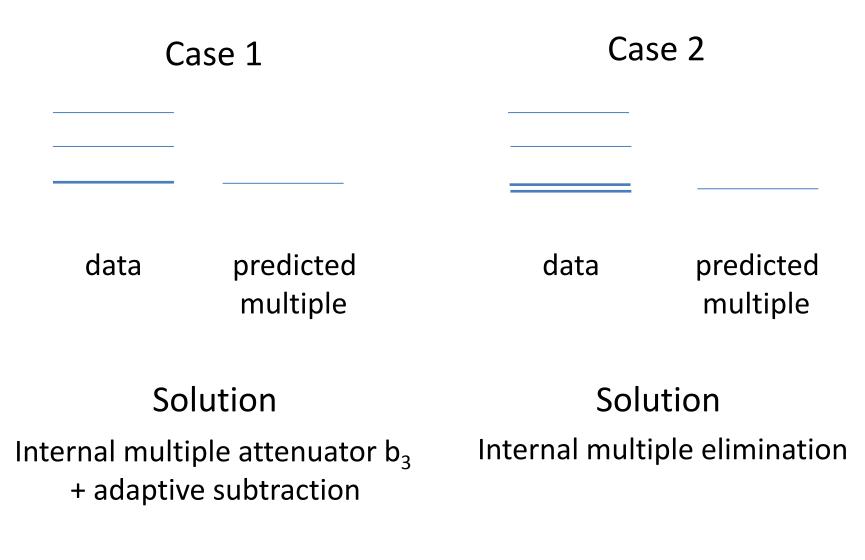




# A method for the elimination of all first order internal multiples from all reflectors in a 1D medium: theory and examples

# Yanglei Zou San Antonio, Texas

May 2, 2013



Yanglei Zou

# Key Points

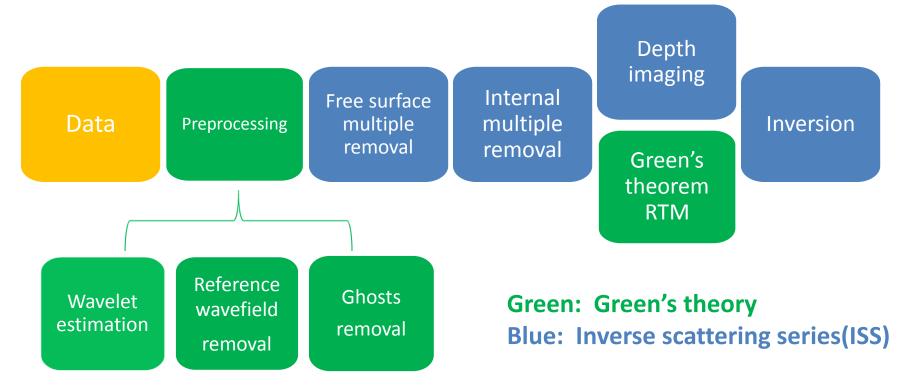
In this presentation, a new method is given to eliminate all first order internal multiples under 1D normal incidence.

This method

- is derived in a reverse engineering way (not seeking higher order terms within inverse scattering series) to construct an algorithm to eliminate first order internal multiples.
- 2. achieves the goal directly in terms of data without determining the earth.

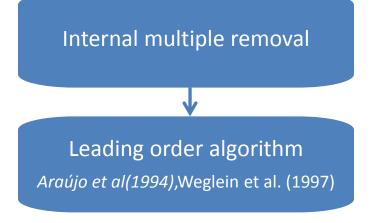
Yanglei Zou Internal Multiple Removal

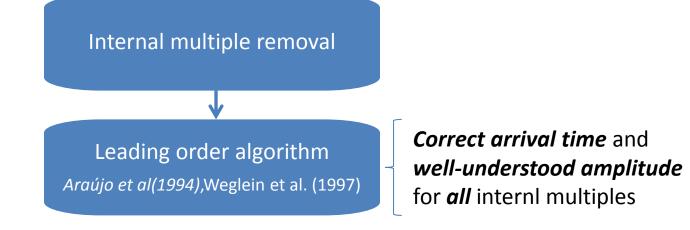
## **Processing Train**

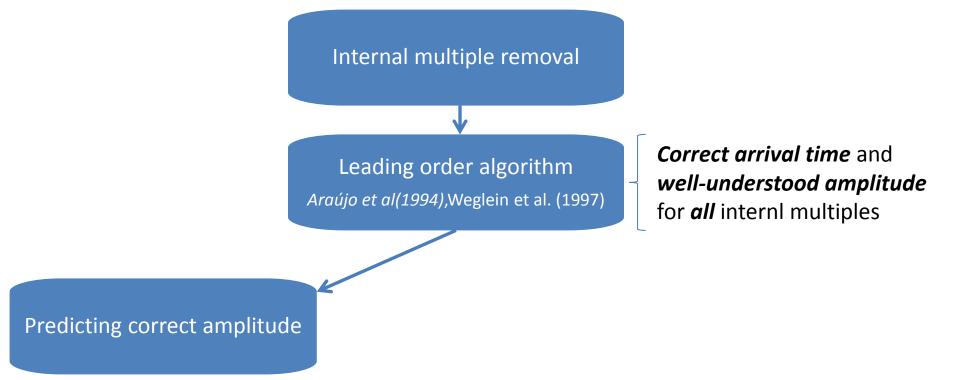


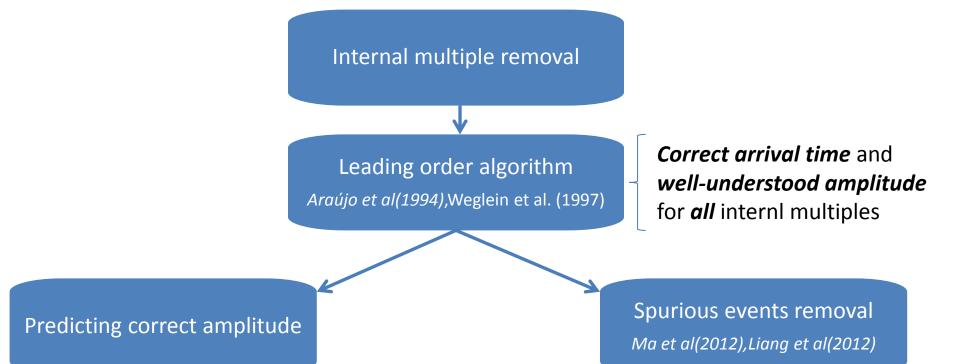
#### Internal multiple removal

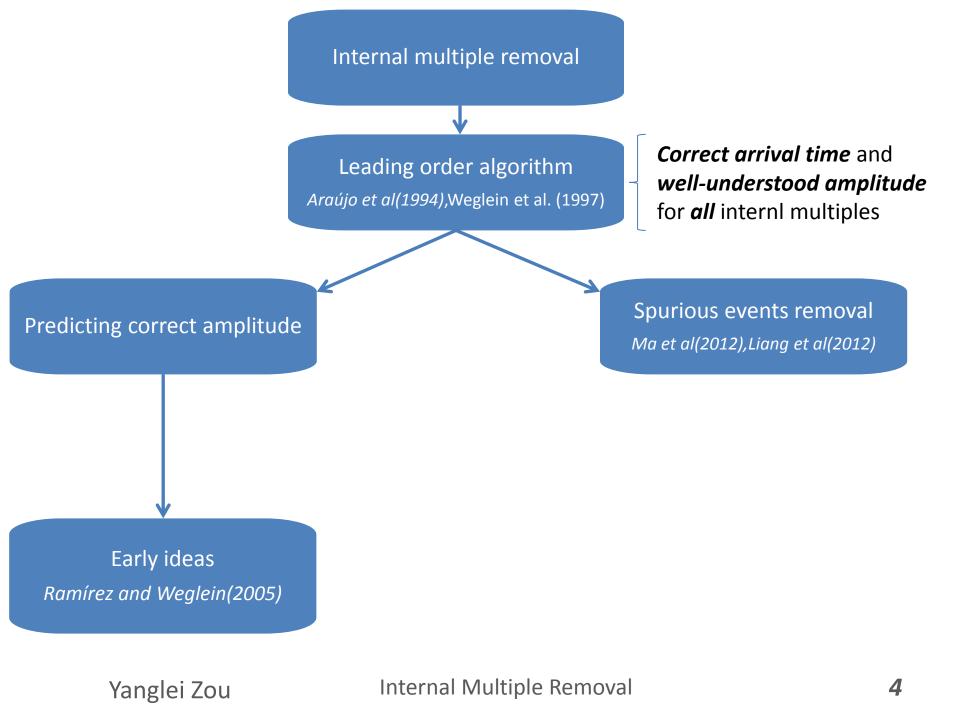
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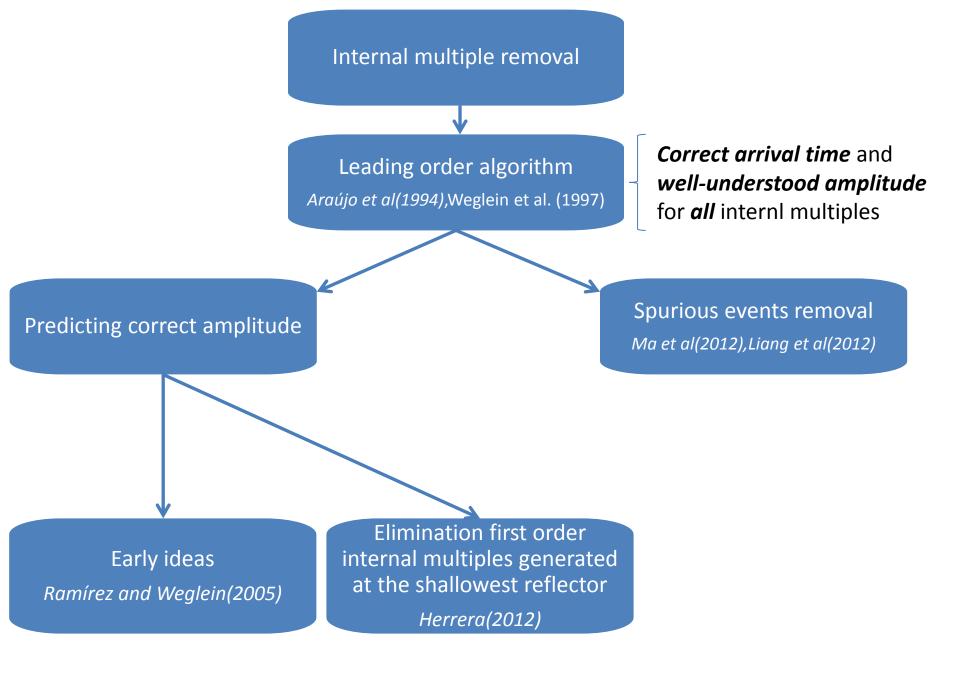




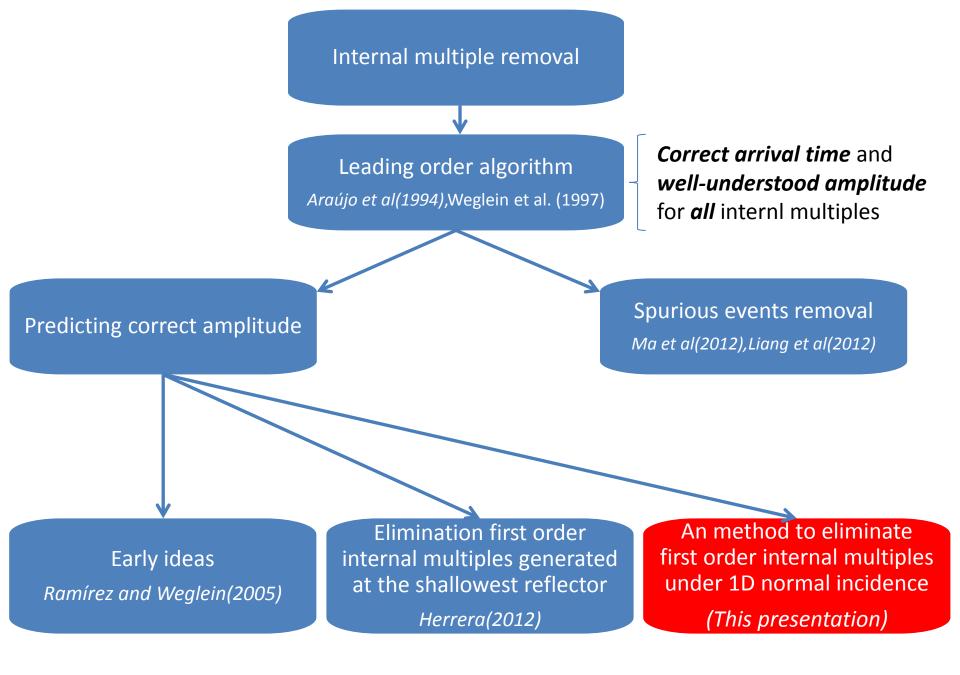








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## Goal



predict all internal multiples with
1)correct amplitude
2)correct time

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predict *all* internal multiples with 1)correct amplitude 2)correct time

## Leading order algorithm





predict *first order* internal multiples with 1)*correct time* 

#### 2) approximate amplitude

*Higher order* internal multiples with *correct time and well-understood amplitude* 

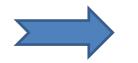
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#### The method in this presentation

data



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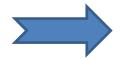
# Goal



predict *all* internal multiples with 1)correct amplitude 2)correct time

## Leading order algorithm





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data

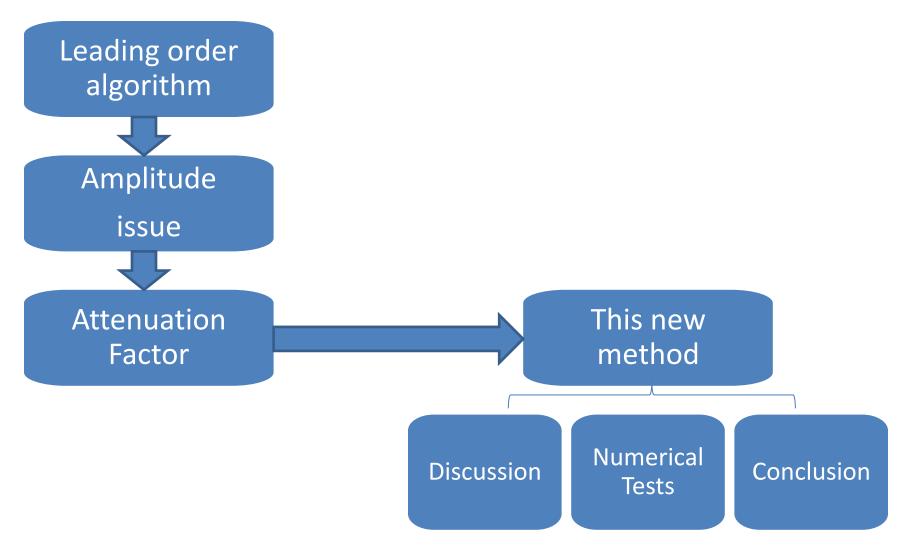


predict *first order* internal multiples with 1)*correct time* 

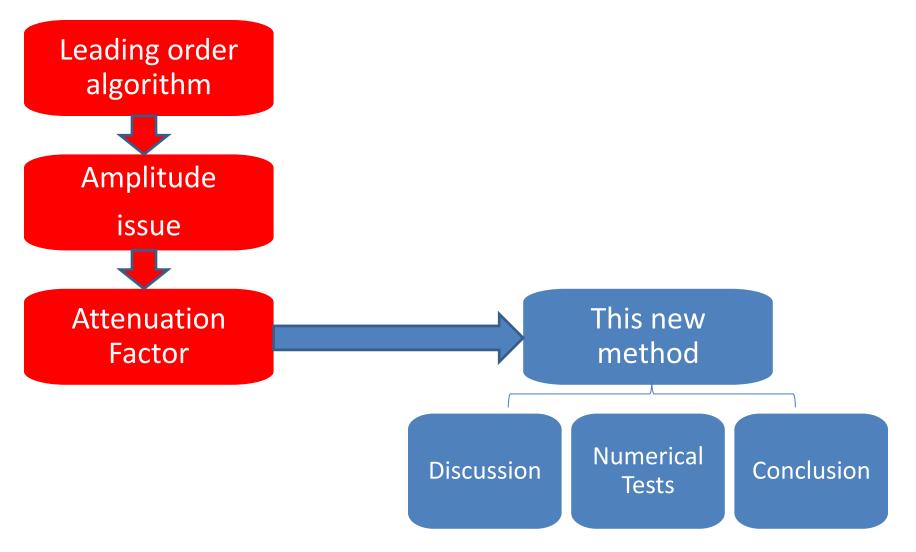
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## The structure of this presentation



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The 1D normal incidence version of the leading order algorithm given by Araújo et al.(1994) and Weglein et al. (1997) is presented as follows:

$$b_3^{IM}(k) = \int_{-\infty}^{\infty} dz e^{ikz} b_1(z) \int_{-\infty}^{z-\varepsilon_2} dz' e^{-ikz'} b_1(z') \int_{z'+\varepsilon_1}^{\infty} dz'' e^{ikz''} b_1(z'') \quad (1)$$

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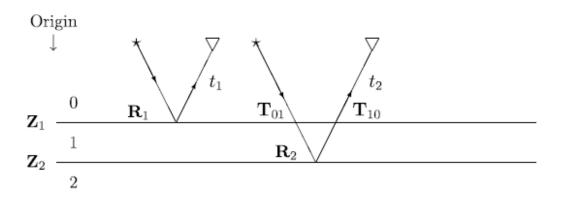
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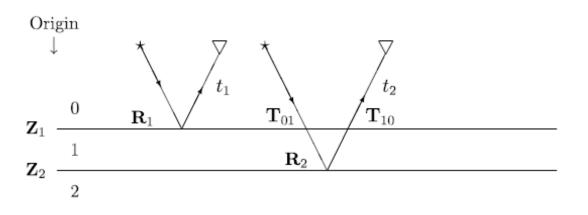
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Consider the simplest one-generator model example that can produce an internal multiple given by Weglein et al.(2003)



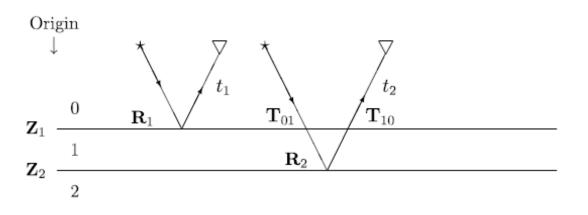
A one dimensional model with two interfaces.

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A one dimensional model with two interfaces.

$$D(t) = R_1 \delta(t - t_1) + T_{01} R_2 T_{10} \delta(t - t_2) + \cdots$$

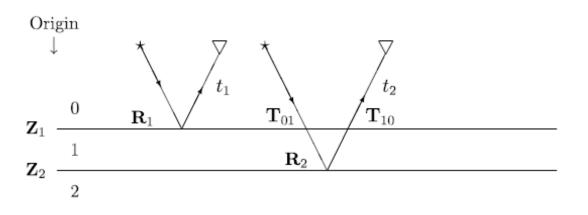


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$$D(t) = R_1 \delta(t - t_1) + T_{01} R_2 T_{10} \delta(t - t_2) + \cdots$$

Make a water speed migration of D(t) with  $k_z = \frac{2\omega}{c_0}$ and pseudo-depths:  $z_1 = \frac{c_0 t_1}{2}$   $z_2 = \frac{c_0 t_2}{2}$ .

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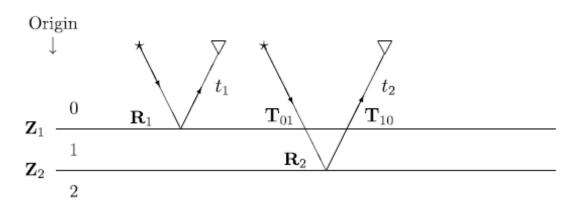
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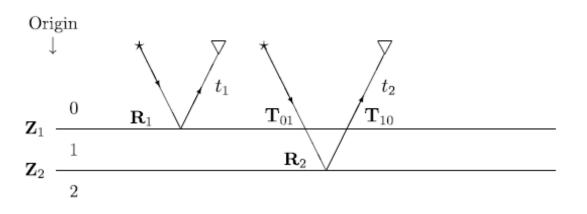
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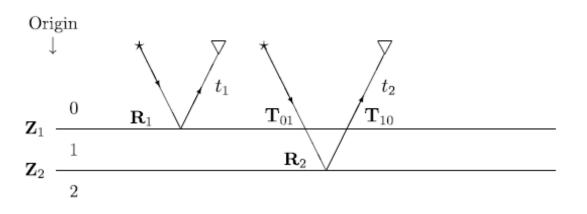
The date is now ready for the internal multiple algorithm.



A one dimensional model with two interfaces.

Substituting  $b_1(z)$  into the algorithm, we can derive the prediction (in the time domain):

$$b_3(t) = R_1 R_2^2 T_{01}^2 T_{10}^2 \delta(t - (2t_2 - t_1))$$



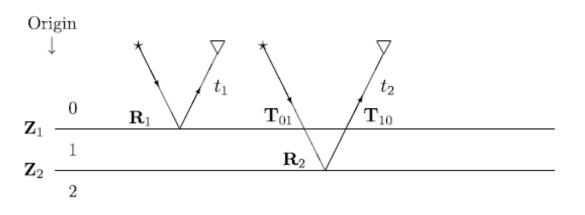
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From the example it is easy to compute the actual first order internal multiple precisely:

$$-R_1 R_2^2 T_{01} T_{10} \delta(t - (2t_2 - t_1))$$



A one dimensional model with two interfaces.

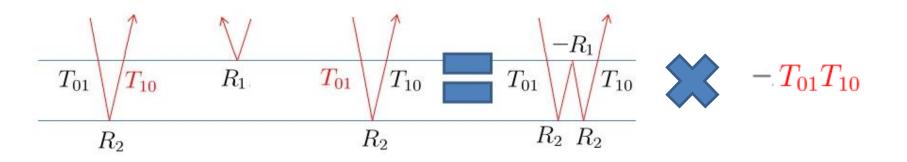
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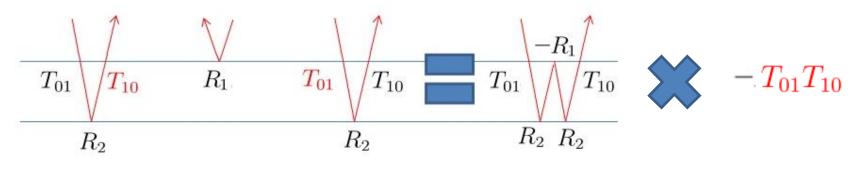
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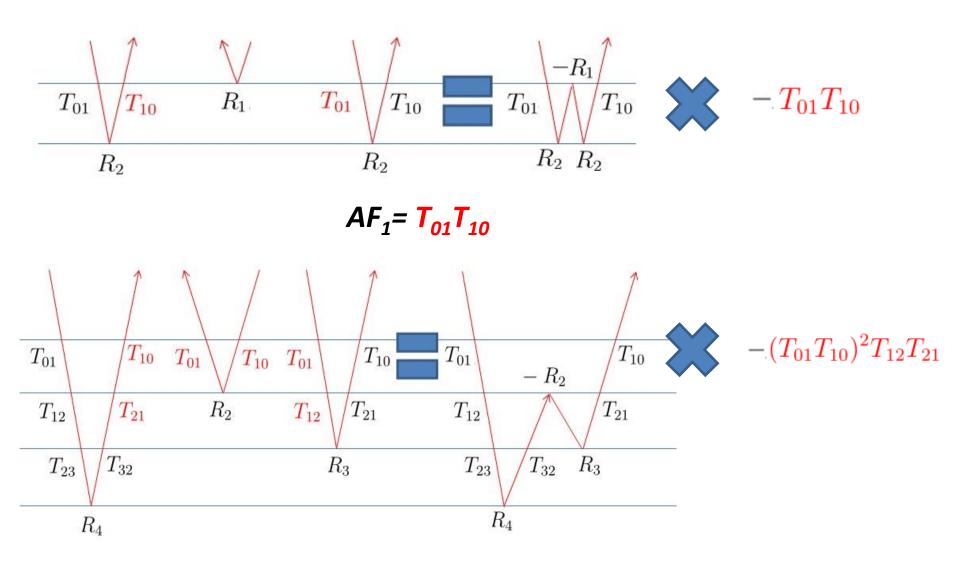
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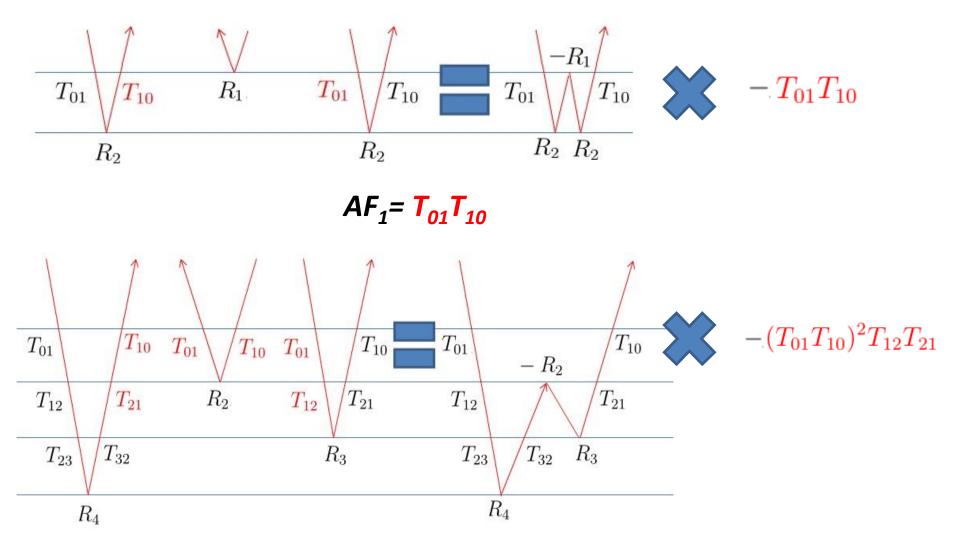
The time prediction is precise, and the amplitude of the prediction has an extra power of  $T_{01}T_{10}$  which is called the **attenuation factor**.





*AF*<sub>1</sub>= *T*<sub>01</sub>*T*<sub>10</sub>





 $AF_2 = (T_{01}T_{10})^2 T_{12}T_{21}$ 

(using transmission coefficients)

$$AF_{j} = \begin{cases} T_{0,1}T_{1,0} & (j=1) \\ \prod_{i=1}^{N-1} (T_{i-1,i}^{2}T_{i,i-1}^{2})T_{j,j-1}T_{j-1,j} & (1 < j < J) \end{cases}$$

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$$AF_{j} = \begin{cases} 1 - R_{1}^{2} & (j = 1) \\ (1 - R_{1}^{2})^{2} (1 - R_{2}^{2})^{2} \cdots (1 - R_{j-1}^{2})^{2} (1 - R_{j}^{2}) & (1 < j < J) \end{cases}$$

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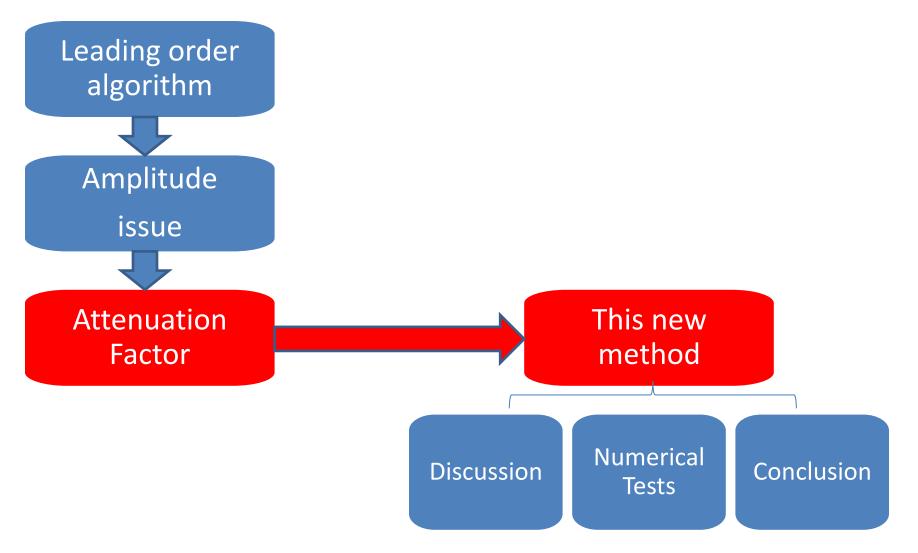
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 $AF_j$  is the attenuation factor for all first order internal multiples with a downward reflection at the j<sup>th</sup> reflector.

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#### The structure of this presentation



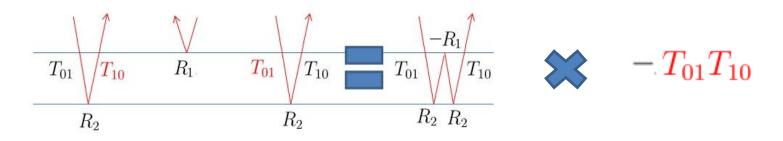
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$$b_3^{IM}(k) = \int_{-\infty}^{\infty} dz e^{ikz} b_1(z) \int_{-\infty}^{z-\varepsilon_2} dz' e^{-ikz'} b_1(z') \int_{z'+\varepsilon_1}^{\infty} dz'' e^{ikz''} b_1(z'')$$

$$b_E^{IM}(k) = \int_{-\infty}^{\infty} dz e^{ikz} b_1(z) \int_{-\infty}^{z-\varepsilon_2} dz' e^{-ikz'} F[b_1(z')] \int_{z'+\varepsilon_1}^{\infty} dz'' e^{ikz''} b_1(z'')$$

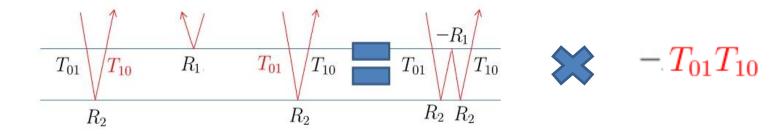
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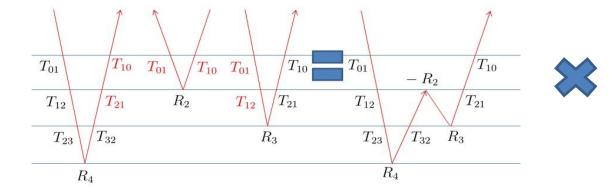
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 $-(T_{01}T_{10})^2T_{12}T_{21}$ 

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$$b_{1}(z) = R_{1}\delta(z-z_{1})$$

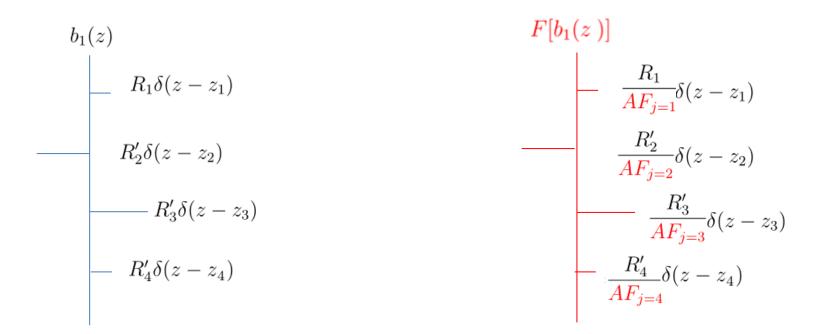
$$R_{2}\delta(z-z_{2}) = R_{3}'\delta(z-z_{3})$$

$$R_{4}'\delta(z-z_{4})$$

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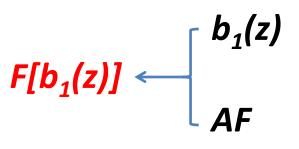
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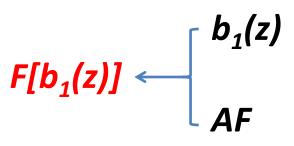
With the definition:  $R'_i = (1 - R_1^2)(1 - R_2^2) \cdots (1 - R_{i-2}^2)(1 - R_{i-1}^2) R_i$ 

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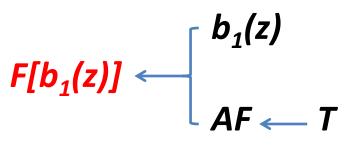
#### *F[b<sub>1</sub>(z)]*



$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \end{bmatrix} \qquad b_1(z)$$

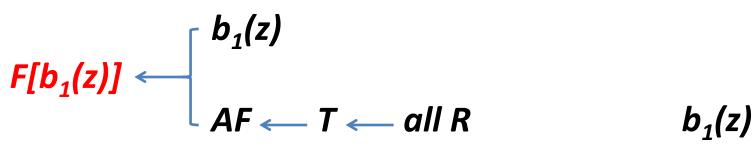


**b**<sub>1</sub>(z)



*b*<sub>1</sub>(*z*)

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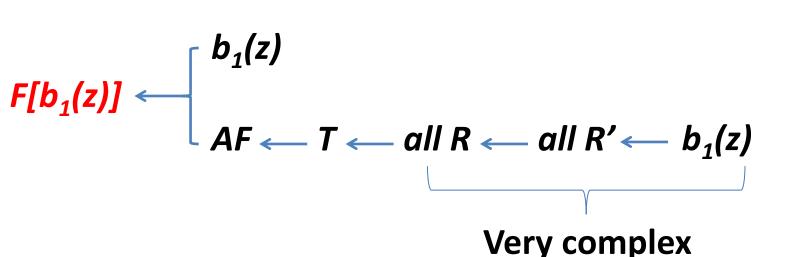


Internal Multiple Removal

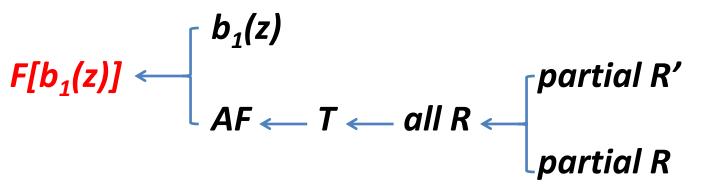
15







$$F[b_{1}(z)] \leftarrow \begin{bmatrix} b_{1}(z) \\ AF \leftarrow T \leftarrow all R \leftarrow a \leftarrow b_{1}(z) \\ Very complex \end{bmatrix}$$



$$F[b_{1}(z)] \leftarrow \begin{bmatrix} b_{1}(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \leftarrow \begin{bmatrix} b_{1}(z) \\ g(z) \end{bmatrix}$$

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# g(z) is a new function defined with R as coefficients. $g(z) = R_1\delta(z - z_1) + R_2\delta(z - z_2) + R_3\delta(z - z_3) + \dots + R_n\delta(z - z_n) + \dots$

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The coefficient of the i<sup>th</sup> term in F[b\_{1}(z)] is:  $\frac{R'_{i}}{AF_{i}}$   
 $\frac{R'_{i}}{AF_{i}} = \frac{R'_{i}}{(T_{01}T_{10})^{2}(T_{12}T_{21})^{2} \cdots (T_{i-2,i-1}T_{i-1,i-2})^{2}(T_{i-1,i}T_{i,i-1})}$ 

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$$\frac{R'_{i}}{(T_{01}T_{10})^{2}(T_{12}T_{21})^{2}\cdots(T_{i-2,i-1}T_{i-1,i-2})^{2}(T_{i-1,i}T_{i,i-1})}$$
$$= \frac{R'_{i}}{(1-R_{1}^{2})^{2}(1-R_{2}^{2})^{2}\cdots(1-R_{i-1}^{2})^{2}(1-R_{i}^{2})}$$
$$= \frac{R'_{i}}{(1-R_{1}R_{1}-R_{2}'R_{2}-\cdots-R'_{i-1}R_{i-1})^{2}(1-R_{i}^{2})}$$

$$F[b_{1}(z)] \leftarrow \begin{pmatrix} b_{1}(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{pmatrix} partial R' \\ partial R \end{pmatrix} \leftarrow \begin{pmatrix} b_{1}(z) \\ g(z) \leftarrow b_{1}(z) \end{pmatrix}$$
  
The coefficient of the i<sup>th</sup> term in F[b\_{1}(z)] is:  $\frac{R'_{i}}{AF_{i}}$ 
$$\frac{R'_{i}}{(T_{01}T_{10})^{2}(T_{12}T_{21})^{2}\cdots(T_{i-2,i-1}T_{i-1,i-2})^{2}(T_{i-1,i}T_{i,i-1})}$$
$$= \frac{R'_{i}}{(1-R_{1}^{2})^{2}(1-R_{2}^{2})^{2}\cdots(1-R_{i-1}^{2})^{2}(1-R_{i}^{2})}$$
$$= \frac{R'_{i}}{(1-R_{1}R_{1}-R'_{2}R_{2}-\cdots-R'_{i-1}R_{i-1})^{2}(1-R_{i}^{2})}$$
In the derivation, I used the expression:

$$\begin{array}{ll} (1-R_1^2)(1-R_2^2)\cdots(1-R_{i-2}^2)(1-R_{i-1}^2) \\ = 1-R_1R_1-R_2'R_2-\cdots-R_{i-1}'R_{i-1} \\ \\ \mbox{Yanglei Zou} & \mbox{Internal Multiple Removal} \end{array}$$

$$F[b_{1}(z)] \leftarrow \begin{bmatrix} b_{1}(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \leftarrow \begin{bmatrix} b_{1}(z) \\ g(z) \leftarrow b_{1}(z) \end{bmatrix}$$
  
The coefficient of the i<sup>th</sup> term in F[b\_{1}(z)] is:  $\frac{R'_{i}}{AF_{i}}$   
 $\frac{R'_{i}}{AF_{i}} = \frac{R'_{i}}{(1 - R_{1}R_{1} - R'_{2}R_{2} - \dots - R'_{i-1}R_{i-1})^{2}(1 - R_{i}^{2})}$ 

$$F[b_{1}(z)] \leftarrow \begin{bmatrix} b_{1}(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \leftarrow \begin{bmatrix} b_{1}(z) \\ g(z) \leftarrow b_{1}(z) \end{bmatrix}$$
  
The coefficient of the i<sup>th</sup> term in F[b\_{1}(z)] is:  $\frac{R'_{i}}{AF_{i}}$   
$$\frac{R'_{i}}{AF_{i}} = \frac{R'_{i}}{(1 - R_{1}R_{1} - R'_{2}R_{2} - \dots - R'_{i-1}R_{i-1})^{2}(1 - R_{i}^{2})}$$
  
Expressions we will use The coefficient of the i<sup>th</sup> term  $\binom{z^{z+\varepsilon}}{\sum_{z-\varepsilon}} dz'g(z')^{2} \longrightarrow R_{i}^{2}R_{1} + R'_{2}R_{2} + \dots + R'_{i-1}R_{i-1}$ 

$$F[b_{1}(z)] \leftarrow \begin{bmatrix} b_{1}(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \leftarrow \begin{bmatrix} b_{1}(z) \\ g(z) \leftarrow b_{1}(z) \end{bmatrix}$$
  
The coefficient of the i<sup>th</sup> term in F[b\_{1}(z)] is:  $\frac{R'_{i}}{AF_{i}}$   
 $\frac{R'_{i}}{AF_{i}} = \frac{R'_{i}}{(1 - R_{1}R_{1} - R'_{2}R_{2} - \dots - R'_{i-1}R_{i-1})^{2}(1 - R_{i}^{2})}$   
Expressions we will use The coefficient of the i<sup>th</sup> term  $(\int_{z-\varepsilon}^{z+\varepsilon} dz'g(z'))^{2} \longrightarrow R_{i}^{2}$   
 $\int_{-\infty}^{z-\varepsilon} dz'b_{1}(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz''g(z'') \longrightarrow R_{1}R_{1} + R'_{2}R_{2} + \dots + R'_{i-1}R_{i-1}$   
 $F[b_{1}(z)] = \frac{b_{1}(z)}{[1 - (\int_{z-\varepsilon}^{z+\varepsilon} dz'g(z'))^{2}][1 - \int_{-\infty}^{z-\varepsilon} dz'b_{1}(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz''g(z'')]^{2}}$ 

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \leftarrow \begin{bmatrix} b_1(z) \\ g(z) \leftarrow b_1(z) \end{bmatrix}$$

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \begin{bmatrix} b_1(z) \\ g(z) \leftarrow b_1(z) \end{bmatrix}$$

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \begin{bmatrix} b_1(z) \\ g(z) \leftarrow b_1(z) \end{bmatrix}$$

$$R_{i} = \frac{R'_{i}}{(1 - R_{1}^{2})(1 - R_{2}^{2})\cdots(1 - R_{i-2}^{2})(1 - R_{i-1}^{2})}$$
$$= \frac{R'_{i}}{1 - R_{1}R_{1} - R'_{2}R_{2} - \cdots - R'_{i-1}R_{i-1}}$$

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \begin{bmatrix} b_1(z) \\ g(z) \leftarrow b_1(z) \end{bmatrix}$$

$$R_{i} = \frac{R'_{i}}{(1 - R_{1}^{2})(1 - R_{2}^{2})\cdots(1 - R_{i-2}^{2})(1 - R_{i-1}^{2})}$$
$$= \frac{R'_{i}}{1 - R_{1}R_{1} - R'_{2}R_{2} - \cdots - R'_{i-1}R_{i-1}}$$

Expression we will use The coefficient of the i<sup>th</sup> term  $\int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' g(z'') \longrightarrow R_1 R_1 + R'_2 R_2 + \dots + R'_{i-1} R_{i-1}$ 

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \begin{bmatrix} b_1(z) \\ g(z) \leftarrow b_1(z) \end{bmatrix}$$

$$R_{i} = \frac{R'_{i}}{(1 - R_{1}^{2})(1 - R_{2}^{2})\cdots(1 - R_{i-2}^{2})(1 - R_{i-1}^{2})}$$
$$= \frac{R'_{i}}{1 - R_{1}R_{1} - R'_{2}R_{2} - \cdots - R'_{i-1}R_{i-1}}$$

Expression we will use The coefficient of the i<sup>th</sup> term  $\int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' g(z'') \longrightarrow R_1 R_1 + R'_2 R_2 + \dots + R'_{i-1} R_{i-1}$   $g(z) = \frac{b_1(z)}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' g(z'')}$ 

Yanglei Zou

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all \ R \leftarrow \begin{bmatrix} partial \ R' \end{bmatrix} \leftarrow \begin{bmatrix} b_1(z) \\ b_1(z) \end{bmatrix} \begin{bmatrix} b_1(z) \\ c \leftarrow b_1(z) \end{bmatrix}$$

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all \ R \leftarrow \begin{bmatrix} partial \ R' \\ partial \ R \end{bmatrix} \leftarrow \begin{bmatrix} b_1(z) \\ g(z) \leftarrow b_1(z) \end{bmatrix}$$

$$F[b_1(z)] \leftarrow b_1(z)$$

$$F[b_1(z)] \leftarrow \begin{bmatrix} b_1(z) \\ AF \leftarrow T \leftarrow all R \leftarrow \begin{bmatrix} partial R' \\ partial R \end{bmatrix} \leftarrow \begin{bmatrix} b_1(z) \\ g(z) \leftarrow b_1(z) \end{bmatrix}$$

$$F[b_1(z)] \leftarrow b_1(z)$$

$$b_E^{IM}(k) = \int_{-\infty}^{\infty} dz e^{ikz} b_1(z) \int_{-\infty}^{z-\varepsilon_2} dz' e^{-ikz'} F[b_1(z')] \int_{z'+\varepsilon_1}^{\infty} dz'' e^{ikz''} b_1(z'')$$

$$F[b_1(z)] = \frac{b_1(z)}{[1 - (\int_{z-\varepsilon}^{z+\varepsilon} dz'g(z'))^2][1 - \int_{-\infty}^{z-\varepsilon} dz'b_1(z')\int_{z'-\varepsilon}^{z'+\varepsilon} dz''g(z'')]^2}$$
(5)

$$g(z) = \frac{b_1(z)}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' g(z'')}$$
(6)

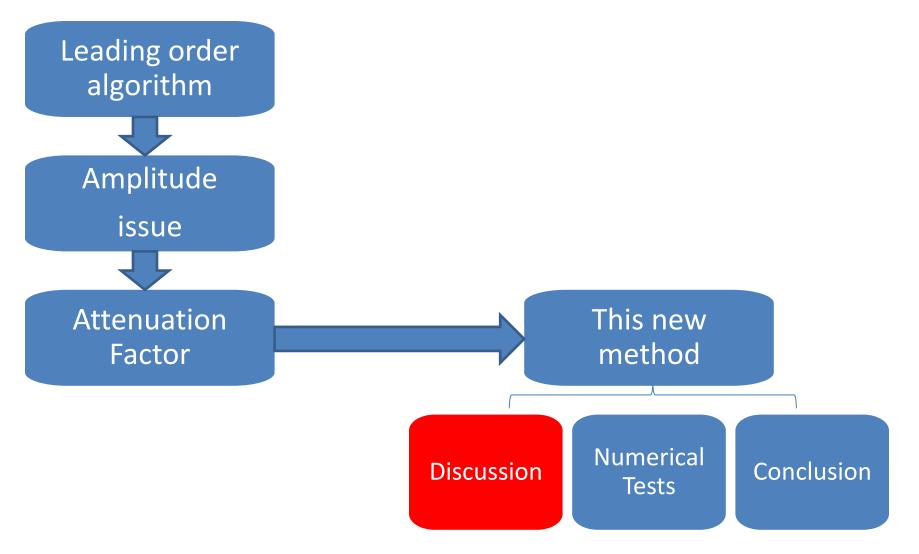
Yanglei Zou

Internal Multiple Removal

20

| First type<br>approximation<br>for equation(6)  | $g(z) = \frac{b_1(z)}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' g(z'')}$ $\approx \frac{b_1(z)}{1 - 0}$ $\approx b_1(z)$   |
|---|---|
| Second type<br>approximation<br>for equation(6) | $g(z) = \frac{b_1(z)}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' g(z'')} \approx \frac{b_1(z)}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' b_1(z'')}$  |
| Third type<br>approximation<br>for equation(6)  | $g(z) = \frac{b_1(z)}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' g(z'')} \\ \approx \frac{b_1(z)}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_1(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' \frac{b_1(z'')}{1 - \int_{-\infty}^{z''-\varepsilon} dz''' b_1(z'') \int_{z''-\varepsilon}^{z'''+\varepsilon} dz^{(4)} b_1(z^{(4)})}}$ |

#### The structure of this presentation



#### Discussion

This method considers only primaries as the input.

## Discussion

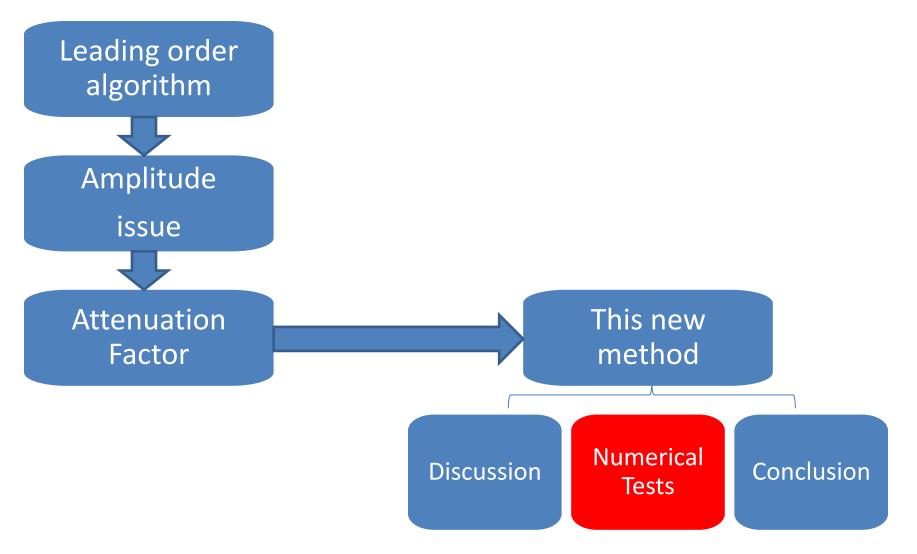
The input data contains both primaries and internal multiples.

| First type<br>approximation  | all IM <sub>j=1</sub> : correct<br>all IM <sub>j&gt;1</sub> : more accurate than<br>internal multiple attenuator |  |
|------------------------------|--|--|
| Second type<br>approximation | all $IM_{j=1}$ and $IM_{j=2}$ : correct<br>all $IM_{j>2}$ : more accurate than the first type                    |  |
| Third type<br>approximation  | Internal multiples<br>arrive <b>after</b> the<br>3 <sup>rd</sup> primary   | all $IM_{j=1}$ , all $IM_{j=2}$ and all $IM_{j=3}$<br>:correct<br>all $IM_{j>3}$ : more accurate than<br>the second type |
|                              | Internal multiples<br>arrive <b>before</b> the<br>3 <sup>rd</sup> primary  | all $IM_{j=1}$ and all $IM_{j=2}$ :correct<br>all $IM_{j>2}$ : more accurate than<br>the second type                     |
|                              |  |  |

#### Discussion

To deal with this problem, we can first run the internal multiple attenuation algorithm, then attenuate the amplitude of internal multiples in the data and then run this method using the new data to eliminate all first order internal multiples.

#### The structure of this presentation



In this section we test 3 different equations under 1D normal incidence:

(1) internal multiple attenuator

(2)First type of approximation of the new method.

(3)Second type of approximation of the new method.

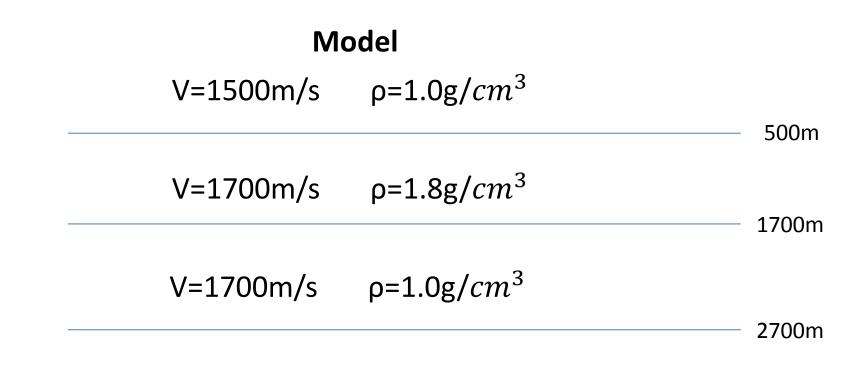
$$b_{3}^{IM}(k) = \int_{-\infty}^{\infty} dz e^{ikz} b_{1}(z) \int_{-\infty}^{z-\varepsilon_{2}} dz' e^{-ikz'} b_{1}(z') \int_{z'+\varepsilon_{1}}^{\infty} dz'' e^{ikz''} b_{1}(z'')$$

$$b_{E}^{IM}(k) = \int_{-\infty}^{\infty} dz e^{ikz} b_{1}(z) \int_{-\infty}^{z-\varepsilon_{2}} dz' e^{-ikz'} F[b_{1}(z')] \int_{z'+\varepsilon_{1}}^{\infty} dz'' e^{ikz''} b_{1}(z'')$$

$$F[b_{1}(z)]_{1T} = \frac{b_{1}(z)}{1 - (\int_{z-\varepsilon}^{z+\varepsilon} dz' b_{1}(z'))^{2}}$$

$$F[b_{1}(z)]_{2T} = \frac{b_{1}(z)}{[1 - (\frac{\int_{z-\varepsilon}^{z+\varepsilon} dz' b_{1}(z')}{1 - \int_{-\infty}^{z-\varepsilon} dz' b_{1}(z')})^{2}][1 - \int_{-\infty}^{z-\varepsilon} dz' b_{1}(z') \int_{z'-\varepsilon}^{z'+\varepsilon} dz'' b_{1}(z'')}$$

Internal Multiple Removal



V=3500m/s  $\rho$ =4.0g/cm<sup>3</sup>

5700m

V=5000m/s 
$$\rho$$
=4.0g/cm<sup>3</sup>

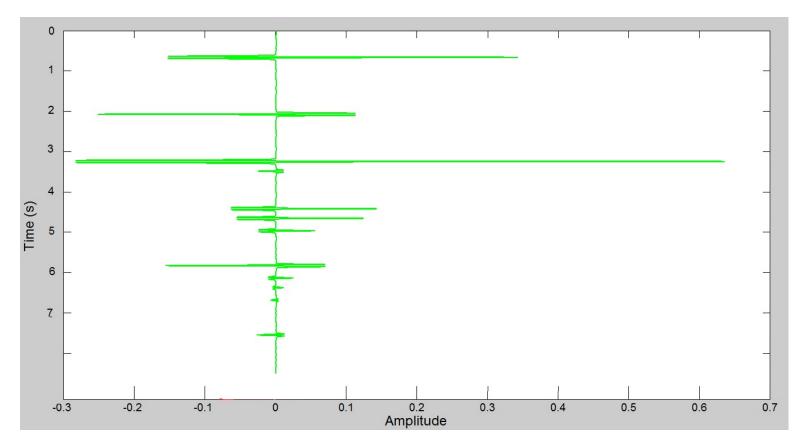
# **Initial Tests**

- A. Test for perfect data.
- B. Test for bandlimited data.
- C. Test for data with white noise.

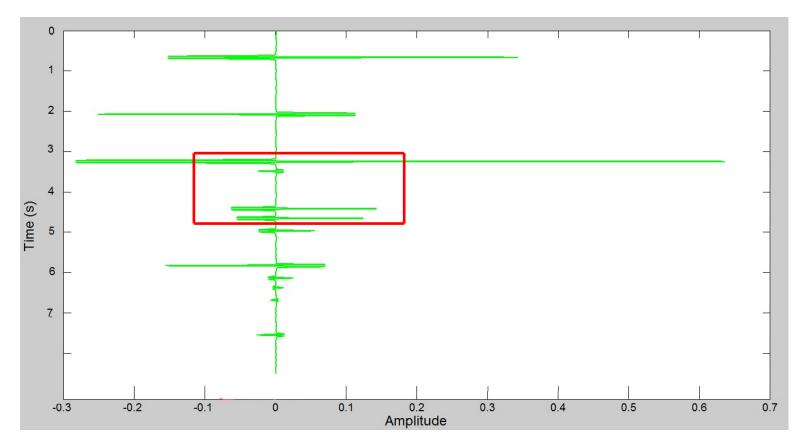
# **Initial Tests**

- A. Test for perfect data.
- B. Test for bandlimited data.
- C. Test for data with white noise.

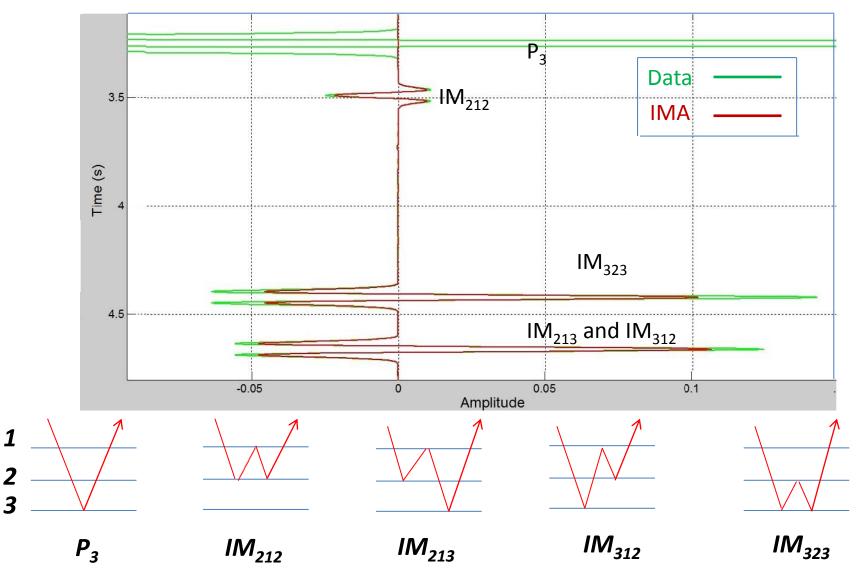
## The input data(1D normal incidence)



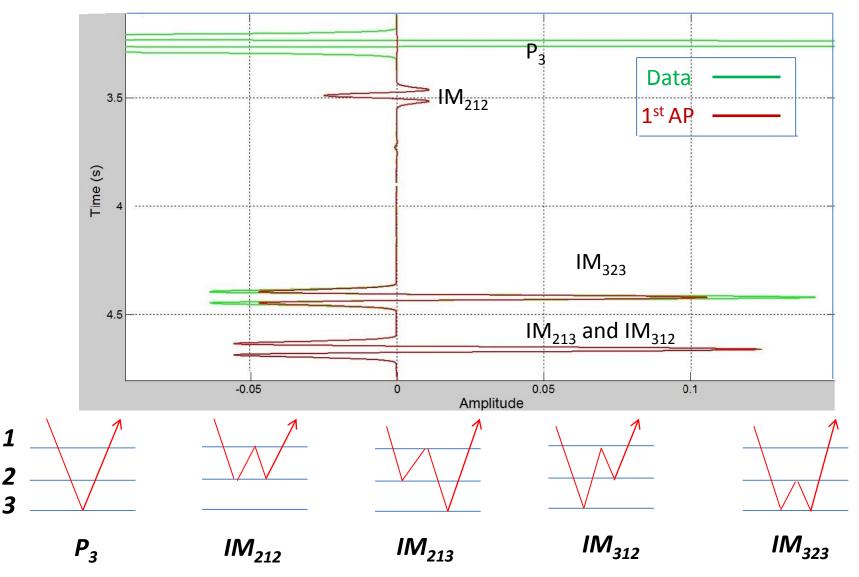
## The input data(1D normal incidence)



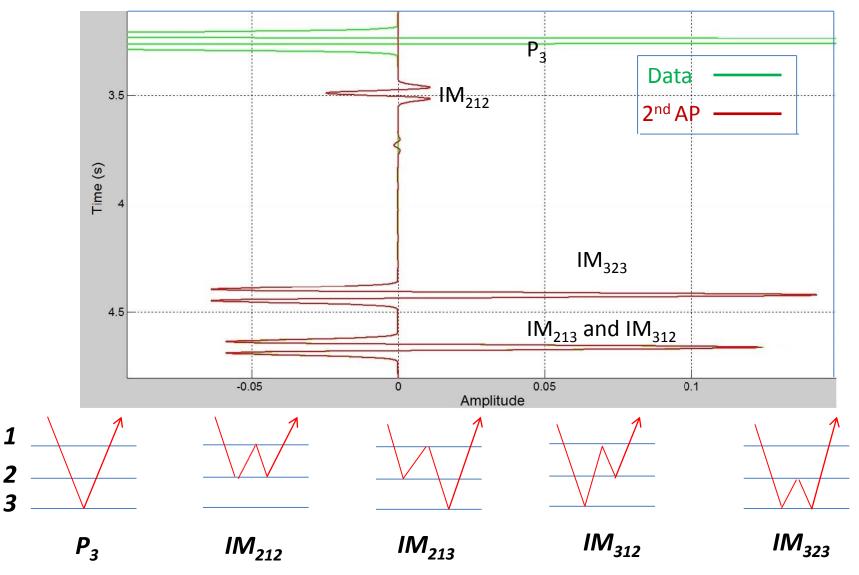
### Internal multiple attenuator



### First type of equation approximation



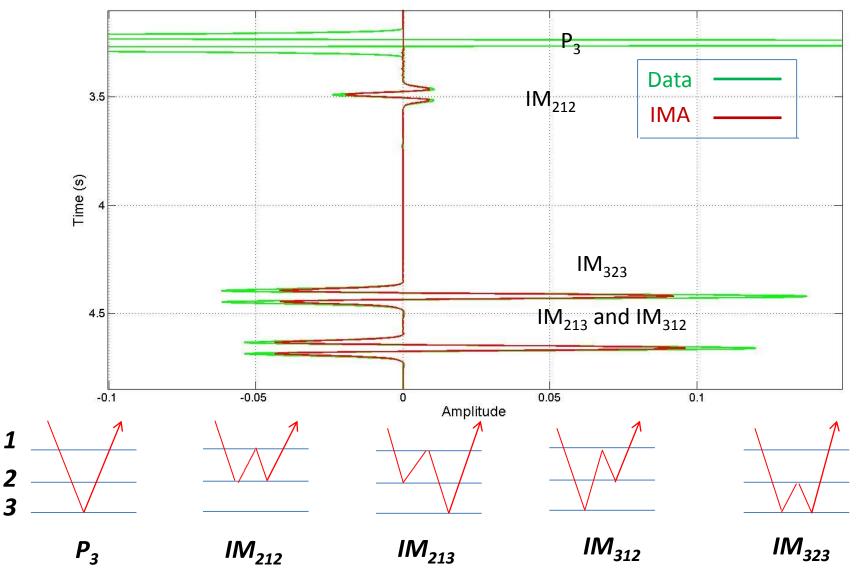
### Second type of equation approximation

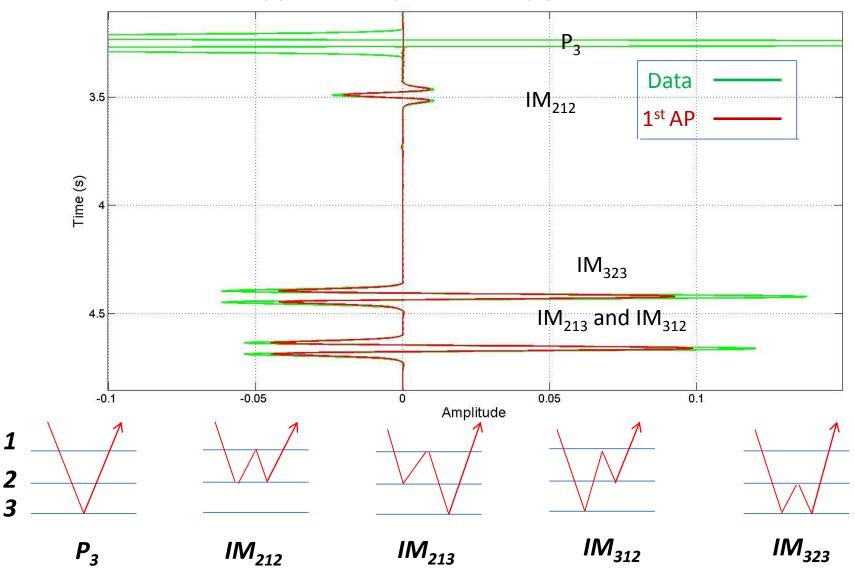


# **Initial Tests**

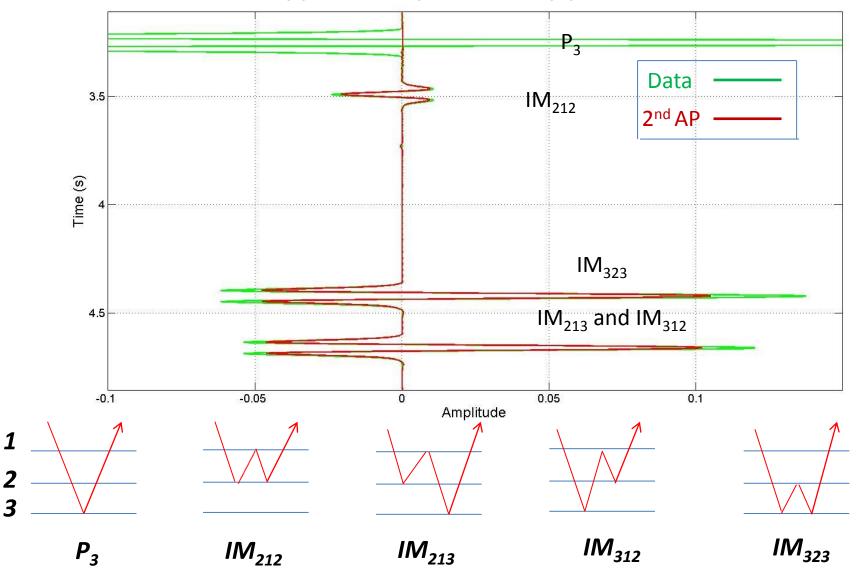
- A. Test for perfect data.
- B. Test for bandlimited data.
- C. Test for data with white noise.

#### Internal multiple attenuator





#### First type of equation approximation



#### Second type of equation approximation

# **Initial Tests**

- A. Test for perfect data.
- B. Test for bandlimited data.
- C. Test for data with white noise.

#### $P_3$ Data 3.5 IM<sub>212</sub> IMA Time (s) IM<sub>323</sub> 4.5 $IM_{213}$ and $IM_{312}$ -0.05 0.05 0.1 -0.1 0 Amplitude IM<sub>312</sub> IM<sub>323</sub> IM<sub>213</sub>

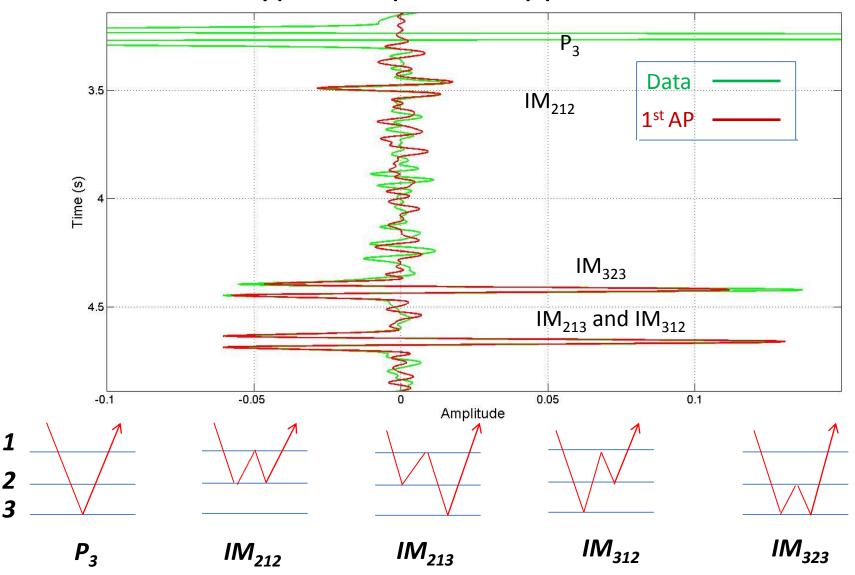
IM<sub>212</sub>

1

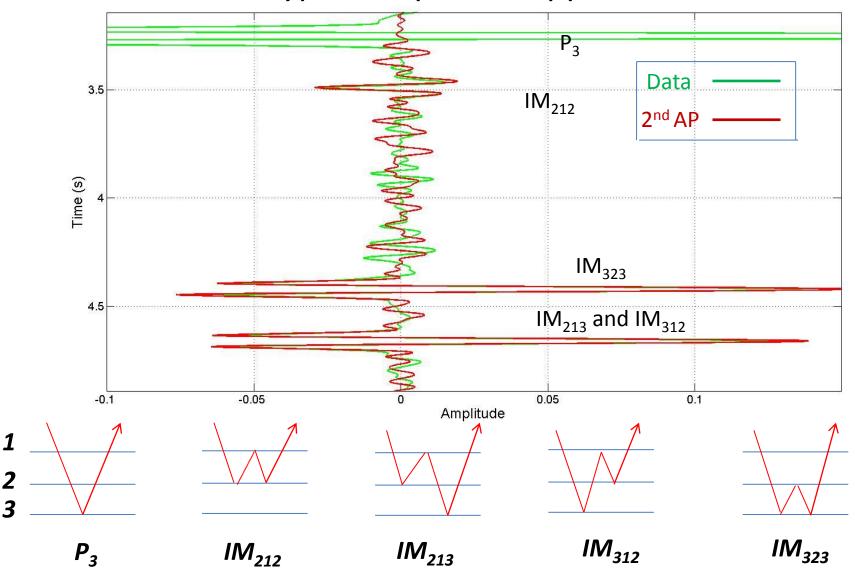
2 3

**P**<sub>3</sub>

### Internal multiple attenuator

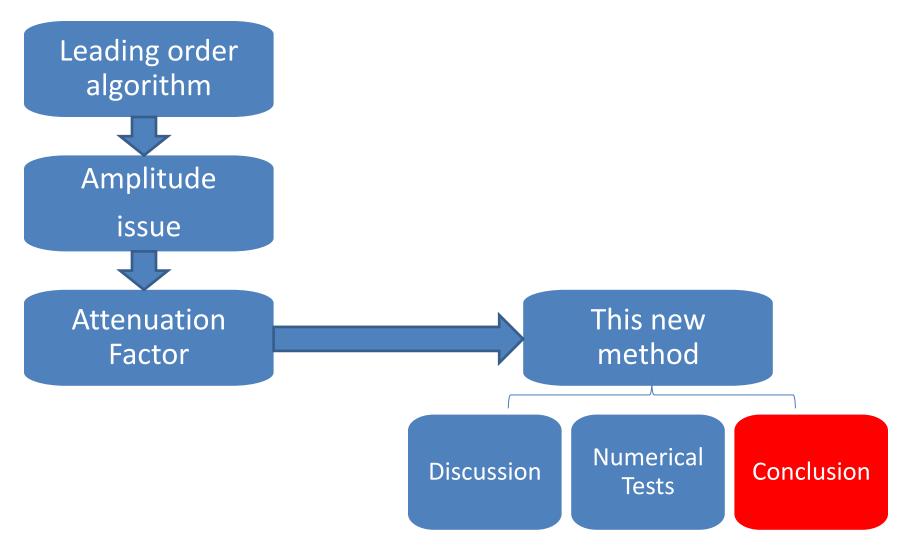


#### First type of equation approximation



#### Second type of equation approximation

#### The structure of this presentation



Yanglei Zou

Internal Multiple Removal

## Conclusion

Predicting correct amplitude of internal multiples is an important problem.

1. In this presentation, a new method is given to eliminate all first order internal multiples under 1D normal incidence directly in terms of data without determining the earth.

 As one of the equations in the method is an integral equation , we can make different types of approximations to it and achieve different levels of delivery by using different orders of approximations.

## Conclusion

3. This method considers only primaries in the data  $(b_1(z))$ .

To address this issue, we can first run the internal multiple attenuation algorithm, then attenuate the amplitude of internal multiples in the data and then run this method using the new data to eliminate all first order internal multiples.

4. From the test we can see, this method is robust to bandwith and noise.

## Conclusion

5. This method is a part of a project which is aimed at predicting correct amplitude and time of all internal multiples. This method is a step within seeking this purpose.

The project may be relevant and provide value when primaries and internal multiples interfere with each other in both on-shore and off-shore data.



#### U.S. Federal Government Research Support



