



MIRART

Weighting Function Implementation

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1. Overview on MIRART
2. Automatic Differentiation
3. Problems and Solutions

MIRART

Modular InfraRed Atmospheric Radiative Transfer

Line shapes: Voigt, VanVleck⊗Doppler, Lorentz

Line data: HITRAN, HITEMP, GEISA, JPL, ...

Continua: H₂O (CKD), CO₂ (Ridgway/GenLn2), N₂, O₂, “dry air” (Liebe)

Geometries: Limb, uplooking, downlooking (refraction optional)

Instruments:

- line shape: FTS (optionally apodized), Heterodyne, Fabry-Perot
- field-of-view: Box, Gauss, Trapez

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Verification: AMIL2DA, IRTMW01, PIRAMHYD, ...

Weighting functions: Automatic differentiation (molecular concentration, temperature)

Numerics — LbL: Voigt= $\Re(\text{cerf})$, wavenumber grid adjusted to level (p , T) and molecule

Numerics — path integrals: numeric quadrature schemes

Implementation: Fortran 77 with BLAS and SLATEC libraries

Interfaces: “classical” input file, Python scripts, web interface (<http://v1.nz.dlr.de>)



Weighting Functions

Weighting functions — Jacobians — derivatives wrt. gas densities/VMR, temperature, ...
required for retrieval applications (nonlinear least squares) and sensitivity analysis

Approaches

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- Finite difference approximations:
 - time consuming
 - step size selection:
(cancellation errors for small steps and truncation errors for large steps)
- Hand coded analytical derivatives:
 - boring and error prone,
 - no “automatic update” of derivative code if forward model is improved
- Symbolic differentiation:
 - impracticable: no lbl ir rad transfer code in MATHEMATICA, MAPLE, ...
 - generally leads to huge expressions
- [Automatic differentiation](#)

Weighting Functions

Automatic Differentiation

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- Even large codes are essentially formulated in terms of elementary mathematical operations (sums, products, powers) and elementary functions
- (In contrast to integration) differentiation is based on simple recipes such as the chain rule
- Differentiation rules can be performed automatically by some kind of precompiler
- ★ Automatic differentiation tools available for Fortran, C, ...
- ★ Automatic differentiation generates exact derivatives

→

ADIFOR



Automatic Differentiation

ADIFOR

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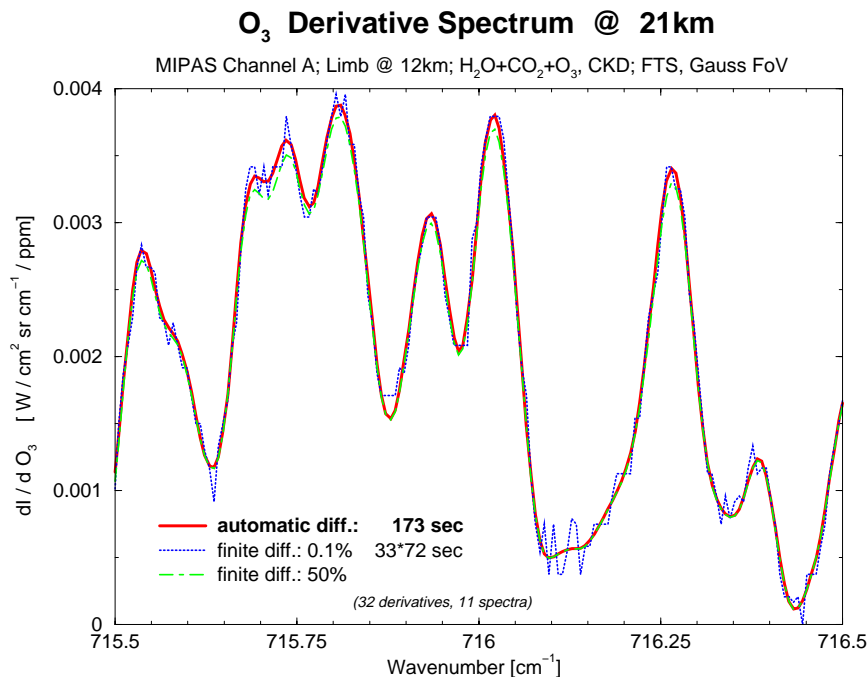
1. Top-level subroutine, independent ('x') and dependent ('y') variable(s), max. number of independent variables
2. Composition file (relevant routines and "main")
3. Generate derivative code
4. Adjust routine calling the top-level subroutine, initialize seed matrix
5. Compile and link (incl. ADIFOR libraries)

```
AD_TOP      = schwarzschild_ils_fov
AD_IVARS    = Density
AD_DVARS    = Radiance
AD_PMAX     = 50
```

```
dummy_main.f
schwarzschild_ils_fov.f
schwarzschild.f
convolve_ils.f
convolve_fov.f
.....
```

```
.....
y = a*sin(b*x+c)
g_y = a*b*cos(b*x+c)
.....
```

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Automatic Differentiation

Problems & Solutions

- ADIFOR significantly increases memory requirements:
For each (intermediate or final) variable a gradient variable is generated:

```
REAL SCALAR    → REAL g_SCALAR(g_p)
REAL VECTOR(N) → REAL g_VECTOR(g_p, N)
```

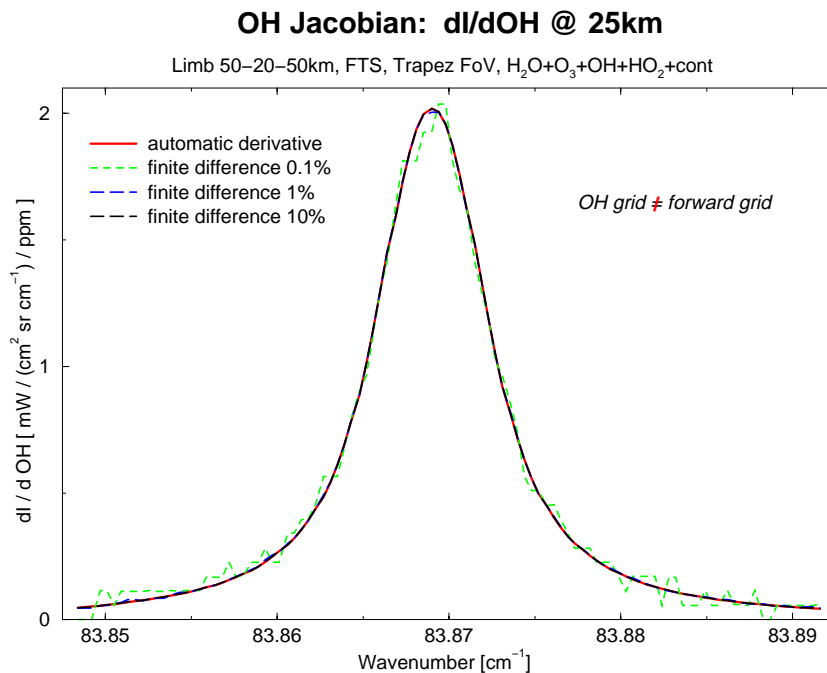
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- ▷ compact storage scheme for cross sections
- ▷ individual executables depending on task:

```
SQUIRRL  — pure forward code
Co1DRAT  — column derivatives          (g_p=MxGas)
MADRAT   — molecular profile derivatives (g_p=MxLevel)
FATRAT   — molecular or temperature profile derivatives (g_p=MxLevel)
```

- Retrieval grid \neq forward grid
 - ▷ interpolate to forward grid “inside derivative code”
- Temperature derivatives: $n = p/kT$ violated in old implementation

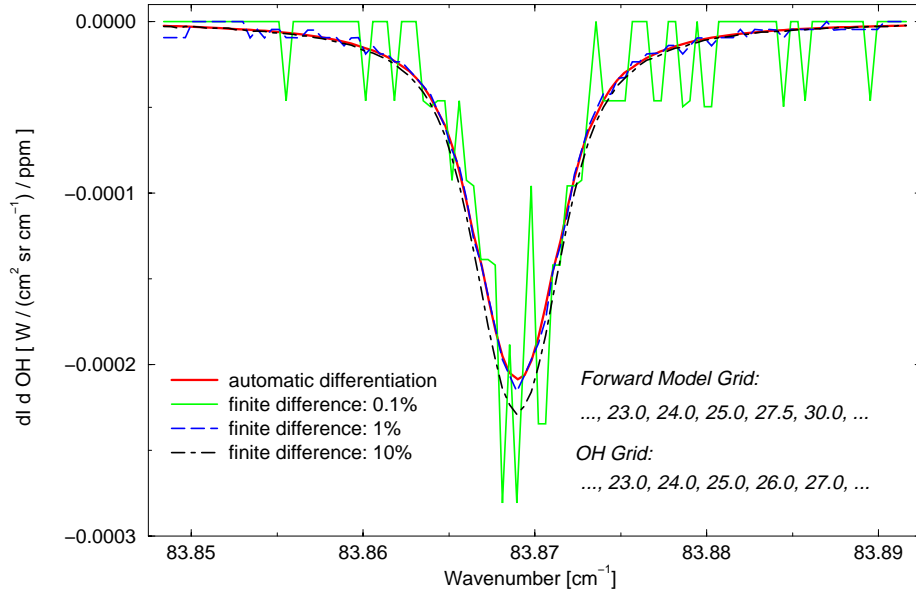
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OH Jacobian: dI/dOH @ 26km

Limb 50–20–50km, FTS, Trapez FoV, $H_2O+O_3+OH+HO_2+cont$



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