ELECTRICITY CONSUMPTION PATTERN DISAGGREGATION BASED ON USER UTILIZATION FACTOR

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To my mother, my family, my supervisors, & my fellow friends,

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ABSTRACT

Non-Intrusive Appliance Load Monitoring (NIALM) technique has been studied intensively by many researchers to estimate the electricity consumption of each appliance in a monitored building. However, the method requires a detailed, secondby-second power consumption data which is commonly not available without the use of high specification energy meter. The common energy meter used in buildings can only capture low frequency data such as kWh for every thirty minutes. This thesis proposes a bottom-up approach for disaggregating kWh consumption of a building. The relationship between the load profile of a building and electricity usage pattern of the occupants were studied and analysed. From the findings, a method based on utilization factor that relates user usage pattern and kWh electricity consumption was proposed to perform load disaggregation. The method was applied on the practical kWh profile data of electricity consumption of Block P19a, Fakulti Kejuruteraan Elektrik, Universiti Teknologi Malaysia. The disaggregated kWh consumption results for air-conditioning and lighting system were validated with the actual kWh consumption recorded at the respective branch circuits of the building. Results from the analysis showed that the proposed method can be used to disaggregate energy consumption of a commercial building into air-conditioning and lighting systems. The proposed method could be extended to disaggregate the energy consumption for different areas of the building.

ABSTRAK

Teknik Pemantauan Perkakas Beban Tidak Bergantung (PPBTB) telah dikaji secara intensif oleh ramai penyelidik untuk menganggarkan penggunaan elektrik setiap perkakas di dalam sesebuah bangunan yang dipantau. Walau bagaimanapun, kaedah ini memerlukan secara terperinci, data penggunaan kuasa setiap saat yang biasanya tidak boleh didapati tanpa menggunakan meter tenaga berspesifikasi tinggi. Meter tenaga yang kebiasaannya digunakan di dalam bangunan hanya boleh merekod data frekuensi rendah contohnya kWh bagi setiap tiga puluh minit. Tesis ini mencadangkan teknik bawah-ke-atas untuk mengasingkan data penggunaan kWh bagi sesebuah bangunan. Hubungan antara profil beban bangunan dan corak penggunaan elektrik oleh penghuni telah dikaji dan dianalisis. Dari hasil kajian, kaedah berdasarkan faktor penggunaan yang menghubungkan corak penggunaan pengguna dan penggunaan elektrik kWh telah dicadangkan untuk melaksanakan pengasingan beban. Kaedah ini diaplikasikan secara praktikal dengan menggunakan profil data kWh di Blok P19a, Fakulti Kejuruteraan Elektrik, Universiti Teknologi Malaysia. Keputusan pengasingan bagi penggunaan kWh untuk sistem penghawa dingin dan pencahayaan telah disahkan dengan penggunaan kWh sebenar yang dicatatkan daripada litar cawangan bangunan masing-masing. Keputusan daripada analisis menunjukkan bahawa kaedah yang dicadangkan boleh digunakan untuk mengasingkan penggunaan tenaga bangunan komersial kepada sistem penghawa dingin dan lampu. Kaedah yang dicadangkan juga boleh digunakan lebih meluas untuk mengasingkan penggunaan tenaga berdasarkan setiap ruang bangunan yang berbeza.

TABLE OF CONTENTS

CHAPTER		TITLE	PAGE
	DEC	LARATION	ii
	DED	ICATION	iii
	ACK	NOWLEDGEMENT	iv
	ABS	ТПАСТ	V
	ABS	TRAK	vi
	TAB	LE OF CONTENTS	vii
	LIST	COF TABLES	Х
	LIST	COF FIGURES	xii
	LIST	COF ABBREVIATIONS	XV
	LIST	COF SYMBOLS	xvii
	LIST	COF APPENDICES	xviii
1	INTI	RODUCTION	1
	1.1	Research Background	1
	1.2	Problem Statement	3
	1.3	Objectives of the Research	4
	1.4	Scope of the Research	4
	1.5	Significance of the research	5
	1.6	Thesis outline	6
2	LITE	ERATURE REVIEW	7
	2.1	Introduction	7

2.2	Energy Management 8				
2.3	Non-Intrusive Appliance Load Monitoring 1				
	2.3.1 Low-Frequency Hardware Installation	13			
	2.3.1.1 Real power and reactive power				
	changes	13			
	2.3.1.2 Real power and reactive power				
	with additional "macroscopic"				
	signatures changes	16			
	2.3.1.3 Changes of real power only	20			
	2.3.2 Higher Frequency Sampling Hardware	26			
	2.3.3 Comparison of different Non-Intrusive				
	Appliance Load Monitoring Method	32			
2.4	Bottom up approach	36			
	2.4.1 Statistical random model	38			
	2.4.2 Probabilistic empirical models	40			
	2.4.3 Time of use based models	41			
2.5	Chapter Summary	44			

3	ME	THODOLOGY	45
	3.1	Introduction	45
	3.2	Implementation Plan	45
	3.3	Data Collection	47
		3.3.1 Load Profile Data	47
		3.3.2 Walk-through audit	50
		3.3.3 Building specification	52
	3.4	Proposed Load Disaggregation method	53
	3.5	Summary of Chapter	61
4	RES	SULTS AND DISCUSSIONS	62
	4.1	Introduction	62
	4.2	Collected load profile data for analysis	63
	4.3	Load disaggregation results	66

		4.3.1	Utilization	factor		66
	4.4	Estimat	ion of Ligh	ting and A	ir-Conditioning	
		Energy	Consumpti	on		69
		4.4.1	Validation	of Load	Disaggregation	
		results				73
	4.5	Electric	ity consum	ption profi	le of individual	81
		area				01
	4.6	Summa	ry of Chapt	er		85
5	CON	CLUSIC	DNS	AND	FUTURE	
	RECO	OMME	NDATION	S		86
	5.1	Conclus	sion			86
	5.2	Recom	mendations	for future v	work	87
REFERENCES						89

Appendices A-C	98-112
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LIST OF TABLES

TABLE NO.

TITLE

PAGE

1.1	Regional and sectoral electricity consumption in	
	Malaysia, 2014	1
2.1	Fragment of a life style matrix	37
2.2	Classification of occupancy pattern for a three-	
	person household	38
3.1	Power input of lighting system according to specified	
	area	52
3.2	Power input of air-conditioning system according to	
	specified area	53
3.3	Electricity usage ratings	56
3.4	Example of utilization factor representation	57
3.5	Utilization factor limits	58
3.6	Example of estimated energy consume by lighting	
	system of overall laboratory in P19a building	59
4.1	Usage ratings defined by occupants of P19a for	
	lighting system	66
4.2	Usage ratings defined by occupants of P19a for air	
	conditioning system	67

4.3	Utilization factor result for lighting system	68
4.4	Utilization factor result for air-conditioning system	68
4.5	Result of load disaggregation on 18 February 2015	69
4.6	Comparison of actual and proposed data on 18	
	February 2015	73
4.7	Table of percentage error for the proposed method on	
	18 February 2015	77
4.8	Table of error for remaining days	80
4.9	Energy consumption of individual area of lighting	
	system on 18 February 2015 (kWh)	82
4.10	Energy consumption of individual area of air-	
	conditioning system on 18 February 2015 (kWh)	84

LIST OF FIGURES

FIGURE NO. TITLE PAGE

1.1	Non-Intrusive Appliance Load Monitoring	
	Division	3
2.1	Types of behavior-based energy efficiency	
	strategies and approaches	8
2.2	Efficiency programs with different types of	
	feedback systems	9
2.3	Intrusive Load Monitoring process using smart	
	meter by measuring each appliance power	
	consumption	11
2.4	Non-intrusive Appliance Load Monitoring	
	disaggregate data to individual appliance power	
	consumption given aggregate data consumption	12
2.5	Power versus time shows step changes for total	
	load due to individual appliance events	15
2.6	Signature Space and appliance cluster (Reactive	
	versus Real Power)	15
2.7	Application of the method used in Hidden Markov	
	Model	19
	(a) LCD State Transition diagram	
	(b) Hidden Markov Model (HMM) graphical	
	representation	

	(c) Appliance HMMs are organized in a particular	
	structure to form a Factorial HMM and define a	
	combined load model.	
2.8	Noise FFT as a features	32
	(i) Power line shows background noise detected.	
	(ii) After a new device is turned on, new signal that	
	create EMI is present.	
	(iii) After background subtraction the new signal	
	features are extracted. The features are amplitude	
	(A), mean (μ) and variance (σ) of the Gaussian fit	
3.1	Implementation plan flow chart	46
3.2	Data collection process	48
	(a) Kyoritsu Kew Power Quality Analyzer used to	
	record power and energy consumption	
	(b) Connection between power analyzer at main	
	power input of air-conditioning system	
3.3	Overview of connection	49
	(a) single line diagram of installed meter	
	(b) diagram of 3-phase 4-wire connection between	
	power analyzer and load	
3.4	Hourly energy consumption of P19 building	49
3.5	Illustration of the proposed method	54
3.6	Flowchart of proposed load disaggregation method	55
4.1	Building's load profile on Wednesday (18/2/15)	63
4.2	Building's load profile on Monday (23/2/15)	64
4.3	Building's load profile on Tuesday (24/2/15)	64
4.4	Building's load profile on Wednesday (25/2/15)	64
4.5	Building's load profile on Thursday (26/2/15)	65
4.6	Building's load profile on Sunday (1/3/15)	65
4.7	Building's load profile on Monday (2/3/15)	65
4.8	Load profile of disaggregated data on 18 February	
	2015	70

4.9	Load profile of disaggregated data for	
	corresponding days 7	72
	(a) 23 February 2015	
	(b) 24 February 2015	
	(c) 25 February 2015	
	(d) 26 February 2015	
	(e) 1 March 2015	
	(f) 2 March 2015	
4.10	Comparison of actual data and proposed data load	
	profile 7	75
	(a) lighting system	
	(b) air-conditioning system	
4.11	Comparison of disaggregated data between actual	
	and proposed data 7	76
4.12	Comparison of disaggregated data between actual	
	and proposed data for corresponding days 7	79
	(a) 23 February 2015	
	(b) 24 February 2015	
	(c) 25 February 2015	
	(d) 26 February 2015	
	(e) 1 March 2015	
	(f) 2 March 2015	
4.13	Load profile of lighting system for individual area 8	33
4.14	Load profile of air-conditioning system for	
	individual area 8	35

LIST OF ABBREVIATIONS

AC	-	Air-conditioning system
ANNOT	-	Automated Electricity Data Annotation
APEC	-	Asia-Pacific Economic Cooperation
CU	-	Concordia University
CHP	-	Combine Heat Power
CPU	-	Central Processor Unit
DHW	-	Domestic hot-water
DSM	-	Demand Side Management
DSP	-	Data Signal Processor
EE	-	Energy Efficiency
EMI	-	Electromagnetic Interference
FFT	-	Fast Fourier Transform
FL	-	Lighting system
FKE	-	Fakulti Kejuruteraan Elektrik
FHMM	-	Factorial Hidden Markov Model
FSM	-	Finite State Machines
Н	-	High
HELP	-	Heuristic End-Use Load Profiler
HMM	-	Hidden Markov Model
HVAC	-	Heating, ventilation and air-conditioner
IALM	-	Intrusive Appliance Load Monitoring
kHz	-	Kilohertz
kW	-	Kilowatt
kWh	-	Kilowatt-hour

MOHE	-	Ministry of Higher Education's
NIALM	-	Non-Intrusive Appliance Load Monitoring
N-N	-	Neural-network
PE	-	Pencawang Elektrik
PF	-	Power factor
RBF	-	Radial Basis Function
RECAP	-	Recognition of electrical Appliances and Profiling
SMLP	-	Simple Method of formulating Load Profile
SMPS	-	Switch Mode Power Supplies
SVM	-	Support Vector Machines
TOU	-	Time of Use
T5	-	Fluorescent Tube Lamp Type T5
Τ8	-	Fluorescent Tube Lamp Type T8
UEC	-	Unit Energy Consumption
UK	-	United Kingdom
UNEP	-	United Nations Environment Program
UTM	-	Universiti Teknologi Malaysia
US	-	United States
VH	-	Very high
VL	-	Very low
VRV	-	Variable Refrigerant Volume
W	-	Watt
W/h	-	Watt per hour

LIST OF SYMBOLS

А	-	Utilization factor
i	-	Area
j	-	Time/hour
μ_j	-	mean electricity consumption at hour j
σ	-	Variance
\mathbf{P}_i	-	Power input
Ej	-	kWh electricity consumption
Ns	-	Total number of areas
PF	-	Power factor
P _{std}	-	Standard deviation for real power

LIST OF APPENDICES

APPENDIX	TITLE	PAGE	
А	List of Author's Publication	95	
В	Schedule of Split-Unit Air-Conditioning System	96	
С	Load Profile of P19a block	106	

CHAPTER 1

INTRODUCTION

1.1 Research Background

Asia-Pacific Economic Cooperation (APEC) Energy Demand and Supply Outlook estimate that electricity demand in Malaysia will significantly rise to 206 TW/h in 2035 from 96.3 TW/h in 2009 [1]. It is said that approximately 30% from the total demand nationwide is contributed by commercial buildings alone [2-4]. Table 1.1 presents regional and sectoral electricity consumption in Malaysia for year 2014. Thus, Malaysia must explore available initiatives to encourage efficient usage of the electricity demand for a better energy management control [5]. University buildings are also one of the high energy consumers and thus, education centres are advised by Ministry of Higher Education's (MOHE) to apply a better energy practice [6-8].

Table 1.1: Regional and	sectoral electricity	y consumption	in Malaysia, 2014

REGION	INDUSTRY GWH	COMMERCIAL GWH	RESIDENTIAL GWH	TRANSPORTATION GWH	AGRICULTURE GWH	TOTAL GWH
PENINSULAR MALAYSIA	46,755	37,108	23,721	261	413.5	108,259
SARAWAK	10,966	2,290	1,896	-	-	15,152
SABAH	1,230	2,043	1,647	-	-	4,919
TOTAL	58,951	41,441	27,264	261	414	128,330

There are many strategies to reduce electricity consumption such as demand side management (DSM) and energy efficiency (EE) program. DSM focuses on consumer demand adjustment in energy usage to control energy consumption at enduser through education or financial incentives. EE program is a system to manage and restraining the increase in energy demand. A building that provides same services for less energy input or more services for the same energy input is said to be energy efficient [2, 9]. There are several options to reduce energy consumption waste by providing consumer with their energy level consumption such as smart meter installation known as intrusive appliance load monitoring and analysis of current and voltage waveform through non-intrusive appliance load monitoring method.

Intrusive Appliance Load Monitoring (IALM) is known as one of the most accurate and reliable systems to recognize load consumption of an individual appliance. A few smart meter will be installed to recognized consumers' energy consumption in a building where each meter directly interacts with each involve appliances and finally decomposes the total energy consumption. Apparently, this method is not the common preferred option due to the high cost of installation that requires smart meter to be set up on each appliance.

Meanwhile, the study on Non-Intrusive Appliance Load Monitoring (NIALM) is widely evolving which focuses on load signature behaviour to perform load disaggregation. This method delivers reliable results which conforms with one of the objectives of feedback system that is to achieve appliance disaggregation. A NIALM is designed to monitor an electrical circuit that contains a number of appliances. NIALM estimates the number and nature of the individual loads, their individual energy consumption, and other relevant statistics such as time-of-day variations by a sophisticated analysis of the current and voltage waveforms of the total load. This method is often used to disaggregate overall consumption into individual energy usage of experimented appliances. Besides, the analysis requires less cost than the IALM technique because access to individual components is unnecessary for installing sensors or making measurements.

The study of NIALM is divided into two categories which are low sampling and high sampling installation. Figure 1.1 shows the NIALM method separated into several divisions, depending on the frequency of input data [10]. Low sampling installation normally analyses the behaviour of power changes and applies macroscopic features due to fundamental period that is 1/60 s or 1/50 s while higher frequency sampling hardware focuses on the study of harmonic and waveforms.

Various studies on NIALM could offer historical usage information to consumers which provide more detail in energy use comparisons and their energy usage per appliance data. Therefore, consumers who are well informed of their current energy usage will be more motivated to change their behaviours toward energy consumption to reduce electricity bill.

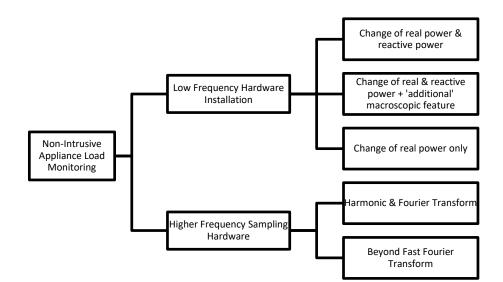


Figure 1.1 Non-Intrusive Appliance Load Monitoring Division [10]

1.2 Problem Statement

Most NIALM method requires high frequency sampling data for load signature tracing before load disaggregation can be made. It also requires a recorder with extensive memory space to store such detail and huge data, which can only be done by specialized high specification energy meter. In practice, most commercial buildings are only equipped with a standard energy meter, which is only capable in capturing hourly kWh data. Applying NIALM method for such data to disaggregate load is impossible. A load disaggregation method for limited data i.e. low frequency sampling kWh data is needed. The method must able to estimate the usage pattern of major electrical equipment in the building that reflects the electricity usage behaviour of the occupants.

1.3 Objective of the research

The objectives of this research are:

- i. To study the relationship between recorded energy consumption (kWh) pattern of a building and the occupant's energy usage behaviour
- ii. To propose a load disaggregation method that reflects occupant's energy usage behaviour through bottom-up based utilization factor
- iii. To validate the proposed method against actual data

1.4 Scope of the research

This thesis proposes Non-Intrusive Appliance Load Monitoring to disaggregate overall load profile into specific appliance energy usage. Due to the requirement of high specification device and extensive amount of data storage, this study makes use of bottom-up concept to evaluate the relationship between load profile pattern and occupant utilization behaviour. For this purpose, block P19a, Faculty of Electrical Engineering, University Teknologi Malaysia (UTM) is used as a test system for this research. The test system is an academic building which consists of administration offices, lecturer rooms, class rooms and laboratories. The developed method is tested against practical data of the test system to validate the results. The experimental analysis focuses on estimating the energy usage pattern of two main equipment that influence the high energy consumption of P19a building, which is lighting and air-conditioning system. Other electrical equipment are neglected due to their very low power rating and their presence in total energy consumption is too small compared to lighting and air-conditioning system.

1.5 Significance of the research

Standard energy meter commonly does not provide individual energy data of the equipment installed in a building. Thus, the prospect of estimating energy consumption of various appliance is definitely appealing. Load disaggregation explored in NIALM requires high frequency data and high specification devices. Therefore, a simplified way of load disaggregation method is introduced in this study by correlating the relationship between load profile pattern and occupant utilization behaviour. The proposed method only requires historical data per hour which is accessible through a standard energy meter and simple calculations are involved so that it can be practically applied by the general consumers. The relevance of this study is to offer historical usage information to consumers with more detail energy use comparisons, their energy usage pattern as well as historical data of regular energy consumption. The disaggregated load information provided by the proposed method can inform consumer of their individual appliance energy usage and identifying which appliances are consuming and contribute the most power from their total electricity bills as well as providing guidance for energy-saving behaviours.

1.6 Thesis outline

This thesis is separated into five chapters. Chapter 1 describes the project overview. This chapter consist of background of the research, problem statement, objectives, scope of research and the significance of the research. Chapter 2 presents the literature review that discusses the concept of non-intrusive appliance load monitoring where the idea to disaggregate electrical appliance data consumption is developed. Chapter 2 also reviews the method that was implemented in previous researches that are related to this study. Later in this chapter, the basic idea of bottom-up concept is briefly reviewed. Chapter 3 present the proposed method to be applied in this analysis. It explains the steps taken throughout the proposed method in this study to meet the research objectives. Chapter 4 presents the results and analyses obtain through the proposed method. The result is presented in simple tables, figures and charts as well as detailed clarification of the findings. In this chapter, the extended application of the proposed method is also presented. Chapter 5 is the concluding section of this thesis. The recommendations for future work are also presented in this chapter.

factor and power input of that building needs to be calculated and observed to fit the user behaviour and characteristic of the building.

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