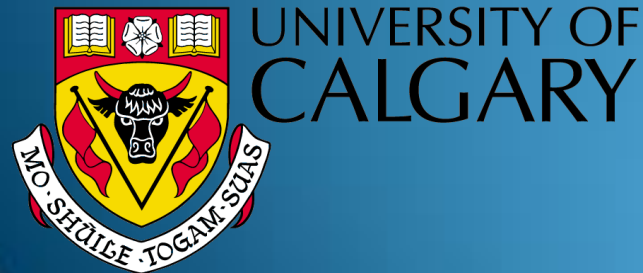


Recovering Low Frequencies For Impedance Inversion By Frequency Domain Deconvolution

Sina Esmaeili

Supervisor: Gary Frank Margrave



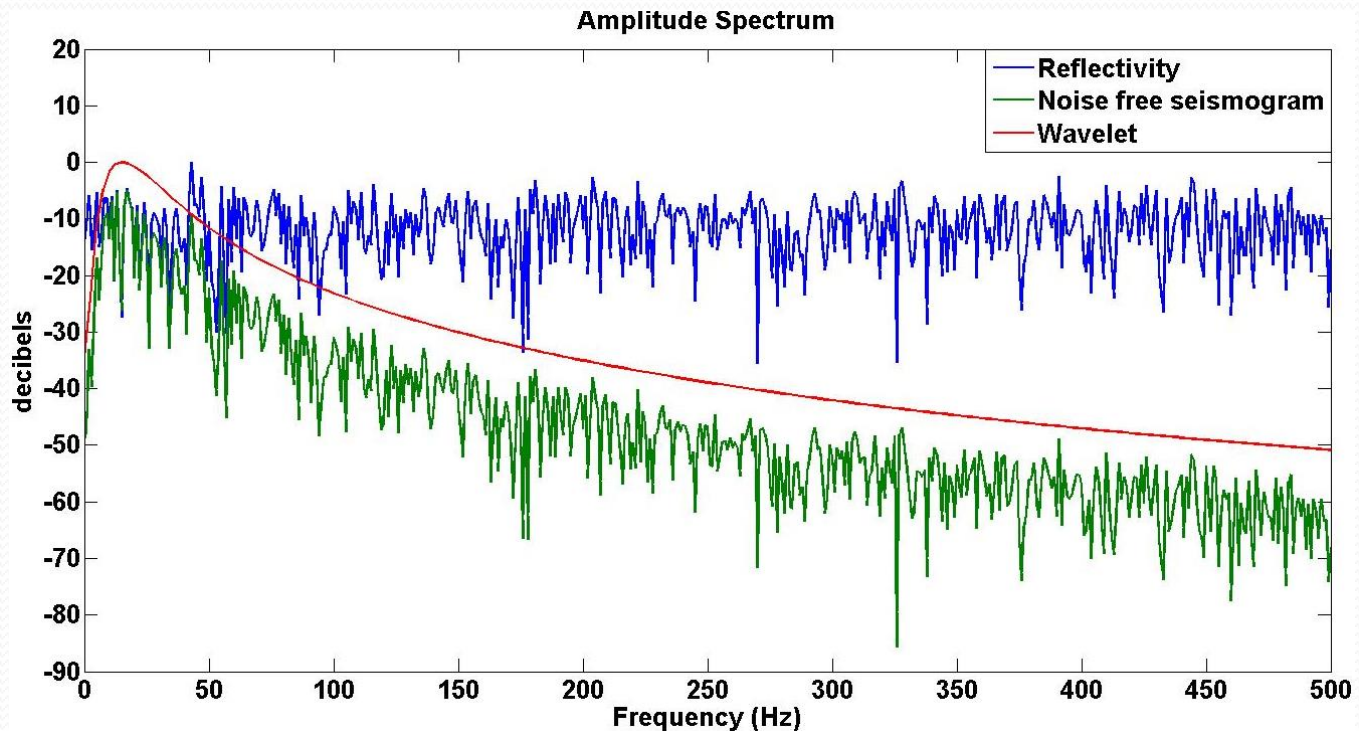
Outline

- Modeling of seismic trace
- Impedance inversion and estimation of reflectivity
- Frequency domain deconvolution
- Spectral white reflectivity estimation (random reflectivity)
- Spectral color reflectivity estimation (well log data)
- Impedance inversion results
- Conclusion

Modeling the Seismic trace

Convolutional Model: $s(t) = r(t) \bullet w(t)$

In frequency domain



Impedance Inversion

In homogenous medium for normal incident wavelet the reflection coefficient can be written as:

$$\rightarrow r_n = \frac{I_{n+1} - I_n}{I_{n+1} + I_n} \left\{ \begin{array}{l} I = v\rho \quad \text{Acoustic Impedance} \\ I_n \quad \text{Acoustic Impedance of layer } n^{\text{th}} \\ I_{n+1} \quad \text{Acoustic Impedance of layer } (n+1)^{\text{th}} \end{array} \right.$$

$$I_{n+1} = I_n \left(\frac{1+r_n}{1-r_n} \right) = I_1 \prod_{i=1}^n \left(\frac{1+r_i}{1-r_i} \right) \approx I_1 \exp \left(2 \sum_{i=1}^n r_i \right)$$

How to estimate the reflectivity?

$$S(t) = r(t) \bullet w(t)$$

We can define $d(t)$ as:

“d” is the perfect
deconvolution operator

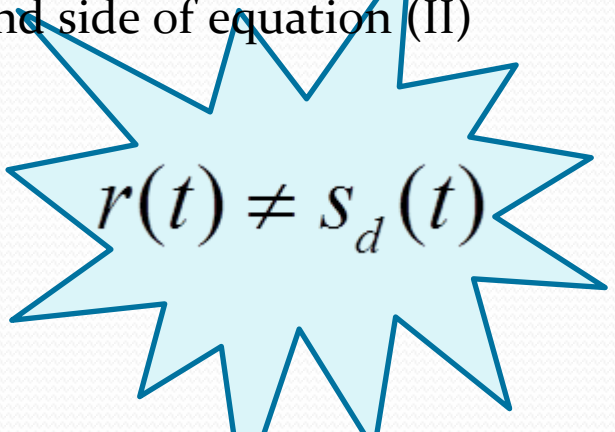
$$w(t) \bullet d(t) = \delta(t) \quad \longrightarrow \quad d(t) = w^{-1}(t) \quad (\text{II})$$

$$r(t) = s(t) \bullet d(t)$$

In practice, because of the **band-limited** nature of the wavelet, it is impossible to find $d(t)$ which makes the right hand side of equation (II) equal to a delta function.

$$w(t) \bullet d(t) = w_d(t)$$

$$s_d(t) = r(t) \bullet w_d(t)$$


$$r(t) \neq s_d(t)$$

Deconvolution

- In mathematics, deconvolution is an algorithm-based process which is used to reverse the effects of convolution on the recorded data
- Deconvolution must estimate the wavelet from the data
- Amplitude spectrum of wavelet estimated by smoothing data amplitude spectrum.
- All spectral shape is attributed to the wavelet (white reflectivity assumption)

Frequency Domain Deconvolution

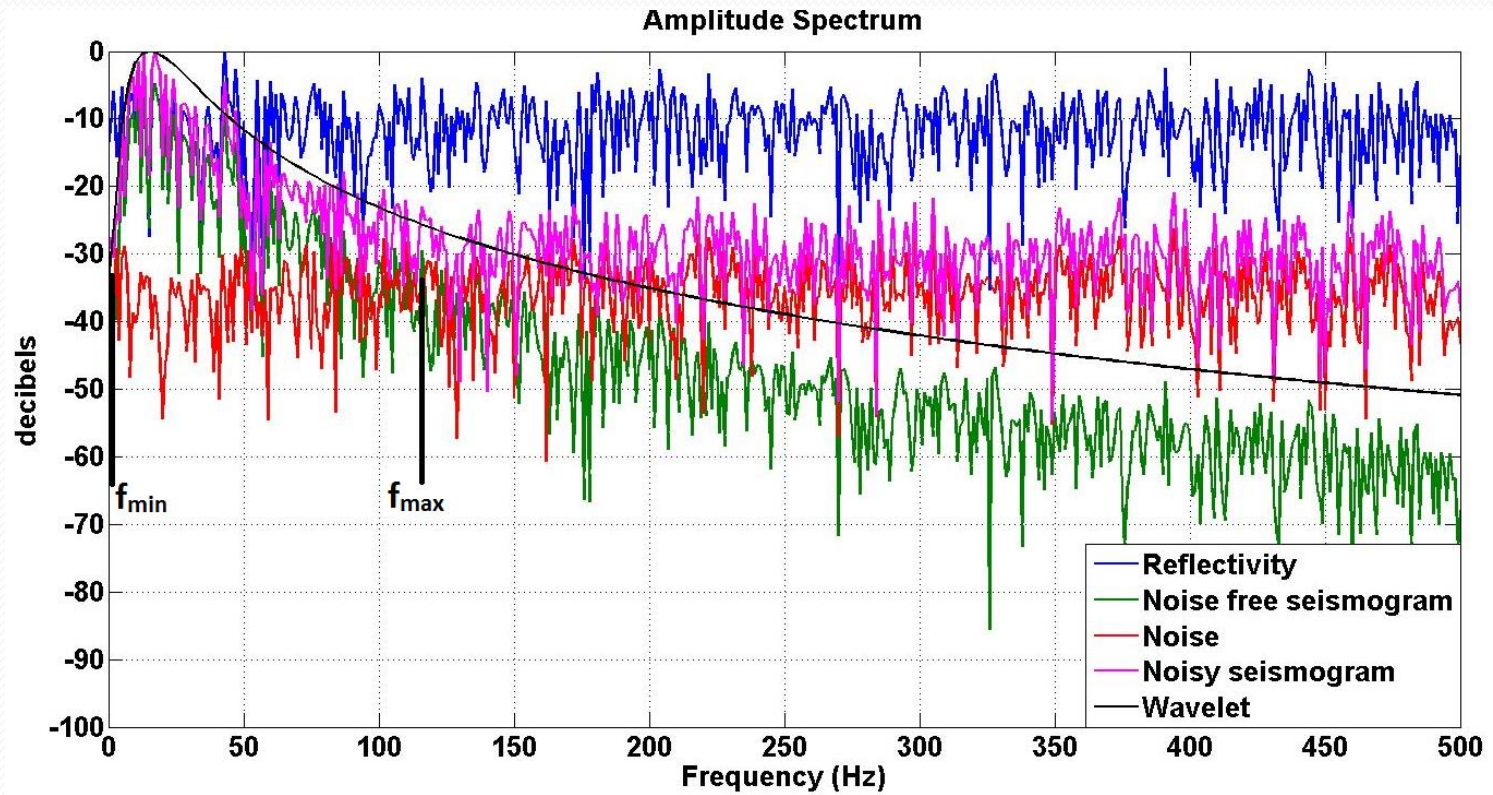
The assumptions:

- The wavelet should be minimum phase
- The wavelet spectrum should be smooth
- The wavelet should be stationary
- The reflectivity should be random so it has a white spectrum

Adding the noise to seismogram:

$$s(t) = r(t) \cdot w(t) + n(t)$$


$$S(f) = R(f)W(f) + N(f)$$



$$f_{\min} \leq f \leq f_{\max}$$

$$\underbrace{|S(f)|}_{\text{Noisy Seismogram}} \approx \underbrace{|R(f)||W(f)|}_{\text{Noise free Seismogram}}$$

The white reflectivity means: $\overline{|R(f)|} \approx 1$

$$\overline{|S(f)|} \approx |W(f)|$$


Frequency Domain Deconvolution

Then the estimated wavelet is defined as:

By comparing with:

$$\left. \begin{aligned} |w(f)|_{estimated} &= \overline{|S(f)|} \\ r(t) &= s(t) \cdot w^{-1}(t) \end{aligned} \right\} |D(f)| = |W(f)|_{estimated}^{-1} = \overline{|S(f)|}^{-1}$$

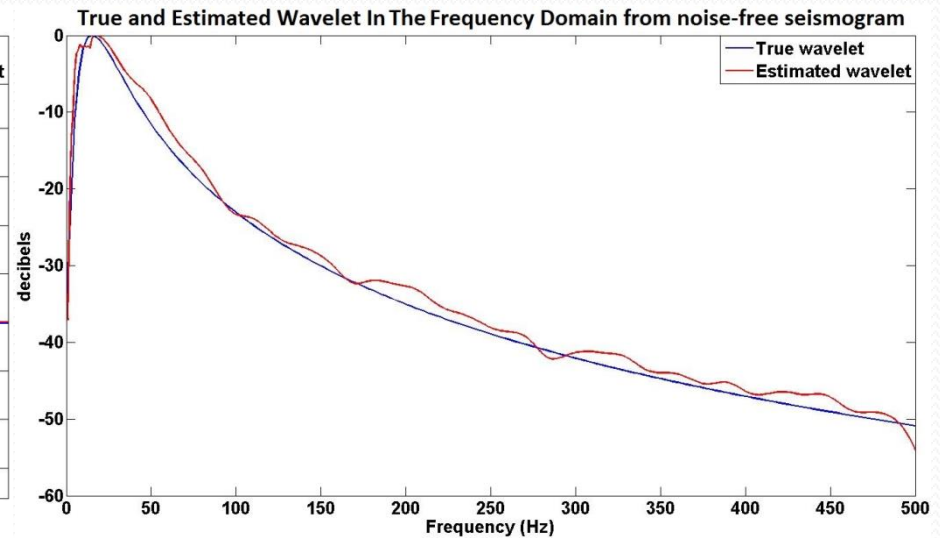
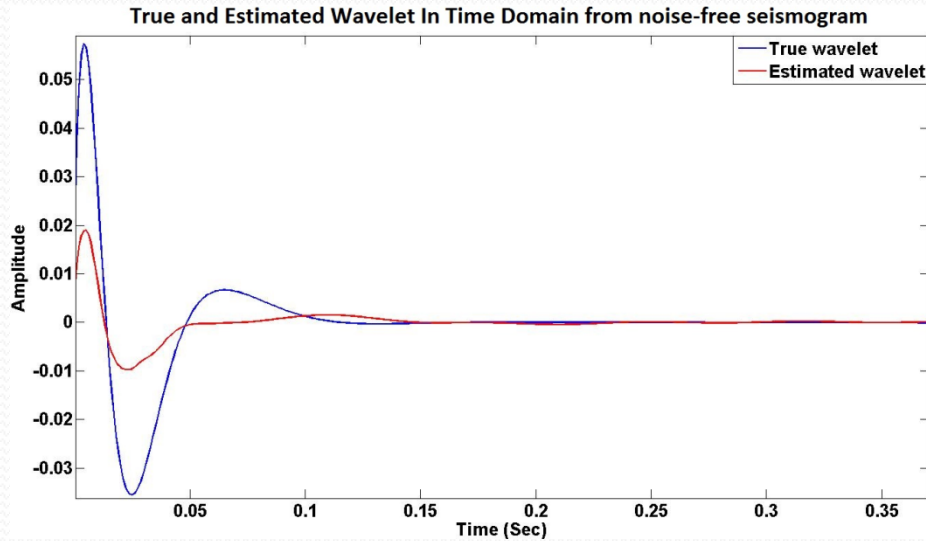
Deconvolution operator can be defined as (Margrave, 2002):

$$D(f) = \frac{1}{|W(f)|_{est} + \mu A_{max}} e^{i\phi_D(f)} \left\{ \begin{aligned} A_{max} &= \max(|W(f)|_{est}) \\ \phi_D(f) &= H(\ln(|D(f)|)) \\ 0.01 &< \mu < 0.000001 \end{aligned} \right.$$

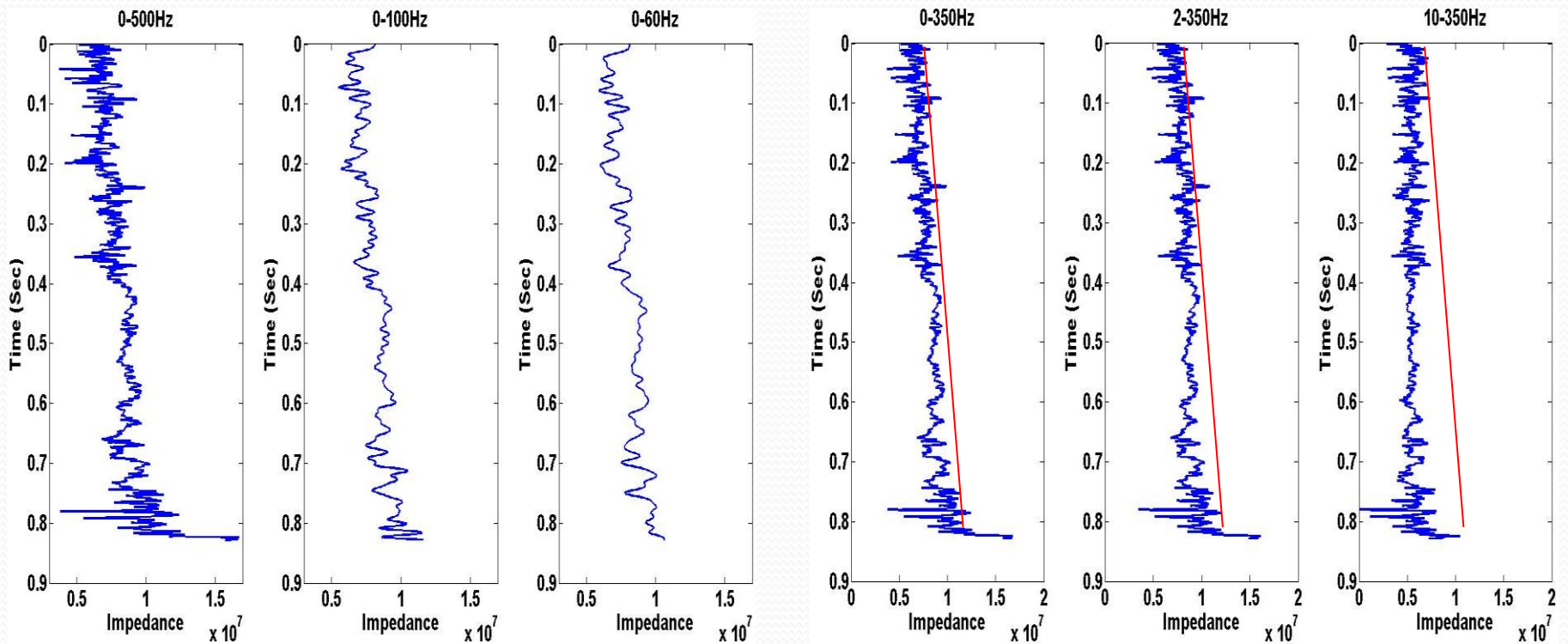
Estimated wavelet

$$S(f)D(f) = R(f)W(f)D(f) = R(f)W_D(f)$$

$$W_D(f) = W(f)D(f) = \frac{W(f)}{|W(f)|_{est} + \mu A_{max}} e^{i\phi_D(f)}$$



Missing High and Low Frequencies

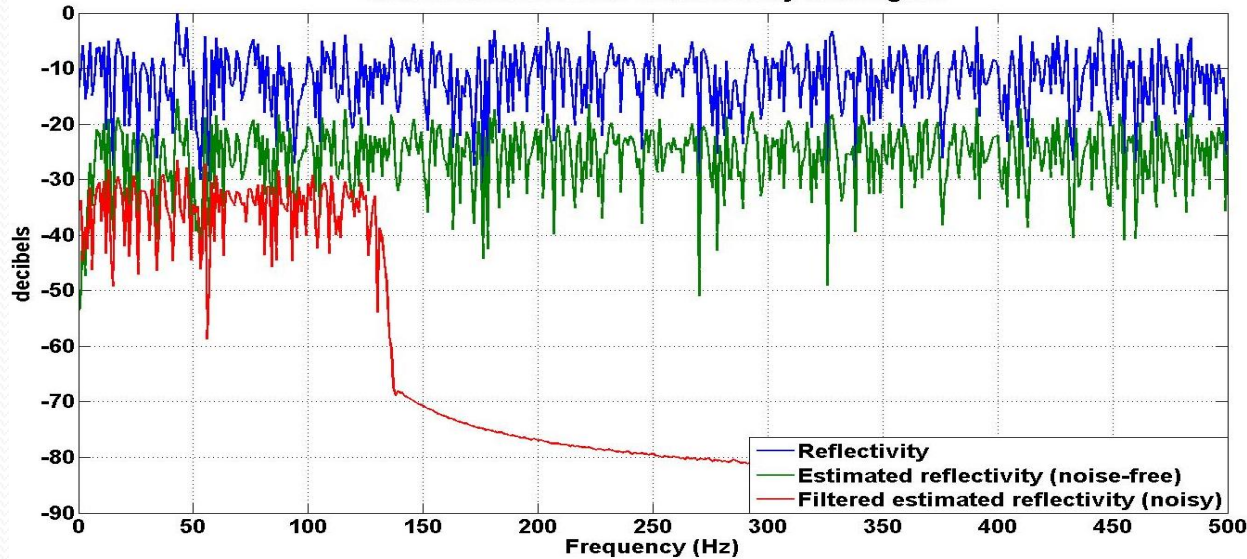


Missing high frequency causes
reduced resolution

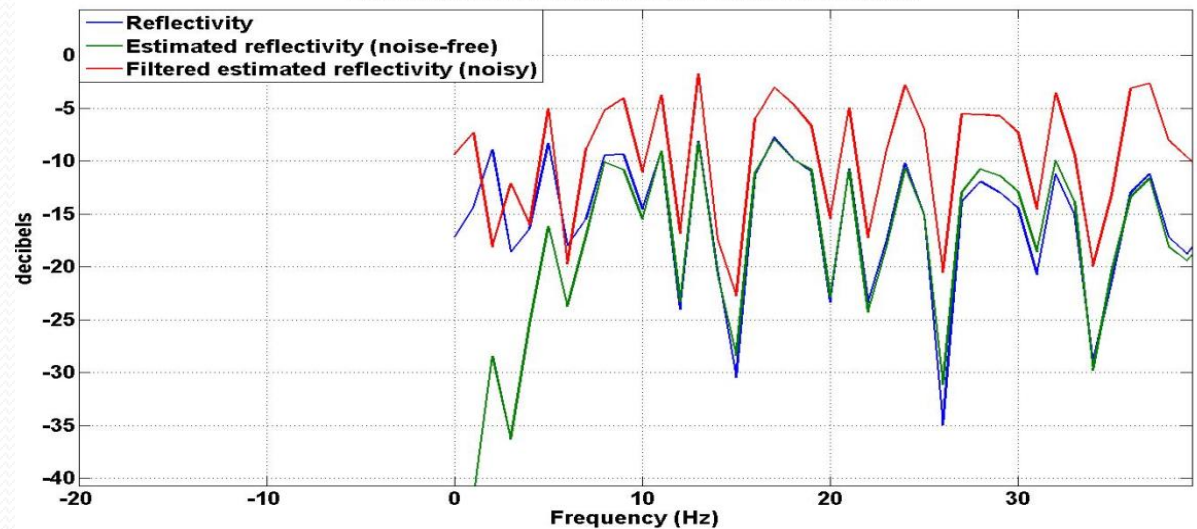
Missing low frequency causes
reduced trend of data

Frequency Domain Deconvolution (Old Version)

Deconvolution of noise free and noisy seismogram

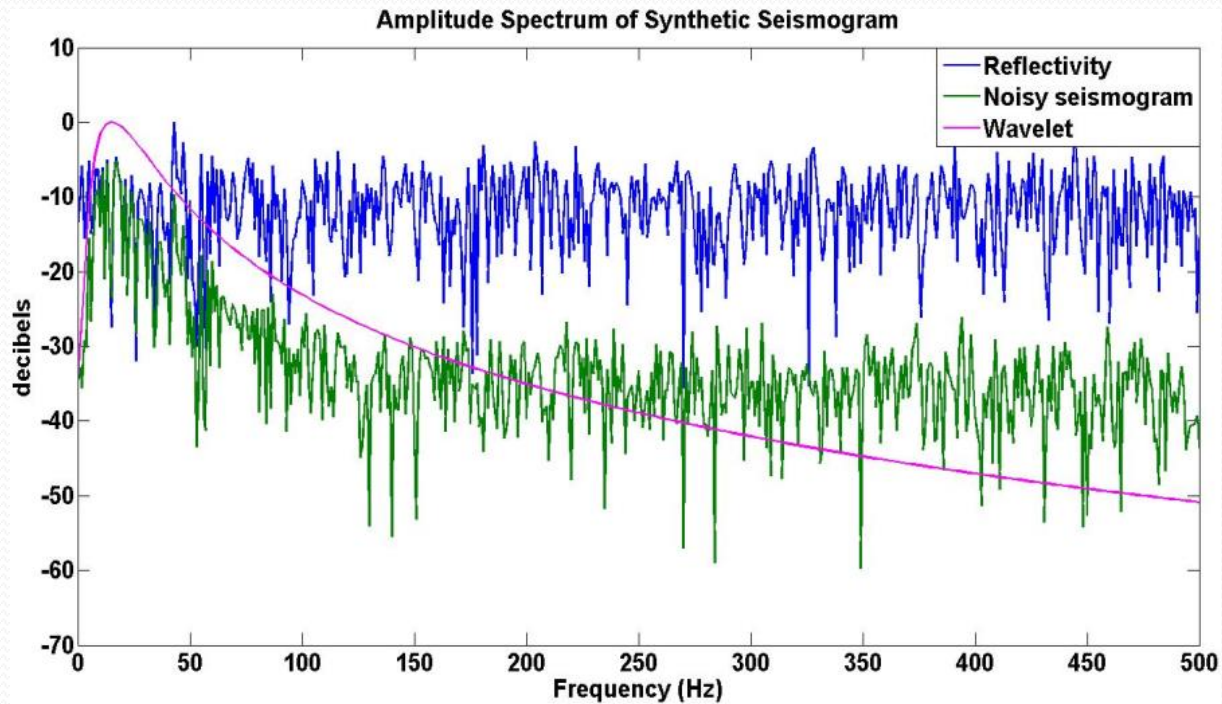


Deconvolution of noise free and noisy seismogram



How possible to improve?

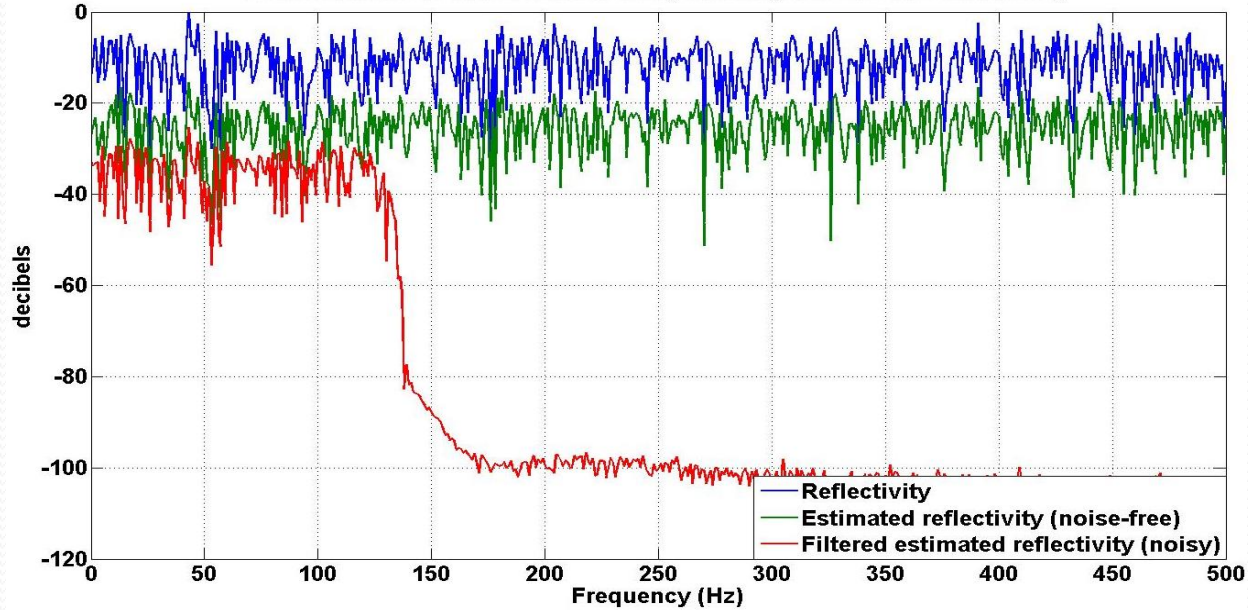
- Improving the process of smoothing amplitude spectrum of seismic trace for estimating wavelet



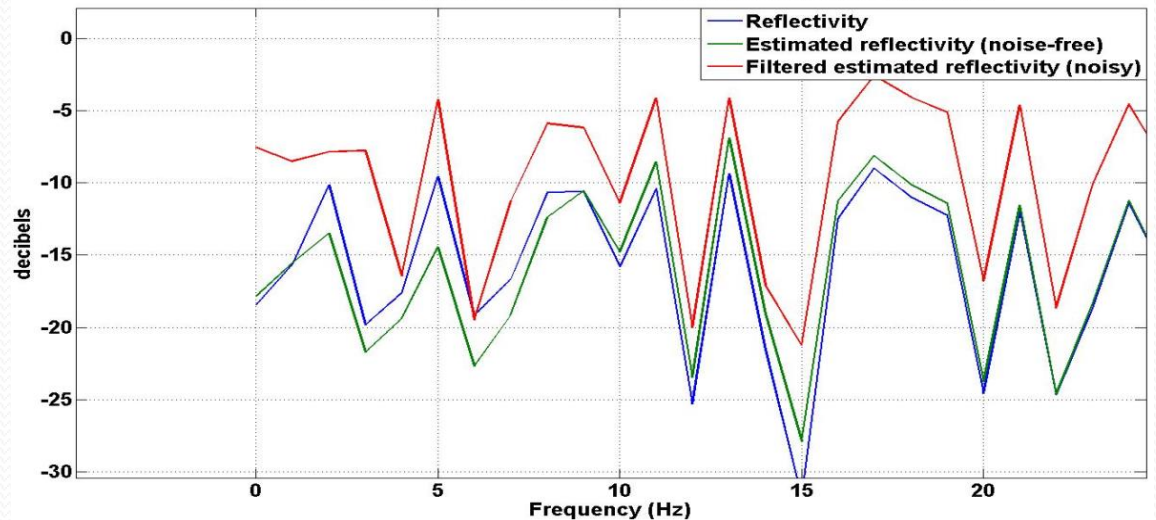
- Applying spectral color operator to Deconvolved data

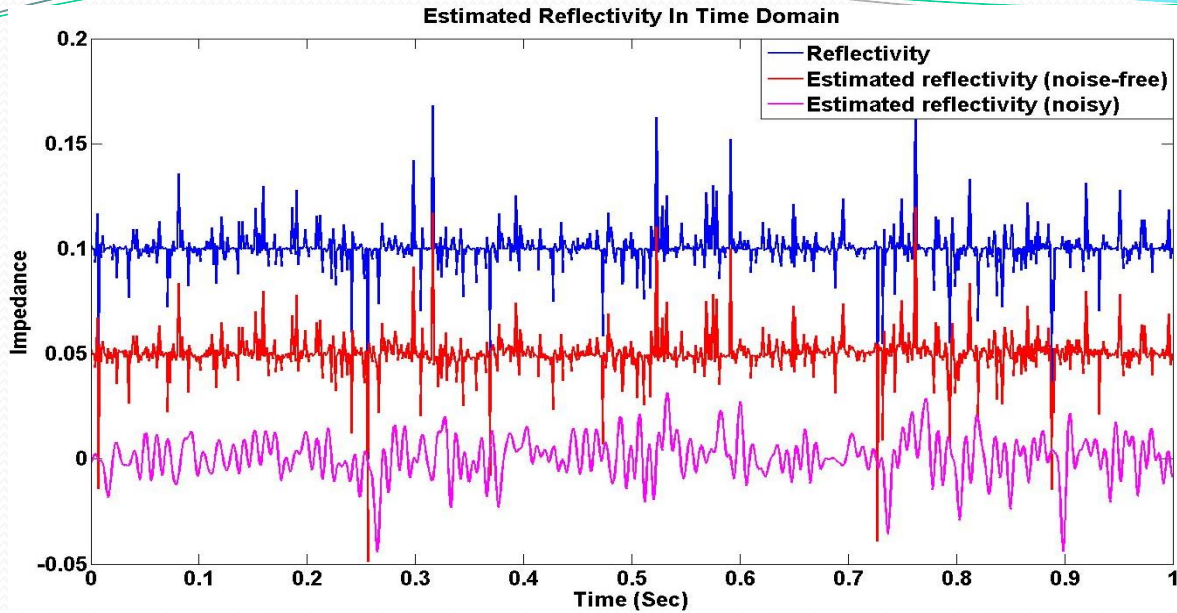
Deconvolution with new smoothing

Deconvolution of noise free and noisy seismogram with new smoothing



Deconvolution of noise free and noisy seismogram with new smoothing



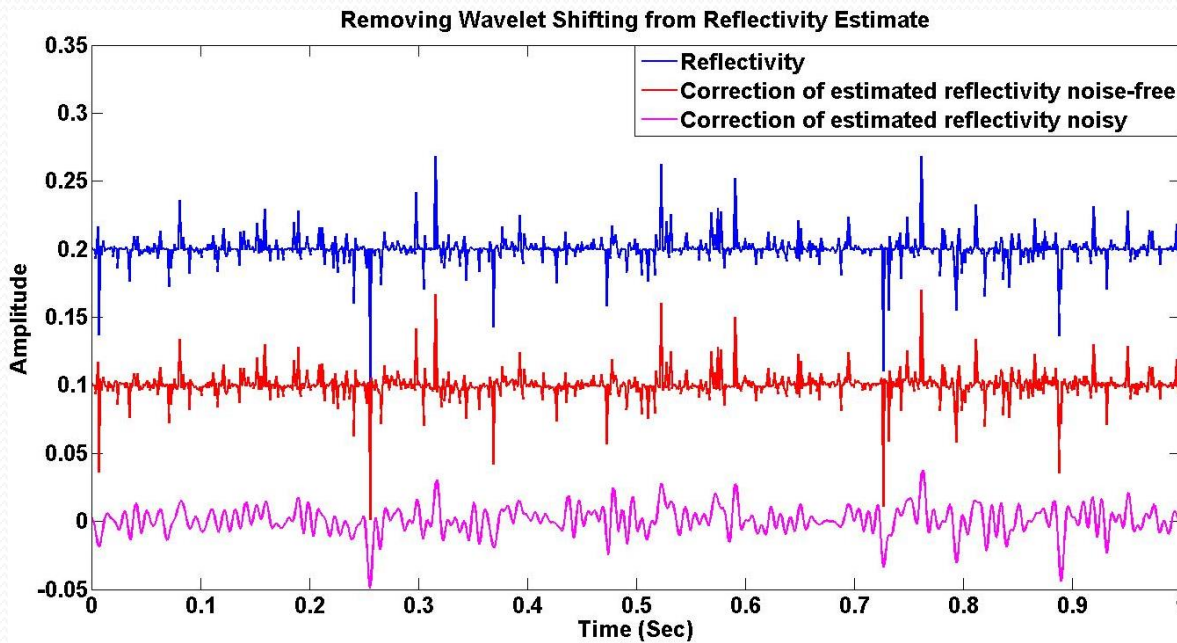


Maximum Correlation = 0.9865

Lag = 0.0000

Maximum Correlation = 0.3950

Lag = 9.5000



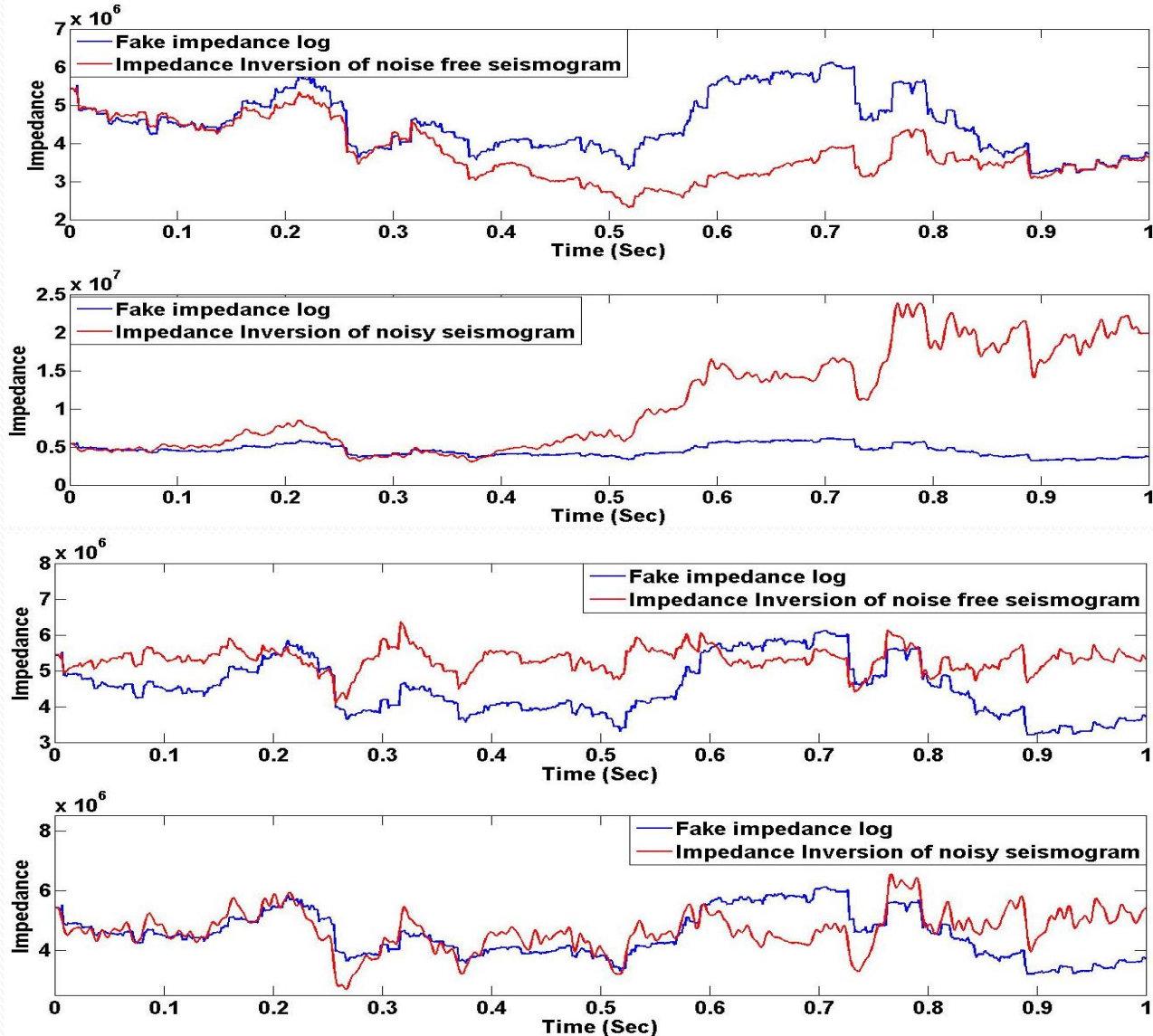
Maximum Correlation = 0.9865

Lag = 0.0000

Maximum Correlation = 0.4900

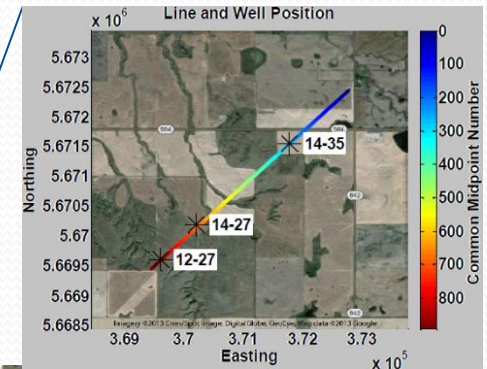
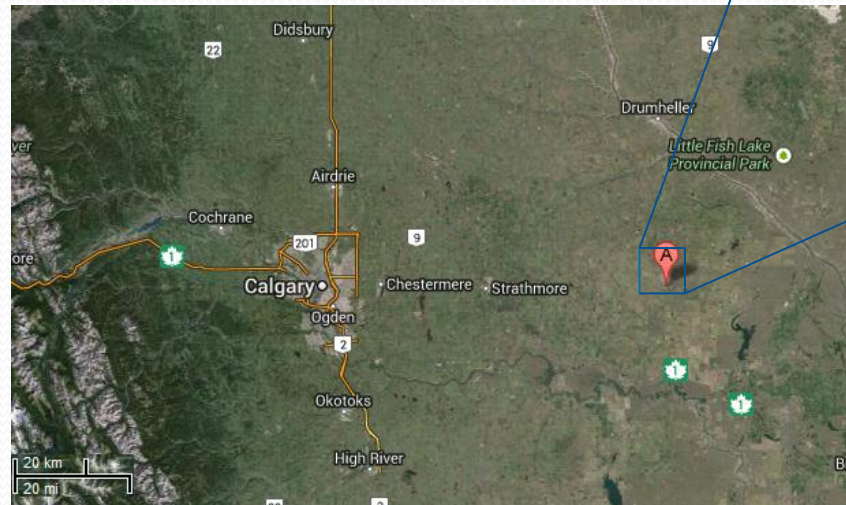
Lag = 0.5000

Inverted Impedance for random reflectivity



Real well log data

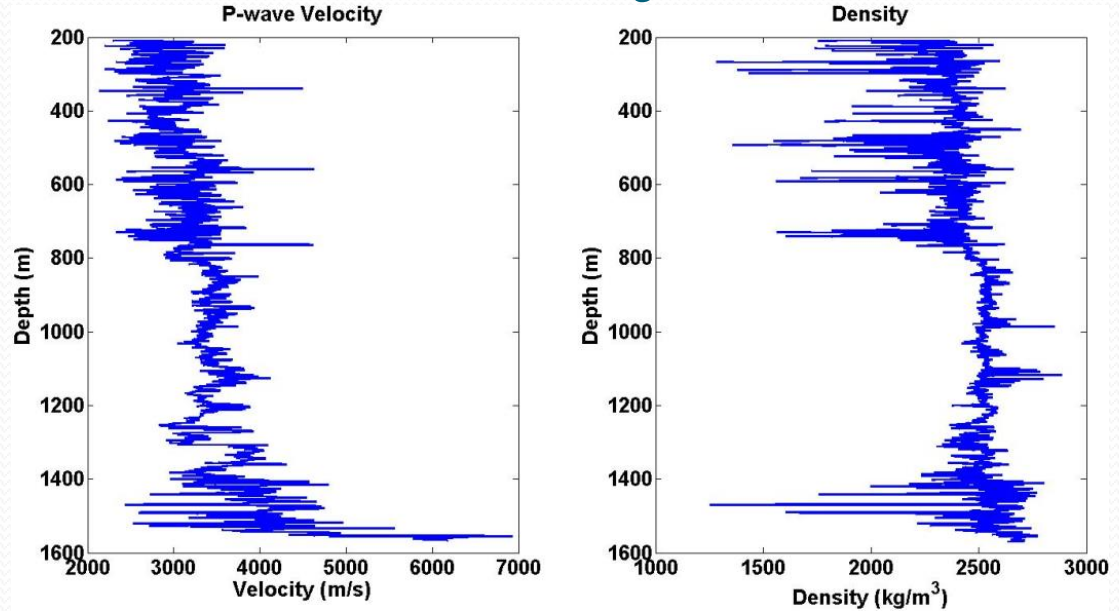
- The Real Data: Sonic and Density logs from the well 12-27 near Hussar



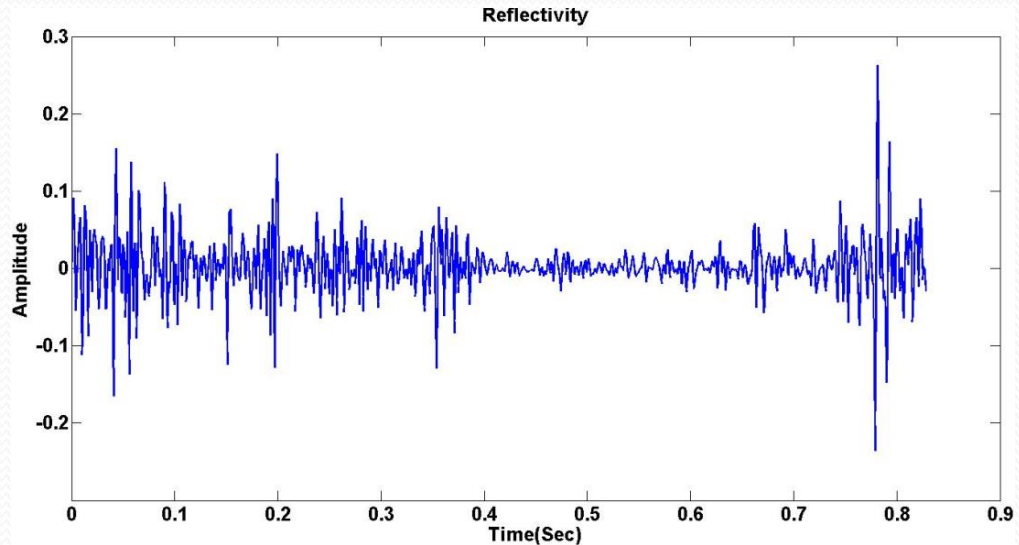
Lloyd, Heather, 2013

Location of the well 12-27 area near Hussar, Alberta, Canada indicated by the red marker

Spectral color reflectivity

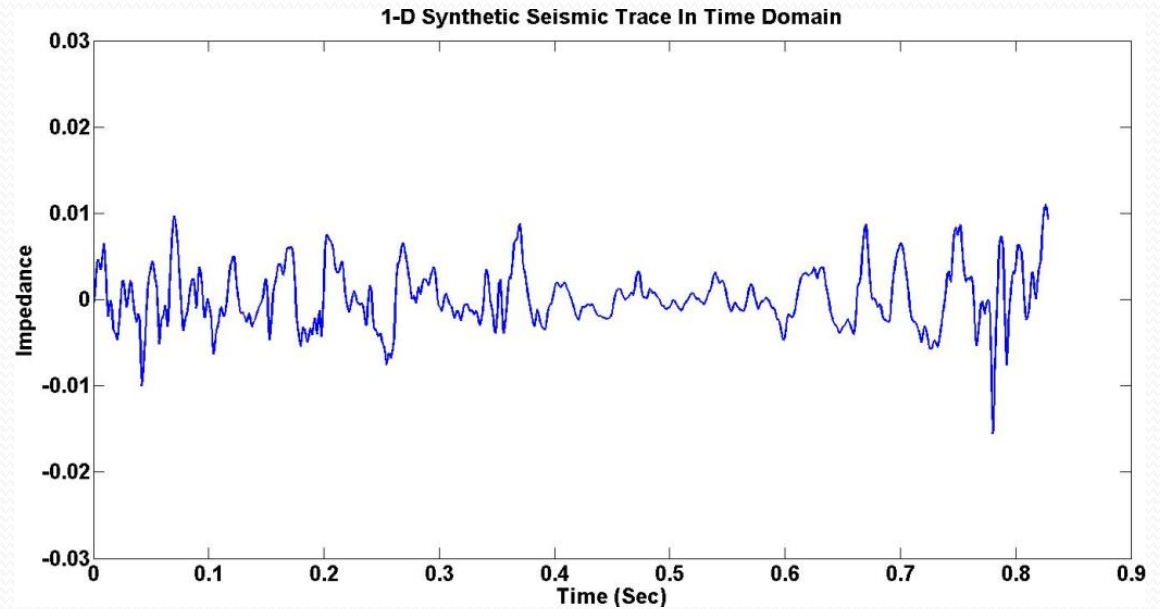


$$r_n = \frac{\rho_{n+1}v_{n+1} - \rho_n v_n}{\rho_{n+1}v_{n+1} + \rho_n v_n}$$

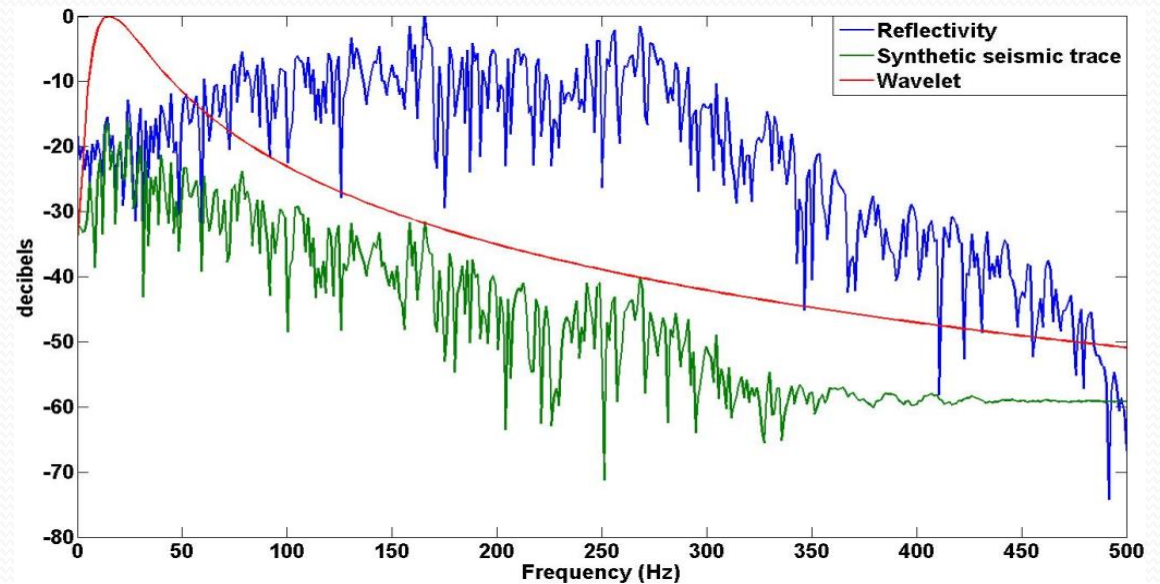


The Synthetic Seismic Trace

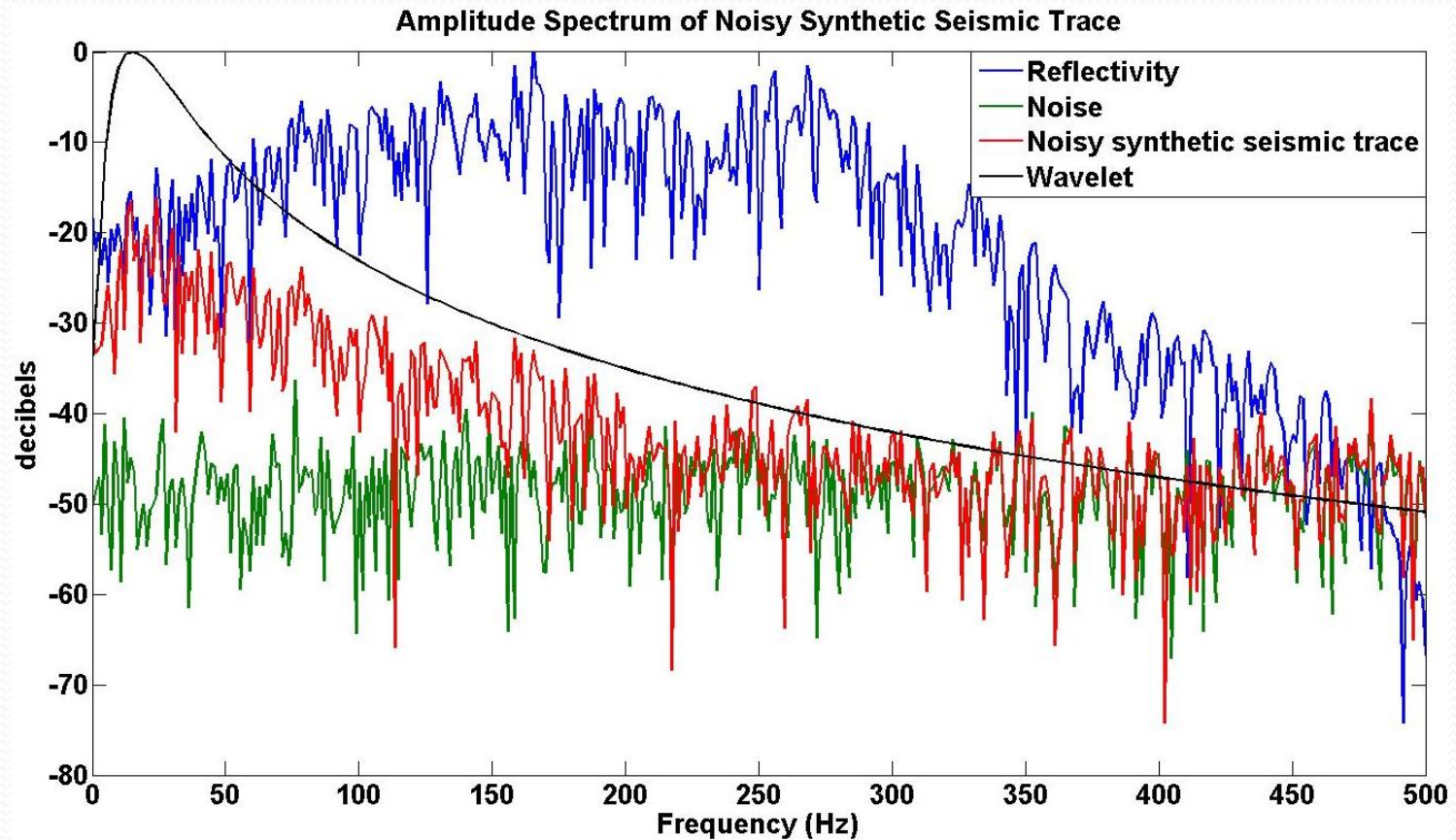
In time domain:



In frequency domain:

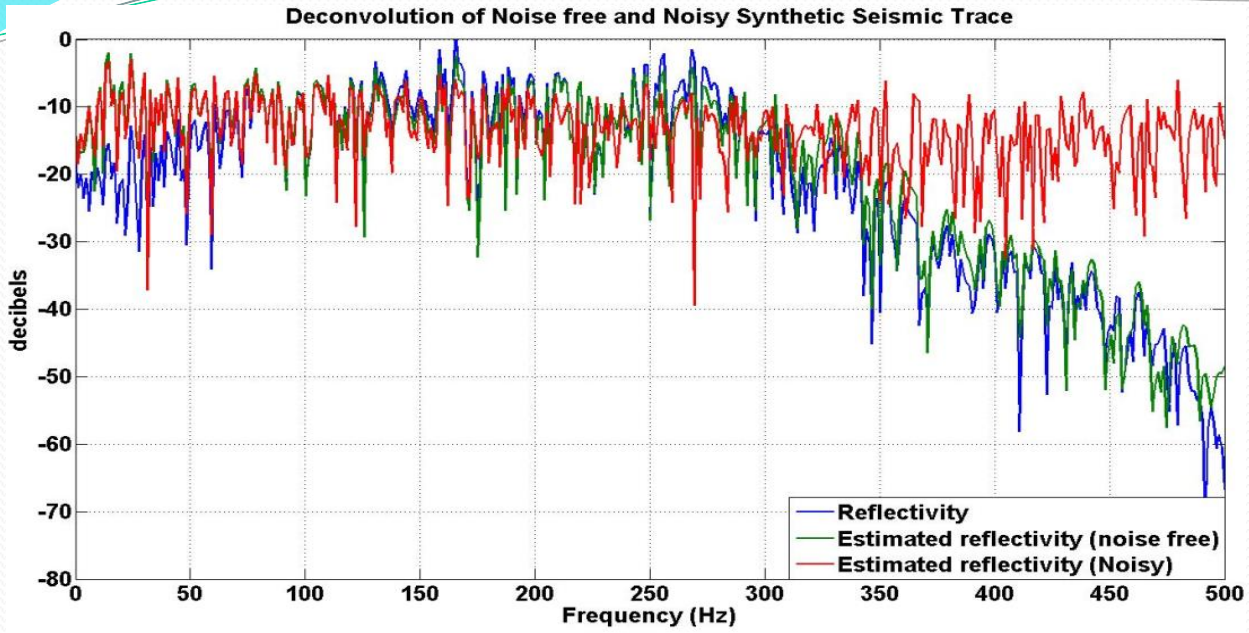


Adding Noise To Data



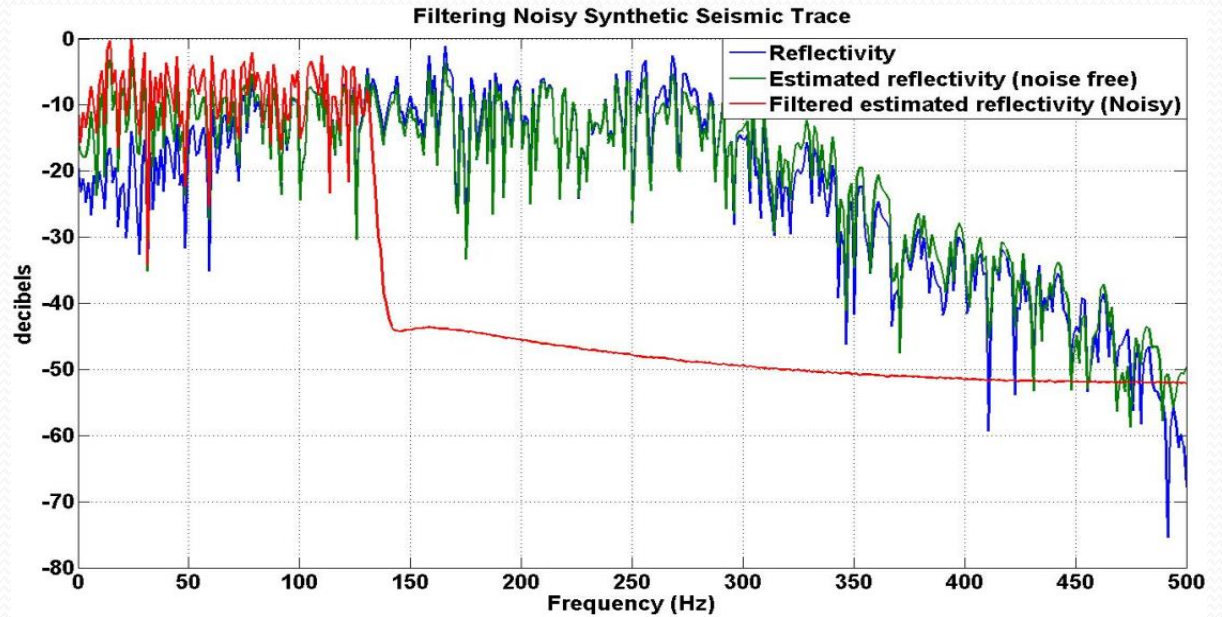


DECONVOLUTION

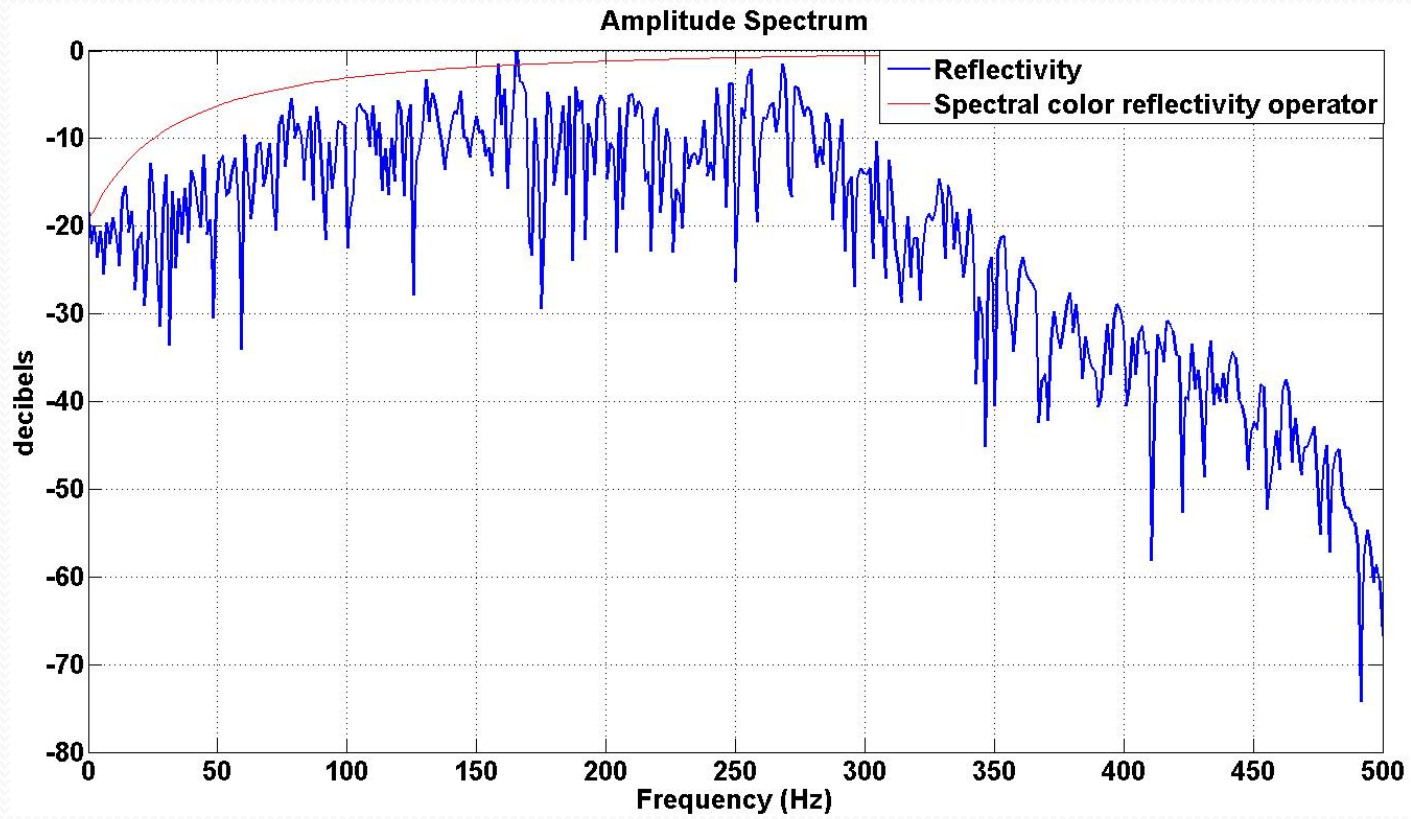


Deconvolution

Filtering

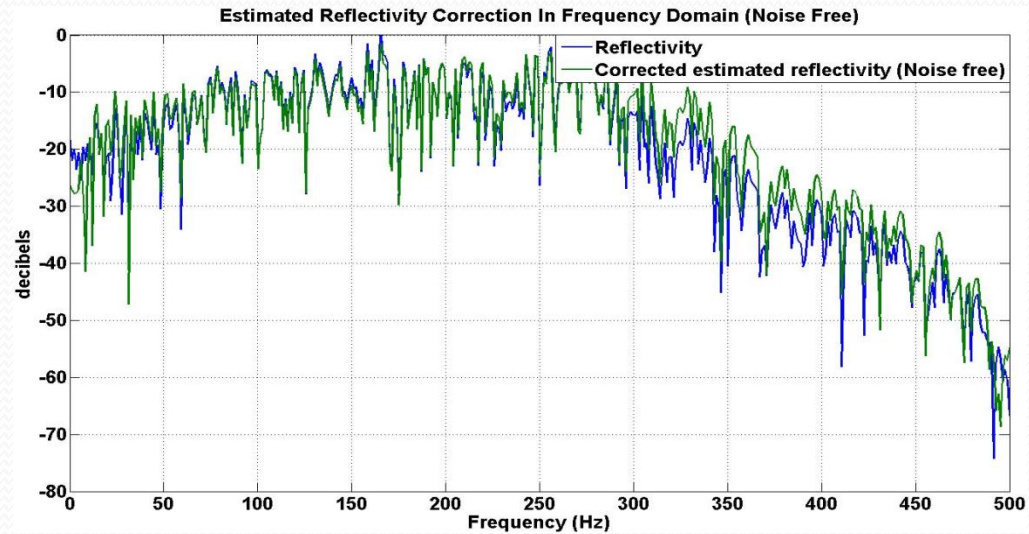


Spectral Color Operator

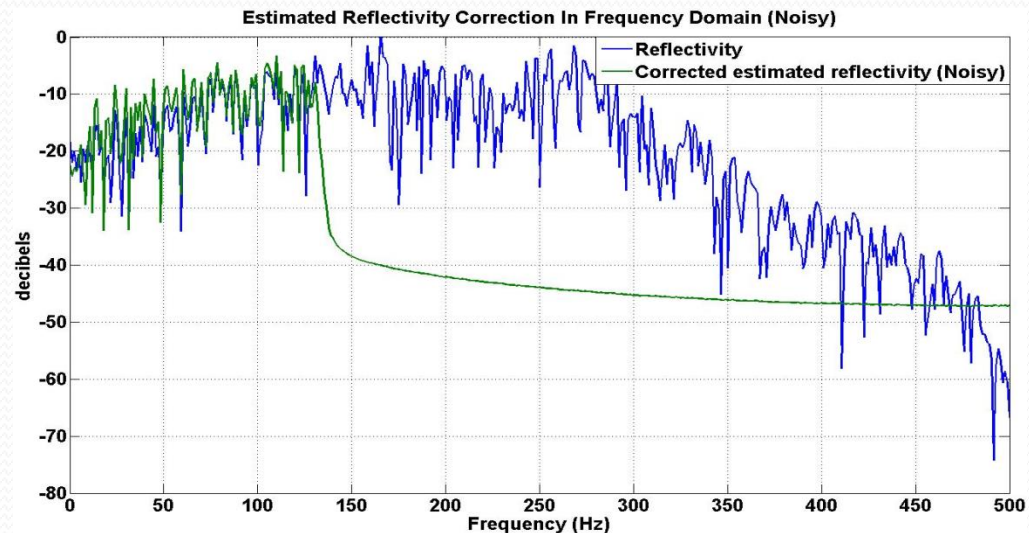


Applying Operator To Data

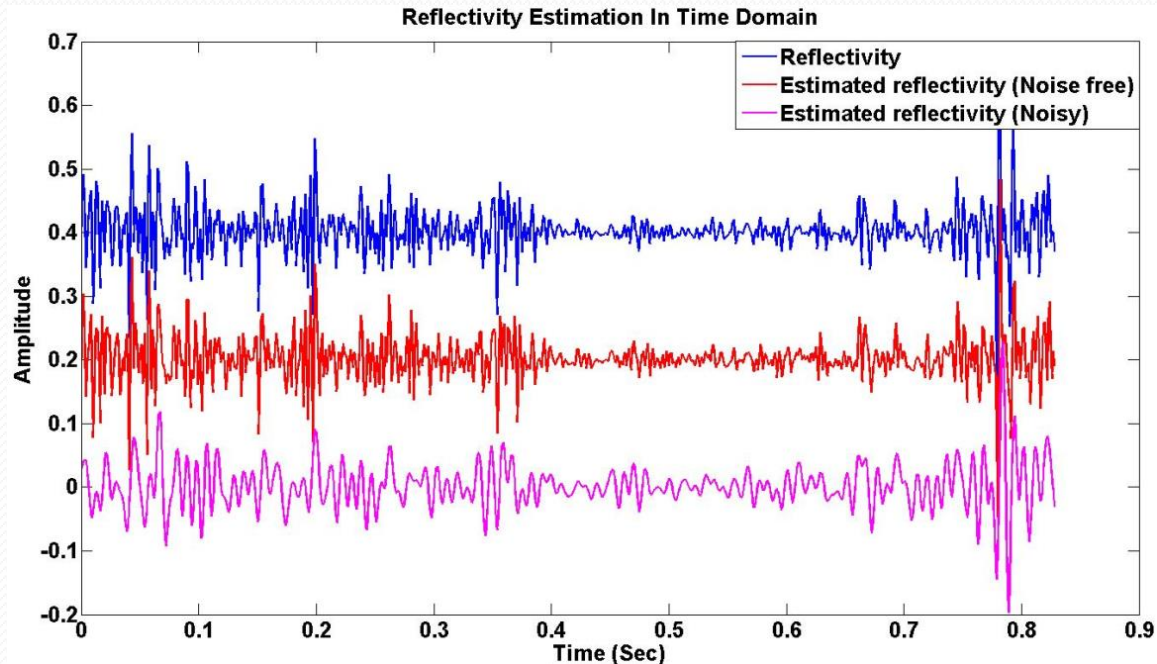
Noise free case:



Noisy case:

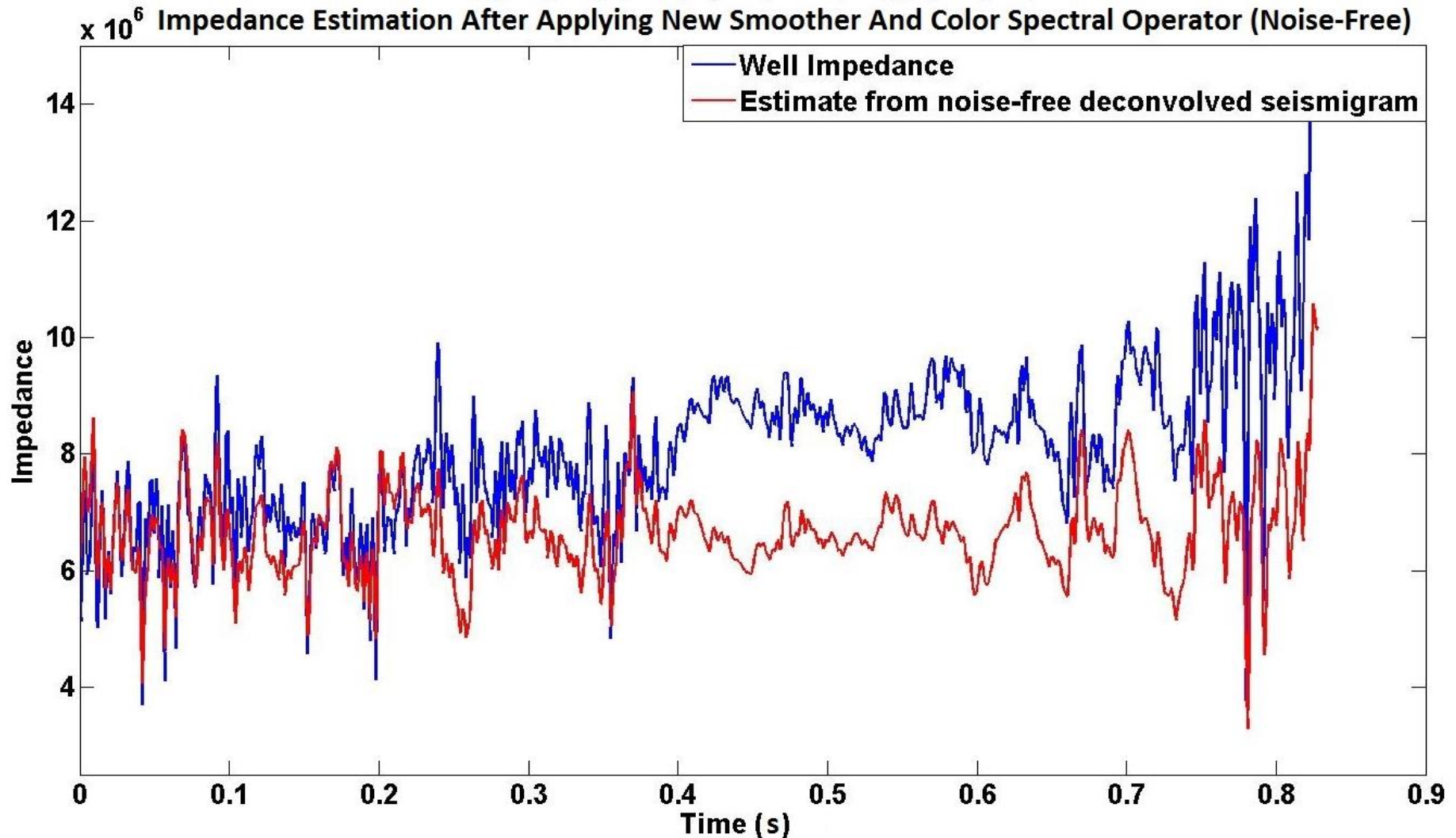


Final Results In Time Domain

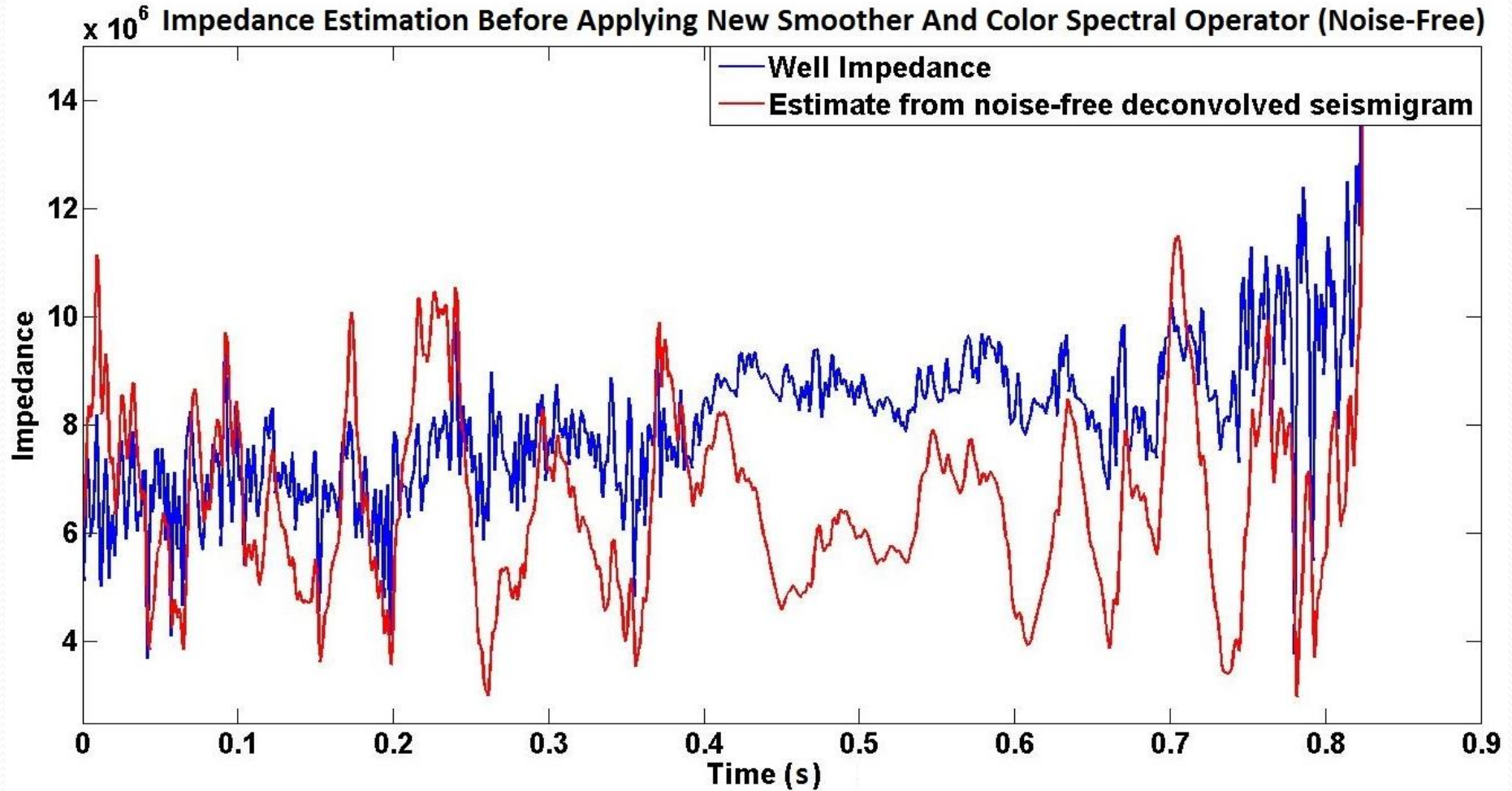


| | Frequency Domain Deconvolution (Old Version) | Frequency Domain Deconvolution (New Version) |
|--|--|--|
| Final estimated reflectivity (Noise Free) | Maximum Correlation = 0.8668 Lag = 0.2000 | Maximum Correlation = 0.8912 Lag = 0.3000 |
| Final estimated reflectivity (Noisy) | Maximum Correlation = 0.3198 Lag = 1.4000 | Maximum Correlation = 0.5698 Lag = 0.5000 |

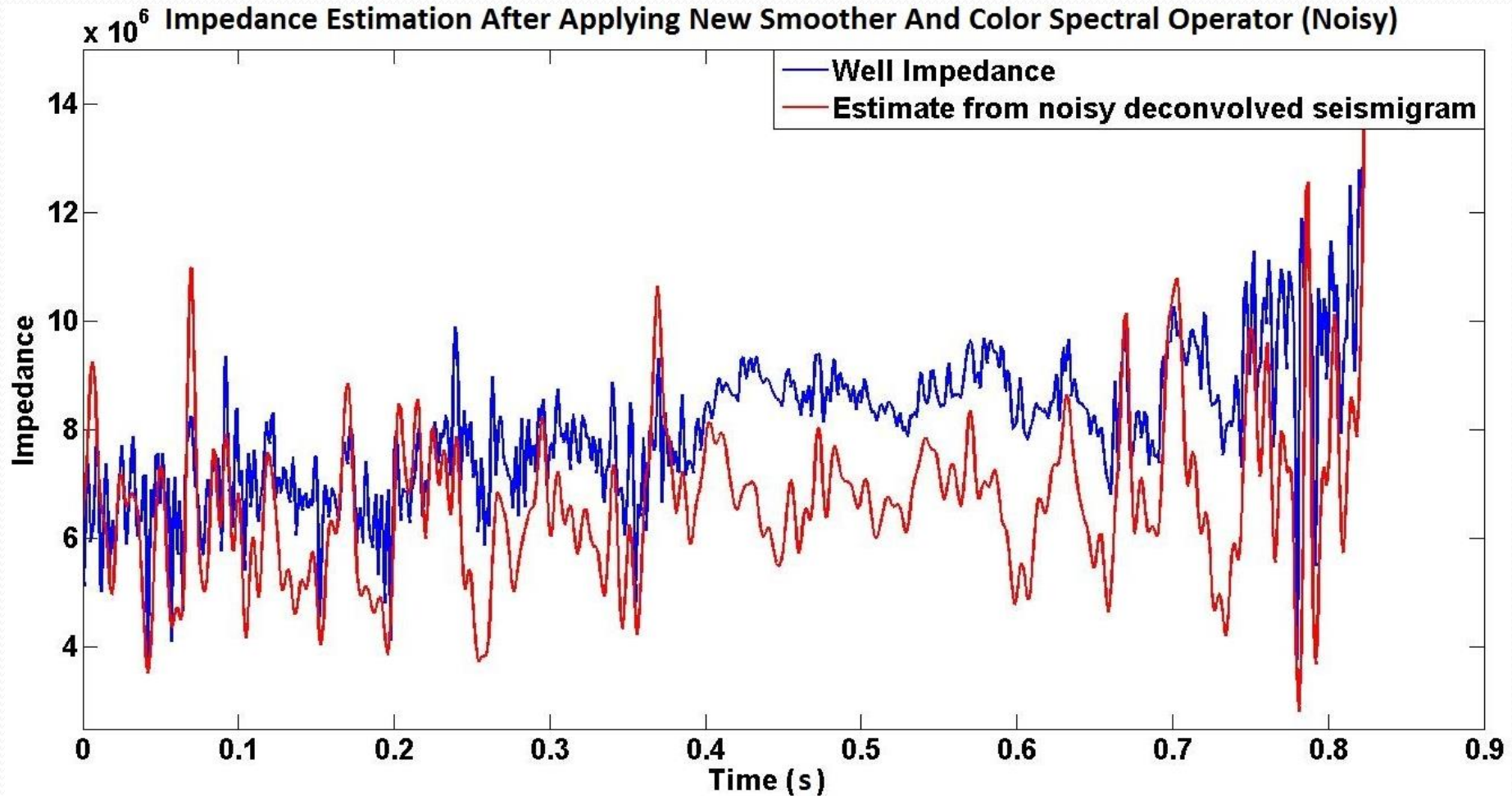
Impedance Inversion for noise-free seismogram



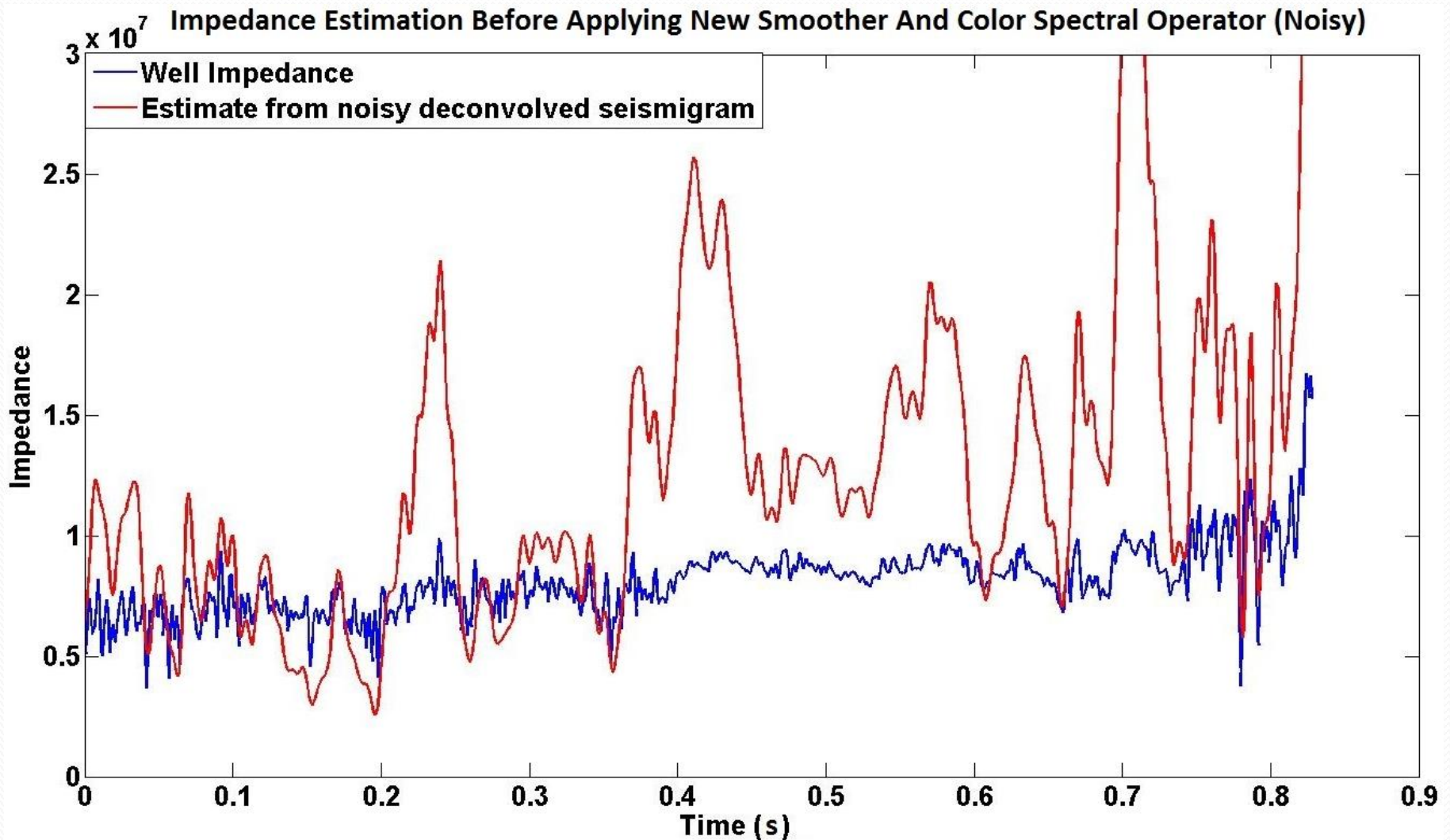
Impedance Inversion for noise-free seismogram



Impedance Inversion for noisy seismogram



Impedance Inversion for noisy seismogram



Conclusion

- The bandlimited nature of wavelet and also noise contamination causes the missing low and high frequency in recorded data.
- The better seismic data smoothing, the more realistic reflectivity estimation we can reach.
- The colored spectrum of low frequencies data could be recovered by spectral color operator.
- Acoustic impedance can be estimated much precisely if the low frequency can be recovered from seismic data.
- For the future work real recorded seismic data will be used

Acknowledgement

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- All the students and staffs in CREWES