

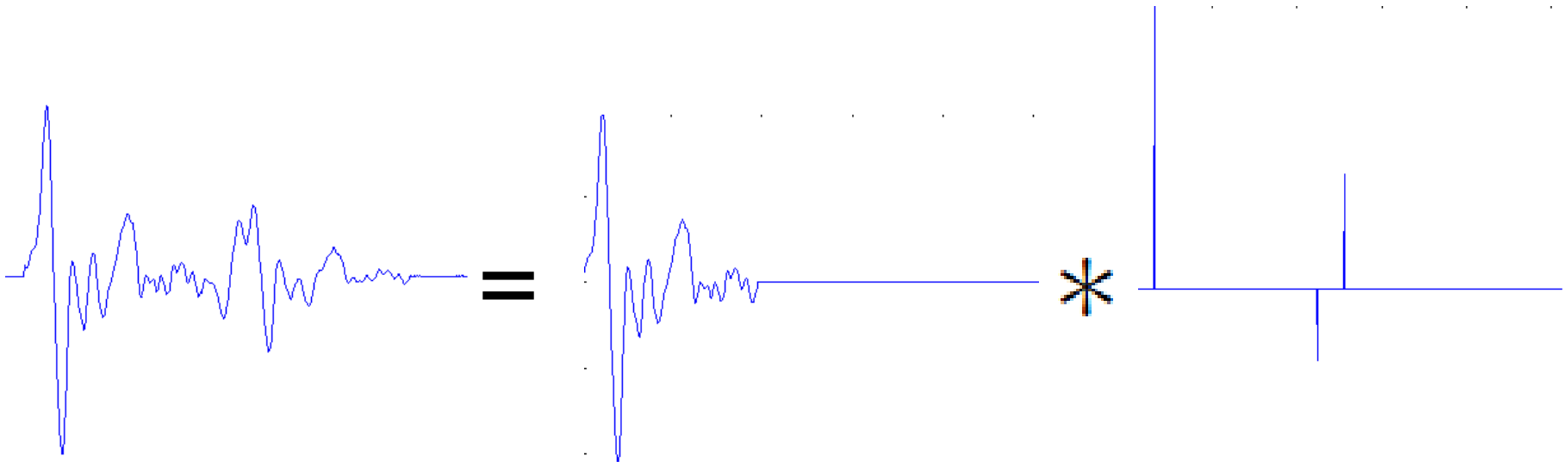
Deconvolution

$$u(t) = s(t) * g(t)$$

$u(t)$ – Recorded seismogram

$s(t)$ – source function

$g(t)$ – receiver function



Produce synthetic seismograms from a synthetic receiver function convolved with P-Codas obtained from windowed P waves of actual data.

Deconvolution

- **Aim:** Produce $g(t)$ by deconvolving source functions from multiple seismograms.
- Construct the problem as an optimization problem.
- Seek a solution which is sparse in the wavelet domain.
- \mathbf{X} is the wavelet transform of $g(t)$
- \mathbf{y} is a vector of seismograms in the time domain.

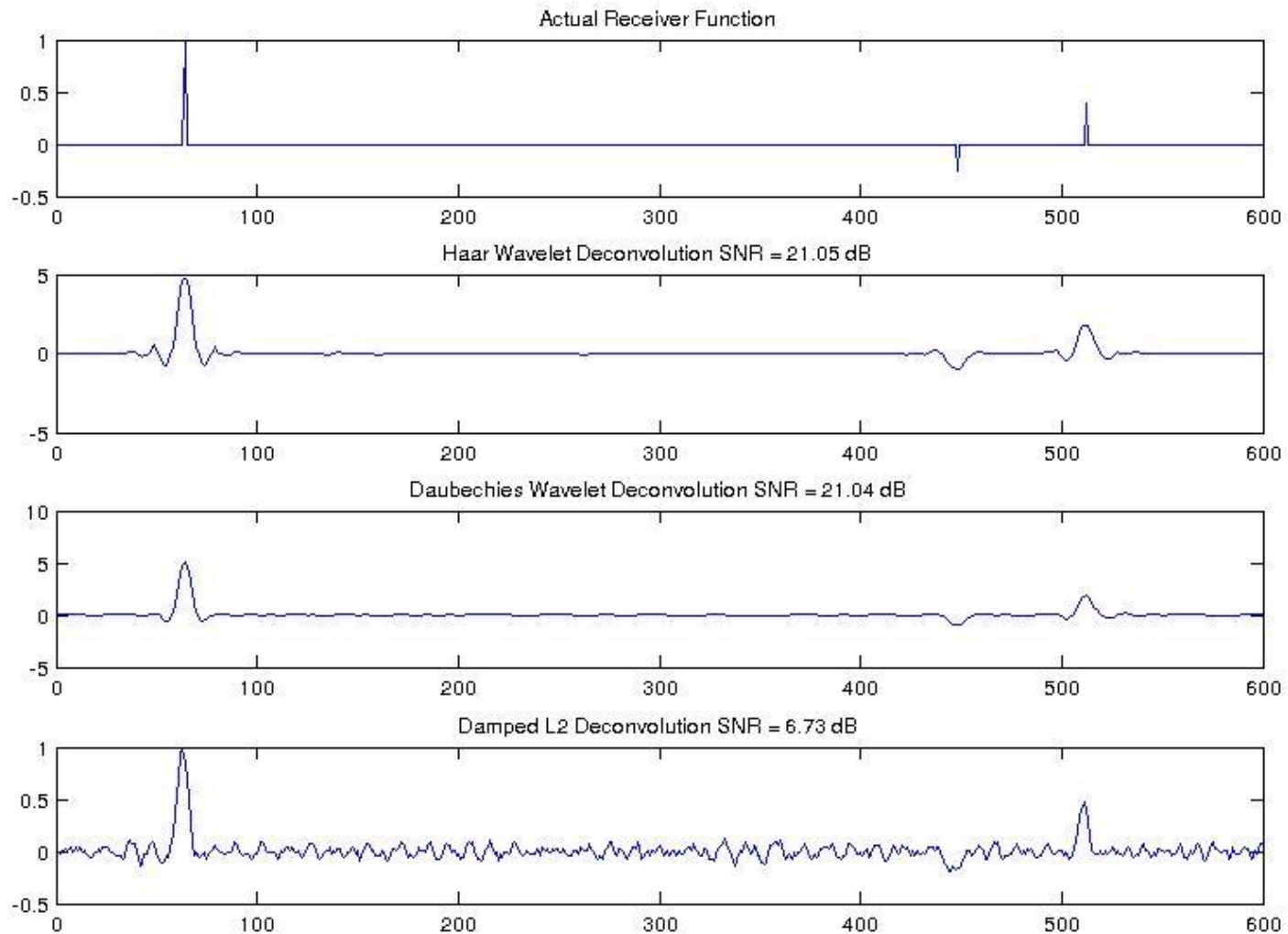
$$\mathbf{y} = \mathbf{A}\mathbf{x}$$

- Seek to minimize $\|\mathbf{x}\|_1$ s.t. $\|\mathbf{A}\mathbf{x} - \mathbf{y}\|_2 \leq \sigma$

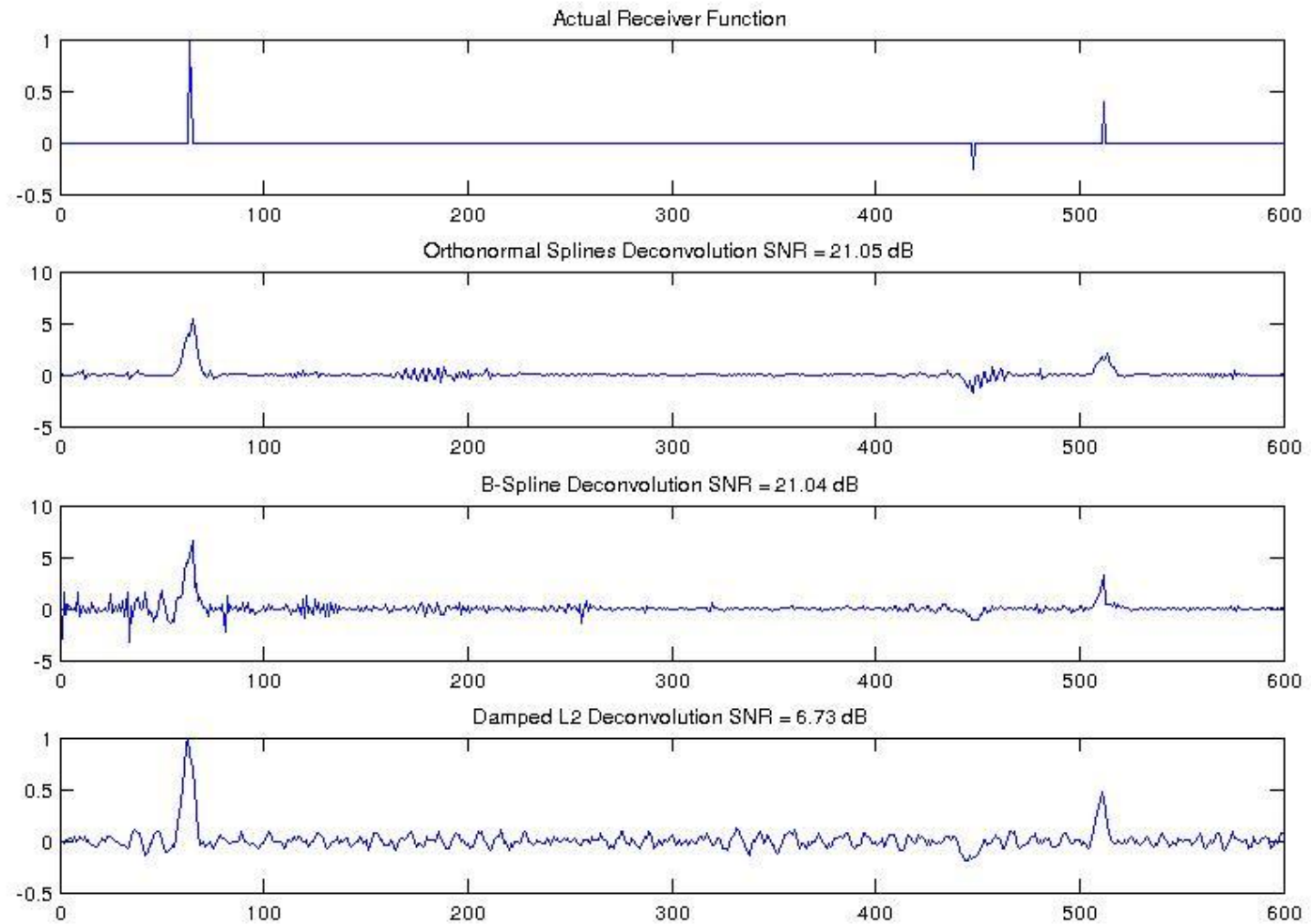
$$\mathbf{y} = \underbrace{[\mathcal{F}^T \mathcal{F}(s) \mathcal{F}(g)] \mathcal{W}^T}_{\text{Convolution operator}} \mathbf{x}$$

Convolution operator

Synthetic Tests – Standard wavelets

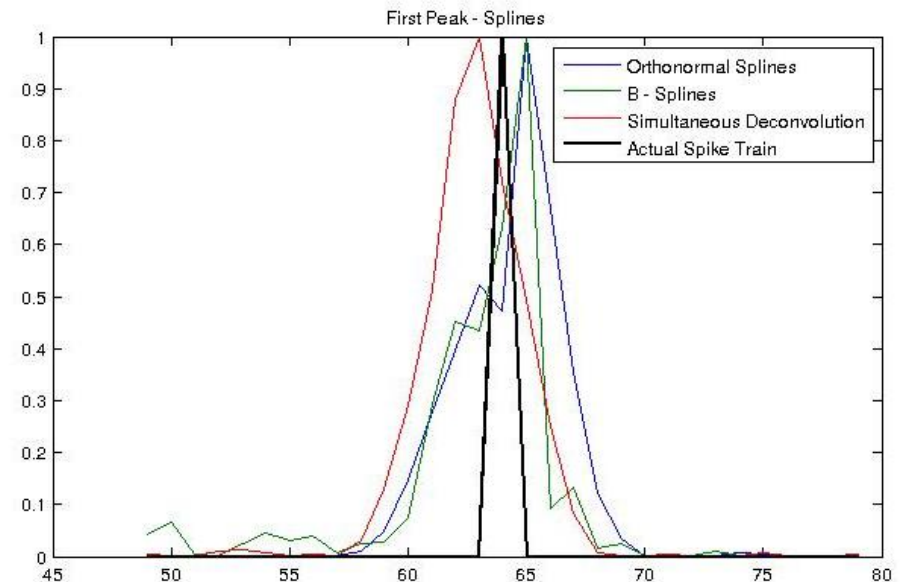
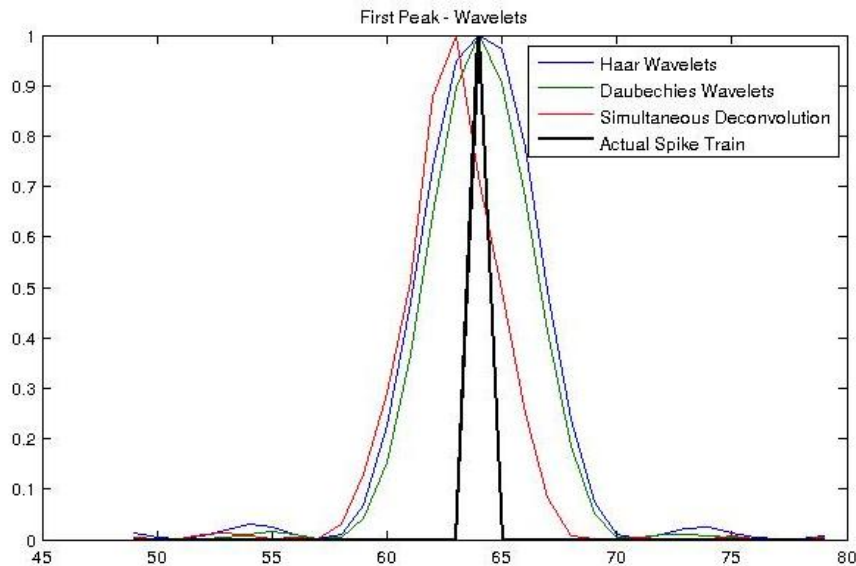


Synthetic Tests – Spline wavelets



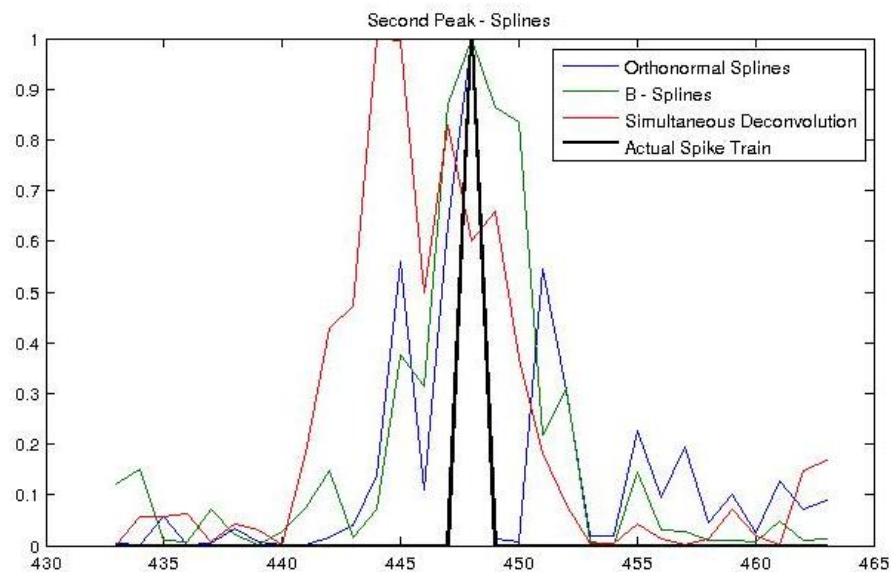
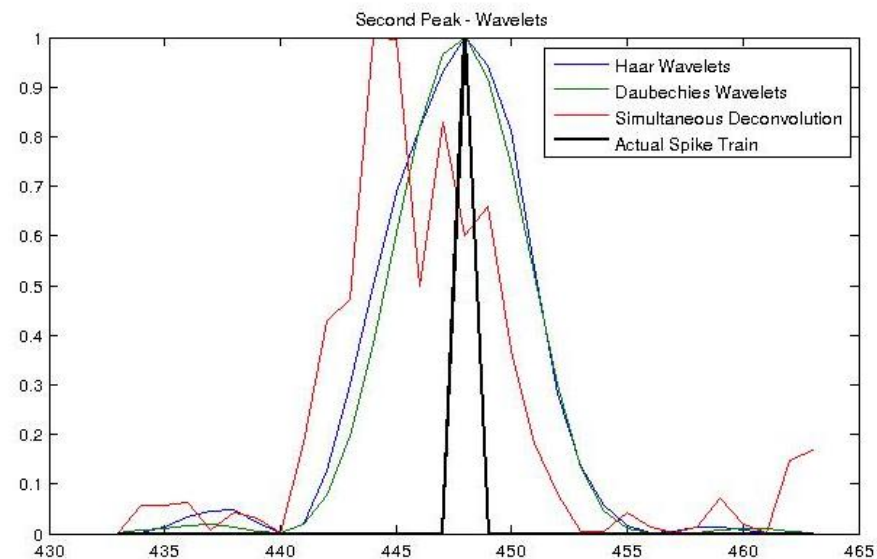
Peak comparison

- SNR is an uninformative measure on the quality of fit in this case.
- Whilst simple wavelets provide a sparse solution we also aim for a sharper peak which allows us to better constrain parameters computed later.
- Compare the normalized peak recoveries for a range of wavelet types for each peak in the receiver function.

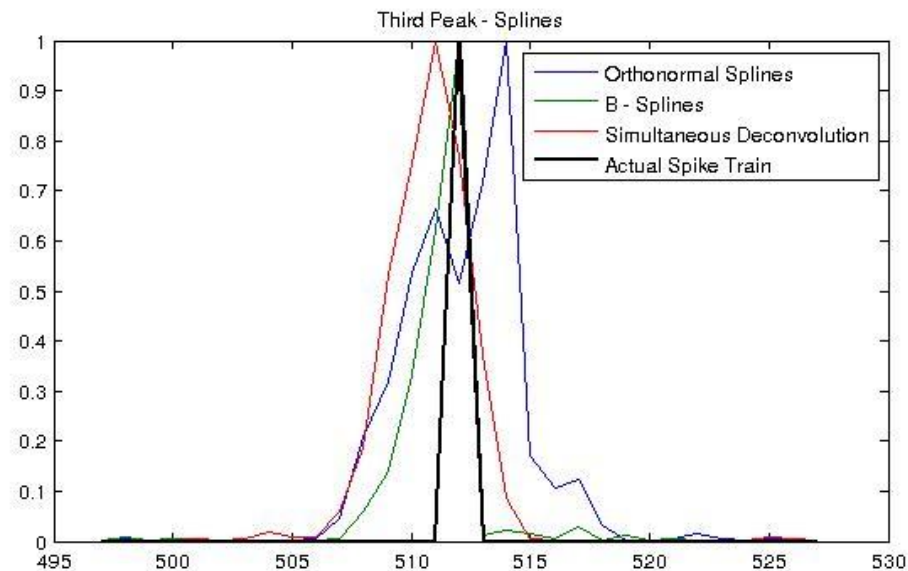
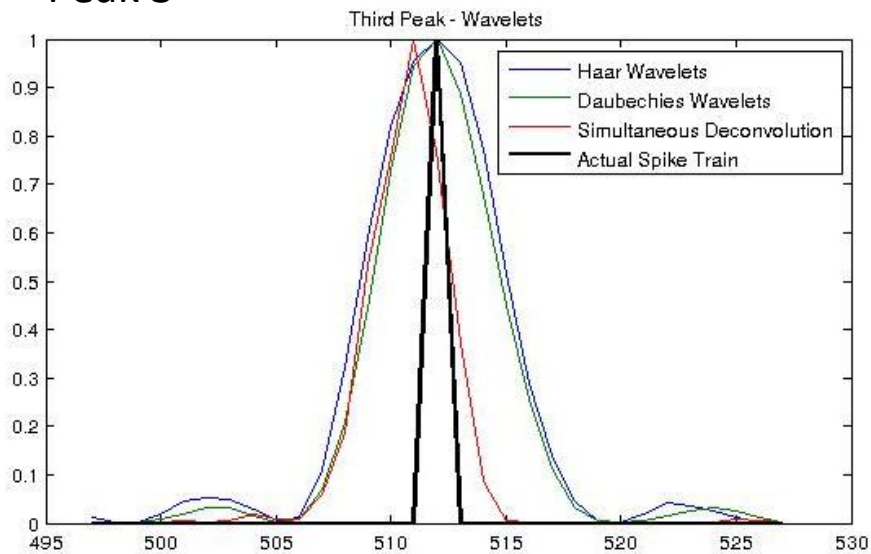


Peak comparison

Peak 2



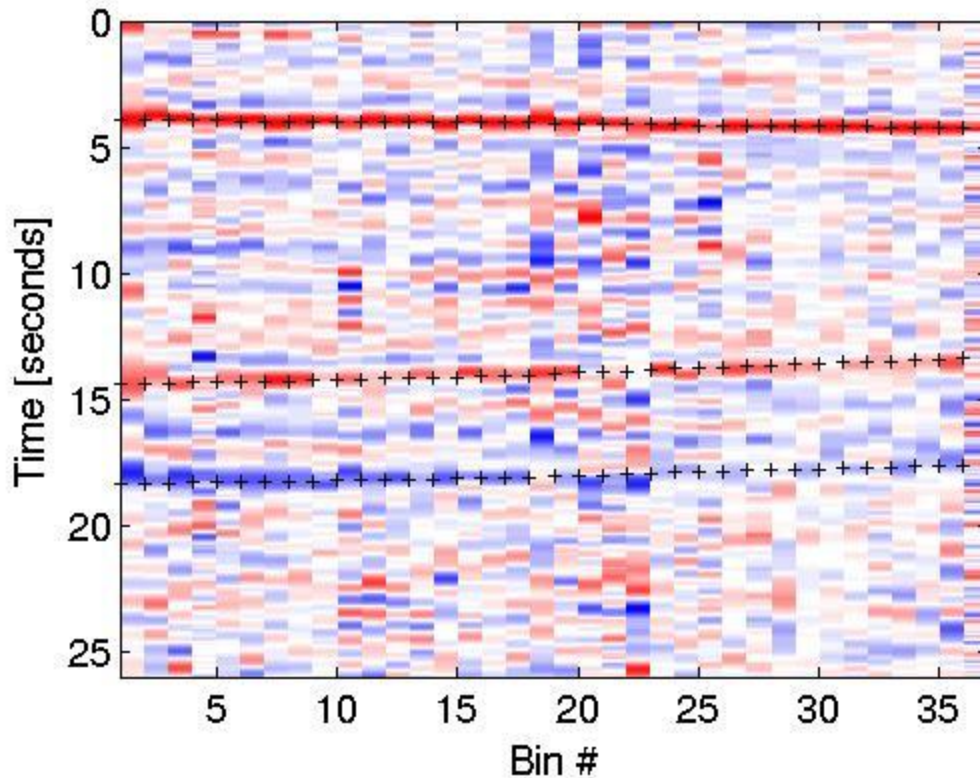
Peak 3



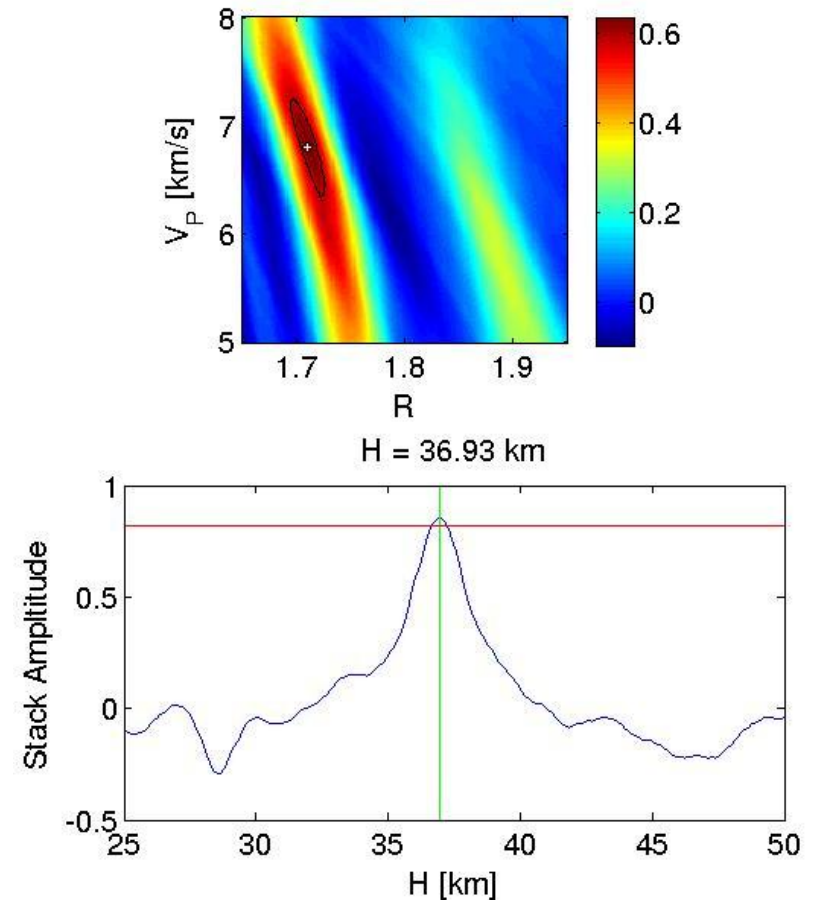
Results – Simultaneous Deconvolution

- Seismograms are binned depending on their slowness, p .
- For each bin an L1 optimization problem is solved producing a receiver function.
- Receiver function traces are collected and plotted by bin.

Receiver Fnc Section -> L2 Deconvolution

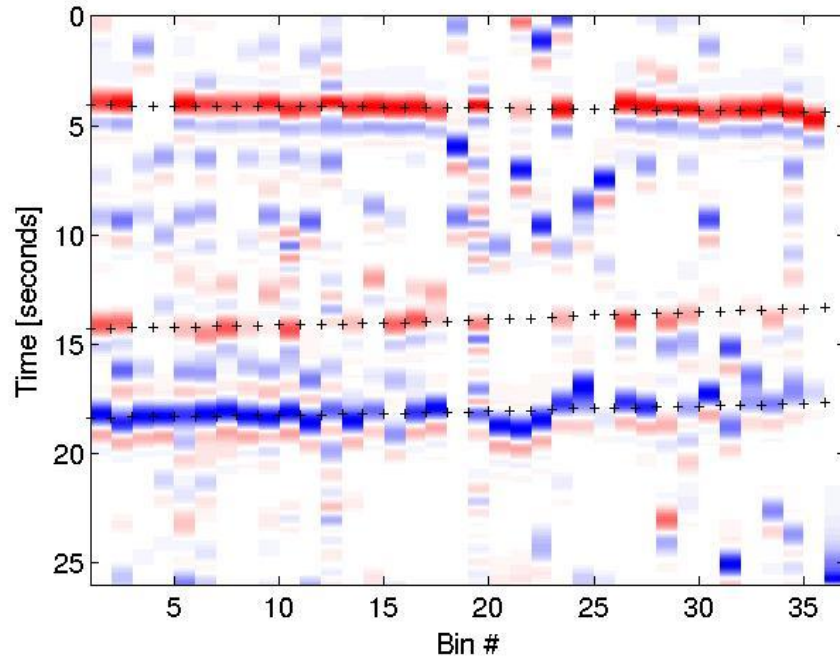


L2 Deconvolution Grid Search, error contour at $\sigma = 0.047$



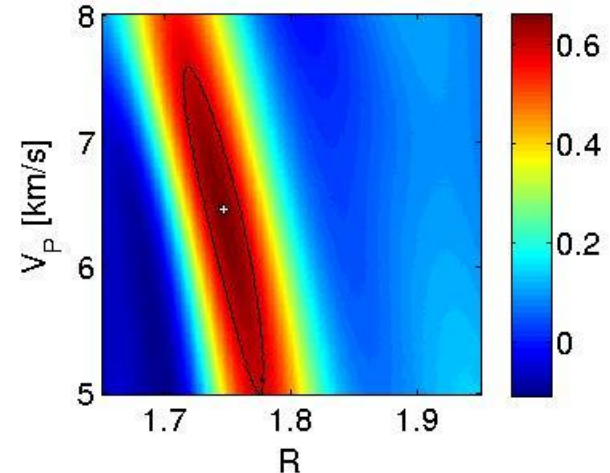
Results – Daubechies Wavelets

Receiver Fnc Section -> Daubechies

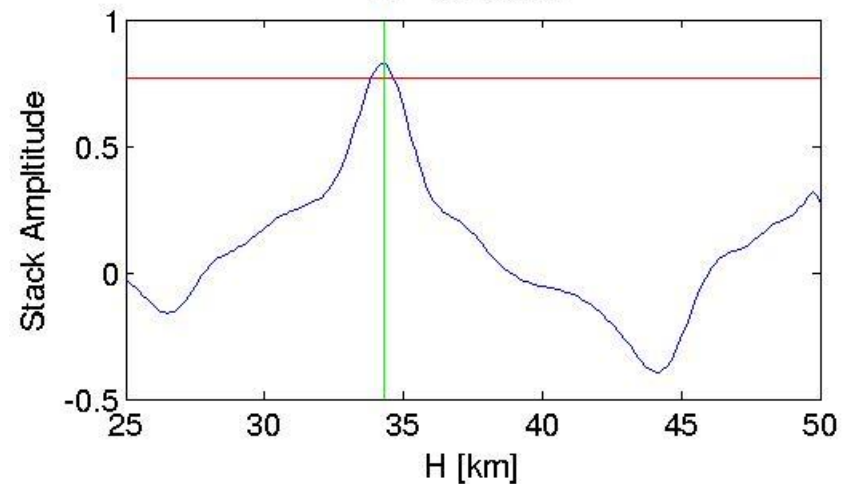


- As earlier observed, Daubechies wavelets provides a much sparser representation but peaks are less tightly constrained.

Daubechies Grid Search, error contour at $\sigma = 0.066$

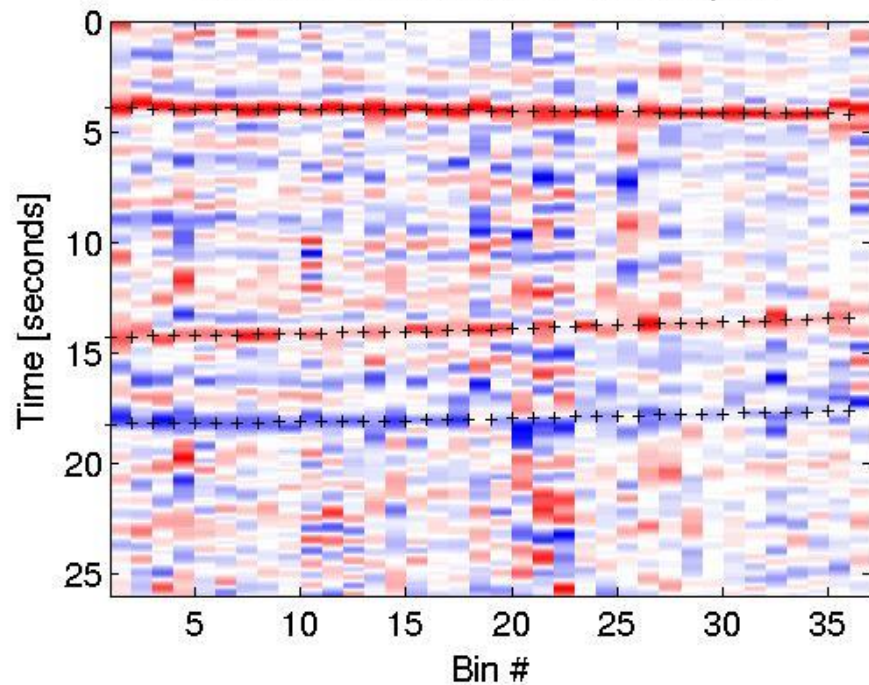


$H = 34.30$ km

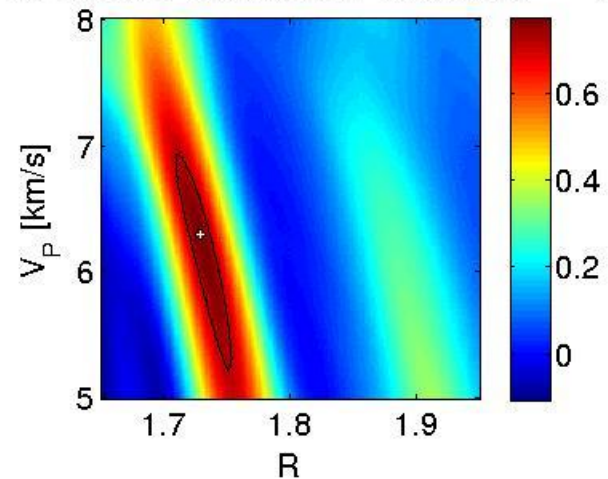


Results – Ortho Spline Wavelets

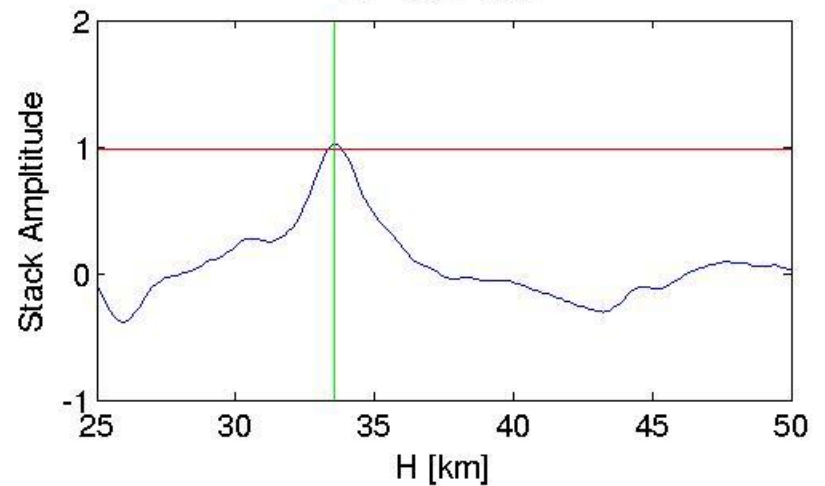
Receiver Fnc Section -> Ortho-Spline



Ortho-Splines Grid Search, error contour at $\sigma = 0.053$

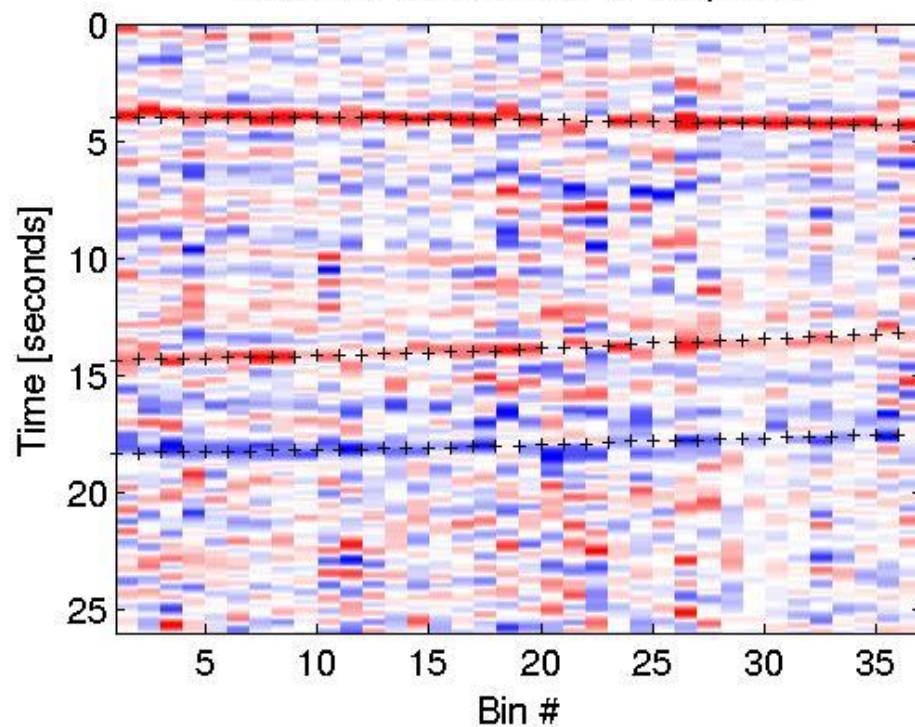


$H = 33.54$ km

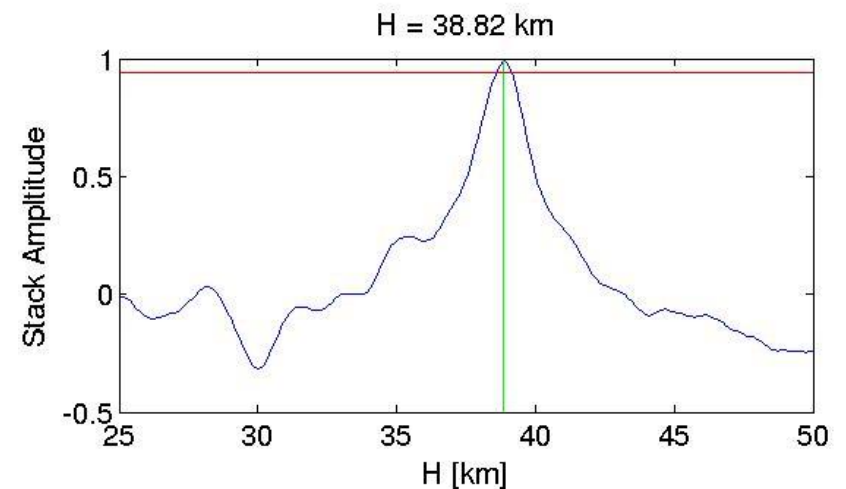
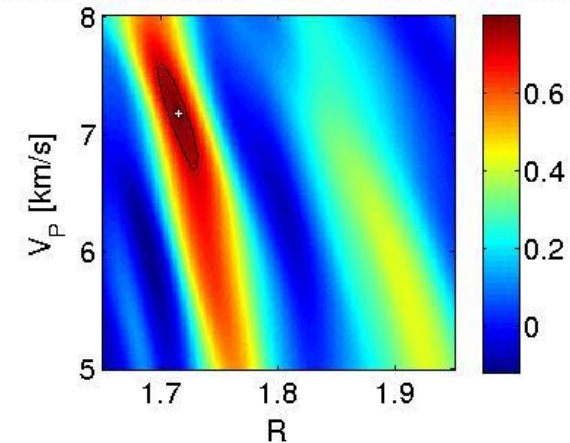


Results – B Spline Wavelets

Receiver Fnc Section -> B-Splines



B-Splines Grid Search, error contour at $\sigma = 0.063$



Summary

- Performed a deconvolution to acquire receiver functions from recorded seismograms.
- Solved an L1 optimization problem and obtained a solution which promoted sparsity for the receiver function in the wavelet domain.
- Compared a range of wavelets with an L2 type algorithm.
- L1 algorithms failed to reproduce results which were as tightly constrained as the L2 method but show promising initial results which can be further developed.