

Assessments of in situ and remotely sensed CO₂ observations in a Carbon Cycle Fossil Fuel Data Assimilation System to estimate fossil fuel emissions

Marko Scholze¹, T. Kaminski², P. Rayner³, M. Vossbeck², M. Buchwitz⁴, M. Reuter⁴, W. Knorr²,
H. Chen¹, A. Agusti-Panareda⁵, A. Löscher⁶, and Y. Meijer⁶

¹ Department of Physical Geography and Ecosystem Science, Lund University, Sweden

² The Inversion Lab, Hamburg, Germany

³ University of Melbourne, Australia

⁴ University of Bremen, Institute of Environmental Physics (IUP), Germany

⁵ European Centre for Medium-Range Weather Forecasts, UK

⁶ ESA, Noordwijk, The Netherlands

EGU 2020, 6 May, Vienna



iLab



ECMWF



Objectives

- Develop a coupled data assimilation system for natural and fossil fuel fluxes based on process models (CCFFDAS)
- Assess the potential of remotely sensed CO₂ observations to constrain fossil fuel CO₂ emissions for an exemplary 1-week period in 2008 in a Carbon Cycle Fossil Fuel Data Assimilation System
- Test design options for a Monitoring and Verification Support (MVS) capacity

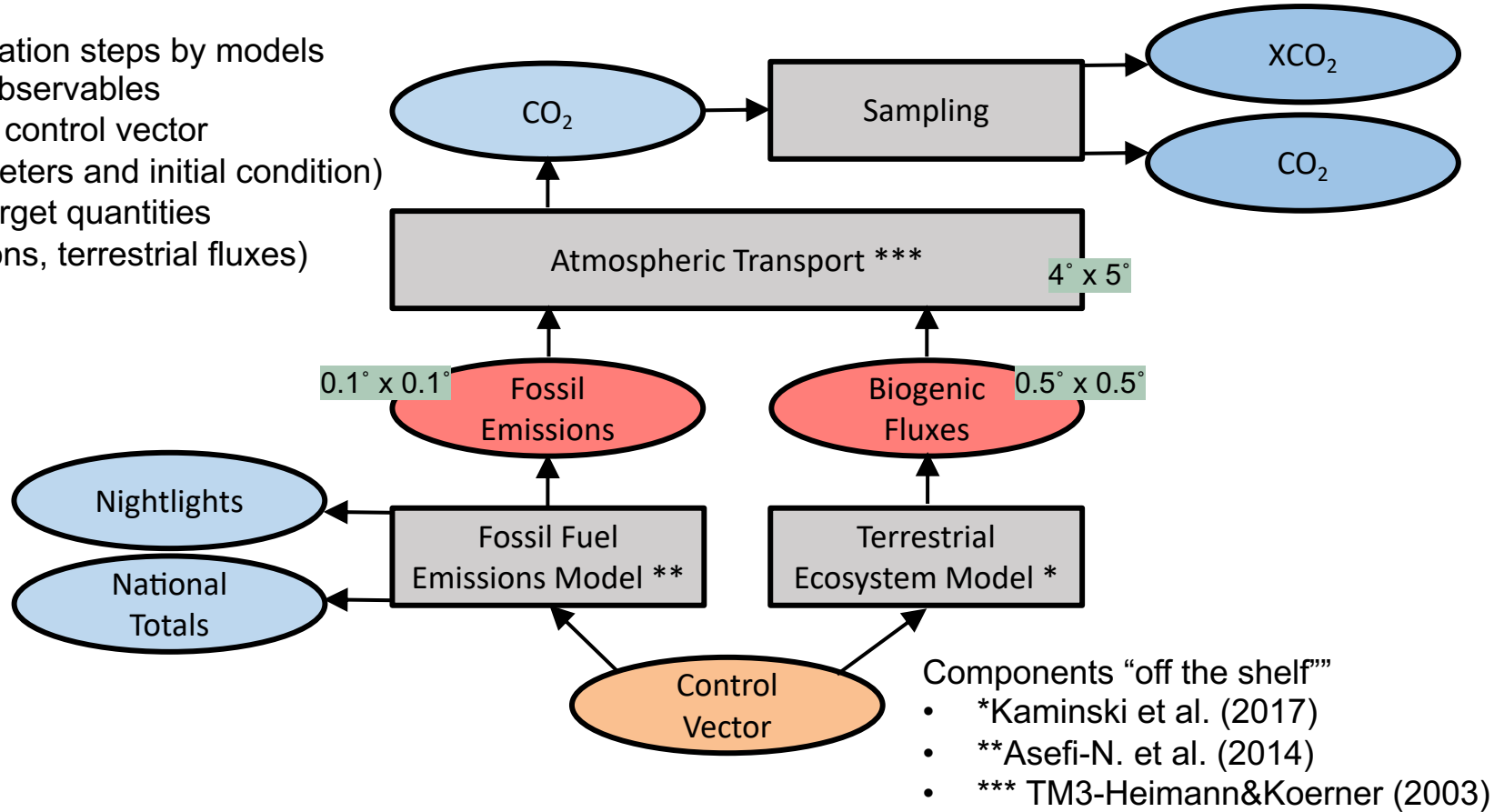
Information flow in CCFFDAS

Boxes: calculation steps by models

Blue ovals: observables

Orange oval: control vector
(model parameters and initial condition)

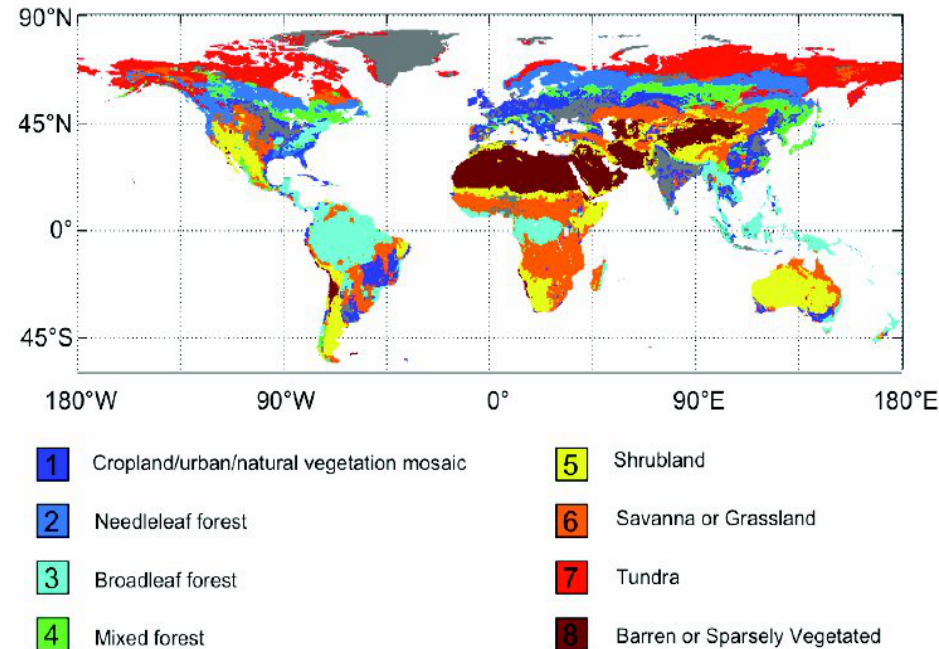
Red ovals: target quantities
(fossil emissions, terrestrial fluxes)



Natural Fluxes

Model based on concept by Knorr and Heimann (1995):

- Globally 0.5 degree
- Running from 2006-2010
- Driven by JRC-TIP FAPAR and WATCH climate
- PFT classification into 8 groups (Knorr et al. 2014)
- Process parameters (for calibration):
 - PFT specific Photosynthetic Light Use Efficiency (ϵ)
 - PFT specific Temperature dependency of RHET(Q10)
 - spatially constant initial atm. XCO₂ value
- TM3 (Heimann and Koerner) fine grid (4x5) for atmospheric transport
- Calibrated against GOSAT XCO₂(Kaminski et al., 2017).



Sectorial Fossil Fuel Emissions

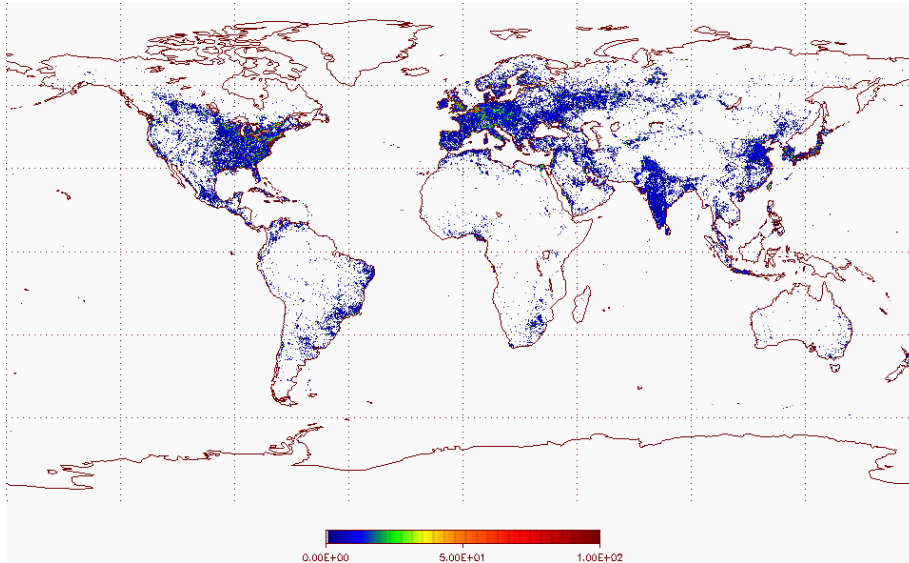
The FFDAS concept allows for making use of the different emission sectors. In this project, we distinguish between two main sectors:

- Electricity production, includes the IEA categories
 - Main activity electricity and heat production
 - Unallocated autoproducers
- Other, includes the IEA categories
 - Other energy industry own use
 - Manufacturing industries and construction
 - Transport, excluding Non-energy use in transport
 - Other

$$F(x, c) = pP(x)g(x)ef(c)$$

Kaya Identity

Observation operators and uncertainties



$$O_{NL} = npP(x)g(x)$$

n: nightlights scaling factor

p: population density scaling factor

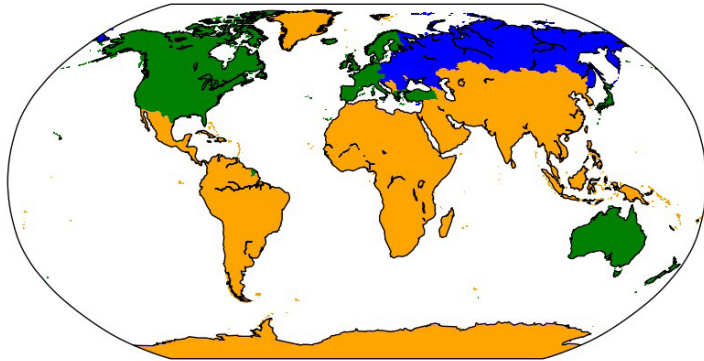
P(x): population density in each grid cell

g(x): GDP per capita in each grid cell

Nightlights
12.5% uncertainty

Nightlights from the Defense Meteorological Satellite Program, for 2008 interpolated from 2006 and 2010 data

Observation operators and uncertainties



- National total uncertainty per country-class (over all sectors)
- Distributed to sectors by partitioning variance in proportion to totals
- Not used in default experiment (weekly time scale), only for sensitivity test

$$O_{\text{elec}} = \sum_{i \in c} u(i) \quad u(i): \text{power plants emission in each country}$$

- from CARMA database
- Uncertainties are estimated from the average of the reported standard error from plants listed in the US Energy Information Administration and/or the US EPA Clean Air Market datasets, floor value of 0.01 MtC/yr

$$O_{\text{other}} = \sum_{x \in c} A(x) p P(x) g(x) e f(c)$$

e: global energy intensity of the economy
A(x): area of grid cell
f(c): carbon intensity of energy production for each country

Quantitative Network Design

Uncertainty $\rightarrow \mathbf{C}(d)^2 = \mathbf{C}(d_{\text{obs}})^2 + \mathbf{C}(d_{\text{mod}})^2.$

What we do know already

$$\mathbf{C}(x)^{-1} = \mathbf{M}'^T \mathbf{C}(d)^{-1} \mathbf{M}' + \mathbf{C}(x_0)^{-1}. \quad (3)$$

$$\sigma(y)^2 = \mathbf{N}' \mathbf{C}(x) \mathbf{N}'^T + \sigma(y_{\text{mod}})^2. \quad (4)$$

What we are after

$$\sigma(y_0)^2 = \mathbf{N}' \mathbf{C}(x_0) \mathbf{N}'^T + \sigma(y_{\text{mod}})^2. \quad (5)$$

Notation:

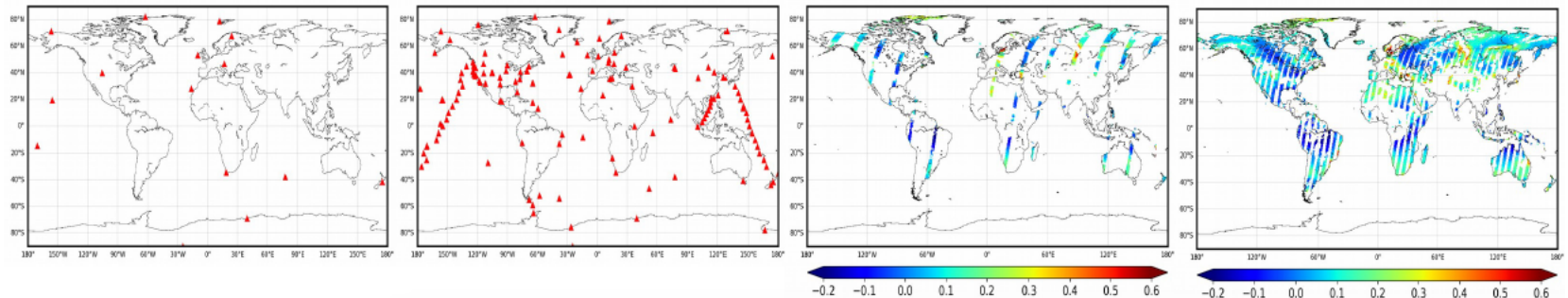
y:	vector of target quantities
d:	vector of observations
x:	vector of unknowns/control variables
d=M(x):	model linking unknowns to observations
y=N(x):	model linking unknowns to target quantities
C:	covariance of uncertainty

M' large (1.5m x 1.5m)
- exploit sparsity
- solve (3) iteratively for selected N'

Jacobians via
Automatic Differentiation
(Hascoet & Pascual, 2013)

Assessments for week June 1-7, 2008

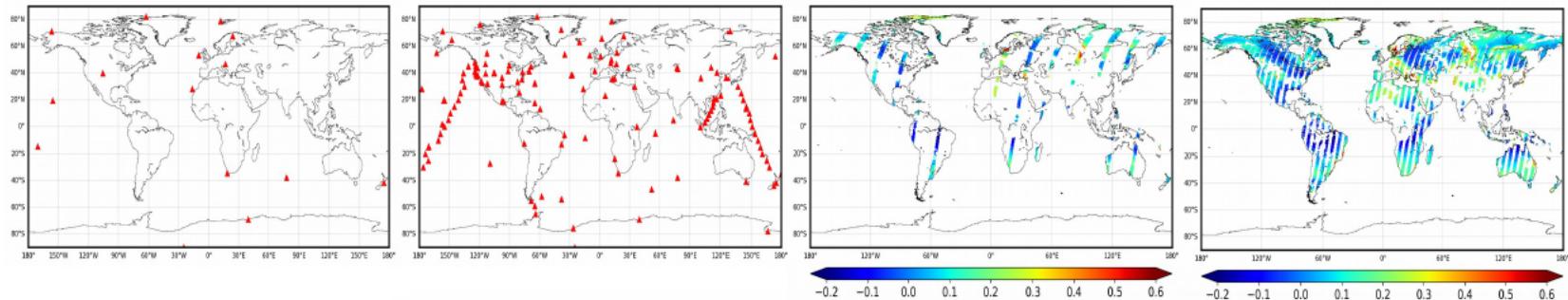
Scenario	Description	Other sector					Electricity generation sector				
		Emission rate uncertainty (MtC/week)									
		AUS	BRA	CHN	DEU	POL	AUS	BRA	CHN	DEU	POL
1	surface 15 sites	9.03	16.70	177.31	12.18	4.70	0.28	0.17	2.36	0.43	0.23
2	surface 141 sites	4.57	8.21	8.29	2.60	2.10	0.28	0.17	2.36	0.43	0.23
3	1 satellite (default)	0.30	0.42	3.43	0.97	0.38	0.27	0.17	2.21	0.43	0.23
4	4 satellites	0.25	0.29	2.38	0.79	0.33	0.26	0.17	2.07	0.43	0.23
5	default with ocean	0.29	0.41	2.93	0.94	0.37	0.27	0.17	2.20	0.43	0.23
6	default with repr. error	0.35	0.68	4.68	1.36	0.62	0.28	0.17	2.28	0.43	0.23
7	default with nat. inventory	0.03	0.16	1.84	0.08	0.05	0.04	0.06	1.43	0.07	0.05
-	national inventory	0.90	1.67	17.73	2.43	0.73	1.15	0.22	16.36	1.76	0.83



Assessments for week June 1-7, 2008

with representation error

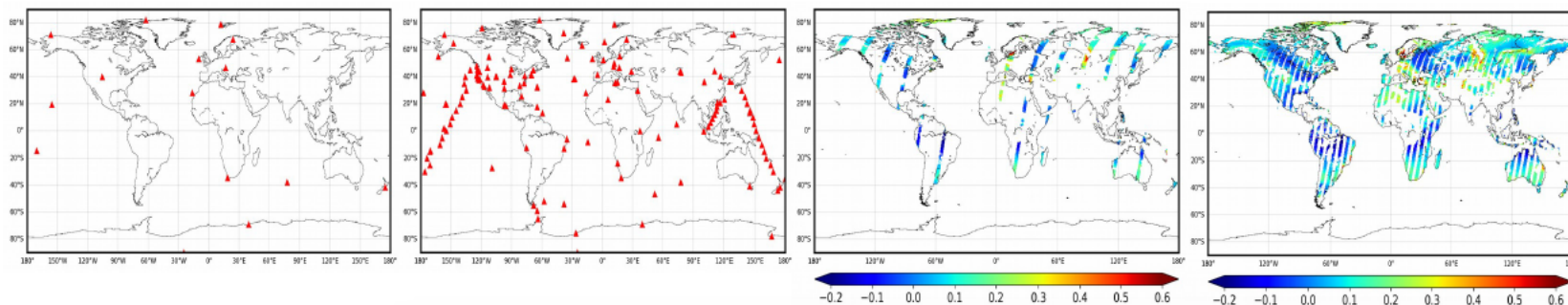
		Other sector					Electricity generation sector				
		Emission rate uncertainty (MtC/week)									
Scenario	Description	AUS	BRA	CHN	DEU	POL	AUS	BRA	CHN	DEU	POL
1	surface 15 sites	9.03	16.70	177.31	12.18	4.70	0.28	0.17	2.36	0.43	0.23
2	surface 141 sites	4.57	8.21	8.29	2.60	2.10	0.28	0.17	2.36	0.43	0.23
3	1 satellite (default)	0.30	0.42	3.43	0.97	0.38	0.27	0.17	2.21	0.43	0.23
4	4 satellites	0.25	0.29	2.38	0.79	0.33	0.26	0.17	2.07	0.43	0.23
5	default with ocean	0.29	0.41	2.93	0.94	0.37	0.27	0.17	2.20	0.43	0.23
6	default with repr. error	0.35	0.68	4.68	1.36	0.62	0.28	0.17	2.28	0.43	0.23
7	default with nat. inventory	0.03	0.16	1.84	0.08	0.05	0.04	0.06	1.43	0.07	0.05
		Annual average weekly emission rate (MtC/week)									
-	national inventory	0.90	1.67	17.73	2.43	0.73	1.15	0.22	16.36	1.76	0.83



Assessments for week June 1-7, 2008

“Synergy” vs. “Verification” mode

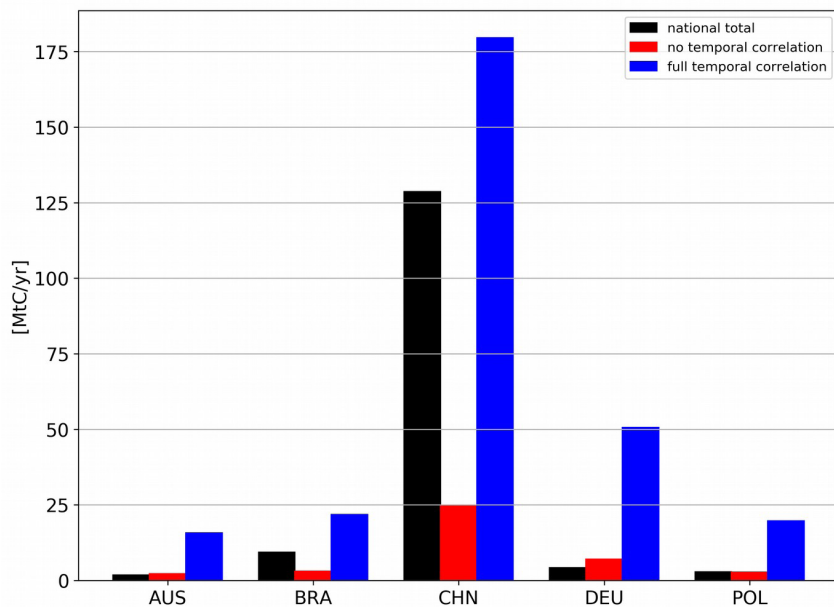
		Other sector					Electricity generation sector				
		Emission rate uncertainty (MtC/week)									
Scenario	Description	AUS	BRA	CHN	DEU	POL	AUS	BRA	CHN	DEU	POL
1	surface 15 sites	9.03	16.70	177.31	12.18	4.70	0.28	0.17	2.36	0.43	0.23
2	surface 141 sites	4.57	8.21	8.29	2.60	2.10	0.28	0.17	2.36	0.43	0.23
3	1 satellite (default)	0.30	0.42	3.43	0.97	0.38	0.27	0.17	2.21	0.43	0.23
4	4 satellites	0.25	0.29	2.38	0.79	0.33	0.26	0.17	2.07	0.43	0.23
5	default with ocean	0.29	0.41	2.93	0.94	0.37	0.27	0.17	2.20	0.43	0.23
6	default with repr. error	0.35	0.68	4.68	1.36	0.62	0.28	0.17	2.28	0.43	0.23
7	default with nat. inventory	0.03	0.16	1.84	0.08	0.05	0.04	0.06	1.43	0.07	0.05
		Annual average weekly emission rate (MtC/week)									
-	national inventory	0.90	1.67	17.73	2.43	0.73	1.15	0.22	16.36	1.76	0.83



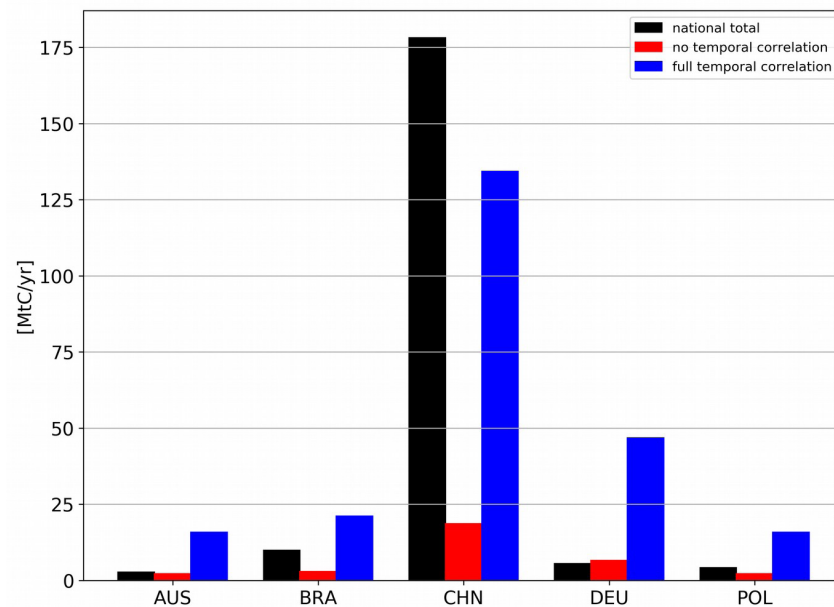
Extrapolation of Posterior Uncertainty to Annual Scale: XCO2 (1 CO2M satellite) + nightlights

Weekly (1st week of June) posterior emission uncertainties scaled to annual values assuming:

- no temporal correlation of weekly uncertainties (red bars)
- full temporal correlation of weekly uncertainties (blue bars)



Other sector



Aggregated over all sectors

Summary and Outlook

- Light Framework developed
- Verification and synergy modes
- Useful to explore (some) design options of
 - Satellite mission
 - Surface network (e.g. ICOS network ...)
 - Inversion component of MVS (control vector, assimilation window, posterior uncertainty ...)
- National and annual scale unc. in range of inventory unc.
- Can refine models and go to higher resolution
- Use further data streams, e.g. radiocarbon as a proxy for fossil fuel, ...
- Use more sectors in fossil fuel model component, e.g. transport, ...