

# Multicomponent internal multiple prediction analysis with elastic stolt-migration, time-stretching, best-fitting by high resolution radon transform

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# Outline

- Multicomponent internal multiple prediction (MIMP) algorithm
- Monotonicity of actual depth and pseudo-depth/vertical-traveltime
- Several approximate solutions for input preparation
- Synthetic implementation and discussion
- Conclusion
- Acknowledgements

# Multicomponent IMP

1.5D formulation of MIMP (Matson, 1997; Sun and Innanen, 2016):

$$b_3^{ij}(k_g, \omega) = - \int_{-\infty}^{+\infty} dz_1 e^{i(\nu^m + \nu^i)z_1} b_1^{im}(k_g, z_1) \int_{-\infty}^{z_1 - \epsilon} dz_2 e^{-i(\nu^n + \nu^m)z_2} b_1^{mn}(k_g, z_2) \\ \times \int_{z_2 + \epsilon}^{+\infty} dz_3 e^{i(\nu^j + \nu^n)z_3} b_1^{nj}(k_g, z_3)$$

Inputs:  $b_1^{ij}(k_g, z) = i2\nu^j D^{ij}(k_g, z)$ ,  $\{i, j\} \in \{P, SV\}$ .

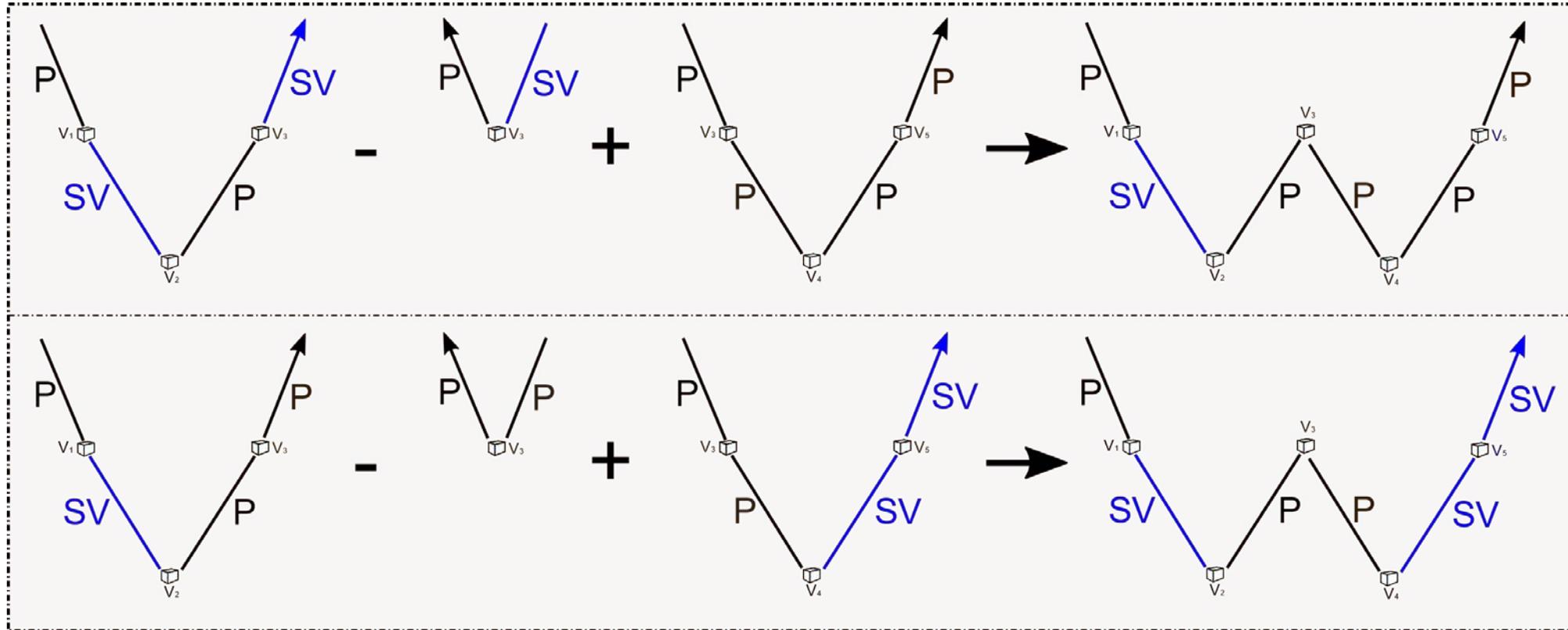
For P-wave source only:

$$b_3^{\dot{P}\dot{P}} = \Theta_1(b_1^{\dot{P}\dot{P}})\Theta_2(b_1^{\dot{P}\dot{P}})\Theta_3(b_1^{\dot{P}\dot{P}}) + \Theta_1(b_1^{\dot{P}\dot{S}})\Theta_2(b_1^{\dot{S}\dot{P}})\Theta_3(b_1^{\dot{P}\dot{P}}) + \Theta_1(b_1^{\dot{P}\dot{P}})\Theta_2(b_1^{\dot{P}\dot{S}})\Theta_3(b_1^{\dot{P}\dot{P}})$$

$$b_3^{\dot{S}\dot{P}} = \Theta_1(b_1^{\dot{S}\dot{P}})\Theta_2(b_1^{\dot{P}\dot{P}})\Theta_3(b_1^{\dot{P}\dot{P}}) + \Theta_1(b_1^{\dot{S}\dot{P}})\Theta_2(b_1^{\dot{P}\dot{S}})\Theta_3(b_1^{\dot{S}\dot{P}})$$

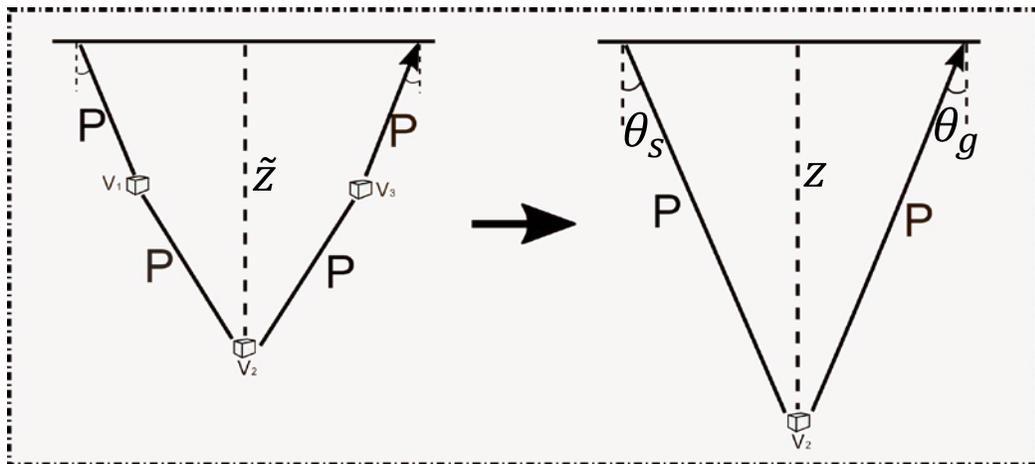
# Multicomponent IMP

Lower-Higher-Lower combinations in actual depth:



# Monotonicity of actual & pseudo-depth

For acoustic cases:



$$z = \frac{v\tau}{\cos\theta_s + \cos\theta_g}$$

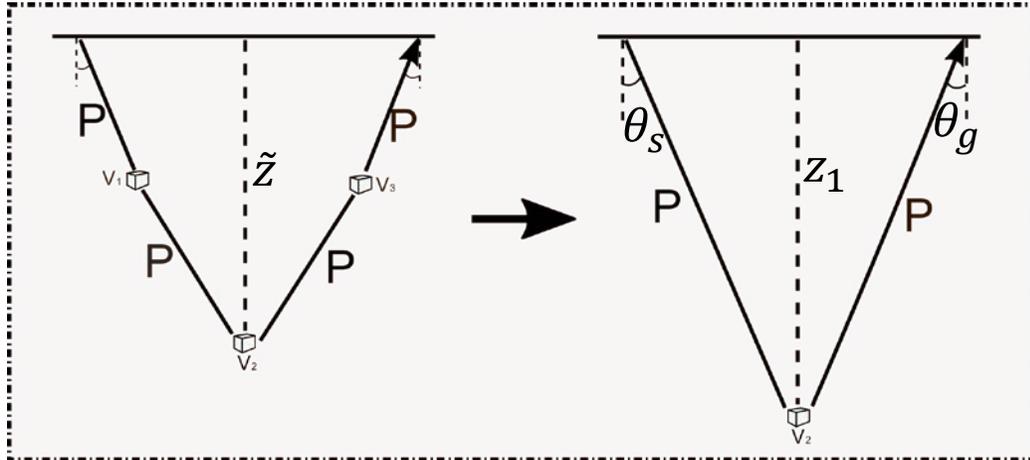
$$\tilde{z}_1 < \tilde{z}_2 \leftrightarrow z_1 < z_2$$

$D(x, t) \rightarrow D(k/p, \omega(k_z)) \rightarrow D(k/p, k_z) \rightarrow D(k/p, z) \leftarrow$  Stolt migration

$D(x, t) \rightarrow D(p, \tau) \leftarrow$   $\tau - p$  transform

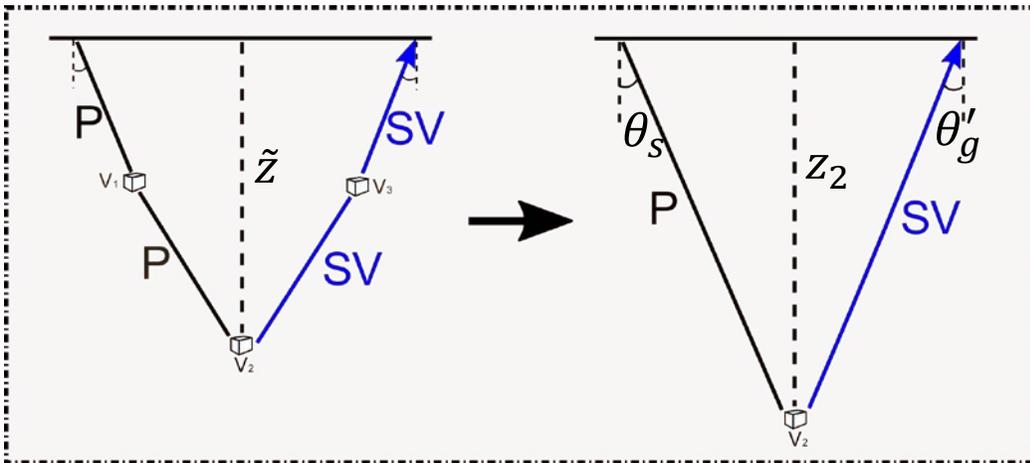
# Monotonicity of actual & pseudo-depth

For elastic cases:



$$z_1 = \frac{\alpha \tau_{pp}}{\cos \theta_s + \cos \theta_g}$$

$$z_1 \approx z_2$$

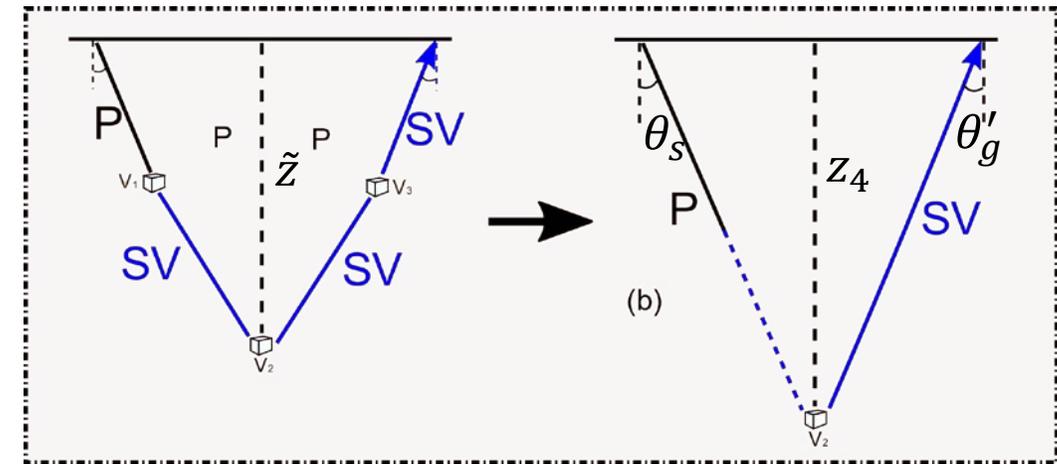
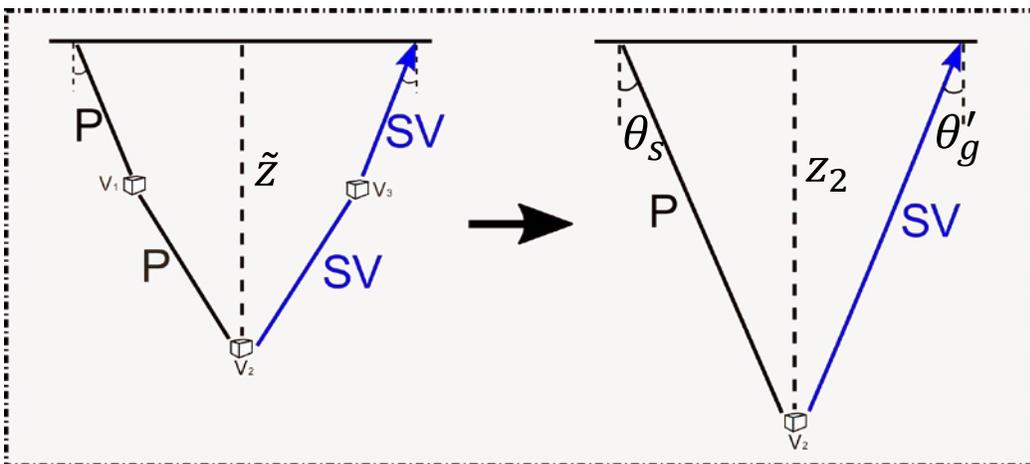
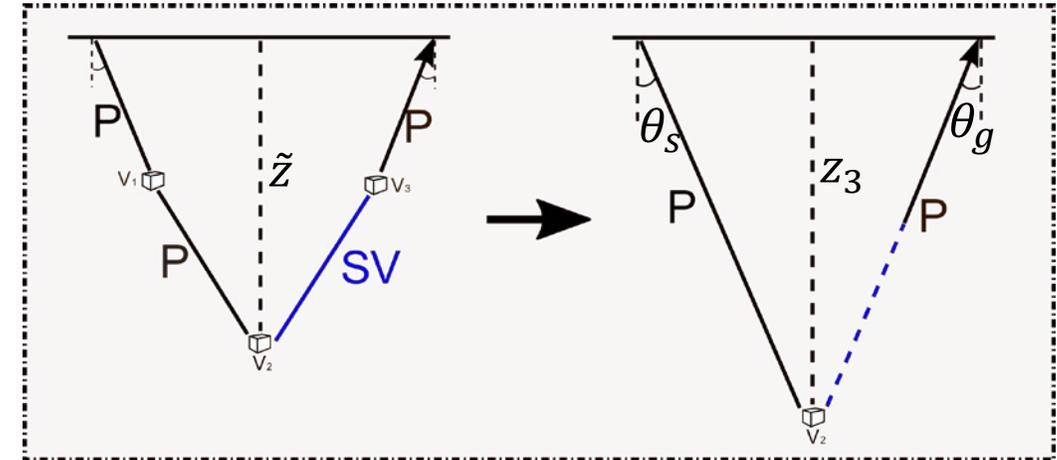
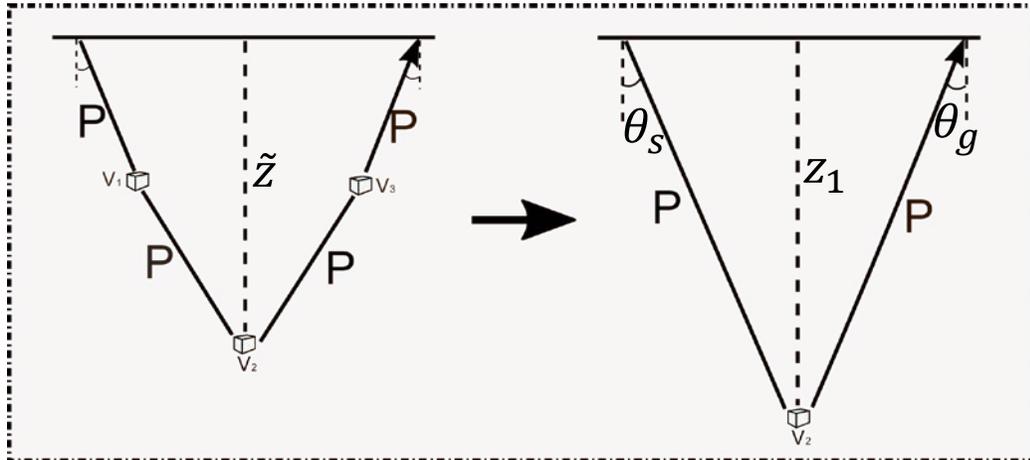


$$z_2 = \frac{v \tau_{ps}}{\cos \theta_s + \cos \theta'_g}$$

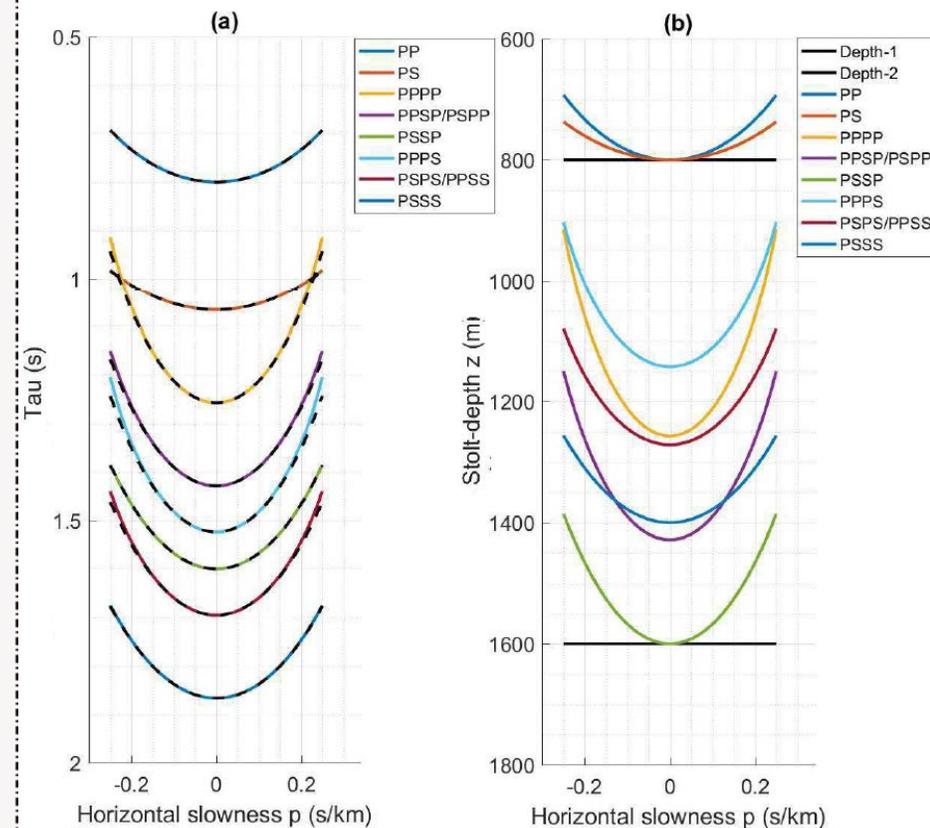
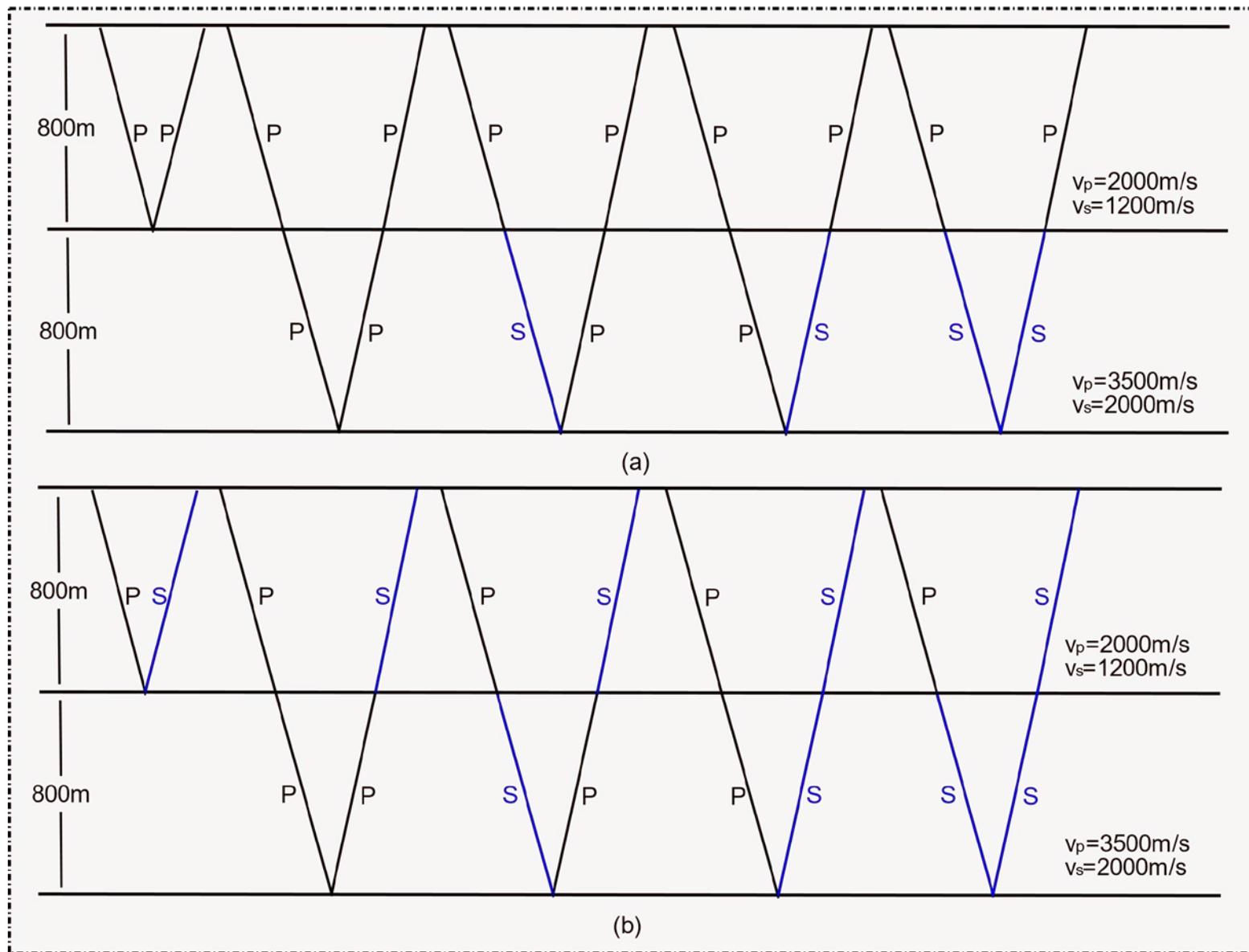
$$v = \frac{\alpha \beta}{\alpha + \beta}$$

# Monotonicity of actual & pseudo-depth

For elastic cases:



# Approximate solutions for input preparation



# Approximate solutions for input preparation

$$b_3^{ij}(p_g, \omega) = - \int_{-\infty}^{+\infty} d\tau_2^{mn} e^{-i\omega\tau_2^{mn}} b_1^{mn}(p_g, \tau_2^{mn}) \int_{\Upsilon(\tau_2^{mn}|\tau_3^{nj})+\epsilon}^{+\infty} d\tau_3^{nj} e^{i\omega\tau_3^{nj}} b_1^{nj}(p_g, \tau_3^{nj}) \\ \times \int_{\Upsilon(\tau_2^{mn}|\tau_1^{im})+\epsilon}^{+\infty} d\tau_1^{im} e^{i\omega\tau_1^{im}} b_1^{im}(p_g, \tau_1^{im})$$

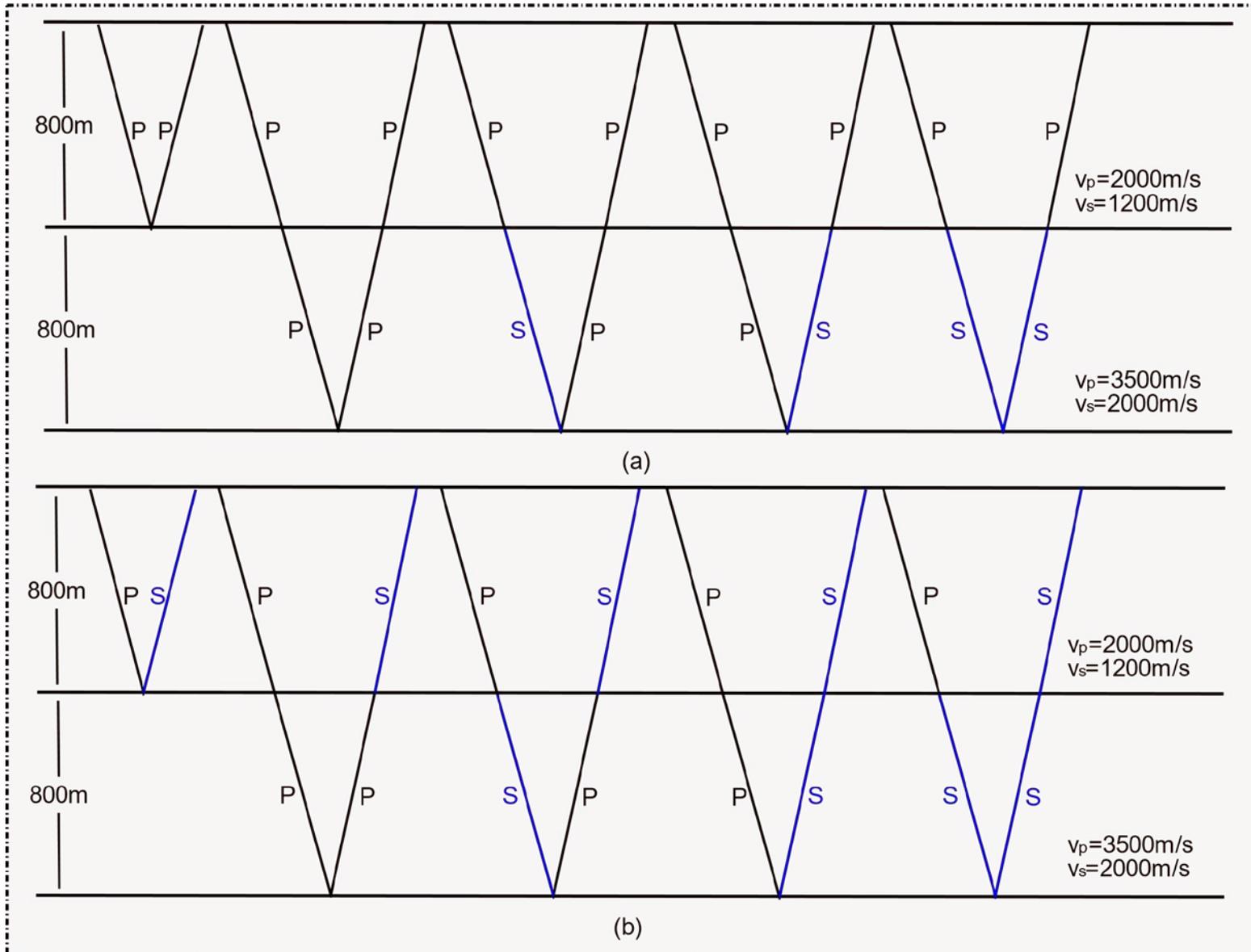
$$\Upsilon(\tau_2^{mn}|\tau_1^{nj}) = \begin{cases} \tau_2^{mn}, & j = m; \\ \frac{\alpha+\beta}{2\beta} \tau_2^{mn}, & j = S \text{ \& } m = P; \\ \frac{2\beta}{\alpha+\beta} \tau_2^{mn}, & j = P \text{ \& } m = S; \end{cases}$$

# Approximate solutions for input preparation

$$b_3^{ij}(p_g, \omega) = -\frac{1}{(2\pi)^2} \int_{-\infty}^{+\infty} d\tau_2^{mn} e^{-i\omega\tau_2^{mn}} b_1^{mn}(p_g, \tau_2^{mn}) \int_{\Gamma(mn,nj)+\epsilon}^{+\infty} d\tau_3^{nj} e^{i\omega\tau_3^{nj}} b_1^{nj}(p_g, \tau_3^{nj}) \\ \times \int_{\Gamma(mn,im)+\epsilon}^{+\infty} d\tau_1^{im} e^{i\omega\tau_1^{im}} b_1^{im}(p_g, \tau_1^{im})$$

$$\Gamma(X, Y) = \begin{cases} \frac{\alpha}{\beta}, & X = SS; \\ \frac{\alpha + \beta}{2\beta}, & X \neq SS \& Y = PS \text{ or } SP; \\ 1, & \text{others.} \end{cases}$$

# Approximate solutions for input preparation



$$t^2 = t_0^2 + \frac{x^2}{v_s^2}$$

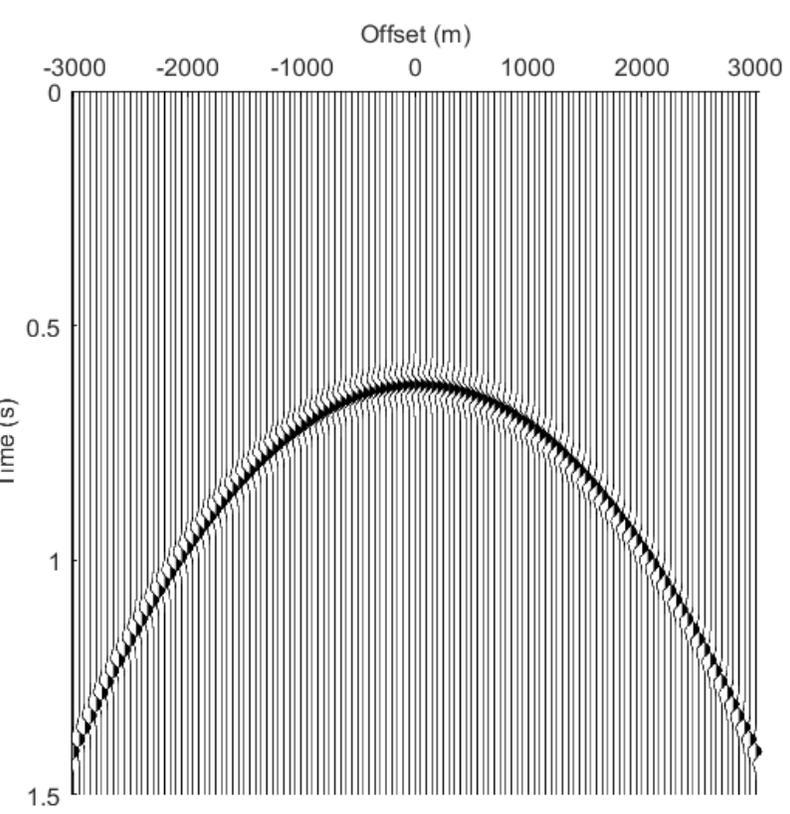
$$m(\tau, v) = \int_{h_{min}}^{h_{max}} d(t = \sqrt{\tau^2 + h^2/v^2}, h) dh$$

$$\mathbf{d} = \mathbf{Lm}$$

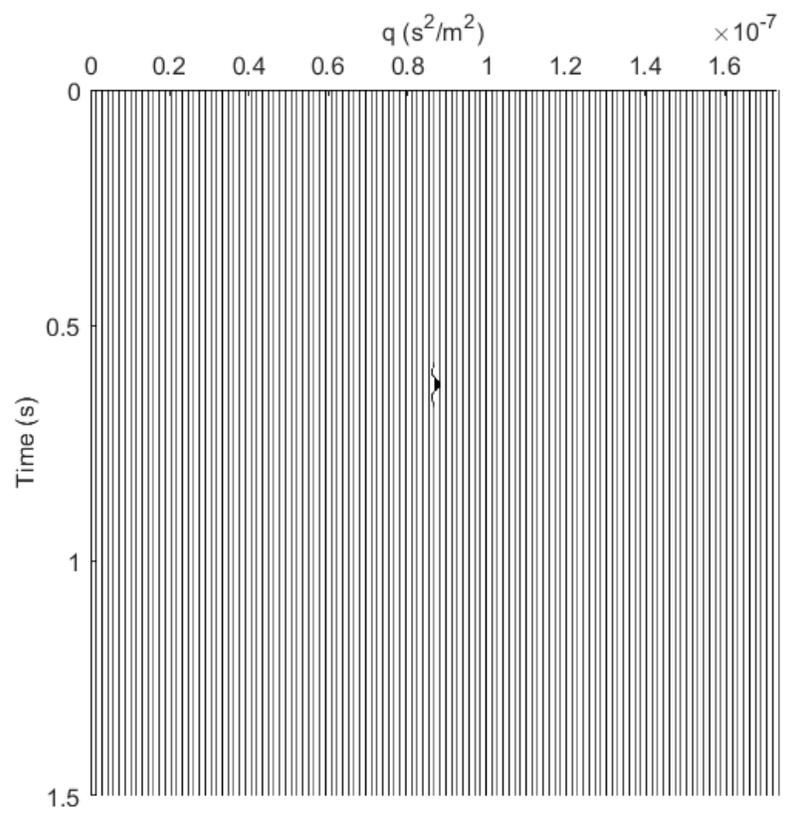
$$J = \| \mathbf{W}_d(\mathbf{Lm} - \mathbf{d}) \|^2 + \| \mathbf{W}_m \mathbf{m} \|^2$$

$$\hat{d}(\tau, p) = \int_{v_{min}}^{v_{max}} m(\tau_0 = \frac{\tau}{\sqrt{1 - p^2 v^2}}, v) dv$$

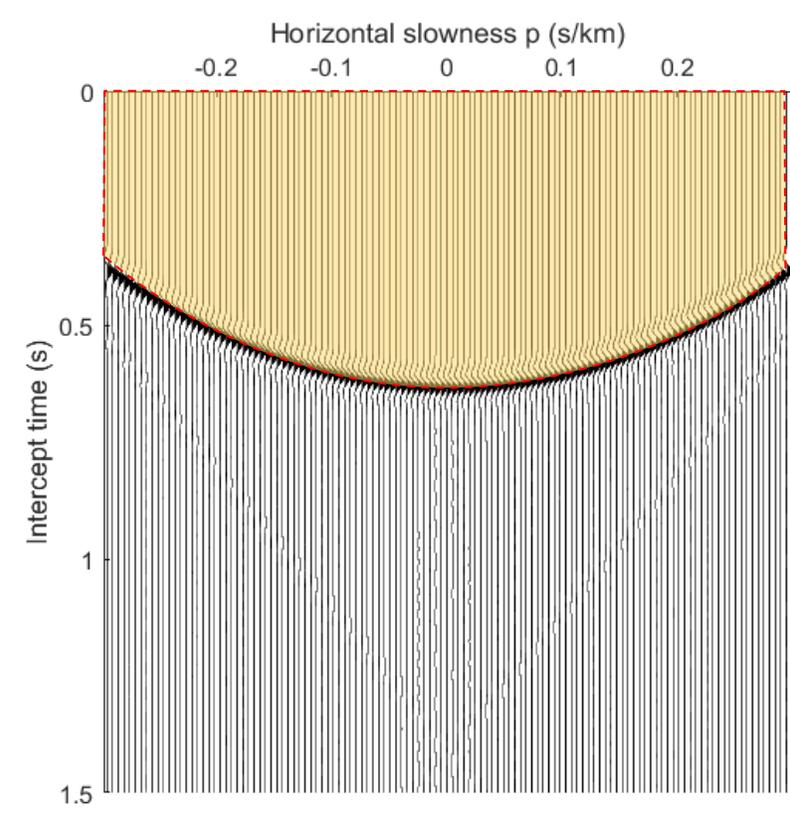
# Approximate solutions for input preparation



$D(x, t)$



$m(v, \tau)$



$\tilde{D}(p, \tau)$

# Approximate solutions for input preparation

## ❖ Approximate solutions for input preparation:

➤ Elastic stolt migration (First mentioned by Matson, 1997)

➤ Time stretching method

$$b_3^{ij}(p_g, \omega) = \int_{-\infty}^{+\infty} dz_1 e^{i(\nu^m + \nu^i)z_1} b_1^{im}(k_g, z_1) \int_{-\infty}^{z_1 - \epsilon} dz_2 e^{-i(\nu^n + \nu^m)z_2} b_1^{mn}(k_g, z_2)$$

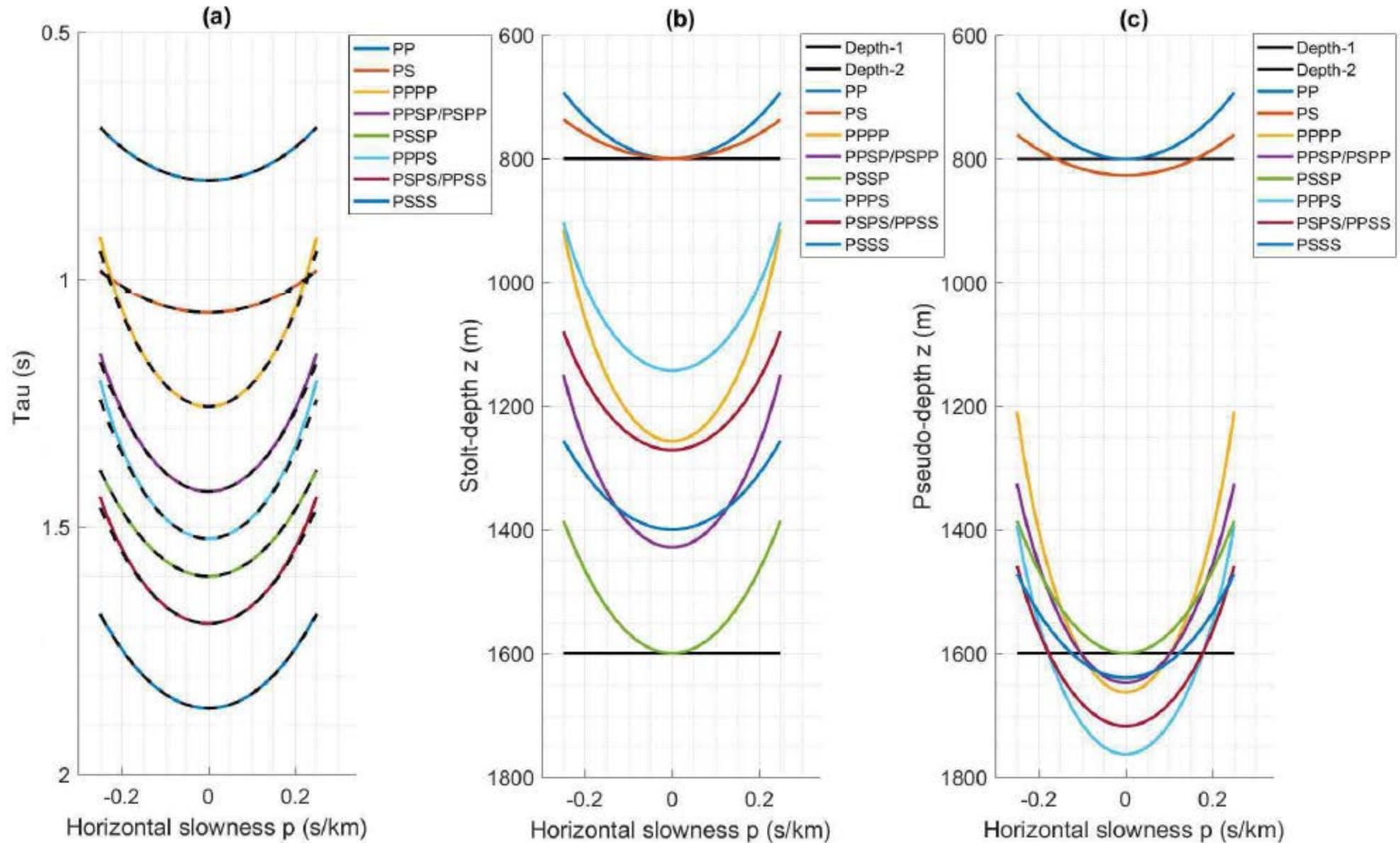
➤ Best-fitting approach by high-resolution hyperbolic radon

$$b_3^{ij}(p_g, \omega) = - \int_{-\infty}^{+\infty} d\tau_1^{im} e^{i\omega\tau_1^{im}} b_1^{im}(p_g, \tau_1^{im}) \int_{-\infty}^{\Gamma(\tau_1^{im}|\tau_2^{mn}) - \epsilon} d\tau_2^{mn} e^{-i\omega\tau_2^{mn}} b_1^{mn}(p_g, \tau_2^{mn})$$

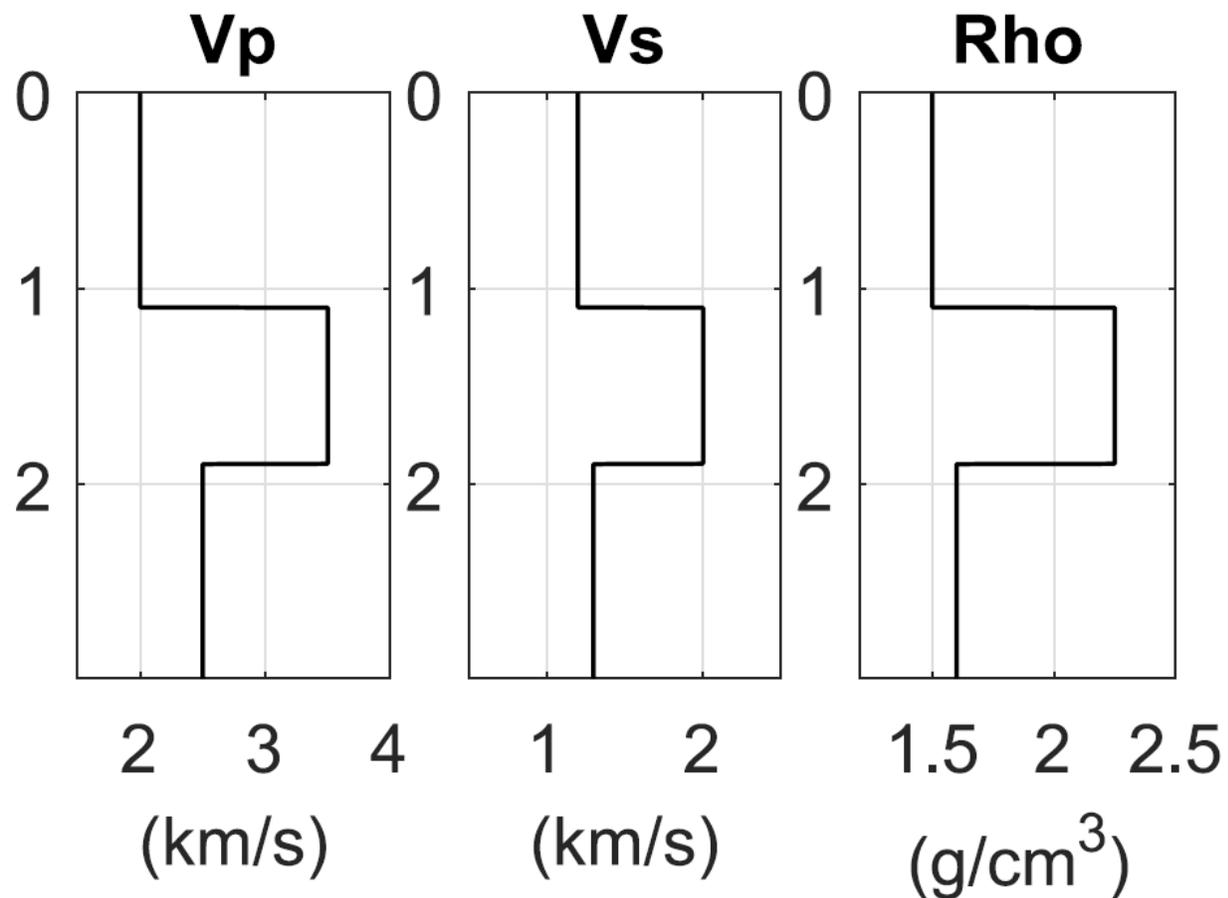
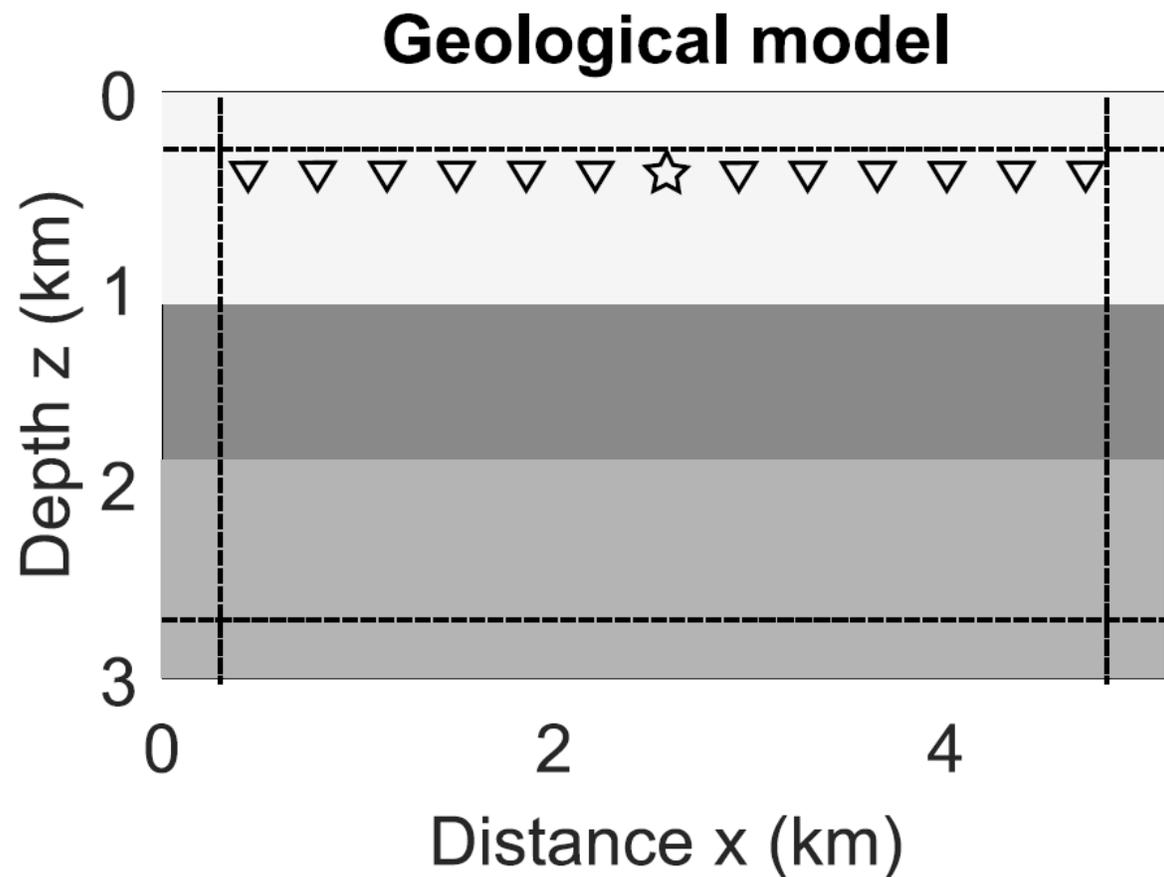
$$\times \int_{\Gamma(\tau_2^{mn}|\tau_3^{nj}) + \epsilon}^{+\infty} d\tau_3^{nj} e^{i\omega\tau_3^{nj}} b_1^{nj}(p_g, \tau_3^{nj})$$

$$\Gamma(\tau_2^{mn}|\tau_1^{nj}) = \frac{v^{mn}(p_g, \tau_2^{mn})}{v^{nj}(p_g, \tau_1^{nj})} \tau_2^{mn} \quad P \ \& \ m = S;$$

# Approximate solutions for input preparation



# Synthetic -- model



# Synthetic -- shot profile

Label-P	Primaries in P-mode
$Pr_1^P$	$\bar{P}\bar{P}$
$Pr_{21}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}$
$Pr_{22}^P$	$\bar{P}\bar{P}\bar{S}\bar{P}$ & $\bar{P}\bar{S}\bar{P}\bar{P}$
$Pr_{23}^P$	$\bar{P}\bar{S}\bar{S}\bar{P}$

Label-P	1st-order IMs in P-mode
$IM_{11}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}$
$IM_{12}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}\bar{P}$
$IM_{13}^P$	$\bar{P}\bar{P}\bar{P}\bar{S}\bar{S}\bar{P}$
$IM_{14}^P$	$\bar{P}\bar{P}\bar{S}\bar{S}\bar{S}\bar{P}$
$IM_{15}^P$	$\bar{P}\bar{S}\bar{S}\bar{S}\bar{S}\bar{P}$

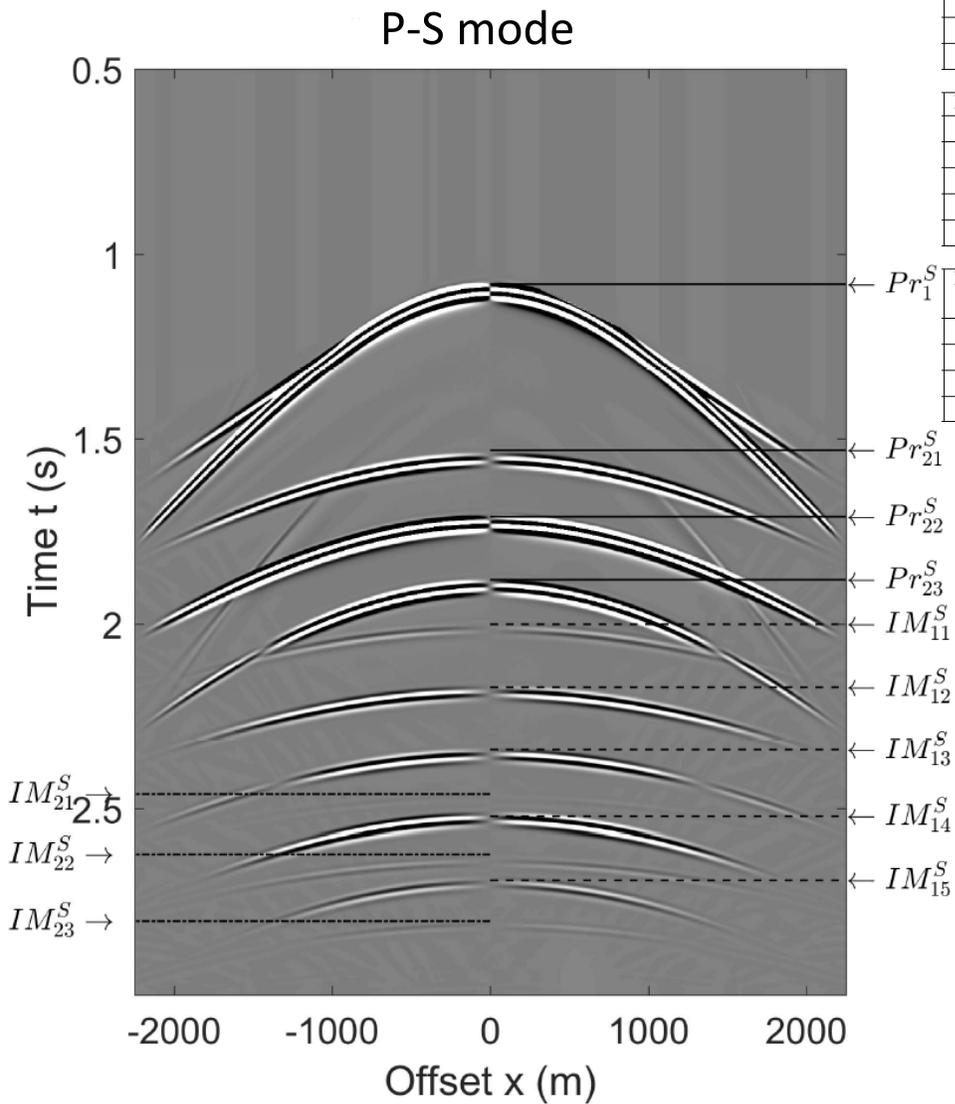
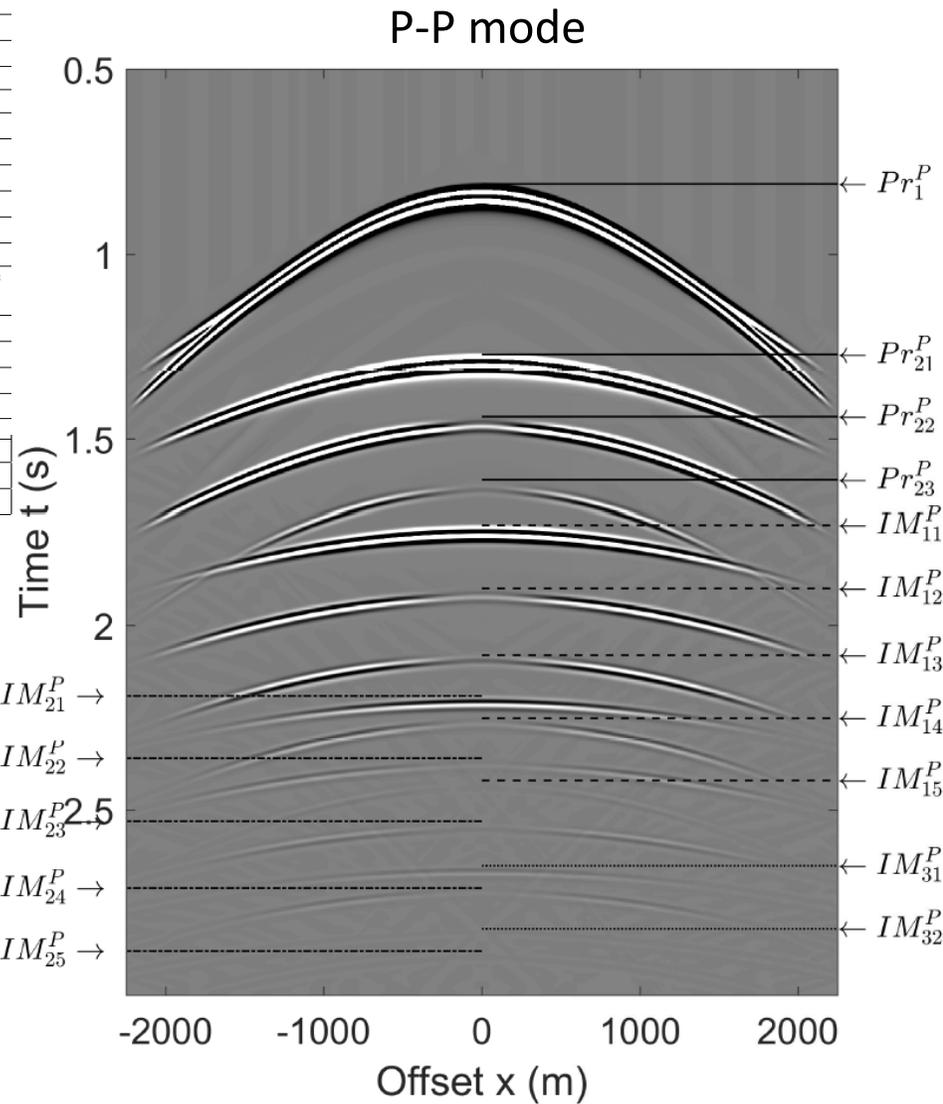
Label-P	2nd-order IMs in P-mode
$IM_{21}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}$
$IM_{22}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}\bar{P}$
$IM_{23}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}\bar{S}\bar{P}$
$IM_{24}^P$	$\bar{P}\bar{P}\bar{P}\bar{S}\bar{S}\bar{S}\bar{P}$
$IM_{25}^P$	$\bar{P}\bar{P}\bar{S}\bar{S}\bar{S}\bar{S}\bar{P}$

Label-P	3rd-order IMs in P-mode
$IM_{31}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}$
$IM_{32}^P$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}\bar{P}$

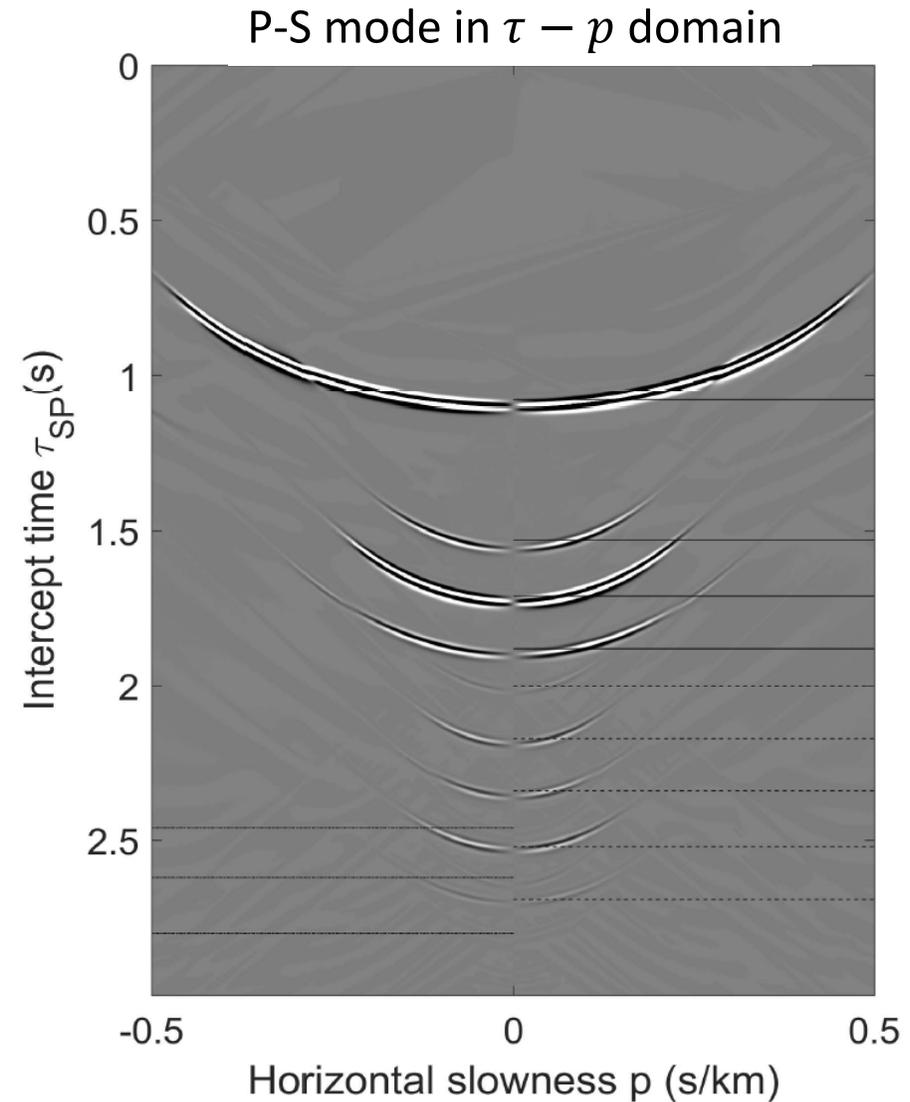
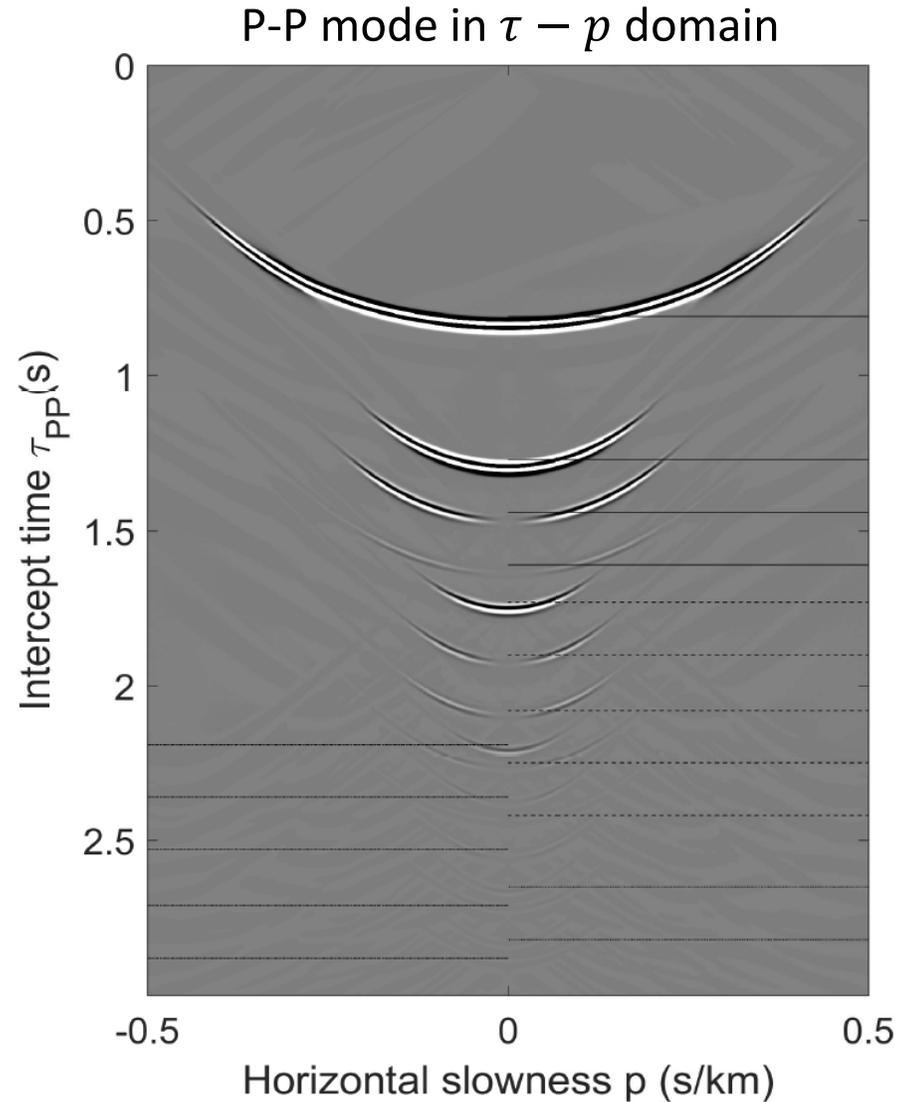
Label-S	Primaries S-mode
$Pr_1^S$	$\bar{P}\bar{S}$
$Pr_{21}^S$	$\bar{P}\bar{P}\bar{P}\bar{S}$
$Pr_{22}^S$	$\bar{P}\bar{P}\bar{S}\bar{S}$ & $\bar{P}\bar{S}\bar{P}\bar{S}$
$Pr_{23}^S$	$\bar{P}\bar{S}\bar{S}\bar{S}$

Label-S	1st-order IMs in S-mode
$IM_{11}^S$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}$
$IM_{12}^S$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}\bar{S}$
$IM_{13}^S$	$\bar{P}\bar{P}\bar{P}\bar{S}\bar{S}\bar{S}$
$IM_{14}^S$	$\bar{P}\bar{P}\bar{S}\bar{S}\bar{S}\bar{S}$
$IM_{15}^S$	$\bar{P}\bar{S}\bar{S}\bar{S}\bar{S}\bar{S}$

Label-S	2nd-order IMs in S-mode
$IM_{21}^S$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}$
$IM_{22}^S$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}\bar{S}$
$IM_{23}^S$	$\bar{P}\bar{P}\bar{P}\bar{P}\bar{S}\bar{S}\bar{S}$



# Synthetic -- linear radon

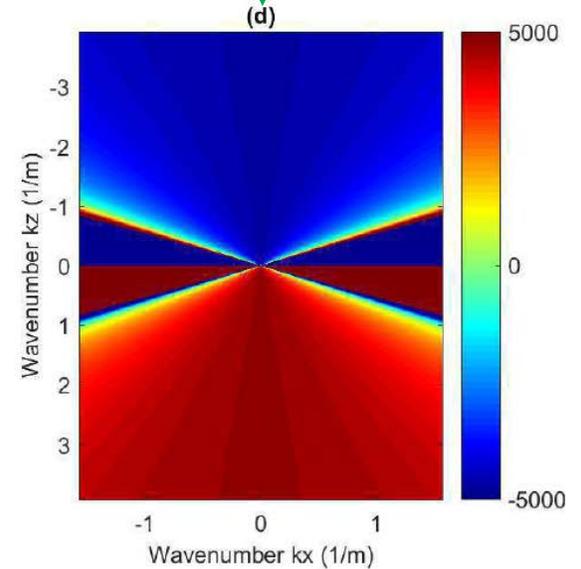
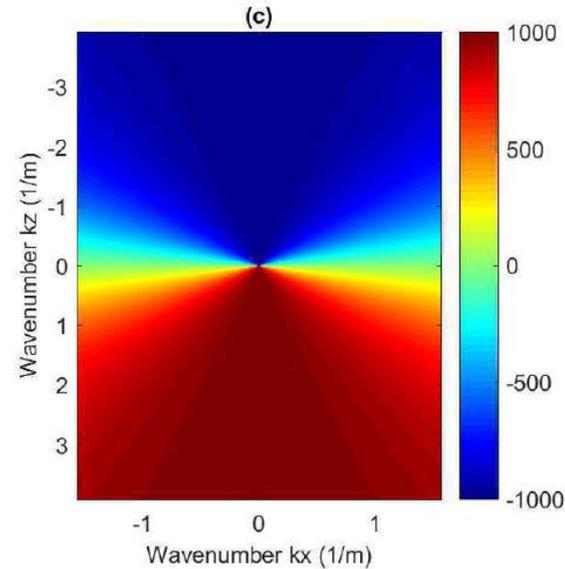
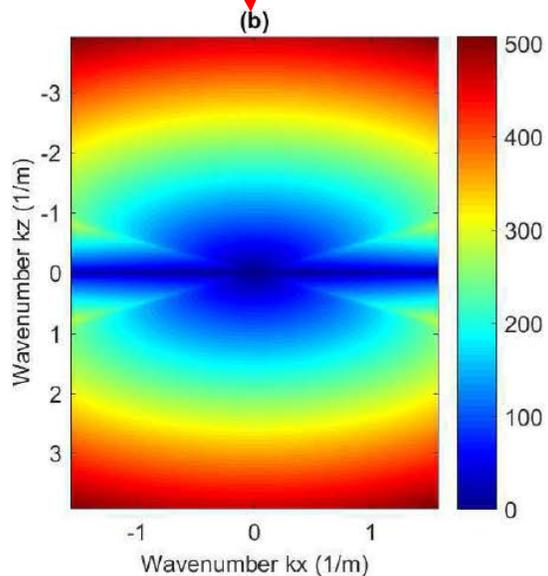
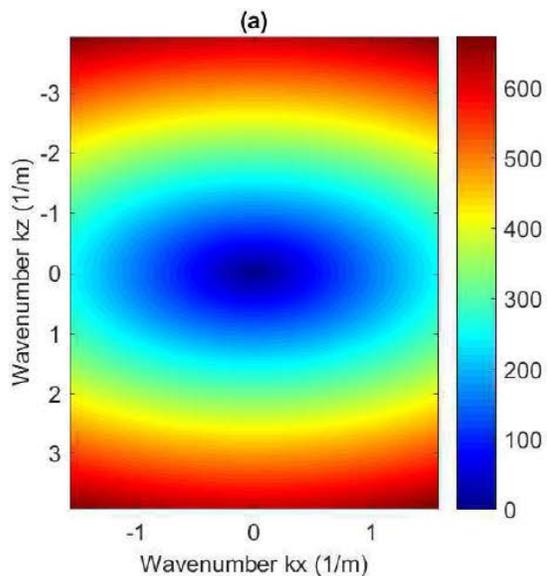


# Synthetic -- elastic stolt-migration

Elastic stolt-migration (Etgen, 1988):

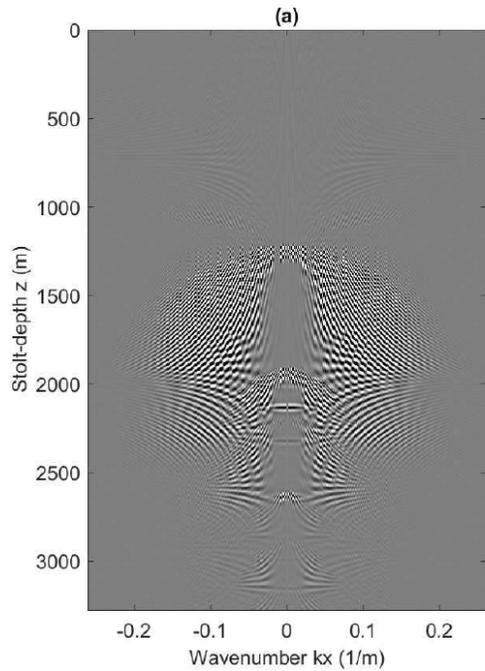
$$R_{PP}(k_m, z) = \int dk_h \int dk_{zP} \left| \frac{d\omega}{dk_{zP}} \right| P(k_m, k_h, w(k_{zP})) e^{ik_{zP}z}$$

$$R_{PSV}(k_m, z) = \int dk_h \int dk_{zSV} \left| \frac{d\omega}{dk_{zSV}} \right| P(k_m, k_h, w(k_{zSV})) e^{ik_{zSV}z}$$

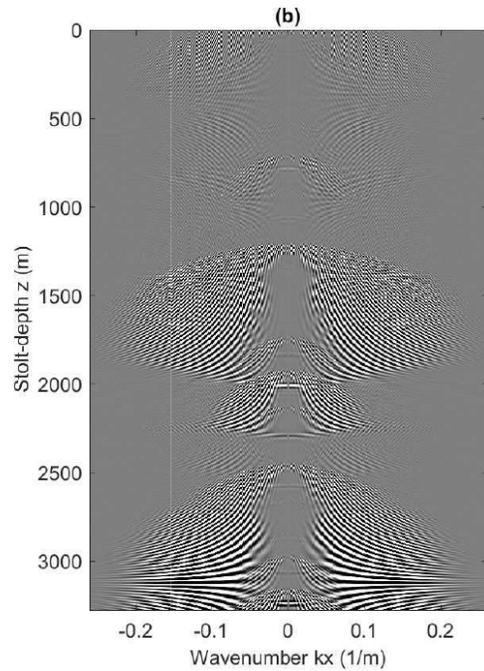


# Synthetic -- elastic stolt-migration

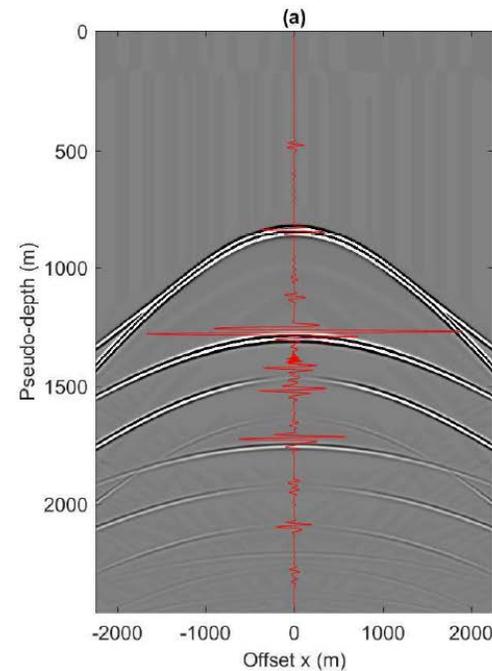
Elastic stolt-migration with wavenumber:



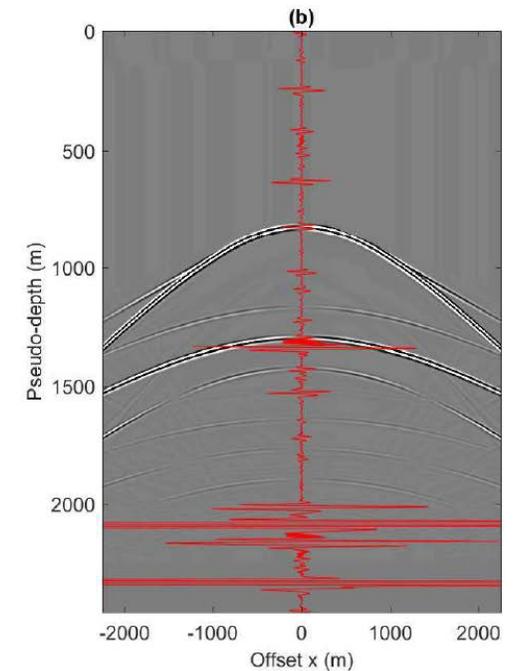
$$R_{PP}(k_x, z)$$



$$R_{PS}(k_x, z)$$



$$R_{PP}(z)$$



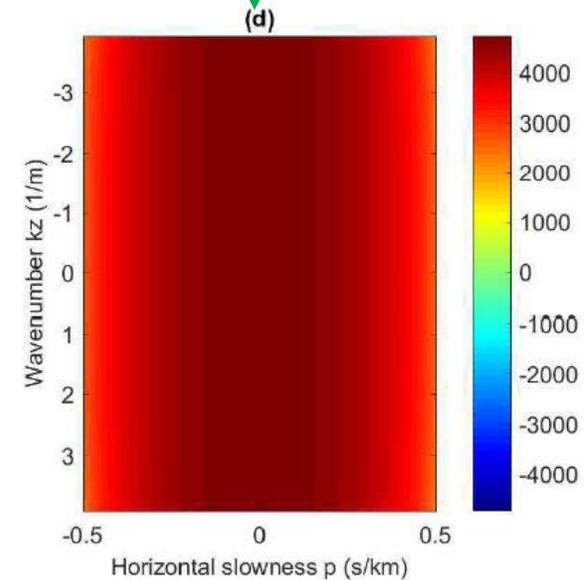
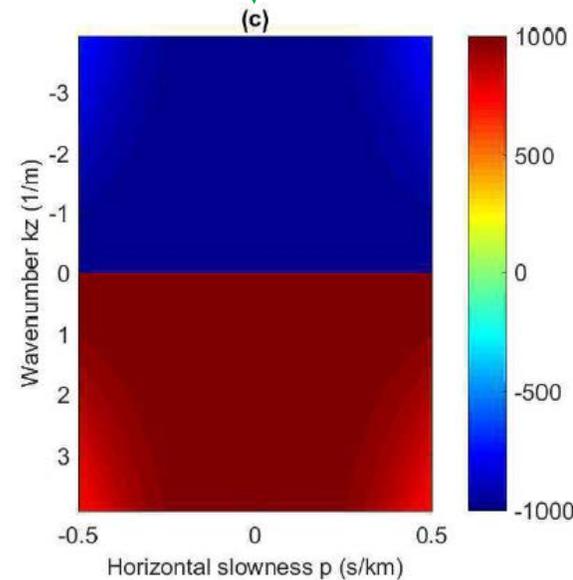
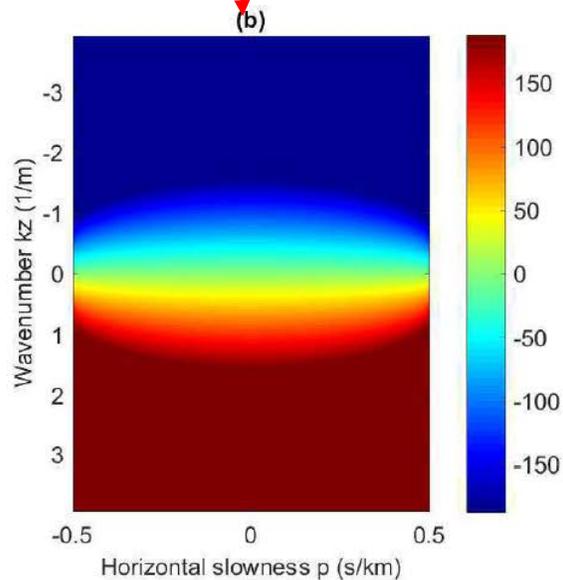
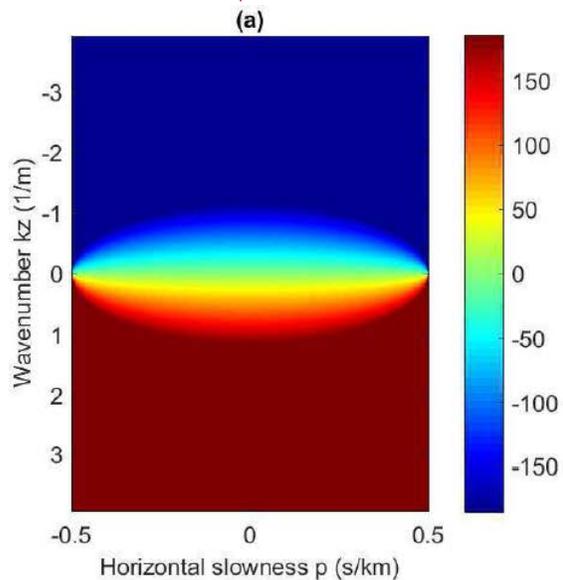
$$R_{PS}(z)$$

# Synthetic -- elastic stolt-migration

Elastic stolt-migration with horizontal slowness:

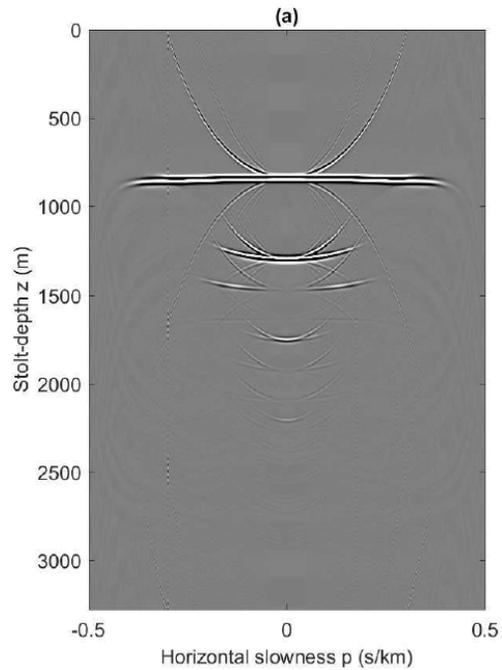
$$R_{PP}(p_m, z) = \int dp_h \int dk_{zP} \left| \frac{d\omega}{dk_{zP}} \right| P(p_m, p_h, w(k_{zP})) e^{ik_{zP}z}$$

$$R_{PSV}(p_m, z) = \int dp_h \int dk_{zSV} \left| \frac{d\omega}{dk_{zSV}} \right| P(p_m, p_h, w(k_{zSV})) e^{ik_{zSV}z}$$

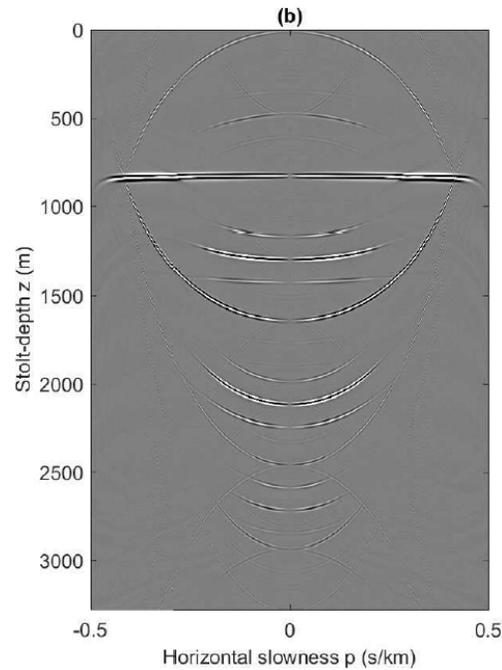


# Synthetic -- elastic stolt-migration

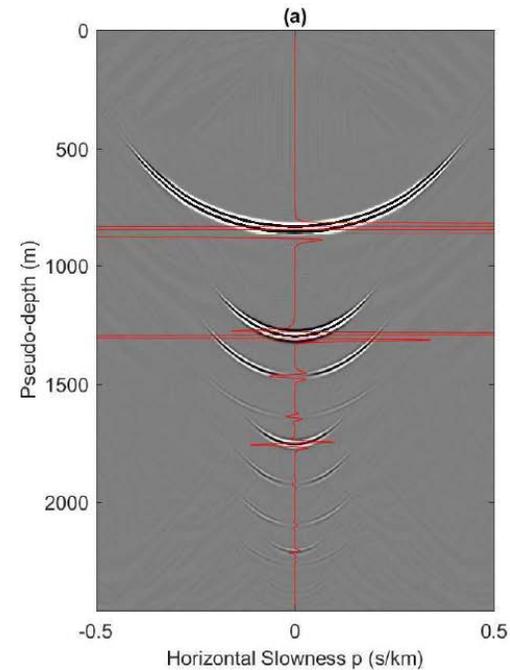
Elastic stolt-migration with horizontal slowness:



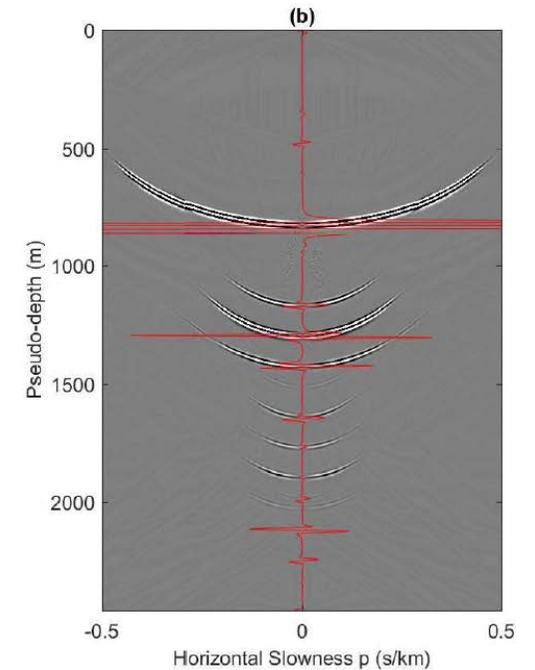
$$R_{PP}(p, z)$$



$$R_{PS}(p, z)$$



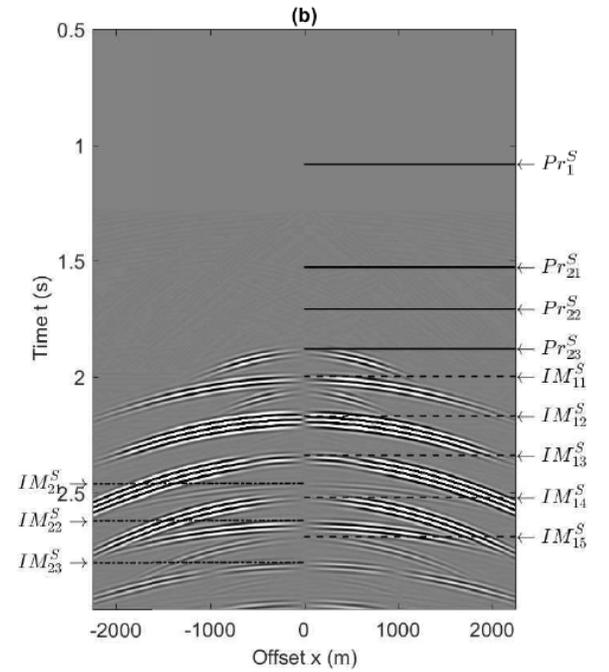
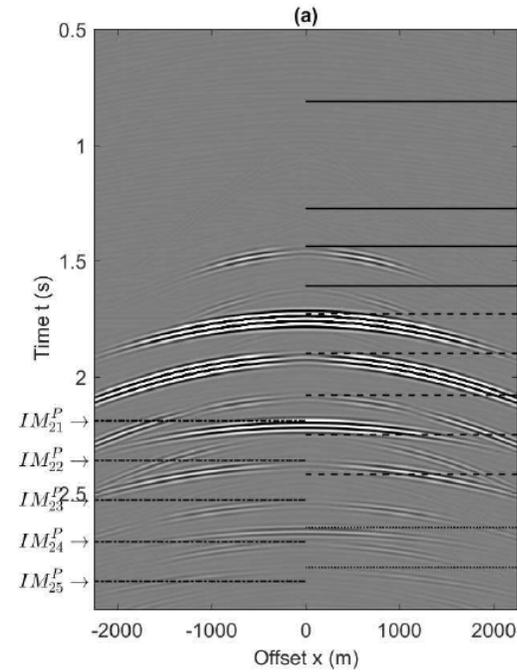
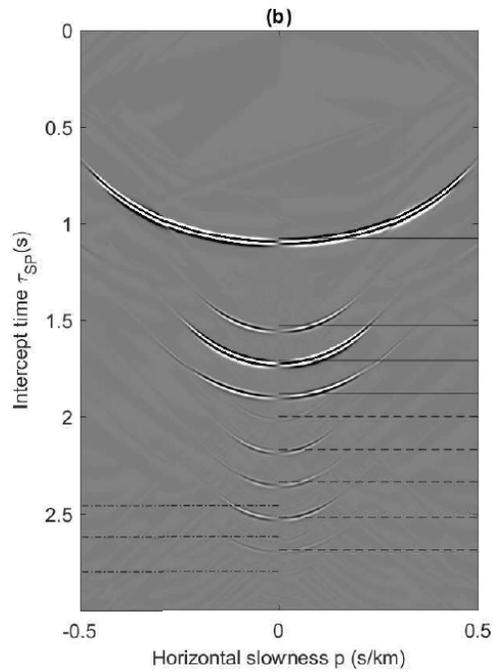
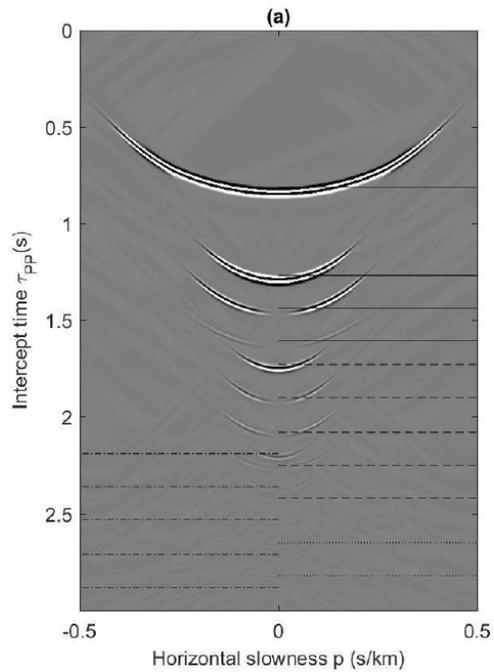
$$R_{PP}(z)$$



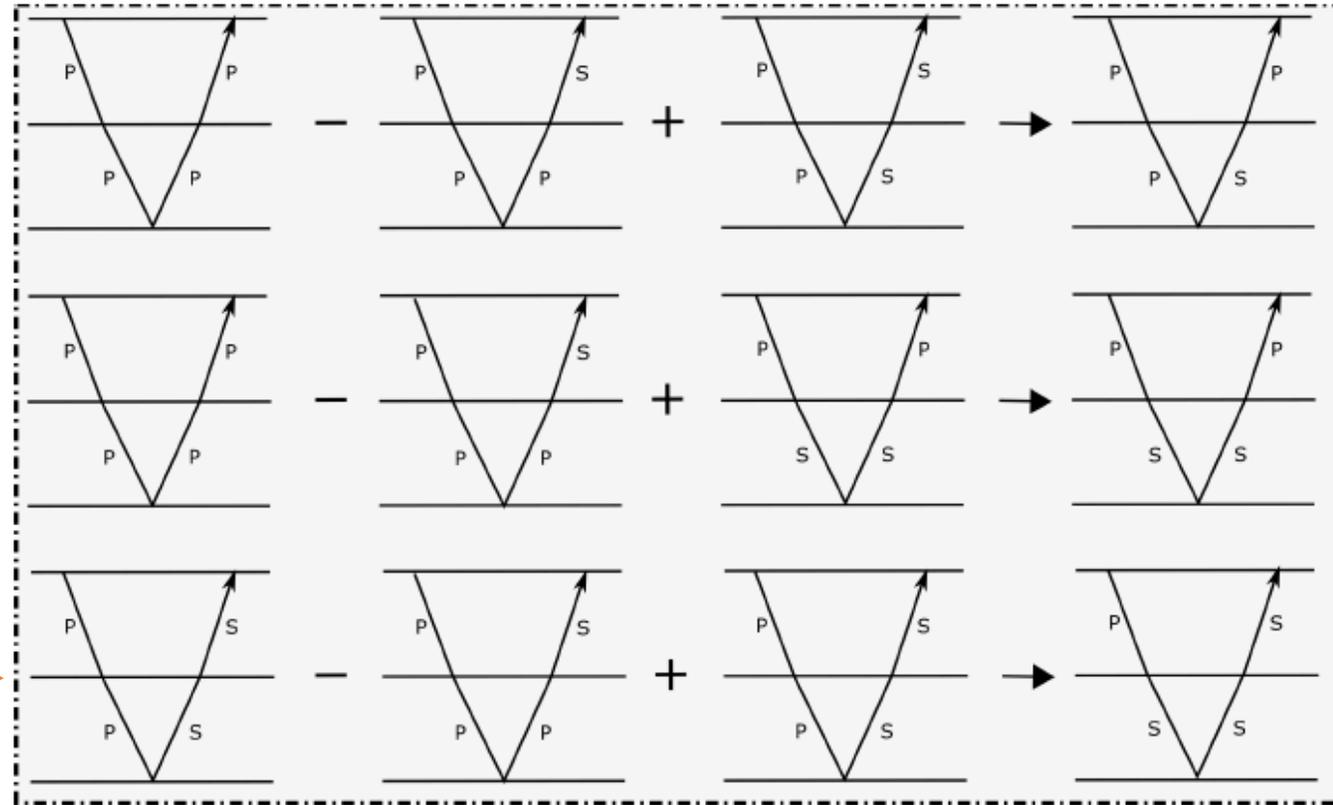
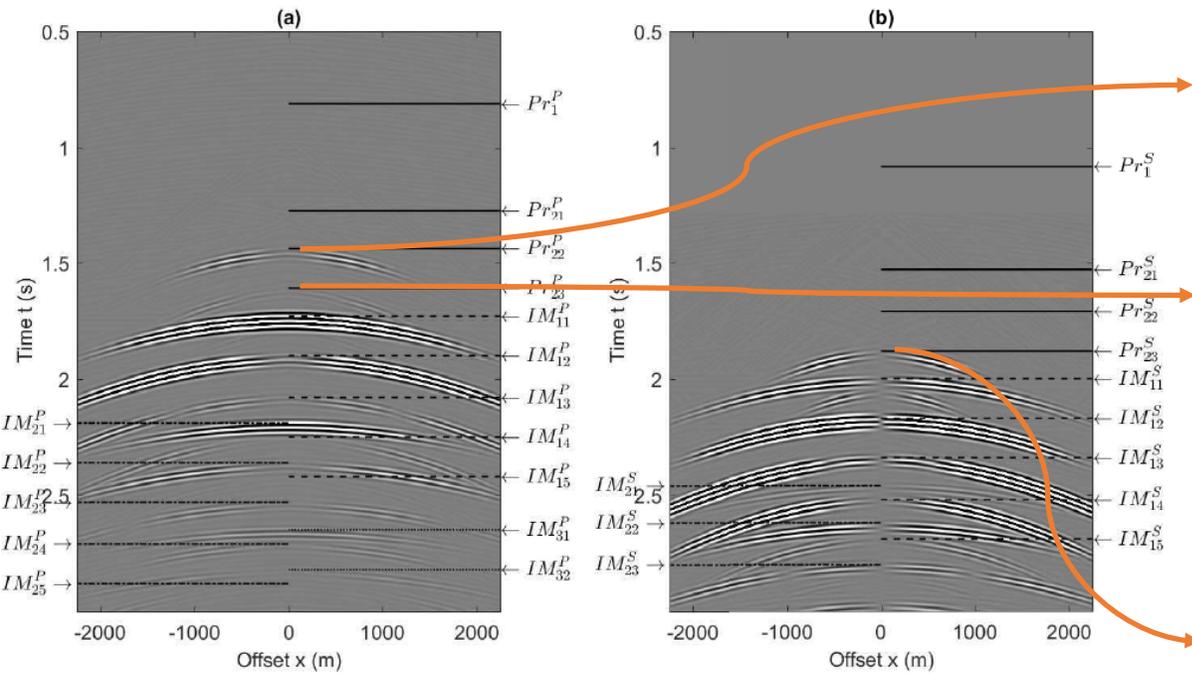
$$R_{PS}(z)$$

# Synthetic -- elastic stolt-migration/time-stretching

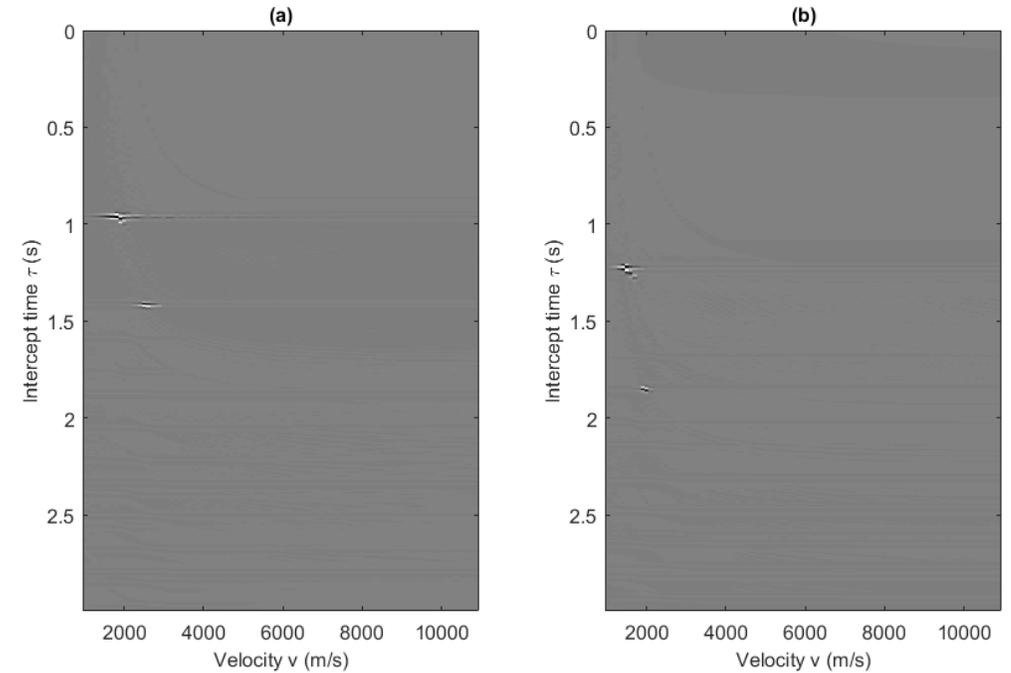
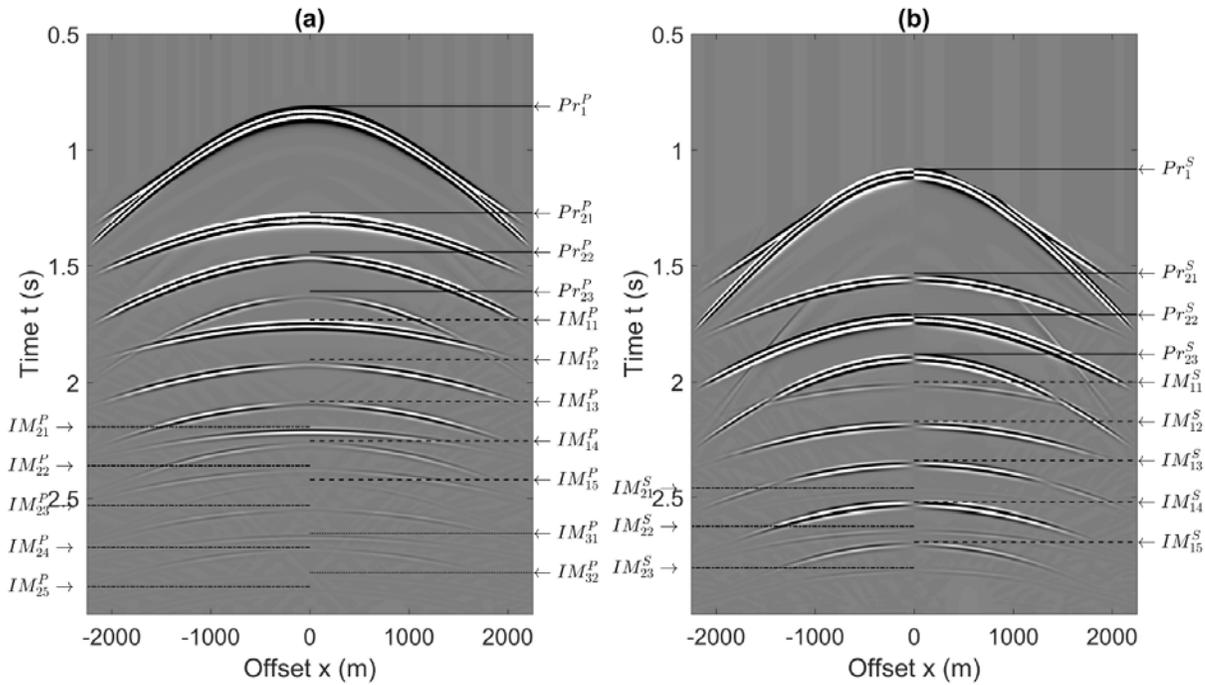
Prediction with time-stretching condition ( $\epsilon = 96ms$ ):



# Synthetic -- elastic stolt-migration/time-stretching



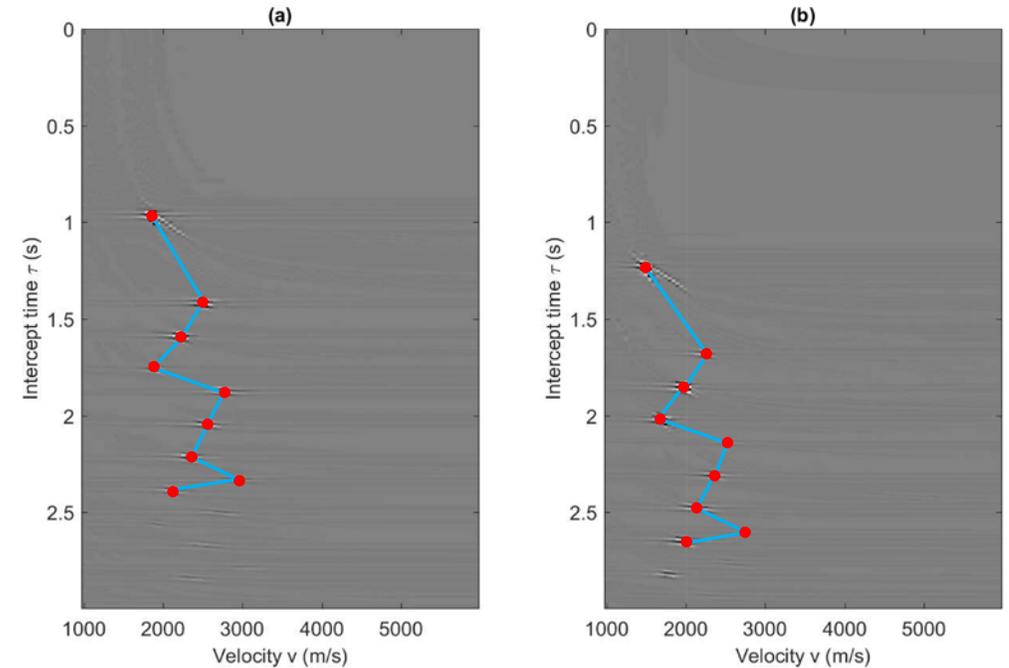
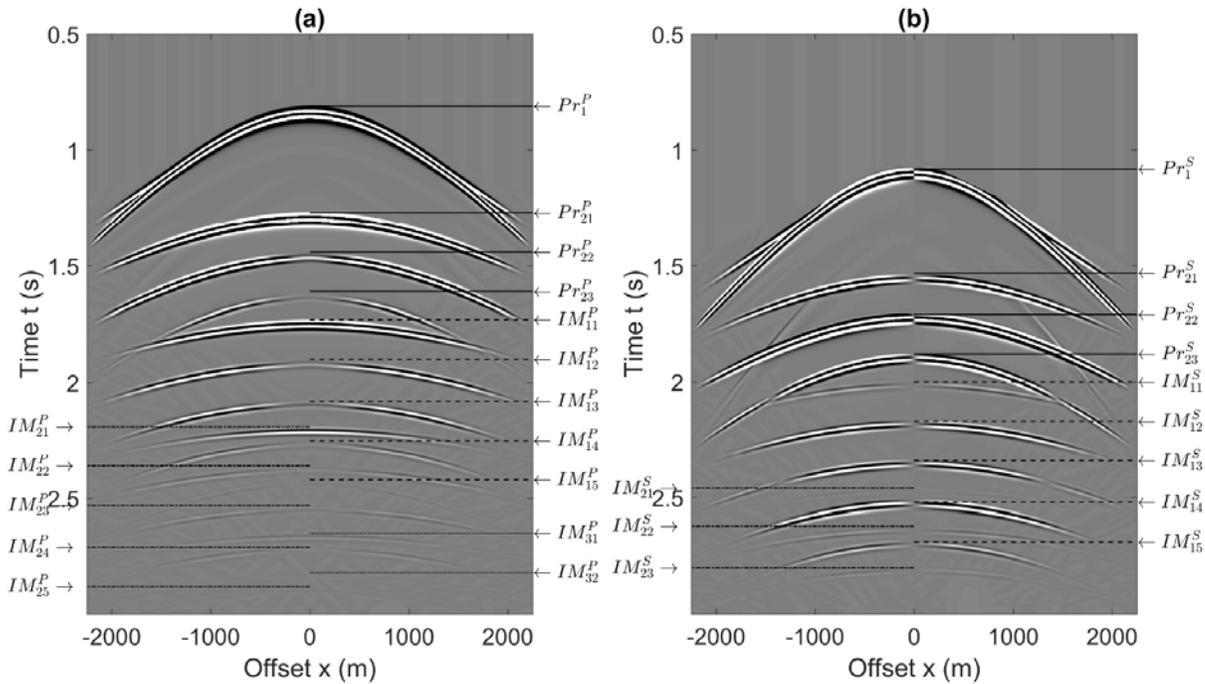
# Synthetic -- best-fitting approach



$$\Delta v_{iq+1} = \Delta v_{iq}(1 + \epsilon) + \Delta v_{min} \text{ if } n/2 < iq < n$$

$$\Delta v_{iq-1} = \Delta v_{iq}(1 + \epsilon) - \Delta v_{min} \text{ if } 1 < iq < n/2$$

# Synthetic -- best-fitting approach



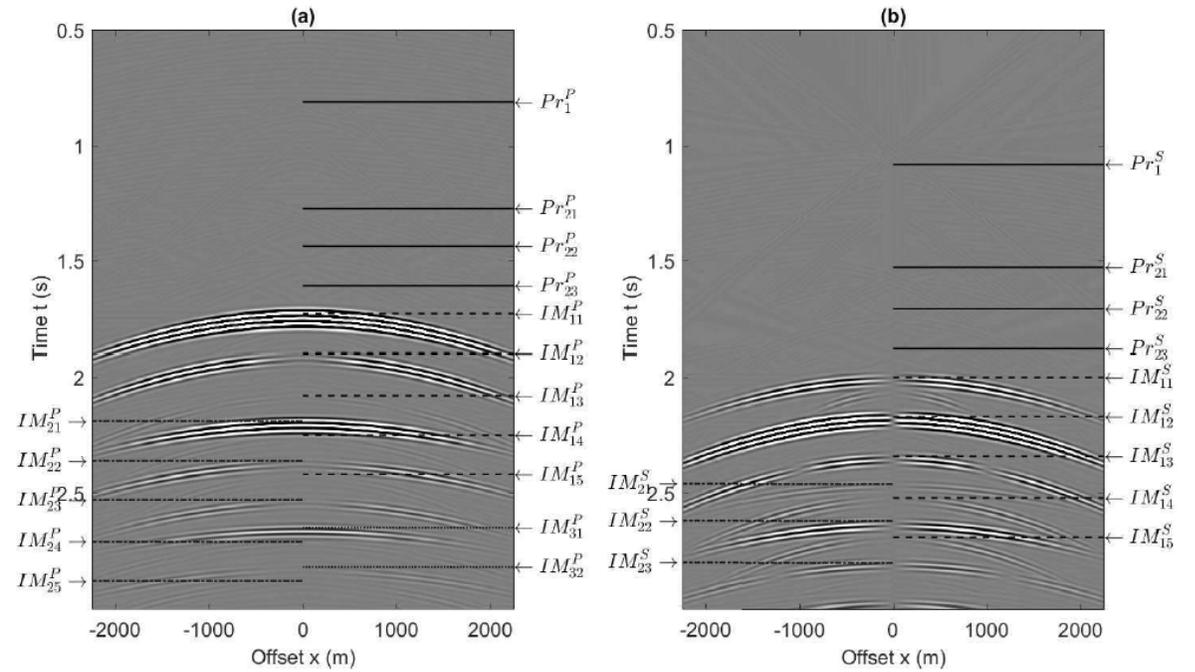
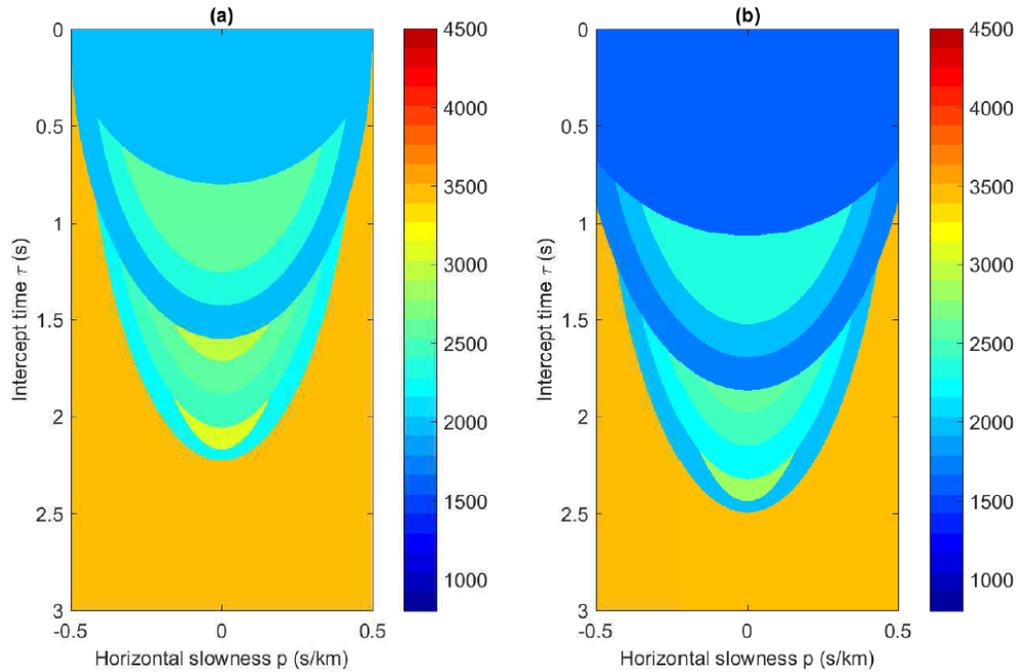
Data = Data \*  $t^{tpow}$ , Here  $tpow = 4$ .

$$\Delta v_{iq+1} = \Delta v_{iq} + \Delta v_{min} \text{ if } n/2 < iq < n$$

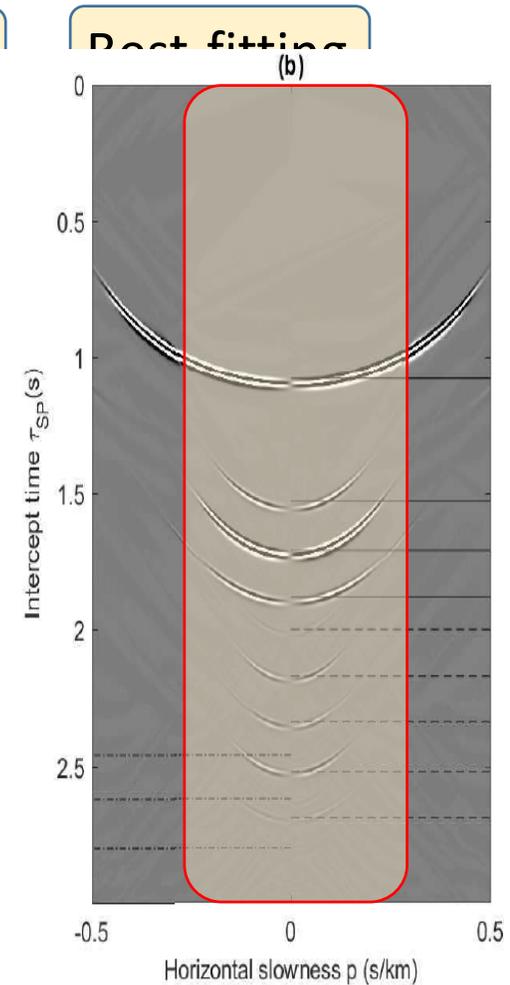
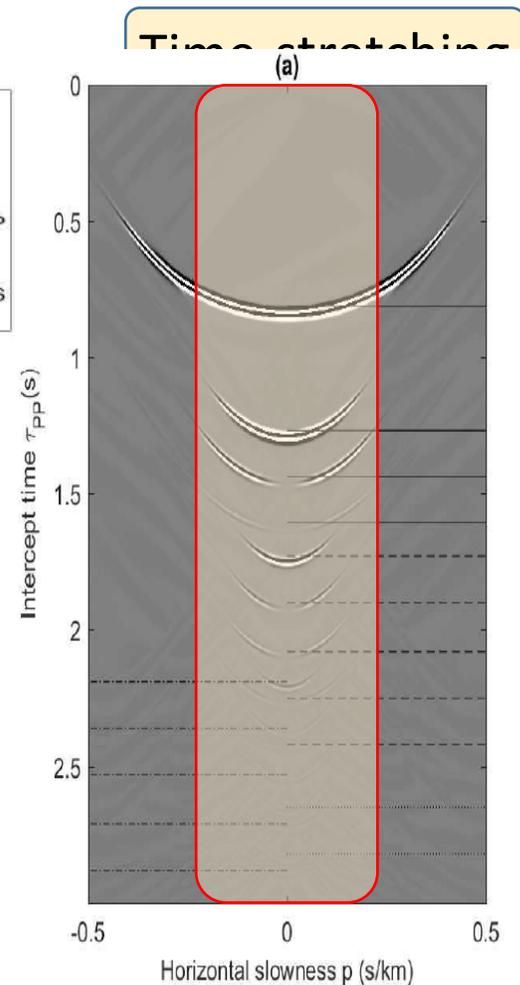
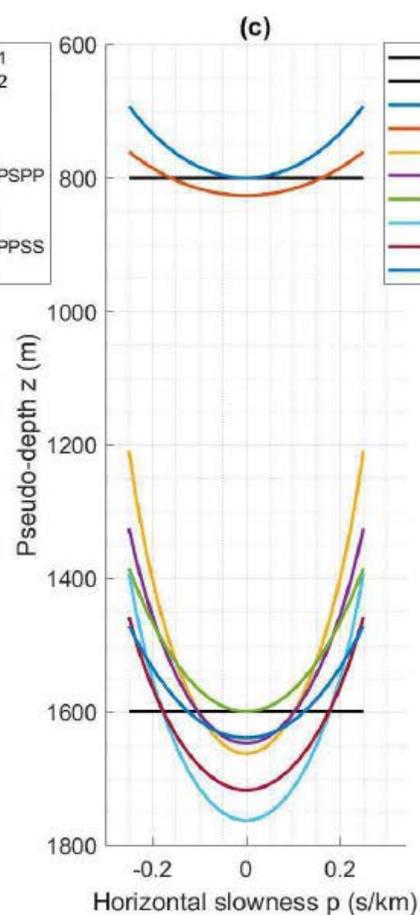
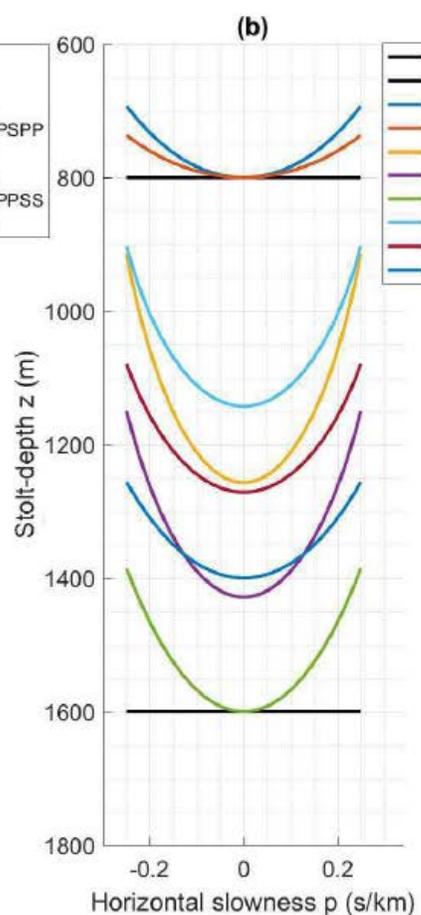
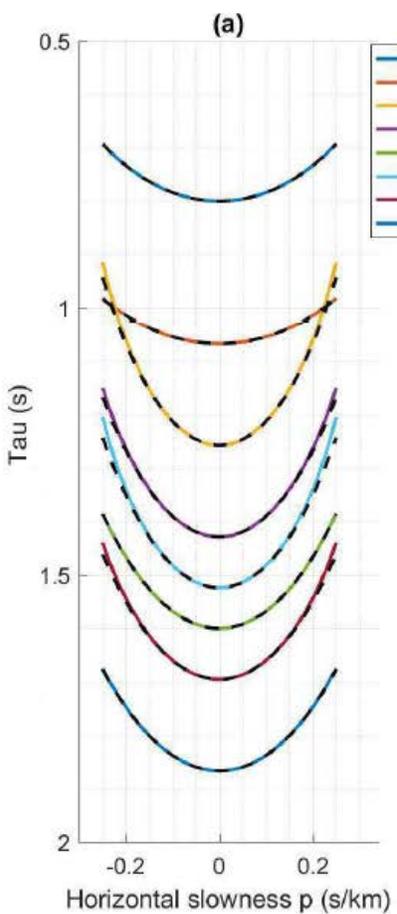
$$\Delta v_{iq-1} = \Delta v_{iq} - \Delta v_{min} \text{ if } 1 < iq < n/2$$

# Synthetic – best-fitting approach

Prediction with best-fitting condition ( $\epsilon = 200ms$ ):



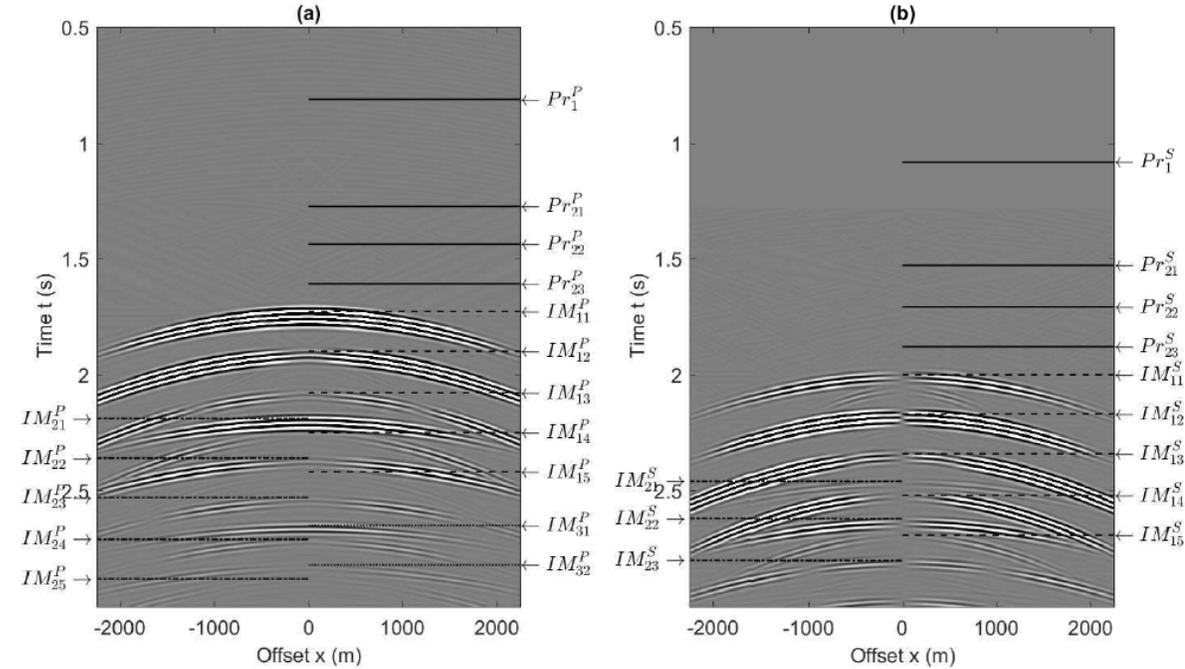
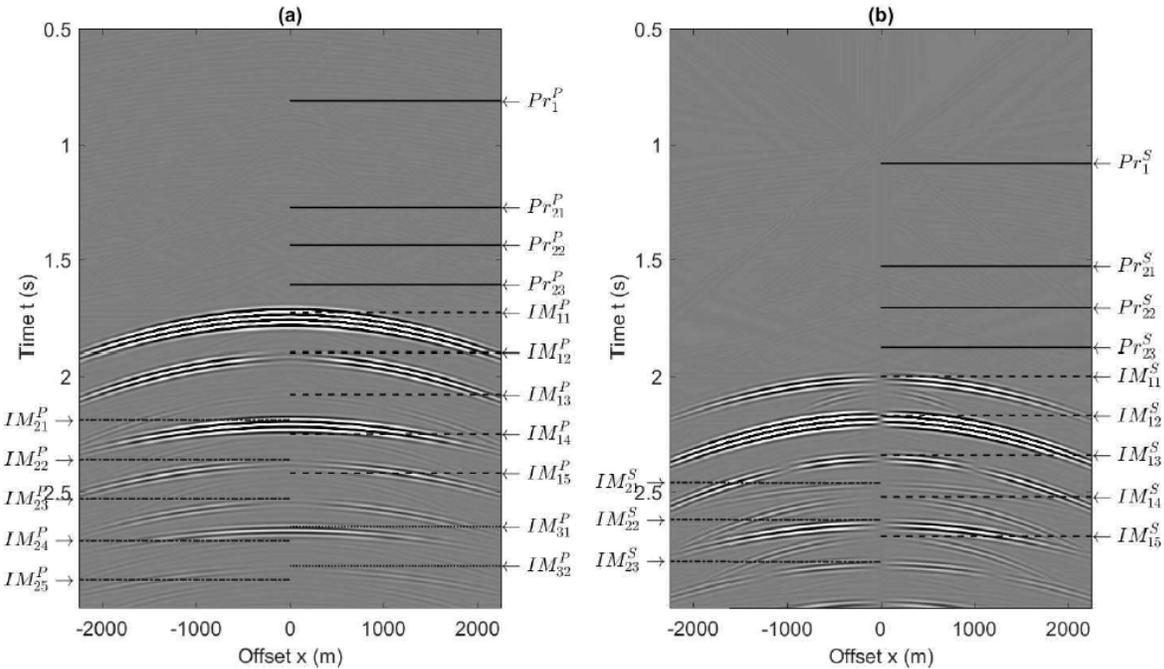
# Approximate solutions for input preparation



# Synthetic – best-fitting & time-stretching

Best-fitting ( $\epsilon = 200ms$ )

Cross-validate ( $\epsilon = 96ms$ )



# Conclusions

- We disclosed and discussed the issues of the multicomponent internal multiple prediction using inverse scattering series.
- Elastic stolt-migration with two background velocities is not suitable for input preparation.
- Time-stretching method retain the monotonic relationship of vertical travelttime and pseudo-depth.
- Best-fitting velocity by high-resolution hyperbolic radon provided an approximate solution which requires a large search parameter.
- Based on the opposite sorting order, the cross-validate condition obtained by time-stretching and best-fitting velocity can produce an exact prediction while allowing a regular constant epsilon.

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# Thank You!