

## Faculty of Electrical and Electronic Engineering Technology



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**Bachelor of Electrical Engineering Technology (Industrial Automation & Robotics) with Honours** 

# ESTIMATING THE WEIGHT OF MANGOES USING IMAGE PROCESSING AND ANALYSIS TECHNIQUES

### ISZUZULDIN AMIRULL BIN MOHD JANUDIN

A project report submitted in partial fulfillment of the requirements for the degree of Bachelor of Electrical Engineering Technology (Industrial Automation & Robotics)



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### UNIVERSITI TEKNIKAL MALAYSIA MELAKA

FAKULTI TEKNOLOGI KEJUTERAAN ELEKTRIK DAN ELEKTRONIK

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### **DECLARATION**

I declare that this project report entitled "Estimating The Weight of Mangoes Using Image Processing and Analysis Techniques" is the result of my research except as cited in the references. The project report has not been accepted for any degree and is not concurrently submitted in the candidature of any other degree.

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### **DEDICATION**

I want to thank everyone who has helped me with this thesis. Furthermore, I am grateful to both my parents and siblings for their constant support and encouragement in helping me finish this bachelor's degree project (BDP). I wouldn't have made it this far without them. Apart from that, I'd want to thank all of my fellow lecturers for their guidance, instruction, and mentoring during my studies. Not to mention all of my dear friends I've met along the way. Finally, I'd like to express my gratitude to everyone who has helped me complete this project, whether directly or indirectly.



#### **ABSTRACT**

One of the most important factors that consumers associate with mango quality is its size. The weight of the Mango is commonly used to assess its size. This project aims to offer research that uses image processing and analysis techniques to estimate the weight of mangoes. The mango picture produced from the image acquisition system was processed and analyzed using the MATLAB program. There have two methods were used to find the number of pixels of mangoes, which are area and perimeter. This project offers an application that can choose which method is most suitable for the estimated weight of mango based on observations of r-squared values. The relationship between mango pixels and mango weights was examined using linear regression as a statistical approach. As a consequence, methodologies for mango weight assessment using image processing and analysis are practical, viable, and successful.

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#### **ABSTRAK**

Salah satu faktor terpenting yang dikaitkan pengguna dengan kualiti mangga ialah saiznya. Berat Mangga biasanya digunakan untuk menilai saiznya. Projek ini bertujuan untuk menawarkan penyelidikan yang menggunakan pemprosesan imej dan teknik analisis untuk menganggarkan berat mangga. Gambar mangga yang dihasilkan daripada sistem pemerolehan imej telah diproses dan dianalisis menggunakan program MATLAB. Terdapat dua kaedah yang digunakan untuk mencari bilangan piksel mangga iaitu luas dan perimeter. Projek ini menawarkan aplikasi yang boleh memilih kaedah yang paling sesuai untuk anggaran berat mangga berdasarkan pemerhatian nilai r-kuasa dua. Hubungan antara piksel mangga dan berat mangga telah dikaji menggunakan regresi linear sebagai pendekatan statistik. Akibatnya, metodologi untuk penilaian berat mangga menggunakan pemprosesan dan analisis imej adalah praktikal, berdaya maju dan berjaya.

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### TABLE OF CONTENTS

		PAGE
DECI	LARATION	
APPR	ROVAL	
DEDI	CATIONS	
ABST	TRACT	i
ABST	`RAK	ii
ACKI	NOWLEDGEMENTS	iii
TABI	LE OF CONTENTS	i
LIST	OF TABLES	iv
	OF FIGURES	v
	OF SYMBOLS	viii
	OF ABBREVIATIONS	ix
LIST	OF APPENDICES	X
	PTER 1 INTRODUCTION	11
1.1	Background	11
1.2 1.3	Problem Statement TI TEKNIKAL MALAYSIA MELAKA	12
1.3	Project Objective Scope of Project	13 13
		13
	PTER 2 LITERATURE REVIEW	15
2.1	Introduction	15
2.2	Image Acquisition and Preprocessing 2.2.1 Digital Camera	15 15
	2.2.2 Color camera	18
	2.2.3 Camera with a video adapter	19
2.3	Image Segmentation	20
	2.3.1 Thresholding method	20
	2.3.2 Interpretation of the form of an image histogram	24
	2.3.3 The traditional method for image enhancement, de-noising, and edge	
	detection	26
	2.3.4 Image blurring	28
	<ul><li>2.3.5 Curve fitting</li><li>2.3.6 Grayscale method</li></ul>	29 30
2.4	Weight Estimation	32
	2.4.1 Estimate weight using volume estimation	32

	<ul><li>2.4.2 Estimating weight using linear regression analysis</li><li>2.4.3 Estimate weight using neural network</li></ul>	35 36
	2.4.4 Estimating weight using visible imaging	38
2.5	Comparison of Previous Studies	39
2.6	Summary	43
	PTER 3	45
3.1	Introduction	45
3.2	Project Implementation	45
3.3	Project Development	50
	3.3.1 Sampling	50
	3.3.2 Image acquisition and pre-processing	51
	3.3.3 Image processing and analysis	52
	3.3.3.1 Grayscale image	53
	3.3.3.2 Graythresh image 3.3.3.3 Remove noise	53 54
	3.3.3.4 Finding area	57
	3.3.3.5 Finding perimeter	58
	3.3.4 Statistical analysis	59
	3.3.4.1 Correlation coefficient	59
	3.3.4.2 Linear regression	60
	3.3.5 Performance evaluation	66
3.4	Component Involves	69
	3.4.1 Web camera with stand	70
	3.4.2 Mini studio box	71
	3.4.3 Scales	72
	3.4.4 Mango	73
3.5	Summary	74
	TER 4 NIVERRESULTS AND DISCUSSIONS IA MELAKA	75
4.1	Introduction	75
4.2	Image Analysis	75 70
4.3	System Implementation	78 70
	4.3.1 Training session	79
4.4	4.3.2 Testing session Data Analysis	80 81
4.4	4.4.1 Area	82
	4.4.2 Perimeter	85
4.5	Discussion	88
4.6	Performance Evaluation	89
4.7	Summary	91
CHAP	PTER 5 CONCLUSION AND RECOMMENDATIONS	92
5.1	Introduction	92
	Conclusion	92
	Recommendation for Future Project Improvement	93
5.4	Commercialization Relevancy	93
REFE	RENCES	95

APPENDICES 97



### LIST OF TABLES

TABLE TITL	Æ	PAGE
Table 2.1 Reference values for weight-based f	fruit classification	34
Table 2.2 Chosen parameter.		37
Table 2.3 Comparison of previous studies.		39
Table 3.1: Example 1 calculating the correlation	on coefficient	62
Table 3.2: Example 1 Calculating the Equation	n of Regression Line	63
Table 3.3: Example 2 calculating the correlation	on coefficient	63
Table 3.4: Example 2 Calculating the Equation	n of Regression Line	64
Table 4.1: Mango area, actual, and estimated set 84		
Table 4.2: Mango perimeter, actual, and es reference set	stimated weight of mangoes in the	87
Table 4.3: Weight estimation equation and the	ey are Respective <b>r2</b> values	89
LINIVERSITI TEKNIKAL	MALAYSIA MELAKA	

### LIST OF FIGURES

FIGURE TITLE	PAGE
Figure 2.1 The image acquisition setup	16
Figure 2.2 Digital photos of Mango	17
Figure 2.3 The setup used to capture the image	17
Figure 2.4 Schematic diagram of the machine vision system for image acquisition	18
Figure 2.5 Schematic diagram of the machine vision system for image acquisition for image acquisition	19
Figure 2.6 Front diffuse illumination scheme	20
Figure 2.7 Binarization on H, S, and V channel images and result comparison	24
Figure 2.8 Bimodal Histogram	25
Figure 2.9 Relation between CB and Cr channel in the YCbCr color space	26
Figure 2.10 Original image and CB channel information	27
Figure 2.11 Histogram of a threshold segmentation and the resulting image	28
Figure 2.12 Top profile of mango	29
Figure 2.13 Threshold with no blurring	29
Figure 2.14 Threshold with blurring window	29
Figure 2.15 Curve fitting the mango boundary with two polynomial equation	30
Figure 2.16 (a) Original image (b) Grey-scale image (c) Binary image (d) Edge detection (e) Contour filling (f) Filtere image	31
Figure 2.17 Pixel value for the segmentation image	31
Figure 2.18 A cross-section of the fruit	32
Figure 2.19 Ratio analysis at different points	33
Figure 2.20 Major and minor axis on fruit image	34
Figure 2.21 Mango dimension measurement	38
Figure 3.1 Flowchart of project implementation	49

Figure 3.2: Image acquisition system	52
Figure 3.3: RGB image or input image	52
Figure 3.4: Binarize and thresholded image	54
Figure 3.5: Fill holes image	55
Figure 3.6: Remove blob on the border-image	56
Figure 3.7: Remove small blob	56
Figure 3.8: Output image	56
Figure 3.9: Number of white region area	57
Figure 3.10: The properties of region prop	58
Figure 3.11: Number of perimeters	59
Figure 3.12: Sketched regression line example 1	65
Figure 3.13 Sketched regression line example 2	65
Figure 3.14: Comparison linear regression line between example 1 and 2	66
Figure 3.15: Flowchart of the overall system	67
Figure 3.16: Flowchart of image acquisition	68
Figure 3.17: Flowchart of image processing MALAYSIA MELAKA	68
Figure 3.18: Statistical Analysis	69
Figure 3.19: ANBIUX USB webcam	70
Figure 3.20: Webcam stand	71
Figure 3.21: Mini studio box	71
Figure 3.22: Mini scale	72
Figure 3.23: Mango	73
Figure 4.1 Original image for web camera	77
Figure 4.2 Adjust YCbCr	77
Figure 4.3 Adjust HSV value	77
Figure 4.4 Algorithm to find number of black pixels Mango	78

Figure 4.5: Image analysis result	7/8
Figure 4.6: Training application interface in MATLAB	80
Figure 4.7: Testing application interface in MATLAB	81
Figure 4.8: The dependency between area and actual weight	84
Figure 4.9: 95% confident interval for intercept and slope of the linear model of the area	85
Figure 4.10: Calculating all the area data using excel	85
Figure 4.11: The dependency between perimeter and actual weight	
Figure 4.12: 95% confident interval for intercept and slope of the linear model of the perimeter	88
Figure 4.13 Calculating all the perimeter data using excel	88
Figure 4.14: Actual weight and estimated weight method 1: area	90
Figure 4.15: Actual weight and estimated weight method 2: perimeter	90
UNIVERSITI TEKNIKAL MALAYSIA MELAKA	

### LIST OF SYMBOLS

δ	_	Voltage angle
t	_	ton
%	_	Percentage
cm	_	Centimeter
mm	-	Millimeter
S	_	Seconds
dB	-	Decibel
T	-	Threshold
ρ	-	Density
v	-	Volume
m	-	Mass
Df	-	Depth factor
g	-	Gram
$cm^3$	- 10	Cubic centimeter
$r^2$	4	Coefficient of Determination
r	<b>5</b> /-	Correlation Coefficient
L	- E	Lenght
W	H -	Width
Н	F.	Height
ŷ	100	Predicted Value
m	_9/	Slope
b	143	y-intercept
$\mu$	مالالك	او بوتر سنتي بكنيكل مليMean
S	-	Sample Standard deviation
$\sum_{i} i$	UNĪVE	EKSITI TEKNIKAL MALAYSIA MELAKA
$\epsilon$	_	Error
$\epsilon_{rms}$	_	Root Mean Square
$\varphi$	-	Accuracy
•		

### LIST OF ABBREVIATIONS

VVoltage

BDP **Bachelor Degree Project** 

MATLAB Matrix Laboratory

Ts. Technologist

Federal Agricultural Marketing Authority of Malaysia **FAMA** 

Universal Serial Bus USB

Joint Photographic Experts Group JРG

XGA **Extended Graphics Array** Charge-coupled device **CCD** 

PLPolarized light IC **Integrated Circuit BMP** Basic Metabolic Panel

Institute of Electrical and Electronics Engineers IEEE

RGB Red, Green, Blue

HSI Hue, Saturation, Intensity

**CMYK** Cyan, Megenta, Yellow and Black

**HSV** Hue, Saturation, and Value

Luminance and Chroma components YCbCr.

2D 2 Dimensional 3D 3 Dimensional

MLP Multilayer perceptron Support Vector Machine **SVM** 

Convolute Neural Network **CNN** 

ANN Artificial Neural Network

Root Mean Square Error **RMSE** Frames per second AL MALAYSIA MELAKA

**FPS** 

LED **Light Emitting Diode** 

**Application** App

### LIST OF APPENDICES

APPENDIX	TITLE	PAGE
Appendix A	(Gantt Chart) BDP 2	97
Appendix B	(MATLAB Application Interface)	98
Appendix C	(Image Acquisition Setup and Demonstration)	99
Appendix D	(Output Image and Result)	101



### **CHAPTER 1**

### INTRODUCTION

### 1.1 Background

One of the essential fruits produced in Malaysia, mangoes were valued at 126,000 t in 2002, representing a 12% share of total exports. Post-harvest handling is a crucial step in the fruit production process since it helps maintain its quality. Cleaning, grading, and packing are all part of the post-harvest handling process. Fruit grading is necessary for identifying the quality of the fruit as well as its cost. Many local farmers are still heavily reliant on human labor to grade their fruits, following the grade recommendations set out by the Federal Agricultural Marketing Authority of Malaysia (FAMA). The fruits are assessed according to their size, maturity index, and the presence of exterior flaws (Teoh and Syaifudin, 2007).

Computer vision systems may be utilized for automated fruit inspection and grading and other applications. These systems have been extensively employed in the food and agricultural industries for review and evaluation because they offer an appropriate speedy, economical, consistent, and objective assessment method (Sun, 2000). According to (Kanali et al, 1998) the use of machine vision to automate the inspection of products not only results in labor savings, but it also enhances the objectivity of the inspection process. Advancements have spurred research on constructing a system to assess the quality of various and processed meals in hardware and software for digital image processing and analysis during the last decade. (Gerrard et al. 1996; Locht et al. 1997, 1996).

Image processing and analysis are at the heart of computer vision, allowing it to perform the necessary categorization and measurement tasks (Krutz et al, 2000). (Nagata et al, 1997) The use of image processing and analysis to sort fresh strawberries based on size and shape was examined. The findings revealed that the system could accurately sort 600 strawberries into three classes of condition and five classes of length with 94–98 percent accuracy.

Size is one of the significant parameters that consumer identifies together with the quality of fruits (Aleixos et al, 2002). According to the Florida Agricultural Marketing Association, weight may affect the size of tropical fruits (i.e. Mango, pineapple, watermelon, papaya and starfruit). Manual weighing of fruits to determine their size is time-consuming, labor-intensive, and expensive. Instead of relying on weight to determine fruit size, image processing and analysis methods in computer vision systems may be utilized to make an automated determination of fruit size. The goal of this project was to estimate the weight of the mangoes and estimate the error to assess the ability of the linear model generated in the linear regression analysis to evaluate the importance of the mangoes, which was the primary purpose.

### 1.2 Problem Statement

The size of the Mango is one of the most critical factors that the customer has identified as being associated with mango grade. According to the Malaysian Federal Agricultural Marketing Authority (FAMA), the size of Mango is defined by its weight. It is, however, difficult to determine the grade of mango fruits because of the difficulty in collecting the fruit at the proper size. Manual weighing of fruits to determine their size is time-consuming, labor-intensive, and expensive.

Furthermore, a computer vision system can automate fruit inspection and grading to achieve the Mango fruit's desired weight. The mango pictures collected from the image capture equipment will be processed and analyzed using the MATLAB program in the computer environment. It is the software's job once image processing and analyzing operations have been completed to determine how many pixels are included inside the mango area of the acquired picture. With the help of the linear regression approach, we will discover a link between mango pixels and mango weights.

### 1.3 Project Objective

In this study, the primary objectives were to investigate several image processing and analysis methodologies for determining the weight of mangoes, specifically:

- a) To develop a vision-based system for mangoes' weight estimation.
- b) To select an appropriate feature from the image analysis that can be used as a parameter in the regression technique.
- c) To perform testing and analysis in term of the performance of the completed

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### 1.4 Scope of Project

In this study, a primitive machine vision system was used to improve mango fruit grading in the country of Malaysia. The methodology is straightforward compared to other state-of-the-art technologies now accessible, but it has limits in analysis and practical use.

a) Image processing and analysis were performed by filtering and thresholding processes using MATLAB software.

- b) Image is captured using a digital camera and transferred to MATLAB software.
- c) The position of the Mango must remain in a static setting during the capturing process.
- d) Just one Mango should be placed at a time.



### **CHAPTER 2**

### LITERATURE REVIEW

### 2.1 Introduction

An overview of estimating the weight of mangoes using image processing and analysis techniques was covered in Chapter 2. This project aims to create image processing and analysis tools to estimate the weight of mangoes, which will assist local producers in classifying the fruits following the FAMA's suggested classification. It was also done in this project to compare and contrast the advantages and disadvantages of past research findings. It discusses the source and supports the claim with facts from similar research or study known as the literature review.

### 2.2 Image Acquisition and Preprocessing

Image Acquisition and Preprocessing for Machine Vision Systems is a detailed, in-UNIVERSITITEKNIKAL MALAYSIA MELAKA

depth guide text covering any element of image acquisition and preprocessing, from scene
lighting to image shaping optics, from CCD and CMOS image capture to image
transformation

### 2.2.1 Digital Camera

(Khairul Adilah and Sharifah Lailee, 2016) Indicate that the method of capturing photographs using a digital camera is known as image acquisition. For this experiment, the distance between the camera and the fruit was adjusted at 30 cm. Until being processed into grayscale files, the captured images were saved in RGB color format. According to (Ganiron,