



The challenge of estimating regional emissions from observations

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Why estimate regional or country emissions from observations?

- As part of the UNFCCC (United Nations Framework Climate Change Convention) each developed country has to report its annual anthropogenic emissions of a range of greenhouse gases (GHG).
 - CO₂, CH₄, N₂O, HFCs, SF₆, PFCs
- Traditional **inventory** approach ('bottom-up').
 - Combines **Activity Data** (activities that result in the emission of a GHG e.g. landfill waste) and **Emission Factors** (links a specific activity to an emission).
 - Sum emissions per sector (industry, agriculture, energy, waste, etc) per gas to estimate an annual country GHG emission total.
- Emissions from observations: **Inversion modelling** ('top-down').
 - Challenge traditional emission inventories.
 - Completely independent.
 - Best practice for Kyoto Protocol although not mandatory.
- Both bottom-up and top-down methods have uncertainty.



Inversion modelling

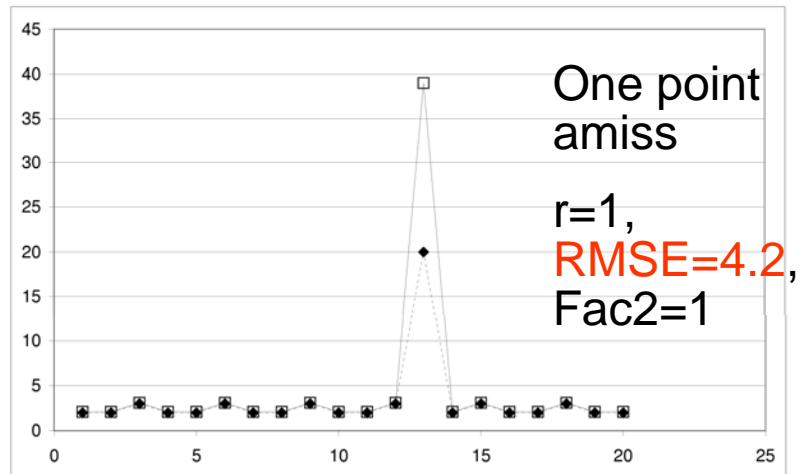
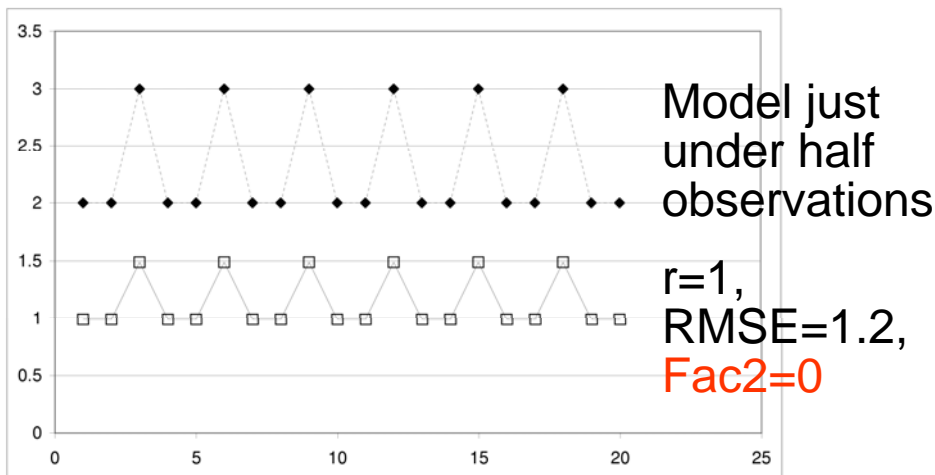
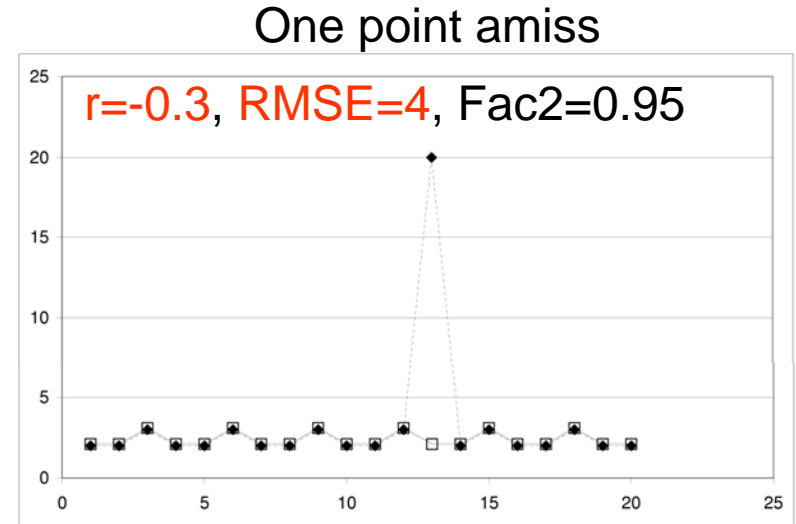
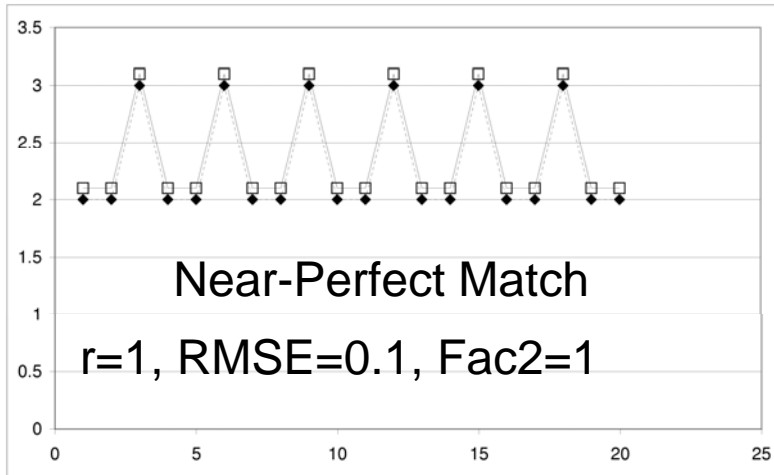
- Information required:
 - How emissions dilute in the atmosphere as they travel from a source region to an observation point.
 - Atmospheric transport (dispersion) model underpinned by meteorology.
 - Precise observations preferably at high temporal resolution.
- Output:
 - Spatial distribution and magnitude of emissions.
- Challenge:
 - Maximising the match in concentration between the modelled (estimated given an emission map) and measured (truth) time-series.



Comparing measured and modelled time-series of concentration

- Inversion Model
 - Uses a grid of emissions, fixed within a time frame (emission map).
 - Estimates the contribution each grid makes to an observation in each time period.
 - Uses a dispersion model and modelled meteorology.
 - Sums contributions from each grid to estimate total time-series of concentration.
 - Searches for the emission map that produces model time-series that has the best statistical match to the observed time-series.
- Statistical Match
 - What statistical function to use to measure the quality of the fit?
 - Common functions (there are many more):
 - root mean square error ($RMSE$),
 - correlation coefficient (r),
 - fraction within a factor of two ($FAC2$) .
 - Each statistic has strengths but also some weaknesses.

Simple artificial examples of statistical measures





Additional knowledge can help the inversion

- Inversion models can use additional information to better constrain the emission map.
 - E.g. knowledge about how the emissions are distributed (*a priori*).
- An emission map is penalised for moving away from the *a priori*.
 - I.e. the benefit in terms of a better time-series fit needs to outweigh the penalty of an emission map more distant from the *a priori*.
- How to price the distance from the *a priori* solution relative to the statistical fit of the modelled time-series to the observations?
 - related to the perceived quality of the *a priori* (subjective).
 - Usually defined as a percentage relative to the *a priori* solution, e.g. 100% uncertainty.
 - If *a priori* estimate is small this does not give much leeway.
 - New or unexpected sources or those significantly different from the *a priori* estimate struggle to be seen in the inversion.
- A **good** *a priori* estimate can significantly improve the robustness of the final solution. A **poor** one is detrimental.
- If inventory data used then inversion solution is **not independent**.



Transport (dispersion) model

- Critical component of inversion system.
 - Describe how emissions dilute with distance and where they go.
- 2 components: Meteorology and Dispersion.
- 3-D wind, temperature and boundary layer information on a grid from Numerical Weather Prediction (NWP) models.
 - Use short-term forecast 0-3 hours corrected for by observations.
 - Resolution varies between models (25-80 km globally up to 1.5 km country scale).
 - NWP models do not 'see' everything. Sub-grid features not represented, i.e. sharp changes in orographic features e.g. steep mountains, valleys or coasts. NWP 'sees' average flow in grid.

Location of measurement station matters

- Flat terrain sites are usually ideal as the flow well represented by NWP.

Cabauw Tower
The Netherlands



Location of measurement station matters

- Coastal stations are affected by (sub-grid scale) land-sea breezes but benefit from a 'clean' well mixed sea sector.

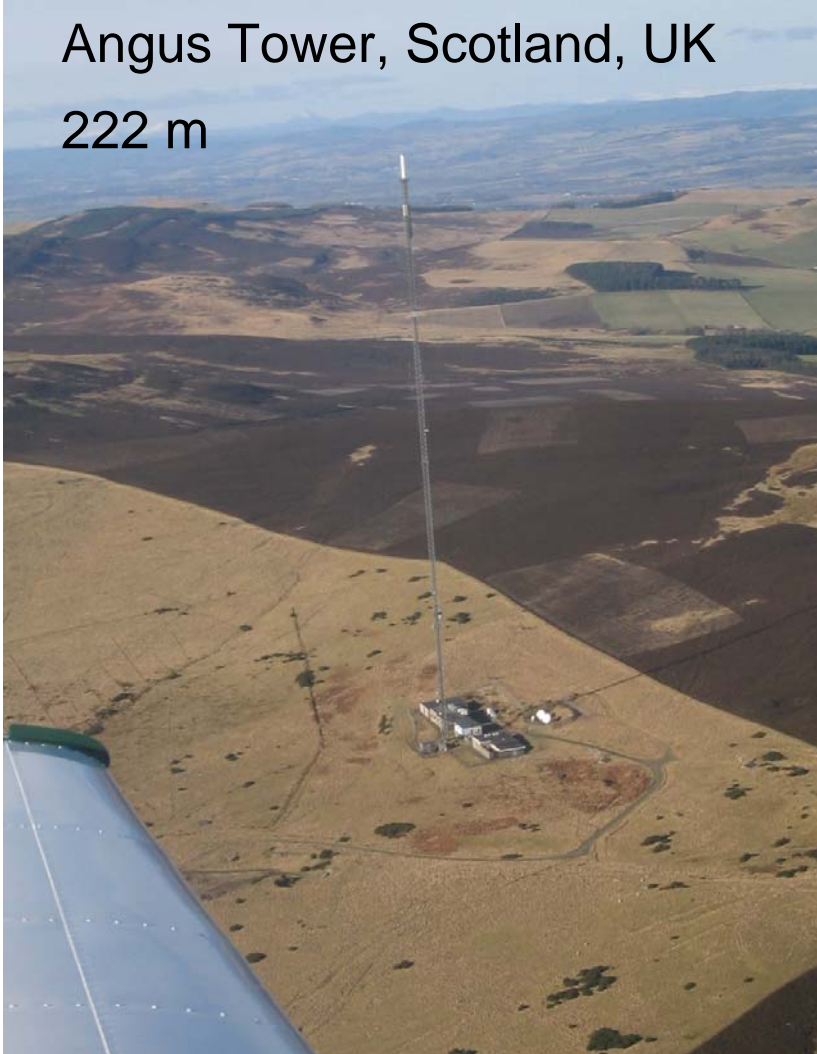
Mace Head

West coast Ireland



Location of measurement station matters

Angus Tower, Scotland, UK
222 m



- Elevated observations (tower) more representative of grid average.
- Ground is heterogeneous and thus complex.
- Potentially difficult to decide whether measurement within Boundary Layer (BL) or not (profile of observations valuable for this).
 - BL notoriously difficult to estimate in NWP models.

Location of measurement station matters

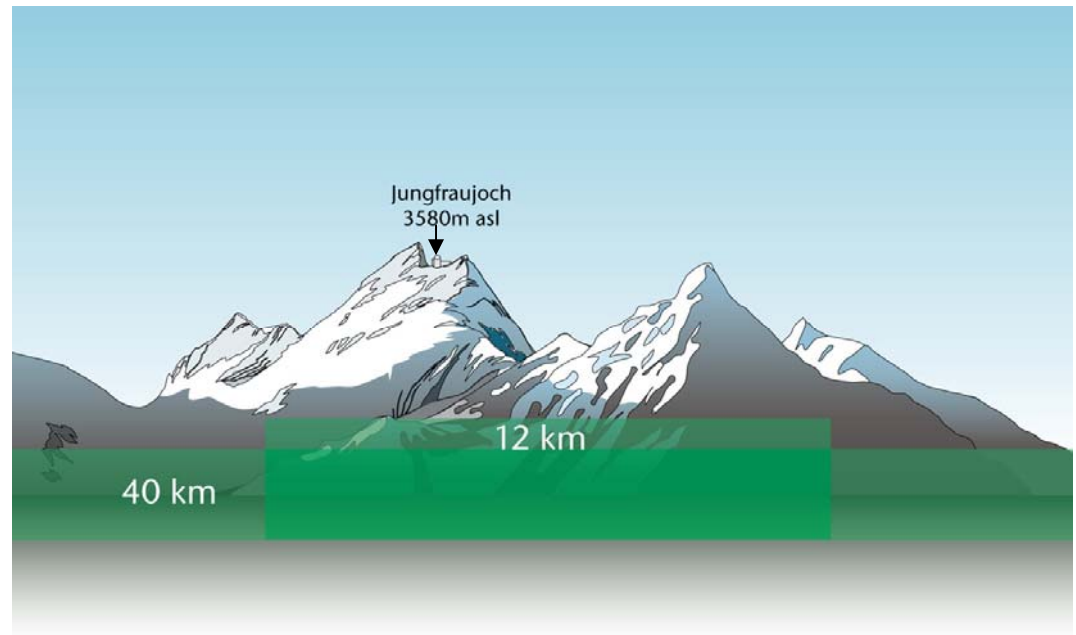
Stations in mountainous areas are very challenging!



Jungfrauoch station in the Swiss Alps.

Where is the measurement station?

- Which point in 3-D model atmosphere is most representative of the observed air?
- Surface station at Jungfraujoch in the Alps (3580 m asl).
 - On saddle between two mountains with valleys on either side.
 - UK Met Office global model 40 km: ground level = 1760 m.
 - UK Met Office North Atlantic European model 12 km: ground = 2110 m.



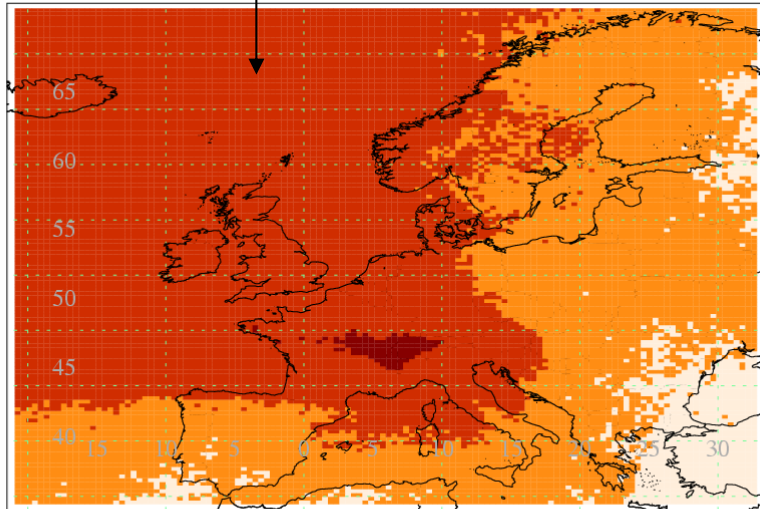


Where is the measurement station?

- Night - flow reaching the station **probably** disconnected from ground (above boundary layer [BL]) and best represented by a model point at 3580 m a.s.l. (i.e. 1.8 km above global model ground).
- Day - station **probably** influenced by upslope surface winds from the valleys and best represented by model point on the model ground.
- When does it switch?
- Difficult to model => Challenging to interpret observations.

Impact of station location

Annual difference of more than 400 3-hr periods



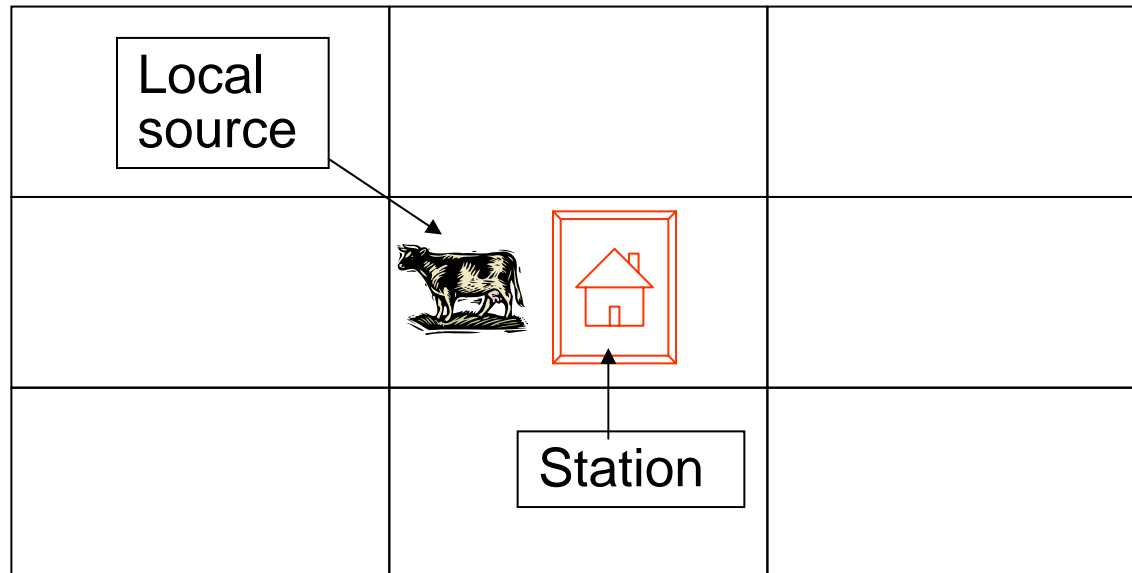
Maximum value = 1.12×10^3



- Difference between whether modelled Jungfraujoch is on ground or above BL is significant.
- Surface regional emissions more readily impact station when modelled on the ground.
- Strong impact on inversion solution.
- Smaller impacts in less severe topography.

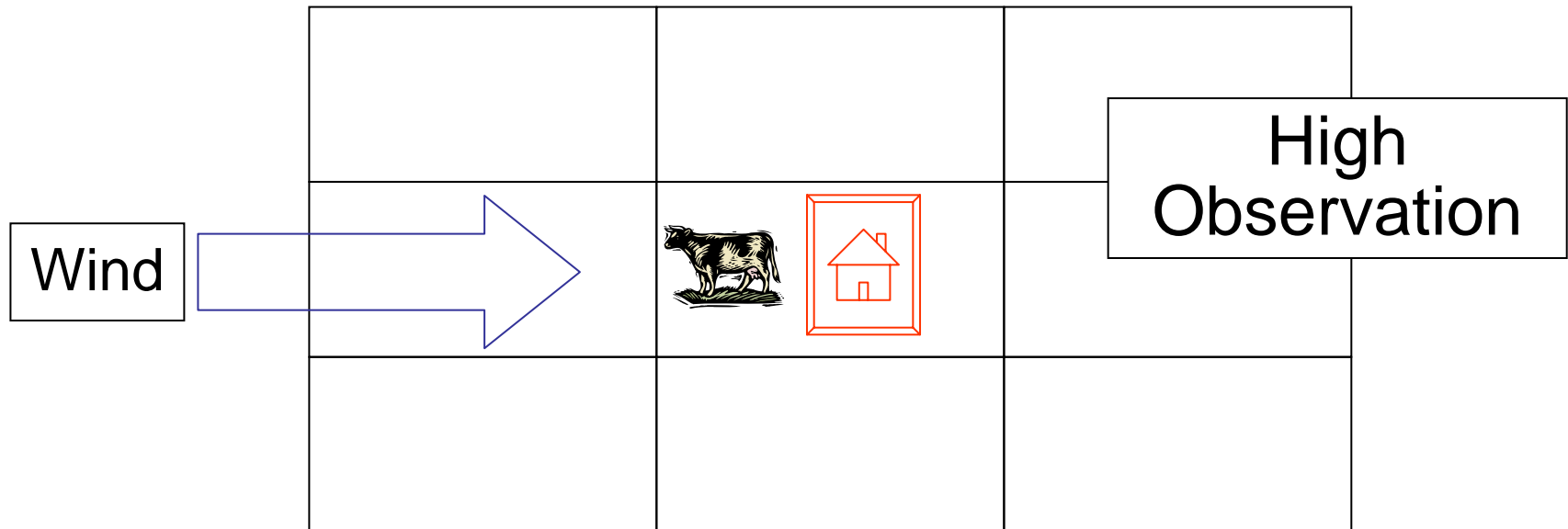
Resolution of inversion solution

- Inversion emission map has a defined horizontal grid and a specified time window. During this time emissions are assumed constant in each grid.
- Intermittent emissions or large sources near to the monitoring station will cause problems for inversion.
- Incorrectly placed emissions will lead to over- or under- estimates.



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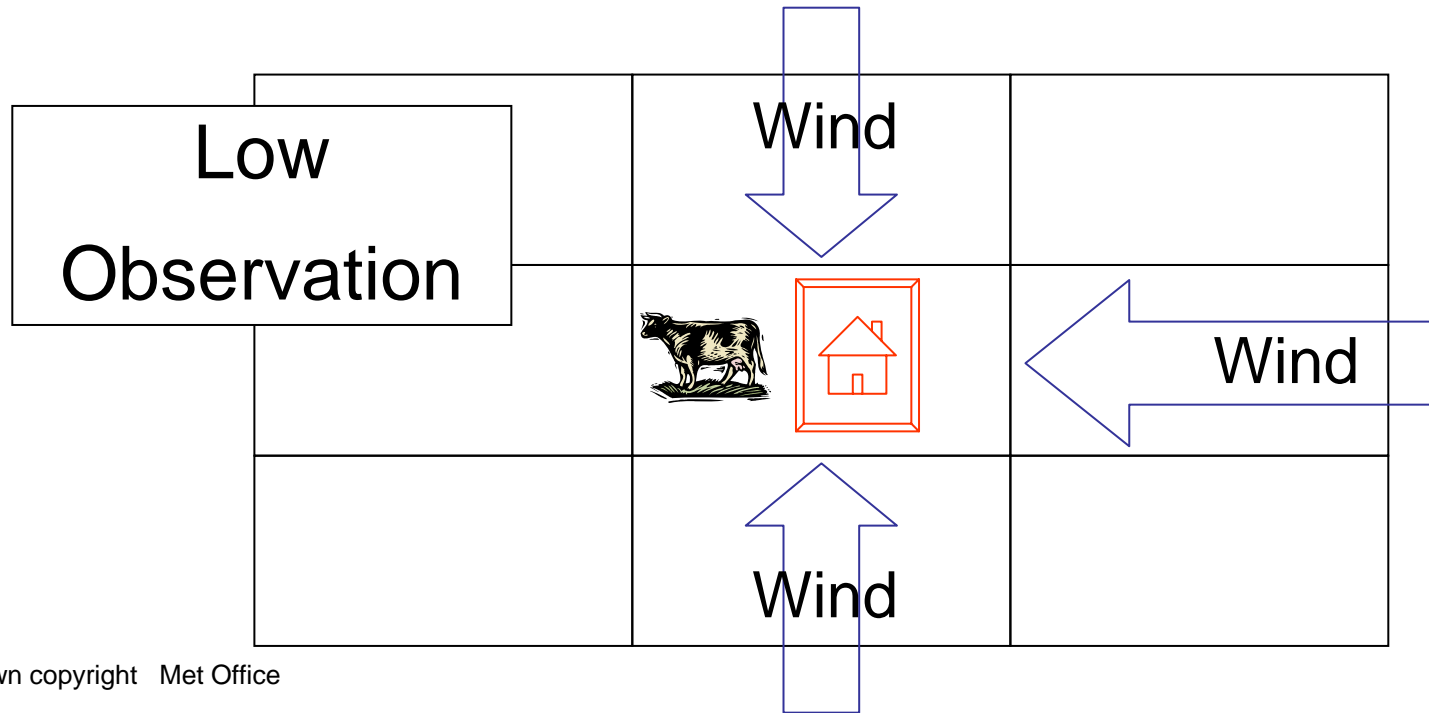




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Resolution of inversion solution

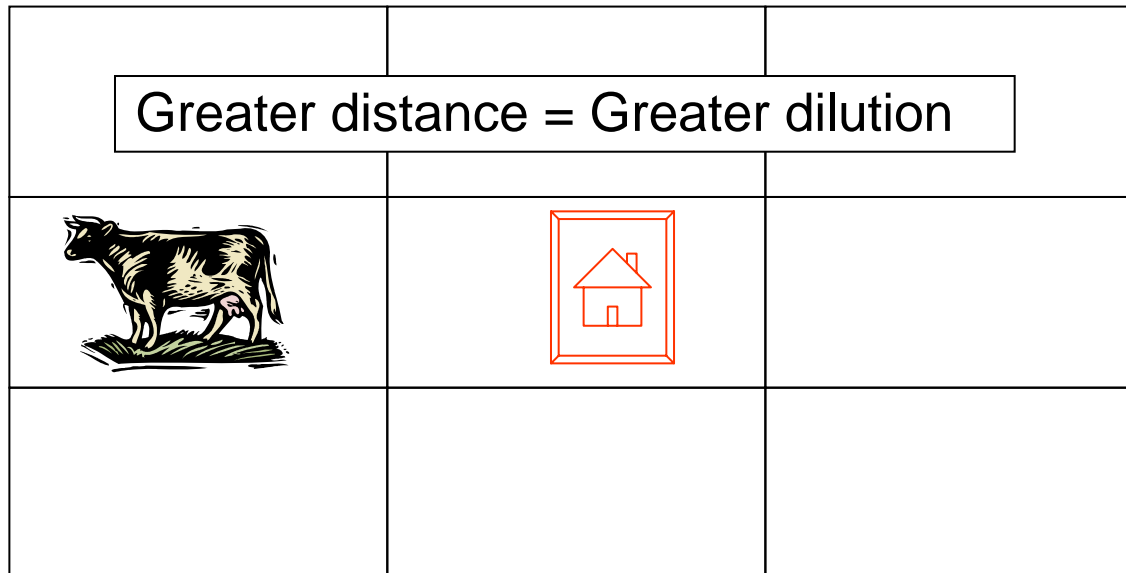
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**Larger
emission
estimated
further
from
station**





Selecting the observations to use in the inversion

- Increasing the number of observations can help the inversion model (better triangulation).
 - Flask (weekly) to high-frequency (hourly).
 - Observations from multiple measurement stations.
 - Improves knowledge about the distribution and magnitude of sources.
- **Multiple stations** – Are the observations inter-comparable? Would both instruments measure the same value if side by side?
- Removing observations that maybe too challenging to model e.g.
 - low wind speed conditions (local sources dominate).
 - day time values in mountainous areas (flow complex).

Uncertainty in inversion models

- Sources of uncertainty:
 - Observation error,
 - Error in modelled meteorology,
 - Error in the transport and dilution of pollution.
 - Error in inversion method (statistical fit)
- Difficult to assess the total uncertainty but vital to try. Possible ways to test robustness of inversion solutions:
 - Randomly perturb or randomly sub-sample observations,
 - Use multiple NWP models (ensemble),
 - Use multiple inversion methods (inversion ensemble),
 - Change uncertainty level when using *a priori* information.



Examples of inversion modelling

- NAME-inversion method:
 - Example: HFC-134a.
 - Principle use: mobile (car) air conditioning units.
- NitroEurope 5-year EU project:
 - Multiple inversion models.
 - Example: CH₄ (methane).
 - Principle emissions: farming, waste, energy.



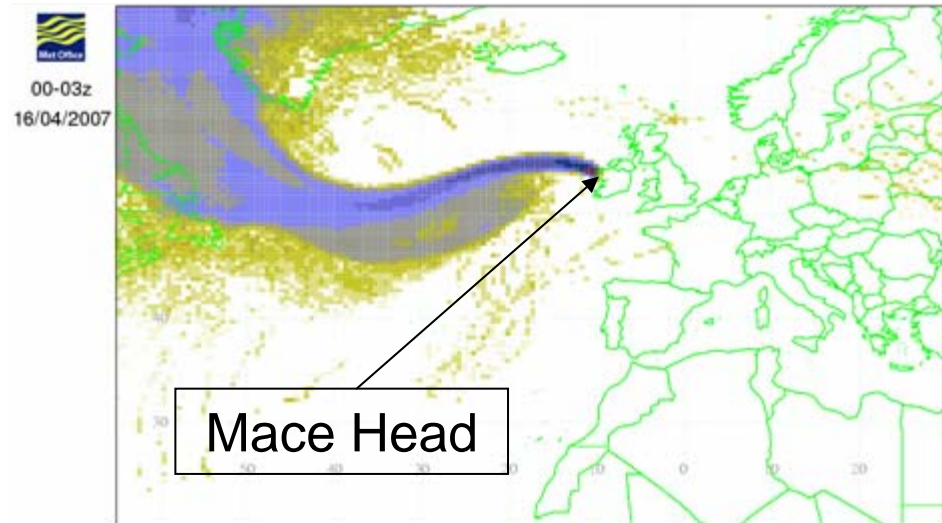
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NAME-inversion method

- NAME model (Lagrangian particle dispersion model).
- Uses 3D meteorological data from UK Met Office NWP and ECMWF models (40-80 km resolution).
- Derive air history map for Mace head for a 3-hour period:
 - Combination of tens of thousands of trajectories.
 - Darker shade means greater contribution from that area.
 - All surface sources within previous 12 days of travel that contribute to an observation during a 3-hour period are recorded.

Maps generated for each 3-hour period:

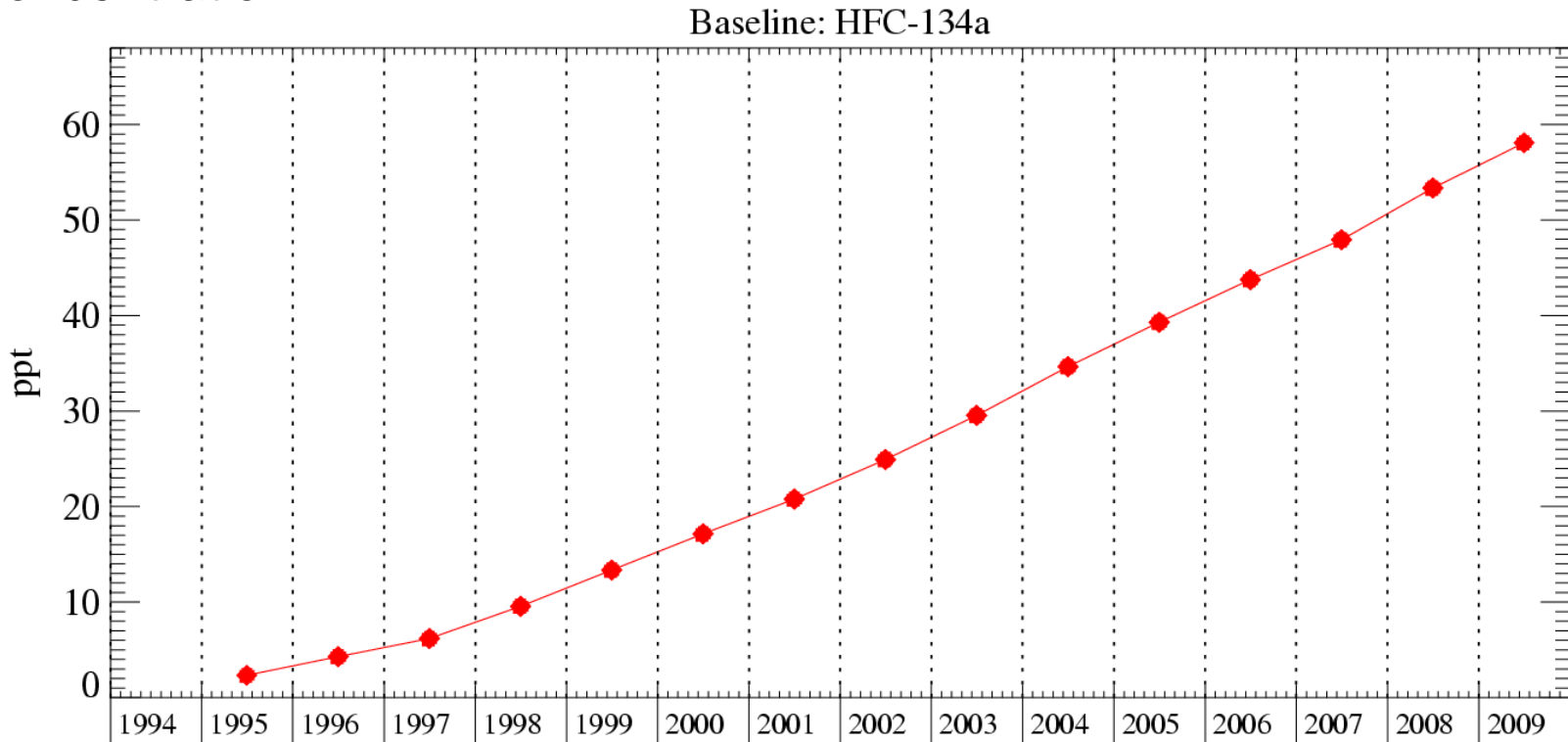
- 1994 onwards



HFC-134a

Northern Hemisphere baseline

- Select observations when air come from Atlantic and wind speed high.
- Smooth 'baseline' data and derive Northern Hemisphere background concentration.

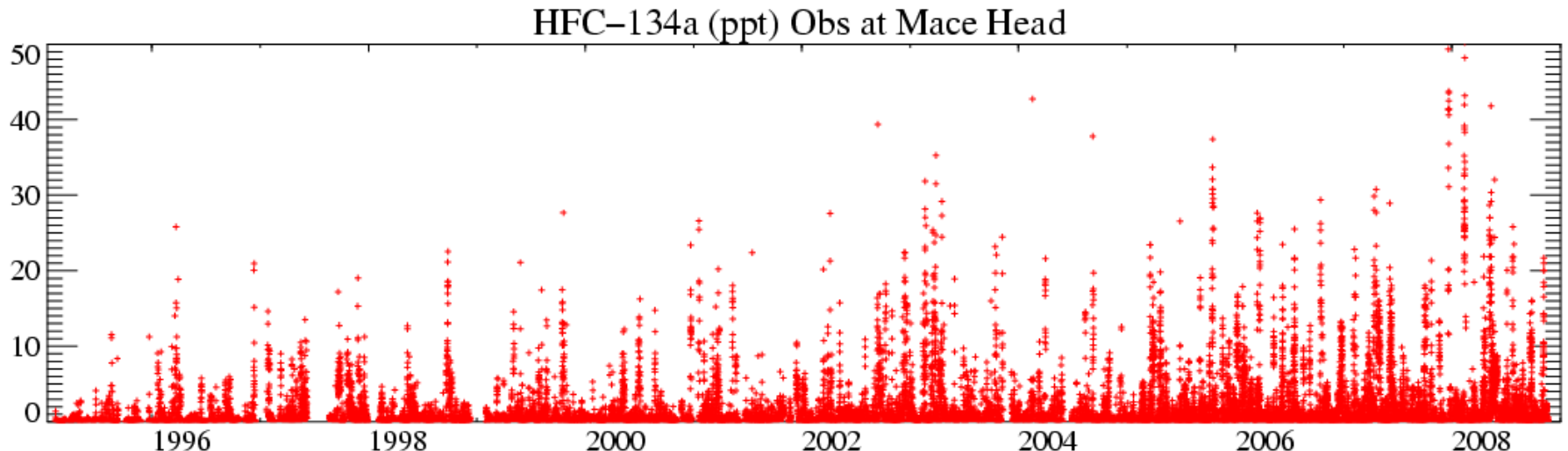




Regional emissions produce observations above baseline

Aim: **Generate emission estimates from ‘polluted’ (above baseline) observations.**

Subtract the baseline concentration from each observation.



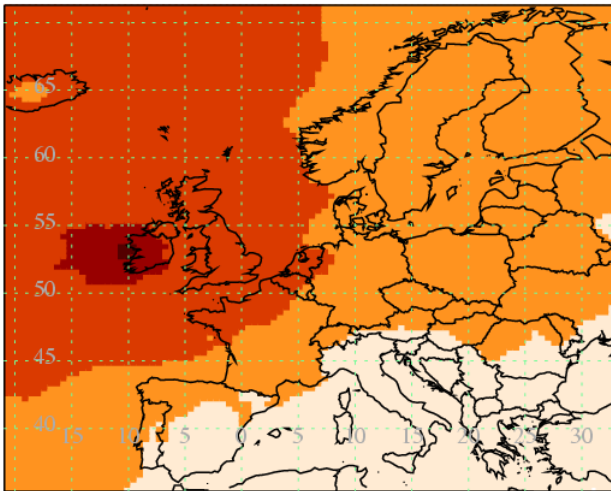
What regions influence Mace Head?

Composite of air-history maps.

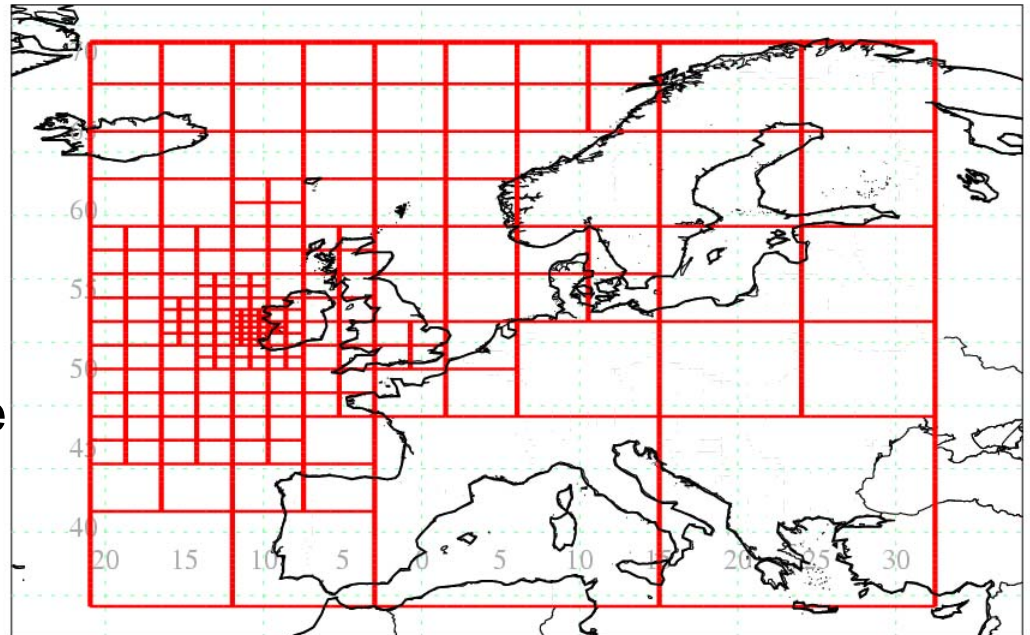
Greatest influence: Ireland, UK, northern France and Benelux countries.

Lesser influence: southern France, Germany, Denmark.

Poor influence: Mediterranean countries.

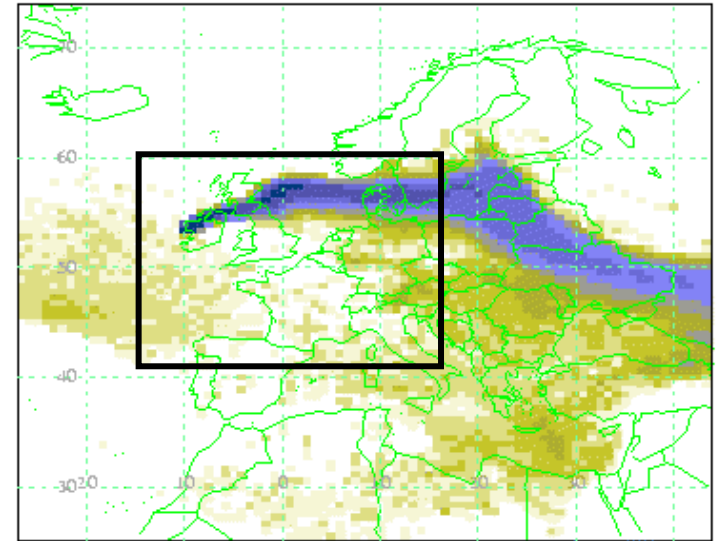


Increase size of solution grid as influence on Mace Head decreases.



NAME-inversion technique

Air Origin Map = Matrix **A**
(N° times x N° grids)



Measurement - Baseline = **m**

Emission Map = **e** (the solution)

Relationship: **A** **e** = **m**

Problem: Minimise **m** - **A** **e**

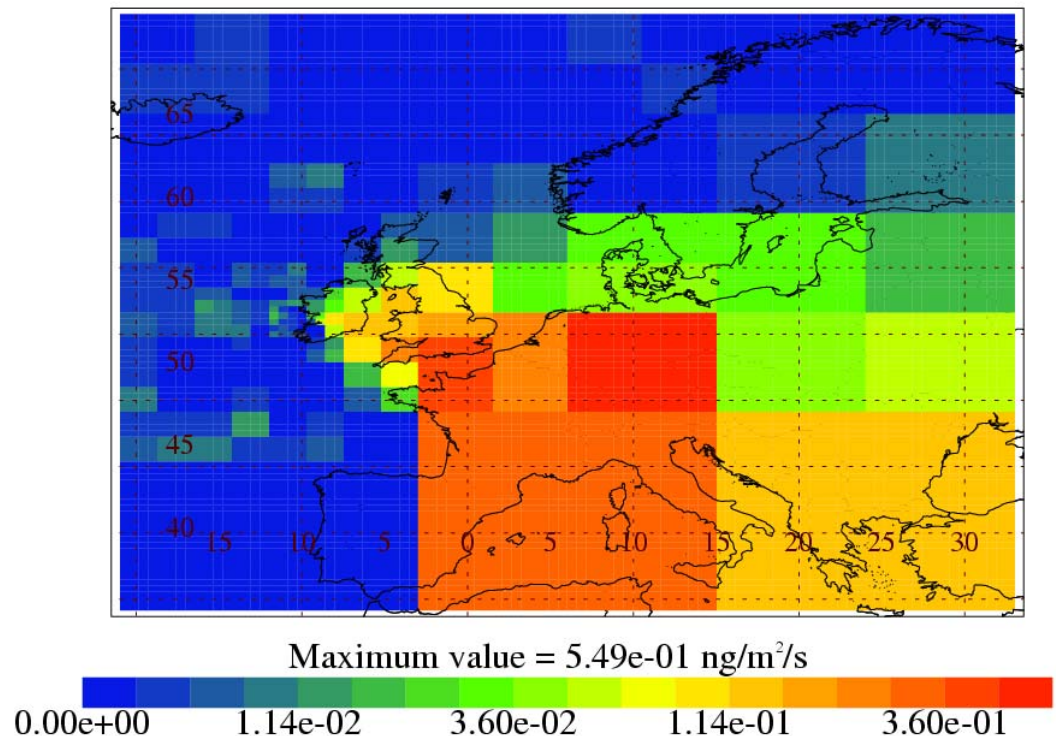
- Remove observations that have a strong **local** influence.
- Scale emissions (**iteration**) to obtain **best-fit** statistical match between model time-series and observations.
- **No prior** information – **Random** initial guess.
- Solve for each **3-yr** period iterating monthly e.g. Jan'05 – Dec'07, Feb'05 – Jan'08, ...
- Repeat **multiple** times, each time start from different random initial guess.
- Apply random '**noise**' to observations (different for each inversion).

NAME-Inversion Results: HFC-134a

Mean emission
distribution of HFC-
134a that best fits the
observations:

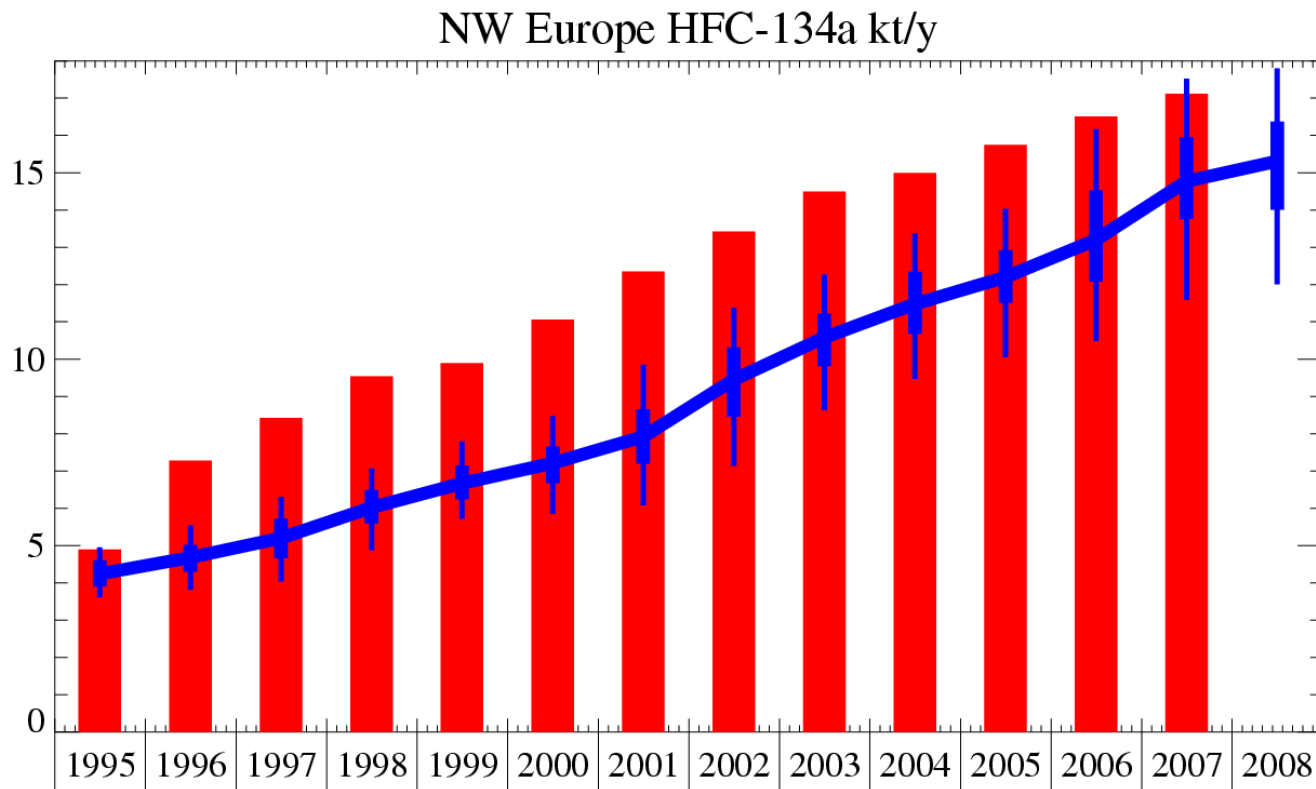
May 2006 – April 2009.

0605-0904 MapT= 42.0 Kt/y



HFC-134a emissions: N.W. Europe

Inventory estimates (UNFCCC 2009)



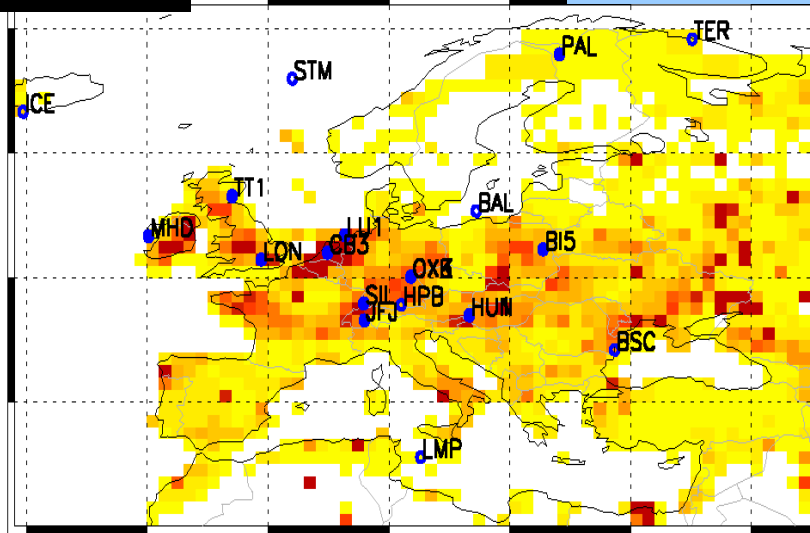
NAME-inversion estimates with uncertainty



NitroEurope: CH₄ inversion 2006

2006

JRC

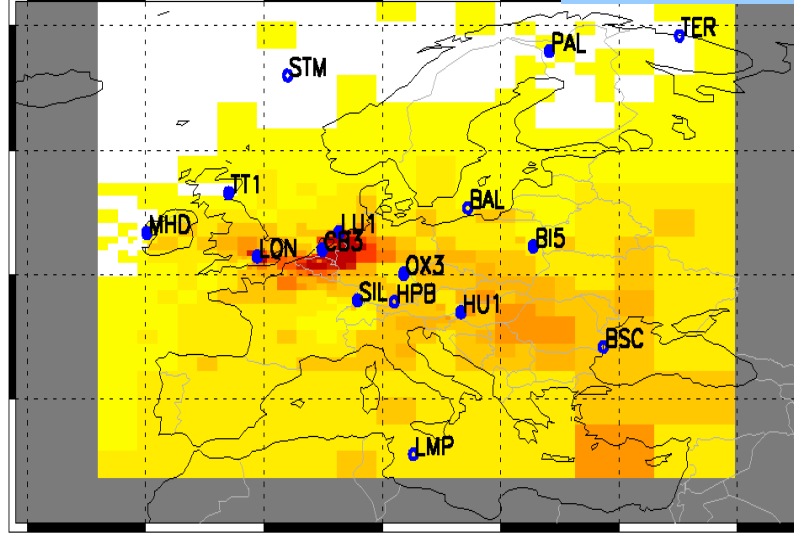


VAR_T33_CH4_25L60_tm5ei_eur_EU305_EDG040_V0205_MS_20051201_20070201

NAME

2006

UKMET



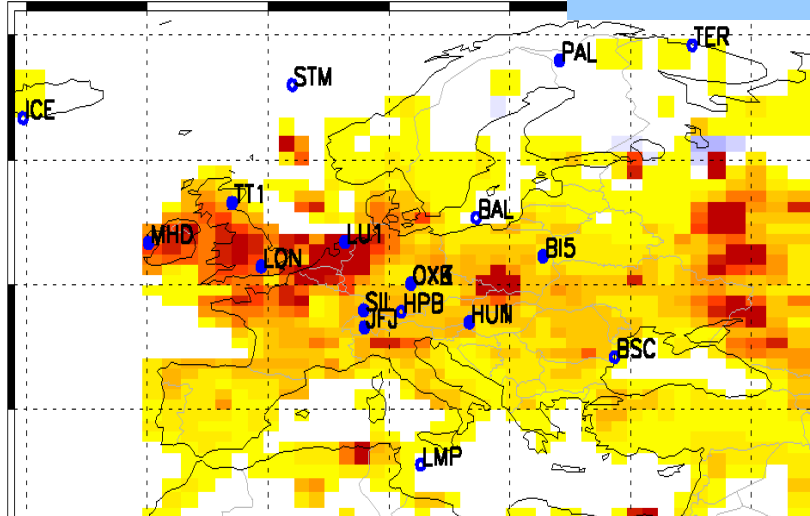
ch4_j_p_2006_mean

35.7
21.4
7.1
-7.1
-21.4
-35.7
CH₄ emission [mg CH₄ / m² / day]

LMDZ

2006

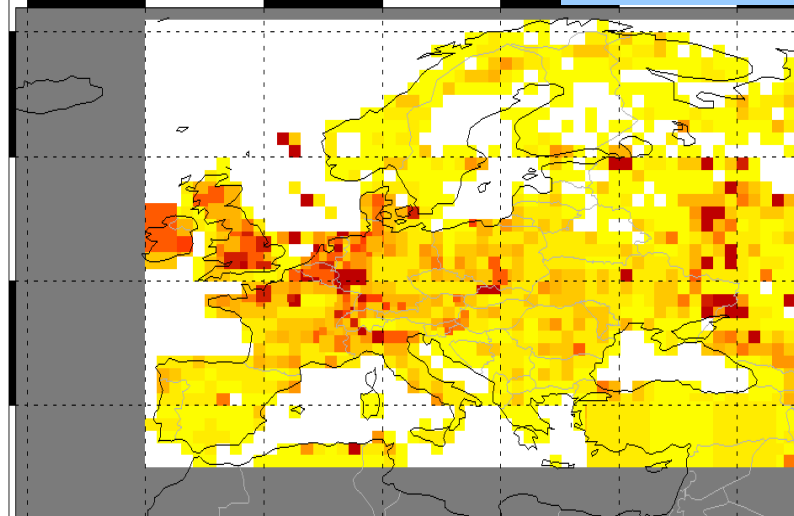
LSCE



COMET

2006

ECN



ml.asc

35.7
21.4
7.1
-7.1
-21.4
-35.7
CH₄ emission [mg CH₄ / m² / day]

Observations: ECN, HMS, UEDIN, CIO-RUG, RHUL, FMI, UBA(D), EMPA, AGAGE, ENEA, NOAA

preliminary results

Conclusions

- Inversion models can be used to estimate regional emissions.
 - Compare with existing inventories.
 - Investigate compliance (verification).
- Important issues to consider:
 - Statistical measures, a priori knowledge, dispersion and meteorological models, location of measurement station, local emissions, observations data selection.
 - Uncertainty in methodology.