

**TIME LAPSE (4D) SEISMIC SIGNAL
DETECTABILITY IMPROVEMENT USING
CO-PROCESSING TECHNIQUE ON CARBONATE
RESERVOIRS**

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by

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LIST OF SYMBOLS

d	The seismic data
L	The Kirchhoff forward modelling operator
r	Subsurface reflectivity
m	The Kirchhoff migration
L^H	The adjoint of the Kirchhoff forward operator
r_{LS}	The subsurface reflectivity
$L^H L$	Hessian operator
$f(r)$	The cost function of least-square migration
Q	Seismic waves absorption compensation quality factor
LQ	Kirchhoff visco-acoustic modelling operator
$(L^H L)^{-1}$	Hessian operator inverse
RMS	Root mean Square.
NRMS	Normalized Root Mean Square.
a_i	Base survey trace.
b_i	Monitor survey trace.
$t_1 - t_2$	A given time window
C_{bm}	Cross-correlation of baseline and monitor trace within a given time window.
C_{bb}	Auto-correlation of baseline trace.
C_{mm}	Auto-correlation of monitor trace.
τ	Tau which equals to 300.
π	Pi is the ratio of the circumference of any circle to the diameter of that circle. In decimal form, the value of pi is around 3.14.
LM	The linear modelling operator L , M is the reflectivity model.
d	The seismic data.
d_0	The baseline survey
d_1	The monitor survey.
m_0 and m_1	The baseline and monitor reflectivity models.
L_0 and L_1	The baseline and monitor modelling operators.
S	Quadratic Cost function
$\widetilde{m}_0$ and $\widetilde{m}_1$	Migrated images of baseline and monitor
$\acute{L}_0$ and $\acute{L}_1$	Base and monitor migration operators.
L_0 and L_1	Base and monitor modelling operators.
H_0 and H_1	The Hessian matrices.
$\Delta \widehat{m}$,	Inverted time-lapse image.

dV	The velocity variation.
$\Delta t/t$	Time variation and two-way travel time.
$\Delta z/z$	The variation in thickness and the thickness.
S_{zz}	The total vertical stress variation.
S_{xx}	Stress in the x-direction.
S_{yy}	Stress in the y-direction.
E_{zz}	The vertical strain.
R	The dimensionless parameter.

LIST OF ABBREVIATIONS

4D	The time-lapse 4 th dimension
5D	Five dimensions; inline/crossline/offset/ azimuth, and time dimension
Co-processing	Same processing sequence applied on base and monitor seismic surveys
COV	Common Offset Vector
DSDR	Source & Receiver distance
Hz/DB	Hertz, measuring frequency unit for. Decibel, amplitude measuring unit
IP	Impedance of P (Compressional seismic waves)
IS	Impedance of S (Shear seismic waves)
KPSDM	Kirchhoff Pre-Stack Depth Migration
LSM	Least Square Migration
LSRTM	Least Square Reverse Time Migration
OBC	Ocean Bottom Cable
OBN	Ocean Bottom Node
OVTs	Offset Vector Tiles
PETREL	Seismic interpretation software
PreSDM	Pre-Stack Depth Migration
PreSTM	Pre-Stack Time Migration
PU	Per Unit
PZ	P (hydrophone receiver which is pressure gradient) and while the collocated geophone is recording Z, the vector displacement of the seabed.
RMO	Residual Moveout
RMS/NRMS	Root Mean Square/Normalized Root Mean Square
RTM	Reverse Time Migration
SW	Saturation Water
TL	Time Lapse
TWT	Two Way Time
USM	Universiti Sains Malaysia
DAS	Distributed Acoustic Sensing
VSP	Vertical Seismic Profile

**PENINGKATAN PENGESANAN ISYARAT SEISMIC SELANG MASA (4D)
MENGUNAKAN TEKNIK PEMROSESAN BERSAMA BAGI
TAKUNGAN KARBONAT**

ABSTRAK

Kajian seismic 4D digunakan untuk memantau prestasi takungan hidrokarbon semasa fasa pengeluaran atau suntikan. Tinjauan seismic 3D yang berulang mesti dapat mengukur isyarat 4D yang dapat mengesan penggantian cecair atau perubahan tekanan yang mungkin berlaku semasa pengeluaran takungan atau suntikan air atau gas dalam sesuatu tempoh masa. Secara praktiknya, isyarat seismic yang digabungkan dengan perubahan sedemikian boleh dilihat dalam takungan klastik sedangkan ia boleh diabaikan dalam takungan karbonat yang berheterogeni. Mengulangi kaji selidik seismic 3D dengan geometri pemerolehan yang sama biasanya sukar di Timur Tengah kerana perubahan alam sekitar (contohnya, rupa bumi kasar, arus air, perubahan bermusim, dan pemasangan pengeluaran lapangan). Oleh itu, tiga medan minyak dan gas di luar pesisir Abu Dhabi di Emiriah Arab Bersatu dengan kebolehlaksanaan pemerolehan seismic yang terhad telah dipilih untuk kajian ini. Dalam kajian ini, pemprosesan bersama seismic 4D untuk tiga kajian kes berbeza daripada input data, sama ada ia bermula dari data medan seismic atau dari pra-pemprosesan dan penyahkonvolusi yang berbeza, atau sama ada ia bermula selepas deconvolusi yang sama pada asas dan memantau survei seismic. Penyelidikan seismic 4D ini bermula dengan kajian kebolehlaksanaan 4D berasaskan kajian 1D untuk menentukan kebarangkalian isyarat 4D dari data log di telaga minyak semasa melakukan pemprosesan bersama seismic 4D tinjauan asas dan pemantauan. Pelengkungan dinamik pengesan 4D dikira untuk tinjauan survei,

menggunakan tinjauan asas sebagai rujukan untuk mendapatkan penyongsangan 4D seismik penentu yang tepat. Penggabungan balak yang baik dan jumlah seismik telah diuji untuk meramalkan jumlah sifat log di lokasi telaga jumlah penyongsangan seismik. Pembelajaran mesin yang diselidiki, Deep Feed Forward Neural Network (DFFNN), juga diuji menggunakan log yang relevan dari enam telaga untuk melatih dan mengesahkan data. Jumlah ketumpatan dan keliangan seismik 3D dicipta untuk tinjauan asas dan pemantauan dan dipadankan dengan ketumpatan dan keliangan balak dari telaga. Penyelidikan inovatif ini telah menggunakan langkah yang tepat untuk membuat pemprosesan bersama dengan seismik 4D untuk menunjukkan kebolehulangan seismik yang rendah. Ia juga telah membuktikan bahawa ketepatan yang kuat dan perubahan tekanan dalam takungan dapat mengesan isyarat seismik 4D lebih baik daripada memastikan kebolehulangan pemerolehan seismik yang sempurna. Bagi tiga lokasi di luar pesisir yang digunakan dalam kajian ini, penyelidikan ini dapat membantu mengoptimumkan penempatan telaga minyak yang lebih yang tepat untuk mencapai kawasan takungan yang baik dan tidak terganggu supaya telaga minyak dapat memberikan penghasilan maksimum dengan menanggulangi air dan gas daripada pengeluaran.

TIME LAPSE (4D) SEISMIC SIGNAL DETECTABILITY IMPROVEMENT USING CO-PROCESSING TECHNIQUE ON CARBONATE RESERVOIRS

ABSTRACT

4D seismic studies are used to monitor the performance of hydrocarbon reservoirs during production or injection phases. Repeated 3D seismic surveys must be able to measure a 4D signal that can detect fluids substitution or pressure changes that may take place during reservoir production or injection of water or gas over a period of time. In practice, the seismic signal allied with such changes could be perceptible in clastic reservoirs whereas it could be negligible in heterogeneous carbonate reservoirs. Repeating 3D seismic surveys with the same acquisition geometry is usually difficult in the Middle East due to environmental changes (e.g., rough terrain, water currents, seasonal changes, and field production installations). Therefore, three oil and gas fields of offshore Abu Dhabi in the United Arab Emirates with scarce seismic acquisition repeatability have been selected for this study. In this study, the 4D seismic co-processing for the three case studies differs from the input data, whether it starts from the seismic field data or from different pre-processing and deconvolution, or whether it starts after the same deconvolution on base and monitor seismic surveys. This 4D seismic research began with a 1D well-based 4D feasibility study to determine the probability of 4D signals from the borehole logs while performing 4D seismic co-processing of the baseline and monitoring surveys. A 4D trace dynamic warping was computed for the monitor survey, using the baseline survey as a reference to obtain an accurate deterministic seismic 4D inversion. Merging well logs and seismic volumes was tested to predicting a volume of log properties at the well locations of the seismic inversion

volume. Supervised machine learning, Deep Feed Forward Neural Network (DFFNN), was also tested using relevant logs from six wells to train and validate the data. 3D seismic volumes of density and porosity were created for the baseline and monitoring surveys and matched to the density and porosity logs from the wells. This innovative research has developed best practices for 4D seismic co-processing for such low seismic repeatability. It has also proven that strong saturation and pressure changes in reservoirs can detect the 4D seismic signal better than ensuring perfect seismic acquisition repeatability. For the three offshore fields used in this study, this research could help optimize the proper placement of infill wells to reach the undrained areas of the reservoirs so that the wells can provide maximum recovery with delayed water and gas breakthrough.

CHAPTER 1

INTRODUCTION

1.1 Introduction

While the extraction of hydrocarbons from carbonates represents a significant portion of the world's energy resource, 4D seismic is more widespread and successful in clastic reservoirs than in carbonate reservoirs (Sarkar et al., 2003; Calvert, 2005). The heterogeneity of carbonate rocks and difficult rock physics challenges, such as the high stiffness of the rock frame, could complicate the use of 4D seismic to monitor fluid changes in carbonate rocks. The seismic industry still believes that seismic acquisition repeatability is the most important success factor for meaningful 4D seismic analysis for both clastic and carbonate reservoirs. Despite the high cost of acquisition, time-lapse processing is an indispensable factor to obtain highly repeatable data (Smith et al., 2019).

Repeatability of seismic acquisition in the development areas cannot be readily guaranteed because there are more rigs, wellhead production platforms, pipelines, and infrastructure that could make it impossible to place the sources and receivers in the same location after the production period is over. In the study areas, 4D seismic surveys were conducted with different geometries of the baseline and monitor, and the 4D seismic analysis went beyond the usual way of focusing only on the seismic amplitude difference between the baseline and monitor. The creation of a petroelastic model and a 1D wellbore-based 4D feasibility study was an exceptional step in predicting the 4D signal from the wellbore information. Simulation of different water and gas injection scenarios shows the effects on impedance of

compressional and shear waves and VP / VS changes. In addition, these changes were matched with changes in fluid and gas saturation in the reservoir over time.

One of the main objectives of this research was to develop an ideal workflow for 4D seismic that can bridge the poor repeatability of seismic acquisition. In this study, joint base and monitor 4D seismic processing was tested using Kirchhoff Pre-Stack Depth Migration (KPSDM), the common seismic best practice, against Kirchhoff Pre-Stack Depth Migration Least Square Migration (LSM). LSM was found to have a lower average Normalized Root Mean Square (NRMS) value compared to KPSDM. Typical NRMS values for offshore fields in the literature range from 10% to 25% (Detomo, 2013).

For optimal calculation of dynamic time and space warping, multiple monitor surveys must be matched with a baseline survey (Hampson et al., 2005). The accuracy of dynamic time warping (DTW) in similarity estimation is being investigated for template matching and clustering of seismic traces (Kumar et al., 2022). In addition, the alignment of baseline and monitor seismic volumes in the time and depth domains has resulted in amplitude and time convergence of these two volumes. Time shifts and amplitude variations between the baseline and monitor surveys were also investigated to track fluid exchange and pressure changes. 3D seismic volumes with velocity variations were also created by scripting the 3D time-shift volumes to create the low-frequency model of the monitor, which was combined with the low-frequency baseline model to obtain the 4D seismic inversion result.

Choosing a robust analysis domain is particularly important in 4D time-lapse studies (Rafael et al., 2017), so 4D seismic inversion can be more effective than

systematic seismic wiggle-trace work in this regard. (Tarantola, 2005) has described seismic inversion as the conversion of seismic reflections into elastic physical properties of the subsurface. Such elastic properties may be related to porosity, lithology, fluid saturation, and geomechanical properties (Frazer et al., 2008). The goal of time-lapse seismic inversion is to predict changes in elastic rock properties, such as acoustic impedance (Daiane et al., 2020). In addition, the goal is to define a reservoir model that has less error between predicted and observed seismic amplitudes (Francis, 2005).

It is a fact that the reflectivity of water-bearing reservoirs increases with water saturation and decreases with decreasing frequency. The fluid in the reservoir induces a low-frequency anomaly in the seismic spectral decomposition, while a high-frequency anomaly is induced by the gas or clay content in the reservoir (Goloshubin et al., 2005). Spectral decomposition was also extracted in the third case of this study to relate frequency changes to changes in fluid and gas saturation in the seismic time course of the baseline and monitor record. Vertical 4D resolution (ideally 1-10 m) is the main challenge in 4D seismic. For carbonate reservoirs, seismic repeatability could be improved by using innovative methods to acquire 4D seismic with reservoir measurements and simulation techniques (Amundsen and Landrø, 2007). In this study, a hybrid theory and data model was used to evaluate 4D seismic inversion results and predict changes in reservoir rock properties, such as density and porosity, over the period of production from baseline and monitor surveys.

To establish a link between the logs and the seismic data at the well sites, Hampson Russell (Emerge) software was used. In geophysics, neural network is used

to quantitatively predict rock properties from seismic data (Downton, 2018). In supervised machine learning, the neural network uses a set of inputs and outputs for a given task and the relationship between inputs and outputs is determined. The disadvantage of this deep learning approach is that there must be enough inputs and outputs to adequately train the network (Hampson et al., 2001).

For Case II, 3D seismic volumes for density and porosity were created. They can be used to determine the best places to fill wells when there are areas of low density and high porosity in the reservoir that are undrained. Porosity and permeability changes in time-lapse production were matched with pressure changes in the reservoir that could explain reactivation of faults in the monitor survey. The changes in reservoir performance were matched with the observed 4D signal in seismic time-lapse for the three case studies in this research. The mechanism of 4D signal generation was investigated using the changes in water and oil saturation and pressure.

1.2 Problem Statement

Seismic acquisition repeatability is a challenge when it comes to maintaining the same shot and receiver position throughout the production period. Enhanced oil recovery (EOR) or CO₂ removal will encounter some difficulties in carbonate reservoirs, which generally have difficult reservoir characterization. The oil and gas industry also still relies on standard seismic imaging algorithms such as pre-stack Kirchhoff time and depth imaging. Both are not forgiving of geometric irregularities and poor repeatability of seismic imaging. Moreover, residual differences in time, phase, amplitude, frequencies, and background noise after cross equalization are the major obstacles to all 4D seismic monitoring surveys. The weak 4D signal is unable

to detect changes in the reservoir due to production or injection activities. Seismic interpretation of 4D seismic in carbonate is not common to date because there are no relevant seismic attributes to help with quantitative interpretation of 4D seismic.

Various studies on 4D seismic consider that 4D in carbonate reservoirs is still a major challenge and is still in the pilot and test phase and is very rarely used at production scale. In addition, the repeatability of seismic images is not optimally guaranteed. 4D is not credible enough to be used for reservoir monitoring. (Lafram et al., 2016) has shown that the 4D signal can interpret water intrusion, gas re-dissolution, and gas cap extension on 4D with the same acquisition repeatability. However, with different acquisition geometries, it was found that the gas cap expansion could not be interpreted, but the water motion could. To date, no complete 4D seismic analysis has been published, demonstrating that 4D is a model for integration: acquisition of 4D feasibility, propagation through 4D seismic co-processing to dynamic 4D trace warping cascaded through 4D seismic inversion, and finally reconciliation of results with reservoir saturation and pressure changes throughout the 3D seismic period.

1.3 Research Objectives

The main purpose of time-lapse (4D) seismic surveys is to monitor reservoir surveillance and optimize well plugging in a production field. It is also to monitor fluid and gas injection in the reservoir to improve the EOR or IOR. To achieve the goals, we have the following objectives:

- i. To determine whether 4D seismic acquisition in carbonate reservoirs can be generally successful and when the repeatability of seismic

acquisition is low in particular. Acquiring repeated surface seismic baseline and monitor surveys in development fields cannot be perfectly guaranteed.

- ii. To develop the best 4D seismic co-processing sequence that could close the gap of low repeatability of seismic acquisition, and to verify whether seismic imaging such as LSM alone can improve the repeatability of seismic acquisition. Nevertheless, seismic co-processing, starting from pre-processing, such as noise reduction, deconvolution, passing through demultiple to LSM pre-stack depth migration, is the best sequence to increase repeatability by joint seismic processing.
- iii. Integrate 4D seismic co-processing results with production and injection data to match 4D seismic signatures with reservoir fluid substitution and pressure changes over seismic time-lapse and understand the mechanism of 4D generation.

1.4 Scope of Study

To evaluate the feasibility of a meaningful 4D seismic study in carbonate, the reservoir background of the three cases in this study was investigated. The results of the 4D pilots from Case II, one of the giant offshore fields in Abu Dhabi, UAE, were the motivation for conducting this study. The analysis of the results of these 4D pilots is detailed later in this study.

1.4.1 Reservoir Background of the Study Area

In the case II, 3D OBC seismic survey was conducted over this producing field in December 2013 and completed in November 2014 (Figure 1.1). It was decided to take the opportunity of the current seismic acquisition to conduct a pilot 4D survey over 24 km² of the area of interest (Phase-1). Three swaths were acquired as monitors, using the same seismic acquisition geometry (base) from 1994 for this 4D study.

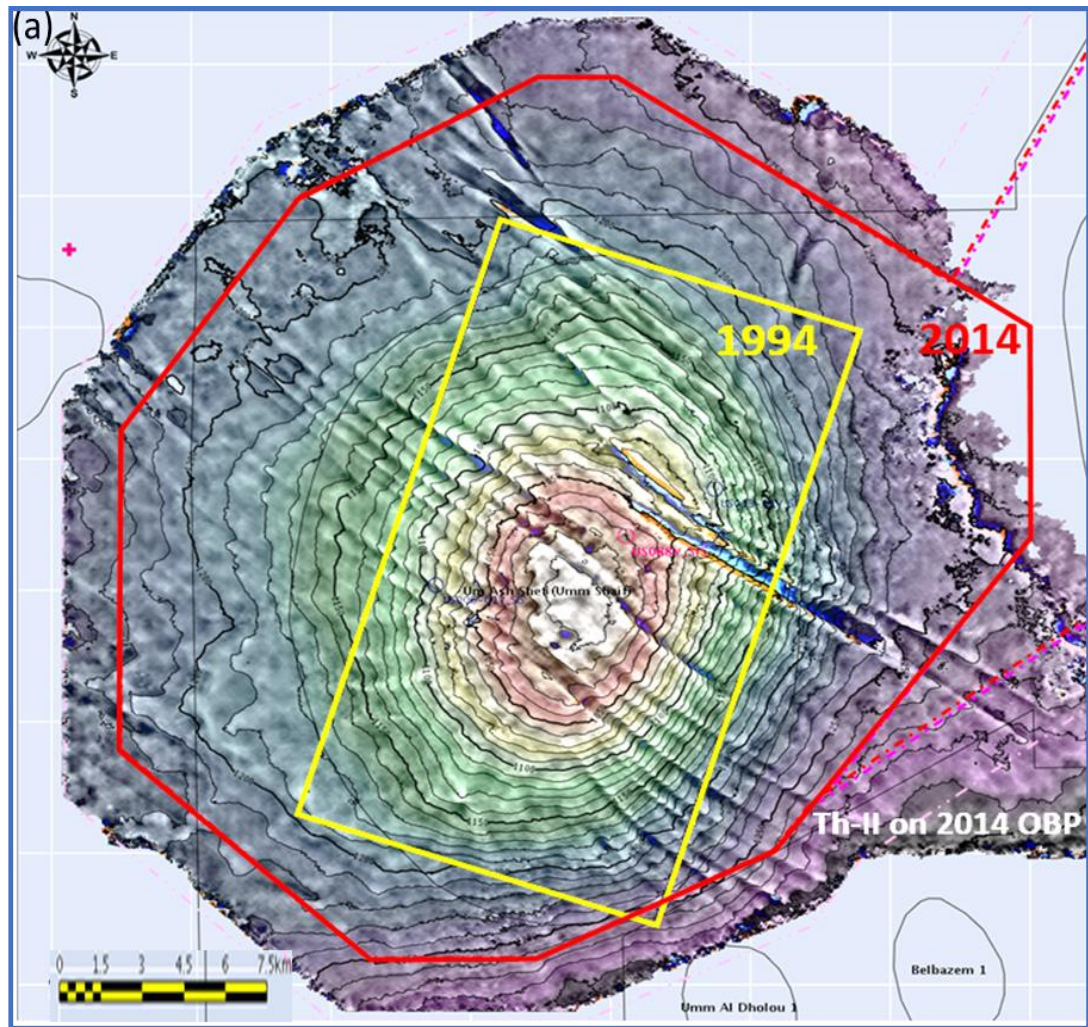


Figure 1.1 Case II; area of study: baseline 1994 3D OBC seismic acquisition (in yellow), and the new seismic acquisition 2013-2014 (red polygon).

This acquisition was an exclusive opportunity to determine the 4D signal in this offshore carbonate environment in the target reservoir of Abu Dhabi, UAE (Lafram et al., 2016). The objective of this Phase 1 4D pilot study was to test whether a reliable 4D signal could be identified above the noise level. This Phase 1 pilot study, using the same repeated geometry of base and monitor, successfully demonstrated that a 4D signal is measurable at the reservoir level of this field after a seismic time-lapse of 20 years of production.

A more sophisticated phase 2 was then initiated, using the new, completely different acquisition design as a monitor over a 57 km² test area. The second phase is a cross-sectional acquisition with a wide azimuth so that the baseline acquisition with a narrow azimuth is not repeated (Figure 1.2). More specifically, shots direction was almost 70° to the Phase 1 shot direction, which is very unfavourable in terms of repeatability. In addition, different sources and sensors were used (an array of 4 velocity meter in the baseline seismic and single accelerometers for the monitor seismic). The layout of the acquisition geometry of the new seismic acquisition consists of orthogonal overlapping zippers with 8 receiver lines and 4 receiver line rolls (Figure 1.3).

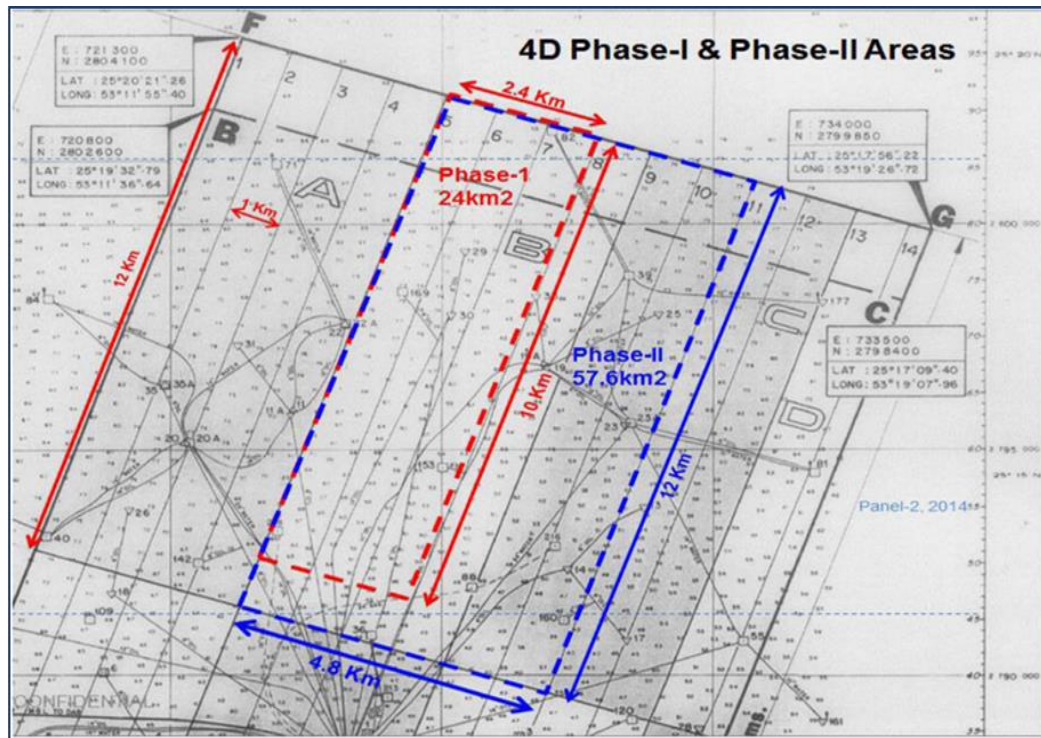


Figure 1.2 Case II; Phase 1 in red polygon and phase 2 in blue polygon outlines on the 1994 seismic acquisition survey (after Lafram et al., 2016).

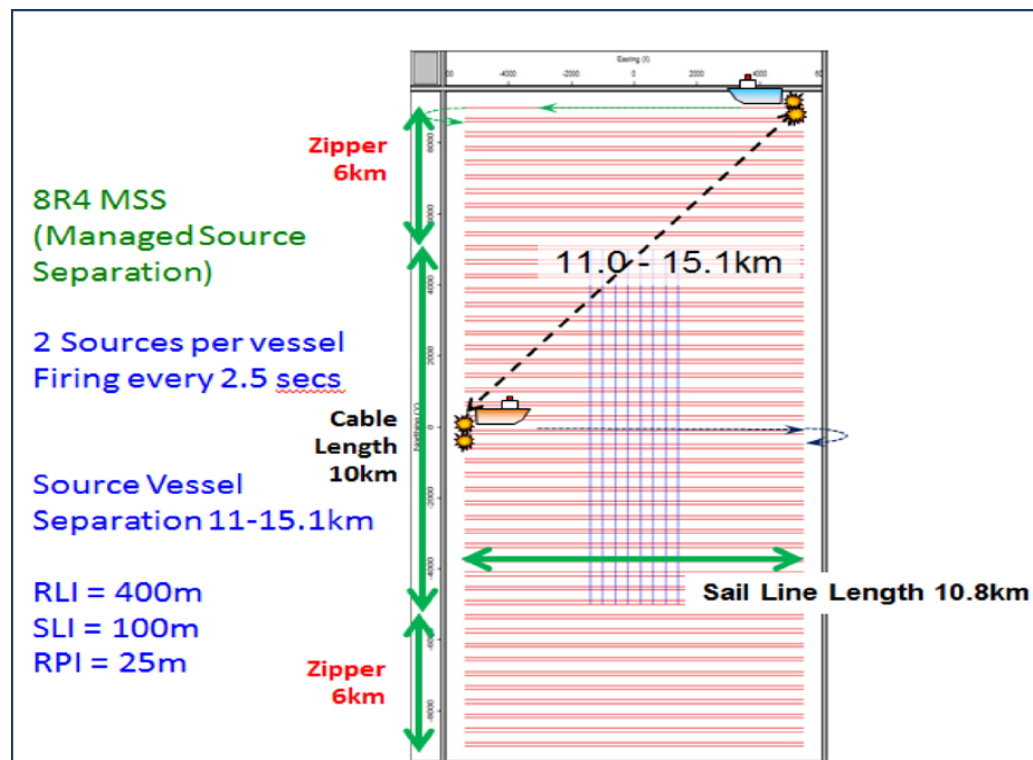


Figure 1.3 Case II; Monitor seismic acquisition survey (2013-2014) spread geometry layout.

The experience gained in Phase 1 also provided a good understanding of the global processing challenges. The interpretation results validated the processing sequence used at that time. Phase 2 needed to address the new challenges described above, and so a processing strategy was developed and updated throughout the project. The first challenge was to find the optimal denoising and de-multiple strategy with the most appropriate data modelling and subtraction in both surveys. In addition, 4D binning with the selection of the most repeatable traces without decimating too many traces was also a challenge. The second major challenge was to mitigate acquisition-related time and amplitude biases, combat the remaining non-repeatable multiples and non-repeatable coherent noise, optimizes muting, and carefully fit very different data.

It has proven to be very beneficial to have a systematic and consistent QC strategy for these very demanding pilots. Much research was conducted before the optimal workflow was developed. For some key steps, it was imperative to advance QC to seismic inversion/warping. As expected, analysis of the final products from Phase 2 shows higher repeatability noise compared to Phase 1, but initial QC after processing has yielded useful information.

1.4.2 Phase 1 Summary

In the Phase 1 summary, the 4D signal was found to be coherent in the reservoir:

- Coherent with the structural scheme (Figure 1.4)
- Coherent with the dynamic data from the wells (gas/water production)
- Coherent with the phenomena expected from the reservoir model.

One of the most important points is that the water rise can be detected, and the upper water surface can be determined at the time of monitoring. The role of the barrier at a certain level in the reservoir is well seen in the 4D data. Other production phenomena (expected from the dynamic model) such as pressure decay, gas rebound, and gas cap expansion can be interpreted in 4D. This experiment demonstrates that 4D seismic imaging in carbonates offshore Abu Dhabi (United Arab Emirates) is feasible under certain conditions for repeatability of data acquisition and with careful planning and execution of processing in a challenging seismic environment. Quality control and correction of data integrity, extensive denoising sequencing, demultiple, and correction of acquisition-related biases was necessary to reveal the 4D signal.

1.4.3 Phase 2 Summary

Several details identified during the interpretation of Phase 1, such as the extent of the gas cap in small heterogeneities and localized areas of overpressure. Those areas are difficult to find in Phase 2 (Figure 1.4); however, the 4D signal associated with water movement is robust and can be interpreted. This 4D signal was calibrated with production wells and the key issues in interpreting the 4D signal associated with water movement are as follows:

- i. The boundaries of areas crossed by water can be mapped separately.
- ii. Water is visible and can be mapped with 4D seismic in the eastern part of the phase 2 area.
- iii. Moreover, 4D shows that water has advanced further south in 2014.

- iv. The 4D interpretation shows that the water can follow a non-matrix path in some areas and finds a preferred path through fractures and structural lineaments.

The full study of the pilot phase, phase I, and phase II clearly shows that the most important parameter limiting repeatability is the ability to place the sensors and shots well, as was the case in the baseline study. This is critical for two reasons: avoidance of wave propagation bias due to local, surface changes in the subsurface and the ability to select the same subset of data for multiple processing steps. In addition, the random noise was very high in the OBC acquisition for both phases (ADNOC internal reports not published).

In Phase 1 of the pilot with the same baseline and monitor acquisition geometry, the NRMS is 13%, while in Phase 2 with a different baseline and monitor acquisition geometry, it is 28%. Seafloor currents, Schulte surface waves, and ship noise are hard on OBC seismic cables laid on the seafloor. While the Ocean Bottom Node (OBN) wireless system is highly recommended for seismic acquisition, especially because of the shallow water in this offshore field in Abu Dhabi, UAE. ADNOC has conducted the world's largest 3D onshore and offshore seismic surveys in Abu Dhabi using very high seismic acquisition parameters specifications. For offshore seismic, a 3D OBN seismic acquisition type was selected to provide seismic fold coverage of 2400 with 6 km of inline and crossline offsets, (Cambois et al., 2019).

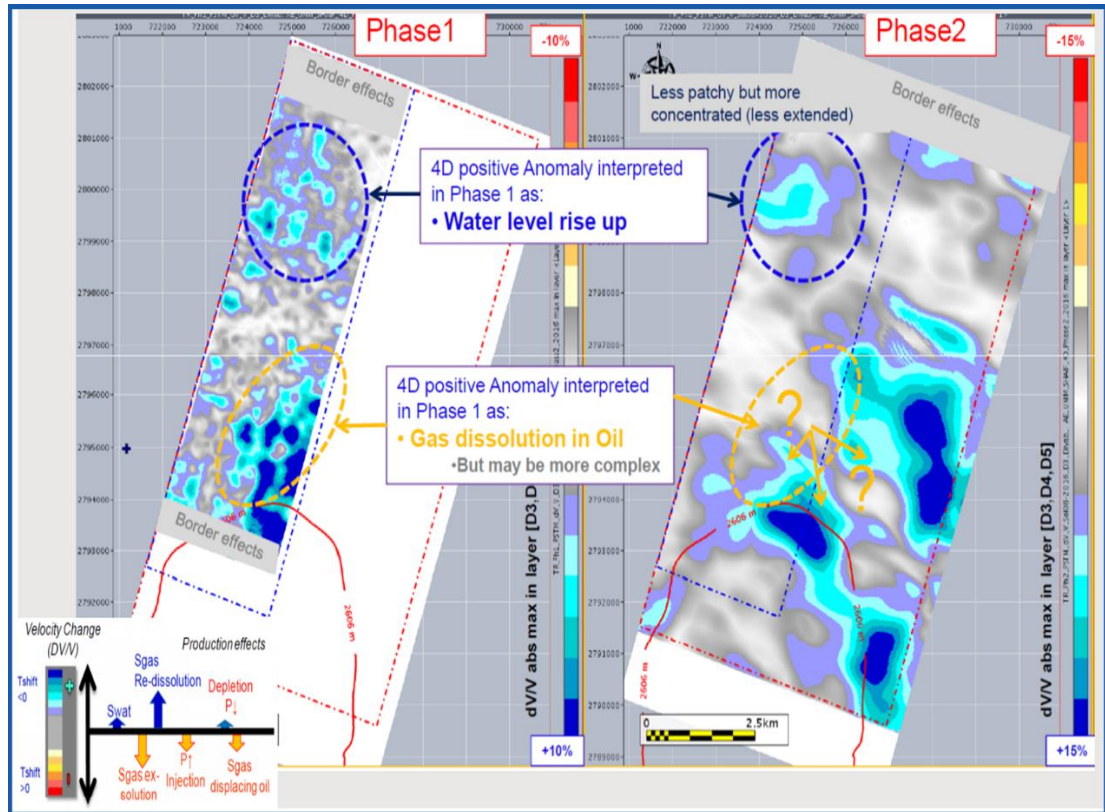


Figure 1.4 Case II; Phase 1 /Phase 2 comparison of the reservoir zonation (after Lafram et al., 2016).

1.5 Novelty of the Research

There are still two perceptions in the industry that 4D seismic in carbonate has very limited success. The first is the heterogeneity of carbonate, which is a high velocity deposit and has very subtle changes in velocity and density. These are the two essential properties required to detect 4D seismic signatures. The second is the repeatability of the seismic acquisition, which must be perfectly accurate to obtain meaningful 4D seismic in general and in carbonate rocks in particular. However, the novelty of this research has proven that there is something other than these perceptions.

The 4D feasibility of this research has shown that 4D signal capture is higher than 4% (the industry standard capture threshold). Second, the development of best practices for 4D seismic studies in general, and the creation of optimal 4D seismic co-processing in the case of low seismic acquisition repeatability in carbonate reservoirs in particular. Optimal 4D seismic co-processing, which is the main objective of this research, could improve the repeatability of seismic acquisition, which is extremely difficult to ensure in the offshore fields of Abu Dhabi (UAE).

The full integration of the three cases of this research, 4D feasibility, joint processing, inversion and matching with reservoir performance in seismic time-lapse is an unprecedented novelty in the industry and an outstanding outcome of this research.

1.6 Layout of Thesis

The background of the research and the research objectives are explained in Chapter 1. Chapter 2 covers the previously published work on 4D seismic co-processing, pre-stack depth migration (KPSDM versus LSM), followed by the theoretical aspects of least-square migration algorithm. In Chapter 2 as well, the 4D seismicity in carbonates and the characteristics of the seismic 4D metrics attributes and their corresponding theoretical background are highlighted. Chapter 3 describes the seismic and borehole data and related documents that form the input for this research, as well as the research methodology of this study. Chapter 4 (Results and Discussion) presents the outcomes of the petroelastic model and the 4D feasibility study. The pilot seismic volumes analysis of case II is illustrated and highlights the

seismic 4D co-processing workflows of this research of the three different offshore cases in Abu Dhabi, UAE.

Chapter 4 then also presents the geometric and metric attributes of 4D seismic for the processing results of the three cases. The results of dynamic 4D warping of the seismic traces and seismic inversion are also explained in Chapter 4. The application of machine learning to support the processing and inversion results is also accentuated in Chapter 4. Finally, the unprecedented research processes are highlighted by matching the outstanding results with reservoir performance, fluid exchange, and pressure changes over time of the 4D seismic surveys. Chapter 5 presents the conclusions of the research, the contribution of the research to the industry, and the recommendations for future seismic 4D investigations in carbonate reservoirs.

CHAPTER 2

LITERATURE REVIEW

2.1 Introduction

The literatures contain many previously published papers on time-lapse seismic monitoring, most of them on clastic reservoirs and only a few on heterogeneous carbonate reservoirs. For the different approaches used in this research, the literatures covering these approaches are highlighted by presenting different algorithms for 4D seismic data processing and pre-stack depth migration with their respective mathematical backgrounds. A review of publications on 4D seismic in carbonate reservoirs and the diagnostic 4D seismic metric attributes that should be investigated as a verifier to improve seismic acquisition repeatability during 4D seismic co-processing, especially when seismic acquisition repeatability is low. Very recent publications from 2022 were also discussed, highlighting the new seismic industry trend of the 4D seismic carbonate for the future.

2.2 4D Seismic Data Processing

The processing of 4D seismic data is challenging when the acquired data set contains inherent variability that obscures the expected time-lapse signal (Fischer et al. 2013). Accurate processing of seismic 4D time-lapse data, either pre-stack or post-stack, with a minimum number of matching filters can greatly improve the 4D seismic data interpretability. Seismic 4D processing pathways can be broadly divided into three main categories (Lumley et al. 2003), as shown in Table 2.1:

Table 2.1 4D Seismic Processing Routes.

Post-Stack Cross Equalization	Pre-Stack Parallel Processing	Simultaneous 4D Pre-Stack Processing
Input data are: 3D seismic migration volumes processed independently.	Input data are: Multi-vintage data sets either pre-stack processed or reprocessed.	The input data are: The multi-vintage data sets are merged or linked in many processing steps.
Need: Cross-equalization processing to improve seismic repeatability in the unmodified zone prior to 4D interpretation.	Need: Process them asynchronously with a similar or identical processing sequence.	Need: They are specially processed simultaneously to extract the processing operators that are derived and applied to the data of multiple vintages.
Data sets are not reprocessed pre-stack	Data sets are processed pre-stack	Data sets are processed pre-stack simultaneously
This type is commonly used to enhance legacy seismic data and clean up residual data from re-shoot or 4D design datasets.	This type is used to optimize the repeatability of the final product and for re-shoot and 4D design datasets.	This type of data collection is commonly used in 4D design projects.

In 4D seismic processing, the non-repeatable noise must be suppressed before the 4D signal is revealed. There are many sources of non-repeatable noise, such as variable amplitude gains, frequency content, static shifts, waveform phase changes, and events that lie between different data vintages (Lumley et al., 2003). A series of matching filters are applied to the data sets to remove sources of such non-repeatable noise (Ross et al., 1996).

4D Seismic data acquisition and processing should ensure repeatability of seismic events with respect to non-reservoir intervals, while the only differential signal could be a phase change of the reservoir fluid. Therefore, data conditioning and compensation of the non-reservoir (overburden level) is extremely important and essential (Mitra et al., 2007). Mitra has also recommended that pre-stack conditioning and post-stack matching are extremely important for the true representation of 4D seismic data.

4D seismic joint processing are successive steps started with 4D seismic trace binning and the corresponding 4D pre-processing. Spatial repeatability is enhanced by 4D binning, while concurrent 4D pre-stack processing develops common operators of processing that use cross-equalization as an essential processing step. Time-lapse (TL) seismic interpretation is divided into two categories based on the characteristics of the interpreted data: Interpretation of data sets and seismic differences between vintages from all TL vintages (Nguyen et al., 2015).

2.3 Pre-Stack Depth Migration

In the presence of inhomogeneous subsurface illumination and irregular acquisition geometry, least-square pre-stack depth migration (LSM) can restore reflectivity with amplitude reliability and resolution likelihood better than the classical Kirchhoff or even RTM algorithm (Shao et al., 2017). LSM is a user-friendly technique for 4D seismic co-processing, but LSM alone cannot improve the repeatability of seismic acquisition. Seismic co-processing shall be started from the base and monitor raw seismic filed data.

A 4D seismic case study in West Africa over a sandstone reservoir in which the 4D seismic analysis was demonstrated from feasibility to reservoir characterization (Webb et al., 2019). However, the 4D co-processing method that fills the gap of poor repeatability of seismic acquisition has not been sufficiently elaborated. Pre-stack depth migration is a standard for seismic imaging, starting from the application of Kirchhoff migration in the early 1990s to the introduction of reverse-time migration (RTM) in the late 2000s (Huang et al., 2017). However, pre-stack depth migration algorithms might induce migration artifacts such as migration swings, narrow seismic bandwidth, and amplitude distortion (Gray, 1997).

This is even true for an advanced imaging algorithm such as RTM (Baysal et al., 1983; Zhang and Zhang, 2009). Narrow seismic frequency range of depth migration often outcomes poor image illumination and resolution, because of the seismic imaging geometry and the processing technology limitations used (Liu et al., 2018). The applications of linear inversion in seismic imaging are widely known (Schuster, 1993; Nemeth et al., 1999; Prucha and Biondi, 2002). Least squares migration (LSM) aims to recover the true reflectivity of the Earth by determining the inverse of the forward modelling operator by minimising the square of the misfit between the observed data and the modelled data (Huang et al., 2017).

The LSM was proposed to overcome of the standard Kirchhoff migration limitations (Trantola, 1987; Schuster, 1993; Nemeth et al., 1999). The LSM method was applied to a 2D line CDP post-stack data in the Gulf of Mexico (courtesy of Mobile) and did show the improving of spatial resolution even when the data are not spatially under-sampled (Nemeth et al., 1999).

Moreover, least-square reverse time migration (LSRTM) is expected to enable relatively high-resolution imaging while maintaining amplitude by encompassing the inversion of the Hessian matrix. LSRTM was outstanding for use in the delineation of the reservoir and 4D seismic imaging compared to out-dated RTM and Fourier finite difference (FFD) migration (Xiao-Dong et al., 2017). Inhomogeneous subsurface illumination and irregular acquisition geometry might make standard Pre-stack depth migration (PSDM), e.g., Kirchhoff/RTM, unable to completely recover reflectivity with the likely amplitude fidelity and resolution, (Shao et al., 2017).

Shan Qu and Dirk Verschuur (2019) have proposed simultaneous joint migration inversion as an active tool combining a joint time-lapse data processing strategy for baseline and monitor surveys with the joint migration inversion method for reservoir monitoring. However, simultaneous joint inversion was not established to overcome the poor repeatability of data acquisition in terms of detectability of the 4D signals.

2.4 Theory of Seismic Imaging by Least-Squares Inversion

To understand the theoretical assumptions of least-square inversion, it is assumed that the observed data, d_{obs} of seismic imaging are assumed to be exclusive of the Earth's reflectivity. The standard pre-stack depth migration produces an estimate of the reflectivity, m as follows:

$$m = L^* d_{obs} \quad (2.4.1)$$

Here the operator L^* is the adjoint, the inverse of the forward modelling operator L . The standard depth migration produces the Earth structural image. However, this result often led to irregular illumination, low bandwidth, and low wavenumber content because migration is not the reverse modelling (Xiao-Dong et al., 2017). Unlike conventional migration (Equation 2.4.1), least square migration (LSM) resolves the modelling path inverse:

$$m = (L^* L)^{-1} L^* d_{obs} \quad (2.4.2)$$

The above explanation in Equation 2.4.2 can be obtained using two separate methods; the first clearly calculates the matrix L^*L and its equivalent inverse, while the second is alternatively an implied method that can be used iteratively to invert the operator L and answering the minimization problematic through:

$$m = \arg \min \frac{1}{2} \|d_{obs} - Lm\|^2 \quad (2.4.3)$$

The iterative LSM algorithm description is summarized in Figure 2.1, which could be fulfilled in a migration/demigration context. Inversion iteration consists of a Born modelling (Cohen et al., 1986) and a migration. The LSM sequence starts with an original pre-stack depth migration, followed by a migrated image in Born's modelling, using the migrated image as the reflectivity model. If the difference between the modelled synthetic data and the observed data is big, the data remaining $d_{obs} - d_{syn}$ is migrated and used to update the image m . In general, the inversion converges when the discrepancy is within a resilient threshold. The path of seismic LSM processing (Dong et al., 2017) is shown in Figure 2.1.

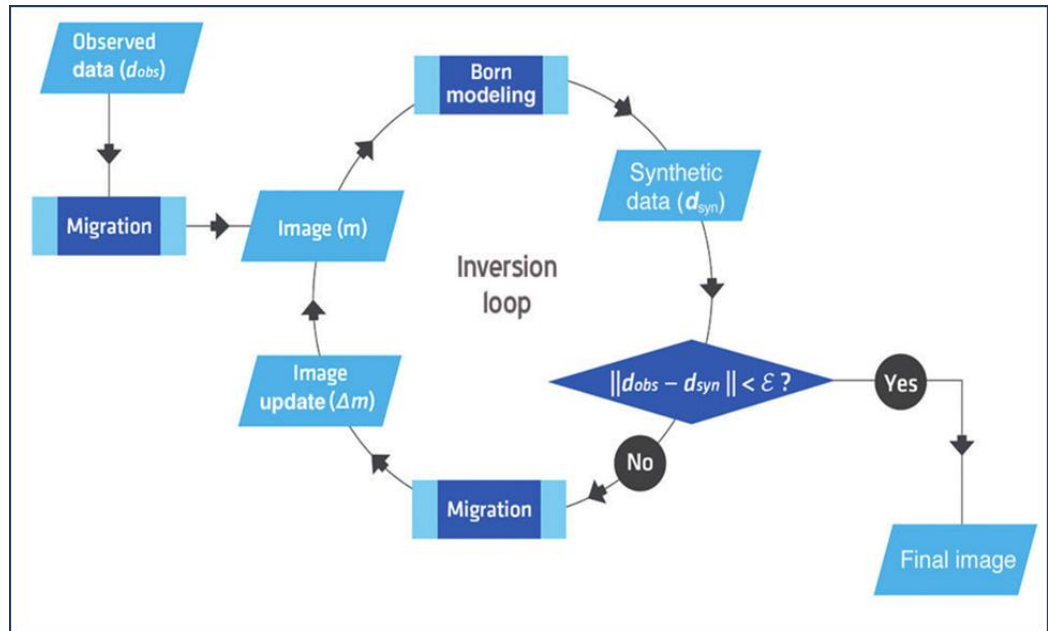


Figure 2.1 An iterative LSM algorithm workflow according to (Dong et al., 2017)

The Linear operator of original imaging Kirchhoff demigration:

$$d = Lr \quad (2.4.4)$$

Where d is the acquired seismic data, L is the acoustic operator of Kirchhoff's modelling, and r is the subsurface reflectivity. Unravelling the inverse problem, $r = L^{-1}d$, produces the desired subsurface reflectivity. Nevertheless, the calculation of this traditional inverse is not possible in seismic acquisition reality. The usual alternate is to apply the adjoint L^H , of the forward operator L , to the acquired seismic data (Claerbout, 1992):

$$m = L^H d \quad (2.4.5)$$

Where, m is the original Kirchhoff migrated image.

This image will have a lot of migration artifacts and illumination problems since the inverse operator is not used. These shortcomings shall be overawed by minimization of a least squares cost function, $f(r) = \|d - Lr\|$. Nemeth et al., (1999) has developed a formulation that ties de-migrated modelled data with observed data to appraise the migrated image. Moreover, the solution of the least-squares normal equations provides the least-squares approximation to the subsurface reflectivity r_{LS} :

$$r_{LS} = (L^H L)^{-1} L^H d \quad (2.4.6)$$

Here, $(L^H L)$ is called the Hessian operator and Equation 2.4.6 can be solved iteratively by using the conjugate gradient approaches. The Earth variable elastic properties could root absorption of seismic waves, resulting in phase distortion and

amplitude attenuation. In straight acoustic migration, these (Q) possessions are accounted for in both pre- and post-migration processing. However, the standard migration can be modified to directly compensate for these effects (Xie et al., 2009) by amplitude improvement and phase distortion elimination.

The least-squares Kirchhoff pre-stack depth migration process, visco-acoustic modelling that includes seismic absorption, the quality factor Q of the medium, could be cascaded on top of the LSM (Perrone et al., 2018). The inclusion of Q in the least squares migration takes a different approach to solving the amplitude data quality problem mentioned previously by altering the modelling in Equation 2.4.4:

$$d = L_Q r \quad (2.4.7)$$

Where, L_Q signifies the Kirchhoff visco-acoustic modelling operator (Wu et al., 2017) and this could solve the normal equations to gain a new type of equation (Equation 2.4.6) that comprises Q -compensation:

$$r_{LS} = (L_Q^H L_Q)^{-1} L_Q^H d \quad (2.4.8)$$

The migration and de-migration runs associated with iteratively solving equations (2.4.6) or (2.4.8) are expensive from a computational perspective, which raises the question: What can be a reasonable, cost-effective method? Let us start by substituting Equation 2.4.5 into Equation 2.4.8 and provide:

$$r_{LS} = (L^H L)^{-1} m \quad (2.4.9)$$

With using this equation, the least squares approximation of reflectivity is an inverse Hessian, $(L^H L)^{-1}$ operator representing the filtered version of the migrated image. The application of the inverse Hessian matrix operator to the migration is to

reduce the image artifacts due to the non-repeated source and receiver locations. Guitton (2004) proposed that could be done with non-stationary matching filters following a de-migration/re-migration process. The effect on the migrated image of the cascade $a) \rightarrow c)$ is precisely the Hessian operator (Guitton, 2004), which is inverse to estimate through matching filters in d), and final application to a) (Table 2.2).

Table 2.2 Math of LSM Seismic Processing Workflow after Perrone et al., 2018

Processing Step	Equation
a) An initial migration, m_0 :	$m_0 = L^H d$
b) A de-migration, d_0 , of the data in a):	$d_0 = L m_0$
c) A subsequent re-migration, m_1 , of the data in b):	$m_1 = (L^H d_0) = (L^H L) m_0$
d) Design matching filters to match c) to a):	$m_0 = (L^H L)^{-1} m_1$
e) Apply the matching filters from d) to the data in a):	$r_{LS} = (L^H L)^{-1} m_0$

The Curvelet domain Hessian operator was developed by both Khalil et al. (2016) and Wang et al. (2016) and is used to improve the stability and structural consistency of the LSM approach in the matching process.

Time-lapse (4D) technique has been developed in the last decade as an important tool for seismic imaging in hydrocarbon reservoir monitoring. Batzle and Wang (1992) outline the rock and fluid relationships importance; while Lumley (1995), Lumley and Rickett (2001), Calvert (2005), and Johnston (2005) have discussed the processing practical implementation. Lefevre et al., (2003),