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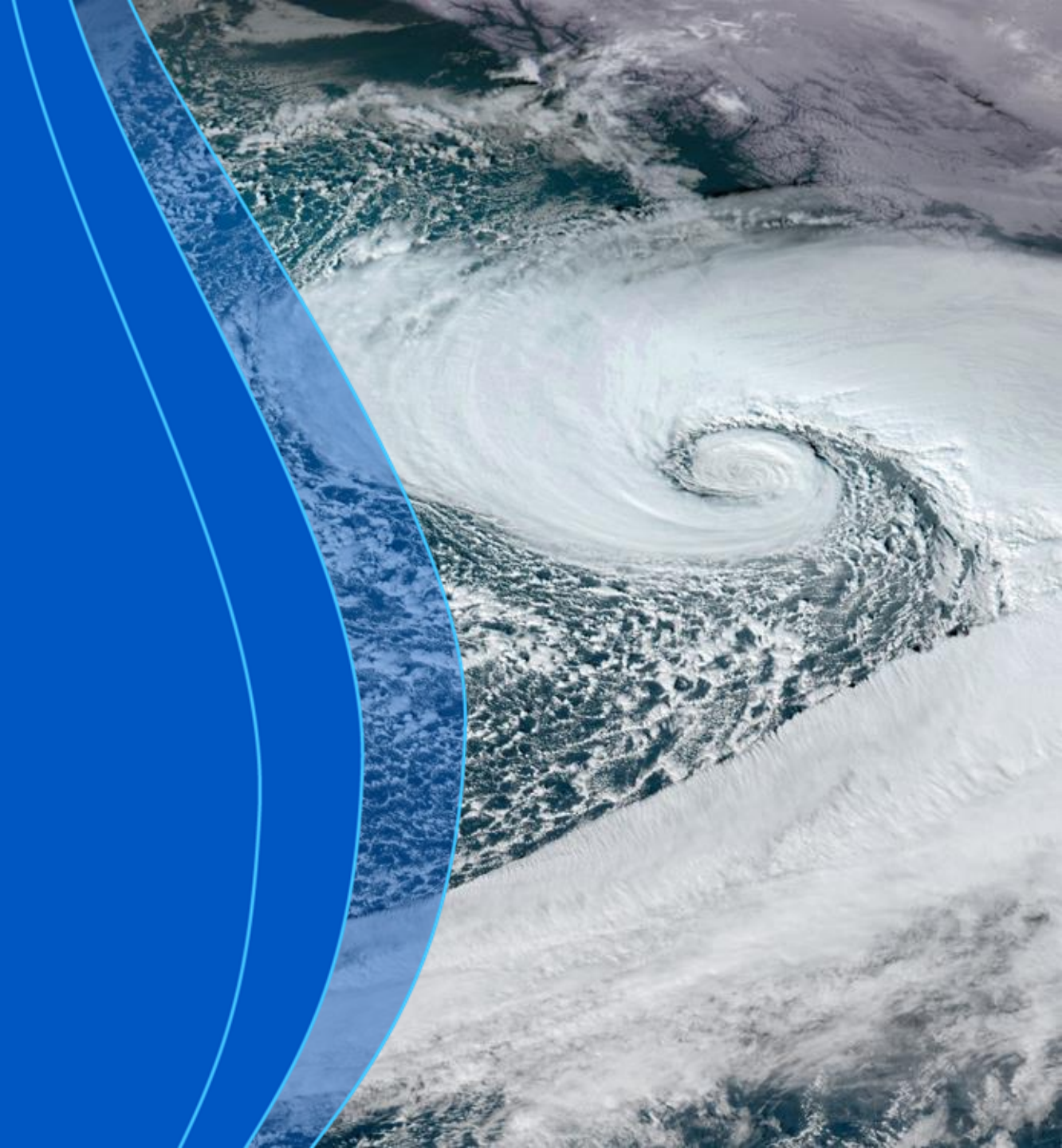
August 13, 2026

MPAS-JEDI 3D/4D_{EnVar}

Materials prepared by Jake (Zhiquan) Liu, I-Han Chen, and Robert Nystrom

Presented by Junmei Ban

MPAS-A and MPAS-JEDI Tutorial, Boulder CO, AUGUST 10-14, 2026



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 to solve

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

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

Distance to background

Distance to observations

B : background error covariance matrix

R : observation error covariance matrix

Incremental Cost Function in JEDI

Original-form:

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

Incremental-form:

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

$$\delta x = x - x_g \quad \delta x_g = x_b - x_g \quad d = y - h(x_g)$$

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 $\rightarrow x_a = x_g + \delta x$

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

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

$$\delta x = x - x_g \quad \delta x_g = x_b - x_g \quad d = y - h(x_g)$$

- Hofx : based on the high-resolution nonlinear background state
- inner-loop minimization : low-resolution
- After converge, the low-res analysis increment is interpolated up to high resolution, and added to the high-res background state, to obtain the final analysis.

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.

$$J(\delta x) = \underbrace{\frac{1}{2}(\delta x - \delta x_g)^T \mathbf{B}^{-1}(\delta x - \delta x_g)}_{\text{Distance to background}} + \underbrace{\frac{1}{2}(H\delta x - d)^T \mathbf{R}^{-1}(H\delta x - d)}_{\text{Distance to observations}}$$

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

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

Two types of background error covariance (B)



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

1. Static B

-> from statistic (e.g. a month of model forecasts), does not vary with time

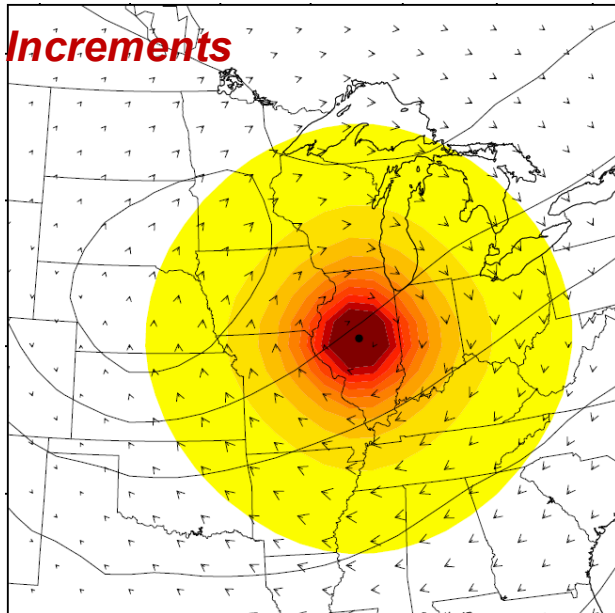
2. Ensemble B

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

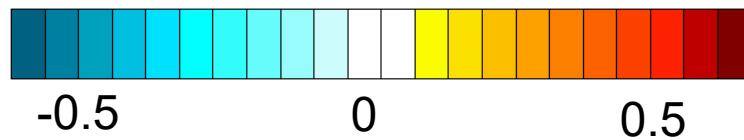
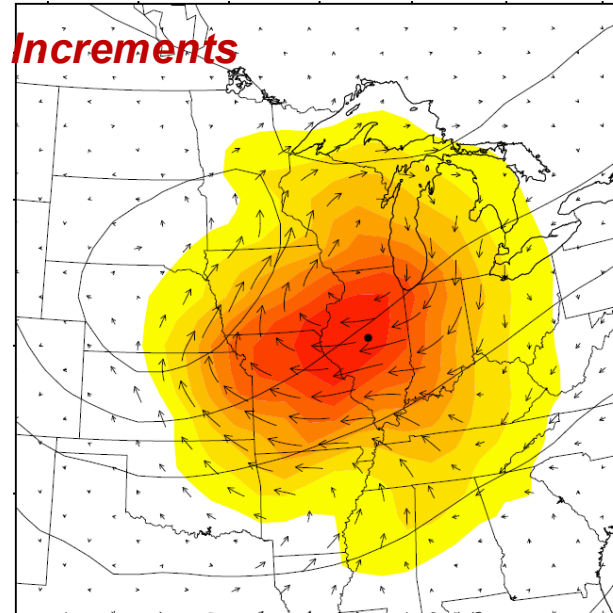
Example to show the **B** effect (Single observation tests)

Assimilate a single Temperature observation

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

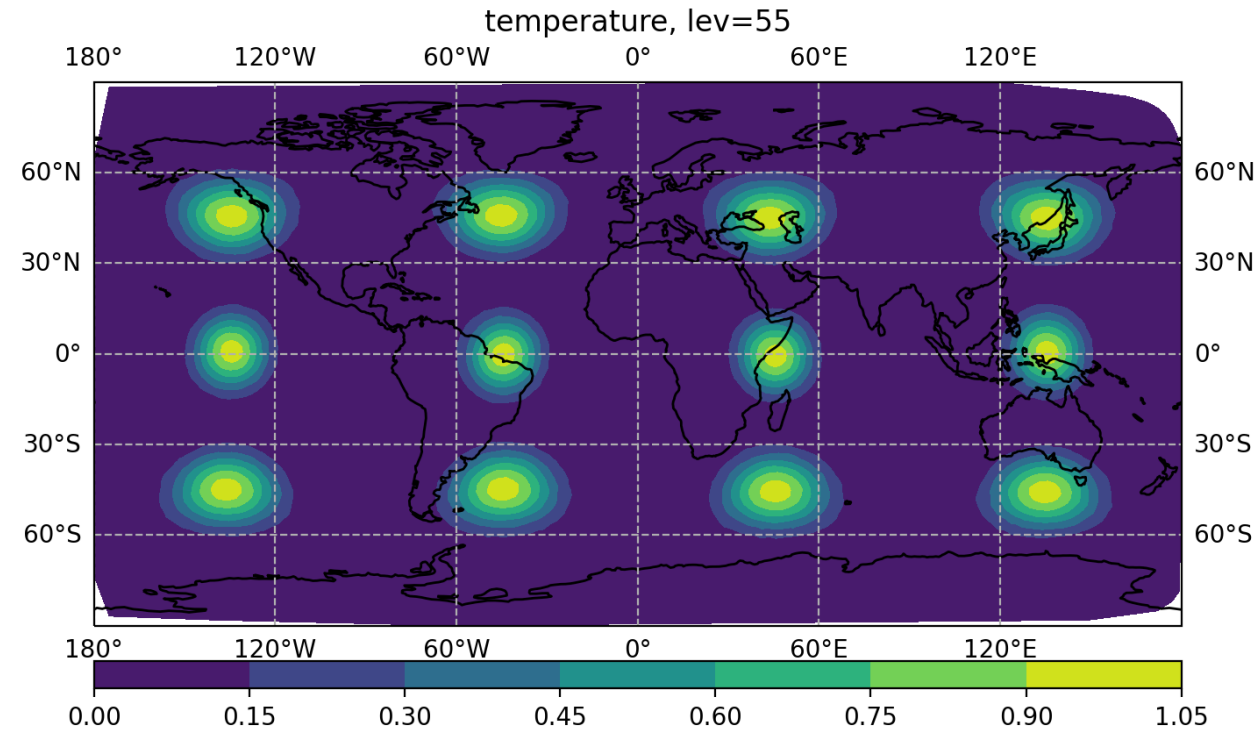
- Sample covariance matrix
- The ensemble mean provides an estimation of the truth
- The perturbations from the mean estimate the uncertainty, which is used to model background-error covariance matrix.
- Ensemble spread is the standard deviation from ensemble mean.

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

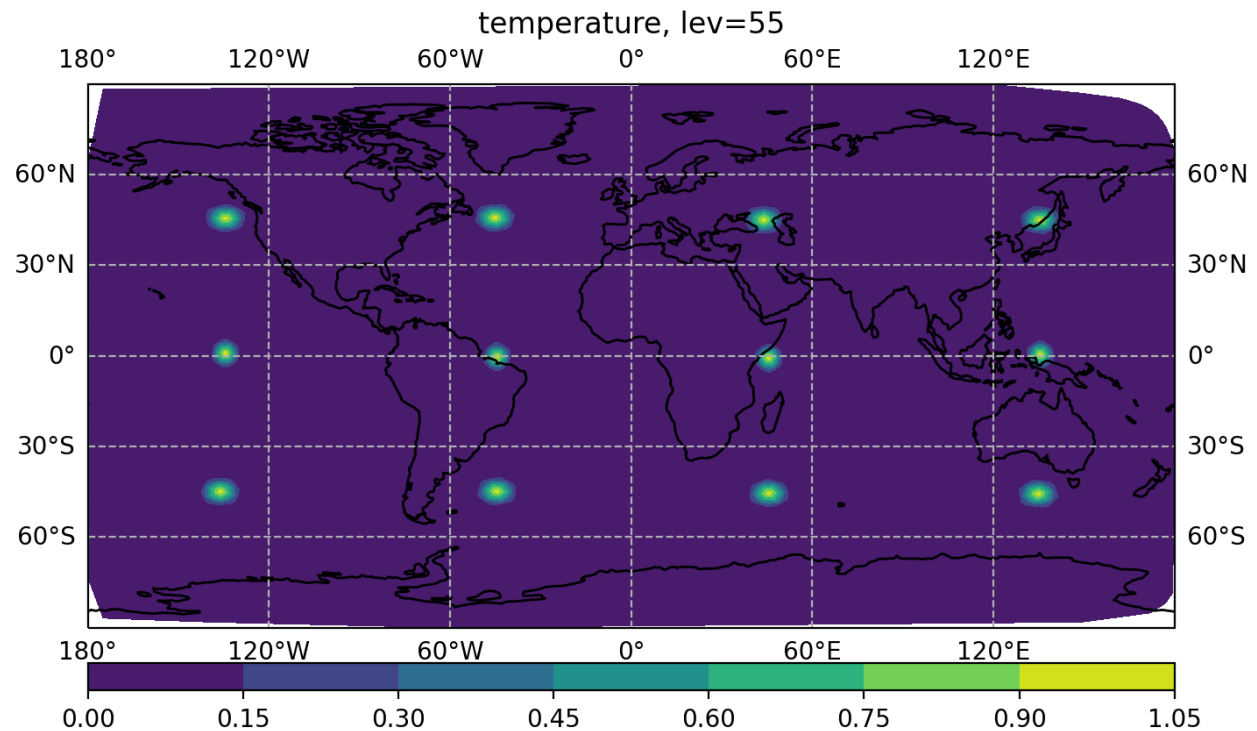


L is a correlation matrix, to reduce spurious correlation due to limited ensemble size

larger localization scales



Smaller localization scale



Influence a a small area

EnVar uses a pure ensemble B to updates a deterministic forecast

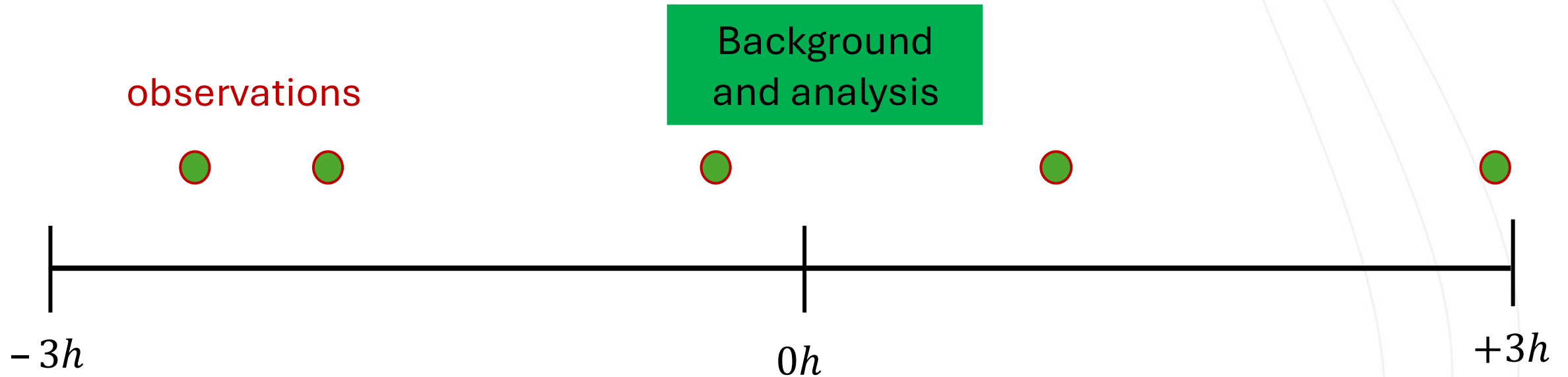
In hybrid methods, B can be a weighted sum of static B (B_s) and ensemble B (B_e).

$$B = \beta_s B_s + \beta_e B_e$$
$$\underbrace{\beta_s}_{=0} + \underbrace{\beta_e}_{=1} = 1$$

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.
- Innovation ($Hx - y_o$) is calculated relative to background at that single time

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Configure the analysis time for 3DEnvvar

```
_member config: &memberConfig
```

```
date: &analysisdate '2018-04-15T00:00:00Z'
```

analysis time (center of window)

```
state variables: &incvars
```

- air_temperature
- water_vapor_mixing_ratio_wrt_moist_air
- eastward_wind
- northward_wind
- air_pressure_at_surface

```
stream name: ensemble
```

```
cost function:
```

```
cost type: 3D-Var
```

```
time window:
```

```
begin: '2018-04-14T21:00:00Z'
```

```
length: PT6H
```

Start of assimilation window
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: [air_temperature, water_vapor_mixing_ratio_wrt_moist_air...]
```

```
filename: "./Data/480km/bg/restart.2018-04-15_00.00.00.nc"
```

```
date: *analysisdate
```

background at analysis time

Configure the ensemble B

background error:

```
covariance model: ensemble
```

set ensemble *B* for 3DEnVar

```
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
```

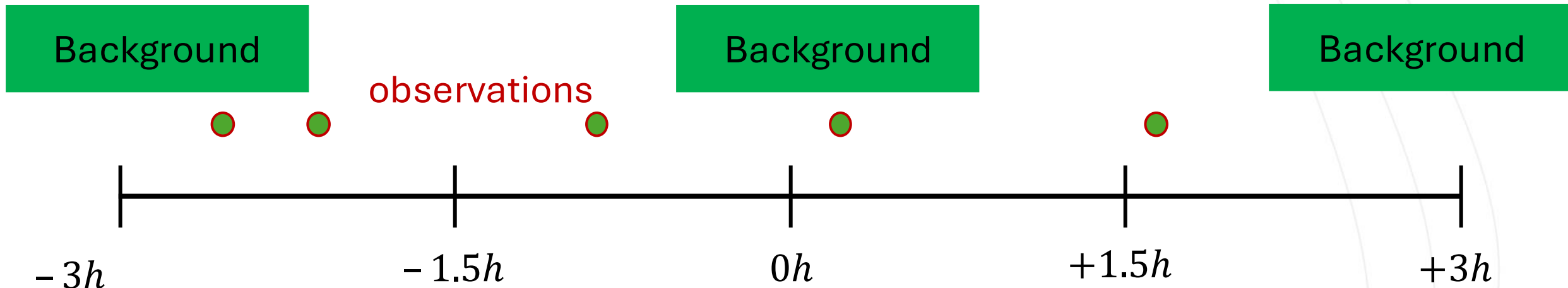
```
    read local nicas: true
```

Specifying members used to compute ensemble B

```
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

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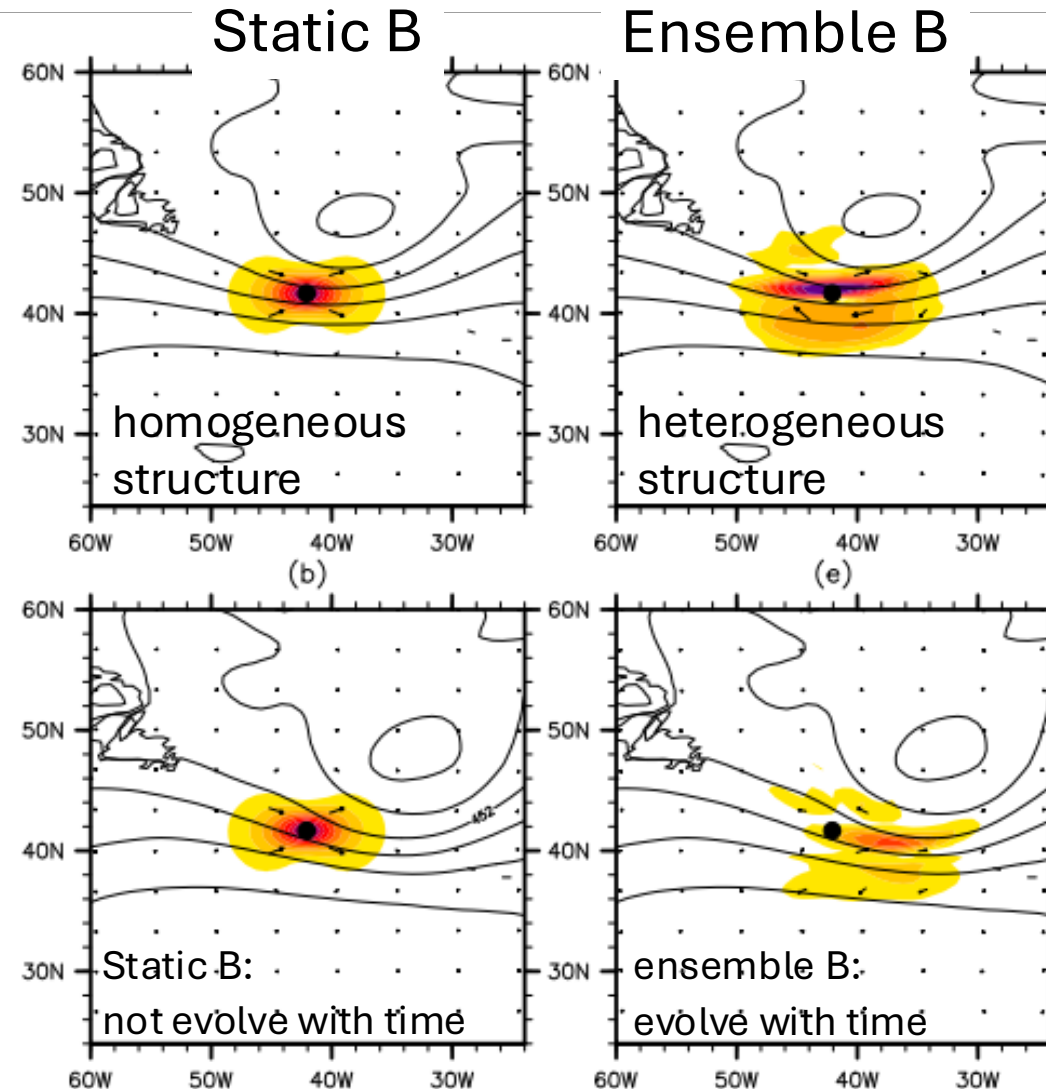
- All observations in 4DEnVar are binned within a smaller subwindow and innovations ($Hx - y_o$) are calculated relative to background valid at that time.
- Ensemble needed at the center of each subwindow (K ensemble required).

The 4D ensemble B is used to propagate the innovation

Analysis increment

Start of window

end of window



increments have moved downstream by the end of the assimilation window

Lorenc et al. (2015)

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6. **Setting up a .yaml file for 4DEnVar**

Configure the analysis times for 4DVar

- `_member config 1: &memberConfig1`

`date: &date1 '2018-04-14T21:00:00Z'`

subwindow1

`state variables: &incvars`

- `air_temperature`

- `water_vapor_mixing_ratio_wrt_moist_air`

- `eastward_wind`

- `northward_wind`

- `air_pressure_at_surface`

`stream name: ensemble`

test/testinput/4denvar_bumploc.yaml

`_member config 2: &memberConfig2`

`<<: *memberConfig1`

`date: &date2 '2018-04-15T00:00:00Z'`

subwindow2

`_member config 3: &memberConfig3`

`<<: *memberConfig1`

`date: &date3 '2018-04-15T03:00:00Z'`

subwindow3

`cost function:`

`cost type: 4D-Ens-Var`

`time window:`

`begin: '2018-04-14T21:00:00Z'`

`length: PT6H`

`subwindow: PT3H`

Background needed for each subwindow



background:

states:

- state variables: `&stvars`

[air_temperature, water_vapor_mixing_ratio_wrt_moist_air, eastward_wind, northward_wind,
air_pressure_at_surface, air_potential_temperature, dry_air_density, u,
water_vapor_mixing_ratio_wrt_dry_air, air_pressure, landmask, seaice_fraction, snowc ...]

filename: `"./Data/480km/bg/restart.2018-04-14_21.00.00.nc"`

date: `*date1`

bg (subwindow 1)

- state variables: `*stvars`

filename: `"./Data/480km/bg/restart.2018-04-15_00.00.00.nc"`

date: `*date2`

bg (subwindow 2)

- state variables: `*stvars`

filename: `"./Data/480km/bg/restart.2018-04-15_03.00.00.nc"`

date: `*date3`

bg (subwindow 3)

Configure the ensemble B

```
background error:
```

```
covariance model: ensemble
```

set ensemble *B* for 4DEnVar

```
localization:
```

```
localization method: SABER
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```
saber central block:
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saber block name: BUMP_NICAS
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active variables: *incvars
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```
read:
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```
io:
```

```
files prefix: Data/bump/mpas_parametersbump_loc
```

```
drivers:
```

```
multivariate strategy: duplicated
```

```
read local nicas: true
```

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 4.4

3DEnVar

```
...  
#PBS -l select=1:ncpus=64:mpiprocs=64:mem=235GB
```

4DEnVar with 3 sub-windows

```
...  
#PBS -l select=3:ncpus=64:mpiprocs=64:mem=235GB
```

- 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>.
- Nystrom, R. G., Snyder, C., Liu, Z., Jung, B. J., Ban, J., & Banos, I. H. (2025). A hybrid four-dimensional variational data assimilation system for the Model for Prediction Across Scales (MPAS-atmosphere): Leveraging the Joint Effort for Data assimilation Integration (JEDI). *Journal of Advances in Modeling Earth Systems*, 17, e2025MS005183. <https://doi.org/10.1029/2025MS005183>