

# OEM Retrievals in ARTS

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## Goals

- Quick and easy setup of retrieval calculations in ARTS
- Good performance



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- Quick and easy setup of retrieval calculations in ARTS
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## Current Status

- OEM implementation based on `invlib` ✓
- Handling covariance matrices ✓
- Computing error statistics ✓
- Setting up retrievals ⚠



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## This Presentation

- General retrieval workflow
- 1D ozone retrieval example



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# The Optimal Estimation Method (OEM)

## Formulation

- Gaussian prior:  $\mathbf{x}_a \in \mathbb{R}^n, \mathbf{S}_x \in \mathbb{R}^{n \times n}$
- Gaussian measurement errors:  $\mathbf{y}_f = \mathbf{F}(\mathbf{x}) \in \mathbb{R}^m, \mathbf{S}_\epsilon \in \mathbb{R}^{m \times m}$

## MAP Estimator

$$\hat{\mathbf{x}} = \arg \min_{\mathbf{x}} (\mathbf{F}(\mathbf{x}) - \mathbf{y})^T \mathbf{S}_\epsilon^{-1} (\mathbf{F}(\mathbf{x}) - \mathbf{y}) + (\mathbf{x} - \mathbf{x}_a)^T \mathbf{S}_x^{-1} (\mathbf{x} - \mathbf{x}_a)$$



# The Optimal Estimation Method (OEM)

## Minimization

- Gauss-Newton:

$$\mathbf{x}_{i+1} = \mathbf{x}_i - \left( \mathbf{K}_{\mathbf{x}_i}^T \mathbf{S}_\epsilon^{-1} \mathbf{K}_{\mathbf{x}_i} + \mathbf{S}_x^{-1} \right)^{-1} \left( \mathbf{K}_{\mathbf{x}_i}^T \mathbf{S}_\epsilon^{-1} (\mathbf{F}\mathbf{x}_i - \mathbf{y}) + \mathbf{S}_x^{-1} (\mathbf{x}_i - \mathbf{x}_a) \right)$$

$$\mathbf{x}_{i+1} = \mathbf{x}_a - \mathbf{S}_x \mathbf{K}_{\mathbf{x}_i}^T \left( \mathbf{K}_{\mathbf{x}_i} \mathbf{S}_x \mathbf{K}_{\mathbf{x}_i}^T + \mathbf{S}_\epsilon \right)^{-1} \mathbf{K}_{\mathbf{x}_i}^T \mathbf{S}_\epsilon^{-1} (\mathbf{F}\mathbf{x}_i - \mathbf{y} - \mathbf{K}_{\mathbf{x}_i}(\mathbf{x} - \mathbf{x}_a))$$



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- Levenberg-Marquardt

$$\mathbf{x}_{i+1} = \mathbf{x}_i - \left( \mathbf{K}_{\mathbf{x}_i}^T \mathbf{S}_\epsilon^{-1} \mathbf{K}_{\mathbf{x}_i} + (1 + \gamma) \mathbf{S}_x^{-1} \right)^{-1} \left( \mathbf{K}_{\mathbf{x}_i}^T \mathbf{S}_\epsilon^{-1} (\mathbf{F}\mathbf{x}_i - \mathbf{y}) + \mathbf{S}_x^{-1} (\mathbf{x}_i - \mathbf{x}_a) \right)$$



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# The Optimal Estimation Method (OEM)

## Computation

- Each optimization step requires solution of a linear system:

$$\underbrace{\left( \mathbf{K}_{\mathbf{x}_i}^T \mathbf{S}_\epsilon^{-1} \mathbf{K}_{\mathbf{x}_i} + \mathbf{S}_x^{-1} \right)}_{\mathbf{M}} \mathbf{a} = \mathbf{b}$$

$$\underbrace{\left( \mathbf{K}_{\mathbf{x}_i} \mathbf{S}_x \mathbf{K}_{\mathbf{x}_i}^T + \mathbf{S}_\epsilon \right)}_{\mathbf{M}} \mathbf{a} = \mathbf{b}$$



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- **Direct solver:**

- Decompose  $\mathbf{M}$  and solve the system forward and back substitution
- Requires explicit computation of  $\mathbf{M}$



# The Optimal Estimation Method (OEM)

## Computation

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$$\underbrace{\left( \mathbf{K}_{\mathbf{x}_i} \mathbf{S}_x \mathbf{K}_{\mathbf{x}_i}^T + \mathbf{S}_\epsilon \right)}_{\mathbf{M}} \mathbf{a} = \mathbf{b}$$

- **Direct solver:**

- Decompose  $\mathbf{M}$  and solve the system forward and back substitution
- Requires explicit computation of  $\mathbf{M}$

- **Conjugate gradient method:**

- Avoids computation of the linear system
- Iterative method, performance depends on  $\mathbf{M}$



# Retrieval Workflow

1. Define retrieval quantities and covariance matrices:

```
retrievalDefInit
Covmat1D(...)           # Create covariance matrix block
retrievalAddAbsSpecies(...) # Add RQ to Jacobian and add block to covmat_sx
...
covmat_blockSetDiagonal(...) # Create diagonal covariance matrix
covmat_seSet(...)          # Set block as covmat_se
retrievalDefClose
```



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2. Define forward model agenda:

```
AgendaSet( inversion_iterate_agenda ){
    ...
}
```



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3. Run oem calculation:

```
OEM(...)
```



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```

2. Define forward model agenda:

```
AgendaSet( inversion_iterate_agenda ){
    ...
}
```

3. Run oem calculation:

```
OEM(...)
```

4. Error analysis:

```
avkCalc      # Averaging kernel matrix
covmat_ssCalc # Smoothing error
covmat_soCalc # Observation noise
```



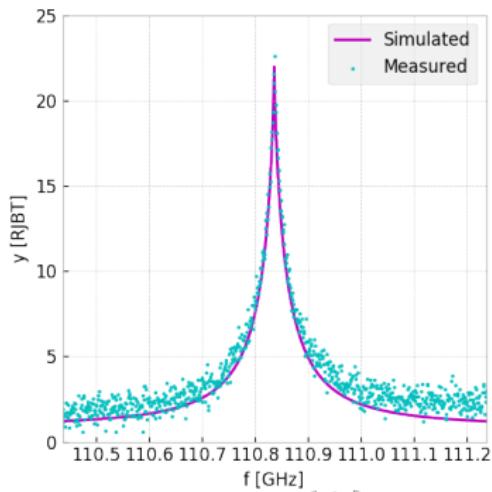
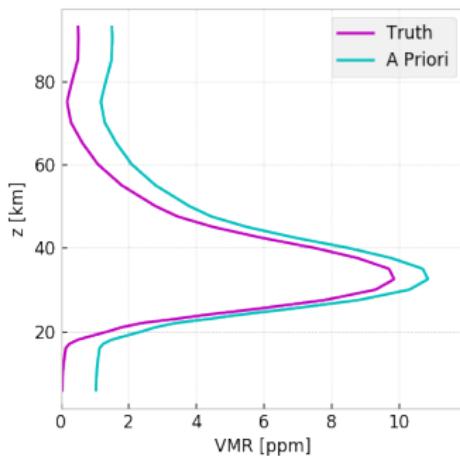
# ARTS Nomenclature

ARTS Name	Mathematical Notation	Description
x	$\mathbf{x} \in \mathbb{R}^n$	State vector
y	$\mathbf{y} \in \mathbb{R}^m$	Measurement vector
yf	$\mathbf{y}_f = \mathbf{F}(\mathbf{x}) \in \mathbb{R}^m$	Simulated measurement vector
covmat_sx	$\mathbf{S}_x \in \mathbb{R}^{n \times n}$	A priori covariance matrix
covmat_se	$\mathbf{S}_\epsilon \in \mathbb{R}^{m \times m}$	Measurement error covariance matrix
avk	$\mathbf{S}_\epsilon \in \mathbb{R}^{n \times n}$	Averaging kernel matrix
covmat_ss	$\mathbf{S}_s \in \mathbb{R}^{n \times n}$	Smoothing error
covmat_so	$\mathbf{S}_o \in \mathbb{R}^{n \times n}$	Retrieval noise
covmat_block		Covariance matrix block
covmat_inv_block		Inverse of covariance matrix block



# 1D Retrieval Example

- Observation of  $O_3$  110.836 GHz
- Ground-based ( $60^\circ$  zenith angle)
- A priori off with 1 ppm
- Noisy measurement with polynomial offset



# 1D Retrieval Example, Retrieval Definition

## Retrieval Definition

- `retrievalDefInit`, `retrievalDefClose`
- Similar to definition of Jacobian quantities

## Covariance Matrices

- `covmat1D`, `covmat1DMarkov`, ... create correlation blocks and store them inside `covmat_block` (`covmat_inv_block`) WSVs
- `retrievalAddAbsSpecies`, `retrievalAddPolyfit`, ... functions add quantity to Jacobian and correlation block in `covmat_block` to `covmat_sx`
- Use `covmat_seSet` to add `covmat_block` to `covmat_se`



# 1D Retrieval Example, Retrieval Definition

```
retrievalDefInit
```

$$\mathbf{S}_x = \left[ \begin{array}{c} \\ \\ \\ \\ \end{array} \right]$$



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# 1D Retrieval Example, Retrieval Definition

```
retrievalDefInit

# Ozone
covmat1D(g1 = z_ret_grid,
          sigma1 = sigma_x,
          lc1 = lcs,
          fname = "gauss")

retrievalAddAbsSpecies(species = "O3",
                      unit = "logrel",
                      g1 = p_ret_grid,
                      g2 = lat_grid,
                      g3 = lon_grid)
```

$$\mathbf{S}_x = \begin{bmatrix} & & \\ & & \\ & & \text{red diagonal} \\ & & \\ & & \end{bmatrix}$$



# 1D Retrieval Example, Retrieval Definition

```
retrievalDefInit

# Ozone
covmat1D(g1 = z_ret_grid,
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retrievalAddAbsSpecies(species = "O3",
                      unit = "logrel",
                      g1 = p_ret_grid,
                      g2 = lat_grid,
                      g3 = lon_grid)

# Polynomial Baseline
covmatSetDiagonal(sigma = sigma_e)
retrievalAddPolyfit(poly_order = 2)
```

$$\mathbf{S}_x = \begin{bmatrix} & & \\ & & \\ & & \text{red diagonal} \\ & & \\ & & \end{bmatrix}$$



# 1D Retrieval Example, Retrieval Definition

```
retrievalDefInit

# Ozone
covmat1D(g1 = z_ret_grid,
          sigma1 = sigma_x,
          lc1 = lcs,
          fname = "gauss")

retrievalAddAbsSpecies(species = "O3",
                      unit = "logrel",
                      g1 = p_ret_grid,
                      g2 = lat_grid,
                      g3 = lon_grid)

# Polynomial Baseline
covmatSetDiagonal(sigma = [10.0])
retrievalAddPolyfit(poly_order = 2)

# S_e
covmatSetDiagonal(sigma = sigma_e)
covmat_seSet
```

$$\mathbf{S}_x = \begin{bmatrix} \text{Red Diagonal Matrix} \end{bmatrix}$$

$$\mathbf{S}_e = \begin{bmatrix} \text{Red Diagonal Matrix} \end{bmatrix}$$



# 1D Retrieval Example, Retrieval Definition

```
retrievalDefInit

# Ozone
covmat1D(g1 = z_ret_grid,
          sigma1 = sigma_x,
          lcl = lcs,
          fname = "gauss")

retrievalAddAbsSpecies(species = "O3",
                      unit = "logrel",
                      g1 = p_ret_grid,
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                      g3 = lon_grid)

# Polynomial Baseline
covmatSetDiagonal(sigma = [10.0])
retrievalAddPolyfit(poly_order = 2)

# S_e
covmatSetDiagonal(sigma = sigma_e)
covmat_seSet

retrievalDefClose
```

$$\mathbf{S}_x = \begin{bmatrix} \text{Red Diagonal Matrix} \end{bmatrix}$$

$$\mathbf{S}_e = \begin{bmatrix} \text{Red Diagonal Matrix} \end{bmatrix}$$



# 1D Retrieval Example, Forwar Model Setup

- Interface between ARTS and OEM implementation
- Should not need to be modified

```
AgendaSet( inversion_iterate_agenda ){

    # Map x to ARTS' variables
    x2artsStandard

    # To be safe, rerun checks dealing with the atmosphere
    atmfields_checkedCalc
    atmgeom_checkedCalc

    # Calculate yf and Jacobian matching x.
    yCalc( y=yf )

    # Add baseline term
    VectorAddVector( yf, yf, y_baseline )

    # This method takes cares of some "fixes" that are needed to get the Jacobian
    # right for iterative solutions. No need to call this WSM for linear inversions.
    jacobianAdjustAfterIteration
}
```



# 1D Retrieval Example, OEM Calculation

## Run OEM Calculation

```
OEM(method="gn_cg")
```

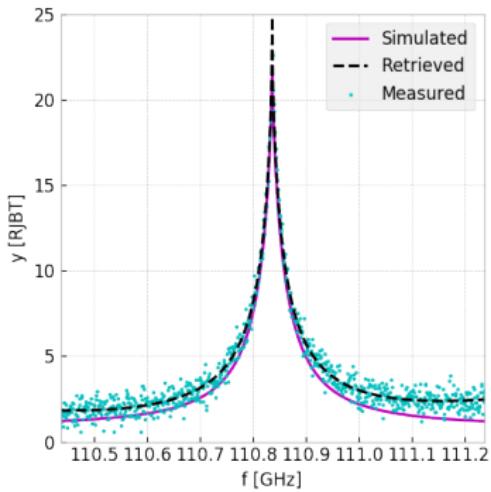
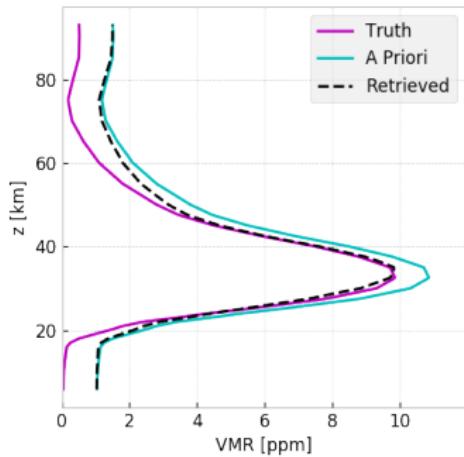
## The Method Argument

Argument String	Minimization Method	Solver	Formulation
"li"	linear	direct	n-form
"li_cg"	linear	CG	n-form
"li_m"	linear	direct	m-form
"li_cg_m"	linear	CG	m-form
"gn"	Gauss-Newton	direct	n-form
"gn_cg"	Gauss-Newton	CG	n-form
"gn_m"	Gauss-Newton	direct	m-form
"gn_cg_m"	Gauss-Newton	CG	m-form
"lm"	Levenberg-Marquardt	direct	n-form
"lm_cg"	Levenberg-Marquardt	CG	n-form



# 1D Retrieval Example, Step 2

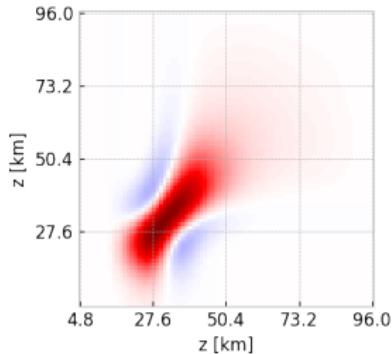
## Results



# Error Analysis

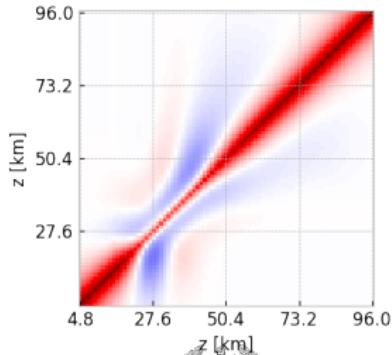
## Averaging Kernel Matrix

```
# Averaging Kernel:  
avkCalc
```



## Smoothing Error

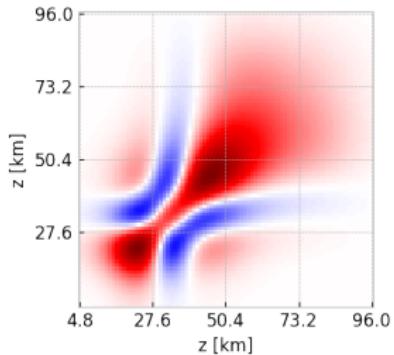
```
# Smoothing Error:  
covmat_ssCalc
```



# Error Analysis

## Observation Noise

```
# Observation Noise  
covmat_soCalc
```



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# Conclusion and Outlook

## Conclusions

- Extensive OEM functionality available in ARTS
- Some work on setting up retrievals remains

## Future Work

- Perform retrievals
- Extend and improve retrieval functionality in arts
- Examples and documentation

