

# kWh Demand Assessment and Forecasting of a Low Voltage Private Academic Institution



Joy R. Antonio, Antonio S. Buendia Jr., Gina B. Narca, Noel T. Florencondia, Lorinda Pascual

**Abstract:** Transformer is one of the crucial components in an electrical system. They are used to convert energy to a utilization level. Load allocation is vital to sustaining the needed energy demand for a growing population, and it is an essential consideration for continuous consumer connectivity of electricity. In this study, the researchers are conducting assessment and forecast of the transformer requirement for a Low Voltage Private Academic Institution. It assessed the kWh demands and kWh allocation from 2005 to 2019 of the three transformers used of the locale. The kWh demand was used to forecast the per semi-annual for the next five years. Six forecasted models were applied and the result were verified by QM for Windows Version 5. Mean Absolute Error (MAE), Mean Square Error (MSE) and Mean Absolute Percent Error (MAPE) were used to select the best forecasting model with the least value. DT (Distribution Transformer) 1 used the 4<sup>th</sup> degree polynomial model because of the least value of MAE, MSE and MAPE of 5340.626, 42306100 and 17.12%, while DT 2 and DT 3 applied the 3<sup>rd</sup> degree polynomial model though it has the second lowest value of MAE, MSE and MAPE of 3142.907, 16748570, 20.44% and 2576.398, 8740315, 11.21% respectively because the projected demand of the 4<sup>th</sup> degree gave a negative value.. The researcher finds that DT 1 and 3 were starting to exceeds the allotted kWh. On the other hand, DT 2 and 4 were plenty of kWh allocation. Relocation of load from DT 1 and 3 to DT 2 may take for considerations by the institution.

**Keywords:** Distribution Transformer), kWh Demand, kWh allocation, Forecasting.

## I. INTRODUCTION

Transformers were widely used in an electrical power system, whether in the generation system, transmission system and distribution system. Transformers function is to transform energy to usable level to utilize by people. In a distribution power system, distribution transformers were usually installed in an operating range of 13,800 primary voltage input to a secondary output voltage of 230 volts. Transformers in the distribution system are classified to be either:

(a). sole-user distribution transformer (if only one consumer entity uses one kWh meter like industrial, commercial, public building) and

(b). commonly-used distribution transformer (if multiple consumers or multiple kWh meters are installed like in most residential areas).

Categorization Sole-user according to the connection of the electric meter.

i. High Voltage if the meter is connected in the primary side of the transformer and also called primary meter.

ii. Low Voltage if the meter is connected in the secondary side of the transformer and also called secondary meter.

Distribution transformer rating is measured in kVA, commonly installed in the distribution system were 10kVA, 15kVA, 25kVA, 37.5kVA, 50kVA. For a transformer, loading capacity is vital for consumer connectivity of electricity. Maximum capacity of it must be 70% [1]. With this, loading of transformer must be taken into account to prevent transformer failure due to exceeding its normal loading [2-3]. About 50% of its useful life is affected when exceeding of normal loading occurs [4], while on the other hand, in below normal loading of transformers do not utilize its use and contribute more core loss [5]. To ensure transformer-costumer connectivity it must be taken to account an effective transformer load monitoring system and transformer data such as rating, connection, and location. [6].

Energy management must show consumers demand and allocation accurately, to inform the consumer about the details of their actual energy usage. [7]. In the recent years, due to the increasing demand of electricity, most of the consumer in distribution utility were driven to improve energy loading management, one of which is academic institution. In Bohol Island State University-Main Campus applied the used of solar power. [8]. Due the increasing number of students, Don Honorio Ventura State University projected the available electric source it can provide. [9]. Computer simulated-audit and analyzed the kWh consumption of each room were conducted by Pamantasan ng Lunsod ng Maynila [10].

Electrical load forecasting has different types, namely:

(a) very-short term forecasting uses few minutes to an hour,

(b) short term forecasting uses a day or a week prediction,

(c) medium-term forecasting is applicable if the prediction within 2 weeks to 3 years range and

(d) long term electrical load forecasting must be taken to account if the forecast interval from 3 years to 50-years,

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Demand for electricity is evidently increase, and this is affecting the loading condition of the transformer.

To give awareness of the needs of the consumers conducting for an assessment the present loading conditions is vital. And forecasting of electrical energy demand can help in determining if the capacity can carry the demand needed. To reduce uncertainties,

planning for additional sources and to the prediction of the phase of electric power must be taken to account [11]. The objective of the study is to assess the present and future loadings of the transformer of low voltage private academic institution. Specifically, the study tends in determining the electrical loading and allocation of each transformer per semi-annual per year and to analyze the historical loading trends. The results are intended to be applied to forecast the percent loading per semi-annual for the next five years.

## Locale of the Study

The locale of the study is a private academic institution founded in 1945.

It offers different programs such as graduate studies, college, junior high school, senior high school and elementary. From 2005 to 2015 the total student enrolled in the institution were less than 2000 until the year 2016, when the institution starts to offer senior high school program, the number of enrolled students increased as shown in Figure 1. In the academic year 2016-2017, an increase of 44.67% of enrolled student had experienced.

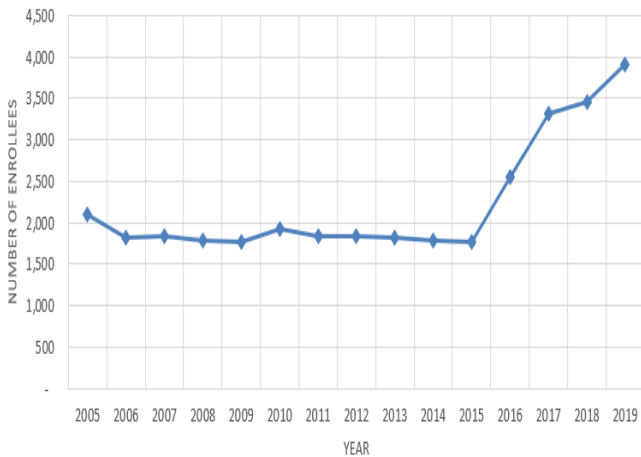


Figure1: Number of Enrolled Student (2005 – 2019)

## Load Demand of the locale

Three (3) distribution transformers installed around the campus with kVA ratings of 15kVA (DT 1), 3 units of 25 kVA (DT 2) and 10 kVA (DT3). From 2005, DT 1 and 2 are being used, and in 2014, DT 3 was installed to cater the increase of energy demand in preparation for the offering of K-12 program. The annual kWh demand is shown in Figure 2. The figure shows that the kWh demand in 2016 of the institution started to increase due to the addition of students enrolled. In academic 2015 – 2016, the energy demand rose to 14.19% in the same way the energy consumption amounting to ₱1,258,679.84. In 2017 – 2018, 22.93% increase for compare in the last academic year. It is undeniable that the demand increased.

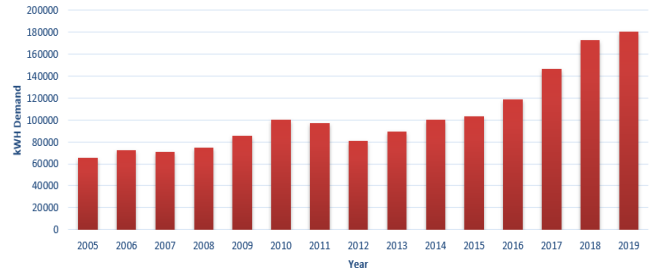


Figure 2: Yearly kWh Demand

## Individual Load Profiling of Transformers

In Figures 3 – 5 show the individual semi-annual kWh demand of each transformer installed in the institution. DT 1 and 2 shows an increasing trend; in the year 2016, when the k-12 programs were offered, the demand increased, each has demand value of 46,545 kWh and 24,150 kWh respectively. It is observable that the kWh demand has continually increased on the next academic years, as seen in Figures 3 and 4. On the other hand, there is a time in which the graph showed a decrease in usage, which is the first half of the year during April and May vacation period when minimal energy is consumed.

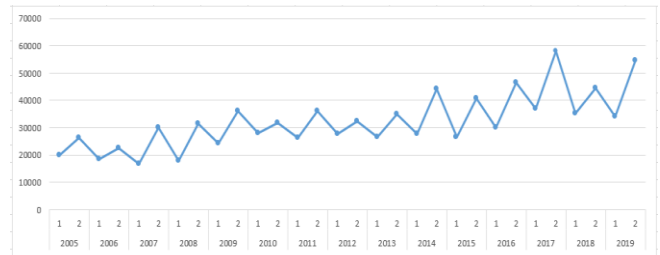


Figure 3: DT 1 Semi-Annual kWh Demand

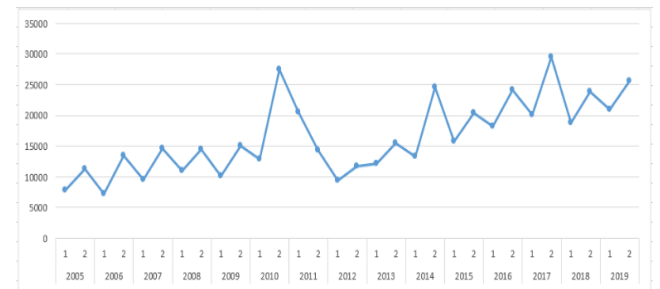


Figure 4: DT 2 Semi-Annual kWh Demand

For the preparation of the offering of the k-12 programs, DT 3 was installed in the year 2014. As the school-year 2016-2017 came, there is an escalating demand experienced as revealed in Figure 5, energy demand for 55,640 kWh for that year. And for the succeeding years, the energy demand is still increasing.

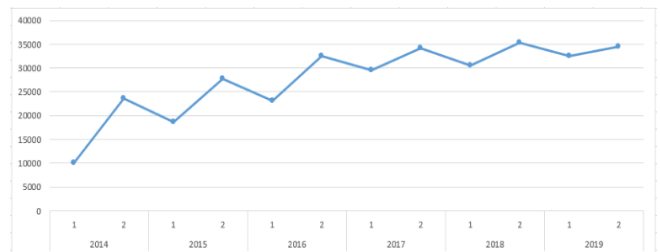


Figure 5: DT 3 Semi-Annual kWh Demand

Forecasting will be performed in the study using the historical energy consumption of the present system to assess the kWh demand. Buildings connected to each meter was not considered in the study since it requires legal coordination with the utility. Hence, the researcher focused on the loading of the transformer installed in the institution to identify the loading condition of each transformer. This will give awareness and information on the utilization of the transformer to the administration of the institution.

## II. METHODOLOGY

### A. Data Collection and Instrument

Data needed in the study such as the kWh consumption per meter and kWh rating of each transformer installed were gathered from the electric utility. 2005 - 2019 kWh consumption and kWh ratings of transformers were the input parameters of the study; this will be used to evaluate the loading demand of the transformer, as shown in Figure 6. The researcher had used QM for Windows for the statistical treatment and Microsoft Excel 2016 for load demand forecast.

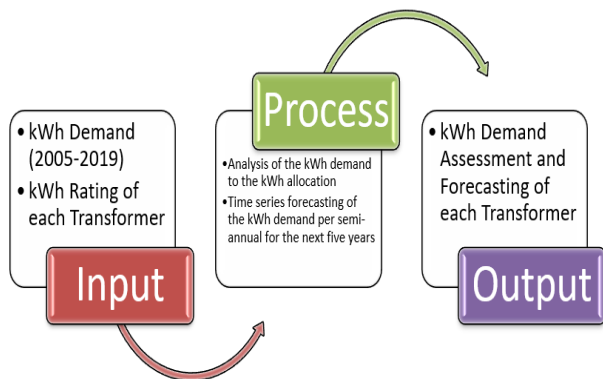


Figure 6: Conceptual Framework of the Study

### B. Assessment of Transformer

From the data gathered from electrical utility, the three (3) distribution transformer and kWh consumption (2005 - 2019) were used to determine the kWh demand per semi-annual of the transformer by using Microsoft Excel 2016.

### C. Forecasting

The load demand values will be used to forecast the per semi-annual for the five (5) years ahead. The study is classified as long-term forecasting. Best loading values will be based for the forecast model. Forecasted data will be gathered using the mathematical forecasted model. The software to be used is QM for Windows for the statistical treatment to identify the Mean Absolute Error (MAE), Mean Squared Error (MSE) and Mean Absolute Percent Error (MAPE). The result will be a criterion to test the accuracy of the forecasted model.

1. Mean Absolute Error (MAE) - It is a measure of how close the forecasted value to the actual value [9], [12-13].

$$MAE = \frac{\sum_{t=1}^t |A_t - F_t|}{t} \quad (1)$$

Where:  $A_t$  = Actual Value  
 $F_t$  = Forecasted Value  
 $t$  = time

2. Mean Square Error (MSE). It is always a positive value and it is closer to the value of zero the better [9], [12-13].

$$MSE = \frac{\sum_{t=1}^t (A_t - F_t)^2}{t} \quad (2)$$

Where:  $A_t$  = Actual Value  
 $F_t$  = Forecasted Value  
 $t$  = time

3. Mean absolute Percent Error (MAPE). Is a measure of accuracy is expressed in percentage of the average of the absolute value of error [12-13].

$$MAPE = \frac{\sum_{t=1}^t |A_t - F_t|}{t} \times 100\% \quad (3)$$

Where:  $A_t$  = Actual Value  
 $F_t$  = Forecasted Value  
 $t$  = time

With this criterion, the forecasted model to be used must be the least value of MAE, MSE and MAPE [13-15].

## III. RESULTS & DISCUSSION

Two (2) sections were provided for the data presentation of the study. The first section is the load demand of each transformer installed in the institution and the assessment of the loading conditions. While the second section is the forecasting analysis of the kWh demand per semi-annual for the next five-years.

### A. Assessment of Transformer

The graphical representation of the kWh demand of the transformers installed is shown in Figures 9-11. These data were taken to account to assess the kWh demand of the four (4) transformers. DT 1 in Figure 7 shows that its kWh demand exceeds the kWh allocation in the last half of year 2017 and 2019. In the academic year 2016-2017, the kWh demand of DT 1 started to reaches the maximum kWh allocation.

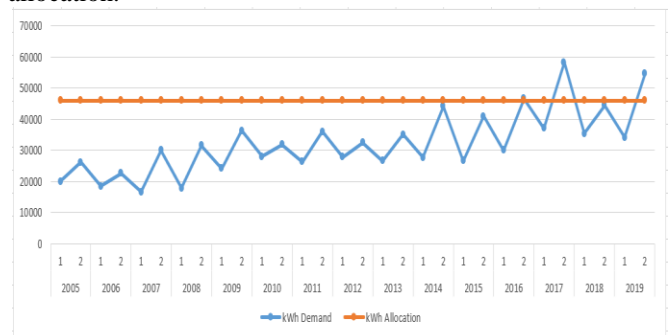


Figure 7: Loading Condition of DT 1

Figure 8 shows that DT 2's loading condition did not reach the kWh allocation. Although it reached to 29250 kWh for the last half of 2017, it can be observed that the gap between the kWh allocation and the kWh Demand is wider.

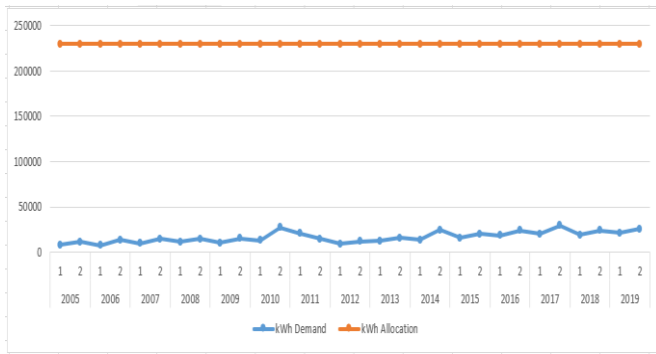


Figure 8: Loading Condition of DT 2

The kWh of DT 3 in Figure 9 shows nearly a normal loading condition. In the last half of 2016, the kWh demand reached the maximum kWh allocation. Starting 2017 the gap of the kWh demand and the kWh allocation became narrower as shown.

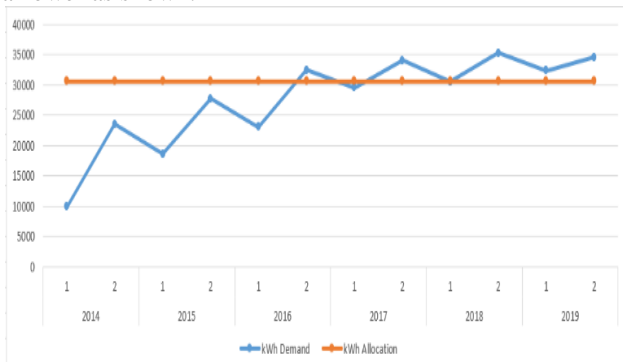


Figure 9: Loading Condition of DT 3

**B. Forecasting per semi-annual for the next 5 years**

Six forecasting model was used for the transformers to test the best of fit of kWh demand shown in Table 1, 2, 3 and 4, respectively. The chosen model is based on the values of the measure of accuracy which are Mean Average Error (MAE), Mean Square Error (MSE) and Mean Absolute Percent Error (MAPE) that were acquired using QM for Windows [14-15]. The lower the values of the MAE, MSE and MAPE the better the fit and useful of the forecasted model.

For DT 1, the 4<sup>th</sup> degree polynomial model had been chosen because it has the least value of MAE, MSE, and MAPE as seen in Table 1.

Table 1: Measure of Accuracy for DT 1

Function	Test of Accuracy		
	MAE	MSE	MAPE
Linear	5405.61	43260480	17.22%
Exponential	5425.468	43066730	17.13%
Logarithmic	6050.986	54448960	20.08%
2nd Degree Polynomial	5412.667	42699600	17.34%
3rd Degree Polynomial	5351.055	42309390	17.16%
4th Degree Polynomial	5340.626	42306100	17.12%

The 4<sup>th</sup> degree polynomial model for the semi-annual kWh demand from 2005 - 2019 is expressed as

$$f(x) = a_1x^4 + a_2x^3 + a_3x^2 + a_4x + a_5 \quad (4)$$

where

$$\begin{aligned} a_1 &= 0.0166 \\ a_2 &= 0.2074 \\ a_3 &= -25.46 \\ a_4 &= 1092.8 \end{aligned}$$

$$a_5 = 18897$$

x = number of semi-annuals

And the per semi-annual for the next five-years forecast using 4<sup>th</sup> degree polynomial model was shown in Figure 10

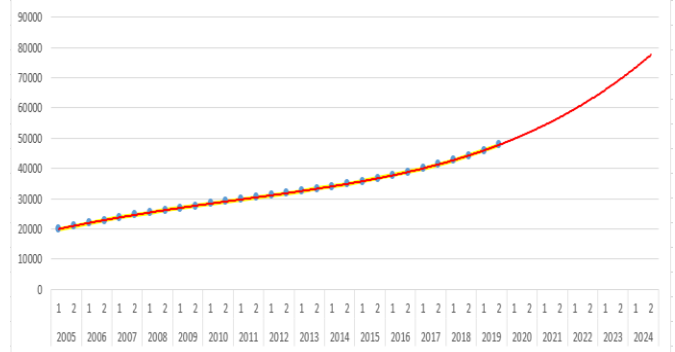


Figure 10: DT 1 Semi-Annual 4<sup>th</sup> degree Polynomial Model

In the case of DT 2, 3<sup>rd</sup> degree polynomial model was chosen second lowest value of MAE, MSE and MAPE compare for the 4<sup>th</sup> degree polynomial model due to the forecasted values of 4<sup>th</sup> degree polynomial was negative in values.

Table 2: Measure of Accuracy for DT 2

Function	Test of Accuracy		
	MAE	MSE	MAPE
Linear	3278.364	17183380	21.36%
Exponential	3128.823	17327910	19.76%
Logarithmic	3607.636	19834880	24.60%
2nd Degree Polynomial	3234.395	17093690	21.12%
3rd Degree Polynomial	3142.907	16748570	20.44%
4th Degree Polynomial	3089.203	16264560	20.03%

For DT 2, the forecasted semi-annual was attained by using the 3<sup>rd</sup> degree polynomial model

$$f(x) = b_1x^3 + b_2x^2 + b_3x + b_4 \quad (5)$$

where

$$\begin{aligned} b_1 &= 1.1604 \\ b_2 &= -49.48 \\ b_3 &= 1039.5 \\ b_4 &= 7575.2 \end{aligned}$$

x = number of semi-annuals

The regression model of the 3<sup>rd</sup> degree polynomial model in semi-annual in five-years ahead can be seen in Figure 11.

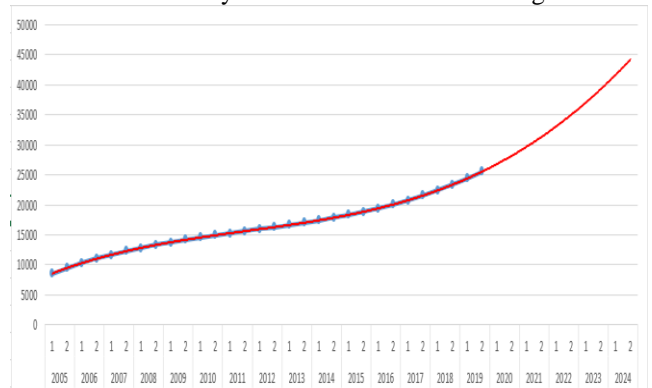


Figure 11: DT 2 Semi-Annual 3<sup>rd</sup> degree Polynomial Model



In Table 3, the lowest MAE, MSE and MAPE is the 4<sup>th</sup> degree polynomial model. But the forecasted value just like what happen to DT 2, the values were negative. The next best model must be the 3<sup>rd</sup> degree polynomial model which is the second lower value in MAE, MSE and MAPE.

**Table 3: Measure of Accuracy for DT 3**

Function	Test of Accuracy		
	MAE	MSE	MAPE
Linear	3392.816	15125570	16.26%
Exponential	3929.657	21564780	17.19%
Logarithmic	2596.677	8307616	11.04%
2nd Degree Polynomial	2738.145	9255675	12.24%
3rd Degree Polynomial	2576.398	8740315	11.21%
4th Degree Polynomial	2576.917	8495130	10.87%

For DT 3 the 3<sup>rd</sup> degree polynomial model for the semi-annual kWh demand is interpret as

$$f(x) = c_1x^3 + c_2x^2 + c_3x + c_4 \quad (6)$$

where

$$c_1 = 23.107$$

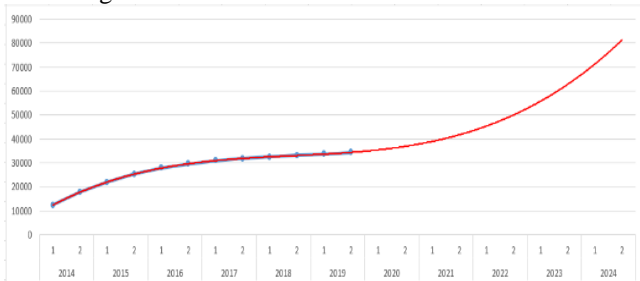
$$c_2 = -680.31$$

$$c_3 = 7211.4$$

$$c_4 = 5925.7$$

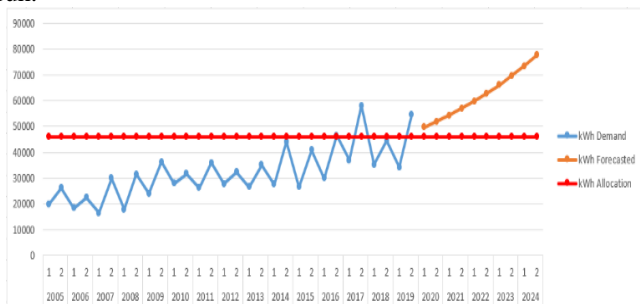
$$x = \text{number of semi-annuals}$$

And the semi-annual kWh demand in five-year ahead prediction using the 3<sup>rd</sup> degree polynomial model can be seen in Figure 12.



**Figure 12: DT 3 Semi-Annual 3<sup>rd</sup> Degree Polynomial Model**

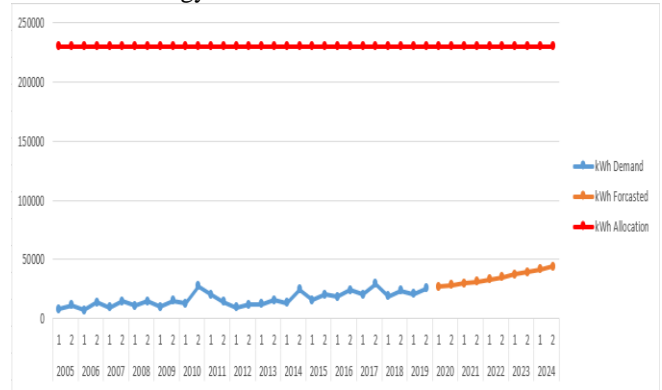
The administration must take into consideration is the gap between the kWh demand and kWh Allocation of the transformer for maximum utilization. The researcher has used the 4<sup>th</sup> degree polynomial model to project the five (5) years ahead semi-annual kWh demand for DT 1, as seen in Figure 13. The graph shows that the transformer will exceed the allocated kWh which will cause power failure in the long run.



**Figure 13: DT 1 kWh Demand, kWh Forecasted and kWh Allocation**

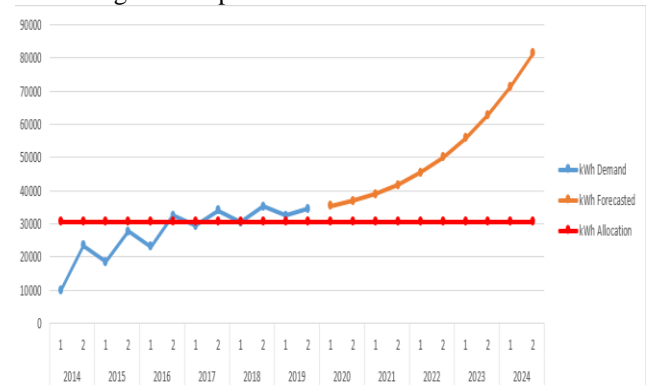
In the case of DT 2, the 3<sup>rd</sup> degree polynomial model applied still there is a large gap between kWh demand before it can reach the kWh allocation as seen in Figure 14. Even in the forecasted per semi-annual kWh demand for

five years ahead there are still more rooms for the additional demand for energy.



**Figure 14: DT 2 kWh Demand, kWh Forecasted and kWh Allocation**

DT 3 just like DT 1 it also exceeds the kWh allocation as shown in Figure 15. For the forecasted values it shows that the transformer will become overloaded that will be a disadvantage to its operation.



**Figure 15: DT 3 kWh Demand, kWh Forecasted and kWh Allocation**

#### IV. CONCLUSION

The time-series analysis shows that from 2005 to 2019 kWh Demand of the four transformers installed were increased evidently. DT1 kWh demand per semi-annual increased at a rate of 1.31 percent and its peak kWh demand reached 58114. This case is an overloaded condition that can cause power interruption and can damage the transformer. DT 2 had increased by 1.44% per semi-annual since this transformer is not overloaded. And even it has a maximum kWh demand of 44523 for the forecasted five (5) years ahead still there is more room for additional loading it can cater. DT 3, on the other hand, increased with a rate of 4.84 percent per semi-annual. In this case, transformer damage may occur on this transformer because it operates in an overloaded condition with a maximum kWh demand value of 35280.

#### RECOMMENDATIONS

For recommendations, quantifying and identifying the electric load of building connected is necessary to provide necessary measures to limit or conserve energy to make the transformers maintain normal loading conditions.

# kWh Demand Assessment and Forecasting of a Low Voltage Private Academic Institution

The two over utilized transformer DT 1 and 3, may be considered to relocate some electric load to the underutilized transformer. For future study, auditing of the connected load may be taken to account to optimize the usage of the transformer.

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## AUTHORS PROFILE



**Joy R. Antonio**, is the youngest and only daughter and a mother of one. She recently moved to Nueva Ecija and finished her degree at Nueva Ecija University of Science and Technology. She passed the board exam titled as a registered Electrical Engineer and is currently affiliated as a service contractor in a government institution, the

Philippine Rice Research Institute Central Experiment Station located at Maligaya, Science City of Munoz, Nueva Ecija under the Department of Agriculture. The Institution aims to develop high-yielding rice varieties and cost reducing technologies so farmers can produce enough rice, available,

affordable and accessible for all Filipino's. With her passion, she was recently promoted as an Engineer II to overlook all electrical works (proposed infra projects, on-going infra projects, and repair and maintenance works) for PhilRice Central Experiment Station and its branches nationwide. She is designated as a focal person for the Energy Efficiency and Conservation under PhilRice CES programmed by the Department of Energy.



**Antonio S. Buendia Jr.**, The author is an engineer from Nueva Ecija's top electric company. He is happily married to his wife Janet with their daughter Mary Antonette. The author, as a child has only a simple dream, to uplift his family's life from poverty. He and his family doesn't have an easy life. They have to work for the day's meal or else they do not have any to put on the table. And because of the dream to be an engineer, he served as a service crew in a fast food chain serving as a part-timer in night and an engineering students during the day. Though a master in multi-tasking, the author is a consistent top student from primary to college where he pursued the Electrical Engineering Course on the top engineering school in Asia-the Nueva Ecija University of Science and Technology. He got his career service recognition as a licensed electrical engineer and a master electrician during the year 2001, in this sense, he soar higher into his career. He is now currently affiliated with the Tarlac Electric Company as a distribution system technical specialist, he's with the company since the year 2002. With his earnest passion for learning, he continued his graduate studies in the same school he earned his bachelors. And as a curious engineer he began starting doing researches with his colleagues, with their recent work "kWh Demand Assessment and Forecasting of a Low Voltage Private Academic Institution". This is just one of the many researches that he will affiliated with.



**Gina B. Narca**, is a Professional License Civil Engineer, a mother of three (3) and presently employed as public servant for more than two (2) decades in one of the reputable government institution in the country. The Philippine Rice Research Institute, Central Experiment Station located at Maligaya, Science City of Munoz, Nueva Ecija under the Department of Agriculture it aims to develop high-yielding rice varieties and cost reducing technologies so farmers can produce enough rice, available, affordable and accessible for all Filipino's. Some awards were received by the author on her unconditional, dedication and hardworking to her tasked, two awards are being received, two times as Outstanding Admin Support Staff and a highest award given by the Head of the Organization or Agency the "Director's Award". The author has a good standing membership for her unduly support in every activity conducted by the organization (National and Provincial Chapter) that gave the members up to date information and technology. The name of the organization is the "Philippine Institute of Civil Engineers, Inc." **PICE**.



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