



University  
of Exeter

# COM1011

## Fundamentals of Machine Learning

# Introduction

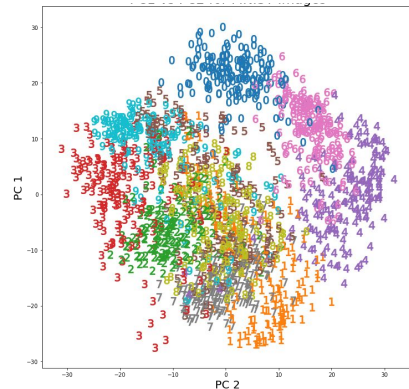
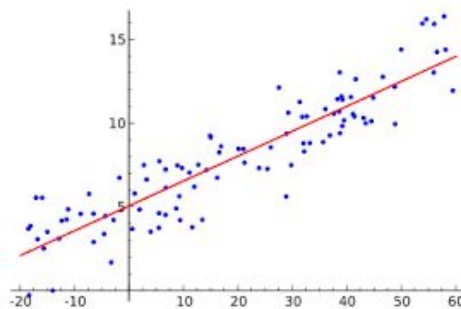
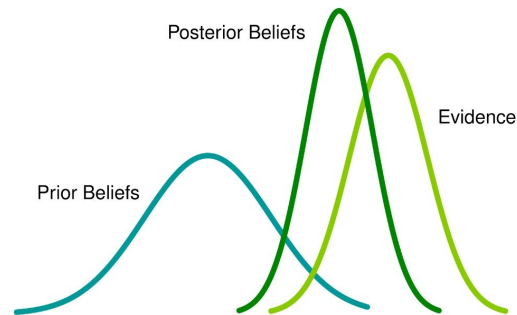
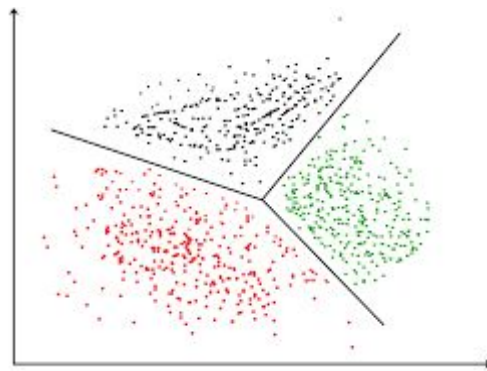
- What this course is about
- What Machine Learning is
- Types of data

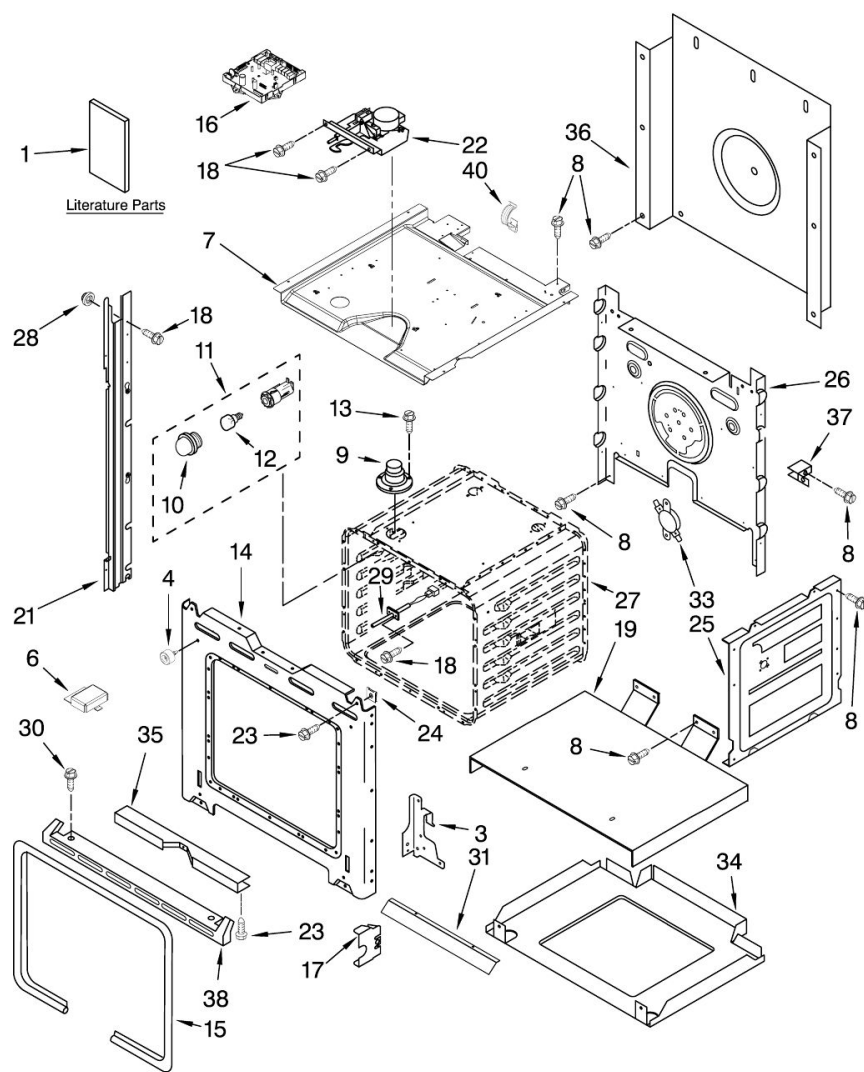
# What this course is about

- Core concepts of Machine Learning (ML), Artificial Intelligence (AI) and Data Science (DS)
- How to make machines actually learn from data
- Strengths and weaknesses of different methods
- All at a conceptual level... mostly

# What this course is about

- Regression
- Classification
- Clustering
- Model selection
- Bayesian statistics
- Dimensionality reduction





# Algorithm

A step-by-step procedure,  
specified to a level of detail  
such that any person  
(or machine) can follow.

# This is basically an algorithm

- 1 Bring a large pan of water to the boil. Heat a grill pan. Toss the courgette strips with the oil and grill them for 2-3 minutes, turning often. Meanwhile, add the pine nuts to a small saucepan over a medium heat and cook for 2-3 minutes until toasted. Transfer to a plate immediately.
- 2 Cook the lasagne sheets for 3-4 minutes in the boiling water, then drain well. Carefully cut each lasagne sheet in half width ways so you have 12 pieces. Meanwhile, cook the spinach in 1 tsp water for 2-3 minutes until the leaves have wilted. Drain well, squeezing out the excess moisture with the back of a spoon.
- 3 Toss the cooked courgette strips in the pesto. Layer them with the lasagne sheets, spinach and dollops of the ricotta. Garnish with the basil and pine nuts.

# This is basically an algorithm

1

Bring a large pan of water to a boil. Grill the courgette strips for 2-3 minutes over a medium heat and

2

Cook the lasagne sheets in boiling water. Drain each lasagne sheet in hot water. Add 1 tsp water for 2-3 minutes. Remove the moisture with the back of a spoon.

3

Toss the cooked courgette strips in the pesto. Layer them with the lasagne sheets, spinach and dollops of the ricotta. Garnish with the basil and pine nuts.



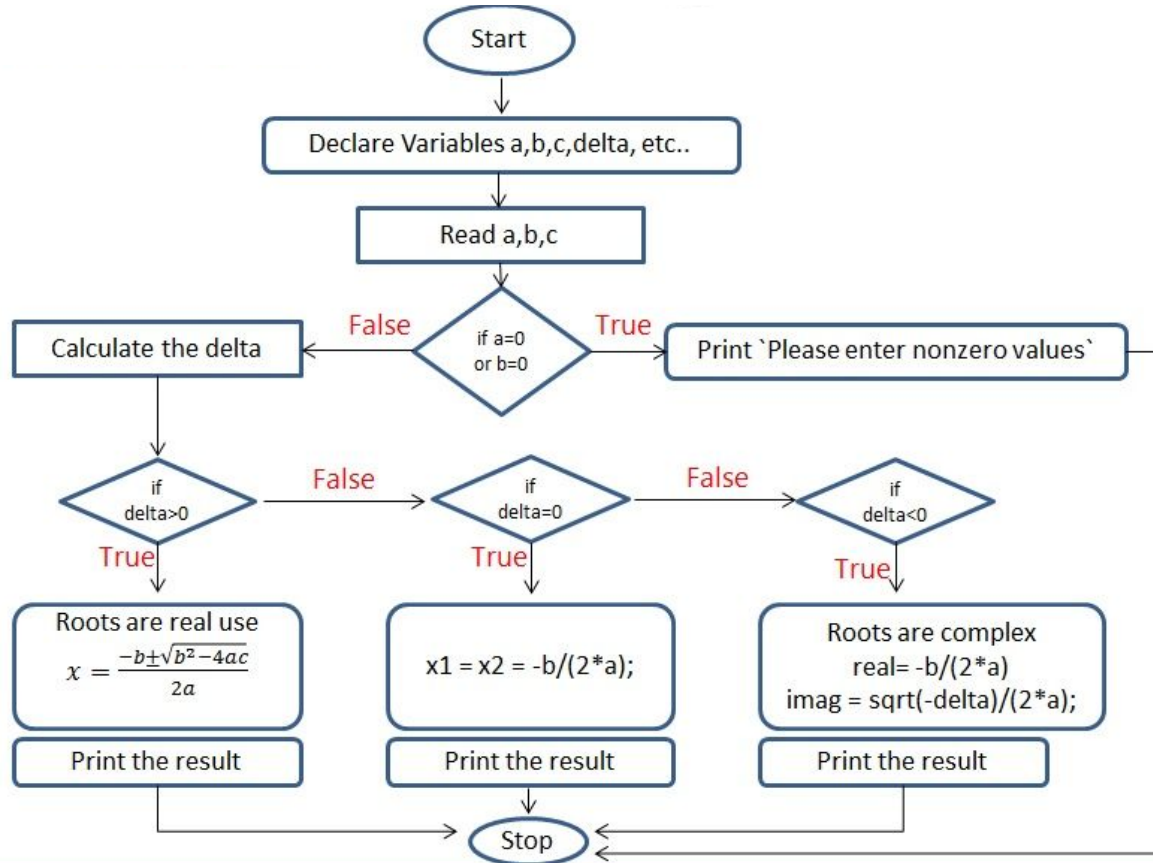
# Algorithms

- Muhammad ibn Musa **al-Khwarizmi**
- Persian mathematician
- Introduced the concept of **algorithm**  
also considered the father of algebra





# An algorithm for solving $ax^2+bx+c=0$



# An algorithm for solving $ax^2+bx+c=0$

```
#include <iostream>
#include <math.h>
using namespace std;

int main()
{
    double a, b, c;
    double result;

    cout << "\t\tQuadratic Equation Program\n\n";

    cout << "Enter a number for 'a' that is not 0: ";
    cin >> a;

    cout << endl << endl << "Now enter a number for 'b': ";
    cin >> b;

    cout << endl << endl << "Finally, enter a number for 'c': ";
    cin >> c;

    result = (-b + sqrt(pow(b, 2) - 4*a*c))/(2*a);

    return 0;
}
```

# Glossary

You can speak of **machine learning algorithms** that use a mathematics + statistics to find patterns about data.

When those patterns look pretty complicated, people call it **artificial intelligence**.

(if they're simple we just don't bother calling it AI)

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( **Forty percent of 'AI startups' in Europe don't actually use AI, claims report**

*Companies want to take advantage of the AI hype*

# Glossary

- **Data science / Data-driven science**
  - a mix of mathematics, statistics, and computer science  
...also often misused as a fancy name for statistics.
- **Machine learning**
  - mathematical foundations and practical applications of systems that learn with data.
    - Zoubin Ghahramani, University of Cambridge, Uber AI

# Machine Learning

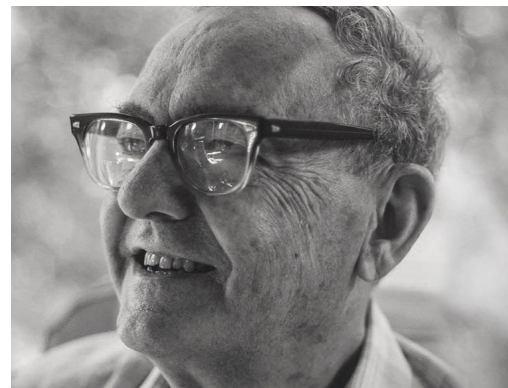
- What is learning?
- If you had to explain what learning is to an alien (or to a robot), how would you explain it?



# Machine Learning

“Learning denotes changes in the system that are adaptive in the sense that they enable the system to do the same task (or tasks drawn from a population of similar tasks) more effectively the next time.”

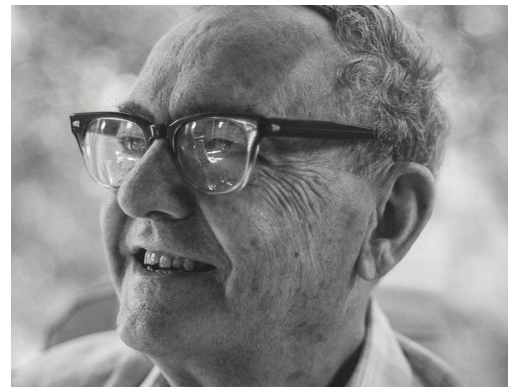
- Herbert Simon, Nobel Prize, Turing Award



# Machine Learning

“Learning denotes changes in the system that are adaptive in the sense that they enable the system to do the same task (or tasks drawn from a population of similar tasks) more effectively the next time.”

- Herbert Simon, Nobel Prize, Turing Award





# Machine Learning

Some kind of  
memory

“Learning denotes changes in the system in the sense that they enable the system to do the same task (or tasks drawn from a population of similar tasks) more effectively the next time.”

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# Machine Learning

“Learning denotes changes in the system in the sense that they enable the system to do the same task (or tasks drawn from a population of similar tasks) more effectively the next time”

Some kind of  
memory

A generalisation  
of the same task

- Herbert Simon, Nobel Prize,



# Machine Learning

“Learning denotes changes in the system in the sense that they enable the system to do the same task (or tasks drawn from a population of similar tasks)

more effectively the next time

Some kind of  
memory

Improvement

Nobel Prize,

A generalisation  
of the same task

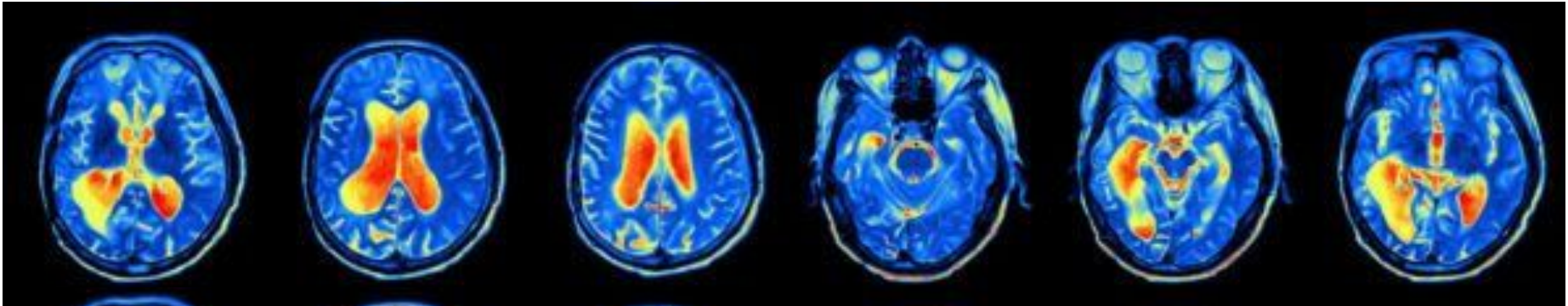


# Examples:

- Classification
  - Medical diagnosis: given a brain scan, can the machine identify regions with abnormal behaviour?
- Planning and acting
  - Playing games, driving a car
- Prediction:
  - Predicting sales, income, customers, ad revenue...

# Data science requires data.

- Classification
  - Medical diagnosis: given a brain scan, can the machine identify regions with abnormal behaviour?



# Quantitative data

It is 180cm tall

Weighs 50 kg

Has 2 dictionaries

Has 3 shelves

Sells for £100



# Qualitative data

It is made of wood

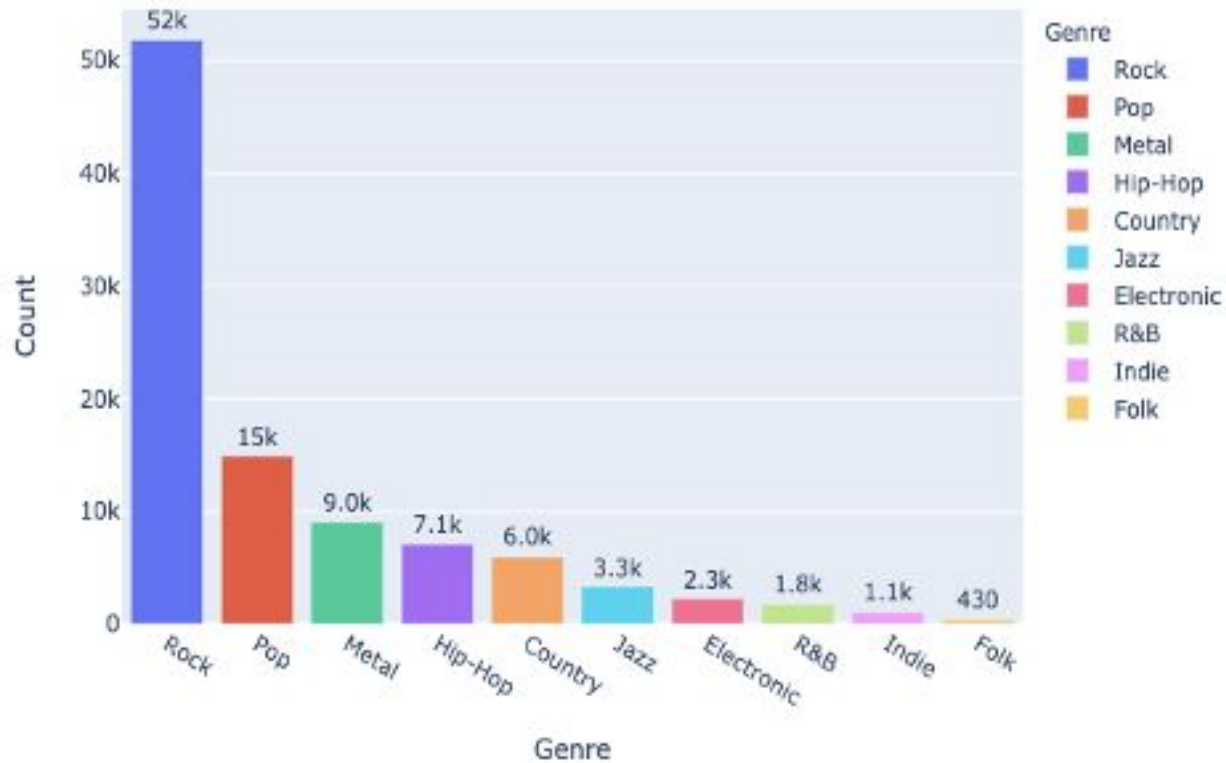
Made in Italy

Brown

Smells like oak

Smooth finish

# Categorical data



# Ordinal data

	Very Good	Good	OK	Poor	Very Poor
Gym	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Pool	<input type="checkbox"/>	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>
Cafe	<input type="checkbox"/>	<input checked="" type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>	<input type="checkbox"/>

	Strongly Disagree	Disagree	Neutral	Agree	Strongly Agree
I'm planning on buying this product again	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>
I'd recommend this product to someone else	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input type="radio"/>	<input checked="" type="radio"/>



# Numerical, continuous

\$\$\$\$



\$\$\$



\$\$\$\$\$\$



\$\$\$



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\$\$\$



# Numerical, discrete

3 rooms



2 rooms



10 rooms



1 rooms



7 rooms



5 rooms



# Mixed

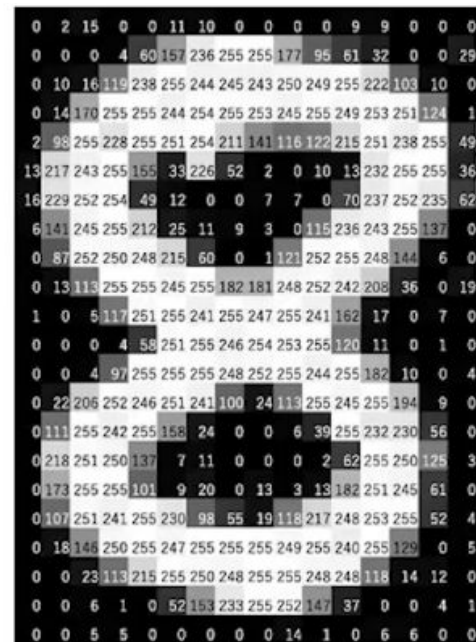
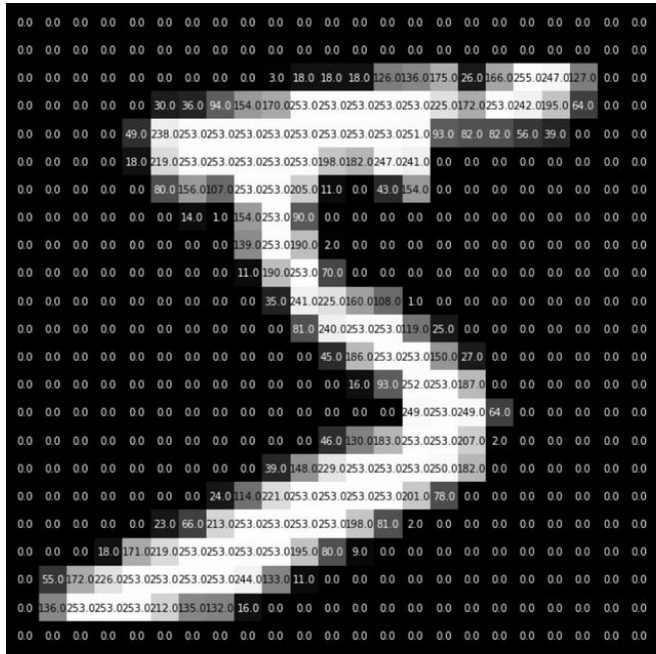
		Type 1	Type 2	Total	HP	Attack	Defense	Sp. Atk	Sp. Def	Speed	Generation	Legendary
			Poison	318	45	49	49	65	65	45	1	False
			Poison	405	60	62	63	80	80	60	1	False
			Poison	525	80	82	83	100	100	80	1	False
			Poison	625	80	100	123	122	120	80	1	False
5	4	Charmander	Fire	309	39	52	43	60	50	65	1	False
6	5	Charmeleon	Fire	405	58	64	58	80	65	80	1	False
7	6	Charizard	Fire	534	78	84	78	109	85	100	1	False
8	6	CharizardMega Charizard X	Fire	634	78	130	111	130	85	100	1	False
9	6	CharizardMega Charizard Y	Fire	634	78	104	78	159	115	100	1	False
10	7	Squirtle	Water	314	44	48	65	50	64	43	1	False
11	8	Wartortle	Water	405	59	63	80	65	80	58	1	False
12	9	Blastoise	Water	530	79	83	100	85	105	78	1	False
13	9	BlastoiseMega Blastoise	Water	630	79	103	120	135	115	78	1	False
14	10	Caterpie	Bug	195	45	30	35	20	20	45	1	False
15	11	Metapod	Bug	205	50	20	55	25	25	30	1	False
16	12	Butterfree	Bug	395	60	45	50	90	80	70	1	False
17	13	Weedle	Bug	195	40	35	30	20	20	50	1	False
18	14	Kakuna	Bug	205	45	25	50	25	25	35	1	False
19	15	Beedrill	Bug	395	65	90	40	45	80	75	1	False
20	15	BeedrillMega Beedrill	Bug	495	65	150	40	15	80	145	1	False
21	16	Pidgey	Normal	251	40	45				56	1	False
22	17	Pidgeotto	Normal	349	63	60				71	1	False
23	18	Pidgeot	Normal	479	83	80	75	70		101	1	False
24	18	PidgeotMega Pidgeot	Normal	579	83	80	80	135	80		1	False
25	19	Rattata	Normal	253	30	56						
26	20	Raticate	Normal	413	55	81						
27	21	Spearow	Normal	262	40	60						
28	22	Fearow	Normal	442	65	90						
29	23	Ekans	Poison	288	35	60						
30	24	Arbok	Poison	438	60	85						
31	25	Pikachu	Electric	320	35	55						
32	26	Raichu	Electric	485	60	90						

POKÉMON



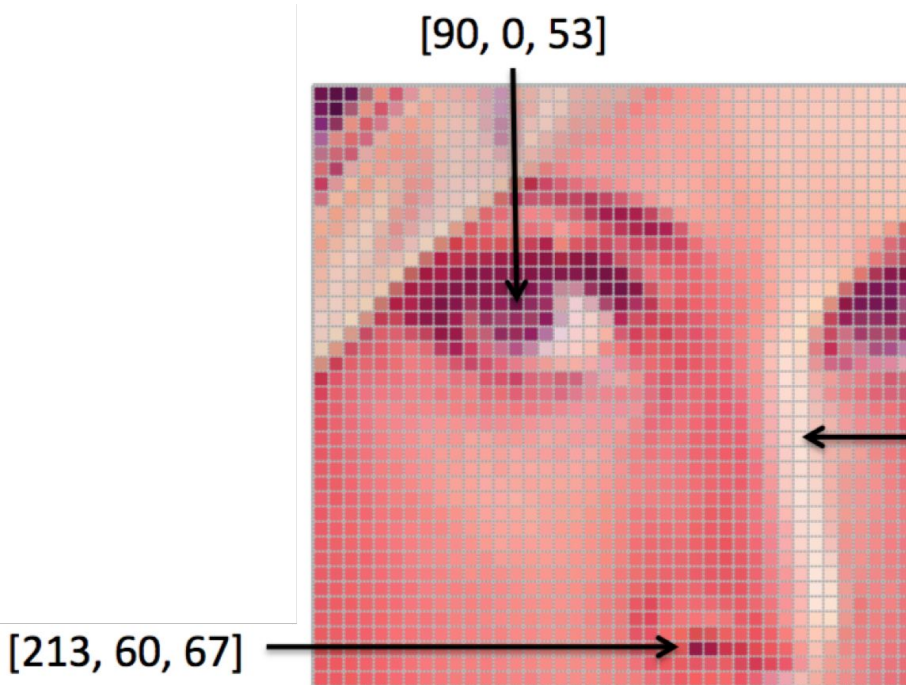
# Image data

Images are usually represented as matrices.  
A 400 x 400 pixels image can be represented by a 400 x 400 matrix.



# Image data

A 400 x 400 pixels image with colours can be represented by three 400 x 400 matrices, for the **red**, **green** and **blue** channels.

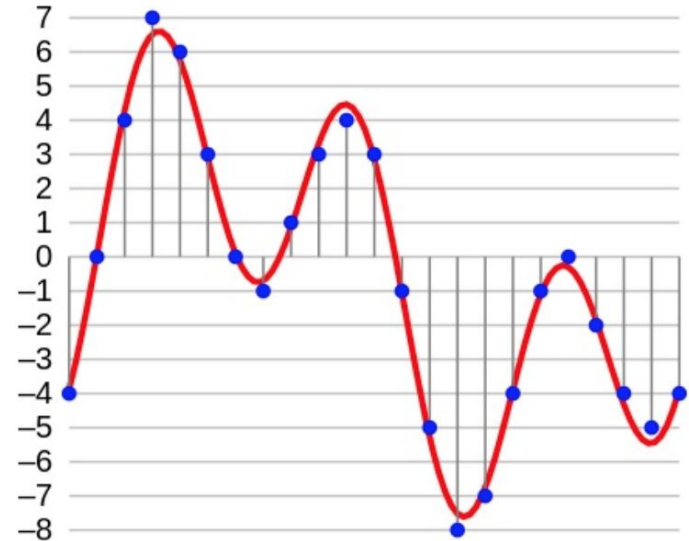




# Text data

```
sales - Notepad
File Edit Format View Help
"Country","Salesperson","Order Amount","Quarter"
"UK","Smith",16753,"Qtr 3"
"USA","Johnson",14808,"Qtr 4"
"UK","Williams",10644,"Qtr 2"
"USA","Jones",1390,"Qtr 3"
"USA","Brown",4865,"Qtr 4"
"UK","Williams",12438,"Qtr 1"
"UK","Johnson",9339,"Qtr 2"
"USA","Smith",18919,"Qtr 3"
"USA","Jones",9213,"Qtr 4"
"UK","Jones",7433,"Qtr 1"
"USA","Brown",3255,"Qtr 2"
"USA","Williams",14867,"Qtr 3"
"UK","Williams",19302,"Qtr 4"
"USA","Smith",9698,"Qtr 1"
"USA","Jones",18978,"Qtr 2"
"UK","Brown",9080,"Qtr 4"
```

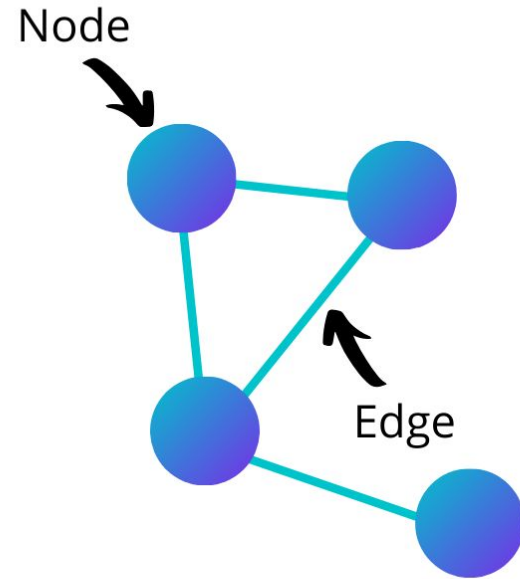
# Sound data



Sound wave

# Network data, or relational data

```
[  
  node1 : node2  
  node2 : node3  
  node1 : node3  
  node4 : node1  
]
```



# Most of the time: mixed data

```
age                job                marital      education      default
Min.      :23.00  management   :454      married :431    primary   : 39    no :642
1st Qu.   :33.00  technician  : 45      single  :139    secondary: 26    yes:  9
Median    :37.00  entrepreneur: 44    divorced: 81    tertiary  :572
Mean      :39.41  self-employed: 34    unknown   : 14
3rd Qu.   :44.50  admin.      : 24
Max.      :75.00  unemployed  : 14
          (other) : 36
```



# Next steps

- Next lecture: Linear regressions, Loss functions and  $R^2$

## Additional material:

- Tom M. Mitchell, [The Discipline of Machine Learning](#)
- Chp. 1, Kevin P. Murphy, M Cambridge Press 2012.
- Chp. 1, Andrew R. Webb, Statistical Pattern Recognition, 2nd edition, Wiley 2002.
- Chp. 1, Dutta et al, Pattern Classification 2nd edition, Wiley 2000
- Chp. 1, Nilsson et al, [Introduction to Machine Learning](#), 1998
- More on ELE