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A New 4D Approach to Compensate Illumination Differences of Time-Lapse Acquisitions

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Summary

Successful time-lapse studies require special care when it comes to the removal of undesirable artefacts caused by the differences in acquisition geometries. By attempting to repeat the source and receiver geometries between surveys as precisely as possible, any subsequent 4D noise is minimized. However, in some cases it is not possible to repeat the survey geometries between vintages. This is the case when a towed streamer survey is compared with an OBS acquisition. The image domain approach for correcting illumination differences between 4D datasets builds on wave equation reflectivity inversion using Point Spread Functions (PSFs). In this two-step least-squares imaging method, the reflectivity is recovered by explicitly computing multi-dimensional PSFs using wave-equation modeling and de-convolving these PSFs with the final migrated image.

We define a 4D formulation which is not dependent on geological and/or reservoir production constraints by introducing the concept of cross-survey PSFs (XPSFs). As shown using synthetic data examples, the joint reflectivity inversion process delivers superior results when compared to separate inversions as it ensures a more robust recovery of the 4D effects. The presented new methodology using cross-survey Point Spread Functions (XPSFs) ensures consistency of the wavefields for recovering the 4D signal.



Introduction

Successful time-lapse (or 4D seismic) studies require special care when it comes to the removal of undesirable artefacts caused by the differences in acquisition geometries. By attempting to repeat the source and receiver geometries between surveys as precisely as possible, any subsequent 4D noise is minimized. However, in some cases it is not possible to repeat the survey geometries between vintages. This is the case when different streamer acquisition azimuths are involved or when a towed streamer survey is compared with an OBS acquisition.

Some solutions to this problem have been described in the literature for implementations in both the data and image domain. For reconciling steamer and OBS data, intelligent 4D binning in the data domain can be performed by selecting trace pairs based on multiple criteria such as the Common Depth Point (CDP) distance and/or pseudo-incidence angle and azimuth similarities at the reservoir level. Dip-angle image filtering has also been suggested in order to better reconcile towed streamer and OBS migrated images by Haacke et al. (2017).

The image domain approach for correcting illumination differences between 4D datasets, that we present here, builds on previously published work concerning wave equation reflectivity inversion using Point Spread Functions (PSFs). In this two-step least-squares imaging method, the reflectivity of depth migrated images is recovered by explicitly computing multi-dimensional PSFs using wave-equation modeling and de-convolving these PSFs with the final migrated image (Valenciano et al., 2006). Ayeni and Biondi (2010) have previously exploited the PSF concept for 4D reservoir monitoring using a target-oriented joint least-squares migration approach with spatial and temporal constraints.

The aim of the study presented here is to define a 4D formulation which is not dependent on geological and/or reservoir production constraints. We modified the previous joint inversion formulations by introducing the concept of cross-survey PSFs (XPSFs). The benefit of working in the image domain is that the illumination variation can be directly evaluated and compensated for at any location in the 4D image regardless the acquisition geometries.

4D reflectivity inversion using XPSFs

As shown by Valenciano et al. (2006), a least-squares inversion framework can be used for inverting the reflectivity in the image domain. The recorded seismic data **d** can be expressed as a linear modeling operator **L** working on the subsurface reflectivity **r**: d=Lr

The reflectivity **r** therefore can be estimated using the Hessian matrix $\mathbf{H} = \mathbf{L'L}$, where **L'** is the migration operator (or adjoint to the modeling operator).

$$\hat{r} = (L'L)^{-1}L'd_{obs} = H^{-1}m_m \tag{1}$$

Where $m_m = migrated \ image$, $\hat{r} = estimated \ reflectivity$, $d_{obs} = observed \ data$

The main difficulty of solving the above equation is the need to explicitly compute the inverse of the Hessian matrix. Using PSFs, the Hessian H can very efficiently be computed in a target-oriented fashion as a set of local imaging responses of a scatter point grid.

4D joint reflectivity inversion summary

In a 4D time-lapse experiment, a matrix system can be defined using \hat{r}_0 , \hat{r}_1 and \tilde{m}_0 , \tilde{m}_1 respectively estimated reflectivity models and migrated images for the base and monitor surveys. By introducing the cross-survey Hessian term H_iH_j, the 4D joint inversion system (with no temporal/geological constraints) can be described as follow:

$$\begin{pmatrix} \begin{bmatrix} H_1 H_0 & 0\\ 0 & H_0 H_1 \end{bmatrix} + \begin{bmatrix} \varepsilon_{01} & 0\\ 0 & \varepsilon_{10} \end{bmatrix}) \begin{bmatrix} \hat{r}_0\\ \hat{r}_1 \end{bmatrix} = \begin{bmatrix} H_1 \tilde{m}_0\\ H_0 \tilde{m}_1 \end{bmatrix}$$
(2)



with regularisation terms: $\varepsilon_{01} = \lambda_{01} \langle (\mathbf{H_0H_1})^* . \mathbf{H_0H_1} \rangle$ and $\varepsilon_{10} = \lambda_{10} \langle (\mathbf{H_1H_0})^* . \mathbf{H_1H_0} \rangle$

The 4D reflectivity change is given by: $\Delta \hat{r} = \hat{r}_1 - \hat{r}_0$

The diagonal of the cross-survey Hessian corresponds to the cross-survey illumination. If one of the surveys does not illuminate a particular reflector sufficiently than the cross-survey illumination becomes very small. In contrast, if the wavefields of the base and monitor are similar at the reflector then the cross-survey illumination is maximum. The cross-survey Hessian will be represented by a set of XPSFs, which are derived from the PSFs. As a result, the illumination common to both datasets is honored to compute the changes in reflectivity caused by production effects.

4D synthetic example 1: Sigsbee2B model

The first 4D synthetic data example has been created using a subset of the Sigsbee2B 2D model. The 4D synthetic seismic data have been generated by modeling two different types of acquisitions, a streamer acquisition for the base survey and an OBN acquisition for the monitor survey (Figure 1).



The baseline towed streamer acquisition is modeled using 500 shots with a shot point distance of 50 m and a 6,000 m long cable with a receiver distance of 20 m. The monitor dataset has been modeled using the same shots geometry as for the base but the receiver network was composed of 120 OBNs positioned on the seabed 200 m apart from each other. The modeled 4D changes at the reservoir are characterized by a 40 m shift of the position of the Oil Water Contact (OWC) and a velocity increase at the reservoir location of 10%. It should be noted that no extra 4D processing steps were applied to the data pre or post-imaging.

The high values in the cross-survey illumination panel shown in Figure 2c define areas where the illuminations of the two acquisitions has been similar. The reservoir has been purposely located beneath the salt body in an area of significant salt thickness variation. The aim is to evaluate the ability of the proposed cross-survey joint reflectivity inversion to recover the true 4D reflectivity without introducing unwanted artefacts, which can be interpreted as 4D noise. The results of the reflectivity inversion are presented in Figure 3 for two different inversion approaches. In the first approach, a least- squares reflectivity inversion process is applied. Despite the very different acquisition geometries of both base and monitor surveys, the inversion process in both cases has been able to recover a meaningful 4D reflectivity response. However, the separate inversion approach has resulted in more reflectivity artefacts around the OWC, which are unrelated to the true 4D signal. The joint inversion approach has produced results that show less artefacts especially in the area outside the reservoir where no production has occurred.





Figure 3 4D reflectivity difference using separate inversion (left) and joint inversion (right)

4D synthetic example 2: SEAM time lapse project model

The second synthetic example has been generated using the 3D/4D reservoir model provided by the SEAM time-lapse project (Oristaglio, 2016; Smit et al., 2017). The 4D modeling involves two non-repeated OBN geometries. The base survey design comprised 60 OBNs laid out in a rectangular shape and the monitor survey included two squares of 30 nodes each that are located to the side of the base survey. The shot carpet is similar for both vintages. A strong 4D difference is expected at the Gas-Oil contact at 3400 m depth due to the large velocity variation as indicated in the Figures 4b and 4c.



Figure **4** *a*) *Design of non-repeated OBN geometries. b*) *Section of the SEAM 4D pilot velocity model. c*) *Depth slice at the top reservoir (3400 m) of the 4D velocity variation.*

Figure 5 displays the 4D results for depth slices at the water bottom (260 m) and at the top of the reservoir (3400 m). Figure 5a (left column) shows the 4D seismic difference results (straight seismic amplitude difference), Figure 5 (centre and right column) show the separate reflectivity inversions and the joint reflectivity inversion results respectively. At the water bottom, it can be observed that the 4D difference around the node locations appear more focused with the reflectivity inversion applied. However, the reflectivity differences do not cancel out for the separate reflectivity inversion case, as the individual illumination corrections are unconnected. In contrast, the jointly inverted reflectivity change (4D reflectivity differences) becomes negligibly small at the water bottom; a consequence of the use of the cross-survey PFSs.



Figure 5 Depth slices difference at the water bottom (top row) and at the top reservoir (bottom row).



The bottom row of images in Figure 5 represents the 4D differences as estimated at the top of the reservoir. The 4D signal is partially imaged by the seismic amplitude difference and presents large lateral amplitude variations. The separate reflectivity inversions have improved the 4D signal but the lack of a 4D constraint has over-boosted the 4D noise outside and inside of the 'D shaped' reservoir area. The joint reflectivity inversion was able to reduce the illumination induced 4D noise further and enable the recovery of a consistent 4D signal amplitude with the same geometry as the original 4D velocity model (Figure 4c).

Conclusion

Image domain 4D reflectivity inversion using multi-dimensional PSFs has been shown to compensate for significant illumination differences when different survey geometries are used for 4D imaging. The proposed methodology is advantageous in 4D studies where the geometry of the different acquisitions cannot be replicated. This is the case when towed streamer surveys with different acquisition directions are used in a 4D experiment or when streamer and OBN (Ocean Bottom Node) surveys are combined. As shown using synthetic data examples, the joint reflectivity inversion process delivers superior results when compared to separate inversions as it ensures a more robust recovery of the 4D effects. The presented new methodology using cross-survey Point Spread Functions (XPSFs) ensures consistency of the wavefields for recovering the 4D signal.

The 4D synthetic data examples shown here use models that produce strong and easy to detect 4D signals. If more subtle 4D effects need to be recovered, it will be paramount to repeat the acquisition geometries between the different time-lapse surveys as closely as possible.

The joint inversion methodology using XPSFs described here, although still beneficial when applied to repeat surveys of largely similar geometry. Combining this 4D imaging inversion technique with an optimum 4D repeated acquisition design will be preferred when subtle 4D effects are to be detected.

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