

MPAS-JEDI 3D/4DEnVar

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Partly based on the materials prepared by I-Han Chen



Overview

- 1. Variational Cost Function**
2. Ensemble Error Covariance Matrix
3. Overview of 3DEnVar
4. Setting up a .yaml file for 3DEnVar
5. Overview of 4DEnVar
6. Setting up a .yaml file for 4DEnVar

The Problem

We want to find the **analysis state** (x) that minimizing a cost function with an optimal fit to the **background** and **observations**.

$$J(x) = \boxed{\frac{1}{2}(x - x_b)^T B^{-1} (x - x_b)}$$

Distance to background

$$+ \boxed{\frac{1}{2}(h(x) - y)^T R^{-1} (h(x) - y)}$$

Distance to observations

Incremental Cost Function in JEDI

Liu et al. (2022)

Original-form

$$J(x) = \frac{1}{2}(x - x_b)^T B^{-1}(x - x_b) + \frac{1}{2}(h(x) - y)^T R^{-1}(h(x) - y)$$

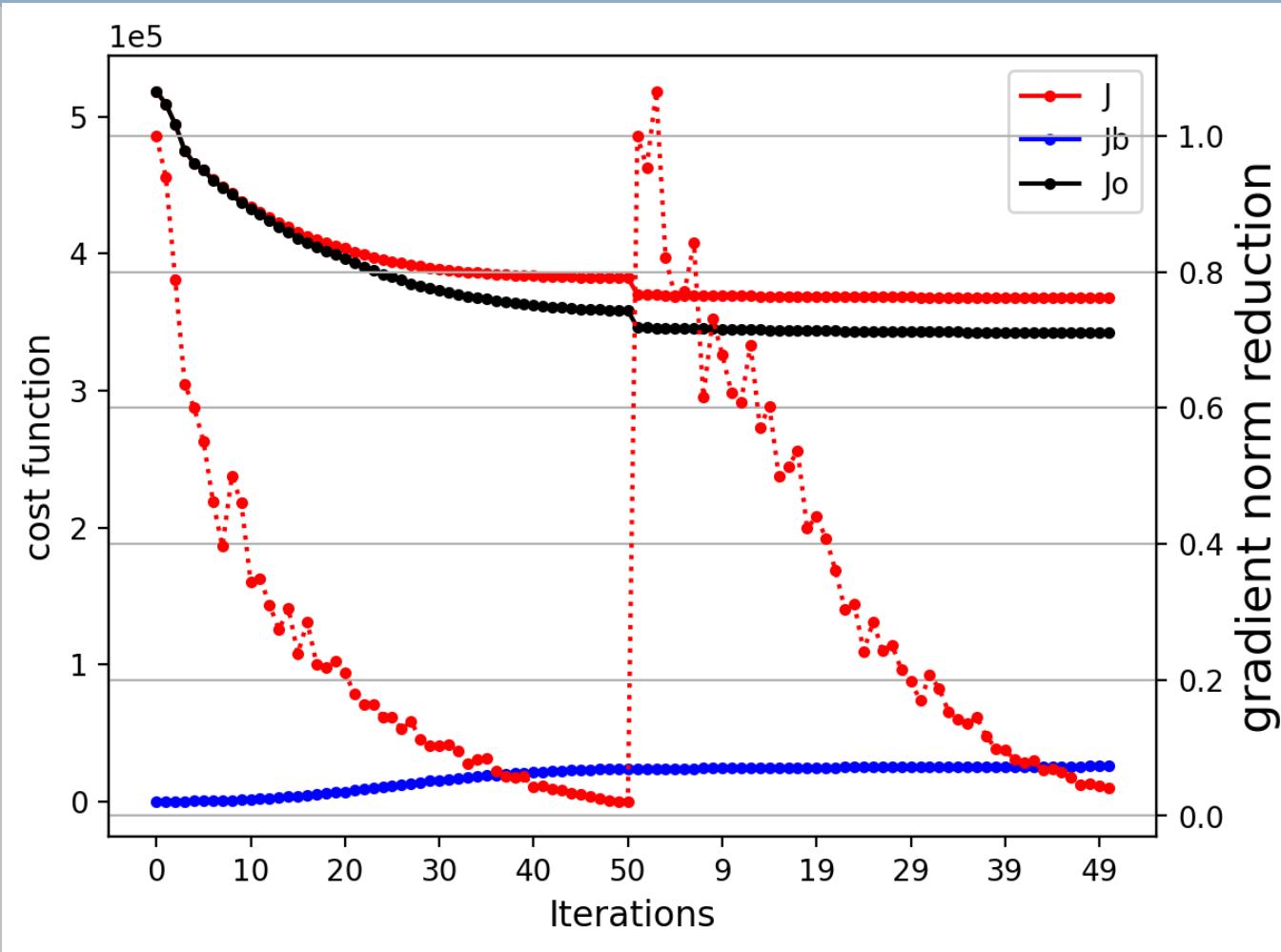
Incremental-form

$$J(\delta x) = \frac{1}{2}(\delta x - \delta x_g)^T B^{-1}(\delta x - \delta x_g) + \frac{1}{2}(H\delta x - d)^T R^{-1}(H\delta x - d)$$
$$\delta x = x - x_g \quad \delta x_g = x_b - x_g \quad d = y - h(x_g)$$

The minimization deals with increments to a known reference state

- Cost function minimizes $\delta x = x - x_g$ instead of the full state (x)
- Start from $x_g = x_b$ and $\delta x_g = 0$
- After minimization-> $x_a = x_g + \delta x$

About inner loop and outer loop



3DEnVar practice
with 2 outer loops

and 50 inner loops
for each outer loop

About dual-resolution

- Analysis increment and ensemble can be at a lower resolution than the background and analysis

Incremental-form

$$J(\delta x) = \frac{1}{2}(\delta x - \delta x_g)^T \mathbf{B}^{-1} (\delta x - \delta x_g) + \frac{1}{2}(\mathbf{H}\delta x - d)^T \mathbf{R}^{-1} (\mathbf{H}\delta x - d)$$
$$\delta x = \mathbf{x} - \mathbf{x}_g \quad \delta x_g = \mathbf{x}_b - \mathbf{x}_g \quad d = \mathbf{y} - \mathbf{h}(\mathbf{x}_g)$$

Practical session 3.3 provides a 120km-240km dual-resolution 3DEnVar test case

Appropriately assign B and R is critical

We want to find the analysis state (x) that minimizing a cost function with **an optimal fit** to the background and observations.

Distance to background

$$J(\delta x) = \frac{1}{2}(\delta x - \delta x_g)^T \mathbf{B}^{-1} (\delta x - \delta x_g)$$

Distance to observations

$$\frac{1}{2}(H\delta x - d)^T \mathbf{R}^{-1} (H\delta x - d)$$

The weighting between the two components is determined by **B** (background error) and **R** (observation error).

- A larger **B** means background is less accurate -> \mathbf{x} will get closer to observation
- A larger **R** means observation is less accurate -> \mathbf{x} will get closer to background

Two types of background error covariance (B)

$$J(\delta x) = \frac{1}{2}(\delta x - \delta x_g)^T \mathbf{B}^{-1} (\delta x - \delta x_g) + \frac{1}{2}(H\delta x - d)^T R^{-1} (H\delta x - d)$$

1. Static B

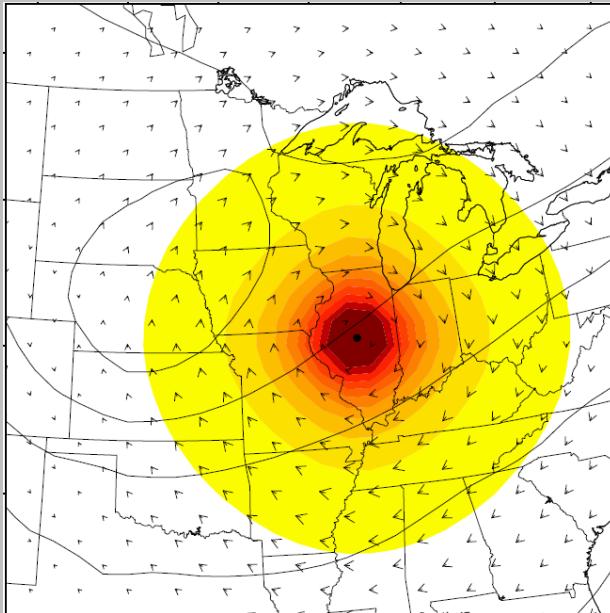
-> from statistic, does not vary with time

2. Ensemble B

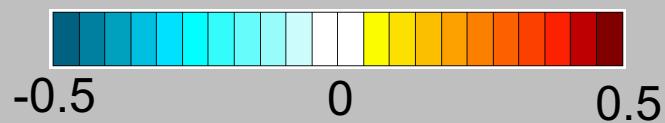
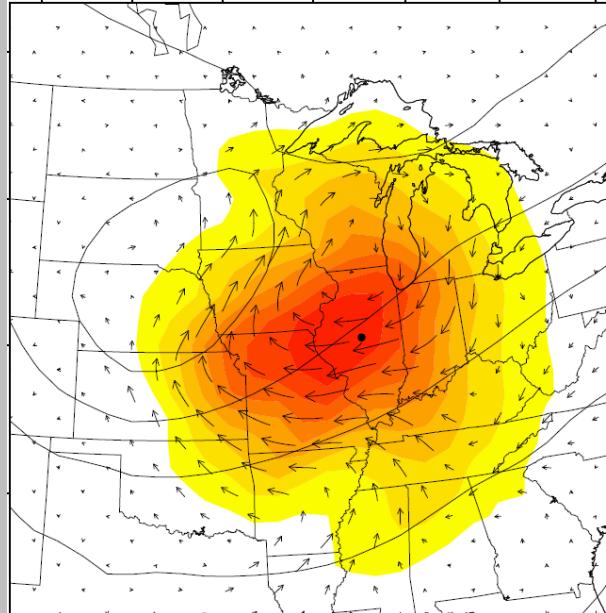
-> flow-dependent, reflect the background error in different time

Example to show the B effect (single observation tests)

Static B



Ensemble B



*Increments of temperature (shaded)
and horizontal winds (vector)*

Ensemble B:

- Errors of the day are sampled
- flow-dependent update

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Derive B matrix from an ensemble of forecasts

$$B_e = \frac{1}{n-1} \sum_{i=1}^n (\mathbf{x}_i - \bar{\mathbf{x}})(\mathbf{x}_i - \bar{\mathbf{x}})^T$$

ensemble size State variable of ensemble mean
 State variable of each ensemble member

$$B_e = \frac{1}{n-1} \sum_{i=1}^n (\delta \mathbf{x}_i)(\delta \mathbf{x}_i)^T$$

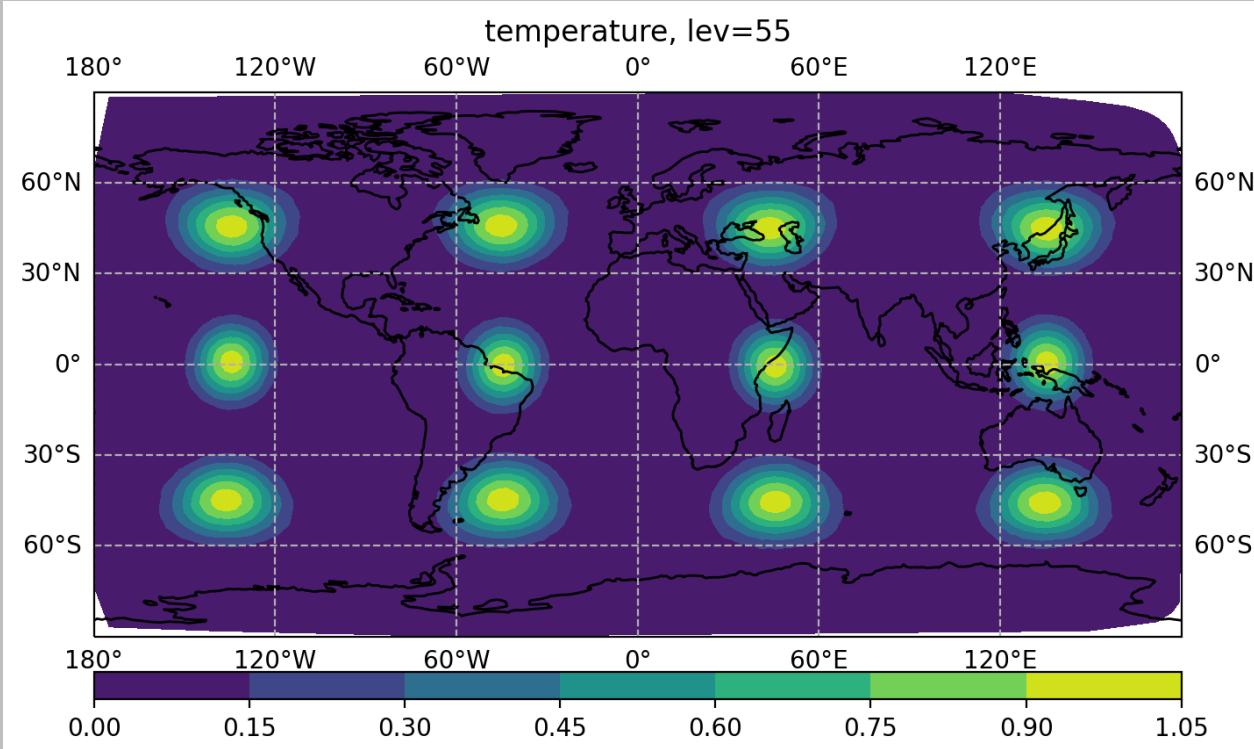
ensemble perturbation

- The ensemble mean provides an estimation of the truth
- The perturbations from the mean estimate the uncertainty, which is used to model background-error covariance matrix.

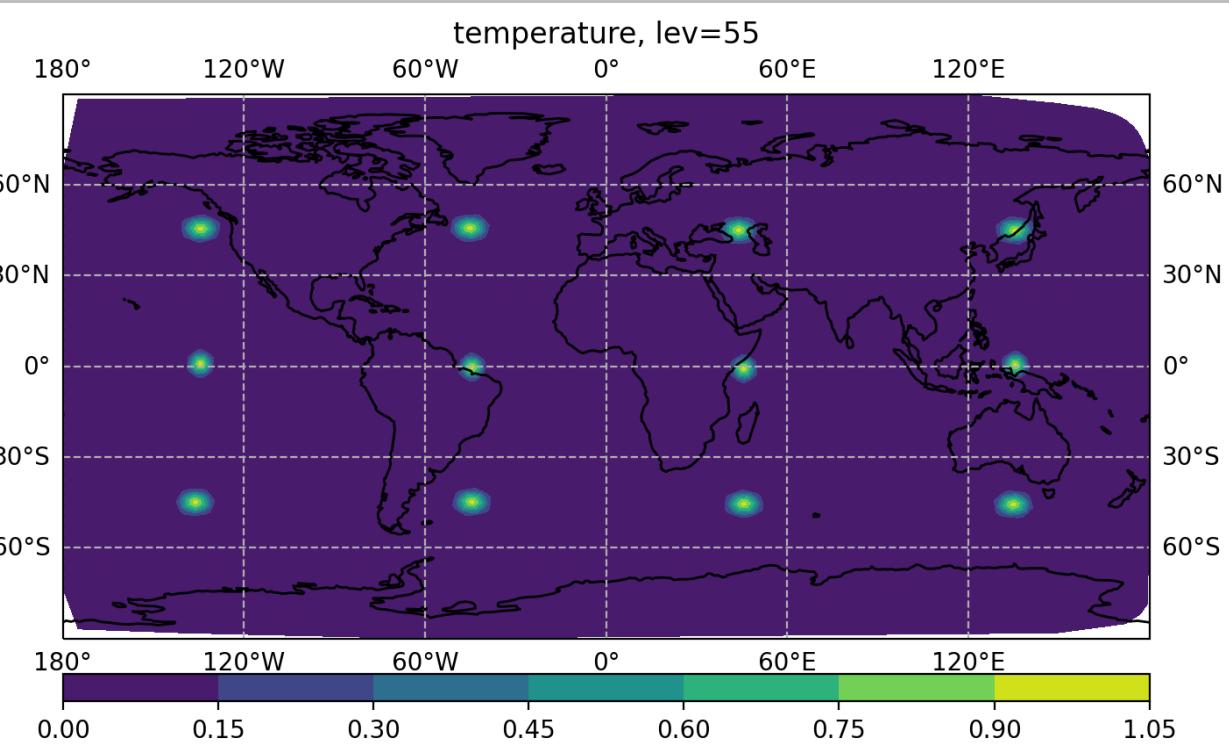
Localization of ensemble B matrix: $B = L \circ B_e$

L is a correlation matrix, to reduce spurious correlation due to limited ensemble size

larger localization scales



Smaller localization scale



EnVar uses a pure ensemble B to updates a deterministic forecast

In hybrid methods, B can be a weighted sum of static B (B_s) and ensemble B (B_e).

$$B = \beta_s B_s + \beta_e B_e$$
$$\beta_s + \beta_e = 1$$

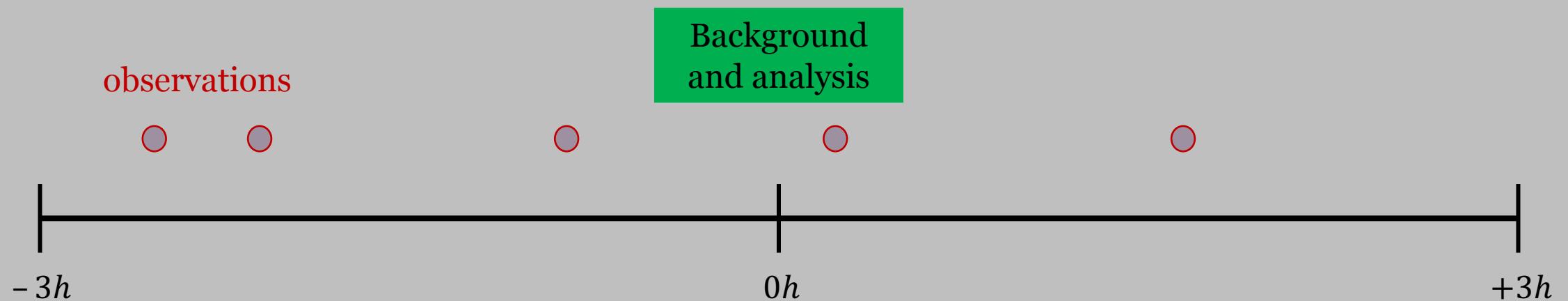
$$=0 \quad =1$$

pure ensemble B

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3DEnVar using a 6h assimilation window



- All observations in 3DEnVar are assumed to be valid at the same time as the background

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Configure the analysis time for 3DEnvar

```
_member config: &memberConfig
date: &analysisdate '2018-04-15T00:00:00Z' analysis time (center of window)
state variables: &incvars
- temperature
- spechum
- uReconstructZonal
- uReconstructMeridional
- surface_pressure
stream name: ensemble
cost function:
cost type: 3D-Var
time window:
begin: '2018-04-14T21:00:00Z' Start of assimilation window
length: PT6H length of assimilation window
geometry:
nml_file: "./Data/480km/namelist.atmosphere_2018041500"
streams_file: "./Data/480km/streams.atmosphere"
deallocate non-da fields: true
analysis variables: *incvars
background:
state variables: [temperature, spechum, uReconstructZonal, uReconstructMeridional, surface_pressure,
theta, rho, u, qv, pressure, landmask, xice, snowc, skintemp, ivgtyp, isltyp,
snowh, vegfra, u10, v10, lai, smois, tslb, pressure_p]
filename: "./Data/480km/bg/restart.2018-04-15_00.00.00.nc" background at analysis time
date: *analysisdate
```

Configure the ensemble B

```
background error:  
covariance model: ensemble  
localization:  
    localization method: SABER  
    saber central block:  
        saber block name: BUMP_NICAS  
    active variables: *incvars  
read:  
    io:  
        files prefix: Data/bump/mpas_parametersbump_loc  
drivers:  
    multivariate strategy: duplicated  
    read local nicas: true  
members:  
- filename: Data/480km/bg/ensemble/mem01/x1.2562.init.2018-04-15_00.00.00.nc  
  <<: *memberConfig  
- filename: Data/480km/bg/ensemble/mem02/x1.2562.init.2018-04-15_00.00.00.nc  
  <<: *memberConfig  
- filename: Data/480km/bg/ensemble/mem03/x1.2562.init.2018-04-15_00.00.00.nc  
  <<: *memberConfig  
- filename: Data/480km/bg/ensemble/mem04/x1.2562.init.2018-04-15_00.00.00.nc  
  <<: *memberConfig  
- filename: Data/480km/bg/ensemble/mem05/x1.2562.init.2018-04-15_00.00.00.nc  
  <<: *memberConfig
```

set ensemble B for 3DEnVar

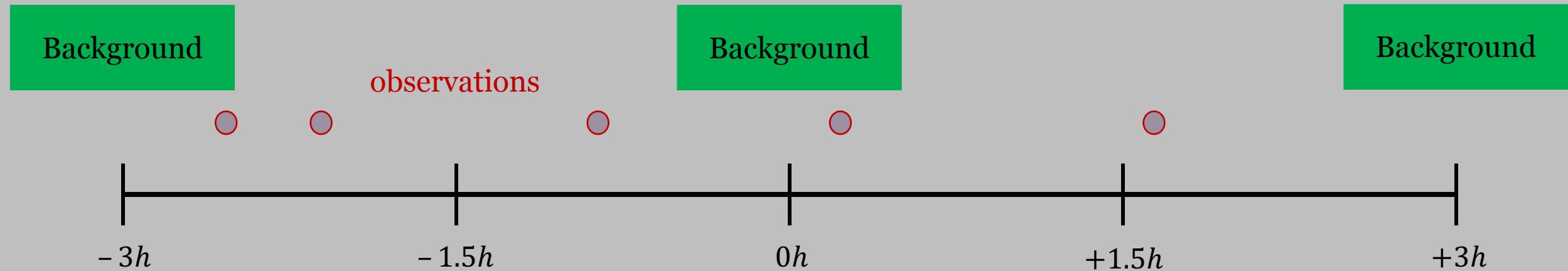
Specifying members used to compute ensemble B

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4DEnVar

$$J(x) = \frac{1}{2}(x - x_b)^T B^{-1}(x - x_b) + \frac{1}{2} \sum_{k=1}^K (Hx_k - y_k)^T R_k^{-1}(Hx_k - y_k)$$

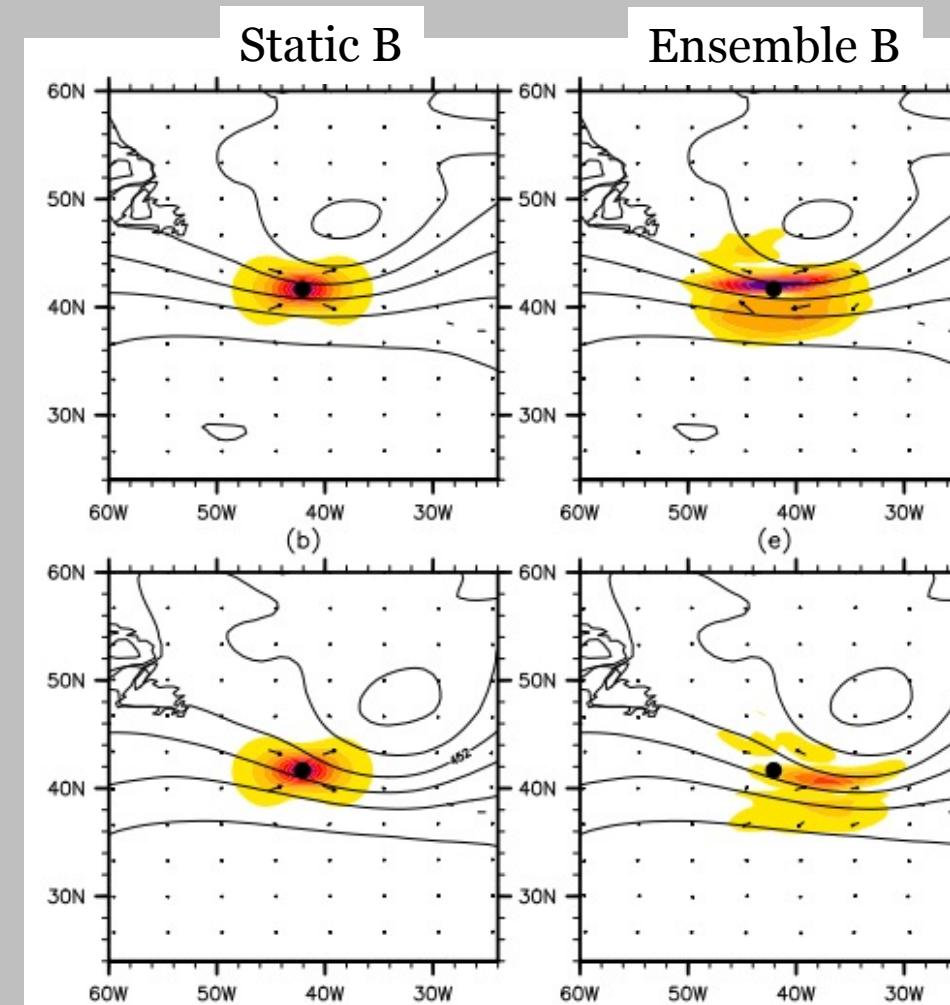


- All observations in 4DEnVar are binned within a smaller subwindow and innovations $(Hx - y_o)$ are calculated relative to background valid at that time.
- Ensemble needed at the center of each subwindow (K ensemble required).

The 4D ensemble *B* is used to propagate the innovation

Start of window

end of window



Lorenc et al. (2015)

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Configure the analysis times for 4DEnvar

```
_member config 1: &memberConfig1
  date: &date1 '2018-04-14T21:00:00Z'
  state variables: &incvars
    - temperature
    - spechum
    - uReconstructZonal
    - uReconstructMeridional
    - surface_pressure
  stream name: ensemble
_member config 2: &memberConfig2
<<: *memberConfig1
  date: &date2 '2018-04-15T00:00:00Z'
_member config 3: &memberConfig3
<<: *memberConfig1
  date: &date3 '2018-04-15T03:00:00Z'
cost function:
  cost type: 4D-Ens-Var
  time window:
    begin: '2018-04-14T21:00:00Z'
    length: PT6H
    subwindow: PT3H
```

subwindow1

test/testinput/4denvar_bumploc.yaml

subwindow2

subwindow3

Background needed for each subwindow

```
background:  
  states:  
    - state variables: &stvars  
      [temperature, spechum, uReconstructZonal, uReconstructMeridional, surface_pressure,  
       theta, rho, u, qv, pressure, landmask, xice, snowc, skintemp, ivgtyp, isltyp,  
       snowh, vegfra, u10, v10, lai, smois, tslb, pressure_p]  
      filename: "./Data/480km/bg/restart.2018-04-14_21.00.00.nc"  
      date: *date1  
    - state variables: *stvars  
      filename: "./Data/480km/bg/restart.2018-04-15_00.00.00.nc"  
      date: *date2  
    - state variables: *stvars  
      filename: "./Data/480km/bg/restart.2018-04-15_03.00.00.nc"  
      date: *date3
```

bg (subwindow 1)

bg (subwindow 2)

bg (subwindow 3)

Configure the ensemble B

```
background error:  
    covariance model: ensemble  
    localization:  
        localization method: SABER  
        saber central block:  
            saber block name: BUMP_NICAS  
            active variables: *incvars  
            read:  
                io:  
                    files prefix: Data/bump/mpas_parametersbump_loc  
            drivers:  
                multivariate strategy: duplicated  
                read local nicas: true
```

set ensemble B for 4DEnVar

ensemble files needed for each subwindow

```
members:  
- states:  
  - filename: Data/480km/bg/ensemble/mem01/x1.2562.init.2018-04-14_21.00.00.nc  
    <<: *memberConfig1  
  - filename: Data/480km/bg/ensemble/mem01/x1.2562.init.2018-04-15_00.00.00.nc  
    <<: *memberConfig2  
  - filename: Data/480km/bg/ensemble/mem01/x1.2562.init.2018-04-15_03.00.00.nc  
    <<: *memberConfig3  
- states:  
  - filename: Data/480km/bg/ensemble/mem02/x1.2562.init.2018-04-14_21.00.00.nc  
    <<: *memberConfig1  
  - filename: Data/480km/bg/ensemble/mem02/x1.2562.init.2018-04-15_00.00.00.nc  
    <<: *memberConfig2  
  - filename: Data/480km/bg/ensemble/mem02/x1.2562.init.2018-04-15_03.00.00.nc  
    <<: *memberConfig3  
- states:  
  - filename: Data/480km/bg/ensemble/mem03/x1.2562.init.2018-04-14_21.00.00.nc  
    <<: *memberConfig1  
  - filename: Data/480km/bg/ensemble/mem03/x1.2562.init.2018-04-15_00.00.00.nc  
    <<: *memberConfig2  
  - filename: Data/480km/bg/ensemble/mem03/x1.2562.init.2018-04-15_03.00.00.nc  
    <<: *memberConfig3  
- states:  
  - filename: Data/480km/bg/ensemble/mem04/x1.2562.init.2018-04-14_21.00.00.nc  
    <<: *memberConfig1  
  - filename: Data/480km/bg/ensemble/mem04/x1.2562.init.2018-04-15_00.00.00.nc  
    <<: *memberConfig2  
  - filename: Data/480km/bg/ensemble/mem04/x1.2562.init.2018-04-15_03.00.00.nc  
    <<: *memberConfig3  
- states:  
  - filename: Data/480km/bg/ensemble/mem05/x1.2562.init.2018-04-14_21.00.00.nc  
    <<: *memberConfig1  
  - filename: Data/480km/bg/ensemble/mem05/x1.2562.init.2018-04-15_00.00.00.nc  
    <<: *memberConfig2  
  - filename: Data/480km/bg/ensemble/mem05/x1.2562.init.2018-04-15_03.00.00.nc  
    <<: *memberConfig3
```

A simpler way to configure ensemble input

```
members from template:  
  template:  
    states:  
      - state variables: *incvars  
        date: 2018-04-14T21:00:00Z  
        filename: ../ensemble/2018041418/%iMember%/mpasout.2018-04-14_21.00.00.nc  
      - state variables: *incvars  
        date: 2018-04-15T00:00:00Z  
        filename: ../ensemble/2018041418/%iMember%/mpasout.2018-04-15_00.00.00.nc  
      - state variables: *incvars  
        date: 2018-04-15T03:00:00Z  
        filename: ../ensemble/2018041418/%iMember%/mpasout.2018-04-15_03.00.00.nc  
    pattern: %iMember%  
    start: 1  
    zero padding: 2  
    nmembers: 20
```

See 4denvar.yaml in practical session 3.4

Nodes/Cores setting

3DEnVar

```
#!/bin/tcsh -f  
#PBS -l select=1:ncpus=64:mpiprocs=64:mem=235GB
```

4DEnVar with 3 sub-windows

```
#!/bin/tcsh -f  
#PBS -l select=3:ncpus=64:mpiprocs=64:mem=235GB
```

References

- Liu, Z., and Coauthors, 2022: Data assimilation for the Model for Prediction Across Scales - Atmosphere with the Joint Effort for Data assimilation Integration (JEDI-MPAS 1.0.0): EnVar implementation and evaluation. *Geosci. Model Dev.*, **15**, 7859–7878, <https://doi.org/10.5194/gmd-15-7859-2022>.
- Lorenc, A. C., N. E. Bowler, A. M. Clayton, S. R. Pring, and D. Fairbairn, 2015: Comparison of hybrid-4DEnVar and hybrid-4DVar data assimilation methods for global NWP. *Mon. Weather Rev.*, **143**, 212–229, <https://doi.org/10.1175/MWR-D-14-00195.1>.