

H9DMML2: Data Mining and Machine Learning II

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| Module Code: | H9DMML2 |
| Long Title | Data Mining and Machine Learning II APPROVED |
| Title | Data Mining and Machine Learning II |
| Module Level: | LEVEL 9 |
| EQF Level: | 7 |
| EHEA Level: | Second Cycle |
| Credits: | 10 |
| Module Coordinator: | MICHAEL BRADFORD |
| Module Author: | Jenette Carson |
| Departments: | School of Computing |
| Specifications of the qualifications and experience required of staff | PhD/MSc degree in a computing or cognate discipline. May have industry experience also. |
| Learning Outcomes | |
| <i>On successful completion of this module the learner will be able to:</i> | |
| # | Learning Outcome Description |
| LO1 | Critically analyse advanced data mining and knowledge discovery methodologies in order to assess best practice guidance when applied to complex data mining problems |
| LO2 | Investigate and evaluate key concepts and advanced data mining techniques and assess when to apply such techniques on complex datasets and problem domains. |
| LO3 | Contextualise, research and utilise current data mining approaches, applications and technologies in order to provide strategies to address processing of datasets with a variety of characteristics |
| LO4 | Critically review and apply appropriate data mining research and assess research methods |
| Dependencies | |
| Module Recommendations | |
| No recommendations listed | |
| Co-requisite Modules | |
| No Co-requisite modules listed | |
| Entry requirements | A level 8 degree or its equivalent in any discipline |

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| Module Content & Assessment | | | |
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| 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. | | | |
| Ensembles 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 volumes required). | | | |
| 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: | Yes | | |
| Assessment Description: 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: | Project | % of total: | 50 |
| Assessment Date: | n/a | Outcome addressed: | 3,4 |
| Non-Marked: | 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: | Terminal Exam | % of total: | 50 |
| Assessment Date: | End-of-Semester | Outcome addressed: | 1,2 |
| Non-Marked: | No | | |
| 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 <i>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.</i> | | | |
| Reassessment Description The repeat strategy for this module is by repeat assessment/project that covers all learning outcomes. | | | |

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| 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 | | | | 48.00 |

| Module Resources | |
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| <i>Recommended Book Resources</i> | |
| <p>Hastie, T., Tibshirani, R. & Friedman, J.. (2016), The Elements of Statistical Learning: Data Mining, Inference, and Prediction (2nd ed), Springer Series in Statistics.</p> <p>James, G., Witten, D., Hastie, T. & Tibshirani, R.. (2017), An Introduction to Statistical Learning: with Applications in R, Springer Texts in Statistics.</p> <p>Kuhn, M. & Johnson, K.. (2013), Applied Predictive Modeling, Springer.</p> <p>Shalev-Shwartz, S. & Ben-David, S.. (2014), Understanding Machine Learning: From Theory to Algorithms, Cambridge University Press.</p> | |
| <i>Supplementary Book Resources</i> | |
| <p>Downey, B.. (2014), Think Stats: Exploratory Data Analysis, (2nd ed).</p> <p>Goodfellow, I., Bengio, Y., & Courville, A.. (2016), Deep Learning, The MIT Press.</p> <p>Hearty, J.. (2016), Advanced Machine Learning with Python, Packt Publishing Ltd.</p> <p>Leskovec, J. Rajaraman, A., & Ullman, J.. (2014), Mining of Massive Datasets, Cambridge University Press.</p> <p>Wickham, H. & Grolemund, G.. (2017), R for Data Science: Import, Tidy, Transform, Visualize, and Model Data, O'Reilly.</p> | |
| <i>This module does not have any article/paper resources</i> | |
| <i>Other Resources</i> | |
| <p>[Website], Datacamp, http://www.datacamp.com</p> <p>[Website], KD Nuggest, http://www.kdnuggets.com</p> <p>[Website], R Bloggers, http://www.r-bloggers.com</p> | |
| Discussion Note: | |