



## Applied Machine Learning (COMP0081)

### Description

#### Aims:

To give a detailed understanding of topics related to efficient implementation of large-scale machine learning with a focus on optimisation in both linear and non-linear machine learning models.

Students will also gain experience in tackling real world problems through solving online machine learning challenges.

A key aim is that students understand the challenges of optimisation and associated time and space complexities of various approaches.

#### Learning outcomes:

On successful completion of the module, a student will:

1. have a good understanding of practical issues arising in implementing machine learning in practice, including engineering challenges as well as the data ethics considerations;

2. become familiar with techniques used in practice to solve real world machine learning problems and will be able to apply these techniques;

#### Content:

Second Order Optimisation methods (Newton and Quasi Newton approaches and Conjugate Gradients);

Methods for solving Large Scale Linear, including Conjugate Gradients;

Automatic Differentiation methods for efficiently computing first and second order gradients;

Classical methods for Regression and Classification including linear and logistic regression;

Clustering Methods for Unsupervised Learning;

Matrix and Tensor Factorisation;

### Key information

<b>Year</b>	2018/19
<b>Credit value</b>	15 (150 study hours)
<b>Delivery</b>	PGT L7, Campus-based
<b>Reading List</b>	<a href="#">View on UCL website</a>
<b>Tutor</b>	<a href="#">Dr Dmitry Adamskiy</a>
<b>Term</b>	Term 2
<b>Timetable</b>	<a href="#">View on UCL website</a>

### Assessment



■	Written examination (main exam period): 70%
■	Coursework: 10%
■	Coursework: 10%
■	Coursework: 10%

### Find out more

For more information about the department, programmes, relevant open days and to browse other modules, visit [ucl.ac.uk](http://ucl.ac.uk)

Visualisation methods including tSNE;  
Ensembling, Gradient Boosting Machines;  
Data Ethics;

**Prerequisites:**

To be eligible to select this module, students must have:

understanding and abilities with Linear Algebra, Multivariate Calculus and Probability at mathematics FHEQ Level 4 (Undergraduate Year 1);

and familiarity with coding a high level language in order to complete assessments (strongly recommend that students are skilled in Python);

*And, must have taken in Term 1:*

Introduction to Machine Learning (COMP0088);  
or Supervised Learning (COMP0078);

*And, are recommended to have taken in Term 1:*

Graphical Models (COMP0080);  
or Probabilistic and Unsupervised Learning (COMP0086);

Please note that this module is not an introduction to machine learning.



## Applied Machine Learning (COMP0081)

### Description

#### Aims:

To give a detailed understanding of topics related to efficient implementation of large-scale machine learning with a focus on optimisation in both linear and non-linear machine learning models.

Students will also gain experience in tackling real world problems through solving online machine learning challenges.

A key aim is that students understand the challenges of optimisation and associated time and space complexities of various approaches.

#### Learning outcomes:

On successful completion of the module, a student will:

1. have a good understanding of practical issues arising in implementing machine learning in practice, including engineering challenges as well as the data ethics considerations;

2. become familiar with techniques used in practice to solve real world machine learning problems and will be able to apply these techniques;

#### Content:

Second Order Optimisation methods (Newton and Quasi Newton approaches and Conjugate Gradients);

Methods for solving Large Scale Linear, including Conjugate Gradients;

Automatic Differentiation methods for efficiently computing first and second order gradients;

Classical methods for Regression and Classification including linear and logistic regression;

Clustering Methods for Unsupervised Learning;

Matrix and Tensor Factorisation;

### Key information

<b>Year</b>	2018/19
<b>Credit value</b>	15 (150 study hours)
<b>Delivery</b>	UGM L7, Campus-based
<b>Reading List</b>	<a href="#">View on UCL website</a>
<b>Tutor</b>	<a href="#">Dr Dmitry Adamskiy</a>
<b>Term</b>	Term 2
<b>Timetable</b>	<a href="#">View on UCL website</a>

### Assessment



■	Written examination (main exam period): 70%
■	Coursework: 10%
■	Coursework: 10%
■	Coursework: 10%

### Find out more

For more information about the department, programmes, relevant open days and to browse other modules, visit [ucl.ac.uk](http://ucl.ac.uk)

Visualisation methods including tSNE;  
Ensembling, Gradient Boosting Machines;  
Data Ethics;

**Prerequisites:**

To be eligible to select this module, students must have:

understanding and abilities with Linear Algebra, Multivariate Calculus and Probability at mathematics FHEQ Level 4 (Undergraduate Year 1);

and familiarity with coding a high level language in order to complete assessments (strongly recommend that students are skilled in Python);

*And, must have taken in Term 1:*

Introduction to Machine Learning (COMP0088);  
or Supervised Learning (COMP0078);

*And, are recommended to have taken in Term 1:*

Graphical Models (COMP0080);  
or Probabilistic and Unsupervised Learning (COMP0086);

Please note that this module is not an introduction to machine learning.