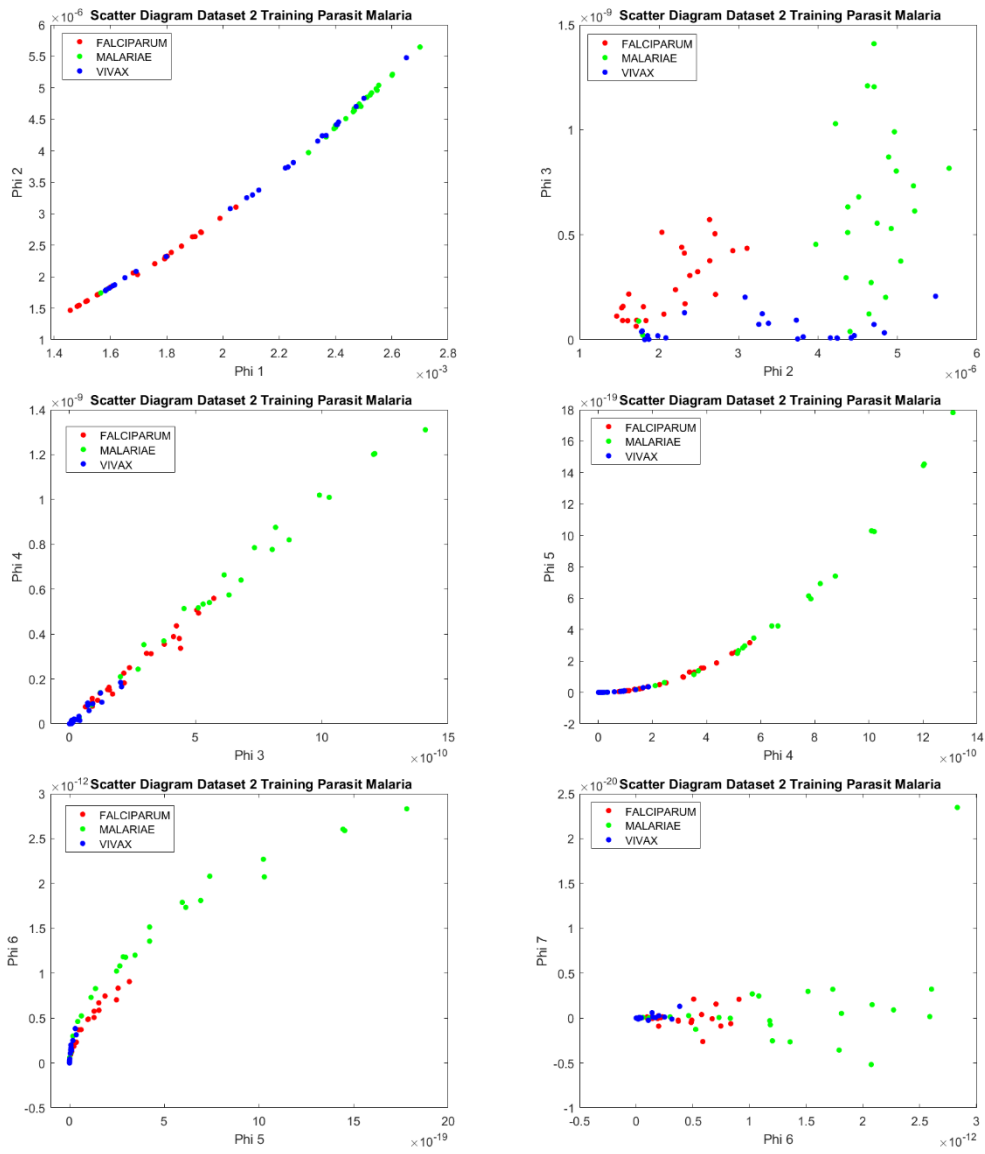
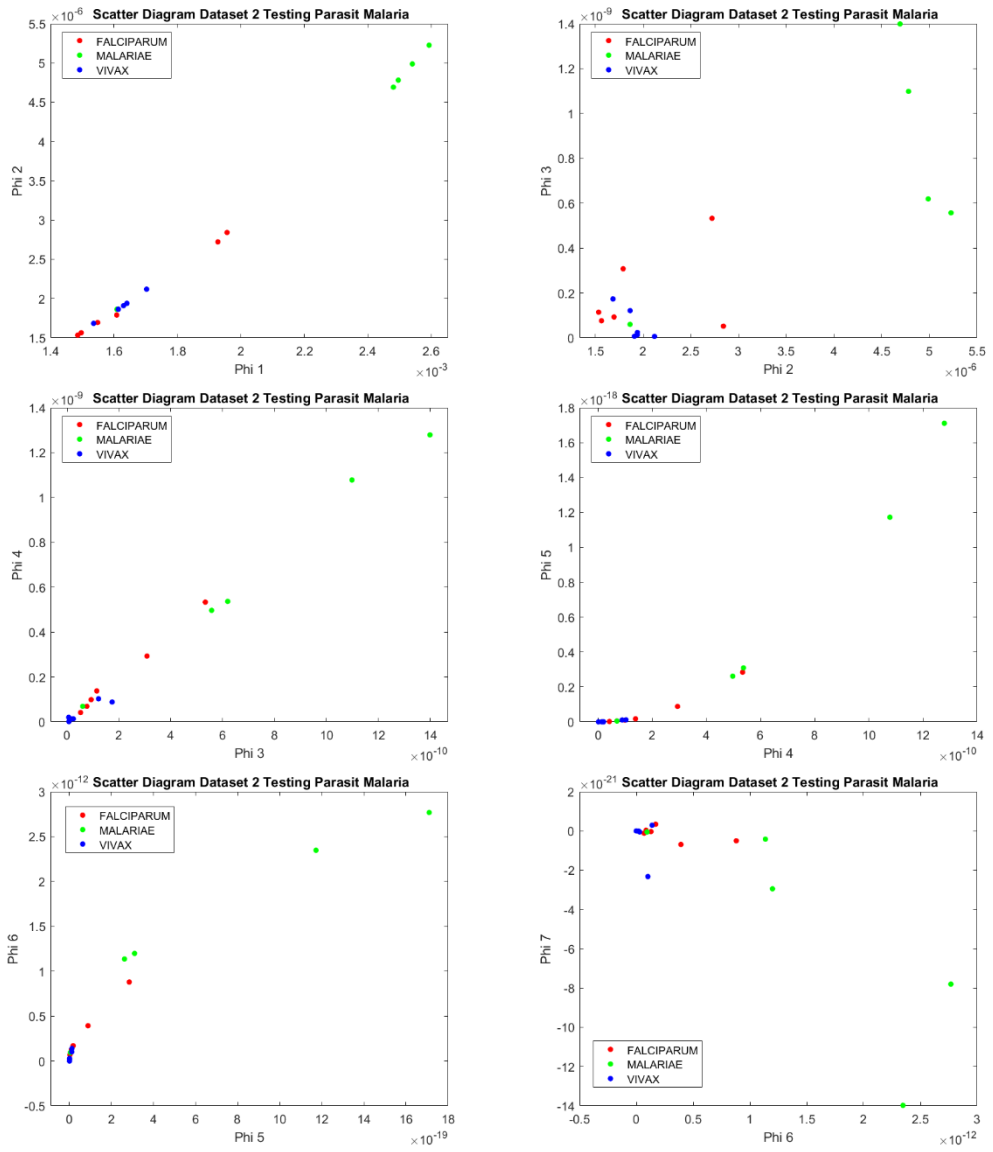


LAMPIRAN

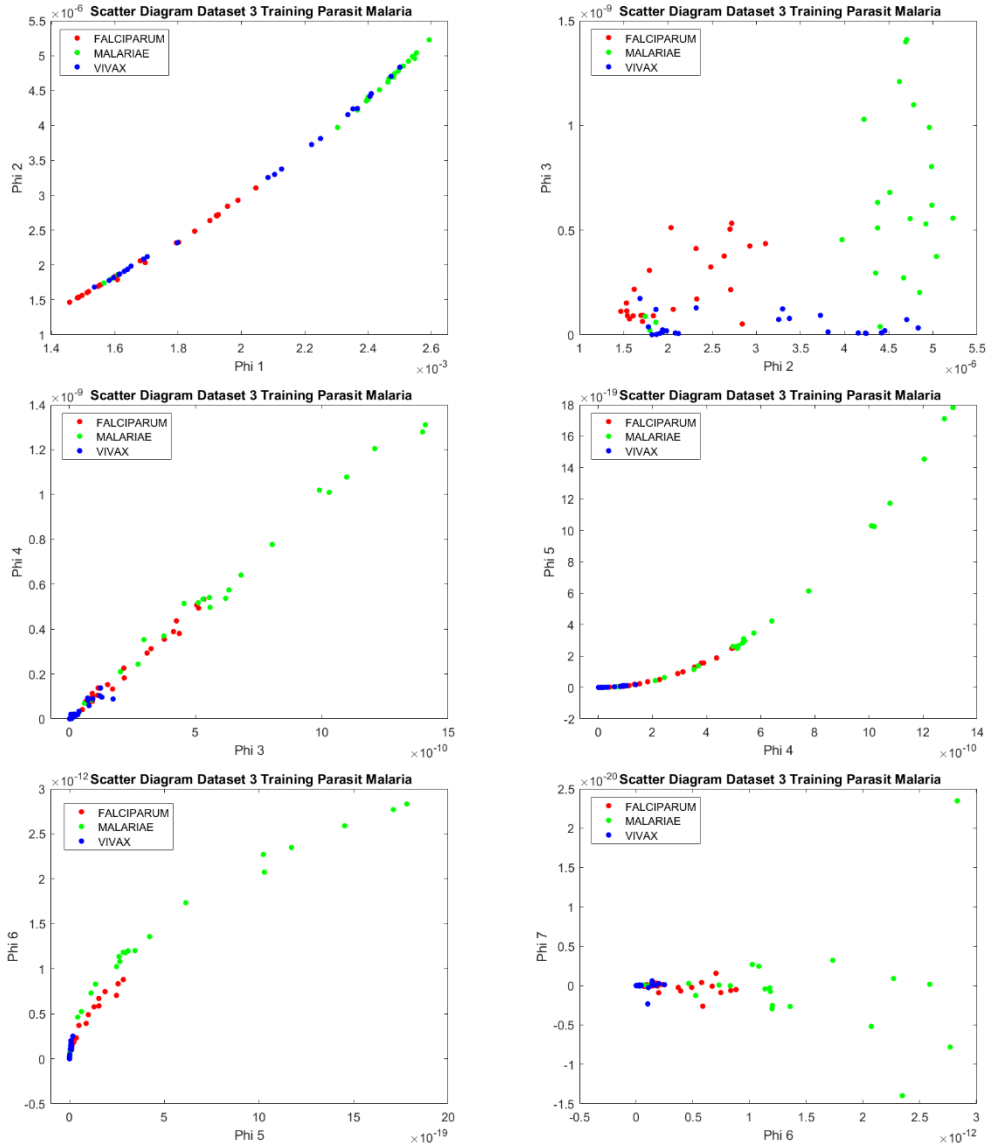
Scatter Diagram Dataset 2 Training



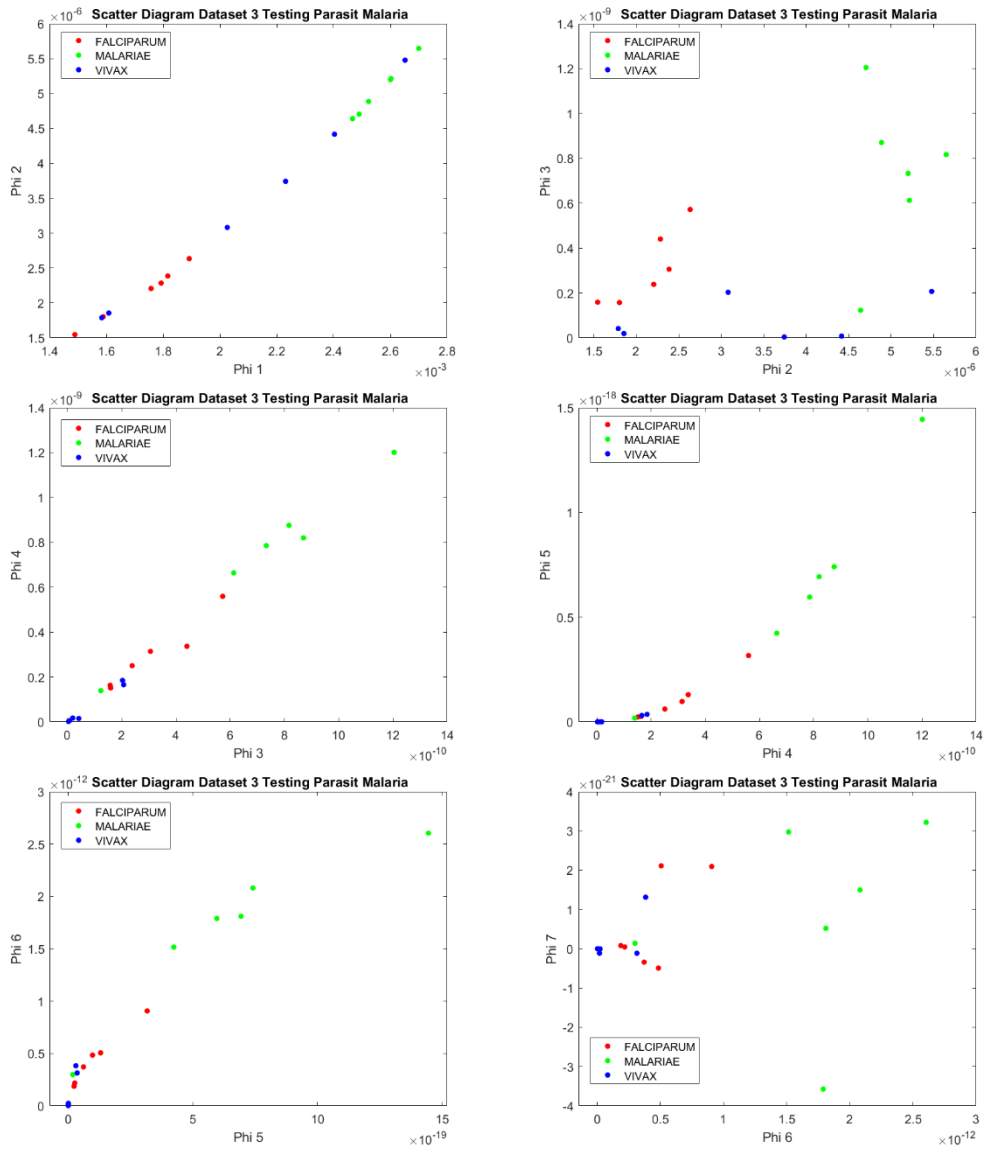
Scatter Diagram Dataset 2 Testing



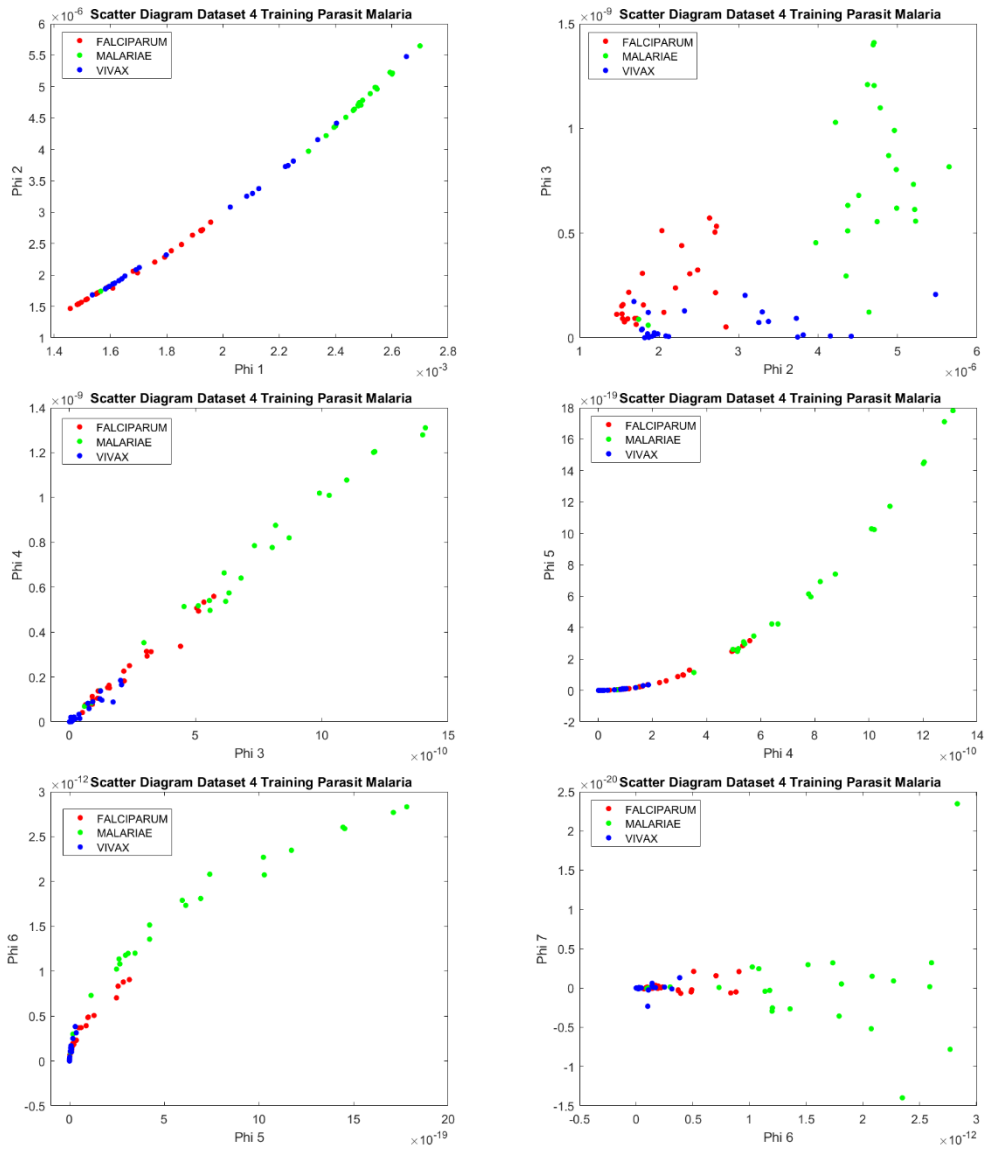
Scatter Diagram Dataset 3 Training



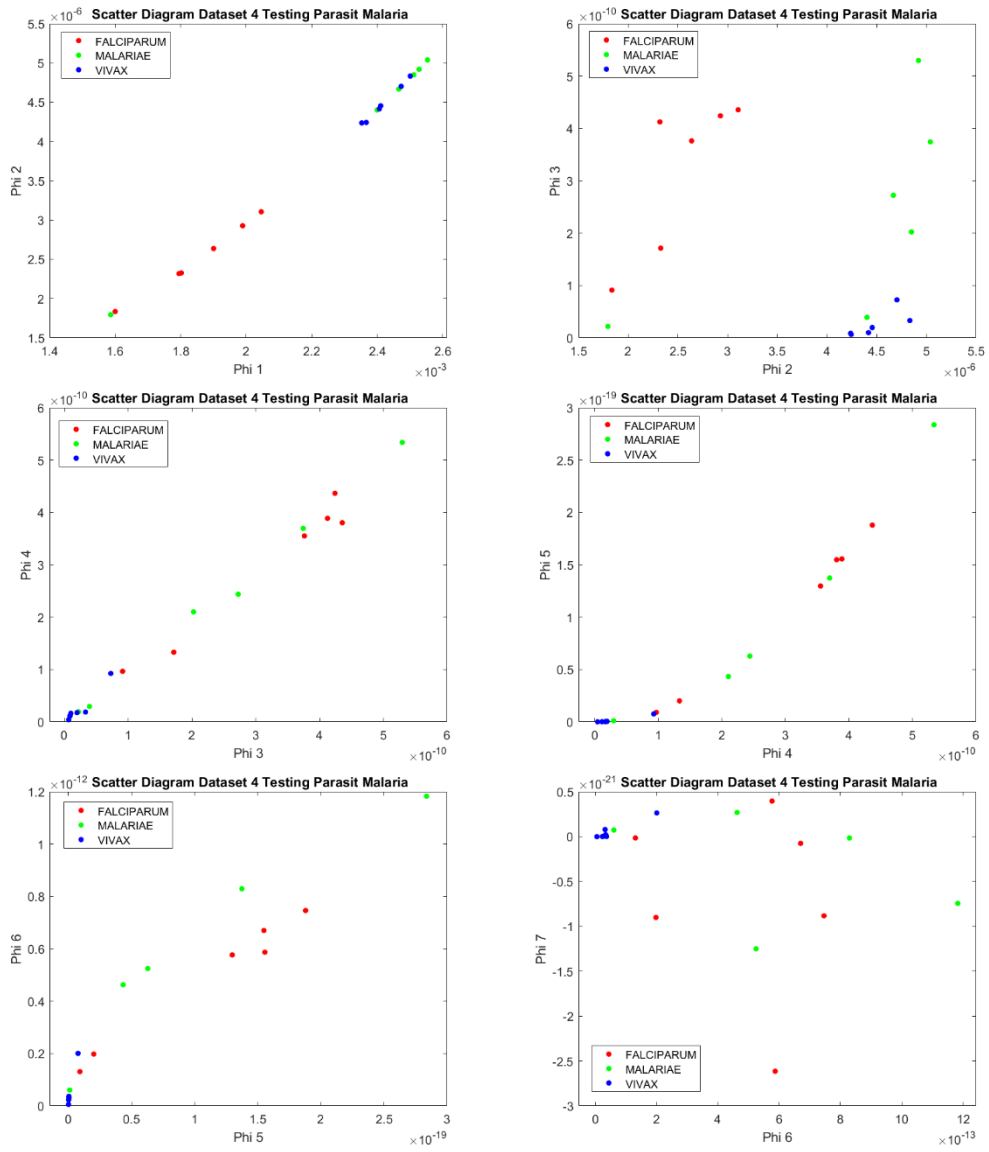
Scatter Diagram Dataset 3 Testing



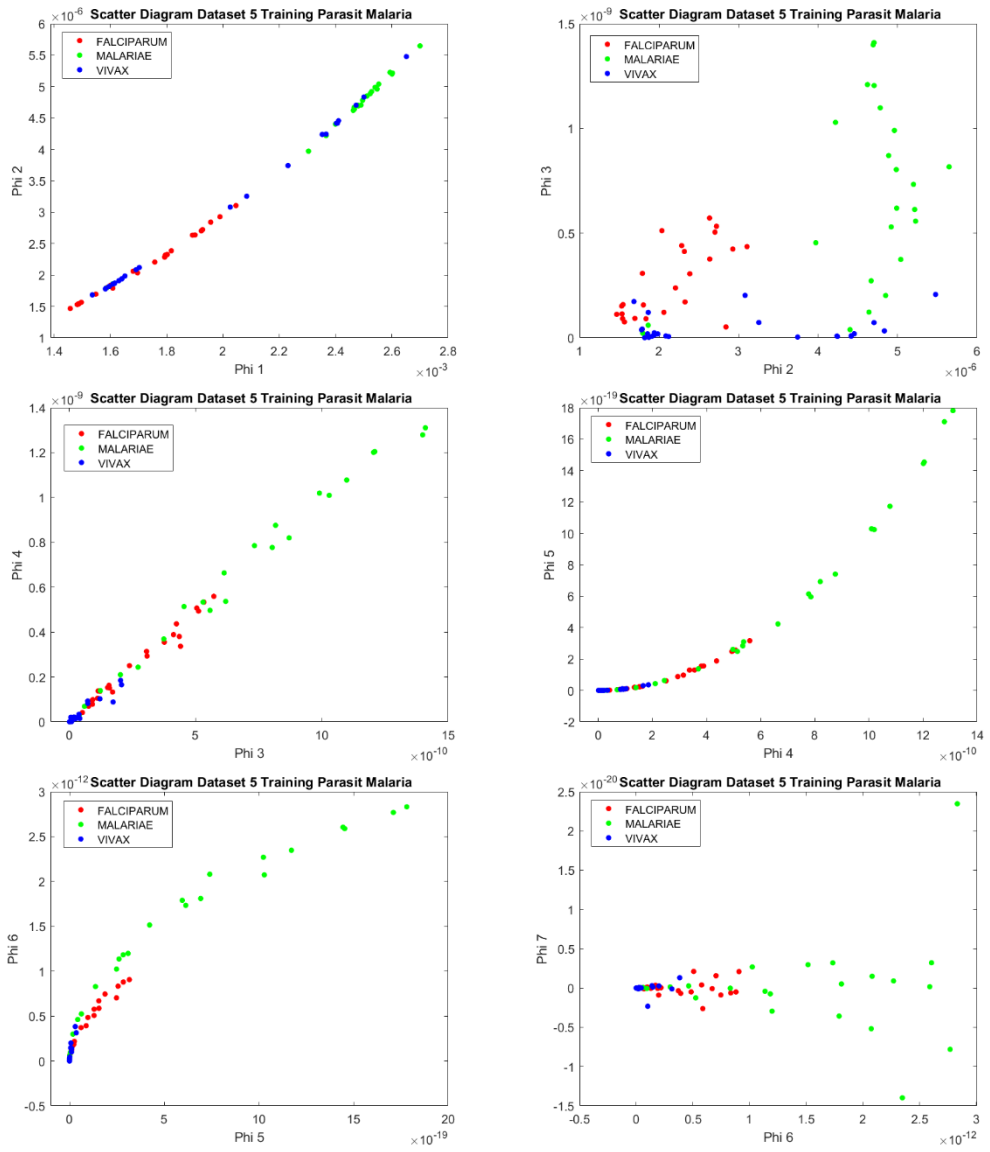
Scatter Diagram Dataset 4 Training



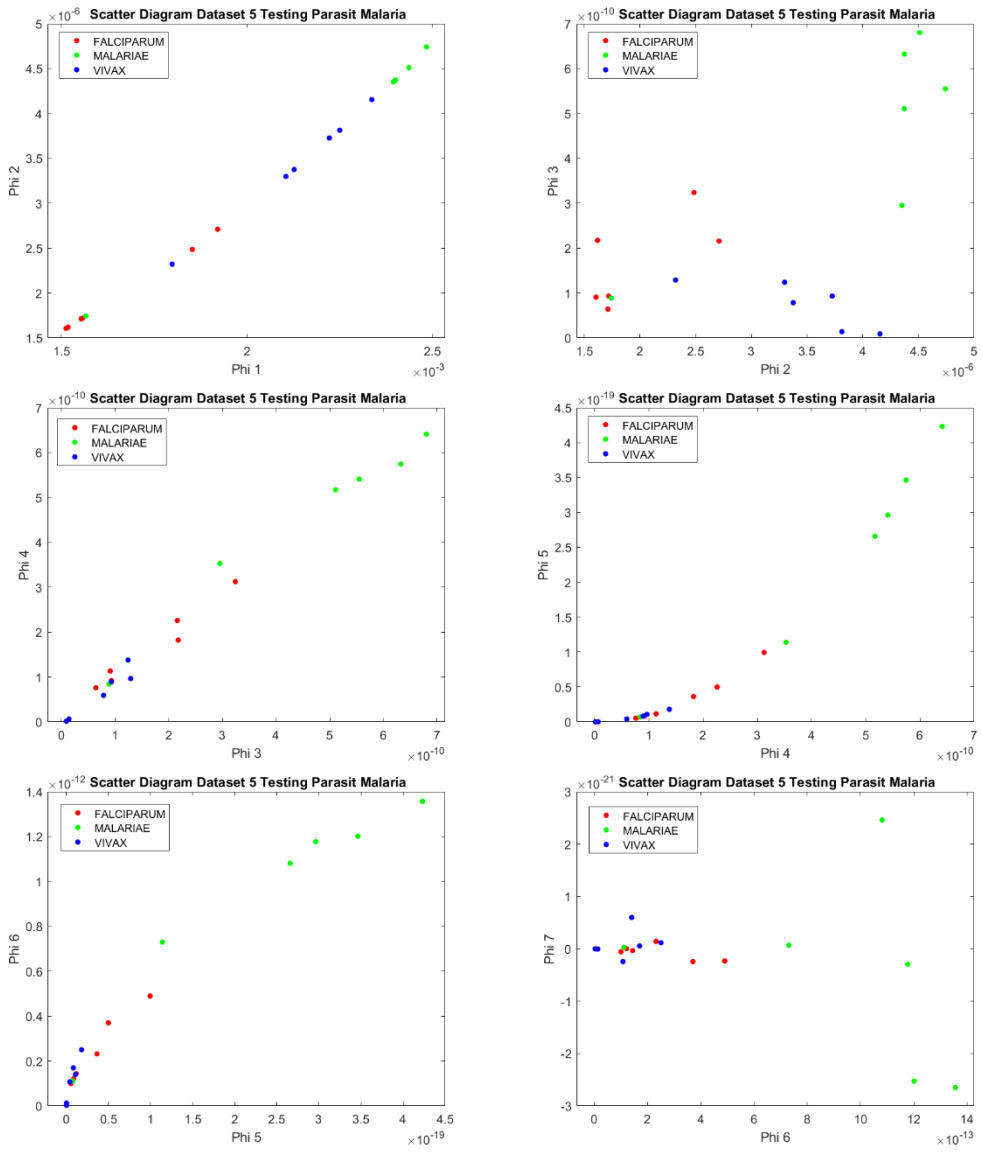
Scatter Diagram Dataset 4 Testing



Scatter Diagram Dataset 5 Training



Scatter Diagram Dataset 5 Testing



Tabel Nilai Ekstraksi Fitur Kelas *Falciparum*

Nama Citra/Kelas	Phi 1	Phi 2	Phi 3	Phi 4	Phi 5	Phi 6	Phi 7
fal(1).bmp	0.001518	1.62E-06	2.17E-10	1.82E-10	3.63E-20	2.32E-13	1.4E-22
fal(10).bmp	0.001901	2.64E-06	3.76E-10	3.55E-10	1.3E-19	5.77E-13	3.96E-22
fal(11).bmp	0.001802	2.33E-06	1.71E-10	1.33E-10	2E-20	1.98E-13	-9E-22
fal(12).bmp	0.002046	3.11E-06	4.36E-10	3.8E-10	1.55E-19	6.7E-13	-7.5E-23
fal(13).bmp	0.001758	2.21E-06	2.38E-10	2.5E-10	6.12E-20	3.71E-13	-3.4E-22
fal(14).bmp	0.001892	2.63E-06	5.72E-10	5.6E-10	3.17E-19	9.06E-13	2.1E-21
fal(15).bmp	0.001793	2.28E-06	4.4E-10	3.37E-10	1.3E-19	5.07E-13	2.11E-21
fal(16).bmp	0.001816	2.39E-06	3.06E-10	3.14E-10	9.73E-20	4.84E-13	-4.9E-22
fal(17).bmp	0.001589	1.8E-06	1.57E-10	1.63E-10	2.6E-20	2.18E-13	4.61E-23
fal(18).bmp	0.001489	1.55E-06	1.59E-10	1.51E-10	2.34E-20	1.87E-13	8.2E-23
fal(19).bmp	0.001497	1.56E-06	7.61E-11	6.92E-11	5.02E-21	8.58E-14	3.72E-23
fal(2).bmp	0.001853	2.49E-06	3.24E-10	3.13E-10	9.94E-20	4.89E-13	-2.3E-22
fal(20).bmp	0.001486	1.53E-06	1.14E-10	1.38E-10	1.73E-20	1.7E-13	3.44E-22
fal(21).bmp	0.001956	2.84E-06	5.19E-11	4.14E-11	1.92E-21	6.84E-14	-1.2E-22
fal(22).bmp	0.001548	1.7E-06	9.28E-11	9.89E-11	9.48E-21	1.29E-13	-3.2E-23
fal(23).bmp	0.001927	2.72E-06	5.33E-10	5.33E-10	2.84E-19	8.8E-13	-5E-22
fal(24).bmp	0.001608	1.79E-06	3.08E-10	2.93E-10	8.81E-20	3.92E-13	-6.9E-22
fal(25).bmp	0.001923	2.7E-06	5.05E-10	5.07E-10	2.57E-19	8.33E-13	-6.2E-22
fal(26).bmp	0.001457	1.47E-06	1.12E-10	1.05E-10	1.14E-20	1.26E-13	7.5E-23
fal(27).bmp	0.001486	1.54E-06	9.15E-11	7.83E-11	6.63E-21	9.65E-14	1.34E-22
fal(28).bmp	0.001482	1.53E-06	1.52E-10	1.53E-10	2.32E-20	1.89E-13	-4.1E-23
fal(29).bmp	0.001681	2.06E-06	1.22E-10	1.37E-10	1.76E-20	1.95E-13	2.44E-22
fal(3).bmp	0.001513	1.61E-06	9.07E-11	1.13E-10	1.15E-20	1.43E-13	-3.5E-23
fal(30).bmp	0.001696	2.04E-06	5.12E-10	4.94E-10	2.48E-19	7.03E-13	1.57E-21
fal(4).bmp	0.001553	1.71E-06	6.39E-11	7.59E-11	5.29E-21	9.92E-14	-5.7E-23
fal(5).bmp	0.001921	2.71E-06	2.16E-10	2.26E-10	4.99E-20	3.7E-13	-2.4E-22
fal(6).bmp	0.001557	1.72E-06	9.3E-11	9.19E-11	8.49E-21	1.2E-13	1.71E-24
fal(7).bmp	0.001795	2.32E-06	4.12E-10	3.89E-10	1.56E-19	5.87E-13	-2.6E-21
fal(8).bmp	0.0016	1.84E-06	9.11E-11	9.65E-11	9.05E-21	1.31E-13	-1.5E-23
fal(9).bmp	0.001989	2.93E-06	4.24E-10	4.37E-10	1.88E-19	7.46E-13	-8.8E-22
Rata-rata	0.001704	2.11E-06	2.49E-10	2.4E-10	8.31E-20	3.63E-13	-2E-23
Standard Deviasi	0.000187	5.07E-07	1.65E-10	1.57E-10	9.41E-20	2.66E-13	8.65E-22

Tabel Nilai Ekstraksi Fitur Kelas *Malariae*

Nama Citra/Kelas	Phi 1	Phi 2	Phi 3	Phi 4	Phi 5	Phi 6	Phi 7
mal(1).bmp	0.002484	4.74E-06	5.55E-10	5.41E-10	2.96E-19	1.18E-12	-3E-22
mal(10).bmp	0.002512	4.85E-06	2.02E-10	2.1E-10	4.33E-20	4.62E-13	2.69E-22
mal(11).bmp	0.002466	4.67E-06	2.72E-10	2.44E-10	6.28E-20	5.24E-13	-1.3E-21
mal(12).bmp	0.0024	4.4E-06	3.92E-11	2.94E-11	9.92E-22	6.02E-14	7.32E-23
mal(13).bmp	0.002467	4.64E-06	1.23E-10	1.39E-10	1.81E-20	2.99E-13	1.39E-22
mal(14).bmp	0.002602	5.21E-06	6.13E-10	6.64E-10	4.23E-19	1.52E-12	2.97E-21
mal(15).bmp	0.002601	5.2E-06	7.33E-10	7.85E-10	5.96E-19	1.79E-12	-3.6E-21
mal(16).bmp	0.0027	5.65E-06	8.17E-10	8.76E-10	7.41E-19	2.08E-12	1.5E-21
mal(17).bmp	0.002523	4.89E-06	8.7E-10	8.2E-10	6.93E-19	1.81E-12	5.21E-22
mal(18).bmp	0.00249	4.7E-06	1.2E-09	1.2E-09	1.44E-18	2.6E-12	3.22E-21
mal(19).bmp	0.001609	1.86E-06	6.01E-11	6.9E-11	4.45E-21	9.41E-14	-5.2E-23
mal(2).bmp	0.001566	1.74E-06	8.86E-11	8.41E-11	7.26E-21	1.11E-13	2.64E-23
mal(20).bmp	0.002481	4.69E-06	1.4E-09	1.28E-09	1.71E-18	2.77E-12	-7.8E-21
mal(21).bmp	0.002496	4.78E-06	1.1E-09	1.08E-09	1.17E-18	2.35E-12	-1.4E-20
mal(22).bmp	0.002593	5.23E-06	5.57E-10	4.97E-10	2.61E-19	1.14E-12	-4.2E-22
mal(23).bmp	0.002541	4.99E-06	6.19E-10	5.37E-10	3.09E-19	1.2E-12	-2.9E-21
mal(24).bmp	0.002541	4.99E-06	6.19E-10	5.37E-10	3.09E-19	1.2E-12	-2.9E-21
mal(25).bmp	0.002544	4.98E-06	8.03E-10	7.77E-10	6.14E-19	1.73E-12	3.21E-21
mal(26).bmp	0.002463	4.62E-06	1.21E-09	1.2E-09	1.45E-18	2.59E-12	1.7E-22
mal(27).bmp	0.002548	4.96E-06	9.9E-10	1.02E-09	1.02E-18	2.27E-12	9.07E-22
mal(28).bmp	0.002479	4.7E-06	1.41E-09	1.31E-09	1.78E-18	2.83E-12	2.35E-20
mal(29).bmp	0.002367	4.22E-06	1.03E-09	1.01E-09	1.03E-18	2.07E-12	-5.2E-21
mal(3).bmp	0.002401	4.37E-06	6.32E-10	5.74E-10	3.46E-19	1.2E-12	-2.5E-21
mal(30).bmp	0.002304	3.97E-06	4.54E-10	5.14E-10	2.48E-19	1.02E-12	2.69E-21
mal(4).bmp	0.0024	4.37E-06	5.11E-10	5.17E-10	2.66E-19	1.08E-12	2.46E-21
mal(5).bmp	0.002395	4.35E-06	2.95E-10	3.53E-10	1.14E-19	7.3E-13	6.74E-23
mal(6).bmp	0.002436	4.51E-06	6.8E-10	6.41E-10	4.23E-19	1.36E-12	-2.6E-21
mal(7).bmp	0.001587	1.79E-06	2.18E-11	1.92E-11	3.94E-22	2.58E-14	9.51E-25
mal(8).bmp	0.002553	5.04E-06	3.74E-10	3.69E-10	1.37E-19	8.29E-13	-1.6E-23
mal(9).bmp	0.002528	4.92E-06	5.3E-10	5.34E-10	2.84E-19	1.18E-12	-7.4E-22
Rata-rata	0.002403	4.47E-06	6.27E-10	6.14E-10	5.27E-19	1.34E-12	-8.9E-23
Standard Deviasi	0.000288	9.66E-07	4.01E-10	3.86E-10	5.34E-19	8.49E-13	5.65E-21

Tabel Nilai Ekstraksi Fitur Kelas *Vivax*

Nama Citra/Kelas	Phi 1	Phi 2	Phi 3	Phi 4	Phi 5	Phi 6	Phi 7
viv(1).bmp	0.002337	4.15E-06	8.84E-12	1.4E-12	4.65E-24	2.56E-15	-1.6E-24
viv(10).bmp	0.002367	4.24E-06	6.54E-12	4.11E-12	2.13E-23	5.15E-15	-4.3E-25
viv(11).bmp	0.002353	4.24E-06	8.55E-12	1.11E-11	1.09E-22	2.29E-14	1.12E-24
viv(12).bmp	0.002501	4.83E-06	3.3E-11	1.88E-11	4.62E-22	3.18E-14	7.83E-23
viv(13).bmp	0.002403	4.42E-06	7.72E-12	5.35E-12	3.44E-23	8.68E-15	-1.2E-24
viv(14).bmp	0.002231	3.74E-06	3.72E-12	1.76E-12	4.47E-24	2.08E-15	-6.3E-25
viv(15).bmp	0.002026	3.08E-06	2.03E-10	1.85E-10	3.58E-20	3.14E-13	-1.1E-22
viv(16).bmp	0.002652	5.48E-06	2.07E-10	1.65E-10	3.05E-20	3.83E-13	1.31E-21
viv(17).bmp	0.001609	1.86E-06	1.91E-11	1.7E-11	3.07E-22	2.28E-14	-5.4E-24
viv(18).bmp	0.001584	1.79E-06	4.14E-11	1.5E-11	3.55E-22	1.79E-14	-1.1E-22
viv(19).bmp	0.001536	1.68E-06	1.73E-10	8.85E-11	1.07E-20	1.01E-13	-2.3E-21
viv(2).bmp	0.001798	2.32E-06	1.29E-10	9.64E-11	1.07E-20	1.4E-13	5.99E-22
viv(20).bmp	0.001613	1.86E-06	1.21E-10	1.03E-10	1.14E-20	1.39E-13	2.93E-22
viv(21).bmp	0.001641	1.94E-06	1.31E-11	1.38E-11	1.85E-22	1.92E-14	-4.6E-25
viv(22).bmp	0.001641	1.94E-06	2.38E-11	1.35E-11	2.42E-22	1.86E-14	-1.1E-23
viv(23).bmp	0.00163	1.91E-06	7.05E-12	1.12E-12	1.15E-24	-6E-16	2.91E-24
viv(24).bmp	0.001703	2.12E-06	6.2E-12	1.99E-11	2.17E-22	2.84E-14	-4.5E-23
viv(25).bmp	0.002084	3.25E-06	7.31E-11	8.23E-11	6.38E-21	1.47E-13	1.15E-22
viv(26).bmp	0.001596	1.82E-06	7.85E-13	4.74E-13	2.89E-25	4.32E-16	-5.1E-27
viv(27).bmp	0.001615	1.87E-06	2.12E-12	2.62E-13	-9E-26	-2.2E-16	1.74E-25
viv(28).bmp	0.001691	2.08E-06	8.92E-12	1.26E-12	1.21E-24	2.43E-16	4.04E-24
viv(29).bmp	0.001652	1.98E-06	1.89E-11	2.09E-11	4.07E-22	2.62E-14	7.91E-23
viv(3).bmp	0.002222	3.73E-06	9.3E-11	8.92E-11	8.12E-21	1.69E-13	5.44E-23
viv(30).bmp	0.001582	1.78E-06	3.82E-11	3.37E-11	1.21E-21	4.41E-14	2.2E-23
viv(4).bmp	0.002127	3.38E-06	7.83E-11	5.9E-11	4E-21	1.07E-13	-2.4E-22
viv(5).bmp	0.002105	3.3E-06	1.24E-10	1.38E-10	1.8E-20	2.5E-13	1.14E-22
viv(6).bmp	0.00225	3.81E-06	1.38E-11	6.23E-12	5.74E-23	1.21E-14	-4.7E-24
viv(7).bmp	0.002406	4.42E-06	9.85E-12	1.62E-11	2.03E-22	3.33E-14	2.17E-23
viv(8).bmp	0.002473	4.7E-06	7.26E-11	9.25E-11	7.57E-21	2E-13	2.63E-22
viv(9).bmp	0.002411	4.45E-06	1.96E-11	1.76E-11	3.27E-22	3.65E-14	1.86E-24
Rata-Rata	0.001995	3.07E-06	5.22E-11	4.39E-11	4.91E-21	7.61E-14	3.33E-24
Standard Deviasi	0.000366	1.19E-06	6.19E-11	5.28E-11	9.01E-21	1.01E-13	5.17E-22

Confusion Matrix Data Training Polynomial SVM

Confusion Matrix Polynomial SVM															
Data set-1	Actual	Fal	<table border="1"> <tr><td>21</td><td>0</td><td>3</td></tr> <tr><td>2</td><td>18</td><td>4</td></tr> <tr><td>3</td><td>1</td><td>20</td></tr> <tr><td>Fal</td><td>Mal</td><td>Viv</td></tr> </table>	21	0	3	2	18	4	3	1	20	Fal	Mal	Viv
21	0	3													
2	18	4													
3	1	20													
Fal	Mal	Viv													
		Mal													
		Viv													
			Predicted												
Data set-2	Actual	Fal	<table border="1"> <tr><td>22</td><td>1</td><td>1</td></tr> <tr><td>1</td><td>20</td><td>3</td></tr> <tr><td>1</td><td>1</td><td>22</td></tr> <tr><td>Fal</td><td>Mal</td><td>Viv</td></tr> </table>	22	1	1	1	20	3	1	1	22	Fal	Mal	Viv
22	1	1													
1	20	3													
1	1	22													
Fal	Mal	Viv													
		Mal													
		Viv													
			Predicted												
Data set-3	Actual	Fal	<table border="1"> <tr><td>22</td><td>0</td><td>2</td></tr> <tr><td>2</td><td>20</td><td>2</td></tr> <tr><td>2</td><td>0</td><td>22</td></tr> <tr><td>Fal</td><td>Mal</td><td>Viv</td></tr> </table>	22	0	2	2	20	2	2	0	22	Fal	Mal	Viv
22	0	2													
2	20	2													
2	0	22													
Fal	Mal	Viv													
		Mal													
		Viv													
			Predicted												
Data set-4	Actual	Fal	<table border="1"> <tr><td>20</td><td>0</td><td>4</td></tr> <tr><td>3</td><td>19</td><td>2</td></tr> <tr><td>3</td><td>1</td><td>20</td></tr> <tr><td>Fal</td><td>Mal</td><td>Viv</td></tr> </table>	20	0	4	3	19	2	3	1	20	Fal	Mal	Viv
20	0	4													
3	19	2													
3	1	20													
Fal	Mal	Viv													
		Mal													
		Viv													
			Predicted												
Data set-5	Actual	Fal	<table border="1"> <tr><td>20</td><td>0</td><td>4</td></tr> <tr><td>1</td><td>18</td><td>5</td></tr> <tr><td>3</td><td>1</td><td>20</td></tr> <tr><td>Fal</td><td>Mal</td><td>Viv</td></tr> </table>	20	0	4	1	18	5	3	1	20	Fal	Mal	Viv
20	0	4													
1	18	5													
3	1	20													
Fal	Mal	Viv													
		Mal													
		Viv													
			Predicted												

Tabel Pengukuran Training Polynomial SVM

Dataset-n	Akurasi	Sensitivitas	Spesifisitas
Dataset 1	81.9	87.5	89.58
Dataset 2	88.9	91.67	95.83
Dataset 3	88.9	91.67	91.67
Dataset 4	81.9	83.33	87.5
Dataset 5	80.6	83.33	91.67
Rata-rata	84.44	87.5	91.25

Confusion Matrix Data Training Linier SVM

Confusion Matrix Linier SVM																			
Data set-1	Actual	<table border="1"> <tr> <td>Fal</td> <td>22</td> <td>0</td> <td>2</td> </tr> <tr> <td>Mal</td> <td>1</td> <td>17</td> <td>6</td> </tr> <tr> <td>Viv</td> <td>6</td> <td>0</td> <td>18</td> </tr> <tr> <td></td> <td>Fal</td> <td>Mal</td> <td>Viv</td> </tr> </table>	Fal	22	0	2	Mal	1	17	6	Viv	6	0	18		Fal	Mal	Viv	Predicted
Fal	22	0	2																
Mal	1	17	6																
Viv	6	0	18																
	Fal	Mal	Viv																
Data set-2	Actual	<table border="1"> <tr> <td>Fal</td> <td>24</td> <td>0</td> <td>0</td> </tr> <tr> <td>Mal</td> <td>1</td> <td>19</td> <td>4</td> </tr> <tr> <td>Viv</td> <td>1</td> <td>0</td> <td>23</td> </tr> <tr> <td></td> <td>Fal</td> <td>Mal</td> <td>Viv</td> </tr> </table>	Fal	24	0	0	Mal	1	19	4	Viv	1	0	23		Fal	Mal	Viv	Predicted
Fal	24	0	0																
Mal	1	19	4																
Viv	1	0	23																
	Fal	Mal	Viv																
Data set-3	Actual	<table border="1"> <tr> <td>Fal</td> <td>22</td> <td>0</td> <td>2</td> </tr> <tr> <td>Mal</td> <td>2</td> <td>18</td> <td>4</td> </tr> <tr> <td>Viv</td> <td>3</td> <td>0</td> <td>21</td> </tr> <tr> <td></td> <td>Fal</td> <td>Mal</td> <td>Viv</td> </tr> </table>	Fal	22	0	2	Mal	2	18	4	Viv	3	0	21		Fal	Mal	Viv	Predicted
Fal	22	0	2																
Mal	2	18	4																
Viv	3	0	21																
	Fal	Mal	Viv																
Data set-4	Actual	<table border="1"> <tr> <td>Fal</td> <td>23</td> <td>0</td> <td>1</td> </tr> <tr> <td>Mal</td> <td>1</td> <td>21</td> <td>2</td> </tr> <tr> <td>Viv</td> <td>4</td> <td>0</td> <td>20</td> </tr> <tr> <td></td> <td>Fal</td> <td>Mal</td> <td>Viv</td> </tr> </table>	Fal	23	0	1	Mal	1	21	2	Viv	4	0	20		Fal	Mal	Viv	Predicted
Fal	23	0	1																
Mal	1	21	2																
Viv	4	0	20																
	Fal	Mal	Viv																
Data set-5	Actual	<table border="1"> <tr> <td>Fal</td> <td>21</td> <td>0</td> <td>3</td> </tr> <tr> <td>Mal</td> <td>1</td> <td>18</td> <td>5</td> </tr> <tr> <td>Viv</td> <td>3</td> <td>1</td> <td>20</td> </tr> <tr> <td></td> <td>Fal</td> <td>Mal</td> <td>Viv</td> </tr> </table>	Fal	21	0	3	Mal	1	18	5	Viv	3	1	20		Fal	Mal	Viv	Predicted
Fal	21	0	3																
Mal	1	18	5																
Viv	3	1	20																
	Fal	Mal	Viv																

Tabel Pengukuran Data Training Linier SVM

Dataset-n	Akurasi	Sensitivitas	Spesifisitas
Dataset 1	79.2	91.67	85.42
Dataset 2	91.7	100	95.83
Dataset 3	84.7	91.67	89.58
Dataset 4	88.9	95.83	89.58
Dataset 5	81.9	87.5	91.67
Rata-rata	85.28	93.33	90.42

Confusion Matrix Data Training Gaussian SVM

Confusion Matrix Gaussian SVM															
Data set-1	Actual	Fal	<table border="1" style="margin: auto; border-collapse: collapse;"> <tr> <td style="padding: 2px;">20</td> <td style="padding: 2px;">3</td> <td style="padding: 2px;">1</td> </tr> <tr> <td style="padding: 2px;">2</td> <td style="padding: 2px;">20</td> <td style="padding: 2px;">2</td> </tr> <tr> <td style="padding: 2px;">2</td> <td style="padding: 2px;">2</td> <td style="padding: 2px;">20</td> </tr> <tr> <td style="padding: 2px;">Fal</td> <td style="padding: 2px;">Mal</td> <td style="padding: 2px;">Viv</td> </tr> </table>	20	3	1	2	20	2	2	2	20	Fal	Mal	Viv
20	3	1													
2	20	2													
2	2	20													
Fal	Mal	Viv													
			Predicted												
Data set-2	Actual	Fal	<table border="1" style="margin: auto; border-collapse: collapse;"> <tr> <td style="padding: 2px;">20</td> <td style="padding: 2px;">4</td> <td style="padding: 2px;">0</td> </tr> <tr> <td style="padding: 2px;">1</td> <td style="padding: 2px;">21</td> <td style="padding: 2px;">2</td> </tr> <tr> <td style="padding: 2px;">1</td> <td style="padding: 2px;">2</td> <td style="padding: 2px;">21</td> </tr> <tr> <td style="padding: 2px;">Fal</td> <td style="padding: 2px;">Mal</td> <td style="padding: 2px;">Viv</td> </tr> </table>	20	4	0	1	21	2	1	2	21	Fal	Mal	Viv
20	4	0													
1	21	2													
1	2	21													
Fal	Mal	Viv													
			Predicted												
Data set-3	Actual	Fal	<table border="1" style="margin: auto; border-collapse: collapse;"> <tr> <td style="padding: 2px;">22</td> <td style="padding: 2px;">1</td> <td style="padding: 2px;">1</td> </tr> <tr> <td style="padding: 2px;">2</td> <td style="padding: 2px;">20</td> <td style="padding: 2px;">2</td> </tr> <tr> <td style="padding: 2px;">3</td> <td style="padding: 2px;">1</td> <td style="padding: 2px;">20</td> </tr> <tr> <td style="padding: 2px;">Fal</td> <td style="padding: 2px;">Mal</td> <td style="padding: 2px;">Viv</td> </tr> </table>	22	1	1	2	20	2	3	1	20	Fal	Mal	Viv
22	1	1													
2	20	2													
3	1	20													
Fal	Mal	Viv													
			Predicted												
Data set-4	Actual	Fal	<table border="1" style="margin: auto; border-collapse: collapse;"> <tr> <td style="padding: 2px;">21</td> <td style="padding: 2px;">0</td> <td style="padding: 2px;">3</td> </tr> <tr> <td style="padding: 2px;">1</td> <td style="padding: 2px;">22</td> <td style="padding: 2px;">1</td> </tr> <tr> <td style="padding: 2px;">3</td> <td style="padding: 2px;">2</td> <td style="padding: 2px;">19</td> </tr> <tr> <td style="padding: 2px;">Fal</td> <td style="padding: 2px;">Mal</td> <td style="padding: 2px;">Viv</td> </tr> </table>	21	0	3	1	22	1	3	2	19	Fal	Mal	Viv
21	0	3													
1	22	1													
3	2	19													
Fal	Mal	Viv													
			Predicted												
Data set-5	Actual	Fal	<table border="1" style="margin: auto; border-collapse: collapse;"> <tr> <td style="padding: 2px;">22</td> <td style="padding: 2px;">1</td> <td style="padding: 2px;">1</td> </tr> <tr> <td style="padding: 2px;">1</td> <td style="padding: 2px;">21</td> <td style="padding: 2px;">2</td> </tr> <tr> <td style="padding: 2px;">4</td> <td style="padding: 2px;">2</td> <td style="padding: 2px;">18</td> </tr> <tr> <td style="padding: 2px;">Fal</td> <td style="padding: 2px;">Mal</td> <td style="padding: 2px;">Viv</td> </tr> </table>	22	1	1	1	21	2	4	2	18	Fal	Mal	Viv
22	1	1													
1	21	2													
4	2	18													
Fal	Mal	Viv													
			Predicted												

Tabel Pengukuran Data *Training Gaussian SVM*

Dataset-n	Akurasi	Sensitivitas	Spesifisitas
Dataset 1	83.3	83.33	91.67
Dataset 2	86.1	83.33	95.83
Dataset 3	86.1	91.67	89.58
Dataset 4	86.1	87.5	91.67
Dataset 5	84.7	91.67	89.58
Rata-rata	85.26	87.50	91.67

Data testing Polynomial Support Vector Machine

Data Aktual	Dataset 1	Dataset 2	Dataset 3	Dataset 4	Dataset 5
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'VIVAX'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'VIVAX'	'VIVAX'	'VIVAX'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'VIVAX'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'VIVAX'	'FALCIPARUM'
'MALARIAE'	'MALARIAE'	'VIVAX'	'VIVAX'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'FALCIPARUM'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'
'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'MALARIAE'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'

Nama Kelas	Nama Label Citra Dataset Pengujian				
'FALCIPARUM'	fal (25).bmp	fal (19).bmp	fal (13).bmp	fal (10).bmp	fal (1).bmp
'FALCIPARUM'	fal (26).bmp	fal (20).bmp	fal (14).bmp	fal (11).bmp	fal (2).bmp
'FALCIPARUM'	fal (27).bmp	fal (21).bmp	fal (15).bmp	fal (12).bmp	fal (3).bmp
'FALCIPARUM'	fal (28).bmp	fal (22).bmp	fal (16).bmp	fal (7).bmp	fal (4).bmp
'FALCIPARUM'	fal (29).bmp	fal (23).bmp	fal (17).bmp	fal (8).bmp	fal (5).bmp
'FALCIPARUM'	fal (30).bmp	fal (24).bmp	fal (18).bmp	fal (9).bmp	fal (6).bmp
'MALARIAE'	mal (25).bmp	mal (19).bmp	mal (13).bmp	mal (10).bmp	mal (1).bmp
'MALARIAE'	mal (26).bmp	mal (20).bmp	mal (14).bmp	mal (11).bmp	mal (2).bmp
'MALARIAE'	mal (27).bmp	mal (21).bmp	mal (15).bmp	mal (12).bmp	mal (3).bmp
'MALARIAE'	mal (28).bmp	mal (22).bmp	mal (16).bmp	mal (7).bmp	mal (4).bmp
'MALARIAE'	mal (29).bmp	mal (23).bmp	mal (17).bmp	mal (8).bmp	mal (5).bmp
'MALARIAE'	mal (30).bmp	mal (24).bmp	mal (18).bmp	mal (9).bmp	mal (6).bmp
'VIVAX'	viv (25).bmp	viv (19).bmp	viv (13).bmp	viv (10).bmp	viv (1).bmp
'VIVAX'	viv (26).bmp	viv (20).bmp	viv (14).bmp	viv (11).bmp	viv (2).bmp
'VIVAX'	viv (27).bmp	viv (21).bmp	viv (15).bmp	viv (12).bmp	viv (3).bmp
'VIVAX'	viv (28).bmp	viv (22).bmp	viv (16).bmp	viv (7).bmp	viv (4).bmp
'VIVAX'	viv (29).bmp	viv (23).bmp	viv (17).bmp	viv (8).bmp	viv (5).bmp
'VIVAX'	viv (30).bmp	viv (24).bmp	viv (18).bmp	viv (9).bmp	viv (6).bmp

Data testing Linier Support Vector Machine

Data Aktual	Dataset 1	Dataset 2	Dataset 3	Dataset 4	Dataset 5
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'VIVAX'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'VIVAX'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'VIVAX'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'MALARIAE'	'MALARIAE'	'VIVAX'	'VIVAX'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'FALCIPARUM'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'
'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'	'FALCIPARUM'
'VIVAX'	'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'

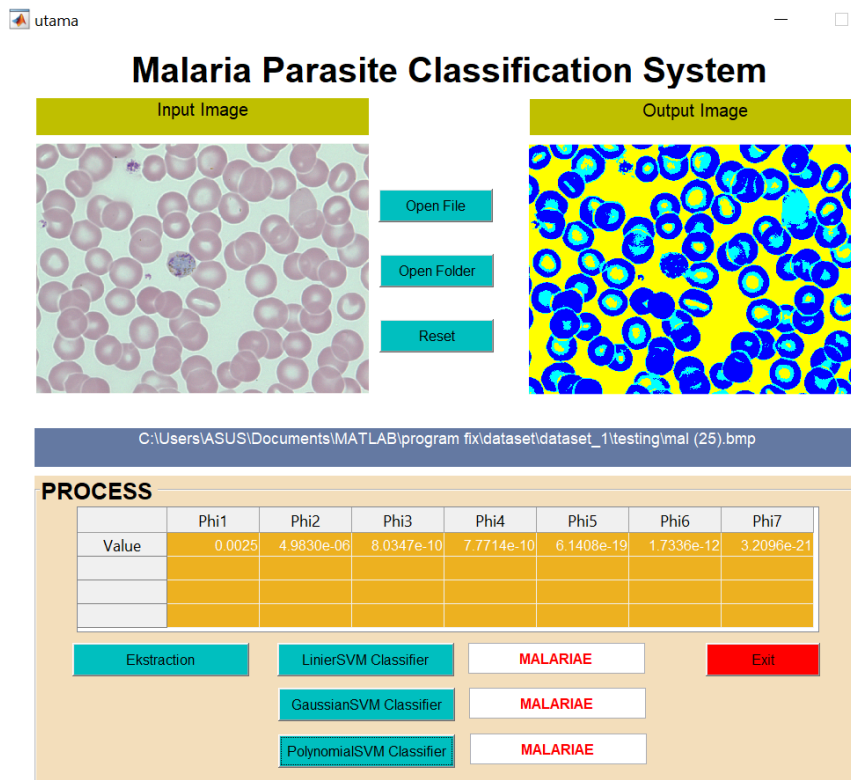
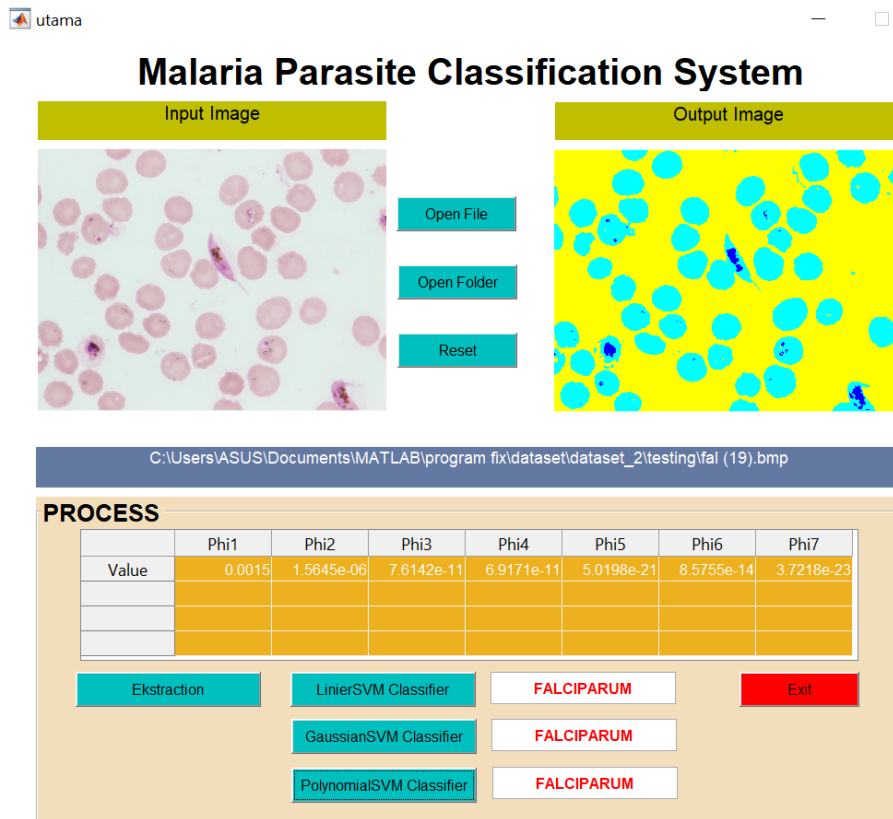
Nama Kelas	Nama Label Citra Dataset Pengujian				
'FALCIPARUM'	fal (25).bmp	fal (19).bmp	fal (13).bmp	fal (10).bmp	fal (1).bmp
'FALCIPARUM'	fal (26).bmp	fal (20).bmp	fal (14).bmp	fal (11).bmp	fal (2).bmp
'FALCIPARUM'	fal (27).bmp	fal (21).bmp	fal (15).bmp	fal (12).bmp	fal (3).bmp
'FALCIPARUM'	fal (28).bmp	fal (22).bmp	fal (16).bmp	fal (7).bmp	fal (4).bmp
'FALCIPARUM'	fal (29).bmp	fal (23).bmp	fal (17).bmp	fal (8).bmp	fal (5).bmp
'FALCIPARUM'	fal (30).bmp	fal (24).bmp	fal (18).bmp	fal (9).bmp	fal (6).bmp
'MALARIAE'	mal (25).bmp	mal (19).bmp	mal (13).bmp	mal (10).bmp	mal (1).bmp
'MALARIAE'	mal (26).bmp	mal (20).bmp	mal (14).bmp	mal (11).bmp	mal (2).bmp
'MALARIAE'	mal (27).bmp	mal (21).bmp	mal (15).bmp	mal (12).bmp	mal (3).bmp
'MALARIAE'	mal (28).bmp	mal (22).bmp	mal (16).bmp	mal (7).bmp	mal (4).bmp
'MALARIAE'	mal (29).bmp	mal (23).bmp	mal (17).bmp	mal (8).bmp	mal (5).bmp
'MALARIAE'	mal (30).bmp	mal (24).bmp	mal (18).bmp	mal (9).bmp	mal (6).bmp
'VIVAX'	viv (25).bmp	viv (19).bmp	viv (13).bmp	viv (10).bmp	viv (1).bmp
'VIVAX'	viv (26).bmp	viv (20).bmp	viv (14).bmp	viv (11).bmp	viv (2).bmp
'VIVAX'	viv (27).bmp	viv (21).bmp	viv (15).bmp	viv (12).bmp	viv (3).bmp
'VIVAX'	viv (28).bmp	viv (22).bmp	viv (16).bmp	viv (7).bmp	viv (4).bmp
'VIVAX'	viv (29).bmp	viv (23).bmp	viv (17).bmp	viv (8).bmp	viv (5).bmp
'VIVAX'	viv (30).bmp	viv (24).bmp	viv (18).bmp	viv (9).bmp	viv (6).bmp

Data testing Gaussian Support Vector Machine

Data Aktual	Dataset 1	Dataset 2	Dataset 3	Dataset 4	Dataset 5
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'VIVAX'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'	'FALCIPARUM'
'MALARIAE'	'MALARIAE'	'FALCIPARUM'	'MALARIAE'	'MALARIAE'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'FALCIPARUM'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'VIVAX'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'
'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'	'MALARIAE'
'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'FALCIPARUM'	'VIVAX'	'VIVAX'	'FALCIPARUM'
'VIVAX'	'VIVAX'	'VIVAX'	'FALCIPARUM'	'MALARIAE'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'MALARIAE'	'VIVAX'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'MALARIAE'	'VIVAX'
'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'	'VIVAX'

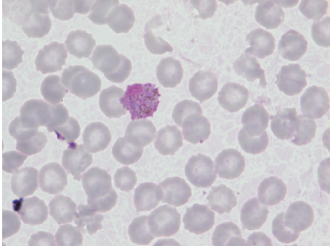
Nama Kelas	Nama Label Citra Dataset Pengujian				
'FALCIPARUM'	fal (25).bmp	fal (19).bmp	fal (13).bmp	fal (10).bmp	fal (1).bmp
'FALCIPARUM'	fal (26).bmp	fal (20).bmp	fal (14).bmp	fal (11).bmp	fal (2).bmp
'FALCIPARUM'	fal (27).bmp	fal (21).bmp	fal (15).bmp	fal (12).bmp	fal (3).bmp
'FALCIPARUM'	fal (28).bmp	fal (22).bmp	fal (16).bmp	fal (7).bmp	fal (4).bmp
'FALCIPARUM'	fal (29).bmp	fal (23).bmp	fal (17).bmp	fal (8).bmp	fal (5).bmp
'FALCIPARUM'	fal (30).bmp	fal (24).bmp	fal (18).bmp	fal (9).bmp	fal (6).bmp
'MALARIAE'	mal (25).bmp	mal (19).bmp	mal (13).bmp	mal (10).bmp	mal (1).bmp
'MALARIAE'	mal (26).bmp	mal (20).bmp	mal (14).bmp	mal (11).bmp	mal (2).bmp
'MALARIAE'	mal (27).bmp	mal (21).bmp	mal (15).bmp	mal (12).bmp	mal (3).bmp
'MALARIAE'	mal (28).bmp	mal (22).bmp	mal (16).bmp	mal (7).bmp	mal (4).bmp
'MALARIAE'	mal (29).bmp	mal (23).bmp	mal (17).bmp	mal (8).bmp	mal (5).bmp
'MALARIAE'	mal (30).bmp	mal (24).bmp	mal (18).bmp	mal (9).bmp	mal (6).bmp
'VIVAX'	viv (25).bmp	viv (19).bmp	viv (13).bmp	viv (10).bmp	viv (1).bmp
'VIVAX'	viv (26).bmp	viv (20).bmp	viv (14).bmp	viv (11).bmp	viv (2).bmp
'VIVAX'	viv (27).bmp	viv (21).bmp	viv (15).bmp	viv (12).bmp	viv (3).bmp
'VIVAX'	viv (28).bmp	viv (22).bmp	viv (16).bmp	viv (7).bmp	viv (4).bmp
'VIVAX'	viv (29).bmp	viv (23).bmp	viv (17).bmp	viv (8).bmp	viv (5).bmp
'VIVAX'	viv (30).bmp	viv (24).bmp	viv (18).bmp	viv (9).bmp	viv (6).bmp

Tampilan GUI untuk klasifikasi PF, PM dan PV



Malaria Parasite Classification System

Input Image

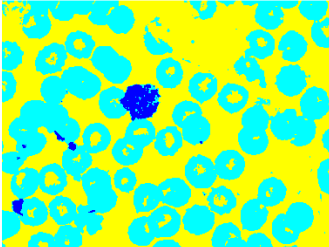


Open File

Open Folder

Reset

Output Image



C:\Users\ASUS\Documents\MATLAB\program fix\dataset\dataset_1\testing\viv (26).bmp

PROCESS

	Phi1	Phi2	Phi3	Phi4	Phi5	Phi6	Phi7
Value	0.0016	1.8196e-06	7.8486e-13	4.7426e-13	2.8931e-25	4.3245e-16	-5.0626e-27

Ekstraktion

LinierSVM Classifier

GaussianSVM Classifier

PolynomialSVM Classifier

VIVAX

VIVAX

VIVAX

Exit

Script Program Sistem Klasifikasi Penyakit Malaria dengan Metode Hu Moment dan Support Vector Machine

A. Script Program Utama

```
%% ----- pilih citra testing-----%%
function pushbutton1_Callback(hObject, eventdata, handles)
[fname, path]=uigetfile('.bmp','pilih citra testing');
fname=strcat(path, fname);
image=imread(fname);
axes(handles.axes1)
imshow(image);
handles.image = image;
guidata(hObject,handles);
set(handles.text3,'string',fname);

%% ----- pilih folder training-----%%
function pushbutton2_Callback(hObject, eventdata, handles)
folder2 = uigetdir('C:\Users','open folder training');
set(handles.text3,'string',folder2);

%% ----- Reset-----%%
function pushbutton3_Callback(hObject, eventdata, handles)
cla (handles.axes1,'reset');
cla (handles.axes2,'reset');
set(findobj('style','pushbutton'),'Value', 0);
set(findobj('style','edit'),'String', {''});
set(handles.text3,'string','');
set(handles.uitable1,'Data','');

%% ----- ekstraksi fitur-----%%
function pushbutton4_Callback(hObject, eventdata, handles)
folder = get(handles.text3,'string');
images = dir([folder,'\*.bmp']);
nimages = length(images);
nimages
B=zeros(nimages,7);
    for i = 1:nimages
        currentimagename = images(i).name;
        imagenames(i,:) = {currentimagename};
        currentimage=imread(fullfile(folder,currentimagename),'bmp');
        axes(handles.axes1);
        imshow(currentimage);
        axes(handles.axes2);
        imshow(currentimage);
        phi = invmoments(currentimage)
        set(handles.uitable1,'Data',(phi));
        B(i,:)=phi;
    end
```

```

sheet=1;
filename='AlldataSeg.xlsx';
xlswrite(filename,B,sheet,'A1:G90');
xlswrite(filename,imagenames,sheet,'I1:I90');

%% ----- linierSVM classifier-----%%
function pushbutton5_Callback(hObject, eventdata, handles)
image = handles.image;
axes(handles.axes2);
phi = invmoments(image)
set(handles.uitable1,'Data',(phi));
VarName1 = phi(1);
VarName2 = phi(2);
VarName3 = phi(3);
VarName4 = phi(4);
VarName5 = phi(5);
VarName6 = phi(6);
VarName7 = phi(7);
phi2 =
table(VarName1,VarName2,VarName3,VarName4,VarName5,VarName6,VarName7);
%-----load dataset-----
load('dataSeg.mat');
trainingData = try2;
inputTable = trainingData;
predictorNames = {'VarName1', 'VarName2', 'VarName3', 'VarName4',
'VarName5', 'VarName6', 'VarName7'};
predictors = inputTable(:, predictorNames);
response = inputTable.FALCIPARUM;
%-----latih classifier-----
template = templateSVM(...
'KernelFunction', 'linear', ...
'PolynomialOrder', [], ...
'KernelScale', 'auto', ...
'BoxConstraint', 1, ...
'Standardize', true);
classificationSVM = fitcecoc(...
predictors, ...
response, ...
'Learners', template, ...
'Coding', 'onevsone', ...
'ClassNames', {'FALCIPARUM'; 'MALARIAE'; 'VIVAX'});
%-----predict-----
predictedlabels = predict(classificationSVM, phi2);
set(handles.edit1,'string',predictedlabels);
f = msgbox(predictedlabels,'your classification');

%% ----- gaussianSVM classifier-----%%
function pushbutton7_Callback(hObject, eventdata, handles)

```

```

image = handles.image;
axes(handles.axes2);
phi = invmoments(image)
set(handles.uitable1, 'Data', (phi));
VarName1 = phi(1);
VarName2 = phi(2);
VarName3 = phi(3);
VarName4 = phi(4);
VarName5 = phi(5);
VarName6 = phi(6);
VarName7 = phi(7);
phi2 =
table(VarName1,VarName2,VarName3,VarName4,VarName5,VarName6,VarName7);
% -----load dataset-----
load('dataSeg.mat');
trainingData = try2;
inputTable = trainingData;
predictorNames = {'VarName1', 'VarName2', 'VarName3', 'VarName4',
'VarName5', 'VarName6', 'VarName7'};
predictors = inputTable(:, predictorNames);
response = inputTable.FALCIPARUM;
% -----latih classifier-----
template = templateSVM('KernelFunction',...
    'gaussian',...
    'PolynomialOrder', [],...
    'KernelScale', 0.66,...
    'BoxConstraint', 1,...
    'Standardize', true);
classificationSVM = fitcecoc(predictors,...
    response,...
    'Learners', template,...
    'Coding', 'onevsone',...
    'ClassNames', {'FALCIPARUM'; 'MALARIAE'; 'VIVAX'});
% -----predict-----
predictedlabels = predict(classificationSVM, phi2);
set(handles.edit2, 'string', predictedlabels);
f = msgbox(predictedlabels, 'your classification');

%% ----- polynomialSVM classifier-----%%
function pushbutton8_Callback(hObject, eventdata, handles)
image = handles.image;
axes(handles.axes2);
phi = invmoments(image)
set(handles.uitable1, 'Data', (phi));
VarName1 = phi(1);
VarName2 = phi(2);
VarName3 = phi(3);
VarName4 = phi(4);

```

```

VarName5 = phi(5);
VarName6 = phi(6);
VarName7 = phi(7);
phi2 =
table(VarName1,VarName2,VarName3,VarName4,VarName5,VarName6,VarName7);
%-----load data-----
load ('dataSeg.mat');
trainingData = try2;
inputTable = trainingData;
predictorNames = {'VarName1', 'VarName2', 'VarName3', 'VarName4',
'VarName5', 'VarName6', 'VarName7'};
predictors = inputTable(:, predictorNames);
response = inputTable.FALCIPARUM;
%-----latih model-----
template = templateSVM(...
    'KernelFunction', 'polynomial', ...
    'PolynomialOrder', 2, ...
    'KernelScale', 'auto', ...
    'BoxConstraint', 1, ...
    'Standardize', true);
classificationSVM = fitcecoc(...
    predictors, ...
    response, ...
    'Learners', template, ...
    'Coding', 'onevsone', ...
    'ClassNames', {'FALCIPARUM'; 'MALARIAE'; 'VIVAX'});
%-----predict-----
predictedlabels = predict(classificationSVM, phi2);
set(handles.edit3,'string',predictedlabels);
f = msgbox(predictedlabels,'your classification');

```

B. Script Program Pendukung (hu moment)

```

function phi = invmoments(image)
I = rgb2gray(image);
thresh = multithresh(I,2);
seg_I = imquantize(I,thresh);
image = label2rgb(seg_I);
imshow(image);
% image = rgb2hsv(image);
% imshow(image);
image = double(image);
phi= compute_phi(compute_eta(compute_m(image)));

%-----
--%
function m = compute_m(image)

```

```

[M, N] = size(image);
[x, y] = meshgrid(1:N, 1:M);

% Turn x, y, and F into column vectors to make the summations a
bit
% easier to compute in the following.
x = x(:);
y = y(:);
image = image(:);

% DIP equation (11.3-12)
m.m00 = sum(image);
% Protect against divide-by-zero warnings.
if (m.m00 == 0)
    m.m00 = eps;
end
% The other central moments:
m.m10 = sum(x .* image);
m.m01 = sum(y .* image);
m.m11 = sum(x .* y .* image);
m.m20 = sum(x.^2 .* image);
m.m02 = sum(y.^2 .* image);
m.m30 = sum(x.^3 .* image);
m.m03 = sum(y.^3 .* image);
m.m12 = sum(x .* y.^2 .* image);
m.m21 = sum(x.^2 .* y .* image);

%-----
--%
function e = compute_eta(m)

% DIP equations (11.3-14) through (11.3-16).

xbar = m.m10 / m.m00;
ybar = m.m01 / m.m00;

e.eta11 = (m.m11 - ybar*m.m10) / m.m00^2;
e.eta20 = (m.m20 - xbar*m.m10) / m.m00^2;
e.eta02 = (m.m02 - ybar*m.m01) / m.m00^2;
e.eta30 = (m.m30 - 3 * xbar * m.m20 + 2 * xbar^2 * m.m10) /
m.m00^2.5;
e.eta03 = (m.m03 - 3 * ybar * m.m02 + 2 * ybar^2 * m.m01) /
m.m00^2.5;
e.eta21 = (m.m21 - 2 * xbar * m.m11 - ybar * m.m20 + ...
2 * xbar^2 * m.m01) / m.m00^2.5;
e.eta12 = (m.m12 - 2 * ybar * m.m11 - xbar * m.m02 + ...
2 * ybar^2 * m.m10) / m.m00^2.5;

```