

DADS Experiment No: 1

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Dept: FY M Tech DSA(2022-24)

Aim:

Implement two pre-processing techniques on standardized data from UCI repository or any other as recommended by the instructor.

What is WEKA?:

WEKA - an open-source software provides tools for data pre-processing, implementation of several Machine Learning algorithms, and visualization tools so that you can develop machine learning techniques and apply them to real-world data. It contains a Collection of visualization tools and algorithms for data analysis and predictive modelling coupled with graphical user interface.

Weka supports several standard data mining tasks, more specifically, data pre-processing, clustering, classification, regressing, visualization and feature selection.

Weka has the following advantages, such as: Free availability under the GNU General Public License. Portability, since it is fully implemented in the Java programming language and thus runs on almost any modern computing platform. A comprehensive collection of data pre-processing and modelling techniques.

WEKA Interface:

Weka GUI Chooser

The entry point into the Weka interface is the Weka GUI Chooser.

It is an interface that lets you choose and launch a specific Weka environment.

1 Screenshot of the Weka GUI Chooser



Weka Explorer

The Weka Explorer is designed to investigate your machine learning dataset.

It is useful when you are thinking about different data transforms and modelling algorithms that you could investigate with a controlled experiment later. It is excellent for getting ideas and playing what-if scenarios.

The interface is divided into 6 tabs, each with a specific function:

The pre-process tab is for loading your dataset and applying filters to transform the data into a form that better exposes the structure of the problem to the modelling processes. Also provides some summary statistics about loaded data.

Load a standard dataset in the data/ directory of your Weka installation, specifically ex. data/ 1year.arff. This is a binary classification problem that we will use on this tour.

No.	Label	Count	Weight
1	0	6756	6756
2	1	271	271

Dataset (website/sample data):

Website: <https://archive.ics.uci.edu/ml/datasets/Polish+companies+bankruptcy+data>

Data Set Information:

The dataset is about bankruptcy prediction of Polish companies. The data was collected from Emerging Markets Information Service (EMIS, [Web Link]), which is a database containing information on emerging markets around the world. The bankrupt companies were analysed in the period 2000-2012, while the still operating companies were evaluated from 2007 to 2013.

Basing on the collected data depends on the forecasting period:- 1stYear the data contains financial rates from 1st year of the forecasting period and corresponding class label that indicates bankruptcy status after 5 years. The data contains 7027 instances (financial statements), 271 represents bankrupted companies, 6756 firms that did not bankrupt in the forecasting period.

Sample data:

	Attr1	Attr2	Attr3	Attr4	Attr5	Attr6	Attr7	Attr8	Attr9	Attr10	...	Attr56	Attr57	Attr58
0	0.088238	0.55472	0.01134	1.0205	-66.5200	0.342040	0.109490	0.57752	1.0881	0.32036	...	0.080955	0.275430	0.91905
1	0.006202	0.48465	0.23298	1.5998	6.1825	0.000000	0.006202	1.06340	1.2757	0.51535	...	0.028591	0.012035	1.00470
2	0.130240	0.22142	0.57751	3.6082	120.0400	0.187640	0.162120	3.05900	1.1415	0.67731	...	0.123960	0.192290	0.87604
3	0.089951	0.88700	0.26927	1.5222	-55.9920	0.073957	0.089951	0.12740	1.2754	0.11300	...	0.418840	0.796020	0.59074
4	0.048179	0.55041	0.10765	1.2437	-22.9590	0.000000	0.059280	0.81682	1.5150	0.44959	...	0.240400	0.107160	0.77048

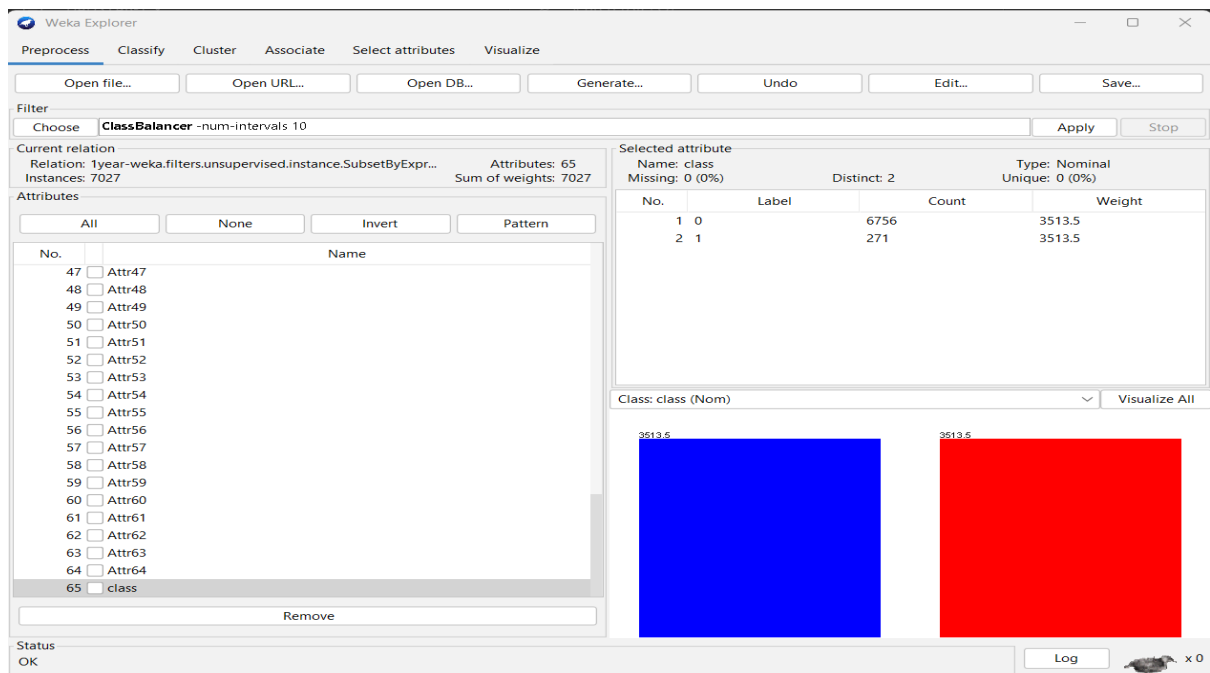
Steps involved in WEKA:

1. Class balancer:

A frequent question of Weka users is how to implement oversampling or under sampling, which are two common strategies for dealing with imbalanced classes in classification problems. This post provides some explanation.

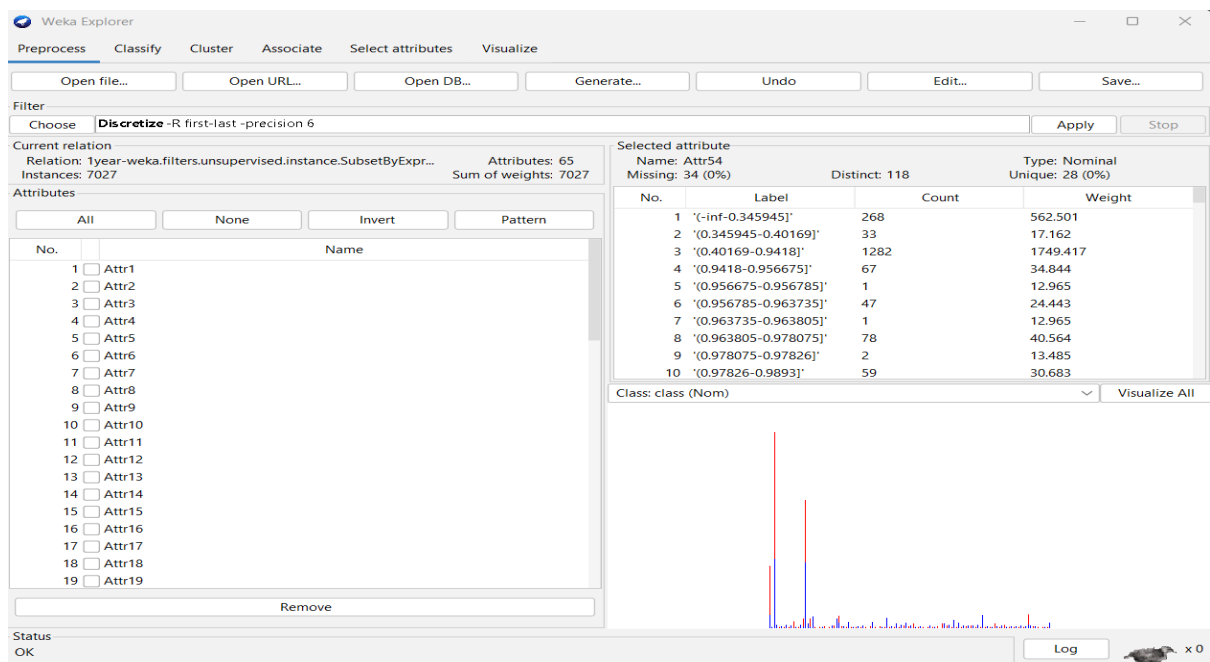
When a binary classification problem has a lot less data in one class than in the other one, e.g., when learning from results of a medical test for which the vast majority of instances have a negative outcome and only a few returns positive, some machine learning algorithms will simply learn to ignore the minority class and classify all cases into the majority class because this will trivially yield high classification accuracy. This kind of classification model is clearly not useful. Two common methods for combating this problem are under sampling of the majority class and oversampling of the minority class respectively.

In our case we use Oversampling for a given dataset of Bankruptcy.



2. Discretize

An instance filter that discretizes a range of numeric attributes in the dataset into nominal attributes. Discretization is by simple binning. Skips the class attribute if set.



Results obtained:

Random Forest gives 73% accuracy.

Weka Explorer

Preprocess **Classify** Cluster Associate Select attributes Visualize

Classifier: Choose **J48 -C 0.25 -M 2**

Test options:

- ☐ Use training set
- ☐ Supplied test set
- ☒ Cross-validation Folds: **10**
- ☐ Percentage split % **66**

More options...

(Nom) class: **Start** **Stop**

Result list (right-click for options):

- 11:30:44 - trees.RandomForest
- 11:30:56 - trees.J48

Classifier output:

Bagging with 100 iterations and base learner

weka.classifiers.trees.RandomTree -K 0 -M 1.0 -V 0.001 -S 1 -do-not-check-capabilities

Time taken to build model: 0.88 seconds

=== Stratified cross-validation ===

=== Summary ===

Metric	Value	Percentage
Correctly Classified Instances	5163.0954	73.4751 %
Incorrectly Classified Instances	1863.9046	26.5249 %
Kappa statistic	0.4695	
Mean absolute error	0.3109	
Root mean squared error	0.3861	
Relative absolute error	62.1707 %	
Root relative squared error	77.2259 %	
Total Number of Instances	7027	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
0	0.993	0.524	0.655	0.993	0.789	0.549	0.981	0.985	0
1	0.476	0.007	0.987	0.476	0.642	0.549	0.981	0.968	1
Weighted Avg.	0.735	0.265	0.821	0.735	0.716	0.549	0.981	0.977	

=== Confusion Matrix ===

a \ b	0	1	Classified as
0	3490.62	22.88	a = 0
1	1841.02	1672.48	b = 1

Status: OK

Log

Next, J48 model gives 73% accuracy which is more than random Forest model.

Weka Explorer

Preprocess **Classify** Cluster Associate Select attributes Visualize

Classifier: Choose **J48 -C 0.25 -M 2**

Test options:

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- ☒ Cross-validation Folds: **10**
- ☐ Percentage split % **66**

More options...

(Nom) class: **Start** **Stop**

Result list (right-click for options):

- 11:30:44 - trees.RandomForest
- 11:30:56 - trees.J48

Classifier output:

Number of Leaves : 2911

Size of the tree : 2966

Time taken to build model: 0.16 seconds

=== Stratified cross-validation ===

=== Summary ===

Metric	Value	Percentage
Correctly Classified Instances	5691.5838	80.9959 %
Incorrectly Classified Instances	1335.4162	19.0041 %
Kappa statistic	0.6199	
Mean absolute error	0.2438	
Root mean squared error	0.4163	
Relative absolute error	48.7639 %	
Root relative squared error	83.2662 %	
Total Number of Instances	7027	

=== Detailed Accuracy By Class ===

	TP Rate	FP Rate	Precision	Recall	F-Measure	MCC	ROC Area	PRC Area	Class
0	0.945	0.325	0.744	0.945	0.833	0.644	0.843	0.781	0
1	0.675	0.055	0.924	0.675	0.780	0.644	0.843	0.818	1
Weighted Avg.	0.810	0.190	0.834	0.810	0.806	0.644	0.843	0.799	

=== Confusion Matrix ===

a \ b	0	1	Classified as
0	3319	195	a = 0
1	1141	2373	b = 1

Status: OK

Log

Conclusion:

WEKA is a powerful tool for developing machine learning models. It provides implementation of several most widely used ML algorithms. Before these algorithms are applied to your dataset, it also allows you to pre-process the data. The types of algorithms that are supported are classified under Classify, Cluster, Associate, and Select attributes. The result at various stages of processing can be visualized with a beautiful and powerful visual representation. This makes it easier for a Data Scientist to quickly apply the various machine learning techniques on his dataset, compare the results and create the best model for the final use.

Reference:

- <https://machinelearningmastery.com/tour-weka-machine-learning-workbench/>
- https://www.tutorialspoint.com/weka/weka_quick_guide.htm
- <https://weka.sourceforge.io/doc.dev/weka/filters/unsupervised/attribute/Discretize.html>
- <https://www.analyticsvidhya.com/learning-paths-data-science-business-analytics-business-intelligence-big-data/weka-gui-learn-machine-learning/>