

Application of the transport models for inverse modeling of the greenhouse gas fluxes.

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and

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Outline

Improvements for the inverse model for GOSAT L4A

• Terrestrial biosphere, ocean carbon cycle, atmospheric transport and fossil fuel emissions

Ongoing developments

- Kalman smoother application with coupled transport.
- Cumulus cloud mixing
- Biospheric model optimization with inversion

Summary





Options	Model					
	NIES-05 ¥SML¥0.5	NIES-08 ¥VL¥1.25	NIES-08 ¥VL¥0.625	NIES-08 ¥Pr¥2.5		
Numerical Scheme	Semi- Lagrangian	Flux-form (flux version)				
		3-order van Leer		2-order Moment		
Resolution, deg	0.5	1.25	0.625	2.5		
Number of vertical levels	47					
Meteo dataset	JMA/GPV dataset					

GPV meteorological dataset with resolution of 0.5×0.5 degrees for 21 pressure levels - special product for GOSAT by JMA (added extra levels in LT, 3 hourly vs standard 6 hourly)

v. 08 VL: Van Leer, 2nd order shape function
v. 05 Pr: Prather, with 2nd order moments
v. 05 SML: Semi-Lagrangian, bilinear interpolation

Tracer transport model performance



Resolution		NIES-08 Val Leer	NIES-08 Prather	NIES-05 SML
2.5°×2.5° (2°×2° for SML)	CPU, sec	10.21	292.90	6.08
	emin	6.22E-04	1.38E-04	-5.37E-03
	emax	-2.86E-03	-3.48E-04	2.16E-03
	err1	-6.66E-03	-5.96E-08	6.09E-02
	err2	1.17E-03	1.02E-03	5.08E-03
	Memory, Gb	0.72	0.77	0.72
1.25°×1.25° (1°×1° for SML)	CPU, sec	55.93	1755.77	20.86
	emin	1.57E-04	-5.95E-05	-0.229E-06
	emax	-3.66E-03	1.03E-06	0.00E+00
	err1	-3.50E-03	4.78E-03	1.93E-02
	err2	8.94E-04	8.48E-03	9.90E-03
	Memory, Gb	0.98	1.10	0.98
0.625°×0.625° (0.5°×0.5° for SML)	CPU, sec	370.975	12683.15	82.20
	emin	3.93E-05	-5.11E-06	-1.04E-07
	emax	-2.44E-03	-5.21E-03	7.95E-08
	err1	-1.75E-03	5.55E-03	1.59E-02
	err2	6.15E-04	1.18E-04	1.74E-02
	Memory, Gb	1.94	2.46	1.94

Tracer transport model: model validation





Tracer transport model: model validation





Seasonal cycle of CO_2 over South Pole (-89.98;-24.80; 2810) (a) and Cold Bay in Alaska (55.20; -162.72; 25) (b) for 2008, ppmv

Land CO₂ fluxes: improved simulation seasonal cycle in atmospheric CO₂ partial column abundance





Land CO₂ fluxes: improved simulation seasonal cycle in atmospheric CO₂ partial column abundance



Seasonal variation of CO_2 vertical profile column, with VISIT ecosystem model before optimization. Red – VISIT Green – Globalview08



Plan – optimize q10 and photosynthetic capacity (and ~ 15 other parameters) with inverse model

Ocean CO2 flux. 4D-var assimilation of the surface ocean pCO2



Objective: Provide ocean CO₂ flux priors for GOSAT L4A inverse model.

Physical : Oceanic TM (Valsala et al., 2008) Chemical: OCMIP-II (Watson and Orr, 2003) Biological: McKinley et al. 2004 4D-var: Ikeda and Sasai (2002)

Currents: ECCO (or GFDL)

DIC/pCO2 observations: LDEO database, Takahashi 2009



Ocean fluxes: assimilation output

ECCO annual mean co2 flux (spin up; mole/m2/yr)





High resolution fossil fuel emissions.





Using large point source data and DMSP lights (1km res) produces good match at 50 km scale with high resolution bottom-up inventory by Vulcan project (K.Gurney) in US.

by Oda & Maksyutov, ICDC8, 2009

Development of a coupled transport model



1. Objective;

Develop a new transport model LPDM (Flexpart) coupled with NIES-TM for use in inverse modeling

2. Progress; Model has been validated using ESRL flask and NIES continuous greenhouse gases observations 1996-2007.

3. Plan

Extend model to using GOSAT column observations

Y. Koyama et al, ICDC8



Fig. 1. CO2 variation at Hateruma for year 2005.



Fig. 2. CO2 and CH4 variations at Hateruma for winter season, year 2001.



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Inverse modeling with Lagrangian transport



1. Objective:

Use each available observation without smoothing filtering (aggregation), reduce response function simulation time (5 to 10 times).

2. Present state:

Implemented fixed lag Kalman smoother to use 3-hourly continuous observations and flasks directly in 64 region monthly inversion with 3-6 month time lag (Bruhwiler etal 2005), tested with 1996 data

3. Plan

Complete 1980-1990 analysis in 2009, rest in 2010

Y. Koyama et al, ICDC8







Improving the algorithms for regional CO2 flux estimation in global scale including:

- Inverse modeling with large number of observations
- Fossil fuel emissions model, high spatial resolution
- Process-based modeling of the terrestrial ecosystem fluxes
- Observation-driven data assimilation system for near real-time surface pCO2 and ocean-atmosphere flux estimation
- High resolution transport modeling