# **H9DMML2: Data Mining and Machine Learning II**

Module Code:		DMML2					
Long Title		Data Mining and Machine Learning II APPROVED					
Title		Data Mining and Machine Learning II					
Module Level:		LEVEL 9					
EQF Level:							
EHEA Level:		Second Cycle					
Credits:		10					
Module Coordinator:		IICHAEL BRADFORD					
Module Author:		lenette Carson					
Departments:		School of Computing					
Specifications of the qualifications and experience required of staff		D/MSc degree in a computing or cognate discipline. May have industry ea	xperience also.				
Learning Ou	tcomes						
On successful completion of this module the learner will be able to:							
#	Learning Outcome	Description					
LO1	Critically analyse admining problems	anced data mining and knowledge discovery methodologies in order to assess best practice guidance when applied to complex data					
LO2	Investigate and evalued domains.	aluate key concepts and advanced data mining techniques and assess when to apply such techniques on complex datasets and problem					
LO3	Contextualise, resea datasets with a varie	arch and utilise current data mining approaches, applications and technologies in order to provide strategies to address processing of ety of characteristics					
LO4	Critically review and	apply appropriate data mining research and assess research methods					
Dependenci	es						
Module Rec	ommendations						
No recommendations listed							
Co-requisite Modules							
No Co-requisite modules listed							
Entry requirements		A level 8 degree or its equivalent in any discipline	A level 8 degree or its equivalent in any discipline				

# **H9DMML2: Data Mining and Machine Learning II**

### **Module Content & Assessment**

### Indicative Content

**General Strategies Revisited** 

Increasing data complexity and size with fundamental methods. . Considerations of Complexity on Computing Requirements

#### **General Strategies Revisited**

Dimensionality Reduction (PCA, MCA, etc.). Feature Engineering. Measuring Predictor Importance

#### **General Strategies Revisited**

Understanding, Detecting and Handling (massive) class imbalance.. Understanding Factors that can Affect Model Performance; e.g. Type III errors, selection bias, measurement errors, improper variable encoding. Ethically assessing biases...

### **Advanced Regression Models**

Regression revision, and penalised models

# **Advanced Regression Models**

Generalised Linear Modelling

#### **Advanced Regression Models**

Automated Linear Modelling via Bagging and Boosting

#### **Ensembles**

Ensembles:. Random Forest. Voting. Stacking.

Bagging and Boosting Methods (e.g. XGBoost, AdaBoost, CART aggregation etc.)

#### Black Box Methods

Support Vector Machines and Support Vector Regression

#### Black Box Methods

Neural Networks:. Classic Topologies and Activation Functions. Back Propagation. Gradient Descent and Stochastic Gradient Descent. Hyperparameter Optimisation techniques

### **Black Box Methods**

Algorithmic Accountability, Ethical issues with black-box methods

## Deep Regression Models

A brief introduction to deep learning applied to regression problems (e.g. GLMNet). Special emphasis to be played on when these methods are(n't) appropriate (e.g. data

Assessment Breakdown	%		
Coursework	50.00%		
End of Module Assessment	50.00%		

#### Assessments

# **Full Time**

Coursework

Assessment Type: Formative Assessment % of total:

Non-Marked

**Assessment Date:** 

n/a

Outcome addressed:

1,2,3,4

Non-Marked:

**Assessment Description:** 

Yes

Formative assessment will be provided on the in-class individual or group activities. Feedback will be provided in written or oral format, or on-line through Moodle. In addition, in class discussions will be undertaken as part of the practical approach to learning.

Assessment Type:

% of total:

50

Assessment Date:

Project n/a

Outcome addressed:

3.4

Non-Marked:

No

No

**Assessment Description:** Propose and execute a research project using data mining techniques as a team of 3-4 participants.

# **End of Module Assessment**

Assessment Type: **Assessment Date:** 

Terminal Exam End-of-Semester % of total:

Outcome addressed:

50 1.2

Non-Marked:

# **Assessment Description:**

The examination will be a minimum of three hours in duration and may include a mix of: short answer questions, vignettes, essay based questions and case study based questions requiring the application of core module competencies. Marks will be awarded based on clarity, appropriate structure, relevant examples, depth of topic knowledge, and evidence of outside core text reading.

No Workplace Assessment

# Reassessment Requirement

Repeat examination

Reassessment of this module will consist of a repeat examination. It is possible that there will also be a requirement to be reassessed in a coursework element.

## Reassessment Description

The repeat strategy for this module is by repeat assessment/project that covers all learning outcomes.

# **H9DMML2: Data Mining and Machine Learning II**

Module Workload									
Module Target Workload Hours 0 Hours									
Workload: Full Time									
Workload Type	Workload Description	Hours	Frequency	Average Weekly Learner Workload					
Lecture	Classroom & Demonstrations (hours)	24	Every Week	24.00					
Tutorial	Other hours (Practical/Tutorial)	24	Every Week	24.00					
Independent Learning	Independent learning (hours)	202	Every Week	202.00					
Total Weekly Contact Hours									

# Module Resources

### Recommended Book Resources

Hastie, T., Tibshirani, R. & Friedman, J.. (2016), The Elements of Statistical Learning: Data Mining, Inference, and Prediction (2nd ed), Springer Series in Statistics

James, G., Witten, D., Hastie, T. & Tibshirani, R.. (2017), An Introduction to Statistical Learning: with Applications in R, Springer Texts in Statistics.

Kuhn, M. & Johnson, K.. (2013), Applied Predictive Modeling, Springer.

Shalev-Shwartz, S. & Ben-David, S.. (2014), Understanding Machine Learning: From Theory to Algorithms, Cambridge University Press.

### Supplementary Book Resources

Downey, B.. (2014), Think Stats: Exploratory Data Analysis, (2nd ed).

Goodfellow, I., Bengio, Y., & Courville, A.. (2016), Deep Learning, The MIT Press.

Hearty, J.. (2016), Advanced Machine Learning with Python, Packt Publishing Ltd.

Leskovec, J. Rajaraman, A., & Ullman, J.. (2014), Mining of Massive Datasets, Cambridge University Press.

Wickham, H. & Grolemund, G.. (2017), R for Data Science: Import, Tidy, Transform, Visualize, and Model Data, O'Reilly.

### This module does not have any article/paper resources

# Other Resources

[Website], Datacamp, http://www.datacamp.com

[Website], KD Nuggest, http://www.kdnuggets.com

[Website], R Bloggers, http://www.r-bloggers.com

## Discussion Note: