Flux-normalized wavefield decomposition and migration of seismic data

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ABSTRACT

Separation of wavefields into directional components can be accomplished by an eigenvalue decomposition of the accompanying system matrix. In conventional pressure-normalized wavefield decomposition, the resulting one-way wave equations contain an interaction term which depends on the reflectivity function. Applying directional wavefield decomposition using flux-normalized eigenvalue decomposition, and disregarding interaction between up- and downgoing wavefields, these interaction terms were absent. By also applying a correction term for transmission loss, the result was an improved estimate of the up- and downgoing wavefields. In the wave equation angle transform, a crosscorrelation function in local offset coordinates was Fourier-transformed to produce an estimate of reflectivity as a function of slowness or angle. We normalized this wave equation angle transform with an estimate of the plane-wave reflection coefficient. The flux-normalized one-way wave-propagation scheme was applied to imaging and to the normalized wave equation angle-transform on synthetic and field data; this proved the effectiveness of the new methods.

INTRODUCTION

Accurate wavefield traveltimes and amplitudes can be described using two-way wave equation techniques like finite-difference or finite-element methods. However, these methods can often be significantly more computationally expensive compared to one-way methods. The computational cost in modeling wavefield extrapolation using full wave equation methods can become a limitation for 3D applications in particular. Ray methods based upon asymptotic theory provide effective alternatives to full wave equation methods; however, their high-frequency approximations restrict their use in complex subsurface velocity. One-way wavefield methods based upon a paraxial approximation of the wave equation provide a robust and computationally cheap alternative approach for solving the wave equation. With wavefield propagators based on one-way methods, we can substantially increase the speed of computations compared to full wavefield methods. Representation of a wavefield using the one-way wave equation permits separation of the wavefield into up- and downgoing constituents. This separation is not valid for near-horizontal propagating waves. Schemes for splitting the wave equation into up- and downgoing parts and seismic mapping of reflectors are discussed by Claerbout (1970, 1971).

Several authors have investigated various methods for amplitude correction to one-way wave equations. Zhang et al. (2003, 2005, 2007) address true-amplitude implementation of one-way wave equations in common-shot migration by modifying the one-way wave equation. This is accomplished by introducing an auxiliary function that corrects the leading order transport equation for the full wave equation. Ray theory applied to the modified one-way wave equations yields up- and downgoing eikonal equations with amplitudes satisfying the transport equation. Full waveform solutions substituted with corresponding ray-theoretical approximations provides true-amplitude in the sense that the imaging formulas reduce to a Kirchhoff common-shot inversion expression.

Kiyashchenko et al. (2005) develop improved estimation of amplitudes using a multi-one-way approach. It is developed from an iterative solution of the factorized two-way wave equation with a right side incorporating the medium heterogeneities. It allows for vertical and horizontal velocity variations and it is demonstrated that the multi-one-way scheme reduces errors in amplitude estimates compared to conventional one-way propagators.

Cao and Wu (2008) reformulate the solution of the one-way wave equation in smoothly varying 1D media based on energy-flux conservation. By introducing transparent boundary conditions and transparent propagators, their formulation is extended to a general heterogeneous media in the local angle domain utilizing beamlet methods.
By decomposing the wavefield into up- and downgoing waves with an eigenvalue decomposition using symmetry properties of the accompanying system matrix, one can derive simplified equations for computing the wavefield propagators. This directional decomposition is consistent with a flux-normalization of the wavefield (Ursin, 1983). Further, by neglecting coupling terms between the up- and downgoing waves, the resulting system matrix can be used as a starting point to derive paraxial approximations of the original wave equation. They also can be used to derive WKBJ approximations of various orders (Bremmer, 1951; van Stralen et al., 1998).

In this paper, we derive initial conditions and one-way propagators for flux-normalized wavefield extrapolation in 1D media and show how this provides accurate amplitude information. We formulate an unbiased estimate of the reflectivity using the wave equation angle transform. Further, we propose an extension to a general heterogeneous media in which the flux-normalization and an approximation to the transmission loss are performed.

We account for the medium perturbations in the downward propagation using Fourier finite-difference methods (Ristow and Rühl, 1994). We apply conventional pressure-normalized and the derived approximations for computing the wavefield propagators. This directional decomposition is accomplished by applying an inverse eigenvector matrix of $A$, denoted $L^{-1}$, on $b$. We define the transformed field vector containing the directional decomposed wavefield by

$$w = \begin{bmatrix} U \\ D \end{bmatrix} = L^{-1}b.$$  

Moreover, upon substitution of $w$, the matrix differential equation 8 transforms to
where an eigenvalue decomposition of $A$ provides the diagonal eigenvalue matrix

$$
\Lambda = L^{-1}AL = \begin{bmatrix}
-p_3 & 0 \\
0 & p_3
\end{bmatrix},
$$

where $p_3$ is the vertical slowness.

**Amplitude-normalized wavefields**

In the conventional pressure-normalized wavefield separation approach, the eigenvector matrix of $A$ is chosen as (Claerbout, 1976; Ursin, 1984, 1987)

$$
L = \begin{bmatrix}
1 & 1 \\
-\frac{1}{Z} & \frac{1}{Z}
\end{bmatrix}.
$$

This leads to the inverse eigenvector matrix

$$
L^{-1} = \frac{1}{2} \begin{bmatrix}
1 & -Z \\
1 & Z
\end{bmatrix}.
$$

With the eigenvector matrix defined in equation 14, the matrix differential equation 12 becomes

$$
\partial_3 w = i\omega \begin{bmatrix}
-p_3 & 0 \\
0 & p_3
\end{bmatrix} w - \gamma(x_3) \begin{bmatrix}
-1 & 1 \\
1 & -1
\end{bmatrix} w,
$$

where

$$
\gamma(x_3) = \frac{1}{2} \partial_3 \log Z(x_3)
$$

is the reflectivity function. Using equation 7 it can be expressed as

$$
\gamma(x_3) = \frac{1}{2} \begin{bmatrix}
1 & \partial \theta \\
\cos^2 \theta & \partial c
\end{bmatrix}.
$$

The wavefield decomposition described in equation 16 is referred to as being pressure-normalized in the sense that the pressure field equals the sum of the up- and downgoing wavefields.

We consider a stack of inhomogeneous layers where $\rho$ and $c$ are continuous functions of $x_3$ within each layer. At an interface between two layers, the boundary condition requires that the wave vector $b$ shall be continuous. For an interface at $x_3 = x_3$, we must have $L_+ w_+ = L_- w_-$ where $L_- = L(x_3)$ is evaluated above the interface, and $L_+ = L(x_3)$ is evaluated beneath the interface (the $x_3$-axis is pointing vertically downward). We therefore have

$$
w_+ = L_+^{-1} L_- w_-. 
$$

Equations 14 and 15 give

$$
L_+^{-1} L_- = \frac{1}{2} \begin{bmatrix}
1 + \frac{Z}{L} & 1 - \frac{Z}{L} \\
1 - \frac{Z}{L} & 1 + \frac{Z}{L}
\end{bmatrix}.
$$

This can be written as (Ursin, 1983, equation 33):

$$
L_+^{-1} L_- = \begin{bmatrix}
T^{-1} u & -T^{-1} v \\
R u & T^{-1} v
\end{bmatrix},
$$

where $T_u$ and $R_u$ are the transmission and reflection coefficients for an upward traveling incident wave at the interface.

**Flux-normalized wavefields**

We now derive an alternative directional decomposition by a flux-normalization of the wavefield. The main advantage of flux-normalizing the wavefield is that we obtain a simpler expression of the corresponding directional decomposed matrix differential equation, as compared to the pressure-normalized approach. Disregarding the interaction between directional components yields a matrix differential equation independent of the reflectivity function.

In order obtain a flux-normalized system of equations, the eigenvector matrix of $A$ is chosen as (Ursin, 1983; Wapenaar, 1998)

$$
\tilde{L} = \frac{1}{\sqrt{2}} \begin{bmatrix}
\sqrt{Z} & \sqrt{Z} \\
-\sqrt{Z} & \sqrt{Z}
\end{bmatrix},
$$

and thus the inverse eigenvector matrix becomes

$$
\tilde{L}^{-1} = \frac{1}{\sqrt{2}} \begin{bmatrix}
\frac{1}{\sqrt{Z}} & -\frac{1}{\sqrt{Z}} \\
\frac{1}{\sqrt{Z}} & \frac{1}{\sqrt{Z}}
\end{bmatrix}.
$$

This provides a flux-normalized representation of the wavefield

$$
\tilde{w} = \tilde{L}^{-1} b,
$$

where $\tilde{b} = (\tilde{U}, \tilde{D})^T$, and where $\tilde{U}$ and $\tilde{D}$ denote the flux-normalized directional components of the wavefield. The wavefield is referred to as flux-normalized in the sense that the energy flux in the $x_3$-direction is propagation invariant (Ursin, 1983; Wapenaar, 1998). The pressure-normalized and the flux-normalized decomposition break down for near-horizontally-traveling waves because the lateral wavenumber $k_3$ approaches to zero in the horizontal direction.

Combining equations 22 and 24 with equation 12 yields the transformed matrix differential equation

$$
\partial_3 \tilde{w} = i\omega \begin{bmatrix}
-p_3 & 0 \\
0 & p_3
\end{bmatrix} \tilde{w} - \gamma(x_3) \begin{bmatrix}
0 & 1 \\
1 & 0
\end{bmatrix} \tilde{w}.
$$

Comparing the flux-normalized system of equations in equation 25 with the conventional pressure-normalized system of equations in equation 16, we see that in equation 25 only the off-diagonal terms (depending on the reflectivity function $\gamma(x_3)$) are present. Further, by neglecting interaction between the flux-normalized directional decomposed components, the flux-normalized matrix differential equation becomes independent of the reflectivity function $\gamma(x_3)$. Finally, we note that
\( \tilde{w}(\omega, k_1, k_2, x_3) = \sqrt{\frac{2}{Z}} w(\omega, k_1, k_2, x_3) \)
\[ = \sqrt{\frac{2k_3}{\rho \omega}} w(\omega, k_1, k_2, x_3). \tag{26} \]

At an interface between two smoothly varying media we have
\( \tilde{w}_+ = L_1^{-1} L \tilde{w}_- \)
with
\[ \tilde{L}_+^{-1} \tilde{L} = \begin{bmatrix} \tilde{T}_{u-1} & R_u \tilde{T}_{u-1} \\ R_u & \tilde{T}_{u-1} \end{bmatrix} \tag{27} \]

Here,
\[ \tilde{T}_u^{-1} = \sqrt{\frac{Z_-}{Z_+} T_u^{-1}} = \frac{Z_+ + Z_-}{2 \sqrt{Z_+ Z_-}} = \frac{1}{2} \left\{ \sqrt{\frac{Z_+}{Z_-}} + \sqrt{\frac{Z_-}{Z_+}} \right\}, \tag{28} \]
and
\[ R_u \tilde{T}_u^{-1} = \frac{Z_- - Z_+}{2 \sqrt{Z_+ Z_-}} = \frac{1}{2} \left\{ \sqrt{\frac{Z_-}{Z_+}} - \sqrt{\frac{Z_+}{Z_-}} \right\}, \tag{29} \]

where \( Z_- \) denotes the impedance at the bottom of the previous thin layer and \( Z_+ \) denotes the impedance at the top of the next layer.

Zhang et al. (2005) use the scaling \( \tilde{w} = \sqrt{k_3} w \) so they are not using flux-normalized variables. With their scaling, the transformed matrix differential equation will only have the same form as equation 25 for a medium with constant density.

**ONE-WAY WAVE EQUATIONS**

We obtain one-way equations for the up- and downgoing waves by neglecting the interaction terms in equations 16 and 21. This gives the zero-order WKBJ approximation (Clayton and Stolt, 1981; Ursin, 1984) obeying the equations
\[ \frac{\partial}{\partial x_3} \begin{bmatrix} U \\ D \end{bmatrix} = \begin{bmatrix} -ik_3 + \gamma & 0 \\ 0 & ik_3 + \gamma \end{bmatrix} \begin{bmatrix} U \\ D \end{bmatrix} \tag{30} \]
with interface conditions
\[ \begin{bmatrix} U \\ D \end{bmatrix}^+ = T_u^{-1} \begin{bmatrix} U \\ D \end{bmatrix}^- \tag{31} \]

In a region with smoothly-varying parameters, the equation
\[ \frac{\partial D}{\partial x_3} = (ik_3 + \gamma) D \tag{32} \]
with \( D(x_3) \) given, has the solution
\[ D(x_3) = D(x_{3b}) \exp \left[ \int_{x_{3b}}^{x_3} (ik_3(\xi) + \gamma(\xi))d\xi \right] \]
\[ = D(x_{3b}) \frac{Z(x_3)}{Z(x_{3b})} \exp \left[ \int_{x_{3b}}^{x_3} (ik_3(\xi))d\xi \right]. \tag{33} \]

The solution to equations 30 and 31 in the zero-order WKB approximation becomes
\[ D(x_3) = D(x_{3b}) T(x_3) \exp \left[ i \int_{x_{3b}}^{x_3} k_3(\xi)d\xi \right] \tag{34} \]
and
\[ U(x_3) = U(x_{3b}) T(x_3) \exp \left[ -i \int_{x_{3b}}^{x_3} k_3(\xi)d\xi \right]. \tag{35} \]

The factor
\[ T(x_3) = \sqrt{\frac{Z(x_{3b})}{Z(x_3)}} \prod_{0 < x_3 < x_3b} T_u^{-1}(x_3) \sqrt{\frac{Z(x_{3b})}{Z(x_3)}} \tag{36} \]
comes from using equations 31 and 33 for each layer, starting at the top. Passing from one layer to the next, the factor \( T_u^{-1}(x_3) \) takes the boundary conditions partly into account (equation 31). The square-root factors come from integrating equation 32 inside each layer using equation 33. This gives
\[ \sqrt{\frac{Z(x_{3b})}{Z(x_3)}} \sqrt{\frac{Z(x_{3b})}{Z(x_{3c})}} \cdots \sqrt{\frac{Z(x_3)}{Z(x_{3c})}}. \tag{37} \]

By combining the square roots for each interface from 1 to \( N \) we obtain equation 36.

Using equation 28 this can be written as:
\[ T(x_3) = \sqrt{\frac{Z(x_{3b})}{Z(x_3)}} \prod_{0 < x_3 < x_3b} \tilde{T}_u^{-1}(x_3) \tag{38} \]
in terms of the flux-normalized transmission coefficients.

For the flux-normalized up- and downgoing waves we obtain from equations 34, 35, and 38 using equation 26:
\[ \tilde{D}(x_3) = \tilde{D}(x_{3b}) \prod_{0 < x_3 < x_3b} \tilde{T}_u^{-1}(x_3) \exp \left[ i \int_{x_{3b}}^{x_3} k_3(\xi)d\xi \right] \tag{39} \]
and
\[ \tilde{U}(x_3) = \tilde{U}(x_{3b}) \prod_{0 < x_3 < x_3b} \tilde{T}_u^{-1}(x_3) \exp \left[ -i \int_{x_{3b}}^{x_3} k_3(\xi)d\xi \right]. \tag{40} \]

These equations could, of course, also have been obtained directly by neglecting the interaction terms in equations 25 and 27.
HETEROGENEOUS MEDIUM

We want to use one-way wave propagators for migration in a heterogeneous medium. Based on the previous discussion, we choose to use flux-normalized variables. The downgoing field from a point source is then represented in the wavenumber-frequency domain by

\[ D_0 = D_0(\omega, k_1, k_2, x_3) = -\frac{2\pi S(\omega)}{ik_3}. \]  

(41)

The inverse Fourier transform of equation 41 with respect to \( k_1 \) and \( k_2 \) is known as the Weyl integral (Aki and Richards, 1980). The Fourier transform of the effective source signature is \( S(\omega) \).

The flux-normalized downgoing field from a point source is represented in the wavenumber domain by

\[ D_0(\omega, k_1, k_2, x_3) = 2\pi i \sqrt{\frac{2}{\rho_0 k_3}} S(\omega). \]  

(42)

In marine seismic data, we may add the effect of the free surface (the ghost) on the downgoing wavefield (Amundsen and Ursin, 1991):

\[ D_0(\omega, k_1, k_2, 0) = 2\pi i \sqrt{\frac{2}{\rho_0 k_3}} \times (\exp[-ik_3 x_3] - R_0 \exp[ik_3 x_3]) S(\omega), \]  

(43)

where \( x_3 \) is the source depth, and the reflection coefficient is theoretically \( R_0 = 1 \).

If the wavefield is acquired by a conventional streamer configuration, only pressure is recorded. The primary upgoing wavefield \( U_0 \) can then be estimated by a demultiple procedure (Amundsen, 2001; Robertsson and Kragh, 2002), where ghost and free-surface multiples are removed from the data. Hence, using equation 26, the flux-normalized upgoing wavefield can be represented by

\[ U_0 = \sqrt{\frac{2}{Z}} U_0. \]  

(44)

The pressure and the vertical displacement velocity can be measured in ocean-bottom seismic acquisition. A recent development (Landro and Amundsen, 2007; Tenghamn et al., 2008) also allows for both these to be measured on a streamer configuration. The flux-normalized upgoing wavefield is then given by

\[ \hat{U}_0 = \frac{1}{\sqrt{2Z}} [P - ZV3]. \]  

(45)

The downward continuation of the wavefields is accomplished by solving the equation

\[ \frac{\partial \hat{w}}{\partial x_3} = \begin{bmatrix} -i \hat{H}_1 & 0 \\ 0 & i \hat{H}_1 \end{bmatrix} \hat{w} \]  

(46)

for \( x_3 > 0 \) with \( \hat{w}(0) = [\hat{U}_0, \hat{D}_0]^T \) given in equations 43–45. Equation 46 is a generalization of equation 25 with the coupling terms neglected. The operator \( \hat{H}_1 \) is the square-root operator satisfying (Wapenaar, 1998)

\[ \hat{H}_1 \hat{H}_1 = (\omega/C)^2 + \frac{\rho}{\rho_0} \frac{\partial}{\partial x_1} \left( \frac{1}{\rho_0} \frac{\partial}{\partial x_1} \right) + \frac{\rho}{\rho_0} \frac{\partial}{\partial x_2} \left( \frac{1}{\rho_0} \frac{\partial}{\partial x_2} \right). \]  

(47)

Dividing the medium into thin slabs of thickness \( \Delta x_3 \) with negligible variations in the preferred direction \( x_3 \) of propagation within each slab, allows us to extend the propagator to a general inhomogeneous medium with small lateral medium variations using, for example, split-step (Stoffa et al., 1990), Fourier finite difference (Ristow and Rühl, 1994), or a phase-screen (Wu and Huang, 1992) approach depending on the size of the medium heterogeneities in the lateral direction (Zhang et al., 2009).

At thin-slab boundaries one may apply a correction term for the transmission loss (see equation 27):

\[ \tilde{T}_u^{-1} = \frac{1}{2} \left[ \sqrt{\frac{Z_+}{Z_-}} + \sqrt{\frac{Z_-}{Z_+}} \right]. \]  

(48)

Cao and Wu (2006) have proposed a similar correction for the downward continuation of pressure.

The downward-continued wavefields can be used in a standard way to create an image. One may also apply a crosscorrelation and a local Fourier transform to compute common-angle gathers (de Bruin et al., 1990; Sava and Fomel, 2003; de Hoop et al., 2006; Sun and Zhang, 2009). This is termed the wave equation angle transform, and a common-image gather for a single shot is

\[ I(p, x, x_3) = \int \int \tilde{U} \left( \omega, \mathbf{x} + \mathbf{h}, x_3 \right) \tilde{U}^* \left( \omega, \mathbf{x} - \mathbf{h}, x_3 \right) e^{-i \mathbf{p} \cdot \mathbf{h}} d\mathbf{h} d\omega, \]  

(49)

where \( \mathbf{h} = (h_1, h_2, 0) \) is the horizontal-offset coordinate and \( \mathbf{p} \cdot \mathbf{h} = p_1 h_1 + p_2 h_2 \). In the Appendix, it is shown that this approach produces an estimate of the plane-wave reflection coefficient for a horizontal reflector, multiplied by the energy of the corresponding downgoing plane wave. To obtain an unbiased estimate of the reflection coefficient, it is necessary to divide by this factor (which we will refer to as the source correction term). The result is

\[ R(p, x, x_3) = \frac{\int \int \tilde{U} \left( \omega, \mathbf{x} + \mathbf{h}, x_3 \right) \tilde{U}^* \left( \omega, \mathbf{x} - \mathbf{h}, x_3 \right) e^{-i \mathbf{p} \cdot \mathbf{h}} d\mathbf{h} d\omega}{\int |\tilde{U}(\omega, \mathbf{p}, x, x_3)|^2 d\omega}. \]  

(50)

where \( \tilde{U}(\omega, \mathbf{p}, x, x_3) \) is the local Fourier transform as defined in the Appendix. It may be necessary to apply a stabilizing procedure as discussed in Vivas et al. (2009). To obtain an estimate of the reflection coefficient for a range of \( p \)-values it is necessary to average the expression in equation 50 over many shots.

NUMERICAL RESULTS

Throughout our numerical examples, we employ a Fourier finite-difference approach to account for lateral medium variations, while correcting for the transmission loss using the minimum velocity within each thin-slab. Further, we consider wave-propagation in a 2D medium. First, the input data to migration are modeled over a medium with density contrasts only; hence, the reflection coefficients are independent of angle. Next, we compare conventional pressure normalization to the flux-normalized approach on
a field data example where we, in a quantitative fashion, compare the estimated reflectivity. For the pressure-normalized approach, we set the transmission correction to unity. Equation 50 is used to output AVP gathers on selected locations.

**Imaging in a lateral invariant medium**

In our first test, we consider a laterally-invariant medium with a constant velocity of 2000 m/s and with density contrasts in depth at 1, 2, and 3 km as illustrated in Figure 1. We choose this model because in this particular case we will have angle-independent reflection coefficients. We create a synthetic split-spread shot gather over the laterally invariant medium using a finite-difference modeling scheme. In Figure 2 we show the modeled shot, and the migrated shot is shown in Figure 3.

To extract AVP or AVA information at a reflector position, we need information from more than one shot because each shot gives limited angle information. A schematic representation of angle information available from one shot is shown in Figure 4. Using information from the wave equation angle-transform, this can further be illustrated by plotting $I(p, x)$ (in gray-scale) overlaid the source correction term (in color-scale) at midpoints $x_{m,1} = -0.5$ km, $x_{m,2} = 0.0$ km, and $x_{m,3} = 0.5$ km shown in Figure 5.

By simulating more shots over one midpoint location $x_m$, we can extract angle information for larger angle coverage as shown schematically in Figure 6. We simulate 100 shots with a shot-distance of 10 m on both sides of $x_m$, in addition to one shot just above $x_m$. This produces the angle coverage shown in Figure 7, where we plot $I(p, x)$ overlaid the corresponding source illumination for the fixed midpoint location $x_m$. We notice that the angle coverage for each reflector in depth is different (as expected).

At each reflector depth, we extract the peak amplitude of $R(\theta, x_m)$ using equation 50 with $\sin \theta = pc$. The result is depicted in Figure 8. We have plotted the AVA response for the reflector at 1 km up to 50°, the reflector at 2 km up to 35°, and the reflector at 3 km up to 25°. In this example, we expect an angle-independent reflectivity, and from the result we see that the reflectivity is recovered relatively accurately for a wide range of angles. Due to a limited aperture, edge effects impact the results, and the largest angles on each reflector are affected.

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**Figure 1.** Densities used in the finite-difference modelling over the lateral invariant model example.

**Figure 2.** A synthetic shot gather from a finite-difference modeling over the lateral invariant model example.

**Figure 3.** Migrated shot from the lateral invariant model with flux-normalized wavefields.

**Figure 4.** Schematic representation of angle information contained in one shot, where $x_{m,1}$ and $x_{m,2}$ are midpoint locations with information around angles $\theta_1$ and $\theta_2$. For each midpoint location, each shot only gives limited angle information.
Marine seismic data example

We apply conventional pressure-normalized and the derived flux-normalized methods to a field data set from the Nordkapp Basin. The basin is located offshore Finnmark, in the Norwegian sector of the Barents Sea. It is an exploration area which exhibits complex geology and is challenging for seismic imaging. We have extracted a subset of a 2D survey which partially covers two salt dome structures. Figure 9 shows the velocity model used in the migration.

The data set is composed by collecting and combining streamer data in two directions, providing a split-spread configuration. Each shot is separated by 25.0 m, and in our example we have included a total of 775 shots. Each streamer has 1296 receivers with a hydrophone distance of 12.5 m and a total offset of about 8100 m on both sides of the source location. Notice that no demultiple is applied in the preprocessing step. In Figure 10, we show one extracted shot which is input to migration.

In the imaging, we have used a source signature comparable to a Ricker wavelet with a peak frequency of 17 Hz. We used 3–35 Hz of the frequency content of the data, and imaged the data down to 10 km. The total aperture of each shot was 16 km. For the pressure-normalized and the flux-normalized wavefield decomposition, we migrate the data set with the same downward continuation scheme and the same imaging condition. That is, we use a third-order Fourier finite-difference migration operator and an imaging condition which estimates the reflectivity by accounting for the source illumination. The flux-normalized migration has an approximation to the transmission loss correction applied at thin-slab boundaries using the minimum velocity at each slab given by the aperture of each migrated shot. In Figures 11 and 12, the pressure-normalized and the flux-normalized migrated sections are shown, respectively. By inspecting and comparing both sections, we see that we have an apparent similar amplitude response.

Figure 5. Slowness coverage $I(p,x)$ from one shot (gray-scale) overlaid the corresponding source correction (color-scale) for (left) $x_{m,1} = -0.5$ km; (middle) $x_{m,2} = 0.0$ km; and (right) $x_{m,3} = 0.5$ km.

Figure 6. Schematic representation of angle coverage $\theta$ at one midpoint location $x_m$ from a range of shots. To extract a larger range of angle coverage, each midpoint location requires several shots.

Figure 7. Angle coverage $I(p,x)$ (gray-scale) from one spatial location $x_m$, overlaid the corresponding source correction (color-scale), where the contribution from multiple shots are included.

Figure 8. Peak amplitudes at each reflector at 1 km (red), 2 km (green), and 3 km (blue) for one spatial location $x_m$.

Figure 9. The velocity model used in the Nordkapp field data example.
To quantify the difference between the migrated sections, we compute the difference between the absolute values of each section. The difference plot is shown in Figure 13. The colors red or black indicate that the flux-normalized image provides higher or lower amplitudes than the pressure-normalized image, respectively. In the shallower part of the difference image, from the surface to about 2 km, the pressure-normalized image appears to be dominating; however, these parts of the sections are also contaminated by low-frequency-migration noise. In the sediment basin between the two salt-domes, that is, below and around a distance of 6 km, no coherent energy appears below 2 km. Around a distance of approximately 14 to 16 km, at about 8 km depth, the flux-normalized images give a higher amplitude response on some parts of a few subsurface reflectors. The peak amplitude difference is around one-tenth of the reflectivity image amplitudes.

Further, we extract a slowness gather from each of the migration approaches corresponding to a lateral position of 14.4 km and these are shown in Figure 14. Figure 14a and 14b shows the output from the pressure-normalized and flux-normalized approach, respectively. The gathers look similar. Next, we extract one event at 7.8 km of depth on these gathers, as shown in Figure 15. For this event, we extract the peak amplitudes for each of the migrated reflectors, and plot these in Figure 16 (top), where the red curve is the flux-normalized peak amplitude and the blue curve is the pressure-normalized peak amplitude. Finally, we take the difference between the normalized peak amplitudes (bottom), where the positive and negative values correspond to higher and lower peak amplitudes when using flux-normalized variables. The plot shows differences between the results from the different approaches, and explains the difference plot in Figure 13.
CONCLUSIONS

By directionally decomposing a wavefield using a flux-normalized eigenvalue decomposition, we have derived initial conditions for pre-stack depth migration of common-shot data. This decomposition simplifies the system of differential equations. Further, by neglecting interaction between directional components, we derive propagators for flux-normalized wavefields where we formulate a transmission-loss compensation approach for flux-normalized wavefield propagation. By using the wave equation angle transform, we formulate an estimate of the plane-wave reflection coefficient for a horizontal reflector. From our 1D numerical example, we show that a flux-normalized directional decomposition provides accurate amplitude information in a medium where the parameters are function of depth only. Finally, we extend our approach to a laterally varying media. From a field data example, we observe some differences in the strength of the estimated reflectivity compared to a pressure-normalized approach.

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APPENDIX A

THE WAVE EQUATION ANGLE TRANSFORM

In terms of local Fourier transforms,

\[
\hat{U}(\omega, k', x, x_3) = \int \hat{U}(\omega, x + \frac{\mathbf{h}}{2}, x_3) e^{-\mathbf{k} \cdot \mathbf{x}} d\mathbf{h}
\]

(A-1)

and

\[
\hat{D}(\omega, k', x, x_3) = \int \hat{D}(\omega, x - \frac{\mathbf{h}}{2}, x_3) e^{-\mathbf{k} \cdot \mathbf{x}} d\mathbf{h}
\]

(A-2)

the wave equation angle-transform in equation can be written

\[
\hat{I}(p, x, x_3) = \int \hat{U}(\omega, k', x, x_3) \hat{D}(\omega, k', x, x_3)
\]

(A-3)

where \( k' = (k'_1, k'_2) \) and \( k = (k_1, k_2) \). Snell’s law is

\[
\hat{U}(\omega, k', x, x_3) = R(p, x, x_3) \hat{D}(\omega, k, x, x_3) \delta(k' - k)
\]

(A-4)

Inserted into equation A-3 this gives

\[
\hat{I}(p, x, x_3) = \left( \frac{1}{2\pi} \right)^2 \int \int \int R(p, x, x_3) |\hat{D}(\omega, k, x, x_3)|^2 e^{i(k \cdot \mathbf{op})} d\mathbf{k} d\omega d\mathbf{h}
\]

(A-5)

where \( k' = k = \mathbf{op} \). Further simplifications give

\[
\hat{I}(p, x, x_3) = \left( \frac{1}{2\pi} \right)^2 \int \int \int R(p, x, x_3) |\hat{D}(\omega, k, x, x_3)|^2 \delta(k - \mathbf{op}) d\omega d\mathbf{h}
\]

(A-6)

and finally

\[
\hat{I}(p, x, x_3) = R(p, x, x_3) \int |\hat{D}(\omega, \mathbf{op}, x, x_3)|^2 d\omega
\]

(A-7)

The derivations above are only approximate, because finite-aperture effects have not been taken into consideration. The result is valid only for a horizontally reflecting plane.

REFERENCES


Pierce, A. D., 1981, Acoustics: An introduction to its physical principles and applications; McGraw-Hill.


