

Reducing the influence of remnant noises on FWI with misfit modification

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NSERC
CRSNG



UNIVERSITY OF CALGARY
FACULTY OF SCIENCE
Department of Geoscience



Normal Distribution Probability Density Function

$$L(\mathbf{d}|\mathbf{m}) = \frac{1}{\sqrt{(2\pi)^N |\mathbf{C}_d|}} \exp\left(-\frac{1}{2} \underbrace{\mathbf{r}^T \mathbf{C}_d^{-1} \mathbf{r}}\right)$$

Misfit function in FWI

$$E = \left[\underbrace{(\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}})}_{\mathbf{r}^T} \underbrace{(\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}})}_{\mathbf{r}} \right].$$

$\mathbf{d}_{\text{obs}}$ as “true data”, but actually noisy



Modified misfit function

$$E = \frac{1}{2N} \left[(\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}})^T \mathbf{C}_D^{-1} (\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}}) \right].$$

The gradient of the objective function

$$\frac{\partial E(\mathbf{m})}{\partial m_i} = \frac{1}{N} \left(\frac{\partial \mathbf{d}_{\text{pre}}}{\partial m_i} \right)^T \boxed{\mathbf{C}_D^{-1} (\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}})}.$$

The artifacts generated by matching seismic data with **low SNR** will be **down weighted**, improving the quality of imaging.



1 Estimation of data covariance matrix

$$c_j = \frac{1}{N} \sum_{k=0}^{N-j-1} (\mathbf{d}_{j+k} - \bar{\mathbf{d}}) (\mathbf{d}_k - \bar{\mathbf{d}}) .$$

Inversion results of the 1st FWI $\approx$ Sample mean $\bar{\mathbf{r}}$

2 Dimension of covariance matrix

Time domain:

$$(N_t * N_r * N_s) * (N_t * N_r * N_s)$$

Frequency domain:

$$(N_f * N_r * N_s) * (N_f * N_r * N_s)$$



Research questions:

How noises can influence the FWI results ?

How modified misfit function can improve the inversion result ?

Acoustic FWI in time domain

Elastic FWI in frequency domain

Random noise
Correlated noise

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graph BT; A[Random noise<br/>Correlated noise] --> B[Acoustic FWI in time domain]; A --> C[Elastic FWI in frequency domain];
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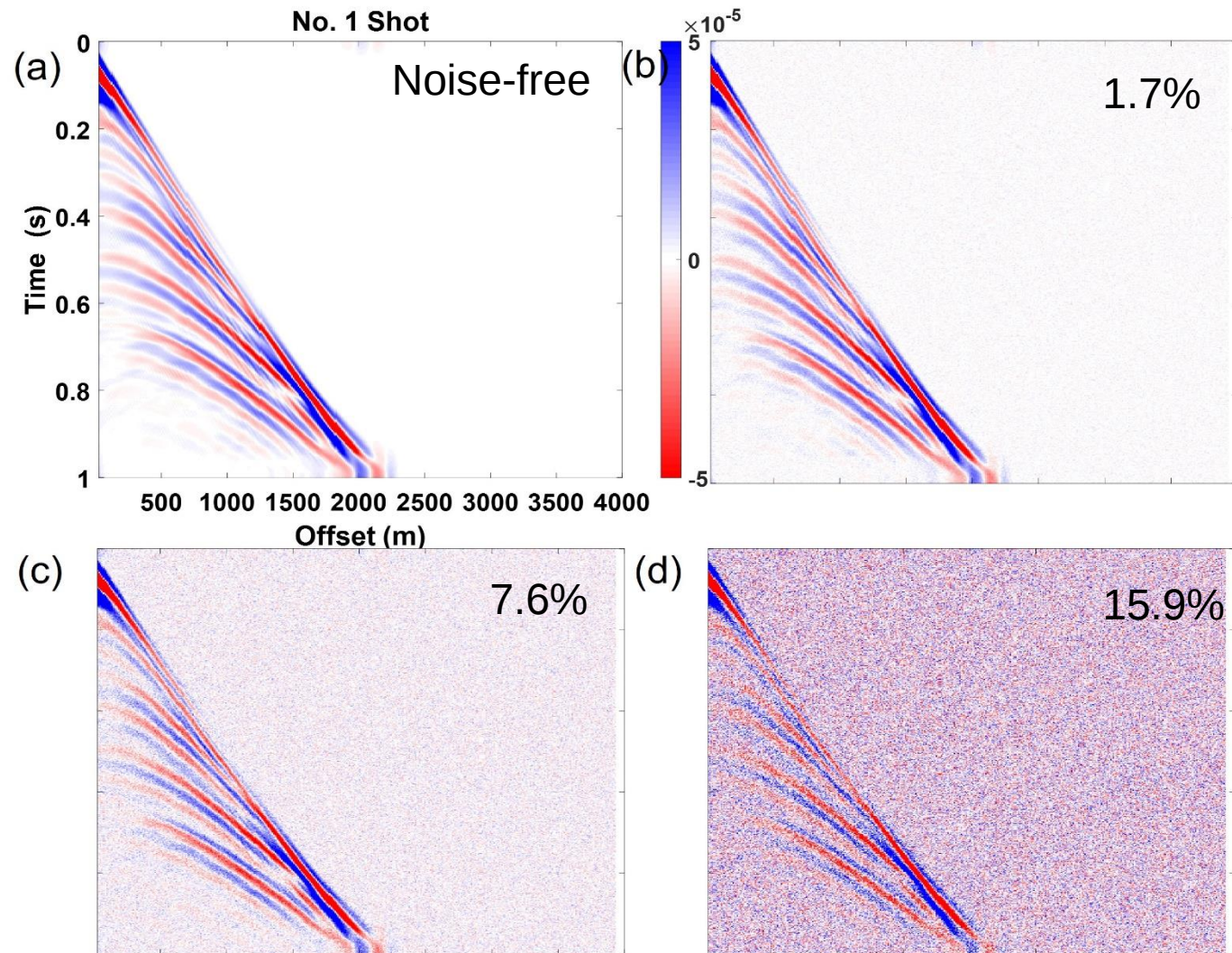


Acoustic FWI in time domain

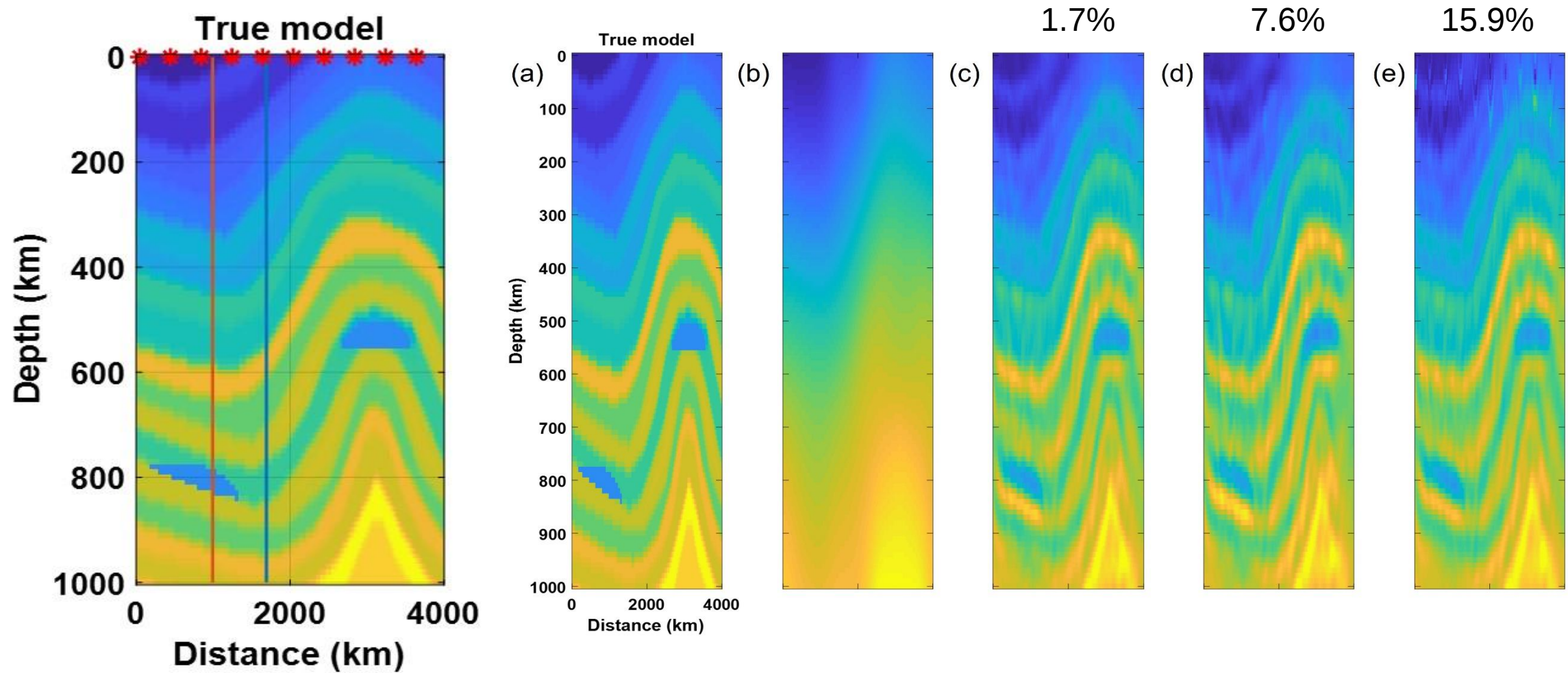
Random noise

Correlated noise

Seismic data with random noises (d_obs)



Inversion results



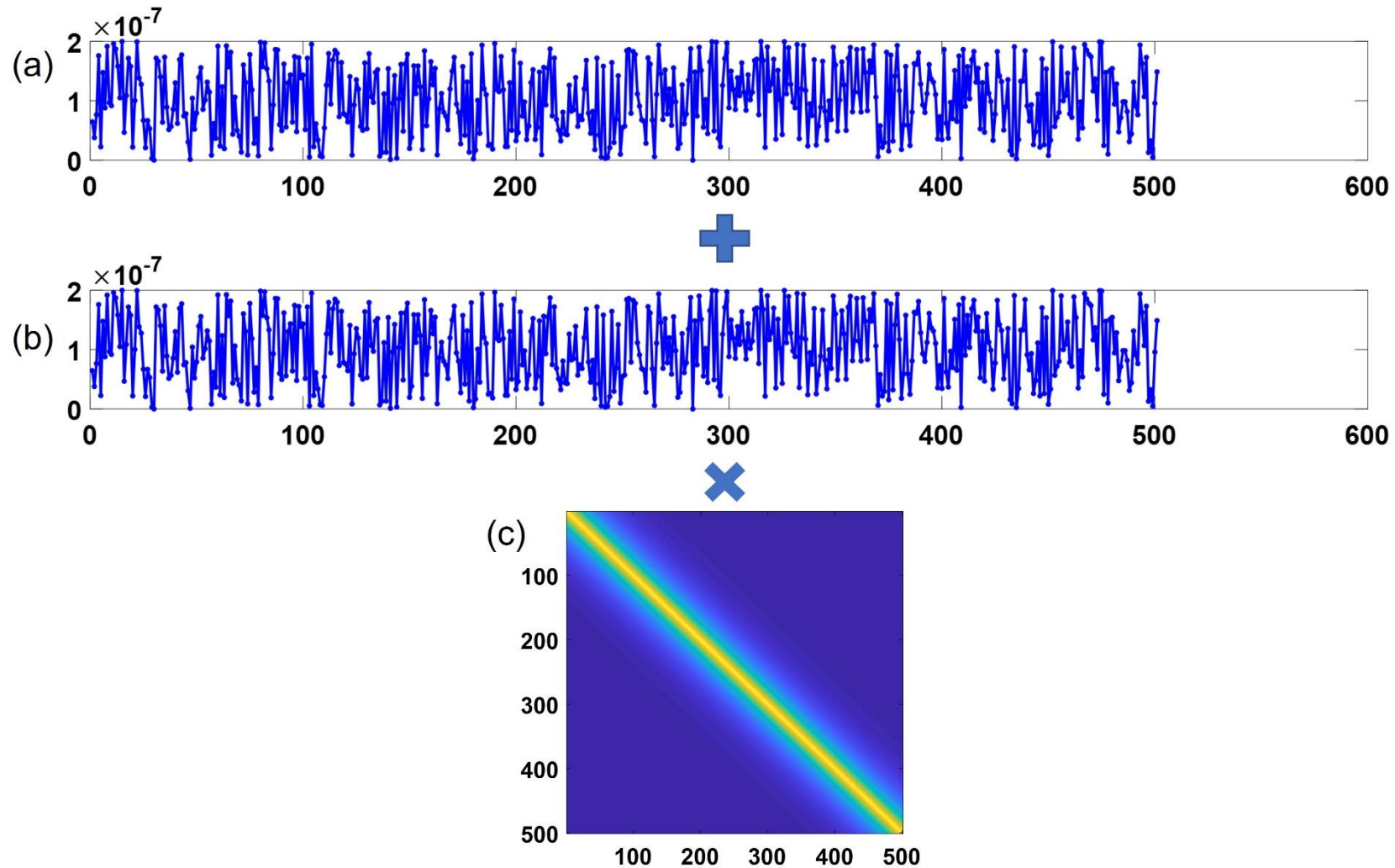


Acoustic FWI in time domain

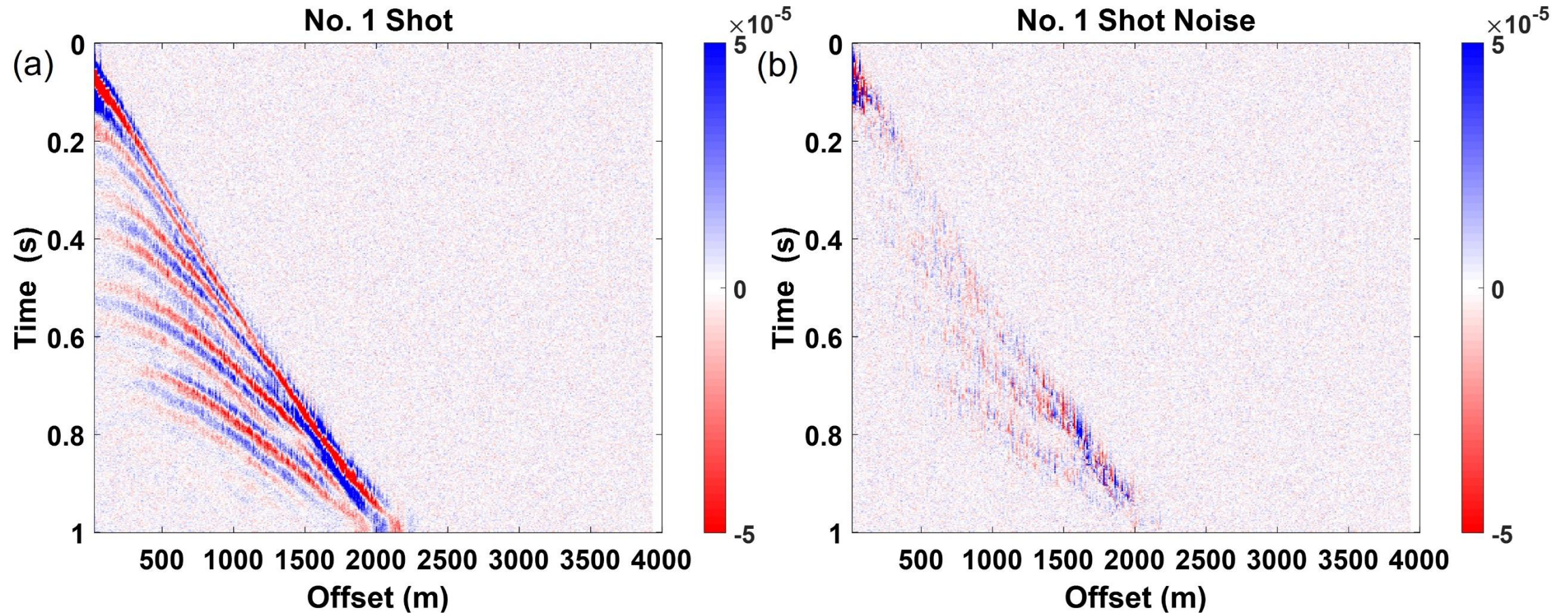
Random noise

Correlated noise

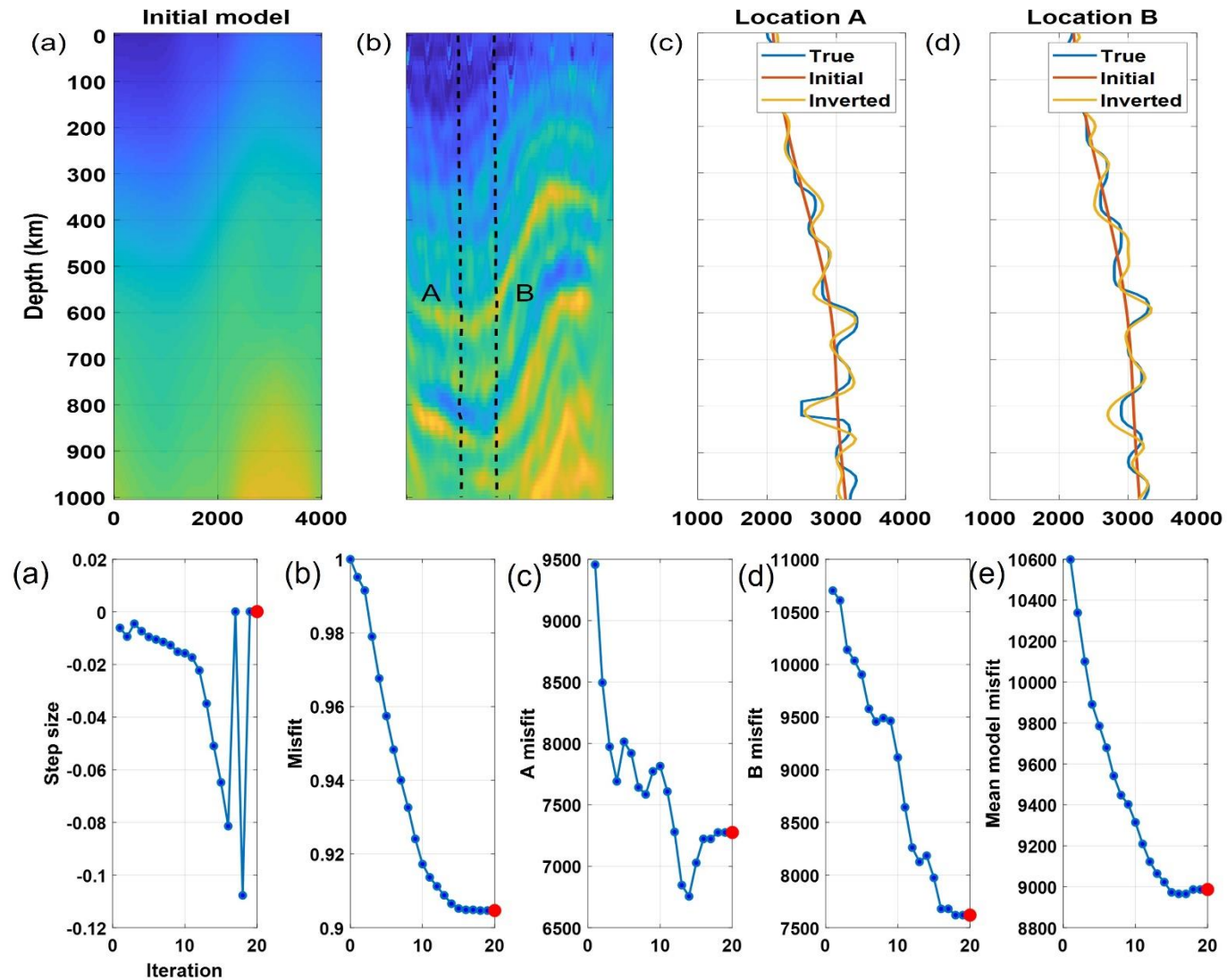
Generation of correlated noise



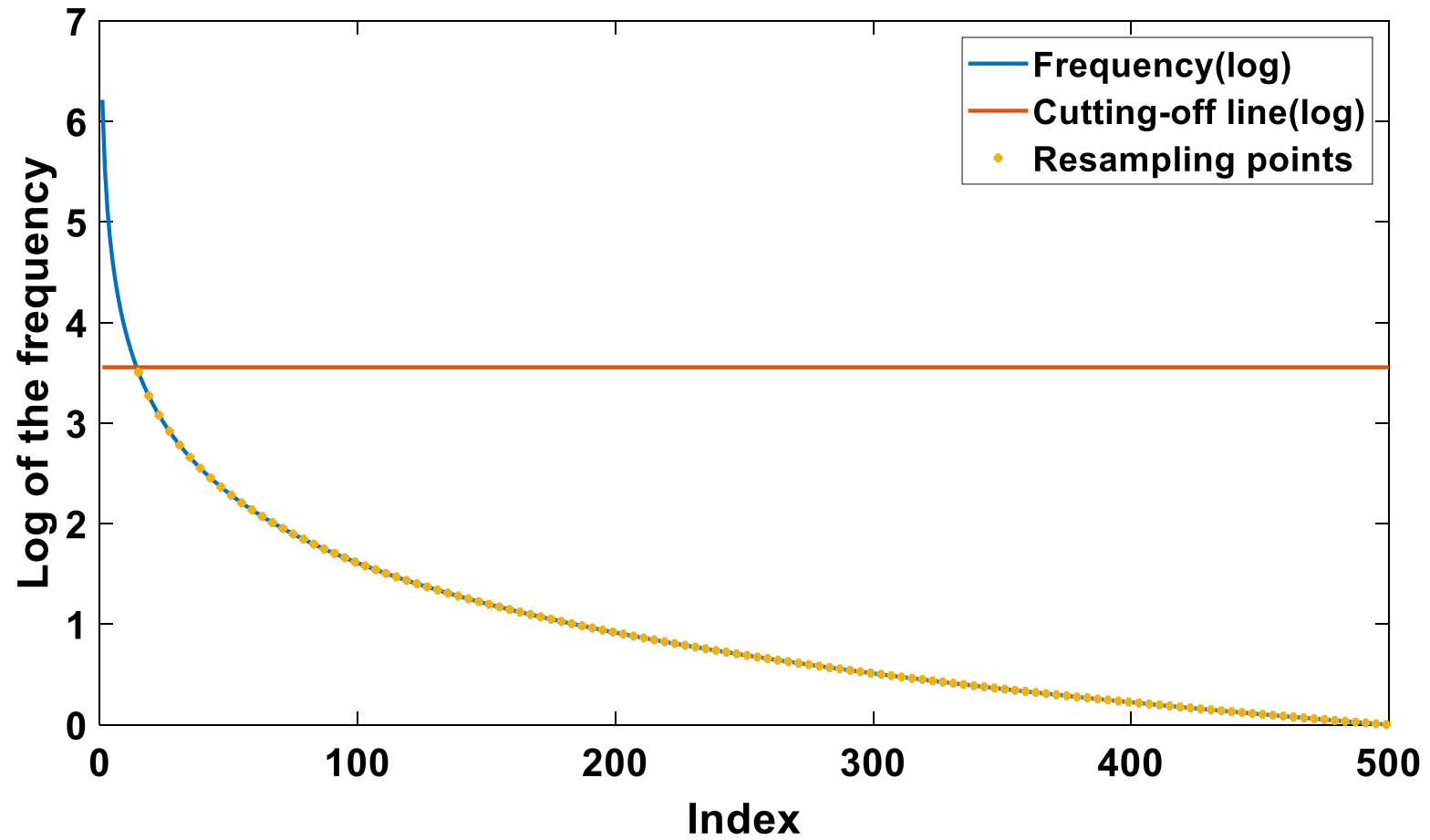
Seismic data with correlated noises



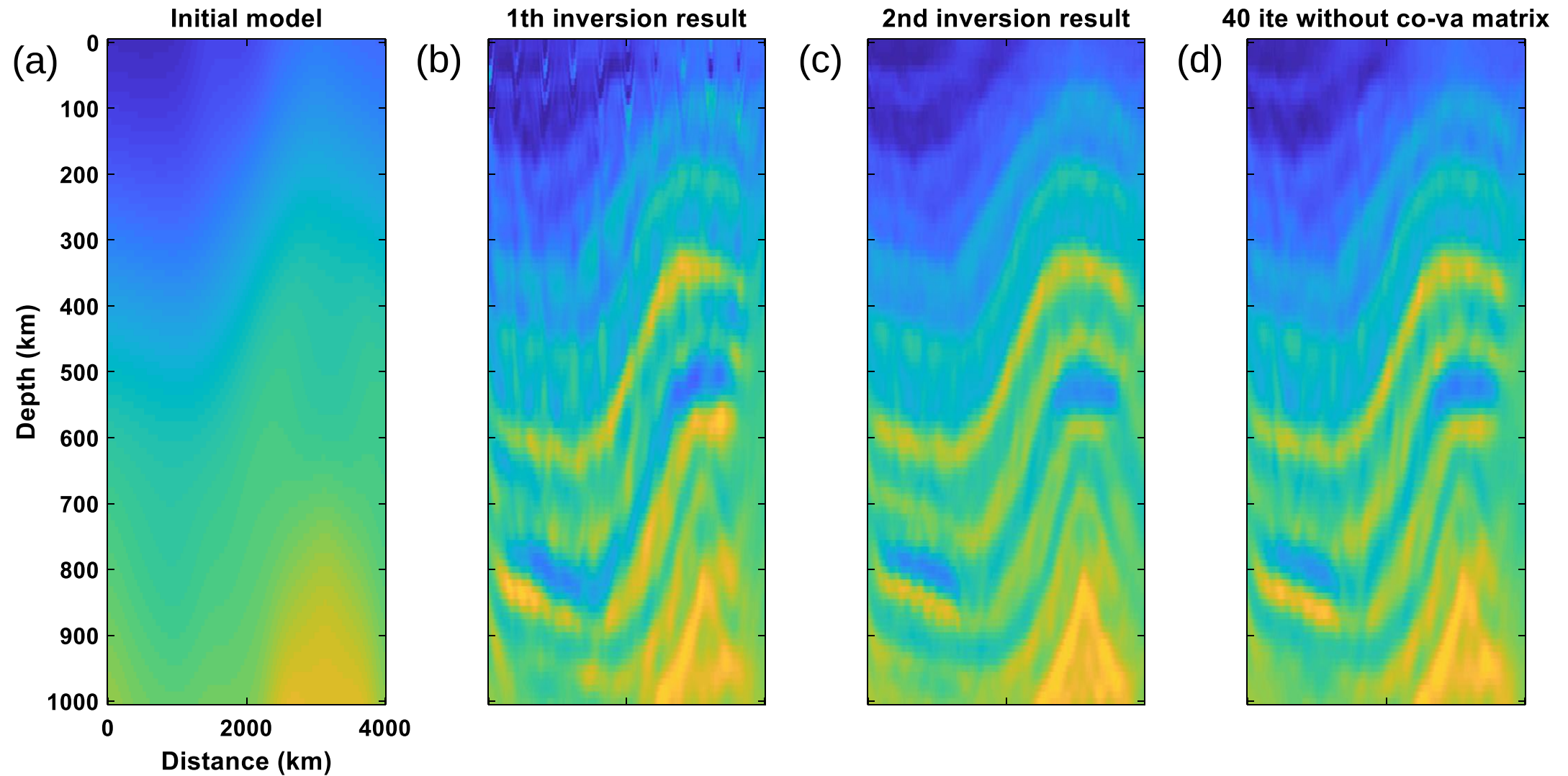
1st FWI inversion result



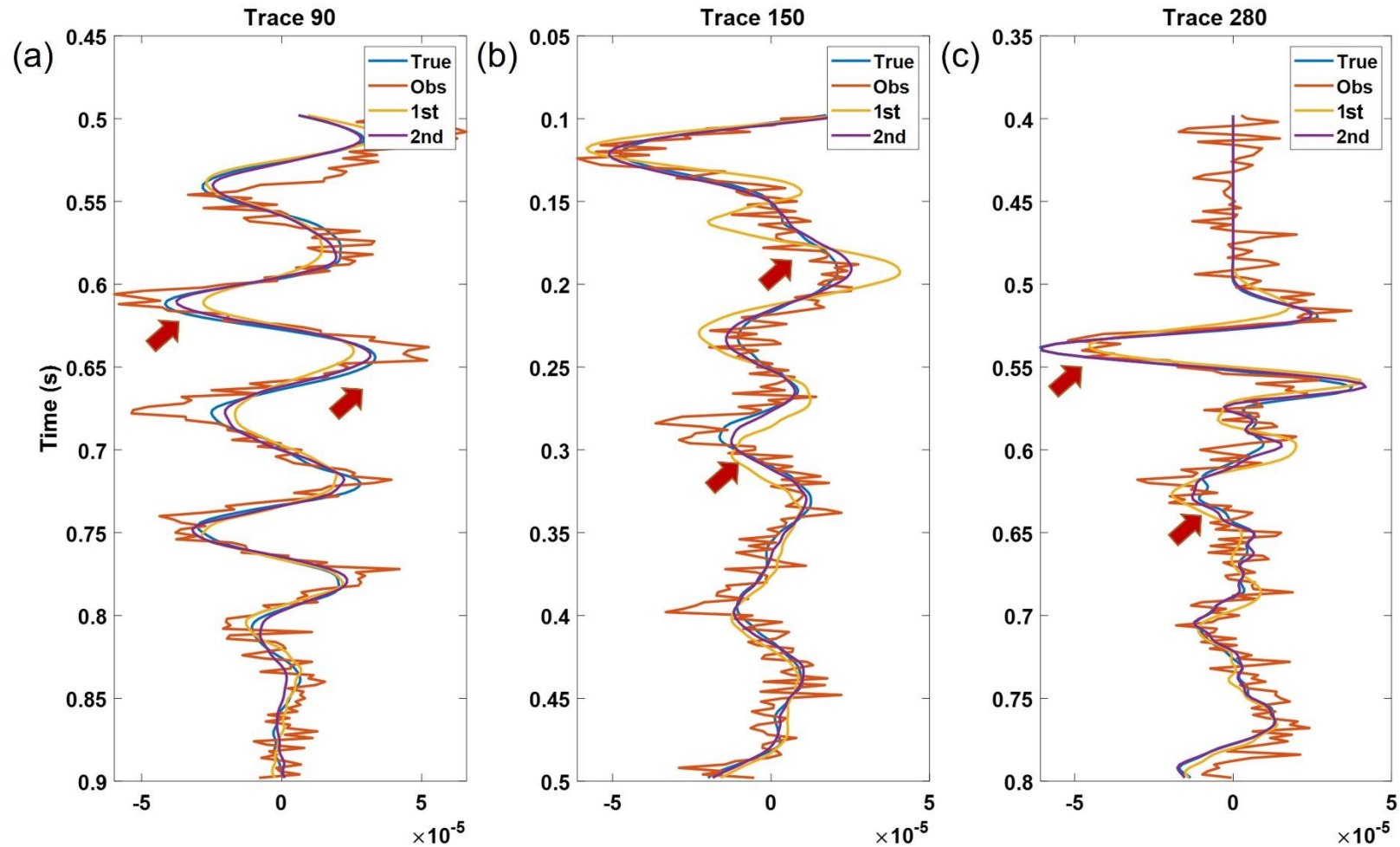
Resampling in frequency domain



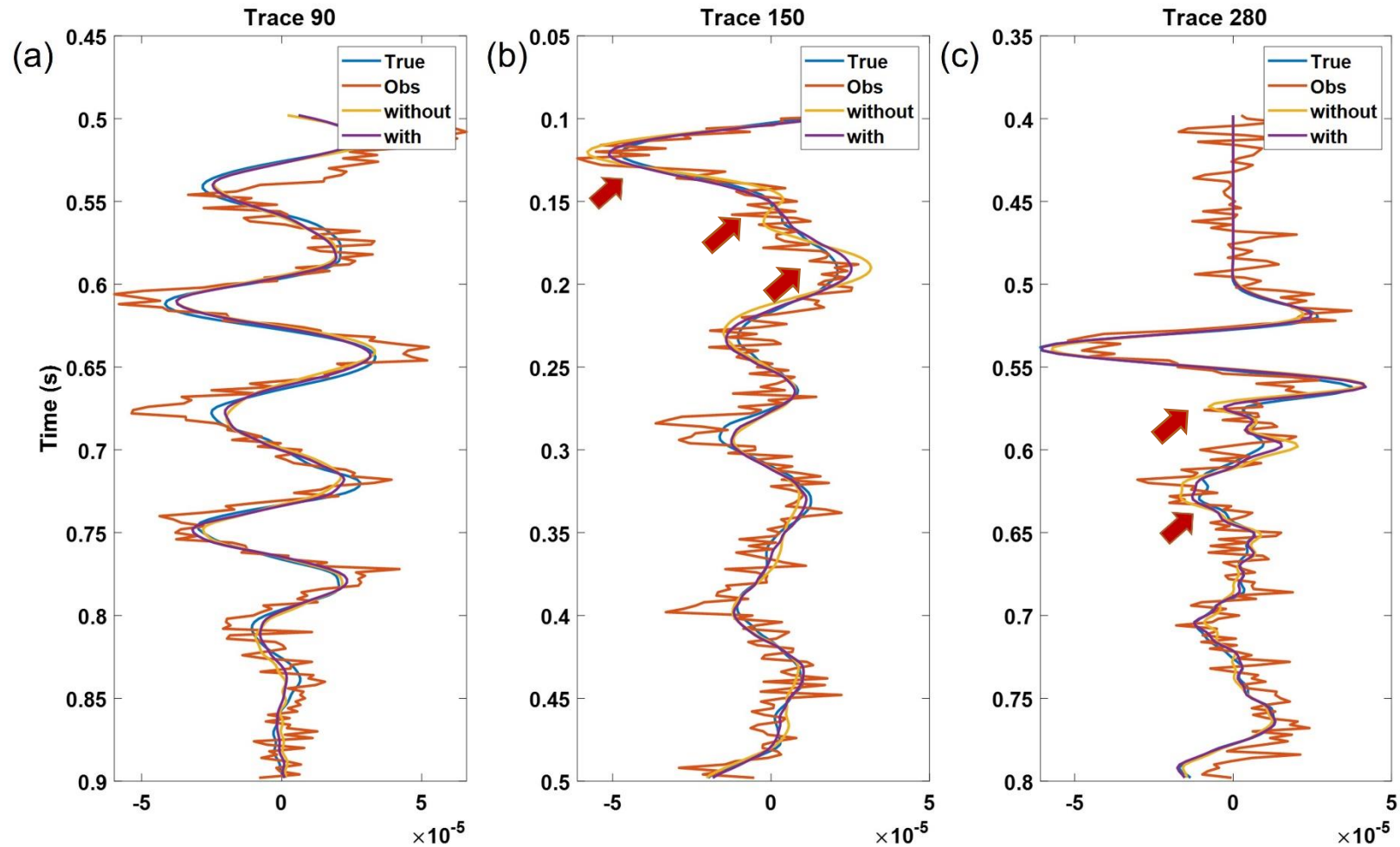
2nd FWI inversion result



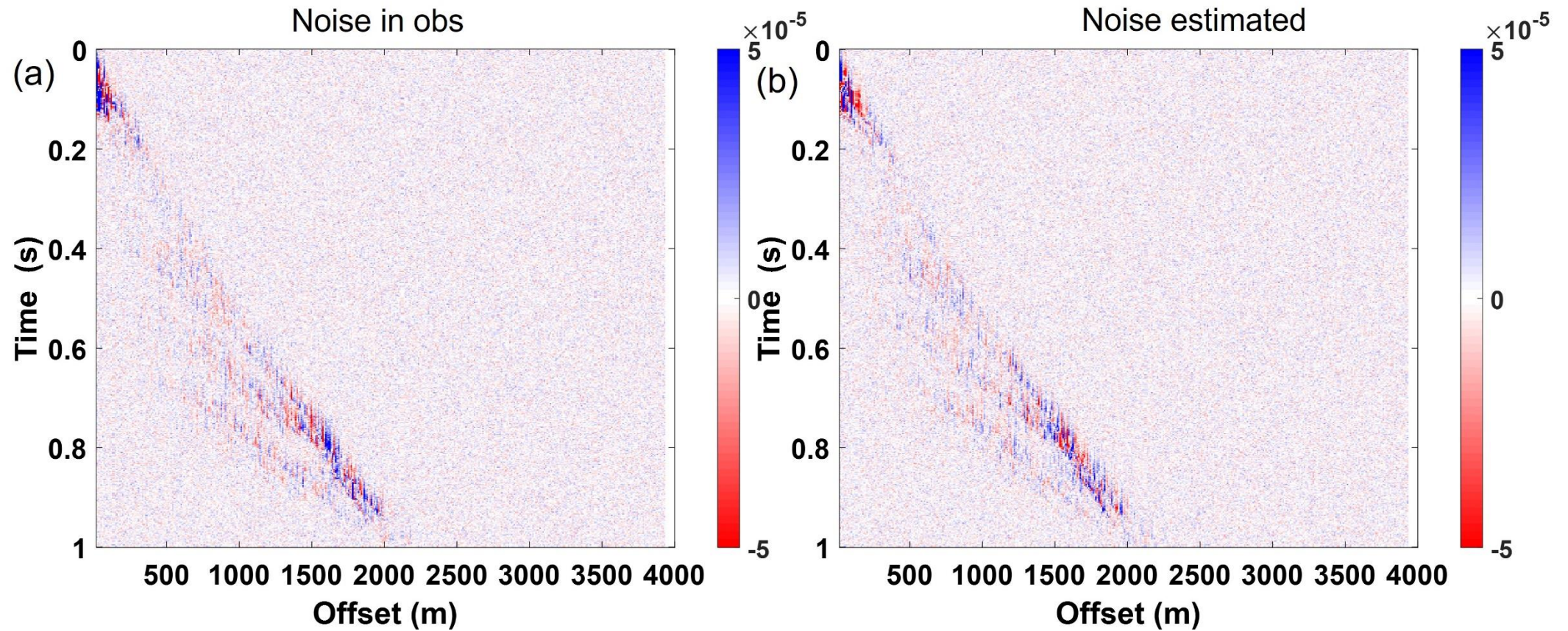
Trace comparison for the 1st inversion



Trace comparison for the 2nd inversion



True noise and estimated noise



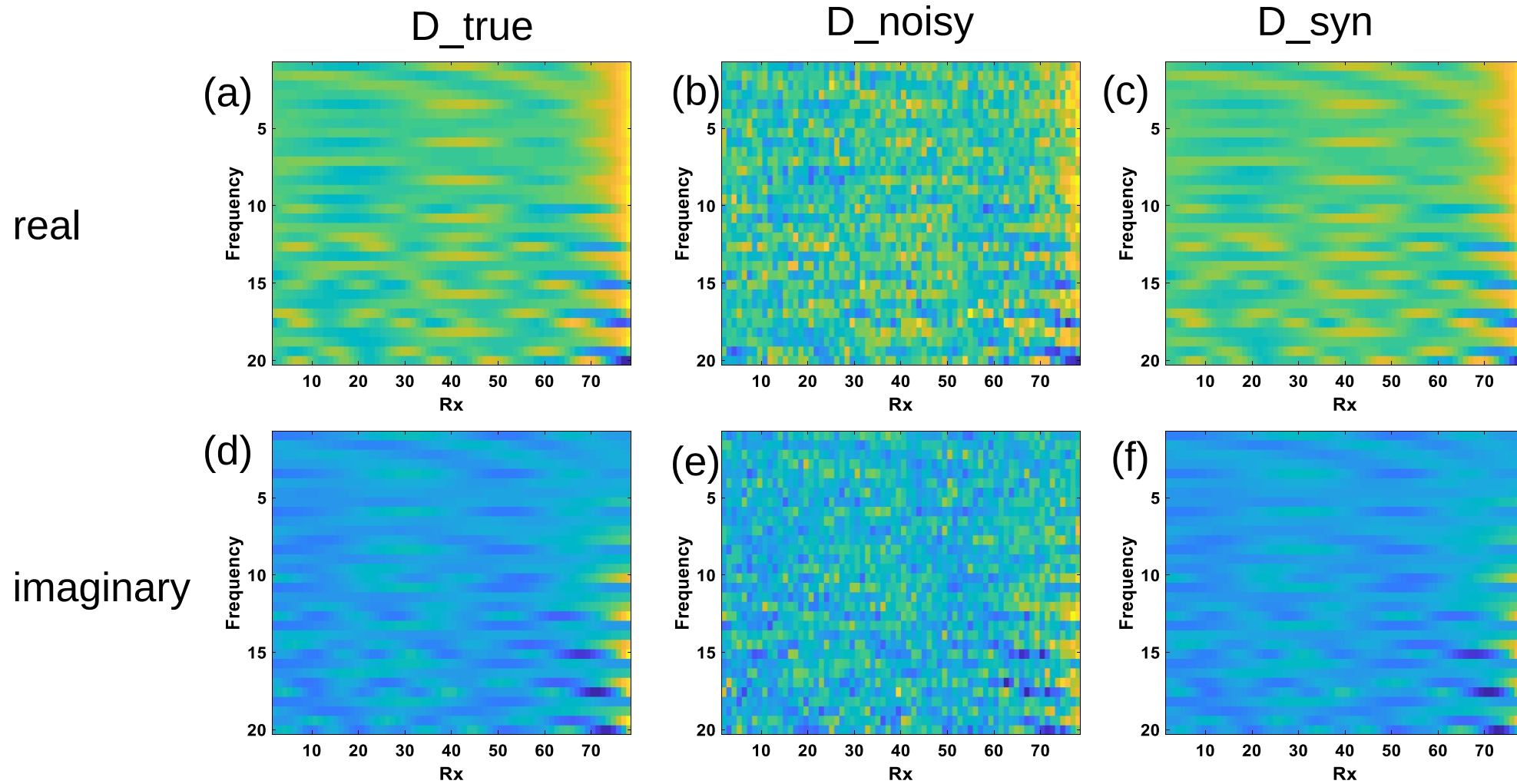


Elastic FWI in frequency domain

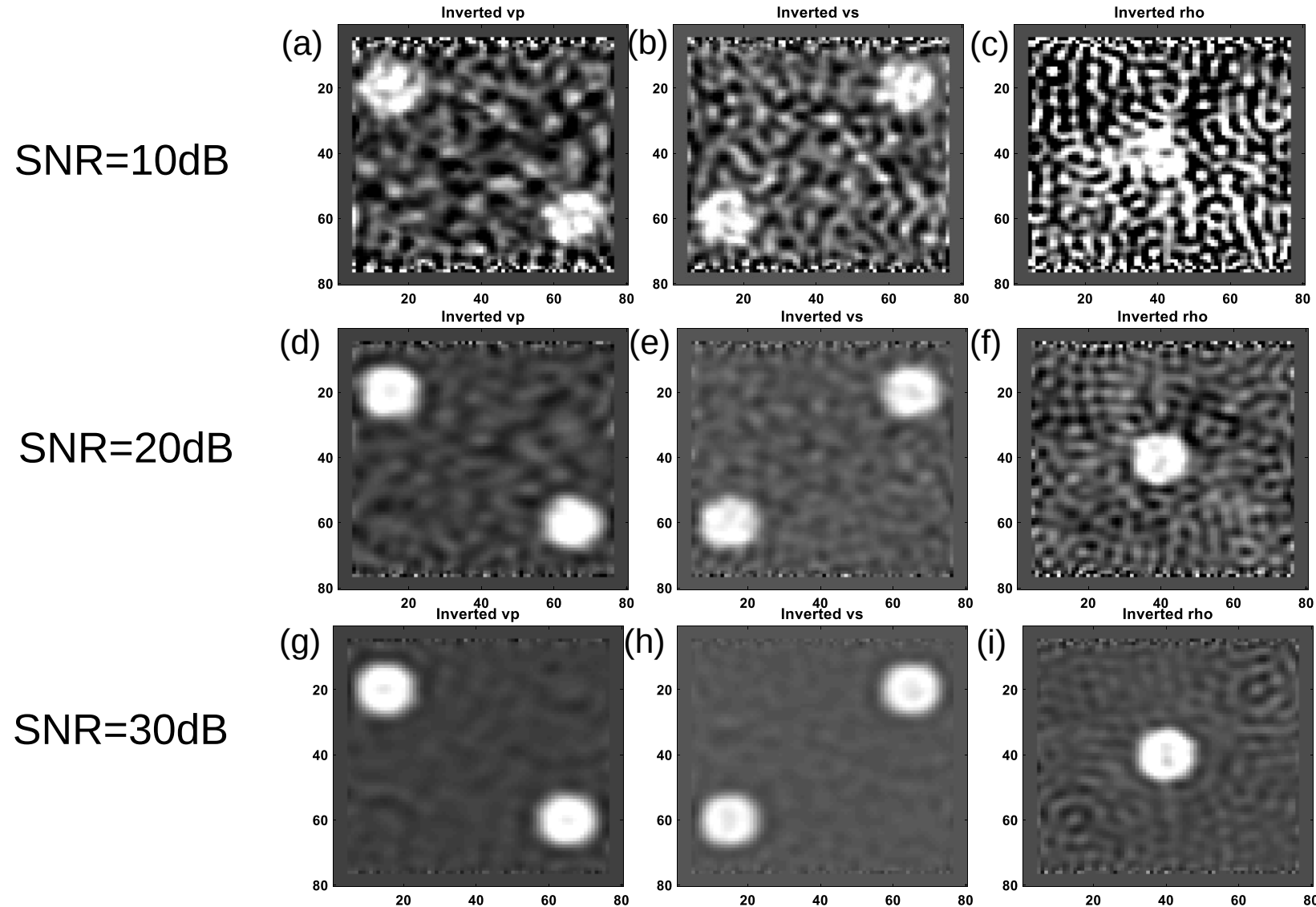
Random noise

Correlated noise

Seismic data with random noises (d_obs)

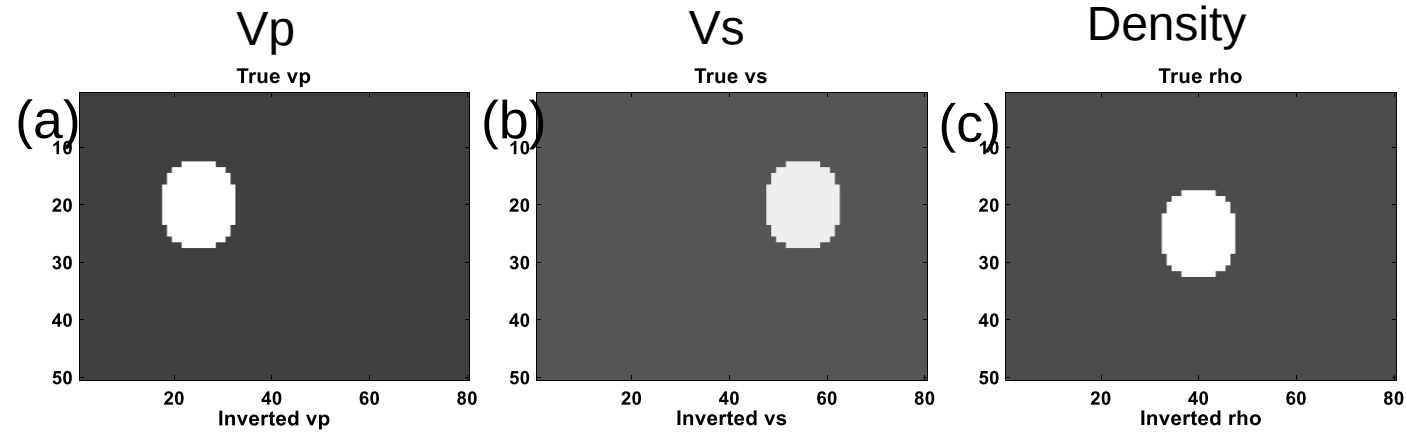


Inversion result of seismic data with different SNR (random noises)

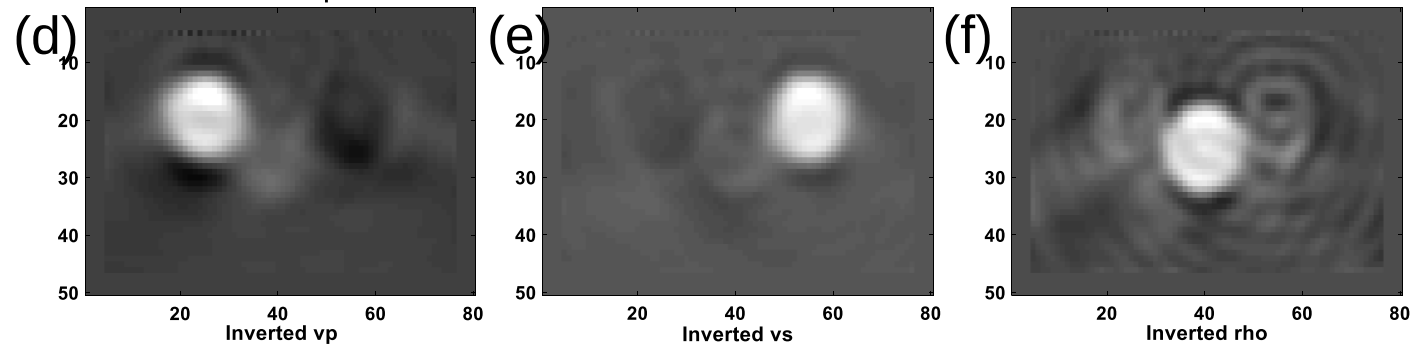


1st FWI inversion result

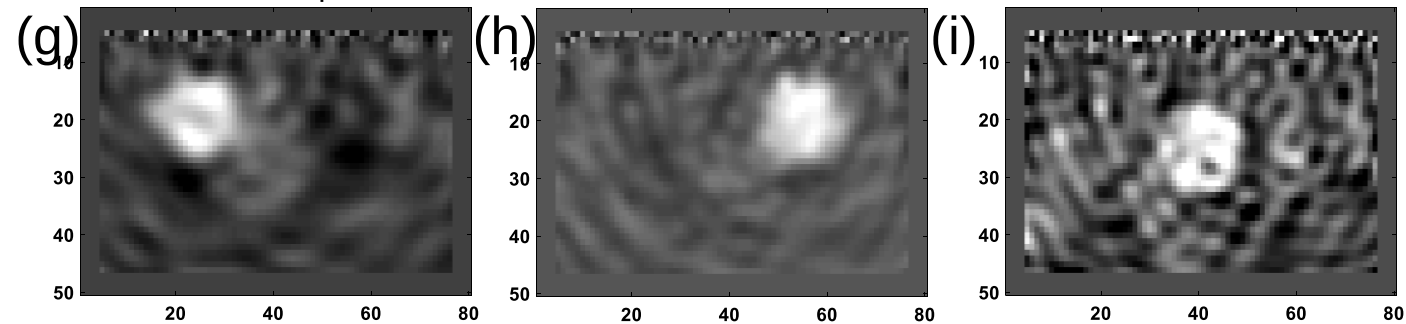
True model



Inverted model of
noise-free data

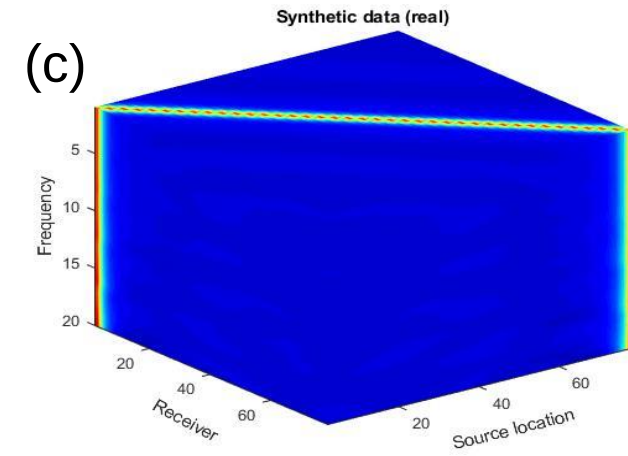
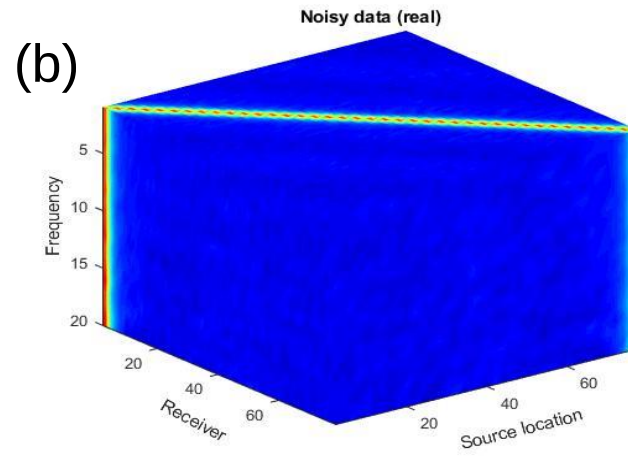
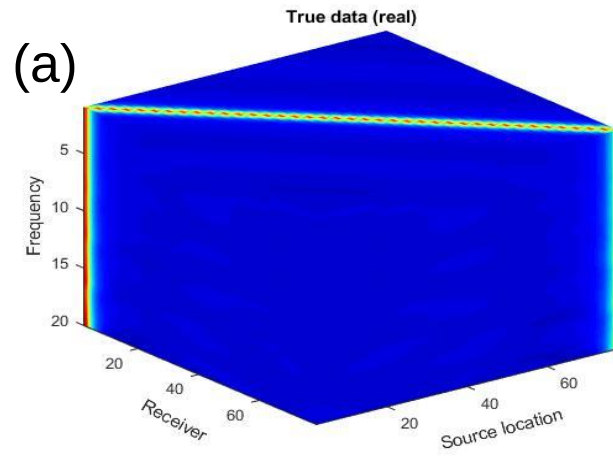


Inverted model of
data with noisy data
SNR=16dB

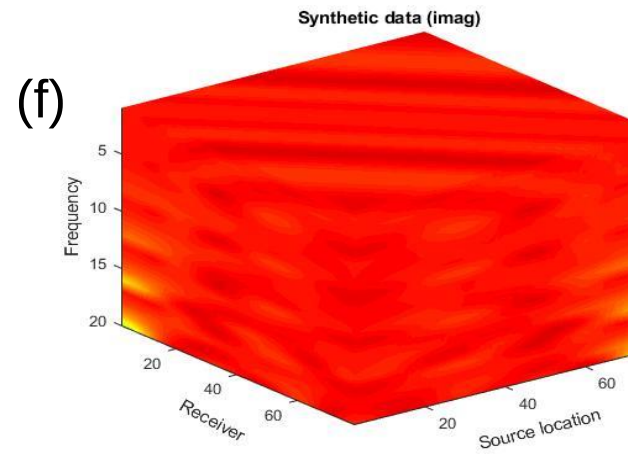
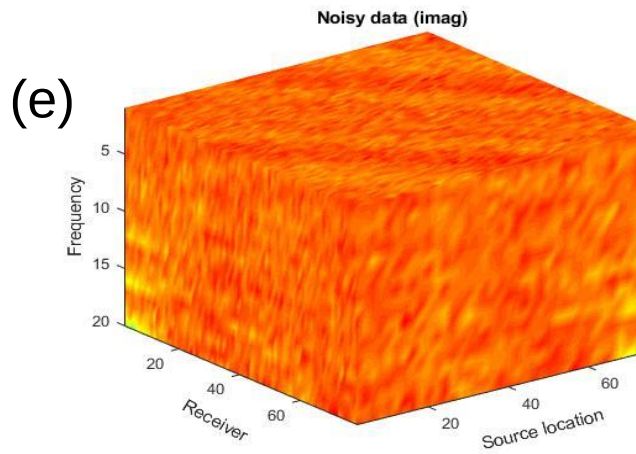
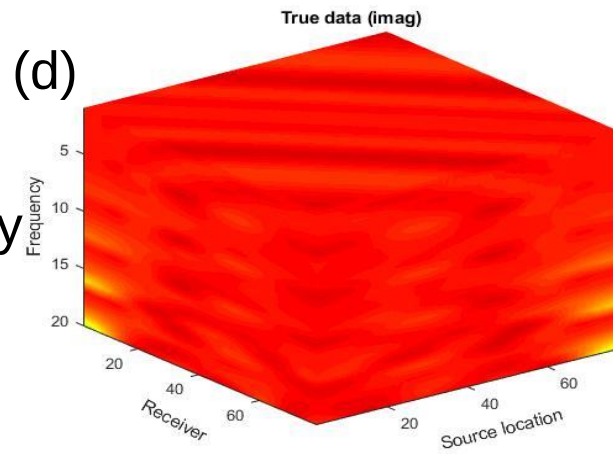


Seismic data in frequency domain

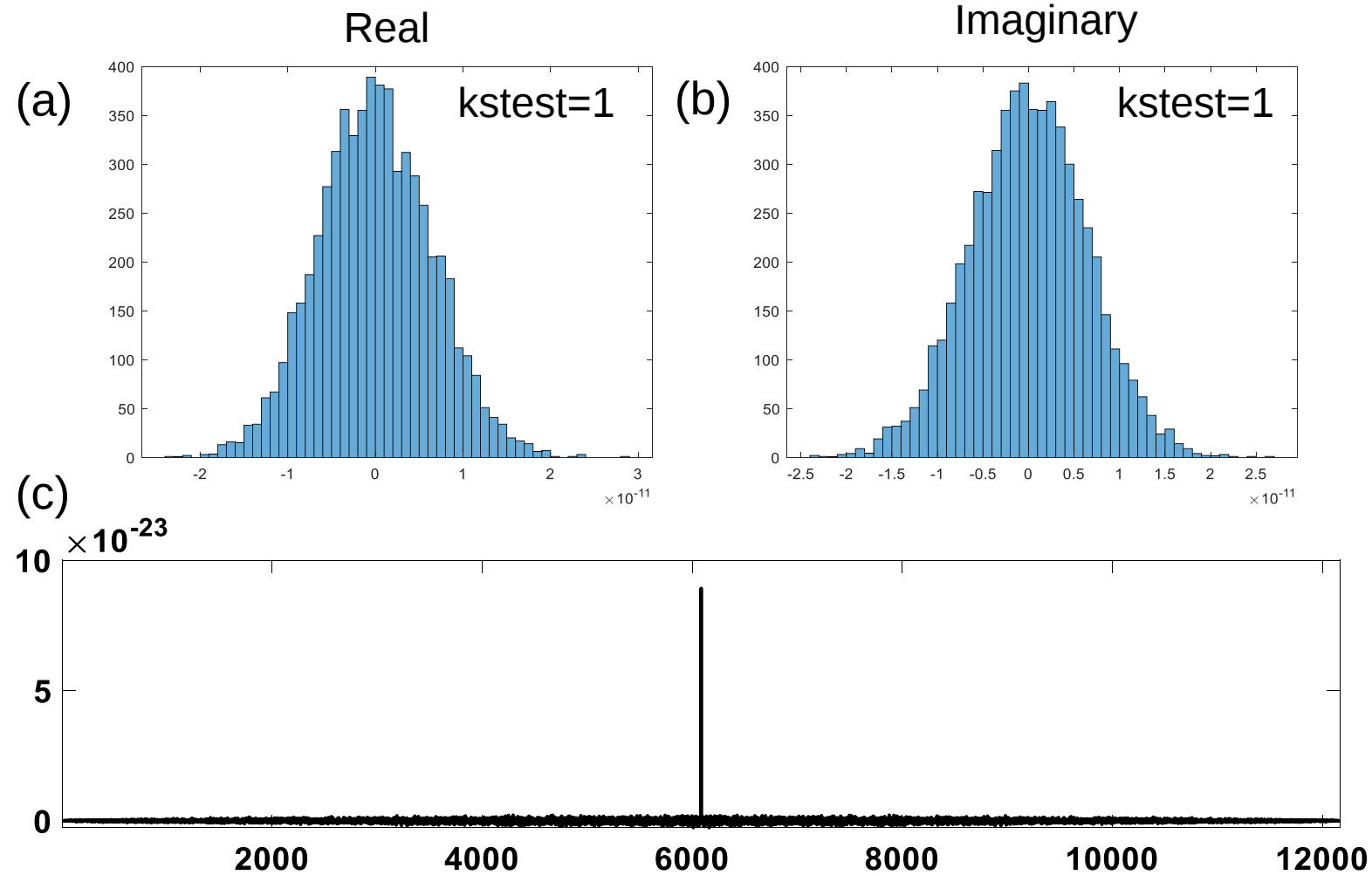
Real



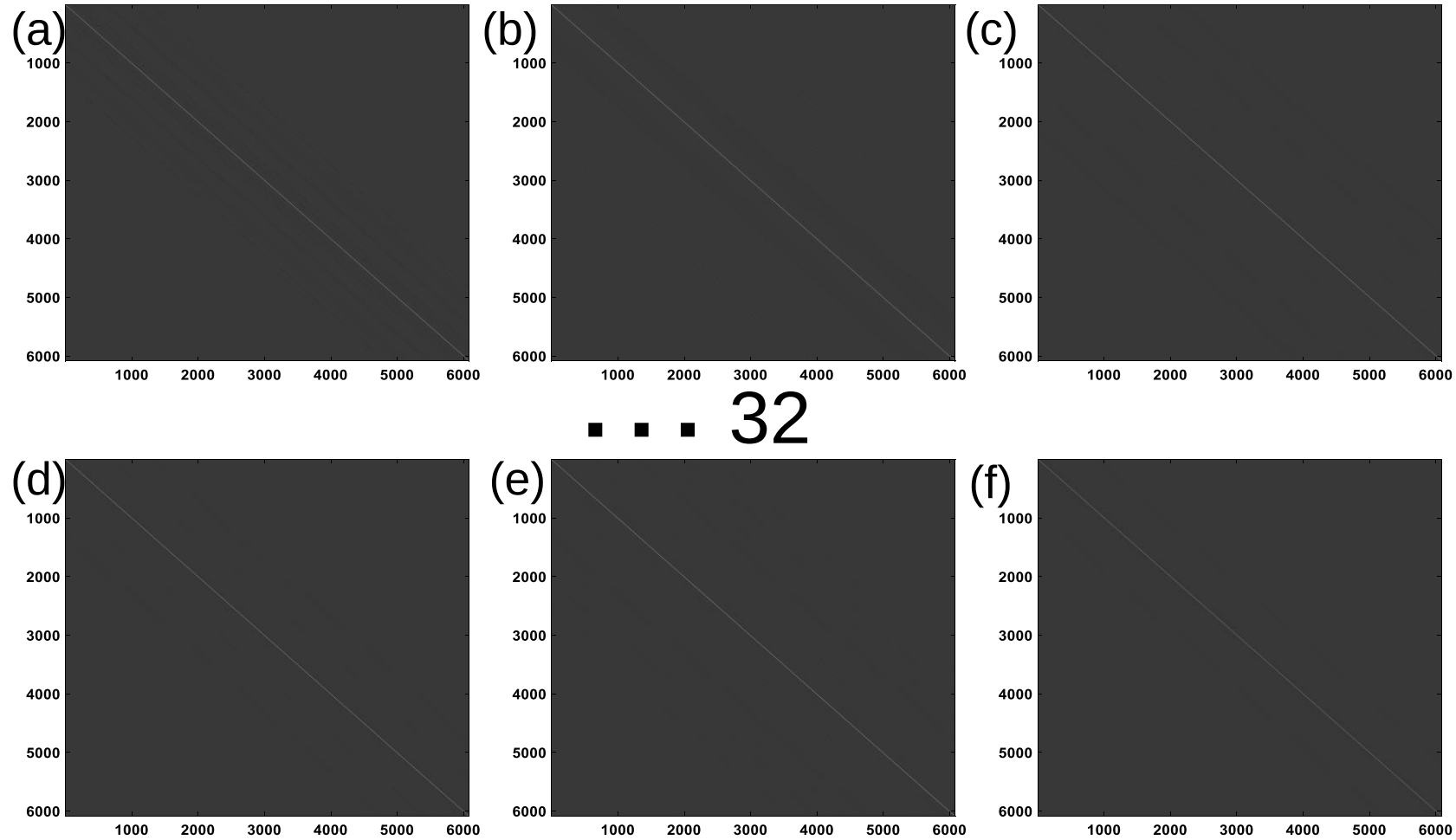
Imaginary



Data residuals

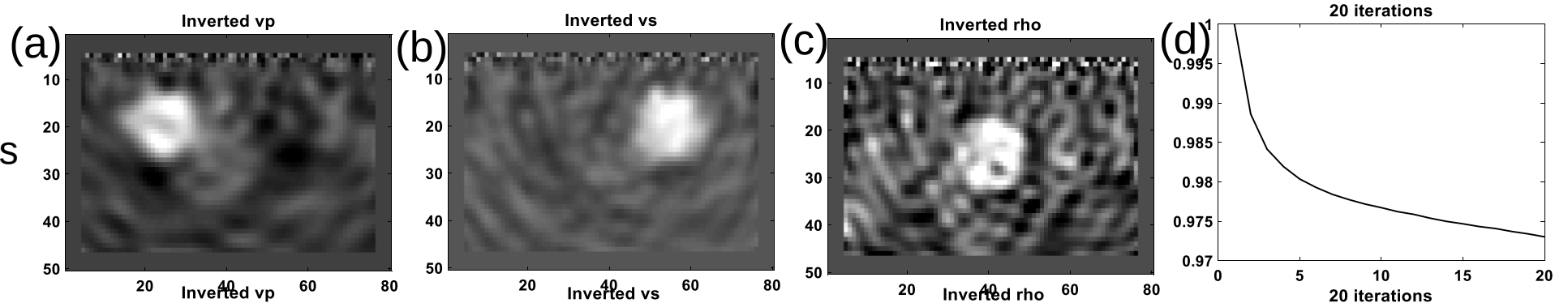


Data covariance matrices

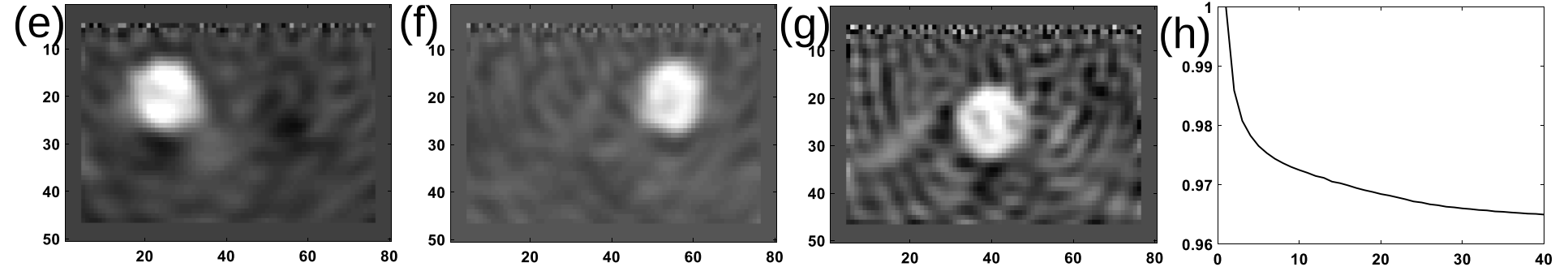


Inversion results

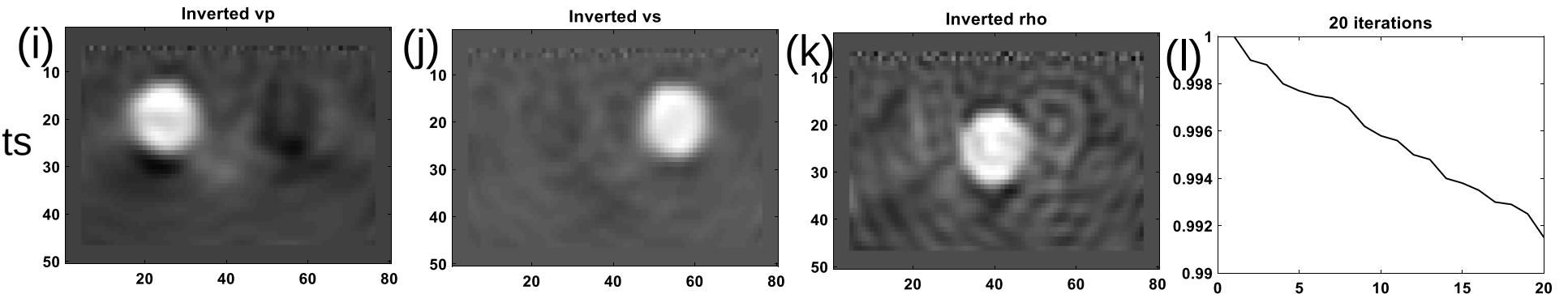
1st inversion results



Reference results



New method results





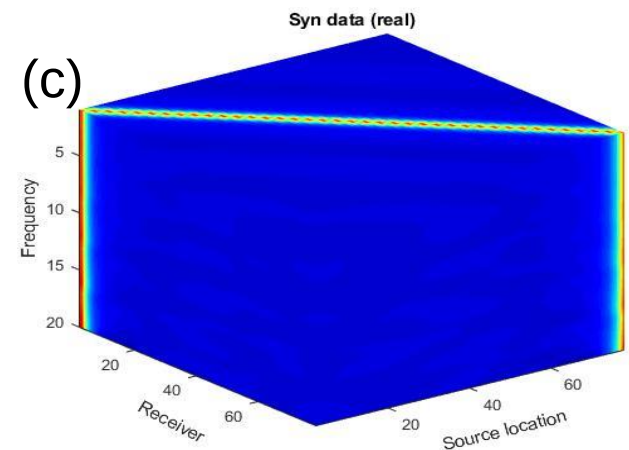
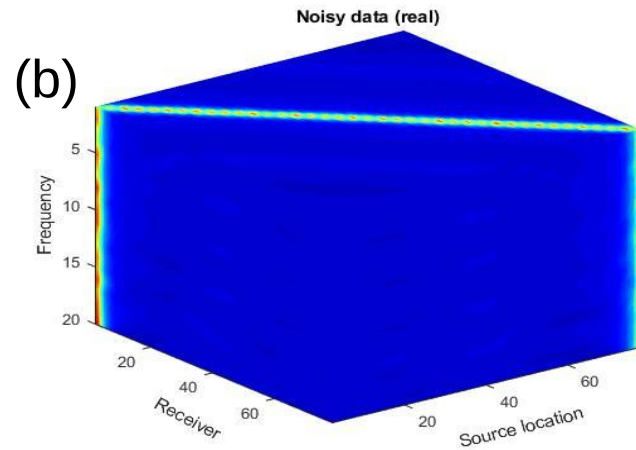
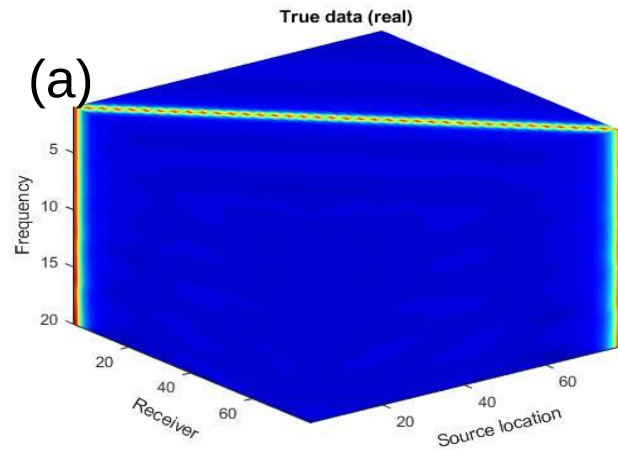
Elastic FWI in frequency domain

Random noise

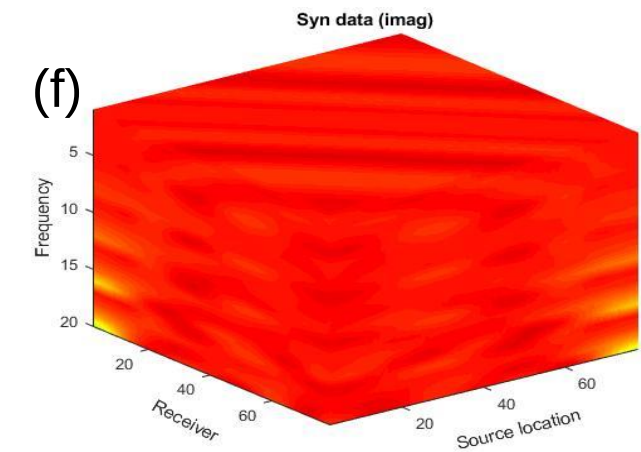
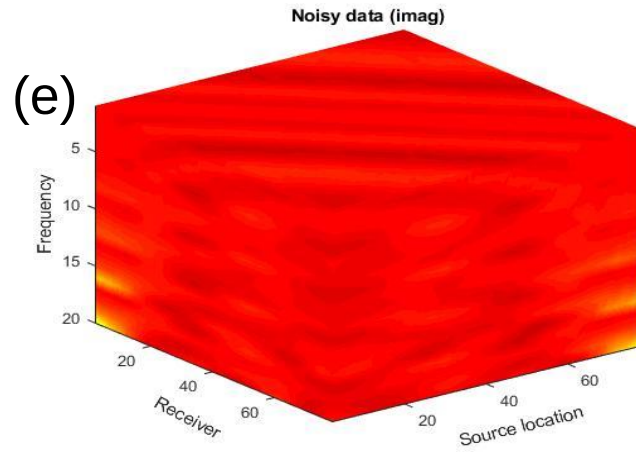
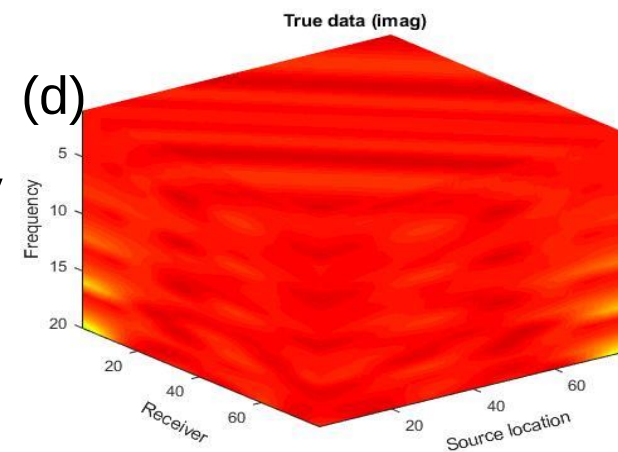
Correlated noise

Seismic data in frequency domain

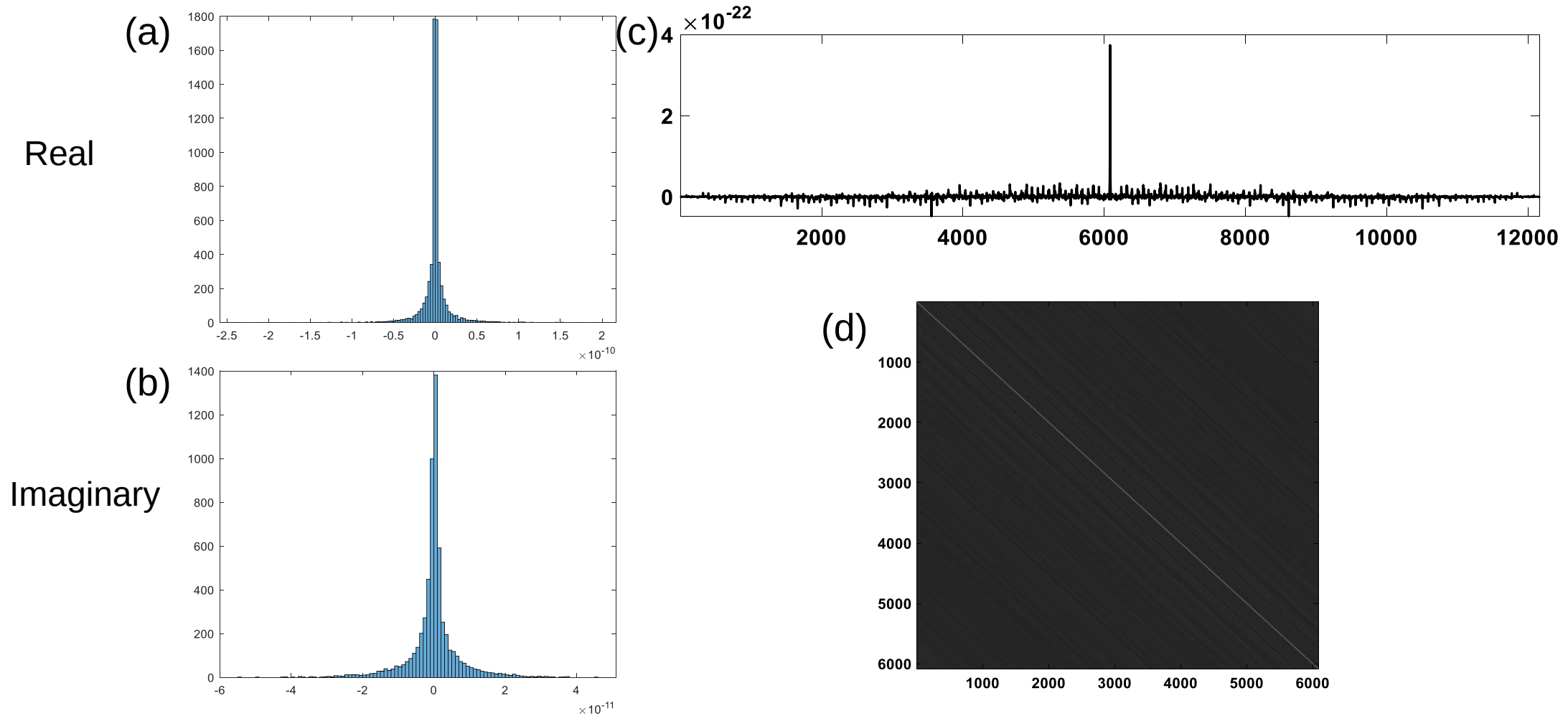
Real



Imaginary

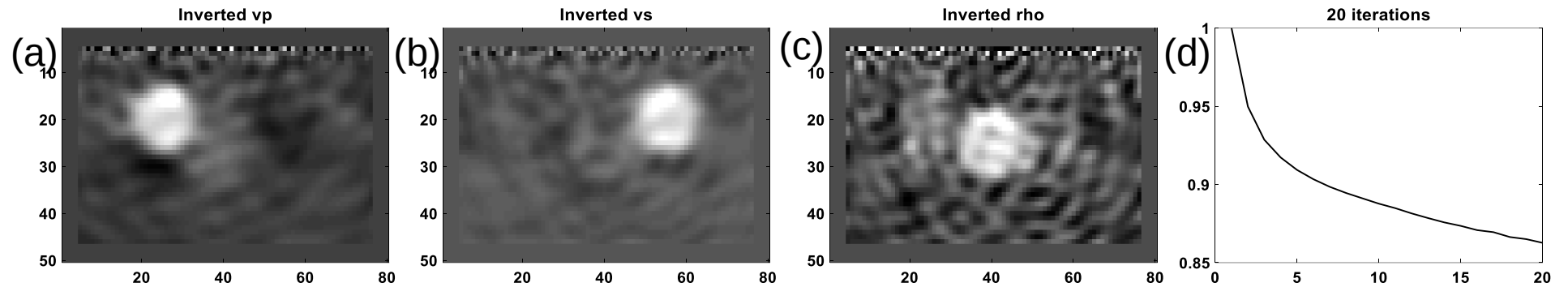


Data residuals and data covariance matrices

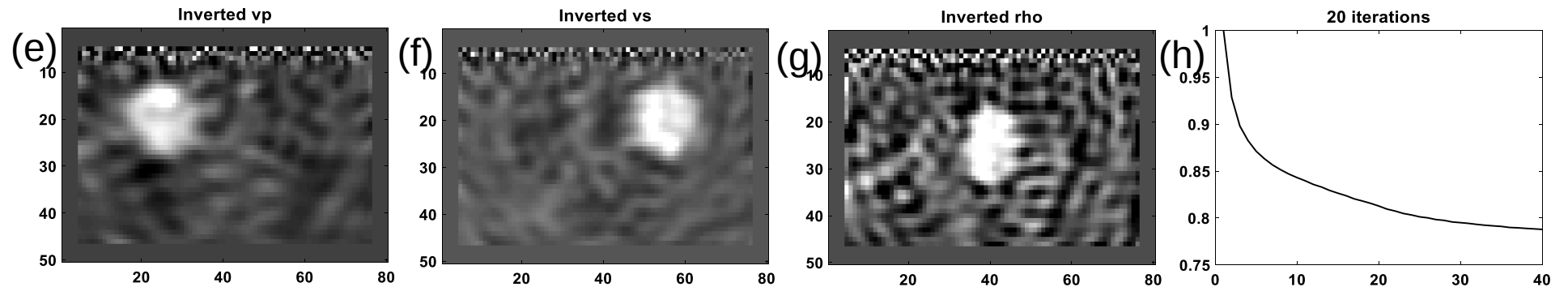


Inversion results

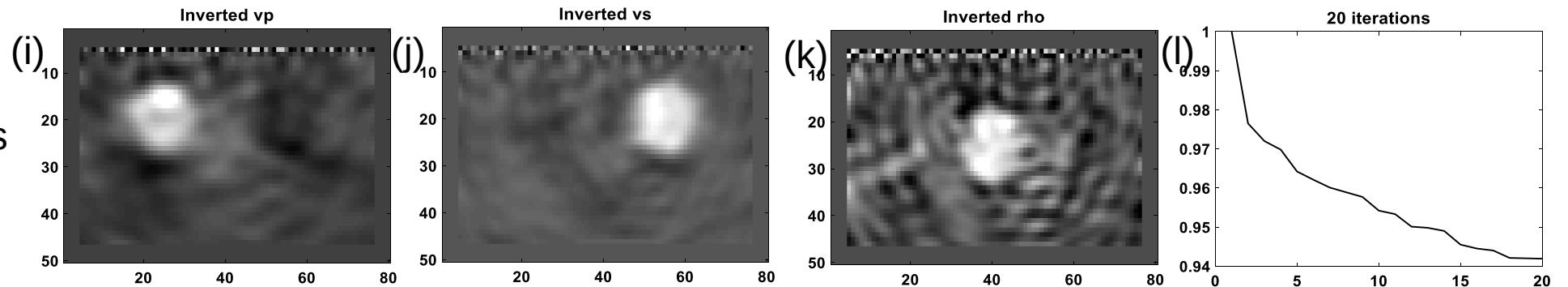
1st inversion results



Reference results



New method results





- Inversion algorithm is more resistant to the influence of random noises.
- The influence can be ignored if the random noise magnitude is within 20 percent of the true data.
- FWI with modified misfit can better estimate the random and correlated noises in the observing data by incorporating the data covariance matrix.
- The new method can yield better imaging with less artifacts.

Further study

- Field data tests need to be conducted to examine the effectiveness of the method.
- Local minimum.



Jan Dettmer, Ninoska

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CREWES sponsors

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CREWES faculty, staff and students







Modified misfit function

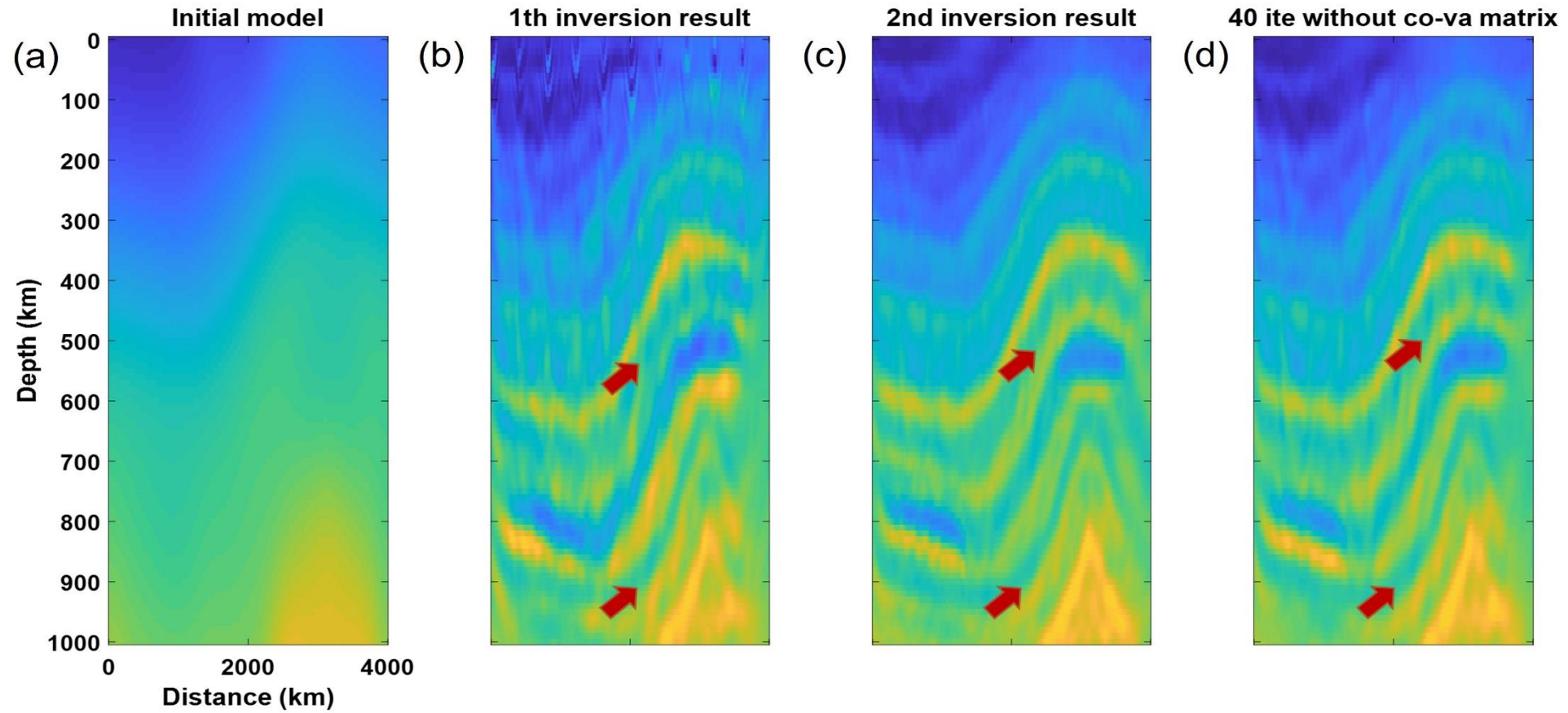
$$E = \frac{1}{2N} \left[(\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}})^T \mathbf{C}_D^{-1} (\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}}) \right].$$

The gradient of the objective function

$$\frac{\partial E(\mathbf{m})}{\partial m_i} = \frac{1}{N} \left(\frac{\partial \mathbf{d}_{\text{pre}}}{\partial m_i} \right)^T \boxed{\mathbf{C}_D^{-1} (\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}})}.$$

The artifacts generated by matching seismic data with **low SNR** will be **down weighted**, improving the quality of imaging.

2nd FWI inversion result







Modified misfit function

$$E = \frac{1}{2N} \left[(\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}})^T \mathbf{C}_D^{-1} (\mathbf{d}_{\text{pre}} - \mathbf{d}_{\text{obs}}) \right].$$

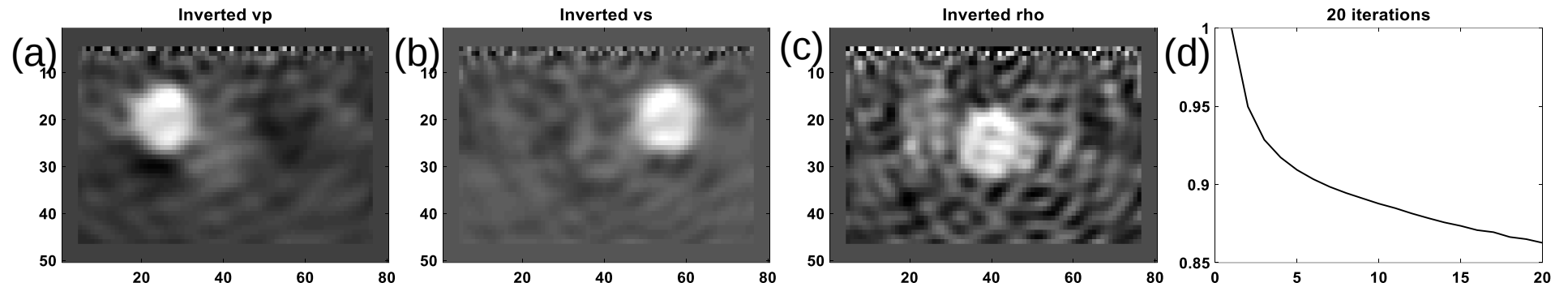
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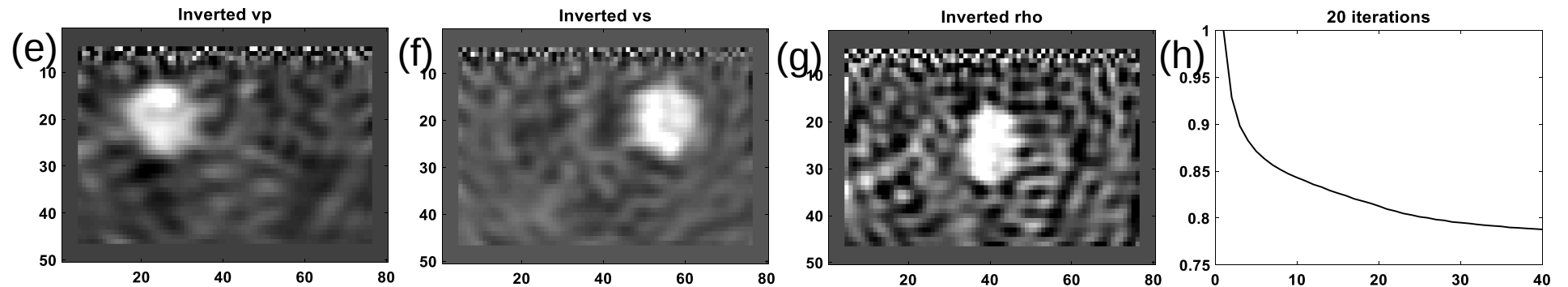
The artifacts generated by matching seismic data with **low SNR** will be **down weighted**, improving the quality of imaging.

Inversion results

1st inversion results



Reference results



New method results

