

COMP47460

Introduction to Machine Learning (Blended Delivery)

Aonghus Lawlor
Deepak Anjwani

School of Computer Science
Autumn 2019



Overview

- Module Outline
- Practical Details
- Machine Learning
 - Common Applications
 - Supervised v Unsupervised Learning
 - Representing Data as Features

Module Outline: Topics Covered

- Introduction and Fundamentals
- Supervised Learning
 - Classification: KNNs, Decision trees, Naive Bayes
 - Regression analysis
- Unsupervised Learning Algorithms
 - k-Means, Hierarchical clustering
- Working with Data
 - Dimensionality reduction, feature selection
- Evaluation and Methodologies
 - Statistical testing
 - Evaluating performance of ML systems
- Further Topics: Ensembles, Recommender systems

Practical Details

The course material will be made available on Moodle following the timetable of COMP47490: Tuesdays & Thursdays

There are 2 Tutorials/Practicals (one in Oct. and one in Nov), We have split them in two sessions each. You can attend either the first or second session of each tutorial/practical (they will be the same).

	Date	Location	SessionA	SessionB
Tutorial 1	Friday 18th Oct	ARD AA1	13:00-14:20	14:30-16:00
Practical 1	Friday 25th Oct	ARD AA1	13:00-14:20	14:30-16:00
Tutorial 2	Friday 8th Nov	ARD AA1	13:00-14:20	14:30-16:00
Practical 2	Friday 15th Nov	ARD AA1	13:00-14:20	14:30-16:00

Notes, assignments, and additional material will be available on the course page for COMP47460

Check that your surname, firstname and student ID number are all correct.

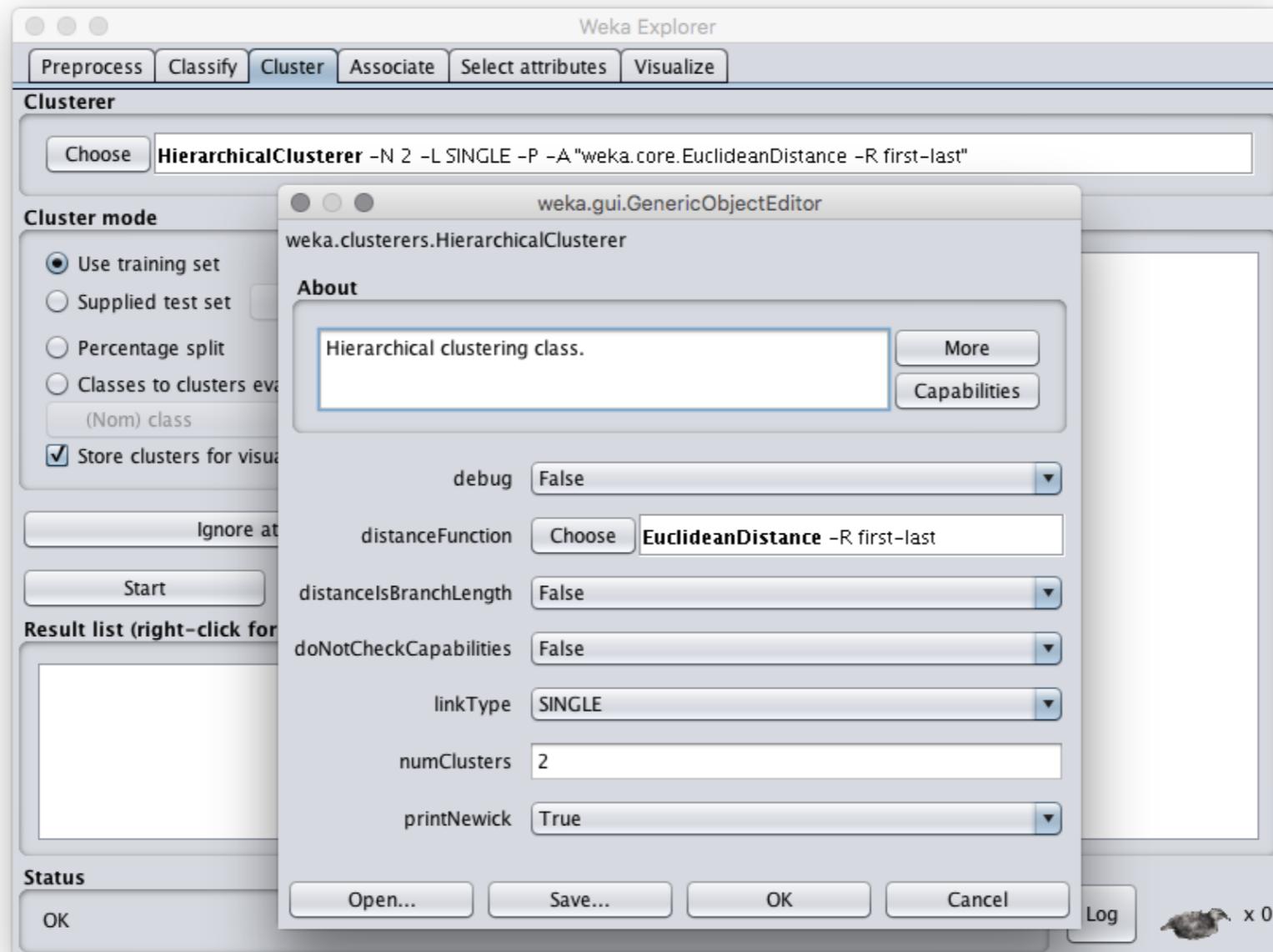
Dimension Reduction
PCA, LDA, LSI

-  10 - Dimension Reduction (Slides)
-  Dimension Reduction technical report
-  11 - Tutorial on Dimension Reduction & Feature Selection
-  Wine data (ARFF)
-  Diabetes data (ARFF)
-  Tutorial on Dimension Reduction and Feature Selection (Solutions)

For all module queries contact aonghus.lawlor@ucd.ie

Practical Details

Tutorials require laptop with Java Weka Toolkit - Version 3.8 Stable



<http://www.cs.waikato.ac.nz/ml/weka>

Practical Details

Module marks are based on assignments + final theory exam:

20%	Assignment 1: Weka + Report
20%	Assignment 2: Weka + Report
60%	End of Semester Theory Exam

Note:

- ! All assignment deadlines are hard deadlines.
1-5 days late: 10% deduction from overall mark
6-10 days late: 20% deduction from overall mark
Not accepted after 10 without extenuating circumstances
- ! Plagiarism will be treated seriously. Any evidence of plagiarism in an assignment or exam will result in a 0 mark.

Practical Details

CS grading scheme applies for this module. Pass mark is 40%.

Grade	Min	Max
A+	95	100
A	90	95
A-	85	90
B+	80	85
B	75	80
B-	70	75
C+	65	70
C	60	65
C-	55	60
D+	50	55
D	45	50
D-	40	45

Grade	Min	Max
E+	35	40
E	30	35
E-	25	30
F+	20	25
F	15	20
F-	10	15
G+	8	10
G	5	8
G-	2	5
NG	0	0

Plagiarism and UCD Computer Science

- Plagiarism is a serious academic offence
- [Student Code, section 6.2] or [UCD Registry Plagiarism Policy] or [CS Plagiarism policy and procedures]
- Our staff and demonstrators are proactive in looking for possible plagiarism in all submitted work
- Suspected plagiarism is reported to the CS Plagiarism subcommittee for investigation
 - Usually includes an interview with student(s) involved
 - 1st offence: usually 0 or NG in the affected components
 - 2nd offence: referred to the University disciplinary committee
 - Student who enables plagiarism is equally responsible

http://www.ucd.ie/registry/academicsecretariat/docs/plagiarism_po.pdf

http://www.ucd.ie/registry/academicsecretariat/docs/student_code.pdf

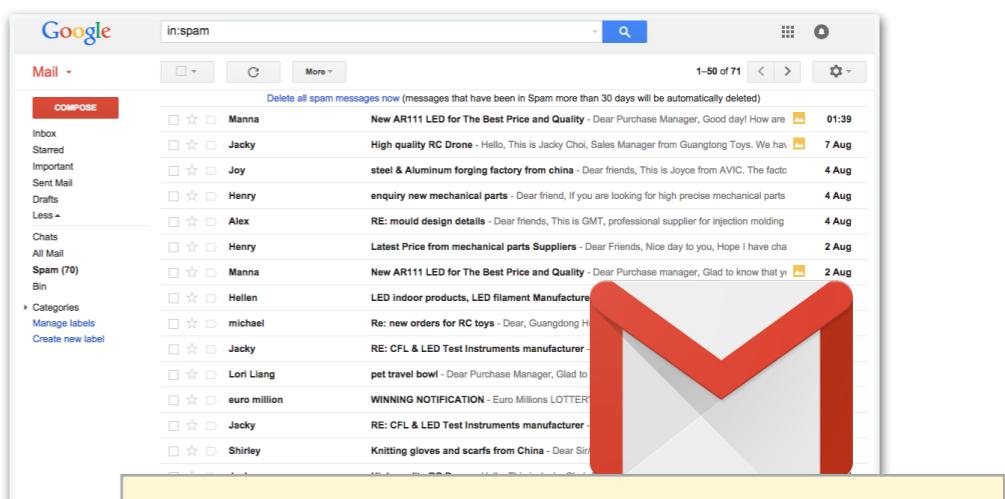
<http://libguides.ucd.ie/academicintegrity>

Why Study Machine Learning?

- Explosion in rich, complex data to analyse - online and offline.
- Significant recent progress in algorithms and theory.
- Computational power is now available.
- Industry demand - Data scientists, Data engineers...
- New applications in many disciplines - Medicine, engineering, humanities...



A screenshot of the Twitter profile for 'Twitter Engineering' (@TwitterEng). The profile features a blue gear icon with a white bird logo. It shows 291 tweets, 1 follower, 861K followers, 8 favourites, and 1 list. The bio reads: 'The official account for Twitter Engineering.' It lists San Francisco, CA as the location and provides a link to engineering.twitter.com. The profile has joined June 2007. Below the bio, there are 11 photos and videos. A yellow callout box contains the following statistics:
316 million monthly active users
500 million tweets per day
5 billion user sessions per day



A screenshot of a Google Mail inbox search results for 'in:spam'. The search results show 71 items, with the first few being from 'Manna', 'Jacky', 'Joy', 'Henry', 'Alex', 'Henry', 'Manna', 'Helen', 'michael', 'Jacky', 'Lori Liang', 'euro million', 'Jacky', and 'Shirley'. A large red Gmail logo is overlaid on the right side of the inbox. A yellow callout box contains the following statistics:
900 million users
Handles 2+ trillion mails per year

Application: Web Search

Submit a query to a search engine, it finds pages relevant to the query, and returns them ranked by relevance.

The screenshot shows a Google search results page for the query "machine learning". The search bar at the top contains the query. Below the search bar, the "All" tab is selected, along with other categories like News, Images, Videos, Books, More, and Search tools. A message indicates there are about 22,800,000 results found in 0.48 seconds. The results are listed in descending order of relevance:

- Machine learning - Wikipedia, the free encyclopedia**
https://en.wikipedia.org/wiki/Machine_learning ▾
Machine learning is a subfield of computer science (more particularly soft computing) that evolved from the study of pattern recognition and computational learning theory in artificial intelligence.
List of machine learning ... · Supervised learning · Computational learning theory
- Machine Learning - Stanford University | Coursera**
<https://www.coursera.org/learn/machine-learning> ▾
About this course: Machine learning is the science of getting computers to act without being explicitly programmed. In the past decade, machine learning has given us self-driving cars, practical speech recognition, effective web search, and a vastly improved understanding of the human genome.
- Machine Learning: What it is and why it matters | SAS**
www.sas.com/en_id/insights/analytics/machine-learning.html ▾
Machine learning is a method of data analysis that automates analytical model building. Using algorithms that iteratively learn from data, machine learning ...
- What is machine learning? - Definition from WhatIs.com**
[whatis.techtarget.com › Topics › Application Development › Programming](http://whatis.techtarget.com/Topics/Application%20Development/Programming) ▾
Machine learning is a type of artificial intelligence (AI) that provides computers with the ability to learn without being explicitly programmed. Machine learning ...
- Intro to Machine Learning Course | Udacity**
<https://www.udacity.com/course/intro-to-machine-learning--ud120> ▾
Intro to Machine Learning explores pattern recognition during data analysis through computer science and statistics using the popular Python language.
- Machine Learning - OpenClassroom - Stanford University**
openclassroom.stanford.edu/MainFolder/CoursePage.php?course=MachineLearning ▾
Course Description. In this course, you'll learn about some of the most widely used and successful machine learning techniques. You'll have the opportunity to ...

Application: Movie Recommendation

Netflix provides personalised recommendations for movies you might like, based on previous user ratings.

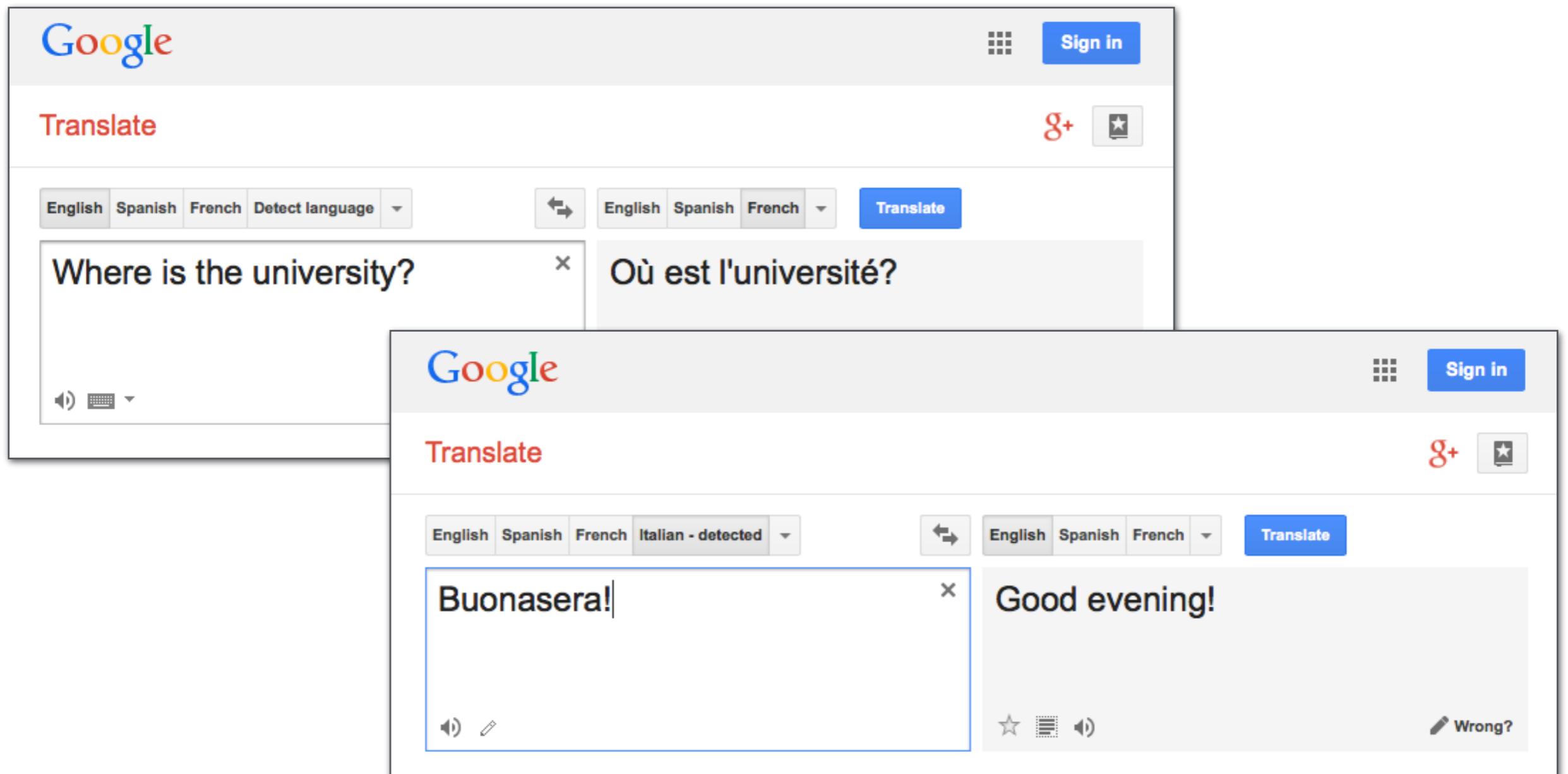
The screenshot shows the Netflix homepage with a red header. The top navigation bar includes links for 'Watch Instantly', 'Browse DVDs', 'Your Queue', and 'Movies You'll ❤️'. A search bar is located at the top right. Below the header, a large banner displays the message 'Congratulations! Movies we think You will ❤️' and encourages users to 'Add movies to your Queue, or Rate ones you've seen for even better suggestions.'

The main content area is divided into several sections:

- Suggestions to Watch Instantly:** This section lists recommended TV shows and movies. Examples include 'INSPECTOR LEWIS', 'Masterpiece Mystery!: Inspector Lewis', 'DROP DEAD Diva', 'Drop Dead Diva', and 'That's What I Am'. Each item has a 'Play' button, a star rating, and a 'Not Interested' link.
- Action & Adventure:** This section lists recommended movies in the Action & Adventure genre. Examples include 'Unstoppable', 'Star Wars: Episode III - Revenge of the Sith', 'LOTR: Fellowship of the Ring: Extended Ed.', and 'Man on Fire'. Each item has an 'Add' button, a star rating, and a 'Not Interested' link.
- Other Sections:** There are also sections for 'Suggestions to Watch Instantly' (with arrows to see more) and 'Action & Adventure' (with arrows to see more).

Application: Machine Translation

Use examples of translated documents to learn how to translate between the two languages.



Application: Entity Recognition

Automatically extracted named entities (e.g. people, places, organisations) from text documents (e.g. news articles).

SPORT FOOTBALL

Home Football Formula 1 Cricket Rugby U Tennis Golf Athletics Cycling All Sport

Arsenal > Results | Fixtures | Table | Live Scores | All Teams | Leagues & Cups

9 August 2015
Last updated at 17:13
GMT
3.1K Share f t

Arsene Wenger: Arsenal boss wants response after defeat

Arsenal will recover from the "nerves" they suffered in their opening-day Premier League defeat by West Ham, their manager Arsene Wenger says.

The Gunners lost 2-0 at the Emirates Stadium on Sunday, despite having 62% of possession and 22 shots.

"We will respond to that accident," Wenger said.

"The players were maybe too nervous and put too much pressure on themselves. Today we have been hurt mentally and it is a good opportunity to respond."

He added: "We were not convincing offensively or defensively. I knew it could be a tricky game. If you can't win the game, make sure you don't lose it."

Wenger whose side finished third last season and won the FA Cup, has been under pressure to add players to his squad, with keeper Petr Cech - a £10m signing from Chelsea - the only summer arrival at the Emirates.

Real Madrid striker Karim Benzema is a reported target for Arsenal but Wenger said new additions would not have aided a performance where his side managed just six shots on target.

Tag colours:

Person

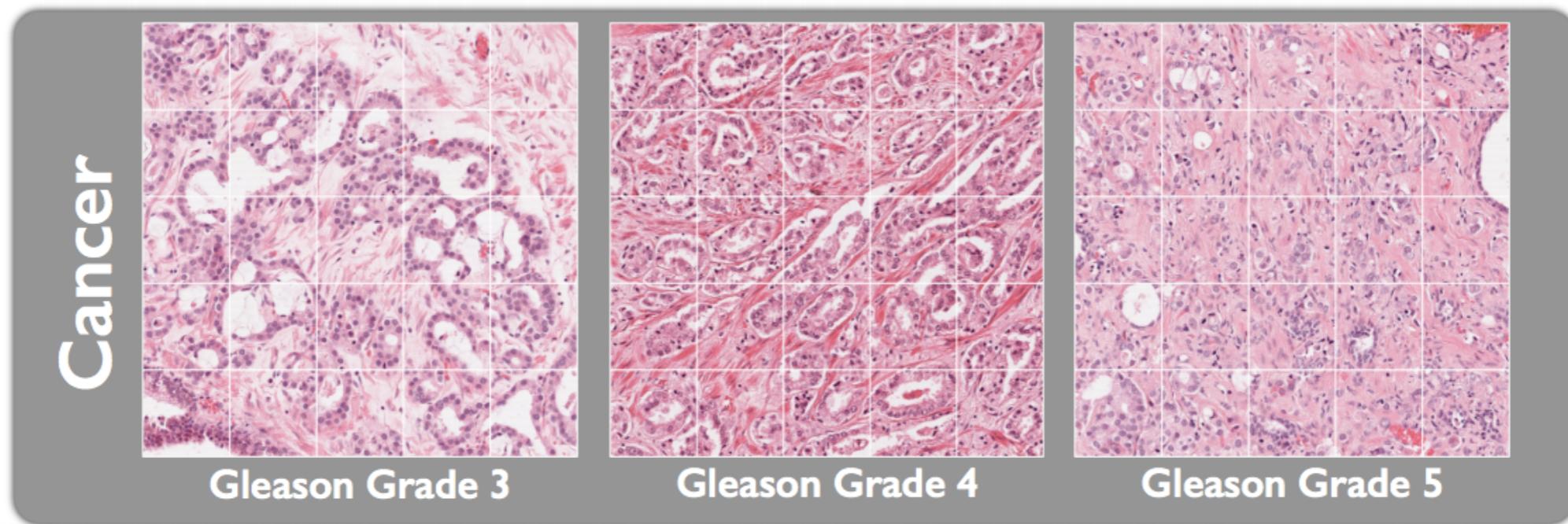
Location

Organisation

Applications in Medicine

Machine Learning provides tools and support solving diagnostic and prognostic tasks in a variety of medical domains, including...

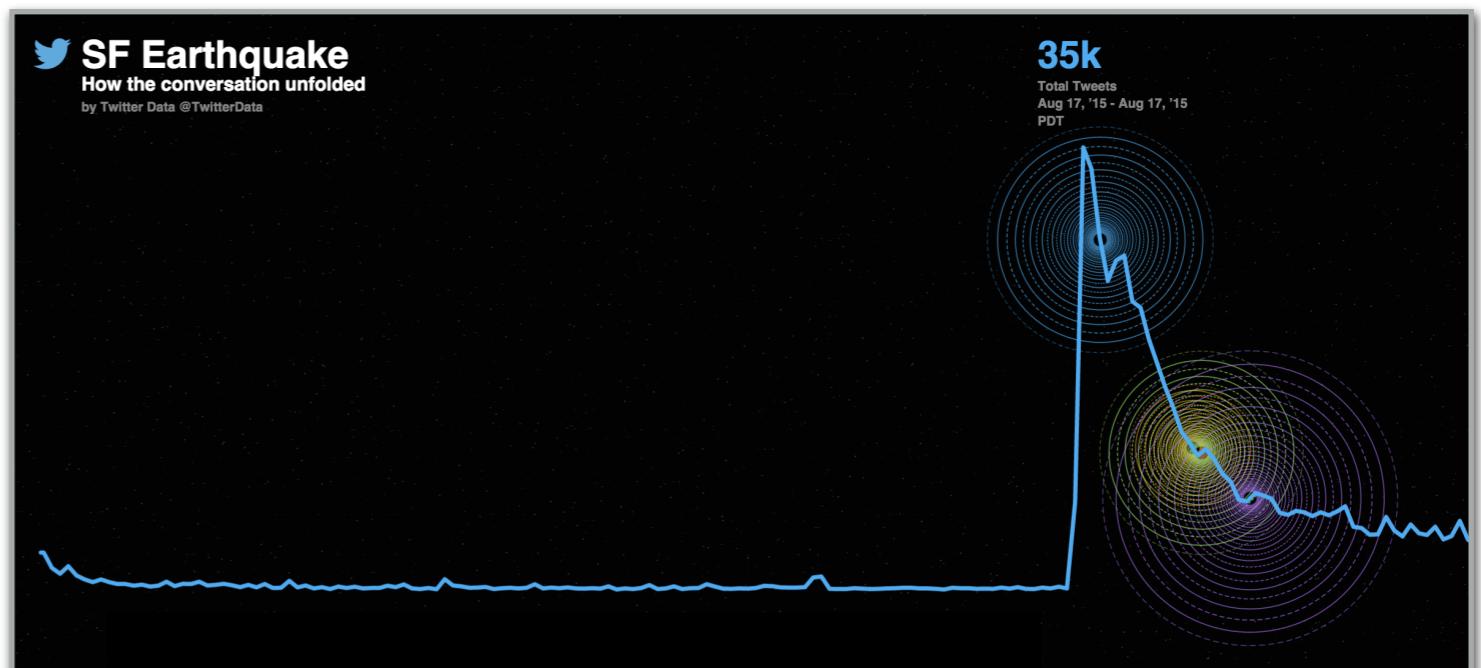
- Disease diagnosis based on previous correct cases
- Prediction of disease progression
- Medical image analysis and understanding
- Hospital information systems



Application: Anomaly Detection

Algorithms to find patterns in data that do not conform to a model of “normal” behaviour in a system. In some systems, these are rare events. In other systems, these are unexpected bursts of activity.

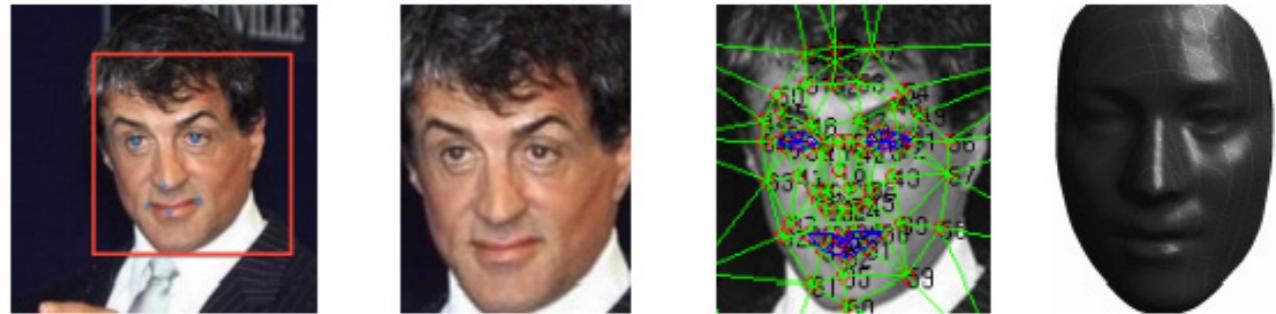
- *Cybersecurity*: Spike in number of false login attempts.
- *Payment systems*: Unusual number of failed or incomplete payments made.
- *Fraud detection*: Unexpected transactions in financial networks.
- *Event detection*:
Sudden spike in volume of social media posts.



Application: Face Recognition

Facebook tags photos by comparing them to profile pictures.

“We currently use facial recognition software that uses an algorithm to calculate a unique template based on someone’s facial features, like the distance between the eyes, nose and ears. This template is based on your profile pictures and photos you’ve been tagged in on Facebook.”



In 2013, Facebook revealed its users have uploaded > 250 billion photos, and are uploading 350 million new photos each day.

Using other cues (e.g. hair style, clothing), allows Facebook to accurately identify people, even when their face is obscured.

<http://www.fastcolabs.com/3028414/how-facebooks-machines-got-so-good-at-recognizing-your-face>

<http://arstechnica.com/?p=695873>

Application: Autonomous Vehicles

- Car manufacturers and researchers are exploring the potential of self-driving cars. Involves analysis of large volumes of sensor data, categorised using ML approaches combined with human labelling.
- 2004: Autonomous cars tried to navigate a 150 mile desert DARPA race. None of the 21 teams finished.
- 2015: Google driverless test vehicles have driven nearly 1 million miles, with no accidents caused by a self-driving car. Prototypes launched on public roads.



<http://googleblog.blogspot.ie/2015/05/self-driving-vehicle-prototypes-on-road.html>

<http://www.nature.com/news/autonomous-vehicles-no-drivers-required-1.16832>

Application: Spam Classification

Apply a learning algorithm to automatically classify incoming emails into *spam* or *non-spam*, based on previous examples of legitimate and spam email.

The screenshot shows a Google Mail interface with a search bar at the top containing "in:spam". The search results are displayed below, showing 1-50 of 71 messages. The messages are listed in descending order of date, with the most recent message at the top. Each message includes the sender's name, subject, and a timestamp. The messages are all from spammers, as indicated by the subject lines which are typical of spam emails.

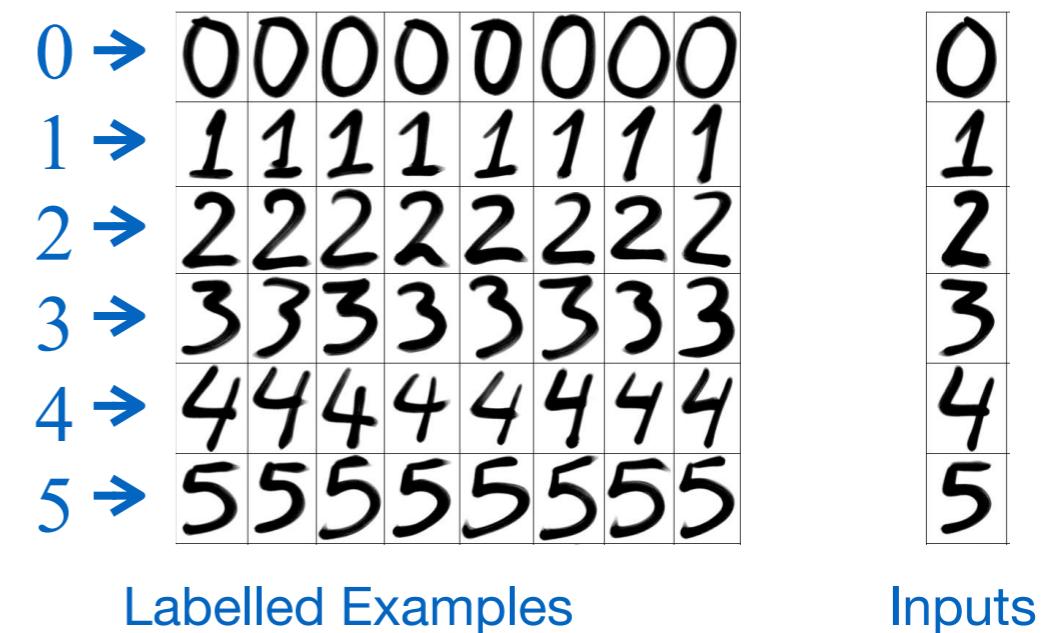
Sender	Subject	Date
Manna	New AR111 LED for The Best Price and Quality - Dear Purchase Manager, Good day! How are	01:39
Jacky	High quality RC Drone - Hello, This is Jacky Choi, Sales Manager from Guangtong Toys. We have	7 Aug
Joy	steel & Aluminum forging factory from china - Dear friends, This is Joyce from AVIC. The factory	4 Aug
Henry	enquiry new mechanical parts - Dear friend, If you are looking for high precise mechanical parts	4 Aug
Alex	RE: mould design details - Dear friends, This is GMT, professional supplier for injection molding	4 Aug
Henry	Latest Price from mechanical parts Suppliers - Dear Friends, Nice day to you, Hope I have cha	2 Aug
Manna	New AR111 LED for The Best Price and Quality - Dear Purchase manager, Glad to know that you	2 Aug
Hellen	LED indoor products, LED filament Manufacturer direct quotation - Dear Sir, How are you? I am	1 Aug
michael	Re: new orders for RC toys - Dear, Guangdong Huanqi Electronic Co., Ltd. which is specialized i	31 Jul
Jacky	RE: CFL & LED Test Instruments manufacturer - Hello, Lisun Group is the leader in CFL & LED	30 Jul
Lori Liang	pet travel bowl - Dear Purchase Manager, Glad to learn you are in the market for pet travel bowl.	30 Jul
euro million	WINNING NOTIFICATION - Euro Millions LOTTERY PROMOTION MADRID OFFICE WINNING N	30 Jul
Jacky	RE: CFL & LED Test Instruments manufacturer - Hello, Lisun Group is the leader in CFL & LED	28 Jul
Shirley	Knitting gloves and scarfs from China - Dear Sir/Madam, We are making Knitting gloves, bands	28 Jul
Jacky	High quality RC Drone - Hello, This is Jacky Choi, Sales Manager from Guangtong Toys. We have	28 Jul
Ivy	RE: Bearings manufacture - Hello, We are specialized in bearings more than 12 years. Our prod	27 Jul

Supervised v Unsupervised Learning

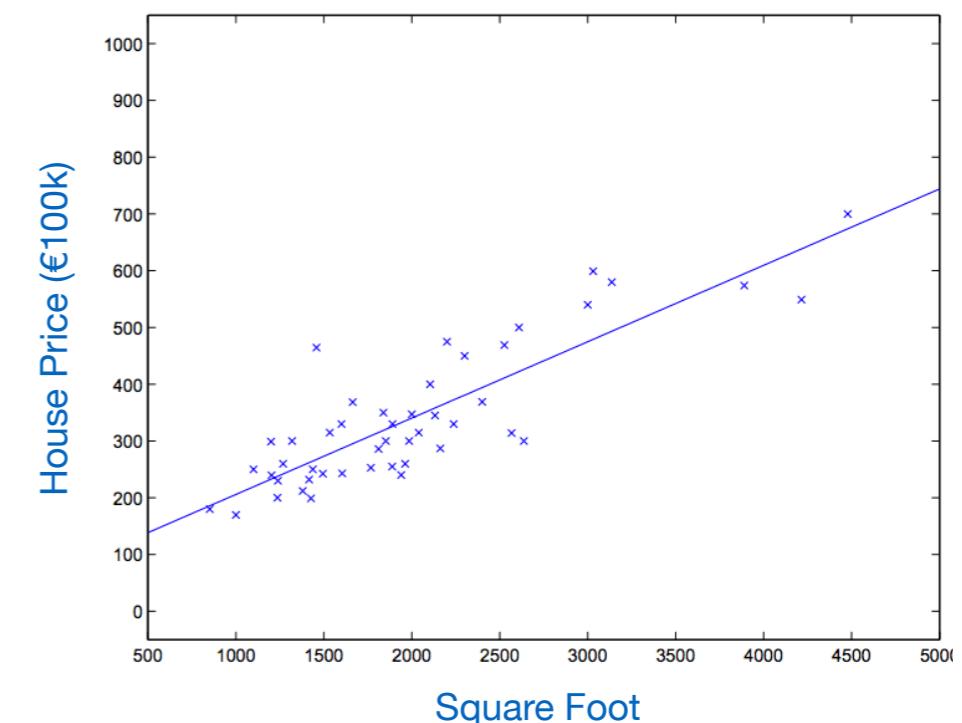
- **Supervised Learning:**
An algorithm that learns a function from examples of its inputs and outputs. It requires manually-labelled example data to learn the correct answer for a given query input.
 - e.g. Classification, Regression algorithms
- **Unsupervised Learning:**
An algorithm that finds patterns in data when no manually labelled examples are available as inputs. More focused on data exploration and knowledge discovery.
 - e.g. Clustering, Graph partitioning algorithms

Supervised Learning

- **Classification:**
Examples represented by a set of features, which help decide the *class* to which a new query input belongs (i.e. output is a label)



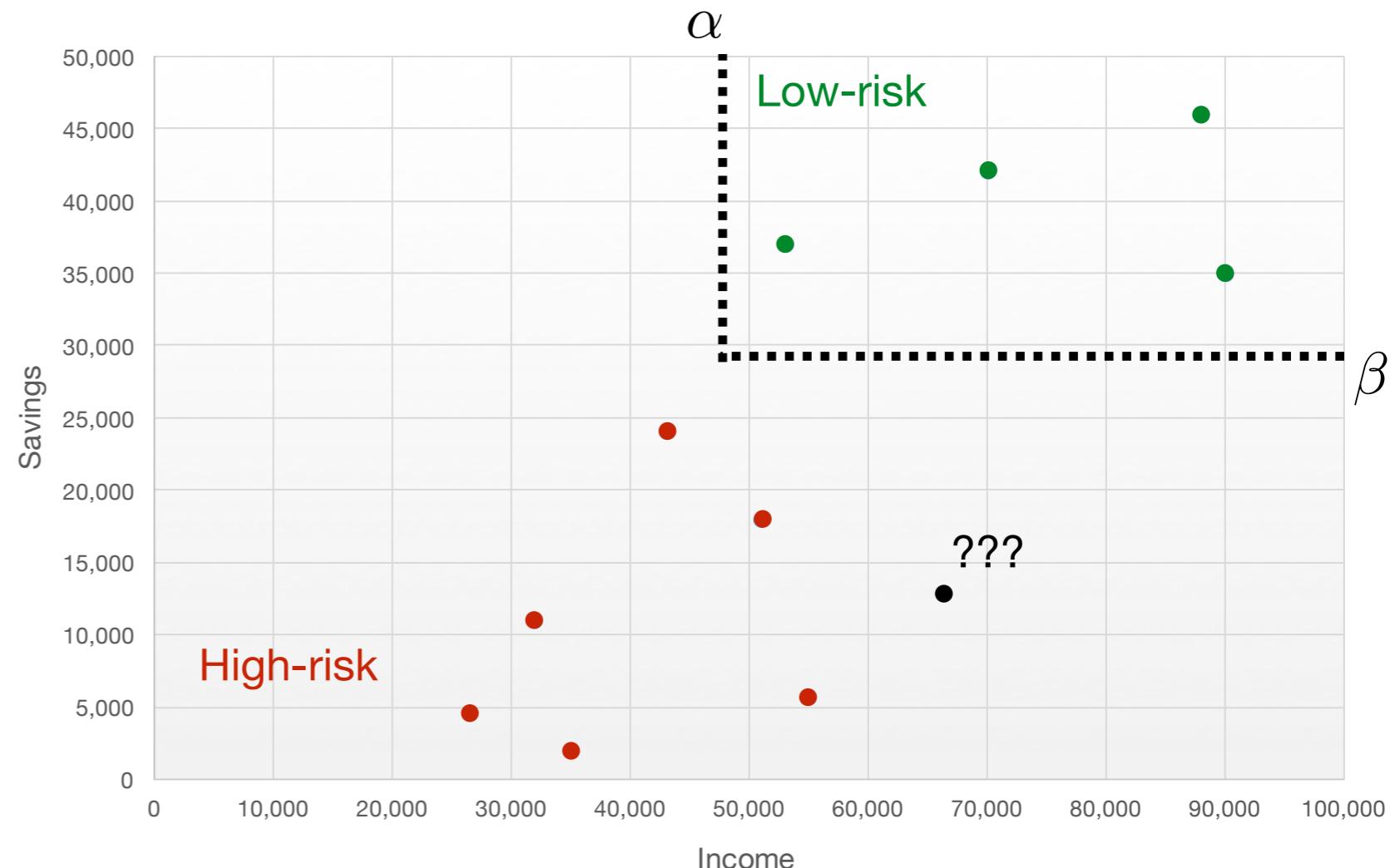
- **Regression:**
Examples characterised by a set of features, which help decide the value of a continuous output variable (i.e. output is a number)



Typical Classification Task

Example: Credit Scoring

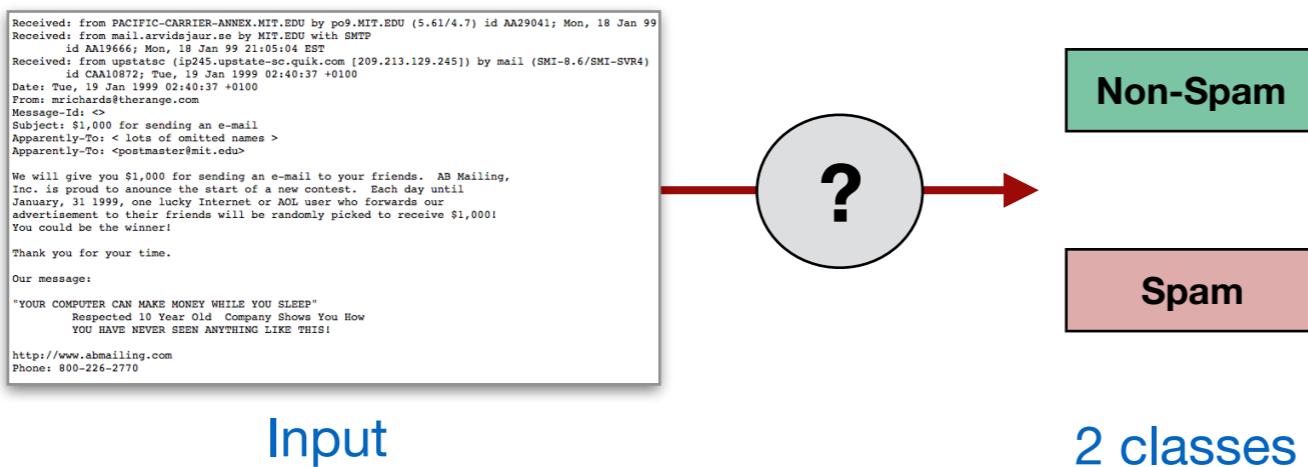
Manually classify customers into two categories (**low-risk** and **high-risk**) based on savings and income data.



- Q. Can we train an algorithm to learn to automatically classify new customers as either **low-risk** or **high-risk**?
i.e. can we learn α and β ?

Classification Tasks

- **Binary Classification:**
 - Assign an input to one of two possible target class labels.



- **Multiclass Classification:**
 - Assign an input to one of M different target class labels.



Classification Tasks

- **Evaluation:** Standard approach for classification tasks is to split the set of examples into a *training set* and the *test set*.
- **Training set:** Examples provided to the classifier to build a model of the data. Each example has been manually assigned a class label.
- **Test set:** Examples held back from the classifier, which are used to evaluate the accuracy of the classifier. Test examples are completely separate from the training set.
- Why not just train on all the data?
 - The test set is used to evaluate how well the model built by the classifier will generalise to new input examples.
 - Using the training data will give us over-optimistic results!

Classification Algorithms

- Many different learning algorithms exist for classification (e.g. k-nearest neighbour, decision tree, neural network, support vector machine).
- Problem dimensions will often determine which classification algorithm will be practically applicable, due to processing, memory, and storage constraints.
 1. Number of input examples N .
 - Sometimes millions of input examples.
 2. Number of feature (dimensions) D representing each input example.
 - Often 10-1000, but sometimes far higher.
 3. Number of target classes M .
 - Often small (binary), but sometimes far higher.

Representing Data

- Examples are represented by one or more features, which can be distinguished by the type and number of values they can take.
 - **Binary:** Takes only two values - a boolean True/False decision
e.g. married={True,False}, test_result={Pass,Fail}
 - **Categorical (Nominal):** A feature that takes values from two or more categories, with no intrinsic ordering to the categories.
e.g. blood_group={A,B,AB,O}, nationality={French,Irish,Italian}
 - **Ordinal:** Similar to a categorical variable, but there is a clear ordering of the variables.
e.g. grade={A,B,C,D,E,F}, dosage={Low,Medium,High}
 - **Continuous:** Numeric measurements, with or without a fixed range for the values.
e.g. temperature, weight, height, latitude, longitude etc.

Typical Classification Task

- Training set with $N=10$ examples (customers). Each is described by $D=5$ features: 3 continuous, 2 categorical
- Each example has one of two class labels = {High-risk, Low-risk}

Example	Income	Savings	Married	Gender	Age	Class
1	35,000	2,000	Y	M	32	High-risk
2	51,000	18,000	N	M	34	High-risk
3	70,000	42,000	Y	F	41	Low-risk
4	26,500	4,500	N	M	22	High-risk
5	32,000	11,000	N	F	25	High-risk
6	53,000	37,000	N	F	39	Low-risk
7	88,000	46,000	Y	M	48	Low-risk
8	55,000	5,700	N	M	55	High-risk
9	90,000	35,000	Y	F	61	Low-risk
10	43,000	24,000	Y	M	33	High-risk

Q. To which class does this new customer belong?

Example	Income	Savings	Married	Gender	Age	Class
X	66,000	13,000	Y	M	44	???

Typical Classification Task

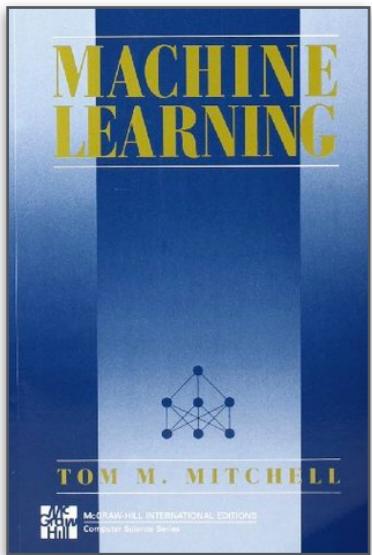
- Training set with $N=10$ examples (fruit). Each is described by $D=4$ features: 3 continuous, 1 categorical
- Each has one of three class labels = {Apple,Pear,Orange}

Example	Height	Width	Taste	Weight	Class
1	60	62	Sweet	186	Apple
2	70	53	Sweet	180	Pear
3	55	50	Tart	152	Apple
4	76	40	Sweet	152	Pear
5	68	71	Tart	207	Orange
6	65	68	Sour	221	Apple
7	63	45	Sweet	140	Pear
8	55	56	Sweet	154	Apple
9	76	78	Tart	211	Orange
10	60	58	Sour	175	Apple

Q. To which class does this new fruit belong?

Example	Height	Width	Taste	Weight	Class
X	63	68	Sweet	168	???

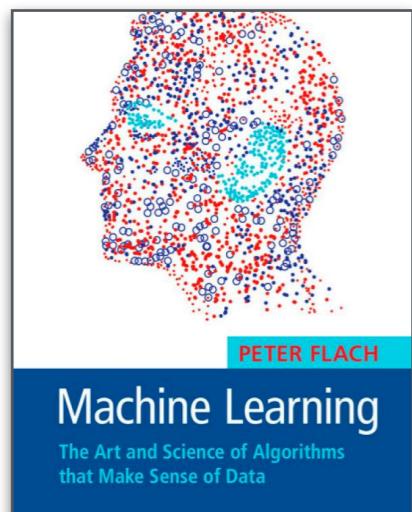
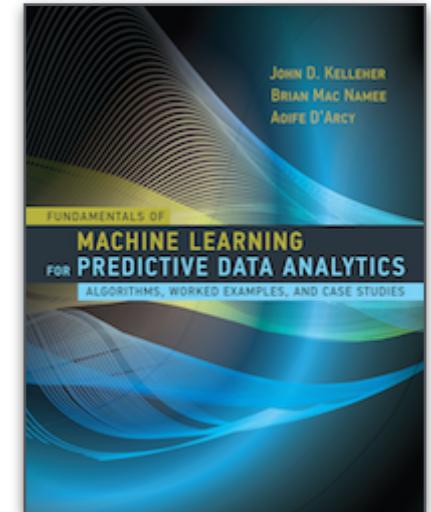
Additional Reading



Machine Learning
McGraw-Hill 1997
Tom M. Mitchell

*Fundamentals of Machine
Learning for Predictive Data
Analytics*

John D. Kelleher, Brian Mac
Namee, Aoife D'Arcy



*Machine Learning: The Art and Science
of Algorithms that Make Sense of Data*
Peter Flach

*Data Mining: Practical Machine
Learning Tools and Techniques, 3rd Ed*
Ian H. Witten, Eibe Frank, Mark A. Hall

