



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



Approved

<b>Programme</b>	<b>LEVEL 5 EXTENDED DIPLOMA IN ARTIFICIAL INTELLIGENCE</b>		
<b>Unit Number/ Unit Title</b>	<b>UNIT 7 ADVANCED MACHINE LEARNING</b>		
<b>Cohort Code:</b>	L05AML-U7		
<b>Unit Level</b>	5		
<b>Total GLH</b>	Total qualification time 200/ Total Guided learning hours 90/ Self-guided learning hours 110		
<b>Credits</b>	20 CATS/ 10 ECTS		
<b>Lecturer</b>			
<b>Start Date</b>		<b>End Date</b>	

<b>Unit Aims</b>	Students will gain hands-on experience with algorithms such as ensemble methods, deep learning, and reinforcement learning. The unit will also emphasize model evaluation, tuning, and ethical considerations in deploying machine learning solutions.
<b>Differentiation Strategies (e.g. planned activities or support for individual learners according to their needs)</b>	<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"><li>1. Progressive tasks</li><li>2. Digital resources</li></ol>

	<p>3. Verbal support          4. Variable outcomes          5. Collaborative learning          6. Ongoing assessment</p> <p>Flexible-pace learning</p>
<b>Equality &amp; Diversity</b>	Variety of teaching techniques will be employed to ensure that the needs of each individual learner are met.
<b>Safeguarding &amp; Prevent</b>	Safeguarding policies and the Prevent duty are strictly observed to ensure the safety, well-being, and inclusivity of all students and staff.
<b>Health &amp; Safety</b>	SIRM H&S policies will be maintained.
<b>Learning Resources</b>	<p style="text-align: center;"><b>Teaching and Learning Materials</b></p> <ul style="list-style-type: none"> <li>• "Deep Learning" by Ian Goodfellow, Yoshua Bengio, and Aaron Courville</li> <li>• "Pattern Recognition and Machine Learning" by Christopher Bishop</li> <li>• "Machine Learning: A Probabilistic Perspective" by Kevin P. Murphy.</li> </ul>

Learning Outcome	Assessment Criteria
<b>LO1. Explore advanced machine learning algorithms.</b>	1.1: Understand advanced supervised learning techniques such as ensemble methods. 1.2: Implement support vector machines and kernel methods.
<b>LO2. Apply deep learning techniques.</b>	2.1: Develop and train deep learning models. 2.2: Utilize convolutional neural networks (CNNs) for image recognition tasks.
<b>LO3. Optimize machine learning models.</b>	3.1: Apply hyperparameter tuning techniques. 3.2: Use model evaluation metrics to optimize performance.
<b>LO4. Develop solutions for complex problems using machine learning.</b>	4.1: Apply advanced machine learning techniques to real-world datasets. 4.2: Evaluate and improve model robustness and generalization.

No	Learning Outcome / Topic	Learning and Teaching Activities	Which assessment criteria does the session relate to?	Day/month/year/ signature
1.	<b>Ensemble Methods Fundamentals</b>	<b>Ensemble Methods Fundamentals</b> Bagging vs. Boosting, bias-variance tradeoff	LO1: Advanced ML Algorithms	
2.	<b>Random Forests Deep Dive</b>	<b>Random Forests Deep Dive</b> Feature importance, OOB error, scikit-learn implementation	LO1: Advanced ML Algorithms	
3.	<b>Gradient Boosting Machines (GBM)</b>	<b>Gradient Boosting Machines (GBM)</b> XGBoost, LightGBM, CatBoost (comparative analysis)	LO1: Advanced ML Algorithms	
4.	<b>Support Vector Machines (SVMs)</b>	<b>Support Vector Machines (SVMs)</b> Hard/soft margins, kernel trick (RBF, polynomial)	LO1: Advanced ML Algorithms	
5.	<b>Kernel Methods Workshop</b>	<b>Kernel Methods Workshop</b> Implementing custom kernels for non-linear problems	LO1: Advanced ML Algorithms	
6.	<b>Neural Networks Review</b>	<b>Neural Networks Review</b> Feedforward networks, backpropagation refresher	LO2: Deep Learning Techniques	
7.	<b>Deep Learning Frameworks</b>	<b>Deep Learning Frameworks</b> TensorFlow vs. PyTorch comparison, GPU utilization	LO2: Deep Learning Techniques	
8.	Half-Term Exam	<ul style="list-style-type: none"> <li>- Review of LO1 topics</li> <li>- Practice questions and mock assessment</li> <li><b>- Half-term assessment</b> based on LO1 (theory)</li> </ul>	LO1 LO2	
9.	<b>Training Deep Networks</b>	<b>Training Deep Networks</b> Vanishing gradients, ReLU, batch normalization	LO2: Deep Learning Techniques	
10.	<b>CNNs for Image Recognition</b>	<b>CNNs for Image Recognition</b> Convolutional layers, pooling, architecture design	LO2: Deep Learning Techniques	
11.	<b>Transfer Learning Lab</b>	<b>Transfer Learning Lab</b> Fine-tuning pre-trained models (ResNet, VGG)	LO2: Deep Learning Techniques	
12.	<b>Advanced CNN Architectures</b>	<b>Advanced CNN Architectures</b> Inception modules, residual connections (ResNet)	LO2: Deep Learning Techniques	

13.	<b>Hyperparameter Tuning</b>	<b>Hyperparameter Tuning</b> Grid search, random search, Bayesian optimization	LO3: Model Optimization	
14.	Final Exam Preparation & Review	- Comprehensive review of all learning outcomes - Practice questions and revision of key topics		
15.	Final Exam	- <b>Final-term assessment</b> covering all learning outcomes (theory and practical elements)		
16.	Feedback & Reflection	- Review of final exam - Individual feedback on performance - Reflective discussion on key learning points		
17.	<b>AutoML Tools</b>	<b>AutoML Tools</b> H2O.ai, AutoKeras (hands-on demo)	LO3: Model Optimization	
18.	<b>Model Evaluation Metrics</b>	<b>Model Evaluation Metrics</b> Precision-recall tradeoff, AUC-ROC, F $\beta$ scores	LO3: Model Optimization	
19.	<b>Model Interpretability</b>	<b>Model Interpretability</b> SHAP values, LIME, feature attribution	LO3: Model Optimization	
20.	<b>Bias Detection &amp; Mitigation</b>	<b>Bias Detection &amp; Mitigation</b> Fairness metrics (demographic parity, equalized odds)	LO3: Model Optimization	
21.	<b>Time Series Forecasting</b>	<b>Time Series Forecasting</b> LSTMs, Prophet (Facebook's model)	LO4: Complex Problem-Solving	
22.	<b>Natural Language Processing (NLP)</b>	<b>Natural Language Processing (NLP)</b> Word embeddings (Word2Vec, GloVe), transformer overview	LO4: Complex Problem-Solving	
23.	Half-Term Exam	<i>Integrated Projects &amp; Assessments</i> <b>Kaggle-Style Competition</b> End-to-end pipeline on a complex dataset (e.g., satellite images)		
24.	<b>Anomaly Detection</b>	<b>Anomaly Detection</b> Autoencoders, GANs for fraud detection	LO4: Complex Problem-Solving	
25.	<b>Reinforcement Learning Primer</b>	<b>Reinforcement Learning Primer</b> Q-learning, Deep Q-Networks (DQN) concepts	LO4: Complex Problem-Solving	
26.	<b>Model Deployment Basics</b>	<b>Model Deployment Basics</b> Flask/Docker for ML APIs, ONNX format	LO4: Complex Problem-Solving	

27.	<b>Model Robustness Challenge</b>	<b>Model Robustness Challenge</b> Adversarial attacks (FGSM), defensive distillation	LO4: Complex Problem-Solving	
28.	<b>Capstone Project</b>	<b>Capstone Project</b> Solve a real-world problem (healthcare, finance, or IoT)	LO4: Complex Problem-Solving	
29.	Final Exam Preparation & Review	LO1, LO2, LO3, LO4	LO1, LO2, LO3, LO4	
30.	Final Exam		LO1, LO2, LO3, LO4	