Multiparameter inverse scattering: A preliminary computational approach

Glen Young, Kris Innanen



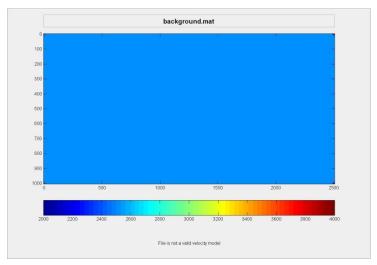
Introduction

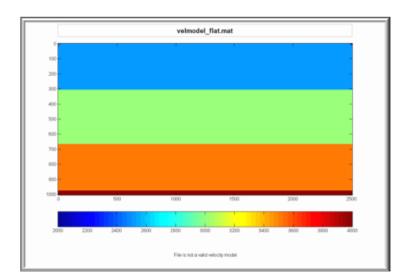
- This study investigates the approach to imaging and inversion based upon the paper by Clayton and Stolt (1981) using inverse scattering methods.
- They discuss two cases, Imaging/inversion in a constant and variable background.
- Here we treat the 2D constant background case only.
- In this study:
- Created synthetic data for 2D targets as input for the 2nd step,
- Test a prototype inverse scattering algorithm using a single parameter only (velocity) as a check on the correctness and accuracy of this algorithm.



Forward Modeling

- Creation of a synthetic 2D velocity model using a gui based tool AFD_VELCREATE.
- Generate the scattered field:
- (1) Forward model in the reference medium (constant velocity)
- (2) Forward model in the layered medium using the CREWES FD routine AFD_SHOTREC to generate 2D shot records.

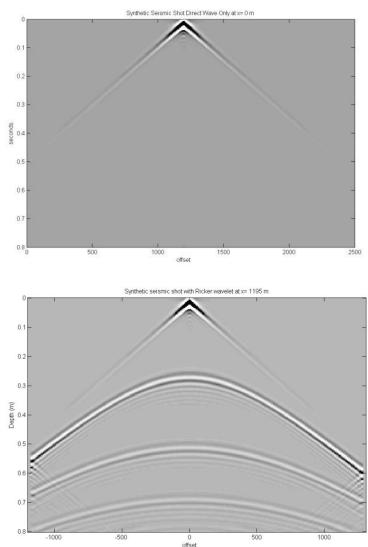






Forward Modeling

- Convolve with a Ricker wavelet.
- Generation of a background and model shotpoint gather from the background and model velocity profiles for each receiver.
- repeated shot gathers created for each receiver point along the desired seismic line with the desired receiver spacing.





Basic Scattering Theory

Lippmann-Schwinger Equation

$$G(\mathbf{r_g}|\mathbf{r_s};\omega) - G_r(\mathbf{r_g}|\mathbf{r_s};\omega) = \int_{V'} G_r(\mathbf{r_g}|\mathbf{r}';\omega) V(\mathbf{r}') G(\mathbf{r}'|\mathbf{r_s};\omega) \mathrm{d}V'$$

Inhomogeneous Helmholtz equation

$$\mathbf{L}\mathbf{G} = \left(\frac{\omega^2}{K} + \nabla \cdot \frac{1}{\rho}\nabla\right)\mathbf{G} = -\delta(\mathbf{r}' - \mathbf{r_s}),$$
$$\mathbf{L}_r\mathbf{G}_r = \left(\frac{\omega^2}{K_r} + \nabla \cdot \frac{1}{\rho_r}\nabla\right)\mathbf{G}_r = -\delta(\mathbf{r}' - \mathbf{r_s}),$$

V is called the scattering potential

$$\mathbf{V} = \mathbf{L} - \mathbf{L}_{\mathbf{r}} = \left(\frac{\omega^2}{K} - \frac{\omega^2}{K_r}\right) + \nabla \cdot \left(\frac{1}{\rho} - \frac{1}{\rho_r}\right) \nabla = \frac{\omega^2}{K_r} a_K + \nabla \cdot \frac{a_\rho}{\rho_r} \nabla,$$

where a_K, a_ρ are

$$a_K = \left(\frac{K_r}{K(\mathbf{r}')} - 1\right), \ a_\rho = \left(\frac{\rho_r}{\rho(\mathbf{r}')} - 1\right),$$

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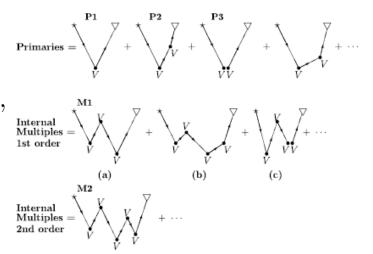
Basic Scattering Theory

Born-Neumann Series

 $G = G_r + G_r V G_r + G_r V G_r V G_r + \dots,$

We define the scattered field as:

$$\Psi_{scattered}(\mathbf{r}) = \mathbf{G} - \mathbf{G}_{\mathbf{r}} \to data$$



Born Approximation: $|\psi_{scattered}| << |\psi_{inc}|$

$$\psi_{scattered} \approx \int_{V} G_{r}(\mathbf{r}|\mathbf{r}')V(\mathbf{r}') \bigg[\psi_{inc}(\mathbf{r}') + (\psi_{scattered}(\mathbf{r}') \to 0)\bigg] \mathrm{d}V'$$

In operator form the linearized series becomes

$$G = G_r + G_r V G_r,$$

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Inversion $D = (G - G_r)S(\omega) = (G_rVG_r)S(\omega).$

 $D'(k_m, k_h, k_z) = \frac{-1}{\rho_r} \frac{D(k_m, k_h, \omega)}{S(\omega) + \epsilon}, \qquad \text{Deconvolve the source wavelet from data.}$

Fourier transform from x_g , x_s coordinates to source-receiver wavenumbers k_g, k_s .

$$D(k_g, k_s, \omega) = \frac{1}{2\pi} \int dx_g \int dx_s e^{-ik_g x_g} D(x_g, x_s, \omega) e^{ik_s x_s},$$

= $\int dx' \int dz' G_r^+(k_g, 0 | x', z'; \omega) V(x', z'; \omega) G_r^+(x', z' | k_s, 0 |; \omega) S(\omega),$
 $k_m = k_g - k_s , k_h = k_g + k_s , x_m = \frac{x_g + x_s}{2}, x_h = \frac{x_g - x_s}{2},$

Inversion is driven by mapping the spectrum of the data to the spectrum of the model.

$$D(k_m, k_z, k_h) = \frac{-\rho_r S(\omega)}{C(k_m, k_z, k_h)} \left[\frac{\omega^2}{\nu_r^2} a_K(k_m, k_z) + F(k_m, k_z, k_h) a_\rho(k_m, k_z) \right]$$



Inversion

When transformation variables are substituted in and simplified we obtain the system of equations which need to be inverted.

$$D(k_m, k_z, k_h) = -\rho_r \bigg[\sum_{i=1}^2 A_i(k_m, k_h, k_z) a_i(k_m, k_z) \bigg] S(\omega),$$

$$A_1 = (k_m, k_h, k_z) = \frac{1}{4} \frac{(k_z^2 + k_h^2)(k_z^2 + k_m^2)}{k_z^4 - k_m^2 k_h^2}, A_2 = (k_m, k_h, k_z) = \frac{1}{4} \frac{(k_z^2 - k_h^2)(k_z^2 + k_m^2)}{k_z^4 - k_m^2 k_h^2},$$

After the deconvolution stage we are left with

$$D'(k_m, k_z, k_h) = \left[\sum_{i=1}^2 A_i(k_m, k_h, k_z)a_i(k_m, k_z)\right],$$

of which we need to determine the $a_i(k_m, k_z)$ through perhaps a least squares method.



Synthetic Examples

- Forward modelling:
- x line length of 2500m , Max depth of 1000m (t=0.8 sec),
- source/receiver spacing of 10m, sampling time of 4ms.
- ➢ F-D calculations: time step is 0.5ms computation grid of 10mx10m.
- A band limited Ricker wavelet at 30Hz and 0.1 seconds was also added.
- Scalar(1 parameter acoustic) model.
- Pre-processing:
- Deconvolution (already know the type of wavelet).
- Subtraction of direct wave from input dataset.
- Inversion:
- Direct Fourier transform of the dataset.
- Fill in model spectra on a regular k_m , k_z grid
- Inverse Fourier transform back into the image space x,z.



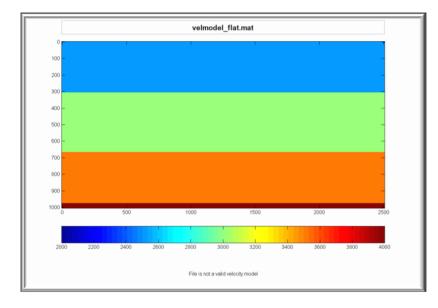


Figure 1: Four Layer Horizontal Velocity Model

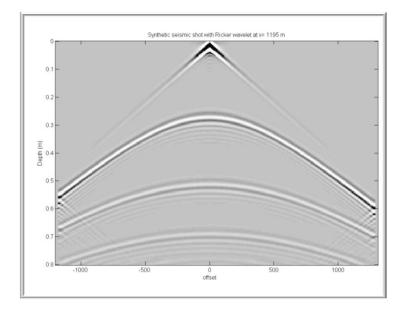
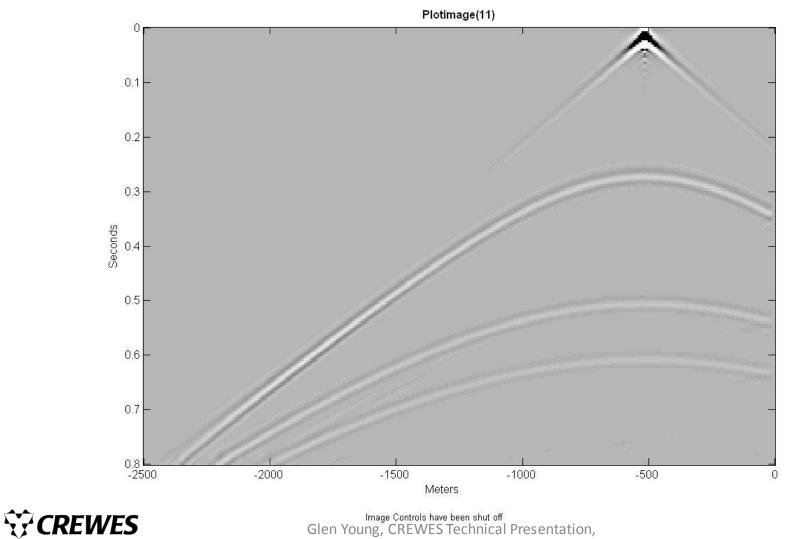


Figure 2: Four Layer Shot Profile w Ricker wavelet



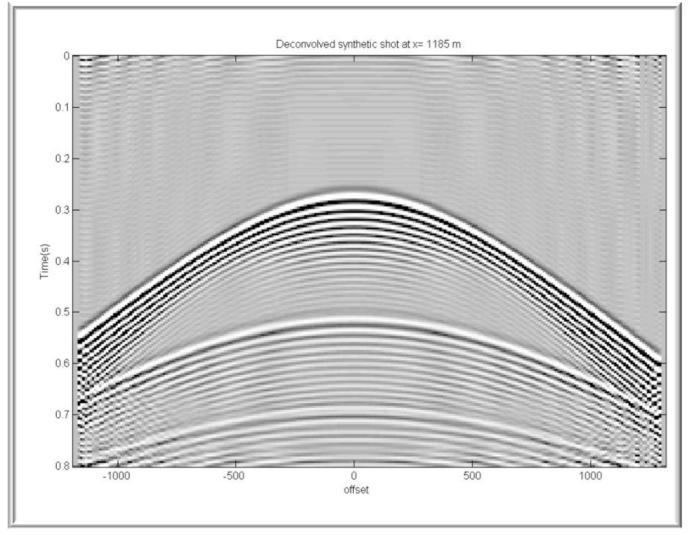




Figure 3: Four Layer, Deconvolved w/o Direct Wave Glen Young, CREWES Technical Presentation,

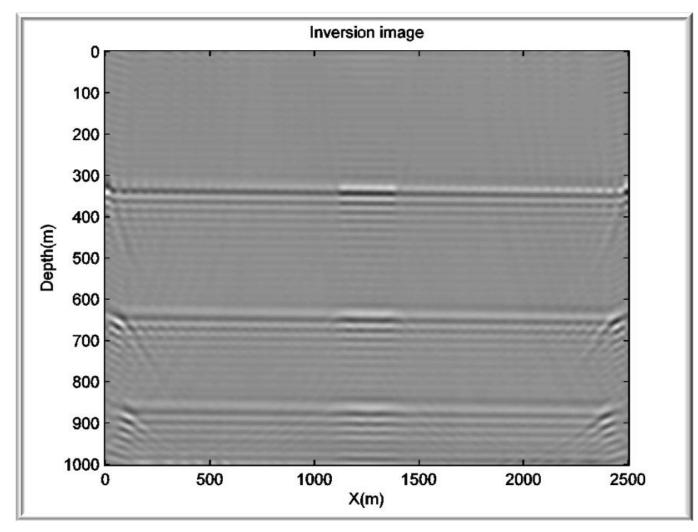


Figure 5: Four layer Horizontal Model, Migrated Image



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Shallow low velocity model

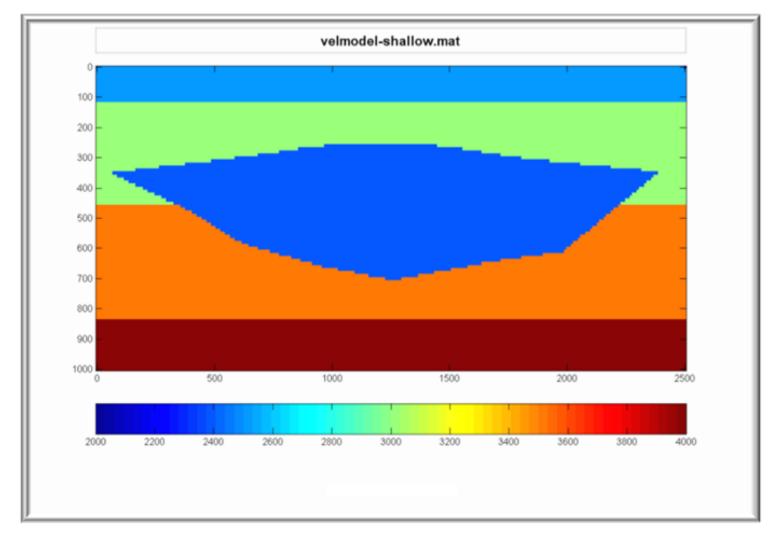




Figure 6: Shallow Lens Velocity Model

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Shallow low velocity model

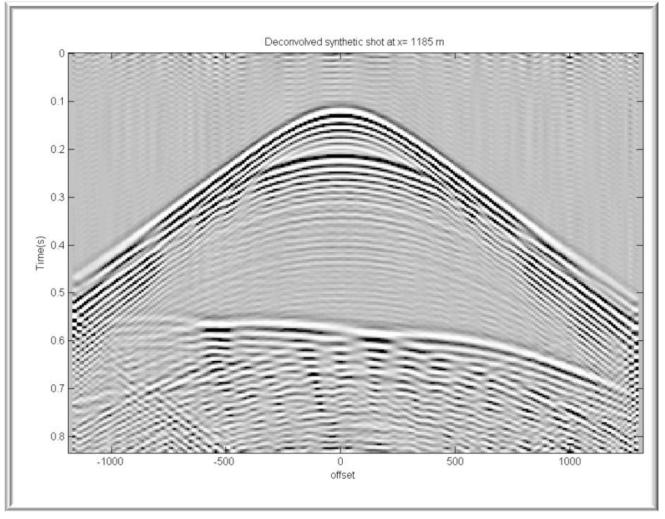


Figure 8: Shallow Lens, Deconvolved w/o Direct Wave



Shallow low velocity model

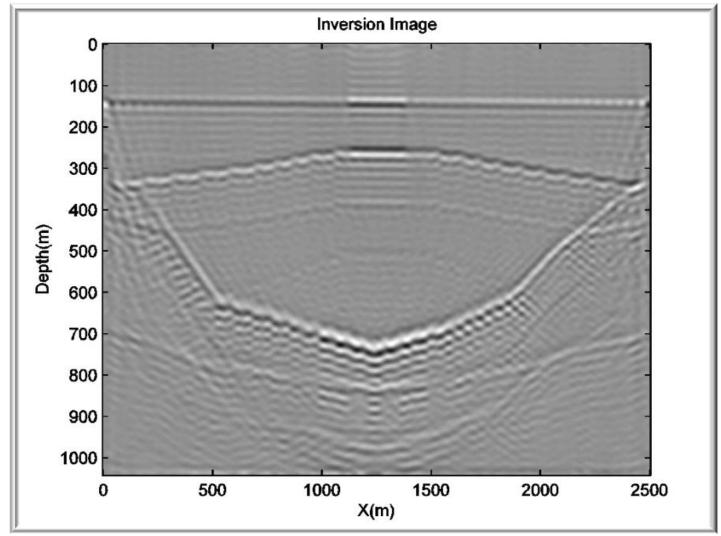


Figure 10: Shallow, Low Velocity Model, Migrated Image



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Anticline velocity model

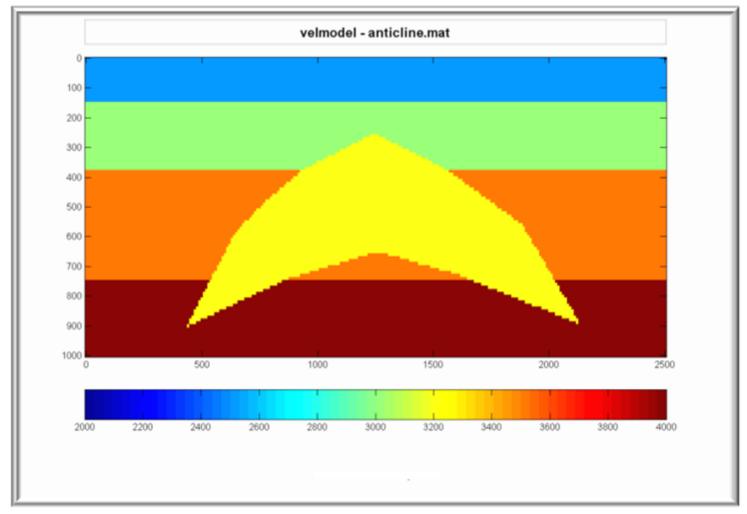


Figure 11: Anticline Velocity Model



Anticline velocity model

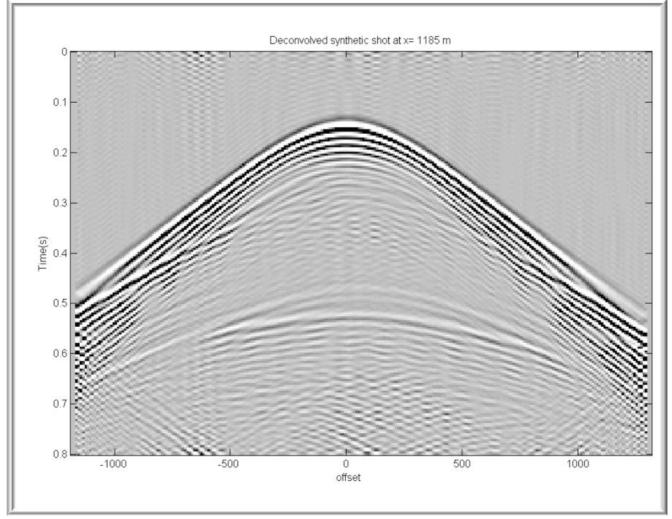


Figure 13: Anticline Deconvolved w/o Direct Wave

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Anticline velocity model

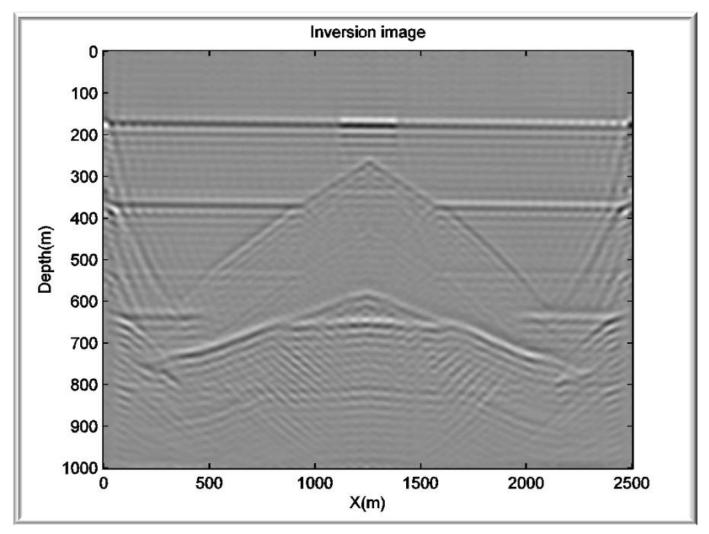


Figure 15: Anticline Velocity Model, Migrated Image

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Conclusion

- Even with the relatively straightforward algorithm for a constant background, we are able to image the models to a fairly high degree of accuracy.
- Our simulations have not taken into account real world effects such as multiples and anisotropic and anelastic media which are variable background effects.
- We need to verify that the amplitudes of the reflections found in these results are true amplitudes.
- Next step incorporate material parameters such as bulk modulus, variable density into the forward model(true two parameter mode).
- test the accuracy of the inversions in producing the correct and physically realistic answers.



Acknowledgements

- Kris Innanen for suggesting this as a "short" summer project.
- Also for lending me his GPR code based on the Clayton and Stolt paper as a template.
- Heather Lloyd for pointing out those nice FD tools I initially used
- Plus pointing out the sources of my F-D errors to this F-D noob!



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The complete derivation from the Clayton and Stolt paper.



Appendix A: Inversion
$$D = (G - G_r)S(\omega) = (G_rVG)S(\omega).$$

 $D'(k_m, k_h, k_z) = \frac{-1}{\rho_r} \frac{D(k_m, k_h, \omega)}{S(\omega) + \epsilon}, \qquad \text{Deconvolve the source wavelet from data.}$

The first step: Fourier transform from x_g , x_s coordinates to source-receiver wavenumbers k_g , k_s .

$$D(k_g, k_s, \omega) = \frac{1}{2\pi} \int dx_g \int dx_s e^{-ik_g x_g} D(x_g, x_s, \omega) e^{ik_s x_s},$$
$$= \int dx' \int dz' G_r^+(k_g, 0|x', z'; \omega) V(x', z'; \omega) G_r^+(x', z'|k_s, 0|; \omega) S(\omega),$$

with the Green's operators for a 2D constant background are given by

$$G_r^+(k_g, 0|x', z'; \omega) = \frac{i\rho_r}{\sqrt{2\pi}} \frac{e^{-i(k_g x' - q_g|z'|)}}{2q_g}, \quad G_r^+(x', z'|k_s, 0; \omega) = \frac{i\rho_r}{\sqrt{2\pi}} \frac{e^{i(k_s x' + q_s|z'|)}}{2q_s},$$

where

$$q_g = \frac{\omega}{\nu_r} \sqrt{1 - \frac{\nu_r^2 k_g^2}{\omega^2}}, \quad q_s = \frac{\omega}{\nu_r} \sqrt{1 - \frac{\nu_r^2 k_s^2}{\omega^2}},$$



$$\mathbf{V} = \frac{\omega^2}{K_r} a_K + \nabla \cdot \frac{a_\rho}{\rho_r} \nabla, \qquad a_K = \left(\frac{K_r}{K(\mathbf{r}')} - 1\right), \ a_\rho = \left(\frac{\rho_r}{\rho(\mathbf{r}')} - 1\right),$$

using the definitions for a_K and a_ρ then integration by parts we get

$$D(k_g, k_s, \omega) = \frac{-\rho_r^2}{2\pi} \int dx' \int dz' \frac{e^{-i[(k_g - k_s)x' - (q_g + g_s)z')]}}{4q_g q_s} \times \left[\frac{\omega^2}{\nu_r^2} a_K(x', z') + (q_g q_s - k_g k_s) a_\rho(x', z')\right] S(\omega),$$

The above equation is of the same form as a double Fourier transform in the x' and z' variables if we do some rearranging the result of the evaluation yields

$$D(k_g, k_s, \omega) = \frac{-\rho_r^2 S(\omega)}{4q_g q_s} \left[\frac{\omega^2}{\nu_r^2} a_K (k_g - k_s, -q_g - q_s) + (q_g q_s - k_g k_s) a_\rho (k_g - k_s, -q_g - q_s) \right],$$



Change of Coordinates to ω , q_g , q_s space

To solve for a_K and a_ρ change to midpoint/offset coordinates from the source/receiver system. We have the midpoint wavenumber $k_m = k_g - k_s$, the half offset wavenumber $k_h = k_g + k_s$ which in x,z domain corresponds to

$$x_m = \frac{x_g + x_s}{2}, \ x_h = \frac{x_g - x_s}{2},$$

and a new independent variable, Solving for ω , q_g and q_s we have the expressions

$$\begin{aligned} k_z &= -q_g - q_s = -\frac{\omega}{\nu_r} \sqrt{1 - \frac{\nu_r^2 k_g^2}{\omega^2}} - \frac{\omega}{\nu_r} \sqrt{1 - \frac{\nu_r^2 k_s^2}{\omega^2}}, \\ \omega(k_m, k_h, k_z) &= -\frac{\nu_r k_z}{2} \sqrt{(1 + \frac{k_m^2}{k_z^2})(1 + \frac{k_h^2}{k_z^2})}, \\ q_g(k_m, k_h, k_z) &= -\frac{k_z}{2} (1 - \frac{k_m k_h}{k_z^2}), \quad q_s(k_m, k_h, k_z) = -\frac{k_z}{2} (1 + \frac{k_m k_h}{k_z^2}), \\ k_m &= k_g - k_s \ , k_h = k_g + k_s \ , x_m = \frac{x_g + x_s}{2}, \ x_h = \frac{x_g - x_s}{2}, \end{aligned}$$



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Now that we have expressions for ω , q_g , q_s the Direct Fourier transform computed in these coordinates and data can be transformed back after the inversion.

$$D(k_m, k_z, k_h) = \frac{-\rho_r^2 S(\omega)}{4q_g q_s} \left[\frac{\omega^2}{\nu_r^2} a_K(k_m, k_z) + (q_g q_s - k_g k_s) a_\rho(k_m, k_z) \right],$$

$$D(k_m, k_z, k_h) = \frac{-\rho_r S(\omega)}{C(k_m, k_z, k_h)} \left[\frac{\omega^2}{\nu_r^2} a_K(k_m, k_z) + F(k_m, k_z, k_h) a_\rho(k_m, k_z) \right],$$

When transformation variables are substituted in and simplified we obtain the system of equations which need to be inverted.

$$D(k_m, k_z, k_h) = -\rho_r \bigg[\sum_{i=1}^2 A_i(k_m, k_h, k_z) a_i(k_m, k_z) \bigg] S(\omega),$$

$$A_1 = (k_m, k_h, k_z) = \frac{1}{4} \frac{(k_z^2 + k_h^2)(k_z^2 + k_m^2)}{k_z^4 - k_m^2 k_h^2}, A_2 = (k_m, k_h, k_z) = \frac{1}{4} \frac{(k_z^2 - k_h^2)(k_z^2 + k_m^2)}{k_z^4 - k_m^2 k_h^2},$$

After the deconvolution stage we are left with

$$D'(k_m, k_z, k_h) = \left[\sum_{i=1}^{2} A_i(k_m, k_h, k_z) a_i(k_m, k_z)\right],$$

of which we need to determine the $a_i(k_m, k_z)$ through perhaps a least squares method.