

# A practical comparison of different parameterizations for anisotropic elastic full-waveform inversion

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# Outline

Introduction

Theory

Models and survey setup

Results

Conclusions

Acknowledgments

# Introduction

- Choice of parameterization can affect results in FWI.
- Apply FWI using different parameterizations to the same problem and examine differences.
- Parameterizations:
  1.  $V_{P0}, V_{S0}, \varepsilon, \delta$
  2.  $V_{P0}, V_{S0}, V_{\text{hor}}, V_{\text{nmo}}$
  3.  $V_{\text{nmo}}, V_{S0}, \eta, \delta$
- Inverted for  $V_{P0}$  in a complex model.
- Multiparameter inversion in a simple model.

# Theory

- In FWI we want to find a parameter model  $\mathbf{m}$  that can produce modeled data  $\mathbf{u}$  which is close to some measured data  $\mathbf{d}$ .
- Apply a numerical wave operator that maps  $\mathbf{m}$  from the model domain into the data domain:

$$\mathcal{L}(\mathbf{m}) = \mathbf{u}. \quad (1)$$

- Ideally, find an inverse operator to map  $\mathbf{d}$  from the data domain to the model domain:

$$\mathbf{m} = \mathcal{L}^{-1}(\mathbf{d}). \quad (2)$$

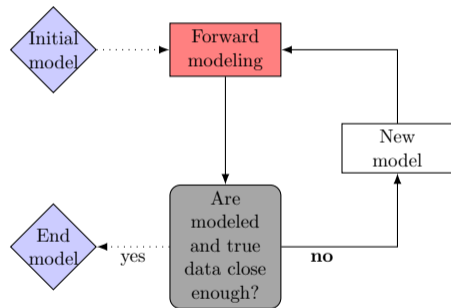
# Theory

- Define a misfit functional:

$$\mathcal{F}(\mathbf{m}) = \frac{1}{2} \sum_{j=0}^{n_s} \sum_{i=0}^{n_r} \|\hat{\mathbf{u}}_{i,j}(\mathbf{m}) - \hat{\mathbf{d}}_{i,j}\|_2^2. \quad (3)$$

- The solution is an extreme point of  $\mathcal{F}(\mathbf{m})$ :

$$\mathbf{m}' = \arg \min_{\mathbf{m}} \mathcal{F}(\mathbf{m}). \quad (4)$$



## Theory

- Update the model iteratively:

$$\mathbf{m}_{k+1} = \mathbf{m}_k - \alpha_k \mathbf{H}_k^{-1} \delta \mathbf{m}_k. \quad (5)$$

- Hessian matrix contains second derivatives of the misfit functional
  - Approximated from previous gradients (L-BFGS)
- Gradients are found via the adjoint method, Mora (1987).

$$\delta \hat{\mathbf{m}}(\mathbf{x}) = \sum_{n_s} \int dt \sum_{n_r} \frac{\partial u_i(\mathbf{x}_S, \mathbf{x}_R, t)}{\partial \mathbf{m}(\mathbf{x})} \delta u_i(\mathbf{x}_S, \mathbf{x}_R, t). \quad (6)$$

$$\delta u_i(\mathbf{x}_S, \mathbf{x}_R, t) = \int_V dV \frac{\partial u_i(\mathbf{x}_S, \mathbf{x}_R, t)}{\partial \mathbf{m}(\mathbf{x})} \delta \mathbf{m}(\mathbf{x}). \quad (7)$$

## Gradients, 3D

$$\delta\rho = -\sum_{n_s} \int dt \dot{u}_j \dot{\Psi}_j,$$

$$\delta c_{11} = -\sum_{n_s} \int dt (u_{1,1} + u_{2,2})(\Psi_{1,1} + \Psi_{2,2}),$$

$$\delta c_{33} = -\sum_{n_s} \int dt u_{3,3} \Psi_{3,3},$$

$$\delta c_{13} = -\sum_{n_s} \int dt \left[ \Psi_{3,3}(u_{1,1} + u_{2,2}) + (\Psi_{1,1} + \Psi_{2,2})u_{3,3} \right],$$

$$\delta c_{44} = -\sum_{n_s} \int dt \left[ (\Psi_{3,1} + \Psi_{1,3})(u_{3,1} + u_{1,3}) \right. \\ \left. + (\Psi_{3,2} + \Psi_{2,3})(u_{3,2} + u_{2,3}) \right],$$

$$\delta c_{66} = -\sum_{n_s} \int dt \left[ (\Psi_{2,1} + \Psi_{1,2})(u_{2,1} + u_{1,2}) \right. \\ \left. - 2(\Psi_{2,2}u_{1,1} + \Psi_{1,1}u_{2,2}) \right].$$

## Gradients for parameterizations, 2D

1)

$$\delta V_{P0} = k_{11}\delta c_{11} + k_{12}\delta c_{13} + k_{13}\delta c_{33},$$

$$\delta V_{S0} = k_{14}\delta c_{13} + k_{15}\delta c_{44},$$

$$\delta \varepsilon = k_{16}\delta c_{11},$$

$$\delta \delta = k_{17}\delta c_{13},$$

2)

$$\delta V_{P0} = k_{21}\delta c_{13} + k_{22}\delta c_{33},$$

$$\delta V_{S0} = k_{23}\delta c_{13} + k_{24}\delta c_{44},$$

$$\delta V_{\text{hor}} = k_{25}\delta c_{11},$$

$$\delta V_{\text{nmo}} = k_{26}\delta c_{13},$$

3)

$$\delta V_{\text{nmo}} = k_{31}\delta c_{11} + k_{32}\delta c_{13} + k_{33}\delta c_{33},$$

$$\delta V_{S0} = k_{34}\delta c_{13} + k_{35}\delta c_{44},$$

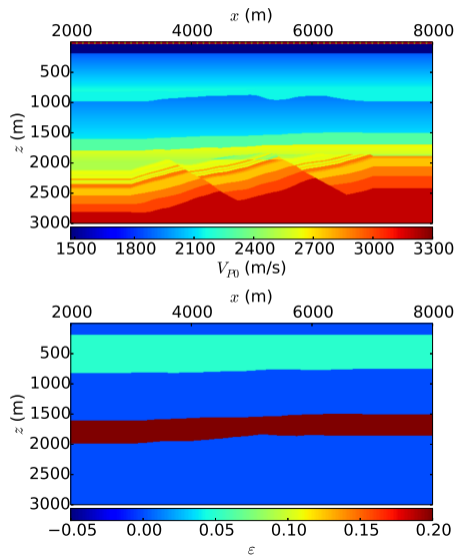
$$\delta \eta = k_{36}\delta c_{11},$$

$$\delta \delta = k_{37}\delta c_{13} + k_{38}\delta c_{44},$$



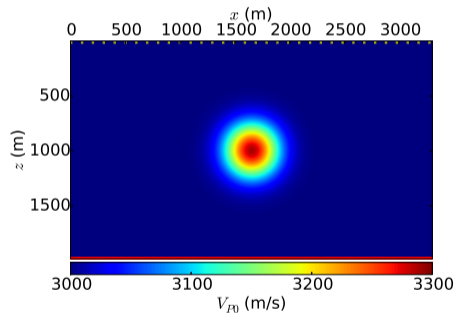
## Complex model

- Synthetic model representative of the Gullfaks field
- 10 km long, 3 km deep
- $1001 \times 300$  grid points
- Total of 101 shots and 1001 receivers
- Source: 5 Hz Ricker wavelet
- Receivers: Pressure
- Gradient muted in the water layer

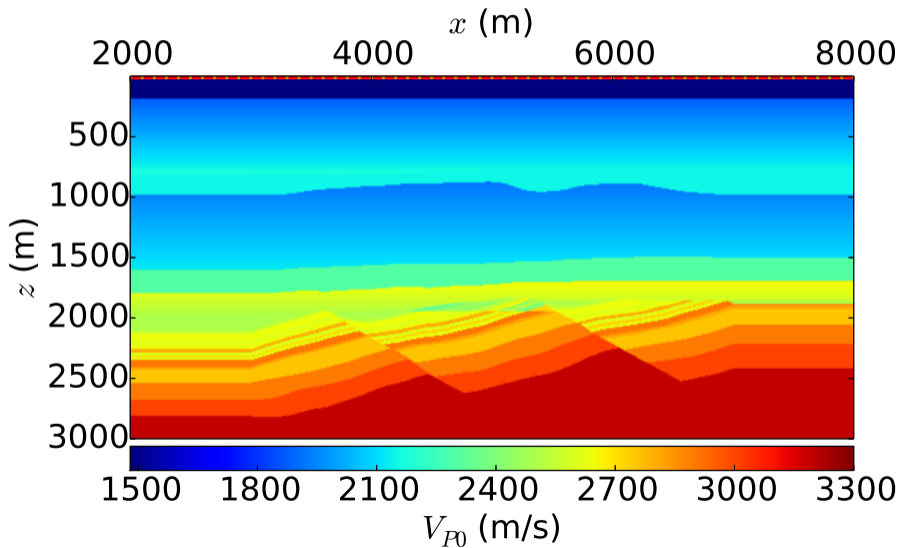


## Simple model

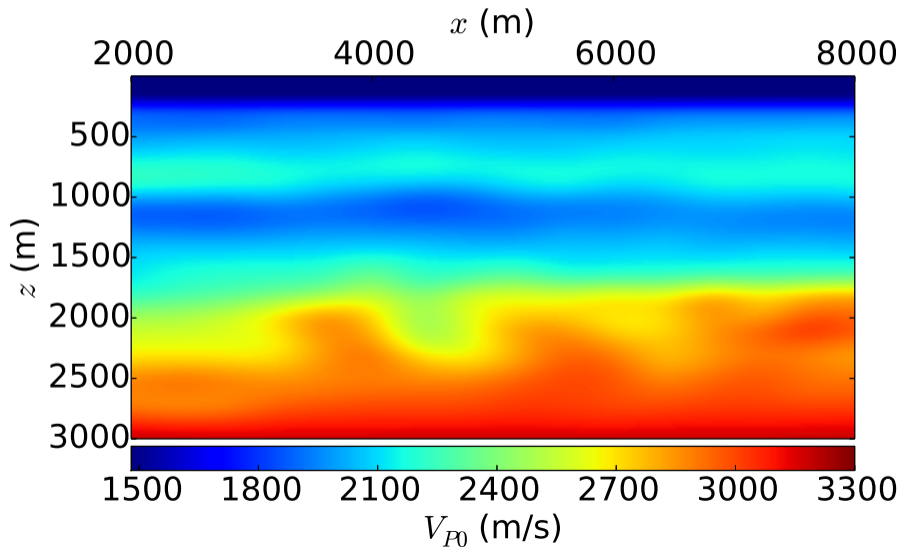
- Homogeneous model with a Gaussian perturbation in  $V_{P0}$  or  $\varepsilon$
- $\sim 3.5$  km long,  $\sim 2$  km deep
- $500 \times 300$  grid points
- Transmission experiment: 34 shots at the top, 500 receivers at the bottom
- Source: 10 Hz Ricker wavelet
- Receivers: P,  $V_z$ ,  $V_x$
- $V_{P0} = 3000$  m/s,  $V_{S0} = 1500$  m/s,  $\varepsilon = 0.1$ ,  $\delta = -0.05$ .



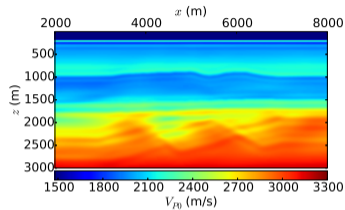
## Complex model



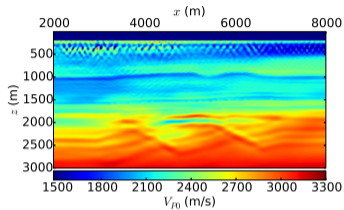
## Starting model



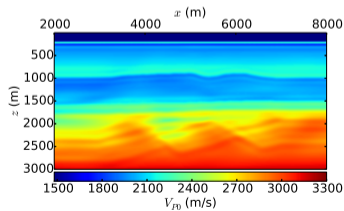
# Complex model



(a) Parameterization 1

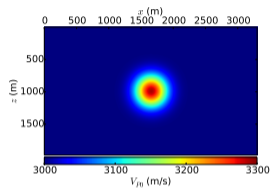


(b) Parameterization 2

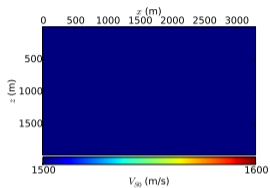


(c) Parameterization 3

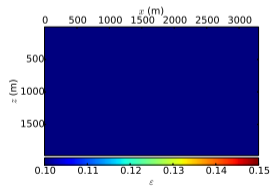
## True models, $V_{P0}$ perturbation



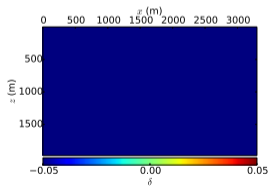
(a)



(b)

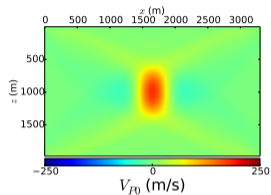


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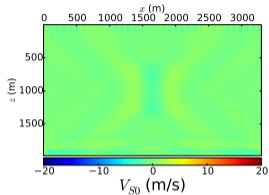


(d)

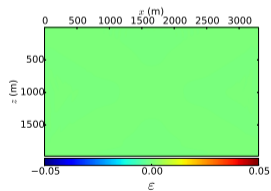
## Simple model $V_{P0}$ , parameterization 1



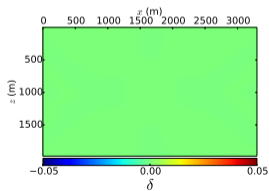
(a)



(b)

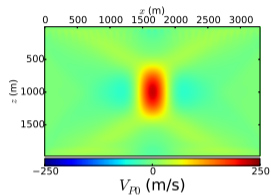


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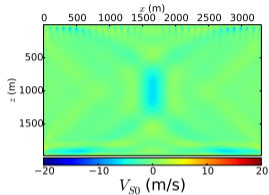


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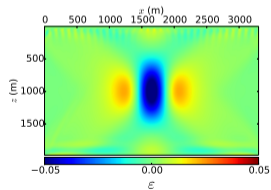
## Simple model $V_{P0}$ , parameterization 2



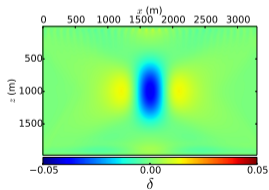
(a)



(b)



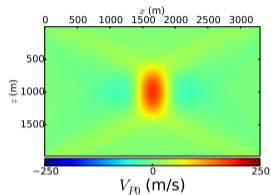
(c)



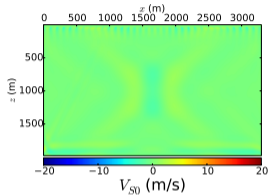
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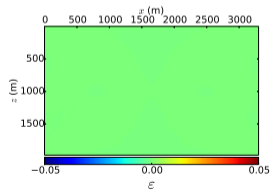
## Simple model $V_{P0}$ , parameterization 3



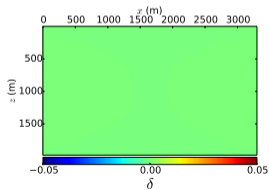
(a)



(b)

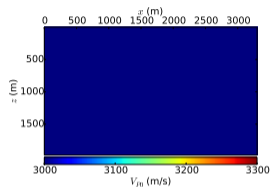


(c)

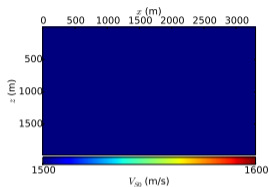


(d)

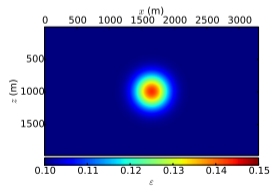
## True models, $\varepsilon$ perturbation



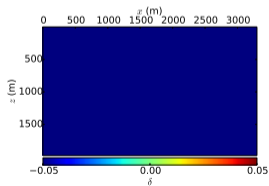
(a)



(b)

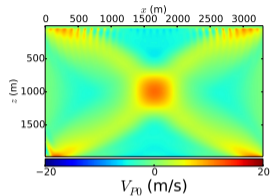


(c)

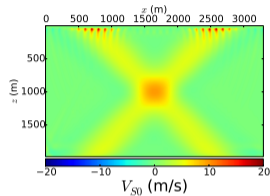


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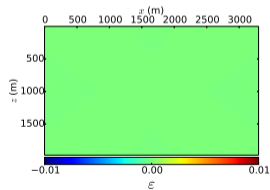
# Simple model $\varepsilon$ , parameterization 1



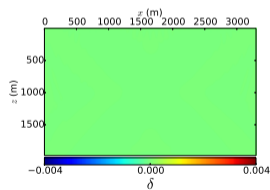
(a)



(b)

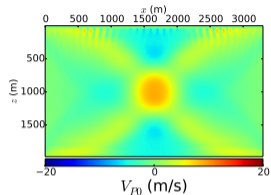


(c)

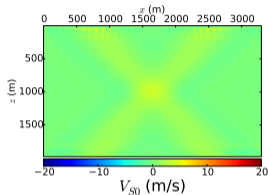


(d)

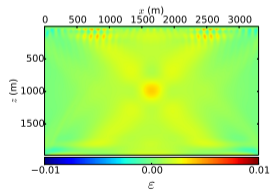
## Simple model $\varepsilon$ , parameterization 2



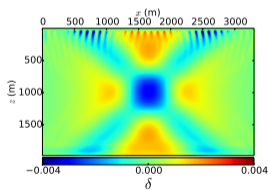
(a)



(b)

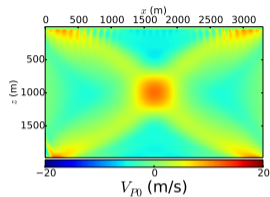


(c)

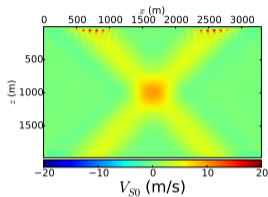


(d)

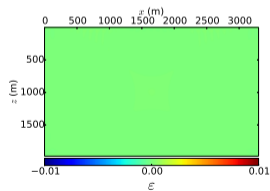
## Simple model $\varepsilon$ , parameterization 3



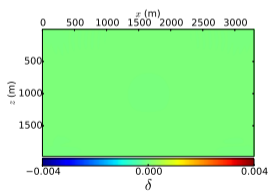
(a)



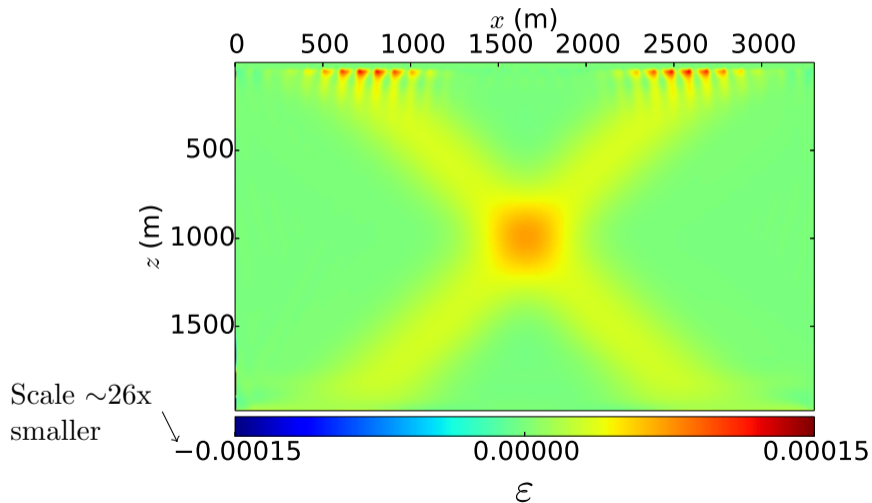
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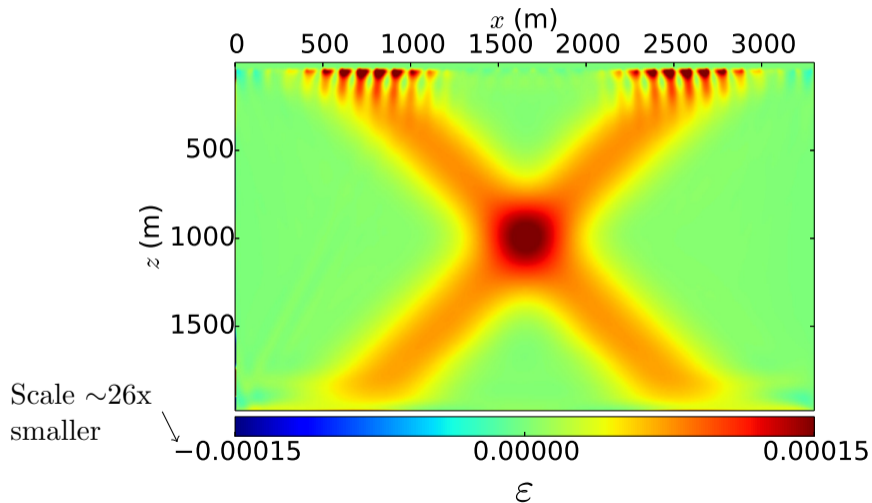
(c)



(d)

$\varepsilon$  update, parameterization 1

## $\varepsilon$ update, parameterization 3



## Conclusions

- Multiparameter FWI of multicomponent transmission data in simple models as well as  $V_{P0}$  inversion in a complex model, using three different model parameterizations in each example.



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- Simple model experiments demonstrate significant crosstalk in parameterization two.

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- Multiparameter FWI of multicomponent transmission data in simple models as well as  $V_{P0}$  inversion in a complex model, using three different model parameterizations in each example.
- $V_{P0}$  inversions show that using a parameterization that contains as much information as possible about the horizontal component of the wavefields is crucial.
- Simple model experiments demonstrate significant crosstalk in parameterization two.
- Parameterizations one and three perform very similarly, and both have problems updating anisotropy parameters in multiparameter inversion, likely as a consequence of the amplitude difference between velocities and anisotropy parameters.

# Acknowledgments

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