# CO3093 Big Data and Predictive Analytics

Credits: 20 Convenor: Dr. E. Tadjouddine Semester: 2<sup>nd</sup>

Prerequisites: Essential: CO1005, CO1008 Desirable: CO2015, CO3091

Assessment: Coursework: 60% Two hours exam in May/June: 40%

Lectures:20hoursClass Tests:2hoursSurgeries:10hoursPrivate Study:98hours

# Subject Knowledge

**Aims** This module is aimed at teaching tools and methods to interrogate data in order to carry out predictive analytics and to evaluate the resulting models.

**Learning Outcomes** Students should be able to: analyse possibly large amount of data; use a Map-Reduce approach in processing data; develop and back-test a predictive model; compare and contrast different types of predictive models; and evaluate a given model.

**Methods** Class sessions together with course notes, recommended textbook, Labs, printed solutions, and some additional hand-outs and web support.

#### **Assessment**

Summative: Coursework assessments including class tests and traditional written problem-based examination.

### **Skills**

**Aims** To teach students scientific writing and problem solving skills.

**Learning Outcomes** Students will be able to: appreciate niche applications wherein predictive analytics can be useful; write a technical report based on design and experiments and solve abstract and concrete problems involving possible vast amount of data (structured or unstructured).

**Methods** Lab sessions together with worksheets.

**Assessment** Summative: Coursework assessments including class tests and traditional written problem-based examination.

**Explanation of Prerequisites** In addition to the pre-recquisite modules above, the basics of calculus will be helpful. The ability to program in Python is desirable but this will be part of the Lab sessions.

## **Course Description**

As we increasingly rely upon the online environment for our daily routines, we leave behind a vast amount of information about us. Commercial and public organisations can use this information to predict our behavior. This module aims to study methods and tools enabling us to identify variables of interest and their relationships from an existing data set in order to develop a statistical model that can predict values of variables of interest. This kind of analysis should give us an insight into individual preferences, and most importantly, what someone is likely to do in a given scenario. Some of the applications include credit bank approval, marketing, stock price predictions, demand forecasting or polical campaining.

In this course, we will also study the importance of good quality data and will rely upon open libraries such as scikit-learn (http://scikit-learn.org/stable/) to implement basic models with much less programming effort. We will also learn how to compare and contrast different models for the same data and objective. As a predictive analysis does not necessarily demand a huge amount of data, we will also discuss the utility or misfortune of the so-called big data and how to process such a large amount of data efficiently by using a distributed approach as in the Apache Spark.

# **Detailed Syllabus**

- Python: the basics and relevant libraries such as numpy, scipy, and pandas and its visualisation tools will be reviewed typically during the first three Lab sessions.
- Big data: its philosophy (seek, store, analyse, and act); structured and unstructured data; data sources, volume, value, and timeliness; and data cleaning and manipulation.
- Basic probability and statistics: random variables, mean, variance, standard deviation; relationship between variables (correlation and regression); probability distributions including Poisson and Gaussian; sampling; hypothesis testing, confidence intervals and significance tests.
- Using predictive models: setting up clear objectives, identifying predictor data and outcome data, and the decision making process. We will use the example of a credit scoring model.
- Types of predictive models: regression and classification models including linear models, logistic regression, random forests, clustering, and finding similar items. These algorithms will be presented in the form of black-box and be used from within a software package, e.g., http://scikit-learn.org/stable/
- Building up a predictive model: data collection, sampling, and the iterative process (understanding the data, modeling, and performance evaluation). We will run through an example e.g., stock price movements or marketing.
- Processing big data: The Apache Spark tool. We will introduce the basics of how to efficiently process large data sets.

# **Reading List**

- [B] David M. Levine and David F. Stephan, Even You Can Learn Statistics and Analytics: An Easy to Understand Guide to Statistics and Analytics (Third Edition);
  ISBN: 978-0133382662, Pearson FT Press.
- [B] Steven Finlay, *Predictive Analytics, Data Mining and Big Data (Business in the Digital Economy); ISBN:* 9781137379276, Palgrave Macmillan.
- [B] Philipp K. Janert, *Data Analysis With Open Source Tools; ISBN:* 9787564126742, O'Reilly Media.
- [B] Ashish Kumar, Learning Predictive Analytics with Python; ISBN 978-1783983261, PACKT Publishing.
- [B] P.N. Tan and M. Steinbach and V. Kumar, *Introduction to Data Mining; ISBN:* 9787111316701, Pearson.
- [B] Anand Rajaraman and Jeffrey D. Ullman, *Mining of Massive Datasets; ISBN:* 9781207015357, Cambridge University Press.
- [B] Viktor Mayer-Schonberger and Kenneth Cukier, *Big data: A revolution that will transform how we live, work and think; ISBN:* 9781848547926, John Murray.
- [B] Frank J. Ohlhorst, *Turning Big Data into Big Money; ISBN:* 9781118147597, Wiley.
- [B] John M. Chambers, Software for Data Analysis: Programming with R (Statistics and Computing); ISBN: 9781441926128, Springer.
- [B] Kevin Sheppard, Introduction to Python for Econometrics, Statistics and Data Analysis (free eBook, August 2014);
  ISBN: N/A, KevinSheppard.com.
- [B] Charles Severance, *Python for Informatics: Exploring Information (free eBook, May 2014); ISBN: N/A*, pythonlearn.com.

**Resources** Course notes, web page, study guide, handouts, lecture, mock tests or examinations, rooms with two OHPs.

**Module Evaluation** Course questionnaires, course review.