

# Multipole Point Representation of Seismic Sources for Joint Model-Source FWI

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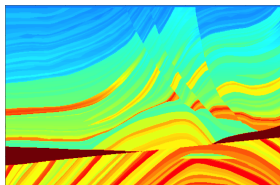
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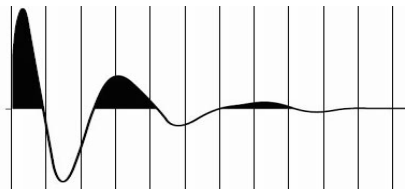
RICE

- 2010 - University of Texas at El Paso (UTEP),  
B.S. in Physics and Applied Mathematics
- 2015 - Rice University,  
M.A. in Computational and Applied Mathematics (CAAM)  
Thesis: "*Discontinuous Galerkin and Finite Difference Methods  
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- present - Rice University,  
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Thesis: "*Joint Model and Minimal-Source Full Waveform  
Inversion for Seismic Imaging Under General (Anisotropic)  
Sources*"

- 1 Multipole Point-Source (MPS) Approximation and Representation
- 2 MPS Numerical Considerations and Results
- 3 Future Work



accurate model estimation



accurate **estimation** & **representation**  
of seismic sources

**Estimation:** joint model-source parameter estimation

**Representation:** multipole point-source (MPS)

- preserve point-source representation
- account for source radiation pattern

# Importance of Modeling Source Anisotropy: Case Study [Minkoff & Symes, 1997]

- plane-wave marine data from Gulf of Mexico
- estimated:
  - P-wave background velocity
  - 3 elastic parameter reflectivities
  - anisotropic source term (airgun array)
- viscoelastic layered model, primary reflections:

$$d(t, p) = f(t, p) * \tilde{r}(t, p)$$

- $t$  = time
- $p$  = slowness
- $d$  = data
- $f$  = source term
- $\tilde{r}$  = reflectivity

- anisotropic source model:

$$f(t, p) = \sum_{i=0}^N f_i(t) L_i(p)$$

- $L_i = i^{th}$  Legendre polynomial

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Better data fit when inverting for anisotropic source:

- fixed isotropic source (airgun model): 55% misfit
  - estimated isotropic source: 53% misfit
  - estimated anisotropic source: 29% misfit
- \* Better match with P-wave impedance well-logs and estimated P-wave reflectivity.
- \*  $v_P/v_S$  reflectivity relation helped detect gas-sand target, ONLY with estimated anisotropic source

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## Multipole Point-Source (MPS) Approximation

Consider acoustic wave equation:

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

General source term  $f$  can be approximated via truncated MPS series centered at some  $\mathbf{x}^*$ , [Santosa & Symes, 2000]:

$$f(\mathbf{x}, t) \approx \sum_{|\mathbf{s}|=0}^N (-1)^{|\mathbf{s}|} f_{\mathbf{s}}(t) \mathbf{D}^{\mathbf{s}} \delta(\mathbf{x} - \mathbf{x}^*),$$

where  $\mathbf{s} = [s_1, s_2, s_3]$  is a multi-array index, and

$$\mathbf{D}^{\mathbf{s}} = \left( \frac{\partial}{\partial x_1} \right)^{s_1} \left( \frac{\partial}{\partial x_2} \right)^{s_2} \left( \frac{\partial}{\partial x_3} \right)^{s_3}.$$

## *Multipole Point-Source (MPS) Approximation*

Number of terms in MPS series depends on

- desired accuracy
- size of source region relative to wavelengths of waves generated
- complexity of source radiation pattern

In most seismic applications, point-source representation is justified

- size of source region  $\sim 10m$
- wavelengths  $\sim 100m$

**My interest:** use of MPS series to represent source terms, modeling anisotropy/directivity of seismic sources.

# Multipole Point-Source (MPS) Representation

- Acoustic equations (velocity-pressure form):

$$\frac{\partial}{\partial t} v_i - \frac{1}{\rho} \frac{\partial}{\partial x_i} p = f_i^{[v]}$$

$$\frac{\partial}{\partial t} p - \kappa \frac{\partial}{\partial x_i} v_i = f^{[p]}$$

- $p(\mathbf{x}, t)$  = pressure field
- $v_i(\mathbf{x}, t)$  = particle velocity field
- $\rho(\mathbf{x})$  = density
- $\kappa(\mathbf{x})$  = bulk modulus

- Linear elasticity equations (velocity-stress form):

$$\rho \frac{\partial}{\partial t} v_i - \frac{\partial}{\partial x_j} \sigma_{ij} = f_i^{[v]}$$

$$\frac{\partial}{\partial t} \sigma_{ij} - C_{ijmn} \frac{\partial}{\partial x_n} v_m = f_{ij}^{[\sigma]}$$

- $\sigma_{ij}(\mathbf{x}, t)$  = stress field
- $C_{ijmn}(\mathbf{x})$  = Hooke's tensor

- source terms:  $f^{[p]}, f_i^{[v]}, f_{ij}^{[\sigma]}$

## Multipole Point-Source (MPS) Representation

Represent source terms as “linear combination” of MPS basis:

$$f^{[p]}(\mathbf{x}, t) \approx \sum_n f_n(t) b^n(\mathbf{x} - \mathbf{x}^*)$$

$$f_i^{[v]}(\mathbf{x}, t) \approx \sum_n f_n(t) b_i^n(\mathbf{x} - \mathbf{x}^*)$$

$$f_{ij}^{[\sigma]}(\mathbf{x}, t) \approx \sum_n f_n(t) b_{ij}^n(\mathbf{x} - \mathbf{x}^*)$$

- $f_n(t)$  - MPS coefficients
- $b^n(\mathbf{x})$  - MPS basis, related to  $\mathbf{D}^s \delta(\mathbf{x})$
- \* Impose constraints on source radiation pattern via choice of MPS basis  $b^n(\mathbf{x})$
- \* Forward map will be linear w.r.t. MPS coefficients  $f_n(t)$

## **Problem:**

- numerically solve PDEs with singular source terms

$$\mathbf{D}^s \delta(\mathbf{x} - \mathbf{x}^*)$$

## **Interested in:**

- approximation of singular sources terms in FD methods, uniform grids
- preserving spatial convergence rate of FD methods

**Solution:** approximate singularity with continuous function of compact support, [Waldén, 1998], [Tornberg & Engquist, 2004], [Pettersson & Sjögreen, 2010]

$$D^s \delta(x) \approx \delta_\varepsilon^s(x) \in C_0([- \varepsilon, \varepsilon])$$

where  $\delta_\varepsilon^s$  satisfies certain properties (discrete moments) dependent on

- $q$  - accuracy order of approximation
- $s$  - derivative order
- $\varepsilon$  - width of support

\* Condition on width of support:

$$2\varepsilon \geq (q + s)h$$

# MPS Numerical Considerations

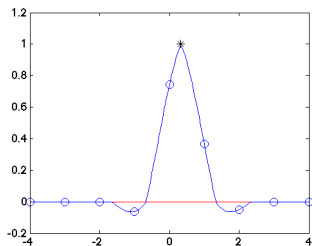


Figure 1:  $\delta_\varepsilon^s(x - \alpha)$  with  $s = 0$ ,  $q = 4$ ,  $\alpha = h/3$ ,  $\varepsilon = 4h$ .

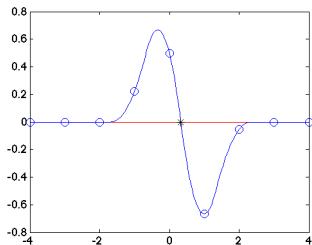
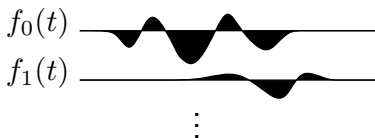


Figure 2:  $\delta_\varepsilon^s(x - \alpha)$  with  $s = 1$ ,  $q = 3$ ,  $\alpha = h/3$ ,  $\varepsilon = 4h$ .

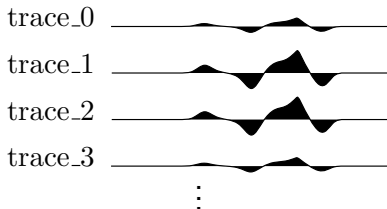
# *IWave Implementation of MPS Representation*

From MPS representation

$$f(x, t) = \sum_n f_n(t) b^n(x - x^*)$$



To time traces at FD grid points





# Numerical Experiments

**Goal:** preserve spatial convergence rates of FD methods

**Setup:**

- solved acoustic equations in velocity-pressure form
- staggered grid FD solver (2-2 and 2-4)
- homogenous medium;  $\rho = 1 \text{ kg/m}^3$ ,  $c = 4 \text{ km/s}$
- source term , with  $s = 0, 1$ ,

$$f^{[p]}(\mathbf{x}, t) = f(t) \left( \frac{\partial}{\partial x_2} \right)^s \delta(\mathbf{x} - \mathbf{x}^*)$$

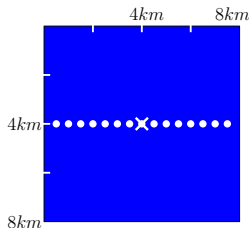


Figure 3: Homogeneous model.

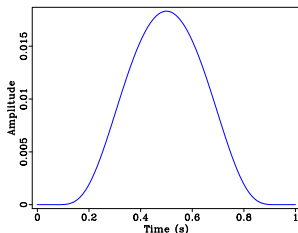


Figure 4: MPS coefficient  $f(t)$ .

Convergence rate calculated approximately by

$$R(\mathbf{x}_r) = \log_2 \left( \frac{\|\rho_h(\mathbf{x}_r, \cdot) - \rho_{h/2}(\mathbf{x}_r, \cdot)\|}{\|\rho_{h/2}(\mathbf{x}_r, \cdot) - \rho_{h/4}(\mathbf{x}_r, \cdot)\|} \right),$$

where norm  $\|\cdot\|$  is chosen to be either  $\|\cdot\|_2$  or  $\|\cdot\|_\infty$  defined by

$$\|\rho(\mathbf{x}_r, \cdot)\|_2 := \sqrt{\Delta t \sum_k |\rho(\mathbf{x}, t_k)|^2},$$

$$\|\rho(\mathbf{x}_r, \cdot)\|_\infty := \max_k |\rho(\mathbf{x}_r, t_k)|.$$

h-refinement [m]:

■ 2-2: 10, 20, 40

■ 2-4: 20, 40, 80

\*  $\Delta t = 0.5 \text{ ms}$ , sufficiently small that spatial error dominates, i.e., approximate semi-discrete solution

# Numerical Results

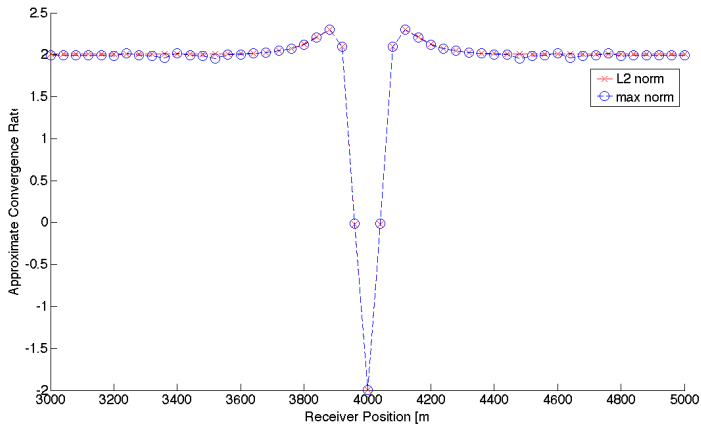


Figure 5: Rates for 2-2 staggered grid scheme with  $\delta(\mathbf{x} - \mathbf{x}^*)$  source term and approximation  $\delta_{\varepsilon_1}^0 \delta_{\varepsilon_2}^0$  with  $q = 2$ .

# Numerical Results

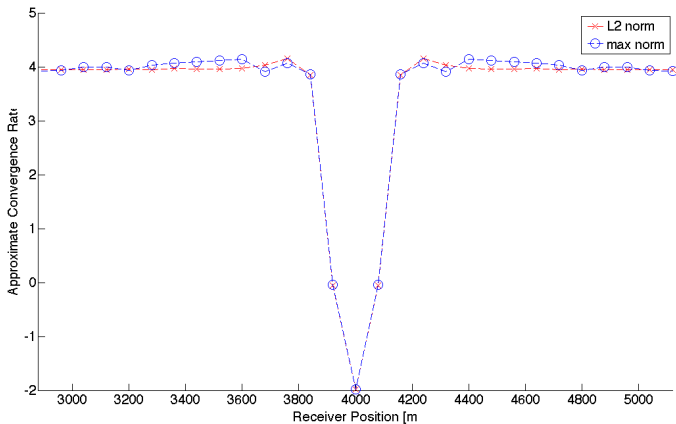


Figure 6: Rates for 2-4 staggered grid scheme with  $\delta(\mathbf{x} - \mathbf{x}^*)$  source term and approximation  $\delta_{\varepsilon_1}^0 \delta_{\varepsilon_2}^0$  with  $q = 4$ .

# Numerical Results

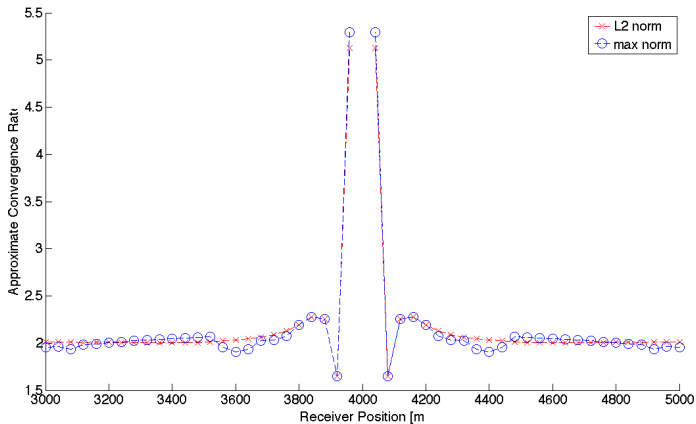


Figure 7: Rates for 2-2 staggered grid scheme with  $\frac{\partial}{\partial x_2} \delta(\mathbf{x} - \mathbf{x}^*)$  source term and approximation  $\delta_{\varepsilon_1}^0 \delta_{\varepsilon_2}^1$  with  $q = 2$ .

# Numerical Results

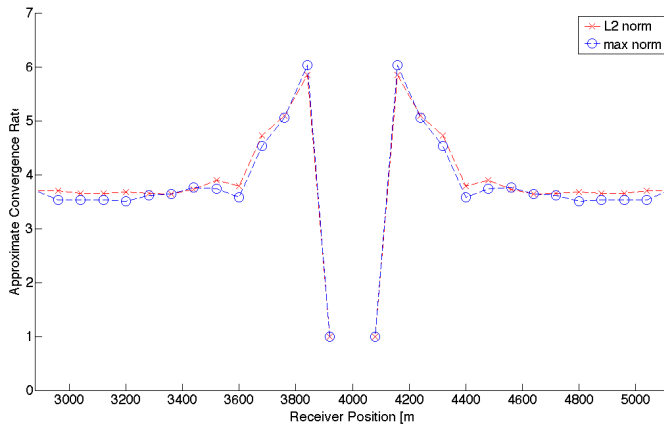


Figure 8: Rates for 2-4 staggered grid scheme with  $\frac{\partial}{\partial x_2} \delta(\mathbf{x} - \mathbf{x}^*)$  source term and approximation  $\delta_{\varepsilon_1}^0 \delta_{\varepsilon_2}^1$  with  $q = 4$ .

# Numerical Results

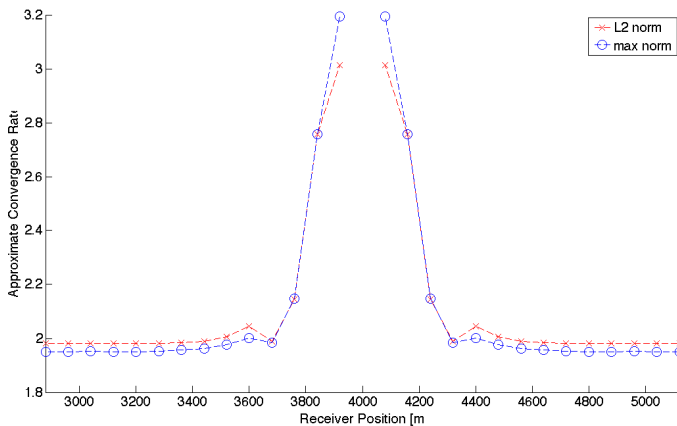


Figure 9: Rates for 2-4 scheme with  $\frac{\partial}{\partial x_2} \delta(\mathbf{x} - \mathbf{x}^*)$  source term and approximation  $\delta_{\varepsilon_1}^0 \delta_{\varepsilon_2}^1$  with  $q = 2$ .

**Idea:** modify domain norm of forward map w.r.t. source such that

$$\|\text{input}\| \sim \|\text{output}\|$$

- better bounded forward map
- better condition on inverse problem for source

Consider  $f^{[p]}$  for acoustics in velocity-pressure form:

$$f^{[p]}(\mathbf{x}, t) = f(t) \left( \frac{\partial}{\partial x_2} \right)^s \delta(\mathbf{x} - \mathbf{x}^*).$$

Resulting data (pressure field) will look like some derivative of MPS coefficient  $f(t)$ ,

$$p(\mathbf{x}, t) \sim f^{s+r}.$$



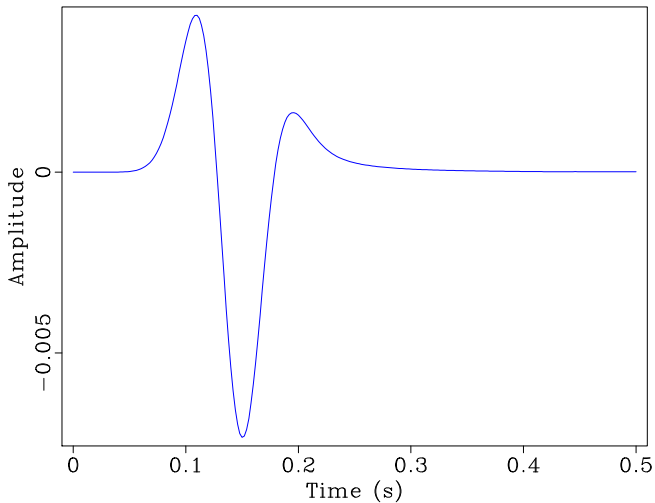


Figure 10: Data trace for source with  $s = 0$ .

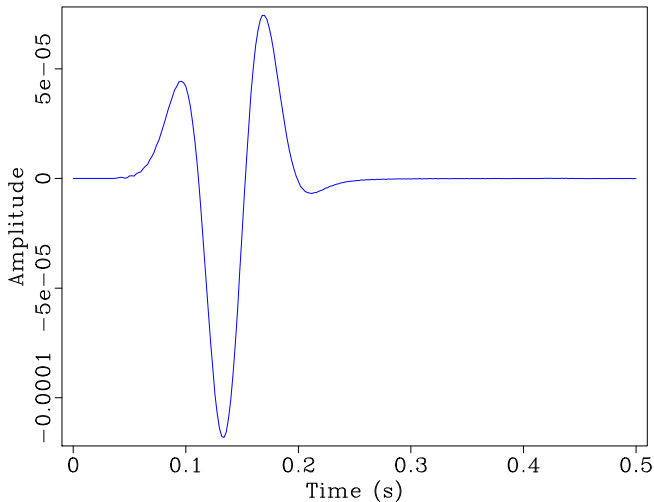


Figure 11: Data trace for source with  $s = 1$ .

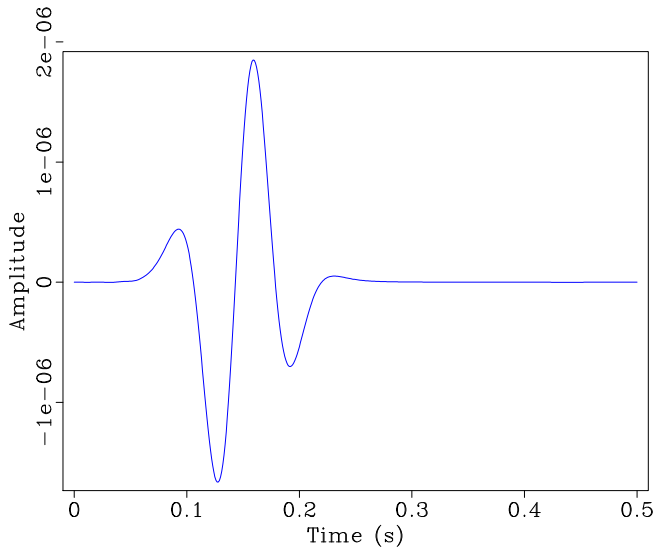


Figure 12: Data trace for source with  $s = 2$ .

Denote forward map  $F : \mathfrak{M} \times \mathfrak{F} \rightarrow \mathfrak{D}$ ,

$$F(\mathbf{m}, \mathbf{f}) \equiv F(\mathbf{m})\mathbf{f} = \mathbf{d}$$

- $\mathfrak{M}$  - model space;  $\mathbf{m}$  - model parameters
- $\mathfrak{F}$  - source space;  $\mathbf{f}$  - MPS coefficients
- $\mathfrak{D}$  - data space;  $\mathbf{d}$  - data vector

Introduce weight  $W$  such that

$$\|W\mathbf{f}\|_{\mathfrak{F}} \sim \|F(\mathbf{m})\mathbf{f}\|_{\mathfrak{D}}.$$

E.g.,

$$W = \left( \frac{d}{dt} \right)^{s+r}$$

### MPS inversions:

- better condition of source estimation from weighted inner product (preconditioning in CG)
- non-uniqueness of MPS representation of seismic sources (null space of forward map)

### MPS + model inversions:

- acoustics  $\rightarrow$  elasticity

Thank You