



# COMP7220

## Data Science and Machine Learning

Session 1, Special circumstances 2021

*School of Computing*

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#### **Notice**

As part of [Phase 3 of our return to campus plan](#), most units will now run tutorials, seminars and other small group activities on campus, and most will keep an online version available to those students unable to return or those who choose to continue their studies online.

To check the availability of face-to-face activities for your unit, please go to [timetable viewer](#). To check detailed information on unit assessments visit your unit's iLearn space or consult your unit convenor.

## General Information

### Unit convenor and teaching staff

Convenor, Lecturer

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Tutor, Lecturer

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### Credit points

10

### Prerequisites

Admission to MRes

### Corequisites

### Co-badged status

COMP8220

### Unit description

This unit begins with conventional machine learning techniques for constructing classifiers and regression models, including widely applicable standard techniques such as Naive Bayes, decision trees, logistic regression and support vector machines (SVMs); in this part, given required prior knowledge of machine learning, we focus on more advanced aspects. We then look in detail at deep learning and other state-of-the-art approaches. We discuss in detail the advantages and disadvantages of each method, in terms of computational requirements, ease of use, and performance, and we study the practical application of these methods in a number of use cases.

## Important Academic Dates

Information about important academic dates including deadlines for withdrawing from units are available at <https://www.mq.edu.au/study/calendar-of-dates>

## Learning Outcomes

On successful completion of this unit, you will be able to:

**ULO1:** Derive algorithms to solve machine learning problems based on an understanding of how machine learning and data science problems are mathematically formulated and analysed.

**ULO2:** Create machine learning solutions to data science problems by identifying and applying appropriate algorithms and implementations.

**ULO3:** Analyse real-world data science problems, identify which methods are appropriate, organise the data appropriately, apply one or more methods, and evaluate the quality of the solution.

**ULO4:** Evaluate one or more approaches to advanced topics in machine learning and data science and report the findings in oral and written form.

## General Assessment Information

### Late Submission

No extensions will be granted without an approved application for Special Consideration. There will be a deduction of 20% of the total available marks made from the total awarded mark for each 24 hour period or part thereof that the submission is late. For example, 25 hours late in submission for an assignment worth 10 marks – 40% penalty or 4 marks deducted from the total. No submission will be accepted after solutions have been posted.

## Assessment Tasks

Name	Weighting	Hurdle	Due
<a href="#">Individual Project</a>	30%	No	Interim: week 6; final: week 13
<a href="#">Practical Exercises</a>	30%	No	Throughout the semester
<a href="#">Major Project</a>	40%	No	Interim: first week of break; final: week 13

### Individual Project

Assessment Type <sup>1</sup>: Project

Indicative Time on Task <sup>2</sup>: 25 hours

Due: **Interim: week 6; final: week 13**

Weighting: **30%**

In contrast to the Major Project, in this one the student will select a dataset from an appropriate domain, and then design and implement a solution to a task on this chosen dataset. The deliverables will be the implementation and a report describing this implementation.

On successful completion you will be able to:

- Derive algorithms to solve machine learning problems based on an understanding of how machine learning and data science problems are mathematically formulated and analysed.
- Create machine learning solutions to data science problems by identifying and applying appropriate algorithms and implementations.
- Analyse real-world data science problems, identify which methods are appropriate, organise the data appropriately, apply one or more methods, and evaluate the quality of the solution.
- Evaluate one or more approaches to advanced topics in machine learning and data science and report the findings in oral and written form.

## Practical Exercises

Assessment Type <sup>1</sup>: Problem set

Indicative Time on Task <sup>2</sup>: 30 hours

Due: **Throughout the semester**

Weighting: **30%**

These will consist of practical exercises set throughout the semester.

On successful completion you will be able to:

- Derive algorithms to solve machine learning problems based on an understanding of how machine learning and data science problems are mathematically formulated and analysed.
- Create machine learning solutions to data science problems by identifying and applying appropriate algorithms and implementations.

## Major Project

Assessment Type <sup>1</sup>: Project

Indicative Time on Task <sup>2</sup>: 30 hours

Due: **Interim: first week of break; final: week 13**

Weighting: **40%**

The student will apply knowledge of conventional machine learning and deep learning to design and implement a solution to a (classification or other) task on a defined dataset. The deliverables

will be the implementation and a report describing this implementation.

On successful completion you will be able to:

- Derive algorithms to solve machine learning problems based on an understanding of how machine learning and data science problems are mathematically formulated and analysed.
- Create machine learning solutions to data science problems by identifying and applying appropriate algorithms and implementations.
- Analyse real-world data science problems, identify which methods are appropriate, organise the data appropriately, apply one or more methods, and evaluate the quality of the solution.
- Evaluate one or more approaches to advanced topics in machine learning and data science and report the findings in oral and written form.

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<sup>1</sup> If you need help with your assignment, please contact:

- the academic teaching staff in your unit for guidance in understanding or completing this type of assessment
- the [Writing Centre](#) for academic skills support.

<sup>2</sup> Indicative time-on-task is an estimate of the time required for completion of the assessment task and is subject to individual variation

## Delivery and Resources

### Classes

- **Classes:** There will be a two hour lecture each week, and additionally a small class that will focus on working through practical tasks.
- **Textbook:** The main textbook for the unit is Aurélien Géron (2019) "[Hands-On Machine Learning with Scikit-Learn, Keras and TensorFlow](#)" (2nd edition; September 2019). This is available through the MQ library (MQ has an arrangement with publisher O'Reilly: you can register at O'Reilly using your MQ email, and get access to the book there). The book comes with source code that is available from <https://github.com/ageron/handson-ml2>. A supplementary source of material for a deeper understanding of the theoretical material is Trevor Hastie, Robert Tibshirani and Jerome Friedman (2009; corrected 12th printing Jan 2017) "The Elements of Statistical Learning: Data Mining, Inference, and Prediction." A freely downloadable pdf is available at the [first author's webpage](#).

## Background Material

- The unit requires a sound background in programming, and particularly Python. If you feel you need a refresher on Python (or an introduction from scratch, as long as you're a quick and independent learner), there's a popular tutorial at <http://learnpython.org/>. This goes all the way from basic programming to the mathematical and data science libraries used by Python, like numpy and pandas. There's also the resources at the Python website at python.org, like the [Beginner's Guide](#).
- For a refresher on linear algebra as it is relevant to machine learning, Jason Brownlee (2018) "Basics of Linear Algebra for Machine Learning" has useful material that's linked to Python data structures. (The book used to have a freely available pdf, but this seems to have disappeared. It is published by [Machine Learning Mastery](#).)

## Unit Webpage and Technology Used and Required

- [iLearn](#) is going to be used as a main web server for the unit.
- The programming language for the unit will be Python. The "conventional" machine learning section will use Python's scikit-learn, and the deep learning section will use TensorFlow and Keras.
- For the most part, programming will be done via Jupyter notebooks. We'll typically be running these notebooks on [Google Colab](#).

## Unit Schedule

Week	Topic	Readings (from Géron)
1	What is Machine Learning?	Ch 1
2	Workflow of a Machine Learning Project	Ch 2
3	Support Vector Machines and Decision Trees	Ch 3-6
4	Ensemble Learning, Random Forests, and Dimensionality Reduction	Ch 7-8
5	Handling Text Data	supplementary notes
6	Classical Reinforcement Learning	supplementary notes
	break	
7-8	Introduction to Artificial Neural Networks: <ul style="list-style-type: none"> <li>• ANN basics</li> <li>• Multi-Layer Perceptrons</li> <li>• The Tensorflow and Keras frameworks</li> </ul>	Ch 10-11

Week	Topic	Readings (from Géron)
9-10	Deep Neural Networks <ul style="list-style-type: none"> <li>• The structure of deep NNs</li> <li>• Convolutional NNs</li> <li>• Practical issues in training NNs</li> </ul>	Ch 11-14, supplementary notes
11-12	NNs for sequences, and advanced topics: <ul style="list-style-type: none"> <li>• Recurrent NNs</li> <li>• Autoencoders</li> <li>• Reinforcement Learning</li> </ul>	Ch 15 and onwards, supplementary notes
13	Unit review	

## Policies and Procedures

Macquarie University policies and procedures are accessible from [Policy Central \(https://policies.mq.edu.au\)](https://policies.mq.edu.au). Students should be aware of the following policies in particular with regard to Learning and Teaching:

- [Academic Appeals Policy](#)
- [Academic Integrity Policy](#)
- [Academic Progression Policy](#)
- [Assessment Policy](#)
- [Fitness to Practice Procedure](#)
- [Grade Appeal Policy](#)
- [Complaint Management Procedure for Students and Members of the Public](#)
- [Special Consideration Policy](#)

Students seeking more policy resources can visit [Student Policies \(https://students.mq.edu.au/support/study/policies\)](https://students.mq.edu.au/support/study/policies). It is your one-stop-shop for the key policies you need to know about throughout your undergraduate student journey.

To find other policies relating to Teaching and Learning, visit [Policy Central \(https://policies.mq.edu.au\)](https://policies.mq.edu.au) and use the [search tool](#).

## Student Code of Conduct

Macquarie University students have a responsibility to be familiar with the Student Code of Conduct: <https://students.mq.edu.au/admin/other-resources/student-conduct>

## Results

Results published on platform other than [eStudent](#), (eg. iLearn, Coursera etc.) or released directly by your Unit Convenor, are not confirmed as they are subject to final approval by the

University. Once approved, final results will be sent to your student email address and will be made available in [eStudent](#). For more information visit [ask.mq.edu.au](http://ask.mq.edu.au) or if you are a Global MBA student contact [globalmba.support@mq.edu.au](mailto:globalmba.support@mq.edu.au)

## Student Support

Macquarie University provides a range of support services for students. For details, visit <http://students.mq.edu.au/support/>

## Learning Skills

Learning Skills ([mq.edu.au/learningskills](http://mq.edu.au/learningskills)) provides academic writing resources and study strategies to help you improve your marks and take control of your study.

- [Getting help with your assignment](#)
- [Workshops](#)
- [StudyWise](#)
- [Academic Integrity Module](#)

The Library provides online and face to face support to help you find and use relevant information resources.

- [Subject and Research Guides](#)
- [Ask a Librarian](#)

## Student Services and Support

Students with a disability are encouraged to contact the [Disability Service](#) who can provide appropriate help with any issues that arise during their studies.

## Student Enquiries

For all student enquiries, visit Student Connect at [ask.mq.edu.au](http://ask.mq.edu.au)

If you are a Global MBA student contact [globalmba.support@mq.edu.au](mailto:globalmba.support@mq.edu.au)

## IT Help

For help with University computer systems and technology, visit [http://www.mq.edu.au/about\\_us/offices\\_and\\_units/information\\_technology/help/](http://www.mq.edu.au/about_us/offices_and_units/information_technology/help/).

When using the University's IT, you must adhere to the [Acceptable Use of IT Resources Policy](#). The policy applies to all who connect to the MQ network including students.

## Changes from Previous Offering

The topics and assessment are broadly similar to 2020 (which were changed significantly in 2019).