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SLIM's findings on the Machar dataset Ning Tu, Tim Lin, Zhilong Fang, with contribution from many other SLIM members



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Imaging seismic structure





Acquisition geometry



- Sources and receivers lies on irregular grid. • # of source is 330 with an approximate spacing of 50 m
- # of receiver is 505 with an approximate spacing of 25 m

18300 (m)

n)		17425(m)
ottom depth 95m)	15512(m)	



Source receiver mask of processed data



acquisition mask



Wave velocity field [provided by BP, with anisotropy correction, after interpolation and smoothing]

Water 0m–95m

Slow rock 95m–3000m



-3000m Salt & fast rock 3000m+



Imaging-RTM

Reverse time migration:

all sources, all freq. up to 30F aliasing), 10m grid spacing

Il sources, all freq. up to 30Hz (higher freq. suffer from spacial



Original seismic data

about 25m spatial sampling, 370 traces





Zoomed in



Time (s)





F-K (2D Fourier) spectrum



F-K nearest neighbor spectrum



spatial aliasing of wavefronts



Remove imaging artifacts

We use a cropped section of the model to reduce the turnaround time:

- first reproduce the artifacts
- then try to remove them by interpolating the data



Artifacts reproduced with cropped model





Desired sampling

Desired sampling after interpolation, 10m spacing

Exploiting spatially irregular sampling

Histogram of trace irregularity

-8m

+8m

Trace interpolation via curevelet-domain basis pursuit

- $\mathbf{x} = \mathbf{D}\mathbf{z}$ (assume **x** is not sparse, but **z** is) $\tilde{\mathbf{x}} = \mathbf{D} \cdot \operatorname{argmin} \|\mathbf{z}\|_1$ subject to $\mathbf{y} = \mathbf{A}\mathbf{D}\mathbf{z}$ Z
- ℓ_1 -norm as convex measure of sparsity
- **y** is data with original spatial sampling
- A is trace restriction ('subsampling operator')
- **x** is a seismic wavefield with desired sampling
- \mathbf{z} is a choice of curvelet coefficients that reconstructs \mathbf{X} • **D** is curvelet synthesis operator

Regularization + Interpolation

- Using non-uniform FFT as measurement operator A Fourier coefficients)
- Fourier coefficients

(maps coarse *non-uniform* spatial grid -> fine *uniform* grid 2D

Curvelet dictionary D constructed directly from uniform 2D

F-K spectrum before interpolation

F-K spectrum after interpolation

F-K spectrum masking

Original seismic data

about 25m spatial sampling, 370 traces

Final interpolated wavefield

10m spatial sampling, 925 traces

Zoomed in

Time (s)

Zoomed in

Time (s)

Reverse-time migrated image

[all freq. in 5-30Hz, all sources, 10m grid distance]

free of artifacts

Reverse-time migrated image

[all freq. in 5-60Hz, all sources, 5m grid distance]

Remarks

- The ringing artifacts in the high-velocity-contrast zone are caused by spacial aliasing in the data.
- We will extend the interpolation and imaging to the entire model (work in progress).

Imaging-L1 migration

- source estimation on the fly

using dimensionality reduction to reduce computational cost

Setup

- use the cropped model to reduce the turnaround time
- no freq. subsampling, 50% source subsampling, 20 iterations

luce the turnaround time urce subsampling, 20 iterations

L1 migration w. source estimation, aliased data

L1 migration w. impulsive source, aliased data

L1 migration w. source estimation, de-aliased data

L1 migration w. impulsive source, de-aliased data

Remarks

- Source wavelet is not important *per se*, but import for correctly inverting the image (a.k.a, nuisance parameter). On-the-fly source estimation greatly improves image quality.

Progress and future plan

- RTM of the entire dataset using frequencies up to 60Hz. Currently in progress.
- promoting migration.
- Working directly on the unprocessed pressure and particle velocity wavefield, with Robust EPSI and imaging-withmultiples implementations.

Optimizing the subsampling strategy for more efficient sparsity-

RTM of the entire model using iWave

Lateral distance(m)

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