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Abstract

We present a new methodology, and preliminary results attained by incorporating bandwidth expansion techniques in acoustic and elastic impedance inv ersions. We discuss the required adaptations in the sparse-spike inv ersion workflow, and present the adv antages of this approach when compared with conv entional inv ersion results.

Introduction

Many internal processes were dev eloped along the last two decades in Petrobras with the goal to improve seismic resolutions. Applications of these processes were limited to qualitativ e analy sis in which f aciologic, lithologic and fluid characterizations are correlated to relative physical quantities. These applications include internal publications regarding the characterization of carbonate reserv oirs of the pre-salt in the Santos and Campos Basins and 4-D analy sis of turbidities in the Campos Basin, f acies mapping of the pre-salt reserv oirs in Campos basin (Cunha e outros, 2013). On the other hand, adv ances on seismic acquisition and processing improv ed the reliability of the seismic amplitude, thereby the quantitative interpretation of turbidite and pre-salt reserv oirs based on impedance v olumes became more trustworthy and integrated with static and dy namic data (Teixeira et al., 2017). Howev er, impedance v olumes are restricted to the bandwidth of the input data and, theref ore, hav e not benef ited f rom the bandwidth expansion processes.

Seismic quantif ication of rock properties relies on the combination of rock phy sics and seismic inv ersion (Av seth et al., 2005; Vernik, 2016). Rock phy sics links the petrophy sical and elastic properties. It ev aluates how saturating fluid, porosity, clay content, pore shape, mineralogy, pressure affect the seismic signal. Sparsespike seismic inv ersion enables the spatial extrapolation of well-deriv ed elastic properties by inv erting the migrated traces into elastic properties (Latimer, 2011). The incorporation of high-f requency bandwidth in seismic inv ersion signif icantly improv es the reserv oir characterization, with quantitativ e characterization of the depositional sy stem geometry , and rock properties with resolution more compatible with the scales inv olv ed in f low simulators.

Iterative Deconvolution and High-Resolution Pseudo-Impedance

The process of residual deconv olution of the seismic wav elet from migrated v olumes provides reflectivity v olumes with an expanded f requency bandwidth. It has a long history of dev elopment in Petrobras. Internally known as iterativ e deconv olution, it was originally dev eloped during the late 1970's by José Tassini and Emilson Ev angelista, and subsequently improv ed by André Romanelli (Rosa, 2018). This process was subject to sev eral refinements along the last two decades, with dev elopments that include the stage of data preconditioning (Machado and Cunha, 2015), deconv olution process itself , and post-conditioning workf lows.

The pre-conditioning stage comprises attenuation of coherent and incoherent noises as well as residual phase correction of the wav elet. The phase correction is based on seismic/sy nthetic fitting at well locations as well as analy sis of some ref erence ev ents such as the water bottom and

Figure 1: Estimating the residual phase correction. Top: HRR data without phase rotation and their average phase spectrum. Bottom: HRR data after 20^o phase rotation and their average phase spectrum. The wavelet asymmetry observed in the original data is corrected after the phase rotation.

igneous intrusions.

Figure 1 illustrates the residual phase correction for the case of using the water bottom ev ent as ref erence. By adjusting a linear trend to the relev ant part of the phase spectrum, it is possible to determine the phase rotation that prov ides a null linear coef f icient. In the particular case of this project, a phase rotation of 20 degrees was necessary. Subsequent correlation with sy nthetic traces at well locations v alidated this correction.

Af ter the residual wav elet deconv olution, it is necessary to simulate the spectral tendency of the subsurface reflectivity series. This is achiev ed by a spectral coloring shaping procedure (Rosa and Ulry ch, 1991).

The spectrally shaped deconv olv ed data is then subject to time integration, which simulates a conversion from ref lectiv ity to relativ e impedance or the so-called pseudoimpedance. The next stage inv olv es a process called High-Resolution Pseudo-Impedance (PSIM), which consists of a combination of temporal and spatial structural f ilters that aim at the attenuation of the low-f requency noise amplif ied by the integration, while preserv ing the signal which is normally discharged in conv entional f iltering processes. Finally, the PSIM data is differentiated with respect to time, resulting in a High-Resolution Ref lectiv ity (HRR) v olume.

Figure 2 compares the (conv entional) input data and the HRR data for 2-D window around Well-1. We verify the

expressive increase in resolution, with continuous ref lectors f ree f rom the usual artif acts amplif ied by typical processes of spectral enhancement, while preserv ing good correlation with the sy nthetic traces.

Inversion of the high resolution data

From 47 wells used in this project, 25 show a seismic-log correlation higher than 60% f or the high-resolution data, considering a time window comprising the entire acoustic impedance log.

The time-depth calibration for broadband data is still more critical than f or conv entional data, since a small mismatch produces a signif icant decrease in the sy nthetic/seismic cross correlation due to the bandwidth expansion.

The usual tools prov ided by commercial inv ersion sof tware for the seismic wav elet estimation, which are based on least-squares minimization between seismic and sy nthetics, were not dev eloped f or broadband data. This method is efficient to estimate wav elets for conventional data, but inadequate for high-resolution data, such as HRR.

The size of the time window required for the wavelet estimation has also to be readdressed. For a satisf actory estimation of the amplitude spectrum, this size needs to be, at least, three times the size of the wav elet, especially to capture the amplitudes associated with the low f requencies. Howev er, not only in this particular project, but in most standard projects, the sonic and density logs have a limited depth range, with an extent that may be inadequate ev en f or conv entional data. To cope with these issues, we opted for an alternative strategy for wavelet estimation, with a combination of automatic tools and manual adjustments.

The first step of the wav elet estimation is based purely on seismic data (without any well-related inf ormation). A large number of traces are used, with a larger time window than usually inv olv ed in conv entional inv ersion, allowing f or a

Figure 2: Top: Conventional data. Bottom: HRR data. Ampli fi ed windows show good correlation of both with respective well derived synthetics (each with their own wavelet).

more robust estimation. **Figure 3** compares the av erage seismic spectrum, the seismic-derived statistical wavelet spectrum, and a manually edited v ersion of the wav elet spectrum to produce a smoother v ersion. The next step comprises the least-squares adjustment of phase and amplitude, using the edited statistical wav elet as a priori inf ormation with a strong weight. With this approach, the least-squares adjustment is used only to guarantee that the seismic amplitudes honor the amplitude from the sy nthetics deriv ed from the well logs. The phase correction was neglectable and thus set to zero.

Figure 4 shows the final wav elets estimated for the conv entional and the high resolution data, as well as their corresponding amplitude spectra.

Figure 3: Comparing seismic spectrum (green) with original statistical wavelet (blue) and its edited version (red).

Figure 4*:* Superposition of final estimated wavelets *conventional data (red) and HRR data (blue). The HRR wavelet is broader, allowing for more low frequency content and with much smaller side lobes.*

Regarding the inv ersion parametrization, since the highresolution data is, in many aspects, different from conv entional data, some considerations were necessary to serv e as a guide f or this project:

- Low-noise input data;
- More signif icant ref lectors when compared to conv entional data;
- Reliable data down to 4 Hz.

Lateral continuity constraints were used to prov ide smoother inv ersion results. Furthermore, we f ollowed the two-step approach for the low-frequency insertion, which prov ides a more laterally continuous result, remov ing subv ertical artif acts (Sobreira et al., 2014; internal report). We used merge f requencies of 4 Hz in the f irst step and 12 Hz in the second step, with the second-step model obtained f rom the output of the f irst-step inv ersion.

Figure 5: Results from three different approaches to acoustic inversion for a session crossing two wells. Hot colors (yellow to red) correspond to low impedances, while cold colors (light green to dark blue) correspond to high impedances. A trapezoidal band-pass filter (3.5Hz-5.5H z - 137.5Hz-162.5Hz) was applied to the seismic traces as well as to the well logs. We verify that the HRI approach shows better correlation with the well logs when compared to the other two approaches.

Analysis and quality control of results

The results were analy zed in 1-D, by comparing the inv erted traces with the logs at the well positions, and in 2- D, by comparing the seismic section along the wells. We also compare the acoustic impedance deriv ed f rom the sonic and density logs with three v ersions of the inv ersion: conv entional inv ersion (**CONV**), pseudo-impedanc e inv ersion (**PSIM**), deriv ed f rom the integration of the highresolution data (which has no low-f requency components), and the high-resolution inv ersion (**HRI**), which blends the interactiv e-deconv olution large bandwidth seismic and the sparse-spike seismic inv ersion. Finally we show the results of blind tests.

Pseudo-wells and sessions

To illustrate the concept that the high-resolution inv ersion produces superior results in terms of resolution and more reliable absolute v alues of acoustic impedances, we constructed **Figure 5**. It compares the results f rom three inv ersions: Conv entional, PSIM, and HRI. For a session that crosses two wells (Well-1 and Well-2), the examination of this f igure v erif ies that:

- The transitions between lay ers are smoother in the conv entional inv ersion. The more abrupt contrast shown in the other two inv ersions is more geologically consistent with the discontinuous character of stratigraphic units.
- The HRI is capable to identify thin (10 m) anhy drite lay ers (dark blue in region **1**), and, although it is present also in PSIM, its amplitude (tonality) shows a small relativ e contrast. We observ e a similar behav ior in other high relativ e impedance f eatures (dark blue lay ers) in region **2**, as well, in region **3**.
- The smaller (seismically related) content of low f requencies in the conv entional inv ersion produces incorrect acoustic impedance (greenish colors) in thicker lay ers, such as the Piçarras Formation indicated in region **3**, as attested by the well log.
- The dark blue feature in region 2 would be

interpreted as a single body in the conv entional inv ersion, but as three independent bodies (and more conform to local lay er attitudes) in the HR inv ersion.

• In general, the HR inversion combines the stratigraphic details prov ided by the PSIM inv ersion with a more rigorous amplitude control prov ided by the sparse-spike inv ersion.

In order to complement these observ ations, **Figure 6** compares impedance logs (f ull bandwidth up to 175Hz) of Well-1 and the traces f rom the HRI and CONV inv ersions. A visual inspection shows that (except for the last 50ms) the HRI inv ersion correlates better with the well log than the CONV inv ersion. The superiority holds in quantitative terms: 0.9 f or HRI and 0.84 f or CONV. It is worthwhile to point out that this result is ev en more surprising when we consider that the CONV inv ersion used the well inf ormation to a larger extent (up to 6Hz) than the HRI inv ersion (up to 4Hz) to construct the low-f requency model.

Figure 6: Acoustic Impedance logs: (red) from Well-1; (blue) from *conventional inversion; (black) from high resolution inversion. The HRI result reproduces the sharp impedance contrasts present in the well log for a bandwidth up to 175Hz.*

Blind Tests

In order to v alidate our conclusions, f iv e wells were chosen to be used as blind tests (not used in any part of the inv ersion workflow). We present, for each of the wells in table 1, two v alues of correlation with the seismic inv ersions at the well locations. Values on the left correspond to f ull bandwidth, and on the right (bold f ace) with a low cut of 6Hz.

Table 1: Correlation values (in %) between well log acoustic *impedances and seismic inverted traces for the two approaches: CONV and HRI*. Values on the left correspond to full bandwi dth , and on the right (bold face) with a low cut of 6Hz.

Because the insertion of the low f requencies deriv ed f rom wells inv olv ed dif f erent merge f requencies, presenting the values with the 6 Hz low cut allows for a comparison in a bandwidth less susceptible to the inf luence of the well deriv ed low f requency model.

For the full bandwidth results, correlations for the HRI are equal or superior compared to CONV in 4 of the 5 wells analy zed, while f or limited bandwidth the HRI correlations are always superior. Notice that almost all correlations decrease when the low f requencies are absent. This result is expected ev en in the case of blind tests since this low f requency band contains inf ormation f rom neighbor wells. In the specif ic case of Well-4, both inv ersions present a relativ ely low correlation due to imaging def iciencies in the region where this well is located.

Positioning an injector at a border of the field

The distance between injectors and producers should be as large as possible in order to optimize the injected fluid pathways inside the reservoir, enhancing the oil production, and av oiding premature water breakthrough. The best positions f or injector wells are ty pically located in lower structures of reserv oirs to maintain the pressure and optimize the oil sweep by differential density and gravity. The lake bathy metry is a key control for facies deposition of pre-salt reserv oirs (Faria et al, 2017). As consequence, the positioning of injector wells in lower structures increases the risk of drilling clay -bearing carbonate f acies, penetrating low porosity and permeability rocks.

Theref ore, an important part of the geological characterization of the pre-salt reserv oirs is the prediction of the most probable regions where non-reserv oir f acies are present, and where respectiv e transition zones, f rom clear to clay -bearing carbonates, occur in order to optimize the drainage.

Figure 7 presents some logs f rom a well located at a region characterized by the occurrence of transitional f acies, near the west border of the field. The arrows in the figure indicate a zone of low porosity due to the presence of clay minerals in the carbonate rocks.

The results f rom the conv entional and the high-resolution impedance inv ersions are presented in a session across this well (**Figure 8**). The gain of resolution achiev ed with the HRI approach signif icantly reduces the uncertainties to characterize the geometries of the two f acies.

In addition to a better delineation of the carbonate lens rich in clay minerals the high resolution inversion has been ef f ective in the delineation of lateral f acies transition as well as the interdigitation of associated f acies. The high porosity

and permeability reserv oir zone (dark blue) as well as the thin clay -rich lay er (light blue to green) are much better def ined, and with a more geologically consistent geometry in the HRI than in the CONV inv ersion. The gradual interdigitation between the two f acies becomes ev ident.

Figure 7: On the left several logs showing a layer with clay minerals indicated by the interval between the two arrows. On the right two versions of the well-based acoustic impedance, one in full bandwidth, and the other with a 65Hz high cut filter. We verify that the limitation of the bandwidth decreases the resolution of the clayish layer not only in terms of impedance contrast (amplitude) but also in terms of thickness and positioning.

Four well locations f or the new injector were considered at first, based in integrated studies of flow simulation, risk analy sis for the presence of clay minerals at the reserv oir and return on inv estments. Flow simulations f av ored two locations, while geological risk analy sis recommended the other two locations. It became critical to perf orm a detailed study using the high-resolution data, to reduce the geological uncertainty and to support a saf e location suited f or the drainage optimization.

The well was drilled and the seismic prediction of highquality reserv oir rocks proved to be suitable for the fluid injection requirements.

Final Comments and Conclusions

We presented in this paper the general workflow for producing high resolution ref lectiv ity (HRR) volumes. We also discussed the required adaptations in the conv entional inv ersion workf low in order to cope with the particularities of the HRR data.

To ev aluate the results we compared three inv ersion approaches: conv entional inv ersion, pseudo-impedanc e inv ersion and high-resolution inv ersion. The adv antages of this new methodology are quite representativ e, as demonstrated by the sessions across some wells. The improv ement in resolution is testif ied by the quality control results, and supported by well data. Blind tests also attest the capability of HRI to capture high f requency inf ormation f rom seismic.

The use of the HRI v olume supported the location of new injectors and producers in the f ield. This approach to inv ersion brings signif icant gains not only in the characterization of the reserv oir internal geometries (and associated depositional sy stem), as well as in quantitative estimation of reserv oir properties with resolution more compatible with f low simulation models.

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Figure 8: (a) Conventional inversion (CONV) (b) High resolution inversion (HRI).