

A FRAMEWORK FOR AUTOMATED QUALITY ASSESSMENT OF
SOFTWARE REQUIREMENT SPECIFICATION BASED ON PART-OF-SPEECH
TAGGING, MULTI-AGENT K-MEANS CLUSTERING AND CASE-BASED
REASONING

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DEDICATION

This thesis is dedicated to:

The sake of Allah, my Creator.

My great teacher and messenger, Mohammed (May Allah bless
and grant him), who taught us the purpose of life;

My great parents, who lead me through the valley of darkness with the light of hope
and support;

My beloved brothers and sisters;

To all my family, the symbol of love and giving;

My friends who encourage and support me;

All the people in my life who touch my heart;

I dedicate this research.



PTTA
PERPUSTAKAAN TUN AMINAH

My reverend mother

“Muna Abdulwaheed Abdulkareem”

Humanitarian, Compassionate, Religious, and a True Mother

“M” is for the million things she gives me,

“O” means only that she is growing old,

“T” is for the tears she sheds to save me

“H” is for her heart of purest gold,

“E” is for her eyes, with love-light shining,

“R” means right, and right she will always be,

Put them all together, they spell “MOTHER”,

A word that means the world to me.



PTTA UTHM
PERPUSTAKAAN TUNKU TUN AMINAH

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ABSTRACT

Software Requirement Specification (SRS) is an imperative process in a Software Engineering (SE) cycle, where its role is to document functional and non-functional requirements and to establish the tasks that a particular system is set to accomplish. Because a badly written SRS has an expensive impact on the entire project, the success or failure of any software product depends on the quality of the SRS document. Recent advancements in the field have explored automated extraction of quality attributes in SRS documents such as the Reconstructed ARM and the Rendex models. However, automating the quality assessment process poses major challenges, which requires advanced Natural Language Processing (NLP) algorithms to extract the quality features, interpreting the context of the features, formulating the assessment metrics, and documenting the shortcomings as well as possible improvements. Recent automated models also attempted to assess the quality of the SRS based on a small number of quality attributes and indicators due to the limitation in extracting quality attributes that require specific indicators from the SRS. To address this gap, this thesis proposes an Automated Quality Assessment of SRS (AQA-SRS) framework by integrating NLP for feature extraction, Multi-Agent System (MAS) with *K*-means for features clustering, and Case-based Reasoning (CBR) for process management. This framework assessed the SRS documents by automatically extracted 11 quality attributes and their corresponding 11 quality indicators through a deep analysis of the SRS textual content. This process is performed through the Multi-Agent *K*-means (MA-*K*-means) model for handling the automatic evaluation of the AQA-SRS framework. The performance of the AQA-SRS framework is evaluated by comparing the results against the state-of-the-art techniques as well as human experts based on two standard SRS datasets. The results showed the AQA-SRS framework reliably handled the assessment of 11 quality attributes and their corresponding 11 quality indicators with Krippendorff's Alpha 0.78 for the agreement with software engineering experts.

ABSTRAK

Spesifikasi Keperluan Perisian (SRS) merupakan satu proses penting di dalam kitaran Kejuruteraan Perisian (SE), yang mana peranannya adalah untuk mendokumentasi keperluan fungsian dan bukan fungsian serta membangunkan tugas bagi mencapai tujuan sesebuah sistem. Oleh kerana SRS yang tidak ditulis dengan baik boleh memberikan impak yang merugikan kepada keseluruhan projek, kejayaan atau kegagalan mana-mana perisian adalah bergantung kepada kualiti dokumentasi SRS. Kemajuan terkini bidang SE telah mengkaji pengekstrakan atribut kualiti daripada dokumen SRS secara automatik seperti model *Reconstructed ARM* dan *Rendex*. Walau bagaimanapun, proses mengautomatikan penilaian kualiti adalah sangat mencabar, serta memerlukan algoritma Pemprosesan Bahasa Tabii (NLP) termaju bagi mengekstrak ciri-ciri kualiti yang diperlukan, menterjemah konteks maksud ciri-ciri tersebut, memformulasi metrik penilaian, dan mendokumentasi sebarang kekurangan serta penambahbaikan. Model automasi terkini juga cuba untuk mengakses kualiti SRS berdasarkan ciri dan petunjuk kualiti yang sedikit disebabkan keterbatasan dalam proses pengekstrakan ciri-ciri yang memerlukan indikator spesifik sesebuah SRS. Bagi menangani kekurangan ini, tesis ini mencadangkan sebuah Kerangka Penilaian Kualiti SRS secara Automatik (AQA-SRS) dengan mengintegrasikan teknik NLP bagi pengekstrakan ciri-ciri, Sistem Multi-Agen (MAS) dengan *K-Means* bagi pengelompokan ciri-ciri, serta Penaakulan berasaskan Kes (CBR) bagi menguruskan proses ini. Kerangka ini menilai dokumen SRS dengan mengekstrak 11 atribut serta 11 indikator kualiti secara automatik melalui analisis bermakna ke atas kandungan teks. Proses ini dijalankan melalui model *Multi-Agent K-means* (MA-K-means) bagi mengurus penilaian automatik kerangka AQA-SRS tersebut. Prestasi kerangka AQA-SRS dinilai dengan membandingkan hasil dengan teknik terkini serta pakar bidang melalui dua dataset piawai SRS. Keputusan menunjukkan kerangka AQA-SRS mengurus penilaian 11 atribut dan 11 indikator kualiti secara dipercayai dengan keseragaman Krippendorff's Alpha 0.78 berbanding pakar Kejuruteraan Perisian.

TABLE OF CONTENTS

	TITLE	i
	DECLARATION	ii
	DEDICATION	iii
	ACKNOWLEDGEMENT	v
	ABSTRACT	vi
	ABSTRAK	vii
	TABLE OF CONTENTS	viii
	LIST OF TABLES	xii
	LIST OF FIGURES	xiii
	LIST OF APPENDIX	xvi
	LIST OF SYMBOLS AND ABBREVIATIONS	xvii
	LIST OF PUBLICATIONS	xviii
CHAPTER 1	INTRODUCTION	1
	1.1 Overview	1
	1.2 Problem Statement	5
	1.3 Research Objectives	6
	1.4 Research Scope	7
	1.5 Research Organization	7
CHAPTER 2	LITERATURE REVIEW	9
	2.1 Introduction	9
	2.2 Requirement Analysis	10
	2.3 Software Requirement Specification	11
	2.3.1 Purpose of SRS	12
	2.3.2 Structure of SRS	14
	2.3.3 Challenges of SRS	14
	2.4 SRS Quality Assessment	16
	2.4.1 SRS Quality Attributes	16

2.4.2	SRS Quality Indicators	17
2.4.3	Overlapping SRS Quality Attributes	20
2.4.4	SRS Quality Assessment Methodology	22
2.5	Automated SRS Quality Assessment Steps	31
2.5.1	Pre-processing	31
2.5.2	Processing	32
2.5.3	Post-Processing	33
2.5.4	Reporting	34
2.6	Previous Methods used in SRS Quality Assessment	35
2.6.1	Natural Language Processing	35
2.6.2	K-Means Clustering Algorithm	37
2.6.3.	Case-Based Reasoning	39
2.6.4	Multi-Agent Systems	42
2.7	Related Work	43
2.7.1.	Text Processing	43
2.7.1.	Quality Analysis	49
2.7.3.	Data Mining	51
2.7.4.	Discussion	53
2.8	Chapter Summary	57
CHAPTER 3	RESEARCH METHODOLOGY	58
3.1	Introduction	58
3.2	Research Methodology	59
3.3	Automated Quality Assessment of SRS Framework	60
3.3.1	Pre-Processing	62
3.3.2	Features Extraction	63
3.3.3	Multi-Agent K-means Clustering	66
3.3.4	Case-Based Reasoning	72
3.3.5	Report Generation	76
3.4	Dataset Description	77
3.4.1	Synthetic Clustering Dataset	77
3.4.2	Wine Dataset	77
3.4.3	Natural Language SRS Dataset	77
3.4.4	Reconstruction ARM dataset	79
3.5	Expert Validation	80

3.6	Evaluation Metrics	82
3.6.1	Pearson's Correlation Coefficient	82
3.6.2	Correlation Coefficient (R^2)	82
3.6.3	Purity	83
3.6.4	Normalized Mutual Information (NMI)	83
3.6.6	Adjusted Rand Index (ARI)	83
3.6.7	Fowlkes–Mallow index (FM)	84
3.6.8	Jaccard index	84
3.6.9	F-Measure	84
3.6.10	Krippendorff's Alpha	85
3.7	Example Scenario	85
3.8	Chapter Summary	89
CHAPTER 4	RESULT AND DISCUSSION	91
4.1	Introduction	91
4.2	Implementation of AQA-SRS Framework	92
4.3	Multi-Agent K -means Clustering Model	95
4.3.1	Determining Optimal Value of K	96
4.3.2	MA- K -Means Clustering Mahanobis Distance	97
4.3.3	MA- K -Means Clustering Quality Evaluation	101
4.4	Statistical Measurement for Extracted Features	105
4.4.1	Descriptive Statistics Analysis of SRS Documents	106
4.4.2	Statistical Analysis of SRS Quality Indicators	108
4.4.3	Correlation Coefficient and Regression Analysis for SRS Documents	109
4.5	Case Base Construction	113
4.6	Evaluation and Validation	119
4.6.1	Expert Validation	119
4.6.2	Comparative Evaluation	121
4.7	Chapter Summary	123
CHAPTER 5	CONCLUSION AND RECOMMENDATION	124
5.1	Introduction	124
5.2	Achievement of Research Objectives	125
5.3	Research Contribution	127
5.4	Recommendation and Future Work	129

REFERENCES	130
APPENDICES	145
VITA	173



LIST OF TABLES

2.1	Various Purposes of SRS	13
2.2	Summary of quality attributes	17
2.3	SRS quality indicators	19
2.4	Verification and validation process criteria	23
2.5	Analysis of related works	54
3.1	Indicators and features extracted from SRS	64
3.2	Automated Metric	66
3.3	Distance methods	69
3.4	Dataset description	78
3.5	Assessed SRS	81
3.6	List of requirements	86
3.7	Pre-processing Requirements	86
3.8	Post-processing requirements	87
3.9	Numeric features	88
3.10	Benchmark Datasets Labelling	89
3.11	Distance	89
4.1	Optimal value of K	97
4.2	Descriptive Statistics for the SRS Documents	107
4.3	Correlation Coefficients	110
4.4	Correlation coefficient	113
4.5	No. of Qa and Qi	122

LIST OF FIGURES

2.1	Changing Cost at Various SDLC Phases	12
2.2	Overlapping <i>Qa</i>	21
2.3	Verification and validation Process	22
2.4	Accumulative effect of errors and faults	25
2.5	Defect and cost in a software project	26
2.6	Cost to correct requirements defects	27
2.7	SRS assessment approaches	28
2.8	SRS assessment techniques	29
2.9	SRS assessment steps	31
2.10	NLP classification	36
2.11	Clustering example	37
2.12	CBR cycle	41
2.13	MAS	42
2.14	Forward reference detection approach	44
2.15	The reconstructed ARM tool steps	45
2.16	Measuring multiple sentences	46
2.17	Quality attributes measurement	46
2.18	Readability measurement	46
2.19	WSD framework	47
2.20	The automatic ambiguity detector framework	48
2.21	Flow diagram	49
2.22	Quantitative measurement framework	50
2.23	Rendex measurement model	50
2.24	Data mining approach	51
2.25	The noise detection approach	52
3.1	Research Methodology	59
3.2	Framework of AQA-SRS	61

3.3	Pre-processing steps	63
3.4	Feature extraction process	63
3.5	MA- K -Means Algorithm	68
3.6	MA- K -means clustering model	68
3.7	Real benchmarking dataset	70
3.8	Set of approximately 3 clusters	71
3.9	Value of differential inertia	72
3.10	CBR construction	73
3.11	Pseudocode of case-based construction	75
3.12	Case-based construction process	76
3.13	Length of the documents/pages	78
3.14	Document distribution terms of structure	79
3.15	Expert validation process	81
3.16	Extracted Features	87
3.17	Overall quality result of GAMMA-J Web Store	88
3.18	Process highlights for GAMMA SRS	89
4.1	AQA-SRS tool	92
4.2	AQA-SRS highlighting example	94
4.3	AQA-SRS assessment Report	95
4.4	Different results based Mahalanobis	97
4.5	Standard Deviation-Mahalanobis, $k = 12$	98
4.6	Elbow-Mahalanobis, $k = 7$	99
4.7	Silhouette-Mahalanobis, $k = 30$	100
4.8	Purity	102
4.9	NMI	103
4.10	ARI	103
4.11	Jaccard	104
4.12	FM	104
4.13	F-Measure	105
4.14	Variance of features (Q_i) in different SRSs	108
4.15	Visualization vagueness for PURE dataset	109
4.16	Regression analysis for PURE dataset	112
4.17	Case-based construction	114
4.18	Mismatching case	115

4.19	Matching case	116
4.20	No. of matching and mismatching	118
4.21	Processing time	119
4.22	No. of Qa and Qi	121



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LIST OF APPENDIX

A	Visualization to the Q^i from the tested PURE dataset	47
B	Tested SRS	50
C	Manual Assessment Results	65



LIST OF SYMBOLS AND ABBREVIATIONS

RE	-	Requirement Engineering
SRS	-	Software Requirement Specification
SQA	-	Software Quality Assurance
SDLC	-	Software Development Life Cycle
SATC	-	Software Assurance Technology Center
Q^a	-	Quality Attributes
Q^i	-	Quality Indicators
AI	-	Artificial Intelligence
NLP	-	Natural Language Process
MAS	-	Multi-Agent System
CBR	-	Case-Based reasoning
ARM	-	Automated Requirement Measurement
AQA-SRS	-	Automated Quality Assessment of SRS
FAQA-metrics	-	Fully Automated Quality Assessment metrics
MA- K -means	-	Multi Agent- K -means
IT	-	Inspection Technique
RT	-	Review Techniques
DBR	-	Defect-based reading
PBR	-	Perspective-based reading
UBR	-	Usage-based reading
CLBR	-	Checklist-based reading techniques

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- (i) **Jubair, M. A.**, Mostafa, S. A., Mustapha, A., & Hafit, H. (2018, August). A survey of multi-agent systems and case-based reasoning integration. In 2018 International Symposium on Agent, Multi-Agent Systems and Robotics (ISAMSR) (pp. 1-6). IEEE.
- (ii) **Jubair, M. A.**, Mostafa, S. A., Mustapha, A., Amana H. & Hassan, M. H. (2019). Fully automated quality assessment metrics for Software requirement specifications. AUS journal, 26(1), 188-197.
- (iii) **Jubair, M. A.**, Mostafa, S. A., Mustapha, A., Salamat, M. A., and Hassan M. H., (2020). Digging Deeper into Quality Assessment for Software Requirement Specifications. Journal of Critical Reviews, JCR. 2020; 7(12): 3869-3875.
- (iv) Mostafa, S. A., Gunasekaran, S. S., Khaleefah, S. H., Mustapha, A., **Jubair, M. A.**, & Hassan, M. H. (2019). A fuzzy case-based reasoning model for software requirements specifications quality assessment. International Journal on Advanced Science, Engineering and Information Technology, 9(6), 2134-2141.

CHAPTER 1

INTRODUCTION

1.1 Overview

Requirement Engineering (RE) is a process that involves a set of activities from collection, analysis, specification, to validation of user requirements in the form of natural language (Davis *et al.*, 2011). The implementation of RE is carried out throughout the early stage of the software development life cycle. In the process of RE, the most crucial factor is the Software Requirement Specification (SRS) documents, which is the main outcome of the process (Aurum & Wohlin, 2003). SRS is a set of requirements that describes the features and properties of the desired software product. It has numerous advantages to the software developers, it represents the functional and non-functional specification of the desired product, states the scope of the project, reduces the development effort, and removes any misunderstanding in the early stage (Wilson *et al.*, 1999).

In the SRSs document, all the expected capabilities and functionalities that must be present in a software system are explicitly stated. In addition to these, the document also spells out the limits of the system. While a requirement can be described as an objective that a system must meet, a specification is a description of how the objective must be met (Jani, 2010). One main challenge in preparing SRS documents is the complex writing structures to describe the requirements (Mostafa & Jani, 2011). A poor requirement of any product leads to failing the products because the quality of SRSs is the determinant of the quality of any kind of product. In addition, the

stakeholders' needs, as well as the limitations, are also reflected in the SRS (Wilson *et al.*, 1997).

Nonetheless, to document requirements, natural language is still the primary means. Requirements in natural language can be created and understood by all stakeholders without additional effort and specific requirements engineering background. However, natural language poses the risk of being imprecise. Poorly written requirements have an expensive impact on the whole project. Incomplete or ambiguous requirements generate additional effort due to unnecessary feedback loops. In the end, bad requirements lead to misinterpretations and finally to the wrong product (Femmer *et al.*, 2017).

Software Quality Assurance (SQA) is a set of processes that utilize to protect the quality of any software delivered by monitoring the processes of software engineering in different stages, which ultimately leads to, or at least gives confidence, high-quality software products (Parnas & Lawford, 2003). SQA expands on the entire software development cycle (e.g., SDLC), which rely on the design of software, coding, testing, and release management. The main focus of the SQA plan is to ensure a system or service aligns with the requirements defined in the SRS. Therefore, SRS quality evaluation is critical to identify the level of quality and faults in the very starting steps of the software development process (Thitisathienkul & Prompoon, 2015). The success of a project is strongly determined by the presence of a set of statements that clearly define the requirements of a system. In other words, the success of software projects is significantly influenced by requirements (Ali *et al.*, 2018).

The quality of SRS must be guaranteed to achieve a successful software project. Besides, there are no hard and fast rules associated with the production of the SRS document. However, there are given authorities that have suggested sets of information and details that must be possessed by a high-quality SRS. One of such authorities includes the Institute of Electrical and Electronics Engineers (IEEE) that has provided the recommended practice for SRS (IEEE-SA Standards Board, 2000). Since the quality of SRS determines the quality of a given product, it is crucial to have a robust requirement that leads to the achievement of quality and highly efficient software (Nordin *et al.*, 2017). For this purpose, Wilson *et al.* (1997) proposed a set of Quality Attributes (Q^a) and Quality Indicators (Q^i) that used to assess the quality of the SRS document. To achieve their goal, they defined a group of word/phrase that indicate the defect in the SRS document and distributed these features by indicators.

These Q^a and Q^i are used to ensure the SRS has precise requirements; the researchers proposed several techniques to detect the defect in SRS document based on the Q^a and Q^i such as the work of Jani and Islam (2012); Alshazly *et al.* (2014); Haque *et al.* (2019), and Femmer *et al.* (2018). These approaches can be categorized into three main categories, which are, automated, semi-automated, and non-automated or manual inspection. Manual inspection is considering a popular technique used to assess the quality of the SRS document. However, it is time-consuming for different reviewers to inspect and integrate the results of the review manually. Since the completion of a review cycle is achieved within several days or even weeks, the author of the requirements must wait a long time before receiving feedback. Additionally, the reviews of different individuals' consequent inconsistency (Saavedra *et al.*, 2013). The implication of these problems affects the assessment quality of the SRS and increases the cost.

To address these issues, different Artificial Intelligence (AI) and statistical methods have been used (i.e. case-based reasoning, natural language process, correlation coefficient, multi-agent system). For instance, in 2012, Jani & Islam, proposed semi-automated methods based on Case-based Reasoning (CBR) that combine the manual and automated methods to reduce the inspection time, cost, and workload. The technique has made a significant contribution in reducing the time, cost, and workload required for the process of assessment. However, there is still a need for improvement because humans (experts) still interact with their methods to assess the quality of the document (Rossanez & Carvalho, 2016). There are several AI and statistical methods that are used to measure the quality of the SRS document automatically (i.e. using fuzzy logic, neural network, genetic algorithm, and so on). These methods employ to propose an automated approach, such as the work of (Génova *et al.*, 2013; Bakar *et al.*, 2016; Thitisathienkul & Prompoon, 2014). However, all the works only considered a limited number of Q^a such as ambiguity, completeness and correctness. This leads to the need for a comprehensive model that deals with different methods or techniques like; pre-processing, features extraction, analysis and assessment to handle a wider range of Q^a (Rine & Fraga, 2015; Hisaszumi *et al.*, 2019; Li *et al.*, 2018; Stephen & Mit, 2017).

In general, some motivations show the need for automatic review, and these motivations are given as follows. First, an automatic review is fast in terms of different aspects of quality; instant feedback is one of the benefits of automatic reviews. For

instance, in the current configuration processes are received by the requirements scout, and then feedback for a paragraph is given about 500ms. This way, it is possible to obtain immediate feedback. Second, an automatic review is low in cost. The high cost of performing manual reviews remains the major problem associated with the process of manual reviews that come along with a comprehensive analysis. Thus, it becomes paramount to have a technique that can be used in obtaining feedback at a low cost; this is surely a promising advantage. Finally, an automatic review is consistent. With the manual method, two different reviewers can provide different results on different occasions for a requirement artifact. Even though this can be advantageous in terms of quality factors such as if the artifact satisfies a given guideline, this subjectivity paves the way for inconsistencies. An automatic review, on the other hand, is capable of providing consistent results for a requirement artifact on different occasions.

To overcome these problems, an automated technique has been proposed in this thesis that includes several statistical and AI methods to minimize the role of the human expert. However, there is no standard Q^a and Q^i that can assess the quality of the SRS in an automated manner. For this reason, this thesis focuses on three main points. Firstly, define a new group of Q^a and Q^i that can be assessed in an automated way. Secondly, construct a framework that is able to assess the quality of the SRS in an automated manner. Finally, this thesis proposes a model based on collaboration between Multi-Agent System and K -means clustering algorithm, that used to clustering the extracted feature based on the corresponding Q^a and Q^i .

This thesis attempts to mimic the steps of the reviewers when assessing the quality of the SRS document. First, the reviewers read the document, and they take a general idea about it. Then, they focus on finding the defect in the SRS document. Next, generate a report that shows the overall quality of the SRS. The proposed work also involves Natural Language Processing (NLP) techniques to understand the SRS document and extract the text from the documents. The AI and statistical methods are used to assess the SRS quality and generate an assessment report. This technique saves time, reduces the cost, workload and produces consistency assessment.

1.2 Problem Statement

Developers and stakeholders can understand requirements written in natural language. However, the usage of natural language in writing the Software Requirement Specification (SRS) along with human writing skills poses the risk of producing low quality and poorly written SRS. The low-quality SRS creates severe problems for the development process of software systems, which eventually causes additional costs and makes unnecessary processing loops (Thitisathienkul & Prompoon, 2015). Therefore, the quality of SRS documents is one of the critical factors for determining project success or failure (Wang *et al.*, 2013; Femmer *et al.*, 2017).

The popular approach of manual inspection of the SRS documents by multiple reviewers and then integrating the review results is a challenging task. One review cycle often takes days or weeks to be completed. Meanwhile, the author of the requirements has to wait a long time before receiving the assessment report. The result of these problems is that reviews are often only performed sporadically or only superficially (Antinyan & Staron, 2017). To tackle these challenges, many researchers proposed different automated models or tools to assess the quality of the SRS document. All existing models attempt to assess the quality of the SRS based on a limited number of quality attributes and indicators, this is due to several factors, including the assessment of each quality attribute requires extracting specific indicators or features and there is an overlapping of features between the quality attributes.

For instance, the work of Siahaan & Umami (2012) introduced an approach to detect the forward reference in the SRS document by using NLP. Carlson & Laplante (2014) reconstructed the Automated Requirement Measurement (ARM) model (Wilson *et al.*, 1997) to increase the quality of the SRS document by assessing three Q^a s which are ambiguity, complete, and understandability. Antinyan & Staron (2017) introduced the Rendex model to assess the understandability of the document. Nonetheless, these models are still lacking in terms of contextual features, which affect the process of constructing relationships between different features, the depth of the analysis, and the assessment process. Existing works neglect contents such as tables, text analysis, and more importantly relationships between the quality features.

In addition, quality indicators or features need to be grouped into clusters in order to make it possible to measure the corresponding Q^a and Q^i . The work of Mezghani *et al.* (2018b) used several NLP operators such as Part-of-Speech (POS) and noun chunking to extract features from requirements. It then uses a K -means clustering algorithm to categorize the features into groups of quality indicators. These indicators are used to assess redundancy and inconsistency among the requirements documents. In their work, the optimal value of k was found based on inertia and statistical gap.

However, the k -means algorithm finds locally optimal solutions concerning the clustering error. It is a fast-iterative algorithm that has been used in many clustering applications. It is a point-based clustering method that starts with the cluster centers initially placed at arbitrary positions and proceeds by moving at each step of the cluster centers to minimize the clustering error. The main disadvantage of the method lies in its sensitivity to the initial positions of the cluster centers. Therefore, to obtain near-optimal solutions using the k -means algorithm several runs must be scheduled differing in the initial positions of the cluster centers (Likas *et al.*, 2011). Also, handling a bigger number of quality indicators and attributes need an advanced algorithm that can draw the boundaries for the shape of clusters and avoid falling in local optimal.

1.3 Research Objectives

The main objective of this research is to propose a framework that utilizes to assess the quality of the SRS document in an automated manner. The following points will achieve the desired objective:

- To propose an Automated Quality Assessment of SRS (AQA-SRS) framework by integrating part-of speech-tagging for textual feature extraction, Multi-Agent and K -means for clustering, and Case-Based Reasoning (CBR) for assessment.
- To formulate a Multi-Agent K -means (MA- K -means) model for handling the measurements of the SRS quality in the AQA-SRS framework.
- To test and evaluate the performance of the AQA-SRS framework for clustering evaluation by implementing the PURE and Reconstruction ARM datasets, as well as the wine and synthetic datasets.

1.4 Research Scope

This work focuses on assessing the quality of the SRS document in an automated way to get a high-quality review and assessment to the SRS in lesser time, lower cost, lesser workload, and higher consistency of the review. The scope of this research concentrates on defining new metrics that comprise a group of quality attributes Q^a (i.e. complete, correct, unambiguity) and Q^i (like; imperative, continuances, directives) that can be assessed in an automated manner.

AQA-SRS comprises four main methods, which feature extraction using Natural Language Processing (NLP) approaches feature clustering using the Multi-Agent Systems (MAS) with K -means algorithm, and Case-based Reasoning (CBR) for process management and evaluation. The Java programming language is used to implement the proposed model. The performance of the AQA-SRS framework will be tested by using unlabeled standard datasets that consist of 79 SRS document and a Reconstruction ARM dataset that contain four pre-assessed SRS documents.

The SRS assessment datasets are prepared to provide an environment for testing particular problems, which are addressed in this work. The evaluation metric to the AQA-SRS framework performance focuses on assessing the reliability by measuring the agreement percentage of the proposed AQA-SRS framework against human experts. The classification of the requirement of functional and non-functional is out of the scope of this research. Additionally, quality attributes (Q^i) that require expert review or semantic analysis are also beyond the scope of this research.

1.5 Thesis Organization

This research presents the work of proposing an automated AQA-SRS framework that can assess the quality of the SRS document in an automated manner. This chapter presented the problem statement, objectives, and scope. The remainder of the chapters are organized as follows:

Chapter Two: This chapter presents a comprehensive literature review of this research. It first presents an introduction of the SRS document, a summary of the quality attributes and indicators, challenges, and vulnerabilities. Also, the chapter

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PTTA UTHM
PERPUSTAKAAN TUNKU TUN AMINAH