

LAB ASSIGNMENT – IV

Data classification using K-nearest neighbor classifier and Bayes classifier with unimodal Gaussian density

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1 a.

	Prediction Outcome	
Label	81	27
True	27	201

Figure 1 KNN Confusion Matrix for K = 1

	Prediction Outcome		
Label	83	25	
True	12	216	

Figure 2 KNN Confusion Matrix for K = 3



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	Prediction Outcome	
Label	82	26
True	9	219

Figure 3 KNN Confusion Matrix for K = 5

b.

Table 1 KNN Classification Accuracy for K = 1, 3 and 5

	Classification
K	Accuracy (in %)
1	83.92
3	88.98
5	89.58

Inferences:

- 1. The highest classification accuracy is obtained with K =5
- 2. The increasing value of K increases the prediction accuracy.
- 3. More the number of nearest neighbors, better the analysis hence accuracy increases.
- 4. The diagonal elements (true values) increase with increasing k.
- 5. As the accuracy increases, so does the number of true predictions .
- 6. The off-diagonal elements decrease with higher k values.
- 7. The increase in accuracy with k decreases the number of false predictions.



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2 a.

	Prediction Outcome	
Label	100	8
True	8	220

Figure 4 KNN Confusion Matrix for K = 1 post data normalization

	Prediction Outcome	
Label	100	8
True	7	221

Figure 5 KNN Confusion Matrix for K = 3 post data normalization

	Prediction	Outcome
True	101	7



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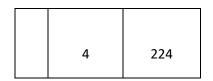


Figure 6 KNN Confusion Matrix for K = 5 post data normalization

b.

Table 2 KNN Classification Accuracy for K = 1, 3 and 5 post data normalization

	Classification
K	Accuracy (in %)
1	95.23
3	95.53
5	96.72

Inferences:

- 1. Normalisation increases accuracy.
- 2. Normalising data eliminates bias due to uneven range of different attributes.
- 3. The highest classification accuracy is obtained with K = 5.
- 4. More the number of nearest neighbors, better the analysis hence accuracy increases.
- 5. The diagonal elements (true values) increase with increasing k.
- 6. As the accuracy increases, so does the number of true predictions .
- 7. The off-diagonal elements decrease with higher k values.
- 8. The increase in accuracy with k decreases the number of false predictions.

3

		iction come
True	96	12



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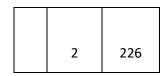


Figure 7 Confusion Matrix obtained from Bayes Classifier

The classification accuracy obtained from Bayes Classifier is 95.83%.

Table 3 Mean for class 0 and class 1

S. No.	Attribute Name	Mean	
		Class 0	Class 1
1.	X_Maximum	286.3321554770318	746.584
2.	Y_Maximum	1711478.0565371024	1445963.75
3.	Pixels_Areas	7268.031802120141	583.512
4.	X_Perimeter	355.6148409893993	52.184
5.	Y_Perimeter	207.15547703180212	43.112
6.	Sum_of_Luminosity	808615.6925795053	61552.412
7.	Minimum_of_Luminosity	53.40282685512368	94.804
8.	Maximum of Luminosity	135.85865724381625	130.184
9.	Length of Conveyer	1382.5159010600707	1486.63
10.	Steel_Plate_Thickness	40.24734982332156	100.434
11.	Edges_Index	0.12644699646643126	0.3888644
12.	Empty_Index	0.44960777385159006	0.4186427999999997
13.	Square Index	0.5932530035335692	0.5103224
14.	Outside_X_Index	0.10817279151943467	0.019853799999999998
15.	Edges_X_Index	0.5658508833922259	0.6256006000000003
16.	Edges_Y_Index	0.5246918727915195	0.8374429999999995
17.	Outside_Global_Index	0.26855123674911663	0.611
18.	LogOfAreas	3.599567137809189	2.2643111999999976
19.	Log X Index	2.048011307420494	1.214075400000001
20.	Log_Y_Index	1.825002826855123	1.2994936
21.	Orientation_Index	-0.3280713780918728	0.131946
22.		-0.10907385159010606	-0.12263200000000007
	Luminosity_Index		
23.	SigmoidOfAreas	0.9158699646643109	0.5270244



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In Fig. 8 and 9 representing covariance matrices for class 0 and class 1 respectively the column numbers and row numbers correspond to attribute with serial number as in Table 3.

X, Maximum Y, Maximum Pixels, Areas X, Perimeter Y, Perimeter Sum, of Lurr/Minimum_of Maximum_of Length_of GSteel_Plate_[Edges_Index Empty_Index Square_Index Outside_X_In Edges_Y_Ind Outside_Glo LogofAreas_Log_X_Index Log_Y_Index Orientation_Luminosity_I SigmoidOfAr 57593.634 -86700603 -349303.67 -15539.489 -8064.1972 -38068098 4246.33381 2211.84499 2606.64364 2047.40245 26.1726577 -9.7545065 7.64282701 -2.2299973 20.4543405 28.0076481 34.6356602 -87.729562 -55.971493 -35.525749 32.6632543 19.3710044 -33.459881 -86700603 2.6343E+12 | -754411187 | -38584596 | -29519236 | -29519236 | -9.799E+10 | -4665083.7 | -8053964.8 | -10797723 | -325686.4 | -55558.111 | 14531.1097 | -93632.707 | 3191.98492 | 6707.55061 | -38653.707 | -133538.2 | 183163.591 | 137803.088 | 46364.0404 | -141235.62 | -57051.381 | 95439.9397 -349303.67 -754411187 | 28362934.2 | 1395370.66 | 857469.928 | 3373269717 | -130039.3 | -4383.8856 | 30347.1821 | -158.48307 | -476.93692 | 368.752327 | 529.978051 | 228.204148 | -931.49974 | -654.23987 | 290.156323 | 2816.52507 | 1451.62832 | 1686.8919 | 371.995881 | -158.52226 | 605.051105 -45539.489 -38584596 1395370.66 | 74685.8334 45819.8402 | 166734113 -6114.6244 45.1368569 | 2140.32352 | 1.3720259 - 22.568996 | 22.2883679 | 32.9466804 | 11.6106668 | -52.151251 | -33.651946 | 22.9282698 | 135.711025 | 69.4065827 | 86.5152844 | 26.9798362 | -5.8278697 | 28.8341696-8064.1972 -29519236 857469.928 45819.8402 28599.2594 103157362 -3579.2863 186.007869 1535.58263 -4.6130617 -12.424968 13.3740311 22.3889453 6.61831524 -32.587 -19.54701 19.0112899 79.7231402 39.1879972 52.7242425 20.915304 -2.3370094 16.3945224 -38068098 -9.7995 ± 10 3373269717 166734113 103157362 4.03485 ± 11 -14678207 10270.4919 3727267.82 -38801.98 -53411.29 43540.7935 6945.529 26038.0089 -112302.15 -74739.591 44593.8719 321540.321 162501.462 197432.138 54471.5075 -14263.373 67039.1397424633381 4665083.7 -13009.3 -6114.6244 -3579.2863 -14678207 1435.62439 454.163522 -143.80075 -2.688645 4.15137072 -2.0602085 1.1099701 -1.5074099 4.21781454 4.82591392 3.3045573 -23.060082 -13.287033 -11.310914 2.99730935 4.69163446 -7.15036342211.84499 -8053964.8 -4383.8856 45.1368569 186.007869 10270.4919 454.163522 359.476405 -7.7353322 -7.2698795 1.95865773 -0.3497528 2.29322064 -0.3561766 -0.052436 1.56550878 3.83951081 -6.0902196 -4.447025 -1.7853542 3.9526147 2.95131967 -2.9104642 2606.64364 - 10797723 30347.1821 2140.32352 1535.58263 3727267.82 - 143.80075 - 7.7353332 2489.1017 40.5811593 1.08805262 0.40379668 3.90272256 - 0.2913207 - 2.6184306 0.08847087 4.97798411 1.11010496 - 0.9431204 2.47784605 5.15358234 - 0.4766416 0.07951839 204.740245 - 325686.4 - 158.48307 | 1.37220259 - 4.6130617 | -38801.98 | -2.688645 - 7.2698795 | 40.5811593 | 6.67618976 | -0.022884 | -0.0183317 | -0.0003323 | 0.00704221 | 0.01551574 | 0.0425024 | 0.07518232 | -0.0511798 | -0.0434922 | -0.0117702 | 0.06357091 | -0.0547973 | 0.01641526 26.1726577 -55558.111 -476.93692 -22.568996 -12.424968 -53411.29 | 4.15137072 | 1.95865773 | 1.08805262 -0.022884 | 0.03137594 | -0.0166964 | 0.00844312 | -0.0065201 | 0.01694349 | 0.02476217 | 0.02510613 | -0.0894748 | -0.0572308 | -0.0401421 | 0.02475006 | 0.01714353 | -0.0303127 | -0.006061 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 | -0.006661 -9.7545065 | 14531.1097 | 368.752327 | 22.2883679 | 13.3740311 | 43540.7935 | -2.0602085 | -0.3497528 | 0.40379668 | -0.0183317 | -0.0106964 | 0.01587908 | 0.0031618 | 0.00588363 | -0.0171605 | -0.0149049 | -0.0015512 | 0.05516606 | 0.03518784 | 0.03445395 | -0.000615 | -0.0044721 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.01697751 | 0.07.64282701 -93632.707 529.978051 32.9466804 22.3899453 69465.529 1.11099701 2.29322064 3.90272256 -0.0003231 0.00849312 0.00849312 0.00849312 0.00849312 0.008693791 0.00184816 0.07012421 -0.0070323 0.0024745 0.02477526 0.07252356 0.01620267 -0.0134611-2.2299973 3191.98492 | 228.204148 | 11.6106668 | 6.61831524 | 26038.0089 | -1.5074099 | -0.3561766 | -0.2913207 | 0.00704221 | -0.0065201 | 0.00588363 | -0.0046098 | 0.0051917 | -0.0026865 | -0.0078948 | -0.0087671 | 0.03156268 | 0.02265726 | 0.01548856 | -0.0093191 | -0.0039082 | 0.00842183 -32.587 | -112302.15 | 4.21781454 | -0.052436 | -2.6184306 | 0.01551574 | 0.01694349 | -0.0171605 | -0.0367916 | -0.026865 | 0.05762837 | 0.02655619 | -0.035449 | -0.1038782 | -0.0436831 | -0.072027 | -0.0402731 | 0.00384707 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | -0.0268723 | 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-0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.034963 | -0.0 34.6356602 -133538.2 | 29.156323 | 22.9282698 | 19.0112899 | 44593.8719 | 3.3045573 | 3.83951081 | 4.97798411 | 0.07518232 | 0.02510613 | -0.0015512 | 0.0701421 | -0.0087671 | -0.035449 | 0.02144634 | 0.19358194 | -0.0481752 | -0.0655075 | 0.01660367 | 0.12789406 | 0.02863604 | -0.0297291 -87.729562 | 183163.591 | 2816.52507 | 135.711025 | 79.7231402 | 321540.321 | -23.060082 | -6.0902196 | 1.11010496 | -0.0511798 | -0.0894748 | -0.0551606 | -0.020323 | -0.03156268 | -0.1038782 | -0.108107 | -0.0481752 | -0.49708655 | -0.2844205 | -0.2571164 | -0.0451129 | -0.0658451 | -0.14708501-55.971493 137803.088 1451.62832 69.4065827 39.1879972 162501.462 -13.287033 -4.447025 -0.0434922 -0.0434922 -0.0518784 -0.0242445 -0.0245726 -0.046831 -0.0667494 -0.0655075 -0.284205 -0.17867695 -0.1343316 -0.0642812 -0.0445669 -0.08865325-35.525749 46364.0404 1686.8919 86.5152844 52.7242425 197432.138 -11.310914 1.7853542 2.47784605 -0.011702 -0.0401421 0.03445395 0.02427526 0.01548856 -0.072027 -0.0527923 0.01660367 0.25371164 0.1343316 0.14662905 0.01841134 -0.0247914 0.07034317 32.6632543 -141235.62 371.995881 26.9798362 20.915304 54471.5075 2.99730935 3.9526147 5.15358234 0.06357091 0.02475006 -0.000615 0.07252356 -0.0093191 -0.0402731 0.02018127 0.12789406 -0.0451129 -0.0642812 0.01841134 0.12295571 0.02940416 -0.028301 19.3710044 -57051.381 -158.52226 -5.8278697 -2.3370094 -14263.373 4.69165446 2.95131967 -0.4766416 -0.0547973 0.01714353 -0.0044721 0.01620267 -0.0039082 0.00384707 0.01540362 0.02863604 -0.0668461 -0.0445669 -0.0247914 0.02940416 0.02583603 -0.0276008 -33.459881 95439.9397 605.051105 28.8341696 16.3945224 67039.1397 -7.1503634 -2.9104642 0.07951839 0.01641526 -0.0303127 0.01697751 -0.0134611 0.00842183 -0.0268723 -0.0334963 -0.0297291 0.14708501 0.08863525 0.07034317 -0.0282501 -0.027291 -0.027291 -0.0282501 -0.027291 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 -0.0282501 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For better view open 'c0.csv' in zip

Figure 8: Covariance matrix for class 0



LAB ASSIGNMENT - IV

Data classification using K-nearest neighbor classifier and Bayes classifier with unimodal Gaussian density

X Maximum! Y Maximum Pixels Areas X Perimeter Y Perimeter Sum of Lurr Minimum of Maximum of Length of OSteel Plate Edges Index Empty Index Square Inde Outside X In Edges X Ind Edges Y Ind Outside Glol LogOfAreas Log X Index Log Y Index Orientation | Luminosity I SigmoidOfAr 258038.019 | 148247097 | 19263.781 | 261359263 | 1901.4483 | 2032754.1 | 1183.6488 | 1180.0175 | 12247.3588 | 2832.3161 | 3.8962885 | 2.4625913 | 11.6303512 | 1.20702483 | 8.39175576 | 4.327484 | 1.0083992 | 15.464289 | 1.16597031 | 1.8.575736 | 2.3.356581 | -10.16544 | 1.4.897115 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66888 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 1.2.66688 | 148247097 3.2968E+12 506836443 29140899.8 9302067.87 5.296E+10 3586455.2 600089.513 -1305440.9 -34372254 36534.7689 -16500.843 -26651.402 18243.7641 54437.8942 -29077.053 -74054.72 74366.2716 89905.8487 -28476.239 -11664.65 -13913.783 -2798.6542 $-19263.781 \quad 506336443 \quad 5121723.53 \quad 201881.166 \quad 135506.594 \quad 532831804 \quad -15218.092 \quad 2762.6531 \quad -29026.474 \quad 2315.24829 \quad -37.452553 \quad 31.8358468 \quad -107.65827 \quad 69.8587161 \quad -87.930551 \quad -125.61698 \quad 30.5783246 \quad 692.874031 \quad 377.582817 \quad 342.899992 \quad 17.3937558 \quad -31.139554 \quad 225.094957 \quad -32.67827 \quad -32.6782$ 261.359263 29140899.8 201881.166 10847.8258 5755.1056 21160999.6 541.64523 203.950044 -2125.8236 185.314774 -0.3720684 3.60420554 -7.9980781 4.80708988 -4.1746181 -10.055025 -3.3170581 37.998743 24.9050097 16.1432399 -5.7225648 -1.0182258 15.1152837 -1901.4483 9302067.87 | 135506.594 | 5755.1056 | 5008.4724 | 14025223.9 -538.58321 | -23.24109 | -1229.6879 | 313.849078 | -1.3455593 | 2.599497 | -6.4127763 | 1.40340759 | -8.1855518 | -2.7111152 | 6.3557856 | 28.17901 | 10.6683422 | 19.7474841 | 9.9134259 | 1.4954339 | 12.2662191 -2032754.1 5.296E+10 532831804 21160999.6 14025223.9 5.5619E+10 -1443015.2 397726.778 -3291478.1 147379.793 -3554.681 3415.54063 -11365.915 7414.94048 -8940.0169 -13523.878 2549.55838 71815.0233 39675.6919 35077.3895 888.039931 -2320.9565 23386.7242 -1183.6488 -3586455.2 -15218.092 | 541.64523 | -538.58321 | -1443015.2 | 775.075735 | 358.481026 | -1115.2991 | -263.23941 | 125855413 | 0.7646066 | 0.29934789 | -0.1577692 | 0.23707928 | -1.2046988 | -2.8329098 | -4.855393 | -1.1122046 | -3.1843305 | -2.8037773 | 3.94402157 | -1.9063732 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -1.2046988 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.803779 | -2.8037 -1180.0175 600089513 2762.6531 203.950044 -23.24109 397726.778 358.481026 454.202549 -543.78148 -252.573 0.64891137 -0.0336688 -0.6267745 0.15812234 0.83431312 -1.4206789 -2.3621483 -0.8789942 1.21832317 -2.1113846 -3.4066518 -2.9141063 -0.709934412247.3588 - 13054409 - 29026.474 | -2125.8236 - 1229.6879 | -3291478.1 | -1115.2991 | -543.78148 | 24015.1835 | 1507.22102 | -0.812817 | -4.7055675 | 5.13441572 | -1.0397644 | 7.17455393 | 3.7878935 | 0.73654309 | -10.228312 | -4.3945841 | -9.485397 | -4.3959683 | -5.6947445 | -7.3639611 -2832.3161 -34372254 2315.24829 185.314774 313.843078 147379.793 -263.23941 -252.573 1507.22102 4839.48461 -1.6835653 0.49900403 -1.0898059 -0.1236502 -2.5489209 2.17793581 5.52186974 2.46303721 -1.6609719 4.56295689 7.27255755 -2.0202143 1.79776855 3.38962885 36534.7689 37.452553 -0.3720684 -1.3455593 -3554.681 1.25855413 0.64891137 -0.812817 -1.6835653 0.09152501 -0.0006192 0.000718962 0.0002905 0.00573767 -0.005633 -0.0180243 -0.0068285 0.0062196 -0.013269 -0.0253706 0.00602743 -0.0007349 -2.4629513 - 16500.843 | 31.8358468 | 3.60420554 | 2.59947 | 3415.54063 | 0.7646066 | -0.0336688 | 4.7055675 | 0.49900403 | -0.0006192 | 0.01925909 | -0.0041018 | 0.00101333 | -0.0127655 | -0.010491 | -0.007875 | 0.0215052 | 0.01803361 | 0.0189361 | -0.012524 | 0.0026876 | 0.0213124 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.007875 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001491 | -0.001116303512 - 26651402 | -107.65827 | -7.9980781 | -6.4127763 | -11365.915 | 0.29934789 | -0.6267745 | 5.13441572 | -1.0898059 | 0.00718962 | -0.0041018 | 0.07921745 | -0.0035408 | 0.02160555 | 0.01592075 | -0.0116421 | -0.0504217 | -0.0218775 | -0.031296 | -0.0166797 | -0.0011201 | -0.0267006 1.20702483 | 18243.7641 | 69.8587161 | 4.80708968 | 1.40340759 | 7414.94048 | -0.1577692 | 0.15812134 | -1.0397644 | -0.1236502 | 0.00029095 | 0.00101333 | -0.0035408 | 0.00307401 | 0.00195839 | -0.005908 | -0.005706 | 0.01366705 | 0.0129821 | 0.00149777 | -0.009452 | -0.002132 | 0.00493196 8.39175576 54437.8942 87930551 4.1746181 8.1855518 8940.0169 0.23707928 0.83431312 7.17455393 2.5489209 0.00573767 0.00127655 0.02160555 0.00195839 0.06479208 0.00136109 0.0660066 0.0602551 0.01281467 0.0028514 0.0104352 0.00366659 0.0422166 -4.1377484 -29077.053 -125.61698 -10.055025 -2.7111152 -13523.878 -1.2046988 -1.4206789 -3.7878935 -1.793581 -0.0056931 -0.01592075 -0.0059908 -0.0136109 -0.04840095 -0.06550003 -0.0269553 -0.0573723 -0.02195446 -0.08604196 -0.0992573 -0.0153896-10.083992 -74054.472 -0.5783246 -3.3170581 6.35527856 -2549.55838 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 -2.8329998 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8995.8487 377.582817 24.9050097 10.6683422 39675.6919 -1.1122046 1.21823217 -4.3945841 -1.6609719 0.0062196 0.01803361 -0.0218775 0.01298821 0.01281467 -0.0750401 0.1143878 0.11620742 0.0132104 -0.103683 0.00073869 0.06081141-18.575736 -28476.229 342.989992 16.1432399 19.7474841 35077.3895 -3.1843305 -2.1113846 -9.485397 4.56295689 -0.013269 -0.013269 -0.013269 -0.00149777 -0.0828514 -0.0219546 -0.1048117 -0.16374976 -0.01332104 -0.16536017 -0.16291606 -0.0175153 -0.0945976-23.356581 -116644.65 17.3937558 -5.7225648 9.9134259 888.039931 -2.8037773 -3.4066518 -4.3589683 7.2725755 -0.0253706 -0.00253706 -0.0027524 -0.0166797 -0.0094552 -0.1014352 0.08604196 -0.2306543 -0.06231963 -0.10683 -0.16291606 -0.30203792 -0.0213964 -0.03861205-10.16544 -13913.783 -31.139554 -1.0182258 -1.4954339 -2320.9565 3.94402157 2.9141063 -5.6947445 -2.0202143 0.00602743 0.0026876 -0.0011201 -0.0002132 0.0036659 -0.0092573 -0.0182531 -0.018757 0.00073869 -0.0175153 -0.0213964 0.0261085 -0.0085037-14.897115 -2798.6542 | 225.094957 | 15.1152837 | 12.2662191 | 233867242 | -1.9063732 | -0.7099344 | -7.3639611 | 1.79776855 | -0.0007343 | -0.021324 | -0.0267006 | -0.0422166 | -0.0153896 | -0.02057885 | -1.4027205 | -0.06081141 | -0.09645976 | -0.03861205 | -0.0085037 | -0.07979023

For better view open 'c1.csv' in zip

Figure 9: Covariance matrix for class 1

Inferences:

- 1. Write the accuracy of Bayes Classifier and state reason why it is lesser / greater than previous classification approaches.
- 2. Infer from covariance matrix the nature of values along the diagonal. State the reason.
- 3. Infer from off-diagonal elements the covariance between attributes. Write 2 pair of attributes with maximum and 2 pair of attributes with minimum covariance.



LAB ASSIGNMENT - IV

Data classification using K-nearest neighbor classifier and Bayes classifier with unimodal Gaussian density

Note: Please write diagonal values of covariance matrices in boldface. On moodle, the template for covariance matrix is uploaded as .docx and .xlsx format. Fill in the values and change the covariance matrices into images. Insert the covariance matrices as images to the document.

4

Table 4 Comparison between classifiers based upon classification accuracy

S. No.	Classifier	Accuracy (in %)
1.	KNN	89.58
2.	KNN on normalized data	96.72
3.	Bayes	95.83%.

Inferences:

1. KNN on normalized data has highest accuracy while KNN has lowest accuracy.