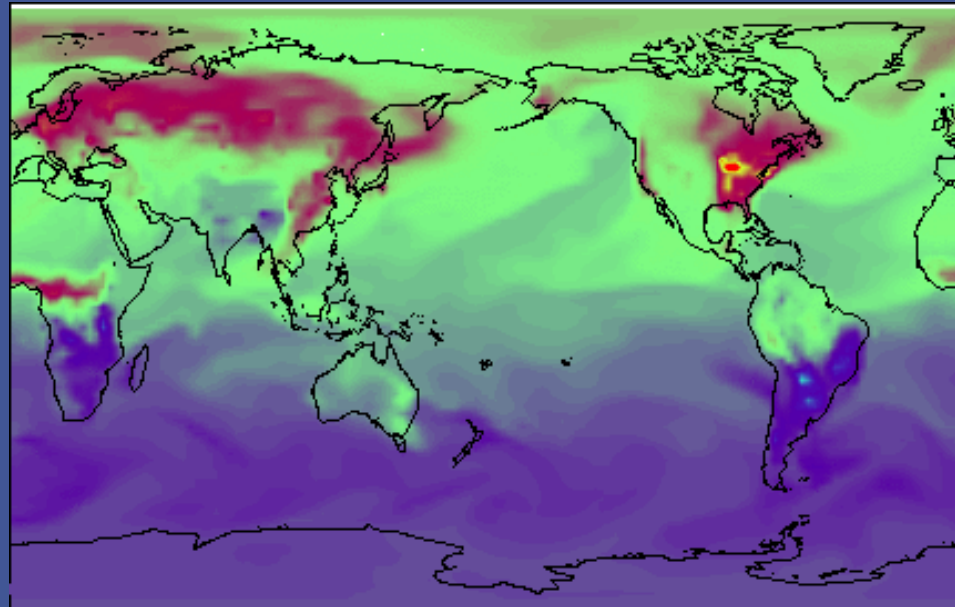


# Offline transport models and the carbon cycle



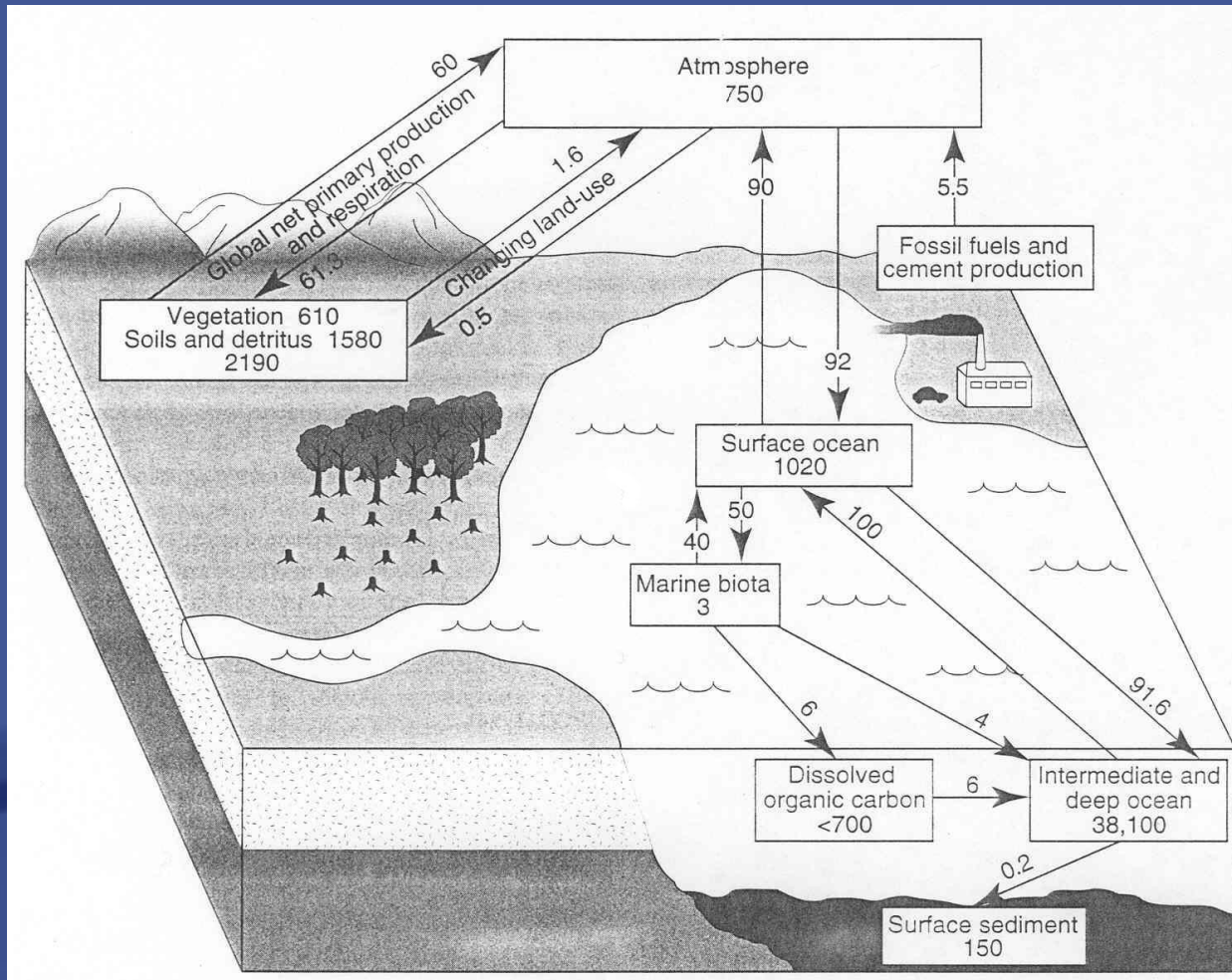
# CO<sub>2</sub> Concentrations at the Surface



# Scientific Questions

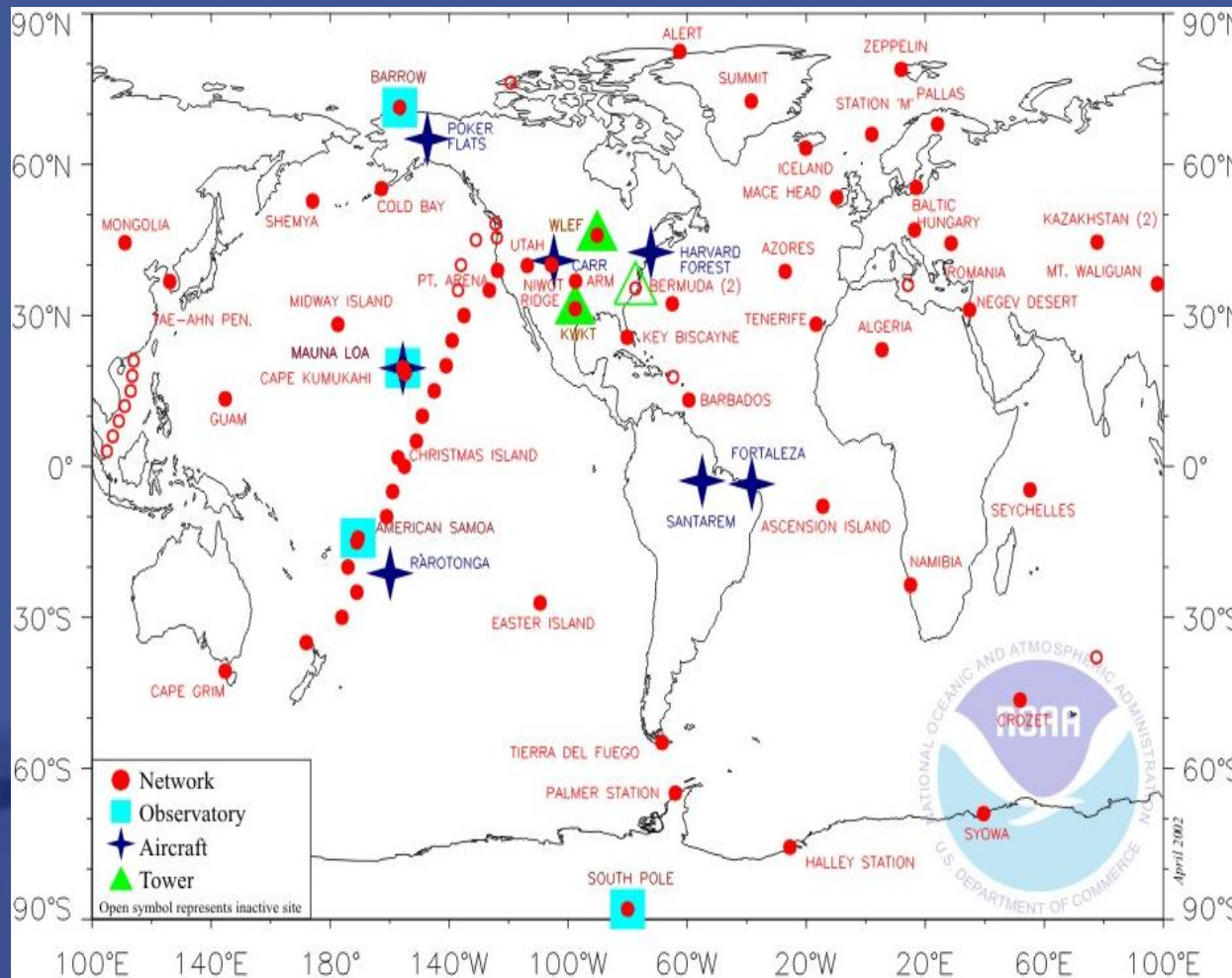
- Examine how atmospheric carbon concentrations have changed due to anthropogenic activities.
- Understand what and how the carbon budget affects the climate and thus the environment.

# Schematic CO<sub>2</sub> Cycle



Observation network  
Concentrations  
Fluxes

# Schematic CO<sub>2</sub> Cycle



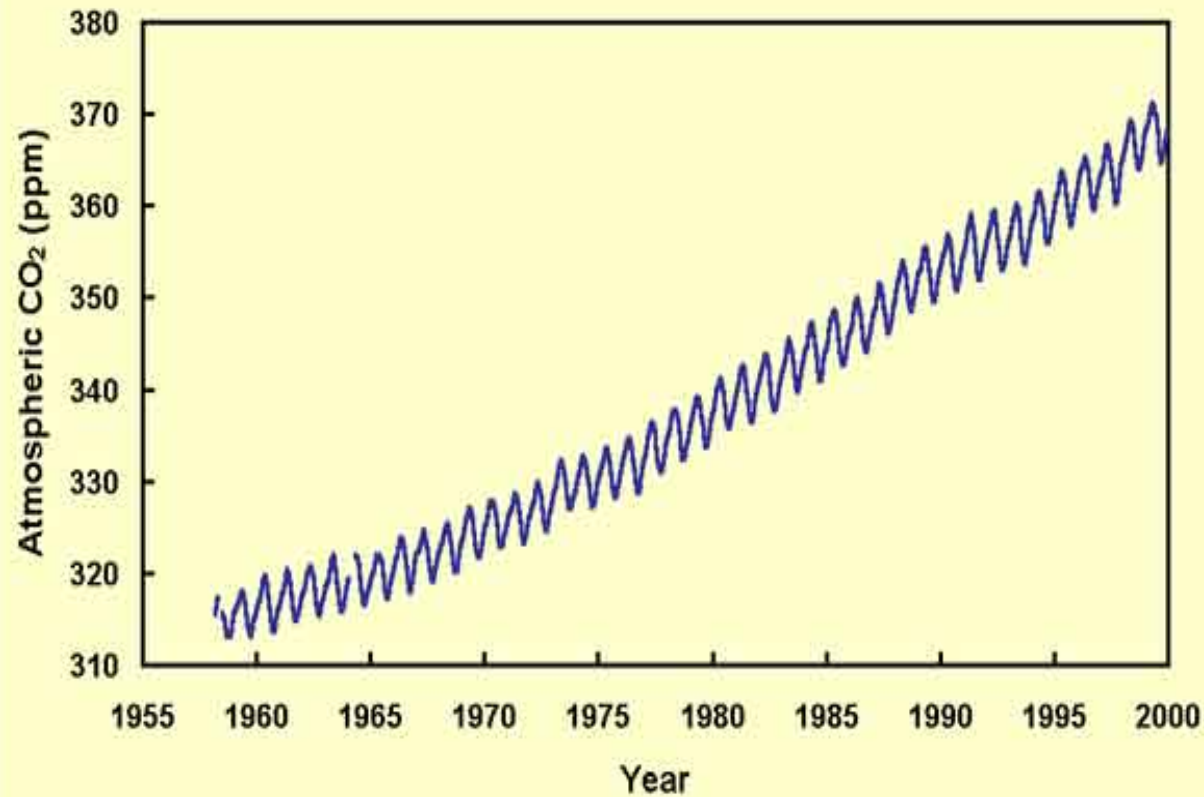
Observation network

Concentrations

Fluxes

# Schematic CO<sub>2</sub> Cycle

Atmospheric CO<sub>2</sub> Concentration, Mauna Loa, HI (1958-1999)

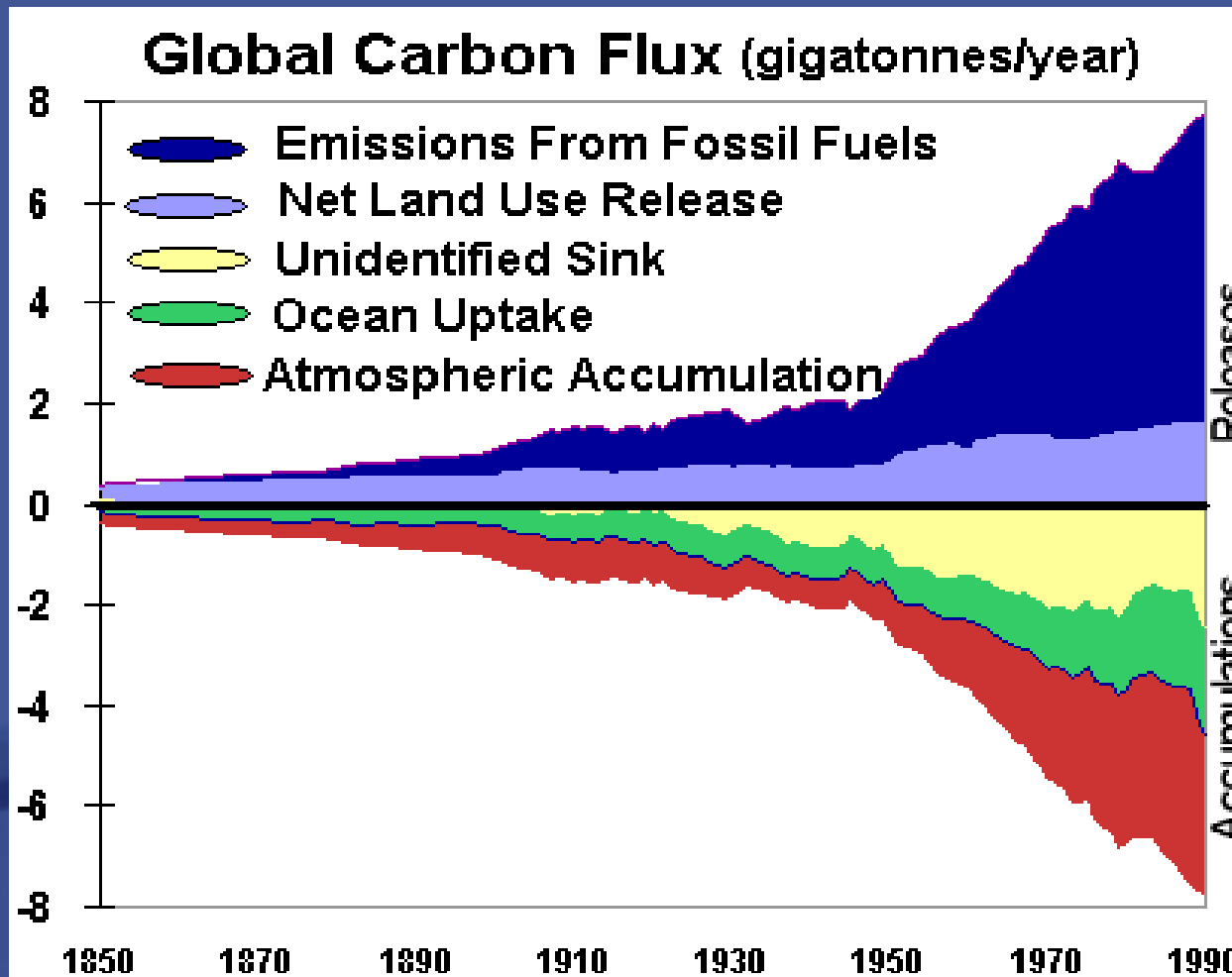


Observation network

Concentrations

Fluxes

# Schematic CO<sub>2</sub> Cycle



Observation network

Concentrations

Fluxes

# Transport Model

A (parametrized chemistry) transport model numerically solves the **constituent continuity equation** over the entire globe.



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- Aerosols
- Liquid water

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$$\frac{\partial x}{\partial t} = -\mathbf{v}\nabla x + P(x) - L(x)$$

$x$ : constituent

$\mathbf{v}$ : velocity

$P(x), L(x)$ : production and loss

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⇒ where do we get  $\mathbf{v}$  from?

⇒ how to we solve the equation?

# General Circulation Model

General circulation models are a **numerical representation** of the atmosphere and its phenomena over the entire Earth.

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## Numerical representation

- Physical laws of flow
  - Simplification
  - Discretization
- Conservation of momentum
  - Conservation of mass
  - Equation of state for ideal gases
  - Conservation of energy
  - Conservation equation for water mass

# General Circulation Model

General circulation models are a **numerical representation** of the atmosphere and its phenomena over the entire Earth.

## Numerical representation

- Physical laws of flow
- Simplification
- Discretization

$$\frac{d\mathbf{v}}{dt} = -\alpha \nabla p - \nabla \phi + \mathbf{F} - 2\boldsymbol{\Omega} \times \mathbf{v}$$

$$\frac{\partial \rho}{\partial t} = -\nabla \cdot (\rho \mathbf{v})$$

$$p\alpha = RT$$

$$Q = C_p \frac{dT}{dt} - \alpha \frac{dp}{dt}$$

$$\frac{\partial \rho q}{\partial t} = -\nabla \cdot (\rho \mathbf{v} q) + \rho(E - C)$$

# General Circulation Model

General circulation models are a **numerical representation** of the atmosphere and its phenomena over the entire Earth.

## Numerical representation

- Physical laws of flow
  - **Simplification**
  - Discretization
- Spherical coordinates
  - Traditional approximation
  - Hydrostatic model
  - Quasi-geostrophic model
  - Barotropic model



# General Circulation Model

General circulation models are a **numerical representation** of the atmosphere and its phenomena over the entire Earth.

## Numerical representation

- Physical laws of flow
- Simplification
- **Discretization**
  - Leapfrog
  - Implicit (semi/fully)
  - Euler (forward/backward)

# General Circulation Model

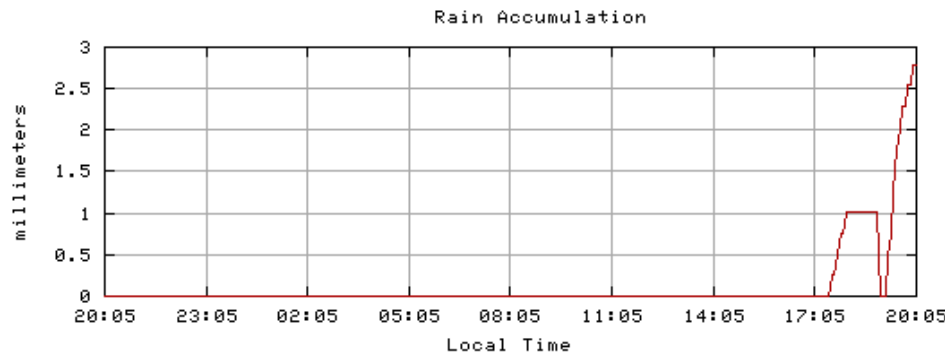
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## Numerical representation

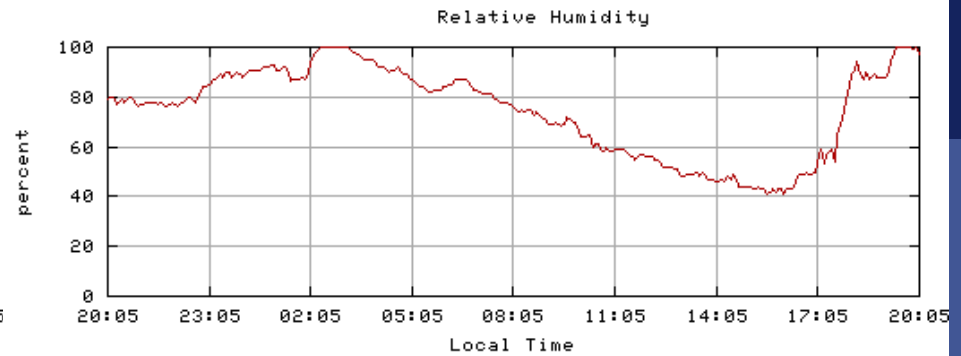
- Physical laws of flow
- Simplification
- Discretization

## Observations: Initial/boundary cond.

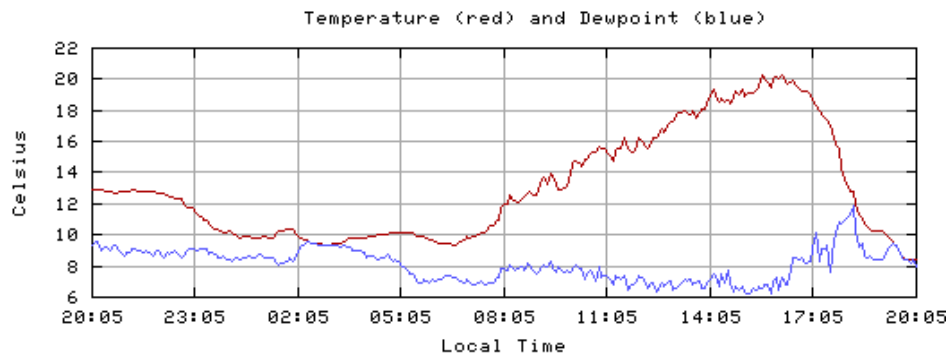
- Surface (towers, ships)
- Altitude (planes, balloons)
- Radar
- Satellites



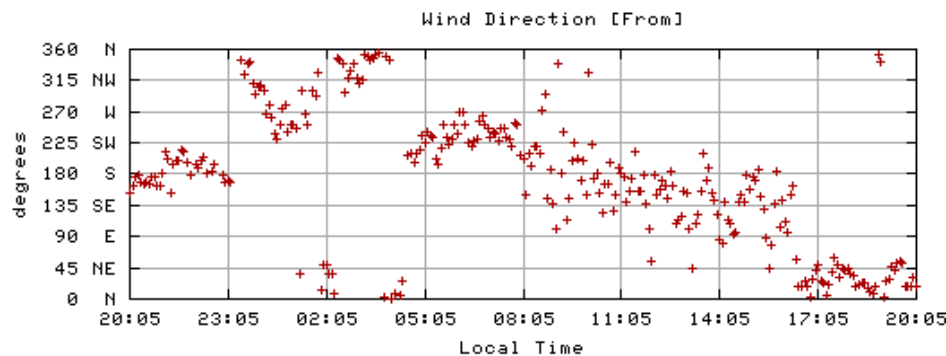
Fri Jun 06 20:08:03 2003



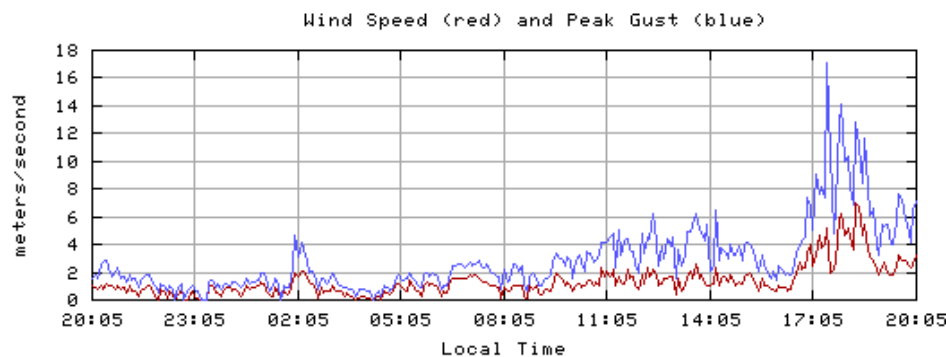
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Fri Jun 06 20:08:02 2003



Fri Jun 06 20:08:01 2003

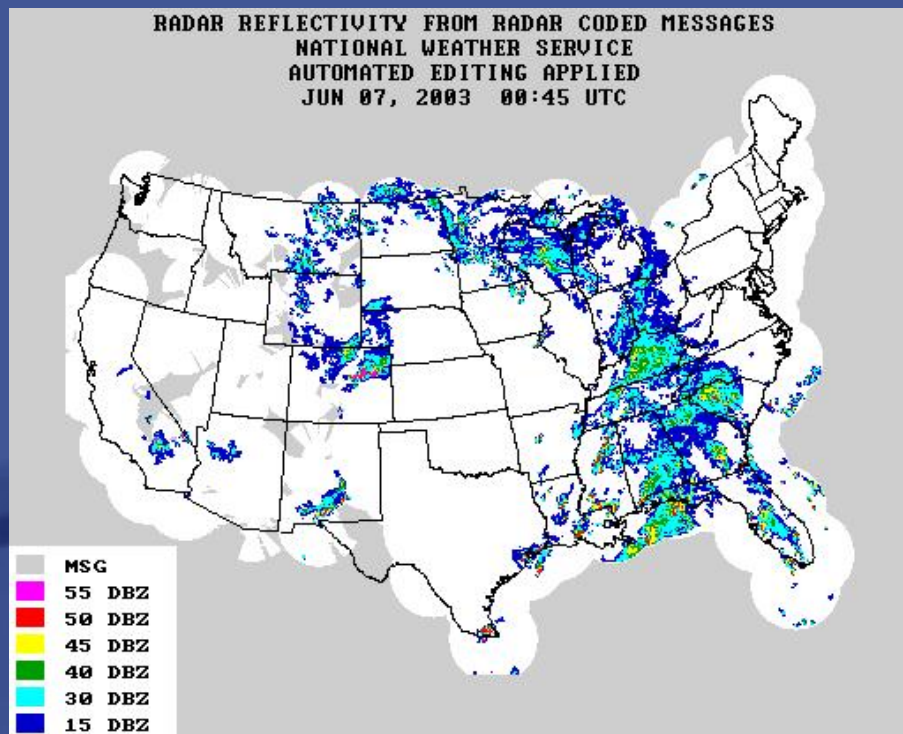
Monitors over the entire Earth.

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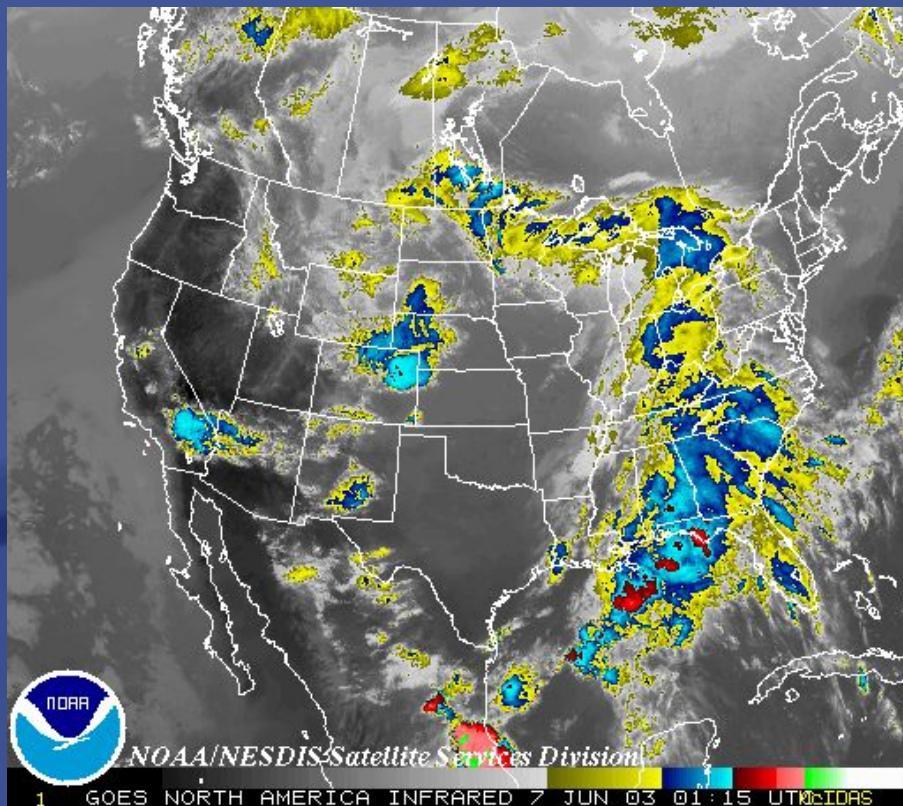


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# General Circulation Model

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## Numerical representation

- Physical laws of flow
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- Discretization

## Model output

Wind, humidity, temperature, pressure and specific density (at a “regular” grid).

## Observations: Initial/boundary cond.

- Surface (towers, ships)
- Altitude (planes, balloons)
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- Satellites

# Offline Transport Model

In an offline transport model, the dynamics of the model are supplied by archived meteorological datasets.

Dynamics = winds, temperatures, surface pressures and cloud mass fluxes.

# Modeling the CO<sub>2</sub> Cycle

We model the CO<sub>2</sub> cycle with:

- $\mathbf{z}_t$  measured concentrations at a given network,
- $\mathbf{x}_t$  actual concentrations (mixing ratios),  
are dynamically constrained to  $\mathbf{x}_{t-\delta}$  and fluxes,
- $\mathbf{u}_t$  surface CO<sub>2</sub> fluxes,  
represent the sources and sinks in the model.

We solve for the unknown fluxes  $\{\mathbf{u}_t\}$



# Inverse Method 1 2

The CO<sub>2</sub> surface flux problem is formulated as an optimality problem (simplified vector notation)

$$\begin{aligned} \min_{\{\mathbf{u}\}} & \sum_t (h(\mathbf{x}_t) - \mathbf{z}_t)^T \mathbf{W}_1 (h(\mathbf{x}_t) - \mathbf{z}_t) \\ & + \sum_t (\mathbf{u}_t - \mathbf{u}_t^{\text{con}})^T \mathbf{W}_2 (\mathbf{u}_t - \mathbf{u}_t^{\text{con}}) \\ & + (\mathbf{x}_0 - \mathbf{x}_0^{\text{con}})^T \mathbf{W}_3 (\mathbf{x}_0 - \mathbf{x}_0^{\text{con}}) \end{aligned}$$

subject to dynamical constraints

where:

$h(\cdot)$  is the measurement function,  
 $\mathbf{W}_i$  are weight matrices,  
superscript 'con' are constraints,  
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# Inverse Method 1 2

Run model forward:

with initial concentrations and forced by a priori fluxes  
↪ obtain modeled measurement history.

Run (adjoint) model backward:

forced by the weighted measurement differences  
↪ get the adjoint to the concentrations,  
↪ get new estimates for the fluxes and  
initial concentrations.

Repeat until convergence is achieved.

# Statistical Work

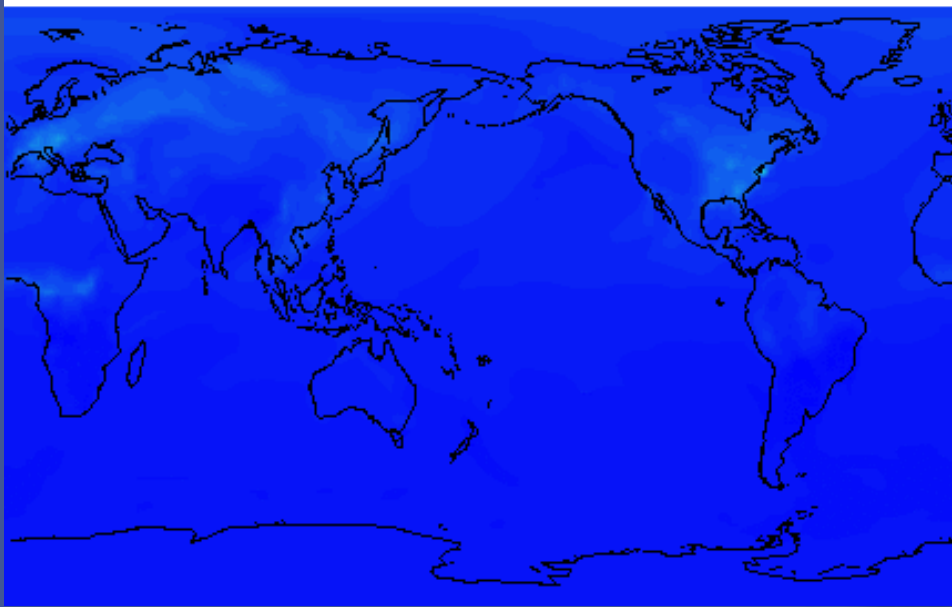
Quantify the uncertainty of carbon sources and sinks on a regional scale.

- Acquaint with transport model
- First quantitative result of a sensitivity analysis
- Setup Gibbs sampler
- ....





# Sensitivity Analysis: Simulation Setup

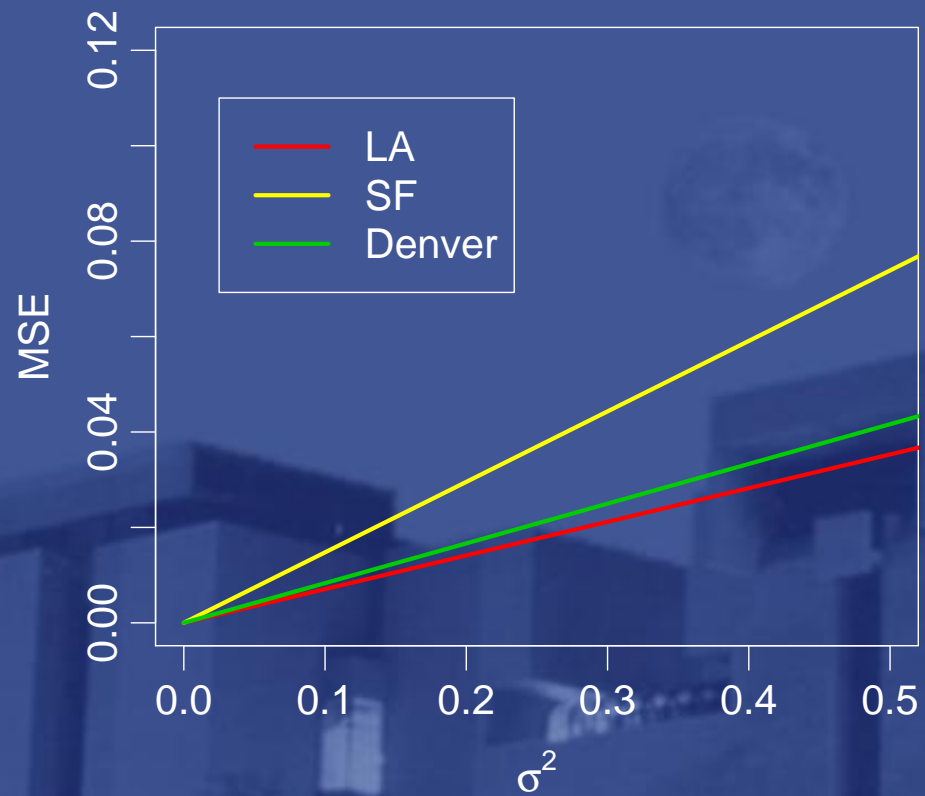


Transport model from NASA,  
Goddard Space Flight Center.  
PCTM, Lin/Rood advection code  
w/ simple vertical mixing schemes.

- Eliminate sources and sinks
- Inject CO<sub>2</sub> “Dirac” point sources at LA, SF and Denver
- Run model forward with measurement and/or transport errors
- Try to recover the intensity with regression techniques after 4 days

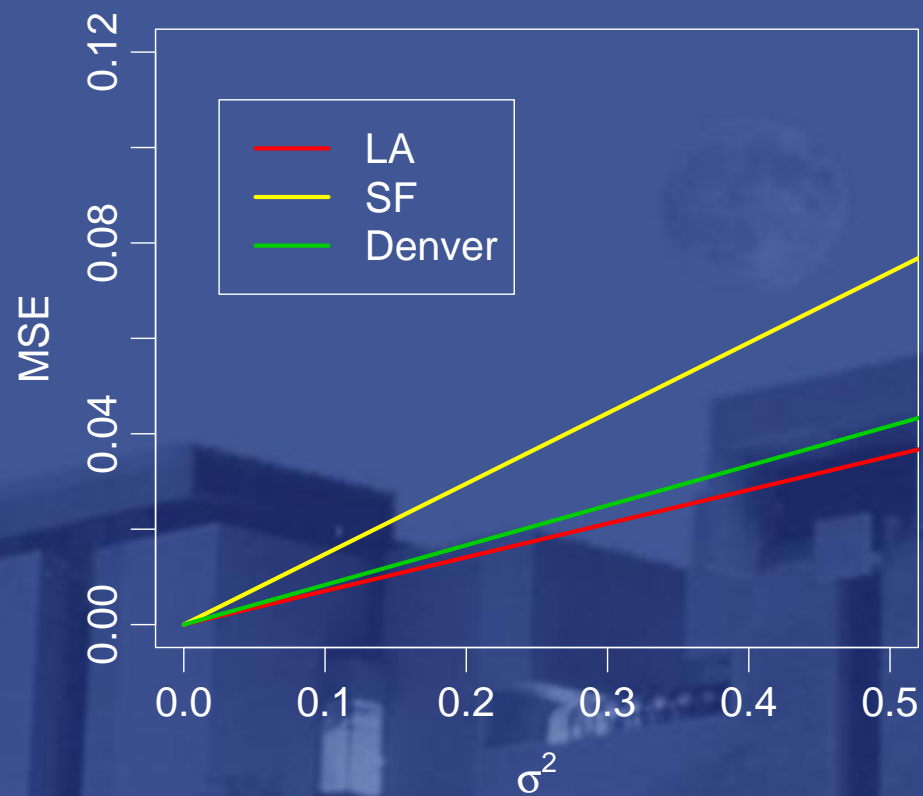
# Sensitivity Analysis: Results

Measurement error

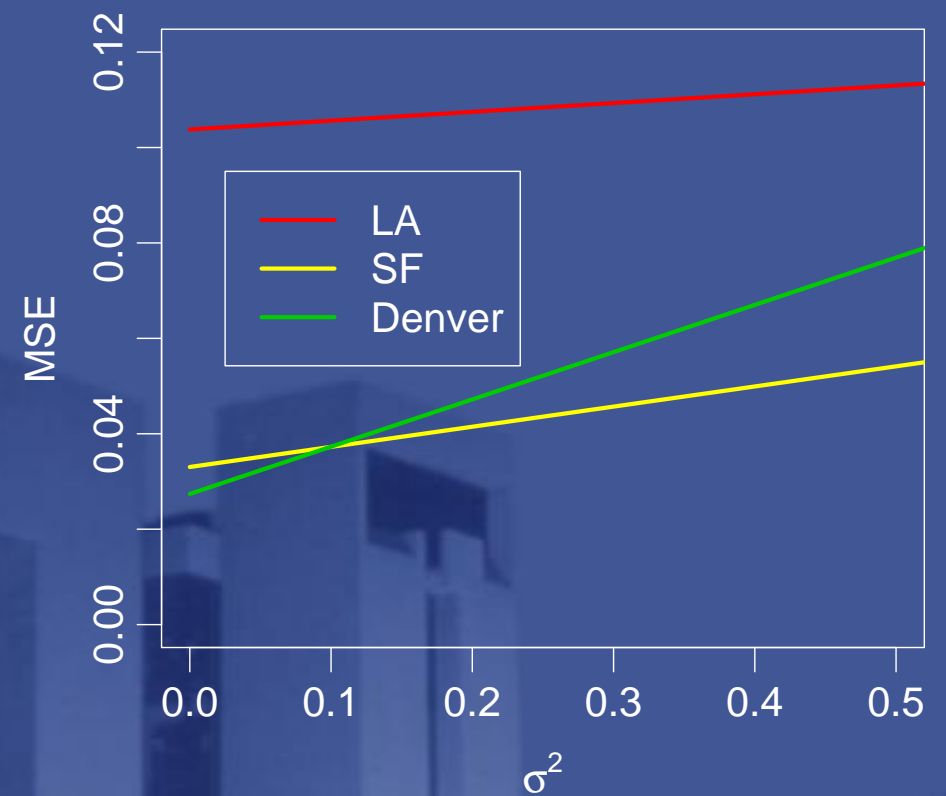


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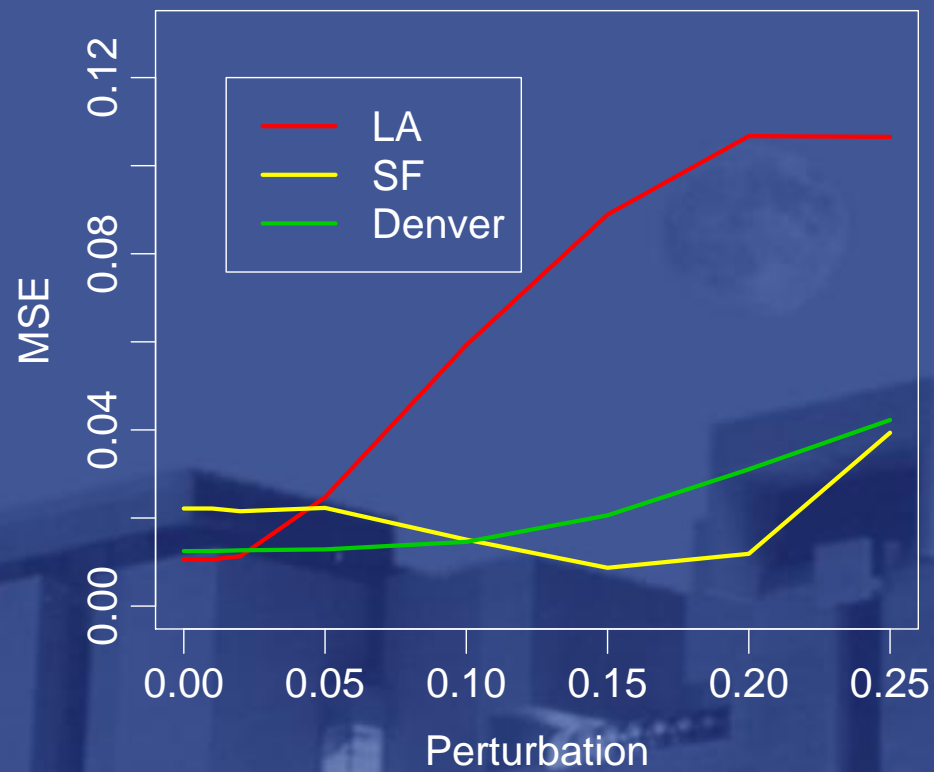
Measurement error



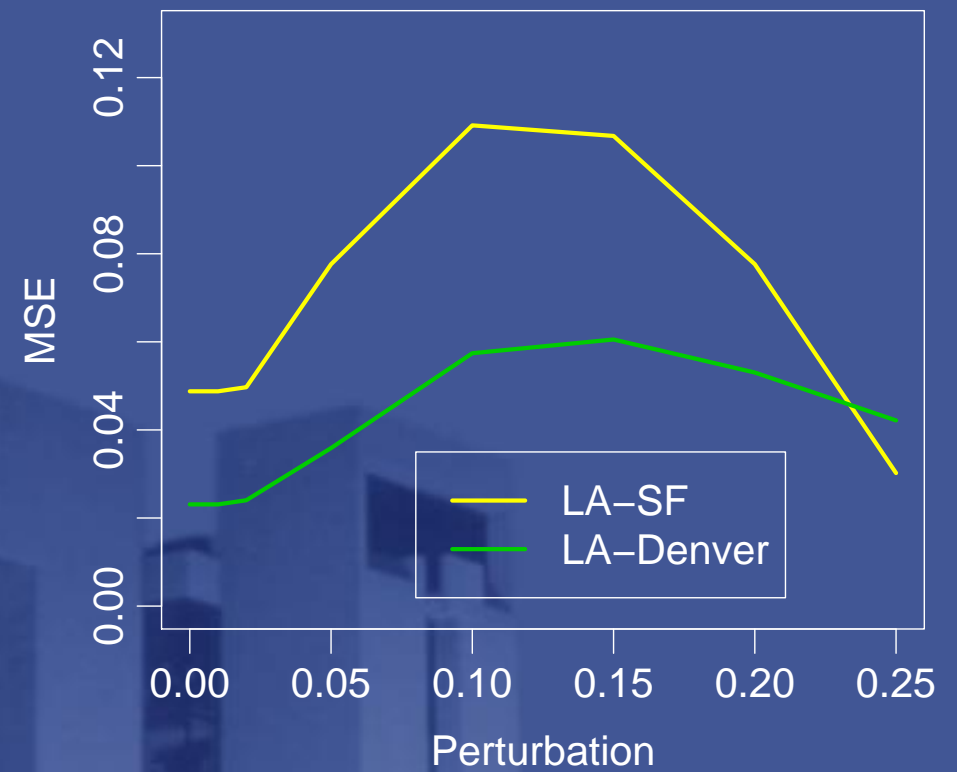
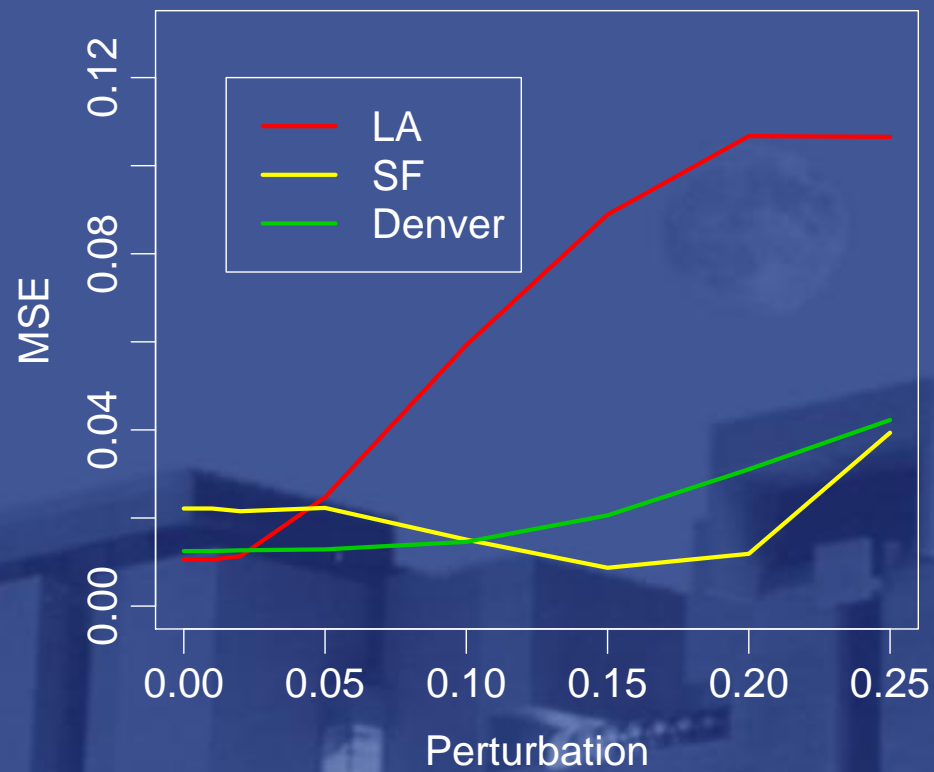
Measurement error and wind perturbation



# Sensitivity Analysis: Results



# Sensitivity Analysis: Results



# Sensitivity Analysis

Regression model

$$\mathbf{y} = \mathbf{K}\boldsymbol{\beta} + \boldsymbol{\varepsilon}$$

$$\mathbf{y} = \mathbf{K}_{\text{pert}}\boldsymbol{\beta}_{\text{pert}} + \boldsymbol{\varepsilon}$$

where

$\mathbf{y}$  concentration anomaly after 4 days

$\mathbf{K}$ ,  $\mathbf{K}_{\text{pert}}$  true and perturbed transport operator

$\boldsymbol{\varepsilon}$  measurement error

The mean squared error is

$$\text{MSE} = \sigma^2(\mathbf{K}^T\mathbf{K})^{-1}$$

$$\text{MSE} = \sigma^2(\mathbf{K}_{\text{pert}}^T\mathbf{K}_{\text{pert}})^{-1} + (\mathbf{M} - \mathbf{I})\boldsymbol{\beta}\boldsymbol{\beta}^T(\mathbf{I} - \mathbf{M}^T)$$

with  $\mathbf{M} = (\mathbf{K}_{\text{pert}}^T\mathbf{K}_{\text{pert}})^{-1}\mathbf{K}_{\text{pert}}^T\mathbf{K}$ .