Porosity and Permeability Prediction through Forward Stratigraphic Simulations Using GPMTM and PetrelTM: Application in Shallow Marine Depositional Settings.

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Abstract

The forward stratigraphic simulation approach is used in this workapplied to predict provide provide and permeability attributestrends in the Volve field, Norway. This was achieved by using spatial data subsurface model. Variograms and synthetic well logs from the forward stratigraphic model were combined with known data to control the distribution of guide porosity and permeability in the 3-D grid.distribution. Building a subsurface property reservoir model that fits data at different locations in a hydrocarbon reservoir is a task associated comes with high levels of uncertainty. An-Therefore, it is critical to generate an appropriate means to minimise property representation uncertainties is to use geologically realistic sediment distribution and or stratigraphic patterns framework to predict guide lithofacies units and related associated petrophysical properties. distribution in a subsurface model. The workflow used areadopted is in three parts; first, simulation of twenty scenarios of sediment transportation and deposition using the geological process modeling (GPMTM) software developed by Schlumberger was used to simulate scenarios of sediment deposition in the model area. Secondly, an estimation of the extent and proportion of lithofacies proportions in the stratigraphic model was done-using the property calculator tool in the PetrelTM software. Finally, porosity and permeability values arewere assigned to corresponding lithofacies-associations in the forward stratigraphic model to produce a forward stratigraphic-based petrophysical porosity and permeability model. Results show a lithofacies distribution that is controlled bymodel, which depends on sediment diffusion rate, sea level variation, flow rate, wave processes, and tectonic events. This observation is consistent with real-world events were-the natural occurrence, where variation in sea level-changes, volume of, sediment inputsupply, and accommodation space-control the kind of stratigraphic sequence formed.sequences. Validation wells-prefixed, VP1 and

VP2 <u>located</u> in the original Volve field petrophysical-model and the forward stratigraphic-based models show a good match in <u>significant similarity</u>, especially in the porosity and permeability attributes at 5 m vertical sample intervals. By reducing the level of property uncertainty between wells through<u>models</u>. <u>These results suggest that</u> forward stratigraphic <u>simulation outputs</u> can be used together with geostatistical modeling, an improved porosity and permeability can be achieved for an efficient field development strategy workflows to improve subsurface property representation in reservoir models</u>.

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1 Introduction

2 The distribution of reservoir properties such as porosity and permeability is a direct function of a complex 3 combination of sedimentary, geochemical, and mechanical processes (Skalinski & Kenter, 2014). The 4 impact of reservoir petrophysics on hydrocarbon field development and depletion well planning and production strategies makes it imperative to use reservoir modeling techniques that present realistic 5 6 property variations invia 3-D models (e.g. Deutsch and Journel, 1999; Caers and Zhang, 2004; Hu & Chugunova, 2008). Typically, reservoir modeling tasks requirerequires continued property modification 7 until an a appropriate match to known subsurface data is obtained. However, acquisition of. Meanwhile, 8 9 subsurface datasetsdata acquisition is costlyexpensive, thus restricts data collection and accurate 10 subsurface property modeling-condition. Several studies, e.g. Hodgetts et al. (2004) and Orellana et al. 11 (2014) have demonstrated that how stratigraphic patterns, and therefore petrophysical attributes can be 12 fairly well understood from in seismic, outcrop data, outcrops, and well logs- are applicable in subsurface modeling. However, this notion is limited by the absence of an accurate and reliable detailed 3-13 14 Definitional depositional model frameworks to guide the distribution of property variability in reservoir 15 unitsmodeling inhibits this strategy (Burges et al. 2008). Reservoir modeling techniques with the capacity to integrate forward stratigraphic simulation outputs with stochastic modeling techniques for subsurface 16 17 property modeling will improve reservoir heterogeneity characterization, because they more accurately 18 produce geological realism than the other modeling methods (Singh et al. 2013). 2013). The use of geostatistical-based methods to represent the spatial variability of reservoir properties have has been 19 20 widely accepted in many exploration and production projects (e.g. Kelkar and Godofredo, 2002). In the 21 geostatistical base modeling methods method, an alternate numerical 3-D model (i.e. realizations) is 22 derived to demonstrateshows different scenarios of property distribution scenarios that can be conditioned are most likely to match well data (Ringrose & Bentley, 2015). Typically, subsurface 23 24 modeling practioners are faced with However, due to cost reservoir modeling practitioners continue to encounter the challenge of getting a lot of obtaining adequate subsurface data to deduce reliable variogram 25

models as a result of cost variograms for subsurface modeling, therefore introducing a significant level of 26 27 uncertainty in a-reservoir modelmodels (Orellena et al. 2014). The advantages of applying geostatistical 28 modeling approaches in populating propoerties in to represent reservoir properties in models is well 29 established (e.g. are discussed in studies by Deutsch and Journel, (1999;), Dubrule, (1998), but). A notable disadvantage is that the geostatistical modeling method tends to confine reservoir property 30 models distribution to known subsurface data and rarely realize produces geological realism to capture 31 32 sedimentary events that have led to reservoir formation (Hassanpour et al. 2013). In effect, the 33 geostatistical modeling technique is unable todoes not reproduce a long-range continuity of continuous 34 reservoir properties that, which are essential for generating realistic reservoir connectivity models 35 (Strebelle & Levy, 2008). Based on lessons from a previous work (e.g. Otoo and Hodgetts, 2019), the The 36 forward stratigraphic simulation approach is againwas applied in this contribution to predictforecast 37 lithofacies-units, porosity, and petrophysical propertiespermeability in a 3-Dreservoir model. An important, based on lessons from Otoo and Hodgetts (2019). A significant aspect of this work is the use 38 39 ofusing variogram parameters from forward stratigraphic-based synthetic wells to populate petrophysical 40 properties, especially within inter well regions of simulate porosity and permeability trends in the reservoir under studymodel. Forward stratigraphic modeling involves the uses morphodynamic rules to 41 42 derive sedimentary depositional patterns to reflectreplicate 3-dimensional stratigraphic 43 observations depositional trends observed in real-data. The approach is driven by the (e.g. seismic). 44 Forward stratigraphic modeling operates on the guiding principle that multiple sedimentary process-based 45 simulations in a 3-D framework will most likely improve our understanding on spatial variation of facies, 46 as well as and therefore petrophysical properties property distribution in a geological system model. The sedimentary system, Hugin formation makes up the main geological process modeling GPMTM 47 48 software (Schlumberger, 2017), which operates on forward stratigraphic simulation principles, replicates

50 <u>study area. The reservoir interval in the Volve field. According to studies under study is within the Hugin</u>

a depositional sequence to provide a 3-dimensional framework to predict porosity, permeability in the

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51 <u>formation. Studies</u> by Varadi et al. (1998); Kieft et al. (2011),) indicate that the Hugin formation is made

78 upconsists of a complex depositional architecture of waves, tidestidal, and riverine riverine fluvial 79 processes; suggesting. This knowledge suggests that a single depositional model will not be adequate to produce a realisiterealistic lithofacies or petrophysical distributions model of the area. Furthermore, the 80 complicated Syn-depositional rift-related faulting system, significantly influenceinfluences the 81 82 stratigraphic architecture (Milner and Olsen, 1998). The Therefore, the focus of this study here is to produce a depositional sequence, which captures subsurface attributes observed in the shallow marine 83 environment by using a forward stratigraphic modeling approach in the GPMTM (Schlumberger, 2017), 84 seismic and use variogram parameters from the forward model to control porosity and permeability well 85 data to guide property representation in a 3-D model modeling. 86

87 Study Area

88 The Volve field (Figure 1), located in Block 15/9 south of the Norwegian North Sea-is Jurassic in age 89 (i.e. late Bajocian to Oxfordian) with, has the Hugin Formation as the main-reservoir unitinterval from which hydrocarbons are produced (Vollset and Dore, 1984). The Hugin formation, which is Jurassic in 90 91 age (late Bajocian to Oxfordian), is made up of shallow marine to marginal marine sandstone deposits, coals, and a significant influence of wave events that tend to control lithofacies distribution in the 92 formation (Varadi et al. 1998; and Kieft et al. 2011). Several studies, e.g. Studies by Sneider et al. (1995),) 93 94 and Husmo et al. (2003) associate sediment deposition ininto the Hugin systemstudy area to-a rift-related 95 subsidence and successive flooding during a large transgression of the Viking Graben within the Middle 96 to Late Jurassic period. Previously it was interpreted to comprise Also, Cockings et al. (1992), Milner and 97 Olsen (1998) indicate that the Hugin formation comprises of marine shoreface, lagoonal and associated coastal plain, back-stepping delta-plain, and delta front-deposits (e.g. Cockings et al. 1992; Milner and 98 99 Olsen, 1998), but. However, recent studies, e.g. by Folkestad and Satur, (2006) suggest the influencealso .00 provide evidence of a stronghigh tidal event, which introduces another dimension in property that requires 101 attention in any subsurface modeling oftask in the reservoirstudy area. The thickness of the Hugin 102 formation is estimated to range between 5 m and 200 m, but can be thicker off-structure and non-existent 103 on structurally high segments as a result of due to post-depositional erosion (Folkestad and Satur, 2006).

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104 Based on studies by Kieft et al. (2011), aA summarised sedimentological delineation within the Hugin 105 formation is presented inderived based on studies by Kieft et al. (2011). In Table 1. Lithofacies, 106 lithofacies-association codes A, B, C, D, and E used in the classification represents represent bay fill units, shoreface sandstone facies, mouth bar units, fluvio-tidal channel fill sediments, and coastal plain facies 107 108 units, respectively. In additionAdditionally, a lithofacies association prefixed code F-was interpreted to consist, which consists of open marine shale units, mudstone with. Within it are occasional siltstone beds, 109 110 parallel laminated soft sediment deformation that locally develop at bed tops. The lateral extent of the 111 code F lithofacies package in the Hugin formation is estimated to be 1.7 km to 37.6 km, but the total 112 thickness have of code F lithofacies is not been completely penetrated known (Folkestad & Satur, 2006).

113 Data and Software

114 This work is based on the description, and interpretation of petrophysical datasets in the Volve field by 115 Statoil, now Equinor. Datasets include 3-D seismic data, and a suite of 24 wells that consist of formation 116 pressure data, core data, petrophysical and sedimentological logs. Previous works such as studies by Folkestad & Satur, (2006) and Kieft et al., (2011) in this reservoir interval show varying grain size, 117 sorting, sedimentary structures, bounding contacts of sediment matrix that play a significant part of the 118 119 reservoir petrophysics. Grain size, sediment matrix, and the degree of sorting will typically drive the 120 volume of the void created, and therefore the porosity and permeability attributes. Wireline-log attributes 121 such as gamma--ray (GR), sonic (DT), density (RHOB), and neutron-porosity (NPHI) were used to 122 distinguish lithofacies units, stratigraphic horizons, and zones that are required to buildessential for 123 building the 3-D property model. Porosity, and permeability models, of the Volve field, were generated in Schlumberger's PetrelTM software. ImportantlyBesides, this workstudy also seeks to produce 124 125 geologicallya realistic depositional architecture that is comparable to a real-worldmodel like the natural 126 stratigraphic framework in a shallow marine environment. Deriving a representative depositional setting. 127 Therefore, obtaining a 3-D dimensional stratigraphic model of the reservoir that shows a similar

- stratigraphic sequence observed in the seismic data allows us to deduce geometrical and variogram
 parameters to serve as input datasets in actual subsurface property modeling.
- 152 The Twenty forward stratigraphic simulations were produced in the geological process modeling (GPMTM) software developed by Schlumberger was used to undertake twenty forward stratigraphic 153 154 simulation in an attempt to replicate theto illustrate depositional processes that resulted in the build-up of 155 the reservoir. Simulations were constrained interval under study. By the fourth simulation, there was a 156 development of stratigraphic patterns that shows similar sequences as those observed in seismic, hence 157 the decision to constrain the simulation to twenty scenarios because the desired stratigraphic sequence and associated sediment patterns were achieved at the fourth simulation. Several process modeling 158 159 software packages exist and have been applied in similar studies; e.g., Delft3D-FlowTM; Rijin & Walstra, (2003); DIONISOSTM; Burges et al. (2008). The geological) are examples of subsurface process modeling 160 (GPMTM)-software was preferred because of the used in similar studies. The availability of the GPMTM 161 software license, and also the ease in integrating of its capacity to integrate stratigraphic simulation outputs 162 intoin the property modeling workflow in PetrelTM, is the reason for using the geological process modeling 163 164 software in this study.

165 Methodology

The workflow (Figure 2a) combines the stratigraphic simulation capacity of the GPMTM-software in different depositional settings, sedimentary processes and the property modeling tools in PetrelTM to predict the distribution of porosity and permeability properties away from wellknown data. Three This involves three broad steps have been used here to achieve this goal;: (i) forward stratigraphic simulation (FSS)-in GPMTM-software (2019.1 version), (ii) lithofacies classification using the calculator tool in PetrelTM, and (iii) lithofacies, porosity; and permeability modeling in PetrelTM (2019.1 version).

Process ModelingForward Stratigraphic Simulation in GPMTM

The GPMTM is commercial software consist of different developed by Schlumberger to simulate clastic 173 and carbonate sedimentation in a deep or shallow marine environment. GPMTM consists of geological 174 175 processes designed to such as steady flow, sediment diffusion, tectonics, and sediment accumulation that 176 rely on physical equations and assumptions to replicate sediment-the process of sedimentation in a geological basin. A realistic realization of a stratigraphic pattern as observed in seismic or well data 177 178 provides a 3-dimensional framework to constrain subsurface property representation that conforms with 179 the real-world property distribution trends. In clastic sedimentation, the movement of sediments relies on 180 equations from the original SEDSIM developed in Stanford University (Harbaugh, 1993). Sediment movement, erosion, and deposition in clastic and carbonate environments. Example, the steady flow 181 182 process is efficient for simulating sediment deposition in fluvial bodies, whilst the unsteady flow process 183 control sediment transportation from the basin slope into deep-water basin setting, largely in theis governed by a simplified Navier Stokes equation. "Simplified" because the Navier-Stokes equation in its 184 185 original form of basinal floor fan units. Previous studies, e.g. define sediment movement in a 3dimensions differential form, while the flow equation in GPMTM is 2-dimensional with an arbitrary input 186 of flow depth. Kieft et al₇. (2011) identified describe the influence of riverine (a combination of fluvial), 187 and wave processes in the genetic structure of sediments in the Hugin formation. These geological 188 189 processes could be very are rapid, depending on accommodation space generated as a result of by sealevel variation, and or sediment composition and flow intensity. Sediment The deposition, of sediments 190 191 into a geological basin and its response to post-depositional sedimentary and or tectonic processes are 192 significant in the ultimate distribution of subsurface lithofacies units; hence the variation of input 193 parameters to increase our chance attaining outputs that fall within acceptable limits of what may exist in 194 the natural order.and petrophysics. Therefore, several input parameters for the forward simulation to attain 195 a stratigraphic output that fits existing knowledge of paleo-sediment transportation and deposition into 196 the study area (see Table 2). The forward simulation generated geologically realisticat all stages portrayed 197 geological realism concerning stratigraphic frameworkssequence, but it also revealed some limitations,

such as instability in the simulator when more than three geological processes and sub-operations run at
 a time. In view of<u>run concurrently. Given</u> this, the diffusion and tectonic processes are<u>remained</u> constant
 features whiles other processes like<u>varying the</u> steady flow, unsteady flow, and sediment accumulation,
 compaction were varied. processes in each simulation run.

202 ParametersSteady & Unsteady Flow Process

The steady flow process in GPM simulates flows that change slowly over a period, or sediment transport
 scenarios where flow velocity and channel depth do not vary abruptly e.g. rivers at a normal stage, deltas,
 and sea currents. Considering the influence of fluvial activities during sedimentation in the Hugin
 formation, it is significant to capture its impact on the resultant simulated output.

207 The unsteady flow process can simulate periodic flows such as turbidites where the occurrence is not 208 regular, and the velocity of flow changes abruptly over time. The unsteady flow process applies several 209 fluid elements driven by gravity and friction against the hypothetical topographic surface. Otoo and Hodgetts (2019) illustrate how the unsteady process in GPMTM attains realistic distribution of lithofacies 210 211 units in a turbidite fan system. Although the steady and unsteady flow governing equations distantly rely 212 on the Navier-Stokes equations, the steady flow is quite distinct, as it uses a finite difference numerical 213 method for faster computation and to also illustrate the frequency of flow that is characteristic in channel flow such as rivers. The finite difference method applies an assumption that flow velocity is constant 214 215 from channel bottom to surface. In contrast, the unsteady flow uses the particle method from SEDSIM3 to solve the sediment concentration in flow and sediment transport capacity (Tetzlaff & Harbaugh 1989). 216 The simplified equation in GPMTM attempts to solve the problem of "shallow-water free-surface flow" 217 218 over an arbitrary topography surface (Tetzlaff, D. personal communication, February 2021). "Shallow 219 water" indicates the instance where only the vertically-averaged flow velocity and flow depth are applied 220 and kept track of as a function of two horizontal coordinates.

221 <u>The equation that control steady and unsteady flow is expressed through:</u>

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$$\frac{\partial h}{\partial t} + \nabla . hQ = 0 \tag{1}$$

223 <u>Where: h is flow depth, t is time, and Q the horizontal flow velocity vector.</u>

24
$$(\frac{\partial Q}{\partial t} = -(g\nabla)H + \frac{c_2}{\rho}\nabla^2 Q - \frac{c_2 Q/Q}{h}$$

225 Where: $\frac{\partial Q}{\partial t}$ is the Lagrangian derivative of flow relative to time, g is gravity, H is the water surface 226 elevation, c₂ is the fluid friction coefficient, ρ is the water density, c₁ is the water friction coefficient and 227 h is the flow depth.

(2)

228 <u>The Manning's equation is applied to relate flow, slope, flow depth and hydraulic radius channels with a</u>
 229 <u>constant cross-section for the steady flow process. Manning's formula states:</u>

$$\underline{\mathbf{V}} = \frac{k}{n} R_h \frac{2/3}{5} S^{1/2}$$
(3)

Where: V is the flow velocity, k is the unit conversion factor, n is the Manning's coefficient which
 depends on channel rugosity, R_h is the hydraulic radius and S is the slope.

As mentioned earlier, the unsteady flow process uses the particle method equation, which relies on the assumption that erosion and deposition depend on the balance between the flow's transport capacity and the "effective sediment concentration". The equation for multiple-sediment transport in flow is given as follows:

$$\underline{\mathbf{A}_{\rm em}} = \sum_{k_S} \frac{l_{KS}}{f_{1k_S}} \tag{4}$$

238 Where: A_{em} is the effective sediment concentration of mixture, l_{ks} is the sediment concentration of each 239 type, and f_{1} , k_{s} is the transportability of each sediment type.

240 <u>The transport capacity of a sediment type is expressed by equations (5) and (6). Let consider</u>

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$$\underline{\mathbf{R}} = (\mathbf{A} - \underline{\mathbf{A}}_{em}) \underline{\mathbf{f}}_{2,\underline{\mathbf{k}}_{S}} \tag{5}$$

242 Where f_{2,k_s} is the erosion-deposition rate coefficient for sediment type k_s . For every sediment type k_s , 243 the formula for transporting sediment of different grain sizes is given as:

244
$$\underbrace{(H-Z)}_{Dt} \stackrel{Dl_{Ks}}{=} \begin{cases} R & if \ R > 0 \ and \ \tau_0 \ge f_{3,k_s} \ and \ k(x,y,z) = K_s \\ or \ R < 0 \ and \ K_s = 1 \ or \ l_{k_s-1} = 0 \\ 0 & otherwise \end{cases}$$
(6)

- 245 <u>Where;</u>
- 246 <u>H is the free surface elevation to sea level, Z is the topographic elevation for sea level, K_{s} is the sediment</u>
- 247 <u>type, l_{ks} is the volumetric sediment concentration of a specific type (k).</u>

248 Sediment Diffusion Process

- 249 <u>The diffusion process replicates sediment movement from a higher slope (source location) and deposition</u>
 250 into a lower elevation of the model area. Sediment diffusion runs on the assumption that sediments are
- transported downslope at a proportional rate to the topographic gradient, making fine-grained sediments
- easily transportable than coarse-grained sediments. Sediment diffusion depends on three parameters: (i)
- sediment grain size and turbulence in the flow, (ii) diffusion curve that serves as a unitless multiplier in
 the algorithm and, (iii) diffusion coefficient. The diffusion coefficient depends, among other variables on
 the type of sediment and "energy" of the depositional environment. In this contribution, the highest depth-
- dependent diffusion coefficient occurs near sea level, where the "energy" is highest over a geological
- time (Dashtgard et al. 2007).

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²⁵⁸ In GPMTM, sediment diffusion is calculated using a simplified expression:

$$\frac{\partial z}{\partial t} = D_i \nabla^2 z + S_n$$
(7)

- where **z** is topographic elevation, D_i is the diffusion coefficient, **t** for time, and $\nabla^2 z$ is the laplacian of z, and S_n is the sediment source term.
- 262 Sediment diffusion (D_i) is estimated by assuming that the grain size for each sediment component (coarse
 263 sand, fine sand, silt, and clay) are known. Also an assumption that these sediment types have a uniform
- diameter (D) in the flow mix (Dade & Friend 1998; and Zhong 2011). In that case, external fore (F_e),
- which consist of drag, lift, virtual mass, and Basset history force is given as:

266
$$\underline{F_e} = \alpha_e \underline{M_e} + \alpha_e \underline{\Phi_D} \cdot \frac{U_{fi} - U_{ei}}{T_p}$$
(8)

267 <u>M_e is the resultant force of other forces with the exception of drag force, T_p stokes relation time, expressed 268 <u>as: T_p = $\rho_{\rho} D^2 / (18\rho_f V_f)$, with ρ_f and V_f as density and viscosity of fluid respectively. Φ_D is a coefficient 269 that accounts for the non-linear dependence of drag force on grain slip Reynolds number (R_p).</u></u>

270
$$\underline{\Phi_{\rm D}} = \frac{\kappa_{\rm P}}{24} C_D \qquad (9), \text{ with } C_{\rm D} \text{ sediment grain coefficient.}$$

With the flow component in place, the diffusion coefficient (D_i) is deduced from the Einstein equation.
 Using an assumption that the diffusion coefficient decreases with increasing grain size and rise in
 temperature, and that the coefficient f is known, the expression for D_i is:

$$\underline{\mathbf{D}}_{\underline{i}} = \frac{K_{B}.T}{f} \tag{10}$$

275 Meanwhile, f is a function of the dimension of the spherical particle involved at a particular time (t). In
 276 accounting for f, the equation for D_i changes into:

$$\underline{D_i} = \frac{K_B T}{6.\pi \eta_0 r}$$
(11)

278 Sediment Accumulation

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279 The sediment accumulation process in GPM is designed to generate an arbitrary amount of sediment 280 representing the artificial vertical thickness of a lithology as interpreted in a well or outcrop data (Tetzlaff, D., personal communication, February 2021). The areal input rates for each sediment type (coarse-281 grained, fine-grained sediments) use the value of the map surface at each cell in the model and multiply 282 283 it by a value from a unitless curve at each time step in the simulation to estimate the thickness of sediments 284 accumulated or eroded from a cell in the model. Sediment accumulation in the GPM software requires 285 other processes such as steady flow and diffusion to account for sediment transport (sediment entering or leaving a cell) before a deposition/year (mm/yr) function to artificially produce the height of sediment 286 287 deposited per cell. The accumulation of sediments in GPM is expressed as:

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 $\underline{\mathbf{A}_{\mathrm{T}}} = \sum_{S=1}^{n} \left[(M_{\nu 1} \ \underline{*} S_{c1}), \quad \mathbf{n} \right]$ (12)

313 <u>Where;</u>

A_T is the total sediment accumulated in a cell over a period, S is the sediment type, M_v is the map value of sediment in each cell, and S_C is the sediment supply curve as a function of topographic elevation.

Boundary Conditions for Forward Stratigraphic Simulation

A realistic<u>Realistic</u> reproduction of stratigraphic patterns <u>in</u> the <u>studymodel</u> area <u>requirerequires</u> input parameters (<u>also known as initial conditions</u>). These include: a hypothetical), such as paleo-topography, sea-_level curves, sediment source location, and distribution curve, tectonic <u>events (i.e. event maps</u> (subsidence and uplift), and sediment mix velocity. The application of these input parameters in the GPMTM-simulator, and their <u>influenceimpact</u> on the resultant stratigraphic framework <u>are explainedis</u> below.

323 Hypothetical Paleo-Surface: The hypothetical paleo-surface, on which topographic for the stratigraphic 324 simulation commences was inferred is from the seismic section. Here, we assumed that (Figure 3), using the 325 assumption that the present day stratigraphic surface, also referred to as the (paleo shoreline in Figure 3a4a) 326 occurred as a result of basin filling through different over geological periodstime. Since the hypothetical topography 327 generated surface obtained from the seismic section have undergone various phases of subsidence and uplifts-over time, , it is significant to note that the paleo topographic surface used in this work does not present represent an 328 329 accurate description of the basin at the period of sediment deposition. To mitigate this; thus presenting another 330 level of uncertainty, 5 in the simulation. To derive an appropriate paleo-topographic for this task, five paleo 331 topographic surfaces (TPr) were generated stochastically, by adding or subtracting elevations from the 332 inferred paleo topographic surface or base topography (see Figure 4g) using the equation:

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 $TPr = Sbs + EM_{-}$ (13)

334 where, Sbs is the base surface scenario (in this instance, scenario 6), and EM an elevation below and 335 above the base surface. In this work, Fo

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The paleo-topographic surface in scenario 3 (figure 3d) was used as the paleo-topographic surface, 4d) is
 selected because it produced a stratigraphic sequences that fit the conceptual knowledge of depositional
 framework as observed inpatterns interpreted from the seismic section (Figure 5d).

365 Sediment Source Location: Based on regional well correlations in previous studies (e.g., Kieft et al. 2011), 366 and seismic interpretation of the basin structure interpreted from seismic data, the sediment entry point for this task wasis, placed in the north-eastern section of the hypothetical paleo-topography. Since the surface. \$67 368 The exact sediment entry point is uncertain, multipleinto this basin is unknown, so three entry points were 369 placed at a 4 mkm radius around the primary location in (Figure 3c), in order) to capture possible sediment source locations- in the model area. The source position is characterised by a positive integers (i.e. integer 370 (values greater than zero) to enable fluid flowsediment movement to other parts of the 371 372 simulation topographic, surface.

Sea Level: Primarily, the <u>The</u> sea_level variation relative to elevation was inferred<u>curve is deduced</u> from published studies and facies description in shallow marine <u>depositional</u> environments (e.g. Winterer and Bosellini, 1981). <u>Considering the limitations in the software, we assumed aTo</u> sea level of <u>was constrained</u> 30 m for short simulation runs, e.g. (5000 to 20000 years to attain stability in the simulator and), but varied <u>it accordingly</u> with <u>the</u> increasing duration of the simulation. (see Table 2). The peak sea-level in the simulation represents<u>depicts</u> the maximum flooding surface, (Figure 5d), and therefore anthe inferred sequence boundary in the geological process model.

380 Diffusion and Tectonic Event Rates: The sediment mix proportion and diffusion rate for the simulation 381 were stochastically inferred from previous studies (e.g. Burges et al., 2008), primarily to attain a 382 prograding and or aggrading clinoforms features that are noticeable in real world geological outcrops. The subsidence and uplift rates were kept constant in most part of the model. The, diffusion rate, and 383 384 tectonic event functions are inferred from published works; e.g. studies such as Walter, (1978;), Winterer and Bosellini, 1981, and(1981), and Burges et al., (2008). The diffusion and tectonic event rates were 385 increased or reduced to produce a stratigraphic model that fit our knowledge of the basin evolution. The 386 387 simulation parameters applied (Table 2) were generated randomly using the in the study area. For

example, in scenario 1 (Figure 6a), the early stages of clinoform development show resemblance to
 interpreted trends in the seismic section (Figure 3b). The process commenced with a diffusion coefficient
 of 8 m2/a, but it varied at each scenario to obtain diffusion coefficients to improve the model. Excluding
 the initial run (Figure 6a) as a guide. The guiding principle for parameter selection is their capacity to
 produce stratigraphic outputs that depict different depositional scenarios in the shallow marine setting. A
 sudden change in subsidence rate tends to constrain coarse to medium sediments at proximal distance to
 source location than in scenarios where the rate of subsidence was made gradual.

The influence of topography (Figure 4d), input parameters in the simulation is evident whenever geological
 processes such as wave events, steady/unsteady flow, diffusion, and tectonic events used curve functions
 to provide variations in the simulation.

398 The sensitivity of input parameters in the forward stratigraphic simulation is notable when there is a slight 399 change of value in sediment diffusion, and tectonic rates or dimension of the hypothetical topographic 400 surfaces.topography. For example, a change in sediment source position has a strong impact on affects the 401 extent and depth to which of sediments are deposited deposition in the basin simulation. Shifting the source 402 point to the mid-section of the topography (the mid-point of the topography in a basin-ward direction) resulted in the accumulation of distal elements that are identical to turbidite lobe systems. This output is 403 404 consistent with morphodynamic experiments (e.g. by de Leeuw et al., (2016)), where abruptsediment 405 discharge of sediments from the basin slope leads to the build-up of basin floor fan units. Stratigraphic 406 patterns generated using different input parameters provides 3-D perspective into subsurface property 407 variations under alternating initial conditions.

408 **Property Classification in Stratigraphic Model**

In our opinion, the most appropriate <u>output is the stratigraphic</u> model in this work is **Figure 5d**. This <u>point</u> <u>of view</u> is because, <u>it produced a stratigraphic sequence that mimics compared to</u> the depositional <u>sequencedescription</u> in the shallow marine depositional environment under study. The stratigraphic model <u>was converted into a 3-D format, 20 m x 20 m x 2 m grid cells in order to be used instudies such as</u>

Folkestad and Satur (2006); Kieft et al. (2011), and the property modeling tool in PetrelTM. Lithofacies, 439 440 porosity, and permeability properties are characterized in the stratigraphic using a rule based approach 441 (Table 3).seismic interpretation presents a similar stratigraphic sequence. Sediment distribution in each 442 time step of the simulation werewas stacked into a single zone framework to attain a simplified model. 443 This was done with the assumptionstrategy assumes that sedimentary processes that lead to the final buildup of genetic related units within zones of the forward stratigraphic architecture model will not vary 444 445 significantly over the simulation period. Property classification in the The stratigraphic model (Figure 5d) was achieved with converted into a 3-D format (20 m x 20 m x 2 m grid cells) for the property calculator 446 toolmodeling in PetrelTM. 447

Facies, porosity, and permeability representation in Petrel.the stratigraphic model was done via a rule 448 based approach in PetrelTM (see **Table 3**). The classification is driven by depositional depth, geologic 449 450 flow velocity, and sediment distribution patterns as indicated in Figure 7. Lithofacies representation in 451 the stratigraphic model was based<u>relied</u> on the sediment grain size pattern, and proximity to sediment 452 source. For example, shoreface lithofacies units were characterized using are medium-to-coarse grained sediments to that are, which accumulate at a proximal distance to the sediment source, whiles. In contrast, 453 454 mudstone units are constrained to the distal parts of the stratigraphic model, where confined to fine-grained sediments accumulate atin the enddistal section of the simulation domain. 455

PorosityUsing knowledge from published studies by Kieft et al. (2011), and wireline-log attributes such 456 as gamma ray, neutron, sonic, and density logs, porosity and permeability variations werein the 457 458 stratigraphic model are estimated from published wireline-log attributes (e.g. Kieft et al., 2011), which is outlined in (Table 1. Based). In previous studies on petrophysical report of the Sleipner Øst, and Volve 459 460 field (StatoilEquinor, 2006), a deduction was made to the effect that high net to gross zones will be 461 associated with; Kieft et al. 2011), shoreface deposits make up the best quality reservoir units; classified as shoreface lithofacies, whiles lagoonal deposits formed the worst reservoir, units, whilst low net-to-gross 462 zones-were interpreted to be connected with high proportions of shale or. With this guide, shoreface 463 sandstone units and mudstone-deposits./shale units in the forward stratigraphic model are best and worst 464

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Fo Fo reservoir units respectively. The porosity and 184 permeability values in Table 4 were derivedare from
 equations in Statoil's petrophysical report of the Volve 185 field (StatoilEquinor, 2016):

$$\mathcal{Q}_{er} = \mathcal{Q}_{D} + \alpha \star (NPHI - \mathcal{Q}_{D}) + \beta - (14)$$

where ϕ_{er} is the estimated porosity range, ϕ_D is density porosity, α and β are regression constants; ranging between -0.02 - 0.01 and 0.28 - 0.4 respectively, *NPHI* is neutron porosity. In instances where NPHI values for lithofacies units is not available from the published references, an average of 0.25 was used.

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 $KLOGH_{er} = 10^{(2 + 8 * PHIF - 5 * VSH)}$ (15)

where $KLOGH_{er}$ is the estimated permeability range, VSH is the volume of clay/shale in the lithofacies unit, and *PHIF*, the fractured porosity. The *VSH* range between 0.01 - 0.12 for the shoreface units, and 0.78 - 0.88 for lagoonal deposits.

499 **Property Modeling in Petrel**TM

The workflow (**Figure 2b**) used for subsurface property (e.g. lithofacies, and petrophysical) modeling in PetrelTM is extended<u>applied</u> to represent lithofacies, porosity, and permeability properties in the forward stratigraphic model. These processes <u>includeinvolve</u>:

503 <u>1.(1)</u> Structure modellingmodeling: identified faults within the study area are modelled modeled 504 together with interpreted surfaces from seismic and well datacorrelation to generate the main 505 structural framework, within which the entire property model will beis built. The procedures 506 involve modification of <u>Here</u>, fault pillars and connecting fault bodies <u>are linked</u> to <u>one another to</u> 507 attainobtain the kind of fault framework interpreted from the seismic and core data.

Pillar gridding: <u>building</u> a "grid skeleton" that is made up of a top, middle and base architectures. Typically, there are pillars which join corresponding corners of every grid cell of the adjacent -grid, forming to form the foundation of for each cell within the model; hence its nomenclature as a corner point gridding. The prominent orientation of faults (I-direction) within the model is area was in an N-S and NE-SW direction, so the "I-direction" was set the major

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direction along which grid cells align.to NNE-SSW to capture the general structural description

of the area.

Horizons, Zones, and Vertical Layering: stratigraphic horizons and subdivisions (zones) 541 (2)(3)delineatesdelineate the geological formation's boundaries. As stratigraphic horizons are 542 543 inserted introduced into the model grid, the surfaces are trimmed iteratively and modified along faults to correspond with displacements across multiple faults. Vertical layering on the other hand 544 545 defines hows the thicknesses and orientation between the layers of the model. In order Layers refers to honoursignificant changes in particle size or sediment composition in a geological 546 formation. Using a vertical layering scheme makes it possible to honor the fault framework, pillar 547 548 grid, and horizons that have been derived. Cell thicknesses are defined. A constant cell thickness 549 of 1 m is used in the model to control the vertical scale, in which subsurface properties such as of \$50 lithofacies, porosity, and permeability attributes are modelled modeling. (3)(4) 551 Upscaling; which: involves averagingthe substitution of finersmaller grid cells in **5**52 orderwith coarser grid cells. Here, log data is transformed from 1-dimensional to assign property 553 values a 3-dimensional framework to the cells and evaluate which discrete value suits eachselected data point. It also encompasses in the generation model. One advantage of coarser grids (i.e. lower 554 resolution grids) in the geological model, in orderupscaling procedure is to make simulation the 555 556 modeling process faster.

557 **Porosity and Permeability Modeling**

The Volve field porosity and permeability model that was built byfrom Equinor for their operations wasare adopted as the base (reference) model. The model, which cover an area of covers 17.9 km² was generated with the reservoir management software (RMS), developed by Irap and Roxar (EmersonTM). The original petrophysical model has a grid dimension of 108 m x 100 m x 63 m, and was compressed by 75.27% of cell size. from an approximated cell size of 143 m x 133 m x 84 m. To achieve a comparable model resolution to as the original Volve field porosity and permeability model, the forward stratigraphic output was, which had an initial resolution of 90 m x 78 m x 45 m, is upscaled to a cell size grid of 107 m x 99 m x 63 m. TwoVariograms being a critical aspect of this work, we submit two options were explored with respect to the use of to extrapolate variogram parameters derived from the forward modelstratigraphic-based synthetic wells.porosity and permeability models. In Option 1-was to assign, the porosity and permeability values were assigned to the synthetic lithofacies wells to correspond to that correlate with known facies-associations as indicated association in the study area (see Table 4.-).

595 The synthetic pseudo wells with comprising porosity and permeability data are placed situated in-between 596 actual-well (known data) locations to guide porosity and permeability property distributions imulation in **5**97 the model. For option 2, the best-fit forward stratigraphic model was populated withchanges by assigning porosity, and permeability attributes. attribute using the general stratigraphic orientation captured in the 598 599 seismic data (NE-SW; 240°). Porosity and permeability pseudo (synthetic) logs arewere then extracted from the forward stratigraphic output to build the porosity and permeability models (Figure 8). The 600 601 second option provided a broader framework for evaluating the reliability of forward stratigraphic 602 simulation on property Porosity modeling is through normal distribution in areas of sparse data. Taking 603 into account, whiles the permeability models were produced using a log-normal distribution and the 604 possibility corresponding porosity property for collocated co-kriging.

605 Considering that vertical trends in options 1 and 2 will most likely produce abe similar trend inwithin a 606 sampled interval, it is our opinion that option 2 will provide presented a viable 3-D representation of 607 property variations in the major and minor directions of the forward stratigraphic model. Ten synthetic 608 wells, (SW), ranging between 80 m and a 120 were m in total depth (TD), are positioned in the forward model to capture the vertical distribution of porosity-permeability at different sections of the forward 609 610 stratigraphic model. Typically, sediment distribution, and associated petrophysical attributes are directly 611 related to depth within the geological model; thus aiding in the analysis of the most likely proportions of 612 subsurface properties that match with observations in known well data-based models.

The forward based synthetic wells (Figure 9 c) with porosity and permeability logsdata were upscaled to
 populated, and distributed into the original structural model using the sequential Gaussian simulation

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640 method. The synthetic wells derived from the stratigraphic model served as an additional control for 641 porosity and permeability modeling in the Volve field. Because the variogram-based modeling approach is efficient in subsurface data conditioning, this idea presents an opportunity to get more wells at no 642 643 additional cost to control porosity and permeability distribution. The variogram model (Figure $10_{\frac{1}{2}}$) of 644 dominant lithofacies units in the formationstratigraphic model served as a guide in the estimation 645 ofestimating variogram parameters from the forward model. Afor porosity and permeability modeling. 646 The variogram has major and minor range of 1400 m and 400 m respectively, and an average sill value of 0.75 derived from forward stratigraphic-based synthetic wells were used to populate porosity and 647 648 permeability properties in the model. Porosity models were derived with a normal distribution, whilst the 649 permeability models were produced using a log-normal distribution and the corresponding porosity 650 property for collocated co-kriging. Out. Six out of fifty model realizations, six realizations that 651 showedshow some similarity to the original petrophysical porosity and permeability model are 652 presented formed the basis of our analysis (Figure 11). The selection of six realizations was on a visual and statistical comparison of zones in the original Volve field model and the stratigraphic-based 653 654 porosity/permeability model. The statistical approach involved summary statistics from the reference model and the stratigraphic-based porosity/permeability model. In contrast, the visual evaluation 655 compared the geological realism of forward stratigraphic-based realizations to the base model. 656

657 **Results**

The stratigraphic model in stage 4 (**Figure 5d iv**) shows the final geometry after 700,-000 years of simulation time. <u>Initial The initial stratigraphic simulation produced a progradation sequence with foreset-</u> like features (**Figure 5d i**). A) and a sequence boundary, which indicates the highest sea level in the model separates the initial simulated output from the next prograding phase (**Figure 5d ii**). <u>Initiation of an</u> aggradation<u>An aggradational</u> stacking pattern starts,commences and becomes prominent in stage 3 (**Figure 5d iii**). <u>This is These aggradational sequences observed in the forward stratigraphic model are</u> consistent with <u>real-world scenario-natural events</u> where sediment supply matchup with accommodation ⁶⁹¹ space generated as a result of the relative constant sea <u>due to sea-</u>level rise within a period. The diffusion ⁶⁹² process in GPMTM was used to define the stratigraphic architecture before introducing additional ⁶⁹³ geological processes such as steady flow, unsteady flow, wave events to capture the range of possible ⁶⁹⁴ depositional styles that have been discussed in published literature (e.g. Folkestad & Satur, 2006; Kieft ⁶⁹⁵ et al., 2011).period (Muto and Steel, 2000; Neal and Abreu, 2009).

696 The impact Impact of the forward stratigraphic simulation on porosity and permeability representation in 697 the reservoir model is evaluated evident by comparing its outcomes to the original Volve field porosity and 698 permeability models of the Volveby using two synthetic well prefixed (VP1 and VP2. The synthetic well are-); sampled at a 5 m intervals vertically to estimate the distribution of porosity and permeability 699 700 attributes along wells. Considering that the original porosity and permeability model-vertical interval. Taking into account the fact that the Volve field petrophysical model (Figure 11a) have undergonewent 701 702 through various phases of history matching to enableobtain a model to improve well planning and guide 703 production strategies in the Volve field, it is reasonable to assume that porosity and permeability 704 distribution in the Volve field petrophysical model will be geologically realistic and less uncertain. AThis 705 view formed the basis for using the porosity and permeability models developed by Equinor as a reference 706 for comparing outputs in the stratigraphic model. Table 5a shows an almost good match in porosity was 707 observed in validation wells that penetrate the model realizations; at different intervals in the forward 708 stratigraphic-based models (i.e. R14, R20, R26, R36, R45, and R49 (Table 5a). The vertical distribution 709 (Figure 12)). An analysis of porosity the well logs in selected the model realizations area shows that a 710 modal distribution range (i.e. large proportion of reservoir porosity is between 0.18 - 0.24) that. Also, the analysis of the forward stratigraphic-based porosity model is consistent with the original model. The 711 712 forward stratigraphic based model have been derived porosity range in the Volve field model (see Figure 713 <u>12).</u>

A notable limitation with an this approach is the assumption that variogram parameters, and stratigraphic inclination within zones will remain remained constant. However, throughout the simulation. The difference in permeability attributes between the original petrophysical permeability model takes into 717 accountand the forward stratigraphic-based type is the application of other measured attributes, which 718 could be the main driver of the differences in permeability estimates noted in parameters in the original 719 model (Table 5b-). Typically, a petrophysical model like the Sleipner Øst and Volve field model will take 720 into account factor in other sources of datadatasets such as special core analysis (SCAL) and other 721 petrophysical evaluations level of cementation, which enhances reservoir petrophysics assessment. 722 Bearing in mind that the forward stratigraphic model did not involve some of this additional information 723 from the reservoir section, so, it is reasonably reliable practicable to suggest that results obtained in the forward stratigraphic-based porosity and permeability models have been-adequately conditioned to 724 known subsurface data. 725

726 **Discussion**

727 The results Results show the influence of sediment transport rate, (or in this example, diffusion rate,), 728 initial basin topography, and proximity to sediment source location on the stratigraphic simulation in thein GPMTM-software. Notably, variations in . Compared to studies such as Muto & Steel (2000) and Neal & 729 Abreu (2009), we observed that a variation in sea--level controls the volume of sediment that could beis 730 retained or transported further into the basin;, therefore controlling the kind of resultant stratigraphic 731 732 sequences that are generated. In a related work-by, Burges et al. (2008), it was established) suggest that; 733 for example, a sediment-wedge topset width wasconnects directly linked to the initial bathymetry, in which the sediment-wedge structure was formed, as well as develops, and the correlation between 734 sediment supply and accommodation rate. This opinion is in line with observations in this workstudy, 735 736 where the initial sediment deposit controls the geometry of subsequent phase phases of depositions-737 Since the in the hypothetical basin. The uncertainty of initial conditions used in this work led to the generation of this basin is uncertain, multiple simulation forward stratigraphic scenarios were carried out 738 739 to account for the range of bathymetries that may have influenced the build-up of sediments sediment transportation to form the Hugin formation. present-day reservoir units in the Volve field. 740

767 The simulation produced well-defined clinoformsloping depositional surfaces in a stratigraphic 768 architecture (clinoforms) and sequence boundaries that depict the pattern observed patterns seen in the 769 seismic data. As indicated in other studies, (e.g. In their work, Allen and Posamentier, (1993;); Ghandour 770 and Haredy, (2019) explained the importance of sequence stratigraphy is vital in the lithofacies 771 characterization of lithofacies in shallow marine settings; hence, the forward stratigraphic simulation outputs provide a good framework to better understand the variation of lithofacies units in the reservoir 772 773 through a 3-D perspective. A, and therefore petrophysical property distribution in sedimentary systems. 774 Also, sediment deposition into a geological basin in the natural order is controlled by mechanical and 775 geochemical processes that modify petrophysical attributes (Warrlich et al. 2010); therefore, using 776 different geological processes and initial conditions to generate depositional scenarios in 3-dimension 777 provides a framework to analyse property variations in a hydrocarbon reservoir. The approach produces 778 a porosity-permeability model that match comparable to the original petrophysical model was produced 779 using synthetic porosity and permeability logs from the forward stratigraphic model as input datasets in 780 the sequential Gaussian simulation algorithm. As mentioned previously, this exercise work did not take into accountinclude variations in the layering scheme that develops in different zones of the stratigraphic 781 782 model. we concede that Under this circumstance, there is a possibility to overestimate and or 783 underestimate porosity and permeability properties as observed property in some sampled intervals of in 784 the validation wells. In view of Therefore, we suggest that the forward stratigraphic simulation outputs 785 such as the example presented in this, it is our suggestion that forward stratigraphic simulation outputs 786 should be applied contribution serve as additional data to understand sediment distribution patterns, and 787 associated vertical and horizontal petrophysical trends in the depositional environment-than using its 788 outputs as an, and not as absolute conditioning data in subsurface property modeling.

The assumptions made <u>inconcerning</u> the type of geological processes, and input parameters to use in the <u>stratigraphic</u> simulation <u>significantlycertainly</u> differ from what <u>may have</u> existed during the period of <u>sediment</u> deposition. <u>ApplyingSo, applying</u> stratigraphic models that fit a basin-_scale description to a <u>relatively</u> smaller scale reservoir <u>context</u> presents another <u>degreelevel</u> of uncertainty in the approach-<u>used</u> Fo

793 here. For example, in their study,. This finding agrees with Burges et al., (2008) shows), where they 794 indicate that the diffusion geological process simulation fits the description of large--scale sediment transportation; suggesting. This view further buttresses the point that an extrapolation of its 795 796 outputs integrating forward stratigraphic simulation into a well-scale framework could produce results has 797 a high chance of producing outcomes that deviate from the real-world architecture. In reality, sediment deposition into a geological basin is also controlled by mechanical and geochemical processes, which 798 799 tend to modify a formations petrophysical attributes (Warrlich et al. 2010), hence, the application of different geological processes and initial conditions to produce different depositional scenarios, from 800 801 which a best fits stratigraphic framework of the reservoir can be selected. Many forward stratigraphic-802 based subsurface modeling studies (e.g. description. In line with observations in Bertoncello et al. (2013;); Aas et al. (2014;); and Huang et al. (2015), have identified and discussed some) in relations to limitations 803 804 with the technique. Considering that similar challenges were faced in this work, caution must be taken in 805 using the outputs from in the forward stratigraphic simulations in real simulation method, it is advisable to use its outputs cautiously in reservoir modeling; as this such outputs from forward stratigraphic models 806 807 could lead to an increase uncertainty in theproperty representation of lithofacies and petrophysical properties. bias in a model. 808

809 The correlation between reservoir lithofacies and petrophysics, and its prediction through reservoir 810 models, have been extensively examined in previous several studies, e.g. (Falivene et al. (...,2006); Hu and 811 Chugunova, (2008), but). Meanwhile, the difference in predicted and outputs most often do not depict the 812 actual reservoir character is less understood. This in large part is due to the absence of a realistic 3-D 813 stratigraphic framework to guide reservoir property representation in geocellulargeological models. It is 814 our opinion that The forward stratigraphic modeling methods provide method, notwithstanding its 815 limitations, provides reservoir modeling practitioners a better platform to generate appropriate 3-D 816 lithofacies models to improve petrophysical property prediction in a reservoir, but its outputs should be 817 used cautiously and together with verifiable subsurface patterns from seismic and well datamodels that 818 reflect the natural variation of reservoir properties.

845 **Conclusion**

846 In this paper, spatial data variogram parameters from a forward stratigraphic simulation is are combined 847 with subsurface data from the Volve field, Norway to constrain porosity and permeability distribution in inter-well regions of the Volve field model-area. As. The caution, for subsurface modeling practitioners 848 849 is that the forward stratigraphic simulation scenarios presented in this contribution do not ultimately prove that spatial and geometrical data derived from forward stratigraphic modeling can be used asmodels are 850 851 absolute input parameters for a real-world reservoir modeling task. Uncertainties in the choice of initial conditionboundary conditions and processes for the stratigraphic simulation led the variation of input 852 853 parameters in order to attain a depositional architecture that is geologically realistic and comparable to 854 the stratigraphic correlation suggested in some published studies of the study area. Significantly, the good The match in porosity obtained from by comparing validation wells in the original and stratigraphic-855 856 based petrophysical model, leads us to the suggestion indicates that an integration of combining variogram 857 parameters from real well data and forward stratigraphic simulation outputs will improve property prediction in inter-well zones. In addition, this This suggestion supports the idea that more conditioning 858 859 data (well data) will increase the chance of producing realistic property distribution in the model area. 860 This work also made some key findings:

- For a-specific application of forward stratigraphic modelingsimulation in GPM[™] and a range of model parameters, the process of sediment transportation and deposition is influenced by based
 on diffusion rate, and proximity to sediment source. This opinion is consistent with several
 published works on sequence stackingstratigraphy and or system tracts in shallow marine settings,
 but. However, further work with different stratigraphic modeling simulators could be useful in
 mitigatingmitigate some of the challenges faced in this work.
- A geologically viable 3 DA lithofacies distribution in the shallow marine Hugin formation that is
 consistent with previous studies was achieved, which produced in the stratigraphic model. This
 position is evident in scenarios where sediment distribution vertically matches with lithofacies
 variation in a sampled interval in an actual well log.

871	Geologically feasible stratigraphic patterns generated in the forward stratigraphic model provide
872	additional confidence in the representation of lithofacies, and therefore porosity and permeability
873	property variations in the depositional setting under study. By reducing the level of property
874	uncertainty between wells, a reliable reservoir model can be generated to guide field planning and
875	development in the hydrocarbon exploration and production industry. The resultant forward
876	stratigraphic-based porosity and permeability model suggests that forward stratigraphic simulation
877	outputs can be integrated into classical modeling workflows to improve subsurface property modeling
878	and well planning strategies.

Future studies will focus on using an artificial neural network approach to classify lithofacies associations in the forward stratigraphic model in an attempt to reduce uncertainties that arise from cognitive or sampling biases in the calculator (or rule-based) approach for estimating lithofacies proportion in a forward stratigraphic model. In addition, efforts will be made in a future contribution to compare the stratigraphic property distribution with ones that are generated more classical methods such as sequential indicator simulation (SIS), and object-based modeling.

885 Data and Code Availability

The datasets <u>used infor</u> this work <u>was obtainedare</u> from Equinor on their <u>operations in</u> Volve field operations, Norway. <u>This The data</u> include: 24 suits of well logs, and 3-D reservoir models in Eclipse and RMS formats. The data, models (eclipse and RMS formats), and the rule-based calculation script to generate lithofacies and porosity/permeability proportions are archived on Zenodo as Otoo & Hodgetts, (2020).

890 **GPM**TM Software

The version (2019.1) version of GPMTM software was used in completing this work after an initial 2018.1 version.
 Available on: https://www.software.slb.com/products/gpm. The software license and code used in the GPMTM
 cannot be provided, because Schlumberger does not allow the code for its software to be shared in publications.

894 Model Availability in PetrelTM

The work started in PetrelTM software (2017.1)), but it was initially used for the task, but completed with PetrelTM software (2019.1);). The software is available on: https://www.software.slb.com/products/petrel. The software runruns on a windowsWindows PC with the following specifications: Processor; Intel Xeon CPU E5-1620 v3 @3.5GHz 4 cores-8 threads, Memory; 64 GB RAM. The computer should be high end, because a lot of processing time is required to execute afor the task. The forward stratigraphic models are achieved in Zenodo as Otoo & Hodgetts, (2020).

901 Author Contribution

Daniel Otoo designed the model workflow, conducted the simulation using the GPMTM software, and
evaluated the results. David Hodgetts converted the Volve field data into Petrel compactible format for
easy integration with outputs from the <u>forward</u> stratigraphic simulation.

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Fig 1. Location map of the Volve field, showing gas and oil fields in quadrant 15/9, Norwegian North Sea (Adapted from Ravasi et al., 2015).





Fig 2. Schematic workflow of processes involved <u>in</u> this work. a. providing information of <u>initial boundary</u> conditions (or input parameters) that were used in the forward stratigraphic simulation in $\text{GPM}^{\text{TM}}_{\frac{1}{2}}$ b. <u>demonstrating demonstrate</u> how the forward stratigraphic were model are converted into a grid that is usable in the PetrelTM environment</sup> for onward 3 D-porosity and permeability modeling.





Fig 3. 3-D seismic section of the study area, from which the hypothetical topographic surface $\frac{\text{was}is}{\text{was}is}$ derived for the simulation. The sedimentary entry point into the basin is located in the North Eastern section, (based on previous study in the model area (e.g., Kieft et al. 2011).





Fig 4. Inferred paleoPaleo topographic surface from seismic, also. Also, illustrating different topographic surface scenarios used in that are produced for the simulation.





Fig 5. a. present-day top and bottom topographic surfaces of the Hugin formation; b. hypothetical topographic surface derived from seismic data; c. geological processes involved in the <u>forward stratigraphic</u> simulation; d. forward stratigraphic models at different simulation time <u>intervals</u>.



Fig 6. Stratigraphic simulation scenarios depicting sediment deposition in a shallow marine framework. **a.** scenario 1 involves equal proportions of sediment input, a relatively low subsidence rate and low water depth, **b.** scenario 10 uses high proportions of fine sand and silt (i.e. 70%) in the sediment mix, abrupt changes in subsidence rate, and a relatively high water depth, c. scenario 15 involves very high proportions of fine sand and silt (i.e. 80%), steady rate of subsidence and uplift in the sediment source area, and a relatively low water depth.





Fig 7 a. Sediment distribution patterns in the geological process modeling software. b. lithofacies classification using the property calculator tool in PetrelTM.



Fig 8. <u>PropertyLithofacies, porosity and permeability</u> characterization in the stratigraphic <u>usingmodel through</u> the property calculator tool in <u>PetrelPetrelTM</u>. Also <u>showing, is</u> a cross-sectional view <u>throughof</u> the <u>model3-D models</u>.





Fig 9. Synthetic wells derived from a forward stratigraphic-driven porosity and permeability models.



model. The average separation distance between the synthetic wells shown in Figure 9c is about 0.9 km apart (maximum and minimum separation distance of 1.3 km and 0.65 km, respectively).



Fig 10. Variogram model of dominant lithofacies units extracted from the FSM forward stratigraphic model. The points indicate the number of lags in the variogram. The distance between these lags is about 100 m. This figure shows the lags between sample pairs for calculating the variogram in the major direction (NE-SW) of the stratigraphic model.



b. Forward Modeling-Based Porosity and Permeability Model Realizations



Fig 11, <u>Comparing original Original</u> Volve field model <u>toys</u> the forward modeling-based models. Realizations 16, 20, 26, 36, 45, and 49 on the left half are porosity models, <u>whilstwhiles</u> realizations 12, 20, 26, 35, 42, and 48 on the right half showare permeability models.

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Figure 12.-Illustrating how; a. <u>12a</u>. Comparing porosity in validation well<u>Well</u> 1 in five stratigraphic-based realizations, and b. the original model at similar vertical intervals.



Figure 12b. Comparing porosity in validation well-Well 2 samples in the synthetic forward five stratigraphic-based model compares to pseudo wells from realizations, and the original Volve field petrophysical model.

at similar vertical intervals	

			Thickness (t);	Wireline-log	
Code	Facies	Description	extent (I)	Attribute	Interpretation
	Al	Parallel-laminated mudstone with occasional siltstone inputs. Monospecific pattern of disorder bivalves parallel to bedding.	t= 30-425 cm l= 6 to 29 km	GR= 41-308 API DT= 225-355 µsm ⁻¹ NPHI= 0.17-0.45 v/v RHOB= 2280-2820 gcm ⁻³	Restricted marine shale
	A2	Interbedded claystone and very fine-grained sandstone; non- parallel and wavy lamination. Scarcely bivalve shells oriented parallel to bedding.	t= 10-725 cm l = 8 km to 13 km	GR= 71-65 API DT= 189-268 µsm ⁻¹ NPHI=? RHOB= 2280-2820 gcm ⁻³	Muddy Shallow bay-fill
A	A3	Fine to medium grained sandstone; moderately to well sorted grains. Wavy bedding, cross bedding, rare wave ripples	t= 60-370 cm 1 <8 km	GR= 18-46 API DT= 199-314 µsm ⁻¹ NPHI= 0.07-0.52 v/v RHOB= 1690-2745 gcm ⁻³	Sandy shallow bay-fill
	A4	Coarse to fine-grained sandstones with alternating upward fining to coarsening trend. Moderately sorted grains. Sparse sedimentary structures.	t= 250-500 cm l = _1.8 km to _4.2 km	GR= 7-35 API DT= 175-230 µsm ⁻¹ NPHI= 0.038-0.146 v/v RHOB= 2280-2820 gcm ⁻³	Marine channel- fill sandstones
	Bl	Upward-coarsening siltstone to fine-grained moderate sorted sandstones, with shell debris, and quartz granules.	t= 30-480 cm l = <2 km	GR= 18-80 API DT= 168-291 µsm ⁻¹ NPHI= 0.038-0.191 v/v RHOB= 2322-2723 gcm ⁻³	Distal lower shoreface
В	В2	Very fine-fine grained, moderate to well sorted sandstone. Fine grained carbonaceous laminae, typically low angle cross beds.	t= 130-440 cm l = 1.7 km - 8 km	GR= 20-56 API DT= 179-277 µsm ⁻¹ NPHI= 0.048-0.168 v/v RHOB= 2314-2696 gcm ⁻³	Proximal lower shoreface
	B3	Coarsening upward, cross laminated, fine to medium grained, well sorted sandstone; consist carbonaceous fragments	t= 425-800 cm l = 1.7 km - 8 km	GR= 15-25 API DT= 250-275 µsm ⁻¹ NPHI= 0.09-0.113 v/v RHOB= 2271-2342 gcm ⁻³	Upper Shoreface
	Cl	Highly bioturbated siltstone to very fine sandstones, which has beds of rounded granules	t= 175-1010 cm 1 = 7.2 km -	GR= 20-80 API DT= 230-260 µsm ⁻¹ NPHI= 0.08-0.169 v/v PHOR= 2327 2521 scm ⁻³	Distal mouth bar

Table 1 Lithofacies-associations in the Hugin formation, Volve Field (after Kieft et al. 20)	11).
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Code	Facies	Description	Thickness (t); Extent (l)	Wireline-log Attribute	Interpretation
		Parallel-laminated mudstone		GR = 41 - 308 API	
		with occasional siltstone inputs.		DT = 225 - 355 µsm ⁻¹	
	AI	Monospecific pattern of disorder	t = 30 - 425 cm 1 = 6 - 29 km	NPHI = 0.17 - 0.45 v/v	Restricted marine shale
		bivalves parallel to bedding.		RHOB = 2280 - 2820 gcm ⁻¹	
		Inter-bedded claystone and very		CP = 17 - 65 API	
		fine-grained sandstone; non-		GR = 17 - 05 AFT	
	A2	parallel and wavy lamination.	t = 10 - 725 cm = 8 - 13 km	DI = 189 - 268 µsm -	Muddy hallow bay fill
		Scarecely bivalve shells oriented		NPHI = r	
Α		parallel to bedding.		RHOB = 2280 - 2820 gcm-1	
		Fine to medium grained		GR = 18 - 46 API	
	A3	sandstone; moderately to well	t = 60 - 370 cm = 1 - 8 km	DT = 199 - 268 µsm ⁻¹	Sandy shallow bay fill
		sorted grain. Wavy bedding,		NPHI = 0.07 - 0.52 v/v	
		cross bedding, rare wave ripples.		RHOB = 1690 - 2745 gcm-1	
		Parallel-laminated mudstone		GR = 7 - 35 API	
	A4	with occasional siltstone inputs.	t = 30 - 425 cm = 6 - 29 km	DT = 175 - 230 µsm ⁻¹	Marine channel fill
		Monospecific pattern of disorder		NPHI = 0.04 - 0.15 v/v	sandstone
		bivalves parallel to bedding.		RHOB = 2280 - 2820 gcm-1	
		Upward coarsening siltstone to		GR = 18 - 80 API	
	B1	fine-grained; moderatley sorted	t = 30 - 480 cm = 1 - 2 km	DT = 168 - 291 µsm ⁻¹	Distal lower shoreface
		sandstone. Shell debris and		NPHI = 0.04 - 0.191 v/v	
		quartz granules.		RHOB = 2322 - 2723 gcm-1	
		Very fine-fine grained sandstone.		GR = 20 - 56 API	
в	B2	Moderate to well sorted; fine	t = 130 - 440 cm = 1.7 - 12 km	DT = 179 - 277 µsm ⁻¹	Proximal lower
-		grained carbonaceous laminae,		NPHI = 0.05 - 0.168 v/v	shoreface
		typically low angle cross beds.		RHOB = 2314 - 2696 gcm-1	
		Coaesening upward, cross		GR = 15 - 25 API	
	83	laminated, fine to medium	t=425-800 cm l=17-8 km	DT = 250 - 275 µsm ⁻¹	Unner shoreface
	00	grainned sandstone; consist of	12-425 000 cm 1-1.7 0 km	NPHI = 0.09 - 0.113 v/v	opper shoreface
		carbonaceous fragments.		RHOB = 2271 - 2342 gcm-1	
		Highly bioturbated siltstone to		GR = 20 - 80 API	
	C1	very fine sandstone, with beds of	t = 175 - 1010 cm = 7.2 - 19.6	DT = 230 - 260 µsm ⁻¹	Distal mouth bar
		rounded granules.	km	NPHI = 0.08 - 0.169 v/v	
с		_		RHOB = 2327 - 2521 gcm-1	
		Very fine to fine grained		GR = 12 - 38 APT	
	C2	sandstone, low angle cross	t = 290 - 775 cm = 1 - 5 km	DT = 167 - 397 µsm -	Proximal mouth bar
		bedding.		NPHI = 0.05 - 0.595 V/V	
		Fining unward coarse to fine		GR = 8 - 134 API	
		grained sandstone. Stacked fining		DT = 225 - 225 usm ⁻¹	Tidal influenced fluvial
	D1	upward beds with rare coarse	t = 740 - 820 cm = 1 - 2 km	NPHI = 0.14 - 0.46 v/v	channel fill sandstone
		grained stringers.		RHOB = 2284 - 2570 gcm-1	
D		Fining upward coarse to medium			
		grained sandstone.		GR = 9 - 34 API	fluxial 1 1 100
	D2	Carbonaceous laminae and	t = 580 cm = < 2 km	DT = 241 - 297 µsm ⁻¹	fluvial channel fill
		fragments. Sharp and cohessive		NPHI = 0.14 - 0.289 v/v	sandstone
		contact at base of bed.		кнов = 2168 - 2447 gcm-1	
		Coal and carbonaceous shale		GR = 8 - 56 API	
E	E1	Basal contact typically parallel	t = 30 - 520 cm = 6 - 19.6 km	DT = 313 - 427 µsm ⁻¹	Coal
-		although maybe undulose.		NPHI = 0.24 - 0.529 v/v	
				RHOB = 1930 - 2225 gcm-1	
		Alternating dark grey		GR = 32 - 60 API	
	50	mudstone/claystone and		DT = 358 - 415 µsm ⁻¹	Constal at 1.5
	E2	siltstone to very fine grained	t = 60 cm I = < 2 km	NPHI = 0.43 - 0.49 v/v	Coastal plain fines
		sandstone. Wavy to non-parallel		RHOB = 1994 - 2148 gcm-1	
		Mudstops with rore silteters		GR = 4 - 194 API	
		heds Parallel lamination soft	t = section tot completely:	OR - 4 - 104 API	
F	F	sediment deformation developed	nenetrated 1 = 1.7 - 36.7 km	$DI = 187 - 450 \mu \text{sm}^2$	Open marine shale
		locally on top of beds	penetrateu i = 1.7 - 50.7 Km	NPHI = 0.114 - 0.618 V/V	
		locally on top of beus.		кпов = 1750 - 2925 gcm-1	

Table 2. Input parameters applied in running the for forward stratigraphic simulations in GPMTM



					Initi	al C	onditi	ons- GP	M Inpu	ut Para	ameters	5		
		Simulation Duration	Sedimer	nt Type Pro	oportio	n (%)	Avg. Water Velocity	Avg. Sediment Velocity	Erodibility	Diffusior Coefficier	Avg. Sea	Turbidite Event Interval	Steady Flow Iteration	Sediment Movement
		(Ma– 0a) Years	Sand (Coarse)	Sand (Fine)	Silt	Clay	(m/a)	(m/a)			Interval (m)	(/years)	(/hrs)	Coefficient
	S1	0.02 - 0	25	25	25	25	0.11	0.03	0.35	0.11	30	2500	10	0.001
	S2	0.25 - 0	25	25	25	25	0.15	0.03	0.45	0.15	70	1000	15	0.012
	S3	0.5 - 0	25	25	25	25	0.11	0.02	0.55	0.11	120	1000	20	0.012
5	S4	0.7 - 0.05	25	25	25	25	0.08	0.02	0.35	0.08	100	500	25	0.0011
l ŭ	S5	1.5 - 0	15	35	30	20	0.15	0.04	0.50	0.15	80	5000	20	0.001
	S6	3.0 - 0	50	25	15	10	0.13	0.04	0.50	0.13	70	5000	30	0.0012
S	S7	3.5 - 0	50	25	15	10	0.11	0.04	0.50	0.11	70	10000	15	0.001
⊢ ĕ	S8	4.0-0	50	25	15	10	0.13	0.04	0.50	0.13	90	5000	20	0.0015
al	S9	4.5 - 0	15	45	25	15	0.1	0.02	0.45	0.1	50	10000	30	0.0012
	S10	5.0 - 0	15	45	25	15	0.12	0.02	0.45	0.12	55	10000	35	0.0013
1 8	S11	5.5 - 0	15	45	25	15	0.12	0.02	0.45	0.12	40	5000	40	0.0013
Ň	S12	6.0 - 0	15	45	25	15	0.1	0.02	0.45	0.1	60	10000	35	0.0011
5	S13	6.5 - 0	10	25	55	10	0.13	0.03	0.48	0.13	100	20000	50	0.0010
	S14	7.0 - 0	10	25	55	10	0.16	0.03	0.48	0.16	40	20000	45	0.0011
10	S15	7.5 – 0	10	25	55	10	0.13	0.03	0.48	0.13	40	20000	40	0.0012
	S16	8.0 - 0	10	25	55	10	0.15	0.03	0.48	0.15	30	10000	30	0.0010
	S17	8.5 - 0	10	25	45	20	0.14	0.02	0.45	0.14	50	50000	50	0.0010
	S18	9.0 - 0	30	30	18	22	0.13	0.02	0.52	0.13	60	25000	35	0.0012
	S19	9.5 - 0	30	40	12	18	0.12	0.02	0.55	0.12	55	25000	20	0.0013
	S20	10.0 - 0	30	42	18	10	0.11	0.01	0.40	0.11	50	5000	15	0.0011
							Sed	iment Pr	operty					
	9	Sediment Type	Diameter	Density	Initial P	orosity	Initial P	ermeability	Compacted	Porosity	Compaction	Compacted Per	meability	Erodibility
	Coa	arse Grained Sand	1.0 mm	2.70 g/cm ³	0.21 r	n³/m³	50	00 mD	0.25 m	3/m3	5000 KPa	50 mD		0.6
	Fi	ne Grained Sand	0.1 mm	2.70 g/cm ³	0.3 m	1 ³ /m ³	10	00 mD	0.15 m	3/m3	2500 KPa	5 mD		0.45
		Silt	0.01 mm	2.65 g/cm ³	0.38 r	n³/m³	5	0 mD	0.12 m	³/m³	1200 KPa	2 mD		0.3
		Clay	0.001 mm	2.65 g/cm ³	0.48 r	n³/m³	5	5 mD	0.05 m	3/m3	500 KPa	0.1 m[0.15

					Initi	al Co	onditio	ons- GP	M Inpu	it Para	meters			
		Simulation Duration	Sedimer	nt Type Pro	portior	n (%)	Avg. Water Velocity	Avg. Sediment Velocity	Erodibility	Diffusion Coefficient	Avg. Sea Level	Turbidite Event Interval	Steady Flow Iteration	Sediment Movement
		(Ma– 0a) Years	Sand (Coarse)	Sand (Fine)	Silt	Clay	(m/a)	(m/a)			Interval (m)	(/years)	(/hrs)	Coefficient
	S1	0.02 – 0	25	25	25	25	0.11	0.03	0.35	0.11	30	2500	10	0.001
	S2	0.25 – 0	25	25	25	25	0.15	0.03	0.45	0.15	70	1000	15	0.012
	S3	0.5 – 0	25	25	25	25	0.11	0.02	0.55	0.11	120	1000	20	0.012
	S4	0.7 – 0.05	25	25	25	25	0.08	0.02	0.35	0.08	100	500	25	0.0011
Ŭ	S5	1.5 – 0	15	35	30	20	0.15	0.04	0.50	0.15	80	5000	20	0.001
	S6	3.0 - 0	50	25	15	10	0.13	0.04	0.50	0.13	70	5000	30	0.0012
S S	S7	3.5 – 0	50	25	15	10	0.11	0.04	0.50	0.11	70	10000	15	0.001
. <u></u>	S8	4.0 - 0	50	25	15	10	0.13	0.04	0.50	0.13	90	5000	20	0.0015
ם	S9	4.5 – 0	15	45	25	15	0.1	0.02	0.45	0.1	50	10000	30	0.0012
	S10	5.0 – 0	15	45	25	15	0.12	0.02	0.45	0.12	55	10000	35	0.0013
Ŭ	S11	5.5 - 0	15	45	25	15	0.12	0.02	0.45	0.12	40	5000	40	0.0013
S	S12	6.0 - 0	15	45	25	15	0.1	0.02	0.45	0.1	60	10000	35	0.0011
	S13	6.5 – 0	10	25	55	10	0.13	0.03	0.48	0.13	100	20000	50	0.0010
	S14	7.0 - 0	10	25	55	10	0.16	0.03	0.48	0.16	40	20000	45	0.0011
	S15	7.5 – 0	10	25	55	10	0.13	0.03	0.48	0.13	40	20000	40	0.0012
	S16	8.0-0	10	25	55	10	0.15	0.03	0.48	0.15	30	10000	30	0.0010
	S17	8.5 – 0	10	25	45	20	0.14	0.02	0.45	0.14	50	50000	50	0.0010
	S18	9.0 - 0	30	30	18	22	0.13	0.02	0.52	0.13	60	25000	35	0.0012
	S19	9.5 – 0	30	40	12	18	0.12	0.02	0.55	0.12	55	25000	20	0.0013
	S20	10.0 - 0	30	42	18	10	0.11	0.01	0.40	0.11	50	5000	15	0.0011
							Sed	iment Pr	operty					
	S	ediment Type	Diameter	Density	Initial P	orosity	Initial Pe	ermeability	Compacted	Porosity (Compaction	Compacted Per	meability	Erodibility
	Coa	rse Grained Sand	1.0 mm	2.70 g/cm ³	0.21 n	n³/m³	50	0 mD	0.25 m ³	³/m³	5000 KPa	50 mD		0.6
	Fir	ne Grained Sand	0.1 mm	2.70 g/cm ³	0.3 m	1 ³ /m ³	10	0 mD	0.15 m ³	³ /m ³	2500 KPa	5 mD		0.45
		Silt	0.01 mm	2.65 g/cm ³	0.38 n	n³/m³	50) mD	0.12 m	³ /m ³	1200 KPa	2 mD		0.3
		Clay	0.001 mm	2.65 g/cm ³	0.48 n	n ³ /m ³	5	mD	0.05 m	³ /m ³	500 KPa	0.1 mD)	0.15

Table 3. Lithofacies classification in the forward stratigraphic model; showing the command used in the property calculator tool in PetrelTM.

		Lithofacies Classification
Facies Code	Lithofacies	Command Used in Petrel's Property Calculator
0	Marine Shale	If(Sand_fine>=0.19 And Sand_fine<=0.21 Or Silt>=0.19 And Silt<=0.2 Or Clay>=0.2 And Clay<=0.21 Or Depth_of_deposition>=-82 And Depth_of_deposition<=-78)
1	Muddy Shallow Bay Fill	If(Sand_fine>=0.36 And Sand_fine<=0.38 Or Silt>=0.18 And Silt<=0.2 Or Clay>0.18 And Clay<=0.19 Or Depth_of_deposition>=-30 And Depth_of_deposition<=-20)
2	Sandy Shallow Bay Fill	If(Sand_coarse>=0.65 And Sand_coarse<=0.73 Or Sand_fine>=0.18 And Sand_fine<=0.22 Or Silt>=0.18 And Silt<=0.2 Or Clay>=0.17 And Clay<=0.18 Or Depth_of_deposition>=-3 And Depth_of_deposition<=0)
3	Channel Fill Sandstone	If(Sand_coarse>=0.5 And Sand_coarse<=0.68 Or Sand_fine>=0.23 And Sand_fine<=0.25 Or Silt>=0.17 And Silt<=0.18 Or Depth_of_deposition>=0 And Depth_of_deposition<=2)
4	Lower Shoreface Units	If(Sand_coarse>=0.19 And Sand_coarse<=0.31 Or Sand_fine>=0.19 And Sand_fine<=0.24 Or Silt>=0.4 And Silt<=0.48 Or Clay>=0.19 And Clay<=0.31 Or Depth_of_deposition>=-83 And Depth_of_deposition<=50)
5	Middle Shoreface Units	If(Sand_coarse>=0.32 And Sand_coarse<=0.53 Or Sand_fine>=0.25 And Sand_fine<=0.32 Or Silt>=0.26 And Silt<=0.32 Or Clay>=0.19 And Clay<=0.21 Or Depth_of_deposition>=-38 And Depth_of_deposition<=-12)
6	Upper Shoreface Units	lf(Sand_coarse>=0.53 And Sand_coarse<=0.72 Or Sand_fine>=0.28 And Sand_fine<=0.33 Or Silt>=0.16 And Silt<=0.21 Or Depth_of_deposition>=-10 And Depth_of_deposition<=6)
7	Distal Mouth Bar Units	If(Sand_fine>=0.23 And Sand_fine<=0.27 Or Silt>=0.38 And Silt<=0.43 Or Clay>=0.19 And Clay<=0.21 Or Depth_of_deposition>=-95 And Depth_of_deposition<=-80)
8	Proximal Mouth Bar Units	If(Sand_coarse>=0.53 And Sand_coarse<=0.71 Or Sand_fine>=0.27 And Sand_fine<=0.32 Or Silt>=0.16 And Silt<=0.21 Or Clay>=0.06 And Clay<=0.07 Or Depth_of_deposition>=-30 And Depth_of_deposition<=-27)
9	Tide Influenced Sandstones	If(Sand_coarse>=0.53 And Sand_coarse<=0.71 Or Sand_fine>=0.26 And Sand_fine<=0.31 Or Silt>=0.35 And Silt<=0.41 Or Depth_of_deposition>=-5 And Depth_of_deposition<=1)
10	Fluvial Channel Sandstones	If(Sand_coarse>=0.54 And Sand_coarse<=0.56 Or Sand_fine>=0.27 And Sand_fine<=0.29 Or Silt>=0.19 And Silt<=0.21 Or Depth_of_deposition>=-2 And Depth_of_deposition<=2)
11	Coal	Estimated as background attribute
12	Coastal plain fines	If(Silt>=0.31 And Silt<=0.43 Or Clay>=0.31 And Clay<=0.35 Or Depositional_depth>=-100 And Depositional_depth<=-40)
13	Marine Mudstone	If(Sand_fine>=0.36 And Sand_fine<=0.38 Or Silt>=0.4 And Silt<=0.52 Or Clay>=0.45 And Clay<=0.78 Or Depth_of_deposition>=-105 And Depth_of_deposition<=-90)

		Lithofacies Classification
Facies Code	Lithofacies	Command Used in Petrel's Property Calculator
0	Marine Shale	If(Sand_fine>=0.19 And Sand_fine<=0.21 Or Silt>=0.19 And Silt<=0.2 Or Clay>=0.2 And Clay<=0.21 Or Depth_of_deposition>=-82 And Depth_of_deposition<=-78)
1	Muddy Shallow Bay Fill	If(Sand_fine>=0.36 And Sand_fine<=0.38 Or Silt>=0.18 And Silt<=0.2 Or Clay>0.18 And Clay<=0.19 Or Depth_of_deposition>=-30 And Depth_of_deposition<=-20)
2	Sandy Shallow Bay Fill	If(Sand_coarse>=0.65 And Sand_coarse<=0.73 Or Sand_fine>=0.18 And Sand_fine<=0.22 Or Silt>=0.18 And Silt<=0.2 Or Clay>=0.17 And Clay<=0.18 Or Depth_of_deposition>=-3 And Depth_of_deposition<=0)
3	Channel Fill Sandstone	If(Sand_coarse>=0.5 And Sand_coarse<=0.68 Or Sand_fine>=0.23 And Sand_fine<=0.25 Or Silt>=0.17 And Silt<=0.18 Or Depth_of_deposition>=0 And Depth_of_deposition<=2)
4	Lower Shoreface Units	If(Sand_coarse>=0.19 And Sand_coarse<=0.31 Or Sand_fine>=0.19 And Sand_fine<=0.24 Or Silt>=0.4 And Silt<=0.48 Or Clay>=0.19 And Clay<=0.31 Or Depth_of_deposition>=-83 And Depth_of_deposition<=50)
5	Middle Shoreface Units	If(Sand_coarse>=0.32 And Sand_coarse<=0.53 Or Sand_fine>=0.25 And Sand_fine<=0.32 Or Silt>=0.26 And Silt<=0.32 Or Clay>=0.19 And Clay<=0.21 Or Depth_of_deposition>=-38 And Depth_of_deposition<=-12)
6	Upper Shoreface Units	If(Sand_coarse>=0.53 And Sand_coarse<=0.72 Or Sand_fine>=0.28 And Sand_fine<=0.33 Or Silt>=0.16 And Silt<=0.21 Or Depth_of_deposition>=-10 And Depth_of_deposition<=6)
7	Distal Mouth Bar Units	If(Sand_fine>=0.23 And Sand_fine<=0.27 Or Silt>=0.38 And Silt<=0.43 Or Clay>=0.19 And Clay<=0.21 Or Depth_of_deposition>=-95 And Depth_of_deposition<=-80)
8	Proximal Mouth Bar Units	If(Sand_coarse>=0.53 And Sand_coarse<=0.71 Or Sand_fine>=0.27 And Sand_fine<=0.32 Or Silt>=0.16 And Silt<=0.21 Or Clay>=0.06 And Clay<=0.07 Or Depth_of_deposition>=-30 And Depth_of_deposition<=-27)
9	Tide Influenced Sandstones	If(Sand_coarse>=0.53 And Sand_coarse<=0.71 Or Sand_fine>=0.26 And Sand_fine<=0.31 Or Silt>=0.35 And Silt<=0.41 Or Depth_of_deposition>=-5 And Depth_of_deposition<=1)
10	Fluvial Channel Sandstones	If(Sand_coarse>=0.54 And Sand_coarse<=0.56 Or Sand_fine>=0.27 And Sand_fine<=0.29 Or Silt>=0.19 And Silt<=0.21 Or Depth_of_deposition>=-2 And Depth_of_deposition<=2)
11	Coal	Estimated as background attribute
12	Coastal plain fines	If(Silt>=0.31 And Silt<=0.43 Or Clay>=0.31 And Clay<=0.35 Or Depositional_depth>=-100 And Depositional_depth<=-40)
13	Marine Mudstone	If(Sand_fine>=0.36 And Sand_fine<=0.38 Or Silt>=0.4 And Silt<=0.52 Or Clay>=0.45 And Clay<=0.78 Or Depth_of_deposition>=-105 And Depth_of_deposition<=-90)

Code	Lithofacies	Average	Density	Estimated	KLOGH
		NPHI	Porosity	Porosity	(mD)
0	Marine Shale	0.17 - 0.45	0.1	0.08 - 0.11	10.02 - 16.1
1	Muddy Shallow Bay Fill	0.17 - 0.42	0.1	0.08 - 0.13	23.85 - 102.3
2	Sandy Shallow Bay Fill	0.07 - 0.52	0.25	0.16 - 0.25	100.0 - 398. 7
3	Channel Fill Sandstone	0.04 - 0.15	0.30	0.18 - 0.22	400.01 - 889.7
4	Distal Lower Shoreface	0.04 - 0.19	0.29	0.1 - 0.23	120.5 - 170.3
5	Proximal Shoreface	0.05 - 0.17	0.31	0.17 - 0.24	80.2 - 412.5
6	Upper Shoreface Units	0.09 - 0.11	0.28	0.21 - 0.26	650.2 - 1023.7
7	Distal Mouth Bar Units	0.08 - 0.17	0.27	0.09 - 0.17	170.5 - 223.1
8	Proximal Mouth Bar	0.05 - 0.59	0.12	0.19 - 0.21	130.5 - 314.3
9	Tide Influenced SS	0.14 - 0.46	0.26	0.15 - 0.20	220.0 - 512.6
10	Fluvial Sandstones	0.14 - 0.29	0.21	0.19 - 0.21	180.5 - 691.8
11	Coal	0.24 - 0.53	0.05	0.001	0.001
12	Coastal Plain Fines	0.43 - 0.49	0.06	0.04 - 0.12	5.2 - 34.6
13	Marine Mudstone	0.16 - 0.42	0.1	0.08 - 0.10	6.0 - 15.2

Table 4. Porosity and Permeability estimate in identified estimates of lithofacies packages in the model area.

Code	Lithofacies	Avg. NPHI	Density Porosity	Estimated Porosity	KLOGH (mD)
0	Marine Shale	0.17 - 0.45	0.1	0.08 - 0.11	10.02 - 16.1
1	Muddy Shallow Bay Fill	0.17 - 0.42	0.1	0.08 - 0.13	23.85 - 102.3
2	Sandy Shallow Bay Fill	0.07 - 0.52	0.25	0.16 - 0.25	100.0 - 398.7
3	Channel Fill Sandstone	0.04 - 0.15	0.3	0.18 - 0.22	400.01 - 889.7
4	Distal Lower Shoreface	0.04 - 0.19	0.29	0.1 - 0.23	120.5 - 170.3
5	Proximal Shoreface	0.05 - 0.17	0.31	0.17 - 0.24	80.2 - 412.5
6	Upper Shoreface	0.09 - 0.11	0.28	0.21 - 0.26	650.2 - 1023.7
7	Distal Mouth Bar	0.08 - 0.17	0.27	0.09 - 0.17	170.5 - 223.1
8	Proximal Mouth Bar	0.05 - 0.59	0.12	0.19 - 0.21	130.5 - 314.3
9	Tidal Influenced Sandstone	0.14 - 0.46	0.26	0.15 - 0.20	220.0 - 512.6
10	Fluvial Sandstones	0.14 - 0.29	0.21	0.19 - 0.21	180.5 - 691.8
11	Coal	0.24 - 0.53	0.05	0.001	0.001
12	Coastal Plain Fines	0.43 - 0.49	0.06	0.04 - 0.12	5.2 - 34.6
13	Marine Mudstone	0.16 - 0.42	0.1	0.08 - 0.10	6.0 - 15.2

Table 5. <u>ComparisonA comparison</u> of a) porosity, and b) permeability estimates <u>from selected intervals</u> in <u>the</u> original <u>petrophysical modelporosity/permeability models</u> and forward modeling-based porosity and permeability models.

	a. \	/alidation W	/ell Position	1			
		Porosity	: GPM-Base	d Model		Porosit	y: Original Model
			Depth (m)				
Models	5 m	10 m	15 m	25 m	35 m	Depth (m)	Average Porosity
R14	0.22	0.24	0.16	0.22	0.16	5	0.2
R20	0.16	0.19	0.26	0.18	0.15	10	0.25
R26	0.18	0.17	0.23	0.16	0.19	15	0.27
R36	0.22	0.21	0.19	0.22	0.21	25	0.16
R45	0.25	0.2	0.23	0.22	0.15	35	0.13
R49	0.21	0.17	0.22	0.17	0.18		
	١	/alidation W	ell Position	2			
		Porosity	: GPM-Base	d Model		Porosit	y: Original Model
			Depth (m)				
Models	5 m	10 m	15 m	25 m	35 m	Depth (m)	Average Porosity
R14	0.17	0.16	0.24	0.15	0.25	5	0.17
R20	0.21	0.22	0.2	0.21	0.23	10	0.21
R26	0.21	0.2	0.21	0.25	0.24	15	0.21
R36	0.2	0.22	0.21	0.21	0.19	25	0.17
R45	0.22	0.19	0.2	0.19	0.21	35	0.19
R49	0.26	0.24	0.23	0.16	0.21		
	b. V	/alidation W	/ell Position	1			
	b. V Pe	/alidation W rmeability	/ell Position Z (mD): GPN	1 A-Based Mo	del	Permeabili	ty Z: Original Model
	b. V Pe	/alidation W rmeability_	/ell Position Z (mD): GPN Depth (m)	1 1-Based Mo	del	Permeabili	ty_Z: Original Model
Models	b. V Pe 5 m	/alidation W rmeability_i 10 m	/ell Position Z (mD): GPN Depth (m) 15 m	1 1-Based Mo 25 m	del 35 m	Permeabili Depth (m)	ty_Z: Original Model Average Perm_Z
Models R14	b. V Pe 5 m 163.95	/alidation W rmeability_i 10 m 312.38	/ell Position Z (mD): GPN Depth (m) 15 m 69.84	1 A-Based Mo 25 m 310.16	del 35 m 508.2	Permeabili Depth (m) 5	ty_Z: Original Model Average Perm_Z 352.74
Models R14 R20	b. V Pe 5 m 163.95 290.84	/alidation W rmeability_2 10 m 312.38 315.09	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66	1 1-Based Mo 25 m 310.16 273.04	del 35 m 508.2 200.63	Permeabili Depth (m) 5 10	ty_Z: Original Model Average Perm_Z 352.74 312.38
Models R14 R20 R26	b. V Pe 5 m 163.95 290.84 375.92	/alidation W rmeability_i 10 m 312.38 315.09 203.81	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23	1 A-Based Mo 25 m 310.16 273.04 189.92	del 35 m 508.2 200.63 348.12	Permeabili Depth (m) 5 10 15	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08
Models R14 R20 R26 R36	b. V Pe 5 m 163.95 290.84 375.92 418.03	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9	1 A-Based Mo 25 m 310.16 273.04 189.92 168.9	del 35 m 508.2 200.63 348.12 370.56	Permeabili Depth (m) 5 10 15 25	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76
Models R14 R20 R26 R36 R45	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6	/alidation W rmeability_2 10 m 312.38 315.09 203.81 203.27 412.67	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66	1 -Based Mo 25 m 310.16 273.04 189.92 168.9 156.71	del 35 m 508.2 200.63 348.12 370.56 305.92	Permeabili Depth (m) 5 10 15 25 35	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2
Models R14 R20 R26 R36 R45 R49	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89	Alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77	1 -Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18	Permeabili Depth (m) 5 10 15 25 35	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2
Models R14 R20 R26 R36 R45 R49	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position	1 -Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18	Permeabili Depth (m) 5 10 15 25 35	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2
Models R14 R20 R26 R36 R45 R49	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89 V Pe	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN	1 -Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del	Permeabili Depth (m) 5 10 15 25 35 35 Permeabili	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model
Models R14 R20 R26 R36 R45 R49	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89 V Pe	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN Depth (m)	1 -Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 -Based Mo	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del	Permeabili Depth (m) 5 10 15 25 35	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model
Models R14 R20 R26 R36 R45 R49 Models	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 3370.89 V Pe 5 m	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i 10 m	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN Depth (m) 15 m	1 A-Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 A-Based Mo 25 m	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del 35 m	Permeabili Depth (m) 5 10 15 25 35 . Permeabili Depth (m)	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model Average Perm_Z
Models R14 R20 R26 R36 R45 R49 Models R14	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89 V Pe 5 m 320.34	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i 10 m 336.22	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN Depth (m) 15 m 151.08	1 A-Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 A-Based Mo 25 m 464.22	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del 35 m 132.98	Permeabili Depth (m) 5 10 15 25 35 35 Permeabili Depth (m)	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model Average Perm_Z 6.6
Models R14 R20 R26 R36 R45 R49 Models R14 R20	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89 V Pe 5 m 320.34 122.66	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i 10 m 336.22 209.15	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN Depth (m) 15 m 151.08 161.3	1 A-Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 A-Based Mo 25 m 464.22 230.58	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del 35 m 132.98 208.48	Permeabili Depth (m) 5 10 15 25 35 Permeabili Permeabili Depth (m) 5 10	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model Average Perm_Z 6.6 883.6
Models R14 R20 R26 R36 R45 R49 Models R14 R20 R26	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89 V Pe 5 m 320.34 122.66 151.48	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i 10 m 336.22 209.15 710.07	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN Depth (m) 15 m 151.08 161.3 175.09	1 A-Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 A-Based Mo 25 m 464.22 230.58 384.49	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del 35 m 132.98 208.48 169.48	Permeabili Depth (m) 5 10 15 25 35 Permeabili Permeabili Depth (m) 5 10 15	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model Average Perm_Z 6.6 883.6 30.3
Models R14 R20 R26 R36 R45 R49 Models R14 R20 R26 R34 R36 R45 R49	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89 V Pe 5 m 320.34 122.66 151.48 184.74	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i 10 m 336.22 209.15 710.07 344.99	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN Depth (m) 15 m 151.08 161.3 175.09 157.08	1 A-Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 A-Based Mo 25 m 464.22 230.58 384.49 420.15	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del 35 m 132.98 208.48 169.48 136.14	Permeabili Depth (m) 5 10 15 25 35 Permeabili Permeabili Depth (m) 5 10 15 25 10 15 25	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model Average Perm_Z 6.6 883.6 30.3 496.99
Models R14 R20 R26 R36 R45 R49 Models R14 R20 R26 R36 R36 R45	b. V Pe 5 m 163.95 290.84 375.92 418.03 337.6 370.89 V Pe 5 m 320.34 122.66 151.48 184.74 91.44	/alidation W rmeability_i 10 m 312.38 315.09 203.81 203.27 412.67 129.33 /alidation W rmeability_i 10 m 336.22 209.15 710.07 344.99 361.04	/ell Position Z (mD): GPN Depth (m) 15 m 69.84 105.66 166.23 190.9 199.66 291.77 /ell Position Z (mD): GPN Depth (m) 15 m 151.08 161.3 175.09 157.08 77.17	1 -Based Mo 25 m 310.16 273.04 189.92 168.9 156.71 175.53 2 	del 35 m 508.2 200.63 348.12 370.56 305.92 551.18 del 35 m 132.98 208.48 169.48 136.14 134.56	Permeabili Depth (m) 5 10 15 25 35 Permeabili Permeabili Depth (m) 5 10 15 25 10 15 25 35	ty_Z: Original Model Average Perm_Z 352.74 312.38 201.08 199.76 508.2 ty_Z: Original Model Average Perm_Z 6.6 883.6 30.3 496.99 156.6

		a. Validation	Well Position 1					
	Depth (m)							
	5 m	10 m	15 m	25 m	35 m			
Models	Measured Porosity							
Original Model	0.2	0.25	0.27	0.16	0.13			
R14	0.22	0.24	0.16	0.22	0.16			
R20	0.16	0.19	0.26	0.18	0.15			
R26	0.18	0.17	0.23	0.16	0.19			
R36	0.22	0.21	0.19	0.22	0.21			
R45	0.25	0.2	0.23	0.22	0.15			
R49	0.21	0.17	0.22	0.17	0.18			
		Validation V	Vell Position 2					
	Depth (m)							
	5 m	10 m	15 m	25 m	35 m			
Models	Measured Porosity							
Original Model	0.17	0.21	0.21	0.17	0.19			
R14	0.17	0.16	0.24	0.15	0.25			
R20	0.21	0.22	0.2	0.21	0.23			
R26	0.21	0.2	0.21	0.25	0.24			
R36	0.2	0.22	0.21	0.21	0.19			
R45	0.22	0.19	0.2	0.19	0.21			
R49	0.26	0.24	0.23	0.16	0.21			
		1						

b. Validation Well Position 1								
	Depth (m)							
	5 m	10 m	15 m	25 m	35 m			
Models	Measured Permeability_Z (mD)							
Original Model	352.74	312.38	201.08	199.76	508.2			
R14	163.95	312.38	69.84	310.16	508.2			
R20	290.84	315.09	105.66	273.04	200.63			
R26	375.92	203.81	166.23	189.92	348.12			
R36	418.03	203.27	190.9	168.9	370.56			
R45	337.6	412.67	199.66	156.71	305.92			
R49	370.89	129.33	291.77	175.53	551.18			

Validation Well Position 2

	Depth (m)							
	5 m	10 m	15 m	25 m	35 m			
Models	Measured Permeability_Z (mD)							
Original Model	6.6	883.6	30.3	496.99	156.6			
R14	320.34	336.22	151.08	464.22	132.98			
R20	122.66	209.15	161.3	230.58	208.48			
R26	151.48	710.07	175.09	384.49	169.48			
R36	184.74	344.99	157.08	420.15	136.14			
R45	91.44	361.04	77.17	382.85	134.56			
R49	134.01	721.73	137.42	636.48	290.06			
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