

Estimation of rock physics properties via FWI of VSP data recorded by accelerometer and fiberoptic sensors

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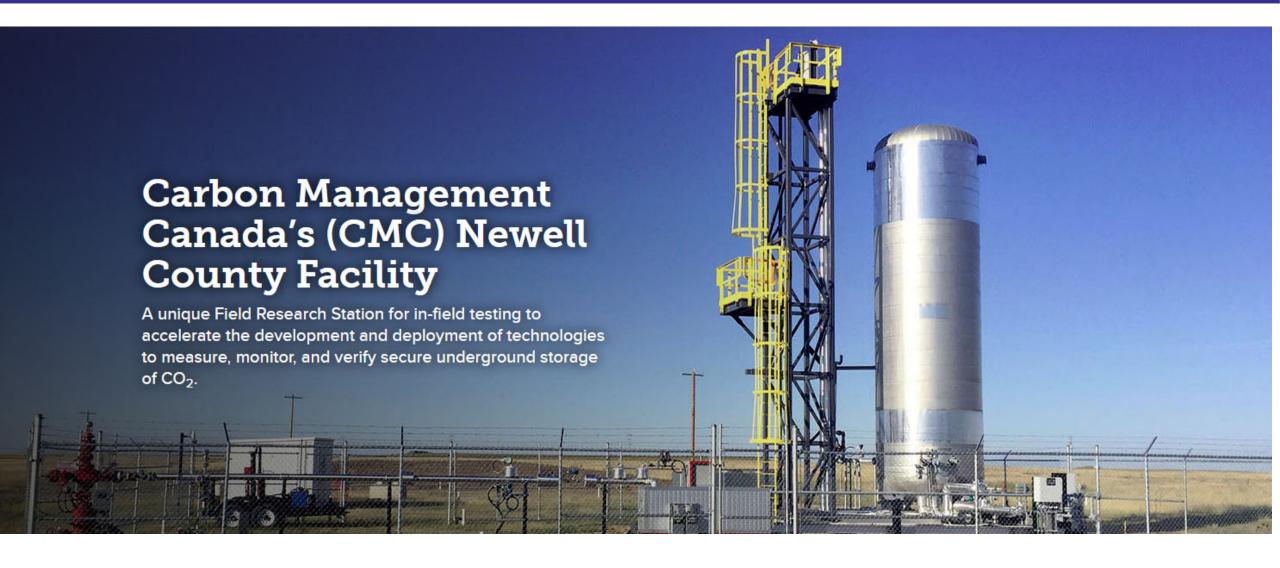


Background

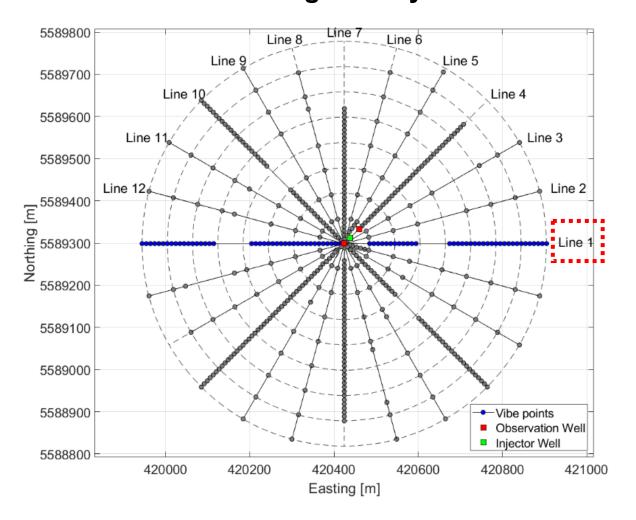
Elastic FWI

Rock physics inversion

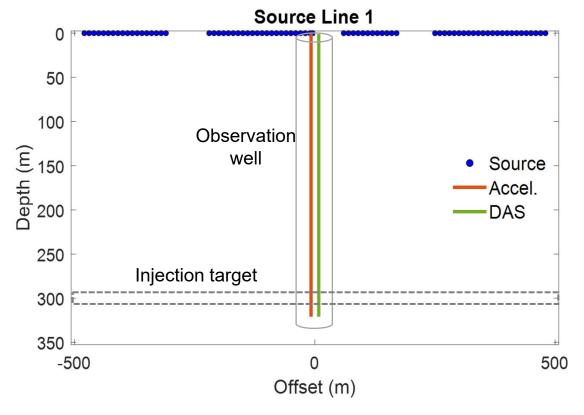




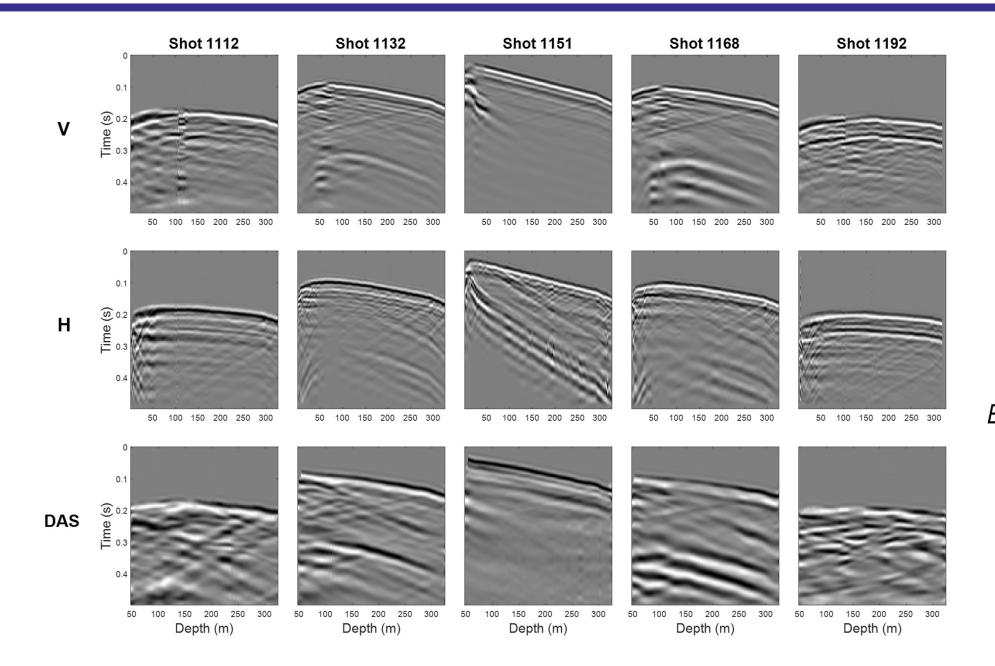
Shot geometry



Section view

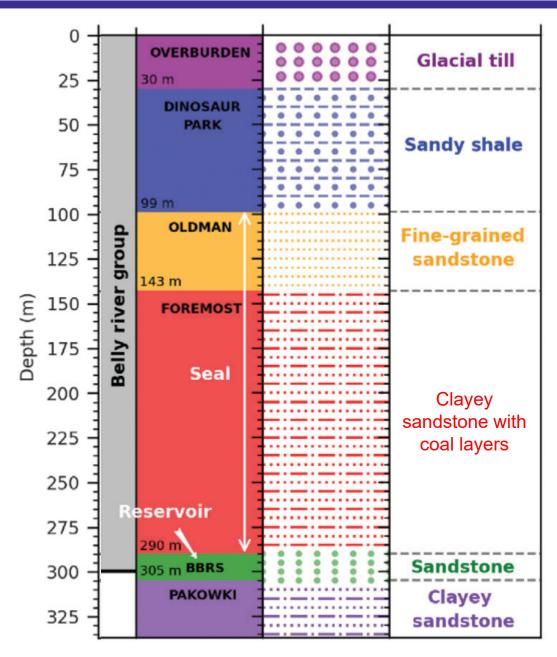


Processed seismic data



Eaid (2021)

Geology Geology



BBRS core (296-298 m)



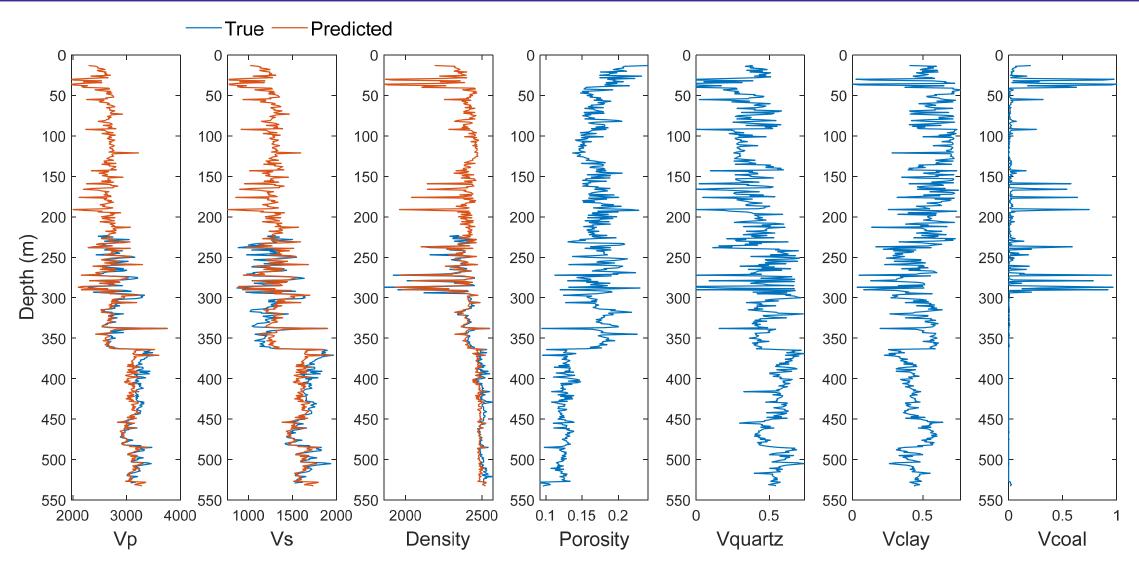
Caprock (288-290 m)



Macquet et al. (2022)

Pan et al (2023)

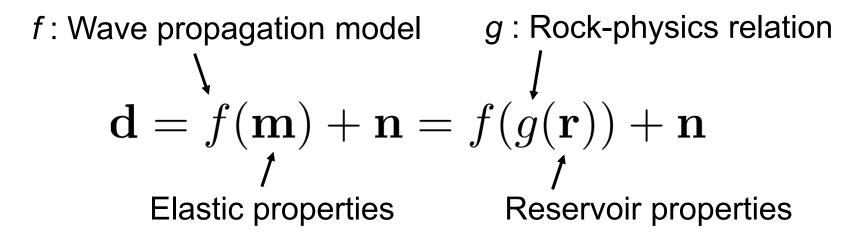
Well-log data

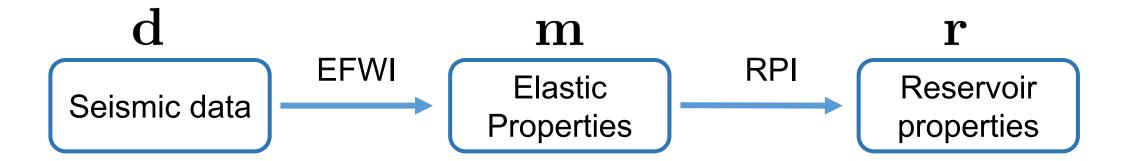


$$(V_{\mathrm{P}}, V_{\mathrm{S}}, \rho) = g(\phi, V_{\mathrm{qu}}, V_{\mathrm{cl}})$$



General problem







Background

Elastic FWI

Rock physics inversion

Constrained optimization problem

$$\min_{\mathbf{m}} E = \frac{1}{2} \|\mathbf{R}\mathbf{u} - \mathbf{d}\|_2^2 \quad \text{subject to} \quad \mathbf{A}(\mathbf{m})\mathbf{u} = \mathbf{f}(\omega),$$
 Receiver matrix Impedance matrix Displacement wavefield

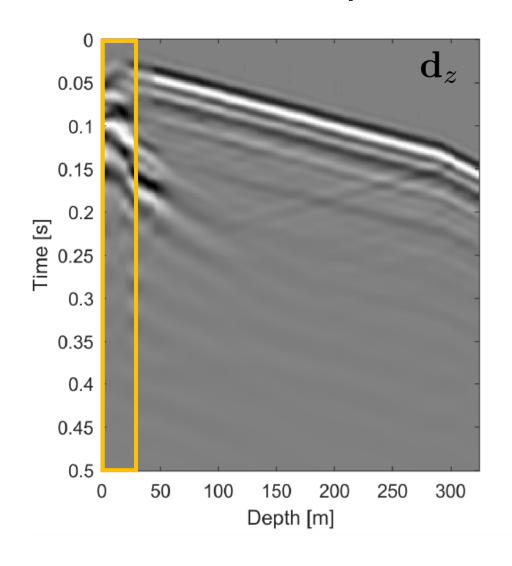
Search direction

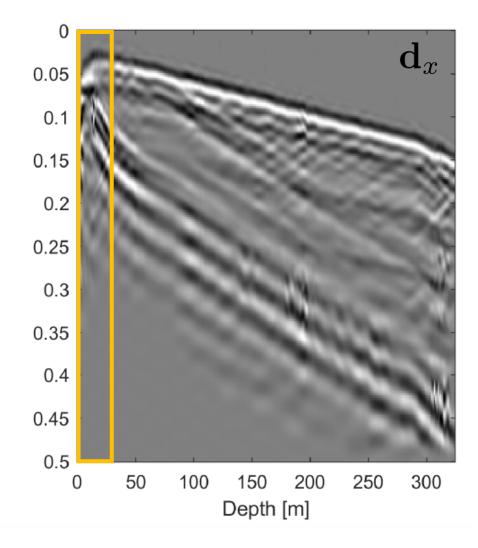
$$\delta \mathbf{m} = -\mathbf{H}^{-1} \ \nabla_{\mathbf{m}} E,$$
Hessian Gradient

rch direction
$$\delta \mathbf{m} = -\mathbf{H}^{-1} \; \nabla_{\mathbf{m}} E, \qquad \text{where} \qquad \nabla_{m_i} E = \Re \langle \frac{\partial \mathbf{A}}{\partial m_i} \mathbf{u} \; , \lambda \; \rangle.$$



Near-surface complications







• Effective source approach (Keating, 2021)

$$\min_{\mathbf{m}^*, \mathbf{f}^*} E = \frac{1}{2} \|\mathbf{R}^* \mathbf{u}^* - \mathbf{d}^*\|_2^2 \quad \text{subject to} \quad \mathbf{A}^* (\mathbf{m}^*) \mathbf{u}^* = \mathbf{f}^*,$$

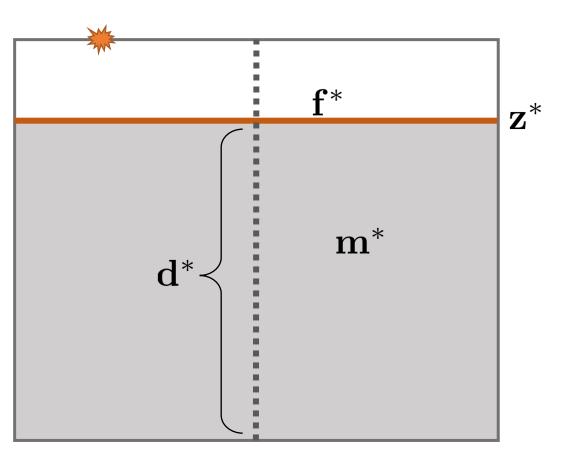


Effective source

Receivers

Near surface

Model domain



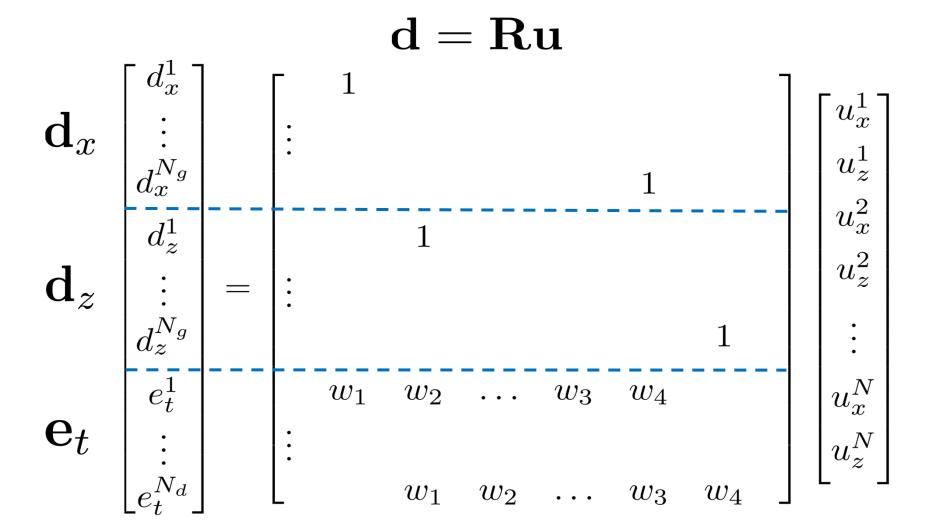


• Incomplete nature of the data

Desirable data qualities for FWI	3C Geophone	DAS
Multicomponent	✓	X
High S/N	✓	X
Strong low frequency content	×	✓
Cost effective dense sampling	×	✓

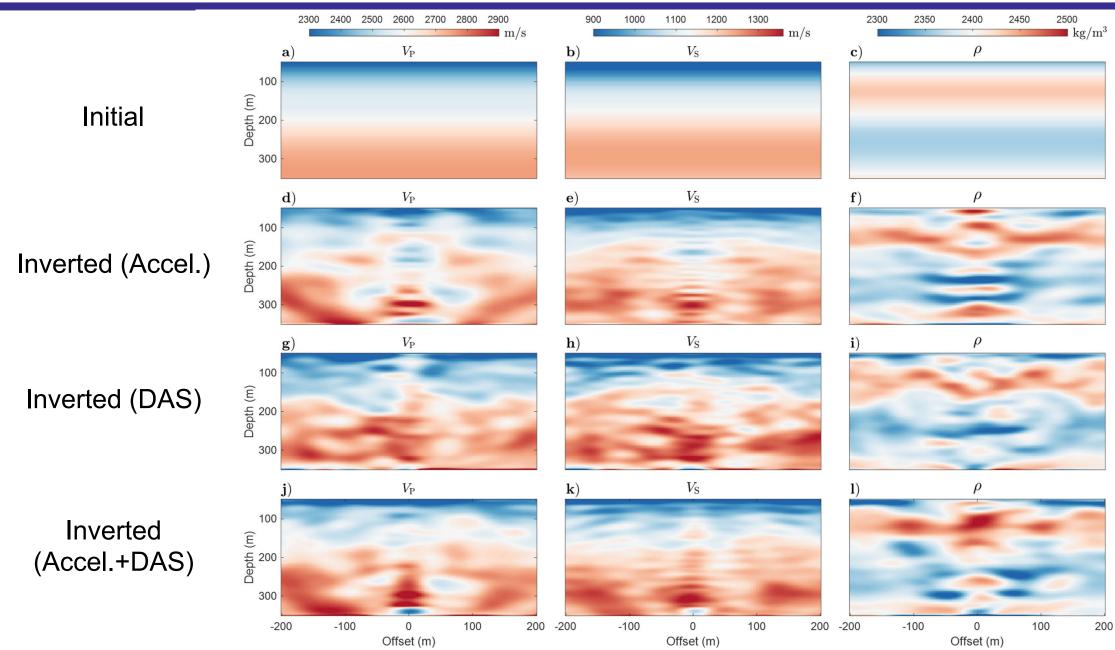


Modeling strategy to include DAS data in FWI (Eaid, 2020)

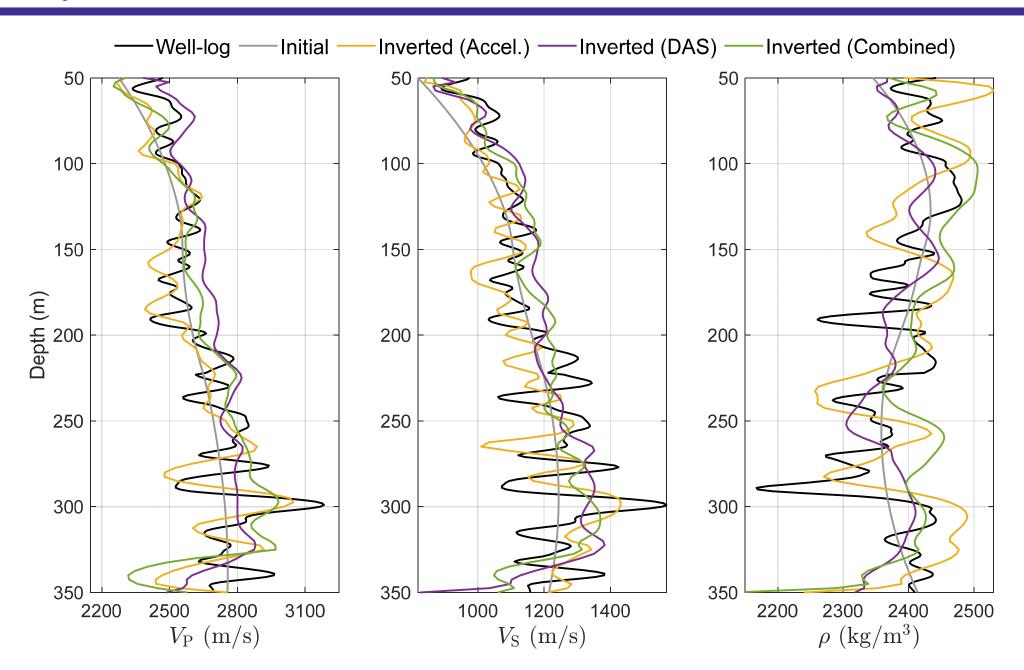




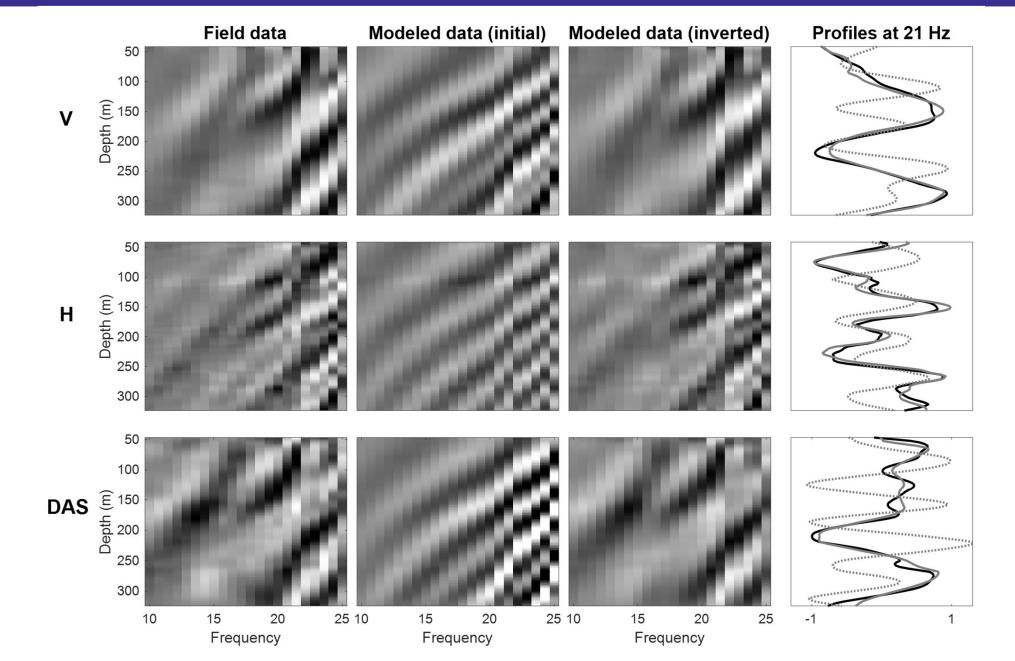
Elastic FWI results



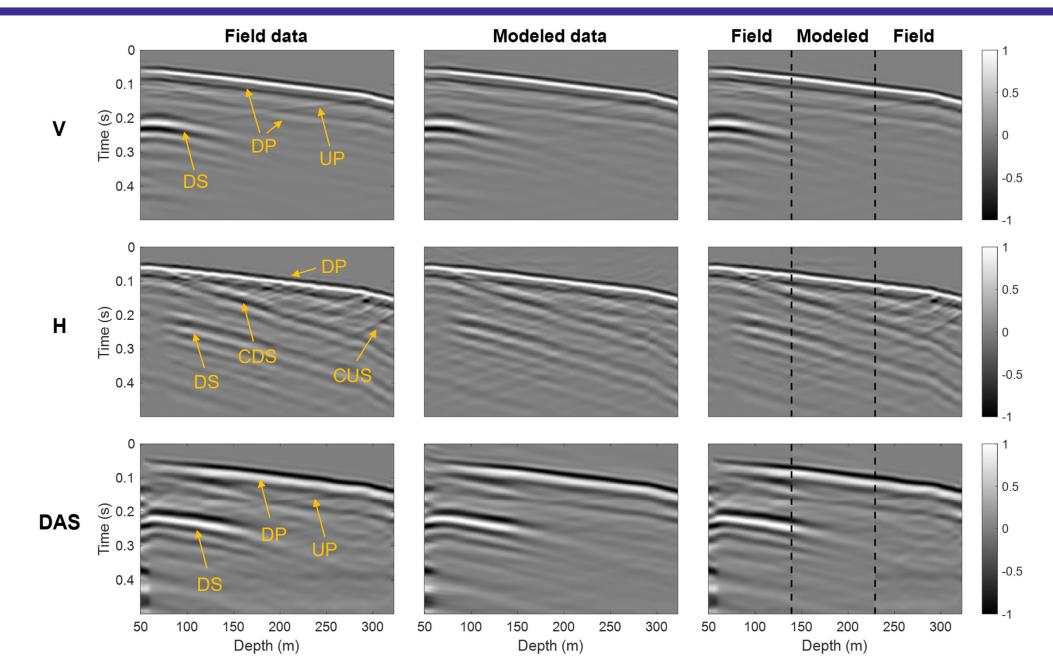
Model profiles at the well location



Data fitting (frequency domain)



Data fitting (time domain)





Background

Elastic FWI

Rock physics inversion



General problem

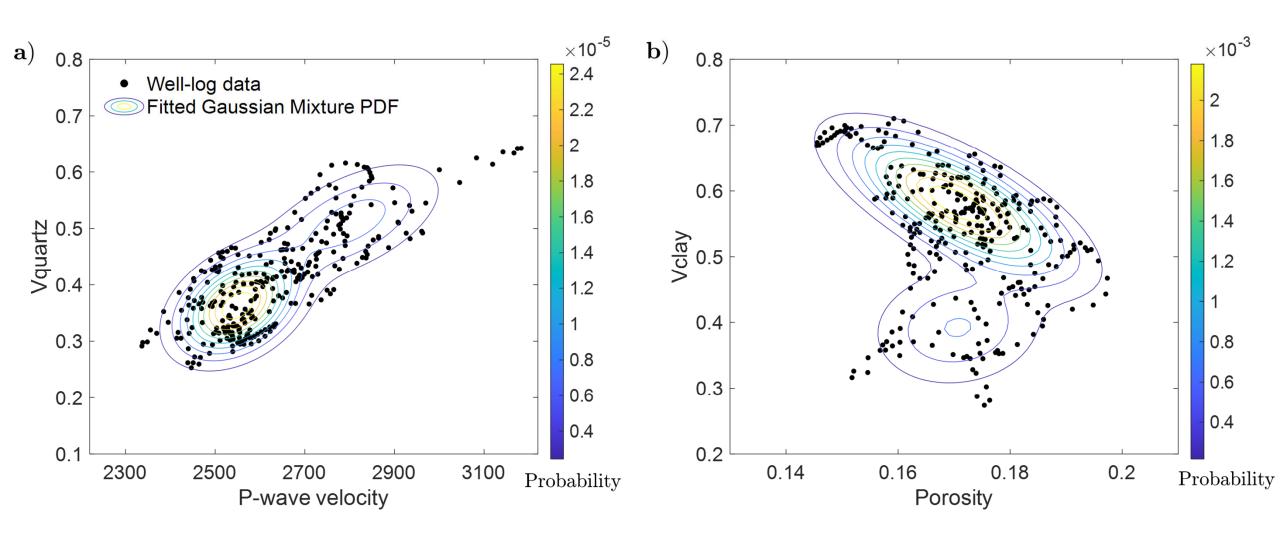
$$\mathbf{m} = g(\mathbf{r}) + \mathbf{e},$$

Bayesian setting

$$P(\mathbf{r}|\mathbf{m}) = \frac{P(\mathbf{r}, \mathbf{m})}{P(\mathbf{m})} = \frac{P(\mathbf{m}|\mathbf{r})P(\mathbf{r})}{P(\mathbf{m})},$$



Rock physics inversion





• Gaussian mixture approach (Grana and Rossa, 2010)

Prior:
$$P(\mathbf{r}) = \sum_{k=1}^{N_f} \lambda_k \mathcal{N}(\mathbf{r}; \boldsymbol{\mu}_r^k, \boldsymbol{\Sigma}_r^k),$$

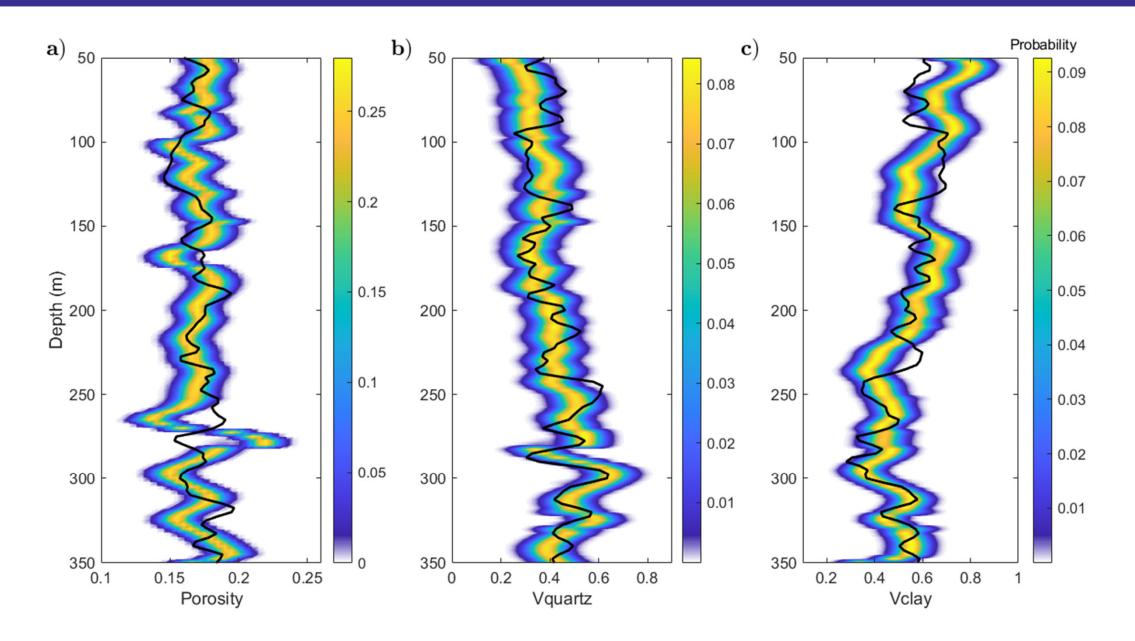
Joint:
$$P(\mathbf{r}, \mathbf{m}) = \sum_{k=1}^{N_f} \lambda_k \mathcal{N}(\mathbf{y}; \boldsymbol{\mu}_y^k, \boldsymbol{\Sigma}_y^k)$$
, where $\mathbf{y} = (\mathbf{r}, \mathbf{m})$,

Posterior:
$$P(\mathbf{r}|\mathbf{m}) = \sum_{k=1}^{N_f} \lambda_k' \mathcal{N}(\mathbf{r}; \boldsymbol{\mu}_{r|m}^k, \boldsymbol{\Sigma}_{r|m}^k),$$

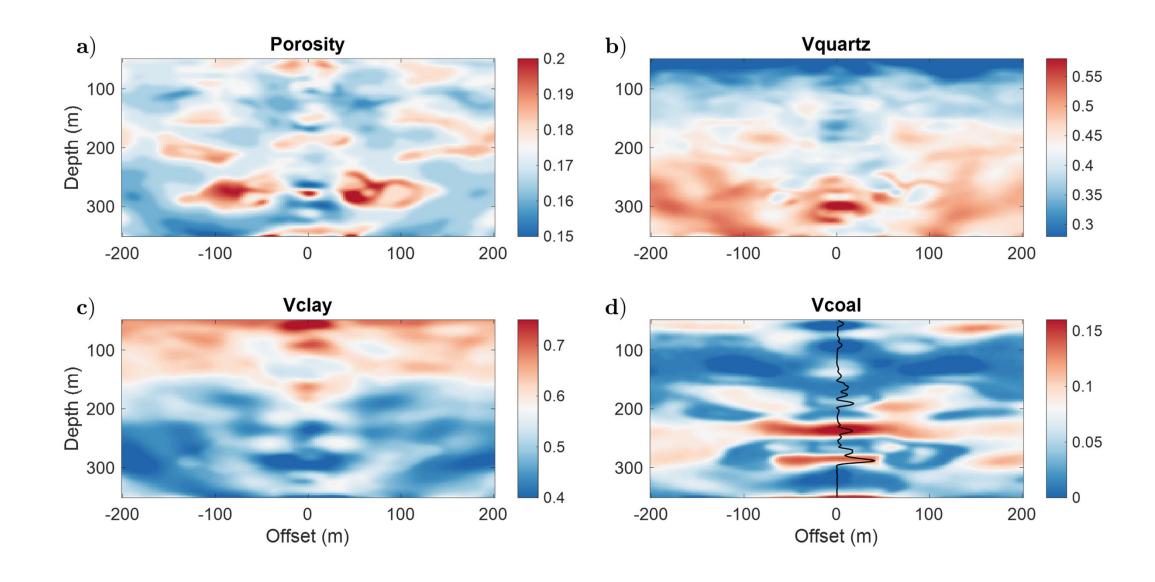
with
$$\begin{aligned} \boldsymbol{\mu}_{r|m}^k &= \boldsymbol{\mu}_r^k + \boldsymbol{\Sigma}_{r,m}^k (\boldsymbol{\Sigma}_{m,m}^k)^{-1} (\mathbf{m} - \boldsymbol{\mu}_m^k) \\ \boldsymbol{\Sigma}_{r|m}^k &= \boldsymbol{\Sigma}_{r,r}^k - \boldsymbol{\Sigma}_{r,m}^k (\boldsymbol{\Sigma}_{m,m}^k)^{-1} \boldsymbol{\Sigma}_{m,r}^k. \end{aligned}$$







MAP solution (2D)



- We applied a sequential inversion scheme combining elastic FWI and rock physics to a VSP dataset including accelerometer and DAS measurements.
- Our key strategies include an effective source approach to cope with near-surface complications, a modeling strategy to simulate DAS data, and a Bayesian Gaussian mixture approach to predict rock properties.
- The proposed methodology can be extended to time-lapse analysis, to predict dynamic reservoir properties such as CO2 saturation and pore pressure.





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