

AOSC 615

Advanced Methods in Data Assimilation

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- Class:
 - Time: Tu.Th. 12:30-01:45
 - Room: 1164 PLS
 - URL: <http://aosc.umd.edu/~ide>

AOSC615: Advanced Data Assimilation

Class work

Lectures

1. Introduction
2. Mathematical Background
3. “3D” Methods
4. Uncertainty evolution
5. “4D” Methods
6. Advanced schemes
7. Special Topics

Projects

- Students will work on implementation of the data assimilation methods using the identical twin experiments

1. Introduction

1. DA Overview
 - a) Background
 - b) Schematics
 - c) Elements with Examples
 - d) Challenges
2. AOSC 615 Overview
 - a) Objectives
 - b) Project Description

- Project I

AOSC615: Advanced Data Assimilation

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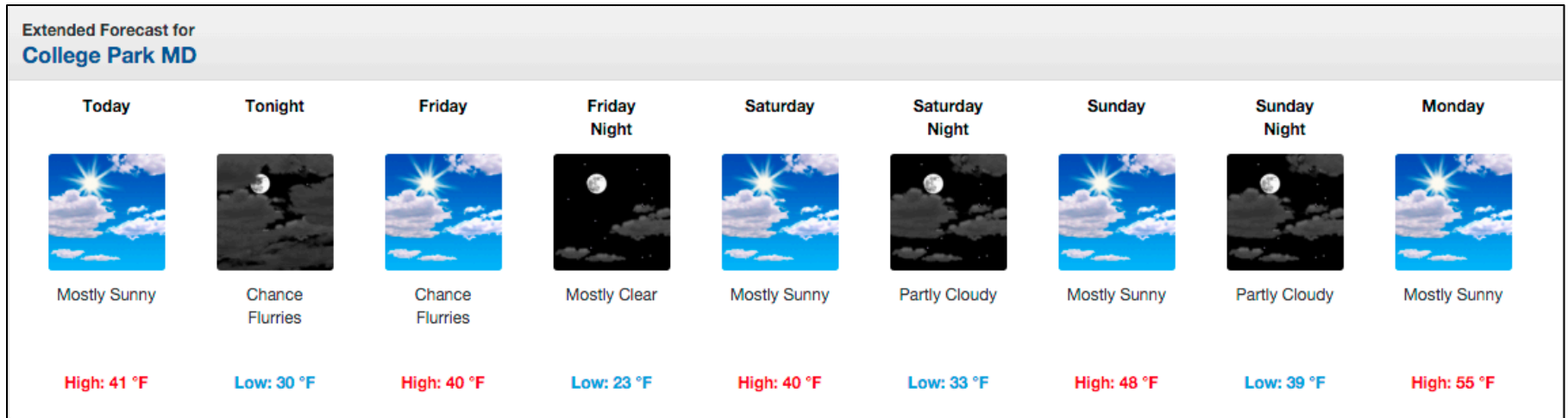
What is Data Assimilation?

- Data assimilation is the problem of **scientific estimation and prediction** (not limited to geophysical or earth systems).
- Data assimilation is an **iterative method** that attempts to combine and make best use of
 - **Model (computational)** of the physical system
 - **Observations**to achieve the “goal”.
- A familiar form is the Numerical Weather Prediction (NWP)

Data Assimilation

- Example: Weather forecast

weather.gov

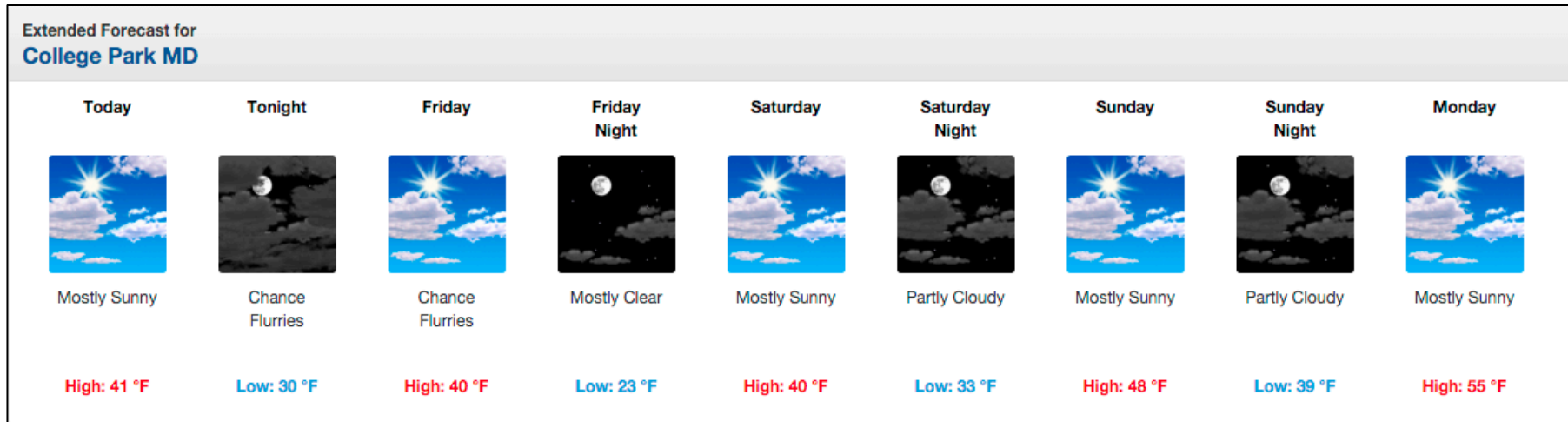


- How long do you trust?
- Why?
 - Nonlinear/complex nature of atmosphere
 - Imperfectness of the model & initial condition

→ **Uncertainty or likelihood needs to be taken into account**

Data Assimilation Example

- Most familiar form: Weather forecast

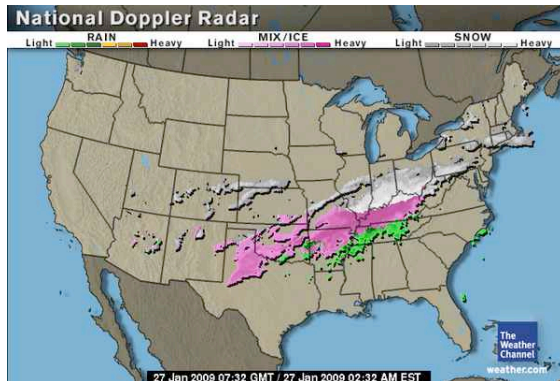


- Iterative method for initialization problem that attempts to combine
 - Model (computational) of the physical system
 - Observationsby incorporating uncertainty for (re-analysis,) monitoring and forecasting

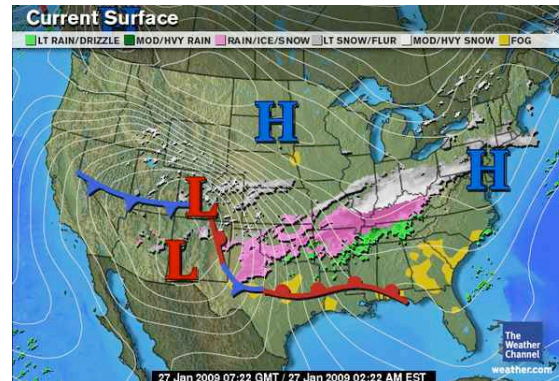
Forecasting

- Data assimilation attempts to estimate
 - States (current states, future states, or even past states)
 - Their likelihood (associated uncertainty)
- Forecast requirement
 - Current condition (initial condition)
 - Forecast model

past conditions



current condition



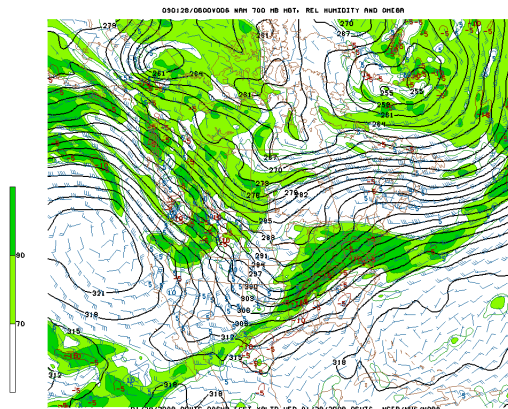
forecast



Forecasting

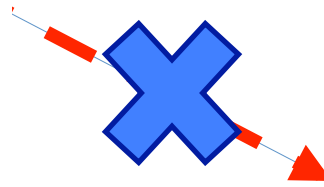
- Available:

model forecast from the past

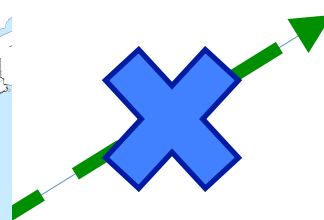
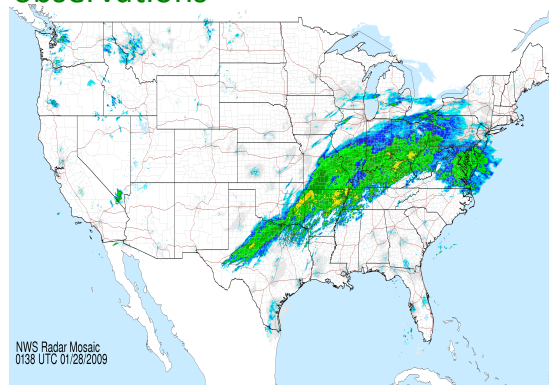


Insufficient representation of the complete current condition

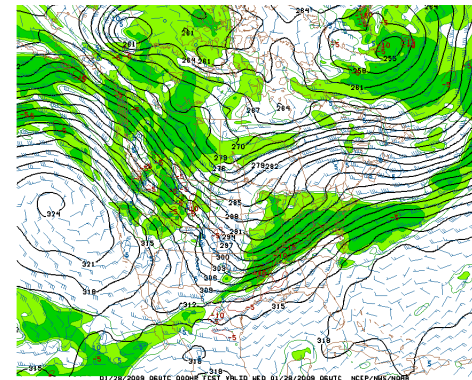
(=IC to the model forecast)
current condition



observations



Analysis (700mb)



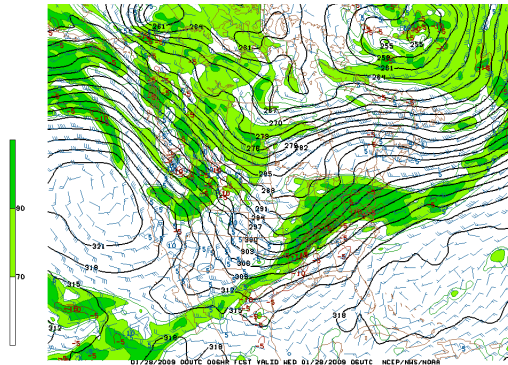
Insufficient to provide with the complete current condition
(=IC to the model forecast)

Schematics of Data Assimilation

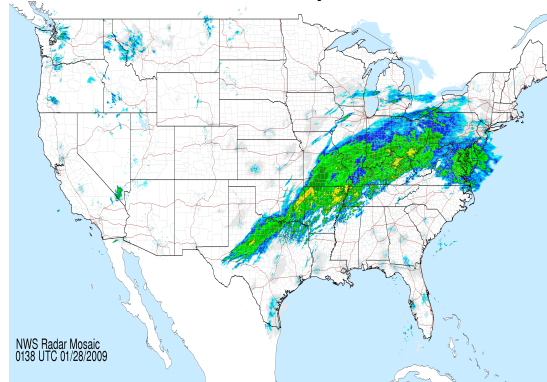
- Data assimilation is an iterative method that attempts to make best use of the **model** and the **observations**

model forecast from the past

6h forecast (700mb)



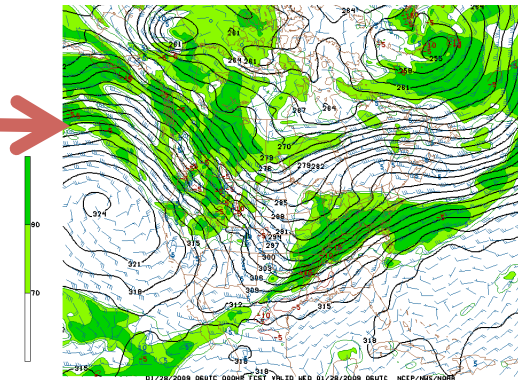
observations of the past condition



Assimilation of
data into
model forecast

current condition

Analysis (700mb)



Schematics of Data Assimilation

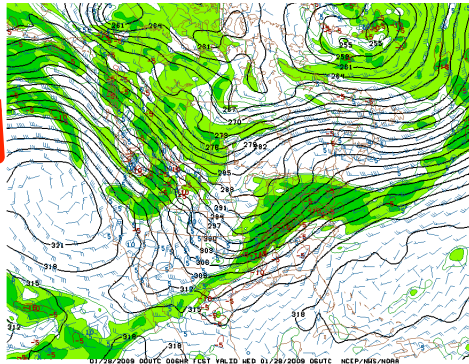
- Data assimilation is an **iterative method**

6hr Assimilation Window

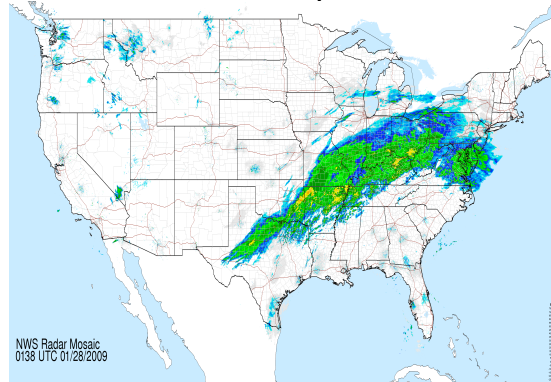
... imagine writing a data assimilation code...

model forecast from the past

6h forecast (700mb)



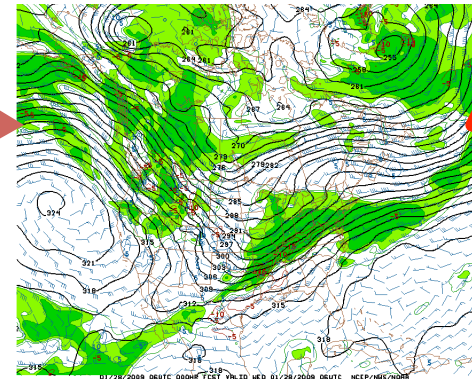
observations of the past condition



Assimilation of
data into
model forecast

current condition

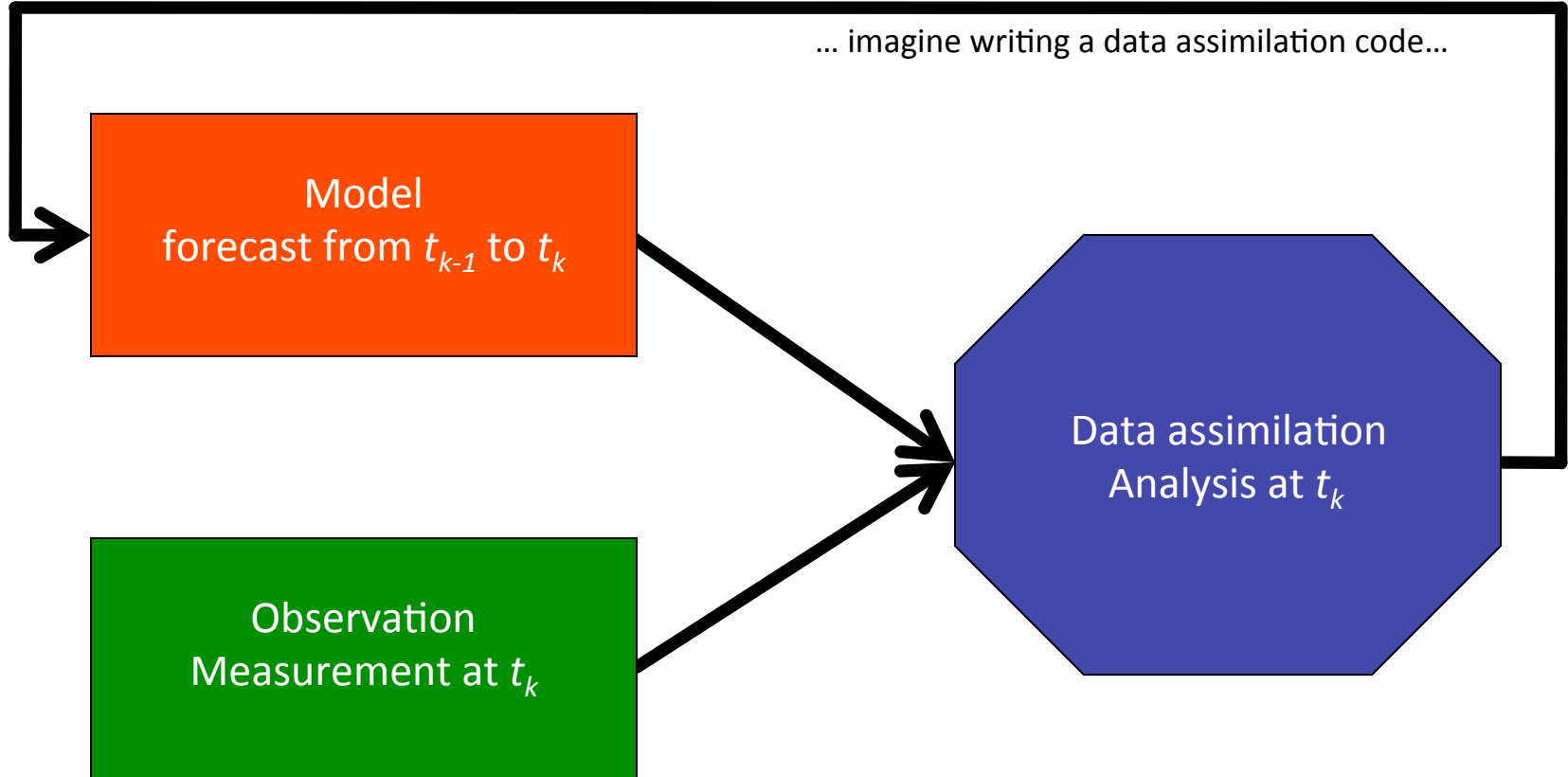
Analysis (700mb)



Data Assimilation Window

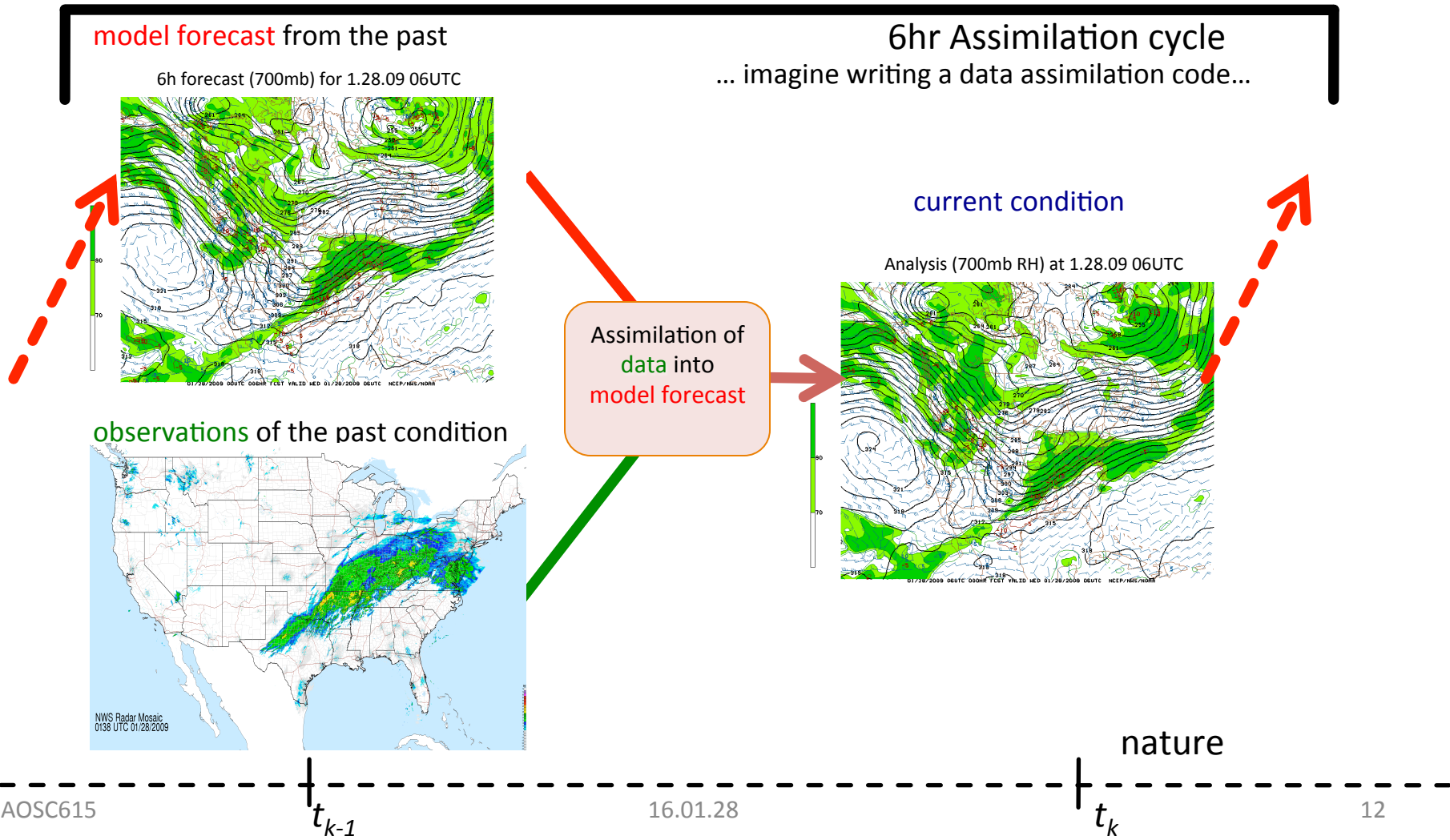
6hr Assimilation Window

... imagine writing a data assimilation code...



Schematic of Data Assimilation

- Data assimilation is an *iterative method for an initialization problem* to make best use of the **model** and the **observations** by **incorporating uncertainties**



1. Introduction

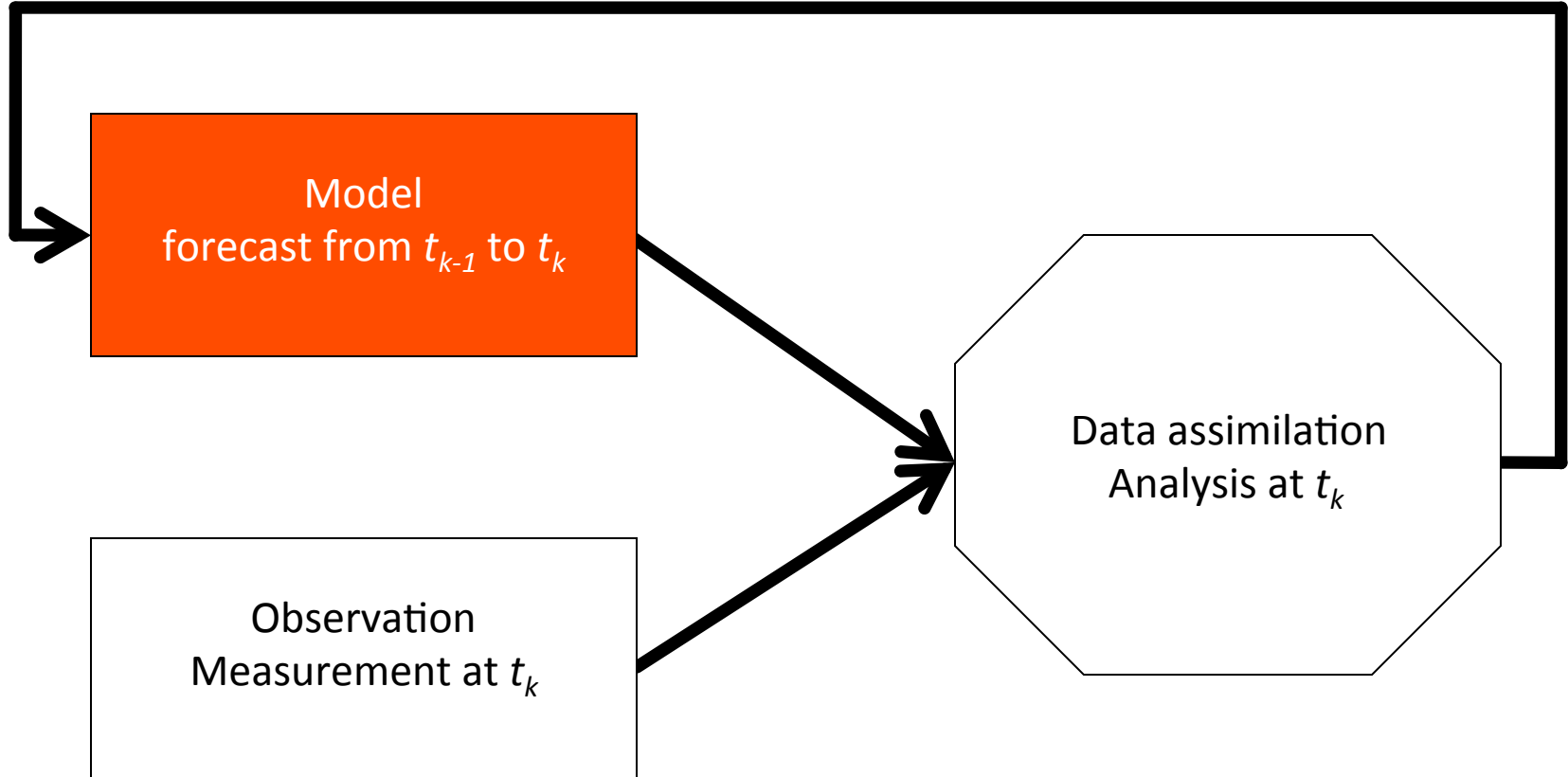
- Data Assimilation Overview
 - Background
 - Schematics of Data Assimilation
 - **Elements of Data Assimilation & Basic Notation**
 - Model
 - Observations
 - Methods
 - Challenges

- AOSC 615 Overview

- Project Overview

Elements of Data Assimilation: Computational Model

Assimilation Window



Elements of Data Assimilation: Computational Model Atmosphere

- Physical Model equations: Primitive equations
 - Approximation & Parametrization
 - Topography & Boundary conditions

$$\frac{d}{dt} \mathbf{v} = -\frac{1}{\rho} \nabla p - \nabla \phi + \mathbf{f} - 2\boldsymbol{\Omega} \times \mathbf{v}$$

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

$$p = \rho R T$$

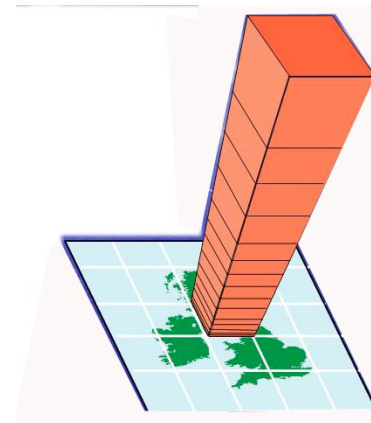
$$Q = C_p \frac{d}{dt} T - \frac{1}{\rho} \frac{d}{dt} p$$

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

- Computational Model
 - Each atmospheric variables are advanced at model grid points or by spectral coefficients
 - High resolution models are computationally expensive to run

$$\text{Ex: } u_{ijl}(t_{k+1}) = u_{ijl}(t_k)$$

+ Δt •discretized time tendency



Grid over England <http://www.climateprediction.net/>

Computational Model of Physical/Dynamical System

- Model variables: \mathbf{x} is N -dimensional vector
 - Atmospheric model (spectral, gridded): $N \sim O(10^{6-7})$
- Model dynamics

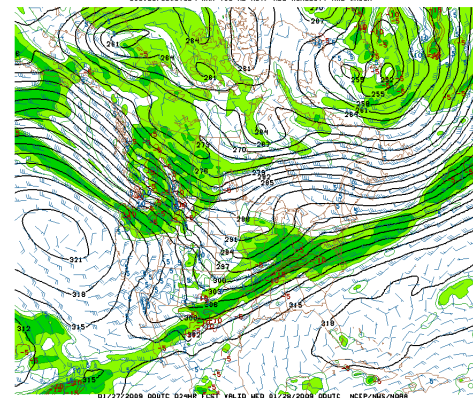
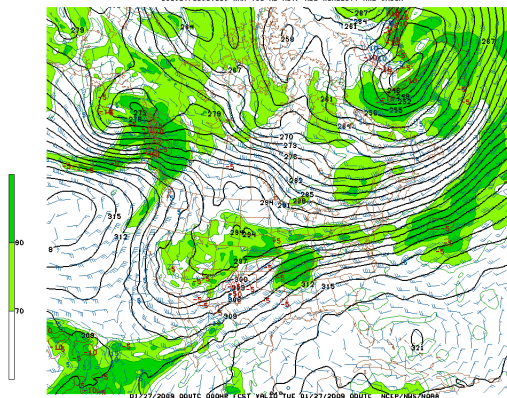
$$\mathbf{x} = \begin{pmatrix} T \\ q \\ D \\ h \\ p_s \end{pmatrix} = \begin{pmatrix} \text{Temperature} \\ \text{moisture} \\ \text{divergence} \\ \text{vorticity} \\ \text{surface pressure} \end{pmatrix}$$

$$\mathbf{x}_{k+1} = \mathbf{m}_{k+1,k}(\mathbf{x}_k)$$

\mathbf{x}_k : model state vector at time t_k

$\mathbf{m}_{k+1,k}$: (forecast) model from time t_k to t_{k+1}

$$\mathbf{x}_k @ t_k = 01 / 27 / 09 \text{ UTC} \longrightarrow \mathbf{x}_{k+1} = \mathbf{m}_{k+1,k}(\mathbf{x}_k) @ t_{k+1} = t_k + 24h$$



<http://www.nco.ncep.noaa.gov/pmb/nwprod/analysis/>

- To run a forecast, N -dimensional vector \mathbf{x}_k must be determined accurately.

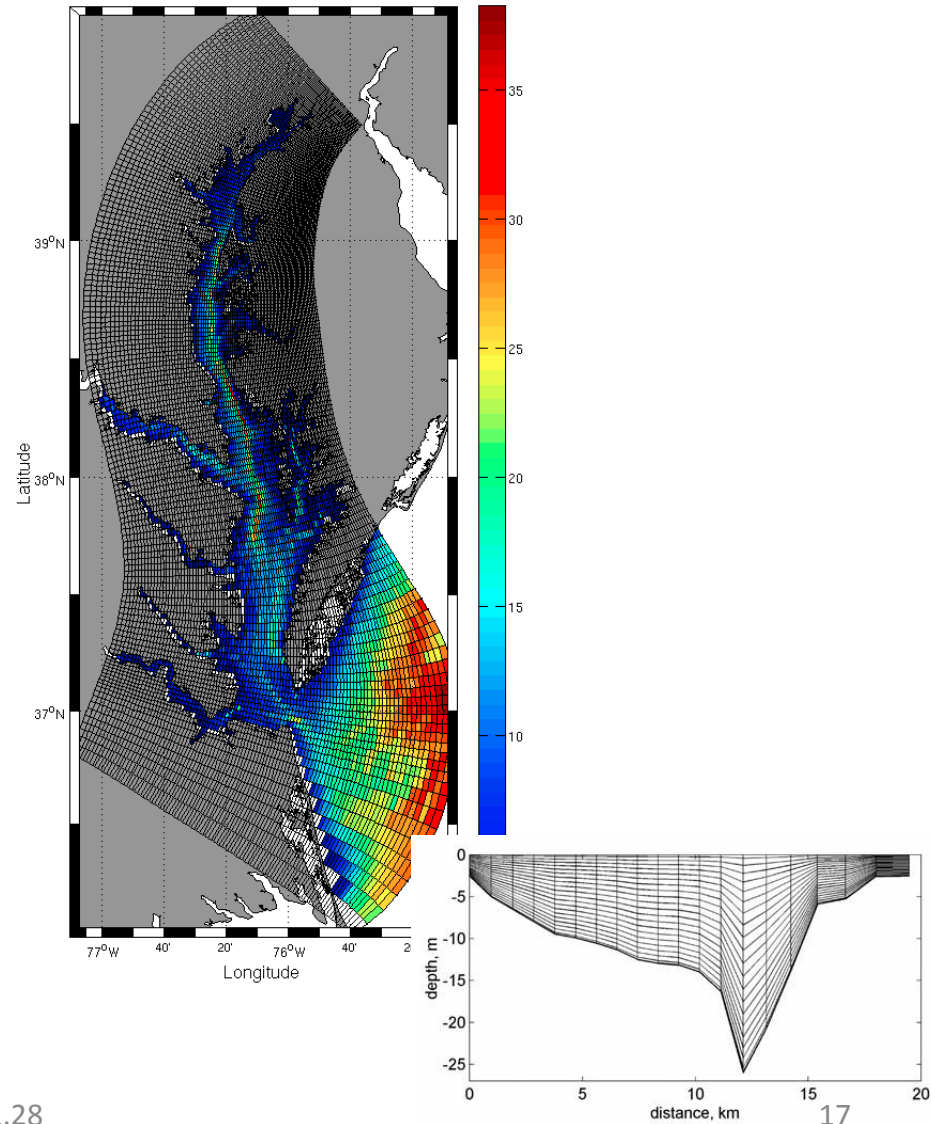
Elements of Data Assimilation: Computational Model Ocean

Regional Ocean Modeling System

- Primitive equation for

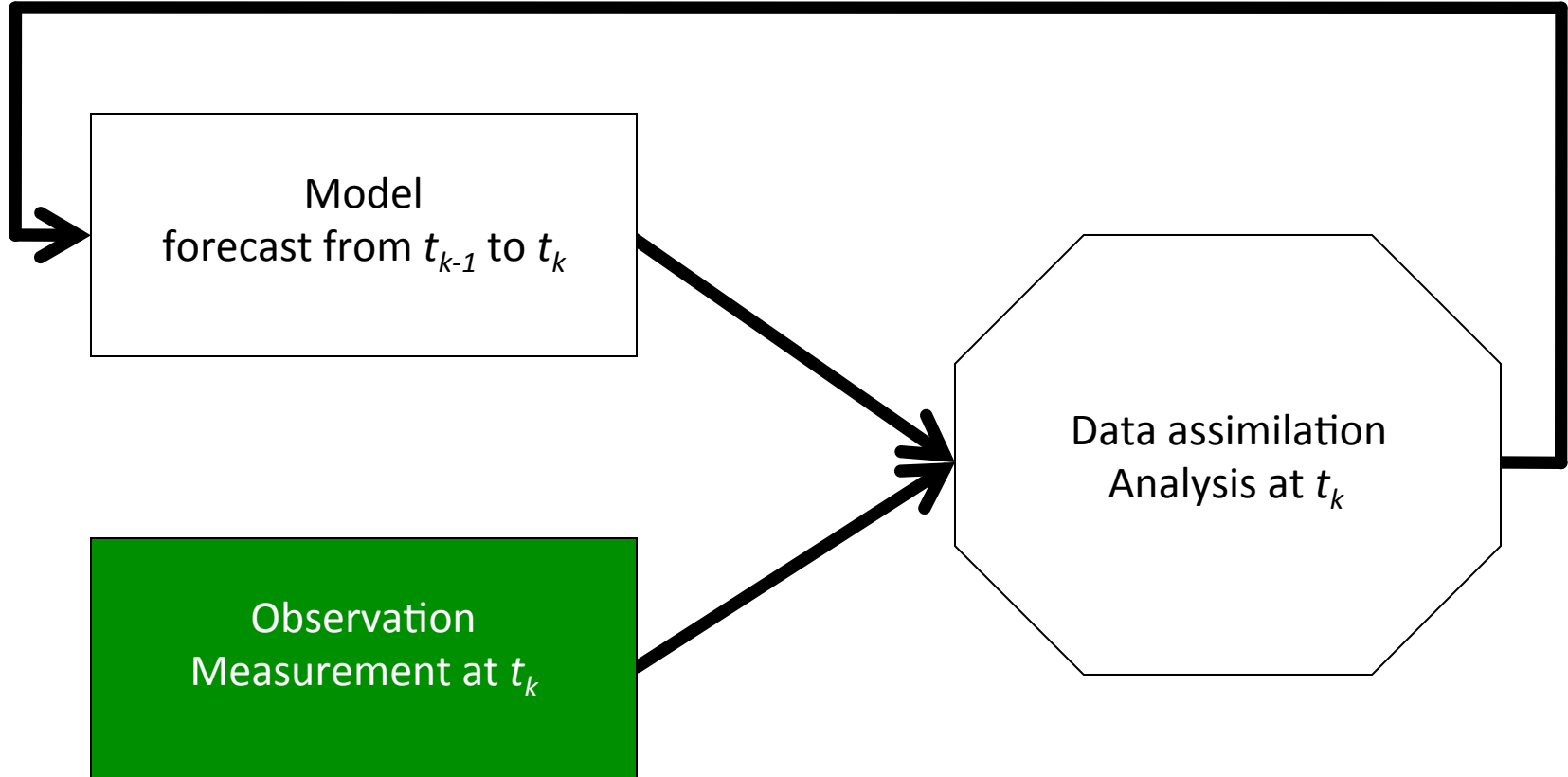
$$\mathbf{x} = \begin{pmatrix} \mathbf{u} \\ \mathbf{v} \\ \mathbf{T} \\ \mathbf{S} \\ \Phi \end{pmatrix}_{r,t} \rightarrow \mathbf{x}_{k+1} = \mathbf{m}_{k+1,k}(\mathbf{x}_k)$$

- Coordinate
 - Horizontal: Orthogonal curvilinear
 - Vertical: Stretched Sigma
- Specification
 - Nested configuration
 - Boundary conditions
 - Atmospheric configuration



Elements of Data Assimilation: Observation

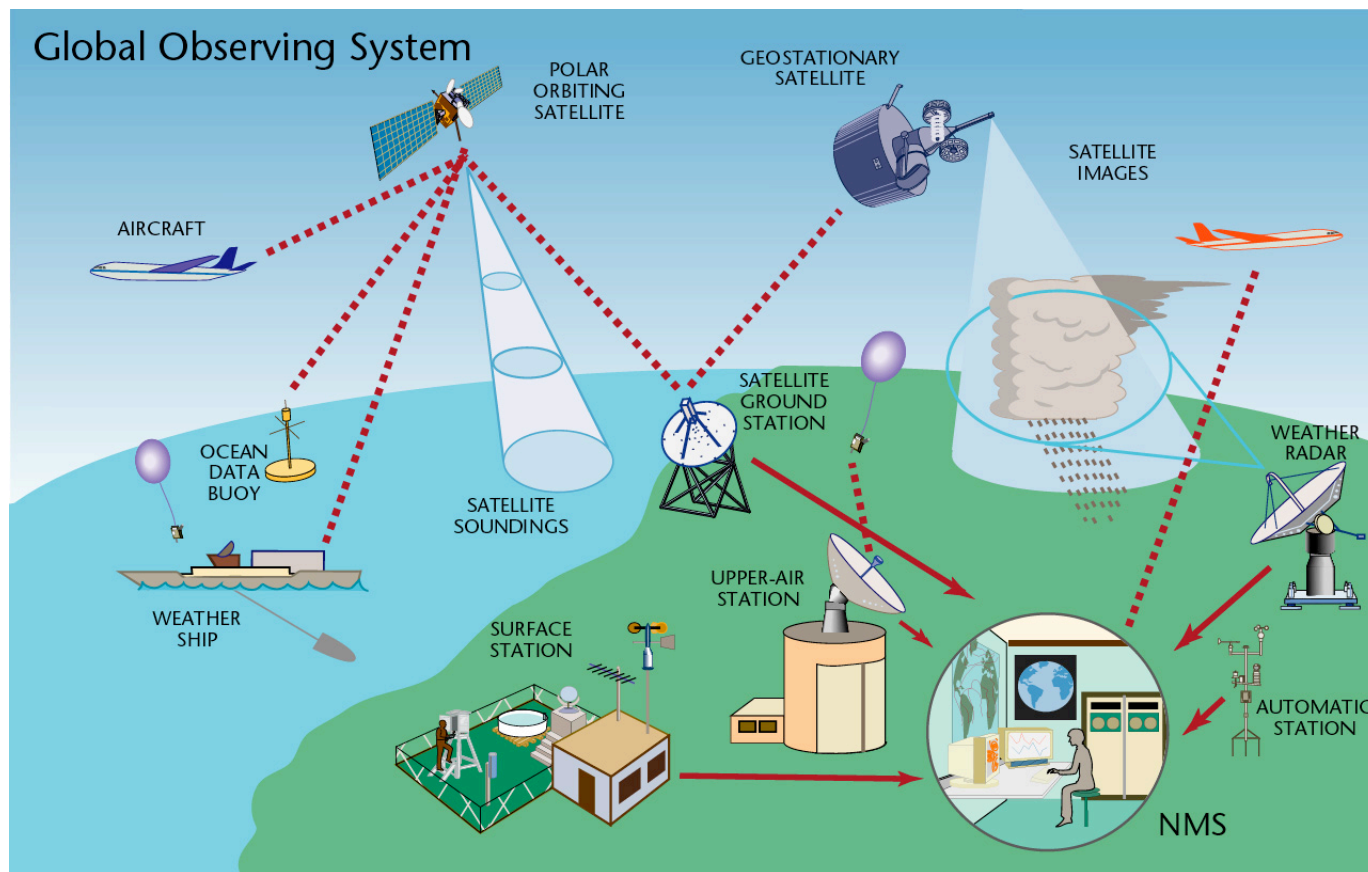
Assimilation Window



Observing System: Atmosphere

- Atmospheric Data Assimilation
 - Observations: Noisy heterogeneous sampling of the evolving state

Instrument Type



<http://www.wmo.int/pages/prog/www/OSY/GOS.html>

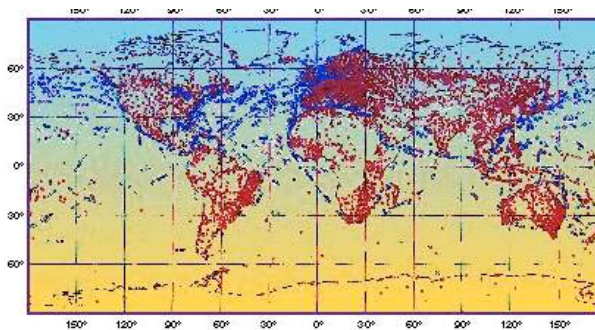
Example of Atmospheric Observation Operators : $y=h(x)$ ("Forward Model")

■ Atmospheric Data Assimilation

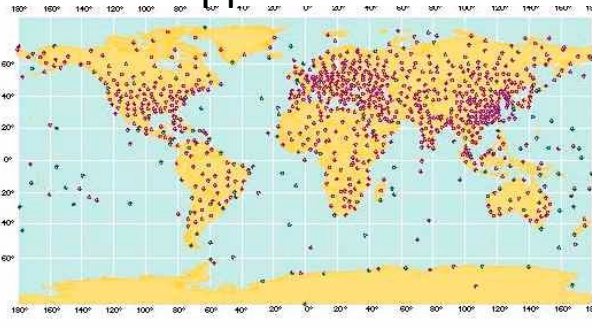
- Observations (y) in-situ or remote sensing
 - Noisy Heterogeneous sampling of the evolving state
 - Poor representation of forward model: $y \leftrightarrow x$

Coverage - Atmosphere

Surface obs



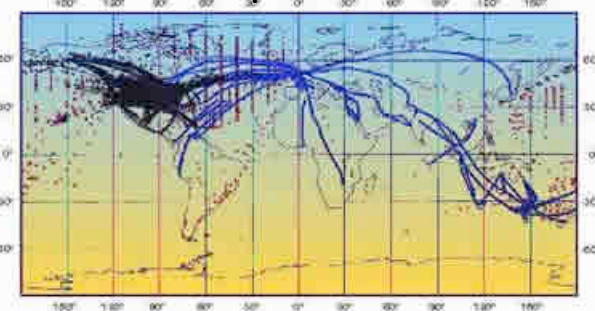
Upper air obs



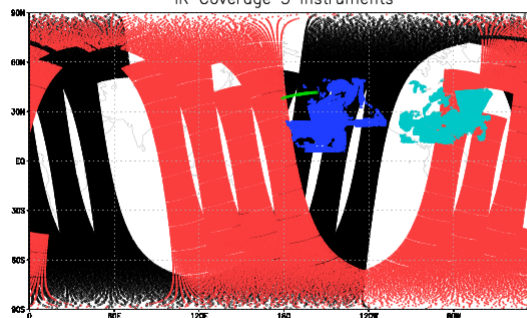
Marine obs



Airplane obs

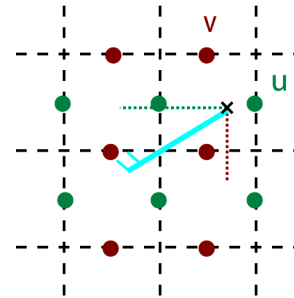
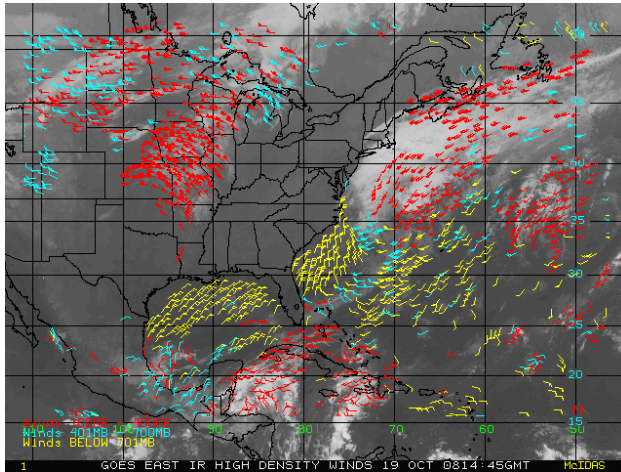


Satellite obs



Example of Atmospheric Observation Operators : $\mathbf{y}=\mathbf{h}(\mathbf{x})$ ("Forward Model")

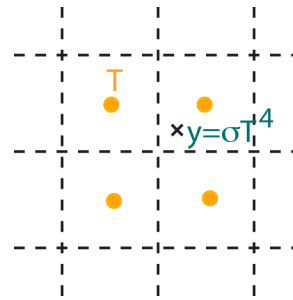
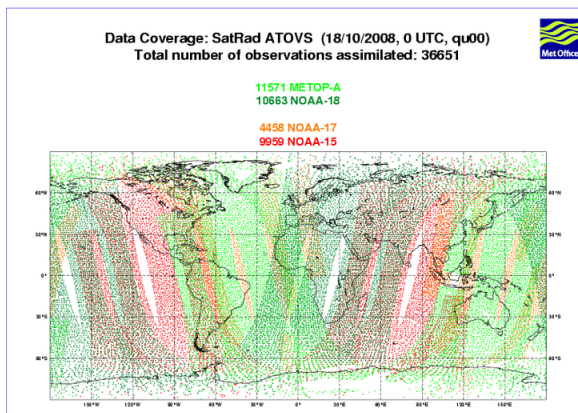
■ Atmospheric Wind Vector



$$\mathbf{y}_{\text{wind @ } r_1} = \begin{pmatrix} U_{\text{wind @ } r_1} \\ V_{\text{wind @ } r_1} \end{pmatrix} = \mathbf{H}_{\text{wind @ } r_1} \mathbf{x}$$

$\mathbf{H}_{\text{wind @ } r_1}$: linear interpolation operator

■ Radiance



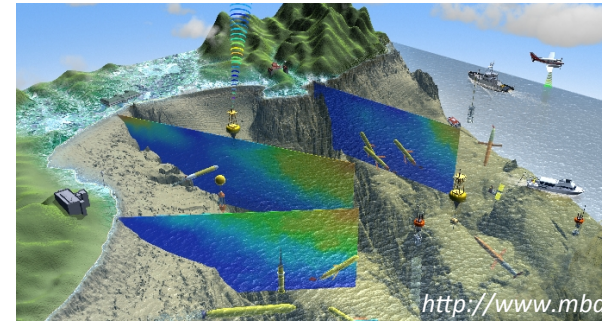
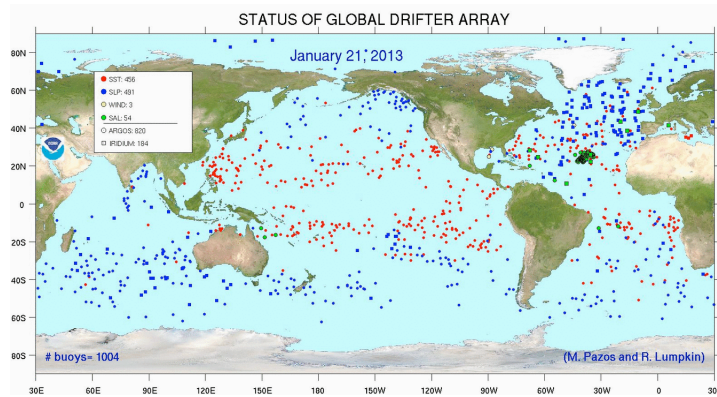
$$\mathbf{y}_{\text{radiance @ } r_2} = \sigma T_{\text{radiance @ } r_2}^4 = \mathbf{h}_{\text{radiance @ } r_2}(\mathbf{x})$$

σ : Stefan-Boltzman constant

$\mathbf{h}_{\text{radiance @ } r_2}$: nonlinear & interpolation operator

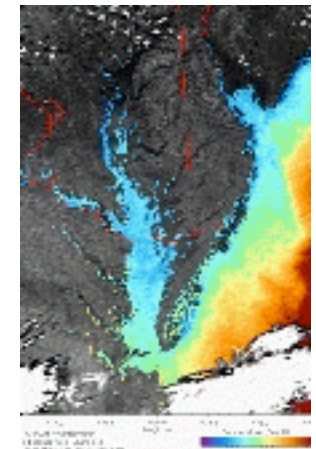
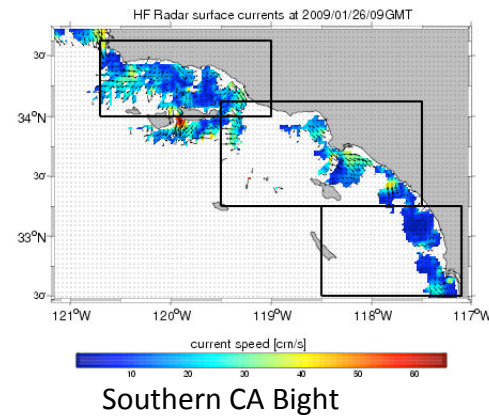
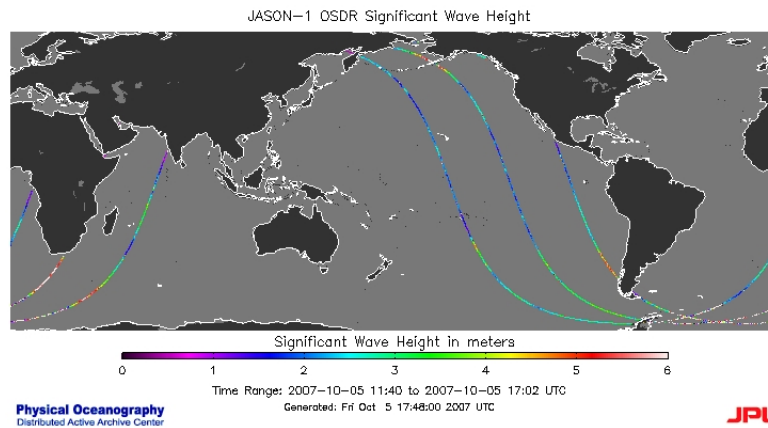
Observing Systems: Ocean

■ In Situ



Monterey Bay

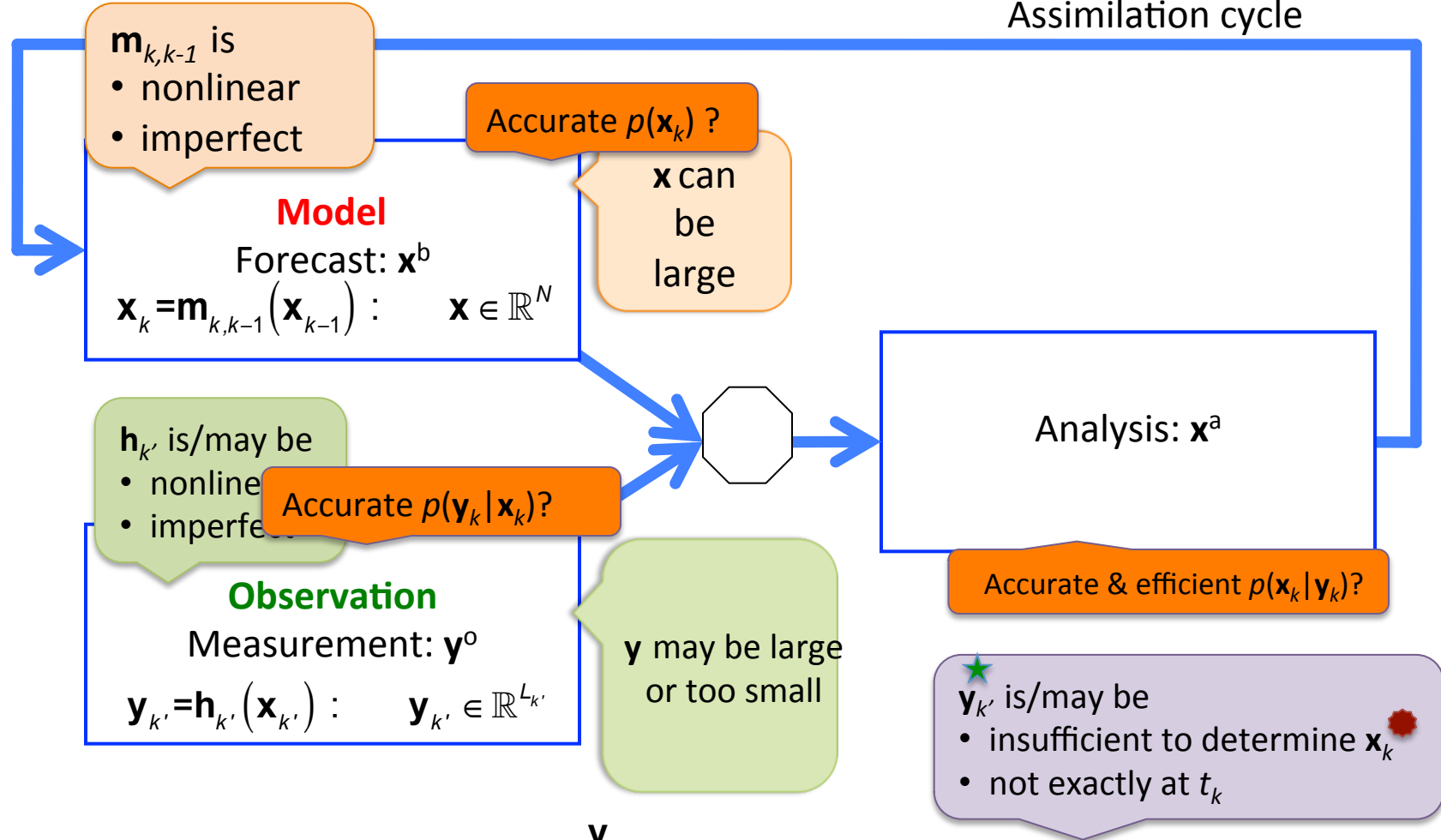
■ Remote Sensing



Chesapeake Bay SST

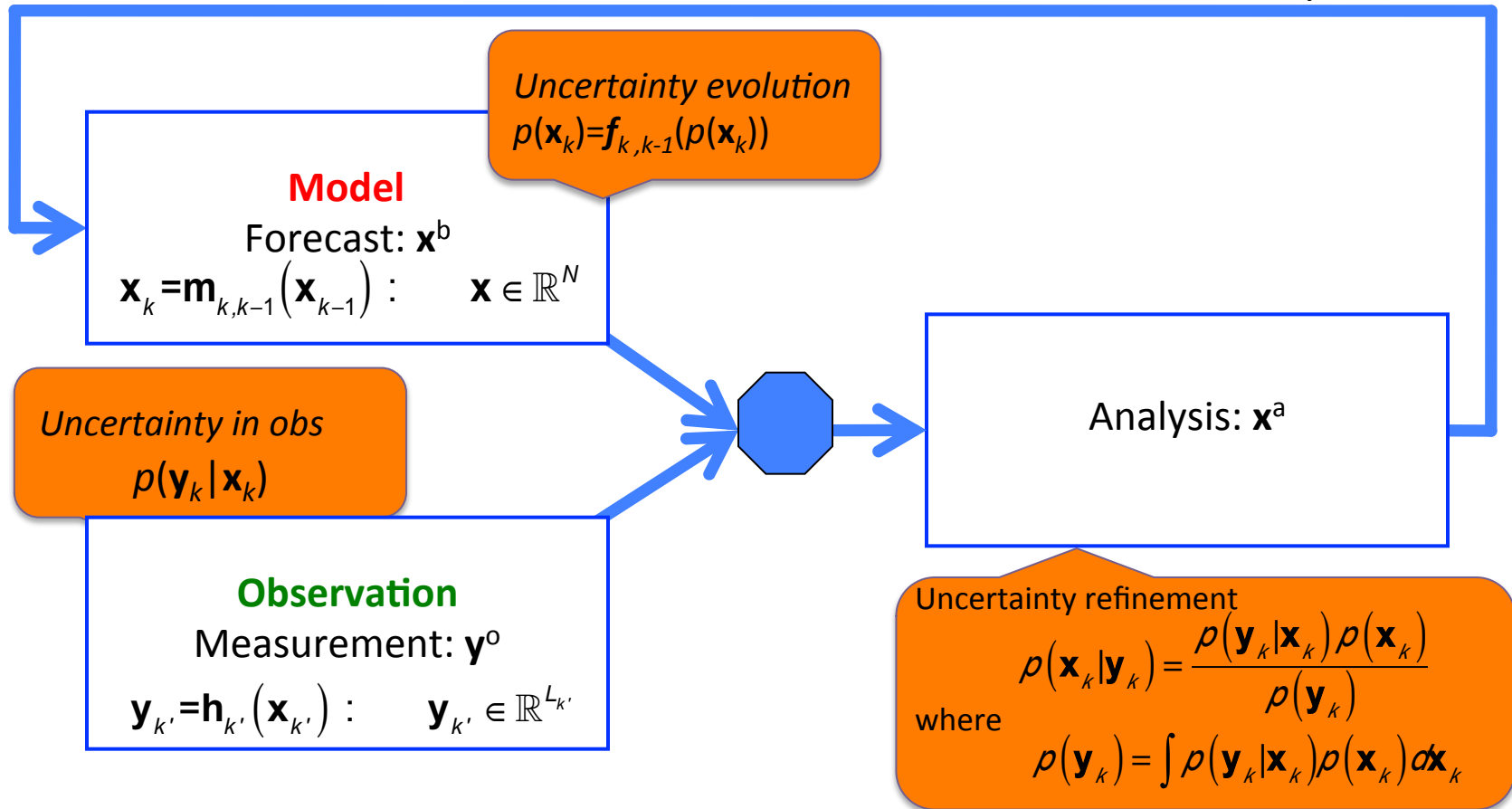
Challenges: Data Assimilation

Assimilation cycle



Bayesian View Point to Data Assimilation

Assimilation cycle

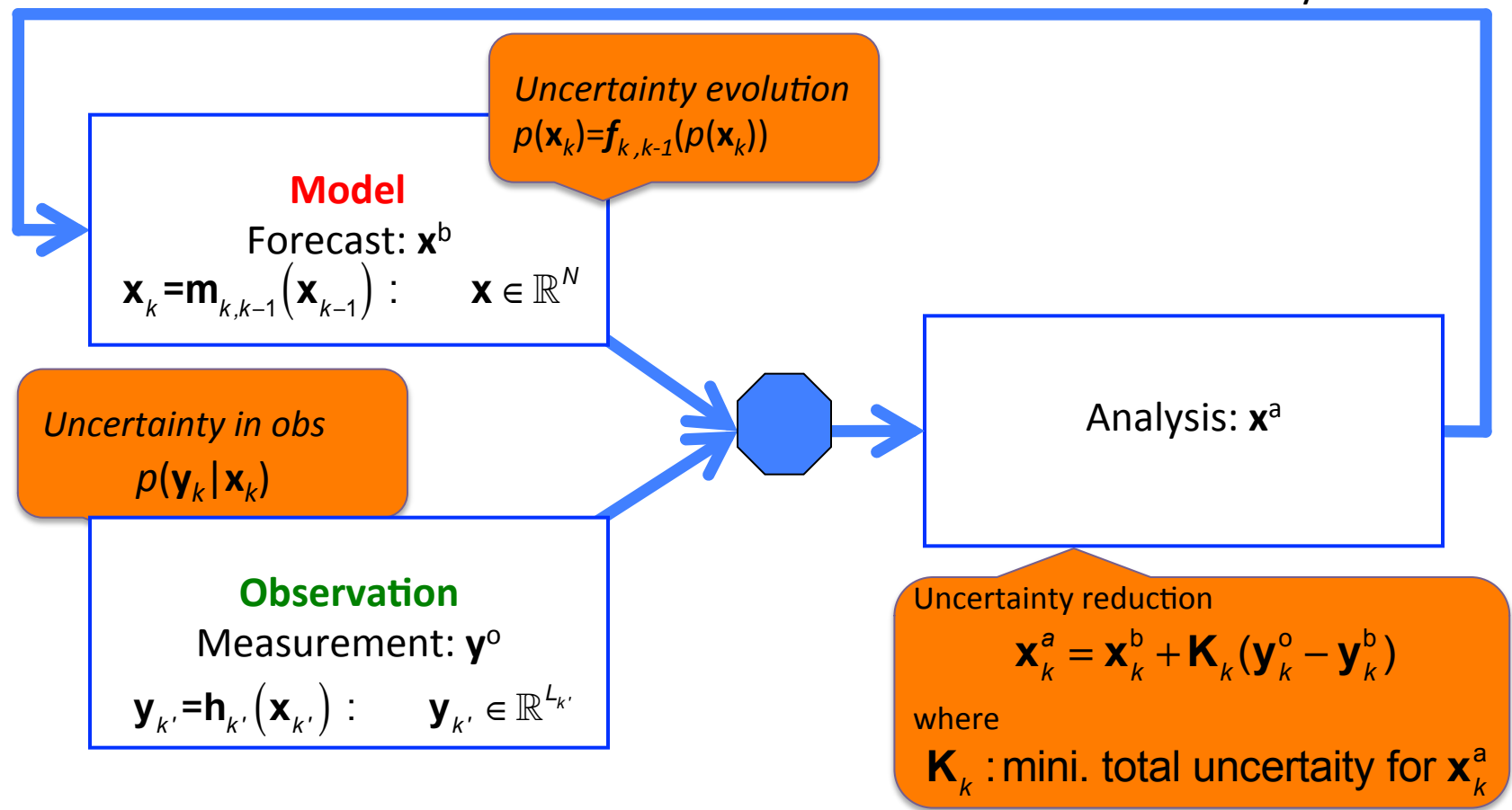


\mathbf{x}_{k-1}

\mathbf{x}_k

Risk-Reduction View Point to Data Assimilation

Assimilation cycle



\mathbf{x}_{k-1}

\mathbf{x}_k

AOSC615: Advanced Data Assimilation

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01.06. Course Objectives

- Data assimilation integrates all available (and useful) sources of information concerning the underlying unknown nature.

This is generally not true

- Primary information sources are (assumed to be correctly given):
 - **Forecast** by the model (dynamical/physical) obtained computationally by making forecast in forward time given the initial conditions
 - **Observation** that samples the nature

- Resulting state is the **analysis**, which is
 - Best possible estimate of the state
 - Best possible initial condition for next forecast

Main focus of AOSC615

- Data assimilation is a scientific problem of
 - Estimation: as truthful as possible to the nature
 - Prediction: as informative as possibleand not particularly limited to geophysical or earth systems.

Course Objectives

- AOSC 615 Objectives
 - Study fundamental concepts and ideas behind data assimilation
 - Learn major methods in data assimilation
 - Develop computer codes for the major methods
 - Foster ability to develop new data assimilation systems

AOSC 615 Format

- Class works
 - Lectures
 - Regular lectures
 - Special lectures by leading experts of data assimilation
 - Projects: Implementation of data assimilation methods
 - Using identical twin experiments
 - Based on the same model (=nature)

- Seminars
 - Students are highly encouraged to attend data assimilation seminars at AOSC, CSCAMM, & ESSIC
 - Announcement will be made as soon as possible
 - Some seminars will substitute regular lectures