

**AN INTELLIGENT DIAGNOSTIC SYSTEM FOR
MALARIA**

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UNIVERSITI MALAYSIA PERLIS

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**AN INTELLIGENT DIAGNOSTIC SYSTEM FOR
MALARIA**

by

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(1141310612)**

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DECLARATION OF THESIS

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TABLE OF CONTENTS

	PAGE
THESIS DECLARATION	i
ACKNOWLEDGEMENT	ii
TABLE OF CONTENTS	iii
LIST OF TABLES	viii
LIST OF FIGURES	xiii
LIST OF ABBREVIATIONS	xxii
LIST OF SYMBOLS	xxv
ABSTRAK	xxix
ABSTRACT	xxxii
CHAPTER 1: INTRODUCTION	
1.1 Introduction	1
1.2 Problem Statements	5
1.3 Research Objectives	7
1.4 Research Scopes	8
1.5 Thesis Structures	9
CHAPTER 2: LITERATURE REVIEW	
2.1 Introduction	11
2.2 Malaria	12

2.2.1	Malaria Parasite Species and Life-Cycle Stages	13
2.2.2	The Causes, Signs and Symptoms of Malaria	16
2.2.3	Malaria Diagnosis	17
2.2.3.1	Clinical Diagnosis	17
2.2.3.2	Microscopic Diagnosis	18
2.2.3.3	Other Current Techniques to Diagnose Malaria	28
2.3	Digital Image Processing	29
2.3.1	Image Enhancement	31
2.3.1.1	Global Contrast Stretching Technique	33
2.3.1.2	Linear Contrast Stretching Technique	35
2.3.1.3	Applications of Image Enhancement for Malaria Images	36
2.3.2	Image Segmentation	38
2.3.2.1	Clustering	41
2.3.2.2	<i>K</i> -means Clustering Algorithm	44
2.3.2.3	Fuzzy <i>c</i> -Means Clustering Algorithm	45
2.3.2.4	Moving <i>k</i> -Means Clustering Algorithm	46
2.3.2.5	Applications of Image Segmentation for Malaria Images	47
2.4	Intelligent Classifier	55
2.4.1	Artificial Neural Network	55
2.4.2	Online Sequential Extreme Learning Machine	58
2.4.3	Random Forest	59

2.4.4	Applications of Intelligent Classifiers for Malaria Diagnosis	63
2.5	Summary	82

CHAPTER 3: METHODOLOGY

3.1	Introduction	84
3.2	Data Collection	87
3.2.1	Image Acquisition Setup	88
3.2.2	Quality of the Captured Images	89
3.3	Image Processing	93
3.3.1	Contrast Enhancement Techniques	96
3.3.1.1	Modified Global and Modified Linear based on Total Pixel Approach	98
3.3.1.2	Modified Global and Modified Linear based on Pixel Level Approach	103
3.3.1.3	Quantitative Measure of Contrast Enhancement Technique	108
3.3.2	Image Segmentation based on RGB and HSI Colour Models	111
3.3.3	Image Segmentation using Adaptive Segmentation Techniques	112
3.3.3.1	Enhanced k -Means Clustering Algorithm	113
3.3.3.2	Cascaded Enhanced k -Means and Fuzzy c -Means Clustering Algorithms	118
3.3.4	Image Filtering using Median Filter	121
3.3.5	Seeded Region Growing Area Extraction Algorithm	122
3.3.6	Proposed Segmentation Method for Plasmodium	124

	Falciparum and Plasmodium Vivax Images	
3.3.7	Evaluation of Image Segmentation Quality	127
3.4	Feature Extraction for Malaria Images	128
3.5	Malaria Diagnosis using Intelligent Classifier	134
3.5.1	Classification using Multilayered Perceptron Neural Network	137
3.5.2	Classification using Online Sequential Extreme Learning Machine	140
3.5.3	Classification using Random Forest	143
3.6	Data Sample and Evaluation of the Classification System	143
3.7	Summary	145
CHAPTER 4: RESULTS AND DISCUSSIONS		
4.1	Introduction	147
4.2	Analysis of Contrast Enhancement Techniques	148
4.2.1	Results of Contrast Enhancement Techniques on Benchmark Images	149
4.2.2	Results of Contrast Enhancement Techniques on Malaria Images	167
4.3	Analysis of Image Segmentation	185
4.3.1	Comparison between RGB and HSI Colour Models	186
4.3.2	Comparison of Adaptive Segmentation Techniques	199
4.3.3	Comparison of Cascaded Clustering Algorithms	228
4.3.4	Validation of the Proposed Segmentation Method for Plasmodium Falciparum and Plasmodium Vivax Images	252

4.4	Performance Comparison of Different Intelligent Classifiers for Diagnosis of Malaria Parasites	266
4.4.1	Classification Results of Malaria Parasites Detection	266
4.4.2	Classification Results of Malaria Parasites Species	278
4.4.3	Classification Results of Malaria Parasites Life-cycle Stages	280
4.4.4	Classification Results of Multiclass Classification for Malaria Parasites Species and Life-cycle Stages	283
4.4.5	Classification Performances among MLP, LM, OS-ELM and Random Forest	286
4.5	The Proposed Intelligent Diagnostic System for Malaria	290
4.5.1	Classification Performance of the Proposed Intelligent Diagnostic System for Malaria	291
4.5.2	The Graphical User Interface for Classification of Malaria and Non-Malaria Images	294
4.6	Summary	305
CHAPTER 5: CONCLUSIONS AND FUTURE RECOMMENDATIONS		
5.1	Conclusions	308
5.2	Research Contributions	313
5.3	Recommendation for Future Work	314
REFERENCES		316
APPENDIX		333
LIST OF PUBLICATIONS		398

LIST OF TABLES

NO.		PAGE
2.1	Morphology of malaria parasites of <i>P. falciparum</i> and <i>P. vivax</i> species	26
2.2	Summary of several research works on detection of malaria parasites using image processing techniques and different intelligent classifiers	68
2.3	Summary of several research works on classification of malaria parasite species and its life-cycle stages using image processing techniques and different intelligent classifiers	75
3.1	The distributions for the number of blood slides and images of both normal and malaria samples	90
3.2	Parameters for MGCS_TPA and MLCS_TPA techniques	99
3.3	Parameters for MGCS_PLA and MLCS_PLA techniques	104
3.4	The distributions of data during training, validation and testing phases	137
3.5	The representative output node of each task	138
3.6	The representative target of each task	142
4.1	The DSM of HE technique for the six malaria images	181
4.2	The DSM of GCS and MGCS techniques for the six malaria images	181
4.3	The DSM of LCS and MLCS techniques for the six malaria images	182
4.4	The DSM values for the seven contrast enhancement techniques that have been applied on 100 malaria images	185
4.5	Segmentation performances based on sensitivity, specificity and accuracy for the segmented malaria images	197
4.6	Segmentation performance based on sensitivity for the segmented images	223
4.7	Segmentation performance based on specificity for the segmented images	224

4.8	Segmentation performance based on accuracy for the segmented images	225
4.9	Segmentation performances based on sensitivity, specificity and accuracy for the overall 100 images	228
4.10	Segmentation performance based on sensitivity for the segmented malaria images	245
4.11	Segmentation performance based on specificity for the segmented malaria images	246
4.12	Segmentation performance based on accuracy for the segmented malaria images	247
4.13	Segmentation performances based on sensitivity, specificity and accuracy for the overall 100 images	249
4.14	Execution time for different clustering algorithms	251
4.15	Segmentation performances for the final segmented malaria images	263
4.16	Analysis of number of training epochs for classification between malaria parasite and non-malaria parasite using MLP_LM network	268
4.17	Analysis of number of hidden nodes for classification between malaria parasite and non-malaria parasite using MLP_LM network	270
4.18	Performances of classification between malaria parasite and non-malaria parasite using MLP_LM network	272
4.19	Analysis of number of hidden nodes for classification between malaria parasite and non-malaria parasite using OS-ELM	273
4.20	Performances of classification between malaria parasite and non-malaria parasite using OS-ELM	276
4.21	Analysis of number of trees for classification between malaria parasite and non-malaria parasite using random forest	276
4.22	Performances of classification between <i>P. falciparum</i> and <i>P. vivax</i> species using MLP_LM network	279
4.23	Performances of classification between <i>P. falciparum</i> and <i>P. vivax</i> species using OS-ELM	279

4.24	Performances of classification among the four different life-cycle stages of <i>P. falciparum</i> using MLP_LM network	281
4.25	Performances of classification among the four different life-cycle stages of <i>P. vivax</i> using MLP_LM network	281
4.26	Performances of classification among the four different life-cycle stages of <i>P. falciparum</i> using OS-ELM	282
4.27	Performances of classification among the four different life-cycle stages of <i>P. vivax</i> using OS-ELM	282
4.28	Classification performance based on Task 5 using MLP_LM network	284
4.29	Classification performance based on Task 6 using MLP_LM network	284
4.30	Classification performance based on Task 5 using OS-ELM	285
4.31	Classification performance based on Task 6 using OS-ELM	285
4.32	The performance comparison among MLP_LM network, OS-ELM and random forest for classification of six different tasks	287
4.33	Classification performances based on classification of six different tasks using MLP_LM network	288
4.34	Classification performance using the intelligent diagnostic system for malaria based on blood images	293
A	The DSM values of the average RGB components for the seven contrast enhancement techniques that have been applied on 100 malaria images	333
B.1	Analysis of number of training epochs for classification between <i>P. falciparum</i> and <i>P. vivax</i> species using MLP_LM network	336
B.2	Analysis of number of hidden nodes for classification between <i>P. falciparum</i> and <i>P. vivax</i> species using MLP_LM network	337
B.3	Analysis of number of hidden nodes for classification between <i>P. falciparum</i> and <i>P. vivax</i> species using OS-ELM	338
B.4	Analysis of number of trees for classification between <i>P. falciparum</i> and <i>P. vivax</i> species using random forest	340

B.5	Analysis of number of training epochs for classification among PF_Ring, PF_Trophozoite, PF_Schizont and PF_Gametocyte stages using MLP_LM network	341
B.6	Analysis of number of hidden nodes for classification among PF_Ring, PF_Trophozoite, PF_Schizont and PF_Gametocyte stages using MLP_LM network	342
B.7	Analysis of number of training epochs for classification among PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using MLP_LM network	343
B.8	Analysis of number of hidden nodes for classification among PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using MLP_LM network	344
B.9	Analysis of number of hidden nodes for classification among PF_Ring, PF_Trophozoite, PF_Schizont and PF_Gametocyte stages using OS-ELM	345
B.10	Analysis of number of hidden nodes for classification among PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using OS-ELM	347
B.11	Analysis of number of trees for classification among PF_Ring, PF_Trophozoite, PF_Schizont and PF_Gametocyte stages using random forest	349
B.12	Analysis of number of trees for classification among PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using random forest	350
B.13	Analysis of number of training epochs for classification among PF_Ring, PF_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using MLP_LM network	351
B.14	Analysis of number of hidden nodes for classification among PF_Ring, PV_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using MLP_LM network	352
B.15	Analysis of number of training epochs for classification among PF_Ring, PF_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont, PV_Gametocyte and others using MLP_LM network	353

B.16	Analysis of number of hidden nodes for classification among PF_Ring, PF_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont, PV_Gametocyte and others using MLP_LM network	354
B.17	Analysis of number of hidden nodes for classification among PF_Ring, PF_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using OS-ELM	355
B.18	Analysis of number of hidden nodes for classification among PF_Ring, PF_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont, PV_Gametocyte and others using OS-ELM	357
B.19	Analysis of number of trees for classification among PF_Ring, PF_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont and PV_Gametocyte stages using random forest	359
B.20	Analysis of number of trees for classification among PF_Ring, PF_Trophozoite, PF_Schizont, PF_Gametocyte, PV_Ring, PV_Trophozoite, PV_Schizont, PV_Gametocyte and others using random forest	360
D.1	The classification results for non-malaria images	368
D.2	The classification results for <i>P. falciparum</i> (ring stage) images	371
D.3	The classification results for <i>P. falciparum</i> (trophozoite stage) images	374
D.4	The classification results for <i>P. falciparum</i> (schizont stage) images	377
D.5	The classification results for <i>P. falciparum</i> (gametocyte stage) images	381
D.6	The classification results for <i>P. vivax</i> (ring stage) images	384
D.7	The classification results for <i>P. vivax</i> (trophozoite stage) images	387
D.8	The classification results for <i>P. vivax</i> (schizont stage) images	390
D.9	The classification results for <i>P. vivax</i> (gametocyte stage) images	394

LIST OF FIGURES

NO.		PAGE
1.1	The countries with ongoing malaria transmission	2
1.2	The confirmed cases per 1000 population/parasite prevalence	4
1.3	The proportion of cases due to <i>P. falciparum</i>	4
2.1	The life-cycle stages of malaria parasite	15
2.2	Thick and thin blood smears	19
2.3	A sample of malaria image	20
2.4	A sample of malaria image	21
2.5	Different stages of <i>P. falciparum</i> inside the thick blood smear sample	21
2.6	Different stages of <i>P. falciparum</i> inside the thin blood smear sample	22
2.7	Different stages of <i>P. vivax</i> inside the thin blood smear sample	22
2.8	Different stages of <i>P. malariae</i> inside the thin blood smear sample	22
2.9	Different stages of <i>P. ovale</i> inside the thin blood smear sample	23
2.10	Different stages of <i>P. knowlesi</i> inside the thin blood smear sample	23
2.11	Life-cycle stages of <i>P. falciparum</i> based on the thin blood smear. 1: normal RBC, 2-10: ring, 11-18: mature trophozoites, 19-26: schizonts, 27-28: macrogametocytes (female gametocyte) and 29-30: microgametocytes (male gametocyte)	24
2.12	Life-cycle stages of <i>P. vivax</i> based on the thin blood smear. 1: normal RBC, 2-6: ring, 7-18: mature trophozoites, 19-27: schizonts, 28-29: macrogametocytes (female gametocyte) and 30: microgametocyte (male gametocyte)	25
2.13	Examples of minimum and maximum values for each R, G and B colour components of an image	35
2.14	MLP network with two hidden layers	57

2.15	An example of a random forest classifier	61
3.1	The development of an automated intelligent diagnostic system for malaria	85
3.2	Image acquisition system	89
3.3	Different stages of malaria parasites inside the <i>P. falciparum</i> blood sample	91
3.4	Different stages of malaria parasites inside the <i>P. vivax</i> blood sample	92
3.5	Normal blood images	92
3.6	A single captured WBC and platelet in malaria images	92
3.7	Samples of the captured malaria images	93
3.8	The proposed procedures for segmentation of malaria image	95
3.9	An example for obtaining the new minimum and maximum values based on MGCS_TPA and MLCS_TPA techniques ($min=139$, $max=202$, $Nmin=161$ $Nmax=194$)	98
3.10	An example for obtaining the $Nmin$ and $Nmax$ values based on MGCS_TPA and MLCS_TPA techniques on an image with 0.1% density of salt-and-pepper noise	102
3.11	An example for obtaining the new minimum and maximum values based on MGCS_PLA and MLCS_PLA techniques ($min=139$, $max=202$, $Nmin=143$ $Nmax=200$)	104
3.12	An example for obtaining the $Nmin$ and $Nmax$ values based on MGCS_PLA and MLCS_PLA techniques on an image with 0.1% density of salt-and-pepper noise	108
3.13	Original malaria image and its segmented malaria parasite and background regions after performing the manual segmentation	109
3.14	An example of gray level histogram for obtaining the c_j value	117
3.15	The proposed image processing steps for segmentation of malaria images	126
3.16	The structure for the classification of individual segmented object inside the malaria blood sample	136

4.1	The original benchmark images	149
4.2	Results of contrast enhancement techniques for Portofino image	154
4.3	Results of contrast enhancement techniques for Ivy Tree image	156
4.4	Results of contrast enhancement techniques for Salzburg image	158
4.5	Results of contrast enhancement techniques for Sailboat image	160
4.6	Results of contrast enhancement techniques for Boats image	162
4.7	Results of contrast enhancement techniques for Zelda image	164
4.8	Saturation histograms for Zelda image after applying contrast enhancement techniques	166
4.9	The original malaria images	168
4.10	Results of contrast enhancement techniques for Normal Gametocyte image	171
4.11	Results of contrast enhancement techniques for Overexposed Gametocyte image	172
4.12	Results of contrast enhancement techniques for Underexposed Gametocyte image	173
4.13	Results of contrast enhancement techniques for Blurred Ring image	174
4.14	Results of contrast enhancement techniques for Blurred Trophozoite image	175
4.15	Intensity histograms for Blurred Trophozoite image after applying contrast enhancement techniques	177
4.16	Results of contrast enhancement techniques for Blurred Trophozoite image with 0.1% density of salt-and-pepper noise	179
4.17	Original malaria images	186
4.18	Results of images of modified global contrast stretching technique	186
4.19	Different colour components of RGB and HSI colour models that have been extracted from malaria images	188

4.20	Results of images after applying KM clustering on different colour components images	190
4.21	Results of malaria images after the colour of KM clustering image has been retrieved based on MGCS image	191
4.22	Results of malaria images after applying median filter and SRGAE algorithms to clustering image	193
4.23	Original malaria images	194
4.24	Final segmented images after applying the proposed segmentation method	196
4.25	Original malaria images	200
4.26	Results of images of modified global contrast stretching technique	200
4.27	Results of images for Ring_1 after applying the five adaptive segmentation techniques	202
4.28	Results of images for Trophozoite_1 after applying the five adaptive segmentation techniques	203
4.29	Results of images for Schizont_1 after applying the five adaptive segmentation techniques	204
4.30	Results of images for Gametocyte_1 after applying the five adaptive segmentation techniques	205
4.31	Results of images for Ring_1 after the colour of segmented image has been retrieved based on MGCS image	208
4.32	Results of images for Trophozoite_1 after the colour of segmented image has been retrieved based on MGCS image	209
4.33	Results of images for Schizont_1 after the colour of segmented image has been retrieved based on MGCS image	210
4.34	Results of images for Gametocyte_1 after the colour of segmented image has been retrieved based on MGCS image	211
4.35	Results of images for Ring_1 after applying median filter and SRGAE algorithms to the segmented image	212

4.36	Results of images for Trophozoite_1 after applying median filter and SRGAE algorithms to the segmented image	213
4.37	Results of images for Schizont_1 after applying median filter and SRGAE algorithms to the segmented image	214
4.38	Results of images for Gametocyte_1 after applying median filter and SRGAE algorithms to the segmented image	215
4.39	A single segmented malaria parasite based on segmentation into 3 clusters using MKM and EKM clustering	216
4.40	Original malaria images	216
4.41	Results of images of modified global contrast stretching technique	217
4.42	The final segmented images of Trophozoite_2 after applying the proposed segmentation method	218
4.43	The final segmented images of Schizont_2 after applying the proposed segmentation method	219
4.44	The final segmented images of Gametocyte_2 after applying the proposed segmentation method	220
4.45	The final segmented images of Gametocyte_3 after applying the proposed segmentation method	221
4.46	Original malaria images	229
4.47	Results of images of modified global contrast stretching technique	229
4.48	Results of images for Ring_1 after applying the three cascaded clustering algorithms	231
4.49	Results of images for Trophozoite_1 after applying the three cascaded clustering algorithms	231
4.50	Results of images for Schizont_1 after applying the three cascaded clustering algorithms	232
4.51	Results of images for Gametocyte_1 after applying the three cascaded clustering algorithms	233
4.52	Results of images for Ring_1 after the colour of clustering image has been retrieved based on MGCS image	233

4.53	Results of images for Trophozoite_1 after the colour of clustering image has been retrieved based on MGCS image	234
4.54	Results of images for Schizont_1 after the colour of clustering image has been retrieved based on MGCS image	235
4.55	Results of images for Gametocyte_1 after the colour of clustering image has been retrieved based on MGCS image	236
4.56	Results of images for Ring_1 after applying median filter and SRGAE algorithms to clustering image	236
4.57	Results of images for Trophozoite_1 after applying median filter and SRGAE algorithms to clustering image	237
4.58	Results of images for Schizont_1 after applying median filter and SRGAE algorithms to clustering image	238
4.59	Results of images for Gametocyte_1 after applying median filter and SRGAE algorithms to clustering image	239
4.60	Original malaria images	240
4.61	Results of images of modified global contrast stretching technique	240
4.62	The final segmented images of Trophozoite_2 after applying the proposed segmentation method	241
4.63	The final segmented images of Schizont_2 after applying the proposed segmentation method	242
4.64	The final segmented images of Gametocyte_2 after applying the proposed segmentation method	243
4.65	The final segmented images of Gametocyte_3 after applying the proposed segmentation method	243
4.66	Original PF_Ring_1 image and the results of images after applying the proposed segmentation method	254
4.67	Original PF_Schizont image and the results of images after applying the proposed segmentation method	255
4.68	Original PF_Gametocyte image and the results of images after applying the proposed segmentation method	255

4.69	Original PV_Ring image and the results of images after applying the proposed segmentation method	256
4.70	Original PV_Trophozoite image and the results of images after applying the proposed segmentation method	256
4.71	Original PV_Gametocyte image and the results of images after applying the proposed segmentation method	257
4.72	Original PF_Ring_2 image and the results of images after applying the proposed segmentation method	257
4.73	Original PF_Trophozoite image and the results of images after applying the proposed segmentation method	258
4.74	Original PV_Schizont image and the results of images after applying the proposed segmentation method	259
4.75	Original and final segmented images for ring stage of <i>P. falciparum</i>	259
4.76	Original and final segmented images for trophozoite stage of <i>P. falciparum</i>	260
4.77	Original and final segmented images for schizont stage of <i>P. falciparum</i>	260
4.78	Original and final segmented images for gametocyte stage of <i>P. falciparum</i>	261
4.79	Original and final segmented images for ring stage of <i>P. vivax</i>	261
4.80	Original and final segmented images for trophozoite stage of <i>P. vivax</i>	261
4.81	Original and final segmented images for schizont stage of <i>P. vivax</i>	262
4.82	Original and final segmented images for gametocyte stage of <i>P. vivax</i>	262
4.83	Original and final segmented images for ring stage of <i>P. falciparum</i> species	263
4.84	Original PF_Gametocyte image and the results of images after applying the proposed segmentation method	265
4.85	A single segmented malaria parasite and WBC in malaria image with its area	265

4.86	Analysis of number of training epochs for classification between malaria parasite and non-malaria parasite using MLP_LM network	270
4.87	Analysis of number of hidden nodes for classification between malaria parasite and non-malaria parasite using MLP_LM network	271
4.88	Analysis of number of hidden nodes for classification between malaria parasite and non-malaria parasite using OS-ELM	275
4.89	Analysis of number of trees for classification between malaria parasite and non-malaria parasite using random forest	278
4.90	The structure of 3-stage classification processes of the intelligent diagnostic system for malaria	290
4.91	The GUI of several facilities that have been embedded inside the proposed intelligent diagnostic system for malaria	300
4.92	The GUI for classification of PV_Schizont image using the proposed intelligent diagnostic system for malaria	302
4.93	The GUI for classification of WBC image using the proposed intelligent diagnostic system for malaria	303
4.94	The GUI for classification of normal blood image using the proposed intelligent diagnostic system for malaria	304
C.1	The system is presently classifying the blood image as malaria image, and further classifies it as ring stage of <i>P. falciparum</i> species	362
C.2	The system is presently classifying the blood image as malaria image, and further classifies it as mature trophozoite stage of <i>P. falciparum</i> species	362
C.3	The system is presently classifying the blood image as malaria image, and further classifies it as schizont stage of <i>P. falciparum</i> species	363
C.4	The system is presently classifying the blood image as malaria image, and further classifies it as gametocyte stage of <i>P. falciparum</i> species	363
C.5	The system is presently classifying the blood image as malaria image, and further classifies it as ring stage of <i>P. vivax</i> species	364

C.6	The system is presently classifying the blood image as malaria image, and further classifies it as mature trophozoite stage of <i>P. vivax</i> species	364
C.7	The system is presently classifying the blood image as malaria image, and further classifies it as schizont stage of <i>P. vivax</i> species	365
C.8	The system is presently classifying the blood image as malaria image, and further classifies it as mature trophozoite stage of <i>P. vivax</i> species	365
C.9	The system is presently classifying the blood image as non-malaria image	366
C.10	The system is presently classifying the blood image as non-malaria image	366
C.11	The system is presently classifying the blood image as non-malaria image	367
C.12	The system is presently classifying the blood image as non-malaria image	367

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LIST OF ABBREVIATIONS

1-NN	1-nearest neighbors
AFKM	Adaptive fuzzy k -means
AI	Artificial intelligence
ANN	Artificial neural network
BP	Back propagation
BFF	Back propagation feed forward
BR	Bayesian regulation
CDC	Centers for Disease Control and Prevention
CART	Classification and regression tree
CT	Computed tomography
DNA	Deoxyribonucleic acid
DSM	Distribution separation measure
EGK'M	Enhanced global K' -means
EKM	Enhanced k -means
EnSFCM	Enhanced suppressed fuzzy c -means
ELM	Extreme learning machine
FLD	Fisher linear discriminant
FAM	Fuzzy ARTMAP
FCM	Fuzzy c -means
GCS	Global contrast stretching
HRP-II	Histidine-rich protein II