Log-validated waveform inversion with wavelet phase and amplitude updating

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Summary

The unknown wavelet represents a challenge that may prevent the successful application of FWI on real seismic data. We propose a methodology to correct the amplitude and phase of the modelled data, and with this information, update the wavelet. In this work we applied a variant of FWI introduced by Margrave et al. (2010) that includes the use of reflections, one way operators and well-logs. We used phase-shift plus interpolation (PSPI) migration with well calibration instead of reverse time migration (RTM) and line search to produce the velocity perturbation. The use of PSPI reduces the computational time, and we take advantage of this fact to implement our methodology. We tested this methodology on synthetic and Hussar datasets obtaining encouraging results.

Workflow

Figure 1 shows the workflow to mitigate the negative effects of an incorrect wavelet in the inversion. The process starts with an estimated wavelet that has similar frequency content than the seismic data. This wavelet does not have the optimal amplitude and phase for reproducing the observed shots. In order to address this problem, we migrate and stack the observed and modelled shots separately. Then we convert both datasets from depth to time by using the current velocity model. The comparison of these reflectivity datasets in time domain provides the elements for the estimation of an amplitude and phase that make the modelled data more similar to the observed data. Next, we take the difference between the observed and the amplitude-and-phase corrected modelled reflectivity to create the gradient. After that, we calibrate the gradient with well information to produce the velocity perturbation. The amplitude and phase corrections are applied to update the wavelet that will be used in the next iteration.

Fig. 1. Log-validated FWI workflow with amplitude-and-phase updating.
Synthetic example

The shots, considered the observed data, were generated by propagating the wavelet in blue in Figure 2 through the velocity model (Figure 3A) by applying constant-density acoustic finite-difference modelling. The initial model for the inversion (Figure 3B) was constructed by applying a Gaussian smoother to the true model. An example of an observed shot located at the middle of the model is shown in Figure 3C. The initial frequency range was from 1 to 6 Hz, and then it was moved up by 1 Hz in each iteration. The calibration well C is located at the middle of the model; its top and base are 400 and 900 m, respectively.

We applied the methodology to update the amplitude and phase of the modelled data using the initial wavelet in red in Figure 2. This wavelet was estimated from the seismic. The inversion result is shown in Figure 4. We are able to recover the true model diminishing most of the negative effects that the wrong wavelet produces. The errors tend to decrease after iteration 10 when the wavelet has been already corrected as shown in Figure 5.
FIG. 5. Evolution of the wavelet for selected iterations. The wavelet has been practically recovered after iteration 10.

FIG. 6. Hussar’s shots used for the inversion. Evolution of the wavelet for selected iterations.

Application on Hussar dataset

Figure 6 shows the shots used for the inversion (Isaac and Margrave, 2011). Radial filtering, Gabor deconvolution, low-pass and FK filtering were applied. Figure 7 shows the inversion result for Hussar dataset. At the zone, where we have significant seismic events from 0.8 seconds (approximately 1000 m of depth), the inverted velocity shows a reasonable agreement with the velocity measure in the wells. However, in the shallow zone where the lack of seismic information is evident, the velocity is not effectively recovered.


In the synthetic example we saw that the amplitude-and-phase wavelet updates don’t have significant changes after the 10 iterations, showing stability in the process. This is not happening for the case of Hussar dataset as it’s shown in 8, where the evolution of the wavelet with iterations is displayed. The latter suggests that the process of updating the modelled reflectivity...
before constructing the gradient has the greatest weight in the process. This methodology seems to be robust enough to be applied on real data.

**Fig. 8. Evolution of the wavelet for Hussar’s inversion.**

**Conclusions**

We proposed a methodology to diminish the negative impact that a wrong initial wavelet produces in the FWI process. Our methodology consists in separating the migration of the observed and modelled data previous to the construction of the gradient. Experiments with synthetic data show that the scheme is stable. We applied the process to the Hussar dataset, obtaining encouraging results. However, the high variability of the wavelet for the case of real data brings certain questions about the validity of the updated wavelet and suggests that the inversion relies on the matching process of the observed and modelled reflectivity datasets. In future work, we will address the issue of the instability of the updated wavelet that we saw in the real seismic data.

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**References**


