

PDAF – Community Software for Ensemble Data Assimilation

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PDAF – Parallel Data Assimilation Framework

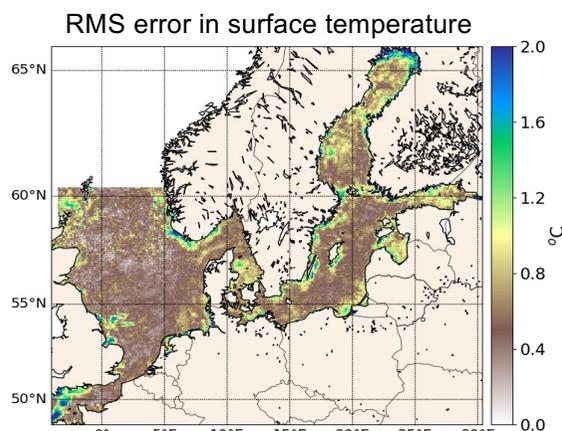
A universal tool for ensemble data assimilation ...

- provide support for parallel ensemble forecasts
- provide assimilation methods (solvers) - fully-implemented & parallelized
- provide tools for observation handling and for diagnostics
- easily useable with (probably) any numerical model
- a program library (PDAF-core) plus additional functions
- run from notebooks to supercomputers (using standards: Fortran, MPI & OpenMP)
- usable for real assimilation applications and to study assimilation methods
- ~550 registered users; community contributions

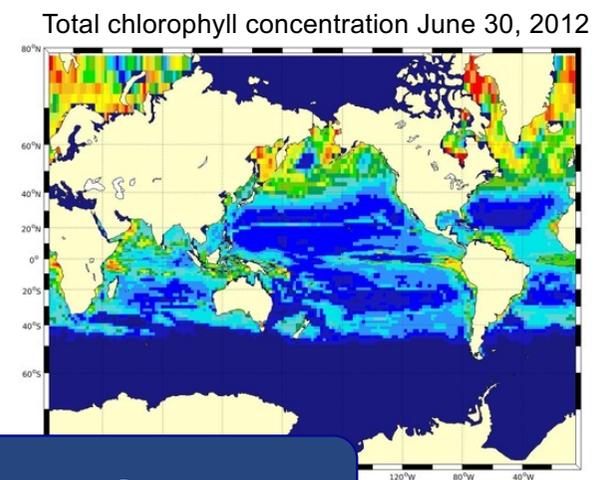
Open source:
Code, documentation, and tutorial available at
<http://pdaf.awi.de>

PDAF Application Examples

HBM-ERGOM:
coupled physics/
biogeochemistry
coastal assimilation
(Goodliff et al., 2019)

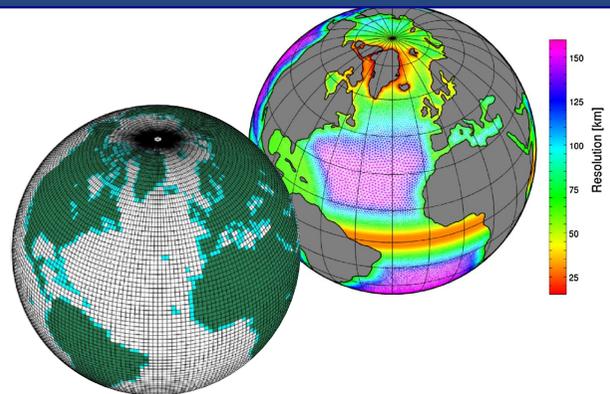


MITgcm-REcoM:
global ocean color
assimilation into
biogeochemical
model
(Pradhan et al., 2019/20)



Different models – same assimilation software

AWI-CM:
coupled atmos.-
ocean assimilation
(Tang et al., 2020
Mu et al., 2020
Nerger et al., 2020)

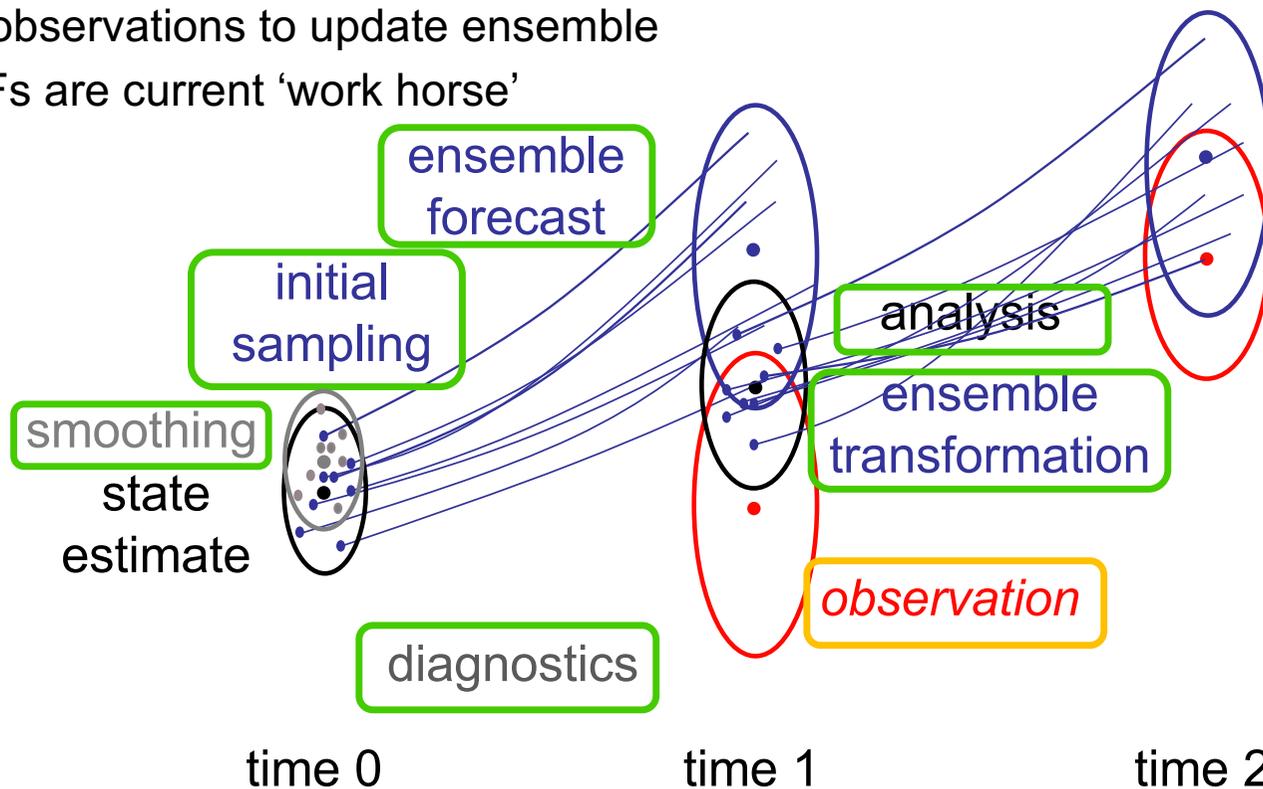


- MITgcm sea-ice assim (*operational*, NMEFC Beijing)
- CMEMS Baltic-MFC (*operational*, SMHI/DMI/BSH)
- HBM North/Baltic Seas (*operational*, BSH)
- NEMO (U Reading, P. J. van Leeuwen)
- SCHISM/ESMF (VIMS, J. Zhang)
- TerrSysMP-PDAF (hydrology, FZ Juelich, U Bonn)
- TIE-GCM (U Bonn, J. Kusche)
- VILMA (GFZ Potsdam)
- Parody geodynamo (IPGP Paris, A. Fournier)

Ensemble Data Assimilation

Ensemble Kalman Filters & Particle Filters

- Use ensembles to represent state and uncertainty
- Propagate ensemble using numerical model
- Use observations to update ensemble
- EnKFs are current 'work horse'



These steps can be implemented in a generic way

We can provide software including the algorithms
→ PDAF

Observation operators are less generic → PDAF also provides some operators and tools for obs. handling

PDAF: User-friendliness

Goal: Enable easy and fast setup of a DA system,
and allow for extension to fully featured system

Assumption: Users know their model

→ let users implement DA system in model context

For users, model is not just a time-stepping operator

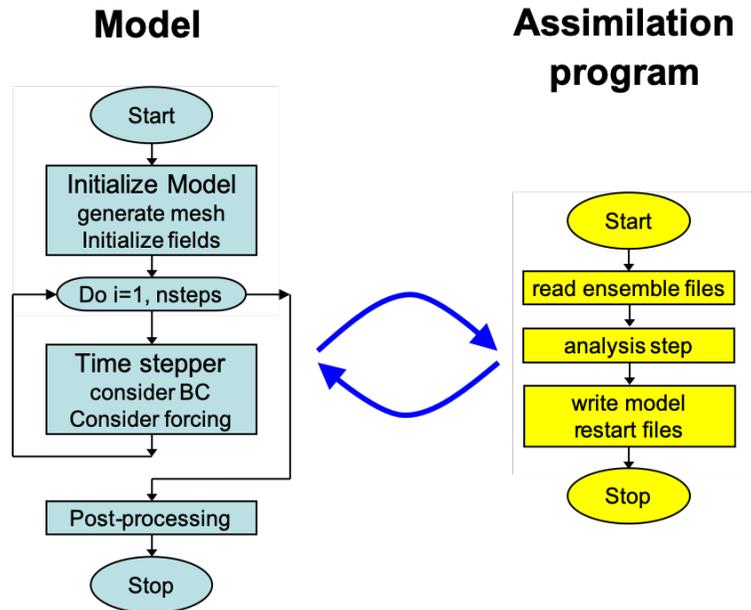
→ let users extend their model for data assimilation

Keep code simple for the user side:

- Define subroutine interfaces to DA code based on arrays
(also simplifies interaction with languages like C/C++/Python)
- No object-oriented programming
(most models don't use it; most model developers don't know it;
many objects we would only have for observations – see later)
- Users directly implement case-specific routines
(no indirect description (XML, YAML, ...) of e.g. observation layout)

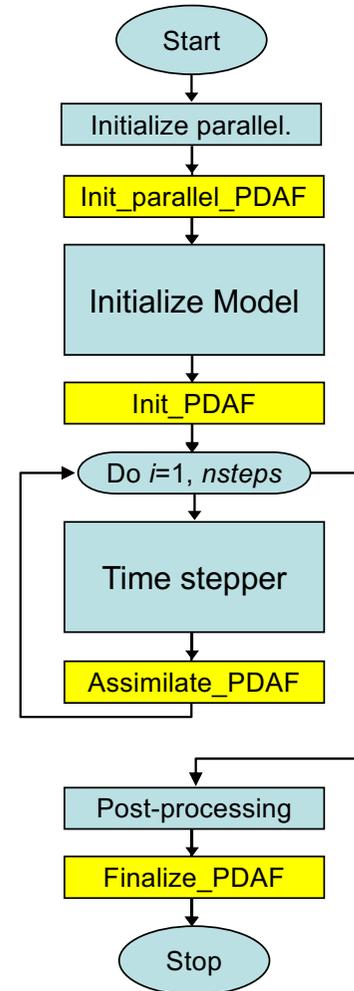
Coupling Model and Assimilation Code: 2 Variants

Offline



- Separate programs for model and DA
- Flexible to run
- Needs frequent model restarts and file output (less efficient than online coupling)

Online



- Augment model with DA functionality
- Insert 4 subroutine calls
 - Ensemble model
 - DA functionality
- Very efficient & highly scalable

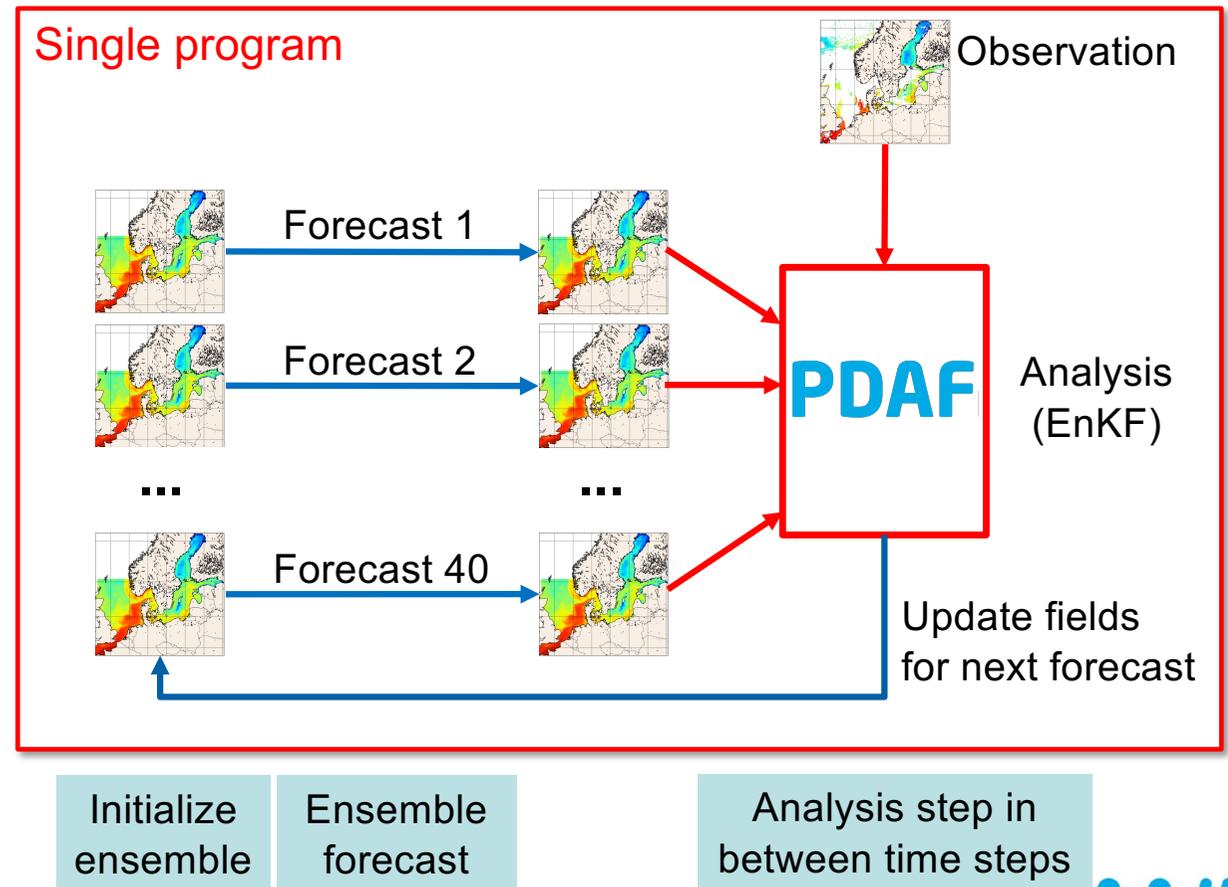
Model code

DA code

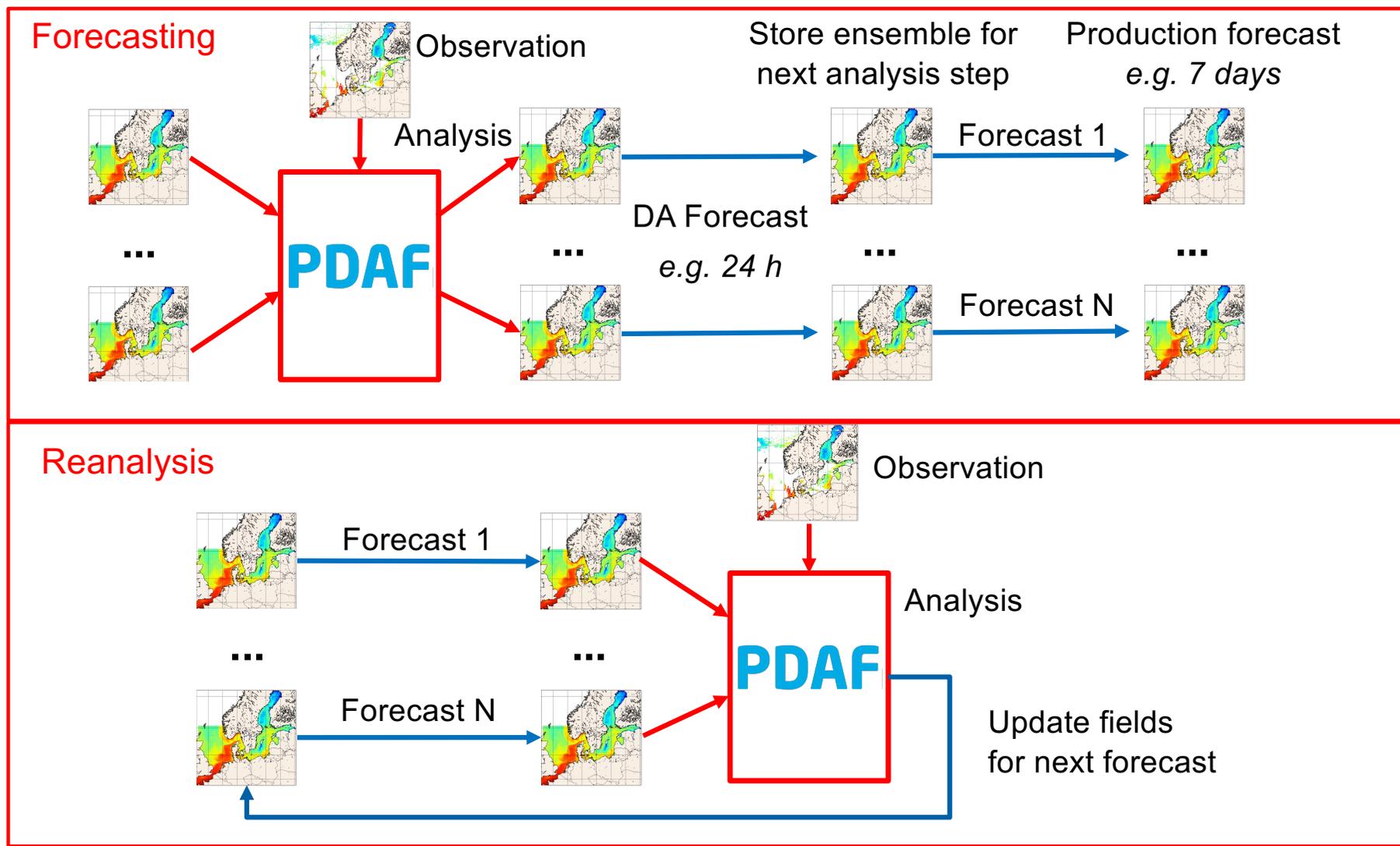
Online-Coupling – Assimilation-enabled Model

Couple a model with PDAF

- Modify model to simulate ensemble of model states
- Insert analysis step/solver to be executed at prescribed interval
- Run model as usual, but with more processors and additional options
- EnOI also possible:
 - Evolve single model state
 - Prescribe ensemble perturbations



Application types



Execution times (weakly-coupled, DA only into ocean)

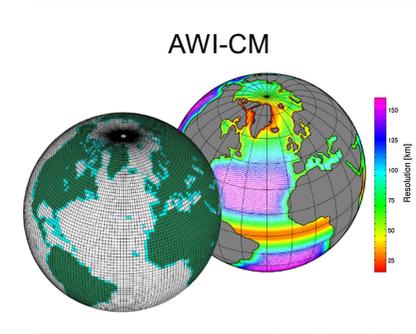
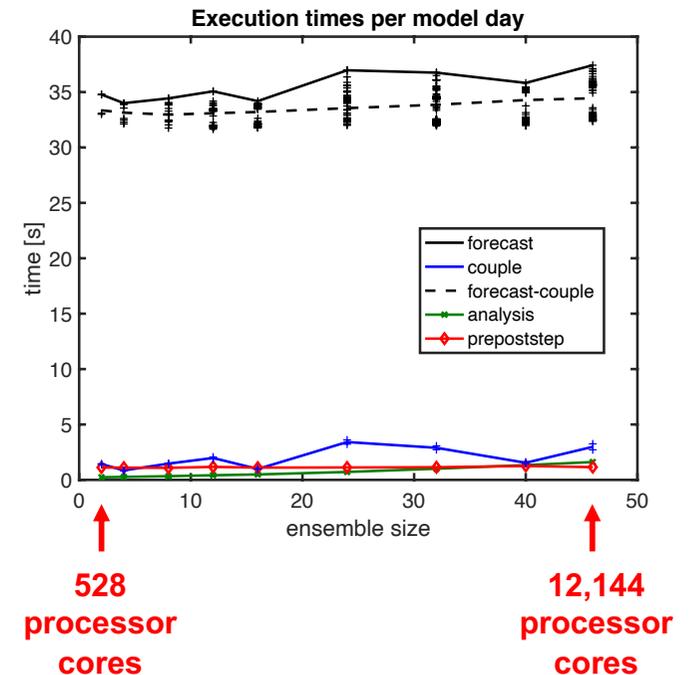
AWI-CM (atmosphere-ocean coupled model)

Processor cores (each model instance)

- ECHAM: 72 / FESOM: 192
- Vary ensemble size
- Increasing integration time with growing ensemble size (11%; more parallel communication; worse placement)
- some variability in integration time over ensemble tasks

Important factors for good performance

- Need optimal distribution of programs over compute nodes/racks (here set up as ocean/atmosphere pairs)
- Avoid conflicts in IO (Best performance when each AWI-CM task runs in separate directory)



Online and Offline Coupling - Efficiency

Offline-coupling is simple to implement but can be very inefficient

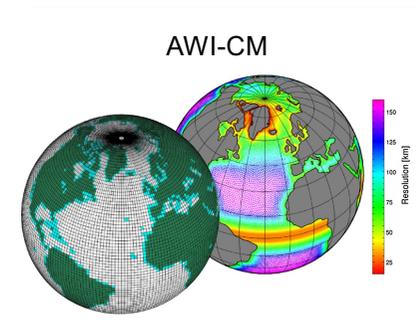
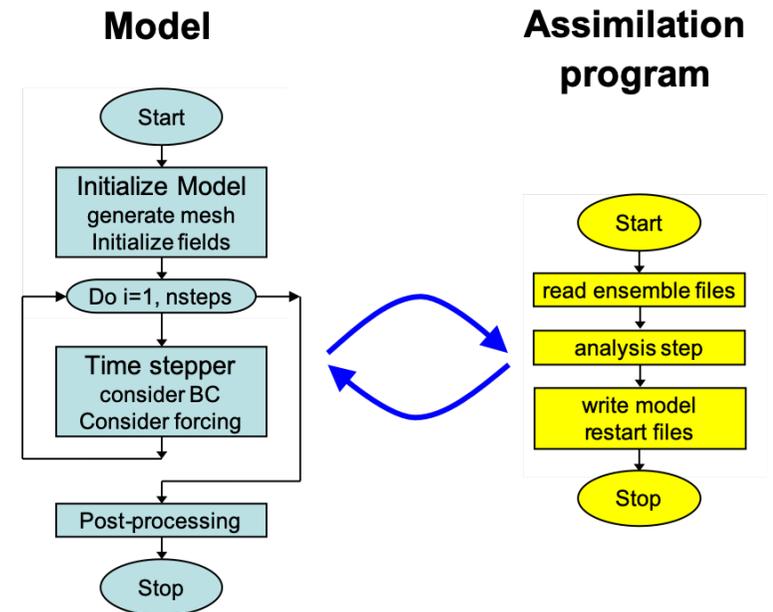
Example:

Timing from atmosphere-ocean coupled model (AWI-CM) with daily analysis step:

Model startup:	95 s	} overhead
Integrate 1 day:	33 s	
Model postprocessing:	14 s	
Analysis step:	1 s	

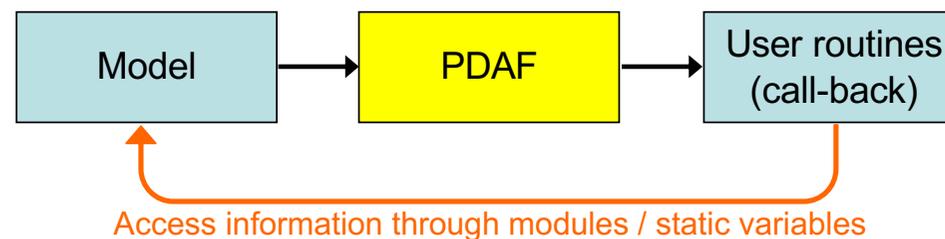
Restarting this model is ~3.5 times more expensive than integrating 1 day

→ avoid this for data assimilation (in particular re-analysis applications)



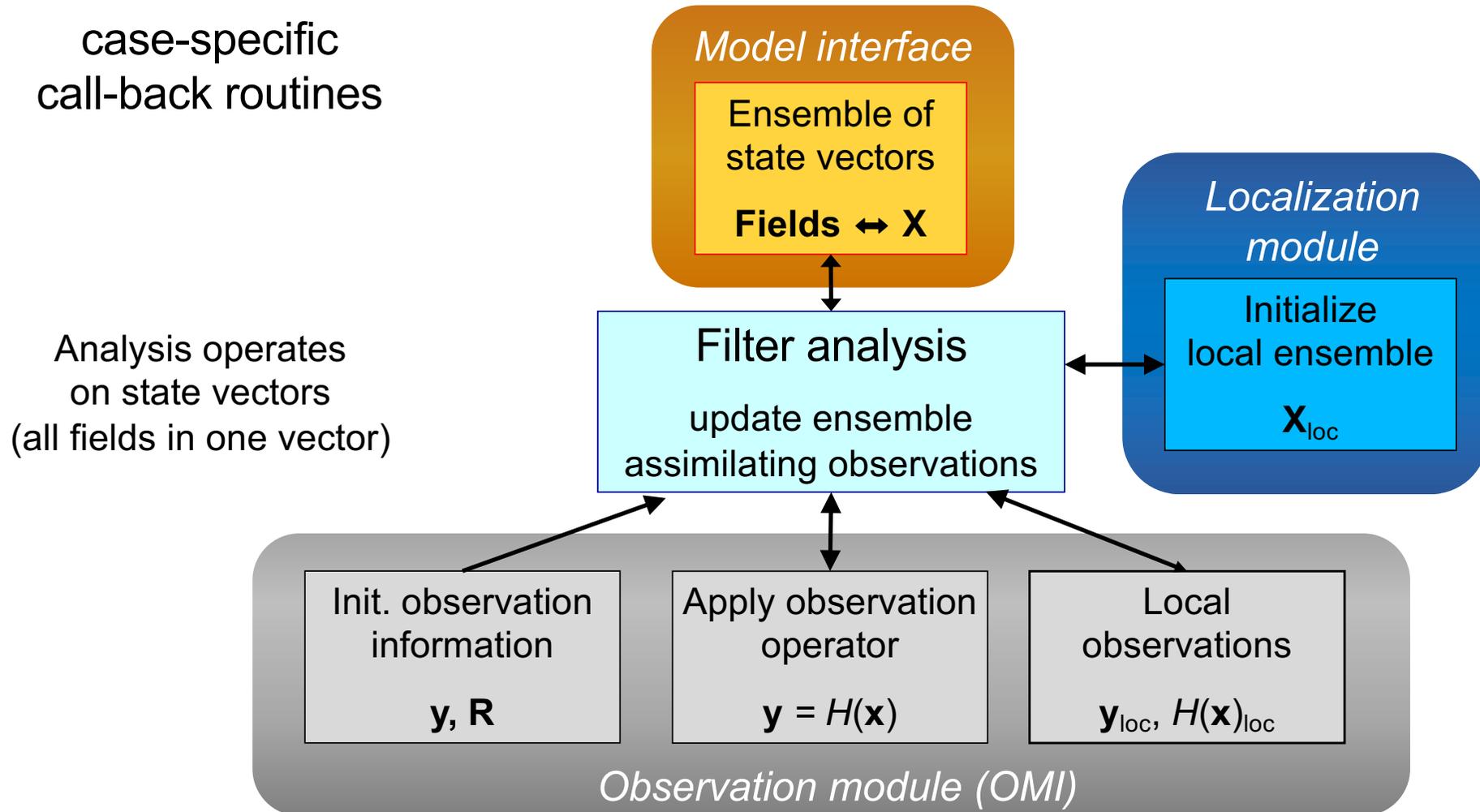
PDAF interface structure

- **Model-sided Interface:** Defined calls to PDAF routines
(called by driver program for offline coupling)
- **Case-related Interface:** User-supplied call-back routines for elementary operations:
 - transfers between model fields and ensemble of state vectors
 - observation-related operations
- **Internal Interface:** Connect to data assimilation methods
- User supplied routines can be implemented as routines of the model and can share data with it (low abstraction level)



Lars Nerger et al. – PDAF - features and developments

Implementing the Ensemble Analysis Step (Solver)



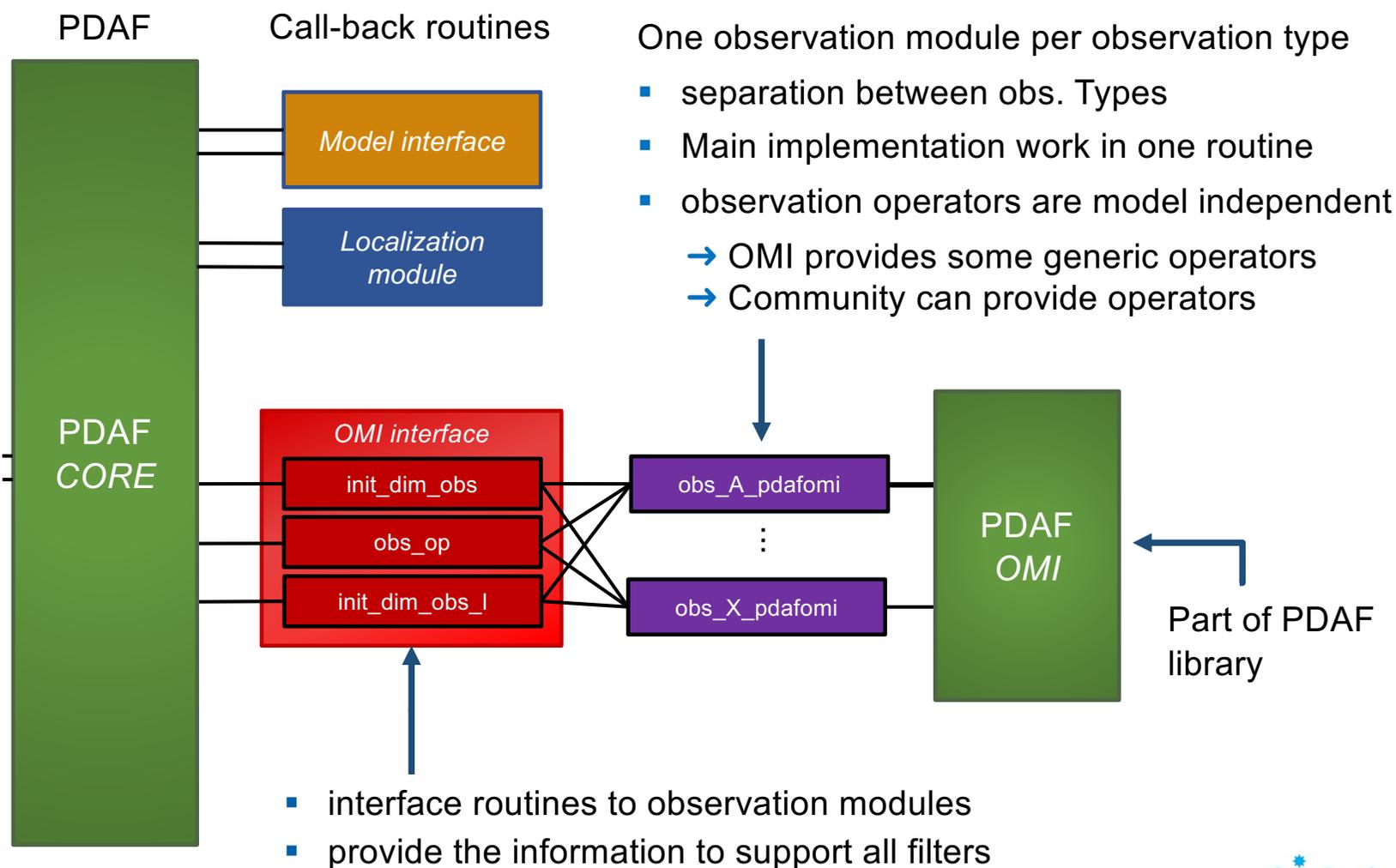
Recent and current developments

OMI: Code structure (Observation Module Infrastructure)

Structure motivated by object-oriented programming. For sake of simplicity not implemented with OOP



Part of PDAF
V1.16



Strongly Coupled DA

Strongly coupled DA:

Assimilate observation of component A into component B

PDAF supports strongly coupled DA:

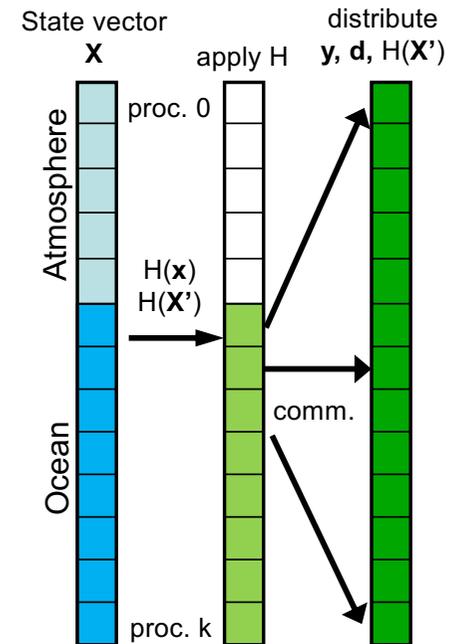
achieved by adapting MPI communicator for the filter processes

- joint state vector decomposed over the processes
- Provide observation operator that only performs MPI communication

need innovation $\mathbf{d} = \mathbf{H}(\mathbf{x}) - \mathbf{y}$
and observed ensemble perturbations $\mathbf{H}(\mathbf{X}')$

Observation operator H links different compartments

1. Compute part of \mathbf{d} and $\mathbf{H}(\mathbf{X}')$ on process 'owning' the observation
2. Communicate \mathbf{d} and $\mathbf{H}(\mathbf{X}')$ to processes for which observation is within localization radius



Observation handling in strongly coupled DA

Part of PDAF
V1.16



Ensemble 3D-Var / Hybrid 3D-Var

Activity in EU-project SEAMLESS

1D Prototype (in development):

- GOTM/FABM + ecosystem models
- DA functionality provided by PDAF

NEMO-ERGOM-PDAF data assimilation

Ensemble/Hybrid 3D-Var

- Some partners (PML, OGS) use 3D-Var
- Integrate in PDAF analogous to EnKFs/PFs
- Focus on infrastructure with optimizers as core

- Possible extension to Ensemble 4D-Var later

Included in
upcoming
release
(Dec. '21)



www.seamlessproject.org

Services based on Ecosystem
data AssiMiLation: Essential
Science and Solutions

PML (UK, lead),
AWI (Germany), IGE (France),
NERSC (Norway), OGS (Italy)
Bolding&Bruggemann (Denmark)



DA Algorithms and models in PDAF

PDAF originated from comparison studies of different filters

Filters and smoothers - *global and localized versions*

- EnKF (Evensen, 1994 + perturbed obs.)
- (L)ETKF (Bishop et al., 2001/Hunt et al. 2007)
- ESTKF (Nerger et al., 2012)
- NETF (Toedter & Ahrens, 2015)
- Particle filter
- *EnOI mode*

Model bindings

- MITgcm
- AWI-CM / FESOM

Toy models

- Lorenz-96 / Lorenz-63

Community provided:

SCHISM/ESMF
TerrSysMP-PDAF

Upcoming (*Dec. '21*):

- Ensemble 3D-Var
- Hybrid 3D-Var
- Hybrid NETF/LETKF

Upcoming:

- NEMO 4 (U Reading)
- GOTM/FABM (BB ApS)

Upcoming (*Dec. '21*):

- Lorenz-2005 II/III

PDAF: Design Considerations

1. Focus on ensemble methods

2. Efficiency:

- Direct (online/in-memory) coupling of model and data assimilation method (file-based offline coupling also supported)
- Complete parallelism in model, DA method, and ensemble integrations
- Provide common DA infrastructure with generic components

3. Ease of use:

- require just standard compilers and libraries, no containers, etc.
- just add subroutine calls into model code when combining with PDAF
- model time stepper not required to be a subroutine
- model controls the assimilation program
- case-specific routines can be implemented like model code
- simple switching between different filters and data sets
- *Separation of concerns*: model, DA methods, observations

Requirements

- Fortran compiler
- MPI library
- BLAS & LAPACK
- make

- PDAF is at least tested (often used) on various computers:
 - Notebook & Workstation: MacOS, Linux (gfortran)
 - Cray XC30/40 & CS400 (Cray ftn and ifort)
 - NEC SX-8R / SX-ACE / SX-Aurora TSUBASA
 - ATOS Bull Sequana X (ifort)
 - HPE Cray Apollo (ARM)
 - Legacy:
 - SGI Altix & UltraViolet (ifort) / IBM Power (xlf) / IBM Blue Gene/Q

Summary - PDAF: A tool for data assimilation

- a program library for ensemble modeling and data assimilation
- provides support for ensemble forecasts, DA diagnostics, and fully-implemented filter and smoother algorithms
- makes excellent use of supercomputers
- separation of concerns: model, DA methods, observations
- easy to couple to models and to code case-specific routines
- easy to add new DA methods
- efficient for research and operational use
- community code for DA methods and observations

PDAF adds DA
functionality to
models

Couple model and
PDAF within days

Get DA capability
in a month

Run DA in known
environment

Access new DA
methods by
updating PDAF

Open source:
Code, documentation, and tutorial available at
<http://pdaf.awi.de>

References

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