gains and losses in AGB (in Mg C ha<sup>-1</sup> a<sup>-1</sup>) in Eastern Europe during 2010-2019 from a) TRENDY S3 (mean of 15 models), b) JPL: Xu et al. 2021, c) L-VOD and d) bias-adjusted WRI: Harris et al. 2021.



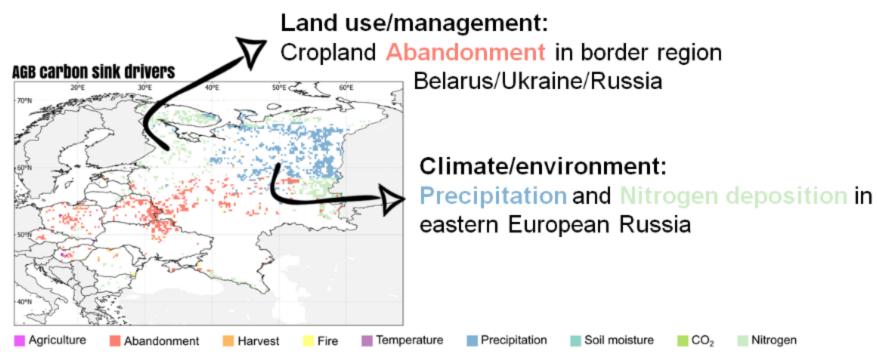
#### 4. What are the drivers?



Spatial trends of possible underlying drivers and influencing factors of AGB carbon change in Eastern Europe during 2010-2019.



#### 4. What are the drivers?

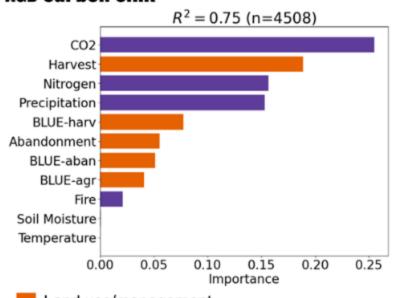


Spatial patterns of drivers of AGB carbon change in Eastern Europe during 2010-2019.



#### 4. What are the drivers?

#### **AGB carbon sink**



Driver analysis based on Random Forest regression

- Climate factors, CO2 fertilization and N deposition as most important drivers
- Land use/management, wood harvest as cause of the decreasing land carbon sink?

Land use/management

Environment

Relative feature importance of driver variables for AGB change across Eastern Europe (Random Forest regression)



# Take-away

- □ Eastern Europe: ~0.46 Gt C a<sup>-1</sup> → 75% of the European carbon uptake
- □ Decline of the carbon sink 2010-2019
- □ Hot spot regions Russian Ural region and ~border Belarus/Ukraine/Russia

#### **Driver analysis**

- □ Land use change (abandonment) in hot spot region
- □ Climate change (CO₂ fertilization, N deposition, precipitation) in European Russia
- Forest management (harvest) as reason for the declining sink?



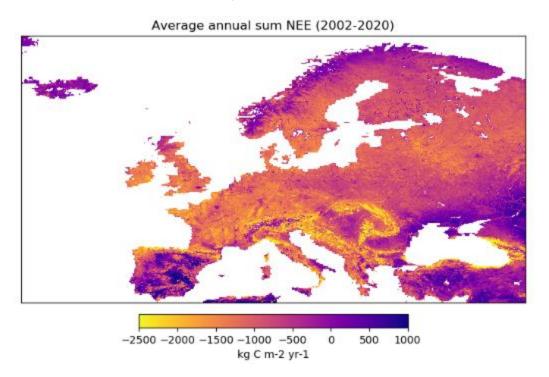
Viadmir Evgenissich Pemphilav/ https://pittlibguides.com/c.php?g=494797&p=3385703



#### WP3: T3.4 - ADDITIONAL PRODUCTS

## ⇒ More data were produced within WP3

• FLUXCOM (data-driven flux estimates): 3-hr;  $\sim 5 km$ 



- Not used yet in the CO2 synthesis because of too large NEE!
- But recent progresses reveal a bias partly linked to shortwave radiation forcing!

 Carbon Cycle Data Assimilation Systems (Fusion of process-based models and observations)



#### T3.4 Carbon Cycle Data Assimilation Systems

#### **BETHY-CCDAS**

#### **ORCHIDEE-CCDAS**

#### **CTDAS**

#### Assimilation of

- Satellite soil moisture
- Satellite L-VOD
- In situ atm. CO2 data
- Nighlights (for FF model)

#### Two-step assimilation

- Fossil Fuel at 0.1 in FFDAS
- Biogenic fluxes at 0.25 in CCDAS

#### Optimisation of

around 70 parameters in CCDAS and 1.5M in FFDAS

#### Assimilation of

- FluxNet in situ data
- Satellite Fluorescence
- In situ atm. CO2 data (using LMDZ model)

#### Optimisation of

- around 150 parameters
- stepwise assimilation of each datastream

#### Dynamic a-priori fluxes

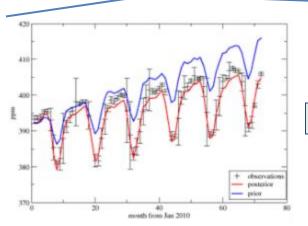
Near real-time statistical data (EU, Eurostat,...) and weather data to fill in short-term variations in fluxes at ~2km

#### **Assimilation of**

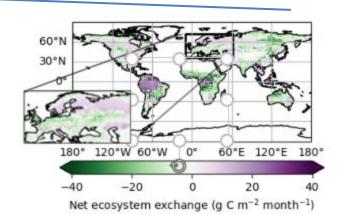
- ICOS CO2 mole fractions
- Δ14C, CO, NO2

#### Optimisation of

~10 params per country, sectorial FF emissions, NEE / PFT





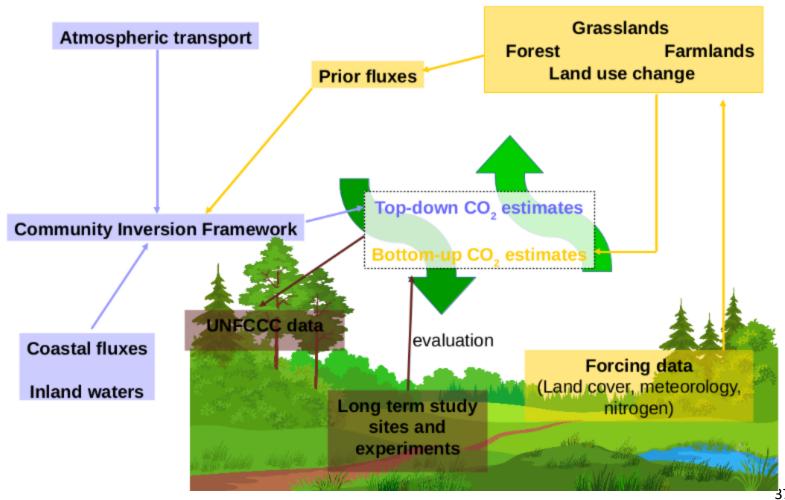


⇒ see **D2.13** 

# WP3 "CO2 land Pre-operational system"

Pre-operational system for CO<sub>2</sub> fluxes from terrestrial ecosystems

VERIFY





# WP3 – STATUS OF LAST DELIVERABLES

D3.7	Attribution analysis	CEA	Draft on-going
D3.14	National forest inventory and high resolution forest cover for eastern Europe	WEnr	Final version ready (proof reading phase)
D3.16	Analysis of NEE and carbon balance of Eastern Europe	CEA	Draft done based on a paper
D3.17	Final version of pre-operational system for terrestrial CO2 verification	UNIABD	Draft on-going



# Thank you for your attention.















































































This project has received funding from the European Union's Horizon 2020 research and innovation programme under grant agreement No 776810



#### WP3 PARTNERS

#### Two co – leaders:

University of Aberdeen (UNIABDN), Commissariat à l'énergie atomique (CEA),

#### 18 partners:

Joint Research Centre (JRC), Karlsruhe Institute of Technology (KIT), Max Planck Society for the Advancement of Science (MPG), Norwegian Institute for Air Research (NILU), International Institute for Applied Systems Analysis (IIASA), Integrated Carbon Observation System (ICOS) ERIC, European Centre for Medium-Range Weather Forecasts (ECMWF), Wageningen Environmental Research (WEnR), University of Bremen (UBremen), University of East Anglia (UEA), Vrije Universiteit Amsterdam (VU), Lund University (ULUND), University of Bergen (UiB), Wageningen University & Research (WUR), Université libre de Bruxelles (ULB), Saint Petersburg State University (SPbU), Ural Federal University (**UrFU**), Ludwig-Maximilian Munich University (**LMU**)



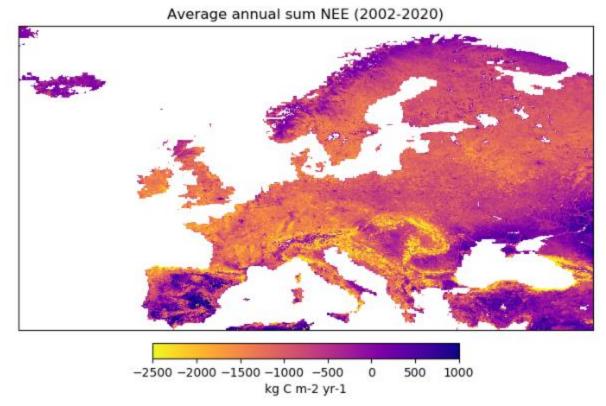
# Additional slides



## WP3 - BOTTOM-UP DATA-DRIVEN BIOGENIC

#### **FLUXES**

# New generation Fluxcom: upscaling of eddy-covariance measurements using Earth observation and machine learning



European flux estimates every 3 hours, 0.05deg

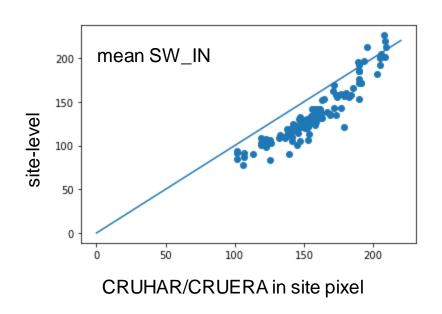
VERIFY GA - WP3 | May 9-11<sup>th</sup> , 2022

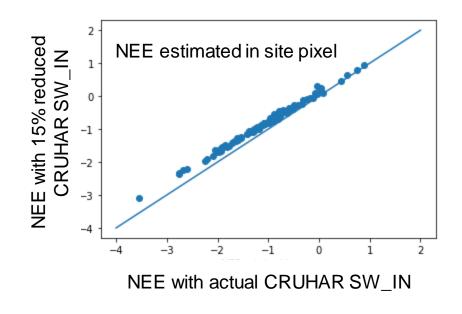


## WP3 - BOTTOM-UP DATA-DRIVEN BIOGENIC

#### **FLUXES**

The importance of meteorological fields: very strong sink in Fluxcom NEE partly related to bias in shortwave incoming radiation:





15% average bias between site-level and gridded incoming radiation

21% biased NEE on average



#### WP3 - COASTAL AIR-SEA CO2 FLUXES

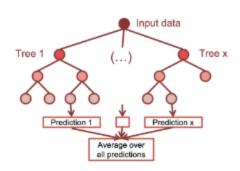




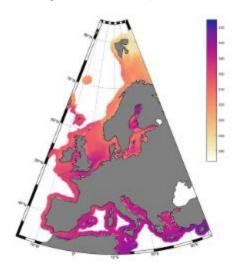
Meike Becker, Are Olsen, University of Bergen

Annual updated coastal CO<sub>2</sub> flux maps on the European shelf (1998-2020):

Fitting a set of driver data (e.g. temperature, chlorophyll concentration, mixed layer depth, ice concentration) against surface ocean  $pCO_2$  observations



pCO<sub>2</sub> / µatm (average 1998-2020)



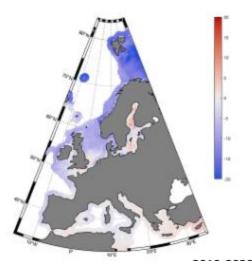
#### Sinks for CO2:

- Barents Sea
- Nordic Seas
- Northern and Central North Sea
- > Atlantic shelf

#### Sources for CO2:

- Southern North Sea and English Channel
- ➤ Baltic Sea
- > Mediterranean

Air-Sea CO<sub>2</sub> Flux / gC m<sup>-2</sup> yr<sup>-1</sup>



2010-2020