

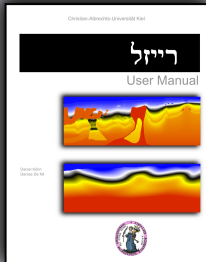
# Introduction to frequency domain modelling and FWI of georadar data with GERMAINE

Daniel Köhn, Denise De Nil, Wolfgang Rabbel

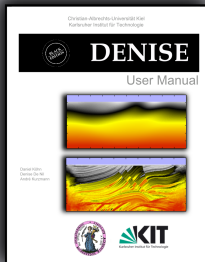
September 27, 2017

# Overview of (seismic) inversion codes

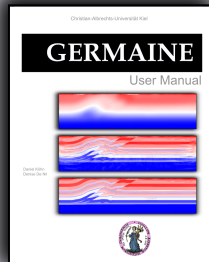
## Software packages



2D first arrival  
traveltime tomography



2D time-domain  
full waveform inversion



2D frequency-domain  
full waveform inversion

- **Forward modelling:** Cartesian finite differences
- **Inverse problem:** gradient-based local optimization
- **MPI parallelization:** shots and domain decomposition
- **Code repository:** <https://github.com/daniel-koehn/>

# Introduction to FDFD modelling and FWI of georadar data with GERMAINE

- 1 GERMAINE installation
- 2 The forward problem
  - Modelling example: homogeneous full space model
  - Modelling example: 2-cross model
- 3 Full Waveform Inversion
  - GERMAINE FWI example: 2-cross model

# GERMAINE installation

# GERMAINE installation requirements

- **C-compiler:** gcc, Intel icc
- **MPI:** OpenMPI, Intel-MPI
- **Optimized BLAS/LAPACK library,** e.g. OpenBLAS or Intel MKL
- **Suite Sparse:**  
<http://faculty.cse.tamu.edu/davis/suitesparse.html>
- **Git** (optional)
- **Seismic Unix:** <http://www.cwp.mines.edu/cwpcodes/>
- **Numpy, SciPy, Matplotlib:** <https://www.scipy.org/>  
e.g. using Anaconda <https://www.anaconda.com/download/>

- ... on the NEC HPC-Linux cluster of RZ Kiel

[Click here](#)

- ... Desktop computer at Department of Geophysics

[Click here](#)

## Get the benchmark model ...

- 1 Clone TE-mode model and parameter files from:

<https://github.com/daniel-koehn/DENISE-Benchmark>

by typing:

```
git clone https://github.com/daniel-koehn/DENISE-Benchmark.git
```

- 2 Copy model and parameter files for the TE-mode model examples from DENISE-Benchmark to GERMAINE

```
cp DENISE-Benchmark/2_cross_TE_model/input_files/GERMAINE* GERMAINE/par/  
cp DENISE-Benchmark/2_cross_TE_model/start/2_cross_TE_* GERMAINE/par/start/  
cp DENISE-Benchmark/2_cross_TE_model/receiver/receiver_2_cross_TE.dat GERMAINE/par/receiver/  
cp DENISE-Benchmark/2_cross_TE_model/source/source_2_cross_TE.dat GERMAINE/par/source/
```

# The forward problem



# The forward problem

## Acoustic frequency domain forward problem

$$\nabla^2 P + \frac{\omega^2}{c^2} P = f$$

with

- $P$  the pressure wavefield [Pa],
- $f$  the source term [Pa/m<sup>2</sup>]
- $c$  the **P-wave velocity** [m/s],
- $\omega$  the angular **frequency** [rad/s].

## TE-mode frequency domain forward problem

$$\nabla^2 E_y + \mu_0 \epsilon_e \omega^2 E_y = -i\omega \mu_0 J_{sy}$$

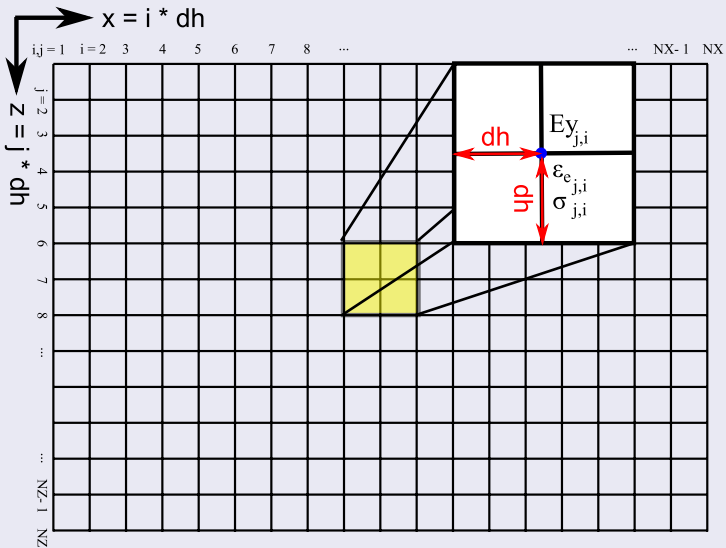
$$\epsilon_e = \epsilon + \frac{i\sigma}{\omega}$$

with

- $E_y$  the electric wavefield [V/m],
- $J_{sy}$  the source current [A/m<sup>2</sup>]
- $\mu_0 = 4\pi \cdot 10^{-7}$  the free space magnetic permeability [H/m]
- $\epsilon$  the **dielectric permittivity** [F/m],
- $\sigma$  the **electrical conductivity** [S/m],
- $\omega$  the angular **frequency** [rad/s].

# Numerical solution of TE-mode equation

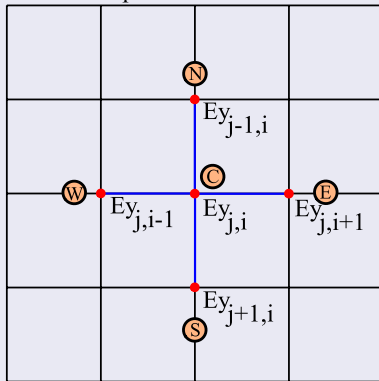
## Discretization of $\epsilon_e$ , $\sigma$ and $E_y$



# Numerical solution of TE-mode equation

## Approximation of Laplace operator by finite-differences

5-point FD Stencil



$$\nabla^2 E_y(\mathbf{x}, t) \approx \frac{E_{y,j,i-1} + E_{y,j,i+1} + E_{y,j-1,i} + E_{y,j+1,i} - 4E_{y,j,i}}{dh^2}$$

with  $dh$  = spatial grid point distance

# Numerical solution of TE-mode equation

## Discretization of Helmholtz equation

Which leads to the discretized wave equation:

$$\frac{E_{y_{j-1,i}} + E_{y_{j,i-1}}}{dh^2} + \left( \mu_0 \epsilon_e \omega^2 - \frac{4}{dh^2} \right) E_{y_{j,i}} + \frac{E_{y_{j,i+1}} + E_{y_{j+1,i}}}{dh^2} = -i\omega \mu_0 J_{sy_{j,i}}.$$

This can be written as matrix equation

$$\mathbf{Ax} = \mathbf{b}$$

where

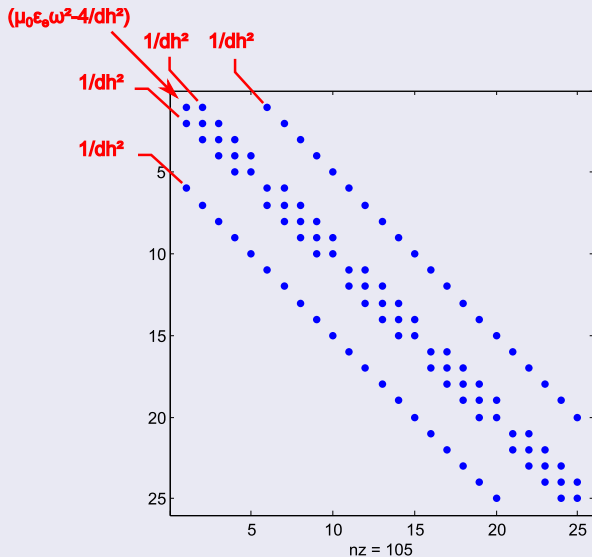
$\mathbf{A}$  = impedance matrix

$\mathbf{x}$  = solution vector

$\mathbf{b}$  = source vector

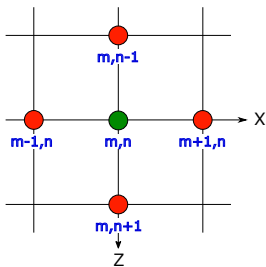
# Numerical solution of TE-mode equation

## Structure of matrix A for a 5 x 5 gridpoint model

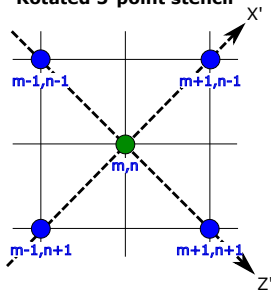


# Numerical solution of TE-mode equation

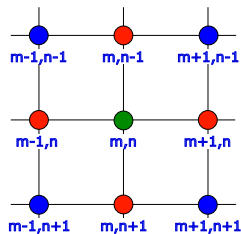
**Classical 5-point stencil**



**Rotated 5-point stencil**



**Mixed-grid 9-point stencil**



adapted from [Jo et al., 1996]

- Minimizing numerical anisotropy and grid dispersion by using mixed-grid 9-point finite-difference scheme with mass-lumping: [Hustedt et al., 2004] [Operto et al., 2009]

# Numerical solution of TE-mode equation

## Perfectly Matched Layers (PMLs)

Introduction to Uni-axial (UPMLs) and Stretched-Coordinate (SC-PMLs):

- Lecture notes (lecture 9):  
<http://emlab.utep.edu/ee5390cem.htm>
- Full lecture available on YouTube:  
[https://www.youtube.com/watch?v=w\\_NnRZ1NuAA](https://www.youtube.com/watch?v=w_NnRZ1NuAA)

## Basic ideas of PMLs

- 1 Introduce artificial attenuation of EM-waves in a boundary frame around the model domain  
**Problem: new artificial impedance contrast**
- 2 Reduce material parameters in boundary frame to match impedances  
**Problem: damping effect depends on incidence angle, frequency and polarization of EM wave**
- 3 Introduce artificial anisotropy in boundary frame which leads to the UPML formulation
- 4 Move PML parameters from RHS to LHS of Maxwell equations and associate them with the curl operator. The PML parameters are now "stretching" the coordinates (SC-PMLs)

# Numerical solution of TE-mode equation

## Solution of the problem $Ax=b$ using a direct solver

- 1 Decompose  $A$  into an upper  $U$  and lower  $L$  triangular matrix:  
 $A = LU$
- 2 Solve the problem  $Ly = b$  by forward substitution
- 3 Solve the problem  $Ux = y$  by backward substitution

In GERMAINE we use the direct solver UMFPACK  
(Unsymmetric MultiFrontal PACKage)

<http://faculty.cse.tamu.edu/davis/suitesparse.html>

## Future implementation of parallel direct solvers

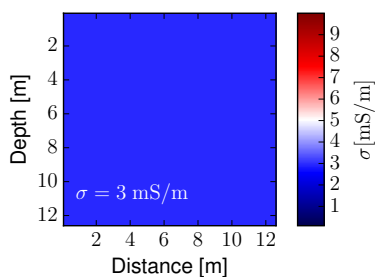
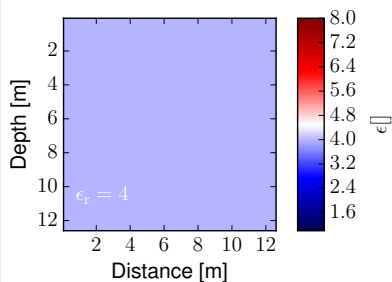
- The currently implemented shot-frequency parallelization is not really efficient. Especially in case of small problems the LU-factorization is the computationally most demanding part
- UMFPACK will be replaced by a parallel solver like MUMPS or PARADISO in a future release



# Modelling example: homogeneous full space model

# Modelling example: homogeneous full space model

Homogeneous model [Meles et al., 2011, Lavoué et al., 2014]

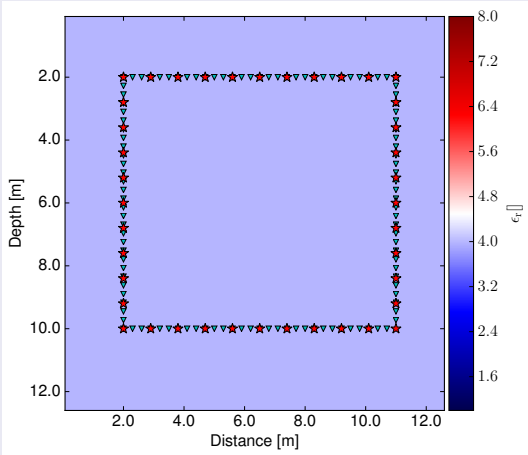


## Model discretization:

- $NX = 180$  gridpoints  $\times$   $NZ = 180$  gridpoints
- + 10 gridpoints PMLs on each side
- $DH = 7$  cm

# Modelling example: homogeneous full space model

## Perfect illumination acquisition geometry



- 40 shots and 117 receivers
- Modelling of frequencies between 50 MHz - 200 MHz

## GERMAINE structure

- **bin**  
This directory contains the executable program GERMAINE
- **include**  
This directory contains header files and global parameters
- **jobs**  
This directory contains Batch-scripts to submit GERMAINE modelling/inversion runs on HPCs with PBS-batch system
- **par**  
Parameter files for GERMAINE modelling/inversion and Jupyter notebooks for visualization
- **src**  
This directory contains the complete source codes

## GERMAINE modelling/FWI files

- **par/GERMAINE\_2\_cross\_TE.inp**  
Defines modelling and fixed FWI parameters
- **par/GERMAINE\_workflow\_2\_cross\_TE.inp**  
Definition of frequencies and variable FWI parameters
- **par/receiver/receiver\_2\_cross\_TE.dat**  
Definition of receiver positions
- **par/source/source\_2\_cross\_TE.dat**  
Definition of source positions
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Model files discretized on Cartesian grid

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Model files discretized on Cartesian grid

- **GERMAINE Modus:**

```
#----- GERMAINE modus -----  
forward_modelling_(yes=0)_FDFWI_(yes=1)_RTM_(yes=2)_(INVMAT) = 0  
#
```

Frequency domain forward modelling	INVMAT = 0
Frequency domain FWI	INVMAT = 1
Reverse Time Migration	INVMAT = 2

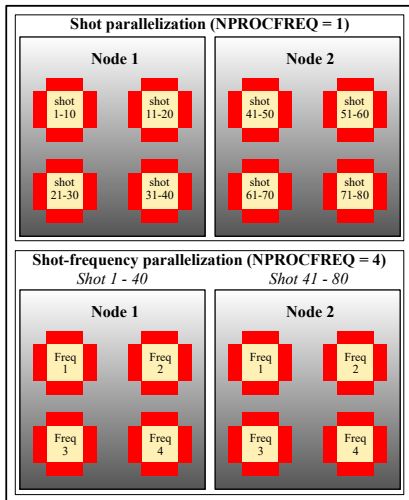
- **GERMAINE Physics:**

```
# ----- GERMAINE Physics -----  
(2D-AC=1;2D-SH=2;2D-TE=4)_(PHYSICS) = 4  
#
```

2D acoustic	PHYSICS = 1
2D elastic SH	PHYSICS = 2
2D TE mode	PHYSICS = 4

# GERMAINE parameter file GERMAINE\_2\_cross\_TE.inp

```
#----- MPI Parallelization -----  
number_of_MPI_processes_for_frequency_parallelization_(NPROCFREQ) = 1
```



## Shot parallelization

- Assume we want to model FD wavefields for 1 frequency and 80 shots on two CPUs with 4 cores, you can distribute the shots over the 8 cores by setting  $NPROCFREQ = 1$

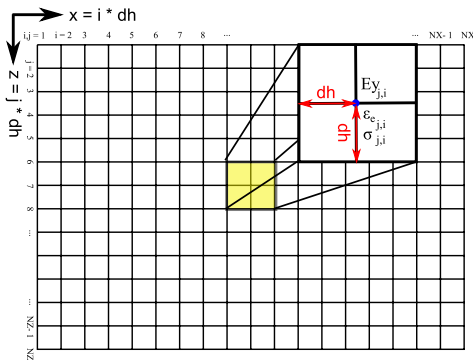
## Shot-frequency parallelization

- If you want to model FD wavefields for 4 frequencies simultaneously, you can assign frequencies to specific cores by setting  $NPROCFREQ = 4$ . In this case the shots are distributed over nodes 1 and 2. If you have more frequencies than cores, the frequencies are distributed over the cores.



- Spatial model discretization

```
#----- 2-D Grid -----
number_of_gridpoints_in_x-direction_(NX) = 180
number_of_gridpoints_in_y-direction_(NZ) = 180
distance_between_gridpoints(in_m)_(DH) = 0.07
```



- **Source geometry and wavelet**

```
#-----Source-----  
# File with source positions:  
SOURCE_FILE = ./source/source_2_cross_TE.dat  
# Read time-domain source wavelet from SU file:  
READ_WAVELET = 0  
WAVELET_NAME = start/wavelet.su  
#
```

Source positions are defined in SOURCE\_FILE

An external time-domain wavelet can be read from a SU file (READ\_WAVELET=1). Frequency domain data for a given frequency of the wavelet is calculated by DFT.

If READ\_WAVELET=0 a spike wavelet is assumed.

- **Model file**

```
#----- Model -----  
read_model_parameters_from_MFILE(yes=1)(READMOD) = 1  
MFILE = start/2_cross_TE_true  
#
```

If READMOD=1, the permittivity  $\epsilon$  [F/m] and conductivity  $\sigma$  [S/m] grids are read from external binary files. MFILE defines the basic file name that is expanded by the following extensions:  
Permittivity: ".eps", Conductivity: ".sig". For this example:

```
start/2_cross_TE_true.eps  
start/2_cross_TE_true.sig
```

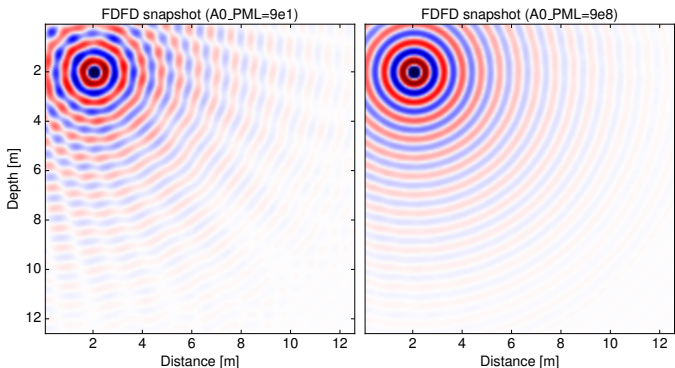
- **Free Surface boundary condition**

```
#-----Free Surface-----  
free_surface_(yes=1)(FREE_SURF) = 0  
#
```

Not implemented for TE-mode.

- PML boundary frame

```
#----- Define PML parameters -----  
width_of_absorbing_frame_(in_gridpoints)_(NPML) = 10  
damping_parameter_(A0_PML) = 9e8  
#
```



PML layers defined outside of the modelling domain.

- **Snapshots**

```
#-----Snapshots-----  
output_of_Ey_field_(SNAP)(yes>0) = 0  
increment_x-direction_(IDX) = 1  
increment_y-direction_(IDY) = 1  
data-format_(SNAP_FORMAT)(ASCII(2);BINARY(3)) = 3  
basic_filename_(SNAP_FILE) = ./snap/2_cross_TE  
#
```

If SNAP=1, the real part of the monochromatic FD wavefield  $E_y$  will be written to file SNAP\_FILE for each shot. This requires that only one stage and frequency is defined in the GERMAINE workflow file. [Details of GERMAINE workflow file](#)

- Receiver

```
#----- Receiver -----  
output_of_seismograms_(SEISMO) = 1  
file_with_receiver_positions_(REC_FILE) = ./receiver/receiver_2_cross_TE  
read_receiver_positions_from_multiple_files_(yes=1)_(READ_REC) = 0  
reference_point_for_receiver_coordinate_system_(REFREC) = 0.0 , 0.0  
#
```

Read receiver positions from ASCII file REC\_FILE.

By setting READ\_REC = 1 different receiver positions for each shot can be read from separate receiver files:

REC\_FILE\_shot\_1.dat, REC\_FILE\_shot\_2.dat ...

- **FD seismograms**

```
#----- FD pressure seismograms -----
Outputfile_for_pressure_seismograms_(PICKS_FILE) = ./seis/2_cross_TE
#
```

Write frequency domain data for all shots at the receiver positions to the IEEE little endian binary file PICKS\_FILE for each stage defined in the GERMAINE workflow file. For details regarding the used data structure I refer to the output in [src/AC/write\\_seis\\_AC.c](#):

```
FILE *fp;
sprintf(pickfile_char,"%s_p_stage_%d.bin",PICKS_FILE,nstage);
fp=fopen(pickfile_char,"wb");

for(k=1;k<=NF;k++){
    for(j=1;j<=nshots;j++){
        for(i=1;i<=ntr;i++){

            index = i + ntr * (j-1) + ntr * nshots * (k-1);
            tmp = precr_vec[index];
            tmp1 = preci_vec[index];

            fwrite(&tmp, sizeof(float), 1, fp);
            fwrite(&tmp1, sizeof(float), 1, fp);

        }
    }
}
fclose(fp);
```



- **Forward modelling log file**

```
# each PE is printing log-information to LOG_FILE.MYID
log-file_for_information_about_progress_of_program_(LOG_FILE) = log/GERMAINE.log
info_of_processing_element_zero_to_stdout_(yes=1/no=0)_(LOG) = 1
```

Write details of forward modelling to log-file LOG\_FILE.

## GERMAINE modelling/FWI files

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Definition of receiver positions
- **par/source/source\_2\_cross\_TE.dat**  
Definition of source positions
- **par/start/2\_cross\_TE\_\***  
Model files discretized on Cartesian grid

- **GERMAINE workflow file (forward modelling)**

```
PRO      FC_low    FC_high    NF    ...  
0.0      50e6      200e6      1     ...  
0.0      60e6      200e6      1     ...  
...
```

Only columns 2-4 are important for FD forward modelling:

**2<sup>nd</sup> column** = lowest frequency modelled with FDFD [Hz]

**3<sup>rd</sup> column** = highest frequency modelled with FDFD [Hz]

**4<sup>th</sup> column** = number of frequencies NF modelled between  
FC\_low and FC\_high

If  $NF > 1$  a frequency sample interval

$$df = \frac{FC_{high} - FC_{low}}{NF - 1}$$

between FC\_low and FC\_high is assumed.

- **GERMAINE workflow file (forward modelling)**

```
PRO      FC_low      FC_high      NF      ...  
0.0      100e6         200e6         1      ...
```

If you want to model monochromatic FD wavefield snapshots, the workflow file has to contain only 1 line and 1 frequency  $NF=1$ .

For the example above, FD snapshots are calculated for a frequency of 100 MHz.

- **GERMAINE workflow file (forward modelling)**

```
PRO      FC_low    FC_high    NF      ...  
0.0      50e6       200e6     1000   ...
```

To produce time-domain data from frequency-domain data you can model a whole bunch of frequencies in 1 stage.

To reduce computation time, I recommend to use the GERMAINE frequency parallelization by setting NPROCFREC in the GERMAINE parameter file to the maximum number of cores available on your computer.

## GERMAINE modelling/FWI files

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Definition of source positions
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Model files discretized on Cartesian grid

- Receiver positions

2.00	2.00
2.30	2.00
2.60	2.00
2.90	2.00
3.20	2.00
3.50	2.00
3.80	2.00
4.10	2.00
...	

**line n** = receiver coordinate of receiver no. n

**1<sup>st</sup> column** = receiver x-coordinate [m]

**2<sup>nd</sup> column** = receiver z-coordinate [m]

## GERMAINE modelling/FWI files

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Definition of frequencies and variable FWI parameters
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Definition of receiver positions
- **par/source/source\_2\_cross\_TE.dat**  
Definition of source positions
- **par/start/2\_cross\_TE\_\***  
Model files discretized on Cartesian grid



- **Source positions**

40

2.00 0.00 2.00 0.00 10.00 1.00 0.00 3.00

2.90 0.00 2.00 0.00 10.00 1.00 0.00 3.00

3.80 0.00 2.00 0.00 10.00 1.00 0.00 3.00

4.70 0.00 2.00 0.00 10.00 1.00 0.00 3.00

5.60 0.00 2.00 0.00 10.00 1.00 0.00 3.00

6.50 0.00 2.00 0.00 10.00 1.00 0.00 3.00

7.40 0.00 2.00 0.00 10.00 1.00 0.00 3.00

...

**1<sup>st</sup> line** = number of sources

**line n+1** = source coordinate of receiver no. n

**1<sup>st</sup> column** = source x-coordinate [m]

**3<sup>rd</sup> column** = source z-coordinate [m]

**Other columns** = can be ignored

## GERMAINE modelling/FWI files

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- **par/source/source\_2\_cross\_TE.dat**  
Definition of source positions
- **par/start/2\_cross\_TE\_\***  
Model files discretized on Cartesian grid

- **Create model files in Matlab**

```
% loop over model grid
for i=1:NX
    for j=1:NZ

        % create your fancy epsilon and sigma models here ...
        eps_model(j,i) = ...
        sig_model(j,i) = ...

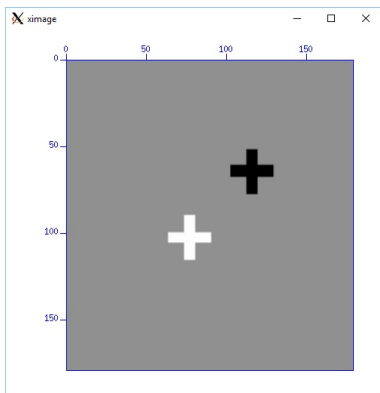
    end
end

% write epsilon model to binary file
file1=['my_epsilon_model.eps'];
fid1=fopen(file1,'w','ieee-le');
fwrite(fid1,eps_model,'float')
fclose(fid1);
```

- **Create model files in Python**

[Example Jupyter notebook for model generation](#)

- **Quick model check with Seismic Unix**



```
ximage n1=180 < start/2_cross_TE_true.eps
```

with  $n1 = NZ$

- **Reading model files in Matlab**

```
% load GERMAINE_epsilon_model
file=['GERMAINE_epsilon_model.eps'];
disp([' loading file ' file]);
fid=fopen(file,'r','ieee-le');
eps_model=fread(fid,[NZ,NX],'float');
fclose(fid);

% plot model
imagesc(eps_model);
```

- **Reading model files in Python**

Example Jupyter notebook to read GERMAINE models

Running GERMAINE for the  
homogeneous model ...

## Running GERMAINE for the homogeneous model ...

- 1 In GERMAINE/par/GERMAINE\_2\_cross\_TE.inp set the following parameters for a forward modelling run of the homogeneous model

```
forward_modelling_(yes=0)_FDFWI_(yes=1)_RTM_(yes=2)_(INVMAT) = 0
MFILE = start/2_cross_TE_init
output_of_P_field_(SNAP)(yes>0) = 1
```

- 2 Copy GERMAINE/par/GERMAINE\_workflow\_2\_cross\_TE.inp to GERMAINE\_workflow\_2\_cross\_TE\_1.inp and remove every line except the first two:

```
PRO  FC_low  FC_high  NF  ...
0.0  50e6    200e6    1   ...
```

- 3 Run GERMAINE forward modelling on e.g. 4 cores of your CPU from the /par directory:

```
mpirun -np 4 ../bin/germaine GERMAINE_2_cross_TE.inp GERMAINE_workflow_2_cross_TE_1.inp
```

# Modelling example: homogeneous full space model

- 4 For each shot defined in source/[source\\_2\\_cross\\_TE.dat](#) the real and imaginary parts of the pressure wavefield at the receiver positions defined in receiver/[receiver\\_2\\_cross\\_TE.dat](#) are written to

[seis/incl2\\_p\\_stage\\_\[fwi\\_stage\].bin](#)

for each frequency (group) defined in the workflow file

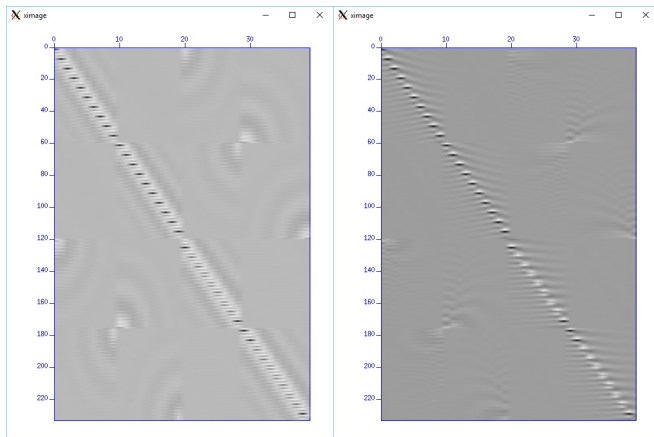
[GERMAINE\\_workflow\\_2\\_cross\\_TE\\_1.inp](#)



# Modelling example: homogeneous full space model

## GERMAINE quick and dirty FD data visualization ...

5 ... with Seismic Unix:



```
ximage n1=234 < seis/2_cross_TE_p_stage_1.bin
```

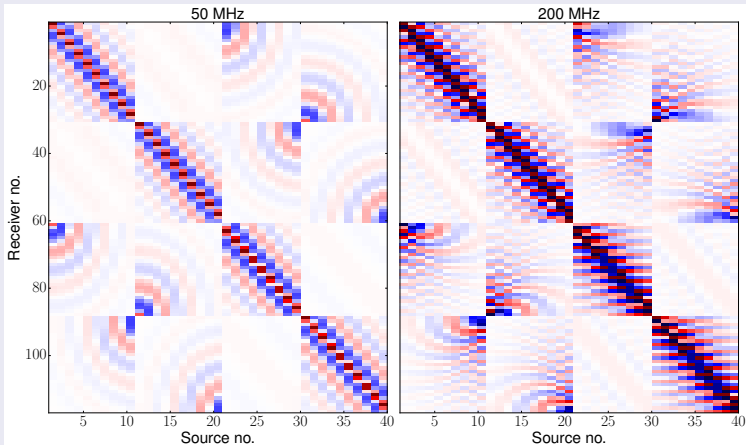
with  $n1 = 2 * \text{number of receivers}$

# Modelling example: homogeneous full space model

## GERMAINE publication-ready FD data visualization ...

5 ... with Matplotlib:

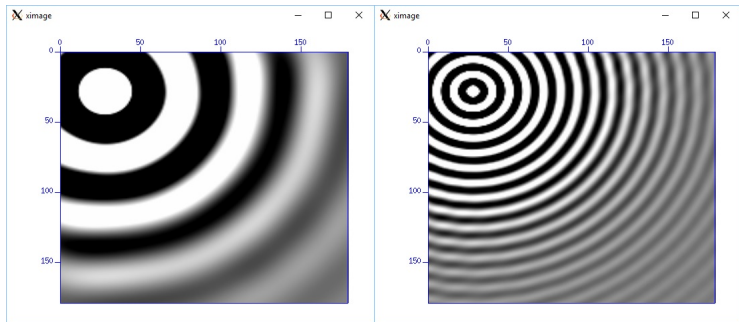
### Frequency domain data $\Re(E_y)$



[Jupyter notebook for frequency domain data visualization](#)

## GERMAINE quick and dirty FD wavefield visualization ...

6 ... with Seismic Unix:



```
ximage n1=180 clip=8e0 < snap/2_cross_TE_shot_1.p
```

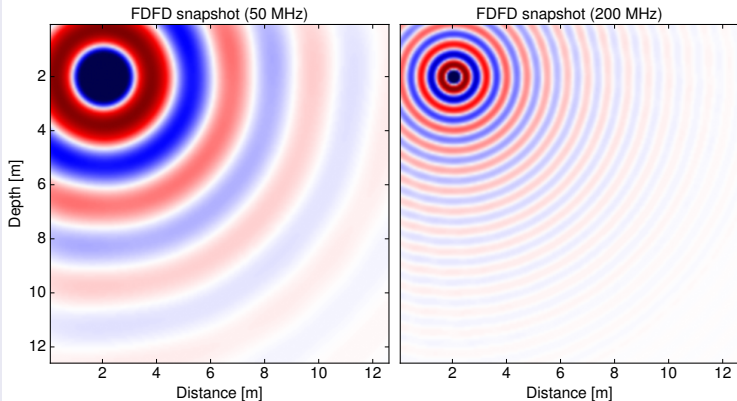
with  $n1 = NZ$

# Modelling example: homogeneous full space model

## GERMAINE publication-ready FD wavefield visualization ...

6 ... with Matplotlib:

Frequency domain data  $\Re(E_y)$



[Jupyter notebook for frequency domain wavefield visualization](#)

## Creating TD radargrams from FD data

- 1 Create a Ricker source wavelet with a centre frequency of 100 MHz using this [Jupyter notebook](#)

- 2 Add SU header to the [wavelet.bin](#) file

```
suaddhead n1=1001 < wavelet.bin | sushw key=dt a=250 > tmp.su
```

where dt denotes the sampling rate [1e-12 s]

- 3 If you use a SU version which supports only XDR SU format, you additionally have to apply:

```
suswapbytes < tmp.su > wavelet.su
```

## Creating TD radargrams from FD data

- 4 Move [wavelet.su](#) to the par/start directory

```
mv par/visu/wavelet.su par/start/
```

- 5 Activate the option to read external source wavelets in GERMAINE\_2\_cross\_TE.inp:

```
# Read time-domain source wavelet from SU file:  
READ_WAVELET = 1  
WAVELET_NAME = start/wavelet.su
```

- 6 Modify the GERMAINE\_workflow\_2\_cross\_TE\_1.inp to model 400 frequencies:

```
PRO    FC_low    FC_high    NF  
0.0    1e6        200e6     400 ...
```

## Creating TD radargrams from FD data

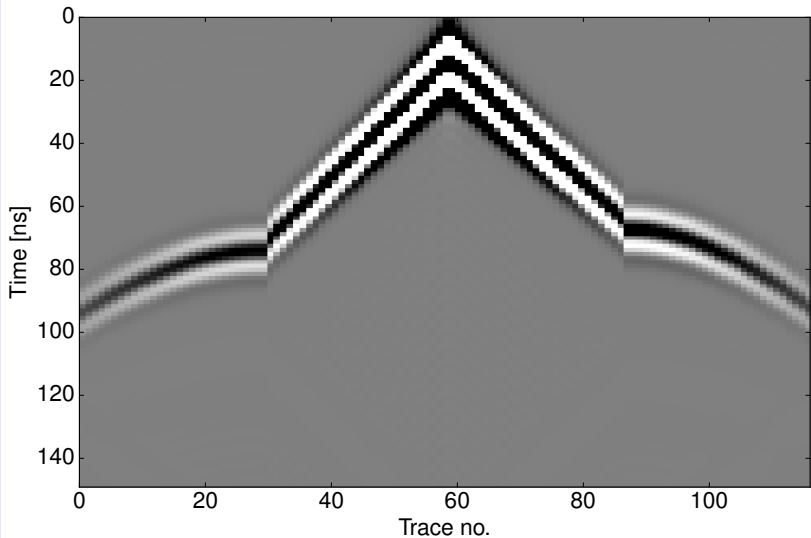
- 7 Run GERMAINE to create FD data

```
mpirun -np 4 ../bin/germaine GERMAINE_2_cross_TE.inp GERMAINE_workflow_2_cross_TE_1.inp
```

- 8 Calculate IFFT and visualization of the resulting TD radargrams with this [Jupyter notebook](#) should lead to TD data like this ...

# Modelling example: homogeneous full space model

Time-domain radargrams  $E_y$





## Analytical solution for 2D homogeneous medium

$$E_y^{\text{analytic}}(x, z, \omega) = \frac{\omega \mu_0 J_{sy}}{4} H_0^{(1)}(kr)$$

$$k = \omega \sqrt{\mu_0 \epsilon_e}$$

$$\epsilon_e = \epsilon + \frac{i\sigma}{\omega}$$

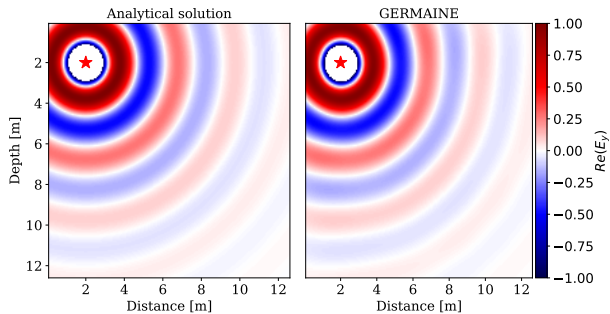
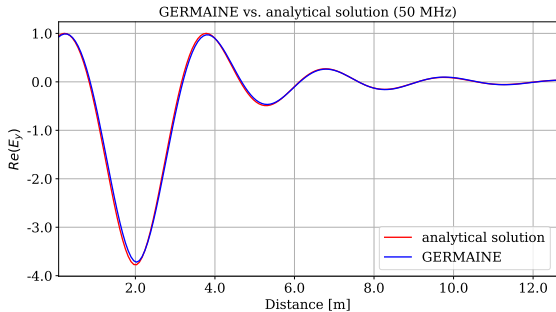
$$r = \sqrt{(x - x_s)^2 + (z - z_s)^2}$$

with

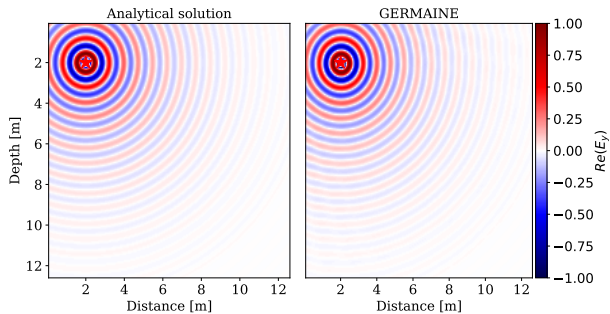
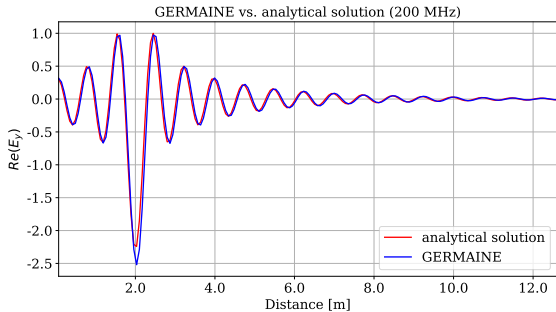
- $E_y$  the electric wavefield,
- $J_{sy} = \delta(x - x_s)\delta(z - z_s)$  the source current
- $x_s, z_s$  the source position
- $\omega$  the angular frequency [rad/s].
- $\epsilon = 3.54 \cdot 10^{-11}$  F/m,
- $\sigma = 3$  mS/m,
- $\mu_0 = 4\pi \cdot 10^{-7}$  H/m
- $H_0^{(1)}$  the Hankel function of first kind and order zero

[Jupyter notebook to compute analytical solution](#)

# GERMAINE vs. analytical solution (50 MHz)

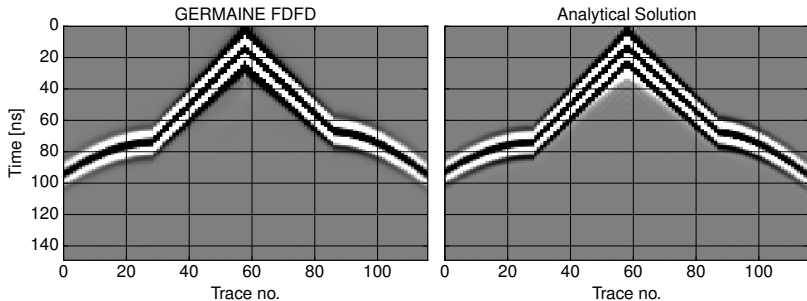


# GERMAINE vs. analytical solution (200 MHz)



# GERMAINE vs. analytical solution

## Comparison of time-domain radargrams



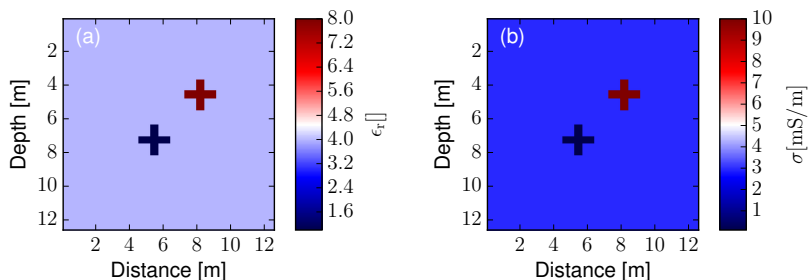
[Jupyter notebook to calculate TD analytical solution at receiver positions](#)

[Jupyter notebook to compare FDFD solution with analytical solution in TD](#)

# Modelling example: 2-cross model

# Modelling example: 2-cross model

2-cross model [Meles et al., 2011, Lavoué et al., 2014]

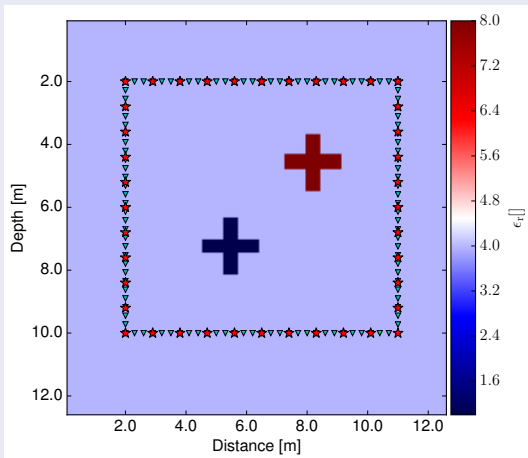


## Model discretization:

- $NX = 180$  gridpoints  $\times$   $NZ = 180$  gridpoints
- + 10 gridpoints PMLs on each side
- $DH = 7$  cm

# Modelling example: 2-cross model

## Perfect illumination acquisition geometry



- 40 shots and 117 receivers
- 7 frequencies [50, 60, 70, 80, 100, 150, 200] MHz

## Running GERMAINE for the 2-cross model ...

- 1 In GERMAINE/par/GERMAINE\_2\_cross\_TE.inp set the following parameters for a forward modelling run of the 2-cross model

```
forward_modelling_(yes=0)_FDFWI_(yes=1)_RTM_(yes=2)_(INVMAT) = 0  
MFILE = start/2_cross_TE_true  
output_of_P_field_(SNAP)(yes>0) = 1
```

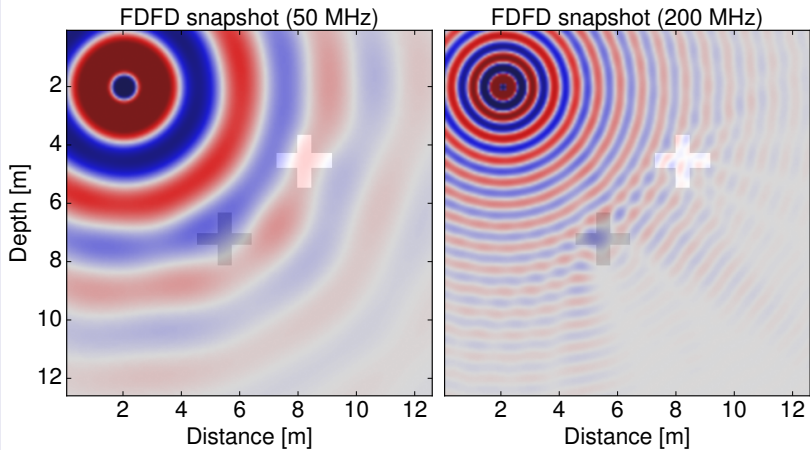
- 2 Run GERMAINE forward modelling on e.g. 4 cores of your CPU from the /par directory:

```
mpirun -np 4 ../bin/germaine GERMAINE_2_cross_TE.inp GERMAINE_workflow_2_cross_TE_1.inp
```



# Modelling example: 2-cross model

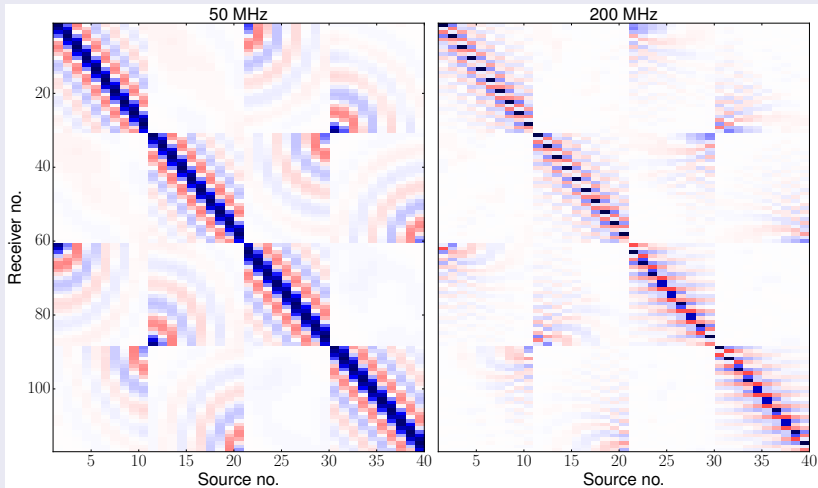
Frequency domain snapshots  $\Re(E_y)$



[Jupyter notebook for frequency domain wavefield visualization](#)

# Modelling example: 2-cross model

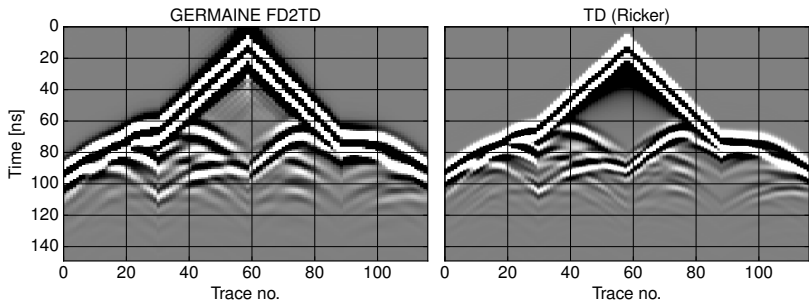
Frequency domain data  $\mathfrak{R}(E_y)$



[Jupyter notebook for frequency domain data visualization](#)

# Modelling example: 2-cross model

## GERAMINE FDFD vs. TD modelling [Irving and Knight, 2006]

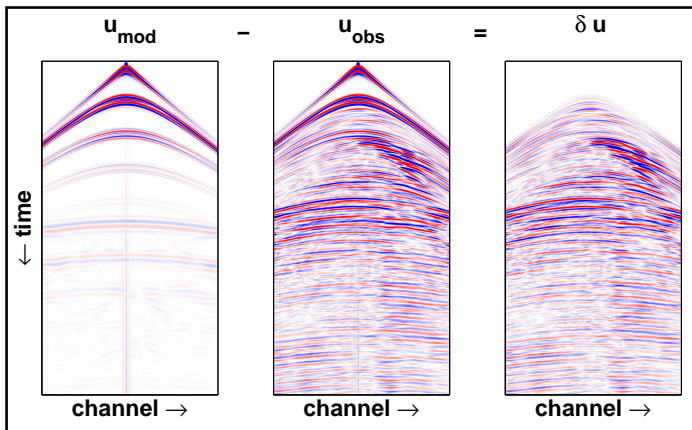


[Jupyter notebook for FD + TD data comparison](#)

# Full Waveform Inversion

# Full waveform inversion

- 1 Define objective function:

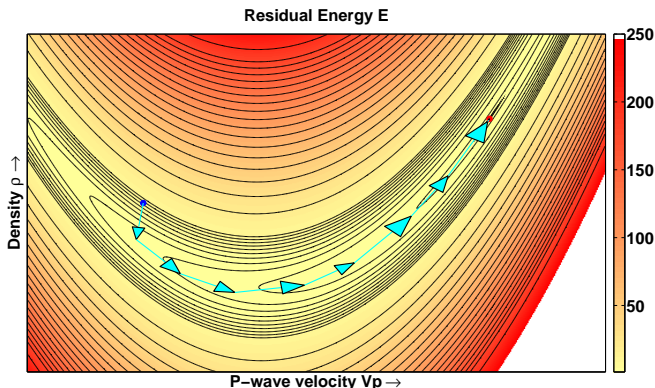


$$E = \frac{1}{2} \delta \mathbf{r}^\dagger \delta \mathbf{r} \text{ with } \delta \mathbf{r} = \mathbf{u}^{\text{mod}} - \mathbf{u}^{\text{obs}}$$

$\mathbf{u}^{\text{mod}}$ ,  $\mathbf{u}^{\text{obs}}$  modelled and field data (time/frequency domain)

# Full waveform inversion

- Minimize objective function by **Newton** optimization method



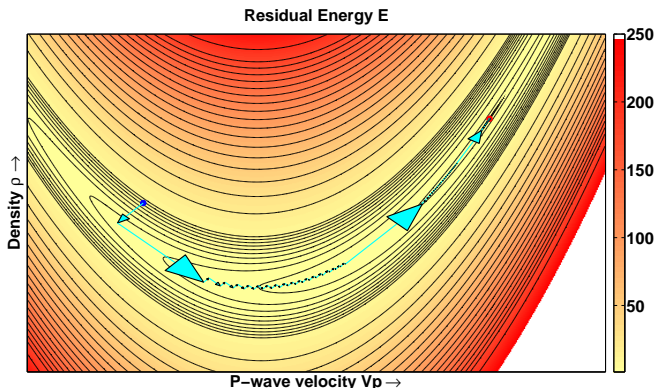
for the model parameters  $\mathbf{m} = [\sigma, \epsilon]$

$$\mathbf{m}^{n+1} = \mathbf{m}^n - \mu^n \left( \mathbf{H}^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)^n$$

with gradient  $\partial \mathbf{E} / \partial \mathbf{m}$ , Hessian  $\mathbf{H}$  and  $\mu$  step-length

# Full waveform inversion

- Minimize objective function by **Steepest Descent** method

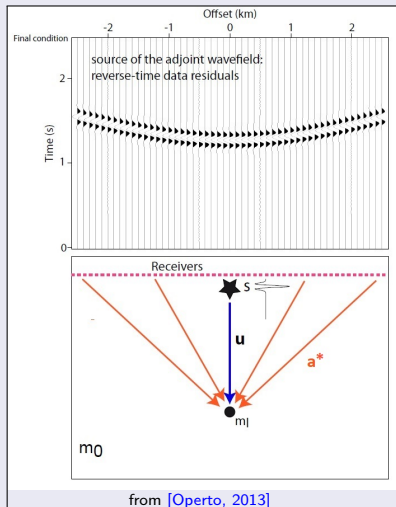


for the model parameters  $\mathbf{m} = [\sigma, \epsilon]$

$$\mathbf{m}^{n+1} = \mathbf{m}^n - \mu^n \left( \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)^n$$

with gradient  $\partial \mathbf{E} / \partial \mathbf{m}$  and  $\mu$  step-length

## Gradient computation



- Efficient gradient calculation by the adjoint-state method:

$$\frac{\partial E}{\partial m_l} = -\Re \left\{ \mathbf{u}^t \left( \frac{\partial \mathbf{A}}{\partial m_l} \right)^t \mathbf{a}^* \right\}$$

with

$\mathbf{u}$  the forward wavefield

$\mathbf{a}$  the adjoint wavefield

$\mathbf{A}$  the impedance matrix

$m_l$  the material parameter  
at grid point  $l$

[Brossier and Virieux, 2011]



# Basic FWI parameters in GERMAINE\_2\_cross\_TE.inp

- **Maximum number of iterations per FWI stage:**

```
#-----  
# GERMAINE inversion parameters  
# -----  
number_of_TDFWI_iterations_(ITERMAX) = 1000
```

- **Location of measured frequency-domain data:**

```
measured_FD_data_(DATA_DIR) = seis/2_cross_TE_true/2_cross_TE
```

- **Location of FWI results:**

```
# ----- Output of inverted models ----- #  
output_of_models_(INV_MODELFILE) = model/modelTest
```

# Basic FWI parameters in GERMAINE\_workflow\_2\_cross\_TE.inp

- You can define a simple FWI workflow similar to [Lavoué et al., 2014]

PRO	FC_low	FC_high	NF	...
0.0	50e6	200e6	1	...
0.0	60e6	200e6	1	...
0.0	70e6	200e6	1	...
0.0	80e6	200e6	1	...
0.0	100e6	200e6	1	...
0.0	150e6	200e6	1	...
0.0	200e6	200e6	1	...

which sequentially inverts the 50 MHz, 60 MHz, 70 MHz, 80 MHz, 100 MHz, 150 MHz and 200 MHz data, respectively.

# GERMAINE FWI example: 2-cross model

# GERMAINE FWI example: 2-cross model

- 1 First, we have to create some synthetic field data for the 2-cross model. In GERMAINE/par/GERMAINE\_2\_cross\_TE.inp set the following parameters for a forward modelling run of the 2-cross model

```
forward_modelling_(yes=0)_FDFWI_(yes=1)_RTM_(yes=2)_(INVMAT) = 0
MFILE = start/2_cross_TE_true
output_of_P_field_(SNAP)(yes>0) = 0
```

- 2 Run GERMAINE forward modelling on e.g. 4 cores of your CPU from the /par directory:

```
mpirun -np 4 ../bin/germaine GERMAINE_2_cross_TE.inp GERMAINE_workflow_2_cross_TE.inp
```

- 3 In GERMAINE/par/seis generate the directory 2\_cross\_TE\_true

```
mkdir 2_cross_TE_true
```

- 4 Move the FDFD data of the true model from /seis to /seis/2\_cross\_TE\_true

```
mv 2_cross_TE_p_stage_* 2_cross_TE_true/
```

# GERMAINE FWI example: 2-cross model

- 5 To run the TE FDFD FWI change the following parameters in GERMAINE/par/GERMAINE\_2\_cross\_TE.inp

```
forward_modelling_(yes=0)_FDFWI_(yes=1)_RTM_(yes=2)_(INVMAT) = 1  
MFILE = start/2_cross_TE_init
```

- 6 Start the 2D TE FDFD FWI with:

```
mpirun -np 4 ../bin/germaine GERMAINE_2_cross_TE.inp GERMAINE_workflow_2_cross_TE.inp
```

As defined in GERMAINE\_workflow\_2\_cross\_TE.inp the FWI is based on a sequential inversion workflow using the 50 MHz, 60 MHz, 70 MHz, 80 MHz, 100 MHz, 150 MHz and 200 MHz FD-data, respectively. The permittivity and conductivity models of the current iteration are saved in:

[model/modelTest\\_eps.bin](#)

[model/modelTest\\_sig.bin](#)

# GERMAINE FWI example: 2-cross model

- 7 Depending on the clock speed of your CPU and number of cores wait approximately a few hours ...

The intermediate results after the FWI of each stage are saved separately in

`model/modelTest_eps_stage_1.bin`

`model/modelTest_sig_stage_1.bin`

...

`model/modelTest_eps_stage_7.bin`

`model/modelTest_sig_stage_7.bin`

# GERMAINE FWI example: 2-cross model

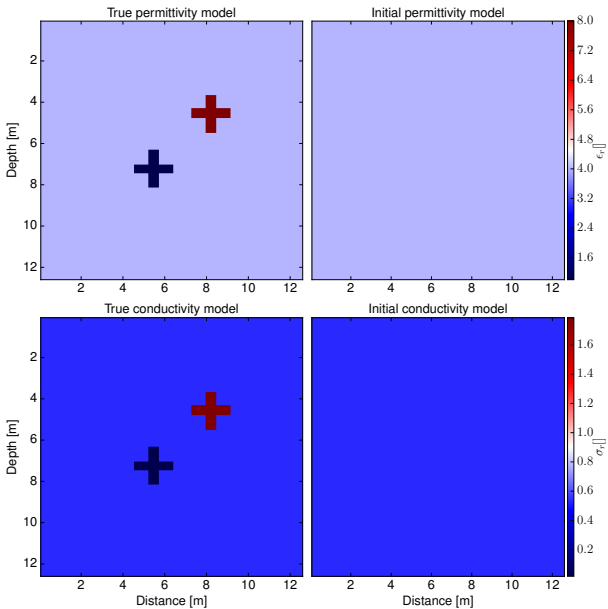
- 8 You can visualize the results with Seismic Unix

```
ximage n1=180 < model/modelTest_eps_stage_7.bin
```

or use this Jupyter notebook

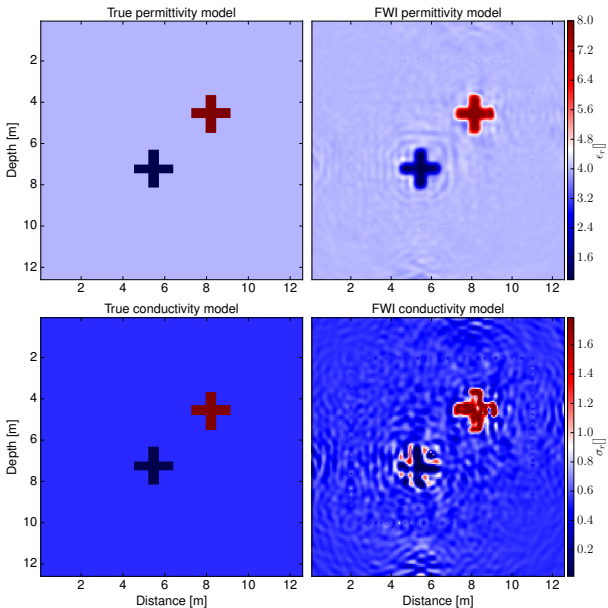
[Jupyter notebook for FWI result visualization](#)

# FWI: initial model





# FWI: result with optimized inversion parameters



# GERMAINE FWI example: 2-cross model

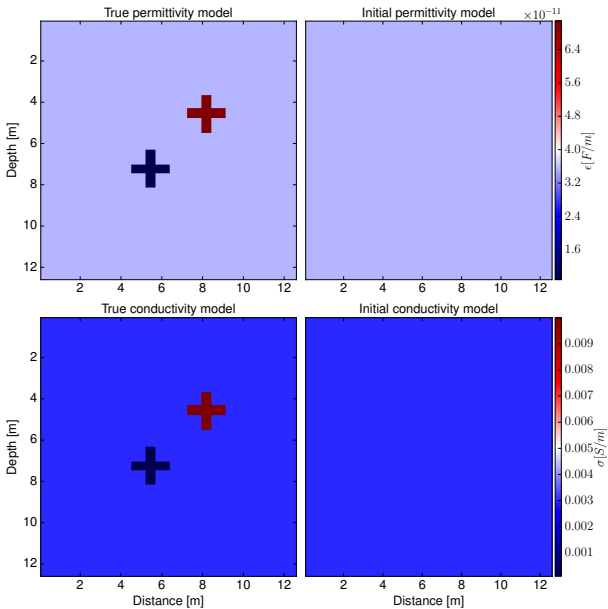
The FWI result might suggest, that multi-parameter FWI is a very easy task. However, its success depends on a lot of parameter tuning.

To discuss the influence of the different parameters on the FWI result, we start with a very simple, naive, non-optimized inversion based on steepest descent optimization:

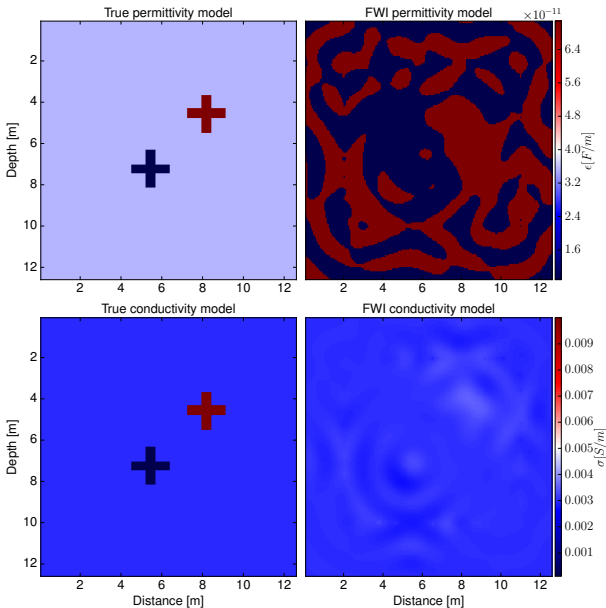
Run the naive 2D TE FDFD FWI with:

```
mpirun -np 4 ../bin/germaine GERMAINE_2_cross_TE_naive.inp GERMAINE_workflow_2_cross_TE_naive.inp
```

# FWI: initial model



# FWI: inversion result with steepest descent method



# FWI of normalized material parameters

- Permittivity and conductivity models vary on different scales  
→ FWI fails to converge

- **Idea:**

Invert normalized material parameters for FWI

$$\epsilon_r = \epsilon / \epsilon_0,$$

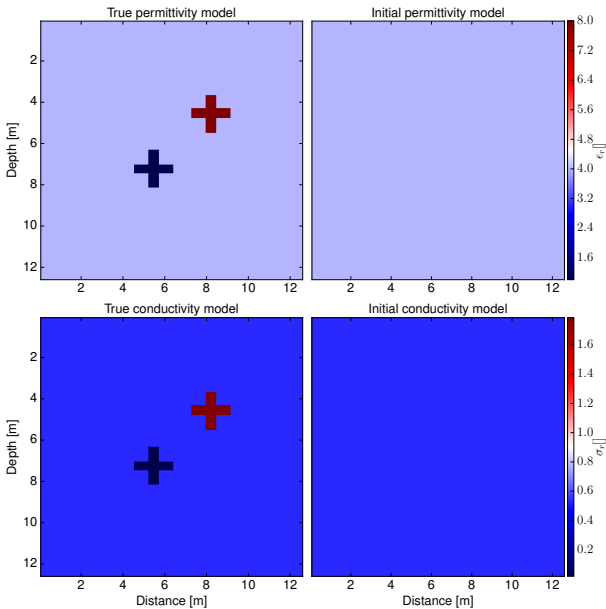
$$\sigma_r = \sigma / \sigma_0$$

with  $\epsilon_0 = 8.85 \cdot 10^{-12}$  F/m and  $\sigma_0 = 5.6 \cdot 10^{-3}$  S/m

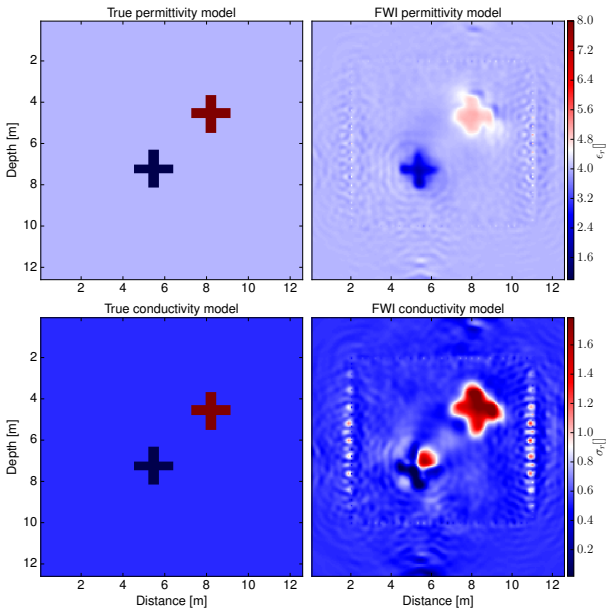
- Definition in GERMAINE parameter file:

```
# -- Normalization factor for material parameters --
norm_sigma_(MAT1_NORM) = 5.6e-3
norm_epsilon_(MAT2_NORM) = 8.85e-12
```

# FWI: initial normalized parameters

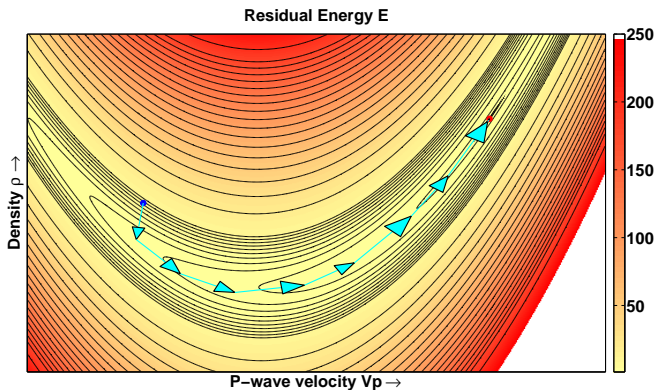


# FWI results using normalized parameters



# Preconditioning gradient

- 2 Minimize objective function by **Newton** optimization method



for the model parameters  $\mathbf{m} = [\sigma, \epsilon]$

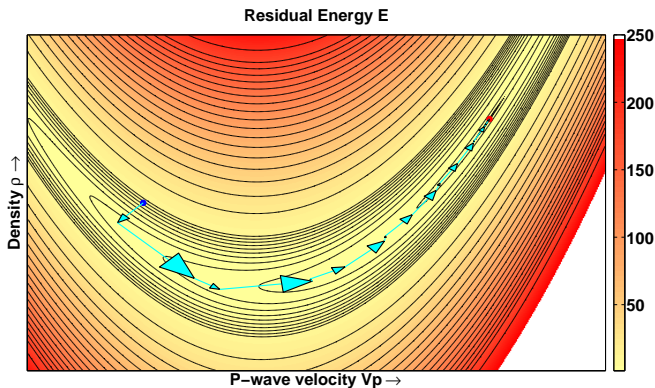
$$\mathbf{m}^{n+1} = \mathbf{m}^n - \mu^n \left( \mathbf{H}^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)^n$$

with gradient  $\partial \mathbf{E} / \partial \mathbf{m}$ , Hessian  $\mathbf{H}$  and  $\mu$  step-length



# Preconditioning gradient

- 2 Minimize objective function by **Preconditioned Gradient**

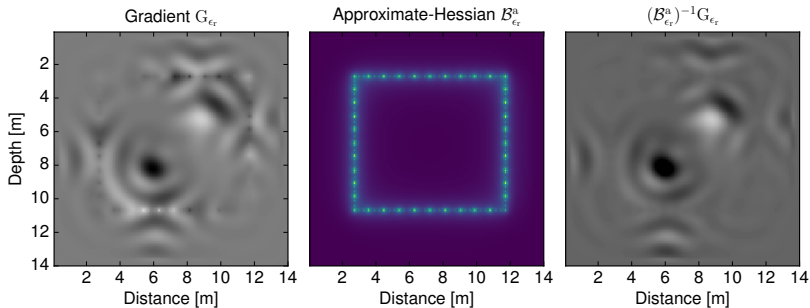


for the model parameters  $\mathbf{m} = [\sigma, \epsilon]$

$$\mathbf{m}^{n+1} = \mathbf{m}^n - \mu^n \left( \mathcal{B}^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)^n$$

with gradient  $\partial \mathbf{E} / \partial \mathbf{m}$ , Hessian approximation  $\mathcal{B}$

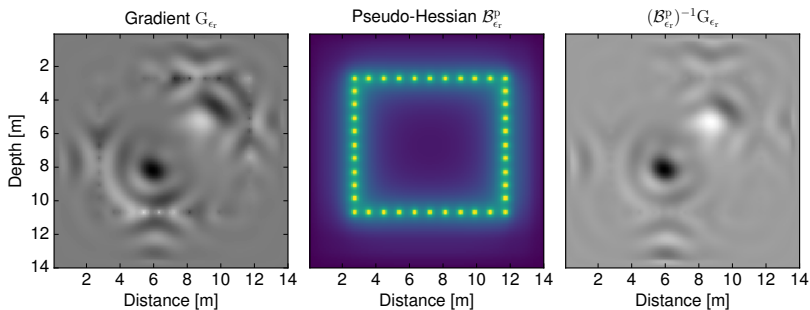
# Preconditioning gradient with Approximate-Hessian



$$\mathcal{B}_{kk}^a = \sum_s \sum_r \left( \mathbf{u}_s^\dagger \left( \frac{\partial A}{\partial \mathbf{m}_k} \right)^\dagger \mathbf{A}^{-1} \mathbf{R}_r^t \right) \left( \mathbf{R}_r \mathbf{A}^{-1} \left( \frac{\partial A}{\partial \mathbf{m}_k} \right) \mathbf{u}_s \right)$$

$$\mathbf{m}^{n+1} = \mathbf{m}^n - \mu^n \left( (\mathcal{B}^a + \{\epsilon_{\text{hess}} \max(|\mathcal{B}^a|)\} \mathbf{I})^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)^n$$

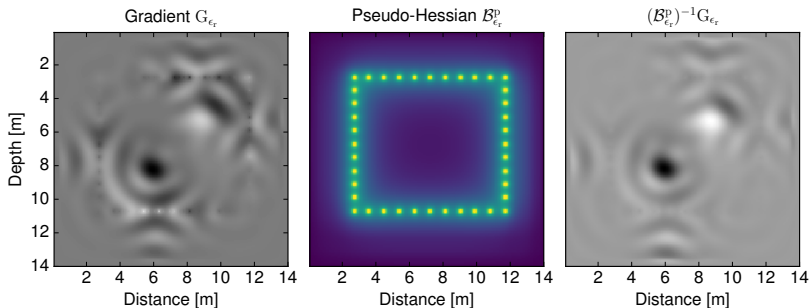
# Preconditioning gradient with Pseudo-Hessian



$$\mathcal{B}_{kk}^P = \sum_s \left( \left( \frac{\partial A}{\partial m_k} \right)_{u_s} \right)^\dagger \left( \left( \frac{\partial A}{\partial m_k} \right)_{u_s} \right)$$

$$\mathbf{m}^{n+1} = \mathbf{m}^n - \mu^n \left( (\mathcal{B}^P + \{\epsilon_{\text{hess}} \max(|\mathcal{B}^P|)\} \mathbf{I})^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)^n$$

# Preconditioning gradient with Pseudo-Hessian

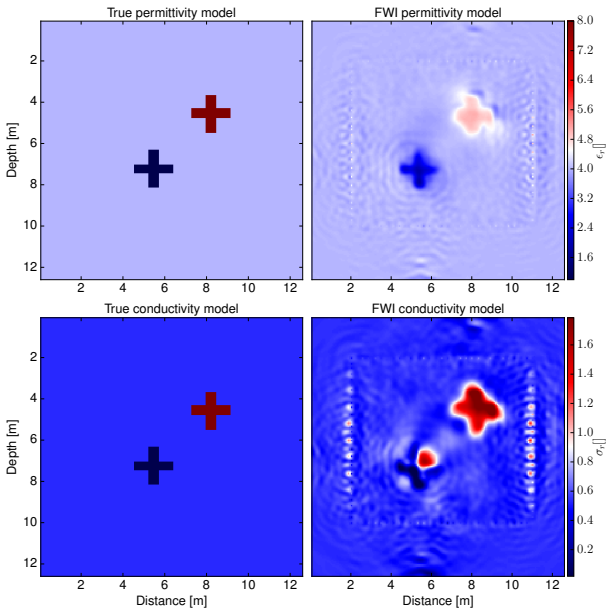


- Definition in GERMAINE parameter file:

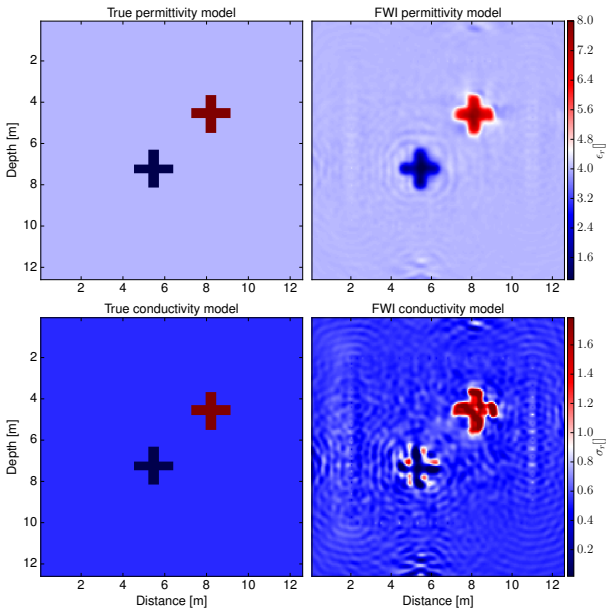
```
HESSIAN_(no=0/app_Hessian=1/pseudo_Hessian=2) = 2
```

```
Regularization_term_for_inverse_Hessian_(EPS_HESS) = 0.01
```

# FWI results using gradient method



# FWI results using gradient with Pseudo-Hessian



# Quasi-Newton I-BFGS optimization

So far we used the very inefficient steepest descent optimization ...

$$\mathbf{m}_{n+1} = \mathbf{m}_n + \mu_n \Delta \mathbf{m}_n$$

with the search direction:

$$\Delta \mathbf{m}_n = - \left( \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)_n$$

or preconditioned search direction:

$$\Delta \mathbf{m}_n = - \left( \mathcal{B}^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)_n$$

Improve Hessian approximation by I-BFGS optimization

## Quasi-Newton I-BFGS algorithm

[Nocedal and Wright, 2006, Métivier and Brossier, 2016]

At iteration step n:

- 1 Compute  $\mathbf{g}_n = \left( \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)_n$
- 2 Compute and store  $\mathbf{s}_n = \mathbf{m}_{n+1} - \mathbf{m}_n$   
Compute and store  $\mathbf{y}_n = \mathbf{g}_{n+1} - \mathbf{g}_n$
- 3  $\mathbf{q} = -\mathbf{g}_n$
- 4 for  $i = n-1$  to  $n-1$  do  
 $\rho_i = \frac{1}{\mathbf{y}_i^T \mathbf{s}_i}$   
 $\alpha_i = \rho_i \mathbf{s}_i^T \mathbf{q}$   
 $\mathbf{q} = \mathbf{q} - \alpha_i \mathbf{y}_i$   
end for
- 5 Compute  $H_n^0 = \frac{\mathbf{s}_{n-1}^T \mathbf{y}_{n-1}}{\mathbf{y}_{n-1}^T \mathbf{y}_{n-1}}$
- 6 Compute  $\mathbf{z} = H_n^0 \mathbf{q}$
- 7 for  $i = n-1$  to  $n-1$  do  
 $\beta_i = \rho_i \mathbf{y}_i^T \mathbf{z}$   
 $\mathbf{z} = \mathbf{z} + \mathbf{s}_i (\alpha_i - \beta_i)$   
end for
- 8  $\Delta \mathbf{m}_n = \mathbf{z}$
- 9 Update model  $\mathbf{m}_{n+1} = \mathbf{m}_n + \mu_n \Delta \mathbf{m}_n$

# Quasi-Newton PI-BFGS optimization

So far we used the very inefficient steepest descent optimization ...

$$\mathbf{m}_{n+1} = \mathbf{m}_n + \mu_n \Delta \mathbf{m}_n$$

with the search direction:

$$\Delta \mathbf{m}_n = - \left( \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)_n$$

or preconditioned search direction:

$$\Delta \mathbf{m}_n = - \left( \mathcal{B}^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)_n$$

Improve Hessian approximation by **PI-BFGS** optimization

## Preconditioned Quasi-Newton **PI-BFGS** algorithm

[Nocedal and Wright, 2006, Métivier and Brossier, 2016]

At iteration step n:

- 1 Compute  $\mathbf{g}_n = \left( \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)_n$
- 2 Compute and store  $\mathbf{s}_n = \mathbf{m}_{n+1} - \mathbf{m}_n$   
Compute and store  $\mathbf{y}_n = \mathbf{g}_{n+1} - \mathbf{g}_n$
- 3  $\mathbf{q} = -\mathbf{g}_n$
- 4 for  $i = n-1$  to  $n-1$  do  
 $\rho_i = \frac{1}{\mathbf{y}_i^T \mathbf{s}_i}$   
 $\alpha_i = \rho_i \mathbf{s}_i^T \mathbf{q}$   
 $\mathbf{q} = \mathbf{q} - \alpha_i \mathbf{y}_i$   
end for
- 5 Compute  $\mathbf{H}_n^0 = (\mathcal{B}_n)^{-1}$
- 6 Compute  $\mathbf{z} = \mathbf{H}_n^0 \mathbf{q}$
- 7 for  $i = n-1$  to  $n-1$  do  
 $\beta_i = \rho_i \mathbf{y}_i^T \mathbf{z}$   
 $\mathbf{z} = \mathbf{z} + \mathbf{s}_i (\alpha_i - \beta_i)$   
end for
- 8  $\Delta \mathbf{m}_n = \mathbf{z}$
- 9 Update model  $\mathbf{m}_{n+1} = \mathbf{m}_n + \mu_n \Delta \mathbf{m}_n$



# Quasi-Newton I-BFGS optimization

- Definition in GERMAINE parameter file

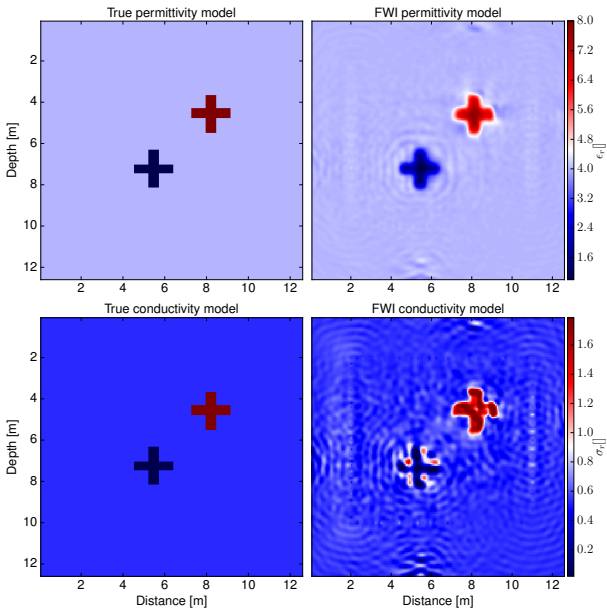
```
# ----- Optimization-Method ----- #  
gradient_method_(PCG=1/LBFGS=2/Descent=3)_(GRAD_METHOD) = 2  
PCG_BETA_(Fletcher_Reeves=1/Polak_Ribiere=2/Hestenes_Stiefel=3/Dai_Yuan=4) = 4  
save_(NLBFGS)_updates_during_LBFGS_optimization = 5
```

- NLBFGS defines how many  $s_n$  and  $y_n$  values are stored during I-BFGS optimization
- Preconditioned quasi-Newton PI-BFGS optimization is used if the Approximate-Hessian (HESSIAN=1) or Pseudo-Hessian (HESSIAN=2) is activated

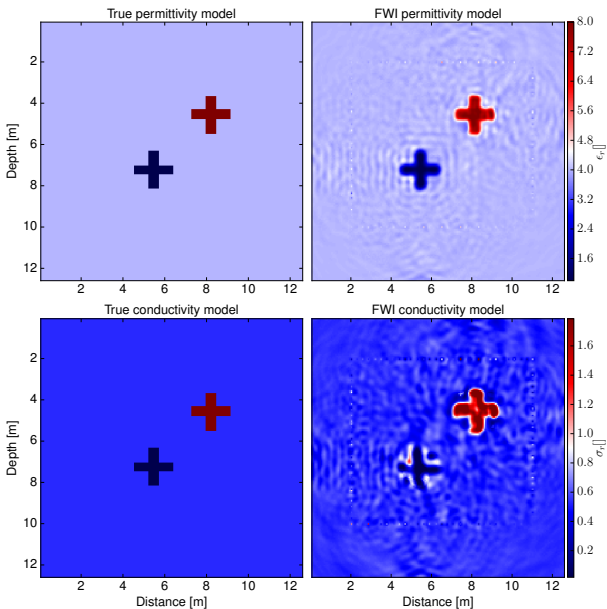
```
HESSIAN_(no=0/app_Hessian=1/pseudo_Hessian=2) = 2  
Regularization_term_for_inverse_Hessian_(EPS_HESS) = 0.01
```

otherwise I-BFGS optimization is applied.

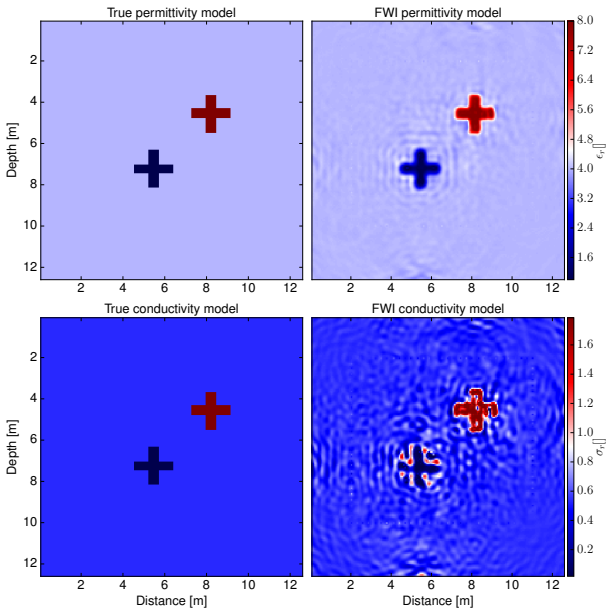
# FWI results using gradient with Pseudo-Hessian



# FWI results using I-BFGS



# FWI results using PI-BFGS with Pseudo-Hessian



# FWI parameter scaling

- To decrease the conductivity update a parameter scaling factor  $\beta_\sigma$  can be introduced, which scales the normalized conductivity [Lavoué et al., 2014]:

$$\tilde{\sigma}_r = \sigma_r / \beta_\sigma$$

- Definition in GERMAINE workflow file:

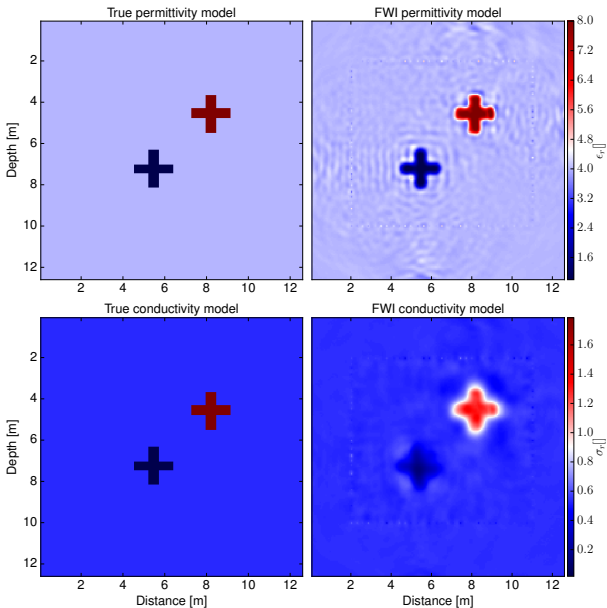
```
... MISFIT BETA_SIGMA BETA_EPSILON ...
... 1      0.25      1.0      ...
... 1      0.25      1.0      ...
... 1      0.25      1.0      ...
```

**11<sup>th</sup> column** =  $\beta_\sigma$  []

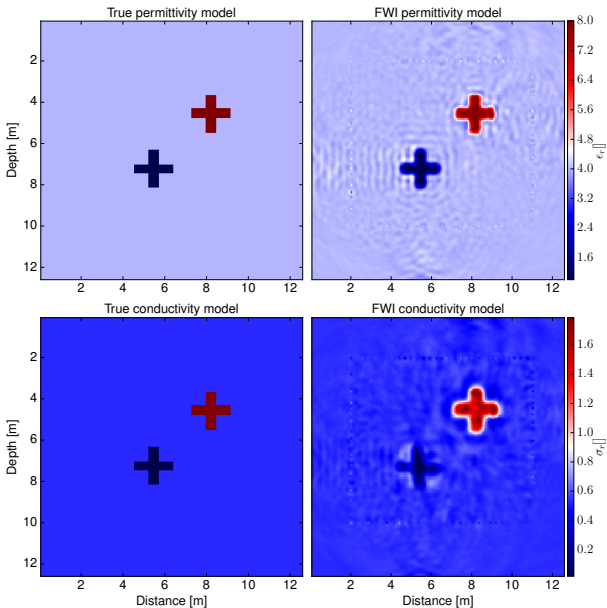
**12<sup>th</sup> column** =  $\beta_\epsilon$  []

- Seem to have nearly no effect on the FWI result when using PI-BFGS instead of I-BFGS optimization

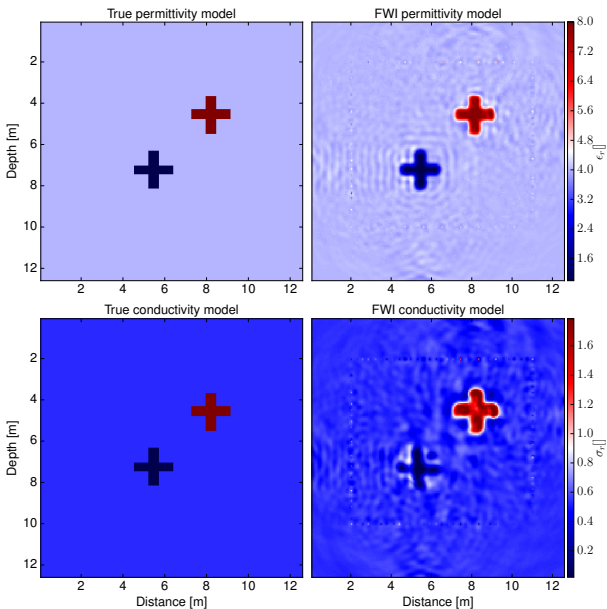
# I-BFGS FWI results with parameter scaling $\beta_\sigma = 0.25$



# I-BFGS FWI results with parameter scaling $\beta_\sigma = 0.5$

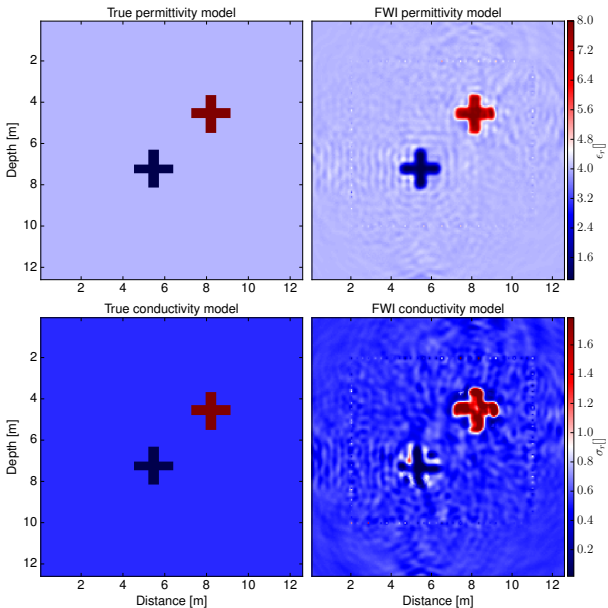


# I-BFGS FWI results with parameter scaling $\beta_\sigma = 0.75$

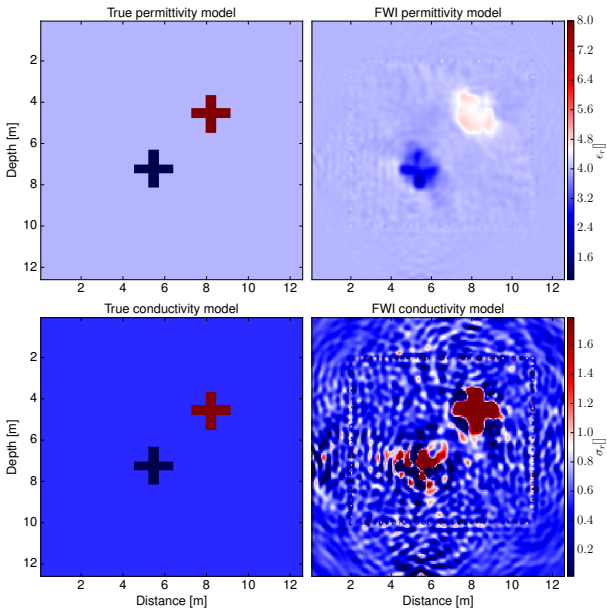




# I-BFGS FWI results with parameter scaling $\beta_\sigma = 1.0$



# I-BFGS FWI results with parameter scaling $\beta_\sigma = 4.0$



# Bound constraints

- Restrict the updated parameters

$$\mathbf{m}^{n+1} = \mathbf{m}^n - \mu^n \left( \mathcal{B}^{-1} \frac{\partial \mathbf{E}}{\partial \mathbf{m}} \right)^n$$

to a bounded part of the parameter space [Métivier and Brossier, 2016]:

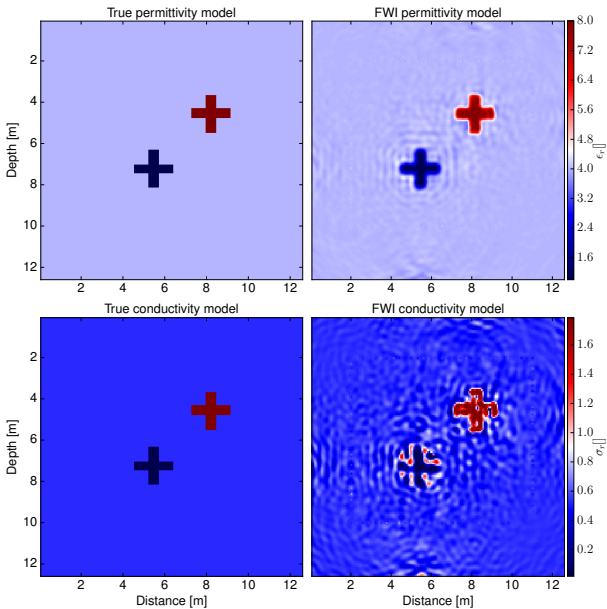
$$m_i^{n+1} = \begin{cases} m_i^{n+1} & \text{if } \text{MAT\_LOW} \leq m_i^{n+1} \leq \text{MAT\_UP}, \\ \text{MAT\_LOW} & \text{if } m_i^{n+1} < \text{MAT\_LOW}, \\ \text{MAT\_UP} & \text{if } m_i^{n+1} > \text{MAT\_UP}, \end{cases}$$

where MAT\_LOW, MAT\_UP denote the lower and upper bounds for the non-normalized material parameters, respectively.

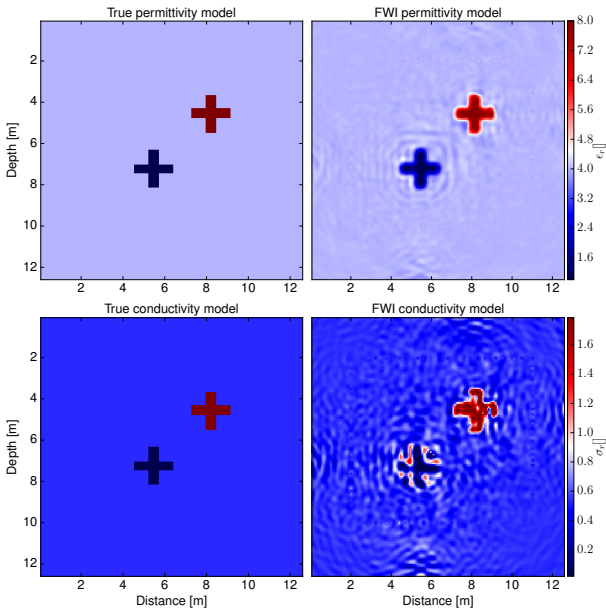
- Definition in GERMAINE parameter file:

```
# ----- Upper and lower bound for model parameters ----- #  
limits_for_sigma_(MAT1_LOW,MAT1_UP) = 0.0001, 0.01  
limits_for_epsilon_(MAT2_LOW,MAT2_UP) = 8.86e-12, 7.0e-11
```

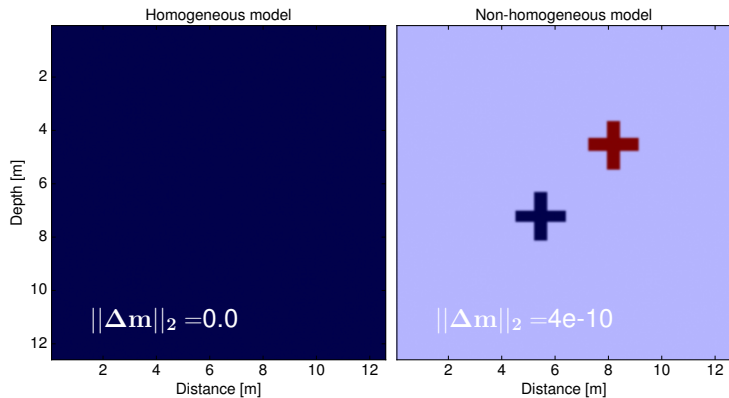
# FWI using PI-BFGS **without** bound constraints



# FWI using PI-BFGS with bound constraints



# Tikhonov regularization



Define a measure for the model roughness, e.g.:

$$\|\Delta \mathbf{m}\|_2,$$

where  $\Delta$  denotes the Laplace operator.

# Tikhonov regularization

- Minimize model roughness in FWI result by adding the Tikhonov regularization term to the objective function:

$$E = \frac{1}{2} \delta \mathbf{r}^\dagger \delta \mathbf{r} + \lambda_\epsilon \|\Delta \epsilon_r\|_2 + \lambda_\sigma \|\Delta \sigma_r\|_2,$$

where the hyperparameter  $\lambda$  balances the contribution of the regularization term with respect to the data misfit term.

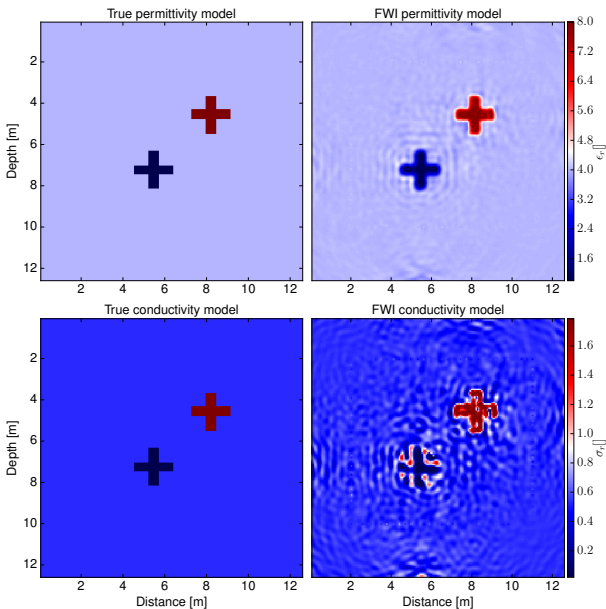
- Flexible definition in GERMAINE workflow file for permittivity and conductivity model:

```
... LAMBDA_SIGMA      LAMBDA_EPSILON
... 1e-4              0.0
... 0.001            1e-4
```

**13<sup>th</sup> column** = hyperparameter  $\lambda_\sigma$

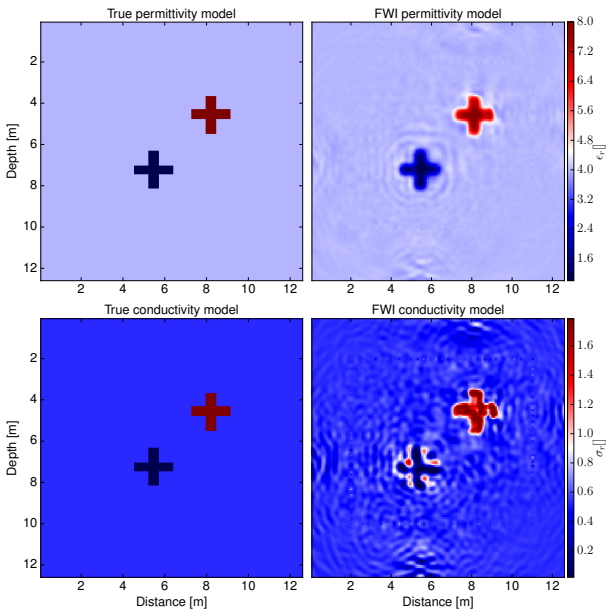
**14<sup>th</sup> column** = hyperparameter  $\lambda_\epsilon$

# Tikhonov regularization: PI-BFGS ( $\lambda_\epsilon = \lambda_\sigma = 0.0$ )

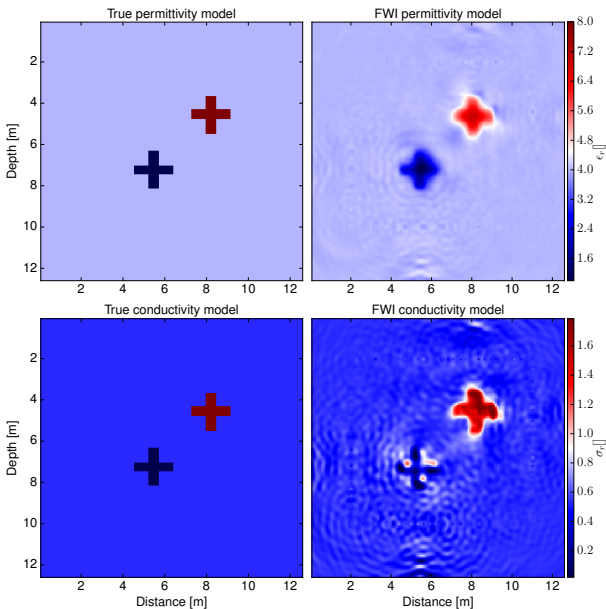




# Tikhonov regularization: PI-BFGS ( $\lambda_\epsilon = \lambda_\sigma = 0.0001$ )



# Tikhonov regularization: PI-BFGS ( $\lambda_\epsilon = \lambda_\sigma = 0.001$ )



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*Geophysics*, 74(5):T75–T95.

# Appendix

# GERMAINE installation on the NEC HPC-Linux cluster

... on the NEC HPC-Linux cluster at Kiel university



- 172 nodes with 2 Intel Xeon Gold 6130 (Skylake-Sp) CPUs (32 cores, clock speed 2.1 GHz), 192 GB DDR4 RAM
- + additional FAT nodes

## More info:

<https://www.rz.uni-kiel.de/de/angebote/hiperf/nec-linux-cluster>

# GERMAINE installation on the NEC HPC-Linux cluster

- 1 Clone GERMAINE source code from:

<https://github.com/daniel-koehn/GERMAINE>

to the \$WORK directory by typing:

```
git clone https://github.com/daniel-koehn/GERMAINE.git
```

- 2 Modify the Makefile in GERMAINE/src by out commenting:

```
# On Linux NEC-cluster with Intel-MPI
```

```
CC=mpiicc
```

```
LFLAGS=-lm -ftz -lstdc++ -lumfpack
```

```
CFLAGS=-O3 -xAVX -fno-fnalias -restrict
```

```
SFLAGS=-L$/sfs/fs2/work-sh2/sungw331/SuiteSparse/lib
```

```
IFLAGS=-I$/sfs/fs2/work-sh2/sungw331/SuiteSparse/include -I../include
```

**Do not forget to comment the unnecessary compiler options**



# GERMAINE installation on the NEC HPC-Linux cluster

- 3 Add the following lines to `.bashrc` file in the `$HOME`-directory:

```
# SuiteSparse directory
export SuiteSparse_DIR='/sfs/fs2/work-sh2/sungw331/SuiteSparse'
export LD_LIBRARY_PATH=$LD_LIBRARY_PATH:$SuiteSparse_DIR/lib
```

- 4 Read `.bashrc` in the running shell:

```
. .bashrc
```

- 5 In `GERMAINE/src` compile the source code with

```
make GERMAINE
```

[Click here to continue ...](#)

# GERMAINE installation on a desktop computer at Department of Geophysics

- 1 Clone GERMAINE source code from:

<https://github.com/daniel-koehn/GERMAINE>

to the HOME directory by typing:

```
git clone https://github.com/daniel-koehn/GERMAINE.git
```

- 2 Modify the Makefile in GERMAINE/src by out commenting:

```
# On Desktop computer @ Geophysics (CAU Kiel) with Debian, OpenMPI and gcc 4.8.2
CC=mpicc
LFLAGS=-lm -lstdc++ -lumfpack
CFLAGS=-O3 -w -fno-stack-protector -D_FORTIFY_SOURCE=0
SFLAGS=-L$/home/dkoehn/libs/SuiteSparse/lib
IFLAGS=-I$/home/dkoehn/libs/SuiteSparse/include -I../include
```

**Do not forget to comment the unnecessary compiler options**

- 3 Add the following lines to `.bashrc` file in your HOME-directory:

```
export OpenBLAS_DIR='/home/dkoehn/libs/OpenBLAS'  
export LD_LIBRARY_PATH=$LD_LIBRARY_PATH:$OpenBLAS_DIR/lib  
  
export LD_LIBRARY_PATH=/home/dkoehn/libs/SuiteSparse/lib:$LD_LIBRARY_PATH
```

- 4 Read `.bashrc` in the running shell:

```
. .bashrc
```

- 5 In `GERMAINE/src` compile the source code with

```
make GERMAINE
```

[Click here to continue ...](#)

# SuiteSparse installation on NEC HPC-Linux cluster

# SuiteSparse installation on the NEC HPC-Linux cluster

- 1 Download SuiteSparse v4.5.5 source code from:  
<http://faculty.cse.tamu.edu/davis/SuiteSparse/SuiteSparse-4.5.5.tar.gz>
- 2 Copy TAR-archive to \$WORK directory on NEC-cluster
- 3 Extract source code:

```
tar -zxvf SuiteSparse-4.5.5.tar.gz
```

- 4 Load Cmake, Intel compiler and MPI modules, which include the MKL library:

```
module load cmake3.9.1 intel17.0.4 intelmpi17.0.4
```

- 5 In the SuiteSparse directory type:

```
make config  
make
```

- 6 Add the following lines to `.bashrc` in `$HOME` directory:

```
# SuiteSparse directory  
export SuiteSparse_DIR='/sfs/fs2/work-sh2/sungw331/SuiteSparse'  
export LD_LIBRARY_PATH=$LD_LIBRARY_PATH:$SuiteSparse_DIR/lib
```

where `SuiteSparse_DIR` has to be adapted to your path.