

MULTIVARIATE APPROACH FOR SHOE SIZE MODELLING OF
MALAYSIAN WOMEN

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DEDICATION

This thesis is dedicated to my wife, who taught me that nothing is impossible if we have the desire to achieve it. It is also dedicated to my parent, who taught me that it is required to accomplish one task at a time in order to complete the largest task.

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ABSTRACT

The problem of finding the best fitting shoes among Malaysian women is due to the absence of a Malaysian women shoe sizing system. Currently, shoe manufacturers in Malaysia use United Kingdom (UK) and United States of America (USA) size which is based on the ISO 9407:2019 standard: Mondopoint shoe sizing system, which is basically developed based on foot length and foot breadth. The existing shoe sizes do not cover the width and girth part that is a unique foot shape for Malaysian women. Improper shoe sizing can interrupt the biomechanics of the foot and ankle causing pain and increasing the risk of falling to the wearer. This may create safety issues especially in the workplace like strained foot muscles, slips and falls, and body fatigue which gradually grow into severe musculoskeletal disorders or rheumatic diseases if they are unattended. These problems and related issues were addressed in this study using statistical modelling of foot measurements for the development of the Malaysian women shoe sizing system. The objectives of this study are to investigate the selection of foot measurements used in modelling foot measurements, to enhance existing clustering technique for modelling shoe size, to model foot size variation with multivariate normal model and piecewise regression, and to develop a process to determine shoe size. A set of foot measurement was measured by Infoot 3-dimensional foot scanner; foot length (FL), foot breadth (FB) and foot's ball girth (BG) were used to represent shoe size. 252 women from University of Malaya (UM) volunteered to have their foot measurements taken using the Infoot 3D foot scanner. The data was then split into 161 training data and 91 testing data. The variations of size were investigated using an improved clustering analysis and the multivariate normal mixture probability distribution where a shoe size model is proposed. The process was repeated to five-foot measurements (FL, FB, BG, instep length (IL), and fibulare instep length (FIL)). As for three-foot measurements, FL is treated independently of FB and BG. Conditional on fixed FL value, subdivisions were identified for selected (FB, BG) combination. Taking advantage of the existence of normal probability distribution, univariate hypothesis testing was performed to test group separation and chi-square test to test goodness fit of subdivisions. An alternative approach called piecewise regression was used for five-foot measurements (FL, FB, BG, IL, FIL) to model shoe size as comparison. The study successfully attained the performance of aggregate loss which is less than benchmark value and coverage percentage of 71.25% for the training data. The proposed shoe size was then verified with testing data and showed high coverage percentage of 82.53%. This study concludes that a modest sample size of 252 women was sufficient to develop a prototype shoe sizing system using the proposed novel approach. In conclusion, this study provides a novel process to create a Malaysian women shoe sizing system that considers ball girth (BG), in which most Malaysian women have unique shape. The proposed system provides a significant improvement to the Mondopoint sizing system for Malaysian women. In addition, it also allows for a test of the significant separation of shoe sizes that is overlooked by most ergonomics by using a multivariate normal mixture model.

ABSTRAK

Masalah dalam mencari kasut yang sesuai di kalangan wanita Malaysia adalah kerana ketiadaan sistem ukuran kasut wanita Malaysia. Pada masa ini, pengeluar kasut di Malaysia menggunakan ukuran United Kingdom (UK) dan Amerika Syarikat (AS) yang mengikut piawai ISO 9407:2019 Sistem ukuran kasut Mondopoint yang pada dasarnya dibangunkan berdasarkan panjang kaki (FL) dan kelebaran kaki (FB). Ukuran kasut sedia ada tidak merangkumi bahagian kelebaran dan lilitan bola kaki yang unik untuk wanita Malaysia. Ukuran kasut yang tidak tepat boleh mengganggu biomekanik kaki dan buku lali sehingga mengakibatkan kesakitan dan meningkatkan risiko jatuh bagi si pemakai. Situasi ini boleh menimbulkan masalah keselamatan terutamanya di tempat kerja seperti ketegangan urat kaki, tergelincir dan jatuh, dan keletihan badan yang boleh mengakibatkan gangguan rangka otot dan penyakit reumatik sekiranya tidak ditangani. Kajian ini menangani masalah dan isu yang berkaitan dengan menggunakan model statistik pengukuran kaki untuk membangunkan sistem ukuran kasut wanita Malaysia. Objektif kajian ini adalah untuk mengkaji pemilihan ukuran kaki yang digunakan dalam model pengukuran kaki, menambah baik teknik kelompok yang sedia ada dalam membentuk model ukuran kasut, membangunkan model berasaskan kepelbagaian ukuran kaki menggunakan model multivariat normal dan regresi *piecewise*, dan seterusnya membangunkan proses penentuan saiz kasut. Satu set ukuran kaki diukur menggunakan mesin pengimbas kaki Infoot 3-dimensi; panjang kaki (FL), kelebaran kaki (FB) dan lilitan bola kaki (BG) digunakan untuk mewakili ukuran kasut. 252 wanita dari Universiti Malaya (UM) secara sukarela melakukan pengukuran kaki mereka menggunakan mesin pengimbas kaki Infoot 3-D. Data tersebut kemudian dibahagikan kepada 161 data latihan dan 91 data ujian. Kepelbagaian ukuran dikaji menggunakan analisis pengelompokan yang ditambah baik dan taburan kebarangkalian campuran multivariat normal di mana model ukuran kasut dicadangkan. Proses ini diulang dengan menggunakan lima ukuran kaki (FL, FB, BG, panjang kekura kaki (IL) dan panjang kekura fibular (FIL)). Bagi tiga ukuran kaki, FL diukur secara berasingan dari FB dan BG. Bagi nilai FL yang tetap, sub-bahagian ditentukan bagi kombinasi (FB, BG) terpilih. Berdasarkan kewujudan taburan kebarangkalian normal, ujian hipotesis univariat dijalankan untuk menguji pemisahan kumpulan dan ujian kuasa-dua untuk menguji kesesuaian sub-bahagian. Pendekatan alternatif yang disebut regresi *piecewise* digunakan untuk lima ukuran kaki (FL, FB, BG, IL, FIL) untuk model ukuran kasut sebagai perbandingan. Kajian ini berjaya mencapai prestasi kerugian agregat yang kurang daripada nilai penanda aras dan peratusan liputan sebanyak 71.25% untuk data latihan. Ukuran kasut yang dicadangkan kemudiannya disahkan dengan data ujian dan menunjukkan peratusan liputan yang tinggi iaitu 82.53%. Kajian ini menyimpulkan bahawa ukuran sampel sederhana dari 252 wanita memadai untuk membentuk prototaip sistem ukuran kasut menggunakan pendekatan baru yang dicadangkan. Kesimpulannya, kajian ini menyediakan proses baru dalam membentuk sistem ukuran kasut wanita Malaysia yang mempertimbangkan BG, di mana kebanyakan wanita Malaysia mempunyai bentuk yang unik. Sistem yang dicadangkan memberikan peningkatan yang signifikan terhadap sistem ukuran Mondopoint untuk wanita Malaysia. Di samping itu, ia juga membolehkan untuk menguji pemisahan saiz kasut yang diabaikan oleh kebanyakan ahli ergonomik dengan menggunakan model multivariat normal.

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LIST OF ABBREVIATIONS

FSS	-	Foot Sizing System
FL	-	Foot Length
FB	-	Foot Breadth
BG	-	Ball Girth
IL	-	Instep Length
FIL	-	Fibulare Instep Length
EM	-	Expectation Maximization
MLE	-	Maximum Likelihood Estimator
LRT	-	Likelihood Ratio Test
AIC	-	Akaike's Information Criterion
BIC	-	Bayesian Information Criterion
CPCC	-	Cophenetic Correlation Coefficient
TEMS	-	Technical Error Measurements
SE	-	Standard Error
MSE	-	Mean Square Error

LIST OF SYMBOLS

d	-	Difference between first and second measurement
N	-	Number of samples
f	-	Probability density distribution
π_r	-	Mixing proportion
θ_r	-	Estimated parameter
$g_r(\cdot)$	-	Component densities of the distribution
r	-	Number of clusters
s	-	Maximum number of clusters
H_0	-	Null hypothesis
H_1	-	Alternative hypothesis
m	-	Number of parameters
μ	-	Mean
σ	-	Standard deviation
Z	-	Expected value
χ^2	-	Chi-square statistic
f_i	-	Observed frequency
e_i	-	Expected frequency
r_{ij}	-	Correlation between foot measurements

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CHAPTER 1

INTRODUCTION

1.1 Research Background

It is very crucial to choose ideal footwear since it plays a major role in protecting the foot in daily activity. Choosing the right footwear can help ensure the long-term health of our feet, as well as our entire body because it will provide essential mechanical support that upholds body weight and structural balance while withstanding the pressure, most of the time. The best fit for a user may depends on their daily use.

The search for the ideal footwear leads to the need for a foot sizing system (FSS) for Malaysian women. The development of FSS for Malaysian women is from two approaches; firstly statistical methods and ideas using foot measurements to represent the foot shape for examples foot breadth (FB) and foot length (FL), and secondly the ideas of expressing or measuring shape using Fourier Descriptor (FD) (Zhang and Lu, 2004). This thesis focuses on the first approach, and only gives introduction of the second approach. The study on the variation of foot measurement is important since it will determine a standard for shoe size determination.

Malaysia is one of the largest producers of footwear in Asia as stated in the MATRADE's report (MATRADE, 2015). This success is due to the ability to deliver high-quality products at reasonable prices as well as extensive marketing. Besides having exports to South East Asia, a few European and the Middle East countries, most of the footwear is sold to the Malaysian market. The Malaysian footwear industry is continuously growing since it is supported by an experienced and skilled workforce with both technical and practical skills (Insights, 2014). Currently, there are 1,000 footwear manufacturers in Malaysia employing some 30,000 people. They are mainly located in the states of Perak, Selangor, and Johor (Insights, 2014).

These number emphasize the significance of the local shoe-making industry and its contribution to the country's economy. For instance, the industry recorded a stunning RM1 billion in sales in 2013 as reported by the Malaysian Tourism Ministry, involving RM621 million footwear exports to the international market as well as RM300 million in the domestic market (Awani, 2013). As comparison, MATRADE statistics show total footwear exports in 2017 slightly increased to RM652.7 million (Shah, 2018). Malaysia was ranked 19th worldwide regarding shoe exports, with Ipoh as the main footwear production center for the country (Awani, 2013).

A 4% increase in footwear export was recorded in the year 2014 for Malaysian footwear manufacturers, of which 90% was for high-end ladies' footwear (MATRADE, 2015). For Malaysia to move from Asia's leading shoe manufacturer to become a global competitor, Malaysian footwear manufacturers should be proactive to adopt new technology and innovation to improve footwear quality and refine their skills and production. Since ladies' footwear dominates the share of footwear production, it is important to concentrate on this market segment.

There are five type of foot measurement being used this research which is foot length (FL), foot breadth (FB), ball girth (BG), instep length (IL) and fibulare instaep length (FIL). Table 1. 1 shows the detail of the main foot measurement being used in this study. Figure 1. 1 and Figure 1. 2 show illustration of foot measurements of foot length, ball girth, foot breadth, instep length, Fibulare instep length and anatomical bone and landmark names (I-Ware Laboratory Co., 2017).

Table 1. 1 Definition of Foot Measurement Used

Foot Measurement	Definition
Foot Length (FL)	The distance between the back point of the heel and the foremost of the longest toe. Refer Figure 1. 1.
Foot Breadth (FB)	The distance between the metatarsal tibial and metatarsal fibulare of the ball cross section projected to the standing surface. Refer Figure 1. 1.
Ball Girth (BG)	Circumference of foot. It measured around the last from the lateral ball point to the medial ball point. Refer Figure 1. 1.
Instep Length (IL)	The distance between the back point of the heel and the first metatarsophalangeal (MTP) protrusion. Refer Figure 1. 1.
Fibulare Instep Length (FIL)	The distance between the back point of the heel and the fifth metatarsophalangeal (MTP) protrusion. Refer Figure 1. 1.

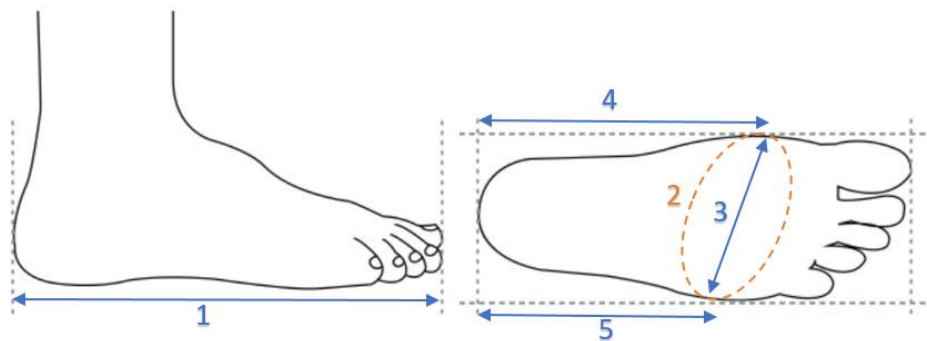


Figure 1. 1 Illustration of Foot Measurements. 1: Foot length, 2: Ball girth, 3: Foot breadth, 4: Instep length, 5: Fibulare instep length

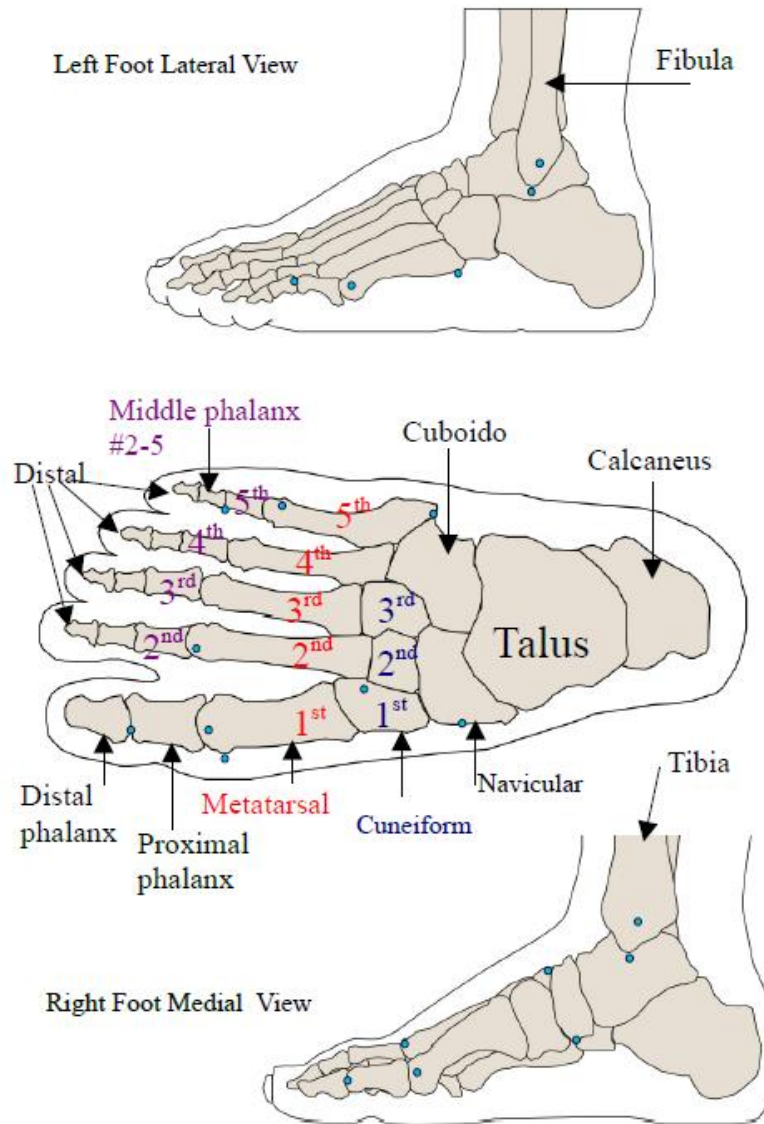


Figure 1. 2 Anatomical Bone and Landmark Names

1.2 Problem Background

The production of shoe was made depends on the shoe last which is a 3-dimensional wooden or plastic mould, that provides human anatomical foot measurement. The shoe last used will determine the overall fit of the shoe and describes the foot shape measurement such as heel width, instep height, forefoot width, and toe box depth. The shoe last was made following the standard sizing system to fit a certain population for the economic and large-scale production purpose in the shoe

industry (Hinojo-Pérez *et al.*, 2016). All of this is made possible only if the following problems are considered.

One of the issues in developing a shoe sizing system is a different group of peoples might have a different sizing system for example; the shoe sizing system for adults and children is different. It also differs for men and women (Mishra *et al.*, 2017). The knowledge of sex-related differences in foot measures is vital to assist proper shoe fit in both men and women (Hong *et al.*, 2011; Krauss *et al.*, 2008; Krauss *et al.*, 2010). Nowadays, there are sizing systems developed particularly for activity-specific, function-specific and fashion-specific shoes. At least two categories of shoe sizing standard exist, either based on stick length or actual foot length (Cheng and Perng, 2000). The stick length method categorized the foot based on the measurement of foot length only where it is measured from the last toe to heel top points. The measurement is longer than the shoe last bottom length. It is widely used in European countries, America and Britain such as the British Sizing System, French (Continental) Sizing System, Euro point Sizing System. The second category is based on foot length and foot girth or foot width. Shoe sizing standards that apply this method include the Japanese Sizing System, Mondopoint Sizing System, and Mainland China sizing system.

Another issue concerns health and injury. The improper shoe size selected by the user exerts constant pressure on a certain area of the foot, mostly on the medial and lateral surfaces of the foot that can be produced from shoes that are too small. While the shoes that are too large can cause the foot sliding within the shoes, and the effect is friction ulcers that occur behind the heel (Most and Sinnock, 1983; Reddy *et al.*, 1989). Besides that, improper shoe sizing can interrupt the biomechanics of the foot and ankle causing pain and increasing the chances of falling (Harrison *et al.*, 2007; Manna *et al.*, 2001). This creates safety issues in the workplace like strained foot muscles, slips and falls, body fatigue and other problems (Witana *et al.*, 2004), which gradually grow into severe musculoskeletal disorders or rheumatic diseases if not addressed. A lot of research have been done from different perspectives such as ergonomics, anthropometry, engineering physiology, industrial design, and foot science to address the issue of injury. Some studies enhanced footwear performance

and safety and produced a positive impact on the user and market. Figure 1. 3 shows the example of the effect of using the wrong shoe size.

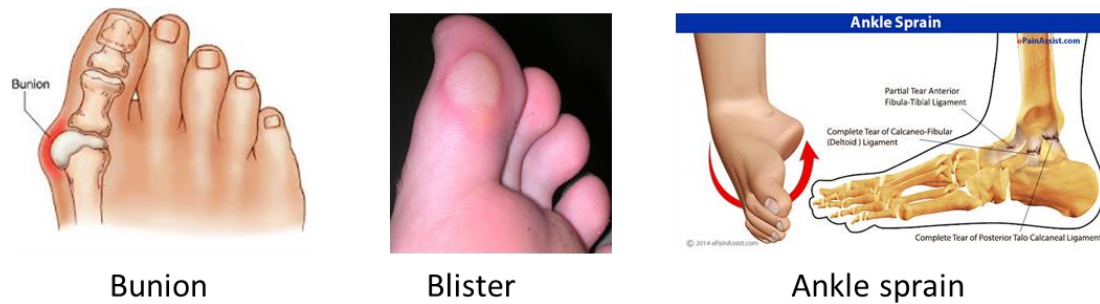


Figure 1. 3 Example of the effect of using the wrong shoe size

A major commercial concern is that of shoe comfort; a problem for both producer and consumer. A shoe's comfort may be influenced by the inside shoe climate and factors such as colour and fashion (Slater, 1986). Other factors that may lead to user satisfaction are foot shape (Hawes *et al.*, 1994), skeletal alignment (Miller *et al.*, 2000) and shoe fit (Witana *et al.*, 2004). Improper fit of footwear may cause slips, falls (Hignett and Masud, 2006) or body fatigue (Lin *et al.*, 2007). Other than that, the material properties influence the shoe's comfort (Goonetilleke, 1999; Mündermann *et al.*, 2002; Yung-Hui and Wei-Hsien, 2005; Zhang *et al.*, 1991).

Several approaches have been introduced to ensure comfortability of shoes produced. Au and Goonetilleke (2007) sought to find the preferred fit in the different regions of the shoe along with the characteristics that distinguish between comfortable and uncomfortable shoes. In the study, the shoes were divided into six regions; the toe region, Metatarso-Phalangeal Joint (MPJ) region, arch region, rearfoot region, Ingress/Egress opening and fastening type. The rear fit region and fastening type do not show a significant difference between different fit ratings. The study concludes that the comfortable shoe does not require a perceived fit in every region of the shoe. The above issues aggravate the shoe design problem where if consider too many foot measurements, it will require a large number of shoe size and is a problem in shoe manufacturing.

Branthwaite *et al.* (2013a) focused on how the toe box shape can affect dorsal and plantar pressure with particular interest around the forefoot in a healthy female population. Three different types of toe box styles were examined; round, squared and pointed. As a result, the shoes with a round toe established the least pressure around the medial aspect of the toe while the pointed shoes distributed least amount of pressure in the lateral toe area. Another finding is that the volume of the foot does not correlate to forefoot pressure. The study also recommends shoe manufacturers or designers to consider footwear design around the toe box to improve fit and reduce pressure.

Even though the consideration of foot function and foot health are important in footwear choice, Branthwaite *et al.* (2013b) consider a different perspective. They claim that footwear choices by younger women are activity-specific and subjects chose the style and design of shoes related to the image they want to portray. The summary for problem background regarding the shoe sizing system shown in Figure 1. 4.

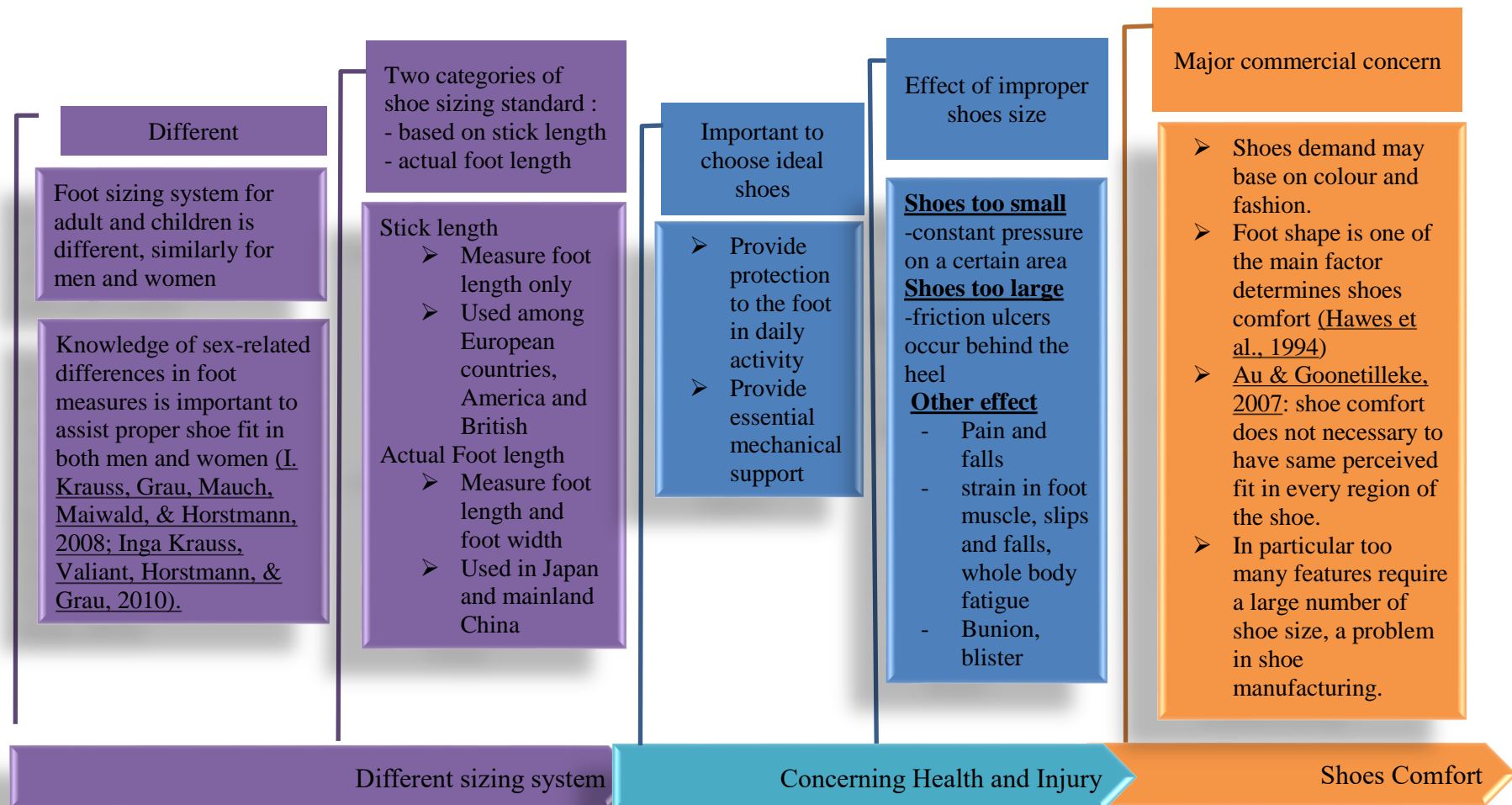


Figure 1. 4 Summary of Problem Background

1.3 Problem Statement

To the best of our knowledge, there is no standard shoe sizing for Malaysian women with little attention paid to model or measure variations in shoe shape. Currently, Malaysia is using the sizing systems of other countries such as ISO 9407:2019 (Standard, 2019) which is the Mondopoint shoe sizing system. Ibrahim and Khedifb (2004) disclosed that most Malaysian women have difficulty choosing the right shoe size as they have a unique foot shape, particularly at the girth and width. This is proven by surveys conducted by Chua *et al.* (2013) and Shariff *et al.* (2014) which revealed that more than 60% of Malaysian women have difficulty choosing the right shoe size and are dissatisfied with the existing selected shoe size. Therefore, it is crucial for Malaysian women to have their shoe sizing system.

This thesis applies statistical methods and ideas in particular multivariate approach to address issues (i) choice of shape features, (ii) the number of shape groups, (iii) classification of subjects using shape features and (iv) general inference from shape features, to determine the foot shape of Malaysian women. Historically, ergonomics and allometry dominate all issues concerning the shape of the human anatomy. Issues (i), (ii), (iii) and (iv) have been addressed using ad-hoc approaches and methods especially on the selection of shape features from statistical views and the separation between different group of shapes (Cheng and Perng, 2000; Hill *et al.*, 2017; Kalebota *et al.*, 2003; Lee and Wang, 2015). This study offers a systematic approach to the issues by using foot measurements to represent foot shape.

Agglomerative hierarchical clustering analysis is commonly used in foot shape classification (Bataller *et al.*, 2001; Lee and Wang, 2015; Mauch *et al.*, 2009). However, different results may be found when using different clustering algorithms or techniques (Everitt, 1981). Because the heuristic method is infrequently conclusive, the model-based clustering technique was developed by Fraley and Raftery (2002) to conform with the standard statistical method and a multivariate normal mixture model. However, it can lead to be over-fitting model if the number of clusters is treated as the number of mixing elements in the multivariate normal mixture model (Baudry *et al.*, 2010). There was no application of the multivariate normal mixture to study the human

foot shape variation being done since most of the classification studies satisfied with the outcome from cluster analysis. Thus, this study utilizes the multivariate normal mixture model to enhance the clustering technique for modelling shoe size.

1.4 Research Objectives

This study aims to propose a process that can help create a Malaysian women shoe sizing system based on multivariate approaches. To achieve the aim, this study has four main objectives which are;

- (a) To investigate the selection of foot measurements in modelling shoe size.
- (b) To enhance the clustering technique for modelling shoe size.
- (c) To model foot size variation with multivariate approaches.
- (d) To develop a process to determine Malaysian women shoe size.

At the end of the research, a novel process is proposed for the development of a Malaysian women shoe sizing system based on multivariate approaches with a view of improving the existing shoe sizing system.

1.5 Scope of The Research

This study will concentrate on women's shoe sizing for ladies above 18 years only. The final closure of growth plates in the feet normally occurs by the age of 18 years old. Hence, the data was randomly collected for women aged 18 years old and above, from University of Malaya staff and students which represents the population fairly well. The data was collected from the previous study by using a research grant funded by The Ministry of Education Malaysia, University of Malaya and Universiti Teknologi Malaysia (Shariff *et al.*, 2014).

In order to address issues of (i) choice of shape features, (ii) the number of shape groups, (iii) classification of subjects using shape features and (iv) general inference from shape features as stated in Section 1.3, and to achieve the objectives as stated in Section 1.4, this study focused on (a) Modelling foot shape using existing statistical technique, (b) Providing critical appraisal of existing technique using statistical ideas or theory, (c) Investigating choice of foot measurement using clustering of variables, (d) Emphasizing on statistical test for multivariate normal mixture; example are means of 2- ellipsoid different, and (e) Proposing a sizing system by applying subdivision of the ellipsoid technique. The approach undertaken in this study is for small sample size analysis. This study focused on the flat type of shoe sizes.

1.6 Significance of The Research

This study proposed a process that complements the Mondopoint sizing system for creating the appropriate shoe sizing system for Malaysian women. The proposed process allows for statistical test to investigate the choice for (i) number of clusters or groups (ii) the significant subdivision of foot measurements.

1.7 Definition of Key Terms

This study contains certain key terms. Although these key terms are discussed in detail in Chapter 2 and Chapter 3, they are briefly introduced in this early chapter of the thesis to allow the reader to make sense of what is presented in the subsequent Chapters. Table 1.2 shows the definition of few key terms used in this study.

Table 1. 2 Definition of Key Terms

Key Terms	Definition
Technical Error Measurements (TEMS)	Index to express the error margin in anthropometry where it represents the accuracy and the quality of the tool used in this study.
Cophenetic Correlation Coefficient (CPCC)	Correlation between original distance matrix and cophenetic distance matrix. An optimal cluster is defined when the large value of the CPCC index.
Expectation Maximization (EM) algorithm	An iterative method to find maximum likelihood estimates of parameters in statistical models in the presence of latent variables.
Ellipsoid	The geometrically representative of a probability mixture model of 3 foot measurements
Akaike's Information Criterion (AIC)	An estimator of out-of-sample prediction error and thereby relative quality of statistical models for a given set of data. It is an information criteria that going to be used in justify number of clusters.
Bayesian Information Criterion (BIC)	An index used in Bayesian statistics to choose between two or more alternative models. It is also an information criteria that going to be used in justify number of clusters.
Likelihood Ratio Test (LRT)	A test to assesses the goodness of fit of two competing statistical models based on the ratio of their likelihoods, specifically one found by maximization over the entire parameter space and another found after imposing some constraint. It is used to justify the number of clusters gained from EM algorithm.

1.8 Overview of The Thesis

This thesis is divided into five chapters. It starts by introducing the topic in Chapter 1 which includes foot size variations introduction, the background of study and a problem statement. This chapter also explains the objectives, scope of the research and data analysis methods that were used in this research.

Chapter 2 reviews the literature relevant to the study. It begins by reviewing the literature concerning the shoe problem, followed by the problems faced by women in Malaysia concerning foot shape. The last section is an overview of the methods of foot shape modelling including clustering analysis, multivariate normal mixture technique, and piecewise regression technique. Chapter 2 also reviewed the validation techniques applied to the proposed shoe sizing system.

Chapter 3 details the methodology of the research. It consists of flowcharts that help the reader understand the method applied in the research. It also offers an overview of the Agglomerative Hierarchical clustering analysis including distance measure and linkage method used in this research, piecewise regression, and multivariate normal mixture model. This study used four types of distance which were Euclidean, Manhattan (City block), Maximum and Canberra. The linkage methods employed in this study are single, complete, group average, centroid, median and Ward linkage.

Chapter 4 starts with providing the explanatory data analysis including descriptive statistical data analysis, the selection of foot measurement, the accuracy of measurement and outlier detection. It also provides the shoe size modelling study through multivariate normal mixture method including; the selection of distance measure and linkage method based on the Cophenetic Correlation Coefficient (CPCC) index, the optimal number of component based on likelihood ratio test (LRT) p-value, Akaike's information criterion (AIC), and Bayesian information criterion (BIC). Chapter 4 also presents the shoe size modelling study through the piecewise regression method including; the selection y -axis variables and x -axis variables by using the cluster of features analysis; and determining breakpoint for piecewise regression.

Chapter 5 starts with providing a blind shoe sizing system. Due to a large number of shoe sizes, this chapter provides a new shoe sizing system created from a multivariate normal mixture model. The one-way ANOVA test and post hoc then give how is the separation between shoe sizes created from a multivariate normal mixture model. Then, it also gives a comparison between the shoe sizing system created from 3-foot measurement and 5-foot measurements. Chapter 5 also gives the validation result of the subdivision proposed by a multivariate normal mixture model besides the validation of using the test dataset. Chapter 5 ends with showing the shoe size created from the piecewise regression model as a comparison.

Chapter 6 conclude the novelty of the study, short comings and future works. It also provide the limitations of the research.

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Appendix A Comparison of CPCC Index of Different Distance Measures for 3 Foot measurements and Its Combination by Different Linkage Method.

(a) Linkage: Average Distance measure

Foot shape combination	Euclidean	Manhattan	Maximum	Canberra
FL	0.7492	0.7492	0.7492	0.7385
FB	0.7756	0.7756	0.7756	0.7750
BG	0.7872	0.7872	0.7872	0.7752
FL and FB	0.7282	0.7141	0.7161	0.7156
FL and BG	0.7248	0.7190	0.7243	0.6773
FB and BG	0.7929	0.7874	0.7808	0.7807
FL, FB and BG	0.7411	0.7422	0.7237	0.7555

(b) Linkage: Single Distance measure

Foot shape combination	Euclidean	Manhattan	Maximum	Canberra
FL	0.6966	0.6966	0.6966	0.6730
FB	0.6780	0.6780	0.6780	0.6550
BG	0.6988	0.6988	0.6988	0.6820
FL and FB	0.6018	0.6143	0.5903	0.5978
FL and BG	0.5828	0.5831	0.6295	0.5658
FB and BG	0.5696	0.5343	0.5991	0.3639
FL, FB and BG	0.5270	0.5171	0.5833	0.4182

(c) Linkage: Complete Distance measure

Foot shape combination	Euclidean	Manhattan	Maximum	Canberra
FL	0.6472	0.6472	0.6472	0.6441
FB	0.6729	0.6729	0.6729	0.7016
BG	0.7567	0.7567	0.7567	0.7284
FL and FB	0.5850	0.6119	0.6244	0.6541
FL and BG	0.5649	0.5783	0.6341	0.5977
FB and BG	0.6188	0.6259	0.7222	0.7476
FL, FB and BG	0.6419	0.6099	0.6099	0.5665

(d) Linkage: Centroid Distance measure

Foot shape combination	Euclidean	Manhattan	Maximum	Canberra
FL	0.7548	0.7548	0.7548	0.7454
FB	0.7719	0.7719	0.7719	0.7371
BG	0.7718	0.7718	0.7718	0.7758
FL and FB	0.7206	0.7092	0.7108	0.7106
FL and BG	0.7393	0.7209	0.6934	0.7091
FB and BG	0.7907	0.7932	0.7879	0.7587
FL, FB and BG	0.7488	0.7499	0.7283	0.7484

(e) Linkage: Median Distance measure

Foot shape combination	Euclidean	Manhattan	Maximum	Canberra
FL	0.6537	0.6537	0.6537	0.7028
FB	0.7059	0.7059	0.7059	0.7452
BG	0.7636	0.7636	0.7636	0.6480
FL and FB	0.4936	0.5423	0.5761	0.5065
FL and BG	0.5263	0.6361	0.6223	0.5780
FB and BG	0.7612	0.6601	0.7135	0.7481
FL, FB and BG	0.5668	0.5893	0.5705	0.6174

(f) Linkage: Ward Distance measure

Foot shape combination	Euclidean	Manhattan	Maximum	Canberra
FL	0.6330	0.6330	0.6330	0.6291
FB	0.6292	0.6292	0.6292	0.6000
BG	0.6126	0.6126	0.6126	0.6251
FL and FB	0.5963	0.6012	0.5358	0.5437
FL and BG	0.5964	0.5639	0.5252	0.5706
FB and BG	0.6101	0.5989	0.6002	0.6940
FL, FB and BG	0.5745	0.5802	0.4893	0.5902

* the yellow box indicates the highest value of CPCC index for a fix foot measurement combination.

Appendix B Comparison of CPCC Index of Different Distance Measures for 5 Foot measurements and Its Combination by Different Linkage Method

Shape combination	Average				Single			
	Euclidean	Manhattan	Maximum	Canberra	Euclidean	Manhattan	Maximum	Canberra
FL	0.749	0.749	0.749	0.739	0.697	0.697	0.697	0.673
FB	0.776	0.776	0.776	0.775	0.678	0.678	0.678	0.655
BG	0.787	0.787	0.787	0.775	0.699	0.699	0.699	0.682
IL	0.788	0.788	0.788	0.788	0.704	0.704	0.704	0.708
FIL	0.782	0.782	0.782	0.777	0.674	0.674	0.674	0.659
FL, FB	0.728	0.714	0.716	0.716	0.602	0.614	0.590	0.598
FL, BG	0.725	0.719	0.724	0.677	0.583	0.583	0.629	0.566
FL, IL	0.763	0.764	0.762	0.753	0.707	0.696	0.697	0.687
FL, FIL	0.769	0.690	0.776	0.673	0.524	0.502	0.592	0.497
FB, BG	0.793	0.787	0.781	0.781	0.570	0.534	0.599	0.364
FB, IL	0.698	0.681	0.707	0.660	0.671	0.660	0.665	0.568
FB, FIL	0.736	0.759	0.750	0.722	0.710	0.637	0.737	0.569
BG, IL	0.763	0.748	0.733	0.706	0.678	0.655	0.663	0.614
BG, FIL	0.787	0.786	0.778	0.752	0.630	0.653	0.570	0.621
IL, FIL	0.788	0.764	0.803	0.754	0.622	0.539	0.697	0.543
FL, FB, BG	0.741	0.742	0.724	0.755	0.527	0.517	0.583	0.418
FL, FB, IL	0.719	0.723	0.699	0.684	0.584	0.579	0.569	0.555
FL, FB, FIL	0.684	0.665	0.765	0.648	0.528	0.480	0.587	0.497
FL, BG, IL	0.714	0.671	0.715	0.692	0.635	0.613	0.669	0.584
FL, BG, FIL	0.753	0.692	0.743	0.711	0.585	0.548	0.606	0.534
FL, FIL,IL	0.731	0.721	0.766	0.661	0.528	0.475	0.622	0.461
FB, BG, IL	0.760	0.766	0.734	0.752	0.603	0.547	0.644	0.470
FB, BG, FIL	0.788	0.781	0.757	0.767	0.594	0.559	0.558	0.476
FB, IL, FIL	0.748	0.745	0.787	0.711	0.628	0.537	0.690	0.517
BG, IL, FIL	0.752	0.727	0.756	0.713	0.636	0.604	0.648	0.545
FL, FB, BG, IL	0.716	0.683	0.718	0.682	0.551	0.524	0.646	0.467
FL, FB, BG, FIL	0.738	0.732	0.742	0.731	0.561	0.508	0.599	0.465
FL, BG,IL, FIL	0.670	0.681	0.752	0.678	0.618	0.554	0.635	0.521
FB, BG,IL, FIL	0.763	0.722	0.755	0.703	0.623	0.555	0.648	0.488
FL, FB,IL,FIL	0.717	0.716	0.769	0.686	0.545	0.496	0.627	0.508
FL,FB,BG,IL,FIL	0.709	0.705	0.755	0.677	0.590	0.524	0.631	0.468
Shape combination	Complete				Centroid			
	Euclidean	Manhattan	Maximum	Canberra	Euclidean	Manhattan	Maximum	Canberra
FL	0.647	0.647	0.647	0.644	0.755	0.755	0.755	0.745
FB	0.673	0.673	0.673	0.702	0.772	0.772	0.772	0.737
BG	0.757	0.757	0.757	0.728	0.772	0.772	0.772	0.776
IL	0.610	0.610	0.610	0.622	0.785	0.785	0.785	0.781
FIL	0.588	0.588	0.588	0.553	0.803	0.803	0.803	0.776
FL, FB	0.585	0.612	0.624	0.654	0.721	0.709	0.711	0.711
FL, BG	0.565	0.578	0.634	0.598	0.739	0.721	0.693	0.709
FL, IL	0.635	0.626	0.713	0.642	0.761	0.726	0.759	0.747
FL, FIL	0.710	0.707	0.650	0.590	0.751	0.679	0.759	0.718
FB, BG	0.619	0.626	0.722	0.748	0.791	0.793	0.788	0.759
FB, IL	0.570	0.509	0.540	0.584	0.729	0.701	0.713	0.681
FB, FIL	0.578	0.605	0.650	0.653	0.769	0.732	0.741	0.699
BG, IL	0.623	0.636	0.729	0.556	0.757	0.730	0.753	0.703
BG, FIL	0.715	0.605	0.561	0.648	0.772	0.767	0.760	0.729
IL, FIL	0.568	0.655	0.527	0.597	0.771	0.747	0.775	0.734
FL, FB, BG	0.642	0.610	0.610	0.566	0.749	0.750	0.728	0.748
FL, FB, IL	0.591	0.585	0.590	0.567	0.746	0.716	0.708	0.708
FL, FB, FIL	0.629	0.579	0.533	0.571	0.716	0.721	0.718	0.700
FL, BG, IL	0.545	0.674	0.593	0.571	0.721	0.685	0.717	0.702
FL, BG, FIL	0.698	0.600	0.583	0.671	0.712	0.690	0.710	0.684
FL, FIL,IL	0.635	0.598	0.689	0.592	0.737	0.724	0.662	0.736
FB, BG, IL	0.619	0.614	0.663	0.629	0.764	0.747	0.755	0.748

FB, BG, FIL	0.618	0.716	0.583	0.662	0.781	0.771	0.758	0.750
FB, IL, FIL	0.663	0.594	0.621	0.641	0.739	0.723	0.751	0.696
BG, IL, FIL	0.647	0.662	0.607	0.552	0.712	0.682	0.745	0.703
FL, FB, BG, IL	0.610	0.591	0.659	0.615	0.726	0.699	0.727	0.690
FL, FB, BG, FIL	0.695	0.584	0.682	0.589	0.731	0.699	0.729	0.688
FL, BG,IL, FIL	0.663	0.562	0.686	0.533	0.679	0.691	0.697	0.706
FB, BG,IL, FIL	0.695	0.640	0.573	0.538	0.715	0.709	0.744	0.703
FL, FB,IL,FIL	0.617	0.597	0.679	0.546	0.707	0.744	0.710	0.713
FL,FB,BG,IL,FIL	0.608	0.549	0.681	0.572	0.683	0.674	0.722	0.678
	Median				Ward			
Shape combination	Euclidean	Manhattan	Maximum	Canberra	Euclidean	Manhattan	Maximum	Canberra
FL	0.654	0.654	0.654	0.654	0.654	0.654	0.654	0.654
FB	0.706	0.706	0.706	0.706	0.706	0.706	0.706	0.706
BG	0.764	0.764	0.764	0.764	0.764	0.764	0.764	0.764
IL	0.680	0.680	0.680	0.680	0.680	0.680	0.680	0.680
FIL	0.759	0.759	0.759	0.759	0.759	0.759	0.759	0.759
FL, FB	0.494	0.542	0.494	0.542	0.494	0.542	0.494	0.542
FL, BG	0.526	0.636	0.526	0.636	0.526	0.636	0.526	0.636
FL, IL	0.715	0.610	0.715	0.610	0.715	0.610	0.715	0.610
FL, FIL	0.676	0.591	0.676	0.591	0.676	0.591	0.676	0.591
FB, BG	0.761	0.660	0.761	0.660	0.761	0.660	0.761	0.660
FB, IL	0.556	0.462	0.556	0.462	0.556	0.462	0.556	0.462
FB, FIL	0.639	0.590	0.639	0.590	0.639	0.590	0.639	0.590
BG, IL	0.432	0.504	0.432	0.504	0.432	0.504	0.432	0.504
BG, FIL	0.652	0.533	0.652	0.533	0.652	0.533	0.652	0.533
IL, FIL	0.652	0.637	0.652	0.637	0.652	0.637	0.652	0.637
FL, FB, BG	0.567	0.589	0.567	0.589	0.567	0.589	0.567	0.589
FL, FB, IL	0.672	0.573	0.672	0.573	0.672	0.573	0.672	0.573
FL, FB, FIL	0.649	0.591	0.649	0.591	0.649	0.591	0.649	0.591
FL, BG, IL	0.692	0.611	0.692	0.611	0.692	0.611	0.692	0.611
FL, BG, FIL	0.569	0.635	0.569	0.635	0.569	0.635	0.569	0.635
FL, FIL,IL	0.518	0.611	0.518	0.611	0.518	0.611	0.518	0.611
FB, BG, IL	0.578	0.601	0.578	0.601	0.578	0.601	0.578	0.601
FB, BG, FIL	0.631	0.734	0.631	0.734	0.631	0.734	0.631	0.734
FB, IL, FIL	0.685	0.510	0.685	0.510	0.685	0.510	0.685	0.510
BG, IL, FIL	0.606	0.519	0.606	0.519	0.606	0.519	0.606	0.519
FL, FB, BG, IL	0.567	0.482	0.567	0.482	0.567	0.482	0.567	0.482
FL, FB, BG, FIL	0.703	0.595	0.703	0.595	0.703	0.595	0.703	0.595
FL, BG,IL, FIL	0.515	0.650	0.515	0.650	0.515	0.650	0.515	0.650
FB, BG,IL, FIL	0.748	0.579	0.748	0.579	0.748	0.579	0.748	0.579
FL, FB,IL,FIL	0.624	0.562	0.624	0.562	0.624	0.562	0.624	0.562
FL,FB,BG,IL,FIL	0.537	0.612	0.537	0.612	0.537	0.612	0.537	0.612

* the green box indicates the highest value of CPCC index for all distance measure of six different linkage method of a certain foot measurement combination.

Appendix C R-Command for Statistical Approach

Cluster of Features to Select Foot Measurements

```
LF1<-read.table("C:\\Faizal Hamzah\\Faizal Hamzah\\PhD\\Proposal\\Foot scan
data\\Left foot_PG\\All features left.txt",header=T)
cor(LF1)
d.corr <- as.dist(1-cor(LF1)^2)
d.corr
#plot dendrogram
hc <- hclust(d.corr, method='average')
plot(as.dendrogram(hc))
#calculate CPCC
d2 <- cophenetic(hc)
cor (d.corr,d2)

cor(LF1)
d.corr <- as.dist(cor(LF1))
d.corr
#plot dendrogram
hc <- hclust(d.corr, method='ward')
plot(as.dendrogram(hc))
#calculate CPCC
d2 <- cophenetic(hc)
cor (d.corr,d2)
```

Outlier Detection From Boxplot, 2D plot and 3D plot

```
#boxplot with outlier label
library(car)
#For combine group with label

Boxplot(LFL ~ hc2, data=temp,
        col = "lightgray",
        main = "Foot Measuremet for Foot Length (FL)",
        xlab = "Group",
        ylab = "FL")
Boxplot(LFB ~ hc2, data=temp,
        col = "lightgray",
        main = "Foot Measuremet for Foot Breadth (FB)",
        xlab = "Group",
        ylab = "FB")
Boxplot(LBGC ~ hc2, data=temp,
        col = "lightgray",
        main = "Foot Measuremet for Ball Girth (BG)",
        xlab = "Group",
        ylab = "BG")
```



```

#To plot 2-D graph based on cluster result
LF1<-read.table("C:\\Faizal Hamzah\\Faizal Hamzah\\PhD\\Proposal\\Foot scan
data\\Left foot_PG\\left11.txt",header=T)
d1 <- dist(LF1,method = "euclidean")
hc <- hclust(d1, method = "average")
hc2<-cutree(hc,k=2)
dframe = data.frame(LF1, hc2)
temp<-cbind(LF1,hc2)
plot(temp$LFL,temp$LFB,pch = 21, bg=c("red","blue")[unclass(temp$hc2)],
main = "Foot Breadth (FB) Vs Foot Length (FL)", xlab="FL", ylab="FB")
plot(temp$LFL,temp$LBGC,pch = 21, bg=c("red","blue")[unclass(temp$hc2)],
main = "Ball Girth (BG) Vs Foot Length (FL)", xlab="FL", ylab="BG")
plot(temp$LBGC,temp$LFB,pch = 21, bg=c("red","blue")[unclass(temp$hc2)],
main = "Foot Breadth (FB) Vs Ball Girth (BG)", xlab="BG", ylab="FB")

#To plot 3-D graph based on cluster result
library(rgl)
??plot3d
plot3d(temp,type='s', col=temp$hc2,size=1)

```

Clustering of Subject

```

library(mixtools)
LF1<-read.table("C:\\Faizal Hamzah\\Faizal Hamzah\\PhD\\Proposal\\Foot scan
data\\Left foot_PG\\left11_3.txt",header=T)
d1 <- dist(LF1,method = "euclidean")
hc <- hclust(d1, method = "average")
d2 <- cophenetic(hc)
cor (d1,d2)
hc1<-cutree(hc,k=2)
library(doBy)
dframe1 = data.frame(LF1, hc1)
summaryBy(LFL~hc1, data=dframe1, FUN=c(mean,sd))
summaryBy(LBGC~hc1, data=dframe1, FUN=c(mean,sd))
summaryBy(LFB~hc1, data=dframe1, FUN=c(mean,sd))
for(i in 4){dframe1[,i]<-factor(dframe1[,i])}Ba#assign hc1 as factor
summary (dframe1)
temp<-cbind(LF1,hc1)
str(temp)
cov(temp[hc1==1,1:3])
cov(temp[hc1==2,1:3])
LF1<-as.matrix(LF1)
a<-boot.comp(LF1, x = NULL, N = NULL, max.comp = 2, B = 100,
sig = 0 ,mix.type = "mvnormalmix", hist = TRUE)
a
a$p.value

```

Multivariate Normal Mixture

```
lambda <- c(0.8625,0.1375)
mu <- list(c(232.15, 94.11, 225.77), c(247.53,107.64,257.82))
sigma <-
list(matrix(c(109.879,27.312,66.058,27.312,20.378,46.609,66.058,46.609,118.710),
3, 3),
matrix(c(134.837,5.218,16.031,5.218,7.643,15.267,16.031,15.267,40.687), 3, 3))
library( mixtools)
set.seed(100)
LF1<-read.table("C:\\Faizal Hamzah\\Faizal Hamzah\\PhD\\Proposal\\Foot scan
data\\Left foot_PG\\left11_3.txt",header=T)
LFLFBFBG<-
mvnnormalmixEM(LF1,lambda=lambda,mu=mu,sigma=sigma,arbvar = TRUE, k=2)
LFLFBFBG<- mvnnormalmixEM(LF1,k=3)
summary (LFLFBFBG)
LFLFBFBG$mu
LFLFBFBG$sigma
LFLFBFBG$lambda
```

Piecewise Regression

```
##Cluster of features
LF1<-read.table("C:\\Faizal Hamzah\\Faizal Hamzah\\PhD\\Proposal\\Foot scan
data\\Left foot_PG\\left all_3.txt",header=T)
##dendrogram by using corr as distance
cor(LF1)
d.corr <- as.dist(cor(LF1))
d.corr
#plot dendrogram
hc <- hclust(d.corr, method='ward')
plot(as.dendrogram(hc))
#calculate CPCC
d2 <- cophenetic(hc)
cor (d.corr,d2)

##piece wise for FB vs BG based on highest R2 value
LF1<-read.table("C:\\Faizal Hamzah\\Faizal Hamzah\\PhD\\Proposal\\Foot scan
data\\Left foot_PG\\left all_3.txt",header=T)
x<-LF1$BG
y<-LF1$FB
#plot normal simple regression
LR<-lm (y~x,data=LF1)
summary(LR)
plot (y~x, pch=16, cex=0.9, xlab="BG", ylab="FB")
abline(lm(y~x, data=LF1), col="red")

#sort BG
Break<- (sort(unique(x))[2:160])
```

```

Break
#working now yeay
mse<-numeric(131)
for (i in 1:131) {
model<-lm(y~(x<Break[i])*x + (x>=Break[i])*x)
mse[i]<-summary(model)[[6]] }
mse <- as.numeric(mse)
mse
plot(mse, pch=16)
#one possible breakpoint 232.5
piecewise <- lm(y ~ x*(x < 232.5) + x*(x > 232.5))
summary(piecewise)

```

Appendix D Mathematical Approach To Subdivide The Ellipsoid

Given $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$, then we used

$$Pr \left[(\underline{x} - \underline{\mu})^T \Sigma^{-1} (\underline{x} - \underline{\mu}) \leq C \right] \quad (D.1)$$

where $C \sim \chi^2(p)$. If $y \leq C$ where $y \sim \chi^2(p)$, then ellipsoid can be represented as;

$$(\underline{x} - \underline{\mu})^T \Sigma^{-1} (\underline{x} - \underline{\mu}) = C \quad (D.2)$$

Estimate of $\underline{\mu}$ is given by

$$\hat{\underline{\mu}} = \bar{\underline{x}} = \frac{1}{n} (\underline{x}_1 + \underline{x}_2 + \dots + \underline{x}_n) \quad (D.3)$$

where \underline{x}_k represent the k-th subject (person),

and the estimate of Σ is given by

$$\hat{\Sigma} = \frac{1}{n-1} \sum_{i=1}^n (\underline{x}_i - \bar{\underline{x}})(\underline{x}_i - \bar{\underline{x}})^T \quad (D.4)$$

In equation (D.1), (6.2), (6.3) and (6.4), $\underline{\mu}$ and Σ from the k-th ellipsoid is replaced with $\underline{\mu}_k$ and Σ_k , $k = 1, 2$. The value of n in (D.4) is replaced by n_k for the k-th ellipsoid, $k = 1, 2$. Subdivide the ellipsoid into small 3-D subdivisions is equivalent to subdivide the respective sphere since it can be shown that,

$$N_p(\underline{\mu}, \Sigma) \rightarrow N_p(\underline{0}, I) \quad (D.5)$$

using $\underline{v} = \underline{z} - w \quad (D.6)$

$$\underline{v} = \Lambda^{-\frac{1}{2}} \left(Q^T [\underline{x} - \underline{\mu}] \right) \quad (D.7)$$

Estimates of Q and Λ is given from $\hat{\Sigma} = \hat{Q} \Lambda \hat{Q}^T$ Result in (D.5) will be illustrated in the following 2 pages.

Suppose $\underline{x} = (x_1, x_2, \dots, x_p)$ where the individual \underline{x} is described by p -foot measurements (for example x_1 equal foot length). Let $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$. Then the following standard theorems may be used.

Theorem D (a);

For $A^T \in M_{qp}$ and $\underline{b} \in \mathbb{R}^q$, let $\underline{y} = A^T \underline{x} + \underline{b}$.

$$\text{Then } \underline{y} \sim N_q(A^T \underline{x} + \underline{b}, A^T \Sigma A) \quad (\text{D. 8})$$

Theorem D (b);

If $\underline{x} \sim N_p(\underline{\mu}, \Sigma)$ and Σ is non-singular. Then,

$$y = (\underline{x} - \underline{\mu})^T \Sigma^{-1} (\underline{x} - \underline{\mu}) \sim \chi^2(p) \quad (\text{D. 9})$$

Also, $y = c$ is an ellipsoid, say $c = \chi_{0.05}^2(p)$

Theorem D (c); Spectral decomposition of symmetric matrices

If A is $(n \times n)$ symmetric matrix with eigen value $\lambda_1 \geq \lambda_2 \geq \dots \geq \lambda_n$ and eigen vector (q_1, q_2, \dots, q_n) .

When $Aq_i = \lambda_i q_i$ and $q_i^T q_i = 1$ with $q_i^T q_j = 0 (i \neq j)$, then

$$A = Q \Lambda Q^T \quad (\text{D. 10})$$

where $Q = [q_1 | q_2 | \dots | q_p]$ is an orthogonal $Q^T Q = 1$

and $\underline{\Lambda} = \text{diag}(\lambda_1, \lambda_2, \dots, \lambda_n)$

ELLIPSOID TO UNIT SPHERE

Let $\underline{x} \sim N(\underline{\mu}, \Sigma)$, $\underline{\mu}$ is a set of mean for foot length, foot breadth and ball girth, and Σ is a set of covariance for foot length, foot breadth and ball girth. We know that $\Sigma = Q \Lambda Q^T$ is an orthogonal.

Then,

$$y = Q^T \underline{x} \sim N_p(Q^T \underline{\mu}, Q^T \Sigma Q) \quad (\text{D. 11})$$

Since $\Sigma = Q \Lambda Q^T$, then

$$\begin{aligned} Q^T \Sigma Q &= Q^T (Q \Lambda Q^T) Q \\ &= \Lambda \end{aligned}$$

$$Q^T \underline{x} \sim N_p(Q^T \underline{\mu}, \Lambda) \quad (\text{D. 12})$$

Let

$$\Lambda^{\frac{1}{2}} = \begin{pmatrix} \lambda_1^{\frac{1}{2}} & 0 & 0 & 0 \\ 0 & \lambda_2^{\frac{1}{2}} & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & \lambda_p^{\frac{1}{2}} \end{pmatrix}$$

$$\Lambda^{\frac{1}{2}} \Lambda^{\frac{1}{2}} = \Lambda = \begin{pmatrix} \lambda_1 & 0 & 0 & 0 \\ 0 & \lambda_2 & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & \lambda_p \end{pmatrix}$$

$$\text{Let } \underline{z} = \Lambda^{-\frac{1}{2}} Q^T \underline{x} \quad (\text{D.13})$$

$$\therefore \underline{z} \sim N_p \left(\Lambda^{-\frac{1}{2}} Q^T \underline{\mu}, I \right) \quad (\text{D.14})$$

Given that

$$\Lambda^{-\frac{1}{2}} \Lambda \Lambda^{-\frac{1}{2}} = \Lambda^{-\frac{1}{2}} \Lambda^{\frac{1}{2}} \Lambda^{\frac{1}{2}} \Lambda^{-\frac{1}{2}}$$

$$\Lambda^{-\frac{1}{2}} \Lambda \Lambda^{-\frac{1}{2}} = I_p I_p = I_p$$

$$\therefore \underline{z} \sim N_p(\underline{w}, I) \quad (\text{D.15})$$

$$\text{where } \underline{w} = \Lambda^{-\frac{1}{2}} Q^T \underline{\mu} \quad (\text{D.16})$$

Finally, let $\underline{v} = \underline{z} - \underline{w}$

$$\therefore \underline{v} \sim N_p(\underline{0}, I) \quad (\text{D.17})$$

Now, we slice up ellipsoid from (D.11) and convert back result in term of \underline{x} . Note that,

$$\underline{v} = \underline{z} - \underline{w} = \Lambda^{-\frac{1}{2}} Q^T (\underline{x} - \underline{\mu}) \quad (\text{D.18})$$

APPLICATION OF UNIT SPHERE FOR SIZING SYSTEM

Let $\underline{v}^T = (v_1, v_2, \dots, v_p)$ and $\underline{x}^T = (x_1, x_2, \dots, x_p)$. We have $\underline{v} = \Lambda^{-\frac{1}{2}} Q^T (\underline{x} - \underline{\mu})$ from (D.13). To get back to \underline{x} we use the following

$$\underline{x} = (Q^T)^{-1} \Lambda^{\frac{1}{2}} \underline{v} + \underline{\mu} \quad (\text{D.19})$$

Q^T is nonsingular, so that Q^{-T} is an unique and there will be one to one correspond between \underline{x} and $\underline{\mu}$. From theorem D (b) and result from (D.17)

$$y = (\underline{v} - \underline{0})^T I^{-1} (\underline{v} - \underline{0}) \sim \chi^2(p) \quad (\text{D.20})$$

The quadratic term in (D.20) and result (D.20), and recall $p = 3$ in section D.1, therefore

$$v_1^2 + v_2^2 + v_3^2 = c \quad (\text{D.21})$$

where $c = \chi_{0.95}^2(3)$ from standard Chi-squared table.

The values of v_1, v_2 and v_3 are data dependent and once they are selected, equation D.19 give the foot measurement.

LIST OF PUBLICATIONS

Hamzah, M. F. M., Rijal, O. M., Mokhtar, N. A., and Noor, N. M. (2017a). Piecewise regression for determining Malaysian woman foot shape. Paper presented at the 2017 IEEE International Conference on Signal and Image Processing Applications (ICSIPA), 379-384.

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