

eScience Tools for Investigating Climate Change

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NILU - Norwegian Institute for Air
Research

Introduction

Norwegian Institute for Air Research (NILU)

“Research for a clean atmosphere”

- composition of the atmosphere
- climate change
- air quality
- hazardous substances

Nordforsk, NeGI and NeIC

The NCoE eSTICC

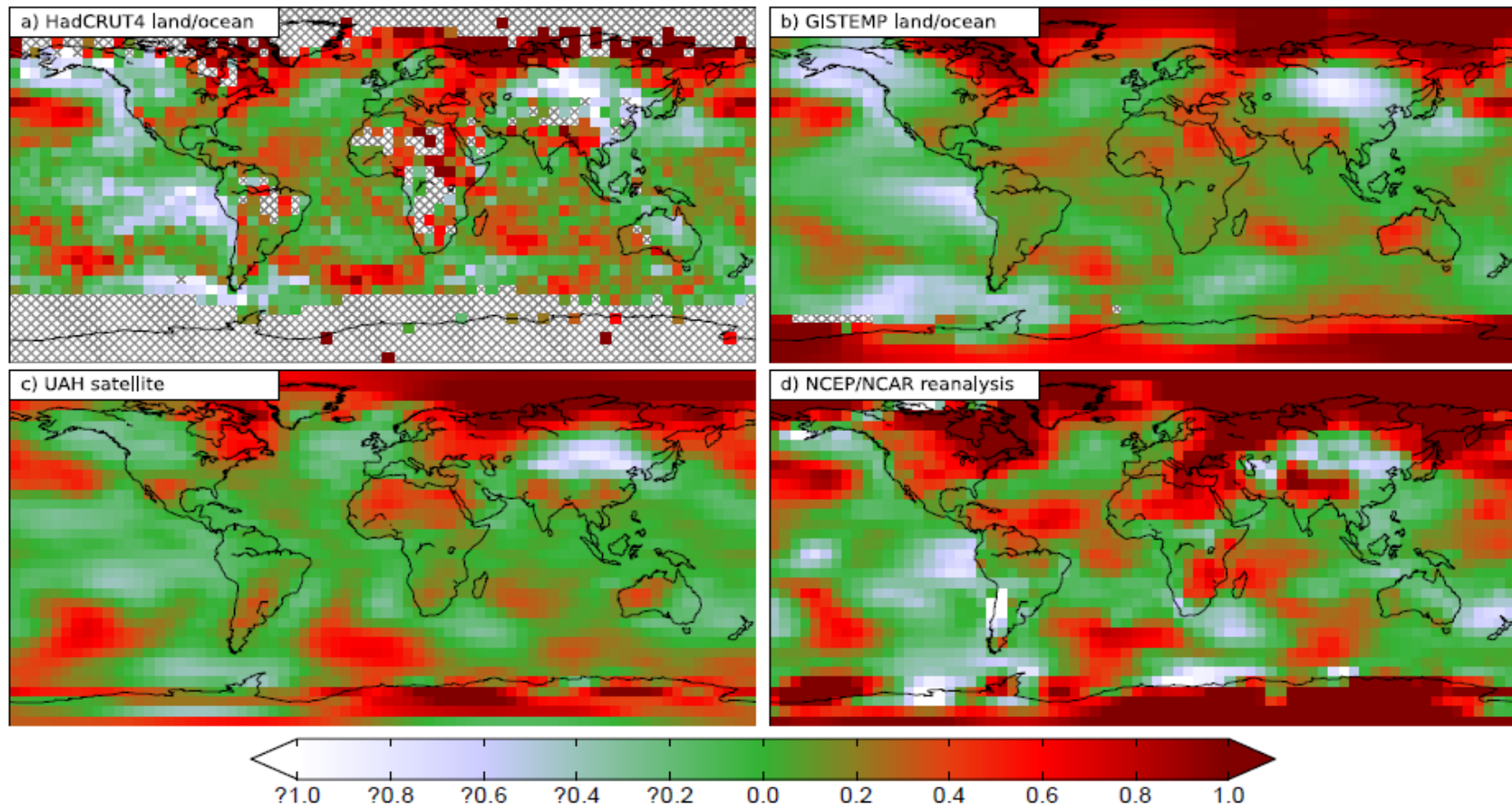
eScience Tools for Investigating Climate Change at High Northern Latitudes



Motivation (1): The Arctic is warming faster than any other place

Underestimate of Arctic warming can probably explain artificial "pause" of global warming (Cowtan and Way, 2013)

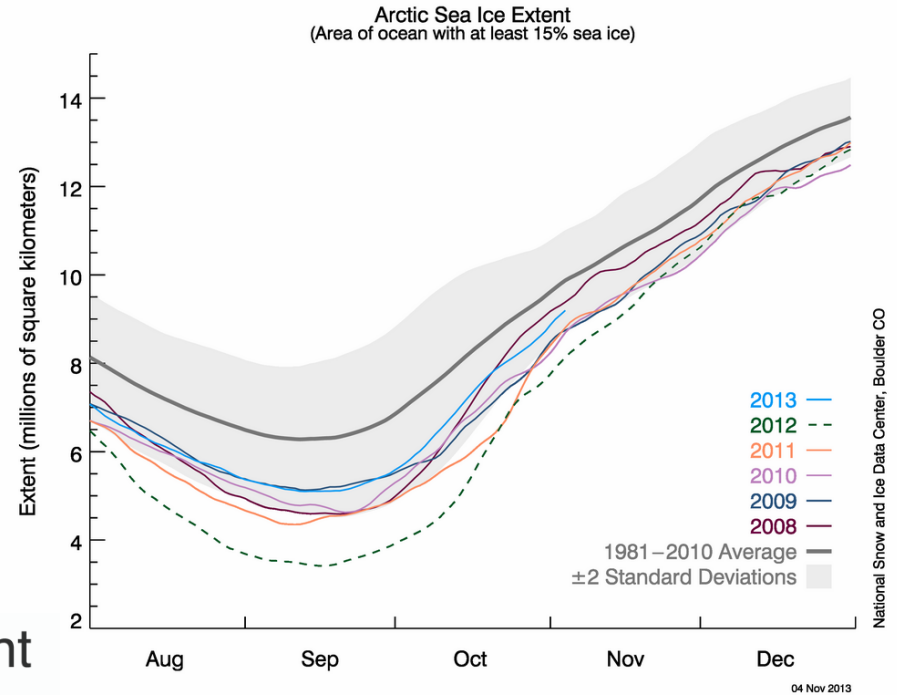
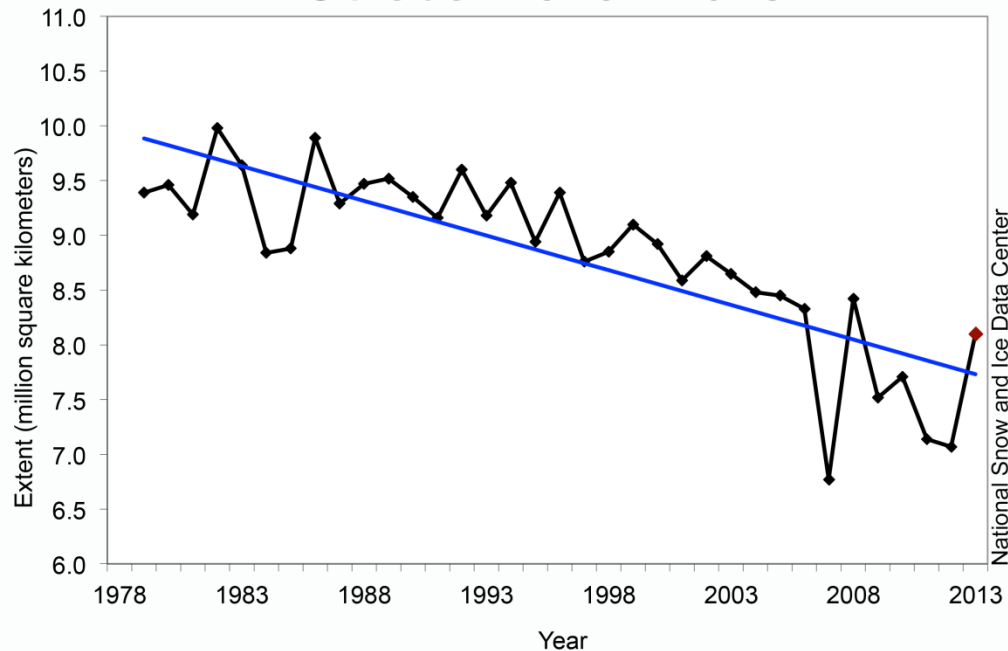
Trend in deg/decade over the last 16 years



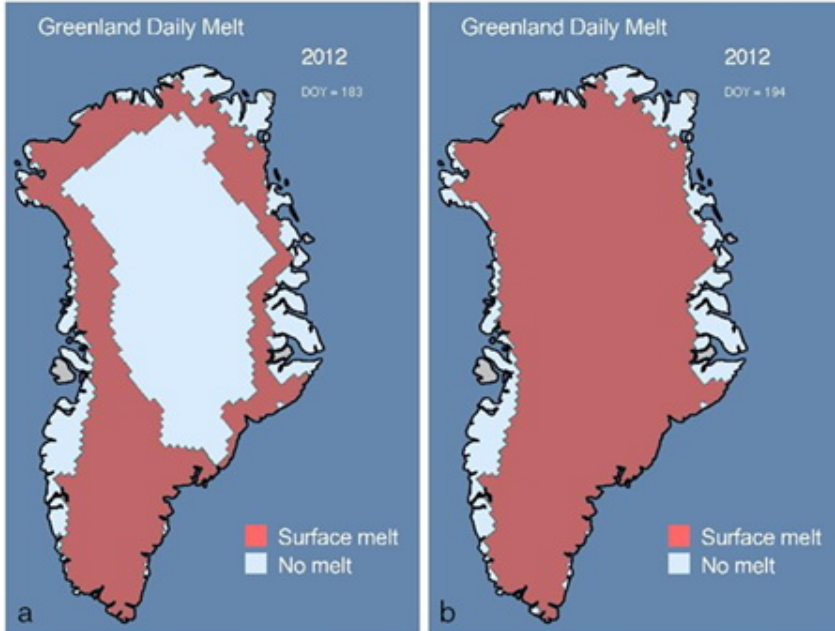
Motivation (2): Sea ice loss



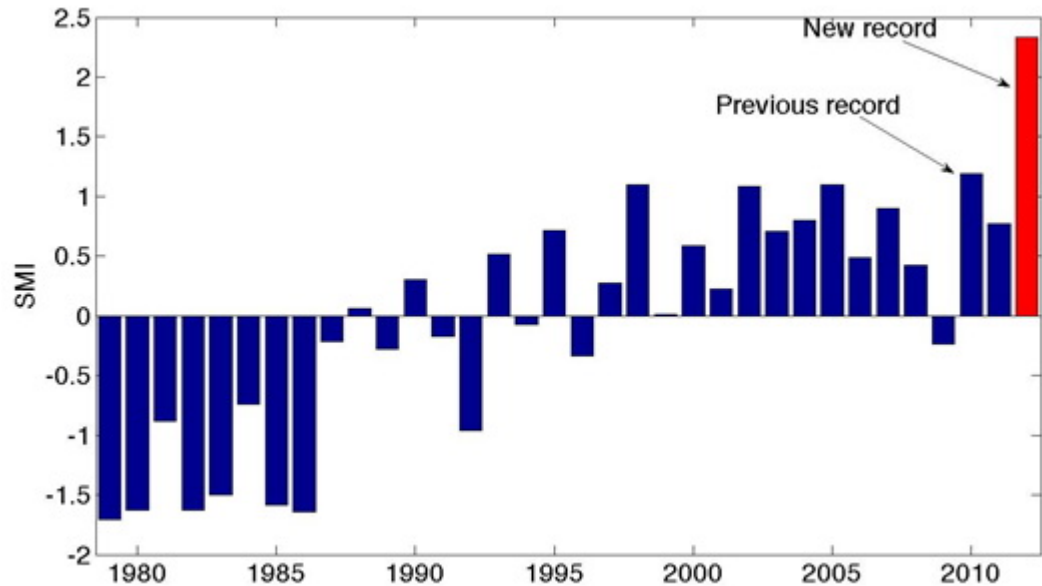
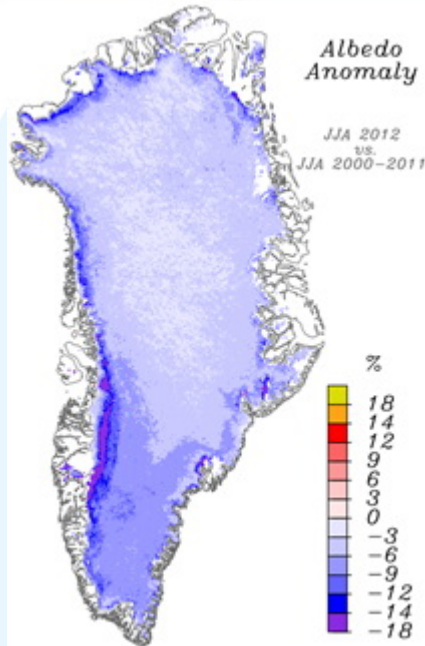
Average Monthly Arctic Sea Ice Extent
October 1979 - 2013



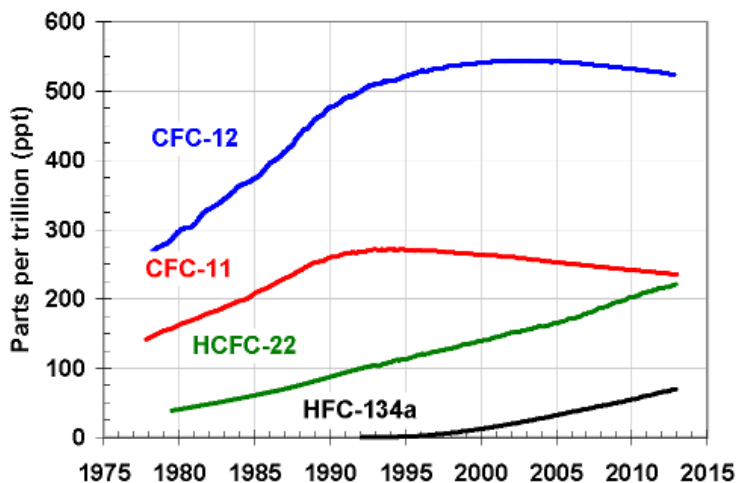
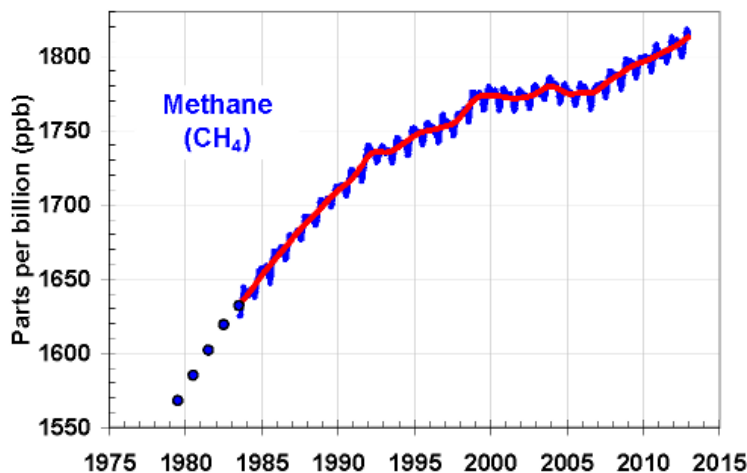
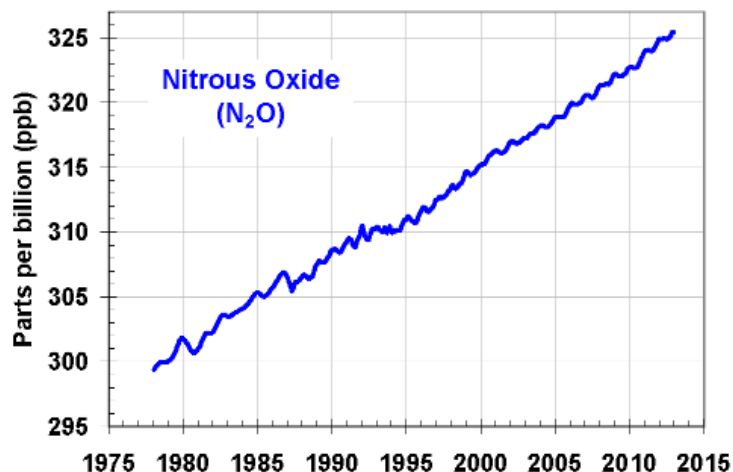
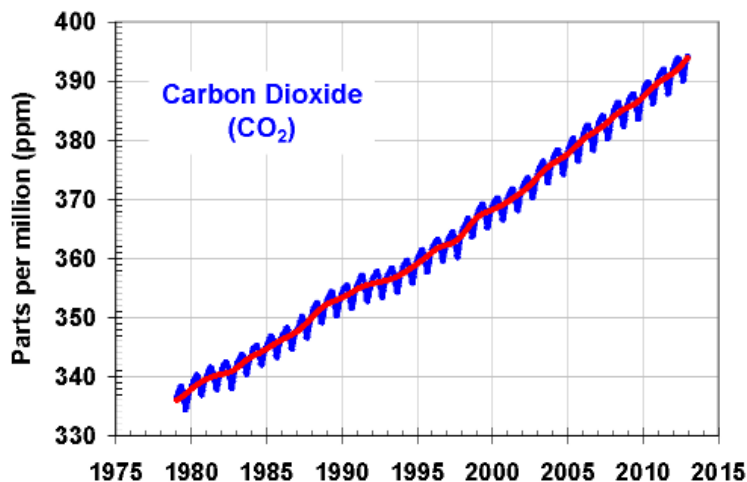
Motivation (3): Ice sheet stability



Greenland melt index:
area-weighted number of days with melting



Motivation (4): Increasing Greenhouse Gas Concentrations



Boreal forests and woodlands store 30% of world's terrestrial carbon; possibly highly sensitive to warming, leading to feedback

Feedback processes

Many are driven by interactions between land, ocean, cryosphere, biosphere and atmosphere

Need to be quantified in Earth system models – but notoriously difficult

Requires interdisciplinary approach

Role of eScience

Data access: Nordic countries host globally important data bases (e.g., World Data Center for Aerosols; AEROCOM; ENSEMBLES), which shall be enhanced

Pre- and Postprocessing of model data becoming a bottleneck

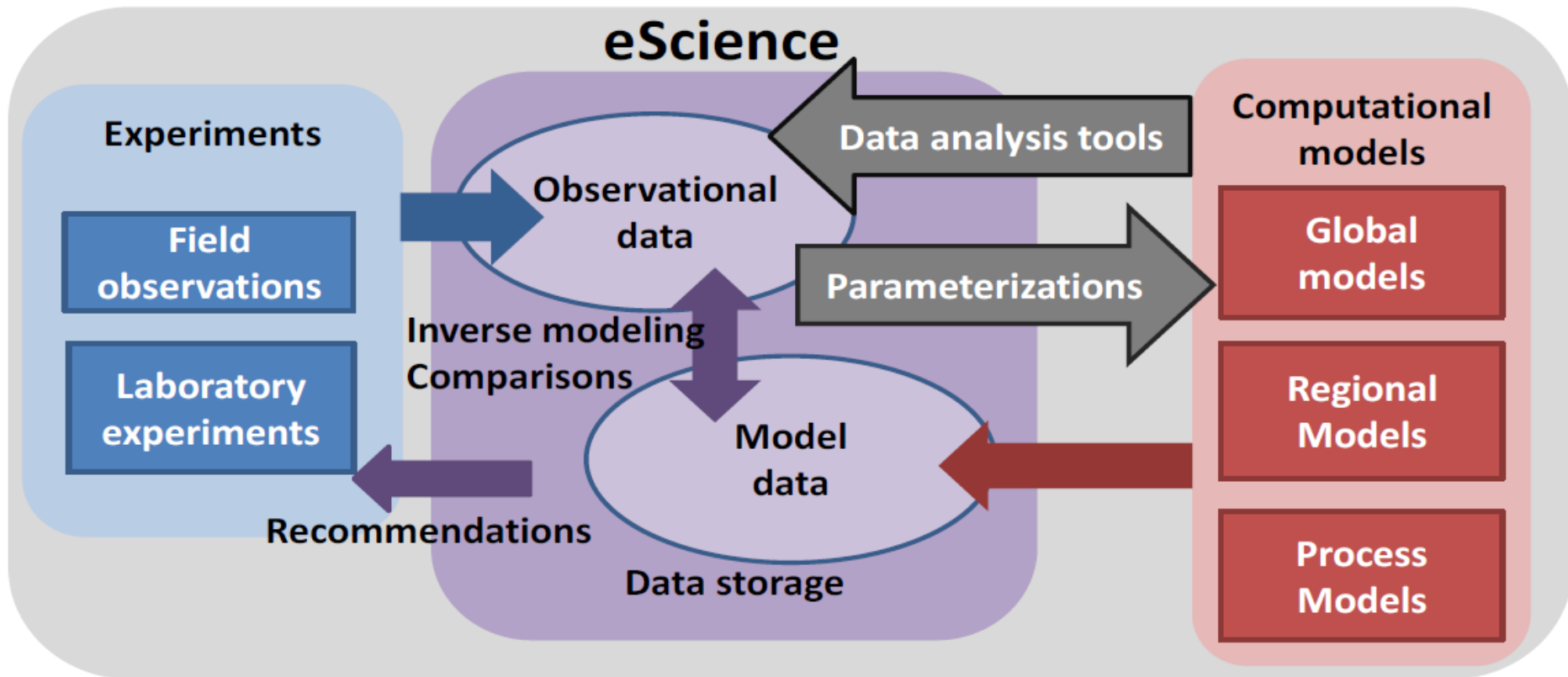
Inverse modeling (greenhouse emission determination): algorithms reach computational limits

Earth System Models: full utilization of computer hardware a challenge (e.g., scalable algorithms)

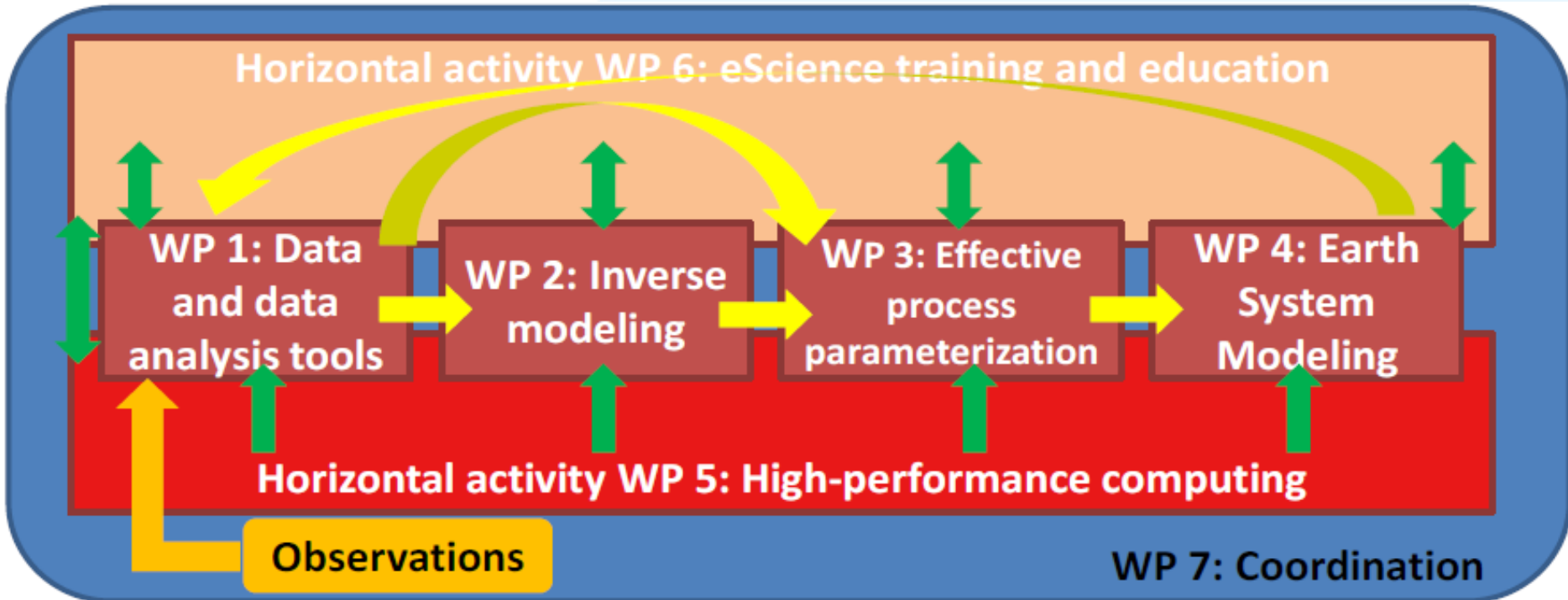
Partnership: 13 partners from 7 Nordic countries, integration across three existing NCoEs (CRAICC, SVALI and DEFROST)

Country	Acronym	Institute	Team leader, WP leaders, some others
NO	NILU	Norwegian Institute for Air Research	<i>A. Stohl, I. Pisso, R. Thompson, M. Cassiani, S. Eckhardt</i>
DK	DMI	Danish Meteorological Institute	Jens H. Christensen , <i>S. Yang, M. S. Madsen</i>
DK	AU	Arctic Research Center, Aarhus University	Mikkel P. Tamstorf , <i>M. Lund</i>
FI	UHEL	University of Helsinki	M. Kulmala , <i>H. Järvinen, A. Lauri, T. Vesala</i>
FI	FMI	Finnish Meteorological Institute	Hannele Korhonen , <i>Tuula Aalto, L. Backman</i>
FI	CSC	CSC- IT Center for Science Ltd	Antti Pursula , <i>Thomas Zwinger, P. Manninen</i>
IS	UICE	University of Iceland	Helmut Neukirchen , <i>Ebba Hvannberg</i>
NO	UoO	University of Oslo	Jon Ove Hagen , <i>Anna Sinisalo, T.V. Schuler</i>
NO	MetNo	Norwegian Meteorological Institute	Michael Schulz , <i>Trond Iversen, Ø. Seland</i>
NO	BCCR	Uni Bjerknæs Centre, Uni Research AS	Helge Drange , <i>Mats Bentsen, Ingo Bethke</i>
SE	SU	Stockholm University	Ilona Riipinen , <i>Peter Tunved, R. Krejci</i>
SE	LU	Lund University	Torben R. Christensen , <i>L. Ström, M. Mastepanov</i>
GL	GCRC	Greenland Climate Research Centre	Søren Rysgaard , <i>John Mortensen</i>

Overall eSTICC concept



eSTICC work packages and information flow



Databases

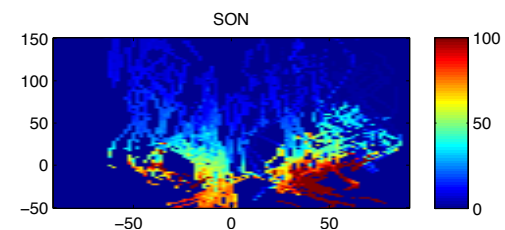
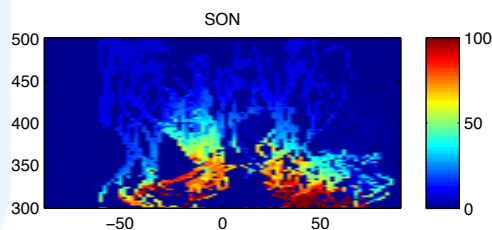
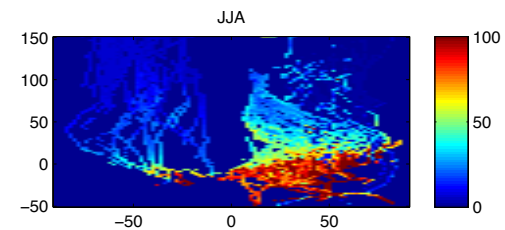
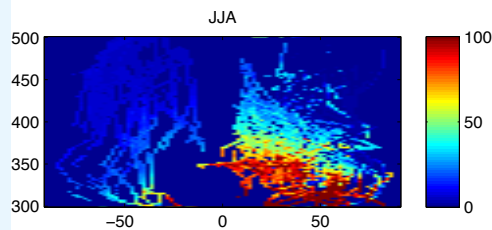
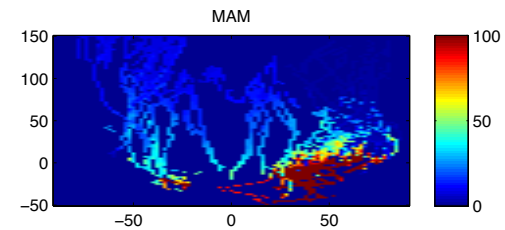
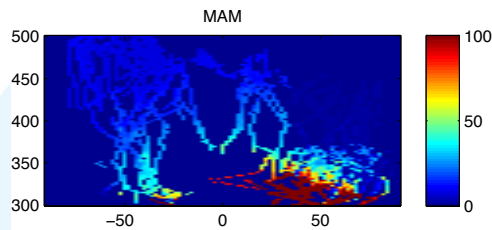
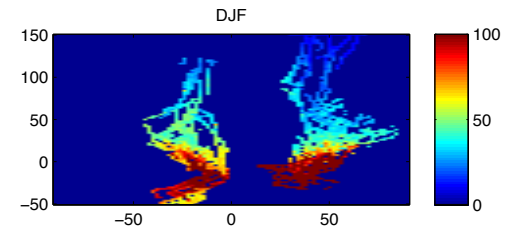
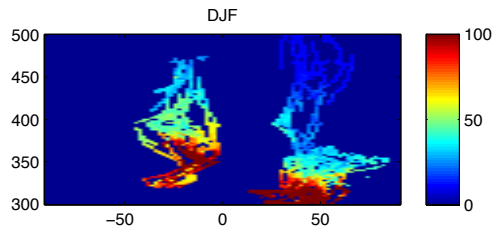
NILU as data centre for observations using different international protocols: databases of atmospheric composition (pH, GHGs, pollutants, aerosols)

EBAS (European Monitoring and Evaluation Programme – EMEP) more than 60 000 datasets (71 different countries, 1060 stations, 608 component types, 23 matrix types, 94 instrument types).

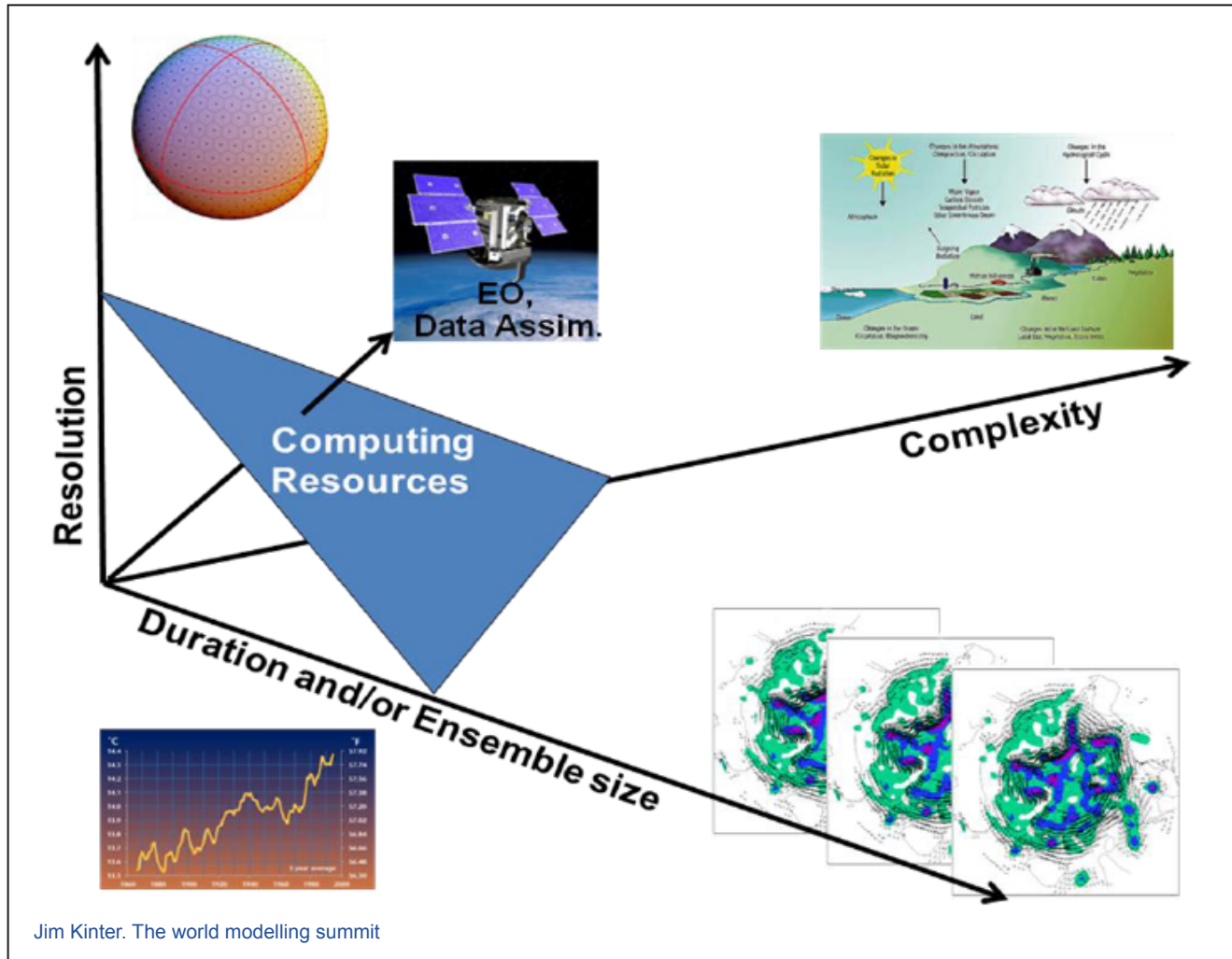
ACTRIS <http://actris.nilu.no> (Aerosols, Clouds, and Trace gases Research InfraStructure Network) more than 500 000 datasets

EVDC - ESA validation data centre (for satellites e.g. Sentinel)

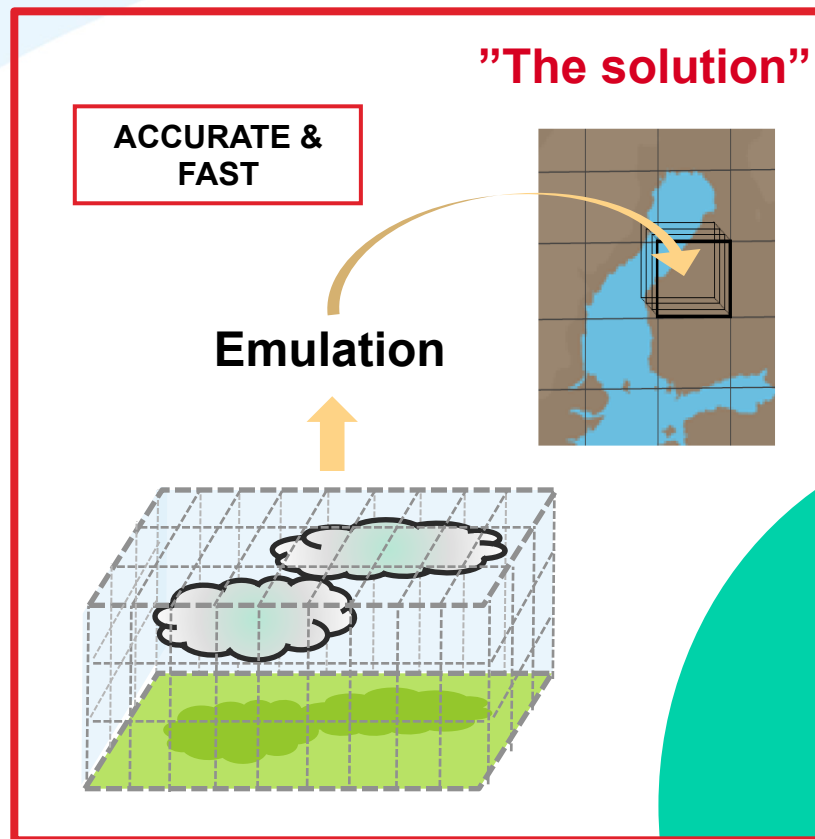
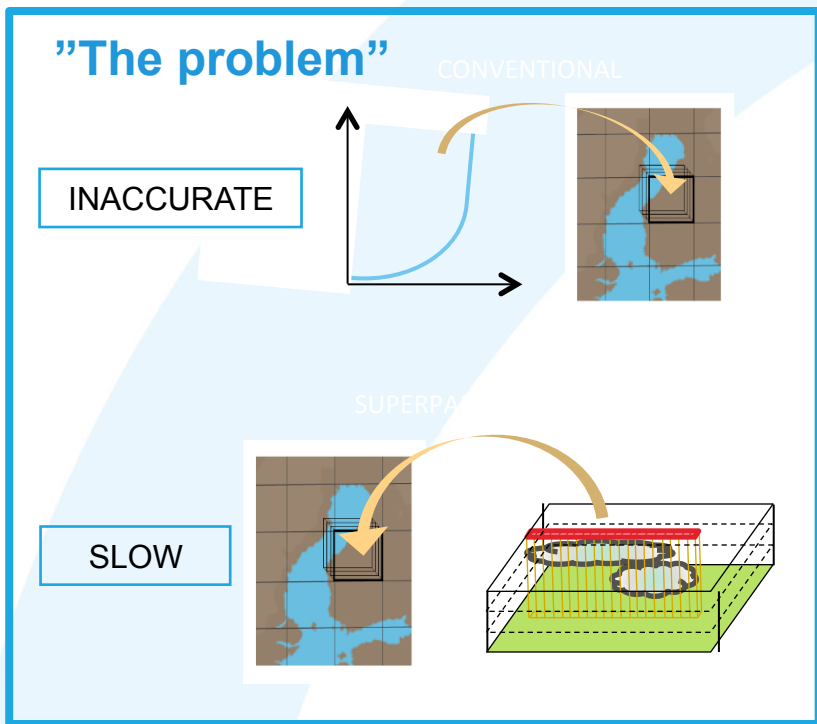
Example of data composite: UTLS CO



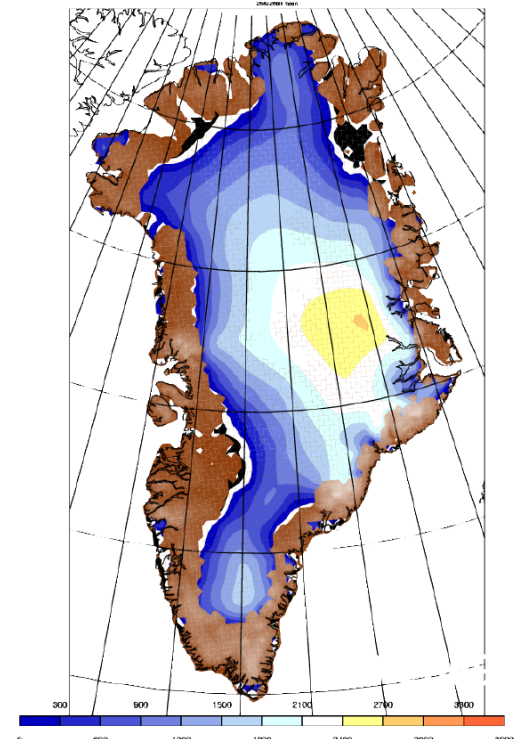
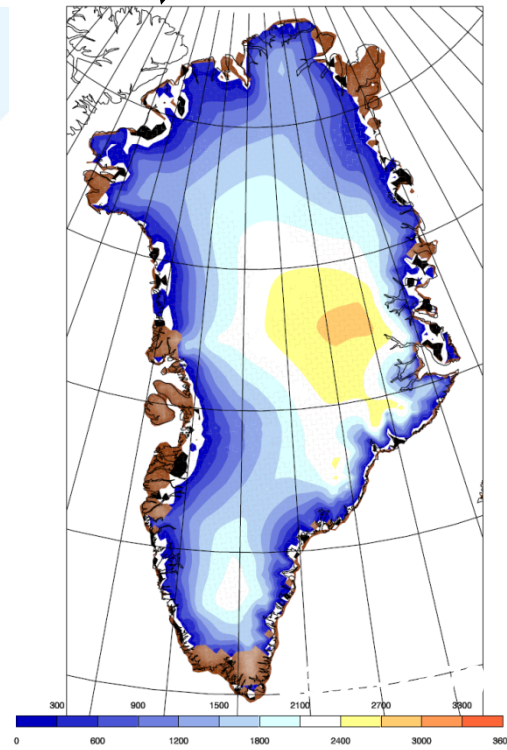
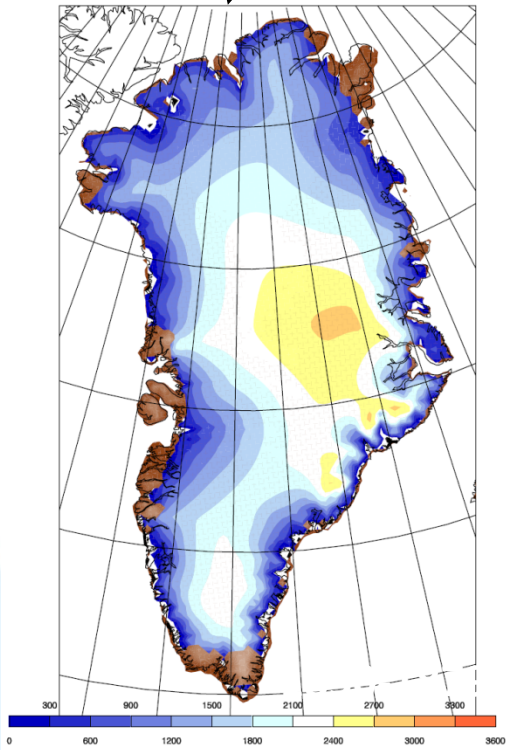
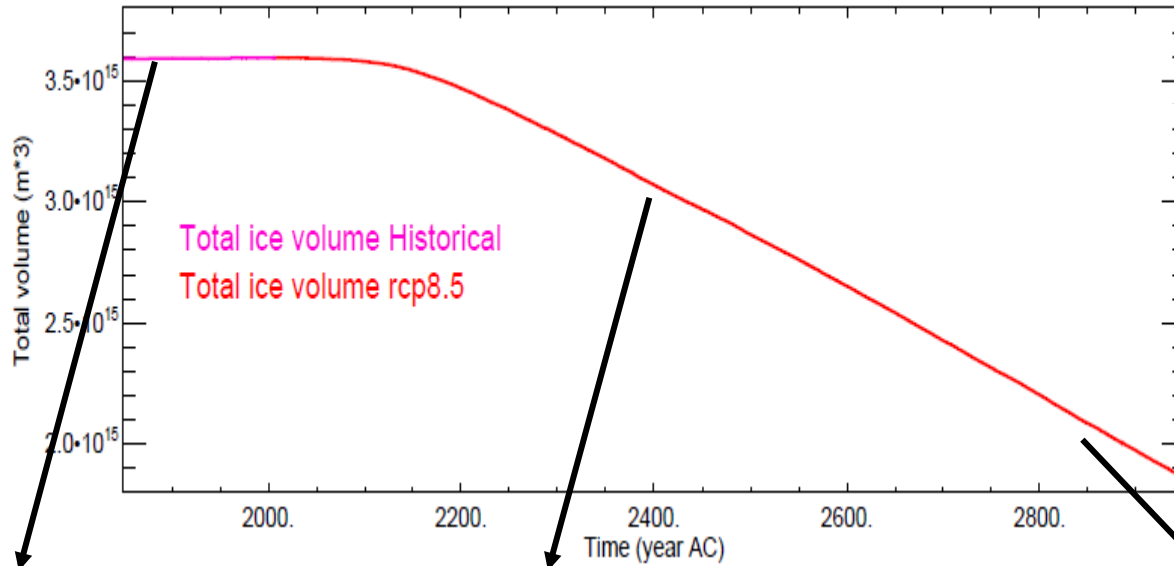
Computing resources for climate modeling are highly dependent on resolution, complexity, duration, ensemble runs and data assimilation



New parametrisations: emulation of sub-grid scale aerosol cloud interactions in climate models (H. Korhonen)



The Greenland ice sheet evolution



FLEXPART web products

<http://niflheim.nilu.no/IgnacioPY/MAGIC/MAGIC.py>

With Sabine Eckhardt and Andreas Stohl

Model set-up

The Lagrangian transport model FLEXPART can be run both forward (from sources) or backward (from measurement stations) in time, whatever is more efficient

e.g.: Backward in time for 20 days

Model output: 4-dimensional emission sensitivity field
(3 space dimensions plus days backward in time)

Mixing ratio = emission sensitivity field x emission flux field

Input data

Input winds from meteorological (ecmwf, ncep), climatic (NorESM) and mesoscale (WRF) models

ECMWF analyses:

Resolution 1x1 degree, but 0.1 x 0.1 degree in the area of interest

Large volume of data (1 file ~ 500 Mb, hourly resolution)

Available products

The model output consists of a gridded 3-d response function to an emission input. The response function (emission sensitivity) gives the contribution (in ppb) a source of unit strength at a particular location would make at the measurement location

The response function is proportional to the total residence time of the particles on the 3-d grid

Available products

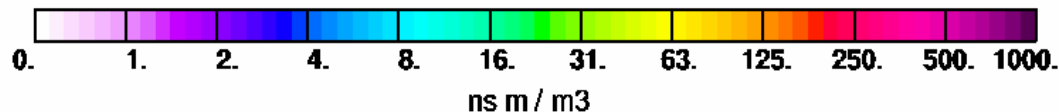
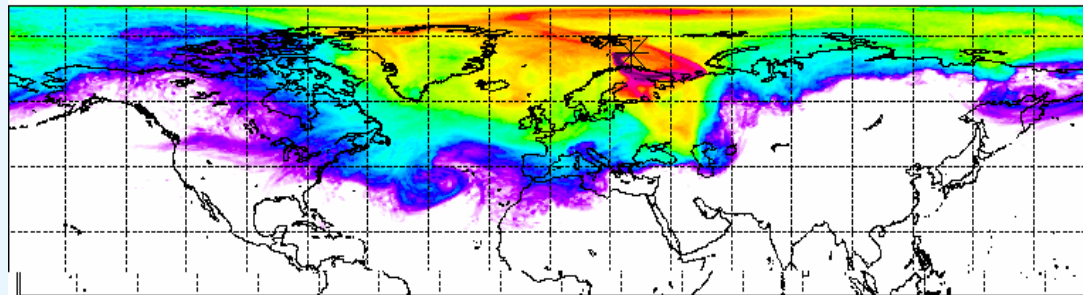
1. Column-integrated emission sensitivity: A first overview of where emissions could have influenced the air

Column-integrated emission sensitivity in global domain for 20140717_Helmer_Har

Start time of sampling 20140717.182135 End time of sampling 20140717.204827

Lower release height 0 m Upper release height 0 m

Aerosol tracer used, meteorological data are from ECMWF

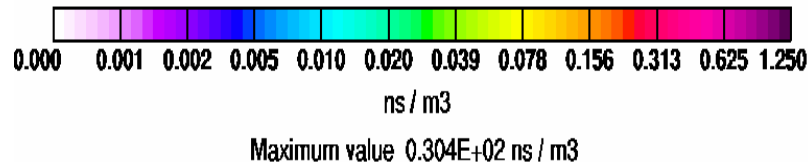
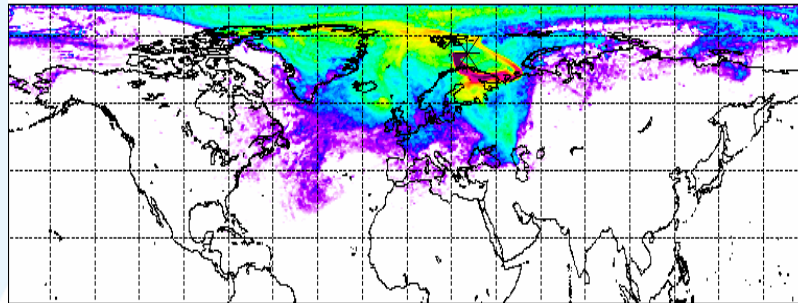


Maximum value 0.361E+04 ns m / m³

Available products

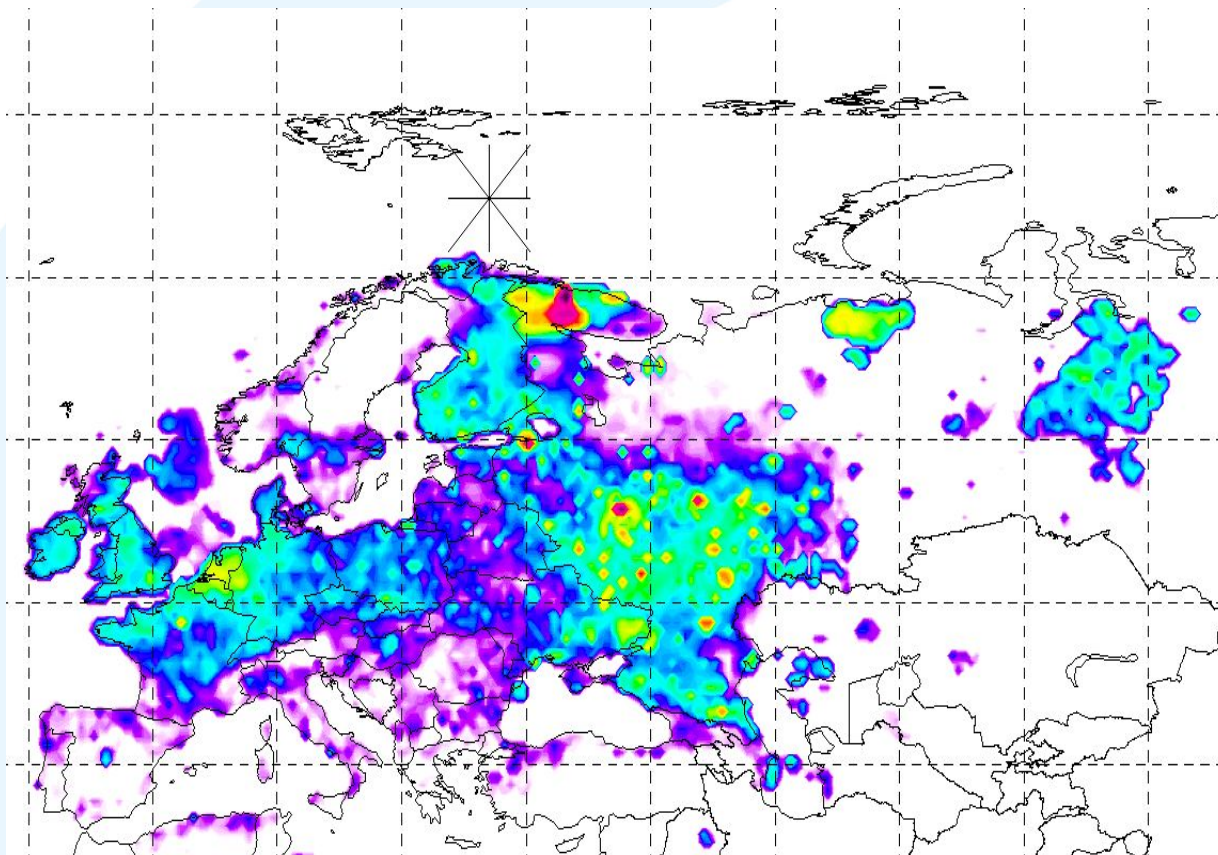
- Footprint residence time: The residence time in the lowest 150 meters, because most anthropogenic emission sources are located there.

Footprint emission sensitivity in global domain for 20140717_Helmer_Hansen
Start time of sampling 20140717.182135 End time of sampling 20140717.204827
Lower release height 0 m Upper release height 0 m
Aerosol tracer used, meteorological data are from ECMWF



Available products

3. Source contribution maps for CH₄: The footprint multiplied with the anthropogenic emission flux from ECLIPSE inventory (17 July)



Inverse flux modeling

Optimal fluxes given by:

$$\min_x [(\mathbf{x} - \mathbf{x}_0)^T \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_0) + (\mathbf{H}\mathbf{x} - \mathbf{y})^T \mathbf{R}^{-1}(\mathbf{H}\mathbf{x} - \mathbf{y})]$$

where:

\mathbf{y} = observations

\mathbf{x}_0 = prior fluxes

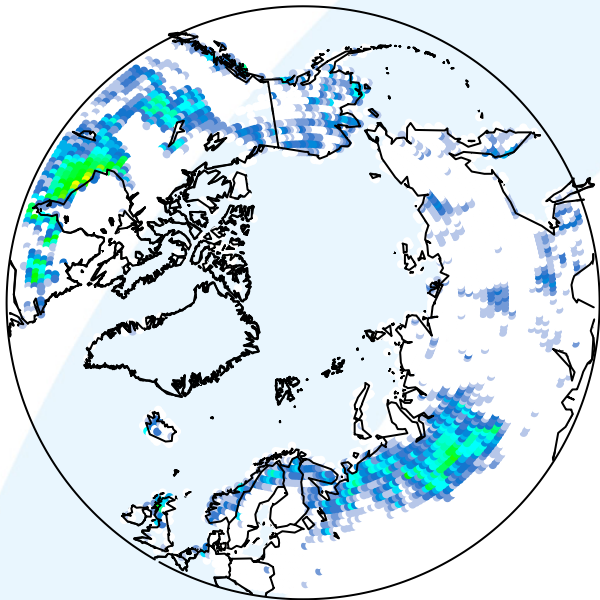
\mathbf{H} = matrix of source-receptor relationships

\mathbf{B} = prior flux error covariance matrix

\mathbf{R} = observation error covariance matrix

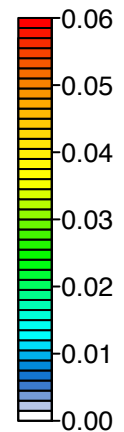
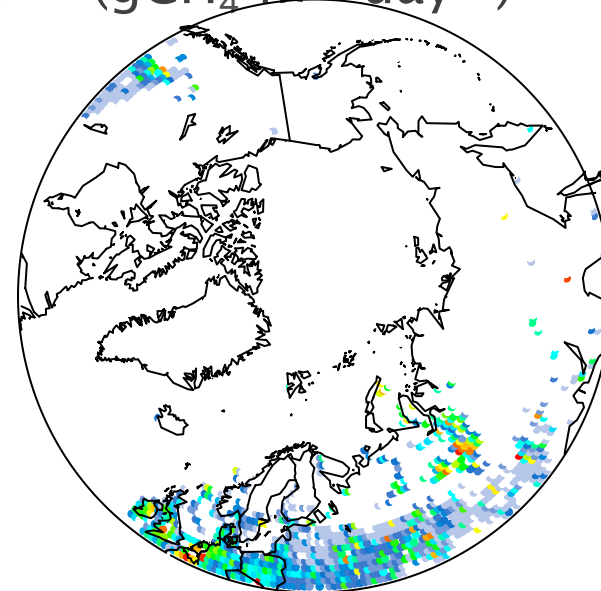
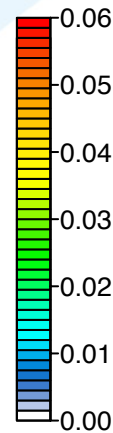
CH₄ Prior flux estimates (R. Thompson)

Annual mean wetlands
(gCH₄ m⁻² day⁻¹)



Bergamaschi et al.
(2007)

Annual mean anthropogenic
(gCH₄ m⁻² day⁻¹)



EDGAR-4.2 FT2010



Kaplan scheme
implemented in the LPJ
DGVM model

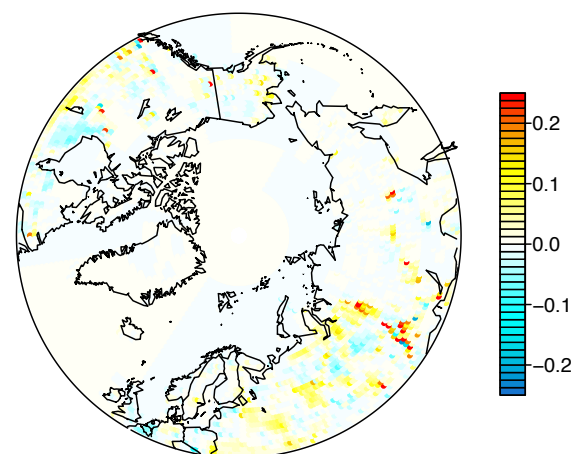
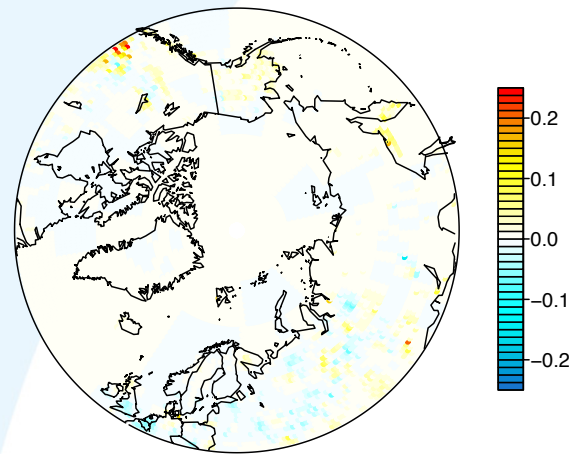
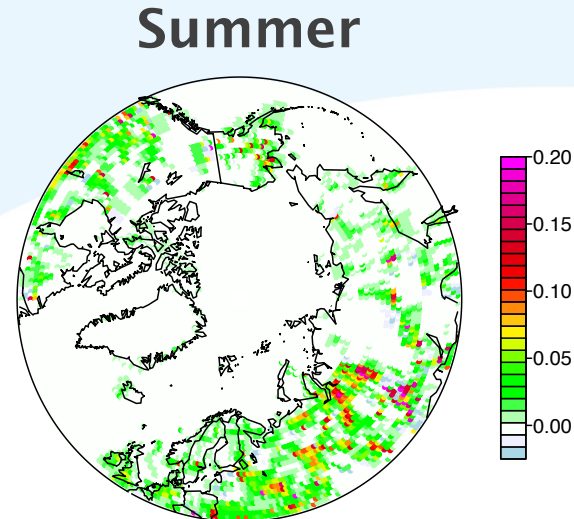
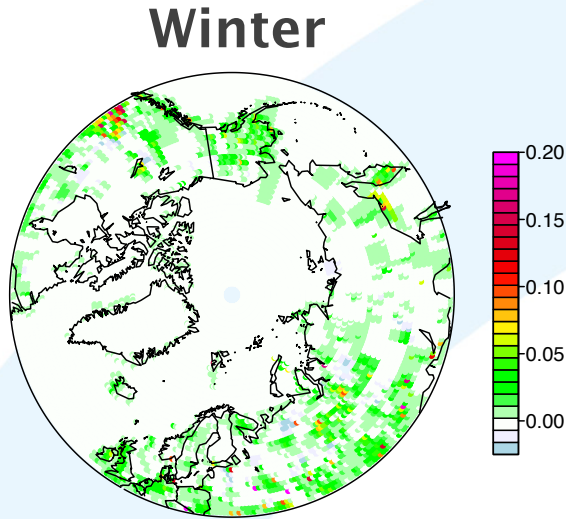
22/09/2015

ESTICC Meeting,
Reykjavik

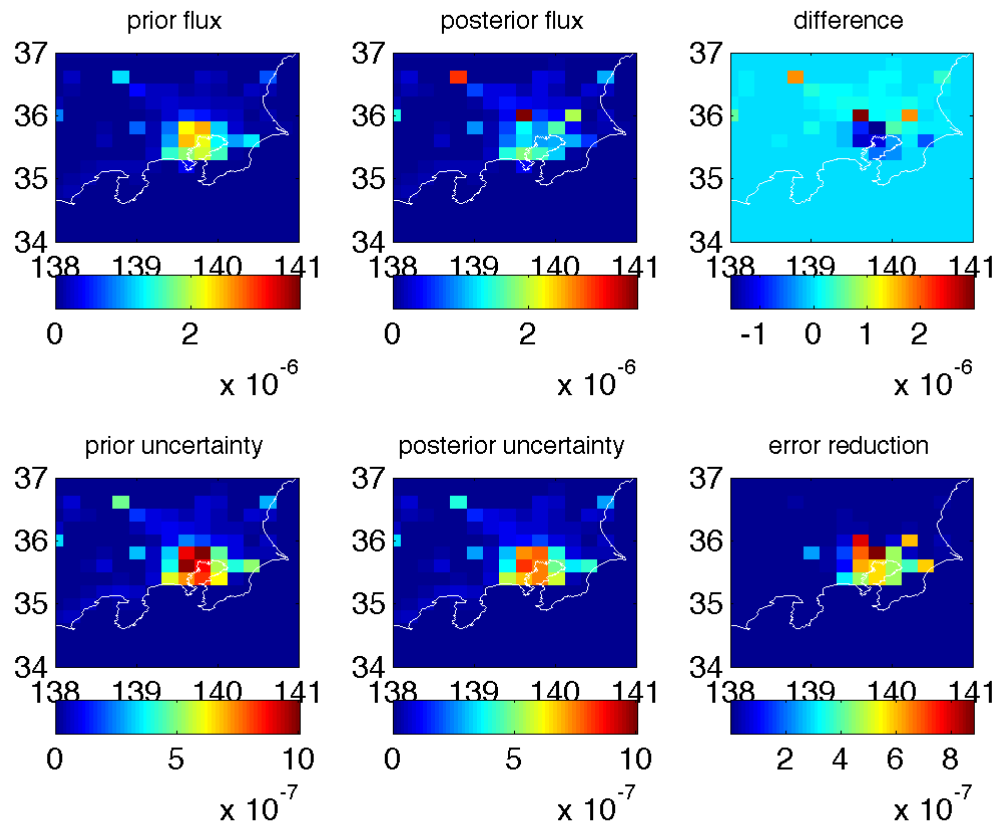
CH₄ Posterior fluxes (R. Thompson)

Posterior - prior
(gCH₄ m⁻² day⁻¹)

Posterior fluxes
(gCH₄ m⁻² day⁻¹)



Anthropogenic CO₂



Summary

- big storage volume
 - ESM modeling -> Lagrangian model input
 - Lagrangian model output
 - Analysis
 - Visualisation
- large speed of transfer/processing required
 - Inverse (and direct) modeling
 - Transfer and sharing of large datasets
- complexity of data
 - Different sources of measurements
 - Database harmonisation including model data (e.g. fluxes, footprints, source contributions)

Outlook

We need to upgrade our coding panoply

Better management of the databases of model output and measurement

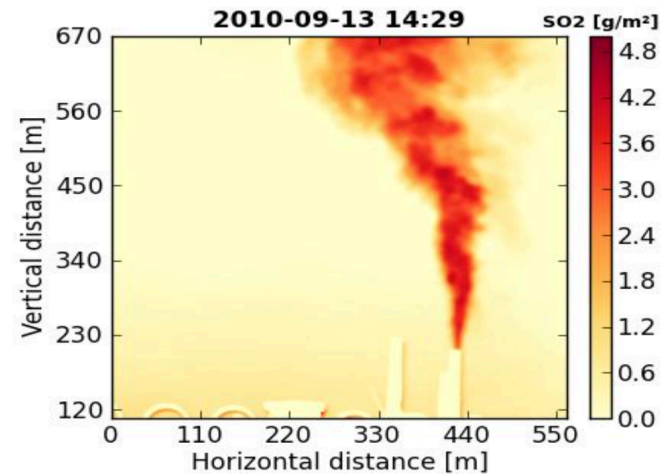
Not always best practices. Other disciplines (e.g. machine learning) could provide some insights on good practices

COMTESSA

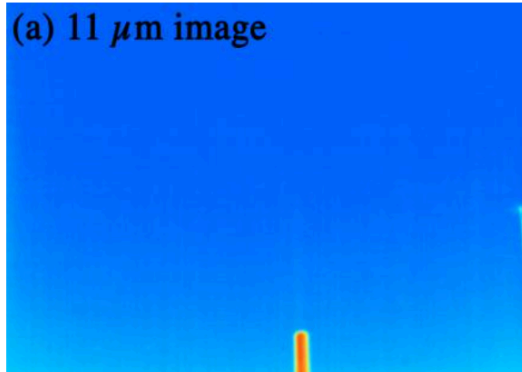
(Camera Observation and Modelling of 4D Tracer Dispersion in the Atmosphere)



Figure 2: Envicam-2 UV (blue) and NicAir-2 IR camera (on tripod), monitoring emissions from Bardarbunga volcano, Iceland.



(a) 11 μm image



(b) 8.6 μm image

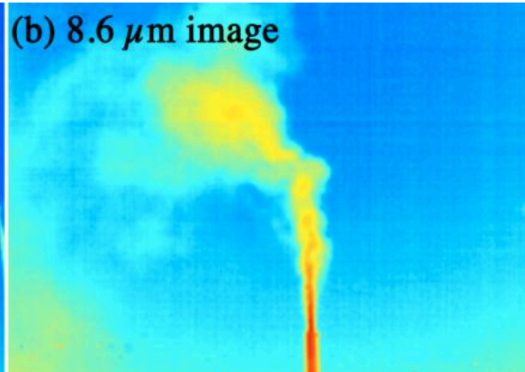


Figure 3. (a) IR camera image obtained at 11 μm of an SO_2 plume from a stack at an industrial smelter (aluminium). (b) Exactly the same plume but imaged at 8.6 μm , where SO_2 has a strong absorption. Combining these images allows the temperature and concentration effects to be separated.

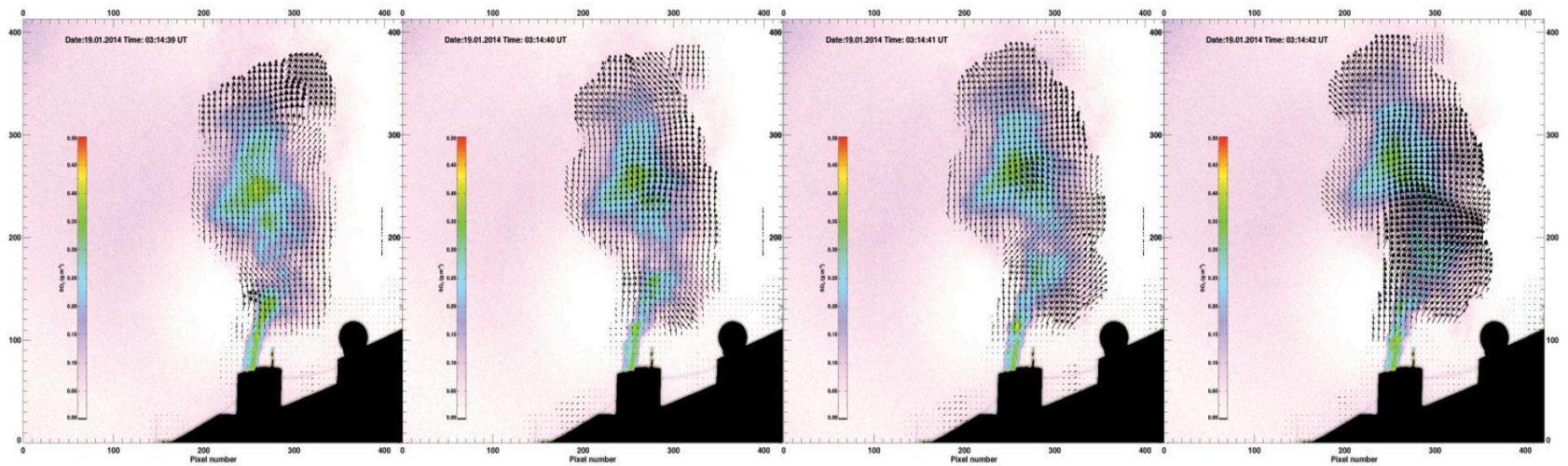
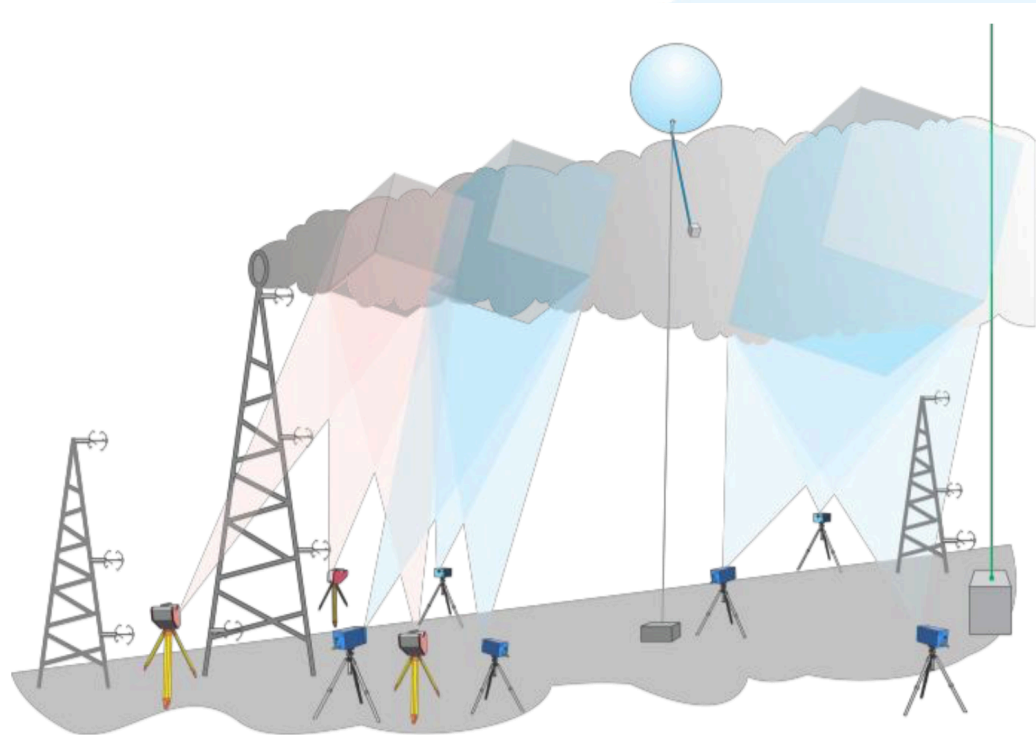


Figure 1. Sequence (~ 1 s time interval) of UV camera retrievals of SO_2 and velocity vectors for a ship plume.



Aim: time series of
3D tomographical
reconstruction
To validate
 1024^3 LES simulations

Fig. 3: Sketch of a plume measurement campaign with both UV cameras (blue) and IR cameras (red), as well as a ceilometer, instrumented masts and a tethered balloon.

Thank you!