

**STATISTICAL PROCESS CONTROL USING MODIFIED ROBUST  
HOTELLING'S  $T^2$  CONTROL CHARTS**

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## Abstrak

Carta Hotelling's  $T^2$  adalah alat yang popular bagi memantau kawalan proses berstatistik. Walau bagaimanapun, carta ini sensitif pada titik terpencil. Bagi mengatasi masalah ini, tiga pendekatan terhadap carta Hotelling's  $T^2$  teguh telah dicadangkan iaitu pendekatan pemangkasan, peWinsoran dan berasaskan median. Kesemua pendekatan ini menggunakan penganggar lokasi teguh dan penganggar skala teguh yang masing-masing menggantikan min biasa dan matriks kovarians. Bagi setiap pendekatan, tiga penganggar skala teguh:  $MAD_n$ ,  $S_n$  dan  $T_n$  diperkenalkan, dan penganggar ini berfungsi sewajarnya mengikut pendekatan. Pendekatan pertama, ditandai sebagai  $T_t^2$ , menggunakan konsep pemangkasan melalui jarak Mahalanobis. Penganggar skala teguh digunakan untuk mengganti matriks kovarians dalam jarak Mahalanobis. Min terpankaskan dan matriks kovarians terpankaskan merupakan penganggar lokasi dan skala bagi carta  $T_t^2$ . Pendekatan kedua,  $T_w^2$ , menggunakan setiap penganggar skala sebagai kriteria Winsor. Pendekatan ini mengaplikasikan penganggar M-satu langkah terubahsuai terWinsor dan kovarians terWinsor yang sepadan, masing-masing sebagai penganggar lokasi dan matriks skala bagi carta  $T_w^2$ . Manakala dalam pendekatan ketiga,  $T_H^2$ , penganggar skala teguh berperanan sebagai matriks skala dengan Hodges-Lehman sebagai penganggar lokasi. Pendekatan ini menggunakan data asal tanpa sebarang pemangkasan atau peWinsoran. Secara keseluruhannya, sembilan carta kawalan teguh telah dicadangkan. Prestasi setiap carta kawalan teguh dinilai berdasarkan kadar penggera palsu dan kebarangkalian mengesan. Bagi mengkaji kekuatan dan kelemahan carta yang dicadangkan, pelbagai keadaan diwujudkan dengan memanipulasi empat pembolehubah iaitu bilangan ciri-ciri kualiti, kadar data terpencil, tahap anjakan min dan sifat ciri-ciri kualiti (bebas dan bersandar). Secara umumnya, carta yang dicadangkan menunjukkan prestasi yang baik dari segi kadar penggera palsu. Dari sudut kebarangkalian mengesan, prestasi kesemua carta yang dicadangkan mengatasi carta Hotelling's  $T^2$  tradisional. Keseluruhannya, kajian mendapati carta Hotelling's  $T^2$  teguh yang dicadangkan boleh dijadikan alternatif yang baik kepada carta tradisional yang dipertikaikan.

Katakunci: Hotelling's  $T^2$ , Carta kawalan, Penganggar teguh

## Abstract

Hotelling's  $T^2$  chart is a popular tool for monitoring statistical process control. However, this chart is sensitive to outliers. To alleviate the problem, three approaches to the robust Hotelling's  $T^2$  chart namely trimming, Winsorizing and median based were proposed. These approaches used robust location and scale estimators to substitute for the usual mean and covariance matrix, respectively. For each approach, three robust scale estimators:  $MAD_n$ ,  $S_n$  and  $T_n$  were introduced, and these estimators functioned accordingly to the approach. The first approach, denoted as  $T_t^2$ , applied the concept of trimming via Mahalanobis distance. The robust scale estimator was used to replace the covariance matrix in Mahalanobis distance. The trimmed mean and trimmed covariance matrix were the location and scale estimators for the  $T_t^2$  chart. The second approach,  $T_w^2$ , employed each scale estimator as the Winsorized criterion. This approach applied Winsorized modified one step M-estimator and its corresponding Winsorized covariance as the location and the scale matrix for  $T_w^2$  chart, respectively. Meanwhile, in the third approach,  $T_H^2$ , the robust scale estimator took the role of the scale matrix with Hodges-Lehman as the location estimator. This approach worked with the original data without any trimming or Winsorizing. Altogether, nine robust control charts were proposed. The performance of each robust control chart was assessed based on false alarm rates and probability of detection. To investigate on the strengths and weaknesses of the proposed charts, various conditions were created by manipulating four variables, namely number of quality characteristics, proportion of outliers, degree of mean shifts, and nature of quality characteristics (independent and dependent). In general, the proposed charts performed well in terms of false alarm rates. With respect to probability of detection, all the proposed charts outperformed the traditional Hotelling's  $T^2$  charts. The overall findings showed that, the proposed robust Hotelling's  $T^2$  control charts are viable alternatives to the disputed traditional charts.

Keywords: Hotelling  $T^2$ , Control chart, Robust estimator

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## List of Abbreviations

***MOM*** Modified One-step *M*-estimator

***HL*** Hodges and Lehmann estimator

***Med*** Median

***MAD<sub>n</sub>*** Median absolute deviation

***S<sub>n</sub>*** A scale estimator

***T<sub>n</sub>*** A scale estimator

**FA** False Alarms

**POD** Probability of Detection

**ARE** Asymptotic Relative Efficiency

***MD*** Mahalanobis Distance



# **CHAPTER ONE**

## **MULTIVARIATE QUALITY CONTROL CHARTS**

### **1.1 Introduction**

The invention of Statistical Process Control (SPC) chart was pioneered by Dr. Walter Shewhart while he was working for Bell Labs in 1920. He aimed to monitor the quality of a process mathematically. Since then, this tool has received tremendous attention and interest from many researchers and practitioners from various fields including statistics, engineering and education to name just a few. There are some definitions of SPC charts tool. We refer to Montgomery (2005), who defined the SPC charts as tool for optimizing the amount of information needed for decision-making purposes. In addition, Nedumaran and Pignatiello (2000) defined the charts as tools to monitor performance or state of the process.

In general, SPC charts are graphical presentations that display the stability of a process. Unlike other common charts, such as bar chart, line chart or pie charts, SPC charts have some main features such as the following:

- (i) The upper limit and lower limit's lines that create a range to where a process output is considered "in control"
- (ii) A center line which located in the middle of the lower and upper limits that reflects the average state of the process.

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