



Using JEDI for Atmospheric Composition applications: A technical introduction

Jérôme Barré¹, Maryam Abdi-Oskoue¹
¹ NASA, GMAO



What this tutorial covers?

1. **Introduction:** Understand the basic concepts in JEDI. The object-oriented benefits
2. **Lecture 1:** Understand the IODA observation space format - learn to use and develop IODA converters
3. **Lecture 2:** Use observations in JEDI. Monitoring demonstration H(x) experiment example
4. **Lecture 3:** Assimilate observations in JEDI. EnVar experiment example

What this tutorial does not cover (yet)?

1. How to build/compile your own JEDI system
2. How to add new feature in JEDI
3. How to use data assimilation workflows

Today: Agenda



Introduction: Understand the basic concepts in JEDI. The object-oriented benefits 5 mins

Questions 5 mins

Lecture 1: IODA observation space format 20mins

Questions 10mins

Lecture 2: Using observations and H(x) experiment 30 mins

Questions 10mins

Lecture 3: Assimilate observations and EnVar experiments 30mins

Closing remarks 5mins

Break 15mins

Open hour, Q&A, trying yourself! 45 mins

Please save your questions for the dedicated question times

Intro: JEDI – Design Principles and Notional Status



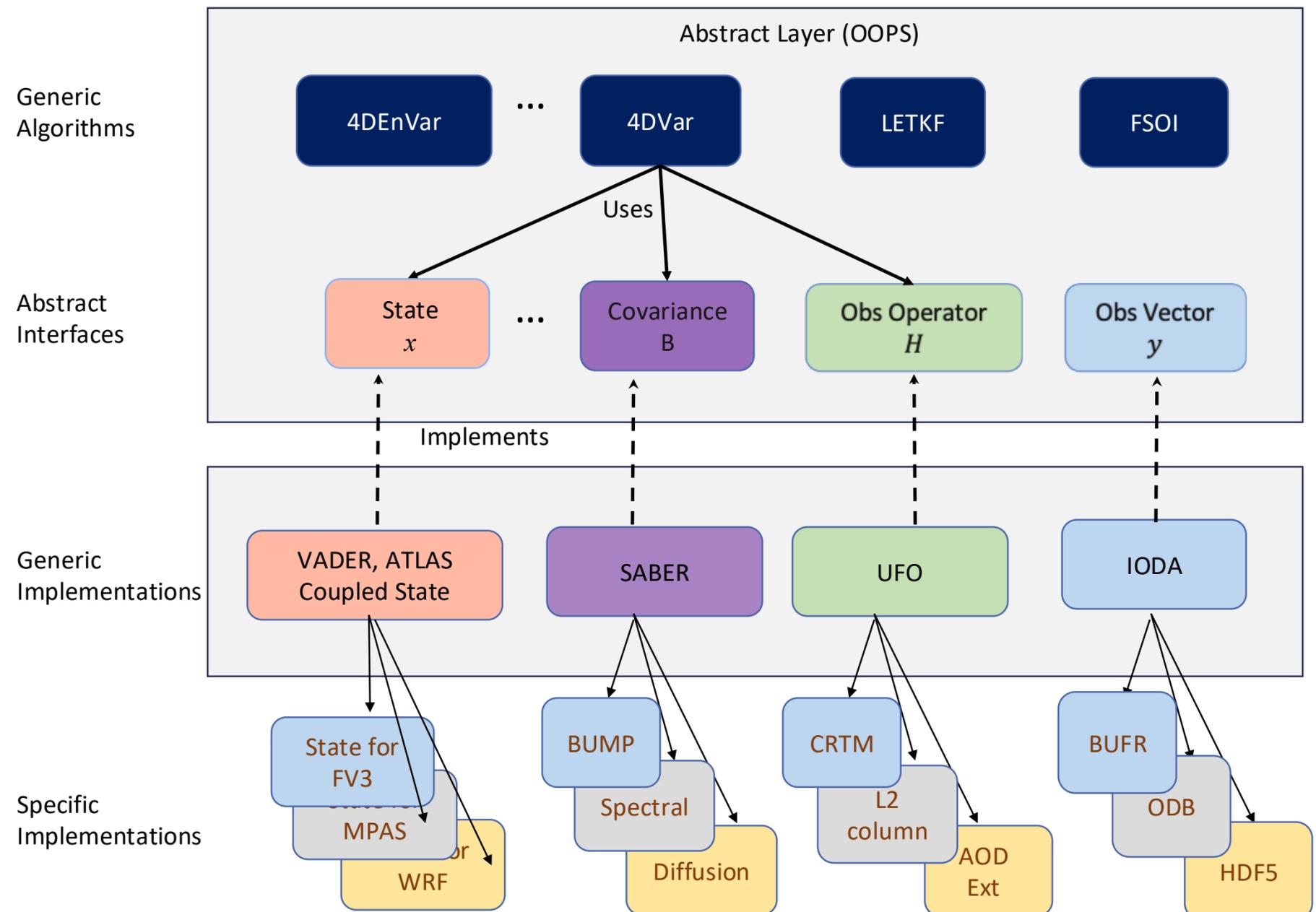
$$\Delta x_a = \mathbf{B}\mathbf{M}^T\mathbf{H}^T(\mathbf{H}\mathbf{M}\mathbf{B}\mathbf{M}^T\mathbf{H}^T + \mathbf{R})^{-1} (y_o - \mathbf{H}(\mathbf{M}(x_b)))$$

JEDI design cleanly separates the concepts of:

- DA algorithms
- model specifics
- observation specifics

Abstract interfaces are used by DA algorithms and implemented by specific models and observations.

Generic code eliminates duplication of development, testing and maintenance effort, across teams & Earth-system components.



Intro: Do I need to know C++ to work with JEDI?



Not really... but if you do, that's a ++

If you are using JEDI there's a high chance that what you need is already coded there. You often don't need to modify core code.

You need to understand the flow of information in JEDI but LLMs (e.g. chat-GSFC) can help you with that!

During this session we will go over concept of C++ classes and manipulation of YAMLS in JEDI

Intro: Object-Oriented JEDI (OOPS) and YAML driven execution



OOPS in JEDI:

- DA concepts are **classes**: e.g. Geometry, State, Model, ObsSpace, CostFunction
- Abstract base classes define required **methods**, which are concrete implementations selected at runtime: ModelInterfaces, ObsOperators, BackgroundError, etc...
- **Algorithms** compose and drive objects inside specific implementations

YAMLs controls the DA "objects":

- YAML "wires" which objects exist in the execution (model, obsop, filters, covariance, solver, etc..) and their options
- **Same YAML patterns** and keys across applications.
- **Same executable** with different experiments: edit YAML to change wiring, functionality and tuning

```
1  cost function:
2    cost type: 3D-Var
3    time window:
4      begin: 2020-09-03T15:00:00Z
5      length: PT6H
6    analysis variables:
7    - volume_mixing_ratio_of_no2
8    geometry:
9      fms initialization:
10     namelist filename: Data/fv3files/fmsmpp.nml
11     akbk: Data/inputs/fv3files/akbk72.nc4
12     npx: 13
13     npy: 13
14     npz: 72
15   background:
16     datetime: 2020-09-03T18:00:00Z
17     filetype: cube sphere history
18     datapath: Data/inputs/geos_c12
19     filename: geos_cf.bkg.%yyyy%mm%dd_%hh%MM%ssz.nc4
20     state variables:
21     - air_pressure_thickness
22     - volume_mixing_ratio_of_no2
23     - volume_mixing_ratio_of_no
24     - volume_mixing_ratio_of_o3
25     - air_pressure_at_surface
26     - water_vapor_mixing_ratio_wrt_moist_air
27     field io names: &field_io_names
28     volume_mixing_ratio_of_no2: N02
29     air_pressure_thickness: DELP
30     volume_mixing_ratio_of_no: "N0"
31     volume_mixing_ratio_of_o3: 03
32     air_pressure_at_surface: PS
```

Intro: Object-Oriented JEDI (OOPS) and YAML driven execution



DA methods require the same limited number of entities. The development of these entities should be reusable. Entities = basic (abstract) classes that define the system. A typical Variational DA YAML for example has general classes (see below). Inside each of those you will have more specific implementations and YAML options. DA abstract classes generally are:

- cost function:**
- cost type:**
- time window:**
- analysis variables:**
- geometry:**
- background:**
- background error:**
- observations:**
- output:**
- variational:**

```
1  cost function:
2    cost type: 3D-Var
3    time window:
4      begin: 2020-09-03T15:00:00Z
5      length: PT6H
6    analysis variables:
7    - volume_mixing_ratio_of_no2
8    geometry:
9      fms initialization:
10     namelist filename: Data/fv3files/fmsmpp.nml
11     akbk: Data/inputs/fv3files/akbk72.nc4
12     npx: 13
13     npy: 13
14     npz: 72
15   background:
16     datetime: 2020-09-03T18:00:00Z
17     filetype: cube sphere history
18     datapath: Data/inputs/geos_c12
19     filename: geos_cf.bkg.%yyyy%mm%dd_%hh%MM%ssz.nc4
20     state variables:
21     - air_pressure_thickness
22     - volume_mixing_ratio_of_no2
23     - volume_mixing_ratio_of_no
24     - volume_mixing_ratio_of_o3
25     - air_pressure_at_surface
26     - water_vapor_mixing_ratio_wrt_moist_air
27   field io names: &field_io_names
28     volume_mixing_ratio_of_no2: N02
29     air_pressure_thickness: DELP
30     volume_mixing_ratio_of_no: "NO"
31     volume_mixing_ratio_of_o3: O3
32     air_pressure_at_surface: PS
```

Intro: Object-Oriented JEDI (OOPS) and YAML driven execution



DA methods require the same limited number of entities. The development of these entities should be reusable. Entities = basic (abstract) classes that define the system. A typical Variational DA YAML for example has general classes (see below). Inside each of those you will have more specific implementations and YAML options. DA abstract classes generally are:

cost function:
cost type:
time window:
analysis variables:
geometry:
background:
background error:
observations:
output:
variational:

The goal of those lectures (2 and 3 particularly) is to go over the YAMLs and understand how they are constructed to perform a DA application

```
1  cost function:
2  cost type: 3D-Var
3  time window:
4  begin: 2020-09-03T15:00:00Z
5  length: PT6H
6  analysis variables:
7  - volume_mixing_ratio_of_no2
8  geometry:
9
10
11
12
13
14
15
16
17  filetype: cube sphere history
18  datapath: Data/inputs/geos_c12
19  filename: geos_cf.bkg.%yyyy%mm%dd_%hh%MM%ssz.nc4
20  state variables:
21  - air_pressure_thickness
22  - volume_mixing_ratio_of_no2
23  - volume_mixing_ratio_of_no
24  - volume_mixing_ratio_of_o3
25  - air_pressure_at_surface
26  - water_vapor_mixing_ratio_wrt_moist_air
27  field io names: &field_io_names
28  volume_mixing_ratio_of_no2: N02
29  air_pressure_thickness: DELP
30  volume_mixing_ratio_of_no: "NO"
31  volume_mixing_ratio_of_o3: O3
32  air_pressure_at_surface: PS
```

Lecture 1: Handling Observations



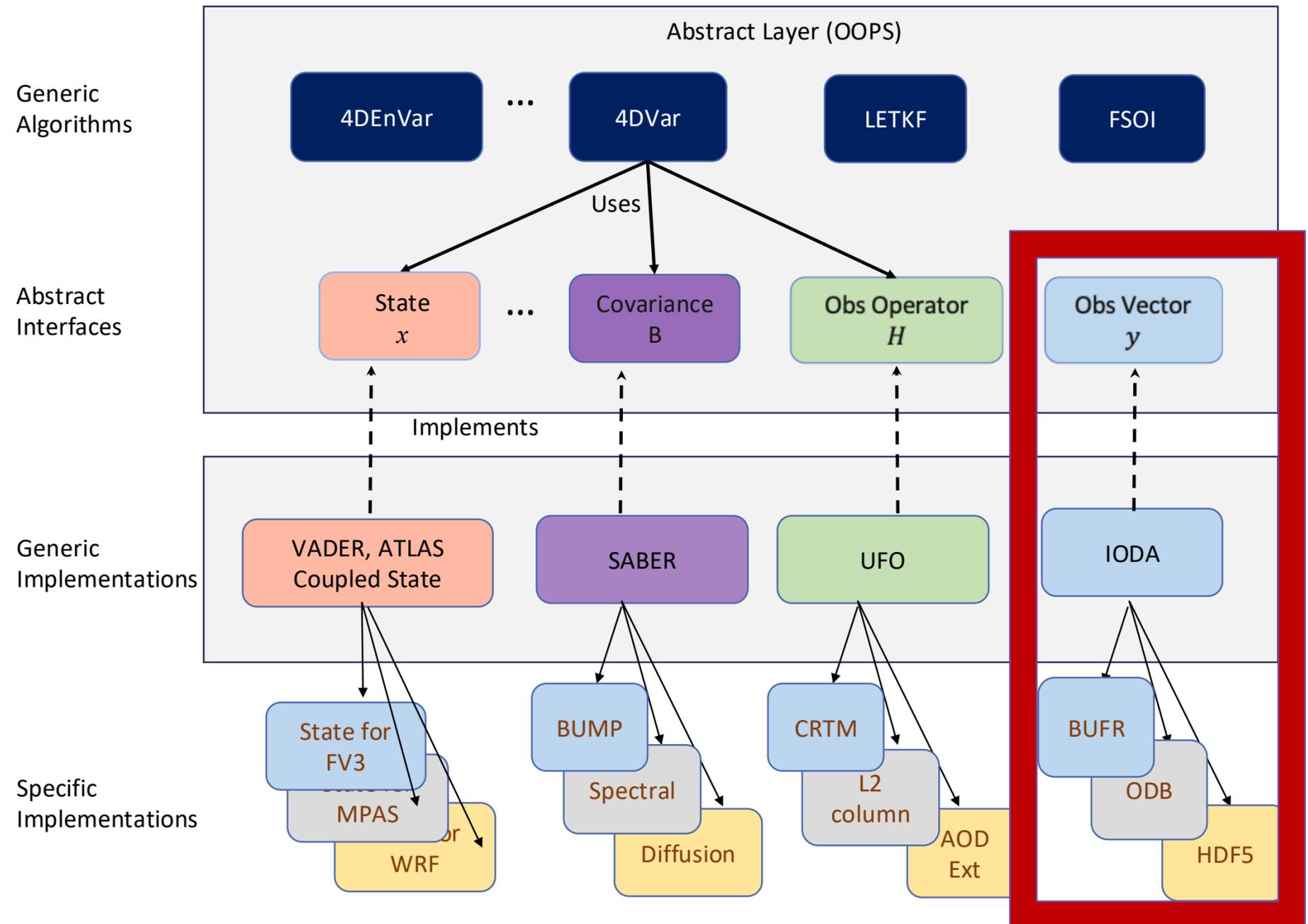
- Unified interface to observations (IODA) with multiple file-format backends
- IODA naming convention for observations & meta-data

YAML parts are something like that

```

observations:
  observers:
    - obs space:
      name: N02
      obsdatain:
        engine:
          type: H5File
          obsfile: Data/obs/tropomi_no2.nc4
      obsdataout:
        engine:
          type: H5File
          obsfile: Data/hofx/tropomi_no2.nc
      simulated variables:
        - nitrogendioxideColumn
    
```

$$\Delta x_a = \mathbf{B} \mathbf{M}^T \mathbf{H}^T (\mathbf{H} \mathbf{M} \mathbf{B} \mathbf{M}^T \mathbf{H}^T + \mathbf{R})^{-1} (y_o - \mathbf{H}(\mathbf{M}(x_b)))$$



Lecture 1: The IODA format: observation and feedback files



```
netcdf airnow_fb_example {
dimensions:
Location = UNLIMITED ; // (8755 currently)
variables:
int64 Location(Location) ;

group: EffectiveError0 {}
group: EffectiveError1 {}
group: EffectiveQC0 {}
group: EffectiveQC1 {}
group: MetaData {}
group: ObsBias0 {}
group: ObsBias1 {}
group: ObsError {}
group: ObsValue {}
group: PreQC {}
group: hofx0 {}
group: hofx1 {}
group: oman {}
group: ombg {}
}
```

In Observation file

Added as Feedback file after H(x)

Added as Feedback file after DA

```
group: MetaData {
  variables:
    int airQualityClassification(Location) ;
        airQualityClassification:_FillValue = -2147483643 ;
    int64 dateTime(Location) ;
        dateTime:_FillValue = -9223372036854775801LL ;
        string dateTime:units = "seconds since 1970-01-01T00:00:00Z" ;
    float latitude(Location) ;
        latitude:_FillValue = -3.368795e+38f ;
        string latitude:units = "degrees_north" ;
    float longitude(Location) ;
        longitude:_FillValue = -3.368795e+38f ;
        string longitude:units = "degrees_east" ;
} // group MetaData
```

For example:

```
group: hofx0 {
  variables:
    float carbonmonoxideInsitu(Location) ;
        carbonmonoxideInsitu:_FillValue = -3.368795e+38f ;
    float nitrogendioxideInsitu(Location) ;
        nitrogendioxideInsitu:_FillValue = -3.368795e+38f ;
    float ozoneInsitu(Location) ;
        ozoneInsitu:_FillValue = -3.368795e+38f ;
} // group hofx0
```

Practical Lecture 1



- <https://mer-a-o.github.io/howtojedi/lecture1/>
- Understanding the IODA format
- Use IODA converters

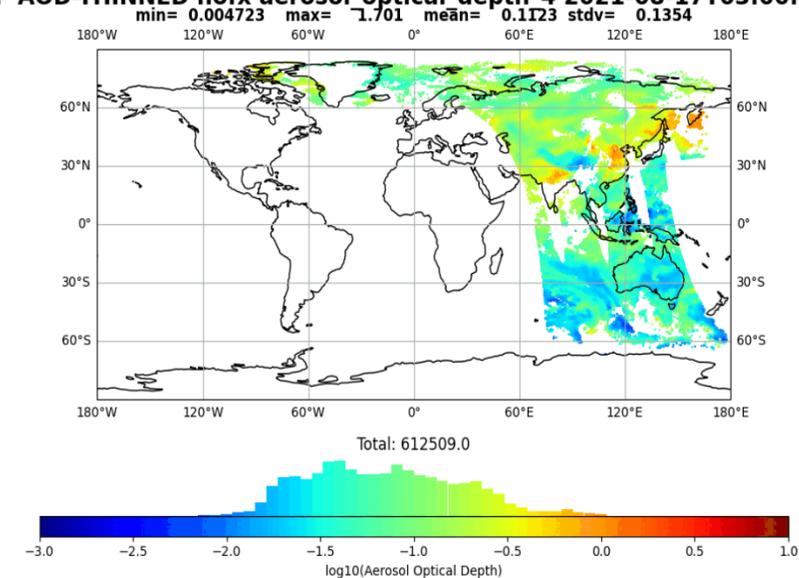
Lecture 1: Tested and assimilated AC observation products in JEDI



Aerosols:

- VIIRS NPP AOD (DT,DB) and L1b
- VIIRS NOAA 20 AOD (DT,DB) and L1b
- MODIS Aqua AOD
- MODIS Terra AOD
- PACE AOD
- AeroNet
- CALIOP Lidar
- Surface: Airnow, Purple Air, Airbase

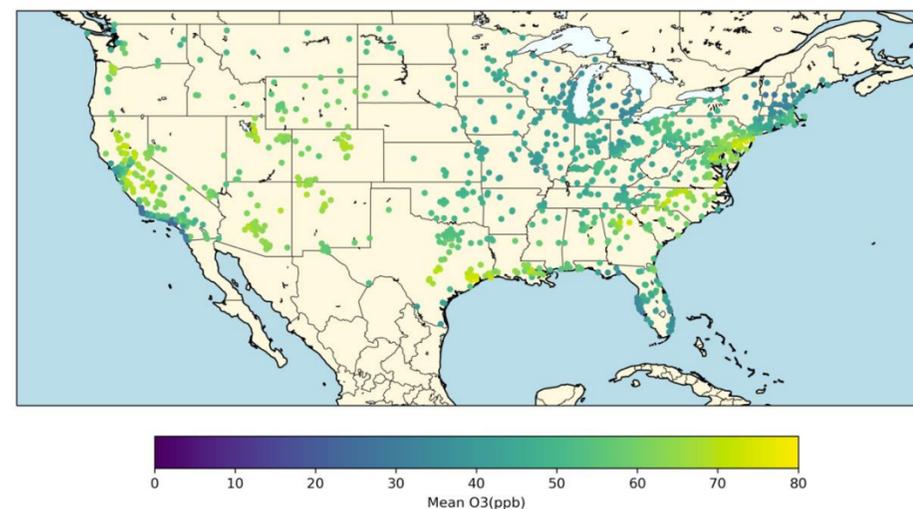
RS NPP AOD-THINNED hofx aerosol optical depth 4 2021-08-17T03:00:00Z PT



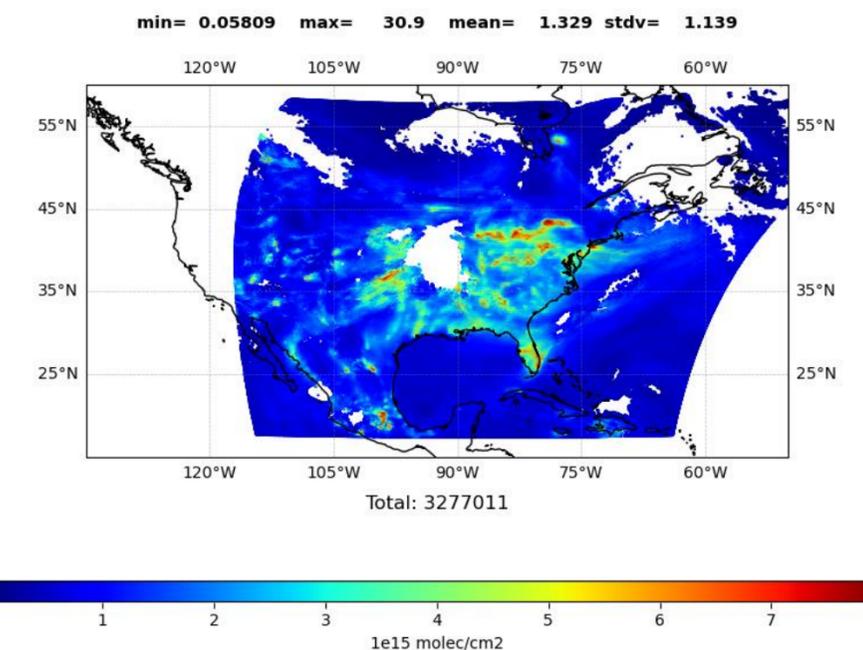
Trace gas:

- TropOMI NO2 (tropospheric and total columns)
- TropOMI CO total column
- MOPITT CO total column
- TEMPO NO2 and HCHO columns
- MLS O3 limb profile
- OMPS LP, NP and TC O3 (limb and nadir)
- PANDORA NETWORK
- Surface: Airnow, Airbase
- Aircraft campaigns

AirNow O3 H(x)



tempo_no2_tropo nitrogendioxideColumn H(x) 20230808T12Z_PT24H



Lecture 2: Using the Observation operators



Large suite of Obs Operators & Quality Control

- CRTM, RTTOV satellite radiances
- Conventional atmospheric obs
- Aerosol optical depth
- Column retrievals
- And many more for NWP, Ocean, Space Weather...

observations:

observers:

- obs space:

<the ioda part that we've just looked at>

obs operator:

name: ColumnRetrieval

nlayers_retrieval: 34

tracer variables:

- volume_mixing_ratio_of_no2

isA priori: false

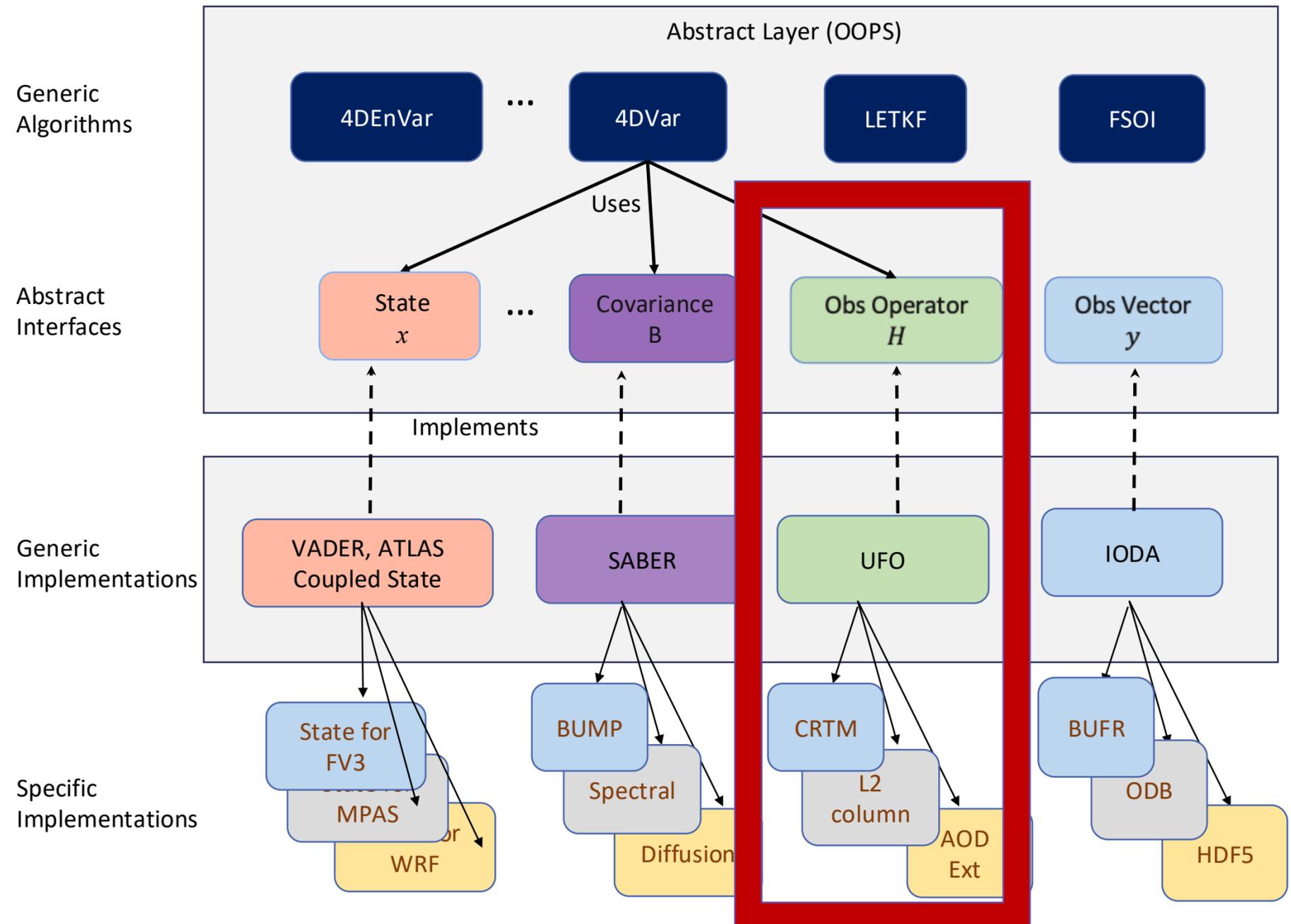
isAveragingKernel: true

stretchVertices: topbottom

obs error:

covariance model: diagonal

$$\Delta x_a = \mathbf{B} \mathbf{M}^T \mathbf{H}^T (\mathbf{H} \mathbf{M} \mathbf{B} \mathbf{M}^T \mathbf{H}^T + \mathbf{R})^{-1} (y_o - \mathbf{H}(\mathbf{M}(x_b)))$$



Lecture 2: Column retrieval operator



It computes the model equivalent to the observation using the retrieval equation as follows:

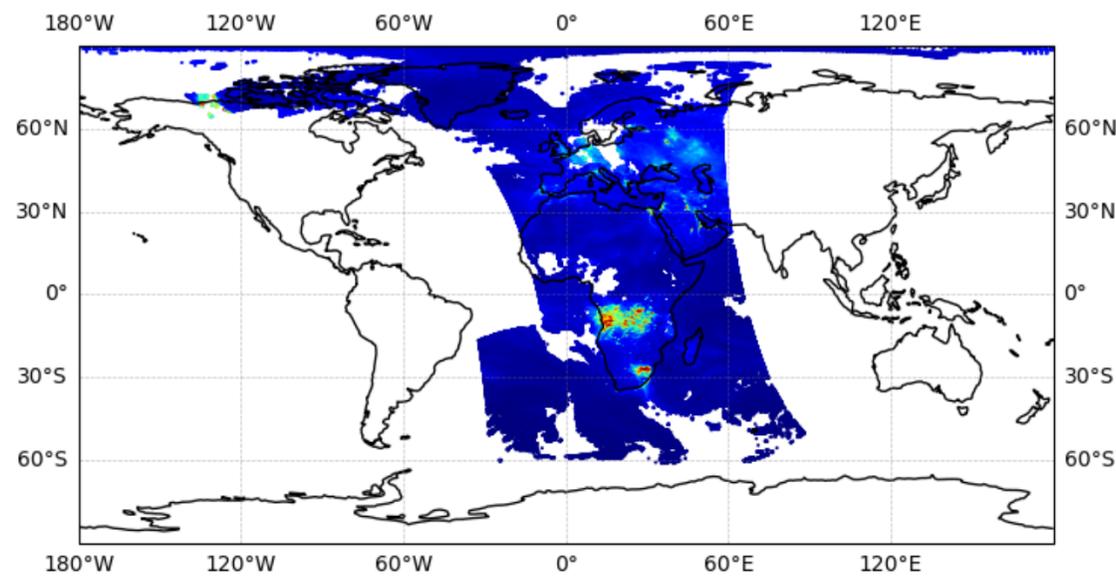
$$\mathbf{X}_c = \mathbf{A}\mathbf{X}_{\text{background}} + (\mathbf{I} - \mathbf{A})\mathbf{X}_{\text{apriori}}$$

The operator computes partial columns $\mathbf{X}_{\text{background}}$ on the retrieval vertical grid and then applies, if needed, the averaging kernel or scattering weights information \mathbf{A} . If needed as well, it can apply an apriori term $\mathbf{X}_{\text{apriori}}$. For each geoval profile the operator computes the partial or total column of any given atmospheric molecule concentration \mathbf{X}_c .

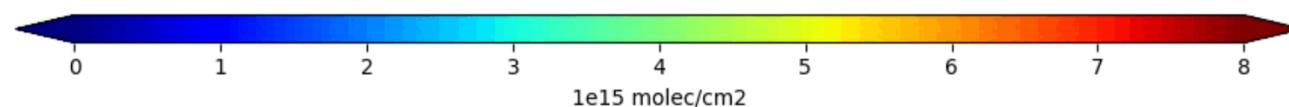
Operator adopted for NWP O3 operational use at NOAA NWS EMC

tropomi_s5p_no2_tropo nitrogendioxideColumn H(x) 20230805T12Z_PT24H

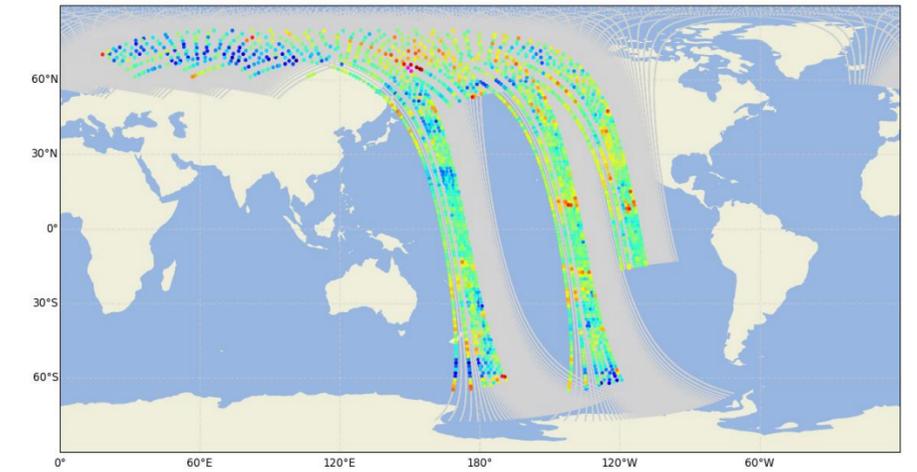
min= 0.001009 max= 71.34 mean= 0.6953 stdv= 1.129



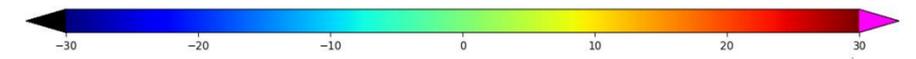
Total: 2522151



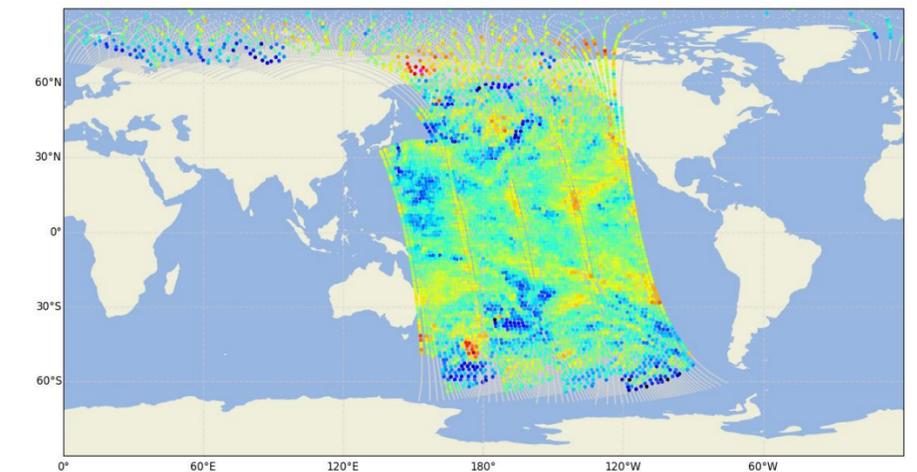
UFO OMI AURA O-F ozoneTotal with QC



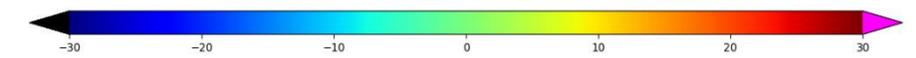
Total Count: 2384 Max: 36.367 Min: -29.219 Mean: -1.030 Std: 7.809



UFO OMPSTC NPP O-F ozoneTotal with QC



Total Count: 6142 Max: 31.031 Min: -34.714 Mean: -1.638 Std: 7.186



Lecture 2: Column retrieval operator



```
obs operator:
  name: ColumnRetrieval
  nlayers_retrieval: 50
  tracer variables: [volume_mixing_ratio_of_co]
  isApriori: false
  isAveragingKernel: true
  stretchVertices: topbottom
  model units coeff: 1e-6
```

TropOMI CO

```
- obs operator:
  name: ColumnRetrieval
  nlayers_retrieval: 1
  tracer variables: [mole_fraction_of_ozone_in_air]
  isApriori: false
  isAveragingKernel: false
  totalNoVertice: true
  stretchVertices: topbottom
  model units coeff: 2.240013904035E-3
```

OMPS O3

retrieval equation as

$X_{apriori}$

retrieval vertical grid

needed, the averaging kernel or scattering weights information A . If needed as

apriori term $X_{apriori}$. For each geoval profile the operator computes the partial or total column of any

```
obs operator:
  name: ColumnRetrieval
  nlayers_retrieval: 34
  tracer variables: [volume_mixing_ratio_of_no2]
  isApriori: false
  isAveragingKernel: true
  stretchVertices: topbottom
  model units coeff: 1e-6
```

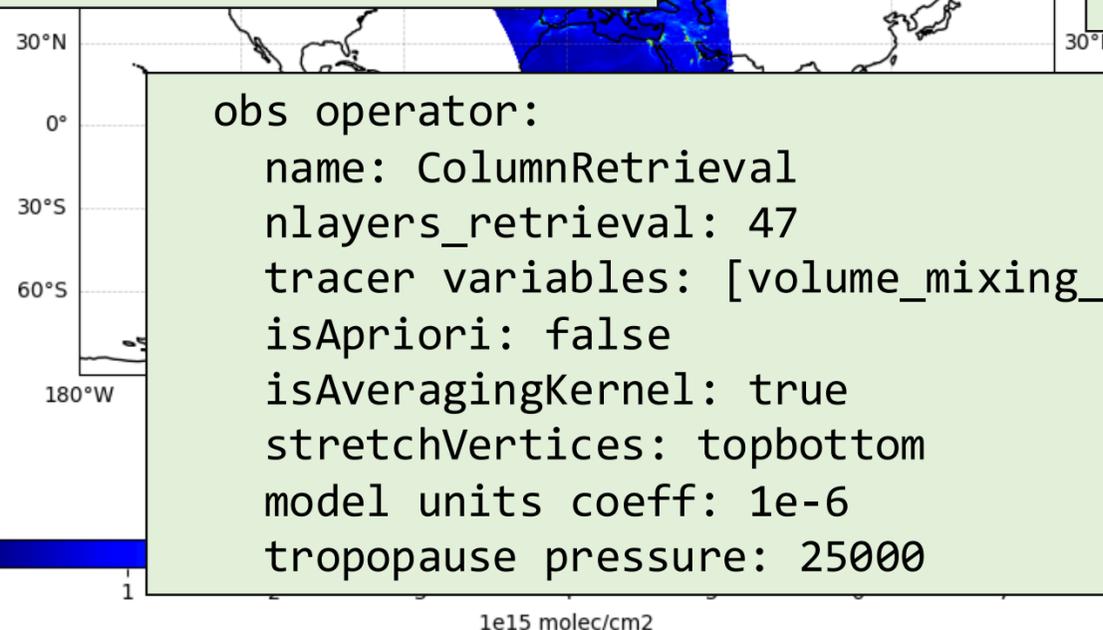
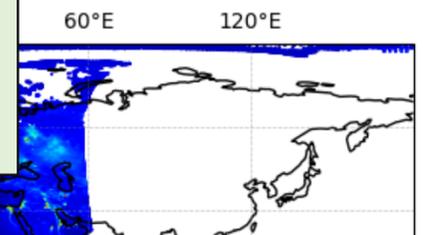
TropOMI NO2

```
obs operator:
  name: ColumnRetrieval
  nlayers_retrieval: 10
  tracer variables: [volume_mixing_ratio_of_co]
  isApriori: true
  isAveragingKernel: true
  stretchVertices: topbottom
  model units coeff: 1e-6
```

MOPITT CO

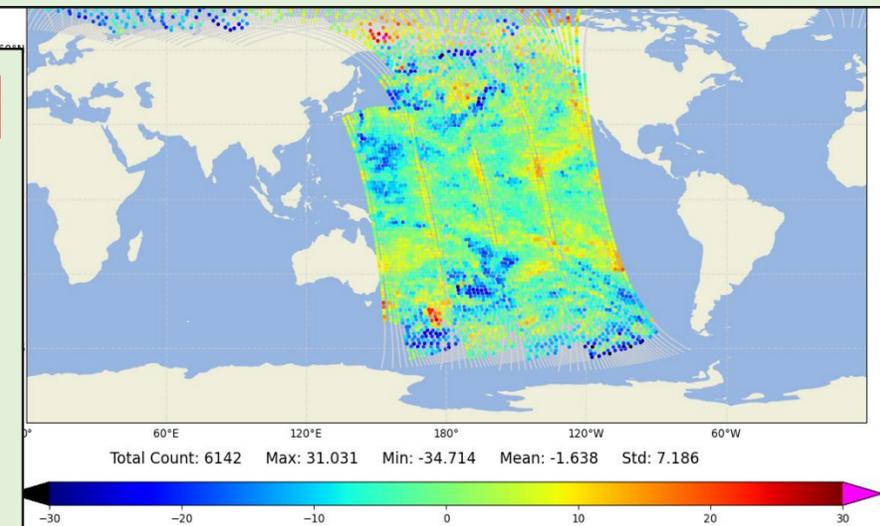
Column H(x) 20230805T12Z_P

0.6953 stdv= 1.129



```
obs operator:
  name: ColumnRetrieval
  nlayers_retrieval: 47
  tracer variables: [volume_mixing_ratio_of_no2]
  isApriori: false
  isAveragingKernel: true
  stretchVertices: topbottom
  model units coeff: 1e-6
  tropopause pressure: 25000
```

TEMPO NO2



Lecture 2: Aerosols operators



$$\Delta x_a = \mathbf{B}\mathbf{M}^T\mathbf{H}^T(\mathbf{H}\mathbf{M}\mathbf{B}\mathbf{M}^T\mathbf{H}^T + \mathbf{R})^{-1} \left(y_o - \mathbf{H}(\mathbf{M}(x_b)) \right)$$

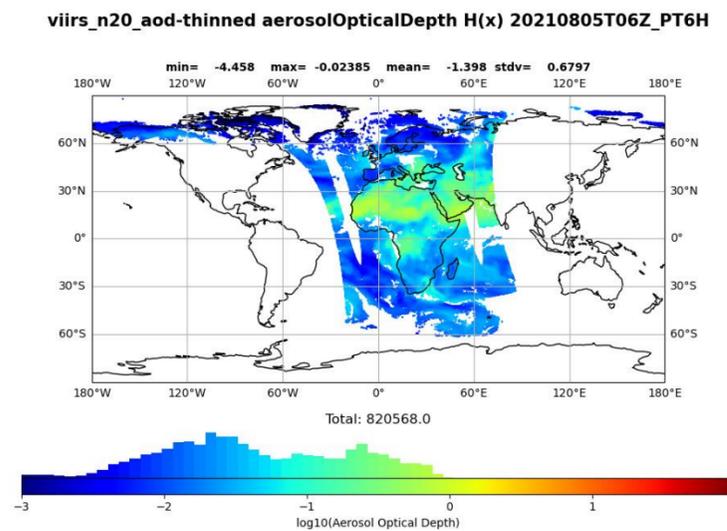
AOD CRTM:

- Extinction calculation within JEDI as part of the observation operator **H**
- Requires all aerosols variables in the state vector to compute AOD with RTM, could be expensive with EnVar at high resolution
- Prognostic variables directly updated to initialize the forecast.
- Different LUTs for optical properties can be loaded and tested within JEDI

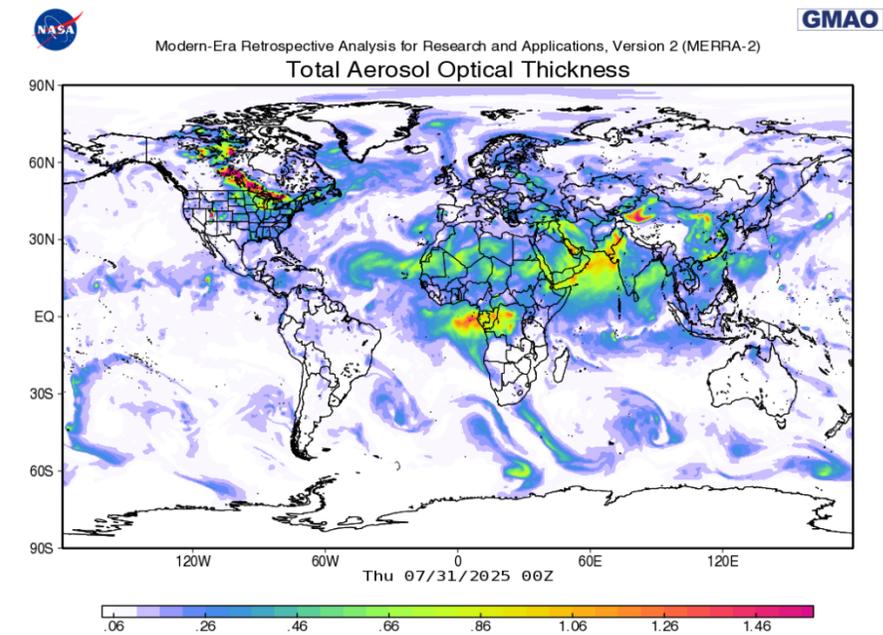
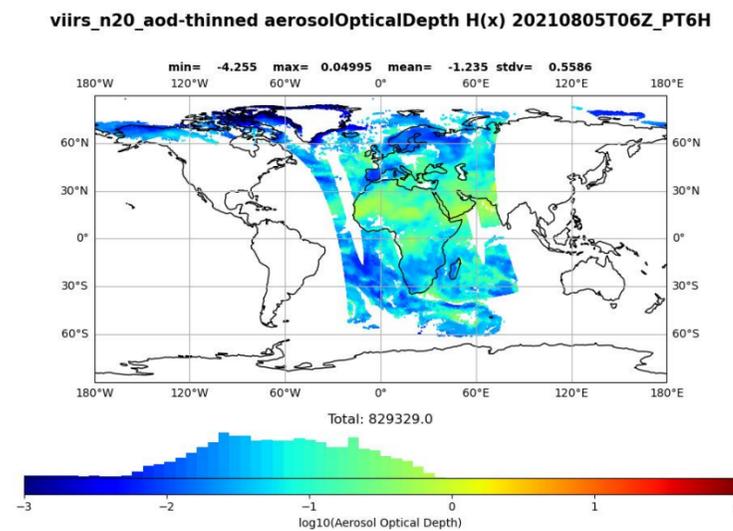
AOD Extinction:

- Extinction calculation within GEOS (or other) as part of the forecast model **M**
- Doesn't require all aerosols variables in the state vector therefore much cheaper for the JEDI part
- Prognostic variables are not updated through JEDI but a posteriori of the analysis

H(x) with default CRTM LUT



H(x) with NASA-GEOS LUT



Lecture 2: Aerosols operators

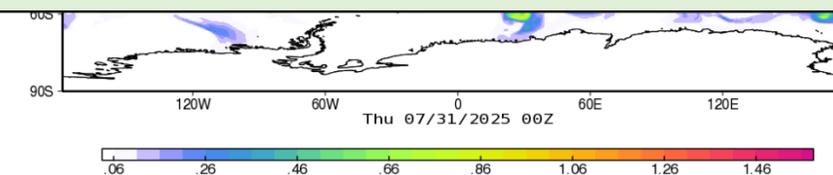


```
observations:
  observers:
    - obs space:
      name: Aod
      obsdatain:
        engine:
          type: H5File
          obsfile: aod_viirs_npp_obs.nc4
      obsdataout:
        engine:
          type: H5File
          obsfile: aod_viirs_npp_3dvar.nc4
      simulated variables:
        - aerosolOpticalDepth
      channels: 4
  obs operator:
    name: AodCRTM
    Absorbers:
      - H2O
      - O3
    obs options:
      Sensor_ID: v.viirs-m_npp
      EndianType: little_endian
      CoefficientPath: Data/crtm/
      AerosolOption: aerosols_gocart_default
```

VIIRS AOD with AODCRTM

```
observations:
  observers:
    - obs space:
      name: Aod
      obsdatain:
        engine:
          type: H5File
          obsfile: nnr_aod_3ch_obs.nc4
      obsdataout:
        engine:
          type: H5File
          obsfile: nnr_aod_3ch_obs_out.nc4
      simulated variables:
        - aerosolOpticalDepth
      channels: 1, 3
  obs operator:
    name: AodExt
    nprofiles: 2
    bkg_wavelengths: [470.0,870.0]
    doing_log_transform_aod: true
    eps_for_log_transform_aod: 0.01
```

MODIS NNR AOD with AODExt

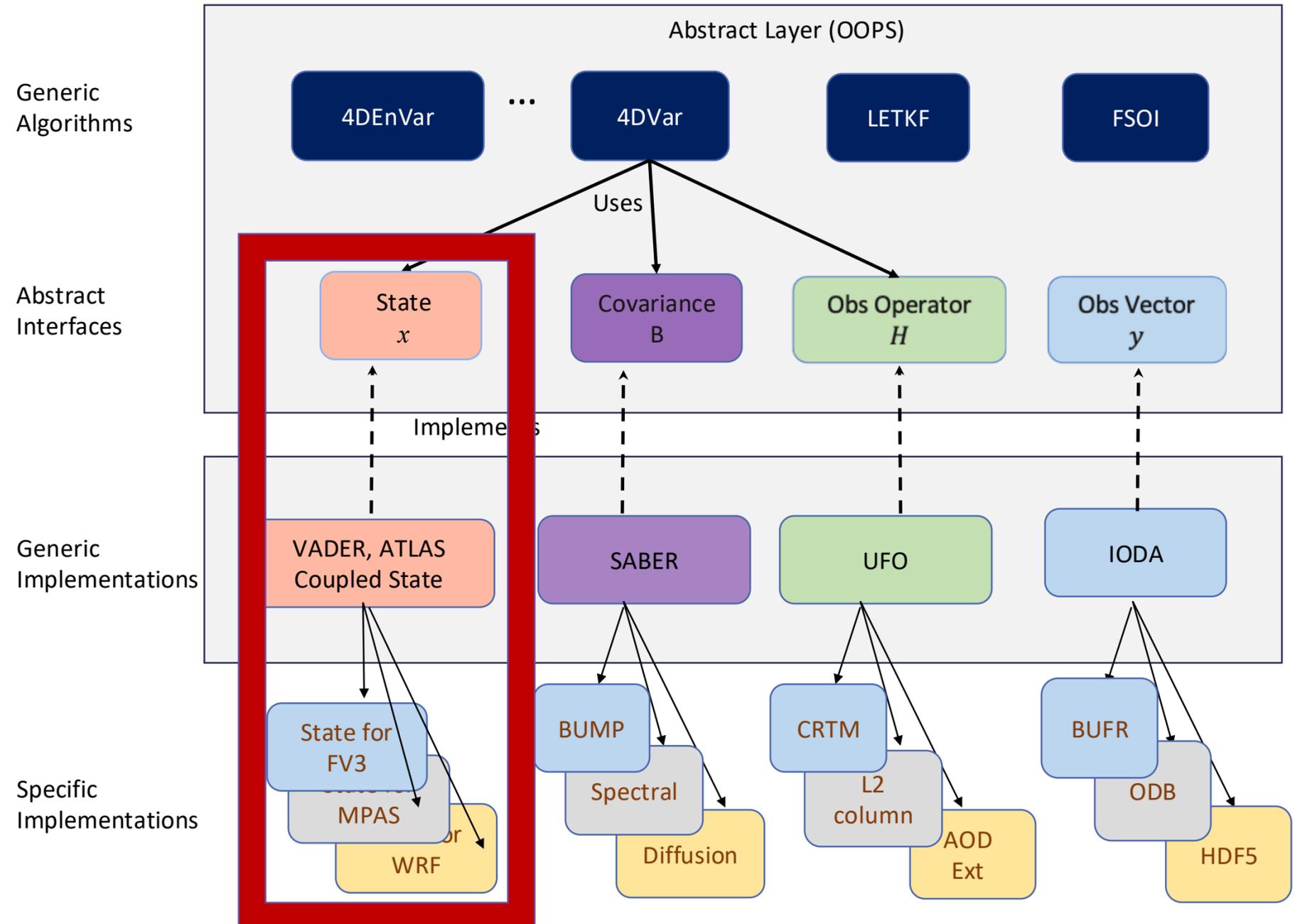


Lecture 2: Model interface



MODEL	TYPE
UFS	Global Atmosphere
GEOS	Global Atmosphere
NEPTUNE	Global atmosphere
MPAS	Global atmosphere
LFRic	Global atmosphere
Unified Model	Global atmosphere
UFS CAM	Regional atmosphere
MPAS-Regional	Regional atmosphere
WRF	Regional atmosphere
UFS GSDChem	Global constituents
GEOS-AERO	Global aerosols
UFS CAM-CMAQ	Regional air quality
MOM6	Global ocean
ROMS	Regional ocean
SIS2	Sea-ice
CICE6	Sea-ice
WW-III	Wave
NOAH-MP	Land and Snow
QG	Toy model
Lorenz 95	Toy model
Shallow Water	Toy model

$$\Delta x_a = \mathbf{B} \mathbf{M}^T \mathbf{H}^T (\mathbf{H} \mathbf{M} \mathbf{B} \mathbf{M}^T \mathbf{H}^T + \mathbf{R})^{-1} (y_o - \mathbf{H}(\mathbf{M}(x_b)))$$



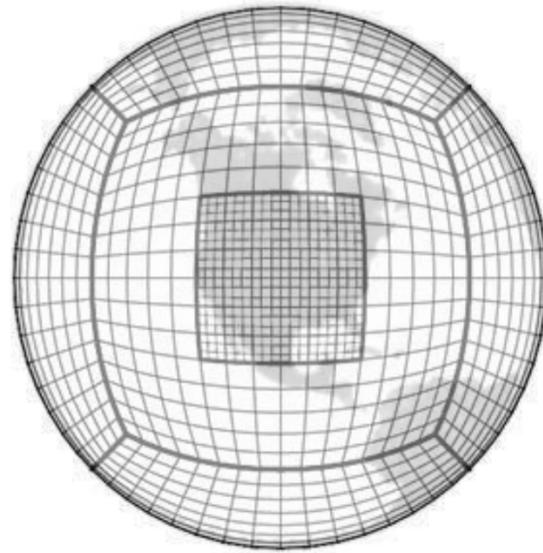
Lecture 2: Model interfaces



JEDI atmospheric composition modeling configurations for DA

NOAA UFS-Chem

Dycore/grid: FV3 – nested domain

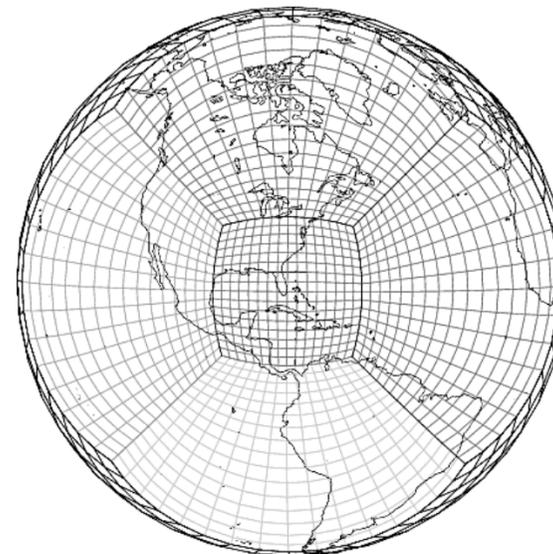


FV3-JEDI interface

JEDI AC Status: Tested and used with JEDI for regional and global AC applications (pre-operational)

NASA GEOS/GEOS-CF

Dycore/grid: FV3 – stretch grid



FV3-JEDI interface

JEDI AC Status: Tested and used with JEDI for global AC applications – tested with stretch grid

NCAR MPAS

Dycore/grid: MPAS – refined grid



MPAS-JEDI interface

JEDI Status: ongoing progress to develop JEDI-MPAS-GOCART for atmospheric composition
Reactive chemistry to come



Model interfaces

FV3-JEDI interface section

```

geometry:
  fms initialization:
    namelist filename: Data/fv3files/fmsmpp.nml
  akbk: Data/inputs/fv3files/akbk72.nc4
  npz: 72
  npx: 13
  npy: 13

state:
  datetime: 2020-09-03T18:00:00Z
  filetype: cube sphere history
  datapath: Data/inputs/geos_c12
  filename: geos_cf.bkg.%yyyy%mm%dd_%hh%MM%ssz.nc4
  state variables:
    - air_pressure_thickness
    - volume_mixing_ratio_of_no2
    - volume_mixing_ratio_of_no
    - volume_mixing_ratio_of_o3
    - air_pressure_at_surface
    - water_vapor_mixing_ratio_wrt_moist_air
  field io names: &field_io_names
    volume_mixing_ratio_of_no2: N02
    air_pressure_thickness: DELP
    volume_mixing_ratio_of_no: "NO"
    volume_mixing_ratio_of_o3: O3
    air_pressure_at_surface: PS
    water_vapor_mixing_ratio_wrt_moist_air: SPHU

```

MPAS-JEDI interface section

```

geometry:
  nml_file: "namelist.atmosphere_2018041500"
  streams_file: "streams.atmosphere"

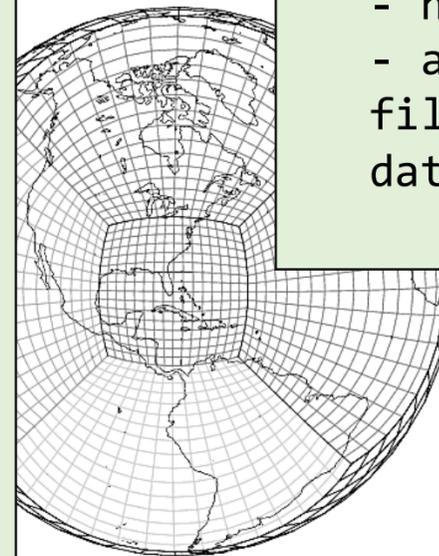
state:
  state variables:
    - air_temperature
    - water_vapor_mixing_ratio_wrt_moist_air
    - eastward_wind
    - northward_wind
    - air_pressure_at_surface
  filename: "restart.2018-04-15_00.00.00.nc"
  date: '2018-04-15T00:00:00Z'

```

position mod

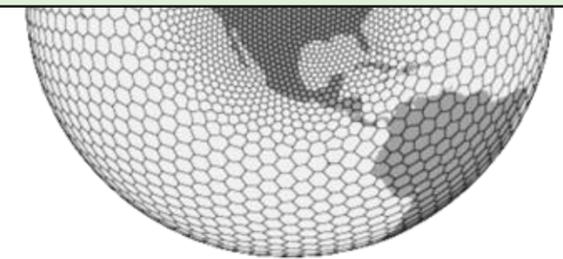
A GEOS/GE

grid: FV3 - s



FV3-JEDI interface

Status: Tested and used
for global AC applications
with stretch grid



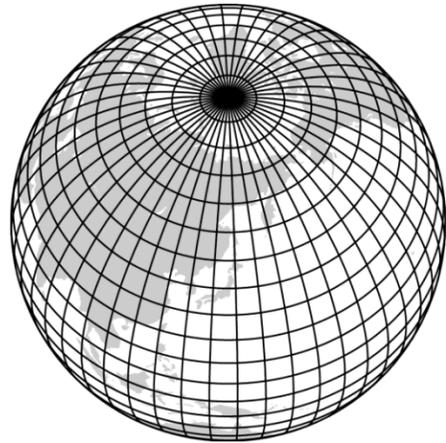
MPAS-JEDI interface

JEDI Status: ongoing progress to
develop JEDI-MPAS-GOCART for
atmospheric composition
Reactive chemistry to come

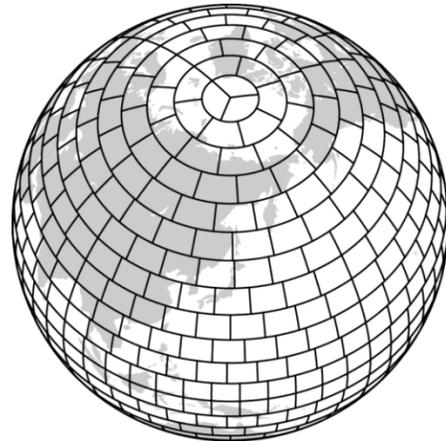
Lecture 2: Model interfaces: Versatile Implementation for Native Data (VIND)



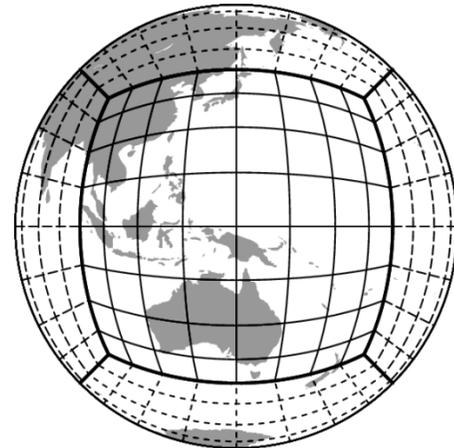
Regular LatLon



Gaussian



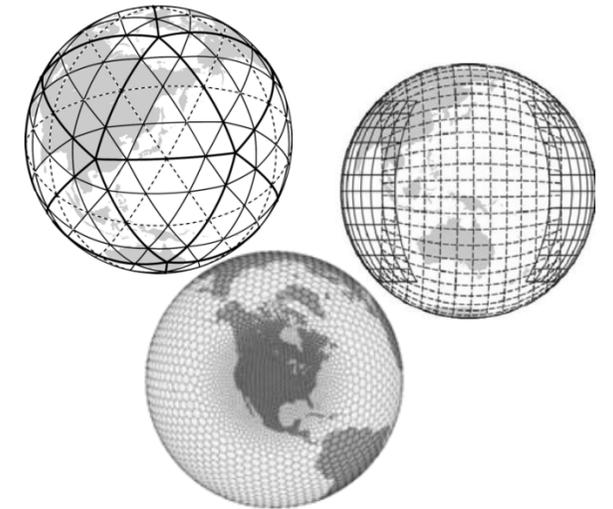
Cubed Sphere



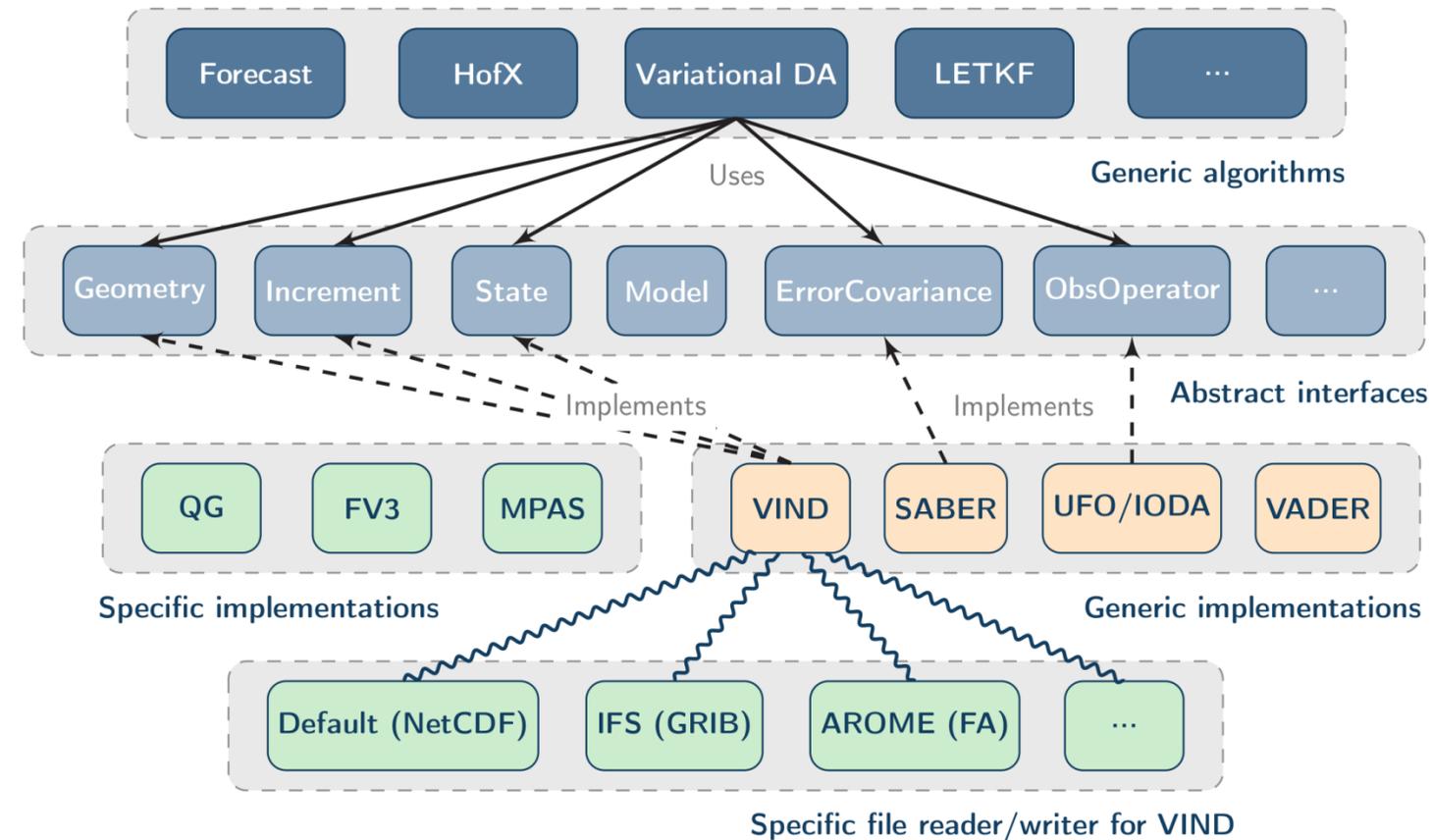
Limited areas models
Including Lambert Conformal



And many other geometries:
Structured and unstructured



VIND builds on ATLAS, an open-source C++ library from ECMWF, which provides data structures for handling fields on a wide range of global and regional grid geometries.



Lecture 2: Model interfaces: Versatile Implementation for Native Data (VIND)



VIND WRF-Chem geometry

```
geometry:
  function space: StructuredColumns
  partitioner: "checkerboard"
  grid:
    type : "regional"
    nx : 222
    ny : 198
    dx : 4000.0
    dy : 4000.0
    lonlat(centre) : [-75.1618,42.90771]
    #lonlat(xmin,ymin): [-81.634705,38.214104]
  projection :
    type : "lambert_conformal_conic"
    longitude0 : -97.0
    latitude0 : 39.0
    latitude1 : 30.0
    latitude2 : 60.0
    radius: 6370000
```

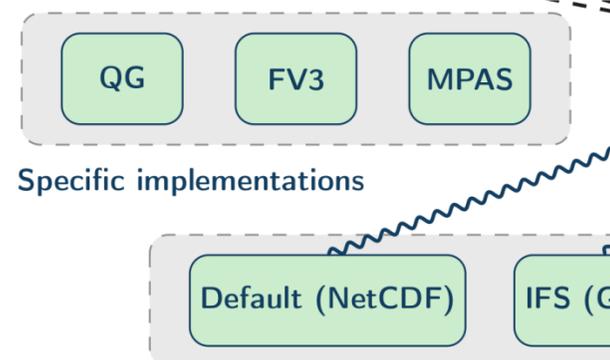
VIND LatLon geometry

```
geometry:
  function space: StructuredColumns
  grid:
    type: regular_lonlat
    nx: 30
    ny: 20
```

VIND regional rotated LatLon geometry

```
geometry: &geom
  function space: StructuredColumns
  grid:
    type: structured
    xspace:
      type: linear
      N : 351
      start : -35.0
      end : 35.0
    yspace:
      type: linear
      N: 271
      start: -27.0
      end : 27.0
  projection :
    type: rotated_lonlat
    north_pole: [190, 39]
    y_numbering: -1
```

on a wide range of global and regional grid geometries.



Lecture 2: Model interfaces: Versatile Implementation for Native Data (VIND)



VIND WRF-Chem geometry

```
geometry:  
  function space: StructuredColumns  
  partitioner: "checkerboard"  
  grid:  
    type : "regional"  
    nx : 222  
    ny : 198  
    dx : 4000.0  
    dy : 4000.0  
    lonlat(centre) : [-75.1618,42.907711]  
    #lonlat(xmin,ymin): [-81.634705,  
    projection :  
      type : "lambert"  
      longitude0 : -90.0  
      latitude0 : 39.0  
      latitude1 : 30.0  
      latitude2 : 60.0  
      radius: 6370000
```

VIND LatLon geometry

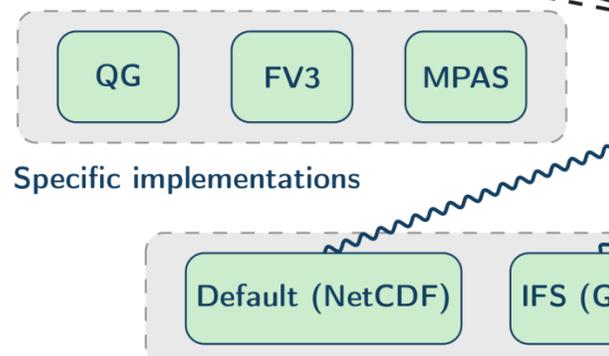
```
geometry:  
  function space: StructuredColumns  
  grid:  
    type: regular  
    nx: 2
```

Regional rotated LatLon geometry

```
function space: StructuredColumns  
type: structured  
xspace:  
  type: linear  
  N : 351  
  start : -35.0  
  end : 35.0  
yspace:  
  type: linear  
  N: 271  
  start: -27.0  
  end : 27.0  
projection :  
  type: rotated_lonlat  
  north_pole: [190, 39]  
  y_numbering: -1
```

Not covered in this tutorial

on a wide range of global and regional grid geometries.



Practical demonstration 2



- <https://mer-a-o.github.io/howtojedi/lecture2/>
- Understand how observation operators and model interfaces are combined
- Running an H(x) experiment
- Looking at the results

Lecture 3: Data assimilation algorithms



DA algorithms available:

- 3DVar, 3DVar-FGAT, 4DEnVar, 4DVar, ...
- LETKF, LGETKF

cost function:

cost type: **3D-Var**

analysis variables:

- volume_mixing_ratio_of_no2

time window:

begin: 2020-09-03T15:00:00Z

length: PT6H

variational:

minimizer:

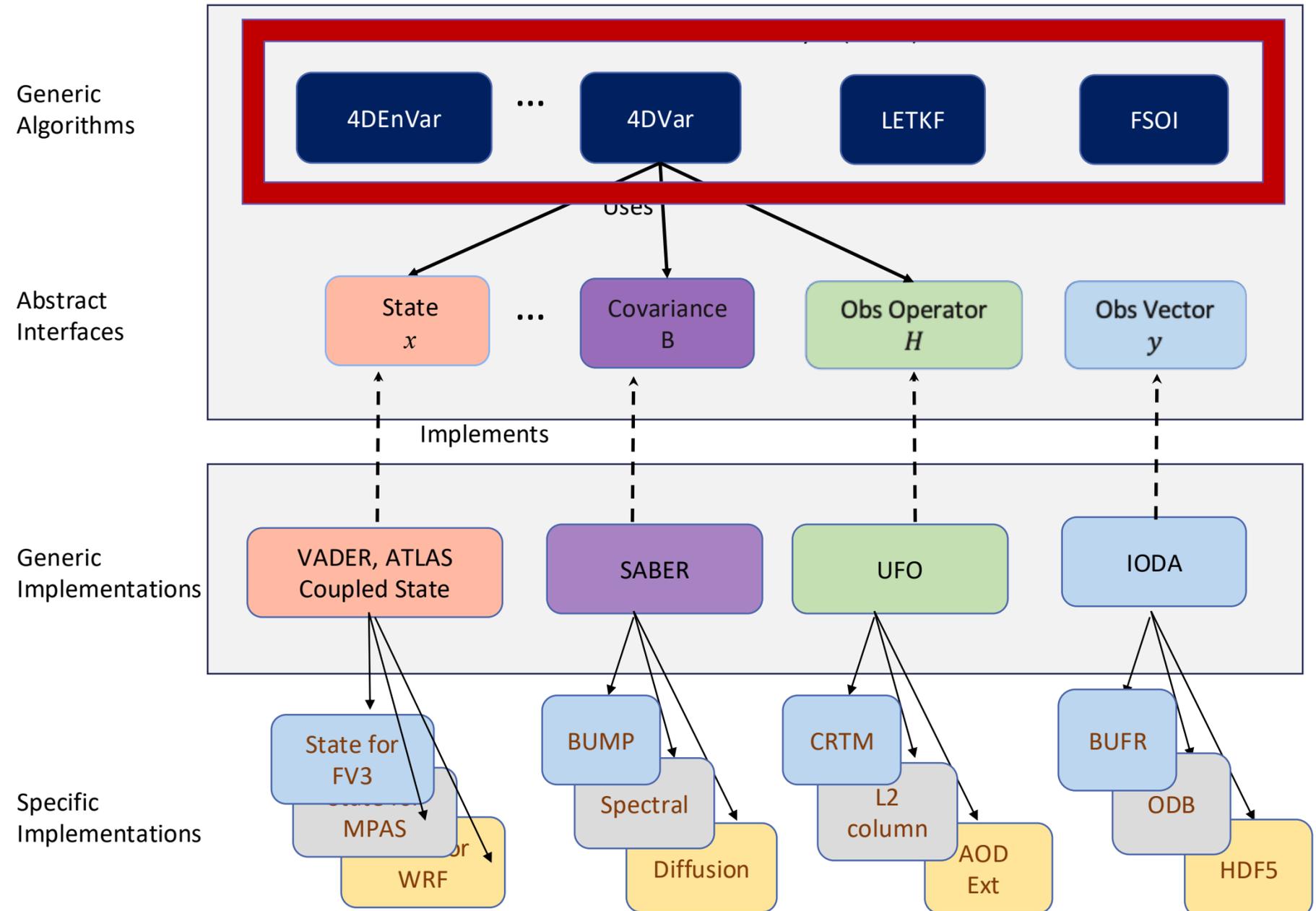
algorithm: DRPCG

iterations:

- ninner: 10

gradient norm reduction: 1e-10

$$\Delta x_a = \mathbf{B} \mathbf{M}^T \mathbf{H}^T (\mathbf{H} \mathbf{M} \mathbf{B} \mathbf{M}^T \mathbf{H}^T + \mathbf{R})^{-1} (y_o - \mathbf{H}(\mathbf{M}(x_b)))$$

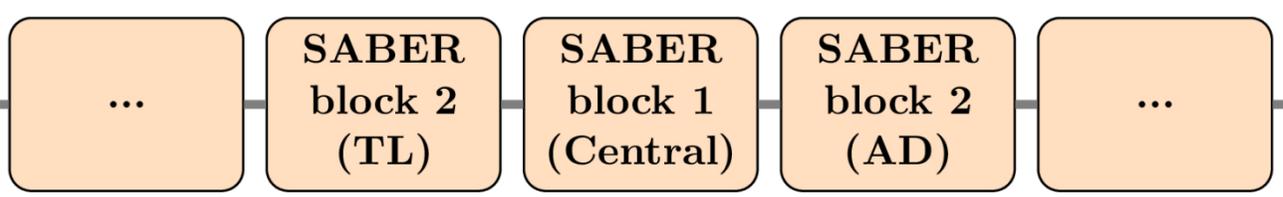


Lecture 3: Background error

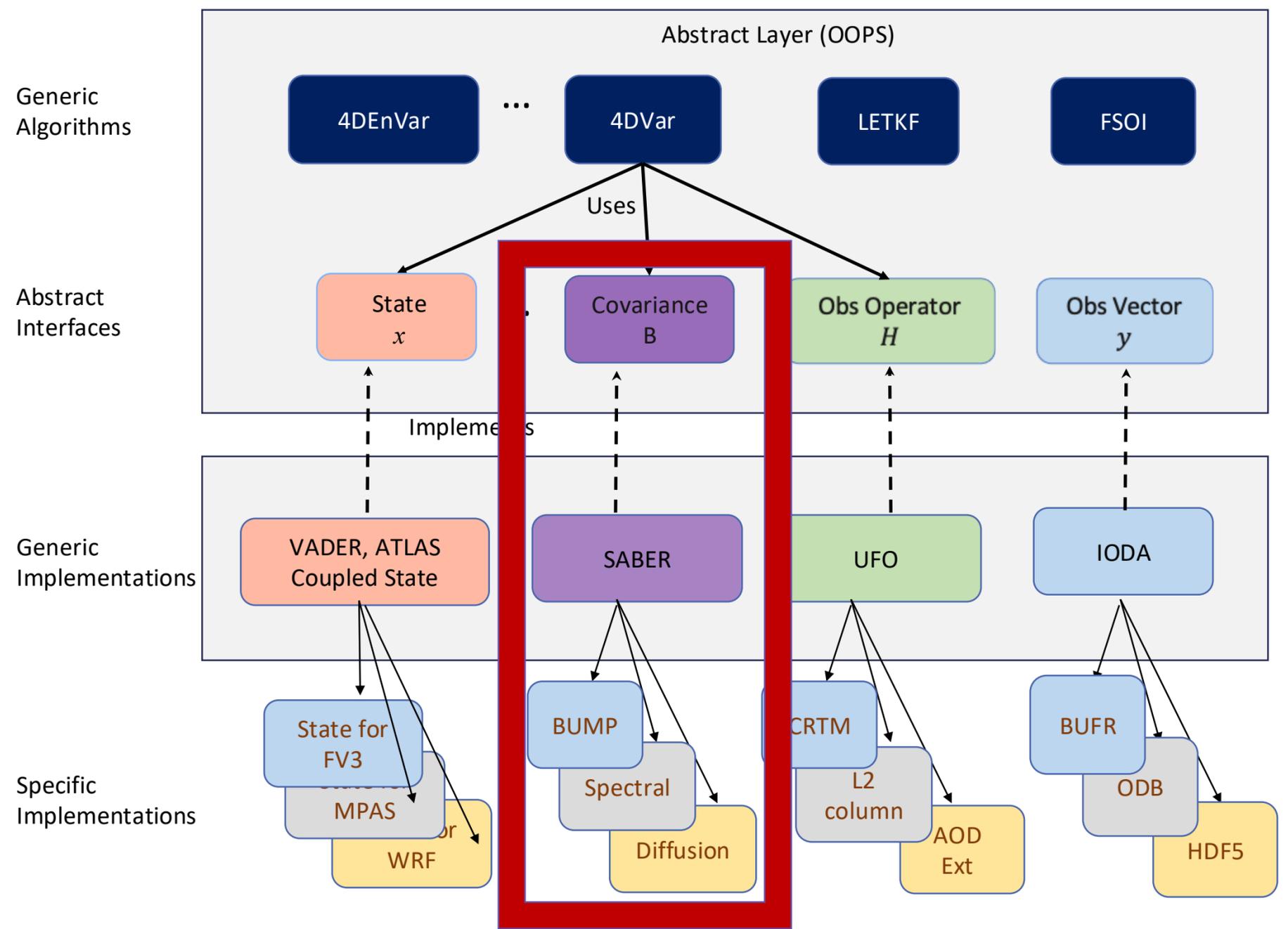


Background Error Covariances represented as a series of generic blocks:

- Balance operator, change of variables
- Auto-correlation operator (BUMP, spectral, diffusion...)
- Hybrid static-ensemble
- Option to import GSI static B



$$\Delta x_a = \mathbf{B}\mathbf{M}^T\mathbf{H}^T(\mathbf{H}\mathbf{M}\mathbf{B}\mathbf{M}^T\mathbf{H}^T + \mathbf{R})^{-1} (y_o - \mathbf{H}(\mathbf{M}(x_b)))$$



Lecture 3: Background error



SABER with BUMP Covariances

```
background error:
  covariance model: SABER

saber central block:
  saber block name: BUMP_NICAS
  read:
    io:
      files prefix: Data/bump/
      alias:
        - in code: volume_mixing_ratio_of_no2
          in file: fixed_2500km_0.3
  drivers:
    multivariate strategy: univariate
    read local nicas: true

saber outer blocks:
- saber block name: StdDev
  read:
    model file:
      datetime: 2020-09-03T18:00:00Z
      set datetime on read: true
      filetype: cube sphere history
      datapath: Data/inputs/geos_c12
      filename: geos_cf.stddev.nox_50_o3_25.nc4
      field io names: *field_io_names
```

$$\Delta x_a =$$



SABER with Diffusion Covariances

```
background error:
  covariance model: SABER

saber central block:
  saber block name: diffusion
  calibration:
    normalization:
      iterations: 1000
  groups:
    - horizontal:
      fixed value: 2500e3
    - vertical:
      levels: 72
      fixed value: 1.0
      as gaussian: true

saber outer blocks:
- saber block name: StdDev
  read:
    model file:
      datetime: 2020-09-03T18:00:00Z
      set datetime on read: true
      filetype: cube sphere history
      datapath: Data/inputs/geos_c12
      filename: geos_cf.stddev.nox_50_o3_25.nc4
      field io names: *field_io_names
```



Lecture 3: Background error

SABER with BUMP Covariances

background error:
covariance model: SABER

saber central block:
saber block name: BUMP_NICAS
read:
io:
files prefix: Data/bump/
alias:
- in code: volume_mixing_ratio_of_no2
in file: fixed_2500km_0.3
drivers:
multivariate strategy: univariate
read local nicas: true

saber outer blocks:
- saber block name: StdDev
read:
model file:
datetime: 2020-09-03T18:00:00Z
set datetime on read: true
filetype: cube sphere history
datapath: Data/inputs/geos_c12
filename: geos_cf.stddev.nox_50_o3_25.nc4
field io names: *field_io_names

$$\Delta x_a =$$

SABER with Diffusion Covariances

background error:
covariance model: SABER
saber central block:

SABER with Identity matrix

background error:
covariance model: SABER
saber central block:
saber block name: ID

diffusion
fixed value: 2500e3
- vertical:
levels: 72
fixed value: 1.0
as gaussian: true
saber outer blocks:
- saber block name: StdDev
read:
model file:
datetime: 2020-09-03T18:00:00Z
set datetime on read: true
filetype: cube sphere history
datapath: Data/inputs/geos_c12
filename: geos_cf.stddev.nox_50_o3_25.nc4
field io names: *field_io_names

Practical demonstration 3

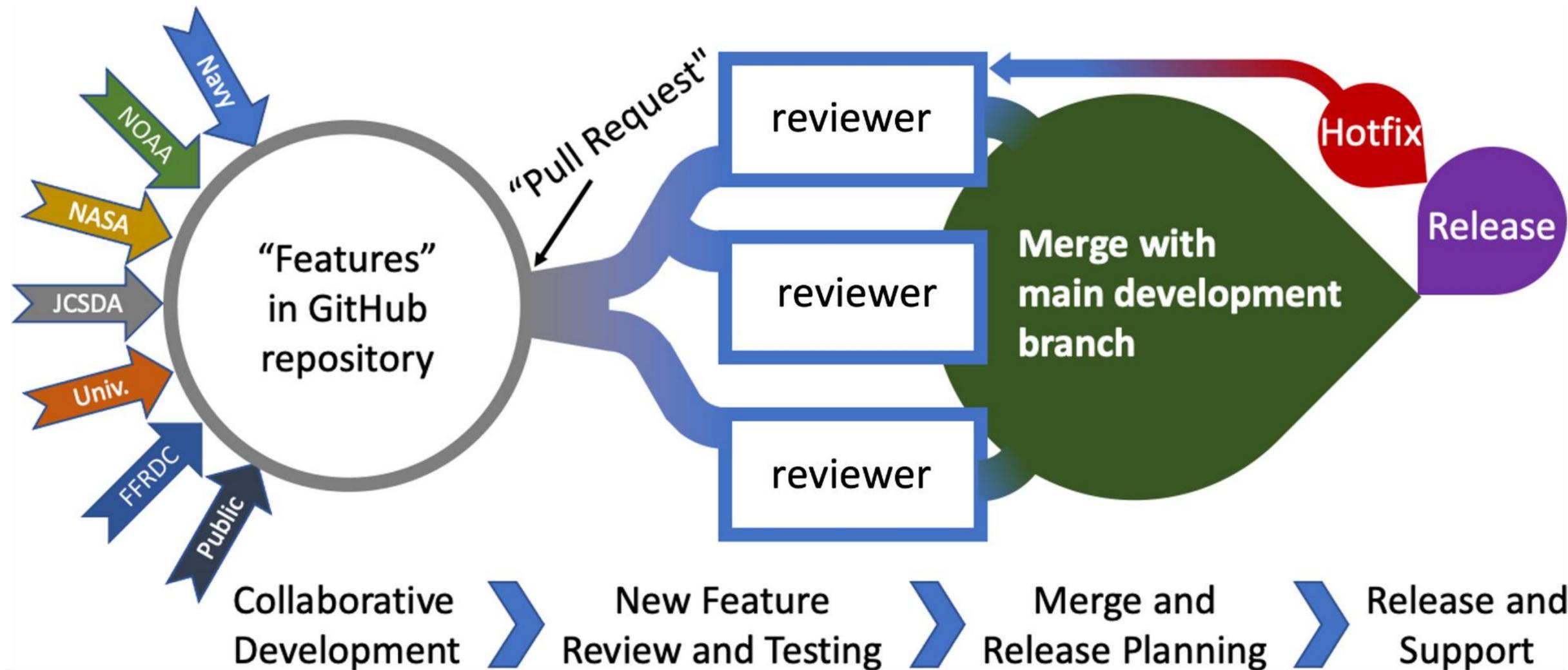


- <https://mer-a-o.github.io/howtojedi/lecture3/>
- Understand the additional elements needed to go from $H(x)$ to Data Assimilation in JEDI
- Running a variational Data Assimilation experiment
- Looking at the results

Closing remarks: CI testing and improving JEDI



- **Test every feature** (generic and specific) or it can **silently break** as other parts evolve.
- **CI tests** prevent regressions and act as a **catalog + runnable examples** (YAML + expected input/outputs).
- **Merge rule:** CI must pass **and a human reviewer** approves.
- **Next tutorial?:** *JEDI for developers* (adding features + writing tests + CI/review workflow).



Using existing JEDI builds on Discover



- Because of the JEDI modularity, you don't necessarily need to build your own JEDI system to use JEDI

see <https://mer-a-o.github.io/howtojedi/discover/>

- You can point to your colleagues builds or the SI team builds:

/discover/nobackup/projects/gmao/advda/jedi_bundles_sles15/current_pinned_jedi_bundle/

- This allows better coordination for joint testing, deployment and then developments.

Closing remarks



- Bi-weekly JEDI office hours starting next week.

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[Join the meeting now](#)

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Passcode: 4vS7kb6q

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