Introduction

Data Assimilation

- Using field data and modeled data to create a basis for prediction
- To create these "means" we use algorithms: **filters** Ensemble Kalman/Particle
- Ability to take dynamic, multi-variable systems and assimilate
- Commonly found in weather forecasting

[1]



Figure 1: General graphic of data assimilation [1]

Background

Kalman Filter

Baye's Rule:

$$P(x_t|\psi_t) \propto P(y_t|x_t)P(x_t|\psi_{t-1})$$

to create Kalman Equation:

$$x^a = x^b - \frac{P^b}{P^b + R}(y - x^b)$$

- Effective for lower dimensional systems
- Higher dimensional systems require an ensemble of x^{b} 's to create an ensemble of x^{a} 's
- Ensemble Kalman Filter

Dual Filtering Data Assimilation of Dynamic Systems

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Methods

• Assimilate the Lorenz '96 model: $\frac{dx_i}{dt} = x_{i-1}(x_{i+1} - x_{i-2}) - x_i + F - \frac{hc}{b} \sum_{j=J(i-1)}^{iJ} \frac{dy_j}{dt} = -cby_{j+1}(y_{j+2} - y_{j-1}) - cy_j + \frac{hc}{b} x_{floor}$ $\left[2\right]$ Figure 2: The Cyclical Lorenz '96 System [2] • Create "field data" y through simulation • Create an ensemble of x^b 's from the Lorenz '96 1 2 3 4 5 6 7 8 9 10 0 1 2 3 4 5 6 7 8 9 **Figure 4**: $P^b = 1$ **Figure 3**: R = 10• Assimilate y to the x^{b} 's to create an ensemble of x^{a} 's (1)Assimilated Ensemble (2)

Conclusion/Continued Research

$$(3)_{j+1} y_j$$

$$or[(j-1)/J] + 1$$
(4)

$$P^b < R$$
 then $x^a \approx x^b$

•
$$R < P^b$$
 then $x^a \approx y$

• Next Step: Developing a Particle Filter to asses the parameter of Lorenz '96:

$$F = f_0 + \theta_k \sin\left(\frac{2\pi}{\theta_k + 1}\right)$$

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[1] Elaine Spiller.

pages 1–61, May 2017.

atmospheric dynamics.

• By filtering, find θ_k 's that reduce error in assimilation of Ensemble Kalman Filter

References





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Data Assimilation: Part 1 Overview and Particle Filters.

[2] Ross M. Leib-Lappen and Christopher M. Danforth. Aggressive shadowing of a low-dimensional model of

Physica D, 241(2012):637–648, December 2011.

[3] Naratip Santitissadeekorn and Christopher Jones. Two stage filtering for joint state-parameter estimation. Monthly Weather Review, 143:2028–2042, December 2014.

