Orbiting carbon observatory (OCO-2) tracks 2-3 Giga tons of carbon release to the atmosphere during the El Nino (Oct 2014 – Feb 2016)

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Introduction

- 2014-2016 has been special time in terms of climate
 - Hot summer in most part of Asia (2016 January set the record high SAT)
 - Large El Nino induced fire occurrence over Southeast Asia in Sep-Oct'15
- CO₂ concentration is known to be affected by anthropogenic and land biospheric activities at interannual timescales
- The present measurements and modelling capabilities do not allow us (accurately enough) to track CO₂ concentration and fluxes regionally
- Here we make an attempt to analyse XCO₂ observations from OCO-2 and estimate global total CO₂ flux anomaly since Oct 2014
- Climate-carbon nexus is important for the projection of future climate scenario due to atmospheric CO₂ change



OCO-2 data processing and screening (e.g., AMF)

ACTM is sampled within 0.5 hr of OCO-2 overpass, and convolved with the OCO-2 a priori and averaging kernel

In the control case, we screen OCO-2 data as:

- 1. Warn level (WL) < 10
- 2. Air mass factor (AMF) < 3.5

All data are gridded in to 2.5°x2.5° grid at monthly intervals

Grids with less than 3 data points are assigned missing value

In sensitivity cases, we have checked the results for WL<5 and AMF < 2.5



ACTM model transport (Patra et al., 2008-2016)

- Based on CCSR/NIES/FRCGC Atmospheric General Circulation model developed in JAMSTEC
- Model Transport is nudged to Japan Meteorological Agency Reanalysis (JRA-55): horizontal winds and temperature
- Fossil fuel and cement production (FFC) : Based in EDGAR4.2FT2012 spatial distribution and CDIAC top-20 country and global totals, the global total emission increased by 0.2 Pg/yr for 2015 and 2016
- Oceanic Exchange (OCN): Based on air-sea pCO₂ measurements and extrapolation to global ocean [Takahashi et al., 2009]
- *Terrestrial Biosphere Flux (CASA)*: Based on Carnegie-Ames-Stanford Approach (CASA) model by Randerson et al. [1997], with a 3-hourly diurnal cycle introduced using JRA (Y. Niwa, pers. comm.)
- Flux Inversion (CYC64 for 2008, IAV84 for 2011)
 - (CYC64; Patra et al., 2011): From the 64-region inverse model using the smoothed surface CO₂ data [GLOBALVIEW-CO₂, 2013] and additional CARIBIC measurements in 2008 [Schuck et al., 2010]. Inversion fluxes validated using the CONTRAIL observations [Machida et al., 2008].
 - (*IAV84; Saeki et al., in prep.*): From a newly developed 84-region inverse model for the period 1992-2011 [GLOBALVIEW-CO₂, 2013]. The CO₂ fluxes have been used in a recent Asian flux assessment study, and validated using CONTRAIL CO₂ observations [Thompson et al., 2016].

• Emissions from Fires (GFAS): The fire-related daily CO₂ emissions from the Global Fire Assimilation System (GFAS) are taken from Kaiser et al. [2012; http://macc.icg.kfa-juelich.de:50080/access? catalogue=MACC_daily_wildfire_emissions], from October 2014 onwards.

3-simulations:

Case 1. FFC + CYC64 (Lnd + Ocn) (highest CO_2 growth rate)

Case 2. FFC + IAV84 (Lnd + Ocn) (slowest CO₂ growth rate)

Case3. FFC + IAV84 + GFAS (fire) (moderate CO_2 growth rate)

FFC: Fossil fuel and cement CYC64 and IAV84: two cases of inversion fluxes

Latitude-time distributions of zonal-mean XCO₂



705 605 505 405 305 205 105 EQ 10N 20N 30N 40N 50N 60N 70N 7

705 605 505 405 305 205 105 EQ 10N 20N 30N 40N 50N 60N 70N

Reference year using ground based observation



ACTM IAV84 run simulated the CO₂ growth for 1/2013 -9/2014 and also for 2012 (not shown), we assume 2011-2013 as the reference year for the CO₂ flux anomaly calculation

NOAA data are taken from : www.esrl.noaa.gov/gmd/ccgg/



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Calculation of CO₂ flux correction

 CO_2 flux corrections from sub-hemispheric atmospheric CO_2 burden change at monthly time interval.

Burden difference (= $\Sigma_i \Sigma_j$ (XCO₂ difference \times area of the grid \times air density) where, i = lon grid (144), j = lat grid (72)

 CO_2 flux correction = $d(Burden \ difference)/dt$

The difference in the burdens between October and September 2014 is assigned to the flux correction for October 2014.





Estimation of CO₂ flux anomaly from empirical relationship with ENSO Index trends



Patra et al., 2005c

Global total CO₂ fluxes: a priori and a poste corrections

Time window	A priori CO ₂ fluxes used for					Patra	CO ₂ flux corrections from		
	ACTM simulations					et al. #	OCO-2 – ACTM differences ^{\$}		
	FFC	CYC64	IAV84	IAV84	GFAS	(2005b)	CYC64	IAV84	IAV84
				+GFAS					+GFAS
Oct-Dec 2014	2.44	3.36	0.80	1.17	0.37		-0.140.42	0.35 - 0.42	0.25 - 0.30
Jan-Dec 2015	9.98	-2.86	-6.29	-4.29	2.00	2.67	-0.280.48	0.95 - 1.59	0.21 - 0.32
Jan-Feb 2016	1.70	0.64	0.10	0.36	0.26	- 2.73	-0.170.31	0.15 - 0.23	0.05 - 0.06
Oct14 -Feb16	14.12	1.14	-5.36	-2.72	2.64		-0.591.21	1.45 - 2.24	0.51 - 0.68

2015 net fluxes for the 3 simulations are: -3.34 (CYC64), -3.70 (IAV84), and -3.97 (IAV84+GFAS)

for ACTM_IAV84 sensitivity cases: -3.91 (AMF<2.5), -3.90 (WL<5)

(these global total sinks are residuals – depends entirely on assumed FFC)

[#] Range estimated from two different fits, with (= $0.3539 + 1.4935 \times MEI$ amplitude change) or without (=- $1.0756 + 2.4579 \times MEI$ amplitude change) the La Niña years

^{\$} Range estimation using two different approximations on area coverage, lower range is for data just over the OCO-2 measurement area, higher values with data coverage extrapolated to the poles

Flux correction time series

- The flux corrections suggest pulsed emissions of ~1month, in agreement with fire counts variability
- Pulsed emission, likely from fires, accounts for 0.7 PgC in 2015, which is 30-44% total CO₂ flux anomaly
- GFAS emissions vary surprisingly slowly throughout the year
- CO₂ emissions from Oct 2015 fire peak goes undetected due to aerosol screening



In support of strong biomass burning emissions in 2015



Sources of uncertainty

- Emissions from fossil fuel and cement (FFC)
 - FFC CO₂ emission increase is assumed to be 0.2 PgC/yr; for no increase in FFC emission, the CO₂ flux anomaly would increase by ~0.3 PgC for the OCO-2 period
 - Constant bias in total FFC emission strength wouldn't change the CO₂ flux anomaly, but affect the regional sources/sinks budgets

Data coverage

- Data screening using AMF and WL affect the estimation of CO₂ flux anomaly marginally
- Extension of observation-model differences to the data void regions affect the flux anomaly calculation and likely misallocate the sources/sinks regionally

More discussion on this in the TransCom-style side meeting in the evenning



Summary and outlook

- We analyzed the XCO₂ from NASA's OCO-2 and JAMSTEC's ACTM during September 2014 and February 2016
- The 2014-2016 El Niño event led to an excess CO₂ release to the atmosphere in the range of 2.24-3.32 PgC yr⁻¹
- A few of the major issues to be dealt with for using XCO₂ data in inversion:
 - Handling of the data gaps in OCO-2 or other passive sensors for long-lived gases is a cause for concern (not serious for the short-lived species as their emission and chemical loss cycle is confined to a particular latitude band)
 - Accounting for chemical production of CO₂ from reduced carbon compounds, CH₄, CO, BVOCs etc. (not serious for in situ surface data, which are influenced most by surface fluxes)
 - Large-scale transport bias in the models should be tracked as the column integrated values are less sensitive to small-scale mixing
 - Low uncertainty in fossil fuel amount are needed for reduction of bias in regional source/sink estimation

Thank you and questions?



Need a smaller mesh to catch CO₂ emissions

Courtesy of Indonesia's Fire Crisis 2015 - The Biggest Environmental Crime of the 21st Century by Erik Meijaard http://jakartaglobe.beritasatu.com

- This work is partly supported by Ministry of Environ. Res. and Tech. Development Fund (grant # 2-1401; Pl. N. Saigusa)
 - CO₂ measurements are available for scientific use at
 - http://ds.data.jma.go.jp/gmd/wdcgg and
 - http://www.esrl.noaa.gov/gmd/dv/ftpdata.html
 - http://tccon.ornl.gov
- All the model results and analysis tools are available unconditionally from the ACTM group; PKP thanks Kentaro Ishijima for pre-processing of JRA-55 meteorological fields

Suggested model improvements

- ACTM is well tested for large-scale transport in troposphere, but issues remain with fast Brewer-Dobson circulation in the stratosphere
- Accounting for CO₂ production from reduced carbon compounds (CO, CH₄, VOCs etc.)
 - As shown here CO₂ is produced in the tropical troposphere and over the source region
 - Chemically CO₂ source are then transported to the higher latitude with a time delay
 - This situation leads to greater CO₂ emission in tropics and greater sink in the extratropical lands by inverse modeling



Why (NASA) satellites are important ?

Lack of globally uniform observations leave the biologically active land areas unconstrained by inverse modelling

Without measurements at the source region, fluxes are misallocated to other regions



We no more have access to "full record" of JAL WPO data

Estimation of CO₂ flux anomaly from empirical relationship with ENSO Index trends



Calendar Year

OCO-2 sensitivity to the AMF



Why it is difficult to track uncertainty in FFC?

Comparison of FFC vs Biosphere



Why no emission peak in October 2015?





Handling of data gaps seasonal sun & clouds

Patra et al., ACP, 2011; BG, 2013

