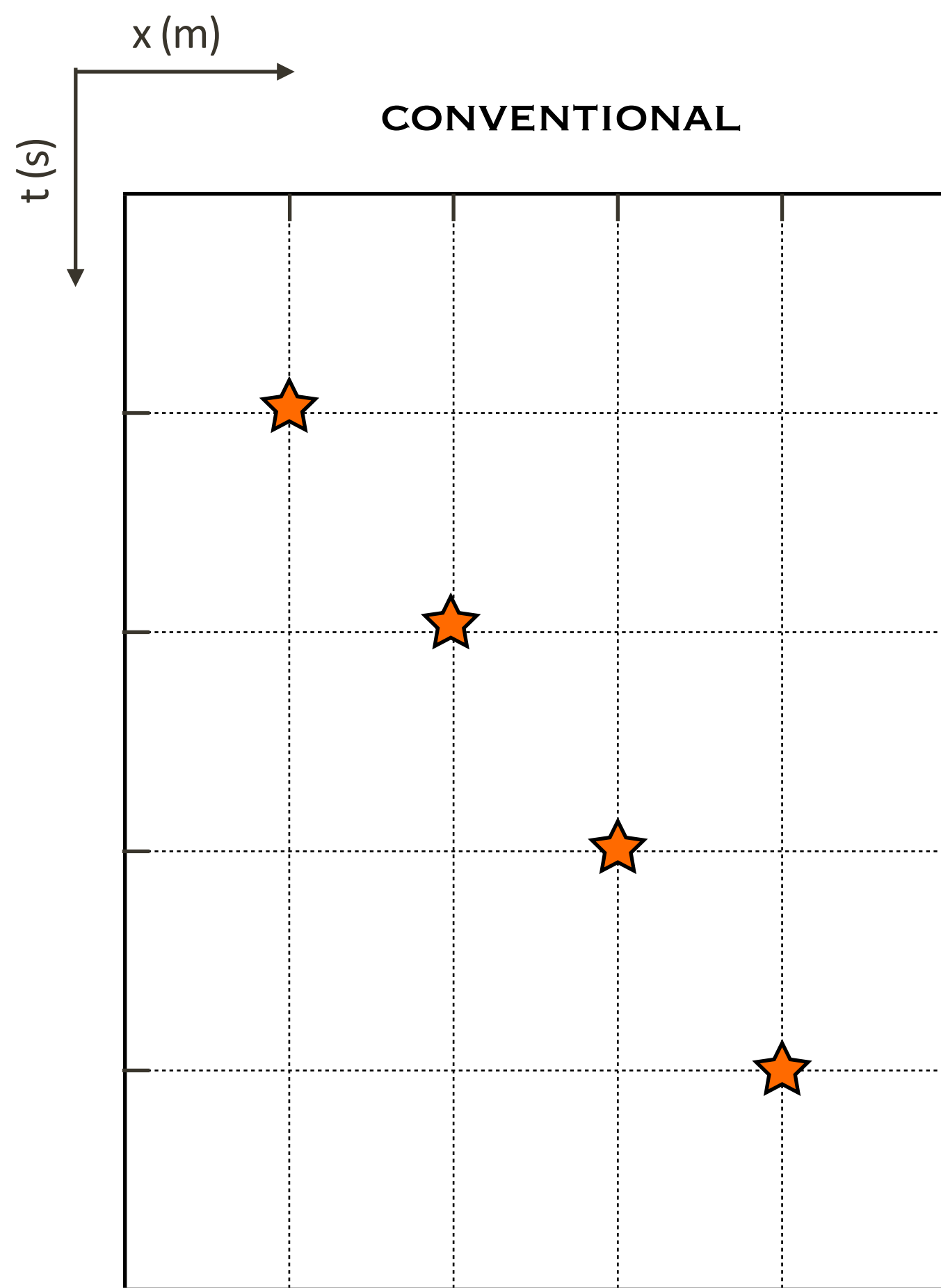


Randomization and repeatability in time-lapse marine acquisition

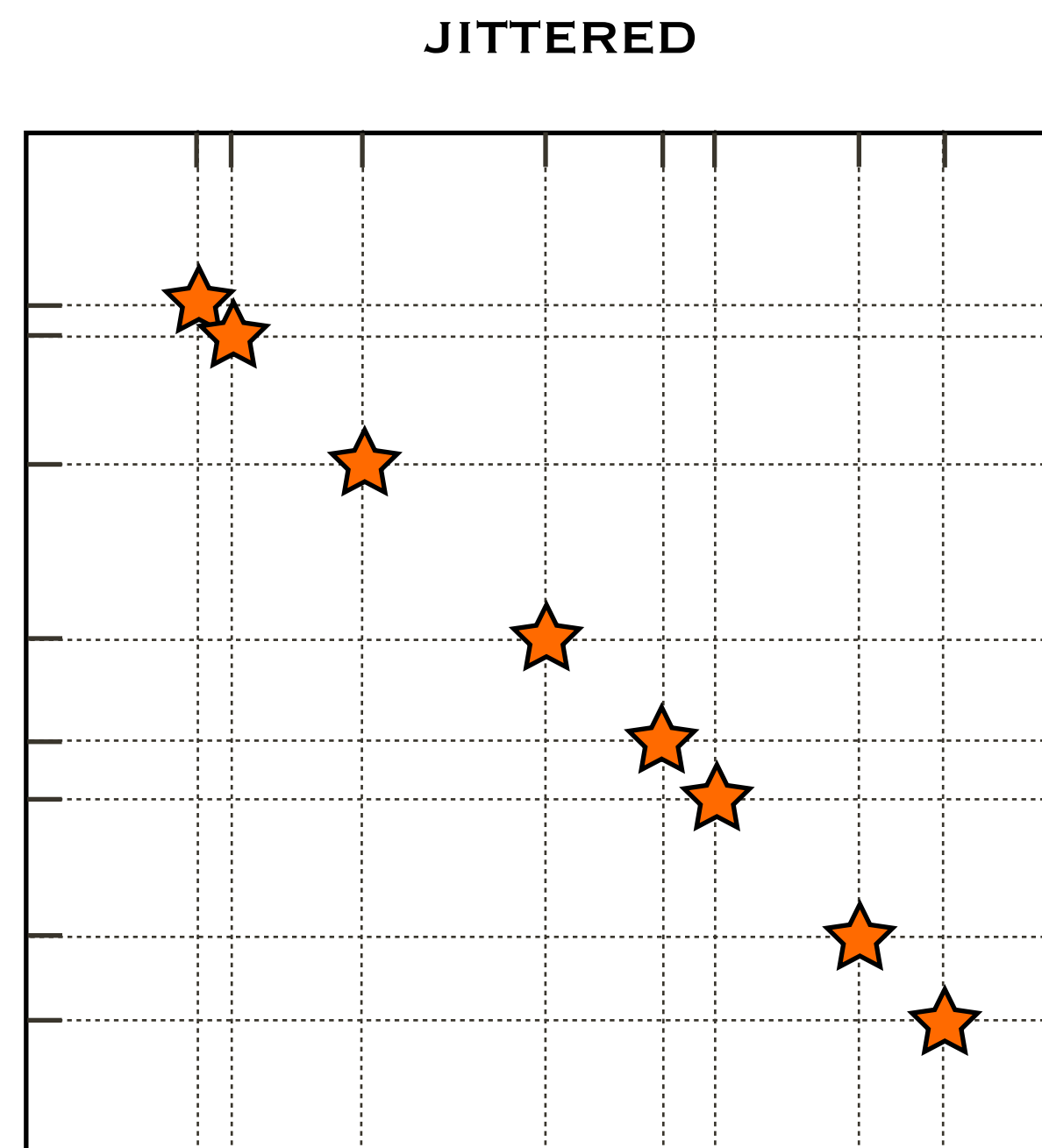
Haneet Wason, Felix Oghenekohwo, and Felix J. Herrmann

Randomized sampling in marine

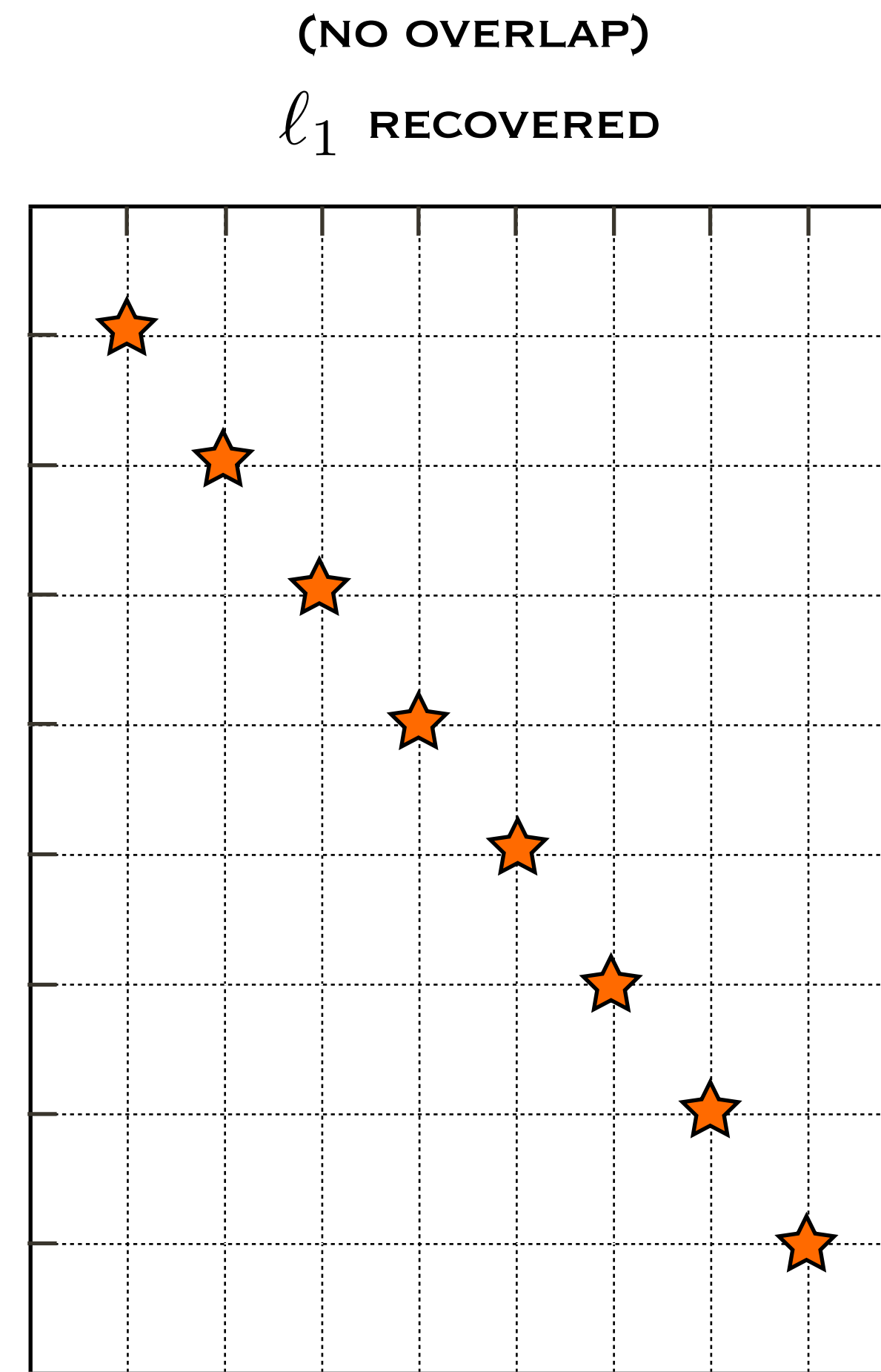
[SEG 2013]



PERIODIC—SPARSE—NO OVERLAP



APERIODIC
COMPRESSED
OVERLAPPING
IRREGULAR



PERIODIC & DENSE

Motivation

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

Motivation

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

Should we repeat in randomized marine acquisition?

Disclaimer

Assumptions:

- ▶ you are a *believer* in *randomized* acquisition & *sparse* recovery
- ▶ *seismic* data & *time-lapse* signal *both* permit *sparse* representations
- ▶ there are *no* calibration *errors* *but* there can be *additive* noise
- ▶ *degree* of repetition refers to *percentage* of a *survey* that is repeated *exactly*

All observations are based on *synthetic* ocean bottom data...

Felix Oghenekohwo, Ernie Esser, and Felix J. Herrmann, "[Time-lapse seismic without repetition: reaping the benefits from randomized sampling and joint recovery](#)", in *EAGE Annual Conference Proceedings*, 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:

- ▶ *repeat **expensive dense** acquisitions & "independent" processing*
- ▶ *compute **differences** between **baseline & monitor** survey(s)*
- ▶ *hampered by **practical** challenges to ensure **repetition***

Felix Oghenekohwo, Ernie Esser, and Felix J. Herrmann, "[Time-lapse seismic without repetition: reaping the benefits from randomized sampling and joint recovery](#)", in *EAGE Annual Conference Proceedings*, 2014.

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Time-lapse seismic

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- ▶ *hampered by **practical** challenges to ensure **repetition***

New compressive sampling paradigm:

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

Compressed sensing *paradigm*

Find representations that reveal structure

- ▶ *transform-domain sparsity* (e.g., Fourier, curvelets, etc.)

Sample to break the structure

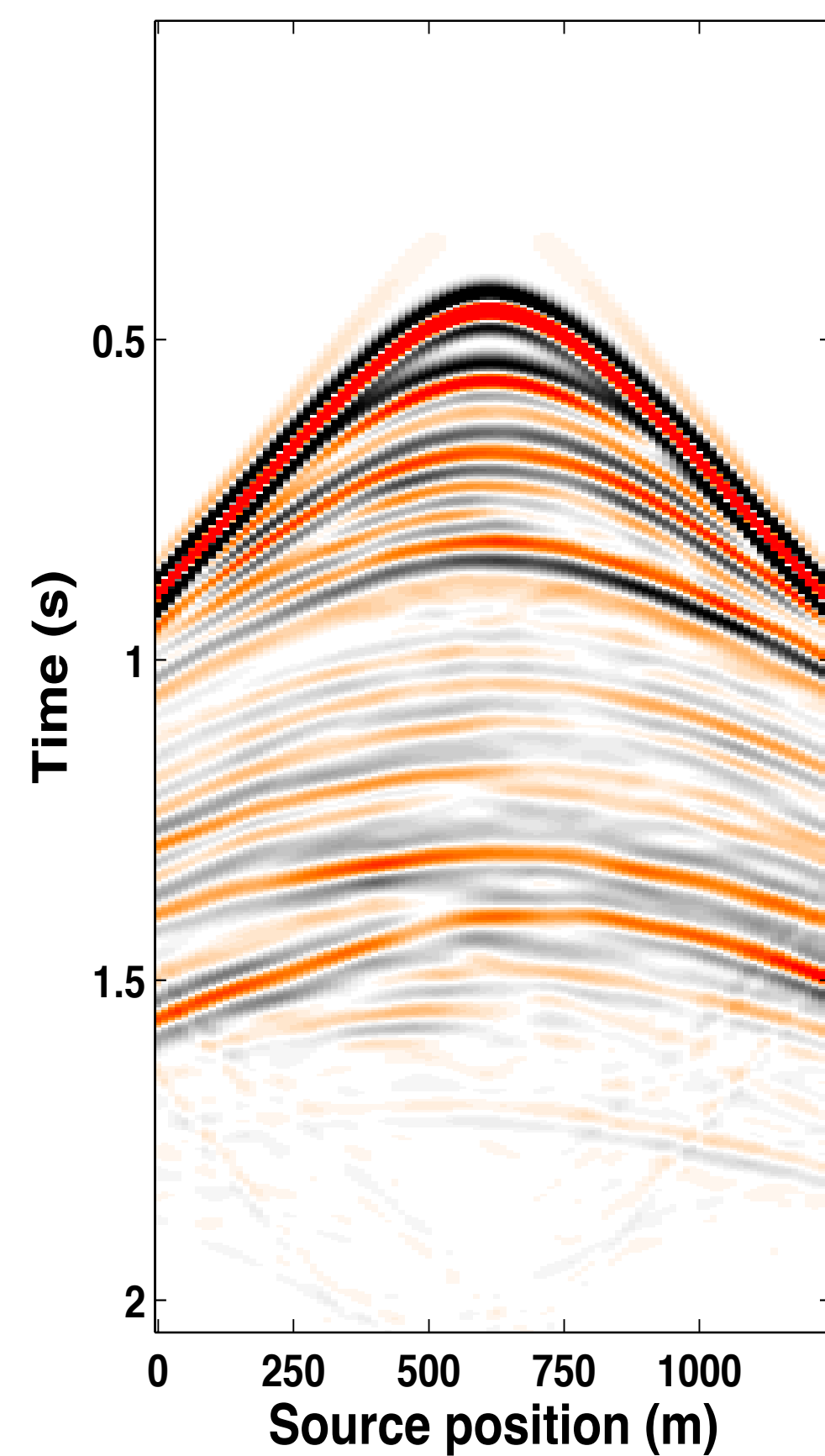
- ▶ *randomized acquisition* (e.g., *jittered* sampling, *time dithering*, *encoding*, etc.)
- ▶ *destroy sparsity*

Recover structure by promoting

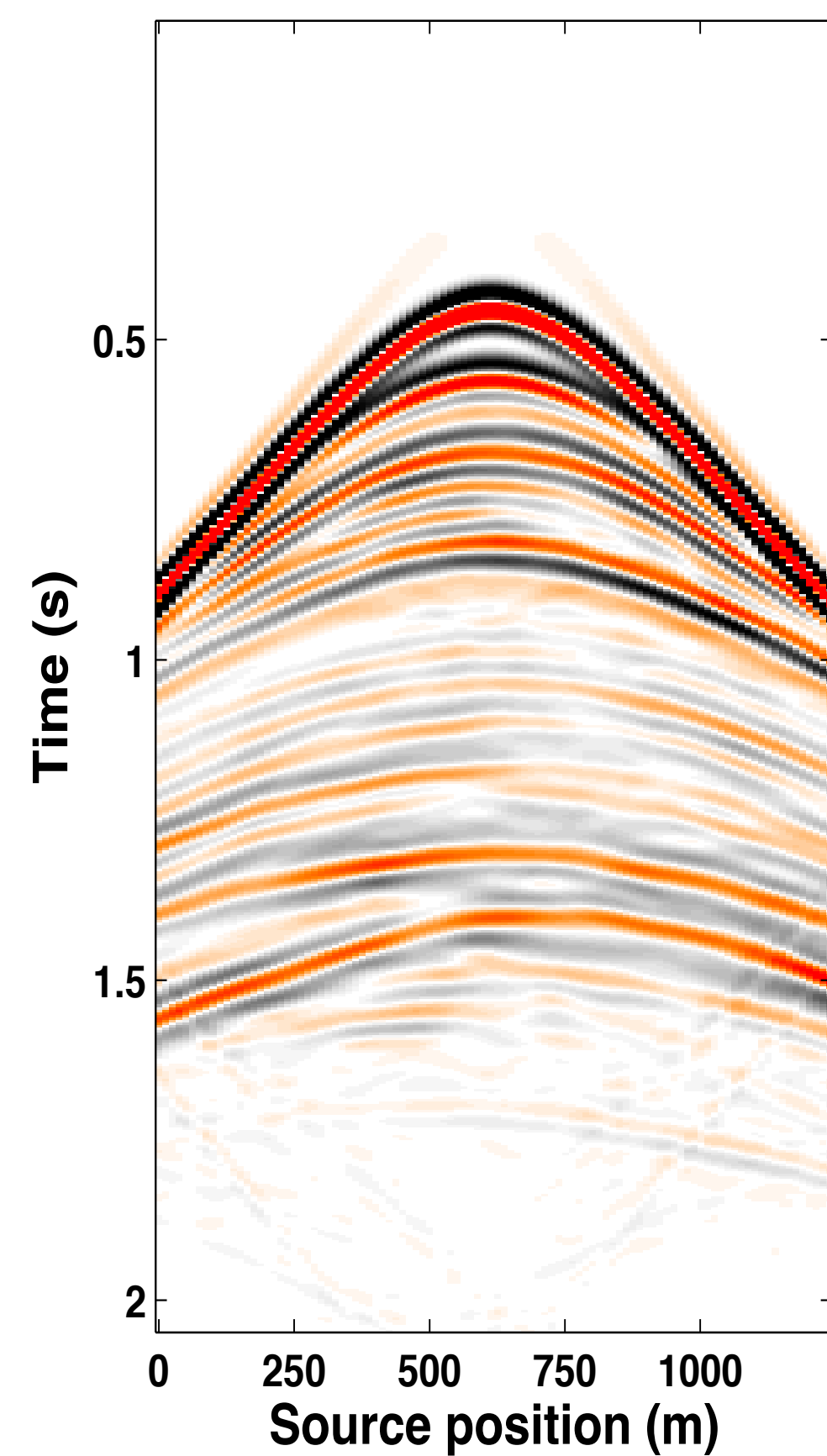
- ▶ *sparsity via one-norm minimization*

Time-lapse data

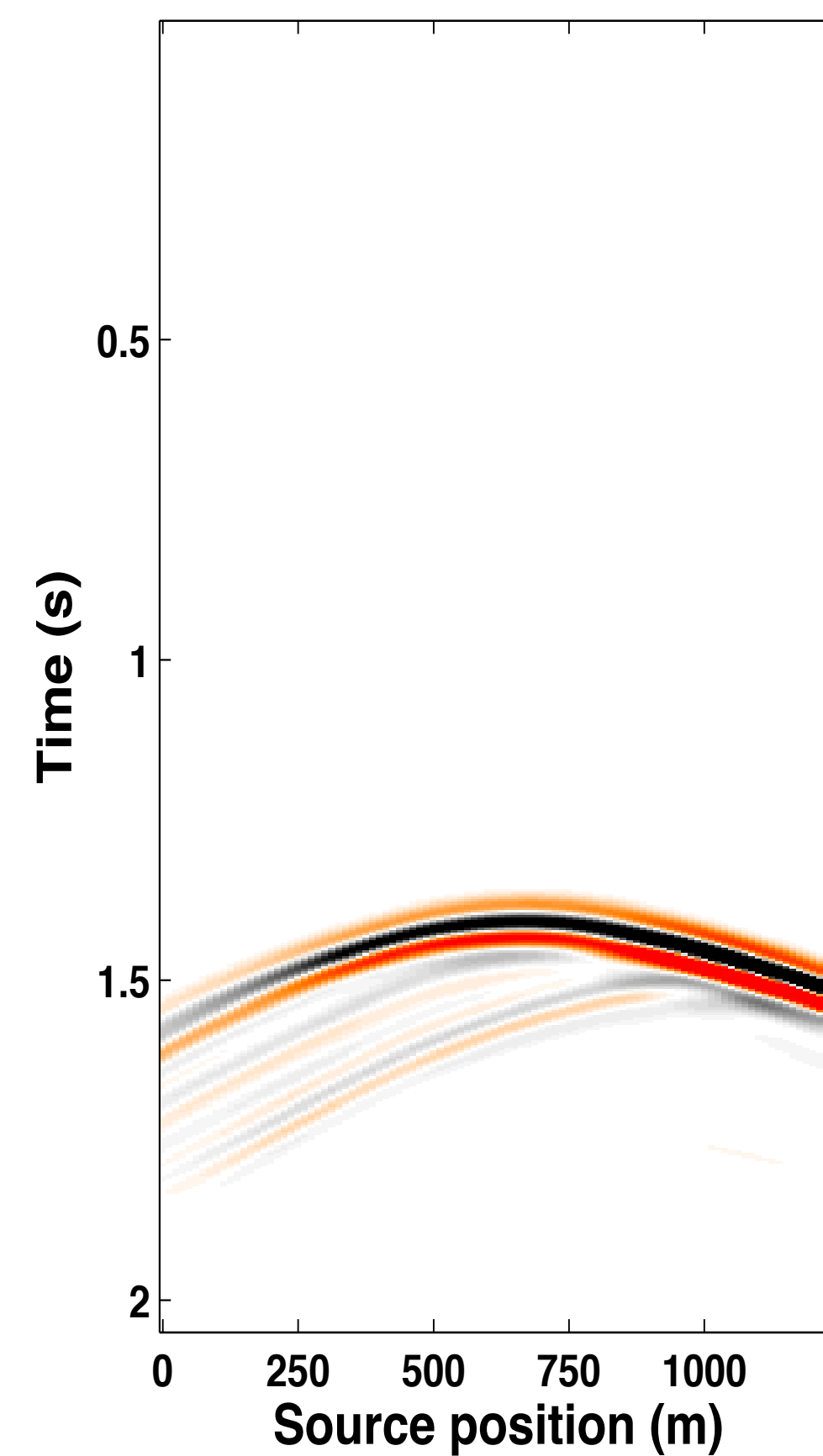
Baseline



Monitor



4-D signal [10 X]

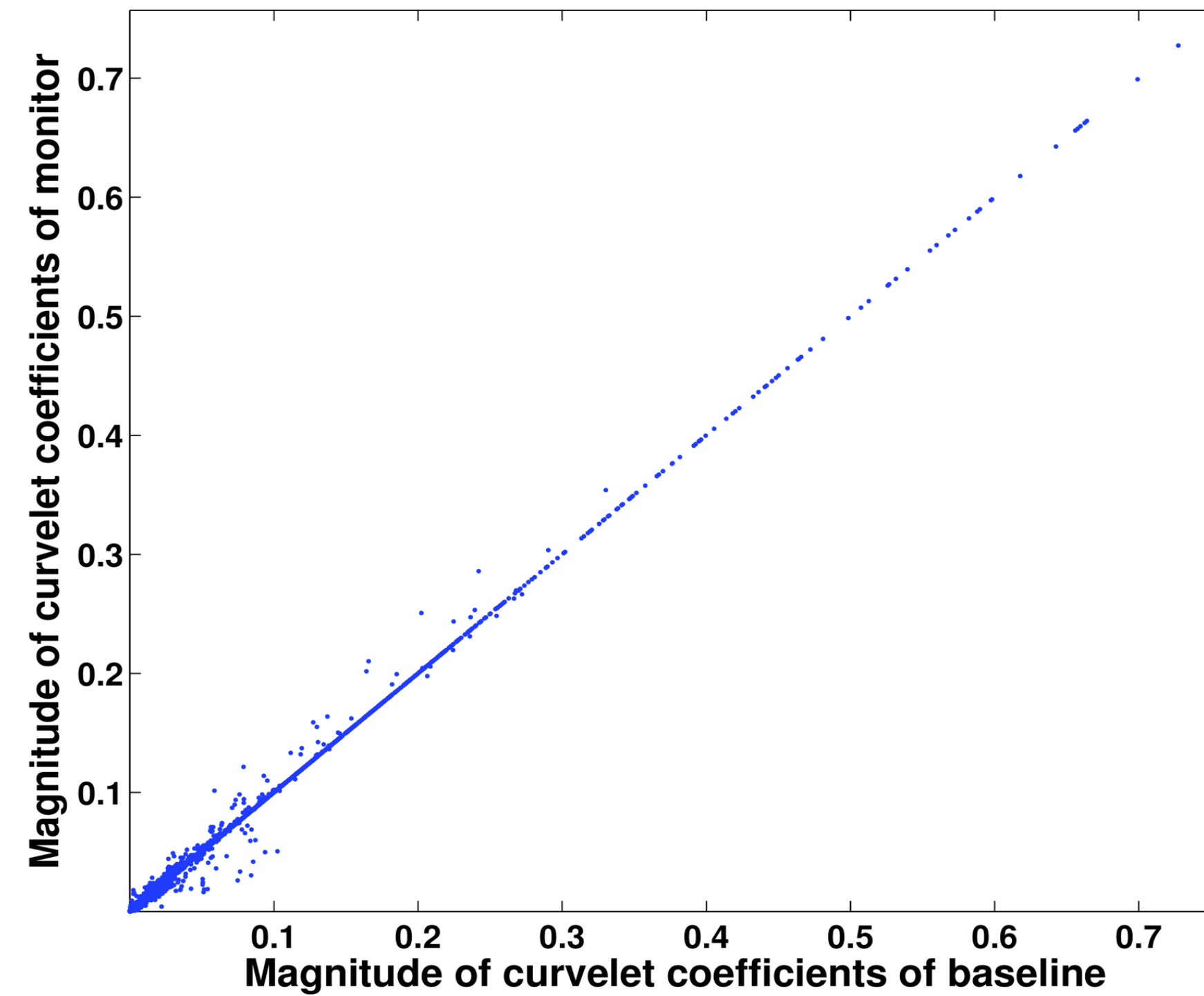
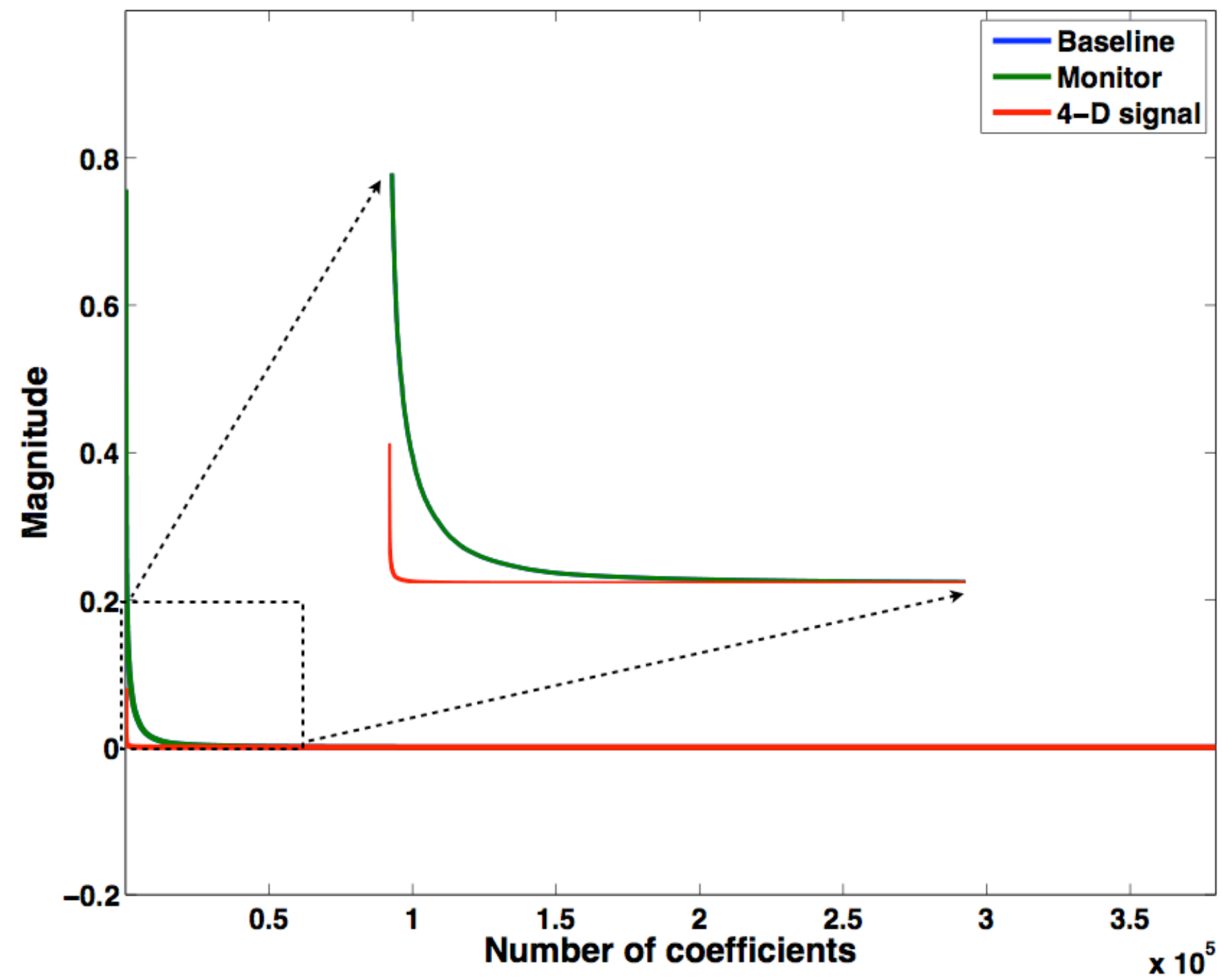


time samples: **512**
receivers: **100**
sources: **100**

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

Sparse structure via *curvelets*

significant correlation between the vintages



Distributed compressed sensing

– joint recovery model (JRM)

vintages

$$\begin{aligned} \mathbf{x}_1 &= \mathbf{z}_0 + \mathbf{z}_1 \\ \mathbf{x}_2 &= \mathbf{z}_0 + \mathbf{z}_2 \end{aligned} \rightarrow \text{differences}$$



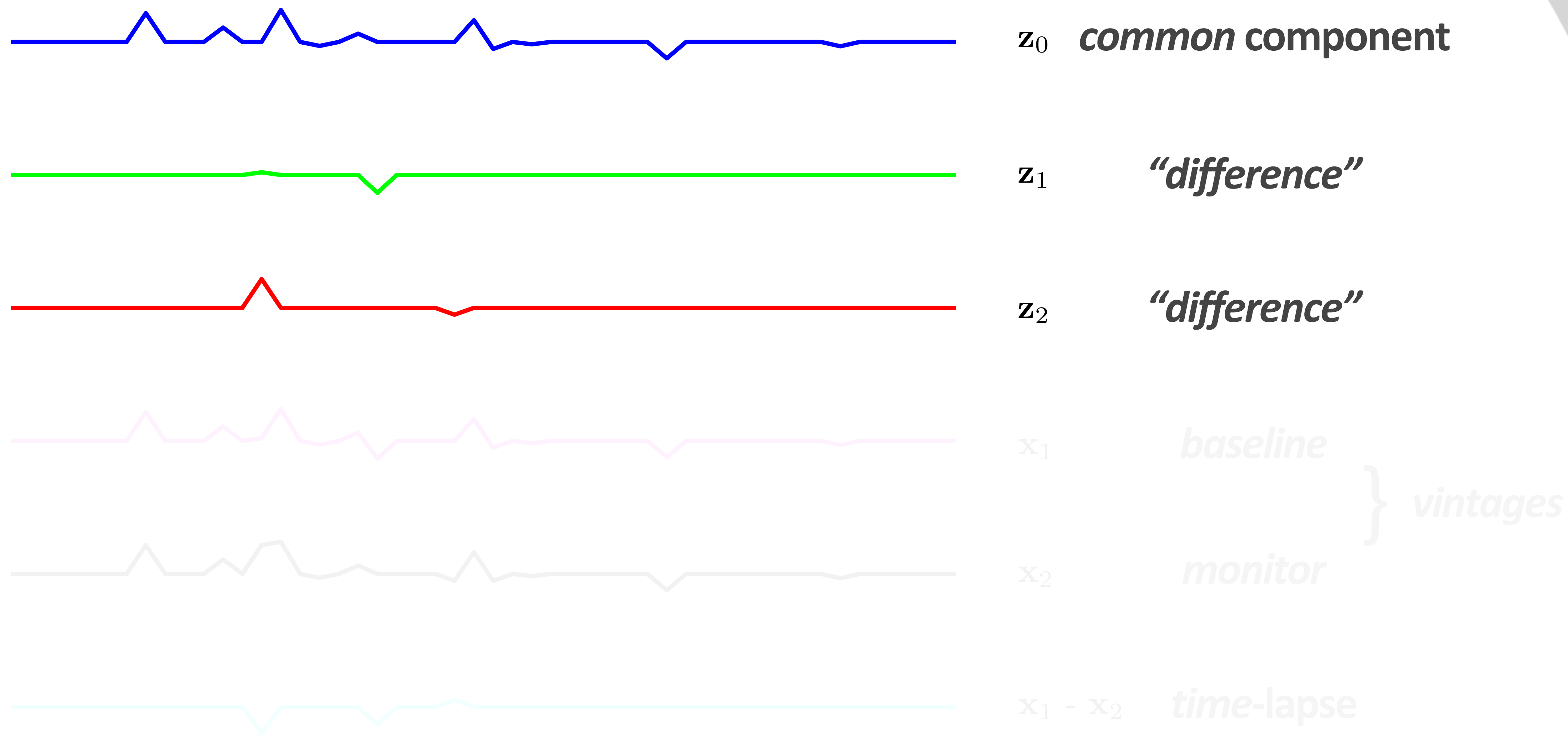
common component

$$\overbrace{\begin{bmatrix} \mathbf{A}_1 & \mathbf{A}_1 & \mathbf{0} \\ \mathbf{A}_2 & \mathbf{0} & \mathbf{A}_2 \end{bmatrix}}^{\mathbf{A}} \overbrace{\begin{bmatrix} \mathbf{z}_0 \\ \mathbf{z}_1 \\ \mathbf{z}_2 \end{bmatrix}}^{\mathbf{z}} = \overbrace{\begin{bmatrix} \mathbf{b}_1 \\ \mathbf{b}_2 \end{bmatrix}}^{\mathbf{b}} \rightarrow \begin{matrix} \text{baseline} \\ \text{monitor} \end{matrix}$$

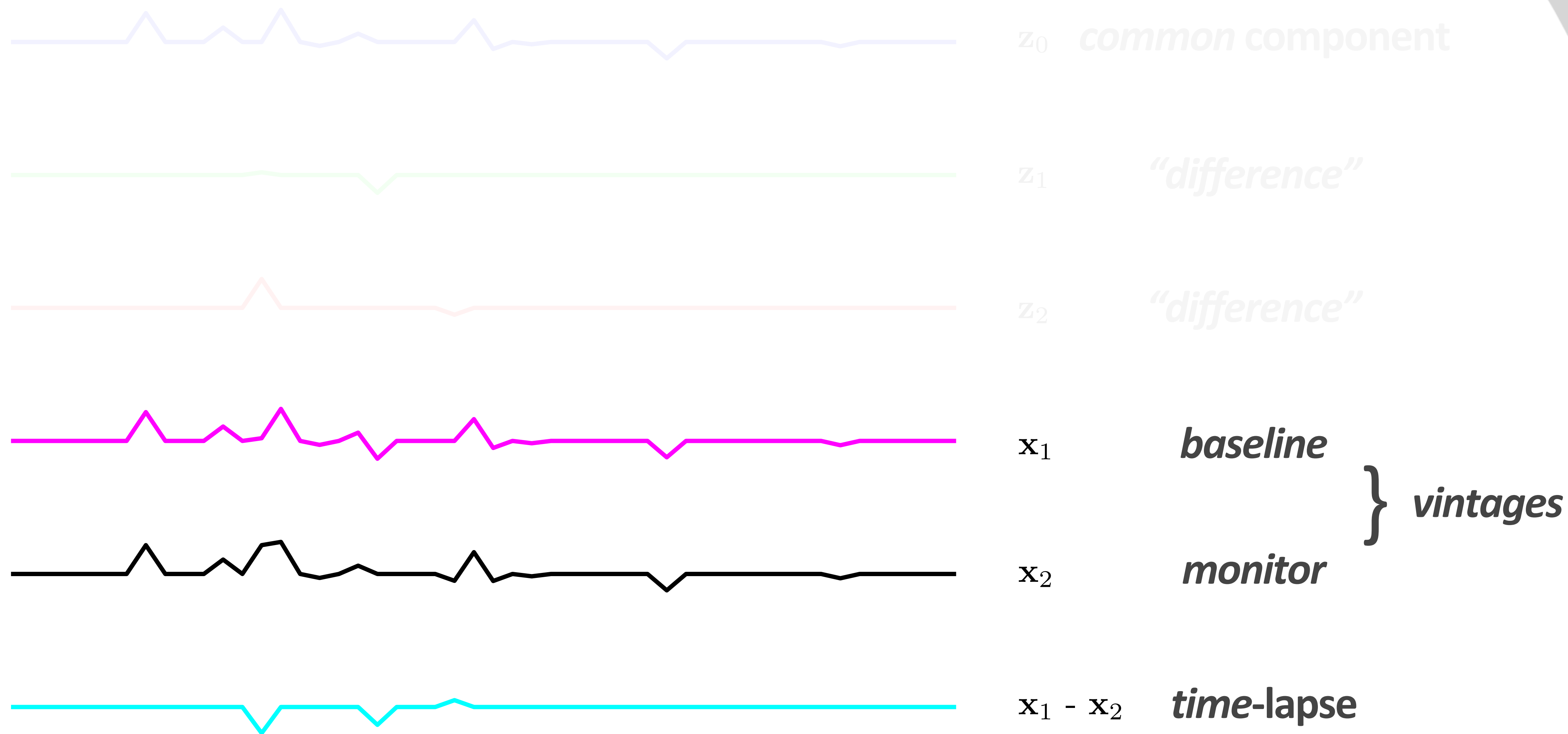
Key idea:

- ▶ use the fact that *different* vintages *share* common information
- ▶ invert for *common* components & *differences* w.r.t. the *common* components with *sparse* recovery

Sparse Joint Recovery Model (JRM)



Sparse baseline, monitor & time-lapse signals



Time-lapse

– w/ & w/o repetition

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

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

Question: To *repeat* or *not* to repeat?

Context

Acquire randomized subsamplings for the baseline and monitor surveys

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

Questions:

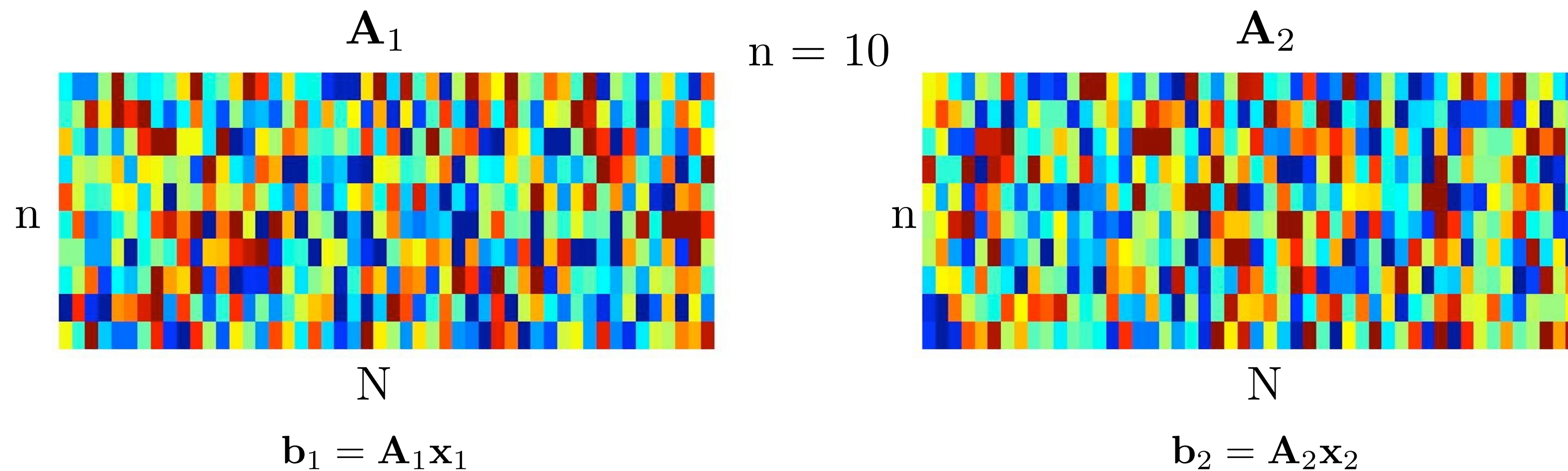
- ▶ Process/recover *independently* or *jointly* to exploit *common* features of *surveys*?
- ▶ Should we *repeat* the *surveys* when doing *randomized subsampling*?

Stylized experiments

Stylized experiments

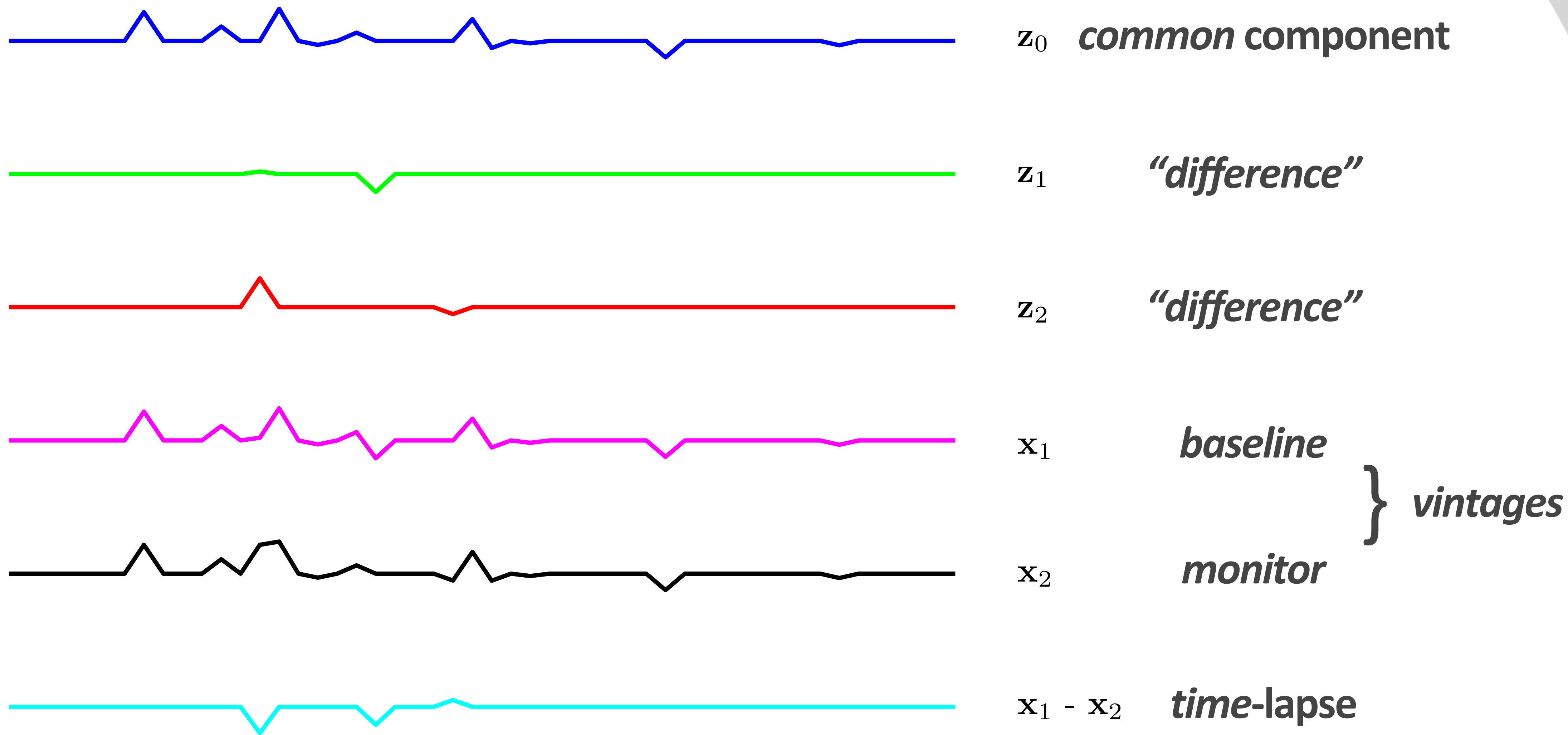
Conduct *many* CS experiments to compare

- ▶ *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

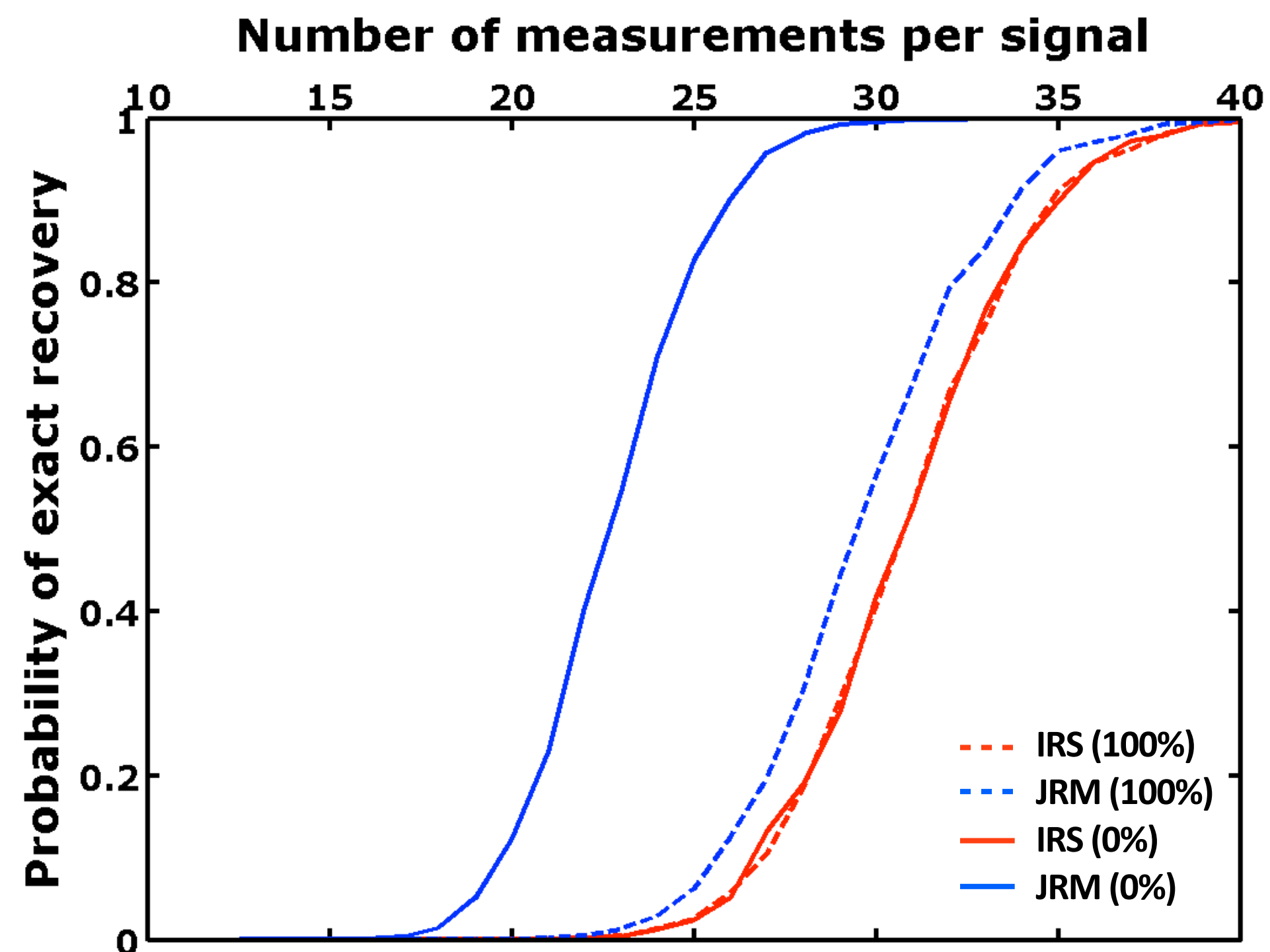
Sparse signals



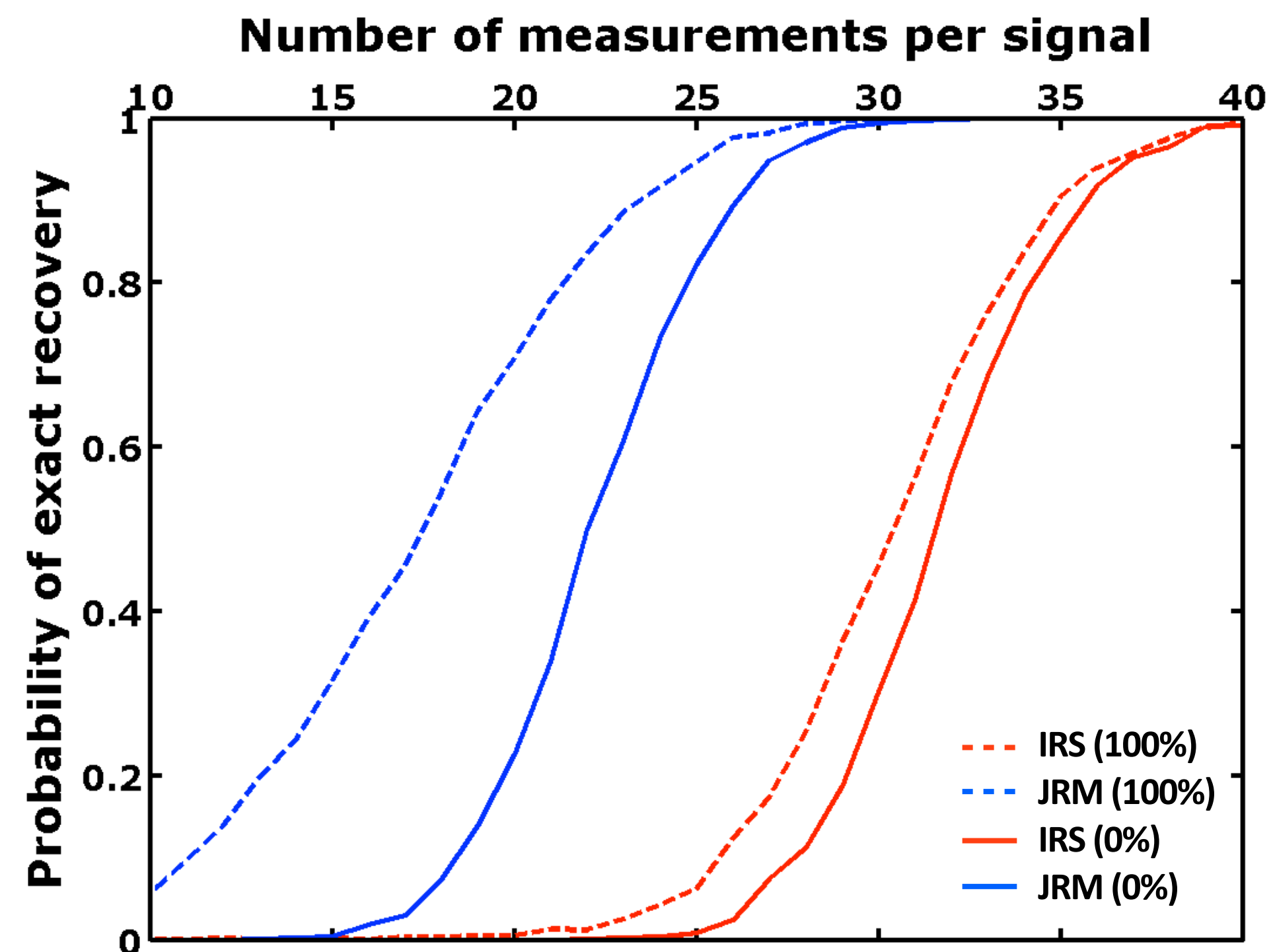
Independent vs. joint recovery

- 100% & 0% overlap in acquisition matrices

100% => "exact" repeatability
(difficult to achieve in practice)



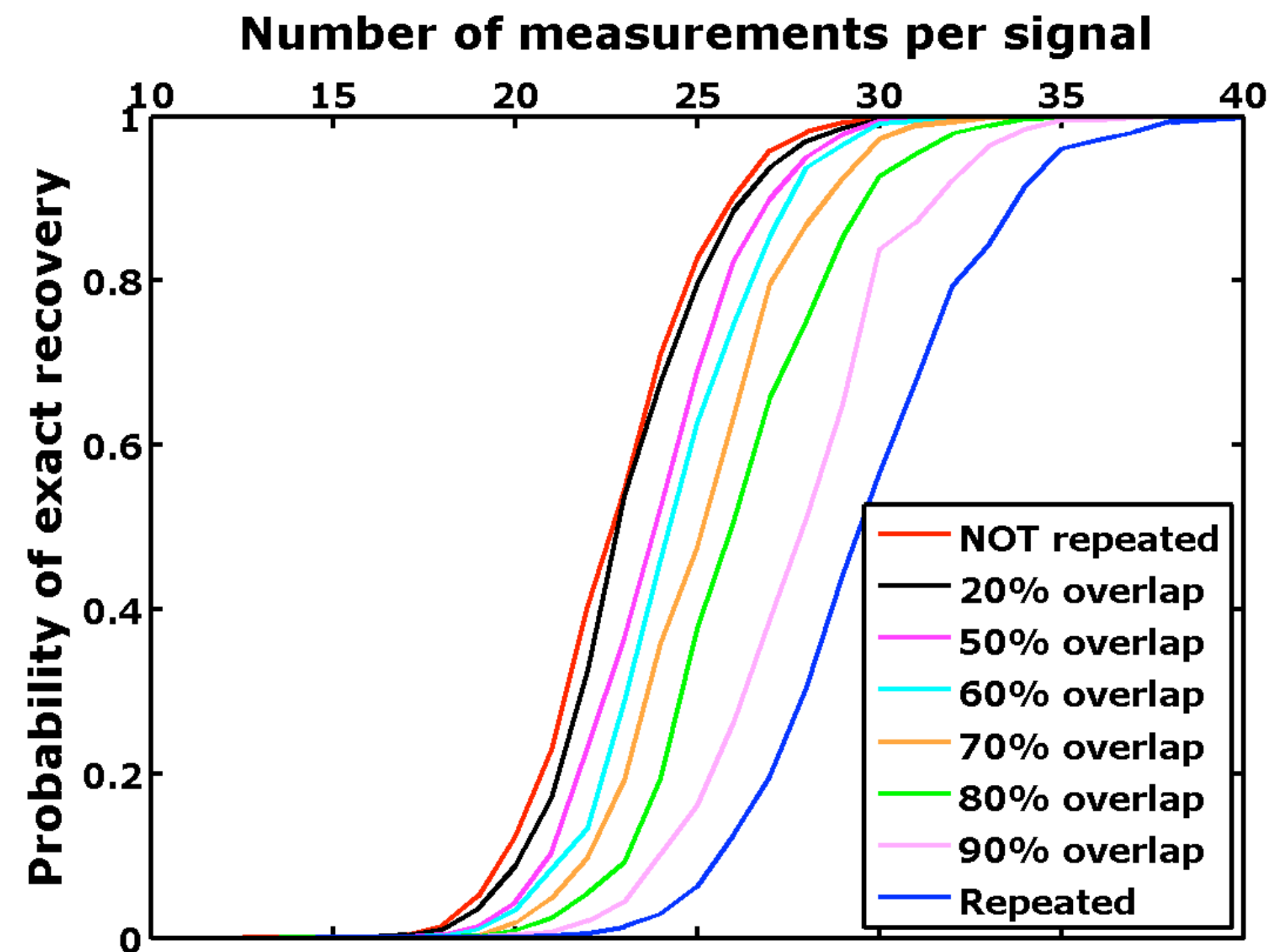
Vintages



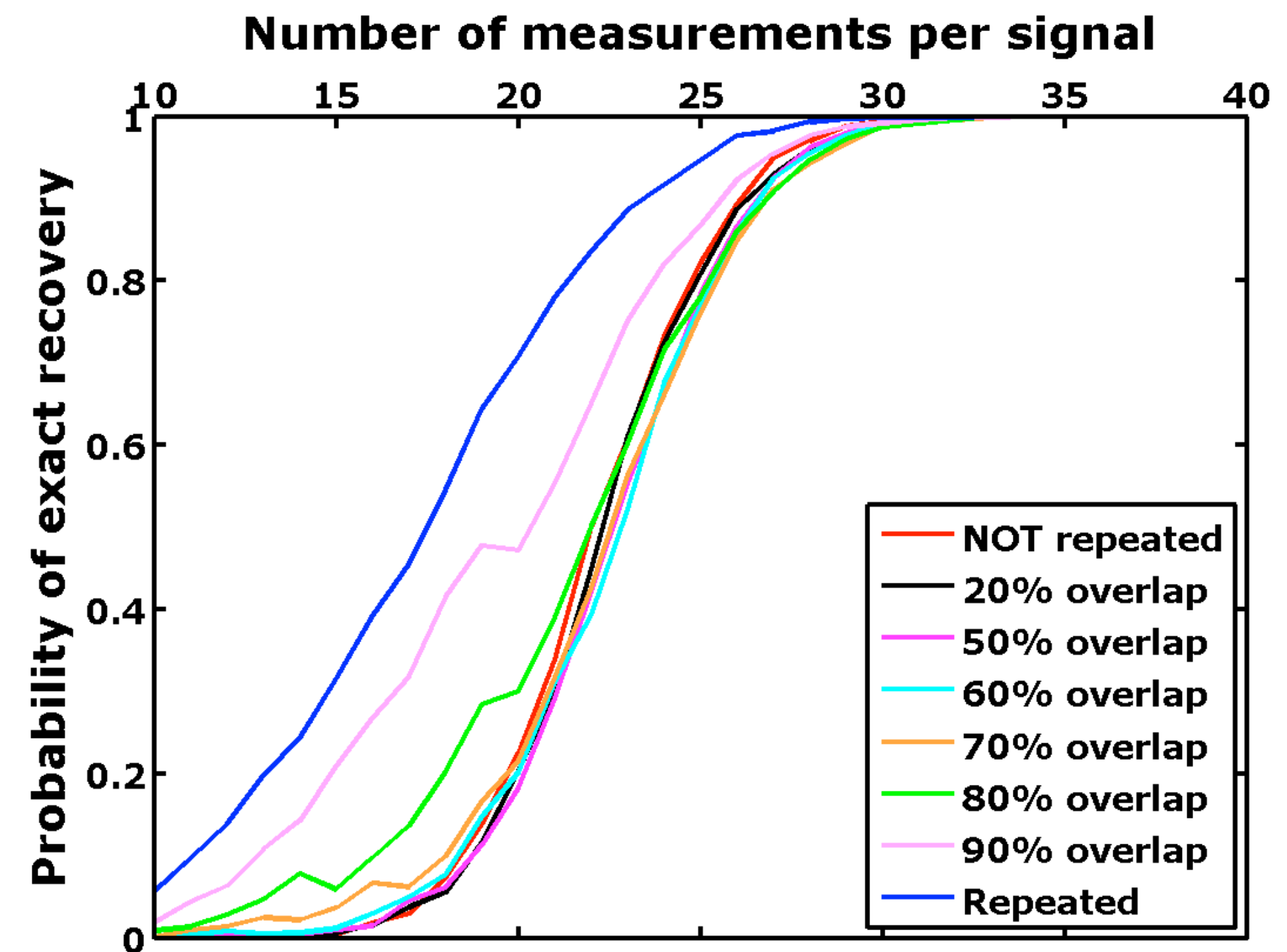
4-D signal

Joint recovery

- **varying %** of overlap in acquisition matrices



Vintages



4-D signal

Observations

Stylized synthetics give *fundamental* insights when recovering signals in 4-D seismic

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

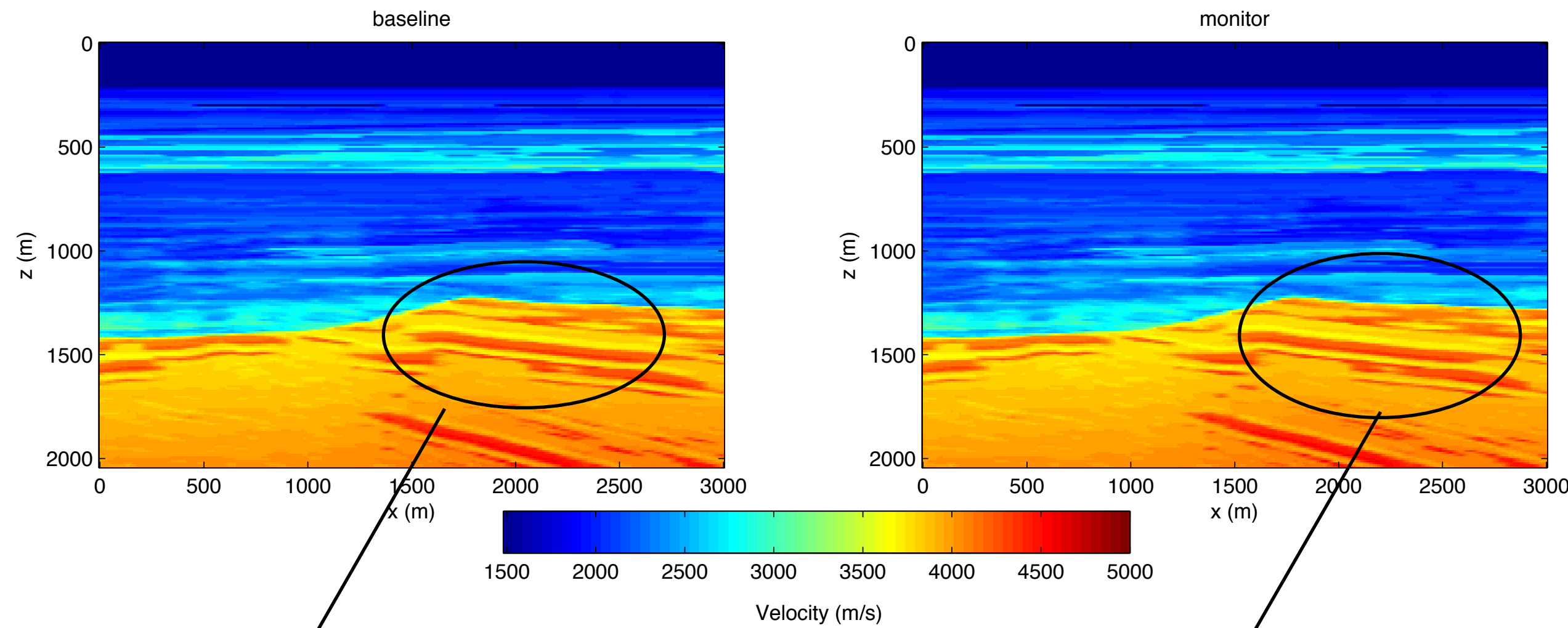
Time-jittered marine acquisition

Method

- ▶ Velocity and density model provided by BG Group, taken as baseline
- ▶ High permeability zone identified at a depth of $\sim 1300\text{m}$
- ▶ Fluid substitution (gas/oil replaced with brine) simulated to derive monitor velocity model
- ▶ Wavefield simulation to generate synthetic time-lapse data
- ▶ scales to 11733300 x 114882048

Baseline Model

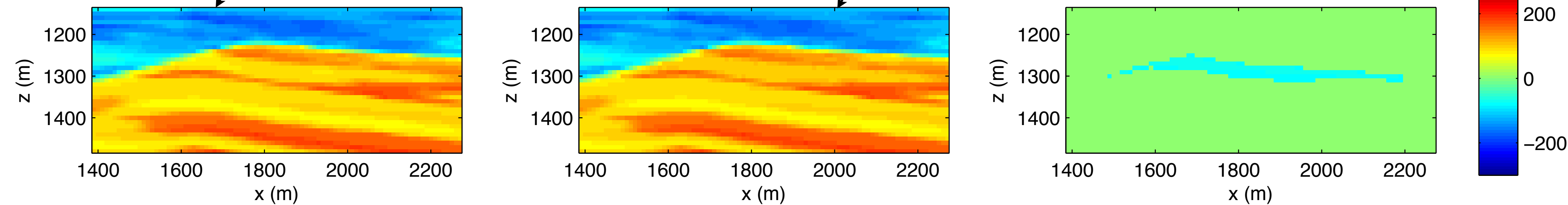
Monitor Model



baseline

monitor

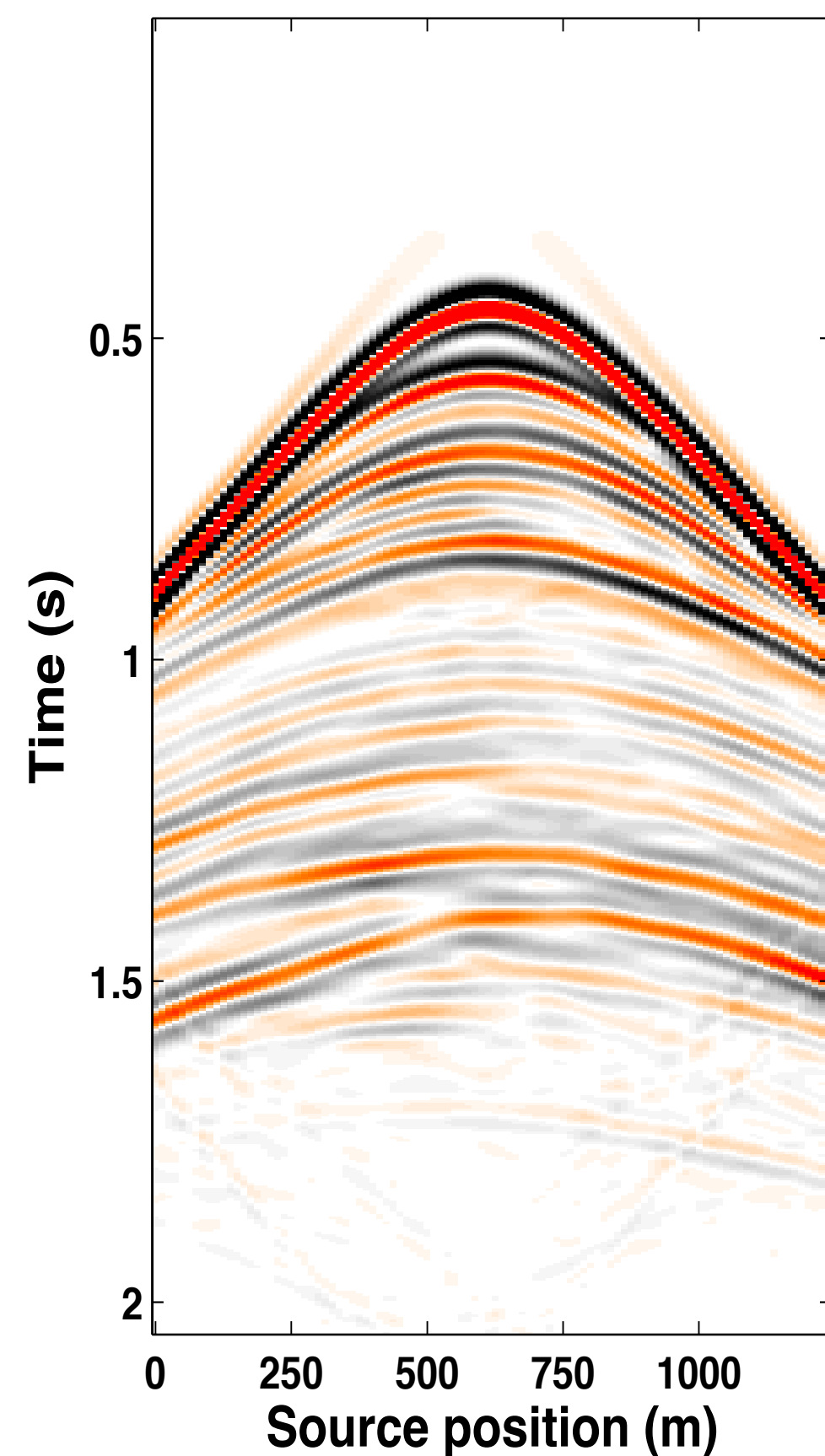
4D (difference)



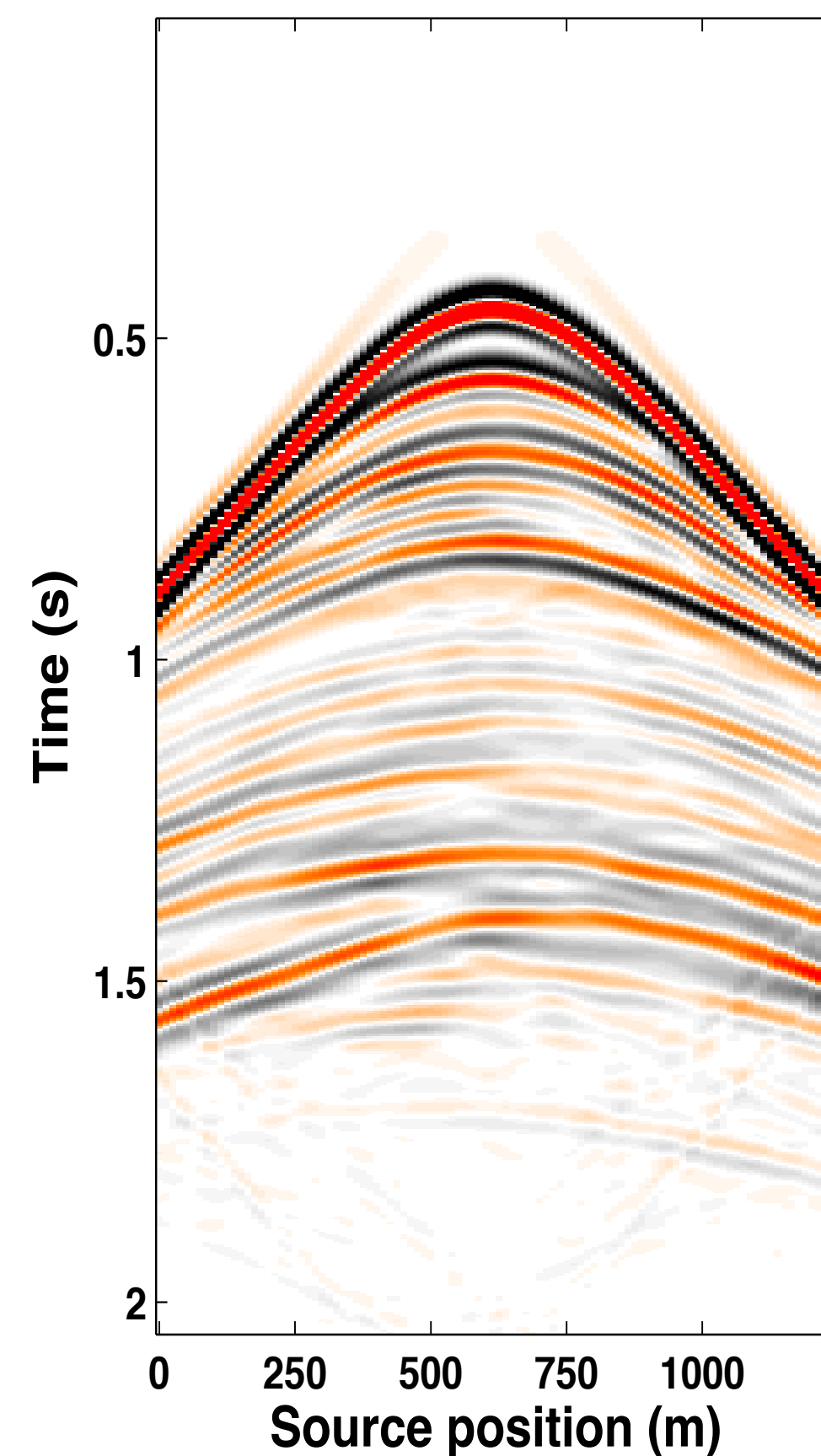
Simulated time-lapse data

– time-domain finite differences

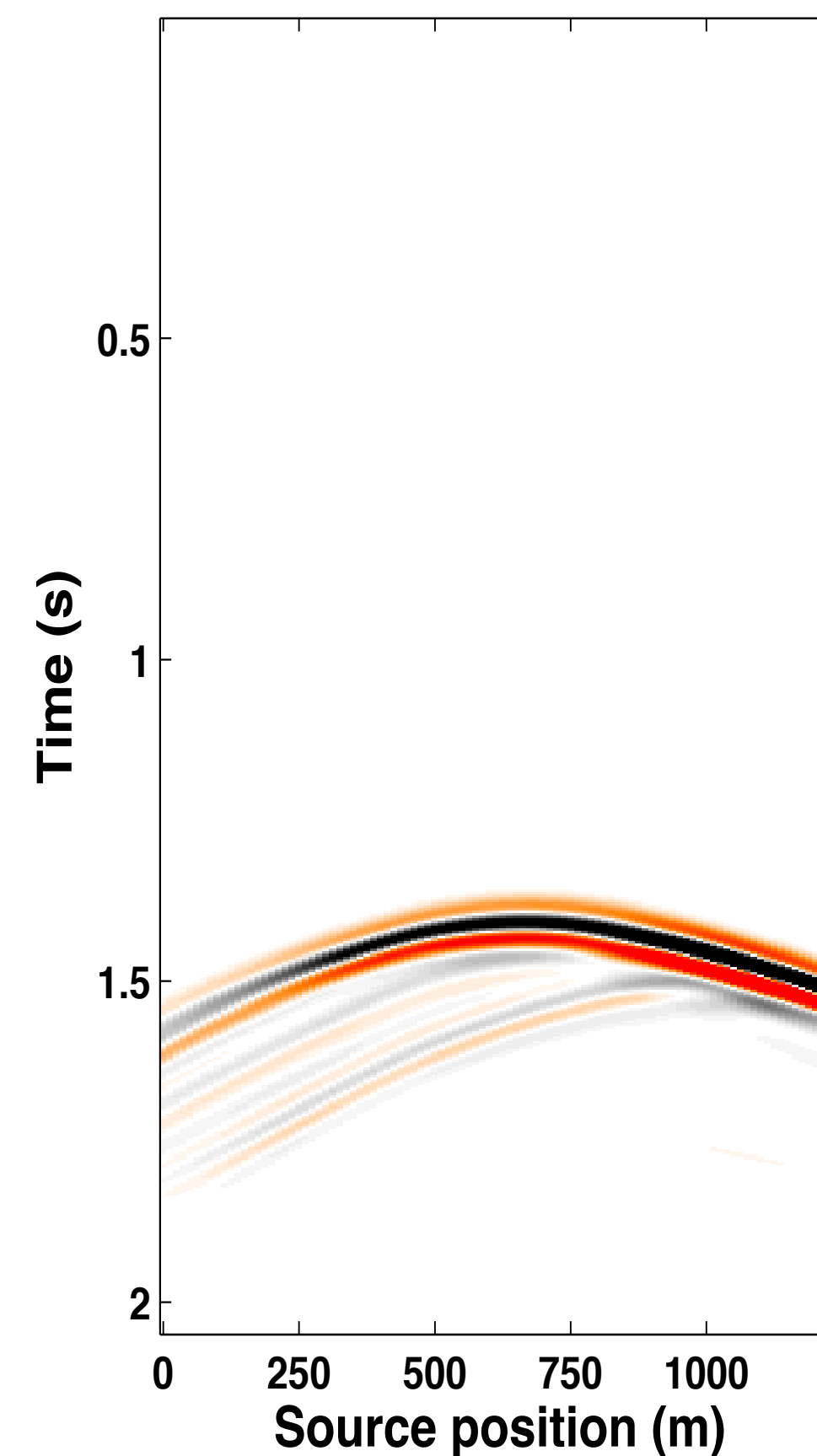
Baseline



Monitor



4-D signal

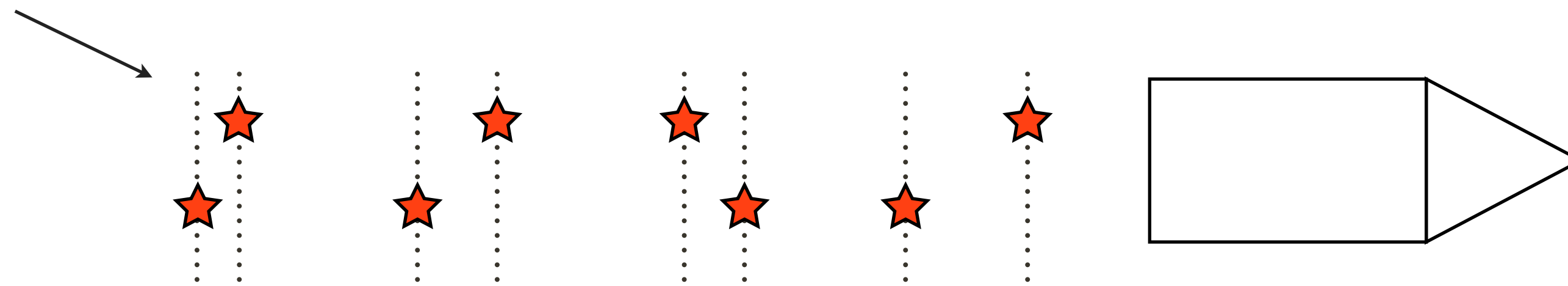


time samples: **512**
receivers: **100**
sources: **100**

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

Time-jittered marine acquisition

irregularly sampled spatial grid



continuous recording
START

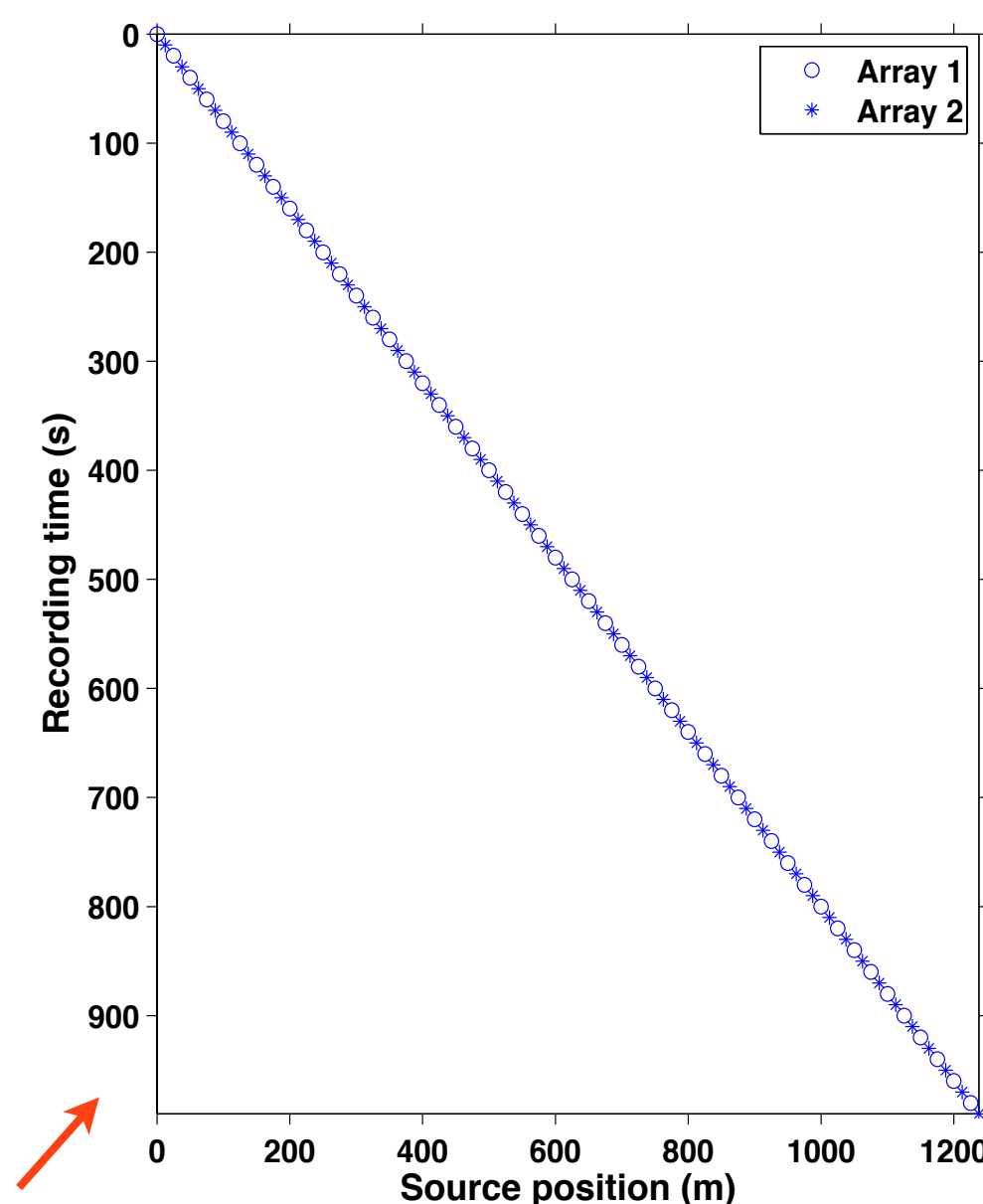
continuous recording
STOP



Conventional vs. *time-jittered* sources

– subsampling ratio = 2, 2 source arrays

conventional



“unblended” shot gathers

number of shots = **100** (per array)

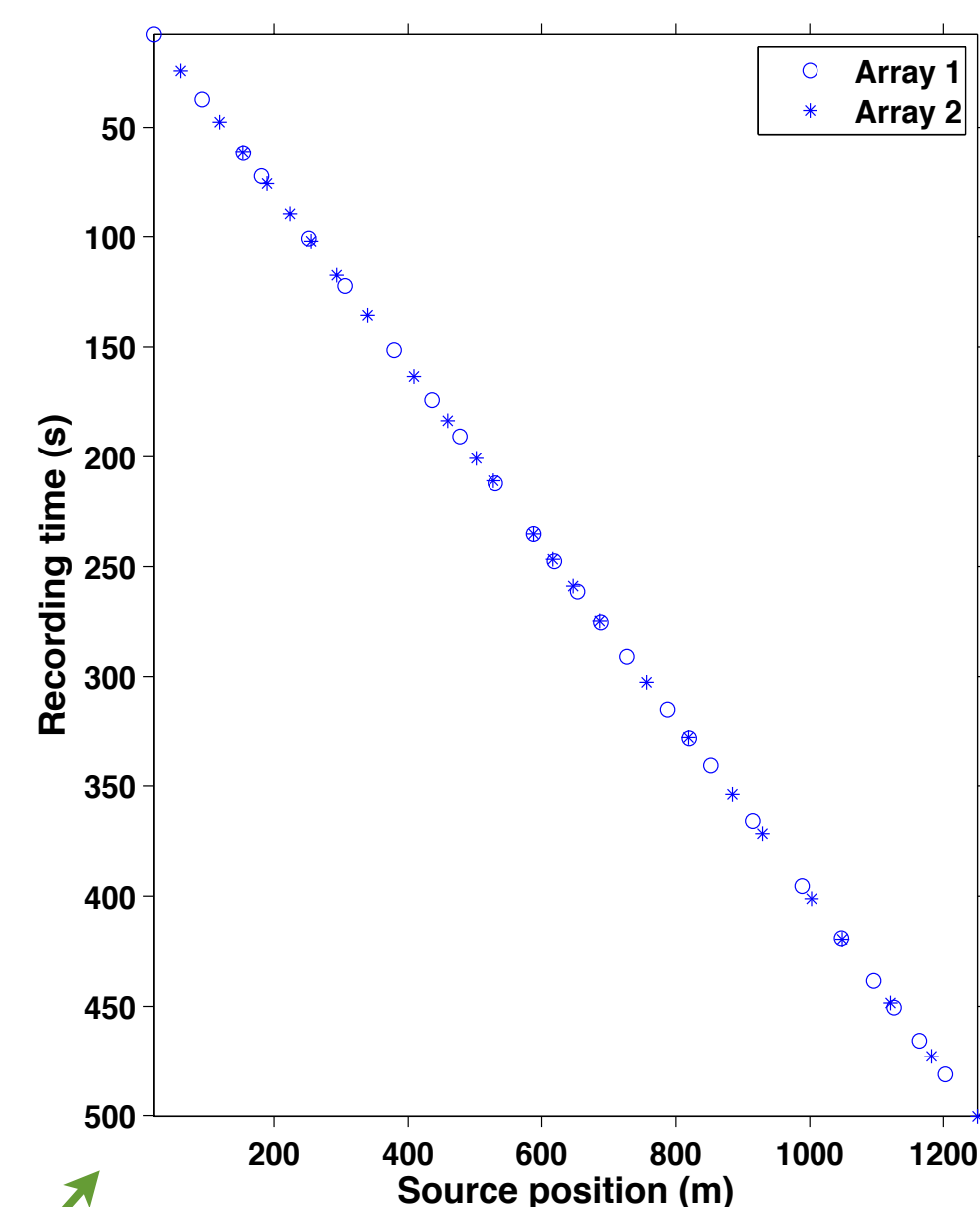
shot record length: 10.0 s

spatial sampling: **12.5 m**

vessel speed: **1.25 m/s**

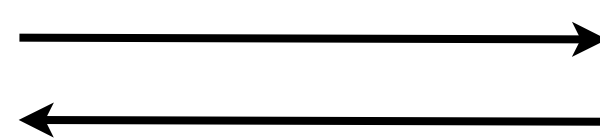
recording time = $100 \times 10.0 = \mathbf{1000.0\ s}$

jittered acquisition 1
(baseline)



[BLENDING & SUBSAMPLING]

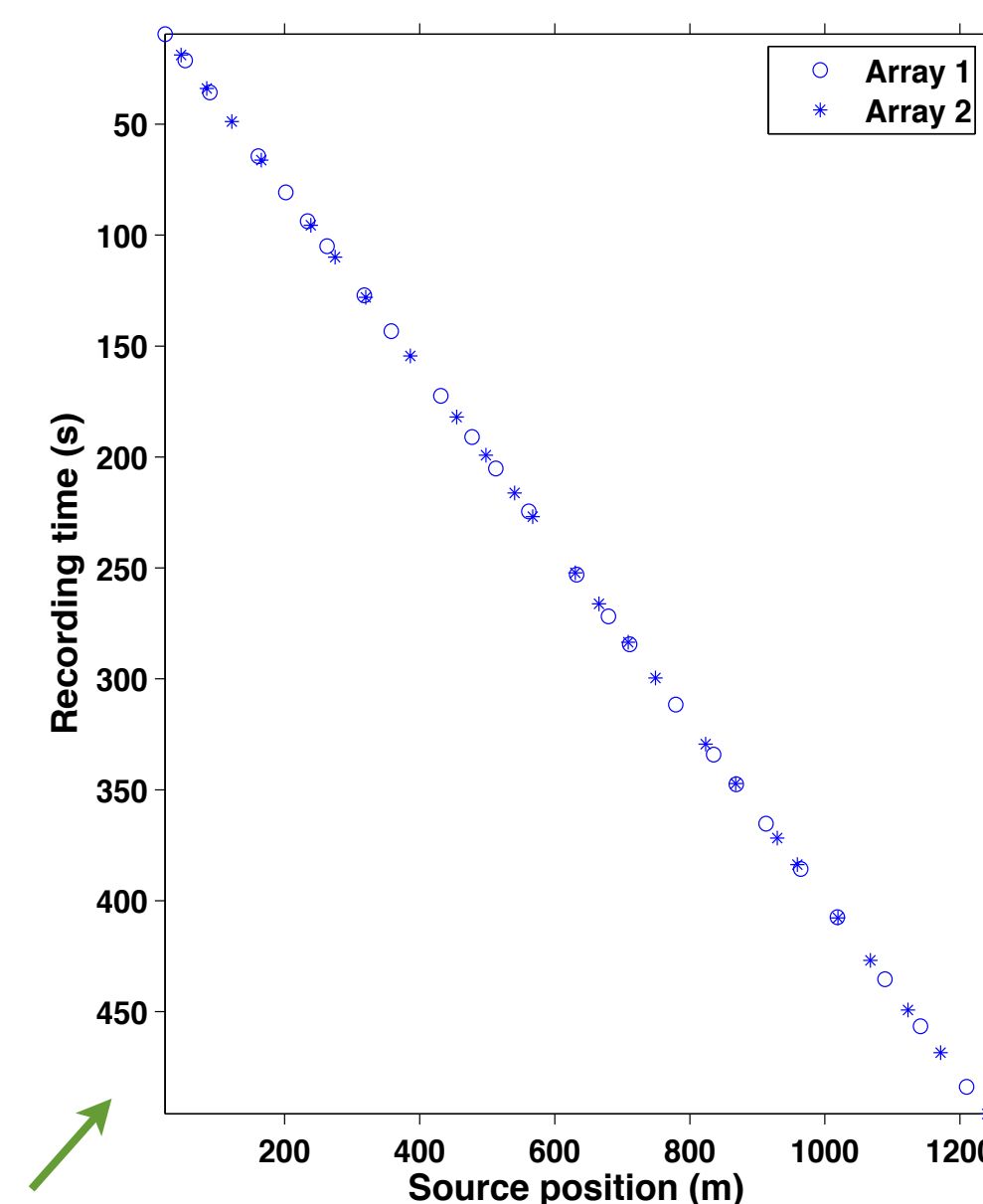
spatial subsampling factor = 2



spatial sampling **increase** factor = 2

[DEBLENDING & INTERPOLATION]

jittered acquisition 2
(monitor)



“blended” shot gathers

number of shots = $100/2 = \mathbf{50}$ (25 per array)

spatial sampling: **50.0 m (jittered)**

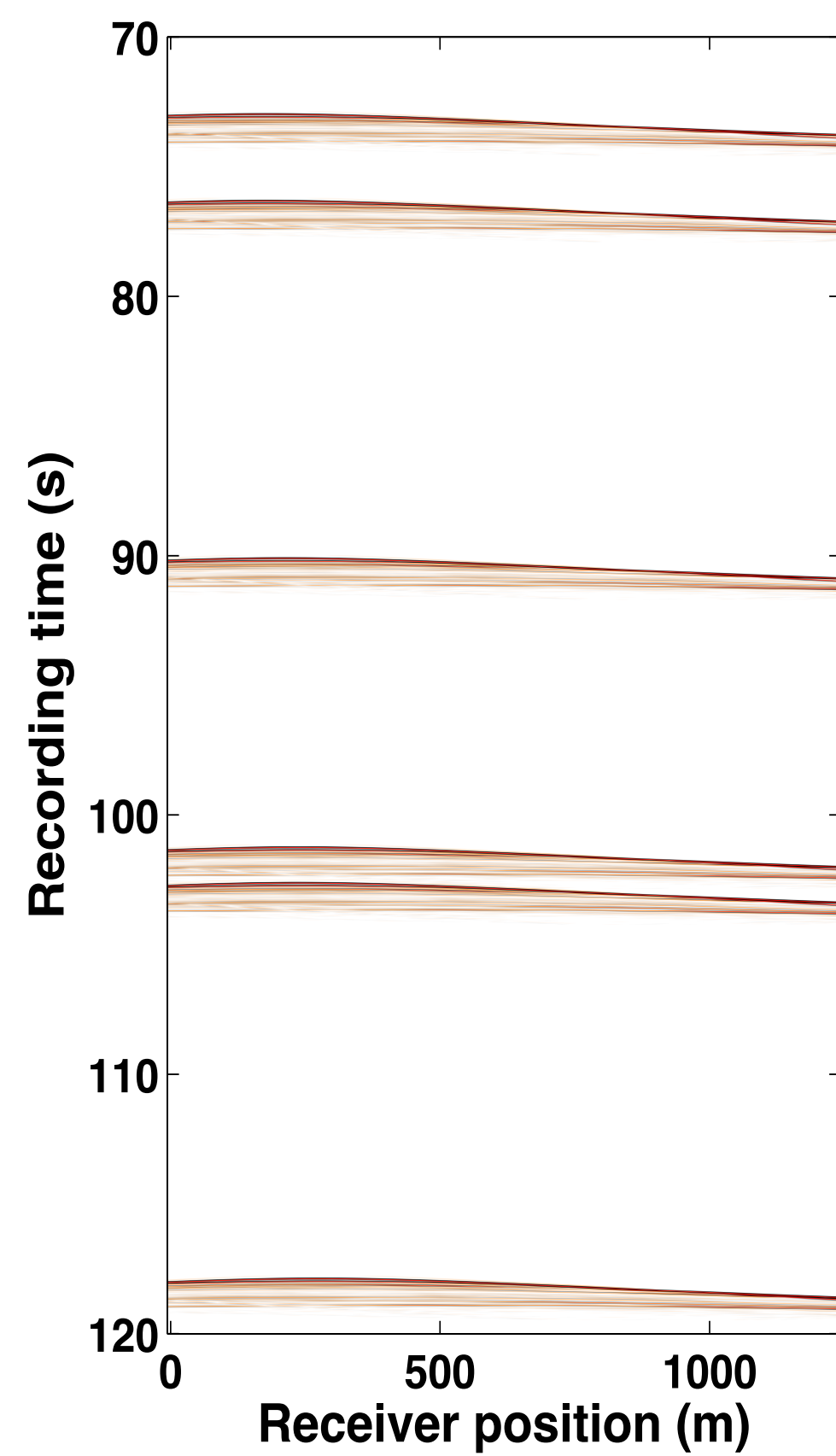
vessel speed: **2.50 m/s**

recording time $\approx 1000.0\ s/2 = \mathbf{500.0\ s}$

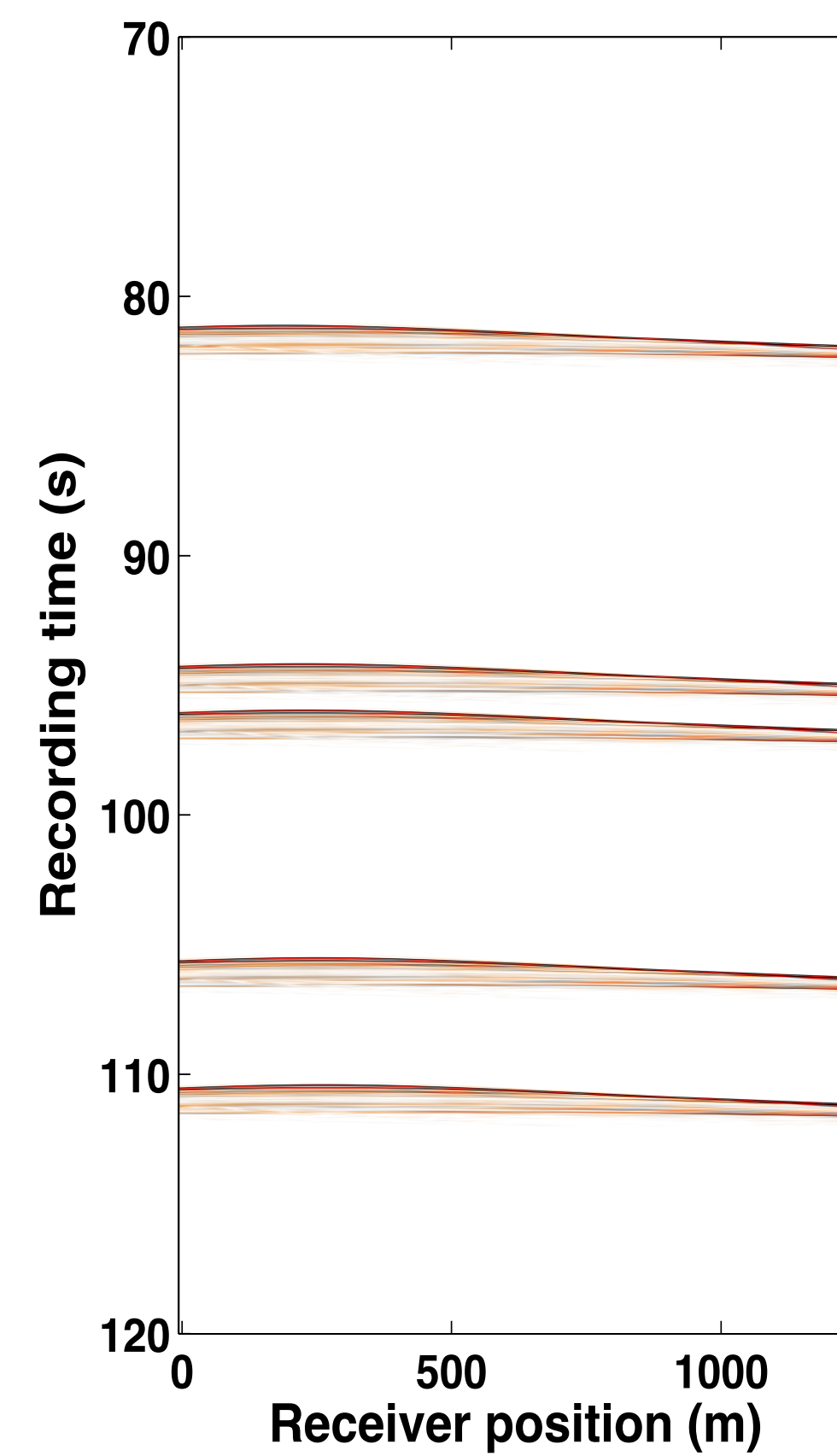
Measurements

– *subsampled and blended*

Baseline



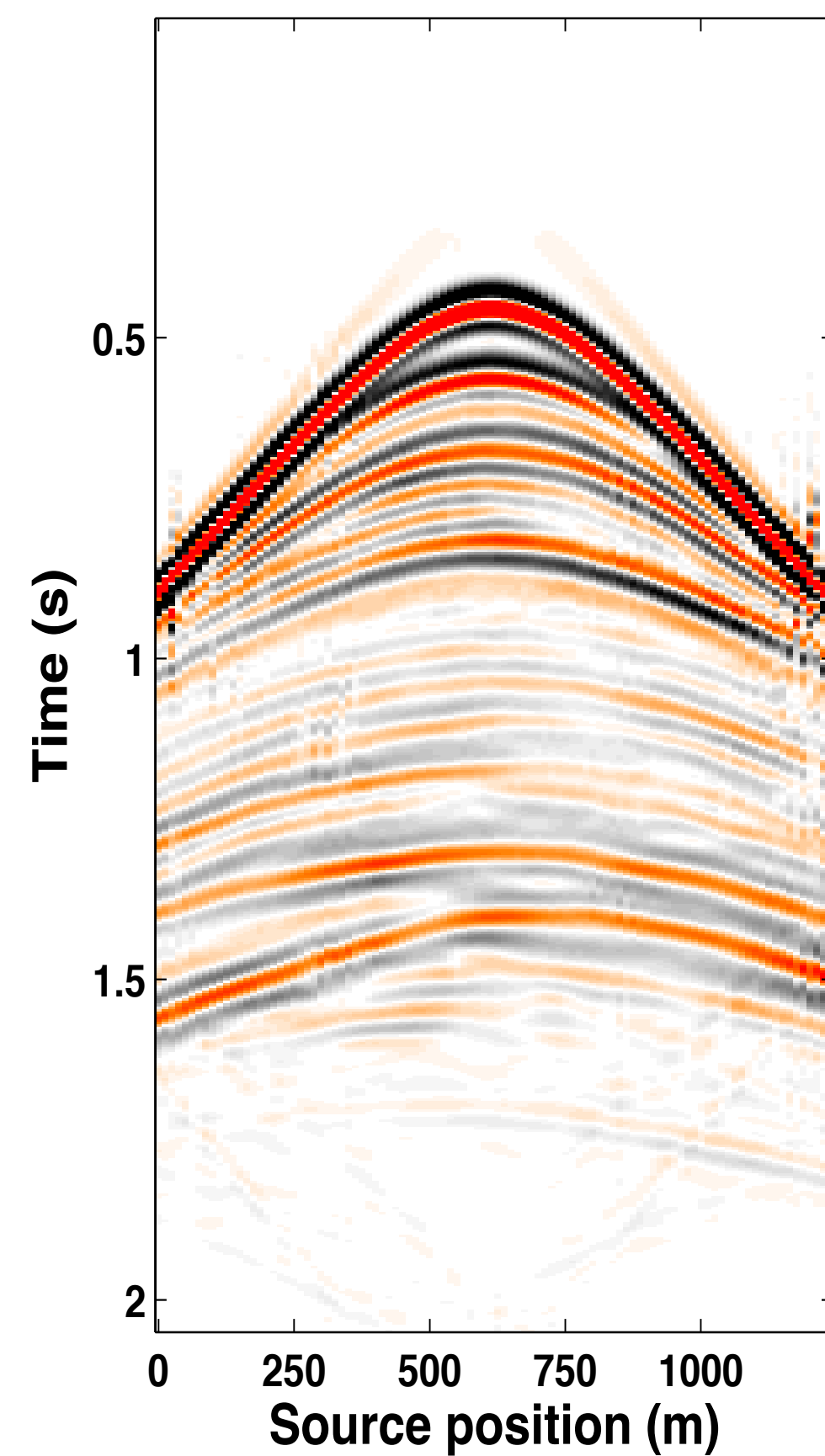
Monitor



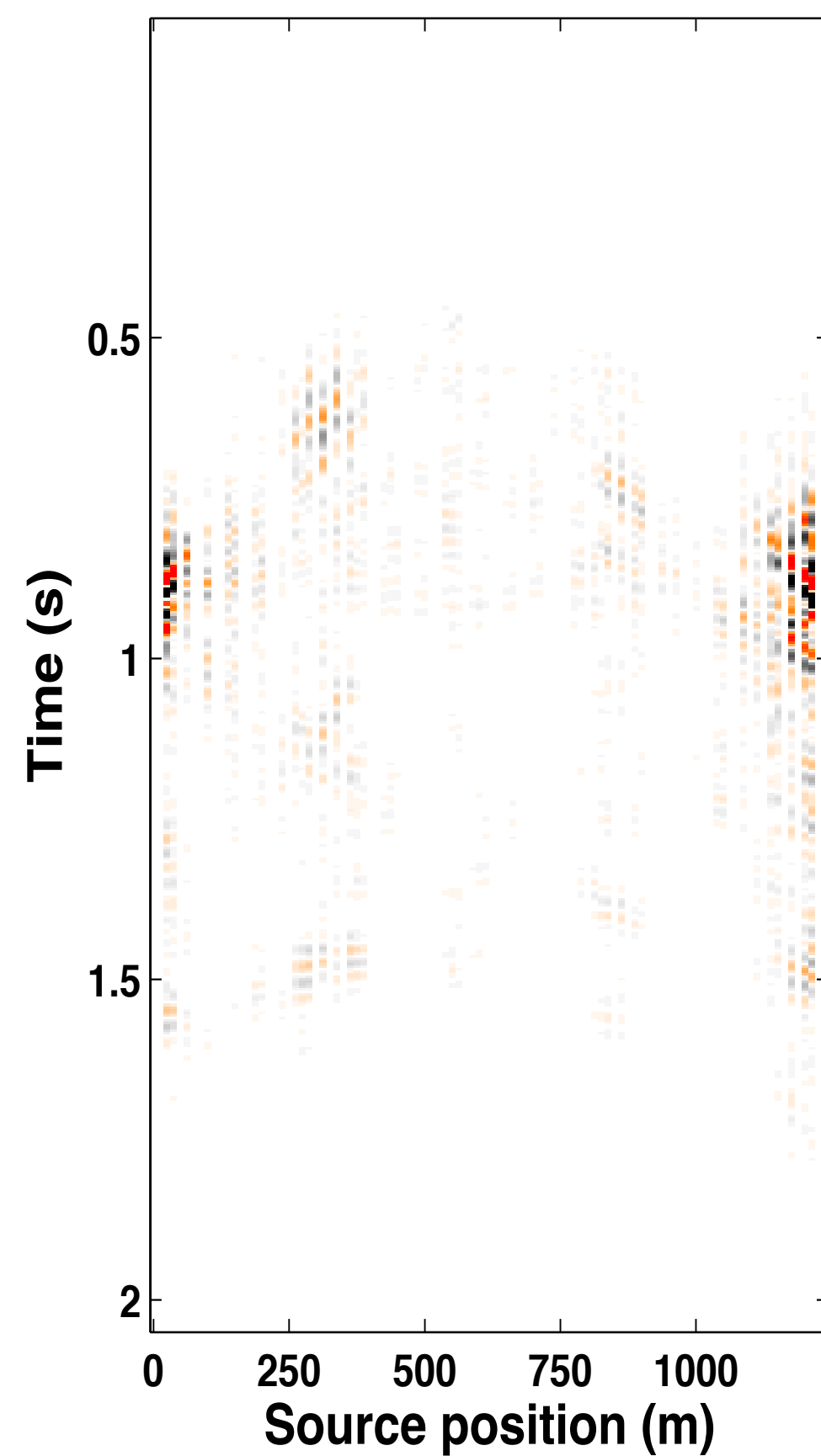
Monitor recovery

- **100%** overlap in acquisition matrices

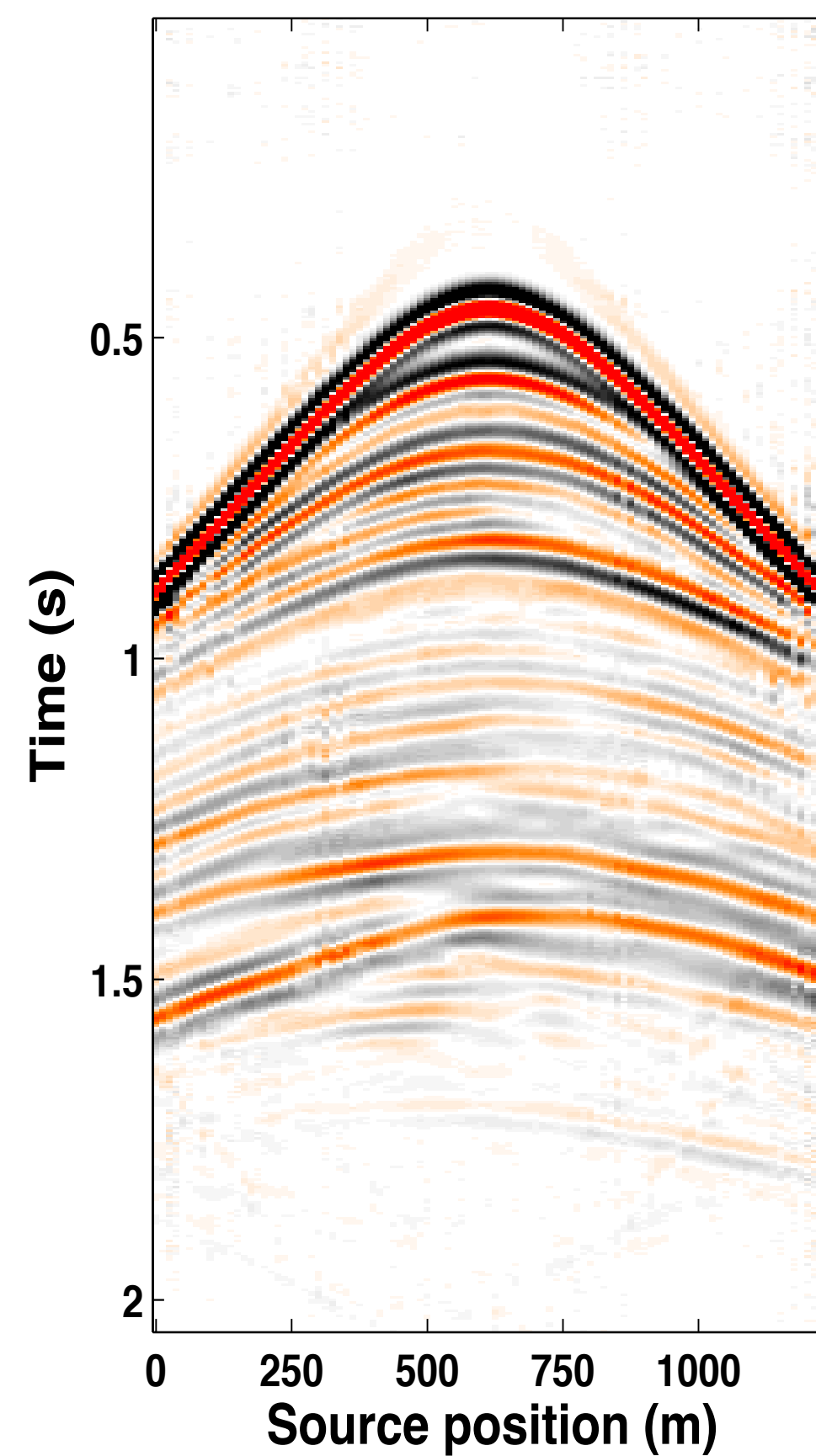
IRS
[11.6 dB]



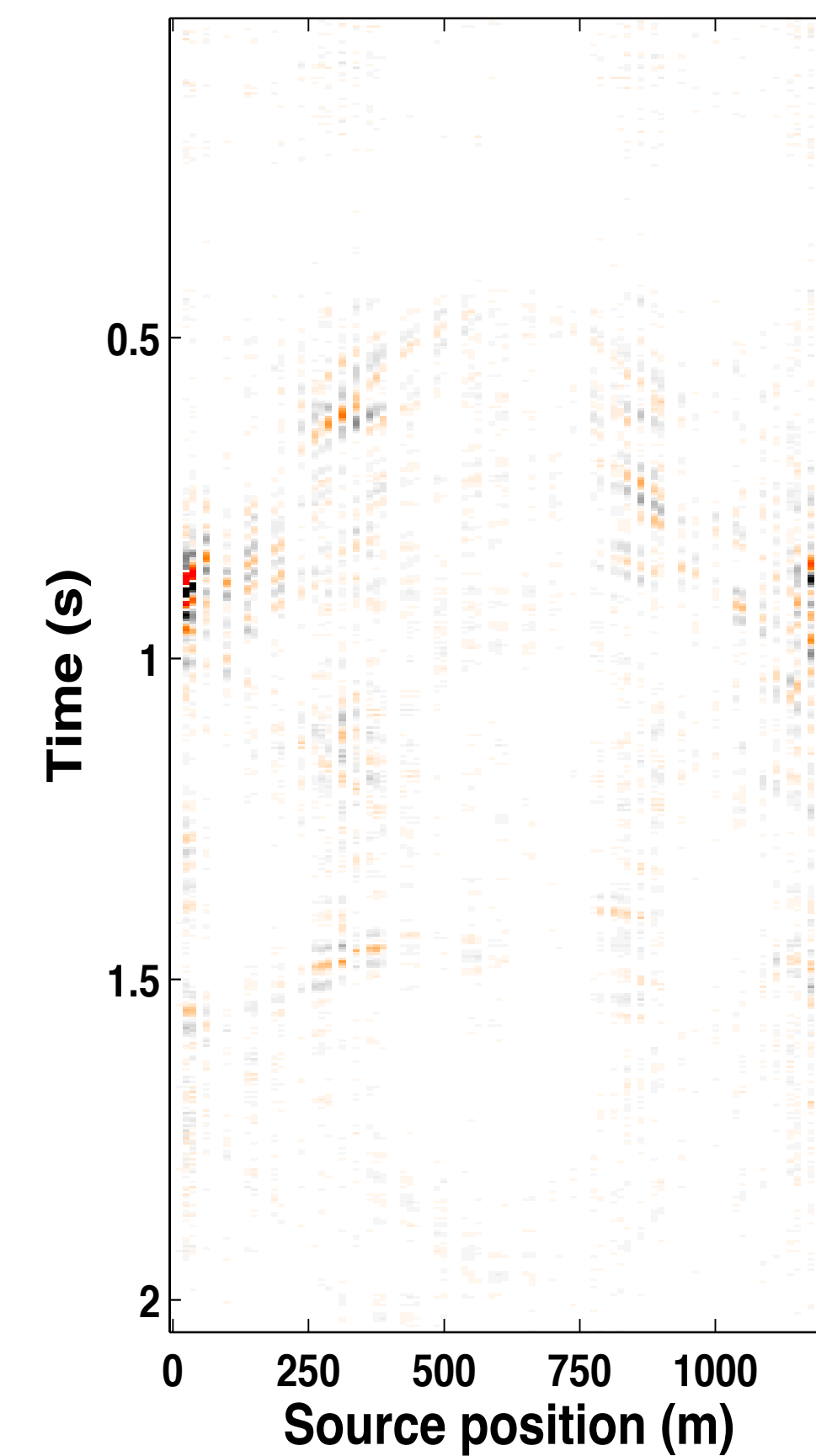
IRS
residual



JRM
[12.1 dB]



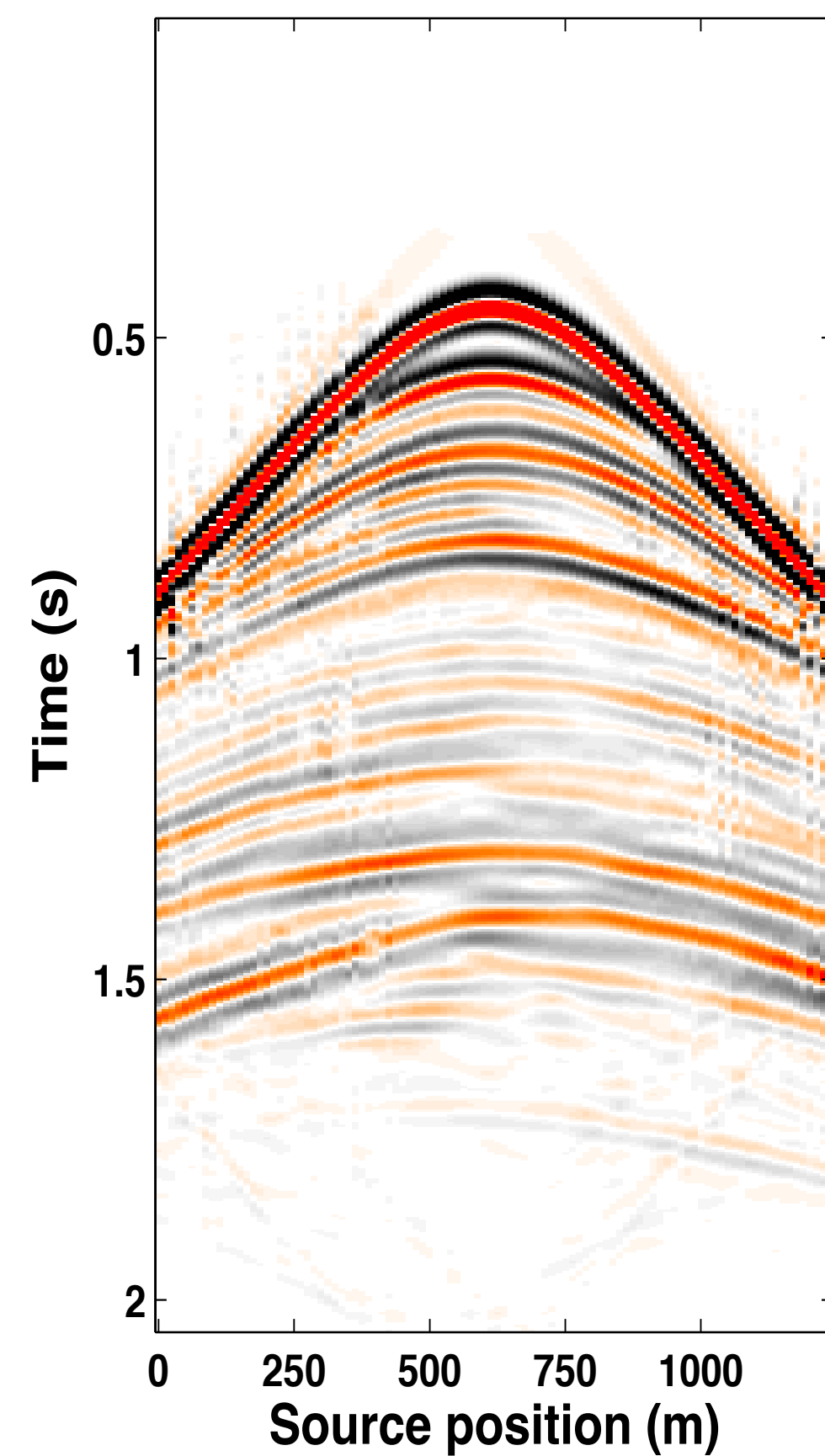
JRM
residual



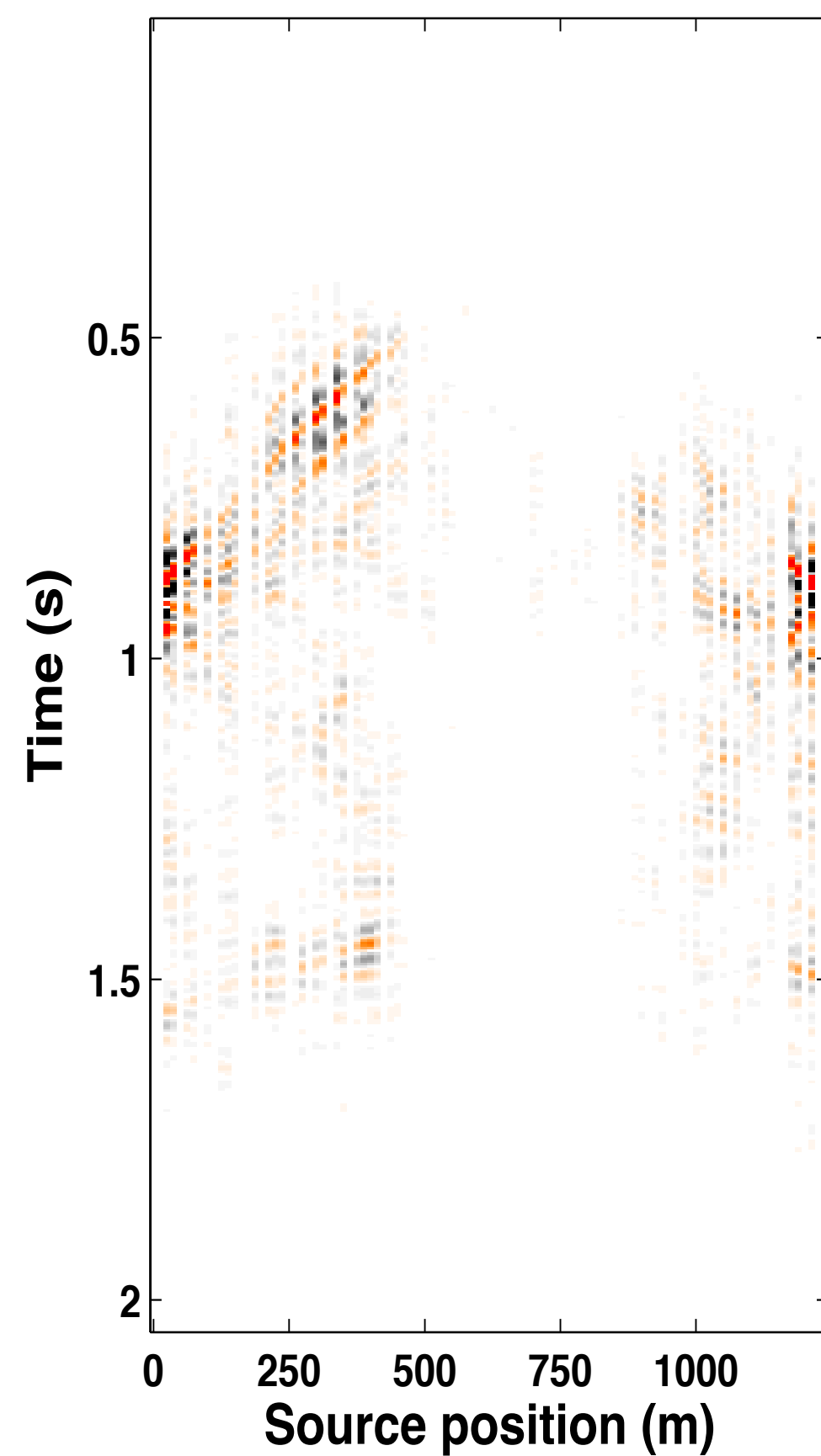
Monitor recovery

- **50%** overlap in acquisition matrices

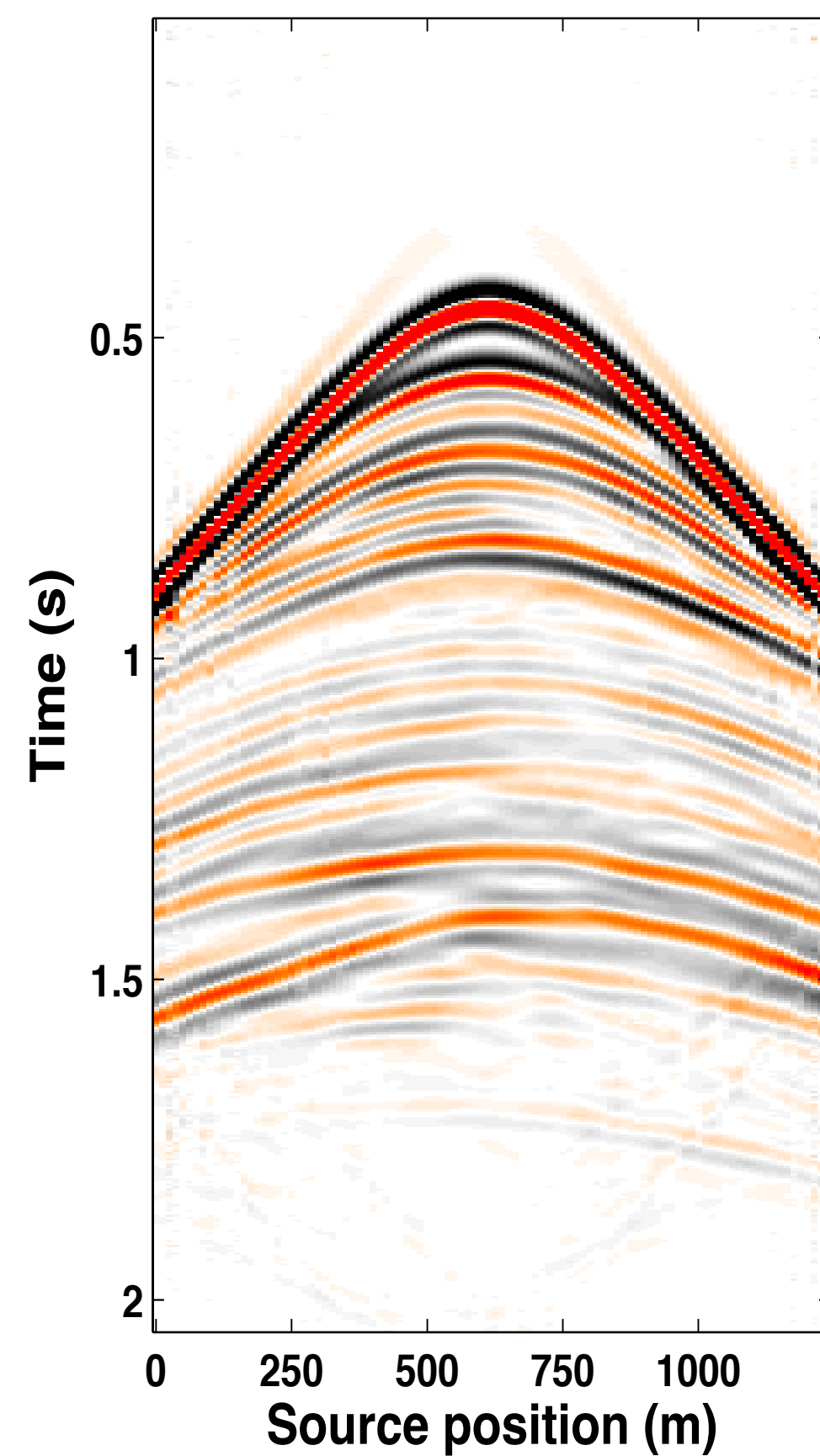
IRS
[11.0 dB]



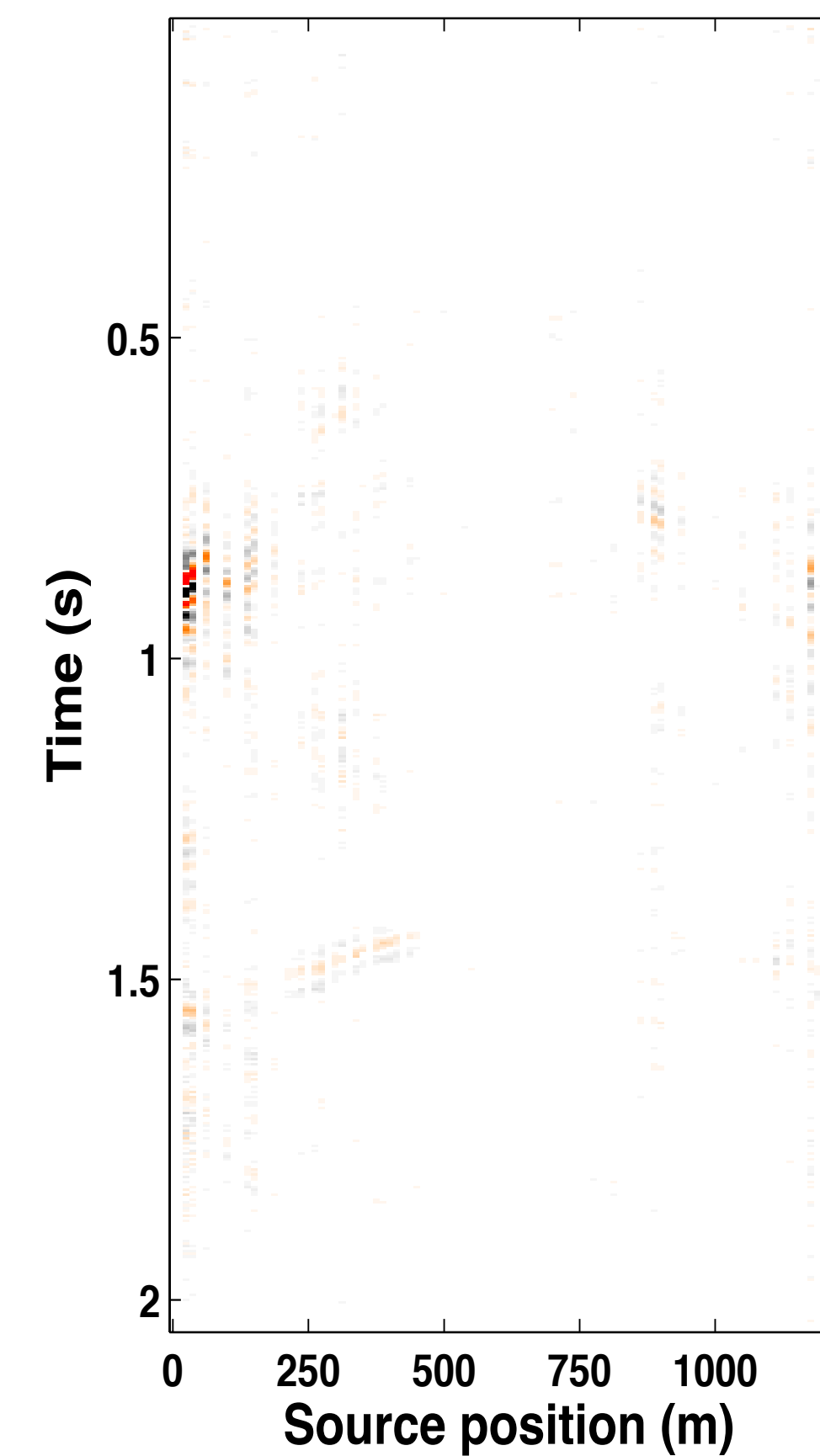
IRS
residual



JRM
[15.8 dB]



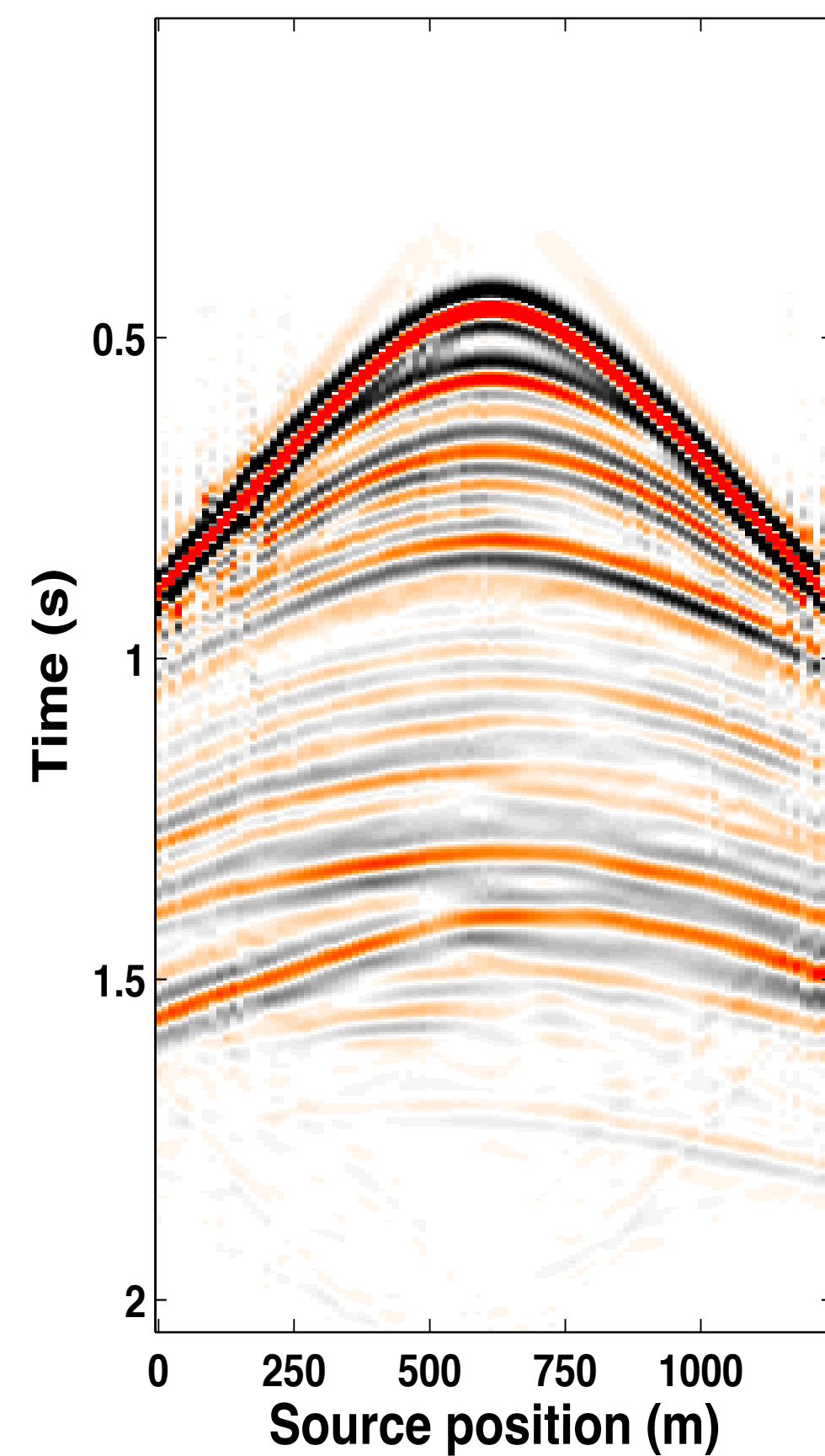
JRM
residual



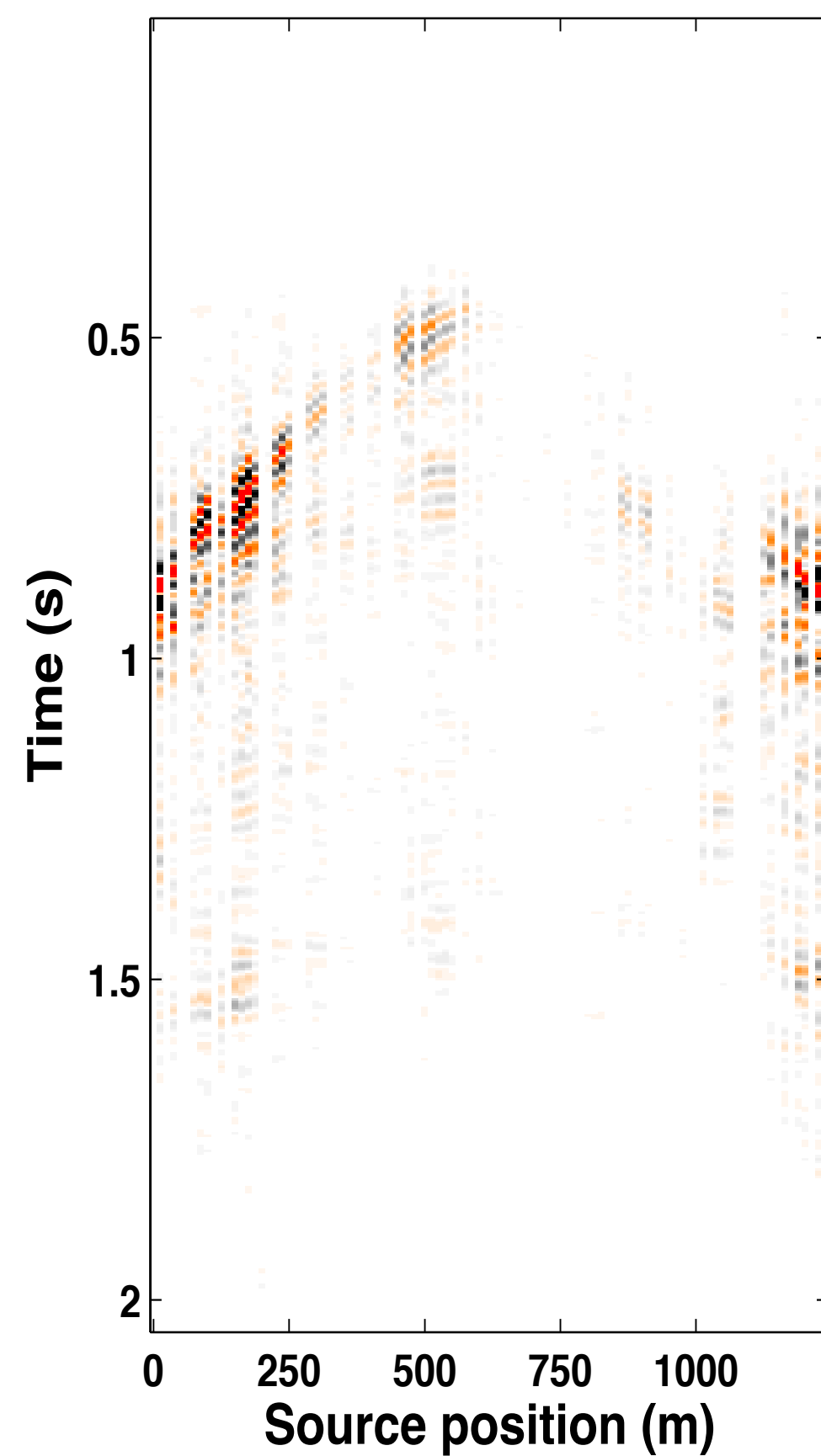
Monitor recovery

- **25%** overlap in acquisition matrices

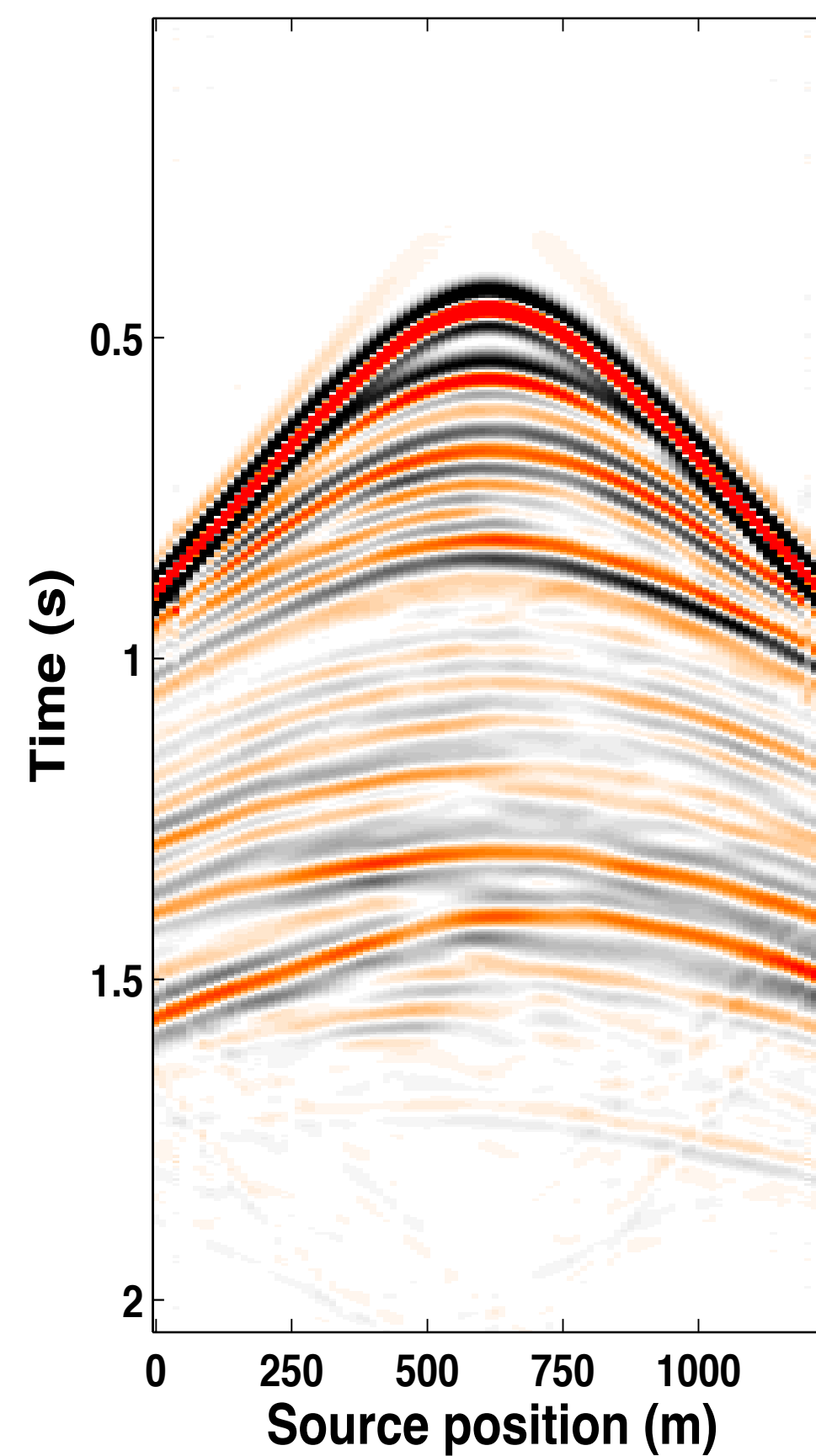
IRS
[10.3 dB]



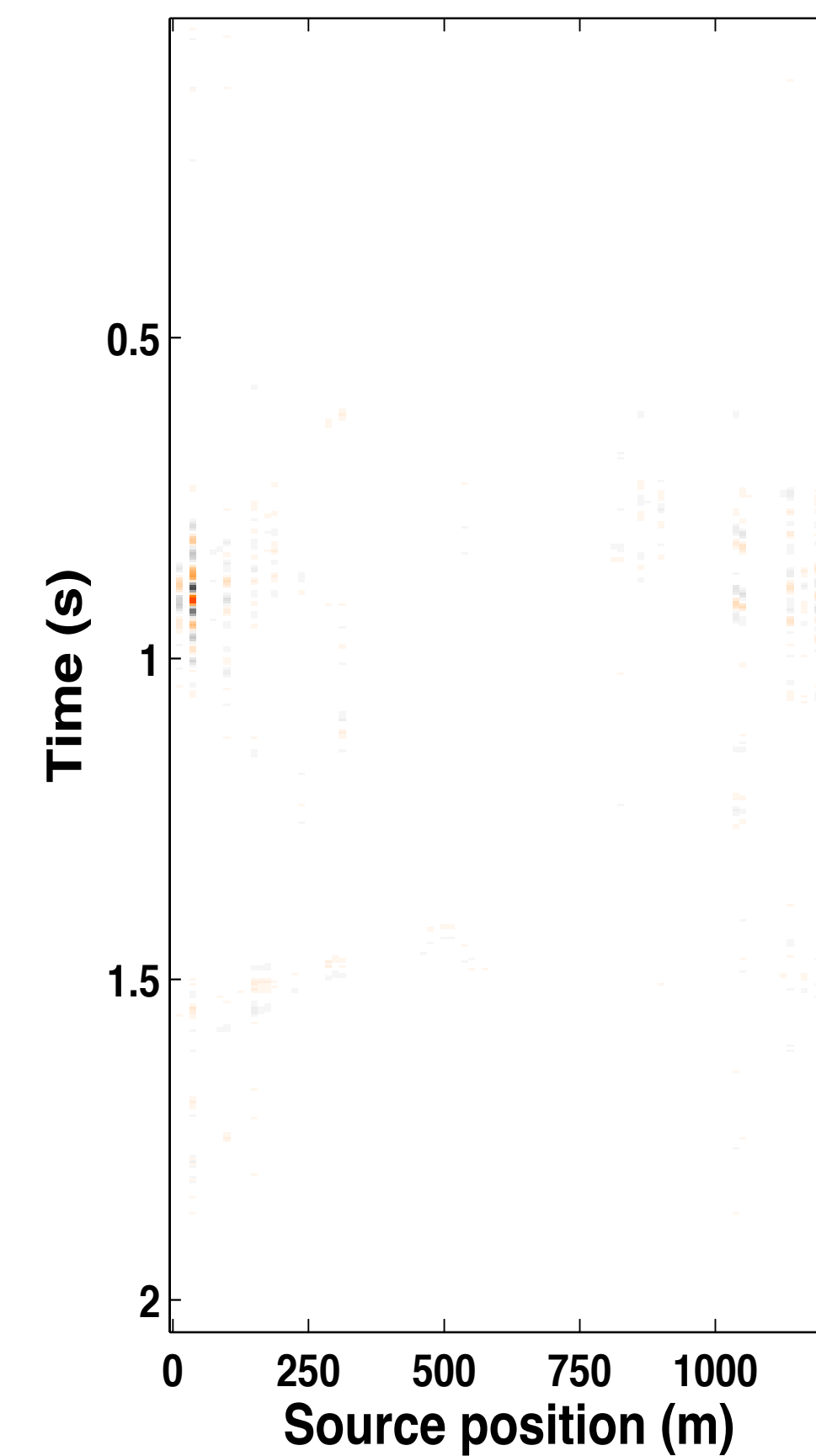
IRS
residual



JRM
[18.4 dB]



JRM
residual

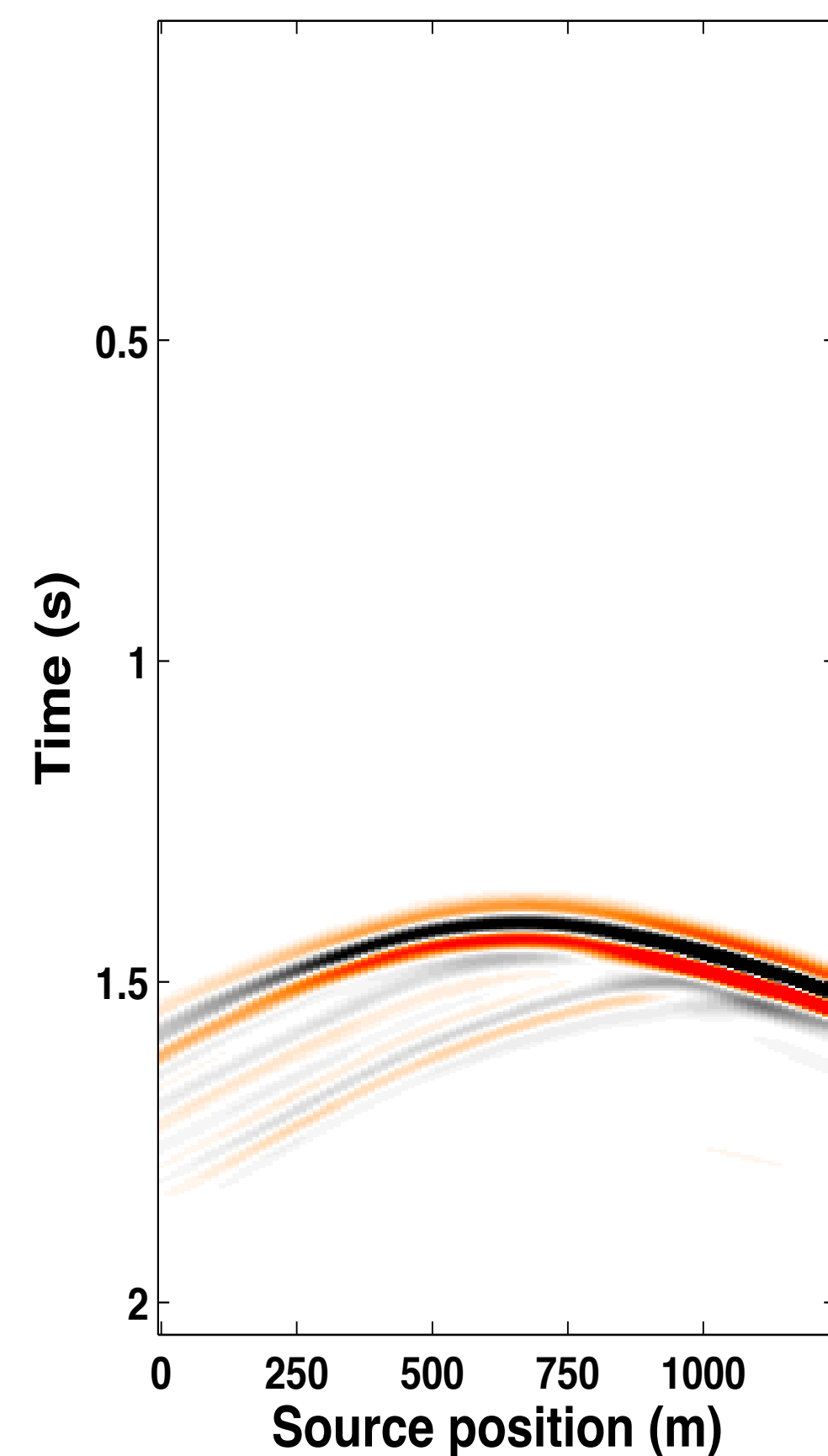


4-D recovery

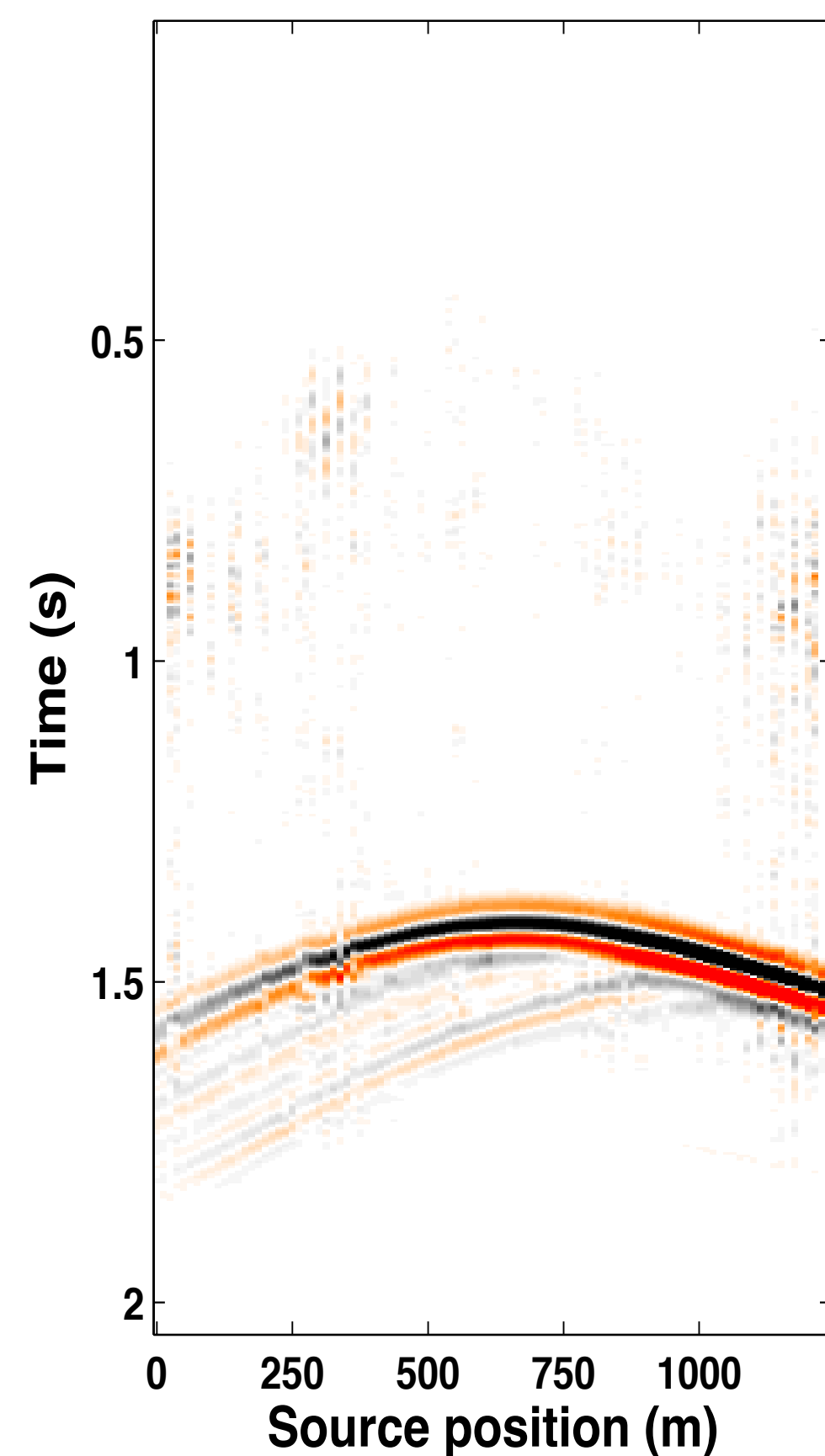
- **100%** overlap in *acquisition* matrices

[colormap scale: 10 X]

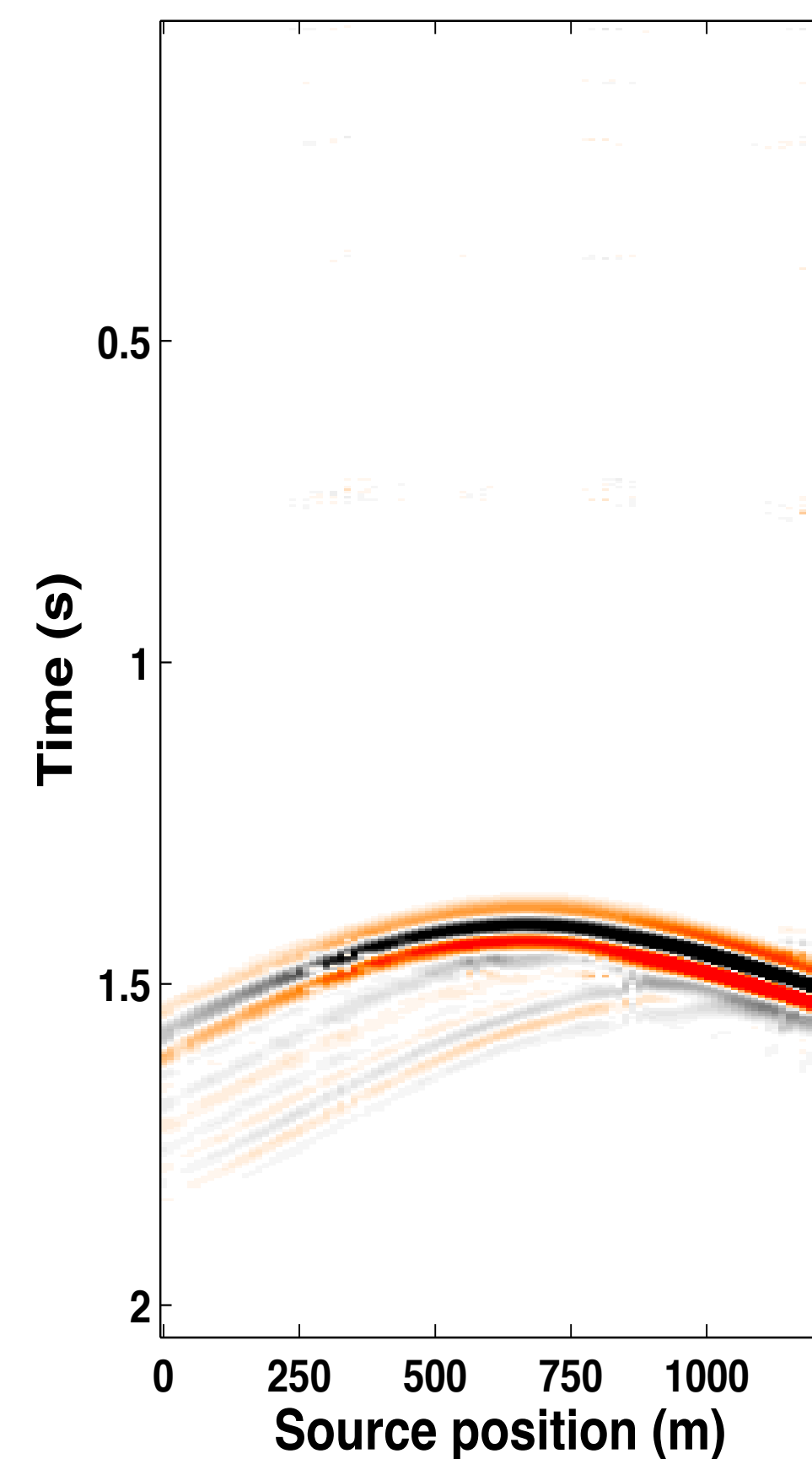
Original



IRS
[10.2 dB]



JRM
[12.5 dB]

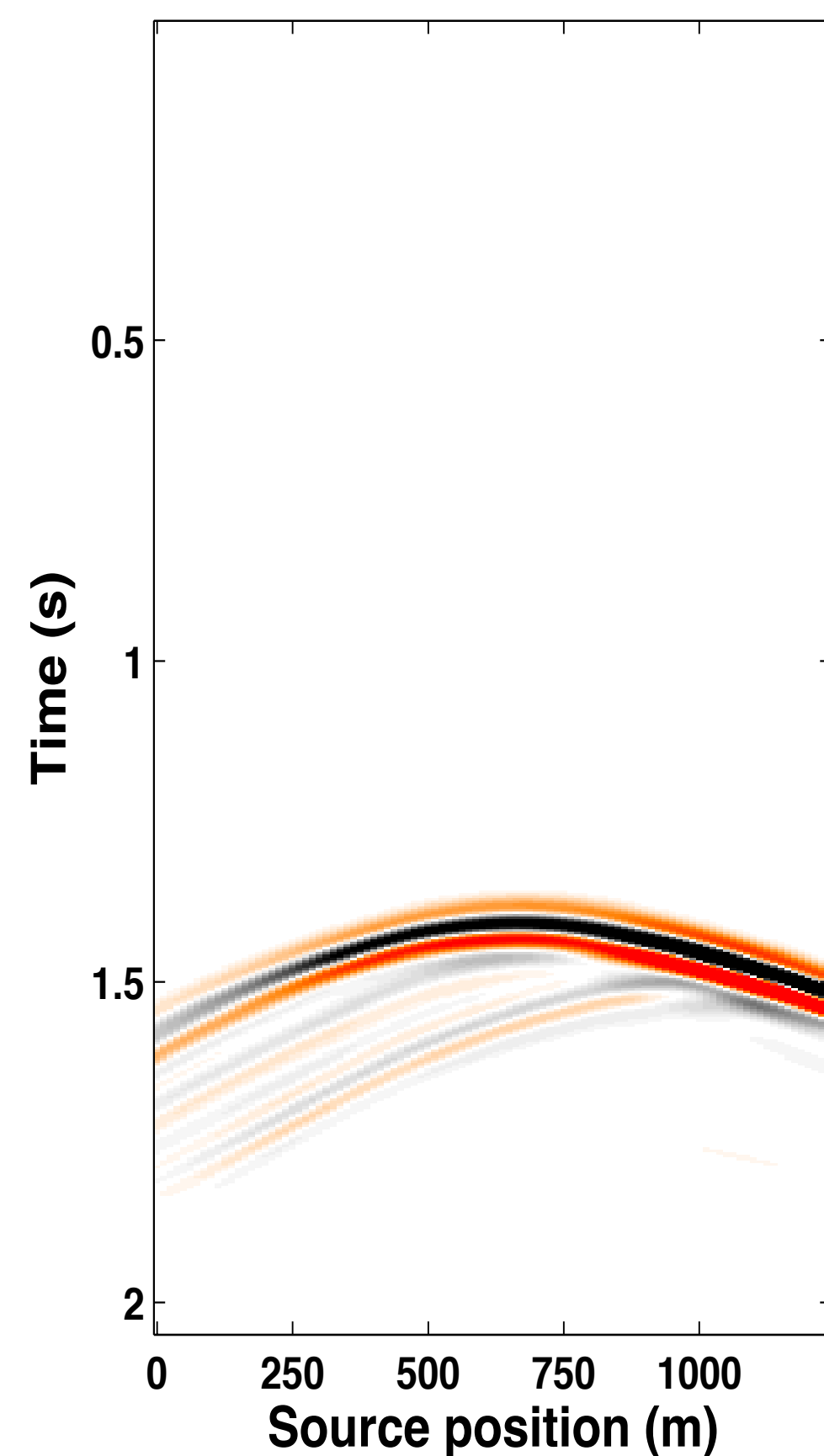


4-D recovery

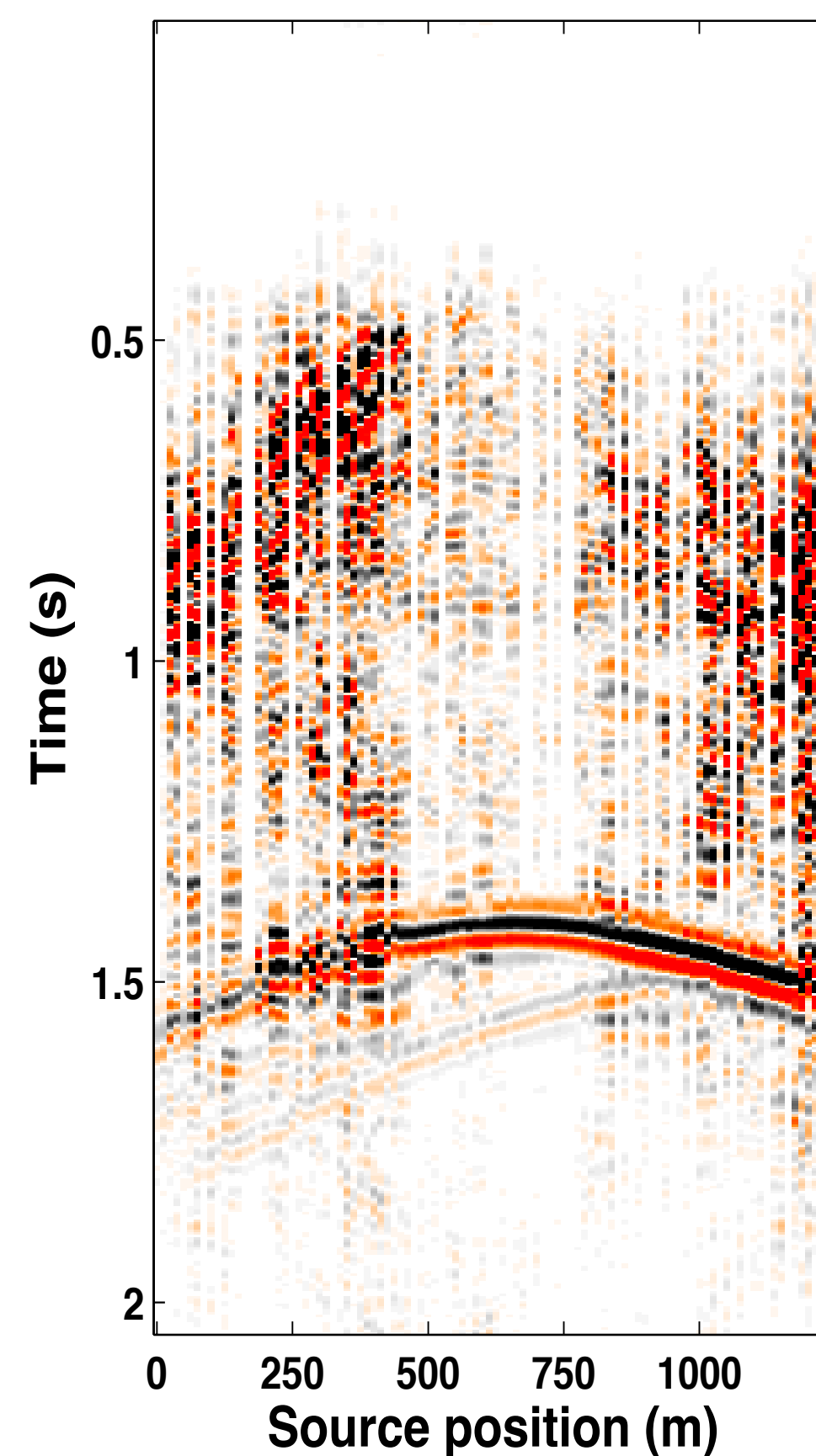
- **50%** overlap in *acquisition* matrices

[colormap scale: 10 X]

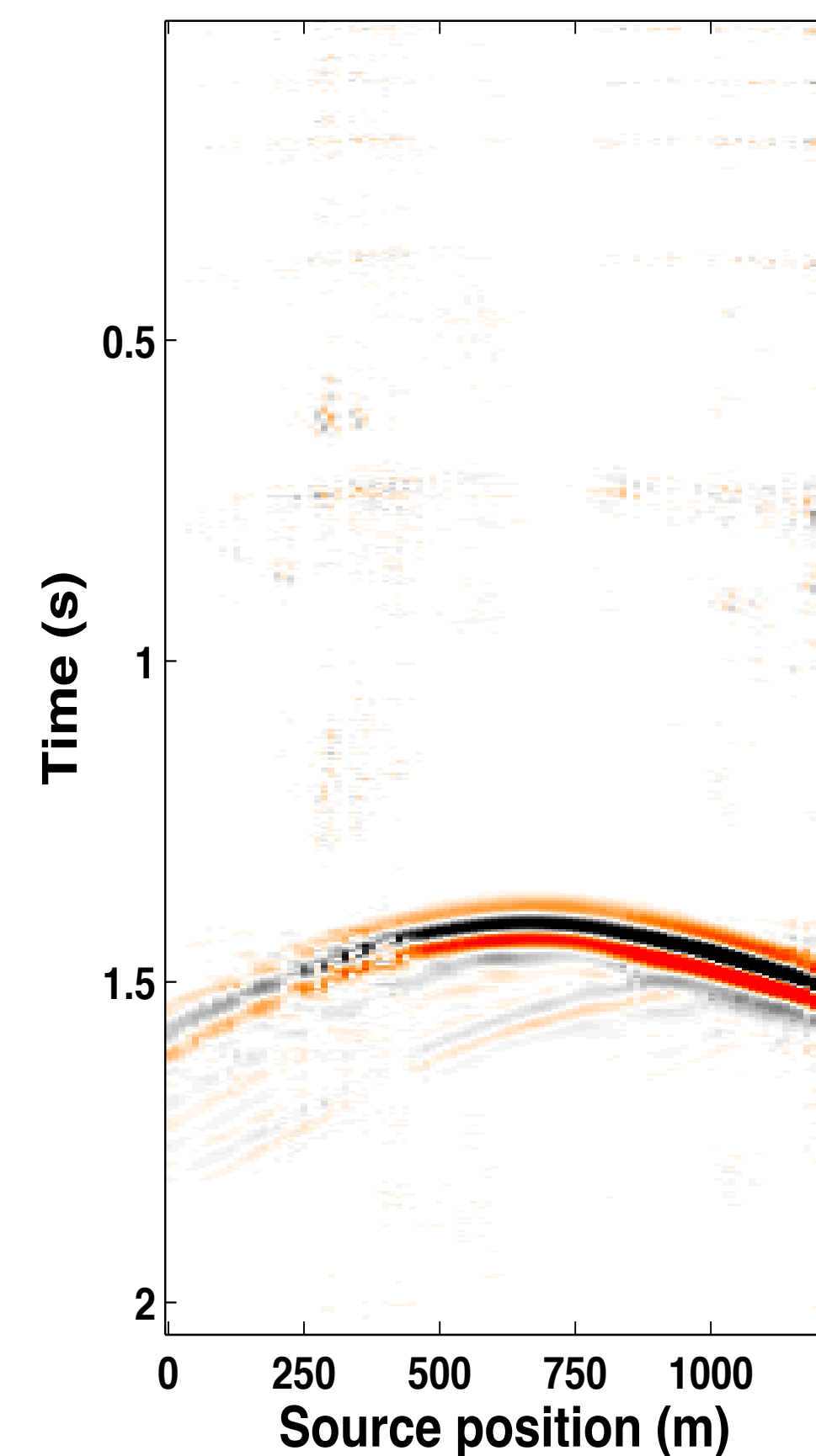
Original



IRS
[-16.0 dB]



JRM
[3.0 dB]

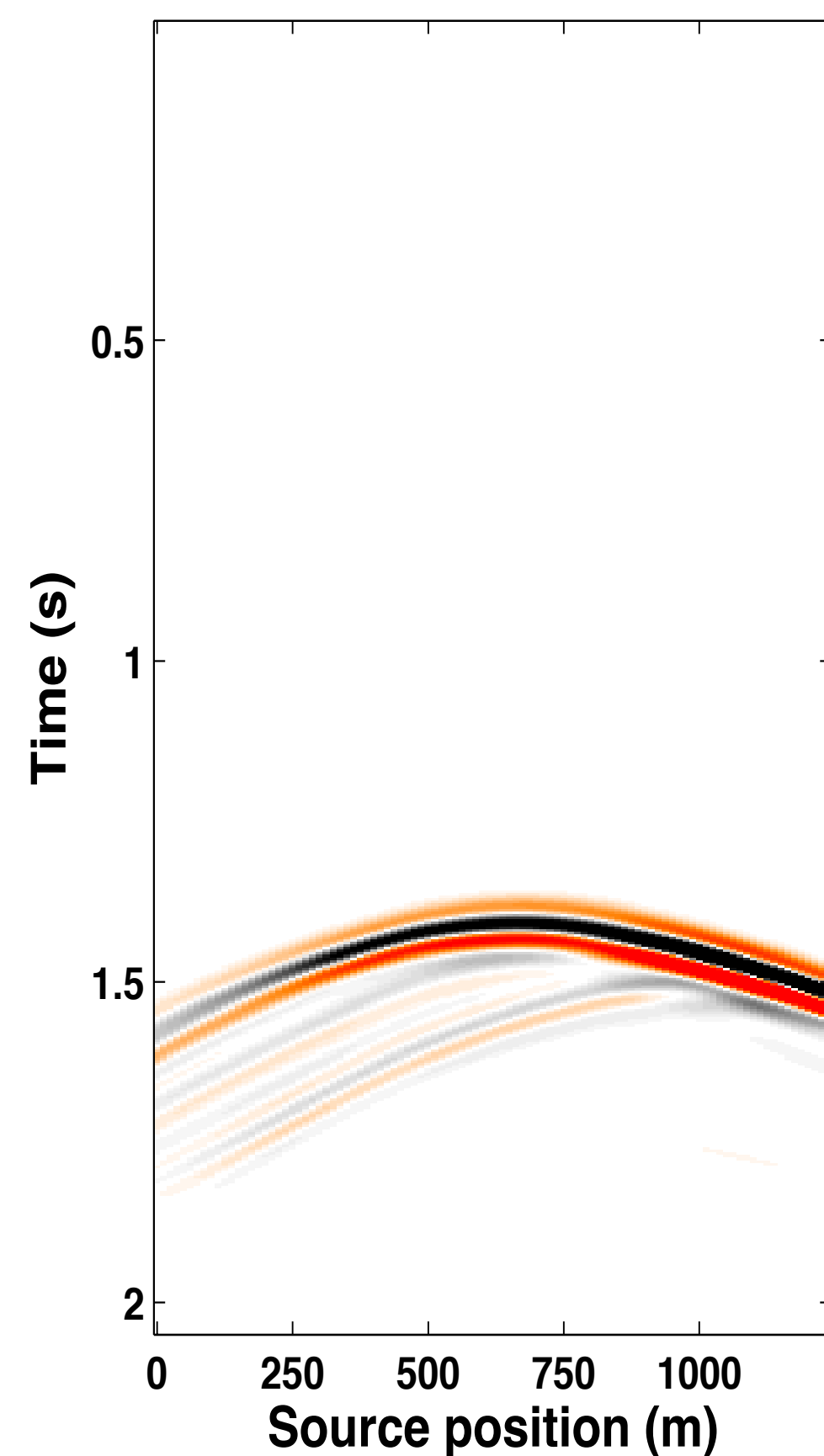


4-D recovery

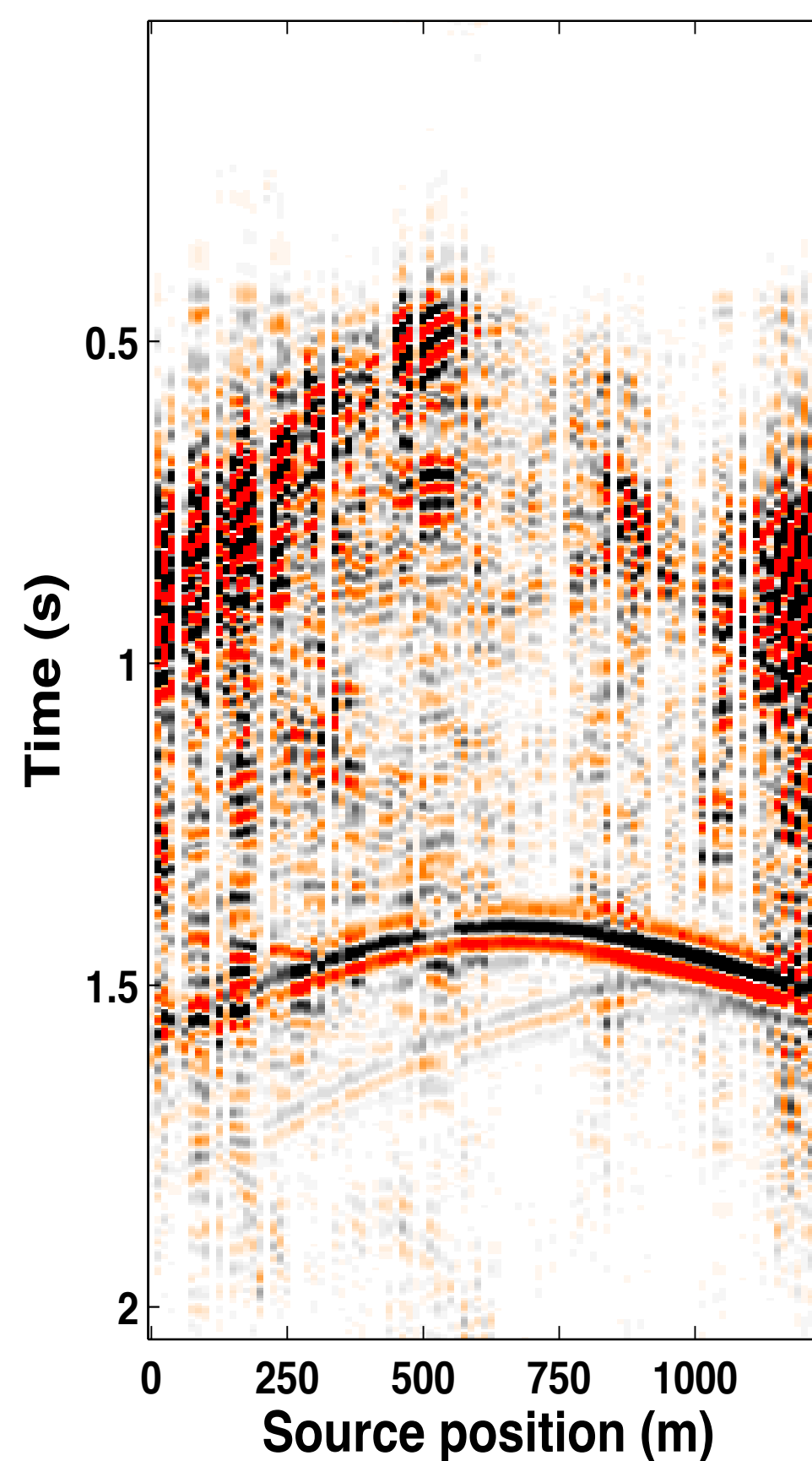
- **25%** overlap in *acquisition* matrices

[colormap scale: 10 X]

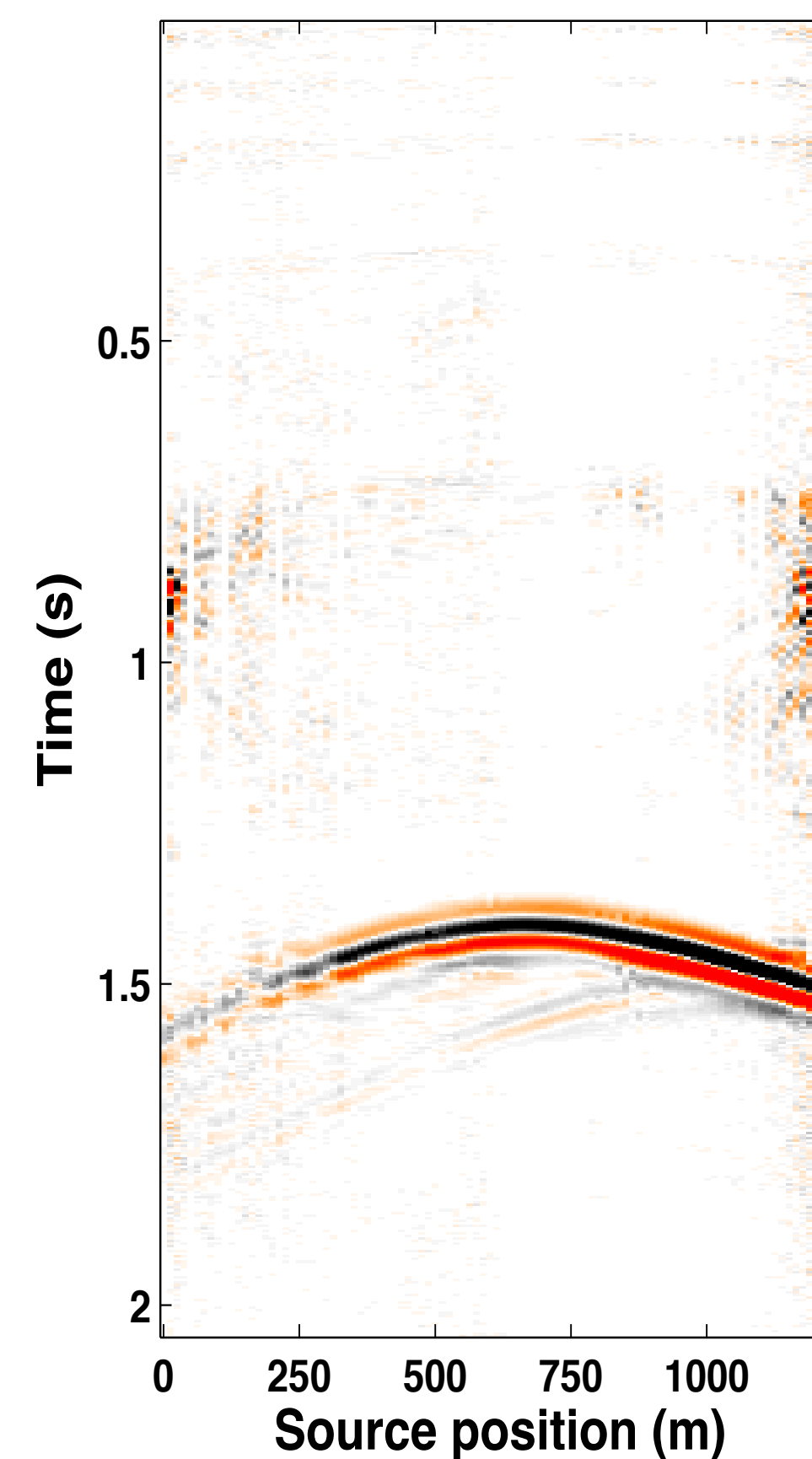
Original



IRS
[-18.5 dB]

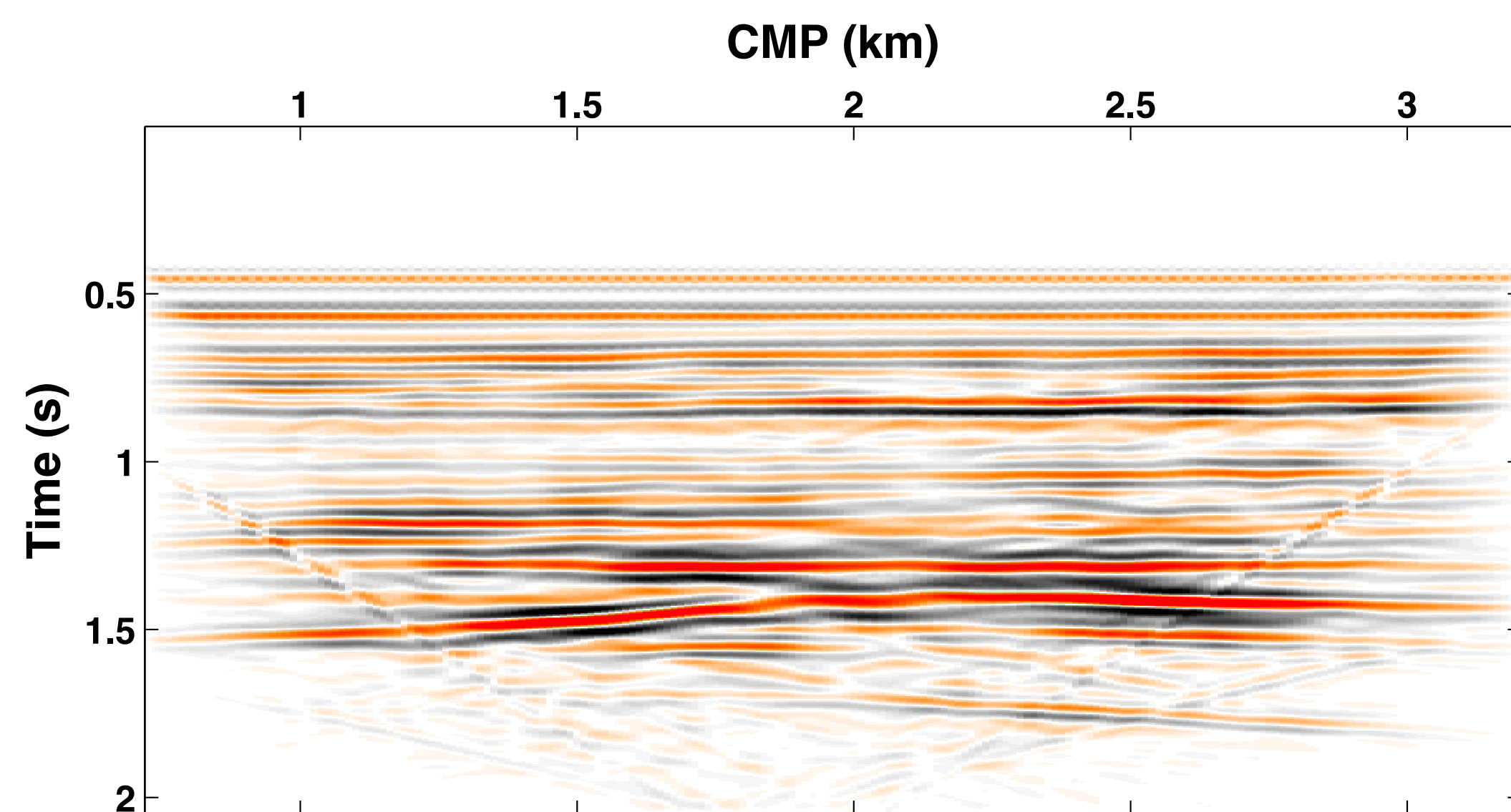


JRM
[-2.1 dB]

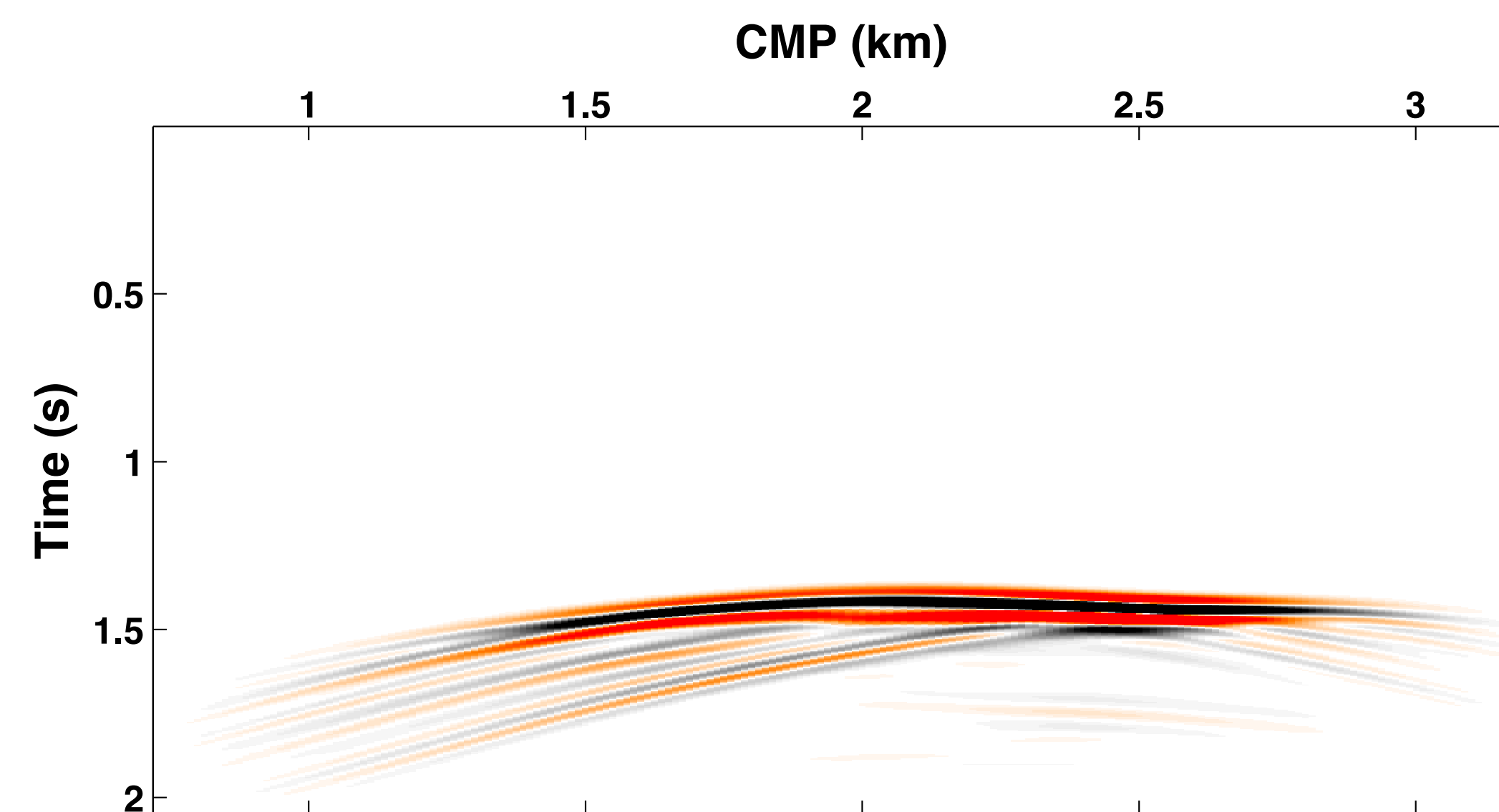


Stacked sections

Baseline



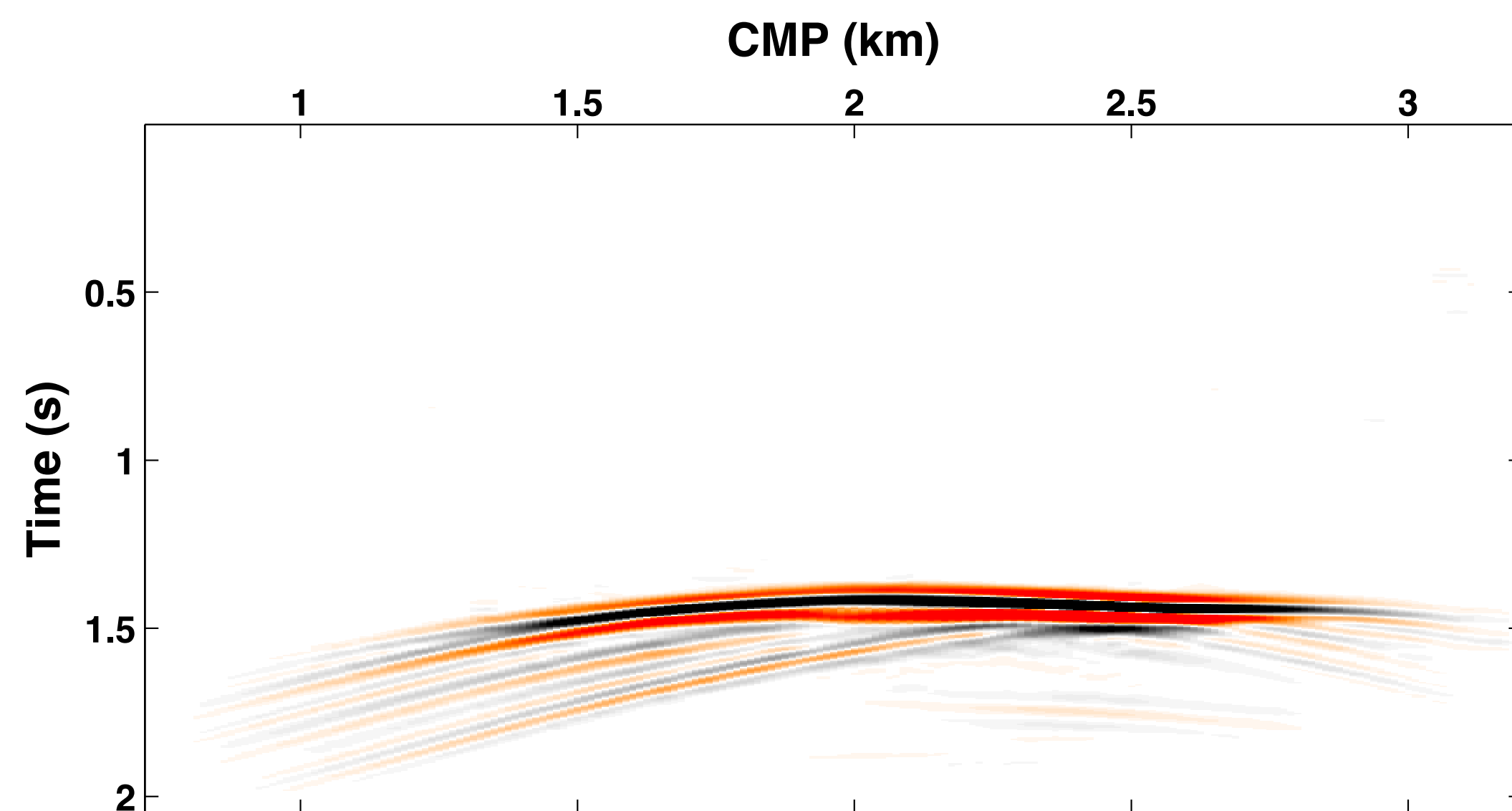
4-D signal [10 X]



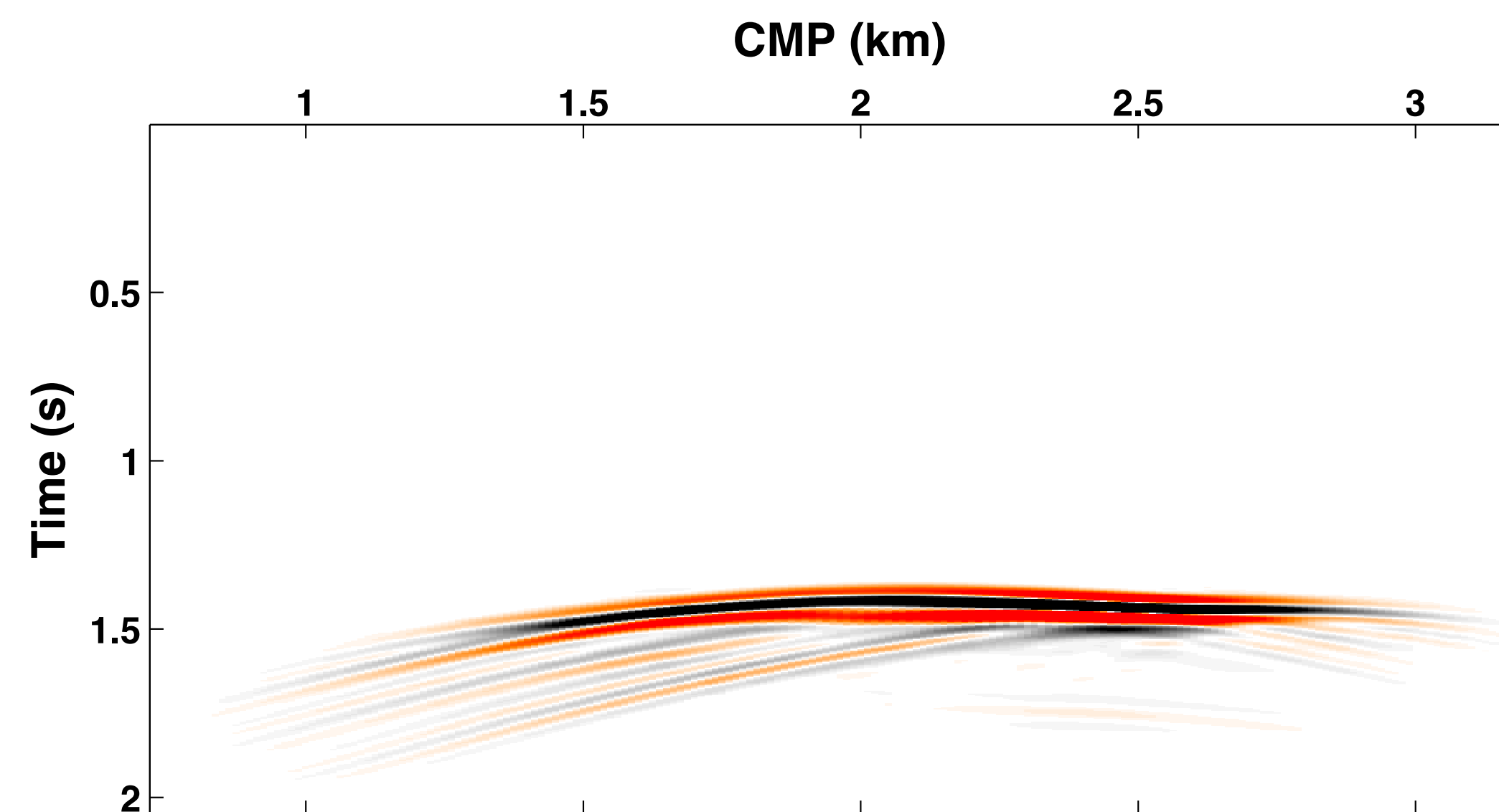
Stacked sections

- **100%** overlap in acquisition matrices

IRS
[25.2 dB]



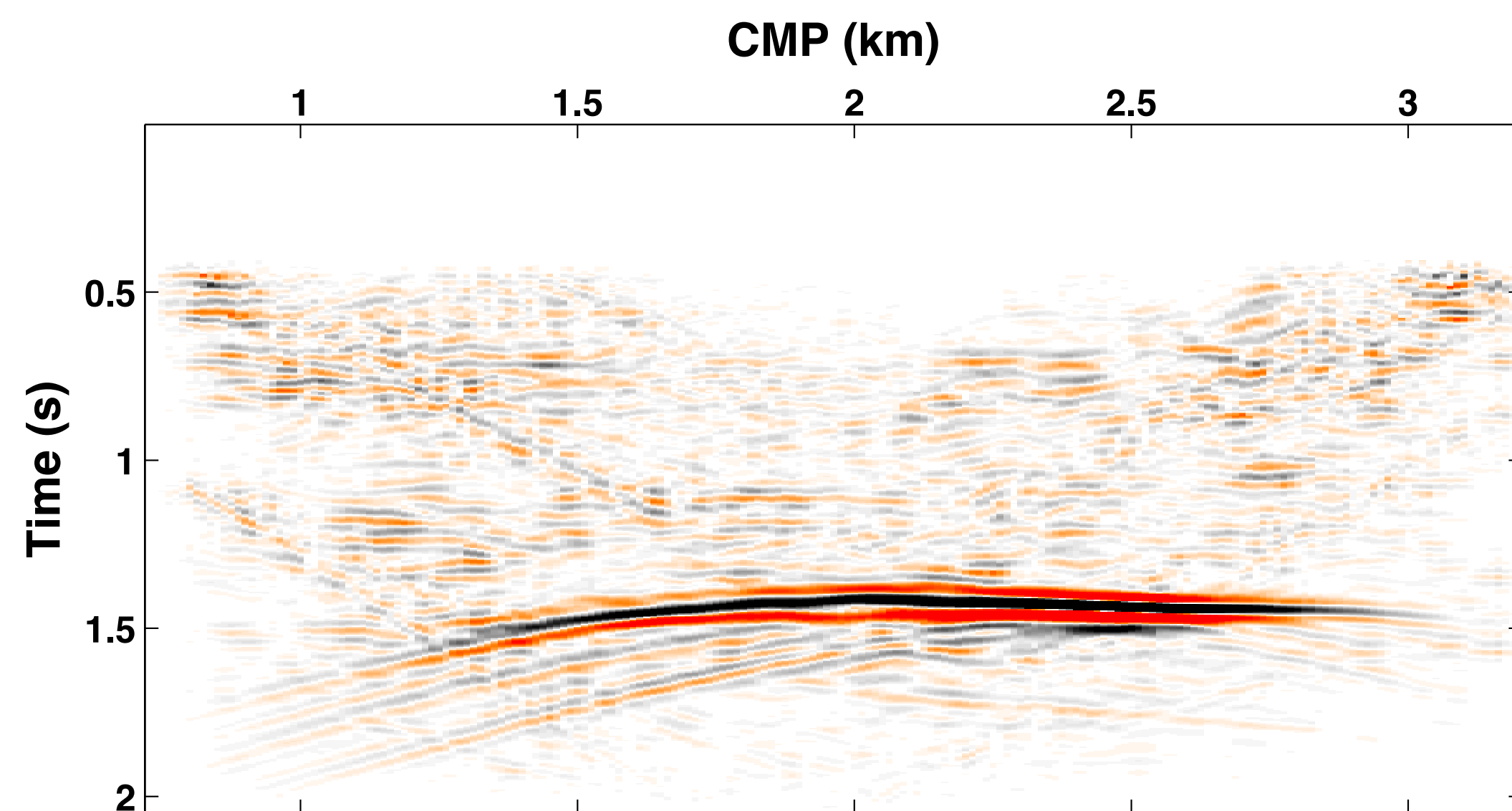
JRM
[23.6 dB]



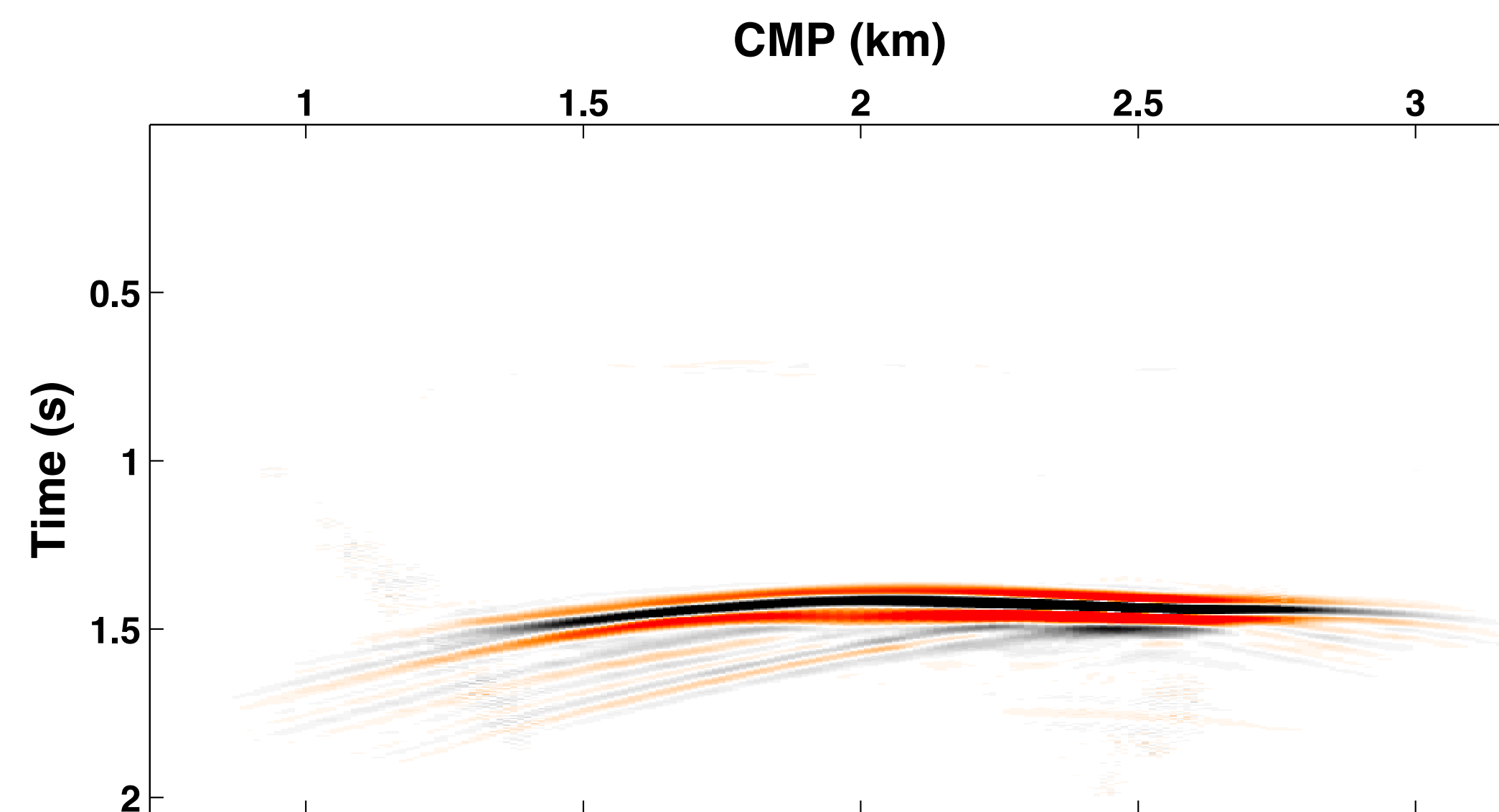
Stacked sections

- **50%** overlap in acquisition matrices

IRS
[9.7 dB]



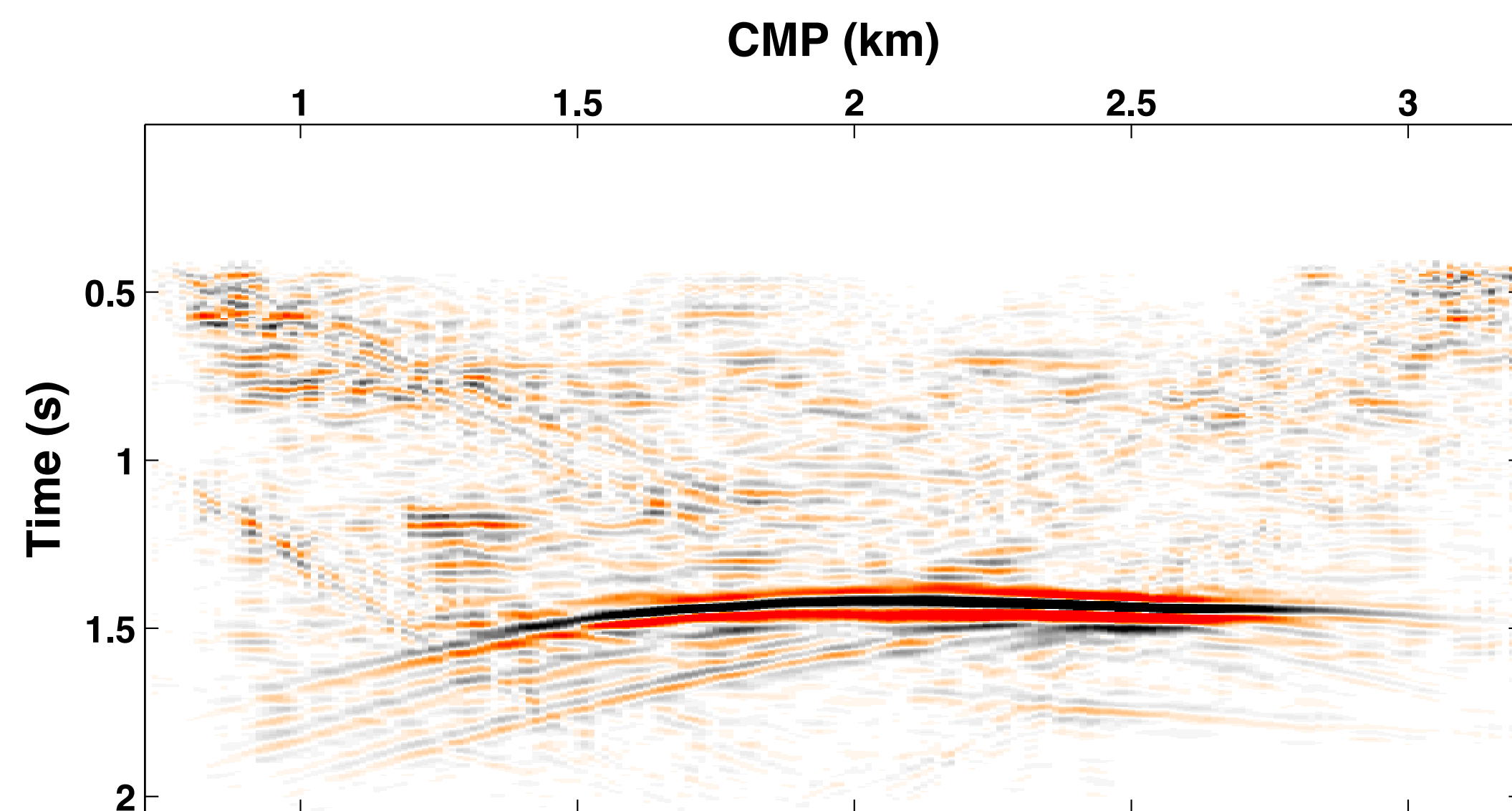
JRM
[17.1 dB]



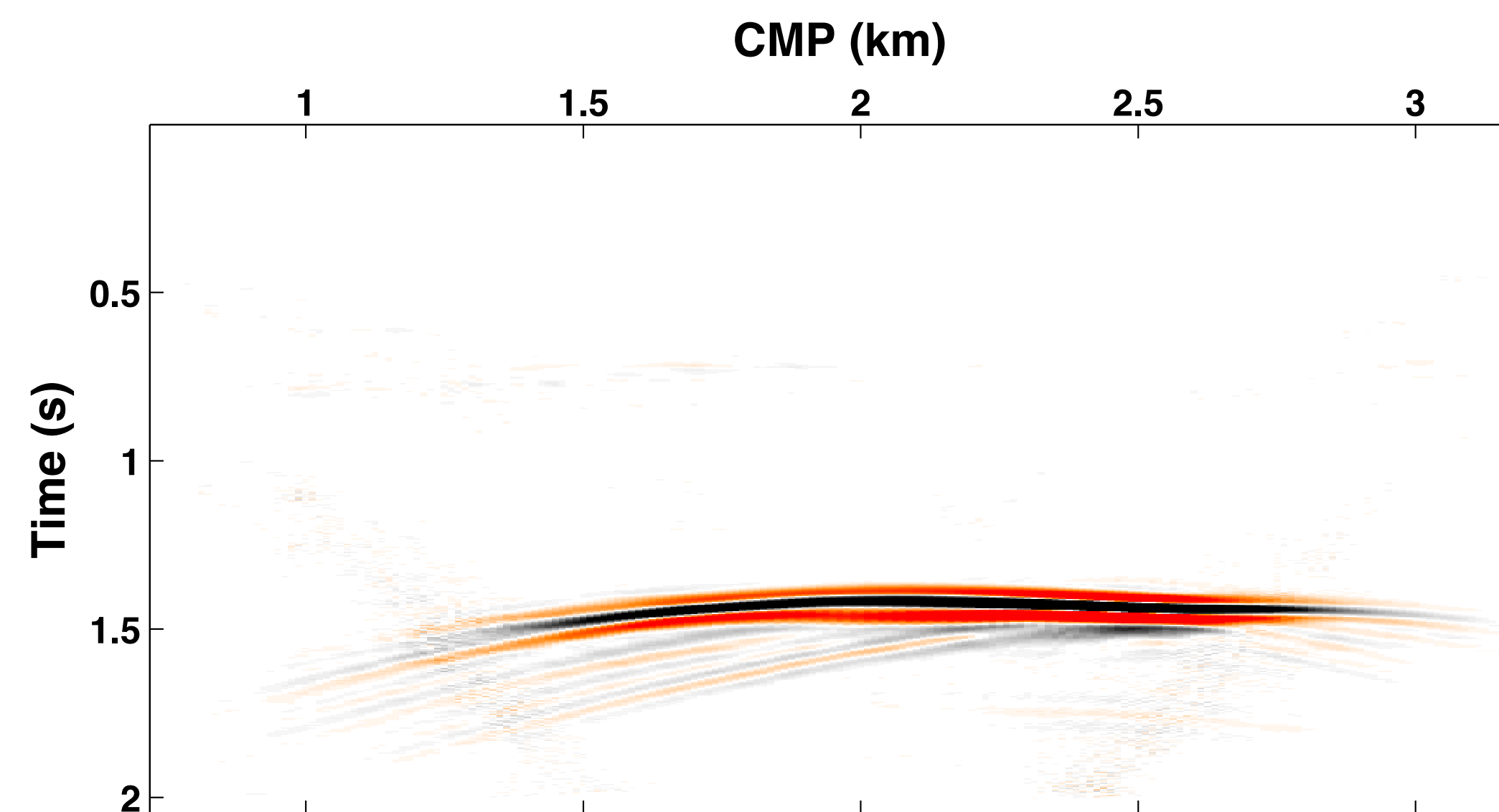
Stacked sections

- **25%** overlap in acquisition matrices

IRS
[9.5 dB]



JRM
[16.0 dB]



SNR (dB)

– average of 5 experiments

overlap	baseline		monitor		4-D signal	
	IRS	JRM	IRS	JRM	IRS	JRM
100%	25.6 ± 1.2	$23.9 \pm \mathbf{1.0}$	25.7 ± 1.1	$24.0 \pm \mathbf{1.0}$	25.0 ± 0.9	$23.4 \pm \mathbf{0.8}$
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	$\mathbf{34.4 \pm 0.9}$	23.5 ± 1.3	$\mathbf{33.6 \pm 0.8}$	8.5 ± 1.3	$\mathbf{15.9 \pm 0.7}$

Observations

Stylized synthetics give *fundamental* insights when recovering signals in 4-D seismic

Seismic synthetics show that we do **not** necessarily have to insist on full *repetition* depending on the *recovery* of the *vintages*

Approach is *trivially* extendable to *multiple* vintages & *image* space*

[*Felix Oghenekowho, “Randomized sampling without repetition in time-lapse seismic surveys”. Wednesday 08:55 AM, Room: 4EF]

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 subsampling*?

✓ no, as long as one samples *sufficiently* to recover **both** vintages *jointly*

✓ yes, if recovery of vintages *fails* and *one* has a *high* degree of *repetition* then the *only* hope is to recover the *difference*, *not* recommended

Acknowledgements

Thank you for your attention!

SINBAD



This work was in part financially supported by the Natural Sciences and Engineering Research Council of Canada Discovery Grant (22R81254) and the Collaborative Research and Development Grant DNOISE II (375142-08). This research was carried out as part of the SINBAD II project with support from the following organizations: BG Group, BGP, BP, CGG, Chevron, ConocoPhillips, ION, Petrobras, PGS, Statoil, Total SA, Sub Salt Solutions, WesternGeco, and Woodside.