

Drill bits optimization using offset wells analysis and ROP modelling

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Article Info

Received: 9 Nov. 2020

Revised: 30 Mar. 2021

Accepted: 30 Apr. 2021

Abstract

Drill Bits optimization is the key factor to improve the overall drilling efficiency and therefore, reduce the cost and the non-productive time. For this, analysis of the offset field data, modelling before drilling and optimization the drilling parameters are highly required to enhance the drilling efficiency. In this paper, 8 ½" hole of three drilled wells is analysed for the rate of penetration (ROP), drilling cost per meter and the drilling duration of each drill bit. The analysis has shown that the variation in drilling efficiency was due to some factor such as, the type and features of the used drill bit, the bottom hole assembly type and the drilling parameters. ROP model (multiple regression equation) is developed by using a reference well's drilling parameters. The drilling parameters that have been used to develop the ROP model are weight on bit (WOB), revolution per minute (RPM) of the drill string, the drilling torque (TRQ), jet impact force (JIF) of the drill bit and the unconfined compressive strength (UCS) of the formation. Parametric sensitivity study has been carried out using this model to interpret the drilling parameters that have the greatest impact on ROP. Finally, drill bits' with the highest performance are recommended to be used based on analysis of the offset wells and ROP model "multiple regression equation" has been approached to predict the ROP values before drilling to be used in optimization of drill bits' performance in the forthcoming drilled wells.

Keywords

ROP model; multiple regression; analysis; Drill bit optimization

Introduction

The objective of the drilling operation is to safely execute the drilling plan at optimum performance with the minimum drilling cost and time. The operators are continuously looking for techniques that maximize operation efficiency and minimize the time to drill each section^(1; 2).

Optimizing the drilling efficiency could only be achieved through three distinct stages; planning and modeling before drilling, monitoring during drilling and analysis the drill bits performance after drilling⁽³⁾.

Drilling in deeper and harsher conditions well requires more advanced drilling technology and equipment. Therefore, the efficiency of drill bits is increased by improving their quality and this will allow a further increase in the rate of penetration⁽⁴⁾.

The challenges in drilling 8 ½" hole section are a lot; the high unconfined compressive strength of conglomerate formation, the torque fluctuation, the high vibrations that were being generated when PDC

bits had been used and the high bit wear in a short time. In addition to the previously mentioned challenges, the operator was suffering from the high cost per every drilled meter and the time consumed in drilling that section.

The target in Sidri area is the conglomerate formation. The thickness of Conglomerate ranges from 900 m to 1500 m and with the same lithology. Conglomerate formation has high-unconfined compressive strength (UCS) that varies between 15,000 – 35,000 psi. Conglomerate formation is composed mainly of 95 % granitic fragment, which had been formed from basement erosion.

One thousand meter of 8 ½" hole was drilled using more than 7 drill bits. Most drilled wells required more than 20 days to drill 8 ½" section. In the three studied wells, the drilling performance was high. The three wells have been analysed and ROP model has been developed.

The objective of this paper is to present analysis about the drill bits' performance of three drilled

wells in Sinai Oil Field, establish ROP prediction mathematical model by using a reference well (well with the highest drill bit performance), then test the accuracy of the ROP model by applying the model on the nearby wells and use the ROP model to predict the most effective drilling parameters on the drilling performance.

The developed ROP model is multiple regression, which characterizes an observation factor by several variables, taking into consideration changes of several properties simultaneously. In this paper, the observation factor (Y) is the rate of penetration. Relevant drilling factors make up the regression variables (X1-5) (5). The ROP model built a relationship between the rate of penetration (ROP) and WOB, RPM, TRQ, JIF and UCS.

Methodology

Analysis of drill bits' performance

The performance of a drill bits could be analysed using ROP values, the length of drilled interval and the cost per meter per every drill bit. Equation (1) is applied on every drill bit run and by comparing the cost per meter values among all the drill bits, it would be easily to define the drill bits that achieved high performance and those achieved poor performance.

$$C = \frac{B+(T+t)*R}{L} \quad (1)$$

C: cost per meter (\$/m) R: Rig cost per hour (\$/hr)

B: refers to bit cost (\$) T: Trip time (hr)

L: refers to the drilled interval (m)

t: rotating time on bottom (hr)

ROP Modelling

Multiple regression workflow has been carried out on 8 1/2" hole section of the reference well (well-X), the parameters used in multiple regression analysis are in the form of WOB, RPM, torque, JIF and the UCS of the drilled formation, together with the observation factor ROP.

The modified jet impact force given in Eq. (2) was therefore chosen to replace the Flow rate and the standpipe pressure (SPP) in the model (6).

$$JIF = \frac{Qx(\sqrt{\rho x P_{bit}})}{58} \quad (2)$$

JIF: Jet impact force (psi)

Q: flow rate (GPM)

ρ : drilling fluid density (PPG)

Pbit: pressure drop across the drill bit

The unconfined compressive strength of formation was included in the equation due to the great impact of the formation strength on the drill bits performance, UCS can

be calculated using interval transit time (Δt) recorded by wire line sonic log (7), given in Eq. (3).

$$UCS(MPA) = \frac{\left(\frac{7682}{\Delta t}\right)^2}{145} \quad (3)$$

UCS: unconfined compressive strength, MPa

Δt : interval transit time

The (Y) range represents the ROP, while(X) range is the drilling variables. The depth is considered only as a reference and is not involved in the analysis. The coefficients are then provided by the analysis. The intercept value is represented by the initial value of coefficients (b0). The other coefficients (b1, b2, b3, b4 & b5) are then multiplied according to their order with the multiple regression equation's variables (X1, X2, X3, X4 & X5). The multiple regression equation that represents the ROP model could be formed as Eq. (4):

$$Y = b_0 + b_1X_1 + b_2X_2 + b_3X_3 + b_4X_4 + b_5X_5 \quad (4)$$

This equation could be written in terms of ROP and drilling parameters as:

$$\text{Modelled ROP} = b_0 + b_1\text{WOB} + b_2\text{RPM} + b_3\text{Torque} + b_4\text{JIF} + b_5\text{UCS} \quad (5)$$

Implementation of the multiple regression workflow in Microsoft Excel software

The multiple regression procedure shown in Figure (1) is applied on the reference well (Well-X). In Microsoft Excel, multiple regression data analysis is carried out by entering Input-Y-range from column (J) which represents the actual ROP values and X-range from columns (E, F, G, H, and I), the multiple regression workflow has been carried out providing the multiple regression coefficients listed in cells Y20 to Y25, shown in Figure (2) "yellow colored". Equation (5) is implemented for row no.4 of drilling data. The equation calculates the modelled ROP for each row by changing the row reference number, shown in Figure (3).

$$K = (\$Y\$20) + (\$Y\$21 * E5) + (\$Y\$22 * F5) + (\$Y\$23 * G5) + (\$Y\$24 * H5) + (\$Y\$25 * I5) \quad (6)$$

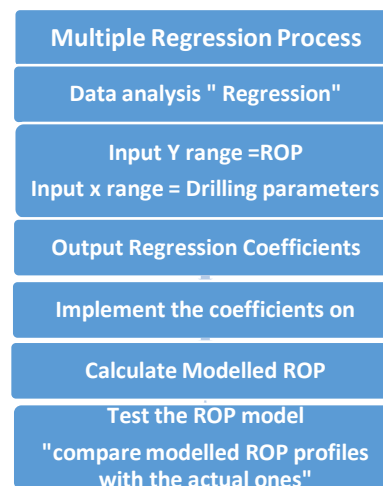


Figure 1 Multiple Regression Procedures flowchart

DEPTH m	FLOW RATE GPM	Pressure Psi	WOB KLBS	RPM	TRQ K Lbs.ft	JIF	UCS psi	ROP MPH
1910	460	283.009	32	81	9	395	18675.96	9.30
1911	460	283.009	38	81	8	395	18679.21	8.28
1912	460	283.009	36	81	8	395	18682.46	9.54
1913	460	283.009	38	90	8	395	18685.71	10.50
1914	460	283.009	37	93	9	395	18688.96	5.04
1915	460	283.009	38	93	8	395	18692.21	4.46
1916	460	283.009	34	93	10	395	18695.46	6.87
1917	460	283.009	36	93	9	395	18698.71	5.64

Figure 2 multiple regression data analysis (Microsoft Excel)

SUMMARY OUTPUT					
Regression Statistics					
Multiple R	0.9802904				
R Square	0.96099269				
Adjusted R Square	0.960823058				
Standard Error	0.393540522				
Observations	1388				
ANOVA					
	df	SS	MS	F	Significance F
Regression	5	5269.747012	1053.95	6805.19929	0
Residual	1382	214.0360644	0.15487		
Total	1387	5483.783076			
Coefficients					
	Standard Error	t Stat	P-value	Lower 95%	Upper 95%
Intercept	12.14776608	0.234962189	38.62	6.341E-222	11.91257
X Variable 1	0.889151953	0.003283358	24.987	5.158E-114	0.0756
X Variable 2	0.066000126	0.000984296	66.9496	0	0.063967
X Variable 3	-0.081773106	0.006085239	-13.2024	1.4409E-377	-0.09228
X Variable 4	0.000870105	0.000560737	0.53848	0.59033448	-0.0008
X Variable 5	-0.000688606	6.10471E-06	-112.16	0	-0.0007

Figure 3 multiple regression equation is applied in Microsoft Excel

Testing the developed ROP model on two nearby wells

After applying Eq. (6) on the reference well (Well-X), the multiple regression equation is tested on two nearby wells (Well-Y & Well-Z) with the same drilling criteria; the bottom hole assemblies and drill bits and similar geological structure. Using the coefficients resulting from multiple regression workflow and with applying the equation on the two wells, predicted ROP values have been computed.

The results (ROP profile) of implementing the multiple regression equation on the two wells are compared with the actual ROP profile to test the reliability of the multiple regression equation. If the predicted ROP profile is similar and close in values to the actual one, the ROP model proves that it could be applicable on that area.

ROP model and parametric sensitivity

A parametric sensitivity analysis is carried out in this study to find out which drilling parameters have a greater impact on the ROP. The drilling parameters that were used in the ROP model (WOB, RPM, torque, JIF & UCS) are increased by 10 % one by one. The new predicted ROP values are compared with the original predicted ROP. By this way, the most effective drilling parameters could be determined, and the optimum drilling parameters can be applied in drilling the next wells. The final step in the parametric sensitivity is to apply the most effective drilling parameters with the increase percentage on the ROP model. The predicted ROP values have been the highest.

Results and discussions

Analysis of the three wells

The three studied wells have been analyzed using bit run cost equation for each drill bit. In addition, ROP values and the length of the drilled interval per each drill bit are also taken into consideration in the analysis study, shown in table 1.

It's clearly obvious the hybrid bit (KM524) had achieved the highest performance in the upper part of 8 ½” hole and the sting blade PDC bit had achieved good performance in the lower part of 8 ½” hole where the compressive strength is higher.

Table 1 analysis of Drill bits performance in wells (X, Y & Z)

	TYPE	IADC	DEPTH IN (m)	DEPTH OUT (m)	AVG. ROP mph	CPM (\$/m)
Well-X	KM524	HYBRID	1910	2491	6.6	312.9
	MSIZI813UQEBPX	STING BLADE	2491	2644	4.5	661.8
	MSZI716NEBPX-J	AXE BLADE	2644	2819	7.6	518.0
	KM524	HYBRID	2819	3109	3.8	578.0
	KM524	HYBRID	3109	3297	2.8	816.1
Well-Y	MSZI716MEBPX-J	PDC	2418	2655	5.1	483.3
	KM524	HYBRID	2655	3105	4.6	440.2
	KM524	HYBRID	3105	3278	3.4	767.9
	KM524	HYBRID	3278	3355	2.4	1347.7
Well-Z	KYM524	PDC	2113	2594	5.1	398.6
	KYM524	PDC	2594	2860	2.8	736.8
	MSZI716MEBPX	PDC	2860	3036	3.4	692.7

ROP Modelling

Well-X is modelled by multiple regression technique. The computed correlation coefficients are provided in table 2. The extracted Coefficients from multiple regression ROP model are reliable. It's noticed that the b1 (WOB), b2 (RPM) & b4 (JIF) are of positive values, on the other hand, b3 (TRQ) and b5 (UCS) are of negative values which is reliable. By implementing

the multiple regression coefficients provided from multiple regression workflow applied on the original well (well-X), a new ROP profile could be predicted and when the new computed ROP profile is compared with the actual ROP profile, the result shows a good match, as shown in Figure (4). Wells (Y & Z) have the same drilling conditions (drill bits, bottom hole assemblies and geological structure) of the reference well (well-X), so the coefficients extracted from well-X ROP model could be applied on these wells to test its reliability and accuracy. The results show that the well-X based model when applied on well-Y predicts excellent match. The predicted ROP appears to match obviously with the actual ROP, although it deviates in some sections. In addition to the clear match between the predicted and actual ROP, the average predicted and actual ROP values using the model are close to each other's (average predicted ROP is 5.03 mph and the average actual ROP is 5.01 mph), shown in Figure (5). In well-Z, the predicted ROP profile appears to correlate very well with the actual one. The average actual ROP is 5.47 mph and the average predicted ROP is 5.57 mph, shown in Figure (6).

Table 2 Calculated regression coefficients based on reference well (X)

Coefficient	Coefficient Value
Intercept, b_0	12.14776608
b_1 (WOC)	0.089151953
b_2 (RPM)	0.066000126
b_3 (TRQ)	-0.081773106
b_4 (JIF)	0.000870105
b_5 (UCS)	-0.000688606

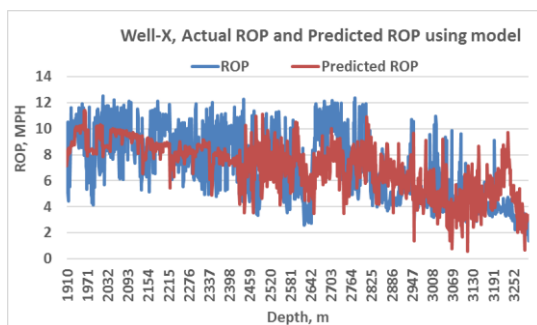


Figure 4 multiple regression method applied on well-X

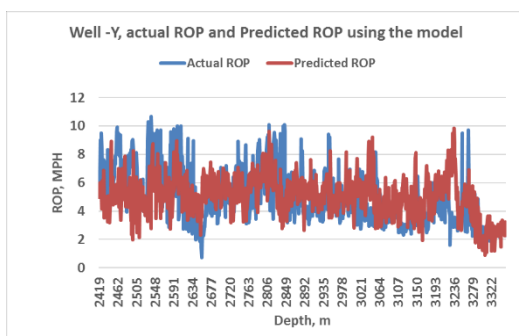


Figure 5 multiple regression method applied on well-Y using coefficients from well-X

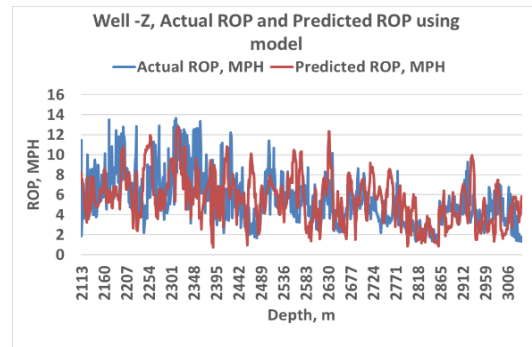


Figure 6 multiple regression method applied on well-Z using coefficients from well-X

Optimization Drilling Parameters

Determining the most effective drilling parameters on the drilling performance is easily carried out by the developed ROP model. As a part of

modelling, predicting which drilling parameters have a noticeable effect on the penetration rate is the utmost target of any ROP model.

The controllable drilling parameters that have been used in the ROP model (i.e. WOB, RPM & JIF) are increased by 10 % one by one, and then the ROP model has been carried out on the new drilling parameters one by one. The new predicted ROP values after increasing the drilling parameters are compared with the original predicted ROP values. The results of increasing the drilling parameters are plotted against the modelled ROP, as shown in Figure (7) which shows the effect of increasing the drilling parameters on ROP values for wells (X, Y & Z).

By finding out the drilling parameters that have the highest effect on the penetration rate, the new averages ROP are calculated, and the saved time and cost could be defined well. From Figure (7), we can conclude that the RPM has the highest impact on the penetration rate then the WOB and the least one is the JIF.

Figure 8 shows the variations in on bottom drilling time when the increase in drilling parameters is applied. The results are an increase in average ROP with 14 % and a reduction in drilling time with 14% when the drilling parameters (WOB, RPM & JIF) are increased with 10%.

Figure 9 shows that the saved cost when the drilling parameters (WOB, RPM & JIF) are increased by 10 % varies from 30 to 48 thousand USD (32-52 \$/m) which represents about 10-14 % of the total drilling cost (the cost has been calculated based on assumption that the average rig rate per hour is 1,800 \$).

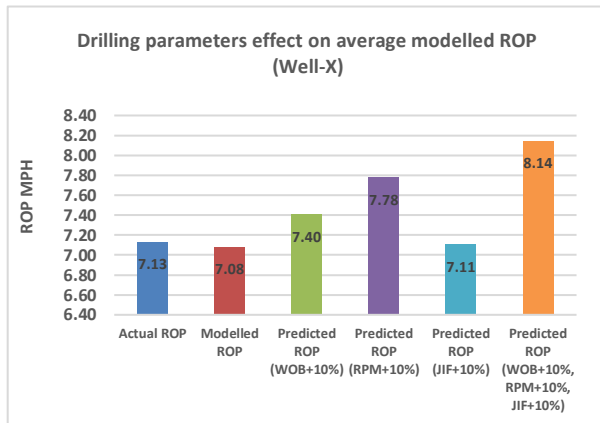


Figure 7 Average ROP values showing the effect of changing drilling parameters (well-X)

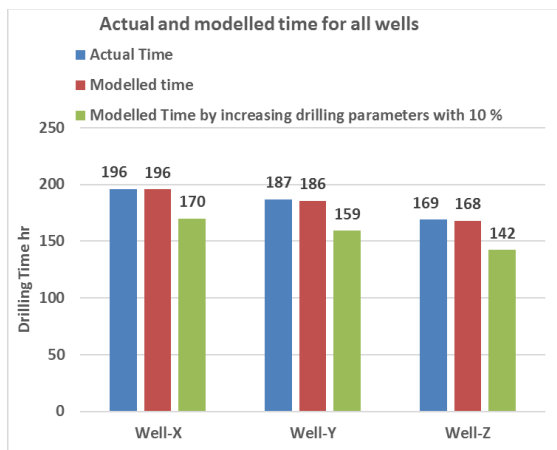


Figure 8 Actual and modelled time for the three wells

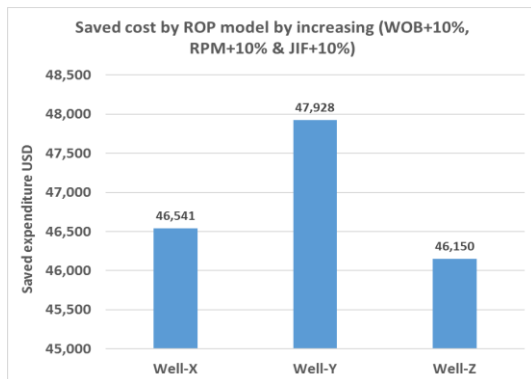


Figure 9 Saved cost when the drilling parameters (WOB, RPM & JIF) are increased by 10 %.

Conclusion

The analysis showed that hybrid bit (kymera: KM524) had achieved high performance in drilling 8.5" hole in Sinai Oil Field. The kymera drill bits offered higher overall ROP (2.5 mph in the very hard formation and 6.6 mph in medium hard formation), lower vibration, better directional control, longer drilled interval (200 m-580 m) and low average cost per drilled meter (350 -700 \$/m).

ROP model has been developed, to predict the ROP values before drilling, to be used in optimization of drill bits' performance in the forthcoming drilled

wells. When the model has been checked and tested on two nearby wells, it has shown high level of accuracy and reliability. The ROP model has been used to optimize the drilling parameters as it relates the ROP value with increasing the drilling parameters. The developed ROP model has used to predict the effectiveness of each drilling parameter and its impact on the ROP values. The ROP model showed that when the drilling parameters (WOB, RPM & JIF) are increased with 10 %, the saved time would be 18-27 hour per and the saved cost would be 30,000 – 48,000 \$ (32-52 \$/m).

Funding sources

This research received no external funding.

Conflicts of interest

There are no conflicts to declare.

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