

Department of Computer Science

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COM3004 Data Driven Computing

Summary	<p>This module is intended to serve as an introduction to machine learning and pattern processing, but with a clear emphasis on applications. The module is themed around the notion of data as a resource; how it is acquired, prepared for analysis and finally how we can learn from it. The module will employ a practical Python-based approach to try and help students develop an intuitive grasp of the sophisticated mathematical ideas that underpin this challenging but fascinating subject.</p>
Session	Autumn 2017/18
Credits	20
Assessment	Assignment (50%) and examination (50%).
Lecturer(s)	<p>Dr Jon Barker</p> <ul style="list-style-type: none">• MOLE (https://vle.shef.ac.uk) <p>()</p>
Resources	<ul style="list-style-type: none">• Unconfirmed practical marks when available (https://www.dcs.shef.ac.uk/intranet/teaching/campus/practicalmarks/com3004.pdf)• Exam Papers, past 2 years (where applicable) (https://www.dcs.shef.ac.uk/intranet/teaching/campus/pastpapers.html)
Aims	<p>This unit aims to:</p> <ul style="list-style-type: none">• provide an accessible introduction to key concepts in machine learning and pattern processing,• demonstrate the application of machine learning in a number of recent research areas,• develop an appreciation of the difficulties involved when trying to extract meaning from naturally occurring data with particular reference to data preprocessing, feature extraction, classifier design and efficient learning,• To prepare students for specialised data-driven subjects at level 3/4 such as natural language processing, speech processing and computational biology.
Objectives	<p>By the end of the unit, a student will be able to</p> <ul style="list-style-type: none">• demonstrate how to extract features from data for use by machine learning (ML) techniques,• demonstrate the ability to analyze and model data using ML techniques,• demonstrate the ability to apply ML in various areas of Computer Science, e.g. in natural language processing, audio/speech processing, biological applications and vision processing,• demonstrate the ability to use Python for scientific computing. <p>Introduction</p> <ul style="list-style-type: none">• overview: classification and feature handling• Python programming <p>Multivariate data</p> <ul style="list-style-type: none">• review: linear algebra/probability• normal distribution <p>Classification</p> <ul style="list-style-type: none">• Bayes decision theory• risk and ROC (receiver operating characteristic)• parameter estimation - maximum likelihood estimation• curse of dimensionality and naive Bayes classifier

Content	<p>Linear classifiers</p> <ul style="list-style-type: none"> • perceptron • XOR problem <p>Instance based approaches</p> <ul style="list-style-type: none"> • nearest neighbour and k-nearest neighbour • template matching and edit distance <p>Feature selection</p> <ul style="list-style-type: none"> • discriminability • feature selection algorithms <p>Feature generation</p> <ul style="list-style-type: none"> • dimensionality reduction • principle components analysis <p>Unsupervised learning and approaches to clustering.</p> <p>Density estimation and mixture modelling.</p> <p>Case study: Analysis of how techniques have been applied in a real system.</p>
Teaching Method	Lectures, problem classes and laboratory classes.
Feedback	Immediately from problem classes. After each assignment stage through debriefing lecture and individual marking.
Recommended Reading	<ul style="list-style-type: none"> • Python Programming, https://en.wikibooks.org/wiki/Python_Programming • Stephen Marsland, "Machine Learning: An Algorithmic Perspective", CRC Press, 2009