Improving Skin Cancer (Melanoma) Detection: New Method

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Abstract

Melanoma, the deadliest form of skin cancer, must be diagnosed early for effective treatment. Rough pigment network and qualities are important signs for melanoma diagnosis using pathologist images. The main focus of this thesis is to improve skin cancer (Melanoma) detection through introducing novel image processing approach for a computer-aided system based on pigment network and elements detection on pathology images. It is important to propose an automated system for differentiating between melanocytic nevi and malignant melanoma. This thesis describes a novel image processing approach for computer-aided pigment network and elements detection on dermoscopy / pathology images. The proposed methods provide meaningful ideas of structures, and extract features for melanoma detection. Additionally, the thesis presents efforts towards prevention of melanoma, by developing a smart system to locate pigment networks.

The thesis aims to cover a complete theoretical model for simulating the processes that takes place when a human interprets an image generated by the eye, through designing a reliable system, that can provide a screening method that "filters" lesions and melanoma in a general practice. The proposed system is to be used with a standard PC with input from a high quality digital camera, dermoscopy / microscopy slides or any other suitable hardware sources. This system analyses the structure of a mole / skin defects, detects cancer, identifies features, makes a decision and provides the result.

The result of the proposed system shows that the Skin Cancer (Melanoma) Detection strategy which uses SVM performs reasonably satisfactorily (accuracy 77.44%, sensitivity 83.60 %, and specify 70.67%). Furthermore, the SVM based wavelet Gabor (SVM-WLG) performs better than the SVM (81.61%, 88.48%, and 74.51 % accuracy, sensitivity, and specify respectively). However, the Swarm-based SVM (SSVM) performs better than the other two algorithms, with average for accuracy, sensitivity, specificity of 87.13%, 94.1% and 80.22%, respectively.

Certificate of Authorship / Originality

I certify that the work in this thesis has not previously been submitted for a degree nor has it been submitted as part of requirements for a degree except as fully acknowledged within the text.

I also certify that the thesis has been written by me. Any help that I have received in my research work and the preparation of the thesis itself has been acknowledged. In addition, I certify that all information sources and literature used are indicated in the thesis.

Mohamed Khaled Abu Mahmoud

November, 2014



To the soul of my father and my mother For encouraging me and pounding me with my education

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Table of Contents

To the soul of my father and my mother3
For encouraging me and pounding me with my education3
Acknowledgments4
Table of Contents 5
Extended Table of Contents7
List of Figures12
List of Tables
CHAPTER 1 19
Introduction19
CHAPTER 2
Human Skin Biology and Cancer
CHAPTER 3 53
Image Processing and Automatic Skin Cancer Diagnosing
CHAPTER 4
Pre-processing and Segmentation
CHAPTER 5 122
Image Representation and Analysis122

CHAPTER 6	150
Experimental Results and Discussion	150
CHAPTER 7	185
Conclusion and Future Work	185
Appendices	189
Bibliography	220

Extended Table of Contents

Table of Contents

To the soul of my father and my mother	3
For encouraging me and pounding me with my educ	cation3
Acknowledgments	4
Table of Contents	5
Extended Table of Contents	7
List of Figures	12
List of Tables	
CHAPTER 1	19
Introduction	19
1.1. Research Aims	20
1.2. Problem Description	20
1.3. Objectives	21
1.4. Methodology	22
1.5. Contribution of the Doctoral Research	23
1.6. Structure and Summary of the Dissertation	27
1.7. Publications Presented During the Doctoral Research	29
CHAPTER 2	32
Human Skin Biology and Cancer	32
2.1. Human Skin Biology	32
2.2. The Cell (Structure and Function)	32
2.2.1. Cell structure	
2.2.2. Cell Cycle and Replication:	33

2.3. Haematoxylin and Eosin (H & E)	
2.4. Skin Cell Structure	
2.4.1. Protection	36
2.4.2. Sensation	36
2.4.3. Thermoregulation	36
2.5. The Layers of the Skin	
2.5.1. Epidermis	37
2.5.2. Dermis	37
2.5.3. Subcutaneous/hypodermis	37
2.6. Melanocyte	
2.7. Physical Properties of Human Skin	
2.8. Human Skin	
2.8.1. Types of Skin	40
2.8.2. Layers of Human Skin	40
2.8.3. Cancer Types	41
2.8.4. Types of Skin Cancer	41
2.8.5. Health Behavior of Cancer Survivors	49
,	
CHAPTER 3	
	53
CHAPTER 3	53 nosing
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag	53 nosing 53
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag	53 nosing 53
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag	53 nosing 53 53
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye	nosing 53 53 53 53 53 53 53 53
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye 3.2.1. Human Visual Perception	nosing 53 53 53 53 53 53 53 53
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye 3.2.1. Human Visual Perception 3.2.2. The Human Eye	nosing 53 53 53 53 53 54 54
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye 3.2.1. Human Visual Perception 3.2.2. The Human Eye 3.2.3. Structure of the Eye	nosing 53 53 53 53 53 54 54 54 55
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye 3.2.1. Human Visual Perception 3.2.2. The Human Eye 3.2.3. Structure of the Eye 3.3. Image Processing & Human skin Imaging	nosing 53 53 53 53 54 54 54 55 55
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye. 3.2.1. Human Visual Perception 3.2.2. The Human Eye. 3.2.3. Structure of the Eye. 3.3.1. Applications of Image Processing	nosing 53 53 53 53 53 54 54 54 55 56 56 58
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye 3.2.1. Human Visual Perception 3.2.2. The Human Eye 3.2.3. Structure of the Eye 3.3.1. Applications of Image Processing 3.3.2. Digital Image Acquisition	nosing 53 53 53 53 53 54 54 54 54 55 56 56 58 59
CHAPTER 3 Image Processing and Automatic Skin Cancer Diag 3.1. Introduction 3.2. Vision System & Structure of the Human Eye 3.2.1. Human Visual Perception 3.2.2. The Human Eye 3.2.3. Structure of the Eye 3.3.1. Applications of Image Processing 3.3.2. Digital Image Acquisition 3.3.3. Digital Camera	nosing 53 53 53 53 53 54 54 54 55 56 56 58 59 60

CHAPTER 4	
Pre-processing and Segmentation	
4.1. Introduction	
4.1.1. Digital Image Representation	93
4.1.2. Image Types	93
4.1.3. General approach of developing a Computer Aided Desig system	
4.2. Pre-Processing Stage	
4.2.1. Wiener Filter:	<i>98</i>
4.2.2. Gabor Filter	100
4.2.3 Median filter:	102
4.2.4. Adaptive Median Filter (AMF):	103
4.3. Image preprocessing: enhancement	
4.3.1. Image enhancement	105
4.3.2. Intensity transformation	106
4.3.3. Histogram Processing	106
4.3.4. Histogram Equalization	108
4.3.5. Special filtering	111
4.4. Image Segmentation	
4.4.1. Edge-based Image Segmentation	114
4.4.2. Edge Detection Operations	114
4.4.3. Thresholding	115
4.4.4. Global Thresholding	116
4.4.5. Optimal Global Thresholding	118
4.4.6. Pixel Classification through Clustering	120
CHAPTER 5	122
Image Representation and Analysis	122
5.1. Introduction	
5.2. Machine Learning	
5.3. Main Concepts	
5.4. Intelligent Image Features Extraction	

5.4.1. Texture:
5.4.2. Texture features:
5.4.3. Grey-Level Co-occurrence Matrix and Features
5.4.4. Gabor filter feature extraction128
5.4.5. Wavelet decomposition of an image129
5.4.6. Discrete Wavelet Transform130
5.4.7. Classifier distance metric equation
5.5. Feature Extraction and Representation131
5.5.1. Methodology used in feature extraction
5.6. Classification and Feature selection
5.6.1. Feature Selection (FS)
5.6.2. Unbalanced data sets
5.7. SVM for Classification and Regression
5.7.1. Linear support vector machine137
5.7.2. Nonlinear support vector machine
5.7.3. Support Vector Regression143
5.8. Swarm based Support Vector Machine for Melanoma Detection 144
5.8.1. Development of a Melanoma Detection based on the Swarm based Support Vector machine144
5.8.2. Optimization of SVM parameters using PSO145
5.8.3. Fitness function for the optimization
CHAPTER 6 150
Experimental Results and Discussion150
6.1. Introduction
6.1.1. Experiment 1
6.1.2. Experiment 2
6.1.3. Experiment 3
6.1.4. Experiment 4171
6.1.5. Experiment 5174
6.1.6. Experiment 6176
CHAPTER 7

Conclusion and Future Work	185
7.1. Conclusion	185
7.2. Future Research	187
Appendices	189
A. Appendix A: Glossary of Pathology Terminology	189
B. Appendix B. Glossary of Pathology Terminology	196
C. Appendix C: Glossary of Machine Learning and Computer Vision Terminology	197
D. Appendix D: Margin between two hyper planes	203
E. Appendix E: Lagrangian dual optimization	204
F. Appendix F: Soft-margin nonlinear support vector machine	206
G. Appendix G: Sequential minimal optimization (SMO) for SVM	208
H. Appendix H: Glossary and Abbreviations	211
Bibliography	220

List of Figures

Figure 2.1 shows, Cells structure; source: www.getting-in.com
Figure 2.2 The eukaryotic cell cycle, G Phases: growing, S phase: synthesis/ DNA replication, G2 Phase: Growing and preparing for Mitosis, M mitosis
Figure 2.3 The cell cycle; G0 terminally differentiated cell G1 gap phase 1 G2 gap phase 2 M mitotic phase S synthesis phase
Source: Wheatear's Functional Histology: A Text and Colour Atlas, 5th Ed35
Figure 2.4 H&E-stained section of skin. High magnification (40xs) view. Haematoxylin stains the nuclei of cells blue to bluish-purple, and eosin stains other cellular elements in the tissues from pink to red [43]
Figure 2.5 Displayed Cross Section & the main layers of Human Skin
Source: nurrashidah2204.blogspot.com
Figure 2.6. Skin Color Distribution around the World
Source: www.gbhealthwatch.com
Figure 2.7. a) Cross section of human skin, b) Structure of Epidermis
Figure 2.8 . The Layers of Skin - the Epidermis and Dermis (At the top, the close up shows a squamous cell, basal cell, and melanocyte)
Figure2.9- a) Basal Cell Carcinoma (BCC), b) Squamous Cell Carcinoma (SCC)42
Figure 2.10 – Melanoma
Figure 2.11: Malignant Melanoma, World Age-Standardised Incidence Rates, World Regions, 2008 Estimates
Figure 2.12. Lentigomaligna Melanoma
Source: melanomaknowmore.com
Figure 2.13. Superficial Spreading Melanoma
Source: melanomaknowmore.com
Figure 2.14. Nodular Melanoma
Source: melanomaknowmore.com
Figure 2.15. Stages of Cancer Development and Metastasis
Source: http://www.cancervic.org.au/about-cancer/advanced-cancer
Figure 2.16: Five-year relative survival from selected cancers by remoteness area, Australia. 2006-2010
Figure 3.1 - Show the function of the eye
Source: http://cdn2.hubspot.net/hub/60407/file-350482438- jpg/images/Master_Eye_logo_short_tag_line_revised_10-13-2013-resized-600.jpg55
Figure 3.2 - Distance of vision

Source:hyperphysics.phy-astr.gsu.edu/hbase/vision/accom.html55
Figure 3.3 - Structure of the Eye
Source:drpion.be/en/bouw-van-het-oog.htm
Figure 3.4 Focal length defined [When parallel rays of light strike a lens focused at infinity, they converge to a point called the focal point. The focal length of the lens is then defined as the distance from the middle of the lens to its focal point.]
Figure 3.5. Displayed, similarities of overall design in the principal features of an optical microscope, a transmission electron microscope and a scanning electron microscope63
Source: www.nslc.wustl.edu63
Figure 3.6 . Principle of CSLM. Note the same optical path is used for the detector and the source. Optics are used to direct the light towards the detector
Figure 3.7. Two-step procedure for the classification of pigmented skin lesions. Adapted from Argenziano.[122]
Figure 3.8 . Algorithm for the determination of melanocytic versus non melanocytic lesions according to the proposition of the Board of the Consensus Netmeeting. Adapted from Argenziano.[122]
Figure 3.9 . A, Macroscopic picture of a superficial spreading malignant melanoma (Breslow thickness 0.52 mm; Clark level II). B, Dermoscopy of A shows (atypical) pigment network and branched streaks and can therefore be considered a melanocytic lesion
Figure 3.10. A, Macroscopic picture of a blue nevus. B, Dermoscopy of A shows steel- blue areas (no pigment network, no aggregated globules, and no branched streaks)67
Figure 3.11. A, Macroscopic picture of a seborrheic keratosis. B, Dermoscopy of A shows comedolike openings (a), multiple milia-like cysts (b), and fissures (c)
Figure 3.12. A, Macroscopic picture of a seborrheic keratosis. B, Dermoscopy of A shows comedolike openings and multiple milia-like cysts
Figure 3.13 . A, Macroscopic picture of a basal cell carcinoma. B, Dermoscopy of A shows maple lifelike areas, ovoid nests, and arborized telangiectasia
Figure 3.14. A, Macroscopic picture of a basal cell carcinoma. B, Dermoscopy of A shows multiple spoke wheel areas
Figure 3.15. A, Macroscopic picture of an angioma. B, Dermoscopy of A shows red lagoons
Figure 3.16. Asymmetry Border Color Diameter Evolution (ABCD-E) rule for Diagnosis of Melanoma
Source: www.webmd.com
Figure 3.17 - Normal moles and Melanomas [Benign and Malignant]
Source: www.skinbychar.com70
Figure 3.18 : Showed: a) Melanoma in skin biopsy with H&E Stain ("This case may represent superficial spreading melanoma."), b) Lymph node with almost complete replacement by metastatic melanoma. The brown pigment is focal deposition of melanin.

13

Figure 3.19. Diagnosis of Melanoma using Three-point Checklist72
Figure 3.20 - Example of pigmented skin lesion. Left: Traditional imaging technique. Right: Dermoscopy imaging technique
Figure 3.21. (a) Macroscopic Image of a Lesion (b) Dermatoscopic Image of the same Lesion. Source: www.jle.com
Figure 3.22 - Show the results of skin lesion boundary tracing algorithm, my data experiment shown [a] Image center of mass, [b] Image process
Figure 3.23 - From left to right: the original image (source image from: http://ijplugins.sourceforge.net/plugins/clustering/), clustered image using 2 clusters (poor), clustered image using 3 clusters (close but one key cluster is missing), clustered image using 4 clusters (optimal), and clustered image using 10 clusters (artifacts are noticeable)
Figure 3.24- Structure of fuzzy image processing
Figure 3.25- Steps of fuzzy image processing
Figure 3.26 . Manual Segmentation by trained pathologists using the 'Aperio Image Scope' software [156]
Figure 3.27 Showed the Radial growth phase melanoma
Figure 3.28 showed the Vertical growth phase melanoma
Figure 3.29 Microscopic satellites, (a). Shows neoplastic group is discontinuous (arrow) from the overlying vertical growth phase component. Shows the melanocytes must have a malignant cytomorphology (b)
Figure 3.30 Regression: Regression of over 75% of the tumor volume of a melanoma is considered a bad prognostic sign
Figure 3.31 : a) An image of a lesion under clinical view (naked eye). b) Shows the same lesion under a dermoscopy with oil immersion
Figure 3.32: a) Colors allow the physician, to some extent, to draw conclusions about the localization of pigmented cells within the skin. Black and brown indicate pigmentation in the epidermis, while grey and blue correspond to pigmented cells within the superficial and deep dermis, respectively [185]
Figure 3.33 : Figures (a, b, c, d, and e) show analogue dermoscopy. All of them, except (a), are attachable to digital cameras to function as digital dermoscopy. Figures (f, g, and h) show DinoLite, Handy scope, and Dermoscopy which are modern digital dermoscopy88
Figure 3.34. The Solar Scan instrument: (a) global appearance, (b) camera, (c) user interface [192]. Source: www.medgadget.com
Figure 4.1 Showed: Four stages CAD system for skin lesions
Figure 4.2 true color Pathological Digital Image
Figure 4.4, calculating the median value of a pixel neighborhood102
Figure 4.5. The basic structure of the MACWM filter
Figure 4.6. Shows the output image of Adaptive median Filtering104
Figure 4.7 Showed: (a) grayscale image, (b) Histogram out107
Figure 4.8. Showed: intensity image using

Figure 4.9. Sobel Edge Detection Image 114
Figure 4.10. Binary Output Image Otsu's Method[114
Figure 4.11, selected thresholds in valleys between peaks
Figure 4.12, obtaining the best possible separation measure for two classes
Figure 4.13, selecting a threshold by visually analyzing a bimodal histogram116
(Principle of histogram peak separation)
Figure 5.1 displayed the main categories: texture elements, regularity, randomness, directionality and regularity. 126
Figure 5.2 , showed, (a) images $(5x \ 5)$ matrix with 3 grey levels 0, 1, and 2. (b) The co- occurrence matrix for d= $(1, 1)$
Figure 5.3, displayed the response from convolving image sample with filter129
Figure 5.4 , (a) Wavelet decomposition of an image (b) Block diagram of the decomposition of an image
Figure 5.5 displayed a block diagram of the discrete wavelet transforms (DWT – Gabor approach)
Figure 5.6, the three principal approaches of feature Mammographic e selection. The shades show the components used by the three approaches: filters, wrappers and embedded methods
Figure 5.7: Two-out-of-many separating lines; (a) with smaller margin and (b) with larger margin
Figure 5.8: Margin m between two supporting hyperplanes
Figure 5.9: Illustration of mapping using a transform Ψ : $\mathbb{R}^2 \to \mathbb{R}^3$ 140
Figure 5.10 : Introducing slack variable ξ in soft-margin SVM
Figure 5.10 : Introducing slack variable ξ in soft-margin SVM
Figure 5.10 : Introducing slack variable ξ in soft-margin SVM
Figure 5.10: Introducing slack variable ξ in soft-margin SVM
Figure 5.10: Introducing slack variable ξ in soft-margin SVM
Figure 5.10: Introducing slack variable ξ in soft-margin SVM
Figure 5.10: Introducing slack variable ξ in soft-margin SVM.142Figure 5.11: Loss Functions.144Figure 5.12: Melanoma detection using swarm based support vector machine.145Figure 5.13: The pseudo of the PSO for the SVM parameter optimization146Figure 5.14: The particles of the PSO.147Figure. 6.2 Skin Cancer Image, (a) Original RGB true colour image153
Figure 5.10: Introducing slack variable ξ in soft-margin SVM. 142 Figure 5.11: Loss Functions. 144 Figure 5.12: Melanoma detection using swarm based support vector machine. 145 Figure 5.13: The pseudo of the PSO for the SVM parameter optimization 146 Figure 5.14: The particles of the PSO. 147 Figure 6.2 Skin Cancer Image, (a) Original RGB true colour image 153 Figure 6.3. The basic structure of the MACWM filter 154 Figure 6.4. A snake with traditional potential forces cannot move into the concave 157
Figure 5.10: Introducing slack variable ξ in soft-margin SVM142Figure 5.11: Loss Functions144Figure 5.12: Melanoma detection using swarm based support vector machine.145Figure 5.13: The pseudo of the PSO for the SVM parameter optimization146Figure 5.14: The particles of the PSO147Figure 6.2 Skin Cancer Image, (a) Original RGB true colour image153Figure 6.3. The basic structure of the MACWM filter154Figure 6.4. A snake with traditional potential forces cannot move into the concave boundary region.156
Figure 5.10: Introducing slack variable ξ in soft-margin SVM
Figure 5.10: Introducing slack variable ξ in soft-margin SVM.142Figure 5.11: Loss Functions.144Figure 5.12: Melanoma detection using swarm based support vector machine.145Figure 5.13: The pseudo of the PSO for the SVM parameter optimization146Figure 5.14: The particles of the PSO.147Figure 6.2 Skin Cancer Image, (a) Original RGB true colour image153Figure 6.3. The basic structure of the MACWM filter154Figure 6.4. A snake with traditional potential forces cannot move into the concave boundary region.156Figure 6.5 A snake with GVF external forces moves into the concave boundary region.157Figure. 6.6. Skin Cancer greyscale image showing: (a) Wiener2 filter has removed the spots effectively (b) noisy image filtered by the median filter157
Figure 5.10: Introducing slack variable ξ in soft-margin SVM.142Figure 5.11: Loss Functions.144Figure 5.12: Melanoma detection using swarm based support vector machine.145Figure 5.13: The pseudo of the PSO for the SVM parameter optimization146Figure 5.14: The particles of the PSO147Figure 6.2 Skin Cancer Image, (a) Original RGB true colour image153Figure 6.3. The basic structure of the MACWM filter154Figure 6.4. A snake with traditional potential forces cannot move into the concave boundary region.156Figure 6.5 A snake with GVF external forces moves into the concave boundary region.157Figure. 6.6. Skin Cancer greyscale image showing: (a) Wiener2 filter has removed the spots effectively (b) noisy image filtered by the median filter158

Root mean square error160
Figure 6.9 . Result of segmentation by threshold with GHE (bottom right) and LHE (bottom left), Top: shows histogram Local and Global results
Figure 6.10.Segmentation by Thresholding (ROI)162
Figure 6.11. Segmentation by SRM163
Figure 6.12. Display of the image and its transform (wavelet coefficients)164
Figure. 6.13: A single curvelet with width $2 - 1$ and length $2 - 12$
Figure 6.14. Rectangular frequency (basic digital) tiling of an image with 5 level curvelet.
Figure 6.15. Curvelet demising image that contains oriented texture and cartoon edges (a) Original (b) Noisy, and (c) Curvelet
Figure. 6.16. The results showed, original image, image after median filter, gray scale image,
Figure. 6.17 A Wavelet Packet decomposition tree
Figure. 6.18 shows the melanoma detection employing SVM (SVMR, SVMP, and SVML); the input is pathology images and the output is melanoma (-1) or benign (+1). 181
Figure A1, showed a physician's hands are seen performing a needle biopsy to determine nature of lump either fluid-filled cyst or solid tumour
[This image was released by the National Cancer Institute, an agency part of the National Institutes of Health, with the ID 1973 (image)]
Figure A2 , the diagnostics showed: Micrograph of a needle aspiration biopsy specimen of a salivary gland showing adenoid cystic carcinoma. (Source: Pap stain. MeSH D044963).
Figure A3, showed a Normal Epidermis and Dermis with Intradermal Nevus 10x-cropped. (Source, Kilbad)
Figure A4 , this image shows a crossection of the vascular tissue in a plant stem. Showed as an example for bright field micrograph. (Source: Wikipedia)
Figure A5, Principle of confocal microscopy
Figure A6 showed the nucleolus is contained within the cell nucleus. (Source: Wikimedia Foundation, Inc.)
Figure A7 displayed the negative of the logarithm (base 10) of the optical Density (OD). (Source: semrock@idexcorp.com)
Figure A8, showing a pathologist examines a tissue section for evidence of cancerous cells while a surgeon observes. (Source: wikipedia.org)
Figure A9, showed: a) A TEM image of the polio virus is 30 nm in size, b) Layout of Optical components in a basic TEM. (Source: Wikimedia)Appendix B: Optical Density of Transmission Microscopy Images
Figure B1. Example optical density image
Figure C1, showed a graph with three connected components

Figure C2, displayed the machine learning and data mining which used to sol	ve problem
in the areas of Classification, Clustering, Regression, Anomaly detection,	Association
rules, Reinforcement learning, Structured prediction, Feature learning, Onlin	ne learning,
Semi-supervised learning, and Grammar induction.	
Figure D1: Margin <i>m</i> between two supporting hyperplanes	

List of Tables

Table 2.1: Breslow's depth 48
Table 2.2 : Survival figures from British Association of Dermatologist Guidelines 200248
Table 2.3 : Observed incidence (2009), and morality (2010) of melanoma of the skin and estimated for 2012. 51
Table 2.4 : Observed incidence (2009), and morality (2010) of non-melanoma of the skin and estimated for 2012
Table 6.1 Sensitivity and Specificity comparison. 152
Table 6.2 Comparison between proposed operators (Sobel, Roberts, Prewitt, Laplacian,Canny and Otsu) and Gradient vector flow (GVF) segmented images.160
Table 6.3 BNN Classification Test with Different Wavelet. 161
Table 6.4. BNN Classification Test for SRM & Thresholding
Table 6.6. Displaying the comparison between wavelet and curvelet features results166
Table 6.7. svm classification for all the dataset (including dermoscopy images) with features extracted from wavelets
Table 6.8. svm classification excluding dermoscopy images with features extracted from wavelets
Table 6.9. svm classification for all the dataset (including dermoscopy images) with features extracted from curvelets
Table 6.10. svm classification excluding dermoscopy images with features extracted from curvelets 170
Table 6.11. sensitivity and specificity statistics [310] 171
Table 6.12. Described sensitivity and specificity
Table 6.13. show result after training or the (svm) network. using (sfs) technique
Table 6.14. the result after training of the (svm) network with using (sfs) technique175
Table 6.15. the result after training of the (svm) network without using (sfs) technique. 175
Table 6.17 The Result after Training of the (SVM) Network, with using (SFS)Technique.
Table 6.18The Result after Training of the (SVM) Network. with using(SVM+WLG+PSO)180
Table 6.19 Described Sensitivity and Specificity [31]