

# Forecast on 2030 Vietnam Electricity Consumption

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**Abstract**—The first but very significant step in electricity system planning is to make an accurate long-term forecast on electricity consumption. This article aims to forecast the consumption for the Vietnam electricity system (GWH) up to 2030. An econometric model with the Cobb Douglas production function is used. The five variables proposed in the forecasting function are GDP, income, population, proportion of industry and service in GDP, and number of households. The forecasting equation is tested in terms of stationary and co-integration to choose meaningful variables and to eliminate the minor ones which account for none or small impacts on the forecast. The results show that: (1) the qualified forecasting equation only includes 3 major variables: the per capita income, the population, and the number of households, (2) with the medium income scenario, the forecasting consumptions in 2020, 2025, 2030 are 230,195 GWH, 349,949 GWH, 511,268 GWH respectively. (3) The GDP and the proportion of industry and service in GDP do not make major impacts on this forecasting in Vietnam. The method and the result of this article are likely a typical example of forecasting electricity consumption in developing countries which have a transforming economy similar to that in Vietnam.

**Keywords**—long-term; forecasting; electricity; consumption; econometric; model; Cobb Douglas; production; function; Vietnam

## I. INTRODUCTION

One of the most important requirements in national electricity system planning is to forecast the long-term electricity consumption for more than 5 or 10 years. The accuracy of the forecast makes a significant contribution to building up an economically optimal scenario for electricity system planning. In fact, it is very challenging to accurately estimate the long-term electricity consumption for the reason that the far-future (more than 5 years) consumption is dependent on some factors such as the GDP growth rate, the population, the transforming of economy structure, the electricity price, the level of applying scientific and technological advances in efficiently using energy, the electricity consumers' behavior and the climate change.

Vietnam is one of the few countries which experienced GDP growth rate more than 6% per year over the last 20 years. As a result, the electricity consumption increases by 15% annually. In 2010 the electricity consumption for more than 5 years ahead was simply predicted by using the value of the elasticity coefficient between the electricity growth rate and the GDP, whose value is 2. However, there have been some considerable gaps between forecasting and actual consumption from 2010 to 2015 [1]. The purpose of this article is to forecast Vietnam's electricity consumption up to 2030 with high reliability by applying the scientific forecasting method which is popular in other developing countries. To make a forecast on long-term electricity consumption, developing countries usually focus on 2 main methods: (1) to predict by using time-series models (TSMs), and (2) to predict by using Artificial Neural Networks (ANNs). Recently, in [2], authors utilized the mean absolute percentage error (MAPE) and the coefficient of determination ( $R^2$ ) to point out that the TSM method brings more accurate predictions in comparison with ANNs method. A new prediction method named Newton was also introduced with promisingly higher accuracy. It should take more time to verify this new method.

This article uses econometric model - the econometric approach combines economic theory and statistical techniques for forecasting electricity consumption. The approach estimates the relationship between electricity consumption and factors influencing the consumption. The relationships are estimated by the TSM method. The most frequently used model is that of Cobb Douglas production function. It is widely utilized in researches from Malaysia [3], Pakistan [4], and China [5]. The common method shared by these researches is to investigate the influential factors or variables impacting the forecast result. In other words, it means to find the independent variables of the equation. Then, the past time-series data of variables are utilized to verify the proposed forecasting equation with major variables. Features of a transforming economy in developing countries which are taken into account as variables in this article are: (1) GDP, (2) income, (3) population, (4) proportion

of industry and service in GDP, and (5) number of households. The input data are collected from 1990 to 2015.

## II. METHOD AND DATA

### A. Applying Cobb Douglas Production Function

Non-linear form of Cobb Douglas production function is given below:

$$EC_t = \varphi G_t^{\beta_1} P_t^{\beta_2} I_t^{\beta_3} X_t^{\beta_4} H_t^{\beta_5} \quad (1)$$

where  $EC_t$  is electricity consumption in year  $t$ .  $G_t$ ,  $P_t$ ,  $I_t$ ,  $X_t$  and  $H_t$  denote GDP, population, income, proportion of industry and service in GDP, and number of households, in the year  $t$ , respectively. The technological parameter is indicated by  $\varphi$ .  $\beta_i$  are returns to scale linked with the five above variables. By logarithmizing the two sides of (1), a linear function emerges:

$$\ln EC_t = \beta_0 + \beta_1 \ln G_t + \beta_2 \ln P_t + \beta_3 \ln I_t + \beta_4 \ln X_t + \beta_5 \ln H_t \quad (2)$$

### B. Testing

There are many methods to test econometric models with the same purpose – to establish a prediction equation with the highest accuracy. For researches proving the causality between variables, the date in time series should be stationary tested (Augmented Dickey Fuller (ADF) or Phillips-Perron (P-P)) to avoid the spurious regression. After stationary test, the causality between variables is proven by using Granger causality test. For forecasting purpose, stationary test only is enough.

#### 1) Stationary Test

In order to ensure the sustainability of the forecasting function, the stationary tests by using ADF unit root test and P-P unit root test are conducted. If there are any non-stationary series, the first difference of the series is calculated and eliminating inappropriate variables is done. If all series are non-stationary, calculating the first difference of all variables, then removing the primary features of the series leads to low  $R^2$  in regressing. The low  $R^2$  means low prediction accuracy. Therefore, the co-integration test is conducted to avoid low  $R^2$ .

#### 2) Cointegration Test

Authors in [6] supposed the linear combination between non-stationary time series can be a stationary series and these non-stationary time series are co-integrated. The non-stationary linear combination is called co-integration equation and can be explained as the balance relationship between variables in long-term. Therefore, if variables co-integrate, regression can be conducted. In this article, Johansen trace test is utilized to test the co-integration. In the trace test (T-test), there are some hypotheses  $H_0$  listed as follows: “None”, means there is no co-integration; “At most 1; 2; 3...”, means there are 1, or 2, 3 ... co-integrations. To decide if we should reject or accept hypothesis  $H_0$ , trace statistic value and critical value of 5% are

compared. If trace statistic < critical value, then  $H_0$  is accepted. If trace statistic > critical value, then  $H_0$  is rejected.

#### 3) $R^2$ test

In addition to stationary and co-integration tests, the research also tests the possibility and eliminates inappropriate variables of prediction equation on the base of two values of  $R^2$ . The high value of  $R^2$  is a signal for intensive relationship.

#### 4) Eliminating inappropriate variables

p-value is used to eliminate inappropriate variables. p-value is the lowest value of  $\alpha$  through which the test result is statistically meaningful. The use of p-value to test the hypothesis is:

$$\text{Reject } H_0 \text{ if } p\text{-value} < \alpha \quad (\alpha=0.05).$$

After testing and eliminating inappropriate variables, a qualified forecasting equation to be proposed.

### C. Data

#### 1) GDP

Vietnam is one country that has experienced annual GDP growth rate of about 6.5% for the last 20 years. There are 3 scenarios of GDP growth rate of Vietnam up to 2030 [1]. Because of low GDP growth rate archived in recent years, the low scenario of GDP growth rate is taken into account in this article. It is likely that the consumption for electricity continues to increase sharply. It is shown that there is partly relationship between the GDP and the electricity consumption. The GDP in Vietnam is extracted from the World Bank data [7].

#### 2) Income

Per capital income and the electricity consumption tend to be positively correlated. As the income increases, families have a tendency to purchase more electrical devices to satisfy their needs. For instances, at a certain level of income, they do not buy air conditioners and heaters. However, when the income reaches a higher level, after considering if they afford paying the electricity consumption, they purchase these devices to make their life more comfortable. As a result, the electricity consumption in that households rises [9]. Therefore, the per capital income is taken into account in this research. Data is extracted from the World Bank data (Atlas method) [7].

#### 3) Population

The population growth rate in Vietnam is around 1.08% [7]. If there is a variable impacting on the electricity consumption, it is the population. The population increases, which enlarges power consumption. The population data are provided by the general statistics office of Vietnam [8].

#### 4) Number of Households

The household separation leads to the increasing consumption for home electric appliances. It creates not only increase of electricity consumption in human livelihood but also that in the industry sector for manufacturing these

appliances. Authors in [10] mention "Households energy consumption accounts for a large share of total energy consumption". However, there is a problem in data collecting that Vietnam and other developing countries encounter. It is the fact that the household data in Vietnam is not fully recorded. This requires establishing a method of calculating the number of households in Vietnam. The household data in Vietnam is scanty and un-intensive. In the years 1999, 2004 and 2009, the accurate numbers of households are 16.66, 19.01 and 22.44 million, respectively. Recognized as the official figures by Vietnam and The United Nations, these numbers are the results from the investigation on population and housing [8, 9]. Moreover, the increasing and decreasing rates of number of households are also important parameters to calculate the numbers of households in the upcoming years. According to [8], the number of households in Vietnam 2014 is 24.27 million, two times that of 25 years ago (1/4/1989), one-and-a-half time that of 15 years ago (1/4/1999), and nearly 2 million households higher than that in 1st April 2009. During the 2009-2014 period, the average increasing rate of households is 1.6% per year period. Recorded by Helgi Library, the number of households in Vietnam reached 26.9 million in 2016, increasing by 2.43 % compared with that of the previous year [11]. Table I shows the number of households in Vietnam.

The comparison between Vietnam and other countries should be made to figure out the similarities in the average size of a household. The average size of Vietnam households in 1999 was 4.6 people/household, equivalent to that of Thailand in 1990 (4.62 people). According to the Thailand Development Research Institute, the size of Thailand's households changed throughout the years as follows: 1985: 4.98, 1990: 4.62, 1995: 4.27, 2000: 3.96, and 2005: 3.7. According to the National Statistical Office of Thailand, the number of households in Thailand reached 20.6 million in 2016. This increase accounts for 0.205%, compared with that of the previous year. The number 20.6 million in 2016 is the highest, while the lowest is 0.6 million in 1960. In comparison with Thailand, those figures of Cambodia and Malaysia respectively are 3.34 million and 7.30 million [11]. According to The Department of Statistics in Malaysia, in 2016, the number of households in Malaysia reached 7.30 million, a rise of 1.74 %, compared with the previous year. This figure is the highest, while the lowest was 1.50 million in 1960. Compared with other neighbor countries of Malaysia, the numbers of households in 2016 were respectively 3.34 million in Cambodia, 1.26 million in Singapore, 20.6 million in Thailand, and 26.9 million in Vietnam [11]. It is easily recognized that there are geographic and socio-economic similarities between Vietnam, Thailand and Malaysia. Even in Japan, a developed country, the number of households keeps increasing at nearly 1%/year, while, population has been decreasing [11]. The statistical charts of the household growth in these countries tend to have exponential shapes. Based on the data in Table I, the trend line has the exponential shape shown in Figure 1. The trend line equation is:

$$y = 4.85E - 23e^{2.71E - 0.2x} \quad (3)$$

TABLE I. NUMBER OF HOUSEHOLDS [8, 9] (MILLIONS)

Year	1999	2004	2009	2014	2016
	16.66	19.01	22.44	24.27	26.9

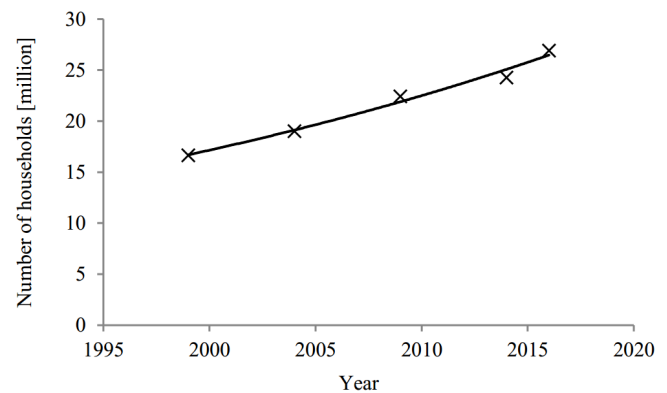


Fig. 1. The trend line of the number of households.

The household data from 1990 to 2015 are shown in Table II. Using the trend line, the prediction on the number of households in the future is shown in Table III.

TABLE II. THE NUMBER OF HOUSEHOLDS CALCULATED BY THE ARTICLE (MILLIONS)

Year	1990	1991	1992	1993	1994	1995	1996
	12.79	13.14	13.50	13.87	14.25	14.64	15.05
Year	1997	1998	1999	2000	2001	2002	2003
	15.46	15.88	16.66	16.77	17.23	17.70	18.19
Year	2004	2005	2006	2007	2008	2009	2010
	19.01	19.20	19.73	20.27	20.83	22.44	21.99
Year	2011	2012	2013	2014	2015		
	22.59	23.21	23.85	24.27	25.18		

TABLE III. THE NUMBER OF HOUSEHOLDS IN THE FUTURE (MILLIONS)

Year	2020	2025	2030
Forecast number of households	28.83	33.02	37.81

##### 5) Proportion of Industry and Service in GDP

In addition to the human livelihood, the proportion of industry and service in GDP are also sectors accounting for a large share of electricity consumption. A feature of a transforming economic in a developing country like Vietnam is that this proportion increases. Relevand data are shown in Table IV [1].

##### 6) Collecting Data in Time Series of 1990-2015

After analyzing factors likely to influence the electricity consumption in Vietnam, the past data of the factors are collected. With full, accurate data and long enough time series, the equation predictions are more exact. According to [12, 13], if  $n$  the data in time series and  $k$  the independent variables in the model, then:  $n-k > 20$ . As mentioned above, the chosen variables include the GDP, the income, the population, the number of households, the proportion of industry and service. Therefore, the forecasting equation contains 5 independent variables and:  $n > 20 + 5$ . This study selects the time series from

1990 to 2015. The time series is relatively long enough variable including 26 values (equivalent to 26 years) to conduct testing. In addition to the domestic statistical data, the country has the chance to assess those data from international organizations. Table IV presents the relevant data.

TABLE IV. SUMMARY OF INPUT DATA

Year	Electricity Consumption (GWh)	GDP (Billion \$)	Population (x10 <sup>6</sup> )	Income (\$/Person)	Industry & Service (%)	Number of Households (Millions)
1990	8,678	6.47	66,017	130	61.3	12.79
1991	9,152	9.61	67,242	110	59.5	13.14
1992	9,654	9.87	68,450	130	66.1	13.50
1993	10,665	13.18	69,645	170	70.1	13.87
1994	12,284	16.29	70,825	200	72.6	14.25
1995	14,636	20.74	71,996	260	72.8	14.64
1996	16,946	24.66	73,157	310	72.2	15.05
1997	19,151	26.84	74,307	350	74.2	15.46
1998	21,665	27.21	75,456	360	74.2	15.88
1999	23,739	28.68	76,597	370	74.6	16.66
2000	26,745	33.64	77,631	400	77.3	16.77
2001	30,187	35.29	78,621	430	78.5	17.23
2002	34,073	37.95	79,538	460	78.7	17.70
2003	38,461	42.72	80,467	510	79.1	18.19
2004	43,414	49.42	81,436	590	80.0	19.01
2005	49,008	57.63	82,392	680	80.7	19.20
2006	53,845	66.37	83,311	760	81.3	19.73
2007	59,159	77.41	84,219	850	81.3	20.27
2008	64,998	99.13	85,119	1,000	79.6	20.83
2009	71,415	106.01	86,025	1,120	80.8	22.44
2010	78,466	115.93	86,933	1,270	69.1	21.99
2011	94,658	135.54	87,860	1,390	69.0	22.59
2012	105,474	155.82	88,809	1,550	70.8	23.21
2013	115,069	171.22	89,760	1,740	71.9	23.85
2014	128,435	186.20	90,729	1,900	72.3	24.27
2015	141,800	193.60	91,704	1,990	73.0	25.18

III. RESULTS

A. Converting Independent Variables

The series of input data are converted into natural logarithms listed in Table V.

B. Testing

1) Stationary test

The stationary test results are shown in Table VI. It is shown that all p-values are larger than  $\alpha$  ( $\alpha=0.05$ ), except for the population variable ( $\ln(P_t)$ ) when P-P test is conducted. In other words, all variables are not stationary. Therefore, the co-integration test is to be conducted instead of calculating the first difference.

2) Co-integration test

The co-integration test results are shown in Table VII. After the test, we accept the hypothesis  $H_0$ . There are three co-integrations with trace statistic method.

3) Testing R2 and p-value

After the co-integration test, the coefficients of (2) are calculated and the results are shown in Table VIII. The condition for choosing the coefficients of equation is that the  $R^2$  almost equals to 1 and p- values of all variables are less than 0.05. In Table VIII, R squared ( $R^2$  testing)=0.997953. It means that 99.8% of dependent variables have impacts on independent variables. However, due to the fact that the p-value is above 0.05, then,  $\ln(G_t)$  is removed.

TABLE V. CONVERTING INDEPENDENT VARIABLES INTO NATURAL LOGARITHMS

Year	$\ln(EC_t)$	$\ln(G_t)$	$\ln(P_t)$	$\ln(I_t)$	$\ln(X_t)$	$\ln(H_t)$
1990	9.07	1.87	11.10	4.87	4.12	2.55
1991	9.12	2.26	11.12	4.70	4.09	2.58
1992	9.18	2.29	11.13	4.87	4.19	2.60
1993	9.27	2.58	11.15	5.14	4.25	2.63
1994	9.42	2.79	11.17	5.30	4.28	2.66
1995	9.59	3.03	11.18	5.56	4.29	2.68
1996	9.74	3.21	11.20	5.74	4.28	2.71
1997	9.86	3.29	11.22	5.86	4.31	2.74
1998	9.98	3.30	11.23	5.89	4.31	2.77
1999	10.07	3.36	11.25	5.91	4.31	2.81
2000	10.19	3.52	11.26	5.99	4.35	2.82
2001	10.32	3.56	11.27	6.06	4.36	2.85
2002	10.44	3.64	11.28	6.13	4.37	2.87
2003	10.56	3.75	11.30	6.23	4.37	2.90
2004	10.68	3.90	11.31	6.38	4.38	2.94
2005	10.80	4.05	11.32	6.52	4.39	2.96
2006	10.89	4.20	11.33	6.63	4.40	2.98
2007	10.99	4.35	11.34	6.75	4.40	3.01
2008	11.08	4.60	11.35	6.91	4.38	3.04
2009	11.18	4.66	11.36	7.02	4.39	3.11
2010	11.27	4.75	11.37	7.15	4.24	3.09
2011	11.46	4.91	11.38	7.24	4.23	3.12
2012	11.57	5.05	11.39	7.35	4.26	3.14
2013	11.65	5.14	11.40	7.46	4.28	3.17
2014	11.76	5.23	11.42	7.55	4.28	3.19
2015	11.86	5.27	11.43	7.60	4.29	3.23

TABLE VI. STATIONARY TESTING OF THE FORECASTING EQUATION

Variable	ADF Test		P-P Test	
	T. statistic	Prob.value	T. statistic	Prob.value
$\ln EC_t$	0.000867	*	-0.026761	*
$\ln G_t$	-2.287395	*	-2.232903	*
$\ln H_t$	0.632598	*	0.721125	*
$\ln I_t$	-1.210096	*	-0.475247	*
$\ln P_t$	-0.641799	*	-9.851650	***
$\ln X_t$	-2.457299	*	-2.457299	*

\* Variables with p-value > 0.05  
 \*\* Variables with 0.02 < p-value < 0.05  
 \*\*\* Variables with p-value < 0.02

TABLE VII. CO-INTEGRATION TEST

Hypothesized No. of CE(s)	Trace Statistic	Critical Value	Prob.
None *	179.6550	95.75366	0.0000
At most 1 *	109.0413	69.81889	0.0000
At most 2 *	52.81937	47.85613	0.0159
At most 3	27.25295	29.79707	0.0956
At most 4	9.796955	15.49471	0.2967
At most 5	0.530471	3.841466	0.4664

\* Denotes rejection of the hypothesis at the 0.05 level

The recalculation of the equation coefficients is shown in Table IX. Similarly,  $\ln X_t$  is removed and the equation coefficients are again re-calculated. The result is shown in Table X, in which all p-values of variables in the equation reach \*\*\* level. In other words, all p-values are less than 0.02 (very qualified).

TABLE VIII. CALCULATION OF THE COEFFICIENTS OF (2)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-69.19764	8.675347	-7.976353	***
$\ln G_t$	-0.110864	0.083668	-1.325048	*
$\ln H_t$	0.682145	0.164572	4.144974	***
$\ln I_t$	0.220460	0.087524	2.518847	**
$\ln P_t$	6.937385	0.861815	8.049738	***
$\ln X_t$	-0.357666	0.151796	-2.356232	**
R-squared	0.998899			
F-statistic	3628.116			
Prob (F-statistic)	0.000000			
AIC	-3.800512			

TABLE IX. CALCULATION OF THE COEFFICIENTS (SECOND TIME)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-62.79885	7.335583	-8.560853	***
$\ln H_t$	0.756167	0.157561	4.799209	***
$\ln I_t$	0.145212	0.067787	2.142167	**
$\ln P_t$	6.327205	0.741462	8.533421	***
$\ln X_t$	-0.283445	0.143599	-1.973864	*
R-squared	0.998802			
F-statistic	4377.180			
Prob (F-statistic)	0.000000			
AIC	-3.793289			

TABLE X. CALCULATION OF THE COEFFICIENTS (THIRD TIME)

Variable	Coefficient	Std. Error	t-Statistic	Prob.
C	-53.82916	6.125857	-8.787205	***
$\ln H_t$	0.974249	0.119498	8.152860	***
$\ln I_t$	0.197233	0.066438	2.968668	***
$\ln P_t$	5.338763	0.581688	9.178048	***
R-squared	0.998580			
F-statistic	5156.169			
Prob (F-statistic)	0.000000			
AIC	-3.700022			

### C. Eliminating Inappropriate Variables

After the eliminating of  $\ln G_t$  and  $\ln X_t$ , (2) becomes:

$$\ln EC_t = \beta_0 + \beta_1 \ln P_t + \beta_2 \ln I_t + \beta_3 \ln H_t \quad (4)$$

where  $\beta_0$ ,  $\beta_1$ ,  $\beta_2$ , and  $\beta_3$  are -53.829, 5.339, 0.197, and 0.974, respectively (see Table X). Equation (4) becomes:

$$\ln EC_t = 5.339 \ln P_t + 0.197 \ln I_t + 0.974 \ln H_t - 53.829 \quad (5)$$

Equation (5) is the qualified forecasting equation.

### D. Forecast on Electricity Consumption to 2030

Future values of population, income, and number of households are predicted and shown in Table XI. Using this data, the natural logarithms, and (5), Vietnam electricity consumption in the future can be forecasted. The results are shown in Table XII.

TABLE XI. FORECAST ON POPULATION, NUMBER OF HOUSEHOLDS AND INCOME

Variable	Note	2020	2025	2030
Population	(thousands) [14]	96,302	99,929	102,886
Income (\$/year) [1]	Low scenario	3,307	4,939	7,205
	Medium scenario	3,370	5,111	7,836
	High scenario	3,485	5,450	8,450
Number of households	(millions)	29.89	34.49	39.79

TABLE XII. FORECAST ON ELECTRICITY CONSUMPTION TO 2030 (GWH)

Electricity Consumption	2020	2025	2030
Low income scenario	229,341	347,597	502,882
Medium income scenario	230,195	349,949	511,268
High income scenario	231,722	354,404	518,923

The results show that in the low income scenario, the electricity consumption in 2020 is around 229,241 GWH, in 2025 is around 347,597 GWH and that in 2030 is 502,882 GWH. In the medium scenario, the electricity consumption in 2020 reaches 230,195 GWH, in 2025 is around 349,949 GWH and that in 2030 is 511,268 GWH. Finally, in the high scenario, the electricity consumptions in 2020, 2025, and 2030 are 231,722 GWH, 354,404 GWH, and 518,923 GWH, respectively. With the income increase from about 2% in case of medium and low scenario in 2020 to about 9% in case of medium and low scenario in 2030, electricity consumptions increases from about 0.4% to about 1.7%, respectively.

## IV. CONCLUSION

By using the econometric model with the Cobb Douglas production function, the proposed forecasting equation (1) is made with five variables: GDP, income, population, proportion of industry and service in GDP and number of households. After ADF, co-integration and  $R^2$  and p-value tests, the qualified forecasting equation (5) has only three variables: income, population, and number of households. This means that GDP, and the proportion of industry and service in GDP do not make a major impact on this forecasting in Vietnam. This is quite a surprising result. The explanation for this can be that income increase is also followed by GDP increase and that the data set of GDP (1990-2015) may not be aligned with the proportion of industry and service in the GDP. The equation of the number of households for Vietnam is proposed by exponential function (3) which is the one of Thailand, Malaysia, and Japan. Electricity consumption in 2030 will be more than two times that of 2020, and more than 3 times that of 2016 (159,450 GWH). This means that billions of US dollars per year will need to be invested to the power system of Vietnam. In details, in case of the medium income scenario, electricity consumption increase about 10%/year in 2017-2020, 8.4%/year in 2021-2025 and 6.8%/year in 2026-2030. This is

quite a good trend, because the increasing is slowing down gradually compared with that of the previous period (2010-2015) which was 10%-15%/year. The method and the results of this research likely are a typical example for the forecasting of electricity consumption of those developing countries which have the same transforming economy with Vietnam.

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