

TASKS DISTRIBUTION IN DRIVER SCHEDULING USING DYNAMIC SET OF BANDWIDTH IN HARMONY SEARCH ALGORITHM WITH 2-OPT

ZATUL ALWANI BINTI SHAFFIEI

DOCTOR OF PHILOSOPHY

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A thesis submitted in fulfillment of the requirements for the degree of Doctor of Philosophy

Faculty of Information and Communication Technology

UNIVERSITI TEKNIKAL MALAYSIA MELAKA

2021

DECLARATION

I declare that this thesis entitled "Tasks Distribution in Driver Scheduling using Dynamic Set of Bandwidth in Harmony Search Algorithm with 2-Opt" is the result of my own research except as cited in the references. The thesis has not been accepted for any degree and is not concurrently submitted in candidature of any other degree.

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APPROVAL

I hereby declare that I have read this thesis and in my opinion this thesis is sufficient in terms of scope and quality for the award of Doctor of Philosophy.

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Supervisor Name	:	Assoc. Prof. Ts. Dr. Zuraida binti Abal Abas
Date	:	

DEDICATION

To my late mother, Tengku Rofiah binti Tengku Zainal

Always miss you, every day!

ABSTRACT

Scheduling is important when dealing with task distributions and time management. In most organisations, the scheduling process is still generated manually. It consumes a lot of time and energy; consequently, the generated schedule is not really efficient. One of the main issues in scheduling is unfair tasks distribution among drivers. A fair schedule is necessary since it determines the quality of service as well as staff or customer satisfaction. Basically, a fair schedule can be defined as a well-balanced distribution of tasks among machines or staff by satisfying most of their constraints and personal preferences. There are two types of constraint to be considered in scheduling, which are hard constraint and soft constraint. This research was focused on driver scheduling problem for university shuttle bus (DSPUSB). Based on previous research using one of metaheuristic algorithms known as harmony search (HS), the generated schedule was still not optimum and cannot be solved maximally as there were too much repetitions of task (shift and route) occurred among drivers. The existing techniques (HS and its variants) have issues in terms of searching strategy (exploration and exploitation), slow convergence rate and high computation time for solving the scheduling problems maximally or near to optimal one. Therefore, a tasks distribution in driver scheduling using dynamic set of bandwidth in harmony search algorithm with 2-opt (SBHS2-opt) was proposed in this research. In the standard HS, the value of bandwidth (BW) parameter was static, while in this research, a dynamic set of bandwidth (BW2) value was formed based on constraints (problem domain). The BW2 value was dynamically changed and determined based on the current solution (with heuristic concept) of each driver every week, whereas the 2-opt swapping, which is normally used in travelling salesman problem, was applied for route constraint based on specific rules. The SBHS2-opt has guided searching strategy using heuristic concept or known as informed search. Knowledge on the problem is needed to assist the searching process and to strengthen the exploitation. There were 33 experiments carried out with different numbers of driver, route and shift. The results produced by SBHS2-opt outperformed 31 experiments out of 33 experiments. Hence, it was clearly shown that these improvements were capable in strengthen the exploitation, increase convergence rate, low computation time and at the same time balance the tasks distribution among drivers. In addition, the statistical analysis using Wilcoxon Rank-Sum Test and Bonferroni-Holm Correction as well as Box-Whisker plotting demonstrated that the SBHS2-opt has a significant difference in most of the experiments and was more stable in searching the best solution compared to HS, improved HS, parameter adaptive HS and step function HS.

PEMBAHAGIAN TUGAS DALAM PENJADUALAN PEMANDU MENGGUNAKAN SET JALUR LEBAR DINAMIK DALAM ALGORITMA PENCARIAN HARMONI DENGAN 2-OPT

ABSTRAK

Penjadualan adalah penting apabila berurusan dengan pembahagian tugas dan pengurusan masa. Dalam kebanyakan organisasi, proses penjadualan masih dihasilkan jadual secara manual. Ianya memakan banyak masa dan tenaga; akibatnya, jadual yang dihasilkan kurang berkesan. Isu utama dalam penjadualan adalah ketidakadilan pembahagian tugas dalam kalangan pemandu. Jadual yang adil perlu kerana ia menentukan kualiti servis dan kepuasan pekerja atau pelanggan. Secara asasnya, jadual yang adil didefinisikan sebagai pembahagian tugas yang seimbang dalam kalangan pekerja atau mesin dengan memenuhi kebanyakan kekangan dan keutamaan peribadi. Terdapat dua jenis kekangan yang perlu dipertimbangkan dalam penjadualan iaitu; kekangan wajib dan kekangan harus. Kajian ini fokus kepada masalah penjadualan pemandu bagi bas ulang alik universiti (DSPUSB). Berdasarkan kajian terdahulu menggunakan salah satu algoritma metaheuristik iaitu pencarian harmoni (HS), jadual yang dihasilkan masih tidak optimum dan tidak dapat diselesaikan secara maksimal kerana terdapat banyak pengulangan tugas (syif dan laluan) yang berlaku dalam kalangan pemandu. Teknik-teknik sedia ada (HS dan variasinya) mempunyai isu-isu dari segi strategi pencarian (eksplorasi dan eksploitasi), kadar penumpuan yang perlahan dan masa komputasi yang tinggi untuk menyelesaikan masalah penjadualan secara maksimal atau dekat dengan optimum. Oleh itu, pembahagian tugas dalam penjadualan pemandu menggunakan set jalur lebar dinamik dalam algoritma pencarian harmoni dengan "2opt" (SBHS2-opt) telah dicadangkan dalam kajian ini. Dalam HS yang asas, nilai parameter jalur lebar (BW) adalah statik, manakala dalam kajian ini; satu set jalur lebar dinamik (BW2) dibentuk berdasarkan kekangan (masalah domain). Nilai BW2 berubah secara dinamik dan ditentukan berdasarkan penyelesaian semasa (dengan konsep heuristik) setiap pemandu pada setiap minggu manakala penukaran "2-opt" yang biasa digunakan dalam masalah perjalanan jurujual telah digunakan untuk kekangan laluan berdasarkan peraturan khusus. SBHS2-opt mempunyai strategi pencarian berpandu yang menggunakan konsep heuristik atau lebih dikenali sebagai pencarian maklum. Pengetahuan tentang masalah diperlukan untuk membantu proses pencarian dan untuk memperkukuhkan eksploitasi. Terdapat 33 eksperimen yang telah dijalankan dengan bilangan pemandu, laluan dan syif yang berbeza. Keputusan yang dihasilkan oleh SBHS2-opt telah mengatasi 31 eksperimen daripada 33 eksperimen. Oleh itu, ianya jelas menunjukkan bahawa penambahbaikan ini mampu mengurangkan, memperkukuhkan eksploitasi, meningkatkan kadar penumpuan, masa komputasi yang rendah dan sekaligus seimbangkan pembahagian tugas dalam kalangan pemandu. Sebagai tambahan, analisis statistik menggunakan kaedah "Wilcoxon Rank-Sum" dan pembetulan "Bonferroni-Holm"; serta plot "Box-Whisker" menunjukkan bahawa SBHS2-opt mempunyai perbezaan yang signifikan dalam kebanyakan eksperimen dan juga lebih stabil dalam mencari penyelesaian terbaik berbanding dengan HS, penambahbaikan HS dan penyesuaian parameter HS.

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TABLE OF CONTENTS

DE	CLAI	RATION	
AP	PROV	VAL	
DE	DICA	ATION	
ABS	STRA	ACT	i
ABS	STRA	AK	ii
AC	KNO	WLEDGEMENTS	iii
TA	BLE (OF CONTENTS	iv
LIS	T OF	TABLES	vii
LIS	T OF	FIGURES	xi
LIS	T OF	ABBREVIATIONS	xiii
LIS	T OF	PUBLICATIONS	xiv
СН	APTI	ER	
1.	INT	RODUCTION	1
	1.1	Introduction	1
	1.2	Research background	3
	1.3	Problem statement	5
	1.4	Research objectives	8
	1.5	Scope of research	9

1.5	Scope of research)
1.6	Research significance	9
1.7	Thesis outline	10
1.8	Summary	13

2.

LIT	FERATURE REVIEW			14
2.1	Introdu	roduction		
2.2	Bus dr	Bus driver scheduling task distribution description model		
	2.2.1	Manual v	versus automated scheduling	22
	2.2.2	Types of	bus service	24
	2.2.3	Common	constraints in bus driver scheduling	27
		2.2.3.1	Shift	28
		2.2.3.2	Route	32
		2.2.3.3	Total of working hours	34
	2.2.4	Pattern re	epresentation in scheduling	36
2.3	Metaheuristic in scheduling			39
	2.3.1	Genetic A	Algorithm	41
	2.3.2	Particle S	Swarm Optimization	44
	2.3.3	Ant Colo	ny Algorithm	48
	2.3.4	Simulate	d Annealing	51
	2.3.5	Tabu Sea	irch	53
	2.3.6	Harmony	V Search Algorithm	56
		2.3.6.1	Improved Harmony Search	60
		2.3.6.2	Global Best Harmony Search	61
		2.3.6.3	Self-Adaptive Global Best Harmony Search	62
		2.3.6.4	Self-Adaptive Harmony Search	63
		2.3.6.5	New Improved Harmony Search Algorithm for	64
			Continuous Optimization Problems	

		2.3.6.6	Research on Adjustment Strategy of PAR in	65
			Harmony Search Algorithm	
		2.3.6.7	Harmony Search Algorithm with Dynamic	66
			Control Parameters	
		2.3.6.8	An Improved Global-Best Harmony Search	67
			Algorithm	
		2.3.6.9	An Analysis of Selection Methods in Memory	68
			Consideration for Harmony Search	
		2.3.6.10	Parameter Adaptive Harmony Search (PAHS)	70
			Algorithm	
		2.3.6.11	Global Dynamic Harmony Search Algorithm	72
		2.3.6.12	Self-Adaptive Harmony PSO Search	73
		2.3.0.12	Algorithm and its Performance Analysis	15
		2.3.6.13	A Differential-based Harmony Search	74
		2.3.0.13	Algorithm	/ 7
		2.3.6.14	Harmony Search with step function of HMCR	75
		2.3.0.14	and fret spacing	15
		2.3.6.15	1 0	79
		2.3.0.13	Modified opposition based learning to improve	19
2.4	Haumia	tia in caba	HS exploration	00
2.4		tic in sched	•	80
	2.4.1		d Search (Heuristic Search) vs Uninformed Search	83
25	0	(Blind S	earch)	0.4
2.5	-	Swapping	1 1 4 1 11 4 11 41	84
2.6			lving task distribution	88
2.7			stification for HS adoption	90
	2.7.1	-	lysis in HS algorithm	91
2.0	2.7.2	-	lysis in problem domain	92
2.8	Summ	ary		93
DEC				05
			DOLOGY AND IMPLEMENTATION	95
3.1				95
3.2		-	g problem of university shuttle bus	96
3.3		-	and methodology	97
	3.3.1	Planning	-	99
		3.3.1.1	Hard constraints and soft constraints of	101
			DSPUSB	
	3.3.2	Analysis	-	102
	3.3.3	Design p		104
		3.3.3.1	Problem formulation and representation of	105
			DSPUSB	
		3.3.3.2	Mathematical modelling and objective function	107
			of DSPUSB	
		3.3.3.3	Tasks distribution in driver scheduling using	116
			dynamic set of bandwidth in Harmony Search	
			Algorithm with 2-opt Swapping	
		3.3.3.4	Dynamic set of Distance Bandwidth	118
		3.3.3.5	2-opt Swapping	121
	3.3.4	Impleme	entation phase	123
		-	-	

3.

			3.3.4.1	Steps and pseudo code of DSPUSB with SBHS2-opt	124
		3.3.5	Evaluati	on and testing phase	151
	3.4	Summa		on and testing phase	151
	5.1	0 unin			101
4.	RES	ULT AN	DISCU	USSION	153
	4.1	Introdu	iction		153
	4.2	Experi	ment setup)	154
	4.3	Experi	ments and	results	155
	4.4	Analys	is of result	t	175
		4.4.1	Stability	of the technique	176
		4.4.2	Solution	availability and constraints requirement	177
		4.4.3	Searchin	g strategy	178
		4.4.4	Speed in	searching the best solution	179
		4.4.5	Statistica	al analysis using Wilcoxon Rank-Sum Test and	180
			Bonferro	oni-Holm Correction	
		4.4.6	Statistica	al analysis using Box-Whisker Analysis	185
	4.5	Summa	ary		190
5.	CON	CLUSI	ON AND I	RECOMMENDATION	192
	5.1	Finding	gs of resea	rch objective	193
	5.2	Summa	ary of cont	ribution	195
	5.3	Limitat	tion of rese	earch	196
	5.4	Future	works		197
	FERE PEND	NCES IX			199

LIST OF TABLES

TAB	LE TITLE	PAGE
2.1	Number of buses assigned to each bus route	33
2.2	Number of worked days for one year (Belén et al., 2012)	35
2.3	2-day legal shift pattern for UKMMC dataset	37
2.4	3-day legal shift pattern for UKMMC dataset	37
2.5	Multi shift working scheme (Rocha et al., 2012)	38
2.6	Scheduling problems solved by GA	43
2.7	Scheduling problems solved by PSO	47
2.8	Scheduling problems solved by ACO	50
2.9	Scheduling problems solved by SA	52
2.10	Scheduling problems solved by TS	55
2.11	PAHS's variants (Kumar et al., 2014)	71
2.12	Basic comparison of Informed Search and Uninformed Search	84
3.1	Constraints for DSPUSB	102
3.2	Example of data analysis of DSPUSB with their similarities and difference	s 103
3.3	Patterns with combination of route and shift	106
3.4	The possible one-month valid patterns	107
3.5	Evaluation of soft constraints violation	110

3.6	Validation of hard constraint 1	112
3.7	Example of demand for route and shift	113
3.8	Validation of hard constraint 2	113
3.9	Validation of soft constraint 1	114
3.10	Validation of soft constraint 2	115
3.11	Notation of BW2 set	119
3.12	Notation of patterns assignment using BW2	120
3.13	Patterns assignment for Week 3 and Week 4 based on set of BW2	120
3.14	Notation for 2-opt Swapping	122
3.15	Parameter setting for SBHS2-opt	125
3.16	HM (feasible schedule) for one month	127
3.17	Soft constraints evaluation	130
3.18	Soft constraints evaluation (cont)	131
3.19	Before application of 2-opt swapping	133
3.20	After application of 2-opt swapping	133
3.21	Pattern evaluation using rand1 and HMCR	138
3.22	Pattern evaluation using rand1, rand2, rand3, rand4 and PAR, BW	140
3.23	Evaluation of each shift for all drivers for Week 1 and Week 2	143
3.24	Pattern adjustment using proposed Dynamic Bandwidth BW2 for Week 3	143
3.25	NCHV after pattern adjustment for Week 3	144
3.26	Evaluation of each shift for all drivers for Week 1, Week 2 and Week 3	144
3.27	Pattern adjustment using proposed Dynamic Bandwidth BW2 for Week 4	144
3.28	Objective function evaluation after pattern adjustment using BW2	145
4.1	Parameter setting for HS variants and proposed approach	154

viii

4.2	List of experiments	156
4.3	Result for Experiment 6 (4 routes, 3 shifts, 12 drivers)	157
4.4	The best and the worst objective function for Experiment 6 (4 routes,	157
	3 shifts, 12 drivers)	
4.5	Result for Experiment 11 (6 routes, 2 shifts, 12 drivers)	160
4.6	The best and the worst objective function for Experiment 11 (6 routes,	160
	2 shifts, 12 drivers)	
4.7	Result for Experiment 12 (6 routes, 2 shifts, 20 drivers)	162
4.8	The best and the worst objective function for Experiment 12 (6 routes,	162
	2 shifts, 20 drivers)	
4.9	Result for Experiment 16 (6 routes, 3 shifts, 20 drivers)	164
4.10	The best and the worst objective function for Experiment 16 (6 routes,	164
	3 shifts, 20 drivers)	
4.11	Result for Experiment 17 (6 routes, 3 shifts, 25 drivers)	166
4.12	The best and the worst objective function for Experiment 17 (6 routes,	166
	3 shifts, 25 drivers)	
4.13	Result for Experiment 25 (7 routes, 3 shifts, 25 drivers)	168
4.14	The best and the worst objective function for Experiment 25 (7 routes,	168
	3 shifts, 25 drivers)	
4.15	Result for Experiment 26 (7 routes, 3 shifts, 50 drivers)	170
4.16	The best and the worst objective function for Experiment 26 (7 routes,	170
	3 shifts, 50 drivers)	
4.17	Result for Experiment 28 (10 routes, 2 shifts, 20 drivers)	172

4.18	The best and the worst objective function for Experiment 28 (10 routes,	172
	2 shifts, 20 drivers)	
4.19	Result for Experiment 32 (10 routes, 3 shifts, 50 drivers)	174
4.20	The best and the worst objective function for Experiment 32 (10 routes,	174
	3 shifts, 50 drivers)	
4.21	Experiments with "0" with the lowest i^{th} to achieve "0"	179
4.22	Statistical test for 4 routes and 2 shifts	181
4.23	Statistical test for 4 routes and 3 shifts	181
4.24	Statistical test for 6 routes and 2 shifts	182
4.25	Statistical test for 6 routes and 3 shifts	182
4.26	Statistical test for 7 routes and 2 shifts	183
4.27	Statistical test for 7 routes and 3 shifts	183
4.28	Statistical test for 10 routes and 2 shifts	184
4.29	Statistical test for 10 routes and 3 shifts	184
4.30	Box-Whisker Analysis	187
5.1	Example of pattern formed	194

LIST OF FIGURES

FIGU	JRE TITLE	PAGE	
1.1	Research contribution in area of scheduling problems and	12	
	optimization algorithm (shaded with grey color)		
2.1	The transportation planning (Liu and Ong, 2004; Lourenço, 2005)	17	
2.2	Tabulation of example of tasks in a shift (De Leone et al., 2010)	29	
2.3	Tabulation of shift types (Musliu et al., 2001)	31	
2.4	Tabulation of sample for demand of drivers (Gartner et al., 2001)	31	
2.5	Tabulation of shift distribution among drivers (Belén et al., 2012)	32	
2.6	Driving, break and rest period (Goel et al., 2012)	35	
2.7	Example of solution using pattern (Hadwan et al., 2013)	37	
2.8	Flow chart of Genetic Algorithm (Dastanpour and Mahmood, 2013)	42	
2.9	Flowchart of PSO algorithm (Maurya, 2018)	46	
2.10	Flowchart of ACO algorithm (Xu et al., 2012)	49	
2.11	Flowchart of Simulated Annealing (Zhan et al., 2016)	51	
2.12	Flowchart of Tabu Search algorithm (Hao et al., 2017)	54	
2.13	Analogy between music improvisation and optimization (Ayvaz, 2009)	56	
2.14	Steps of Harmony Search Algorithm (Lee and Geem, 2005)	58	
2.15	(a) Metaheuristic searching strategy; (b) Heuristic searching strategy	81	
2.16	(a) Before 2-opt swapping; (b) After 2-opt swapping (Kuang, 2012)	85	

2.17	Example of crossover point (Kaya et al., 2011)	87	
3.1	Flow structure of entire process of the research design 98		
3.2	The procedure passes off in planning phase 10		
3.3	Brief procedure of implementation phase in proposed approach		
3.4	Case 1; Same pattern between potential group and affected group		
	in same week		
3.5	Case 2; Same pattern between potential group and affected group after	136	
	1 st checking		
3.6	Case 3; Same pattern between potential group and affected group after	137	
	2 nd checking		
3.7	Case 4; Different pattern between potential group and affected group	137	
	for all weeks		
3.8	Process of updating HM	146	
3.9	Flow chart of SBHS2-opt	147	
4.1	Line graph for Experiment 6	158	
4.2	Line graph for Experiment 11	161	
4.3	Line graph for Experiment 12	163	
4.4	Line graph for Experiment 16	165	
4.5	Line graph for Experiment 17	167	
4.6	Line graph for Experiment 25	169	
4.7	Line graph for Experiment 26	171	
4.8	Line graph for Experiment 28	173	
4.9	Line graph for Experiment 32	175	
4.10	Box-Whisker plotting of each experiment	189	

LIST OF ABBREVIATIONS

BW	:	Bandwidth
BW2	:	Dynamic Set of Bandwidth
CSAHS2-opt		Constrained Self-Adaptive Harmony Search Algorithm with
		2-opt Swapping
DSP	:	Driver Scheduling Problem
DSPUSB	:	Driver Scheduling Problem for University Shuttle Bus
GHS	:	Global Best Harmony Search
HM	:	Harmony Memory
HMCR	:	Harmony Memory Consideration Rate
HMS	:	Harmony Memory Size
HS	:	Harmony Search
IHS	:	Improved Harmony Search
PAHS	:	Parameter Adaptive Harmony Search
PAR	:	Pitch Adjustment Rate
SAHS	:	Self-Adaptive Harmony Search

xiii

LIST OF PUBLICATIONS

Proceeding or Conference

- Shaffiei, Z.A., Abas, Z.A., Fadzli, A., and Abdul, N., 2014. Optimization in Driver 's Scheduling for University. *International Symposium on Research in Innovation and Sustainability 2014 (ISoRIS'14) 15-16 October 2014, Malacca, Malaysia*, 2014 (October), pp.15–16.
- Abas, Z., Shaffiei, Z., Rahman, A.F.N.A., and Samad, A., 2014. Using Harmony Search for Optimising University Shuttle Bus Driver Scheduling for Better Operational Management. *International Conference on Innovative Trends in Multidisciplinary Academic Research* (ITMAR-2014), 1, pp.614–621.

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 2016. An Optimized Intelligent Automation for University Shuttle Bus Driver Scheduling Using Mutual Swapping and Harmony Search. *International Journal of Computer Science and Information Security*, 14 (8), pp.875–884.

CHAPTER 1

INTRODUCTION

1.1 Introduction

Scheduling is very important when dealing with task distributions and time management in any organizations. There are many fields that required scheduling such as transportation (bus, train and flight scheduling) (Limlawan et al., 2011; Yaghini et al., 2015; Constantino et al., 2017), medical field (nurse scheduling) (Glass and Knight, 2010; Ramli et al., 2016a), manufacturing (machine scheduling) (Zamli, 2014; Che et al., 2016), and education (course and examination scheduling) (Babaei et al., 2014; Larabi Marie-Sainte, 2015). Without a schedule, the task distribution among the staffs and machines will be chaotic.

Generally, scheduling is the process of arranging, controlling and assigning tasks or workloads to machines or staffs (crews) based on some constraints. Normally, in most scheduling problem, constraints are divided into two types which are; hard constraint and soft constraint (Shaffiei et al., 2016, 2018). Hard constraint is a constraint that is compulsory to be fulfilled in order to make sure a feasible schedule can be produced (Cheang et al., 2003; Pillay and Banzhaf, 2010; Ramli et al., 2013a; Yaghini et al., 2015). A feasible schedule is an initial schedule that can be applied to machines or staffs but it is not good enough since the soft constraint is not taking into account. Soft constraint is a constraint that is not necessary to be fulfilled, but, by minimizing the violation of this constraint, an optimum schedule can be generated (Al-Betar et al., 2012; Belén et al., 2012; Hadwan et al., 2013; Anwar et al., 2014).

In some papers, there is another type of constraint to be solved named, precedence or priority constraint. The precedence constraint can be defined as a dependent constraint; where the next task cannot be executed if the previous task which is included in precedence constraint still not been accomplished. This type of constraint typically illustrated in acyclic directed graph (Skutella and Uetz, 2005; Widmer et al., 2010; Brucker et al., 2011). Apart from that, the personal preference also been considered in solving scheduling problem. Personal preference is referring to a self-requirement or request, it is not compulsory to be satisfied, for examples, in a nurse scheduling, a nurse requests to off-duty during first weekend; in a train driver scheduling, a driver requests to work overtime on every Thursday. If these examples of preferences were fulfilled, it will give a high satisfaction to that nurse and driver, though, a schedule still can be used if their preferences were violated (Legrain et al., 2015; Muramudalige and Bandara, 2018; Pillay and Qu, 2018). Constraints or preferences might be varies among countries and institutions (Ahmad, 2015).

In fact, scheduling process is quite complex and complicated to be solved especially if it is done manually and involving the large number of machines, tasks or staffs. However, in most organizations, the scheduling process is still generated manually. It is consuming time, energy and resources; consequently, a generated schedule is not really efficient. One of the main issues in scheduling is fairness. Basically, fairness can be defined as a wellbalanced distribution of tasks among machines or staffs by satisfying most of their constraints and personal preferences. It is almost impossible to achieve a perfect schedule with a well-balanced distribution of tasks. A feasible schedule does not have to be fair enough since only hard constraints are fulfilled. Therefore, the soft constraints or personal preferences are preferred to be satisfied to guarantee a fair schedule. Fairness in a schedule is absolutely necessary since it determines the quality of service or performance and staffs' or customers' satisfaction.

1.2 Research background

The scheduling of drivers is needed especially for public transport such as express buses, taxi, flight and train. It is a worldwide problem. Driver scheduling problem (DSP) that can be classified into NP-hard problems (Lenstra and Rinnooy Kan, 1981; Li, 2005) is one of the complex problems in transportation field. Basically, it can be defined as the process of distributing tasks or duties to drivers based on vehicle schedule within particular period, typically a week or a month (Belén et al., 2012). In some papers, the driver scheduling also referred as a crew scheduling as in (Limlawan et al., 2011; Ulrich, 2015; Kasirzadeh et al., 2017). According to Belen et al. (2012), usually in driver scheduling process, there are two phases involved which are staffing phase; estimation of the number of drivers needed to cover the needs of working hours, and scheduling phase; development of calendars of work to cover the estimation of drivers acquired in the phase of the staffing (Belén et al., 2012).

The vehicle scheduling and driver scheduling are closely related. In planning the vehicle scheduling, the driver scheduling or also known as crew scheduling should be emphasized to carry out the duties, tasks or shift based on the vehicle schedule that has been provided. Some researchers solved both schedules using sequential approach which is solving the vehicle schedule first, followed by driver schedule (Oughalime et al., 2009). Based on vehicle schedule, for example, a flight has to return from Kuala Lumpur to Jakarta three times in a day (vehicle schedule). To carry out this task, a few pilots should be

assigned to the vehicle schedule (driver or crew schedule). Zhao in (Zhao, 2006) stated that, it is necessary to assign shift and duties to drivers, so that every bus (or any vehicle) has a driver at all times.

Normally, the number of driver assigned for vehicle schedule is based on passengers' demand. The route with higher demand or the route that is more popular such as in a main city will be assigned with larger number of drivers. This is to ensure that the vehicle schedule can be covered. For example, an express bus from Melaka to Kuala Lumpur needs 8 trips in a day, while from Melaka to Shah Alam needs 3 trips in a day. This is because of the route of Melaka to Kuala Lumpur has higher demand by passengers and more popular since Kuala Lumpur is a main city compared to the route of Melaka to Shah Alam. Therefore, the number of driver assigned to the route of Melaka to Kuala Lumpur is larger than the number of driver assigned to the route of Melaka to Shah Alam.

As a matter of facts, problems that always been focused by researchers in DSP are the drivers' wages (Chen and Li, 2010), number of working days, shift (working hours and overtime), route (multiple depot) and break time assignment (Ramli et al., 2013b). It must be noted that these problems are related to fairness issue. They are very crucial to be solved in an optimum way since it will affect the quality of service and customers' satisfaction. In worse situation, the drivers that working non-stop, overtime and more than fixed working hours can lead to fatigue and consequently can cause an accident as reported in (Goel et al., 2012; Bowden and Ragsdale, 2018). Based on some news reported, mostly, this case happened for truck and express bus that involves with the long journey.

Unlike in most universities, shuttle bus services are provided for students to facilitate students in attending the classes, lectures, co-curricular activities or any activities held in their campus. Thus, the driver scheduling is needed to ensure that the duties or tasks are