Differential Semblance Migration Velocity Analysis via Reverse Time Migration: Gradient Computation

Chao Wang

The Rice Inversion Project Rice University

February 20, 2009



Introduction

- 2 RTM Formula and DSMVA Objective Function
- Gradient Computation



5 Future Work





- Differential Semblance Velocity Analysis
 - Smoothness
 - Convexity
- Reverse Time Migration
 - Dip limitation
- DSMVA-RTM



Acoustic wave equation with constant density

$$\left(\frac{1}{c^2(\mathbf{x})}\frac{\partial^2}{\partial t^2} - \nabla^2\right) p(\mathbf{x}, t; \mathbf{x}_s) = f(t)\delta(\mathbf{x} - \mathbf{x}_s)$$

$$\begin{split} \mathbf{x} &= \text{position vector} \\ \mathbf{x}_{s} &= \text{position of the point source} \\ c(\mathbf{x}) &= \text{velocity} \\ p(\mathbf{x}, t; \mathbf{x}_{s}) &= \text{pressure} \\ \text{Source time function } f(t) &= \delta(t) \\ \text{Two-way wave equation operator } \mathbf{L} &:= \frac{1}{c^{2}(\mathbf{x})} \frac{\partial^{2}}{\partial t^{2}} - \nabla^{2} \end{split}$$



Source Wavefields (forward in time):

Receiver Wavefields (backwards in time):

$$\mathbf{L}R(\mathbf{x}, t; \mathbf{x}_s) = \int d(\mathbf{x}_r, t; \mathbf{x}_s) \delta(\mathbf{x} - \mathbf{x}_r) d\mathbf{x}_r$$
$$R \equiv 0 \text{ for } t > t_{max}$$



Write

$$c(\mathbf{x}) = v(\mathbf{x}) + \delta v(\mathbf{x}).$$

Forward Born Modeling:

$$F[v]\delta v = \delta S|_{surface}$$

Adjoint of Forward Modeling:

$$F^*d = I$$



Migration formula:

$$I(\mathbf{x},h) = \int S(\mathbf{x}+h,t;\mathbf{x}_s)R(\mathbf{x}-h,t;\mathbf{x}_s)\,d\mathbf{x}_s\,dt$$

DSMVA-RTM objective function:

$$J[v] = \frac{1}{2} \|PI\|^2$$



Gradient Computation

For $u = P^*PI$, DI^*u gives the gradient. Introducing g_r solving

$$\mathbf{L}g_r(\mathbf{x}, t; \mathbf{x}_s) = \int R(\mathbf{x} - 2h, t; \mathbf{x}_s)u(t - h, h) dh$$
$$g_r \equiv 0 \text{ for } t > t_{max}, BC$$

Introducing g_s solving

$$Lg_s(\mathbf{x}, t; \mathbf{x}_s) = \int S(\mathbf{x} + 2h, t; \mathbf{x}_s) u(t+h, h) dh$$
$$g_s \equiv 0 \text{ for } t < 0, BC$$

The gradient of DSMVA-RTM is

$$\nabla_{\mathbf{v}}J = \int \frac{2}{\mathbf{v}^3} \left(\frac{\partial^2 S}{\partial t^2} g_r + \frac{\partial^2 R}{\partial t^2} g_s \right) \, d\mathbf{x}_s \, dt$$







One-way wave equation migration:

- Pros: low computational cost
- Cons: dip limitation

Two-way wave equation migration:

- Pros: no dip limitation
- Cons: high computational cost
- Cons: amplitude correction required



- Amplitude Correction (see Rami's work)
- Modification of existing RTM code based on SEAMX, TSOpt to compute objective function and to verify its convexity
- Future modification to create gradient
- Comparison with DS based on downward continuous extrapolation



Perturbation of objective function:

$$\delta J = \frac{1}{2} \langle \delta(PI), PI \rangle + \frac{1}{2} \langle PI, \delta(PI) \rangle$$
$$\implies \delta J = \langle \delta c, Re((DI)^* P^* PI) \rangle$$

Gradient of objective function:

$$\nabla_c J = Re\{(DI)^*(P^*PI)\}$$

For arbitrary u(x, h), we have

$$\langle \delta I, u \rangle = \langle DI \delta c, u \rangle = \langle \delta c, DI^* u \rangle$$

When $u = P^* PI$, $DI^* u$ gives the gradient.



$$\begin{split} \langle \delta I, u \rangle &= \int \overline{\delta I} u(x, h) \, dx \, dh \\ &= \int \delta S(x + h, x_s, t) \overline{R}(x - h, x_s, t) u(x, h) \, dx \, dx_s \, dt \, dh \\ &+ \int S(x + h, x_s, t) \overline{\delta R}(x - h, x_s, t) u(x, h) \, dx \, dx_s \, dt \, dh \\ &= \int \delta S(x, x_s, t) \{ \int \overline{R}(x - 2h, x_s, t) u(x - h, h) \, dh \} \, dx \, dx_s \, dt \\ &+ \int \overline{\delta R}(x, x_s, t) \{ \int S(x + 2h, x_s, t) u(x + h, h) \, dh \} \, dx \, dx_s \, dt \end{split}$$



Introducing g_r solving

$$\mathbf{L}g_{r}(\mathbf{x}, t; \mathbf{x}_{s}) = \int R(\mathbf{x} - 2h, t; \mathbf{x}_{s})u(t - h, h) dh$$
$$\int_{\Sigma} \nabla Sg_{r} - S \nabla g_{r} = 0$$
$$\frac{\partial S}{\partial t}g_{r} - S \frac{\partial g_{r}}{\partial t}(\mathbf{x}, 0; \mathbf{x}_{s}) = 0$$
$$\frac{\partial S}{\partial t}g_{r} - S \frac{\partial g_{r}}{\partial t}(\mathbf{x}, t_{max}; \mathbf{x}_{s}) = 0$$

Introducing g_s solving

$$\mathbf{L}g_{s}(\mathbf{x}, t; \mathbf{x}_{s}) = \int S(\mathbf{x} + 2h, t; \mathbf{x}_{s})u(t + h, h) dh$$
$$\int_{\Sigma} \nabla Rg_{s} - R\nabla g_{s} = 0$$
$$\frac{\partial R}{\partial t}g_{s} - R\frac{\partial g_{s}}{\partial t}(\mathbf{x}, 0; \mathbf{x}_{s}) = 0$$
$$\frac{\partial R}{\partial t}g_{s} - R\frac{\partial g_{s}}{\partial t}(\mathbf{x}, t_{max}; \mathbf{x}_{s}) = 0$$

Then,

$$\begin{aligned} \langle \delta I, u \rangle &= \int (\delta S \mathbf{L} g_r + \delta R \mathbf{L} g_s) \, dx \, dx_s \, dt \\ &= \int ((\mathbf{L} \delta S) g_r + (\mathbf{L} \delta R) g_s) \, dx \, dx_s \, dt \\ &= \int \frac{2\delta c}{c^3} \frac{\partial^2 S}{\partial t^2} g_r + \frac{2\delta c}{c^3} \frac{\partial^2 R}{\partial t^2} g_s \, dx \, dx_s \, dt \end{aligned}$$

Thus

$$\nabla_{c}J = \int \frac{2}{c^{3}} \left(\frac{\partial^{2}S}{\partial t^{2}}g_{r} + \frac{\partial^{2}R}{\partial t^{2}}g_{s}\right) dx_{s} dt$$

