

# Review of tomographic methods

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**NSERC  
CRSNG**



**UNIVERSITY OF CALGARY**  
FACULTY OF SCIENCE  
Department of Geoscience



- Classical reflection tomography
- Prestack depth migration (PSDM) tomography
- Stereotomography
- Compare synthetic test results from classical stereotomography and adjoint stereotomography
- Conclusions

$$t_{raypath} = \int_{raypath} s(x, z) dl$$

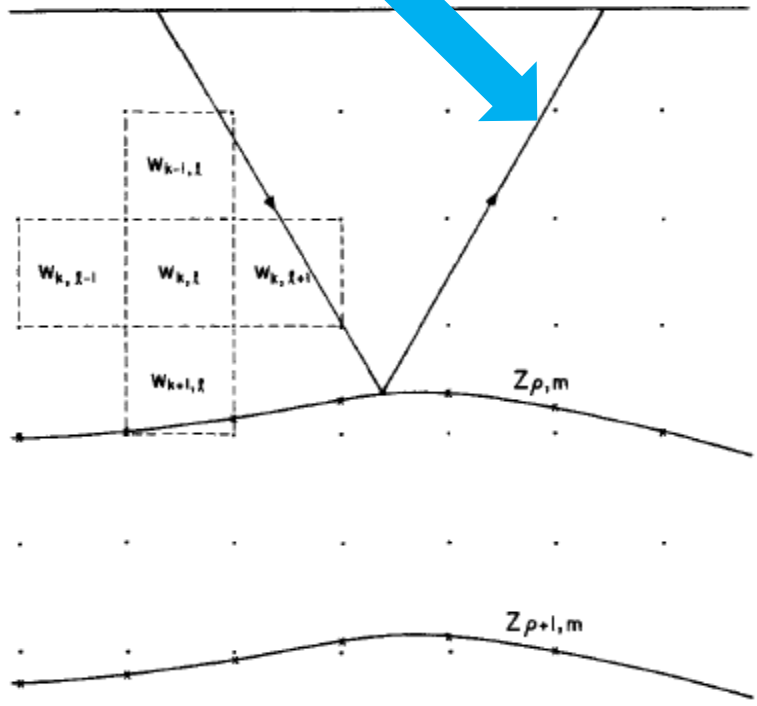
$$T = L S$$

$$\Delta T = L \Delta S$$

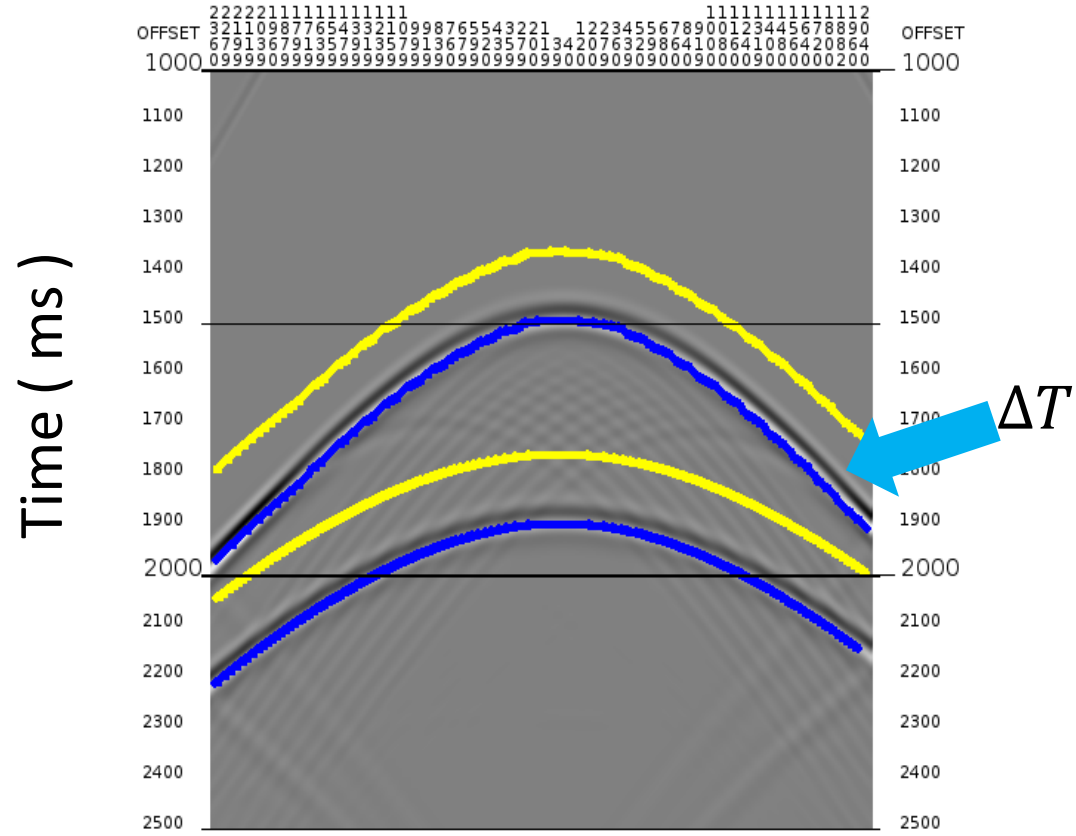
$$L = \begin{bmatrix} L_{1,1} & L_{1,2} & L_{1,3} & L_{1,4} & \dots & L_{1,m} \\ L_{2,1} & L_{2,2} & L_{2,3} & L_{2,4} & \dots & L_{2,m} \\ L_{3,1} & L_{3,2} & L_{3,3} & L_{3,4} & \dots & L_{3,m} \\ \dots & \dots & \dots & \dots & \dots & \dots \\ l_{n,1} & l_{n,2} & l_{n,3} & l_{n,4} & \dots & l_{n,m} \end{bmatrix}$$

$$\Delta T = \begin{bmatrix} \Delta t_1 \\ \Delta t_2 \\ \Delta t_3 \\ \dots \\ \Delta t_n \end{bmatrix}$$

$$\Delta S = \begin{bmatrix} \Delta s_1 \\ \Delta s_2 \\ \Delta s_3 \\ \dots \\ \Delta s_m \end{bmatrix}$$



Bishop (1985)



$$t_{raypath} = \int_{raypath} s(x, z) dl$$

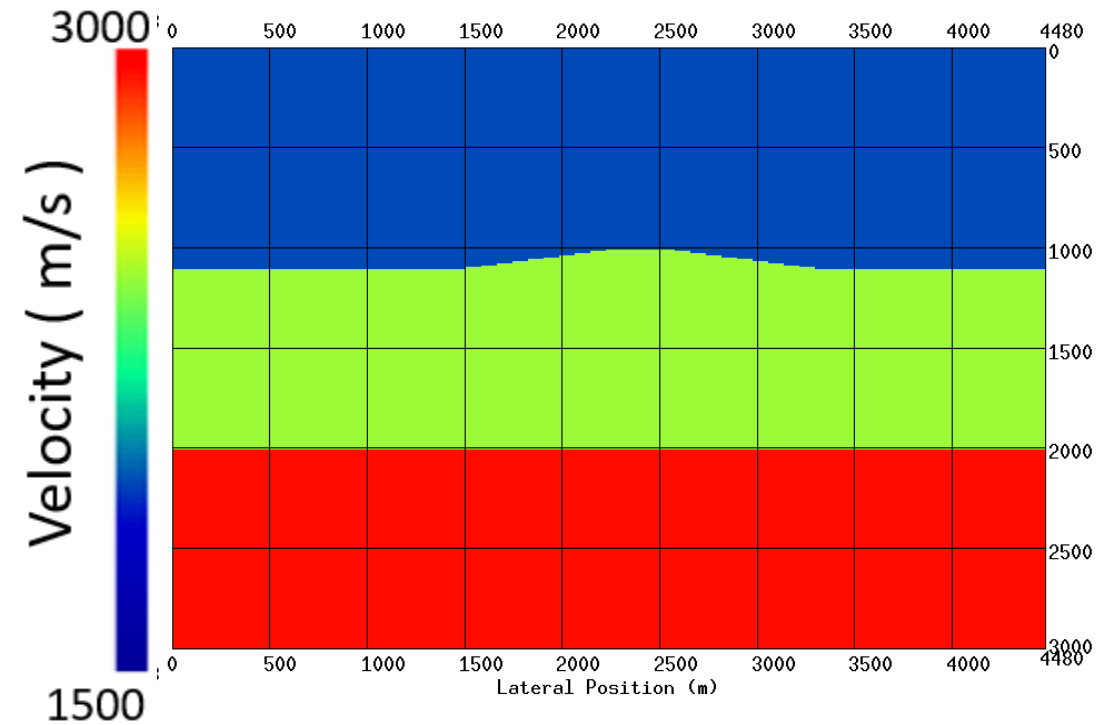
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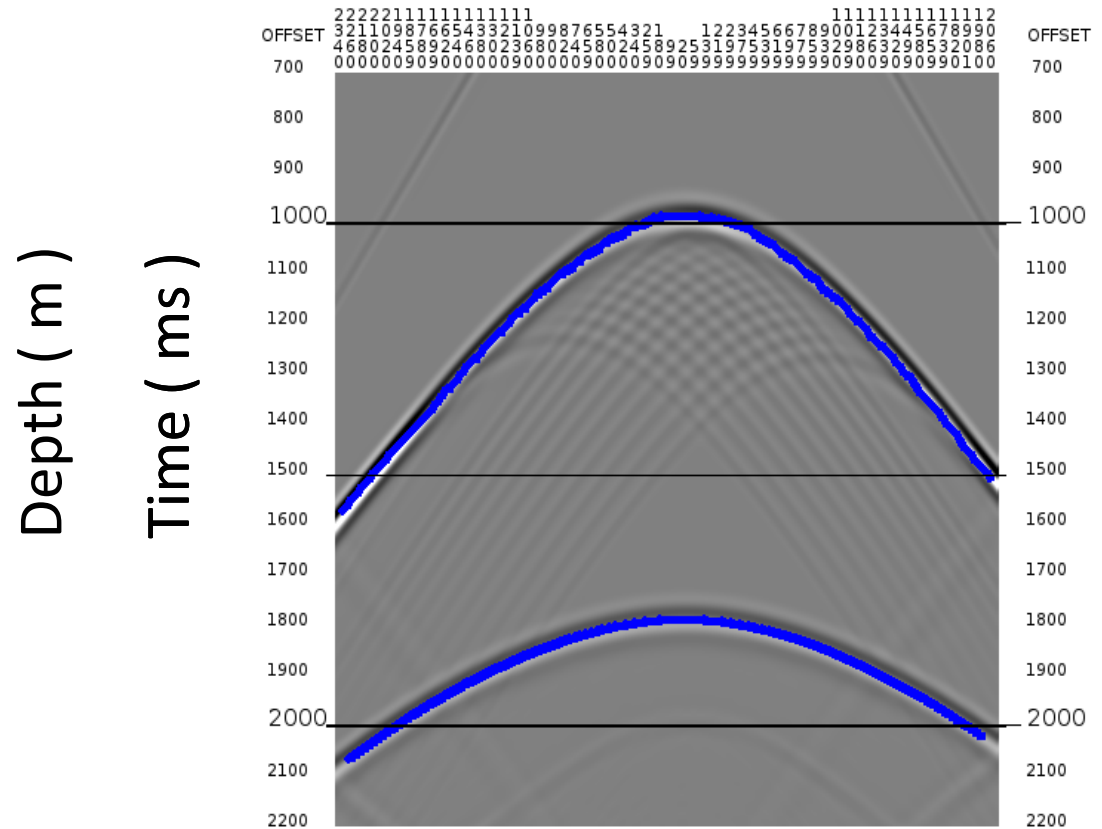
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True Model



$$t_{raypath} = \int_{raypath} s(x, z) dl$$

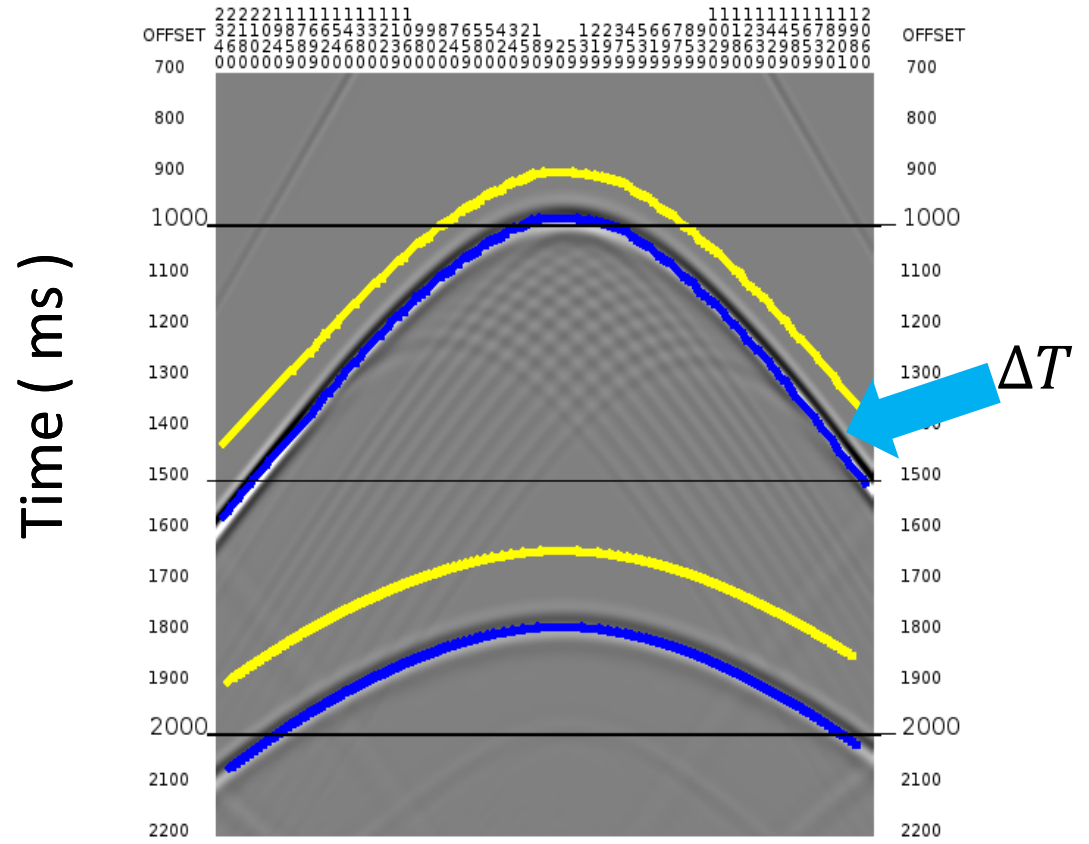
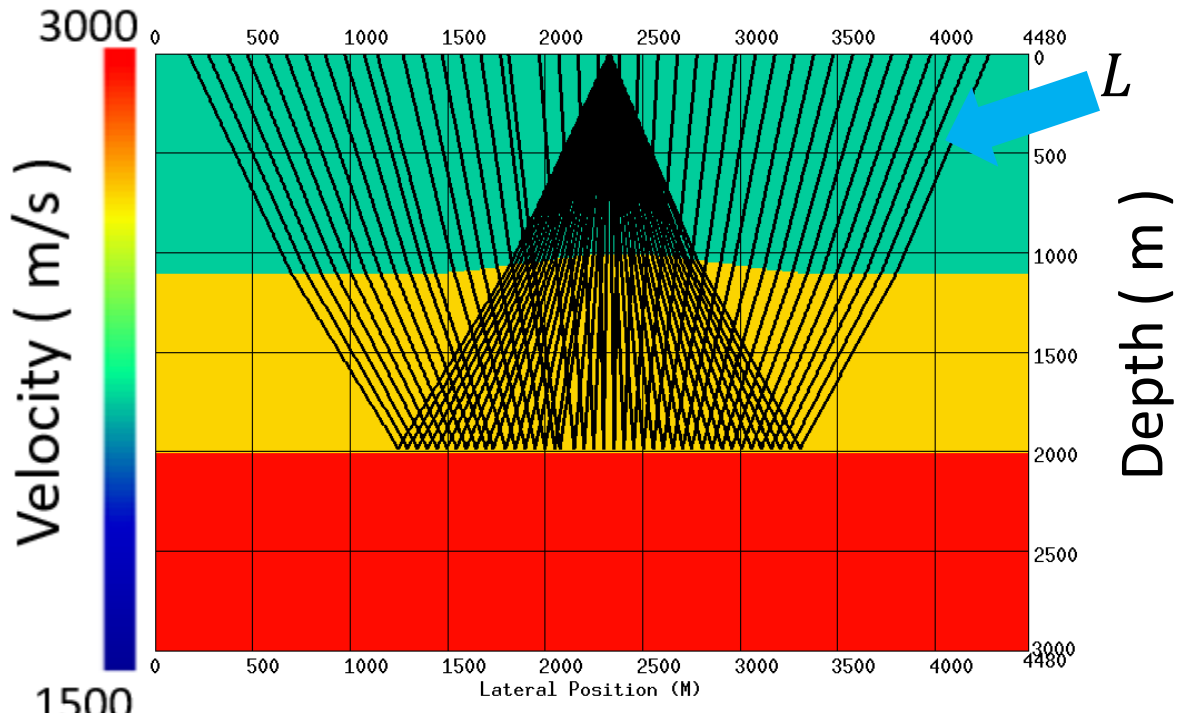
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$$t_{raypath} = \int_{raypath} s(x, z) dl$$

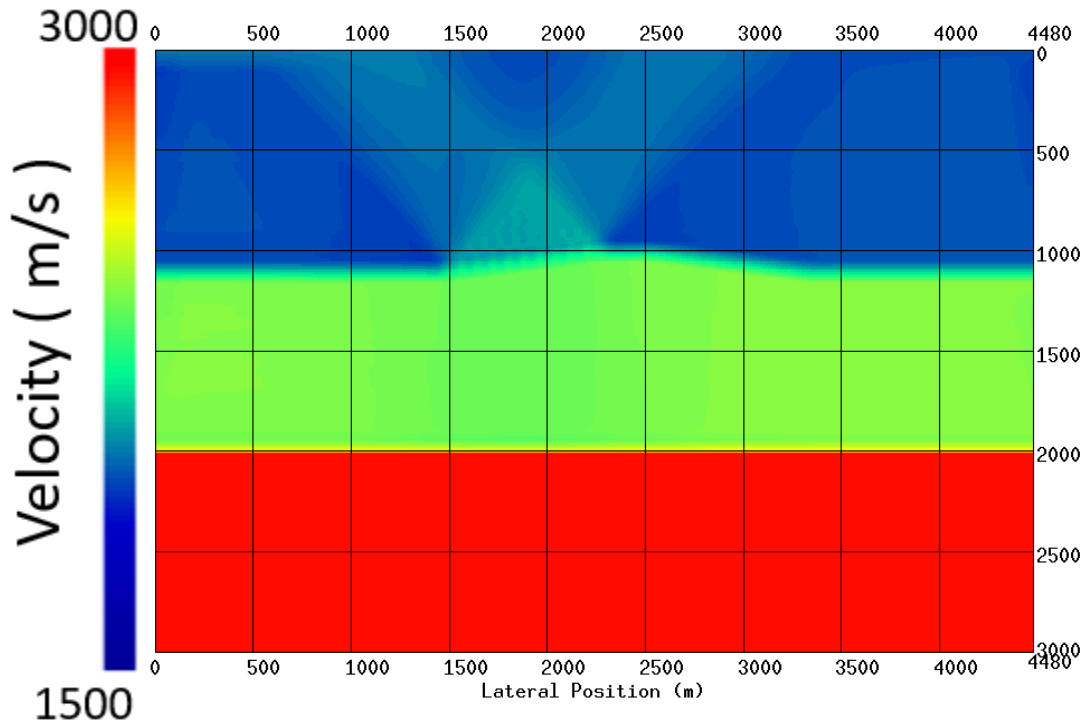
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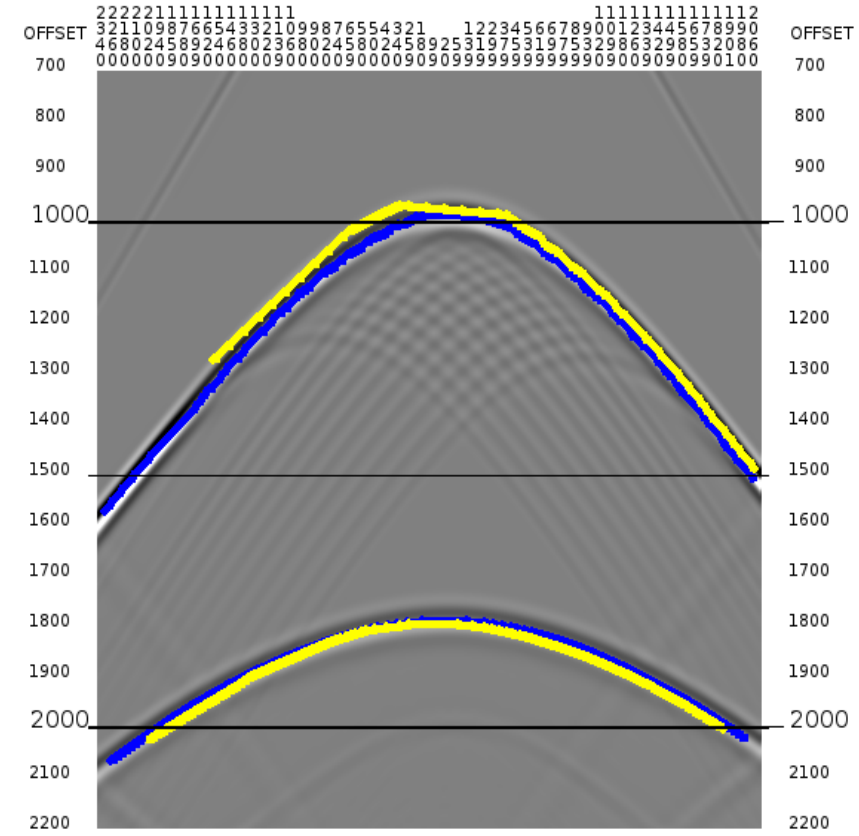
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Velocity model after 40 internal iterations



$$t_{raypath} = \int_{raypath} s(x, z) dl$$

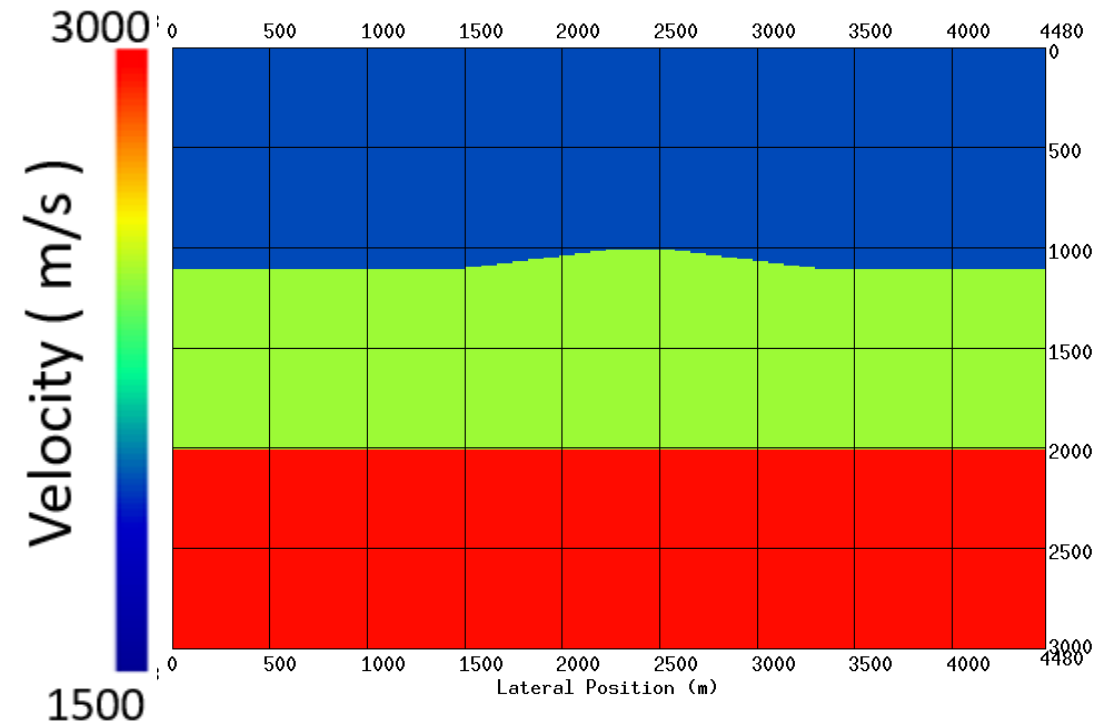
$$T = L S$$

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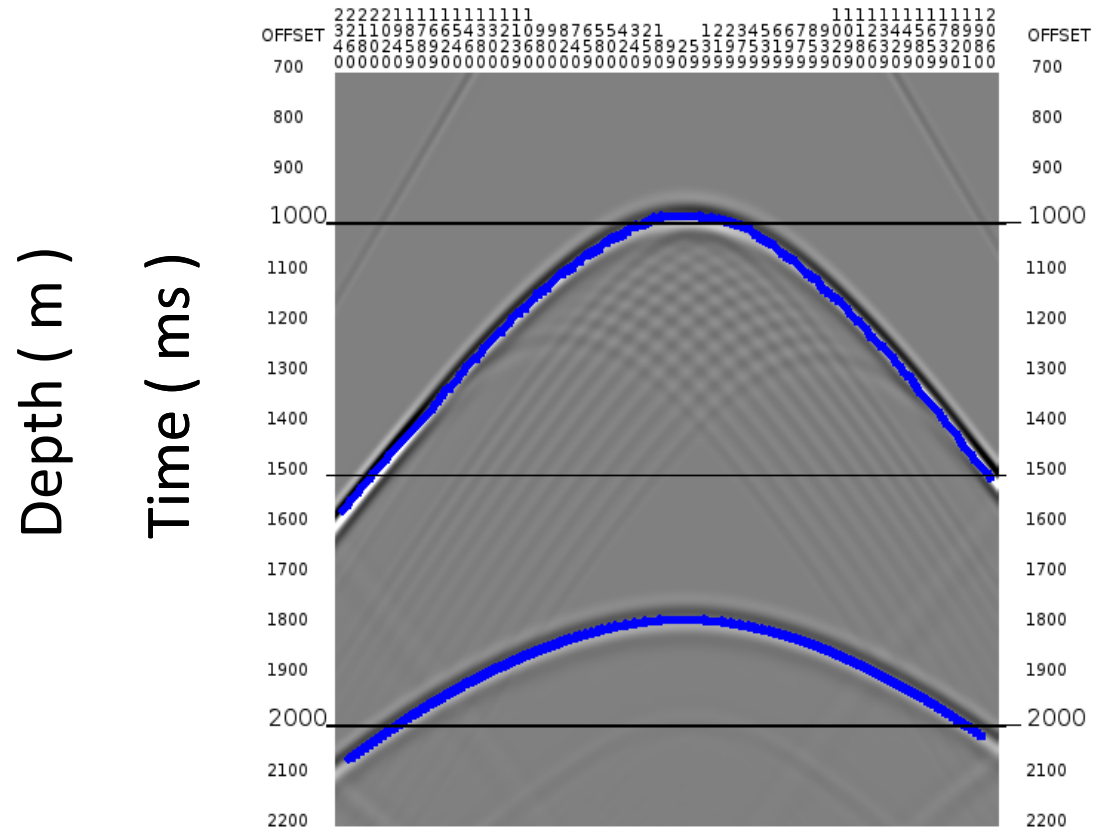
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$$\Delta T = \begin{bmatrix} \Delta t_1 \\ \Delta t_2 \\ \Delta t_3 \\ \dots \\ \Delta t_n \end{bmatrix}$$

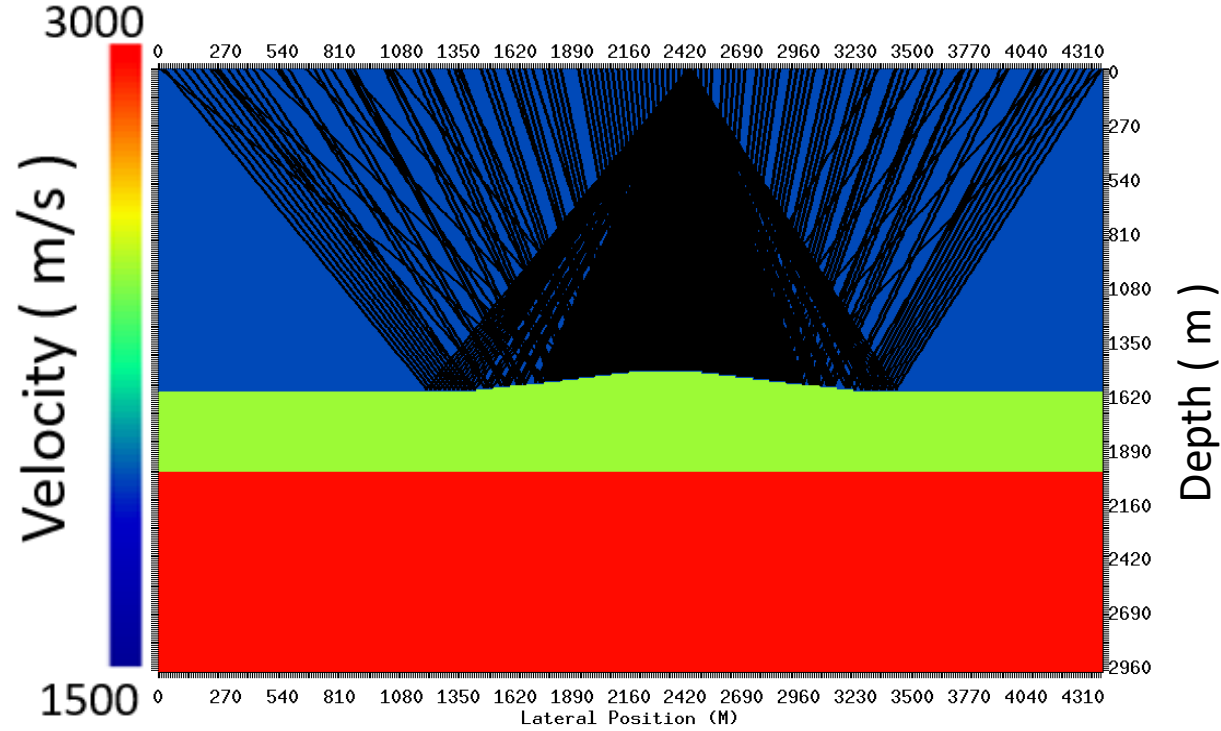
$$\Delta S = \begin{bmatrix} \Delta s_1 \\ \Delta s_2 \\ \Delta s_3 \\ \dots \\ \Delta s_m \end{bmatrix}$$



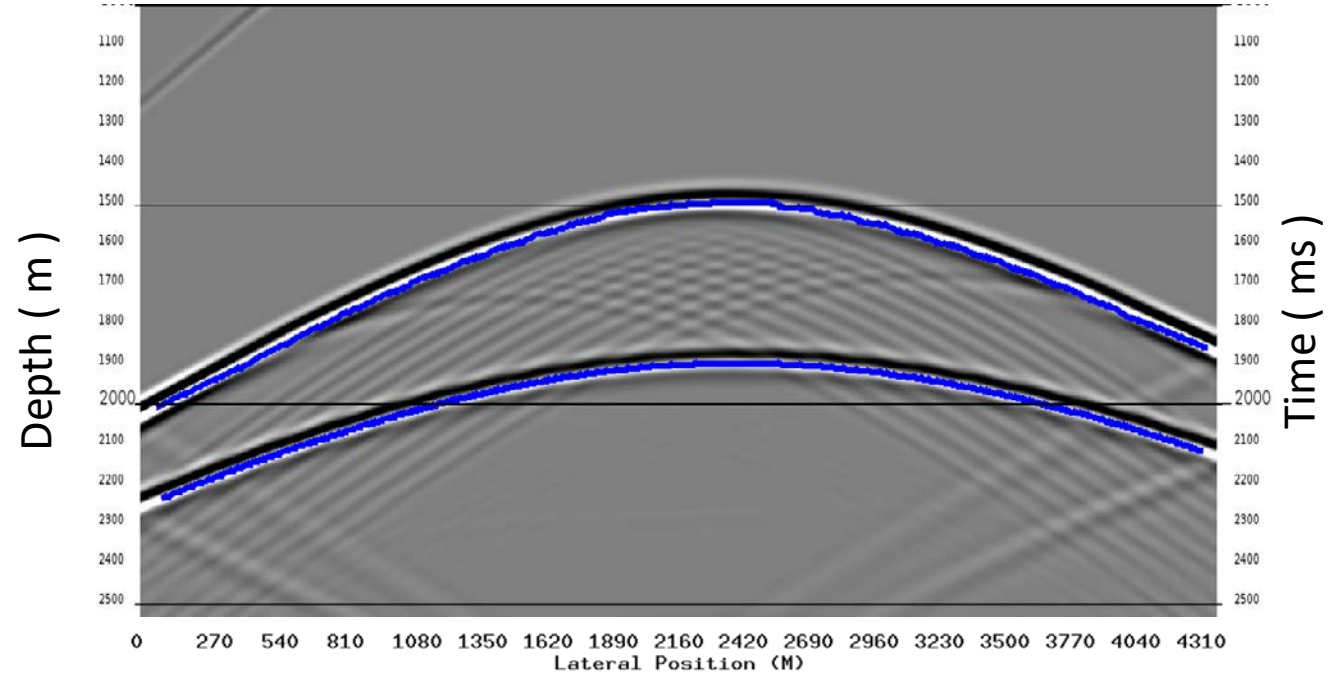
True Model



# Traveltime picking difficulties

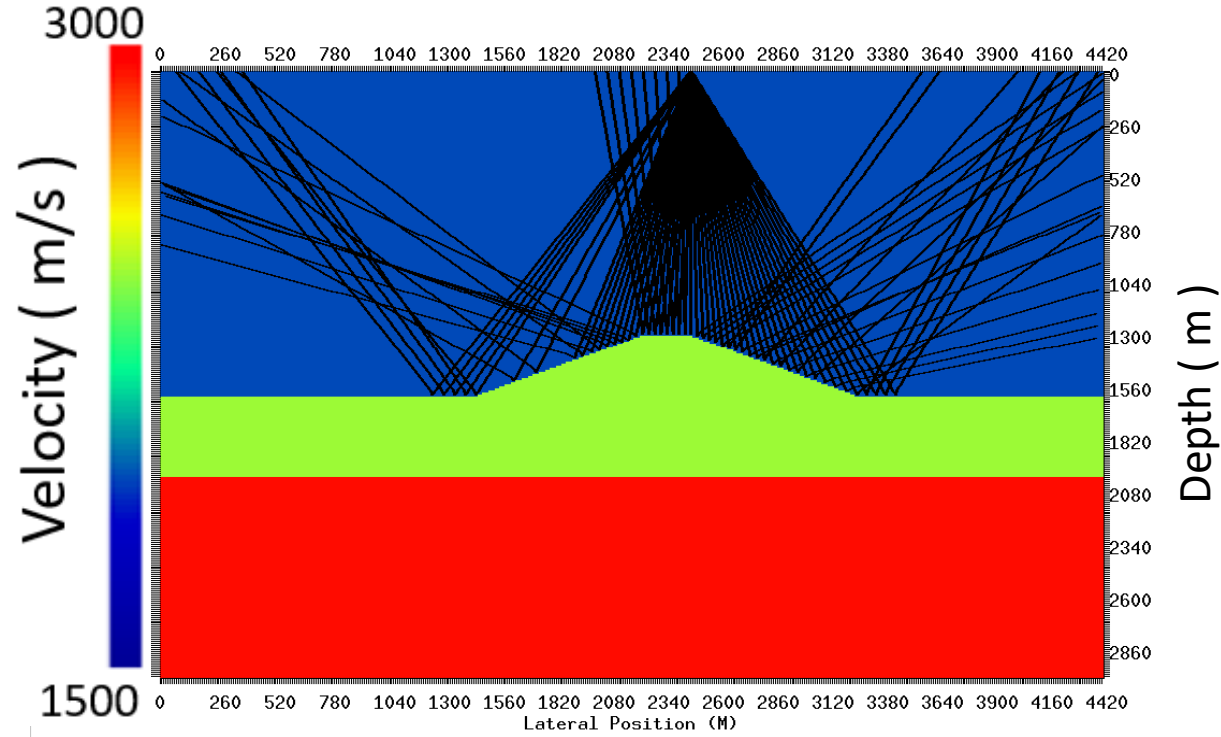


Common Shot Gather

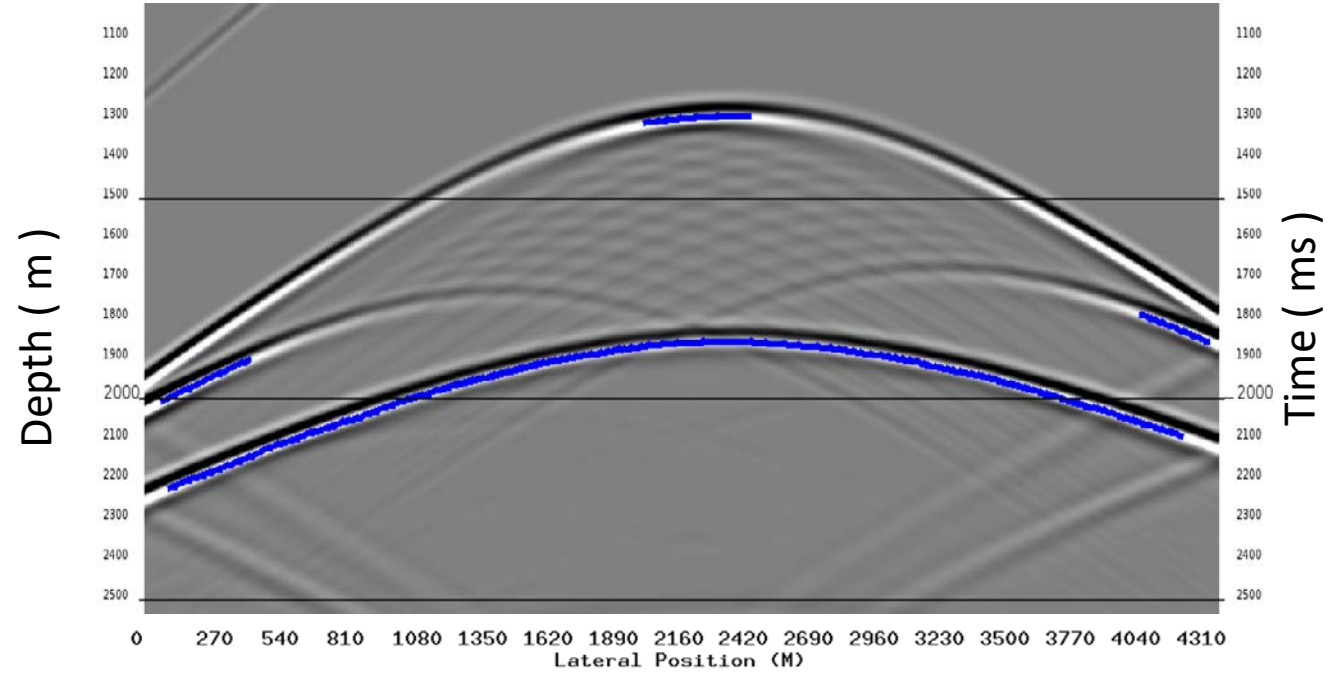




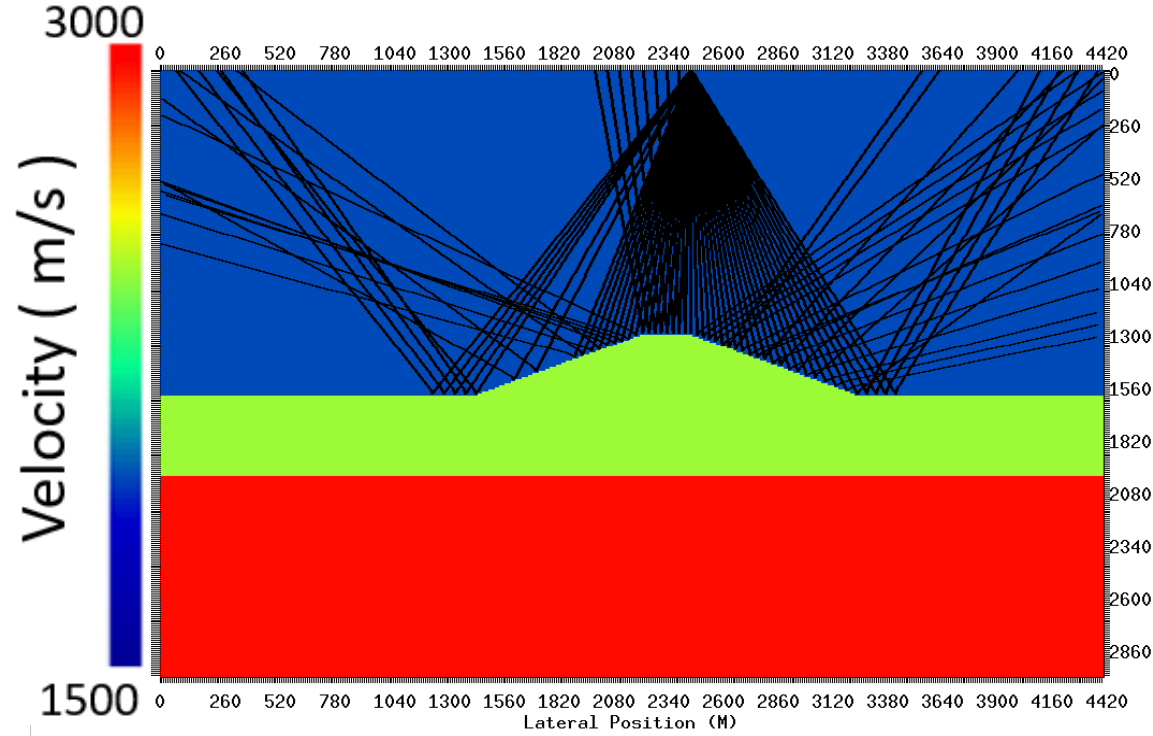
# Traveltime picking difficulties



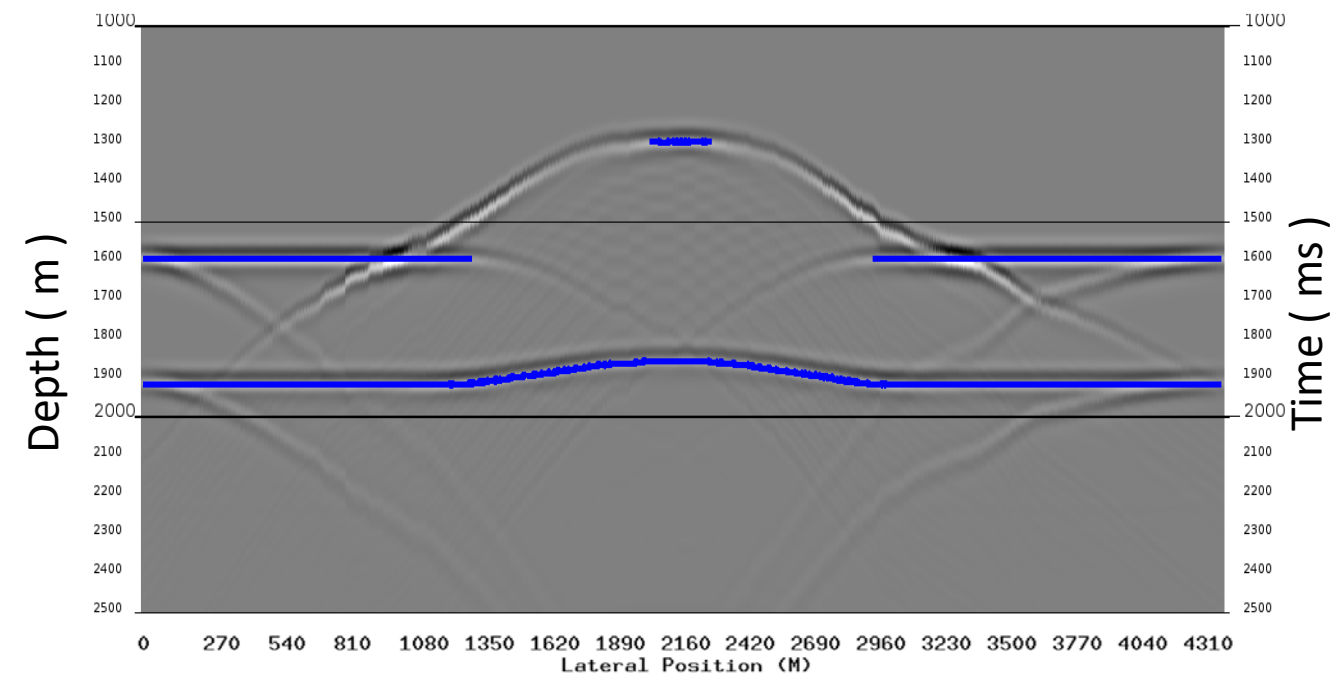
Common Shot Gather



# Traveltime picking difficulties



Common Offset Gather



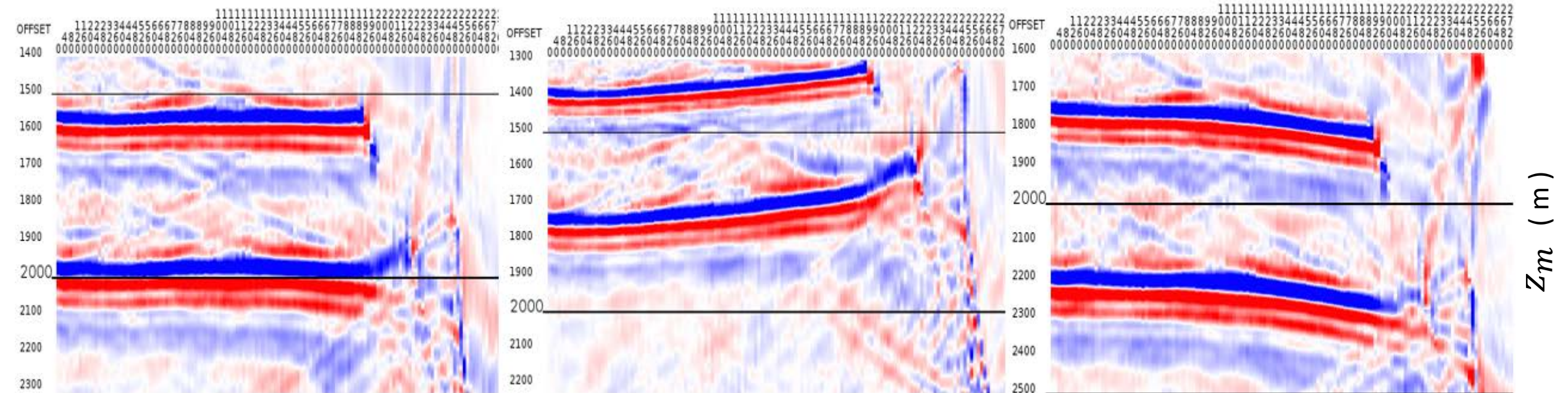


## Residual moveout within a Common Image Gather (CIG)

### True Velocity

### Slow Velocity

### Fast Velocity





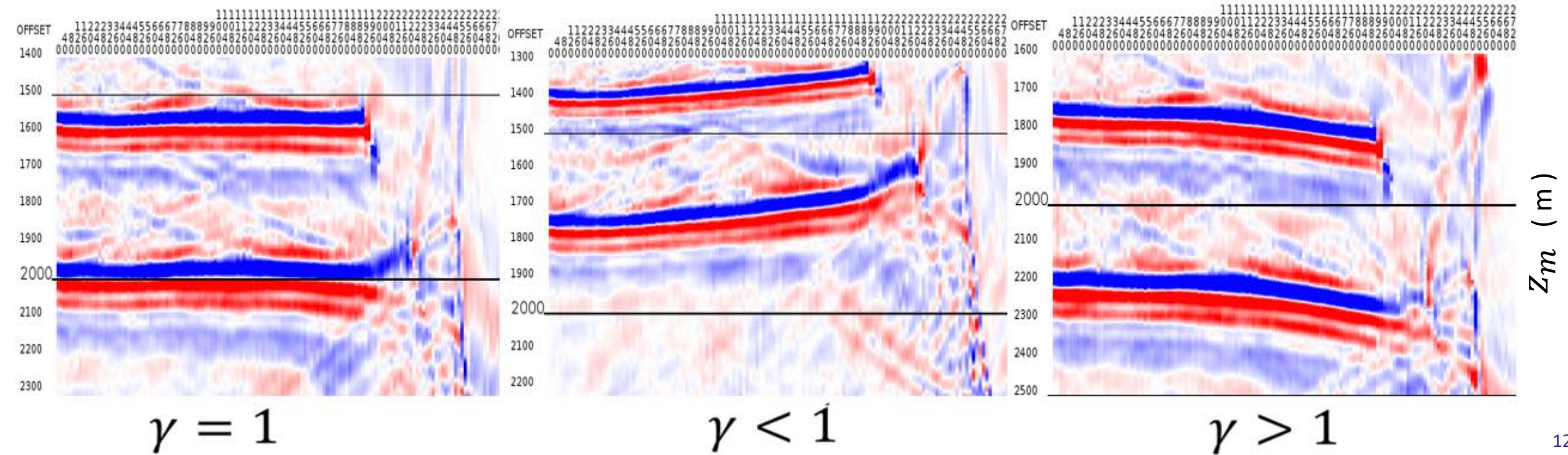
## Residual moveout is scanned automatically

$$z_m = \sqrt{\gamma^2 z^2 + (\gamma^2 - 1)x^2} \quad \gamma = \tilde{V}_m / \tilde{V} \quad \text{Al-Yahya (1989)}$$

### True Velocity

### Slow Velocity

### Fast Velocity

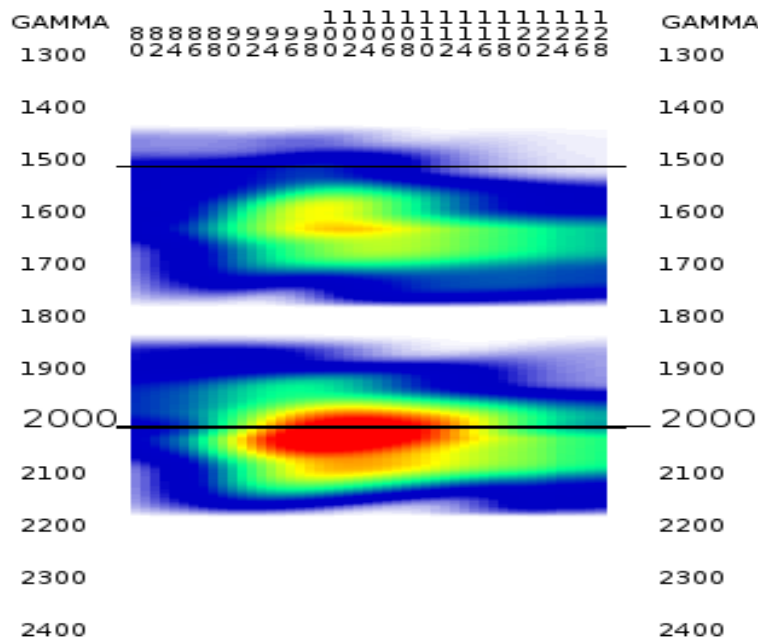




## Residual moveout is scanned automatically

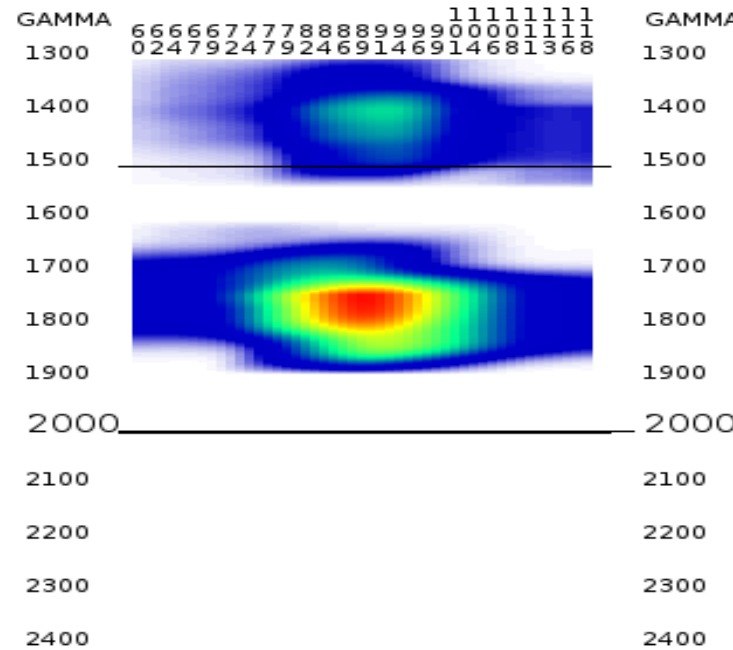
$$z_m = \sqrt{\gamma^2 z^2 + (\gamma^2 - 1)x^2} \quad \gamma = \tilde{V}_m / \tilde{V} \quad \text{Al-Yahya (1989)}$$

### True Velocity



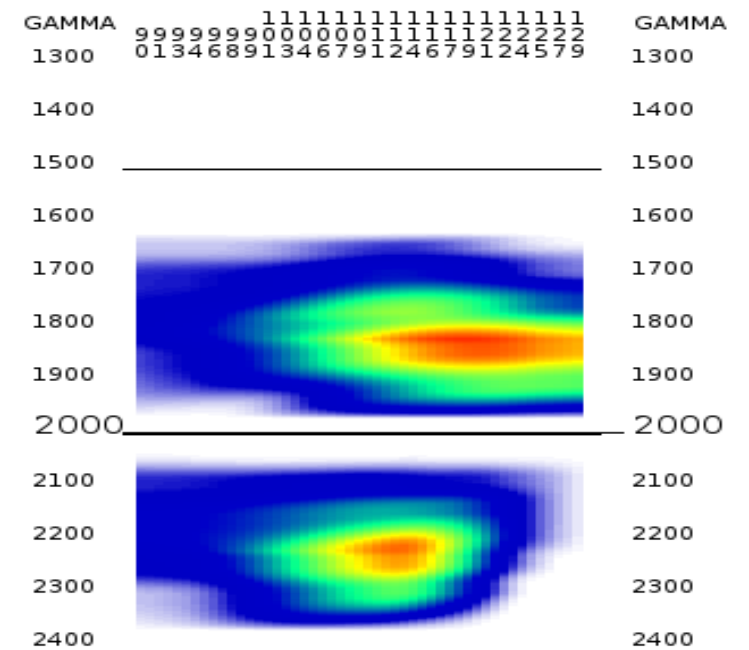
$\gamma = 1$

### Slow Velocity



$\gamma = 0.89$

### Fast Velocity



$\gamma = 1.13$

$z_m$  (m)



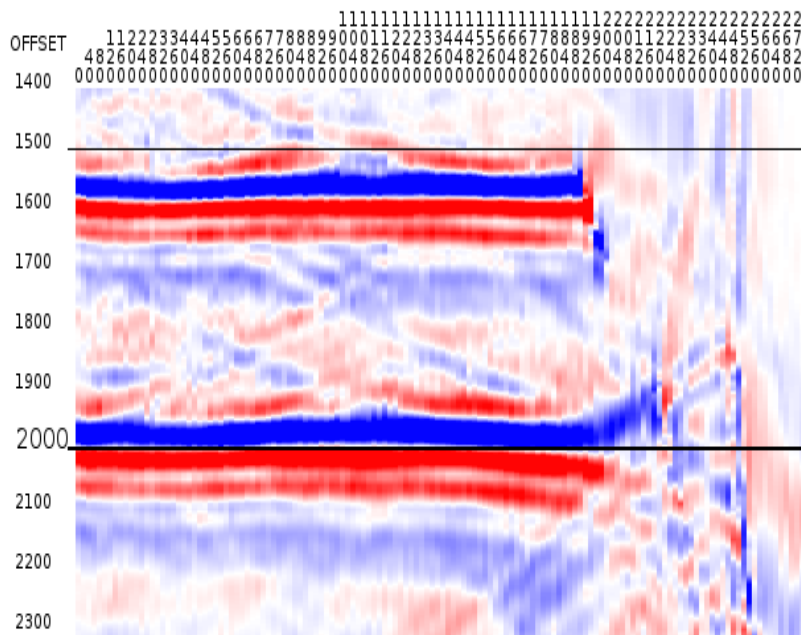
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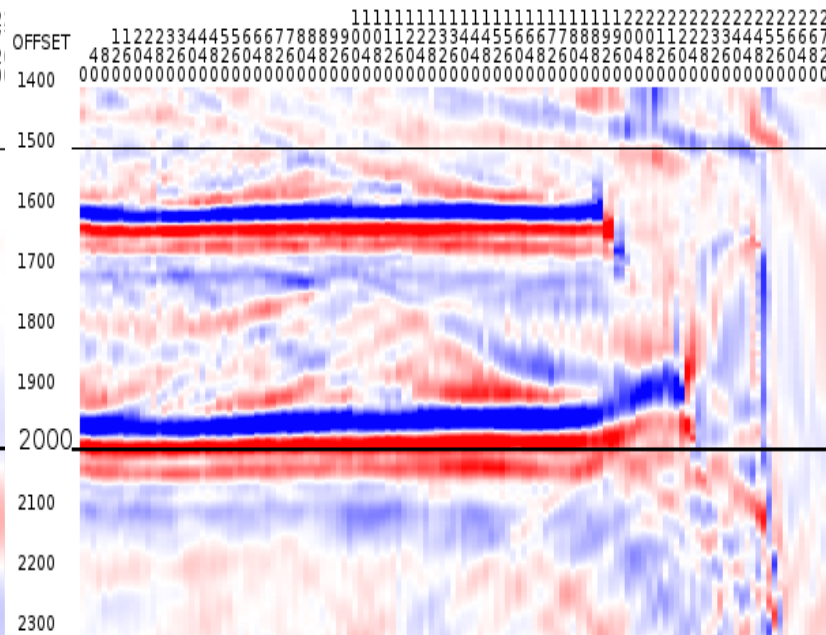
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### Slow Velocity

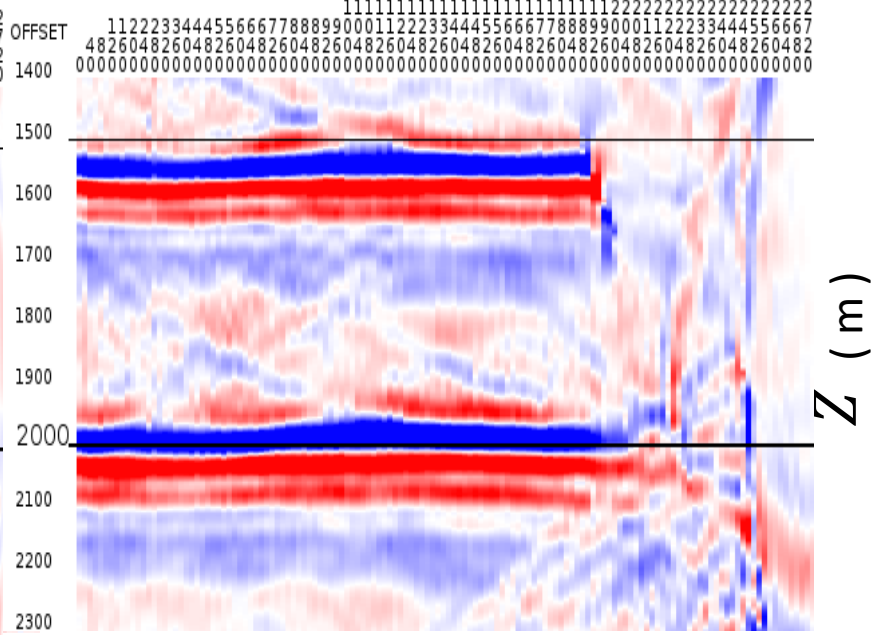
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$\gamma = 1$



$\gamma = 0.89$

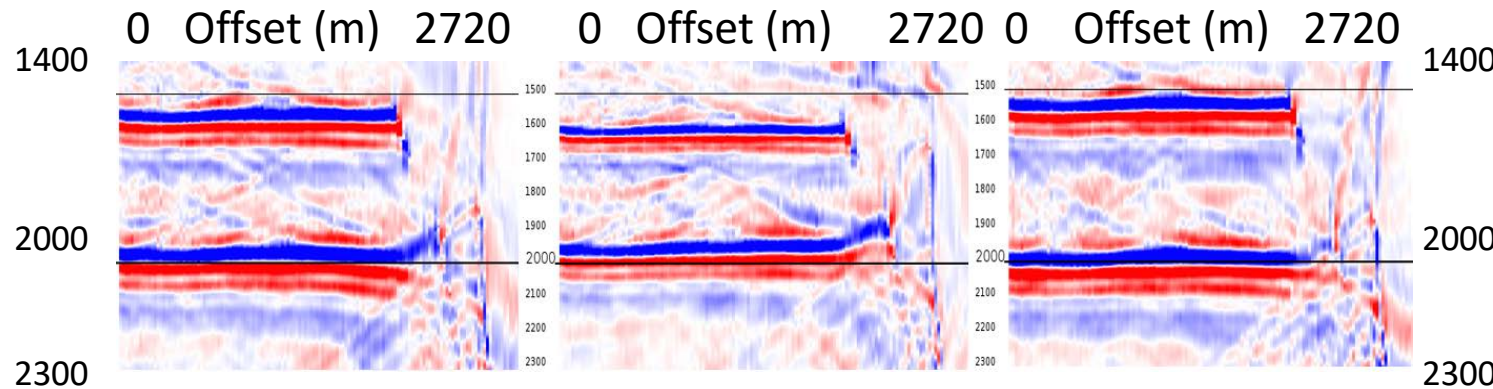
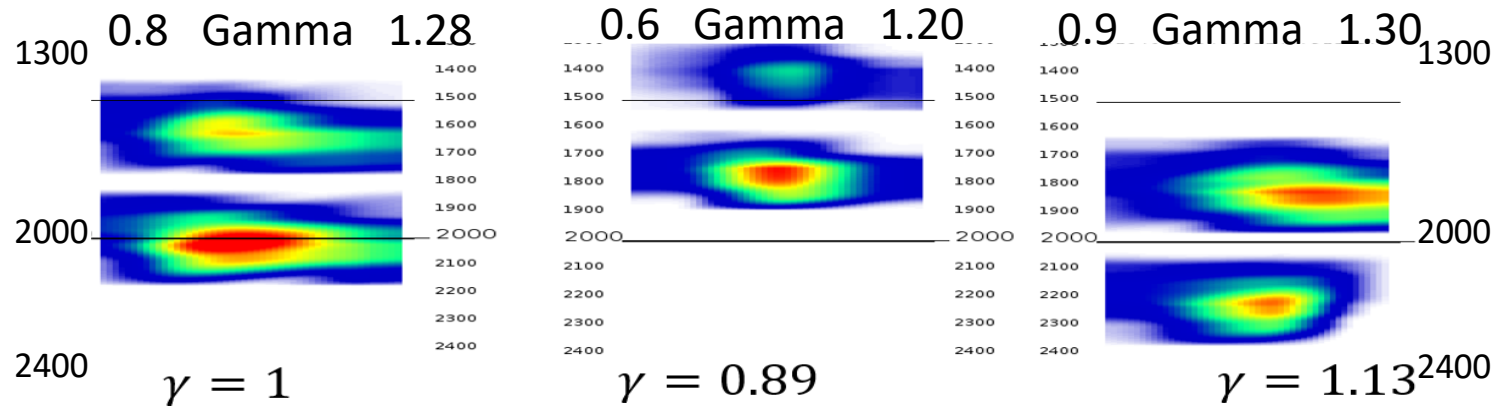
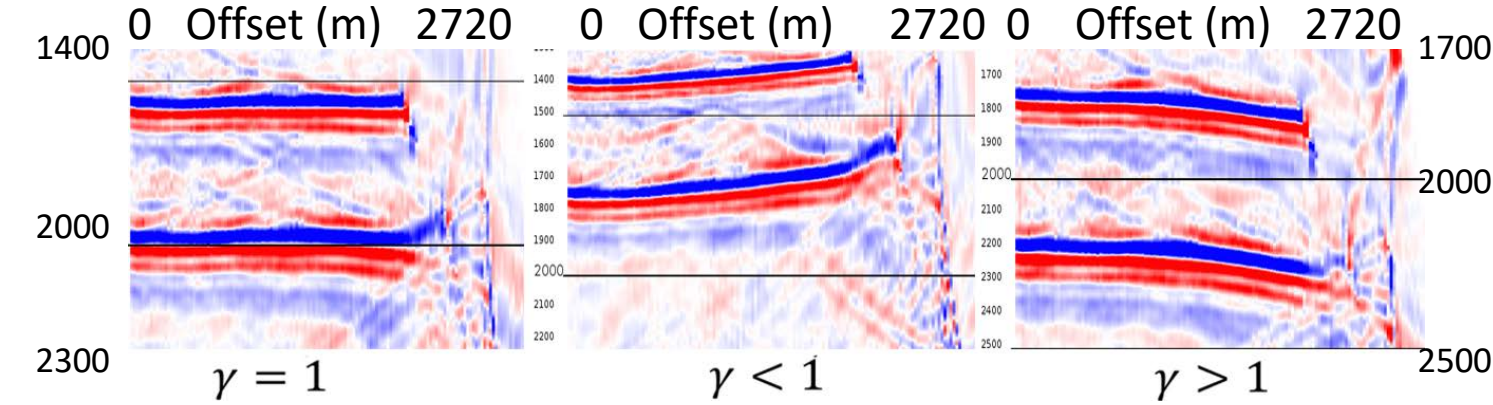


$\gamma = 1.13$

Z (m)



# PSDM tomography



1. Create migrated common image gathers.

Pick analysis windows and locations

Adjust residual moveout to reflect velocity update of upper layers.

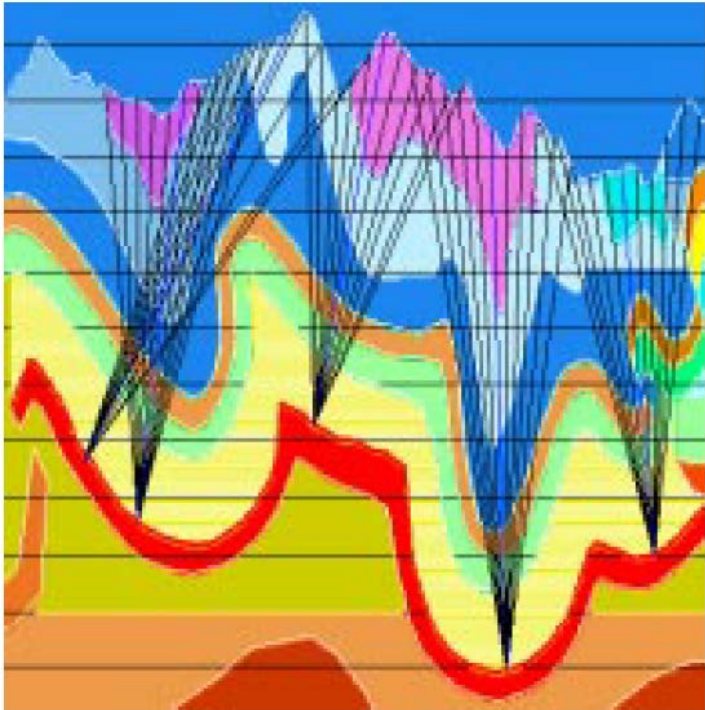
2. Create and pick gamma scans

3. Apply gamma picks to CIG gathers



Update velocity using  $\gamma$

$$\gamma = \tilde{V}_m / \tilde{V}$$



Gray (2000)



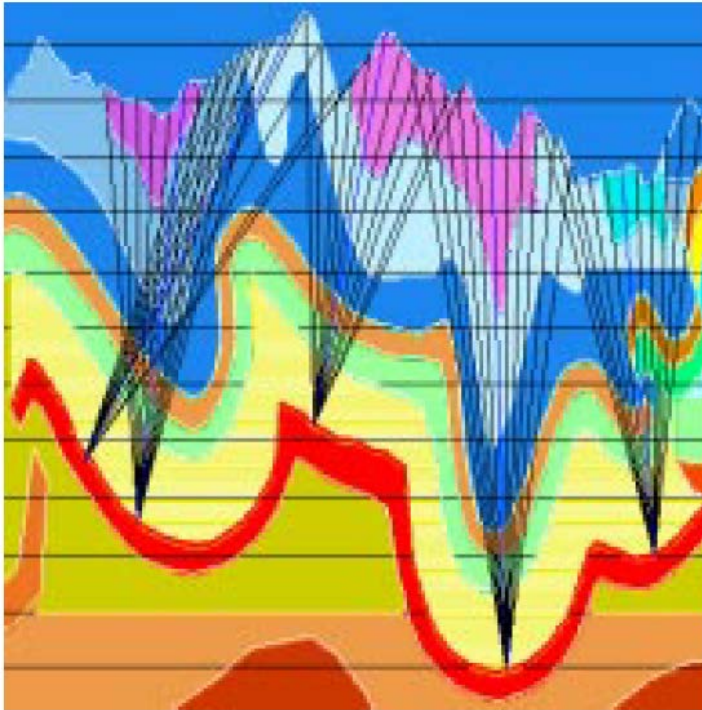


Update velocity using  $\gamma$

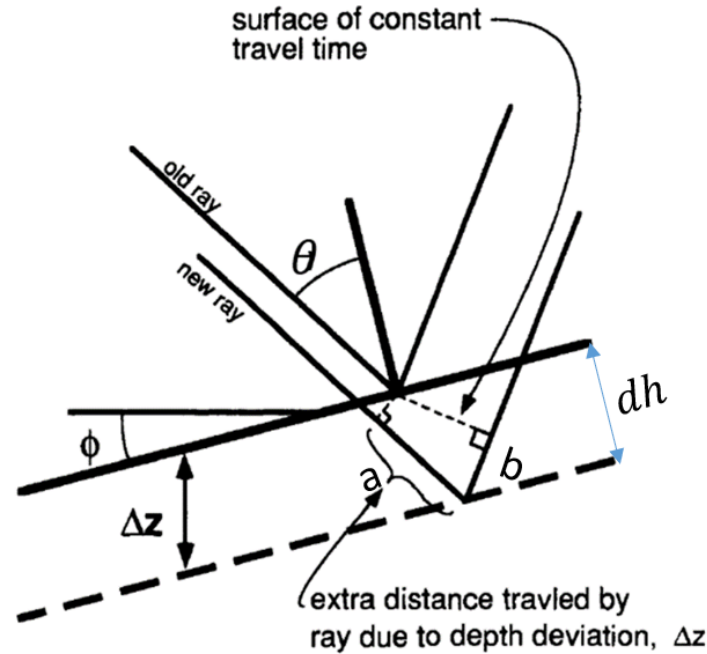
$$\gamma = \tilde{V}_m / \tilde{V}$$

or  $\Delta z$

$$z_m = \sqrt{\gamma^2 z^2 + (\gamma^2 - 1)x^2}$$



Gray (2000)



Stork (1991,1992)

$$\Delta t = (a + b) \cdot s$$

$$\Delta t = 2 \cdot s \cdot dh \cdot \cos(\theta)$$

$$\Delta t = 2 \cdot s \cdot \Delta z \cdot \cos(\phi) \cdot \cos(\theta)$$

$$L \Delta s = \Delta t$$

$$L \Delta s = 2 \cdot s \cdot \Delta z \cdot \cos(\phi) \cdot \cos(\theta)$$

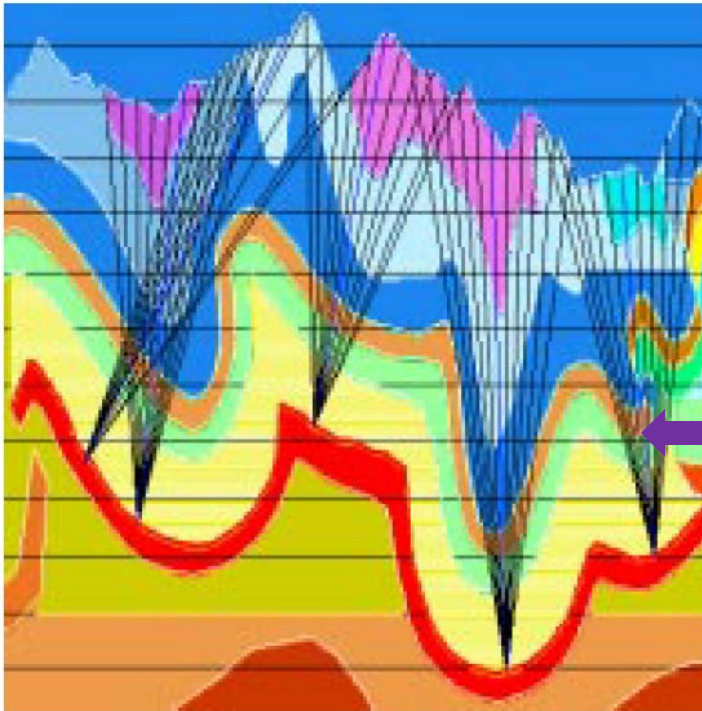


Update velocity using  $\gamma$

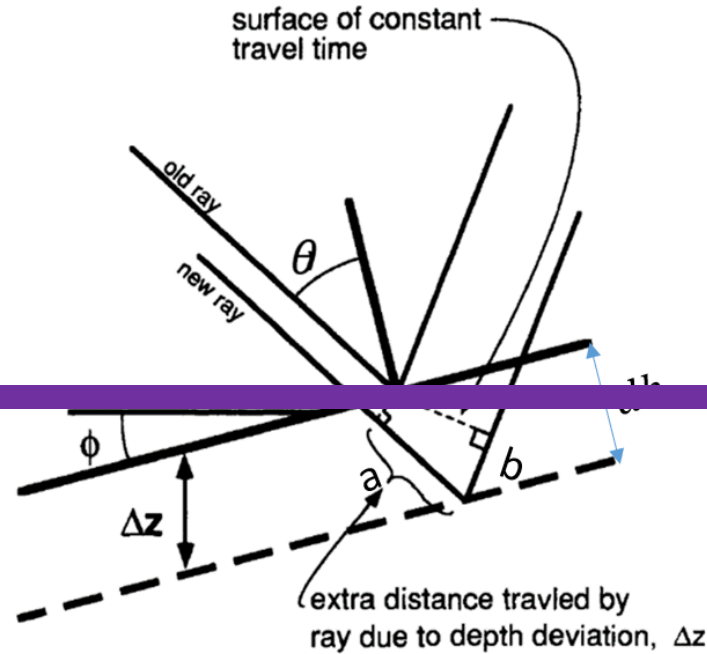
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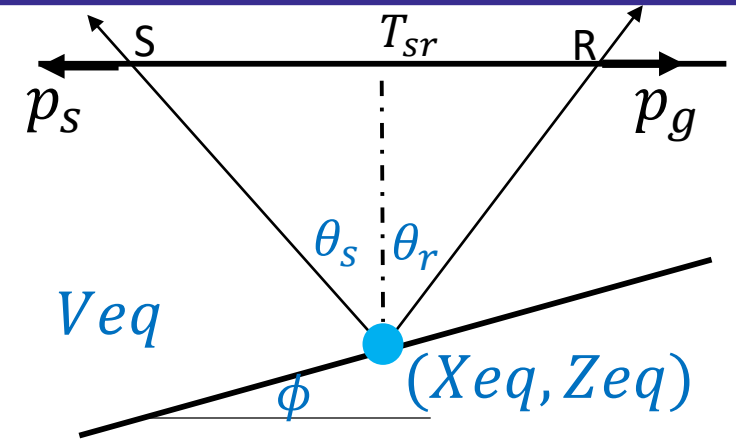
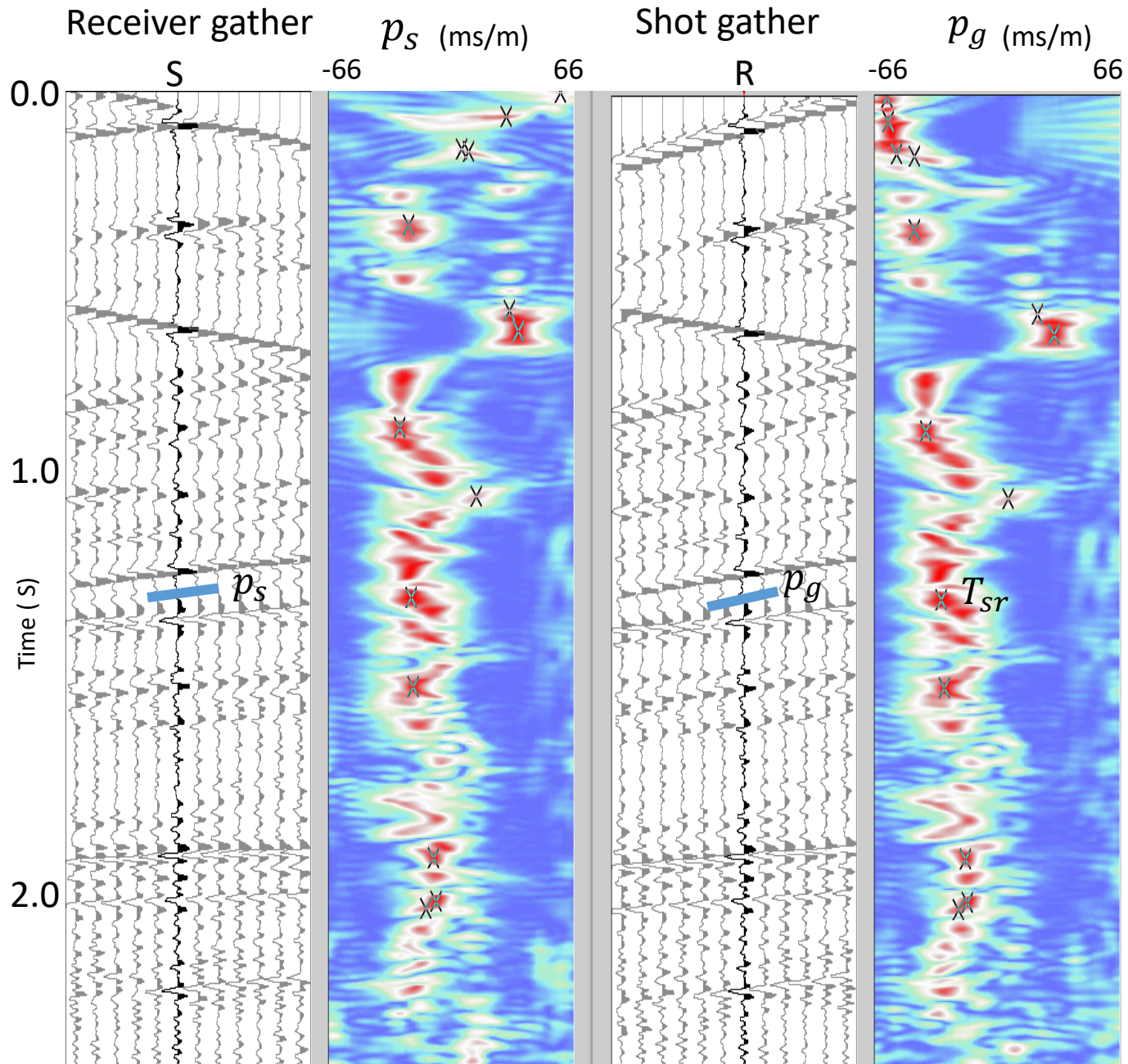
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$$L \Delta s = 2 \cdot s \cdot \Delta z \cdot \cos(\phi) \cdot \cos(\theta)$$



# Stereotomography

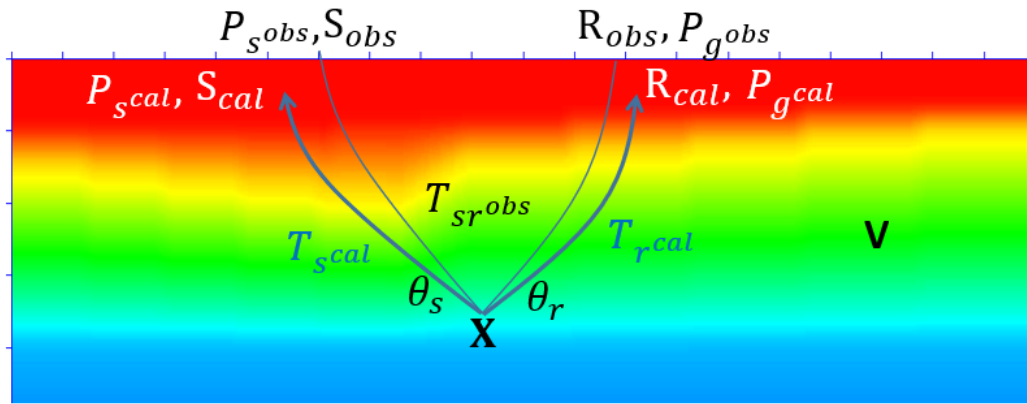


Sword 1987

Billette et al. 1998,2003



# Stereotomography and Adjoint stereotomography



Stereotomography Billette and Lambaré 1998

Model space:  $\mathbf{m} = [(X, \Theta_s, \Theta_r, T_s, T_r)_{i1=1,N}, [V]_{i2=1,M}]$

Data space :  $\mathbf{d} = [S, R, P_s, P_g, T_{sr}]_{j=1,N}$

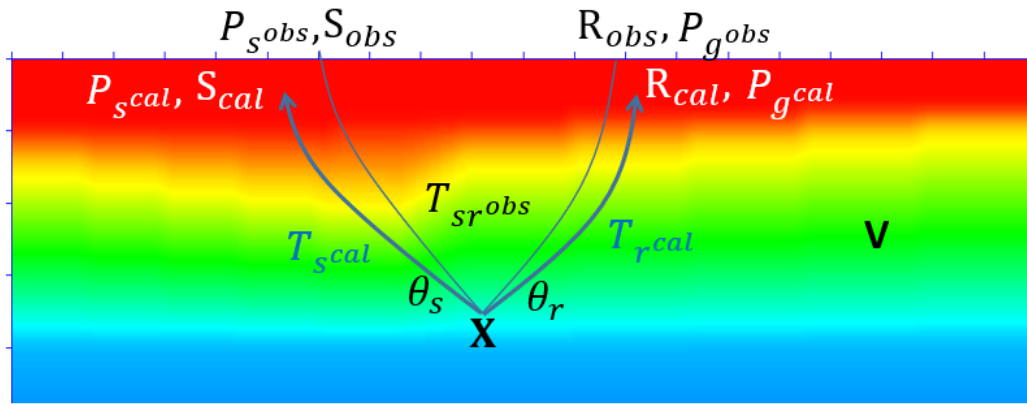
Fréchet derivative:  $A_{ij} = \frac{\partial(S, R, P_s, P_g, T_{sr})}{\partial(X, \Theta_s, \Theta_r, T_s, T_r, V)}$

Inversion:  $\mathbf{A} \Delta \mathbf{m} = \Delta \mathbf{d}$

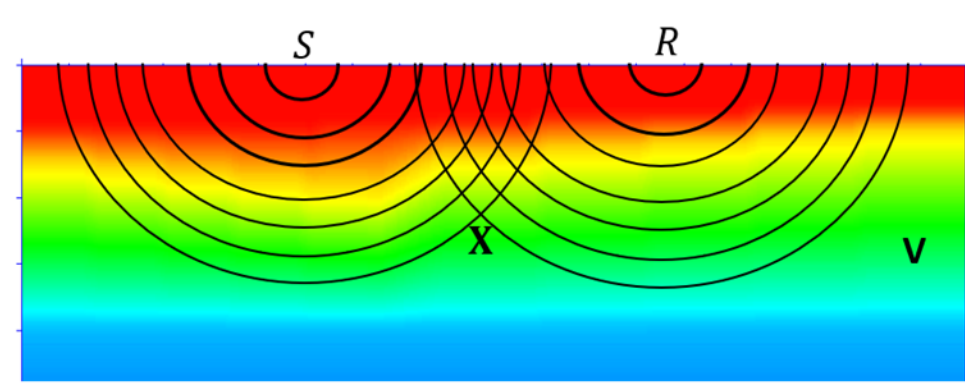
$$\mathbf{A}[6N, 6N+M]$$



# Stereotomography and Adjoint stereotomography



Stereotomography Billette and Lambaré 1998



Adjoint stereotomography Tavakoli 2017

Model space:  $\mathbf{m} = [(X, \theta_s, \theta_r, T_s, T_r)_{i1=1,N}, [V]_{i2=1,M}]$

Data space :  $\mathbf{d} = [S, R, P_s, P_g, T_{sr}]_{j=1,N}$

Fréchet derivative:  $A_{ij} = \frac{\partial(S, R, P_s, P_g, T_{sr})}{\partial(X, \theta_s, \theta_r, T_s, T_r, V)}$

Inversion:  $\mathbf{A} \Delta \mathbf{m} = \Delta \mathbf{d}$

$\mathbf{A} [6N, 6N+M]$

Model space :  $[X_{j=1,N}, [V]_{i=1,M}]$

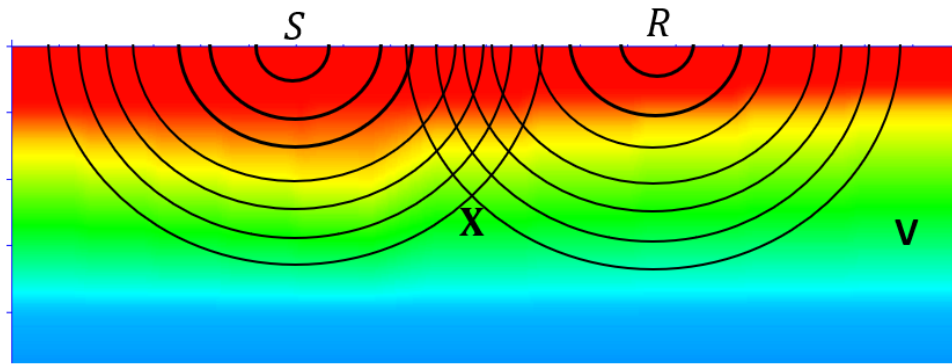
Data space :  $[T_{sr}, P_s, P_g]_{j=1,N}$

Inversion :  $\mathbf{m}_{k+1} = \mathbf{m}_k + \alpha_k \frac{\partial J}{\partial \mathbf{m}}$

$\frac{\partial J}{\partial \mathbf{m}} [2N+M]$



# Adjoint stereotomography



Model space :  $[X_{j=1,N}], [V]_{i=1,M}$

Data space :  $[T_{sr}, P_s, P_g]_{j=1,N}$

Inversion :  $\mathbf{m}_{k+1} = \mathbf{m}_k + \alpha_k \frac{\partial J}{\partial \mathbf{m}}$

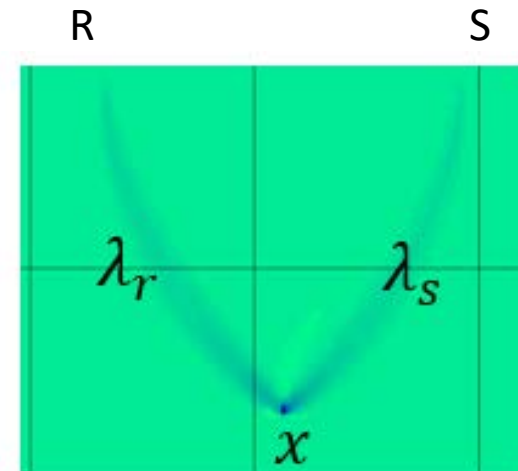
Adjoint Stereotomography Tavakoli 2017

$$\frac{\partial J}{\partial v(x)} = - \frac{1}{v(x)^3} \sum (\lambda_r + \lambda_s)$$

$\lambda_s, \lambda_r$  = adjoint state variables correspond to traveltimes for  $T_s$  and  $T_r$

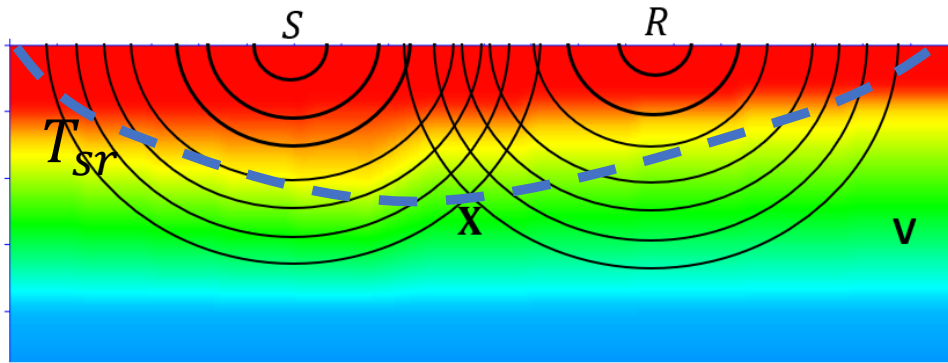
$$(\partial / \partial x, \partial / \partial z) \cdot (\lambda_s \cdot \nabla T_s) = S(\Delta T_{sr}, \Delta P_s)$$

$$(\partial / \partial x, \partial / \partial z) \cdot (\lambda_r \cdot \nabla T_r) = S(\Delta T_{sr}, \Delta P_g)$$





# Adjoint stereotomography



Model space :  $[X_{j=1,N}], [V]_{i=1,M}$

Data space :  $[T_{sr}, P_s, P_g]_{j=1,N}$

Inversion :  $\mathbf{m}_{k+1} = \mathbf{m}_k + \alpha_k \frac{\partial J}{\partial \mathbf{m}}$

Adjoint Stereotomography Tavakoli 2017

$$\frac{\partial J}{\partial X} = \Delta T_{sr} \nabla(T_s + T_r) + \frac{\Delta p_s}{2\Delta s} \nabla(T_{s+1} - T_{s-1}) + \frac{\Delta p_g}{2\Delta r} \nabla(T_{r+1} - T_{r-1})$$

Normal to the wavefront

$T_{sr}$

Move  $x$  normal to  $T_{sr}$

Normal to source- $x$  ray path

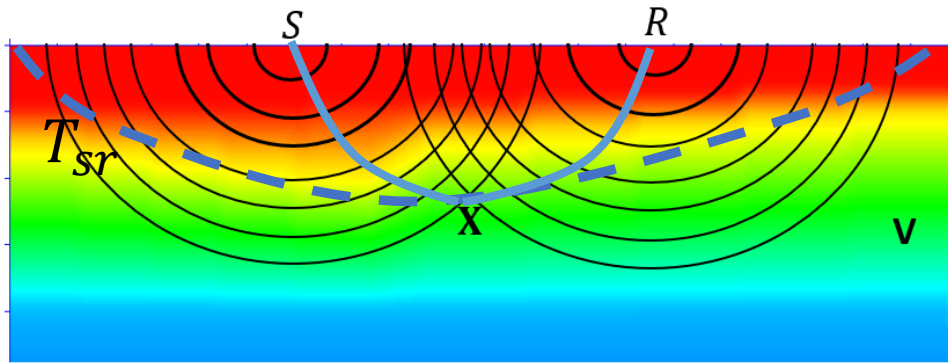
Move  $x$  normal to source- $x$  ray path

Normal to receiver- $x$  ray path

Move  $x$  normal to receiver- $x$  ray path



# Adjoint stereotomography



Model space :  $[X_{j=1,N}], [V]_{i=1,M}$

Data space :  $[T_{sr}, P_s, P_g]_{j=1,N}$

Inversion :  $\mathbf{m}_{k+1} = \mathbf{m}_k + \alpha_k \frac{\partial J}{\partial \mathbf{m}}$

Adjoint Stereotomography Tavakoli 2017

$$\frac{\partial J}{\partial X} = \Delta T_{sr} \nabla(T_s + T_r) + \frac{\Delta p_s}{2\Delta s} \nabla(T_{s+1} - T_{s-1}) + \frac{\Delta p_g}{2\Delta r} \nabla(T_{r+1} - T_{r-1})$$

Normal to the wavefront

$T_{sr}$

Move  $x$  normal to  $T_{sr}$

Normal to source- $x$  ray path

Move  $x$  normal to source- $x$  ray path

Normal to receiver- $x$  ray path

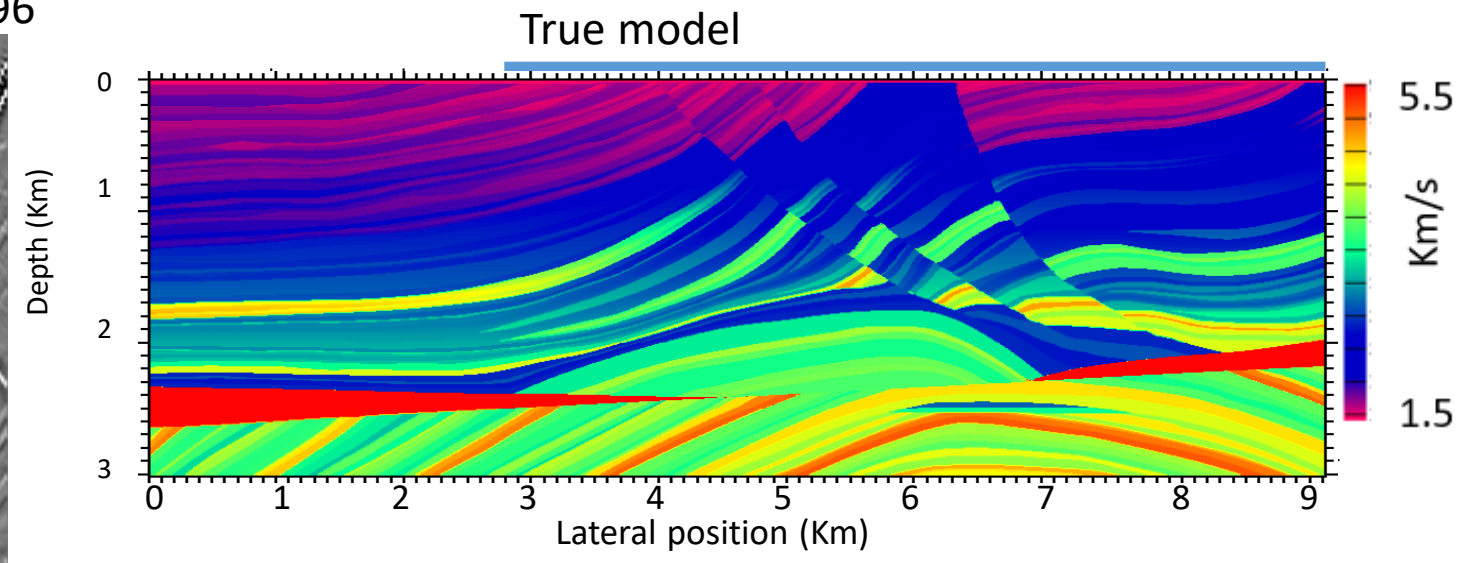
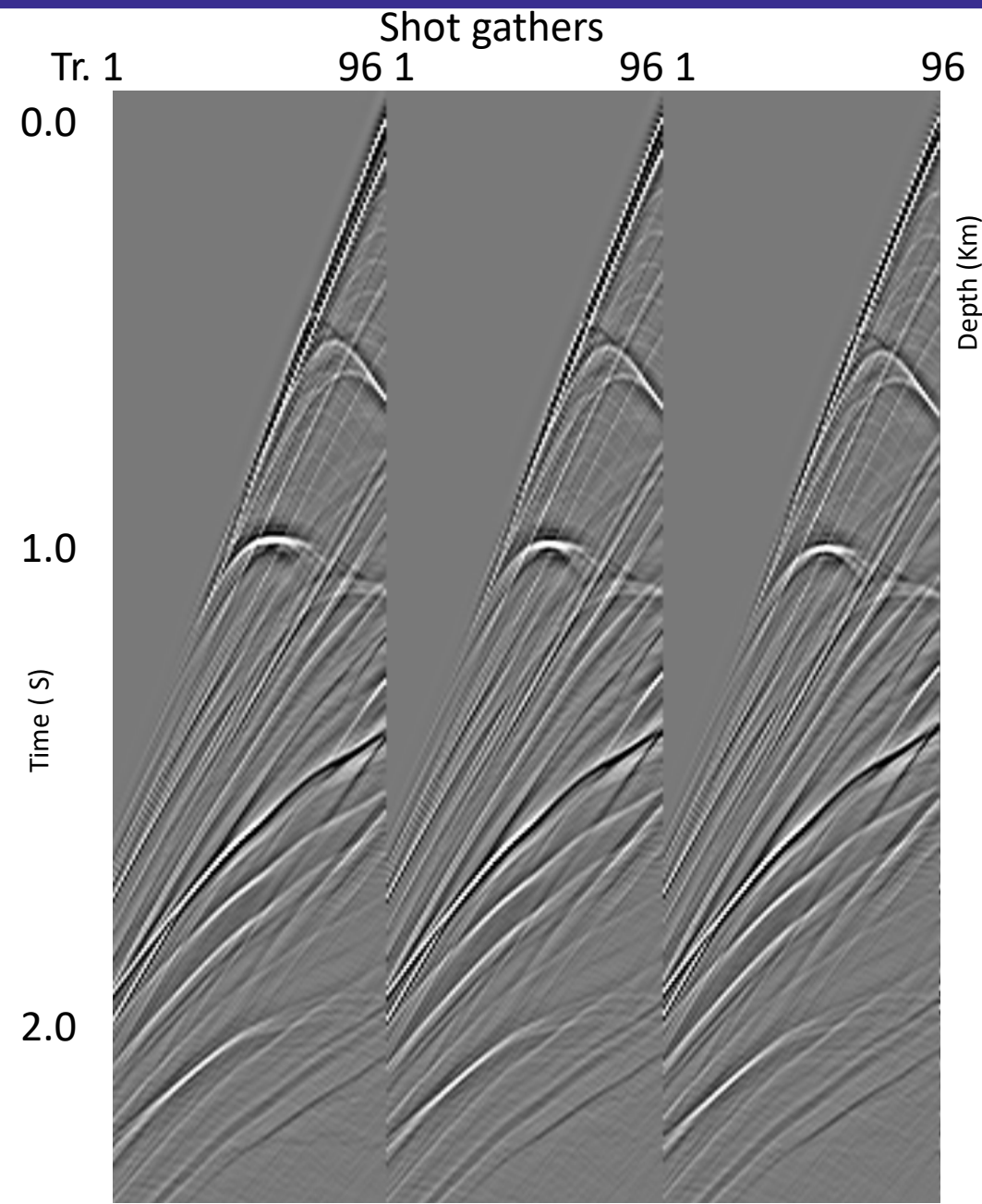
Move  $x$  normal to receiver- $x$  ray path

$$\Delta x = \alpha_x \frac{\partial J}{\partial x}$$





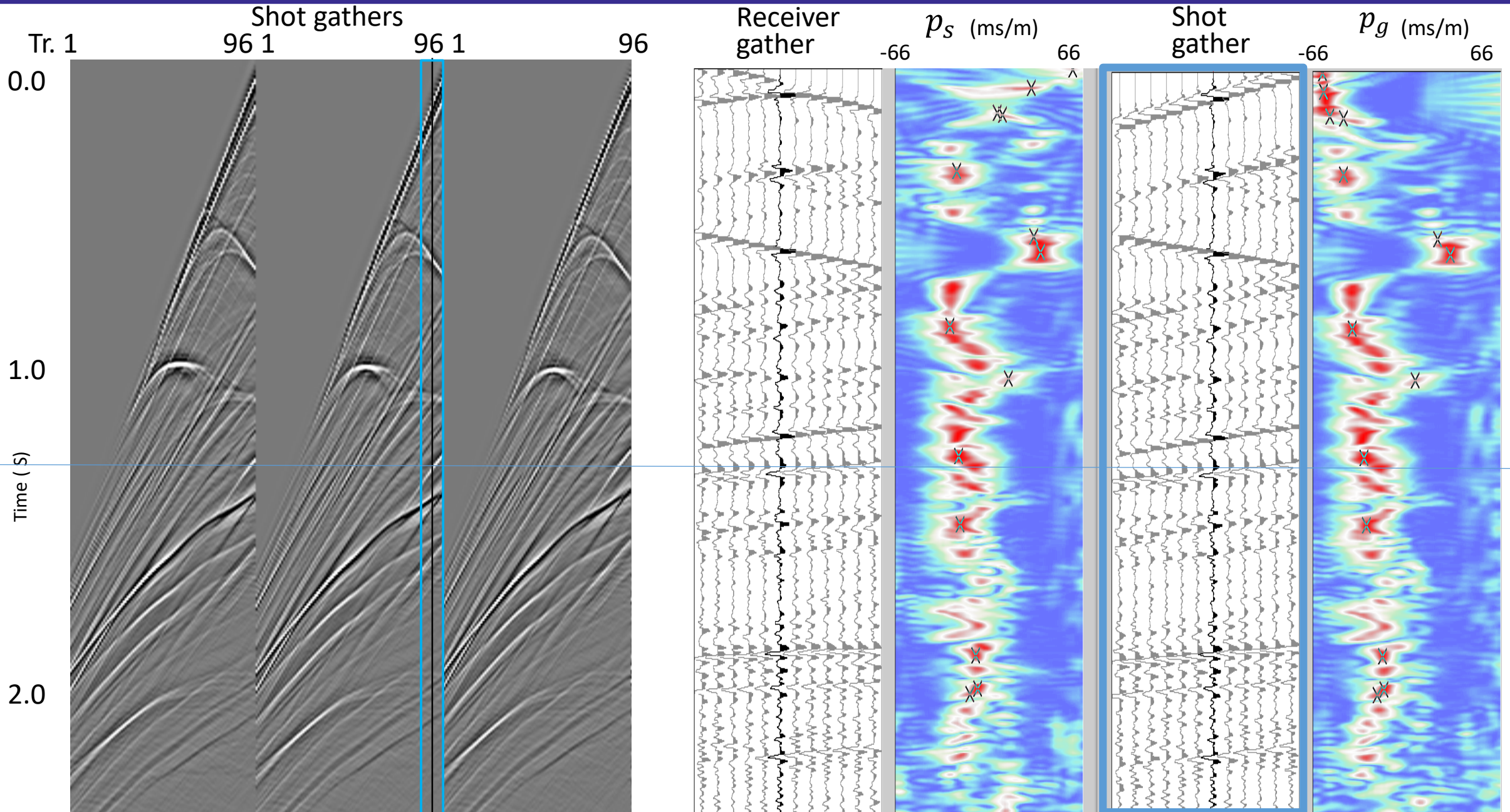
# Synthetic example



- 300 shots
- 96 channels per shot
- 25 m shot and receiver spacing

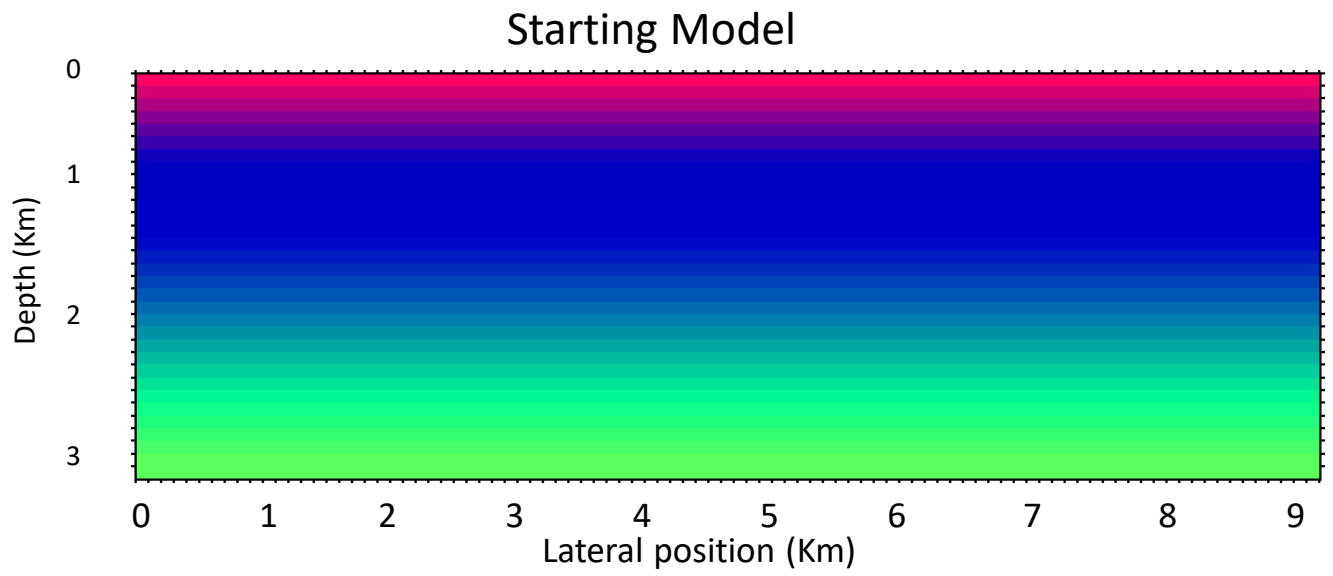
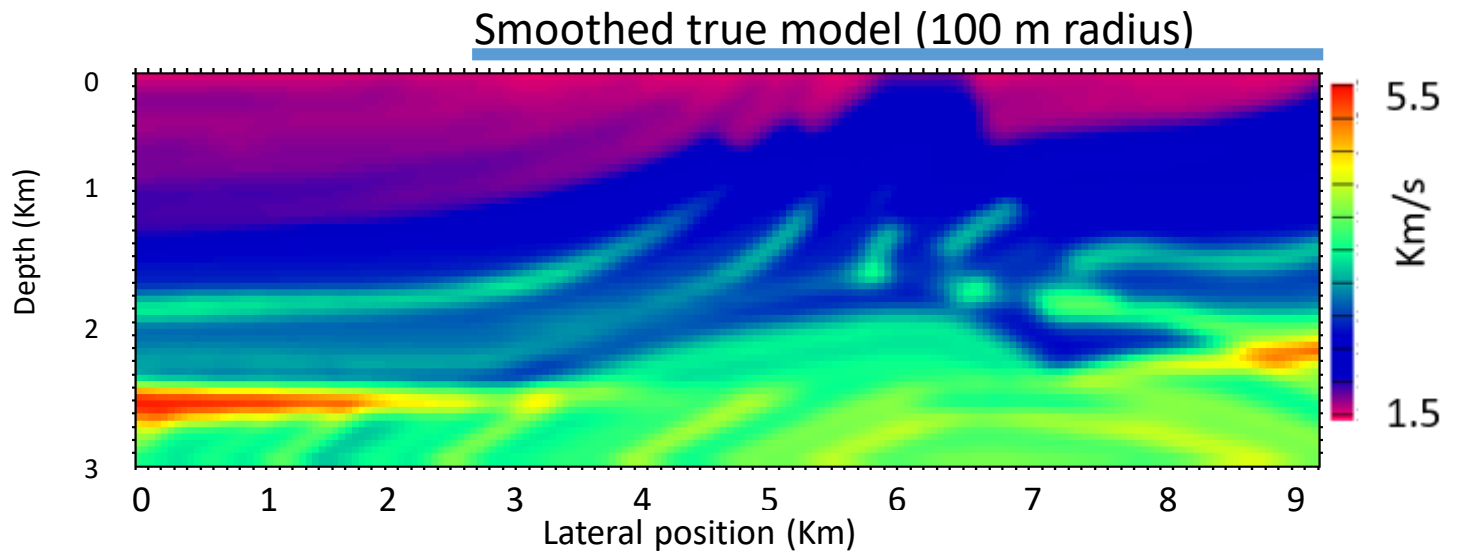
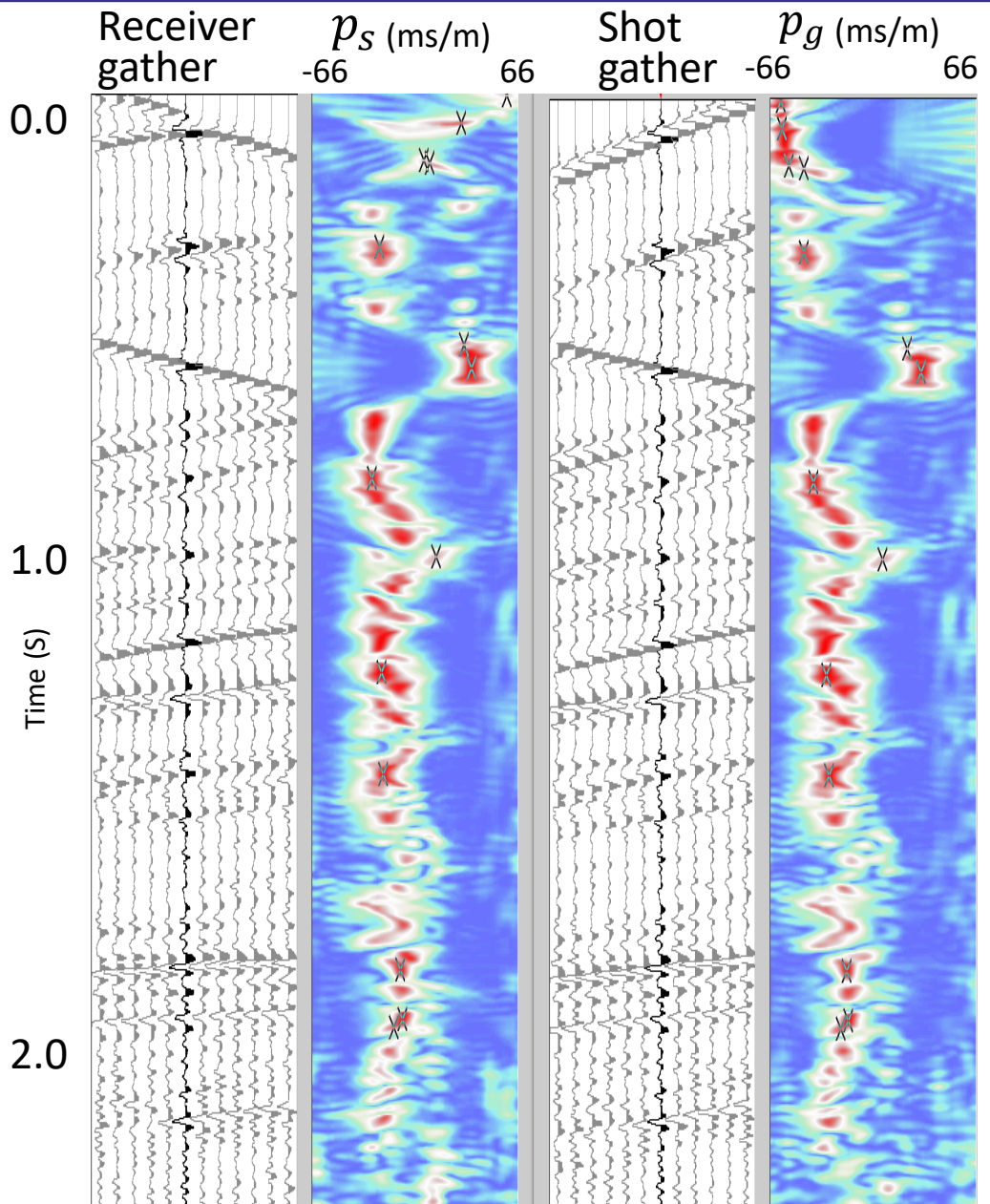


# Synthetic example





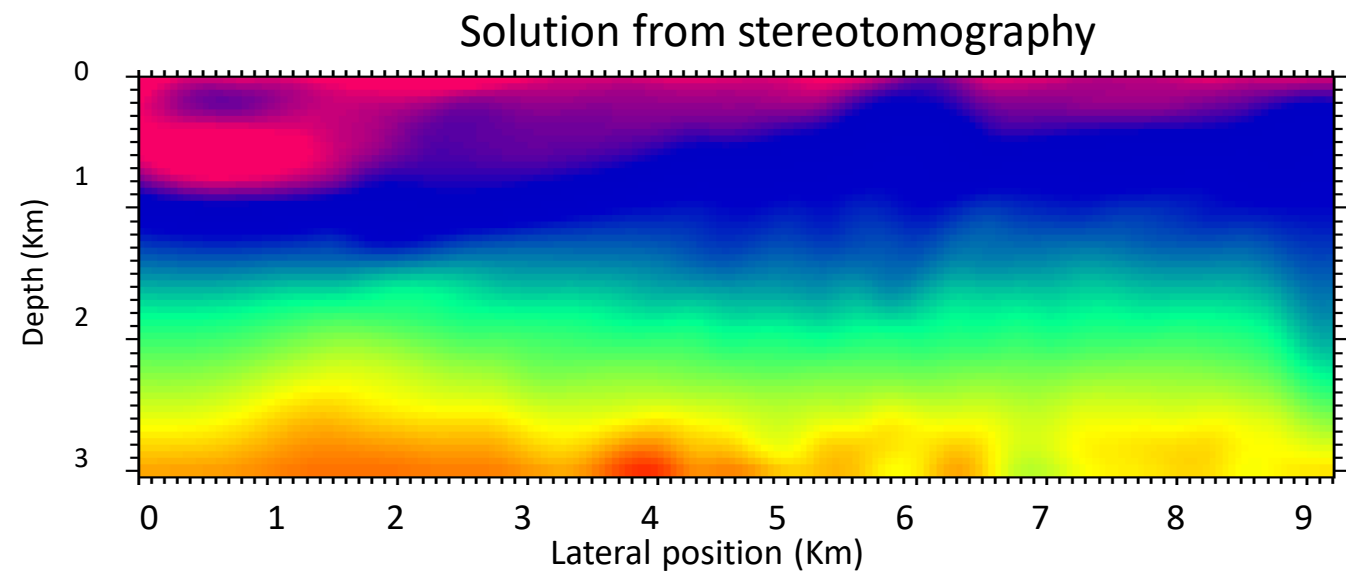
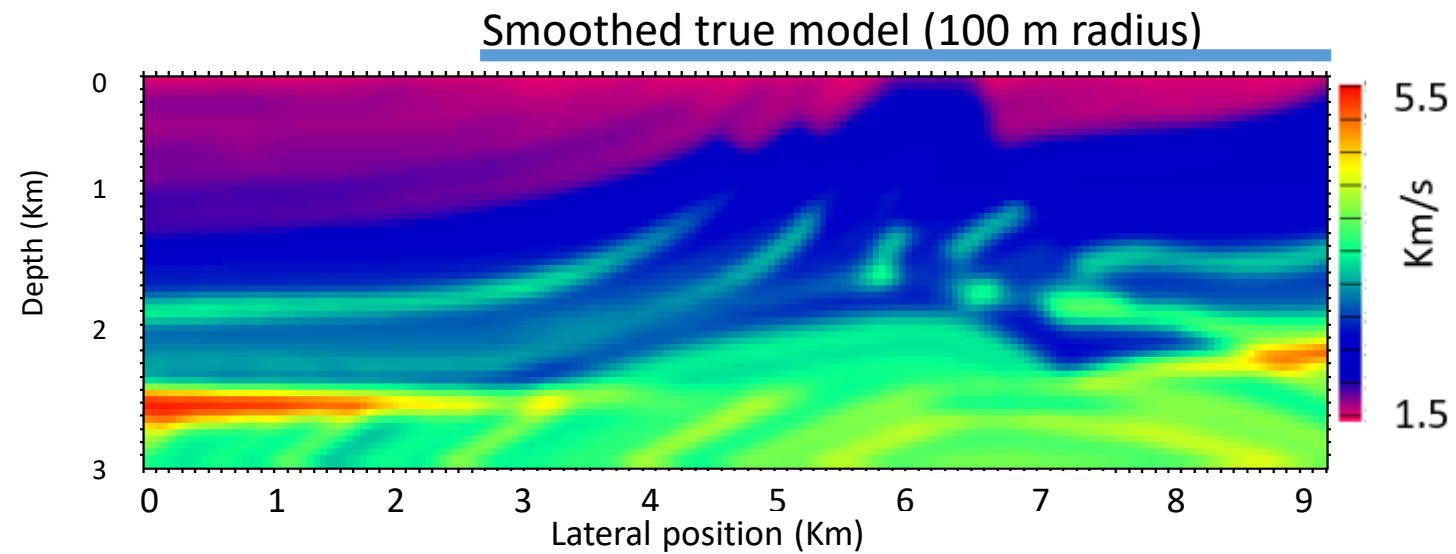
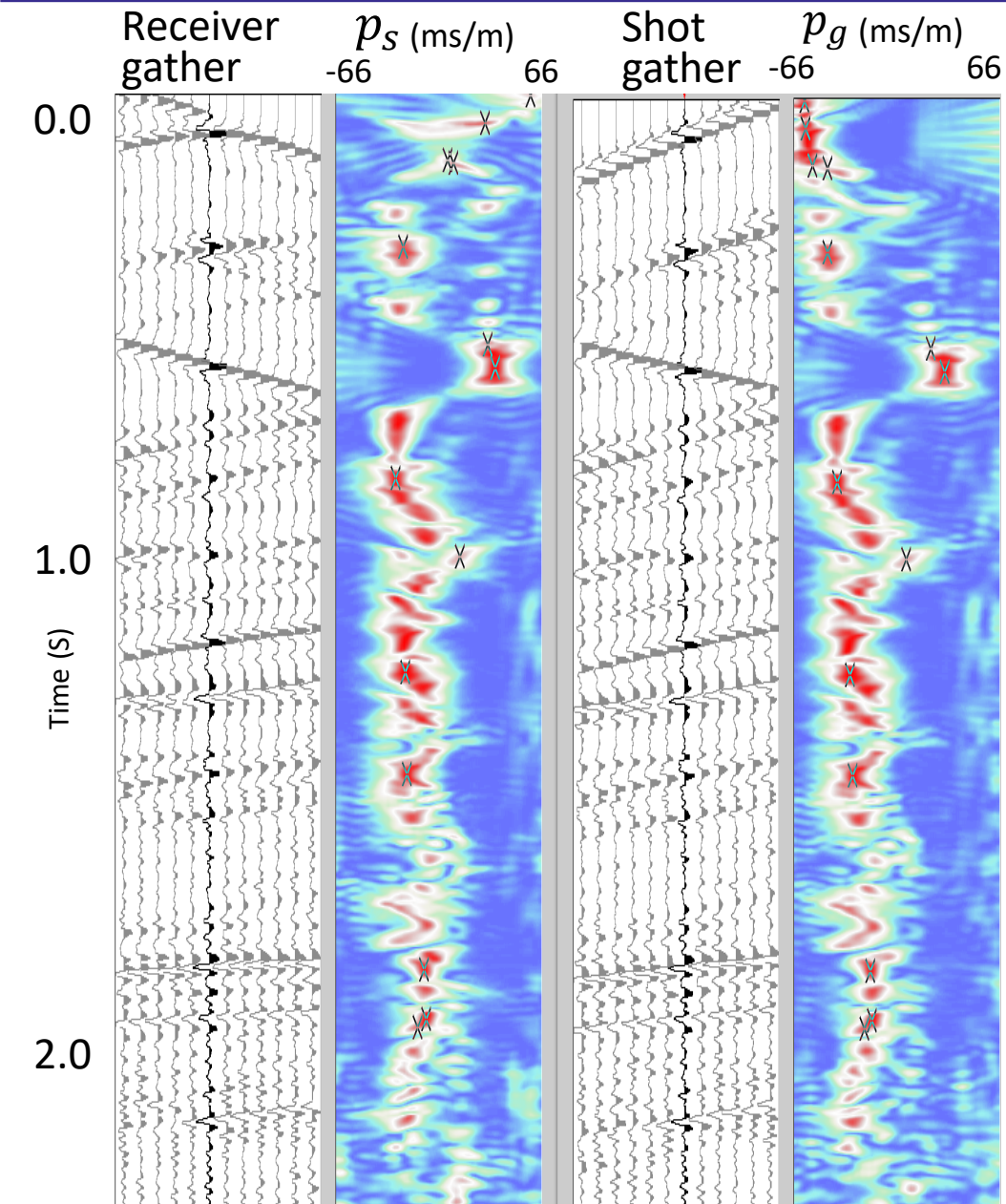
# Synthetic example





# Synthetic example

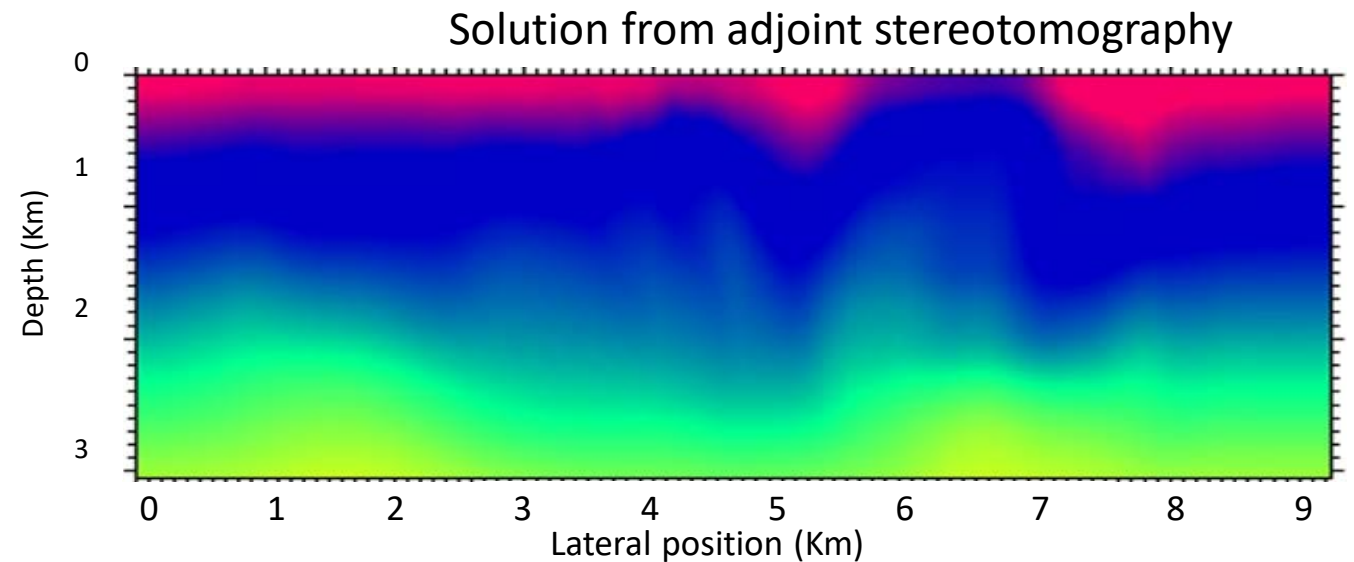
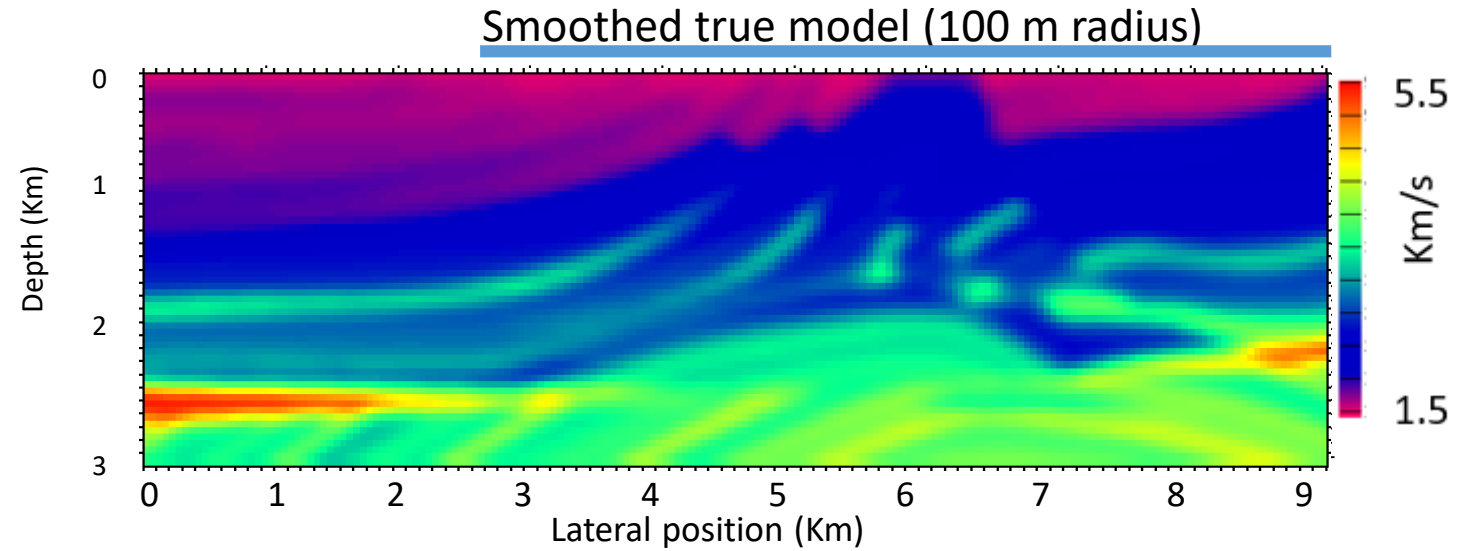
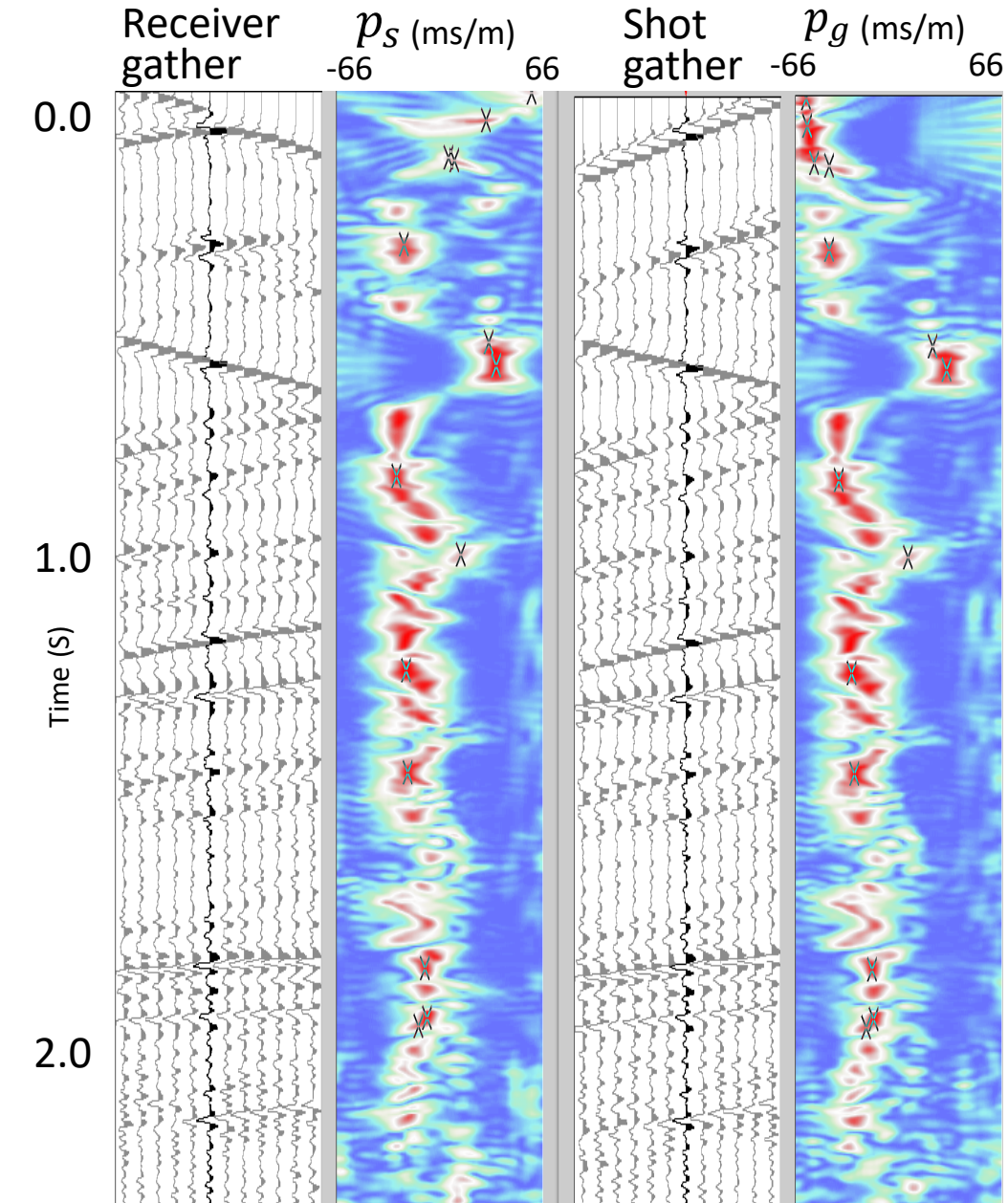
# Stereotomography





# Synthetic example

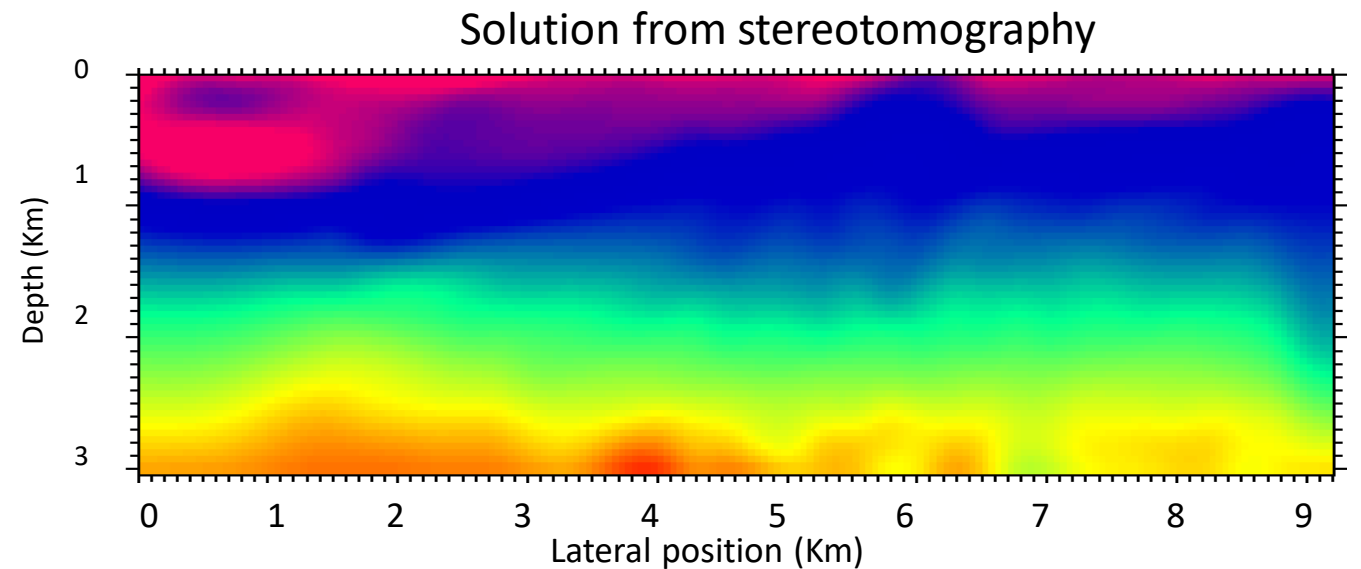
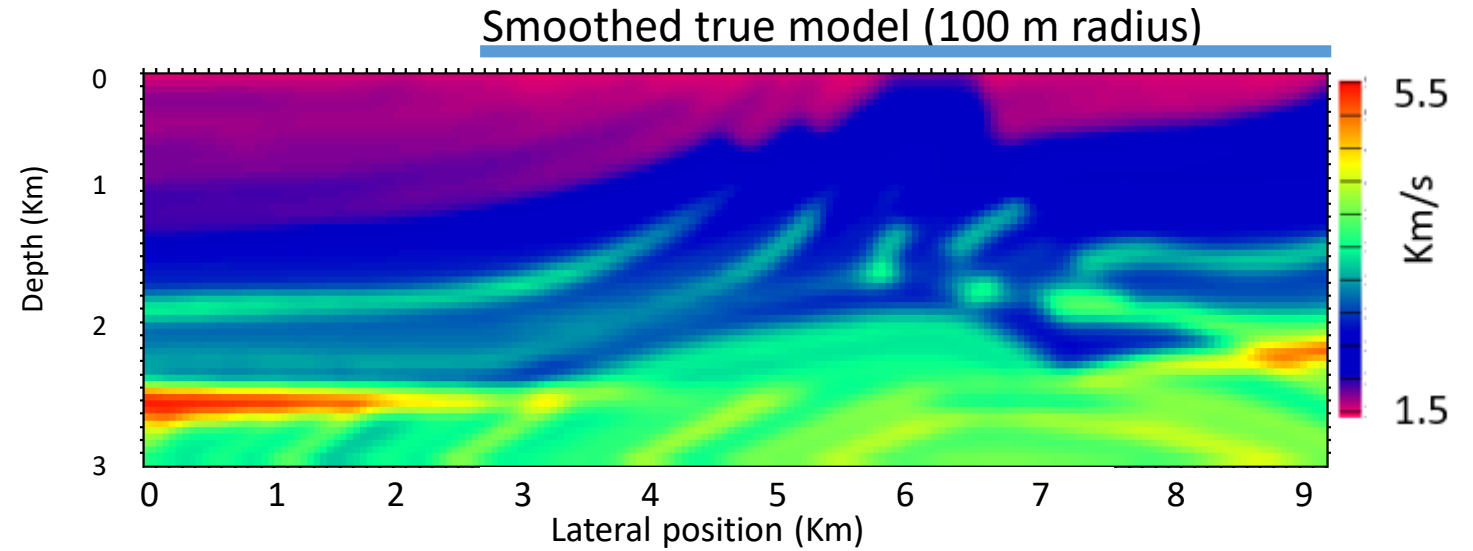
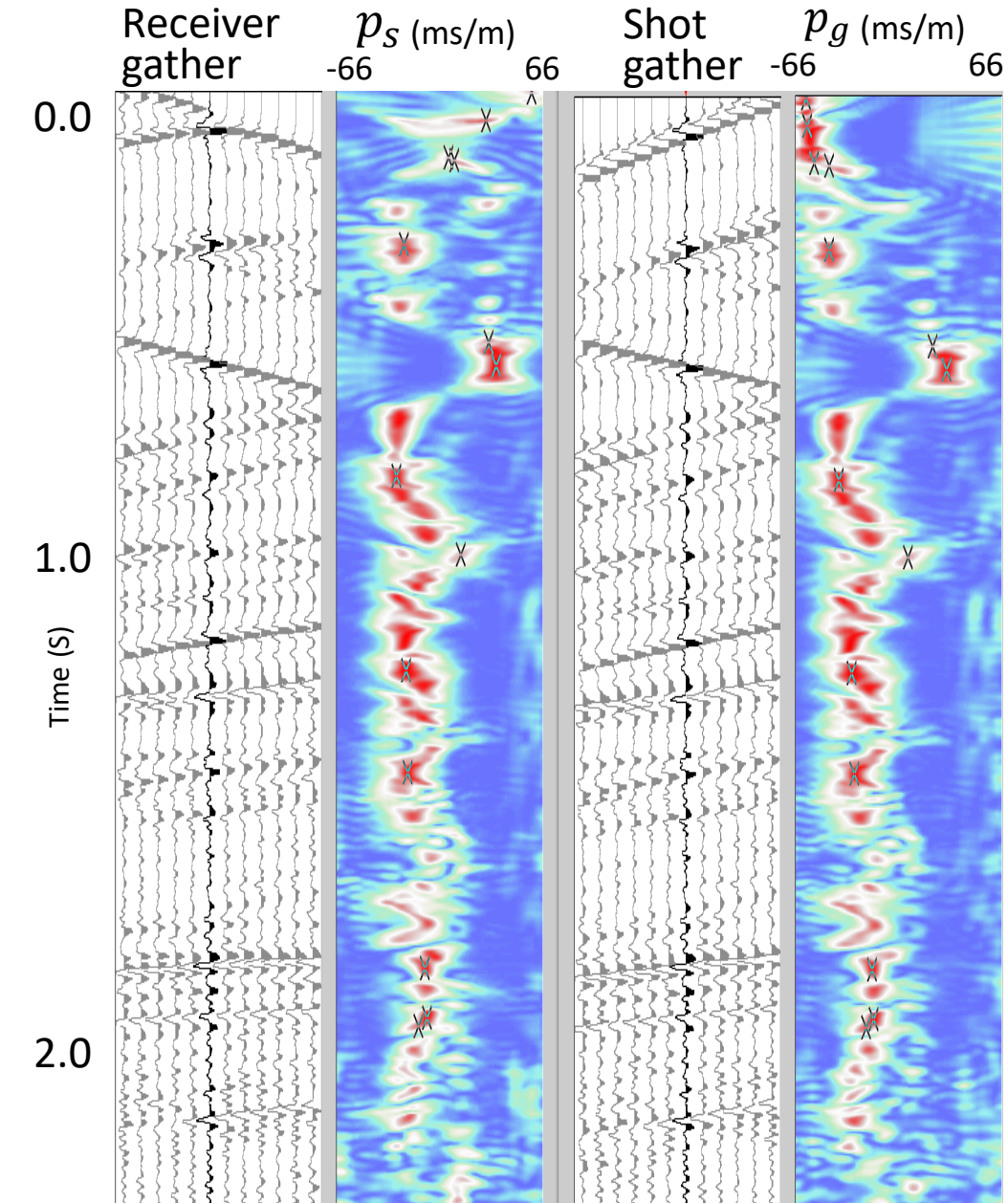
# Adjoint stereotomography





# Synthetic example

# Stereotomography





- Reviewed the classical reflection tomography, PSDM tomography and stereotomography
- PSDM tomography and stereotomography has picking advantage over the classical reflection tomography
- Stereotomography methods may be more efficient in building a macro velocity model than PSDM tomography, because depth migration step is not required
- We found the solution from our implementation of adjoint stereotomography matches the long wavelength component of the reference model
- Solution from classical stereotomography seems to have better resolution



- Improving the resolution of adjoint stereotomography
- Using solution from classical stereotomography and adjoint stereotomography as starting model for FWI
- Including anisotropic parameters in adjoint stereotomography
- Tests on real data including multi-component data





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Paris School of Mines for the classical stereotomography software