



A Hybrid Approach Based on Artificial Neural Network and Integrated Production Modeling for Gas Lift Optimization

By

Mazen Mohamed Bahaa El-Din Hussein Hamed

A Thesis Submitted to the
Faculty of Engineering at Cairo University
in Partial Fulfillment of the
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MASTER OF SCIENCE
In
GAS PRODUCTION ENGINEERING

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in PETROLEUM ENGINEERING NATURAL GAS ENGINEERING AND PRODUCTION

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Title of Thesis:

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Key Words:

Integrated production Modeling; Gas lift Optimization; artificial neural network modeling; estimating well bottom hole pressure; predicting well fluid rate.

Summary:

An artificial neural network model was developed to predict the values of the bottom hole flowing pressure and the total fluid rate per each well using the available field parameters like the water cut samples, static pressure surveys, reservoir gas oil ratio, the well head temperature and pressure in addition to the gas injection rate and gas injection pressure. This developed ANN used in building accurate individual well models on PROSPER and a full field network model gathering all the individuals' models with the surface network. This creates an integrated production model aiming to perform field wide gas lift optimization.

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Dedication

I would like to dedicate this thesis to my parents who stands beside me giving me the continuous assistance and support through my whole life. They are the main reason for any success in my life after the God almighty. They deserve a lot, there is nothing can express my love and pride toward them.

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Nomenclature

Symbols

C Learning constant that determine the rate of learning

D Inside pipe diameter, Inch

E Least square error

E_K Dimensionless kinetic-energy pressure gradient

F Friction factor

PI Productivity Index, BPD/PSI

 P_{wf} Well bore flowing pressure, PSI

P_r Reservoir pressure, PSI

P_B Bubble point pressure, PSI

PBTD Plugged Back Total Depth, Ft

Q max Maximum liquid production, BPD

Q_f total fluid rate of the well, BFPD

H_L Liquid holdup

SCF/STB Standard cubic feet/Stock tank barrel

STBPD Stock Tank Barrel Per Day

O_i Resulted neuron output

ID Inner Diameter, Inch

R Learning signal

RB/STB Reservoir barrel/Stock tank barrel

wi(t) Old values of t weight vector at time level t

wi(t + 1) New values of weight vector at time level t+1

Δwi Weight adjustment

θ Inclination angle from vertical

 ρ_{mn} Mixture density, PCF

 ρ_L Liquid density, PC

 ρ_n No-slip density,PCF

 ρ_{mn} No-slip mixture density, PCF

V_m Mixture velocity, FT/SEC

Abbreviations

ANN Artificial neural network

BPNN Back propagation neural network

BHFP Bottom-Hole Flowing Pressure

BFPD Barrel fluid per d y

BOPD Barrel oil per day

CO₂ Carbon dioxide

FVF(Bo) Formation Volume Factor

GOR Gas Oil Ratio, SCF/STB

IPM Integrated Production Modeling

PVT Pressure, Volume, Temperature

H₂S Hydrogen sulfide

N₂ Nitrogen

VLP Vertical Lift Performance