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## Lab 02 - Classification & Clustering

### Data Mining - Term I/2021-2022

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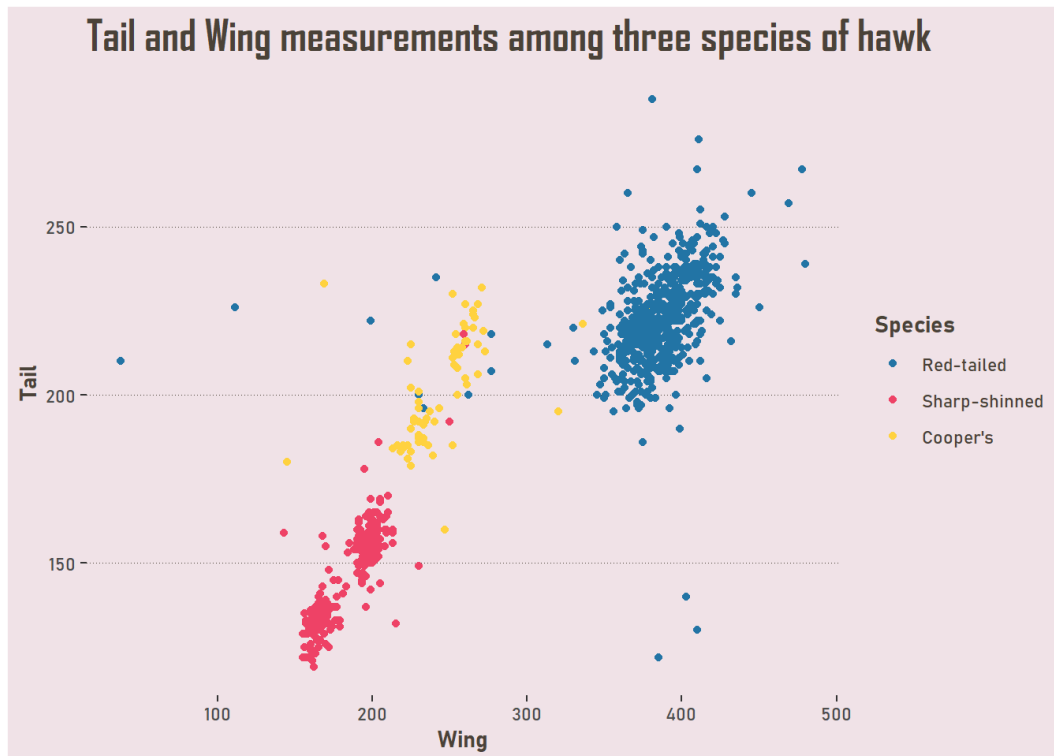
#### Objectives

- In this lab, students need to explore WEKA's data classification utilities by using two interfaces Explorer and Experimenter.
- Additionally, students get to improve programming skills by implementing basic clustering algorithms from scratch.

#### Rules

- Maximum 2 members/group.
- Duration: 3 weeks (more details on Moodle).
- Submission directory: **<StudentID>** or **<StudentID1>\_<StudentID2>** depending on the number of members in your group, containing the following files:
  - preprocess.py: preprocess script (if necessary).
  - Results.xls: results summary of experiments A-D.
  - RawResults.csv: output of experiment D.
  - Observations.pdf: group's observations and answers for given questions.
  - Lab02-Clustering.ipynb: students' work on Clustering part.
- Compress the submission directory into a **zip** file then submit it on Moodle.
- **Similar works will be marked 0 for the whole course.**
- Cite references properly.

# 1 CLASSIFICATION



Specification for the dataset **hawks**: [Hawks: Measurements on Three Hawk Species in Stat2Data: Datasets for Stat2](#). The target is to classify each data sample into one of the three species, namely red-tailed, sharp-shinned, and Cooper's.

## 1.1 Using Weka Explorer to classify data

For each of the experiments A-C below, use WEKA Explorer to perform classification by following methods (with default parameters): 1) NaiveBayesSimple; 2) Id3; and 3) J48. For each of the methods applied on the dataset, perform the following evaluation methods (see “Test options” in “Classify” window of Weka Explorer): a) “Use training set”; b) “Cross-validation” with 10 folds; and c) “Percentage split” with proportion of 66%. Write down the results of each run into an Excel file “Result.xls”, which includes the follow information:

- (a) Experiment type (A-C);
- (b) Name of the input dataset;
- (c) Classification method;
- (d) Evaluation strategy;
- (e) The proportion of correctly classified samples (the total and for each specific class accuracy).

Experiments:

- (A) Classify data points using methods above.
- (B) Discretize all non-class attributes into 10 buckets with **equal widths**: use function "Filter" in window "Preprocess" of the Explorer interface, select 'filters' → 'unsupervised' → 'attribute' → 'Discretize'. Use default parameters for 'Discretize' filter. After ensuring all non-class attributes are discrete, perform classification on the newly processed data set with classification algorithms and evaluation strategies described earlier.
- (C) Discretize all non-class attributes into 5 buckets with **equal depths** by using 'Discretize' filter and appropriate parameters. After ensuring all non-class attributes are discrete, perform classification on the newly processed data set with classification algorithms and evaluation strategies described earlier.

## 1.2 Data classification using Weka Experimenter

- (D) With this experiment, you must use WEKA Experimenter interface. Perform data classification with NaiveBayesSimple and J48 (default parameters). For each method, run 10-fold cross validation 10 times, and write down the results into a file called "RawResult.csv". From these results, compute the average accuracy of each method and add into the "Result.xls" file mentioned earlier: a) Experiment D; b) name of the data set, c) classification method; d) evaluation strategy; and e) average proportion of correctly classified samples after  $10 \times 10$  runs.

**Note:** the given data set contains missing values, so you must handle it before classifying (you can use any reasonable method). If you use python to preprocess it, submit an extra **preprocess.py** file. Otherwise if you use Weka to preprocess, please report the steps and methods used.

## 1.3 Assess the experiment results

After successfully performing the experiments, you should spend time assessing the results.

Specifically, you should at least be able to answer below questions:

- Which classification method typically has the best result?
- Which method does not work well and why?
- Why should we use the discretized version of the data set instead of the original one?

- Do the discretization process and method affect the classification results? If yes then how?
- Which evaluation strategy tends to overestimate the accuracy and why?
- Which evaluation strategy tends to underestimate the accuracy and why?

Answer those questions and make new observations into a “Observations.pdf” file.

## 2 CLUSTERING

In this section you must learn to use the **Jupyter notebook**, then do as instructed in the file **Lab02-Clustering.ipynb**.

If you have any questions related to the exercise, feel free to email me at *ktoan271199@gmail.com*.

Best regards.