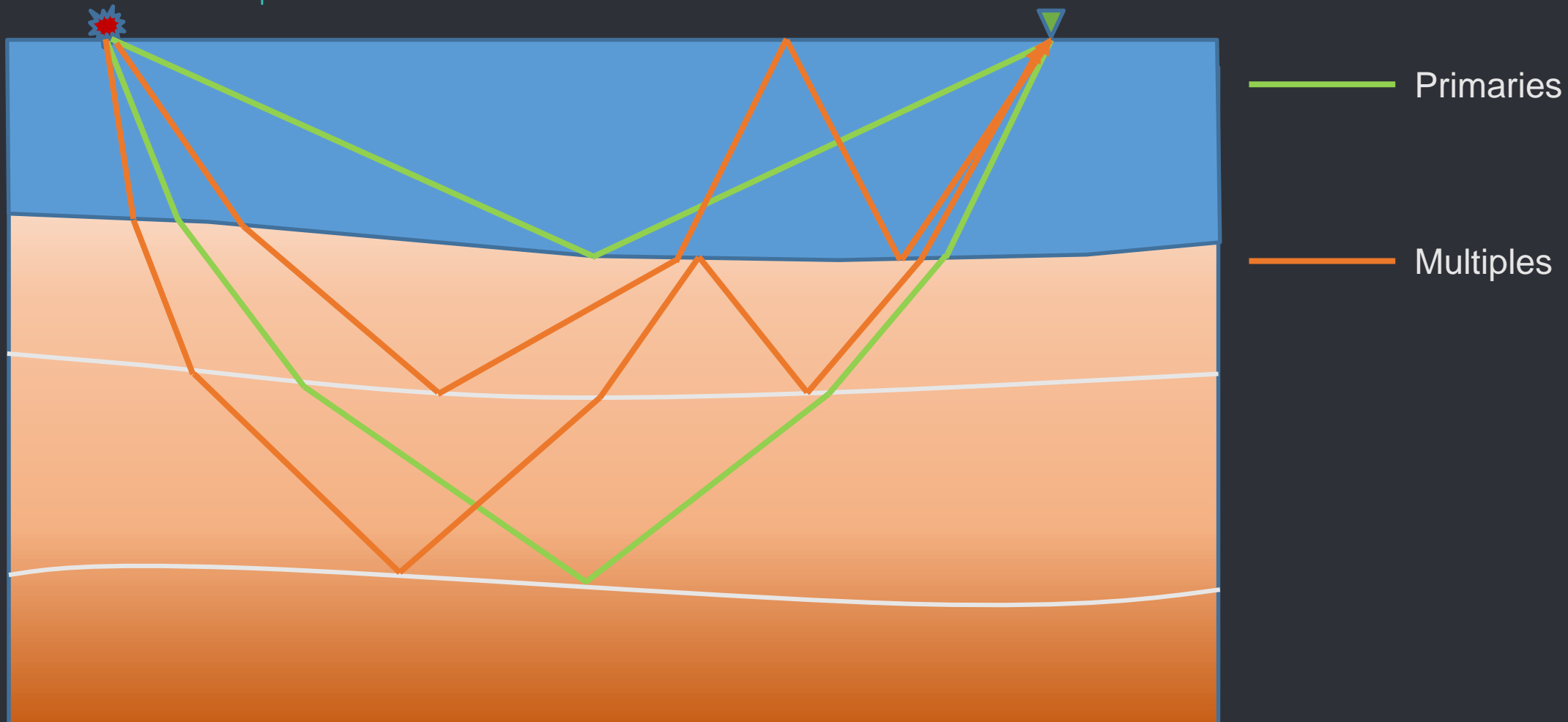




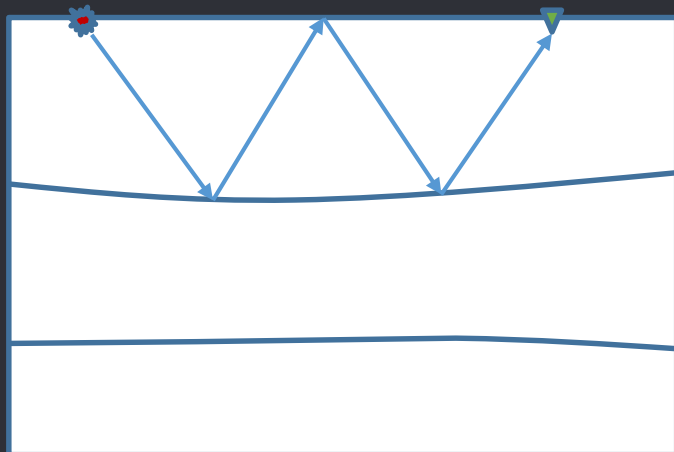
Zero-Offset Demultiple – Efficient Demultiple of Near-Offset Marine Seismic Data in RadExPro

Problem of Multiple Waves

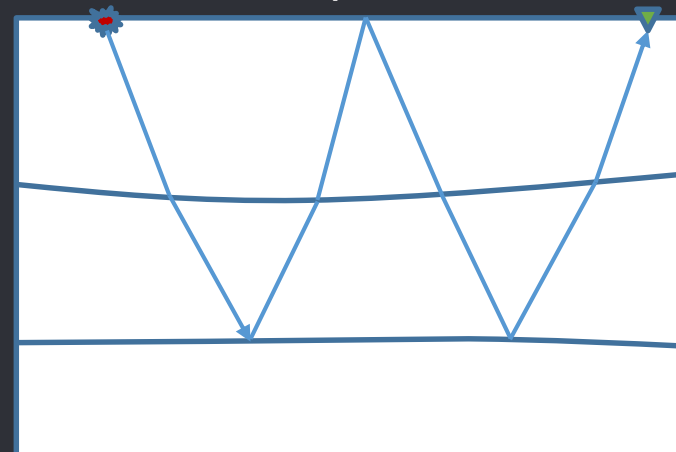


Types of Multiples

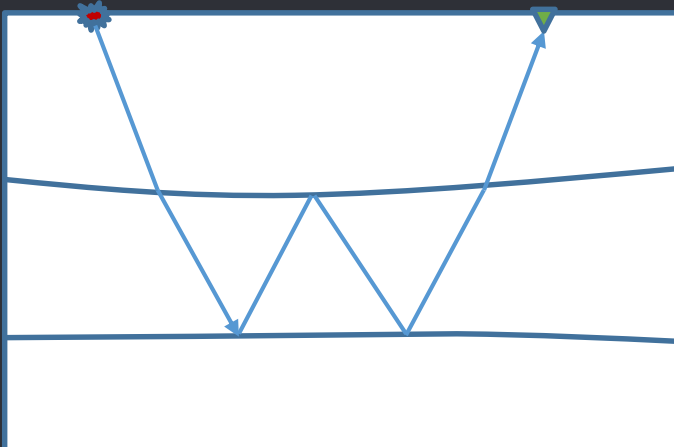
Water bottom multiple



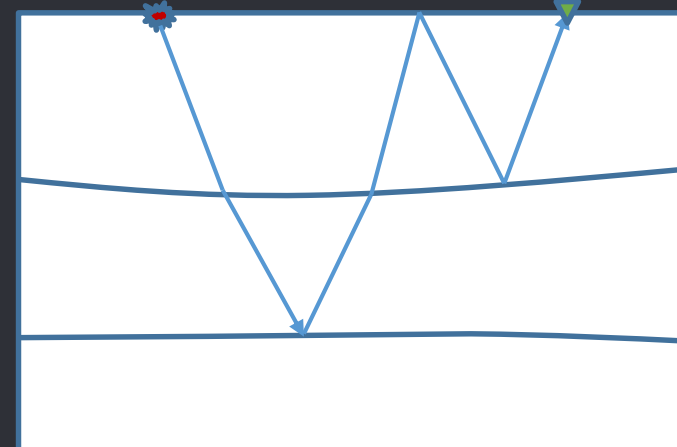
Reflector multiple



Internal multiple



Reflector peg-leg multiple



Multiple Elimination

- **Main Multiple Elimination Techniques**

1. Periodicity of Multiples – **Deconvolutions**

2. Different Move-Out between Multiples and Primaries -- **Radon/F-K/Tau-Pi demultiple, slant-stack, etc.**

3. Wavefied Prediction and Subtraction – **SRME**

Multiple Elimination

○ Main Multiple Elimination Techniques

~~1. Periodicity of Multiples – **Deconvolutions**~~

-- very limited usability: flat seafloor only, very shallow water, not very efficient

~~2. Different Move-Out between Multiples and Primaries -- **Radon/F-K/Tau-Pi demultiple, slant-stack, etc.**~~

-- fail for near offset seismic data (Yilmaz, 1989)

3. Wavefield Prediction and Subtraction – **SRME**

-- work well for multi-channel data, time-consuming

HR/URH specific – **Zero-Offset Demultiple**

-- for single-channel near-offset data

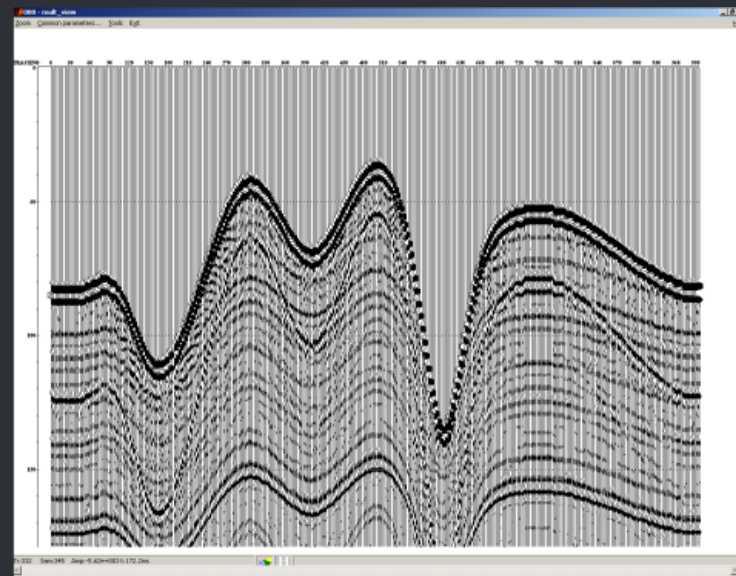
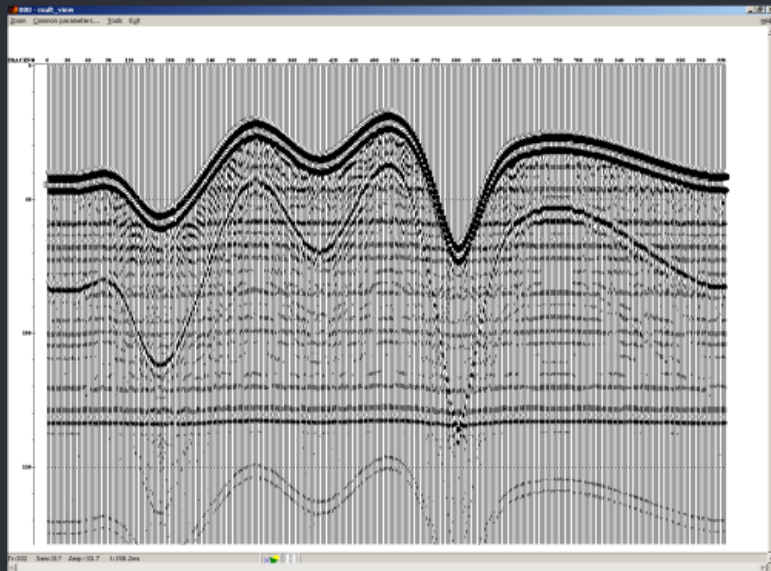
Multiple Elimination: Zero-Offset Demultiple

Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

- Statics shift to the time of seafloor reflection – **model of peg-leg multiples**



Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

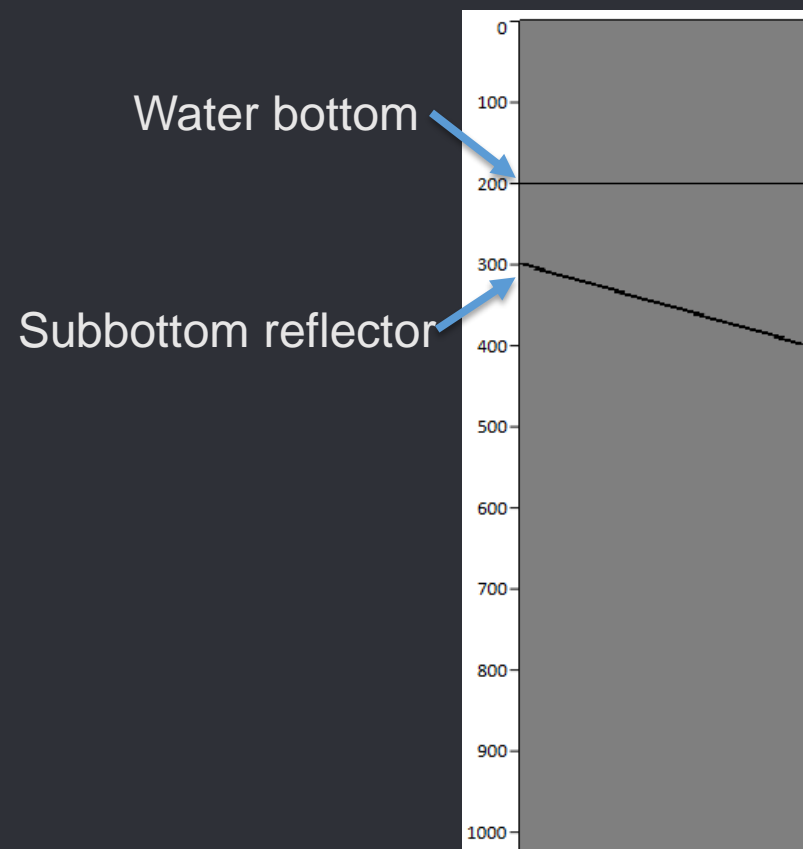
- Autoconvolution of each trace

Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

- Autoconvolution of each trace

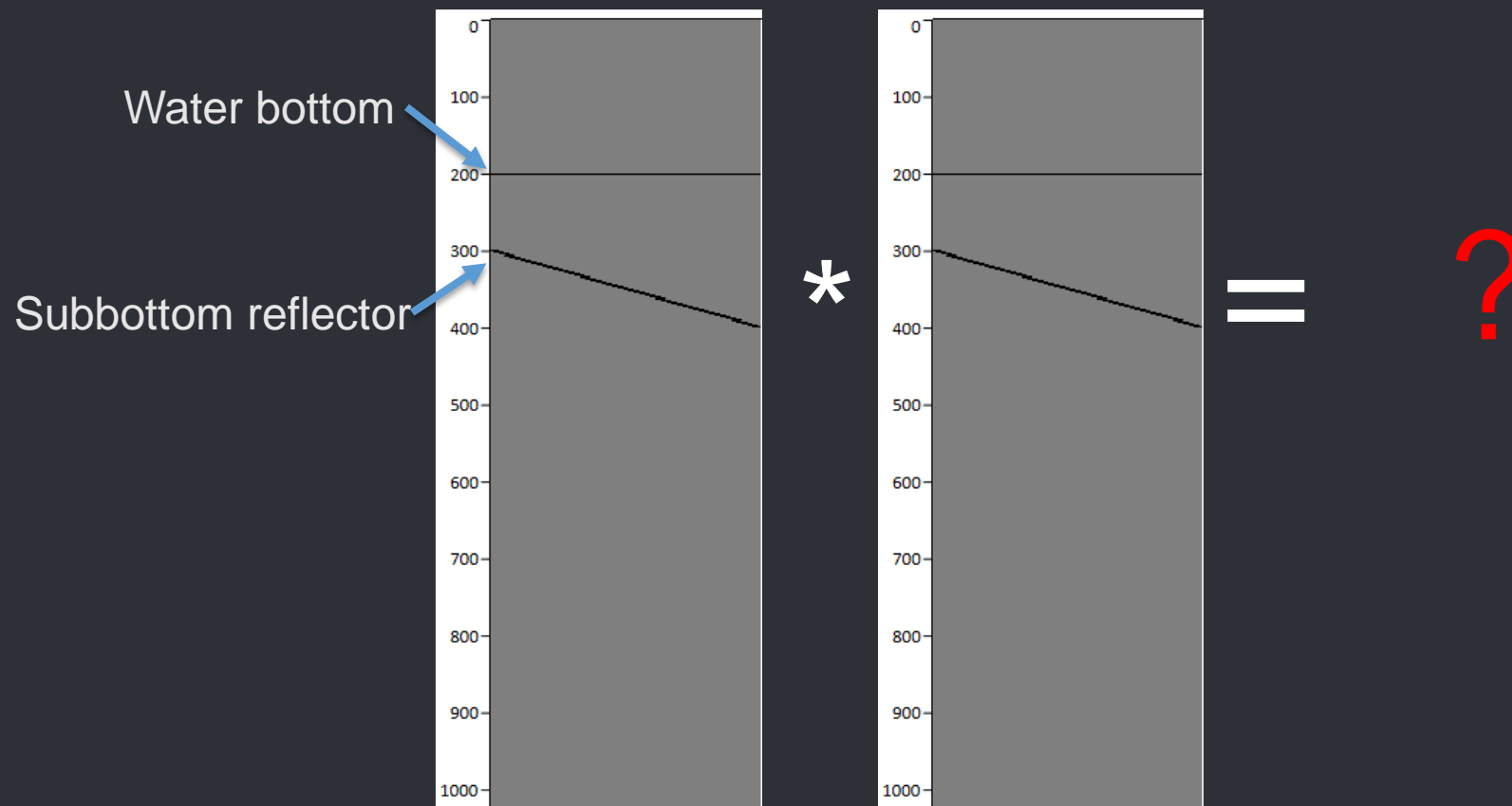


Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

- Autoconvolution of each trace

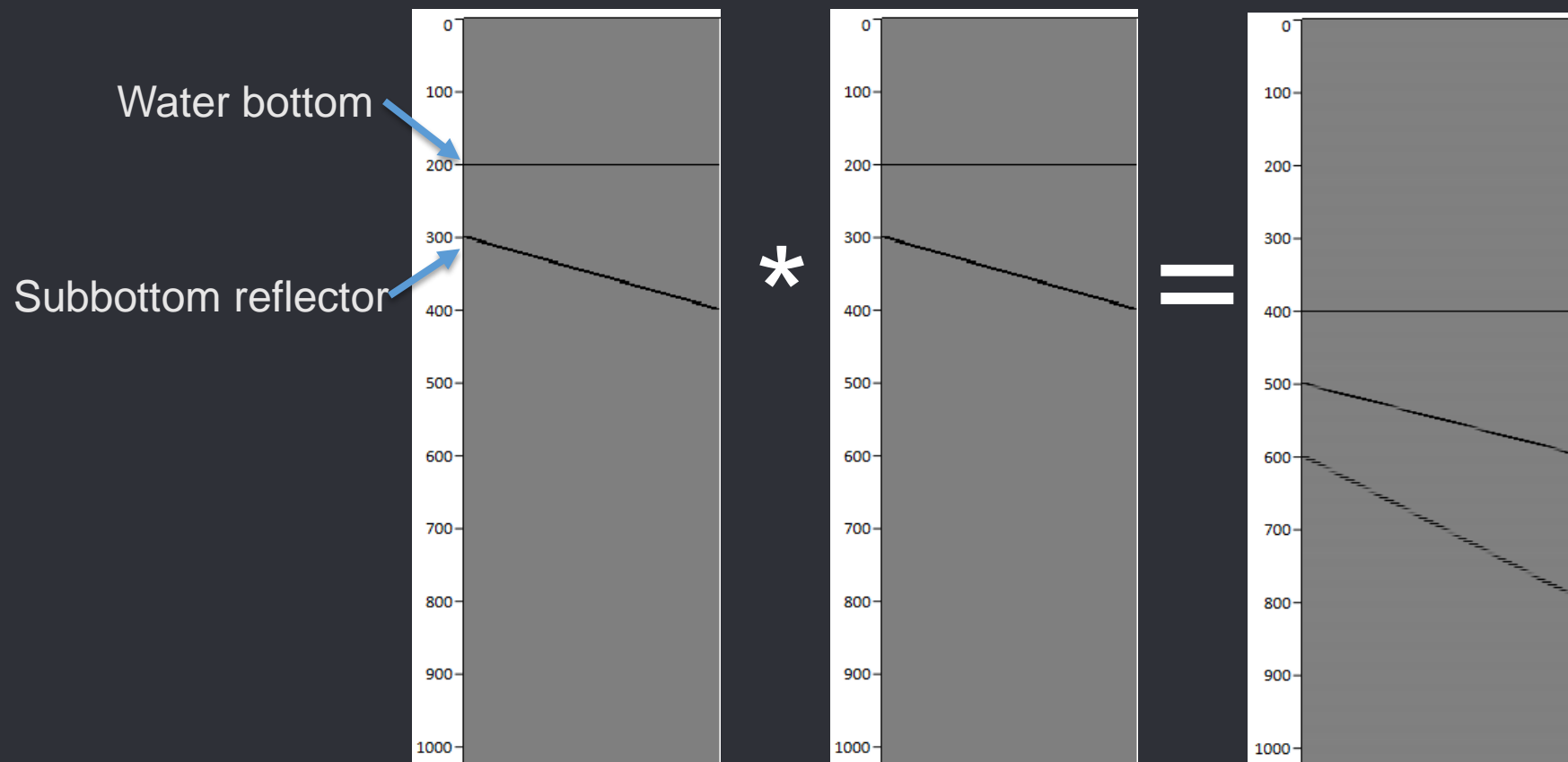


Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

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Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

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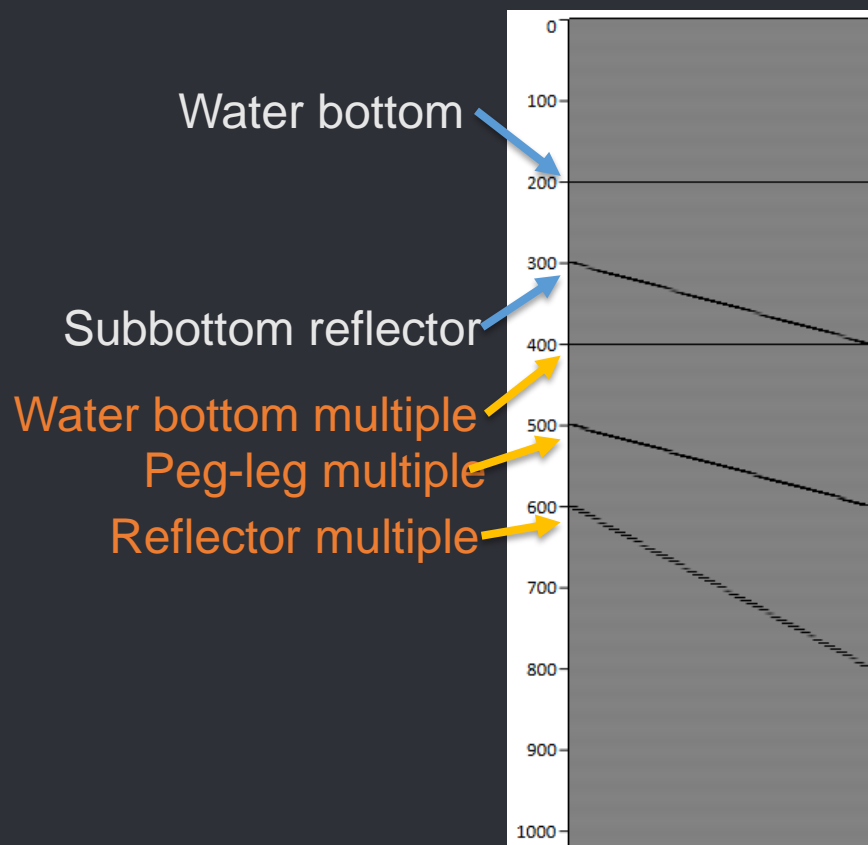


Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

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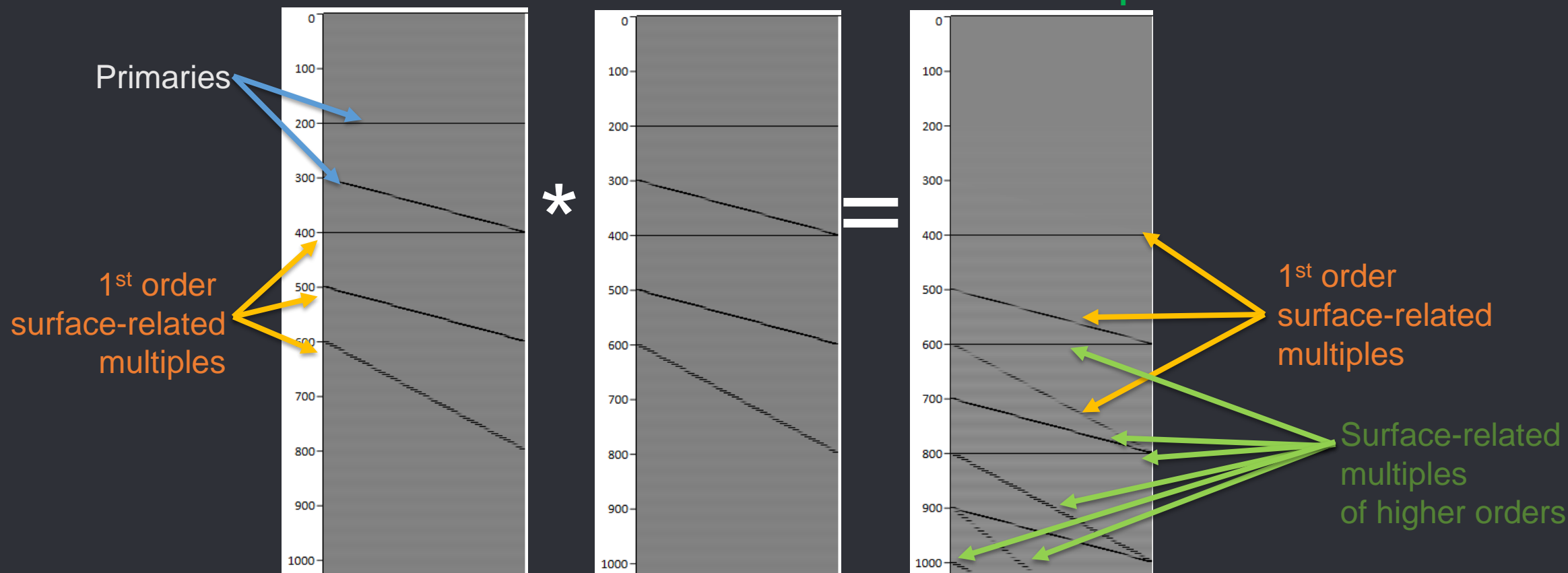


Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

- Autoconvolution of each trace – **model of all surface-related multiples**

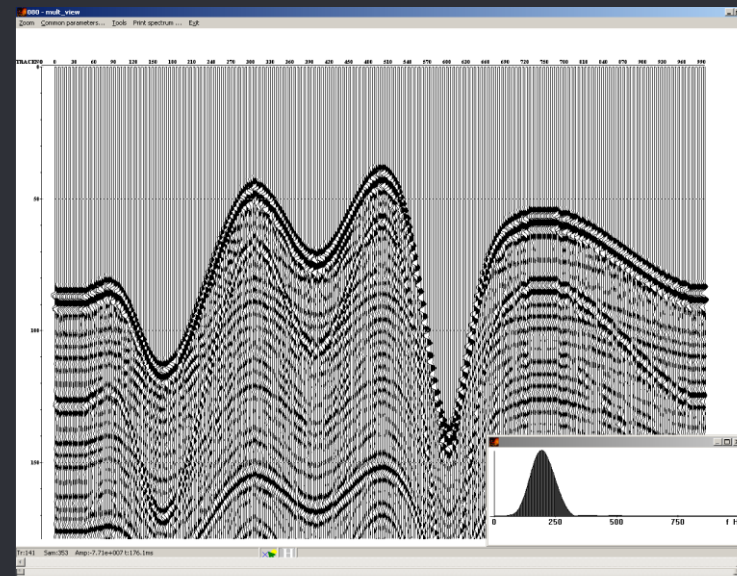
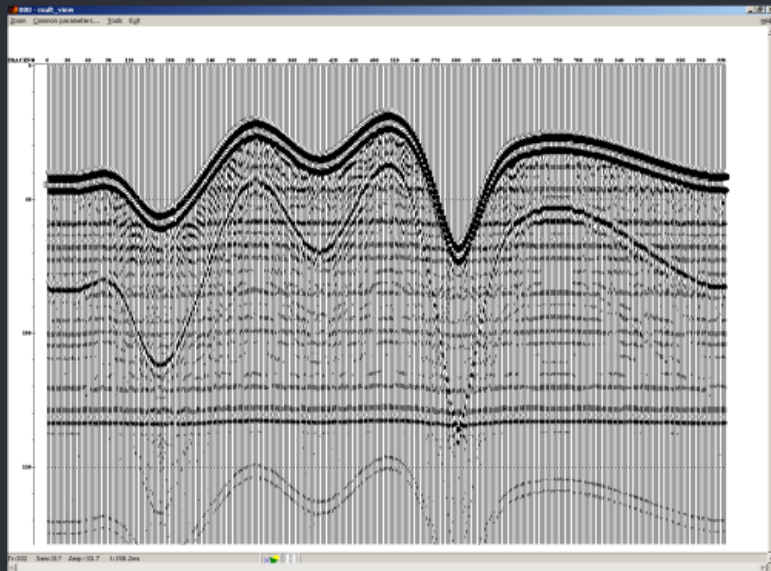


Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

1. An approximate model of multiples created from data itself:

- Autoconvolution of each trace – **model of ALL surface-related multiples**



Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

2. Model is subtracted from the data

Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

2. Model is subtracted from the data

- Our model is **inaccurate**, both in kinematics and in dynamics.

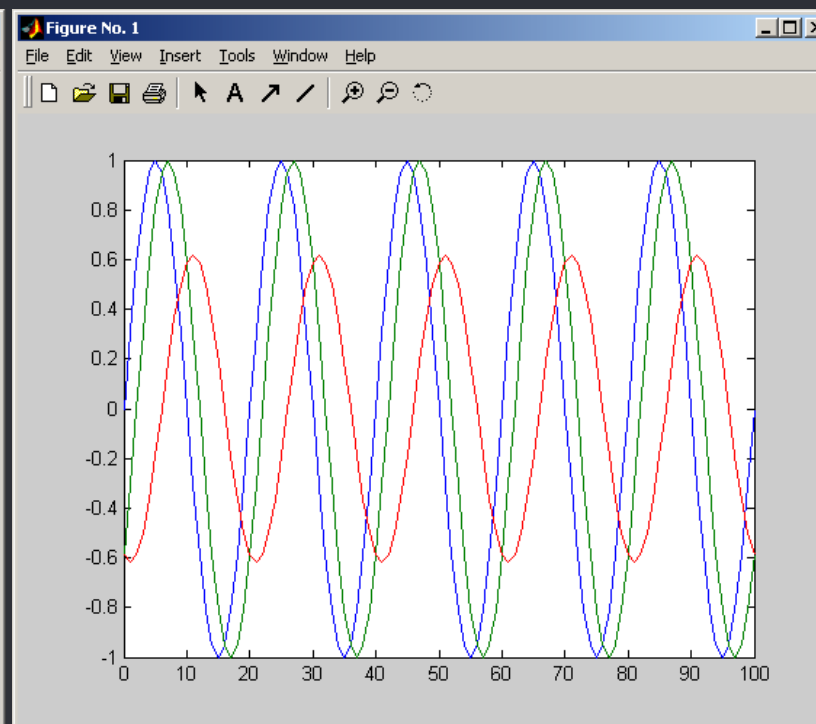
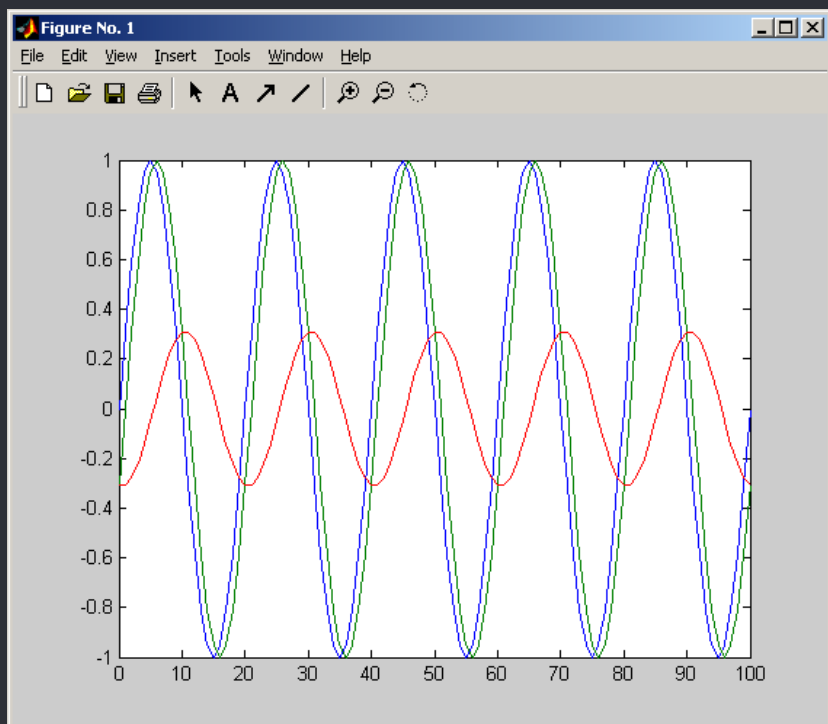
Can we simply subtract it?

Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

2. Model is subtracted from the data

- Subtracting with inaccurate kinematics:

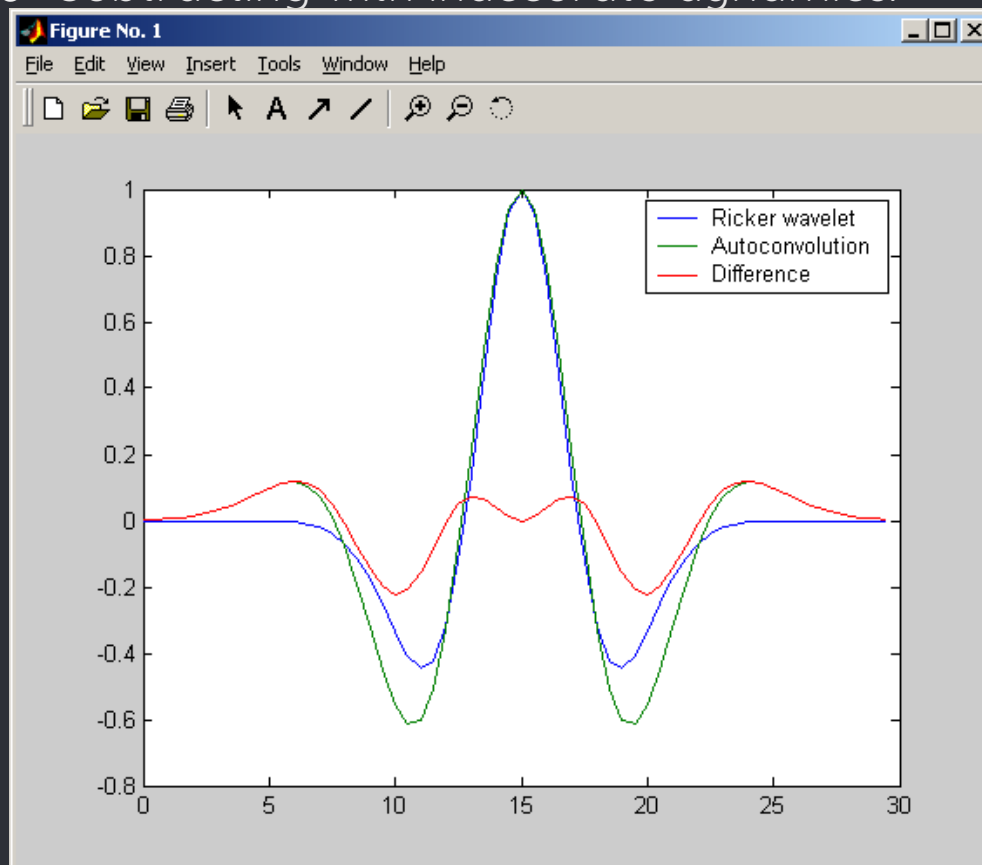


Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

2. Model is subtracted from the data

- Subtracting with inaccurate dynamics:



- Autoconvolution results in change of wavelet
- Adequate amplitude decay compensation is difficult to achieve

Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple theory

2. Model is to be subtracted from the data **ADAPTIVELY**

Zero-offset demultiple theory – Adaptive Subtraction Algorithm

The task of adaptive subtraction of the model of multiples from the initial wavefield is posed in the following way: for each trace it is required to define $f(x,t)$ minimizing in RMS sense the following functional:

$$J = \sum_t \left(Z(x,t) - \sum_{k=-M}^M f(x-k,t) * K(x-k,t) \right)$$

$f(x,t)$ are filters instead of coefficients, this description of distortions is essentially more general and includes waveform fluctuations caused by, particularly, frequency-dependent attenuation as well as amplitude variations

$Z(x,t)$

Current trace of the original wavefield

$K(x,t)$

Initial model of multiples for this trace

x

Current trace number

k

Index of a neighboring trace from the current trace (from $-M$ to M)

t

TWT time

This task can be solved with the help of standard techniques, particularly with the help of Wiggins-Robinson-Levinson algorithm for multi-channel filters.

As a result of this step:

For each trace X we calculate a filter $f(x,t)$

This filter minimizes everything that is in common between the original trace and the model of multiples at traces within the neighborhood $(x-M, x+M)$

STRONG SUPPRESSION OF MULTIPLES

As a result of this step:

For each trace X we calculate a filter $f(x,t)$

This filter minimizes everything that is in common between the original trace and the model of multiples at traces within the neighborhood $(x-M, x+M)$

STRONG SUPPRESSION OF MULTIPLES

Final step:

We assume that the filters $f(x,t)$ shall not be changing too abruptly from trace to trace.

So we average the filters over N neighboring traces:

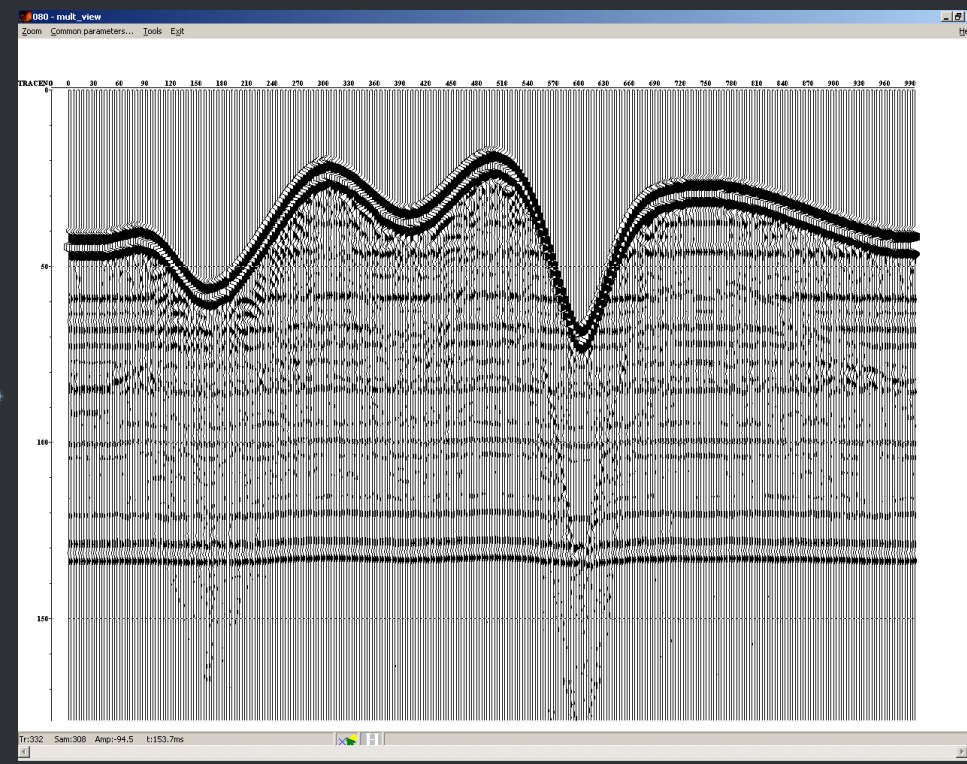
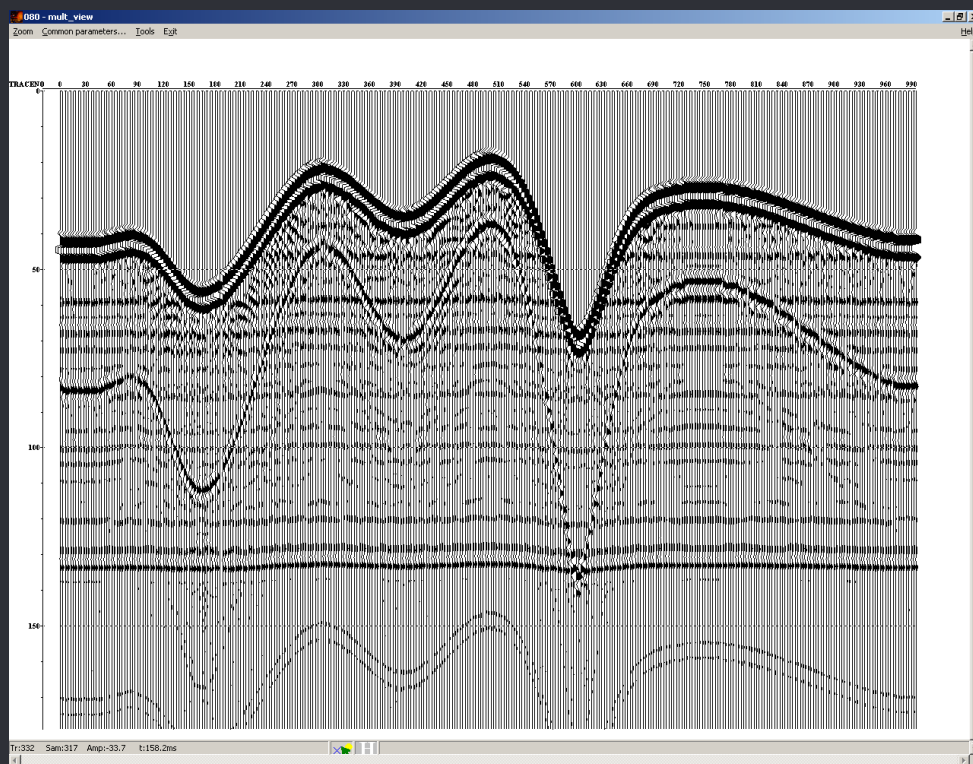
$$f_{final}(x,t) = \frac{\sum_{j=-N/2}^{N/2} f(x-j,t)}{N+1}$$

THIS STEP HELPS PRESERVING PRIMARIES

Multiple Elimination: Zero-Offset Demultiple

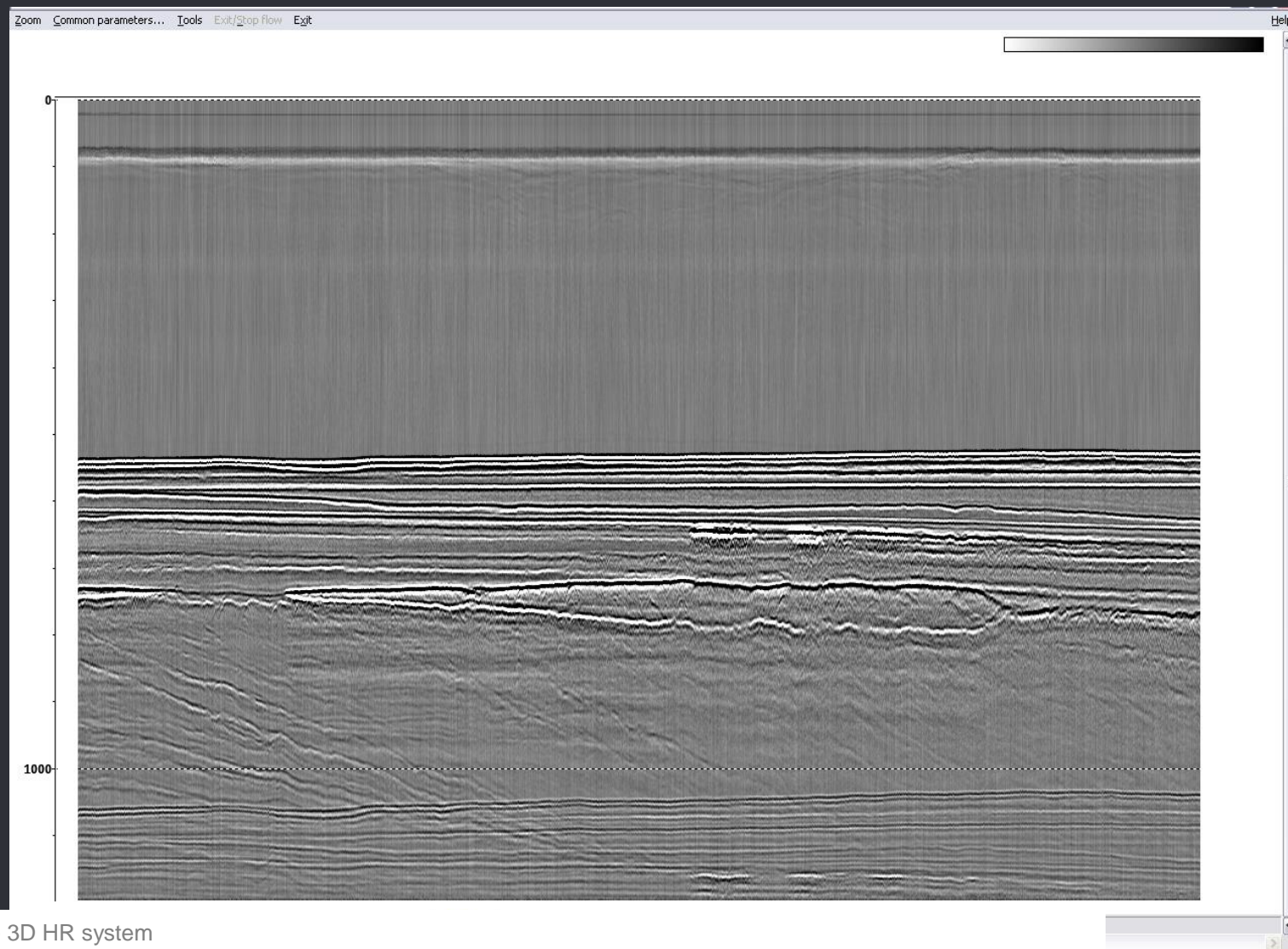
Zero-offset demultiple theory

2. Model is subtracted from the data **ADAPTIVELY**



Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple examples: before



Multiple Elimination: Zero-Offset Demultiple

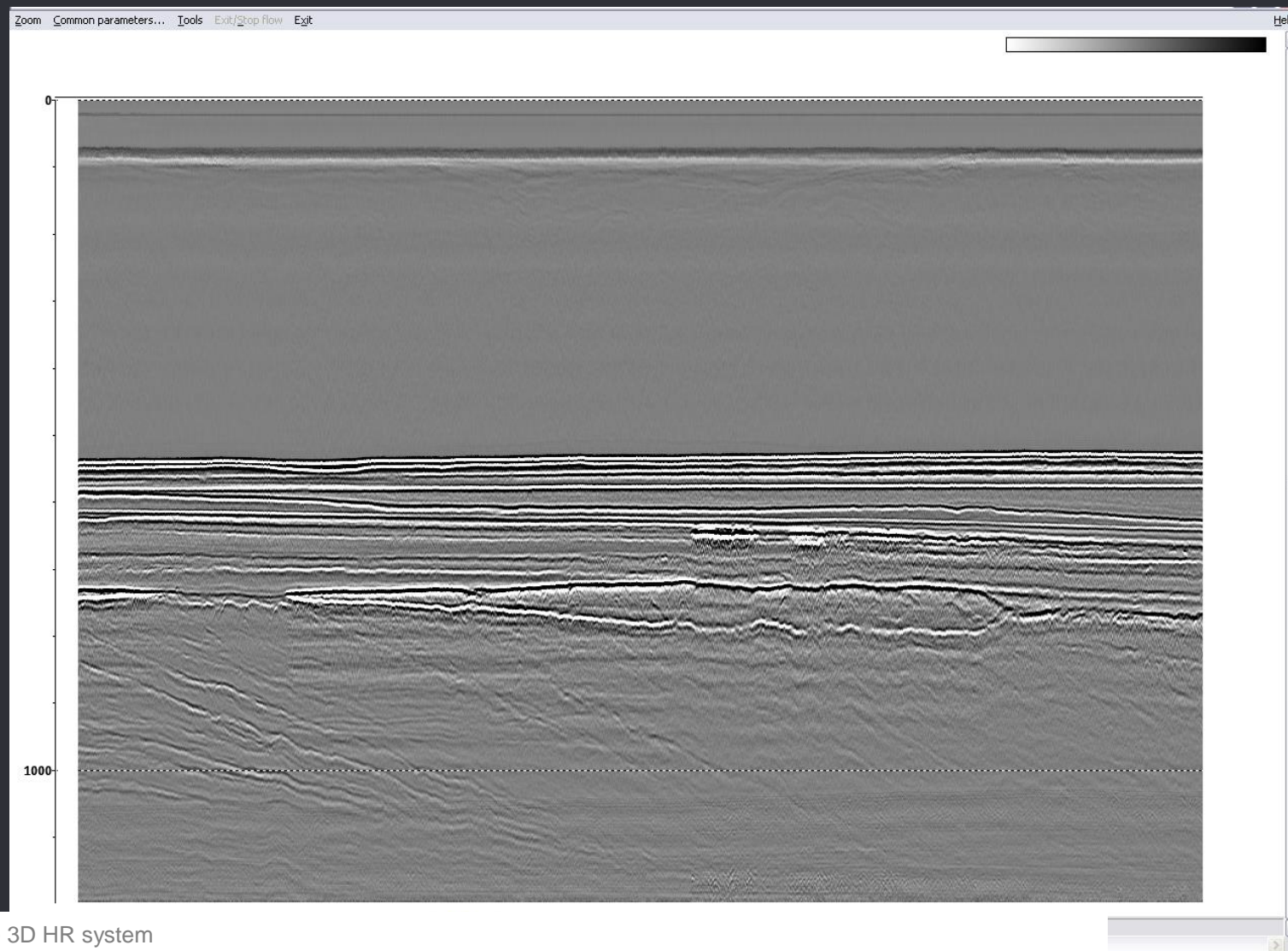
Zero-offset demultiple examples: multiple model



Data acquired by P-Cable 3D HR system

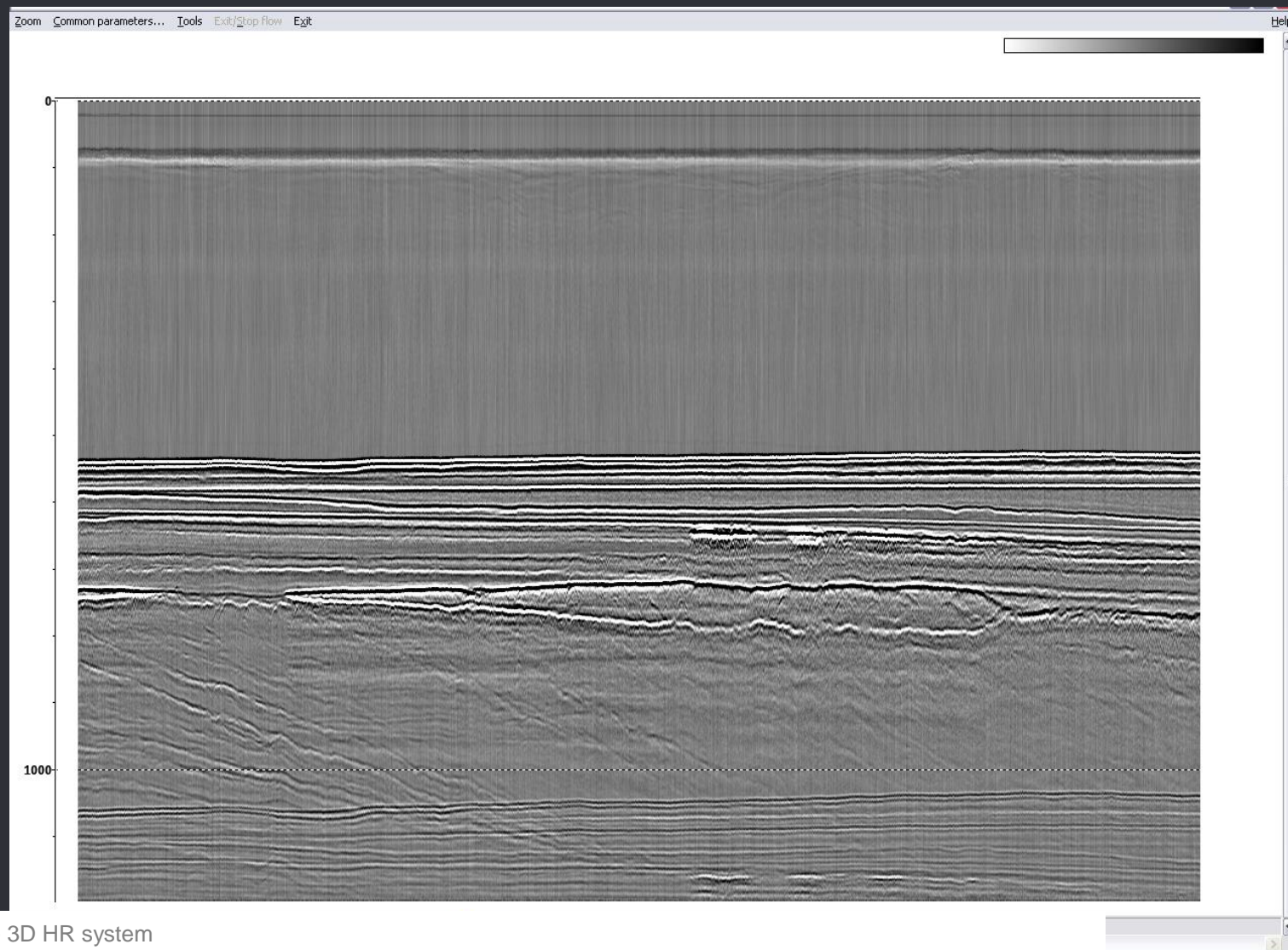
Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple examples: subtraction result



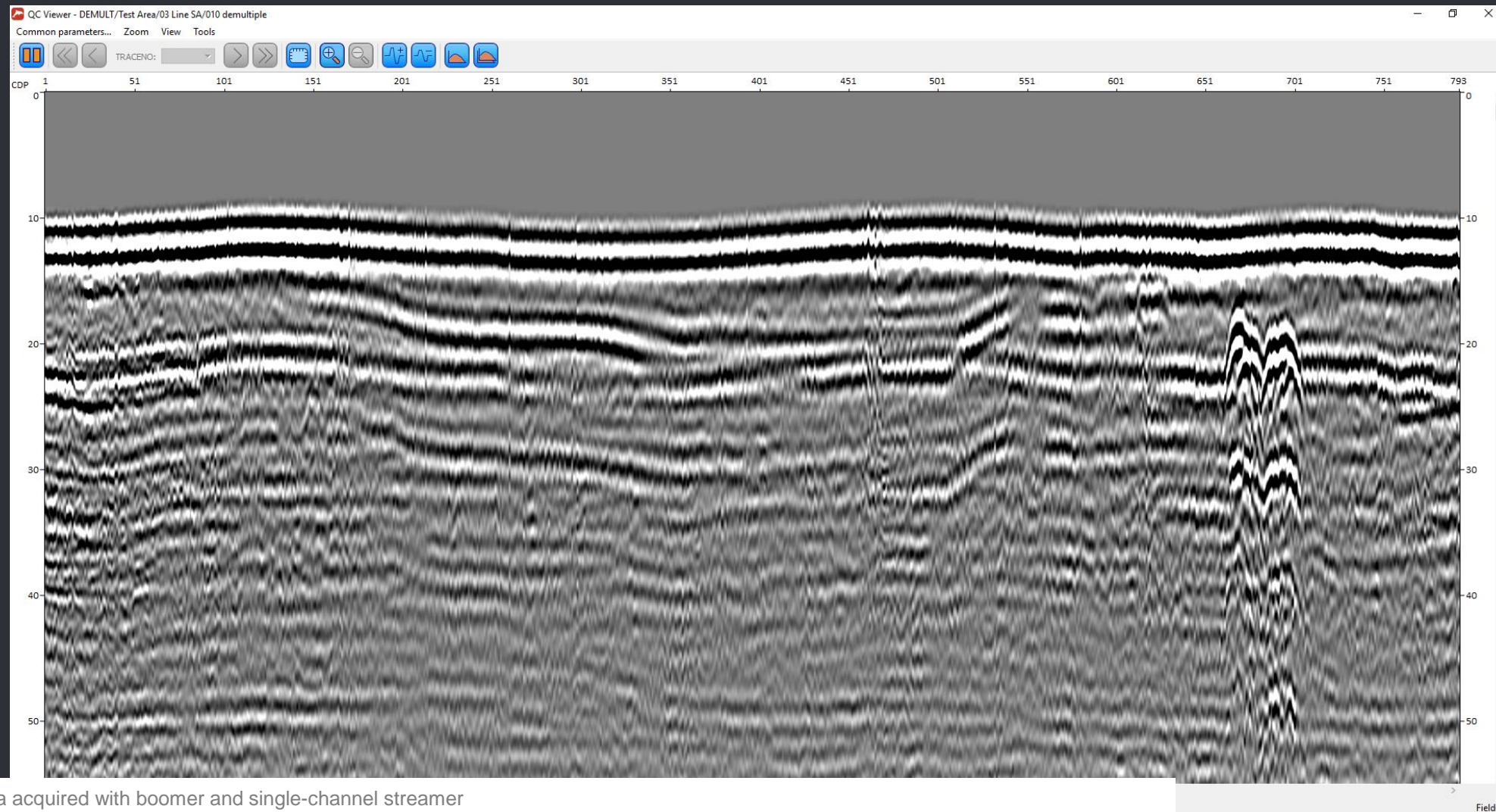
Multiple Elimination: Zero-Offset Demultiple

Zero-offset demultiple examples: before



Multiple Elimination: Zero-Offset Demultiple

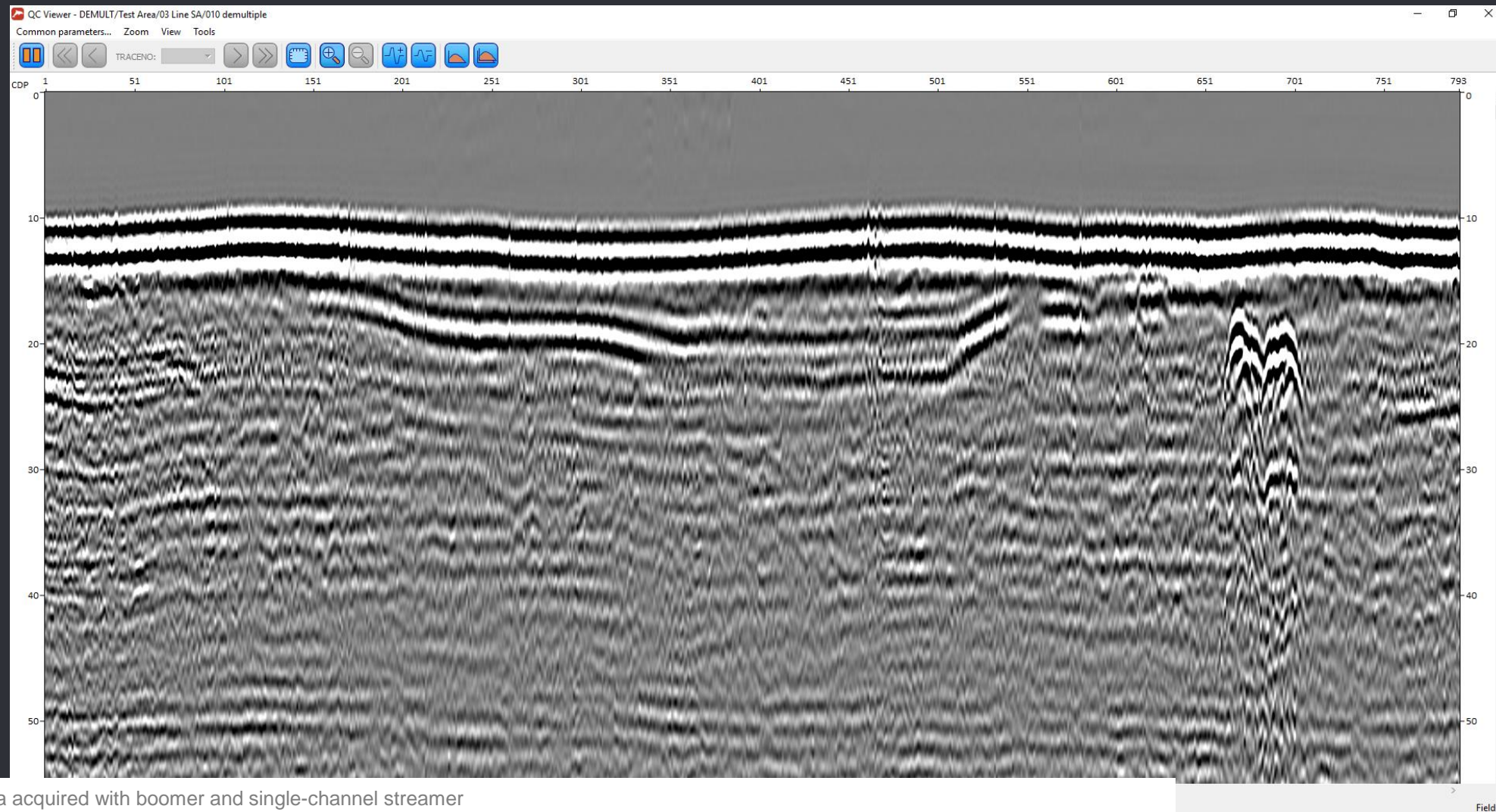
Single channel boomer data – before multiple elimination



Data acquired with boomer and single-channel streamer

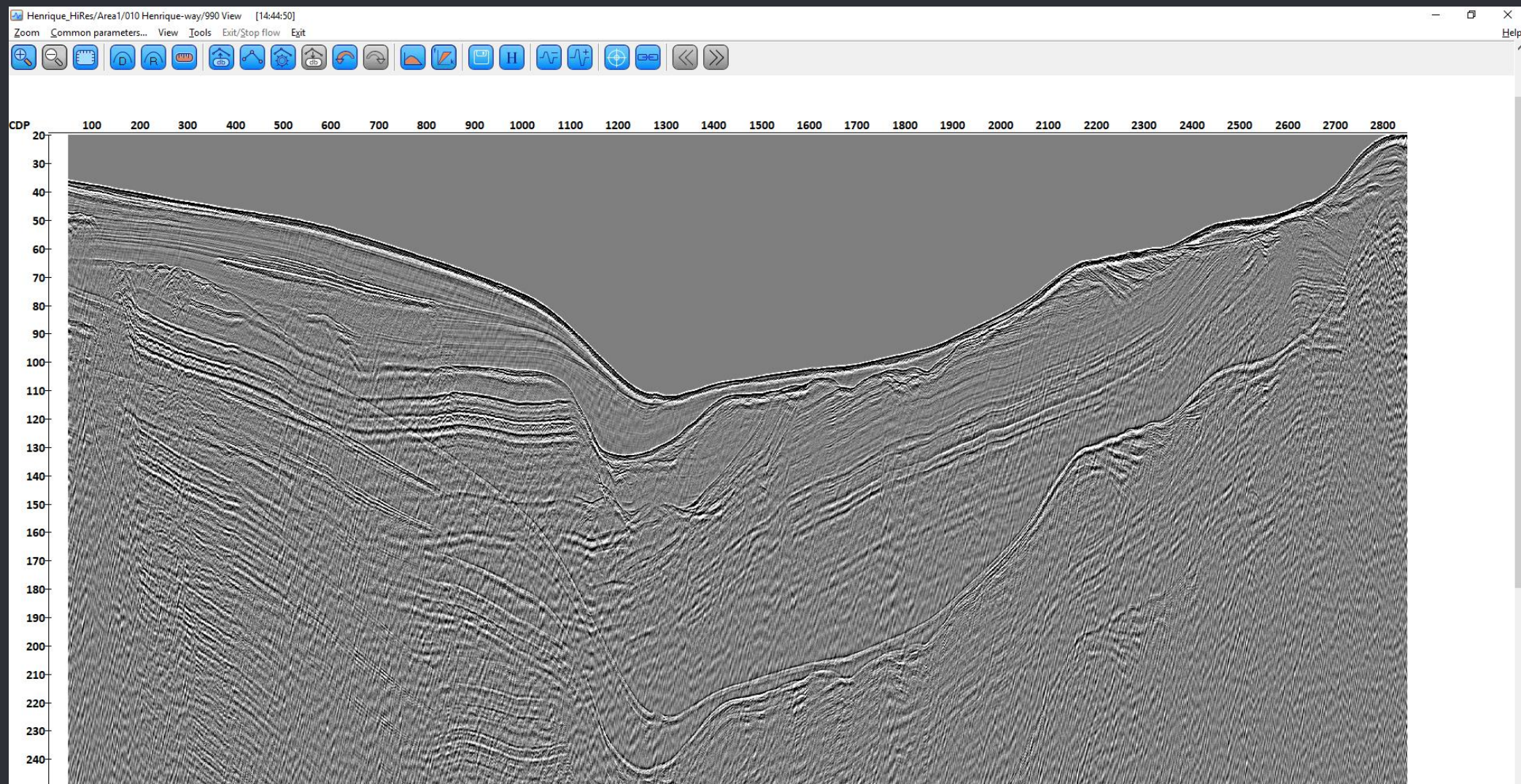
Multiple Elimination: Zero-Offset Demultiple

Single channel boomer data – after zero-offset demultiple



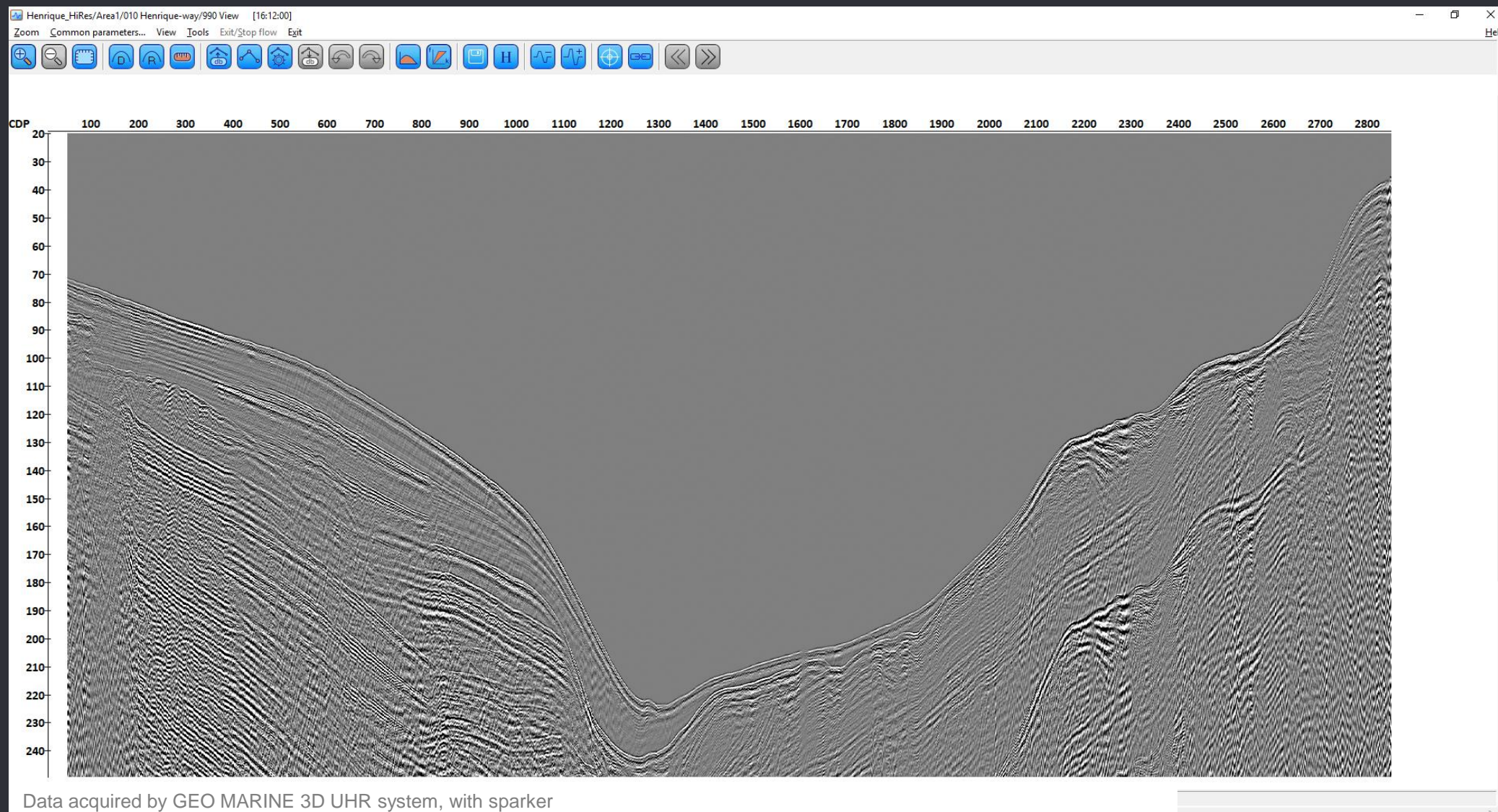
Multiple Elimination: Zero-Offset Demultiple

Before (stack of 16 channels)



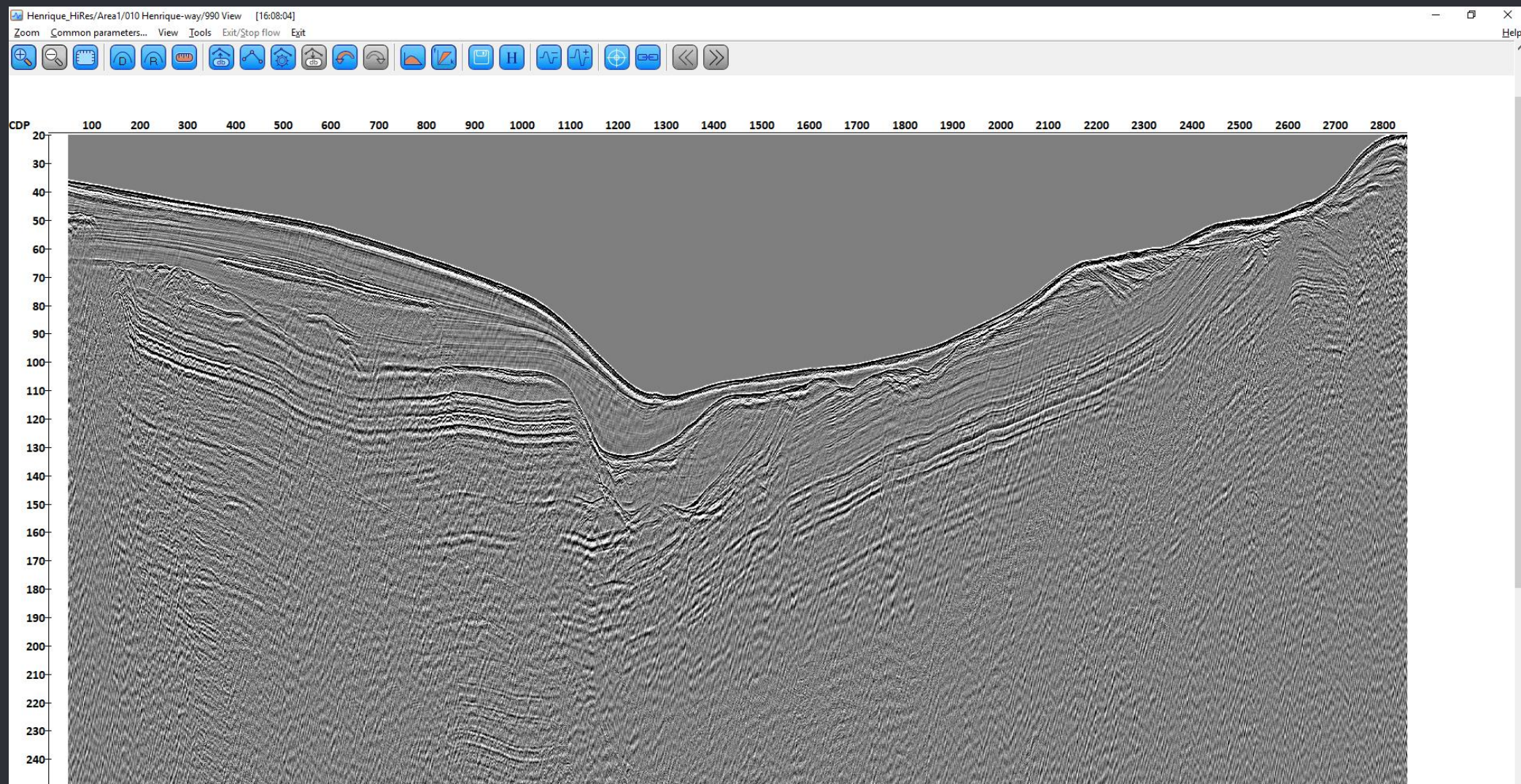
Multiple Elimination: Zero-Offset Demultiple

Model of Multiples (autoconvolution – post-stack)



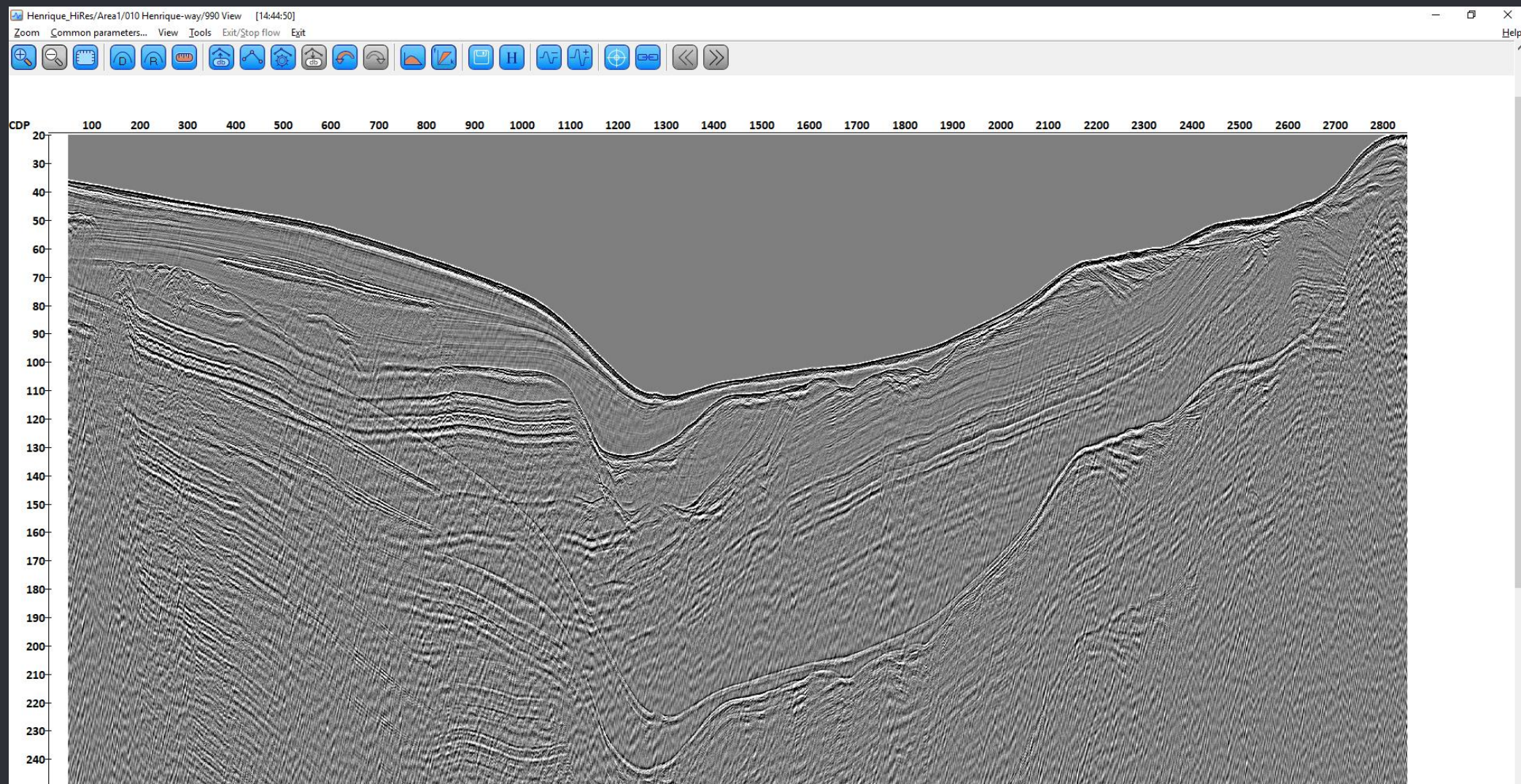
Multiple Elimination: Zero-Offset Demultiple

Subtraction Resultt (post-stack)



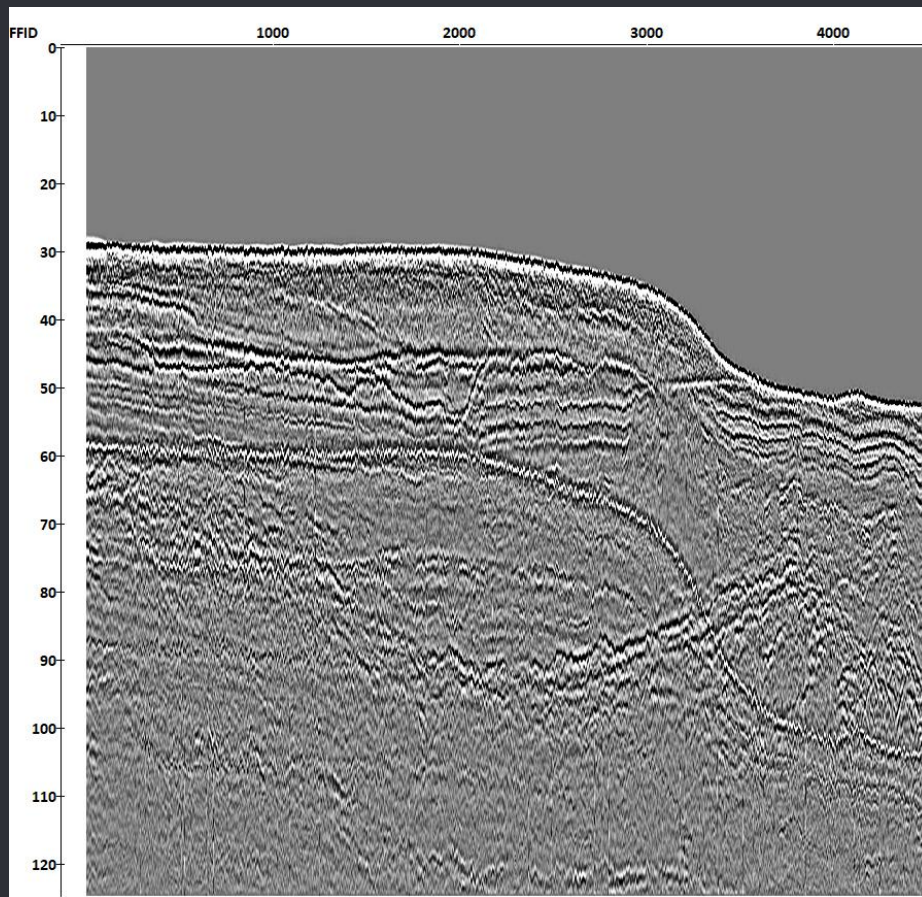
Multiple Elimination: Zero-Offset Demultiple

Before (stack of 16 channels)



Multiple Elimination: Zero-Offset Demultiple

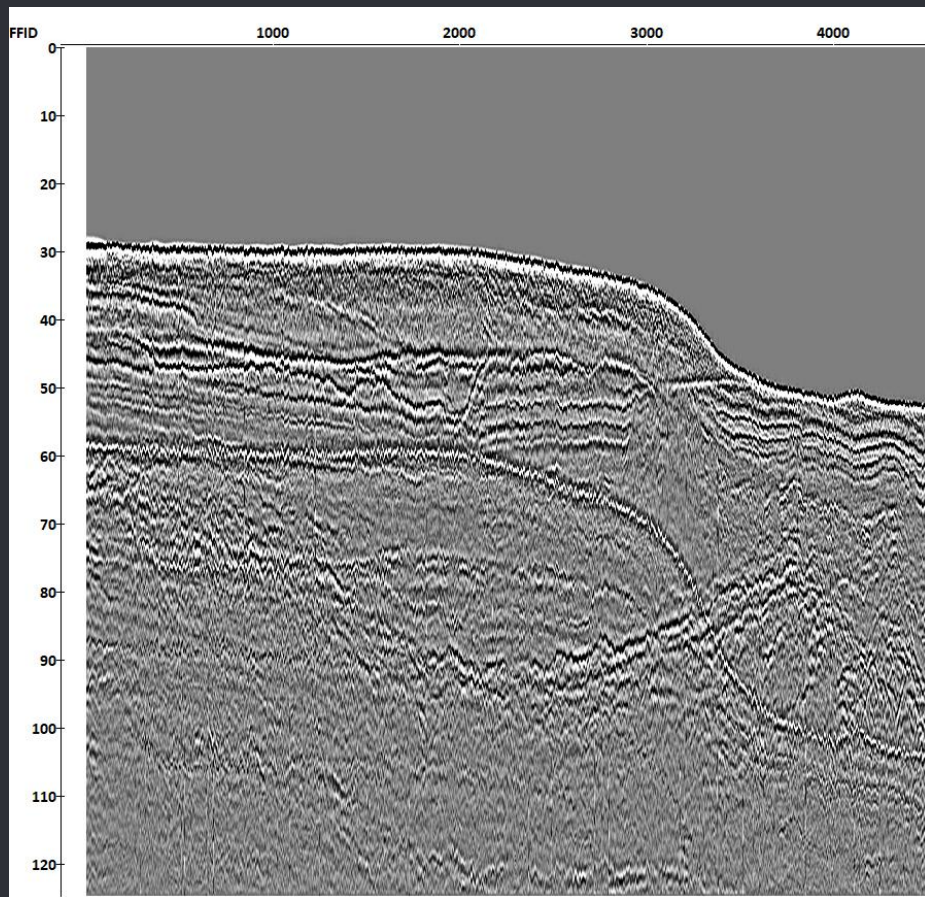
Data disturbed by sea swelling



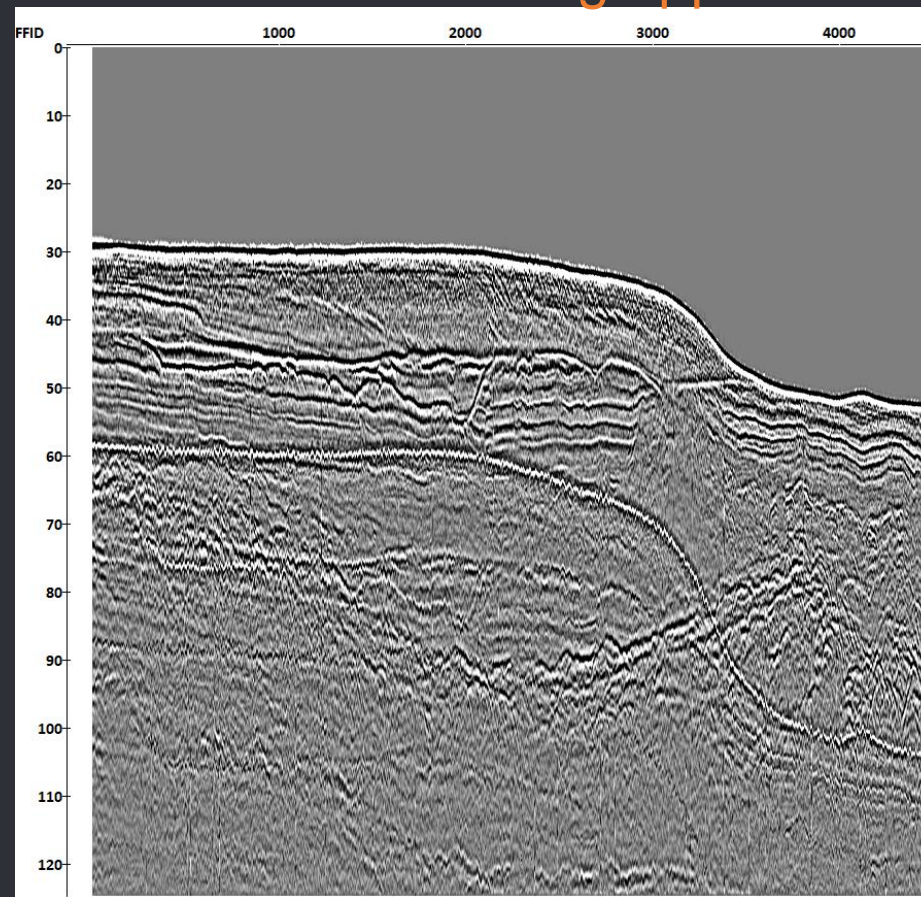
Data acquired with Bubble Pulser (by Falmouth Scientific) and single-channel streamer

Multiple Elimination: Zero-Offset Demultiple

Data disturbed by sea swelling



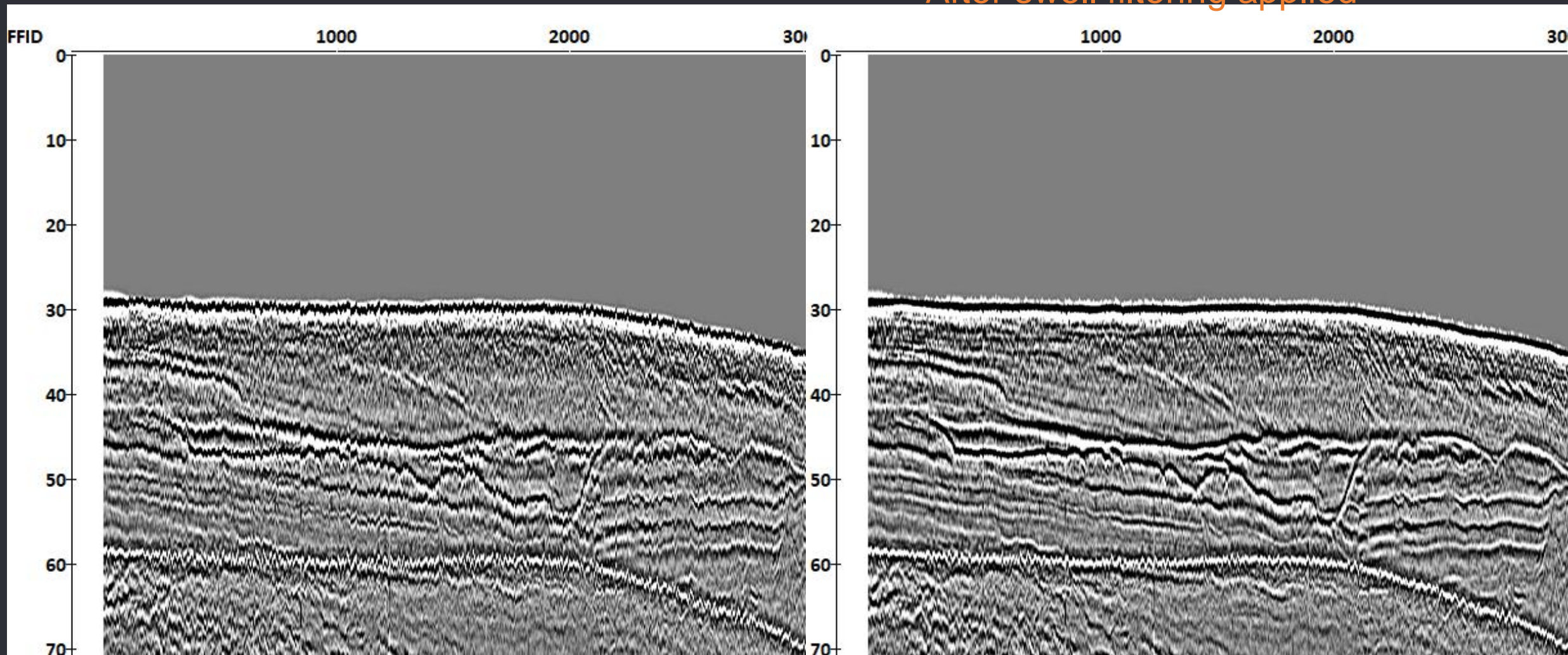
After swell filtering applied



Multiple Elimination: Zero-Offset Demultiple

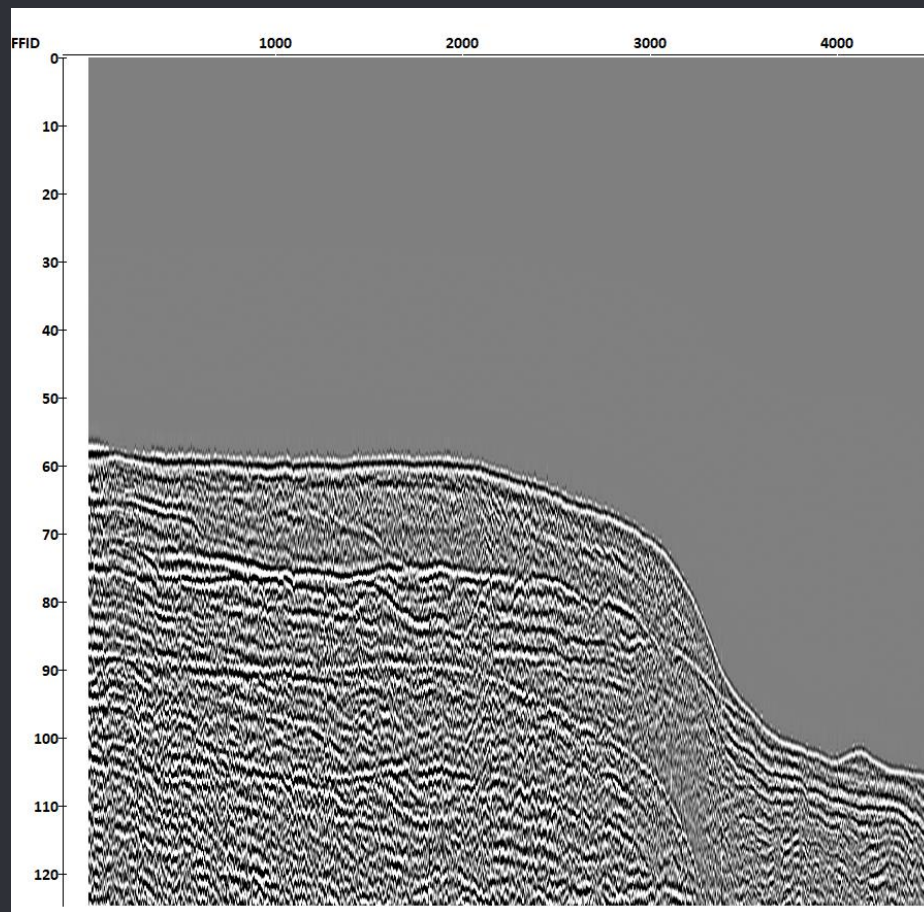
Data disturbed by sea swelling

After swell filtering applied



Multiple Elimination: Zero-Offset Demultiple

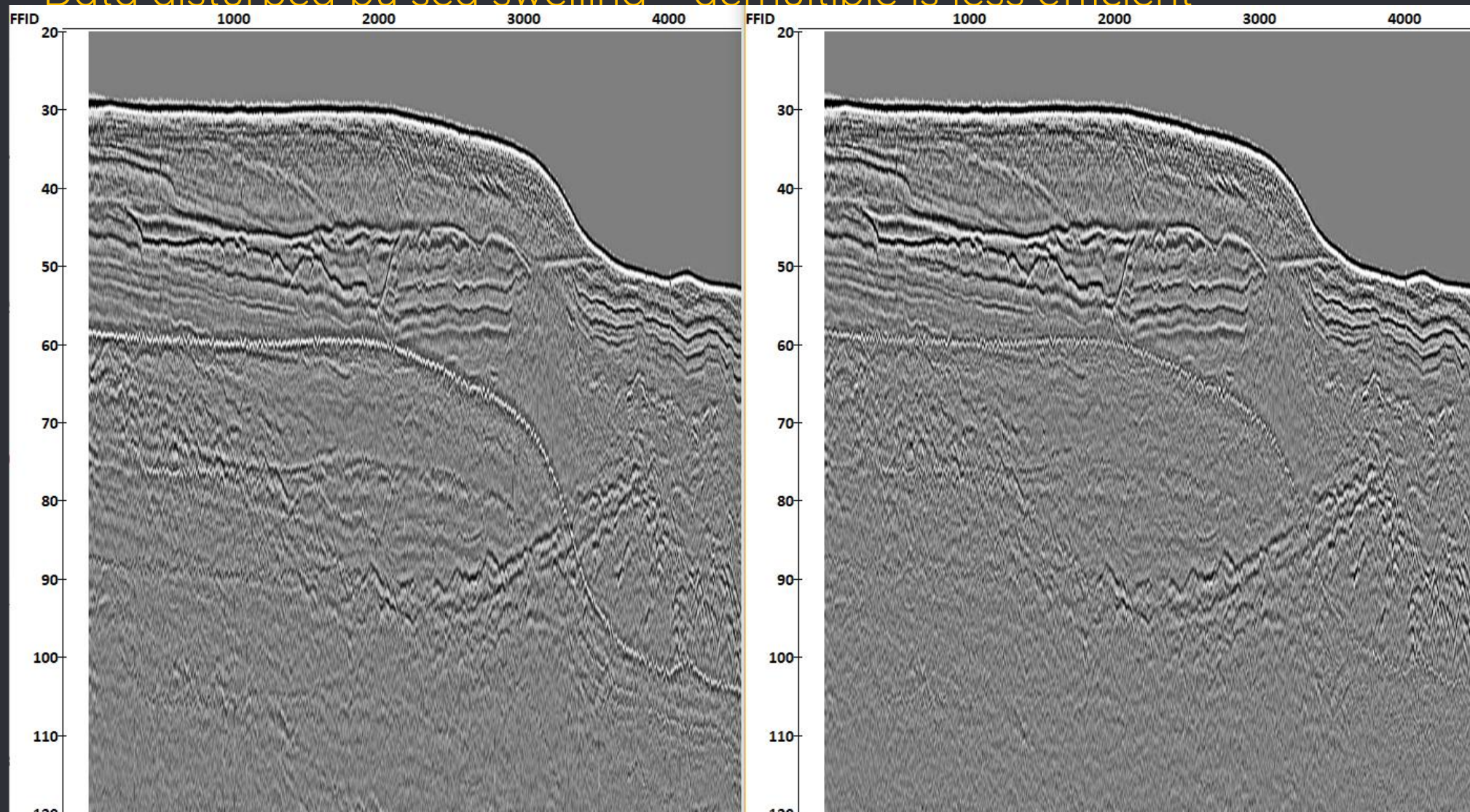
Data disturbed by sea swelling – model of multiples



Data acquired with Bubble Pulser (by Falmouth Scientific) and single-channel streamer

Multiple Elimination: Zero-Offset Demultiple

Data disturbed by sea swelling – demultiple is less efficient



- Zero-Offset Demultiple technique based on approximate modeling of multiples followed by adaptive subtraction, was implemented and tested.
- It was shown that the method can be very efficient for near-offset HR/UHR marine seismic data acquired with different types of sources.
- Greater offsets and sea swelling reduce the efficiency of the algorithm

Thank you for attention!

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