

# Inverse modeling of greenhouse gas and aerosol precursor emissions using satellite measurements *Status and future perspectives*

S. Houweling (IMAU/SRON)  
and M. Schaap (TNO)

The background of the slide features a scenic landscape. The top half shows a bright sunset or sunrise with a warm orange and yellow glow against a blue sky with scattered white clouds. The bottom half shows a misty, mountainous region with green, forested slopes and deep valleys, partially obscured by a light haze.

# Outline

- Brief introduction to inverse modeling
- Current status: CO<sub>2</sub>
- Future perspectives: CO<sub>2</sub>
- Application to aerosol precursors

# Inverse modelling: Principle

Obs:



Cabauw tower

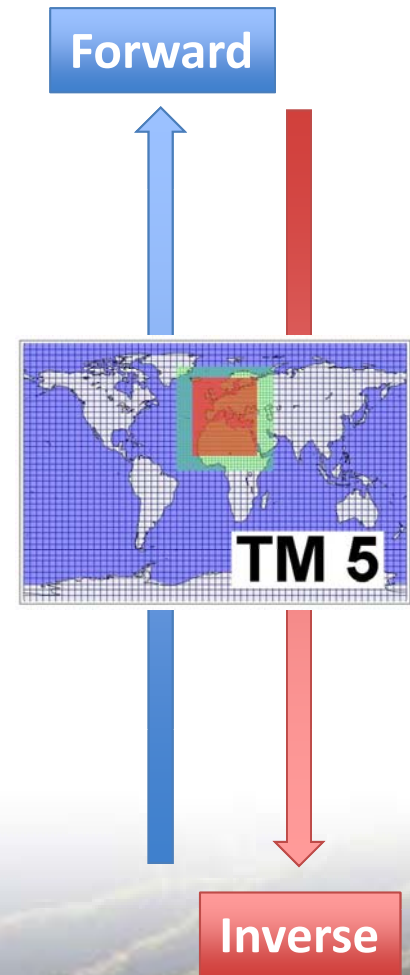


Geophysica

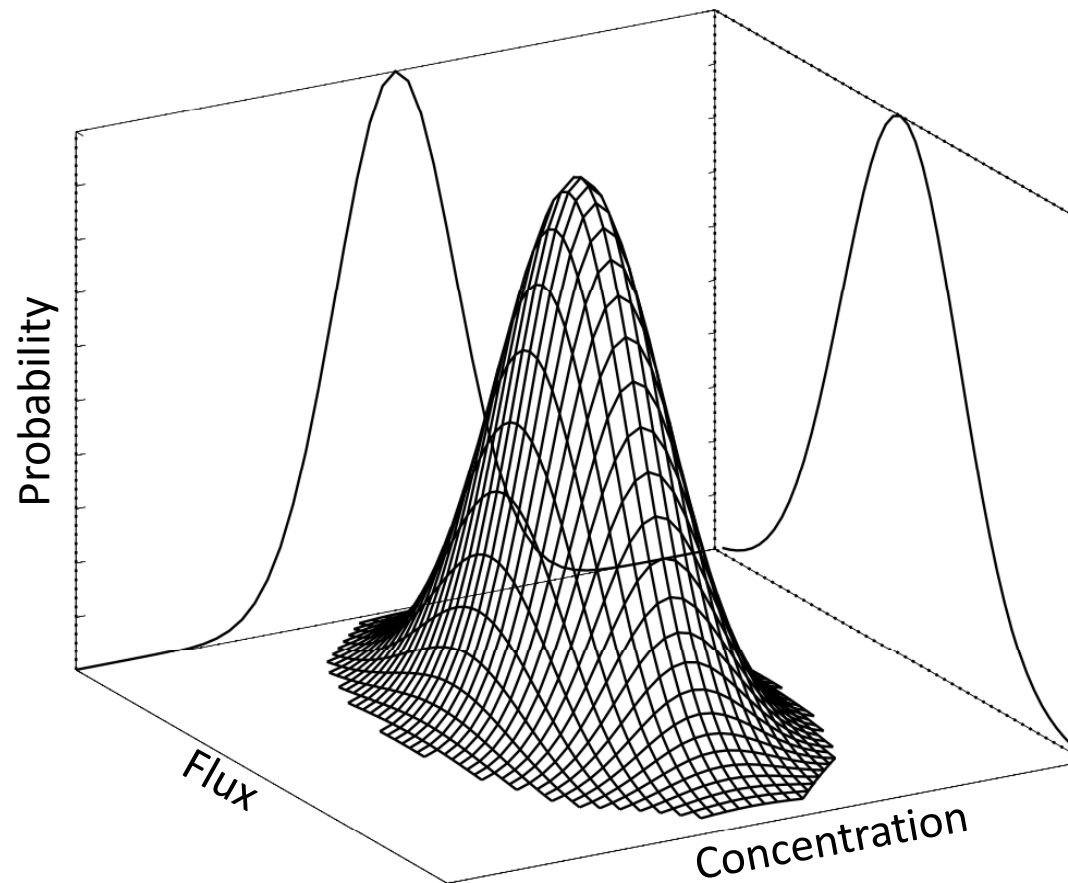


GOSAT

Flux:

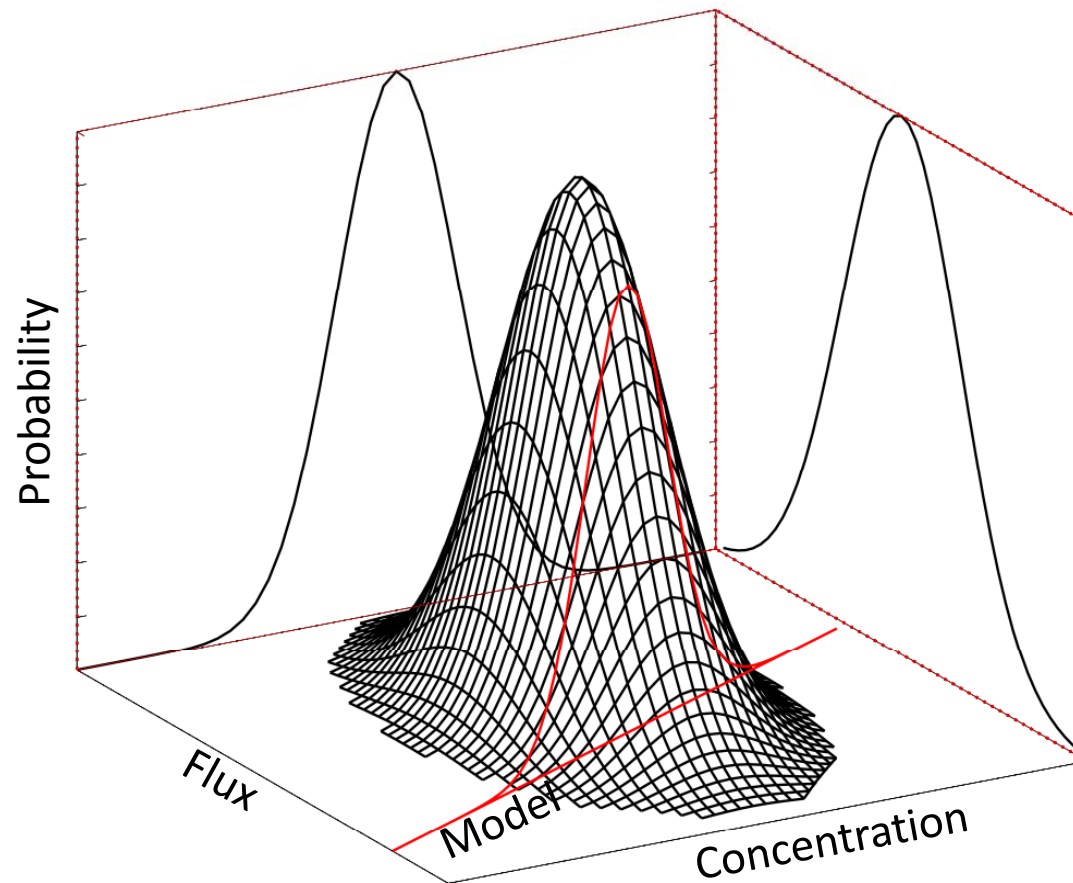


# Optimization problem: A priori

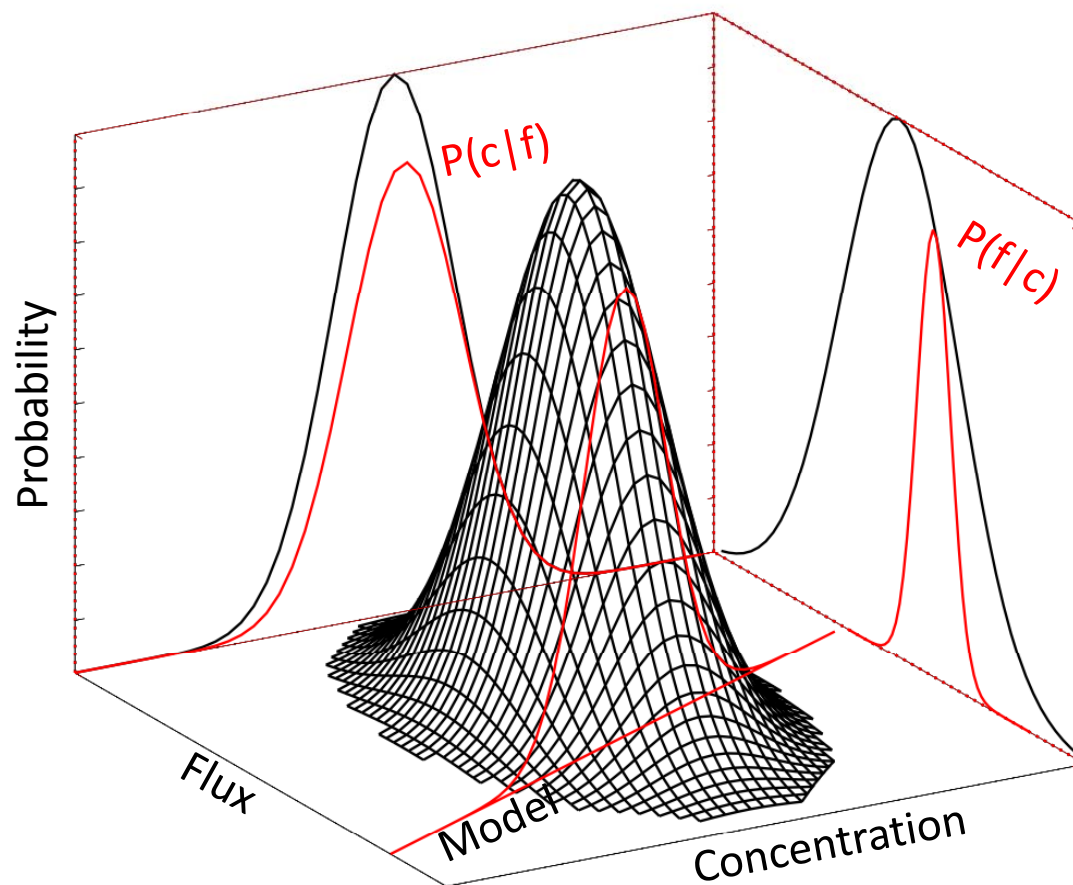




# Optimization problem: + Model

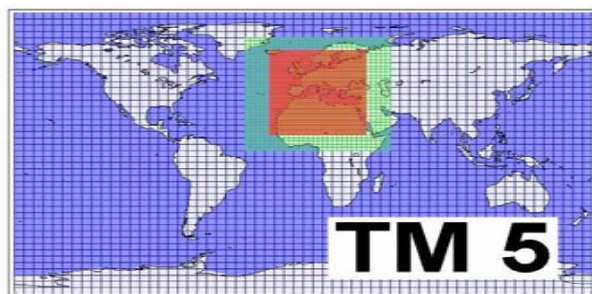


# Optimization problem: Posterior

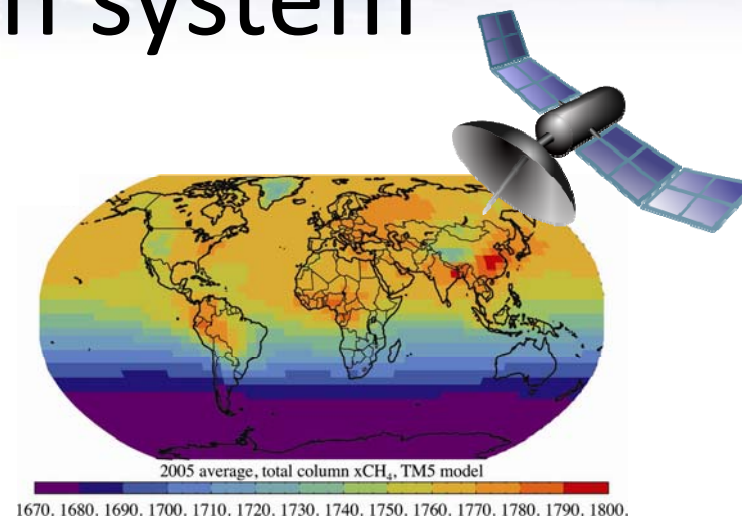
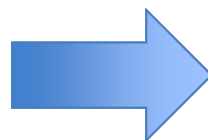


$$S(f) = \frac{1}{2} ((c_{obs} - M(f))^t C_c^{-1} (c_{obs} - M(f)) + (f - f_{pri})^t C_f^{-1} (f - f_{pri}))$$

# Data assimilation system



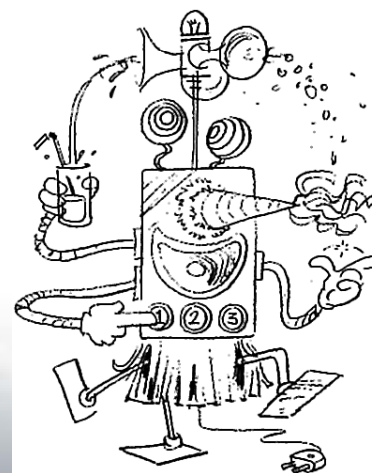
Transport model



Model ⇌ measurements



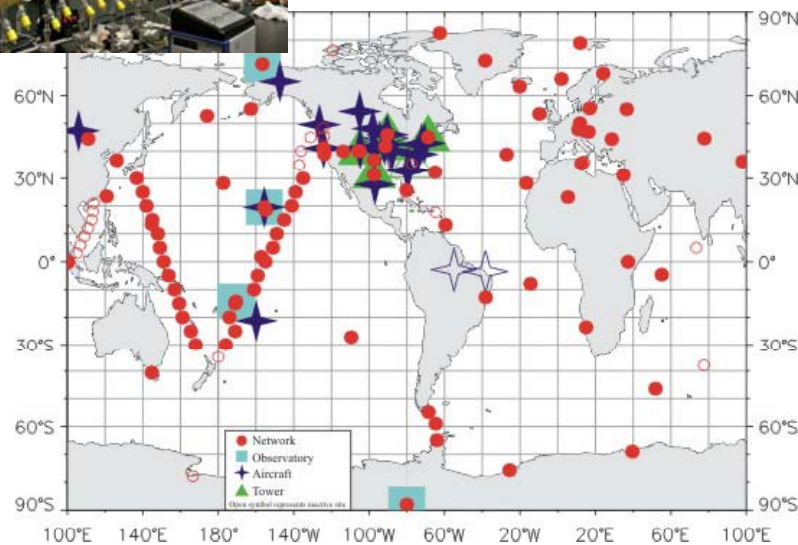
Emissions



Optimization algorithm



# CO<sub>2</sub> inversion: Flask data

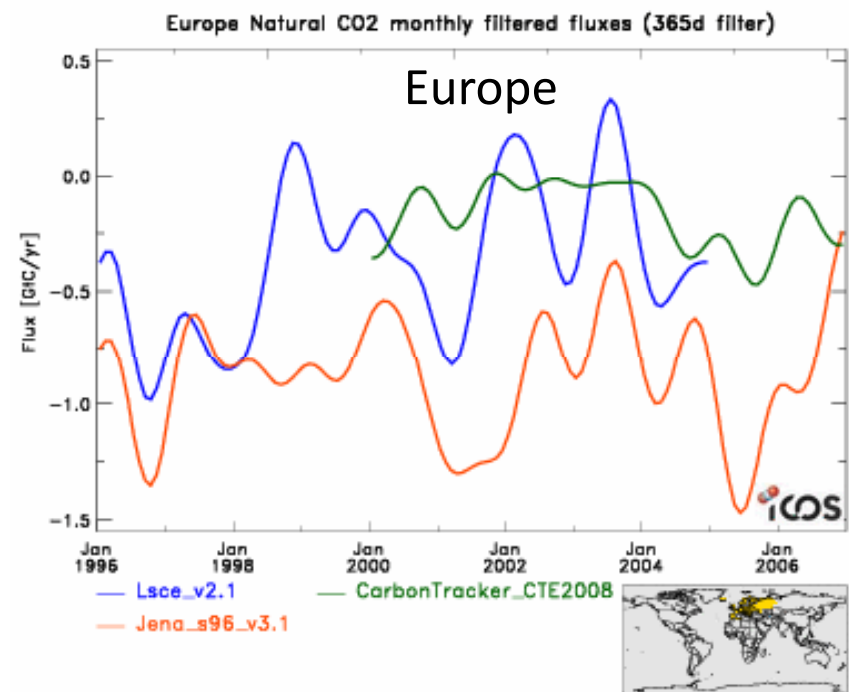
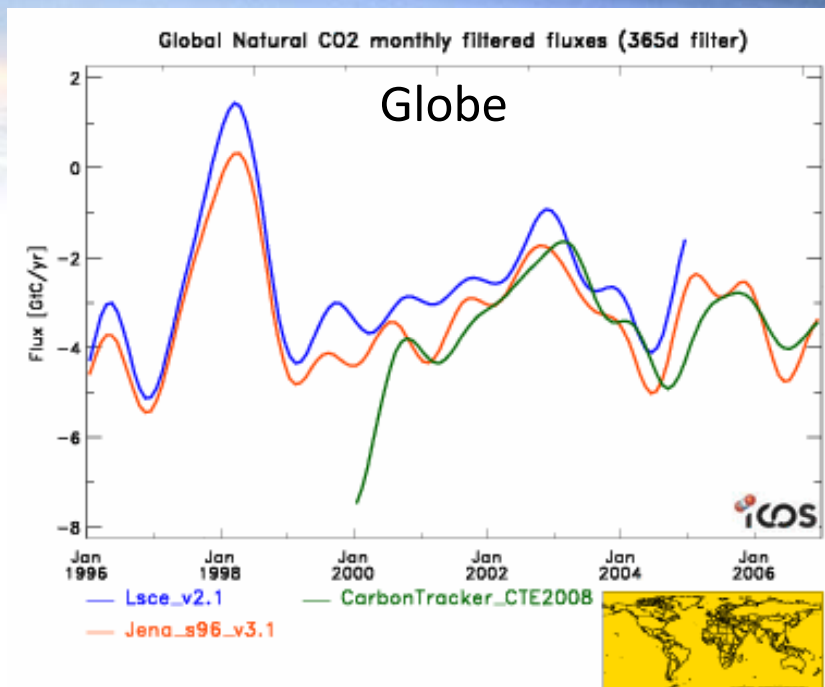


NOAA ESRL flask sampling network

<http://www.carboscope.eu>



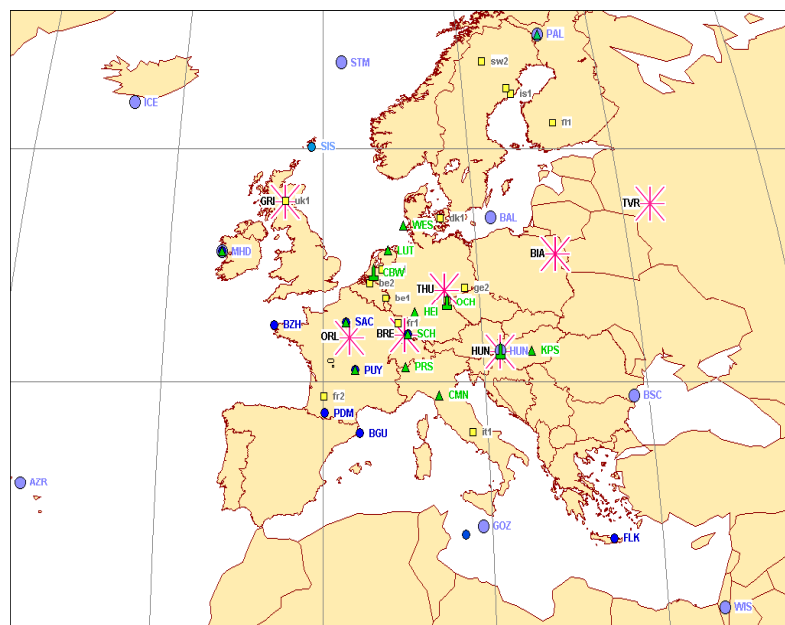
23-3-2010, IPCC TFI uncertainties meeting



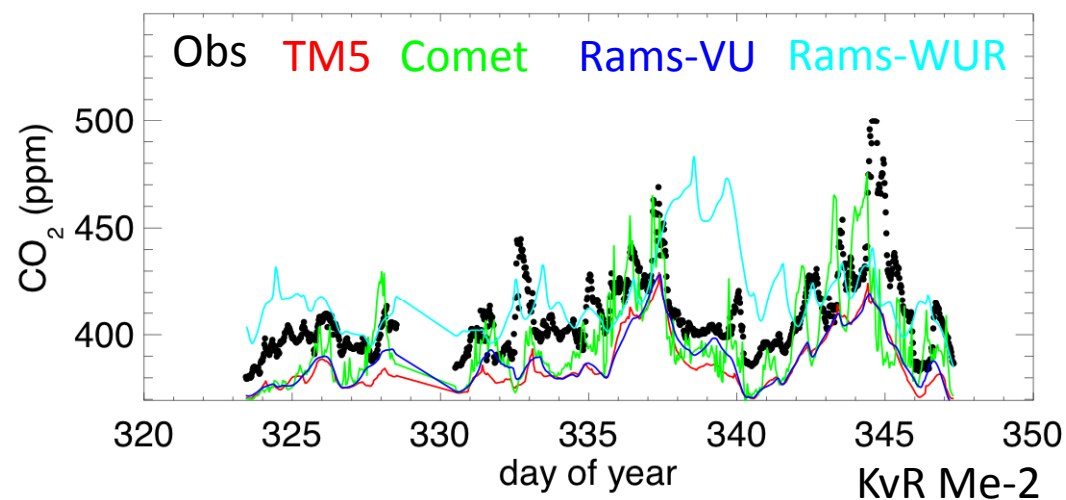
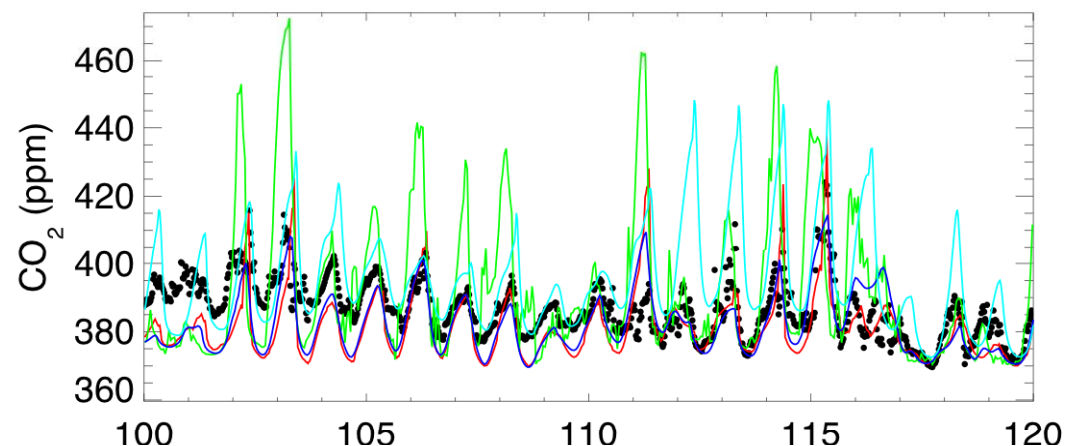


# Solution 1: Dense regional networks

European GHG measurement network



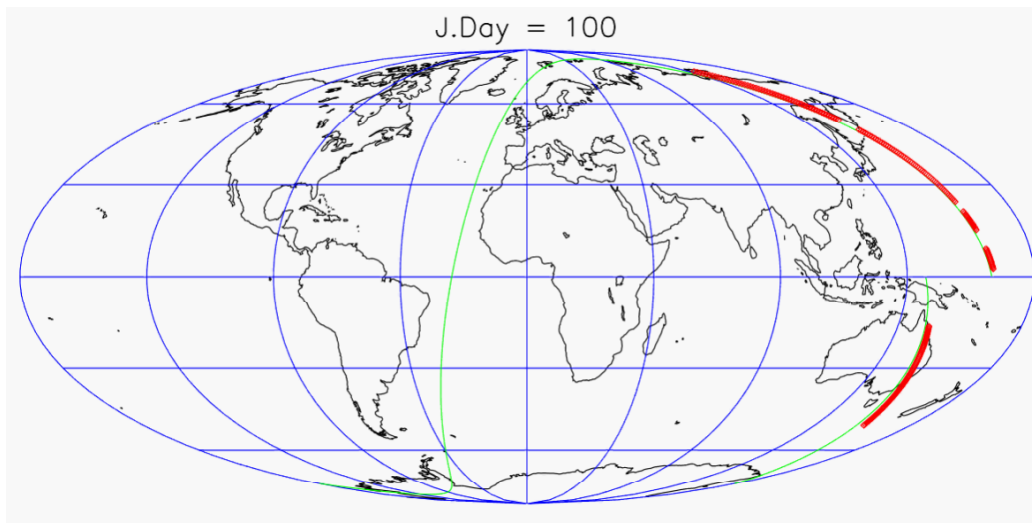
Cabauw tower 120m, NL



Peters et al. (GCB, 2010): NEE for Europe, 2001-2007 =  $-165 \pm 450 \text{ Tg C yr}^{-1}$

# Solution 2: satellites

Courtesy F-M Breon



**Advantage:** data coverage

**Disadvantage:** challenge to reach the required precision on  $XCO_2$  of  $\sim 1\text{ppm}$

**Important:** Vertical sensitivity varies Between various techniques

Thermal IR

ShortWave IR

IASI

TES

AIRS

GOSAT

Sciamachy

OCO-2?

GOSAT-2?

Active sensor?

2000

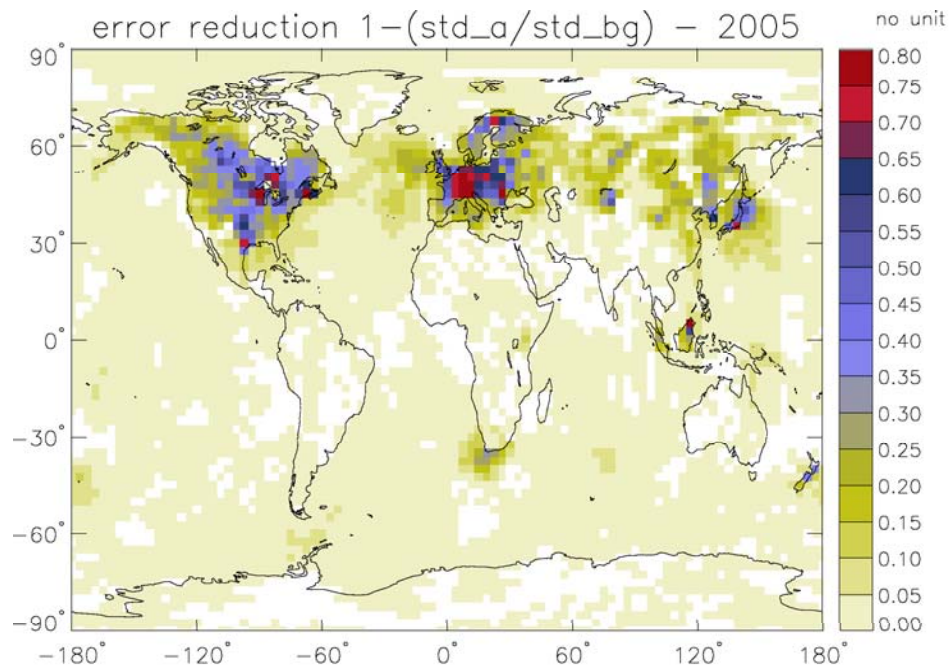
2005

2010

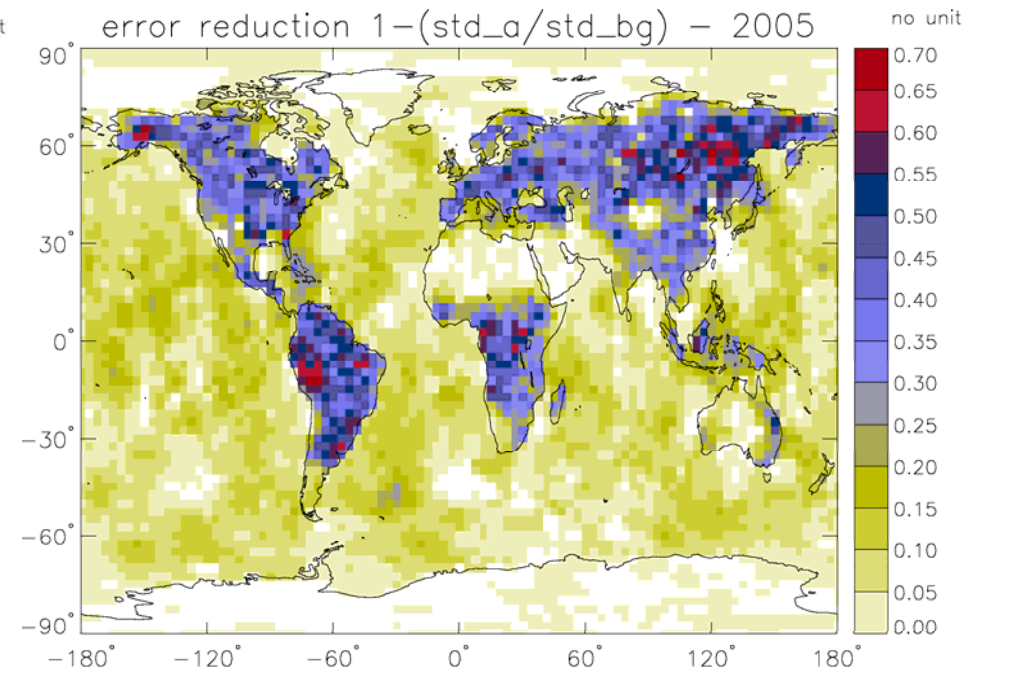
future

# Performance simulation

Surface network



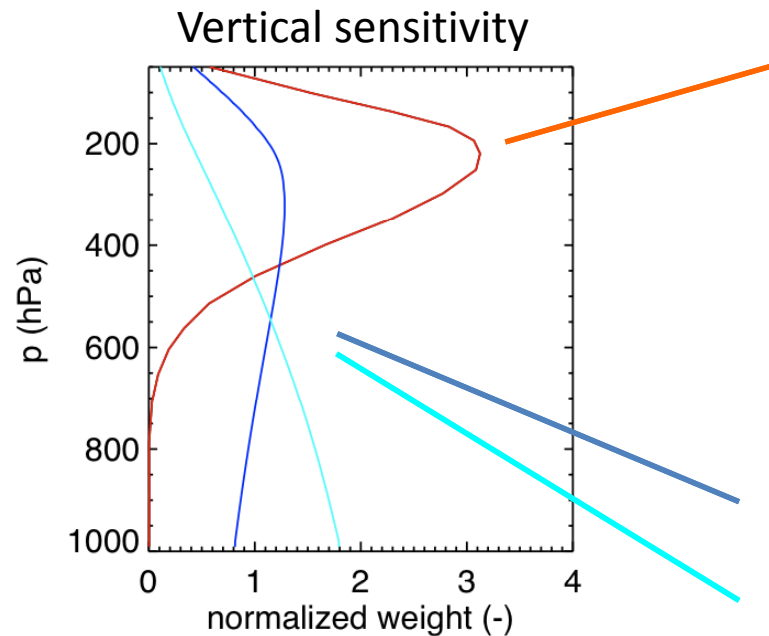
OCO



Courtesy Katja Hungershoefer  
ESA A-SCOPE study

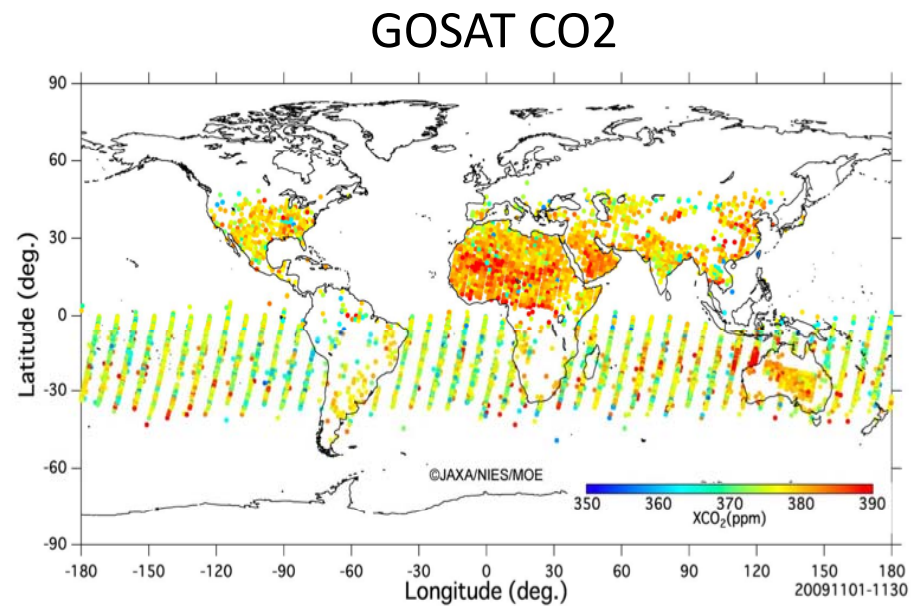
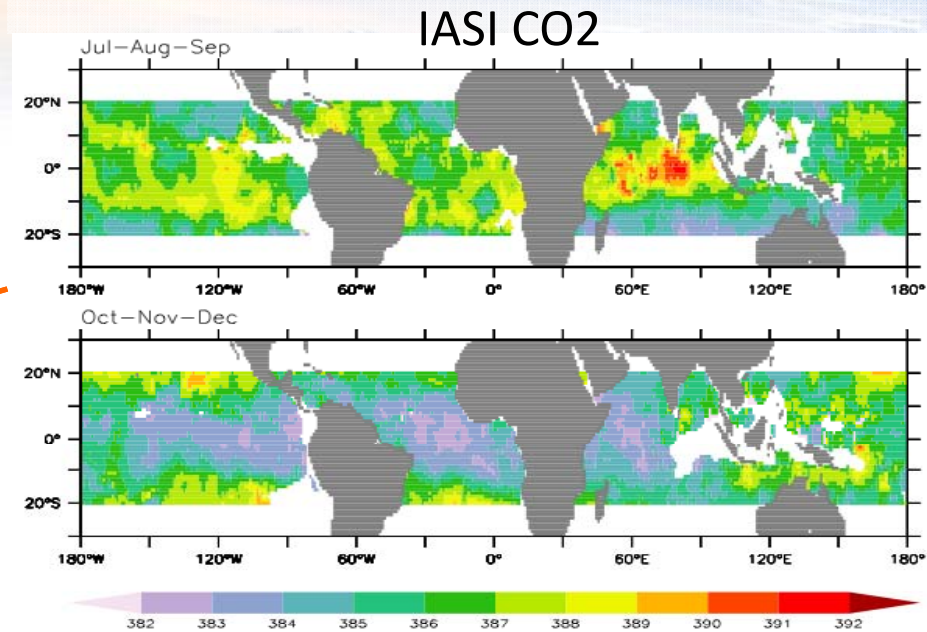


# Status: 1 Obs.



IASI: more accurate

GOSAT: better sensitivity & performance will improve



Crevoisier et al., ACP 2009

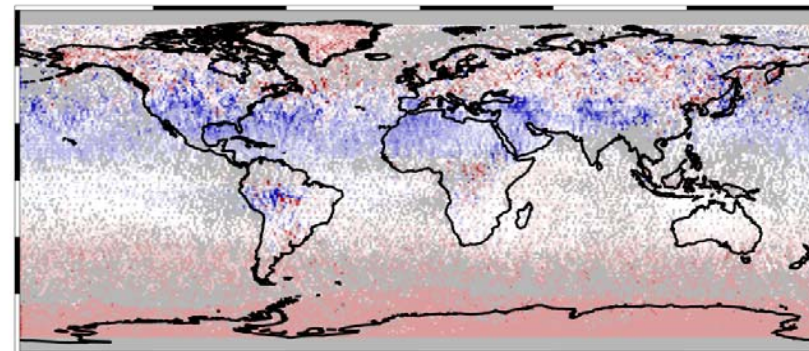
Courtesy Watanabe, 2010



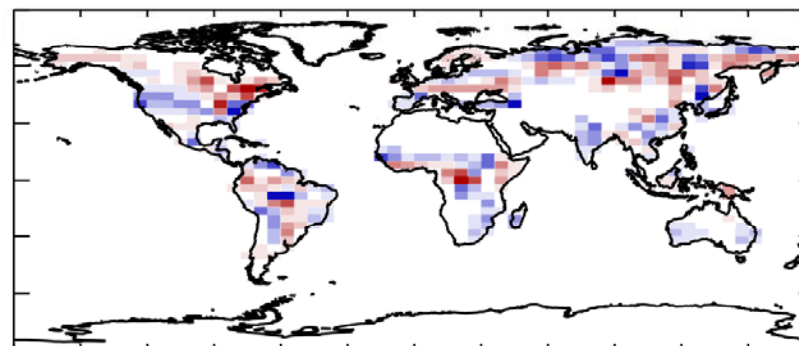
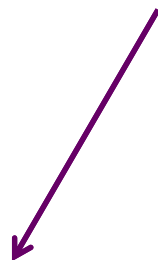
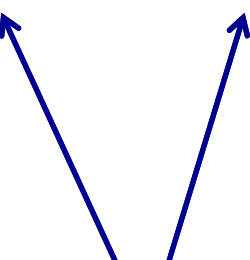
# Status: 2. Models

LMDZ – TM5, August 2005

Simulated satellite samples



-1.5 -1.3 -1.0 -0.8 -0.6 -0.3 -0.1 0.1 0.3 0.6 0.8 1.0 1.3 1.5 ppm



-17.3 -14.7 -12.0 -9.3 -6.7 -4.0 -1.3 1.3 4.0 6.7 9.3 12.0 14.7 17.3 20.0 Tg/grid

Results translate into  
sizable annual flux uncertainties ...

ESA A-SCOPE study

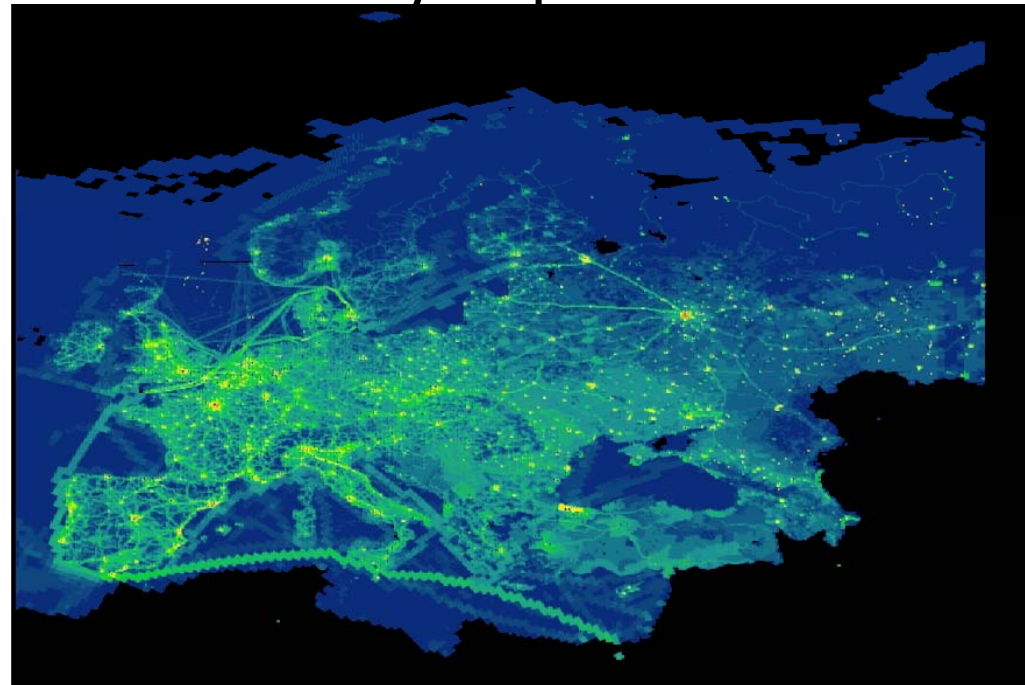
# Conclusions

- **In-situ**: Tall tower networks improve inversion performance on regional scales
- **Satellites**: Could extend such a performance globally.
- **Hurdles**: Satellite data inversions challenge measurement techniques and models.

# Can we estimate emissions from in-situ and satellite data in combination with modeling?

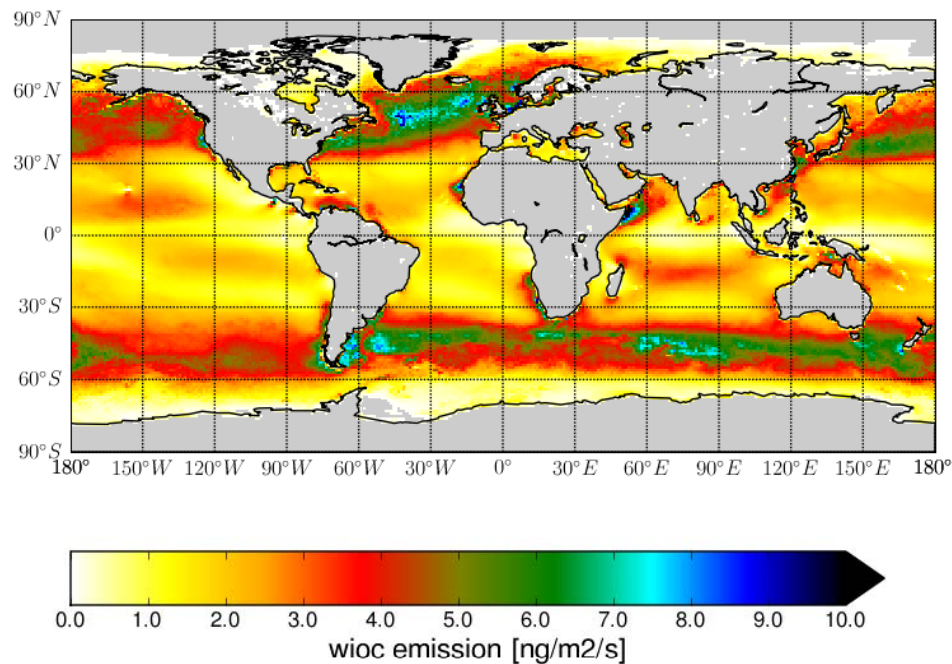
- Main interest is in annual total emissions over an entity (country)
- For modelling applications spatial distribution and timing of emissions are very important

How can  
satellite data  
help to improve  
these maps?

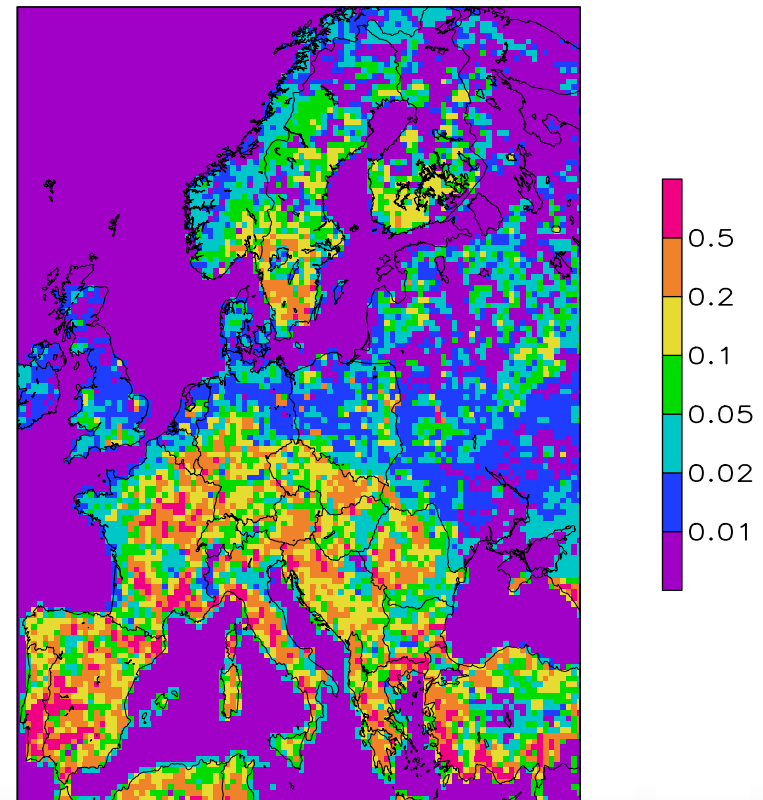


# Directly as input: LAI, FRP, ocean colour, whitecap, etc

Marine emissions



Isoprene

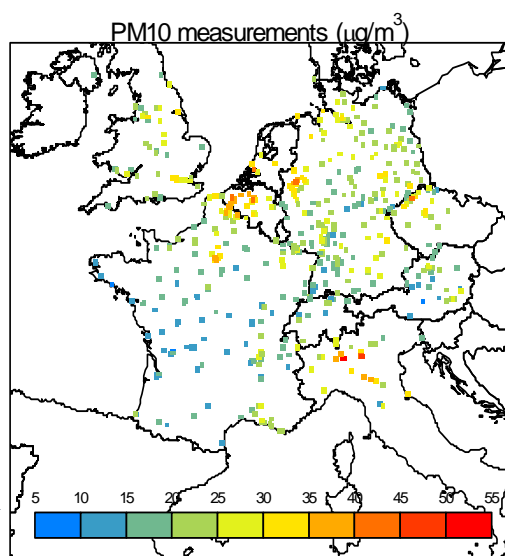


Natural & Biogenic emissions – calculated online in models

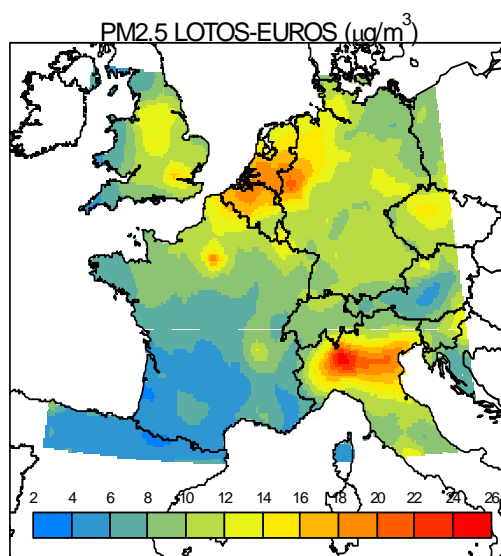


# Indirectly via data assimilation: a challenge!

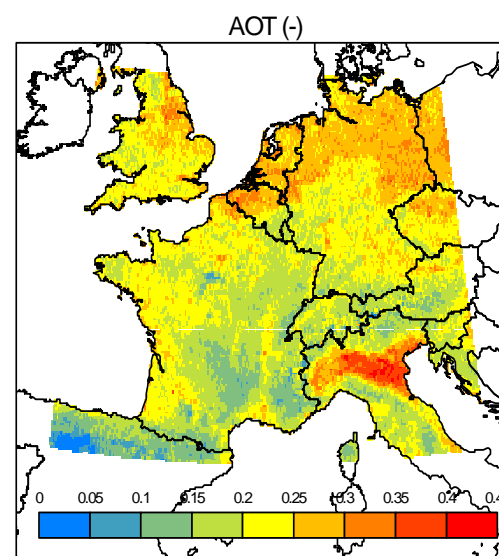
- To disentangle the uncertainty due to the emission input from other model uncertainties
  - The assimilation “blames” all errors to a limited amount of parameters
- To keep the system realistic and balanced
- To combine different sources of data – multi component



Ground based



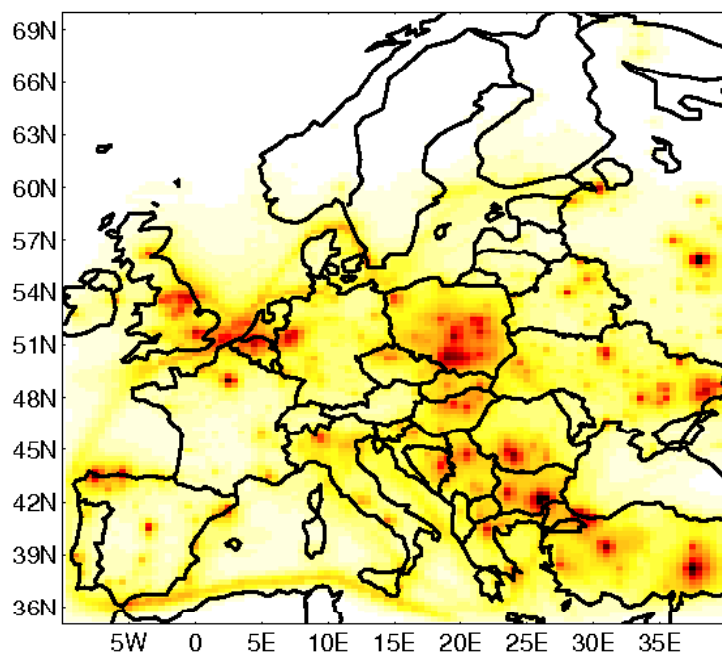
Model



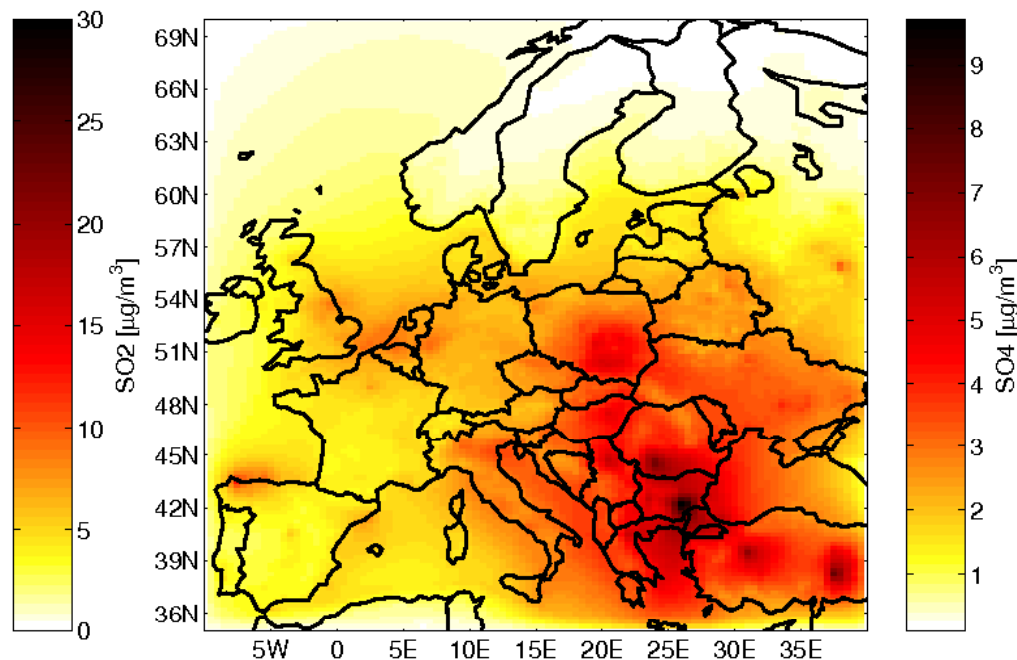
Satellite

# Assimilation of SO<sub>2</sub> and SO<sub>4</sub> – a case study

SO<sub>2</sub>

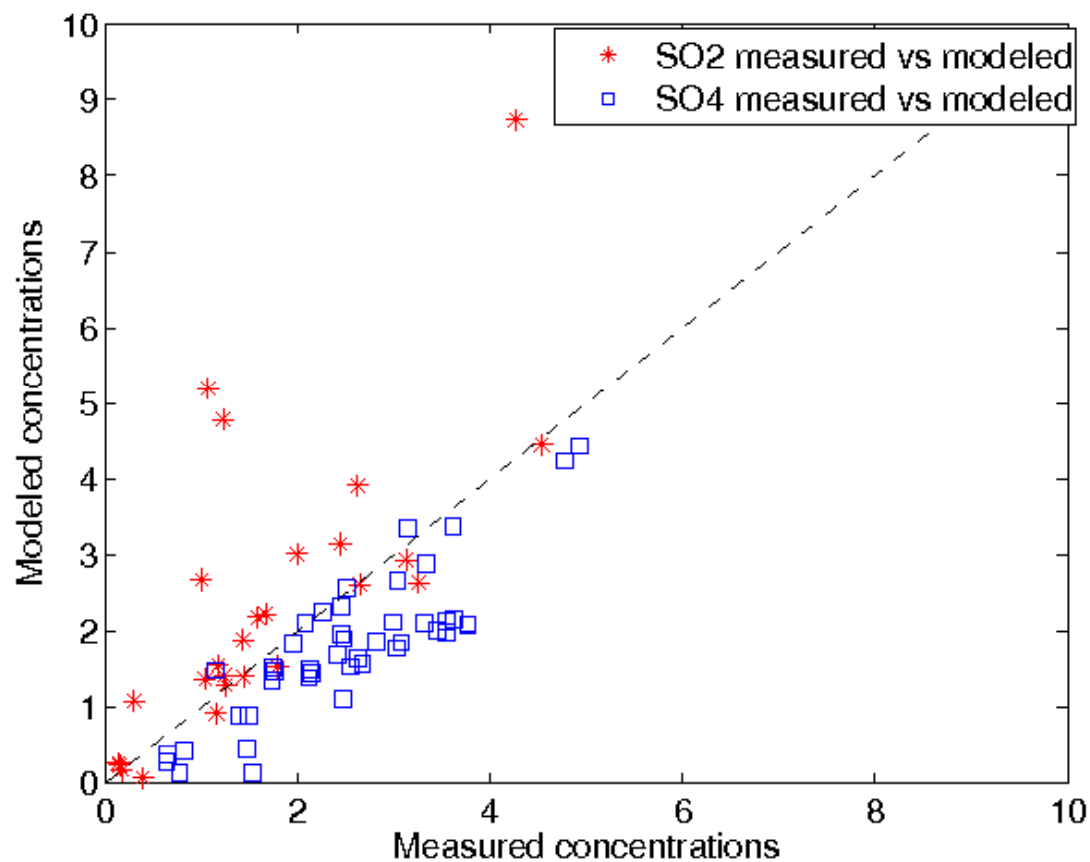


SO<sub>4</sub>



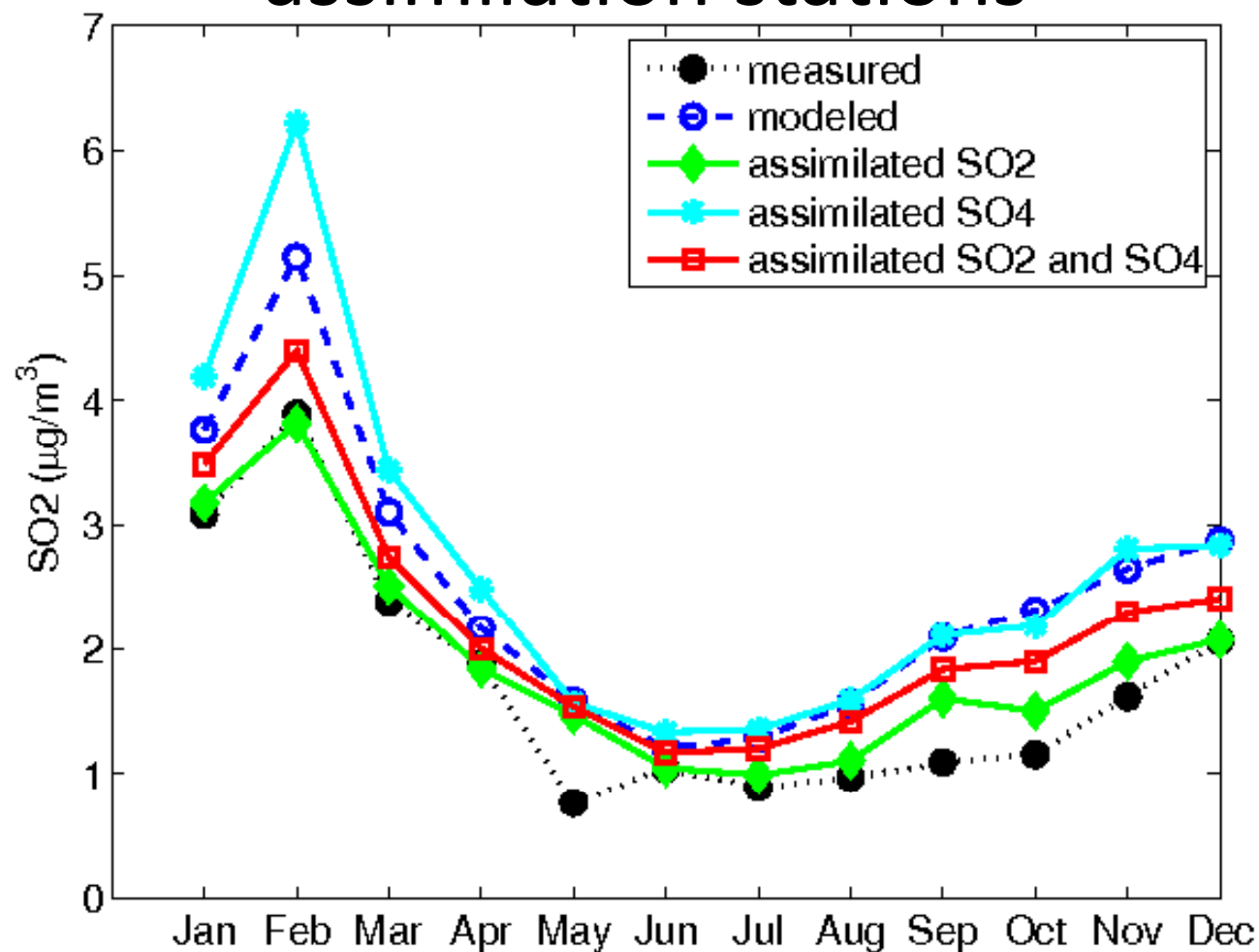
Annual mean for 2003

# Modelled annual mean concentrations SO<sub>2</sub> and SO<sub>4</sub>



	SO2	SO4
OBS	1.6	2.5
MOD	2.3	1.8
OBS/ MOD	1.4	0.7

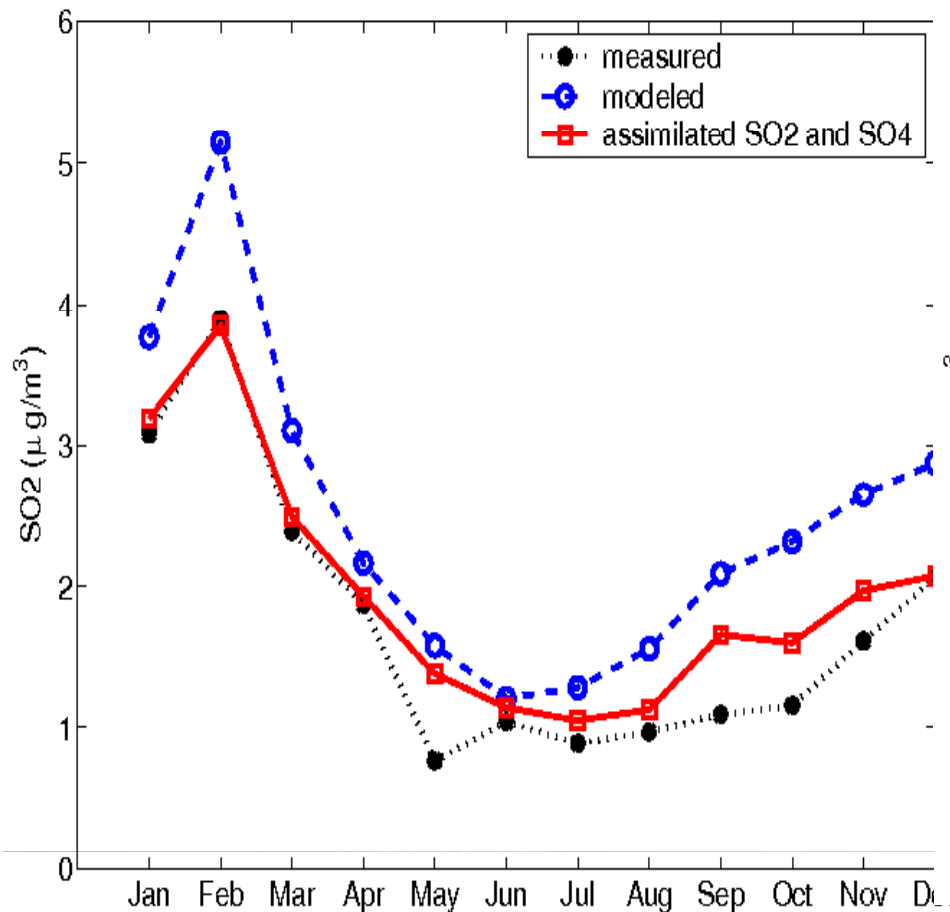
# Results: SO<sub>2</sub> annual cycle over all assimilation stations



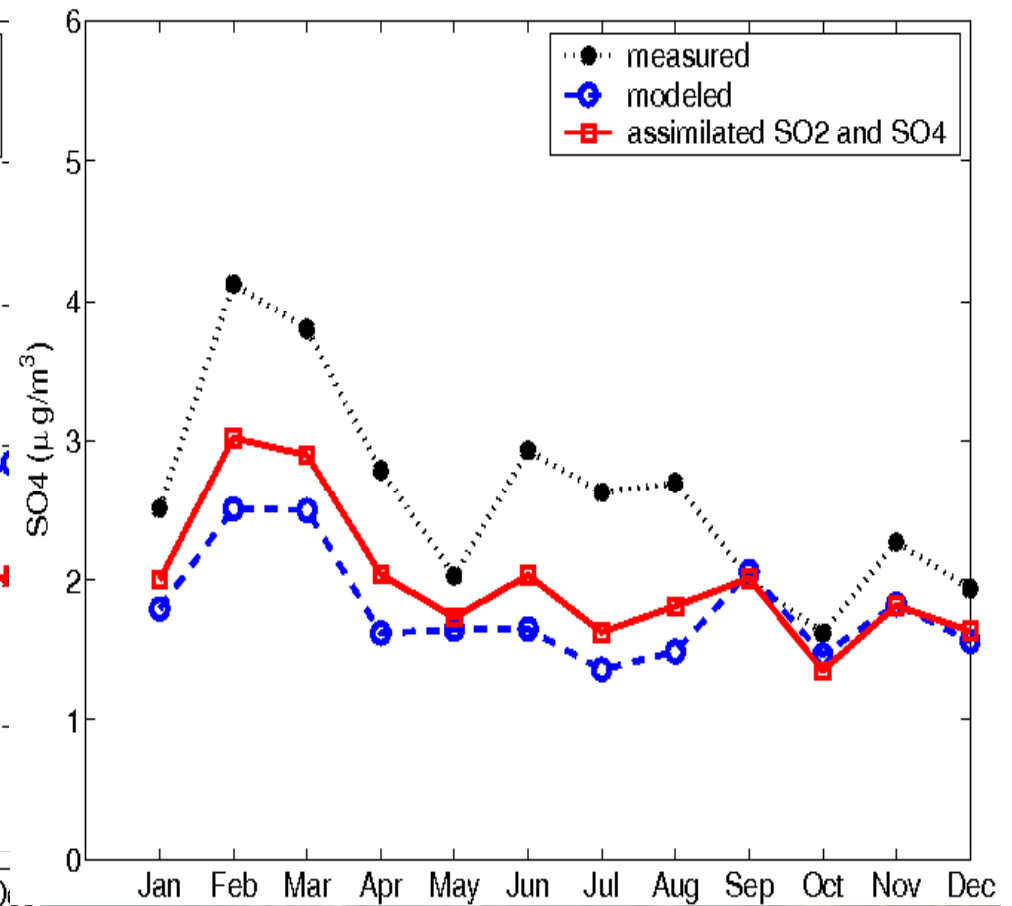


# Results: SO<sub>2</sub> & SO<sub>4</sub> annual cycle over all stations by including uncertain conversion rates

SO<sub>2</sub>

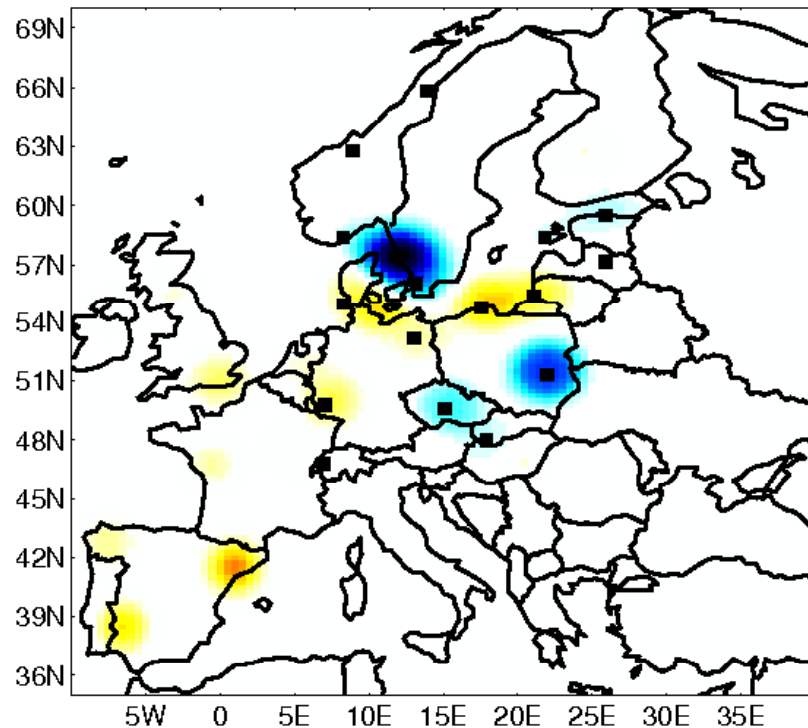


SO<sub>4</sub>

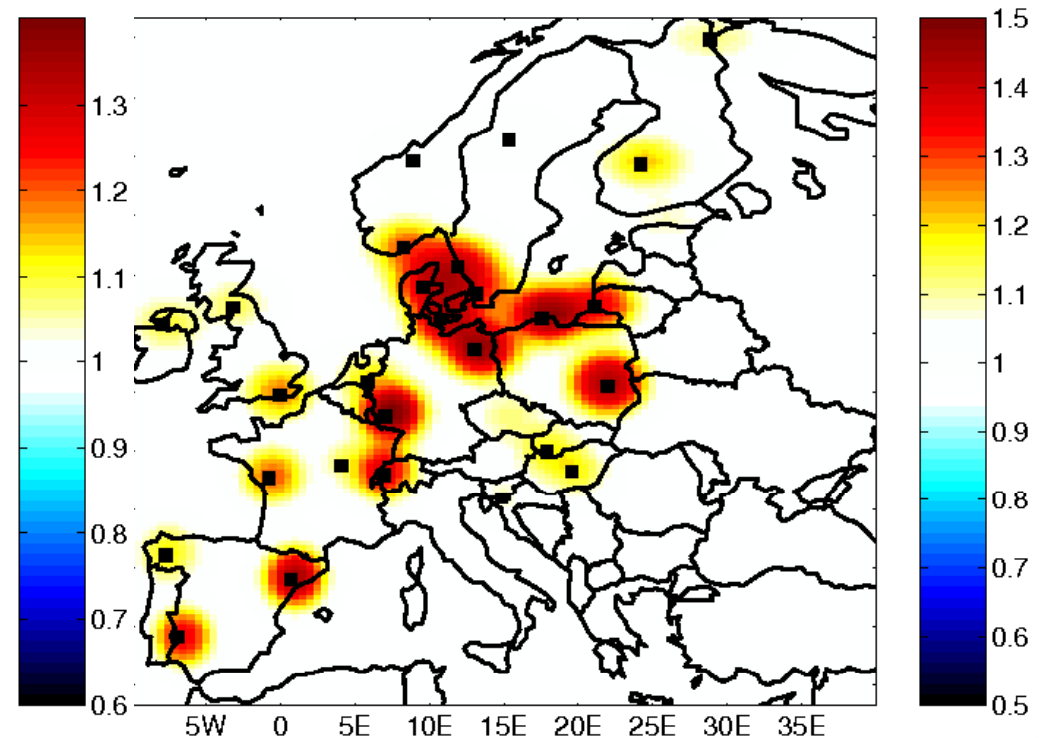


# Annual mean estimated multiplication factors

Emissions



Reaction rate



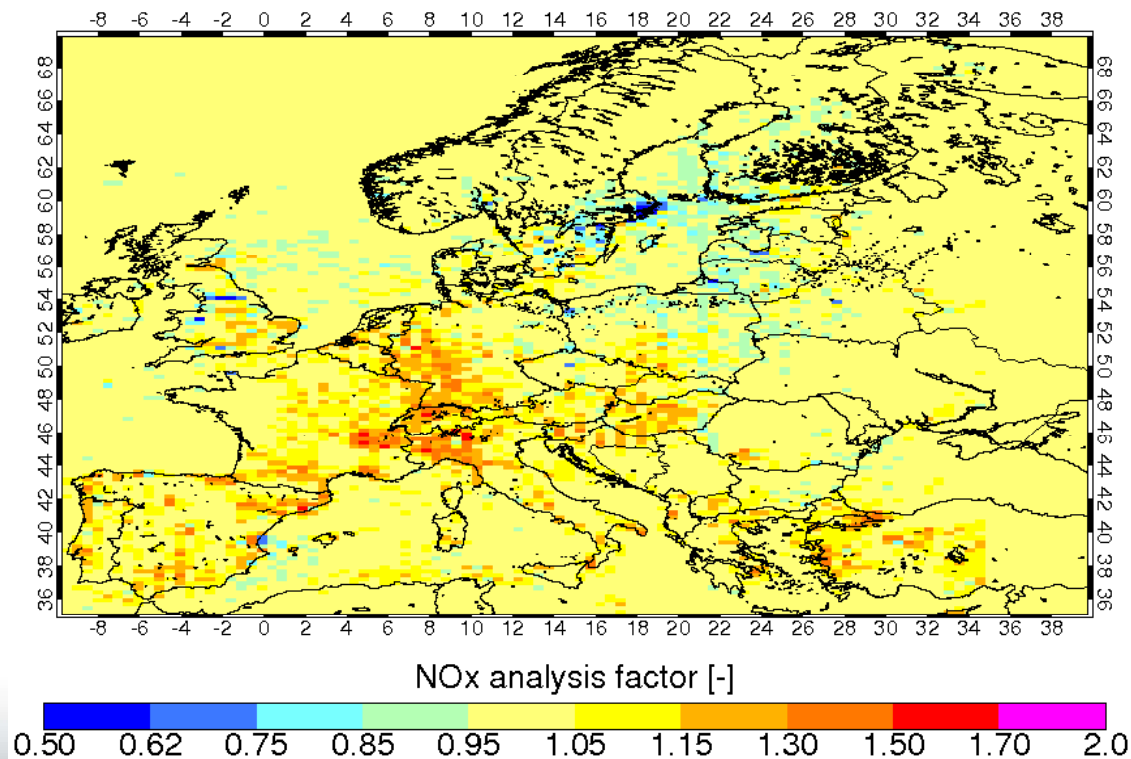
Also after acknowledging the shortcomings of the model it indicates that the shipping emissions and those in Poland may be too high

# Assimilation OMI NO2 measurements with LOTOS-EUROS

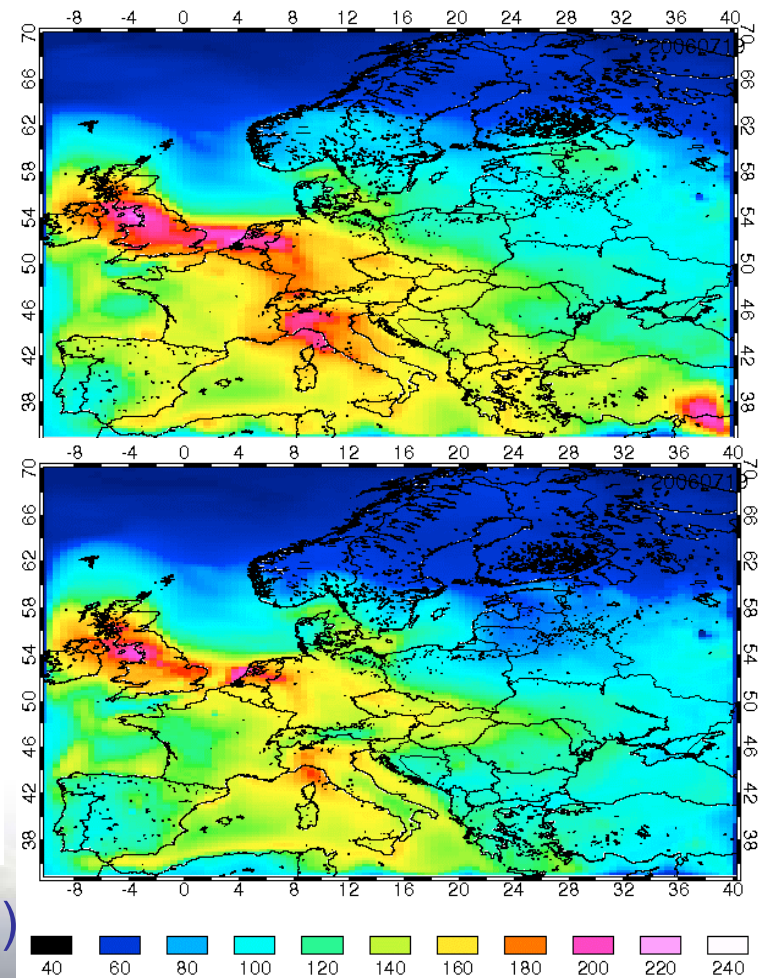
## Impact for ozone at the surface

Lotos-Euros, date 20060717

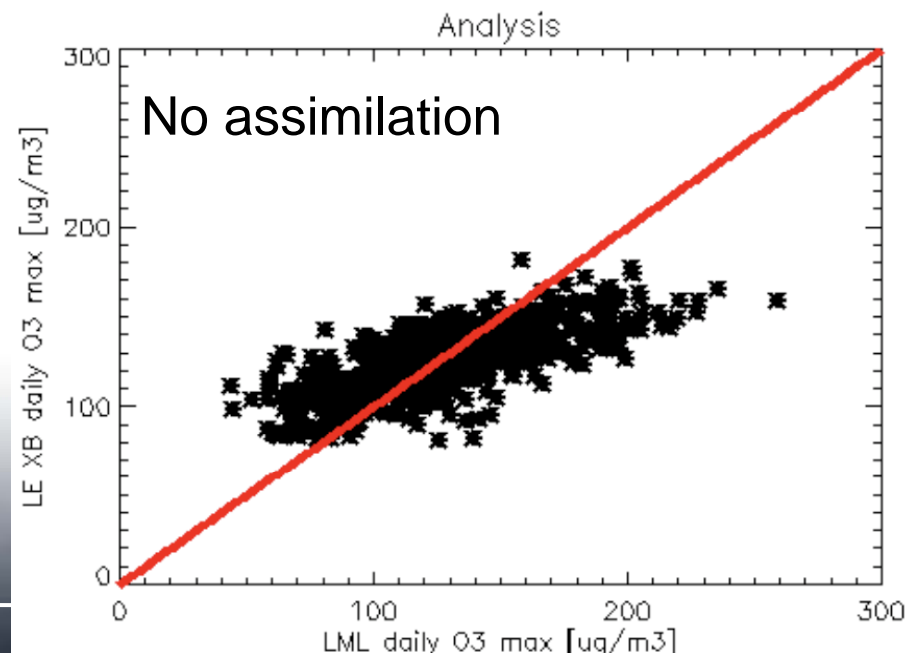
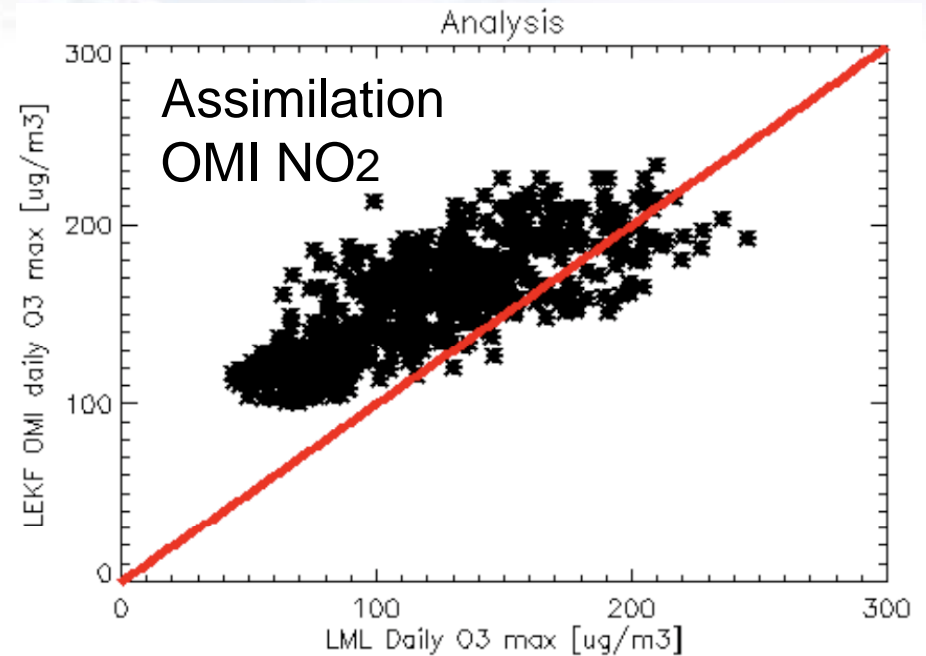
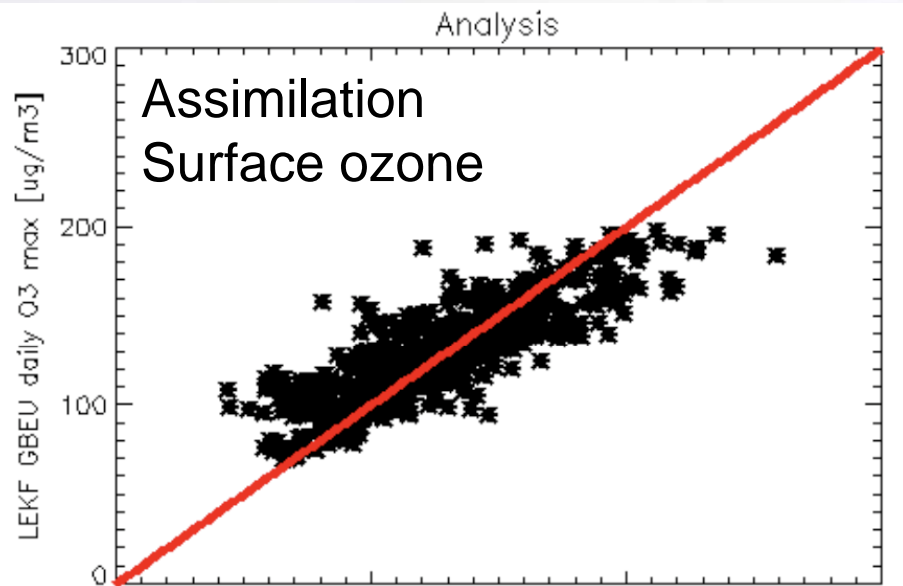
KNMI/RIVM/TNO



Analysis NOx emissions / inventory (yellow=1)



# Impact of assimilation on ozone peak value



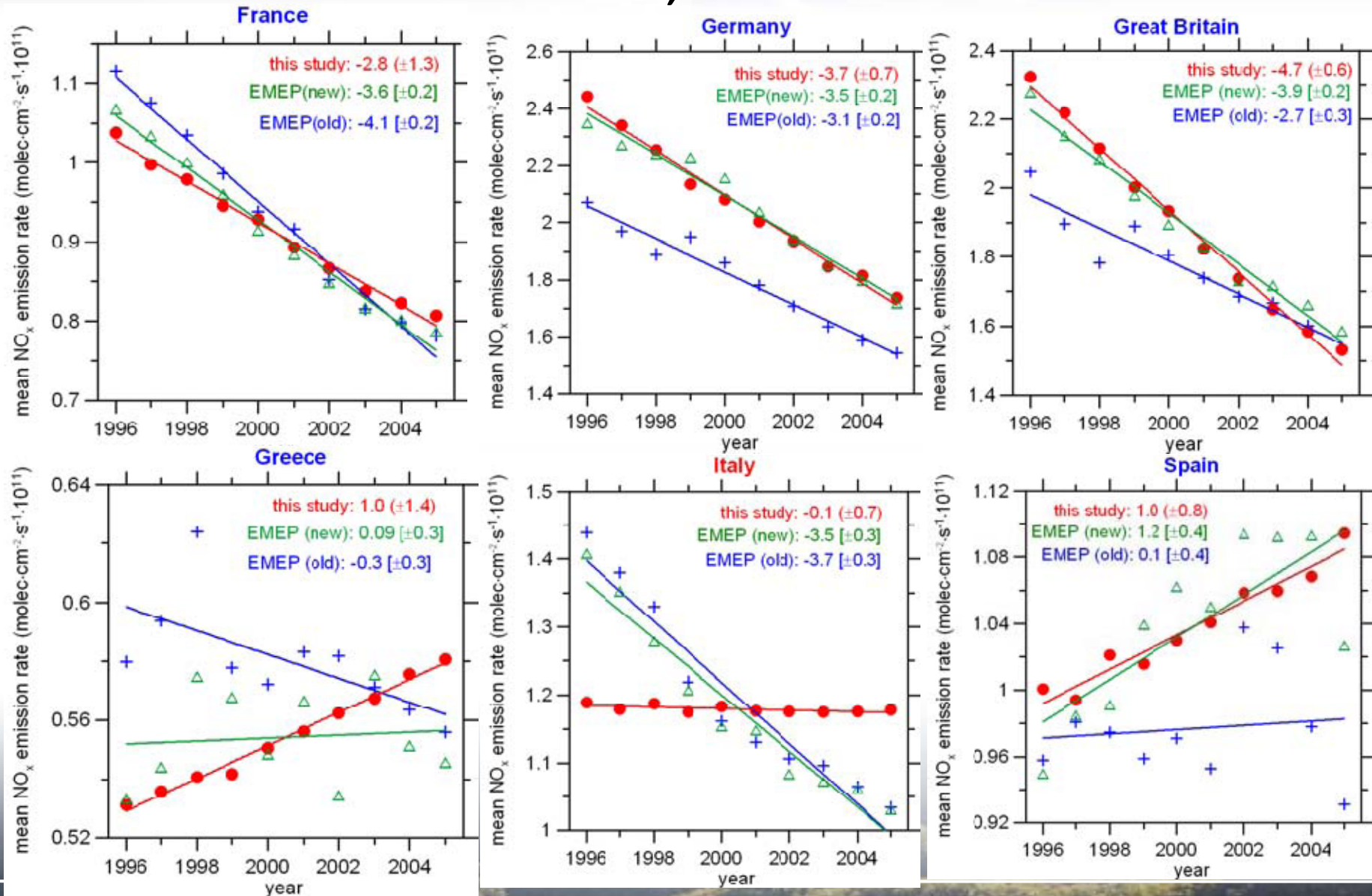
Is the NO<sub>x</sub> lifetime OK?

NO<sub>2</sub> bias in the model effects  
ozone negatively

Note, OMI NO<sub>2</sub> may be ~25%  
high



# Estimating NO<sub>x</sub> emission trends: Kononov et al., 2006



# Where can we use our present capabilities to provide information on emissions?

- Search for trends in the parameter estimates?
  - Does the EO data indicate that the emission trend is not as expected?
  - The system does the meteo correction, etc for you.
- To indentify locations of new and significant emission sources
  - The areas with consistently high model-measurement deviations
- To identify time profiles – needed: geostationary data
- Emission estimates
  - Only in hotspot locations, mostly qualitatively
- Direct variables such as land use, LAI, Fire Radiative Power, White cap, etc