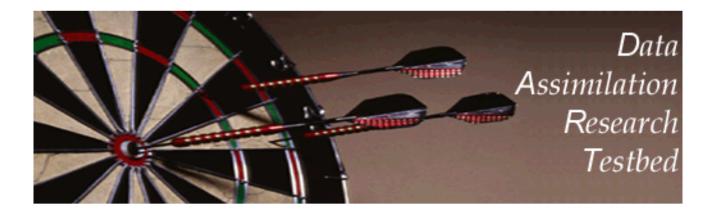
#### Data Assimilation Research Testbed Tutorial



Section 10: Regression and Non-linear Effects

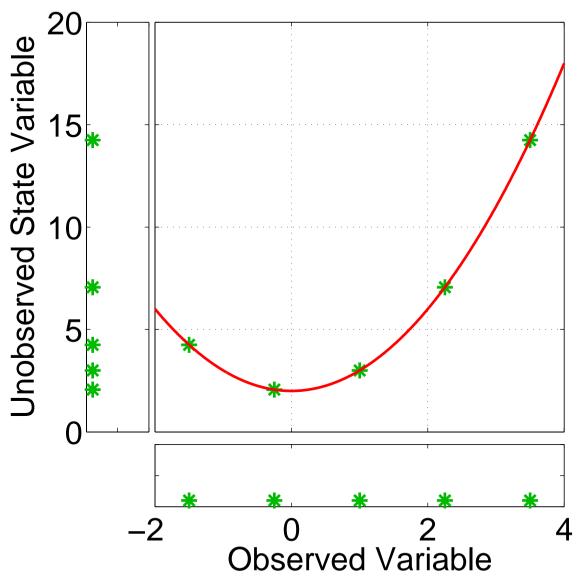
Version 2.0: September, 2006

# Ensemble filters: Updating additional prior state variables:

Two primary error sources:

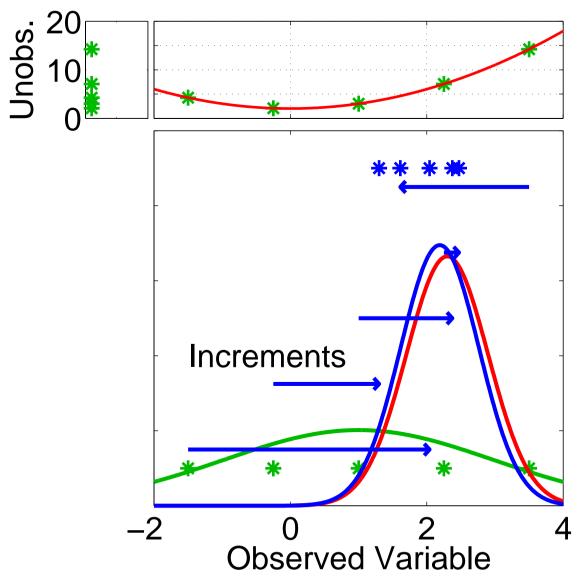
- 1. Linear approximation is invalid.
  Substantial nonlinearity in 'true' relation over range of prior.
- 2. Sampling error due to noise (we've already looked at this). Even if linear relation, sample regression coefficient imprecise.

May need to address both issues for good performance.



Suppose prior sample has NO noise.

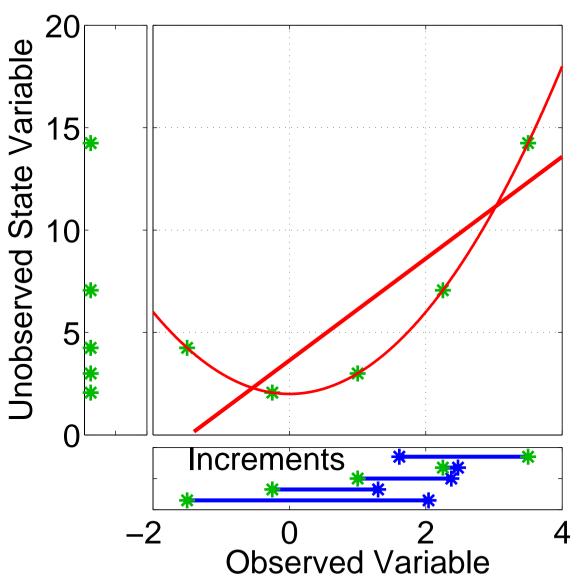
But, relation between un/observed variables is non-linear.



Suppose prior sample has NO noise.

But, relation between un/observed variables is non-linear.

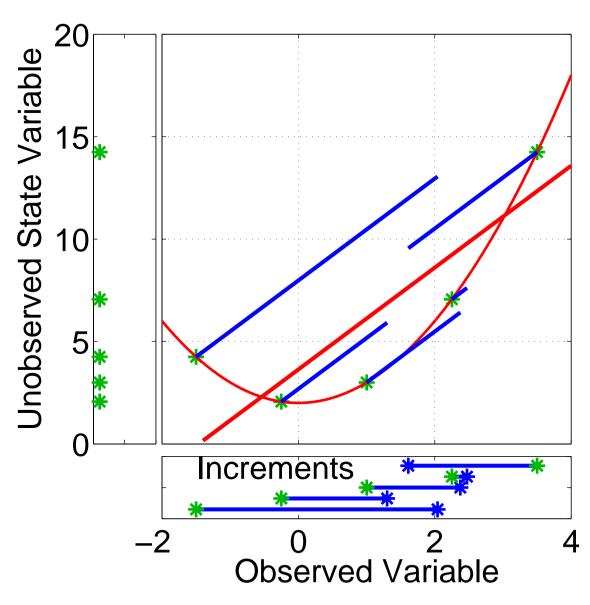
Update observed sample and compute increments.



Suppose prior sample has NO noise.

But, relation between un/observed variables is non-linear.

Regression error varies with value of observed variable.

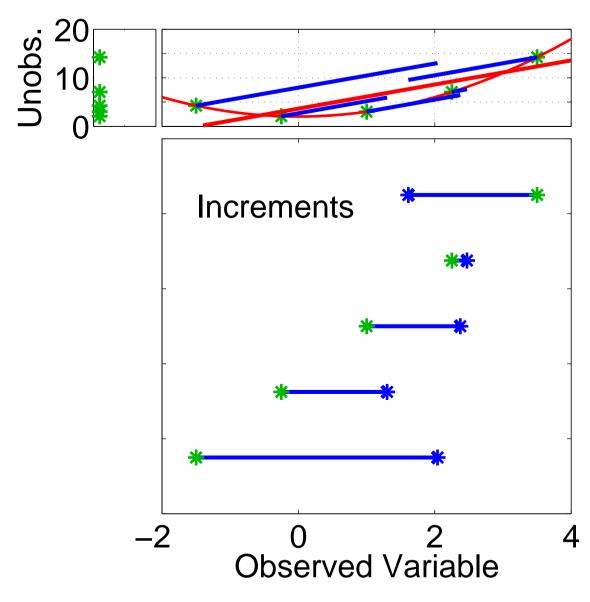


Suppose prior sample has NO noise.

But, relation between un/observed variables is non-linear.

Regression error varies with value of observed variable.

Smaller increments have smaller expected errors.

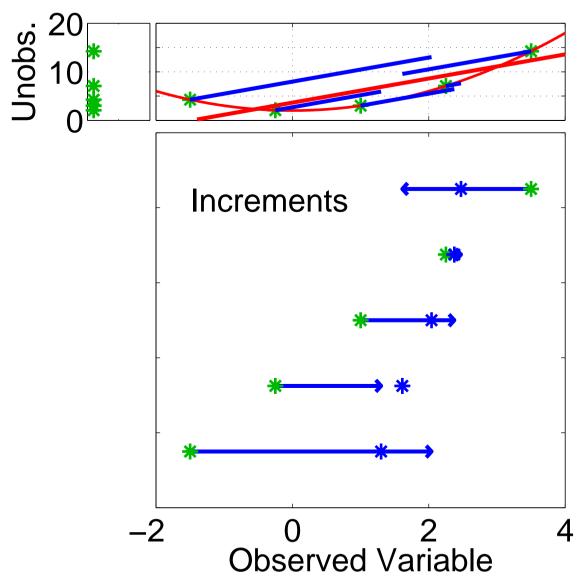


Suppose prior sample has NO noise.

But, relation between un/observed variables is non-linear.

Pairing between prior and posterior sample of observed variable can be viewed as arbitrary.

Posterior is same sample no matter how it is paired.

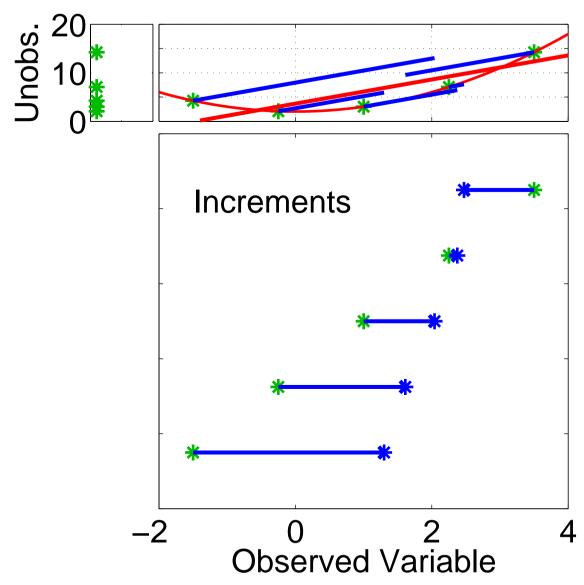


Suppose prior sample has NO noise.

But, relation between un/observed variables is non-linear.

Can minimize increments by changing pairing.

Sorting prior and posterior and pairing samples minimizes one norm of increment size (could do other methods)

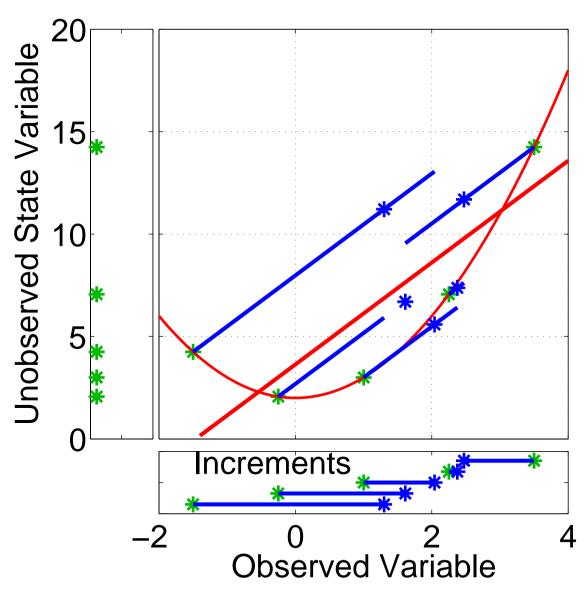


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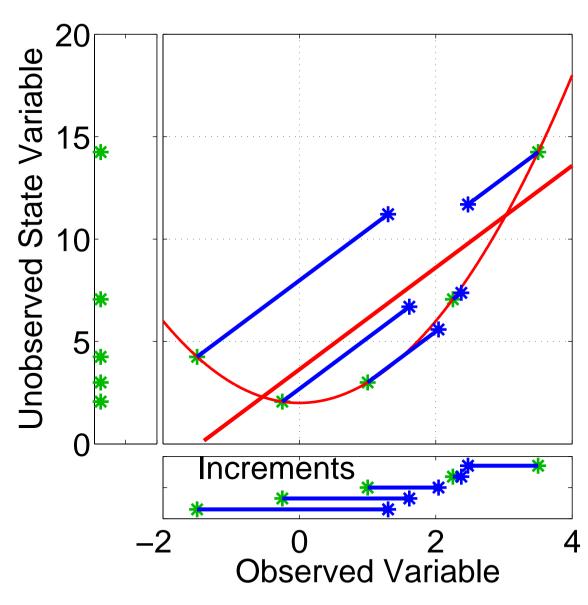


Suppose prior sample has NO noise.

Sorting prior and posterior and pairing samples minimizes one norm of increment size.

Resulting regression error is minimized.

Impact of sorting can be very large when posterior selected by 'random' algorithms.



Suppose prior sample has NO noise.

Sorting prior and posterior and pairing samples minimizes one norm of increment size.

Resulting regression error is minimized.

Impact of sorting can be very large when posterior selected by 'random' algorithms.

Can see this impact nicely in 9var model.

```
Try filter_kind = 2 in assim_tools_nml with:

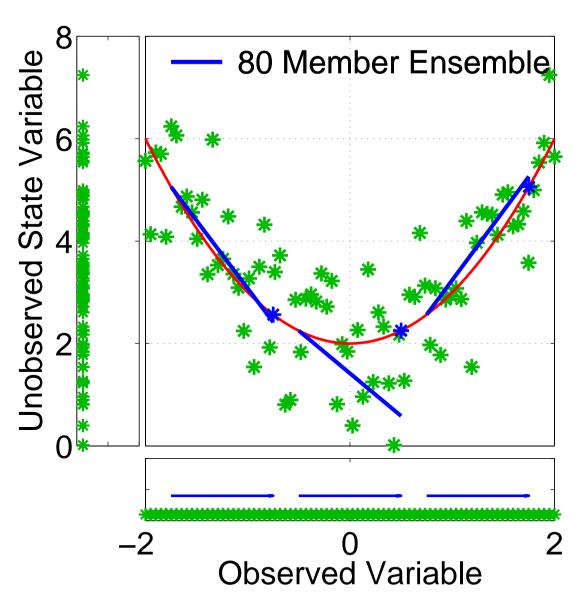
sort_obs_inc = .true. (increments minimized) and

sort_obs_inc = .false.
```

Compare the performance.

There may be surprises in other low-order models when trying this.

# Nonlinear relations between variables: Local regression



Prior sample is noisy.

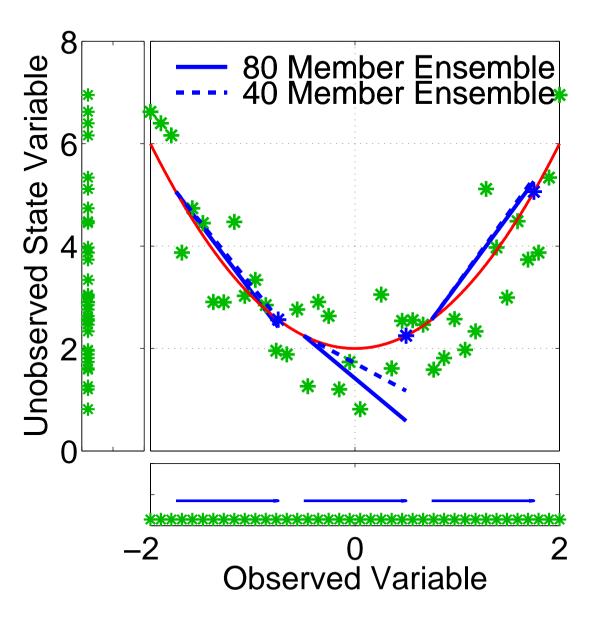
Un/observed relation is non-linear.

Doing global regression would be BAD here.

Can do regression only for points that lie in range of update increment.

Could also pick local sets in other ways.

### Nonlinear relations between variables; Local regression



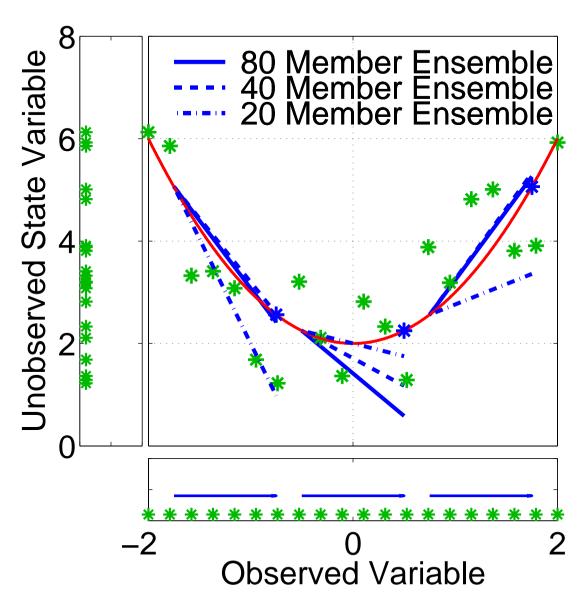
Prior sample is noisy.

Un/observed relation is non-linear.

For larger ensembles, local regressions can work well.

Error is largest where signal is weakest (near bottom of parabola here).

### Nonlinear relations between variables; Local regression



Prior sample is noisy.

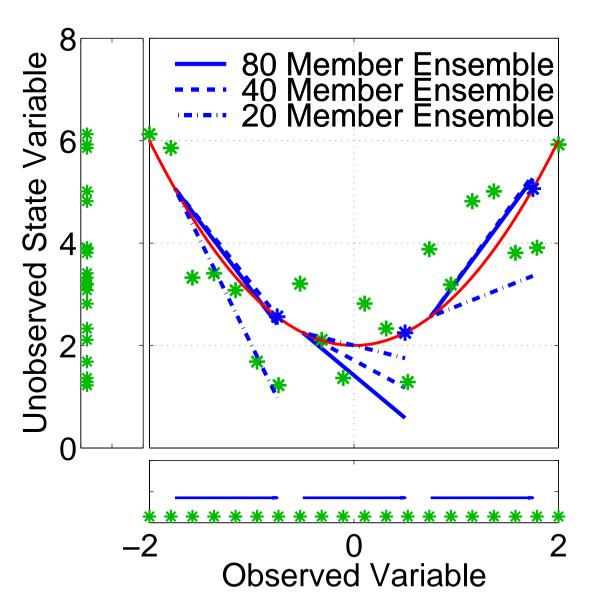
Un/observed relation is non-linear.

As sample size decreases, error grows.

(Except where it was rotten to start).

Applications where local regression is useful are unknown to me.

### Nonlinear relations between variables; Local regression



Prior sample is noisy.

Un/observed relation is non-linear.

Serious issues may exist if local regression is used with multiple unobserved state variables.