

Calculating greenhouse gas emissions at global and regional scales using atmospheric measurements

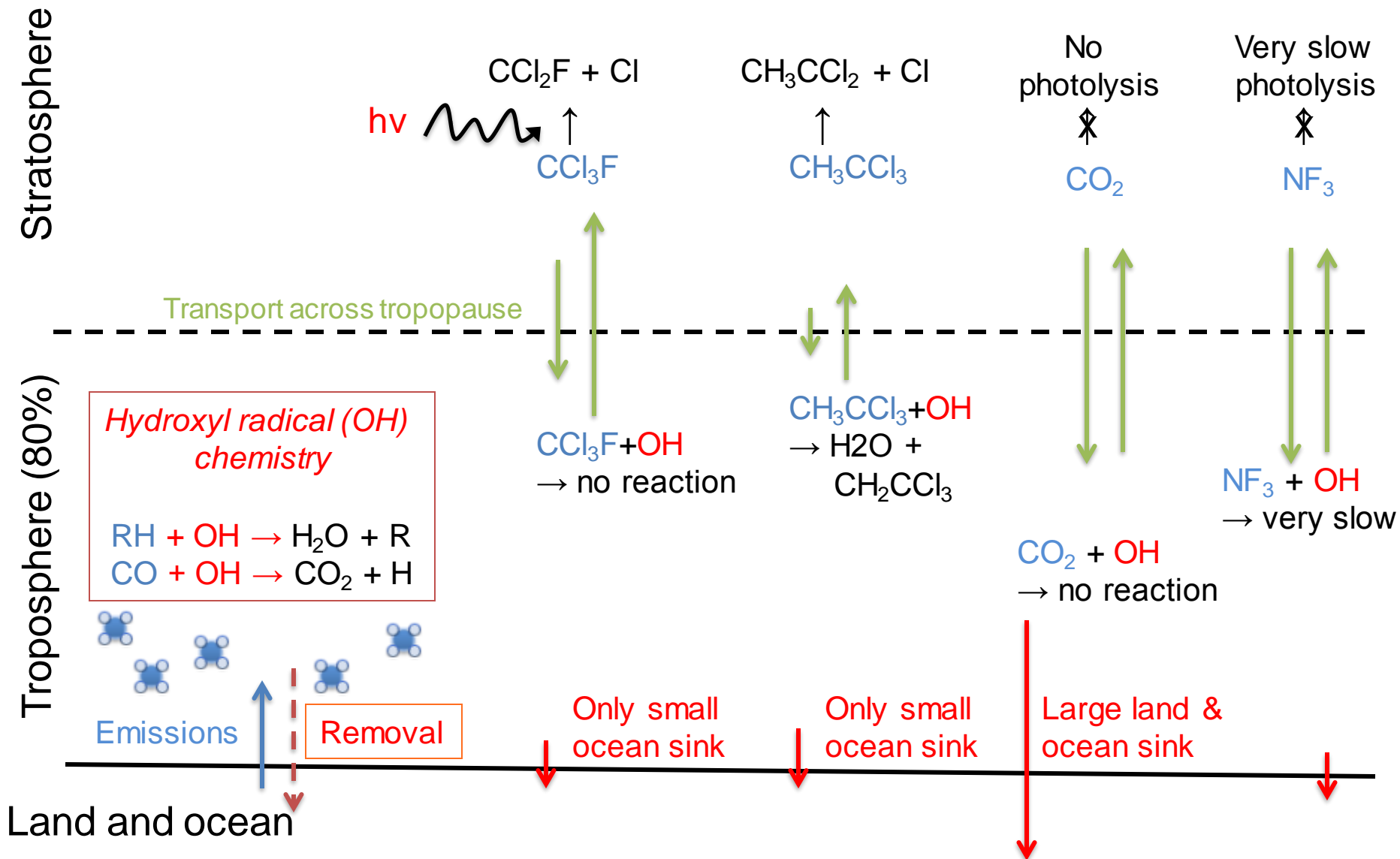
Tim Arnold

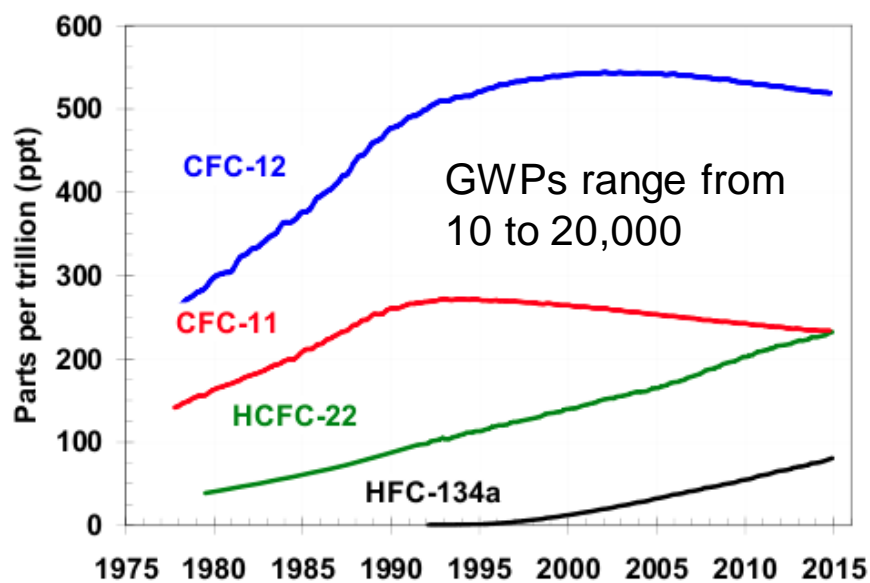
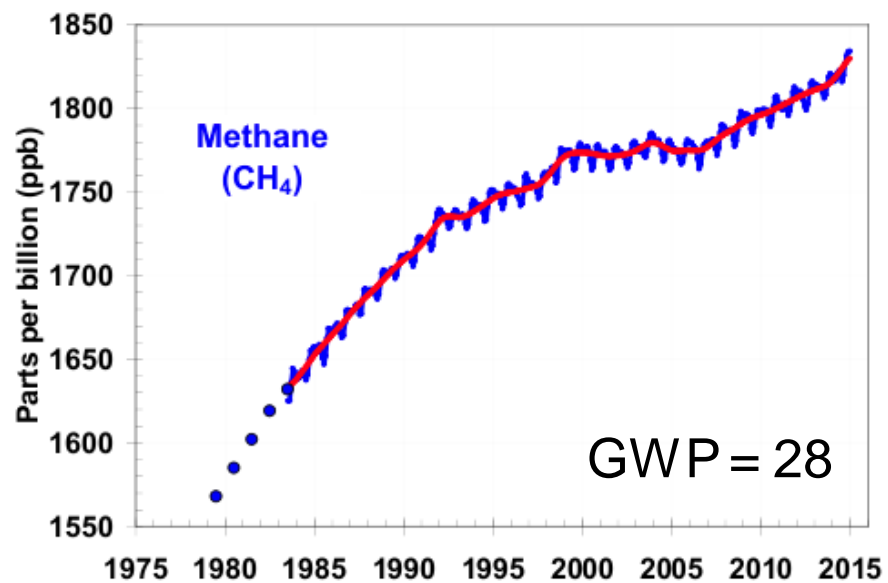
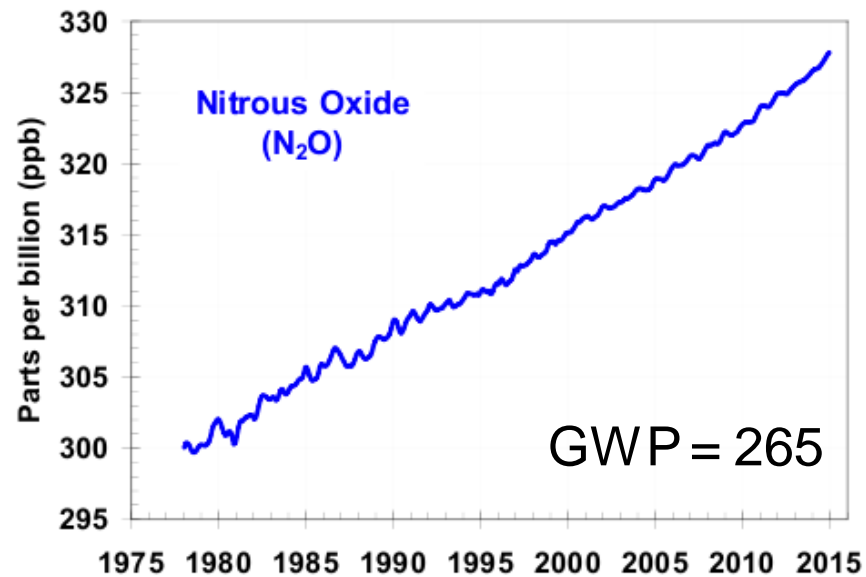
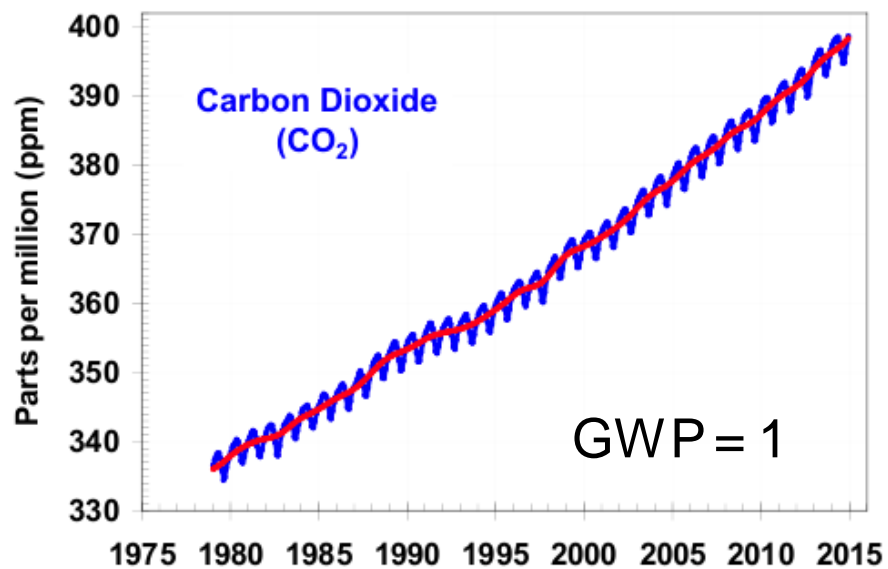
National Physical Laboratory

University of Edinburgh

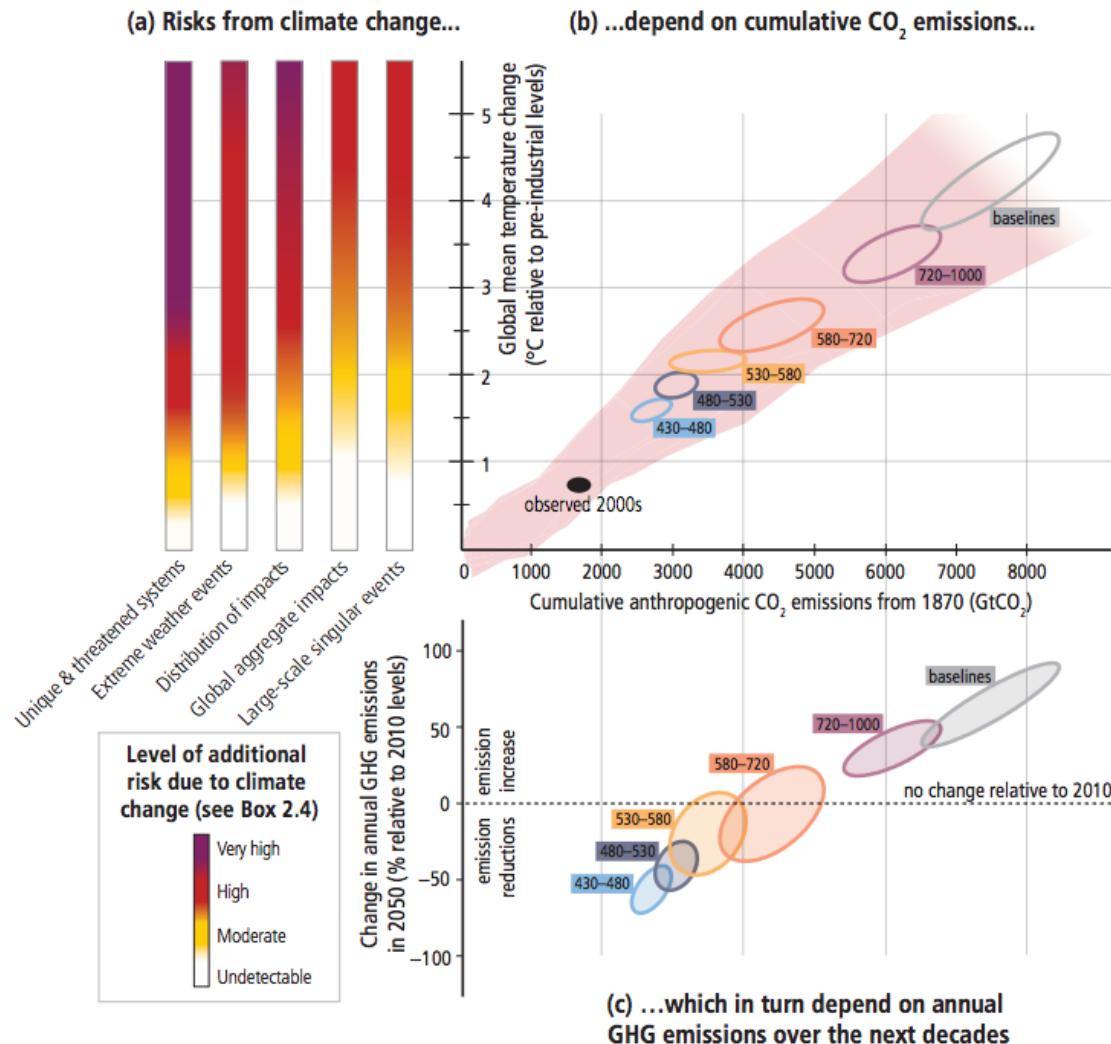
The Middle House, Mayfield
2017 GHG Summer School, 8th August 2017

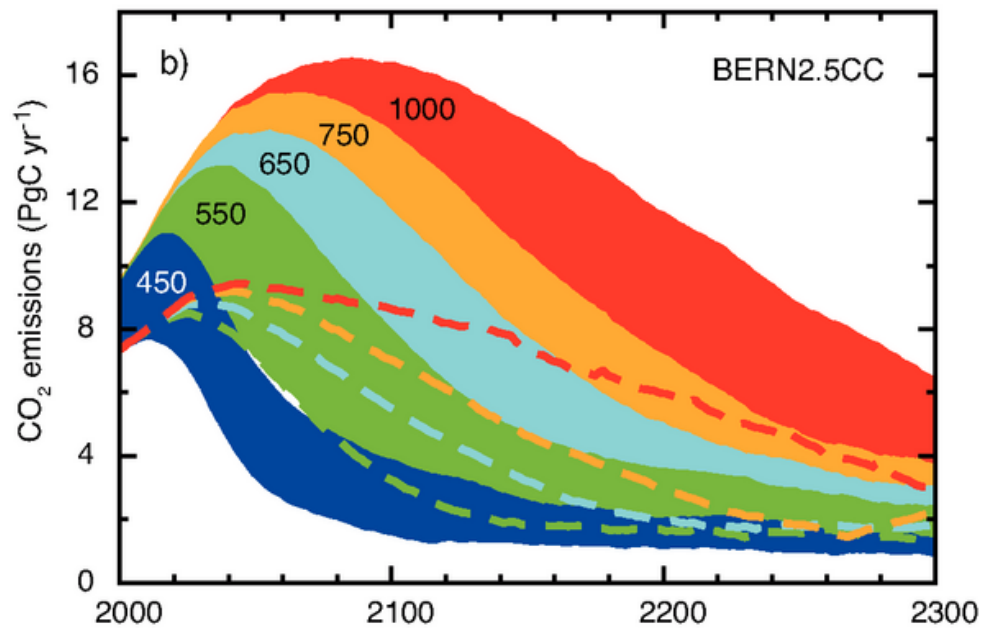
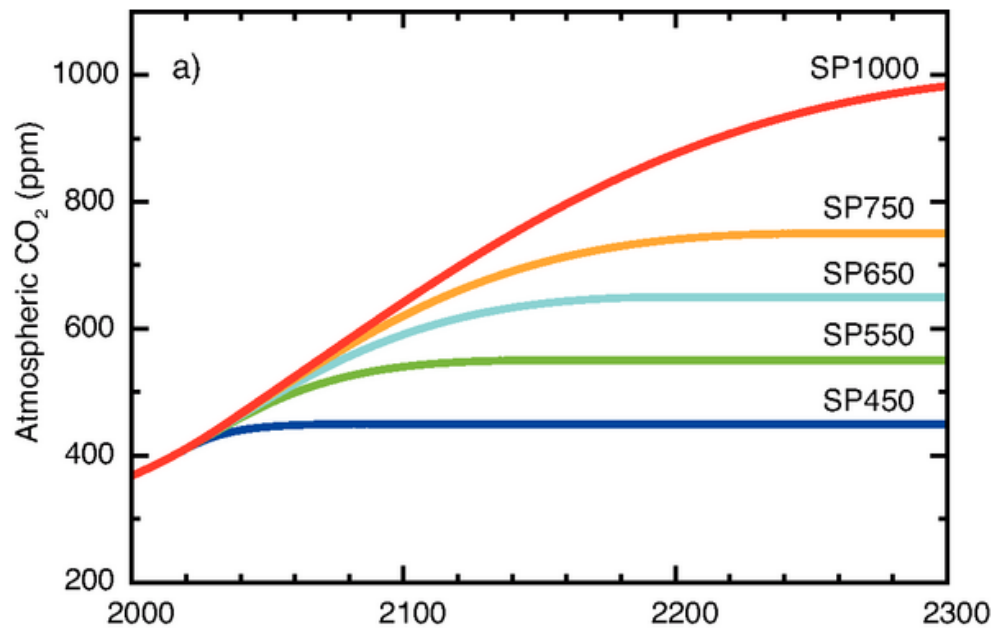
GLOBAL ATMOSPHERIC CHEMISTRY



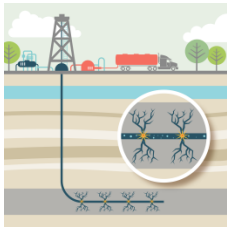


Motivation for immediate and significant emissions reductions





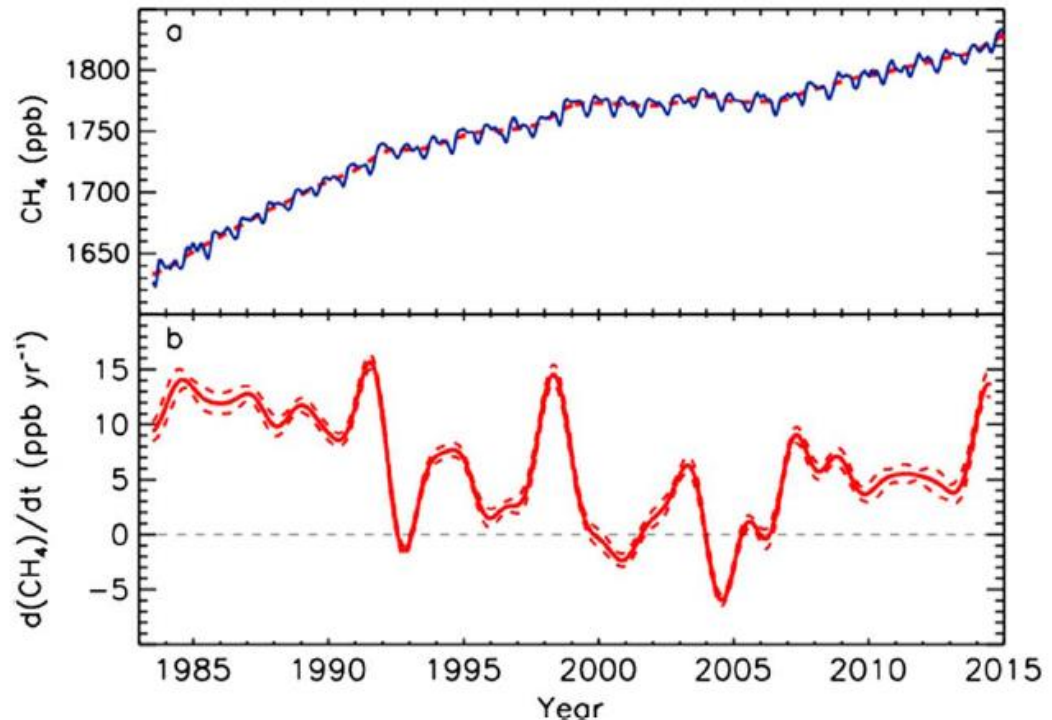
Why atmospheric monitoring can play an increasingly important role



Uncertainty in scaling

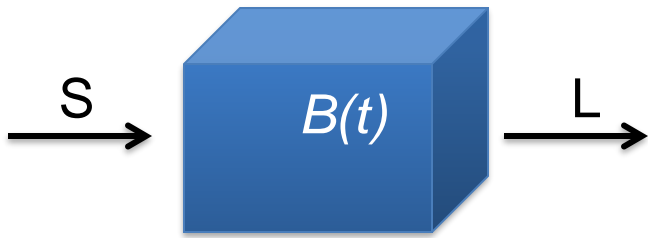


- Provides a reality check – what's actually happening
- It is up to date – to direct timely policy
- Independent



A simple example of emissions verification from atmospheric measurements

1-box model of the
atmosphere



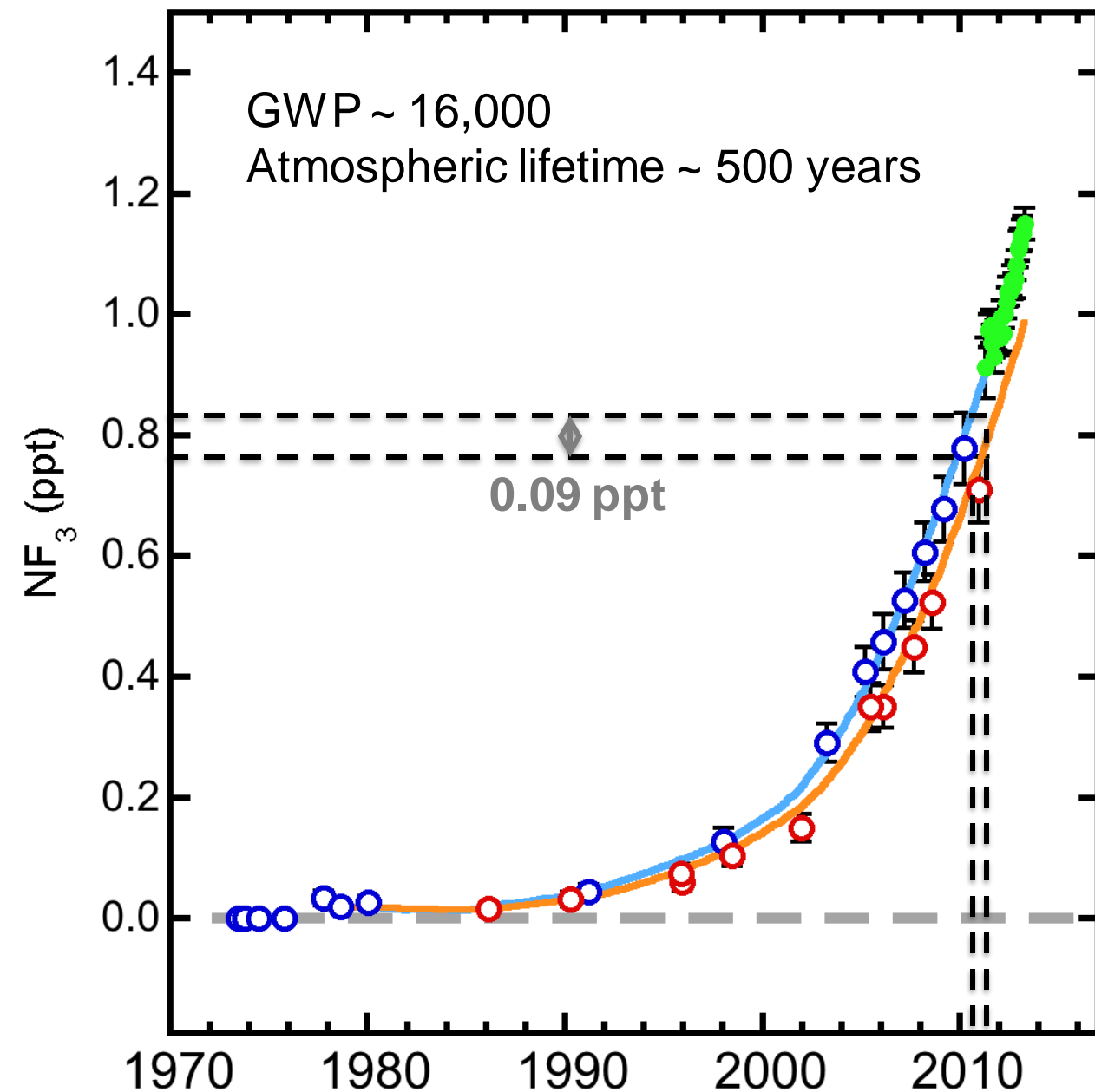
L is the removal rate e.g. Gg / year

S is the source rate

$B(t)$ is the atmospheric burden at time t

$$\frac{dB(t)}{dt} = S - L \qquad L = \frac{B(t)}{t}$$

$$\frac{dB(t)}{dt} = S - \frac{B(t)}{t}$$

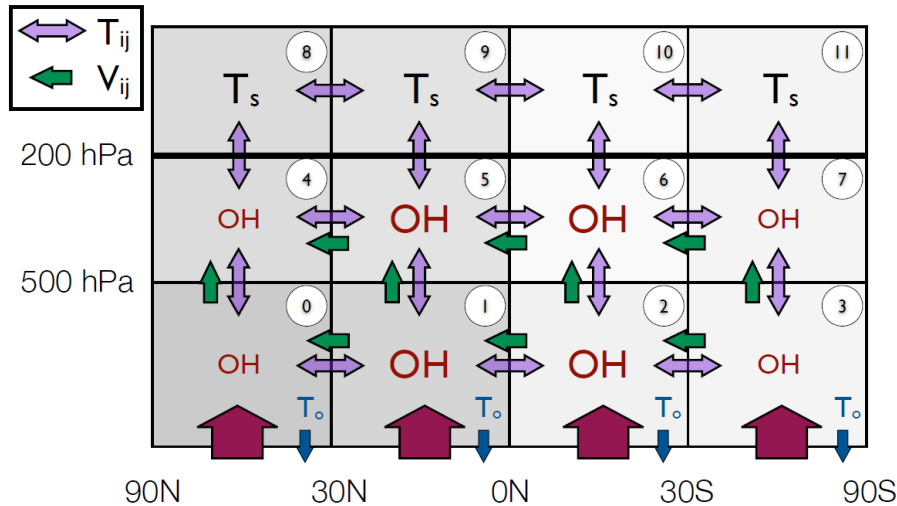


$$\frac{dB(t)}{dt} = S - \frac{B(t)}{t}$$

$$\frac{B(t)}{t} \rightarrow 0$$

$$\frac{dB(t)}{dt} = S$$

Bayesian inversions and more complex models



$$\mathbf{y} = \mathbf{M}\mathbf{x}$$

$\mathbf{y} = (y_1 \quad \cdots \quad y_m)$ is the vector of simulated mole fractions

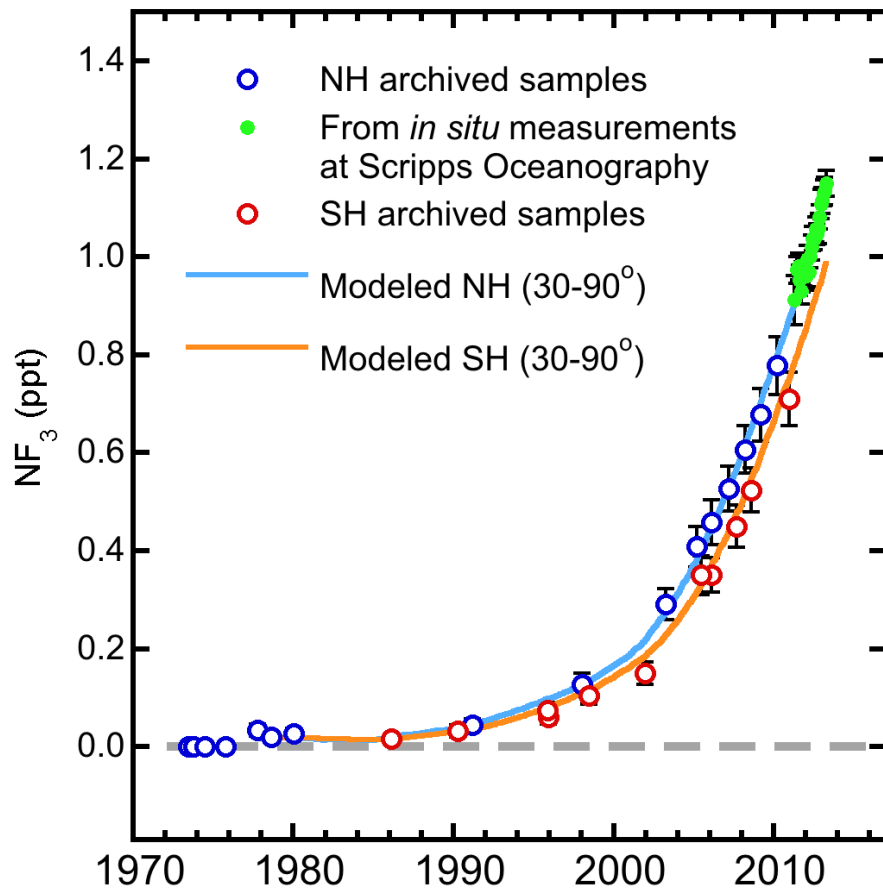
$\mathbf{x} = (x_1 \quad \cdots \quad x_n)$ is the state vector

\mathbf{M} is the sensitivity matrix (with dimension $m \times n$)

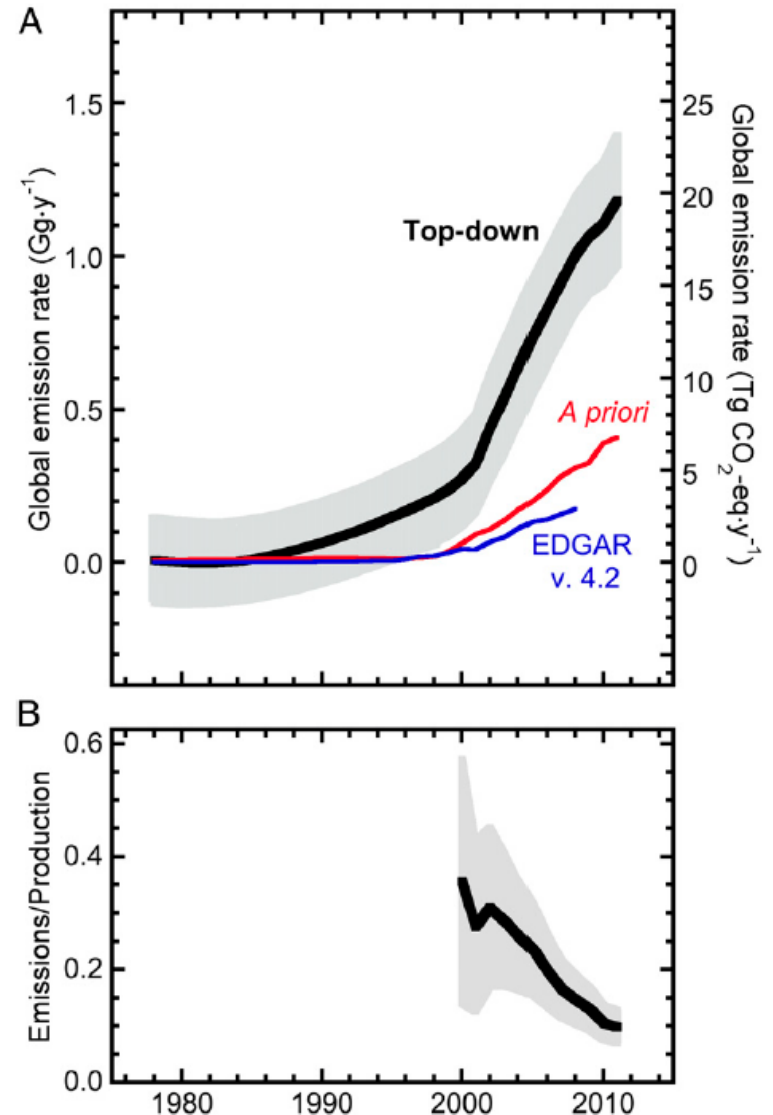
$$\mathbf{M} = \begin{pmatrix} M_{1,1} & \cdots & M_{1,n} \\ \vdots & \ddots & \vdots \\ M_{m,1} & \cdots & M_{m,n} \end{pmatrix}$$

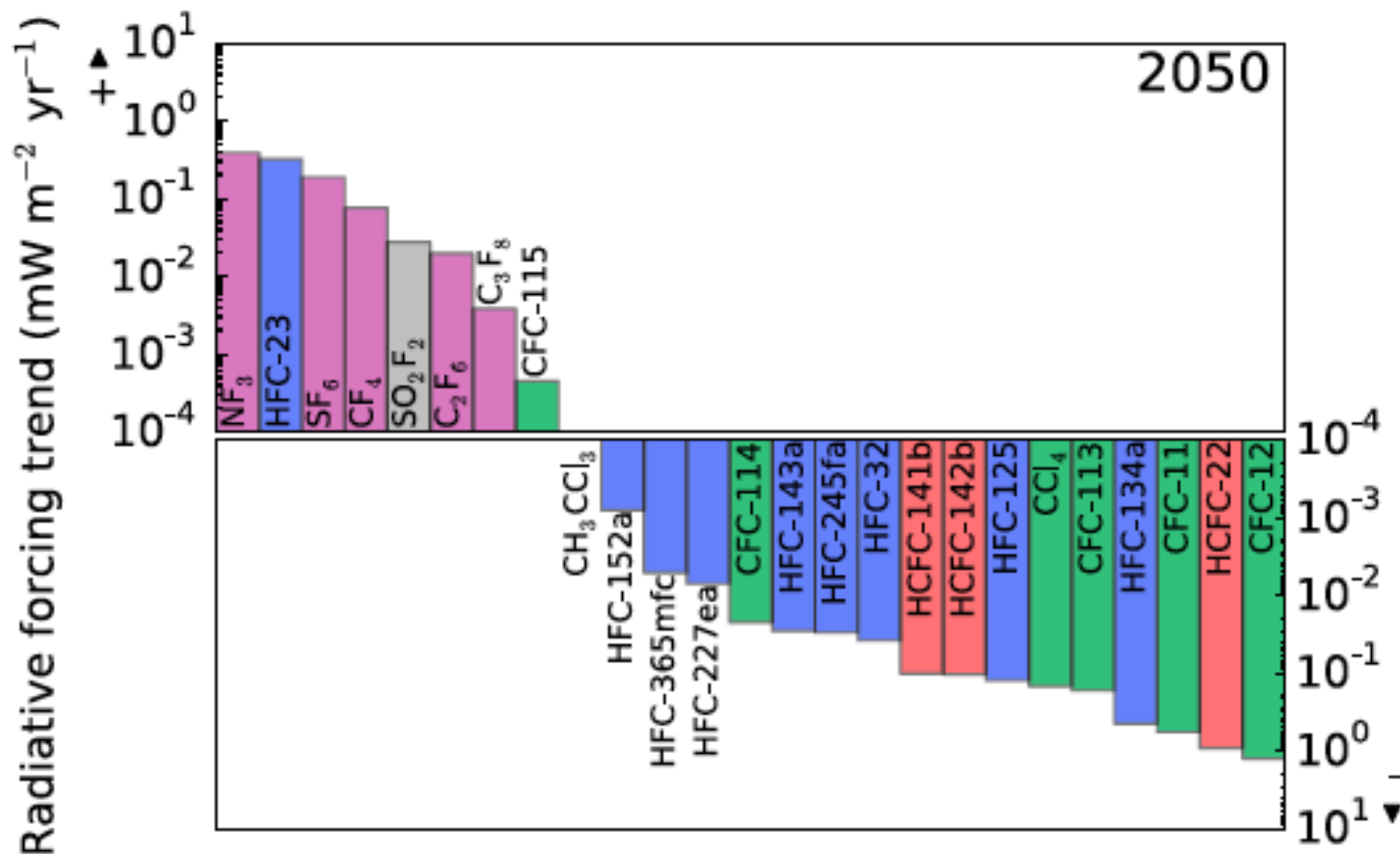
$$J = \frac{1}{2}(\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_b) + \frac{1}{2}(\mathbf{M}\mathbf{x} - \mathbf{y}_o)^T \mathbf{R}^{-1}(\mathbf{M}\mathbf{x} - \mathbf{y}_o)$$

Atmospheric measurements in verification



Arnold et al., 2012, *Analytical Chemistry*
Arnold et al., 2013, *PNAS*

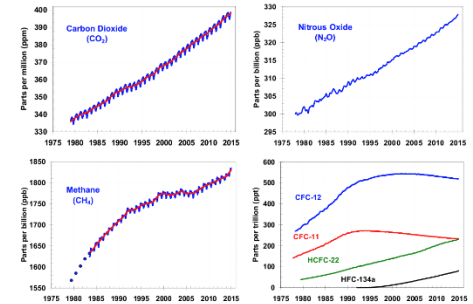




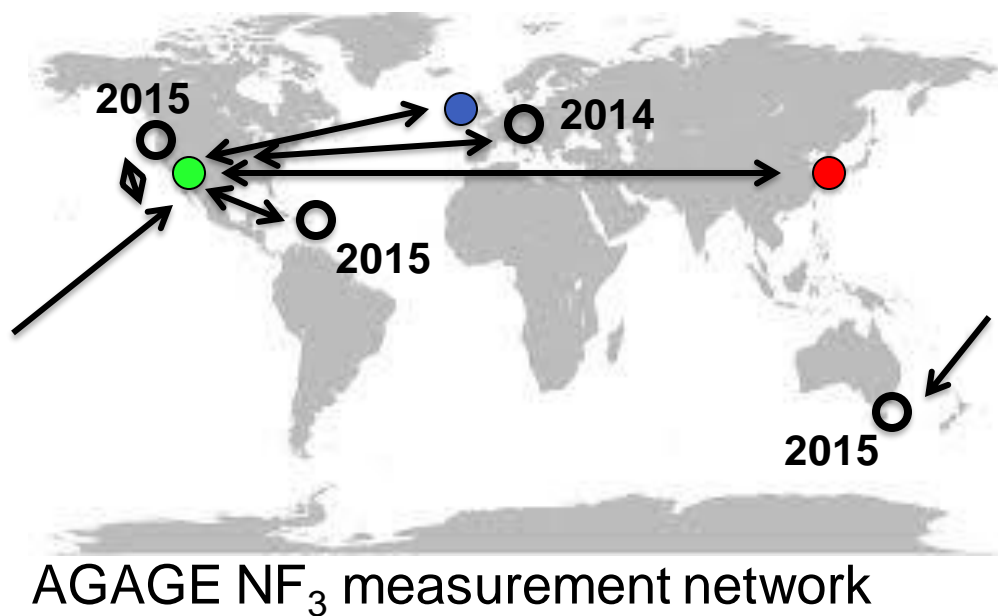
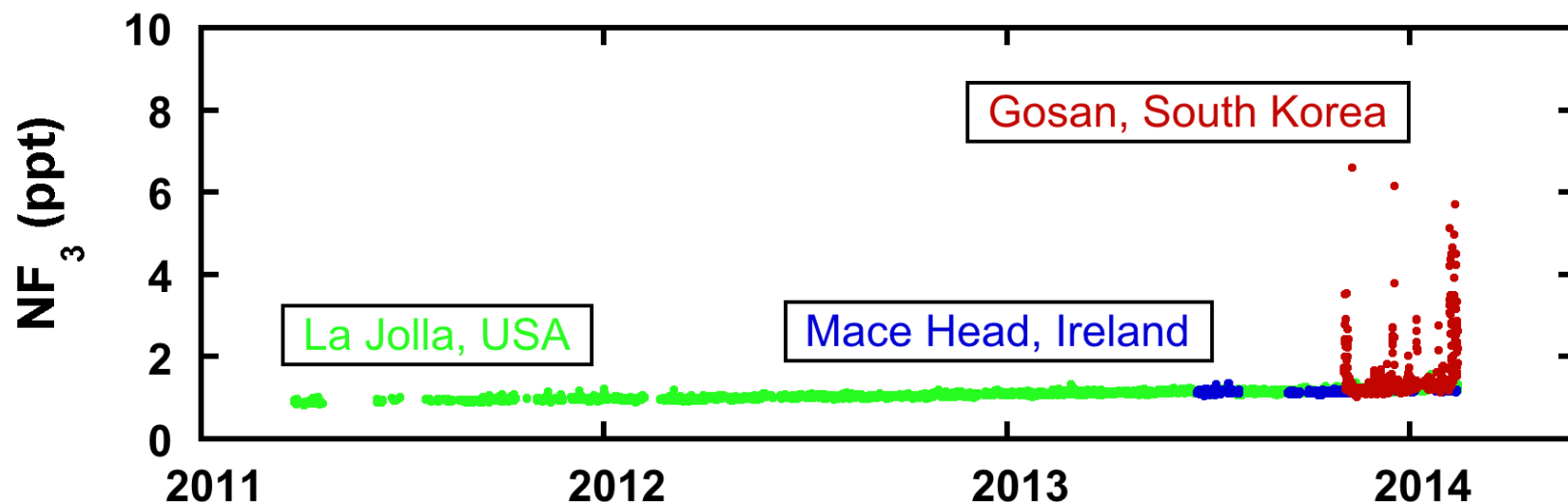
Atmospheric monitoring between the source and the well-mixed atmosphere



Understanding
emission sources
over
 $10\text{-}10^2\text{-}10^3\text{ km}$
using continuous
measurement
methods



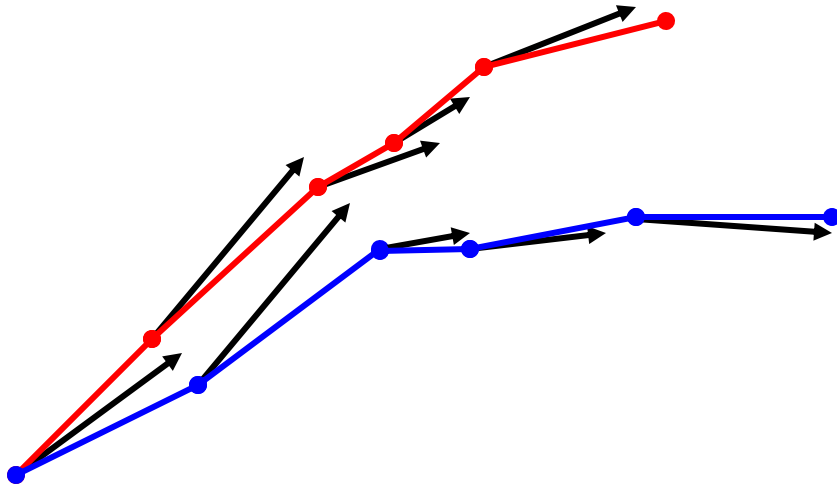
Higher spatial coverage



Calibration and intercomparison – vital exercises in high-precision networks

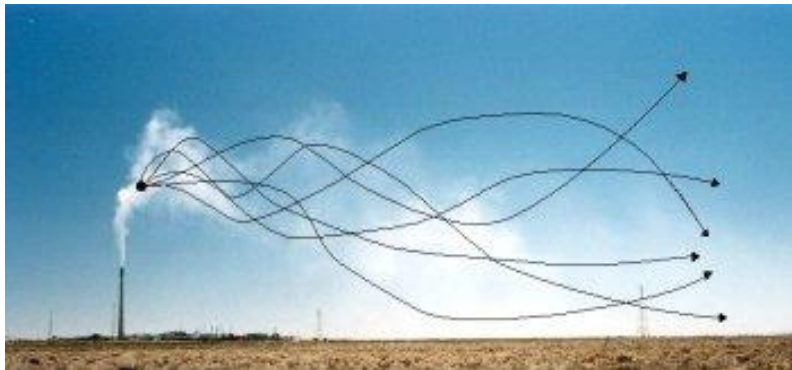
AGAGE now making 20,000 calibrated measurements per year globally

The tool of choice – Lagrangian particle dispersion models

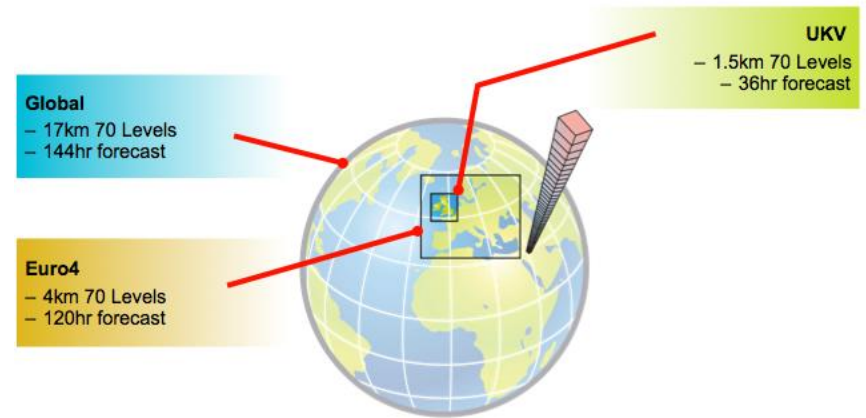


Black arrows show wind direction

Red and blue lines show movement of particles

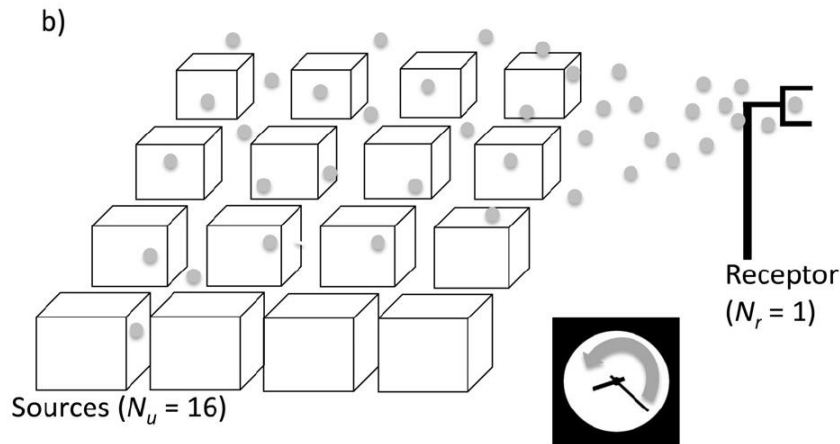


Met Office NAME model



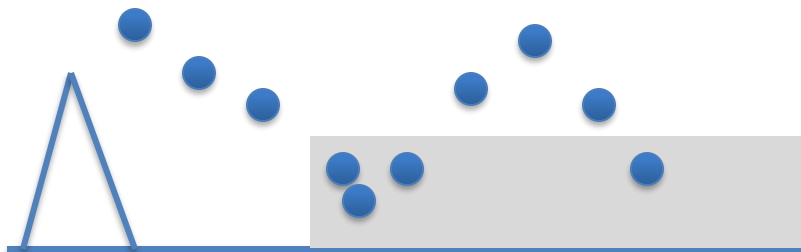
Met Office Unified Model (NWP)

Modelling of atmospheric mixing ratios at a regional scale



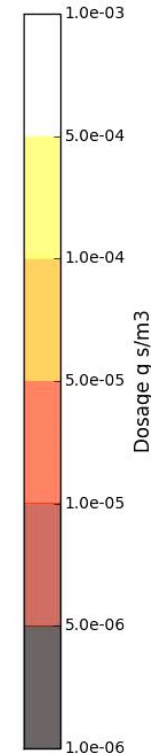
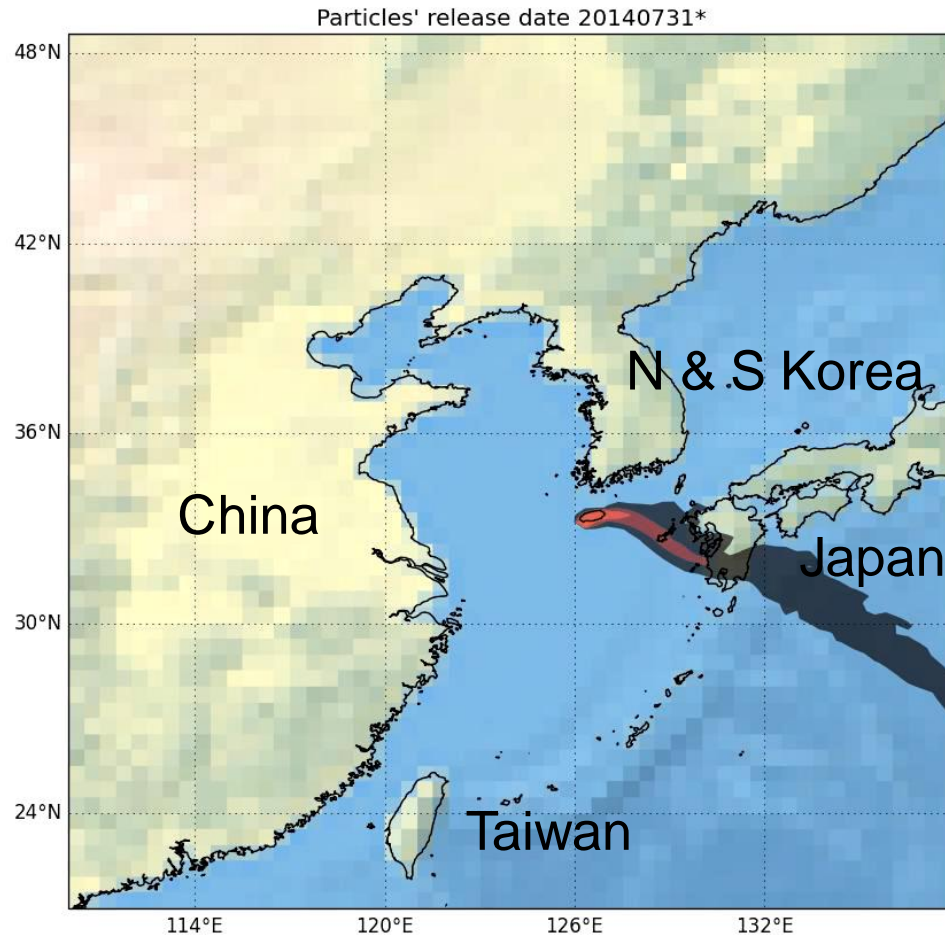
(Lin, *Lang. Mod. Atmos.*, 2012)

We release 20,000 **inert** particles over one hour from the measurement site and follow backwards for 30 days



Time integrated
concentration in each grid
box

Modelling of atmospheric mixing ratios at a regional scale

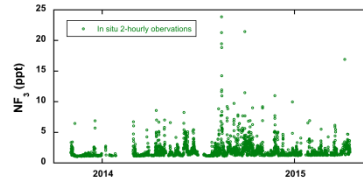


**Sensitivity
matrix**

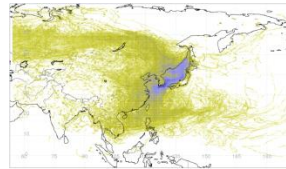
The effect of
emissions from a
specific grid to a
concentration
change at the
receptor

Combining model and observation data

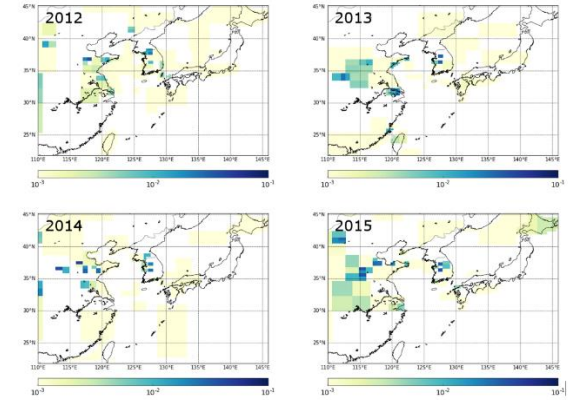
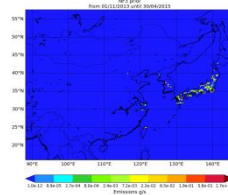
Measurements (y)



Sensitivity matrix (M)



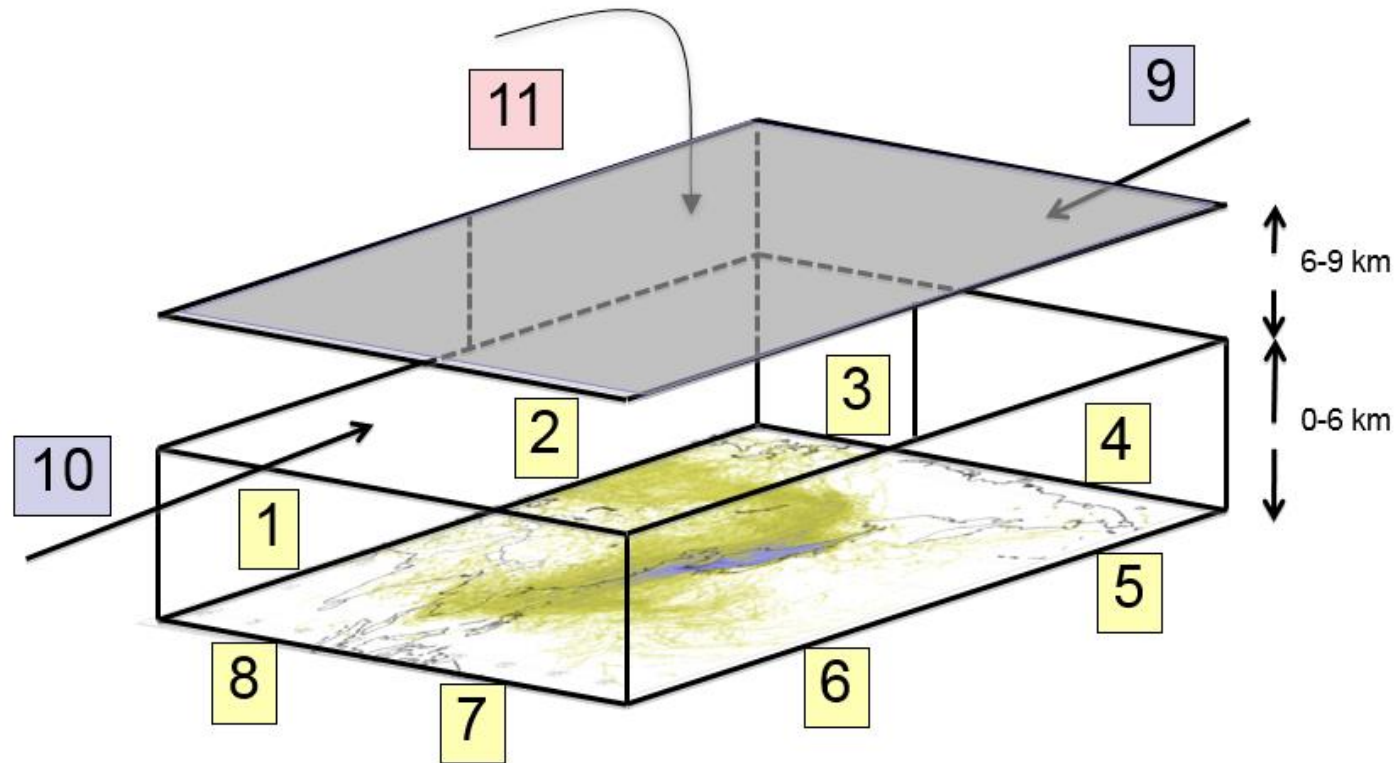
Prior estimate (x_b)



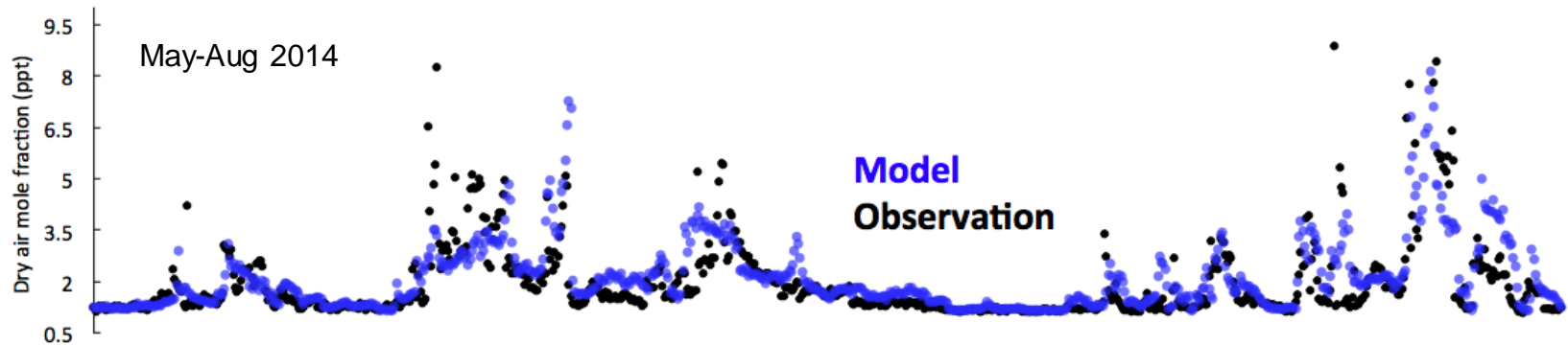
Posterior
estimates (x)

$$J = \frac{1}{2}(\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1}(\mathbf{x} - \mathbf{x}_b) + \frac{1}{2}(\mathbf{M}\mathbf{x} - \mathbf{y}_o)^T \mathbf{R}^{-1}(\mathbf{M}\mathbf{x} - \mathbf{y}_o)$$

Accounting for the variable/uncertain background atmospheric contribution



The biggest issue is in understanding how well the model behaves at each measurement time

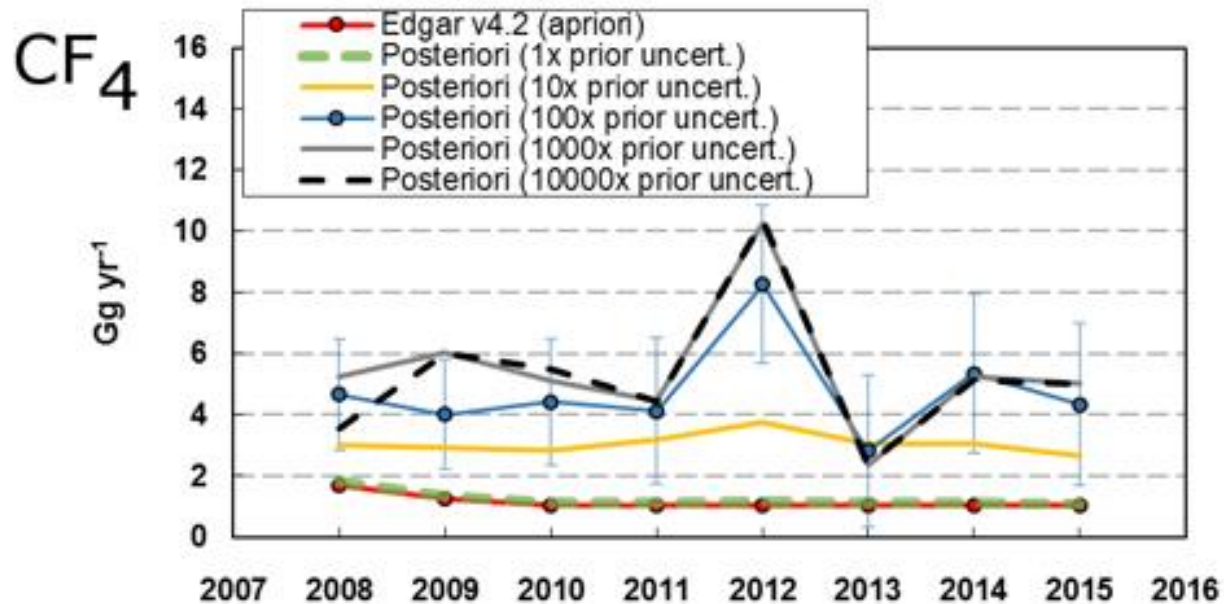


$$J = \frac{1}{2} (\mathbf{x} - \mathbf{x}_b)^T \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_b) + \frac{1}{2} (\mathbf{M}\mathbf{x} - \mathbf{y}_o)^T \mathbf{R}^{-1} (\mathbf{M}\mathbf{x} - \mathbf{y}_o)$$

$$\sigma_{model} = \sigma_{baseline} \times factor_{BLH} \times factor_{Topography} \times factor_{Inlet}$$

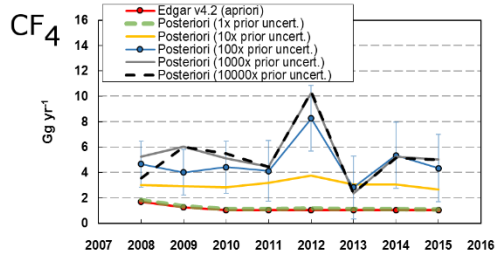
.. And how well is prior information understood?

China

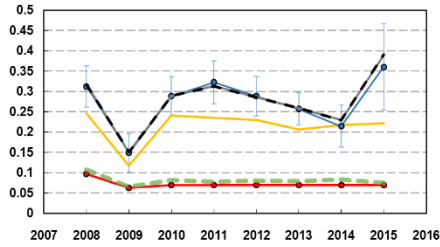


in prep

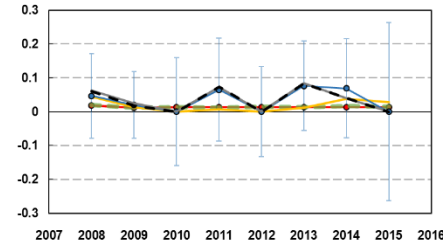
China



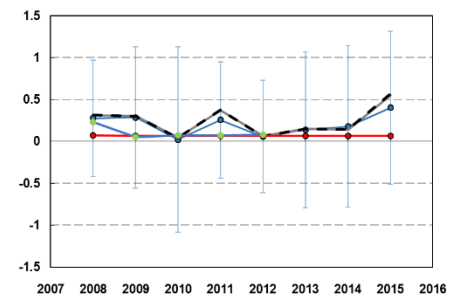
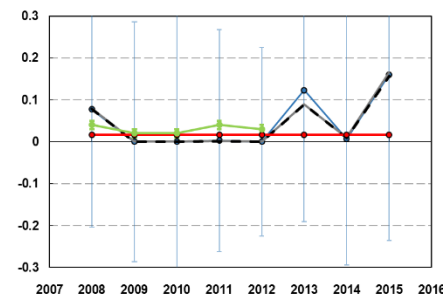
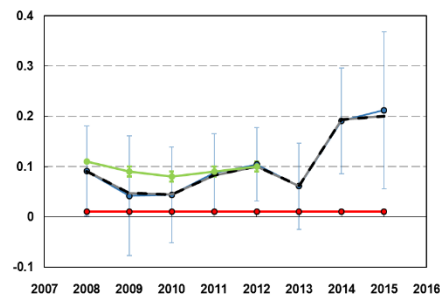
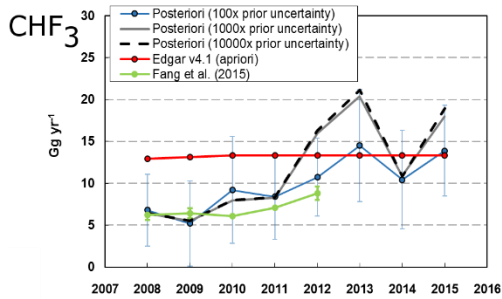
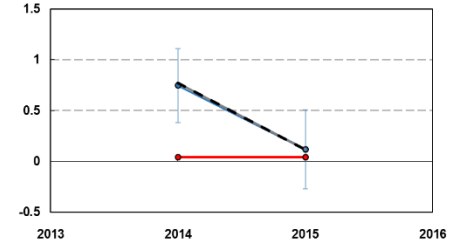
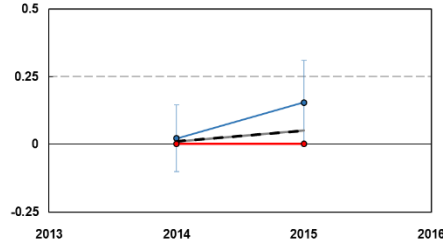
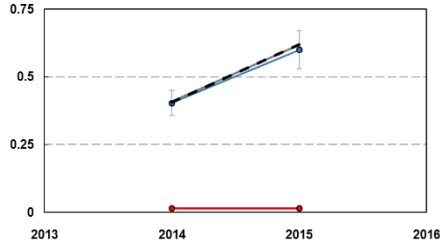
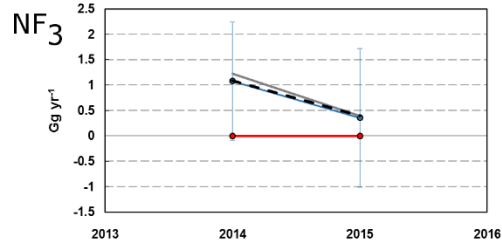
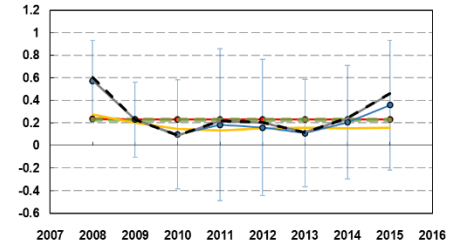
South Korea



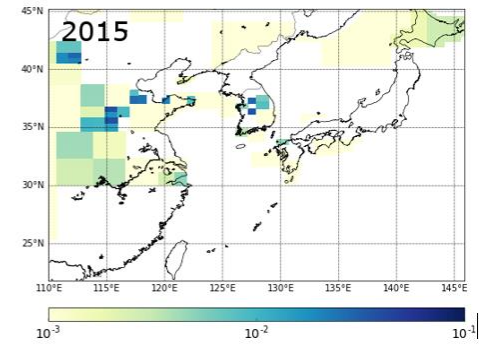
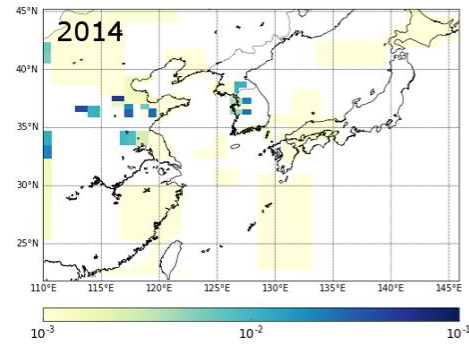
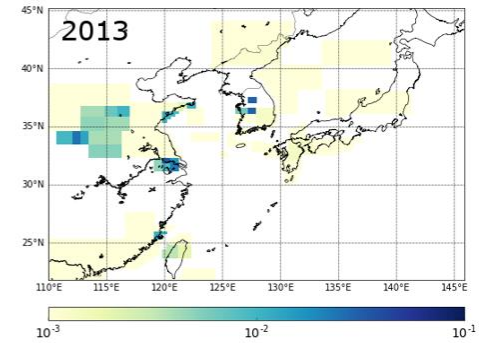
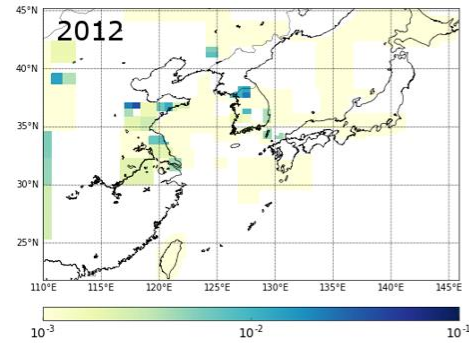
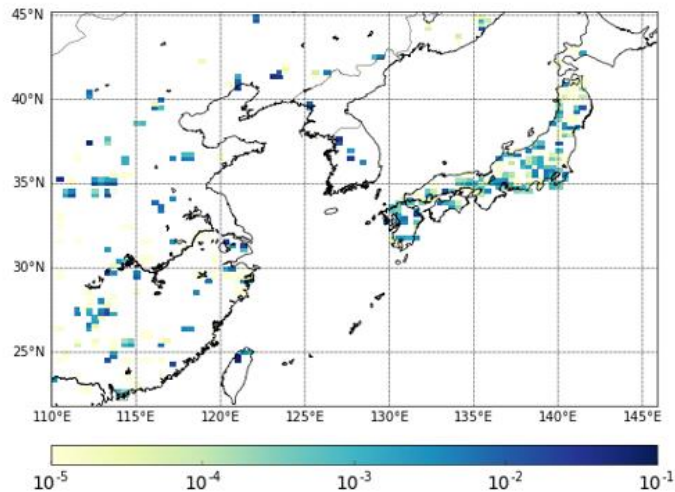
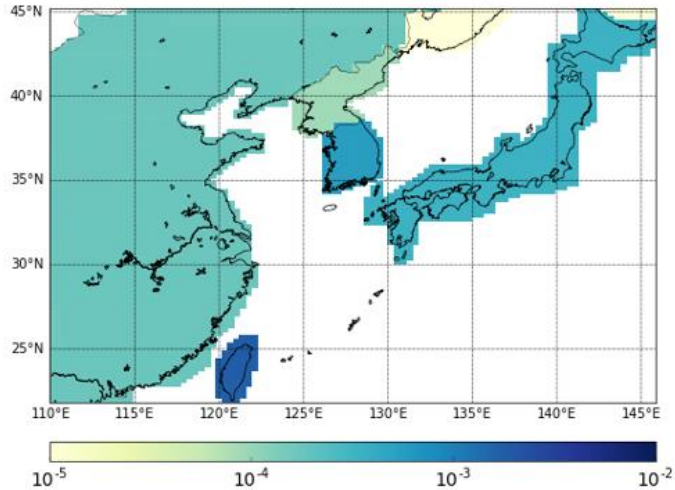
North Korea



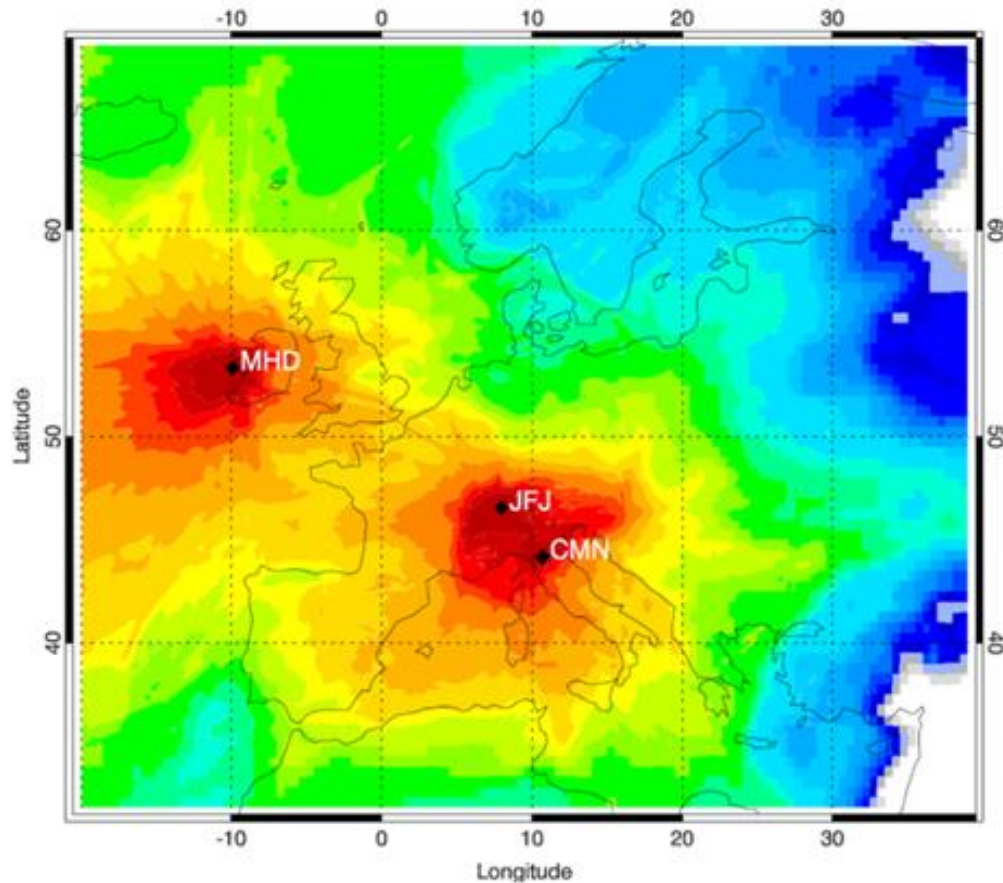
Japan



Uncertainty at higher spatial resolution



Model intercomparison project for SF₆, HFC-125 and HFC-134a



Brunner, Arnold et al., 2017, ACPD

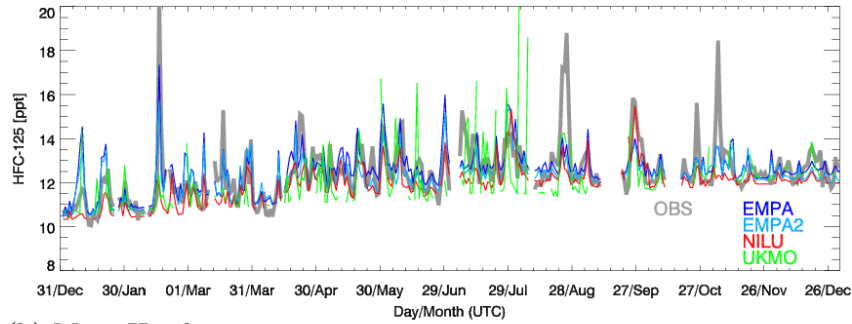
Model intercomparison project

Table 1: Overview of inversion systems

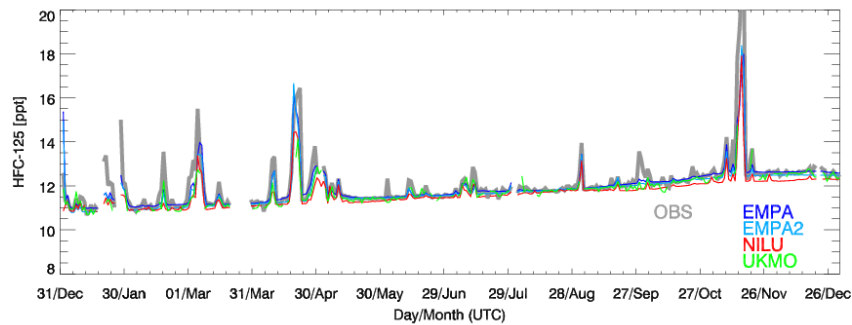
Characteristic	EMPA	EMPA2	NILU	UKMO
Approach	Extended Kalman Filter (ExKF)	Bayesian	Bayesian	Bayesian
Transport model	FLEXPART	FLEXPART	FLEXPART	NAME
Meteorology	ECMWF analyses 0.2°x0.2°, 3hrly	ECMWF analyses 0.2°x0.2°, 3hrly	ECMWF analyses 0.2°x0.2°, 3hrly	UKMO analyses 0.352° x 0.234°, 3hrly
Computational domain	Nested, global	Nested, global	Nested, global	45°W - 40°E, 25°N - 80°N
Inversion grid	0.1°x0.1°min., reduced according to residence time	0.1°x0.1°min., reduced according to residence time	1°x1° over land, reduced over ocean and far eastern boundary	0.352° x 0.234° min., reduced acc. to residence time and within country boundaries
State vector length (e=emiss., b=background, o=other)	1092 (1083e + 3b + 6o)	M1: 522 e + 84 b M2: 522 e + 84 b M3: 405 e + 56 b FLAT: 522 e + 56 b	M1: 1140e M2: 1140e M3: 1140e FLAT: 1140e	~150e + 11 b
Assimilation time resolution	3-hourly means	3-hourly means	Daily means	3-hourly means once per day
Spatial correlation of prior	500 km	None	200 km over land 1000 km over sea	None
Backwards mode run time	5 days	5 days	10 days	19 days
Prior background mole fractions	None, continuously estimated by ExKF	60-day REBS window, biweekly reference points	See Thompson and Stohl (2014) and description below.	Mace Head baseline for all sites, see Manning et al. (2011)
Temporal correlation of observation error	Red-noise Kalman filter	None	None, assumed negligible for daily means	None, assumed negligible with one value per day
Key references	Brunner et al., 2012	Stohl et al. 2009, Vollmer et al., 2009	Thompson and Stohl 2014	Manning et al., 2011

Model intercomparison project

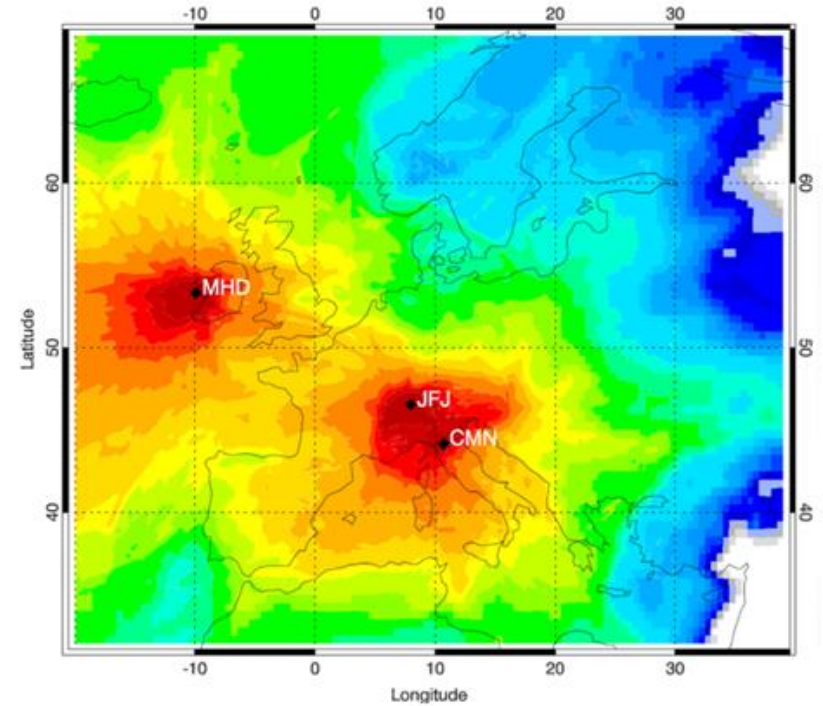
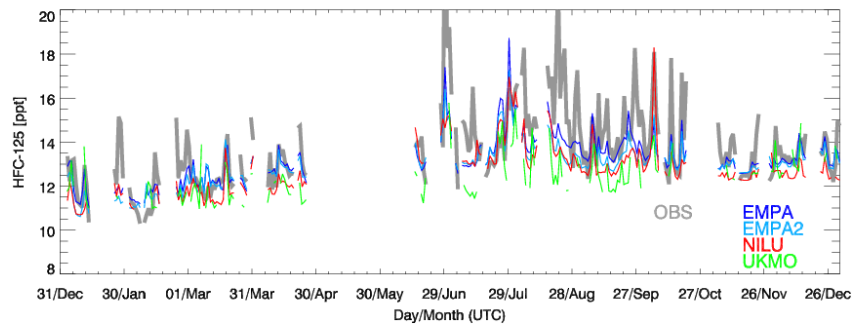
(a) Jungfraujoch



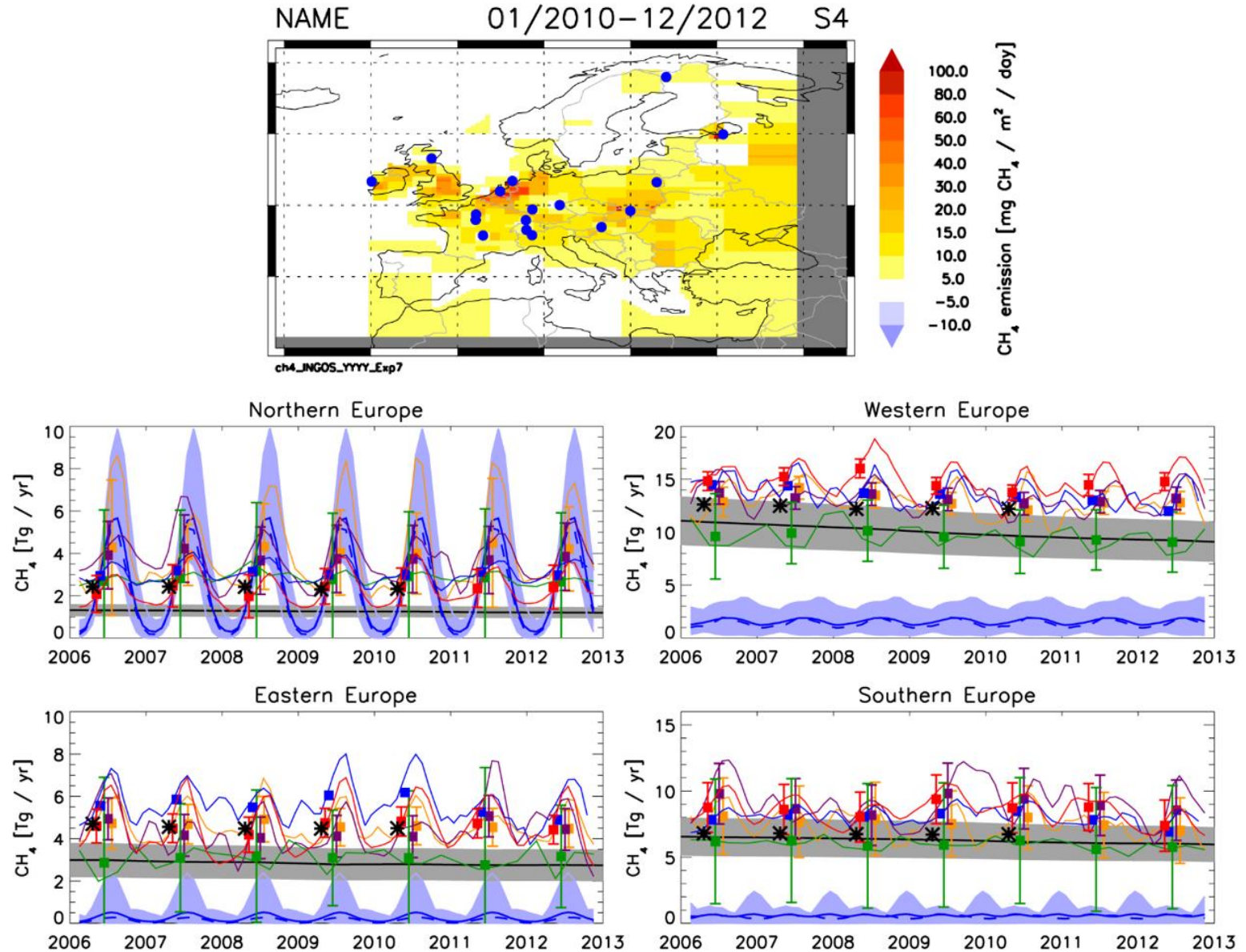
(b) Mace Head



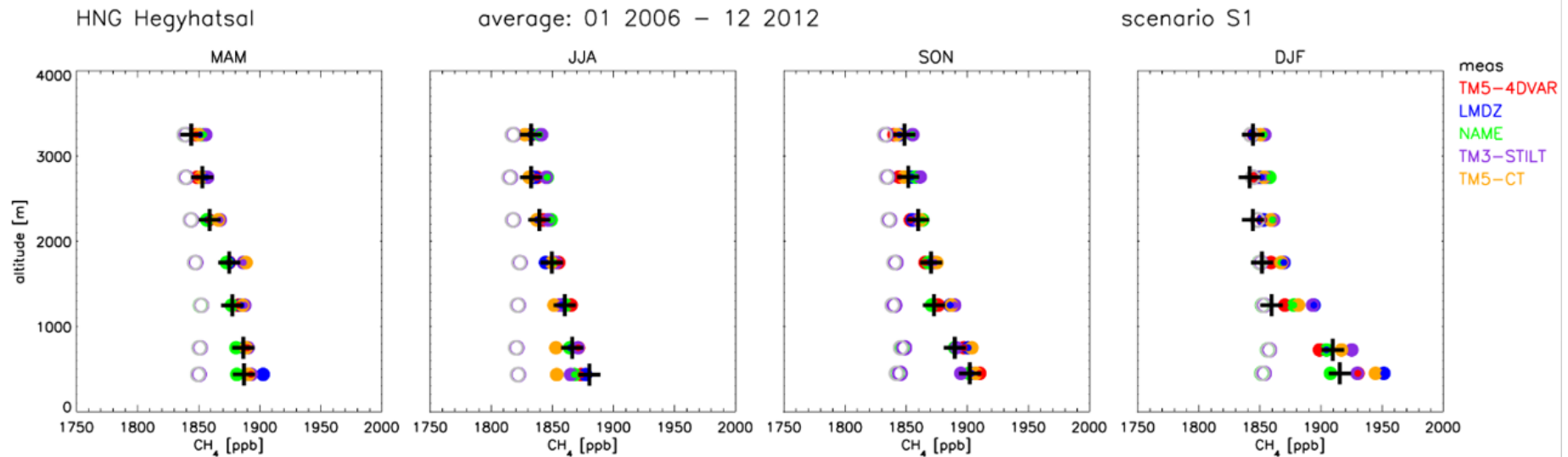
(c) Monte Cimone



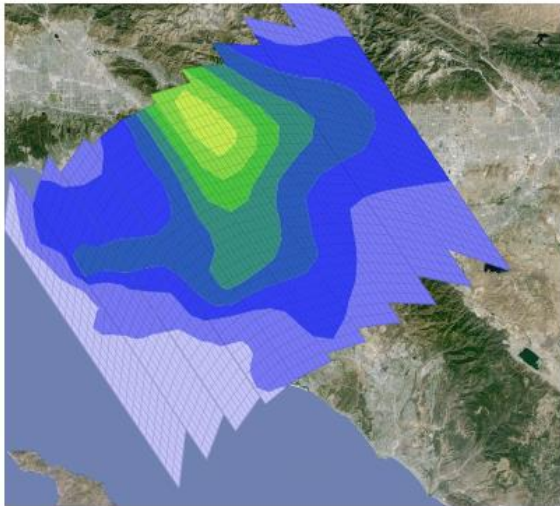
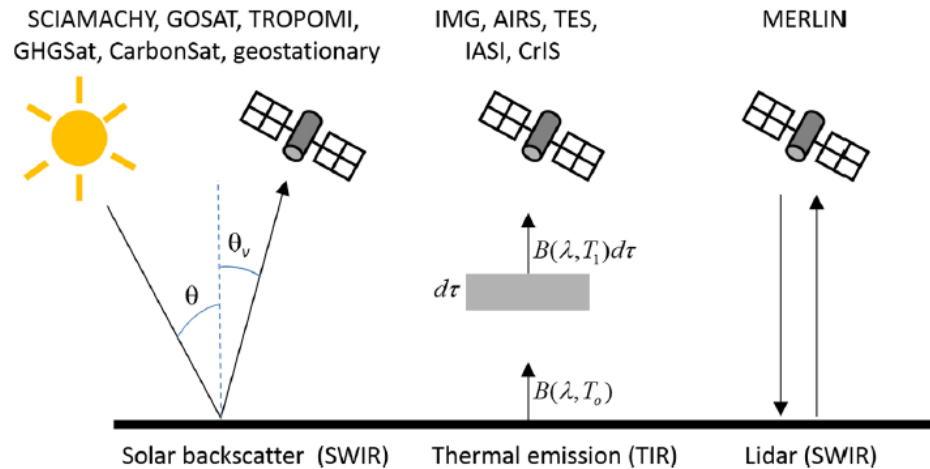
European methane model intercomparison project



Model validation using aircraft measurements



Need to prepare for satellite observations



Launch ~2018
Column CO₂ measurements over cities

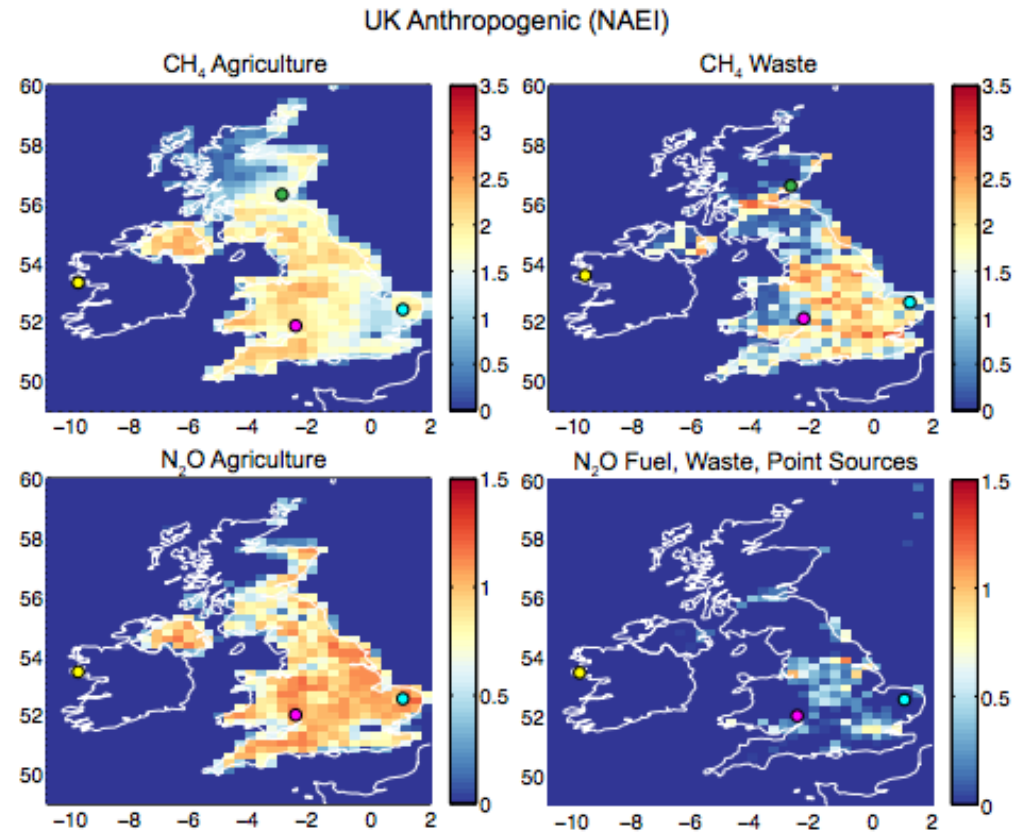
Nitrous oxide and methane have complex mixed source emissions

CH₄

- Wetlands (microbial)
- Fossil fuels
- Biomass burning
- Landfills
- Ruminants
- Rice cultivation

N₂O

- Soils
- Ocean
- Biomass burning
- Sewage
- Fossil fuel burning



Ganesan et al., *ACP*, 2015

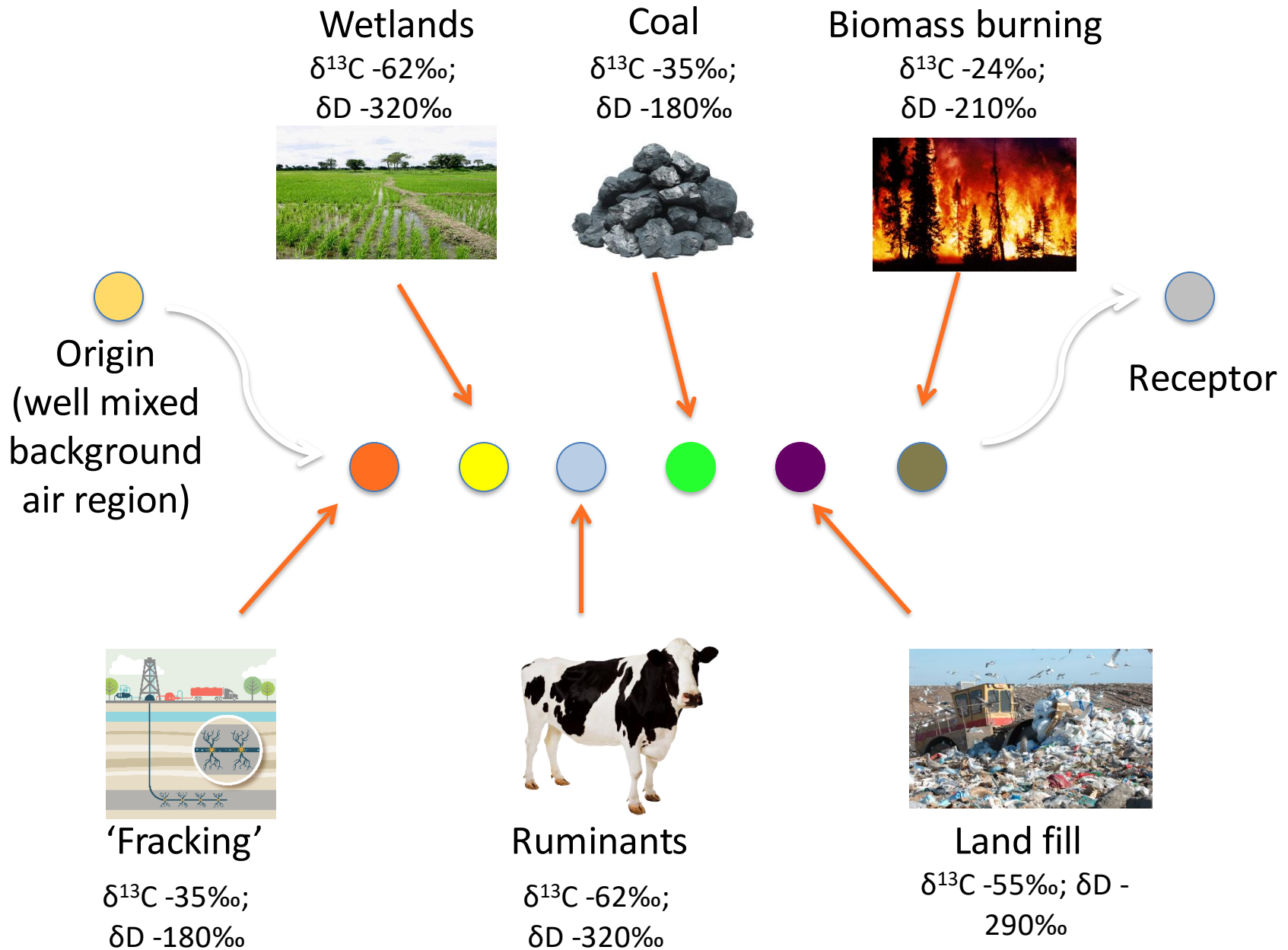
A large range of CH₄ emissions:

1.65–2.67 Tg yr⁻¹ (Ganesan et al., 2015)

1.7–3.0 Tg yr⁻¹ (MO for DECC)

1.8–3.0 Tg yr⁻¹ (Bergamaschi et al., *in prep*)

Methane sources have different isotopic signatures



Laboratory development of isotopologue measurements

CH_4 made of $^{12}\text{CH}_4$, $^{13}\text{CH}_4$, $^{12}\text{CH}_3\text{D}$.. and 7 more!

Preconcentration
Technique

Laser spectroscopy

