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**MODIFICATION OF S_1 STATISTIC WITH HODGES-LEHMANN
AS THE CENTRAL TENDENCY MEASURE**



**MASTER OF SCIENCE (STATISTICS)
UNIVERSITI UTARA MALAYSIA
2018**



Awang Had Salleh
Graduate School
of Arts And Sciences

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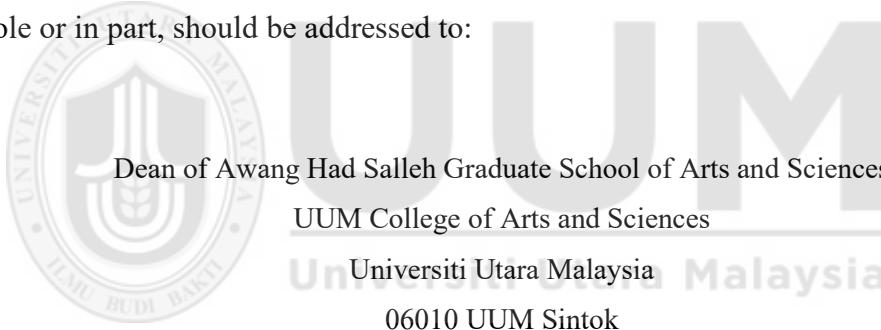
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Abstrak

Andaian kenormalan dan kehomogenan varians adalah merupakan perkara penting bagi prosedur parametrik seperti dalam pengujian kesamaan kecendurungan memusat. Sebarang ketidakpatuhan andaian tersebut boleh meningkatkan kadar Ralat Jenis I yang serius, yang akan mengakibatkan penolakan hipotesis nol yang tidak betul. Prosedur parametric seperti *ANOVA* dan ujian-*t* sangat bergantung pada andaian yang sukar ditemui dalam data sebenar. Sebaliknya, prosedur tak berparameter tidak bergantung pada taburan data tetapi prosedur tersebut kurang kuasanya. Untuk mengatasi isu yang dinyatakan, prosedur teguh adalah dicadangkan. Statistik S_1 adalah salah satu prosedur teguh yang menggunakan median sebagai parameter lokasi untuk menguji kesamaan kecenderungan memusat di antara kumpulan, dan ia membabitkan data asal tanpa perlu memangkas atau mentransformasi data untuk mencapai kenormalan. Kajian terdahulu terhadap S_1 menunjukkan kekurangan keteguhan dalam beberapa keadaan di bawah reka bentuk seimbang. Oleh itu, objektif kajian ini adalah menambahbaik statistik S_1 asal dengan menggantikan median kepada penganggar *Hodges-Lehmann*. Penggantian juga dilakukan terhadap penganggar skala menggunakan varians bagi penganggar *Hodges-Lehmann* serta beberapa penganggar skala teguh yang lain. Bagi memeriksa kekuatan dan kelemahaman prosedur yang dicadangkan dalam mengawal Ralat Jenis I, beberapa pemboleh seperti jenis taburan, bilangan kumpulan, saiz kumpulan yang seimbang dan tidak seimbang, varians yang sama dan tidak sama, dan sifat pasangan telah dimanipulasikan. Hasil kajian menunjukkan kesemua prosedur yang dicadangkan adalah teguh merentasi semua keadaan bagi setiap kes kumpulan. Selain itu, tiga prosedur yang dicadangkan iaitu $S_1(MAD_n)$, $S_1(T_n)$ dan $S_1(S_n)$ menunjuk prestasi yang lebih baik berbanding prosedur S_1 asal di bawah taburan pencong yang ekstrem. Secara keseluruhan, prosedur yang dicadangkan menunjukkan keupayaannya mengawal peningkatan Ralat Jenis I. Oleh yang demikian, objektif kajian ini telah tercapai apabila tiga daripada prosedur yang dicadangkan menunjukkan peningkatan keteguhan di bawah taburan terpencong.

Katakunci: Statistik S_1 , *Hodges-Lehmann*, penganggar skala teguh, ralat Jenis I, taburan terpesong

Abstract

Normality and variance homogeneity assumptions are usually the main concern of parametric procedures such as in testing the equality of central tendency measures. Violation of these assumptions can seriously inflate the Type I error rates, which will cause spurious rejection of null hypotheses. Parametric procedures such as ANOVA and *t*-test rely heavily on the assumptions which are hardly encountered in real data. Alternatively, nonparametric procedures do not rely on the distribution of the data, but the procedures are less powerful. In order to overcome the aforementioned issues, robust procedures are recommended. S_1 statistic is one of the robust procedures which uses median as the location parameter to test the equality of central tendency measures among groups, and it deals with the original data without having to trim or transform the data to attain normality. Previous works on S_1 showed lack of robustness in some of the conditions under balanced design. Hence, the objective of this study is to improve the original S_1 statistic by substituting median with Hodges-Lehmann estimator. The substitution was also done on the scale estimator using the variance of Hodges-Lehmann as well as several robust scale estimators. To examine the strengths and weaknesses of the proposed procedures, some variables like types of distributions, number of groups, balanced and unbalanced group sizes, equal and unequal variances, and the nature of pairings were manipulated. The findings show that all proposed procedures are robust across all conditions for every group case. Besides, three proposed procedures namely $S_1(MAD_n)$, $S_1(T_n)$ and $S_1(S_n)$ show better performance than the original S_1 procedure under extremely skewed distribution. Overall, the proposed procedures illustrate the ability in controlling the inflation of Type I error. Hence, the objective of this study has been achieved as the three proposed procedures show improvement in robustness under skewed distributions.

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Keywords: S_1 statistic, Hodges-Lehmann, robust scale estimators, Type I error, skewed distributions.

Acknowledgement

First of all, I would like to thank God for giving me the chance to complete the thesis which I have spent five years in studying Master of Sciences (Statistics) as a part time student. This is truly a blessing to me. Besides, I would like to extend my appreciation to my supervisor, Associate Professor Dr. Sharipah Soaad Syed Yahya and co-supervisor, Dr Aishah Ahad who have given continuous guidance, patience and support to me. They always be there for me whenever I face difficulty in my journey of writing my thesis and generating data using statistical computer software. I appreciate their help so much. In addition, I would like to thank Universiti Utara Malaysia (UUM) too for approving my master study's application and few staffs in Awang Had Salleh Graduate School who assisted me in the process of submission.

I am deeply grateful to my family, my fiance, Chan Jin Swan and my best undergraduate roommate, Nurull Salmi Md Dazali, my Indonesia friend, Fera, my colleague, Wong Sock Leng and my mentor, Wern Lu who encourage me throughout this study by giving infinite motivation. Due to all the support I get, I manage to complete my study. I would like to give my grateful appreciations to all of them.

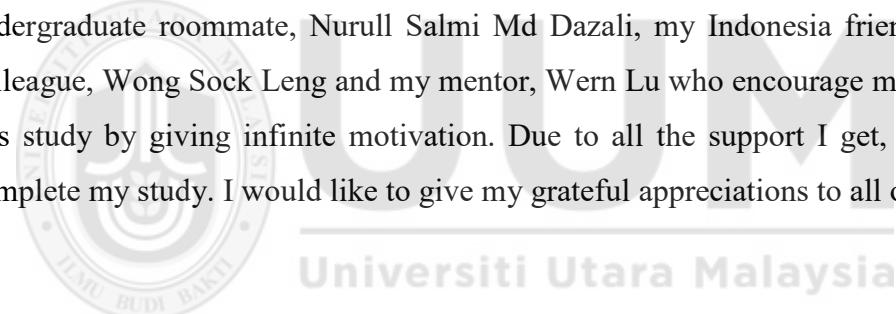


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List of Abbreviations

ANOVA	Analysis of Variance
HL	Hodges-Lehmann Estimator
MAD_n	Median Absolute Deviation about the Median
SAS	Statistical Analysis Software
SAS/IML	Statistical Analysis Software/Interactive Matrix Language
SPSS	Statistical Package for Social Science



CHAPTER ONE

INTRODUCTION

1.1 Introduction

In most research, hypothesis testing has been used as a method of decision making with the help of primary and secondary data that can be obtained from sources such as observations, experiments, journals, articles, reference books and many other sources. The researchers are required to identify the statement of null hypothesis which is usually corresponds to a situation of equality or “no difference” and it is assumed as true hypothesis until receiving an evidence that shows otherwise. Alternative hypothesis is known as the negation of null hypothesis (Sullivan, 2004). Due to the statistical nature of a test, two types of error are determined, Type I error and Type II error. Type I error occurred in the situation where by the null hypothesis is rejected when it is true. In contrast, Type II error existed when the null hypothesis is failed to reject when it is false. There is an inverse relationship between the two errors such that an increase in Type I error will decrease Type II error and vice versa. Furthermore, when Type II error increases, the statistical power of a test will decrease, causing less detection of a test effect. Thus, these two errors need to be in control. A good statistical procedure should be able to control the errors. However, working with Type I error is easier than Type II error as the earlier is usually set in advance by the researcher while the latter is harder to know as it requires estimating the distribution of the alternative hypothesis (Ramsey, 2001).

In order to achieve a good test, we need an appropriate procedure which is able to control Type I error rate and increase the power at the same time. We do not want to

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