

# **Least-squares reverse-time migration with salt-dimming**

**SEP Sponsor Meeting  
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SEP 149**

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**Stanford Exploration Project**

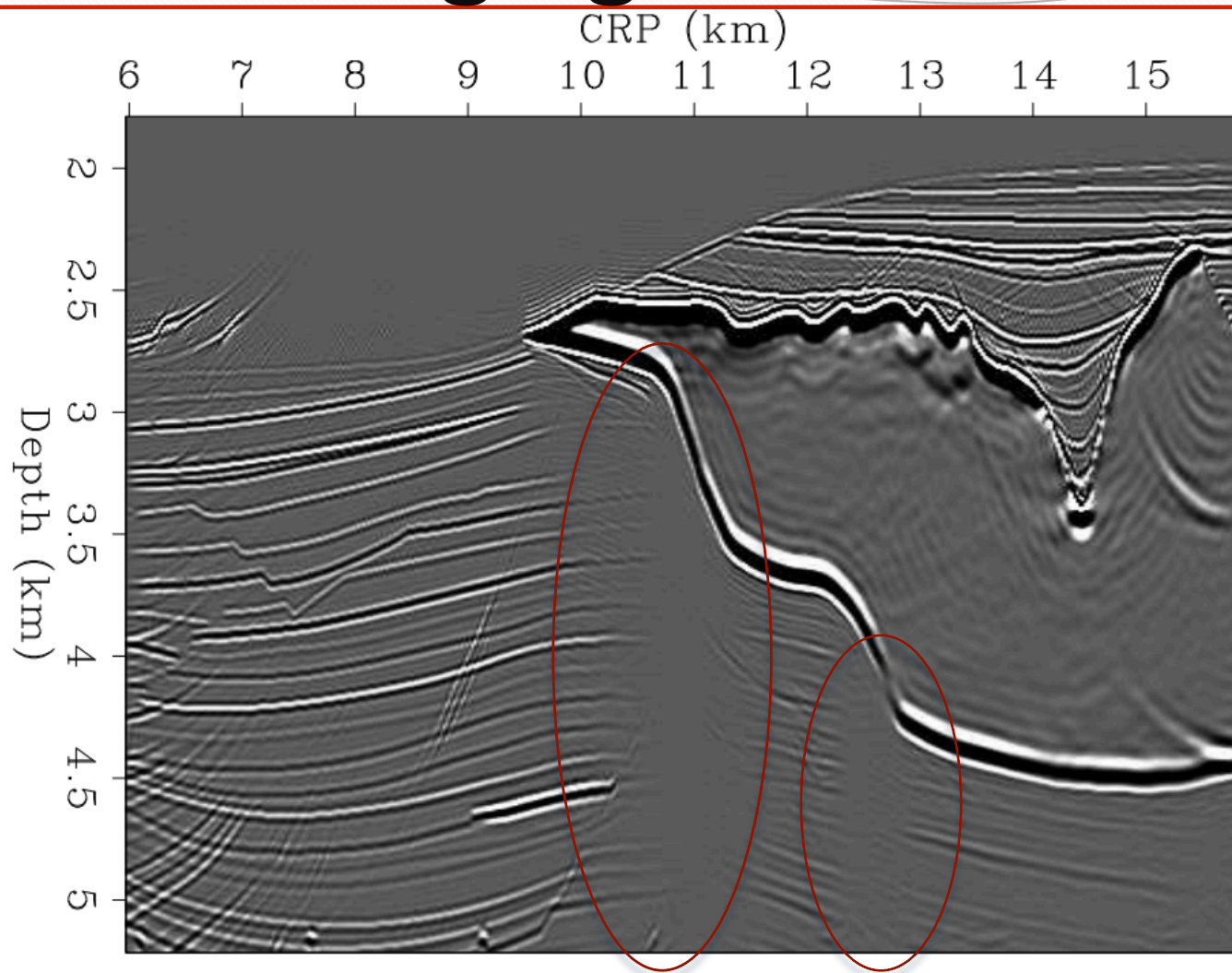
# Overview

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- **Introduction**
- **Background data**
- **Theory and examples**
- **Discussion**
- **Conclusion**



# Subsalt Imaging



Clapp (2005)

# What is least-squares migration (LSM) ?

$$\mathbf{d}^{\text{mod}} = \mathbf{L} \mathbf{m}$$

**L** Forward modeling operator

**d<sup>mod</sup>** modeled data

**d** field data

**m** Image model

Obtain the best image by minimizing the difference between the modeled data and the field data

$$S(\mathbf{m}) = || \mathbf{L} \mathbf{m} - \mathbf{d} ||^2$$





**Each LSM is associated with a type of migration**

**Migration operator**

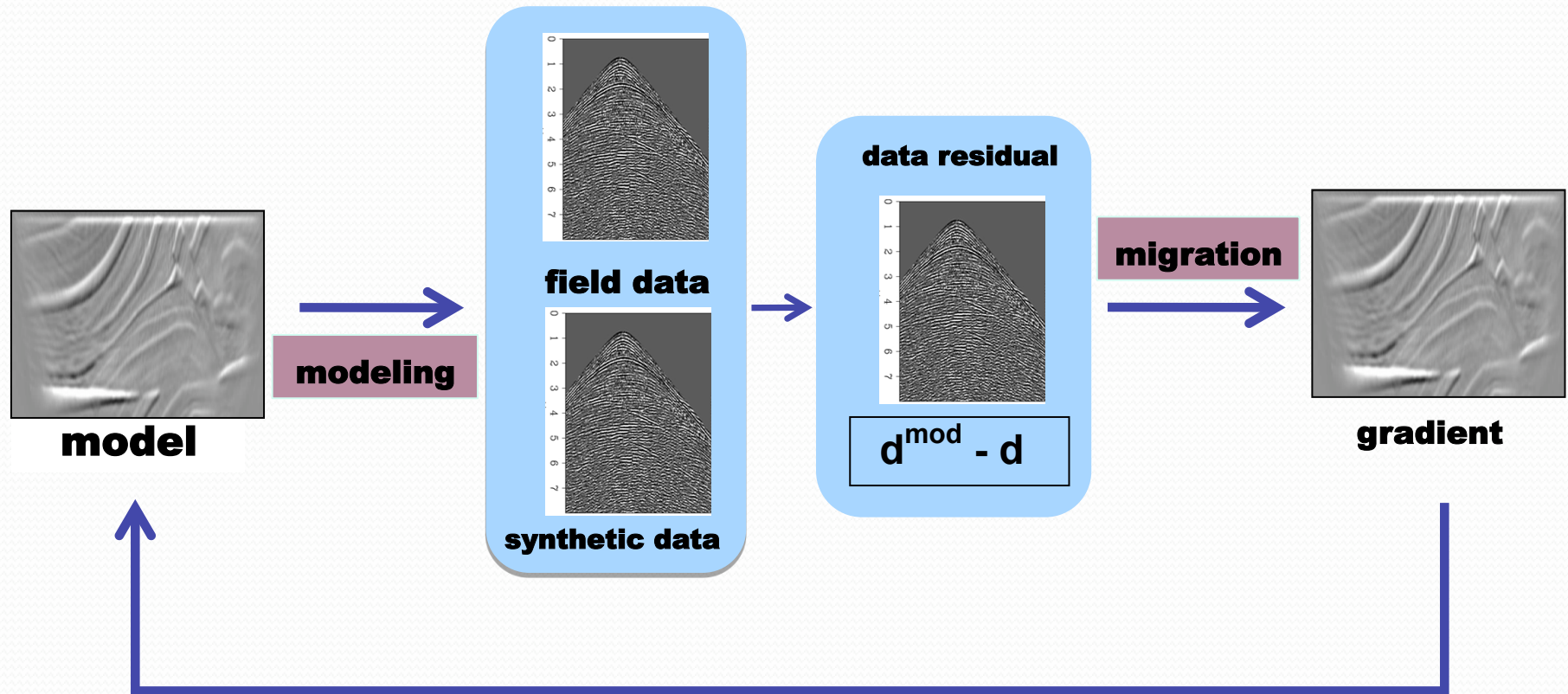
**=**

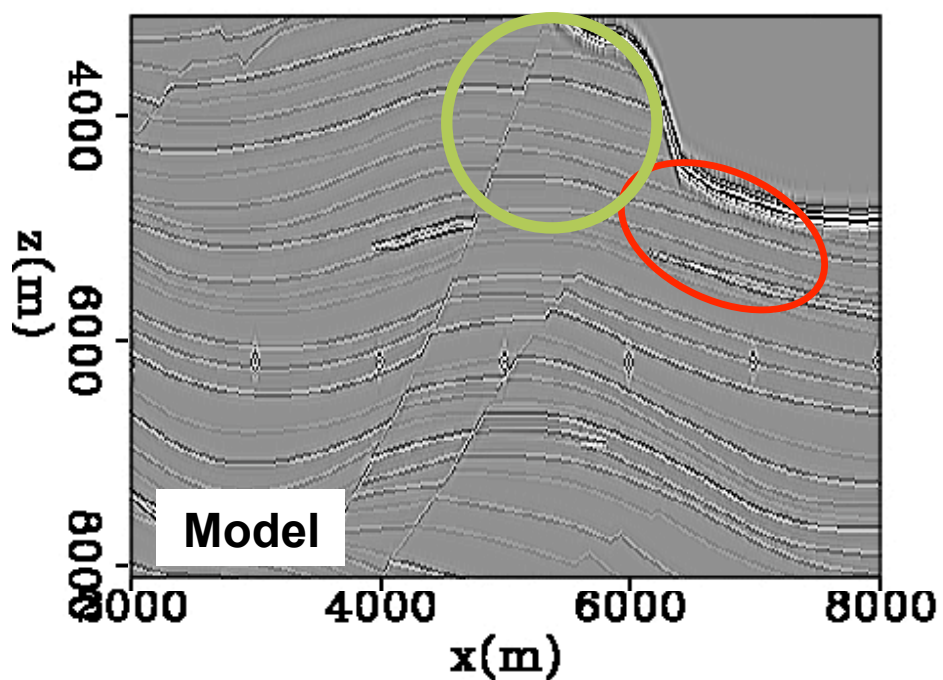
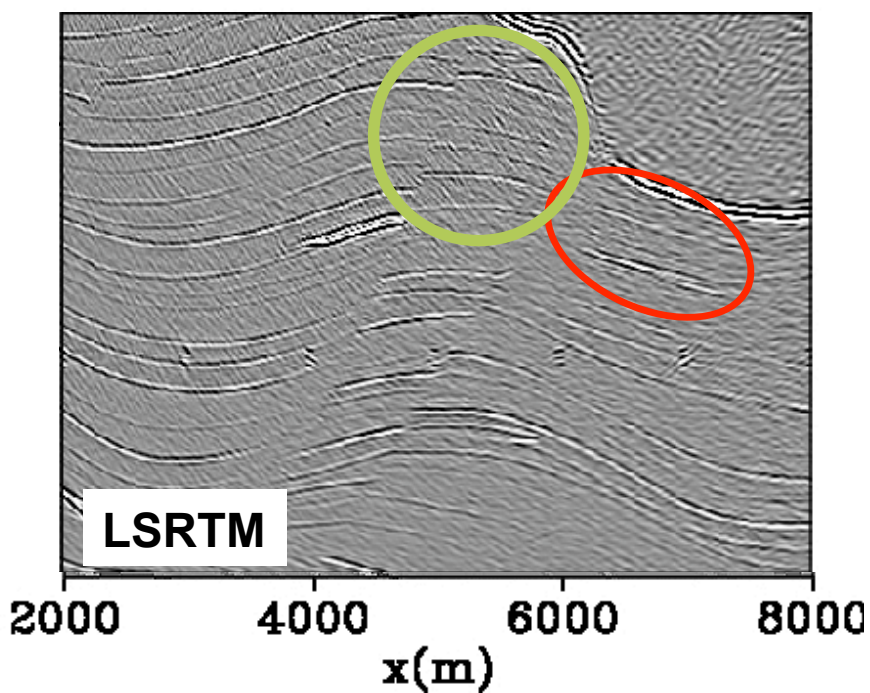
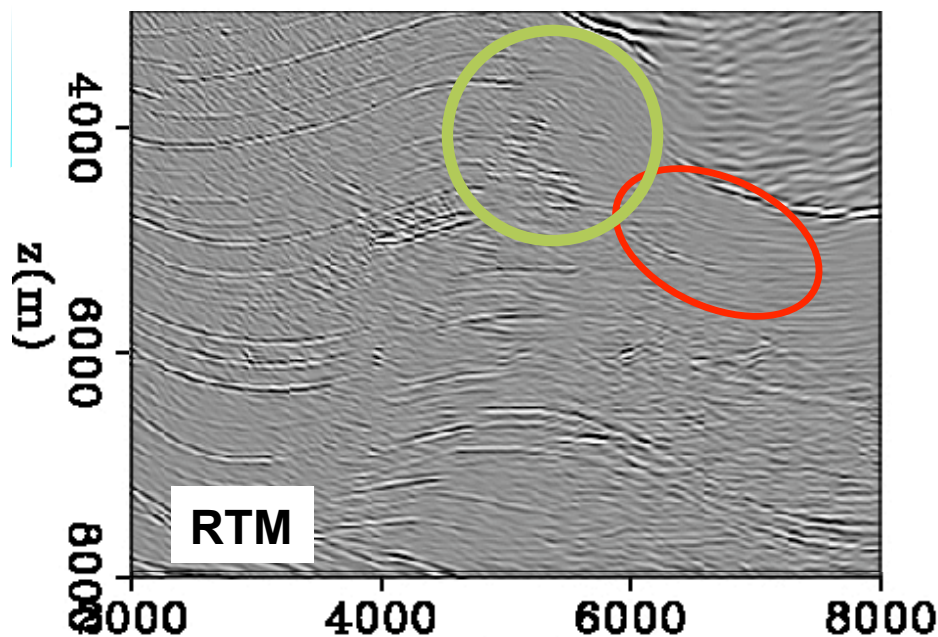
**Adjoint of the Forward modeling  
operator**



# LSM Workflow

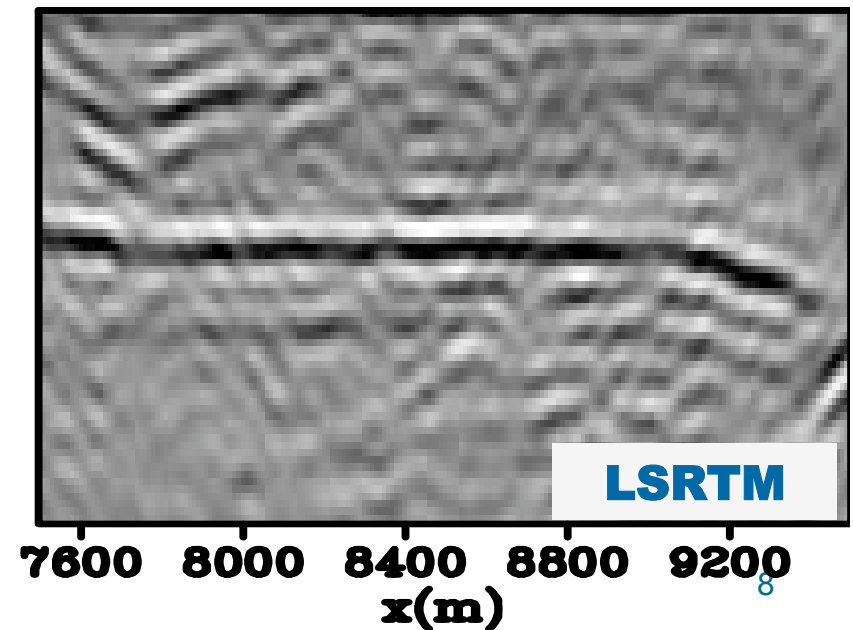
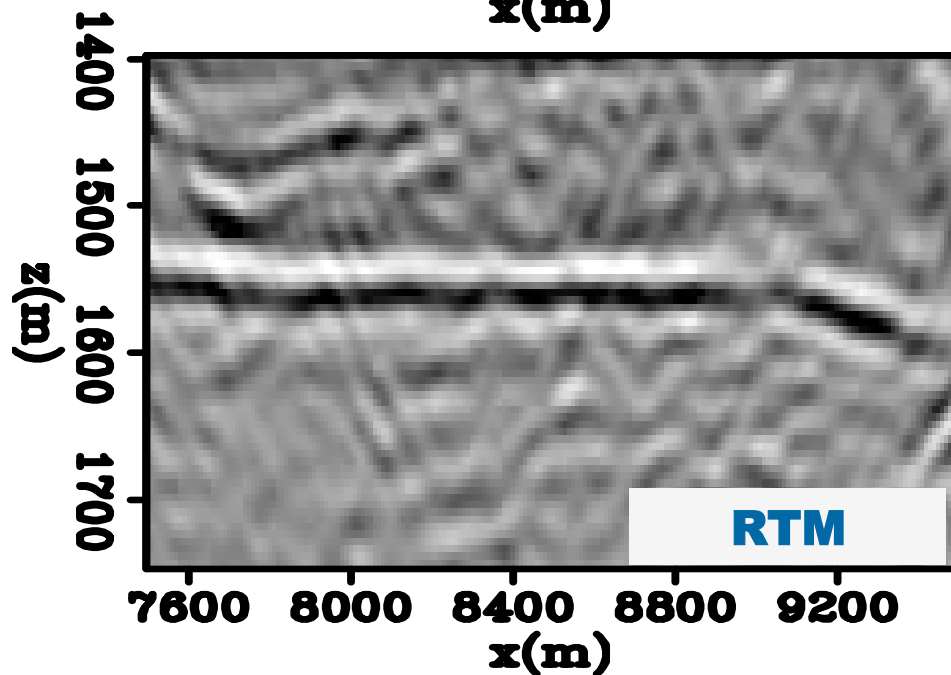
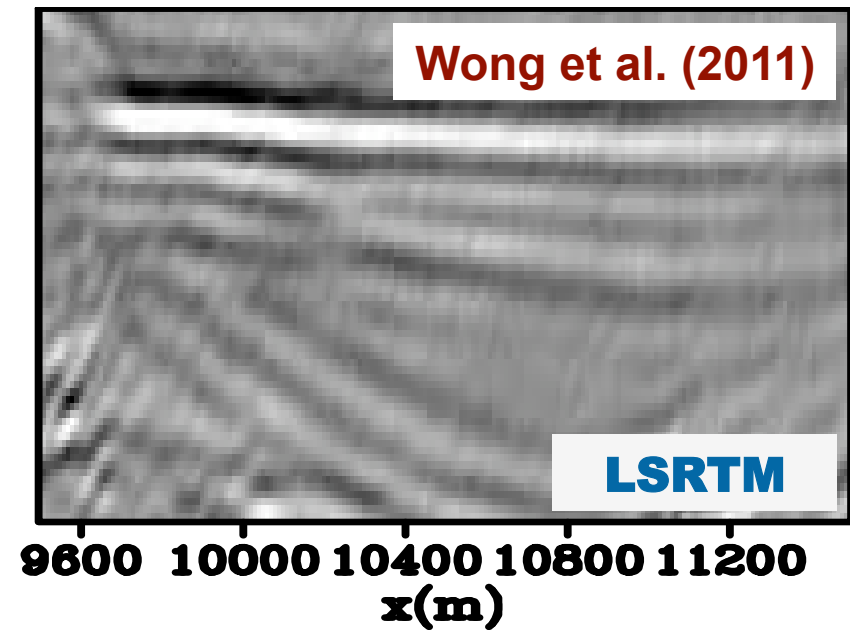
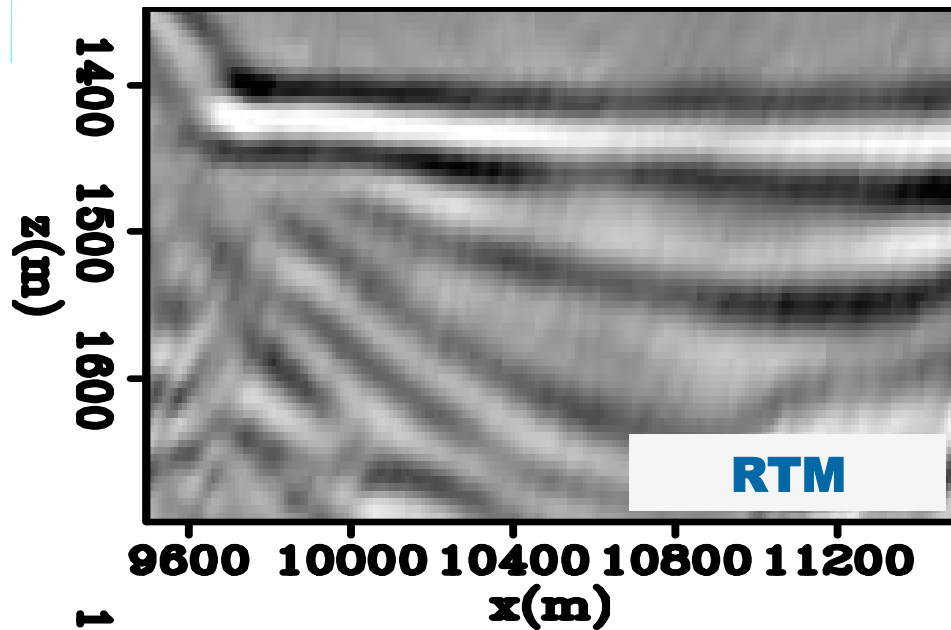
- Iterative inversion by conjugate gradient





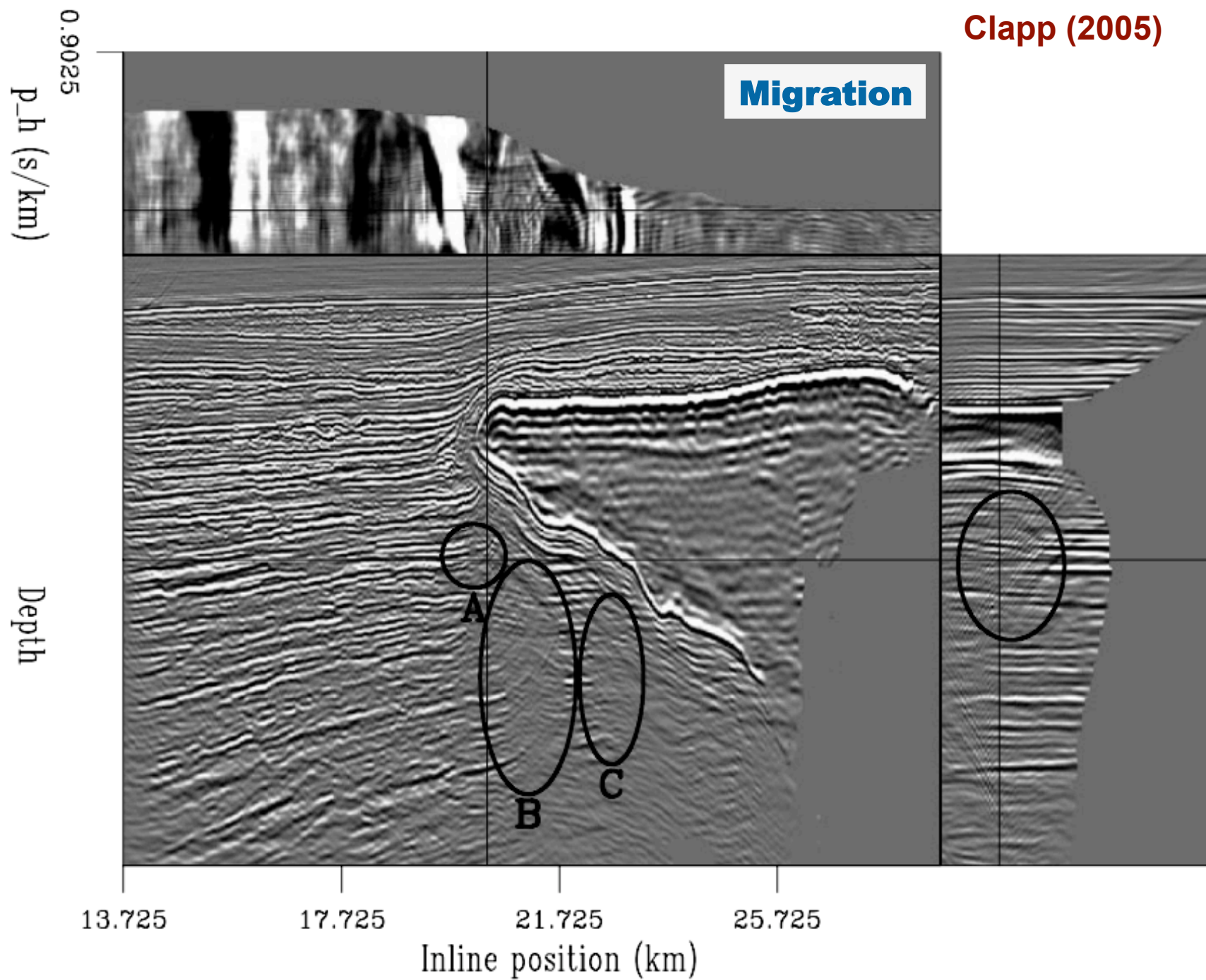
1. Better relative amplitude information
2. Fewer migration artifacts
3. Less acquisition footprints

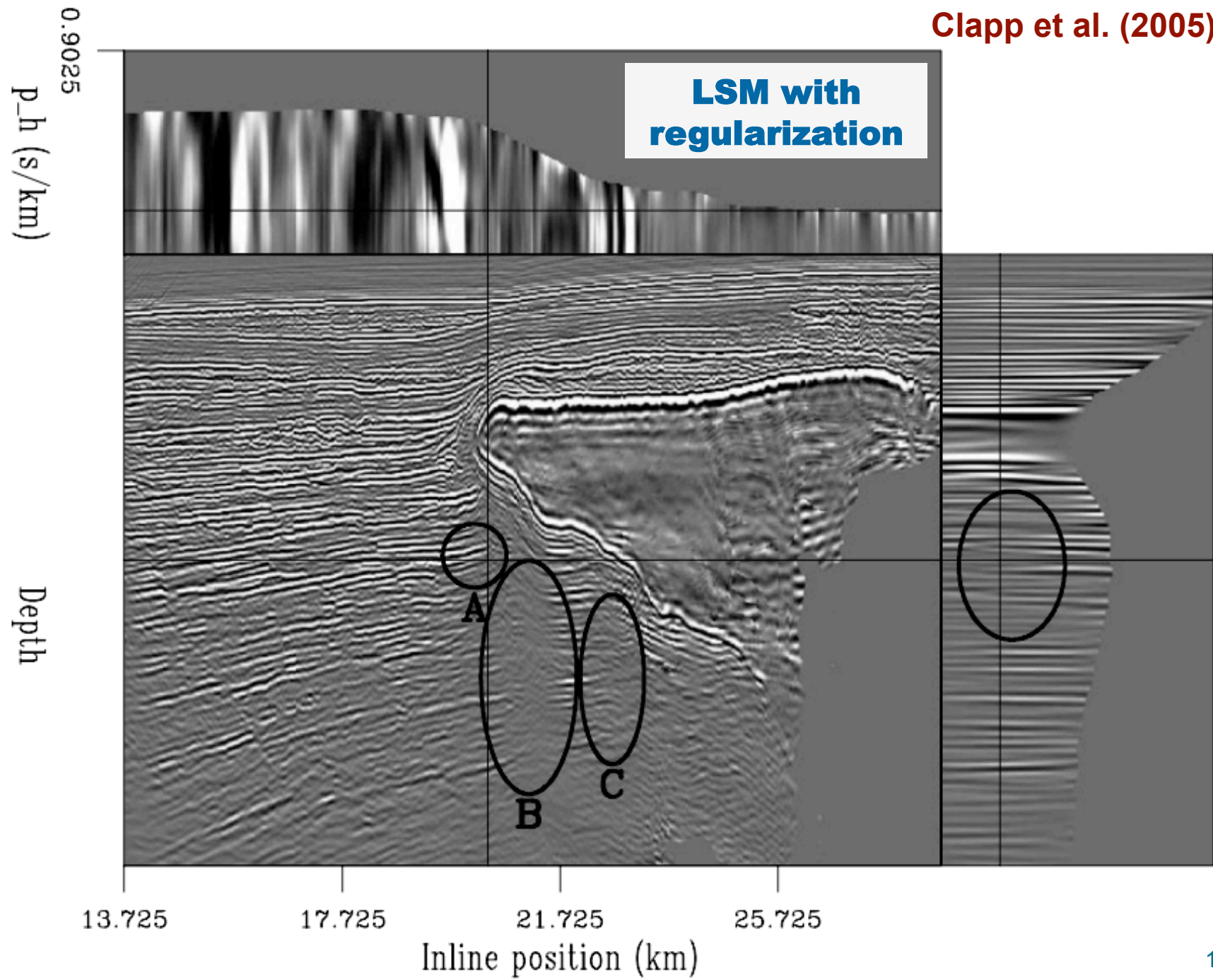
# Comparison of RTM and LSRTM





Clapp (2005)





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# The issue with background data

- Applying LSM in regions like Gulf of Mexico
  - Sharp velocity contrast
  - Background data term needs to be subtracted from observed data
  - Affects LSM result, especially for deep reflectors





# LSM Parameterization

Linearizing the wave equation with respect to model  $m$

$$s(\mathbf{X}) = s_o(\mathbf{X}) + \Delta s(\mathbf{X})$$

$$m(\mathbf{X}) = \Delta s(\mathbf{X}) s_o(\mathbf{X})$$

$s(\mathbf{X})$  true slowness

$s_o(\mathbf{X})$  migration slowness

$m(\mathbf{X})$  model

$\Delta s(\mathbf{X})$  slowness perturbation



# Forward modeling in LSM

The forward modeling operator is linearized with respect to  $m(\mathbf{x})$

$$\begin{aligned} d^{mod} &= F(s_o^2 + m) \\ &\approx F(s_o^2) + \mathbf{L}m \end{aligned}$$

**L**

Linearized forward modeling operator

$d^{mod}$

Synthetic data

$F(s_o^2)$

Background data

# LSM objective function

The forward modeling operator is linearized with respect to  $m(\mathbf{x})$

$$\begin{aligned} S(\mathbf{m}) &= \|d^{mod} - d^{obs}\|^2 \\ &= \|\mathbf{L}m - (d^{obs} - F(s_o^2))\|^2 \end{aligned}$$

**L**

Linearized forward modeling operator

$d^{mod}$

Synthetic data

$d^{obs}$

Observed data

$F(s_o^2)$

Background data

# LSM objective function

The forward modeling operator is linearized with respect to  $m(\mathbf{x})$

$$\begin{aligned} S(\mathbf{m}) &= \|d^{mod} - d^{obs}\|^2 \\ &= \|\mathbf{L}m - (d^{obs} - F(s_o^2))\|^2 \end{aligned}$$

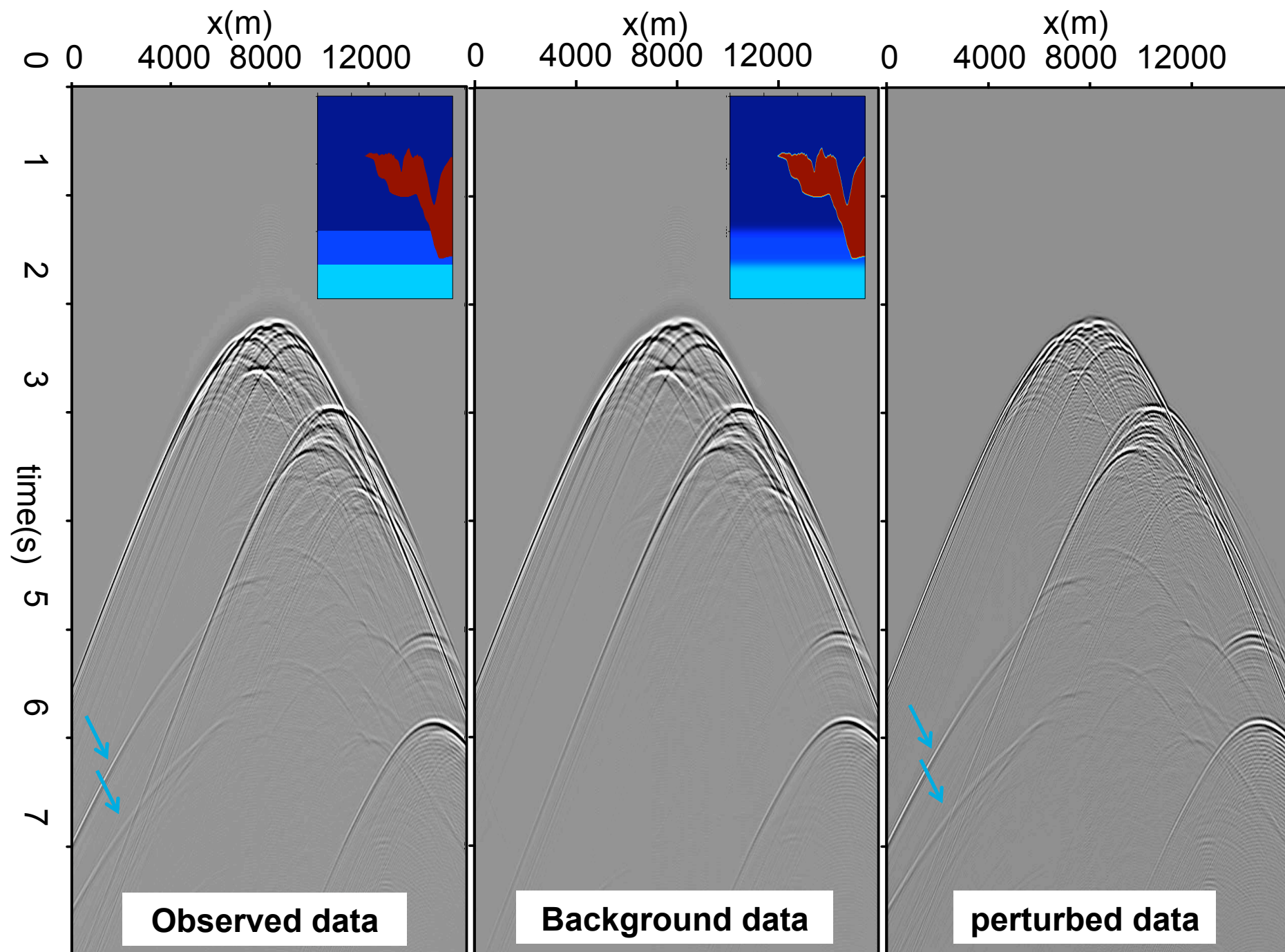
Perturbed data

**L** Linearized forward modeling operator

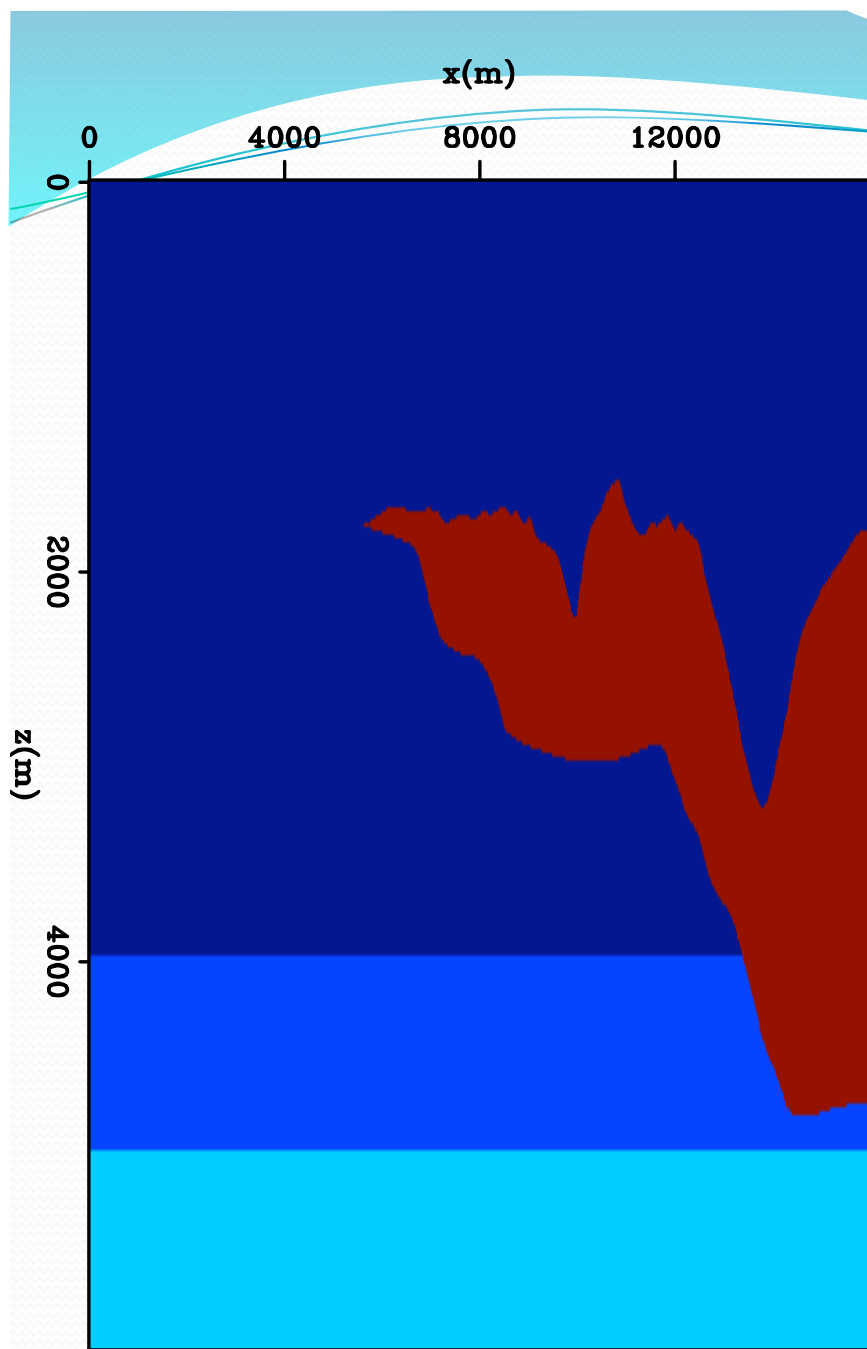
$d^{mod}$  Synthetic data

$d^{obs}$  Observed data

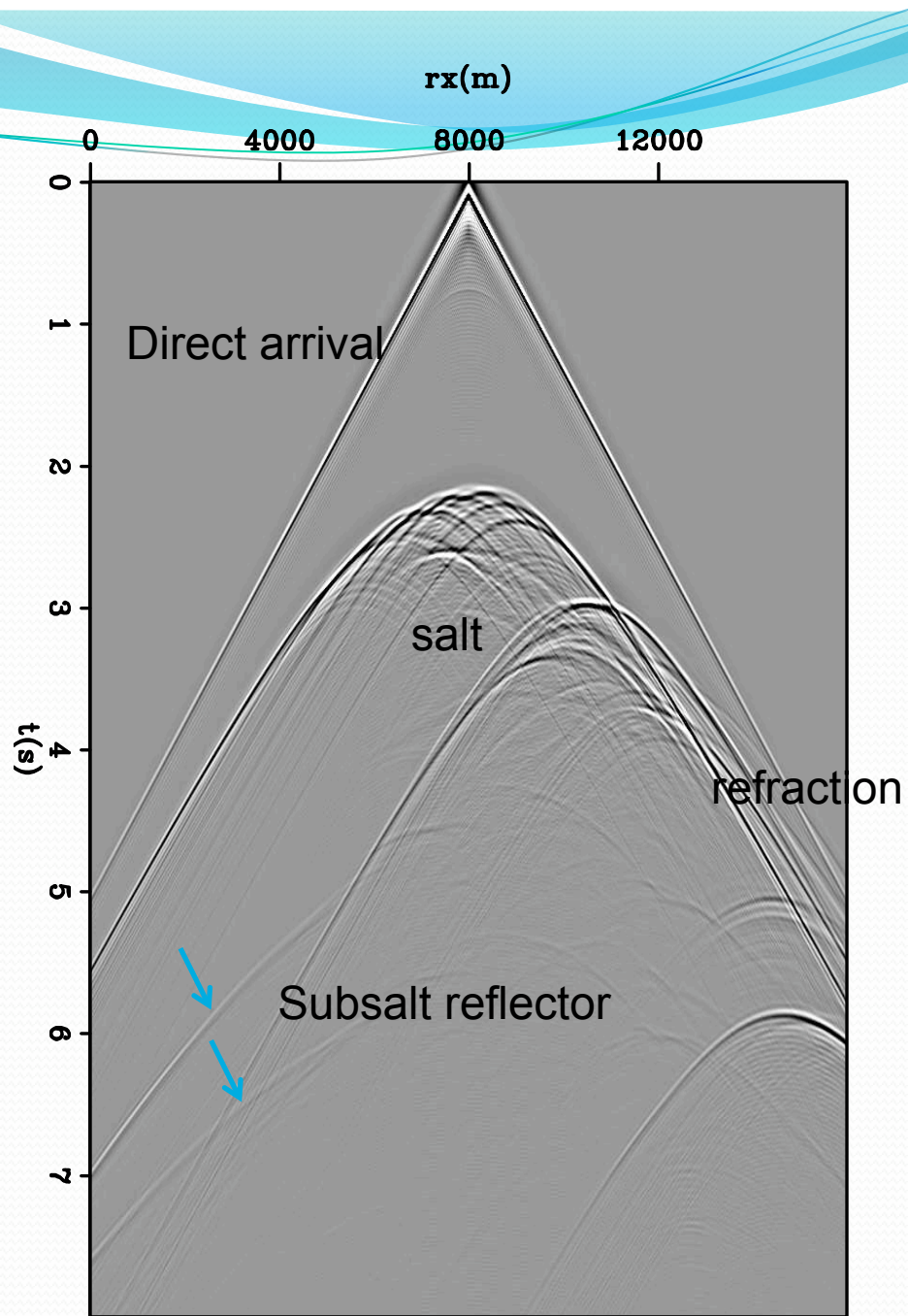
$F(s_o^2)$  Background data





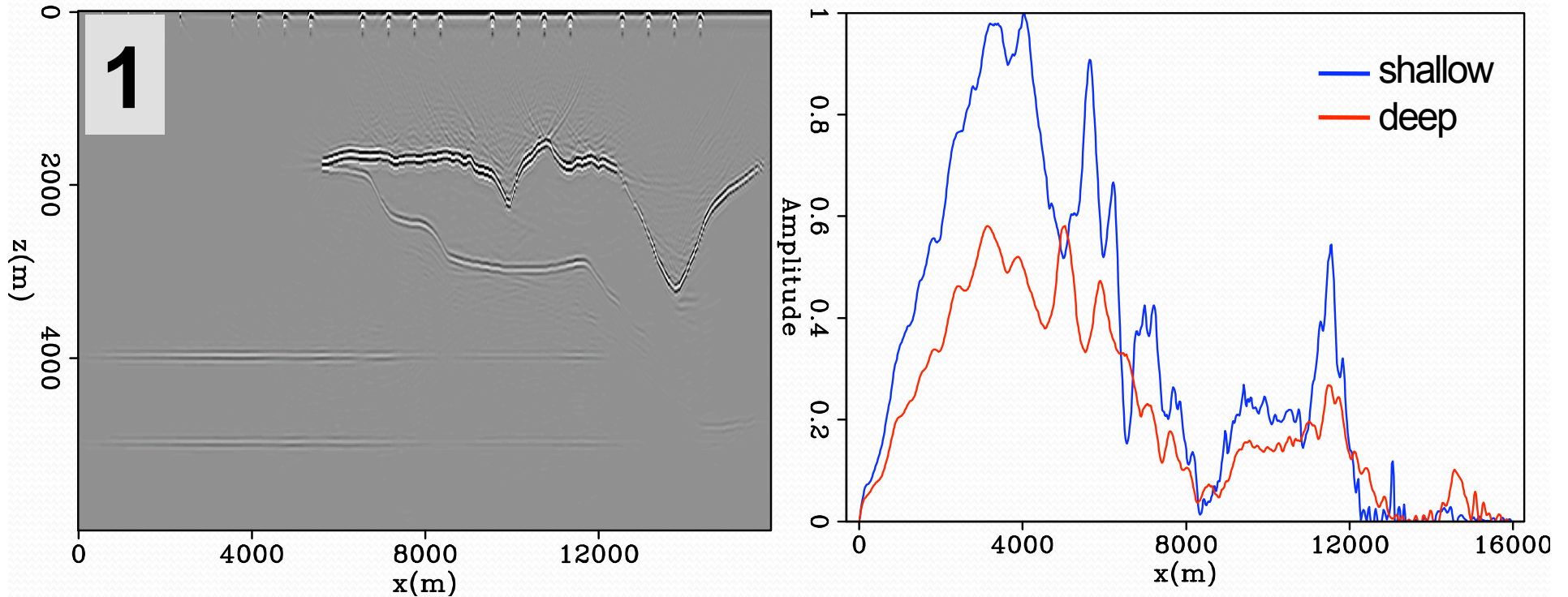


Salt+layer



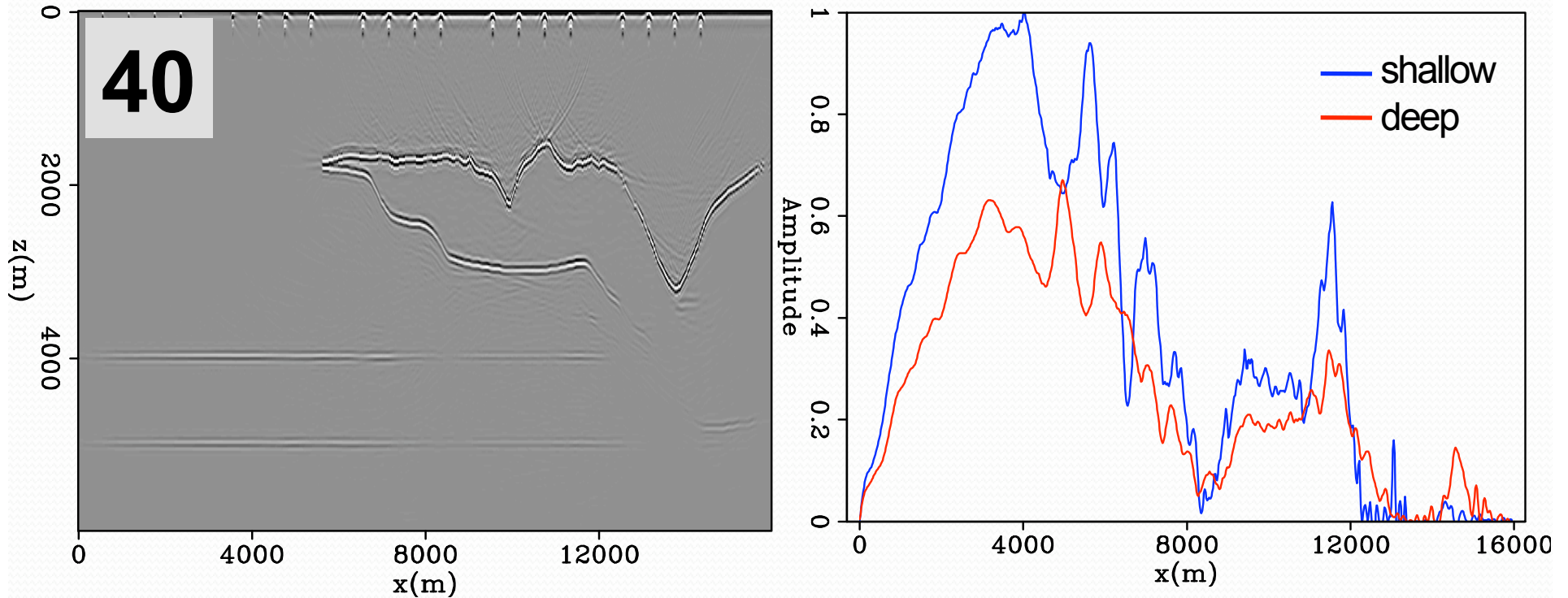
Salt+layer

# Test 1: LSRTM ignoring background data



$$S_1(\mathbf{m}) = \|\mathbf{L}m - d^{obs}\|^2$$

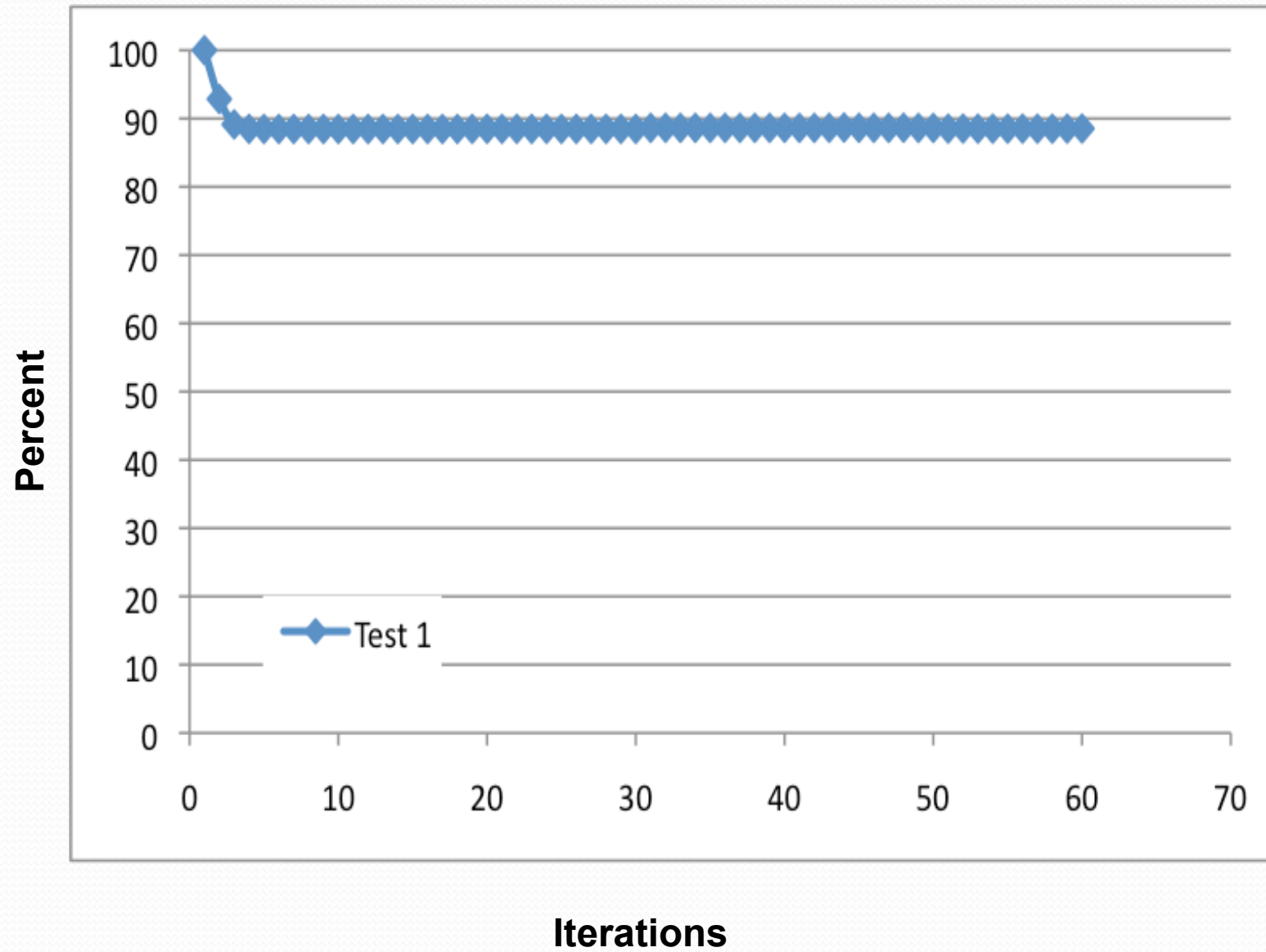
# Test 1: LSRTM ignoring background data



$$S_1(\mathbf{m}) = \|\mathbf{L}m - d^{obs}\|^2$$



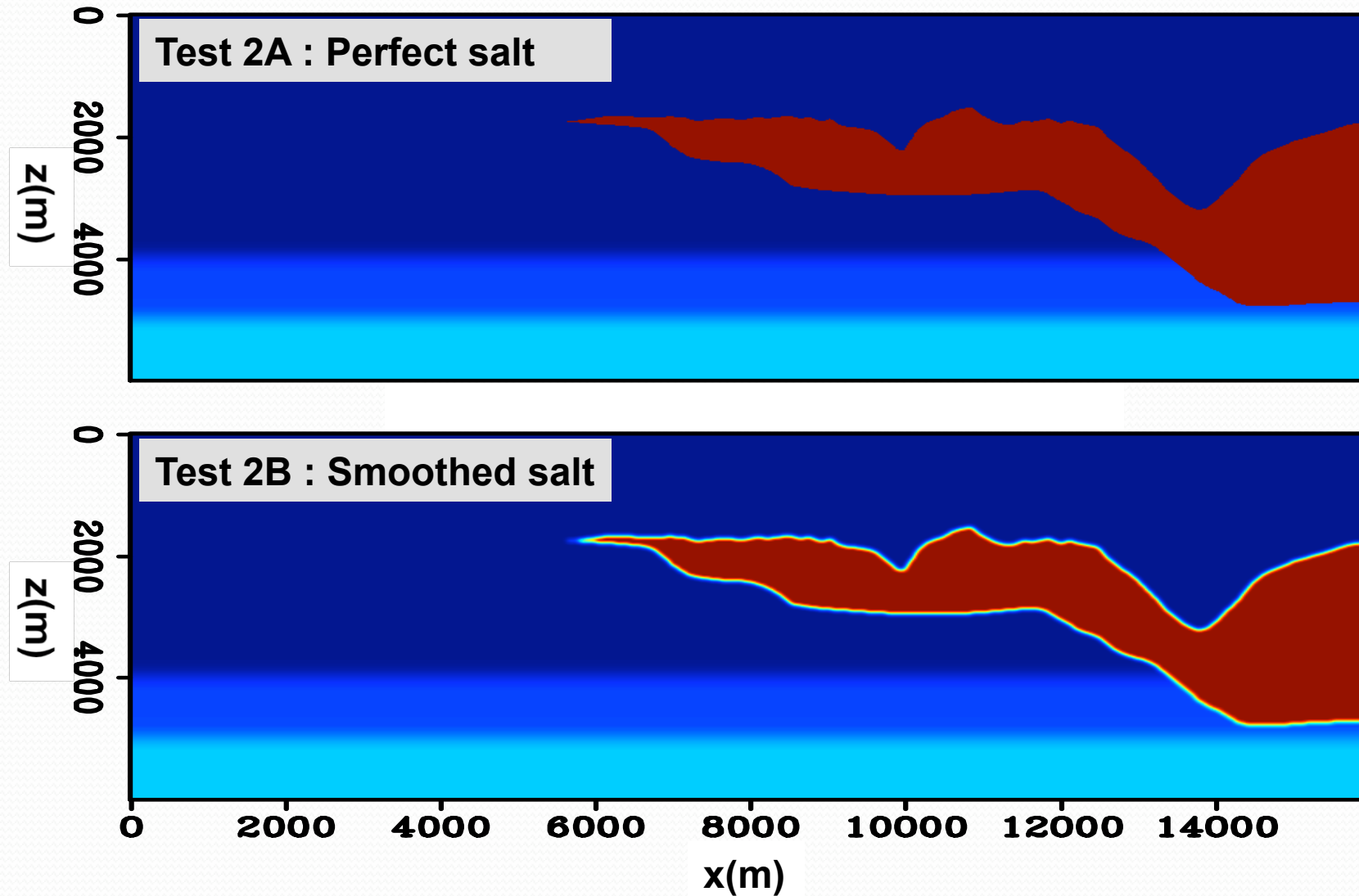
## Residual Curve for Test 1



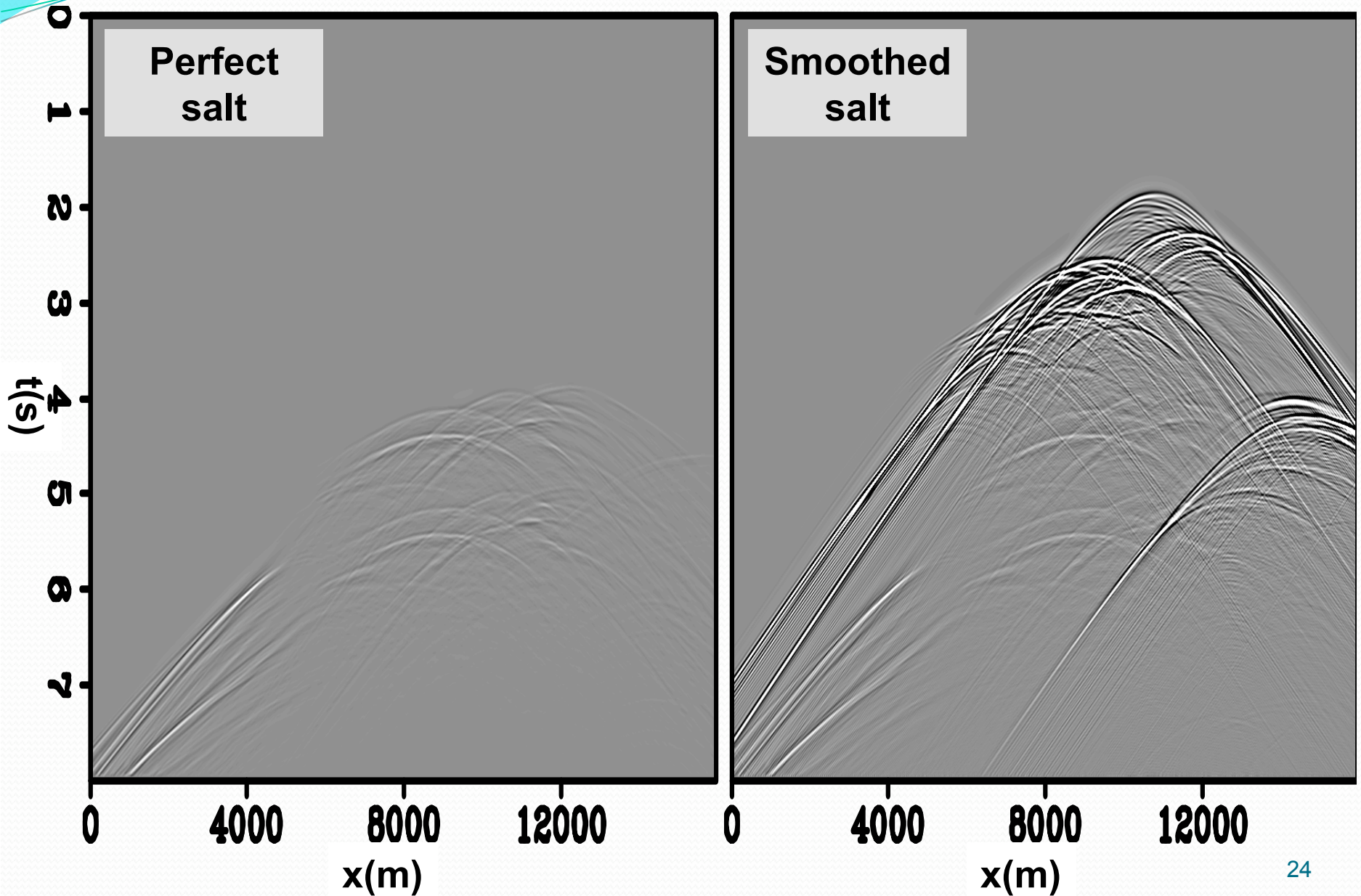
# Let use the correct objective funtion

$$S_2(\mathbf{m}) = \| \mathbf{L}m - (d^{obs} - F(s_o^2)) \|^2$$

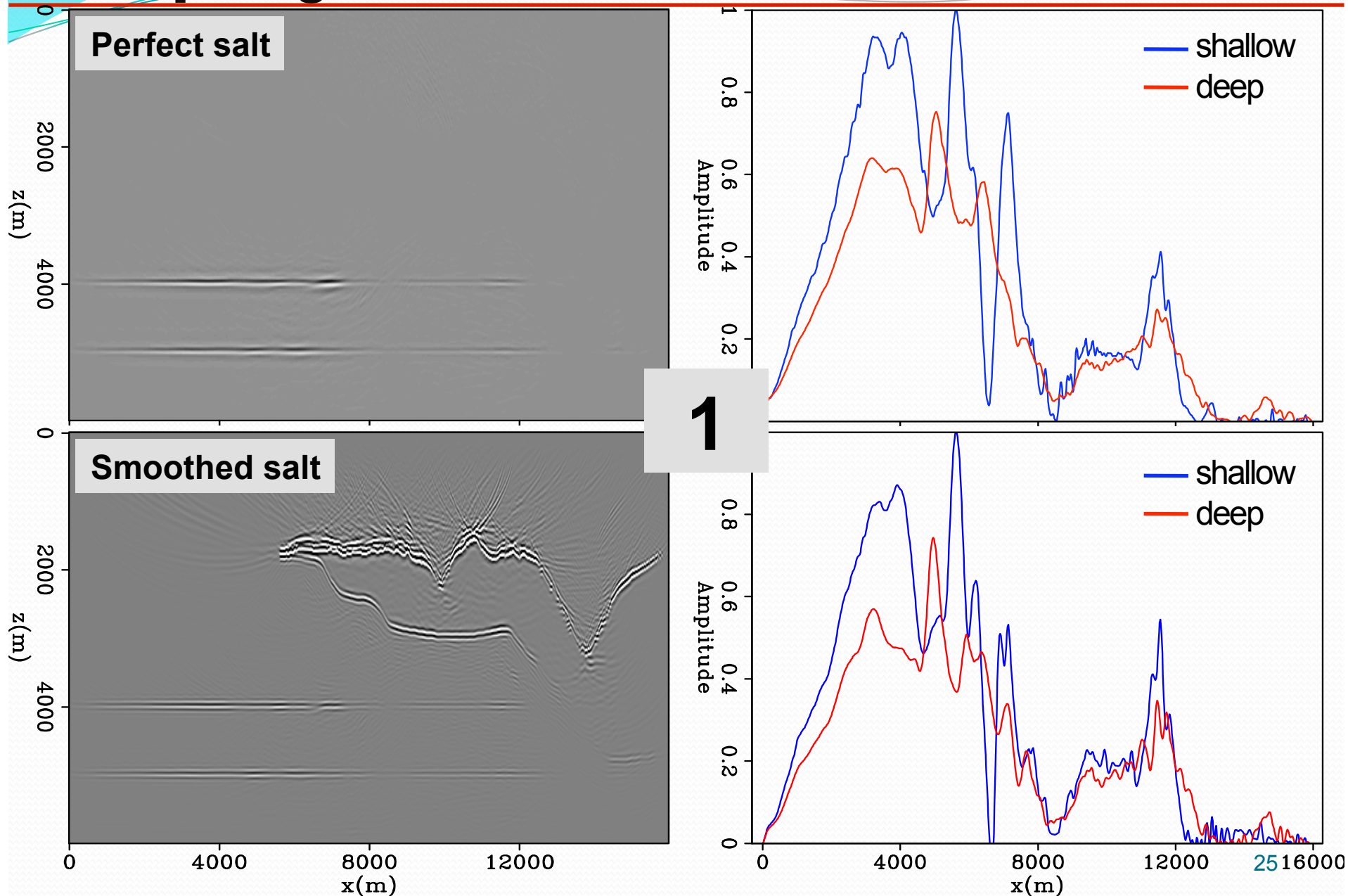
## Test 2 – Subtract background data with varying degree of accuracy



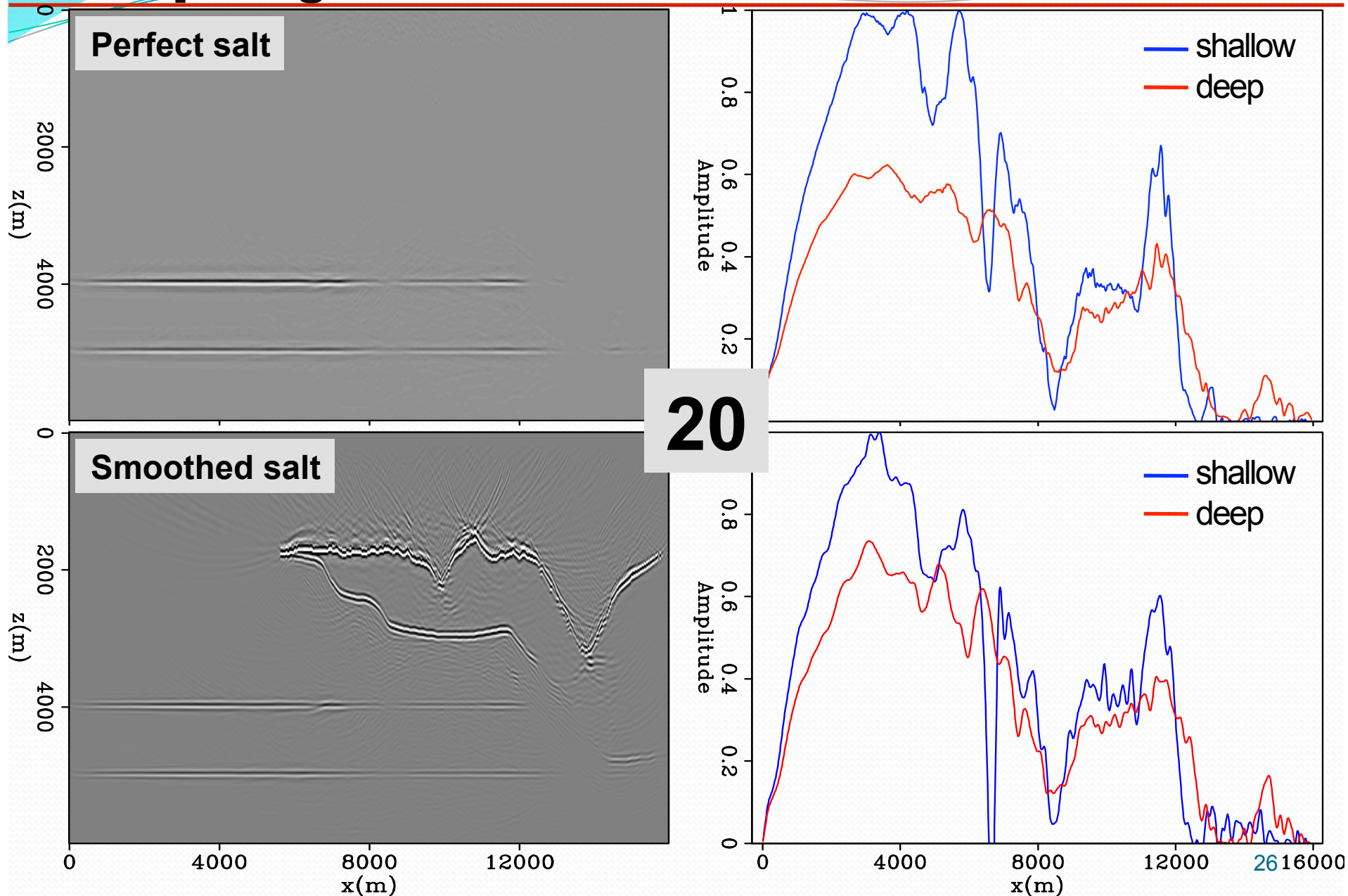
## Input for inversion: perturbed data



# Comparing between two LSRTM

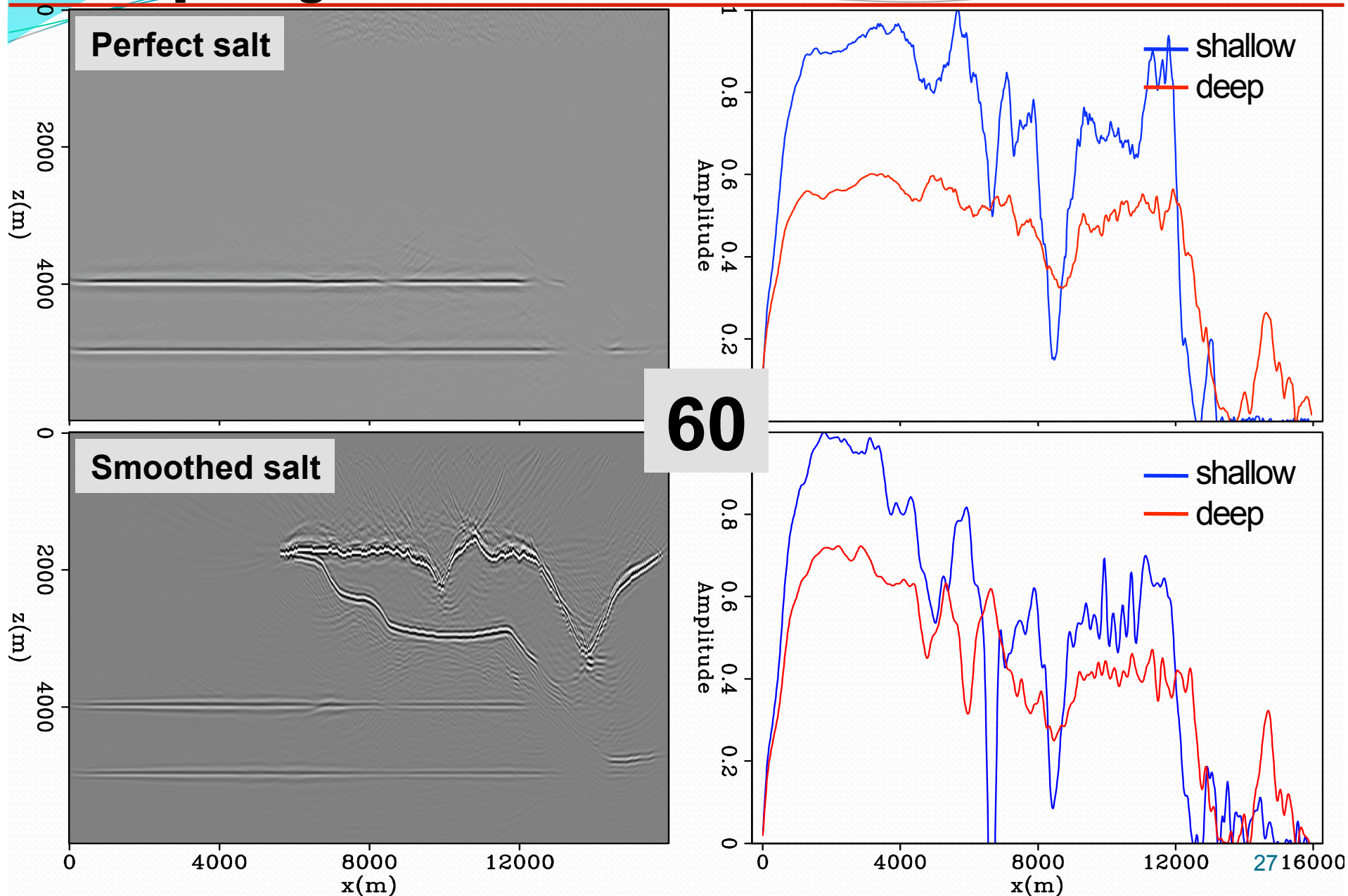


# Comparing between two LSRTM

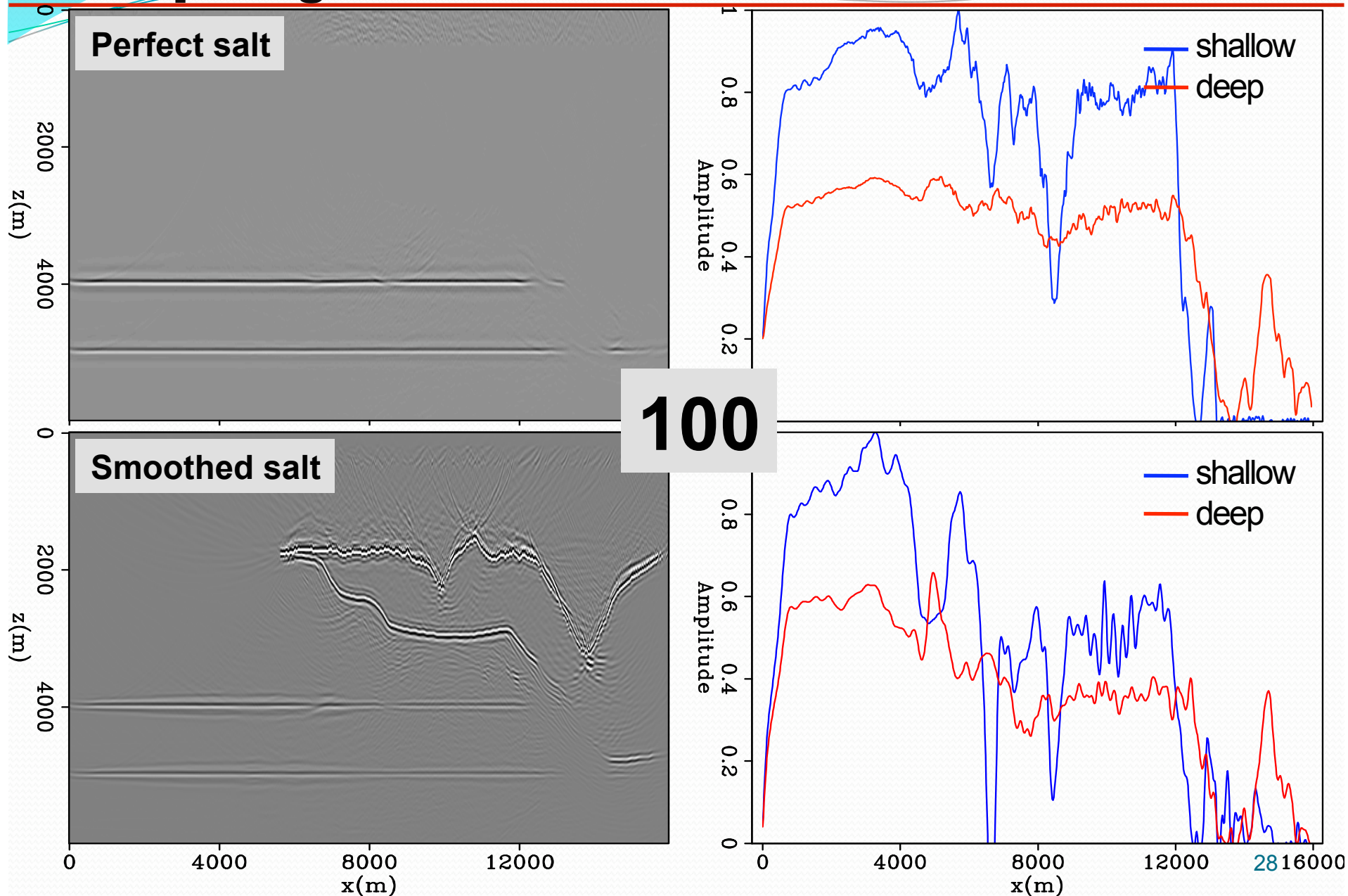




# Comparing between two LSRTM

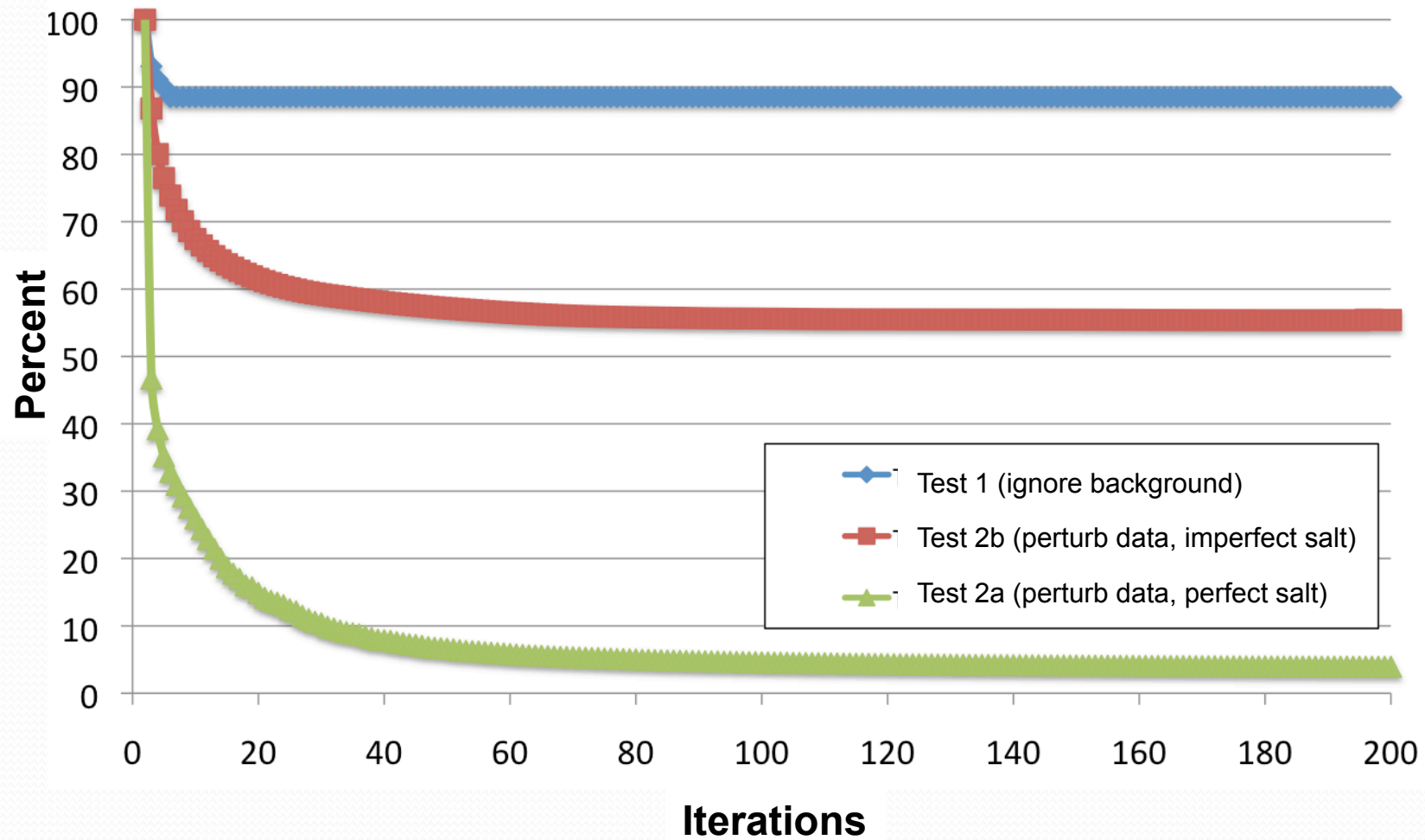


# Comparing between two LSRTM

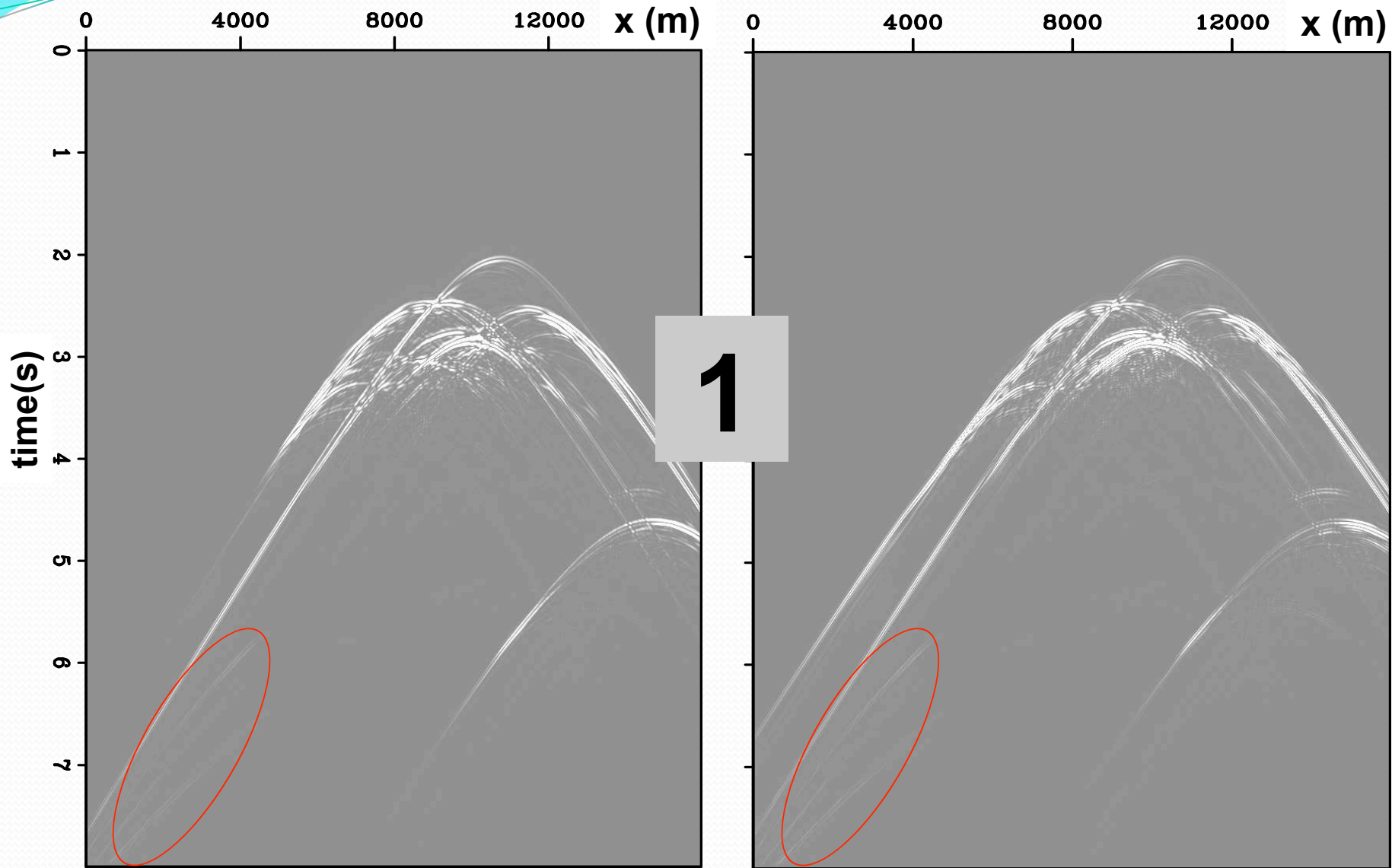




# Convergence curve



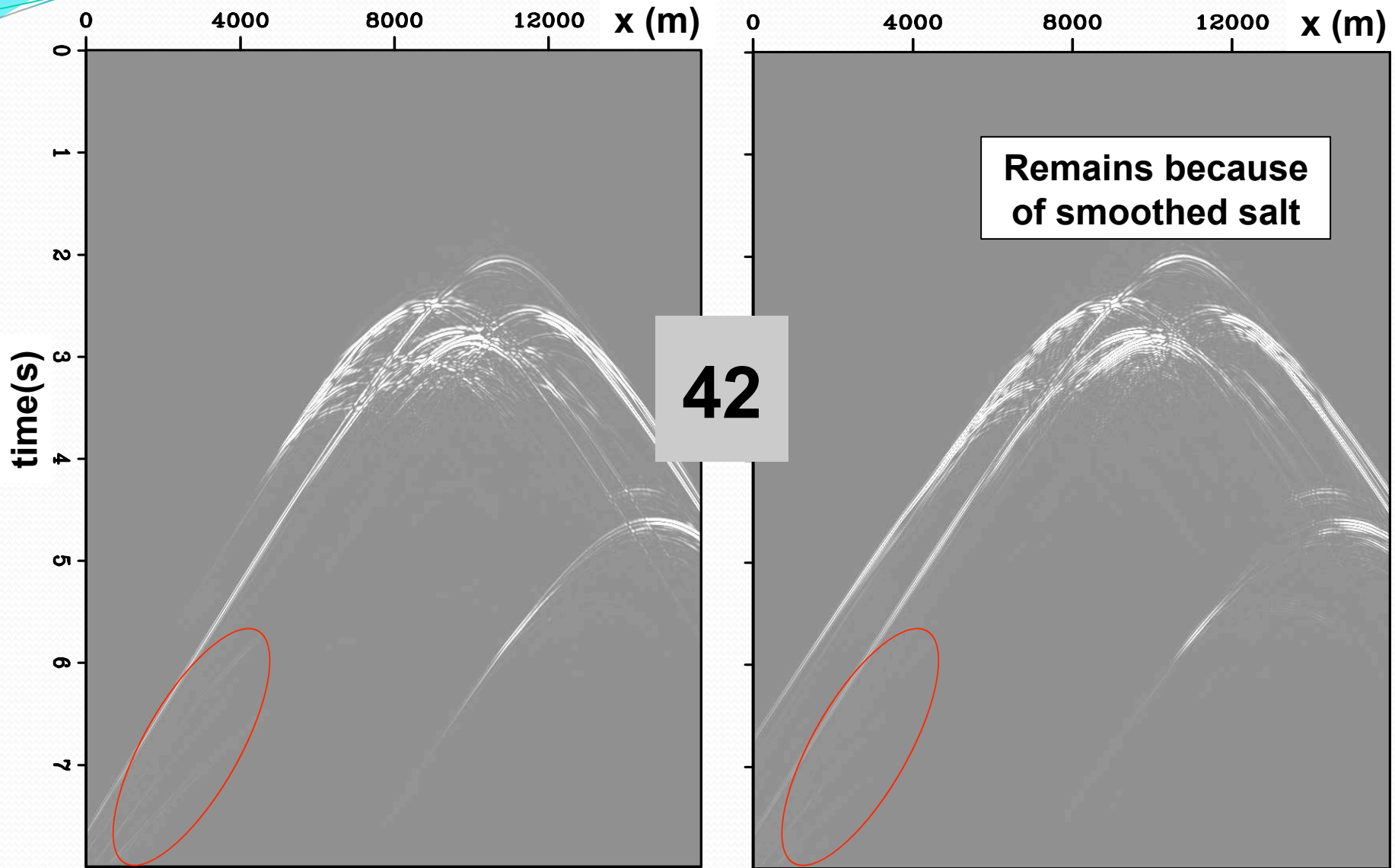
# Data residual squared for the smoothed salt test



Test 1: ignore background data

Test 2: subtract background data

# Data residual squared for the smoothed salt test



Test 1: ignore background data

Test 2: subtract background data

# Summary

- When the background data is significant



$$S_1(\mathbf{m}) \quad \neq \quad \|\mathbf{L}m - d^{obs}\|^2$$

$$S_2(\mathbf{m}) \quad = \quad \|\mathbf{L}m - (d^{obs} - F(s_o^2))\|^2$$

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# Challenges when applying to field dataset

- Applying this to field dataset is non-trivial
  - the background data is just an approximation

$$d^{obs} - F(s_o^2)$$

- Perhaps LSM in the image space can handle this problem better

# LSRTM with salt dimming

$$S_3(\mathbf{m}) = \|W_s(\mathbf{L}m - d^{obs})\|^2$$

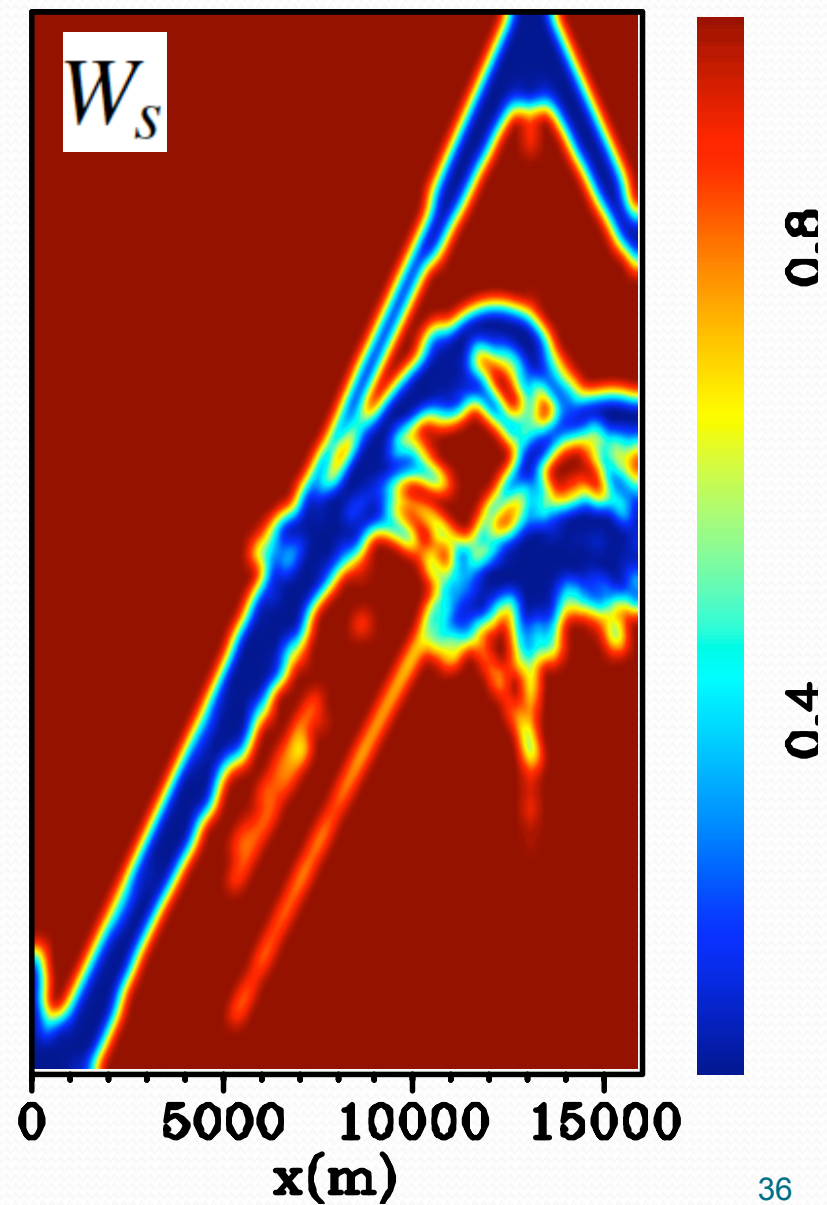
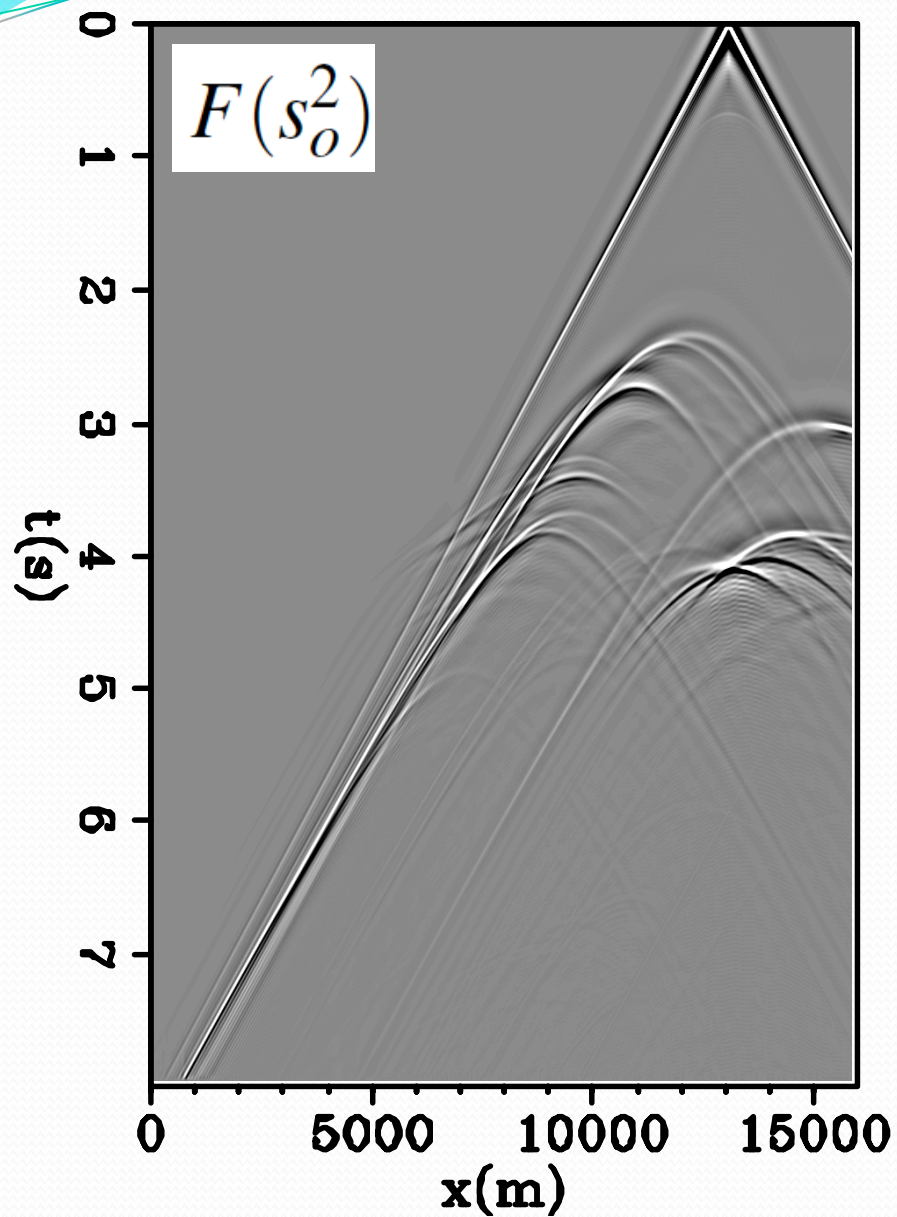
$W_s$  data weighting function that down-weights the salt reflection energy.

- Get the most out of as few iterations as possible while addressing the background data problem



# Calculating

$W_s$





# LSRTM result with and without salt dimming

Test 1 : ignore background

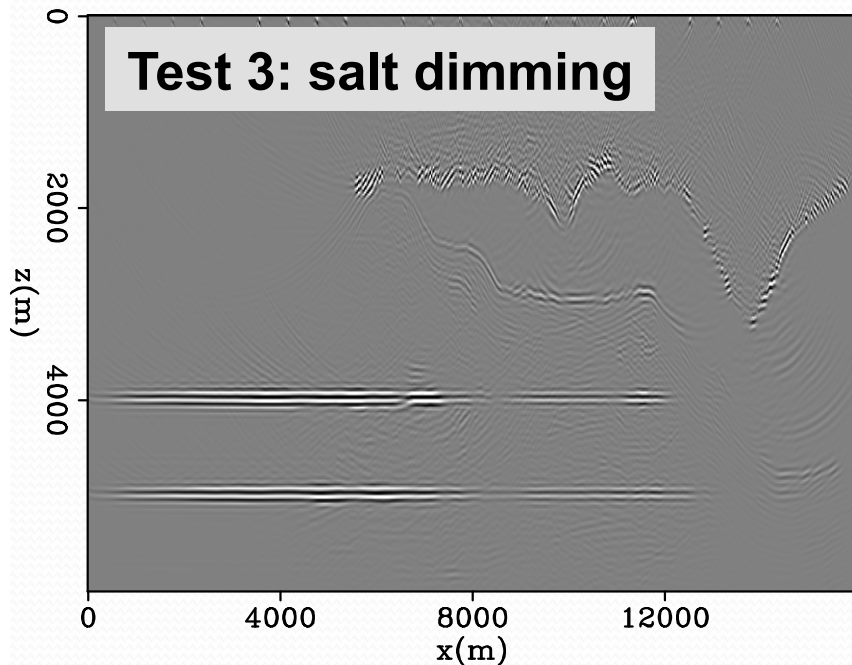
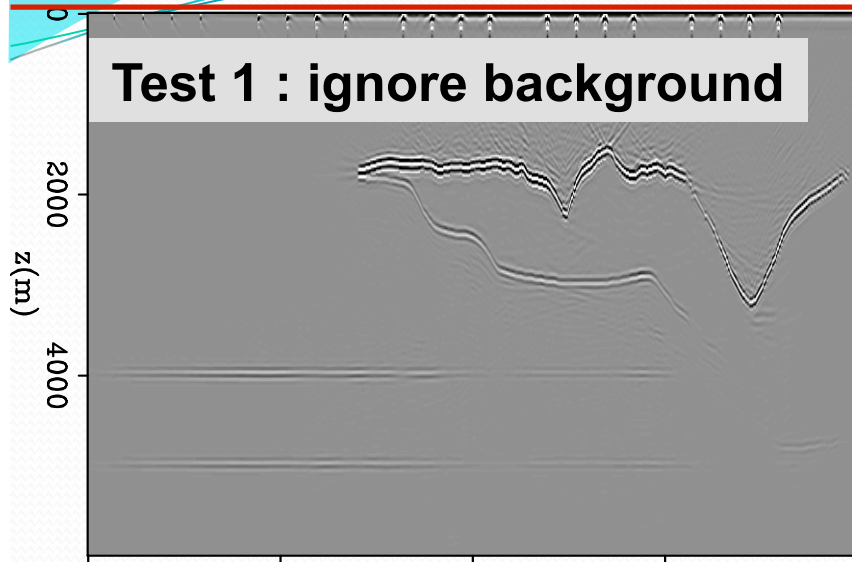
$$S_1(\mathbf{m}) = \|\mathbf{L}m - d^{obs}\|^2$$

using observed data

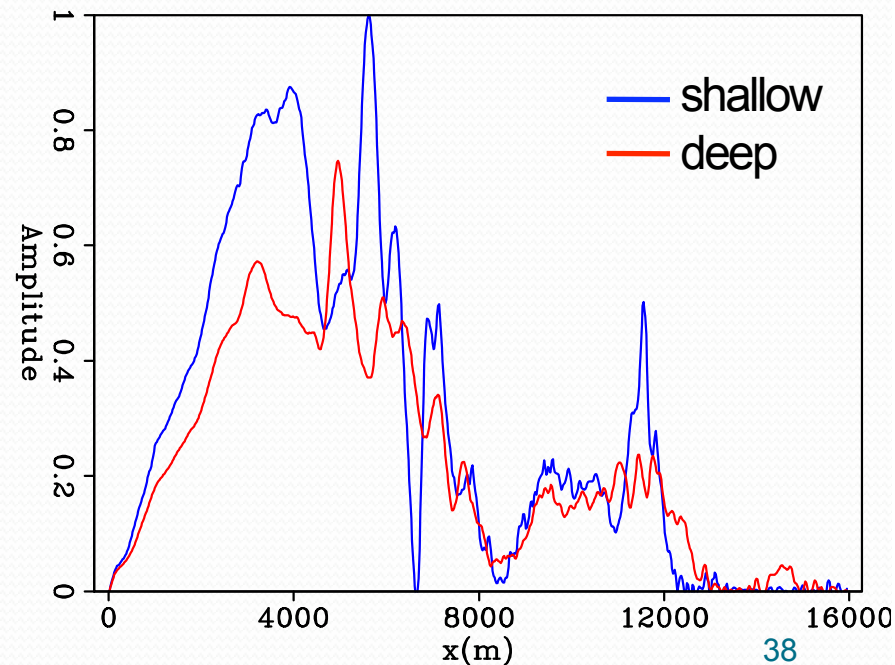
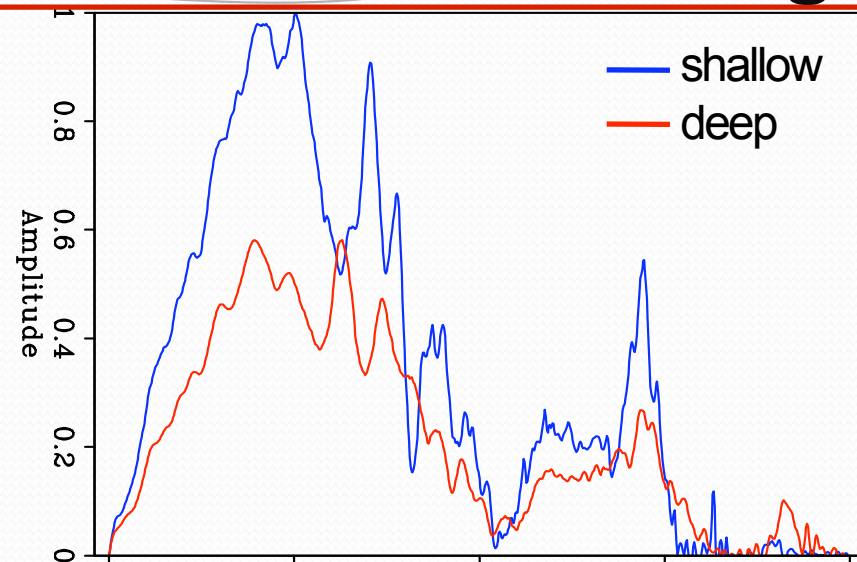
Test 3: salt dimming

$$S_3(\mathbf{m}) = \|W_s(\mathbf{L}m - d^{obs})\|^2$$

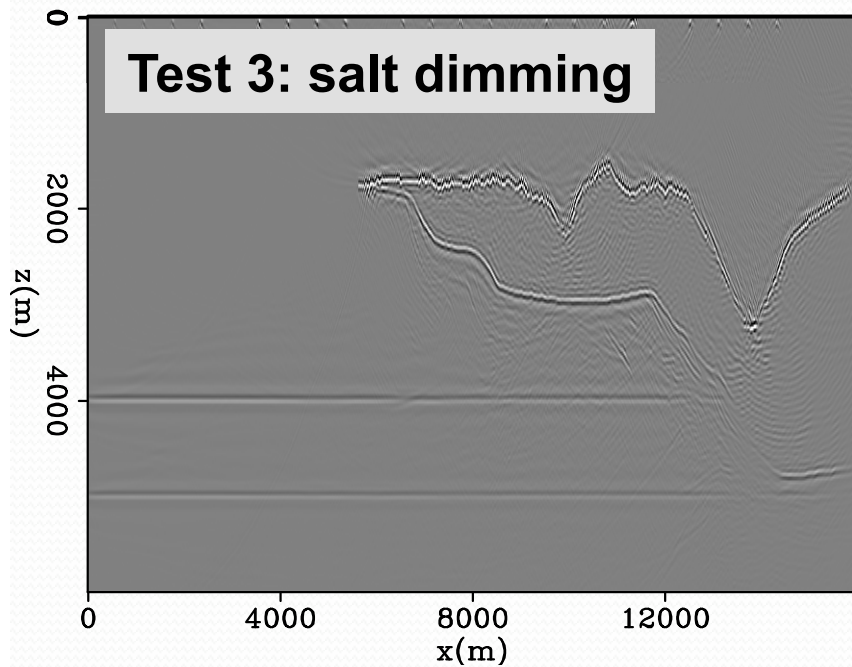
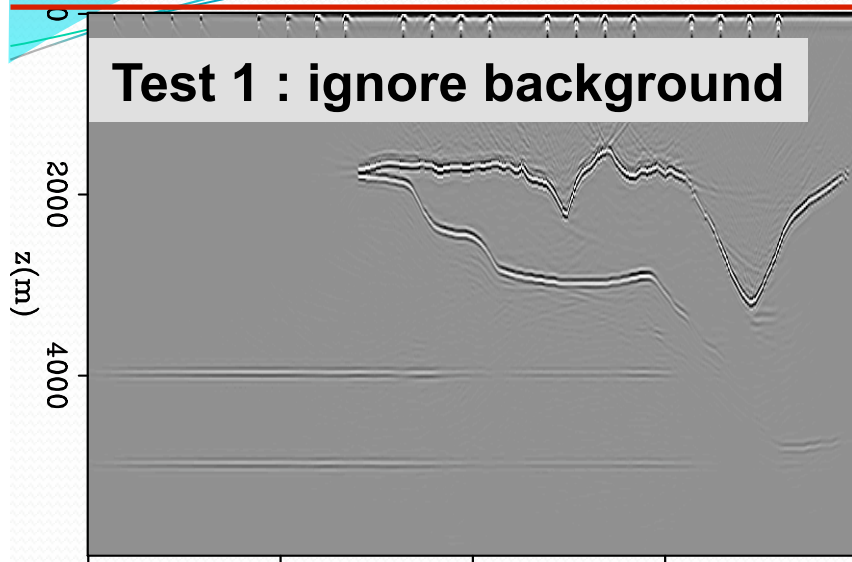
# LSRTM result with and without salt dimming



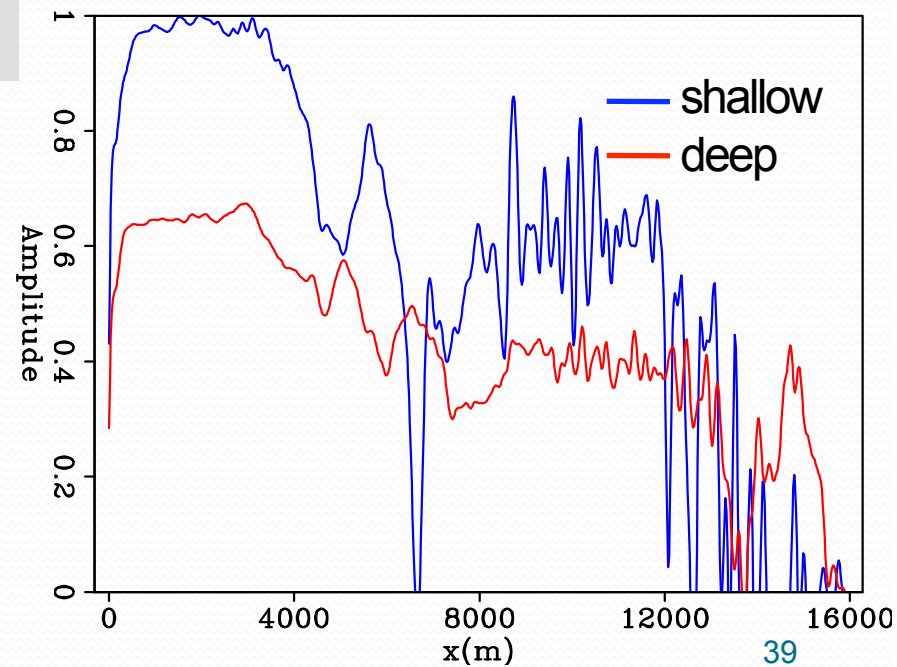
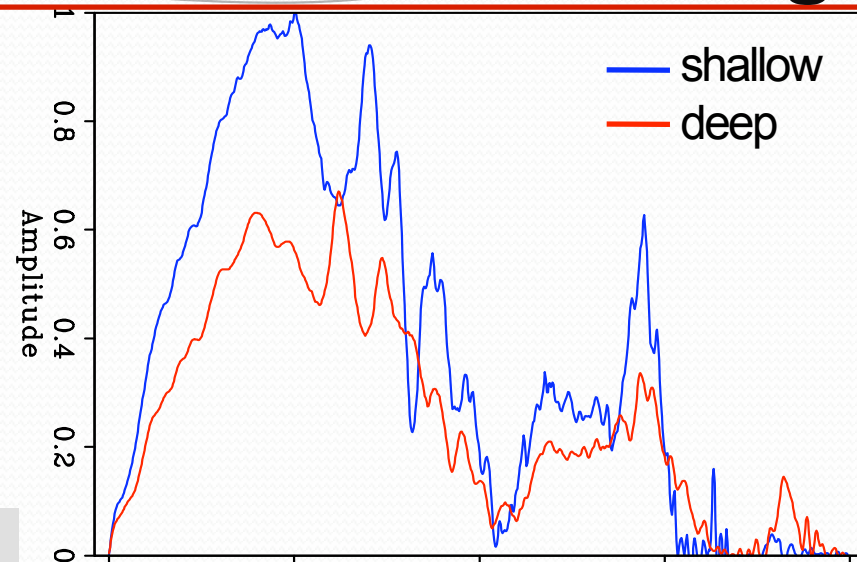
1



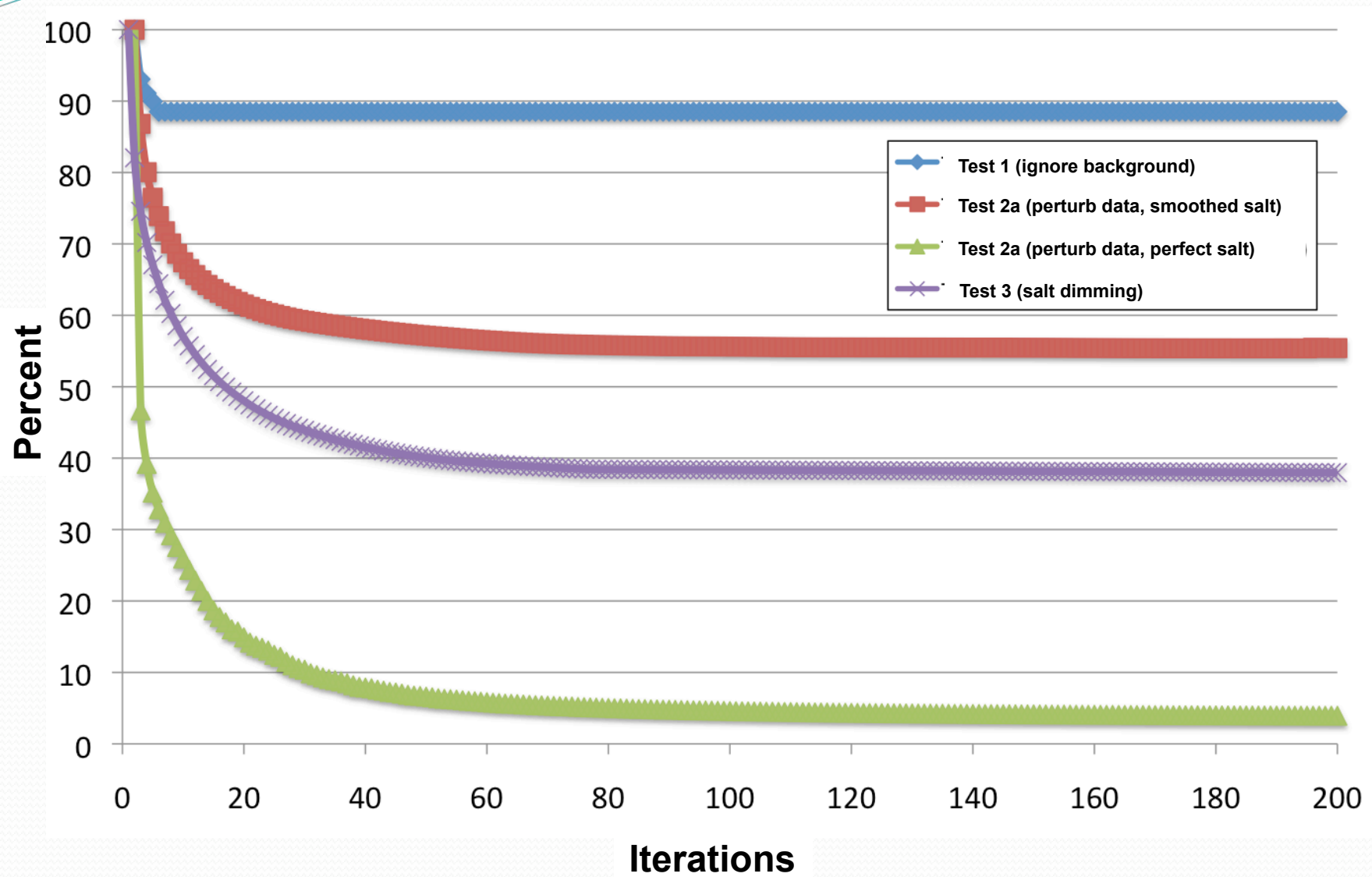
# LSRTM result with and without salt dimming



40



# Convergence Curves



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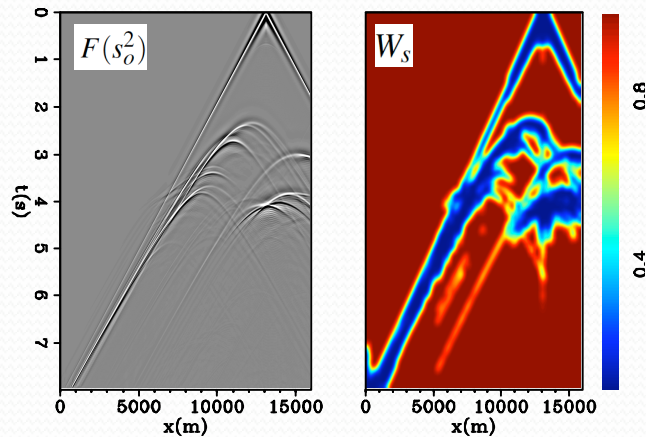
# Conclusion

- Subtracting the background data in LSM is important but non-trivial to apply on field datasets



$$S_2(\mathbf{m}) = \|\mathbf{L}m - (d^{obs} - F(s_o^2))\|^2$$

- Salt-dimming is a viable solution to address this issue



$$S_3(\mathbf{m}) = \|\mathbf{W}_s(\mathbf{L}m - d^{obs})\|^2$$



# Acknowledgement

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- Biondo Biondi and Bob Clapp for helpful discussions



# Backup Slides

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# Some successful examples

- **LSM – using one-way operator**

- Kuehl, H. and Sacchi M., 2002, Robust AVP estimation using least-squares wave-equation migration: SEG Technical Program Expanded Abstract, 21, 281
- Clapp, M. R. Clapp, and B. Biondi, 2005, Regularized least-squares inversion for 3-D subsalt imaging: SEG Technical Expanded Abstract, 24, 1814-1817

- **LSRTM**

- Dai, W., C. Boonyasiriwat, and G. Schuster, 2010 – 3D multi-source least-squares reverse time migration: SEG Technical Expanded Abstract, 29, 3120-3124
- Wong, M., Biondo B., and Ronen S., 2010, Joint inversion of up- and down-going signal for ocean bottom data, SEG Expanded Abstracts,
- Dai, W., X. Wang, and G. Schuster, 2011, Least-squares migration of multisource data with a deblurring filter: Geophysics, 76, R135-R146
- Gang, Y. and Jakubowicz M., 2012 – Least-squares Reverse-Time Migration
- Dong et al., 2012, Least-squares reverse time migration: towards true amplitude imaging and improving the resolution
- Wong, M., Biondo B. and Ronen S., 2012, Imaging with multiples using linearized full-wave inversion