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Randomization and repeatability in time-lapse marine acquisition Haneet Wason, Felix Oghenekohwo, and Felix J. Herrmann



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Randomized sampling in marine [SINBAD 2013]





Motivation

What are the implications of randomization in time-lapse seismic?

Should we repeat in randomized marine acquisition?



Felix Oghenekohwo, Haneet Wason, and Felix J. Herrmann, "Compressive 4D---economic time-lapse seismic with randomized subsampling and joint recovery", submitted to Geophysics, October 2014.

Haneet Wason, and Felix J. Herrmann, "Time-jittered ocean bottom seismic acquisition", in SEG Technical Program Expanded Abstracts, 2013, p. 1-6. Hassan Mansour, Haneet Wason, Tim T.Y. Lin, and Felix J. Herrmann, "Randomized marine acquisition with compressive sampling matrices", Geophysical Prospecting, vol. 60, p. 648-662, 2012

Time-lapse seismic

Current acquisition *paradigm*:

- compute differences between baseline & monitor survey(s)
- hampered by practical challenges to ensure repetition

New compressive sampling paradigm:

- cheap subsampled acquisition, e.g. via time-jittered marine subsampling
- may offer *possibility* to *relax* insistence on *repeatability*
- exploits insights from distributed compressed sensing

repeat expensive dense acquisitions & "independent" processing



Time-lapse data

Baseline

Monitor



5

4-D signal [10 X]



time samples: **512** receivers: **100** sources: **100**

sampling time: **4.0 ms** receiver: **12.5 m** source: **12.5 m**



Sparse structure via curvelets



6

significant correlation between the vintages





Distributed compressed sensing -joint recovery model (JRM)





Key idea:

- use the fact that different vintages share common information
- components with *sparse* recovery

• invert for *common* components & *differences* w.r.t. the *common*





Time-lapse seismic -w/&w/orepetition

In an *ideal* world $(\mathbf{A}_1 = \mathbf{A}_2)$

- expect good recovery when difference is sparse
- but relies on "exact" repeatability...

In the *real* world $(\mathbf{A}_1 \neq \mathbf{A}_2)$

- no absolute *control* on *surveys*
- calibration errors
- noise...

• JRM simplifies to recovering the difference from $(\mathbf{b}_2 - \mathbf{b}_1) = \mathbf{A}_1(\mathbf{x}_2 - \mathbf{x}_1)$



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9

• JRM simplifies to recovering the difference from $(\mathbf{b}_2 - \mathbf{b}_1) = \mathbf{A}_1(\mathbf{x}_2 - \mathbf{x}_1)$

What does repetition really mean?



Context

Acquire randomized subsamplings for the baseline and monitor surveys

Questions:

- Should we repeat the surveys when doing randomized subsampling?

Aim: recovery of **both** vintages & time-lapse signal from incomplete data

Process/recover independently or jointly to exploit common features of surveys?



Stylized experiments



Stylized experiments

Conduct *many* CS experiments to compare

- *joint* vs *parallel* recovery of signals and the difference
- recovery with same, partially or completely independent matrices
- random acquisition with different numbers of samples



run 2000 different experiments & compute probability of recovery



Sparse signals



z ₀ con	<i>common</i> component				
z 1	<i>"difference"</i>				
\mathbf{Z}_2	<i>"difference"</i>				
\mathbf{x}_1	baseline				
\mathbf{x}_2	monitor				
\mathbf{x}_1 - \mathbf{x}_2	<i>time-lapse</i>				



Independent vs. joint recovery - 100% & 0% overlap in acquisition matrices



Vintages





4-D signal



Joint recovery - varying % of overlap in acquisition matrices



Vintages



4-D signal



Observations

The Joint Recovery Model (JRM) always gives superior results exploits shared information between the vintages

Aim: recovery of **both** vintages & time-lapse signal from incomplete data

Question:

Process/recover *independently* or *jointly* to exploit *common* features of surveys? processing jointly leads to improved recovery of both vintages & time-lapse signal





Synthetic seismic case study

Time-jittered marine acquisition on the grid

% repetition => "exact" repetition

No calibration errors





18

Method



- High permeability zone identified at a depth
- Fluid substitution (gas/oil replaced with brine) simulated to derive monitor velocity model
- Wavefield simulation to generate synthetic
- scales to 11733300 x 114882048



Simulated time-lapse data - time-domain finite differences

Baseline

Monitor



19



time samples: **512** receivers: 100 sources: **100**

sampling time: **4.0 ms** receiver: 12.5 m source: **12.5 m**



Time-jittered marine acquisition

irregularly sampled spatial grid



continuous recording START



continuous recording *STOP*



Conventional vs. time-jittered sources -subsampling ratio = 2, 2 source arrays



21

jittered acquisition 2 (monitor)



"blended" shot gathers number of shots = 100/2 = 50 (25 per array) spatial sampling: **50.0 m (jittered)** vessel speed: 2.50 m/s recording time \approx 1000.0 s/2 = (500.0 s)



Measurements - subsampled and blended

Baseline



22





Monitor recovery - 100% overlap in acquisition matrices

IRS [11.6 dB]

IRS residual







Monitor recovery - 50% overlap in acquisition matrices

IRS [11.0 dB]

IRS residual







Monitor recovery - 25% overlap in acquisition matrices

IRS [10.3 dB]

IRS residual







4-D recovery - 100% overlap in acquisition matrices





4-D recovery - 50% overlap in acquisition matrices





-

Source position (m)

[colormap scale: 10 X]



4-D recovery - 25% overlap in acquisition matrices



[colormap scale: 10 X]



Stacked sections

Baseline



29

4-D signal [10 X]





Stacked sections - 100% overlap in acquisition matrices





30



JRM [24.2 dB]





Stacked sections - 50% overlap in acquisition matrices





JRM [20.0 dB]





Stacked sections - 25% overlap in acquisition matrices











SNR (dB) for stacked sections - average of 5 experiments

overlap	baseline		monitor		4-D signal	
	IRS	JRM	IRS	JRM	IRS	JRM
100%	25.6 ± 1.2	23.9 ± 1.0	25.7 ± 1.1	24.0 ± 1.0	25.0 ± 0.9	23.4 ± 0.8
50%	25.6 ± 1.2	30.9 ± 1.3	24.3 ± 0.9	30.6 ± 1.4	10.1 ± 1.4	18.1 ± 0.9
25%	25.6 ± 1.2	34.4 ± 0.9	23.5 ± 1.3	33.6 ± 0.8	8.5 ± 1.3	15.9 ± 0.7



Observations

depending on the recovery of the vintages

Questions:

Process/recover *independently* or *jointly* to exploit *common* features of surveys? processing jointly leads to improved recovery of both vintages & time-lapse signal

Should we *repeat* the surveys when doing *randomized sub*sampling?

Seismic synthetics show that we do **not** necessarily have to insist on full repetition



Observations

depending on the recovery of the vintages

Questions:

Process/recover *independently* or *jointly* to exploit *common* features of surveys? processing jointly leads to improved recovery of both vintages & time-lapse signal

Should we *repeat* the surveys when doing *randomized sub*sampling?

What does repetition "in-the-field" mean?

Seismic synthetics show that we do **not** necessarily have to insist on full repetition



Notion of repetition

Time-jittered marine acquisition off the grid

With & without calibration errors



Randomized sampling in marine





Randomized sampling in marine



4-D recovery - 50% overlap in acquisition matrices, no calibration errors

4-D recovery - 50% overlap in acquisition matrices, calibration errors \approx 1.0 m (avg.)

4-D recovery - 50% overlap in acquisition matrices, calibration errors \approx 2.8 m (avg.)

4-D recovery - JRM - 50% overlap in acquisition matrices, w/ & w/o calibration errors

no error [12.2 dB]

error ≈ 1.0 m [8.5 dB]

42

error ≈ 2.8 m [3.8 dB]

4-D recovery - JRM - 50% overlap in acquisition matrices

no error [12.2 dB]

error ≈ 1.0 m [8.5 dB]

0% overlap

[2.0 dB]

error ≈ 2.8 m

[3.8 dB]

On the contrary,

calibration errors improve recovery of the vintages!

Monitor recovery - 50% overlap in acquisition matrices, no calibration errors

IRS [11.6 dB]

IRS residual

JRM [13.9 dB]

JRM residual

Monitor recovery - 50% overlap in acquisition matrices, calibration errors \approx 1.0 m (avg.)

IRS [11.4 dB]

IRS residual

JRM [14.5 dB]

JRM residual

Monitor recovery - 50% overlap in acquisition matrices, calibration errors \approx 2.8 m (avg.)

IRS [11.4 dB]

IRS residual

JRM [15.5 dB]

JRM residual

Monitor recovery - JRM - 50% overlap in acquisition matrices

no error [13.9 dB]

error ≈ 1.0 m [14.5 dB]

0% overlap

[18.3 dB]

error ≈ 2.8 m

Monitor residual - JRM - 50% overlap in acquisition matrices

no error

error ≈ 1.0 m

0% overlap

error ≈ 2.8 m

Observations

In the given context of randomized subsampling, calibration errors deteriorate recovery of the time-lapse signal

- *improve* recovery of the *vintages*

"Exact" repeatability of the surveys seems essential for good recovery of the time-lapse signal

Observations

depending on the recovery of the vintages

Questions:

Process/recover *independently* or *jointly* to exploit *common* features of surveys? processing jointly leads to improved recovery of both vintages & time-lapse signal

Should we *repeat* the surveys when doing *randomized sub*sampling? no, as long as one samples sufficiently to recover both vintages jointly calibration errors do not allow "exact" repeatability which is essential for good

- recovery of the time-lapse signal

Seismic synthetics show that we do **not** necessarily have to insist on full repetition

Future work

Application to field datasets

Software release: Time-jittered marine acquisition "off-the-grid"

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