SCHEDULE OPTIMIZATION OF FREIGHT VEHICLE FLEET USING DATA ANALYTICS

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Degree of Master of Science in Computer Science

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University of Moratuwa

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Thesis submitted in partial fulfillment of the requirements for the degree Master of Science in Computer Science

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Signature of the supervisor: Date:

Abstract

Schedule Optimization of Freight Vehicle Fleet Using Data Analytics

Schedule optimization is a key decision process of fleet management. However, truck and driver scheduling in multi-plant goods distribution is a complex problem due to geographically distributed customer sites and plants, heterogeneity in trucks, driver behavior, varying traffic conditions, and constraints such as working and resting hours for drivers. Moreover, we need to satisfy conflicting objectives such as maximizing order coverage and minimizing of the overall costs. At present context, the scheduling process is typically handled by a fleet manager who is responsible for assigning both the trucks and drivers to meet the confirmed jobs/orders of a given day. Such scheduling usually happens on the evening of the day prior to the order delivery date. As an NP-complete problem, assigning most suitable pair of vehicle and driver while satisfying both company and customer becomes difficult in a situation where there is an increment of total number of orders. We propose an automated, heuristic-based truck and driver scheduling solution which comprises of a rule checker and a scheduler. Rule checker imposes constraints and conditions such as driver and truck availability, delivery time constraints, and operating and resting hours. A scheduler that applies simulated annealing is proposed to cover as many orders as possible while minimizing the overall cost. The utility of the proposed solution is tested using a workload derived from a real-world bulk-cement distribution company. The results show good coverage of orders where the coverage increased by more than 10% compared to manual scheduling while minimizing the total cost by 35%. Furthermore, the solution has flexibility to tolerate exceptions due to breakdowns, traffic congestion, and extreme weather conditions without a considerable impact on most of the already assigned pairs of vehicle and driver to orders.

Keywords: Heavy Goods Distribution, Multi-plant, Simulated Annealing, Truck and Driver Scheduling

Dedication

I am dedicating this thesis to:

My loving parents, Mr. & Mrs. Keerthisinghe who are my pillars of success

My beloved brother, Mr. Chinthaka M. Keerthisinghe who is my admirer

My dearest husband, Mr. Dinesh Madusanke who stands by me with light of hope and support

All the teachers and friend who encourage and support me.

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Table of Contents

Abstrac	rt	iii
Dedicat	ion	iv
Acknov	vledgement	v
List of l	Figures	viii
List of	Tables	ix
List of A	Abbreviations	X
1. IN	TRODUCTION	1
1.1.	Motivation	1
1.2.	Research problem	2
1.3.	Research objectives	3
1.4.	Outline	3
2. LIT	ΓERATURE REVIEW	5
2.1.	Vehicle Scheduling	6
2.2.	Driver/ Crew Scheduling	9
2.3.	Vehicle and Driver Scheduling	12
2.4.	Meta-heuristics	14
2.4.1.	. Simulated Annealing	14
2.4.2.	. Genetic Algorithm	16
2.5.	Summary	16
3. PR	OBLEM FORMULATION	18
3.1.	Characteristics of the problem	18
3.2.	Problem parameters	19
3.3.	Constraints and Conditions	21
3.4.	Optimization problem	23
4. SO	LUTION APPROACH	25
4.1.	Rule Checker	25
4.2.	Order Scheduler	26
5. PE	RFORMANCE ANALYSIS	29
5.1.	Descriptive Analysis of a Case: Real Bulk Cement Distribution	29

5.1	.1.	Trip Summary	. 31
5.1	.2.	Trip Distribution	. 33
5.1	.3.	Distance Profile	. 34
5.1	.4.	Vehicle Profile	. 37
5.2.	Wo	rkload Creation	. 39
5.3.	Res	sults	. 42
6. SU	JMM	ARY AND FUTURE WORK	. 52
6.1.	Cor	nclusions	. 52
6.2.	Res	search Limitations	. 53
6.3.	Fut	ure Work	. 54
Referer	ices		. 55

List of Figures

Figure 2-1: Simulated Annealing procedure
Figure 2-2: Genetic Algorithm procedure
Figure 3-1: Bulk cement delivery process
Figure 4-1: Visual representation of solution approach
Figure 4-2: Solution process
Figure 5-1: Geographical representation of plants and sites
Figure 5-2: Trip distribution by date
Figure 5-3: Trip distribution by the day of the week
Figure 5-4: Average trip distribution by the day of the week
Figure 5-5: Trip distribution by time zone
Figure 5-6: Distance travelled by the date of the month of June 2016
Figure 5-7: Truck vs. fuel mileage.
Figure 5-8: Truck vs. average operating hours
Figure 5-9: Impact of cooling rate in order coverage
Figure 5-10: Impact of cooling rate in cost per km
Figure 5-11: Order coverage and cost per km for Wednesday orders (Cooling rate
0.9)
Figure 5-12: Order coverage and cost per km for Sunday orders (Cooling rate 0.9). 45
Figure 5-13: Order coverage throughout the week
Figure 5-14: Cost per km throughout the week
Figure 5-15: Delay vs. number of affected orders
Figure 5-16: Impact of delayed orders against days

List of Tables

Table 2-1: Vehicle scheduling and routing problem classification	. 10
Table 3-1: Problem characteristics.	. 18
Table 3-2: Order related parameters.	. 20
Table 3-3: Truck related parameters.	. 20
Table 3-4: Driver related parameters.	. 21
Table 3-5: Solution related parameters.	. 21
Table 5-1: Dataset summary.	. 30
Table 5-2: Trip summary with outliers.	. 32
Table 5-3: Trip summary without outliers.	. 32
Table 5-4: Fuel mileage distribution.	. 33
Table 5-5: Vehicle profile	. 37
Table 5-6: Summary of the workload creation.	. 39
Table 5-7: Truck details.	. 40
Table 5-8: Driver details	.41
Table 5-9: Impact of cooling rate.	. 42
Table 5-10: Results for the whole week	. 46
Table 5-11: Impact to the orders due to random delays	. 48
Table 5-12: Impact of population size.	. 50
Table 5-13: Impact of crossover probability.	. 50
Table 5-14: Impact of mutation probability.	. 51
Table 5-15: Impact of number of iterations.	. 51

List of Abbreviations

ANN Artificial neural network

ANS Artificial Neural Systems

BCD Bulk Cement Delivery

DF Deficit Function

DSS Decision Support System

GA Genetic Algorithm

IBK Naive Bayes classifier

ILP Integer Linear Programming

J48 Decision tree

LP Linear Programming

ML Machine Learning

NB K nearest neighbor

PART Rule based algorithm

RMC Ready Mix Concrete

RO Relief Opportunity

SA Simulated Annealing

SMO Support vector machine