

# Sensitivity Analysis and Uncertainty in EFWI Using the Hessian Matrix

ROSE meeting, 24th April 2018

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# Outline

- ▶ Brief introduction to (Elastic) Full Waveform Inversion
- ▶ Present the Frechét derivative
- ▶ Adjoint theory for calculating the Hessian
- ▶ Calculate the Hessian
  - ▶ Gradient model
  - ▶ Gullfaks model

# Motivation

- ▶ Measure the resolution of FWI
- ▶ Quantify parameter cross-talk
- ▶ Investigate possibility of a Newton solver

# Elastic wave propagation

Elastic waves can be described by the wave-operator  $\mathbf{L}(\mathbf{u}, \mathbf{m})$ :

$$\mathbf{L}(\mathbf{u}, \mathbf{m}) = \rho(\mathbf{x})\ddot{\mathbf{u}}(\mathbf{x}, t) - \nabla \sigma(\mathbf{x}, t) = \mathbf{f}(\mathbf{x}, t), \quad (1)$$

- ▶ Space-coordinates  $\mathbf{x} \in G \subset \mathbb{R}^3$
- ▶ Time  $t \in [0, T] \subset \mathbb{R}$
- ▶ Displacement field  $\mathbf{u}(\mathbf{x}, t)$
- ▶ Driving force  $\mathbf{f}(\mathbf{x}, t)$
- ▶ Model  $\mathbf{m}(\mathbf{x}, t)$
- ▶ Density  $\rho$
- ▶ Stress tensor  $\sigma(\mathbf{x}, t)$
- ▶ Stiffness tensor  $\sigma_{ij} = C_{ijkl} \partial_k u_l$

# Full Waveform inversion<sup>1</sup>

1. Guess a starting model (**m**) based on other information.

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<sup>1</sup>Tarantola 1984; Mora 1987.

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3. Compare the synthetic recording to the real.

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5. Apply the model update.

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1. Guess a starting model (**m**) based on other information.
2. Do a synthetic run.
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4. Use this information to calculate a gradient update.
5. Apply the model update.
6. Repeat.

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## Misfit – How good is our model?

A measure of how our model performs compared to a reference model, the misfit function

$$\Psi = \Psi(\mathbf{u}(\mathbf{m}, \mathbf{x}_r), \mathbf{d}_0) \quad (2)$$

- ▶ Receiver location  $\mathbf{x}_r$
- ▶ Reference recording  $\mathbf{d}_0$

## Misfit

Introduce the Jacobian  $\mathbf{J}$  as

$$\nabla_m \Psi(\mathbf{m} + \delta \mathbf{m}) = \mathbf{J}(\mathbf{m} + \delta \mathbf{m}) \quad (3)$$

Linearise around  $\mathbf{m}$  resulting in

$$\mathbf{J}(\mathbf{m} + \delta \mathbf{m}) \simeq \mathbf{J}(\mathbf{m}) + \nabla_m \mathbf{J}(\mathbf{m}) \delta \mathbf{m} = 0. \quad (4)$$

Thus introducing the Hessian

$$\mathbb{H}(\mathbf{m}) = \nabla_m \mathbf{J}(\mathbf{m}) = \nabla_m \nabla_m \Psi(\mathbf{m}). \quad (5)$$

## Iterative methods<sup>2</sup>

- ▶ The model update  $\delta \mathbf{m}$  can then be obtained by solving

$$\mathbb{H}(\mathbf{m})\delta \mathbf{m} = -\mathbf{J}(\mathbf{m}) \quad (6)$$

- ▶ Iff  $\mathbf{H}$  is invertible we can “simply” solve

$$\delta \mathbf{m} = -\mathbb{H}^{-1} \mathbf{J}.$$

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- ▶ A common approximation is

$$\delta \mathbf{m} \simeq \alpha \mathbf{J},$$

and a line search for the optimal  $\alpha \in \mathbf{R}$ .

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## Jacobian

By backward propagating the misfit kernels we can obtain the adjoint field  $\mathbf{u}^\dagger$  which we can use to calculate the Jacobian using the Frechét derivative

$$\mathbf{J} = \mathbf{J}(\mathbf{u}^\dagger, \mathbf{u}) = \int_T \mathbf{u}^\dagger \nabla_m \mathbf{L}(\mathbf{u}, \mathbf{m}) dt, \quad (7)$$

which boils down to cross-correlating the adjoint and forward fields.

## Perturbed wavefields

Perturbed forward field

$$\delta \mathbf{u} = \lim_{\nu \rightarrow 0} \frac{1}{\nu} \left[ \mathbf{u}(\mathbf{m} + \nu \delta \mathbf{m}) - \mathbf{u}(\mathbf{m}) \right], \quad (8)$$

Perturbed adjoint field

$$\delta \mathbf{u}^\dagger = \lim_{\nu \rightarrow 0} \frac{1}{\nu} \left[ \mathbf{u}^\dagger(\mathbf{m} + \nu \delta \mathbf{m}) - \mathbf{u}^\dagger(\mathbf{m}) \right], \quad (9)$$

## Hessian<sup>3</sup>

The Hessian acting on a model perturbation  $\delta \mathbf{m}$  can be split up into three components

$$\mathbb{H}\delta \mathbf{m} = \mathbf{H}_1(\mathbf{u}^\dagger, \delta \mathbf{u}) + \mathbf{H}_2(\delta \mathbf{u}^\dagger, \mathbf{u}) + \mathbf{H}_3(\mathbf{u}^\dagger, \mathbf{u}). \quad (10)$$

These components can be written out as

$$\mathbf{H}_1(\mathbf{u}^\dagger, \delta \mathbf{u}) = \int_T \mathbf{u}^\dagger \nabla_m \mathbf{L}(\delta \mathbf{u}, \mathbf{m}) dt, \quad (11)$$

$$\mathbf{H}_2(\delta \mathbf{u}^\dagger, \mathbf{u}) = \int_T \delta \mathbf{u}^\dagger \nabla_m \mathbf{L}(\mathbf{u}, \mathbf{m}) dt, \quad (12)$$

$$\mathbf{H}_3(\mathbf{u}^\dagger, \mathbf{u}) = \int_T \mathbf{u}^\dagger \nabla_m \nabla_m \mathbf{L}(\mathbf{u}, \mathbf{m})(\delta \mathbf{m}) dt. \quad (13)$$

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<sup>3</sup>Fichtner and Trampert 2011.

# Hessian

The first two terms  $\mathbf{H}_1$  and  $\mathbf{H}_2$  can be calculated in a similar way as the Jacobian by replacing the relevant fields.

The last term  $\mathbf{H}_3$  can be calculated by recycling the Jacobian calculations as

$$\mathbf{H}_3 = \begin{bmatrix} 0 & \rho^{-1} J_{V_p} & \rho^{-1} J_{V_s} \\ \rho^{-1} J_{V_p} & v_p^{-1} J_{V_p} & 0 \\ \rho^{-1} J_{V_s} & 0 & v_s^{-1} J_{V_s} \end{bmatrix} \begin{bmatrix} \delta \rho \\ \delta v_p \\ \delta v_s \end{bmatrix} \quad (14)$$

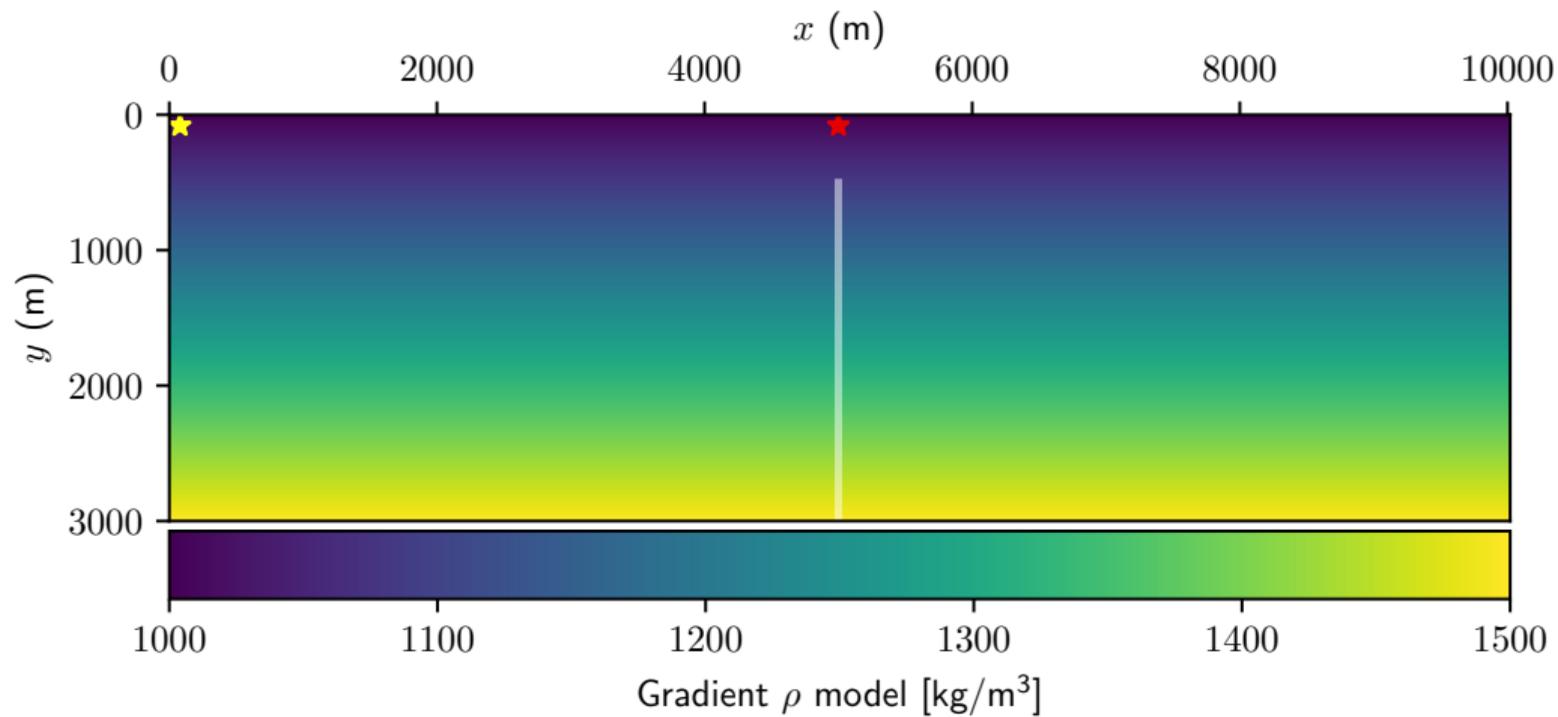
# Constructing the Hessian

$$\mathbf{H}^m(x_i)\delta m(x_j) = \mathbf{H}_i^m \delta_j^m$$

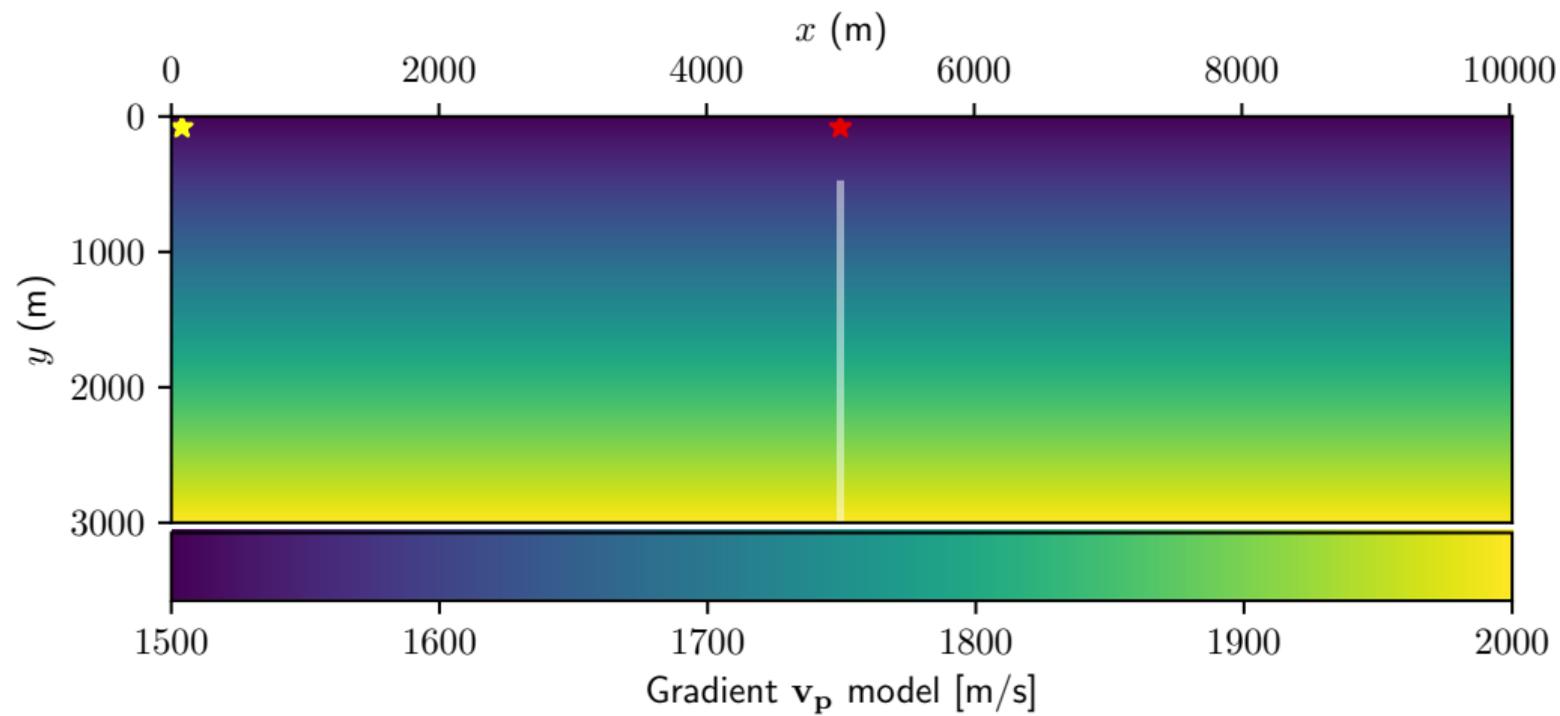
$$\left[ \begin{array}{ccc|cc|ccc} \mathbf{H}_0^\rho \delta_0^\rho & \mathbf{H}_0^\rho \delta_1^\rho & \cdots & \mathbf{H}_0^\rho \delta_0^{V_p} & \cdots & \mathbf{H}_0^\rho \delta_0^{V_s} & \cdots \\ \mathbf{H}_1^\rho \delta_0^\rho & \mathbf{H}_1^\rho \delta_1^\rho & \ddots & \vdots & \ddots & \vdots & \ddots \\ \mathbf{H}_2^\rho \delta_0^\rho & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots \\ \hline \mathbf{H}_0^{V_p} \delta_0^\rho & \cdots & \cdots & \mathbf{H}_0^{V_p} \delta_0^{V_p} & \cdots & \mathbf{H}_0^{V_p} \delta_0^{V_s} & \cdots \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots \\ \hline \mathbf{H}_0^{V_s} \delta_0^\rho & \cdots & \cdots & \mathbf{H}_0^{V_s} \delta_0^{V_p} & \cdots & \mathbf{H}_0^{V_s} \delta_0^{V_s} & \cdots \\ \vdots & \vdots & \ddots & \vdots & \ddots & \vdots & \ddots \end{array} \right]$$

(15)

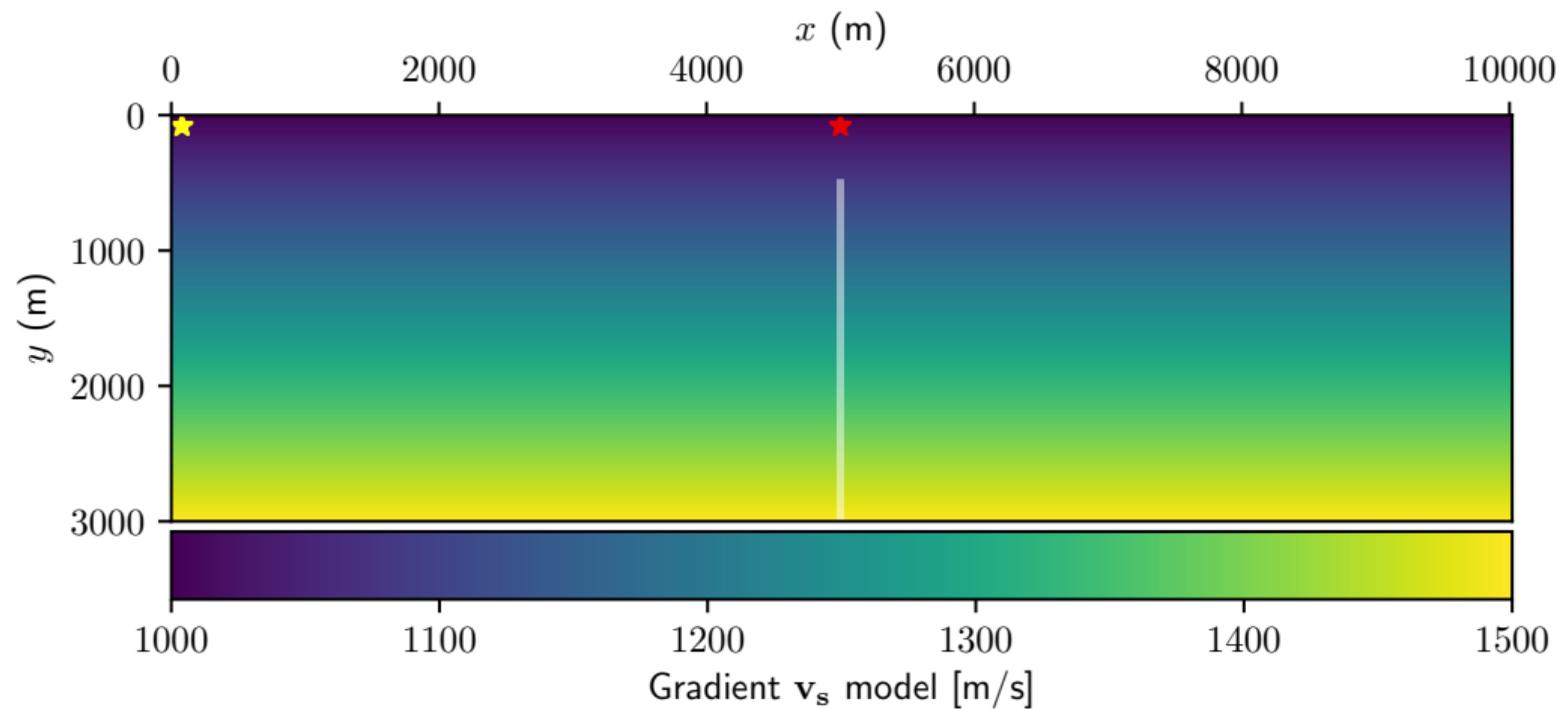
# Gradient model



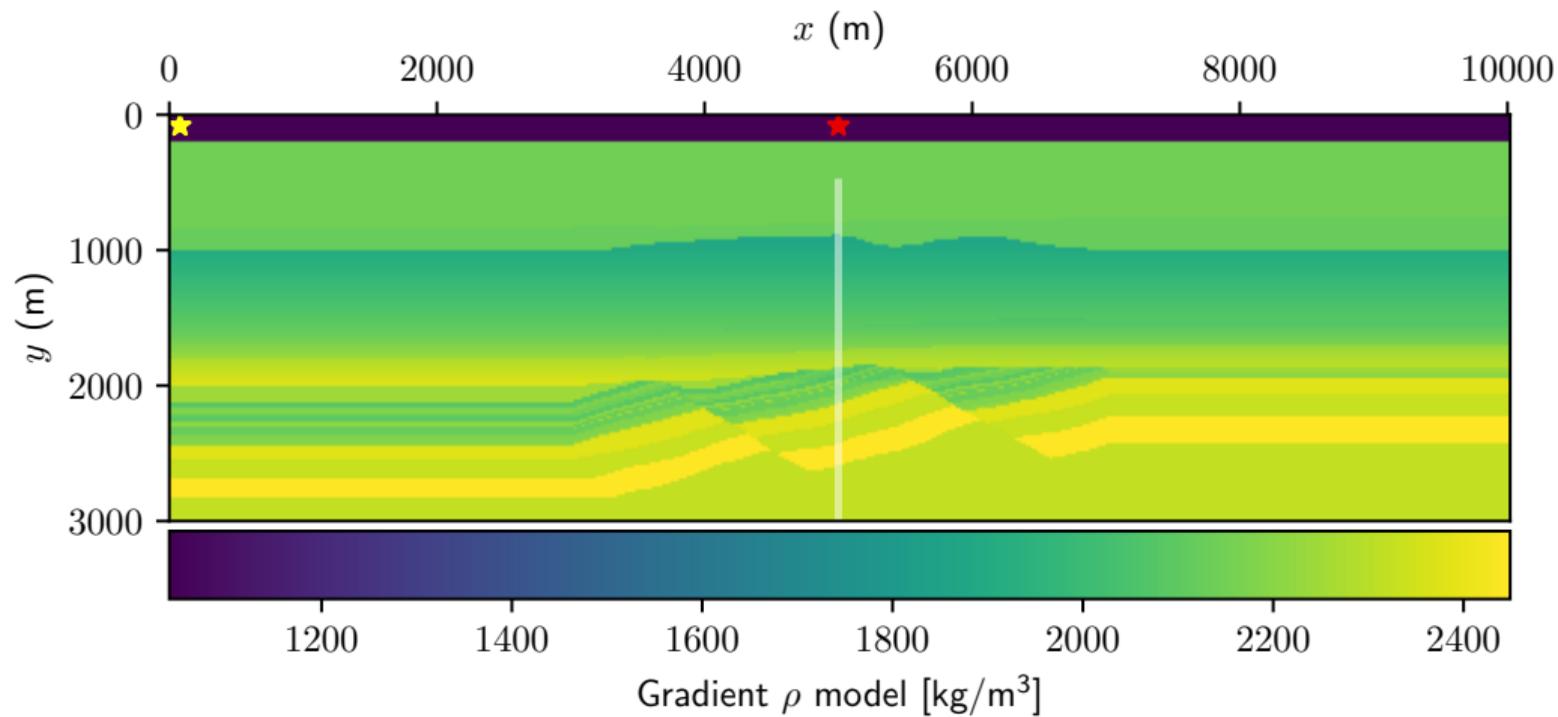
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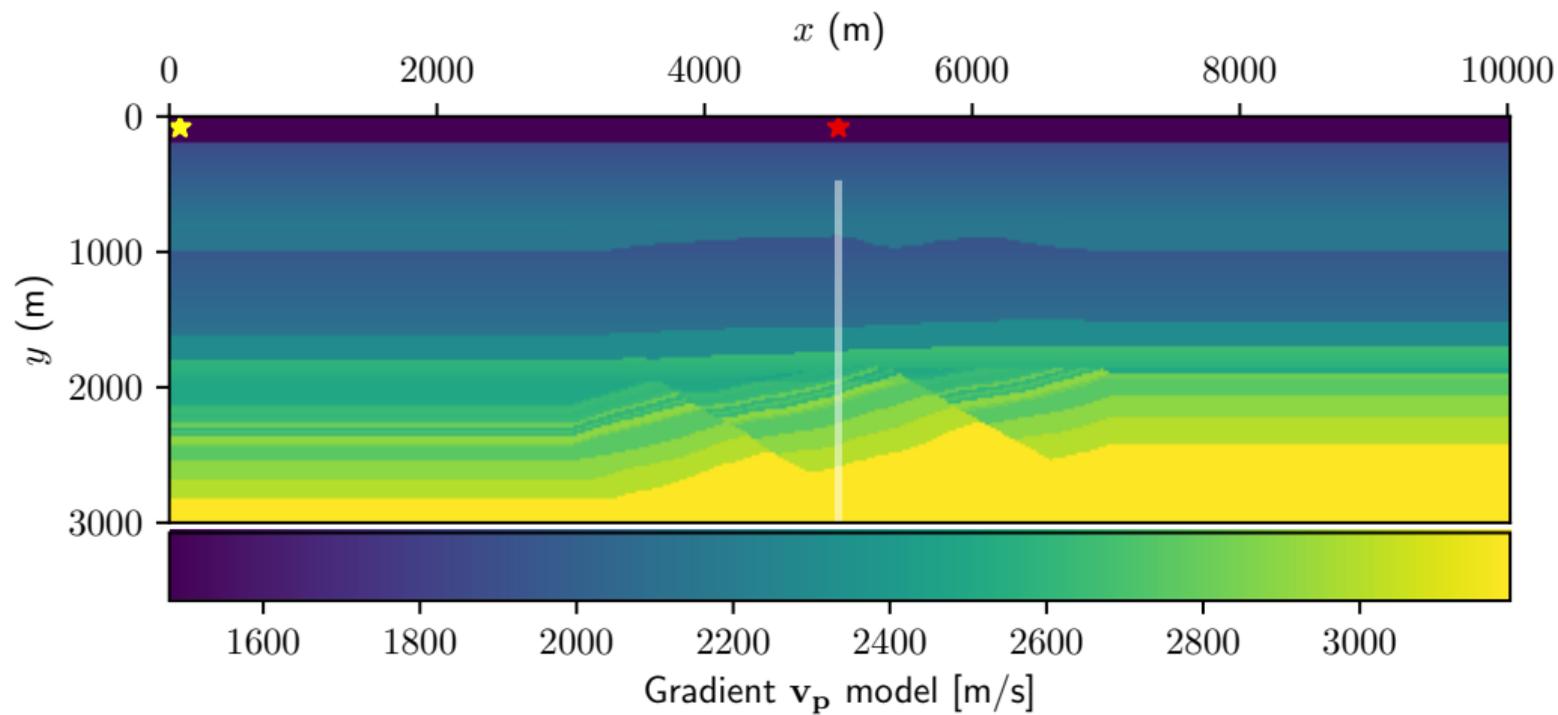
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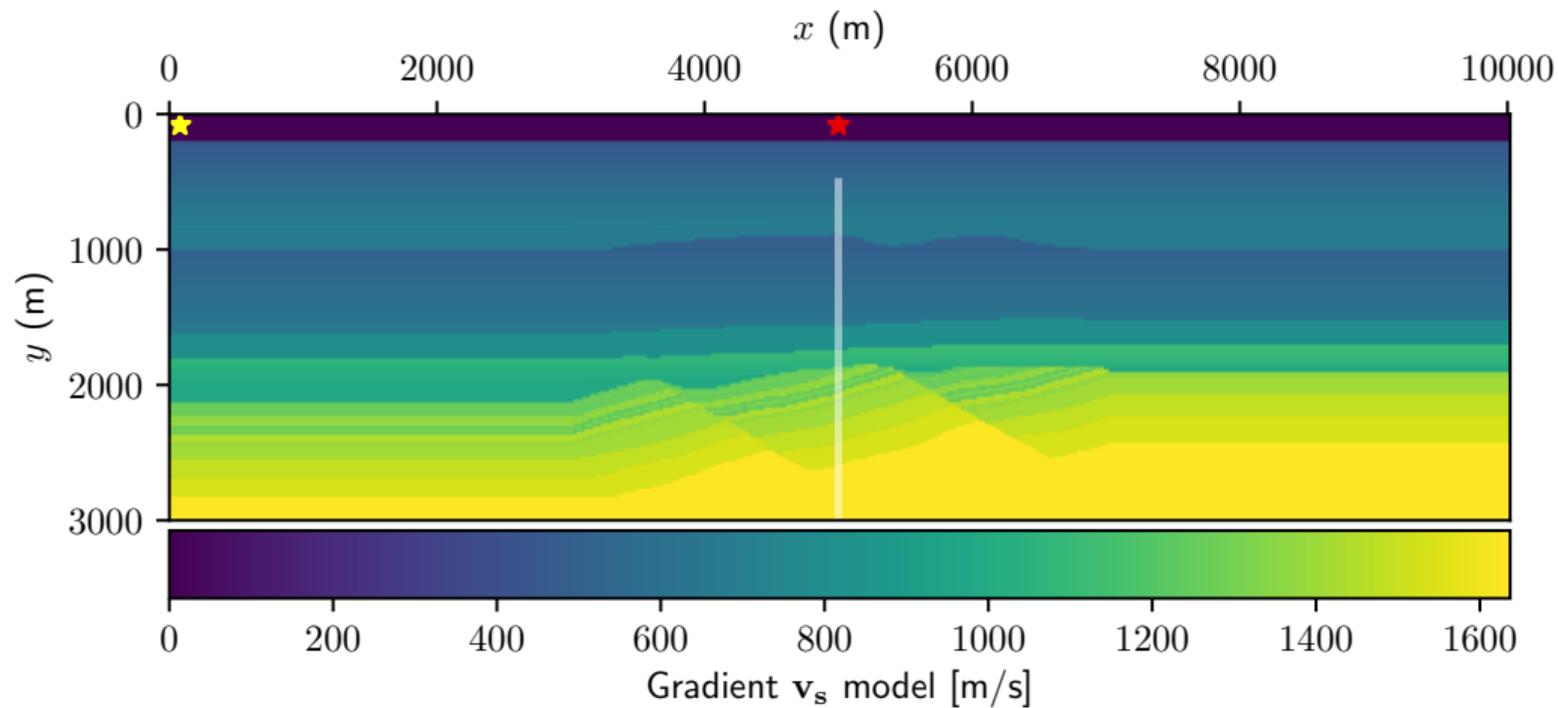
# Gulfaks model



# Gullfaks model



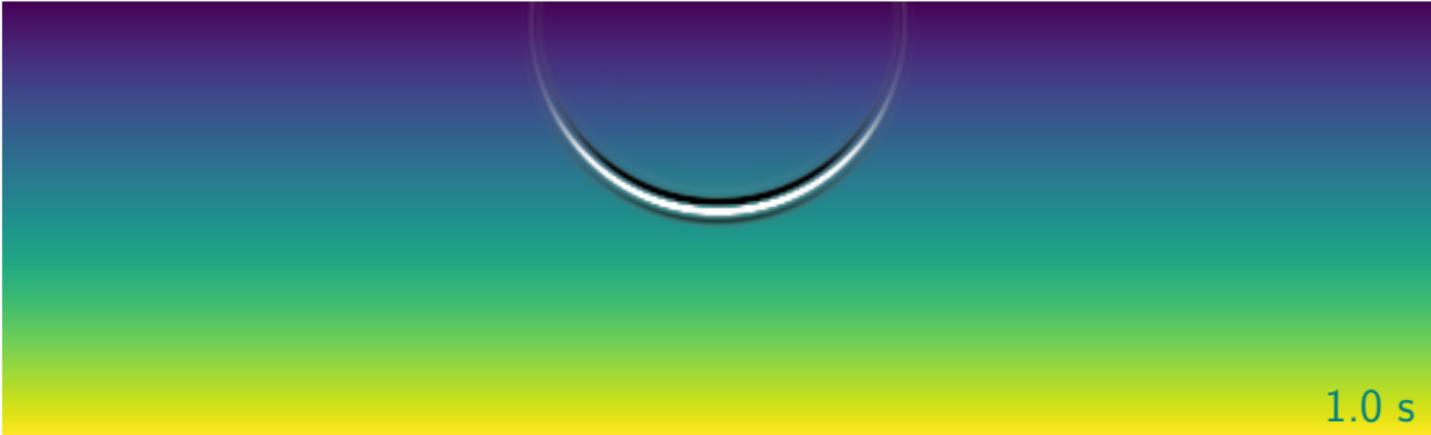
# Gulfaks model



# Gradient model — Shot 1



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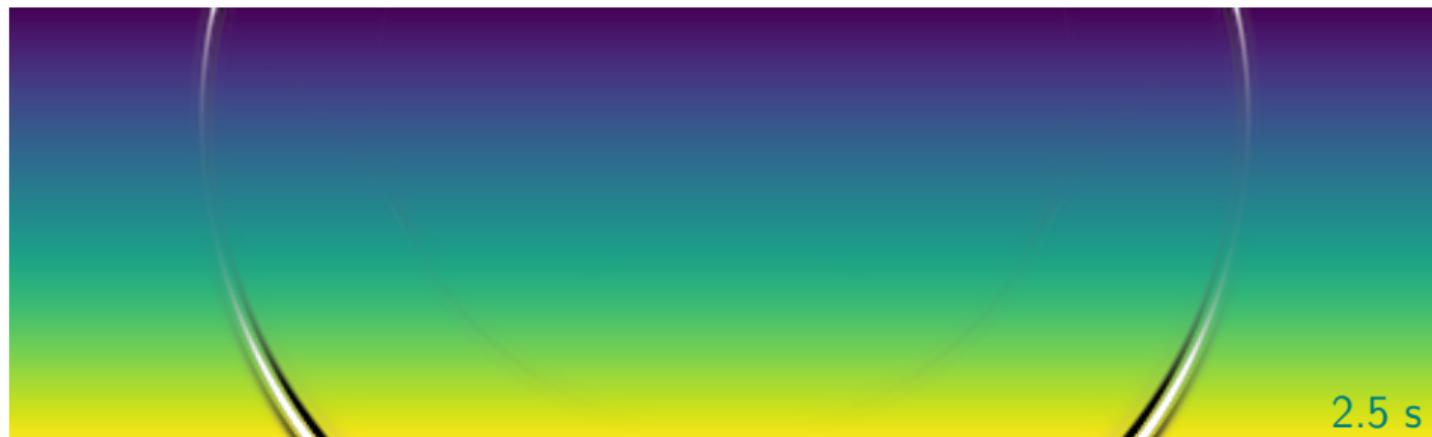
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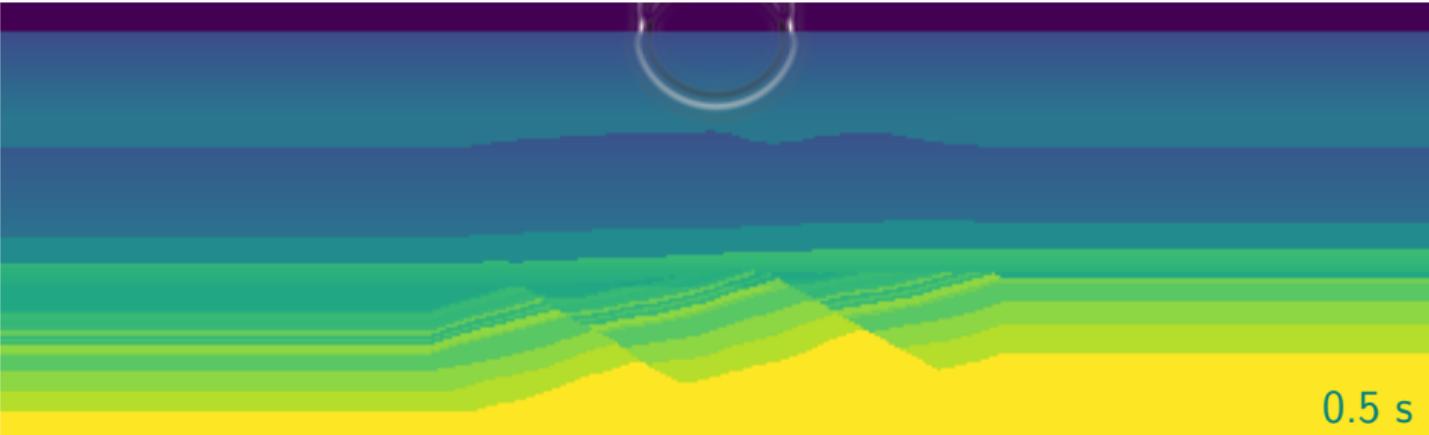
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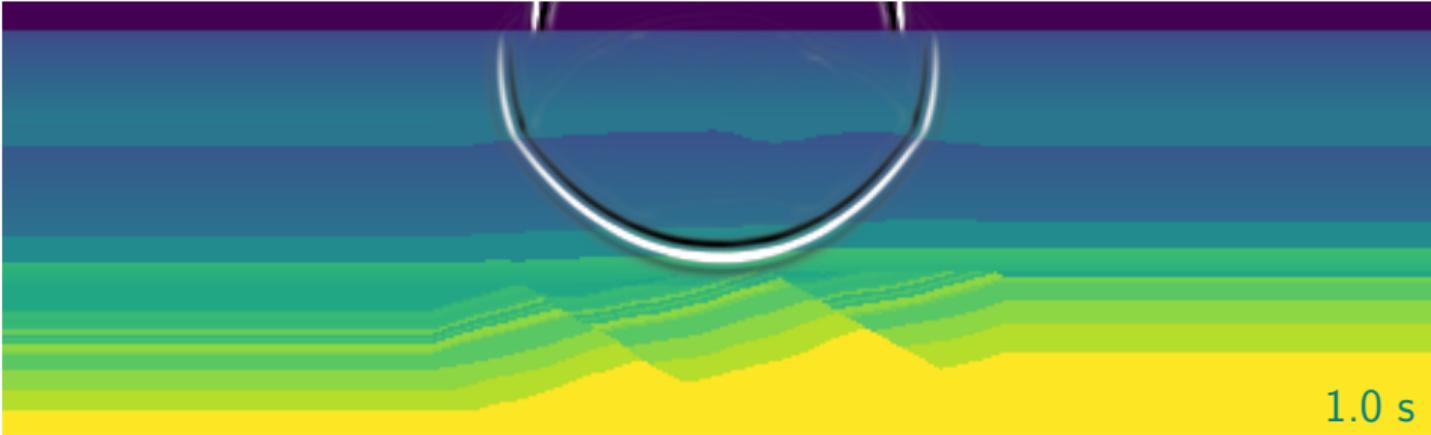
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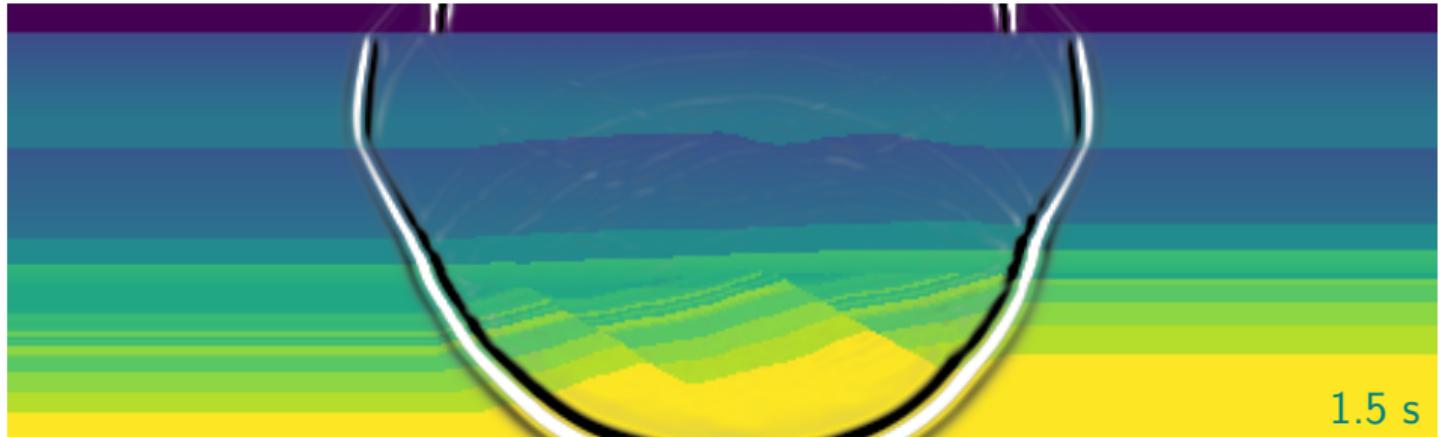
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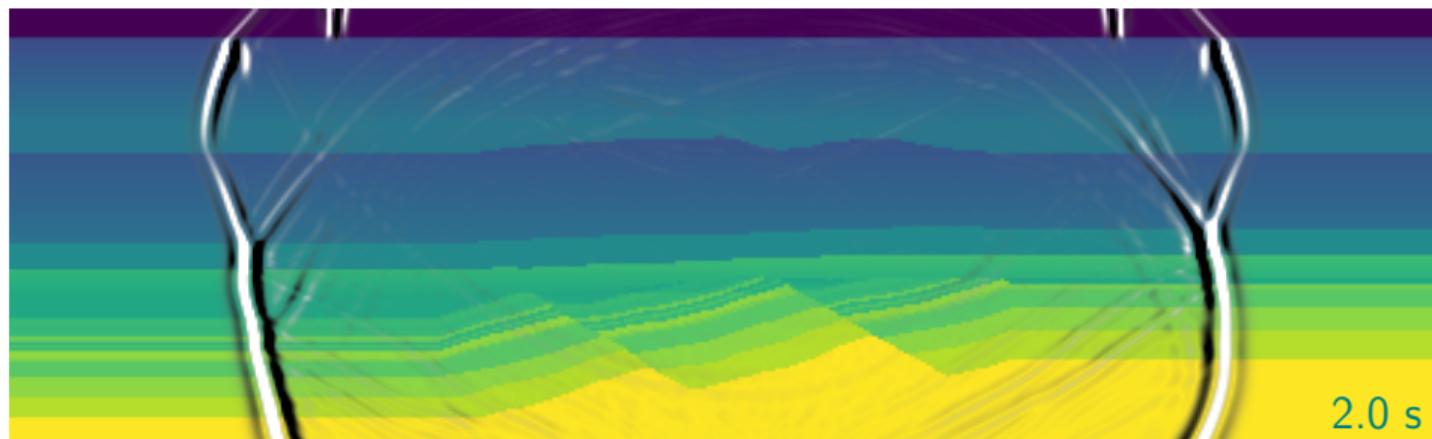
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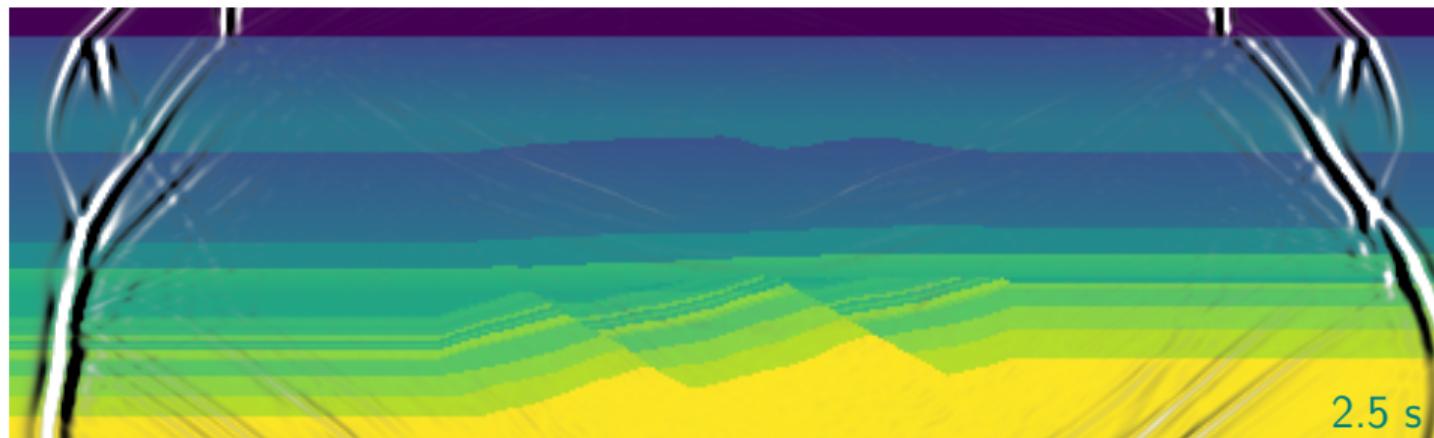
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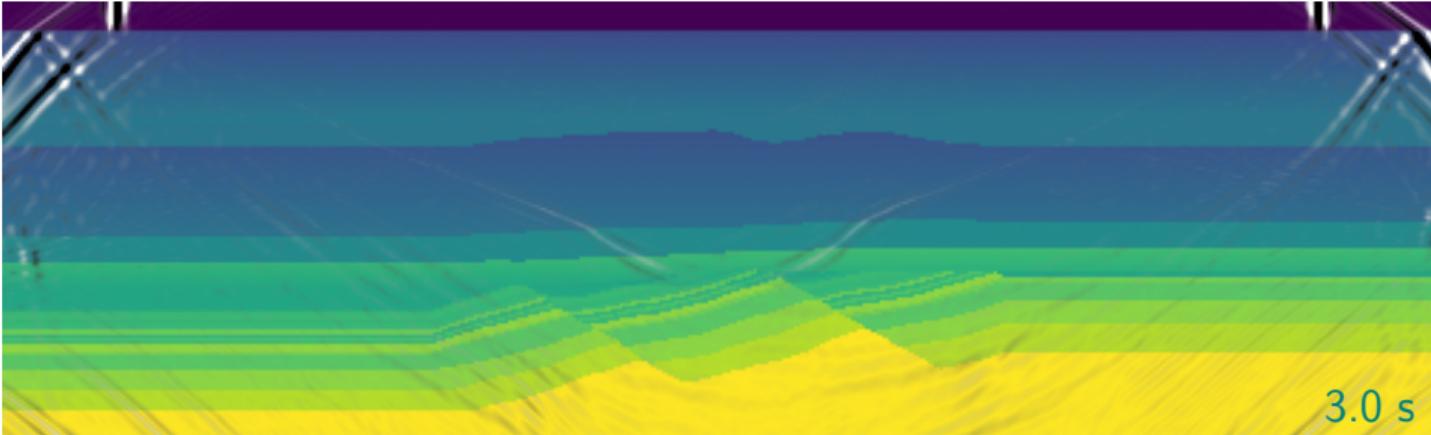
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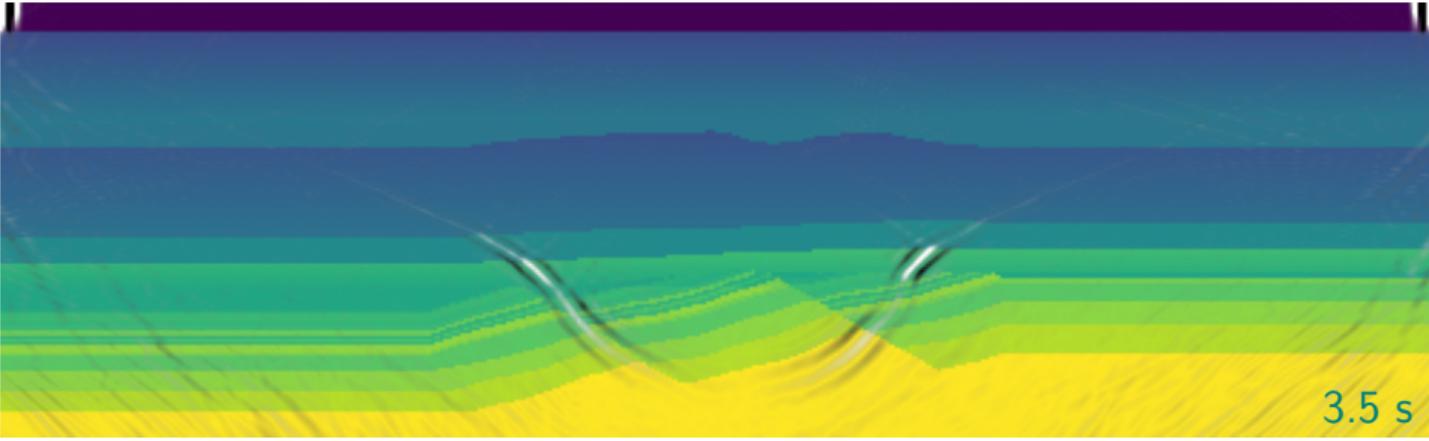
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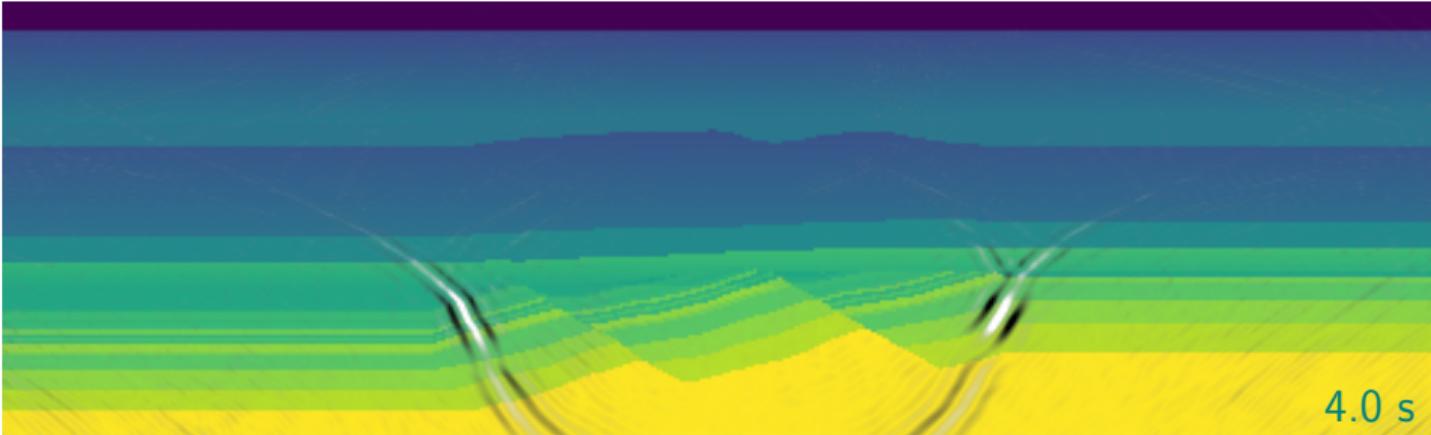
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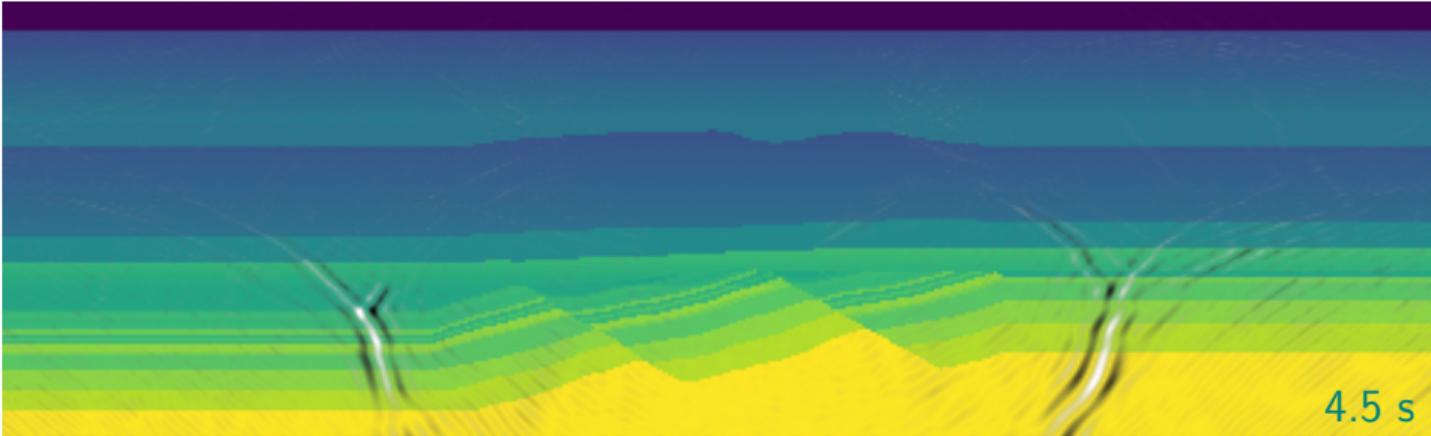
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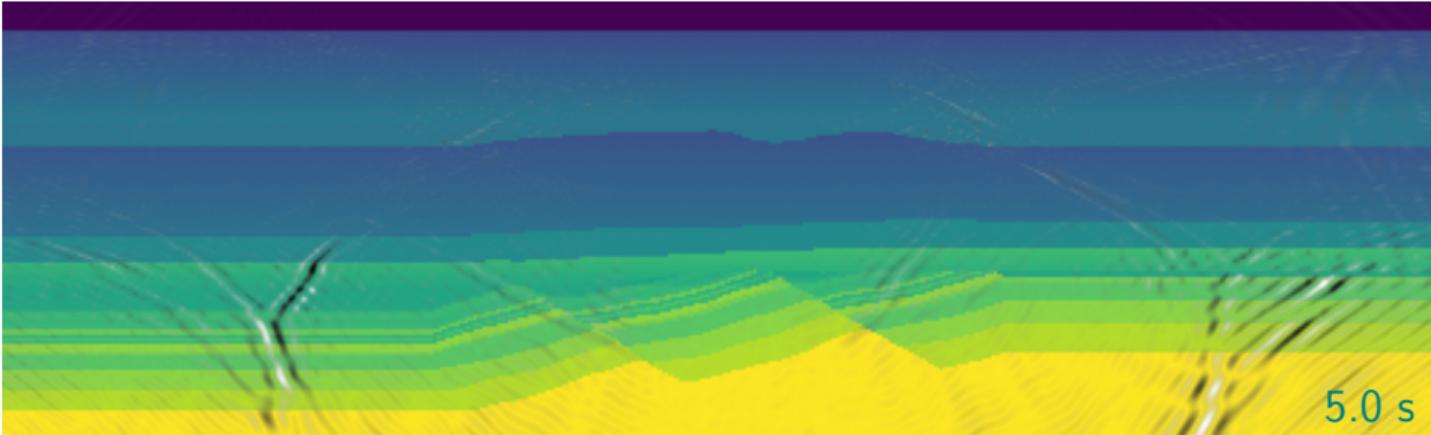
# Gullfaks model — Shot 1



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# Gullfaks model — Shot 1



# Gradient model — Shot 2



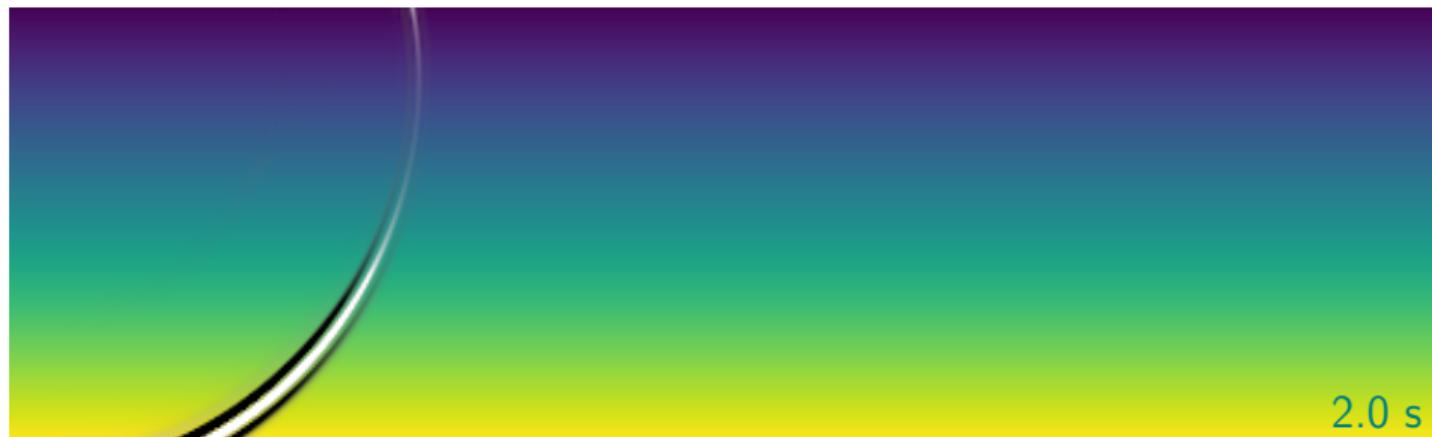
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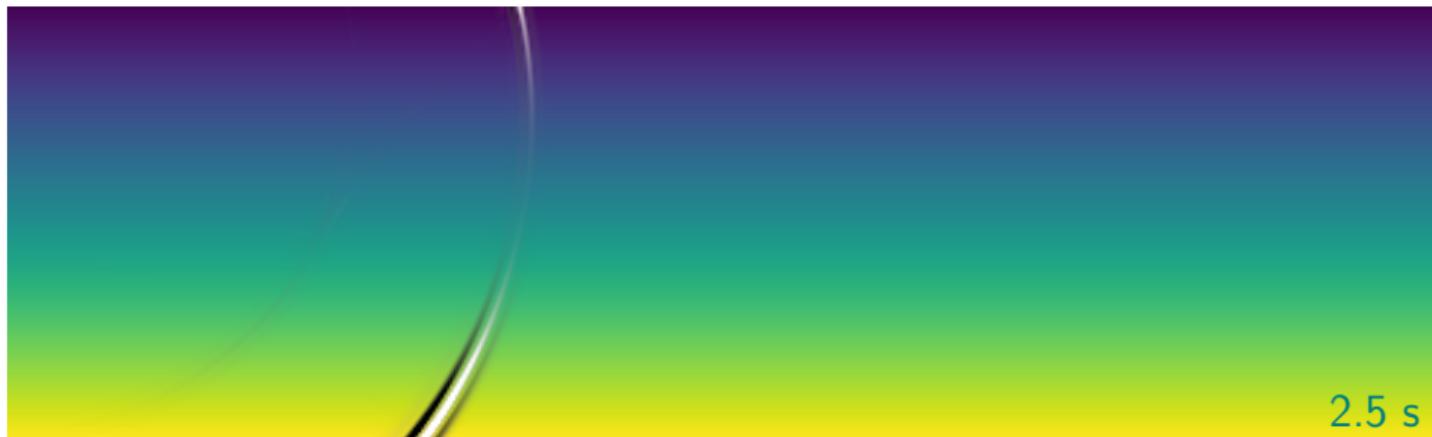
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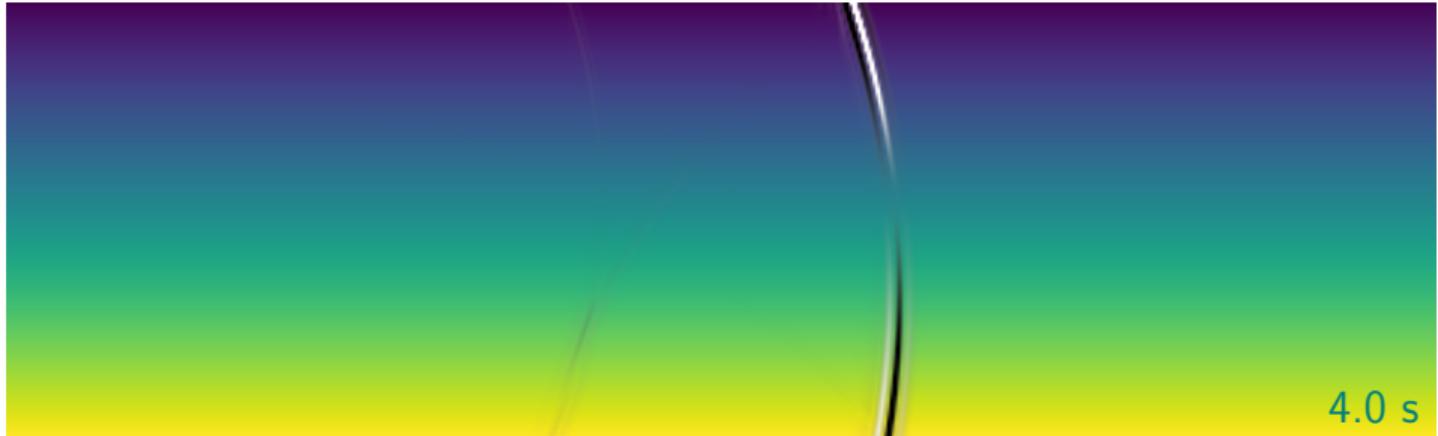
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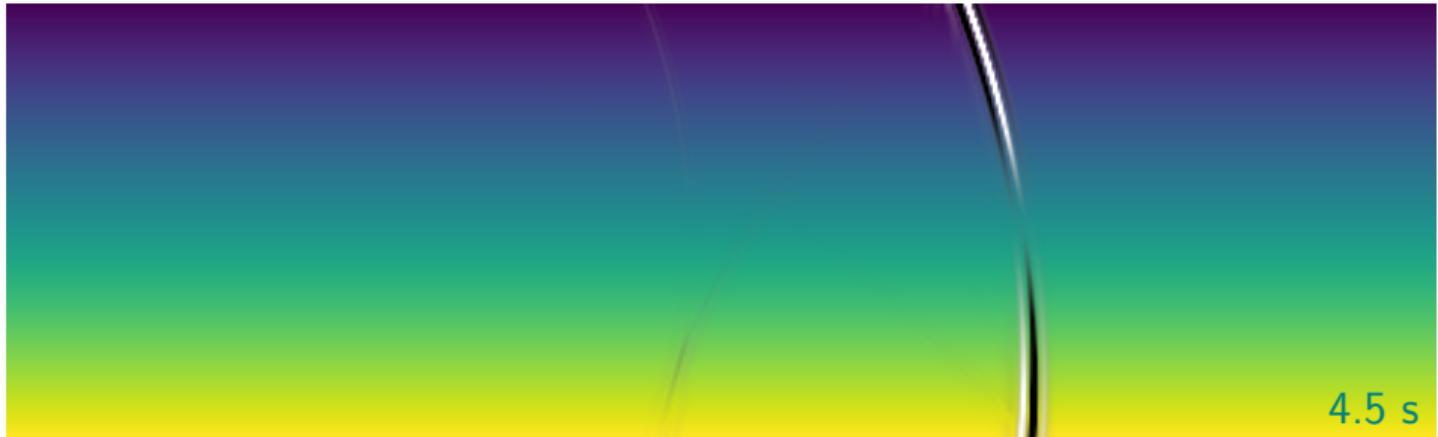
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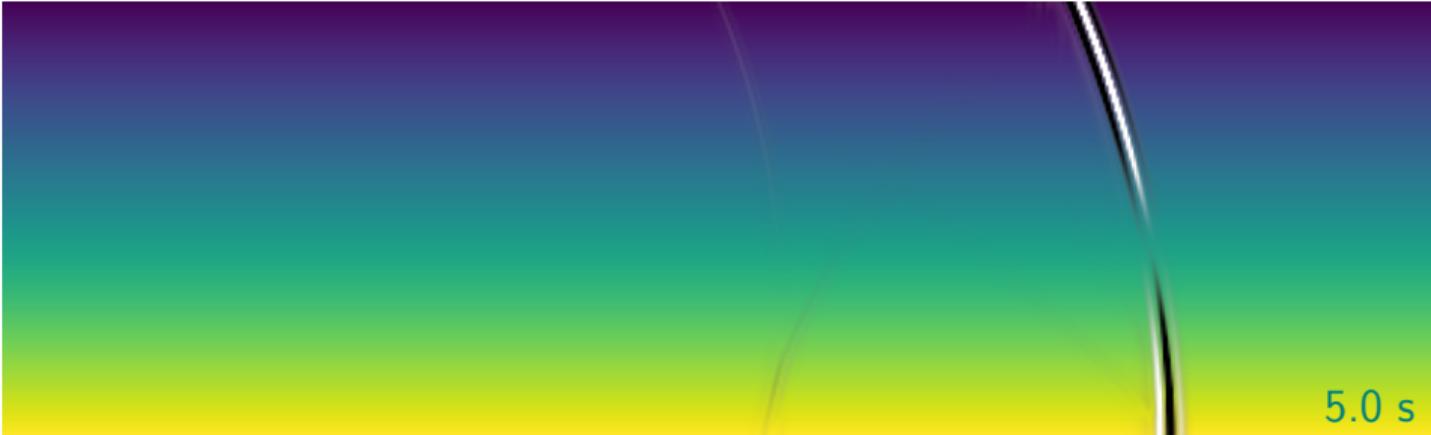
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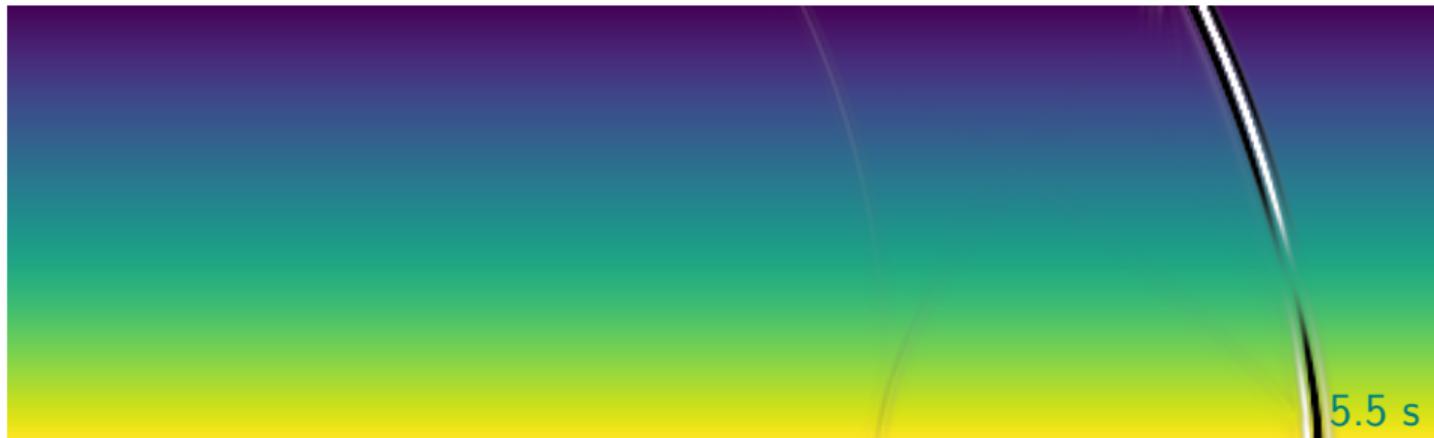
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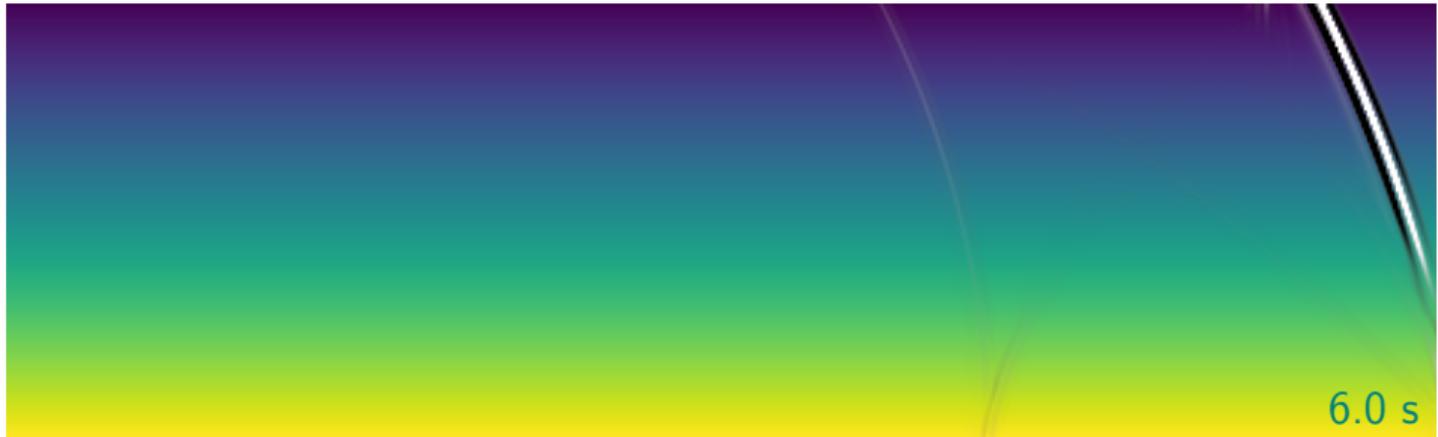
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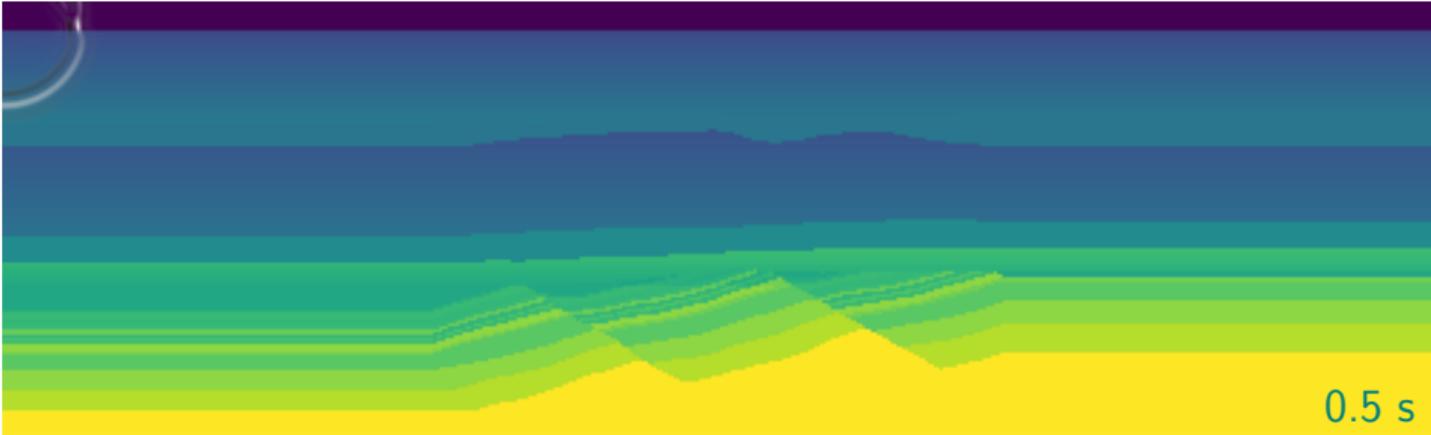
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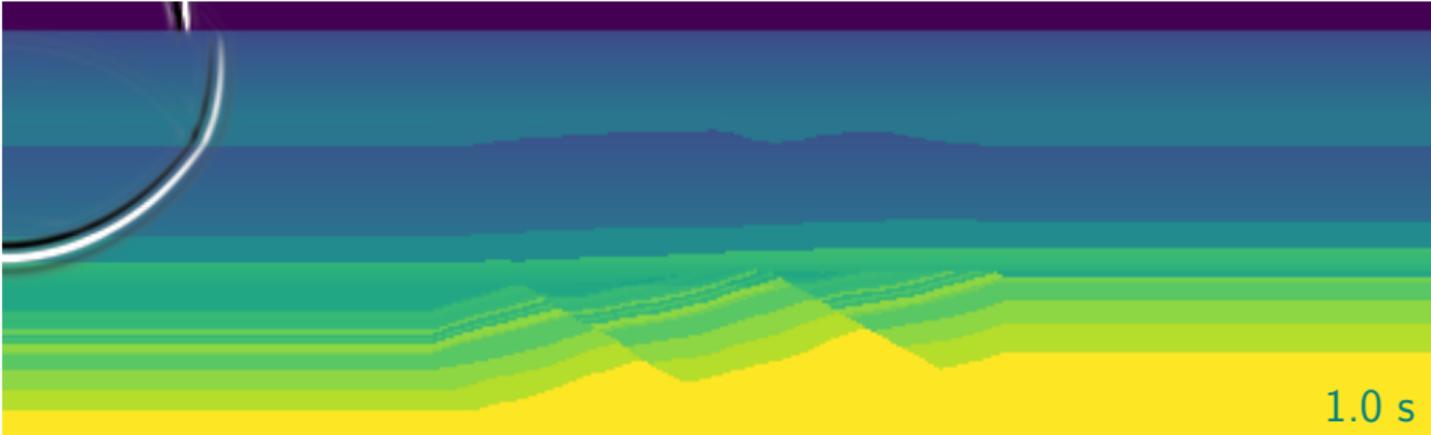
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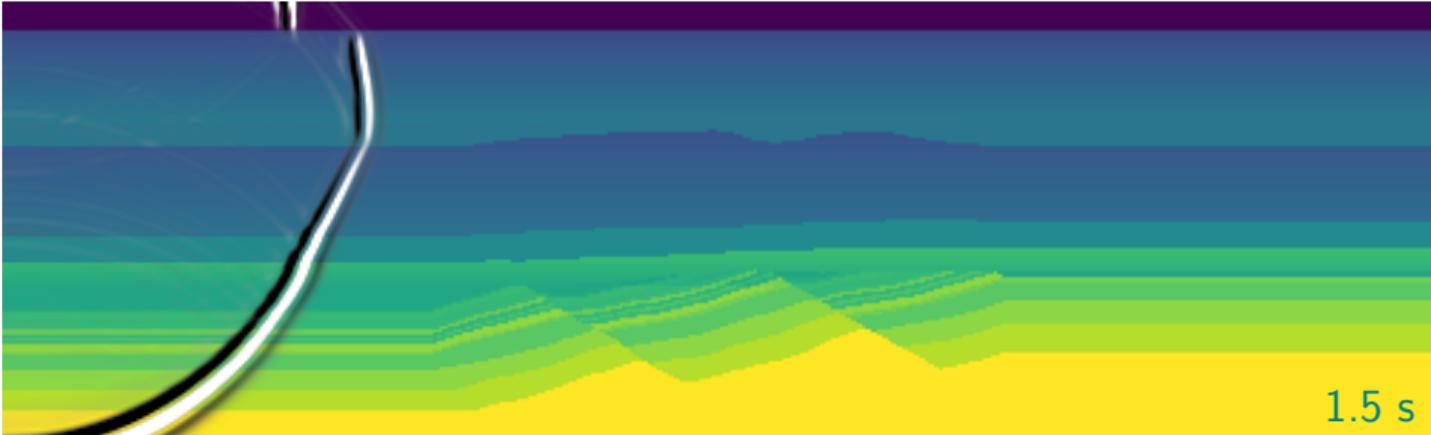
# Gullfaks model — Shot 2



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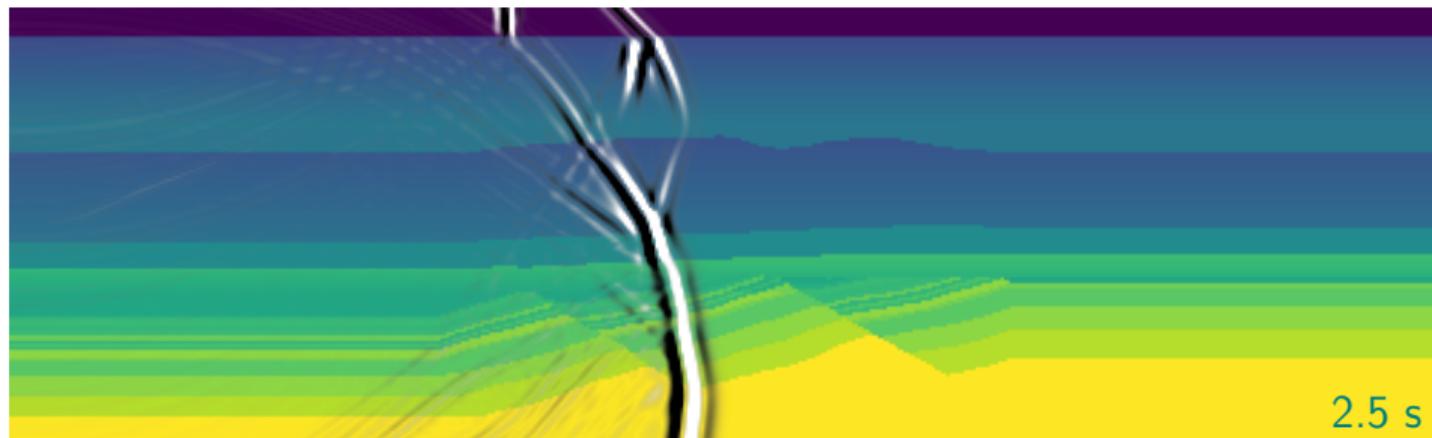
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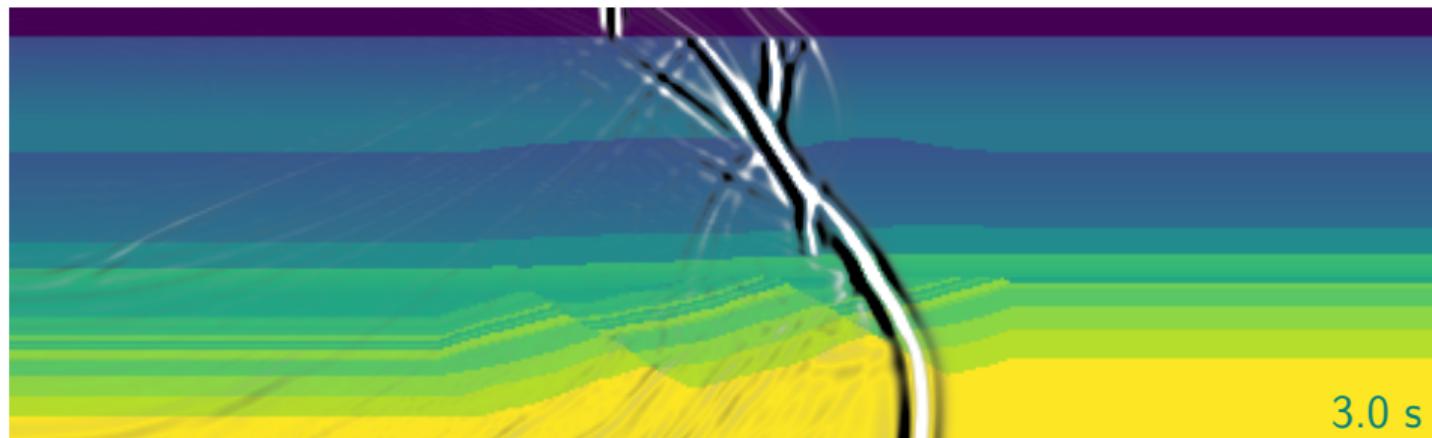
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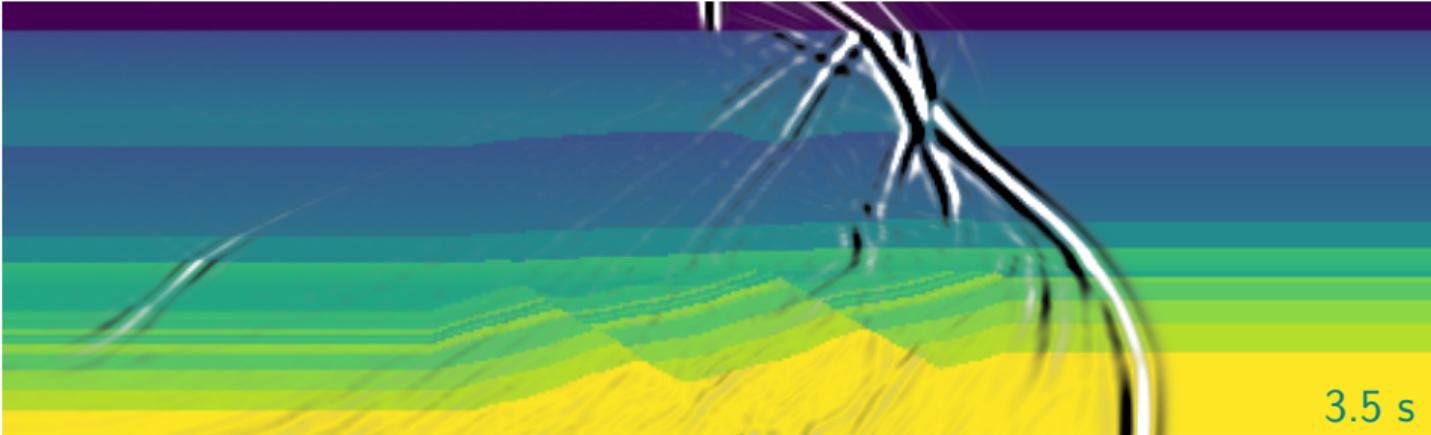
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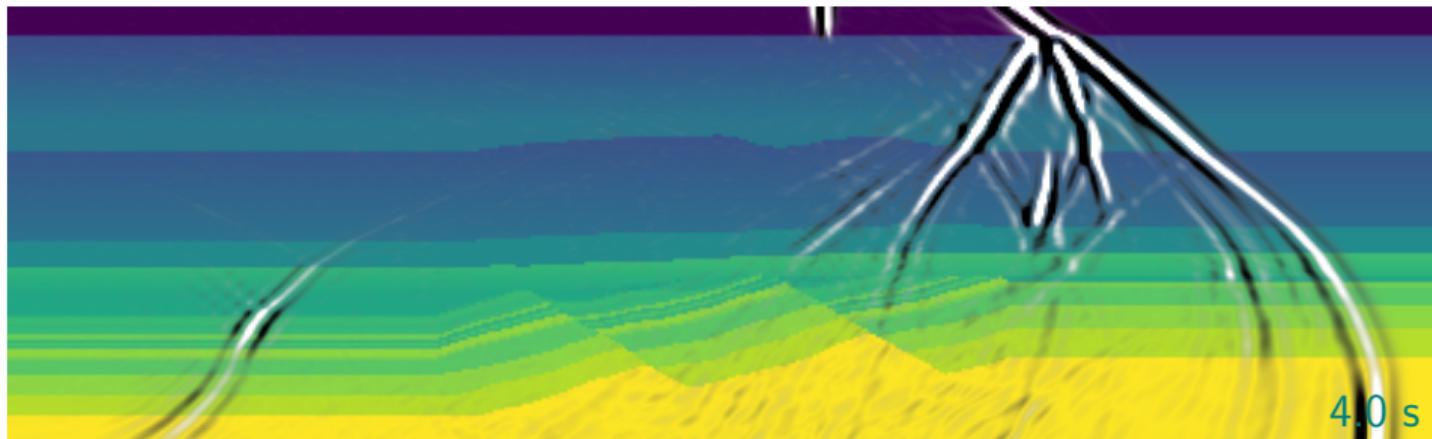
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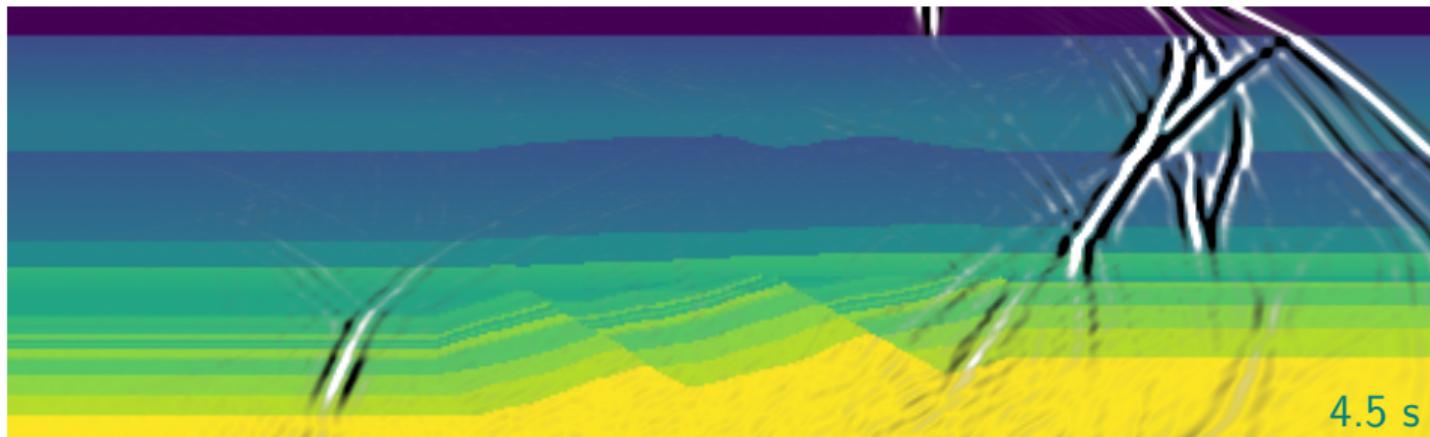
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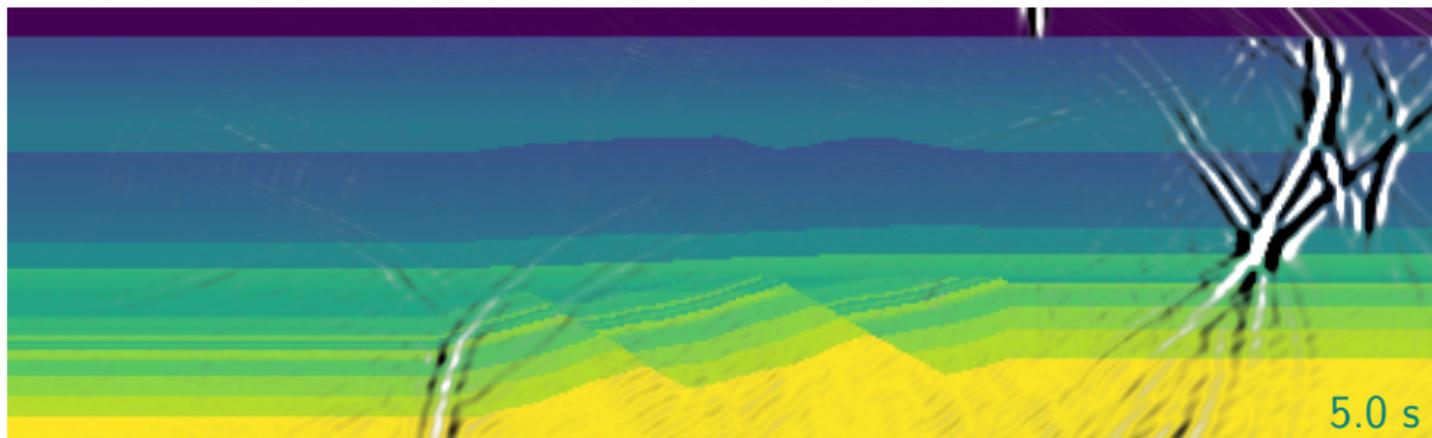
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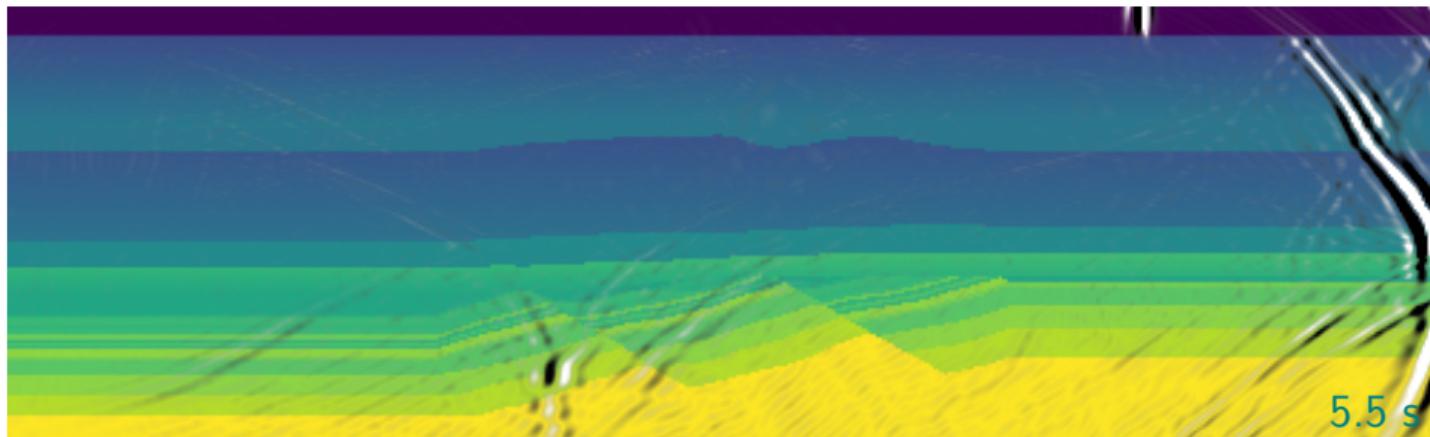
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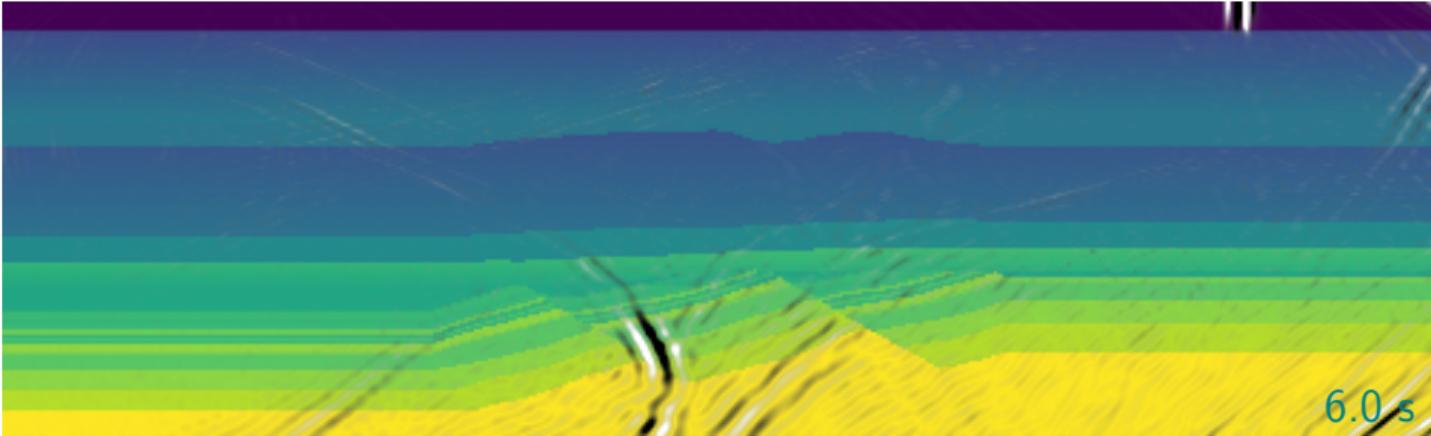
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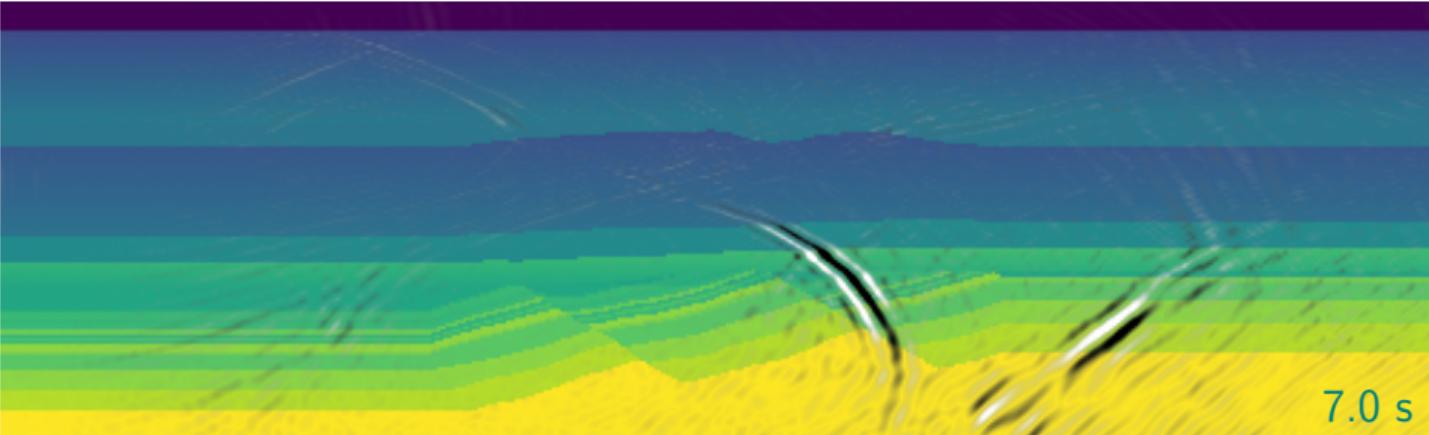
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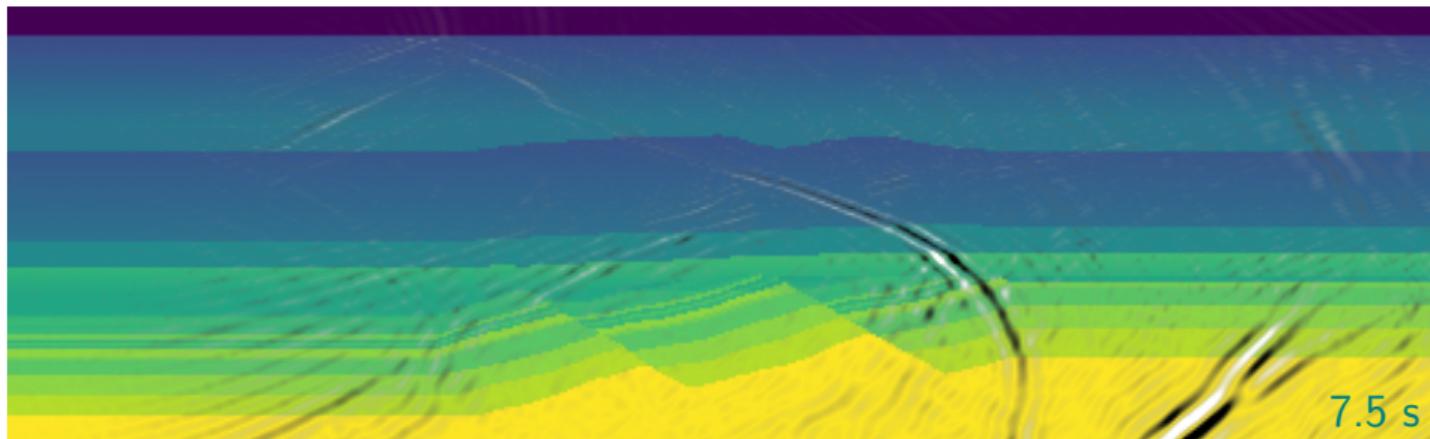
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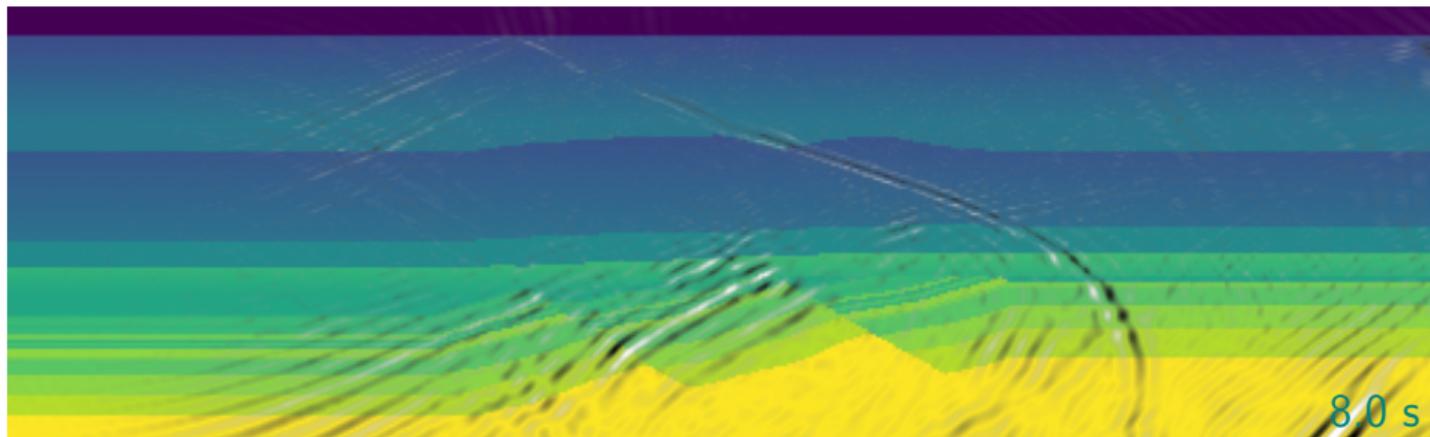
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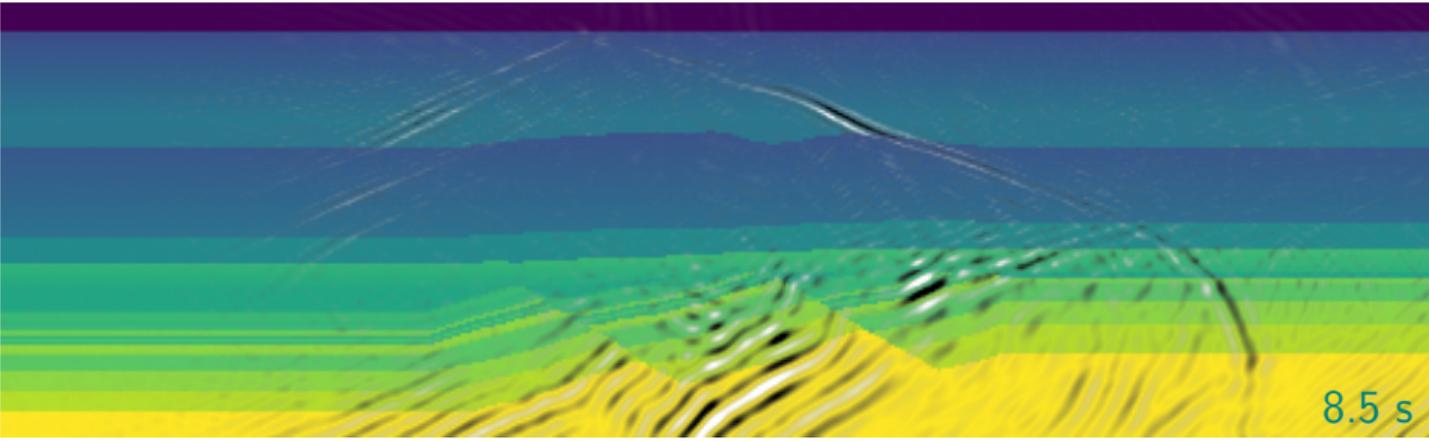
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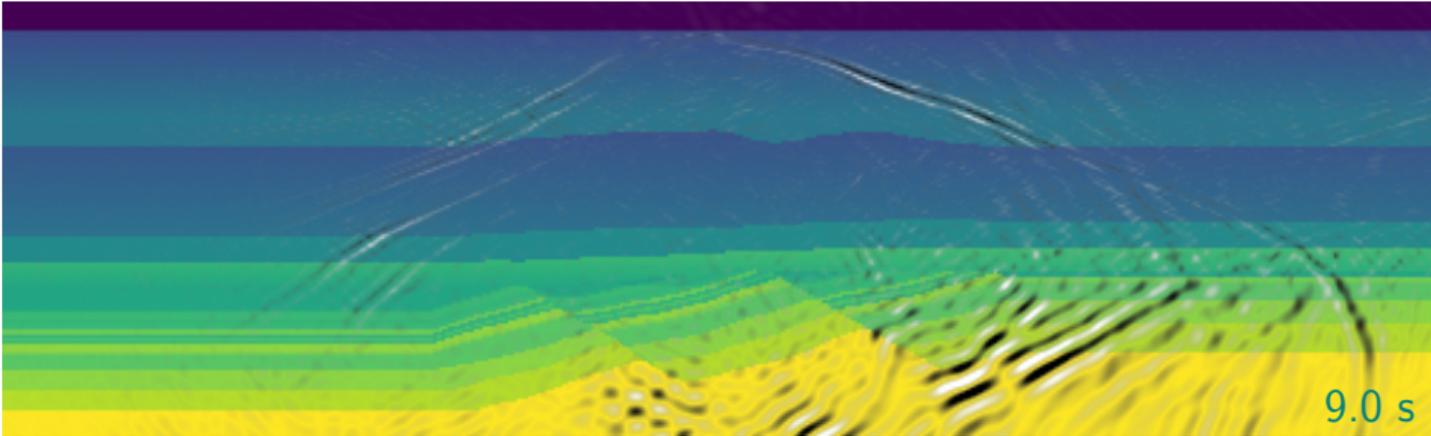
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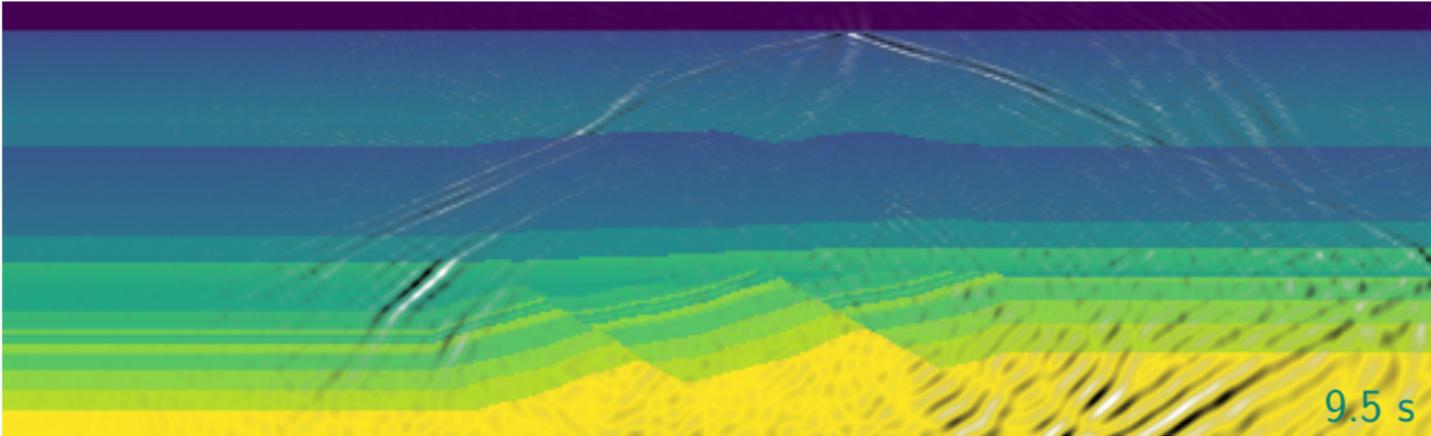
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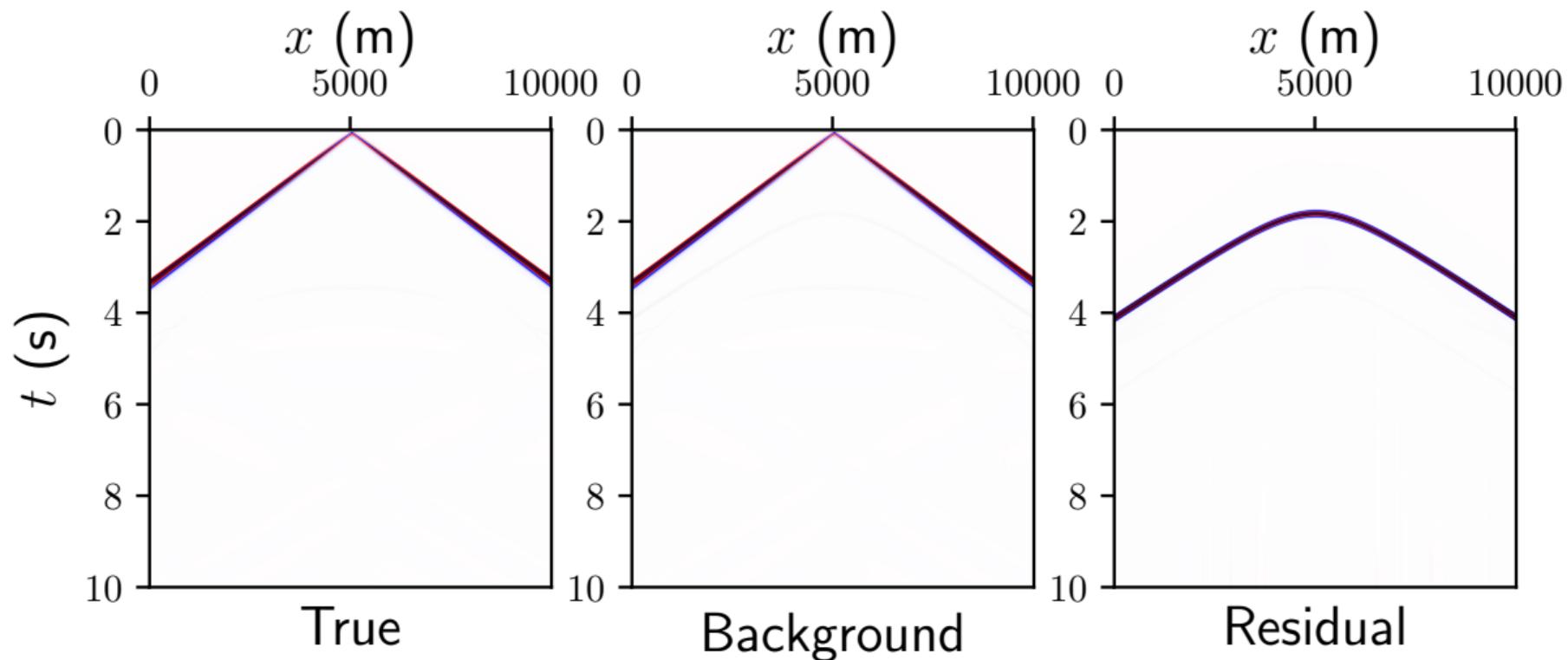
# Gullfaks model — Shot 2



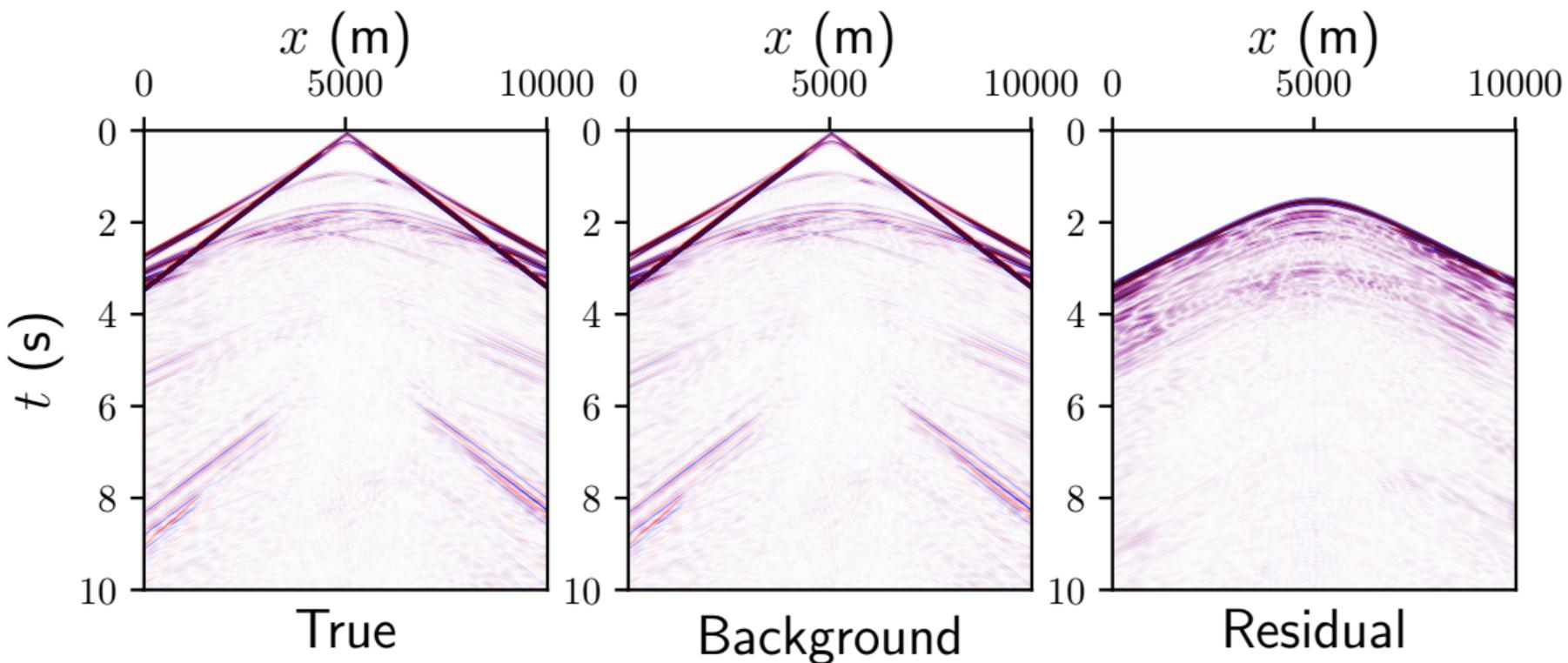
# Gullfaks model — Shot 2



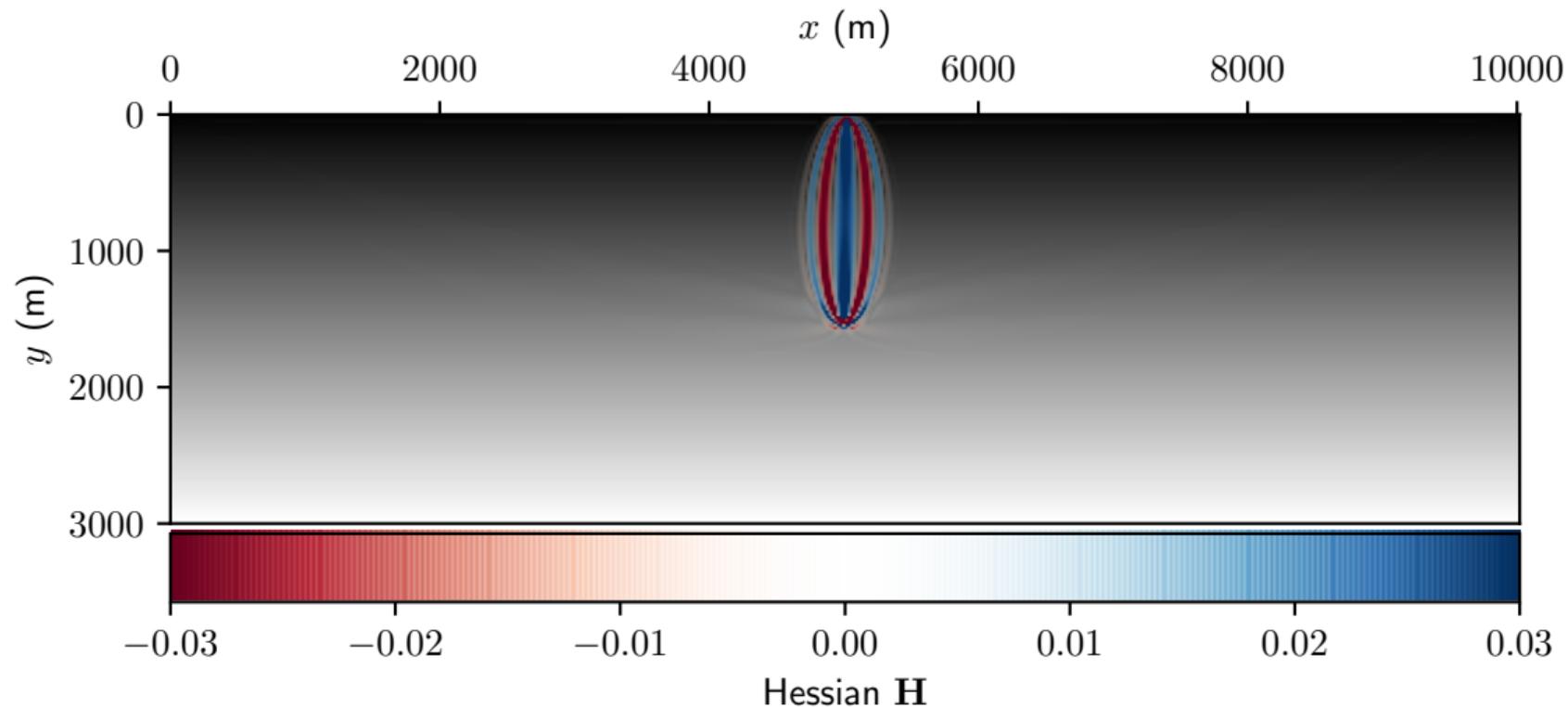
# Gradient recording shot 1, perturbation at 1500 m



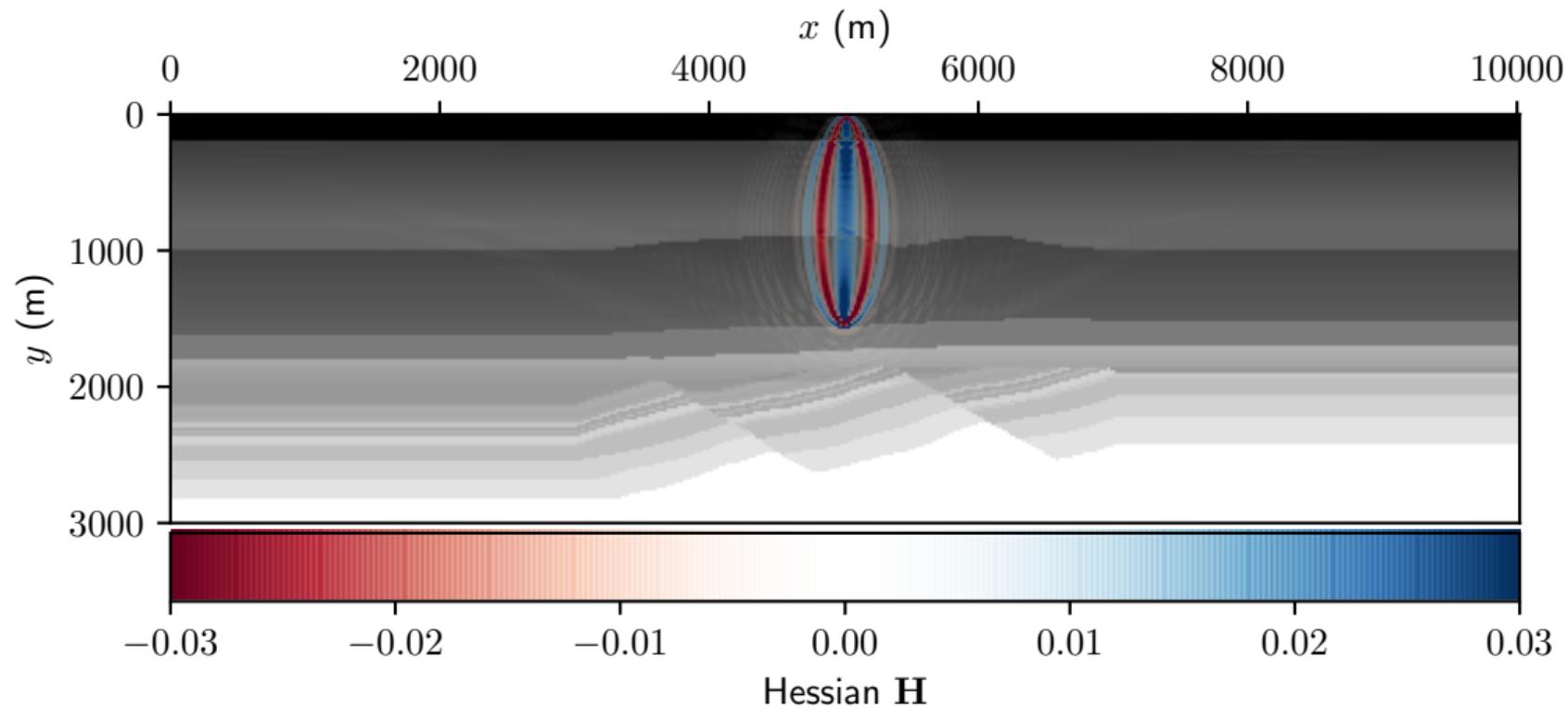
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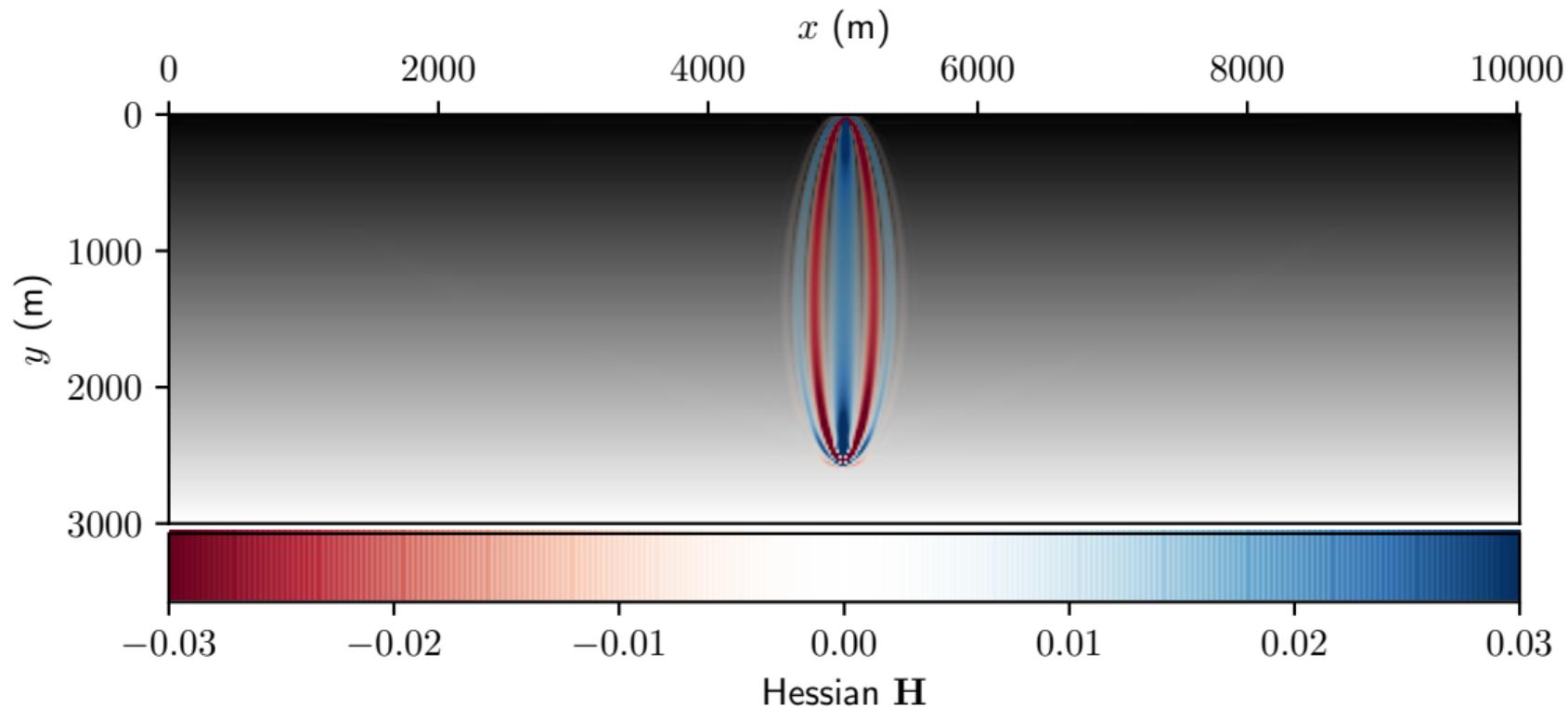
# Gradient Hessian shot 1, perturbation at 1500 m



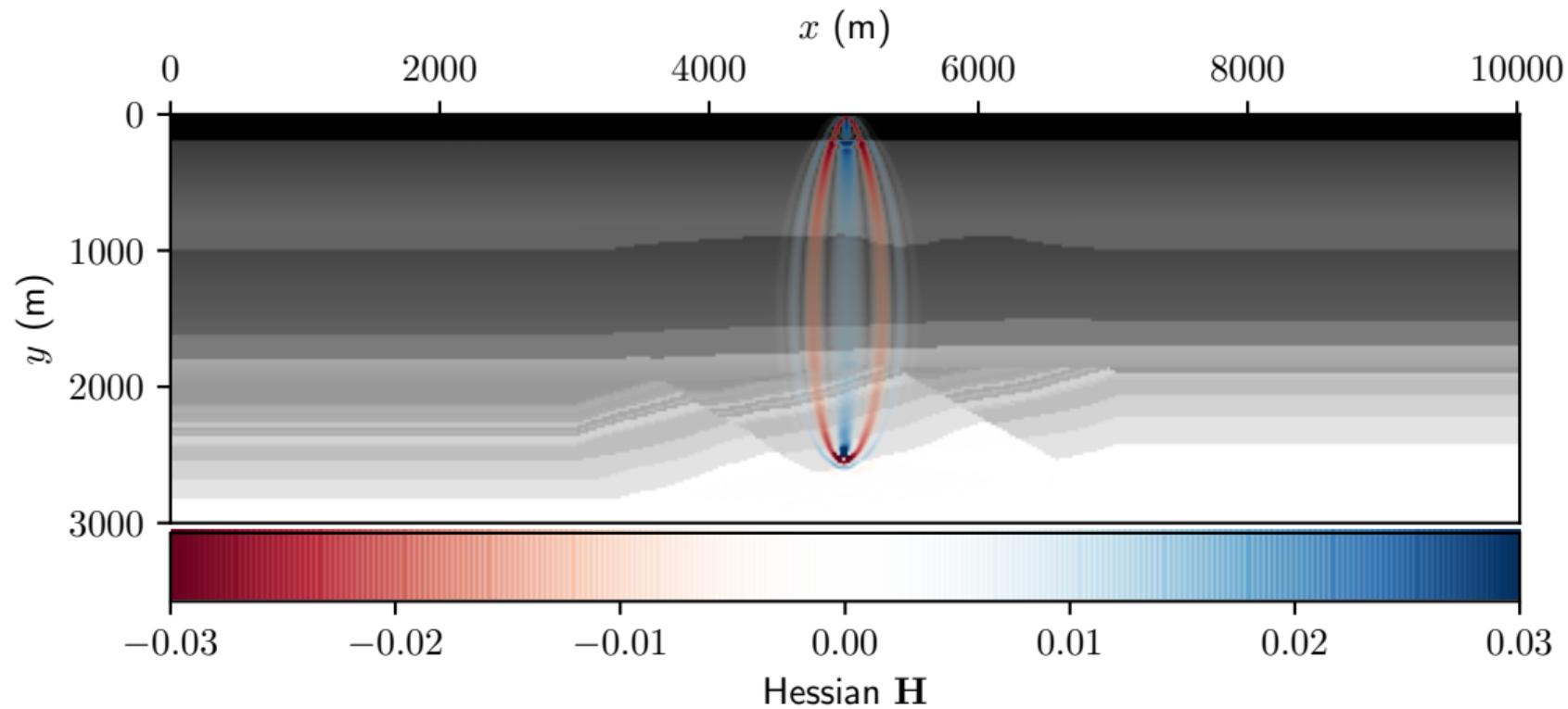
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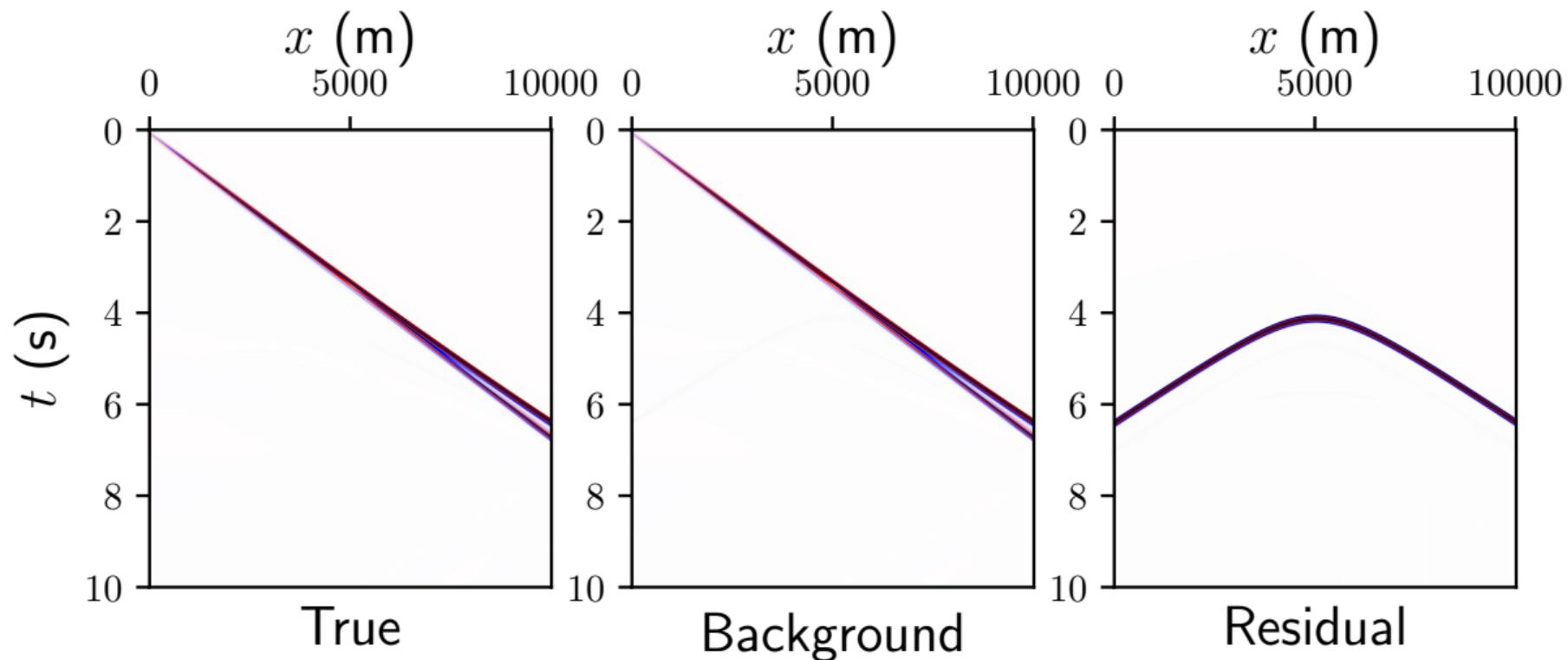
# Gradient Hessian shot 1, perturbation at 2500 m



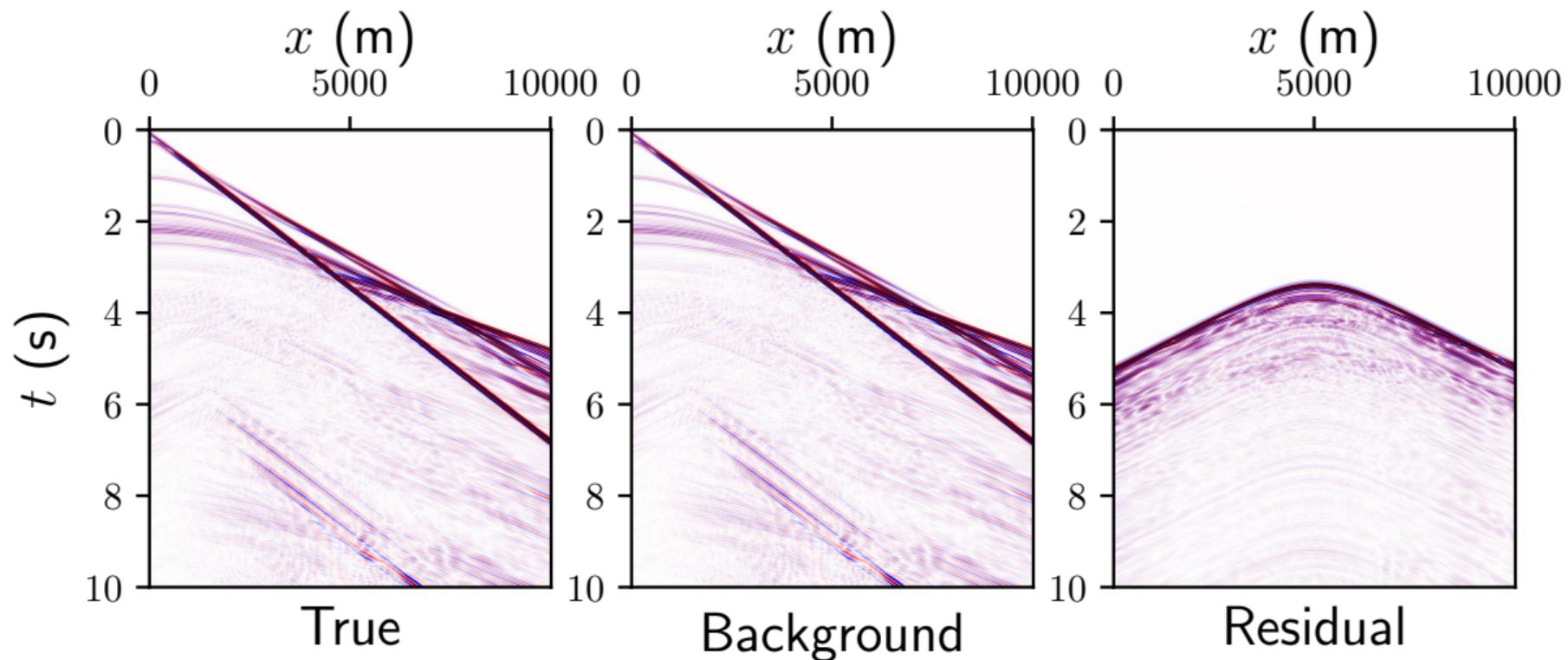
# Gulfaks Hessian shot 1, perturbation at 2500 m



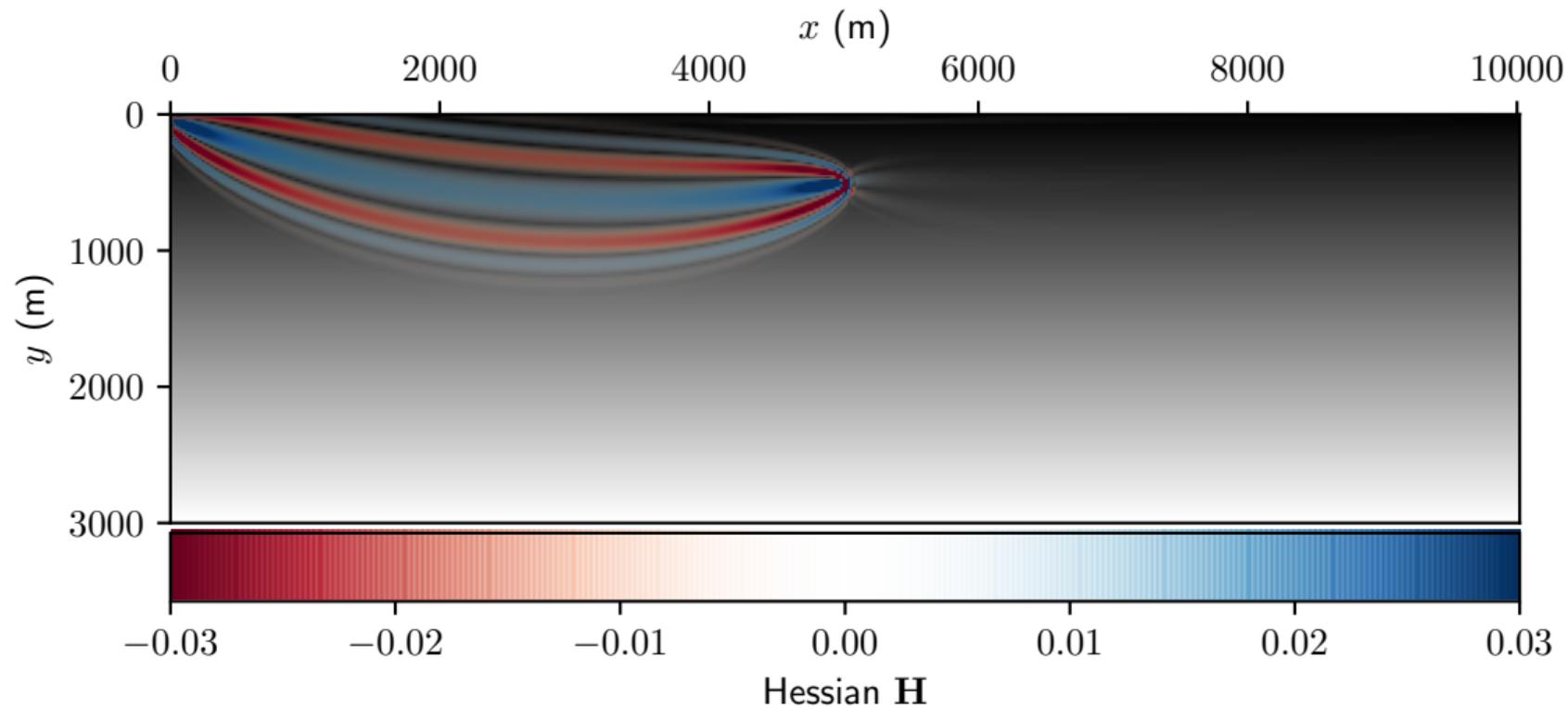
# Gradient recording shot 2, perturbation at 1500 m



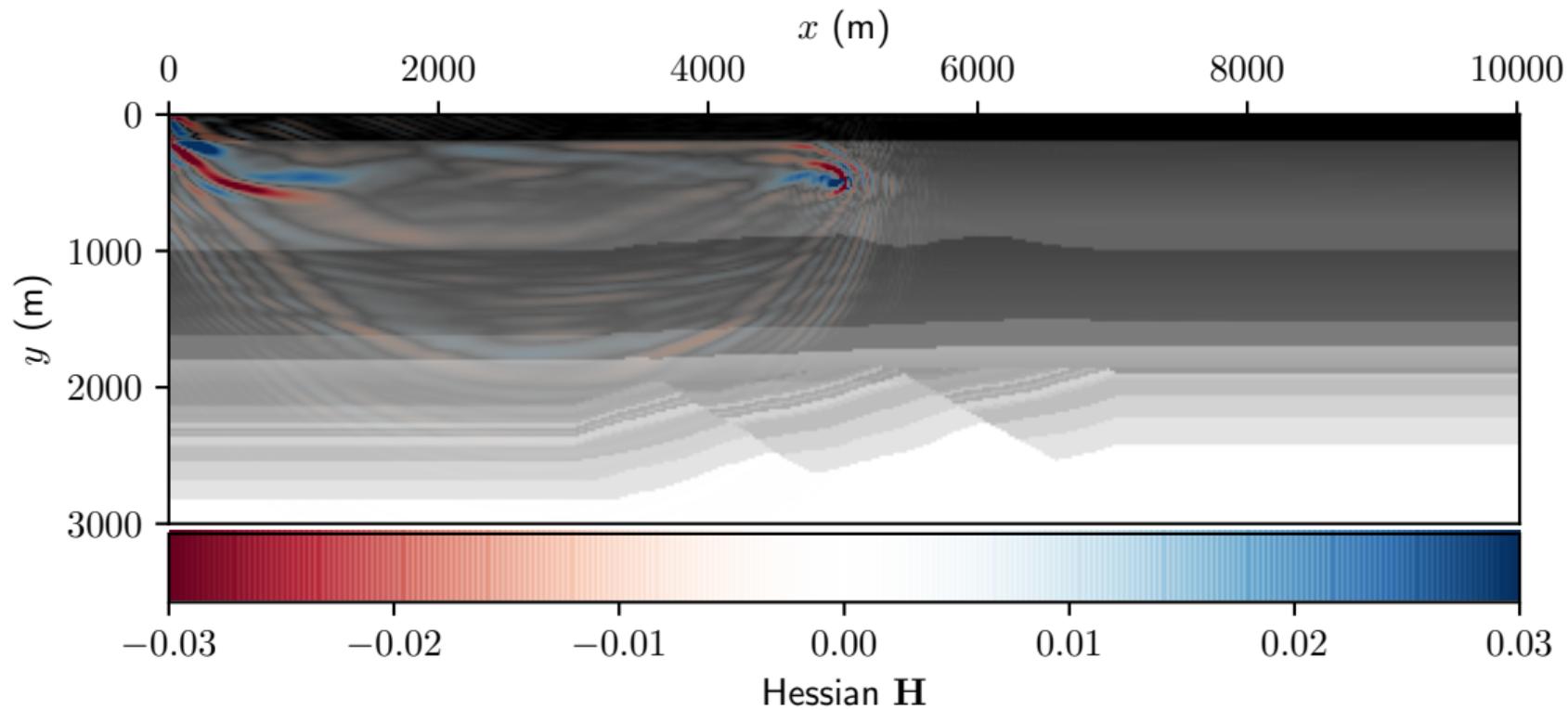
## Gulfaks recording shot 2, perturbation at 1500 m



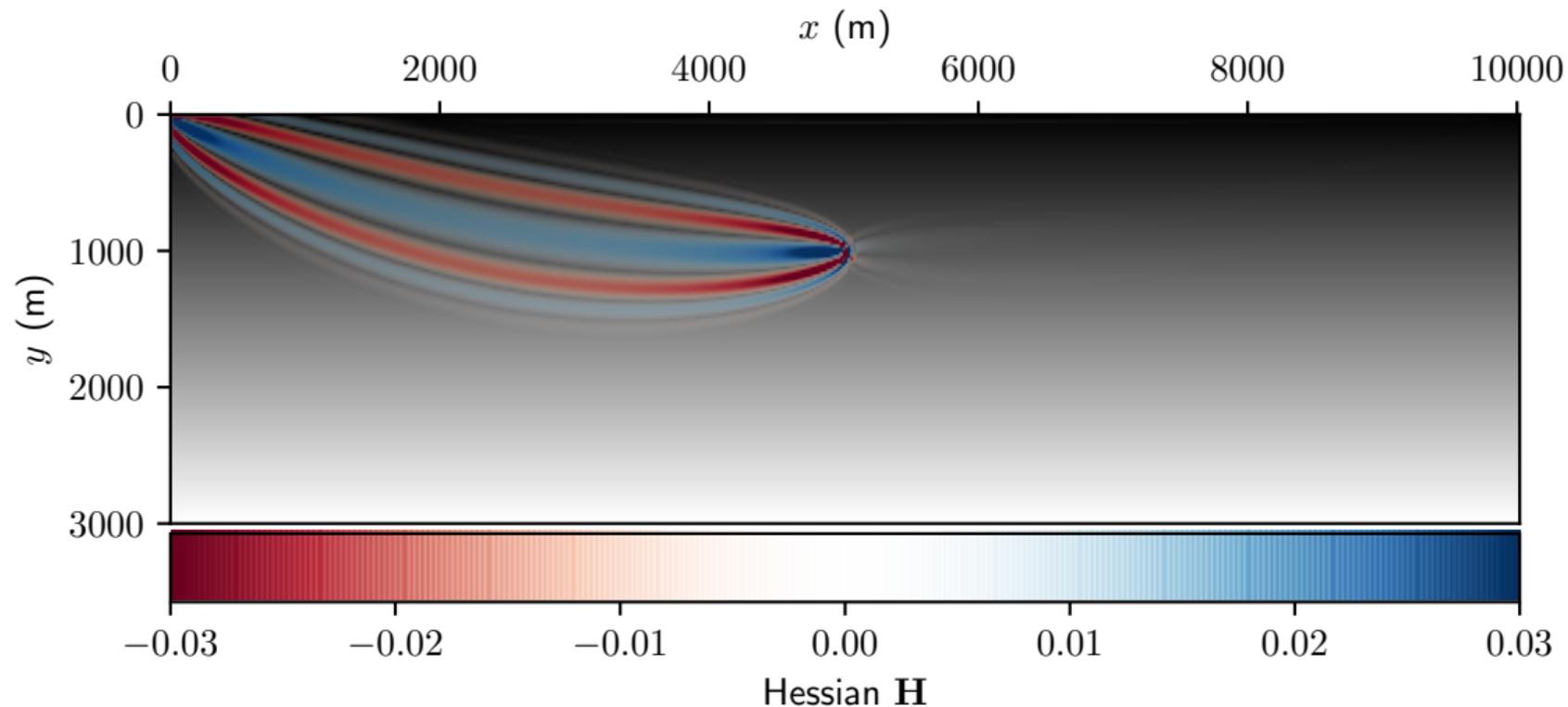
## Gradient Hessian shot 2, perturbation at 500 m



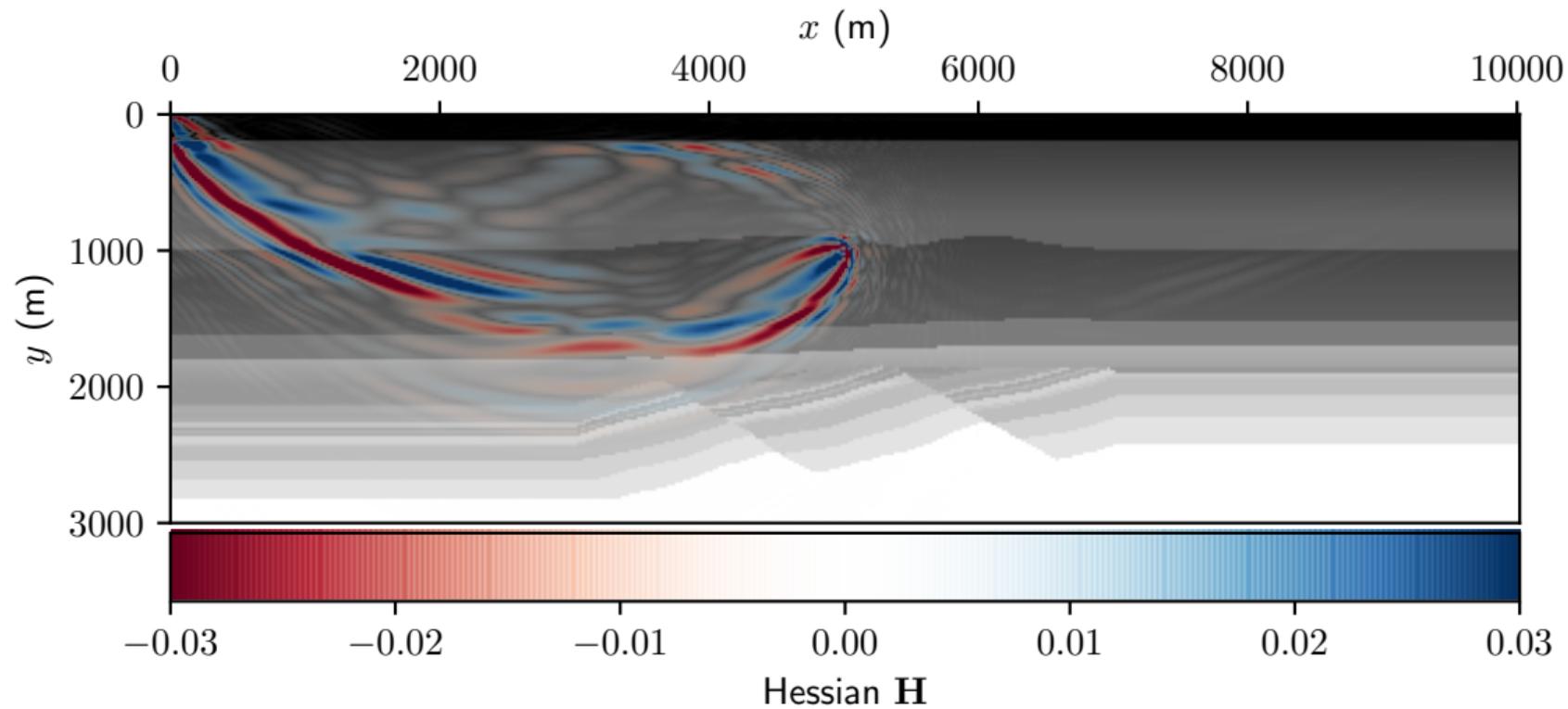
# Gulfaks Hessian shot 2, perturbation at 500 m



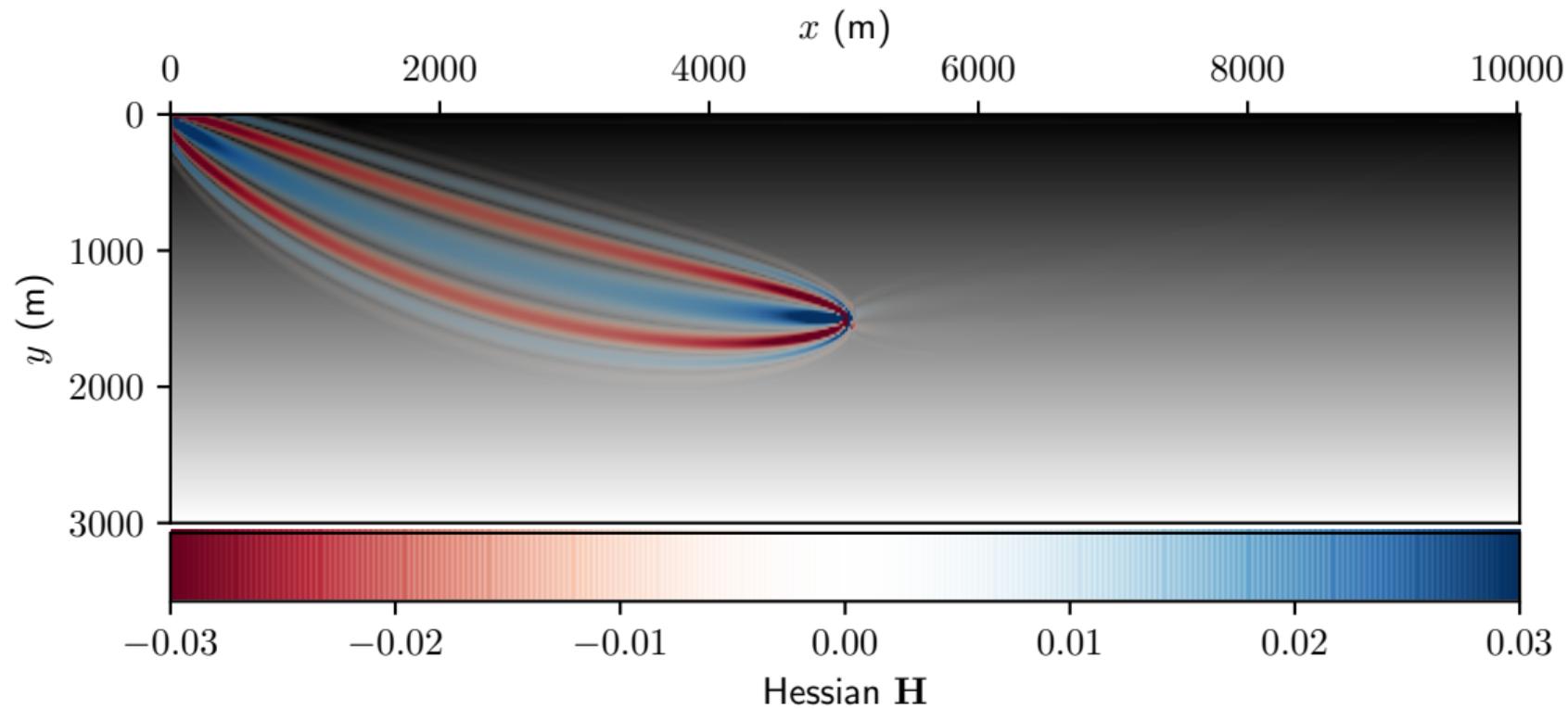
## Gradient Hessian shot 2, perturbation at 1000 m



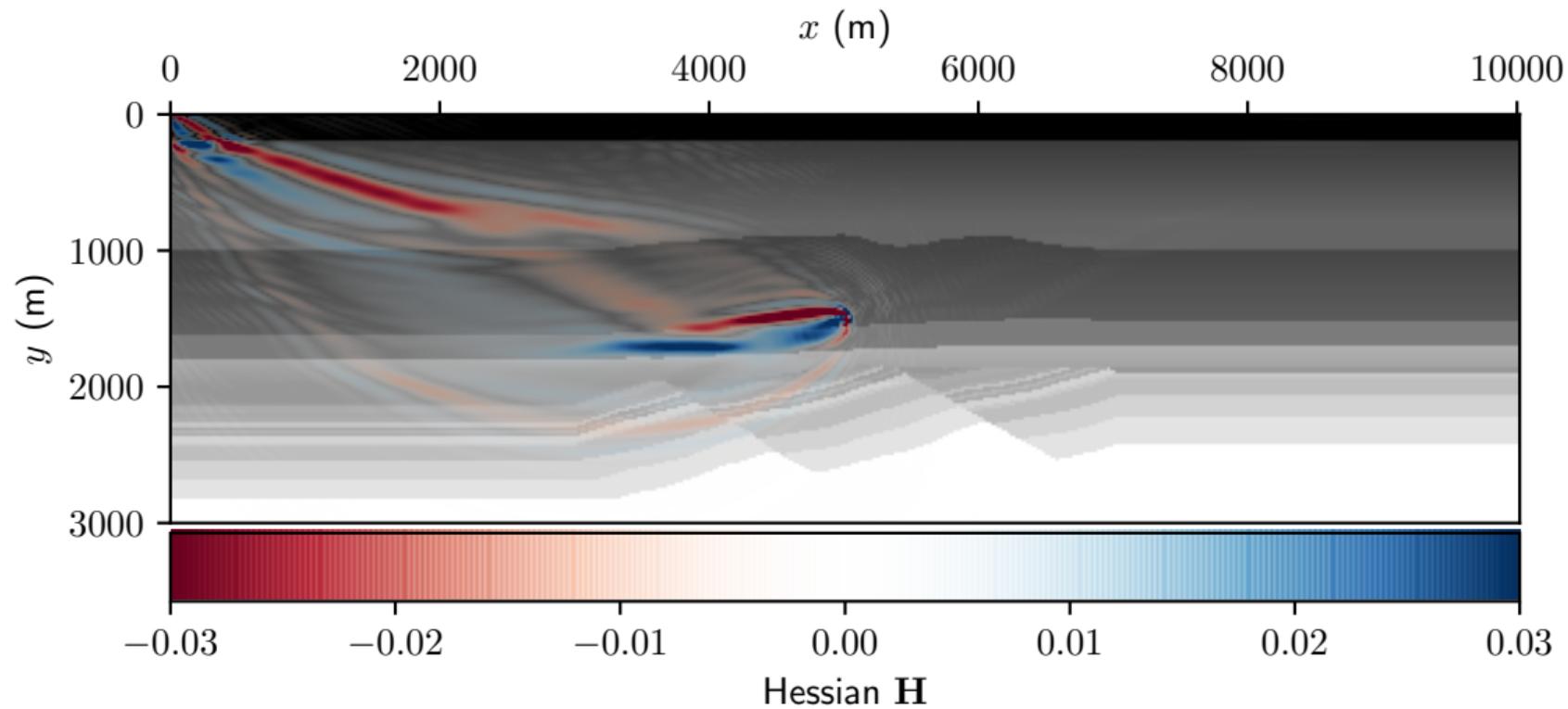
# Gulfaks Hessian shot 2, perturbation at 1000 m



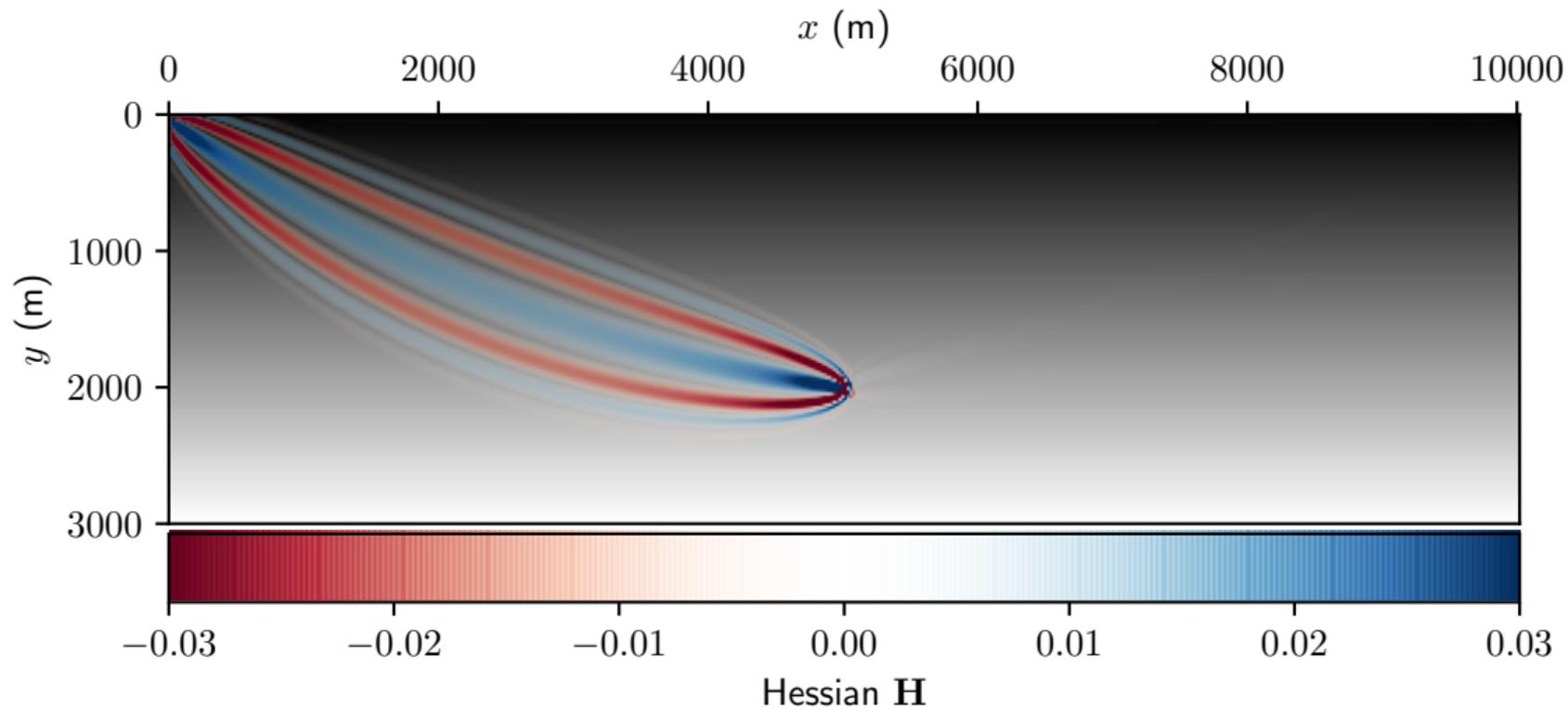
## Gradient Hessian shot 2, perturbation at 1500 m



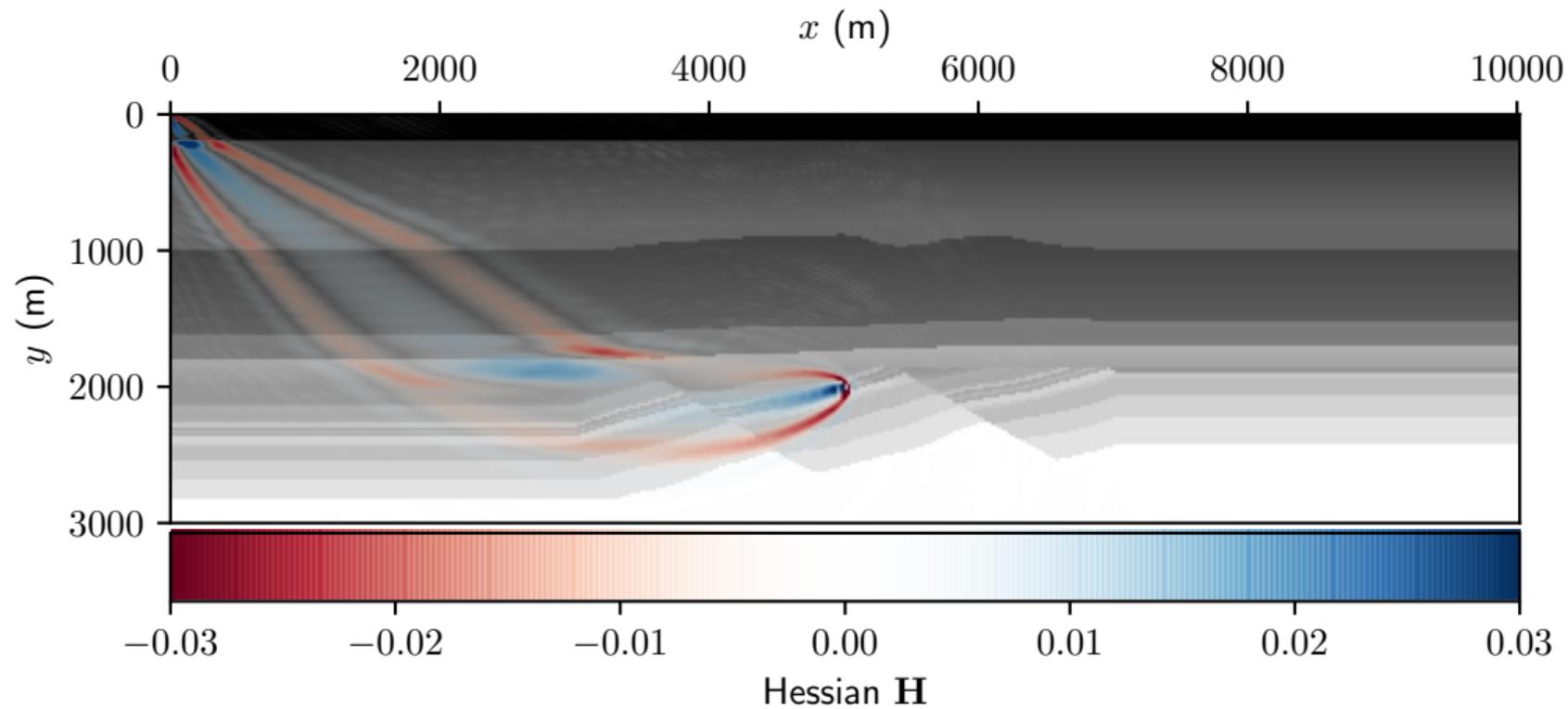
# Gulfaks Hessian shot 2, perturbation at 1500 m



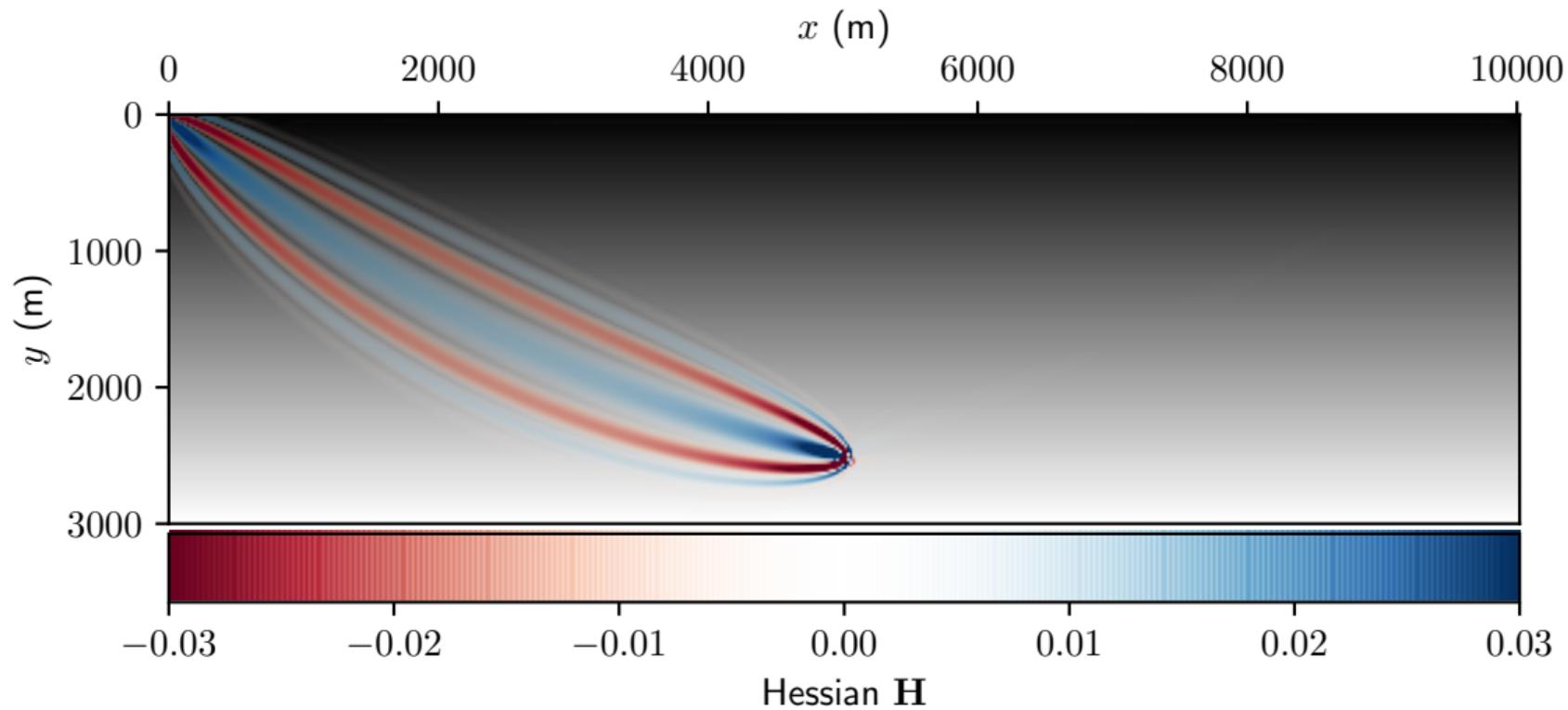
## Gradient Hessian shot 2, perturbation at 2000 m



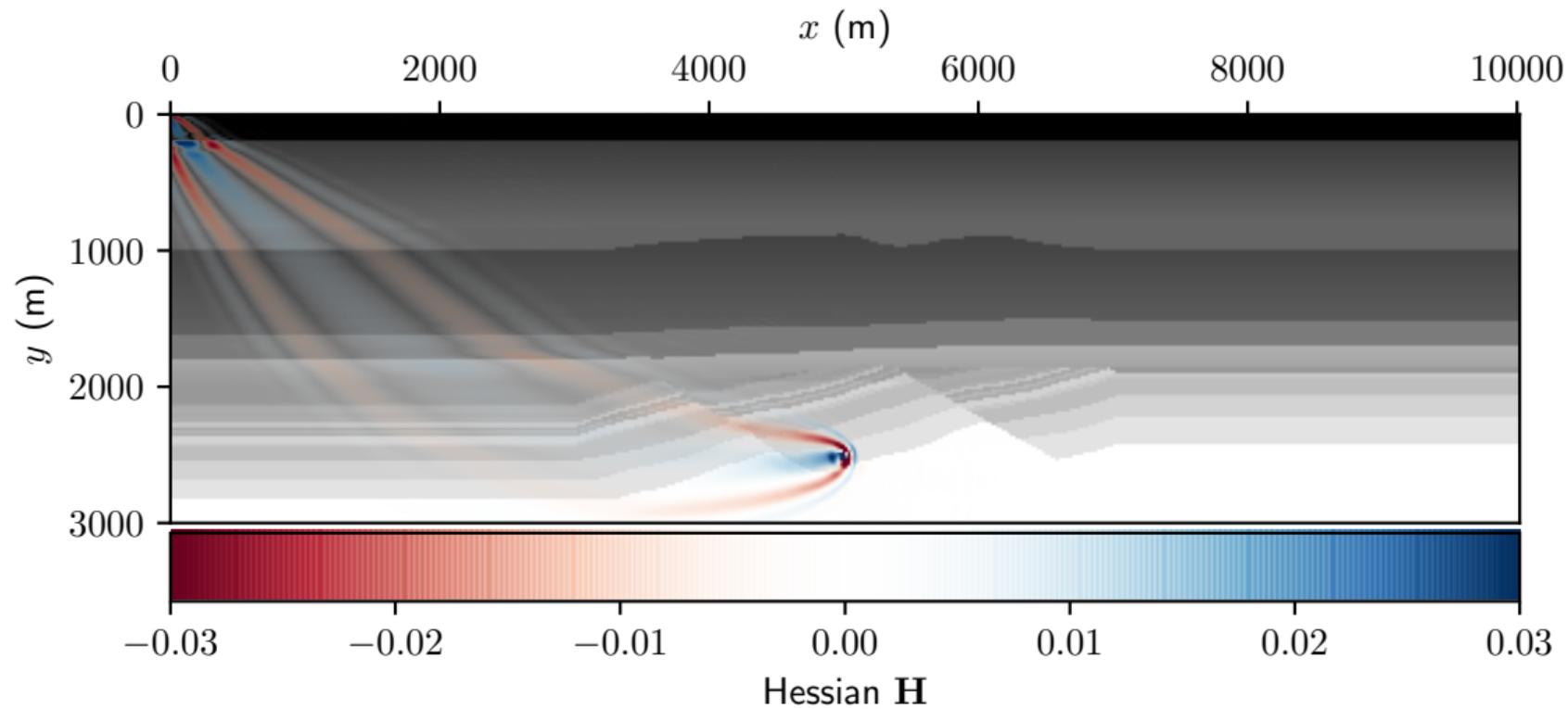
# Gulfaks Hessian shot 2, perturbation at 2000 m



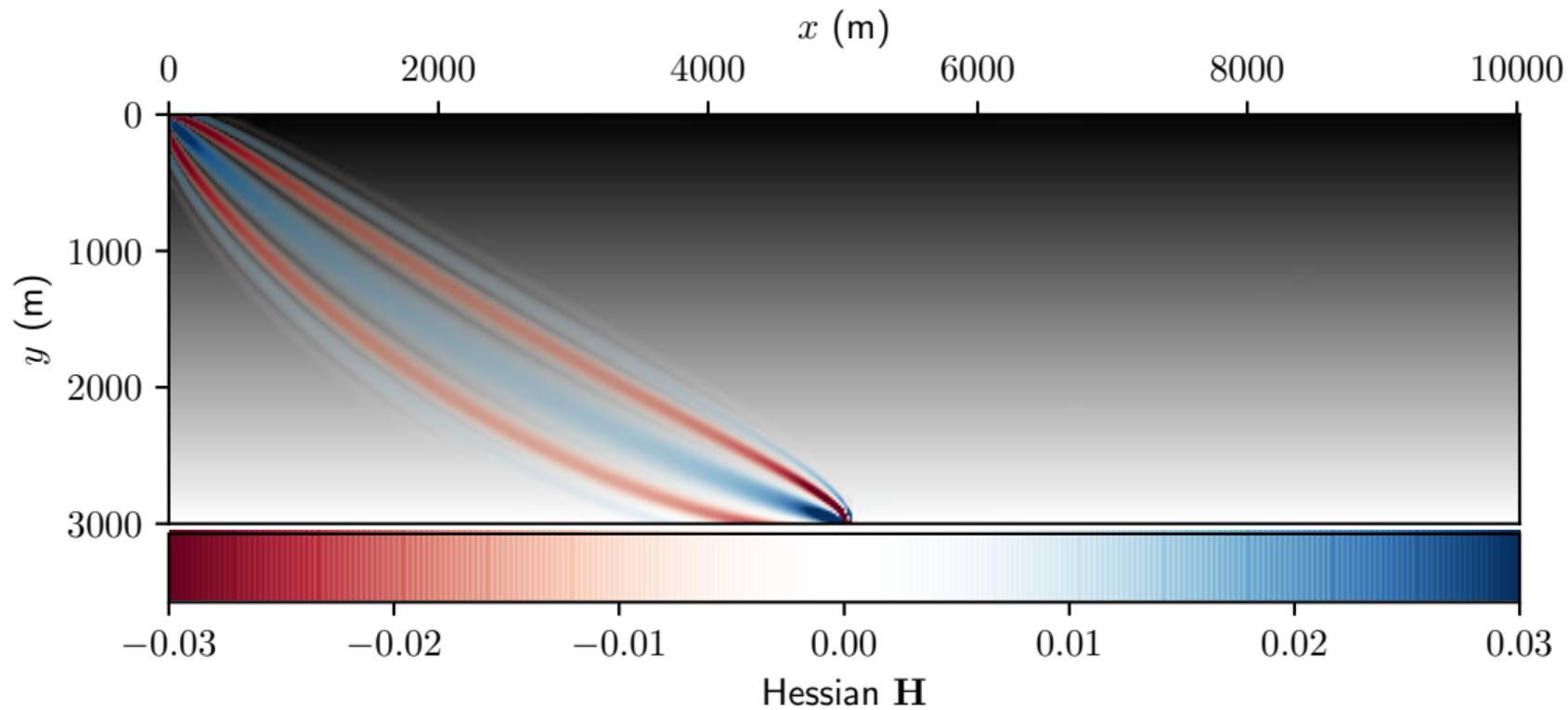
## Gradient Hessian shot 2, perturbation at 2500 m



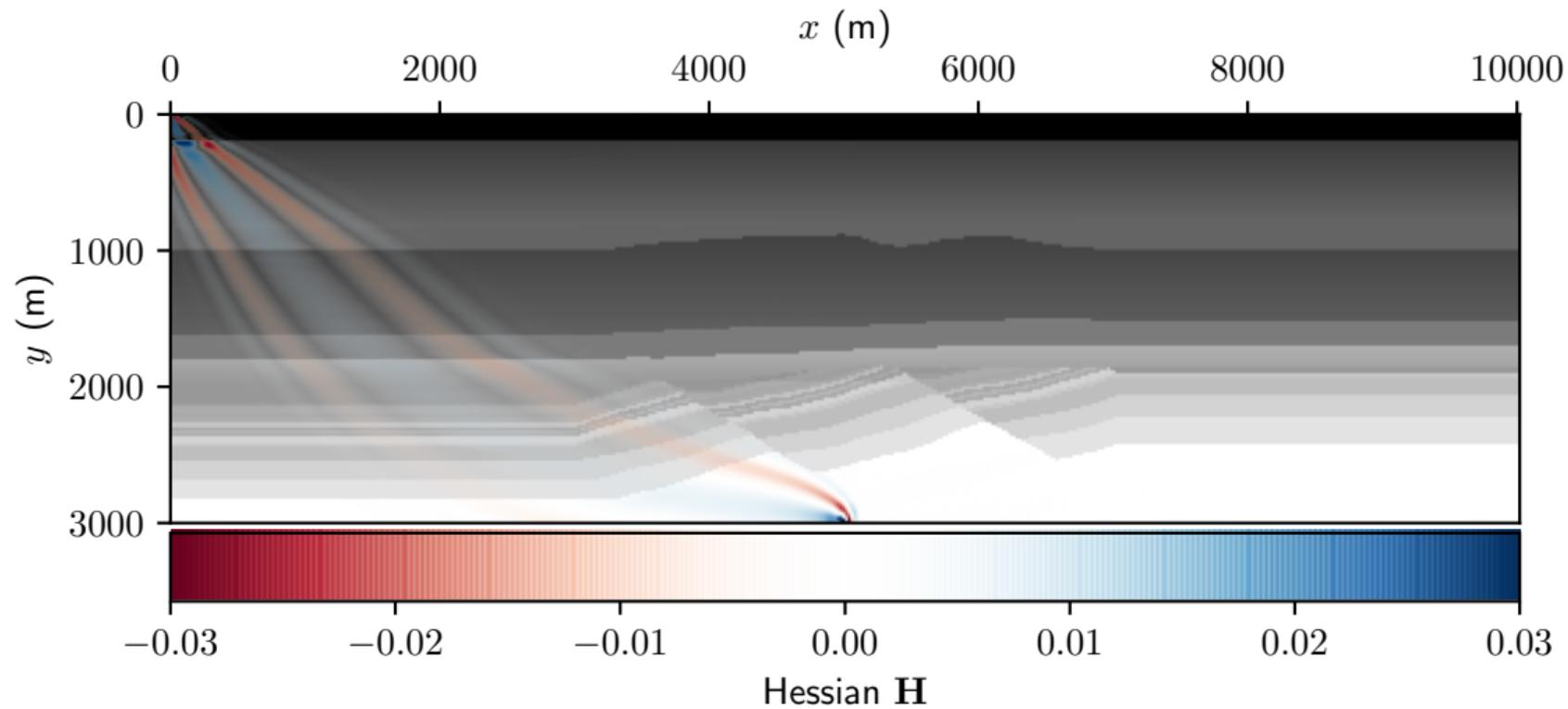
# Gulfaks Hessian shot 2, perturbation at 2500 m



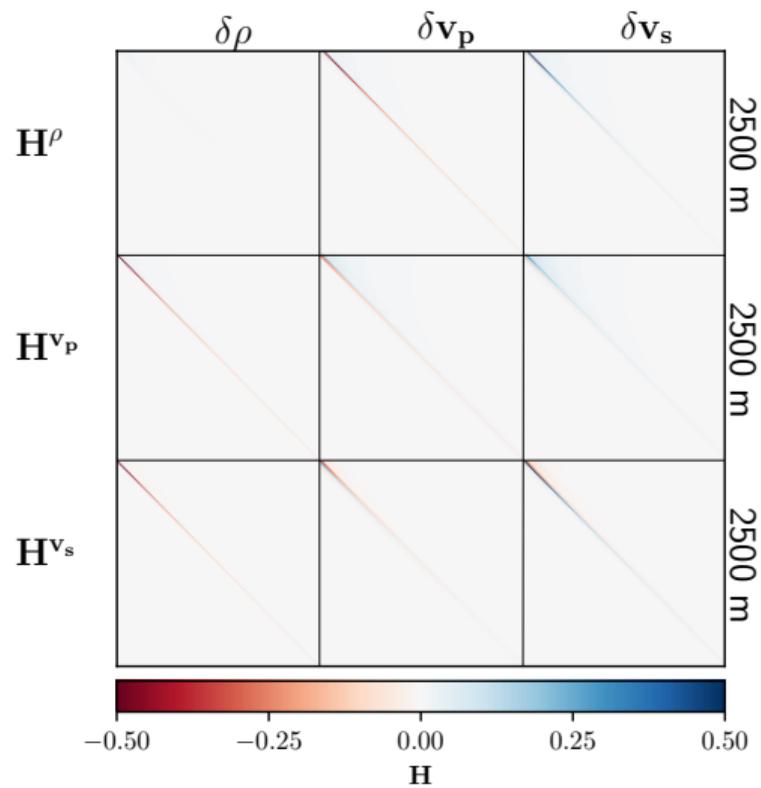
## Gradient Hessian shot 2, perturbation at 3000 m



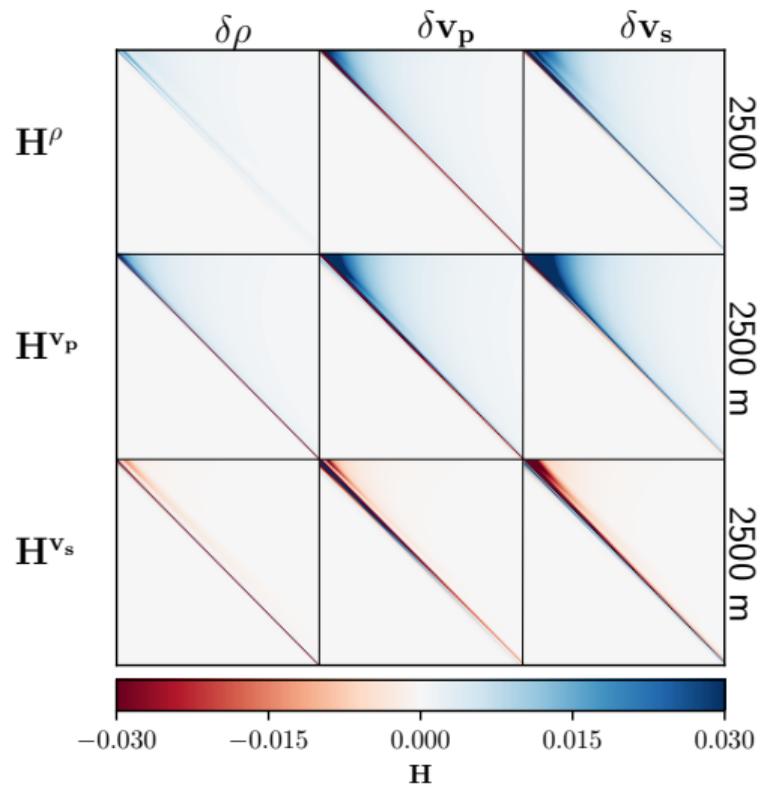
# Gulfaks Hessian shot 2, perturbation at 3000 m



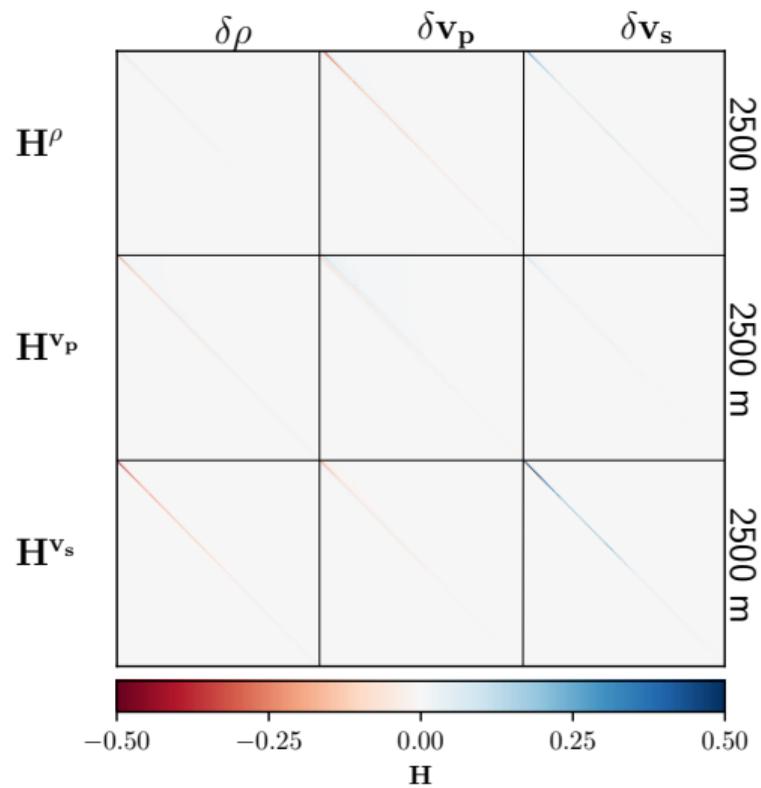
# Gradient model Hessian constructed from shot 1



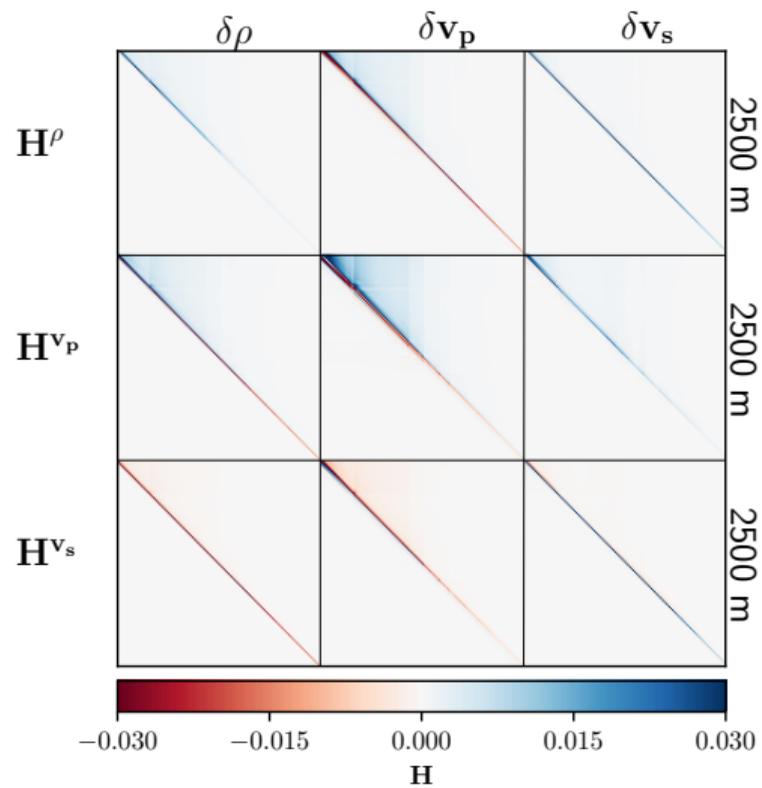
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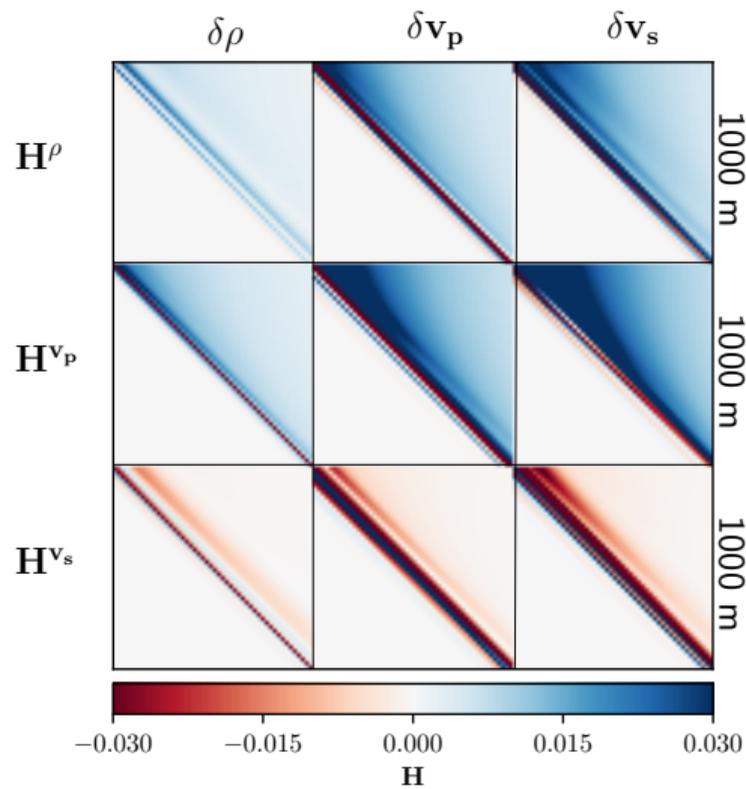
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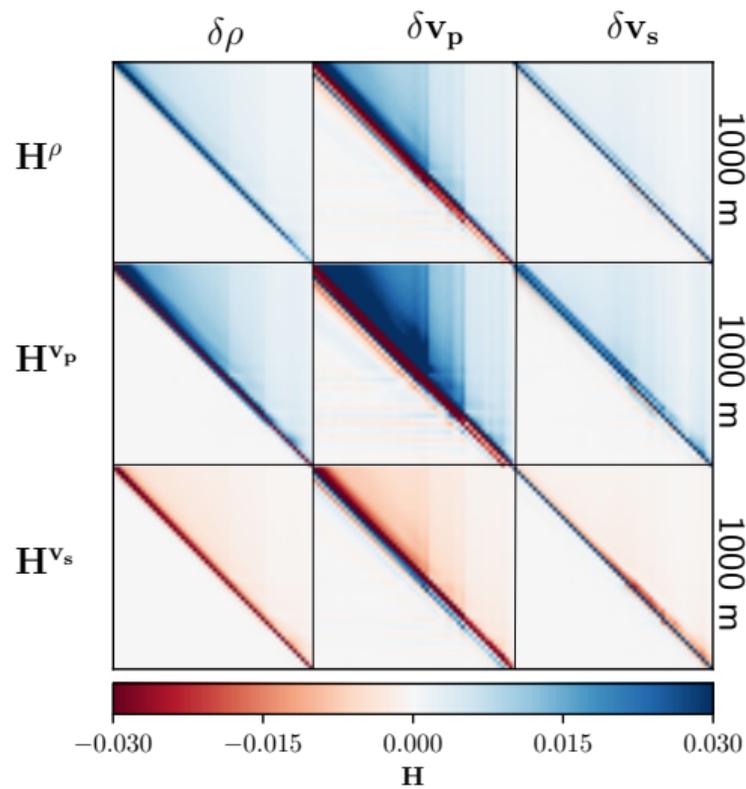
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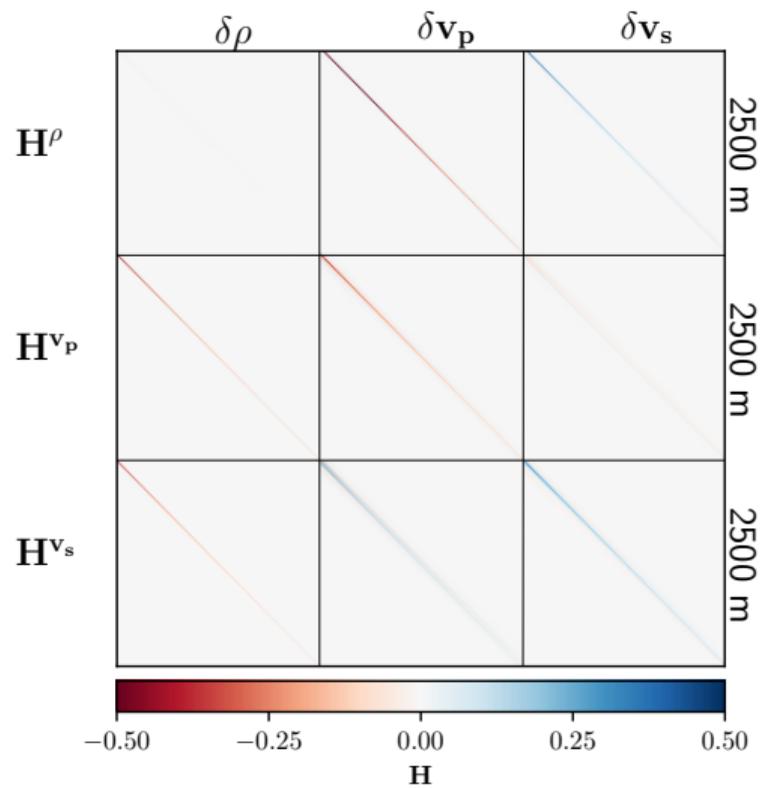
# Gradient model Hessian constructed from shot 1 – 1km-2km zoom



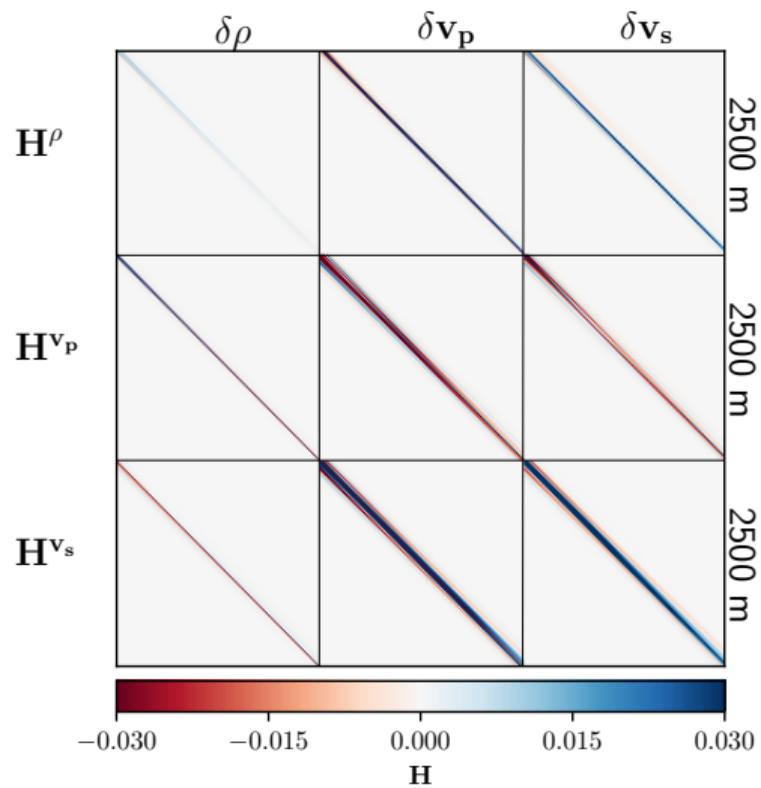
# Gulfaks model Hessian constructed from shot 1 – 1km-2km zoom



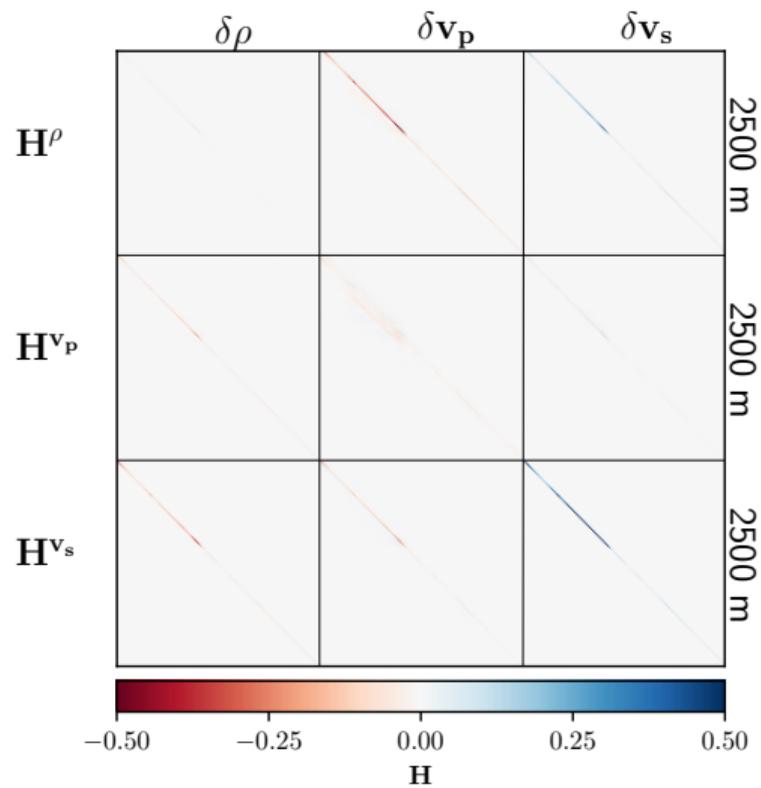
# Gradient model Hessian constructed from shot 2



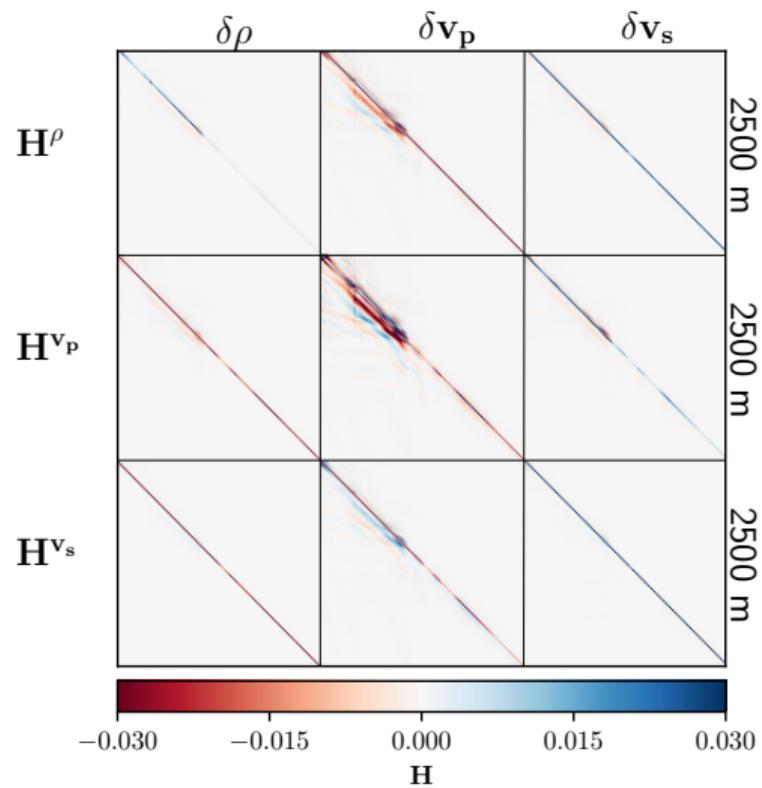
# Gradient model Hessian constructed from shot 2



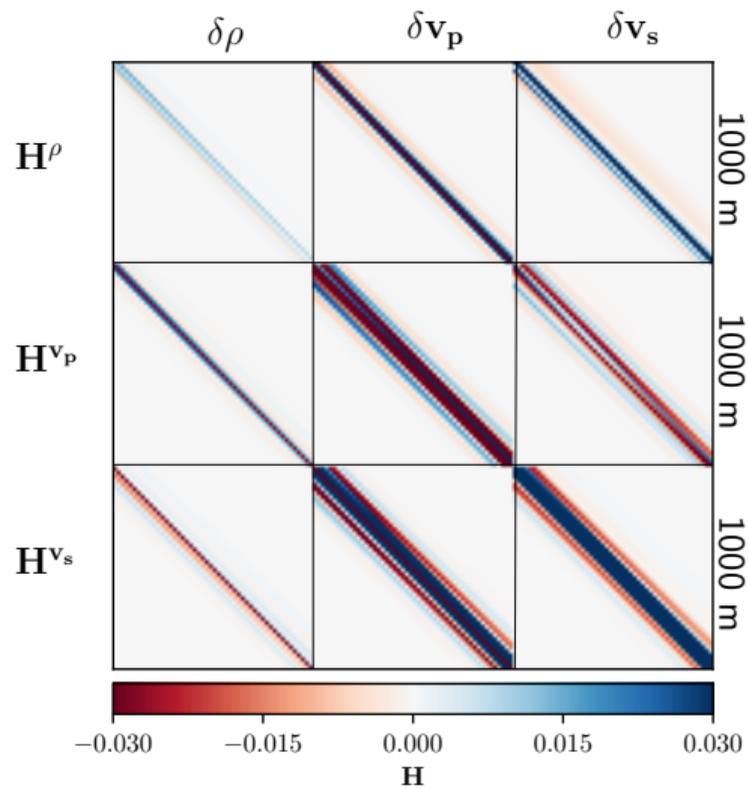
# Gulfaks model Hessian constructed from shot 2



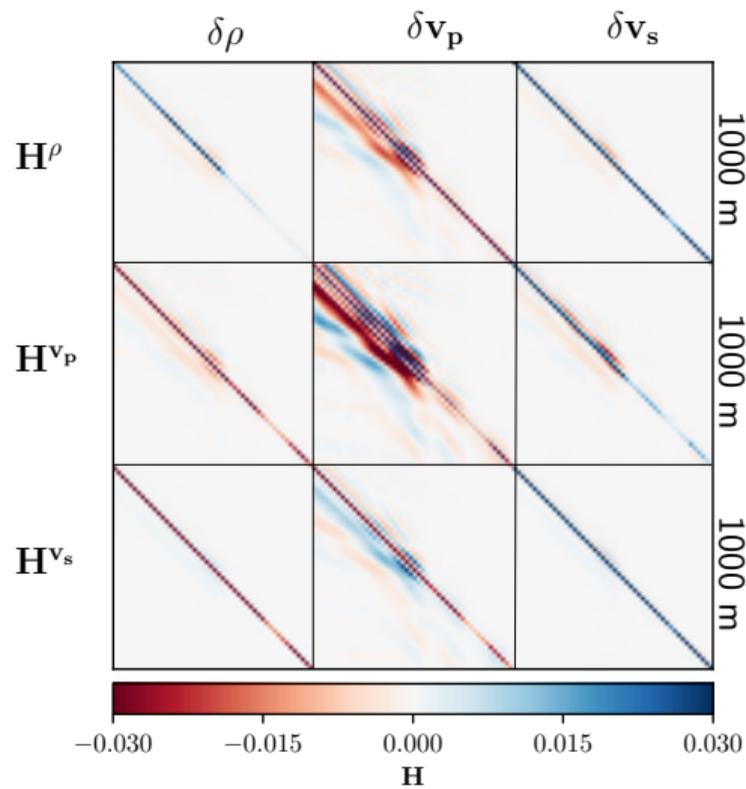
# Gulfaks model Hessian constructed from shot 2



# Gradient model Hessian constructed from shot 2 – 1km-2km zoom



# Gulfaks model Hessian constructed from shot 2 – 1km-2km zoom



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- ▶ We are able to construct the Hessian in a constrained area.
- ▶ Still computationally expensive.
- ▶ Illustrates the influence zone.
- ▶ From the constructed Hessian we can see a strong cross-talk between parameters.
- ▶ Low recovery of density in the given geometries.

## Future work

- ▶ Explore different shot-receiver geometries.
  - ▶ Sum over shots.

## Future work

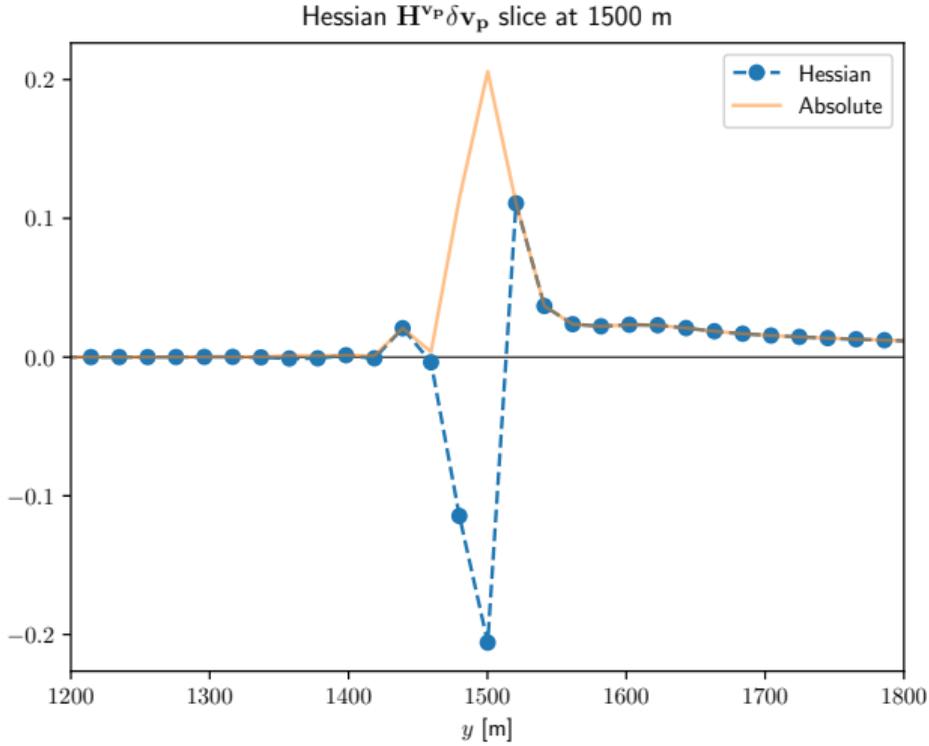
- ▶ Explore different shot-receiver geometries.
  - ▶ Sum over shots.
- ▶ Implement a full-Newton solver.

# References

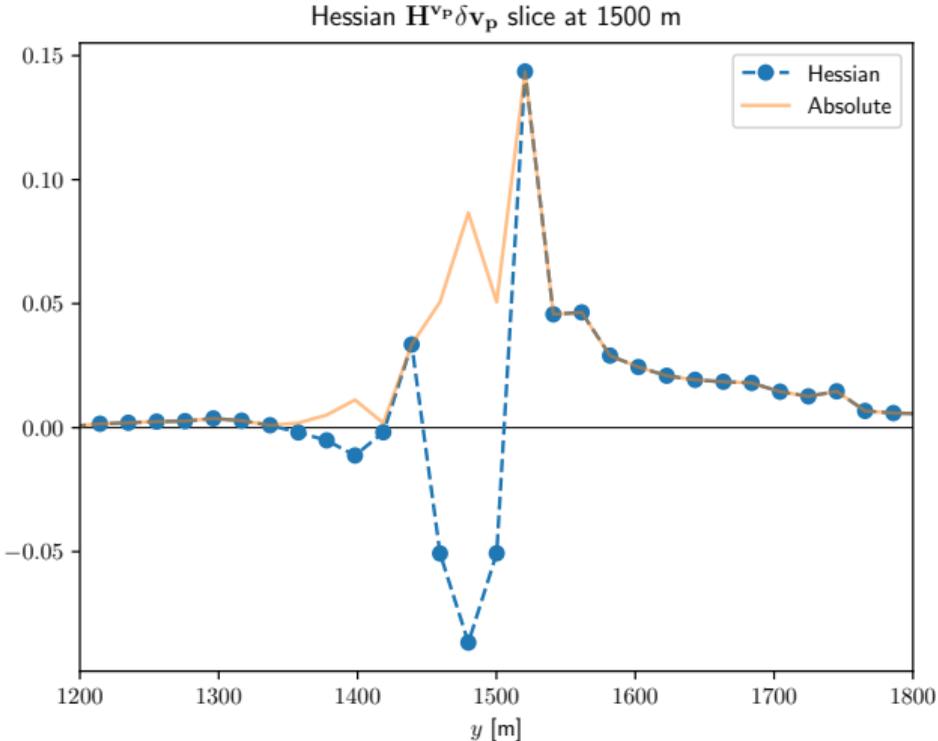
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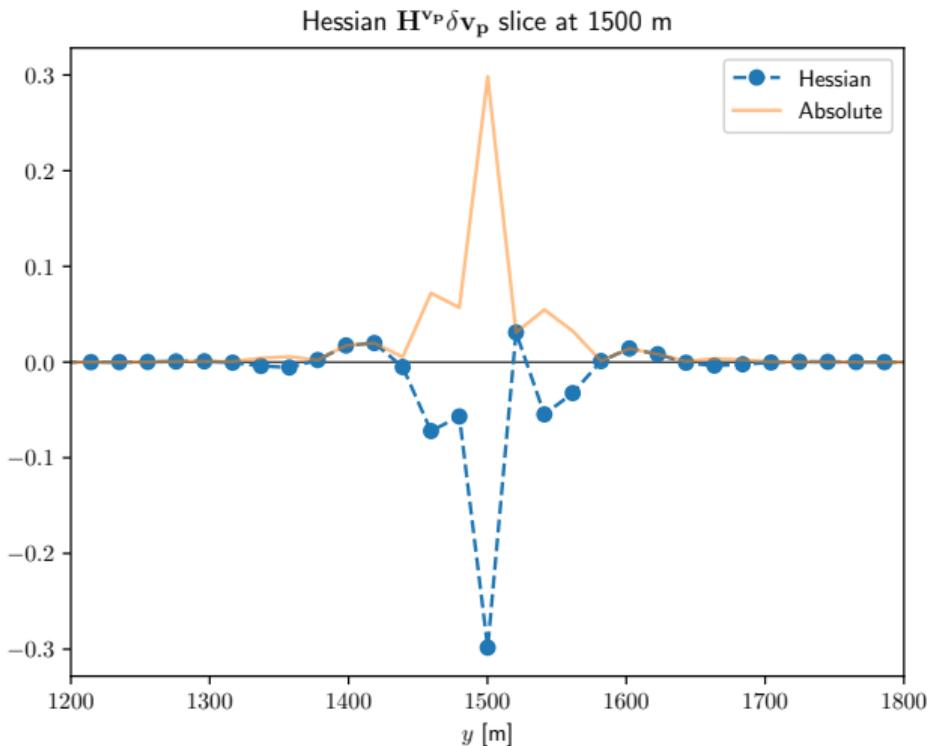
# Gradient model Hessian slice shot 1 – 1500m



# Gulfaks model Hessian slice shot 1 – 1500m



# Gradient model Hessian slice shot 2 – 1500m



# Gulfaks model Hessian slice shot 2 – 1500m

