

COST-EFFECTIVE MODEL-BASED TEST CASE GENERATION AND
PRIORITIZATION FOR SOFTWARE PRODUCT LINE

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DEDICATION

To:

Ma Abah, the two main supporters that always concern on this journey

My husband, the secret of my happiness

*The kids, Umar Aisy and Aisyah Aira, you are my sun, my moon and my
thousand stars.*

Thanks for all your patience and understanding

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ABSTRACT

In Software Product Line (SPL), testing is used to manage core assets that comprised variability and commonality in effective ways due to large sizes of products that continue to be developed. SPL testing requires a technique that is capable to manage SPL core assets. Model-based Testing (MBT) is a promising technique that offers automation and reusability in test cases generation. However, there are difficulties to ensure testing in MBT can achieve good test cases generation results based on cost (size of test suite, total execution time) and effectiveness (coverage criteria, fault detection rate) measures. This is due to lack of trade-off between cost and effectiveness in test cases generated in MBT for SPL. This study aims to increase quality of test cases based on cost and effectiveness by using generation and prioritization approaches for MBT in SPL. This study focuses on three parts to enhance quality of test cases. First, test model development based on traceability link. In order to improve test cases quality, this study focused on implementation of hybrid-based and hyper-heuristic based techniques to generate test cases. This is followed by Test Cases Prioritization (TCP) technique that is based on dissimilarity-based technique with string distance. These test cases generation and prioritization approaches are evaluated by using two benchmarks - one test object and one real object. The results are compared with other prominent approaches. The mapping approach showed 10.27% and 32.39% f-measure improvement against existing approach on e-shop object, respectively. For test cases generation using hybrid-based approach, the proposed approach outperformed existing approaches with 11.66% coverage, 17.78% average execution time, and 45.98% average size of test suite on vending machine object. The hyper-heuristic based approach NSGA-II-LHH outperformed other proposed low-level heuristic approaches with 12.00% improvement on coverage, 46.66% average execution time and 42.54% average size of test suite. Furthermore, evaluation of TCP approaches showed fault detection improvement of 21.60%, 10.40% and 12.20% and total execution time improvement of 48.00%, 22.70% and 31.80% in comparison with three existing approaches. The results revealed that proposed model transformations, test cases generation and prioritization approaches significantly improve cost and effectiveness measure in MBT for SPL.

ABSTRAK

Dalam Barisan Keluaran Perisian (SPL), ujian digunakan untuk menguruskan aset teras yang dibahagikan kepada kepelbagaian dan persamaan. Ujian Berasaskan Model (MBT) ialah teknik yang memungkinkan automasi dan guna semula kes ujian. Walau bagaimanapun, terdapat dua masalah yang menyebabkan MBT sukar mencapai penjanaan kes ujian yang bagus dari segi kos (saiz suit ujian, jumlah masa pelaksanaan) dan keberkesanan (kriteria liputan dan kadar kesalahan dikesan) iaitu masalah ketidakseimbangan antara kos dan keberkesanan kes ujian yang dijana oleh MBT. Objektif kajian ini ialah untuk meningkatkan kualiti kes ujian dari segi kos dan keberkesanan dengan menggunakan kaedah penjanaan dan keutamaan dalam MBT untuk SPL. Kajian ini memfokuskan kepada tiga bahagian dalam usaha untuk meningkatkan kualiti kes ujian. Pertama, pembangunan model ujian berasaskan pautan kebolehsesanan. Untuk memperbaiki kualiti kes ujian, kajian ini menumpukan kepada pelaksanaan berasaskan hibrid dan hiper-heuristik untuk menjana kes ujian. Setelah itu, Kes Ujian Keutamaan (TCP) yang berasaskan teknik ketidaksamaan dengan jarak rangkaian dijalankan. Ujian penjanaan dan keutamaan ini dinilai menggunakan dua penanda aras – sebuah objek ujian dan sebuah objek sebenar. Keputusan kajian dibandingkan dengan kaedah terdahulu. Teknik pemetaan menunjukkan peningkatan 10.27% dan 32.39% ukuran f untuk objek e-kedai. Untuk ujian penjanaan menggunakan kaedah berasaskan hibrid, kaedah yang dicadangkan menunjukkan 11.66% liputan, 17.78% masa purata pelaksanaan dan 45.98% saiz purata suit ujian untuk objek mesin layan diri. Ujian berasaskan hiper-heuristik NSGA-II mengatasi pendekatan lain dengan 12.00% liputan, 46.66% masa purata pelaksanaan dan 42.54% saiz purata suit ujian. Selain itu, penilaian terhadap kaedah TCP menunjukkan peningkatan kadar pengesanan kesalahan 21.60%, 10.40% dan 12.20% manakala jumlah masa pelaksanaan menurun sebanyak 48.00%, 22.70% dan 31.80% berbanding tiga pendekatan sedia ada. Hasil kajian mendapati kaedah transformasi model, kaedah penjanaan kes ujian dan kaedah keutamaan berjaya menurunkan kos serta meningkatkan keberkesanan dalam MBT untuk SPL.

TABLE OF CONTENTS

	TITLE	PAGE
	DECLARATION	iii
	DEDICATION	iv
	ACKNOWLEDGEMENT	v
	ABSTRACT	vi
	ABSTRAK	vii
	TABLE OF CONTENTS	viii
	LIST OF TABLES	xvii
	LIST OF FIGURES	xix
	LIST OF ABBREVIATIONS	xxiii
	LIST OF SYMBOLS	xxiv
	LIST OF APPENDICES	xxvi
CHAPTER 1	INTRODUCTION	27
	1.1 Overview	27
	1.2 Problem Background	29
	1.3 Problem Statement	33
	1.4 Research Question	36
	1.5 Research Goals and Objectives of the Study	37
	1.6 Research Scope	38
	1.7 Significance of the study	38
	1.8 Research Organization	40
CHAPTER 2	LITERATURE REVIEW	43
	2.1 Introduction	43
	2.2 Software Product Line Testing	43
	2.2.1 SPL Testing Process	44
	2.3 Model-based Testing in SPL	45
	2.3.1 Challenges in MBT for SPL	47

	2.3.1.1	Variabilities	48
	2.3.1.2	Reusability of Test Model	48
	2.3.1.3	Large Numbers of Products	49
	2.3.1.4	Cost and Effectiveness Trade-off for Test Cases	49
2.4		Traceability Dimension for Variability Management in SPL	50
	2.4.1	Requirement Traceability Directions	51
		2.4.1.1 Forward Traceability	52
		2.4.1.2 Backward Traceability	52
		2.4.1.3 Bidirectional Traceability	52
	2.4.2	Traceability Link Techniques in MBT for SPL	53
		2.4.2.1 Metamodel	53
		2.4.2.2 Model Transformation	54
	2.4.3	Model Transformation Types for Traceability in SPL	55
		2.4.3.1 Model to Model Transformation	55
		2.4.3.2 Model to Code Transformation	56
		2.4.3.3 Model to Text Transformation	56
	2.4.4	Summary of Works on Mapping based on Traceability	57
2.5		Test Cases Generation in MBT for SPL	60
	2.5.1	Issues in Test Case Generation in SPL	61
		2.5.1.1 Model	61
		2.5.1.2 Redundant Test Cases	62
		2.5.1.3 Effectiveness of Testing	62
		2.5.1.4 Cost of Testing	63
	2.5.2	Types of Test Model	63
		2.5.2.1 UML Model	64
		2.5.2.2 Variability Model	64
		2.5.2.3 Mathematical Model	67
	2.5.3	MBT Test Cases Generation Technique in SPL	67

	2.5.3.1	Coverage-based	67
	2.5.3.2	Pairwise-based	68
	2.5.3.3	Search-based	69
	2.5.3.4	Adaptive Random	69
	2.5.3.5	Fault-based	70
	2.5.3.6	Hybrid-based	71
	2.5.4	Comparison of Test Cases Generation Works	72
2.6		Regression Testing in MBT for SPL	76
2.7		Test Cases Prioritization in MBT for SPL	76
	2.7.1	Goal of Test Cases Prioritization	78
	2.7.1.1	Increase Effectiveness (Fault Detection)	78
	2.7.1.2	Minimize Execution Time	78
	2.7.1.3	Types of Technique	79
	2.7.2	Test Cases Prioritization Approaches in SPL	79
	2.7.2.1	Weight-based Prioritization	80
	2.7.2.2	Similarity / Dissimilarity based Prioritization	80
	2.7.2.3	Delta-Oriented based Prioritization	81
	2.7.2.4	Statistical-based Prioritization	81
	2.7.2.5	Hybrid-based Prioritization	82
	2.7.3	String Distance for Test Cases Prioritization (TCP)	83
	2.7.3.1	Jaccard Distance	83
	2.7.3.2	Hamming Distance	84
	2.7.3.3	Dice Similarity Distance	85
	2.7.3.4	Jaro-Winkler Distance	85
	2.7.3.5	Summary of String Distance	87
	2.7.4	Types of Prioritization Algorithm in SPL	89
	2.7.4.1	Local Maximum Distance	89
	2.7.4.2	Global Maximum Distance	89
	2.7.4.3	All-Yes Configuration	90
	2.7.5	Comparison of Test Cases Prioritization Studies	91

2.8	Overall Discussion	94
2.9	Chapter Summary	96
CHAPTER 3	RESEARCH METHODOLOGY	99
3.1	Introduction	99
3.2	Research Process	99
3.3	Research Methodology Phase	102
3.3.1	Phase 1: Theoretical Foundation	102
3.3.2	Phase 2: Research Framework and Identification of Variables	102
3.3.3	Phase 3: Proposed Model Mapping Approach	103
3.3.4	Phase 4: Proposed Test Cases Generation Techniques	104
3.3.5	Phase 5: Proposed Test Cases Prioritization	105
3.3.6	Phase 6: Empirical Evaluation on Test Objects	106
3.3.7	Phase 7: Thesis Writing and Research Contribution	106
3.4	Research Framework	106
3.5	Test Cases Generation Algorithm	111
3.5.1	Floyd Warshall Algorithm	111
3.5.2	Branch and Bound Algorithm	112
3.5.3	Best First Search (BFS)	113
3.5.4	Strength Pareto Evolutionary Algorithm 2 (SPEA 2)	115
3.5.5	Particle Swarm Optimization (PSO)	116
3.6	Test Cases Prioritization Algorithm	117
3.6.1	Local Maximum Distance Algorithm	117
3.7	Dissimilarity Measure Based on String Distance	118
3.8	Experimental Setup	120
3.8.1	Test Objects	120
3.8.1.1	Benchmark Test Objects	120
3.8.1.2	Vending Machine	122
3.8.1.3	e-shop	123

3.8.1.4	Real Test Objects	124
3.8.2	Objectives Function	127
3.8.3	Parameter Settings	128
3.8.4	Method to Validate Approach	128
3.8.4.1	Model Consistency for Model Mapping Approach	129
3.8.4.2	Pareto Analysis	130
3.8.4.3	Wilcoxon Rank-Sum	131
3.8.4.4	Kruskal Wallis Test	132
3.8.4.5	Average Percentage Fault Detection (APFD)	132
3.9	Chapter Summary	133
CHAPTER 4	MODEL TO TEXT TRANSFORMATION MAPPING APPROACH	135
4.1	Introduction	135
4.2	Development Plan for Model Mapping Approach	135
4.2.1	Mapping Model Definition	136
4.2.2	The Model Mapping Experimental Plan	141
4.2.3	The Model Mapping Approach Experimental Design	143
4.2.3.1	Model to Text Transformation	144
4.2.3.2	.xml Document Types	144
4.2.3.3	.xml File Example	145
4.2.3.4	Traceability Rules	146
4.2.3.5	Traceability Generation	148
4.2.3.6	Traceability Queries	149
4.2.3.7	Traceability Views	150
4.3	Experimental Results	152
4.3.1	Analysis of Model Mapping Approach	152
4.3.2	Comparison of Proposed Approach with Existing Approaches	155
4.3.3	Percentage improvement	157
4.3.4	Overall Discussion	158

4.4	Connection of Proposed Approach with Other Objectives	160
4.5	Chapter Summary	160
CHAPTER 5	A HYBRID HEURISTIC ALGORITHMS FOR MBT TEST CASES GENERATION FOR SPL	163
5.1	Introduction	163
5.2	Motivation	163
5.3	The Overview of Proposed Generation Approach	165
5.4	The Proposed Hybrid Test Cases Generation Algorithm (FWA-BBA-BFS)	166
5.4.1	The definition and process flow of FWA-BBA-BFS	168
5.4.2	The hybrid FWA-BBA-BFS generation example	171
5.5	Experimental results	180
5.5.1	Analysis result of proposed algorithm FWA-BBA-BFS	180
5.5.1.1	Analysis Effectiveness (Coverage) Metric Measure	181
5.5.1.2	Analysis of Cost (Average Size of Test Suite) Measurement	182
5.5.1.3	Average Cost (Execution Time) of FWA-BBA-BFS	183
5.5.2	Discussion on Cost and Effectiveness Metrics Implemented on the Proposed Hybrid Algorithm (FWA-BBA-BFS)	183
5.6	Comparison between the Proposed and Existing Algorithms	185
5.6.1.1	Percentage improvement	187
5.7	Technical Analysis	189
5.7.1	Difference of proposed and existing algorithms	189
5.7.2	Connection of proposed technique with other objectives	192
5.8	Chapter Summary	192

CHAPTER 6	MULTI-OBJECTIVES TEST CASE GENERATION USING HYPER-HEURISTIC ALGORITHM	195
6.1	Introduction	195
6.2	Motivation	195
6.3	Hyper-Heuristic in Test Case Generation Approach	196
6.3.1	Selection and Acceptance Mechanism, ISR-MCF	199
6.3.1.1	ISR-MCF Selection Operator	200
6.3.1.2	ISR-MCF with Low-Level Heuristic	202
6.3.2	Domain Barrier – Define Objectives Functions	204
6.3.3	Low-Level Heuristic Algorithms	204
6.3.3.1	NSGA-II with Low-level Heuristic	205
6.3.3.2	SPEA 2 with Low-level Heuristic	207
6.3.3.3	PSO with Low-level Heuristic	209
6.3.3.4	Integration of Crossover and Mutation with Low-Level Heuristics Algorithm	211
6.3.3.5	Crossover and Mutation of Low-Level Heuristic	212
6.4	Experimental Results	213
6.4.1	Analysis Results	214
6.4.1.1	Analysis of Maximization Results	215
6.4.1.2	Analysis of Minimization Results	219
6.4.1.3	Discussion of Overall Proposed Algorithm Results	222
6.5	Comparison between the proposed and existing algorithm	226
6.5.1	Statistical Analysis of Proposed Hyper-Heuristic with Existing Hyper-Heuristic Approach	226
6.5.2	Difference between Proposed Hyper-Heuristic with Existing Hyper-Heuristic Approach	232
6.5.3	Connection between the Proposed Technique with other Objectives	235

6.6	Chapter Summary	236
CHAPTER 7	THE ENHANCEMENT OF PRIORITIZATION ALGORITHM AND STRING DISTANCE TECHNIQUE	237
7.1	Introduction	237
7.2	Motivation	237
7.3	The Proposed Prioritization Algorithm	238
7.3.1	Phase (I): Dissimilarity Measure: Dice with Jaro-Winkler Algorithm	239
7.3.2	Phase (II): Test Cases Prioritization: Enhance Local Maximum Distance Algorithm	244
7.4	Example of Measurement by using Enhance Dice-Jaro-Wrinkler Measure	246
7.5	Experimental Results	248
7.5.1.1	Overall Prioritization Results	248
7.6	Comparison between the Proposed and Existing Algorithms	252
7.6.1	Percentage Improvement of Average APFD and Execution Time	255
7.6.2	Difference between the Proposed with Existing Algorithm	257
7.7	Chapter Summary	258
CHAPTER 8	THE VALIDATION OF TEST CASES GENERATION AND PRIORITIZATION	259
8.1	Introduction	259
8.2	Technical Analysis	260
8.2.1	Statistical Test Setup	260
8.2.2	Normality Test	261
8.2.3	Kruskal-Wallis Assumptions	264
8.2.4	Kruskal Wallis Non-Parametric Test	265
8.2.4.1	Kruskal-Wallis Test Results	266
8.2.4.2	Mean Rank Kruskal-Wallis Post-Hoc Test Analysis	272
8.2.5	Discussion	276

8.3	Chapter Summary	278
CHAPTER 9	CONCLUSION AND FUTURE WORKS	281
9.1	Summary	281
9.2	Research Contribution	282
9.3	Threats to Validity	284
	9.3.1 Internal Validity	284
	9.3.2 Construct Validity	284
	9.3.3 External Validity	285
9.4	Future Works	285
REFERENCES		287
APPENDICES		302
LIST OF PUBLICATIONS		311

LIST OF TABLES

TABLE NO.	TITLE	PAGE
Table 2.1:	Summary of existing studies on traceability in MBT for SPL	58
Table 2.2:	Comparison of existing studies on MBT for SPL	74
Table 2.3:	Comparison of regression testing categories	77
Table 2.4:	Summary of String Distance	88
Table 2.5:	Summary of existing study for TCP in SPL	93
Table 3.1:	Summary of Test Objects Selected	121
Table 3.2:	List of Objective Functions	127
Table 3.3:	Summary of Parameters Setting for Test Cases Generation Algorithm	129
Table 4.1:	Types of traceability relationship defined	147
Table 4.2:	Description of dictionary link for synchronize synonym	148
Table 4.3:	Artefacts Summary of Proposed Mapping Approach	152
Table 4.4:	Details of mapping results	153
Table 4.5:	Average summary of comparison approaches	156
Table 5.1:	Example of vending machine object test cases generated	173
Table 5.2:	List of score values of vending machine objects	178
Table 5.3:	Sample of test cases measure based on tendency and score values of vending machine object	178
Table 5.4:	Result of FWA-BBA-BFS implementation	181
Table 5.5:	Total execution time of FWA-BBA-BFS	183
Table 5.6:	Comparison summary of existing and the proposed algorithm	191
Table 6.1:	List of mutation and crossover	211
Table 6.2:	Average results based on coverage and cost-effectiveness measurements	215
Table 6.3:	Summary of strength and weaknesses of proposed hyper-heuristic based technique	223

Table 6.4:	Comparison between Low-Level Heuristics Implemented	225
Table 6.5:	Fixed values for two-tailed hypothesis testing	228
Table 6.6:	Summary of two-tailed t-test for PSO-LLH vs Strickler et al., (2017)	229
Table 6.7:	Summary of two-tailed t-test for NSGA-II-LLH vs Strickler et al. (2017)	230
Table 6.8:	Summary of two-tailed t-test for SPEA 2-LLH vs Strickler et al. (2017)	230
Table 6.9:	Comparison between hybrid heuristic and hyper-heuristic based technique	235
Table 7.1:	Dissimilarity Distance Results based on Vending Machine Test Object	249
Table 7.2:	Results of Proposed Dice-Jaro-Wrinkler Distance on e-shop	251
Table 7.3:	Mutant version for test model	251
Table 7.4:	Summary for APFD of Original Version vs Mutant Version Vending Machine Object	252
Table 7.5:	Summary for APFD of Original Version vs Mutant Version e-shop Object	252
Table 7.6:	Summary Rate of Fault Detected	252
Table 8.1:	List of Comparison Types with Goal	262
Table 8.2:	Skewness and Kurtosis Distribution of Educational Robotic Object	263
Table 8.3:	Assumption details for Kruskal-Wallis Test based for Result Validation	252
Table 8.4:	Kruskal-Wallis Hypothesis Summary	265
Table 8.5:	Kruskal-Wallis Test Mean Ranks	271
Table 8.6:	Kruskal-Wallis with Post Hoc hypothesis	272
Table 8.7:	Kruskal Wallis with Post Hoc Analysis	274
Table 8.8:	Kruskal-Wallis with Holm Bonferroni Correction Results	275

LIST OF FIGURES

FIGURE NO.	TITLE	PAGE
Figure 1.1:	Overview of Challenges in MBT for SPL	29
Figure 1.2:	The recognized problems	35
Figure 2.1:	SPL Testing Process (Reis <i>et al.</i> 2006)	44
Figure 2.2:	W-Model of SPL (Lee <i>et al.</i> 2012)	45
Figure 2.3:	Overall MBT process adapted from (Utting 2010)	46
Figure 2.4:	Model based testing process in SPL	47
Figure 2.5:	Example of Feature Model from Simmonds, 2015	66
Figure 2.6:	Pseudocode Local Maximum Distance Algorithm by Henard <i>et al.</i> (2015)	90
Figure 2.7:	Global Maximum Distance Algorithm by Henard <i>et al.</i> (2015)	90
Figure 2.8:	Pseudocode for All-Yes Config Algorithm by Al-Hajjaji <i>et al.</i> (2016)	91
Figure 3.1:	Research Process Flowchart	101
Figure 3.2:	Adopted experimental process (Wohlin <i>et al.</i> 2012)	104
Figure 3.3:	Research Framework	110
Figure 3.4:	Basic pseudocode of matrix generated based on Floyd Warshall Algorithm	112
Figure 3.5:	Flowchart of Branch and Bound Algorithm	112
Figure 3.6:	Flowchart of traverse nodes in Best First Search	114
Figure 3.7:	Workflow of NSGA-II	115
Figure 3.8:	SPEA 2 flowchart process	116
Figure 3.9:	PSO flowchart process	117
Figure 3.10:	Flowchart for Local Maximum Distance Algorithm	118
Figure 3.11:	Vending Machine Feature Model	123
Figure 3.12:	Vending Machine State Machine	123
Figure 3.13:	E-Shop Feature Model	124

Figure 3.14: E-Shop State Machine	125
Figure 3.15: Educational Robotics Feature Model	126
Figure 3.16: Educational Robotics State Machine	126
Figure 4.1: Development plan process for mapping approach	137
Figure 4.2: Model mapping relationship	140
Figure 4.3: Model Mapping Design Approach	142
Figure 4.4: .xml file format example from statechart model of vending machine	145
Figure 4.5: Traceability Link Generator	148
Figure 4.6: Algorithm to query traceability relationship	151
Figure 4.7: Comparison of average f-measure of proposed mapping with existing approaches	157
Figure 4.8: Percentage improvement of proposed mapping approach against existing approaches	158
Figure 5.1: Overview of proposed approach for test cases generation in SPL	166
Figure 5.2: Flowchart of FWA-BBA-BFS process	167
Figure 5.3: FWA-BBA-BFS hybrid proposed algorithm	172
Figure 5.4: Initialization step based on matrix form	173
Figure 5.5: First step to traverse nodes of the proposed algorithm	174
Figure 5.6: Example of branch traverse of FWA-BBA-BFA for TP1	179
Figure 5.7: Average effectiveness (coverage) of FWA-BBA-BFS	182
Figure 5.8: Average size of test suites	182
Figure 5.9: Average coverage for proposed algorithm compare with existing studies	186
Figure 5.10: Average size of test suite of proposed algorithm compared with existing works	187
Figure 5.11: Average time execution of proposed algorithm with existing works	187
Figure 5.12: Percentage Improvement of Coverage Criteria between Proposed Hybrid-Based Approach vs Existing Approaches	188

Figure 5.13:	Percentage Improvement of Average Execution Time between Proposed Hybrid-Based Approach vs Existing Approaches	188
Figure 5.14:	Percentage Improvement of Average Size of Test Suites between Proposed Hybrid-Based Approach vs Existing Approaches	189
Figure 6.1:	Conceptual component of hyper-heuristic algorithm with selection and acceptance mechanism	197
Figure 6.2:	Flowchart overview of hyper-heuristic test case generation approach	198
Figure 6.3:	ISR-MCF flowchart details process	200
Figure 6.4:	Pseudocode of ISR-MCF Selection Operator	203
Figure 6.5:	NSGA-II-LLH details flowchart process	206
Figure 6.6:	SPEA 2-LLH details flowchart process	208
Figure 6.7:	PSO-LLH details flowchart process	210
Figure 6.8:	Pseudocode of crossover and mutation operators	213
Figure 6.9:	Pareto distribution of NSGA-II-LLH of vending machine	216
Figure 6.10:	Pareto distribution of NSGA-II-LLH of e-shop object	216
Figure 6.11:	Pareto distribution of PSO-LLH of vending machine object	217
Figure 6.12:	Pareto distribution of PSO-LLH of e-shop object	217
Figure 6.13:	Pareto distribution of SPEA 2-LLH of vending machine object	218
Figure 6.14:	Pareto distribution of SPEA 2 of e-shop object	218
Figure 6.15:	Pareto distribution of NSGA-II-LLH for size of test suite and execution time based on vending machine object	219
Figure 6.16:	Pareto distribution of NSGA-II-LLH for size of test suite and execution time based on e-shop machine object	220
Figure 6.17:	Pareto distribution of PSO-LLH for size of test suite and execution time based on vending machine object	220
Figure 6.18:	Pareto distribution of PSO-LLH for size of test suite and execution time based on e-shop object	221

Figure 6.19:	Pareto distribution of SPEA 2-LLH for size of test suite and execution time based on vending machine object	221
Figure 6.20:	Pareto distribution of SPEA 2-LLH for size of test suite and execution time based on e-shop object	222
Figure 6.21:	Percentage improvement and decaying of proposed algorithm against existing algorithm on vending machine object	231
Figure 6.22:	Percentage improvement and decaying of proposed algorithm against existing algorithm on e-shop object	231
Figure 7.1:	The Proposed Prioritization Algorithm	239
Figure 7.2:	The Proposed Enhance Dice-Jaro-Winkler Distance	243
Figure 7.3:	The proposed enhancement in Local Maximum Distance	245
Figure 7.4:	Pseudocode of Proposed Local Maximum Distance Algorithm	246
Figure 7.5:	Comparison Results of Proposed TCP Approach against Existing Approach based on average APFD	254
Figure 7.6:	Comparison Results of Proposed TCP approach against Existing Approach based on average execution time	255
Figure 7.7:	Average APFD based on Percentage Improvement of Proposed Prioritization Approach against Existing Approaches	256
Figure 7.8:	Average Execution Time on Percentage Improvement of Proposed Prioritization Approach against Existing Approaches	256
Figure 8.1:	Validation for test case generation and prioritization approaches	261
Figure 8.2:	Independent Samples Kruskal-Wallis Test Boxplot for Time Generation against Types of Algorithms	266
Figure 8.3:	Independent Samples Kruskal-Wallis Test Boxplot for Average Coverage against Types of Algorithms	267
Figure 8.4:	Independent Samples Kruskal-Wallis Test Boxplot for Average Size of Test Suite against Types of Algorithms	268
Figure 8.5:	Independent Samples Kruskal-Wallis Test Boxplot for Average Prioritization Time against Types of Algorithms	269
Figure 8.6:	Independent Samples Kruskal-Wallis Test Boxplot for Average APFD against Types of Algorithms	270

LIST OF ABBREVIATIONS

APFD	-	Average Percentage Faults Detected
BBA	-	Branch and Bound Algorithm
BFS	-	Best First Search
CIT	-	Combinatorial Interaction Testing
Dice-Jaro-Winkler	-	Dice-Jaro-Winkler
EA	-	Enterprise Architect
EMOA	-	Evolutionary Multi-Objectives Algorithms
FDC	-	Fault Detection Capability
FWA-BBA-BFS	-	Floyd Warshall Algorithm-Branch and Bound Algorithm- Best First Search
FM	-	Feature Model
GA	-	Genetic Algorithm
ISR	-	Improvement Selection Rules
ISR-MCF	-	Improvement Selection Rules-Modified Choice Function
LLH	-	Low-Level Heuristic
MBT	-	Model-Based Testing
NSGA-II	-	Non-Dominated Sorting Genetic Algorithm II
NSGA-II-LLH	-	Non-Dominated Sorting Genetic Algorithm II-Low-Level Heuristic
PEMOA	-	Preference-based + Evolutionary Multi-Objective Algorithm
PSO	-	Particle Swarm Optimization
PSO-LLH	-	Particle Swarm Optimization-Low-Level Heuristic
SPEA 2	-	Strength Pareto Evolutionary Algorithm 2
SPEA 2-LLH	-	Strength Pareto Evolutionary Algorithm 2-Low-Level Heuristic
SPL	-	Software Product Line
TCP	-	Test Case Prioritization
UML	-	Unified Modelling Language
XML	-	Extensible Mark-up Language

LIST OF SYMBOLS

$Access_m$	-	Accessibility Matrix
α	-	Best generation of individuals
C	-	Cost
C_P	-	Crossover Probability
β	-	Generation and previous generation comparison
\rightarrow	-	Implies
\in	-	Intersection
$JW(T_1, T_2)_j$	-	Jaro-Winkler between two test cases
$Jaro(T_1, T_2)_{d_{jw}}$	-	Jaro between two test cases
<i>Queue</i>	-	List of queue
\wedge	-	Lebesgue measure
G	-	Maximum Generation
M_P	-	Mutation Probability
Max_{PT}	-	Maximum Path
Min_{PT}	-	Minimum Path
$!$	-	Not Equal
x	-	Number of objectives
ts_n	-	Number of states covered
ts_t	-	Number of states in test model
$score_{+ve}, score_{-ve}$	-	Positive and Negative score
$Q_{next} \in TP_i$	-	Queue list of Test Cases
$r \in R_i$	-	Reference vector
$\{INF 0\}$	-	Set of irrelevant values
f	-	Set of features
w	-	Set of nodes
c	-	Set of Constraints Considerations
P_S	-	Set of Population Size
p	-	Set of Pareto Front
$[s, i], [s, j]$	-	Two states comparison

$\sum score$	-	Total score
tm	-	Test model
$\sum PM_{score}$	-	Total Permutation score
$\sum_{i=1}^n Max\ Score$	-	Total Maximum Permutation
$\sum_{i=1}^n Min\ Score$	-	Total Minimum Permutation
n_x	-	Total number of test cases in test suite
$\Delta time_h$	-	Total execution time
$TS(x)$	-	Total Coverage
δ	-	Time elapsed
w	-	Weight coefficient

LIST OF APPENDICES

APPENDIX	TITLE	PAGE
Appendix A	Coverage values for Hybrid-based Technique	302
Appendix B	Size of Test Suite values for Hybrid-based Technique	303
Appendix C	Coverage values for Hybrid-based Technique	304
Appendix D	Size of Test Suite values for Hybrid-based Technique	307
Appendix E	Total Execution time values for Hybrid-based Technique	308
Appendix F	Number of Faults detected from Test Cases Prioritization Approach	309
Appendix G	Total Execution time values from Test Cases Prioritization Approach	310

CHAPTER 1

INTRODUCTION

1.1 Overview

A Software Product Line (SPL) is one of the paradigms for systematic reuse that guides organization to develop products from core assets rather than develop products from scratch. There are two major activities in SPL that focus of core assets development in domain engineering and product development in application engineering. Development of core assets is based on identification of reusable assets. In order to develop reusable core assets, SPL must have ability to exploit commonality and manage variability. Due to the explosion in the number of products, SPL requires an exhaustive testing technique to manage products. SPL testing is aimed to minimize testing effort while at the same time produce effective testing results. One of the most promising techniques is Model-Based Testing (MBT), which offers systematic automation in test cases generation (Reuys *et al.* 2010). There are two main steps, which are to obtain requirements to be presented in the test model and derive test cases. It offers automated, rigorous and systematic testing early in the software life cycle stage (modelling stage).

Among the challenges of MBT in SPL is to have a test model development that consists of variability and commonality. Cooperation between variability model with test model helps to realize this goal, which requires realization of variability, thus test model can represent SPL core assets. Realization of variability and commonality is important to be well defined in test model because the model will be used for test cases generation. This makes SPL different compared with single system engineering, since SPL features variability property, which is important to be reflected in test model. Researchers addressed mapping linkage to trace between Feature Model (FM) with requirement model. The link between two models is also known as a traceability link. Approaches have been proposed that described the implementation of traceability link

in SPL scope (Cichos *et al.* 2011), (Oster 2012), (Wang *et al.* 2013), however, they do not provide any ways to generate the trace link. Many studies proposed traceability link based on model to code transformation (Shen *et al.* 2008), (Zheng *et al.* 2017) and metamodel (Czarnecki and Helsen 2006), (Rose *et al.* 2012), (Machado 2014), (Font *et al.* 2017). Nevertheless, there is lack of standard semantics definition for trace link and it is represented in poor expressibility of the link and causes the misconception of variability interpreted in the test model (Vale *et al.* 2017). One of the ways to overcome the presentation of mapping linkage is to have a clear prerequisite traceability process with proper guidelines. These guidelines can be used to represent model transformation in traceability between variability and commonality with requirement model.

The second difficulty is regarding the quality of test cases generated from test model artefacts. The quality of test cases covers two main aspects, which are the cost of testing and effectiveness of test cases. The multi-objectives technique for MBT in SPL testing is used to handle trade-off between cost and effectiveness measures in generation and optimization techniques (Henard *et al.* 2015), (Wang *et al.* 2015), (Abbas *et al.* 2016). A test case is classified as a good test case if trade-off can be balanced between cost and effectiveness. The single-objectives measure, for example, coverage criteria, can be used to represent effectiveness of test cases; however, lack of cost measure caused the testing cost, for example total execution time and size of test suite, to be ignored (Hemmati *et al.* 2010), (Devroey 2014).

In recent years, the multi-objectives technique has been proposed to cover multiple test cases quality measure in SPL. Effectiveness of testing in MBT for SPL is commonly measured by using coverage criteria. The third challenge is related with lack of studies that implemented a multi-objective criterion that proved effective in fault detection rate in test suites in MBT statechart for SPL. Effectiveness in fault detection can be discovered efficiently by using Test Cases Prioritization (TCP) Technique (Kazmi *et al.* 2017b). However, the lack of TCP implemented by using multi-objective caused the fault to be unable to be revealed earlier. Thus, it highlights the need for test cases generation and prioritization in order to balance a trade-off between cost and effectiveness measure. Studies by (Henard *et al.* 2013), (Arrieta *et*

al. 2016), (Egyed *et al.* 2016) defined optimization based on generation and prioritization problem by using Search-Based and Heuristic-Based Technique. However, the cost-effectiveness of test suite is still in early phase since there are arguments between the selection of efficient techniques that can give the best cost and effectiveness for SPL testing. It showed the importance of optimization problem of maximizing effectiveness and minimizing the cost for MBT in SPL testing. Summary of challenges in MBT for SPL is illustrated in Figure 1.1.

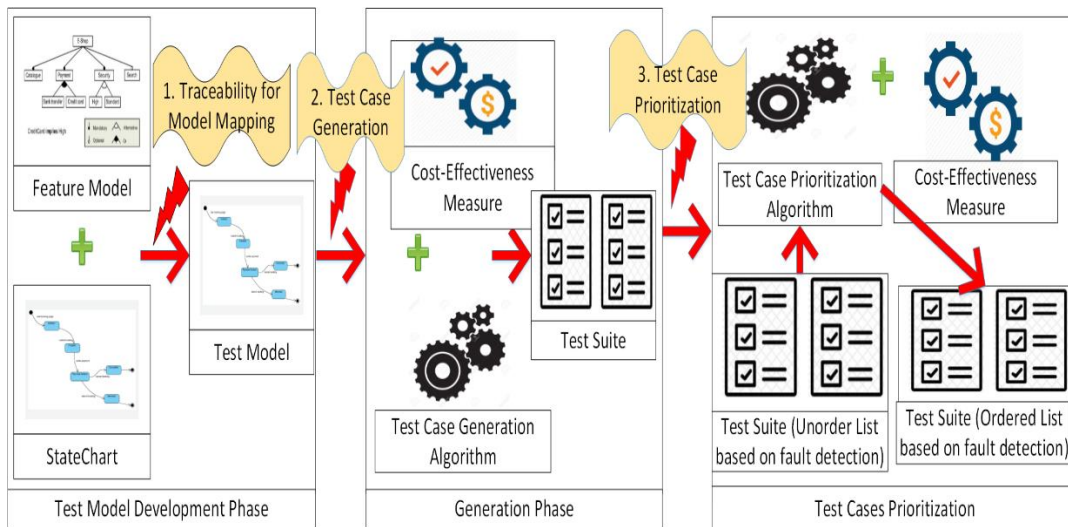


Figure 1.1: Overview of Challenges in MBT for SPL

1.2 Problem Background

In SPL testing context, the explosion in the number of possible products caused exhaustive testing to be infeasible. This issue gives challenge to select relevant subset of product for testing. A basic way to conduct testing for SPL is by using standard technique which are used in single system to be applied for SPL products. However, this takes higher cost and time consumption to evaluate every single product. Thus, there are techniques have been proposed previously to handle issues in SPL testing including MBT. The basic idea of MBT is to systematically minimize effort by exploiting knowledge of core assets. MBT for SPL is used to capture behavior of SPL.

MBT process starts with the test model development which is build based on requirement specifications. Then, test selection criteria will be defined to derive good test cases. A good test case is one that can detect faults earlier with higher effectiveness measure such as structural-based criteria (Utting 2006). In the scope of SPL testing, previous implementation of MBT faced issues of test model development (Ajila and Kaba 2004), (Czarnecki and Helsen 2006), (Heidenreich *et al.* 2008), (Shen *et al.* 2008), (Abbas *et al.* 2016), redundancy in test suites (Perrouin *et al.* 2010), (Wang *et al.* 2015), (Lackner 2017), and quality measure, which are effectiveness of test cases (Devroey 2014), (Machado 2014), (Lackner 2015), (Al-Hajjaji *et al.* 2017) and cost of testing (Hemmati *et al.* 2010), (Ensan *et al.* 2012), (Papadakis *et al.* 2016).

At first, the test model needs to be developed based on core assets reflected in requirement model. This development is related to the process of collection of core assets into the test model. Traceability link is commonly used to handle the different types of models. However, in the scope of SPL testing for derivation of test cases, it required a mapping between models in domain engineering because test cases should be generated in domain engineering before being considered ready to be reused in application engineering. In traceability link for MBT in SPL, approaches have been proposed related to the metamodel (Cavalcanti *et al.* 2011), (Font *et al.* 2017), (Steghöfer *et al.* 2019), model transformation (Czarnecki and Helsen 2006), (Heidenreich *et al.* 2008), (Rose *et al.* 2012), (Jessica *et al.* 2013), (Olsen and Oldevik 2013). Existing studies proposed model transformation approach to trace features in FM with statechart (Czarnecki and Helsen 2006), (Heidenreich *et al.* 2008). These model transformations proposed a process of conversion between FM into test model. However, it faced a problem with scalability due to manual mapping process that required tester to select the features to map with states. Furthermore, these approaches did not fulfill the needs of SPL because the dependency constraints of FM existed were ignored. This caused problem to maintain and evolve the large size of SPL core assets due to tester demand for good quality of test cases based on cost and effectiveness measures. Furthermore, there is a lack of comprehensive steps in development of traceability link. It is very important to have a clear view of traceability link in order to make sure the commonality and variability is undoubtedly defined in test model.

In MBT, redundancy of generated test cases also leads to a large size of test suite and increase in total execution time. This is related with the cost (size of test suite, total execution time) and effectiveness (coverage and fault detection) of test cases that can cause scalability issue. Studies by (Reuys *et al.* 2010), (Cichos *et al.* 2011) and (Lackner 2017) applied coverage-based measures, for example all-states and all-transition to generate test cases. The Ensan *et al.* (2012) discussed the importance of trade-off between different coverage types. However, according to (Ur *et al.* 2018), balance trade-off is not only represented in single measurement types, but needs to be measured between cost and effectiveness of test cases.

For example, once tester has achieved the maximum target to find defects, studies have been unable to ensure that these activities significantly reduce the testing effort and cost. Furthermore, as highlighted by (Inozemtseva and Holmes 2014), a good coverage test suite does not guarantee that the test suite is effective enough. This fact is acceptable in SPL context. The Devroey (2014) proposed a single effectiveness measure aimed to produce a good coverage-based generation. However, it faced scalability issue due to the longer execution time for test case generation. Most existing approaches in MBT for SPL focusing on test case measures were based on effectiveness (coverage). However, due to cost (size of test suite and total execution time), few cases have been evaluated (Cichos *et al.* 2011), (Arrieta *et al.* 2014), (Devroey *et al.* 2014), (Papadakis *et al.* 2016). This brings room for improvement in terms of trade-off between cost and effectiveness measure to improve the test case results. Furthermore, the implementation of Search-Based Technique (SBT) in MBT for SPL was designed to handle trade-off as optimization problem. SBT was based on heuristic technique that consists of cost and effectiveness measures to be transformed into objective functions. The goal of implementation of SBT in MBT for SPL is to generate test cases that balanced trade-off between cost and effectiveness measure.

In order to improve effectiveness (faults detection) in test cases, another technique, which is Test Case Prioritization (TCP), is applied to enhance the fault detection rate. For MBT in SPL, TCP is also used as a technique to improve effectiveness based on earlier fault detection. The lack of TCP implementation made

it difficult for faults to be revealed as soon as possible, thus making it possible for wrong products to be executed.

Machado (2014) improved the effectiveness of test cases by prioritizing fault in test model by using mutation testing. Existing studies proposed a similarity-based technique by using string distance method to prioritize test cases from FM and delta-oriented architecture model (Henard *et al.* 2016), (Al-Hajjaji *et al.* 2017), (Sahak 2018). Devroey *et al.* (2017) also proposed generation and prioritization for SPL. However, in terms of TCP, they have reused TCP algorithm proposed by (Henard *et al.* 2016) in test cases generated from mathematical model, for example Markov Chain model.

Moreover, in MBT for SPL, there is a lack of study that implemented TCP based on another test model, especially behavior model statechart. In terms of technique, similarity-based by using string distance to evaluate similarity of test cases is still in early phase. This is due to the lack of extensive view of efficiency of string distance implementation. Only a few string distances, which is Jaccard Distance and Jaro-Winkler hybrid with Hamming Distance, have been discovered. There is no justification on the evaluation of string distance applied in SPL. The string distance can help to reveal faults by measuring the similarity of test cases. The lack of implementation of string distance for similarity measure caused fault techniques to inefficiently detect faults.

In order to handle TCP, a technique was required to reorganize test cases sequence. There are three algorithms that have been previously discussed in similarity-based, which are Local Maximum Distance, Global Maximum Distance and all-yes config. However, there is a lack of evidence to show that the proposed algorithm is the best algorithm to discover faults as early as possible. Furthermore, the cost of TCP (total execution time) for similarity-based technique also required evaluation to make sure the proposed technique is capable to balance trade-off between cost and effectiveness measure. The limited multi-objectives in TCP approach caused difficulties to revealed faults. However, in terms of execution times, it is still important to be evaluated to ensure time can be minimized.

Due to these reasons, three key issues have been identified in this study to be resolved in order to produce a good quality of test cases with trade-off between cost and effectiveness measures. The first involves process of test model development with variability consideration based on FM. The lack of inadequacy of model mapping caused variability and commonality in SPL to be not completely reflected in test model artefacts. Another concern in this study is the demand of SPL testing that aims to obtain test cases with trade-off between cost and effectiveness measure. However, there is a lack of techniques concerned with multi-objective test case measure in SPL test case generation and prioritization. In terms of test case generation, a good technique that can balance trade-off is seen to be important way to enhance test case quality. This is because the existing studies validate test cases based on single measure without implementation or other validation that is also important to be discovered. There is a lack of previous studies that discover fault detection rate from test cases generated in test model artefacts. However, TCP is one of the important techniques to discover faults as early as possible compared with other existing techniques. Thus, test case generation and prioritization were the main factors to achieve trade-off between cost and effectiveness in MBT for SPL.

1.3 Problem Statement

Based on analysis from existing studies in MBT for SPL, there is a demand of test cases to fulfill a good quality measure in terms of cost and effectiveness. In order to enhance cost and effectiveness of test cases in MBT for SPL, there are three important parts that need to be highlighted, which are statechart test model, generation and prioritization of test cases.

First challenge is related to statechart test model development since it is the basic preparation to conduct SPL testing by using MBT. It requires a test model to represent SPL core assets based on requirements of products. The mapping between models is a challenge with respect to accurate generation of test cases. The implementation of single model, for example FM, can represent variabilities and commonalities in simplest ways. FM is based on taxonomic form commonly

represented as symbols. Mapping with other models such as requirement model can make FM information to be representable (Czarnecki and Antkiewicz 2005). The traceability link is required to map between two model components. In MBT, it is used to reflect SPL core assets in the test model for test case generation.

Model transformation approach is used to create traceability link between two models. However, the implementation of model transformation approach to conduct traceability mapping caused lack of comprehensive views of creating traceability in clear ways (Anquetil *et al.* 2012), (Vale *et al.* 2017). This is important since it involves variability of products. FM constraints need to be validated in order to prevent an invalid generation (Lochau *et al.* 2012).

The second challenge is related to test cases generation from statechart test model. It is associated with implementation of single measure criteria for test cases generation, for example, coverage criteria. It is used to discover coverage that can be covered by proposed approach based on test model artefacts. Nevertheless, it ignores the remaining validation measure, for example, the cost of testing is also important to be validated in test case generation. The single criteria is not able to represent it as a good test case (Inozemtseva and Holmes 2014), (Kazmi *et al.* 2017b). It required a trade-off between cost (size of test suite, total execution time) and effectiveness (coverage, fault detection capability) to ensure that the generated test cases gain a good quality measurement (Ahmed *et al.* 2020). Most existing approaches in MBT for SPL are concerned with a single objective at a time. This remains valid for certain cases; however, it does not reflect real-life problems in testing. However, trade-off issues can be improved by implementing multi-objective optimization for SPL testing (Henard *et al.* 2013), (Wang *et al.* 2014), (Abbas *et al.* 2016).

In order to enhance effectiveness of testing, earlier fault detection also need to be measured. This is because the test cases generated from MBT are not considered test cases based on fault detection. This has caused test cases in test suite to be listed randomly without any consideration of fault detection rate. TCP is the one of the techniques that can reveal faults earlier by reordering test cases based on fault detection rate. However, there is a lack of studies that measured faults in MBT for

SPL. This has caused faults to be not considered as validation measure to evaluate effectiveness of test cases. Furthermore, in SPL, the proposed TCP approach highlights the concern of test cases from FM and mathematical model (Markov Chain) (Egyed *et al.* 2016), (Henard *et al.* 2016), (Schaefer *et al.* 2016), (Al-Hajjaji *et al.* 2017). There is a lack of TCP approach used to evaluate test cases from UML statechart test model artefacts. This is due to the previous studies that highlighted concern on effectiveness measure of UML statechart test model artefacts by using coverage criteria, for example structural-based coverage. This has led to faults to not be considered an effectiveness measure for test cases generated from MBT in SPL.

The summary of problem to be highlighted in the study is summarized as per Figure 1.2.

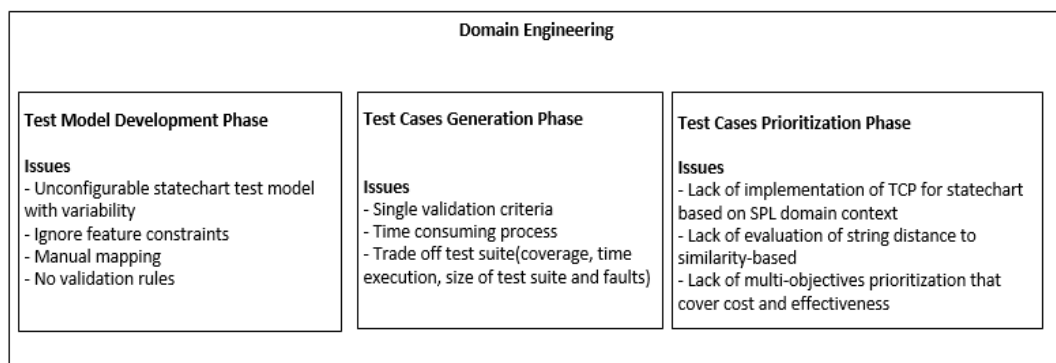


Figure 1.2: The recognized problems

To evaluate effectiveness measure based on earlier fault detection rate, similarity and dissimilarity based prioritization approach is shown as a good evaluation method to evaluate faults. It starts with measuring distance between test cases in the test suite. Then, the proposed similarity or dissimilarity-based approach will reorder test cases based on fault detection rate. In order to ensure test cases can be evaluated based on cost and effectiveness measure, it also requires a balance trade-off based on cost (total execution time) and effectiveness (fault detection rate) to make sure the TCP fulfills the software testing demand. Multi-objective optimization is required to ensure the trade-off issues in generation and prioritization approach can be improved. This is because multi-objective optimization offers different evaluation measure at a single

time. In addition, the technique can be tuned to ensure cost and effectiveness of test cases can be maximized and minimized.

1.4 Research Question

The aim of this research is to solve issues within the concept of MBT in SPL while targeting the accuracy of managing variability and enhancing the multi-objectives criteria in test cases generation and prioritization.

Derived from the research problem, the following are the formulated research questions:

“How to minimize cost (size of test suite, total execution time) and maximize effectiveness (coverage and faults) measures for generation and prioritization approach in MBT based on statechart test model for SPL”

To answer the main research questions, the following sub-question is addressed:

RQ1: How can the traceability mapping approach for MBT in SPL be improved?

RQ 1.1 How can the traceability mapping approach be reflected in explicit mapping between SPL statechart test model and test artefacts?

RQ 1.2: What are the techniques involved in explicit mapping between SPL statechart test model and test artefacts?

RQ2: How can the effectiveness of existing test case generation algorithms be increased?

RQ 2.1: Can the hybrid-heuristic optimization problem handle trade-off between cost and effectiveness in test case generation for MBT in SPL?

RQ 2.2: Does the hyper-heuristic optimization problem handle trade-off between cost and effectiveness in test case generation for MBT in SPL?

RQ3: How can earlier fault detection in test cases by using prioritization technique be discovered?

RQ3.1: How can the existing similarity-based prioritization technique for accelerating fault with minimal execution time be enhanced?

RQ 3.2: How can the existing similarity measurement by using string-based method for similarity-based prioritization be improved?

1.5 Research Goals and Objectives of the Study

The main goal of this research is to obtain increase in the quality of test cases based on cost (size of test suite, total execution time) and effectiveness (coverage, fault detection capability) by using generation and prioritization approach for MBT in SPL. This approach is able to ensure that the derivable test cases can minimize the cost and maximize effectiveness due to demand in SPL testing.

To achieve the mentioned goals, the objective of this study are as follows:

Objective 1: To propose a model transformation approach that can be used to provide explicit traceability mapping in MBT SPL.

Objective 2: To propose hybrid test cases generation technique with trade-off between cost and effectiveness measurement.

Objective 3: To propose a test case prioritization algorithm based on similarity-based with string distance measurement.

1.6 Research Scope

The proposed research is mainly focused on MBT for SPL with three main scopes, which are traceability in model and test case generation and prioritization. Details of the research scope are as follows:

i) Model-based testing for SPL

Model-based in this study covers the scope of testing that starts with test model development. It only covers representations of variability and requirements based on behavior in the scope of test model statechart. The implementation of other test model types are excluded from research.

ii) Traceability on MBT for SPL

The traceability link is based on the model transformation for SPL. Another type of traceability link, which is metamodel transformation, is excluded from the research since the focus is to map between two test models, which required obtaining all the related model components that can be used to derive test cases.

1.7 Significance of the study

SPL core asset management is an important element that needs to be considered before testing is conducted. It is found that constraints validation (Lochau *et al.* 2014) and tracing process (Rose *et al.* 2012) are important elements in order to accurately map test model component. Previous studies do not reflect comprehensive traceability process for SPL in clear way since it implemented a single traceability approach. Thus, the objective is to propose a model transformation approach that can be used to provide explicit traceability mapping in MBT SPL.

The second contribution of this study is to facilitate a good test case evaluation based on cost and effectiveness measures. This is due to the demand in SPL testing to

optimize the cost (size of test suite, execution time) (Wang *et al.* 2015), maximize based on coverage (Devroey *et al.* 2014) and fault detection ability (Al-Hajjaji *et al.* 2017). Due to the lack of multi-objectives optimization in MBT for SPL, this study proposed a test case generation approach with aim to enhance quality of test cases in terms of cost and effectiveness. There are two different techniques compared in the proposed approach, which are hybrid-heuristic and hyper-heuristic based for multi-objectives optimization. In terms of test cases generation, the proposed approach will cover the generation of test cases with maximum coverage and optimal cost (size of test suite and execution time). Another concern is due to the lack of trade-off between cost (time execution) and effectiveness (fault detection rate) in test cases from MBT for SPL. Thus, the enhancement of prioritization approach based on prioritization technique and string distance is used to overcome the problem of fault measurements in test cases from MBT. In this light, following points are significant for current research.

- I. This research will help to give a complete traceability process based on model transformation and formal method that is used to create traceability link between variability model and requirement model.
- II. This research helps to improve the process of mapping between models by using traceability based on automated query.
- III. This research provided the test case generation by using hybrid search-based testing with consideration of cost (size of test suite, total execution time) and effectiveness (coverage) measurements.
- IV. This research attempts to find the suitable test cases generation technique that can balance trade-off between cost and effectiveness measure.
- V. This research attempts to discover faults in test cases by using enhanced prioritization technique and string distance based on dissimilarity measure.

The cost (total execution time) and effectiveness (fault detection) are considered to evaluate the prioritization algorithm proposed.

1.8 Research Organization

The thesis is organized in nine chapters as follows:

Chapter 1: Introduction: It gives an overview on the structure of the research. It described the background of the study that brings the research problem to be discussed and explained. Main sections, which are research question and research objectives, are presented. The importance of this research is also elaborated in this chapter.

Chapter 2: Literature Review: It gives a discussion of the prior studies conducted in the field of MBT and SPL. All related studies are classified, explained and analyzed based on the problem defined in Chapter 1. The advantages and disadvantages of the existing study are also presented.

Chapter 3: Research Methodology: It provides the descriptions of the research framework, research process, case studies, techniques utilized in the current research. Details of each section are described in this chapter.

Chapter 4: The proposed FM_STATE Model Mapping. This chapter described the proposed model mapping approach based on the traceability link of FM and statechart. The design, implementation and results of the study are discussed in this chapter.

Chapter 5: A Hybrid Heuristic Algorithm for MBT Test Cases Generation in SPL. This chapter provides the proposed technique based on the three hybrid algorithms, which are Floyds Warshall Algorithm (FWA), Branch and Bound Algorithm (BBA) and Best First Search (BFS). The design, implementation, results, and existing study comparison are discussed in this chapter.

Chapter 6: Multi-objectives Test Case Generation by using Hyper-Heuristic Algorithm for MBT in SPL will describe the hyper-heuristic approach for MBT in SPL used to generate test cases. This chapter describes the design, implementation and experimental results of the proposed technique.

Chapter 7: Test Case Prioritization by using Hybrid Search Based Algorithm with Similarity Distance describe the TCP technique implemented for MBT statechart. This chapter includes design, implementation and experimental results. This experiment is related to the two proposed generation algorithms.

Chapter 8: The test cases generation from Chapter 5 and 6 with TCP from Chapter 7 are used to evaluate the performance of the proposed approach. The strength and weaknesses of the proposed techniques are also described.

Chapter 9: Conclusion and Future Work. The conclusion of the research is presented while highlighting findings, issues and future works.

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