

# Task 6.5

Multi-disciplinary data assimilation to  
advance process understanding in  
Arctic greenhouse gas exchange

MPG, UiB, NIVA, AU, UB

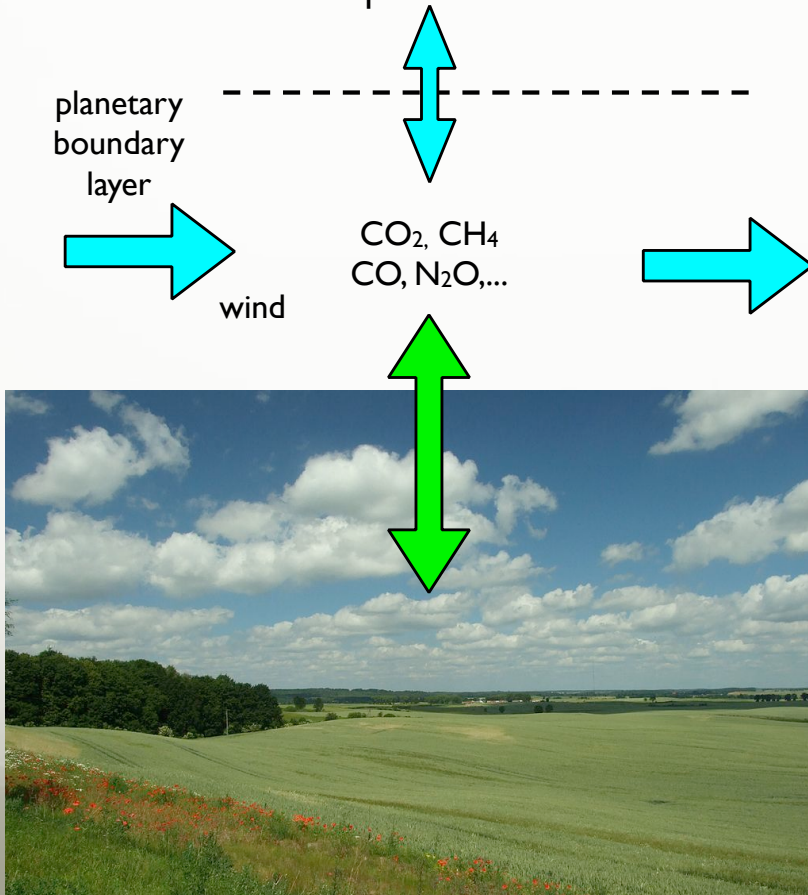
# Overall objective

## **Multi-disciplinary data assimilation that improves quantifying regional Arctic carbon budgets**

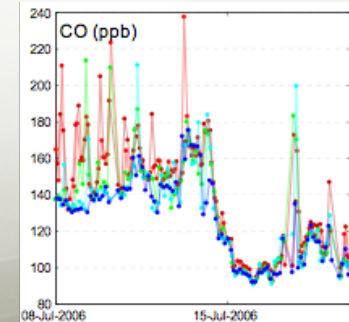
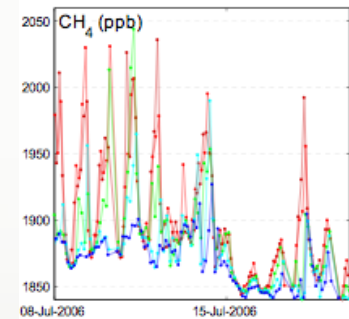
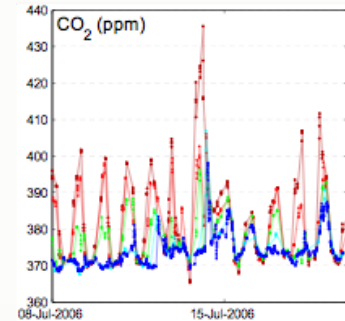
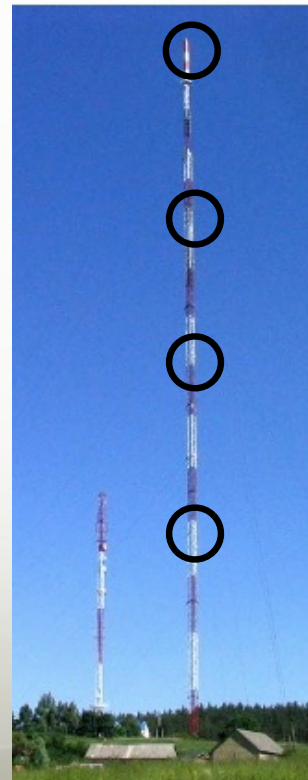
- (Atmospheric) inverse modeling of greenhouse gases, including a geostatistical data assimilation component
- Inter-disciplinary GHG assessment in the North Atlantic ocean region

# Atmospheric inversion: atmosphere as 'integrator'

atmospheric transport constitutes a  
natural integrator of surface-  
atmosphere fluxes

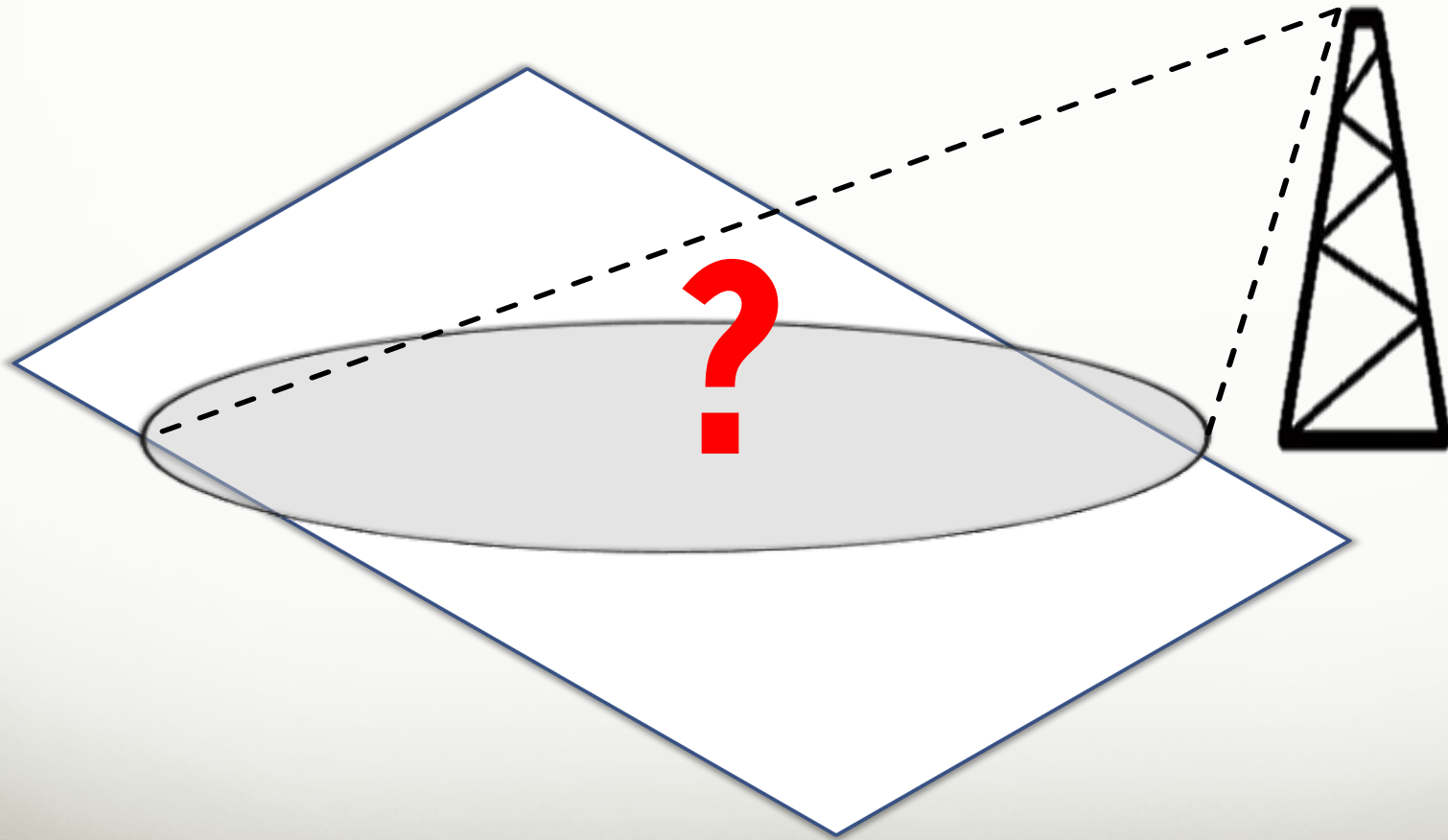


Bialystok tower  
(Eastern Poland)  
Trace gas  
concentration  
measurements

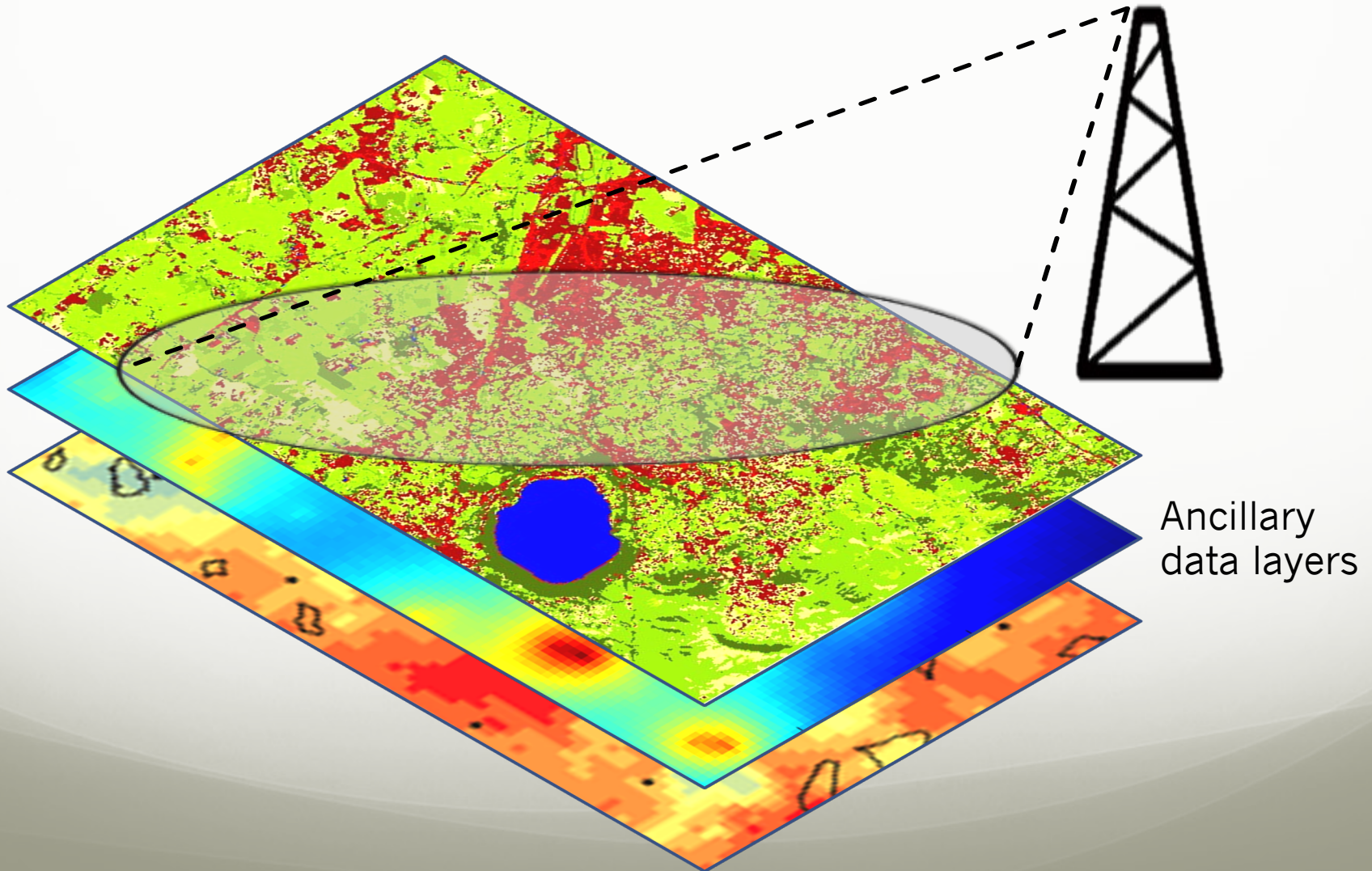


source: Popa et al., 2010

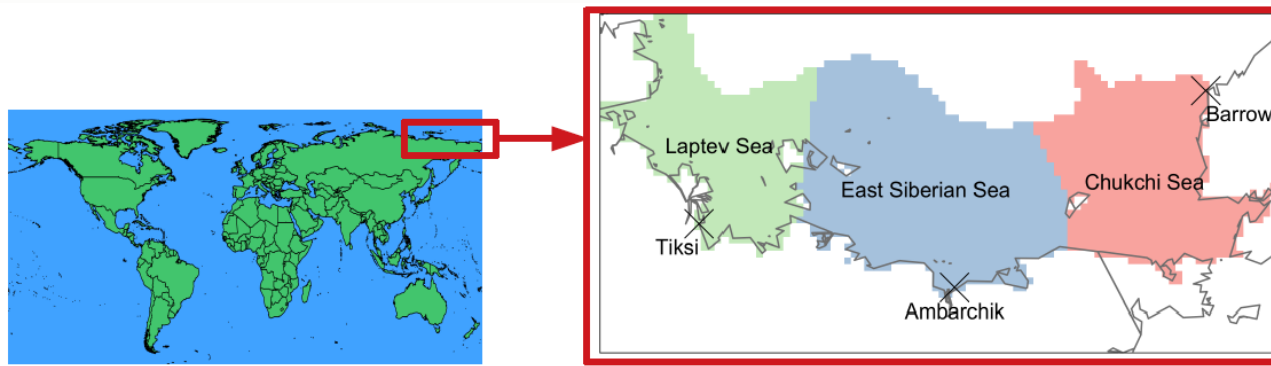
# Under-constrained problem



# Add information for structuring solution space



# Pilot study: East Siberian Arctic Shelf



East Siberian Arctic Shelf

Candidate parameters

considered for prior ocean CH<sub>4</sub> flux:

Category	variable	Process described
Bathymetry	ocean depth	Ebullition
	shelf edge	
Meteorology	wind speed <sup>2</sup>	Turbulent diffusion
Sea ice	ice-free ocean fraction	Ice barrier
	sea ice growth	Brine rejection
	sea ice melt	Accumulation below ice

# Geostatistical CH<sub>4</sub> inversion for ESAS: major findings

- Total emissions rather low: 0.4 – 1.5 Tg CH<sub>4</sub> yr<sup>-1</sup>
- Emissions mostly from shallow waters
  - Emissions distribution, or mixing of CH<sub>4</sub>
- No emission spike during ice melt
  - Biological consumption, ocean currents, etc.
- Large emissions in fall/winter possible
  - Mixing due to sea ice growth, and/or storms?
- Ancillary data layers helped structuring results, link emission patterns to specific processes

# Extension: embed new INTAROS data

e.g. seamless data set of atmospheric total water vapor (TWV) over Arctic ocean and sea ice (UB)

## Based on

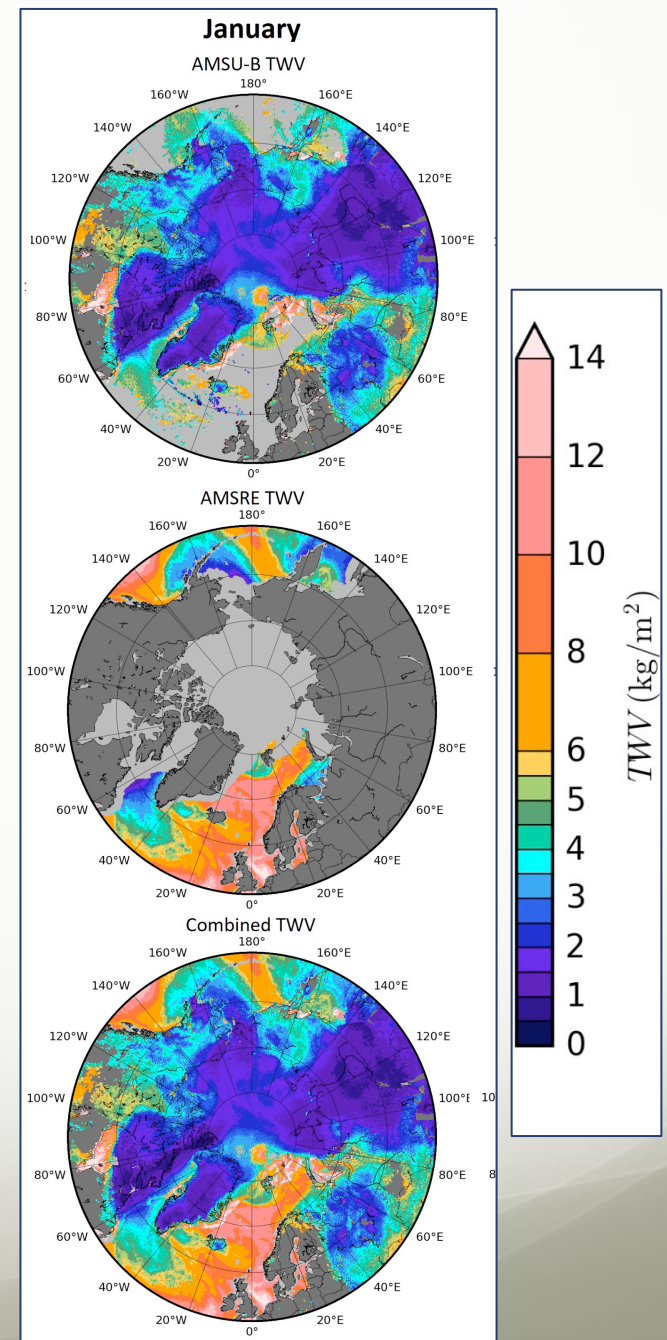
- *AMSU-B/MHS microwave humidity sounders.* (Triana Gomez et al., AMT (2020) )
- *AMSR-E/AMSR-2 microwave imagers.* (Wentz and Meissner (2002) global ocean TWV product)

## Merged product

- Data years 2008-2009 produced, daily NETcdf maps

## Challenges

- *with sounders:*
  - Screen out opaque ice clouds values by image processing
  - Merging of results from three sub-algorithms, with different TWV range and different channel combinations
- *Merging sounder and imager retrieval into TWV fields with homogeneous statistics & no jumps along merging lines.*

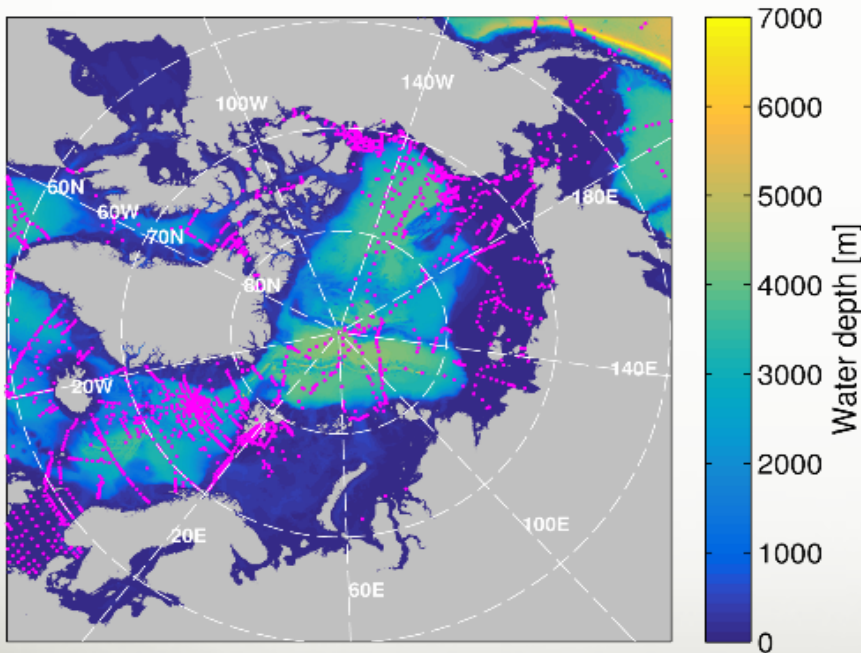




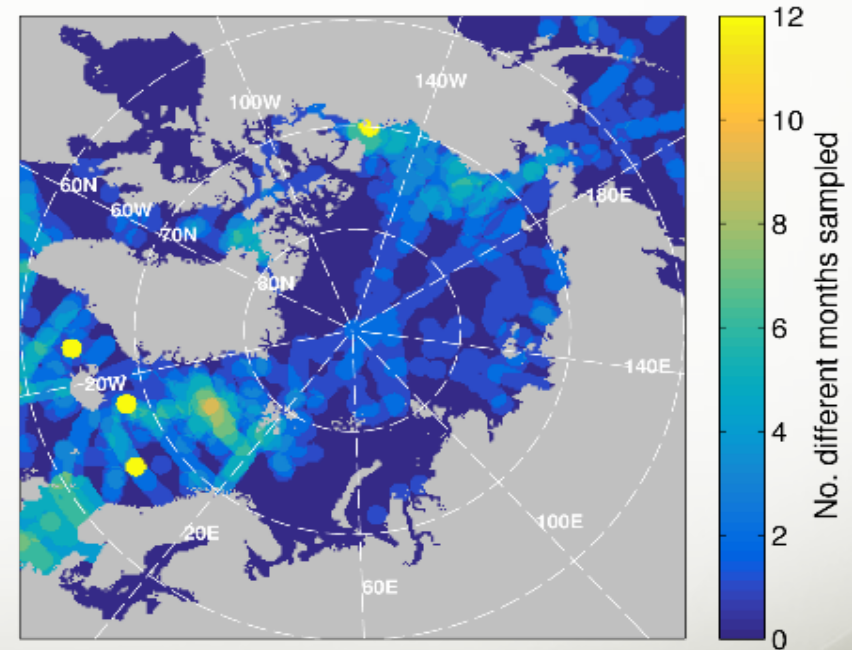
# Data aggregated in Task 2.2

Task 2.2: “Exploitation of existing in situ datasets for Arctic seawater carbonate system chemistry, nutrients, and phytoplankton biomass” (Aggregation and QC/QA of existing GLODAPv2 dataset together with CARINA, WOD, and ICES dataset into “GLODAPv2E” INTAROS dataset)

a) pH GLODAPv2E xy coverage

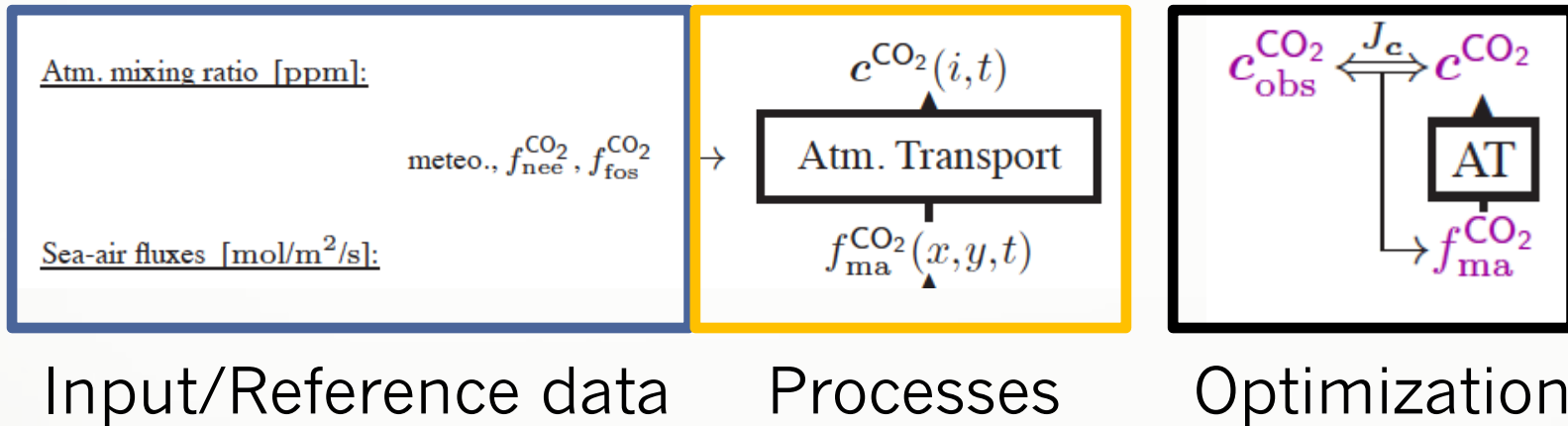


b) pH GLODAPv2E seasonal coverage

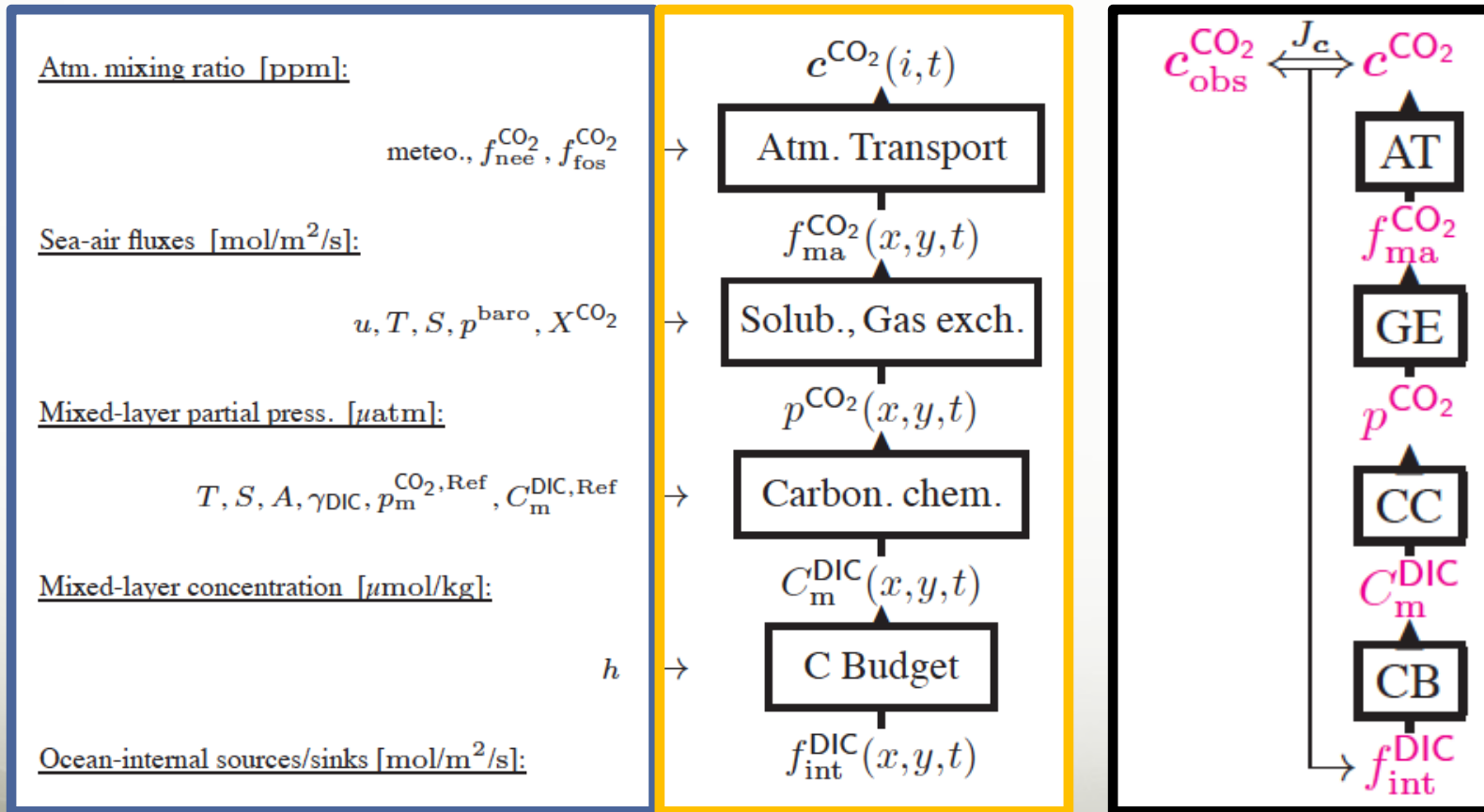


- Relatively good carbonate chemistry coverage in Norwegian Sea and central Arctic
- Poor carbonate chemistry coverage in Barents Sea/Kara Sea
- Relatively poor seasonal coverage throughout the Arctic

# Standard atmospheric inversion



# Generalized data assimilation framework

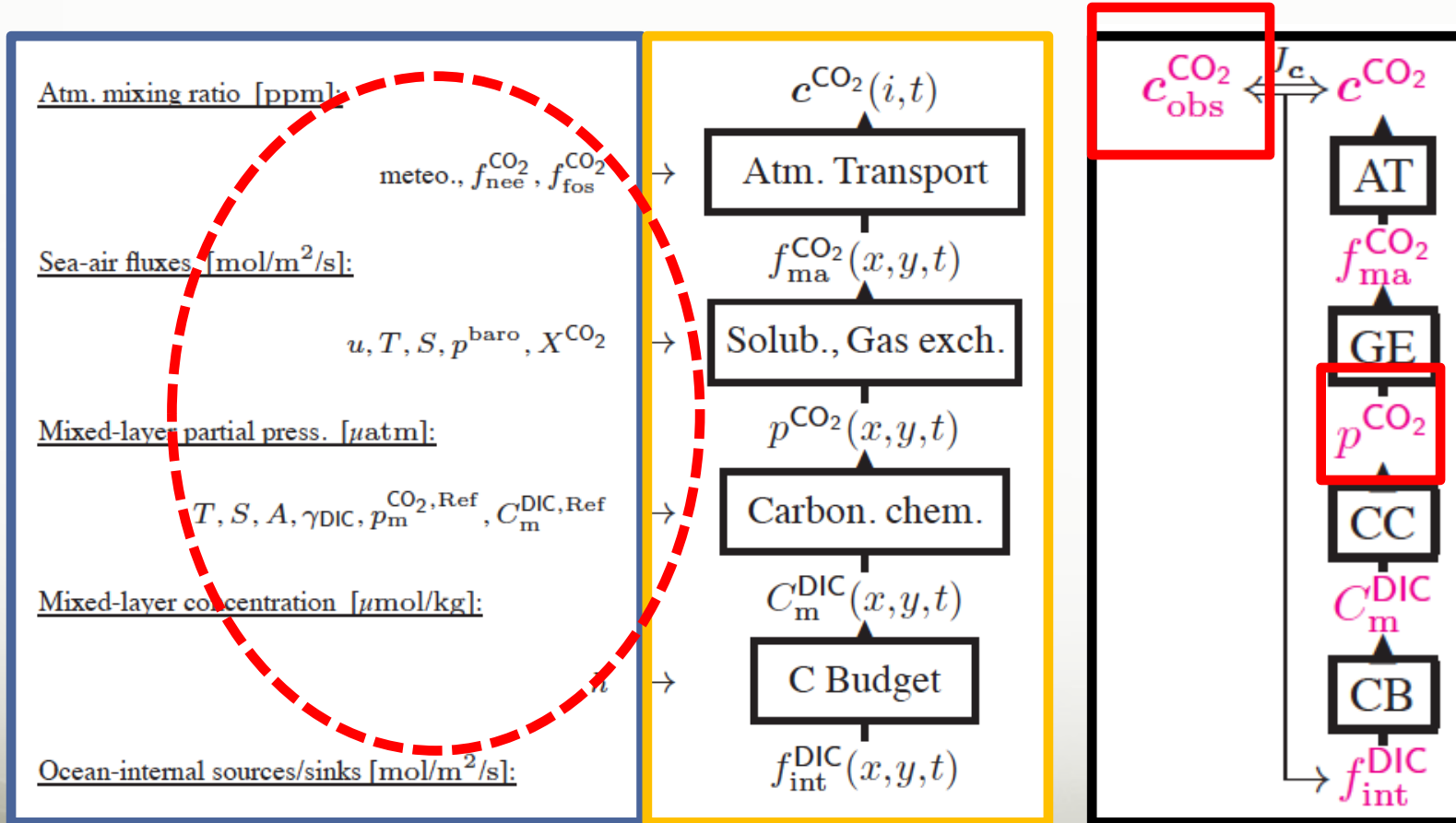


Input/Reference data

Processes

Optimization

# Multi-disciplinary data assimilation



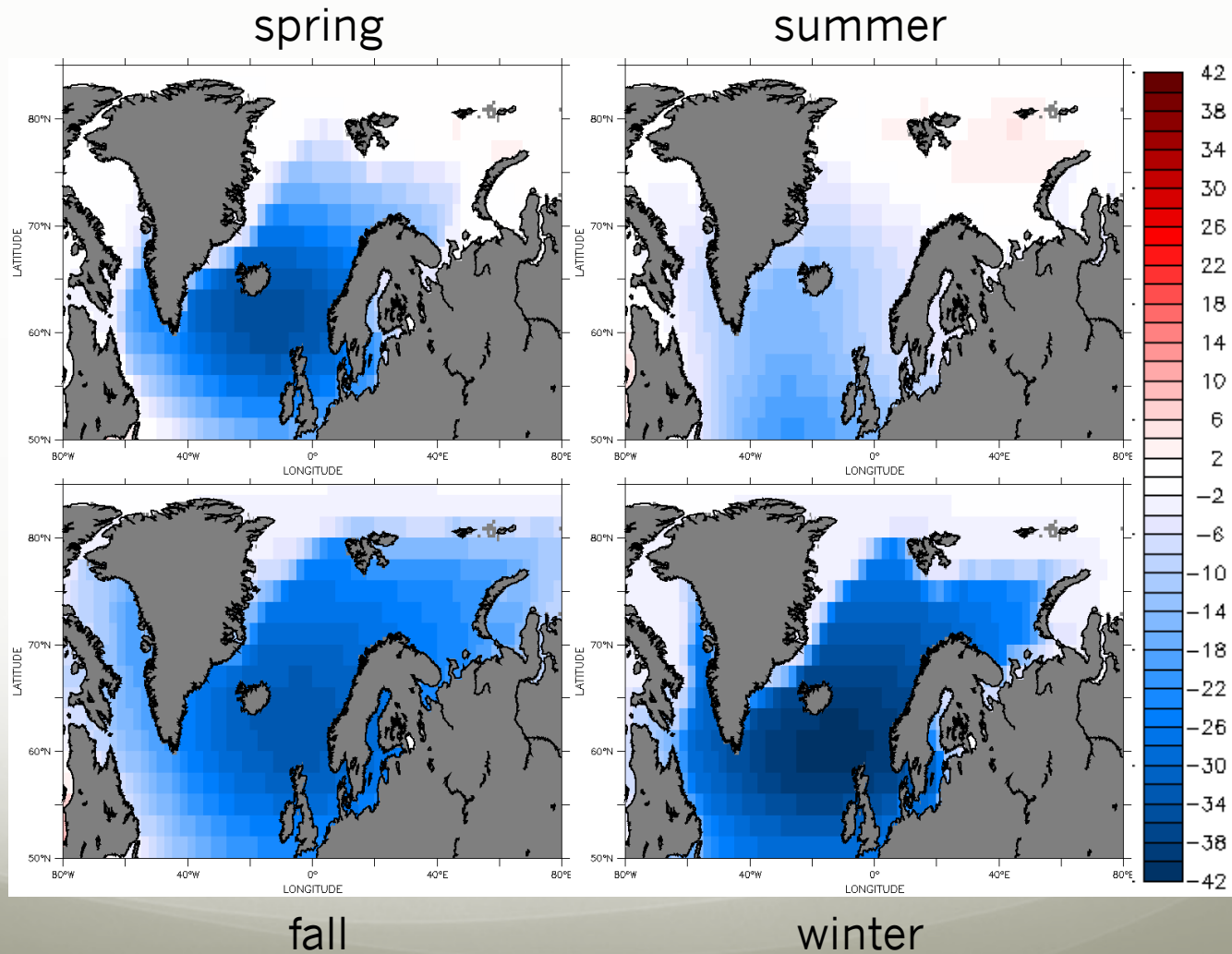
Input/Reference data

Processes

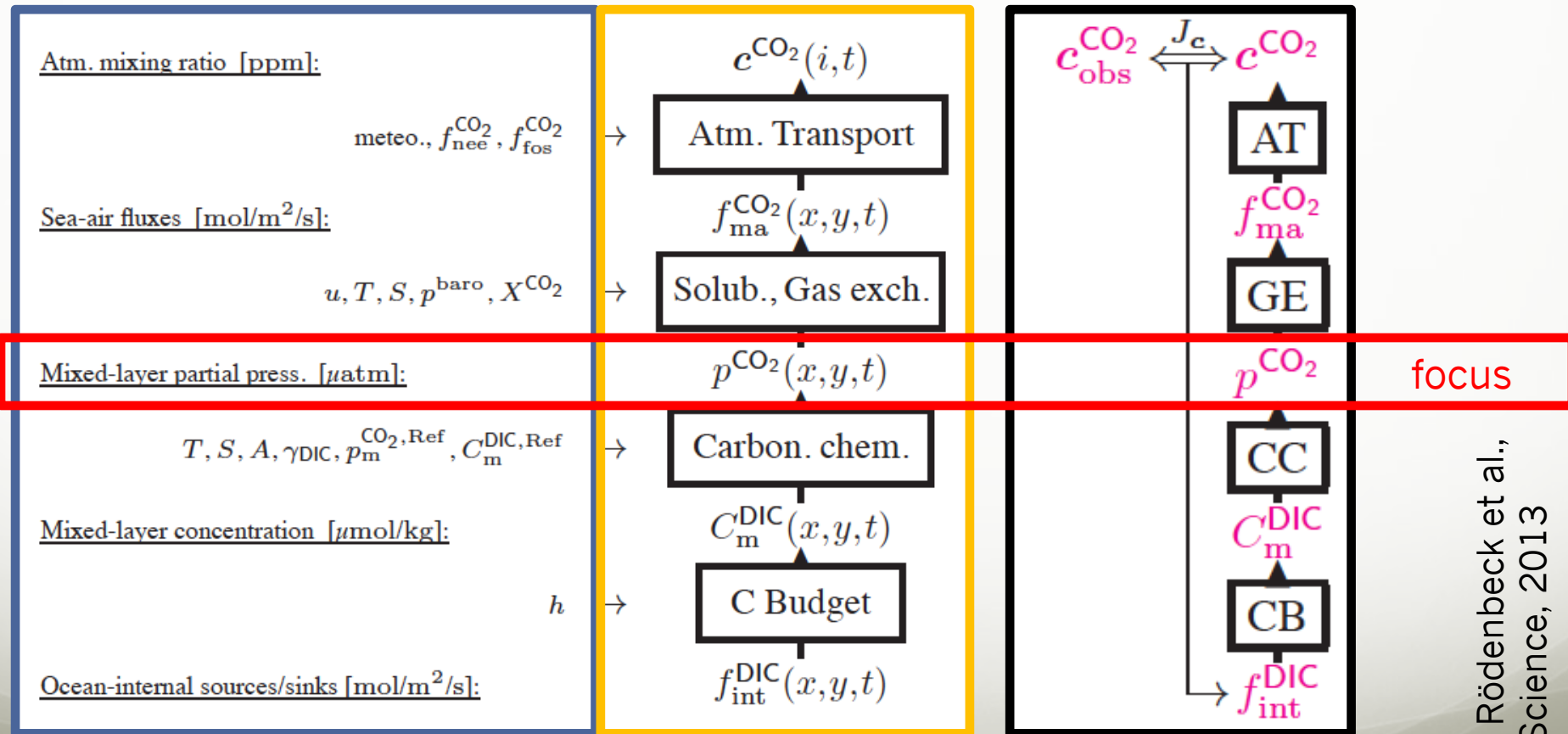
Optimization

# Regional CO<sub>2</sub> inversions, focusing on ocean domains

Ocean-Atmosphere CO<sub>2</sub> flux  
[gC/m<sup>2</sup>/yr]



# Generalized data assimilation framework



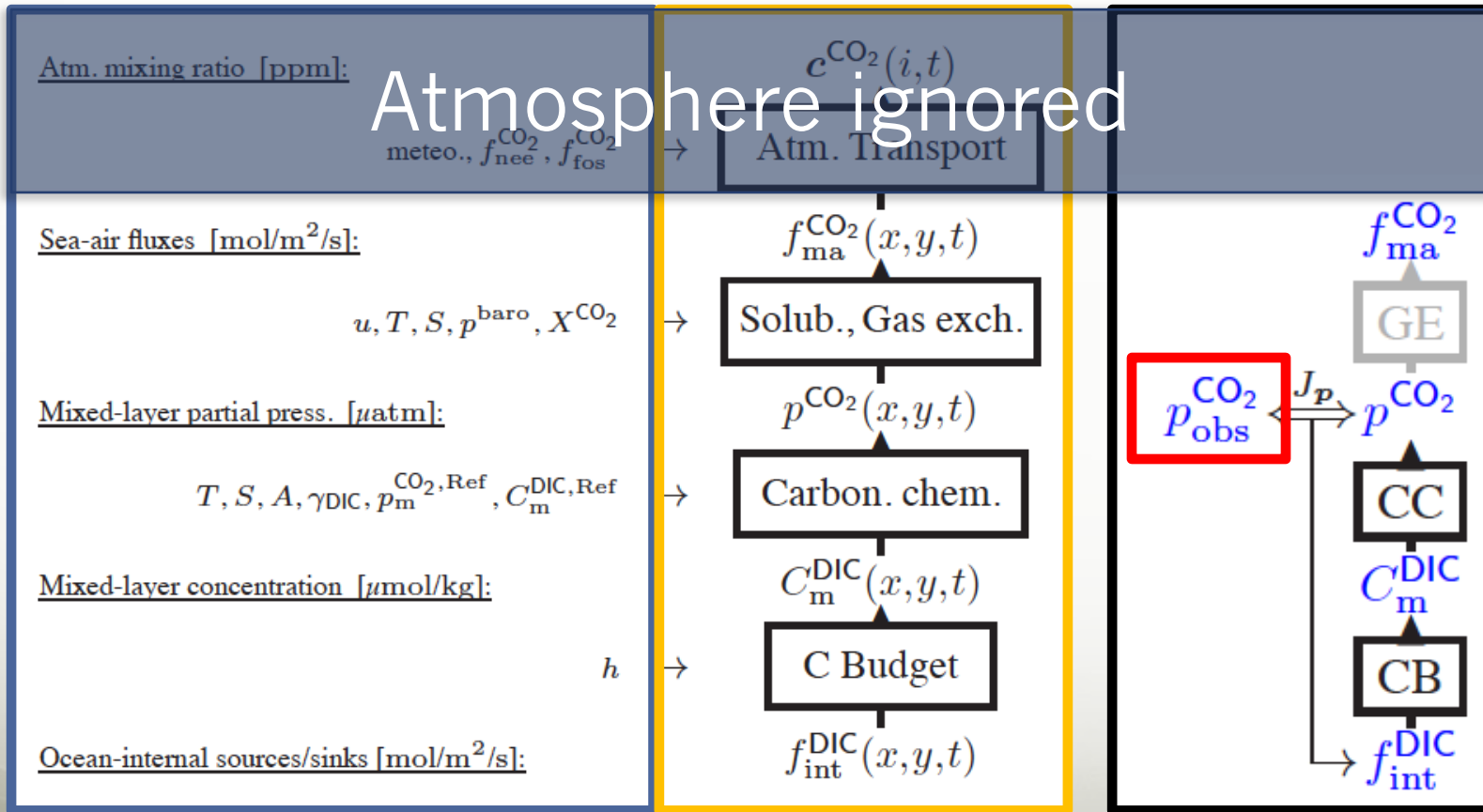
Input/Reference data

Processes

Optimization

# 1: optimize ocean components (pCO<sub>2</sub> inversion)

Atmosphere ignored

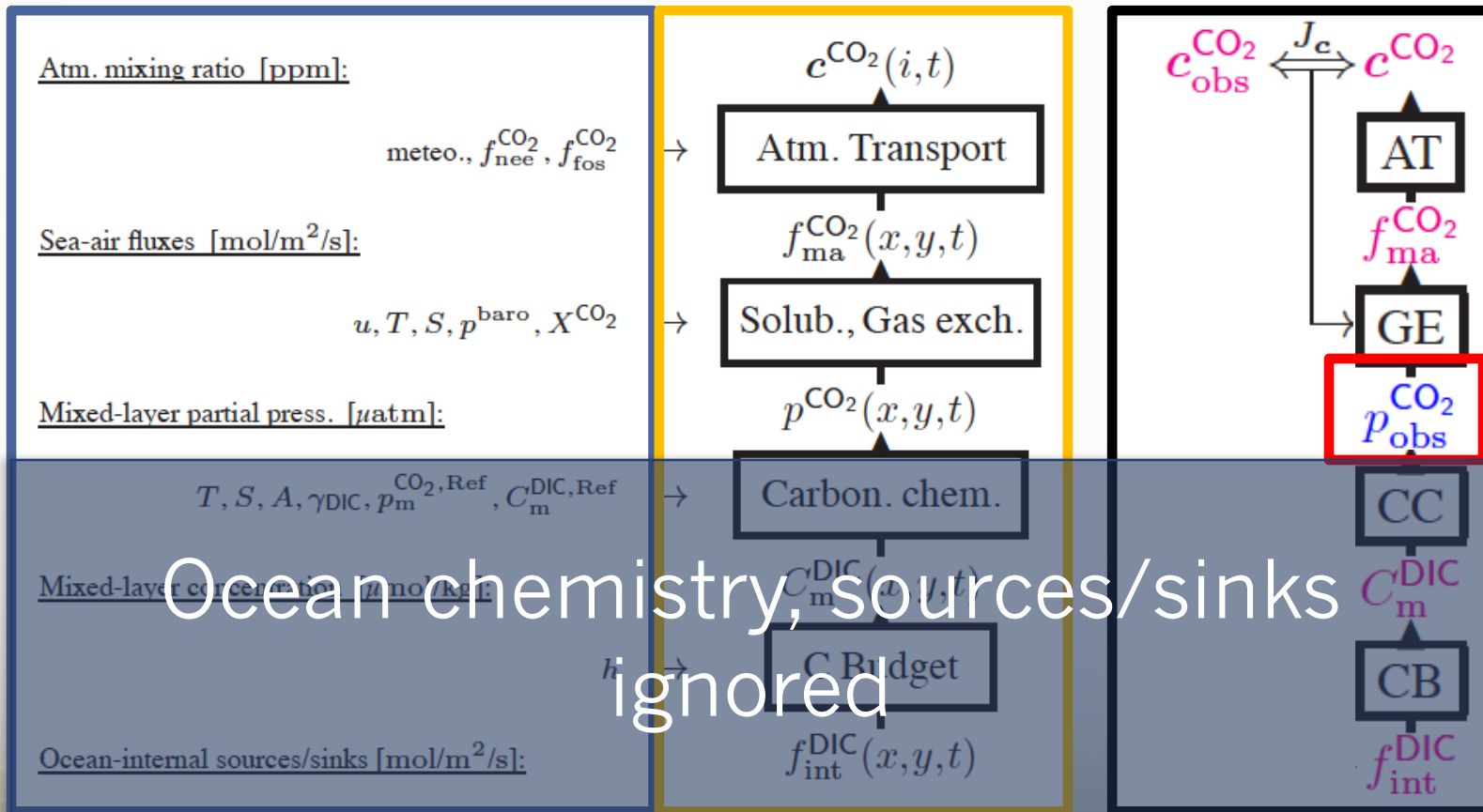


Input/Reference data

Processes

Optimization

# 2: optimize gas exchange coefficients



Input/Reference data

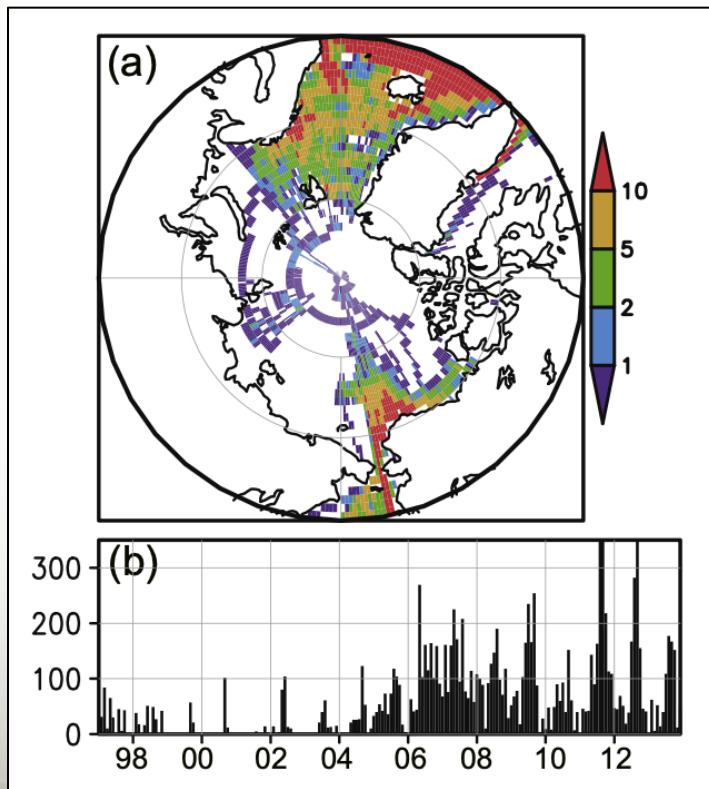
Processes

Optimization

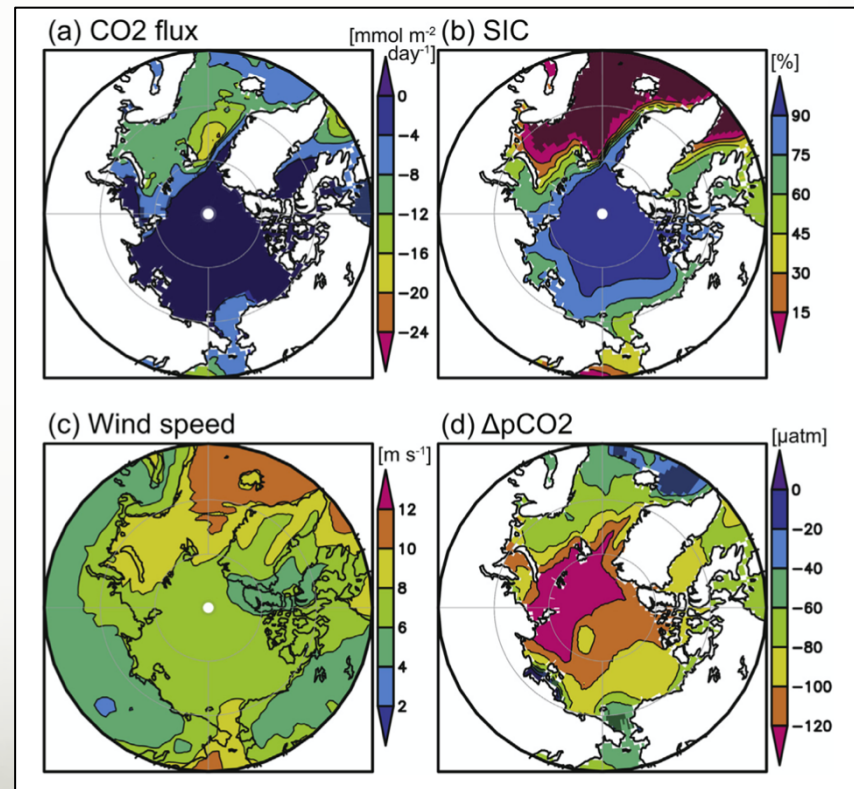


# Mapping air-sea CO<sub>2</sub> fluxes

## Data coverage



## Long-term integration

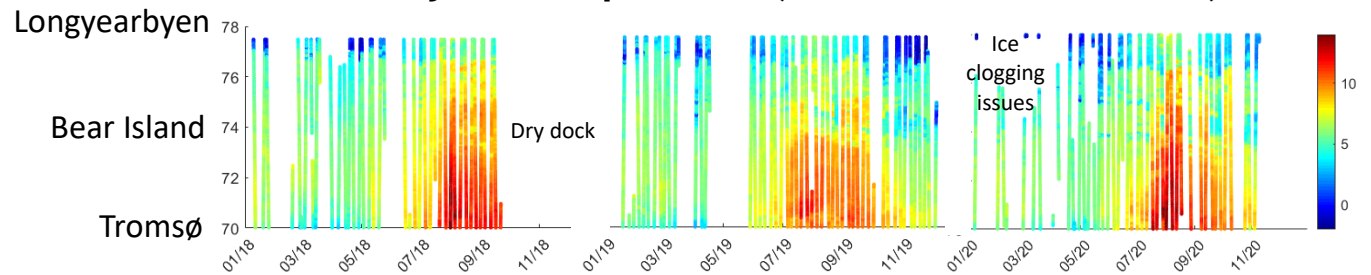


figures: Yasunaka et al.,  
Polar Science, 2016

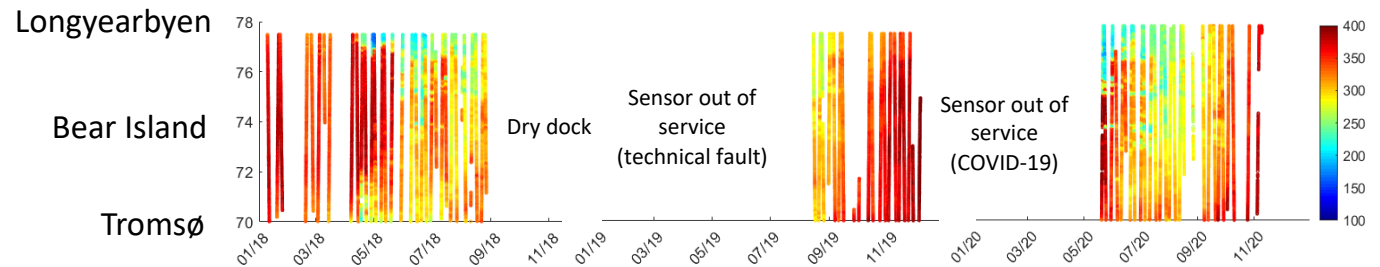
# New NIVA pCO<sub>2</sub> observations from INTAROS



## FerryBox temperature (-2 to 14 °C; n=329782)



## FerryBox fCO<sub>2</sub> (100-400 μatm; n=266230; ±~10 μatm)



# Utilizing new INTAROS pCO<sub>2</sub> observations

A self-organizing neural network mapping technique can produce estimates of surface water pCO<sub>2</sub> by resolving the nonlinear and often discontinuous relationships among proxy parameters, such as chl-a, SSS, SST, MLD and geographical location.

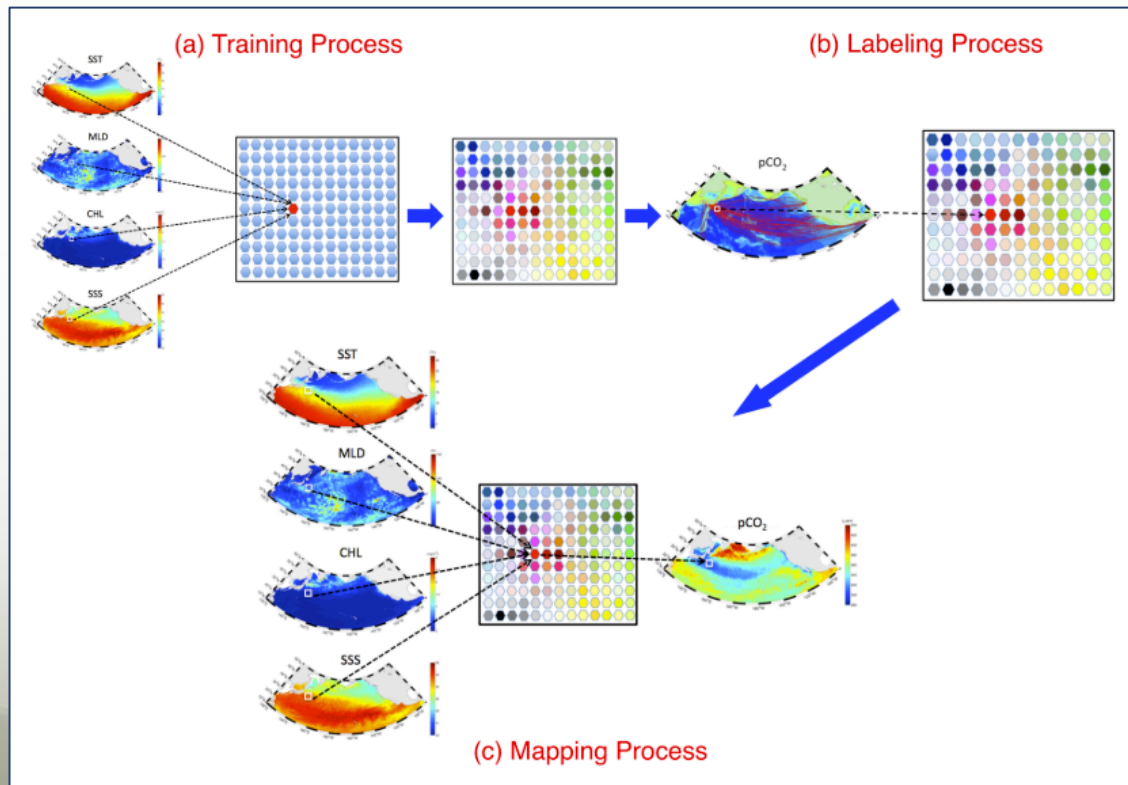


Figure: Nakaoka et al.,  
Biogeosciences, 2013

# Utilizing new INTAROS pCO<sub>2</sub> observations

New pCO<sub>2</sub> observations from NIVA's FerryBox system, combined with other data products from the INTAROS Data Catalogue, will be used to test the validity of self-organizing map techniques in the Barents Sea region.

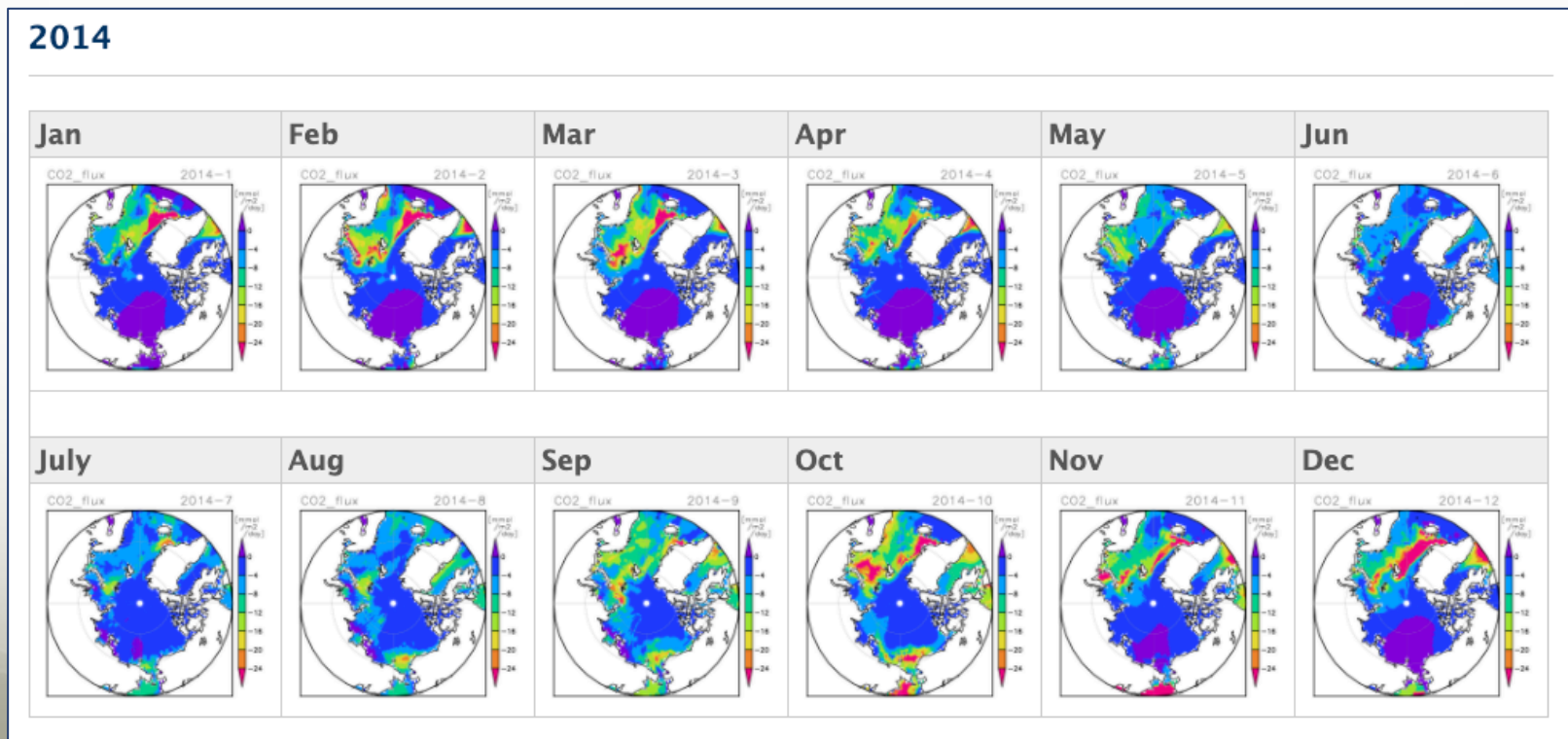


Figure: Yasunaka et al. 2018

# Work plan 2021

- Test new gridded datasets from INTAROS database in regional geostatistical inversion
- Finalize new INTAROS pCO<sub>2</sub> products (WP5 collaboration)
  - Use data to investigate spatio-temporal variability in ocean carbon processes (chemistry, sources/sinks)
  - Couple pCO<sub>2</sub> data with atmospheric observations to investigate variability in gas transfer coefficients in ocean surface layer
- Correlate observed process variability with ancillary data layers from INTAROS database