

A Model for Screening Oil Reservoirs for Carbon Dioxide Flooding

¹Ifeanyi Chukwu, P. C., ¹Isehunwa, S. O., Akpabio, J. U.

¹Department of Petroleum Engineering, University of Ibadan, Ibadan, Nigeria.

²Shell Petroleum Development Company of Nigeria, Port Harcourt, Rivers State, Nigeria

ABSTRACT

The use of Carbon Dioxide (CO₂) flooding as an Enhance Oil recovery (EOR) method is well established in the petroleum industry. An important component of this process is the screening criteria used in assessing the prospective reservoirs. Most of the common screening models passively qualify reservoirs as good or poor candidates for CO₂ EOR. However, the huge upfront capital investment involved in CO₂ EOR projects makes it important for the development of more robust screening criteria. This study used detailed reservoir simulation and Design of Experiment (DOE) techniques to evaluate the key parameters that affect the performance of oil reservoirs undergoing CO₂ flooding and develop a proxy model for the performance prediction. The key parameters affecting performance were established as reservoir heterogeneity, dip, mobility ratio of the CO₂ to oil, injection rate, volume of CO₂ available and the nature of the reservoir fluids. The proxy model was validated with detailed reservoir simulation and comparison with existing correlations.

Keywords: Reservoir performance, CO₂ Flooding, Enhance Oil Recovery, Design of Experiment, Reservoir simulation.

I. INTRODUCTION

The displacement and recovery of oil by CO₂ injection has been extensively applied and studied in the industry since early 1950s. Injection of CO₂ into oil reservoirs can enhance oil recovery and at the same time mitigate the problem of increased CO₂ concentration in the atmosphere. One major advantage of CO₂ as a choice fluid in EOR projects is that at most reservoir conditions, it is a supercritical fluid with high solvency power to extract hydrocarbon components and displace oil miscibly.

Carbon dioxide flooding is a capital intensive project and highly technical in execution. There is therefore the need for robust screening criteria for investigating a reservoir's suitability for the process. Good screening could reduce risk and uncertainty. However, it could be difficult to develop screening criteria that captures all the variables that inclusively contribute to the overall effect of a successful CO₂ flood. This could be responsible for the low performance in some fields that seem to meet the major screening criteria.

The interaction between CO₂, rocks and reservoir fluids varies with type of rock and fluids as well as pressure and temperature. In addition, CO₂ shows more complex phase behavior with reservoir oil than many other solvents. The relatively high solubility in water and the associated reduction in pH will affect the reservoir chemistry depending on the PVT conditions, reservoir fluid and rock composition, Griggs and Siagian (1998).

II. DESCRIPTION AND MODEL DEVELOPMENT

Compositional simulation and pseudo-miscible black oil models

have been widely used to reproduce CO₂ displacement processes. ECLIPSE compositional simulator was used in this study.

A total of 104 reservoir-to-fluid designs to bolster the effect of the contributing factors to the overall performance of CO₂ were investigated. The design entailed building a 30 by 30 by 20 grid cells in the X, Y and Z axis respectively of 100 by 100 by 25ft resulting to 18000 cells. The model has a constant porosity of 25% and a water saturation end point of 10% across all the regions. All the cells are active with no faults. However, the models are hypothetical models built to the Niger delta geologic representation possessing rock and petro physical properties obtainable in the region. A total of 24 different crudes from shell Nigeria fields in the Niger Delta were obtained and characterized using ECLIPSE PVT_i to reproduce fluid performance in the simulator. Other designed parameters are as follows:

Reservoir Depth: To generate a robust scenario, the analysis was carried out at three different depths: 5000ft, 10,000ft and 15,000ft. This serves as a guide in obtaining the formation pressure, equation 1, at various reservoir depths which control injection pressure.

$$p_f = 0.052\rho h \quad (1)$$

This has a direct translation on the choice of injection rates of the CO₂ since operation must be within safety limits to ensure that the formation parting pressure is not exceeded.

For the three depths investigated, the temperature gradients were estimated using equation 2.

$T_g=1.67 \text{ }^\circ\text{F}/100\text{ft}$
(2)

Reservoir Dip: Five different dip angles: 0° , 5° , 15° , 30° and 40° were used.

(C) Modeling Reservoir Permeability: The reservoir permeability was divided into two: (a) Homogeneous permeability distribution; where all the permeabilities of cells in the in the X and Y directions are similar; and the Z axis permeability is constant all through the reservoir (b) Heterogeneous permeability distribution where the permeability distribution differs across the reservoir in both the horizontal and vertical directions.

PETREL pre-processor was used for designing seven different reservoir heterogeneity profiles. A normal distribution profile was used to populate the permeability in the X, Y and Z axis of the 18000 cells as shown in Table A.1.

A permutational selection of each of these cases is done and imported into ECLIPSE compositional simulator with the corresponding angle of dip included prior to importing the fluid. PVT simulation model for EOS tuning process was performed using the Peng-Robinson EOS.

The performance model developed here is a fractional-flow based screening model. It is based on the Koval method, Koval (1963), for predicting recovery in a secondary CO_2 – flood, to model secondary miscible flooding process modified by Claridge (1972) for aerial sweep in an inverted five –spot pattern.

Minimum Miscibility Pressure: In this study, MMP for the twenty four fluids investigated were estimated using Equation 3.

$$MMP = 15.988 * T^{(0.774206+0.0011039)+MWC_{5+}} \quad (3)$$

Where:

T = Temperature in $^\circ\text{F}$

MWC_{5+} = The molecular weight of pentane and heavier hydrocarbons in the reservoir’s oil.

From material balance, the voidage rate is computed as follows:

$$\frac{dv}{dt} = q_o (B_t + (R_p - R_{si})B_g) \quad (4)$$

Where:

$\frac{dv}{dt}$ = Reservoir voidage rate, RM^3/Day

q_o = Oil production rate at start of CO_2 injection, (SM^3/Day)

B_t = Two phase formation volume factor, $(\text{RM}^3 / \text{SM}^3)$

R_s = Solution gas ratio $(\text{SM}^3 / \text{SM}^3)$

R_p = Produced gas oil ratio $(\text{SM}^3 / \text{SM}^3)$

B_g = Gas formation volume factor $(\text{RM}^3 / \text{SM}^3)$

An injection rate that corresponds to the calculated voidage rate was adopted. The surface equivalence of the voidage rate was obtained as follows:

$$Q_{inj.rate} = \frac{\text{Void Rate}}{B_{g.CO_2} @ P_{fi}} \quad (5)$$

Where:

$Q_{Inj.Rate}$ = Injection rate, SM^3/Day

Void Rate = Reservoir Voidage Rate, RM^3/Day

$B_{g.CO_2} @ P_{fi}$ = CO_2 formation volume factor at formation pressures (P_{fi}), RM^3/SM^3

The Hydrocarbon Pore Volume Injected (HCPV) and Fractional Pore Volume (FPV):

The hydrocarbon pore volume injected was estimated by equation 6:

$$HCPV = Q_{Inj.Rate} * t_{eff} \quad (6)$$

Where,

$$t_{eff} = 25 * 365 - t_{initial} \quad (7)$$

Where:

$Q_{Inj.Rate}$ = CO_2 injection rate, SM^3/Day

t_{eff} = Time of effective CO_2 injection, Day

$t_{initial}$ = Time before inception of CO_2 injection, Day

$$FPV = \frac{HCPV}{RSPV} \quad (8)$$

Where:

FPV = Fractional Pore Volume, ratio

$HCPV$ = Hydrocarbon Pore Volume, RM^3

$RSPV$ = Reservoir Pore Volume, RM^3

CO₂ Mobility Ratio:

Since the viscosity of CO₂ at reservoir conditions is much lower than that of most oils, viscous instability will limit the sweep efficiency of the displacement and therefore, oil recovery Campbell (1985).

$$M = \frac{K_{rcos_2} / u_{CO_2}}{K_{rowg} / u_o}$$

(9)

Where:

M = Mobility ratio of CO₂ to oil

K_{rcos_2} = Relative permeability of rock to CO₂

K_{rowg} = Relative permeability of rock to oil in the presence of water and gas

μ_{CO_2} = Viscosity of CO₂, cp

μ_o = Viscosity of oil, cp

Proxy Model Development

A ‘fold over’ Plackett-Burman design, which has been observed to be very efficient in screening models was used (Yeten *et al.*, (2005). The implicit assumption is that all interactions are negligible (Myers, *et al.*, 2008). The main contribution effects out of the 28 parameters examined on the response were obtained from the contrast of the average of the highs and the lows, or mathematically expressed as follows:

Simplifying Equation 11 will yield:

$$A = \frac{1}{ADD RECOV + 1000002} \tag{11a}$$

$$C = 100003E - 06 \tag{11b}$$

$$B1 = -1.32294E - 11 * \phi - 5.38567E - 11 * NTG \tag{11c}$$

$$B2 = -1.29548E - 14 * Kx + 7.03298E - 15 * Ky + 6.24671E - 14 * Kz + 1.76294E - 11 * \frac{K_v}{K_h} - 1.77156E - 14 * HH + 2.15642E - 14 * VH \tag{11d}$$

$$B3 = -1.33092E - 13 * API - 1.66683E - 12 * \mu_o \tag{11e}$$

$$B4 = -5.30945E - 15 * MMP - 2.07089E - 12 * Pc + 1.62500E - 14 * Pi + 5.2738E - 15 * P_b + 9.6858E - 15 * P_{ref} \tag{11f}$$

$$Effect = \frac{\sum Y_+}{n_+} - \frac{\sum Y_-}{n_-}$$

(10)

Where:

N= Number of data points collected at each level

Y= Associated responses

The half-normal probability curve was used to identify the sensitivities of the factors along with their interactions. Half probability plot was used to take the absolute values of the effects as shown in Fig. B.2.

The inverse transformation principle was adopted and applied over the sample spaces available to generate a final equation for the additional recovery obtainable from the Carbon dioxide flooding as:

$$A = C + B1 + B2 + B3 + B4 + B5 + B6 \tag{11}$$

Where:

A = Additional Recovery inverse

C = Constant

B1 = Porosity

B2 = Permeability and Heterogeneity

B3 = Fluid property

B4 = Pressures

B5 = Reservoir orientation and Mobility

B6 = Gas conditions at start of CO₂ injection

$$B5 = 1.12521E-17 * V_r - 1.50031E-13 * Dip + 7.50462E-17 * D - 5.33625E-16 * K_{ro} + 2.85318E-16 * M \quad (11g)$$

$$B6 = -1.52557E-14 * R_g - 6.73741E-16 * R_p + 7.74225E+13 * S_o + 1.23132E-12 * HPVi - 1.6768E-17 * Q_i + 2.57234E-14 * T \quad (11h)$$

Where:

- * ADD RECOV = Additional recovery
- * ϕ = Porosity of the formation
- * NTG = Net to gross ratio
- * K_x = Horizontal permeability in the x axis
- * K_y = Horizontal permeability in the y axis
- * K_z = Vertical permeability
- * K_v/K_h = Vertical to horizontal permeability ratio
- * HH = Heterogeneity index in the horizontal direction
- * VH = Heterogeneity index in the vertical direction
- * API = Fluid gravity
- * μ_o = Fluid viscosity
- * MMP = Minimum miscibility pressure
- * P_c = Capillary pressure
- * P_i = Reservoir initial pressure
- * P_b = Reservoir bubble pressure
- * P_{ref} = Reference pressure for CO₂ injection
- * V_r = Reservoir voidage rate at inception of CO₂ injection
- * Dip = Reservoir angle of dip
- * D = Formation depth
- * K_{ro} = Oil relative permeability at inception of CO₂ injection

- * M = CO₂ to oil mobility ratio at inception of CO₂ injection
- * R_s = Solution gas at inception of CO₂ injection
- * R_p = Produced gas at inception of CO₂ injection
- * S_o = Oil saturation at inception of CO₂ injection
- * HPVi = HCPV of CO₂ introduced to the formation
- * Q_i = CO₂ injection rate
- * T = Reservoir temperature

Model Validation

Validation of the model developed was done by comparing the prediction of the model to the predictive performance of Claridge Correlation. The predictions were then compared with results of ECLIPSE® CO₂ compositional simulation runs. A program dedicated to the prediction of CO₂ EOR performance, based on the principles of the generated model, was presented and its attendant results obtained from imputation of reservoir and fluid parameters was juxtaposed with simulated values from ECLIPSE compositional simulator.

The results are shown in Tables (1-3) and figures (1-3)

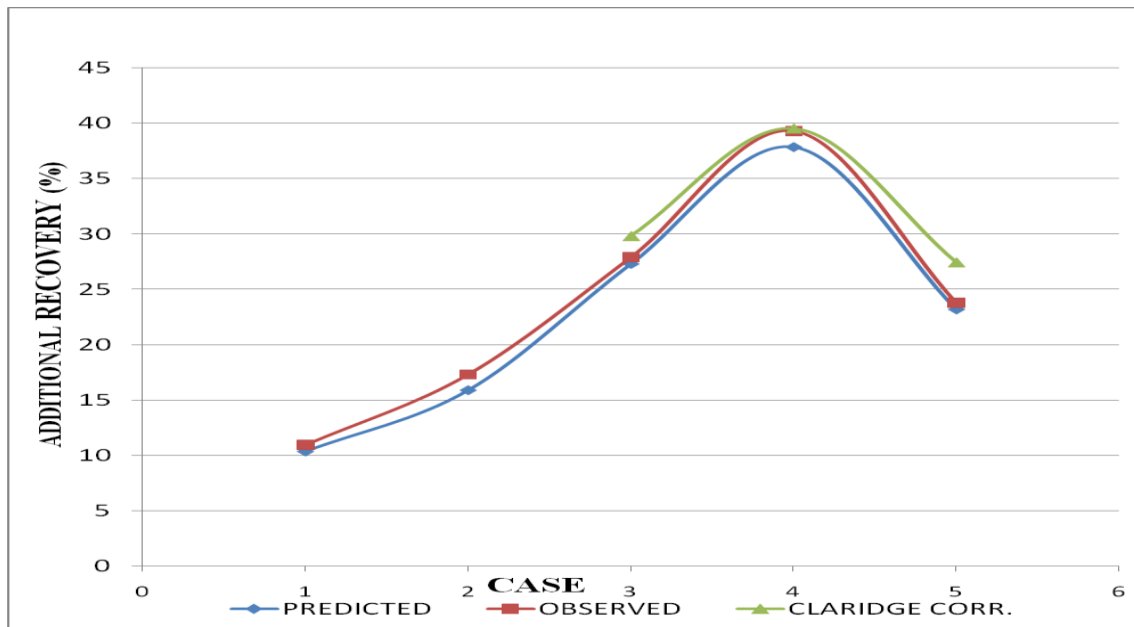


Figure 1: Models Predictions Vs ECLIPSE Observed performance @ 5000ft & 110°F

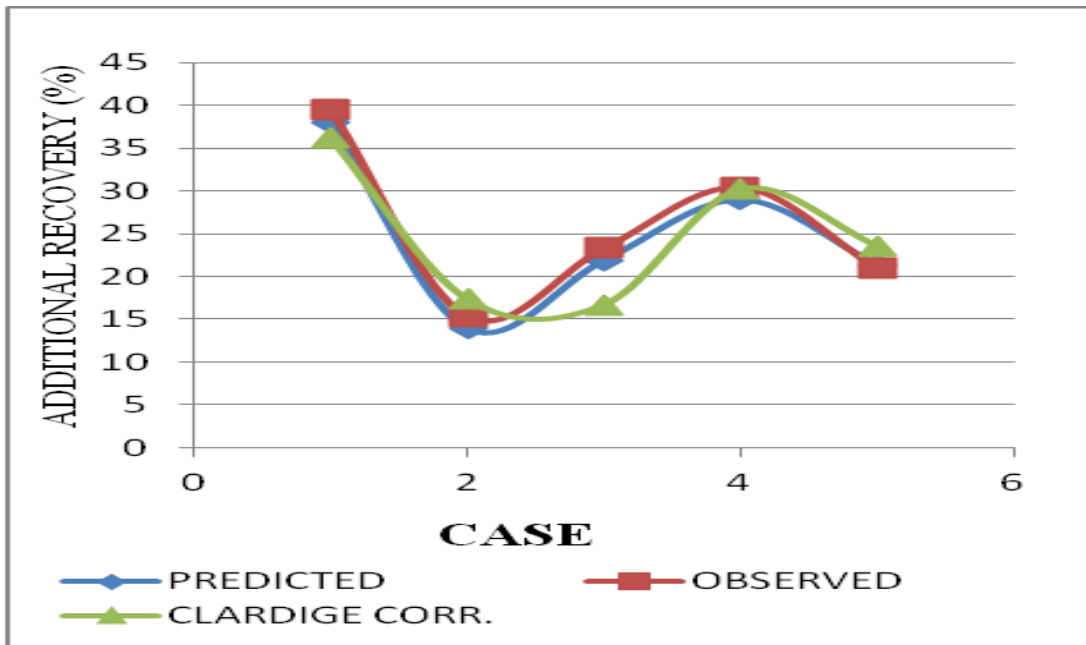


Figure 2: Models Predictions Vs ECLIPSE Observed performance @ 10000ft & 187°F

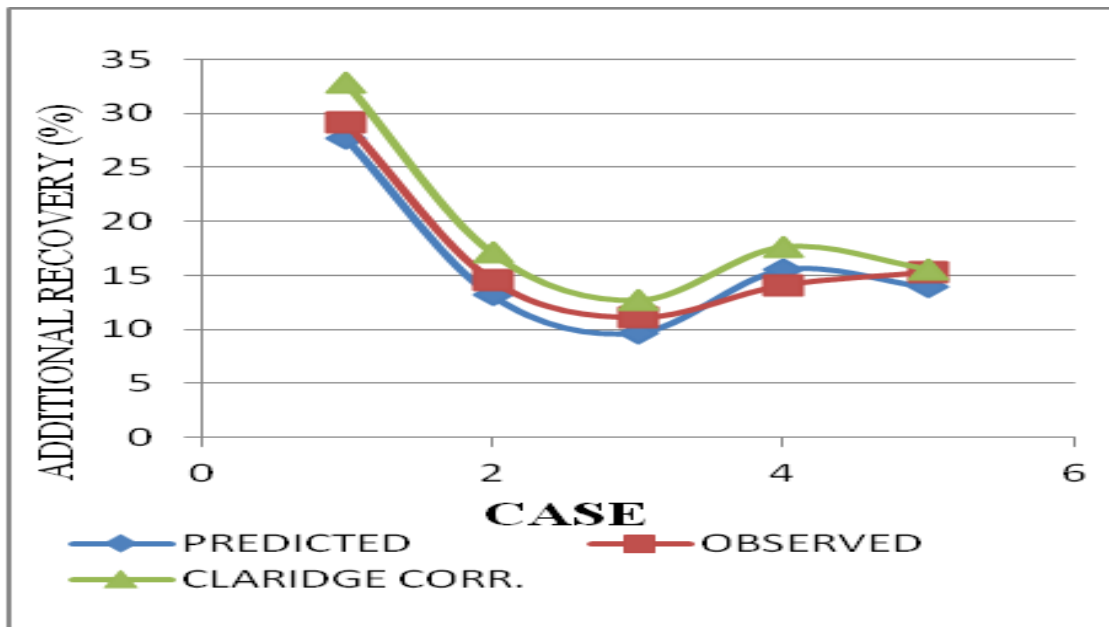


Figure 3: Models Predictions Vs ECLIPSE Observed performance @ 15000ft & 280°F

Analysis of Model Predictions

Tables 1, 2 and 3 presents the observed data from ECLIPSE® and the predicted data from this study and Claridge correlation.

From the tables, the average deviation is calculated using:

$$Average\ Deviation = \frac{1}{n} \sum_{i=1}^5 x_i |-\bar{x}| \quad (21)$$

Where:

n = no of runs

x = calculated difference

\bar{x} = mean of calculated difference

The deviations in the 5000ft, 10000ft and 15000ft are 0.399%, 0.906% and 0.586% respectively for the generated model in this study. The Claridge Correlation exhibits deviations of 1.145%, 1.11% and 1.59% respectively.

Table 1: Analysis of Models Precisions @ 5000ft

RUNS	PREDICTED	CLARIDGE CORR.	OBSERVED	DIFF. BTW MODEL & OBSERVED	DIFF. BTW CLARIDGE CORR. & OBSERVED
1	10.3791	46.724	10.9632	0.584095792	-
2	15.9116	35.48	17.3275	1.415867637	-
3	27.3159	29.802	27.9	0.584073637	1.902
4	37.8675	39.5119	39.2834	1.415926817	0.2285
5	23.2059	27.432	23.79	0.584085276	3.642

Table 2: Analysis of Models Precisions @ 10000ft

RUNS	PREDICTED	CLARIDGE CORR.	OBSERVED	DIFF. BTW MODEL & OBSERVED	DIFF. BTW CLARIDGE CORR. & OBSERVED
1	27.7371	32.81	29.153	1.415925876	3.657
2	13.2033	17.126	14.6192	1.415913736	2.50682
3	9.7033	12.6904	11.1192	1.415903825	1.5712
4	15.5259	17.6324	14.11	1.415887313	3.5224
5	13.9141	15.59	15.33	1.415886733	0.26

Table 3: Analysis of Models Precisions @ 15000ft

RUNS	PREDICTED	CLARIDGE CORR.	OBSERVED	DIFF. OF MODEL & OBSERVED	DIFF. OF CLARIDGE CORR. & OBSERVED
1	38.024	36.166	39.44	1.415955007	3.274
2	13.9141	17.35	15.33	1.415886733	2.02
3	21.8841	16.71	23.3	1.415909302	6.59
4	28.9841	30.1874	30.4	1.415929408	0.213

5	21.3359	23.546	20.92	-	0.415906597	2.626
---	---------	--------	-------	---	-------------	-------

III. CONCLUSION

This work presented a model that can be used to screen Carbon dioxide flood projects. The results have shown that degree of heterogeneity of the reservoir, reservoir dip, mobility ratio of the CO₂ to oil, injection rate, volume of CO₂ introduced to the formation and injection well configurations are important parameters that could affect the performance of CO₂ EOR projects. A *priori-knowledge* of extra recovery expected to be obtained if CO₂ EOR is embarked on is no doubt a prudent approach and promotes best engineering practices.

Nomenclature

- BOPD = Barrel of Oil Produced per day
- CO₂ = Carbon Dioxide
- DOE = Design of Experiment
- EOR = Enhanced Oil Recovery
- EOS = Equation of State
- HC = Hydrocarbon
- IPM = Integrated Petroleum Management
- MMP = Minimum Miscibility Pressure
- MW = Molecular Weight
- P_{fi} = Formation Pressure
- RM³ = Reservoir Cubic Meter
- RSM = Response Surface Methodology
- SM³ = Standard Cubic Meter
- WAG = Water Alternating Gas

Acknowledgement

The Authors wish to thank the staff of the Department of Petroleum Engineering and the Office of Petroleum Engineering Shell Chair, University of Ibadan for their support in the course

of this project. Special thanks to Schlumberger, for the learning Center in the University of Ibadan for the use of Eclipse to facilitate this project.

REFERENCES

Campbell, B. T., "Flow Visualization for CO₂/Crude Displacements," *SPEJ*, October 1985, p665-687.

Claridge, E. L., 1972; Prediction of Recovery in Unsteady Miscible Flooding. *J. SPE*, 12(2), 143-155

Eclipse Reference Manual, 2008. Schlumberger.

Grigg and Sigan. "Understanding and exploiting four-phase flow in low-temperature CO₂ floods". SPE paper 39790, presented at the SPE Permian Basin Oil & Gas Recovery Conference, Midland, Texas 25 - 27 March 1998.

Koval, E.J., 1963: A method for Predicting the Performance of Unstable Miscible Displacement in Heterogeneous Media, *J. SPE*, 3(6), 145-154. *Trans AIME* 228.

Myers, R.H., Montgomery, D.C., and Anderson-Cook, C. "Response Surface Methodology: Process and Product Optimization Using Designed Experiments." Third Edition. New York City: John Wiley and Sons, Inc. 2008.

Yeten, B., Castellini, A., Guyaguler, B and Chen, W. H. 2005."A Comparison Study on Experimental Design and Response Surface Methodologies", SPE 93347 presented at the 2005 SPE Reservoir Simulation Symposium held in Houston, Texas, USA, 31 January 2005-2 February 2005.

APPENDIX

Table A.1 Design Cases of Populated Heterogeneity

Property	1	2	3	4	5	6	7
Kx (md)	100-400	250-310	80-600	200-800	150-750	90-200	300-1000
Ky(md)	100-400	250-310	80-600	200-800	150-750	90-200	300-1000

Kz(md)	10-	5-	12-	2-	1-	3-	2-
	20	9	33	7	15	30	18

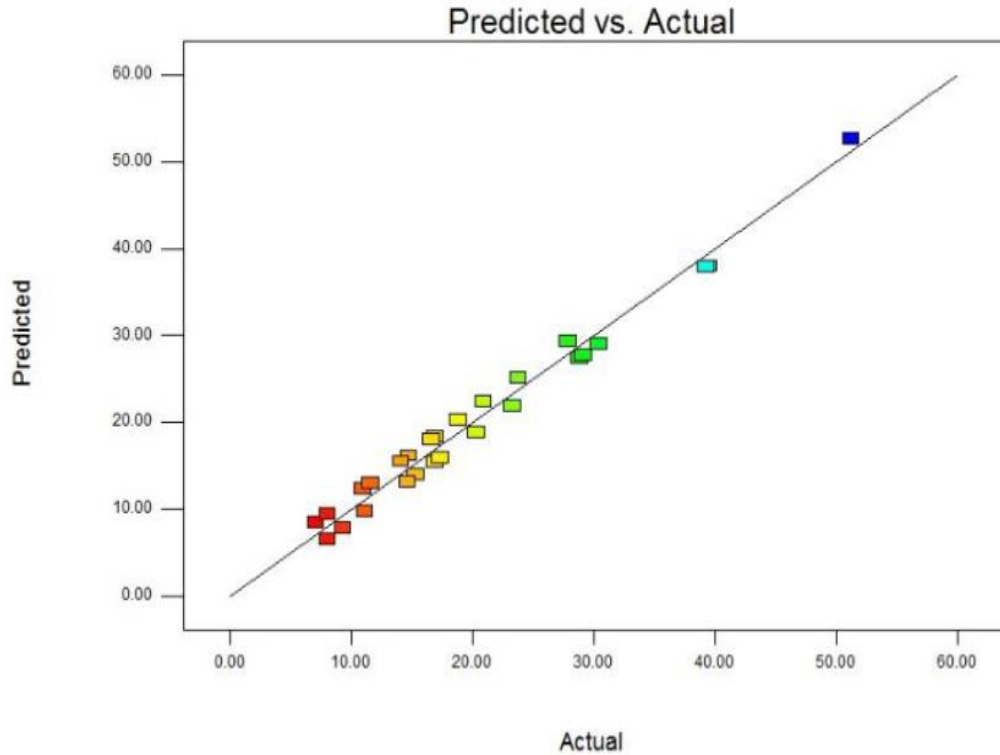


Figure A1: Predicted Vs Actual Profile for Factors

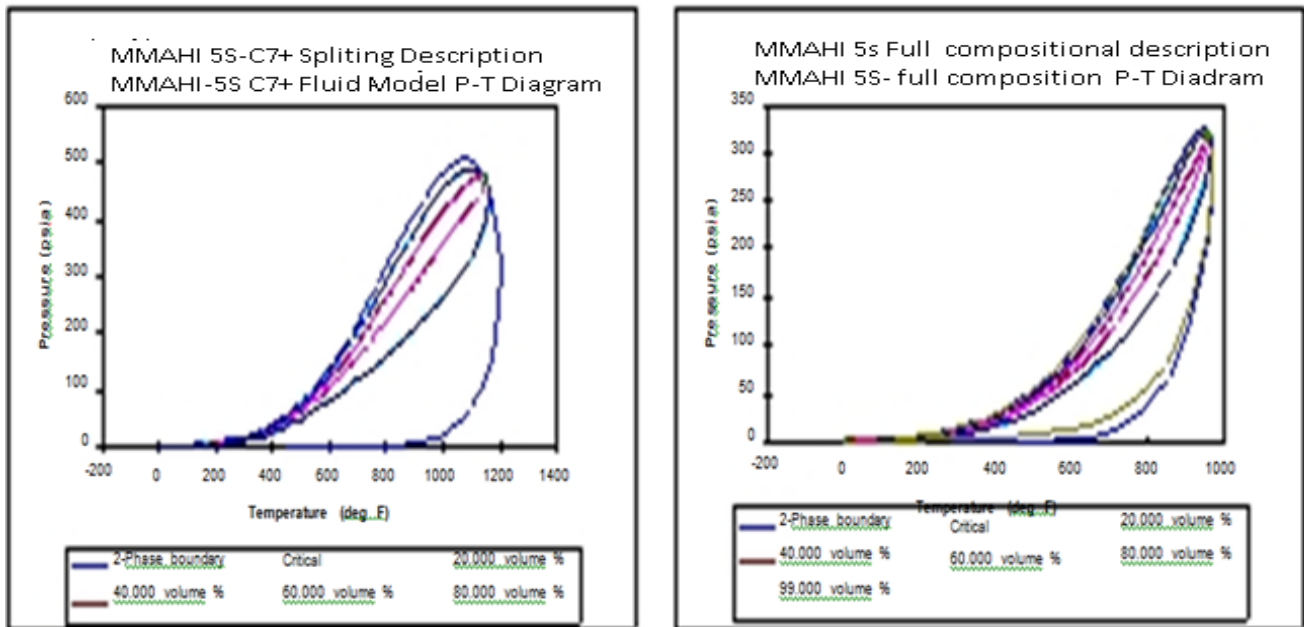


Figure B.1 Descriptions of a C₇₊ and Full Fluid Compositional Model

Figure B.2 Predicted Vs Actual Profile for Factors

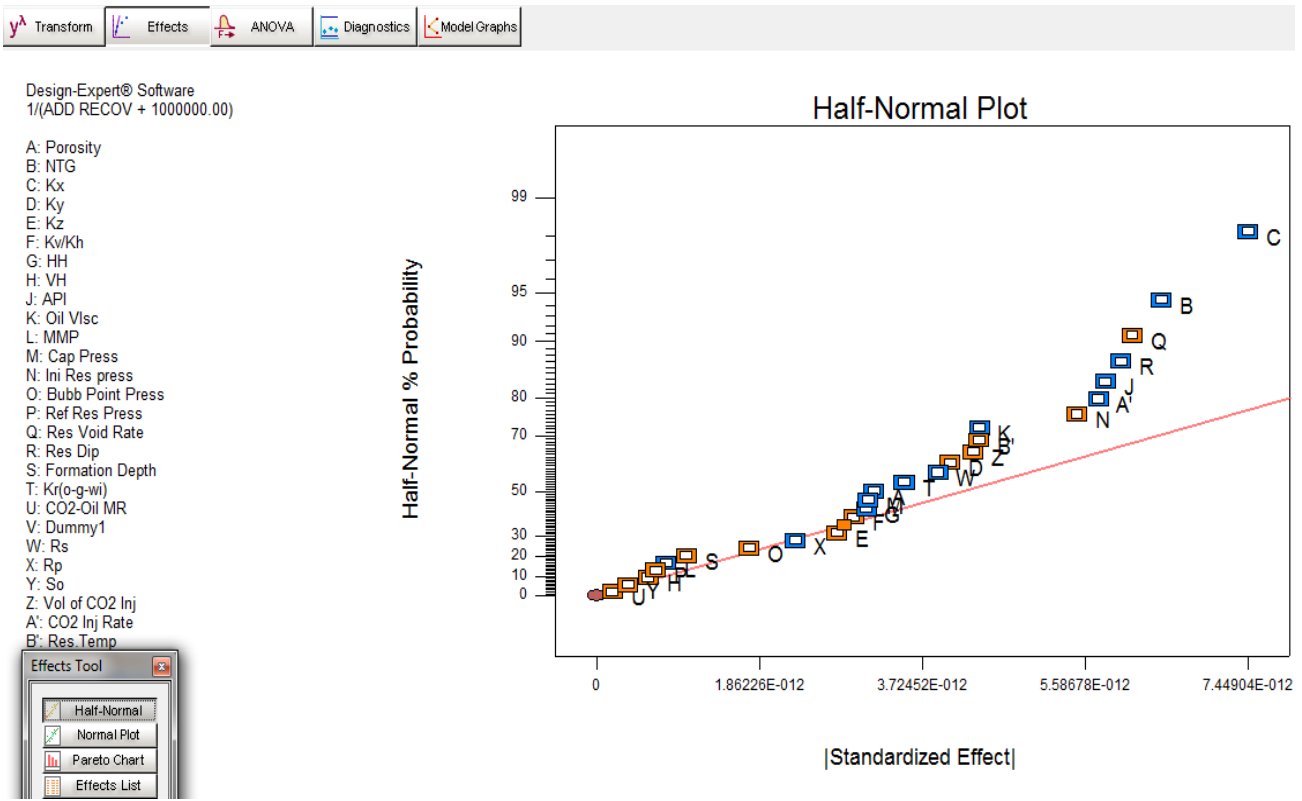


Figure B.3 Half Normal Plot for Effect of Factors in CO₂ Injection

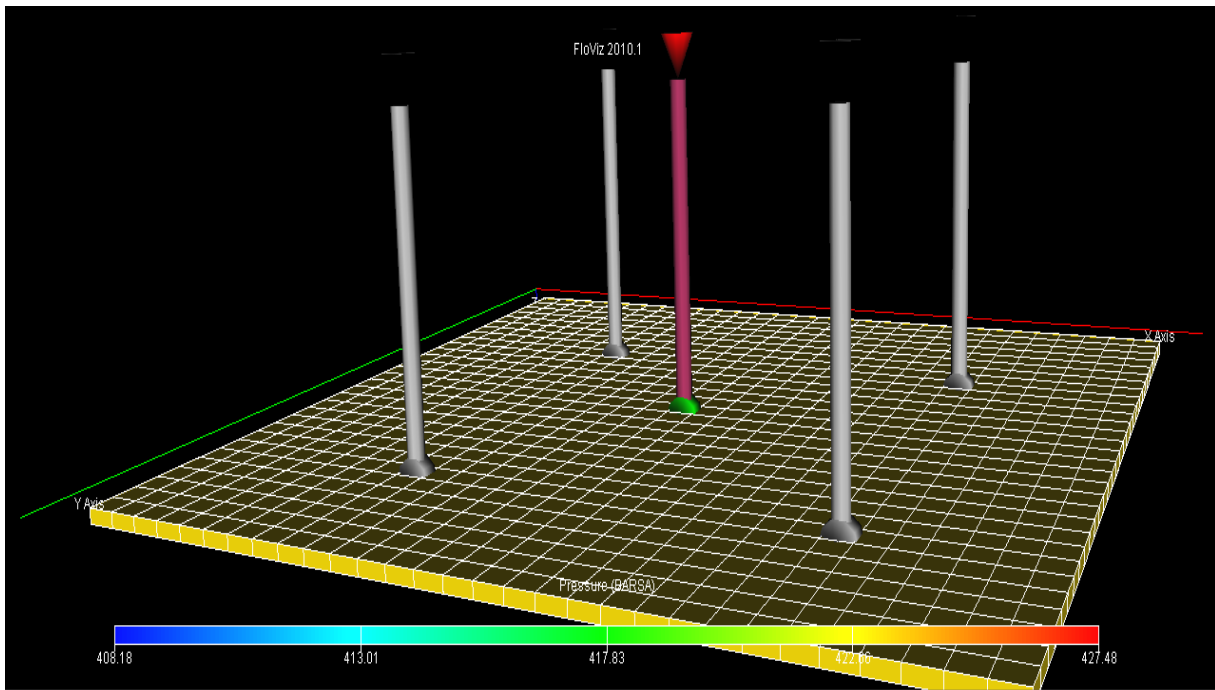


Figure B.4 Well Configurations for a Horizontal Reservoir (0° Dip)