

Workshop
RapidMiner & Python:
Data Cleaning
Supervised Learning

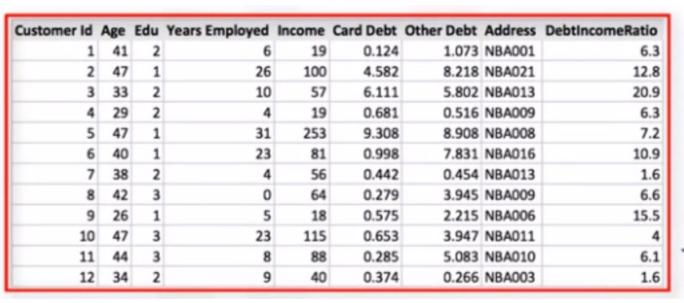
Elizabeth Savochkina | 6th June



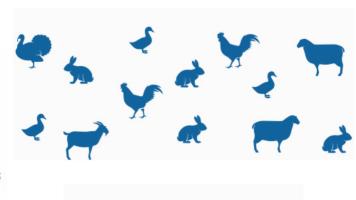
We have historical data from patients and some

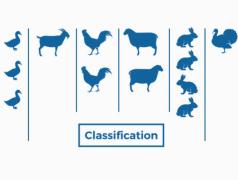
| J | Class | Mit | NormNucl | BlandChrom | BareNuc | SingEpiSize | MargAdh | UnifShape | UnifSize | Clump | ID |
|----|-----------|-----|----------|------------|---------|-------------|---------|-----------|----------|-------|---------|
| 1 | benign | 1 | 1 | 3 | 1 | 2 | 1 | 1 | 1 | 5 | .000025 |
| l | benign | 1 | 2 | 3 | 10 | 7 | 5 | 4 | 4 | 5 | .002945 |
| l | malignant | 1 | 1 | 3 | 2 | 2 | 1 | 1 | 1 | 3 | .015425 |
| Ш | benign | 1 | 7 | 3 | 4 | 3 | 1 | 8 | 8 | 6 | 016277 |
| l | benign | 1 | 1 | 3 | 1 | 2 | 3 | 1 | 1 | 4 | 017023 |
| 11 | malignant | 1 | 7 | | 10 | 7 | 8 | 10 | 10 | 8 | 017122 |
| Ш | benign | 1 | 1 | 3 | 10 | 2 | 1 | 1 | 1 | 1 | 018099 |
| Ш | benign | 1 | 1 | 3 | 1 | 2 | Н | 2 | 1 | 2 | 018561 |
| I | benign | 5 | 1 | 1 | 1 | 2 | 1 | 1 | 1 | 2 | .033078 |
| П | benign | 1 | 1 | 2 | 1 | 2 | 1 | 1 | 2 | 4 | .033078 |

We know little about the data and the outcomes that are expected.

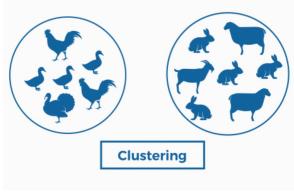








Clustering analysis: Clustering/grouping datapoints that somewhat similar in



Introduction

- Understanding AI methods requires both a theoretical and a practical approach.
- Many tools available to implement AI methods
- Two options in this course: RapidMiner and Python.
- Let's take a quick dive into both, and then you choose!

Overview of the Tools



? python™

Easy to use

Intuitive

Drag-and-drop building blocks

Has its limitations

Powerful and flexible

Active community

Huge number of resources

Free and open source

Involves coding

Recap: Data Cleaning

• Replace missing values with 'worst case' value e.g '0'

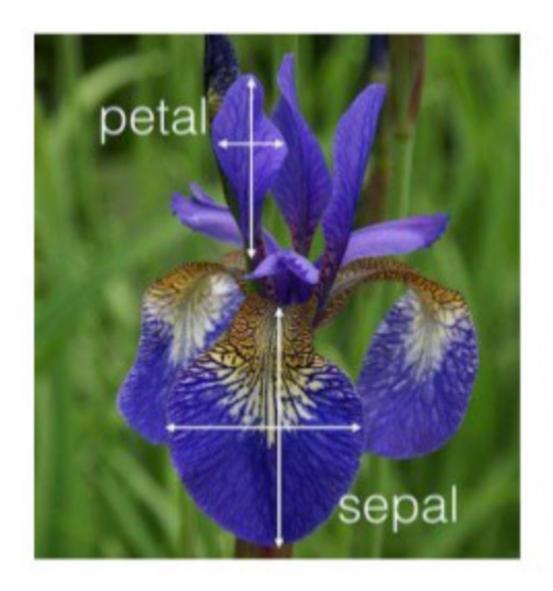
Replace missing value with average of other values in the same column

Example 1: Data Cleansing

Let's go through this morning's example on data cleaning again, first with Python then with RapidMiner

Example 2: Supervised Learning

Iris flower classification



Training / test data

Features

Labels

| Sepal length | Sepal width | Petal length | Petal width | Species |
|-----------------|----------------|-----------------|----------------|-----------------|
| 5.1 | 3.5 | 1.4 | 0.2 | Iris setosa |
| 4.9 | 3.0 | 1.4 | 0.2 | Iris setosa |
| 7.0 | 3.2 | 4.7 | 1.4 | Iris versicolor |
| 6.4 | 3.2 | 4.5 | 1.5 | Iris versicolor |
| 6.3 | 3.3 | 6.0 | 2.5 | Iris virginica |
| 5.8 | 3.3 | 6.0 | 2.5 | Iris virginica |