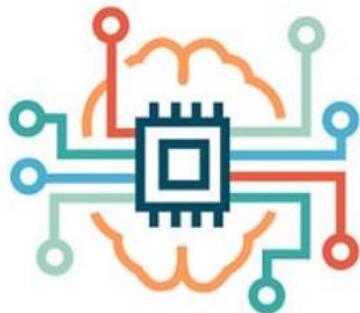


COMP1804
Applied Machine Learning



Lecture 1: Course Introduction





COMP1804

Learning Outcomes

- On completing this module successfully, you will be able to:
 - A. Rationalise appropriate scenarios for machine learning applications and evaluate the choice of machine learning methods for given application requirements.
 - B. Demonstrate competency in using appropriate libraries/toolkits to solve given real-world machine learning problems and develop and evaluate suitable application.
 - C. Understand and apply the relevant input data preparation and processing required for the machine learning models used, and quantitatively evaluate and qualitatively interpret the learning outcome.
 - D. Recognise and critically address the ethical, legal, social and professional issues that can arise when applying machine learning technologies.





Course Contacts

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- Use your Greenwich email account to email
otherwise your emails my be caught by our spam-filter
- Put COMP1804 in the subject line
- We will aim to reply within 48 hours during term-time.
If you don't get a reply by then, send again!





COMP1804

structure

- **Structure**
 - One hour weekly lecture: pre-recorded, 09:00-10:00 Fridays
 - Two hour weekly lab and Q&A: online, 10:00-12:00 Fridays
- **Course Website**

Moodle: COMP-1804-M01-2020-21 Applied Machine Learning





Essential Resources and Further Reading

- **Further reading (ebooks/guide):**

- Real World Machine Learning (eBook)
Brink, H., Richards, J., and Fetherolf, M. - www.allitebooks.in/real-world-machine-learning/
- Hands-On Machine Learning with Scikit-Learn and TensorFlow: Concepts, Tools, and Techniques to Build Intelligent Systems
Géron, A., O'Reilly Media

- **Useful resources:**

- [UCI Machine Learning Repository](https://archive.ics.uci.edu/ml/index.php)
- [UCI KDD Archive](https://kdd.ics.uci.edu/databases/kddcup99/kddcup99.html)
- [Statlib](https://lib.stat.cmu.edu/)
- [Delve](https://www.cs.toronto.edu/~delve/)
- [Kaggle](https://www.kaggle.com/) for datasets, project example and online competitions
- [SeedBank](https://ai.google/research/seeder/) (a collection of ML examples by Google)





General Resources: Journals

- IEEE Transactions on Pattern Analysis and Machine Intelligence (TPAMI)
 - International Journal of Computer Vision (IJCV)
 - Journal of Machine Learning Research
 - Machine Learning
 - Neural Computation
 - Neurocomputing
 - Neural Networks
 - IEEE Transactions on Neural Networks and Learning Systems (TNNLS)
 - Pattern Recognition
 - Annals of Statistics
 - Journal of the American Statistical Association
- etc.





General Resources: Conferences

- IEEE Conference on Computer Vision and Pattern Recognition (CVPR)
- Neural Information Processing Systems (NeurIPS)
- International Conference on Machine Learning (ICML)
- International Conference on Learning Representations (ICLR)
- Conference on Artificial Intelligence (AAAI)
- European Conference on Computer Vision (ECCV)
- International Conference on Computer Vision (ICCV)
- International Joint Conferences on Artificial Intelligence (IJCAI)
- European Conference on Machine Learning (ECML)
- Uncertainty in Artificial Intelligence (UAI)
- International Conference on AI & Statistics (AISTATS)
- Computational Learning Theory (COLT)
- International Conference on Artificial Neural Networks (ICANN)

etc.





COMP1804

assessment

- **Logbook (20%)**
 - individual work
 - four selected exercises throughout the course
 - all exercises should be completed and uploaded by 19/03/2021
(advice: complete each exercise after each lab or within a week)

- **Practical Coursework (80%)**
 - could be work in groups but students should upload individual assignments
 - involves practical development and report in IEEE-style paper
 - is due by 09/04/2021





Avoiding Plagiarism

- Discussing the course, helping each other with tutorials and coursework is great! (the goal should be to understand better, not copy work)

BUT!!!

- Copying from another student, submitting same coursework report, allowing someone to copy your work, or getting someone else to do your work
- Copying from Web, book etc. without clear and explicit acknowledgement
- Not acknowledging who you got help from/did help in your tutorial/coursework

IS PLAGIARISM

- We do check for it and penalties are severe :(
- Please don't put yourself **or** your friends at risk
- If in doubt - ask the staff





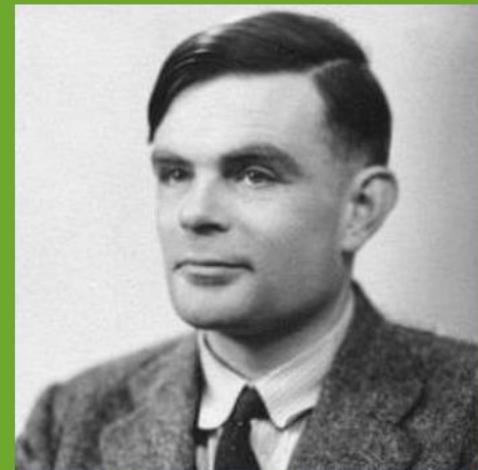
But...

What is machine learning?



“Instead of trying to produce a programme to reproduce the adult mind, why not rather try to produce one which simulates the child’s?”

Alan Turing, 1950





So... Machine Learning

The COOLEST TOPIC IN SCIENCE

- “A breakthrough in machine learning would be worth ten Microsofts” (Bill Gates, Chairman, Microsoft)
- “Machine learning is the next Internet” (Tony Tether, Director, DARPA)
- “Machine learning is the hot new thing” (John Hennessy, President, Stanford)
- “Web rankings today are mostly a matter of machine learning” (Prabhakar Raghavan, Dir. Research, Yahoo)
- “Machine learning is going to result in a real revolution” (Greg Papadopoulos, CTO, Sun)
- “Machine learning is today’s discontinuity” (Jerry Yang, CEO, Yahoo)





How would you teach someone to distinguish...
between a puppy and a muffin?





Puppy or Muffin?



Machine Learning is learning the rules
plus experience.



Some definitions of Machine Learning

- A definition (by Tom Mitchell):
“How can we build computer systems that automatically improve with experience, and what are the fundamental laws that govern all learning processes?”
- More technically:
“A computer program is said to learn from experience E with respect to some class of tasks T and performance measure P, if its performance at tasks in T, as measured by P, improves with experience E”





Why shall the Machine “Learn”? (I)

In general, the machine learns to address the case where the human has not found or is not able to find the law to explain the observed data, such as:

- the human is not able to explicitly encode his/her expertise in computer algorithms
→ speech/face recognition
- the human has no expertise
→ e.g., automatic navigation on Mars
- problems of which the solution needs to be adapted to particular cases
→ user biometrics

etc.





Why shall the Machine “Learn”? (II)

Machine Learning provides smart ways to analyse vast volumes of data. Nowadays there is an abundance of data to learn from, in almost all application domains.

By developing fast and efficient algorithms and data-driven models for real-time processing of data, Machine Learning can produce fast and accurate results and analysis.





What's happening in the industry & a timeline... (I)

- 1959 The term ‘machine learning’ was coined by computer scientist Arthur Samuel
- 1964 Eliza, the first chatbot, is developed at MIT
- 1997 IBM’s Deep Blue defeated Garry Kasparov, the world’s chess champion
- 2009 Chris Bishop: “Even the most sophisticated computers can’t tell a cat from a dog!”
- 2011 IMB’s Watson Q&A machine beat humans on Jeopardy! game;
Google Brain used Deep Neural Networks to identify and categorise objects;
Apple integrates Siri, a personal voice assistant, into the iPhone
- 2012 How many computers do you need to identify a cat? 1600
- 2014 Facebook developed DeepFace, a deep-learning facial recognition system;
Youtube recognizes cats from videos





What's happening in the industry & a timeline... (II)

- 2015 Microsoft & Google beat humans in image recognition
- 2016 Machine Learning algorithms outperform inexperienced radiologists;
Google's AlphaGo defeated a Go world champion
- 2018 Google AI can spot advanced breast cancer more effectively than humans.
It's up 99% accurate under the right conditions.





Today: Revolution & Technologies

- Moore's Law + GPUs
- Cloud computing
- More and richer data
- Github + Open source
- New and refined techniques





APIs and Open Source Tools

Google

Machine Learning as an API



Cloud
Vision API



Cloud
Speech API



Cloud
Natural Language
API



Cloud
Translation API



Cloud Video
Intelligence API

Use your own data to train models



TensorFlow



Cloud Machine
Learning Engine



APIs and Open Source Tools

Amazon



Recommendations

[AMAZON PERSONALIZE »](#)



Advanced Text Analytics

[AMAZON COMPREHEND »](#)



Forecasting

[AMAZON FORECAST »](#)



Image and Video Analysis

[AMAZON REKOGNITION »](#)



Document Analysis

[AMAZON Textract »](#)



Voice

[AMAZON POLLY »](#)





APIs and Open Source Tools

Microsoft



Machine Learning service
AI + MACHINE LEARNING



Translator Text API
AI + MACHINE LEARNING



Machine Learning Studio
ANALYTICS



Face API
AI + MACHINE LEARNING



Batch
COMPUTE

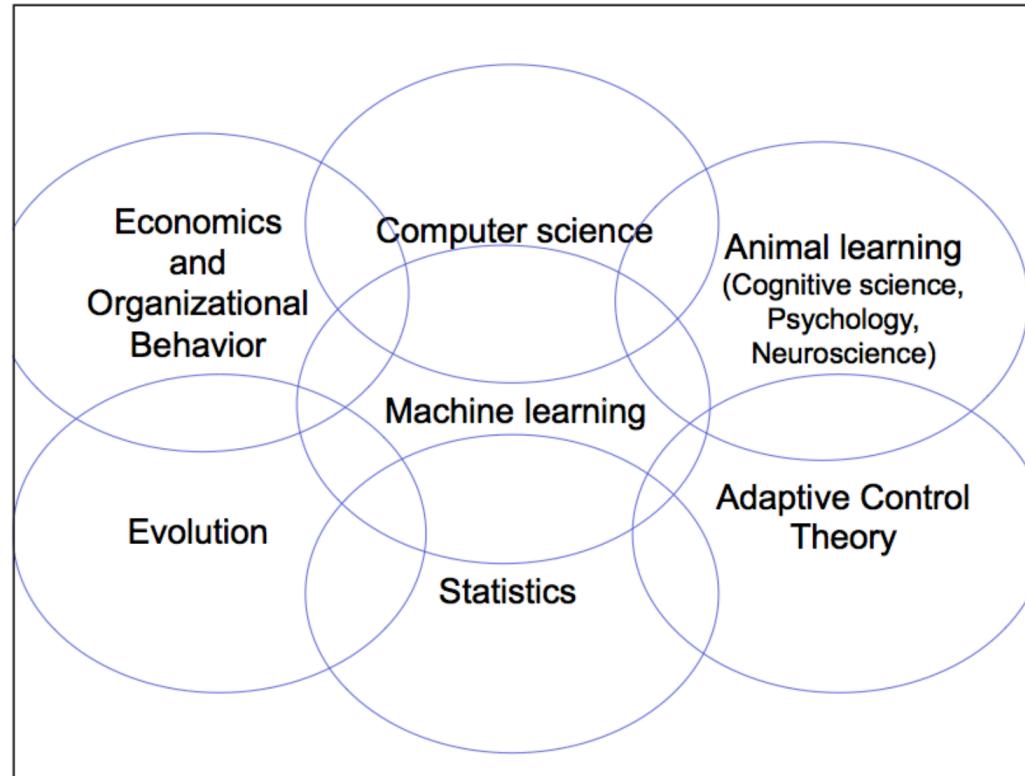


Data Factory
DATABASES





Many Communities relate to Machine Learning





Machine Learning is Everywhere

- Speech Recognition, Natural Language Processing
 - Computer Vision
 - Robotics
 - Medical Outcomes' Analysis
 - Computational Biology
 - Sensor Networks
- etc.

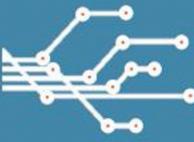




Case Study 1

Speech Recognition





Case Study 2

Machine Translation

Google Translate <https://translate.google.com/#auto/fr/According%20to%20Tom%20Mitchell%3A%20%22Machine%20Learning%20is%20the%20study%20of%20compute>

Apps LaTeX Cookbook Latex Math Symbols LaTeX/Theorems - ... 金山词霸 C++ Free Dictionary 法语助手 English to French, It... UPEM Journals Other bookmarks

Google Translate

English Spanish French English - detected ↗

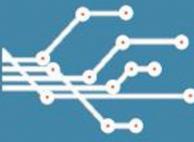
According to Tom Mitchell: "Machine Learning is the study of computer algorithms that improve automatically through experience."

Key question: how to design a computer which is able to learn automatically, from example data and/or past experience, so as to optimize a performance criterion?

Selon Tom Mitchell: "Machine Learning est l'étude des algorithmes informatiques qui améliorent automatiquement par l'expérience."

Question clé: comment concevoir un ordinateur qui est capable d'apprendre automatiquement, à partir de données d'exemple et / ou l'expérience passée, de manière à optimiser un critère de performance?





Case Study 3

Spam Detection



Dear Sir.

First, I must solicit your confidence in this transaction, this is by virtue of its nature as being utterly confidential and top secret. ...



TO BE REMOVED FROM FUTURE MAILINGS, SIMPLY REPLY TO THIS MESSAGE AND PUT "REMOVE" IN THE SUBJECT.

99 MILLION EMAIL ADDRESSES FOR ONLY \$99



Ok, I know this is blatantly OT but I'm beginning to go insane. Had an old Dell Dimension XPS sitting in the corner and decided to put it to use, I know it was working pre being stuck in the corner, but when I plugged it in, hit the power nothing happened.

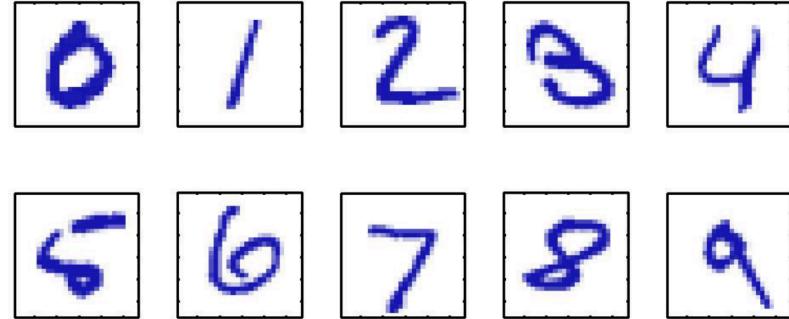




Case Study 4

Computer Vision: Optical Character Recognition (OCR)

Handwritten digit recognition: >99% accuracy



Applications: recognising addresses, checks, books, pen input, etc.

Main difficulty: wide variability of same number/character

7210414959
0690159734
9665407401
3134727121
1742351244





Case Study 5

Computer Vision: Face Recognition

Training examples of one person



Testing images



Main difficulty: large intra-class variations, due to various factors:

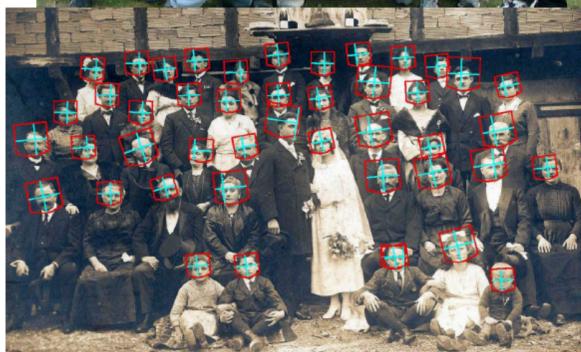
- viewpoint change
 - illumination change
 - hair-style change
 - age change
- etc.





Case Study 6

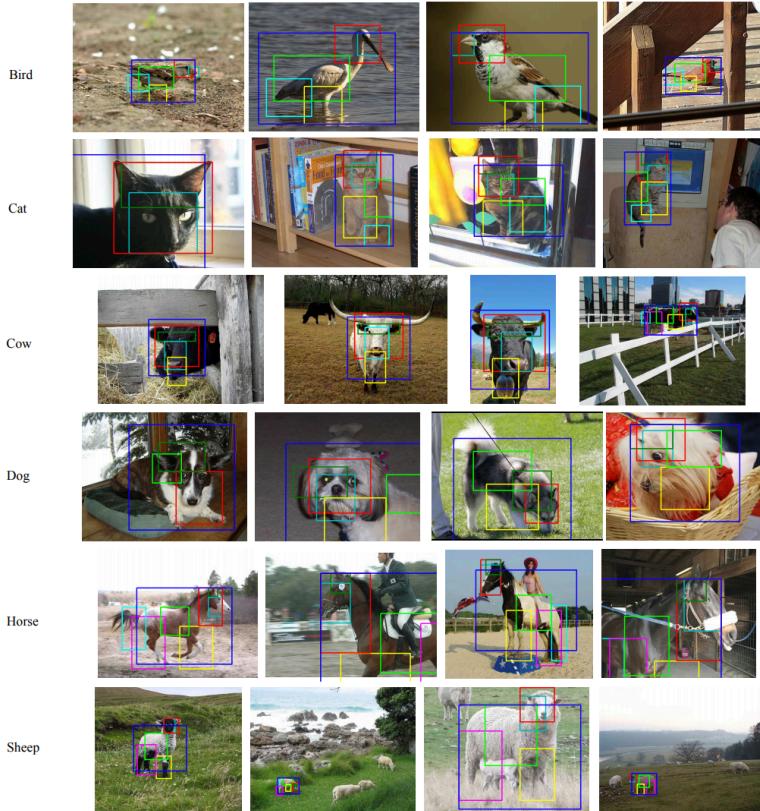
Computer Vision: Face Detection





Case Study 7

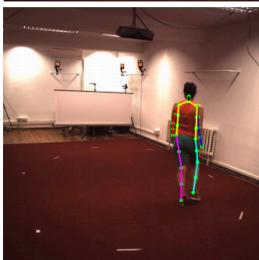
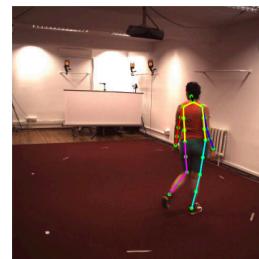
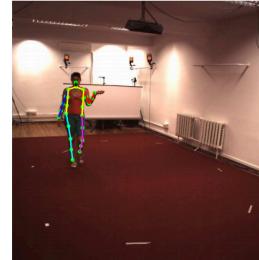
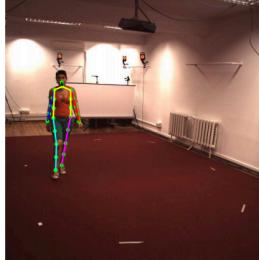
Computer Vision: Object Detection





Case Study 8

Computer Vision: Pose Estimation





Case Study 9

Computer Vision & Robotics

- Self Driving Cars: e.g., Google Prototype:



- Autonomous Helicopter Control: e.g., <http://heli.stanford.edu/>



- Navigation of Mars Lander, . . .





However there is still work to be done (I)

Face recognition



Not always perfect!





However there is still work to be done (II)



"A refrigerator filled with lots of food and drinks"



"A yellow school bus parked in a parking lot"





Back to the definition

“computer systems that automatically improve with experience”

"A computer program that ... improves its performance at tasks in T, as measured by performance measure P, improves with experience E"

- So WHAT is “experience”?

lines of text

Some lines of text, randomly arranged. Each letter, or symbol, has some kind of meaning. This is what we mean when we say “text”. This is what we mean when we say “experience”.

labeled images



games played





Representing data

- Machine learning algorithms (typically) only see numbers
- Typically, we create one vector representing each experience
 - Either with raw data values (pixel, characters) or preprocessed data (words, colours, shapes)
 - Vectors organised in a table

Lorem ipsum dolor sit amet, consetetur sadipscing etiam. Etiam non semper, sed ut labore et dolore magna aliqua. Ut enim ad minim veniam, quis nostrud exercitation ullamco laboris nisi ut aliquip ex ea commodo consequat. Duis aute irure dolor in reprehenderit in voluptate velit esse cillum dolore eu fugiat nulla pariatur. Excepteur sint occaecat cupidatat non proident, sunt in culpa qui officia deserunt mollit animus et ea modius. Sed ut perspiciatis unde omnis iste natus error sit voluptatem accusamus et iusto odio dignissimos sed quoque. Nisi ut aliquid eum neque. Quod autem vel sim illud, qui procul deinceps est, nonnullam etiam. Quod autem vel sim illud, qui procul deinceps est, nonnullam etiam.

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Machine Learning Problems terminology

- Data is often presented in tables
- Columns are called input variables or features or attributes.
- The columns we are trying to predict (outcome and time) are called output variables/targets.
- A row in the table is called a training example or instance.
- The whole table is called a data set.

tumor size	texture	perimeter	...	outcome	time
18.02	27.6	117.5		N	31
17.99	10.38	122.8		N	61
20.29	14.34	135.1		R	27
...					



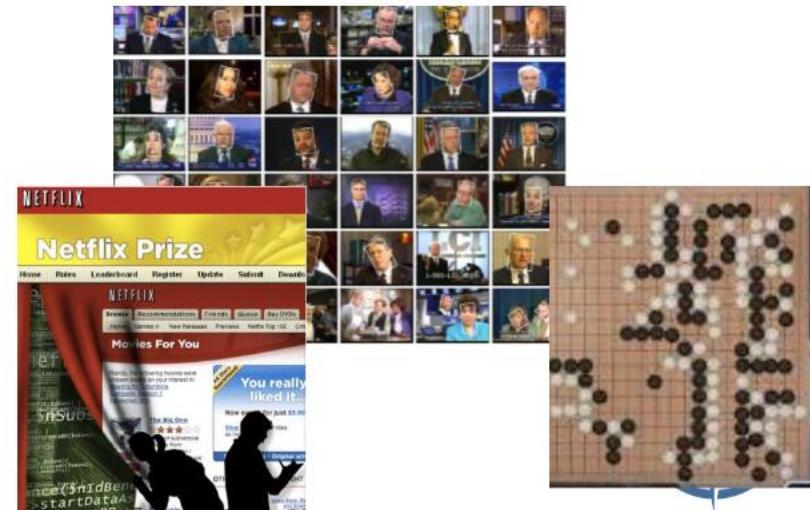


Back to the definition

“computer systems that automatically improve with experience”

"A computer program that ... improves its performance at **tasks** in T, as measured by performance measure P, improves with experience E"

- So **WHAT** is “**tasks**”?
- We’ve seen some examples
- Can we categorise them?





Main types of machine learning

- **Supervised learning**
 - Classification
 - Regression
- **Unsupervised learning**
- **Semi-Supervised learning**
- **Reinforcement Learning**





Thank you for the attention





Special Mention on Plagiarism

Please check and think what is the problem in the following slide





So... Machine Learning

The COOLEST TOPIC IN SCIENCE

- A breakthrough in machine learning would be worth ten Microsofts
- Machine learning is the next Internet
- Machine learning is the hot new thing
- Web rankings today are mostly a matter of machine learning
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