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Data Article

Data based investigation of the energy metering type, billing and usage of sampled residents of Ota Community in Nigeria



Aderibigbe Israel Adekitan^{a,*}, Bukola B. Adetokun^b, Alex Aligbe^a, Tobi Shomefun^a, Abidemi Orimogunje^a

^a Department of Electrical and Information Engineering, Covenant University, Ota, Nigeria

^b Department of Electrical Engineering, Pan African University Institute for Basic Sciences, Technology and

Innovation (PAUISTI), Jomo Kenyatta University of Agriculture and Technology (JKUAT), Kenya

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ABSTRACT

Accurate energy metering and billing is a challenge in some developing countries. In Nigeria, the issues of inadequate power generation, transmission and distribution infrastructure are recurrent problems, coupled with inefficient energy metering which is a major problem that results in residential energy consumers being billed unfair energy charges by the Electricity Distribution Companies (DISCOs) for unused energy, and this has been termed "crazy bill". For the energy sector to be effective, energy bills should be based on the actual energy usage and likewise customers must pay for used energy. To achieve this, the Nigerian Electricity Regulatory Commission (NERC) recommended the installation of prepaid meters for all customers, but as at today, there is no full compliance with this regulation. Power supply is grossly unreliable, and this has affected power quality due to frequent load shedding and power outages. The dataset presented in this article captures the type of apartment, the type of electrical appliances used by occupant, the average monthly energy bill paid for electricity, the use or non-use of alternative energy sources, the type of alternative energy sources used, and the type of energy meter used by sampled residents of the Ota community in Ogun

* Corresponding author.

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E-mail address: aderibigbe.adekitan@covenantuniversity.edu.ng (A.I. Adekitan).

State, Nigeria. The dataset was acquired using an investigative questionnaire to survey the residential consumers within the sampled space.

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Specifications Table

Subject area More specific subject area Type of data How data was acquired	Electrical Engineering Energy Metering and Billing, Electrical Equipment Profiling Tables, figures and spread sheet file Data acquisition using an investigative questionnaire, structured with targeted questions to determine respondent's energy consumption and the mode of energy metering and billing
Data format	Raw, filtered, analysed
Experimental factors	Residents of estates where energy dependence is solely on off-grid, estate-owned power generation schemes were excluded from the study. Only residents that use the public power supply from the Power Dis- tribution Company were sampled
Experimental features	Frequency distributions and statistical analysis were performed to illustrate metering and billing practices, available electrical energy sources, average DISCO power supply duration and the consumers' electrical equipment profile
Data source location Data accessibility	Residents of Ota community, around Covenant University in Nigeria The dataset is available in a spreadsheet file attached to this article

Value of the data

- The dataset presents the energy metering, billing and usage of a sampled community in Nigeria which is representative of the general consumer experience in Nigeria, and as such, the dataset may serve as an indicator for the level of supply of prepaid meters by Electricity Distribution Companies (DISCOs) to residential consumers, in line with the directives of the Nigerian Electricity Regulatory Commission (NERC).
- The dataset may be of interest to researchers studying the billing experience of consumers that are still using analog meter for electricity metering.
- The tables, frequency distribution, and figures presented can provide vital insights which may enable these data to be compared with similar data collected in other geographical locations within the country for billing and metering pattern recognition based on consumer's location.
- These data may be useful for future studies comparing different methodological approaches to consumer billing and tariff rate determination.
- The availability of this data may trigger similar evidenced based empirical research studies [1], and this may create platforms for extensive collaboration.

1. Data

The data captures the different modes of electricity billing in Nigeria, and these are metered (prepaid and analog meters) and unmetered (estimated billing). Estimated billing often results in exorbitant charges termed "crazy bill" which is usually far above the actual energy consumption, and this is unfair to the customer [2]. The analog meter was the only alternative to estimated billing until



Fig. 1. The types of apartment surveyed.



Fig. 2. The number of occupants per household.



Fig. 3. Energy meter availability in households.

in recent years with the advent of prepaid meters [3,4]. Most of the analog meters were installed decades ago and are no longer accurate while some have been tampered with to slow their reading or to stop it completely from metering the energy consumed. NERC has mandated all DISCOs to install



Fig. 4. The type of energy meter installed.



Fig. 5. Sharing of one energy meter with landlord or neighbor.

prepaid meters for all their customers but this is yet to be fully complied with, as DISCOs complain of lack of funds to procure the needed prepaid meters [5]. The data captures the proportion of the sampled residents that are on prepaid meters, it also reflects the opinion of the customers of their current monthly energy charges, it reflects the average hours of power supply by the DISCO to the residents due to insufficient power generation and load shedding [6–10], and also, it presents the profile of the commonly used electrical equipment, used within the community. According to [11,12] the type of electrical equipment used and the behavioral energy usage trend determines a customer's monthly electricity bill. Figs. 1 and 2 describe the types of accommodation sampled and the number of people per household, Figs. 3–9 captures the mode of energy billing, the view of energy consumers of DISCO charges, and the use of alternative energy sources to compensate inadequate public power supply [13]. Table 1 presents the descriptive statistics of the energy cost data while Figs. 10–14 present the boxplots of responses to questions on energy charges and average power supply duration. Fig. 15 shows the variation in Naira between DISCO's monthly charges and the expected fair usage



Fig. 6. Mode of energy billing and payment.



Fig. 7. The use of alternative energy sources by Respondents.



Fig. 8. The type of alternative energy source used.



Fig. 9. Respondents' opinion on DISCO monthly charges.

Table 1Descriptive statistics of energy cost.

	Monthly cost of running generator	Average DISCO monthly charges	Respondents cost estimate of actual consumption	Difference between DISCO charge and consumer's expectation
Mean	6559.94	3675.85	1956.19	2446.17
Sum	1,023,350.00	779,280.00	410,800.00	364,480.00
Min	700.00	400.00	200.00	200.00
Max	40,000.00	26,000.00	20,000.00	24,000.00
Range	39,300.00	25,600.00	19,800.00	23,800.00
Variance	44,655,820.00	11,270,112.55	3,698,621.78	7,528,734.59
Standard	6682.50	3357.10	1923.18	2743.85
Deviation				
Standard	535.03	230.57	132.71	224.79
Error of				
Mean				
Median	5000.00	3000.00	1500.00	2000.00
Mode	5000.00	2000.00	1000.00	2000.00
Count	156	212	210	149

estimate by each respondent. Figs. 16–18 detail the opinion of the respondent on the quality of the voltage supply and a summary of the electrical equipment used in the community. Various statistical analyses were performed on the dataset using methods similar to those found in [14,15]. The statistical model of Fig. 19 was analyzed using Partial Least Squares approach to Structural Equation Modeling (SEM) [16]. Table 2 shows the Variance Inflation Factor (VIF) while Table 3 depicts the direct relationship of the hypothesis considered. Table 4 shows the establishment of the discriminant validity. The F square values are shown in Fig. 20 while the path coefficient histograms are illustrated in Figs. 21–23.



Fig. 10. A box plot of the average monthly expenses on generator.



Fig. 11. A box plot of the average DISCO Monthly Charges.

2. Experimental design, materials and methods

These data were obtained by means of an investigative instrument captured by the questionnaire shown in the Appendix. 214 households were investigated and the obtained data fairly represent the various types of apartment available. The instrument is divided into seven parts labelled A to G. Part A was used to obtain the data regarding the type of apartment and the number of occupants. Part B covers methods of billing, metering information, and alternative back up generation used by consumers. Data on the average monthly payment was obtained in Part C, while Part D obtained the average daily duration and voltage level quality of electricity supply. In Part E, information relating to occupation and monthly income were gathered while Part F captured the electrical appliances used in



Fig. 12. A box plot of respondents' opinion of fair energy charge.



Fig. 13. A box plot of the difference between DISCO Charges and Consumers' Estimate.

each apartment. Lastly, the behavioural energy usage pattern is obtained from Part G. The collected data was processed, and various statistical analyses were carried out to determine the influence of power supply quality, the average duration of power supply per day, households' monthly expense on energy, the use of energy meters and the availability of alternative energy sources on the anticipated households energy demand, using reflective constructs.



Fig. 14. A box plot of the average number of hours of DISCO daily supply.



Fig. 15. A chart showing the difference between DISCO Charges and Consumers' Estimate.



Fig. 16. Scatter diagram of the average number of hours of DISCO daily supply.



Fig. 17. Respondents' opinion of the quality of DISCO voltage supply.



Fig. 18. Electrical equipment distribution across sampled households.



Fig. 19. Statistical measurement model for the energy consumers' data.

Table 2 Collinearity Statistics - Variance Inflation Factor (VIF).

	Energy metering and alternatives	Household usage	Monthly energy expenses	Power quality and duration
Energy metering and alternatives Housebold usage		1.100		
Monthly energy expenses Power quality and duration	1.002 1.002	1.009 1.098		1.000

VIF < 5 - Acceptable (No strong indication of multicollinearity).

Table 3

Direct Relationships for Hypothesis testing.

Hypothesis	Relationship	Std. beta	Std. error	t-value	p Values	95% CL LL	95% CL UL
H1	Energy Metering and Alternatives - > Household Usage	0.222	0.071	3.101**	0.002	0.106	0.335
H2	Power Quality and Duration - > Household Usage	0.012	0.069	0.246	0.806	- 0.095	0.124
H3	Monthly Energy Expenses - > Household Usage	0.321	0.073	4.139	0	0.199	0.446
H4	Monthly Energy Expenses - > Power Quality and Duration	- 0.052	0.073	0.726	0.468	- 0.173	0.061
Н5	Power Quality and Duration - > Energy Metering and Alternatives	0.289	0.063	4.441**	0	0.184	0.382
H6	Monthly Energy Expenses - > Energy Metering and Alternatives	0.069	0.11	0.75	0.454	- 0.12	0.22

CL LL - Confidence Limit Lower Limit.

CL UL - Confidence Limit Upper Limit.

** p < 0.05 – Significant.

Table 4

Discriminant validity check using Fornell-Larcker Criterion.

	Energy metering and alternatives	Household usage	Monthly energy expenses	Power quality and duration
Energy metering and alternatives	0.789			
Household usage	0.236	0.86		
Monthly energy expenses	0.064	0.293	0.849	
Power quality and duration	n 0.29	0.065	- 0.048	0.766



Fig. 20. F square: effect size impact indicator (F square < 0.02 - No effect [17,18]).



Energy Metering and Alternatives -> Household Usage











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Appendix

DEPARTMENT OF ELECTRICAL AND INFORMATION ENGINEERING COVENANT UNIVERSITY					
ELECTRICITY BILLING SYSTEM SURVEY Dear Respondent, This questionnaire is <u>strictly for the purpose of academic research</u> Please kindly provide response to each question objectively and sincerely. Every detail provided shall be treated as confidential. Thank you.					
State					
A1. What type of house/apartment do you reside? 1 room □ 1 room and parlour □ 1 room and parlour self-contained □ 2-bedroom flat □ 4-bedroom flat □ Duplex □ Others □ please specify A2. How many of you live in this apartment?					
B1. Do you have any meter installed in your house/apartment? YES □ NO □ B2. If YES, is it a PREPAID METER? YES □ NO □ What type is your PREPAID METER? OLD type □ NEW type □ B3. Do you share meter with landlord/tenant/neighbour? YES □ NO □ B4. If you do not have prepaid meter, have you/your landlord ordered/paid for one from PHCN? YES □ NO □ B5. Since when did you/your landlord ordered/paid for the prepaid meter? B6. How do you pay for electricity consumed? Prepaid meter □ Analog PHCN meter □ Estimated PHCN bill □ Payment to PHCN officials without bill/receipt □ Do Not pay □ Others □ Please specify B7. Do you have alternative means of generating electricity apart from PHCN supply? YES □ NO □ B8. If YES, which type of alternative generating set do you have? Diesel Generator □ Petrol Generator □ Battery and Inverter □ Solar panel, battery and Inverter □ Others □ Please specify					
C1. On the average, how much do you pay to PHCN for electricity used per month? C2. Do you think this amount is too much for you? YES D NO D C3. How much do you think you should pay for electricity you consume each month?					
D1. On the average, how many hours of electricity supply do you use/enjoy per day? D2. What is the quality of voltage supply to your house? Very low □ Low □ Normal □ High □ Very high □ D3. How often do you experience the condition in D2 above? Always □ Almost all the time □ Sometimes □ Rarely □ Never □					
 E1. Which of the following best describes your occupation? Self-establishment □ Others □ Please specify E2. How much do you/your household earn per month on the averation of the self self self self self self self sel	employed Government employee Employee of private				
G. Please kindly indicate the extent to which you agree/disagree with S/N Questions 1. I often keep my electric bulb ON when sleeping 2. I often keep my electric bulb ON when out of the house 3. I prefer to cook with electric cooker when there is power supply 4. I switch ON the fan/AC only when the room is hot and stuffy 5. Whenever I am in the room, I switch ON the fan/AC 6. I leave my freezer/fidge ON all the time 7. I switch off all electrical appliances whenever I leave my house 8. I know about energy efficient devices/equipment	h each of the following statements. Strongly Agree Agree Strongly Disagree Disagree NA Image: Image				
7. I use eus gy europai devites equipment					

Transparency document. Supporting information

Transparency document associated with this article can be found in the online version at https://doi.org/10.1016/j.dib.2018.07.047.

Appendix A. Supplementary material

Supplementary data associated with this article can be found in the online version at https://doi.org/10.1016/j.dib.2018.07.047.

References

- S.I. Popoola, A.A. Atayero, T.T. Okanlawon, B.I. Omopariola, O.A. Takpor, Smart campus: data on energy consumption in an ICT-driven university, Data Brief 16 (2018) 780–793.
- [2] D.F. Arawomo, Electricity billing systems and household electricity use behaviour in Ibadan, Nigeria, Energy Sustain. Dev. 40 (2017) 77–84.
- [3] K.C. O'Sullivan, H.E. Viggers, P.L. Howden-Chapman, The influence of electricity prepayment meter use on household energy behaviour, Sustain. Cities Soc. 13 (2014) 182–191.
- [4] F.M. Mwaura, Adopting electricity prepayment billing system to reduce non-technical energy losses in Uganda: lesson from Rwanda, Uti. Policy 23 (2012) 72–79.
- [5] M.O. Oseni, Assessing the consumers' willingness to adopt a prepayment metering system in Nigeria, Energy Policy 86 (2015) 154-165.
- [6] A.I. Adekitan, B. Adetokun, T. Shomefun, A. Aligbe, Cost implication of line voltage variation on three phase induction motor operation, Telkomnika (Telecommun. Comput. Electron. Control) 16 (2018).
- [7] I.A. Samuel, J. Katende, C.O.A. Awosope, A.A. Awelewa, Prediction of voltage collapse in electrical power system networks using a new voltage stability index, Int. J. Appl. Eng. Res. 12 (2017) 190–199.
- [8] A. Olukoju, 'Never Expect Power Always': electricity consumers' response to monopoly, corruption and inefficient services in Nigeria, Afr. Aff. 103 (2004) 51–71.
- [9] C.A. Awosope, Nigeria electricity industry: issues, challenges and solutions, in: Covenant University 38th Public Lecture, Public Lecture Series. vol. 3, ed, 2014.
- [10] T.R. Ayodele, A.S.O. Ogunjuyigbe, B.B. Adetokun, Optimal capacitance selection for a wind-driven self-excited reluctance generator under varying wind speed and load conditions, Appl. Energy 190 (2017) 339–353.
- [11] H. Wilhite, H. Nakagami, T. Masuda, Y. Yamaga, H. Haneda, A cross-cultural analysis of household energy use behaviour in Japan and Norway, Energy Policy 24 (1996) 795–803.
- [12] M.H. Ishak, A.H.M. Iman, M. Sapri, Theoretical postulation of energy consumption behaviour assessment in Malaysian higher education institutions, Procedia Soc. Behav. Sci. 65 (2012) 891–896.
- [13] A.F. Adenikinju, Electric infrastructure failures in Nigeria: a survey-based analysis of the costs and adjustment responses, Energy Policy 31 (2003) 1519–1530.
- [14] A.I. Adekitan, T. Shomefun, T.M. John, B. Adetokun, A. Aligbe, Dataset on statistical analysis of jet A-1 fuel laboratory properties for on-spec into-plane operations, Data Brief 19 (2018) 826-834.
- [15] A.I. Adekitan, O. Omoruyi, Stock keeping accuracy: a data based investigation of storage tank calibration challenges, Data Brief (2018).
- [16] C.M. Ringle, S. Wende, J.-M. Becker, "SmartPLS 3", Boenningstedt: SmartPLS GmbH, (http://www.smartpls.com), 2015.
- [17] J. Cohen, Statistical Power Analysis for the Behavioral Sciences, 2nd ed., Erlbaum Associates, Hillsdale, 1988.
- [18] J. Henseler, C.M. Ringle, R.R. Sinkovics, The use of partial least squares path modeling in international marketing, in: New Challenges to International Marketing, ed: Emerald Group Publishing Limited, 2009, pp. 277–319.