



**UNIVERSITI PUTRA MALAYSIA**

***SOFTWARE METRICS SELECTION MODEL FOR PREDICTING  
MAINTAINABILITY OF OBJECT-ORIENTED SOFTWARE USING  
GENETIC ALGORITHMS***

**ABUBAKAR DIWANI BAKAR**

**FSKTM 2016 8**



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GENETIC ALGORITHMS**

By

**ABUBAKAR DIWANI BAKAR**

**Thesis Submitted to the School of Graduate Studies, Universiti Putra Malaysia,  
in Fulfilment of the Requirements for the Degree of Doctor of Philosophy**

**March 2016**

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## DEDICATION

This thesis is dedicated to my ever caring parents my late father Diwani Bakar, my mother Amina Kassim Omar Al-bahsany, Abeida Ahmed Alawy Al-baalawy and my late uncle Said Ali Yussuf Al-baalawy, and to my progeny Abdulsalaam, Amina, Said Aisha, Ilhaam, Zuvena and Abeida.



Abstract of thesis presented to the Senate of Universiti Putra Malaysia in fulfilment of the requirement for the degree of Doctor of Philosophy

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**March 2016**

**Chairman : Abu Bakar Sultan, PhD**  
**Faculty : Computer Science and Information Technology**

Software development life cycle maintenance has been advocated as the critical part that consumes more time and resources. To understand the magnitude of the task to maintain the software product, software metrics have been used to make quantification based on their respective software features. To predict software maintainance, the proper metrics need to be selected to avoid the duplication or the outlying of the potential metrics. This is because on one hand, the individual metrics deals with only a single feature of the object-oriented systems, while on the other hand; the suites either contain duplicate metrics of the same goal or lack some important metrics that match the common attributes in the software products. The latest effort to solve this selection problem is the development of the metrics selection model that uses genetic algorithm (GA). However, the process failed to state clearly the encoding strategy in its initial stage.

This thesis clarifies the issue using the objective method to develop the GA metric selection model for predicting the maintainability of object-oriented systems. The study proposes the use of software metric thresholds in the classification process during the GA representation. The software metric thresholds were used as indication for identifying unsafe design in software engineering. To evaluate this technique, an experiment was conducted on two geospatial systems developed using Java programming language where the Chidamber and Kemerer (CK) metrics were used. The proposed technique was also compared to the ranking results from the experts. The comparison results obtained when compared with those of Principal Component Analysis and the complete software metric suite were very promising. Moreover, the three techniques show significant differences in both treatments when compared using analysis of variance.

Abstrak tesis yang dikemukakan kepada Senat Universiti Putra Malaysia sebagai memenuhi keperluan untuk ijazah Doktor Falsafah

**MODEL PEMILIHAN METRIK PERISIAN UNTUK MERAMAL  
KEBOLEHSELENGGARAAN PERISIAN BERORIENTASI-OBJEK  
MENGUNAKAN ALGORITMA GENETIK**

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Penyelenggaraan dalam kitaraan hayat pembangunan perisian telah dikenal pasti sebagai bahagian yang kritikal yang menggunakan lebih banyak masa dan sumber. Untuk memahami peri pentingnya tugas penyelenggaraan produk perisian, metrik perisian telah digunakan untuk membuat pengiraan berdasarkan ciri-ciri relatif perisian. Untuk meramalkan kebolehsenggaraan perisian, metrik yang sesuai perlulah dipilih untuk mengelakkan pertindihan atau mengelakkan metrik yang berpotensi daripada tersingkir. Perkara ini adalah kerana di satu pihak, setiap metrik terlibat dengan hanya satu ciri daripada sistem berorientasikan objek, manakala di pihak yang lain pula, kumpulan metriks tersebut sama ada mengandungi metrik pendua dengan matlamat yang sama ataupun ia kekurangan beberapa metrik penting yang sepadan dengan sifat-sifat umum bagi produk-produk perisian. Usaha terkini untuk menyelesaikan masalah pemilihan ini ialah pembangunan model pemilihan metrik yang menggunakan teknik algoritma genetic (GA). Walau bagaimanapun, teknik tersebut gagal menyatakan dengan jelas strategi pengekodan di peringkat awal proses tersebut.

Tesis ini memperkenalkan satu cara untuk menjelaskan isu ini menggunakan kaedah objektif untuk membangunkan model pilihan metrik GA untuk meramalkan penyelenggaraan sistem berorientasikan objek. Kajian ini mencadangkan penggunaan *threshold* metrik perisian dalam proses klasifikasi semasa perwakilan GA. *Threshold* metrik perisian digunakan sebagai sistem penggera untuk mengenal pasti reka bentuk yang tidak selamat dalam kejuruteraan perisian. Untuk menilai teknik ini, eksperimen telah dijalankan ke atas dua sistem geospatial yang telah dibangunkan dengan menggunakan bahasa pengaturcaraan Java, di mana metrik Chidamber dan Kemerer (CK) digunakan. Model ini juga telah dibandingkan dengan keputusan perarafan dari pakar-pakar. Keputusan yang diperolehi amat memberangsangkan apabila dibandingkan dengan Analisis Komponen Utama dan perisian kumpulan metrik metrik yang lengkap. Selain itu, ketiga-tiga teknik menunjukkan perbezaan yang ketara dalam kedua-dua rawatan apabila mereka dibanding menggunakan analisis varians.

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This thesis was submitted to the Senate of the Universiti Putra Malaysia and has been accepted as fulfilment of the requirement for the degree of Master of Science. The members of the Supervisory Committee were as follows:

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## LIST OF ABBREVIATIONS

AHF	Active hiding factor
ACO	Afferent Coupling
AIF	Attribute inheritance factor
ANOVA	Analysis of Variance
CBO	Coupling between object
CC	Cyclomatic complexity
CMC	Class method complexity
COF	Coupling factor
CK	Chidamber and Kemerer
CKJM	Chidamber and Kemerer Java metric
CTA	Coupling through abstract data type
CTM	Coupling through message passing
DIT	Depth of inheritance tree
EAs	Evolution Algorithms
EP	Evolution programming
ES	Evolution strategy
FN	False Negative
FP	False Positive
GA	Genetic algorithm
GIS	Geographic Information System
IEEE	Institute of Electrical and Electronics Engineering
ISO/IEC	International standard organisation/ International Electronic Commission
ISO/IEEE	International standard organisation/Institute of Electrical and Electronics Engineering
LCOM	Lack of cohesion metric
LDA	Linear Discriminant analysis
LOC	Lines of code
MEM	Maintainability estimation model
MHF	Method hiding factor
MIF	Method inheritance factor
MOOD	Model for object-oriented design
NAC	Number of ancestor classes



NCBC	Number of catch blocks per class
NDC	Number of descendent class
NLM	Number of attributes and methods number of local methods
NOC	Number of child
NOM	Number of methods
NOO	Number of Operator Overridden
NORM	Number of Overridden Methods
NPF	Number of Public Fields
NPM	Number of public method
NSM	Number of Static Methods
OGC	Open geospatial consortium
OOS	Object-oriented systems
PC	Principal Component
PCA	Principal component analysis
PF	Polymorphism factor
QMOOD	Quality model of object-oriented design
RFC	Reference for class
SAS	Statistical analysis software
TN	True Negative
TP	True Positive
WMC	Weight method per class

# CHAPTER 1

## INTRODUCTION

### 1.1 Background

This chapter explores the method of selecting software metrics to predict the degree of object-oriented software maintainability and related issues. The chapter begins with the background of the information concerning the maintenance phase for software, followed by the explanation on software measurements. The chapter also gives a list of contributions of this study, followed by an outline of the remaining chapters of the thesis. The research main argument here is that software metrics thresholds might be a useful categorisation strategy for evolutionary computation techniques.

As any other engineering measurements, software metrics have become necessary tools for developing quality software products through understanding different quality objectives. The importance of quantifying the quality attributes in software engineering has been identified by Galin (2004) as the baseline for planning and predicting the quality of software for its improvement. Galin points out that “software development methods and measurements are two important allies to ensure that the quality of the software product is fulfilled”. Fenton and Pfleeger (1997) go further to describe the best software developers and practitioners as those who use metrics to prove the quality of the software they design before releasing it for public use. Fenton and Pfleeger add that “this practice is used as a means of minimizing defects.”

In his recent book titled *Why the Program Fails*, Zeller (2009) insists on measuring the software products. He finds that there is no software that contains zero defects, and the results of these defects are the source of increasing software costs. Many writers, Douce and Layzell, 1997, for example, have claimed that “much of those costs occur during the maintenance phase of the software development life cycle. About 50% to 70% of time and resources are estimated to be used in the maintenance phase of software development process.” In solving the problems of cost, efforts, and time during software maintenance, it is according to Fenton and Neil, 1999, crucial to understand the extent of the handling of the software prior to its adoption. To Fenton and Neil, this can also help practitioners understand the quality of software products based on their quality attributes, which are used in quality modelling to identify the post-release fault proneness.

The nature of any software products, especially those developed using object-oriented concepts is that the code is in fact designed using more than one design attribute. For example, developers use different properties like inheritance, coupling, cohesion and other object-oriented features in the same software product. This approach is similar to the philosophy behind the development of the software

metrics, whereby one software metric measures only one attribute. Consequently, it implies that there is a duplication of some metrics, through which more than one metric can be used to measure the same attribute. Therefore, to predict one quality attribute requires several metrics. Then, they need to combine the magnitude, which is essential to predict the constituent software product features.

Software metrics suite was the first attempt to counter the problem of software metrics classification (software metrics selection). Boehm (1975) and Chidamber and Kemerer (1994) were the first researchers to introduce the so-called metric suites, and later followed by other researchers like Abreu and Melo (1996), Elish and Al-Khiaty (2013) and many more. The suites faced the problem of rigidity, which pushed practitioners either to reduce software metrics to remove redundant metrics or to add more metrics to accommodate the missing ones. Michura et al. (2013), for example, made an extension of CK metric suite to conform to his requirements.

In this case where there are dozens of metrics which can be used to measure one particular software quality attribute, there is a need to have in place mechanism that will help practitioners select software metrics that only suite measurement goals. Wang et al. (2011), suggest the elimination of the number of software metrics that are available for the particular measurement property. In their study, they reduced 98% of the metrics they collected for measuring the quality of software product. Their conclusion is that “the average of three software metrics is good enough to predict the quality of software. Thus, the selection of few suitable metrics out of dozens metrics is recommended to increase performance and to avoid redundant measurement. Therefore, the question that is raised here is how to obtain those few suitable software metrics for predicting the quality of a particular software product, out of dozens of metrics.

The evolution computational intelligence approach, as search-based strategies solves the problems associated with metrics suites. The approaches for selecting the best group of software metrics for predicting the quality of software proved to be promising. Vivanco (2010) uses genetic algorithms as a search-based strategy to solve the mentioned problems. His representation methods of encoding does not states clearly in his model while literature insists on using real data in encoding the chromosome. As by Baggen et al. (2012), insist the use of automatic and rea-world data can build the culture of trust for the final model that it has considered the quality decision on the development of that model.

This study used software metrics threshold (benchmark) ranking strategy in encoding the chromosome to propose the Genetic Algorithm software metrics selection model for predicting object-oriented software maintainability. The use of software metrics thresholds facilitates the performance and improves the accuracy of the model.

## 1.2 Problem Statement

The selection of the appropriate software metrics in predicting the quality of software is one of the crucial tasks in software engineering paradigm. The problem is more complicated in the availability of dozens of metrics and the inability of proposed groups of software metrics called suites. These collections of metrics failed to give clear and trusted software metrics classification goals for predicting the quality of software due to variability of attributes and features from one software product to another. As Altidor and his colleague in 2009 suggested that “the issue has forced some practitioners to make some modification to satisfy their needs” as Gray, 2008; Michura, 2013, argue. In that regard, there is therefore a need for the practitioners to propose models that can just select list of software metrics for particular software product.

Evolutionary computation, which is used in exhaustive selection problems, is the latest and hard effort employed in software metrics selection. Vivanco, (2010), uses genetic algorithms as a search-based strategy to solve the problem of rigidity identified in software metric suites. Although promising results were obtained, the representation method of encoding was not clearly stated.

The aim of this study is to propose the Genetic Algorithm (GA) software metrics selection model to predict object-oriented software maintainability. In this case, software metrics thresholds were proposed at the presentation phase of the GA. Because of the sensitivity of the availability of software metric thresholds and other challenges, study used the Chidamber and Kemmerer (CK) metrics suite as the representative for the object-oriented maintainability metrics. The results were validated and compared with real maintainability data ranked by experts in order to gain more confidence on proposed solution. The proposed model intends to help practitioners in selecting the most appropriate metrics according to the attributes in particular software, which will then facilitate faster and more accurate adoption of the maintainable software. Out of dozens of software, practitioners will be able to select only the appropriate list of metrics for particular software, which is then used for predicting the maintainability of that software.

## 1.3 Research Questions

To understand the problem explained above, the reseserch has been guided by the following questions:

1. Which software metrics are suitable to measure the object-oriented software maintainability based on their available thresholds in assumption that software product structure differ from one product to product?

To answer question (1) the review of related work has been conducted together the most appropriate list of software metrics that can be used to predict maintainability of particular object-oriented software products.

2. What are the thresholds for the software metrics identified in the first research question (1)?

Another two reviews were conducted to gather information on the most appropriate software metrics that can be used to predict the maintainability of particular object-oriented software products and to identify their thresholds (benchmark). Results were based on theoretical and experimental validation process in the related literatures.

#### **1.4 Research Objectives**

The objective of this research is to propose GA-based software metrics selection model to predict object-oriented software maintainability.

In achieving this main objective, the study proposes the following sub-objectives:

1. To standardise software metrics thresholds values for the six metrics in Chidamber and Kemmerer (CK) suite to determine the acceptance values.
2. To generate objective thresholds (benchmark) encoding strategy for the representation phase in the GA.
3. To implement GA-based software metrics selection model to predict maintainability of object-oriented software.

#### **1.5 Scope and Limitations of the Research**

This research proposed the software metrics selection model using evolutionary algorithm method. GA, which is one of the evolution strategies, is used to classify the software metrics in predicting maintainability of object-oriented software systems. The object-oriented software system is software technology, which involves many measurable features in software engineering. Geographical Information Software (GIS)-based software was used as case studies. Two geospatial software systems have been used to provide the researcher the chance to compare the software metrics selection performances within the same development context. This extends confidence prior to using software products that have been developed using another technology. Moreover, the geospatial software used was developed using open-source software technology. Apart from universal advantages of open-source technology, open access privillage of source code makes it easier to generate software metrics values.

## 1.6 Contributions of the Research

The use of software metrics thresholds (benchmark) in metrics selection model has the intuitional impacts in software engineering. Having used the software metrics thresholds, the study has successfully developed the software metrics selection model, which facilitates the selection of most effective software metrics in predicting maintainability to accomplish the adoption of the quality software. The use of the threshold prediction model seems to be a promising input in the software engineering paradigm. This encoding strategy has employed the practical use of empirical data that has also shown the great achievement in model performance. Therefore, this new introduced technique supports evolutionary algorithms representation encodings process. In addition, the use of GA as a search technique has put in the philosophy of traditional evolution from biology into the software metrics paradigm using software metrics thresholds.

In addition, upon knowing the maintainability efforts for particular softwares, software practitioners will probably be informed about technical and business aspects of the software product to be adopted. Practitioners who work as software maintainer will easily review the performance of the software product to ensure that the quality criteria have been met. For organizations, Baggen et al. (2012), indicate that “the maintainability models might guide the maintainers in organizing a better way of maintaining the resources”. As for the individual practitioners, they will probably make wiser decisions in adopting the suitable software product that is easily maintained based on available resources.

Moreover, the model can be used by the organisations to evaluate their existing software system to see whether it conforms to the dynamic quality standards from time to time due to error correction or incorporation of the new requirements.

## 1.7 Organization of the Thesis

The next chapter discusses the previous related literature on software maintenance, software metrics, software metrics threshold and evolutionary algorithms techniques. Chapter 3 discusses methodology used during the research phases. Chapter 4 discusses the preliminary results and provides a discussion on the software metrics thresholds encoding strategy. Chapter 5 presents the proposed models and their validation process. Chapter 6 presents the study findings and discussions. Finally, Chapter 7 concludes the research.

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