Handling Non-Normal Data in Individual and Moving Range Control Charts: An Application to Raw Rubber Manufacturing Process Data

by

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Handling Non-Normal Data in Individual and Moving Range

Control Charts: An Application to Raw Rubber Manufacturing

Process Data



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Candidate's Declaration

The work described in this thesis was carried by me under the supervision of Prof. B. M. S. G. Banneheka and Dr. Keminda Herath and a report on this has not been submitted in whole or in part to any university or any other institution for another Degree/Diploma. I certify that all additions and amendments have been incorporated in the thesis in accordance with the comments and suggestions of the examiners.

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"Where I am Today, That's Because of You All"

Affectionately Dedicated

То

All My Respectable Teachers,

My Beloved Grandparents, Parents and Sister

My Dearest Husband

TABLE OF CONTENTSi
LIST OF ABBREVIATIONSvi
LIST OF TABLESvii
LIST OF FIGURESix
ACKNOWLEDGMENTSxv
ABSTRACTxvii
1.0 INTRODUCTION1
1.1 Quality Improvement Using Statistical Process Control1
1.2 Quality Systems in Raw Rubber Manufacturing Process in Sri Lanka2
1.3 Non-Normality Issue on Data When Constructing Individual and Moving Range (I-MR) Control Charts4
1.4 Research Objectives5
1.4 Research Objectives51.5 Structure of the Thesis
1.5 Structure of the Thesis5
1.5 Structure of the Thesis
1.5 Structure of the Thesis 5 2.0 BACKGROUND AND LITERATURE SURVEY 7 2.1 Statistical Basis of Control Chart 7 2.1.1 Role of the Control Chart in Statistical Process Control 8 2.1.2 Nonrandom Patterns of the Data Plotted on the Control Chart 10 2.1.3 Analyzing the Performance of a Control Chart 11

TABLE OF CONTENTS

2.3.2 Interpreting the I-MR Charts
2.3.2.1 Interpreting the Moving Range Chart23
2.3.2.2 Interpreting the Individual Chart24
2.3.3 Operating Characteristic Curve and Average Run Length for I Chart24
2.3.3.1 Operating Characteristic (OC) Curve for I Chart24
2.3.3.2 Average Run Length (ARL) for I Chart25
2.3.4 Assumptions for I-MR Control Charting in Statistical Quality Control25
2.3.4.1 Assessing the Normality Assumption of Data27
2.3.4.2 Solutions for Non-normality Issue of Data
2.4 Use of SPC Techniques in Rubber Industry31
2.4.1 Rubber Industry in Sri Lanka31
2.4.2 Use of Control Charts in Rubber Industry
2.5 Applications of the Use of Control Charts for Non Normal Data33
3.0 METHODOLOGY
3.1 The Factory Data36
3.2 Use of I-MR Charts for Non-normal Data45
3.3 Methodology Developed to Evaluate the Performance of Control Charts48
3.3.1 Data Generated from Non-Normal Distributions
3.3.2 Data Generated from Empirical Distribution Function (EDF) of Factory Data
3.3.3 Evaluation Procedure to Compare the Performance of I-MR Charts using Simulation Study

3.3.3.1 Theoretical Power Curve of Standard 3-Sigma I Chart When
Mean is Shifted52
3.3.3.2 Theoretical Power Curve of Standard 3-Sigma I Chart When
Standard Deviation is Shifted
3.3.3.3 Theoretical Power Curve of Standard 3-Sigma I Chart When
Both Mean and Standard Deviation are Shifted53
3.3.3.4 Steps for Estimation of Power of I Chart for Simulated Data
When the 'In Control' Mean is Shifted54
3.3.3.5 Steps for Estimation of Power of I Chart of Factory Data When
the 'In Control' Mean is Shifted57
3.3.3.6 Steps for Estimation of Power of I Chart When the 'In Control'
Standard Deviation is Shifted
3.3.3.7 Steps for Estimation of Power of I Chart When the Both 'In Control' Mean and 'In Control' Standard Deviation are
Shifted
4.0 RESULTS AND DISCUSSION
4.1 Distributional Properties of the Factory Data65
4.1.1 Normality Assumption of Factory Data65
4.1.2 Fitting Distribution to the Factory Data
4.2 Box-Cox Transformation <i>versus</i> Johnson Transformation for Factory
Data68
4.3 Qunatile-Quantile (Q-Q) Plot to Assess Normality of Transformed Factory
Data70
4.4 I-MR Charts Developed at Phase I for the Quality
Characteristics75

4.5	Comparative	Performance	of	I-MR	Charts	Using	Simulation
	Study	•••••	•••••	•••••	•••••	••••••	89
	4.5.1 Estimated I	Probability Type	I Erro	or and AI	RL ₀ of I-M	IR Charts	
	4.5.2 Compariso	n of Power Curv	e of I	Chart w	ith the Th	eoretical	Power Curve
	of Standar	d 3 Sigma I Cha	rt			· · · · · · · · · · · · · · · · · · ·	94
	4.5.2.1 Sim	nulation Study 1	(wher	n 'in cont	rol' mean	is shifted	
		nulation Study 2					
	shif	fted)					101
		nulation Study 3					
				,			
	4.5.3 Power Cu Runs		•		-		
4.6	Summary	••••••••••••	•••••	•••••	•••••	•••••	118
5.0 RV	WUI WEB APPL	ICATION FOR	R STA	TISTIC	AL PRO	CESS CO	ONTROL
	1 Introduction						
5.2	2 Requirements t	o Use Rwui We	b App	olication	•••••	•••••	119
5.3	3 Rwui Web App	lication for Stat	istica	l Process	s Control	(SPC)	120
	5.3.1 How to En	ter Data to Web	Appli	cation?			
	5.3.2 How to Us	e First Web Page	e?				121
	5.3.3 How to Us	e Second Web P	age?	•••••			130
	5.3.4 How to Us	e Third Web Pag	ge?				137
	5.3.5 Detailed In	struction Page					143

5.3.7 Further Work of the Rwui Web Application for SPC	143
6.0 CONCLUDING REMARKS	144
6.1 General	144
6.2 Limitations of the Study	147
6.3 Further Work	148
REFERENCES	149
APPENDICES	159
Appendix 1 - List of Publications and Communications from Thesis	159
Appendix 2 - List of People who have Importance of this Research	161
Appendix 3 - Data Sets	162
Appendix 4 - R built-in function for Johnson Transformation	178

LIST OF ABREVIATIONS

AD	-	Anderson-Darling
ARL	-	Average Run Length
CL	-	Center Line
СМ	-	Cramer-von Mises
CUSUM	-	Cumulative Sum
DMAIC	-	Define, Measure, Analyze, Improve and Control
EWMA	-	Exponentially Weighted Moving Average
HDD	-	Hard Disk Drive
I- chart	-	Individual chart
I-MR	-	Individual and Moving Range
ISO	-	International Organization for Standardization
KS	-	Kolmogorov-Smirnov
LCL	-	Lower Control Limit
LF	-	Lilliefors
MCUSUM	-	Multivariate Cumulative Sum
MEWMA	-	Multivariate Exponentially Weighted Moving Average
MR-chart	-	Moving Range chart
OC	-	Operating Characteristics
OCP	-	Out of Control Points
PC	-	Pearson chi-square
Q-Q	-	Quantile - Quantile
RRISL	-	Rubber Research Institute of Sri Lanka
RSS	-	Ribbed Smoked Sheet
Rwui	-	R Web User Interface
SF	-	Shapiro-Francia
SPC	-	Statistical Process Control
SW	-	Shapiro-Wilk
TSR	-	Technically Specified Rubbers
UCL	-	Upper Control Limit

LIST OF TABLES

Table 2.1	Functions associated with the three families of the Johnson System	31
Table 3.1	Variables identified which can be used for control charts, their definitions, units of measurements, engineering limits and limits used Payagala rubber factory	l by 39
Table 3.2	Recommended dosages for the chemicals	45
Table 4.1	Normality tests of different quality parameters for individ observations	lual 67
Table 4.2	p-values of AD test for fitting different types of distributions to factory data	the 68
Table 4.3	Normality tests of different quality parameters for transformed data	70
Table 4.4	Reliable control limits of I-MR control charts derived on Johnson transformed data of the quality characteristics for both sites	82
Table 4.5	Percentage of number of samples generated from different non nor distributions which can be converted to normal using two types of transformations, Box-Cox and Johnson	mal 90
Table 4.6	Estimated probabilities of type I error and ARL ₀ of I chart develo using the data generated from different non-normal distributions using factory data	-
Table 4.7	Estimated probabilities of type I error and ARL ₀ of MR chart develo using the data generated from different non-normal distributions using factory data	•
Table 4.8	Estimated probabilities of type I error and ARL ₀ of I chart develo using the data generated from gamma and log normal distributions we respect to Box-Cox transformation and Johnson transformation	•

vii

Table 4.9Estimated probabilities of type I error and ARL0 of MR chart developed
using the data generated from gamma and log normal distributions with
respect to Box-Cox transformation and Johnson transformation94

LIST OF FIGURES

Figure 2.1	Role of control chart in process improvement	9
Figure 3.1	Process map of latex crepe manufacturing process up to productio coagulum	on of 37
Figure 3.2	Process map of milling and subsequent operations of latex crepe manufacturing	38
Figure 3.3	Probability density functions for (a) Gamma distribution, (b) Expo	onential
	distribution, and (c) Log normal distribution with respect to d values of the relevant parameters	lifferent 49
Figure 4.1	Histograms of different quality parameters for individual obsercollected at two sites	rvations 66
Figure 4.2	Q-Q plots for (a) original data,(b) Johnson transformed data, and (transformed data for site 2-Formic (1/100kg DRC)	(c) back 72
Figure 4.3	Q-Q plots for (a) original data, (b) Johnson transformed data, back transformed data for site1-Na ₂ S ₂ O ₃ (kg/100DRCkg) and Q- for (d) original data,(e) Johnson transformed data, and (f) back transformed data for site2-NaHSO ₃ (kg/100DRCkg)	
Figure 4.4	Q-Q plots for (a) original data, (b) Johnson transformed data, back transformed data for site1-Formic (l/100DRCkg)	and (c) 74
Figure 4.5	Q-Q plots for (a) original data, (b) Johnson transformed data, back transformed data for site1-bleaching agent (g/100DRCkg) a plots for (d) original data, (e) Johnson transformed data, and (transformed data for site2-bleaching agent (g/100DRCkg)	nd Q-Q
Figure 4.6	I-MR control charts developed initially at phase I for site 1 J transformed factory data of $Na_2S_2O_3$ (kg/100DRCkg)	Johnson 76
Figure 4.7	I-MR control charts developed initially at phase I for site 1 J transformed factory data of Formic (l/100DRCkg)	Johnson 77

Figure 4.8	I-MR control charts developed initially at phase I for site 1 Johnson transformed factory data of bleaching agent (g/100DRCkg) 78
Figure 4.9	I-MR control charts developed initially at phase I for site 2 Johnson transformed factory data of NaHSO ₃ (kg/100DRCkg) 79
Figure 4.10	I-MR control charts developed initially at phase I for site 2 Johnson transformed factory data of Formic (l/100DRCkg) 80
Figure 4.11	I-MR control charts developed initially at phase I for site 2 Johnson transformed factory data of bleaching agent (g/100DRCkg) 81
Figure 4.12	I-MR control charts with reliable control limits for site1-Na ₂ S ₂ O ₃ (kg/100DRCkg) 83
Figure 4.13	I-MR control charts with reliable control limits for site1- Formic (l/100DRCkg) 84
Figure 4.14	I-MR control charts with reliable control limits for site1-bleaching agent (g/100DRCkg) 85
Figure 4.15	I-MR control charts with reliable control limits for site2-NaHSO ₃ (kg/100DRCkg) 86
Figure 4.16	I-MR control charts with reliable control limits for site2- Formic (l/100DRCkg) 87
Figure 4.17	I-MR control charts with reliable control limits for site2-bleaching agent (g/100DRCkg) 88
Figure 4.18	Comparison of estimated power curves of I charts for simulated data from (a) gamma(2,.5), (b) gamma(3,.5), (c) gamma(9,2) and (d) gamma(7.5,1) to the theoretical power curve of 3-sigma I chart, $1 - (\Phi(3-h) - \Phi(-3-h))$ 97
Figure 4.19	Comparison of estimated power curves of I charts for simulated data from (a) $exp(.5)$, (b) $exp(1)$, and (c) $exp(1.5)$ to the theoretical power

curve of 3-sigma I chart, $1 - (\Phi(3 - h) - \Phi(-3 - h))$ 98

- Figure 4.20 Comparison of estimated power curves of I charts for simulated data from (a) $\ln N(0,.25^2)$, (b) $\ln N(0,.5^2)$, and (c) $\ln N(0,1)$ to the theoretical power curve of 3-sigma I chart, $1 (\Phi(3 h) \Phi(-3 h))$ 99
- Figure 4.21 Comparison of estimated power curves of I charts for factory data of (a) site 1 Na₂S₂O₃ (kg/100kg DRC), (b) site 1 Formic (l/100kg DRC) (c) site 1 Bleaching agent (g/100kg DRC), (d) site 2 NaHSO₃ (kg/100kg DRC), (e) site 2 Formic (l/100kg DRC) and (f) site 2 Bleaching agent (g/100kg DRC) to the theoretical power curve of 3-sigma I chart, $1 - (\Phi(3 - h) - \Phi(-3 - h))$ 100
- Figure 4.22 Comparison of estimated power curves of I charts for simulated data from (a) gamma(2,.5), (b) gamma(3,.5), (c) gamma(9,2) and (d) gamma(7.5,1) to the theoretical power curve of 3-sigma I chart, $1 (\Phi[3/h] \Phi[-3/h])$ 102
- Figure 4.23 Comparison of estimated power curves of I charts for simulated data from (a) exp(.5), (b) exp(1), and (c) exp(1.5) to the theoretical power curve of 3-sigma I chart, $1 (\Phi[3/h] \Phi[-3/h])$ 103
- Figure 4.24 Comparison of estimated power curves of I charts for simulated data from (a) $\ln N(0,.25^2)$, (b) $\ln N(0,.5^2)$, and (c) $\ln N(0,1)$ to the theoretical power curve of 3-sigma I chart, $1 (\Phi[3/h] \Phi[-3/h])$ 104
- Figure 4.25 Comparison of estimated power curves of I charts for factory data of (a) site 1 Na₂S₂O₃ (kg/100kg DRC), (b) site 1 Formic (l/100kg DRC) (c) site 1 Bleaching agent (g/100kg DRC), (d) site 2 NaHSO₃ (kg/100kg DRC), (e) site 2 Formic (l/100kg DRC) and (f) site 2 Bleaching agent (g/100kg DRC) to the theoretical power curve of 3-sigma I chart, $1 (\Phi[3/h] \Phi[-3/h])$ where h is a ratio of new to old process standard deviation. 105
- Figure 4.26 Comparison of estimated power curves of MR charts for simulated data from (a) gamma(3,.5), (b) exp(1.5), (c) lnN(0,.5²) and (d) factory data / site 2- formic(l/100kg DRC) 106

- Figure 4.27 Comparison of estimated power curves of I charts for simulated data from (a) gamma(2,.5), (b) gamma(3,.5), (c) gamma(9,2) and (d) gamma(7.5,1) to the theoretical power curve of 3-sigma I chart, $1 - (\Phi[(3 - h_{\mu})/h_{\sigma}] - \Phi[(-3 - h_{\mu})/h_{\sigma}])$ 109
- Figure 4.28 Comparison of estimated power curves of I charts for simulated data from (a) exp(.5), (b) exp(1), and (c) exp(1.5) to the theoretical power curve of 3-sigma I chart, $1 - (\Phi[(3 - h_{\mu})/h_{\sigma}] - \Phi[(-3 - h_{\mu})/h_{\sigma}])$ 110
- Figure 4.29 Comparison of estimated power curves of I charts for simulated data from (a) $\ln N(0, 25^2)$, (b) $\ln N(0, .5^2)$, and (c) $\ln N(0, 1)$ to the theoretical power curve of 3-sigma I chart, $1 (\Phi[(3 h_{\mu})/h_{\sigma}] \Phi[(-3 h_{\mu})/h_{\sigma}])$ 111
- Figure 4.30 Comparison of estimated power curves of I charts for factory data of (a) site 1 Na₂S₂O₃ (kg/100kg DRC), (b) site 1 Formic (l/100kg DRC) (c) site 1 Bleaching agent (g/100kg DRC), (d) site 2 NaHSO₃ (kg/100kg DRC), (e) site 2 Formic (l/100kg DRC) and (f) site 2 Bleaching agent (g/100kg DRC) to the theoretical power curve of 3-sigma I chart,1 – $(\Phi[(3 - h_{\mu})/h_{\sigma}] - \Phi[(-3 - h_{\mu})/h_{\sigma}])$ 113
- Figure 4.31 Comparison of estimated power curves of MR charts for simulated data from (a) gamma(3,.5), (b) exp(1), (c) lnN(0,0.5²), and (d) factory data / site 2-NaHSO₃ (kg/100kg DRC) 114
- Figure 4.32 Estimated power curves of I charts for simulated data from (a) gamma(2,.5), (b) factory data / site 1-Na₂SO₃ (kg/100kg DRC) when the process mean is shifted by h where h is the magnitude of the shift in the process mean.
- Figure 4.33 Estimated power curves of I charts for simulated data from (a) gamma(2,.5), (b) factory data / site 1-Na₂SO₃ (kg/100kg DRC) when the process standard deviation is shifted according to h which is a ratio of new to old process standard deviation.

- Figure 4.34 Estimated power curves of I charts for simulated data from (a) gamma(2,.5), (b) factory data / site 1-Na₂SO₃ (kg/100kg DRC) when both process mean and process standard deviation are shifted according to hMU and hSIGMA respectively where hMU is the magnitude of the shift in the process mean and hSIGMA is a ratio of new to old process standard deviation. 117
- Figure 5.1 Screenshot showing the welcome page of the first web page 124
- Figure 5.2Screenshot showing the first web page after uploading data files when
user runs this application for the first time125
- Figure 5.3 Screenshot showing the result page which includes first MR chart developed for the data and the description of data points beyond the control limits 126
- Figure 5.4Screenshot showing the first web page after uploading data files when
user runs this application after first time running127
- Figure 5.5 Screenshot showing the result page which includes reliable (final) MR chart developed for site1-Na₂SO₃ data 128
- Figure 5.6Screenshot showing the links appears on the web page for each output
generated during the process of developing reliable MR chart129
- Figure 5.7 Screenshot showing the welcome page of the second web page 132
- Figure 5.8Screenshot showing the second web page after uploading data files when
user runs this application for the first time133
- Figure 5.9 Screenshot showing the result page which includes first I chart developed for the data and the description of data points beyond the control limits and violating runs 134
- Figure 5.10Screenshot showing the second web page after uploading data files when
user runs this application after first time running135

- Figure 5.11Screenshot showing the result page which includes reliable (final) I chart
developed for site1-Na2SO3 data136
- Figure 5.12 Screenshot showing the welcome page of the third web page 139
- Figure 5.13 Screenshot showing the third web page after uploading a data file which includes data generated from current process 140
- Figure 5.14 Screenshot showing the result page which includes reliable I-MR charts together with relevant data values which were generated from 'in control' process 141
- Figure 5.15 Screenshot showing the result page which includes reliable I-MR charts together with relevant data values which were generated from 'out of control' process 142

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Manage Prabhashrini Dhanushika

ABSTRACT

This research focused on control charting for quality assurance of crepe rubber manufacturing process in Sri Lanka. There is lack of interest in applying quality tools for process monitoring in crepe rubber industry due to the crude nature of this process. However, some researchers have identified individual and moving range (I-MR) control charts as a suitable statistical process control (SPC) technique for rubber data. When applying control charting in SPC, normality assumption of data should be considered since performances of the control charts are very much sensitive to departures from normality. Rubber data also show a significant deviation from normal assumption. No comprehensive study has been done to establish a proper practice for laymen to use I-MR charts in crepe rubber industry when data are non-normal. Therefore, the main objective of this research was to find a solution for non-normality issue of data on I-MR control charting and hence to introduce a feasible way to apply I-MR charts for process improvement in crepe rubber manufacturing industry.

We have identified Johnson transformation as a suitable method to handle nonnormality issue of data; other simple transformations and distribution fitting did not work for rubber data. By incorporating Johnson transformation, a feasible way to use I-MR charts was developed. Evaluation of the method was based on Monte Carlo simulation and probability of type I error, average run length and the power of control charts were estimated under the simulation study. We have evaluated the performances of I-MR charts for non normal data generated from different non normal distributions.

xvii

In the case of factory (rubber) data, data were generated from the empirical distribution of factory data in order to evaluate the performance of I-MR charts with respect to the practical situation. In this research, data were collected from two sites where the crepe rubber manufacturing process runs; Dartonfield rubber factory and Payagala rubber factory. Data collected during the period of two years commencing from year 2004 at Dartonfield factory and data collected during the period from September 2010 to August 2011 at Payagala factory were used.

The results of simulation study clearly indicate the high sensitivity of I-MR charts to non-normality. I chart developed for Johnson transformed data produces relatively low values for probability of type I error. Small values of average run length when the process is actually in control (ARL₀) for original data strongly show the bad impact of not addressing non normality issue for control charting in SPC. Same figures which are shown for I chart, can be seen for MR chart also, but comparatively MR chart gets lower values for ARL₀ for Johnson transformed data than I chart obtains. It indicates the low effectiveness of MR chart for process monitoring when compared to the effectiveness of I chart. Further, I chart developed for Johnson transformed data produces a power curve which is aligned with the theoretically accepted power curve of standard 3-sigma I chart. Therefore, the method of applying I-MR charts by incorporating Johnson transformation can be recommended for non normal data in any industry.

Finally, a user friendly web application that can be used by any industry person to develop I-MR charts for process monitoring task when the process generates non normal data has been created by using Rwui (R web user interface) which is a web application which can be used to create user friendly web interfaces for R scripts.

xviii