



Generic parallelization strategies for data assimilation



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Outline



- OpenDA
- Coupling of model to DA method
- Parallelism of model and EnKF
- Parallelism for black box models
- Conclusions

OpenDA

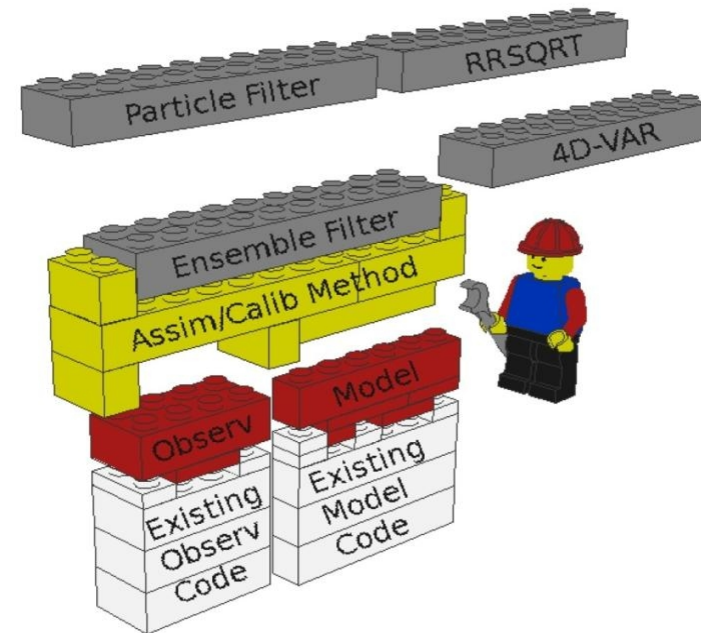
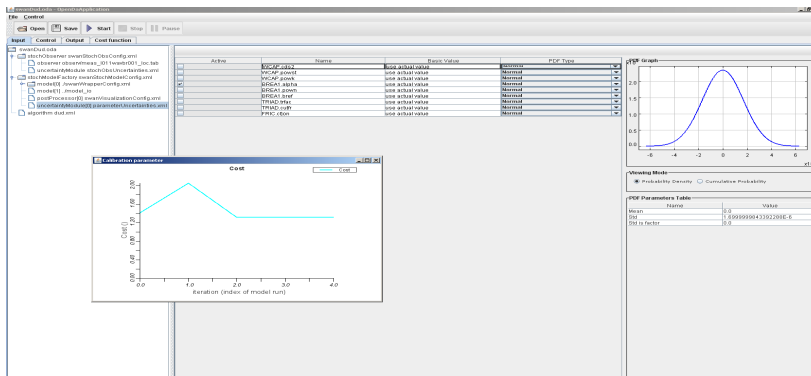


- What is OpenDA?
 - A generic toolbox for data-assimilation
 - Set of interfaces that define interactions between components
 - Library of data-assimilation algorithms
 - Open source
- Why OpenDA?
 - More efficient than development for each application
 - Shared knowledge between applications
 - Development of algorithms with e.g. universities
 - Easier to test

OpenDA



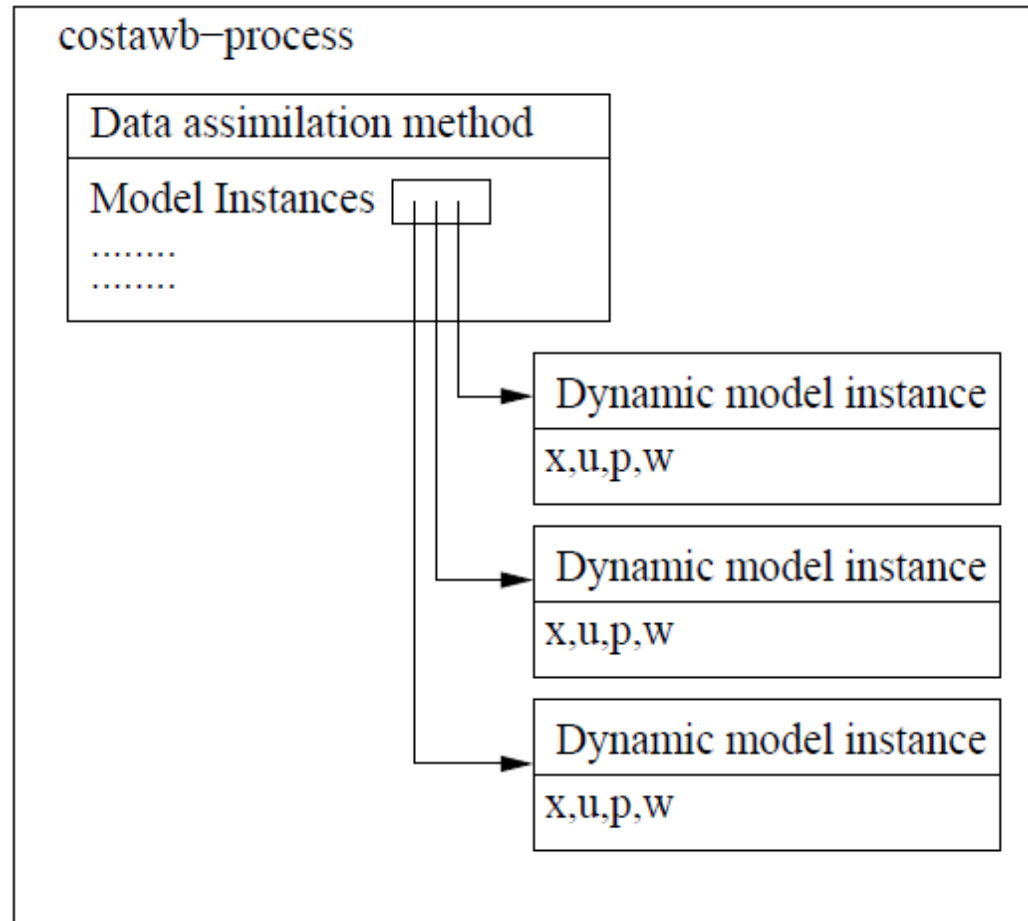
- Object oriented design
 - Classes, software building blocks
 - Interface (set of functions suitable for all models, observations, etc)



- Formal form of a model

$$\frac{dx}{dt} = M(x(t), u(t), p, w(t))$$

- State of model instance $x(t), u(t), p, w(t)$
- Instance state cannot be directly changed only through the methods like:
 GetState, AxyState, Compute...
- Algorithm has no knowledge on model internals



- Only parallelize model steps:
 - Not scalable, often sufficient

$$\xi_i^f(t_k) = M(\xi_i^a(t_{k-1})) + w_i(t_k)$$

$$x^f(t_k) = \frac{1}{N} \sum \xi_i^f(t_k)$$

$$E^f(t_k) = \begin{bmatrix} \xi_1^f(t_k) - x^f(t_k), \xi_2^f(t_k) - x^f(t_k), \dots, \xi_N^f(t_k) - x^f(t_k) \\ \vdots \end{bmatrix}$$

$$\xi_i^a(t_k) = \xi_i^f(t_k) + K(t_k)[y(t_k) - H(t_k)\xi_i^f(t_k) + v_i(t_k)]$$

Semi parallel
Communication
volume

$$C_m + nN$$

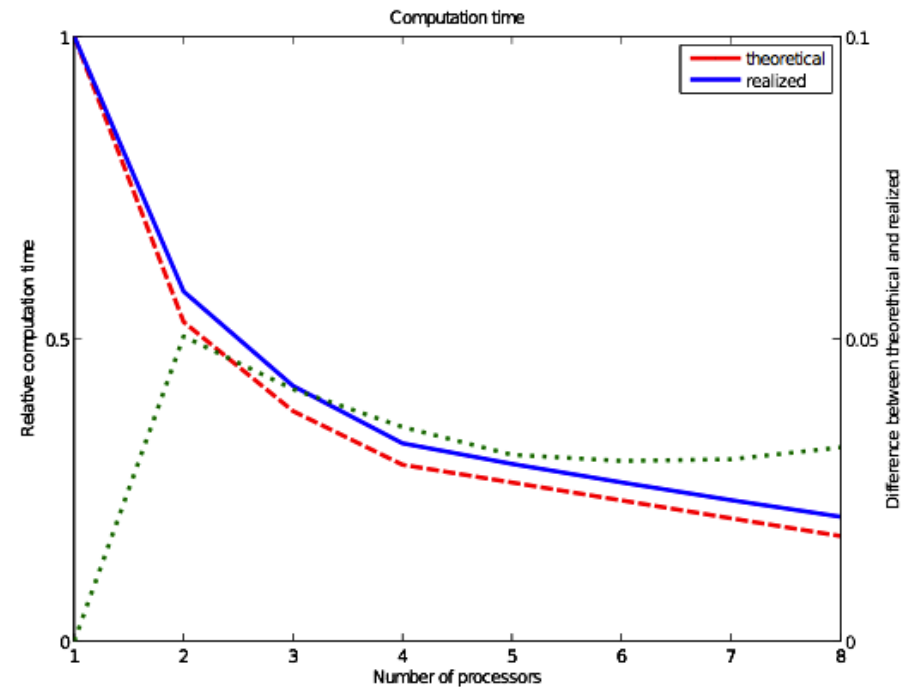
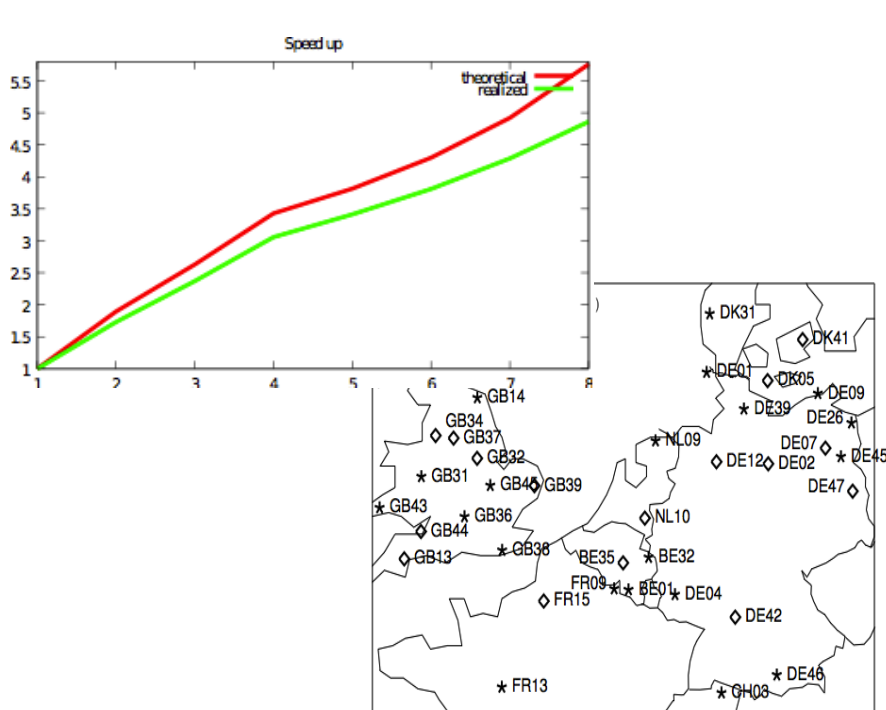
0

$$nN$$

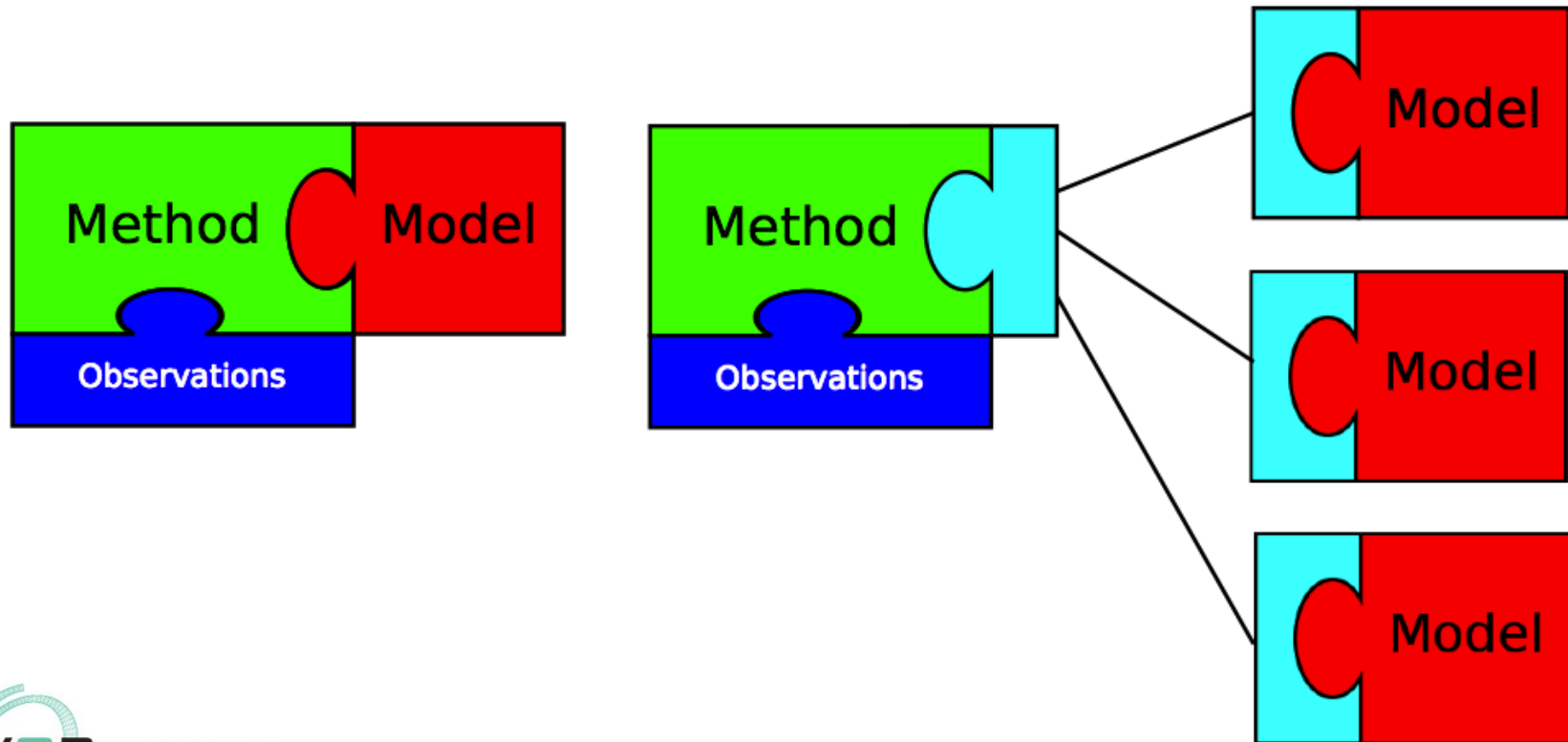
EnKF semi parallel



- Lotos-euros air quality model



- Generic semi parallel due to OO concepts

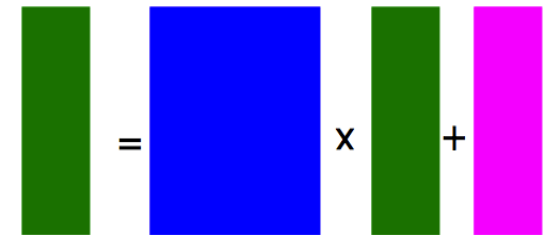


EnKF parallel



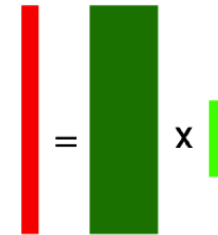
- Stochastic model time steps

$$\xi_i^f(t_k) = M(\xi_i^a(t_{k-1})) + w_i(t_k)$$



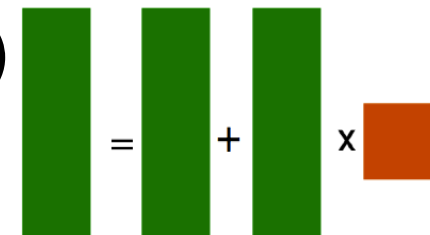
- Mean of Ensemble

$$x^f(t_k) = \frac{1}{N} \xi_i^f(t_k)$$



- Update (for all ensemble members)

$$\xi_i^a(t_k) = \xi_i^f(t_k) + K(t_k)[y(t_k) - H(t_k)\xi_i^f(t_k) + v_i(t_k)]$$

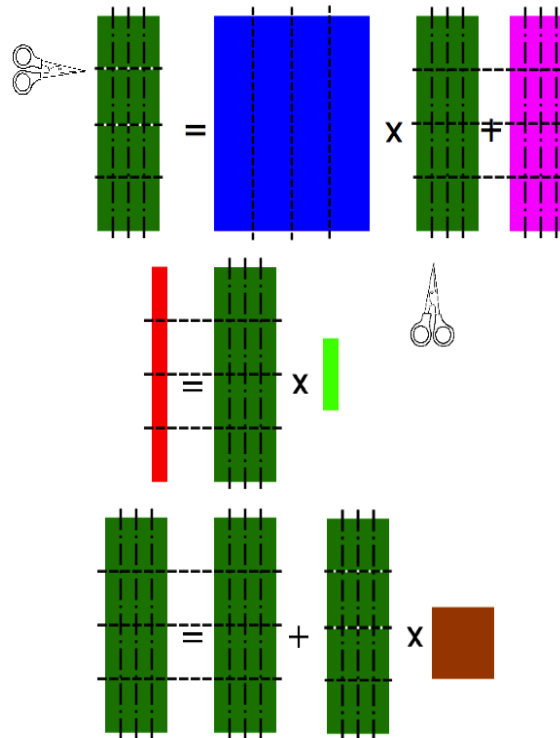


EnKF parallel



Column wise distribution

Separate	Combined
$\frac{nN}{p}$	0
$n \log_2(p)$	$n \log_2(p)$
$\frac{nN}{p} + nN \log_2(p-1)$	$nN \log_2(p-1)$



Row wise distribution

Separate	Combined
$C_m + \frac{nN}{p}$	C_m
0	0
$\frac{nN}{p}$	0

EnKF parallel



Column wise distribution

Separate

Combined

$$\frac{nN}{p}$$

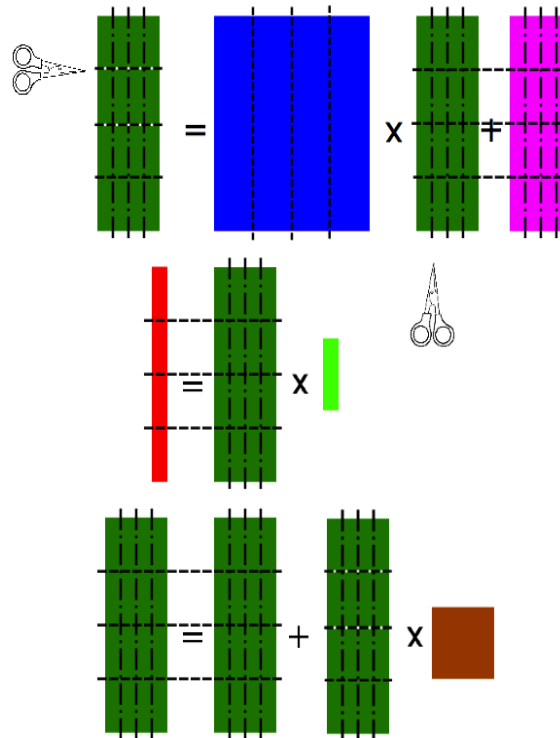
0

$$n \log_2(p)$$

$$n \log_2(p)$$

$$\frac{nN}{p} + nN \log_2(p-1)$$

$$nN \log_2(p-1)$$



Row wise distribution

Separate

Combined

$$C_m + \frac{nN}{p}$$

C_m

0

0

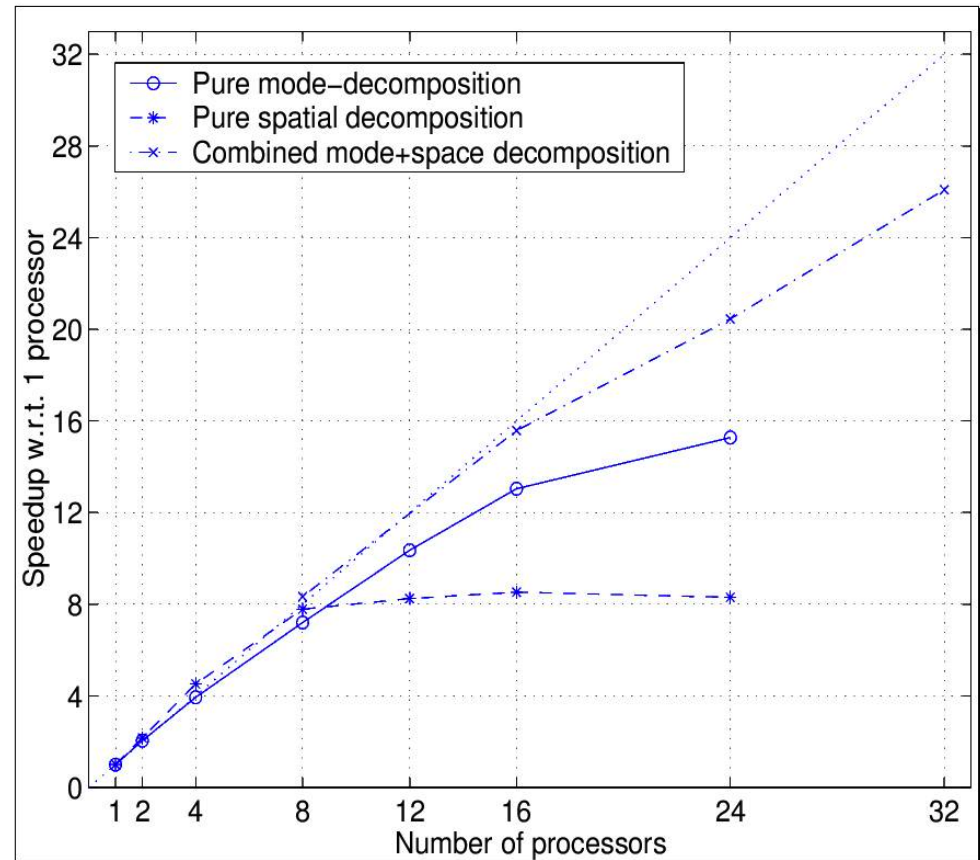
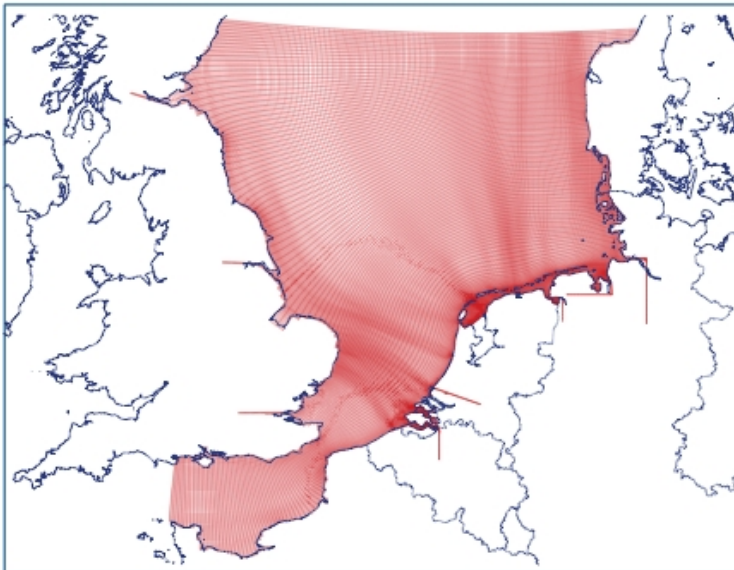
$$\frac{nN}{p}$$

0



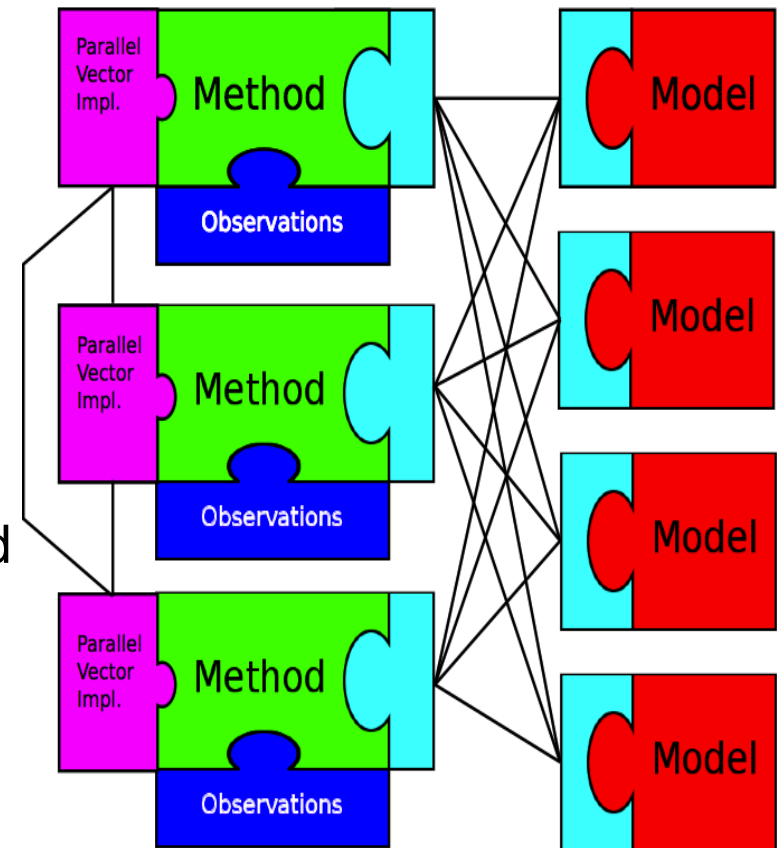
EnKF parallel

- WAQUA shallow water model
- Comparison of parallelization strategies for RRSQRT (Roest et al.)



EnKF parallel

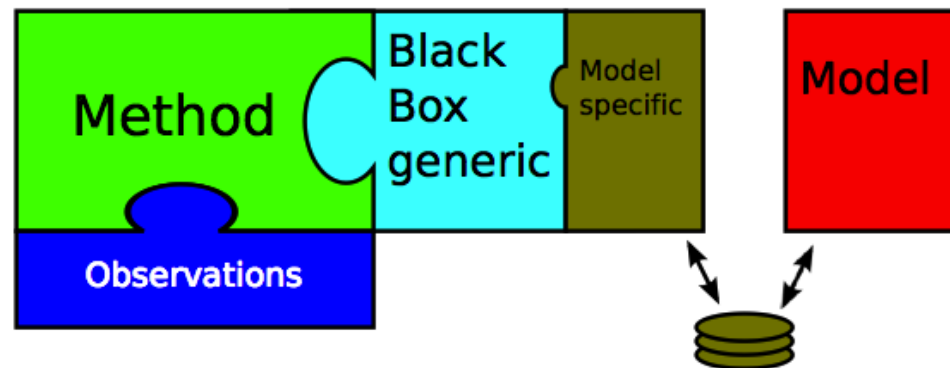
- Full parallel using OO concepts:
 - All filters run same code
 - State vectors are distributed (parallel vector)
 - Operations on parallel vectors by a parallel vector implementation
 - NO need to change the model and filter code
 - All complexities hidden in generic support layers/implementations



Parallel computing with black box models



- Black box models
 - No change to model code
 - Data exchange using files
 - Note: disk can be slow/more data written than needed



Parallel computing with black box models



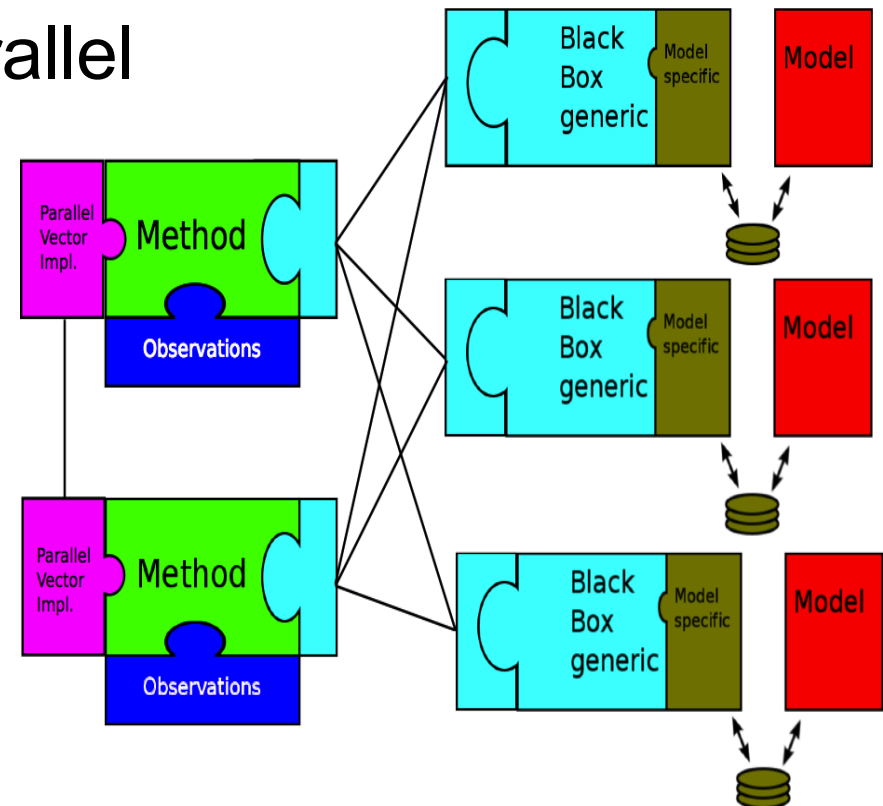
- Swan model for wind generated waves
 - Operational model + DA for the north sea
 - Black box model
 - 1 hour, 8 cpu's 10 min
50% IO
 - EnKF implementation (?)
 - Parallelization of filter



Parallel computing with black box models



- Black box model is normal model
 - Semi parallel+full parallel
 - Note disk speed,
 - Use local disks:
 - Faster
 - No sequential bottleneck



Conclusions



- Generic parallelization strategies due to object oriented programming concepts.
- Single filter implementation for sequential as parallel computing
- EnKF like algorithms need combination of parallel strategies
- Black box models and IO can be parallelized as well