

Inverse modeling of CO emissions

With a focus on biomass burning



Pim Hooghiemstra & Maarten Krol

Outline

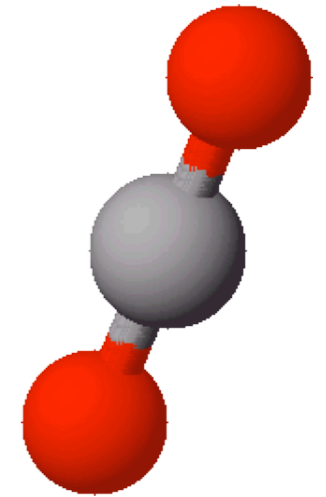
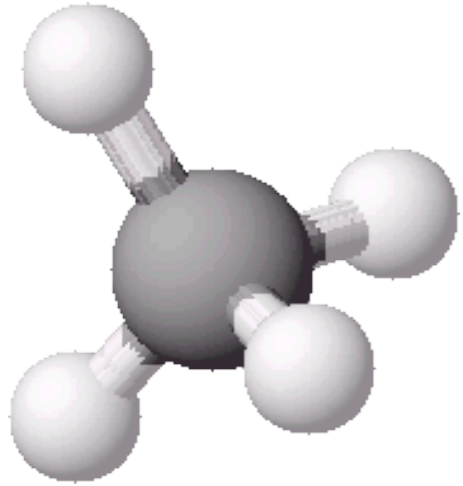
- CO from biomass burning (COBB):
Introduction to the project
- 4DVAR - Optimization of CO emissions
- Results and experiments
- Conclusions & Outline

Project Goal:

Study the magnitude, trend and variability in biomass burning.

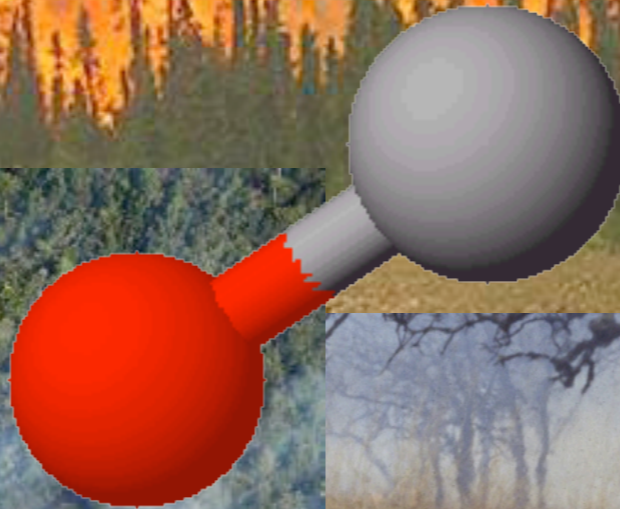
Fire emitted species: CO₂, CH₄ (greenhouse gases) CO, NO_x, NMVOCs, aerosols

Boreal forest fire



Savannah fire

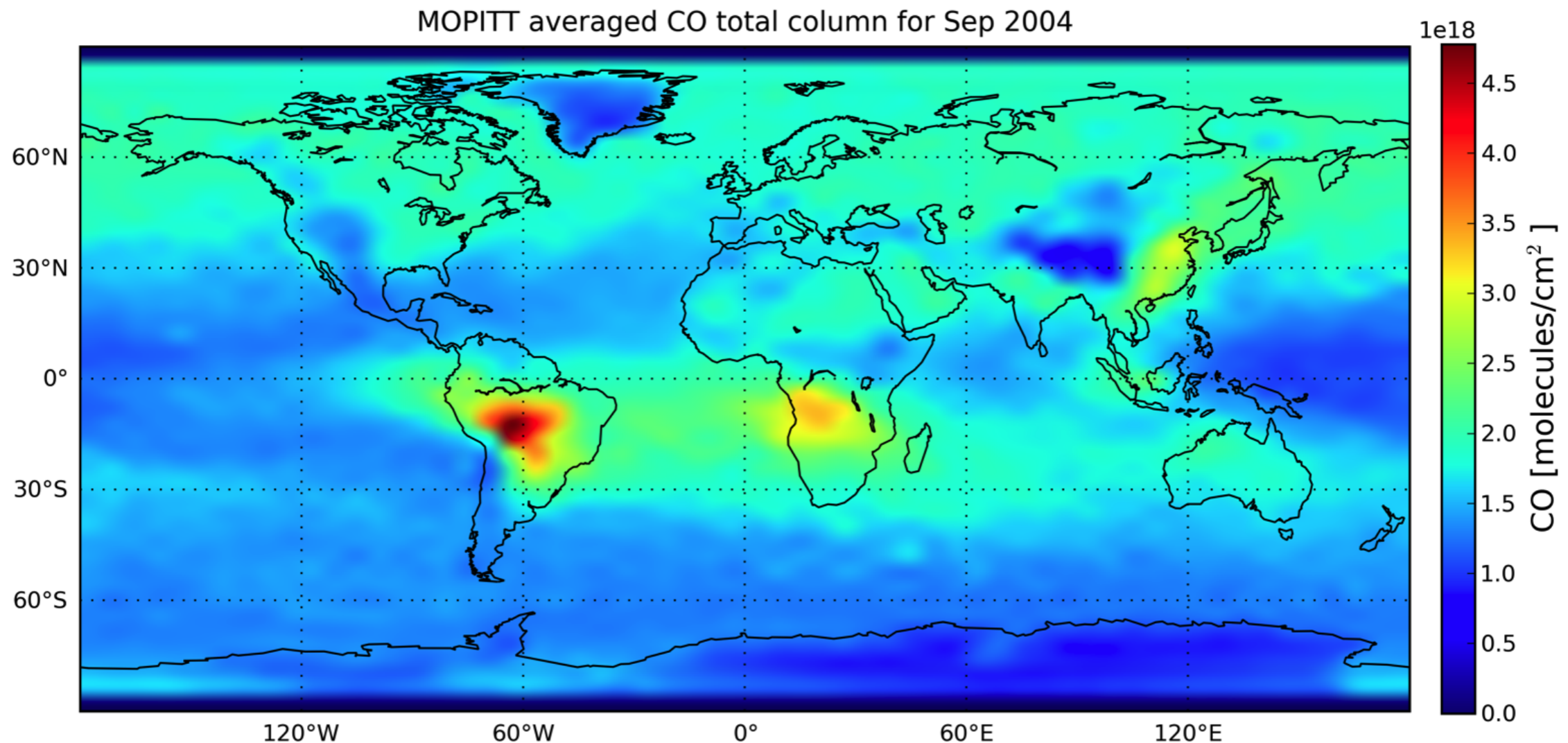
Deforestation fire



Why do we use CO as a tracer?

CO: 2 month lifetime meaning

- concentration gradient due to emissions
- intercontinental transport of CO



CO: sources & sinks

- Fossil fuel & biofuel combustion
 - Biomass burning
-

2D emissions

Annual emissions = 1000-1500 Tg CO

- Oxidation of methane
 - Oxidation of NMVOCs
-

3D emissions

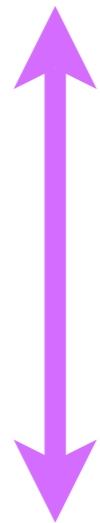
Annual emissions = 1000-1500 Tg CO

Oxidation of CO by OH main sink: 90%
Additional 10% by dry deposition

How do we compute emissions of CO?

Bottom-up:

Using burnt area, fuel loads, emission factors and upscaling



Large discrepancies

Top-down:

Using atmospheric measurements to constrain emissions

Top-down approach

“Adjust emissions in such a way that the misfit between the model and observations is minimal”

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from a bottom-up inventory

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Chemistry Transport Model (CTM):
TM5

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surface sites, ship, aircraft, satellite

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Chemistry Transport Model (CTM):

TM5

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surface sites, ship, aircraft, satellite

Find the vector \mathbf{x} that minimizes J :

$$\mathcal{J}(\mathbf{x}) = (\mathbf{y} - \mathbf{H}\mathbf{x})^\top \mathbf{R}^{-1} (\mathbf{y} - \mathbf{H}\mathbf{x})$$

Problem of not enough measurements...

Far less measurements available than variables to optimize:

Problem is ill-conditioned.

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$$\mathcal{J}(\mathbf{x}) = \frac{1}{2} (\mathbf{x} - \mathbf{x}_b)^\top \mathbf{B}^{-1} (\mathbf{x} - \mathbf{x}_b) + \frac{1}{2} (\mathbf{y} - \mathbf{H}\mathbf{x})^\top \mathbf{R}^{-1} (\mathbf{y} - \mathbf{H}\mathbf{x})$$

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\mathbf{x} = unknown emissions

\mathbf{x}_b = prior emissions

\mathbf{B} = a priori error covariance matrix

\mathbf{y} = observations

\mathbf{H} = TM5 model

\mathbf{R} = observation error covariance matrix

- Run the model with prior emissions ($\mathbf{H}\mathbf{x}_b$) and compare the output with the observations \mathbf{y} .

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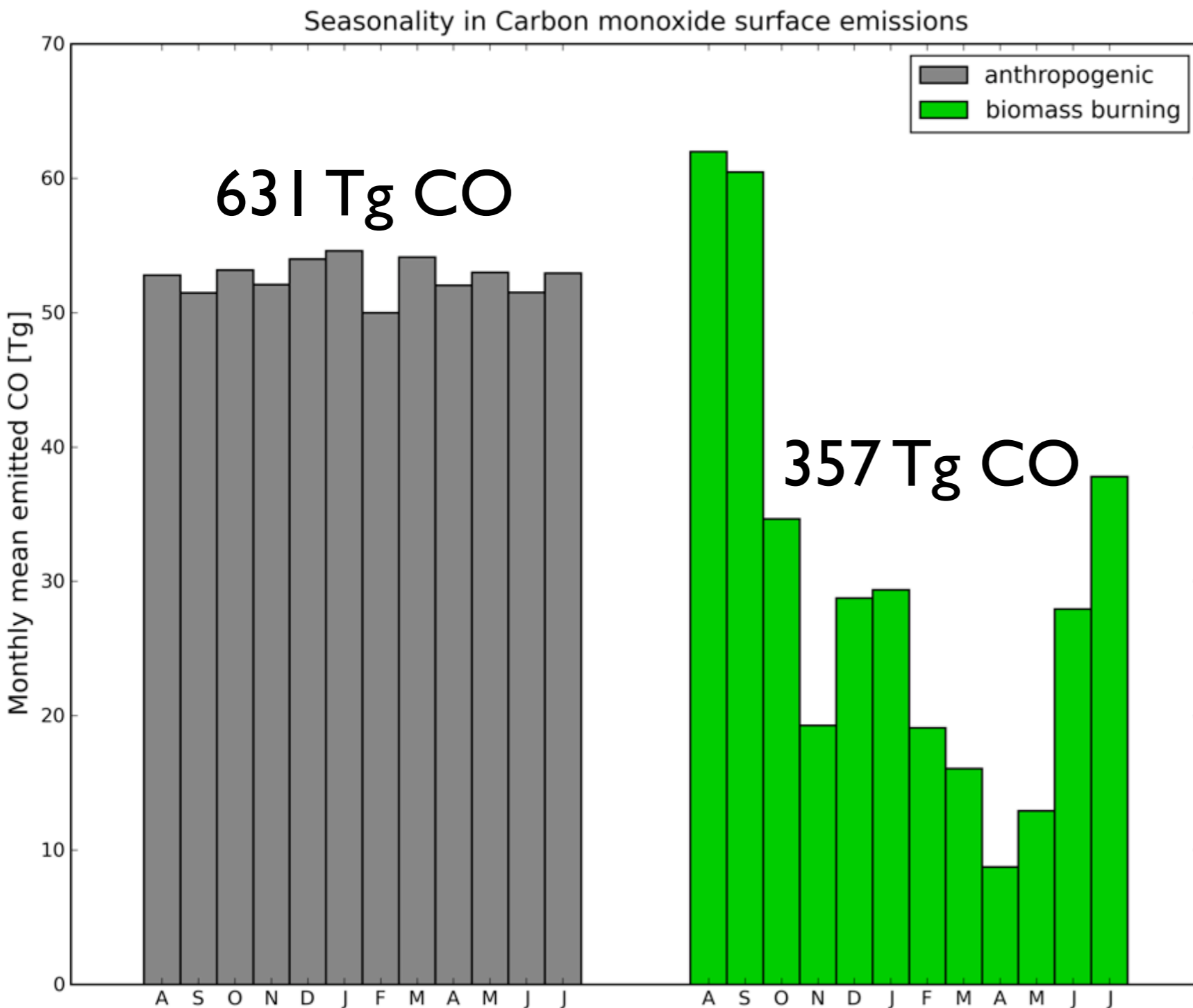
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- **Validate results: compare optimized emissions with a set of independent observations.**

Run the model with prior emissions (Hx_b) and compare the output with the observations y .

Prior emissions:



Direct sources:

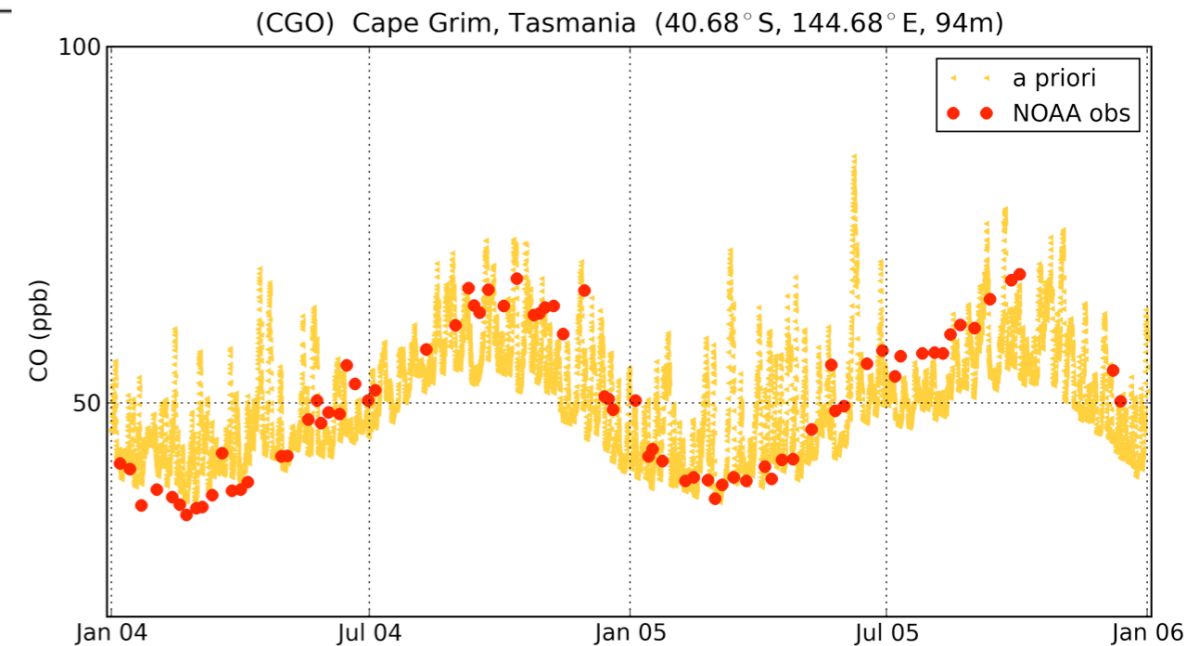
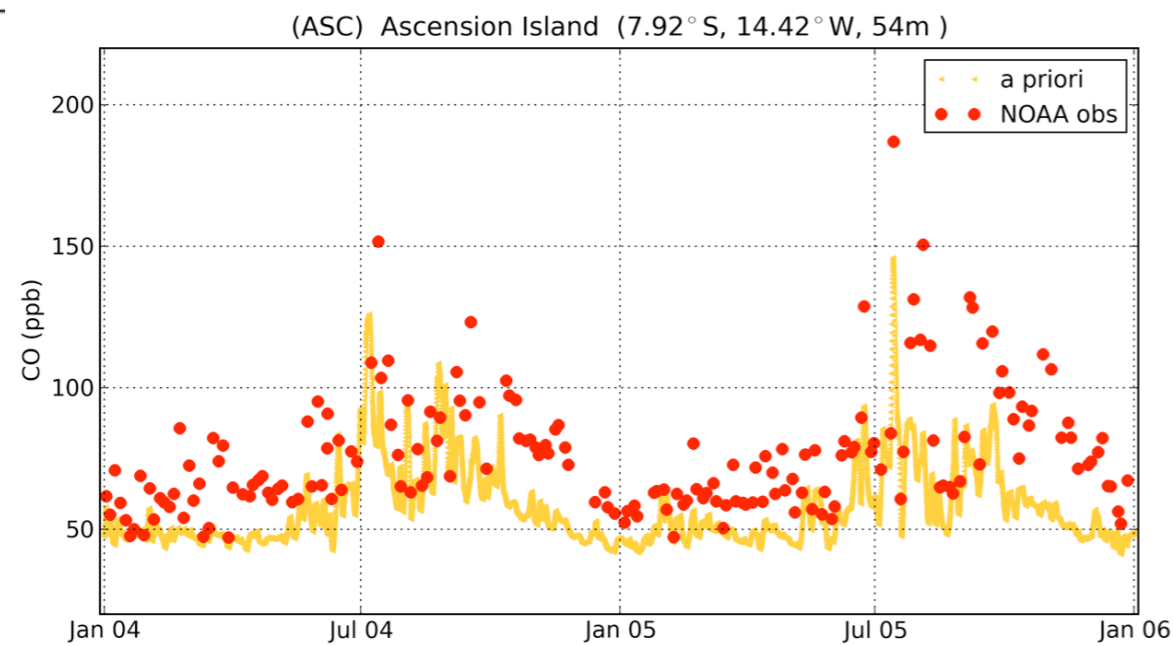
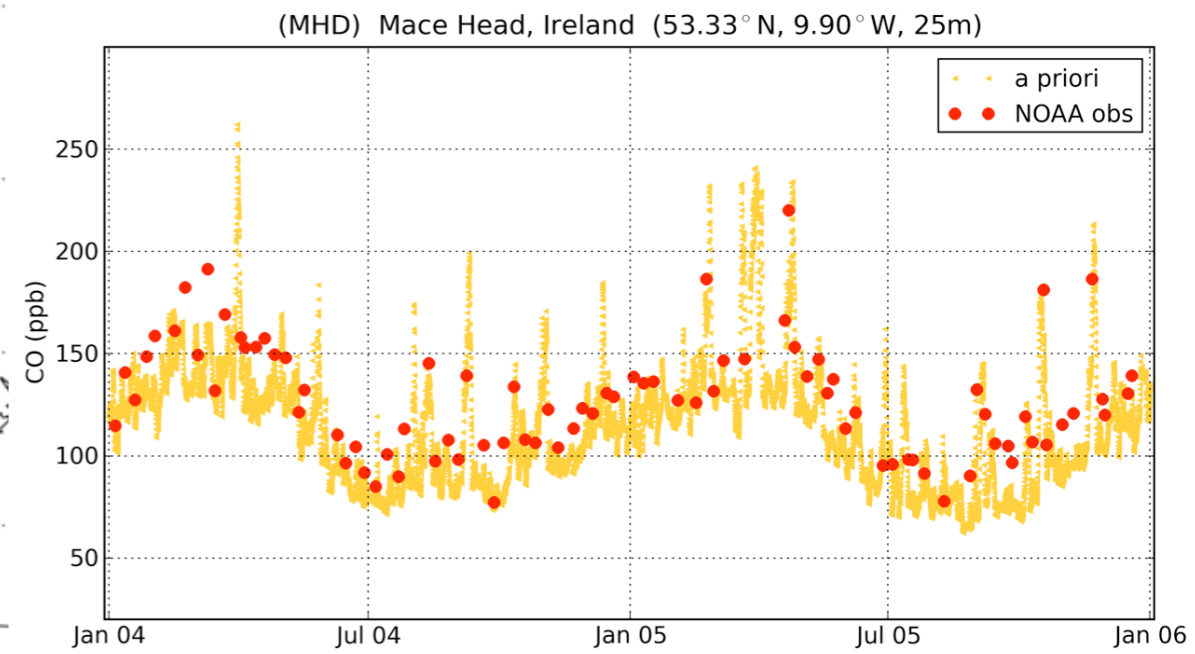
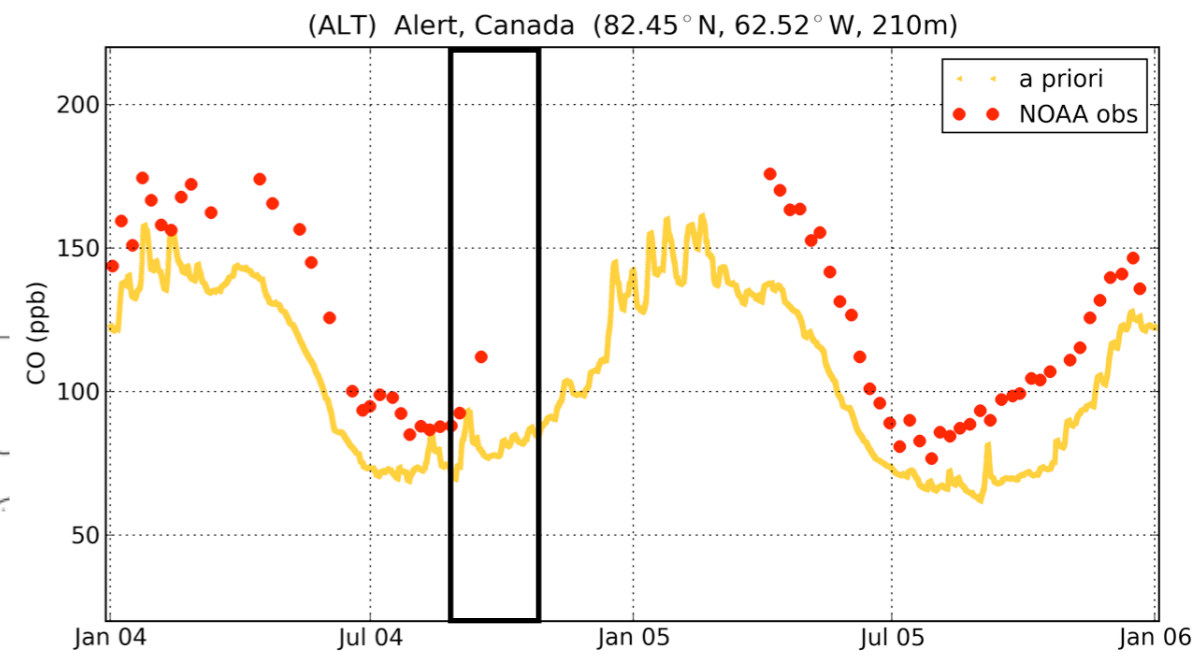
- Fossil fuel/biofuel combustion (anthropogenic) EDGARv3.2
- Biomass burning GFED3

Indirect sources:

- CO from oxidation of CH₄
- CO from oxidation of NMVOC (not yet)

Forward simulation 2004-2006

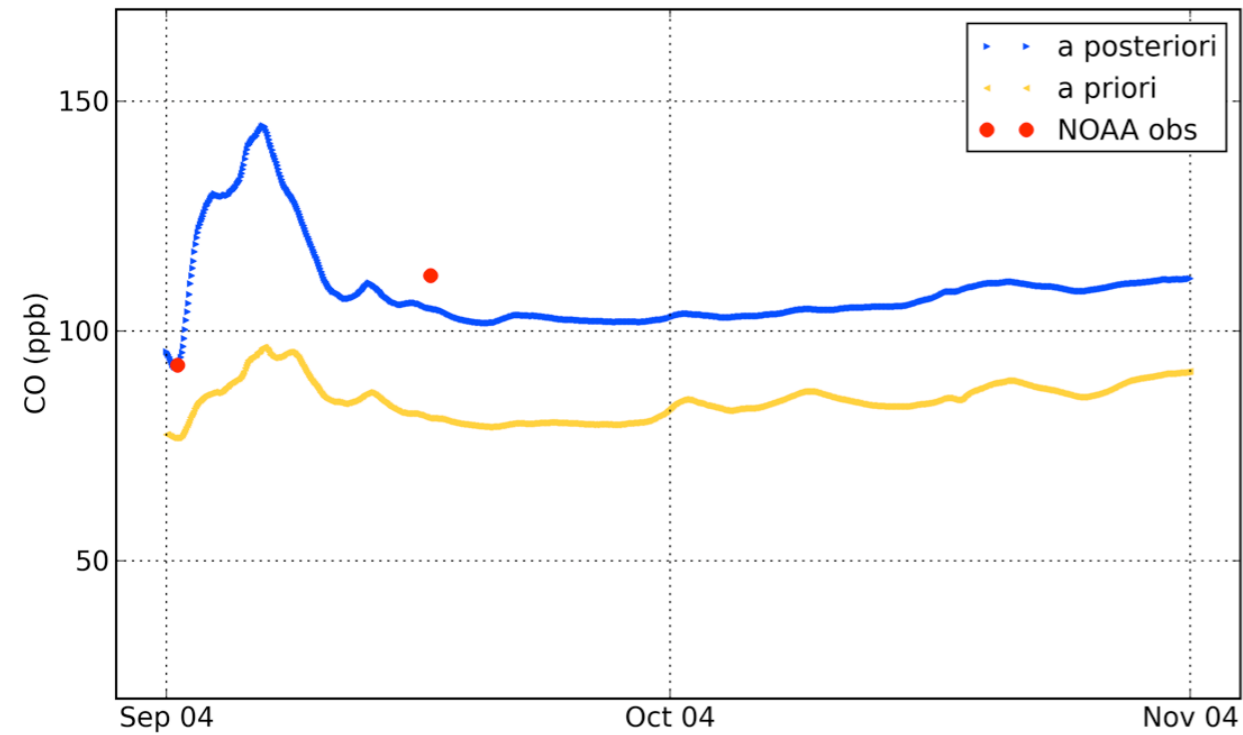
Surface network stations



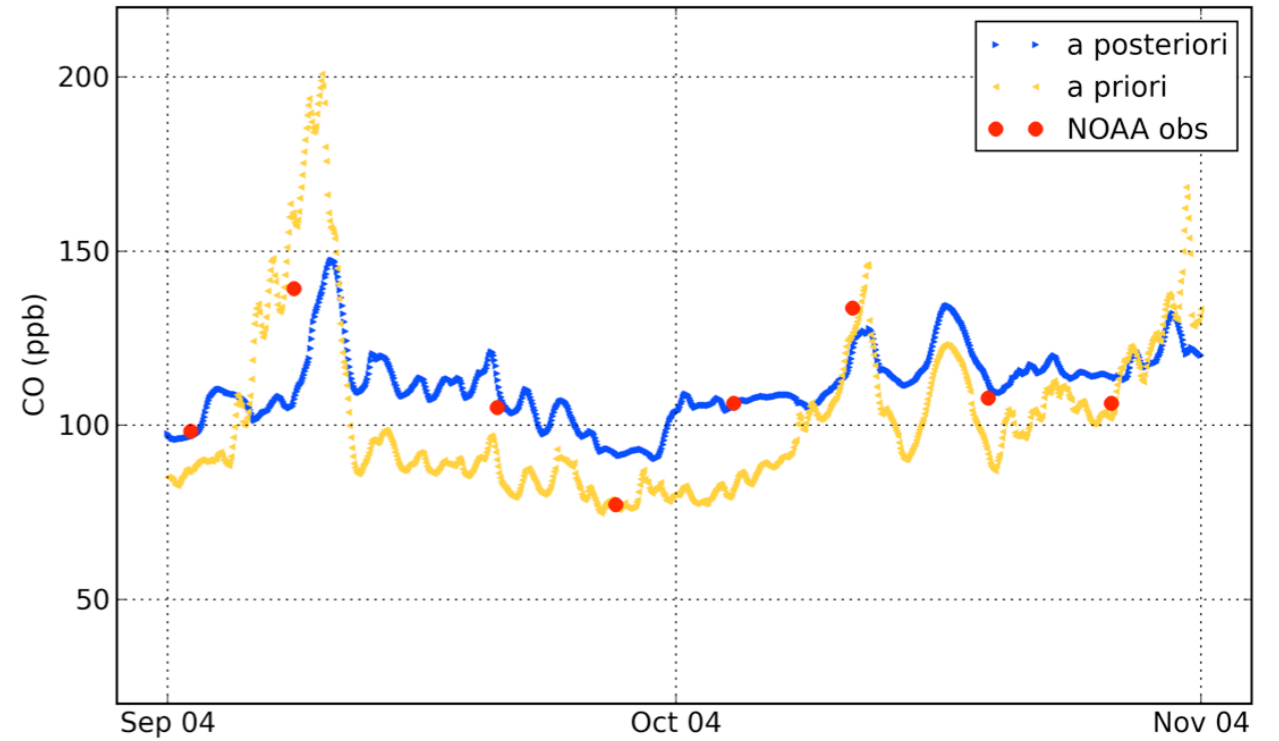
Optimize emissions iteratively: start with $x=x_b$
change x to find minimum of $J(x)$

Improved fit with observations

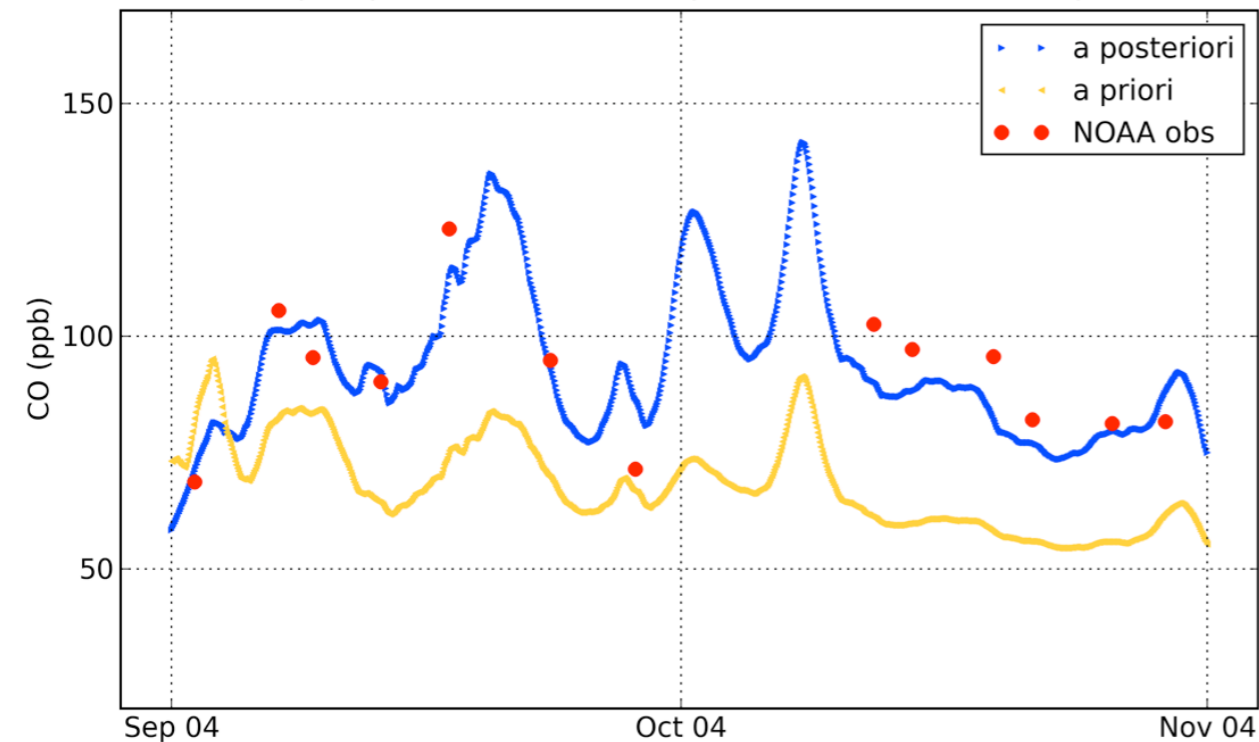
(ALT) Alert, Canada (82.45° N, 62.52° W, 210m)



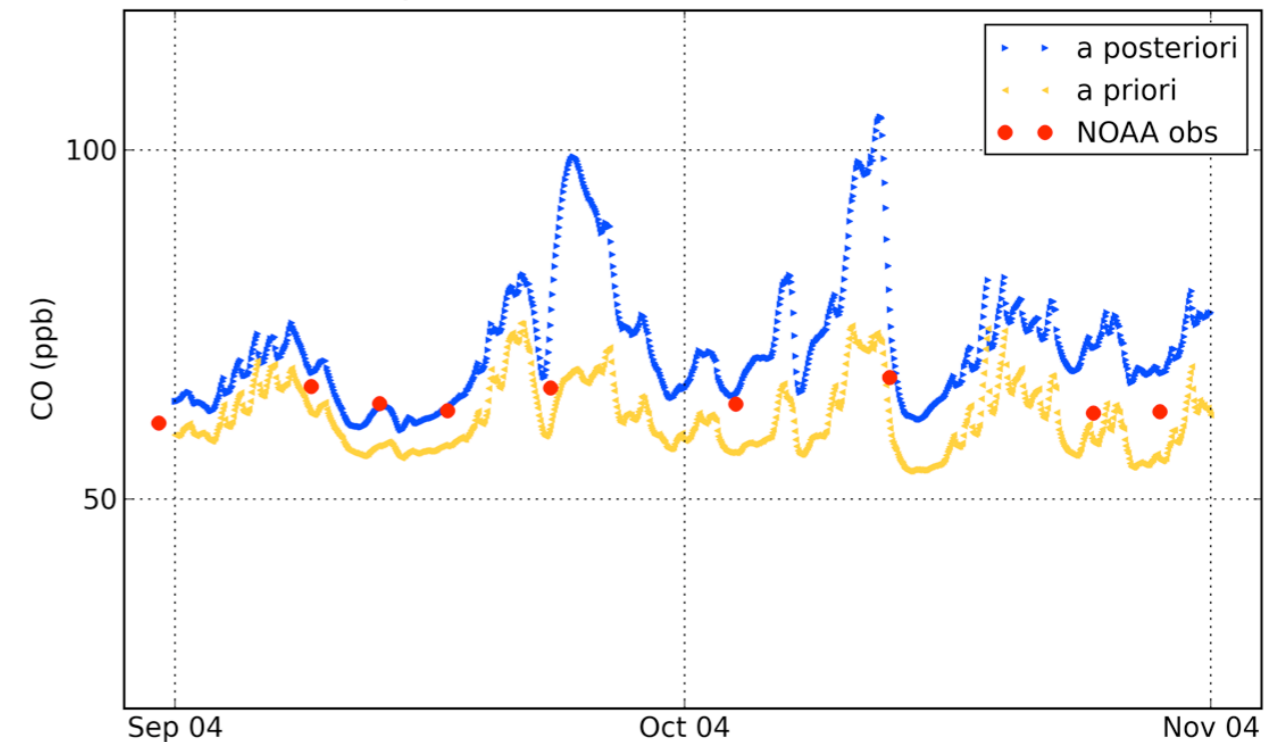
(MHD) Mace Head, Ireland (53.33° N, 9.90° W, 25m)



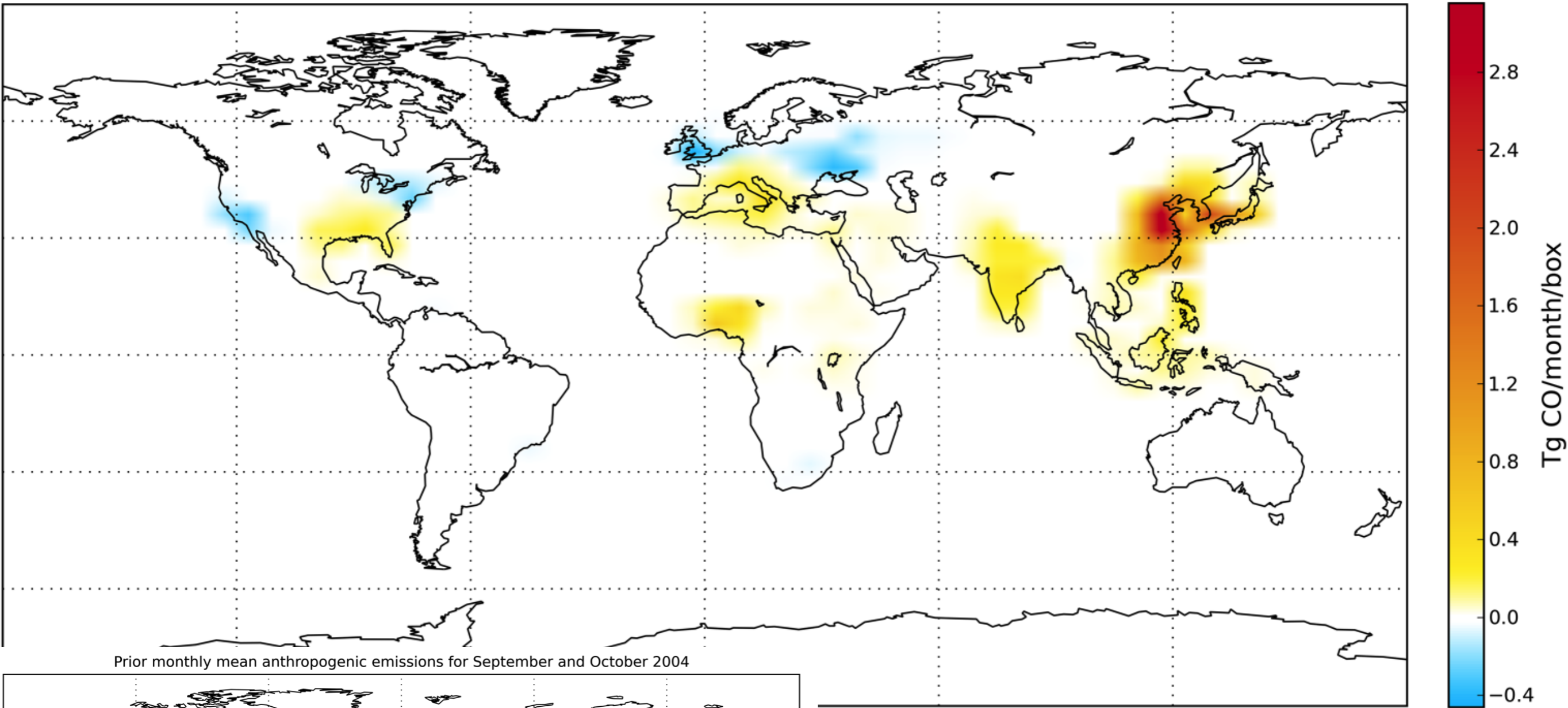
(ASC) Ascension Island (7.92° S, 14.42° W, 54m)



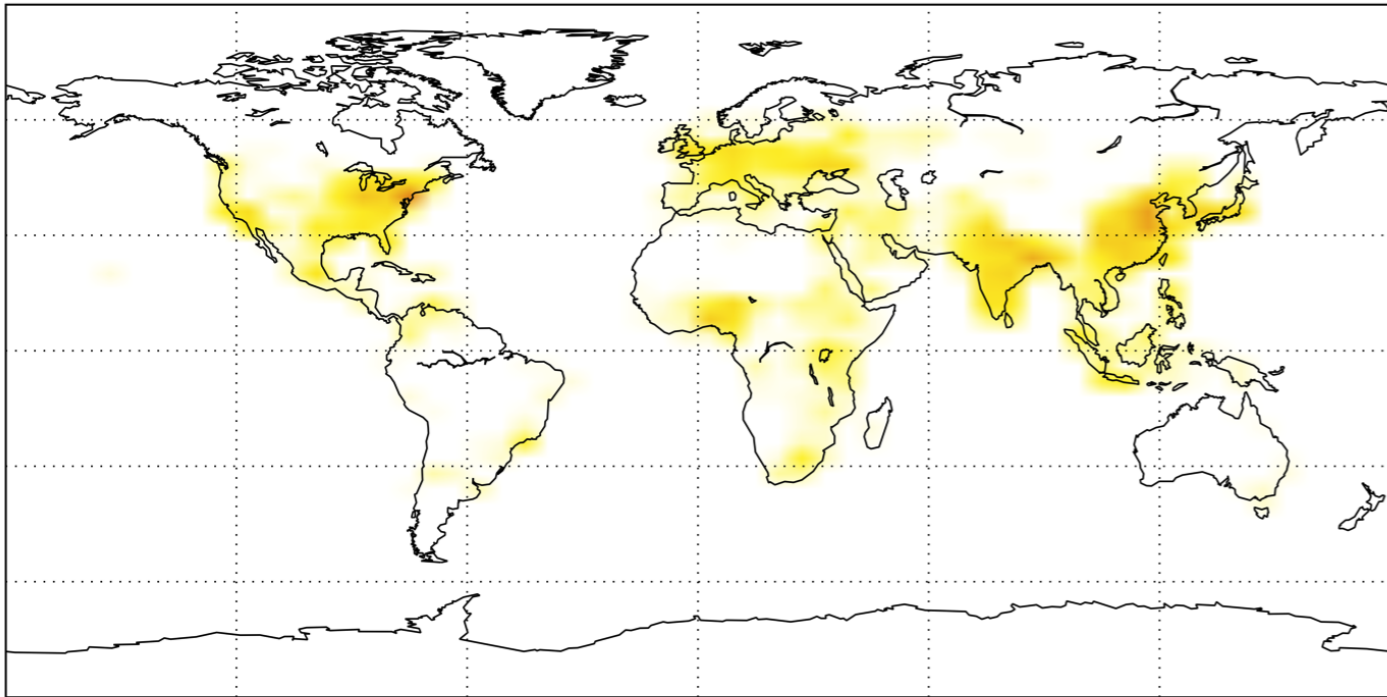
(CGO) Cape Grim, Tasmania (40.68° S, 144.68° E, 94m)



Monthly mean increment in anthropogenic emissions for September and October 2004



Prior monthly mean anthropogenic emissions for September and October 2004



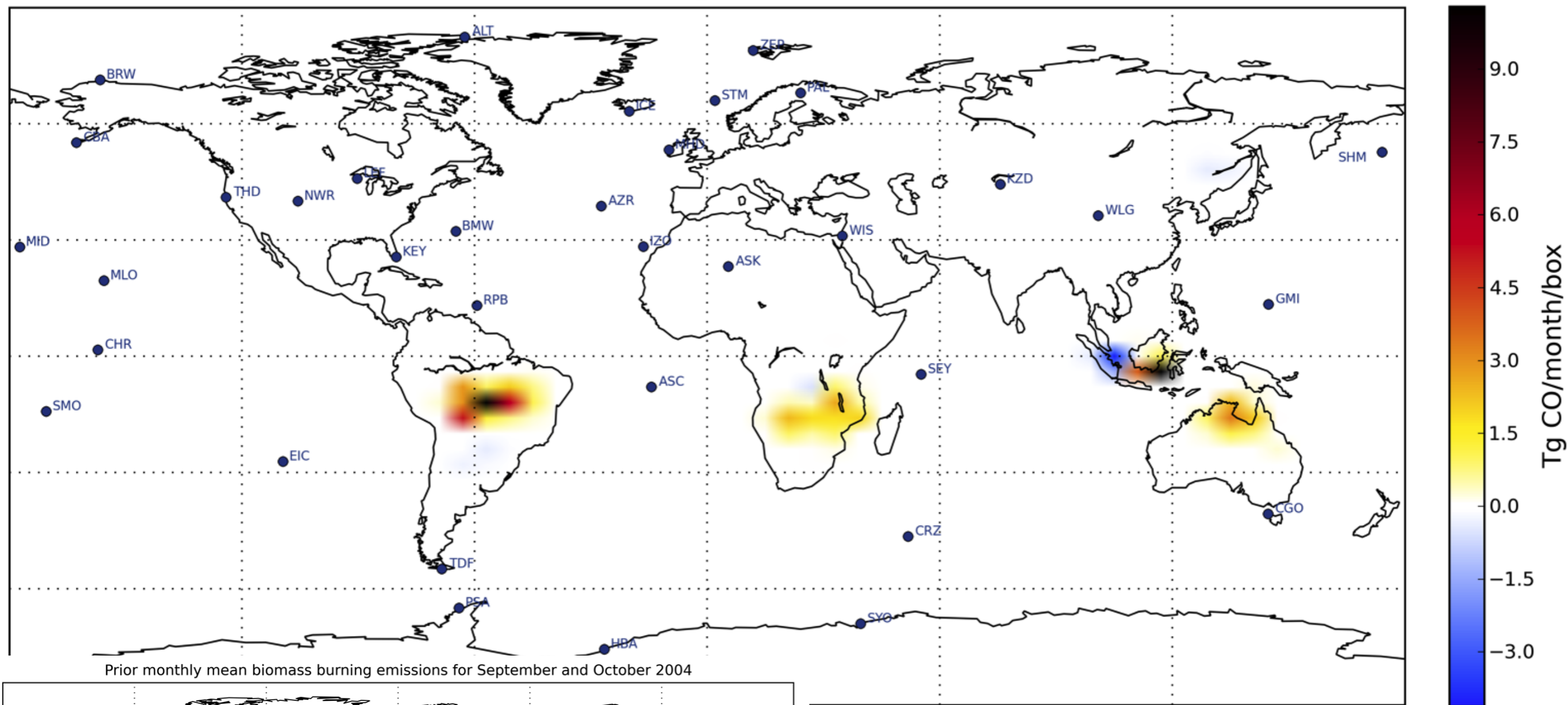
Top: anthropogenic emission increment

Bottom: prior anthropogenic emission

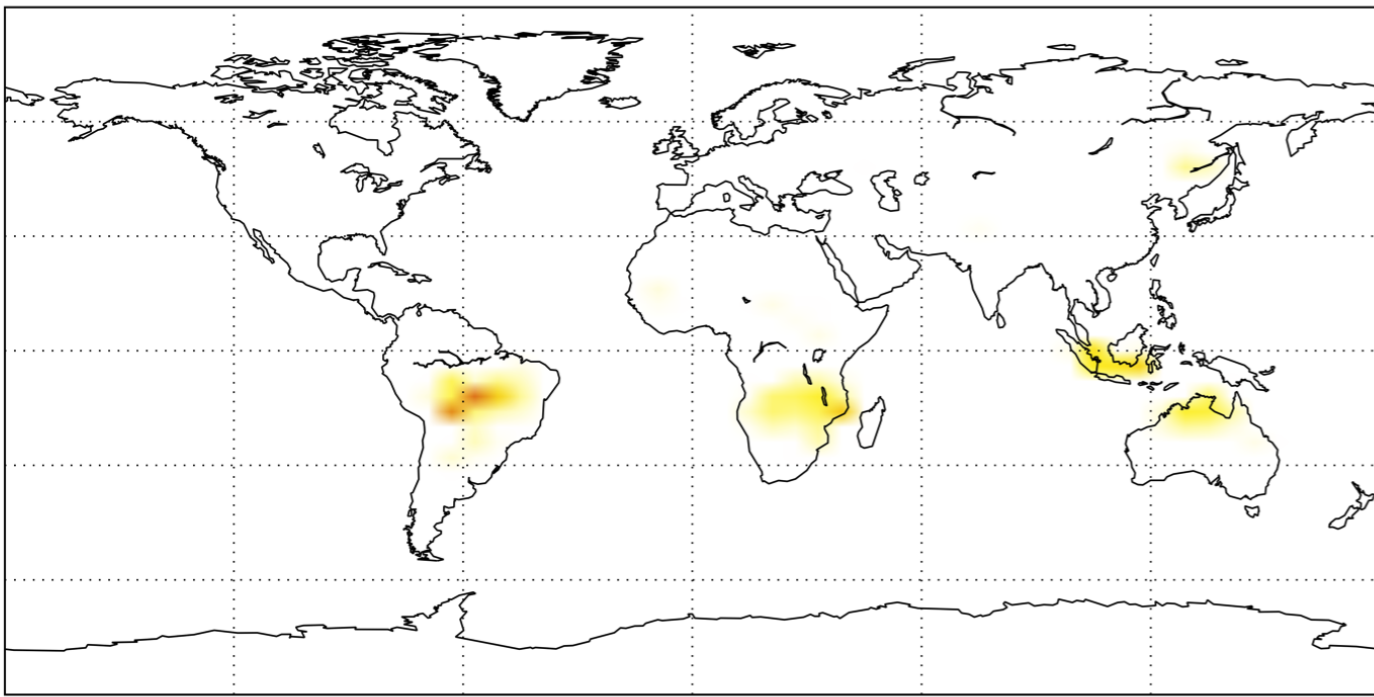
60°E

120°E

Monthly mean increment in biomass burning emissions for September and October 2004



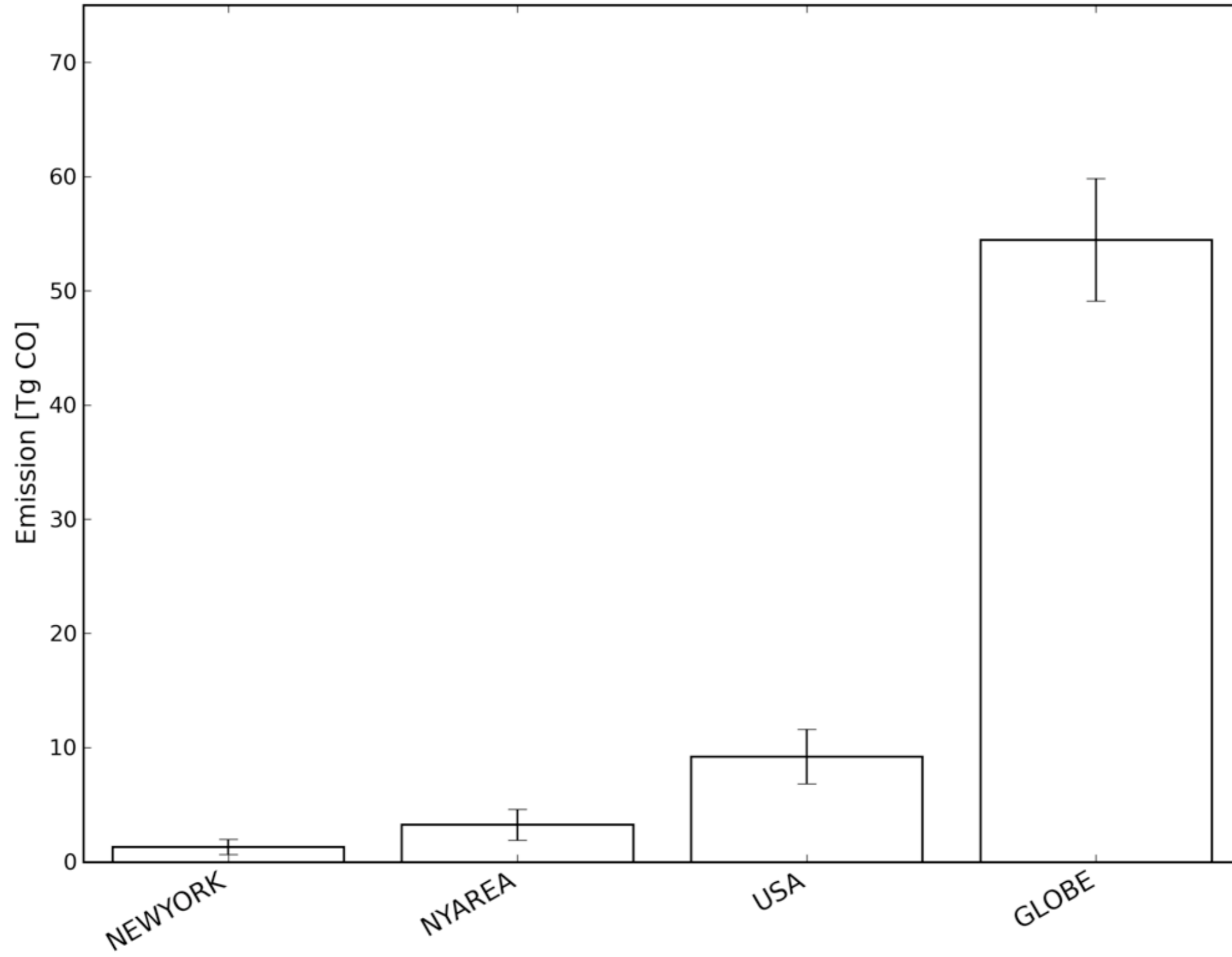
Prior monthly mean biomass burning emissions for September and October 2004



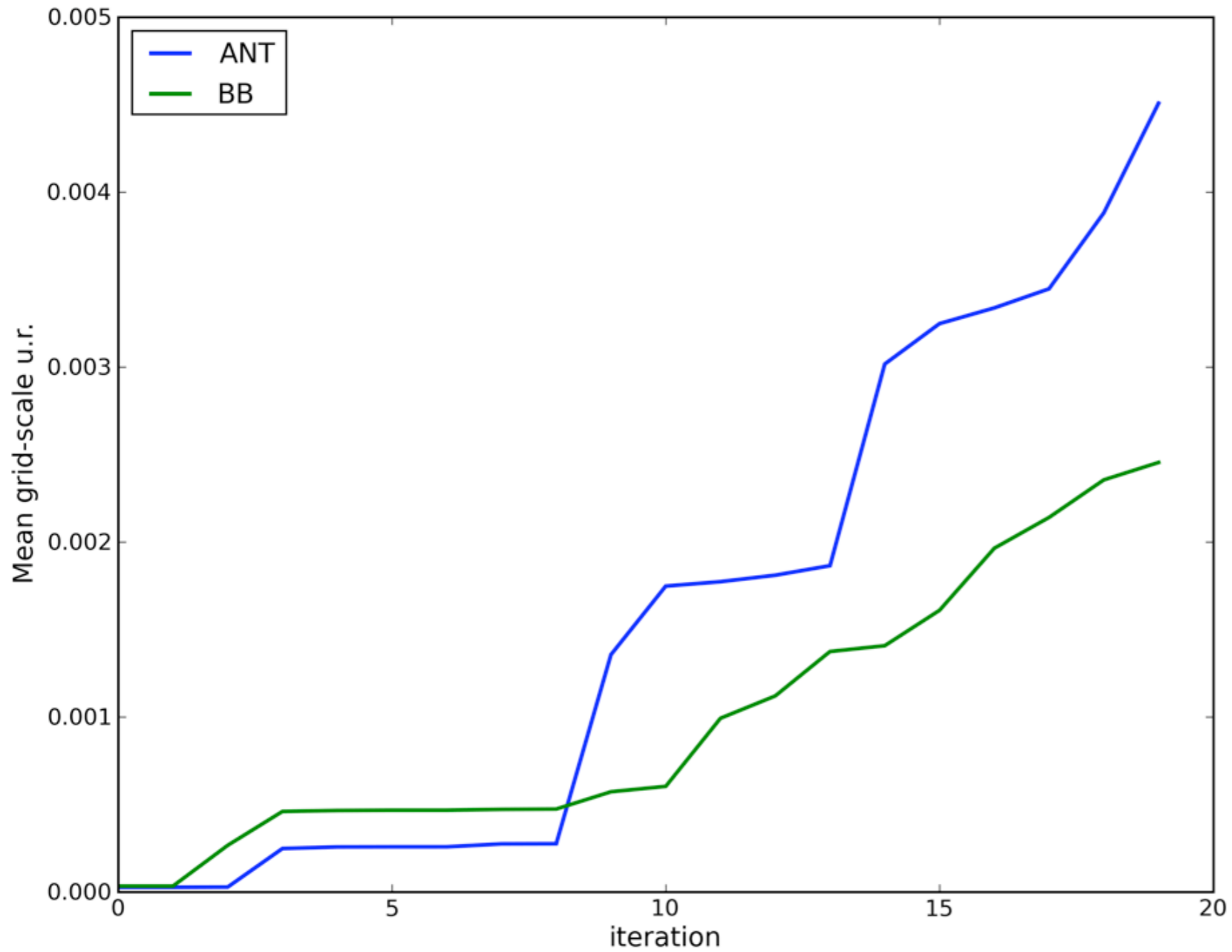
Uncertainty reduction: What about the errors?

Aggregated prior emission

Aggregated anthropogenic emissions January 2004

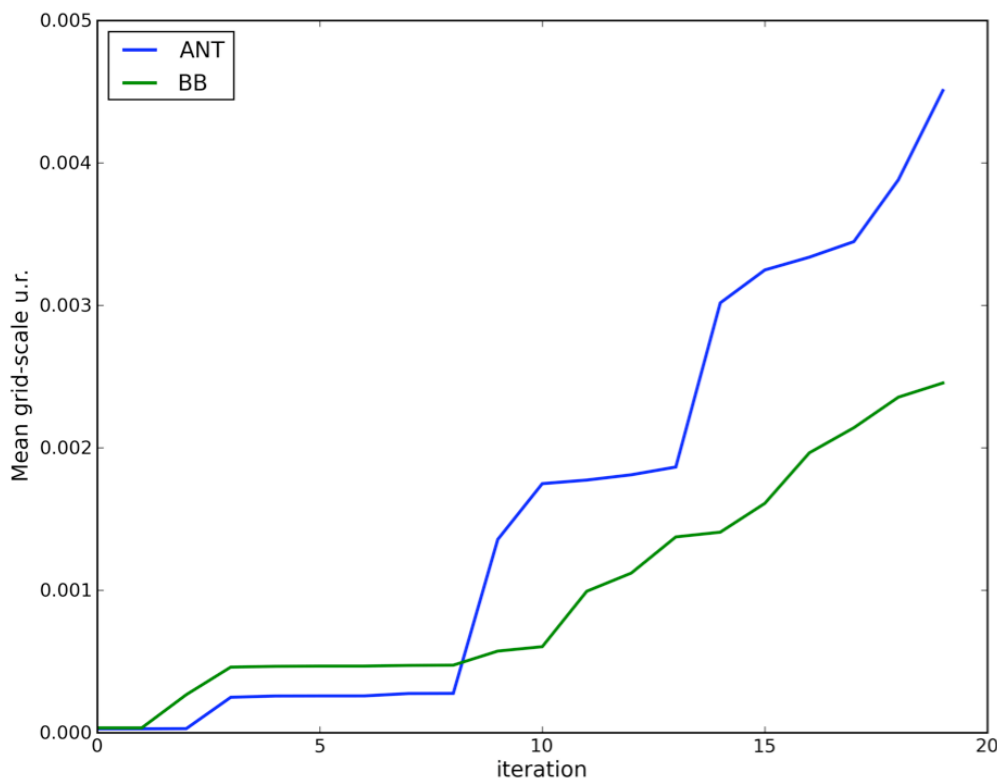


Uncertainty reduction

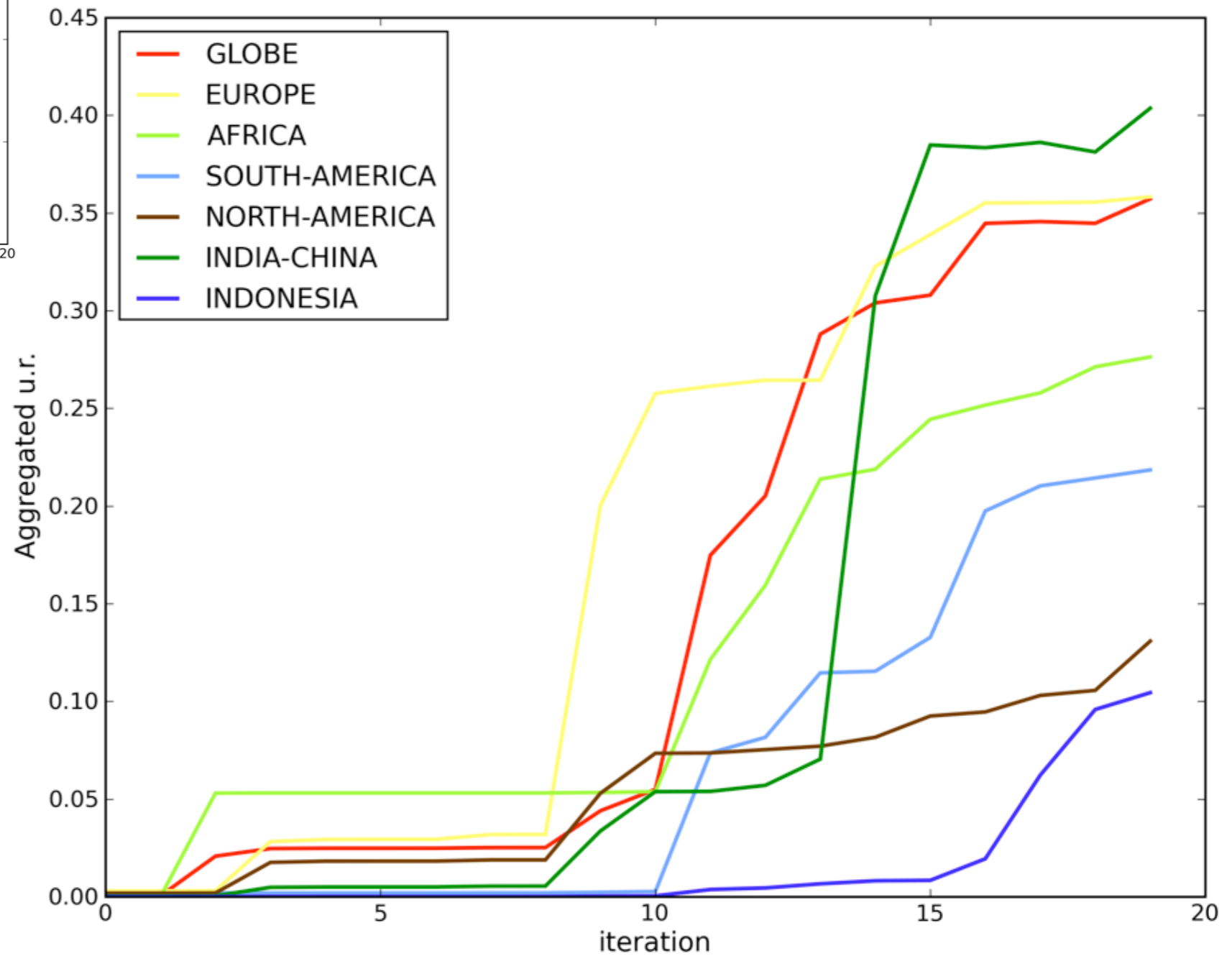


Convergence off 99% only

Uncertainty reduction

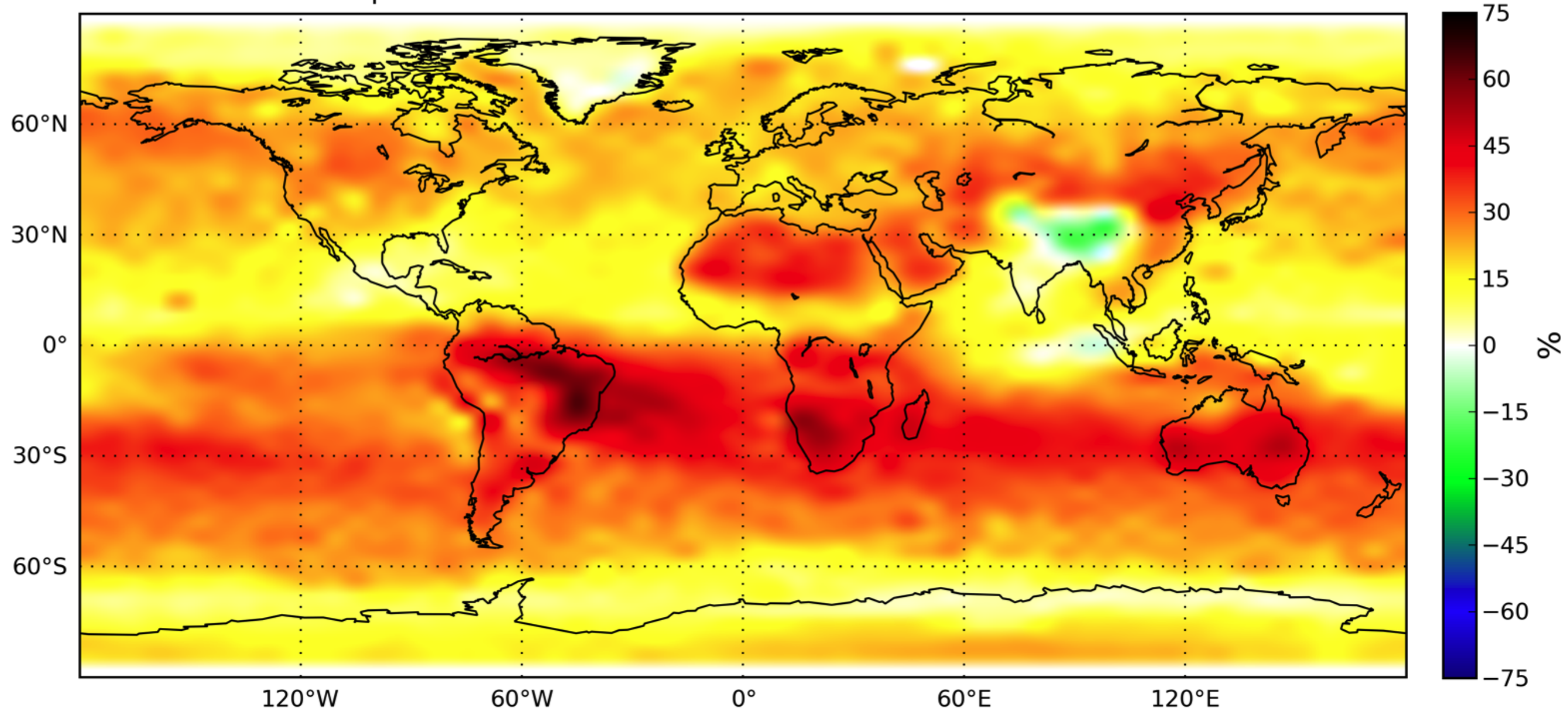


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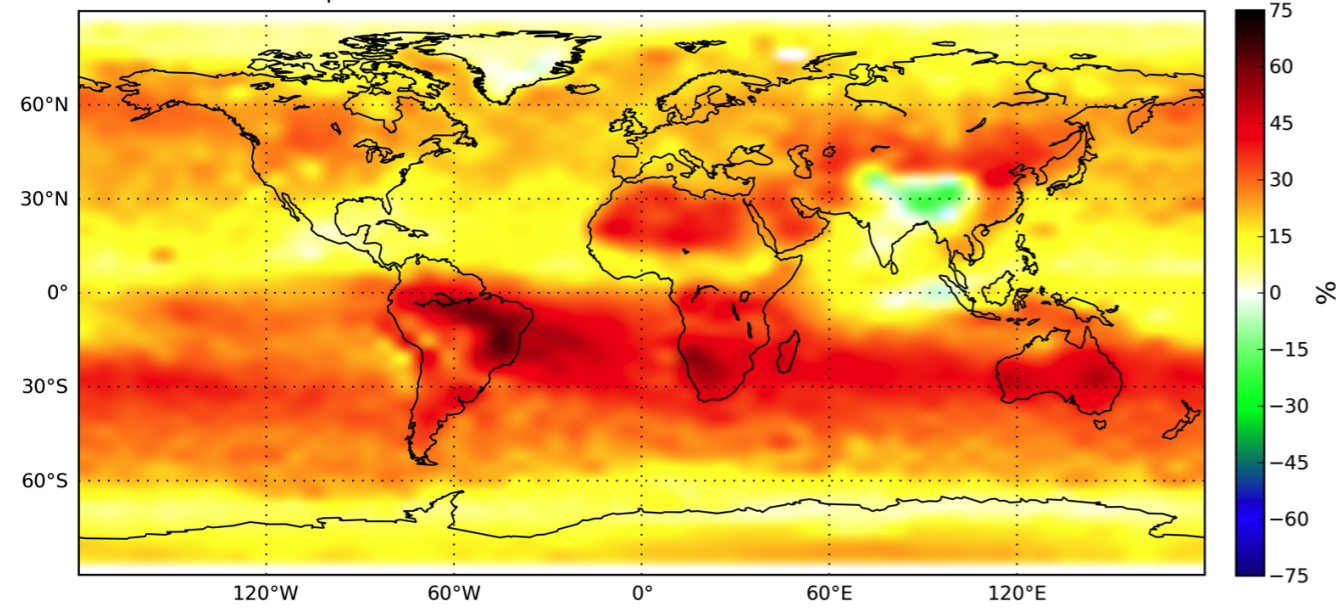
Validate results: compare optimized emissions with a set of independent observations

A priori difference in CO total column for MOPITT and TM5



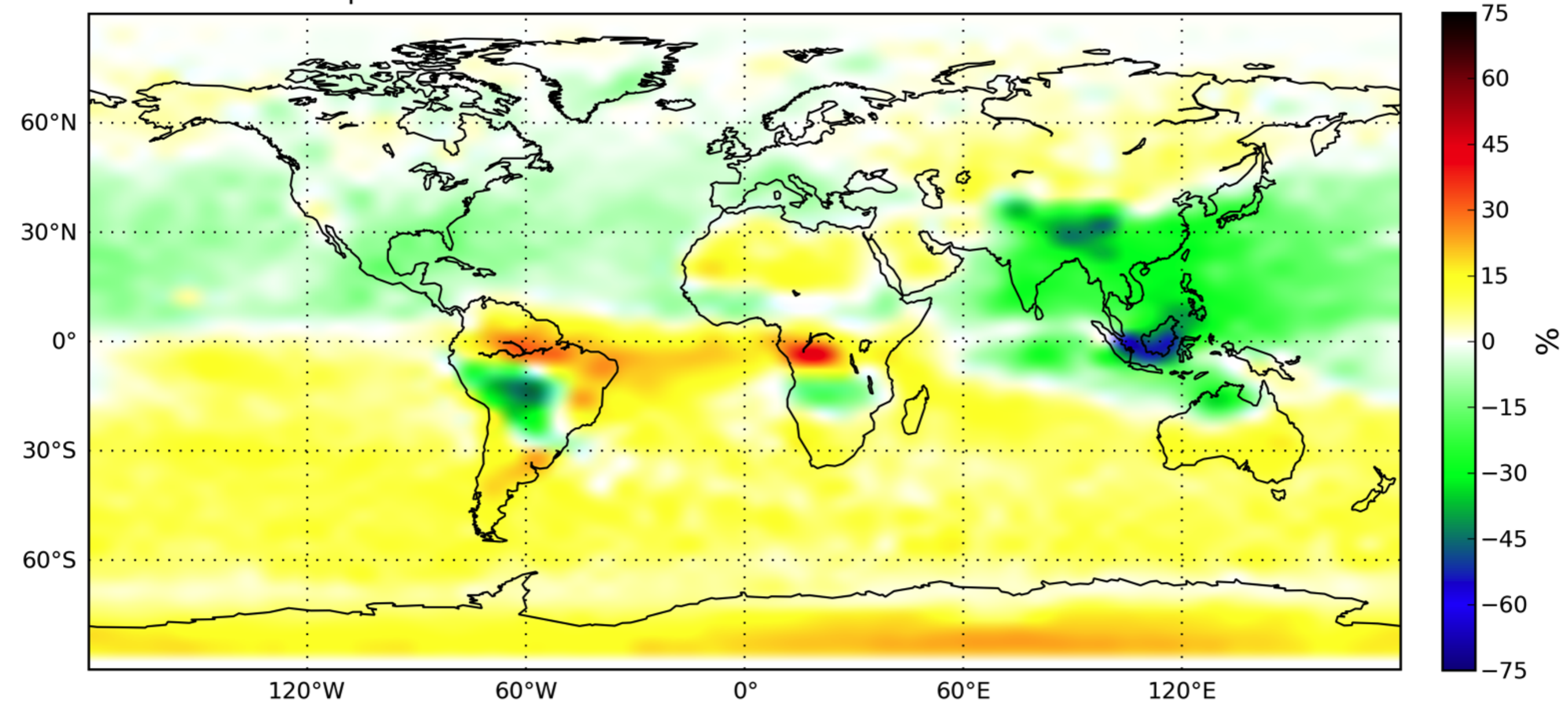
2-monthly mean CO total column: difference between MOPITT and TM5 (%)

A priori difference in CO total column for MOPITT and TM5



Global mean difference decreases from 22% to 8%, large local differences remain.

A posteriori difference in CO total column for MOPITT and TM5



CONCLUSIONS

- TM5-CO version nearly complete, CO from NMVOC to be implemented.
- Forward model agrees well with observations, too low on NH up to 25%.
- Inversion using station data improve agreement between observations and the model.
- Validation with MOPITT V4 shows that the optimized emissions are better in line, but large local differences remain.

Next...

- Include CO source from NMVOC oxidation.
- Apply a vertical distribution of biomass burning emissions.
- Optimize emissions for 1 or 2 years.
- Use satellite data (MOPITT/SCIAMACHY) to constrain emissions more.