

Data Mining Lab 1:

Data sets: characteristics, formats, repositories Introduction to Rattle (R) and Weka (Java)

I. Data sets

I.1. Data sets characteristics and formats

The data to be processed can be

- **structured** (e.g. data matrix, tables from relational databases) ← used during this lab
- semi-structured (e.g. XML files, web logs)
- unstructured (e.g. text documents)

Structured data:

- set of instances (records)
- each instance contains values corresponding to several attributes (features)
- the attributes can be of different types:
 - qualitative or categorical (their values are symbolic objects, e.g. symbols or strings):
 - nominal (e.g. nationality, gender, religion, marital status etc)
 - logical/ binary (e.g. presence or absence of a specific characteristic)
 - ordinal (e.g. level of satisfaction (“low”, “medium”, “high”), qualitative grade (“insufficient”, “sufficient”, “good”, “very good”, “excellent”))
 - quantitative (their values are numbers taking values in discrete sets or continuous intervals)
 - integer (e.g. number of children, age in years, number of hits of a web page – usually values obtained by counting)
 - real (e.g. temperature, height, weight – usually values obtained by measuring)
- possible operations on the values of the attributes:
 - equality check, count of occurrences: nominal, logical and binary, ordinal, integer, real
 - comparison, ranking: ordinal, integer, real
 - comparison, ranking, addition, subtraction: integer, real (these are also called interval data)
 - comparison, ranking, addition, subtraction, multiplication, ratio: real (these are also called ratio data)

I.2. Data sets repositories

UCI Machine Learning repository (<http://archive.ics.uci.edu/ml/>)

- more than 400 datasets grouped by categories
- a dataset usually contains files with the description of data (“names”) and csv files containing the instances (“data”)

Kaggle Platform (<https://www.kaggle.com/>)

- more than 12000 datasets of different sizes (from several hundreds of KB to more than on GB), various formats (csv, json, sqlite, BigQuery), different domains

- solutions (kernels): Python notebooks, R scripts
- forum of discussions
- active competitions

KDD competitions (<http://www.kdd.org/kdd-cup>)

- data for annual Data Mining and Knowledge Discovery competitions organized by ACM

Exercise 1:

Select and analyze the characteristics (number of instances, number and types of attributes, tasks for which it has been used) of a data set from the UCI. Criteria to be used in the selection:

- Size:
 - Small number of attributes of instances (e.g. Iris dataset)
 - Small number of attributes and a large number of instances (e.g. DBWorld emails dataset)
 - Large number of attributes and a large number of instances
- Problem category:
 - Classification
 - Clustering
 - Regression

Exercise 2: Analysis of some Kaggle datasets:

- Titanic data set (<https://www.kaggle.com/c/titanic>). Identify: number of instances, number and types of attributes. Examples of processing steps applied to the dataset
 - <https://www.kaggle.com/hiteshp/head-start-for-data-scientist>
<https://www.kaggle.com/vin1234/best-titanic-survival-prediction-for-beginners>
- Google jobs (<https://www.kaggle.com/niyamatalmass/google-job-skills>). What kind of knowledge could we extract from the dataset
- Visit the current (and recent) competitions at Kaggle

II. Introducing Rattle = R Analytical Tool To Learn Easily = A Graphical User Interface for Data Mining using R (<https://rattle.togaware.com/>)

Rattle is a small platform, implemented in R (and based on existing R packages) used to learn how to apply basic data mining methods (without the need to search for the corresponding R package). It offers a simple to use GUI.

Rattle allows to:

- Load datasets in various formats (e.g. csv, dataset from R, arff (Attribute-Relation File Format - Weka standard) etc)
- Attribute and instance selection
- Data exploration (e.g. summarization, visualization)
- Data transformation
- Construct/ evaluate/ export models for classification, clustering, association analysis

Installation: `>install.packages("rattle")`

Usage:

```
>library(rattle)
>rattle()
```

Processing:

Data loading: **Data** ->fill Filename + upload -> Execute

Setting attributes: input/ target

Data exploration: **Explore**

Summary – statistical measures

Distributions – various visualization (BoxPlot, Histogram etc)

Correlation – analysis of the correlation between attributes (based on correlation coefficients: Pearson, Spearman, Kendall)

Principal Components – identification of maximal variance directions

Test: statistical tests

Transform: scaling(**Rescale**) / fill in missing values (**Impute**) / type conversions (**Recode**) / instance removal (**Cleanup**)

Model: classification and regression models based on: decision trees (**Tree**), ensemble of decision trees (**Forest**), support vector machines (**SVM**), logistic regression (**Linear**), neural networks (**Neural Net**)

Cluster: data clustering – partitional algorithms (**Kmeans**), agglomerative hierarchical algorithms (**Hierarchical**), biclustering (**BiCluster**)

Associate: extraction of association rules

Remarks:

- After a processing task is selected and configured press **Execute**
- Some tasks requires the installation of R packages (Rattle will do this)
- All processing steps can be exported in R scripts which can be used later in R (see **Log** tab)

Exercise 3:

1. Load the **Iris** dataset in Rattle and analyze the summary:
 - a. Number of instances
 - b. Number of attributes
 - c. For each attribute: possible values and frequency
2. Use **View/Edit** to visualize the data matrix and make changes
3. Use **Explore** to analyze:
 - a. The distribution of values of each attribute in each class
 - b. The correlation between the attributes; Which attributes are strongly correlated? How could be used this information?

Exercise 4: For the datasets from files **carr.arff** and **autoMPG.arff** analyze:

- a. The types of the attributes
- b. The distribution of data per classes
- c. The distribution of attributes' values per classes
- d. The correlation between the numerical / ordinal attributes (where it is possible)

III. Introduction to Weka <http://www.cs.waikato.ac.nz/ml/weka/>

III.1. What is Weka?

WEKA = Waikato Environment for Knowledge Analysis

Free open-source software workbench developed at the Waikato University incorporating several implementations of machine learning algorithms to be used for different data mining tasks:

- Data visualization
- Data pre-processing (75 implemented algorithms)
- Attribute selection (around 25 implemented algorithms)
- Classification (more than 100 algorithm currently implemented)
- Clustering (around 20 implemented algorithms)
- Association rules

It is implemented in Java and it runs on: Windows, Linux, Mac

III.2. How can it be installed?

Just download from <http://www.cs.waikato.ac.nz/ml/weka/downloading.html>

Last stable version: 3.8.* (currently 3.8.2)

III.3. Which are the main components / ways of using Weka?

- **Graphical User Interface:**
 - **Explorer** – used to apply specific data mining tasks to datasets
 - **Experimenter** – used to conduct comparative analysis of different methods on different datasets
 - **KnowledgeFlow** – graphical interface used to define flows of tasks
 - **Workbench** – it subsumes all the major Weka GUIs
- **Simple CLI** - command line interface
- **Java API**

III.3.1. Explorer:

- The data to be processed can be loaded:
 - From file (**Open File**) – common formats: **arff** (Weka format), **csv** (comma separated values)
 - From a web site (**Open URL**) or from a database (**Open DB**)
- Some examples of data could be generated randomly by using **Generate**
- The loaded dataset can be visualized as a data table and edited using **Edit**
- Categories of processing tasks:
 - Visualization
 - Data preprocessing
 - Attribute selection
 - Classification
 - Clustering
 - Association rules

III.3.2. Experimenter:

- Allows the statistical comparison of several methods associated to a given task (e.g. classification) by using several data sets
- Provides the result of paired t-test (with corrections for multiple comparisons)

III.3.3. Knowledge Flow:

- **Typical workflow:** “data source” ->”filter”->”classifier”->”evaluator”

- **Example**

- **Arff loader** - [data set](#) connection to
- **Cross validation foldMaker** – [training set](#), [test set](#) connection to
- **Naïve Bayes** – [batch classifier](#) connection to
- **Classifier Performance Evaluation** – [text](#) connection to
- **Text viewer**

Remark: the connections are selected by right button click on the node icon

- **Flow activation:**

- Right button click on Arff and [Start loading](#)

III.3.4. Command line interface:

- **Set the weka environment variable for java**

- setenv WEKAHOME c:\Program Files\Weka-3-6
- setenv CLASSPATH \$WEKAHOME/weka.jar:\$CLASSPATH

- **use a weka function**

- java weka.classifiers.j48.J48 -t \$WEKAHOME/data/iris.arff

III.3.5. Java API:

- Add weka.jar to class path
- Example of using the J48 classifier

```
import weka.core.Instances;
import weka.classifiers.trees.J48;
...
Instances data = ... // from somewhere
String[] options = new String[1];
options[0] = "-U"; // unpruned tree
J48 tree = new J48(); // new instance of tree
tree.setOptions(options); // set the options
tree.buildClassifier(data); // build classifier
```

III.3.6. ARFF data format – Attribute Relation File Format (this is the standard format of datasets to be processed in Weka)

- Header:

- Comments (%)
- Identifier of the data set: [@relation dataset_name](#)
- List of attributes (each attribute is characterized by a name and a type):
[@attribute attr_name type](#)

Remark: in the case of discrete attributes (which do not have many values) the type is the set of possible values; in the case of discrete numeric attributes the type may be specified by [integer](#); in the case of continuous numeric attributes the type is specified by [real](#)

- Data matrix:

- specified by [@data](#)

- each row contains a data instance; the values of the attributes are separated by comma

Example 1:

```
@relation car
@attribute buying {vhigh,high,med,low}
@attribute maint {vhigh,high,med,low}
@attribute doors {2,3,4,5more}
@attribute persons {2,4,more}
@attribute lug_boot {small,med,big}
@attribute safety {low,med,high}
@attribute class {unacc,acc,good,vgood}
@data
vhigh,vhigh,2,2,small,low,unacc
vhigh,vhigh,2,2,small,med,unacc
vhigh,vhigh,2,2,small,high,unacc
vhigh,vhigh,2,2,med,low,unacc
vhigh,vhigh,2,2,med,med,unacc
```

Example 2:

```
@relation 'autoPrice.names'
@attribute symboling real
@attribute normalized-losses real
@attribute wheel-base real
@attribute length real
@attribute width real
@attribute height real
@attribute curb-weight real
@attribute engine-size real
@attribute bore real
@attribute stroke real
@attribute compression-ratio real
@attribute horsepower real
@attribute peak-rpm real
@attribute city-mpg real
@attribute highway-mpg real
@attribute class real
@data
2,164,99.8,176.6,66.2,54.3,2337,109,3.19,3.4,10,102,5500,24,30,13950
2,164,99.4,176.6,66.4,54.3,2824,136,3.19,3.4,8,115,5500,18,22,17450
1,158,105.8,192.7,71.4,55.7,2844,136,3.19,3.4,8.5,110,5500,19,25,17710
1,158,105.8,192.7,71.4,55.9,3086,131,3.13,3.4,8.3,140,5500,17,20,23875
```

Exercise 5:

- Load the [Iris data set](#) in Weka (by clicking [Open file](#) button) and analyze its summary:
 - Number of instances
 - Number of attributes
 - For each attribute: possible values and their corresponding counts
- Use [Edit](#) to see the data matrix and to make changes on the data values.

- c. Use [Visualize All](#) to see the distribution of the values of each of the attributes in different classes (the class attribute is implicitly the last attribute). Identify a pair of attributes which are able to discriminate the classes (e.g. (sepal width, petal length)). How can be intuitively estimated the discriminative ability?
- d. Use the appropriate Weka filter to remove the attributes which are not considered relevant for the classification.

Exercise 6:

1. Follow the same steps for the [car data set](#) (car.arff)
Remark: The main difference between iris and car datasets is the fact that the first one contains numerical attributes while the second one contains nominal/ordinal attributes.
2. Load the [autoMpg.arff](#) file (it contains information on cars and on their consumption (mpg=miles per gallon)) and analyze its content using [Edit](#).
[Remark 1](#): Most of attributes (including the class) are numeric; if the class attribute is numeric then the data are appropriate for prediction tasks (estimate the values of “miles per gallon” depending on the characteristics of the care).
[Remark 2](#): The empty (gray) fields in the table correspond to missing values.
 - a) Use [Visualize All](#) to see which of the attributes is correlated with the mpg value
3. Load the [supermarket.arff](#) file (it contains data useful for market-basket analysis)

Exercise 7: compare the performance of several classification tools on two datasets (more details on the classification methods will be presented in Lecture 3 and Lab 3)

1. Select [Experimenter](#) from the starting Weka panel
2. Click [New](#) button to create a new experiment
3. Add datasets: [iris.arff](#) and [breast-cancer.arff](#)
4. Add algorithms: [zeroR](#), [oneR](#), [naiveBayes](#), [J48](#) (oneR and zeroR are from the “Rules” group, naiveBayes is from the “Bayes” group and J48 is from the “Tree” group)
5. [Run](#) the experiment
6. [Analyze](#) the results:
 - a. Click on [Experiment](#)
 - b. [Perform](#) the statistical test
 - c. Interpret the results of the statistical test (using zeroR as reference method)
 - i. v/* should be interpreted as: v= number of cases (datasets) on which the current method is better than the reference method; /= number of cases (datasets) on which the current method is not significantly different than the reference method; *= number of cases (datasets) on which the current method is worse than the reference method;