

# The use of airborne and ground based atmospheric observations in carbon cycle research

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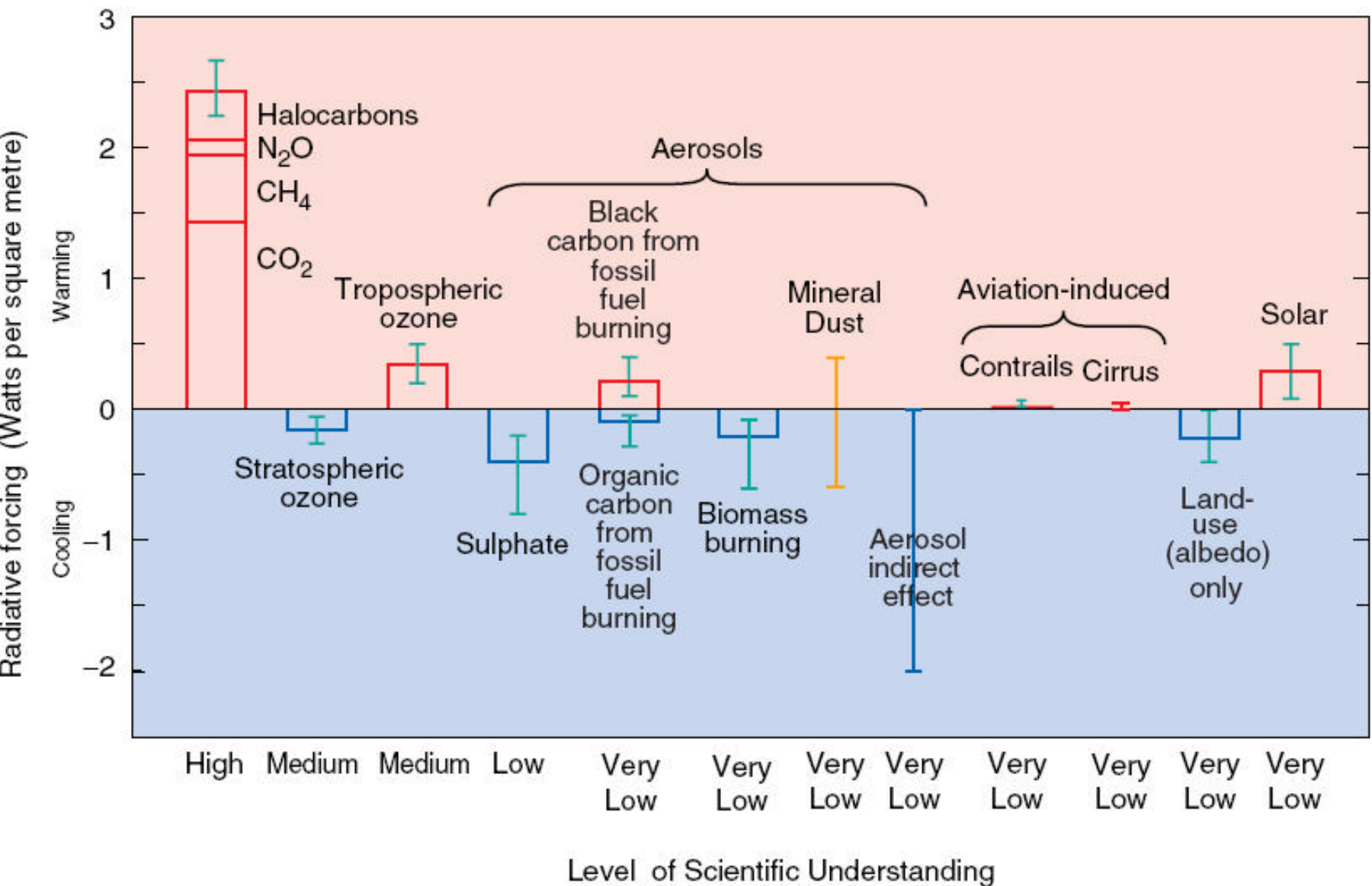


Presented at the  
ECMWF Seminar on  
Global Earth-System Monitoring  
5-9 September 2005

# Overview

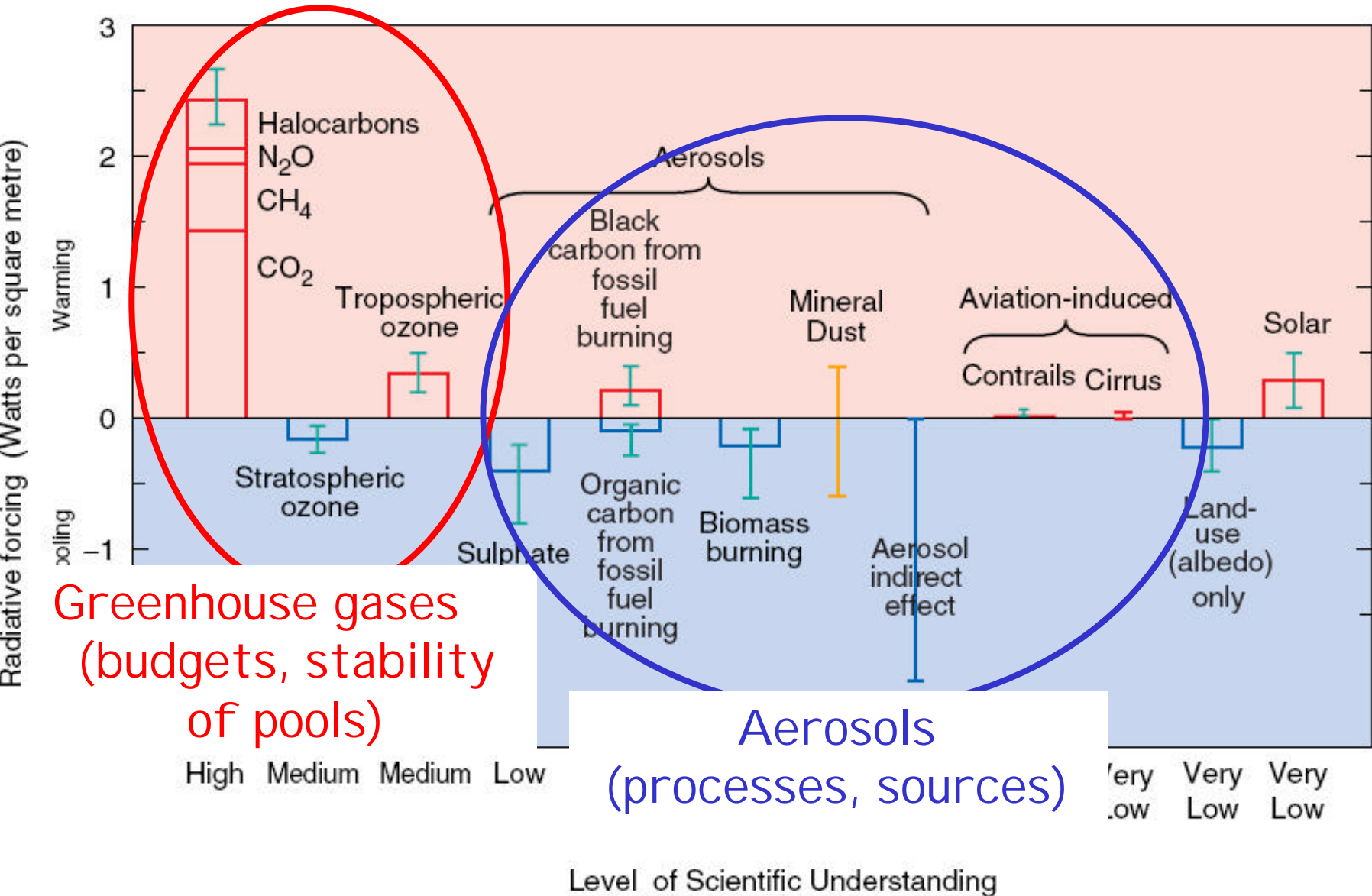
- Introduction: Why Carbon Cycle Research
- Global Observations of atmospheric CO<sub>2</sub>:
  - from remote islands to places nearby
  - from decades to seconds
  - Challenge: continental boundary layer, closeness to strong sources & sinks
- Airborne measurements of tracer distributions
  - Surface fluxes on regional scales (biosphere-atmosphere exchange, ...)
  - Transport processes (tropospheric mixing, convection, ...)
- Hypothesis: Airborne intensives provide paradigm datasets to
  - help design & test (falsify/validate) modeling frameworks
  - help integrate ground based data into data assimilation systems

# The global mean radiative forcing of the climate system for the year 2000, relative to 1750





# The global mean radiative forcing of the climate system for the year 2000, relative to 1750

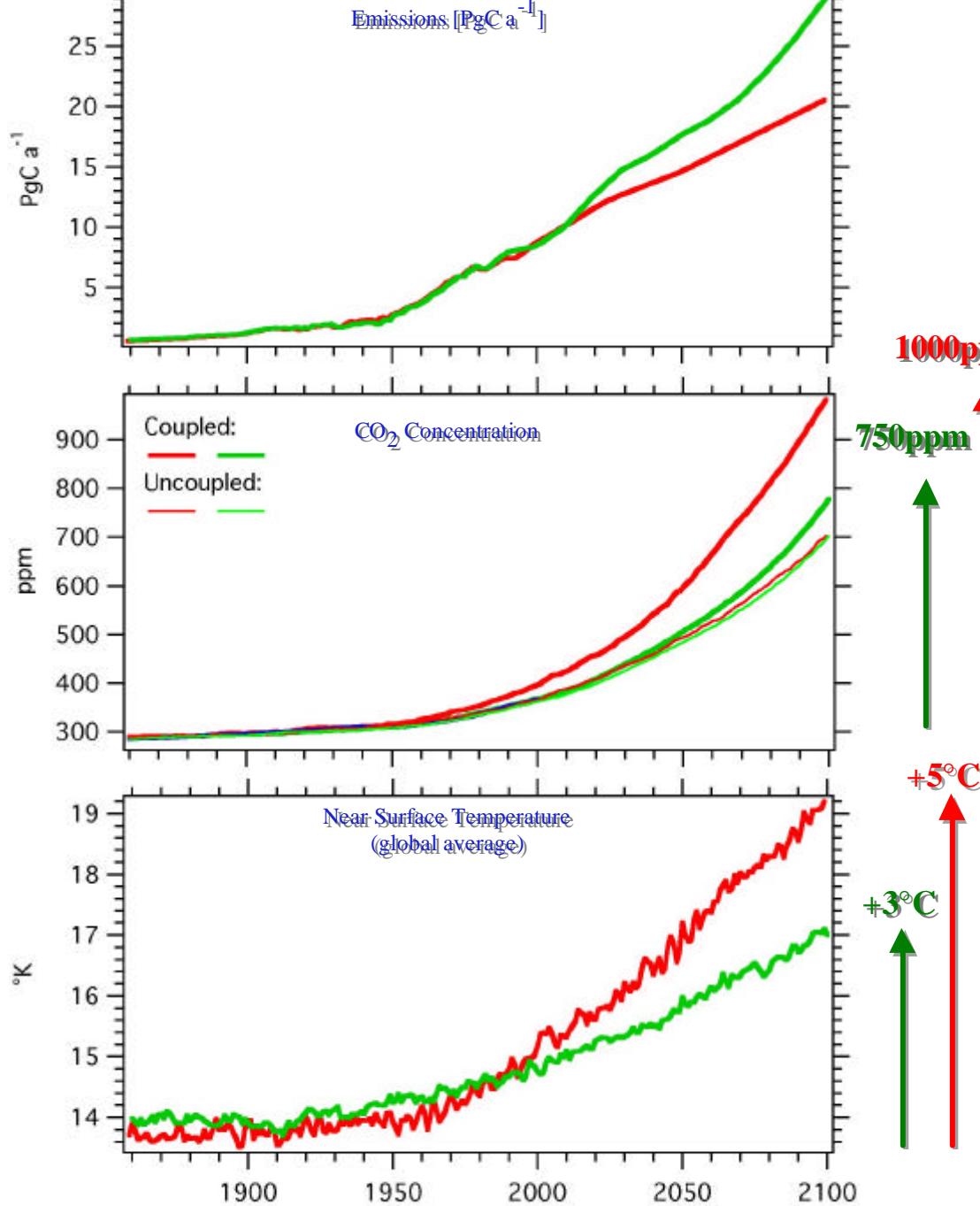


# Fundamental Carbon Cycle Questions

- Where and through which process is the excess anthropogenic carbon being taken up by land and ocean?
- What and how large are the key feedback links between the carbon cycle and the physical climate system?
- What is the carbon budget of a particular region (continent, country)?

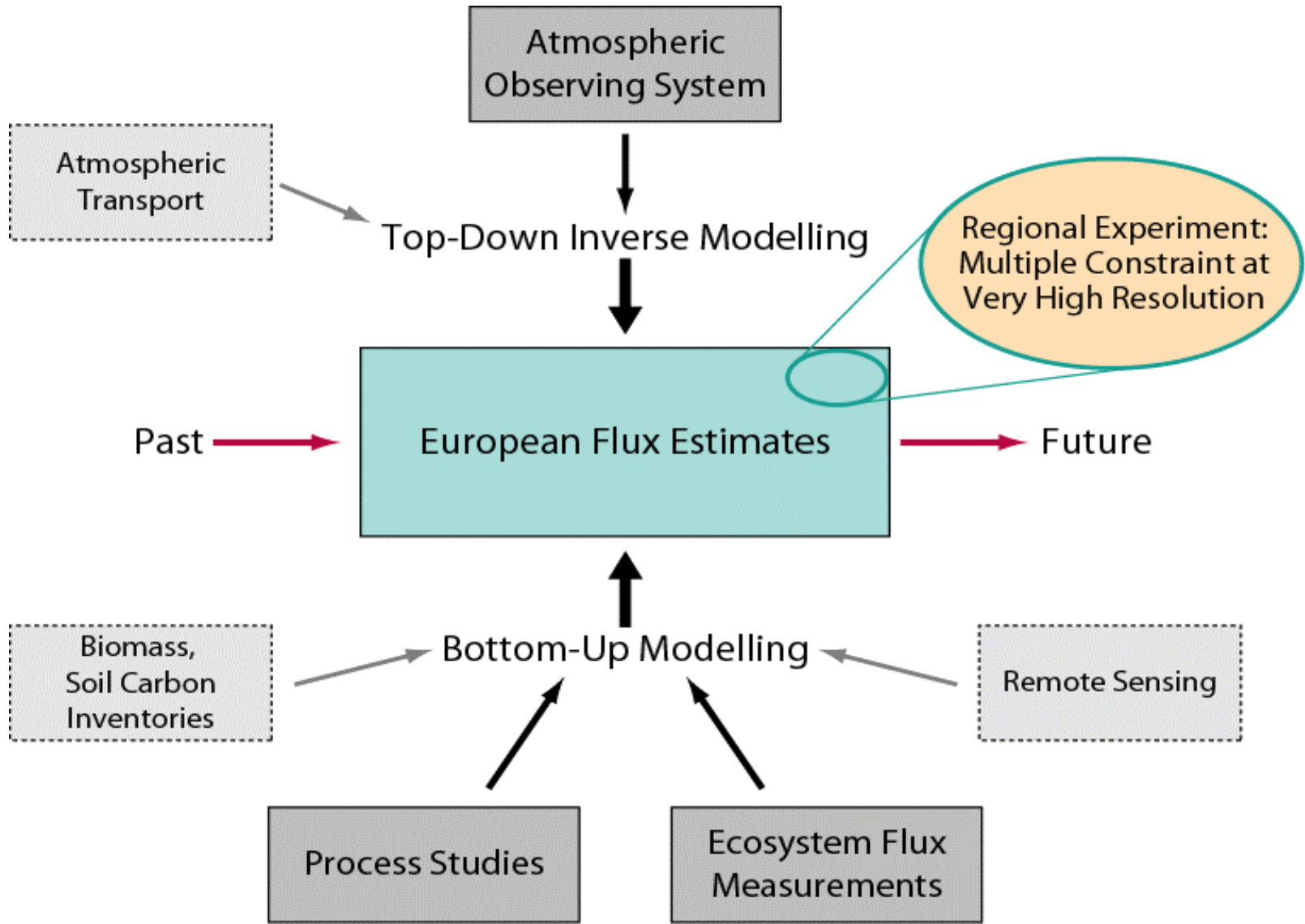
# First Scenarios Calculated with Coupled Carbon Cycle - Climate Models

**Hadley** — (red line)  
**IPSL** — (green line)

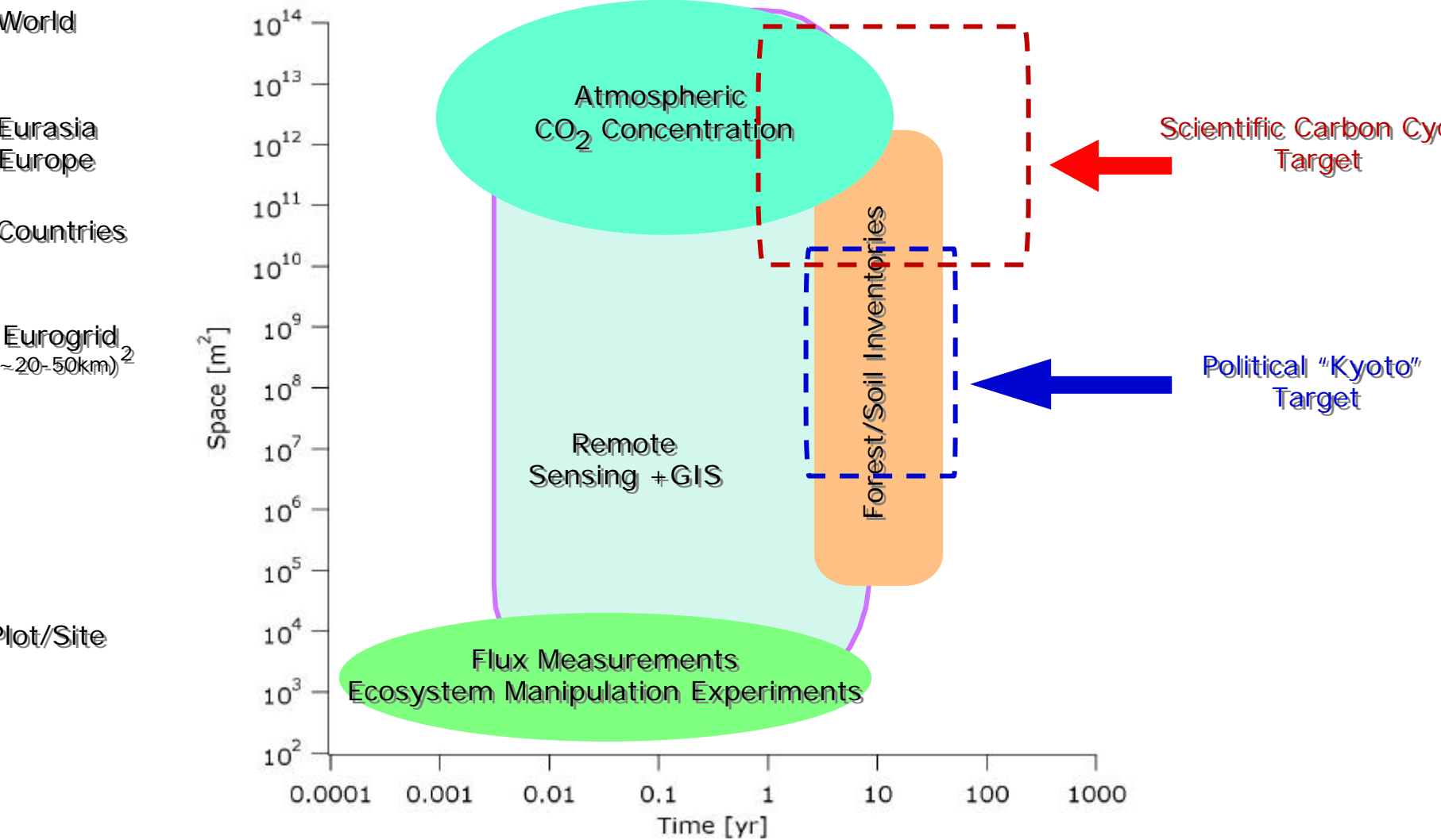


*Cox et al. 2001, Dufrene et al., 2001  
IPCC 2001*

# Estimating Regional Carbon Balances: Top-Down vs. Bottom-Up Approach

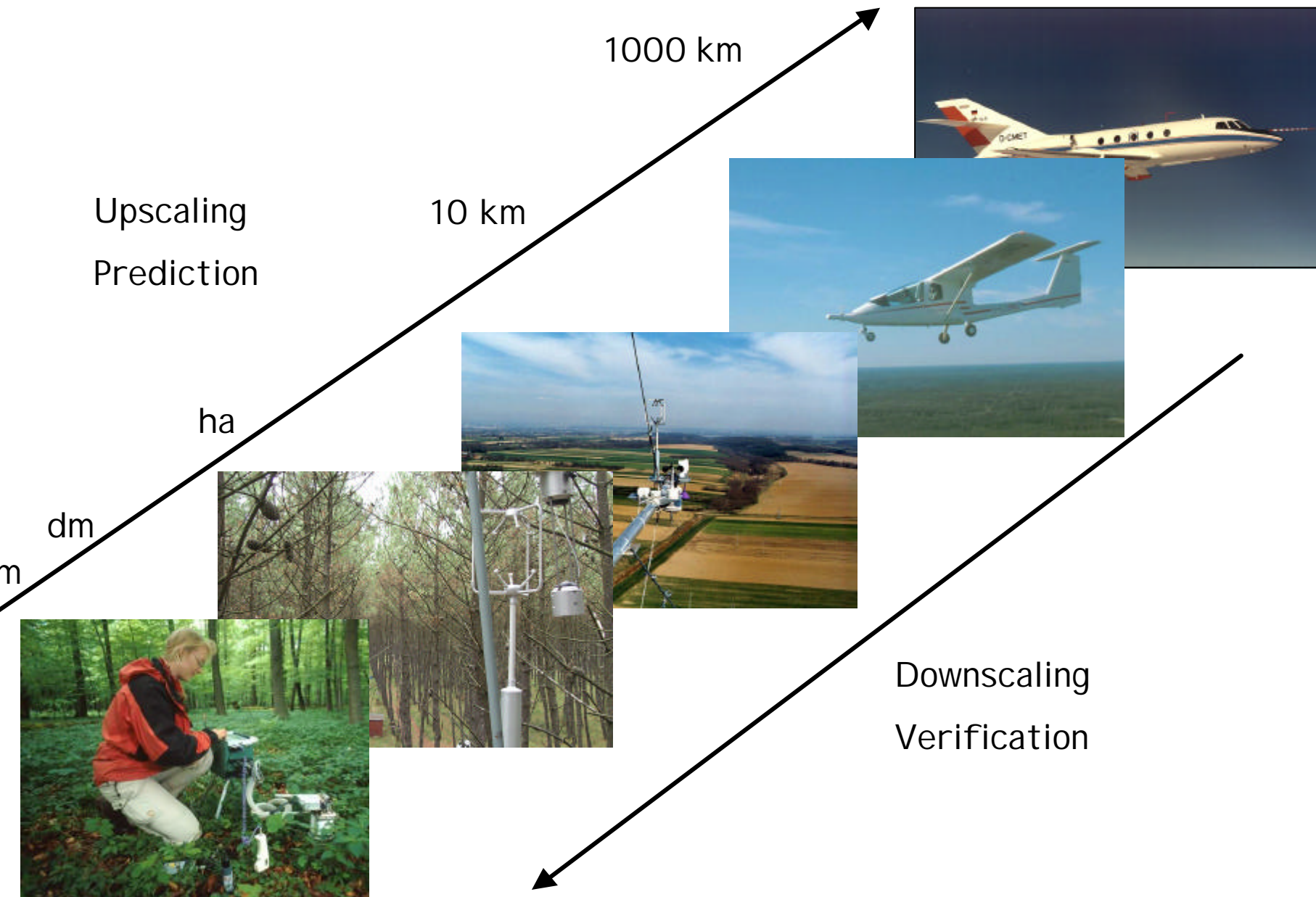


# Carbon Cycle Observing Systems

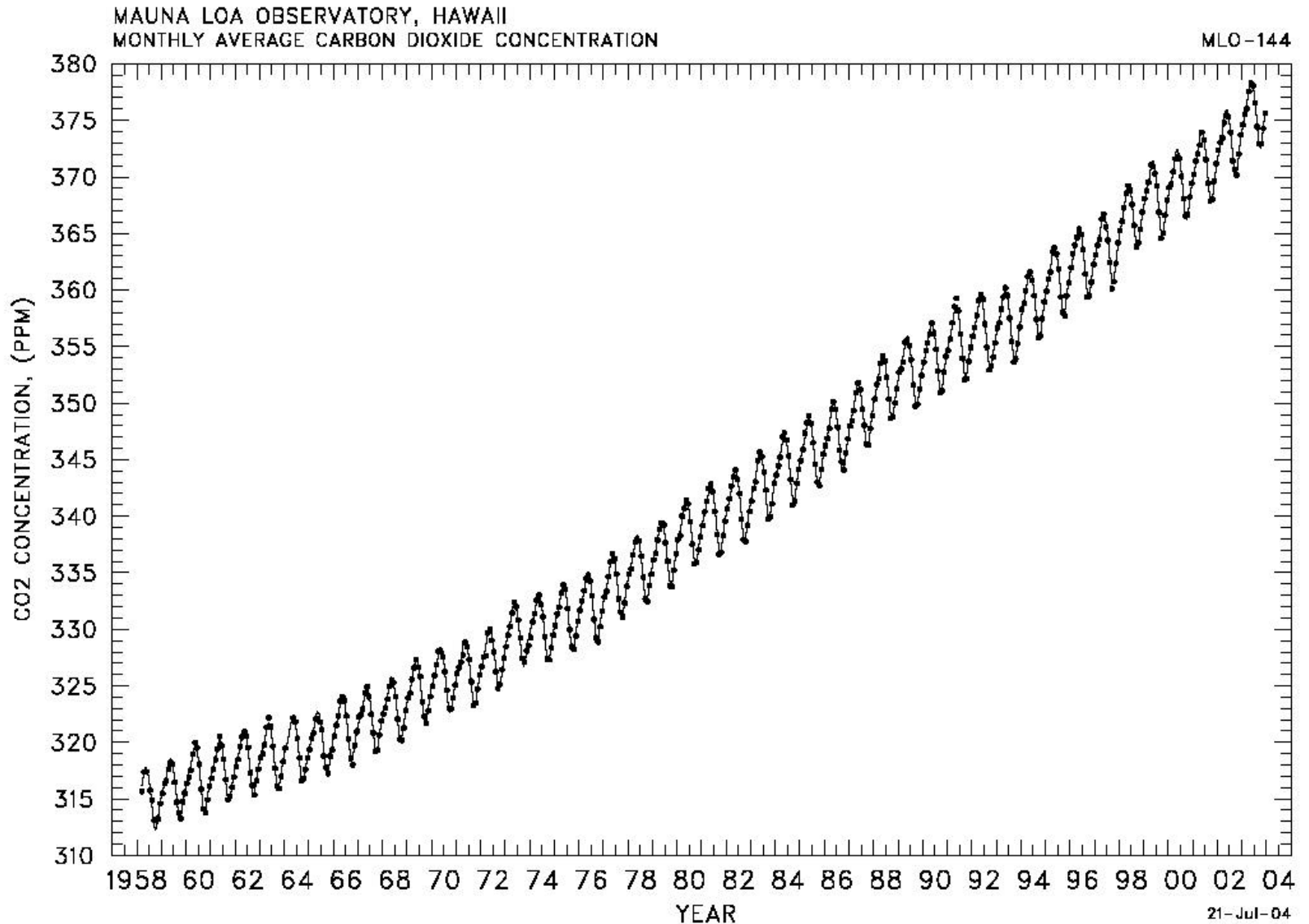




# CarboEurope integrated approach to deliver multiple constraints on the C balance



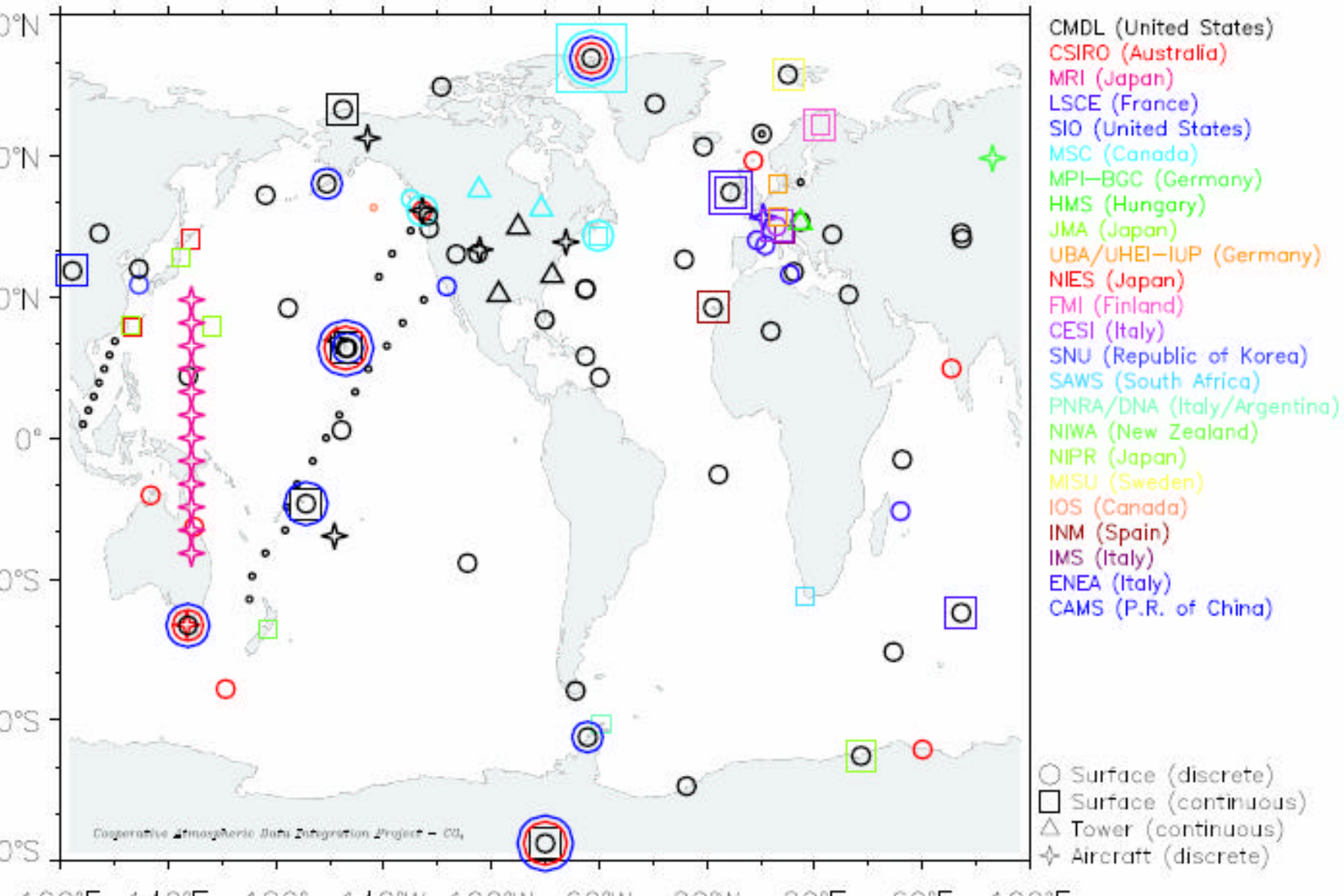
# Global network for atmospheric CO<sub>2</sub>





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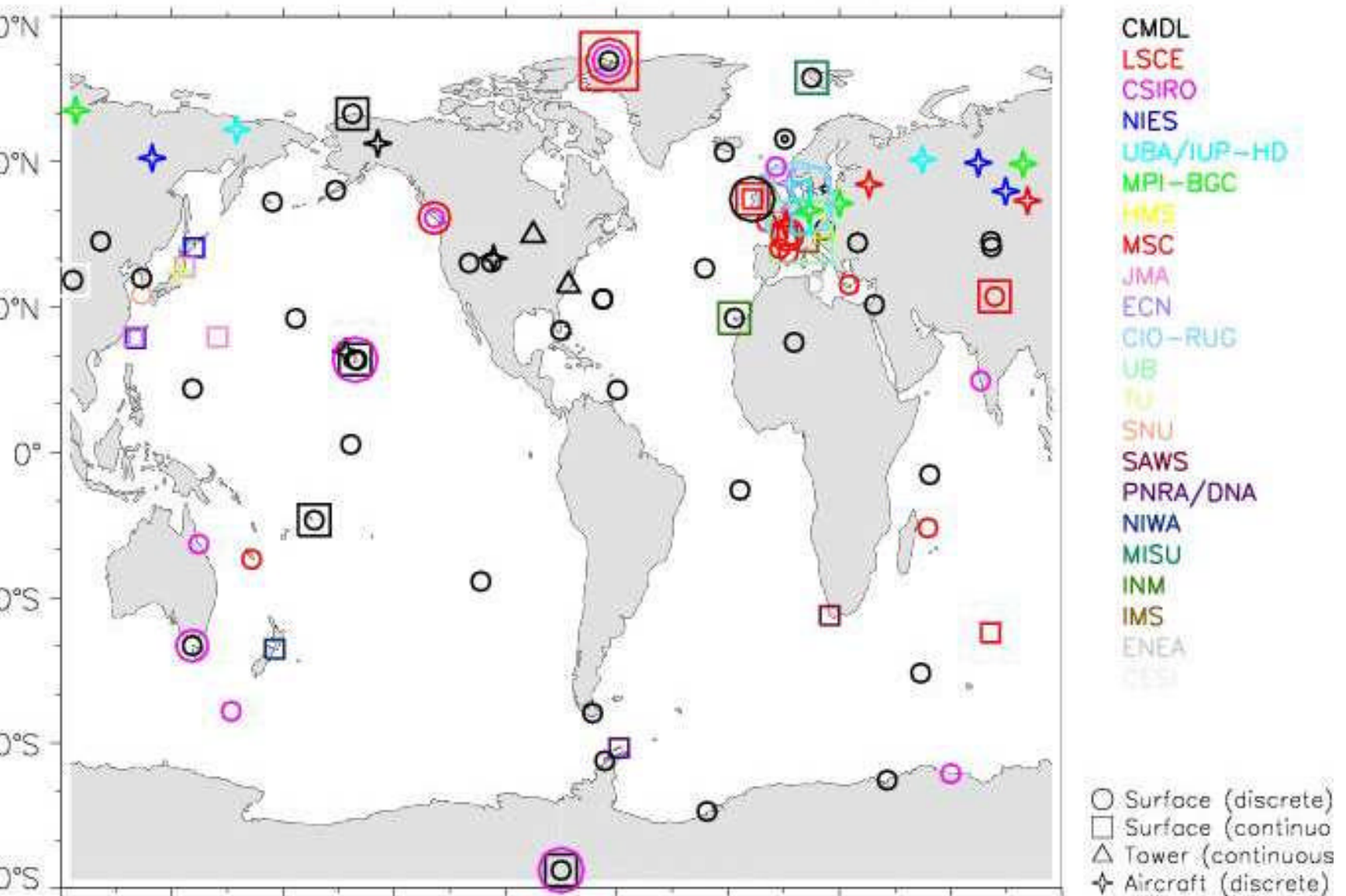
GLOBALVIEW-CO<sub>2</sub>, 2005



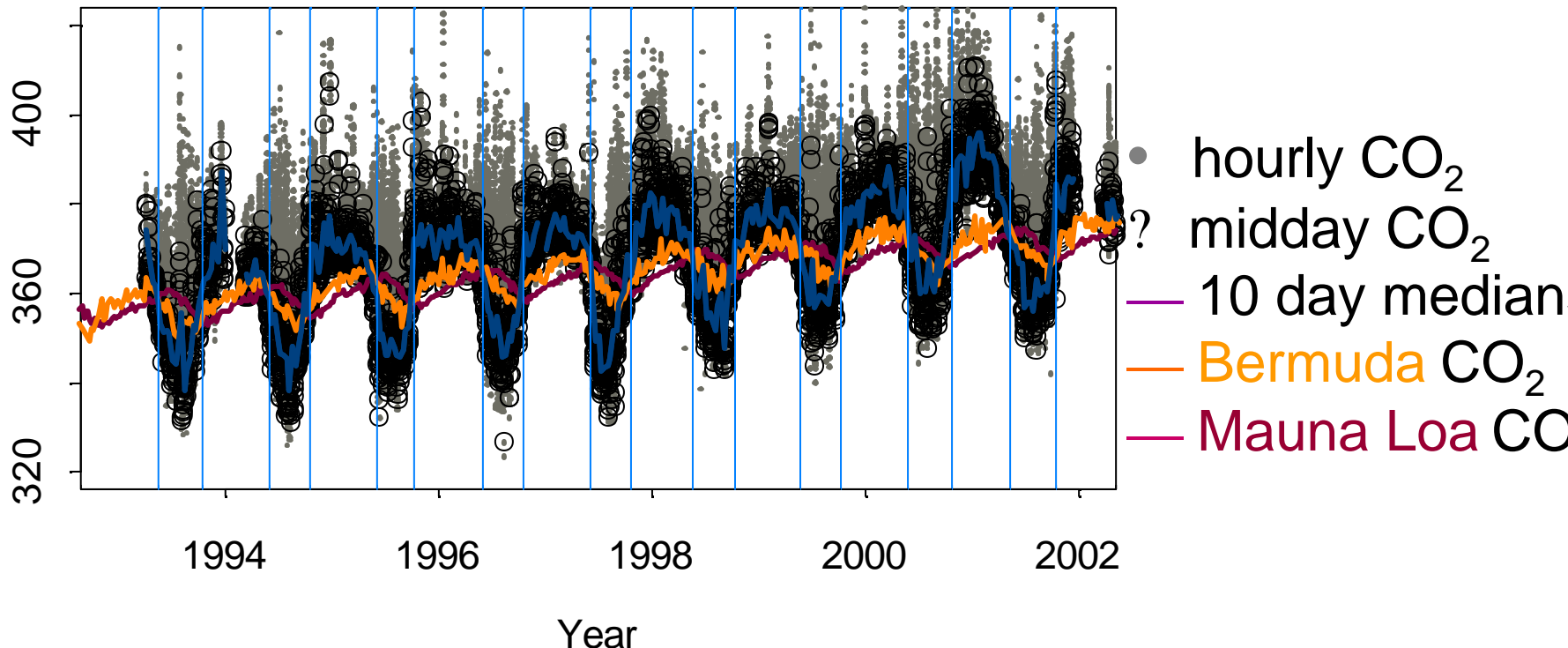
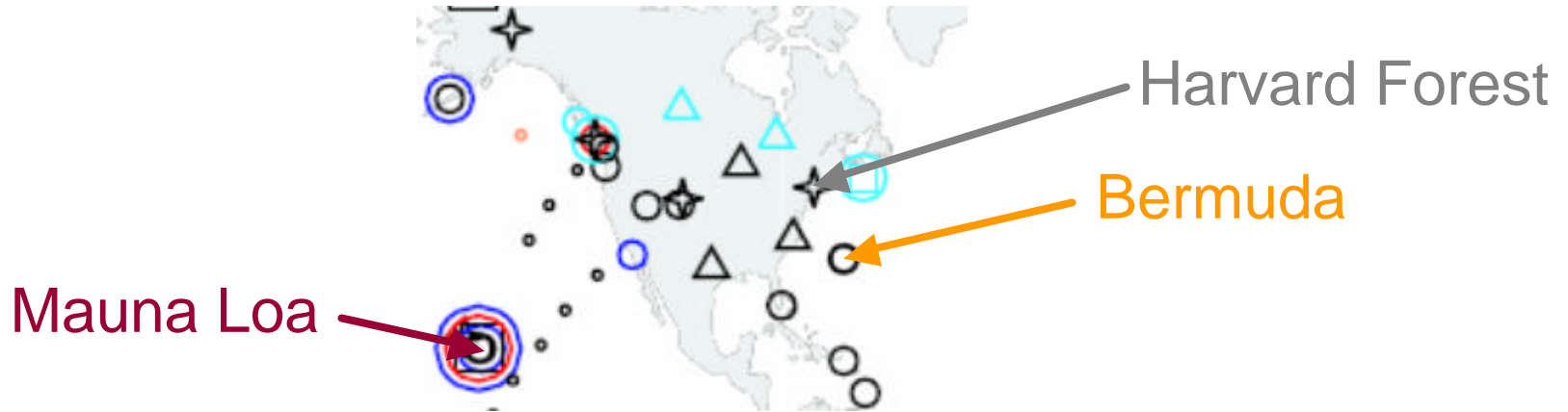


# Global network for atmospheric CO<sub>2</sub>

## GLOBAL NETWORK – Future

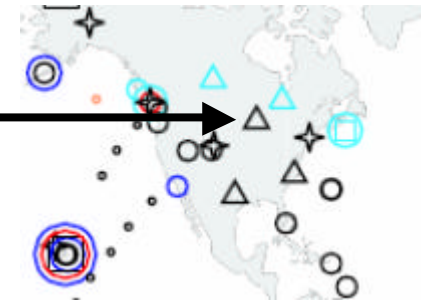
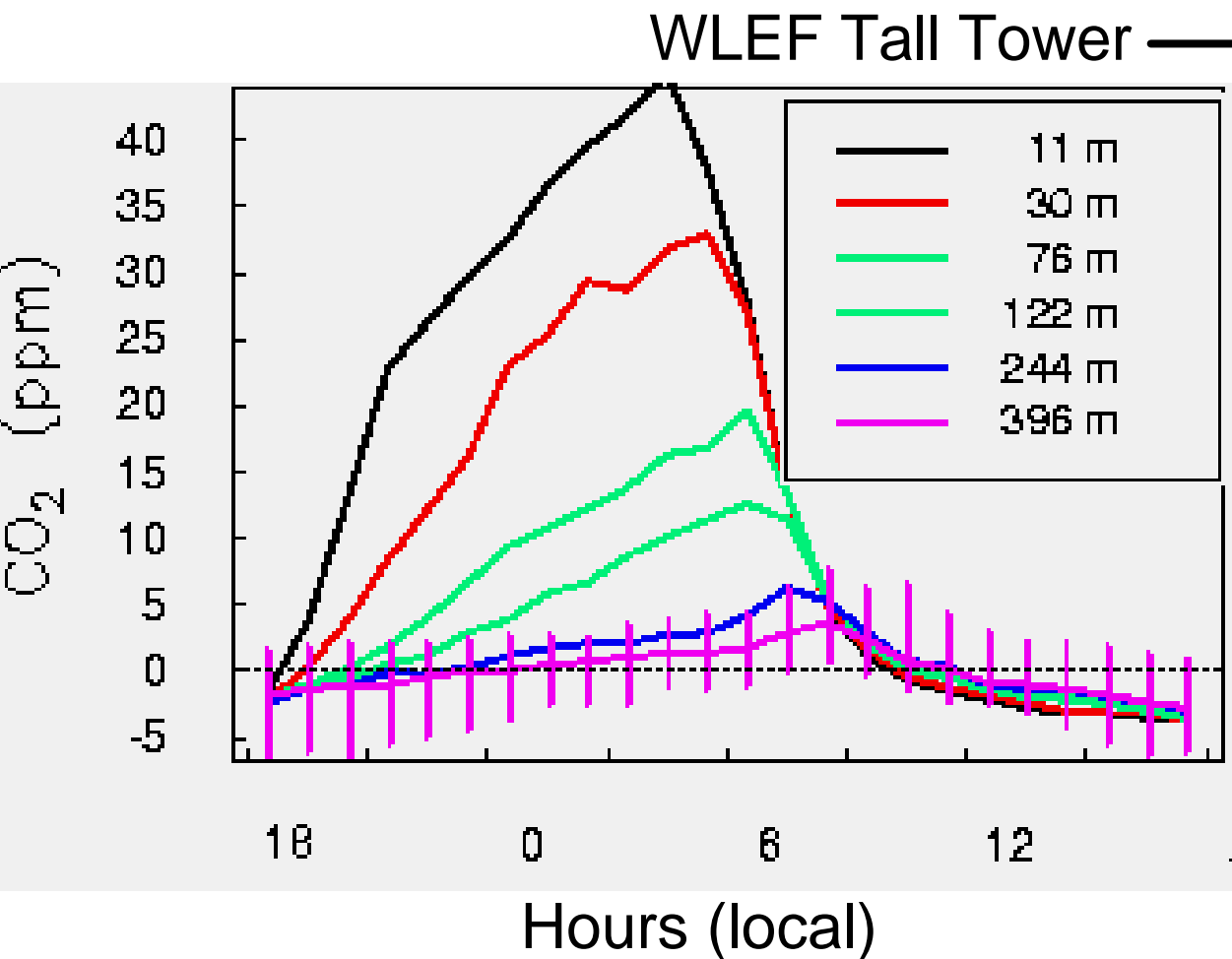


# Continental boundary layer: Harvard Forest Environmental Monitoring Site



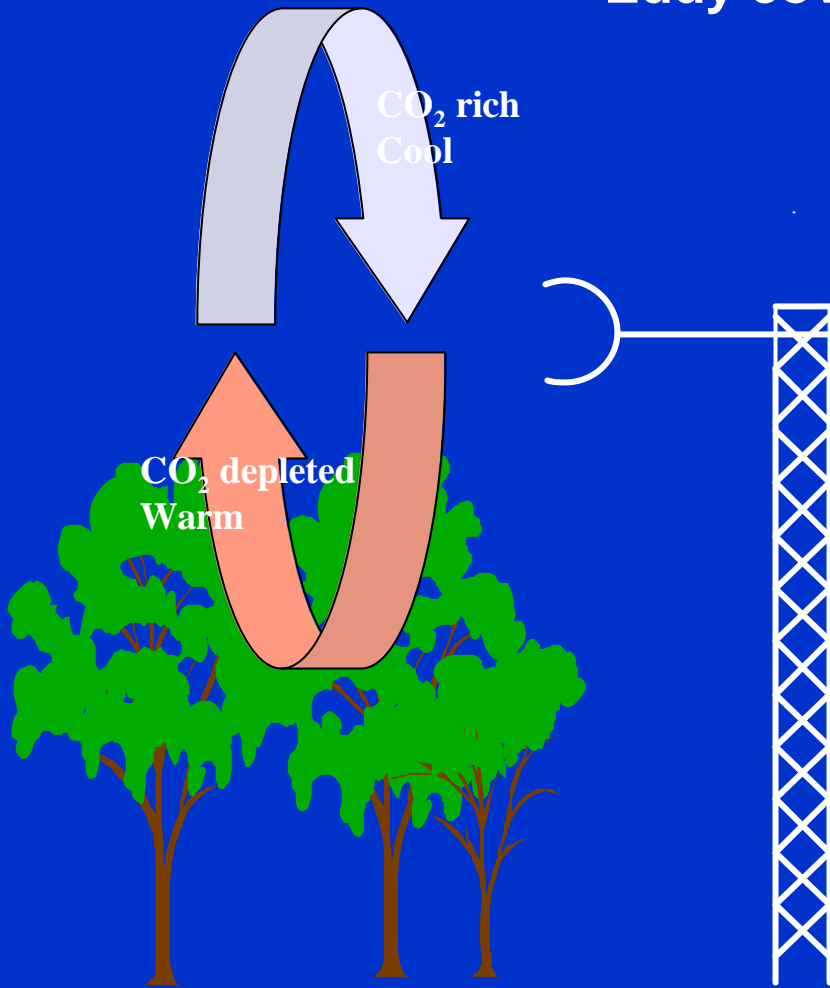
# Continental boundary layer:

diurnal cycle for different heights at a tall tower

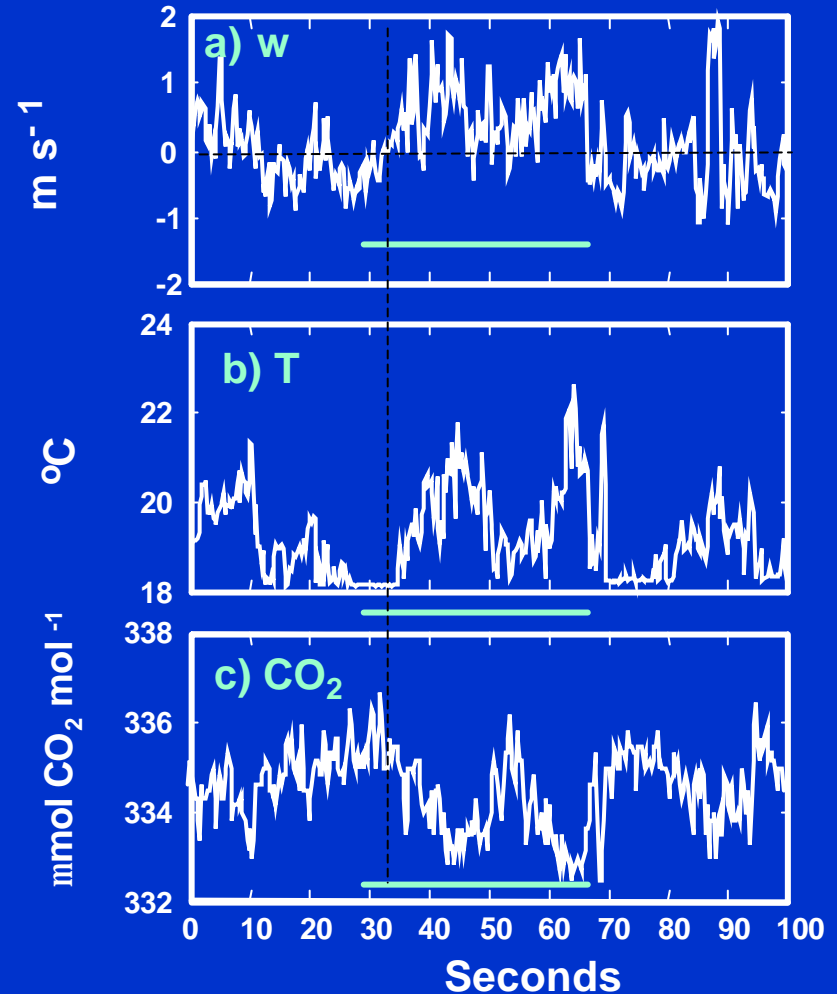


July average  
diurnal cycle  
(top level  
24h mean  
subtracted)

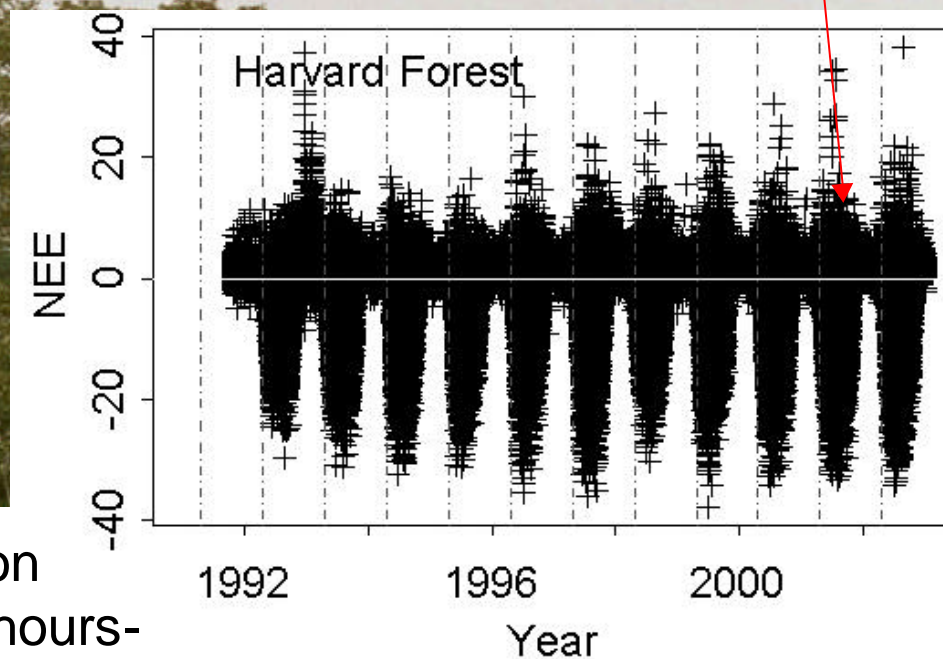
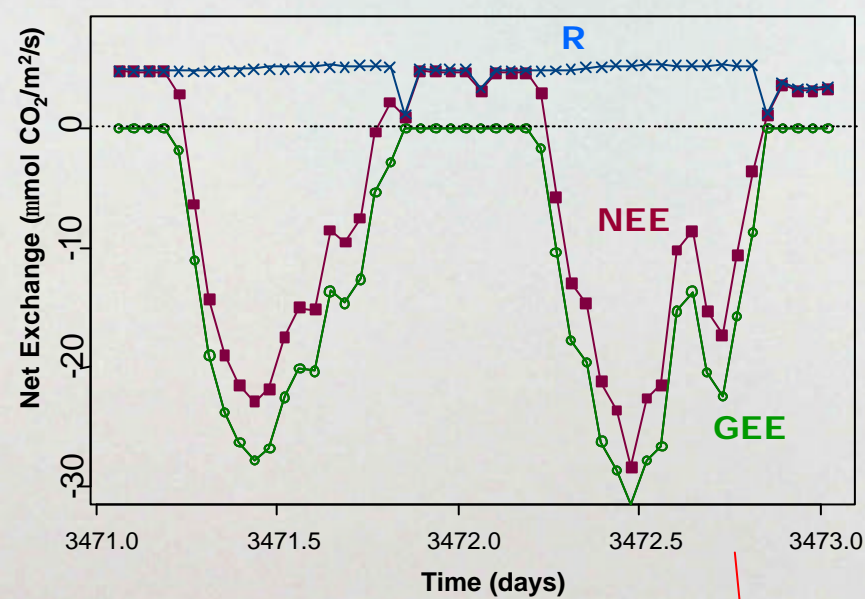
# Eddy covariance



$$\text{Flux}_{\text{CO}_2} = w' \text{CO}_2'$$

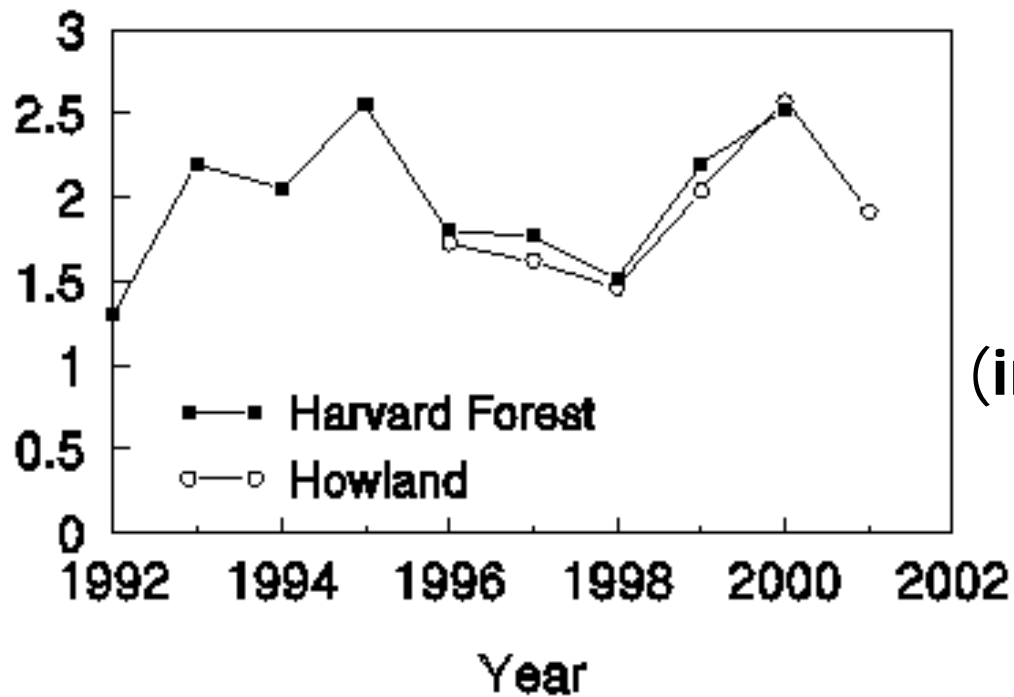






Flux towers give detailed information on atmosphere-biosphere exchange, for hours-to-decade, including annual sums, for ~100 ha

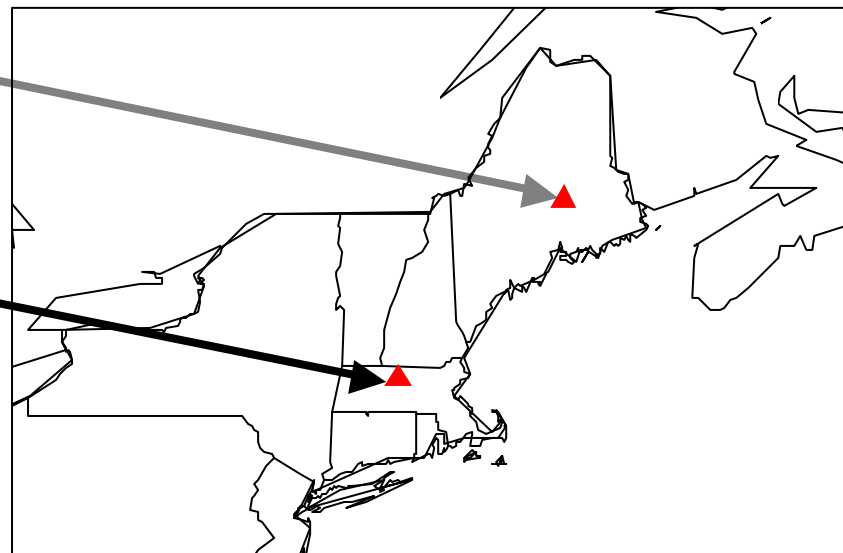
Ecosystem C uptake ( $\text{t ha}^{-1} \text{y}^{-1}$ )

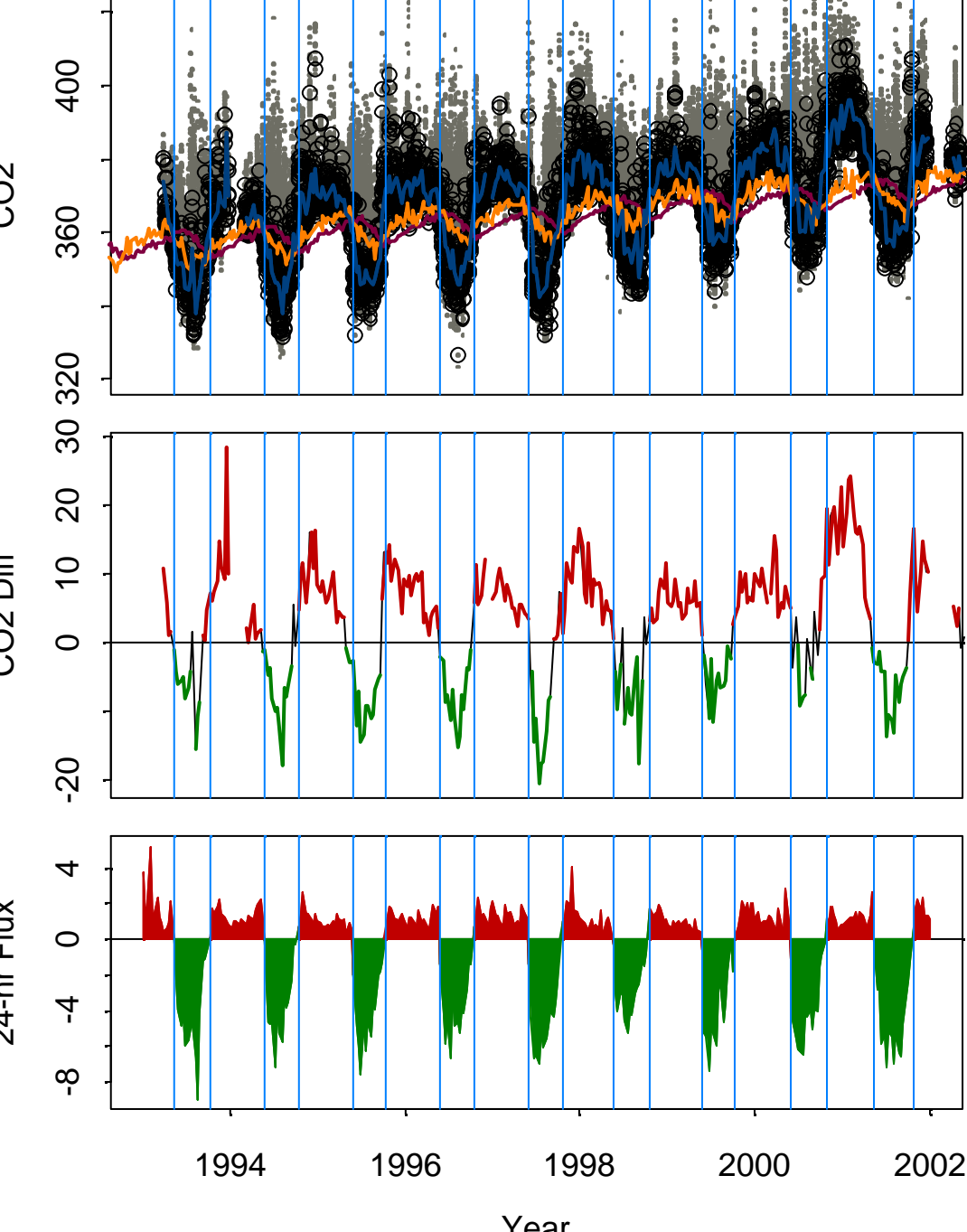


2 AmeriFlux sites  
400 km distance  
Similar IAVs  
(interannual variations)

Howland Forest

Harvard Forest



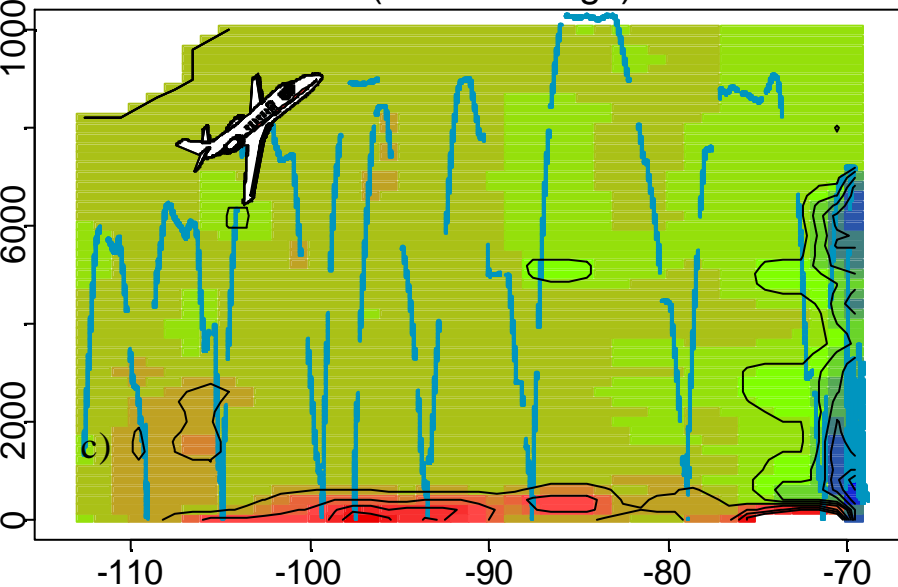


- hourly CO<sub>2</sub>, HF 30 m
- ? midday CO<sub>2</sub>
- 10 day medians
- Bermuda CO<sub>2</sub>
- Mauna Loa CO<sub>2</sub>

Difference  
 Harvard Forest CO<sub>2</sub>  
 – Bermuda CO<sub>2</sub>

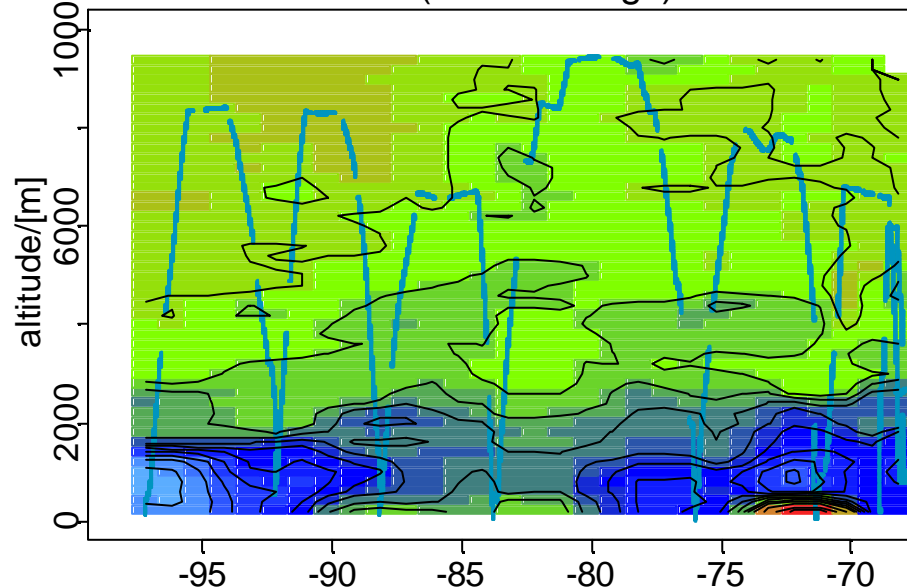
10-day median of the  
 daily mean CO<sub>2</sub> flux at  
 Harvard Forest ( $\mu\text{mole m}^{-2}\text{s}^{-1}$ )

Maine => Maine (southern legs) 8/6-11/2000



longitude/[deg]  
370 CO<sub>2</sub> [ppm]

Maine => NoDak (northern legs) 8/16&19/2000



altitude/[m]  
370 CO<sub>2</sub> [ppm]  
longitude/[deg]

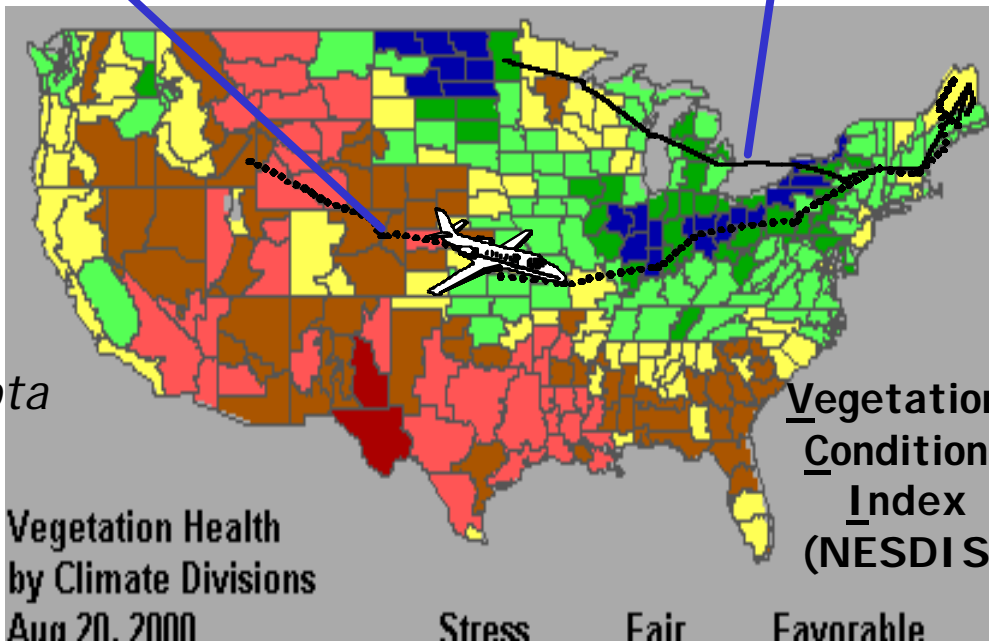
# CO<sub>2</sub> Budget and Rectification Airborne Study "COBRA 2000"

Funding: NOAA, NASA, NSF, and DoE



Harvard  
U. North Dakota  
Scripps  
NOAA CMDL  
Penn State  
CSU

Gerbig et al. 2003a]



Vegetation Health by Climate Divisions Aug 20, 2000

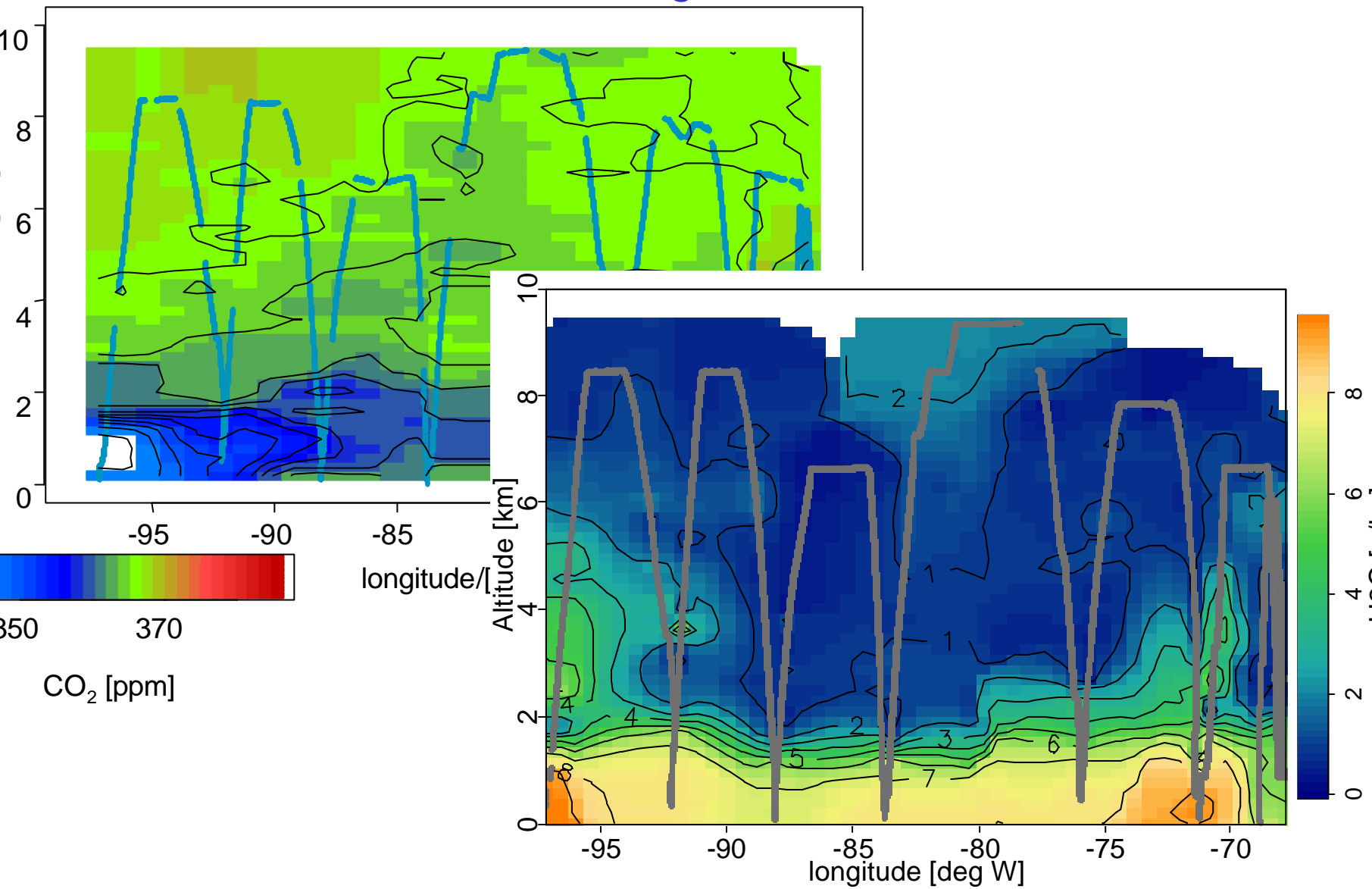
Vegetation Condition Index (NESDIS)

Stress Fair Favorable

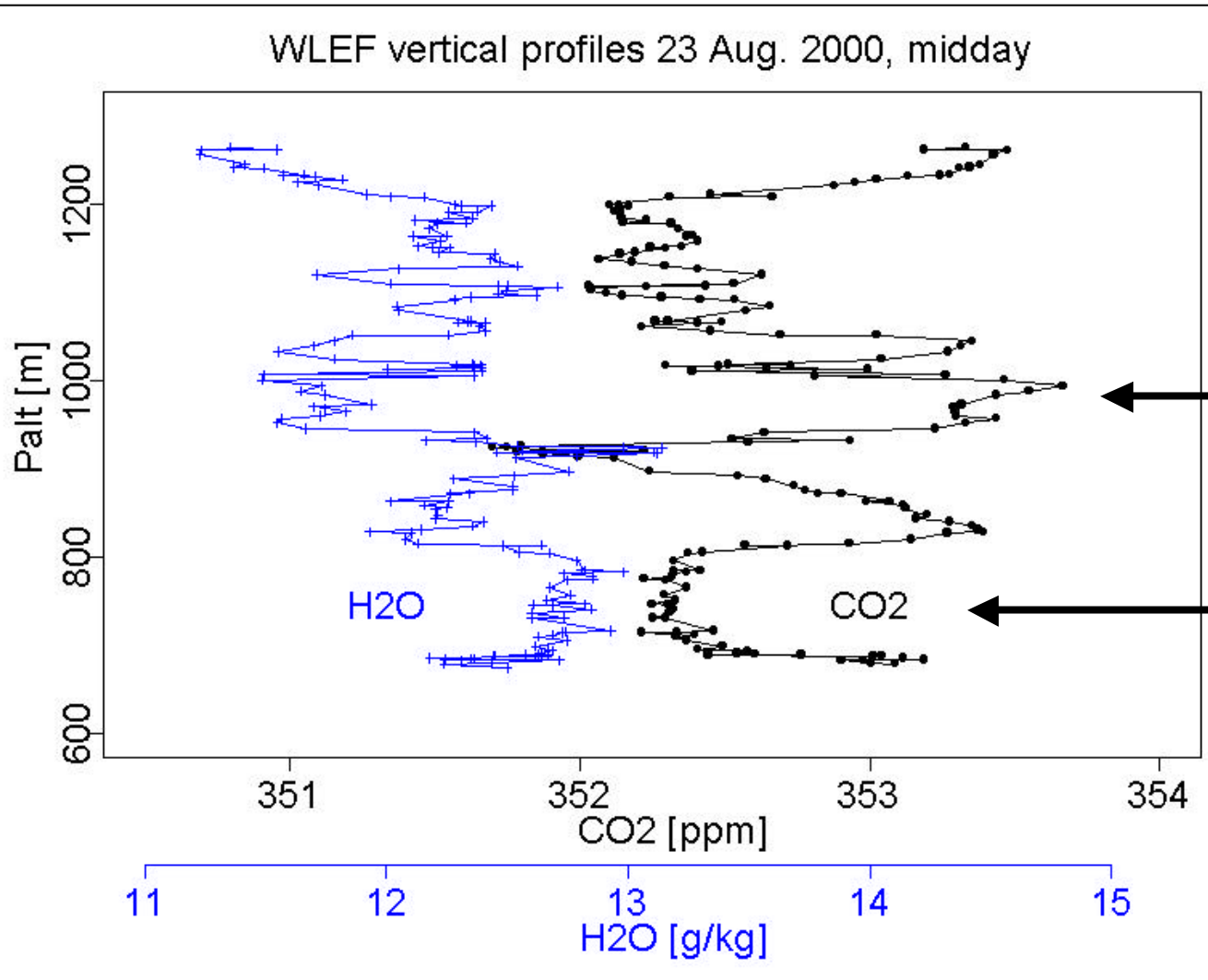


# COBRA-2000 Northern Transect

18-19 August 2000



# What models don't need to resolve



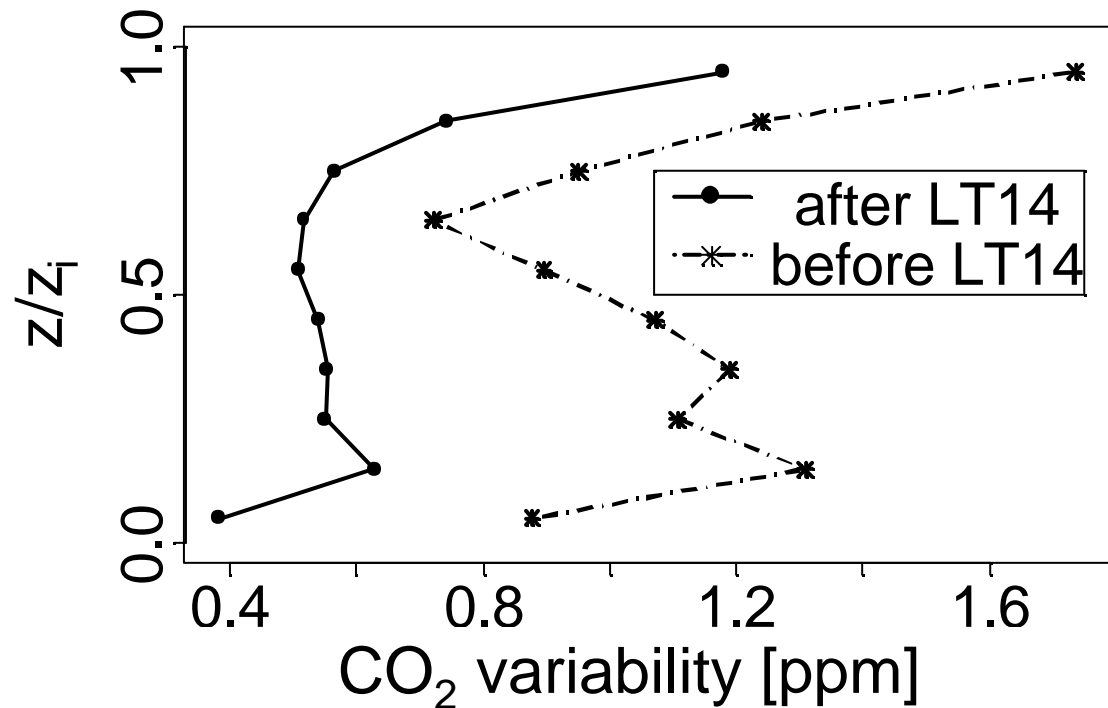
Dry layer, high  
CO<sub>2</sub>, entrained  
from above CBL

Moist layer, low  
CO<sub>2</sub>, from  
surface

“unresolvable” eddies  
⇒ “resolvable” mixed layer mean

# What models don't need to resolve

~ 100 profiles  
(COBRA,  
summer 2000)



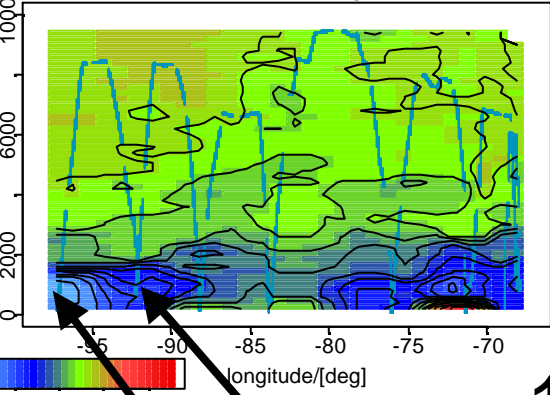
Uncertainty of mixed layer mean  
due to eddies:

**0.2 ppm**

(~ e.g. 25 layers, 1 ppm stdev. each)

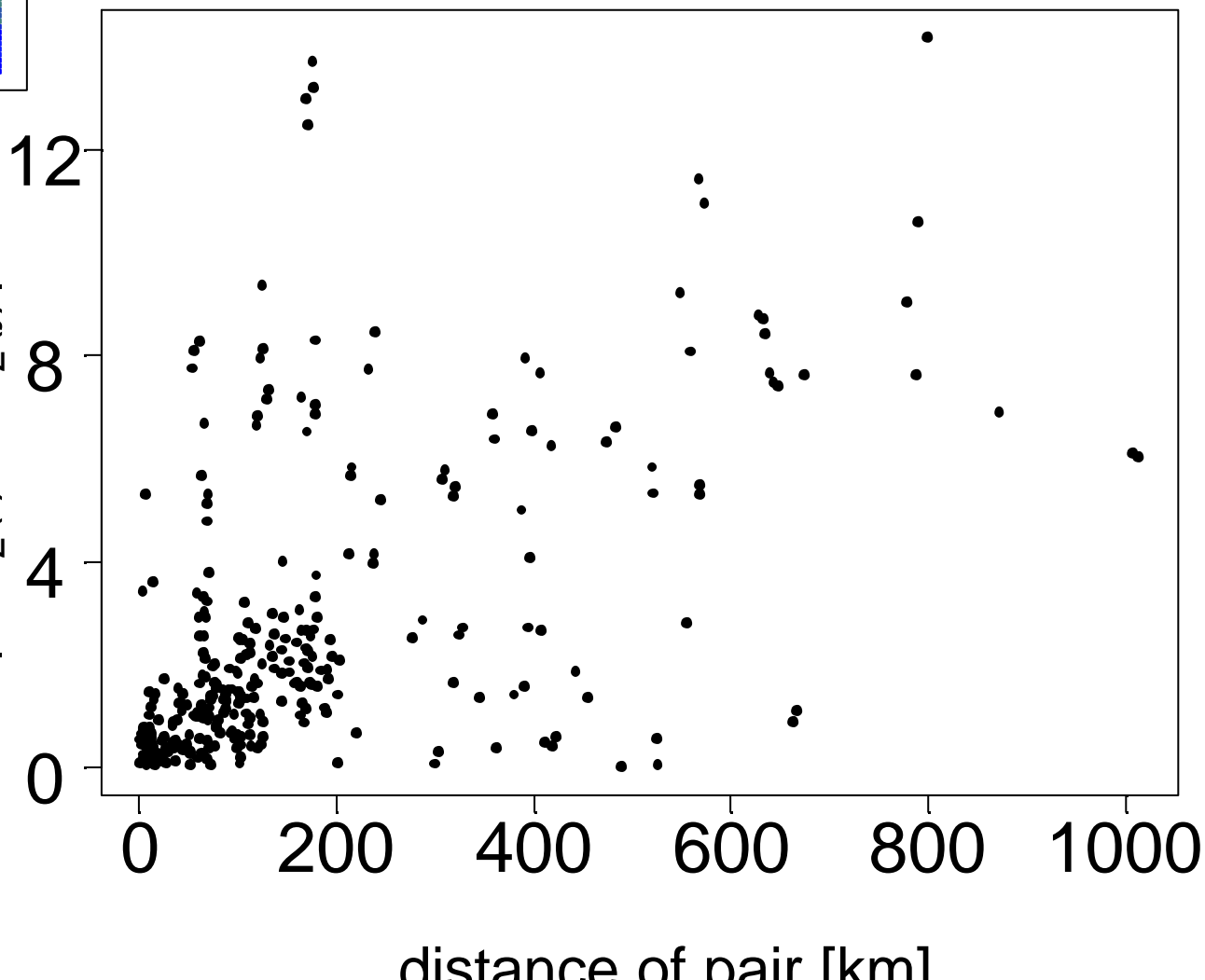
# Grain size of atmospheric CO<sub>2</sub> or "how well-mixed is the atmosphere?"

Maine => NoDak (northern legs) 8/18&19/2000



Spatial  
differences of  
pairs of mixed  
layer profiles  
measured  
within 3 hours  
of each other

$$|\text{CO}_2(i) - \text{CO}_2(j)|$$





# Grain size of atmospheric CO<sub>2</sub>: Variogram

Variogram for a given "distance bin" ( $h$ =average distance):

$$2\mathbf{g}(h) = \text{var}\left(\text{CO}_2(s_i) - \text{CO}_2(s_j)\right) \text{ with } h = \overline{|s_i - s_j|}$$

classical Variogram:

$$2\hat{\mathbf{g}}(h) = \frac{1}{N(h)} \sum_{N(h)} \left(\text{CO}_2(s_i) - \text{CO}_2(s_j)\right)^2$$

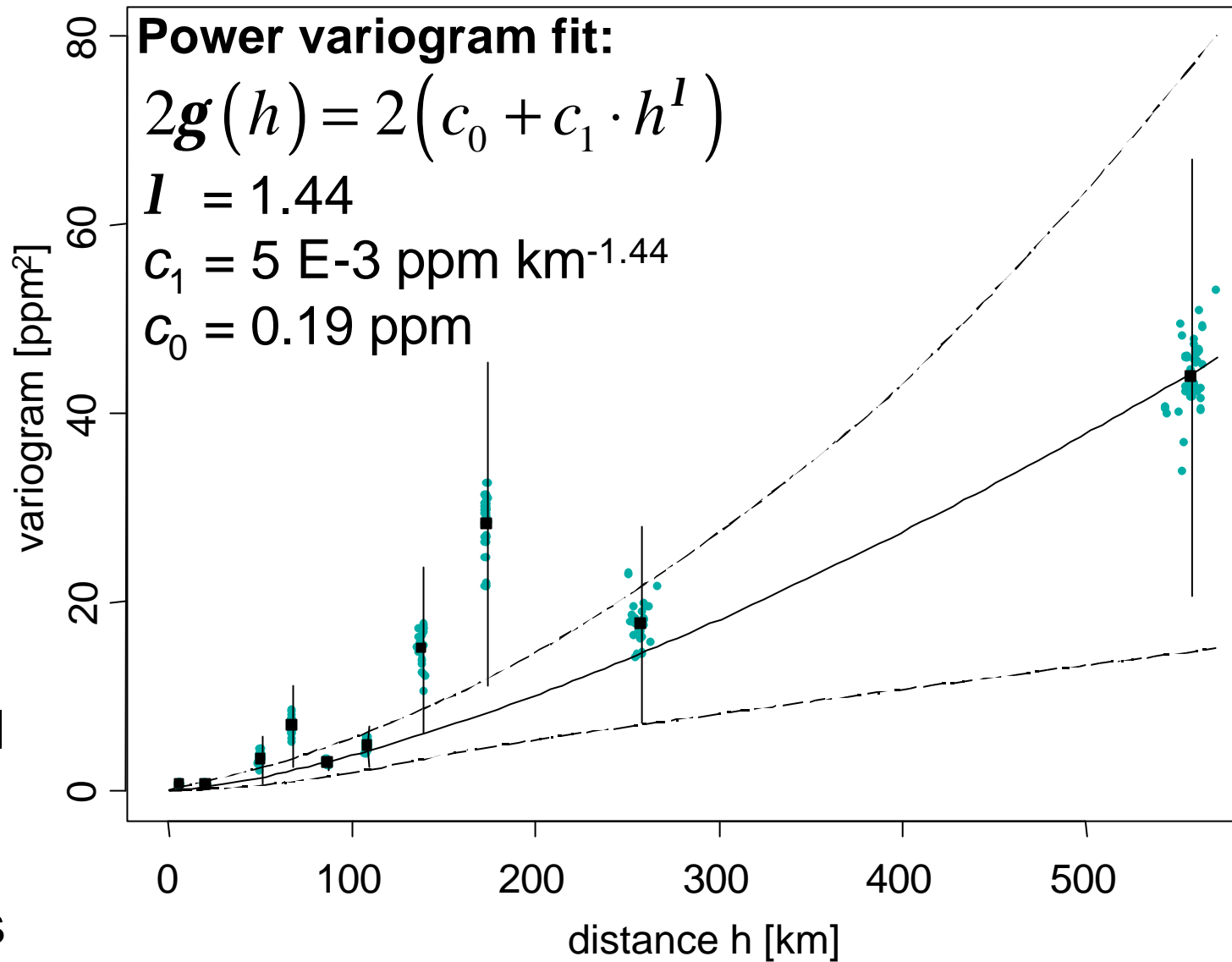
with  $N(h)$ : Number of pairs

robust Variogram:

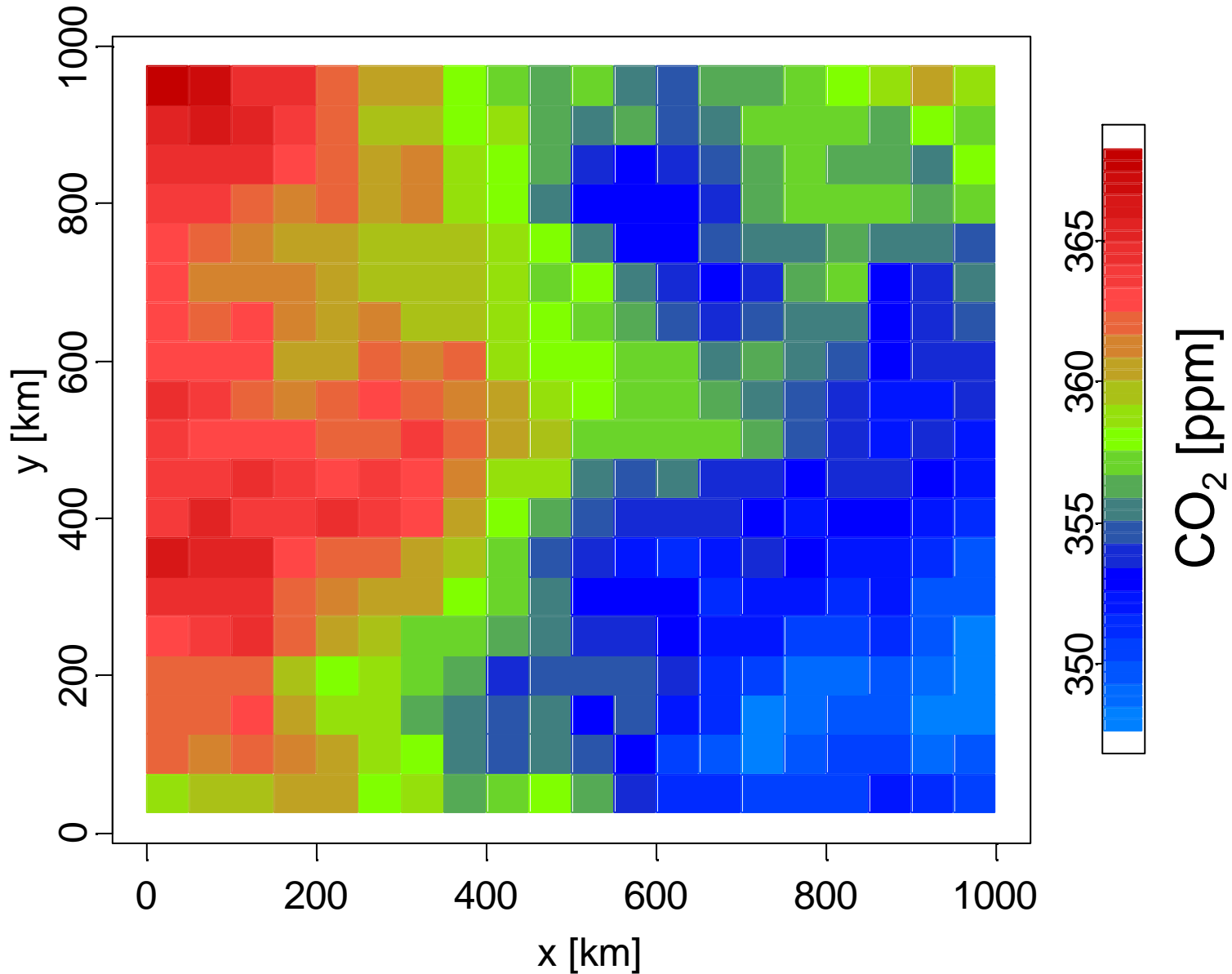
$$2\bar{\mathbf{g}}(h) = \frac{\left\{ \frac{1}{N(h)} \sum_{N(h)} \left| \text{CO}_2(s_i) - \text{CO}_2(s_j) \right|^{1/2} \right\}^4}{0.457 + 0.494 / N(h)}$$

# Fitting of power variogram model

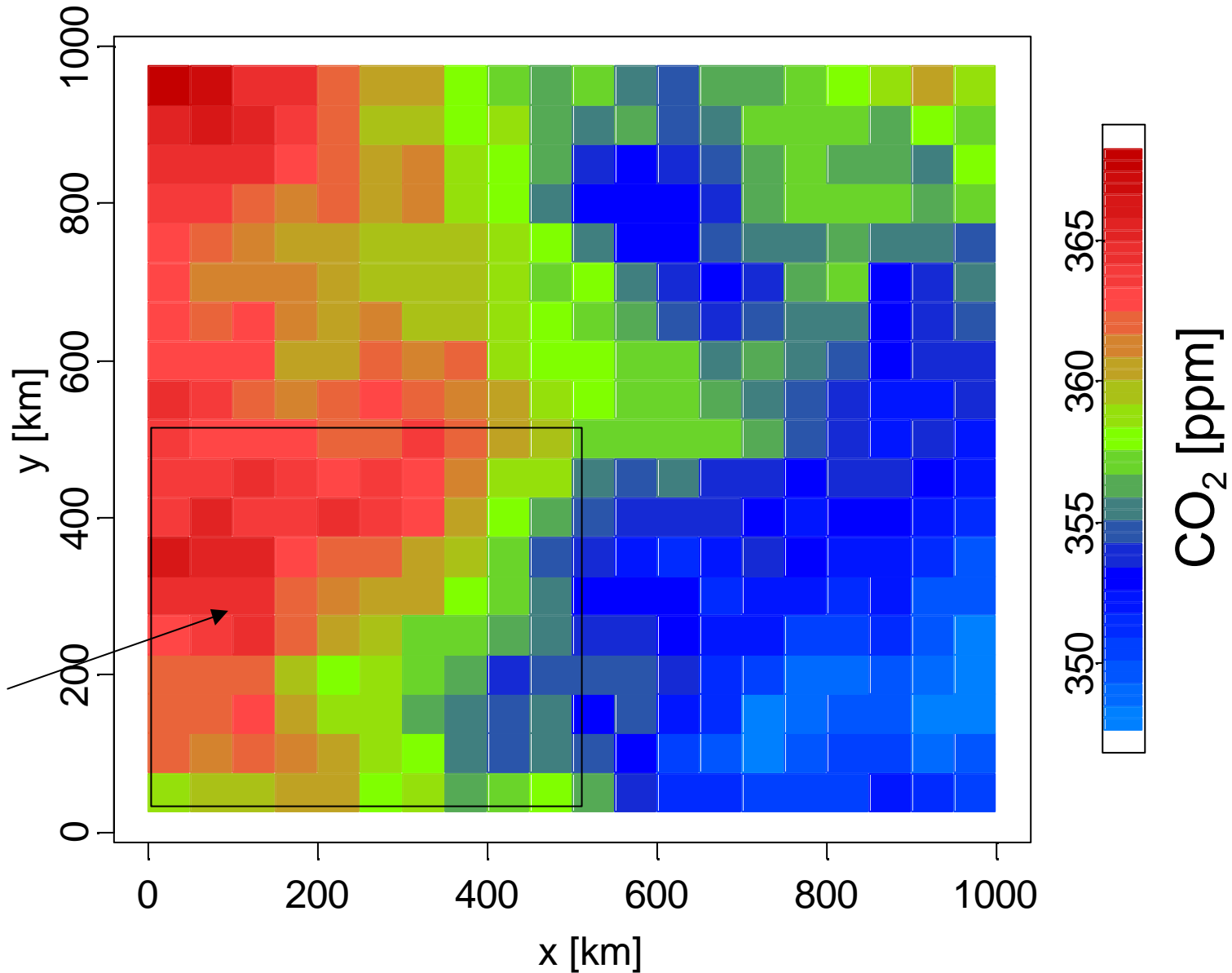
Variogram:  
variance of  
differences of  
pairs of mixed  
layer profiles  
measured  
within 3 hours  
of each other



# Spatial simulation for CO<sub>2</sub>



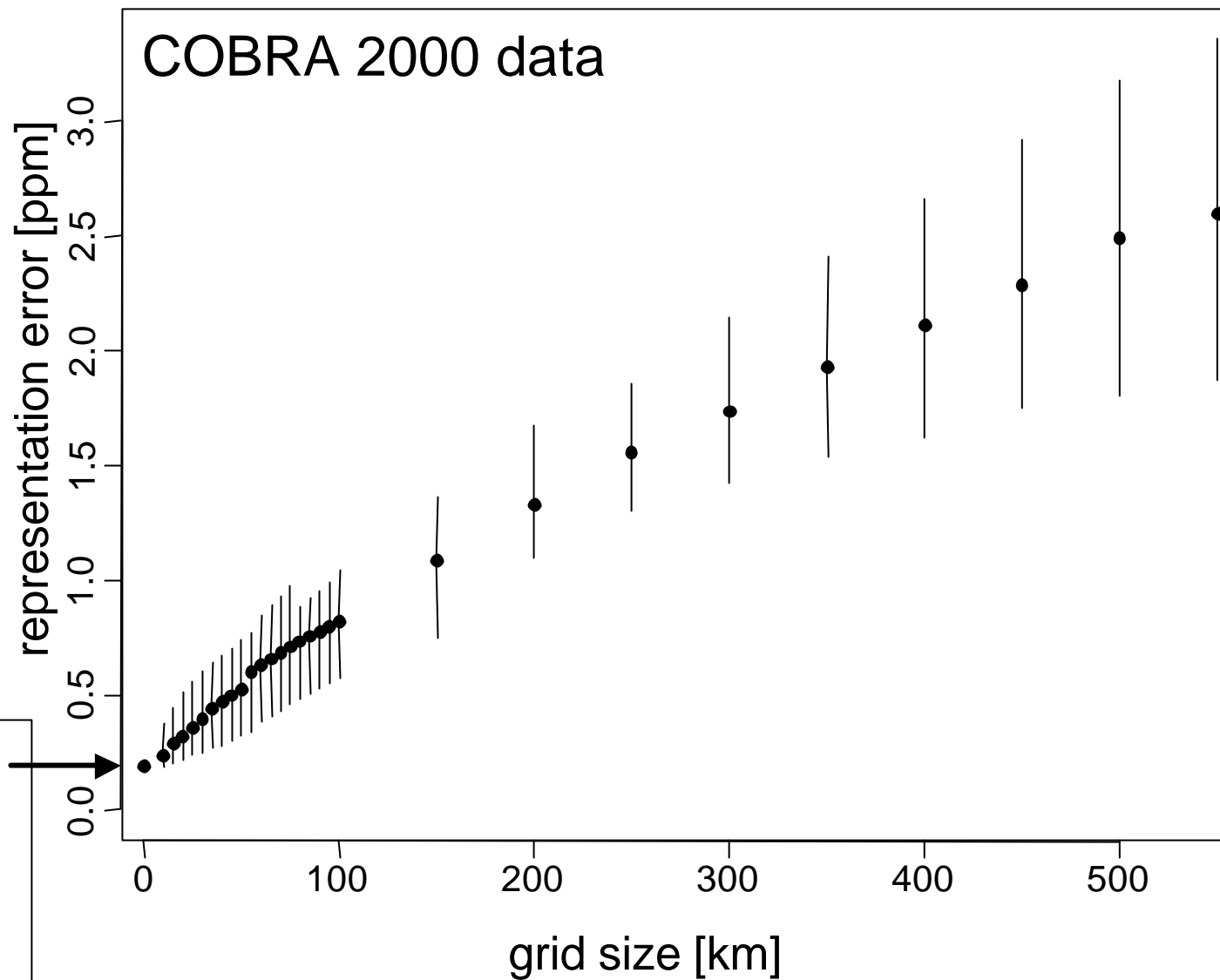
# Spatial simulation for CO<sub>2</sub>



stddev(CO<sub>2</sub>)  
within each  
subgrid of  
size  $\Delta x \Delta y$

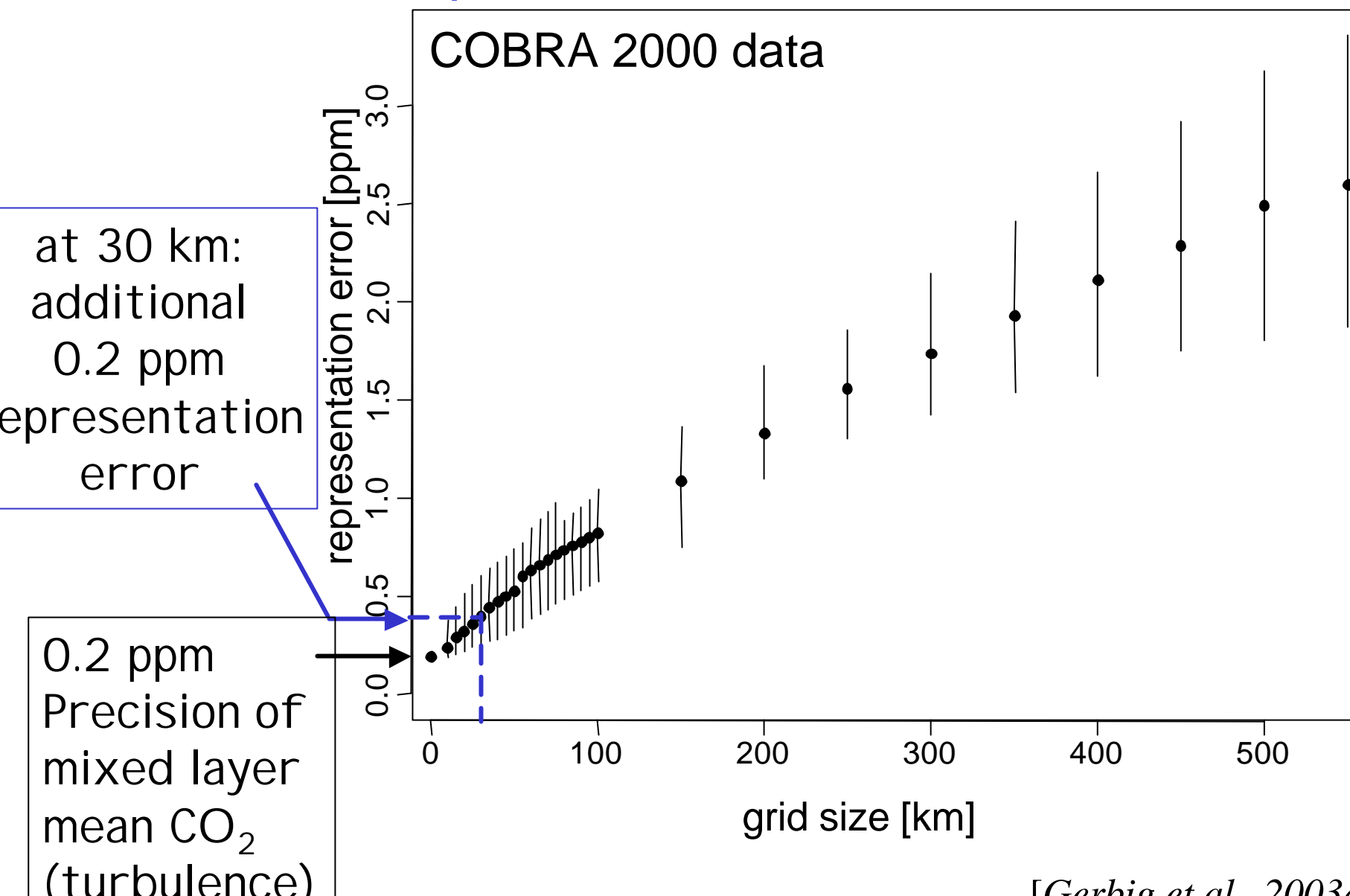


# "Grain size" of atmospheric CO<sub>2</sub>: Representation error

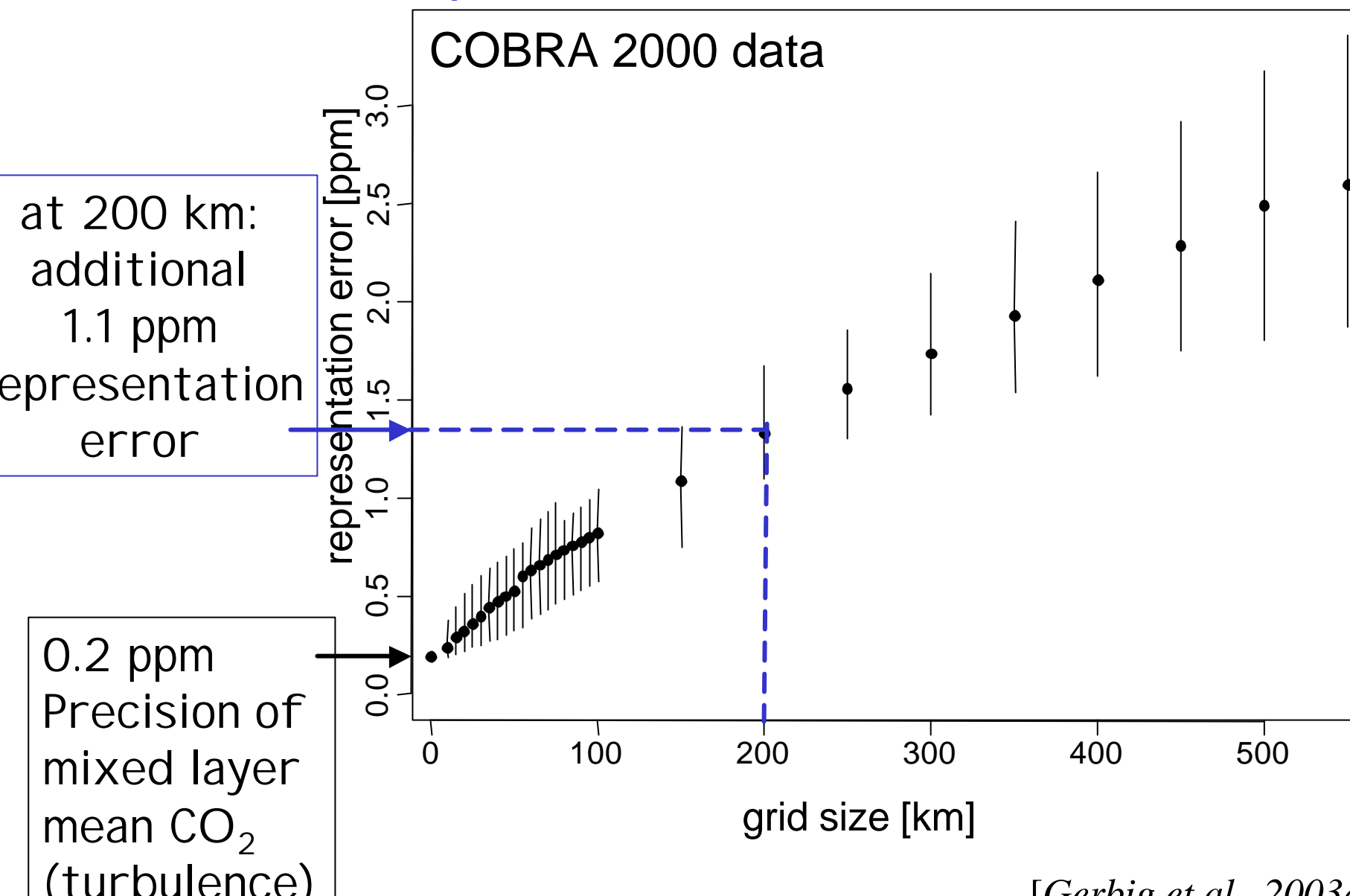


0.2 ppm  
Precision of  
mixed layer  
mean CO<sub>2</sub>  
(turbulence)

# "Grain size" of atmospheric CO<sub>2</sub>: Representation error



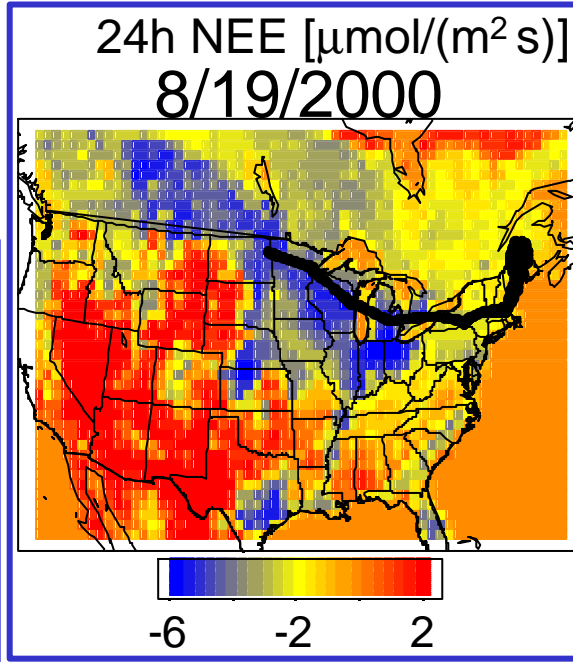
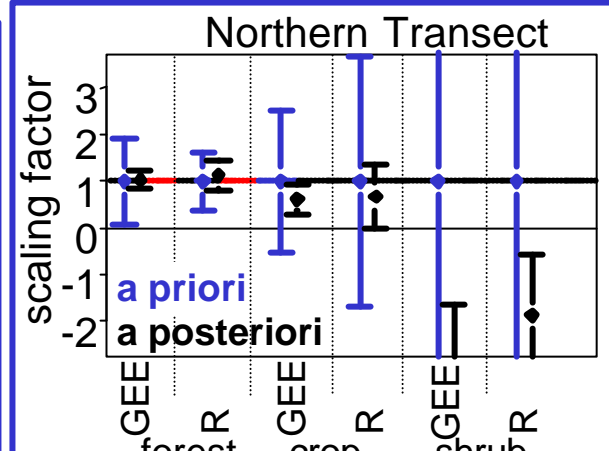
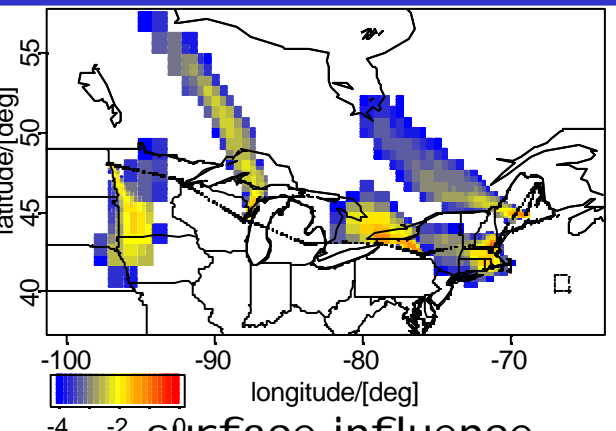
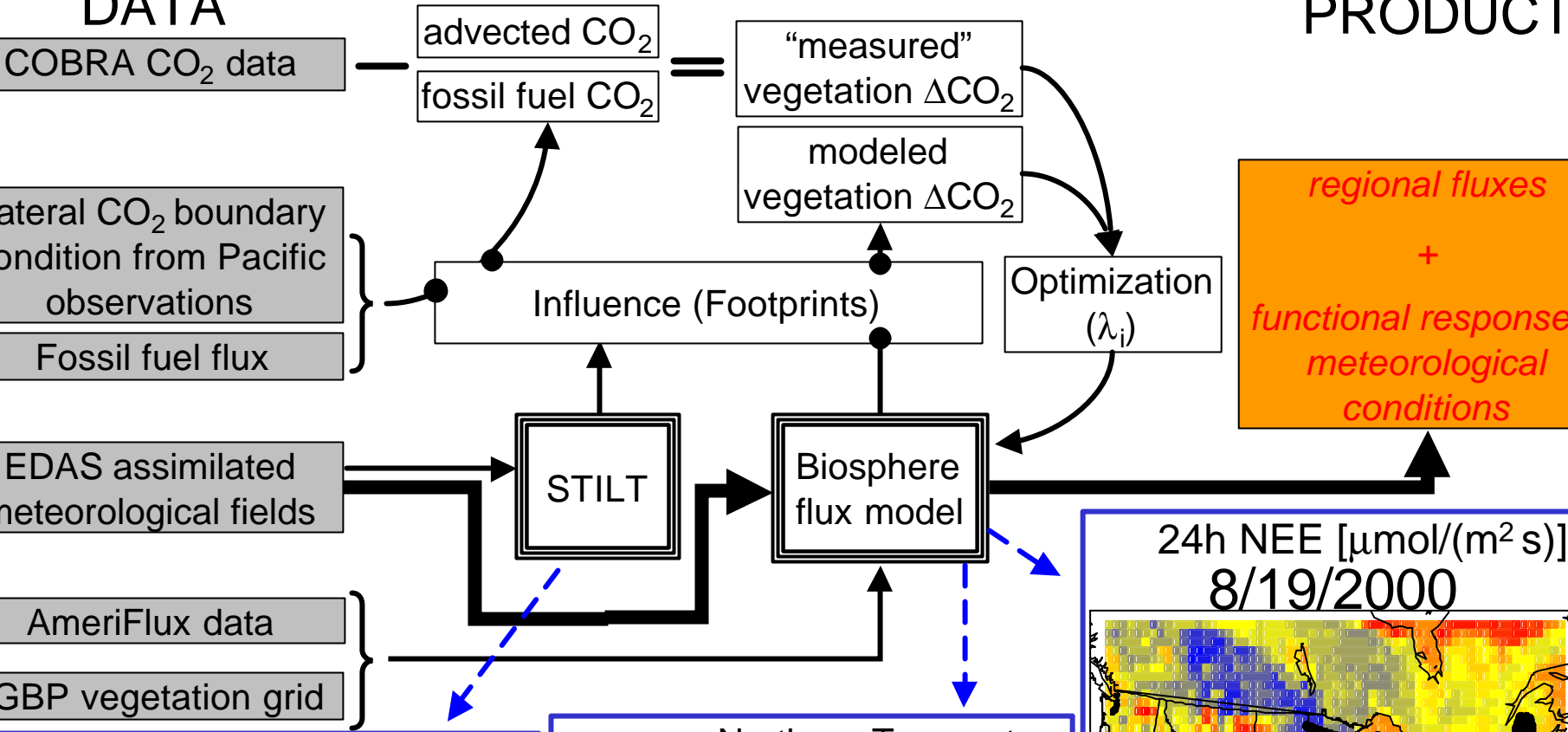
# "Grain size" of atmospheric CO<sub>2</sub>: Representation error



# Receptor Oriented Atmospheric Model "ROAM"

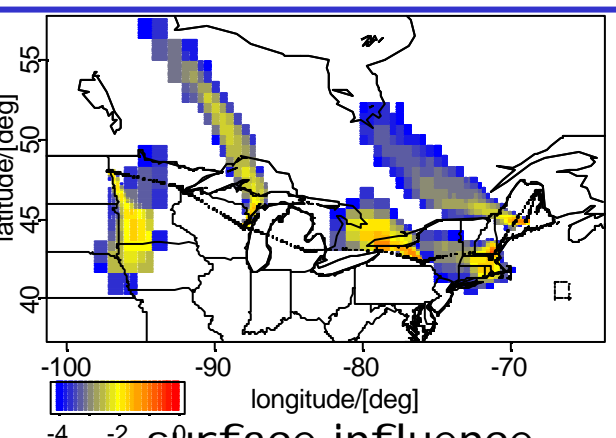
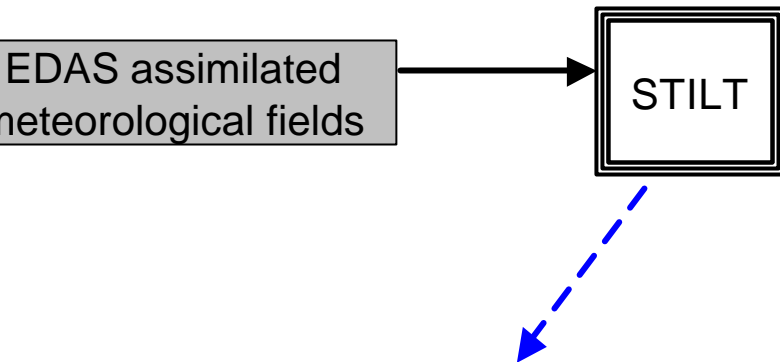
DATA

PRODUCT



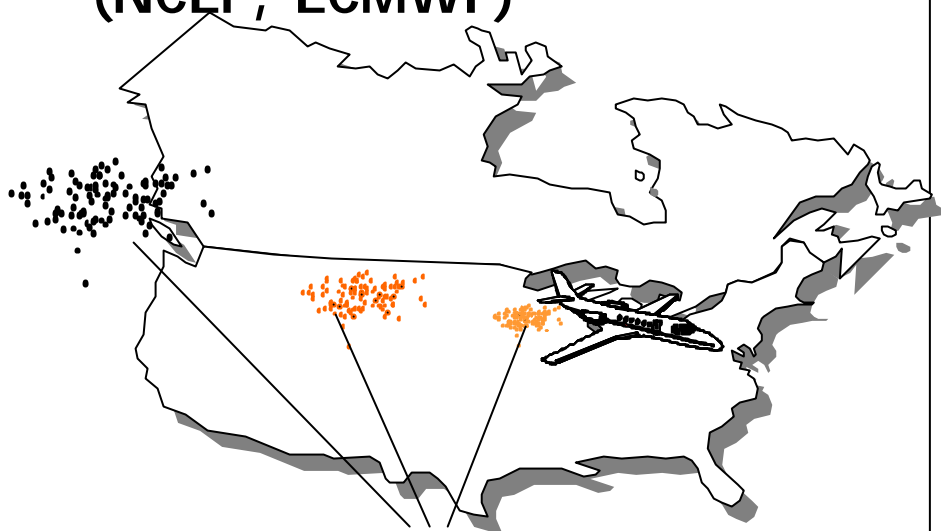


# Receptor Oriented Atmospheric Model "ROAM"



## Stochastic Time Inverted Lagrangian Transport Model

- driven by assimilated or forecasted winds (NCEP, ECMWF)



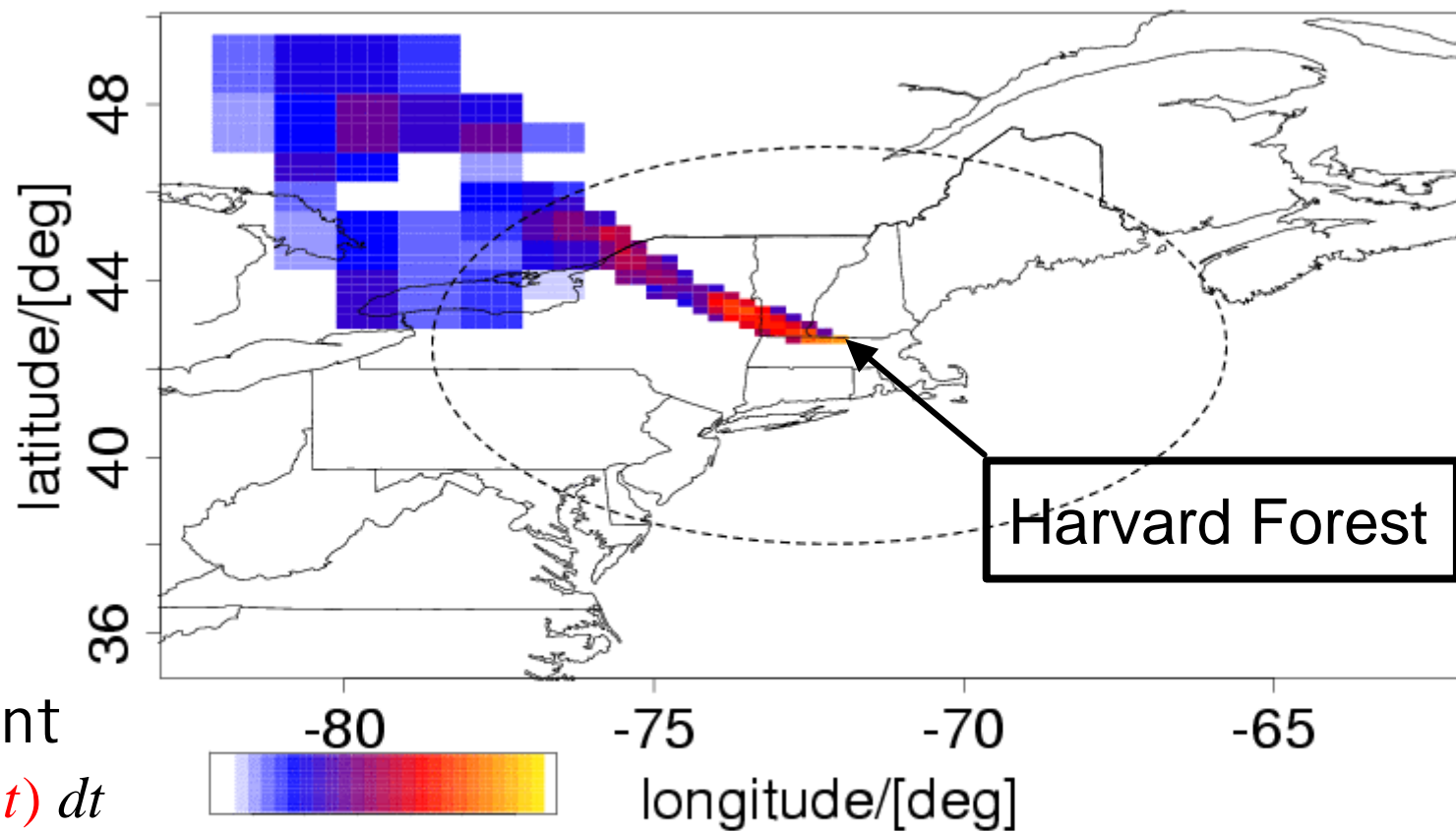
Particle location at different times before arriving at aircraft

[Lin et al., 2003], [Gerbig et al., 2003b]

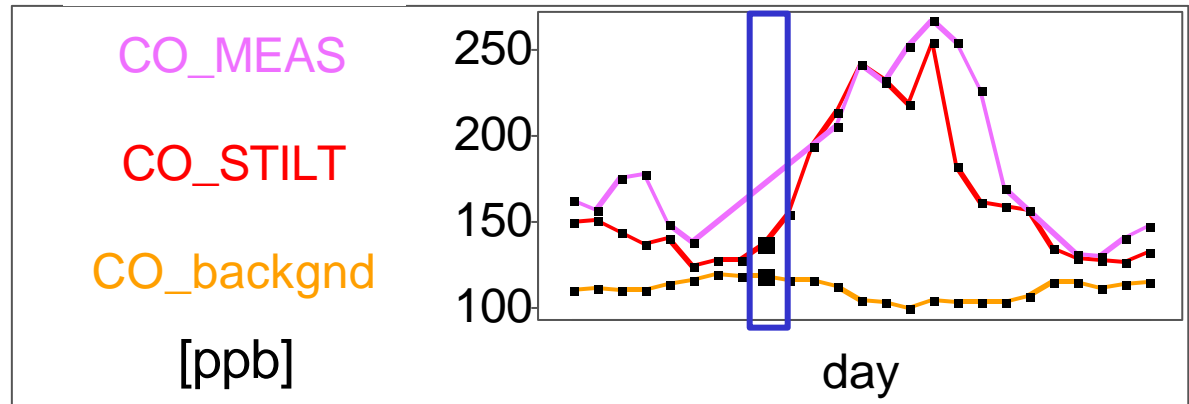
# What does a (tall) tower "see"?

## STILT

Stochastic Time  
Inverted  
Lagrangian  
Transport Model  
[Gerbig et al.,  
2003b]  
[Lin et al., 2003]



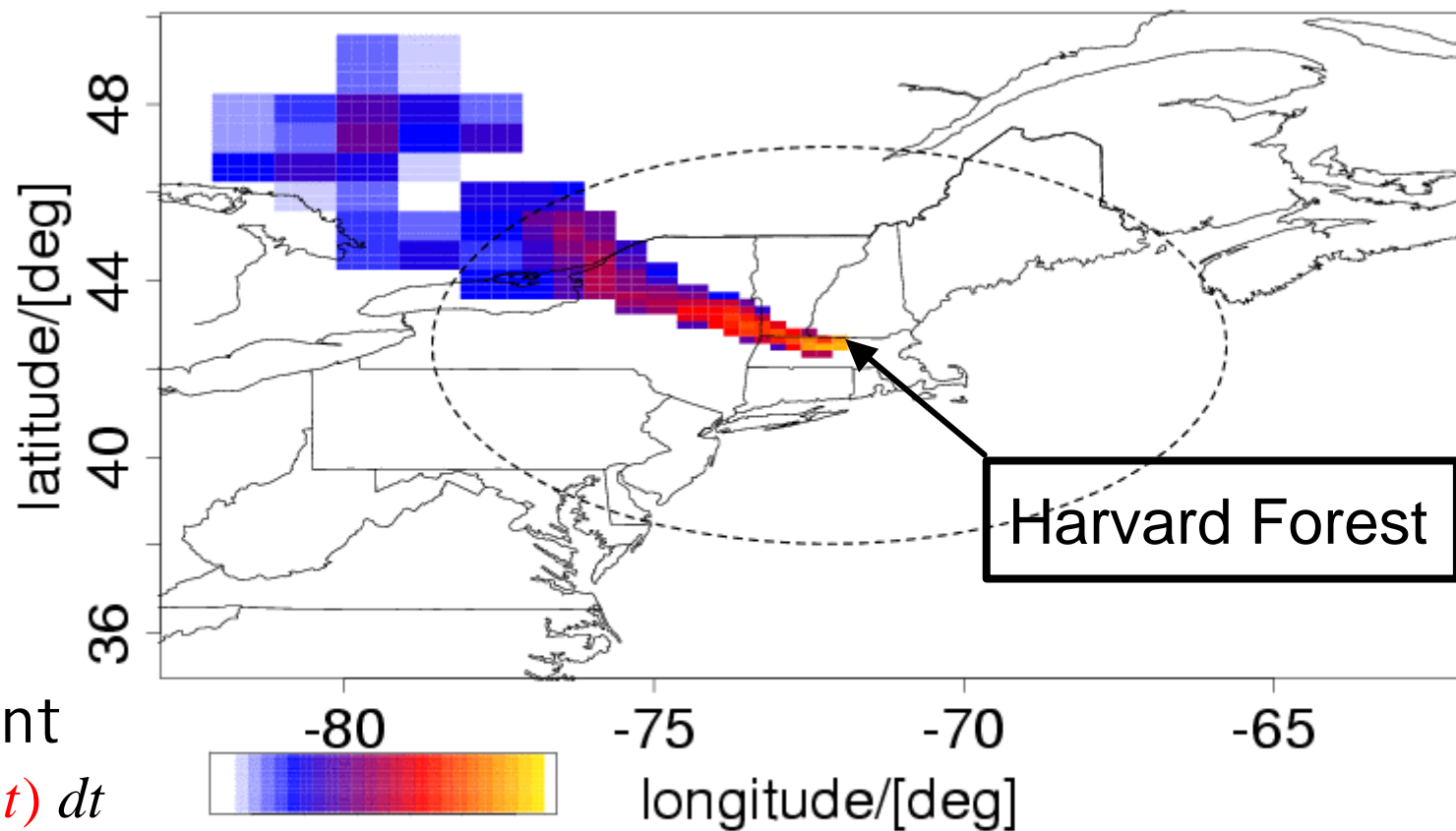
Footprint  
 $\int I(\mathbf{x}_r, t_r | \mathbf{x}, t) dt$



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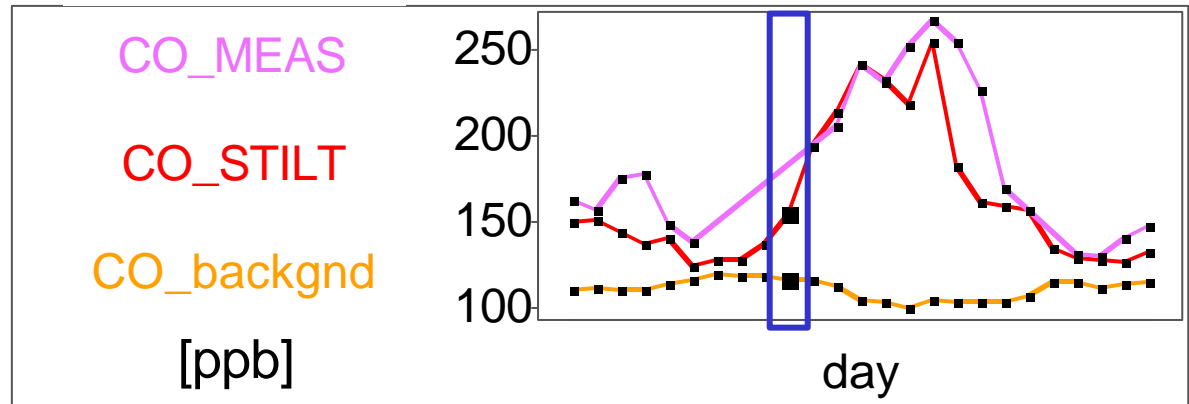
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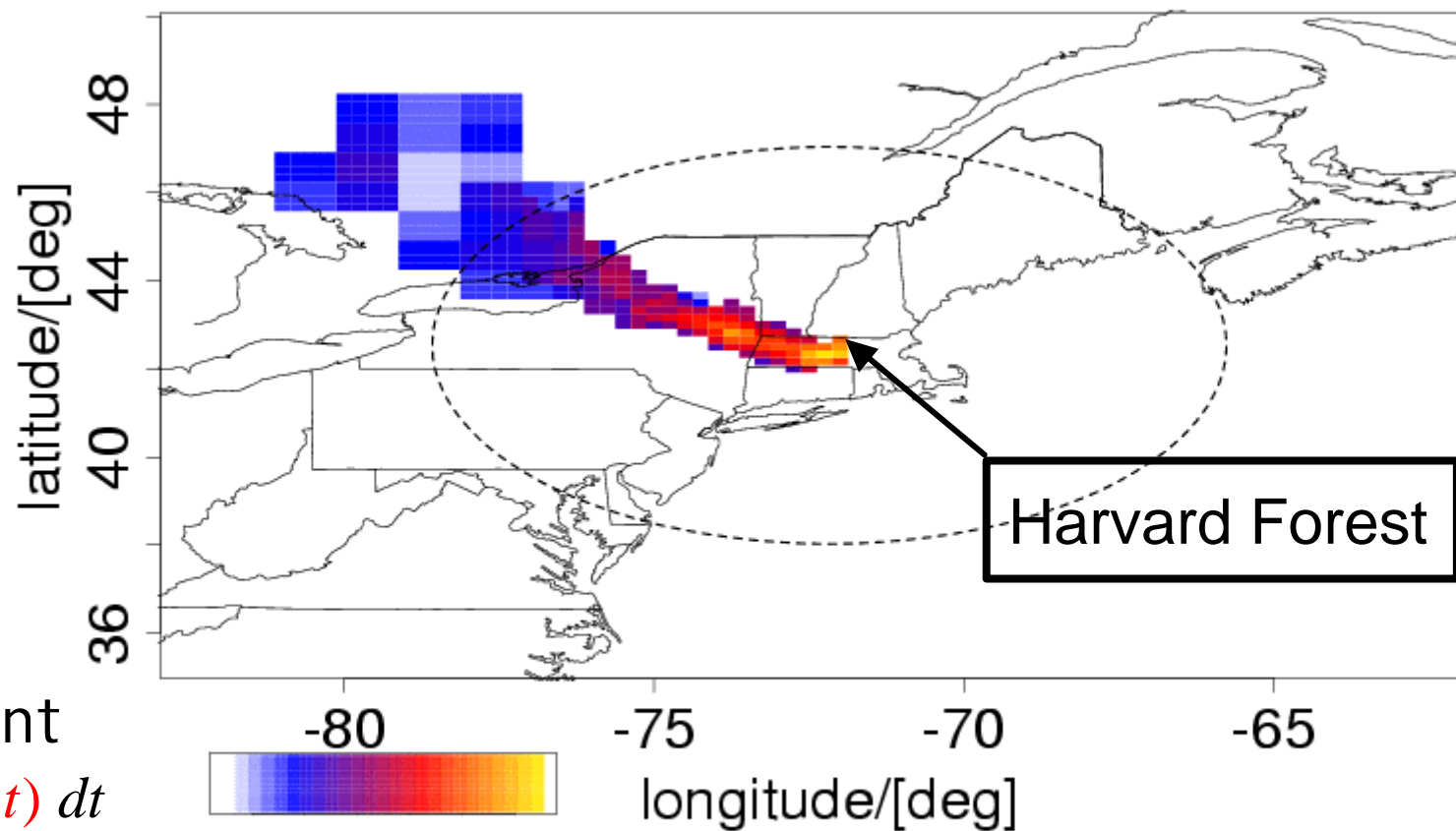
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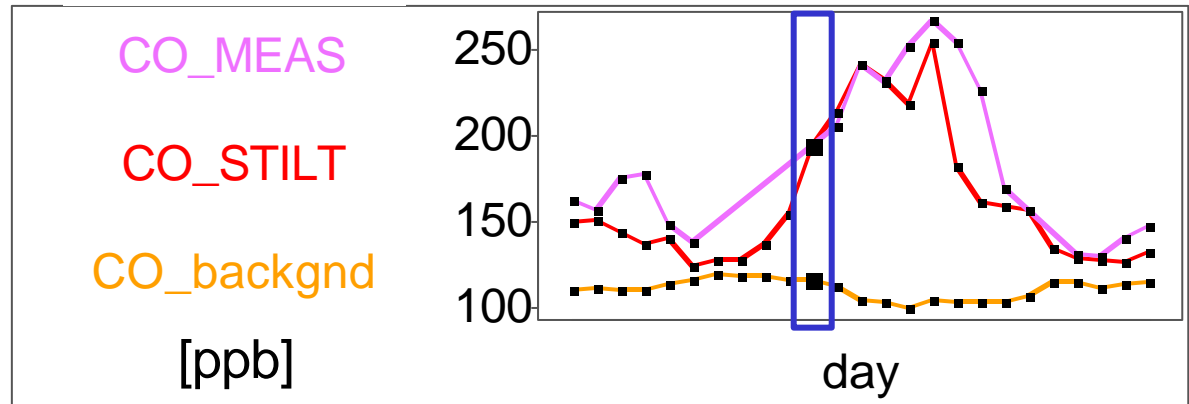
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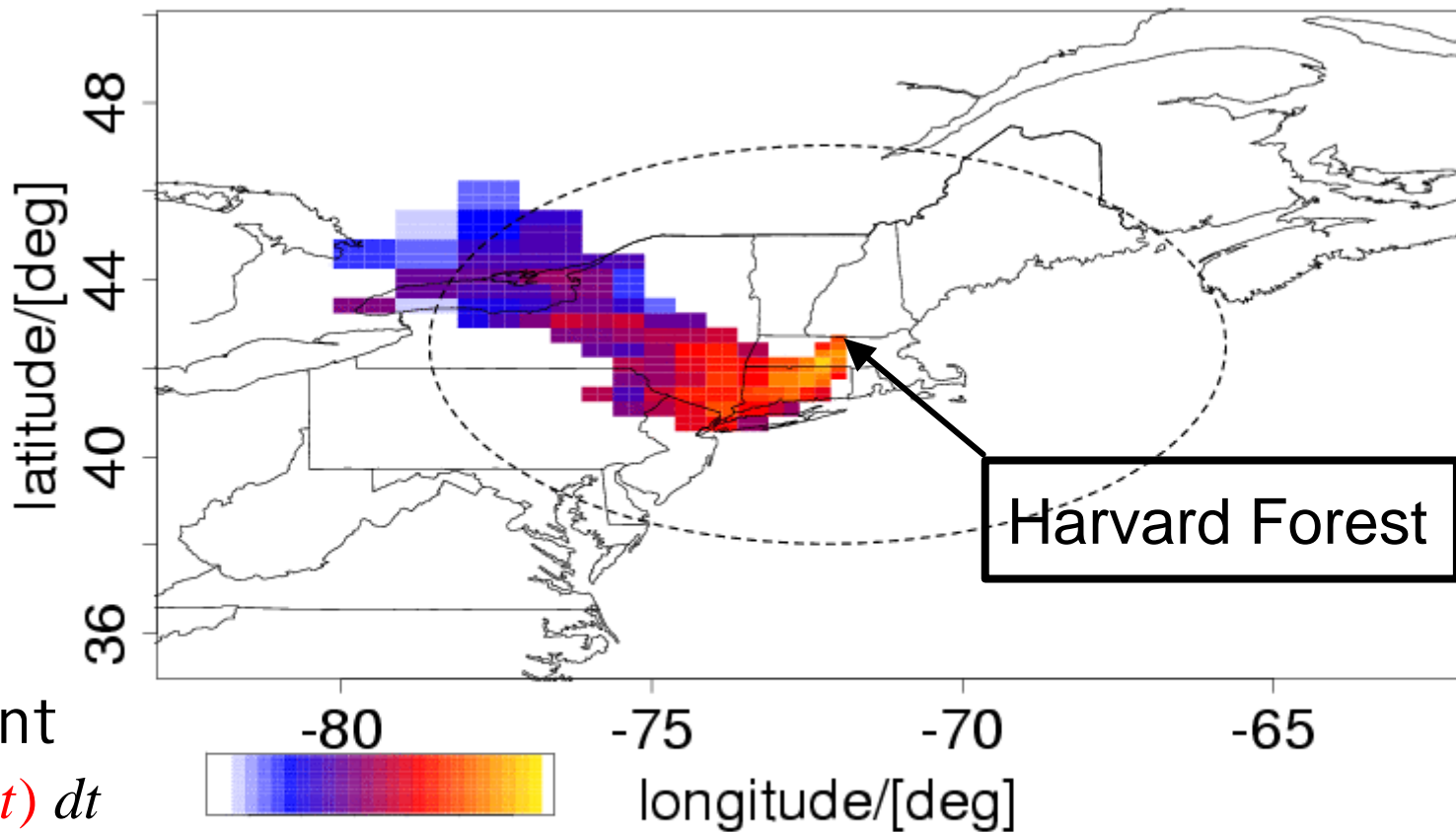
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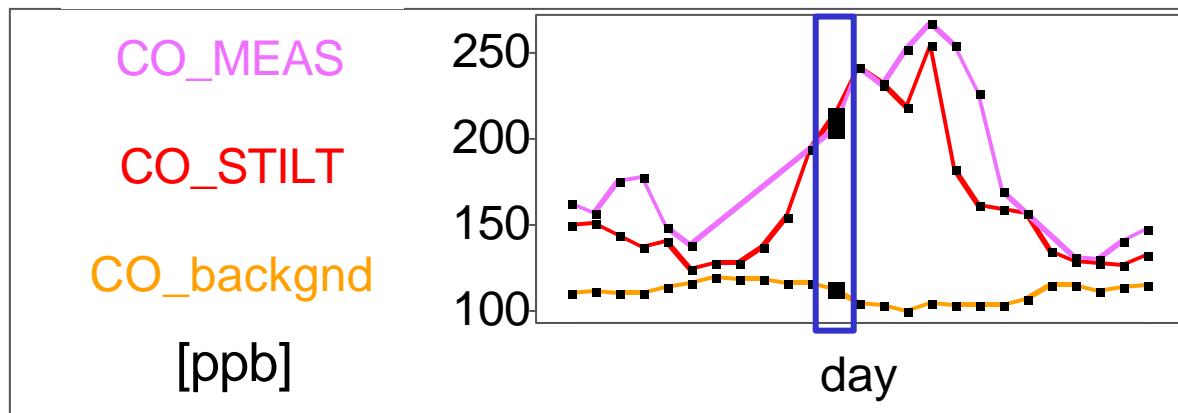
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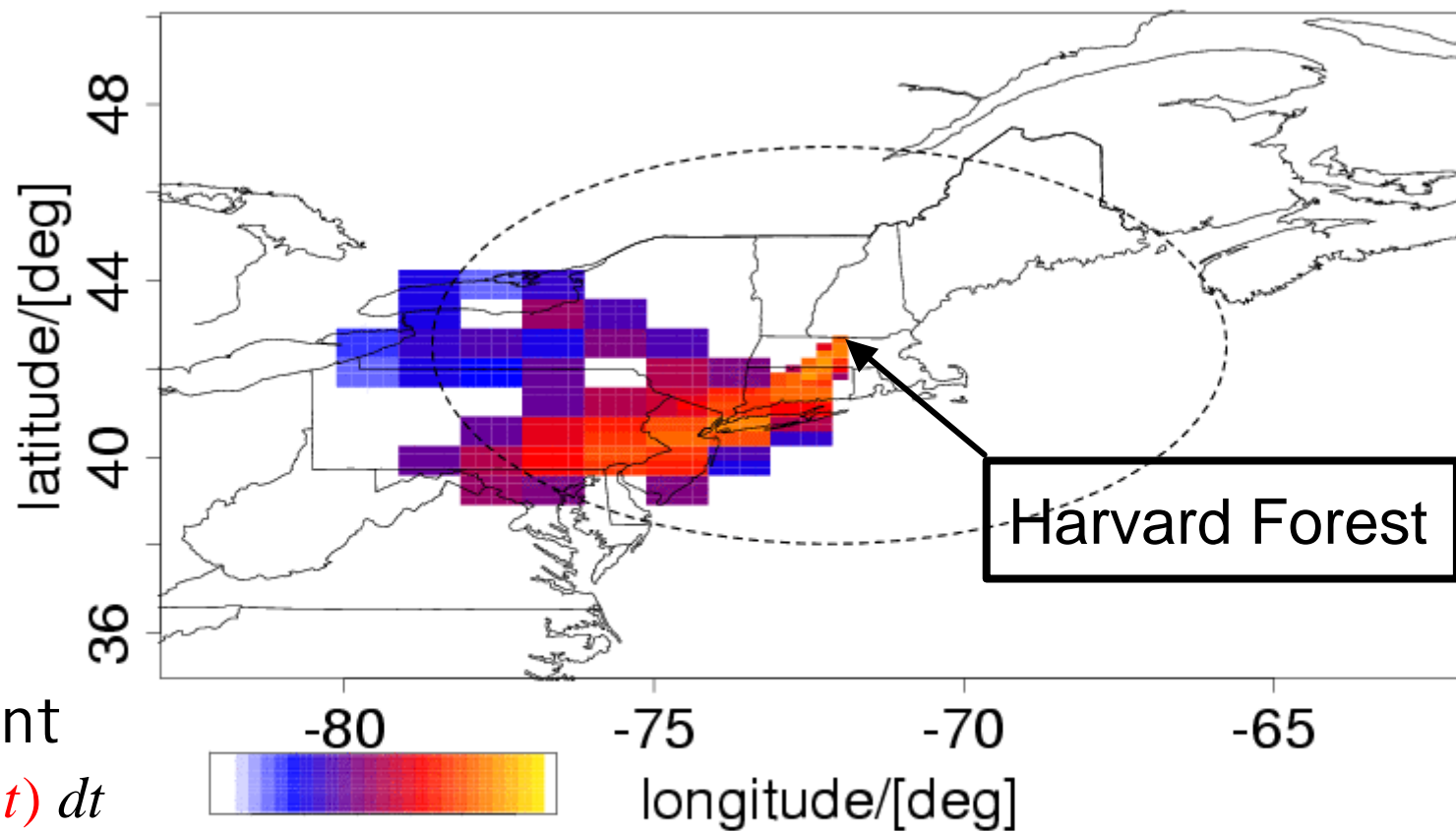




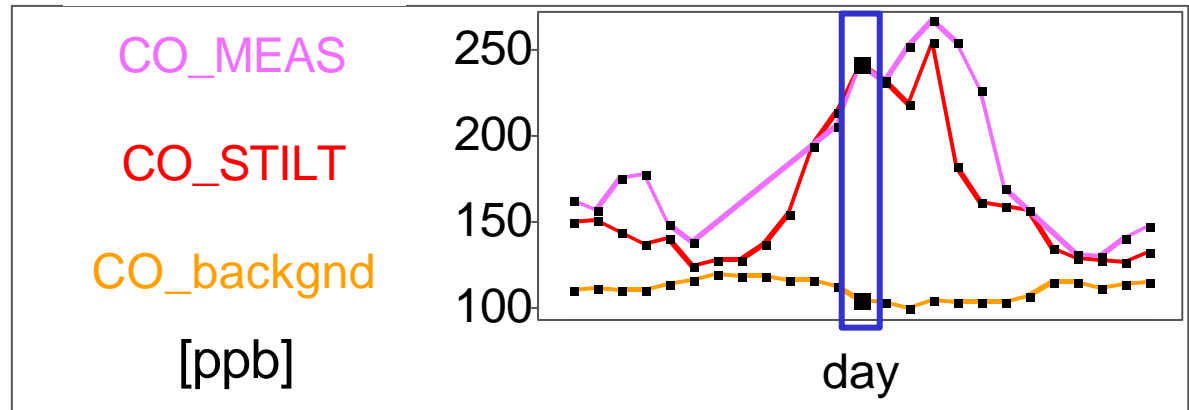
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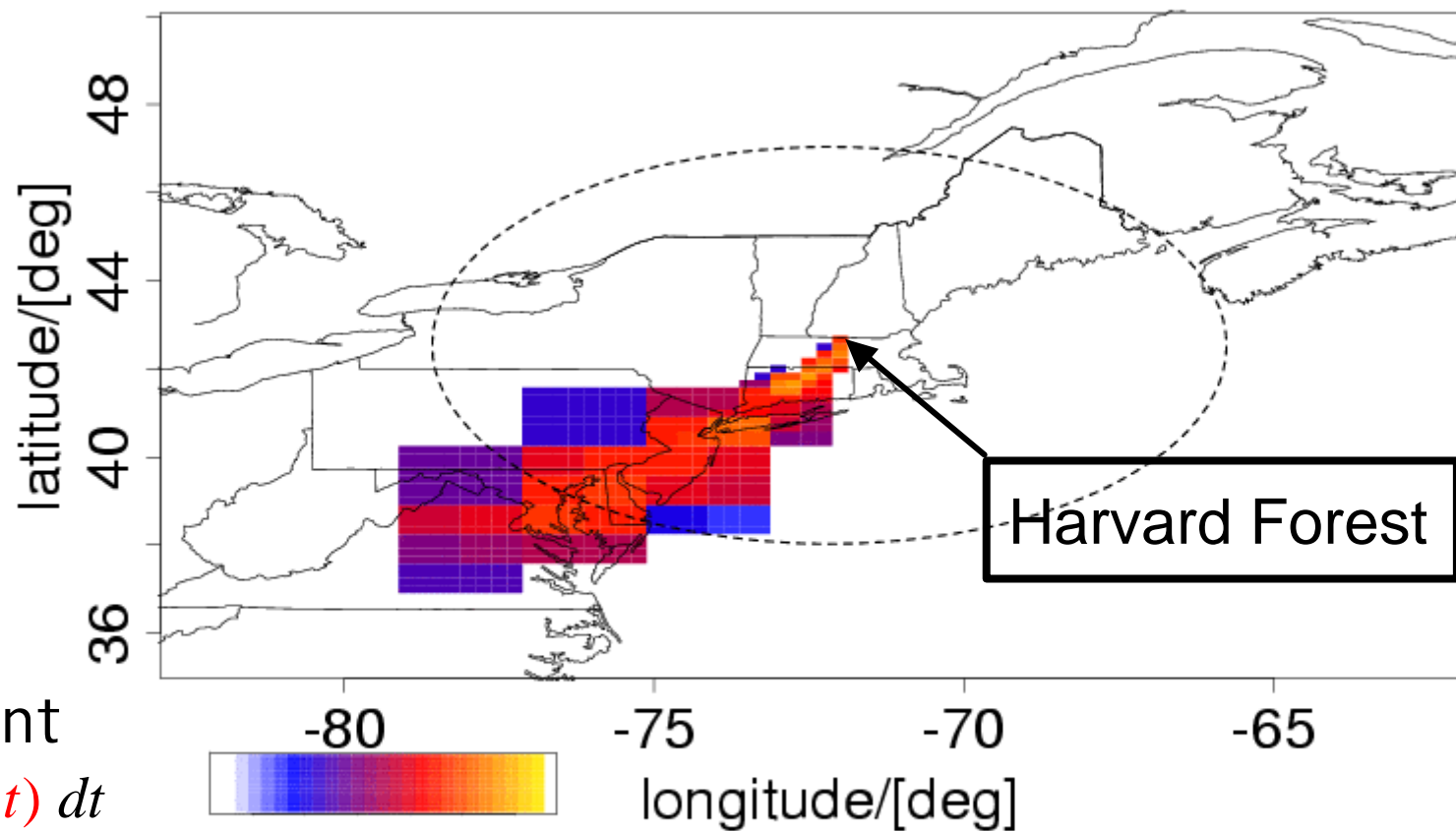
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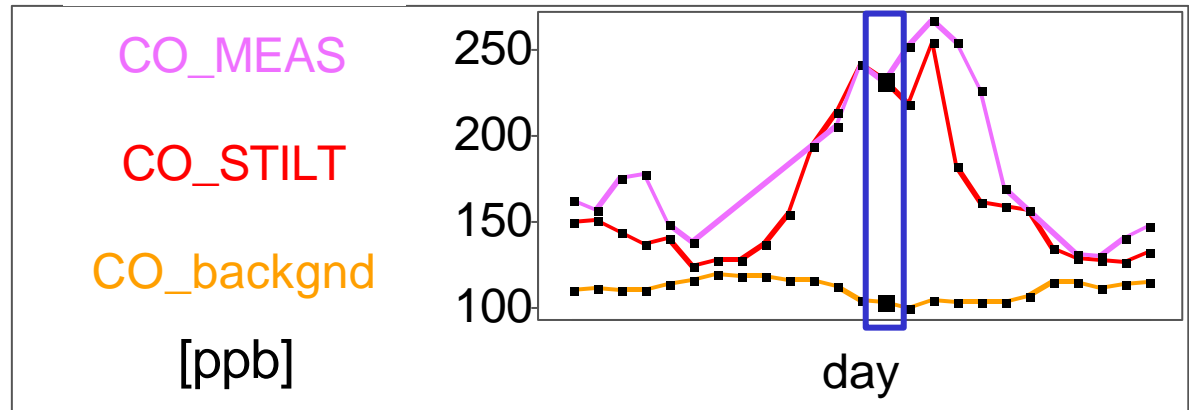
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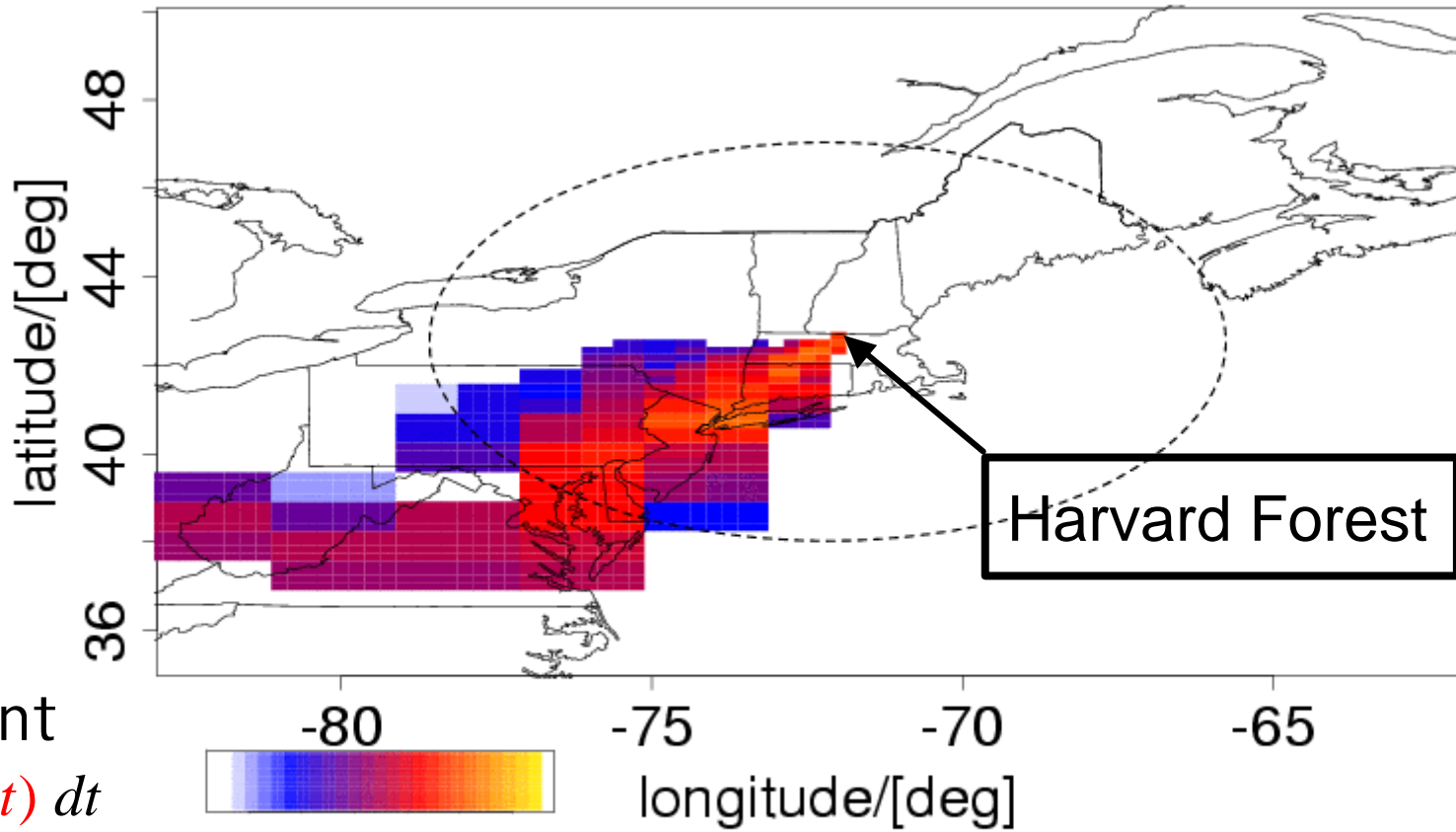
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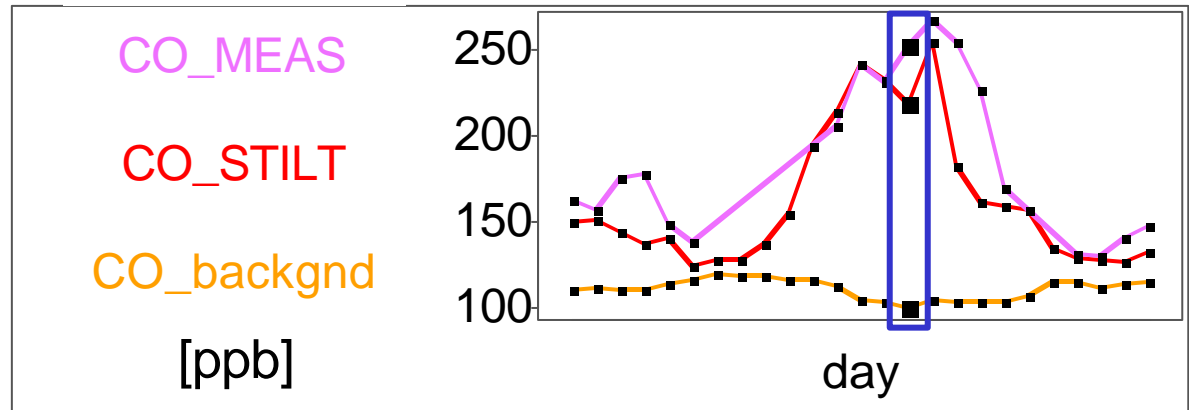
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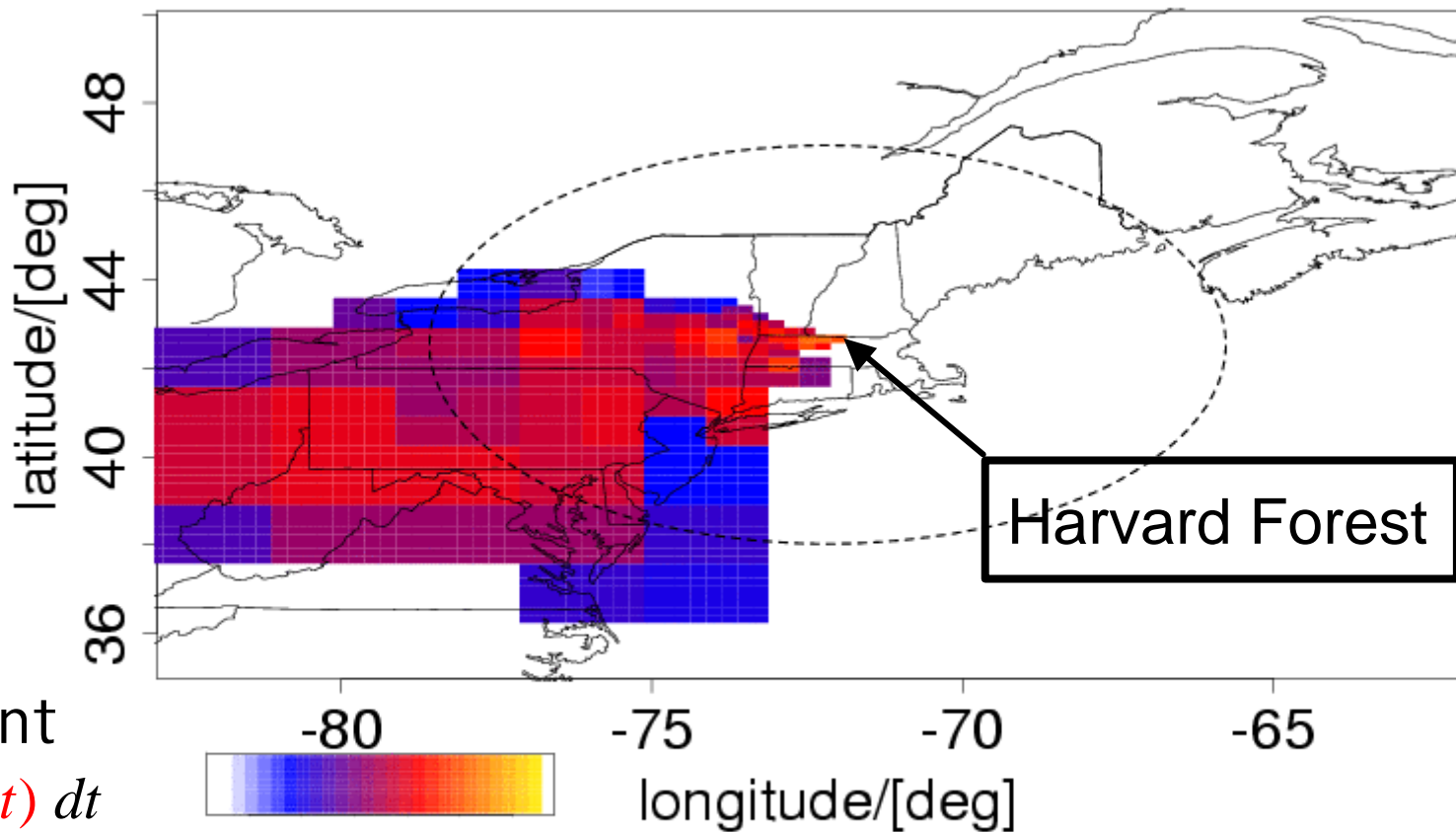
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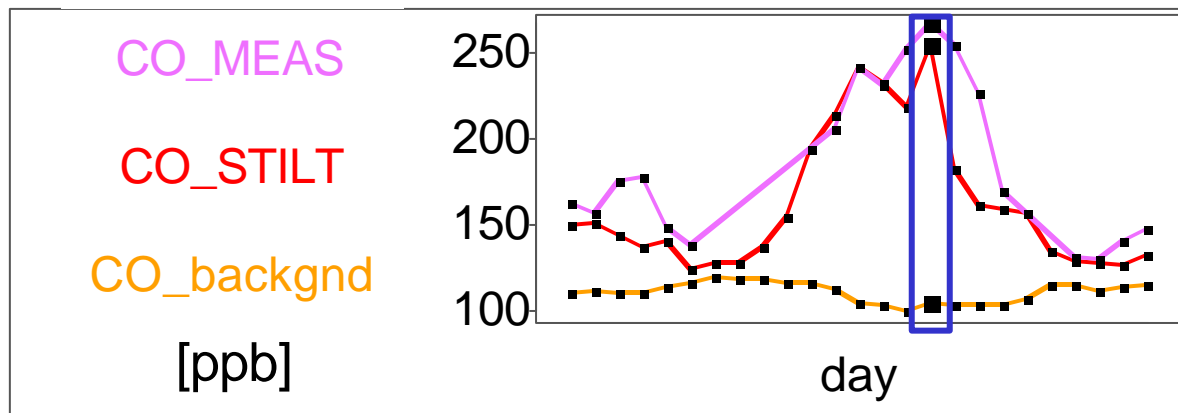
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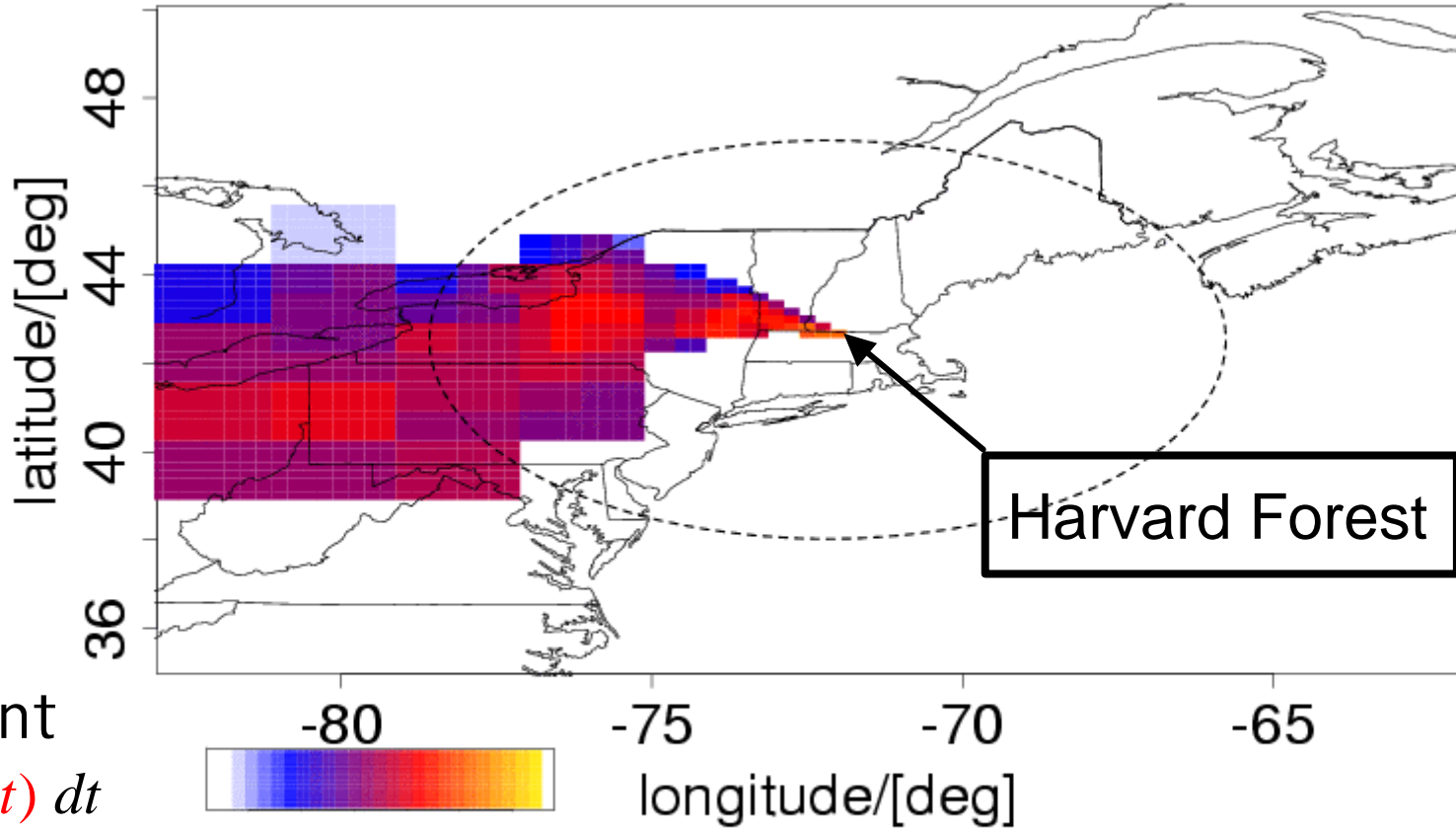
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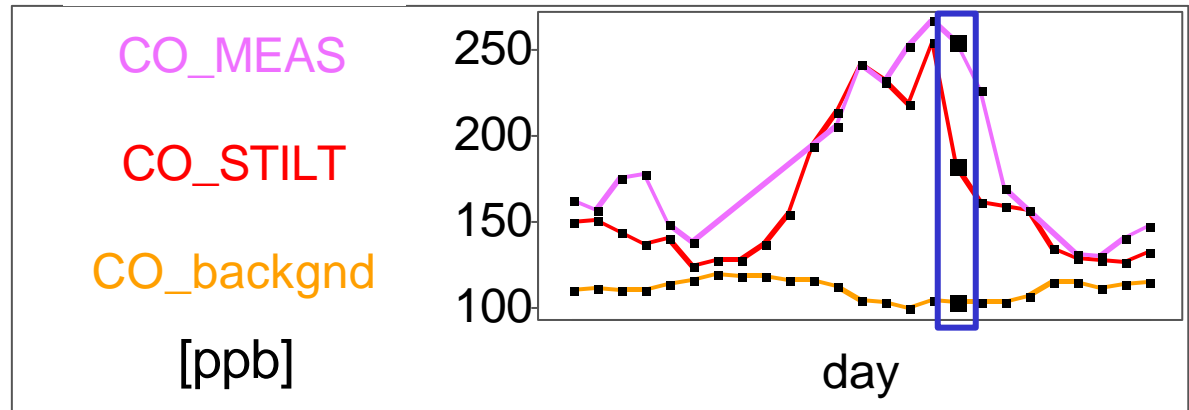
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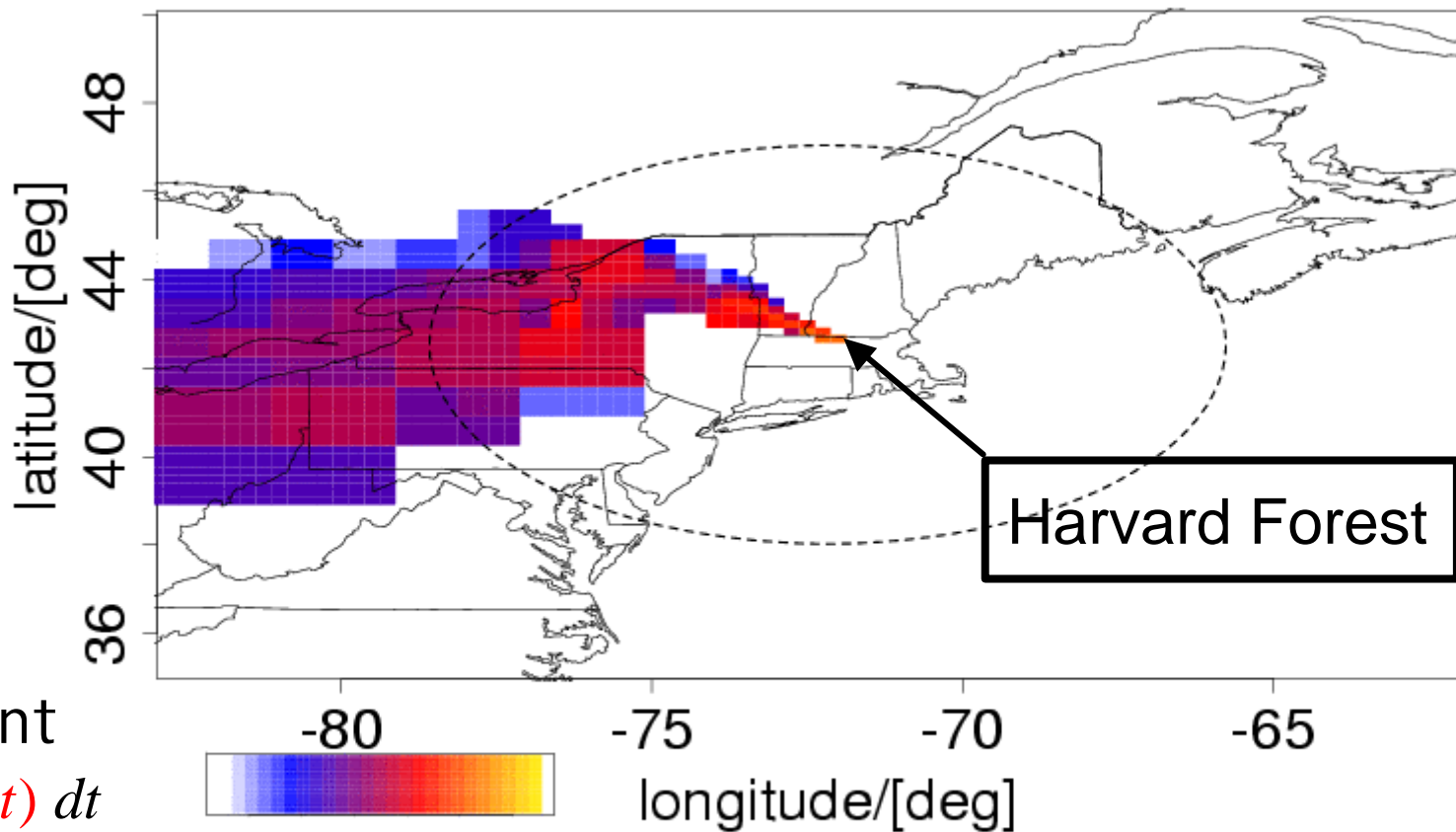




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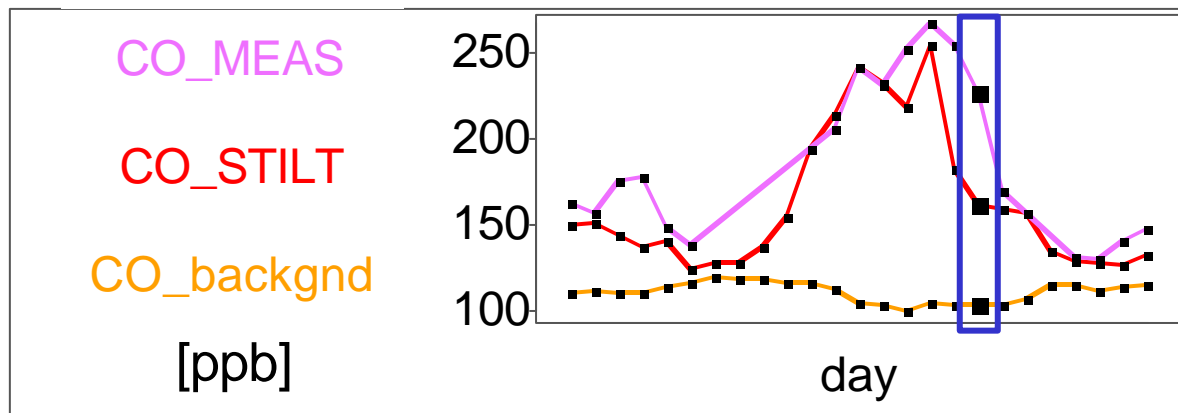
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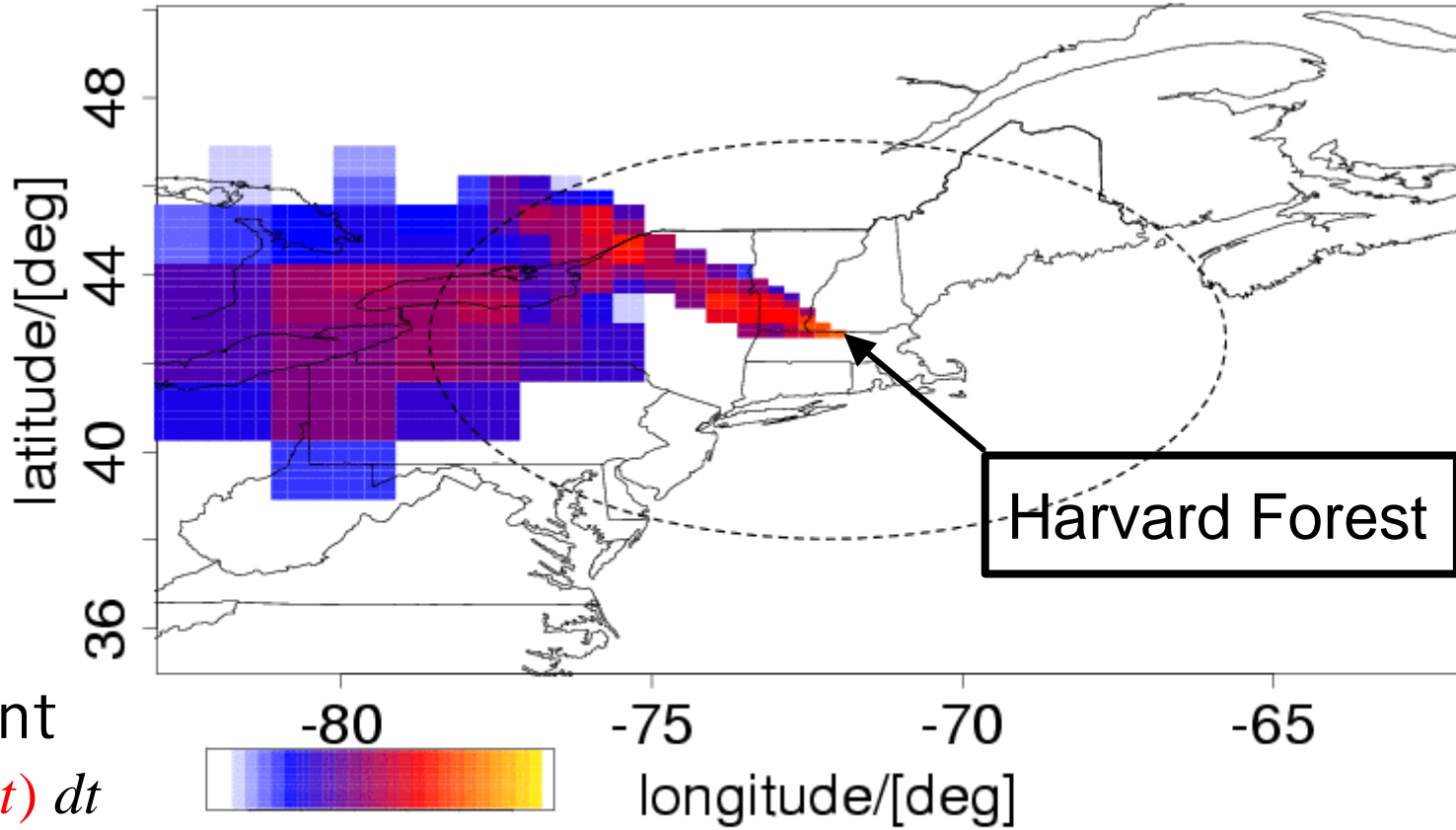
$$\int I(\mathbf{x}_r, t_r | \mathbf{x}, t) dt$$



# What does a (tall) tower "see"?

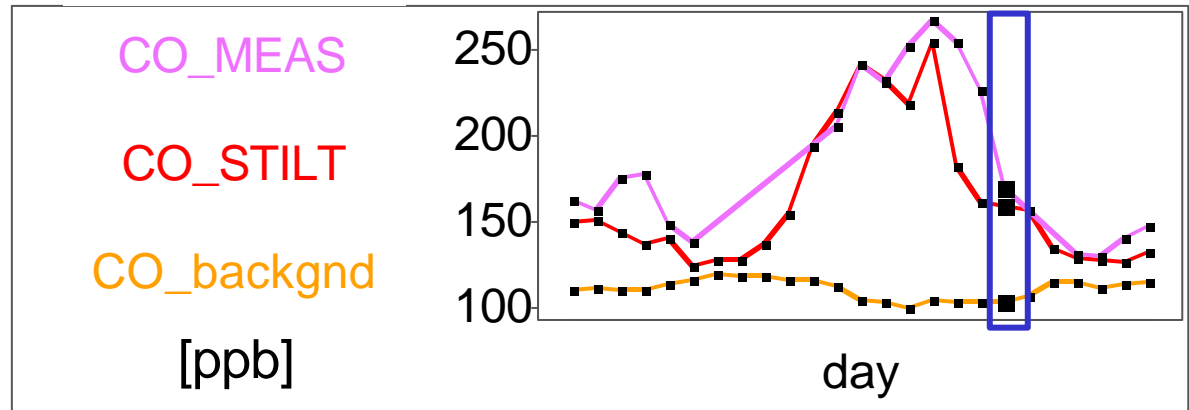
## STILT

Stochastic Time  
Inverted  
Lagrangian  
Transport Model  
[Gerbig *et al.*,  
2003b]  
[Lin *et al.*, 2003]



Footprint

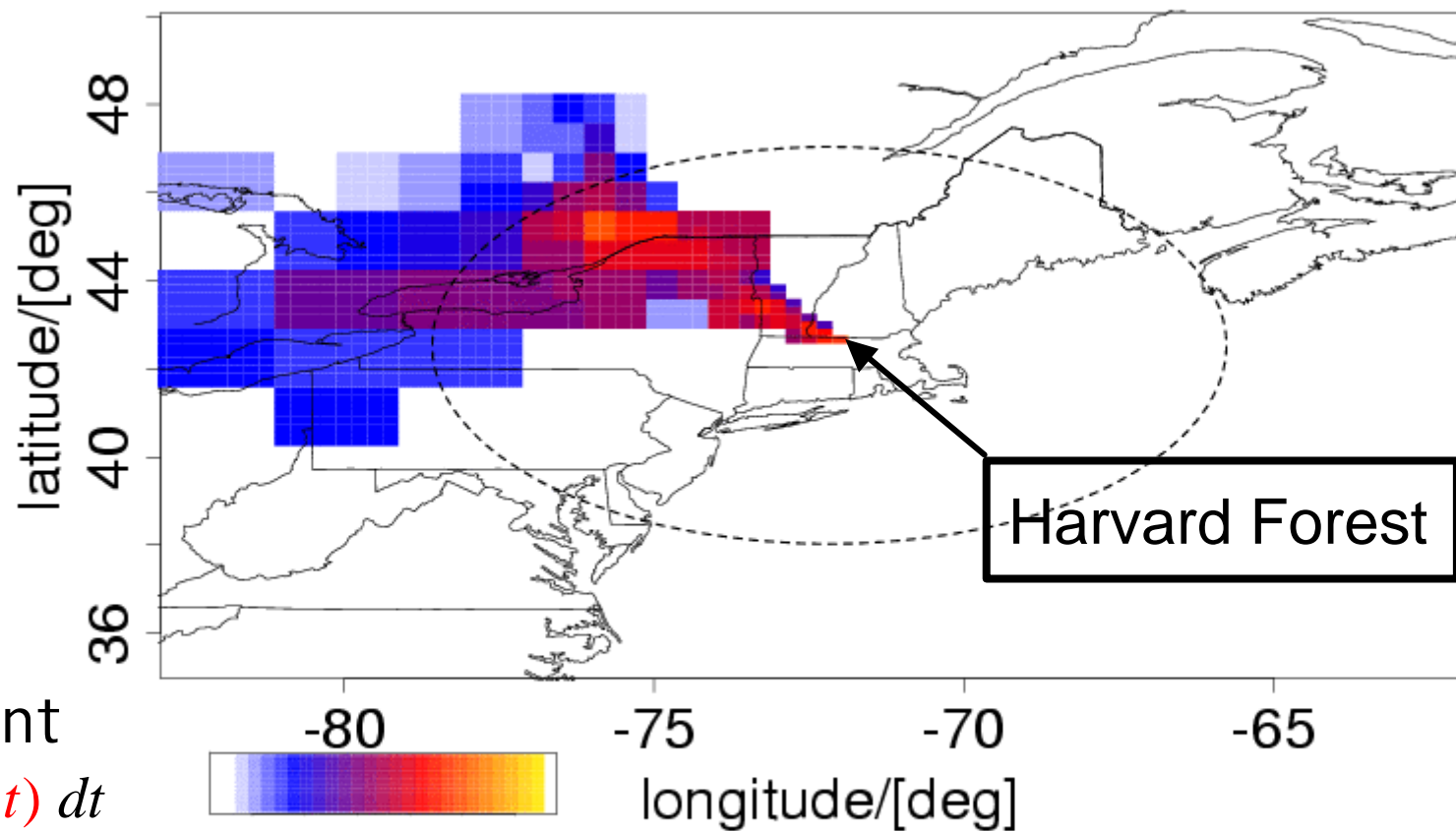
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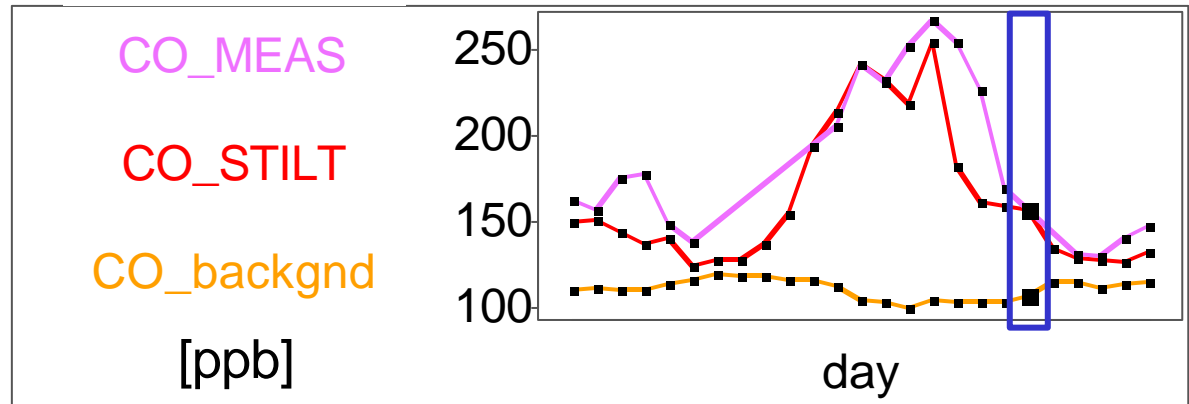
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[Gerbig et al.,  
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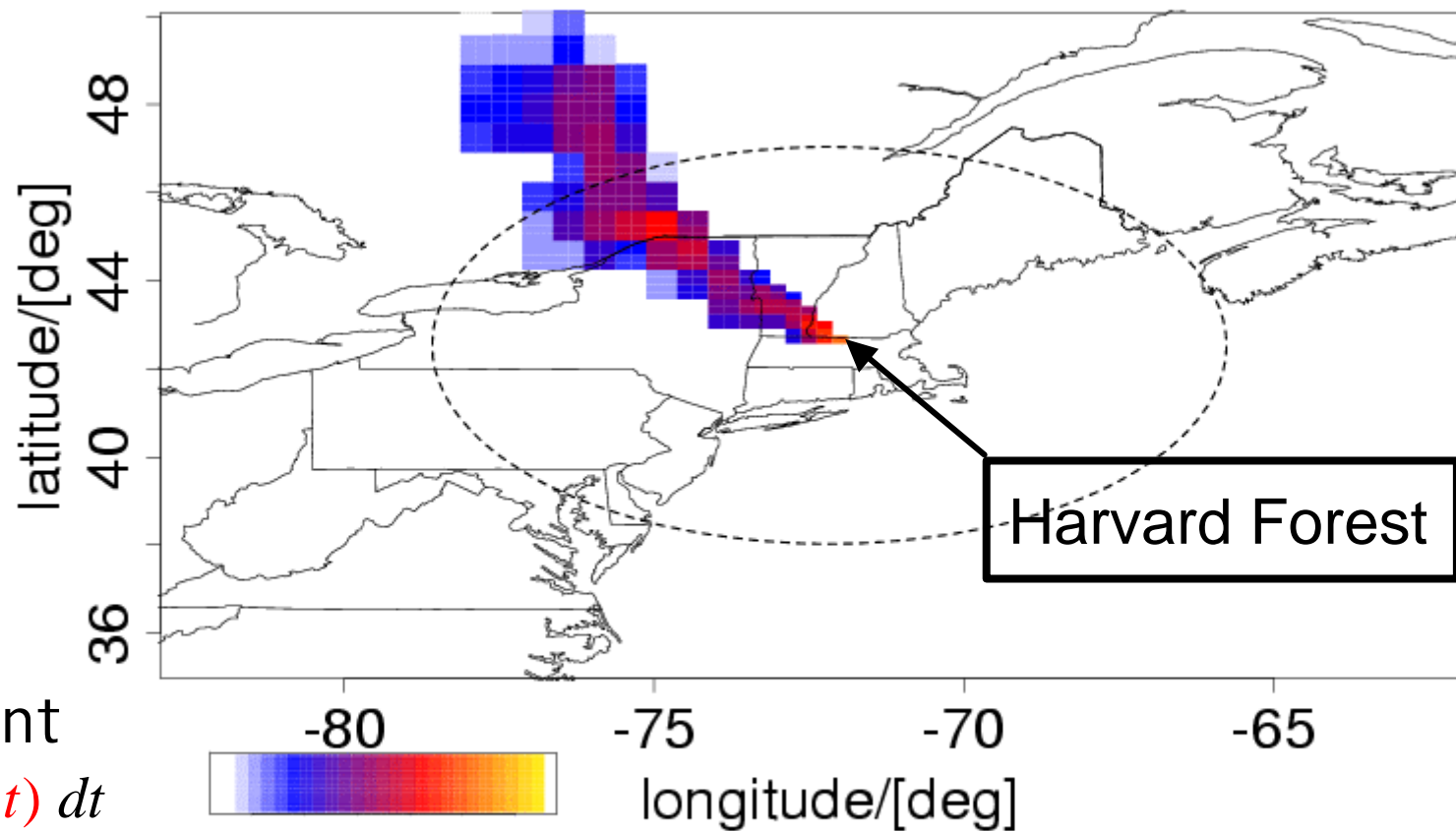
Footprint  
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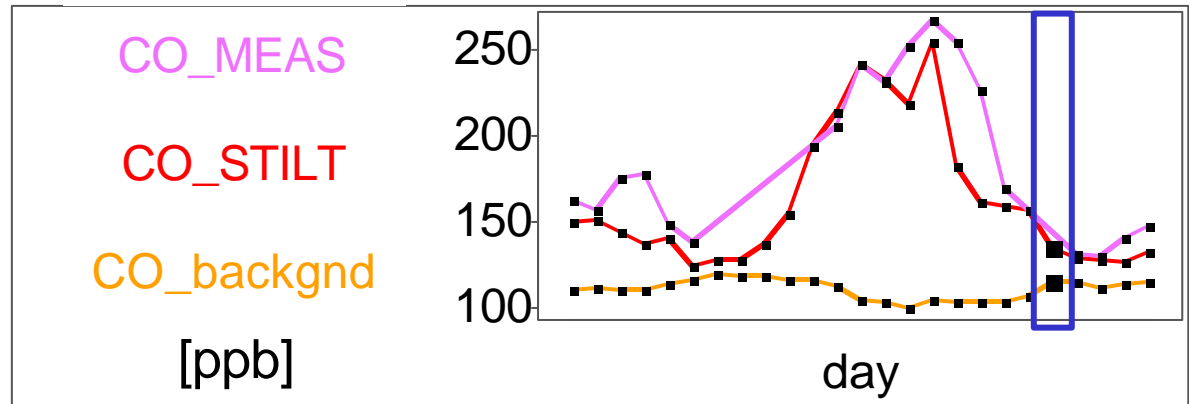
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Footprint

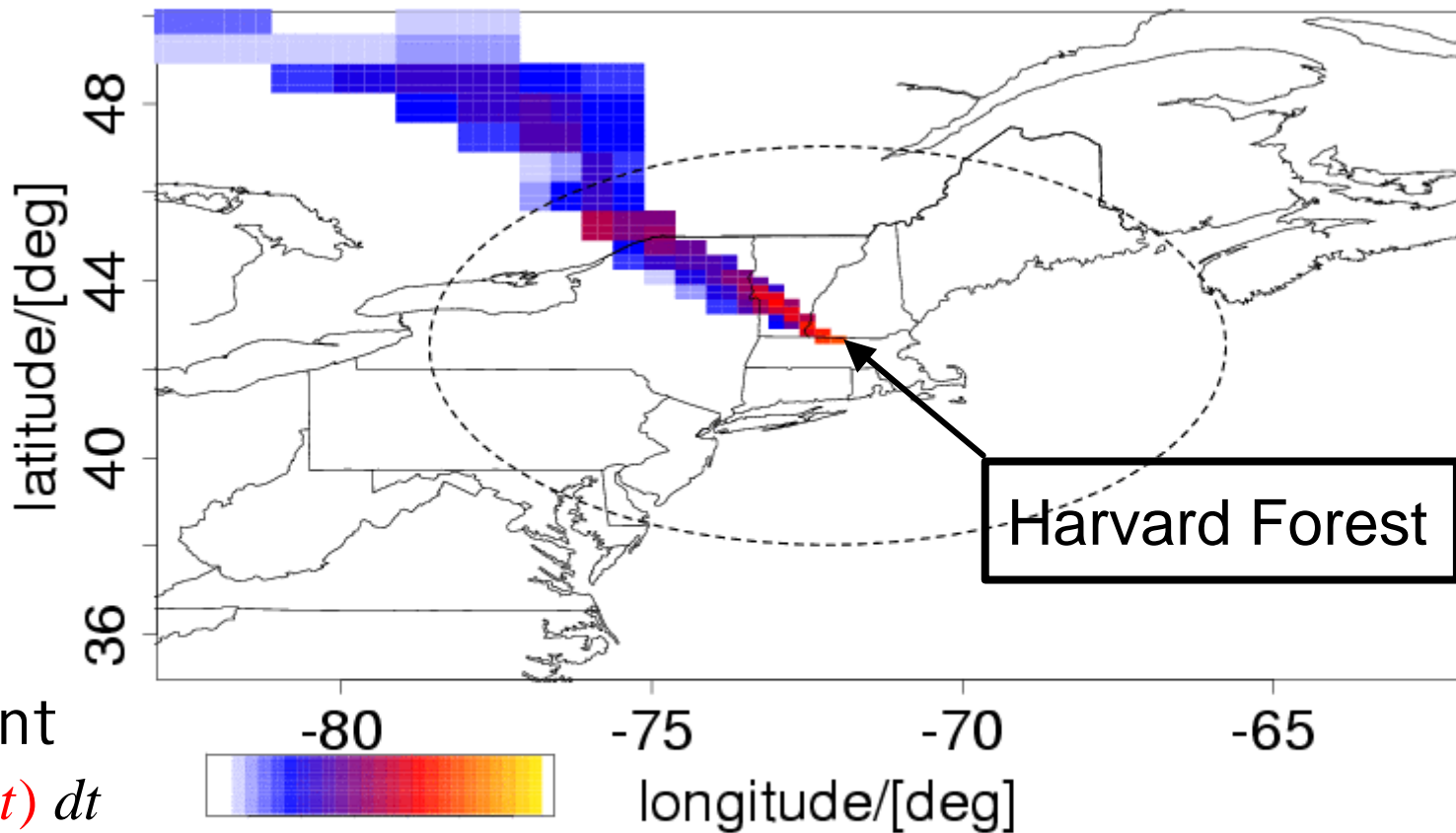
$$\int I(\mathbf{x}_r, t_r | \mathbf{x}, t) dt$$



# What does a (tall) tower "see"?

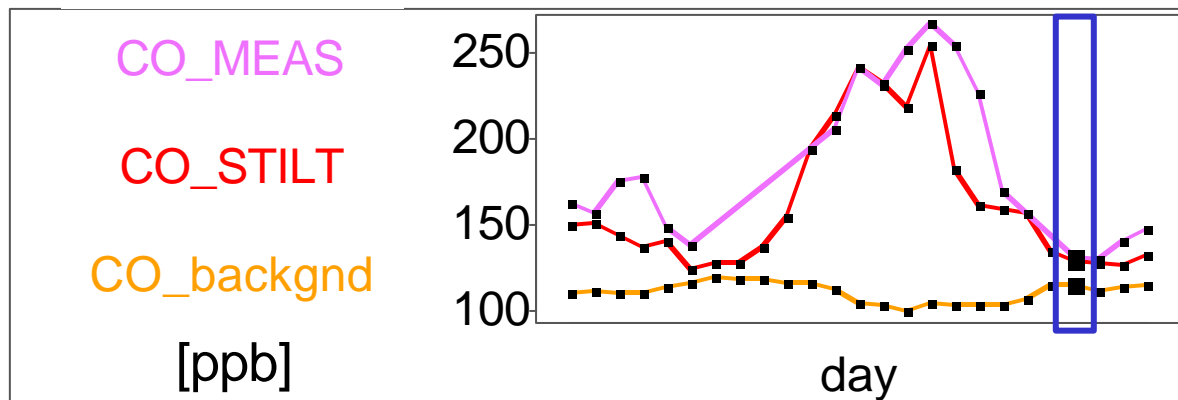
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Stochastic Time  
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[Gerbig *et al.*,  
2003b]  
[Lin *et al.*, 2003]



Footprint

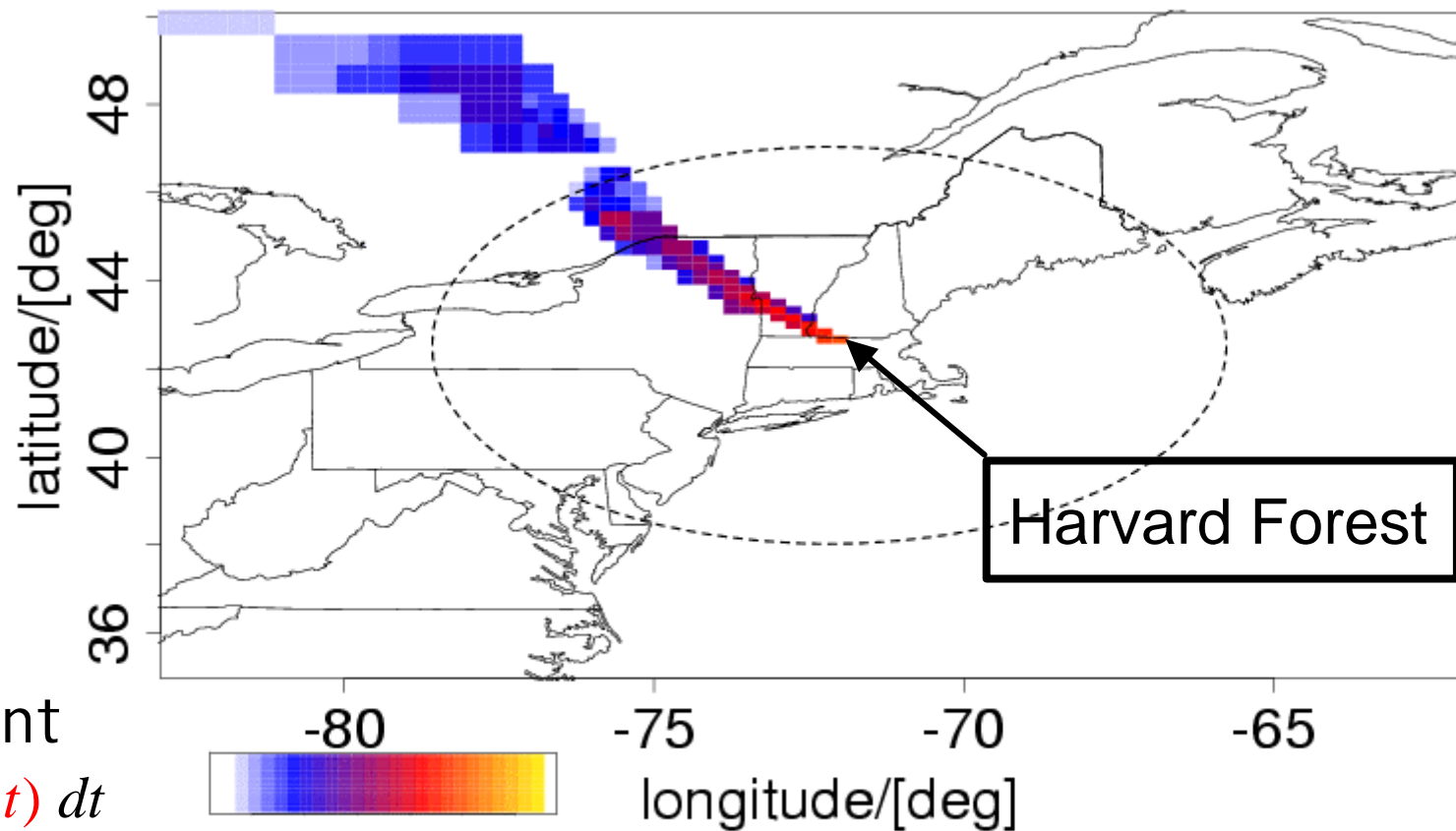
$$\int I(\mathbf{x}_r, t_r | \mathbf{x}, t) dt$$



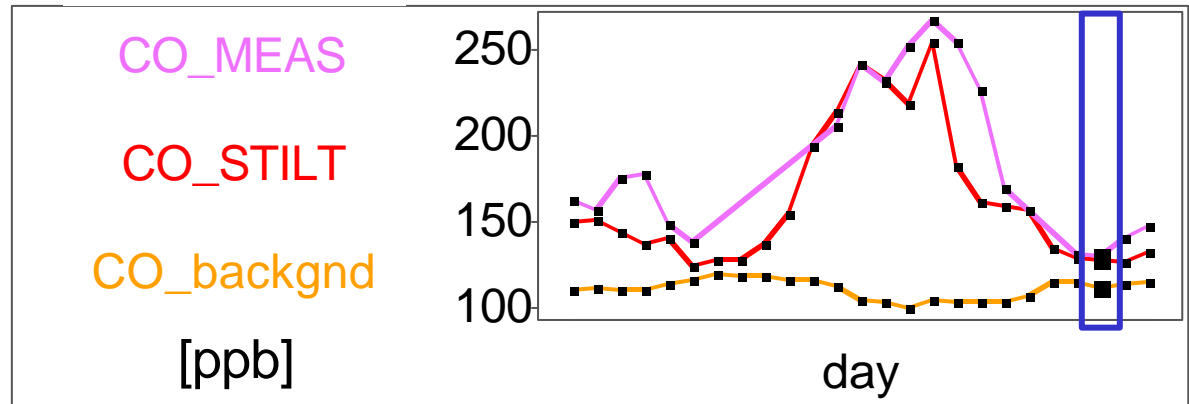
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Footprint  
 $\int I(\mathbf{x}_r, t_r | \mathbf{x}, t) dt$

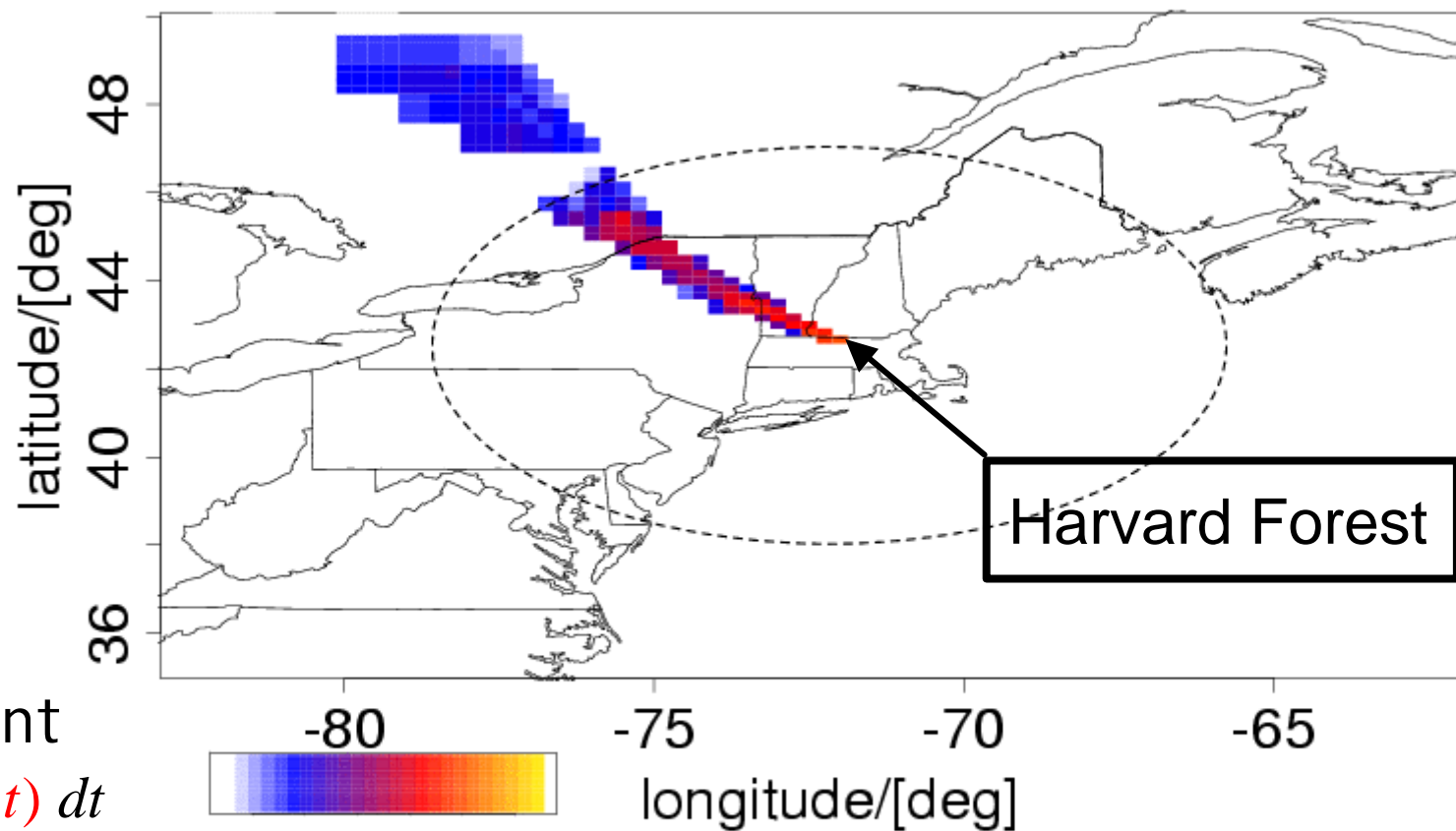




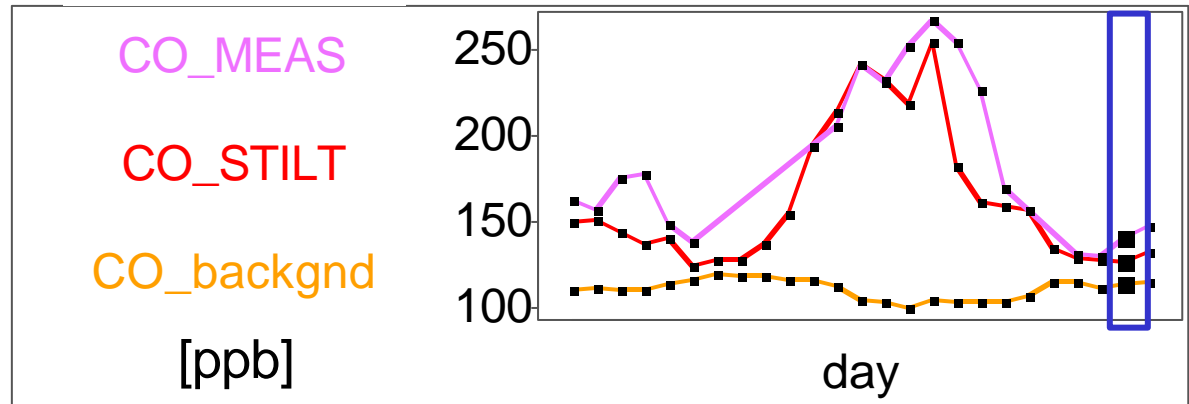
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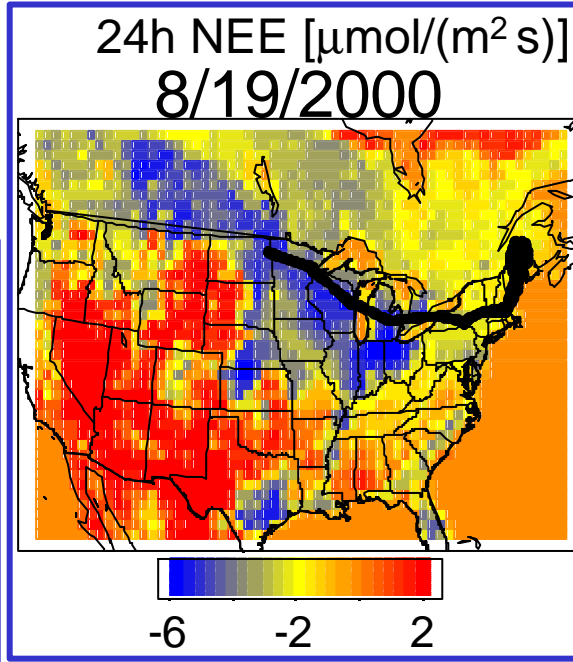
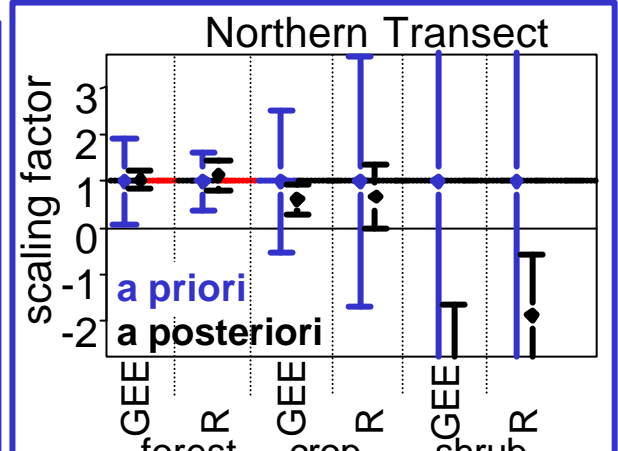
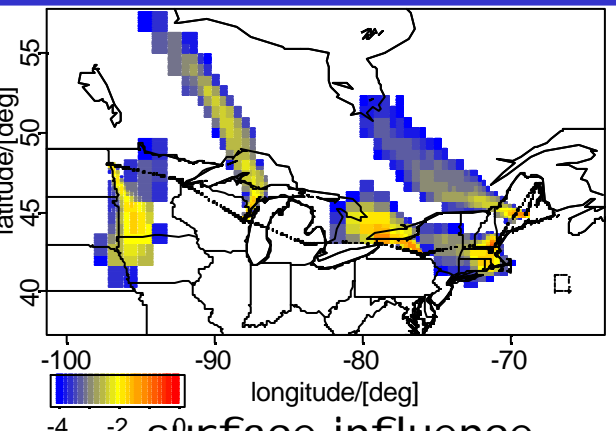
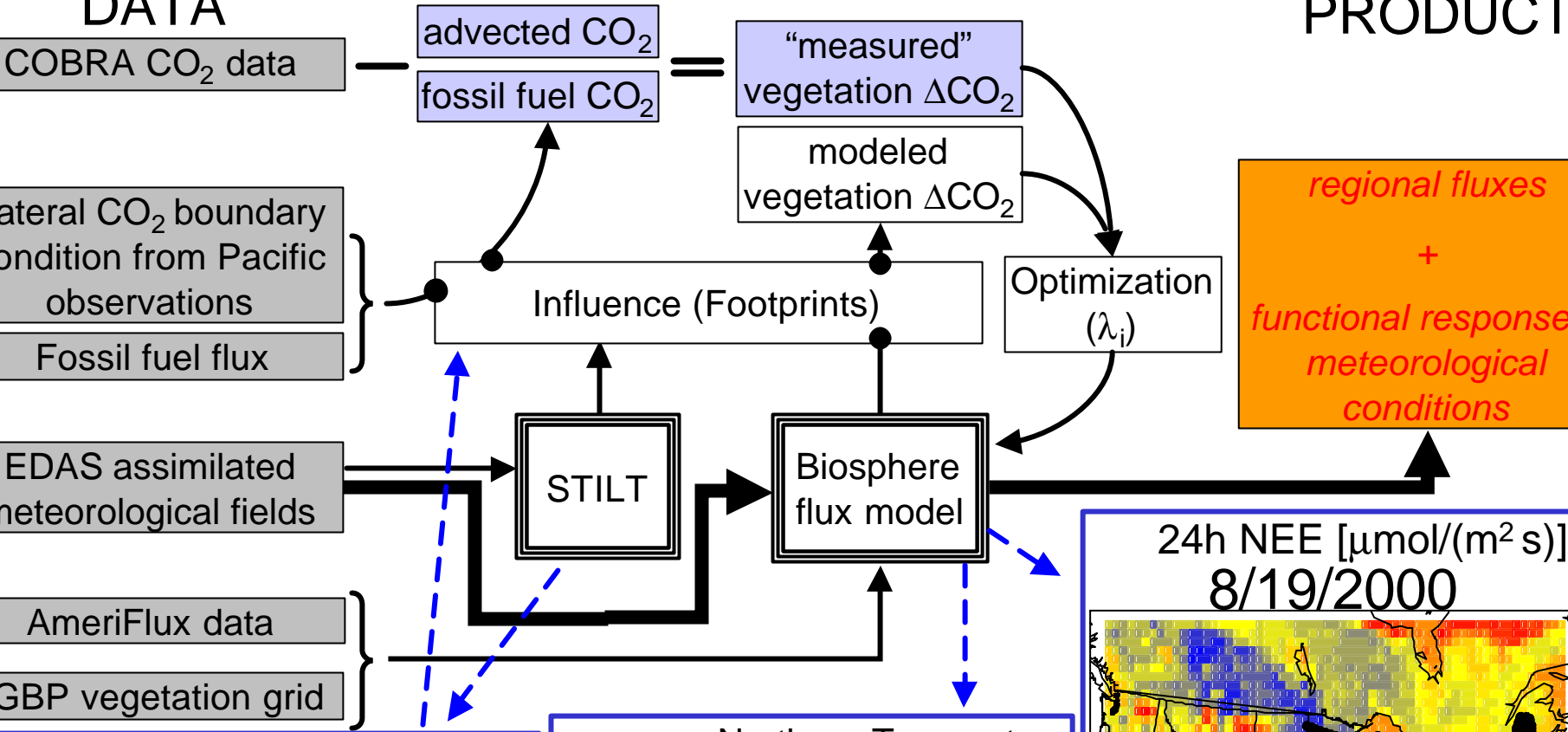
Footprint  
 $\int I(\mathbf{x}_r, t_r | \mathbf{x}, t) dt$



# Receptor Oriented Atmospheric Model "ROAM"

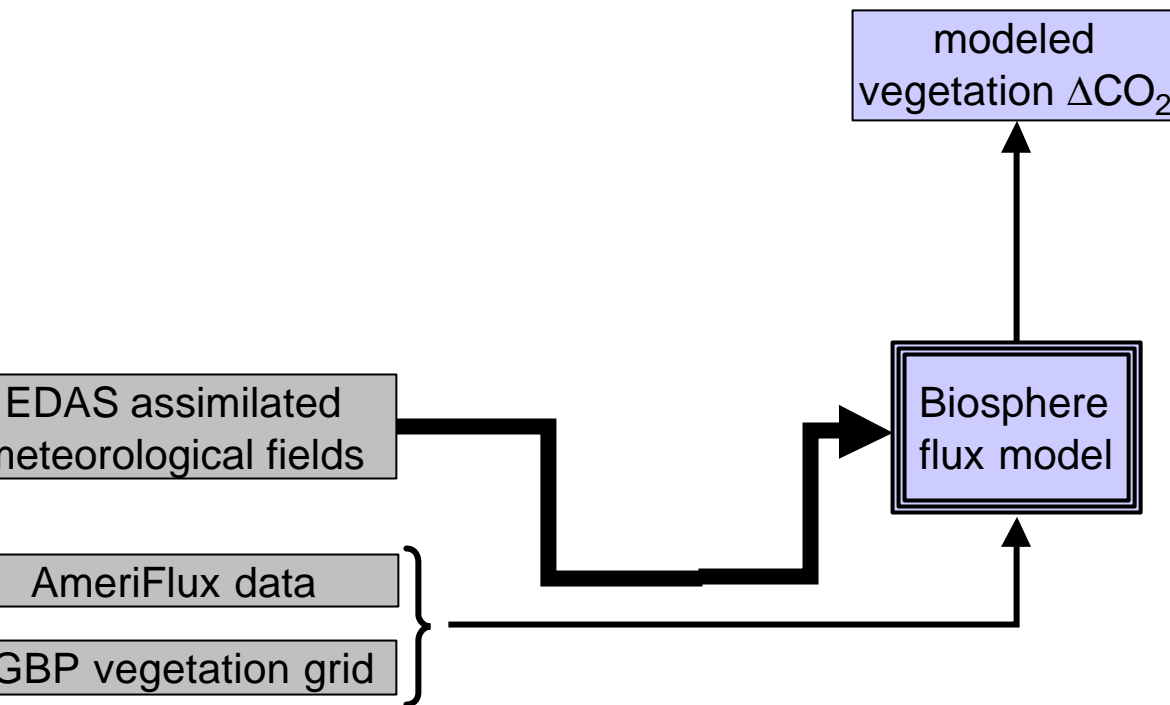
DATA

PRODUCT

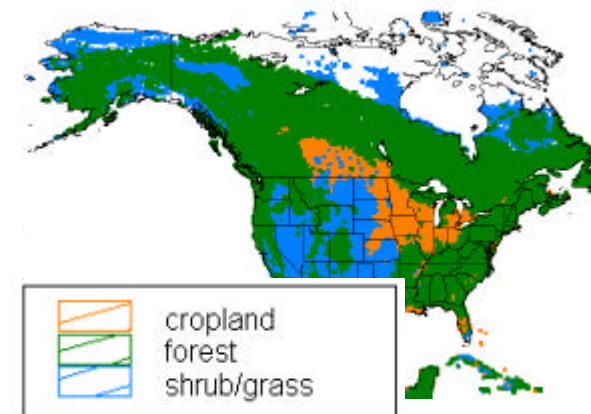


[Gerbic et al. 2002]

# Receptor Oriented Atmospheric Model "ROAM"

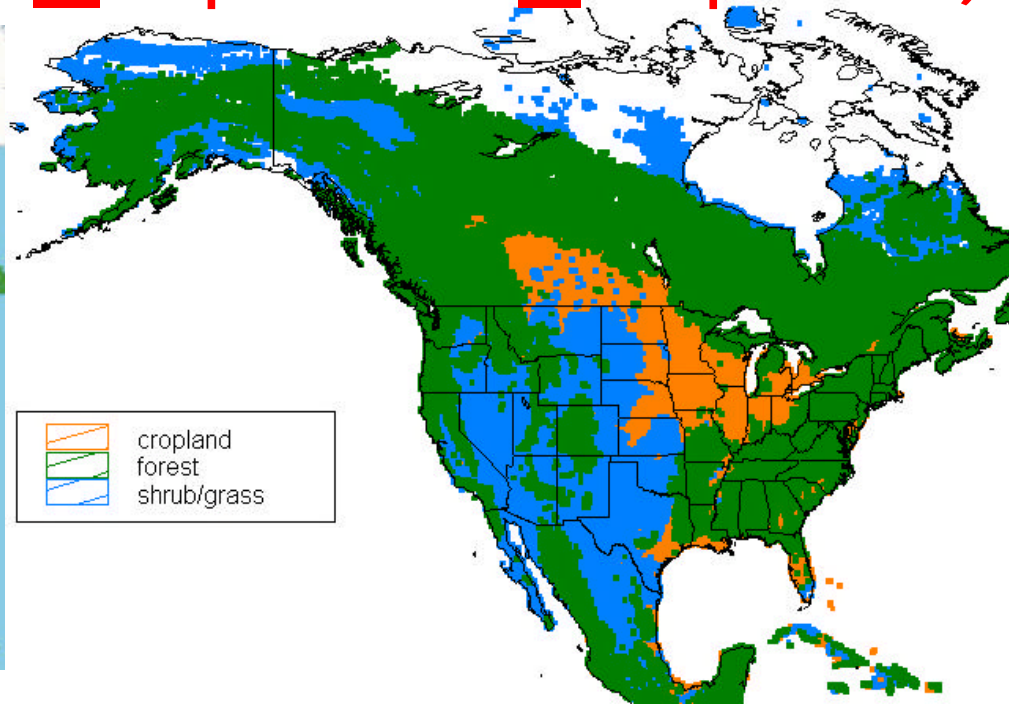
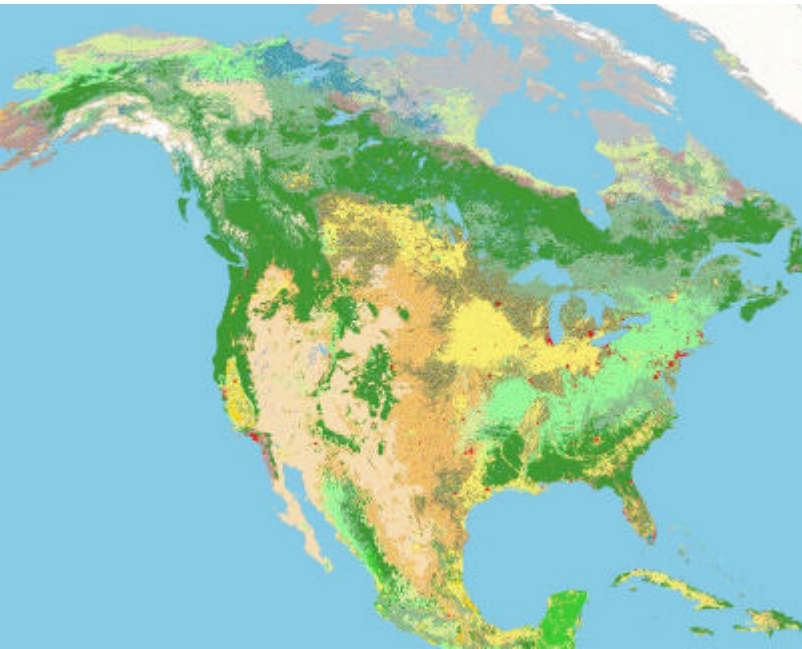


The **GSB** (Greatly Simplified Biosphere)



+ *Radiation and Temperature Sensitivity at ~ 15 eddy flux sites (AmeriFlux)*

# The GSB (Greatly Simplified Biosphere)



GBP Terrestrial Vegetation Map:  
17 classes

Minimal Terrestrial Vegetation Map:  
3 classes

~10-15 useful eddy flux sites (AmeriFlux)  
(mostly NE and SE forests)

$$NEE = I_{ir} b_i T + I_{ip} A_i SWR / (G_i + SWR)$$
  
( $b$ ,  $A$ ,  $\Gamma$ ) <sub>$i$</sub>  from eddy flux data,  $T$  and  $SWR$  from EDAS  
 $\lambda$  factors for upscaling, with a priori uncertainty. [Gerbig et al 2003]

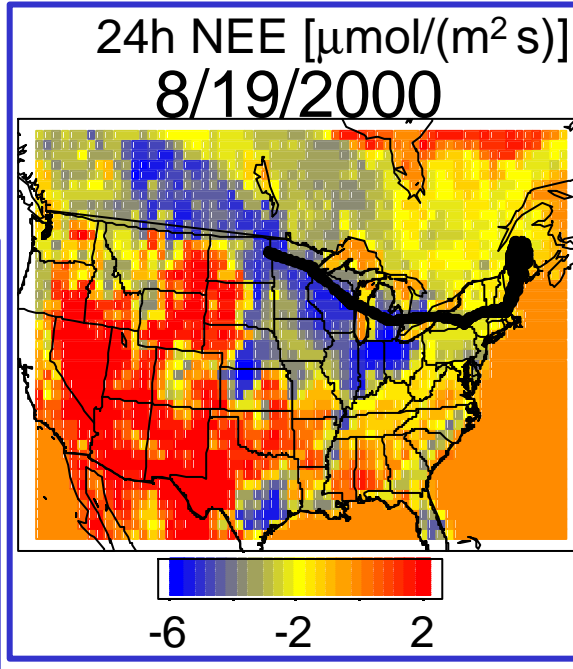
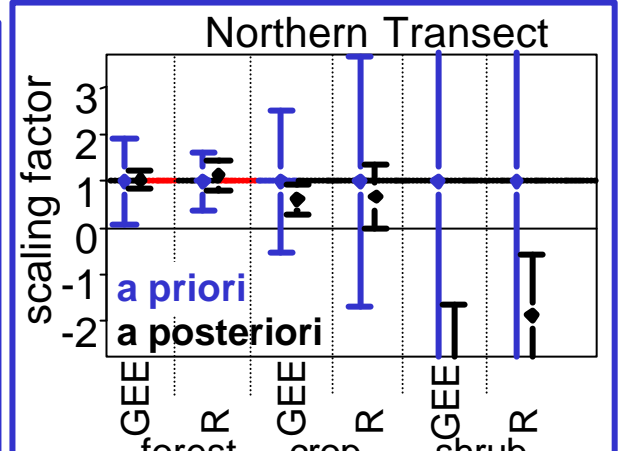
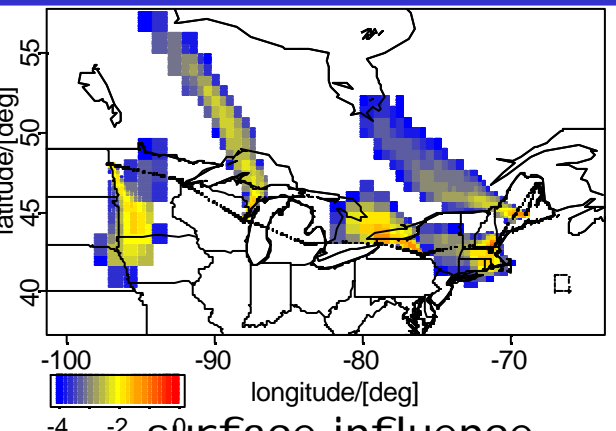
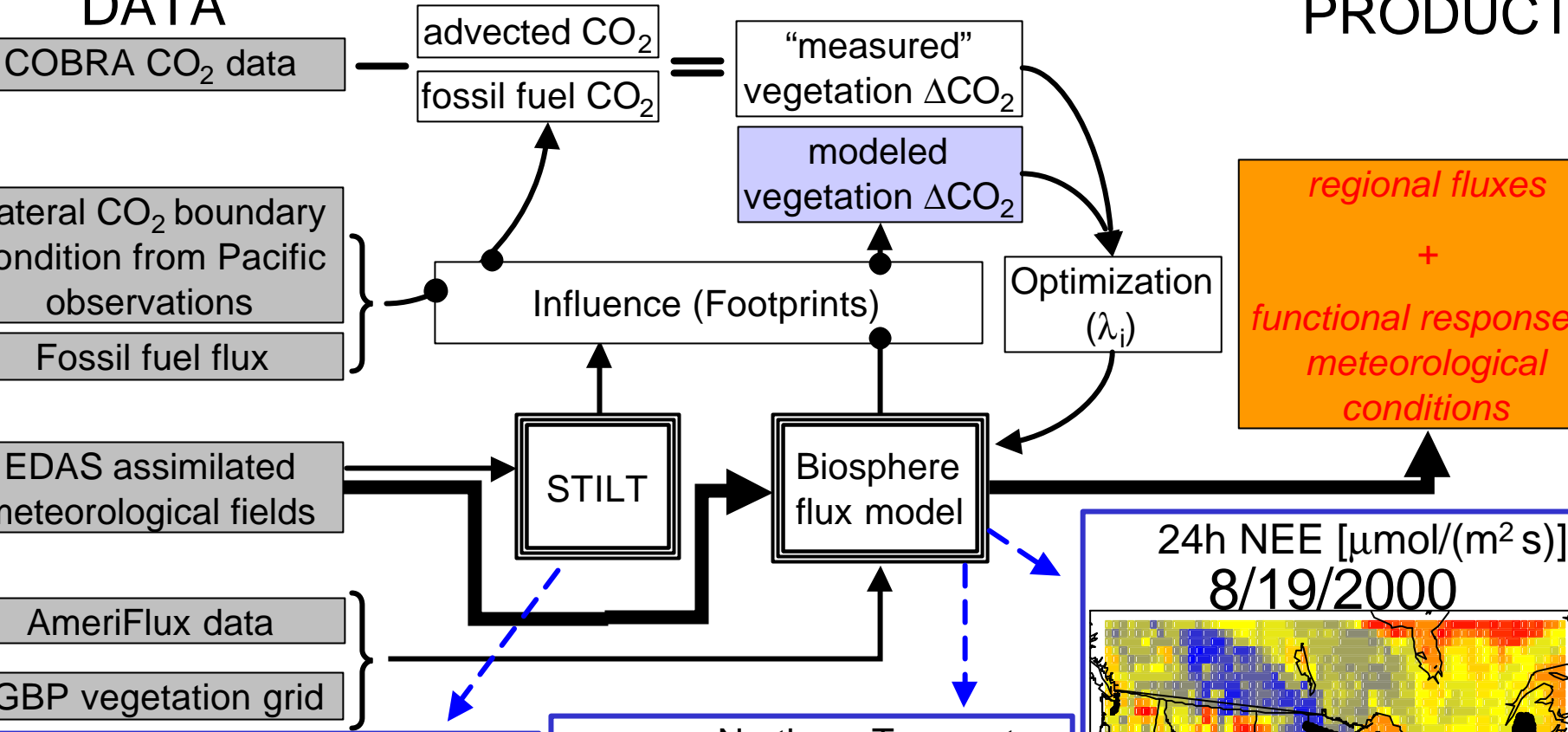
# The GSB (Greatly Simplified Biosphere)

- + captures dominant patterns of variability in space (vegetation cover) and time (light sensitive)
- Not very detailed, only diagnostic

# Receptor Oriented Atmospheric Model "ROAM"

DATA

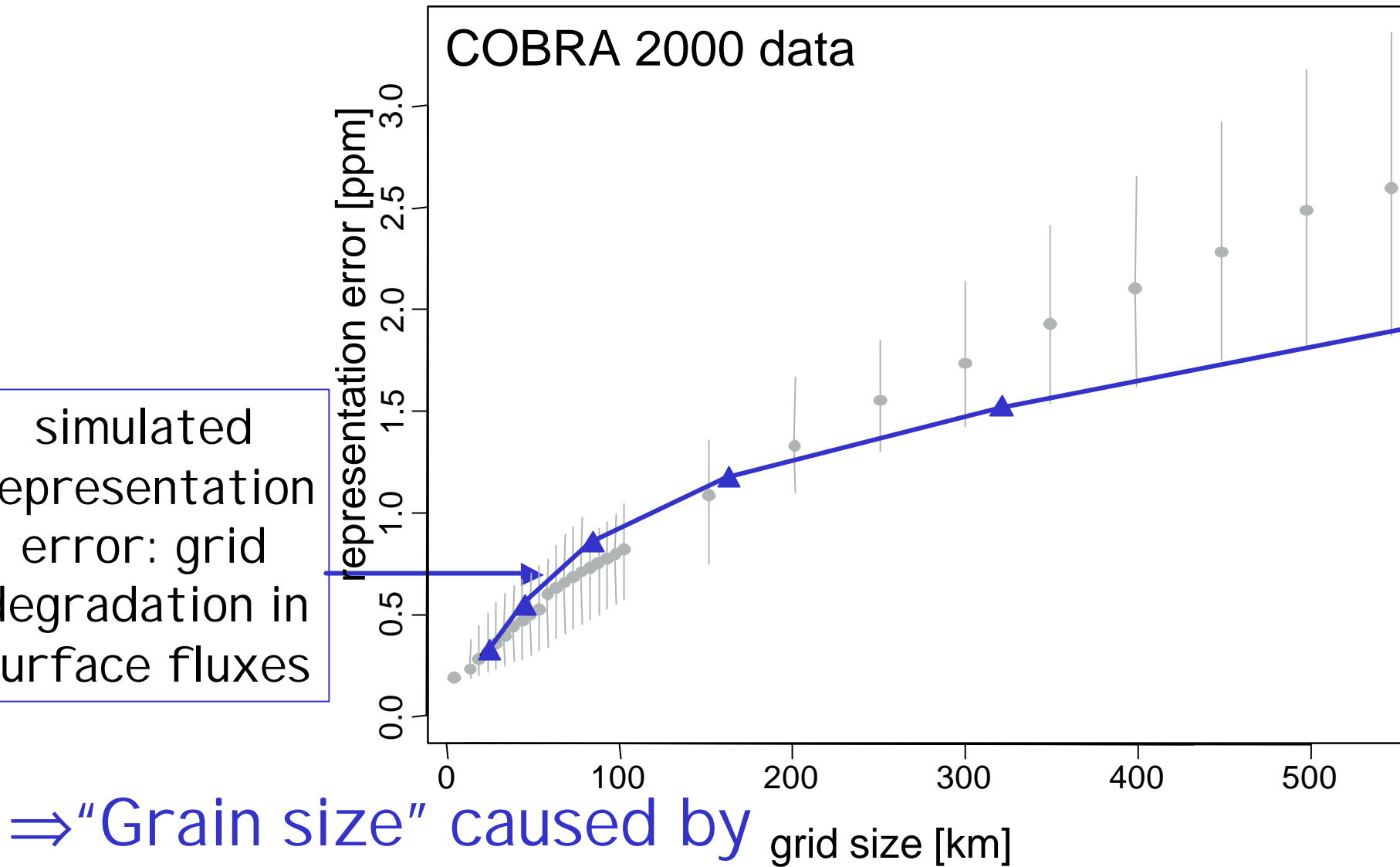
PRODUCT



[Gerbic et al., 2002]



# "Grain size" of atmospheric CO<sub>2</sub>: Representation error

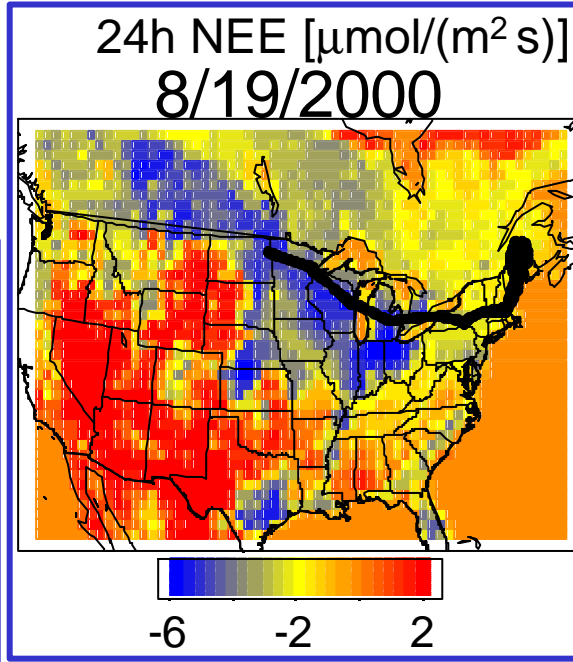
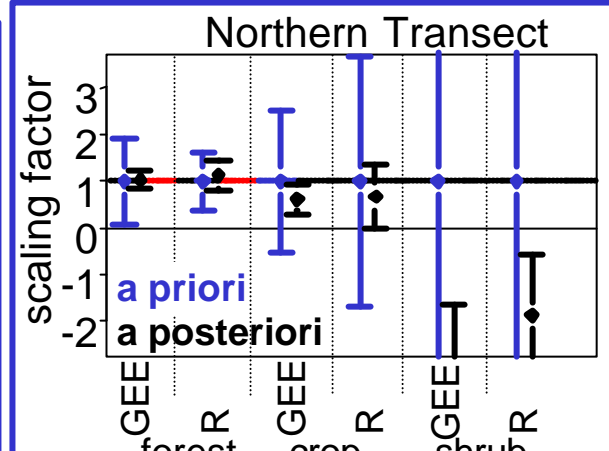
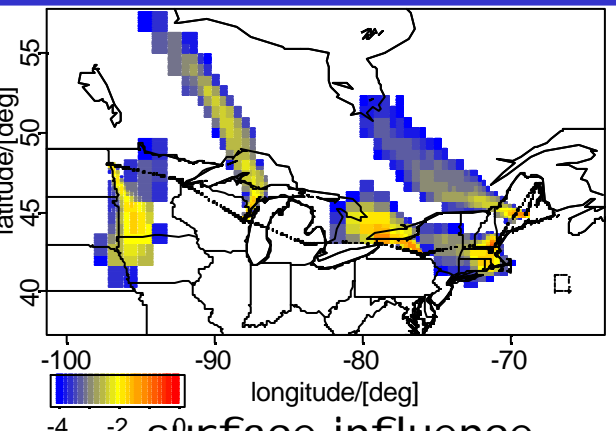
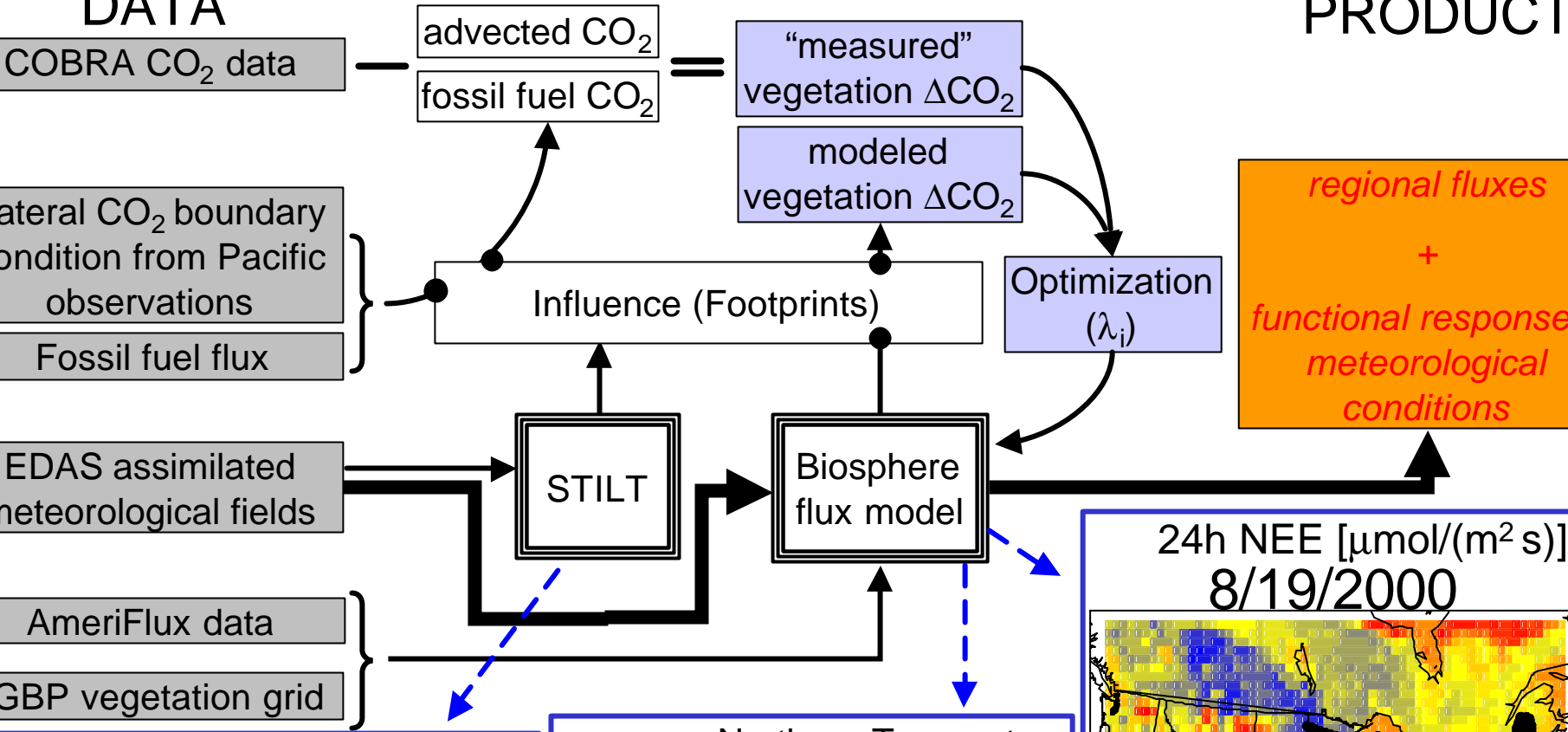


⇒ "Grain size" caused by patterns in surface flux

# Receptor Oriented Atmospheric Model "ROAM"

DATA

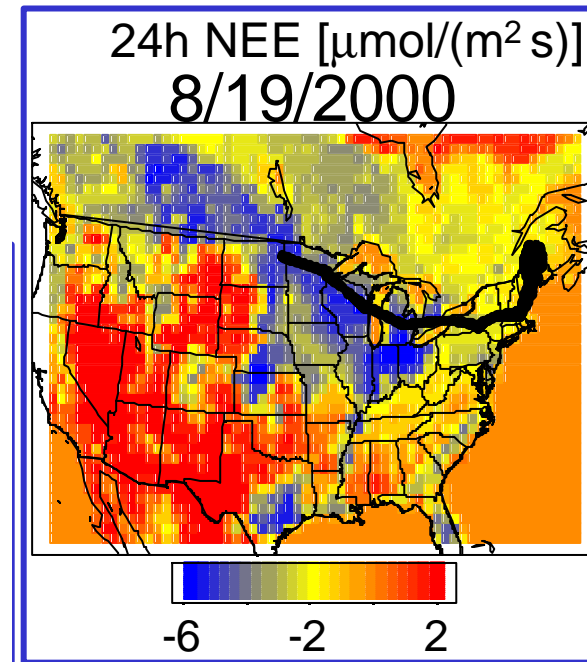
PRODUCT



[Gerbic et al. 2002]

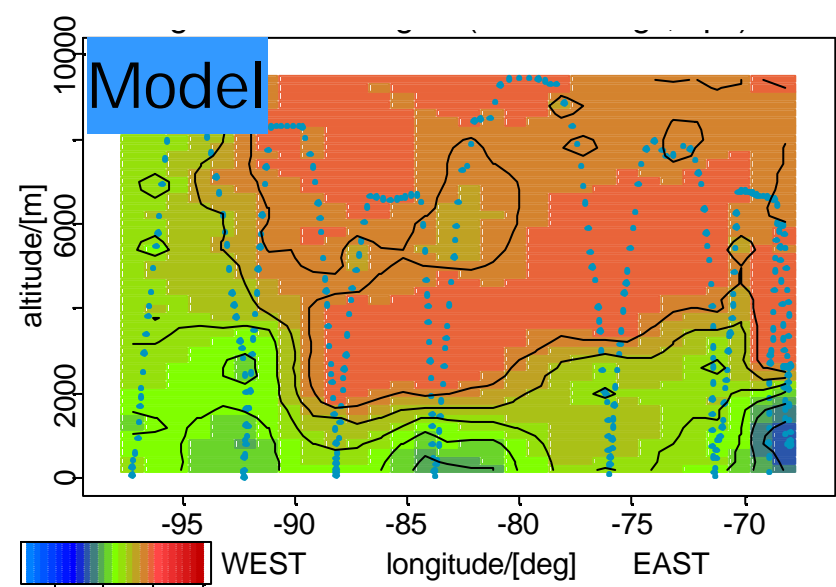
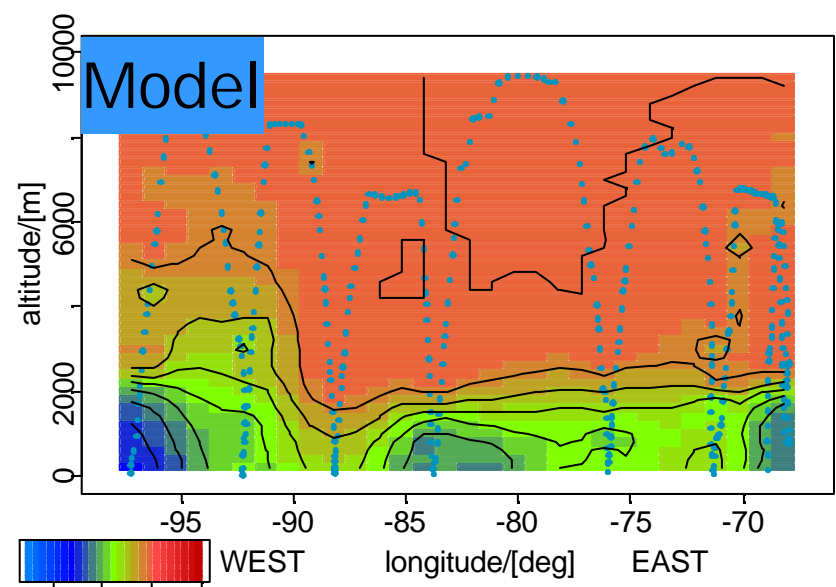
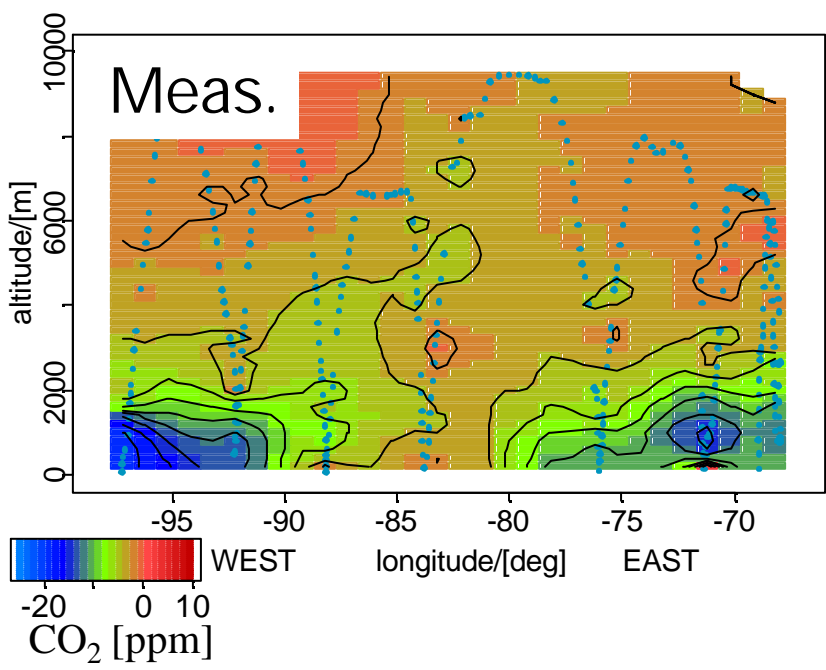
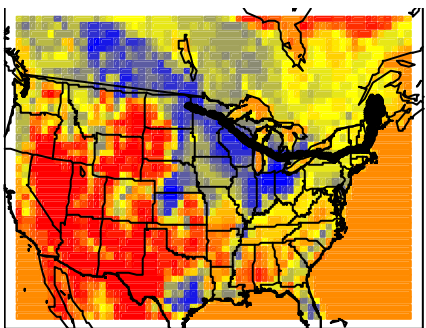
# Receptor Oriented Atmospheric Model "ROAM"

- Does it work? Is this realistic?
- Compare spatial tracer distribution with observations to validate (or falsify)



# Large-scale biospheric CO<sub>2</sub> distribution

COBRA 2000  
northern survey

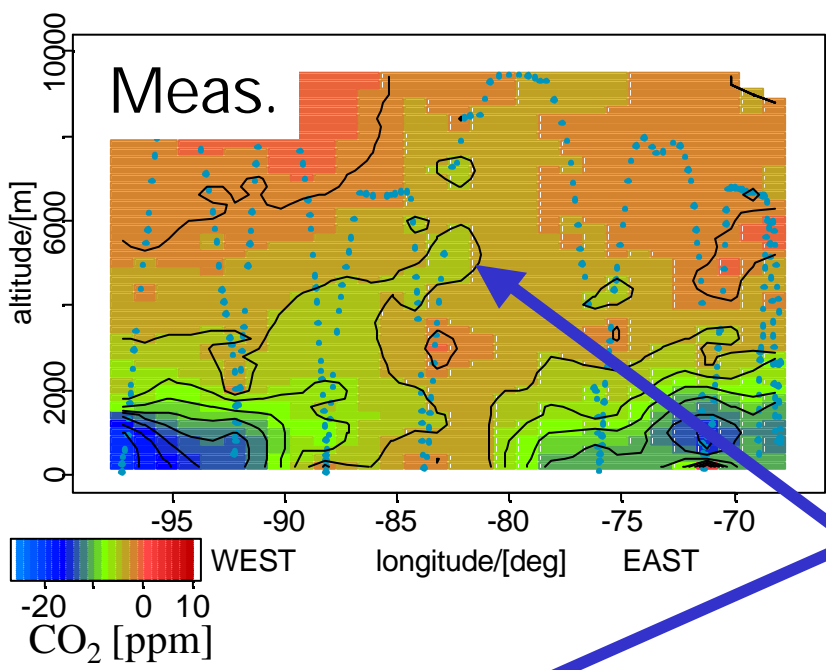


no subar copy

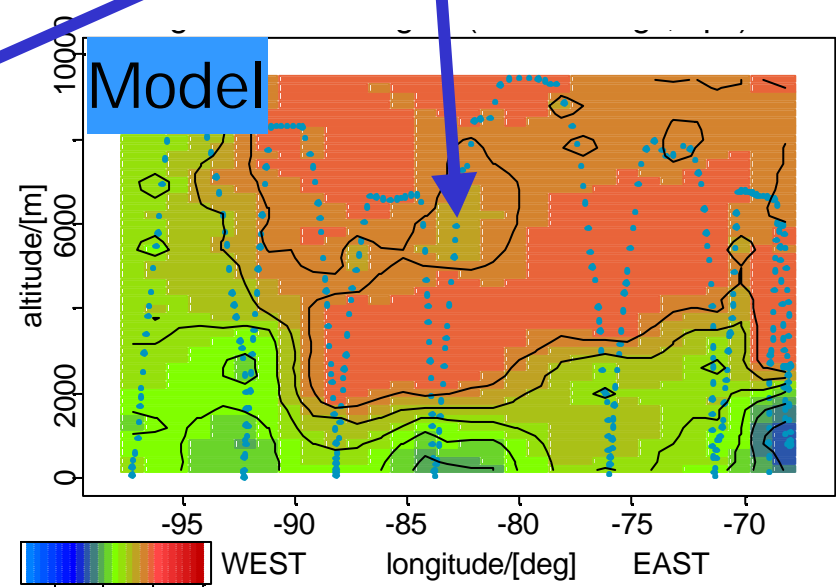
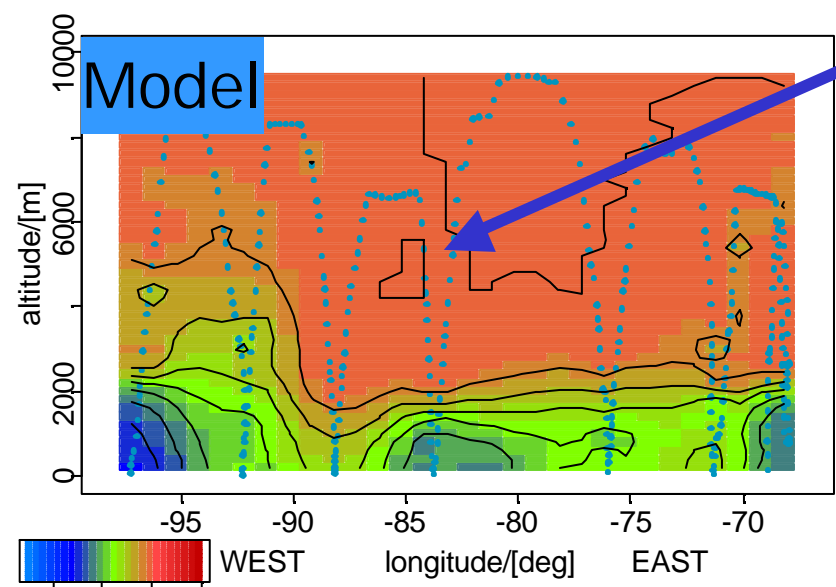
excessive convection

# Large-scale biogenic CO<sub>2</sub> distribution

COBRA 2000  
northern survey

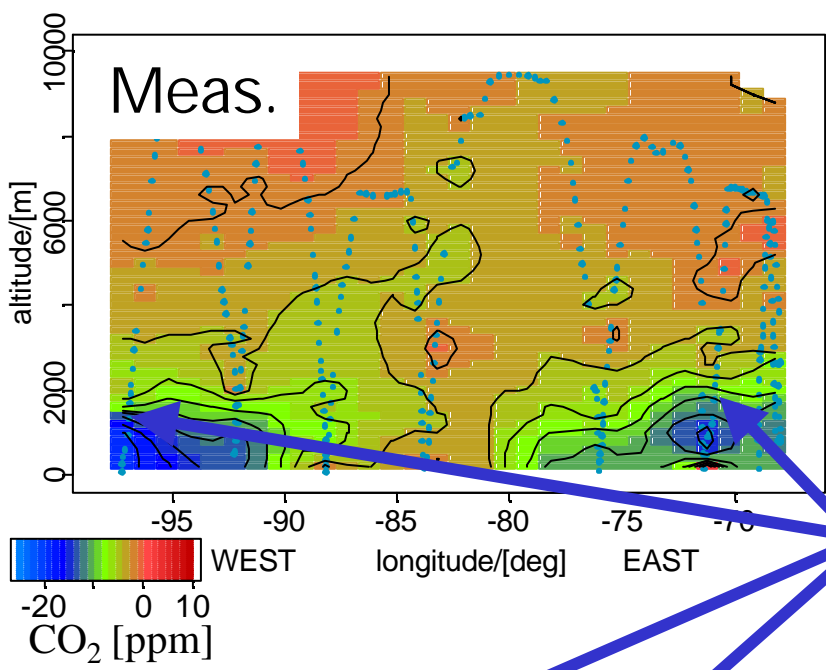


vertical mixing,  
convection...

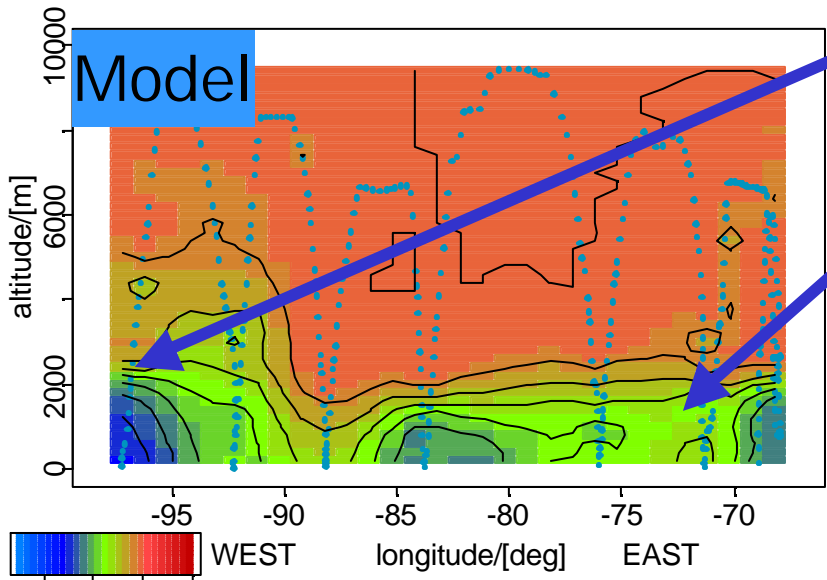


# Large-scale biogenic CO<sub>2</sub> distribution

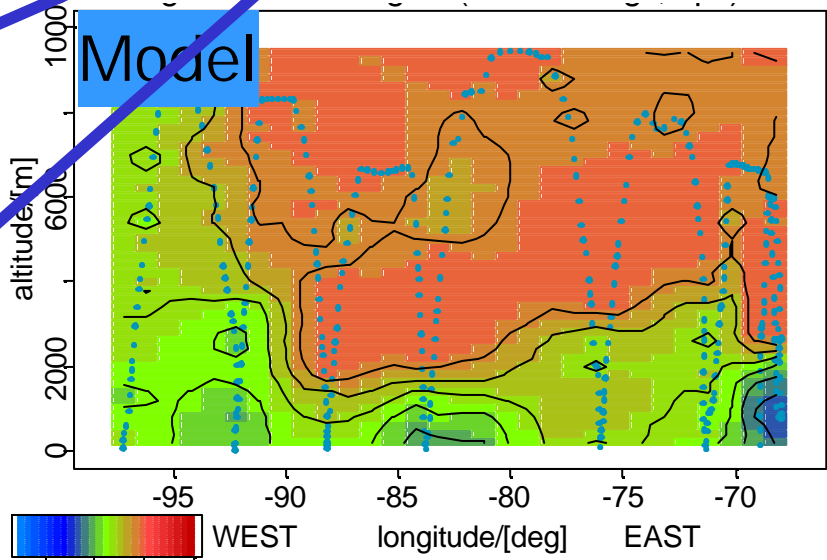
COBRA 2000 northern survey



vertical mixing, "Z<sub>i</sub>" ...



no subar convy



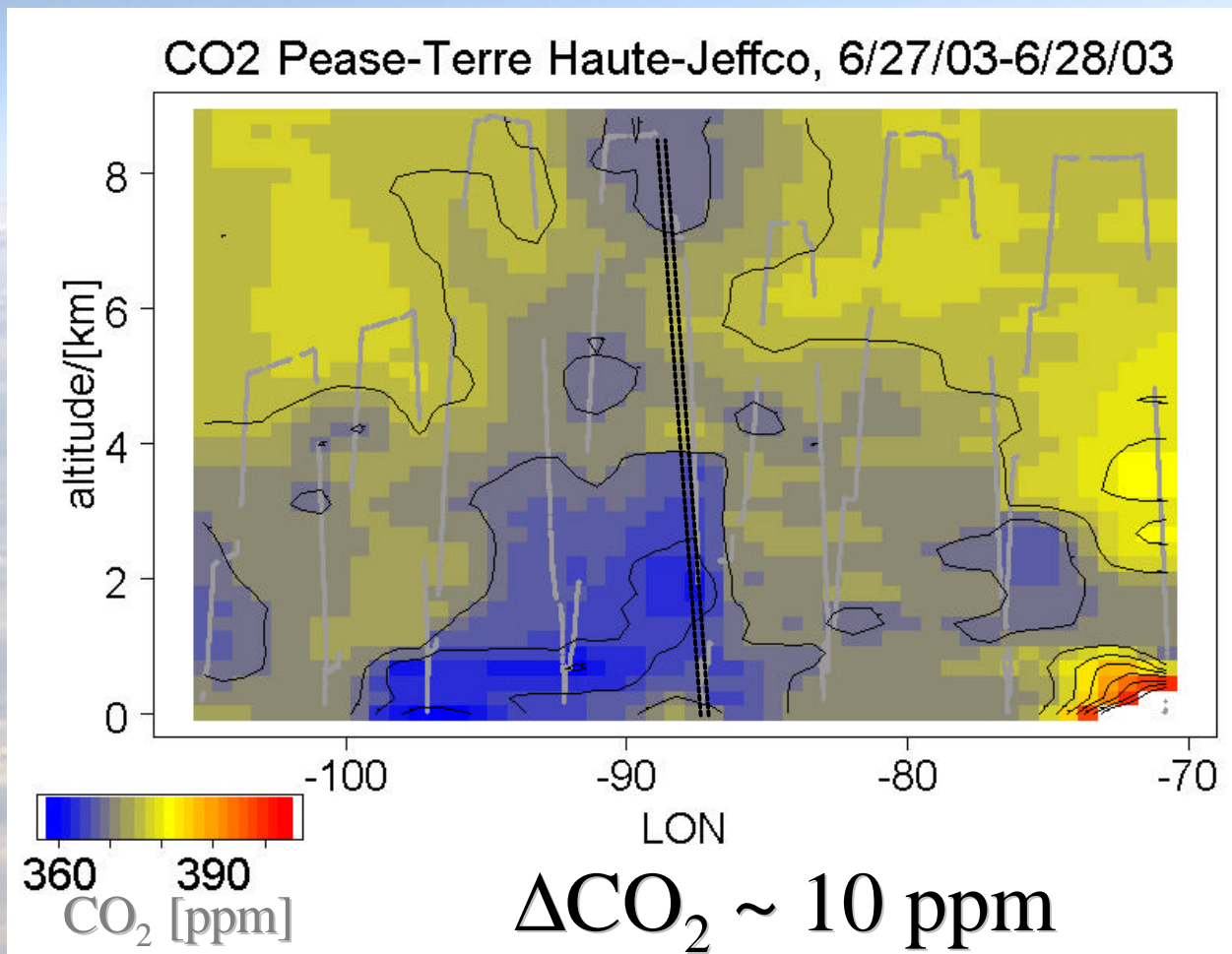
excessive convection



# Constraints on Convective Fluxes

COBRA-2003, June 2003

CO<sub>2</sub> Measurements over the U.S.



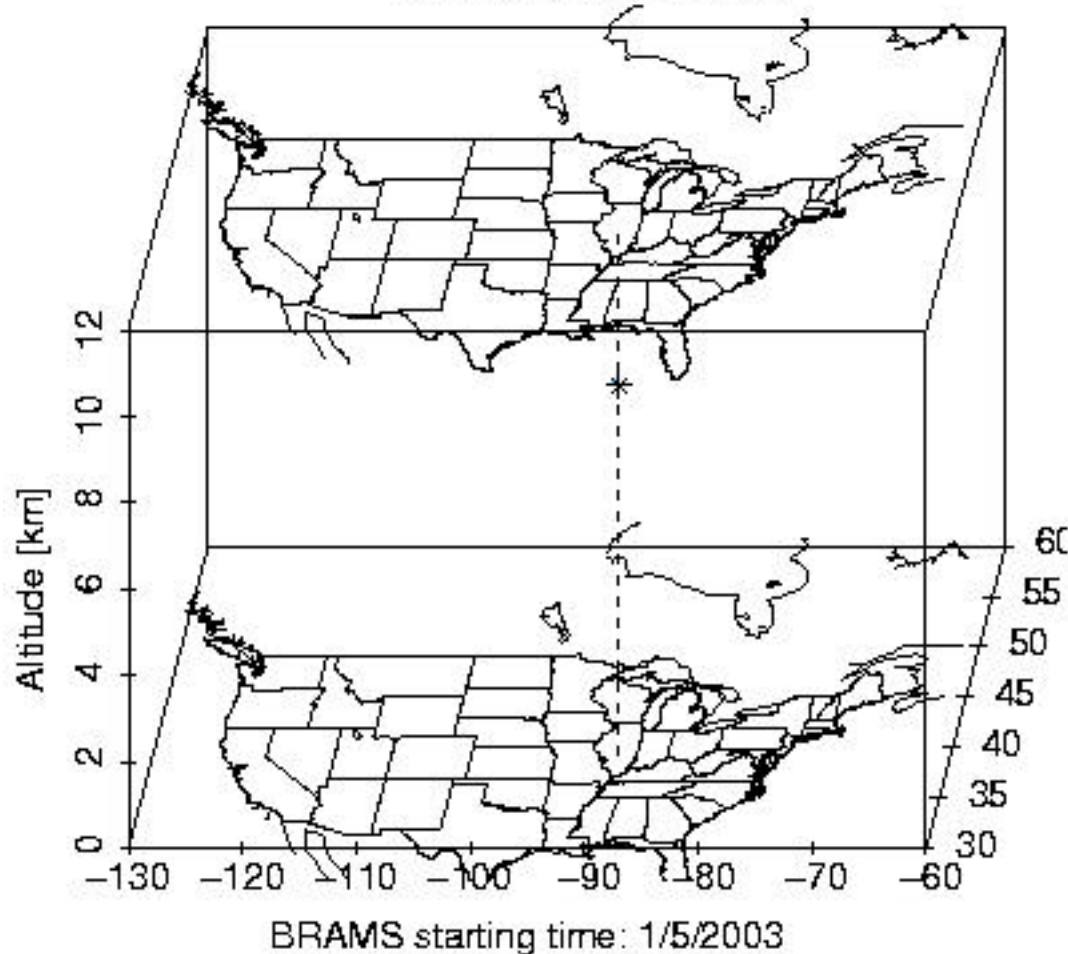


# STILT-BRAMS: convection

Stochastic Time Inverted  
Lagrangian Transport Model  
coupled to  
Brazilian Regional  
Atmospheric Modeling  
System

backward particle motion,  
receptor at top of  
COBRA profile  
6/28/03, ~18:00

BRAMS-STILT modeled trajectories  
Arrival time: 2003x06x28x17x39.30Nx089.10Wx08550  
hours before arrival: 1

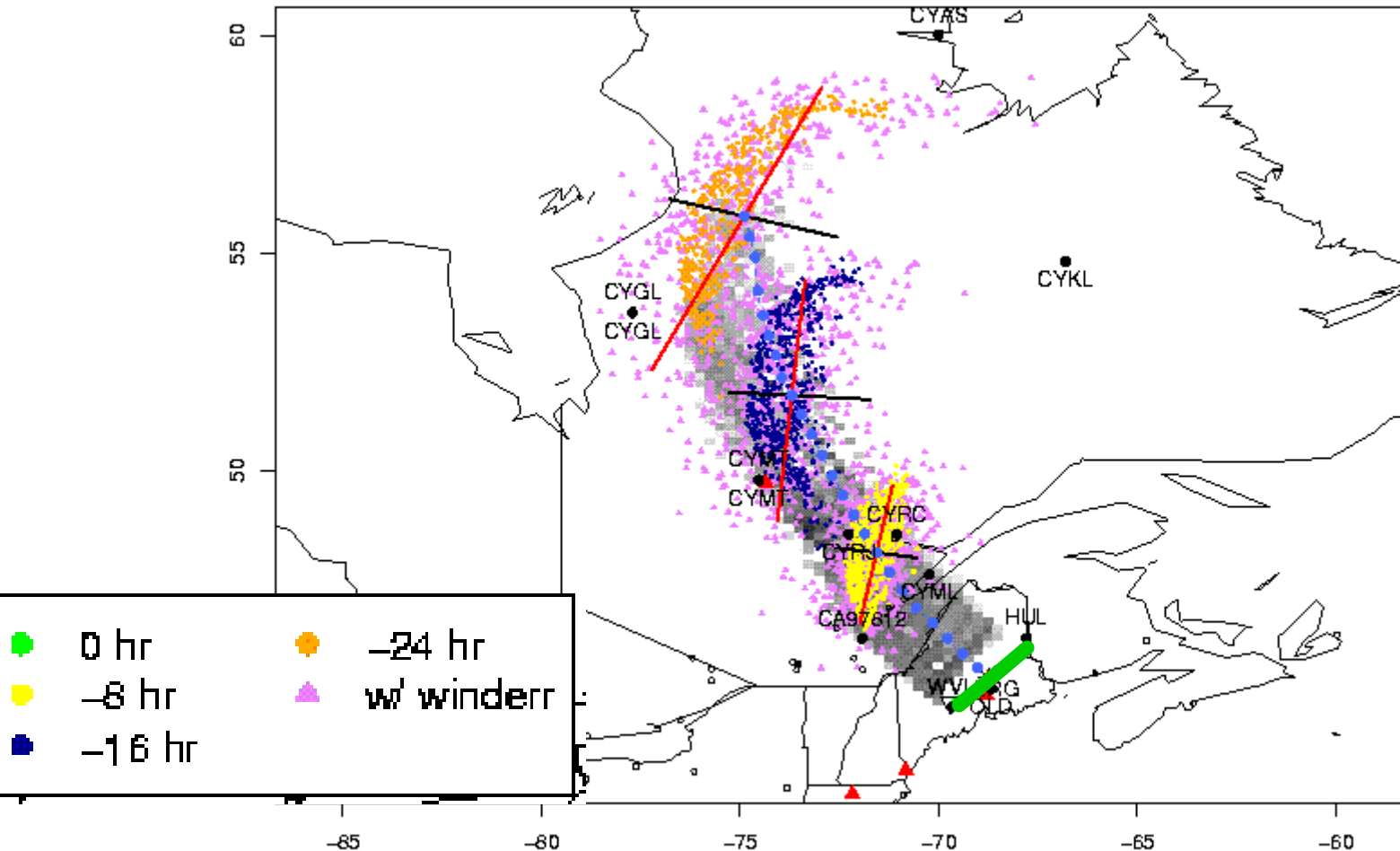


+ fluxes + boundary condition

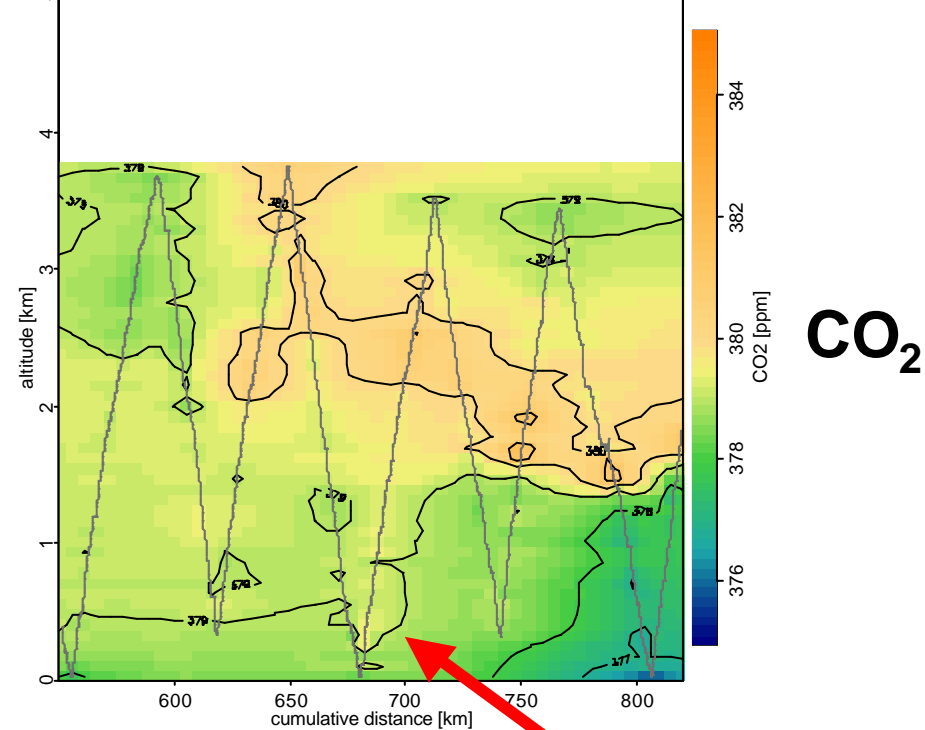
# Forecasting of air mass history: Lagrange Experiment

ETA12 061120

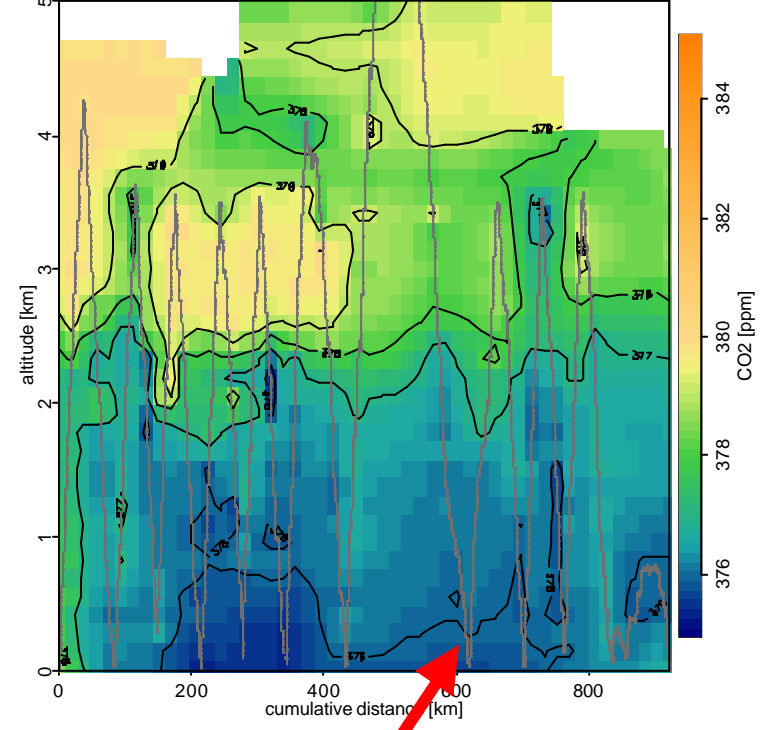
file used:061018; receptor xsec length: 200 km; winderr: 3.81 m/s



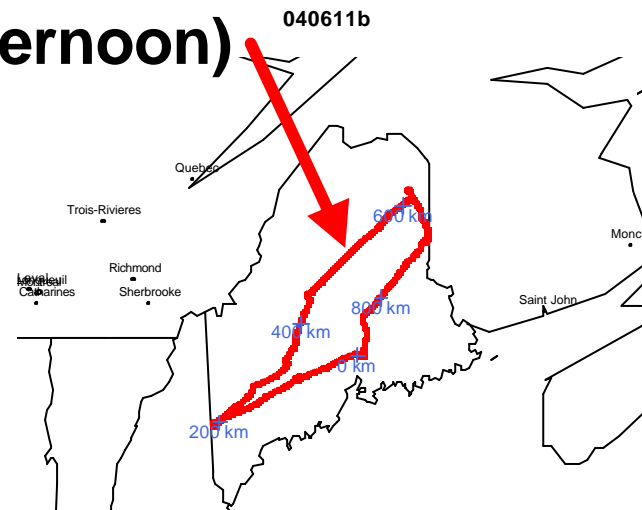
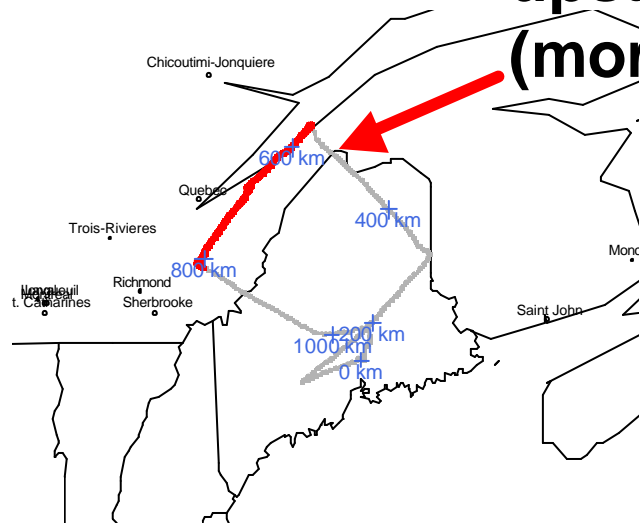
STILT-ETA



**upstream  
(morning)**



**downstream  
(afternoon)**



**COBRA Maine ( June 11, 2004)**

# Concluding Remarks

Future observational network:

- More continental sites that are closer to processes
- Vertical distribution:
  - CMDL: rental aircraft
  - IAGOS: Integration of routine Aircraft measurements into a Global Observing System
  - remote sensing (ground based and satellite based)

Airborne intensive data can provide

- Constraints on fluxes / terrestrial processes
  - Tight constraint on regional scale: Lagrangian experiments
- "Testbed" for a modeling framework
  - we can only learn from discrepancies models vs. measurements (mixing, convective redistribution)

How we can learn: Interplay between modeling and experiment

- Model => Measurement: utilize the little flexibility we have in the experiments (many constraints by sensors/physics, platforms)
- Measurement => Model (example: grain size): models are more flexible than we often think, need to design models to match measurements (thus they become falsifiable)