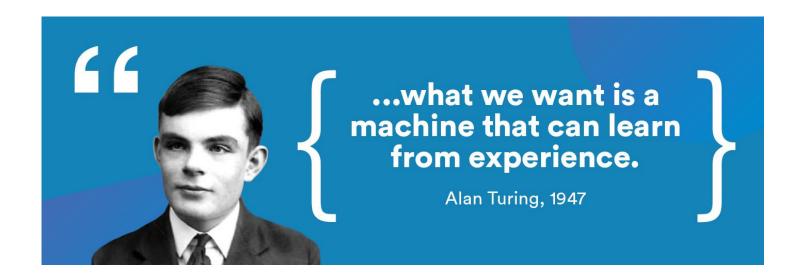
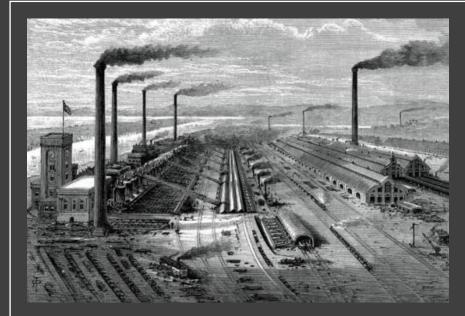
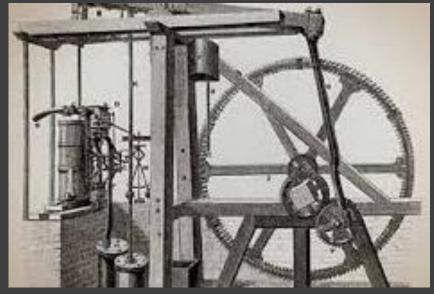
## Introduction to Machine Learning

Somebody



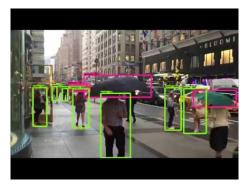
- "A breakthrough in machine learning would be worth ten Microsofts" (Bill Gates, Chairman, Microsoft)
- "Machine learning is the next Internet" (Tony Tether, Director, DARPA)
- "Al is one of the most important things humanity is working on. It is more profound than, I don't know, electricity or fire." Sundar Pichai (Google CEO)



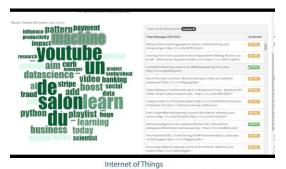




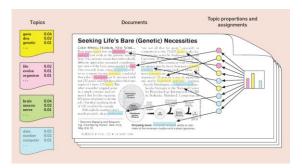












#### Recommender Systems



### Self-driving cars







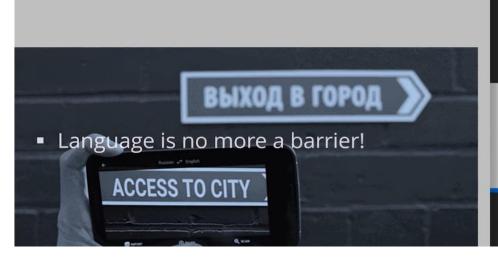




- Recommends ads based on your search
- Machine Learning is used in generating



- Driverless Cars!
- Tesla's Al is driven by Nvidia's H/W focusing mainly on unsupervised learning
- Crowdsources data from all of its vehicles and its drivers - internal & external sensors











NETFLIX

**Recommender System** 

#### **Machine Learning at Apple**

- Smartphone with facial recognition
- Core of the face detection machine learning algorithms





## Google

- Google: 3.5 billion search queries every day.2.
- Facebook: 350 million photos are uploaded to Facebook each day.
   Facebook generates 4 petabytes of data every day.
- Every day, 306.4 billion emails are sent, and 5 million Tweets are made in Twitter:
- Astronomy: Satellite data is in hundreds of PB.

## Big data Era

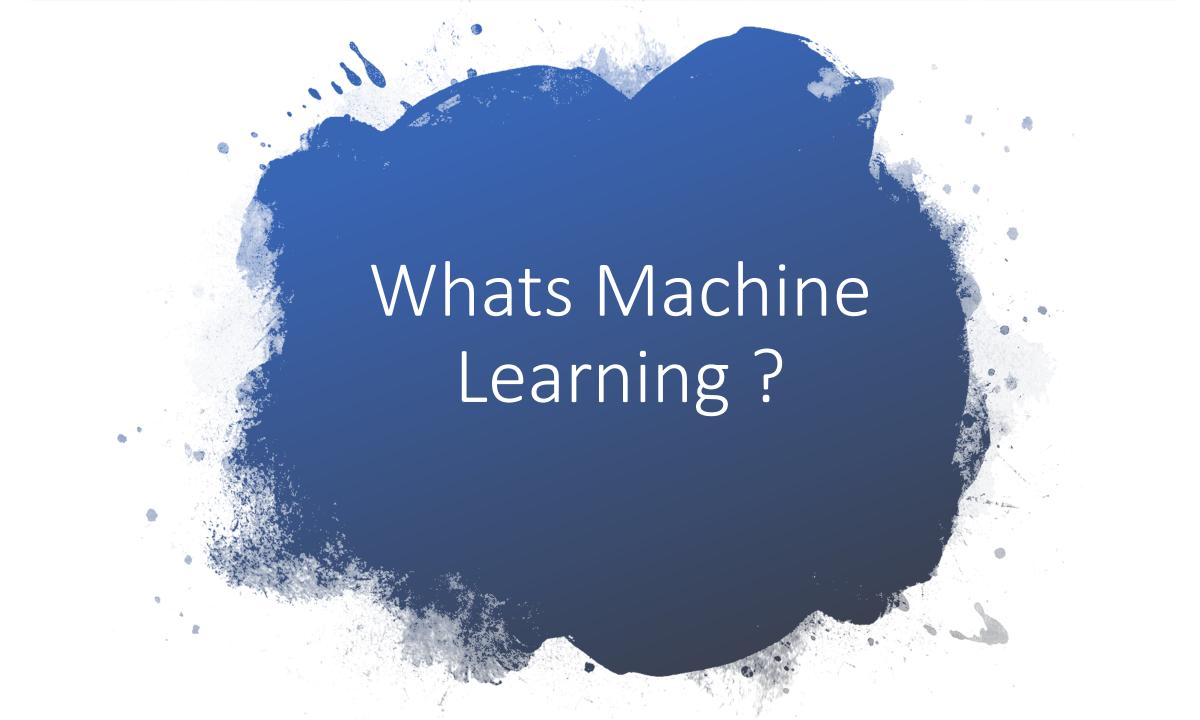
- •1.7MB of data is created every second by every person during 2020.
- •In the last two years alone, the astonishing **90%** of the world's data has been created.
- •2.5 quintillion bytes of data are produced by humans every day.
- •463 exabytes of data will be generated each day by humans as of 2025.
- •95 million photos and videos are shared every day on Instagram.
- •By the end of 2020, **44 zettabytes** will make up the entire digital universe.
- •Every day, **306.4 billion emails** are sent, and **5 million Tweets** are made.

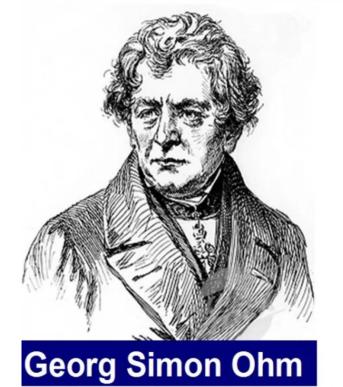


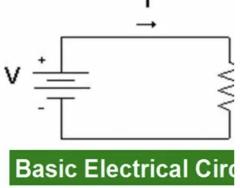
## Why machine learning: Data every where!

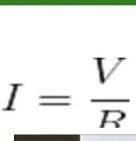
"We are drowning in information and starving for knowledge." — John Naisbitt.

- This deluge of data calls for automated methods of data analysis, which is what machine learning provides.
- Defined as a "set of methods that can automatically detect patterns in data, and then
  use the uncovered patterns to predict future data, or to perform other kinds of
  decision making under uncertainty" Kevin Murphy (Machine Learning: A
  Probabilistic Perspective)

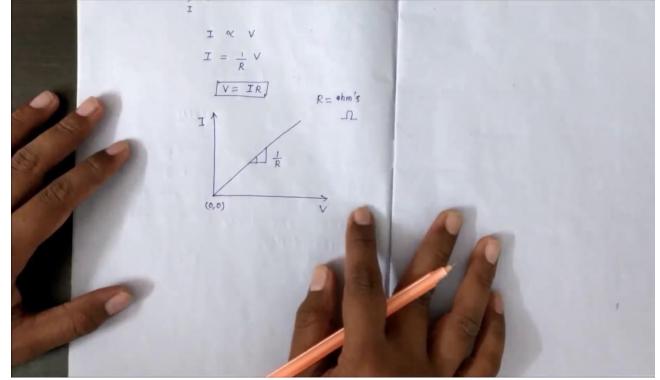












## Estimate Resistance

• Connect the variable voltage supply to both the ends of the rheostat. Connect the ammeter in series of the rheostat. Connect the voltmeter in parallel of the rheostat. Start measuring the voltage and current as you move the rheostat moving hand from minimum position to the maximum position in the steps of constant increase in current.

Potential difference V (in volt)	0.5	1.0	1.5	2.0	2.5
Current I (in ampere)	0.2	0.4	0.6	0.8	1.0

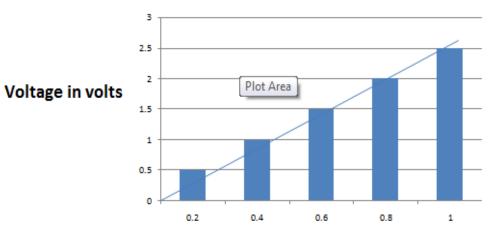
## Estimating Resistance is machine learning!

Potential difference V (in volt)	0.5	1.0	1.5	2.0	2.5
Current I (in ampere)	0.2	0.4	0.6	0.8	1.0

• A currentvoltage characteristic or IV curve (currentvoltage curve) is a relationship, typically represented as a chart or graph, between the electric current through a circuit, device, or material, and the corresponding voltage, or potential difference across it. In the graph, the voltage is plotted along the y-axis and the current is plotted along the x-axis.

V = R I

resistance is 2.5 ohms!

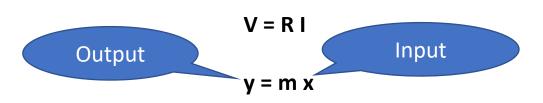


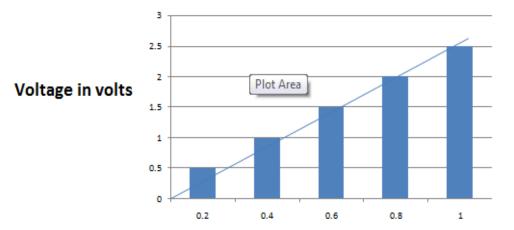
**Current in amps** 

## Estimating Resistance is machine learning!

Potential difference V (in volt)	0.5	1.0	1.5	2.0	2.5
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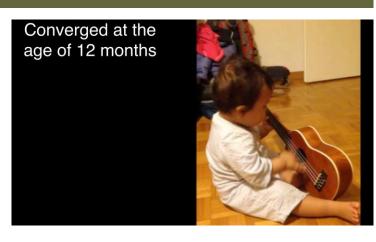


Current in amps

## What's machine learning?

Human Learning at the age of 6 months.





Transfer skills at the age of 14 months





- Algorithms which let machines learn like humans from observations!
- To discover the fundamental principles of learning from data and use them to develop algorithms that can learn like living beings!
- Programming computers to optimize a performance criterion using example data or past experience(Ethem Alpay din, Machine Learning, 2010)
- How do we create computer programs that improve with experience? (Tom Mitchel)

## Machine learning

## **Traditional Programming**



Machine Learning, Data Mining, Knowledge Discovery,
Artificial Intelligence, Statistical Learning, Pattern Recognition,
Computational Learning



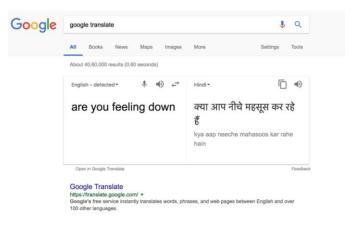




## ML based applications

-Machine learning has become prominent approach to solve problems in AI domains like computer vision, language and speech processing

#### Face detection





Viola-Jones method.

#### Jeopardy! (2011): Humans vs. IBM Watson



By Rosemaryetoufee (Own work), via Wikimedia Commons

Natural Language Understanding and information extraction!

#### Speech recognition

- Virtual assistants: Siri (Apple), Echo (Amazon), Google Now, Cortana (Microsoft).
- "They" helps get things done: send an email, make an appointment, find a restaurant, tell you the weather and more.
- Leverage deep neural networks to handle speech recognition and natural language understanding.



# ML based applications

- Machine learning has become prominent approach to solve problems in AI domains like computer vision, language and speech processing
- Early approaches to AI was based on logic but applications has to face a lot of uncertain situations and has to perform well on unseen situations.
- Machine learning focused on developing algorithms which could perform well on future unseen data (generalization performance) which differentiates it from statistics

#### Go (2016): Lee Sedol versus Google AlphaGo





(Left) By LG Electronics, via Wikimedia Commons (Right) By Google DeepMind, via
Wikimedia Commons

Deep Learning, reinforcement learning, and search algorithms!

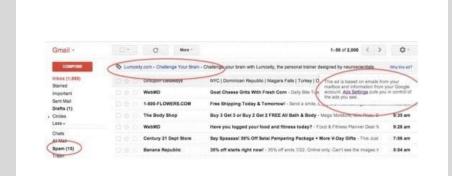
### Chat bots



https://botkit.a

#### Autonomous driving



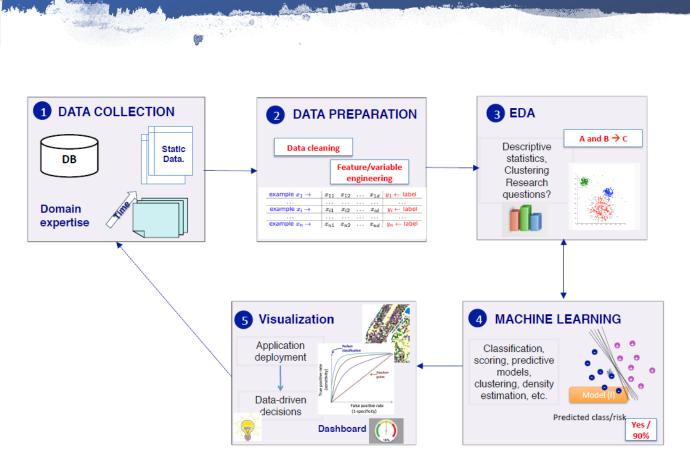


## Machine learning is interdisciplinary

- Science (Astronomy, neuroscience, medical imaging, bio-informatics)
- Environment (energy, climate, weather, resources)
- Retail (Intelligent stock control, demographic store placement)
- Manufacturing (Intelligent control, automated monitoring, detection methods)
- Security (Intelligent smoke alarms, fraud detection)
- Marketing (promotions, ...)
- Management (Scheduling, timetabling)
- Finance (credit scoring, risk analysis...)
- Web data (information retrieval, information extraction, ...)



## ML in practice



- Understanding domain, prior knowledge, and goals
- Data integration, selection, cleaning, pre-processing, etc.
- Learning models
- Interpreting results
- Consolidating and deploying discovered knowledge
- Loop

## Overview of Machine learning

#### **Supervised learning**

- Predict an output y when given an input x
- For categorical y : classification.
- For real-valued y: regression.

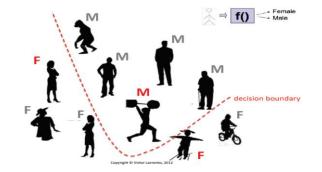
#### **Unsupervised learning**

- Create an internal representation of the input, e.g. clustering, dimensionality reduction
- This is important in machine learning as getting labels is often difficult and expensive

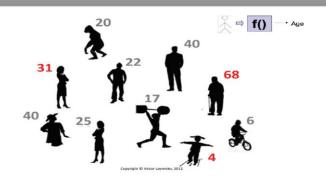
#### Other settings of ML

- Reinforcement learning (learning from "rewards")
- Semi-supervised learning (combines supervised + unsupervised)
- Active learning, Transfer learning, Structured prediction

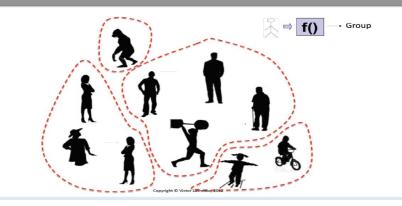
## Classification (Supervised Learning)

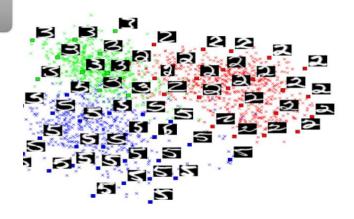


## Regression (Supervised Learning)



## Clustering (Unsupervised Learning)





Input  $\mathbf{x} \in \mathcal{R}^d$ 

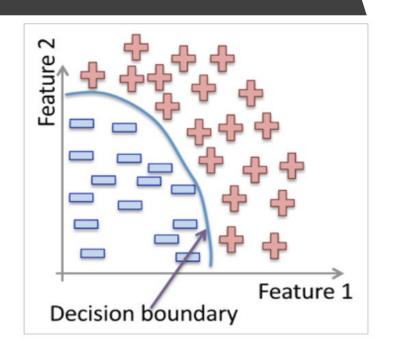
Input

and output y a label. Learn a function

 $f: \mathbf{x} \to y$ 

Output

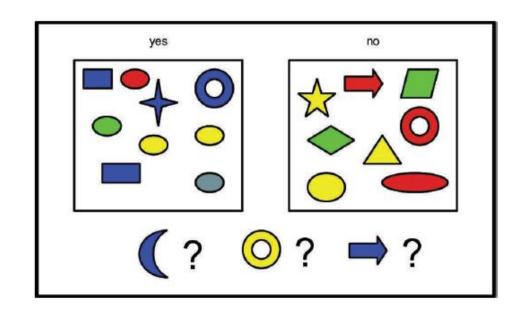
# Supervised learning (classification)

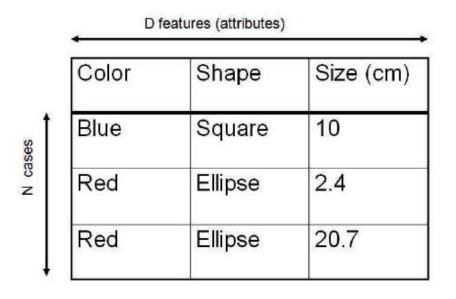


example $x_1 \rightarrow$	X 11	X12	 X 1d	y <sub>1</sub> ←-label
example <i>x</i> <sub>i</sub> →	<i>X i</i> 1	Xi2	 x <sub>id</sub>	y <sub>i</sub>
example <i>xn</i> →	<i>X</i> <sub><i>n</i>1</sub>	X <sub>n</sub> 2	 X <sub>nd</sub>	<i>yn</i> ←−label

fruit	length	width	weight	label
fruit 1	165	38	172	Banana
fruit 2	218	39	230	Banana
fruit 3	76	80	145	Orange
fruit 4	145	35	150	Banana
fruit 5	90	88	160	Orange
fruit n				

## Supervised learning (Classification )





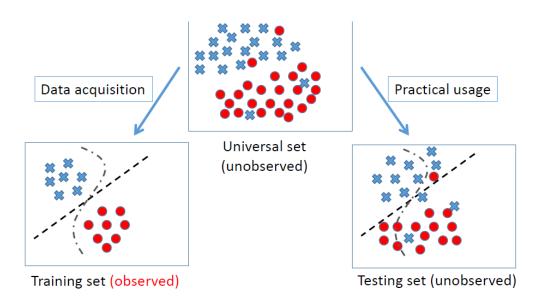
Label
1
1
0

Input  $\mathbf{x} \in \mathcal{R}^d$ and output y a real value. Learn a function  $f: \mathbf{X} \to y$ **Estimating Price of a house** Supervised learning (Regression) Output Price (\$10'000) \$70'000 Size of the House

Input

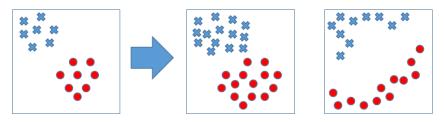
Goal is to learn a function which maps inputs to outputs so that it will predict well on future data points — Generalization performance

## Training and Testing ML models



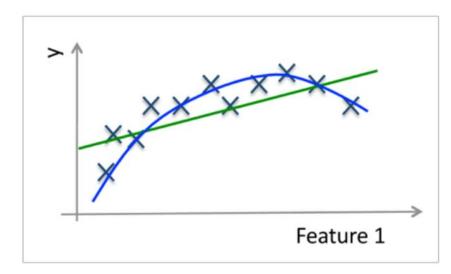
#### No free lunch rule:

- Training set and testing set may not come from the same distribution
- Need to make some assumptions or bias

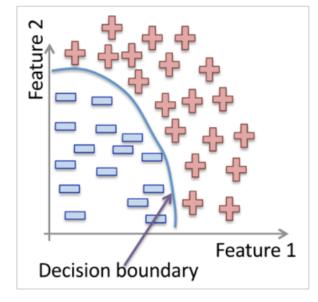


# Machine learning algorithms

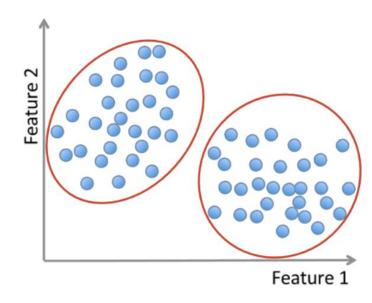
### Regression:



**Method :** Linear regression, support vector regression, gaussian process regression



**Methods:** Support Vector Machines, neural networks, decision trees, K-nearest neighbors, naive Bayes, etc.



**Methods**: K-means, gaussian mixtures, hierarchical clustering, spectral clustering, etc.

## Brief Syllabus: Foundations of ML

- Overview of machine learning, Basic probabilistic and non-probabilistic models, supervised learning, Non-parametric modeling, Model selection, Unsupervised Learning, Representation Learning, Online Learning, Reinforcement Learning, Ensemble methods, Learning with Sequential Data.
  - Main references:
  - Bishop, Christopher M. Pattern Recognition and Machine Learning. Springer, 2006.
  - Murphy, K. P. (2013). Machine learning: a probabilistic perspective. Cambridge, Mass. [u.a.]: MIT Press
  - Hastie, T., R. Tibshirani, and J. H. Friedman. The Elements of Statistical Learning: Data Mining, Inference and Prediction. New York, NY: Springer, 2001
  - Mitchell, Tom. Machine Learning. New York, NY: McGraw-Hill, 1997
  - Alpaydin, Ethem Introduction to Machine Learning. MIT Press, 2014.

#### ML Resources

- MOOCs
  - Coursera, EdX, Udacity
- Conferences/Journals
  - JMLR, Machine Learning, IEEE Transactions on Neural Networks and Learning Systems, IEEE Transactions on Pattern Analysis and Machine Intelligence, Annals of Statistics
  - ICML, NIPS, KDD, IJCAI, AAAI, ICDM

### **ML** Datasets

- UCI Repository: http://www.ics.uci.edu/~mlearn/MLRepository.html
- Statlib: http://lib.stat.cmu.edu/
- Kaggle
- Many more...

## Foundations of ML

- Assume mathematical and programming basics
- Functions, Logarithms and Exponentials
- Vectors, Dot Products, Orthogonality
- Matrices, Matrix Operations, Linear Transformations,
- Eigendecomposition
- Calculus, Differentiation, Integration
- Probability and Statistics
- Functional Analysis, Hilbert Spaces
- Python
- Numpy, Scipy numerical/scientific computing, linear algebra
- Matplotlib for plotting
- Scikitlearn for machine learning

## Math

- Chapter 2 of Pattern recognition and machine learning by Christopher Bishop
- Part 1 of Deep Learning book: http://www.deeplearningbook.org/

Mathematics for Machine Learning https://mml-book.github.io/

- Essence of linear algebra:
   http://youtu.be/kjBOesZCoqc
- Essence of calculus: https://goo.gl/Hnk1jA
- Programming
- Practice Python
- https://try.jupyter.org/
- https://docs.python.org/3/tutorial/
- Video Tutorials:

https://www.youtube.com/watch?v=cpPG0bKHYKc

• Play with Numpy, Matplotlib, scikitlearn

## Foundations of ML: Evaluation

Tentative Evaluation scheme:

Regular online quizzes (50%) + Assignment and Project (50%)

