# ECS607/766 - Data Mining - 2015/16

Lab 1: Data Exploration & Visualization 01/10/2015

#### Introduction

The outcome from the lab is to be familiar with Weka and basic exploratory data analysis. This session is a warm-up that will not count toward your final grade.

# 1. Getting Started

- 1. Download the dataset for this week (qmplus: LondonCars.csv)
  (This is data contains all the cars for sale on Autotrader.com. It was collected in Summer 2014 by an MSc student who worked on a data mining project)
- 2. Invoke Open Office or Excel Spreadsheet
- 3. Invoke Weka.
  - -Windows: Computer → teaching → weka → weka.jar
  - -Linux: search  $\rightarrow$  Weka

### 2. Explore the Dataset - Open Office

- 1. Load the dataset LondonCars.csv in a Spreadsheet (Open Office or Excel)
- 2. Basic exploratory questions:
  - a. How many instances does the dataset have?
  - b. How many attributes? What are they?
  - c. What are the possible values for Body Style & External Color?
  - d. What is the minimum, maximum, average and median price?
  - e. Why might the median price be different than the average price?
  - f. What is the most common year of car?
  - g. What is the ratio of 2-door to 4-door cars? Hint: Try countif function.
  - h. What is the average price of a Honda car versus a Mercedes-Benz car? Hint: Try averageif function.

# 3. Explore the Dataset - Weka

In this part we will explore the same data with Weka data mining software. Some things will be easier to do here than with a spreadsheet program. This type of exploratory analysis is typical when you are faced with a new dataset that you need to understand before trying to solve a specific problem.

- 1. Start Weka Explorer
- 2. Open the dataset: Open  $\rightarrow$  Select csv  $\rightarrow$  LondonCars.csv. You can click edit to

see the raw data in a spreadsheet style.

- a. Again, find out how many instances, and attributes the dataset has?
- b. By clicking on an attribute in the "Attributes" panel, you can see its type, possible values (discrete), or statistics (continuous) in the "Selected Attribute" panel.
- c. Which attributes are continuous (numeric) or discrete (nominal)?
- d. What are the possible body styles? Which is the most and least common body style?
- e. What are the most and least popular external body colors?
- f. What is the minimum, maximum and average values for mileage and price?
- g. Looking at the histogram for those attributes, do they look Gaussian (bell curve) distributed? Are there outliers?
- 3. If you use the Class pulldown in the Selected attribute panel. You can color the histogram of one attribute according to another attribute.
- 4. Select Body style attribute and Body style class. Note the color corresponding to each body type.
  - a. Select Make attribute, keeping body style coloring.
  - b. Which car company only makes SUV style cars?
  - c. Which car company makes the most Coupe style cars?
- 5. The #doors attribute has been interpreted as numeric. How many unique doors are there? To use it as a class for coloring, we should convert it to nominal.
  - a. Click Filter → Unsupervised → NumericToNominal. Click the box next to the filter and **choose attribute index** = 11 (index of doors) and press apply. Be careful not to apply to all attributes, or they will all become nominal.
  - b. Click the doors attribute and observe the coloring of each door configuration.
- 6. Now select body style attribute and color it by doors class.
  - a. Which car types always come with 2 doors? Which always come with 4 doors?
  - b. Which car types come in a mix of door configurations?
  - c. For each class attribute, you can click Visualize All to see each attribute histogram in those colors. See what other patterns you can find.

# 4. Finding Correlations

- 1. Switch to the visualization view in Weka Explorer. This can plot any pair of attributes against each other. Click on any panel to bring up a plot. On these plots every instance (record or row) in the database is shown as a point. If you click on any point it will bring up a window showing you the details of that instance (car).
- 2. Select X: Mileage, Y: Price. What do you observe about the relation between the two?
  - a. Is there any correlation? Is it linear?
  - b. You can simultaneously color the plot by any other variable.
  - c. What do you observe when coloring by price?

- d. When coloring by Engine size, what is the impact of 8, 6 or 4 cylinder engines?
- 3. Plot X: External Color, against Y: Price
  - a. Which color looks like they get the highest valuation overall?
  - b. Which color gets the lowest valuation overall? Use the Jitter slider to slightly spread out all the points that are on top of each other.
- 4. Looking at the least valuable color:
  - a. How many cars with that color are shown?
  - b. Click on a car (data point) of that category, what make is it?
  - c. Is it safe to conclude that cars with that color are generally very likely to be cheap?

# 5. Making Predictions

Lets try to build a car-price predictor. This would be useful for a used car business to know how much to offer to pay for a used car, and how much to sell it for.

- 1. Switch back to the Preprocess view. Remove all attributes besides the numeric ones: Year, Mileage, Price. (Click the nominal attributes and then press Remove). In later exercises we will get back to predictions using nominal inputs.
- 2. Switch to classify view.
  - a. Under classifier. Choose Classifiers  $\rightarrow$  Functions  $\rightarrow$  Linear Regression.
  - b. Make sure Price is selected as the target variable to predict in the drop-down box, and press start.
- 3. Observe the results in the output panel
  - a. The regression model discovers the values of A, B, and C in an equation of the form Price = A\*Year+B\*Mileage+C. It can then use this equation to predict the price of a new car.
  - b. What is the value of every Year of car age?
  - c. How much value does every mile of driving loose?
  - d. The mean absolute error is the average difference between the predicted price of each car and the true price. What is it in this case?
  - e. Do you think this an acceptable level of prediction accuracy for a used car business? What could we do to improve the accuracy?