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# The bridge from orthogonal to redundant transforms in weighted $\ell_1$ minimization

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Seismic Laboratory for Imaging and Modelling Consortium Meeting

December 3rd, 2013

# About myself and time with SLIM



- 2006-2011: Bachelor of Science at University of Calgary, Math and Statistics
- 2011-2013: Master of Science with SLIM, University of British Columbia, Mathematics
- Summer 2013: Internship with Total

# Rest of my talk

Major messages:

- The D-RIP framework (by Needell et al and Shidong et al) and it's role in seismic interpolation via  $\ell_1$ -synthesis
- Analysis and synthesis and choosing the best possible analysis operator
- **③** Improvement of this framework in weighted  $\ell_1$ -synthesis
- This framework provides design criteria for sampling geometries and potentially provide optimal jitter sampling schemes.

### Basic workflow



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# Seismic Acquisition



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# Data Regularization



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# Data Regularization



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### Seismic Wavefield Reconstruction Missing Traces





Figure: Shot gather with missing traces

Image: A mathematical states and a mathem

# Analysis and Synthesis Formulations

**Synthesis**: Find a sparse representation of a wavefield whose reconstruction fits the data.

$$\hat{f}_{synthesis} = D \cdot rgmin_{ ilde{x}} || ilde{x}||_{1,w}$$
 subject to  $||AD ilde{x} - y||_2 \le \epsilon$ 

**Analysis**: Find a wavefield whose forward transform coefficients are sparse and fit the data.

$$\hat{f}_{analysis} = rgmin_{ ilde{f}} ||D^{\dagger}\tilde{f}||_{1,w}$$
 subject to  $||A\tilde{f} - y||_2 \le \epsilon$ 

where  $D^{\dagger}$  is the pseudoinverse of D.

### Seismic Wavefield Reconstruction Example





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### Seismic Wavefield Reconstruction Example





December 3rd, 2013 11 / 21

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# Seismic Wavefield Reconstruction Example with only missing sources - Frequency 11 hz



December 3rd, 2013 12 / 21

### General and Optimal dual analysis

Yulong, Tiebin, Shidong Li, 2012 [2][1], Hargreaves and Yilmaz, 2013

But we have an infinite amount of choices of the analysis operator. So choose  $\tilde{D}$  such that  $\tilde{D}^*f$  is sparse. The natural problem to solve is then:

$$\hat{f}_{general} = rgmin_{ ilde{f}} || ilde{D}^* ilde{f}||_{1,w} \;\; ext{s.t.} \; ||A ilde{f} - y||_2 \leq \epsilon$$

Find an 'optimal' analysis operator:

$$\hat{f}_{optimal} = \argmin_{\tilde{f}, D\tilde{D}^* = I} ||\tilde{D}^*\tilde{f}||_{1, w} \text{ s.t. } ||A\tilde{f} - y||_2 \le \epsilon$$

Equivalence of optimal dual  $\ell_1$ -analysis and synthesis

$$\widehat{f}_{synthesis} = D \cdot rgmin_{\widetilde{x}} ||\widetilde{x}||_1 \;\; subject \; to \;\; ||AD\widetilde{x} - y||_2 \leq \epsilon$$

$$\hat{f}_{optimal} = \mathop{\arg\min}_{\tilde{f}, D\tilde{D}^* = I} ||\tilde{D}^*\tilde{f}||_1 \ s.t. ||A\tilde{f} - y||_2 \le \epsilon$$

$$\hat{f}_{synthesis} = \hat{f}_{optimal}$$

Further still, this theorem can be generalized to the weighted case.

# Recovery conditions

Suppose y = ADx

- **Synthesis** (RIP): A condition imposed on the composition *AD* for error bounds on the recovery of *x*
- Analysis (D-RIP): A condition for recovery imposed on A for error bounds on the recovery of f = Dx

Note: Focus shifts on acquisition geometry

### How equivalence helps us

- Traditional CS: AD does not satisfy RIP due to coherency in D, so these conditions don't explain why synthesis does well.
- $\bullet$  If A satisfies D-RIP then optimal dual  $\ell_1\text{-analysis}$  does well  $\rightarrow$  synthesis does well

Mathematical caveat: The D-RIP bounds depend on us knowing the upper frame bound of optimal dual apriori, but this is not possible. However we find this bound to be independent of the signal in practice.

### Seismic Wavefield Reconstruction



December 3rd, 2013 17 / 21

# Build support information iteratively

Mansour, Herrmann, Yilmaz 2013[3]



- Recovery first frequency slice and estimate support
- Recovery second frequency slice using previous support estimate
- Opdate support estimate and continue . . .

# Gulf of Suez Experiment



# References

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