

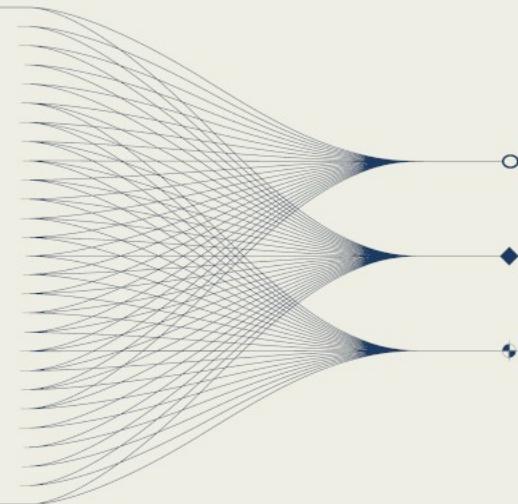
Locating aftershocks using sparse PSM networks

Rolf Häfner, Manfred Joswig and Marco Walter

Sonicona GbR, Germany – www.sonicona.com

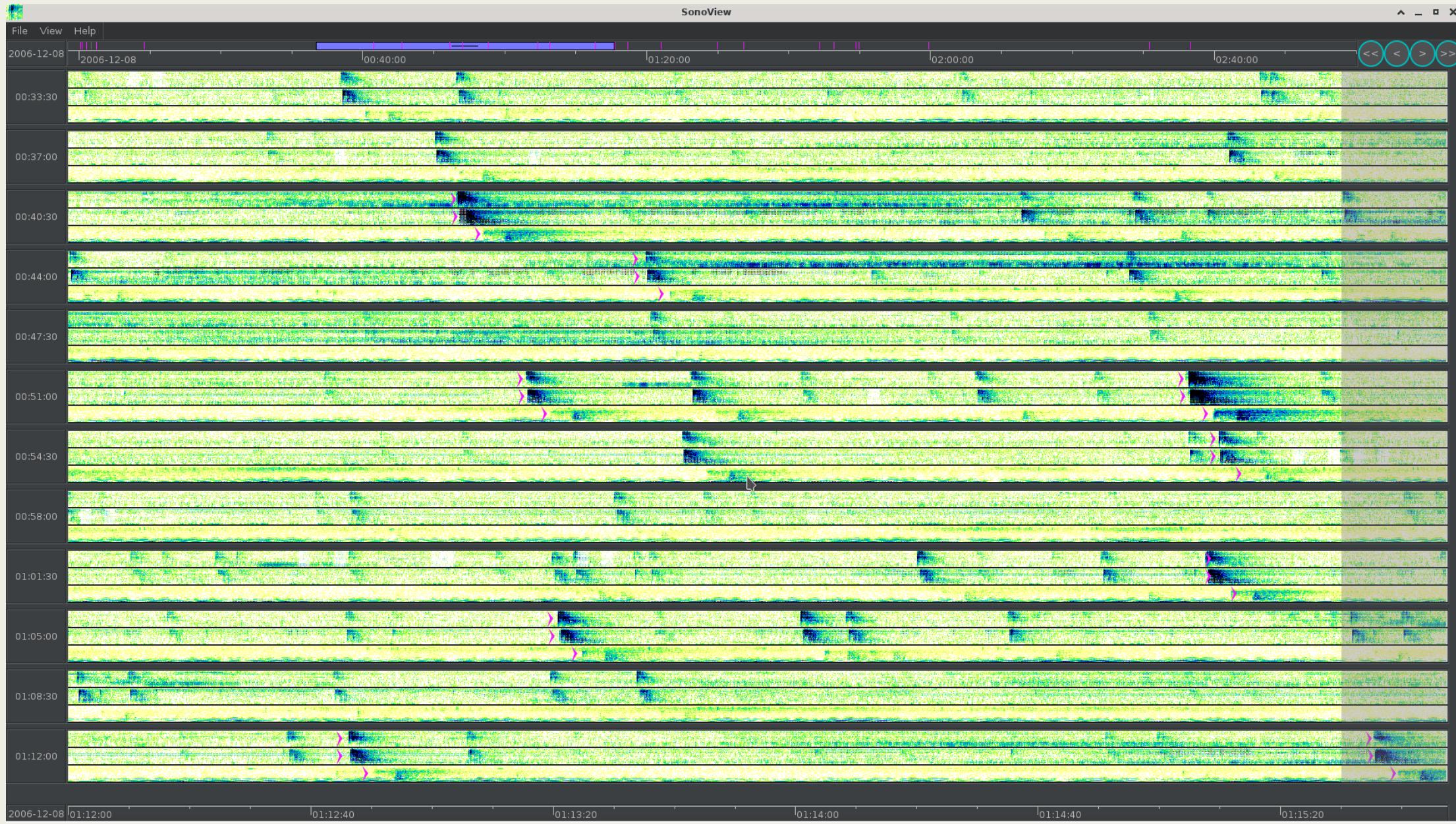


Presentation Date: 11 September 2025, 17:15



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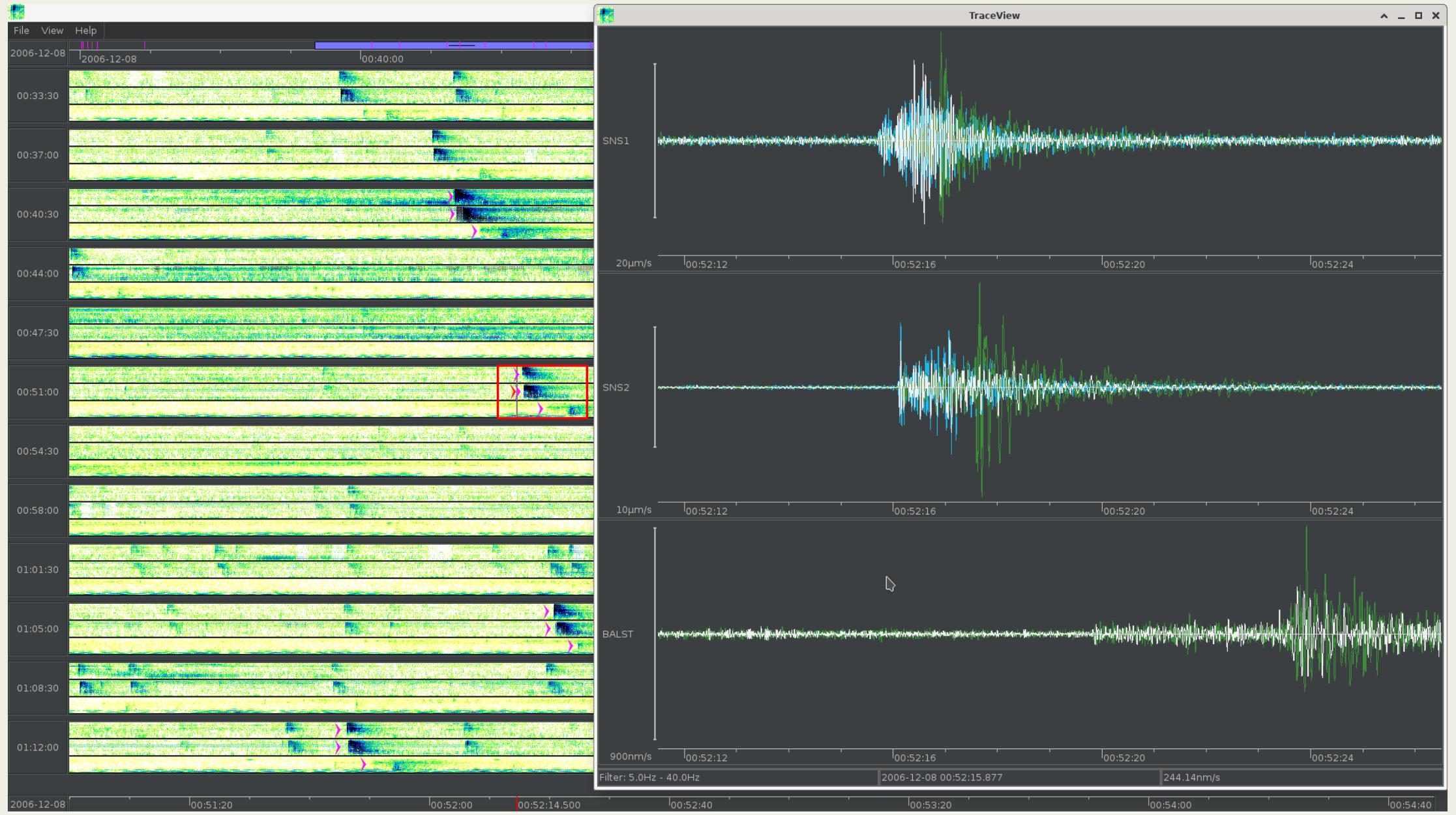
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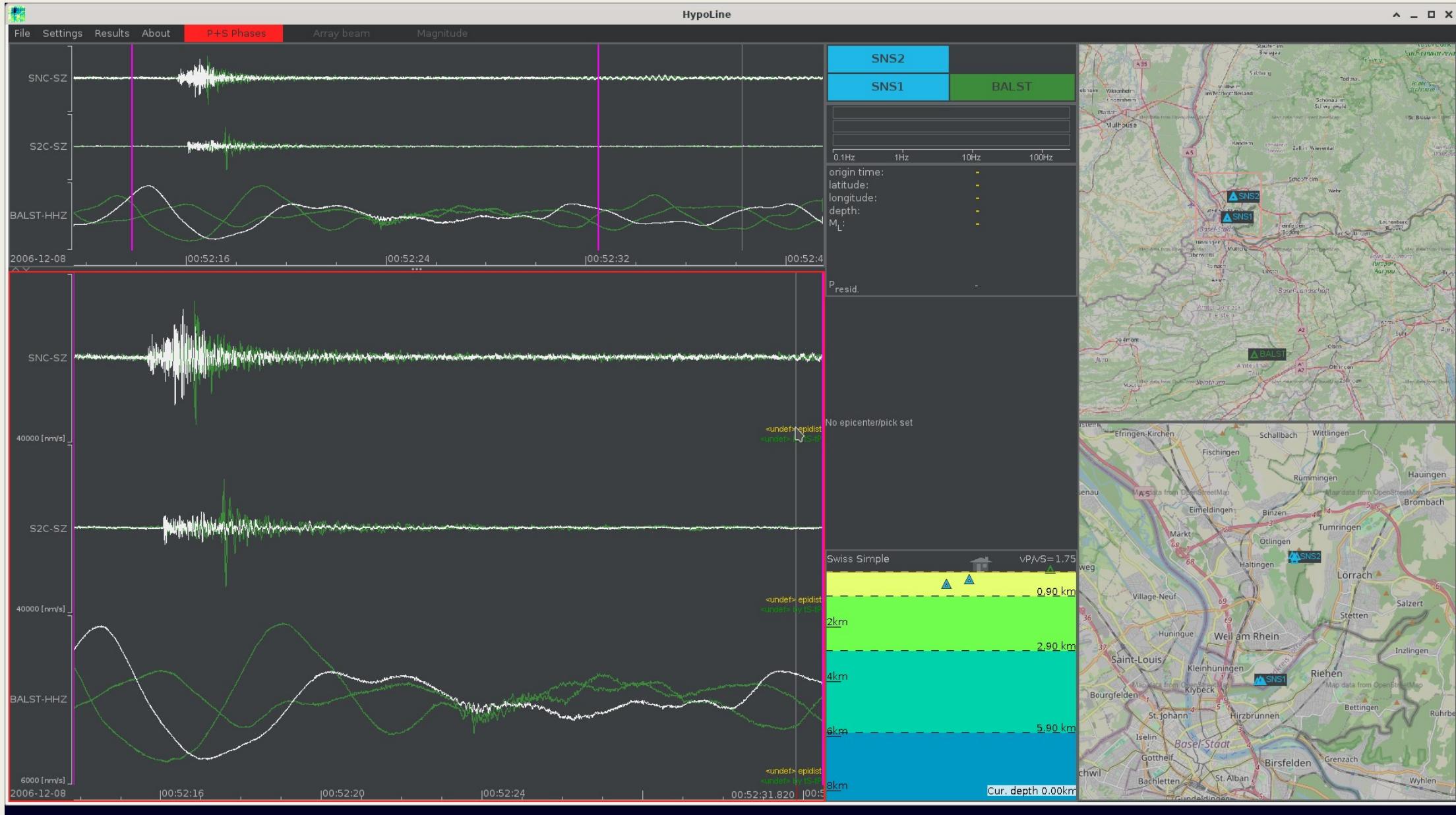
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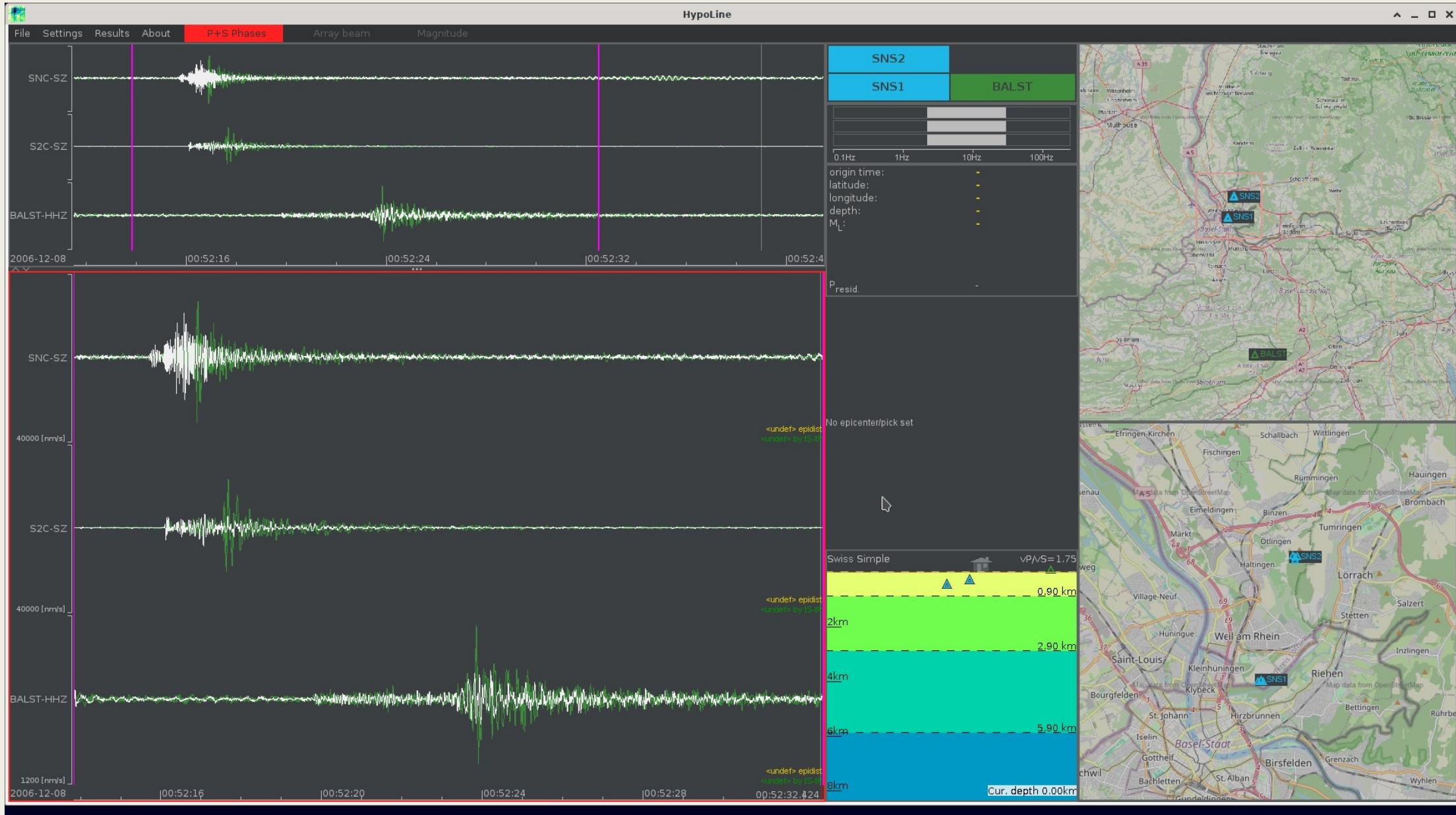
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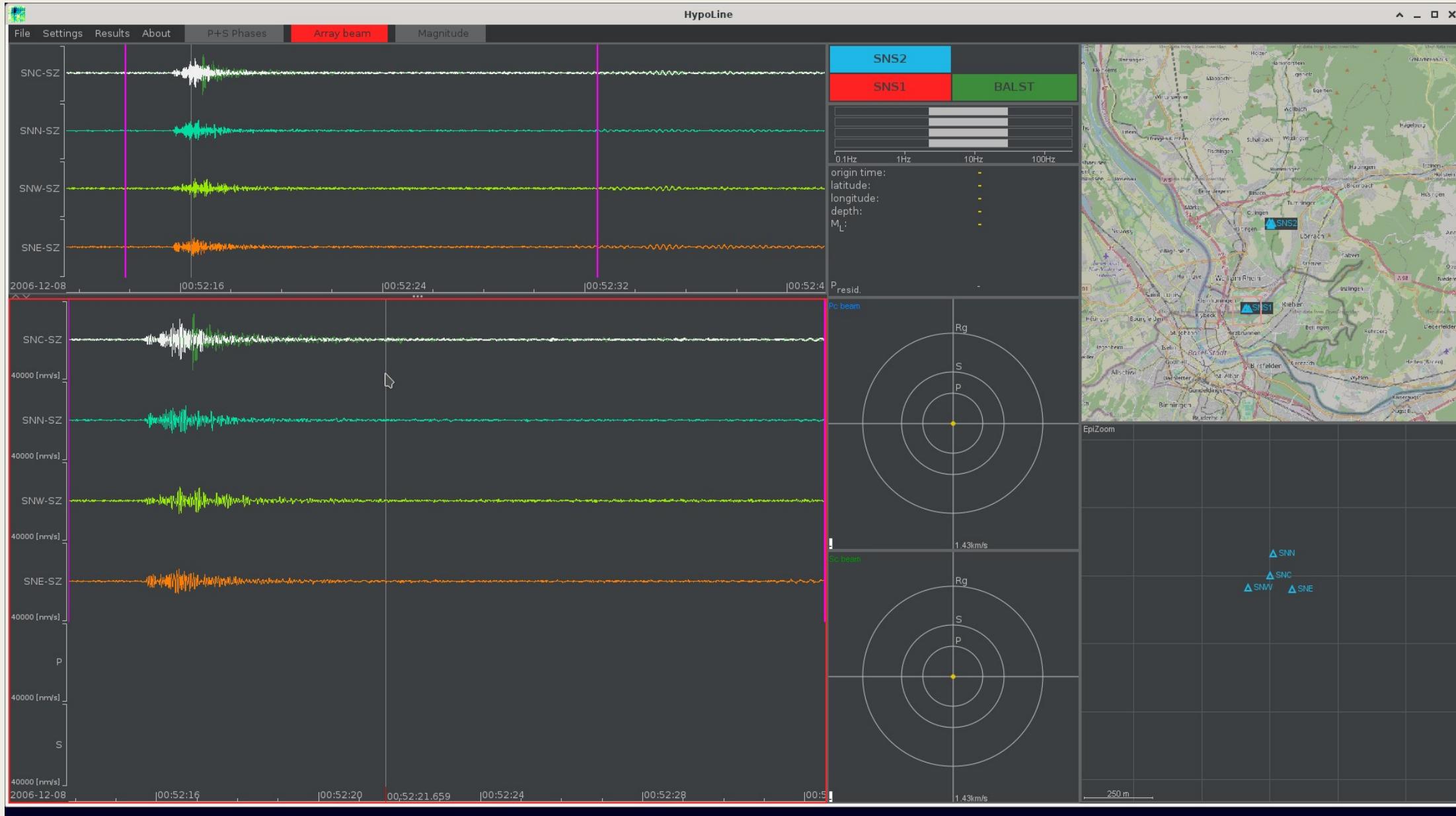
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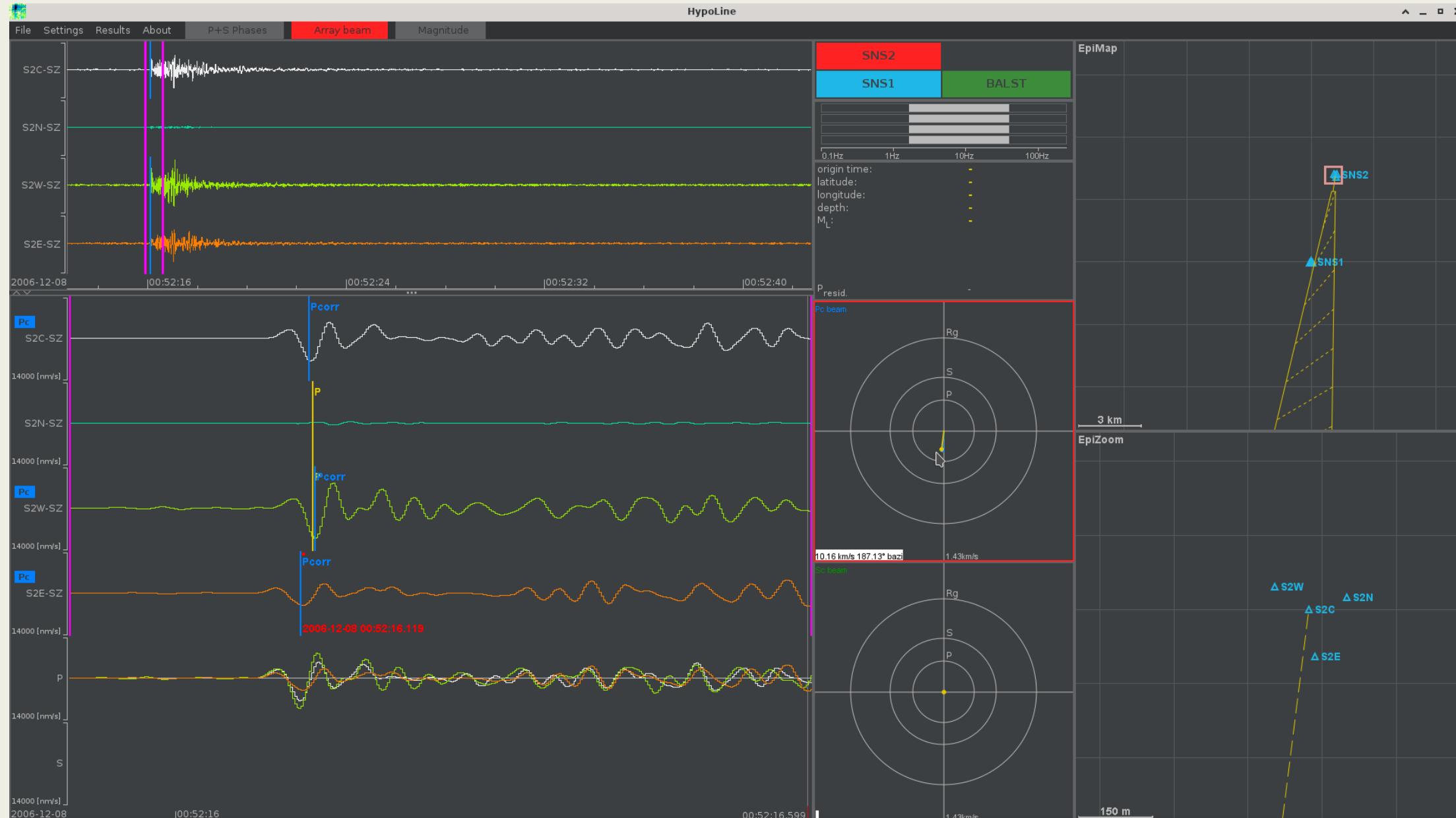
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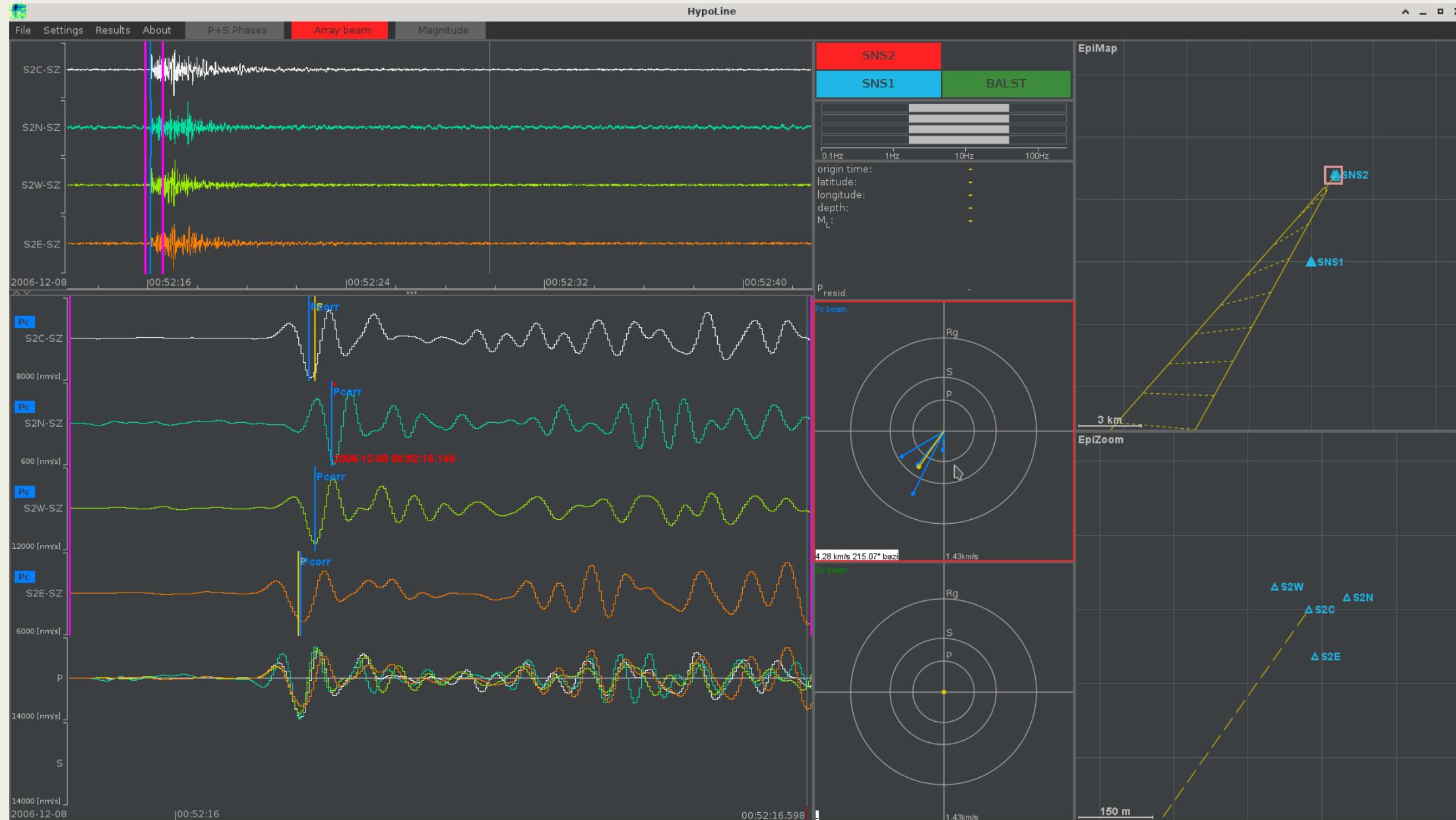


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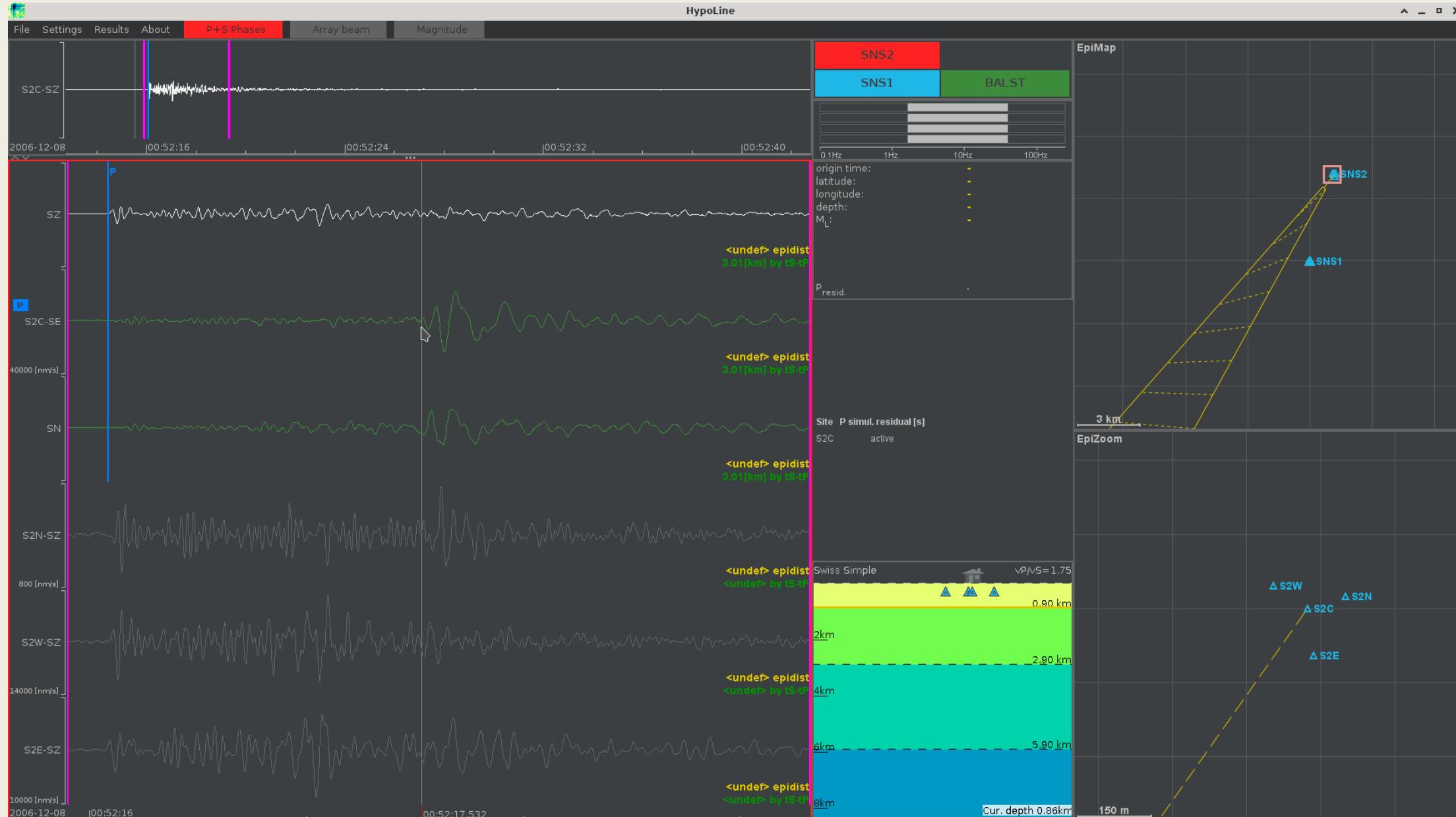
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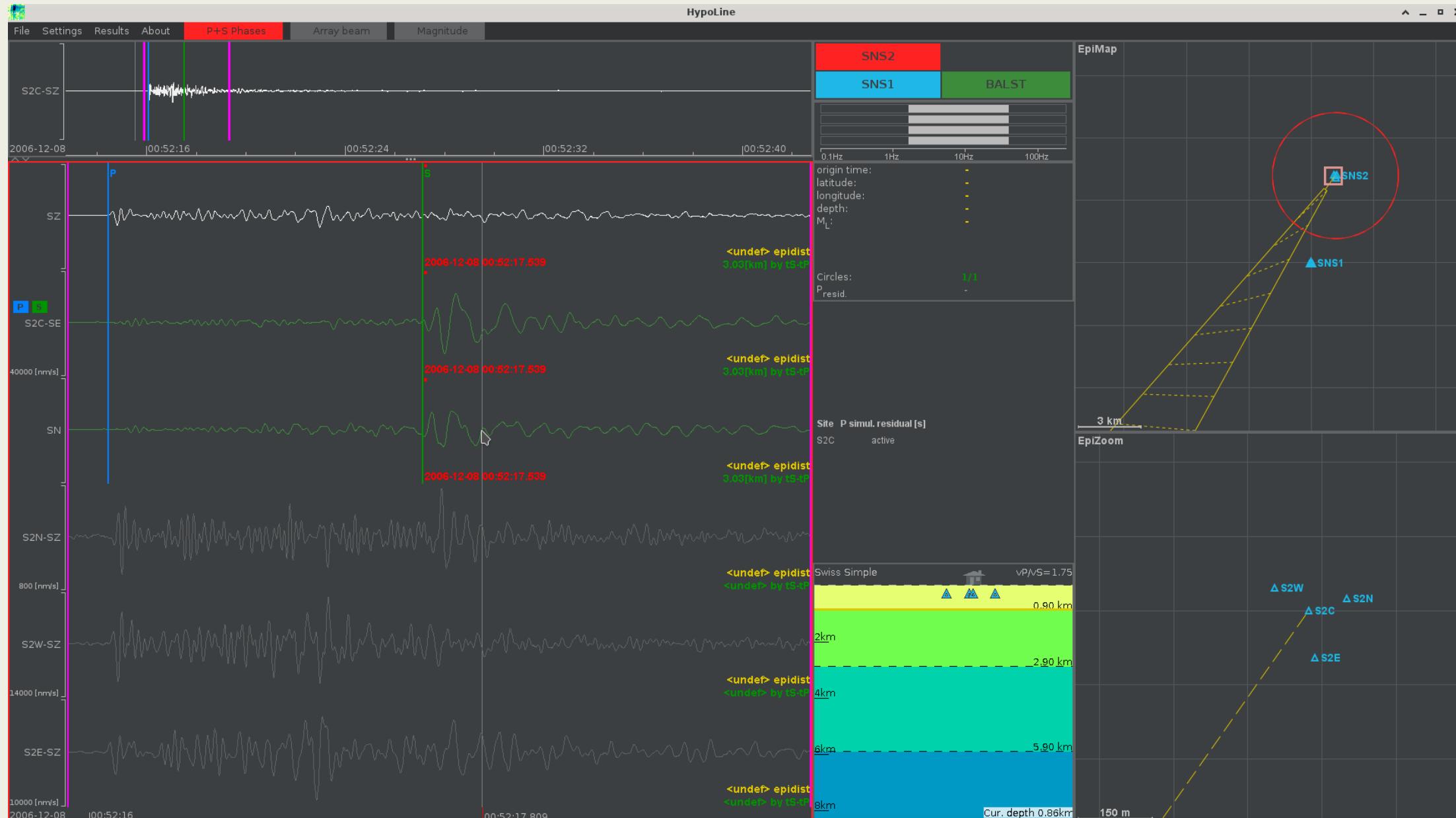
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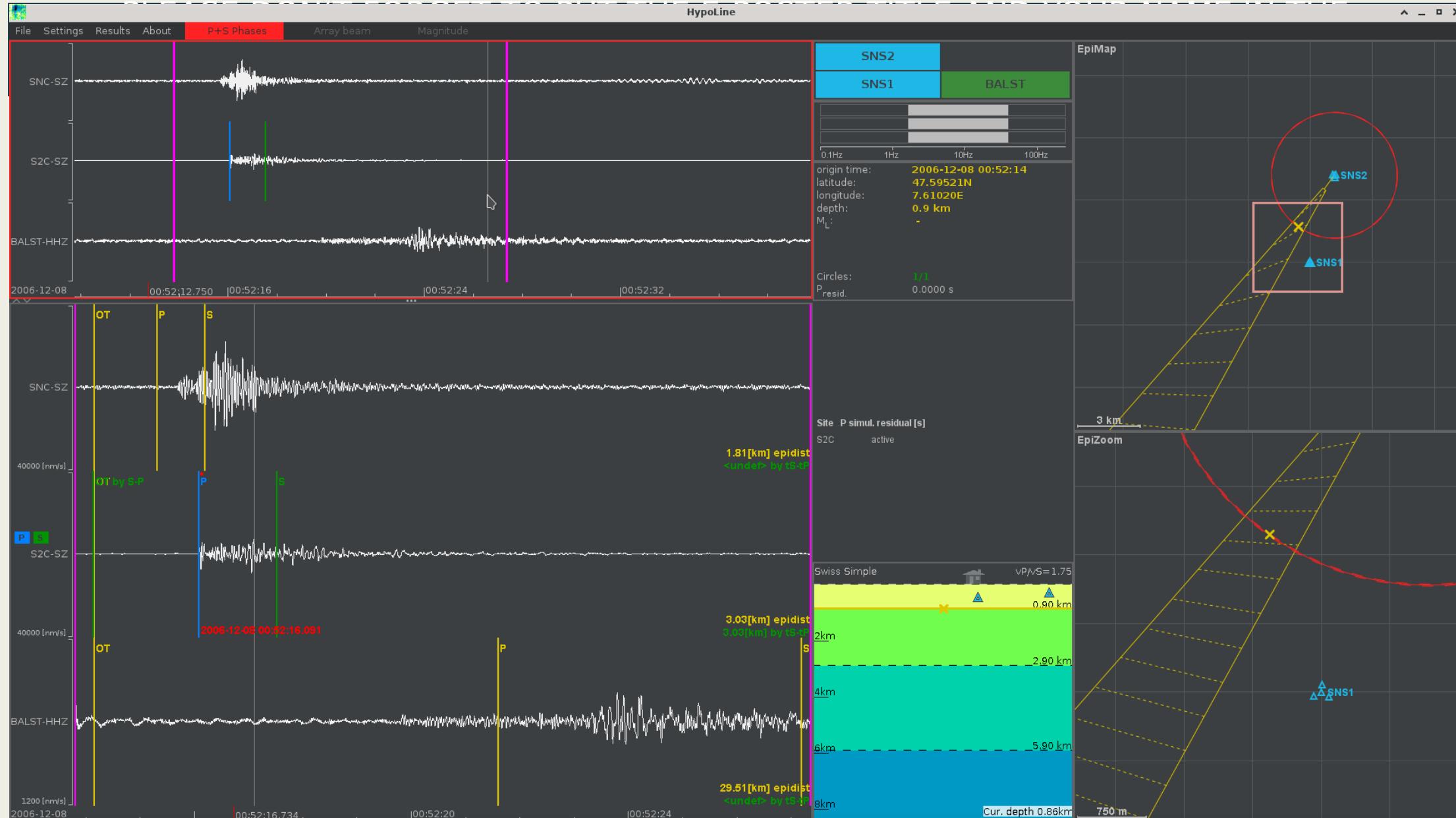
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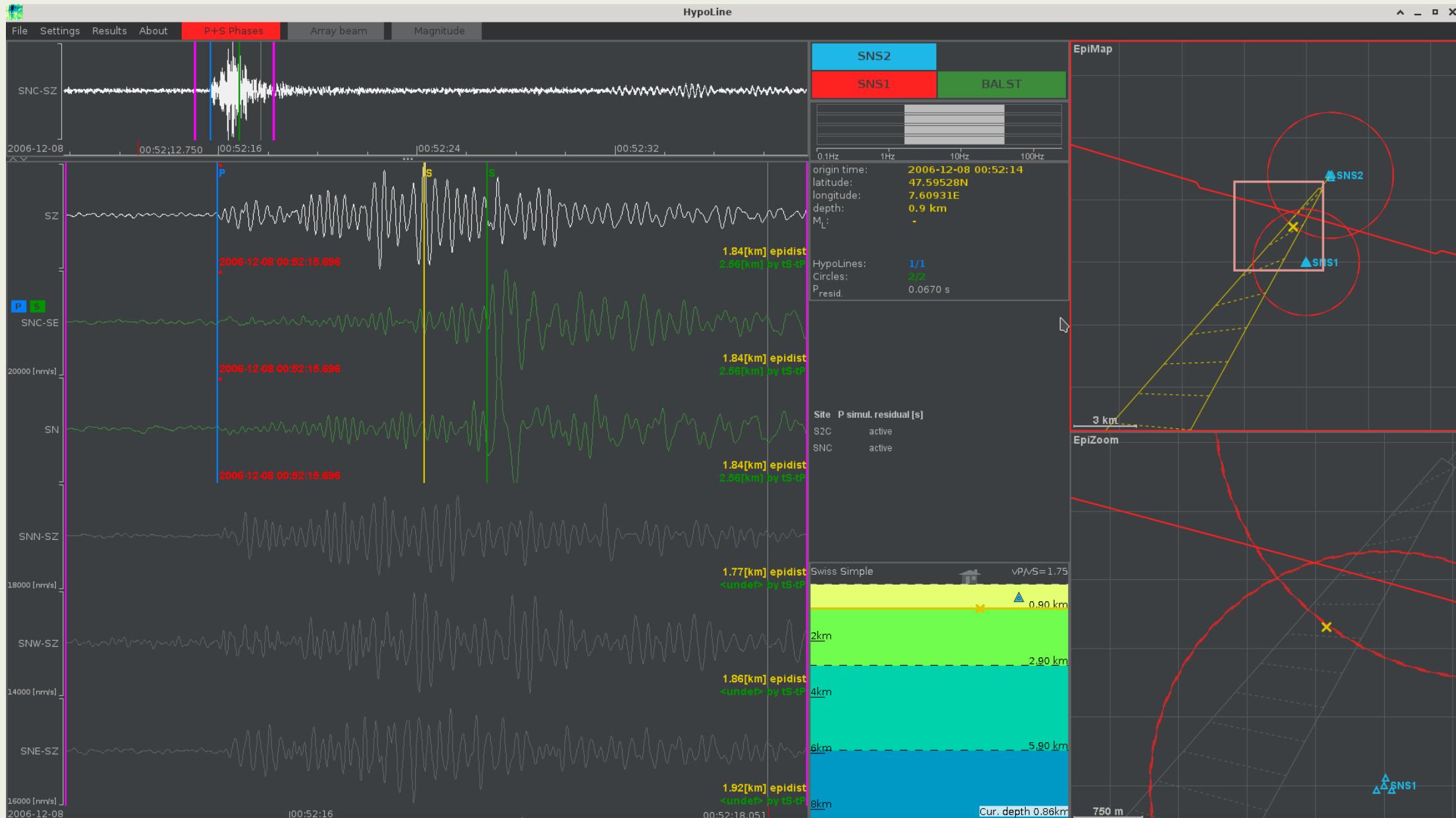
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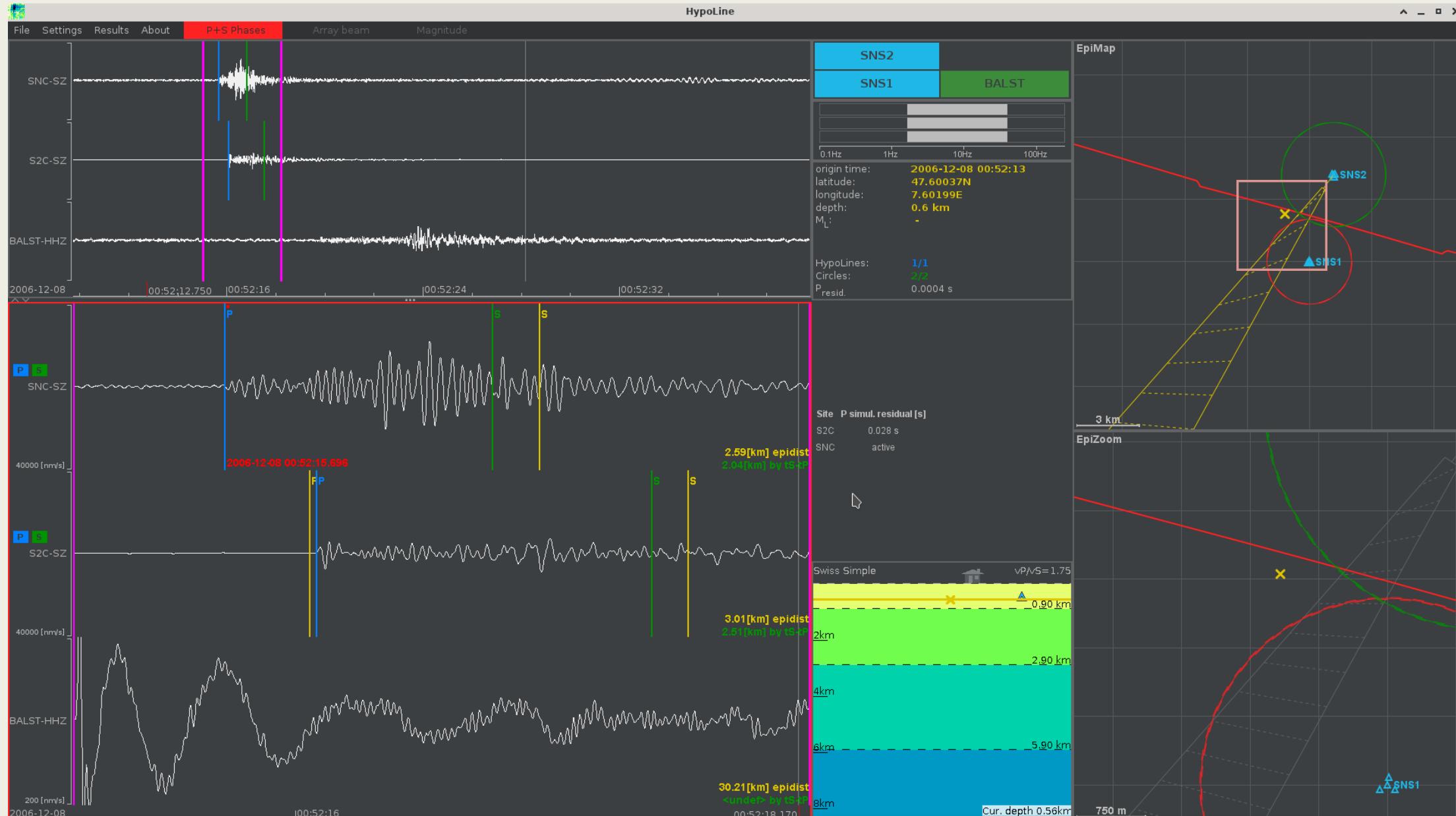
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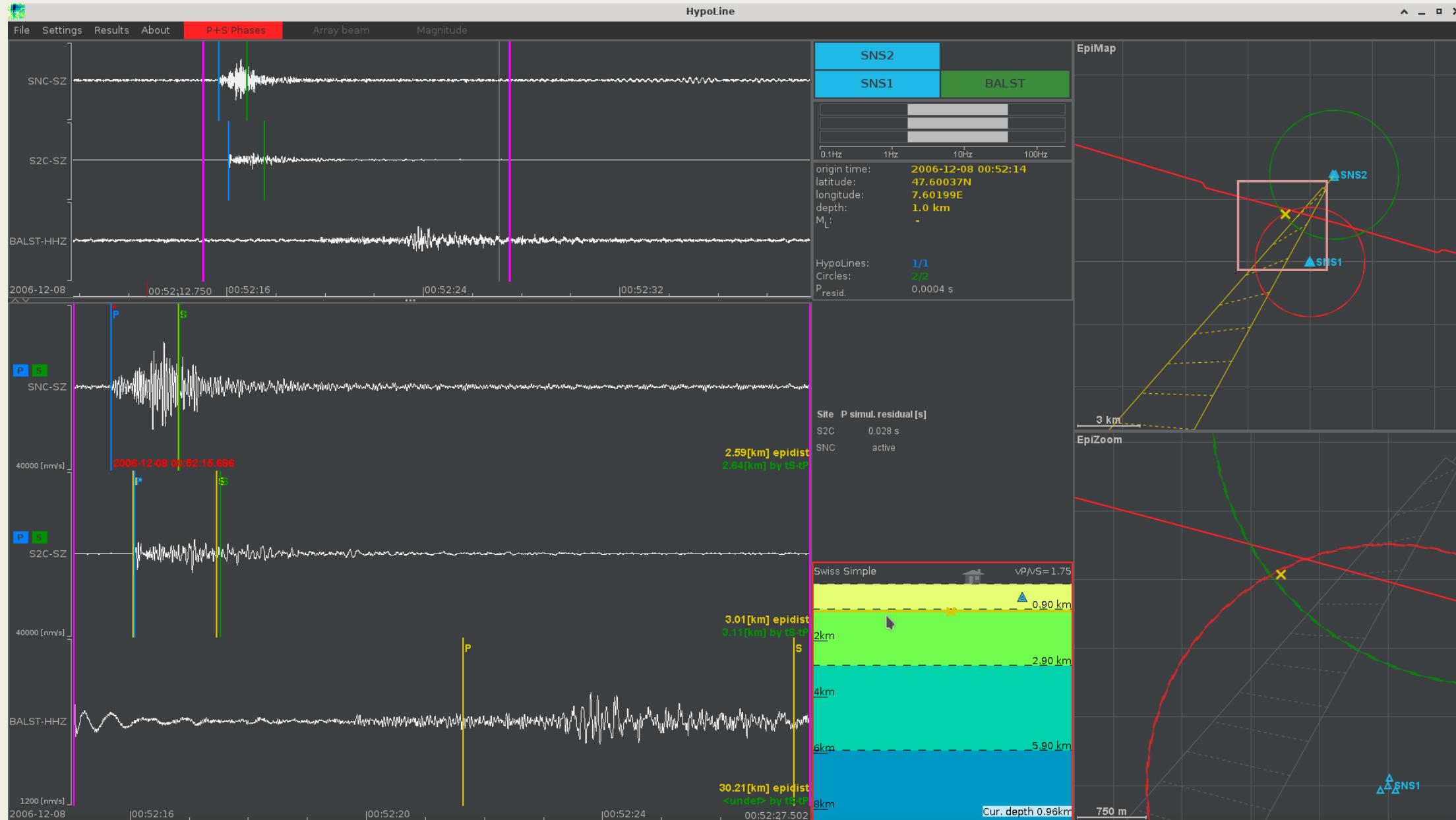
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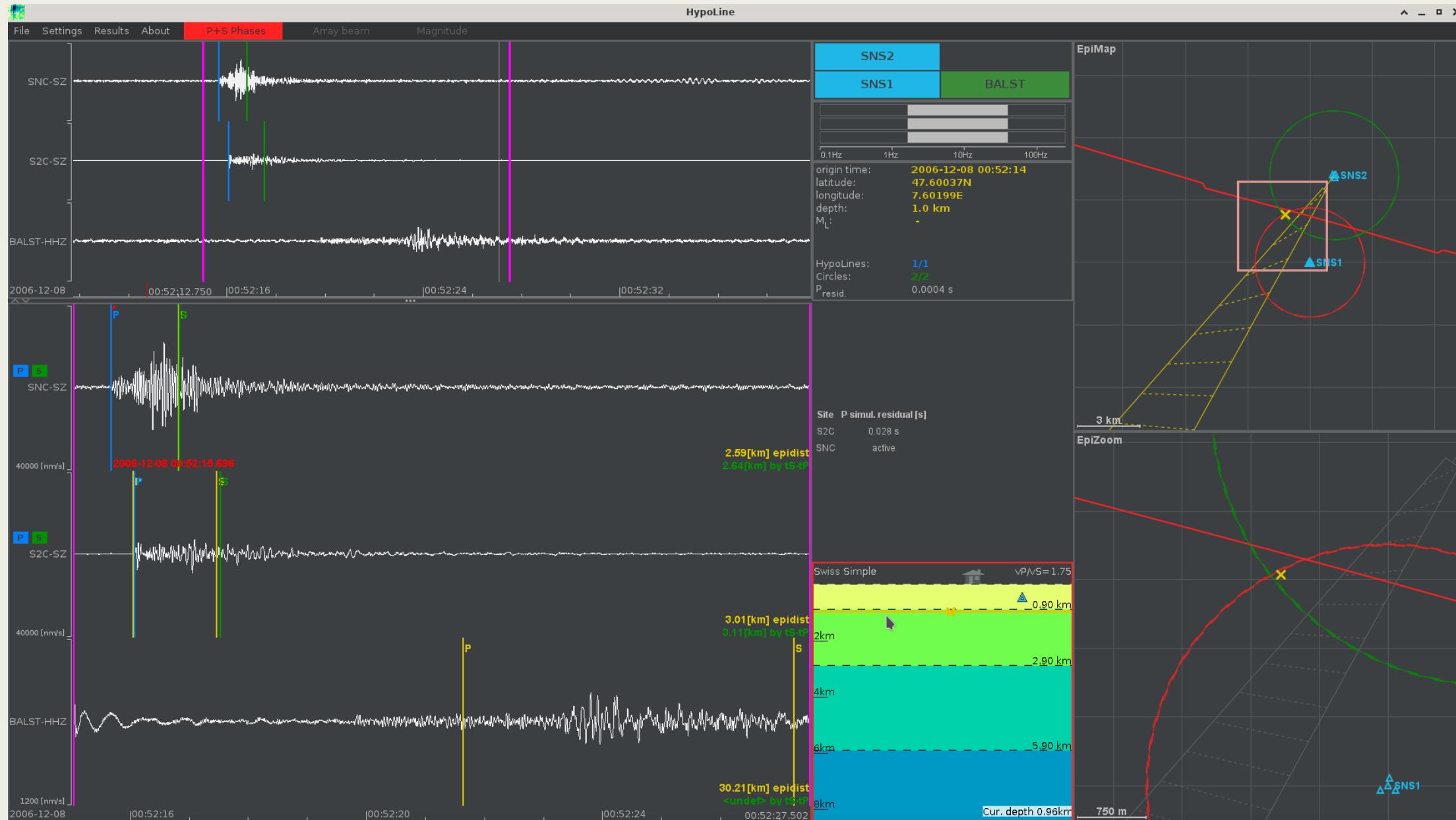
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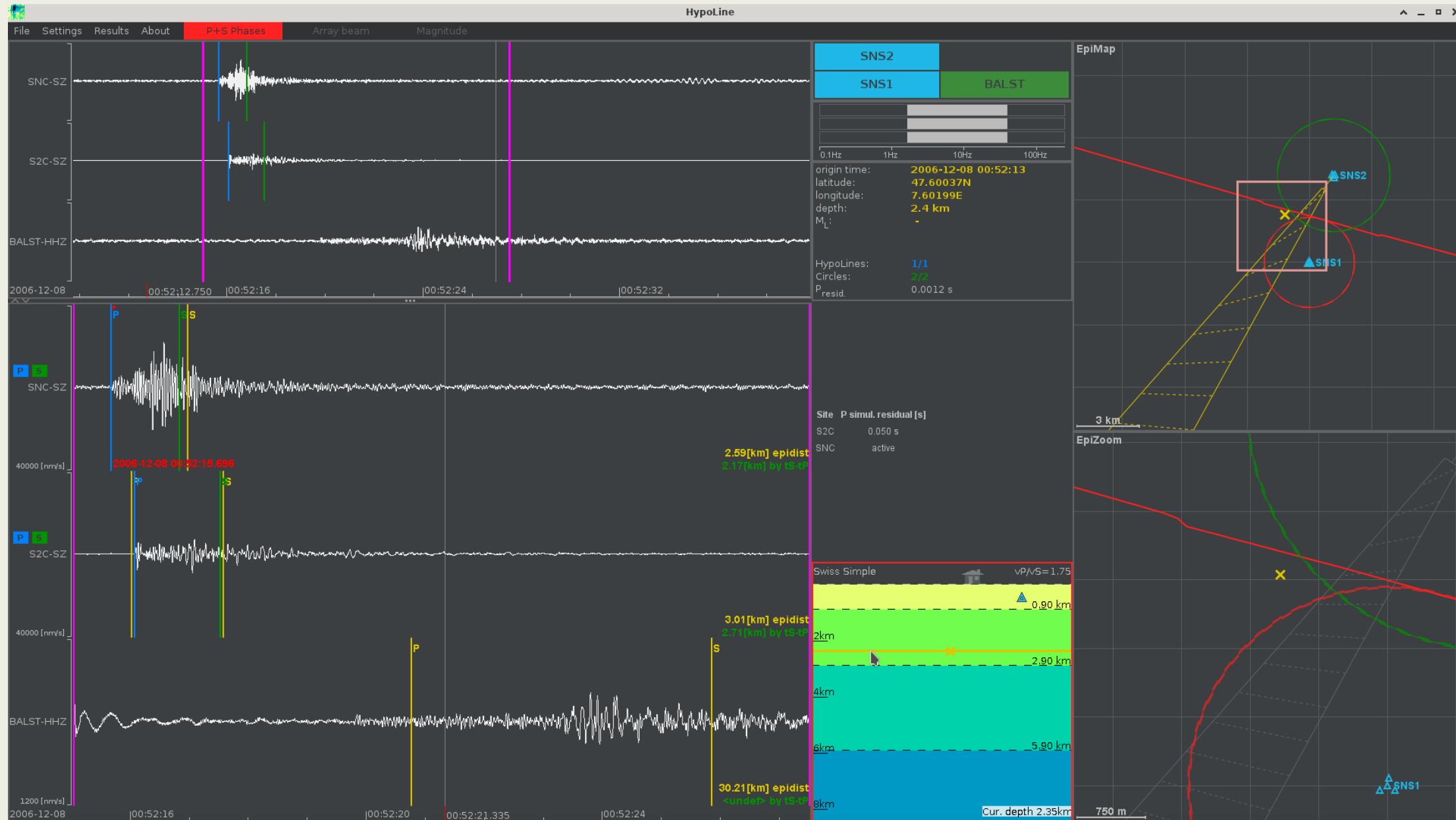
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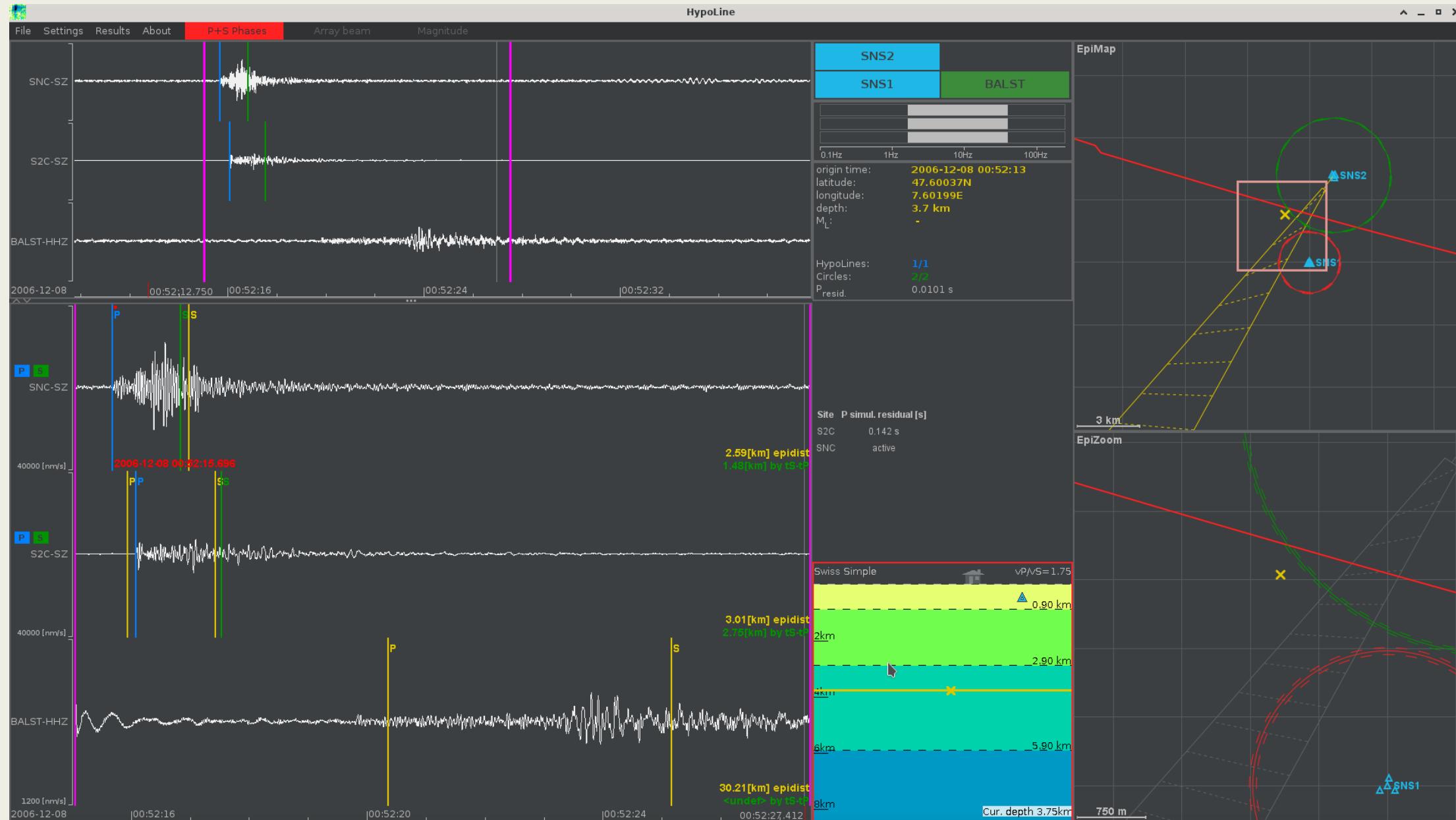
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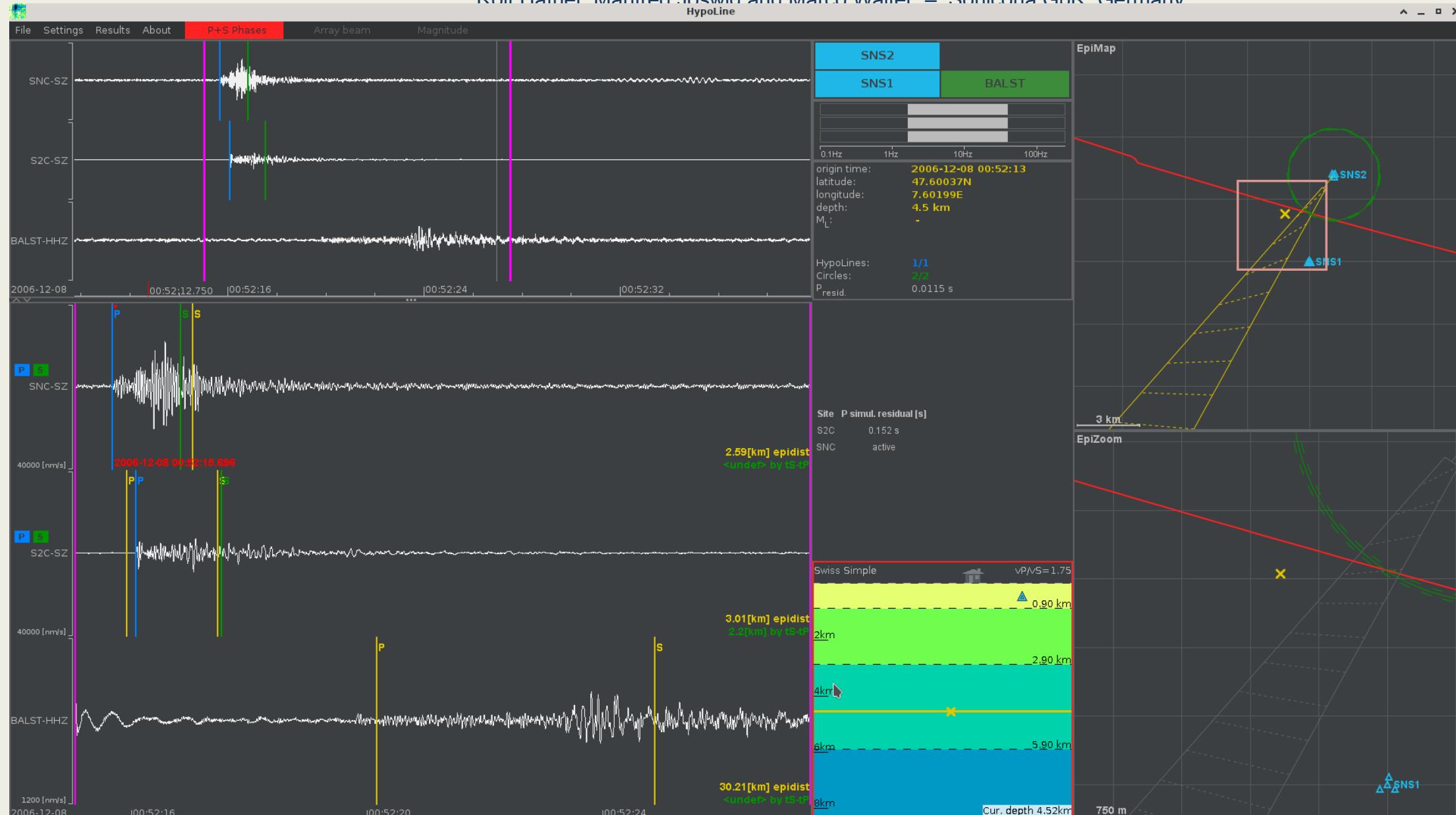
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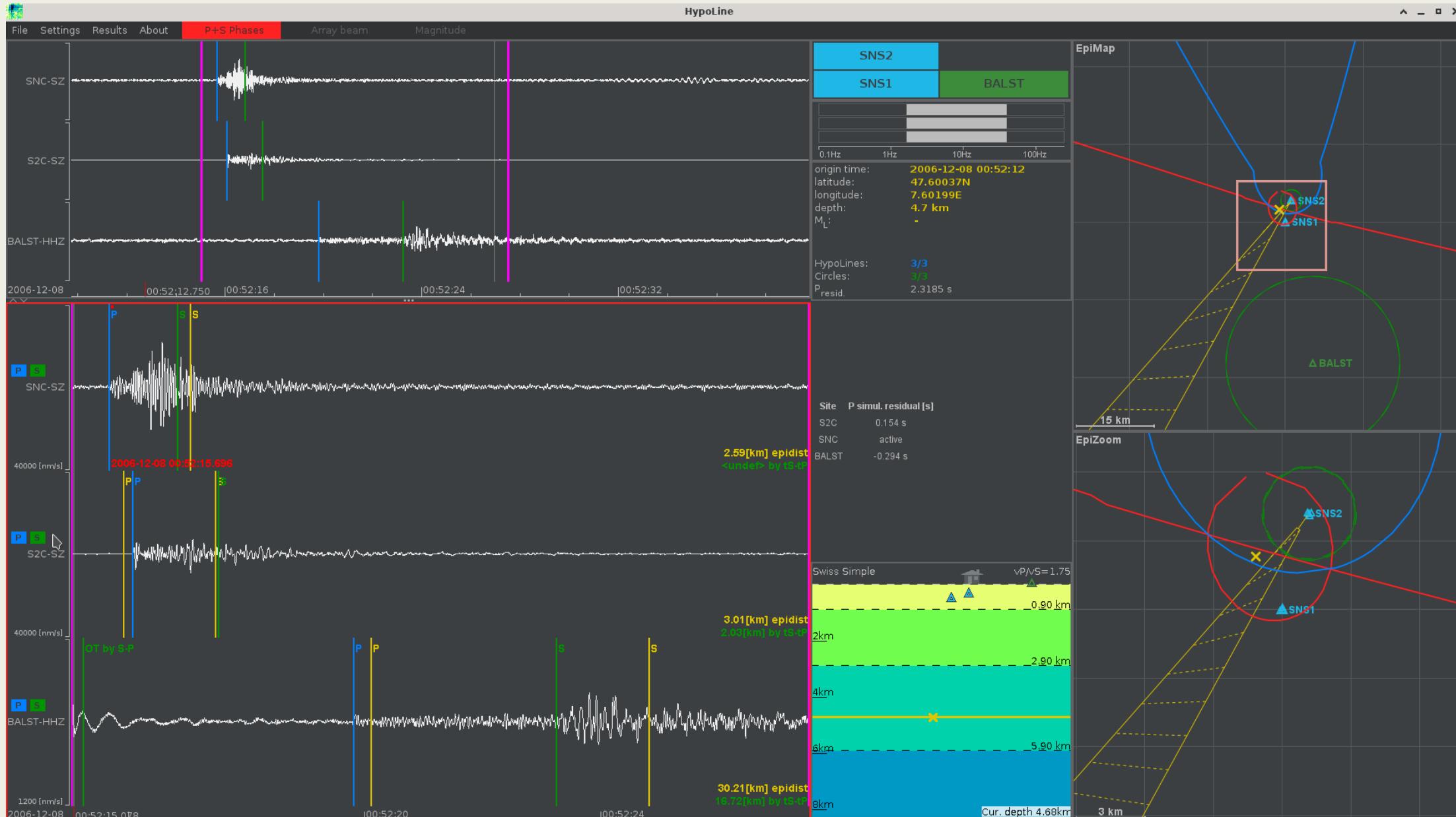
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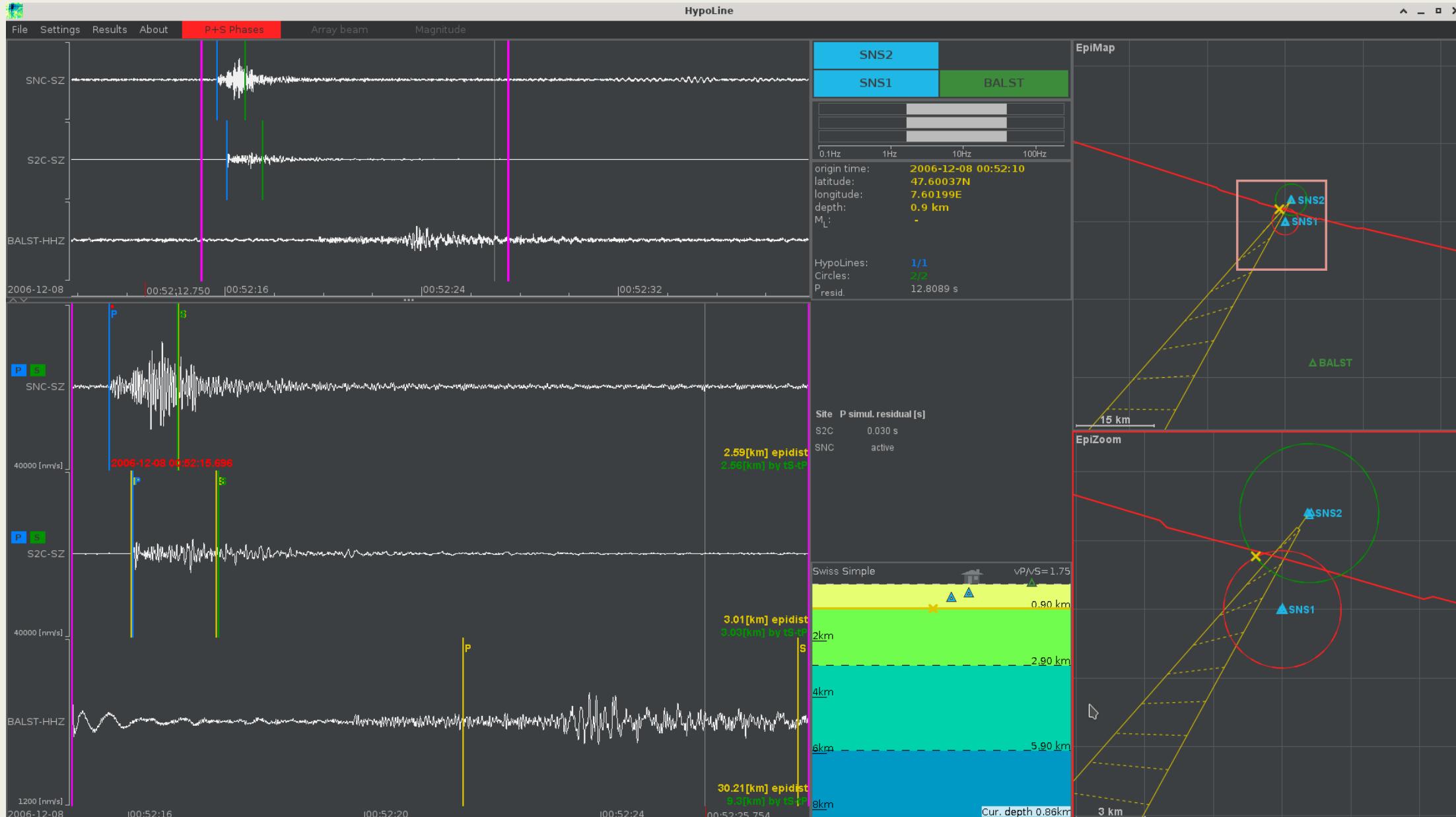
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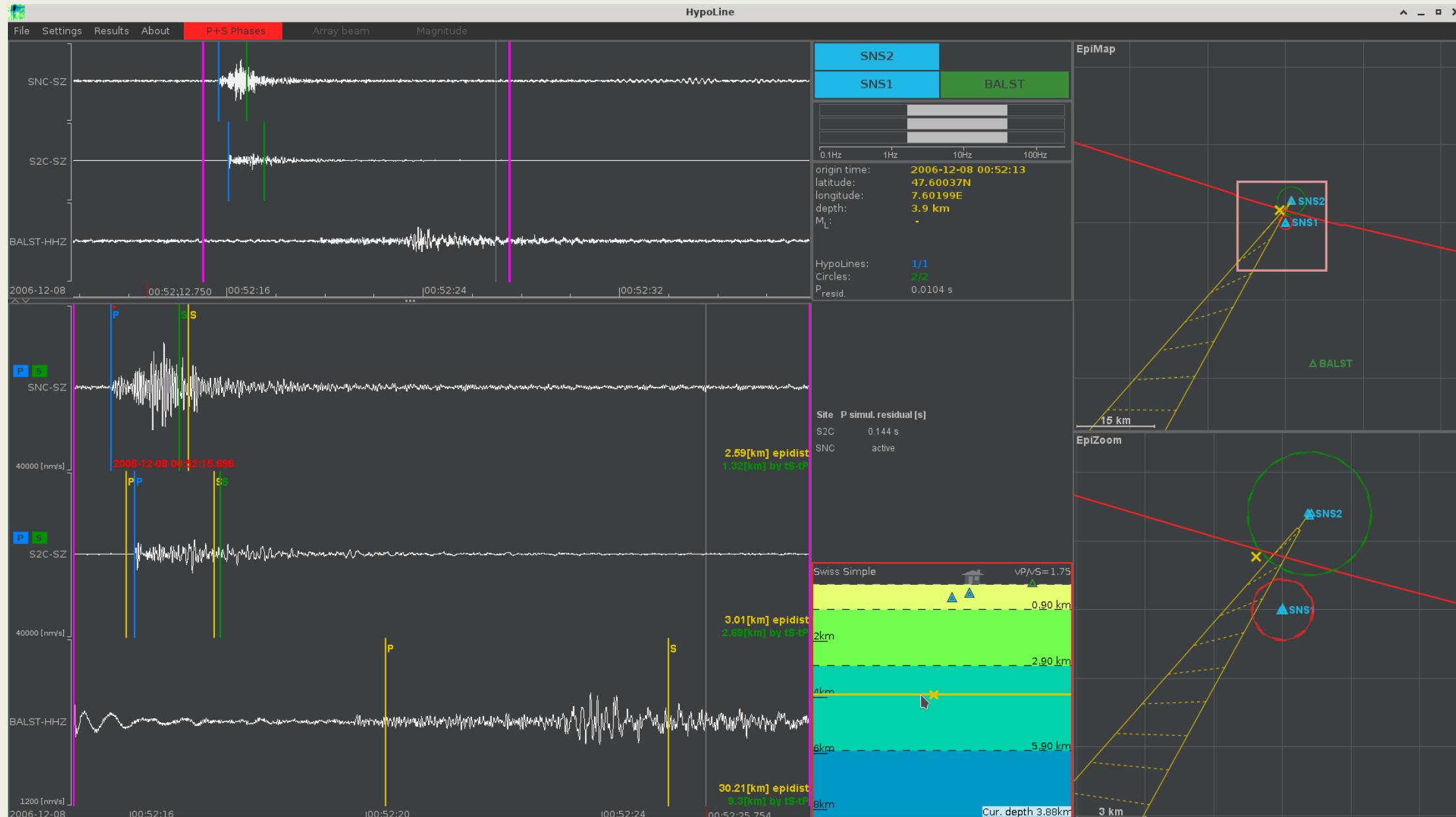
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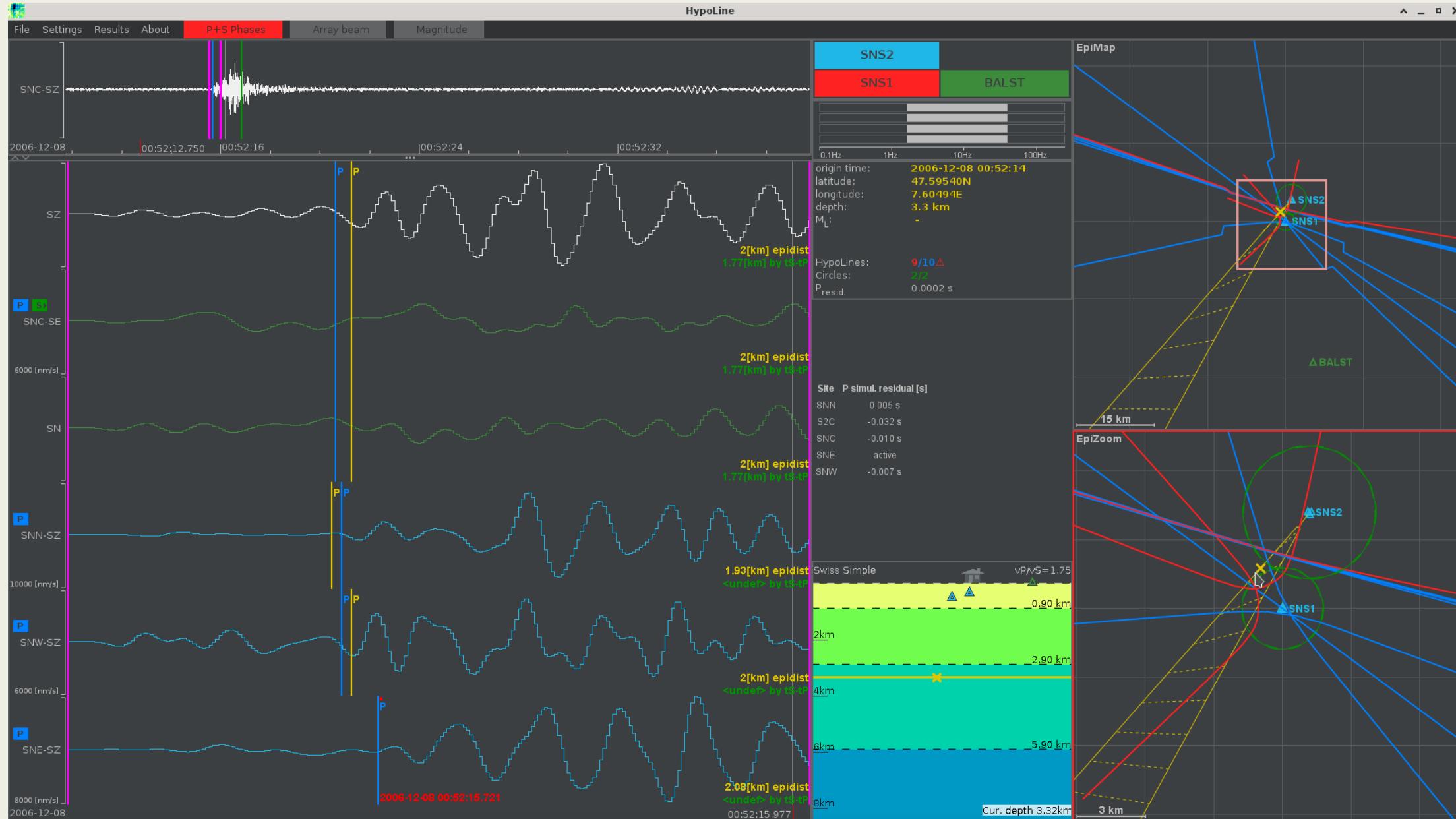
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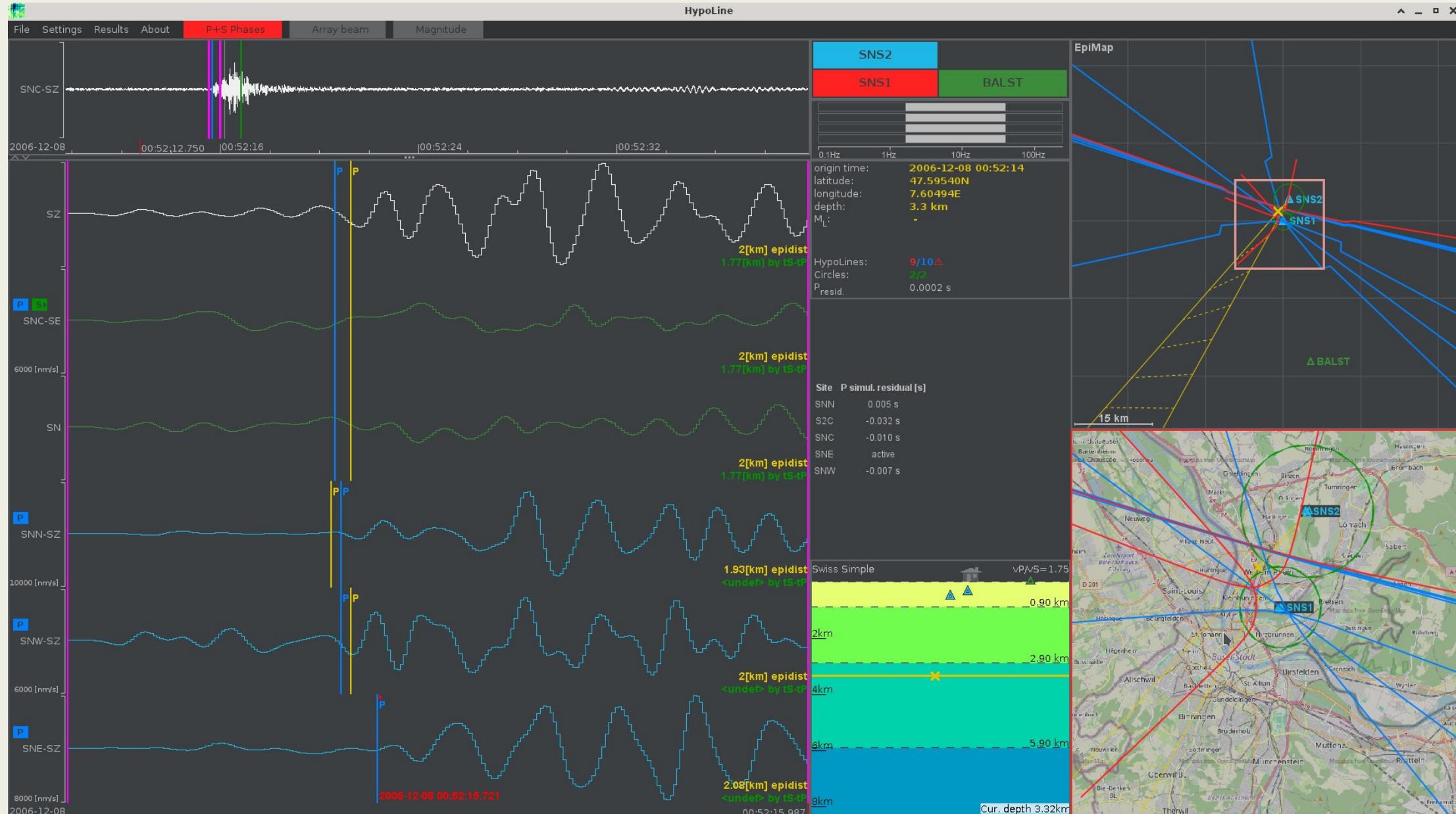
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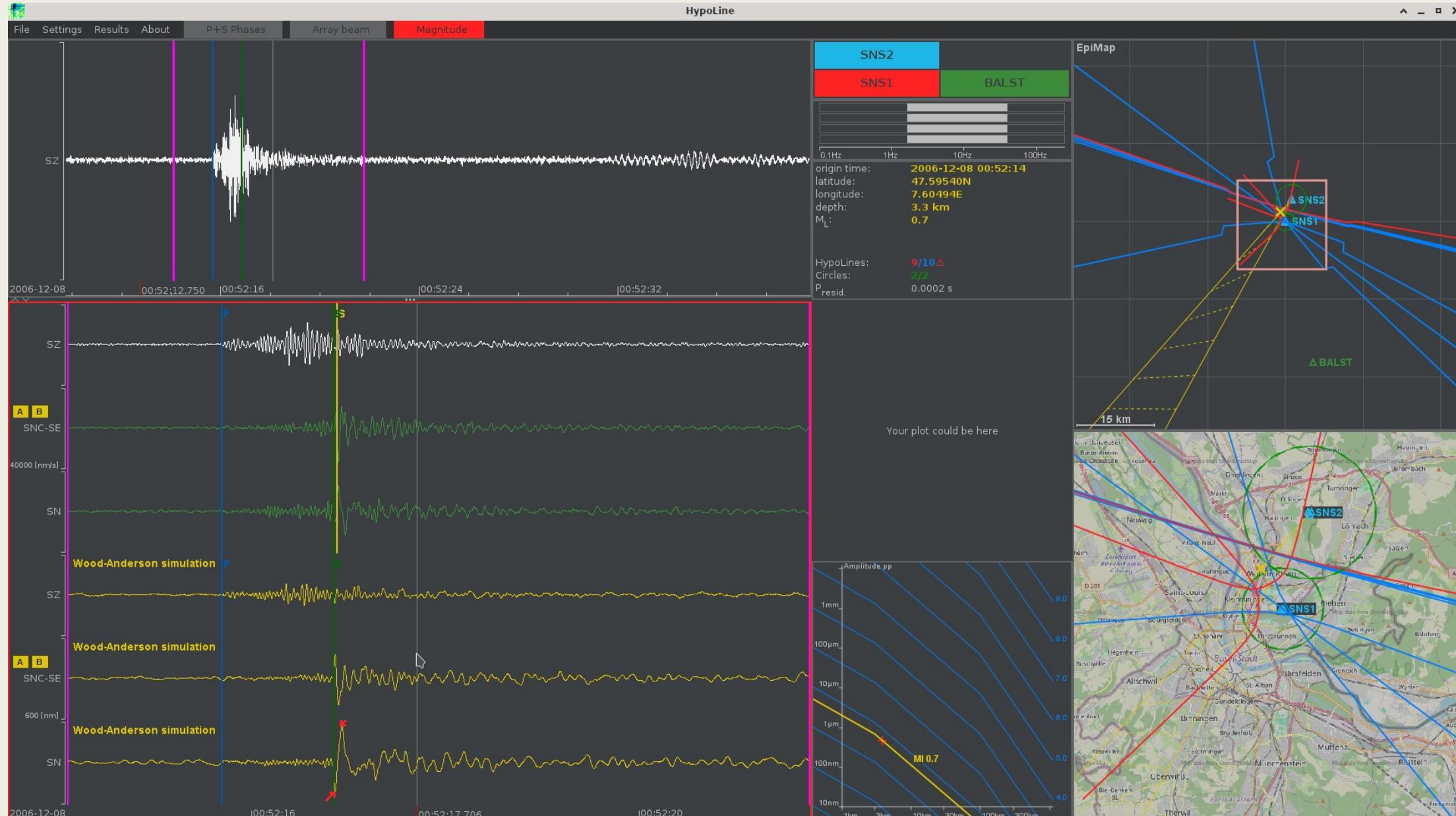
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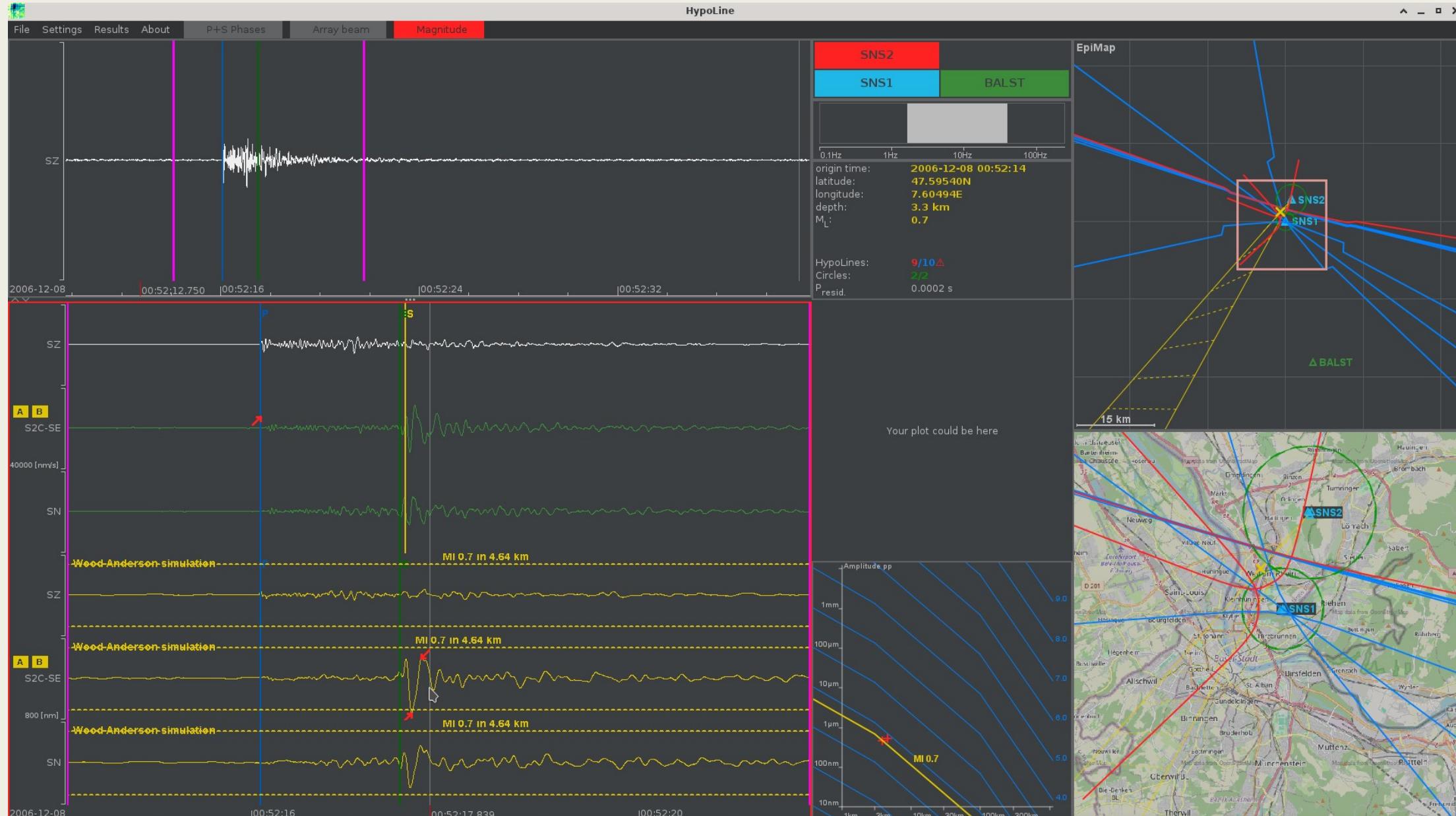
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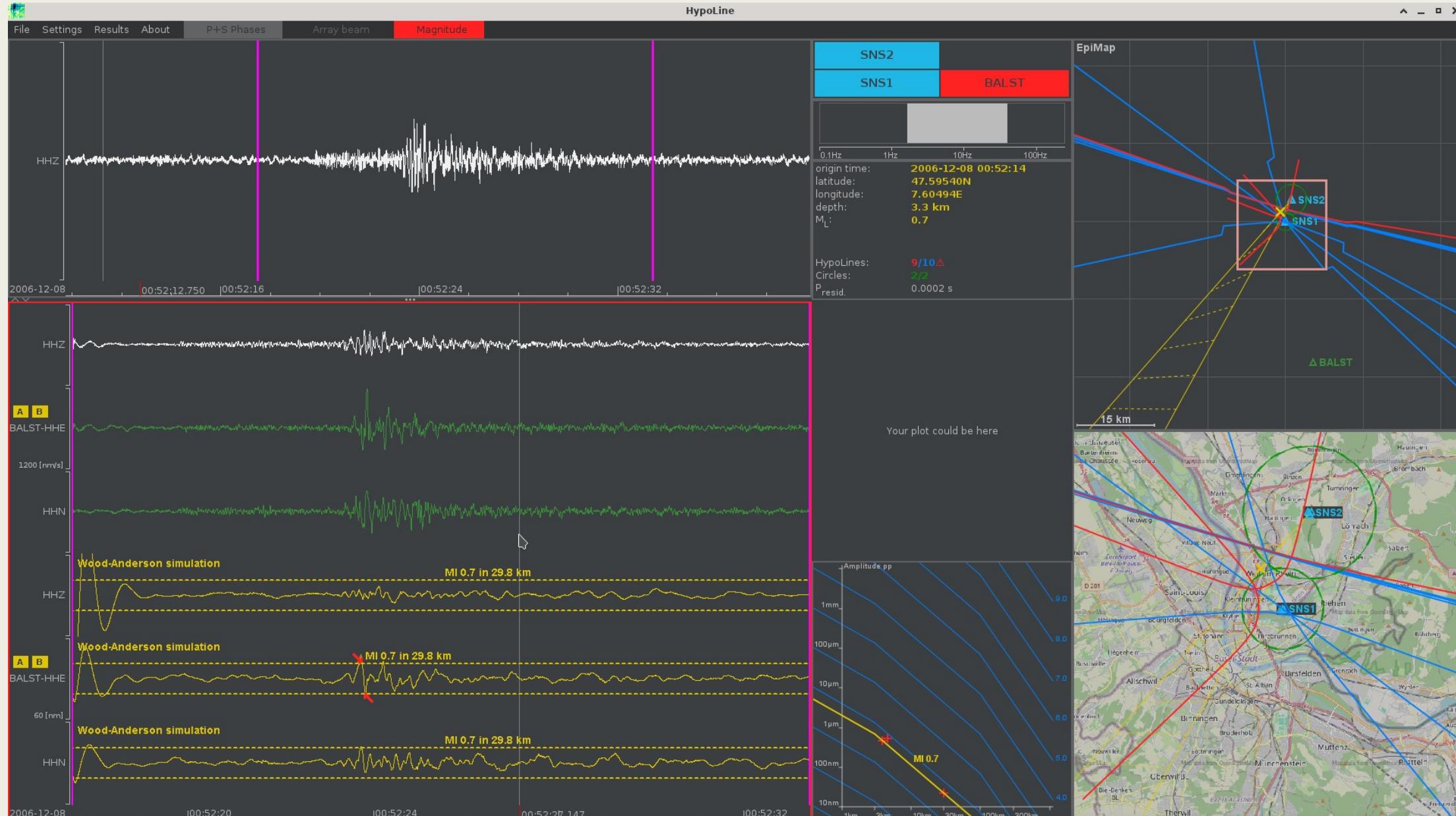
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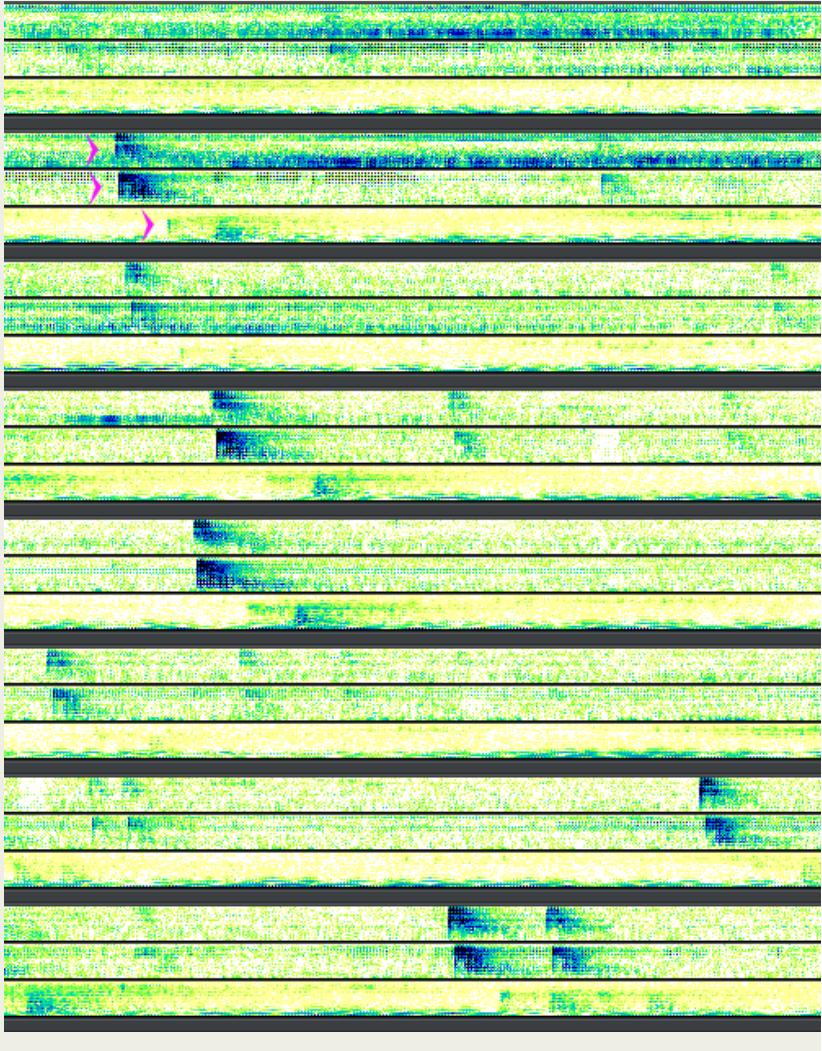


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How can we automate event detection with no training data and limited time for detector calibration?



DIMENSIONALITY CONVERSION

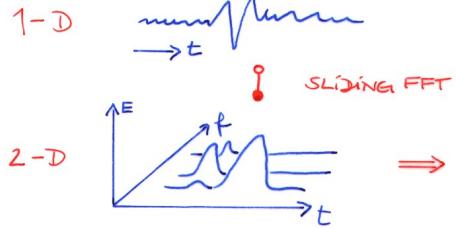


IMAGE ENHANCEMENT

STATISTICS OF NOISE

$$M(f) = M_{50}(f) = \text{median}(E(f,t))$$

$$S(f) = M_{75}(f) - M_{50}(f)$$

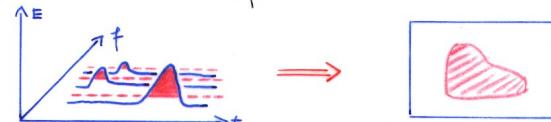
DECISION

$$E(f,t) - (M(f) + S(f)) \begin{cases} \geq 0 : E(f,t) \\ < 0 : 0 \end{cases}$$

ENERGY DETECTOR PER FREQ BAND

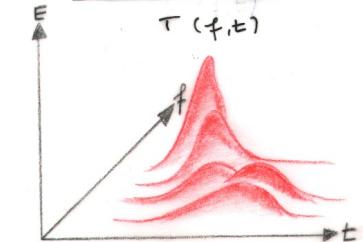
OFFSET REMOVAL AND PREWHITENING

$$D(f,t) = \frac{E(f,t) - M(f)}{S(f)}$$

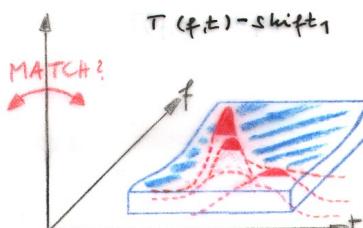
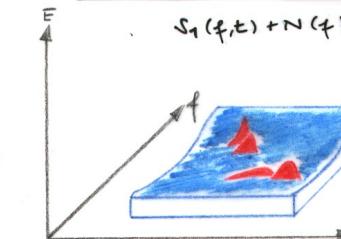


SONOGRAM CALCULATION

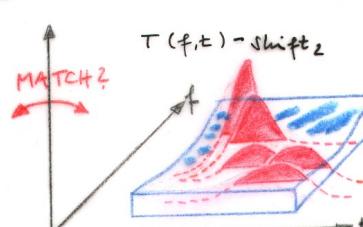
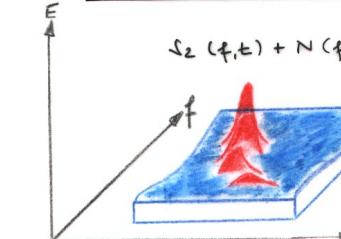
ORIGINAL TEMPLATE



WEAK SIGNAL IN NOISE



STRONG SIGNAL IN NOISE



ADAPTATION OF TEMPLATES



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Conclusion

When using the PSM Toolkit, each user input is directly visualized.

Each new constraint or alteration of a constraint is visualized.

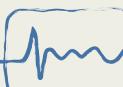
Each simulated onset, amplitude, etc is backprojected on the data after defining a source.

There are no black boxes!

You want to try?

Download the free demo of the Nanoseismic Suite, including three example datasets

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Conclusion

