

ARTIFICIAL NEURAL NETWORKS APPLICATION IN COMBINED CARBON
DIOXIDE REFORMING AND PARTIAL OXIDATION OF METHANE

RUZINAH ISHA

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This work is lovingly and respectfully in honour of my wonderful parents,
Isha Amat Ikram and Zaliha Moh,
as well as my beloved husband and lovely daughter,
Syed Izam Syed Mokhtar and Sharifah Nur Irdina Syed Izam Al-Yahya.

“Many of life’s experiences cultivate unforgettable lessons and rewards.”

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ABSTRACT

A catalytic carbon dioxide reforming of methane with the effect of oxygen was carried out on 1 wt% of Rhodium (Rh) on Magnesium Oxide (MgO) and ZSM-5 catalysts. The effect of parameters on the methane conversion, synthesis gas selectivity and H₂/CO ratio were studied. Three main parameters: temperature, O₂/CH₄ ratio and catalyst weight in 100 ml/min total feed flow rate, have been identified as the major factors that control the process. The results indicated that Rh/MgO showed better catalyst reactivity and stability even though at temperature higher than 800°C. Thus, the optimization of the combined CORM and partial oxidation of methane over Rh/MgO catalyst was carried out. The optimization study was performed with the help of experimental design and two mathematical approaches: empirical polynomial and artificial neural network. Empirical polynomial models were employed to analyze the effect of parameters on the response factor and the correlation coefficient, *r*, was above 85%. However, the feed forward neural network correlation coefficient was more than 95%. The feed forward neural network modeling approach was found to be more efficient than the empirical model approach. The optimum condition for maximum methane conversion was obtained at 850°C with O₂/CH₄ ratio of 0.14 and 141 mg of catalyst resulting in 95% methane conversion. A maximum of 40% hydrogen selectivity was achieved at 909°C, 0.23 of O₂/CH₄ ratio and 309 mg catalyst. The maximum H₂/CO ratio of 1.6 was attained at 758°C, 0.19 of O₂/CH₄ and 360 mg catalyst. The utilization of neural network in predicting the reaction for other catalyst was also tested by introducing other reaction data in the network. The result showed a feed forward neural network was able to predict the output of the reaction even for different reaction or catalysts.

ABSTRAK

Kesan oksigen terhadap mangkin 1 % berat Rhodium di dalam Magnesium Oksida (MgO) dan ZSM-5 dikaji untuk proses pembentukan semula metana dengan menggunakan gas karbon dioksida (CORM). Kesan parameter terhadap penukaran metana, pemilihan gas sintesis dan nisbah H_2/CO diselidiki. Tiga parameter utama: suhu, nisbah reaktan (O_2/CH_4) dan berat mangkin dalam 100 ml/min jumlah halaju suapan dikenalpasti sebagai faktor utama yang mengawal proses tersebut. Keputusan menunjukkan bahawa Rh/MgO mempunyai keaktifan dan kestabilan yang lebih baik walaupun pada suhu melebihi $800^\circ C$. Oleh itu, optimasi untuk kombinasi CORM dan oksidasi separa metana menggunakan mangkin Rh/MgO dijalankan. Proses optimasi dibuat dengan bantuan rekabentuk eksperimen dan dua pendekatan matematik: polinomial empirik dan rangkaian saraf buatan. Model polinomial empirik yang diterbitkan untuk menganalisa kesan parameter terhadap faktor keluaran dan pekali kolerasinya, r adalah melebihi 85%. Walau bagaimanapun, pekali kolerasi untuk suapan hadapan rangkaian saraf adalah melebihi 95%. Oleh itu, suapan hadapan rangkaian saraf adalah lebih efisien daripada model polinomial empirik. Penukaran metana tertinggi sebanyak 95% dihasilkan pada suhu $850^\circ C$ dengan nisbah O_2/CH_4 sebanyak 0.14 dan 141 mg mangkin. Pemilihan hydrogen secara maksima sebanyak 40% boleh dicapai pada suhu $909^\circ C$, nisbah O_2/CH_4 sebanyak 0.23 dan 309 mg mangkin. Nisbah maksima H_2/CO sebanyak 1.6 dihasilkan pada suhu $758^\circ C$ dengan nisbah O_2/CH_4 sebanyak 0.19 dan 360 mg mangkin digunakan. Penggunaan suapan hadapan rangkaian saraf untuk meramal tindakbalas bermangkin yang lain juga dikaji untuk tindakbalas yang berbeza di dalam rangkaian yang sama. Keputusan menunjukkan suapan hadapan rangkaian saraf berupaya meramal tindakbalas walaupun untuk tindakbalas atau mangkin yang berlainan.

TABLE OF CONTENTS

CHAPTER	TITLE	PAGE
	THESIS STATUS DECLARATION	
	SUPERVISORS' DECLARATION	
	TITLE PAGE	i
	DECLARATION OF ORIGINALITY	ii
	DEDICATION	iii
	ACKNOWLEDGEMENT	iv
	ABSTRACT	v
	ABSTRAK	vi
	TABLE OF CONTENTS	vii
	LIST OF TABLES	xi
	LIST OF FIGURES	xiii
	LIST OF SYMBOLS	xvi
	LIST OF APPENDICES	xviii
1	INTRODUCTION	1
	1.1 General Introduction	1
	1.2 Background and Problem Statement	5
	1.3 Research Objectives	7
	1.4 Scope of Study	7
2	LITERATURE REVIEW	
	2.1 Introduction	9
	2.2 Catalytic Natural Gas Utilization	12
	2.2.1 Steam Reforming of Methane	13

2.2.2	Partial Oxidation of Methane	14
2.2.3	Carbon Dioxide Reforming of Methane	16
2.2.4	The Combination of Methane Reforming Reaction	17
2.3	Reforming Methane Technology	21
2.4	Design of Experiments	24
2.4.1	Response Surface Methodology	25
2.4.2	Central Composite Design	26
2.5	Artificial Neural Network	26
2.5.1	Neural Network Topology	28
2.5.2	Network for Small Experimental Data set	30
2.5.3	Retraining Method for Neural Network	31
2.6	Statistical Analysis	33
3	RESEARCH METHODOLOGY	
3.1	Introduction	35
3.2	Experimental Methodology	37
3.2.1	Catalyst Preparation	37
3.2.2	Catalyst Characterization	38
3.2.2.1	Nitrogen Adsorption	38
3.2.2.2	Temperature Program Reduction (TPR)	38
3.2.3	Catalyst Testing	38
3.3	Data Analysis	40
3.4	Equilibrium Thermodynamic Simulation Development	41
3.5	Model Development	42
3.5.1	Input and Output Variables Selection	43
3.5.2	Empirical Model	43
3.5.3	Neural Network Modeling	45
3.6	Model Comparison and Optimization	48
3.7	Neural Network Application in Predicting Two Reaction	48

4	EFFECT OF PROCESS VARIABLES	
4.1	Introduction	52
4.2	Thermodynamic Simulation Result	53
4.3	Rh Catalyst Characteristics	58
4.3.1	Surface Area	58
4.3.2	Temperature Programmed Reduction (TPR)	59
4.4	Effect of Rh Support	61
4.5	Effect of Oxygen	63
4.6	Effect of Operating Temperature	67
4.7	Effect of Catalyst Weight in Total Flow Rate	74
4.8	Catalyst Stability Test	76
4.9	Conclusion	77
5	MODEL DEVELOPMENT AND OPTIMISATION	
5.1	Introduction	79
5.1.1	Polynomial Empirical Model Development	81
5.1.2	Neural Network Model Development	88
5.1.2.1	Selection of Neural Network Topology	89
5.2	Statistical Evaluation of Model Fitness	101
5.3	Optimisation	105
5.4	Conclusion	110
6	ARTIFICIAL NEURAL NETWORK IN HELPING CATALYST EXPERIMENTATION	
6.1	Introduction	111
6.2	Feed Forward Neural Network Development	116
6.2.1	Retraining Neural Network	116
6.2.2	Data Combining Neural Network	118
6.2.3	Stacked Neural Network	121
6.3	Statistical Evaluation of Models Fitness	130
6.4	Network Structure Comparison	132

6.5	Conclusion	134
7	CONCLUSION AND RECOMMENDATIONS	
7.1	Introduction	135
7.2	Recommendations	137
	REFERENCES	139
	APPENDICES	147

LIST OF TABLES

TABLE NO.	TITLE	PAGE
1.1	Gas reserves in Malaysia 1999	3
1.2	Sources of synthesis gas and their applications	4
2.1	Summary of combined process of synthesis gas production	19
3.1	The levels of variables chosen for trials	44
3.2	Input variables of data used in this study	49
4.1	Nitrogen Adsorption data for Rh/ZSM-5 and Rh/MgO	58
4.2	The Catalytic Reactivity at 675°C and CO ₂ /CH ₄ ratio equal to 1 with 200mg catalyst	61
4.3	The catalytic reactivity of Rh/ZSM-5 and Rh/MgO at 675°C with 200 mg catalyst.	64
4.4	The H ₂ /CO product ratio in different temperature of Rh/MgO at O ₂ /CH ₄ =0.17.	73
5.1	Input variables and output response factor for model development data and model validation data.	80
5.2	Central composite design with observed and predicted values	82
5.3	Significance of regression coefficient for methane conversion	84
5.4	Significance of regression coefficient for hydrogen selectivity	86
5.5	Significance of regression coefficient for H ₂ /CO ratio	87
5.6	Training and Validation Error of MISO networks for methane conversion in various number of hidden layer nodes	91
5.7	The Detailed MISO and MIMO structure	92
5.8	Training and Validation Error of MISO networks for various	93

	number of nodes in hidden layer.	
5.9	Training and Validation Error of MIMO networks for various number of nodes in hidden layer	94
5.10	Experimental results and predicted result from neural network model for training data and validation data	97
5.11	The value of biases and weight of FFNN model for MISO network	99
5.12	The value of biases and weight of FFNN model for MIMO network	100
5.13	ANOVA for each responses for polynomial empirical model and neural network models	104
5.14	Optimisation result from NN	106
6.1	Data 1 specification of CORM reaction for Rh/MgO catalyst	114
6.2	Data 2 specification of POM reaction for Co/Ni catalyst	115
6.3	Validation RMSE in Different number of nodes in hidden layer	116
6.4	Validation RMSE of data 1 and data 2 network after introducing data 2 in data 1 network	117
6.5	RMSE of Data Combining NN in different single hidden layer nodes	119
6.6	Validation RMSE of five subnet of three layer stacked NN	125
6.7	Validation RMSE of three subnet of three layer Stacked NN	126
6.8	Validation RMSE of one subnet in stacked NN	126
6.9	Stacked NN validation RMSE and Validation Coefficient Correlation, r in different hidden layer node structure	128
6.10	Stacked NN validation RMSE value for network topology of 6-6-4 hidden layer nodes structure	129
6.11	Overall ANOVA for each network structure	131
6.12	The training and validation RMSE of retrained NN, combined NN and stacked NN	132

LIST OF FIGURES

FIGURE NO.	TITLE	PAGE
1.1	The world proven reserves of natural gas versus consumption from the year of 1980 to 1999	2
1.2	General application of synthesis gas (syngas) to higher value added product	5
2.1	World energy consumption by fuel type for 1970-2020	9
2.2	Consumption of natural gas in volume %	10
2.3	Various direct and indirect routes for the production of useful chemicals from natural gas	11
2.4	Shell middle distillate synthesis (SMDS) in Bintulu, Sarawak, Malaysia.	15
2.5	Common laboratory reactors: operation modes and reacting system	21
2.6	Graph of the information flow in a feed forward neural network	29
2.7	General scheme of the stacked NN	31
2.8	General scheme of the retraining process.	32
3.1	The flow chart of research procedure	36
3.2	The procedure of the catalyst preparation via impregnation method	37
3.3	Experimental rig set up	39
3.4	Feed Forward Neural Networks Topology	47
3.5	General scheme of neural network structure. (a) Retraining NN (b) Data combining NN (c) Stacked NN	50

4.1	The thermodynamic profile of combined CORM and POM reaction for methane conversion	54
4.2	The thermodynamic profile of combined CORM and POM reaction for hydrogen yield and carbon monoxide yield	55
4.3	The thermodynamic profile of combined CORM and POM reaction for H ₂ /CO ratio	56
4.4	The TPR profile of Rh/MgO and Rh/ZSM-5 catalysts	59
4.5	The Rh/MgO catalyst reactivity at 675°C with 200 mg catalyst	66
4.6	CH ₄ conversion and H ₂ /CO product ratio in different temperature for (*) Rh/ZSM-5 and (O) Rh/MgO catalysts at O ₂ /CH ₄ ratio equal to 0.17.	68
4.7	H ₂ yield and CO yield versus temperature at O ₂ /CH ₄ ratio equal to 0.17 ((*) Rh/ZSM-5 and (O) Rh/MgO)	69
4.8	The Rh/MgO catalyst reactivity in various temperature and O ₂ /CH ₄ =0.17	71
4.9	The methane conversion and H ₂ /CO catalysts reactivity in various catalyst weights at 750°C	74
4.10	The hydrogen yield and carbon monoxide yield catalysts reactivity in various catalyst weights at 750°C	75
4.11	Rh/MgO catalytic reactivity at 900°C for more than 16 hours	77
5.1	Pareto chart of standardized effects of methane conversion.	84
5.2	Pareto chart of standardized effects of hydrogen selectivity	86
5.3	Pareto chart of standardized effects of H ₂ /CO ratio	88
5.4	Feed Forward Neural Networks Topology of MISO Neural Networks	90
5.5	Comparison of Predicted and Experimental value for training data.	102
5.6	Comparison of Predicted and Experimental value for validation data	103
5.7	Contour surface plot of methane conversion as a function of temperature, °C and O ₂ /CH ₄ when 141mg catalyst used in the reaction	107

5.8	Contour surface plot of hydrogen selectivity as a function of temperature, °C and O ₂ /CH ₄ when 309mg catalyst used in the reaction	108
5.9	Contour surface plot of H ₂ /CO as a function of temperature, °C and O ₂ /CH ₄ when 360mg catalyst used in the reaction	109
6.1	The Observed value versus Predicted value of training data and validation data of data combination NN	121
6.2	Five subnetwork of three layer stacked NN	122
6.3	Three subnetwork of three layer stacked NN	123
6.4	One subnetwork of three layer stacked NN	123
6.5	Schematic Diagram of Stacked NN with topology 6-6-2	124
6.6	The Observed value versus Predicted value of training data and validation data of stacked NN for network topology 6-6-4.	130
7.1	The Schematic Diagram of Combined Reforming of Methane Rig Set up	137

LIST OF SYMBOLS

ANOVA	-	Analysis of Variance
BET	-	Brunauer Emmett and Teller equation, m ² /g
Btu	-	British thermal unit
C1 Chemistry	-	Methane chemistry
C ₂	-	Acetylene, ethylene and ethane hydrocarbons
C ₂ ⁺	-	Higher Hydrocarbons
COE	-	Crude oil equivalent
CORM	-	Carbon dioxide reforming of methane
d	-	Pore Diameter, nm
Det	-	Detector
DME	-	Dimethylether
EIA	-	United States Energy Information Administration
F	-	Total flow rate pass through catalyst, mL/min
F-T	-	Fischer-Tropsch synthesis
GC	-	Gas chromatography
GHG	-	Greenhouse gas
GHSV	-	Gas hourly space velocity
GTL	-	Gas to liquid
HHCs	-	Higher Hydrocarbons
LPG	-	Liquefied petroleum gas
MTG	-	Methanol to gasoline
n	-	Mole
NA	-	Nitrogen absorption
OCM	-	Oxidative coupling of methane
P	-	Partial pressure

POM	-	Partial oxidation of methane
ppm	-	Part per million
r	-	Correlation Coefficient
r	-	Heating rate, ° C/min
RMSE		Root Mean Square Error
RT	-	Retention time
RWGS	-	Reverse water gas shift reaction
S	-	Selectivity, %
SASOL	-	Suid-Afrikaans Sintetiese Olie
SMDS	-	Shell middle distillate synthesis
SRM	-	Steam reforming of methane
T	-	Temperature
t	-	Time
TCD	-	Thermal conductivity detector
TOS	-	Time on stream
TPR	-	Temperature-programmed reduction
W	-	Weight of catalyst, g
WGS	-	Water gas shift reaction
X	-	Conversion, %
XRD	-	X-ray diffraction
Y	-	Yield, %
ZSM-5	-	Zeolite Socony Mobil number 5
ΔG	-	Gibbs free energy
ΔH	-	Heat, or enthalpy
ΔS	-	Entropy

LIST OF APPENDICES

APPENDIX	TITLE	PAGE
A	Gas Chromatography result for combined CORM and POM	147
B	Calculation method of reactant conversion and product yield and selectivity	149
C	Equilibrium thermodynamic Simulation program	151
D	A multiple input single output feed forward neural network for methane conversion program	152
E	A multiple input single output feed forward neural network for hydrogen selectivity program	155
F	A multiple input single output feed forward neural network for H ₂ /CO ratio program	158
G	A multiple input single output feed forward neural network for retraining structure program	161
H	A multiple input single output feed forward neural network for data combination structure program	172
I	A multiple input single output feed forward neural network for stacked structure program	177

CHAPTER 1

INTRODUCTION

1.1 General Introduction

C1 chemistry refers to the utilization of single carbon-bearing molecules, such as carbon monoxide, synthesis gas (a mixture of carbon monoxide (CO) and hydrogen (H₂), carbon dioxide (CO₂), methane (CH₄), and methanol (CH₃OH), for the production of valuable chemicals, petrochemical intermediates, and ultra-clean transportation fuels. C1 chemistry could also be used to produce high-purity hydrogen or premium chemicals from synthesis gas or from methanol. The main resources of C1 molecules are natural gas, coal, biomass, and organic wastes.

Figure 1.1 exhibits the profile of natural gas reserves versus natural gas consumption from 1980-1999. It shows that the rate of natural gas discovery is higher than the rate of natural gas consumption. Thus, it is necessary to develop a conventional natural gas utilization technologies not only for combustion to supply energy requirement, but also to convert the natural gas to higher value added products for example the production of environment friendly dimethyl ether (DME) as liquid fuel and also for ammonia production.

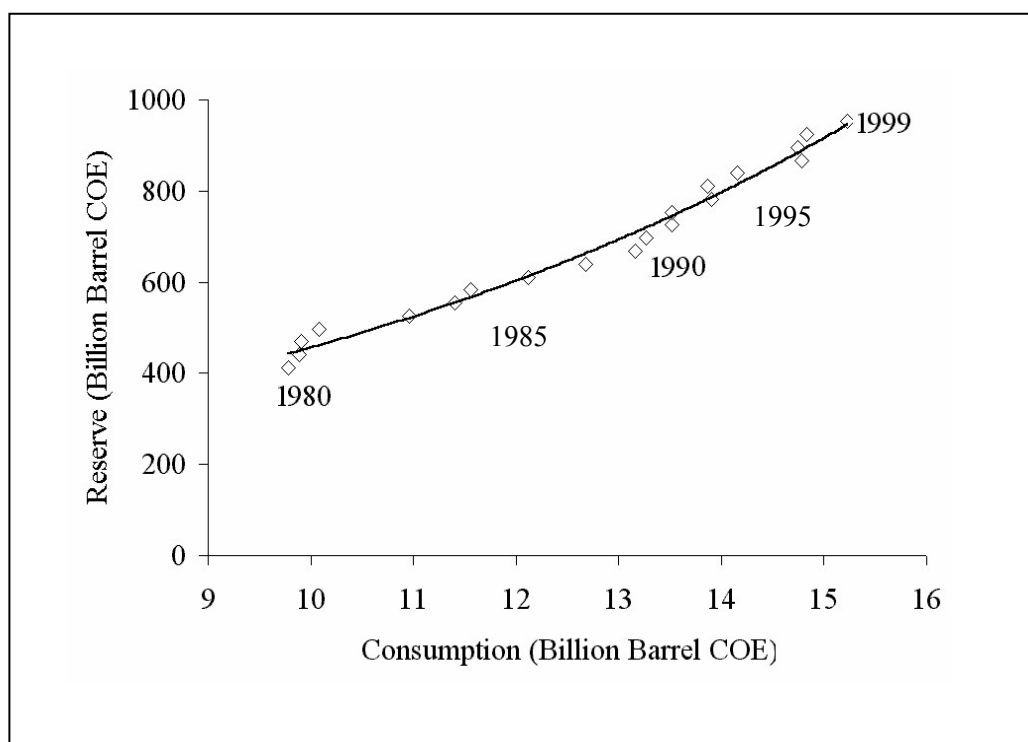


Figure 1.1 The world proven reserves of natural gas versus consumption from the year of 1980 to 1999 (United States Energy Information Administration, 2000).

COE = Crude Oil Equilibrium

Most natural gas reserves in Malaysia are situated at remote areas. As shown in Table 1.1, Malaysia has about 85.8 trillion standard cubic feet of natural gas and ranks as the 11th world's largest natural gas reserve (Sarmidi *et al.*, 2001). It is reported that Malaysia still has many remote gas reserves waiting for exploitation. (Kementrian Tenaga, Komunikasi dan Multimedia Malaysia, 1999). Besides that, the numbers of global proven natural gas reserves have increased by a factor of four over the last 25 years (Bitter, 1997).

Table 1.1: Gas reserves in Malaysia (Kementerian Tenaga, Komunikasi dan Multimedia Malaysia, 1999)

Region	Trillion standard cu ft
Peninsular Malaysia (Duyung, Jerneh, etc.)	34.4
Sabah	43.7
Sarawak	7.7
Total	85.8

- 1 trillion = 1×10^{12}
- 1 cu ft = 0.028317 cu meter

In addition, CO₂ is released to the atmosphere when solid waste, fossil fuels (oil, natural gas and coal) and wood are burned. Furthermore, CO₂ and CH₄ are also found in biogas and flue gas. For example, in a well designed chamber, biodecomposable materials in agricultures industries and waste water treatment process is decomposed to biogas and the gas mainly consisted of CH₄ and CO₂.

The main reason to utilize CO₂ is it can help to reduce the green house effect. Malaysia's greenhouse gas emissions totalled 144 million tonnes in term of CO₂ equivalent in 1994. Net emissions, after accounting for sinks of 68 million tonnes, amounted to 76 million tonnes CO₂ equivalent. On a per capita basis, net emissions amounted to 3.7 tonnes CO₂ equivalent. In terms of greenhouse gas, CO₂ accounted for 67.5% and nitrous oxide (N₂O) 0.1% of total CO₂ equivalent emissions. The fuel combustion energy sector accounted for 86.7% of the total CO₂ emissions and traditional biomass fuels accounted for 86.4% of the total N₂O emissions (Ministry of Science, Technology and Environment, 2000).

Synthesis gas is a very important petrochemical feedstock. It is the first step in a serial reaction to convert gas from hydrocarbons (e.g. natural gas, bio-gas etc.) into liquid form (e.g. gasoline, kerosene, diesel, petroleum wax etc) and higher value products (e.g. ammonia, alcohol etc). Synthesis gas can be produced via direct and indirect conversion. However, indirect conversion of natural gas to liquid fuels via synthesis gas is more efficient than scheme presently known for direct conversion

(Aasberg-Petersen *et al.*, 2001). There are three types of process involved in indirect conversion of natural gas; Steam Reforming of Methane (SRM), Partial Oxidation of Methane (POM) and Carbon Dioxide Reforming of Methane (CORM) as shown Table 1.2.

Table 1.2: Sources of synthesis gas and their applications (Ross *et al.*, 1996; Bitter, 1997)

	Reaction	$\Delta H_{R,298K}$ kJ/mol)	Applications
CORM	$\text{CH}_4 + \text{CO}_2 \Rightarrow 2\text{CO} + 2\text{H}_2$	+247	Oxoalcohols, (1.1) Polycarbonates, Formaldehyde
POM	$\text{CH}_4 + \frac{1}{2}\text{O}_2 \Rightarrow \text{CO} + 2\text{H}_2$	-36	Methanol (1.2) synthesis, Fischer-tropsch Synthesis
SRM	$\text{CH}_4 + \text{H}_2\text{O} \Rightarrow \text{CO} + 3\text{H}_2$	+206	H_2 production, (1.3) Ammonia Synthesis

Therefore, efficient and cheaper large scale process plant is required to accommodate the future use of synthesis gas. Although catalysis is only one of the several key factors for these developments, it plays an important role for advanced reactor designs, feedstock flexibility, and control of carbon formation. Thus, the catalysis of reforming reactions has long been a field of research in order to get industrial economical process output and applicable catalysis.

Today, the route to liquid, higher value products from gas is through conversion via synthesis gas (synthesis gas) as shown in Figure 1.2. The main gas-to-liquids GTL interest now is in Fischer–Tropsch (F–T) synthesis of hydrocarbons. While synthesis gas for GTL can be produced from any carbon-based feedstock such as hydrocarbons, coal, petroleum coke and biomass, the lowest cost routes to synthesis gas so far are based on natural gas. Thus, the focus for GTL has been largely on associated gas, so-called stranded or remotely located gas reserves, and larger gas reserves that are not currently being economically exploited.

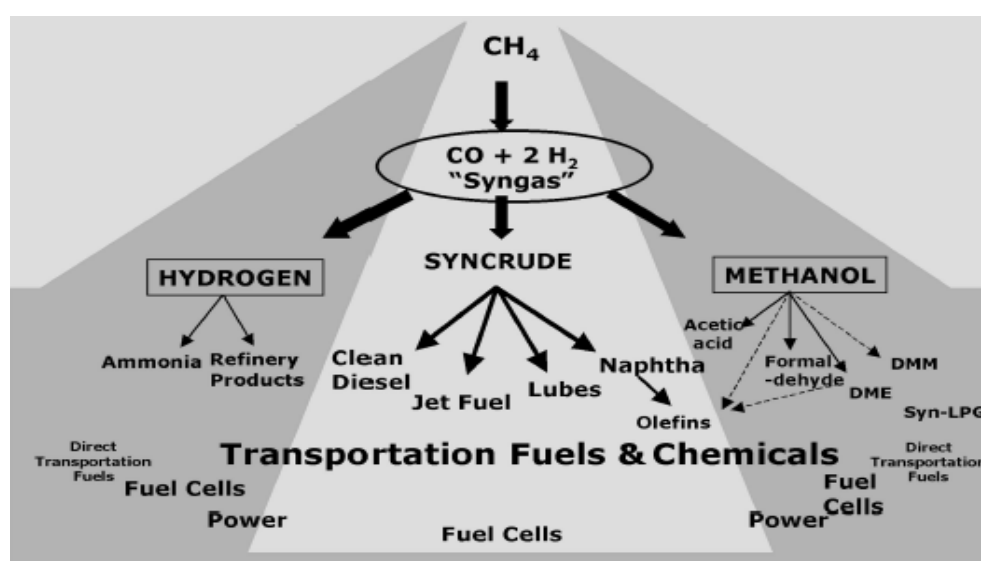


Figure 1.2 General application of synthesis gas (syngas) to higher value added product. (Fleisch *et al.*, 2002)

1.2 Research Background and Problem Statement

CORM is an interesting route of synthesis gas production. It is the most favourable due to its lower cost process compared to other catalytic reforming of methane processes due to the lower feedstock requirements (Ross *et al.*, 1996). Thus, many researchers have been involved in investigating to produce synthesis gas via catalytic CORM. The major obstacle preventing the commercialization of CORM is catalyst deactivation due to coking or sintering. Therefore, many

researchers modify on catalysts by loading the support with metal in order to reduce the catalyst deactivation. Not only that, combination reaction of methane reforming reaction such as SRM and POM or POM and CORM is energetically favourable when combining endothermic and exothermic reaction and produces a controllable H_2/CO ratio.

The most common catalyst support used for reforming reaction is $\gamma-Al_2O_3$. There are limited literatures about the usage of ZSM-5 and MgO catalyst in reforming reaction. ZSM-5 is a mixture of SiO_2 and Al_2O_3 . It has a unique pore structure and high surface area material. However, MgO is an orthogonal crystal structure with Mg and O elements and has a small surface area. Most of the Group VIII metals such as Fe, Ni, Ru, Rh and Pt are more catalytically active towards reforming reaction. However, Rh not only exhibits high activity and stability, but also has a high resistance to the coke formation on the catalyst surface (Mark and Meier, 1997). Chen *et al.* (1997) illustrated that Rh is the best promoter among other noble for the $Ni_{0.03}Mg_{0.97}O$. Wang and Au (1997) suggested that high metal surface material is potentially to have a higher degree of metal dispersion. In addition, Chin (2002) found that Rh metal loaded catalyst shows excellent reactivity and stability compared to Ni catalysts on ZSM-5 support. Furthermore, the Rh metal loaded on MgO support exhibits extraordinary catalyst reactivity and stability (Wang and Ruckenstein, 2000). Thus, the capability of Rh metal loaded on ZSM-5 and MgO support in CORM and combination reaction of CORM and POM are very interesting field to investigate.

The development of mathematical model that represent the process can be used to study the process or reaction behaviour. Nevertheless, traditional method of developing the mathematical model via kinetics study is cumbersome. A new application on modelling such as empirical model and artificial neural networks model development can be employed. Empirical model developed via experimental design techniques in order to interpret experimental data. Artificial neural network is a new non-linear statistical technique inspired by models developed for explanation of human brain activity. Even though, detailed mechanistic models are used much more often than empirical models or artificial neural network for interpretation of catalytic processes, experimental design and mathematical modelling techniques are

the mathematical tools normally used to optimize a process because they allow the gathering of maximum process information with reduced number of experiments (Larentis *et al.*, 2001).

The hypotheses of this research are:

1. combination reaction of CORM and POM shows better reactivity than CORM reaction itself.
2. neural networks model approach gives smaller error than empirical model and be able to predict the process output although with limited data.

1.3 Research Objectives

Computer technology such as experimental design and artificial neural network can be used to reduce time spent and experimental cost in research. Thus, the objectives of this research are:

1. to study the reactivity of Rh supported catalysts in CORM and combined CORM and POM reaction.
2. to develop model based on experimental design and artificial neural network
3. to apply neural network in predicting the catalyst performance in different reaction even with limited data available.

1.4 Scope of Study

The research focuses on modelling via different approaches in catalytic CORM in order to optimize the process conditions. The catalyst is prepared via wetness incipient impregnation. Rhodium (Rh) metal is used on ZSM-5 and MgO

supports. Catalyst testing is carried out in fixed bed 9.1 mm inner diameter and 30 cm long vertical quartz tube reactor. The effect of temperature, catalyst weight in 100 ml/min of total flow rate and ratio of O₂/CH₄ are studied. 16 experiments are carried out according to the arrangement set by the design of experiment. Three operating parameters are evaluated to reach the optimization point of methane conversion, hydrogen selectivity and H₂/CO ratio. The product of the combined CORM and POM is analyzed by TCD gas chromatography.

Two mathematical models approach are developed to represent the process and predict the methane conversion, hydrogen selectivity and H₂/CO ratio for best catalyst reactivity. The second order polynomial empirical model is developed via STATISTICA v.6 software using Response Surface Methodology. The normalized variables are employed to compare the relative importance of the model. A feed forward neural network is developed as well in which neural networks toolbox from Matlab V.6.1 is used to model and simulate the process. A network with one input layer, one or more hidden layers and an output layer consisting of nodes that interconnect the layers is composed. Both models are compared and analyzed via ANOVA.

The CORM and POM reaction data with different catalysts application are used to test the ability of neural network to predict a few reactions in a network. A feed forward neural networks is developed. Three different networks are constructed namely; retraining NN, data combination NN and stacked NN. Neural network toolbox from Matlab V 6.1 is used to model and simulate the process. These three networks are compared and analyzed via ANOVA.

Besides employment of Design of Experiments, attempt to utilize Genetic Algorithm (GA) on reforming of methane process for modelling and optimization purposes should be done. There is an interest to apply Genetic Algorithms which is well suited in the field of combinatorial catalysis to discover optimal catalyst compositions within a limited number of experiments (Rodemerck *et al.*, 2004). In addition, a hybrid process modelling and optimization of Neural Network and Genetic Algorithms may become appropriate in modelling process. (Nandi *et. al.*, 2003).

Many researchers usually neglect the thermal and pressure effects in the process while simulating the process. Thus, simulation work should consider isothermal and non-isothermal operation in order to analyze the role of heat management in the process. In addition, pressure drop should be considered to picture the real reforming process.

For an economically viable and environmentally benign process, utilisation of methane and carbon dioxide from biomass degradation plant or incineration plant can be used as a reactant even though Natuna Oil Field has a special composition of natural gas composition in the world in which 70% and 30% of the composition are carbon dioxide and methane respectively. However, H_2SO_4 in natural gas is still available even though in small amount. Thus, separation H_2SO_4 is required. So, an integrated process from biomass degradation or incineration plant implementing a system that recycle the unreacted reactant is also necessary to provide an effective and efficient process. Therefore, more studies need to be done before scaling it up to pilot scale process.

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