



Module Specification

Machine Learning

Version: 2026-27, v3.0, Approved

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Part 1: Information

Module title: Machine Learning

Module code: UFCFAS-15-2

Level: Level 5

For implementation from: 2026-27

UWE credit rating: 15

ECTS credit rating: 7.5

College: College of Arts, Technology and Environment

School: CATE School of Computing and Creative Technologies

Partner institutions: None

Field: Computer Science and Creative Technologies

Module type: Module

Pre-requisites: Artificial Intelligence I 2025-26

Excluded combinations: None

Co-requisites: Artificial Intelligence II 2026-27

Continuing professional development: No

Professional, statutory or regulatory body requirements: None

Part 2: Description

Overview: This module will introduce students to a range of contemporary machine learning techniques, their principles, characteristics and applications to real-world problems, and how to select, apply and evaluate solutions based on these techniques using the state-of-the-art software tools.

Features: Not applicable

Educational aims: The module places an emphasis on the practical applications of Machine Learning techniques to real-world problems, enabling students to gain hands-on experience in developing machine learning solutions to real-world problems.

Outline syllabus: Topics are likely to include but are not limited to:

Introduction

Overview of supervised learning, a range of contemporary machine learning techniques; Example applications for such techniques (e.g. NLP, computer vision, text mining); Potential wider societal and ethical implications will be highlighted.

Key technical issues in using ML techniques to solve real-world problems

Training and testing – stopping, imbalance, missing variables, bias and variances in data sets, features, pre-processing, etc. i.e. the realities of handling data

Formulating application problems, involving various types of data and problems with implications of algorithm choices, e.g. text, images, generating different types of learned models, e.g. classification vs regression

Kernel methods

What are kernel methods? Why do we need them?

Support Vector Machines (SVMs) – a high-level overview of SVMs

What is Support Vector Machine? How does it work? How to implement it? How to tune parameters of SVM?

Applications of kernel methods, e.g. classification and regression

Ensembles: bagging/boosting. Random forests

What are ensemble methods? Why do we need them?

What is bagging/boosting? How does it work? How to implement it?

What are random forests? How do they work? How to implement them?

Convolutional Neural Networks (CNNs)

What are CNNs? Why do we need them? How does a CNN work (convolution, ReLu, pooling, finding optimal weights (e.g. backpropagation), hyperparameters)?

How to implement it?

Deep learning tricks: Concepts of overfitting and regularization, data augmentation, dropout, early stopping

Applications of CNNs, image classification, NLP (e.g. text classification).

Part 3: Teaching and learning methods

Teaching and learning methods: Lectures will introduce the core concepts and algorithmic essences of each topic listed in the syllabus. Each topic will be explained and illustrated with intuitive examples, expanded with developing the practical solutions to the real-world problems.

Practical classes will provide supervised activities to cover different problem domains, with an emphasis on developing students' hands-on experience in solving real-world problems using contemporary machine learning techniques.

Scheduled teaching and learning study hours include lectures , practical classes/tutorials/project supervision/software demonstration.

Independent learning study hours include engaged with essential and exploratory reading, practical study, assignment preparation and completion etc.

Module Learning outcomes: On successful completion of this module students will achieve the following learning outcomes.

MO1 Select and apply machine learning algorithms to formulate solutions to different types of machine learning problems.

MO2 Implement and evaluate contemporary machine learning solutions to application problems using contemporary frameworks.

MO3 Demonstrate an awareness of the ethical and societal implications of machine learning solutions.

Hours to be allocated: 150

Contact hours:

Independent study/self-guided study = 114 hours

Face-to-face learning = 36 hours

Reading list: The reading list for this module can be accessed at [readinglists.uwe.ac.uk](https://uwe.rl.talis.com/modules/ufcfas-15-2.html) via the following link <https://uwe.rl.talis.com/modules/ufcfas-15-2.html>

Part 4: Assessment

Assessment strategy: A series of lab-based practical exercises with one-to-one demonstrations to tutors will provide rich opportunities for formative feedback.

At both first sit and resit, the summative assessment will consist of:

An individual project creating and evaluating a machine learning solution to a real-world application problem. Assessment of this will be an online assignment submission of a 6-page report following the given formatting guidelines, describing problem identification, analysis and formulation, algorithm selection and application, software implementation including testing and evaluation, and documentation of the project.

The resit coursework will take the same form as the first sit.

Assessment tasks:

Project (First Sit)

Description: An individual project requiring the production of software and a written report.

Weighting: 100 %

Final assessment: Yes

Group work: No

Learning outcomes tested: MO1, MO2, MO3

Project (Resit)

Description: An individual project requiring the production of software and a written report.

Weighting: 100 %

Final assessment: Yes

Group work: No

Learning outcomes tested: MO1, MO2, MO3

Part 5: Contributes towards

This module contributes towards the following programmes of study:

Computer Science (Artificial Intelligence) {Foundation} [GCET] BSc (Hons) 2024-25

Computer Science {Foundation} [Frenchay] BSc (Hons) 2024-25

Computer Science (Artificial Intelligence) {Foundation} [GCET] DipHE 2024-25

Computer Science (Artificial Intelligence) {Foundation} [GCET] BSc (Hons) 2024-25

Computer Science {Foundation} [Frenchay] BSc (Hons) 2024-25

Computer Science [Frenchay] BSc (Hons) 2025-26

Computer Science [Phenikaa] BSc (Hons) 2025-26

Computer Science (Artificial Intelligence) [NepalBrit] BSc (Hons) 2025-26

Computer Science [Villa] BSc (Hons) 2025-26

Computer Science {Dual} [Taylors] BSc (Hons) 2025-26