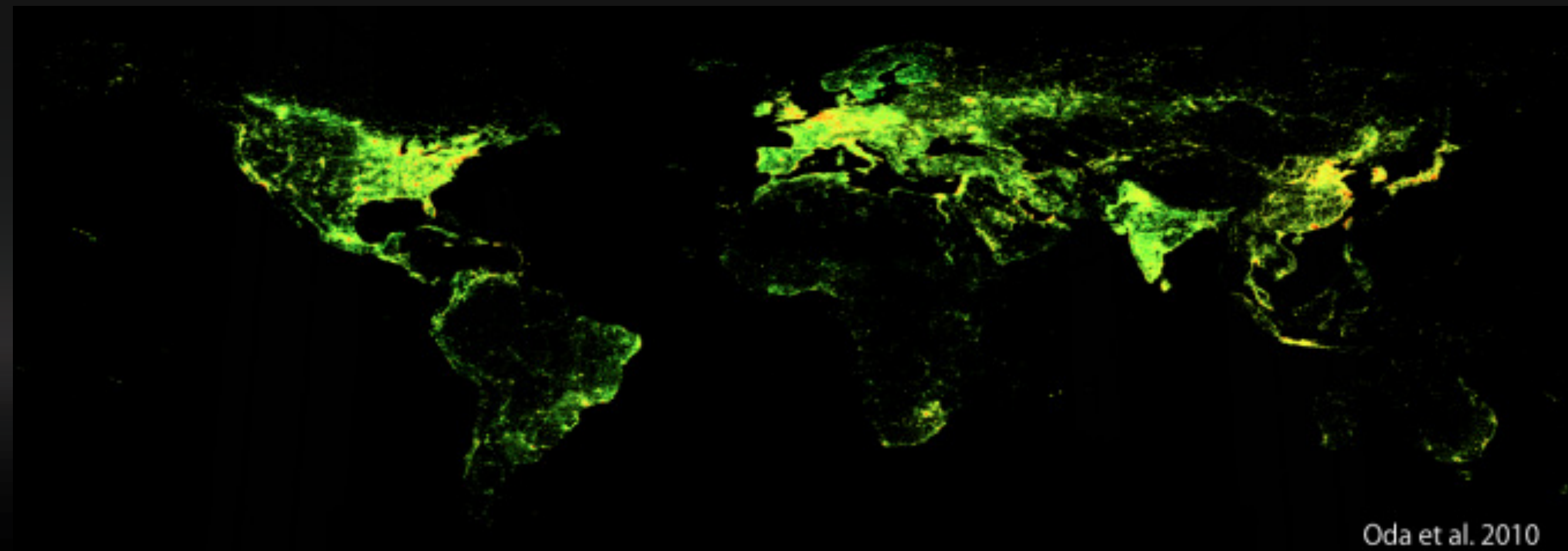


# ASSESSING UNCERTAINTIES IN GRIDDED EMISSIONS: A CASE STUDY FOR FOSSIL FUEL CARBON DIOXIDE (FFCO<sub>2</sub>) EMISSION DATA

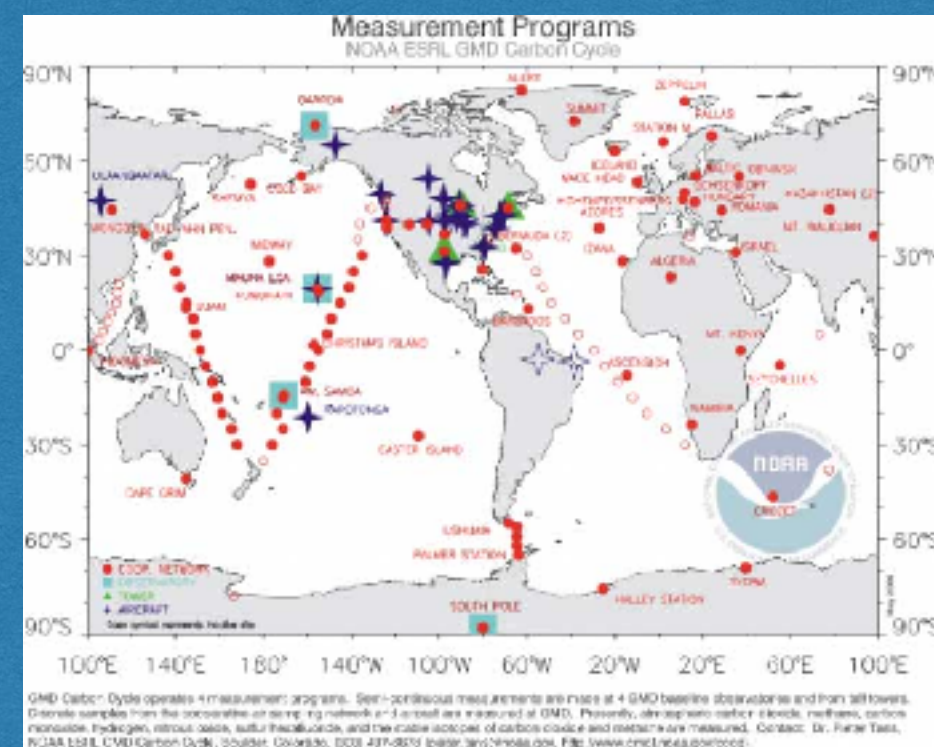
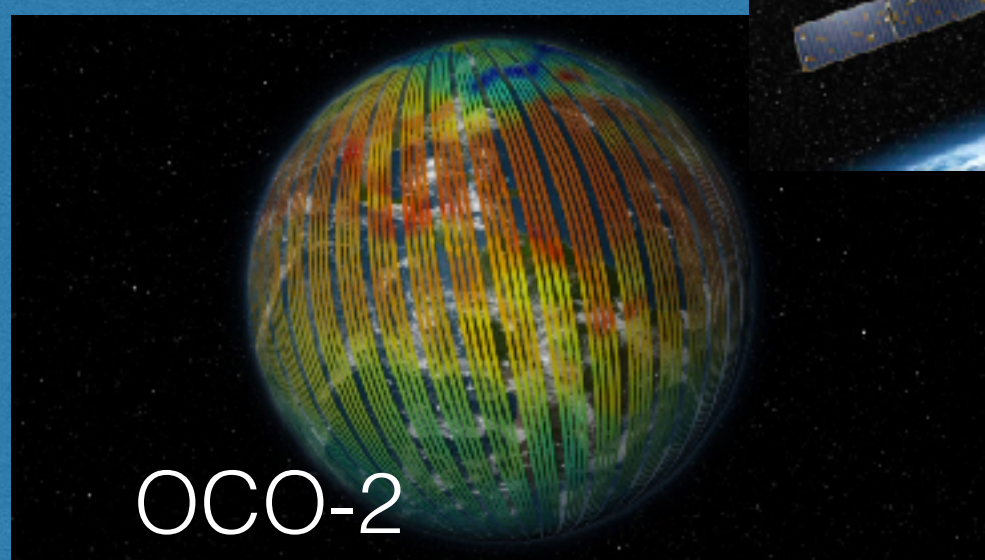


T. Oda<sup>1,2</sup>, L. Ott<sup>2</sup>, T. Lauvaux<sup>3</sup>, S. Feng<sup>3</sup>, R. Bun<sup>4</sup>, M. Roman<sup>2</sup>, D. F. Baker<sup>5</sup> and S. Pawson<sup>2</sup>

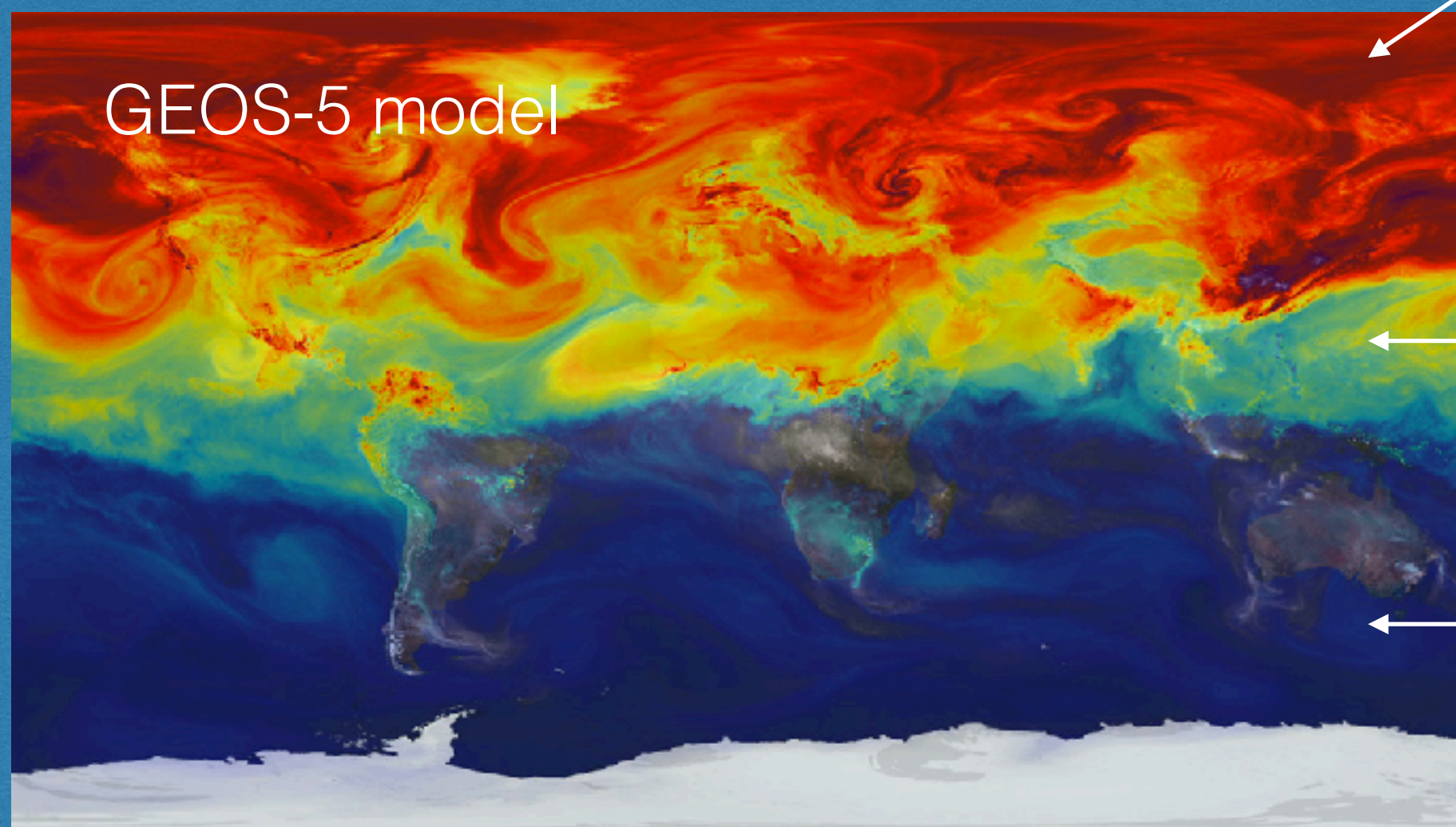
1:Universities Space Research Association, 2:NASA Goddard Space Flight Center 3:Penn State, 4:Lviv Polytechnic National University,  
5:Colorado State University,

# FFCO2 AS A PERFECT QUANTITY

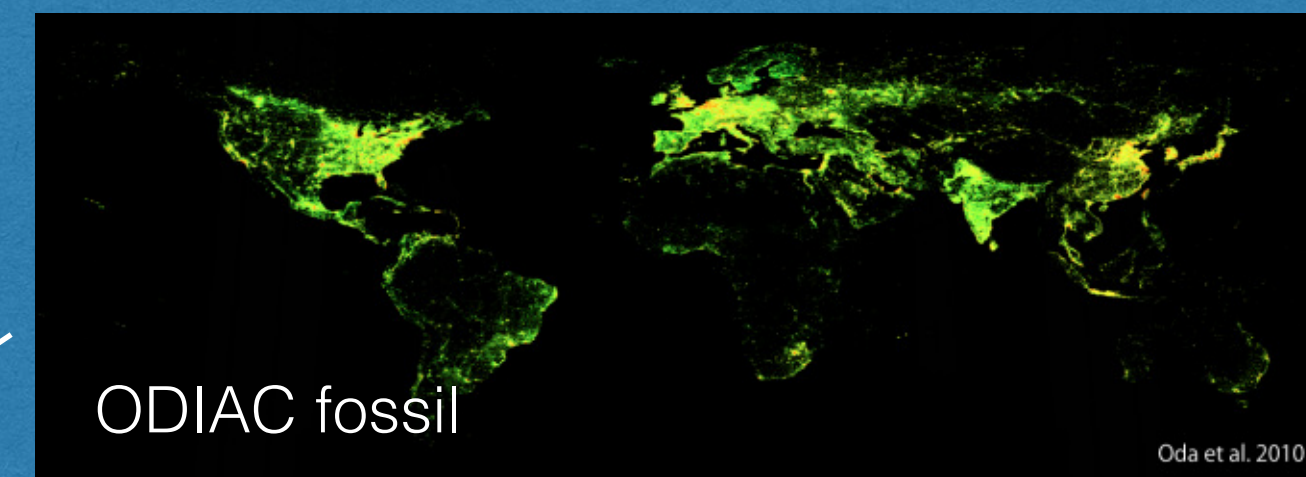
Atmospheric C growth = FFCO2 +/- natural process



Ground-based network

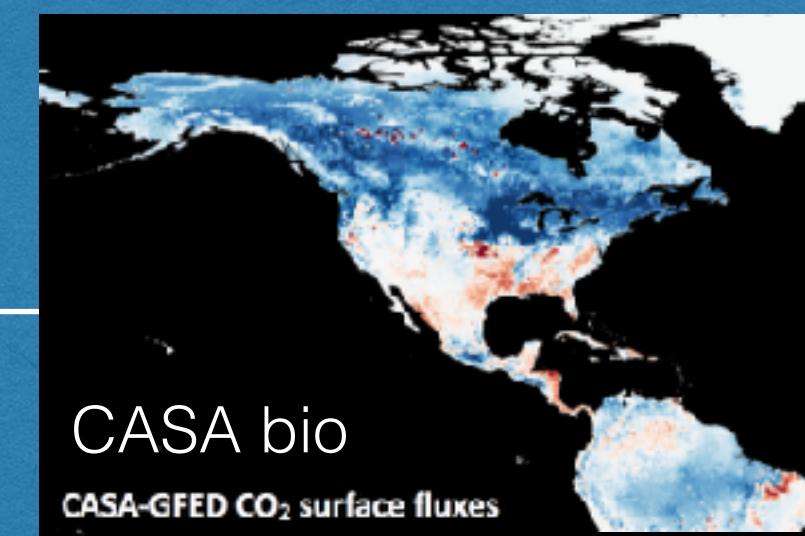


High-res atmospheric model



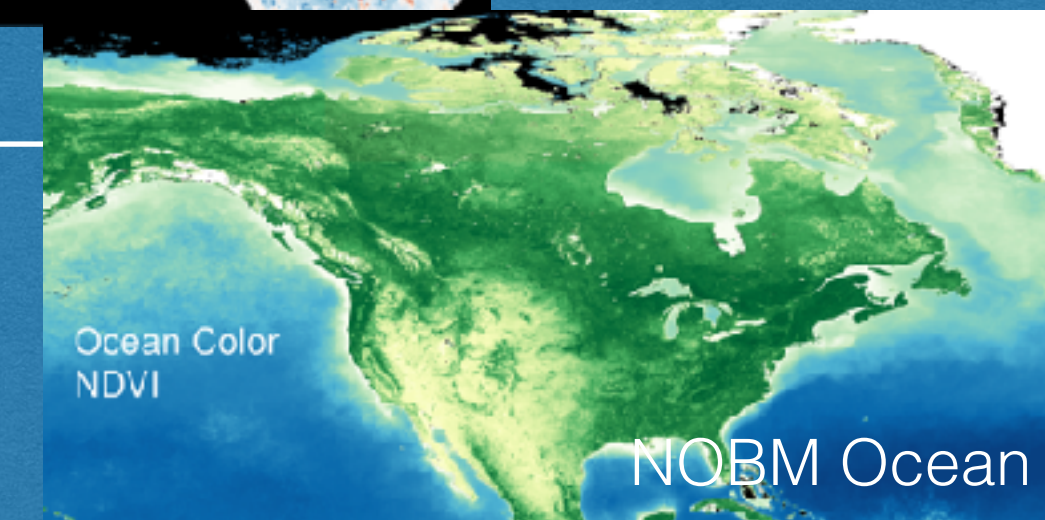
ODIAC fossil

Oda et al. 2010



CASA bio

CASA-GFED CO2 surface fluxes



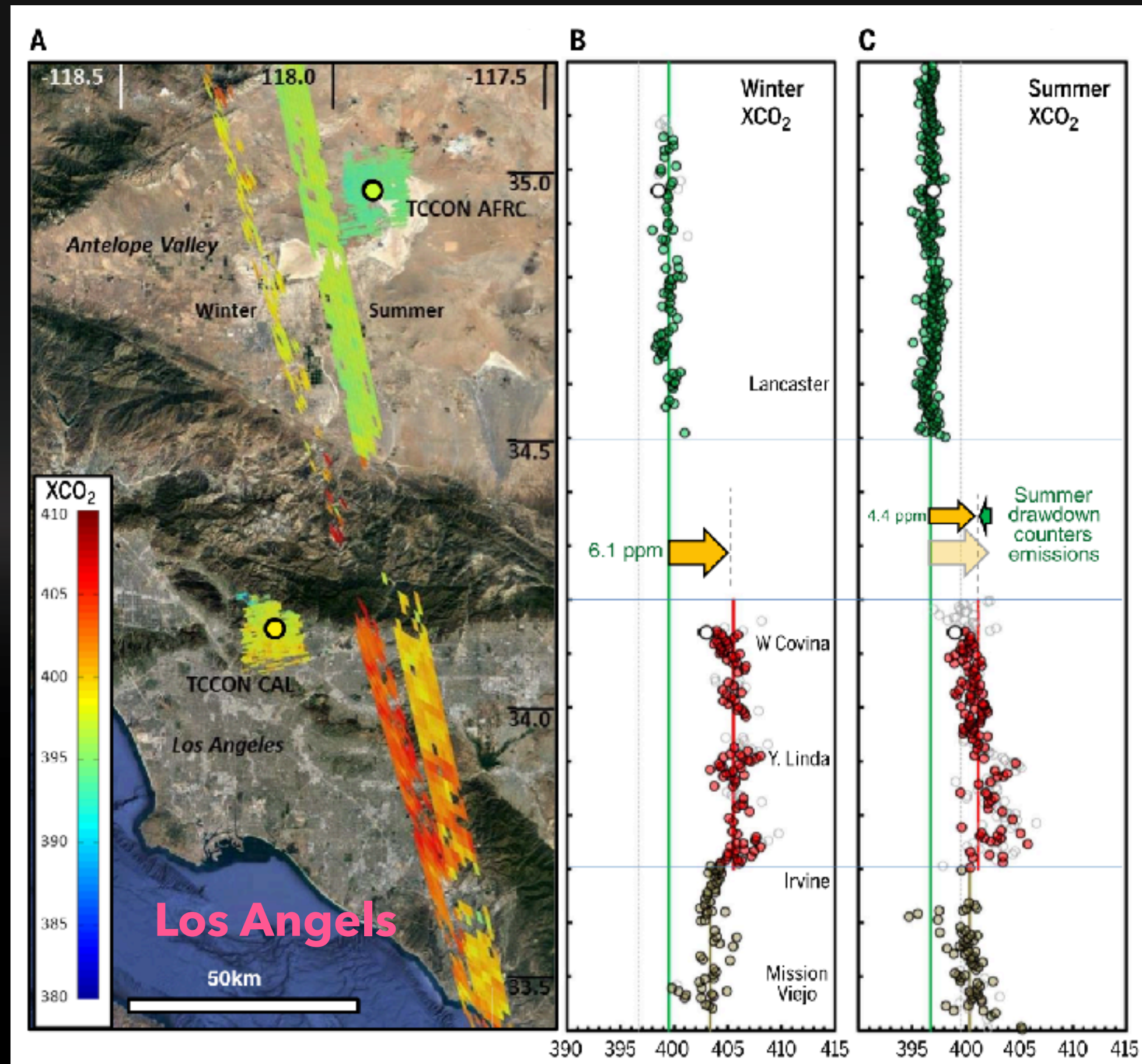
Ocean Color NDVI

NOBM Ocean

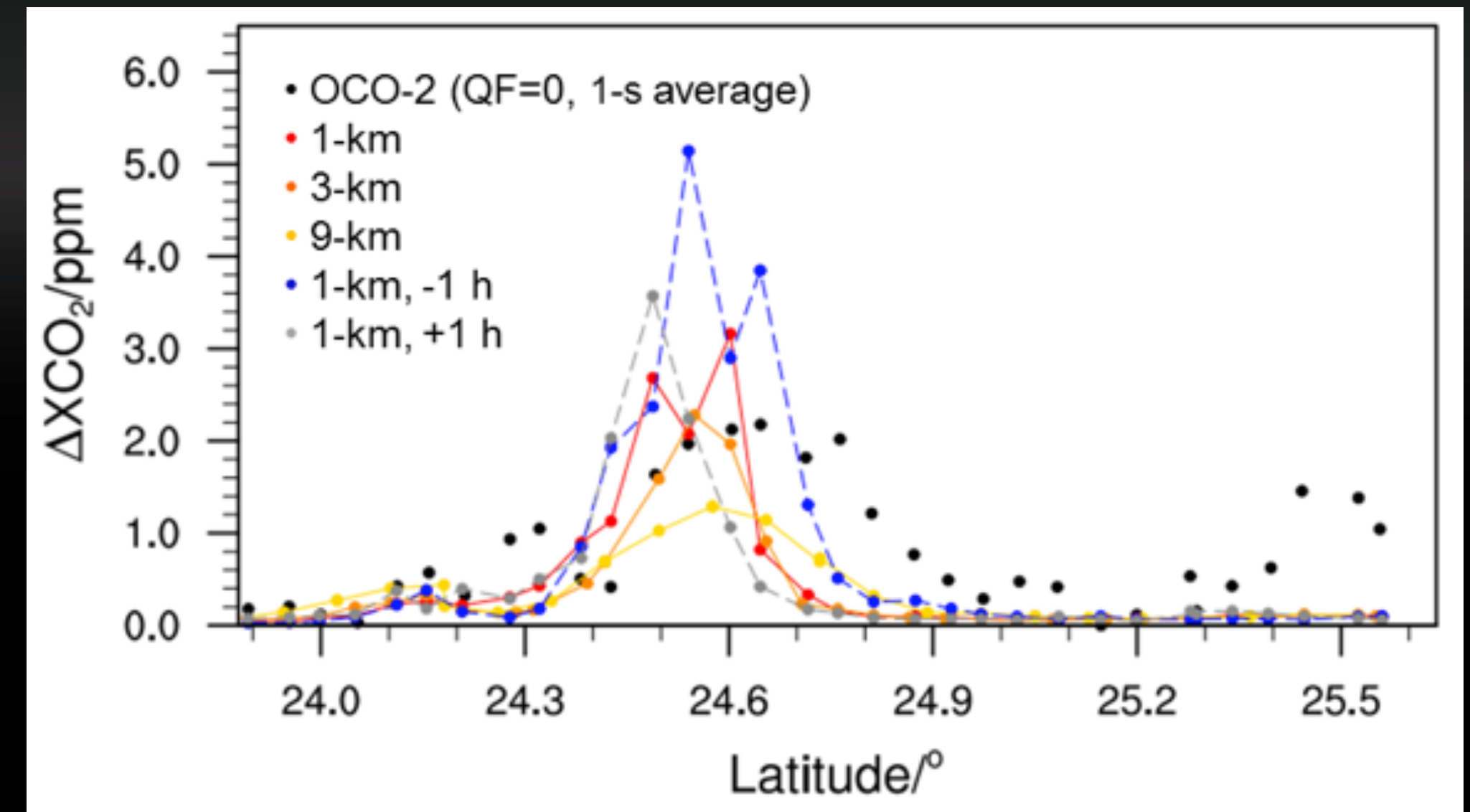
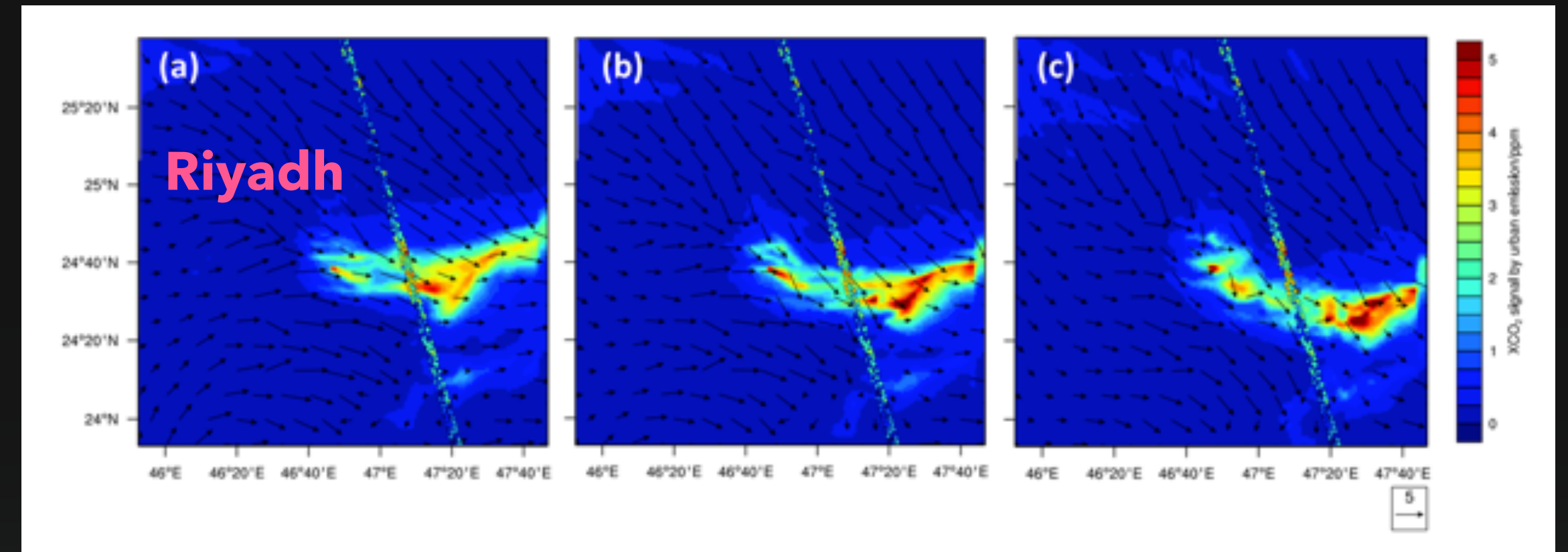
Fluxes contained by satellite obs

OTT ET AL. CMS GEOS-CARM MODELING SYSTEM

# FFCO<sub>2</sub> SEEN FROM SPACE

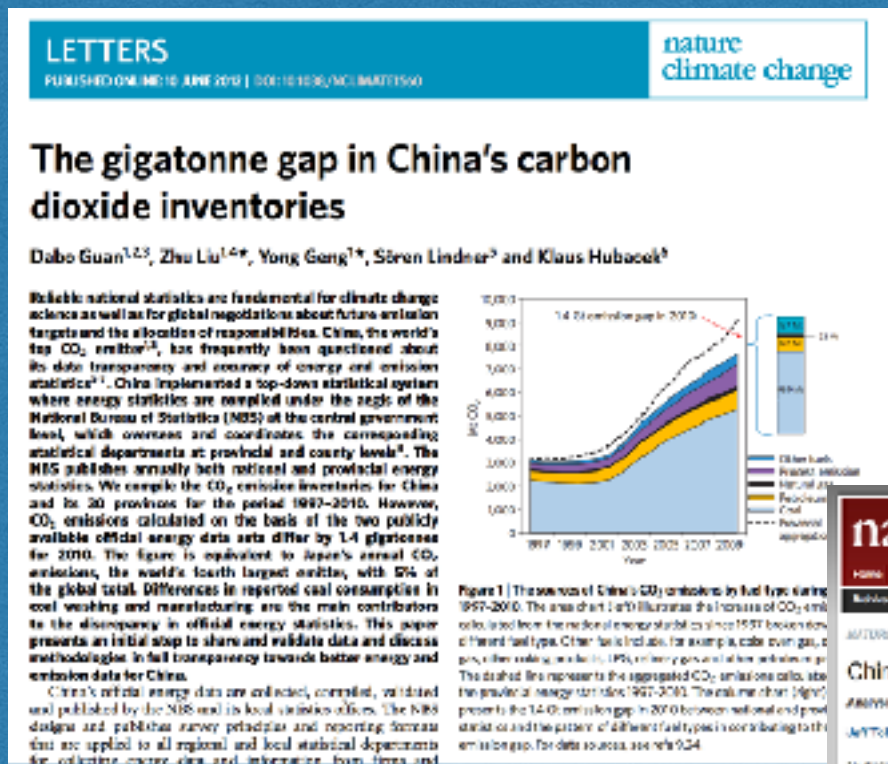


SCHWANDNER ET AL. 2017 SCIENCE



YE ET AL SUBMITTED TO ACP

# SOURCES OF UNCERTAINTIES IN GRIDDED DATA

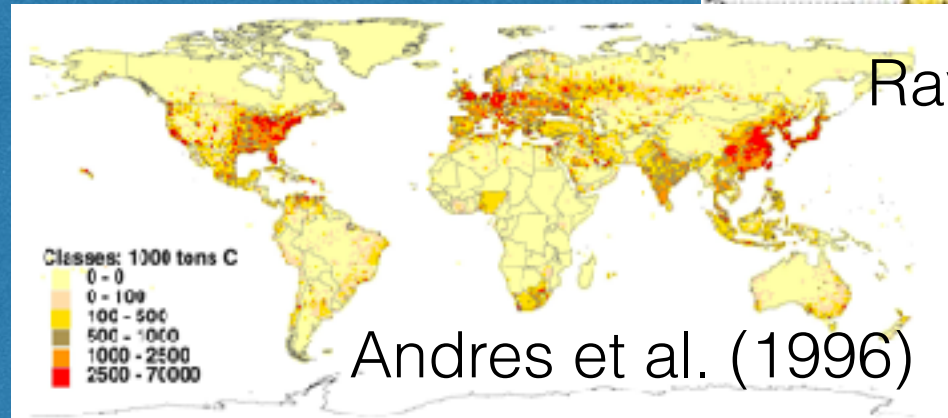
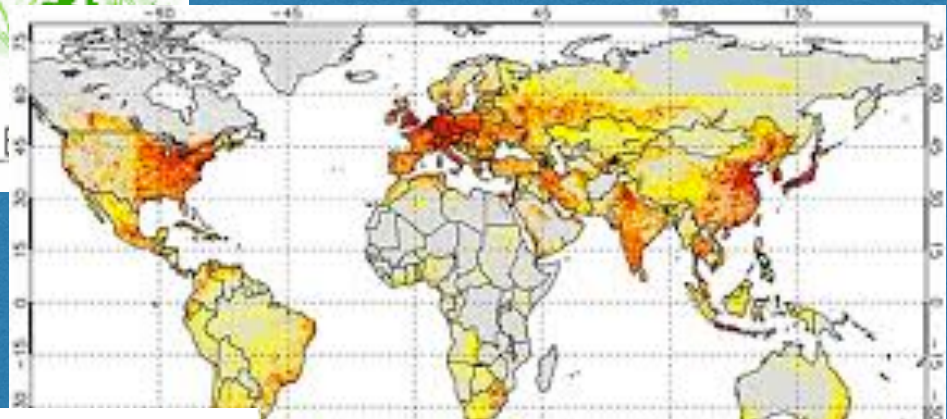
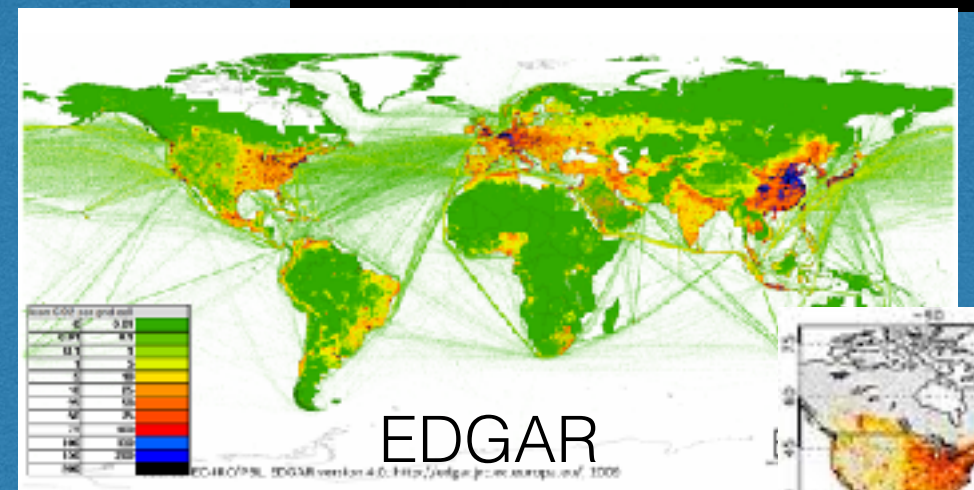
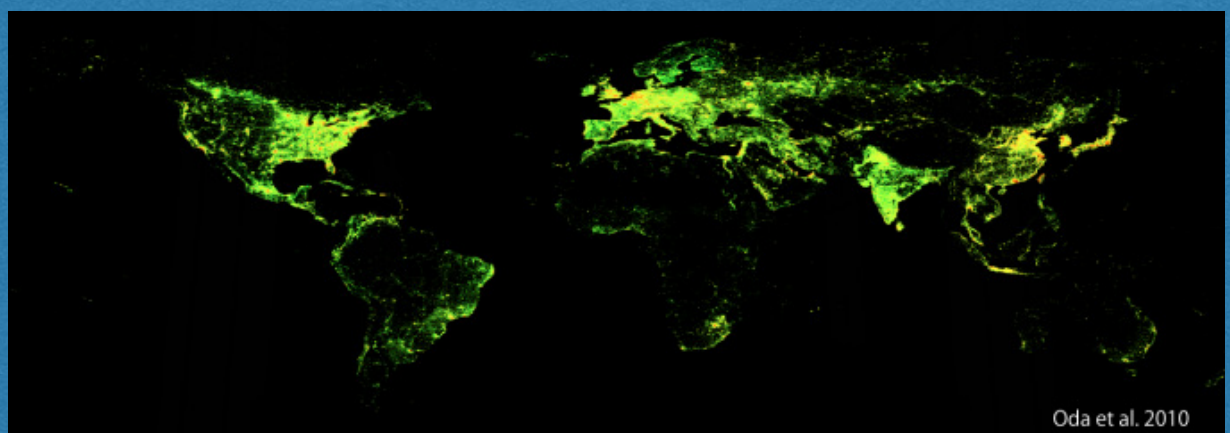


Activity data  
(Guan et al. 2012)

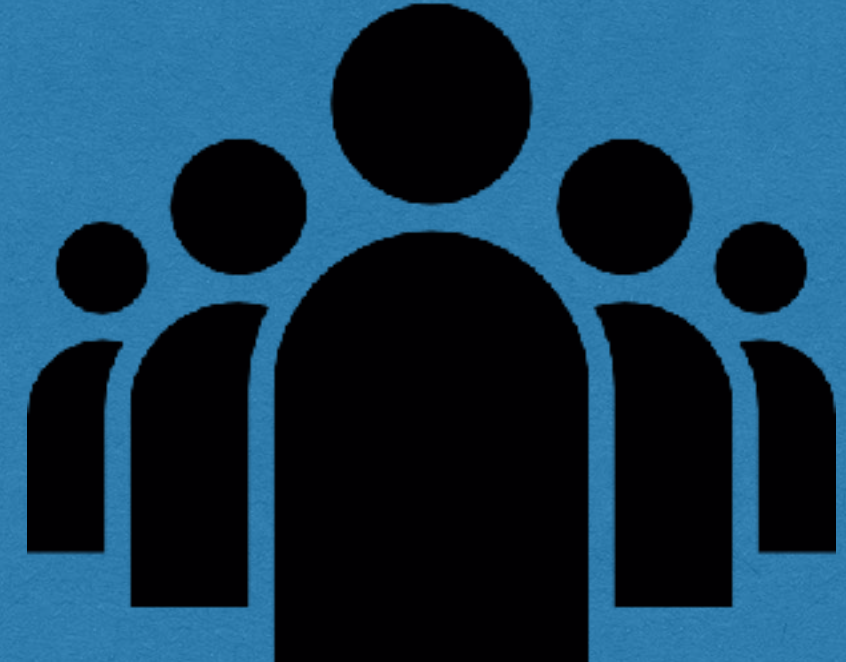


Emission factor  
(Liu et al. 2015)

Uncertainty/bias in emission calculation



Uncertainty/bias due to distribution methods



Users

**Emission calculation and disaggregation are often completely independent steps.**

# FFCO2 IN FLUX INVERSIONS

Biogeosciences, 10, 6699–6720, 2013  
 www.biogeosciences.net/10/6699/2013/  
 doi:10.5194/bg-10-6699-2013  
 © Author(s) 2013. CC Attribution 3.0 License.



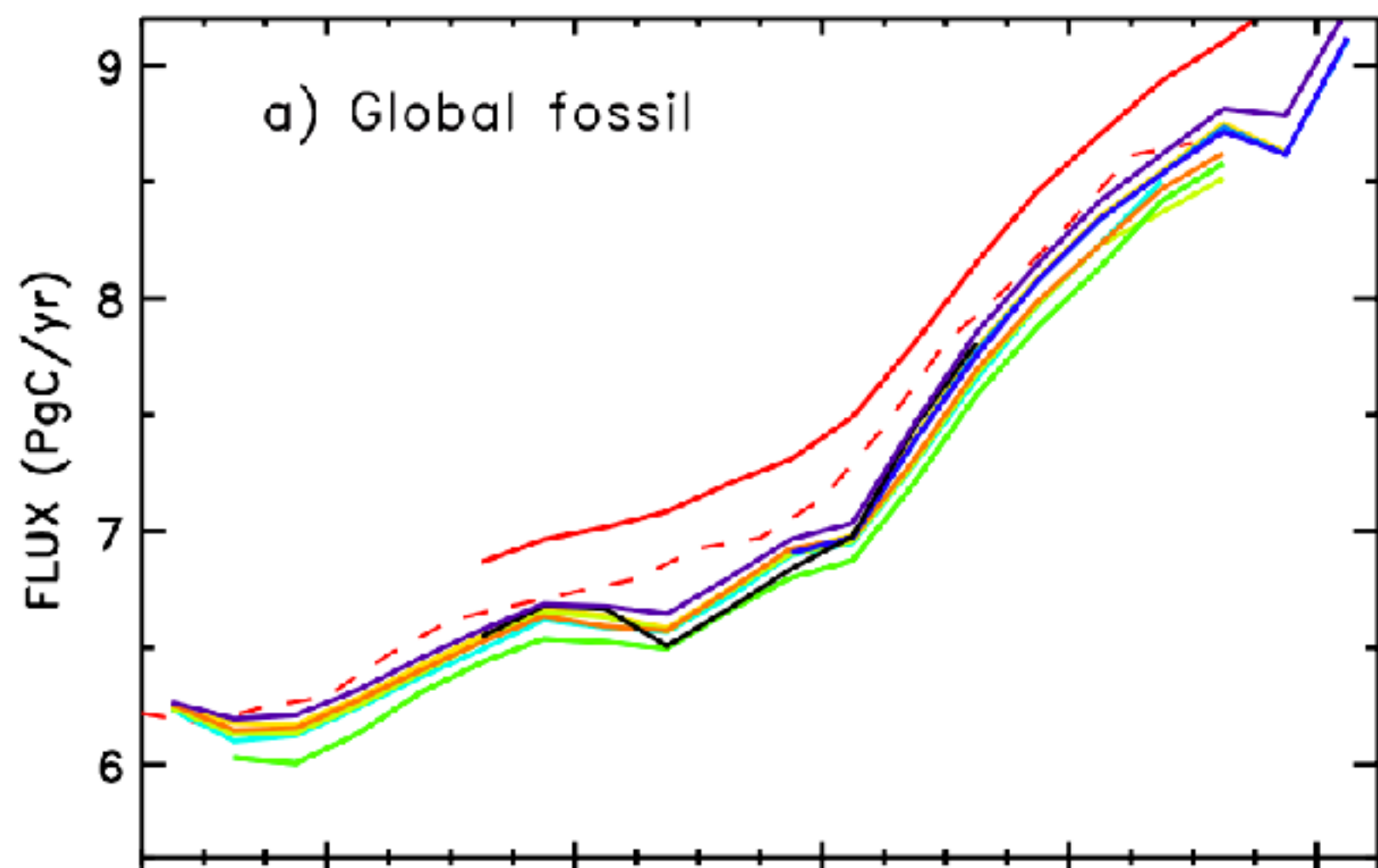
## Global atmospheric carbon budget: results from an ensemble of atmospheric CO<sub>2</sub> inversions

P. Peylin<sup>1</sup>, R. M. Law<sup>2</sup>, K. R. Gurney<sup>3</sup>, F. Chevallier<sup>1</sup>, A. R. Jacobson<sup>4</sup>, T. Maki<sup>5</sup>, Y. Niwa<sup>5</sup>, P. K. Patra<sup>6</sup>, W. Peters<sup>7</sup>, P. J. Rayner<sup>1,8</sup>, C. Rödenbeck<sup>9</sup>, I. T. van der Laan-Luijckx<sup>7</sup>, and X. Zhang<sup>3</sup>

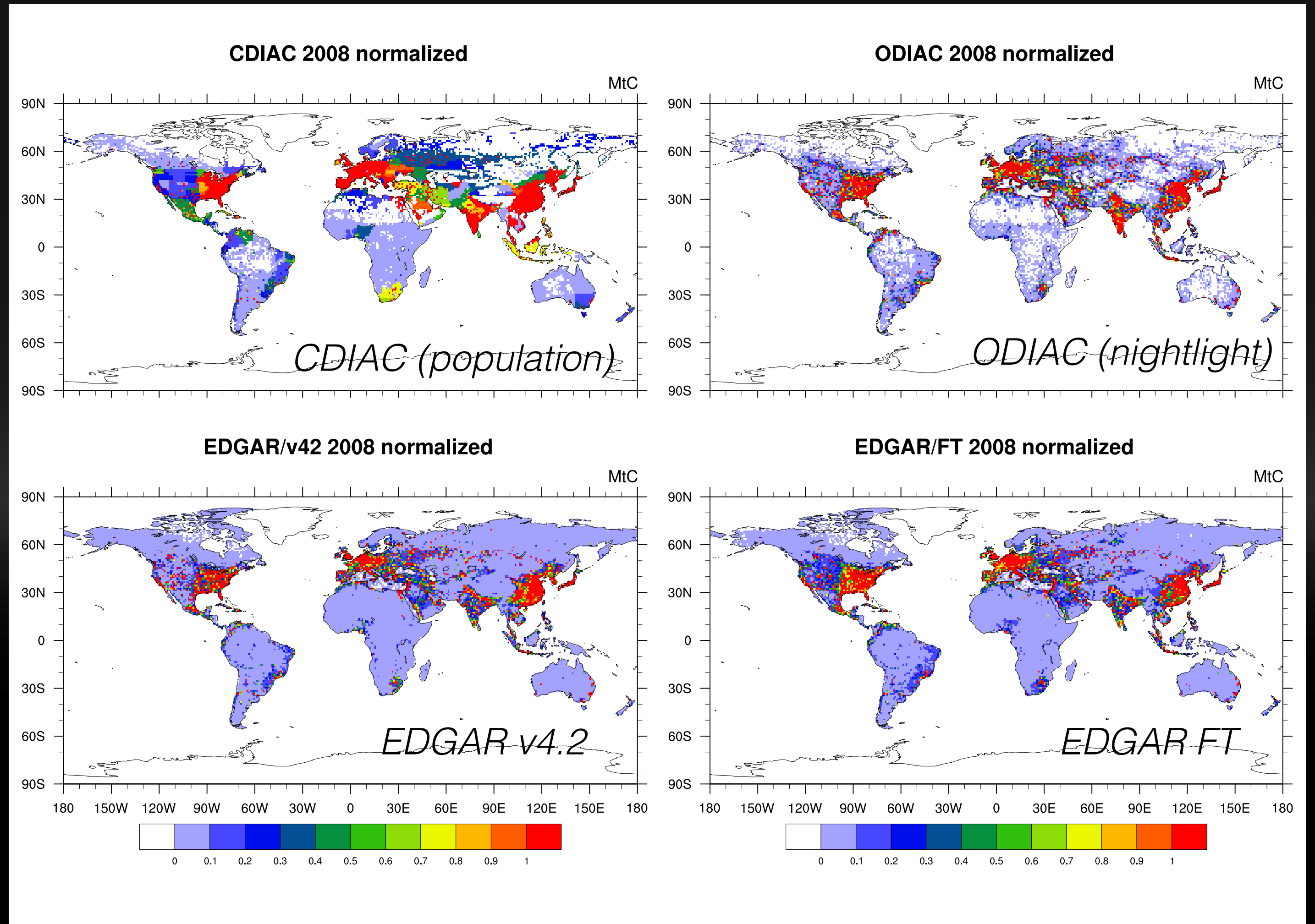
<sup>1</sup>Laboratoire des Sciences du Climat et de l'Environnement, UMR8212, Gif sur Yvette, France  
<sup>2</sup>Centre for Australian Weather and Climate Research, CSIRO Marine and Atmospheric Research, Aspendale, Australia  
<sup>3</sup>School of Life Sciences/Global Institute of Sustainability, Arizona State University, Tempe, USA  
<sup>4</sup>University of Colorado Boulder and NOAA Earth System Research Laboratory, Boulder, Colorado, USA  
<sup>5</sup>Metecological  
<sup>6</sup>Research Institu  
<sup>7</sup>Dept. of Meteor  
<sup>8</sup>School of Earth  
<sup>9</sup>Max-Planck-Ins

Correspondence  
 Received: 30 Jan  
 Revised: 4 Augu

Abstract. Atmos  
 bon fluxes from  
 ments, usually i  
 mates. Eleven se

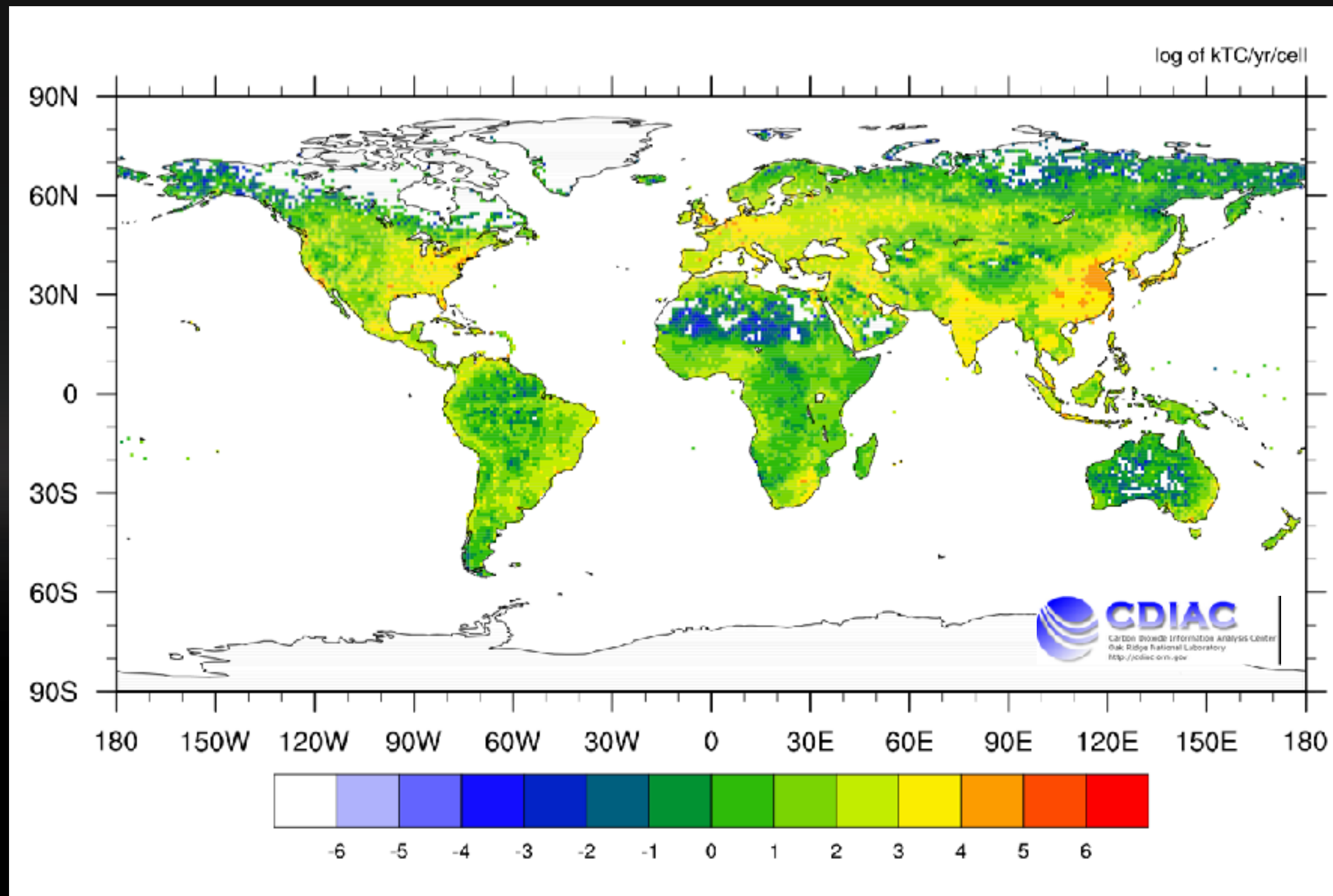


PEYLIN ET AL. (2013) BG

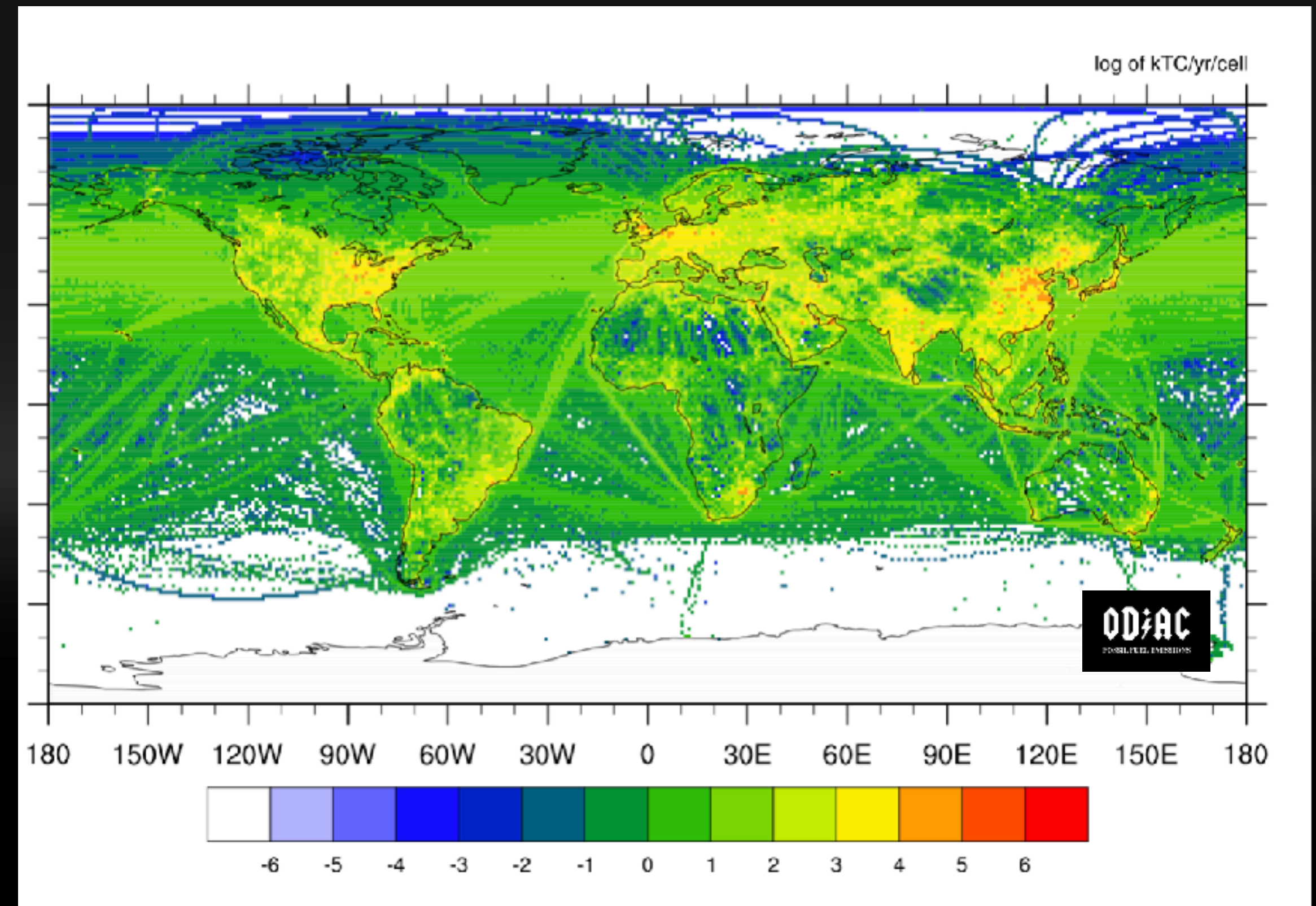


ODA ET AL. (2015); ODA ET AL. IN PREP

# REPRESENTATIONS ERRORS



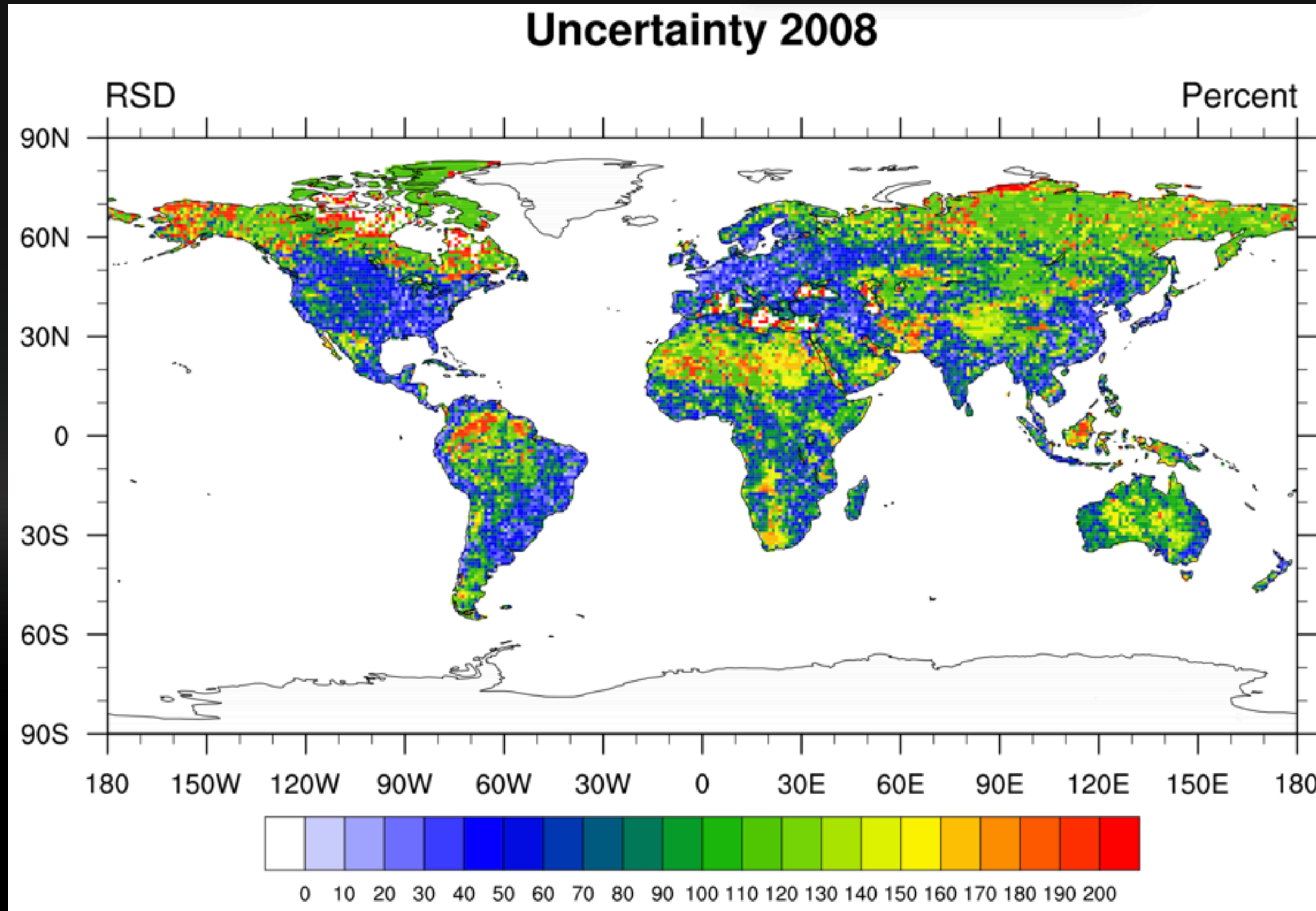
CDIAC 1DEG GRIDDED EMISSION DATA (ANDRES ET AL. 1996)



ODIAC EMISSION DATA AGGREGATED TO 1DEG

ODA ET AL. ESSD ACCEPTED

# SPATIAL UNCERTAINTY ESTIMATES



$$\delta E_{i,j}/E_{i,j} = \sqrt{(\delta M_{Total}/M_{Total})^2 + (\delta W_{i,j}/W_{i,j})^2} \quad (3)$$

ODA ET AL. 2015; ODA ET AL. IN PREP

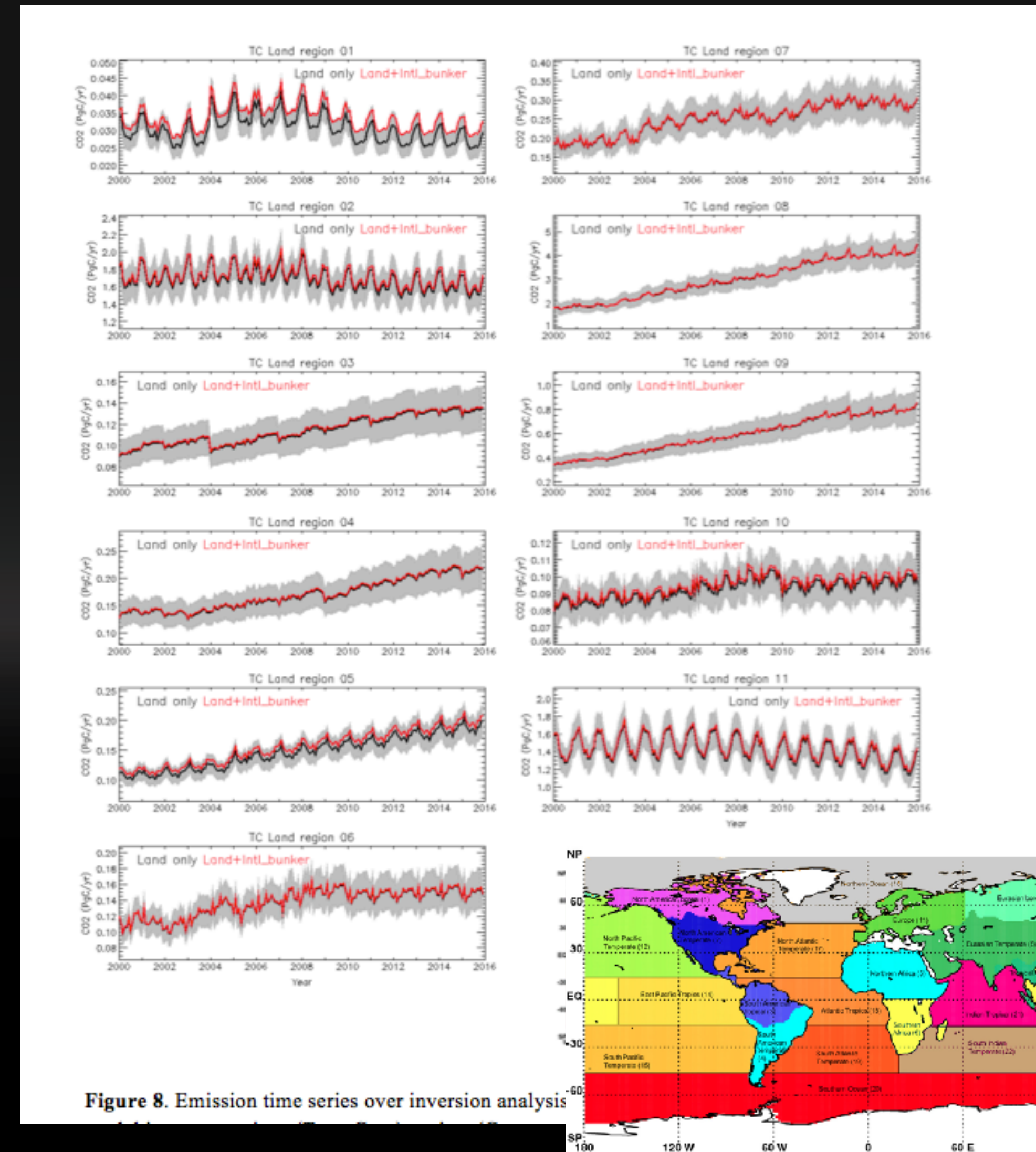
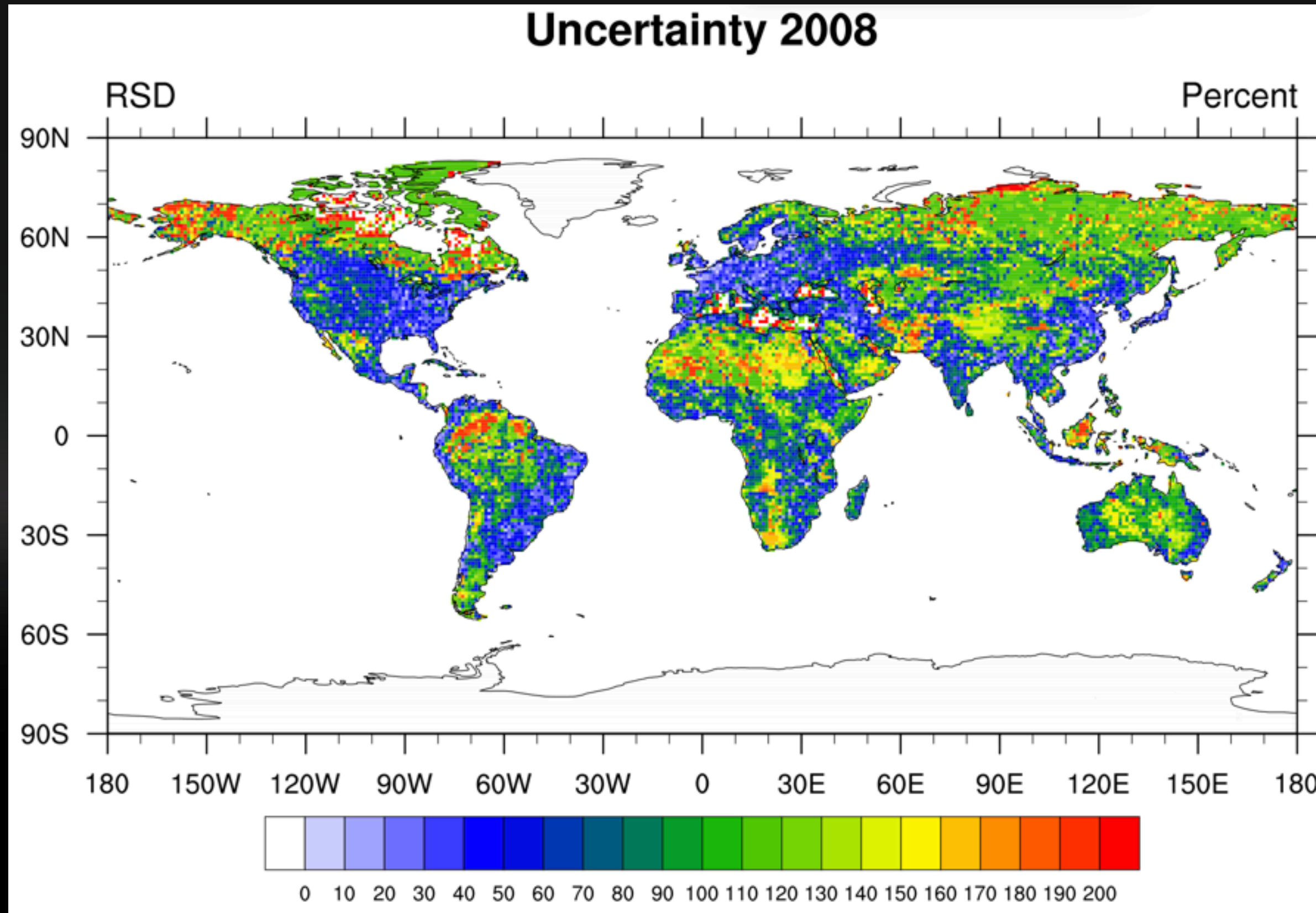


Figure 8. Emission time series over inversion analysis

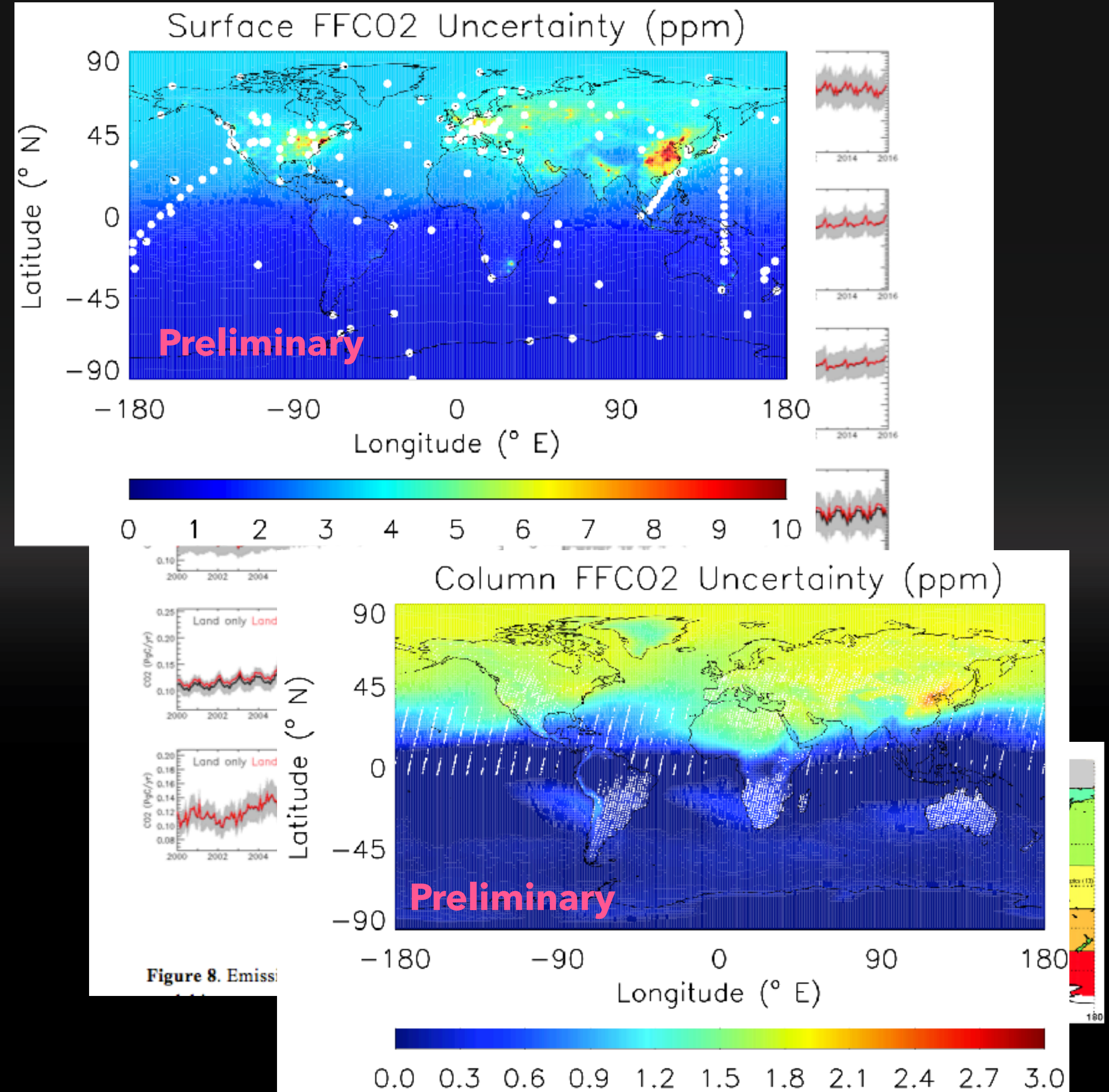
ODA ET AL. ESSD ACCEPTED

# SPATIAL UNCERTAINTY ESTIMATES



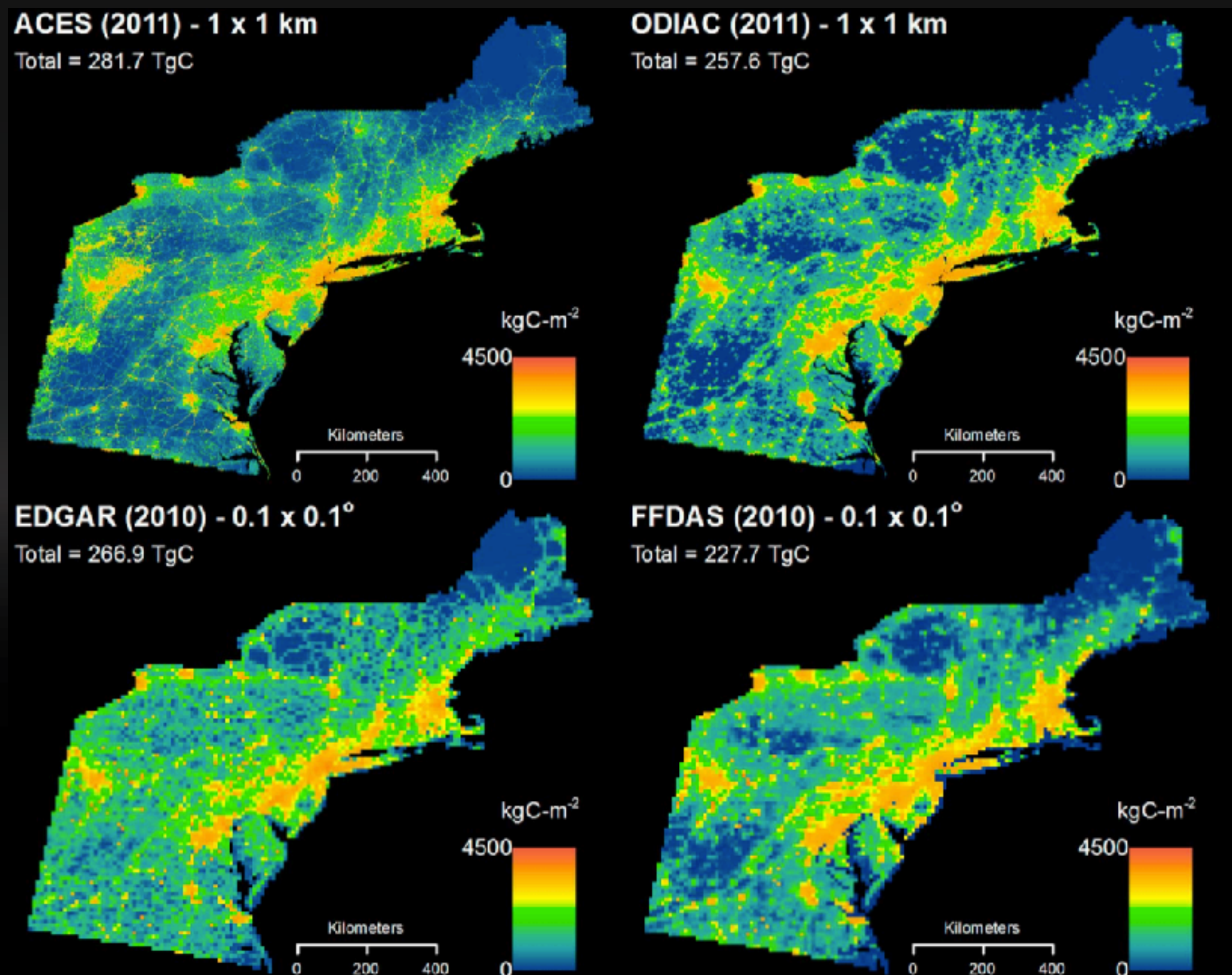
$$\delta E_{i,j}/E_{i,j} = \sqrt{(\delta M_{Total}/M_{Total})^2 + (\delta W_{i,j}/W_{i,j})^2} \quad (3)$$

ODA ET AL. 2015; ODA ET AL. IN PREP

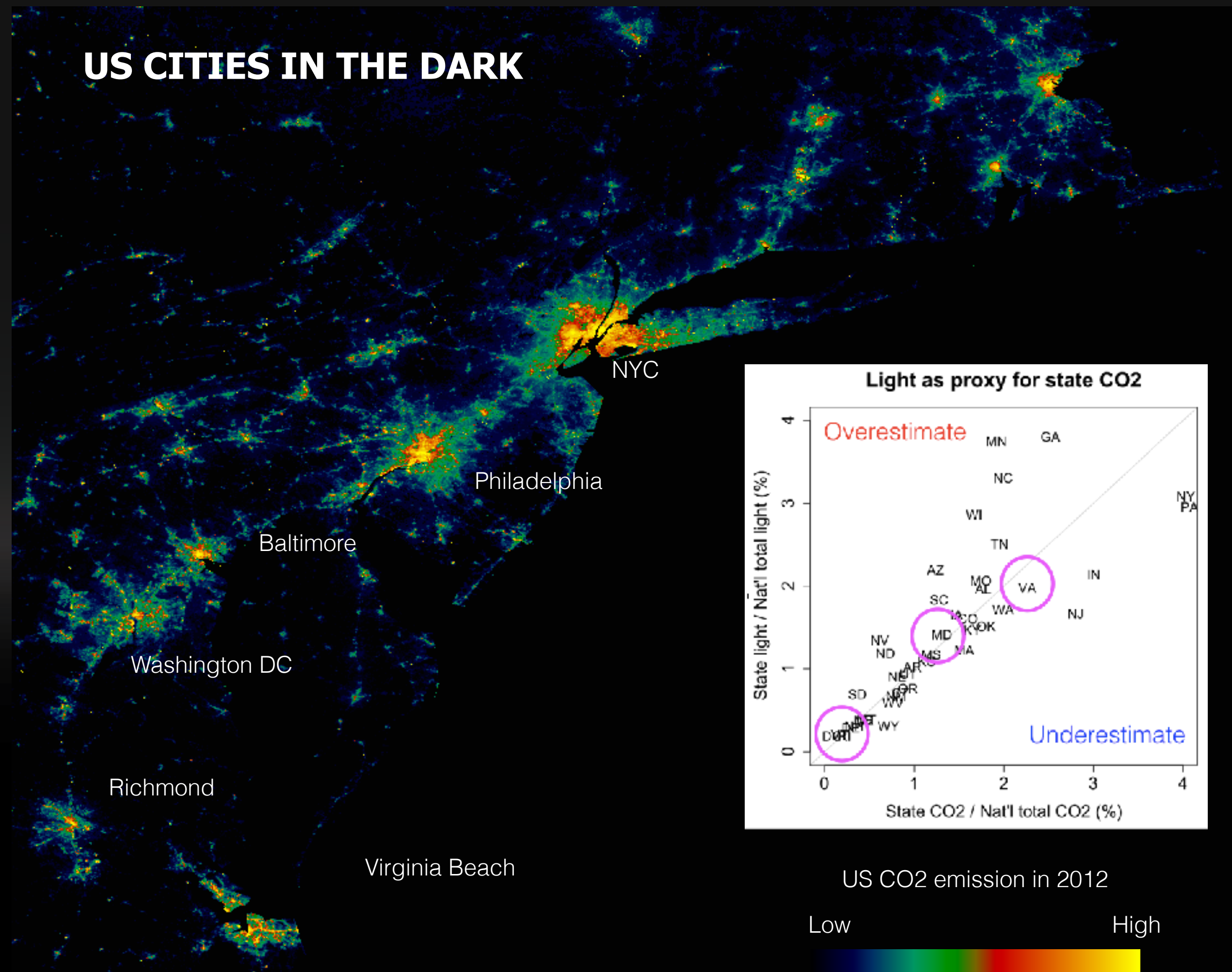




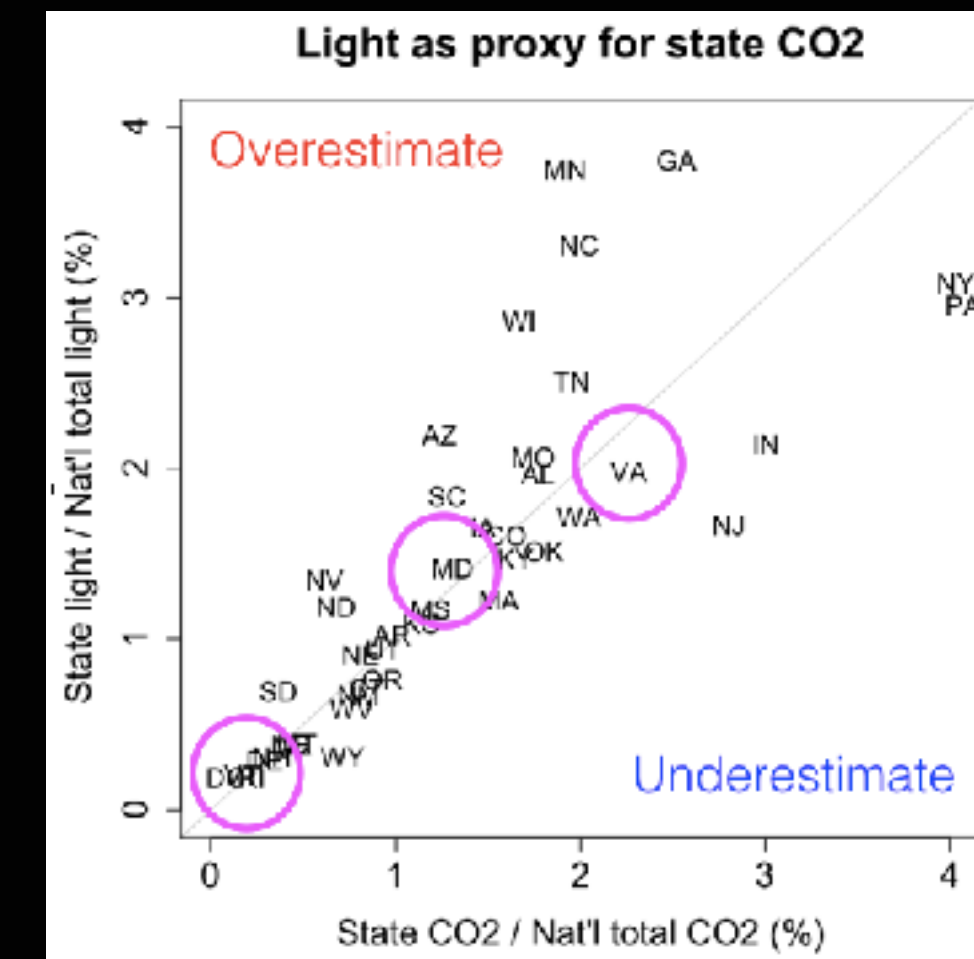
# GRANUALITY IN GRIDDED EMISSIONS



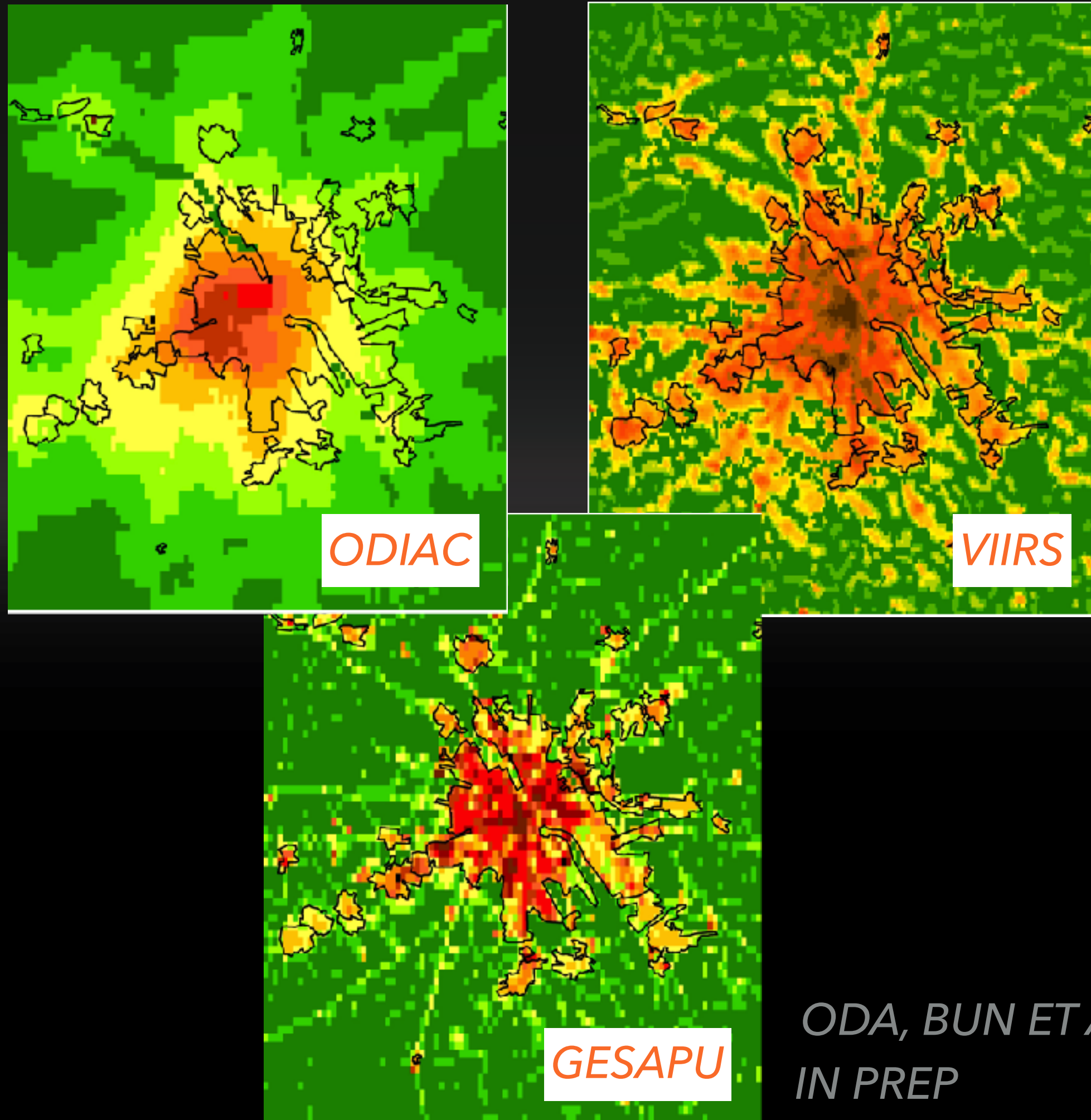
GATELY AND HUTYLA (2017) JGR



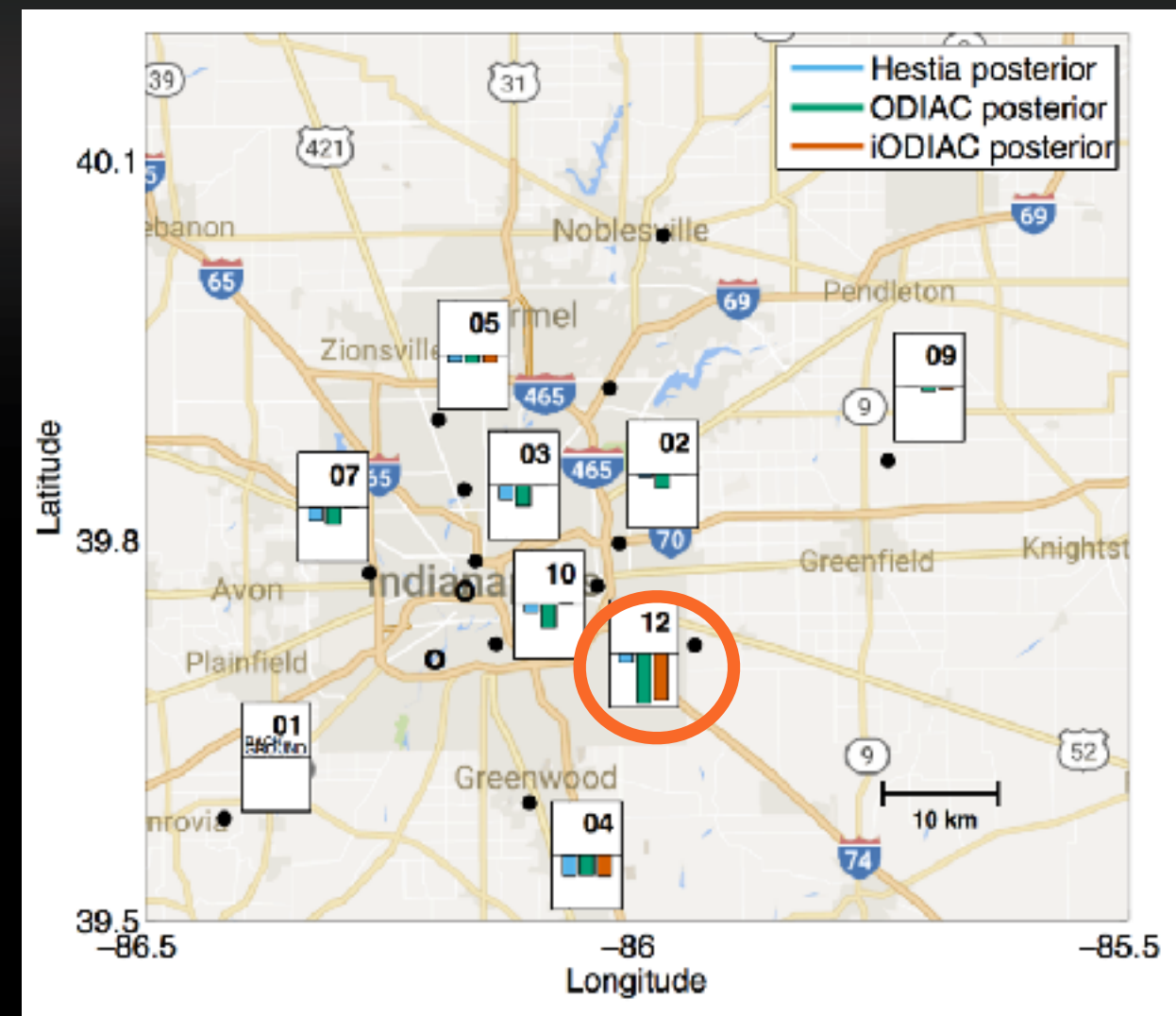
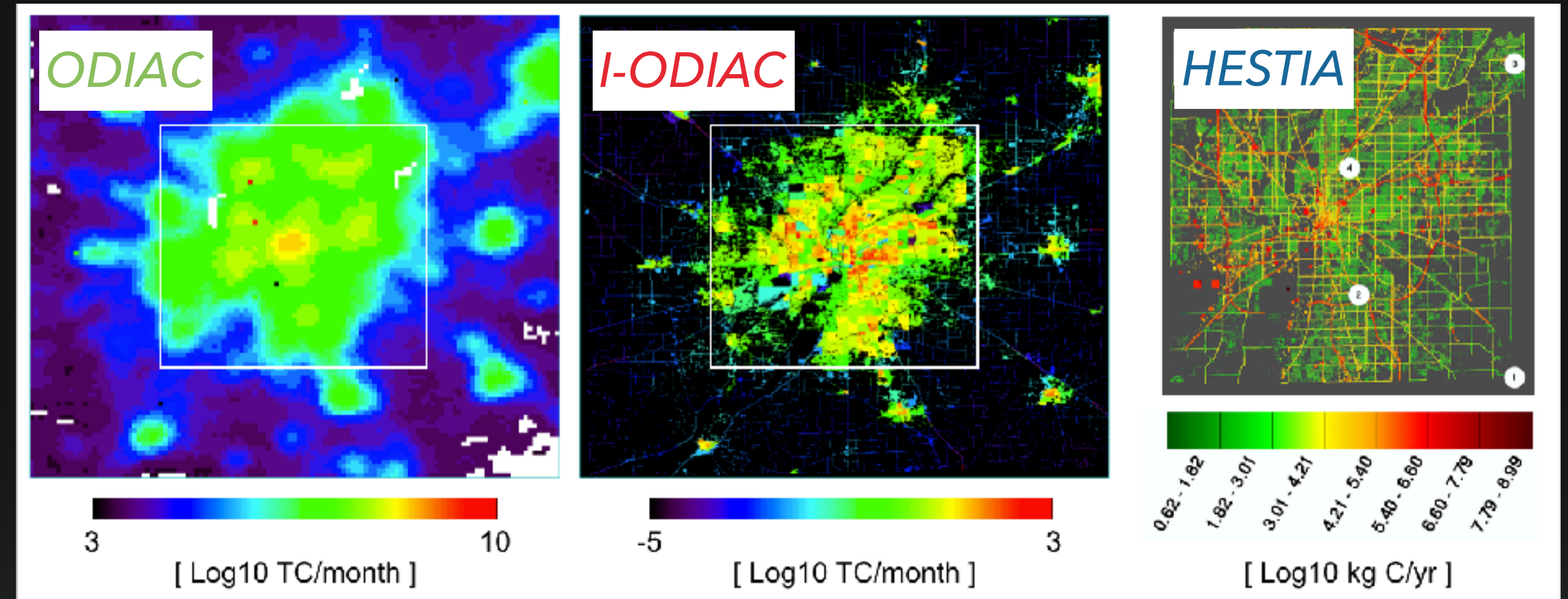
ODA, ROMAN, WANG ET AL IN PREP FOR CMB



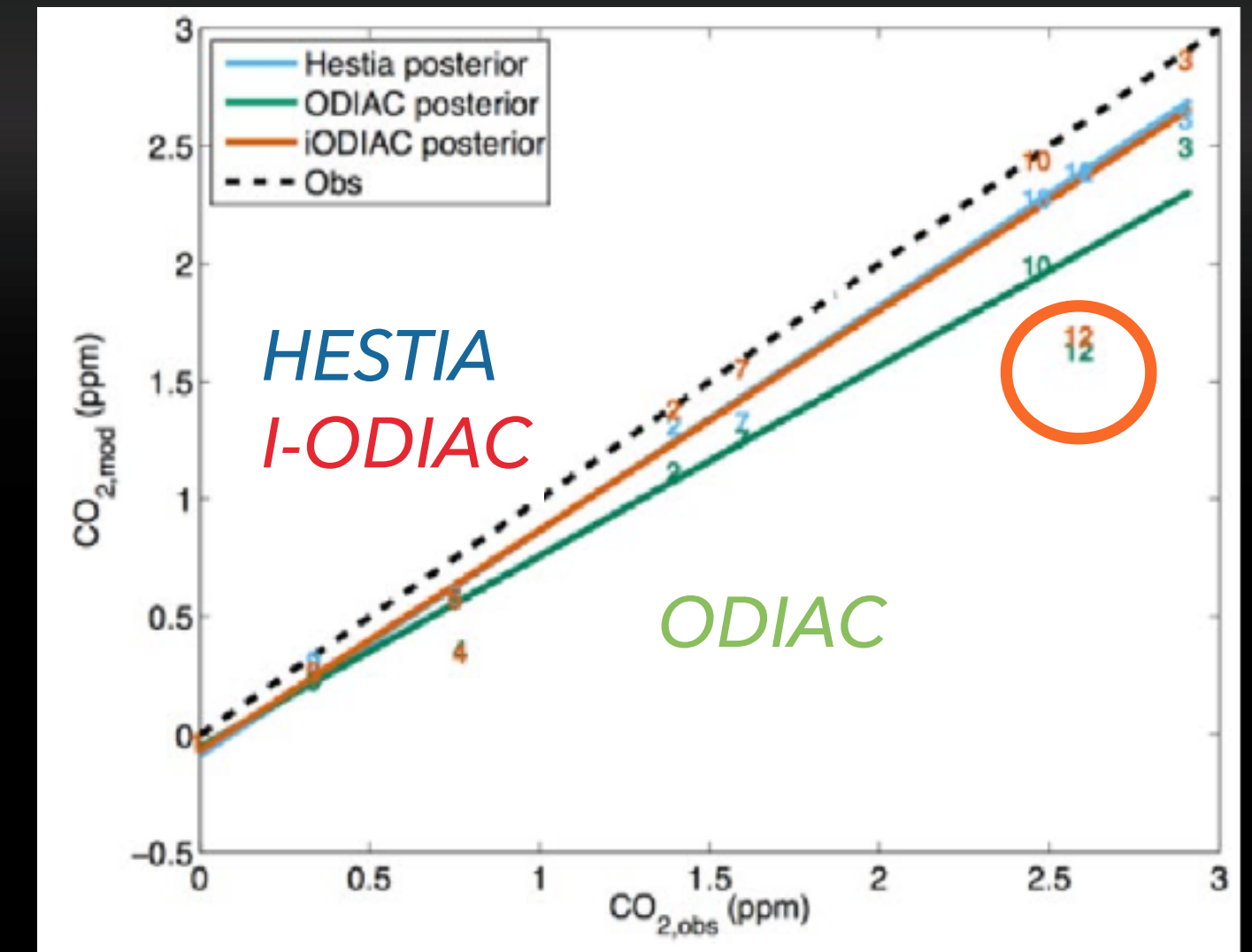
# PROXY APPROACH IS STILL USEFUL



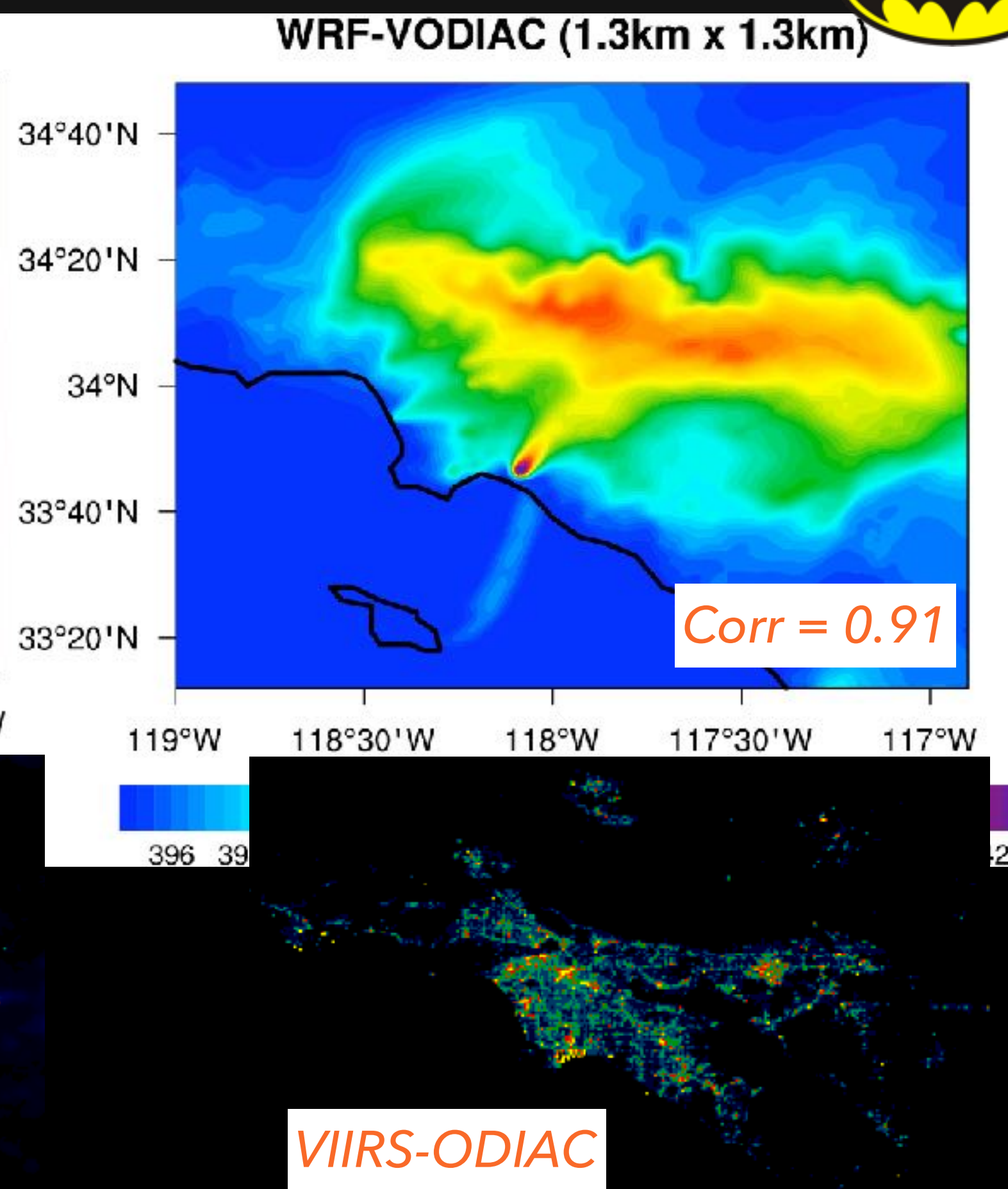
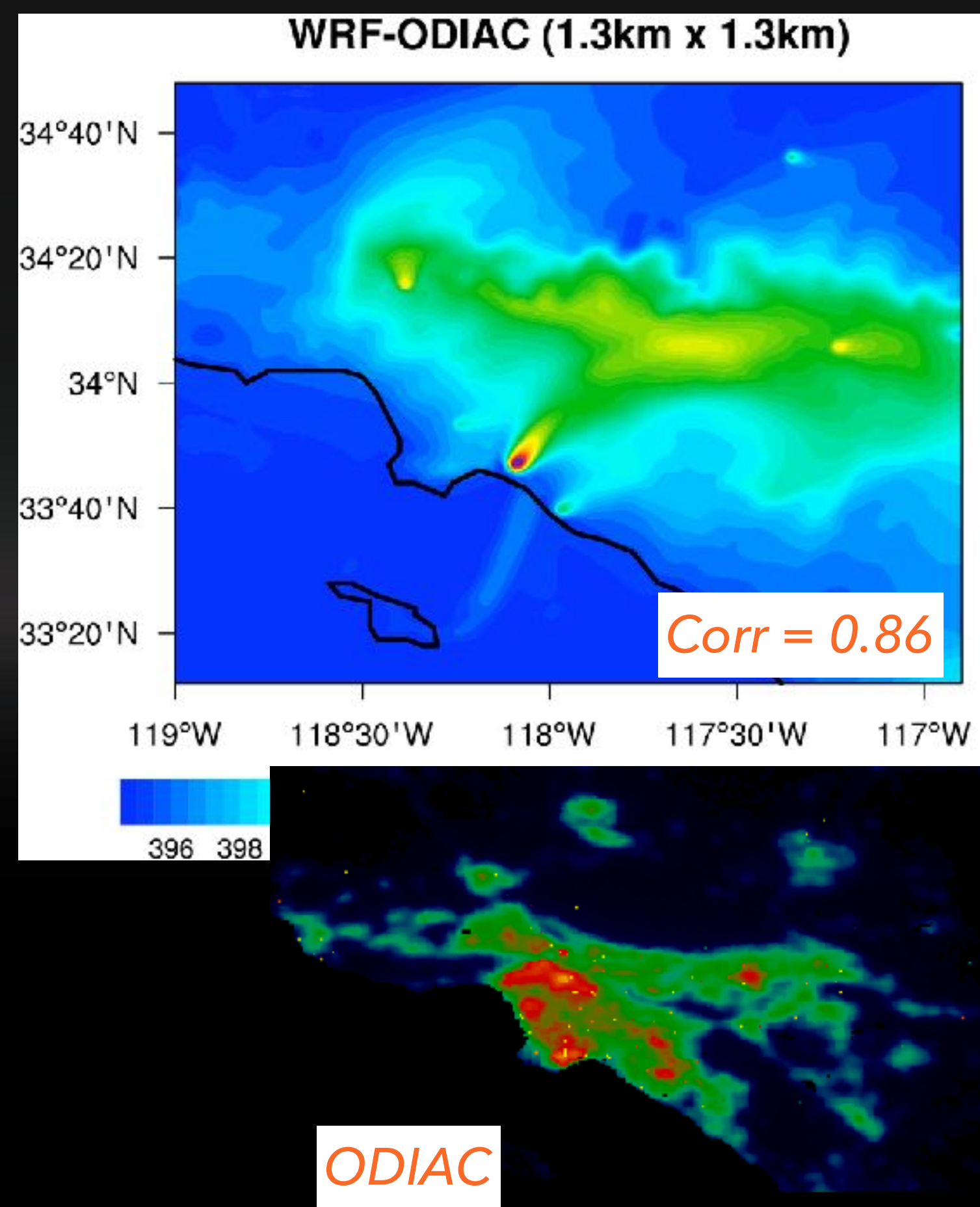
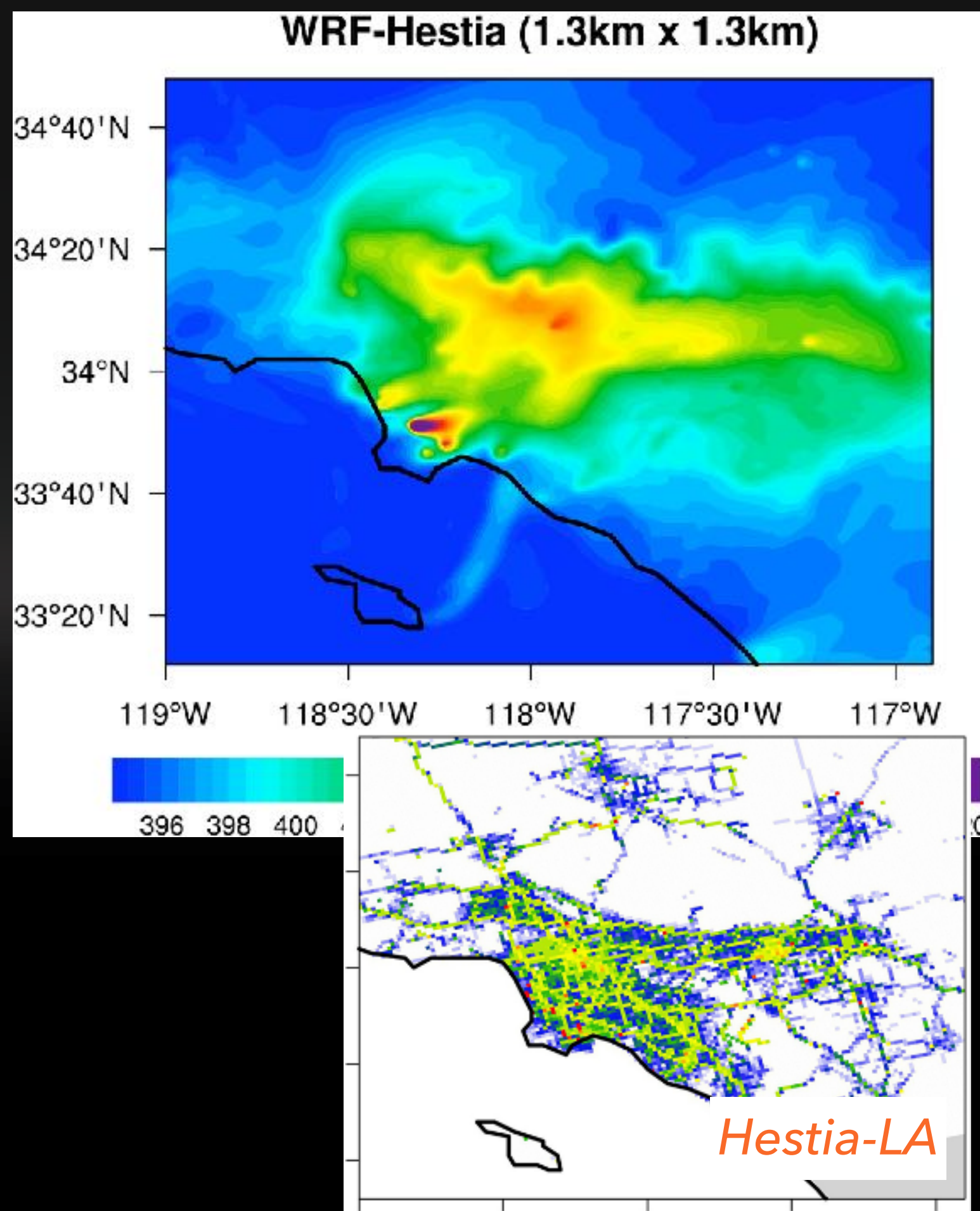
ODA, BUN ET AL.  
IN PREP



LAUVUAX ET AL. 2016 JGR; ODA ET AL. 2017 ELEMENTA



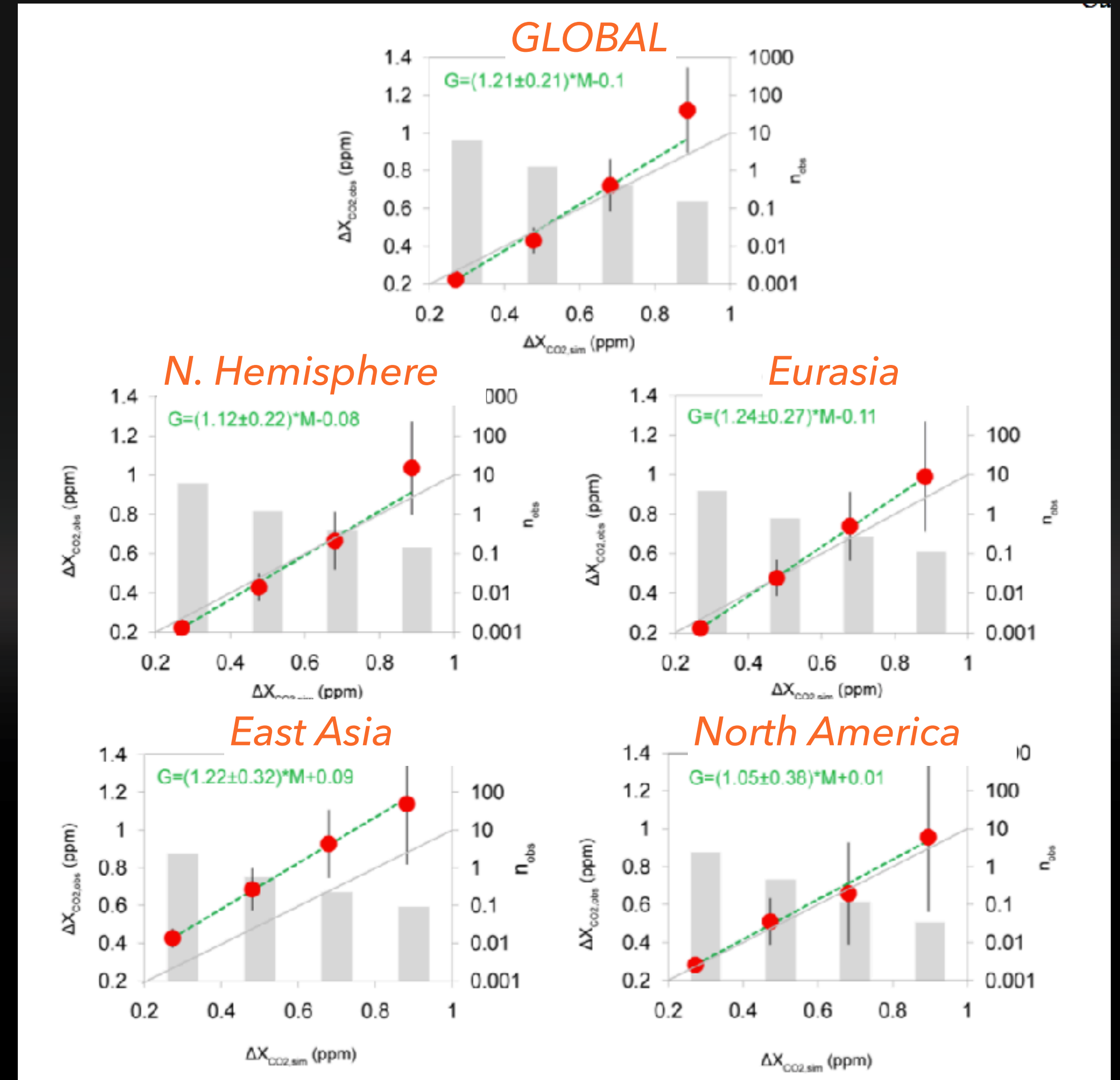
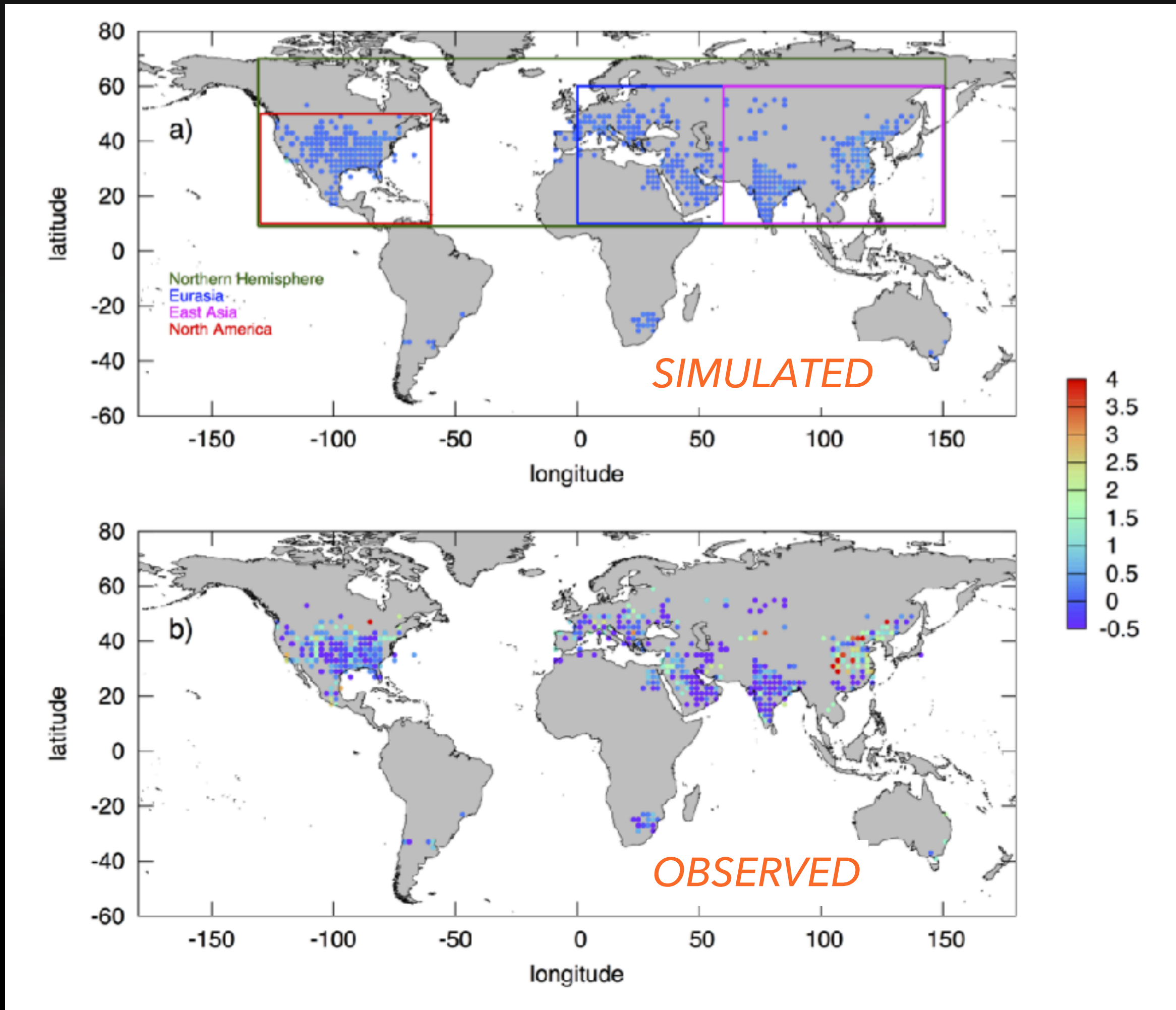
# ATMOSPHERIC CO2 AS A METRIC



FENG, LAUVAUX ET AL 2016 ACP

ODA, FENG, LAUVAX, JUST COMPLETED A FEW DAYS AGO

# TOWARDS GLOBAL CO2 MONITORING

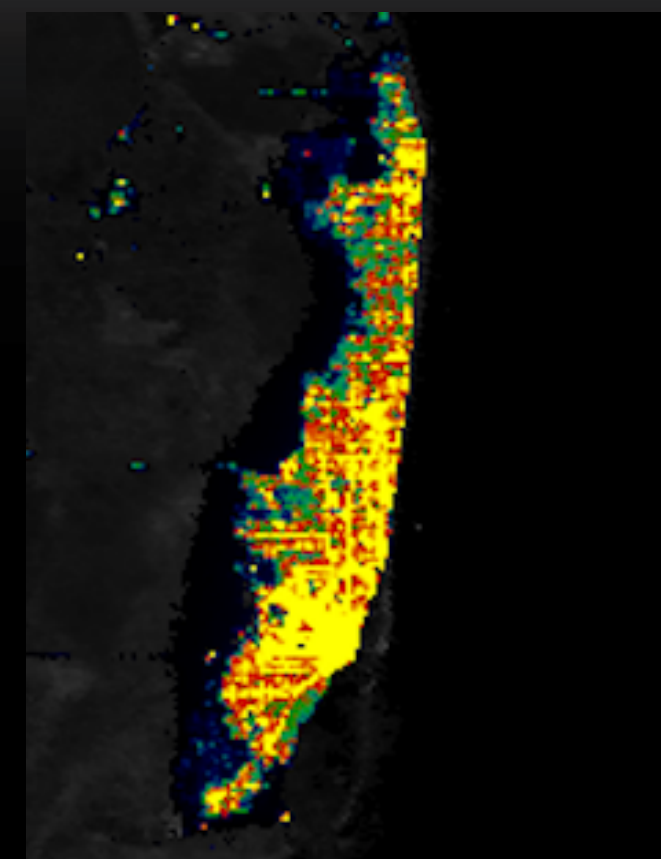
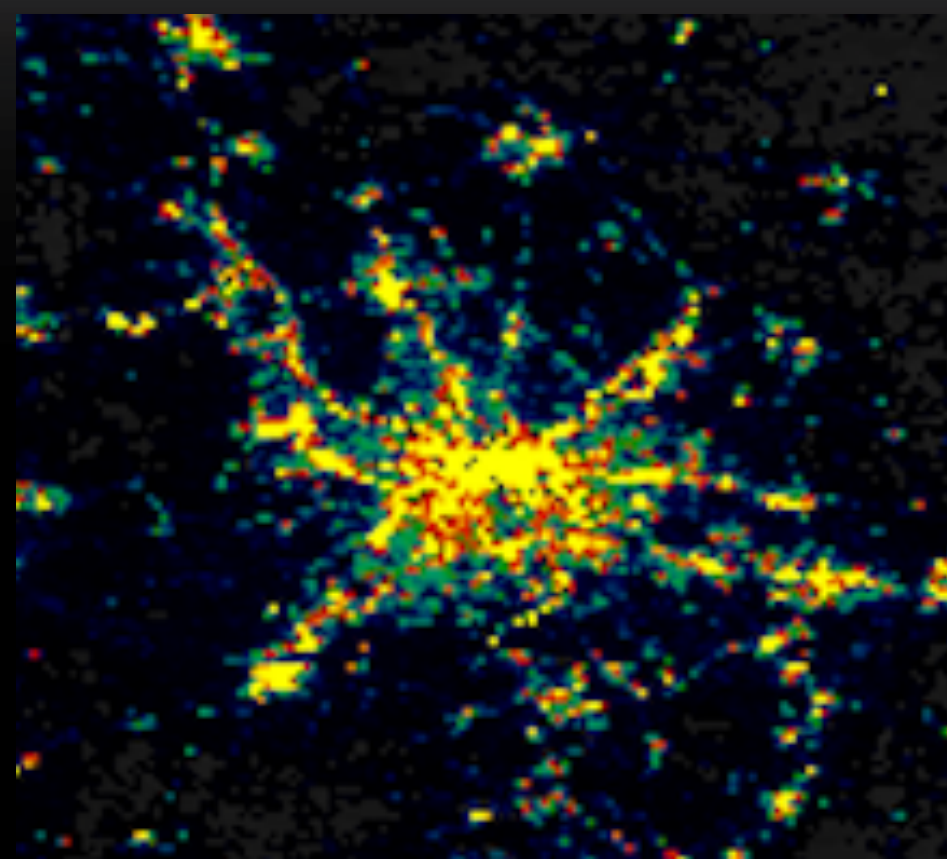
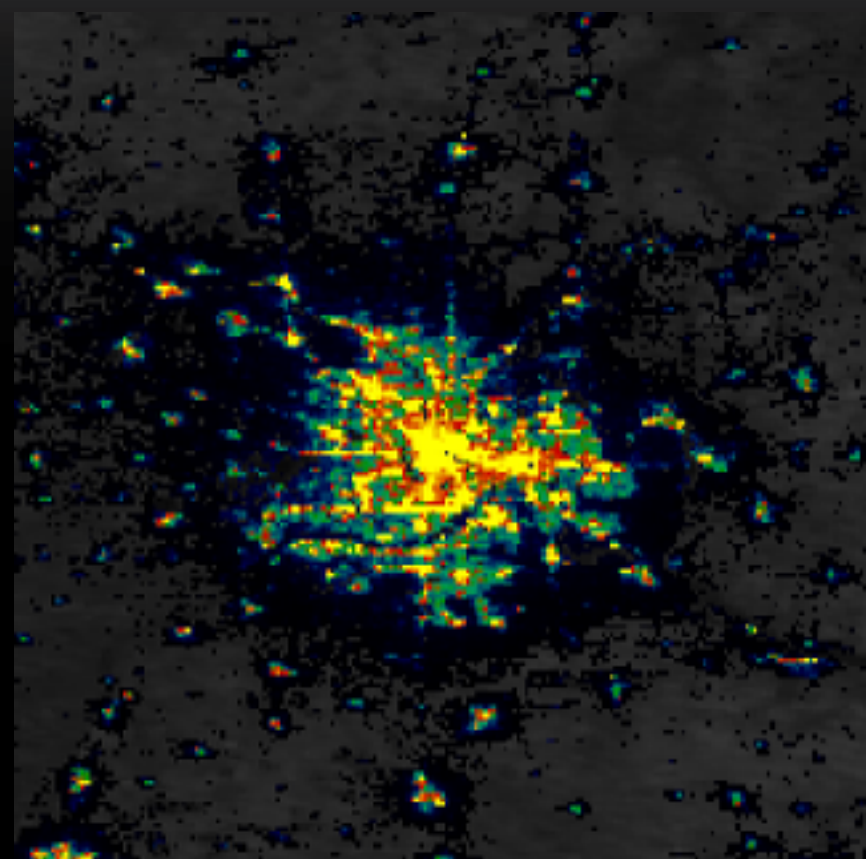
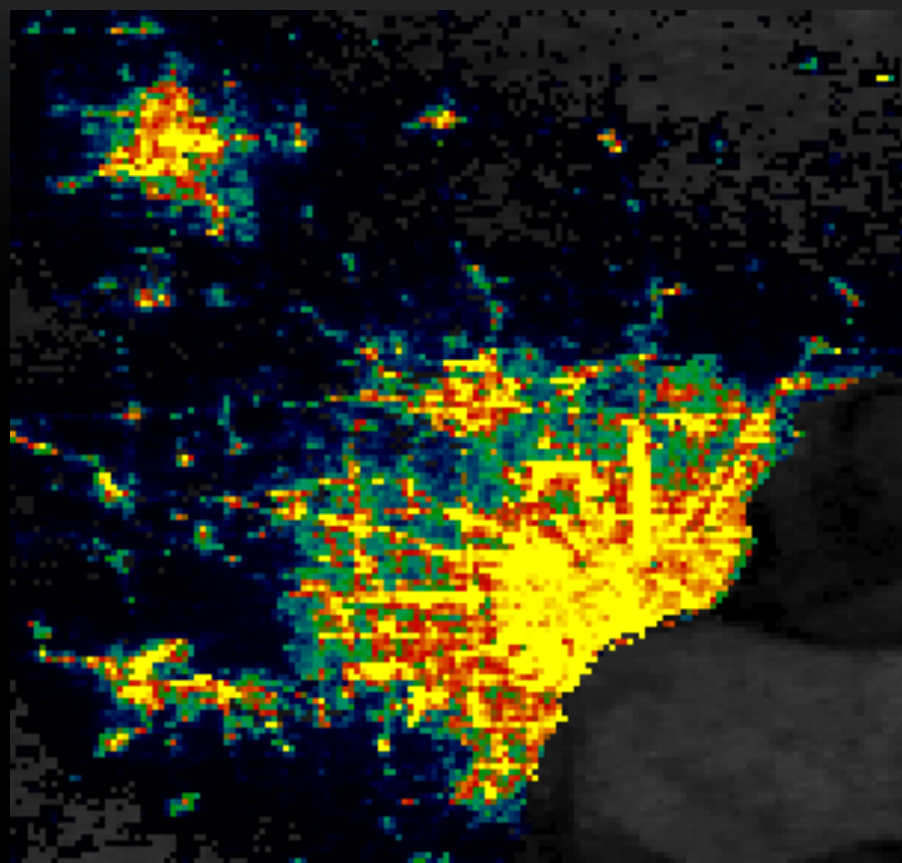
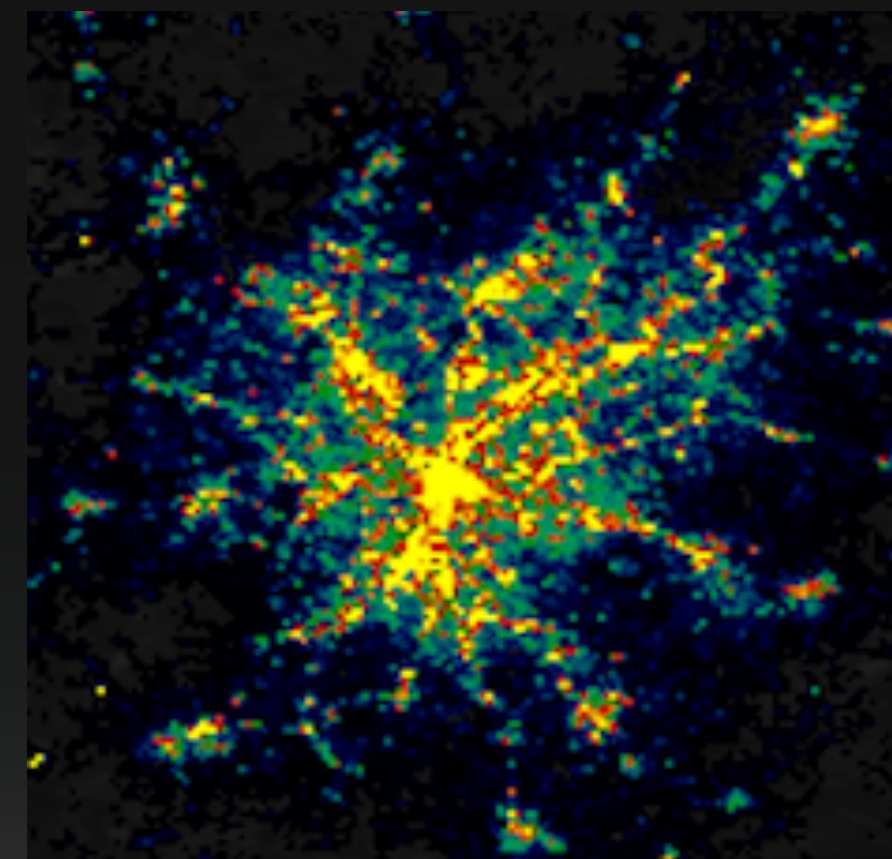
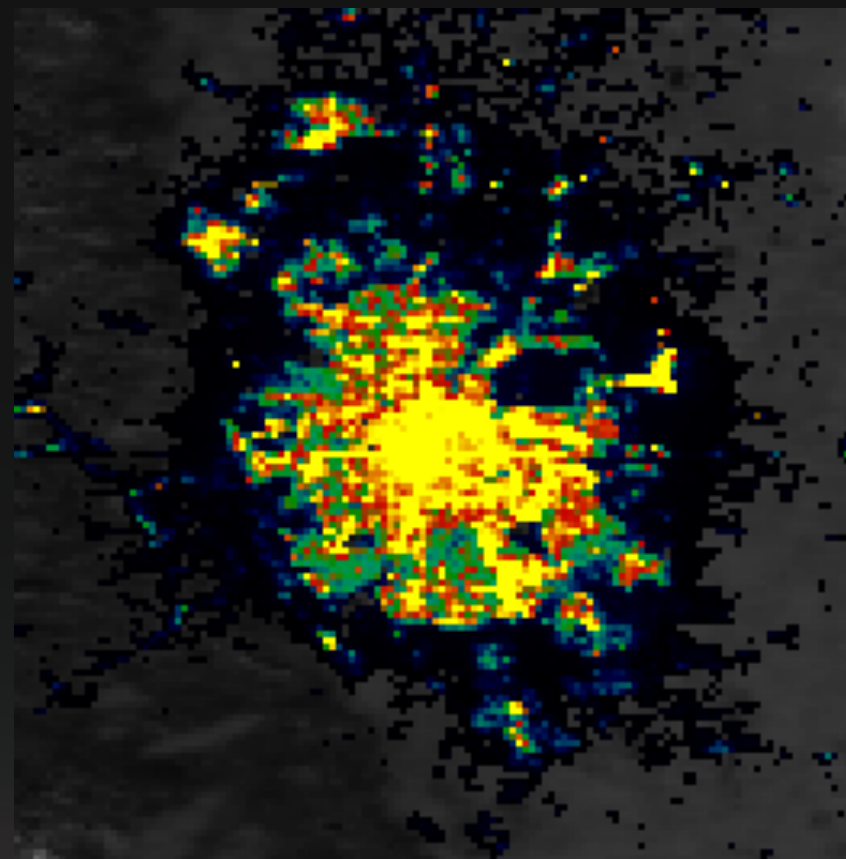
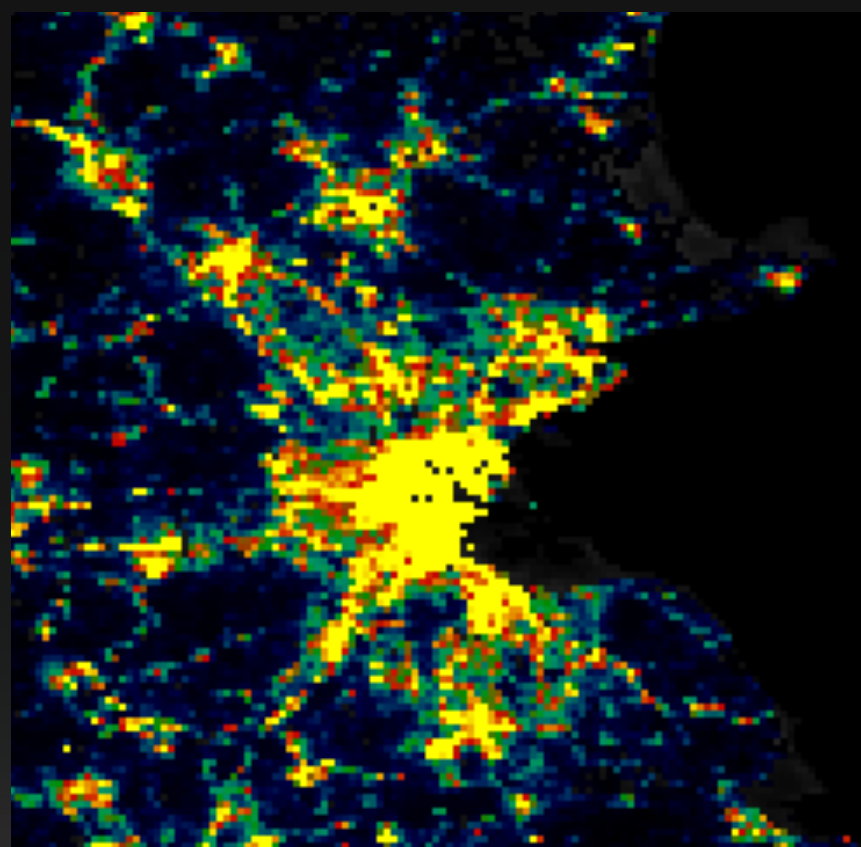
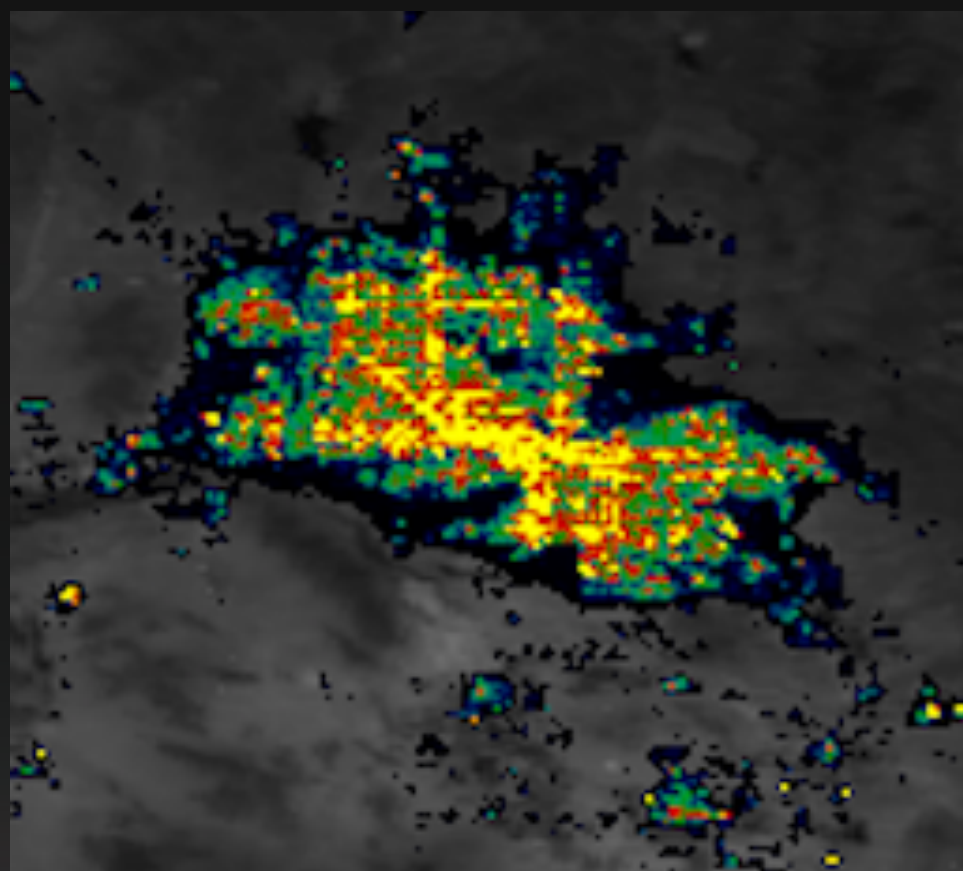


JANARDANAN, MAKSYUTOV, ODA ET AL 2016 GRL

# DISCUSSIONS/SUMMARY/TAKE HOME

- ▶ *The granularity of emission data matters - The use of inter-inventory differences for uncertainty analysis needs to be done at a “appropriate” spatial resolution with acknowledging the fundamental limitations.*
- ▶ *With better granularity, we are still stuck in the same problem - Fine-grained emission data are extremely useful. But those are also emission data (not mechanistic models), and not measurements. Fine-grained emission data also share the same difficulties and limitations.*
- ▶ *New opportunity from CO2 world - New CO2 measurements (dense ground network and satellites) should place us to a better position to objectively quantify disaggregation errors and emission spatial uncertainties using atmospheric simulations, and produce improved emission patterns.*
- ▶ *Proxy-based downscaling remains useful - Using atmospheric CO2 data and modeling, we can tell what is a “good” proxy and possibly what the true emissions look like. New NASA’s Nighttime Environmental Product will be a great proxy data to produce improved emission distributions globally.*

# CAN YOU NAME THEM?



QUESTION/COMMENT? [TOM.ODA@NASA.GOV](mailto:TOM.ODA@NASA.GOV)