

# **Analysis of Least Square and Exponential Regression Techniques for Energy Demand Requirement (2013-2032)**

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**Abstract** This paper considered a long term electric power load forecast for twenty years (20 years) projection, in Nigeria power system using least-square regression and exponential regression model. The model is implemented in Matlab platform with a plot in residential load demand, commercial load demand and industrial load demand in (MW). In the quest for analysis and predicting the energy (power) demand (MW) requirement for a projection period of (2013 - 2032), data are collected between (2000 - 2012), from the Central Bank of Nigeria (CBN), and National Bureau of statistics (NBS). The results obtained shows that energy generated from the respective generating station including Egbin thermal power station Lagos, Sapele thermal power station etc. are grossly inadequate. This mismatch is a major problem in power system planning and operation. The result also shows that there is deviation between predicted energy demand (MW) and available power (or capacity allocated). The predicted energy demand into the projected future of 20years is 45 5,870.2MW. The paper work also extended the prediction form into: least-square, exponential regression model. Evidently, the comparism plot for linear and exponential model which shows similar predicting pattern: particularly least-square exhibit linear behavior, while exponential shows non-linear behaviour, the linear model gives more accurate result as compared to the exponential.

Keywords: exponential regression, least square, energy demand, load, energy, demand, long term forecast

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# 1. Introduction

The generating section are strategically located across the geopolitical-zone in Nigeria with different generating station capacity, people gradually drift from rural to urban cities which has led to an excessive demand of electricity due to the fast growing rate of industries, economic development and increasing population of the residence which has also led to epileptic power supply, power failure, fluctuation and total power outage. It has brought to a loss of energy utilization by the consumers and utility companies.

Peak load forecasting play major role in electrical power system operation, unit commitment and energy scheduling (Amin-Naseri and Soroush, 2008). Energy demand forecasting presents the firsts step in planning and developing for future generation, transmission and distribution facilities. One of the primary tasks of electric utility is to accurately predict energy demand requirements at all times, especially for a long term planning.

Based on the outcomes of such forecasts utilities coordinates their resources to meet the forecasted energy demand, thereby engaging a least-cost Energy management plan and follow-up which are subject to numerous uncertainty that is, in planning for future capacity resource needs an information and operation of the existing generation resources, in order to predicts future capacities and the power system serves one main function that is, to supply energy to the respective customers, which are residential, commercial and industrial consumers with electrical energy as economically and reliable condition. Another responsibility of power utilities is to recognize the needs of their customers Demand and the supply the necessary energies.

Evidently, limitations of energy resources in additions to environmental factors, requires electric energy to be used, more efficiently power plant and transmission lines to be constructed.

# 1.1. Aim of this Paper

The aim of this paper is to conduct the analysis of the load forecast and energy demand.

# **1.2.** Objectives of this Paper

- (i). The objective of this paper is to carry out using engineering techniques, to analyze the behaviour of the energy demand forecast with the aid of data collection.
- (ii). To investigate Energy demand profile of the existing capacity and the energy consumption pattern

- (iii). To analyse and forecast the energy demand response for a projection of 25 years ahead.
- (iv). To recommends to the regulatory agencies for implementing mismatches between the load allocations and the required capacity.

The study shall consider the consumption pattern for residential, commercial and industrial Energy demand forecast, for a projection of twenty-five years ahead.

## **1.3. Significance of the Paper**

The contents of this paper will be of great benefits particularly, to the electricity utilities, regulatory agencies and Nigeria at large, owing to the fact that development of electricity infrastructure is undoubtedly a capital intensive project that needs a serious attention. Hence, Energy demand forecast shall be taken as the first priority for future expansion planning program. Therefore to keep Nigeria abreast with other developing countries which have exhibited in every standard a substantial growth in economic development this means that, the existing gap between the electric power generation and energy demand requirement must be bridged.

# 2. Problem Statement

Sustainable supply of electric power is a prerequisite for energy generation, transmission and distribution to foster all forms of Economic development in the country. The Nigerian power system is not generating enough electric power, this inadequacy has led to:

- (i) Extra ordinary line losses,
- (ii) Load shedding,
- (iii) Failure and collapse of power system,
- (iv) Reduction in quality of electric power

Therefore there is need to overcome this challenges of poor supply of power to the end users at all times.

## 3. Material and Method

## 3.1. Least Square

The materials required for the analysis of electricity (demand) prediction in this paper is the load-capacity (allocation) and capacity-utilization data of previous years from Central Bank of Nigeria (CBN), National Beaureau of Statistics (NBS) and Power Holding Company of Nigeria (PHCN).

The paper strongly need to investigate the deviation of the capacity allocated to that of the capacity utilization on the view to analyse the rate of load consumption pattern with respect to capacity allocation by the Central Controlling Body: National Control Centre of Nigeria.

The analysis of demand forecasts are required for expansion, controlling and scheduling of power systems. The forecast help in determining the optimal-mix of generating capacities and which power plant to operate in a given period, so as to minimize costs and secure demand. The study is essential to be able to predict/forecast the quantity of power needed by Nigeria owing to the declining nature of the Nigerian power system supply and plan for future network expansion, in order to reduce cost of energy generation, stop load shedding and reduce power outages to minimum. Energy consumption data are collected for resident, commercial and industrial:

Table 1. Table of Energy	Consumption (MW)
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ENERGY CONSUMPTION (MW)					
Year	Industrial	Commercial	Residential	Total	
2000	1011.60	2346.00	4608.40	8688.90	
2001	1987.20	2439.00	7714.80	9034.40	
2002	1830.00	3297.60	7668.50	12842.40	
2003	1659.80	3583.00	7668.50	12866.60	
2004	1605.00	3830.30	7725.30	13160.60	
2005	1615.50	3851.00	7760.00	13226.60	
2006	1575.00	3900.80	7650.00	13125.80	
2007	1530.50	3915.00	7860.30	13305.80	
2008	1502.50	3852.00	7910.05	13264.55	
2009	1585.00	3865.50	8075.00	13525.50	
2010	1589.40	3925.80	8205.20	13720.40	
2011	1615.50	4004.70	8285.60	13905.80	
2012	1648.00	4025.40	8350.00	14023.40	

Source: Central Bank of Nigeria Statistical Bulletin and National Bureau of Statistics (NBS).

# Calculation and Analysis using Parabola (2<sup>nd</sup> degree polynomial)

This method employs the least-square technique used in developing a curve that describes the relationship between two or more variables. For example, capacity allocation (A), capacity utilization (U), and difference between the two capacities as errors (E) etc.

That is a given pair of polynomial data can be represented between two sets as:

$$Y = a_0 + a_1 x + a_2 x^2 + a_3 x^3 + \dots + a_n x_n \qquad (3.1)$$

This can be represented in other form as:

$$Y_{1} = a_{0} + a_{1}x_{1} + a_{2}x_{1}^{2} + a_{3}x_{1}^{3} + \dots + a_{n}x_{n}^{n}$$

$$Y_{2} = a_{0} + a_{1}x_{2} + a_{2}x_{2}^{2} + a_{3}x_{2}^{3} + \dots + a_{n}x_{n}^{n}$$

$$Y_{3} = a_{0} + a_{1}x_{3} + a_{2}x_{3} + \dots + a_{n}x_{n}^{n}$$

$$Y_{n} = a_{0} + a_{1}x_{1} + a_{2}x_{n}^{2} + a_{2}x_{n}^{3} + a_{3}x_{n}^{3} \dots - a_{n}x_{n}^{n}$$

$$(3.2)$$

On summing up the columns (equation. 3.2) we have:

$$\sum_{i=1}^{n} y_{1} = na_{0} + a_{1} \sum_{i=1}^{n} x_{1} + a_{2} \sum_{i=1}^{n} x_{1}^{2} + a_{3} \sum_{i=1}^{n} x_{i}^{3} + \dots + a_{n} \sum_{i=1}^{n} x_{n}$$
(3.3)

The above equation from the basis for the least-square method of  $2^{nd}$  – degree polynomial curve fit. Considering (equation 3.3), their increasing order of sequences, it can be placed in matrix formation e.g. Hence multiplying equation (3.3) by *x*:

$$\sum_{i=1}^{n} y_i x_i = a_0 \sum x_i + a_1 \sum x_i^2 + a_2 \sum x_i^3 + a_3 \sum x_i^4$$

$$\dots + a_n \sum x_i^{n+1}$$
(3.4)

• Multiply (3.4) by x:

$$\sum_{i=1}^{n} y_i \times_i^2 = a_0 \sum x_i^2 + a_1 \sum x_i^3 + a_2 \sum x_i^4 + a_3 \sum x_i^5 + (3.5)$$
$$\dots + a_n \sum x_i^{n+2}$$

Hence, in matrix form:

Since the equation of a straight line:

$$y = a + bx \tag{3.7}$$

This can be obtained using, determinate by matrix operation; using Crammer rule:

$$\begin{bmatrix} n & \sum x_i \\ \sum x_i & \sum x_i^2 \end{bmatrix} \begin{bmatrix} a_0 \\ a_1 \end{bmatrix} = \begin{bmatrix} \sum y_i \\ \sum y_i x_i \end{bmatrix}$$
(3.8)

Similarly,

$$a = \frac{\sum y_i \sum x_i^2 - \sum x_i (y_i x_i)}{n \sum x_i^2 - (\sum x_i)^2}$$
(3.9)

This means that,

$$\begin{vmatrix} n & \sum x_i \\ \sum x_i & \sum x_i^2 \end{vmatrix} a_0 = \begin{vmatrix} \sum y_i & \sum x_i \\ \sum y_i x_i & \sum x_i^2 \end{vmatrix}$$
(3.10)

Several methods or techniques are usually used and applied in the analysis of load forecasting for energy demand. The research work rely on the least-square and exponential regression techniques this is because of its numerous advantages, measurements error problems, the issues of large variation of data collection, missing observation of data etc. Therefore this work will consider the comparison between the two techniques that is preferred in the analysis.

The least square method is one of the mathematical tools used in developing a curve that describe the relationship between two variables. A polynomial of any degree can be established using least square method including the straight line form. A given pair of data can be represented by a polynomial that can best fit the relationship between two set of values, as given in the Table 2.

Table 2. A pair of Values

Х	$X_1$	$\mathbf{X}_2$	 X <sub>n-1</sub>	X <sub>n</sub>
Y	$\mathbf{Y}_1$	Y <sub>2</sub>	 $\mathbf{Y}_{n-1}$	y <sub>n</sub>

$$Y = a_0 + a_1 x + a_2 X^2 + a_3 X^3 + \dots + a_n X^n \quad (3.11)$$

If the above polynomial fits the pair of data, it means that every pair of data will satisfy the equation (polynomial).

$$\begin{split} Y_{1} &= a_{0} + a_{1}x_{1} + a_{2}X_{1}^{2} + a_{3}X_{1}^{3} + \dots + a_{n}X_{1}^{11} \\ Y_{2} &= a_{0} + a_{1}x_{2} + a_{2}X_{2}^{2} + a_{3}X_{2}^{3} + \dots + a_{n}X_{2}^{11} \\ Y_{3} &= a_{0} + a_{1}x_{3} + a_{2}X_{3}^{2} + a_{3}X_{3}^{3} + \dots + a_{n}X_{3}^{11} \\ Y_{n} &= a_{0} + a_{1}x_{n} + a_{2}X_{n}^{2} + a_{3}X_{n}^{3} + \dots + a_{n}X_{1}^{11} \end{split} (3.12)$$

Summing all the equations above, we have

$$\sum_{l=1}^{n} y_{l} = na_{o} + a_{l} \sum_{l=1}^{n} x_{l} + a_{2} \sum_{l=1}^{n} x_{l}^{2} +$$

$$\dots + a_{n} \sum_{l=1}^{n} x_{l}^{n}$$
(3.13)

Summation of the sign ( $\Sigma$ ) with i = 1. Multiplying equation 3.13 by x<sub>i</sub>, we obtain.

$$a_0 \sum x_i + a_1 \sum x_i^2 + a_2 \sum x_i^3 + a_3 \sum x_i^4 + \dots + a_n \sum x_i^{n+1}$$
(3.14)

Multiplying 3.14 by x<sub>i</sub>, we obtain

$$a_0 \sum x_i^2 + a_i \sum x_i^3 + a_2 \sum x_i^4 + a_3 \sum x_i^5 + ..... + a_n \sum x_i^{n+2} = \sum y_i x_i^2$$
(3.15)

For the nth time, the expression will be:

$$a_0 \sum x_i^n + a_i \sum x_i^{n+1} + a_2 \sum x_i^{n+2} + \dots + a_n \sum x_i^{n+n} = \sum y_i x_i^n$$
(3.16)

(n + 1) equation. In matrix form, this becomes

$$\begin{bmatrix} n & \sum x_{1} & \sum x_{1}^{2} & \sum x_{1}^{3} & \dots & \sum x_{1}^{n} \\ \sum x_{i} & \sum x_{1}^{2} & \sum x_{1}^{3} & \sum x_{1}^{4} & \dots & \sum x_{i}^{n+1} \\ \sum x_{1}^{2} & \sum x_{1}^{3} & \sum x_{1}^{4} & \sum x_{1}^{5} & \dots & \sum x_{1}^{n+2} \\ \sum x_{1}^{3} & \sum x_{1}^{4} & \sum x_{1}^{5} & \dots & \sum x_{1}^{n+3} \\ \sum x_{1}^{n} & \sum x_{1}^{n+1} & \sum x_{1}^{n+2} & \sum x_{1}^{n+3} & \sum x_{1}^{n+n} \end{bmatrix} \begin{bmatrix} a_{0} \\ a_{1} \\ a_{2} \\ a_{3} \\ a_{n} \end{bmatrix} = \begin{bmatrix} \sum y_{i} \\ \sum y_{i} x_{i} \\ \sum y_{i} x_{i}^{2} \\ \sum y_{i} x_{i}^{2} \\ \sum y_{i} x_{i}^{3} \\ \sum y_{i} x_{i}^{n} \end{bmatrix} (3.17)$$

We can now use the least square expression obtained in approximating the relationship between two variables using polynomial of any degree.

We can use the above expression 3.18 to approximate a straight line equation by extracting  $2 \times 2$  matrix from the main matrix as shown below:

$$\begin{bmatrix} n & \sum x_i \\ \sum x_i & \sum x_1^2 \end{bmatrix} a_0 = \begin{bmatrix} \sum yi & \sum x \\ \sum y_i & \sum x_i^2 \end{bmatrix}$$
(3.18)

The equation of a straight line is given by  $y = a_0 + a_1 x$ . We can apply any adequate method in solving the two equations in expression 3.18 to get the values of  $a_0$  and  $a_1$ . Applying crammer's rule, this becomes:

$$\begin{bmatrix} n & \sum x_i \\ & \\ \sum x_i & \sum x_1^2 \end{bmatrix} \begin{bmatrix} a_0 \\ a_0 \end{bmatrix} \begin{bmatrix} \sum y_i & \sum x_i \\ & \\ \sum y_i x_i & \sum x_1^2 \end{bmatrix}$$
(3.19)

$$a_0 = \frac{\sum y_1 \sum x_1^2 - \sum x_1 (\sum y_1 x_1)}{n \sum x_1^2 - (\sum x_1)^2}$$
(3.20)

$$\begin{bmatrix} n & \sum x_i \\ & \\ \sum x_i & \sum x_1^2 \end{bmatrix} \begin{bmatrix} a_1 = \end{bmatrix} \begin{bmatrix} n & \sum x_i \\ & \\ \sum x_i & \sum y_i x_i \end{bmatrix}$$
(3.21)

$$a_{1} = \frac{n \sum y_{1} x_{1} - \sum x_{1} \sum y_{1}}{n \sum x_{1}^{2} - (\sum x_{1})^{2}}$$
(3.22)

The polynomial of the second degree (quadratic equation) can be determined in a similar manner. The matrix formation for such is given as:

$$\begin{vmatrix} n & \sum x_i & \sum x_1^2 \\ \sum x_i & \sum x_1^2 & \sum x_1^3 \\ \sum x_1^2 & \sum x_1^3 & \sum x_1^4 \\ \end{vmatrix} \begin{vmatrix} a_0 \\ a_1 \\ a_2 \end{vmatrix} = \begin{vmatrix} \sum y_i \\ \sum y_i & x_i \\ \sum y_i & x_1^2 \end{vmatrix}$$
(3.23)

The matrix is 3 x 3 since the polynomial of the second degree is of the form

$$Y = a_0 + a_1 x + a_2 x^2 \tag{3.24}$$

### **Commercial Demand**

This analysis and procedure is repeated in the same manner for the case of residential and industrial load demand forecast.

Table 3. The Commercial Demand Analysis

Year	X	Commercial Demand (MW)y	xy	x <sup>2</sup>
2000	-6	2346.00	-14076.00	36
2001	-5	2439.00	-12195.00	25
2002	-4	3297.60	-13190.40	16
2003	-3	3583.00	-10749.00	9
2004	-2	3830.30	-7660.60	4
2005	-1	3851.00	-3851.00	1
2006	0	3900.80	0.00	0
2007	1	3915.00	3915.00	1
2008	2	3852.00	7704.00	4
2009	3	3865.50	11596.50	9
2010	4	3925.80	15703.20	16
2011	5	4004.70	20023.50	25
2012	6	4025.40	24152.40	36
Total	0	46836.10	21372.60	182

The gradient of the trend line

$$b = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2} = 117.43$$

Trend line value when x = 0:  $a = \frac{\sum y}{n} - \frac{b \sum x}{n} = 3602.77$  Trend equation Y = a + b = 3602.77 + 117.43x

The trend values from the above equation gives the values actual commercial demand and is given in the table below:

Year	Commercial Demand y(MW)	Trend value Y (MW)
2000	2346.00	2898.19
2001	2439.00	3015.62
2002	3297.60	3133.05
2003	3583.00	3250.48
2004	3830.30	3367.91
2005	3851.00	3485.34
2006	3900.80	3602.77
2007	3915.00	3720.20
2008	3852.00	3837.63
2009	3865.50	3955.06
2010	3925.80	4072.49
2011	4004.70	4189.92
2012	4025.40	4307.35
Total	46836.10	46836.01

#### • To calculate the accuracy of commercial forecast

The mean absolute deviation (MAD) = Actual - Forecast)

$$\frac{\sum Actual - Forecast)}{N} = 0.00692 \text{ MW}$$

Table 5. Table of Values for Commercial Demand Forecasted

Year	x	Industrial Demand (MW)y	xy	x <sup>2</sup>
2000	-6	1011.60	-6069.60	36
2001	-5	1987.20	-9936.00	25
2002	-4	1830.00	-7320.00	16
2003	-3	1659.80	-4979.40	9
2004	-2	1605.00	-3210.00	4
2005	-1	1615.50	-1615.50	1
2006	0	1575.00	0.00	0
2007	1	1530.50	1530.50	1
2008	2	1502.50	3005.00	4
2009	3	1585.00	4755.00	9
2010	4	1589.40	6357.60	16
2011	5	1615.50	8077.50	25
2012	6	1648.00	9888.00	36
Total	0	20755.00	483.10	182

Table 6. Trend Values for Residential Load Demand

Year	Residential Demand(MW) y	Trend Value Y (MW)
2000	4608.40	6691.83
2001	7714.80	6851.93
2002	7668.50	7012.03
2003	7668.50	7172.13
2004	7725.30	7332.23
2005	7760.00	7492.33
2006	7650.00	7652.43
2007	7860.30	7812.53
2008	7910.05	7972.63
2009	8075.00	8132.73
2010	8205.20	8292.83
2011	8285.60	8452.93
2012	8350.00	8613.03
Total	99481.68	99481.59

## • To calculate the accuracy of residential forecast

The mean absolute deviation (MAD)

$$=\frac{\sum Actual - Forecast)}{N} = 0.09/13 = 0.00692$$

### • The Predicted Residential Demand

The forecast value can be determined by adding the trend line value (160.10MW) to the preceding load demand to get the current years forecast demand also by using the trend equation.

The gradient of the trend line

$$b = \frac{n \sum xy - \sum x \sum y}{n \sum x^2 - (\sum x)^2} = 2.65$$
$$a = \frac{\sum y}{n} - \frac{b \sum x}{n} = 1596.53$$

Trend equation Y = a + bx = 1596.53 + 2.65x

The trend values and actual industrial demand are shown in Table 7 below;

Year	Industrial Demand y(MW)	Trend value Y(MW)
2000	1011.60	1580.63
2001	1987.20	1583.28
2002	1830.00	1585.93
2003	1659.80	1588.58
2004	1605.00	1591.23
2005	1615.50	1593.88
2006	1575.00	1596.53
2007	1530.50	1599.18
2008	1502.50	1601.83
2009	1585.00	1604.48
2010	1589.40	1607.13
2011	1615.50	1609.78
2012	1648.00	1612.43
Total	20755.00	20754.89

#### • To calculating the accuracy of industrial forecast

The mean absolute deviation (MAD)

$=\frac{\sum Actual}{\sum Actual}$	– Forecas N	<u>t)</u>	= 20755.00 -	20754.89/13
= 0.00846	MW			
Year 2000:	$\frac{4608.40}{99481.68}$	×	100% =	4.63%
Year 2001:	$\frac{7714.80}{99481.68}$	×	100% =	7.75%
Year 2002:	$\frac{7668.50}{99481.68}$	×	100% =	7.71%
Year 2003:	$\frac{7668.50}{99481.68}$		100% =	7.71%
Year 2004:	$\frac{7781.30}{99481}$	×	100% =	7.77%
Year 2005:	$\frac{7760.00}{99481}$	×	100% =	7.80%
Year 2006:	7650.00 99481.68	×	100% =	7.69%

Year 2007: $\frac{7860.30}{99481.68}$	×	100% =	7.90%
Year 2008: 7910.08/99481.68	×	100% =	7.95%
Year 2009: $\frac{8075.00}{99481.68}$	×	100% =	8.12%
Year 2010: $\frac{8205.00}{99481.68}$	×	100% =	8.25%
Year 2011: $\frac{8285.60}{99481.68}$	×	100% =	8.33%
Year 2012: $\frac{8350.00}{99481.68}$	×	100% =	8.39%

#### Table 8. Presentation of Residential Demand Contribution (%)

Year	Residential Energy Demand (MW)	Percentage of Residential Demand Contribution (%)	
2000	4608.40	4.63	
2001	7714.80	7.75	
2002	7668.50	7.71	
2003	7668.50	7.71	
2004	7725.30	7.77	
2005	7760.00	7.80	
2006	7650.00	7.69	
2007	7860.30	7.90	
2008	7910.08	7.95	
2009	8075.00	8.12	
2010	8205.00	8.25	
2011	8285.60	8.33	
2012	8350.00	8.39	
Total	99481.68	100	

#### **Regression Exponential Analysis Method**

In the case of least-square method, approximation are conducted between two variables, using polynomial of any degree (linear, quadratic, cubic etc). The two variables are expected to fit the function to the set of data, when the predication equation is given as:

$$y_t = a_0 + a_1 x$$
 (3.7)

Where:

 $a_0$  = specify intercepts

 $a_1$  = specific slope

 $y_t$  = specify dependent variable

x = independent variable

 $y_t$ : The estimated trend value for a given period t,

By the application of least square regression through Crammers rule, the constants, a and b can be determined from equation (3.25)

$$\begin{pmatrix} n & \sum x_i \\ \sum x_i & \sum x_i^2 \end{pmatrix} \begin{pmatrix} a_0 \\ a_1 \end{pmatrix} = \begin{pmatrix} \sum y_i \\ \sum y_i x_i \end{pmatrix}$$
(3.25)

Which are obtained from the matrix formation as:

$$\begin{array}{ccc}
n & \sum x_i \\
\sum x_i & \sum x_i^2
\end{array} | a_0 = \begin{vmatrix} \sum y_i & \sum x_i \\
\sum y_i x_i & \sum x_i^2 \end{vmatrix} \tag{3.26}$$

$$a_{0} = \frac{\sum y_{i} \sum x_{i}^{2} - \sum x_{i} (y_{i} x_{i})}{n \sum x_{i}^{2} - (\sum x_{i})}$$
(3.27)

or

$$a_0 = \frac{\sum y_i}{n} - \frac{b\sum x}{n} \tag{3.28}$$

Similarly,

$$\begin{vmatrix} n & \sum x_i \\ \sum x_i & \sum x_i^2 \end{vmatrix} a_1 = \begin{vmatrix} n & \sum y_i \\ \sum x_i & \sum y_i x_i \end{vmatrix}$$
(3.29)

or

$$a_{1} = \frac{n \sum y_{i} x_{i} - \sum x_{i} \sum y_{i}}{n \sum x_{i}^{2} - \left(\sum x_{i}\right)^{2}}$$
(3.30)

or

$$a_{1} = \frac{n \sum y_{i} x_{i} - \sum x_{i} \sum y_{i}}{n \sum x_{i}^{2} - \left(\sum x_{i}\right)^{2}}$$
(3.31)

The least-square method is used in fitting trend line to a time series. The technique is one of the basic tools used in developing a curve that describes the relationship between variables. This technique can be applied to any *n*-degree polynomials.

#### • Exponential Regression Analysis

Considering the curve behaviour, while plotting capacity allocation, capacity utilization and error, the curve of the data trend is non-linear in nature. Therefore

there is need to apply the non-linear model: Exponential regression model which can give a good result between the exponential relationship of the variables considered in the study.

This is the therefore achieved by estimating the baseload and the annual growth rate, given as:

$$Y_1 = A\ell^{Bx} \tag{3.32}$$

Converting equation (3.32) to a straight line on (linear), we take a rational logarithms of both sides of equation (3.32).

That is,

$$In Y = In \ a + bx \tag{3.33}$$

To solve 'a' and 'b', make a summation to both sides,

$$\sum InY = \sum Ina + b\sum x \tag{3.34}$$

Multiply, equation (3.34) by variable "x" Hence

$$\sum x \ln Y = n \ln a \sum x + b \sum x^2$$
(3.35)

From equation (3.34) and (3.35) form your matrixfunction trough crammer's rule, we have:

$$\begin{pmatrix} n & \sum x \\ \sum x & \sum x^2 \end{pmatrix} \begin{pmatrix} \ln a \\ b \end{pmatrix} = \begin{pmatrix} \sum \ln y \\ \sum x \ln y \end{pmatrix}$$
(3.36)

# 3.2. Using Exponential Regression Analysis Method

• To determine the values of 'a' and "b" respectively, according to the given equations.

YEAR (N)	Year Index (X)	(X <sup>2</sup> )	Residential (Y) capacity allocation (MW)	In Y	X In Y	XY
2000	-6		4608.40	8.4356	-506136	-27650.40
2001	-5	36	7714.80	8.9508	-44.754	-38574.0
2002	-4	25	7668.50	8.9449	-35.779	-30674.00
2003	-3	16	7668.50	8.9449	-26.8347	-23005.50
2004	-2	9	7720.30	8.95160	-17.9032	-15450.60
2005	-1	4	7760.00	8.9567	-8.9567	-7760.00
2006	0	0	7650.00	8.9425	0	0.00
2007	1	1	7860.30	8.9696	8.969	7860.30
2008	2	4	7910.08	8.9759	17.9518	15820.16
2009	3	9	8075.00	8.9965	26.9815	24225.00
2010	4	16	8205.20	9.01252	36.050008	32826.80
2011	5	25	8285.60	9.02227	45.11135	41428.00
2012	6	36	8350.00	9.030	54.18	50100.00
	$(\Sigma X) = 0$	$(\Sigma X^2) = 182$	$\Sigma Y = 99481.68$	$\Sigma Y = 99481.68$	$\Sigma X \text{ In } Y = 116.13379$	ΣXY=29139.76

 Table 9. Parameter determination using Analysis of Exponential Regression

Substituting, the values obtained in the Table 9 into equation (3.36)

$$\begin{pmatrix} n & \sum x_i \\ \sum x_i & \sum x_i^2 \end{pmatrix} \begin{pmatrix} \ln a \\ b \end{pmatrix} = \begin{pmatrix} \sum \ln y \\ \sum x \ln y \end{pmatrix}$$
(3.36)

Where:

$$(\sum x_i) = 0,$$
  
 $(\sum x_i^2) = 182$   
 $(\sum y_i) = 99481.68$   
 $(\sum x_i y_i) = 29139.76$   
(3.37)

$$\sum \ln y = 116.13379$$
  

$$\sum x \ln y = 4.410458$$
  

$$n = 13$$
  
(13 0) (ln a) (116.13379)  
(3.38)

$$\begin{pmatrix} 13 & 0 \\ 0 & 182 \end{pmatrix} \begin{pmatrix} \text{III} & a \\ b \end{pmatrix} = \begin{pmatrix} 110.13379 \\ 4.410458 \end{pmatrix}$$

The system equation becomes:

$$13 \times \ln a + 0 \times b = 116.13379 \tag{3.39}$$

$$0 \times \ln a + 182 \times b = 4.410458 \tag{3.40}$$

1 ( 12270

$$13Ina = 110.13379$$
 (3.41)

(2 11)

$$182b = 4.410458 \tag{3.42}$$

or

$$b = \frac{4.410458}{182} \tag{3.43}$$

$$b = 0.02423$$
 (3.44)

$$\frac{13\ln a}{13} = \frac{116.13379}{13} \tag{3.45}$$

$$In a = 8.9333$$
 (3.46)

Therefore,

$$Y = A\ell^{BX}$$

$$In Y = In A + Bx$$
(3.47)

or

$$InY = 8.933 + 0.02423X \tag{3.48}$$

$$Y = \ell \times p \ (8.933 + 0.02423X) \tag{3.49}$$

But,

$$\ell \times p \ (8.933) = 7570.95 \tag{3.50}$$

Therefore,

$$a = \underline{7570.95MW} \tag{3.51}$$

Hence, substituting the values of "a" and "b" into the equation as:

$$Y = A\ell^{BX} = 7570.95\ell^{0.02423x} \tag{3.52}$$

$$In \ y = In \ a + bx \tag{3.53}$$

But, slope is the antilog of the value (0.02423) = 1.02452

Hence,  $b = (1.02452 - 1.0) = 0.02452 \times 100\%$ 

$$b = 2.452\% \cong 2.5\% \tag{3.54}$$

Therefore, b = growth rate = 2.5%

$$Y = 7570.95 + 2.5x \tag{3.55}$$

This means substituting values of "x" into the relation y = a + bx to obtain other value of y.

# The predicted load "Y", using exponential regression function $Y = A\ell^{BX}$ are presented as:

 $Y = A\ell^{BX}$ , the predicted value will be: Y = a + bxWhen growth rate b  $= \frac{2.5}{100} \times 7570.95 = 189.2738MW$ 

For 2013 - New predicted value (MW) = 7570.95 + 189.2738 = 7769.22MW For 2014 - New predicted value (MW) = 7760.22 +189.2738 = **7949.497W** For 2015 - New predicted value (MW) = 7949.498 + 189.2738 = **8138.7718MW** For 2016 - New predicted value (MW) = 8138.7718 +189.2738 = **8328.0456MW** For 2017 - New predicted value (MW) = 8328.0456 + 189.2738 = **8517.319MW** For 2018 - New predicted value (MW) = 8706.5932MW For 2019 - New predicted value (MW) = 8706.5932 + 189.2738 = **8895.867MW** For 2020 - New predicted value (MW) = 8895.867 + 189.2738 = 9085.1408MW For 2021 - New predicted value (MW) = 9085.1408 + 189.2738 = **9274.4146MW** For 2022 - New predicted value (MW) = 9274.4146 +189.2738 = **9463.6884MW** For 2023 - New predicted value (MW) = 9463.6884 + 189.2738 = **9652.9622MW** For 2024 - New predicted value (MW) = 9652.9622 +189.2738 = **9842.236MW** For 2025 - New predicted value (MW) = 9842.236 + 189.2738 = **10031.5098MW** For 2026 - New predicted value (MW) = 10031.5098 +189.2738 = **10220.7836MW** For 2027 - New predicted value (MW) = 10220.7836 + 189.2738 = **10410.0574MW** For 2028 - New predicted value (MW) = 10410.0574 + 189.2738 = 10599.3312MW For 2029 - New predicted value (MW) = 10599.3312 + 189.2738 = **10788.605MW** For 2030 - New predicted value (MW) = 10788.605 +189.2738 = 10977.8788MWFor 2031 - New predicted value (MW) = 10977.8788 + 189.2738 = **11167.1526MW** For 2032 - New predicted value (MW) = 11167.1526 + 189.2738 = **11356.4264MW** 

**Predicted load (Y) in the case of using least-square method** (Y = a + bx)

The gradient of the trend-line:

Y = a + bx

Where a = 7652.43MWb = 160.10 (which is 2.092% of the growth rate) That is, Y = 7652.43 + 160.10x $\frac{2.092}{2} \times 7652.43 MW = 160.10$ 100 =160.10MW (capacity addition) For 2013 - New predicted value: =  $7652.43 + 2.093\% \times$ 1 = 7812.53 MW For 2014 - New predicted value: =  $7652.43 + 2.093\% \times$ 2 = 7972.63 MW For 2015 - New predicted value: =  $7652.43 + 2.093\% \times$ 3 = 8132.73MW For 2016 - New predicted value: =  $7652.43 + 2.093\% \times$ 4 = 8292.83MW For 2017 - New predicted value: =  $7652.43 + 2.093\% \times$ 5 = 8452.93 MW For 2018 - New predicted value: =  $7652.43 + 2.093\% \times$ 6 = 8613.03 MW For 2019 - New predicted value: =  $7652.43 + 2.093\% \times$ 7 = 8773.13MW For 2020 - New predicted value: =  $7652.43 + 2.093\% \times$ 8 = 8933.23MW For 2021 - New predicted value: =  $7652.43 + 2.093\% \times$ 9 = 9093.33MW For 2022 - New predicted value: =  $7652.43 + 2.093\% \times$ 10 = 9253.43 MW For 2023 - New predicted value: =  $7652.43 + 2.093\% \times$ 11 = 9413.53 MW For 2024 - New predicted value: =  $7652.43 + 2.093\% \times$ 12 = 9573.63 MW For 2025 - New predicted value: =  $7652.43 + 2.093\% \times$ 13=9733.73MW For 2026 - New predicted value: =  $7652.43 + 2.093\% \times$ 14 = 9893.83MW For 2027 - New predicted value: =  $7652.43 + 2.093\% \times$ 15 = 10053.93MW For 2028 - New predicted value: =  $7652.43 + 2.093\% \times$ 16 = 10214.03 MW For 2029 - New predicted value: =  $7652.43 + 2.093\% \times$ 17 = 10374.13 MW For 2030 - New predicted value: =  $7652.43 + 2.093\% \times$ 18 = 10534.23MW For 2031 - New predicted value: =  $7652.43 + 2.093\% \times$ 19 = 10694.33MW

For 2032 - New predicted value: = 7652.43 + 2.093% × 20 = 10854.43MW

Modified form of Exponential Regression Analysis

$$Y = A\ell^{BX} \tag{3.56}$$

$$In Y = In A + In\ell^{Bx}$$
(3.57)

or

$$In Y = In A + Bx \tag{3.58}$$

Therefore the modified form of exponential demand equation may be expressed as:

$$Y = P_{D_i} = \ell^{a+b} \left( x_i - x_0 \right)$$
(3.59)

or

 $Y = P_{D_i} = \ell^{a+bX_i} \tag{3.60}$ 

Where :

 $X_i$  becomes;

$$X_i = x_i - x_0 \tag{3.61}$$

Now, taking natural log of equation (3.5): We obtain as:

$$\ln P_{Di} = \ln\left(\ell^{a+bX_i}\right) \tag{3.62}$$

or

or

 $In P_{Di} = a + bX_i \tag{3.63}$ 

 $Y_i = In P_{D_i}$  $Y_i = a + bX_i$ (3.64)

Where:

$$Y_i = In P_{D_i} \tag{3.65}$$

The utilization load demand is represented as expected demand while allocated load is represented as available power (MW):

For considering the historical data for energy demands for consecutive year are

$$P_{D_1}, P_{D_2}, P_{D_3} \dots P_{D_n}$$
 (3.66)

$$Y_1 = In P_{D_1} \tag{3.67}$$

$$Y_2 = In P_{D_2}$$
  
:  
$$Y_n = In P_{D_n}$$
 (3.68)

• In order to predicts the demand correctly, the sum of square of error should be minimum:

Summation 
$$(S) = \sum_{i=1}^{n} \left[ Yi - (a+b \times i) \right]^2$$
 (3.69)

• For S to be minimum, the conditions are:

$$\frac{\partial s}{\partial a} = 0 \text{ and } \frac{\partial s}{\partial a} = 0$$
 (3.70)

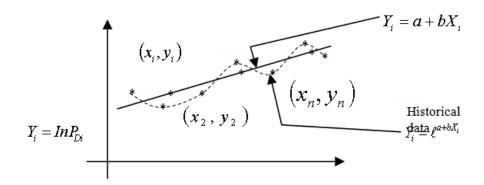


Figure 1. Showing the historical data of linear and non-linear energy demand

Differentiating equation (3.14) with respect to 'a' we or have:

$$O = \frac{\partial s}{\partial a} = \sum_{i=1}^{n} 2 \left[ Yi - (a + bX_i) \right] (-1) = 0$$

or

$$O = 2\sum_{i=1}^{n} Y_{i} = 2\sum_{i=1}^{n} (a + bX_{i})$$

$$\sum_{i=1}^{n} Y_{i} = \sum_{i=1}^{n} (a + bX_{i})$$
(3.71)

$$\sum_{i=1}^{n} Y_i = na + b \sum_{i=1}^{n} X_i$$
(3.72)

• Similarly, differenting equation 3.14) with respect to 'b' we have:

$$\frac{\partial s}{\partial a} = \sum_{i=1}^{n} 2 \left[ Yi - (a+b \times i)(-1)X_i \right] = 0$$
(3.73)

or

$$\sum_{i=1}^{n} Y_i X_i = \sum_{i=1}^{n} (a+b \times i) x_i$$
(3.74)

$$\sum_{i=1}^{n} Y_i X_i = a \sum_{i=1}^{n} X_i + b \sum_{i=1}^{n} X_i^2$$
(3.75)

• Hence, the conditions for the sum of least square of the error (deviation), to be minimum are given by the two equations:

Let us consider

$$\sum_{i=1}^{n} X_i = 0, \tag{3.76}$$

in order to determine 'a' and 'b'

or

• Using the equation (3.21) into equation (3.17) and (3.20):

We have:

$$a_{New} = \frac{1}{n} \sum_{i=1}^{n} Y_{i}$$

$$\sum_{i=1}^{n} Y_{i} = na + b \sum_{i=1}^{n} X_{i} = 0$$
(3.77)

Similarly,

$$b_{New} = \frac{\sum_{i=1}^{n} Y_i X_i}{\sum_{i=1}^{n} X_i^2}$$

 $a = \frac{1}{n} \sum_{i=1}^{n} Y_i$ 

or

or

$$\sum_{i=1}^{n} Y_i X_i = a \sum_{i=1}^{n} X_i + b \sum_{i=1}^{n} X_i^2 = 0$$
(3.78)

$$b = \frac{\sum_{i=1}^{n} Y_i X_i}{\sum_{i=1}^{n} X_i^2}$$

• Hence for load prediction we have:

$$P_{D_i} = \ell^{a+b(x_i - x_0)} = \ell^{a+bX_i}$$
(3.79)

Where:

 $\ell = 2.718$ 

*a* : intercept

*b* : slope or gradients

 $x_i$ : projected look ahead period

 $X_{i_{\ensuremath{New}}}$  : difference between projected year and base year (MW)

 $x_0$ : base year (reference year) = 2006

B: base – MW =  $1 \times 10^3 = 1000 MW$ where: where n = 13

$$a = \frac{1}{n} \sum_{i=1}^{n} Y_i = \frac{1}{13} \sum_{i=1}^{13} Y_i$$
$$= \frac{1}{13} \times 26.333456$$
$$a = 2.02565$$

and

$$b = \frac{\sum_{i=1}^{n} Y_i x_i}{\sum_{i=1}^{n} x_i^2} = \frac{\sum_{i=1}^{13} Y_i x_i}{\sum_{i=1}^{13} x_i^2}$$
$$b = \frac{-1.754837982}{182}$$
$$b = -9.641966934 \times 10^{-3}$$
$$b = 0.02422$$

Modified Form of Exponential Regression Calculation and Analysis for load demand for the year 2013, when our base year is equal 2006

**Case 1: Residential** 

$$Y_i = PD_i = \ell^{a+b(X_i)}$$
$$Y_i = PD_i = \ell^{a+b(x_i - x_0)}$$

Where  $X_i = (x_i - x_0)$ 

- But  $x_i$  = predicted energy demand = 2013.
- $x_0 = based year = 2006$

Note:  $\ell = 2.718$ 

We can find our peak load demand from

$$PD_i = 1000$$

But  $PD = 1000\ell^{a+b(x_i-x_0)}$ That is when  $x_0 = 2006$  (datum case) or base year  $x_i = 2013$  (prediction case) Hence;  $PD_{2013} = 1000 \ell^{2.02565+0.0130249(2013-2006)}$ 

$$PD_{2013} = 1000 \ell^{2.02565 + 0.0130249(7)}$$
$$PD_{2013} = 1000 \ell^{2.02565 + 0.0911743}$$

$$PD_{2013} = 1000\,\ell^{2.1168243}$$

or

 $\begin{aligned} PD_{2013} &= 1000 \times 2.718^{2.01168243} \\ PD_{2013} &= 1000 \times 8.302899728 \\ PD_{2013} &= 8302.899728 \\ PD_{2013} &= 8302.899728 \\ PD_{2013} &= 8305MW \ capacity \end{aligned}$ 

### Modified Form of Exponential Regression Calculation and Analysis For load demand for the year 2014, when our base year is equal 2006

That is; When  $x_0 = 2006$  (datum case) or base year  $x_i = 2014$  (prediction case) Therefore;  $PD_{2014} = 1000 \ell^{2.02565 + 0.0130249(2014 - 2006)}$   $PD_{2014} = 1000 \ell^{2.02565 + 0.0130249(8)}$  $PD_{2014} = 1000 \ell^{2.02565 + 0.1041992}$ 

$$PD_{2014} = 1000 \,\ell^{2.1298492}$$
  
But  $\ell = 2.718$   
 $2.718^{2.1298492}$ 

 $PD_{2014} = 1000 \times 8411.74016$ 

 $PD_{2014} \approx 8415 MW \ capacity$ 

This process is continued in the same similar manner

Total summation of predicted load in the case of residential load (2000 – 2012)

That is,

 $\Sigma$  predicted residential energy demand = (7009.723+7101.6150 + 7194.708 + 7289.02 + 7384.571 + 7481.37 + 7579.44+7678.80 + 7779.46+7881.44 +7984.76+ 8089.43+ 8195.47)

$$\sum_{predicted} = 98.649,807MW$$

Similarly, the actual allocated Residential load between (2000 – 2012) becomes:

$$\begin{split} \Sigma_{Actual} &= (4608.40 + 7714.80 + 7668.50 + 7668.50 + \\ 7725.30 + 7760.00 + 7650.00 + 7860.30 + 7910.05 + \\ 8075.00 + 8205.20 + 8285.60 + 8350.00) = \end{split}$$

$$\sum_{Actual} = 99.481.65MW$$
$$M_{AD} = \frac{\sum (Actual - Forecast)}{N}$$
$$= \frac{99481.65 - 98649.807}{13}$$
$$= \frac{831.843}{13}$$
$$= 63.9879MW$$

#### Case 1: Validation: Residential load demand

Capacity Allocation @ 2000 = 4608.40MW (or available)

Load forecast (prediction) @ 2000;

$$Y_{i,2000} = P_{D_{2000}} = \ell^{a+b(x_i)}$$

$$Y_{i,2000} = P_{D_{2000}} = 2.718^{2.02565(1.7548379821)(-6)}$$

$$Y_{i,2000} = P_{D_{2000}} = 2.718^{2.025187111+10.52902789(x_i-x_0)}$$
Where  $x_i = x_i - x_0$   
 $a = 2.025187111, \ b = -1.7548379821$ 

 $x_0 = \text{base year} = 2006$ 

$$x_i = \text{predicted year} = 2000$$

$$Y_{i,2000} = P_{D_i,2000} = 283049.2437MW$$

### @ 2001 (Prediction)

 $P_{D2001} = 1000\ell^{2.025187111 + -1.7548379821(-5)}$ 

or

$$P_{D2001} = 1000\ell^{(2.025187111+8.77418991)}$$

$$P_{D2001} = 1000\ell^{10.79937702}$$

or

 $P_{D2001} = 1000 \times 2.718^{10.79937702}$ 1000×48935.44671  $P_{D2001} = 48935446.17MW$ @ 2002 (Prediction)  $P_{D2002} = 1000\ell^{2.025187111 + -1.7548379821(2002.-2006)}$ or  $P_{D2002} = 1000\ell^{2.025187111 + -1.7548379821(-4)}$ or  $P_{D2002} = 1000\ell^{2.025187111+7.019351928}$ or  $P_{D2002} = 1000\ell^{9.044539039}$ or  $P_{D2002} = 1000 \times 2.718^{9.044539039}$  $P_{D2002} = 1000 \times 8464204.057 MW$ or  $P_{D2002} = 8464204.057MW$ @ 2003 (Prediction)  $P_{D2003} = 1000\ell^{2.02518711 + -1.7548379821(2003. -2006)}$ or  $P_{D2003} = 1000\ell^{2.02518711 + -1.7548379821(-3)}$ or  $P_{D2003} = 1000 \times 2.718^{(2.025187111+5.264513946)}$ or  $P_{D2003} = 1000 \times 2.718^{7.289701057}$ or  $P_{D2003} = 1000 \times 2.718^{1464.025672}$  $P_{D2003} = 1000 \times 1464.025672$ or  $P_{D2003} = 1464025.672MW$ @ 2004 (Prediction)  $P_{D2004} = 1000\ell^{2.02518711 + -1.7548379821(2004. -2006)}$ or  $P_{D2004} = 1000 \times 2.718^{2.02518711 + -1.7548379821(-2)}$ or  $P_{D2004} = 1000 \times 2.718^{2.02518711 + 3.509675964}$ or  $P_{D2004} = 1000 \times 2.718^{5.534863075}$ or  $P_{D2004} = 1000 \times 253.2277286$ or  $P_{D2004} = 253227.7286MW$ @ 2005 (Prediction)  $P_{D2005} = 1000\ell^{2.025187111 + -1.7548379821(2005. -2006)}$ or  $P_{D2005} = 1000\ell^{2.025187111 + -1.7548379821}$ 

 $P_{D2005} = 1000\ell^{2.621344574}$ or  $P_{D2005} = 1000 \times 2.718^{2.621344574}$  $P_{D2005} = 1000 \times 13.75046693$ or  $P_{D2005} = 13750.46693MW$ @ 2006 (Prediction)  $P_{D2006} = 1000\ell^{2.025187111 + -1.7548379821(2006. -2006)}$ or  $P_{D2006} = 1000\ell^{2.025187111 + -1.7548379821(-0)}$ or  $P_{D2006} = 1000\ell^{2.025187111+0}$ or  $P_{D2006} = 1000 \times 2.718^{2.025187111}$ or  $P_{D2006} = 1000 \times 7.581120857$ or  $P_{D2006} = 7581.120857MW$ @ 2007 (Prediction)  $P_{D2007} = 1000\ell^{2.025187111 + -1.7548379821(2007. -2006)}$ or  $P_{D2007} = 1000\ell^{2.025187111 + -1.7548379821(1)}$ or  $P_{D2007} = 1000\ell^{0.270349129}$ or  $P_{D2007} = 1000 \times 2.718^{0.270349129}$ or  $P_{D2007} = 1000 \times 1.310385145$  $P_{D2007} = 1310.385145MW$ @ 2008 (Prediction)  $P_{D2008} = 1000 \times \ell^{2.025187111 + -1.7548379821(2008. -2006)}$ or  $P_{D2008} = 1000 \times 2.718^{2.025187111 + -1.7548379821(2)}$ or  $P_{D2008} = 1000 \times 2.718^{2.025187111 + 3.509675964}$ or  $P_{D2008} = 1000 \times 2.718^{-1.484488853}$ or  $P_{D2008} = 1000 \times 2.718^{-1.484488853}$ or  $P_{D2008} = 1000 \times 0.226653029$  $P_{D2008} = 226.6530296MW$ @ 2009 (Prediction)

 $P_{D2009} = 1000 \times \ell^{2.025187111 + -1.7548379821(2009. -2006)}$  or

$$P_{D2009} = 1000 \times \ell^{2.025187111 + -1.7548379821(3)}$$
 or 
$$P_{D2009} = 1000 \times \ell^{2.025187111 + -5.264513946}$$

or

 $P_{D2009} = 1000 \times 2.718^{-3.239326835}$ 

or

$$\begin{split} P_{D2009} &= 1000 \times 0.039203432 \\ P_{D2009} &= 39.20343268 MW \end{split}$$

## @ 2010 (Prediction)

 $P_{D2010} = 1000 \times \ell^{2.025187111 + -1.7548379821(2010. -2006)}$  or

 $P_{D2010} = 1000 \times \ell^{2.025187111 + -1.7548379821(4)}$  or

$$P_{D2010} = 1000 \times \ell^{2.025187111 + -1.7548379821}$$

or

```
P_{D2010} = 1000 \times \ell^{-4.994164817}
```

or

 $P_{D2010} = 1000 \times 2.718^{-4.994164817}$ 

 $P_{D2010} = 1000 \times 6.780889434 \times 10^{-3}$ 

or

or

 $P_{D2010} = 6.780889434MW$ 

## @ 2011 (Prediction)

 $P_{D2011} = 1000 \times \ell^{2.025187111+-1.7548379821(2011.-2006)}$ or  $P_{D2011} = 1000 \times \ell^{2.025187111+-1.7548379821(5)}$ or  $P_{D2011} = 1000 \times \ell^{2.025187111+-8.77418991}$ or  $P_{D2011} = 1000 \times \ell^{-6.749002799}$ or  $P_{D2011} = 1000 \times 2.718^{-6.749002799}$ or  $P_{D2008} = 1000 \times 1.17286825 \times 10^{-3}$   $P_{D2011} = 8089.43MW$ @ **2012 (Prediction)**  $P_{D2012} = 1000 \times \ell^{2.025187111+-1.7548379821(2012.-2006)}$ 

or  $P_{D2012} = 1000 \times \ell^{2.025187111 + -1.7548379821 \times 6}$  or  $P_{D2012} = 1000 \times \ell^{2.025187111 + -10.52902789}$  or  $P_{D2012} = 1000 \times \ell^{-8.503840774}$  or  $P_{D2012} = 1000 \times 2.718^{-8.503840774}$ 

 $P_{D2012} = 1000 \times 2.028671829 \times 10^{-4}$ 

 $P_{D2012} = 0.202867182MW$ 

Total summation of predicted load in the case of Residential load (2000 – 2012)

That is,

$$\sum_{predicted} = 98.649,807MW$$

Similarly, the actual allocated Residential load between (2000 – 2012) becomes:

$$\begin{split} \Sigma_{Actual} &= (4608.40 + 7714.80 + 7668.50 + 7668.50 + \\ 7725.30 + 7760.00 + 7650.00 + 7860.30 + 7910.05 + \\ 8075.00 + 8205.20 + 8285.60 + 8350.00) = \end{split}$$

$$\sum_{Actual} = 99.481.65MW$$

$$MAD = \frac{\sum (Actual - Forecast)}{N}$$

$$= \frac{99481.65 - 98649.807}{13}$$

$$= \frac{831.843}{13}$$

$$= 63.9879MW$$

**Standard Error Predication** 

Standard error of prediction

$$\sum y_x = \sqrt{\frac{\sum (Y - Y_r)^2}{n}}$$
$$\sum y_x = \sqrt{\frac{\sum (56,907,966.86)}{13}}$$
$$= \sqrt{4377535.992}$$
$$= 2092.256MW$$
$$\approx 2092MW$$

Analysis of Allocation (Available demand) in MW wrt to energy demand prediction (forecast)

$$Y_i = P_{Di} = \ell^{a+b(x_i)}$$
$$Y_i = P_{Di} = \ell^{a+b(x_i - x_0)}$$

Where  $x_i = (x_i - x_0)$ Where  $x_i$  = Predicted Energy Demand  $x_0$  = base year = 2006  $x_i = 2000$   $\ell = 2.718$   $a = \frac{1}{n} \sum_{i=1}^{n} Y_i = \frac{1}{13} \sum_{i=1}^{13} Y_i = 1.267639604$ a = 1.267639604

$$b = \frac{\sum_{i=1}^{n} Y_i X_i}{\sum_{i=1}^{n} X_i^2} = \frac{\sum_{i=1}^{13} Y_i X_i}{\sum_{i=1}^{n} X_i^2}$$

b = 9.368497571 = 0.051475261

$$b = 0.051475261$$

 $x_0$  = base year (Reference year) 2006 B = base – MW = 1×10<sup>3</sup> = 1000MW

## **Commercial Demand**

Capacity allocation @ 2000 = 2346.00 (or available) Load forecast (prediction) @ 2000

$$Y_{i,2000} = \frac{P_D}{2000} = \ell^{a+b(x_i)}$$

$$Y_{i,2000} = P_{D_{2000}} = 2.718$$
Where  $X_i = x_i - x_0$ 
 $a = 1.267639604, b = 0.051475261$ 
 $x_0 = base year = 2006$ 
 $x_i = Predicted year = 2006$ 
 $Y_{i,2000} = 1000 \times P_{Di,2000}$ 
 $= 2.718^{(1.267639604+0.051475261)}$ 
 $Y_{i,2000} = 1000 \times P_{Di,2000}$ 
 $= 2.718^{(1.267639604+0.051475261(2000-2006))}$ 
 $Y_{i,2000} = P_{Di} = 1000 \times 2.718^{(1.267639604+0.051475261(-6))}$ 
 $Y_i = P_{Di} = 1000 \times 2.718^{(1.267639604+0.051475261(-6))}$ 
 $Y_{i,2000} = P_{Di} = 1000 \times 2.718^{(1.267639604+0.051475261(-6))}$ 
 $Y_{i,2000} = P_{Di} = 1000 \times 2.718^{(1.267639604+0.051475261(-6))}$ 
 $Y_{i,2000} = P_{Di} = 1000 \times 2.718^{(0.267639604+0.051475261(-6))}$ 
 $P_{D2001} = 1000 \ell^{(1.267639604+0.051475261(-5))}$ 
 $P_{D2001} = 1000 \ell^{(1.267639604+0.051475261(-2002-2006))}$ 
 $P_{D2002} = 1000 \ell^{(1.267639604+0.051475261(2002-2006))}$ 
 $P_{D2002} = 1000 \ell^{(1.267639604+0.051475261(-4))}$ 
 $P_{D2002} = 1000 \ell^{(1.267639604}$ 
 $= 1000 \times 2.718^{(1.267639604}$ 
 $P_{D2002} = 1000 \ell^{(1.267639604})$ 
 $P_{D2002} = 1000 \ell^{(1.267639604}$ 
 $P_{D2002} = 1000 \ell^{(1.267639604})$ 
 $P_{D2002} = 1000 \ell^{(1.267639604})$ 
 $P_{D2002} = 1000 \ell^{(1.267639604})$ 
 $P_{D2002} = 1000 \ell^{(1.26763960$ 

2003 (Prediction)  $P_{D2003} = 1000\ell^{1.267639604 + 0.051475261(2003 - .2006)}$  $P_{D2003} = 1000\ell^{1.267639604 + 0.051475261(-3)}$  $P_{D2003} = 1000\ell^{1.267639604 - 0.046425783}$  $P_{D2003} = 1000 \times 2.718^{1.221213821}$  $P_{D2003} = 1000 \times 2.718^{1.221213821}$  $=1000 \times 3.390872288$ = 3390.872288MW  $P_{D2003} = 3390.872288MW$ 2004 (Prediction)  $P_{D2004} = 1000\ell^{1.267639604 + 0.015475261(2004 - .2006)}$  $P_{D2004} = 1000 \times 2.718^{1.267639604 + 0.051475261(-2)}$  $P_{D2004} = 1000\ell^{1.267639604 - 0.030950522}$  $P_{D2004} = 1000 \times 2.718^{1.236689082}$  $P_{D2004} = 3443.749528MW$ 2005 (Prediction)  $P_{D2005} = 1000\ell^{1.267639604 + 0.01547526(2005 - .2006)}$  $P_{D2005} = 1000 \times 2.718^{1.267639604 + 0.05147526(-1)}$  $P_{D2005} = 1000 \times 2.718^{1.267639604 - 0.01347526}$  $P_{D2005} = 1000 \times 2.718^{1.252164344}$  $P_{D2005} = 1000 \times 3.497451339$  $P_{D2005} = 3497.451339MW$ 2006 (Prediction)  $P_{D2006} = 1000\ell^{1.267639604 + 0.01547526(2006 - .2006)}$  $P_{D2006} = 1000 \times 2.718^{1.267639604 + 0.01547526(0)}$  $P_{D2006} = 1000 \times 2.718^{1.267639604 + 0}$  $=1000 \times 2.718^{1.267639604}$  $P_{D2006} = 3551.990568MW$ 2007 (Prediction)  $P_{D2007} = 1000\ell^{1.267639604 + 0.01547526(2007 - .2006)}$  $P_{D2007} = 1000 \times 2.718^{1.267639604 + 0.05147526(1)}$  $P_{D2007} = 1000 \times 2.718^{1.267639604 - 0.01347526}$  $P_{D2007} = 1000 \times 2.718^{1.283114864}$  $P_{D2007} = 1000 \times 3.607380282 = 3607.380282MW$  $P_{D2007} = 3607.380282MW$ 2008 (Prediction)  $P_{D2008} = 1000\ell^{1.267639604 + 0.01547526(2008 - .2006)}$  $P_{D2008} = 1000 \times 2.718^{1.267639604 + 0.01547526(2)}$  $P_{D2008} = 1000 \times 2.718^{1.267639604 + 0.03095052}$  $P_{D2008} = 1000 \times 2.718^{1.298590124}$  $P_{D2008} = 1000 \times 3.663633742$  $P_{D2008} = 3663.633742MW$ 2009 (Prediction)  $P_{D2009} = 1000\ell^{1.267639604 + 0.01547526(2009 - .2006)}$ 

 $Y_i$ 

 $Y_i$ 

 $x_0$  $x_i$ 

 $P_{D2009} = 1000 \times 2.718^{1.267639604 + 0.05147526(3)}$  $P_{D2009} = 1000 \times 2.718^{1.267639604 + 0.04642578}$  $P_{D2009} = 1000 \times 2.718^{1.314065384}$  $P_{D2009} = 1000 \times 3.720764419 = 3720.764419$  $P_{D2009} = 3720.764419MW$ 2010 (Prediction)  $P_{D2010} = 1000\ell^{1.267639604 + 0.01547526(2010 - .2006)}$  $P_{D2010} = 1000 \times 2.718^{1.267639604 + 0.05147526(4)}$  $P_{D2010} = 1000 \times 2.718^{1.267639604 - 0.06190104}$  $P_{D2010} = 1000 \times 2.718^{1.329540644}$  $P_{D2010} = 1000 \times 3.77878599$  $P_{D2010} = 3778.78599MW$ 2011 (Prediction)  $P_{D2011} = 1000\ell^{1.267639604 + 0.01547526(2011 - .2006)}$  $P_{D2011} = 1000 \times 2.718^{1.267639604 + 0.05147526(5)}$  $P_{D2011} = 1000 \times 2.718^{1.267639604 - 0.0773763}$  $P_{D2011} = 1000 \times 2.718^{1.345015904}$  $P_{D2011} = 1000 \times 3.83771235 = 3837.71235MW$  $P_{D2011} = 3837.71235MW$ 2012 (Prediction)  $P_{D2012} = 1000\ell^{1.267639604 + 0.01547526(2012 - .2006)}$  $P_{D2012} = 1000 \times 2.718^{1.267639604 + 0.05147526(6)}$  $P_{D2012} = 1000 \times 2.718^{1.267639604 - 0.09285156}$  $P_{D2012} = 1000 \times 2.718^{1.36049116}$  $P_{D2012} = 3897.557592MW$ Total summation of predicted load in the case of Commercial load (2000 - 2012) That is, Σ Predicted Commercial energy demand 2608.273809 2746.036366 +3551.990568 + 3390.872288 + 3443.749528 3497.451339 3551.990568 3607.380282 3663.633742 +++3720.764419 +3778.785991 +3837.71235 3897.557592 = 45,298.19884MW Σ =45,296.19884MWpredicted

Similarly, the actual allocated Commercial load between (2000 - 2012) becomes:

 $\Sigma_{\text{Actual}} = 2346.00 + 2439.00 + 3297.60 + 3583.00 +$ 3830.30 + 3851.00 + 3900.80 + 3915.00 + 3852.00 +3865.50 + 3925.80 + 4004.70 + 4025.40 = 46, 836.1MW

$$\sum_{Actual} = 46,836.1MW$$

$$MAD = \frac{\sum (Actual - Forecast)}{N}$$

$$= \frac{46,836.1 - 45,29619884}{13}$$

$$= \frac{1539.9}{13} = 118.45MW$$

### **INDUSTRIAL DEMAND**

Analysis of Allocation (Available demand) in MW wrt to energy demand prediction (forecast)

$$Y_{i} = P_{Di} = \ell^{a+b(x_{i})}$$

$$Y_{i} = P_{Di} = \ell^{a+b(x_{i}-x_{0})}$$
Where  $x_{i} = (x_{i} - x_{0})$ 
Where  $x_{i}$  = Predicted Energy Demand  
 $x_{0}$  = base year = 2006  
 $x_{i}$  = 2000  
 $\ell = 2.718$ 

$$a = \frac{1}{n} \sum_{i=1}^{n} Y_{i} = \frac{1}{13} \sum_{i=1}^{13} Y_{i} = 0.457859173$$

$$b = \frac{\sum_{i=1}^{n} Y_{i}X_{i}}{\sum_{i=1}^{n} X_{i}^{2}} = \sum_{i=1}^{12} X_{i}^{2}$$

$$b = \frac{1.004552243}{182} = 5.519517819 \times 10^{-3}$$

$$k_{0}$$
 = base year (Reference year) 2006

 $B = base - MW = 1 \times 10^3 = 1000 MW$ Capacity allocation @ 2000 = 1011.60 (or available) Load forecast (prediction) @ 2000

$$Y_{i,20000} = \frac{P_D}{2000} = \ell^{a+b(x_i)}$$
$$Y_{i,2000} = P_{D_{2000}} = 2.718$$

Where 
$$X_i = x_i - x_0$$
  
 $a = 0.457859173$ ,  $b = 5.519517819 \times 10^{-3}$   
 $x_0 = \text{base year} = 2006$   
 $x_i = \text{Predicted year} = 2006$   
 $Y_{i,2000} = 1000 \times P_{Di,2000} = 2.718^{(0.457859173 + 5.519517819 \times 10^{-3})}$ 

$$Y_{i,2000} = 1000 \times P_{Di,2000}$$
  
= 2.718<sup>0.457859173+5.519517819×10<sup>-3</sup>(2000-2006)</sup>

 $Y_{i,2000} = P_{Di}$ 

+

+

$$=1000 \times 2.718^{0.457859173+5.519517819 \times 10^{-3}(-6)}$$

$$Y_i = P_{Di} = 1000 \times 2.718^{0.457859173 - 0.033117106}$$

$$Y_{i,2000} = P_{Di} = 1000 \times 2.718^{0.424742067}$$

 $Y_{i,2000} = P_{Di} = 1000 \times 1.529128596$ 

 $Y_{i,2000} = P_{Di_{2000}} = 1529.128596MW$ 

2001 (Prediction) Industrial Demand

$$P_{D2001} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3} (2001 - .2006)}$$

 $P_{D2001} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(-5)}$ 

 $P_{D2001} = 1000\ell^{0.457859173 + 0.027597589}$  $P_{D2001} = 1000\ell^{0.0430261584}$  $=1000 \times 2.718^{1.537591103}$  $P_{D2001} = 1537.591103MW$ 2002 (Prediction) Industrial Demand  $P_{D2002} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(2002 - .2006)}$  $P_{D2002} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(-4)}$  $P_{D2002} = 1000 \times 2.718^{0.457859173 + 0.022078071}$  $P_{D2002} = 1000 \times 2.718^{0.435781102}$  $=1000 \times 1.546100444$  $P_{D2002} = 1546.100444MW$ 2003 (Prediction) Industrial Demand  $P_{D2003} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3} (2003 - .2006)}$  $P_{D2003} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(-3)}$  $P_{D2003} = 1000 \times 2.718^{0.457859173 + 0.016558553}$  $P_{D2003} = 1000 \times 2.718^{0.459227364}$  $=1000 \times 1.58277518$  $P_{D2003} = 1582.77518MW$ 2004 (Prediction) Industrial Demand  $P_{D2004} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(2004 - .2006)}$  $P_{D2004} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(-2)}$  $P_{D2004} = 1000 \times 2.718^{0.457859173 + 0.011039035}$  $P_{D2004} = 1000 \times 2.718^{0.463746882}$ =1000×1.589944005  $P_{D2004} = 1589.944005MW$ 2005 (Prediction) Industrial Demand  $P_{D2005} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3} (2005 - .2006)}$  $P_{D2005} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(-1)}$  $P_{D2005} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}}$  $P_{D2005} = 1000 \times 2.718^{0.470266365}$ =1000×1.600342399  $P_{D2005} = 1600.342399MW$ 2006 (Prediction) Industrial Demand  $P_{D2006} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(2006 - .2006)}$  $P_{D2006} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(0)}$  $P_{D2006} = 1000 \times 2.718^{0.457859173}$  $P_{D2006} = 1000 \times 1.609199074$  $P_{D2006} = 1609.199074MW$ 2007 (Prediction) Industrial Demand  $P_{D2007} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(2007 - .2006)}$  $P_{D2007} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(1)}$ 

 $P_{D2007} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}}$  $P_{D2007} = 1000 \times 2.718^{0.481305432}$  $P_{D2007} = 1000 \times 1.618104703$  $P_{D2007} = 1618.104703MW$ 2008 (Prediction) Industrial Demand  $P_{D2008} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(2008 - .2006)}$  $P_{D2008} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(2)}$  $P_{D2008} = 1000 \times 2.718^{0.457859173 + 0.011039035}$  $P_{D2008} = 1000 \times 2.718^{0.486824208}$ =1000×1.627058416  $P_{D2008} = 1627.058416MW$ 2009 (Prediction) Industrial Demand  $P_{D2009} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3} (2009 - .2006)}$  $P_{D2009} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(3)}$  $P_{D2009} = 1000 \times 2.718^{0.457859173 + 0.016558553}$  $P_{D2009} = 1000 \times 2.718^{0.49234447}$  $P_{D2009} = 1000 \times 1.636064105$  $P_{D2009} = 1636.064105MW$ 2010 (Prediction) Industrial Demand  $P_{D2010} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(2010 - .2006)}$  $P_{D2010} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(4)}$  $P_{D2010} = 1000 \times 2.718^{0.457859173 + 0.022078071}$  $P_{D2010} = 1000 \times 2.718^{0.497863988}$  $P_{D2010} = 1000 \times 1.645118416$  $P_{D2010} = 1645.118416MW$ 2011 (Prediction) Industrial Demand  $P_{D2011} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3}(2011 - .2006)}$  $P_{D2011} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(5)}$  $P_{D2011} = 1000 \times 2.718^{0.457859173 + 0.027597589}$  $P_{D2011} = 1000 \times 2.718^{0.503383506}$  $P_{D2011} = 1000 \times 1.654222836$  $P_{D2011} = 1654.222836MW$ 2012 (Prediction) Industrial Demand  $P_{D2012} = 1000\ell^{0.457859173 + 5.519517819 \times 10^{-3} (2012 - .2006)}$  $P_{D2012} = 1000 \times 2.718^{0.457859173 + 5.519517819 \times 10^{-3}(6)}$  $P_{D2012} = 1000 \times 2.718^{0.457859173 + 0.033117106}$  $P_{D2012} = 1000 \times 2.718^{0.508903024}$ =1000×1.663377641  $P_{D2012} = 1663.377641MW$ Total summation of predicted load in the case of Industrial load (2000 – 2012) That is,

$$\sum_{redicted} = 20,839.02692MW$$

p

Similarly, the actual allocated Commercial load between (2000 – 2012) becomes:

$$\begin{split} \Sigma_{Actual} &= 1011.60 + 1987.20 + 1830.00 + 1659.80 + \\ 1605.00 + 1615.50 + 1575.00 + 1502.50 + 1585.00 + \\ 1589.40 + 1615.50 + 1648.00 = 17.756.70906 MW \end{split}$$

$$\sum_{Actual} = 17,756.70906MW$$
$$MAD = \frac{\sum (Actual - \Pr edicted)}{N}$$
$$= \frac{17,756.70906 - 20,839.02692}{13}$$
$$= \frac{-3082.31786}{13} = -237.1013738MW$$

## OW CHART SHOWING THE ACTIVITIES OF LEAST AND EXPONENTIAL REGRESSION ANALYSIS

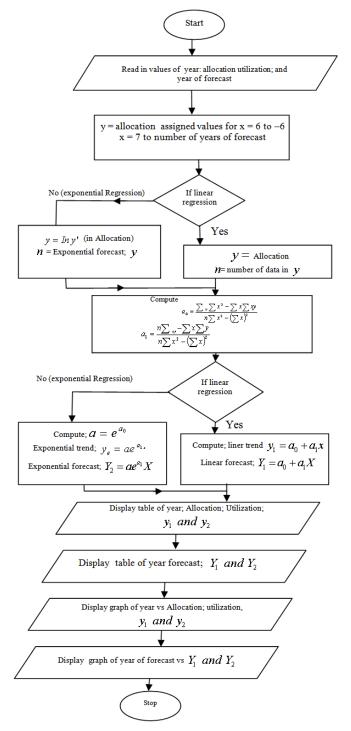


Figure 2. Flow chart showing the activities of least and exponential regression analysis

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# 4. Results and Discussion

The maximum load calculated with exponential  $Y = A \ell^{BX}$ regression. function is given as:  $(Y = 7570.95\ell^{0.02423x})$  with percent growth rate of 2.5%, while the maximum load calculated for least-square regression function is given as: (Y = a + bx) with growth 2.093%. Therefore the maximum rate of load: (7570.95MW) with 2.5% growth rate from exponential regression analysis method is more realistic and appropriate because of its ability to capture the energy requirement for consumers at the receiving end, on the view with steady energy growth-rate.

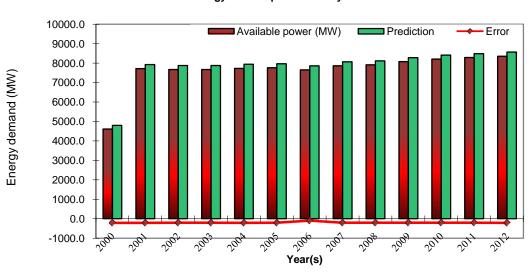
Load forecasting analysis is a major problem in power system planning and operations. This is because it provides then necessary information for: customer service and billing, electricity pricing and tariff planning etc. thereby giving an insight into future expansion planning. The result obtained shows that the utilization capacity (MW) does not give positive reflection of the capacity allocations (MW) when conducted in Matlab/Java platform especially in the case of residential, commercial and industrial plot (MW).

This mean that there is a deviation between installed capacity and that of utilization capacity of the consumer at the receiving end which need to be match in order to avoid overload and system collapse.

The paper work also extended to the linear model (lgast-square regression) and exponential regression plot which exhibit and confirms similar relationship in residential, commercial and industrial plot.

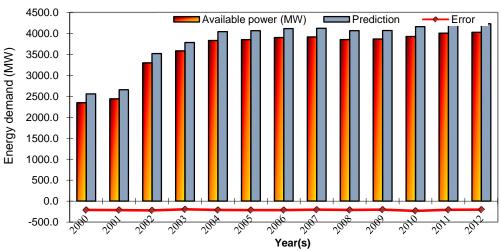
Evidently, the comparism plot for linear model and exponential regression show different behaviour, the least square model display linear graph while the exponential graph exhibit non-linear behaviour which is recommend because of the nonlinear relations between the capacity allocation (MW) and that of capacity in utilization.

#### EXPONENTIAL REGRESSION ANALYSIS



#### Chart of Residential energy consumption versus years

Figure 3. Chart of Residential energy consumption versus years



#### Chart of Commercial energy consumption versus years

Figure 4. Chart of Commercial energy consumption versus years

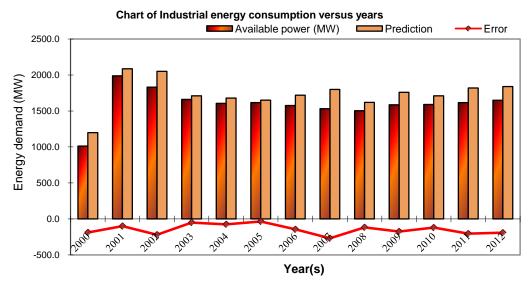
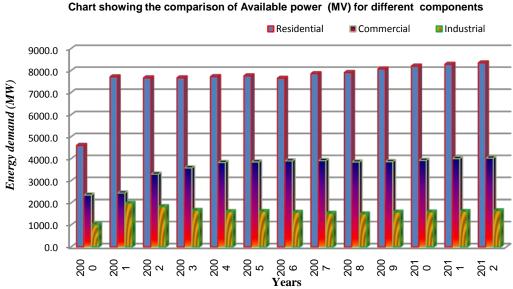


Figure 5. Chart of Industrial energy consumption versus years





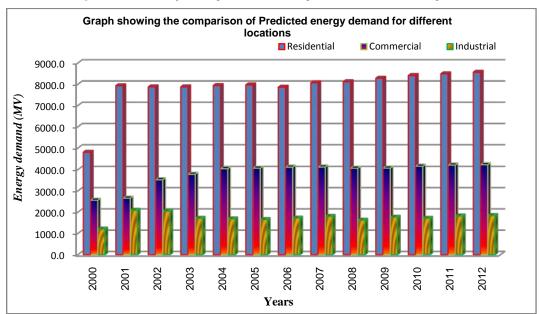
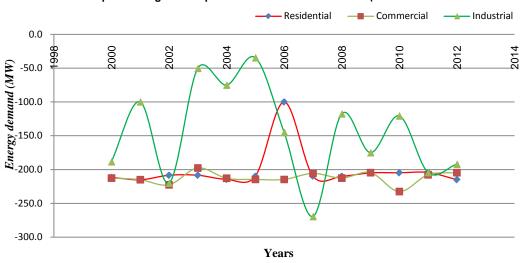


Figure 6. Chart showing the comparison of Available power (MV) for different components

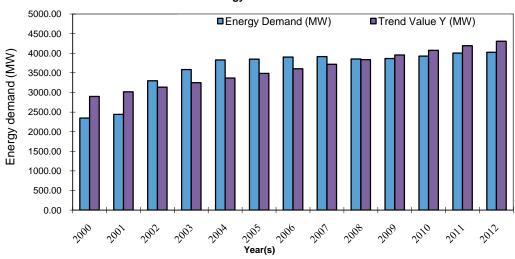
Figure 7. Graph showing the comparison of Predicted energy demand for different locations



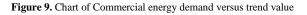
Graph showing the comparison of Error for different components

Figure 8. Graph showing the comparison of Predicted energy demand for different locations

## LEAST SQUARE REGRESSION ANALYSIS



#### Chart of Commercial energy demand versus trend value



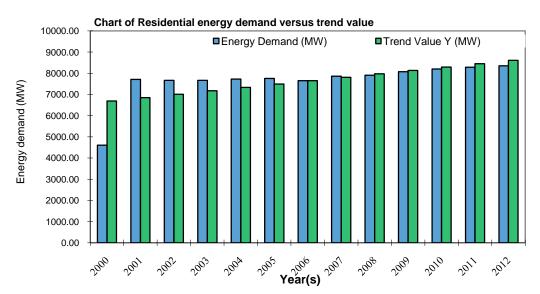


Figure 10. Chart of Residential energy demand versus trend value

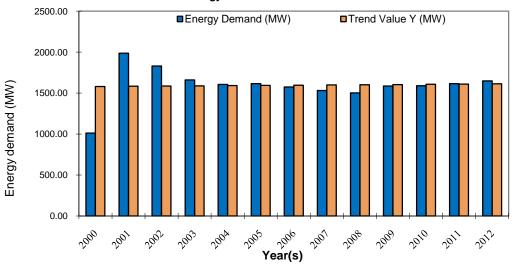


Chart of Industrial energy demand versus trend value

Figure 11. Chart of Industrial energy demand versus trend value



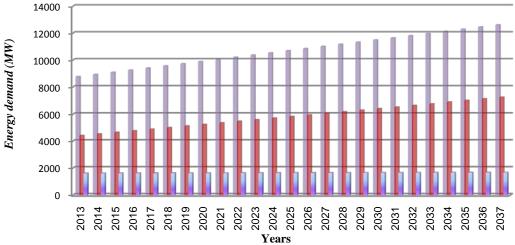
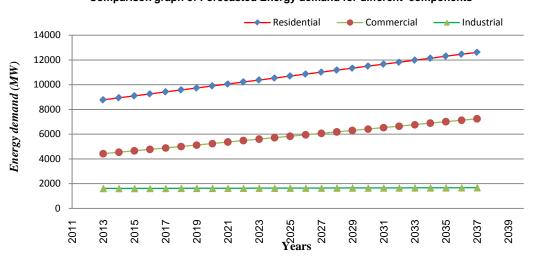
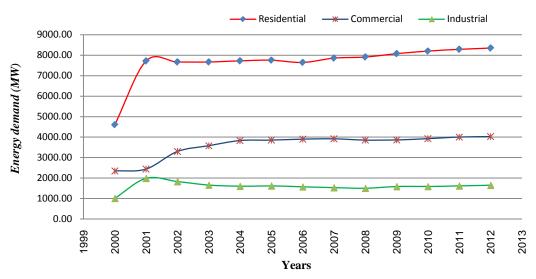


Figure 12. Chart showing the comparison of Predicted energy demand for different components



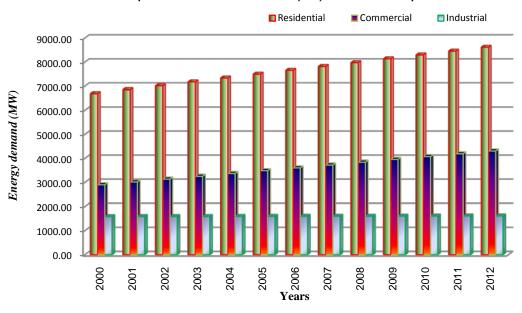
Comparison graph of Forecasted Energy demand for different components

Figure 13. Comparison graph of Forecasted Energy demand for different components



Graph showing the comparison of Energy demand for different components

Figure 14. Graph showing the comparison of Energy demand for different components



Comparison chart of Trend value Y (MW) for different components

Figure 15. Comparison chart of Trend value Y (MW) for different components

# 5. Conclusion

The electricity load forecast is a comprehensive survey of electrical demand and supply at the receiving end, in order to identify areas of inadequacies resulting to mismatches, which may negatively impact an overloads on transmission and distribution network. This work carried out load forecast using the analysis of least-square regression and exponential regression and validated in Matlab and Java programming platform.

Data are collected from (2000 - 2012) Central Bank of Nigeria (CBN) and National Bureau of statistics which serves as the input data into the least-square and exponential regression model for prediction into the projected future 2032. The result obtained shows that there is mismatch between installed capacity (MW) and utilization capacity (MW) in the case of residential, commercial and industrial load demand. Evidently, the result suggested that there is need to bridge the gap in order to enhance high reliability and efficiency in power system operation.

This work also identified the relationship between leastsquare model and exponential model plot on Matlab and Java program environment. Since the deviation between the installed and utilization capacity (MW) are non-linear, the exponential regression plot is therefore recommended, because of the non-linear behaviour of the capacity and load demand requirement.

Therefore, strategies for expansion programme is required for engaging distribution generator (DG) or decentralized generator via distribution outlet, on the view to solve overloads problems. This suggestion and recommendation if implemented will improve the electricity supply and demand in Nigeria.

# 6. Recommendation

Owing to the finding obtained in the course of this research the following recommendation are made:

- To meet the energy demand requirement, additional capacity generation will be required.
- (ii) Additional injection station and substation must be build to cater and care for rapid increase in load demand.
- (iii) Generation expansion program must be in place to accommodate annual growth of energy consumption.

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