



**INTERNATIONAL QUALIFICATIONS
AND ASSESSMENT CENTRE (IQAC)**



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| Programme | LEVEL EXTENDED DIPLOMA IN ARTIFICIAL INTELLIGENCE | | |
| Unit Number/ Unit Title | UNIT 5 MACHINE LEARNING FUNDAMENTALS | | |
| Cohort Code: | L04MLF-U5 | | |
| Unit Level | 4 | | |
| Total GLH | Total qualification time 200/ Total Guided learning hours 90/ Self-guided learning hours 110 | | |
| Credits | 20 CATS/ 10 ECTS | | |
| Lecturer | | | |
| Start Date | | End Date | |

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| Unit Aims | The unit will cover key machine learning concepts such as supervised, unsupervised, and reinforcement learning, along with practical training on popular algorithms and frameworks. Students will develop skills in data preprocessing, model evaluation, and optimization to build effective and reliable machine learning solutions. This foundation will enable learners to apply machine learning techniques to real-world problems across various domains. |
| Differentiation Strategies <i>(e.g. planned activities or support for individual learners according to their needs)</i> | <p>The total number of students to be in the lesson is approximately 20. This is a multicultural group of students predominantly between the ages of 24 – 45, with numerous ethnic, gender, and creed background. These are UK academic level 5 students; hence it is assumed that they have practical, theoretical, or technological knowledge and understanding of a subject or field of work to find ways forward in broadly defined, complex contexts. These students must be able to generate information, evaluate, synthesise the use information from a variety of sources. Various approaches to addressing the various identified students needs will be adopted throughout the lesson. Such will include:-</p> <ol style="list-style-type: none">1. Progressive tasks2. Digital resources3. Verbal support4. Variable outcomes |

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| | 5. Collaborative learning 6. Ongoing assessment 7. Flexible-pace learning |
| Equality & Diversity | Variety of teaching techniques will be employed to ensure that the needs of each individual learner are met. |
| Safeguarding & Prevent | Safeguarding policies and the Prevent duty are strictly observed to ensure the safety, well-being, and inclusivity of all students and staff. |
| Health & Safety | SIRM H&S policies will be maintained. |
| Learning Resources | Teaching and Learning Materials |
| | <ul style="list-style-type: none"> • "Hands-On Machine Learning with Scikit-Learn, Keras, and TensorFlow" by Aurélien Géron • "Machine Learning Yearning" by Andrew Ng • "Pattern Recognition and Machine Learning" by Christopher Bishop. |

| Learning Outcome | Assessment Criteria |
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| LO1. Understand the basic concepts of machine learning | 1.1. Define machine learning and its types (supervised, unsupervised, and reinforcement learning). 1.2. Explain key machine learning algorithms and their applications. |
| LO2. Implement supervised learning algorithms. | 2.1. Develop regression models to predict continuous outcomes. 2.2. Implement classification algorithms to predict categorical outcomes. |
| LO3. Apply unsupervised learning techniques. | 3.1. Perform clustering analysis to group similar data points. 3.2: Apply dimensionality reduction techniques to simplify data. |
| LO4. Evaluate machine learning models. | 4.1: Use cross-validation techniques to assess model performance. 4.2: Interpret confusion matrices and other evaluation metrics. |

| No | Learning Outcome / Topic | Learning and Teaching Activities | Which assessment criteria does the session relate to? | Day/month/year/ signature |
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| 1. | Introduction to Machine Learning (ML) | Introduction to Machine Learning (ML) Definition, importance, and applications of ML. | LO1: Understand the Basic Concepts of Machine Learning | |
| 2. | Types of Machine Learning | Types of Machine Learning Supervised, unsupervised, and reinforcement learning with examples. | LO1: Understand the Basic Concepts of Machine Learning | |
| 3. | Key Terminologies in ML | Key Terminologies in ML Features, labels, training data, testing data, model, overfitting, underfitting. | LO1: Understand the Basic Concepts of Machine Learning | |
| 4. | Overview of Machine Learning Algorithms | Overview of Machine Learning Algorithms Linear regression, decision trees, k-means, neural networks (brief). | LO1: Understand the Basic Concepts of Machine Learning | |
| 5. | Real-world Applications of ML | Real-world Applications of ML Healthcare, finance, marketing, autonomous systems, etc. | LO1: Understand the Basic Concepts of Machine Learning | |
| 6. | Introduction to Supervised Learning | Introduction to Supervised Learning Labeled data, regression vs. classification. | LO2: Implement Supervised Learning Algorithms | |
| 7. | Linear Regression | Linear Regression Concept, cost function, gradient descent. | LO2: Implement Supervised Learning Algorithms | |
| 8. | Half-Term Exam | <ul style="list-style-type: none"> - Review of LO1 and LO2 topics - Practice questions and mock assessment - Half-term assessment based on LO1 (theory) | | |
| 9. | Building a Regression Model | Building a Regression Model Hands-on implementation (e.g., predicting house prices). | LO2: Implement Supervised Learning Algorithms | |

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| 10. | Logistic Regression for Classification | Logistic Regression for Classification Sigmoid function, decision boundary. | LO2: Implement Supervised Learning Algorithms | |
| 11. | Decision Trees & Random Forests | Decision Trees & Random Forests Splitting criteria (Gini, entropy), ensemble methods. | LO2: Implement Supervised Learning Algorithms | |
| 12. | Support Vector Machines (SVM) | Support Vector Machines (SVM) Hyperplanes, kernel trick. | LO2: Implement Supervised Learning Algorithms | |
| 13. | Naive Bayes Classifier | Naive Bayes Classifier Probabilistic approach for text classification. | LO2: Implement Supervised Learning Algorithms | |
| 14. | Final Exam Preparation & Review | - Comprehensive review of all learning outcomes - Practice questions and revision of key topics | | |
| 15. | Final Exam | - Final-term assessment covering all learning outcomes (theory and practical elements) | LO1 LO2 | |
| 16. | Feedback & Reflection | - Review of final exam - Individual feedback on performance - Reflective discussion on key learning points | | |
| 17. | Model Training & Testing | Model Training & Testing Splitting data (train-test-validation sets). | LO2: Implement Supervised Learning Algorithms | |
| 18. | Introduction to Unsupervised Learning | Introduction to Unsupervised Learning Clustering vs. dimensionality reduction. | LO3: Apply Unsupervised Learning Techniques | |
| 19. | K-Means Clustering | K-Means Clustering Algorithm steps, choosing 'k' (elbow method). | LO3: Apply Unsupervised Learning Techniques | |
| 20. | Hierarchical Clustering | Hierarchical Clustering Agglomerative vs. divisive approaches. | LO3: Apply Unsupervised Learning Techniques | |

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| 21. | DBSCAN (Density-Based Clustering) | DBSCAN (Density-Based Clustering) Core points, noise, and applications. | LO3: Apply Unsupervised Learning Techniques | |
| 22. | Principal Component Analysis (PCA) | Principal Component Analysis (PCA) Variance, eigenvectors, reducing dimensions. | LO3: Apply Unsupervised Learning Techniques | |
| 23. | Half-Term Exam | Case Study: End-to-End ML Project From data preprocessing to model deployment. | LO1 LO2 LO3 | |
| 24. | t-SNE for Visualization | t-SNE for Visualization Non-linear dimensionality reduction. | LO3: Apply Unsupervised Learning Techniques | |
| 25. | Model Evaluation Metrics | Model Evaluation Metrics Accuracy, precision, recall, F1-score. | LO4: Evaluate Machine Learning Models | |
| 26. | Confusion Matrix | Confusion Matrix TP, TN, FP, FN, ROC curves. | LO4: Evaluate Machine Learning Models | |
| 27. | Cross-Validation Techniques | Cross-Validation Techniques k-fold, stratified k-fold, LOOCV. | LO4: Evaluate Machine Learning Models | |
| 28. | Bias-Variance Tradeoff and Hyperparameter Tuning | Bias-Variance Tradeoff Overfitting vs. underfitting solutions. Hyperparameter Tuning Grid search, random search. | LO4: Evaluate Machine Learning Models | |
| 29. | Final Exam Preparation & Review | LO1, LO2, LO3, LO4 | LO1, LO2, LO3, LO4 | |
| 30. | Final Exam | | LO1, LO2, LO3, LO4 | |