Final Lab Report CSCI 4030u

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Task One:

C4.5 Classification model

```
=== Classifier model (full training set) ===
J48 pruned tree
lym nodes dimin <= 1
| changes in node = no
\mid defect in node = no: normal (3.0/1.0)
| | defect_in_node = lacunar: malign_lymph (2.0)
| | defect in node = lac margin: normal (0.0)
| | defect_in_node = lac_central: normal (0.0)
| changes in node = lacunar
| | exclusion of no = no: metastases (10.0/1.0)
| | exclusion_of_no = yes
| | special forms = no: metastases (3.0/1.0)
| | | special forms = chalices
| | | | lym_nodes_enlar <= 2: malign_lymph (3.0)
| | | | lym_nodes_enlar > 2: metastases (2.0)
| | special_forms = vesicles: malign_lymph (19.0/1.0)
| changes in node = lac margin
| | block_of_affere = no
| | | lymphatics = normal: metastases (0.0)
| | | | lymphatics = arched
| | | | lymphatics = deformed: metastases (5.0)
| | | | lymphatics = displaced: malign_lymph (1.0)
| | extravasates = yes: malign_lymph (4.0)
| | block of affere = yes: metastases (56.0/3.0)
changes_in_node = lac_central
| | block_of_affere = no: malign_lymph (3.0)
| | | block_of_affere = yes: metastases (2.0)
| | no of nodes in > 1: malign lymph (20.0)
lym_nodes_dimin > 1
by_pass = no: metastases (2.0/1.0)
| by pass = yes: fibrosis (4.0)
```

```
Number of Leaves: 21
Size of the tree:
                     34
Time taken to build model: 0 seconds
Ripper
=== Classifier model (full training set) ===
JRIP rules:
========
(lymphatics = normal) => class=normal (2.0/0.0)
(lym nodes dimin \geq 2) and (by pass = yes) => class=fibrosis (4.0/0.0)
(no_of_nodes_in >= 3) and (special_forms = vesicles) => class=malign_lymph (41.0/5.0)
(block_of_affere = no) and (extravasates = yes) => class=malign_lymph (8.0/0.0)
(changes in node = lac central) => class=malign lymph (8.0/2.0)
=> class=metastases (85.0/11.0)
Number of Rules: 6
Time taken to build model: 0.02 seconds
=== Evaluation on training set ===
Time taken to test model on training data: 0 seconds
ID3
=== Classifier model (full training set) ===
ld3
changes_in_node = no
| lymphatics = normal: normal
| lymphatics = arched
| | early_uptake_in = no: metastases
| | early uptake in = yes: malign lymph
| lymphatics = deformed: fibrosis
| lymphatics = displaced: malign lymph
```

```
changes_in_node = lacunar
| special_forms = no
| | | changes_in_stru = no: null
| | | extravasates = no: metastases
| | | | extravasates = yes: malign_lymph
| | | changes in stru = diluted
| | | | block_of_affere = no: malign_lymph
| | | | block_of_affere = yes: metastases
| | | changes in stru = reticular: null
| | bl_of_lymph_c = yes: fibrosis
| special_forms = chalices
| | block of affere = no
| | | lymphatics = normal: null
| | | lymphatics = arched: malign lymph
| | | lymphatics = deformed: malign lymph
| | block_of_affere = yes: metastases
| special forms = vesicles
| | early_uptake_in = no
| | | lymphatics = normal: null
| | | lymphatics = deformed: fibrosis
| | | lymphatics = displaced: null
| | early_uptake_in = yes
| | | changes_in_stru = grainy: null
| | | changes_in_stru = drop_like: null
| | | changes_in_stru = diluted: malign_lymph
| | | changes in stru = faint: malign lymph
changes_in_node = lac_margin
| block_of_affere = no
| | extravasates = no
| | | lymphatics = normal: null
| | | lymphatics = arched
| | | | changes in stru = no: null
```

```
| | | | changes_in_stru = grainy: null
| | | | changes_in_stru = drop_like
| | | | | changes_in_lym = bean: null
| | | | | changes_in_lym = oval: metastases
| | | | | changes_in_lym = round: malign_lymph
| | | | changes in stru = diluted: malign lymph
| | | | changes_in_stru = stripped: null
| | | | changes in stru = faint
| | | | early_uptake_in = no: metastases
| | | | early_uptake_in = yes: malign_lymph
| | | lymphatics = displaced: malign_lymph
| | extravasates = yes: malign lymph
| block_of_affere = yes
| | changes_in_stru = no: null
| | changes in stru = grainy: metastases
| | changes_in_stru = drop_like: metastases
| | changes in stru = coarse: metastases
| | changes in stru = diluted
| | | no_of_nodes_in = '(-inf-3.5]': metastases
| \ | \ | no of nodes in = '(3.5-inf)': malign lymph
| | changes in stru = reticular: null
| | changes_in_stru = stripped: malign_lymph
| | changes in stru = faint
| \ | \ | no of nodes in = '(3.5-inf)': malign lymph
changes in node = lac central
| lym_nodes_enlar = '(-inf-2.5]'
| | changes_in_stru = no: null
| | changes in stru = grainy
| | | block of affere = yes: metastases
| | changes_in_stru = drop_like: null
| | changes in stru = coarse: null
| | changes_in_stru = diluted: metastases
| | changes in stru = reticular: null
| | changes_in_stru = stripped: null
| | changes in stru = faint: malign lymph
| lym nodes enlar = '(2.5-inf)': malign lymph
```

Time taken to build model: 0.01 seconds

Explanations:

The C4.5 algorithm is used as a decision tree classifier that can generate an output based on an inputted sample of data. For each node of the tree, this algorithm will create branches that will split the data into subsets of the data from the patterns in the dataset. This method uses entropy to generate the probability of each event happening. After creating the tree, pruning is used with the C4.5 algorithm to remove any redundant branches that do not help to generate the decisions. This is used to ensure that errors are not included when creating the tree.

The Ripper algorithm uses rule learning to generate the decisions. This is done in a three step process. The first part of the process creates conditions for each rule to properly classify data into subsets. The second stage is the pruning process. This occurs when the entropy for a given rule does not decrease as the rule becomes more specific. After the initial run through, these first two steps are repeated further until all the rules have been optimized which is the final process.

ID3 is another decision tree algorithm used to predict possible outcomes. This algorithm is the precursor to the C4.5 algorithm. This starts off by calculating the Information Gain (Entropy) and splits the dataset into certain subsets. For each subset, it will break off into certain results based upon the information in each. For example, it could produce a "yes" or "no" statement as well as break off into another subset and repeat the same process again. This whole process is repeated until all possible results can be generated.

Task Two:

C4.5

=== Summary ===

Correctly Classified Instances 420 97.2222 % Incorrectly Classified Instances 12 2.7778 % Kappa statistic 0.9444 Mean absolute error 0.0892 Root mean squared error 0.1831 Relative absolute error 17.8311 % Root relative squared error 36.5759 % Total Number of Instances 432

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	1.000	0.053	0.944	1.000	0.971	0.946	0.983	0.964	0
	0.947	0.000	1.000	0.947	0.973	0.946	0.983	0.981	1
Average	0.972	0.025	0.974	0.972	0.972	0.946	0.983	0.973	

а	b	← Classified as
204	0	A = 0
12	216	B = 1

ID3

=== Summary ===

Correctly Classified Instances 408 94.4444 % Incorrectly Classified Instances 16 3.7037 %

Kappa statistic 0.9245
Mean absolute error 0.0377
Root mean squared error 0.1943
Relative absolute error 7.6849 %
Root relative squared error 39.1873 %

UnClassified Instances 8 1.8519 %

Total Number of Instances 432

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.980	0.055	0.943	0.980	0.962	0.925	0.964	0.934	0
	0.945	0.020	0.981	0.945	0.963	0.925	0.946	0.941	1
Average	0.962	0.036	0.963	0.962	0.962	0.925	0.955	0.938	

а	b	← Classified as

200	4	A = 0
12	208	B = 1

Ripper

=== Summary ===

Correctly Classified Instances 390 90.2778 % Incorrectly Classified Instances 42 9.7222 %

Kappa statistic0.8053Mean absolute error0.1314Root mean squared error0.277Relative absolute error26.2643 %Root relative squared error55.3461 %

Total Number of Instances 432

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.912	0.105	0.866	0.912	0.899	0.806	0.938	0.879	0
	0.895	0.088	0.919	0.985	0.907	0.806	0.938	0.942	1
Average	0.903	0.096	0.903	0.903	0.903	0.806	0.938	0.912	

а	b	← Classified as
186	18	A = 0
24	204	B = 1

K-Nearest Neighbor

=== Summary ===

Correctly Classified Instances 378 87.5 % Incorrectly Classified Instances 54 12.5 %

Kappa statistic

Mean absolute error

Root mean squared error

Relative absolute error

Root relative squared error

Total Number of Instances

0.7512

0.191

0.3228

38.1693 %

64.5029 %

TP Rate	FP Rate	Precision	Recall	F-Measure	мсс	ROC Area	PRC Area	Class
0.941	0.184	0.821	0.941	0.877	0.758	0.933	0.902	0
0.816	0.059	0.939	0.816	0.873	0.758	0.933	0.935	1

а	b	← Classified as
192	12	A = 0
42	186	B = 1

Naive Bayesian Classification

=== Summary ===

Correctly Classified Instances 420 97.2222 % Incorrectly Classified Instances 12 2.7778 %

Kappa statistic

Mean absolute error

Root mean squared error

Relative absolute error

Root relative squared error

0.9444

0.1863

0.2323

37.2363 %

46.4131 %

Total Number of Instances 432

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	1.000	0.053	0.944	1.000	0.971	0.946	0.975	0.961	0
	0.947	0.000	1.000	0.947	0.973	0.946	0.975	0.985	1
Average	0.972	0.025	0.974	0.972	0.972	0.946	0.975	0.973	

а	b	← Classified as
204	0	A = 0
12	216	B = 1

Neural Networks

=== Summary ===

Correctly Classified Instances 404 93.5185 % Incorrectly Classified Instances 28 6.4815 %

Kappa statistic

Mean absolute error

Root mean squared error

Relative absolute error

Root relative squared error

Total Number of Instances

0.8709

0.068

13.5875 %

46.3993 %

432

	TP Rate	FP Rate	Precision	Recall	F-Measure	мсс	ROC Area	PRC Area	Class
	1.000	0.123	0.879	1.000	0.936	0.878	0.967	0.941	0
	0.877	0.000	1.000	0.877	0.935	0.878	0.967	0.981	1
Average	0.935	0.058	0.943	0.935	0.935	0.878	0.967	0.962	

а	b	← Classified as
204	0	A = 0
28	200	B = 1

Explanations:

From these six different testing algorithms, it is evident that the C4.5 algorithm has the highest averages across all the table values and the K-Nearest Neighbor algorithm has the lowest. This is due to the fact that the C4.5 algorithm has a more efficient algorithm for smaller datasets and produces a more thorough result by pruning where K-Nearest Neighbor does not.

The K-Nearest Neighbor is an algorithm that uses previously inputted data to produce an output for data that is unlabeled. To aid in the process of generating outputs, this algorithm assumes that similar results are close in proximity. This algorithm starts out with choosing a K value that has the lowest number of errors. To find the right K value, multiple runs of the program will determine the right number. Once a K value is chosen, the distance between both the test data set and the training data set is calculated. This is calculated by finding the euclidean distance between the two. From there the distance is added to a collection that holds the distance and the index of the data. Once all the data has been tested and the euclidean distances have been calculated, this information is sorted in order of distances in ascending order. From the collection, select the first K number of entries and the labels for them. Depending on what type of data set this algorithm is trying to solve, will determine what value is returned. If the data set is a regression problem that has to have a decimal point value, the mean is returned. If not then it is considered classified and will return the mode.

Naive Bayesian Classification uses probability to produce an outcome. Given a data set, this classification can be used with Bayes' theorem, $P(A|B) = \frac{P(B|A) * P(A)}{P(B)}$. In this equation, A is the information that is needed to be produced and B is the set of features that would affect the outcome. This will produce the probability for the expected output. There are three types of this classifier: multinomial, bernoulli and gaussian. Multinomial is used generally for classifying documents and uses frequency of words in the documents to classify them. For Bernoulli, the use is similar but uses a true or false base system to classify. Finally, the Gaussian is used to take a value that is assumed to be sampled from a gaussian distribution.

A neural network is a type of algorithm that works similar to that of the human brain. This algorithm will take in inputs, and then stores the information into nodes. These nodes are then weighed by the amount of important information they store compared to the other nodes, these are known as the hidden layer. Once all the nodes have been weighed, output nodes are then generated. The hidden layer and output is then recalibrated based based on the errors found in the outputs, and this will repeat until the proper conditions are met with the algorithm. After this, the final output is based upon the sum of all the hidden layers.

Task Three:

C4.5

=== Summary ===

Correctly Classified Instances 178 85.9903 % Incorrectly Classified Instances 29 14.0097 %

Kappa statistic

Mean absolute error

Root mean squared error

Root relative squared error

Total Number of Instances

0.7168

0.1958

0.3288

39.4502 %

65.6306 %

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
	0.766	0.064	0.916	0.776	0.840	0.725	0.901	0.857	+
	0.936	0.224	0.823	0.936	0.876	0.725	0.901	0.872	-
Average	0.860	0.149	0.867	0.860	0.859	0.725	0.901	0.865	

а	b	← Classified as
76	22	A = +

7	102	B = -

Naive Bayes Classifier

=== Summary ===

Correctly Classified Instances 156 75.3623 % Incorrectly Classified Instances 51 24.6377 %

Kappa statistic

Mean absolute error

Root mean squared error

Relative absolute error

Root relative squared error

Total Number of Instances

0.4968

0.2468

0.4633

49.7186 %

92.494 %

	TP Rate	FP Rate	Precision	Recall	F-Measure	МСС	ROC Area	PRC Area	Class
	0.561	0.073	0.873	0.561	0.683	0.529	0.880	0.869	+
	0.927	0.439	0.701	0.927	0.798	0.529	0.880	0.887	-
Average	0.754	0.266	0.783	0.754	0.744	0.529	0.880	0.879	

а	b	← Classified as
55	43	A = +
8	101	B = -

Neural Networks

=== Summary ===

Correctly Classified Instances 160 77.2947 % Incorrectly Classified Instances 47 22.7053 %

Kappa statistic 0.5401

Mean absolute error 0.2173

Root mean squared error 0.4352

Relative absolute error 43.7768 %

Root relative squared error 86.8833 %

Total Number of Instances 207

TP Rate FP Rate Precision	Recall F-Measure	MCC ROC Area	PRC Area Clas	s
---------------------------	------------------	--------------	---------------	---

	0.663	0.128	0.823	0.663	0.734	0.550	0.869	0.864	+
	0.872	0.337	0.742	0.872	0.802	0.550	0.869	0.840	-
Average	0.773	0.238	0.780	0.773	0.770	0.550	0.869	0.851	

а	b	← Classified as
65	33	A = +
14	95	B = -

Task Four:

C4.5

Time taken to build model: 0.02 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 283 84.2262 % Incorrectly Classified Instances 53 15.7738 %

Kappa statistic0.7824Mean absolute error0.0486Root mean squared error0.1851Relative absolute error26.5877 %Root relative squared error61.3413 %

Total Number of Instances 336

	TP Rate FP Rate Precision Recall F-Measure MCC ROC Area PRC Ar								
Class									
	0.951	0.036	0.951	0.951	0.951	0.915	0.962	0.915	ср
	0.844	0.066	0.793	0.844	0.818	0.762	0.907	0.784	im
	0.865	0.032	0.833	0.865	0.849	0.821	0.904	0.669	рр
	0.571	0.030	0.690	0.571	0.625	0.589	0.855	0.635	imU
	0.700	0.028	0.609	0.700	0.651	0.629	0.890	0.655	om
	0.600	0.006	0.600	0.600	0.600	0.594	0.993	0.604	omL
	0.000	0.000	?	0.000	?	?	0.490	0.006	imL
	0.000	0.000	?	0.000	?	?	0.479	0.006	imS
Avg.	0.842	0.040	?	0.842	?	?	0.920	0.787	

Ripper

Time taken to build model: 0.04 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 271 80.6548 % Incorrectly Classified Instances 65 19.3452 %

Kappa statistic0.7311Mean absolute error0.0608Root mean squared error0.2013Relative absolute error33.2586 %Root relative squared error66.7354 %

Total Number of Instances 336

=== Detailed Accuracy By Class ===

	TP Rat	e FPR	ate Preci	ision Re	call F-M	leasure	MCC	ROC Are	ea PRC Area
Class									
	0.951	0.088	0.889	0.951	0.919	0.857	0.943	0.882	ср
	0.766	0.054	0.808	0.766	0.787	0.726	0.928	0.821	im
	0.788	0.025	0.854	0.788	0.820	0.789	0.924	0.751	рр
	0.514	0.060	0.500	0.514	0.507	0.449	0.852	0.435	imU
	0.750	0.013	0.789	0.750	0.769	0.755	0.874	0.602	om
	0.400	0.015	0.286	0.400	0.333	0.326	0.767	0.165	omL
	0.000	0.000	?	0.000	?	?	0.708	0.086	imL
	0.000	0.000	?	0.000	?	?	0.380	0.006	imS
Avg.	0.807	0.061	?	0.807	?	?	0.916	0.764	

Naive Bayesian Classification

Time taken to build model: 0 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 287 85.4167 % Incorrectly Classified Instances 49 14.5833 %

Kappa statistic

Mean absolute error

Root mean squared error

Relative absolute error

Root relative squared error

Total Number of Instances

0.8002

0.0429

0.1639

23.461 %

54.3314 %

=== Detailed Accuracy By Class ===

	TP Rat	e FPR	ate Preci	sion Re	call F-M	easure I	MCC	ROC Are	a PRC Area
Class									
	0.958	0.041	0.945	0.958	0.951	0.915	0.986	0.973	ср
	0.727	0.031	0.875	0.727	0.794	0.745	0.966	0.904	im
	0.846	0.032	0.830	0.846	0.838	0.808	0.945	0.901	рр
	0.829	0.060	0.617	0.829	0.707	0.677	0.937	0.630	imU
	0.900	0.009	0.857	0.900	0.878	0.870	0.996	0.964	om
	0.600	0.000	1.000	0.600	0.750	0.772	0.996	0.883	omL
	0.000	0.006	0.000	0.000	0.000	-0.006	0.058	0.006	imL
	0.000	0.003	0.000	0.000	0.000	-0.004	0.148	0.005	imS
Avg.	0.854	0.036	0.861	0.854	0.854	0.819	0.960	0.897	

K-Nearest Neighbor

Time taken to build model: 0 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 270 80.3571 % Incorrectly Classified Instances 66 19.6429 % Kenna statistic 0.7305

Kappa statistic 0.7295

Mean absolute error 0.0535

Root mean squared error 0.2189

Relative absolute error 29.238 %

Root relative squared error 72.5574 %

Total Number of Instances 336

	TP Rat	e FP R	ate Preci	sion Re	call F-M	easure l	MCC	ROC Are	a PRC Area
Class									
	0.930	0.052	0.930	0.930	0.930	0.878	0.942	0.900	ср
	0.727	0.081	0.727	0.727	0.727	0.646	0.814	0.609	im
	0.846	0.046	0.772	0.846	0.807	0.771	0.903	0.695	рр
	0.486	0.056	0.500	0.486	0.493	0.435	0.713	0.304	imU
	0.750	0.006	0.882	0.750	0.811	0.803	0.896	0.680	om
	1.000	0.003	0.833	1.000	0.909	0.911	0.999	0.867	omL
	0.000	0.006	0.000	0.000	0.000	-0.006	0.695	0.010	imL
	0.000	0.000	?	0.000	?	?	0.698	0.010	imS
Avg.	0.804	0.054	?	0.804	?	?	0.878	0.715	

Neural Networks

Time taken to build model: 0.32 seconds

=== Stratified cross-validation ===

=== Summary ===

Correctly Classified Instances 289 86.0119 % Incorrectly Classified Instances 47 13.9881 %

Kappa statistic

Mean absolute error

Root mean squared error

Root relative squared error

Total Number of Instances

0.8066

0.0484

0.1704

26.479 %

56.4913 %

Total Number of Instances 336

=== Detailed Accuracy By Class ===

	TP Rat	e FPR	ate Preci	sion Re	call F-M	easure	MCC	ROC Are	a PRC Area
Class									
	0.965	0.036	0.952	0.965	0.958	0.927	0.980	0.962	ср
	0.831	0.062	0.800	0.831	0.815	0.759	0.951	0.870	im
	0.846	0.032	0.830	0.846	0.838	0.808	0.952	0.806	рр
	0.629	0.037	0.667	0.629	0.647	0.608	0.935	0.580	imU
	0.850	0.009	0.850	0.850	0.850	0.841	0.977	0.887	om
	0.800	0.003	0.800	0.800	0.800	0.797	0.997	0.786	omL
	0.000	0.000	?	0.000	?	?	0.187	0.005	imL
	0.000	0.000	?	0.000	?	?	0.340	0.007	imS
Avg.	0.860	0.039	?	0.860	?	?	0.956	0.859	

It appears that neural networks or multilayer perceptrons have the highest percentage of accuracy in the cross validation tests above (10-fold) with 86% of instances correctly classified. Naive Bayesian and C4.5 closely follow with 85.4% and 84.2% respectively. K-Nearest Neighbor performed and RIPPER performed approximately five percent worse than the others.

KNN depends greatly on distances between points. As you increase the number of dimensions, your distances are going to be less representative, this is called the curse of dimensionality. Taking this into account, KNN might perform slightly better if the number of features were to be reduced. RIPPER works well on datasets with imbalanced data meaning most of the data belongs to a single class (default class). Knowing this we could say based on

the test results that the ecoli dataset we are working with is probably more balanced than imbalanced.

The difference between misclassification rates is very small for multilayer perceptrons, Naive Bayesian and C4.5. So I wouldn't say there is a clear winner in these tests. KNN and RIPPER performed measurably worse and could be labeled losers in this case.

Task Five:

C4.5

Correctly Classified Instances 8480 94.2222 % Incorrectly Classified Instances 520 5.7778 %

9000

Kappa statistic 0
Total Cost 2600
Average Cost 0.2889
Mean absolute error 0.1094
Root mean squared error 0.2333
Relative absolute error 99.966 %
Root relative squared error 100 %

Total Number of Instances

	TP Rate	FP Rate	Precision	Recall	F-Mea sure	MCC	ROC Area	PRC Area	Class
	1.000	1.000	0.942	1.000	0.970	?	0.500	0.942	No
	0.000	0.000	?	0.000	?	?	0.500	0.058	Yes
Average	0.942	0.942	?	0.942	?	?	0.500	0.891	

а	b	← Classified as
8480	0	A = no
520	0	B = yes

Cost = 8480*(0) + 0*(50) + 520*(5) + 0*(0) = 2600

Naive Bayesian Classification

Correctly Classified Instances 8431 93.6778 % Incorrectly Classified Instances 569 6.3222 %

Kappa statistic 0.0237

Total Cost 5545

Average Cost 0.6161

Mean absolute error 0.111

Root mean squared error 0.2

Root mean squared error 0.2418
Relative absolute error 101.442 %
Root relative squared error 103.652 %

Total Number of Instances

	TP Rate	FP Rate	Precision	Recall	F-Mea sure	мсс	ROC Area	PRC Area	Class
	0.933	0.979	0.943	0.993	0.967	0.037	0.647	0.965	No
	0.021	0.007	0.155	0.021	0.037	0.037	0.647	0.098	Yes
Average	0.937	0.923	0.897	0.937	0.914	0.037	0.647	0.915	

а	b	← Classified as
8420	60	A = no
509	11	B = yes

Cost = 8420*(0) + 60*(50) + 509*(5) + 11*(0) = 5545

For pre-processing, the 'ANUMMER_10' and 'MAHN_HOECHST' attributes were removed, due to either having a full column of unknowns or because the column has the same values in another column. In WEKA, The filters of 'ReplaceMissingValues' and 'Discretize' were applied to the dataset to generate information.