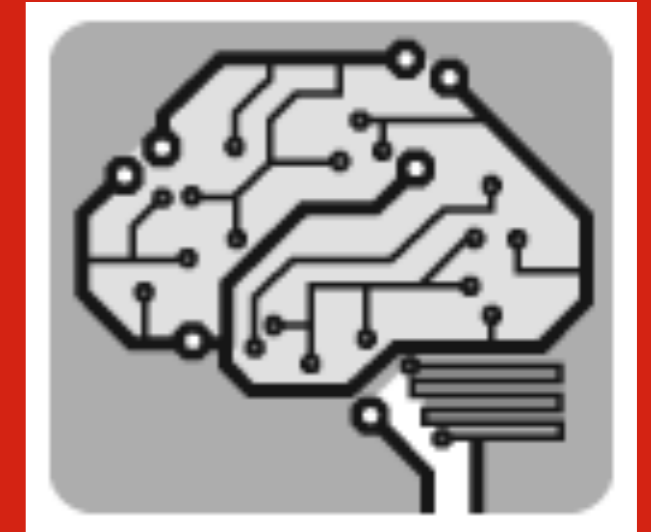


# MLCC 2017

## Machine Learning Crash Course

Universita' di Genova, Summer, 2017



*Instructor:* Lorenzo Rosasco

*Organizers:* Gian Maria Marconi, Fabio Anselmi,

*Workshop organizer:* Raffaello Camoriano

Dibris



The Center for Brains,  
Minds & Machines



Massachusetts  
Institute of  
Technology

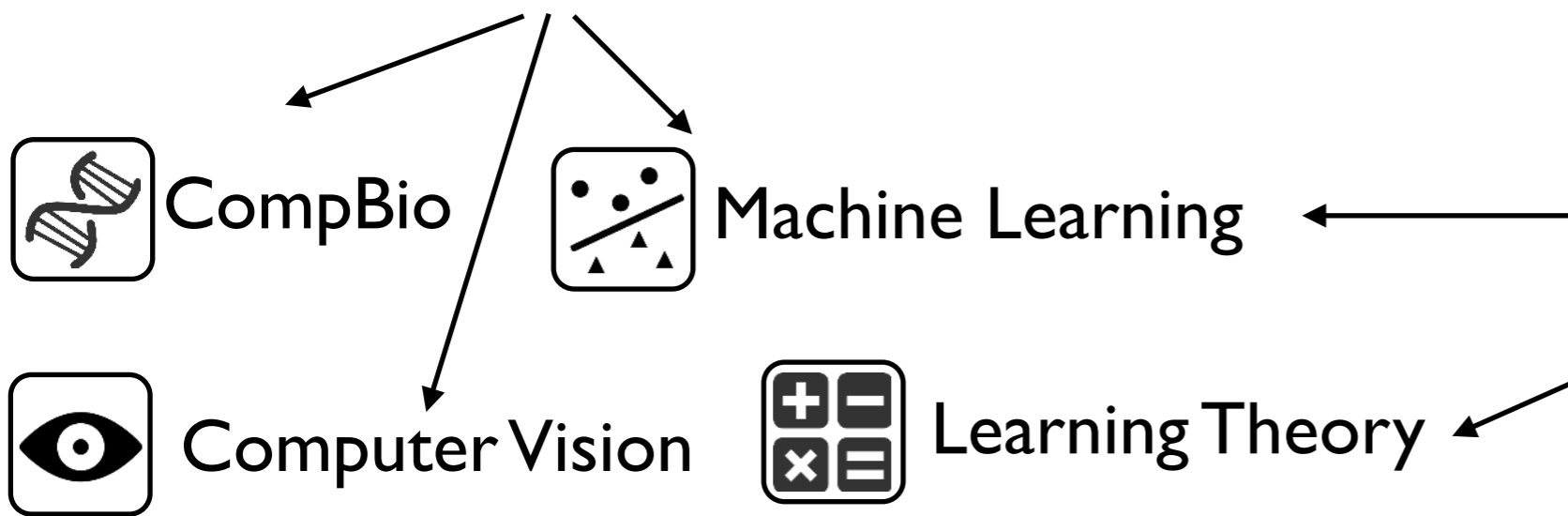


ISTITUTO ITALIANO DI TECNOLOGIA

LABORATORY FOR COMPUTATIONAL AND STATISTICAL LEARNING

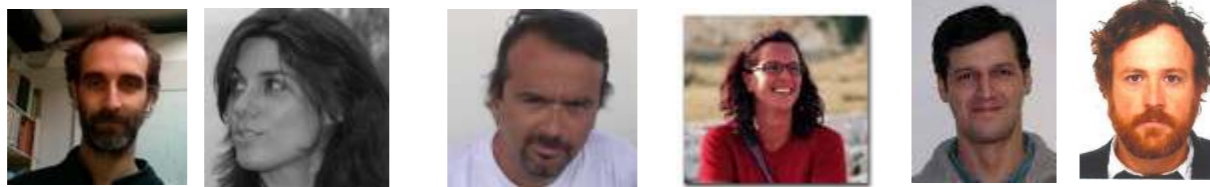


# Dibris



Laboratory for Computational & Statistical Learning

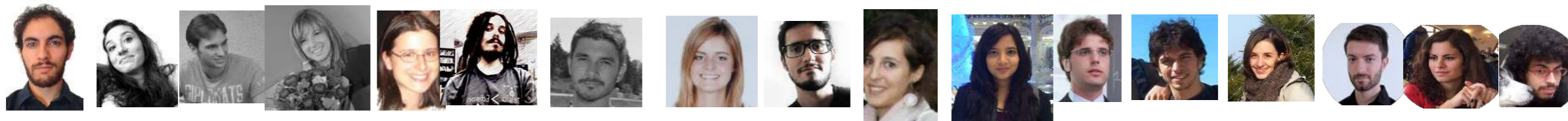
6+3 Faculty



7 PostDoc



~15 PhD+  
master





# From RegML to MLCC

## RegML- Regularization Methods for Machine Learning

(baby 9.520@MIT)

- 2010, 35 attendees
- 2011, 50 attendees
- 2012, 50 attendees (@BISS)
- 2013, 85 attendees
- 2014, 95 attendees
- 2016, 120 attendees
- 2017, 80 attendees (@OSLO)

ADVANCED



## MLCC- Machine Learning Crash Course

(baby ISML2@DIBRIS)

- 2014, 85 attendees
- 2015, 120+ attendees
- 2017, 120+ attendees

INTRO





## ML Desert Island Compilation

An introduction to **essential** Machine Learning:

- Concepts
- Algorithms



# Course at a Glance

CLASS	DAY	TIME	SUBJECT
1	26/06	9:30 - 11:00	Introduction to Machine Learning
2	26/06	11:30 - 13:00	Local Methods and Model Selection
3	26/06	14:00 - 16:00	Laboratory 1: Local Methods for Classification
4	27/06	9:30 - 11:00	Regularization Networks I: Linear Models
5	27/06	11:30 - 13:00	Regularization Networks II: Kernels
6	27/06	14:00 - 16:00	Laboratory 2: Regularization Networks
Talk	28/06	9:30 - 10:10	<i>Pietro Leo</i> , Executive Architect & CTO, IBM Italy
Talk	28/06	10:10 - 10:35	<i>Enrico Ferrari</i> , R&D Manager, Rulex Inc.
Talk	28/06	10:35 - 11:00	<i>Matteo Santoro</i> , CEO, Camelot Biomedical Systems s.p.a.
	28/06	11:00 - 11:30	Coffee Break
Talk	28/06	11:30 - 12:10	<i>Giovanni Zappella</i> , Machine Learning Scientist, Amazon Development Center Germany
Talk	28/06	12:10 - 12:35	<i>Luca Nardelli</i> , Founder and CTO @ Erya, Horus Technology
Talk	28/06	12:35 - 13:00	<i>Alessandro Verri</i> , Professor, University of Genoa
	28/06	Afternoon	Free
7	29/06	9:30 - 11:00	Dimensionality Reduction and PCA
8	29/06	11:30 - 13:00	Variable Selection and Sparsity
9	29/06	14:00 - 16:00	Laboratory 3: PCA and Sparsity
10	30/06	9:30 - 11:00	Clustering
11	30/06	11:30 - 13:00	Data Representation: Deep Learning

Day 1: Local Methods and Model Selection

Day 2: Regularization and nonparametrics

MLCC Workshop!

**COMPANIES!**

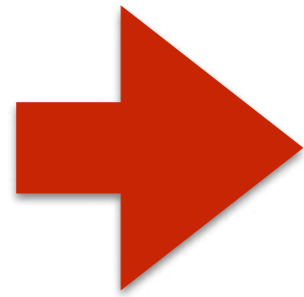
Day 3: Dimensionality Reduction and Sparsity

Day 4: DL & clustering

**Note:** Wed afternoon is vacation!

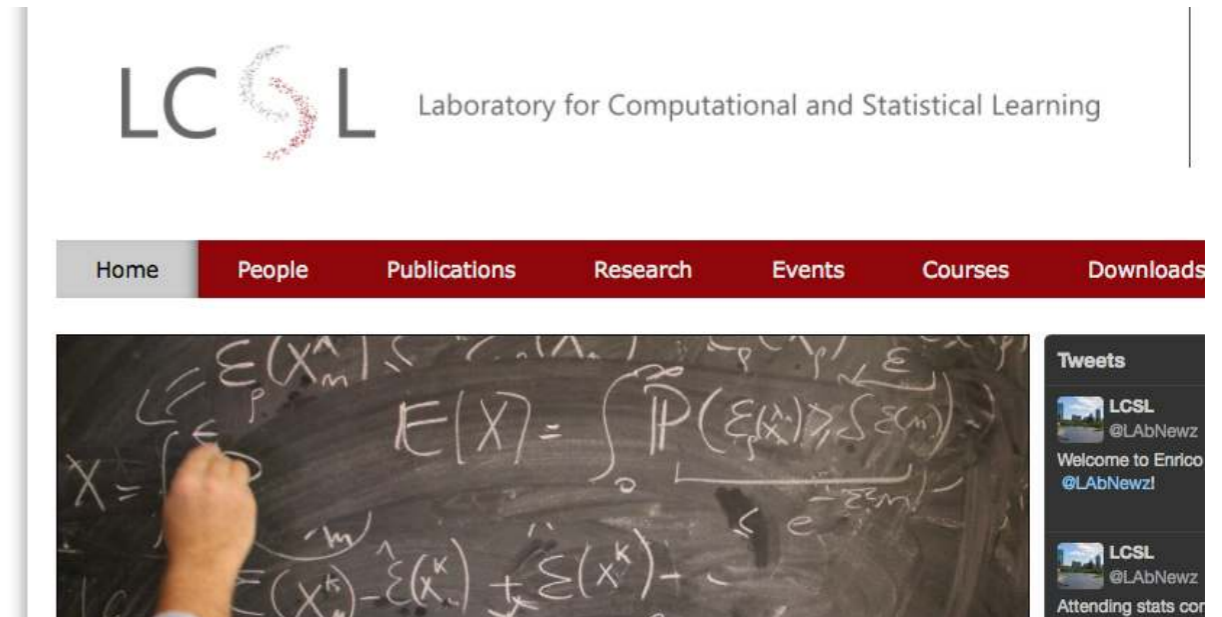
# Prerequisites and References

**Prerequisites:** The mathematical tools needed for the course are basic probability, calculus and linear algebra.



**References**

[lcs.mit.edu](http://lcs.mit.edu)



- L. Rosasco. [Introductory Machine Learning Notes](#).

## References:

- T. Hastie, R. Tibshirani, and J. Friedman. **The Elements of Statistical Learning: Prediction, Inference and Data Mining**. Second Edition, Springer Verlag, 2009 (available for free from the author's website).

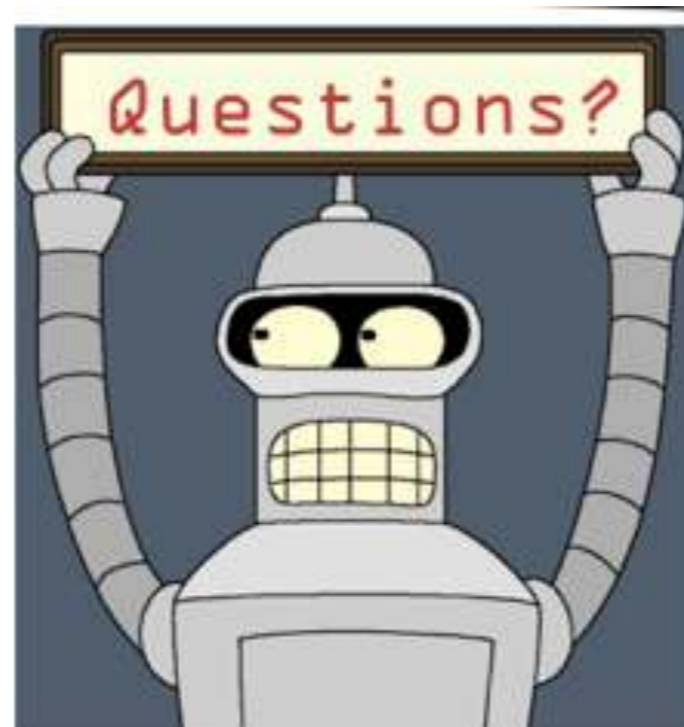
## Further readings :

- T. Poggio and S. Smale. **The Mathematics of Learning: Dealing with Data**. Notices of the AMS, 2003
- [Pedro Domingos](#). **A few useful things to know about machine learning**. Communications of the ACM [CACM Homepage archive](#). Volume 55 Issue 10, October 2012 Pages 78-87.
- ....

## Useful Links

- MIT 9.520: Statistical Learning Theory and Applications, Fall 2013 (<http://www.mit.edu/~9.520/>).
- Stanford CS229 Machine Learning Autumn 2013 (<http://cs229.stanford.edu>). See also the Coursera version (<https://www.coursera.org/course/ml>).

**ASK**



**+attendance!**

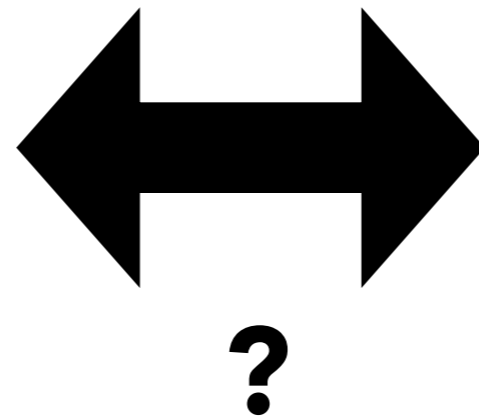


- A quick tour of machine learning
- Basic statistical learning theory
  
- Local algorithms
- Model selection



# What is (Machine) Learning?

**Intelligent Systems**



**Data Science**

# (Artificial) Intelligence: A Working Definition

## Turing test



Alan Turing 1912-1954



## Ingredients for AI

- natural language processing
- knowledge representation
- automated reasoning
- machine learning
  
- computer vision
- robotics to manipulate

# A Glimpse to the Past

1943

[Arturo Rosenblueth](#), [Norbert Wiener](#) and Julian Bigelow coin the term "[cybernetics](#)". Wiener's popular book by that name published in 1948.

.....

1948

[John von Neumann](#) (quoted by [E.T. Jaynes](#)) in response to a comment at a lecture that it was impossible for a machine to think: "You insist that there is something a machine cannot do. If you will tell me *precisely* what it is that a machine cannot do, then I can always make a machine which will do just that!". Von Neumann was presumably alluding to the [Church-Turing thesis](#) which states that any effective procedure can be simulated by a (generalized) computer.

...

1950

[Alan Turing](#) proposes the [Turing Test](#) as a measure of machine intelligence.

1950

[Claude Shannon](#) published a detailed analysis of [chess](#) playing as [search](#).

1955

The first [Dartmouth College summer AI conference](#) is organized by [John McCarthy](#), [Marvin Minsky](#), [Nathan Rochester](#) of [IBM](#) and [Claude Shannon](#).

.....

Late 1990s

[Web crawlers](#) and other AI-based information extraction programs become essential in widespread use of the [World Wide Web](#).

1997

The [Deep Blue](#) chess machine ([IBM](#)) beats the world [chess](#) champion, [Garry Kasparov](#).

.....





10/15 years ago



# How are we doing now?



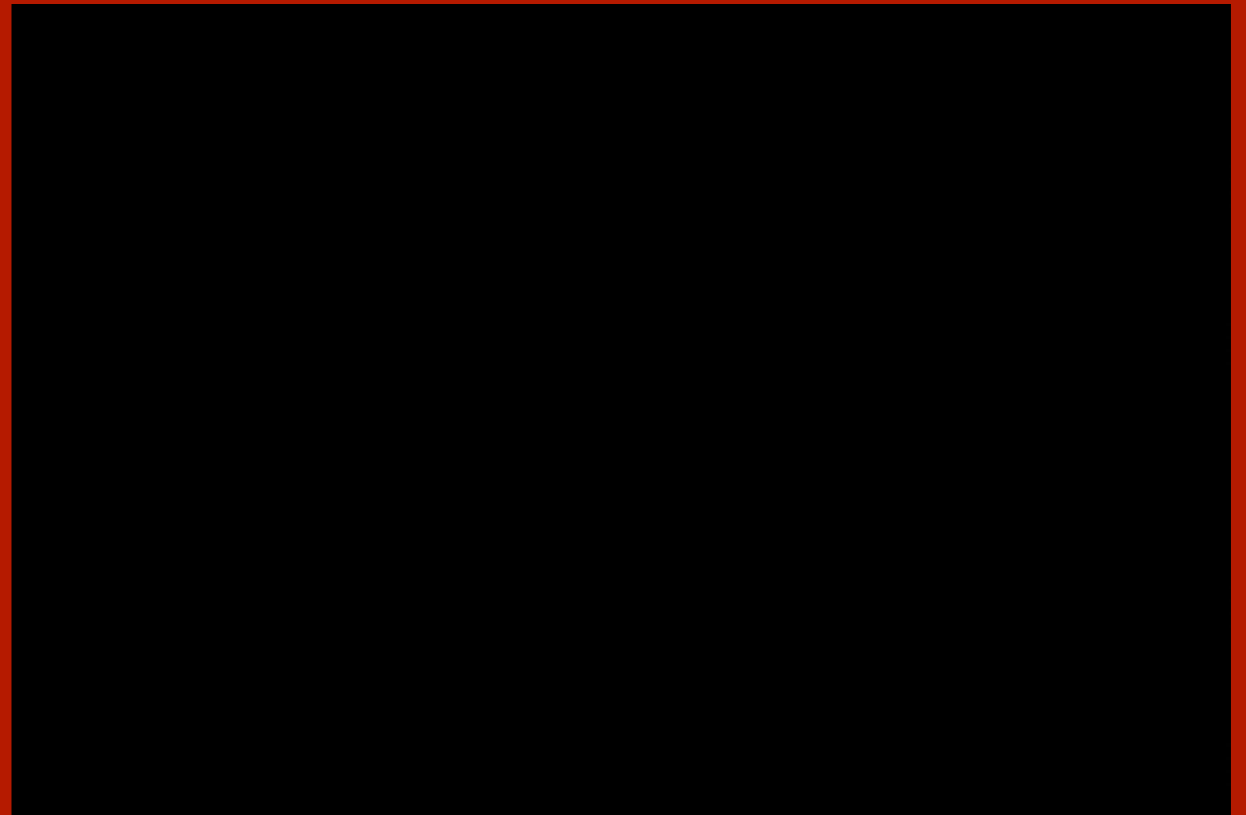


# Pedestrians Detection at Human Level Performance



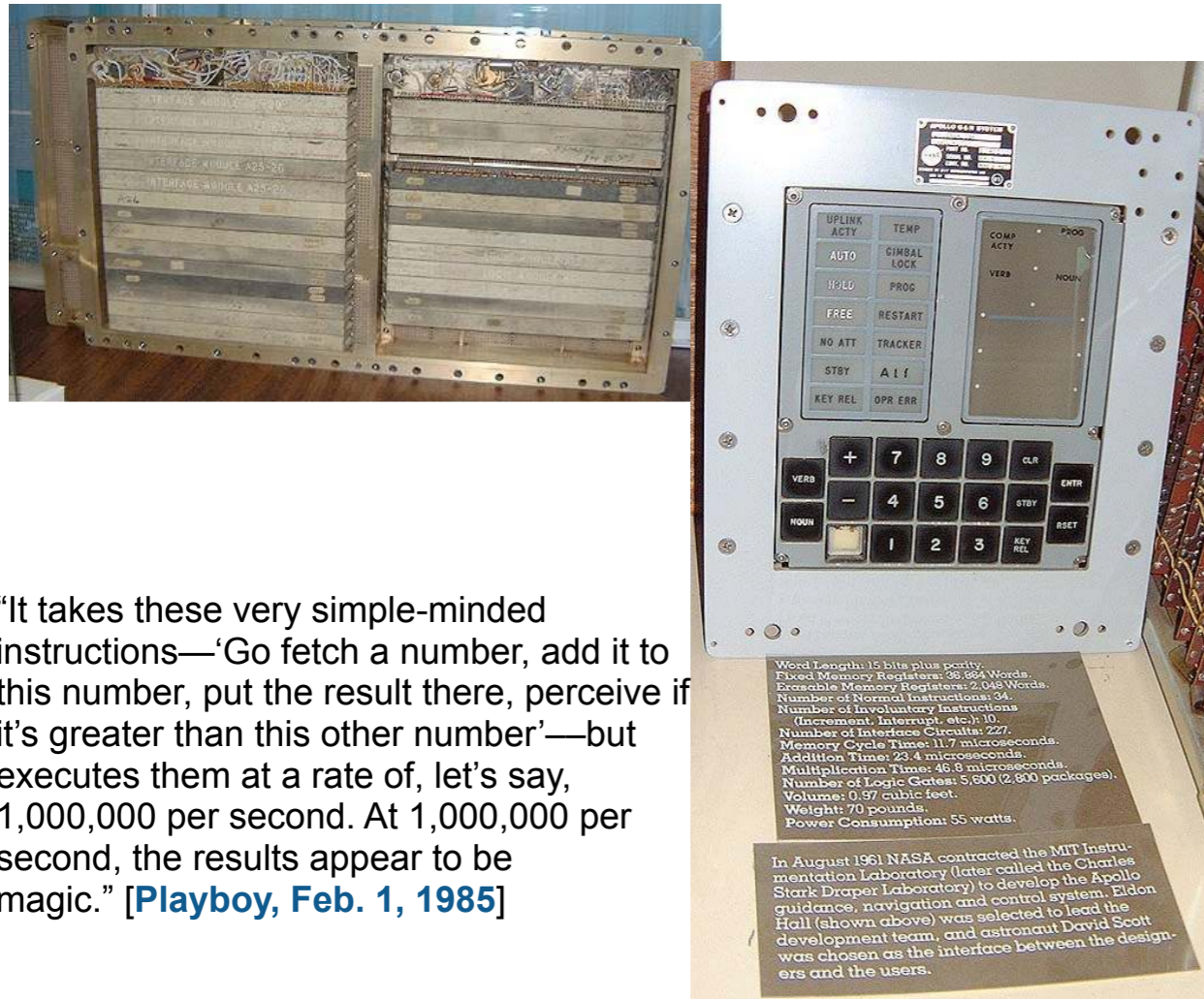


# Speech Recognition



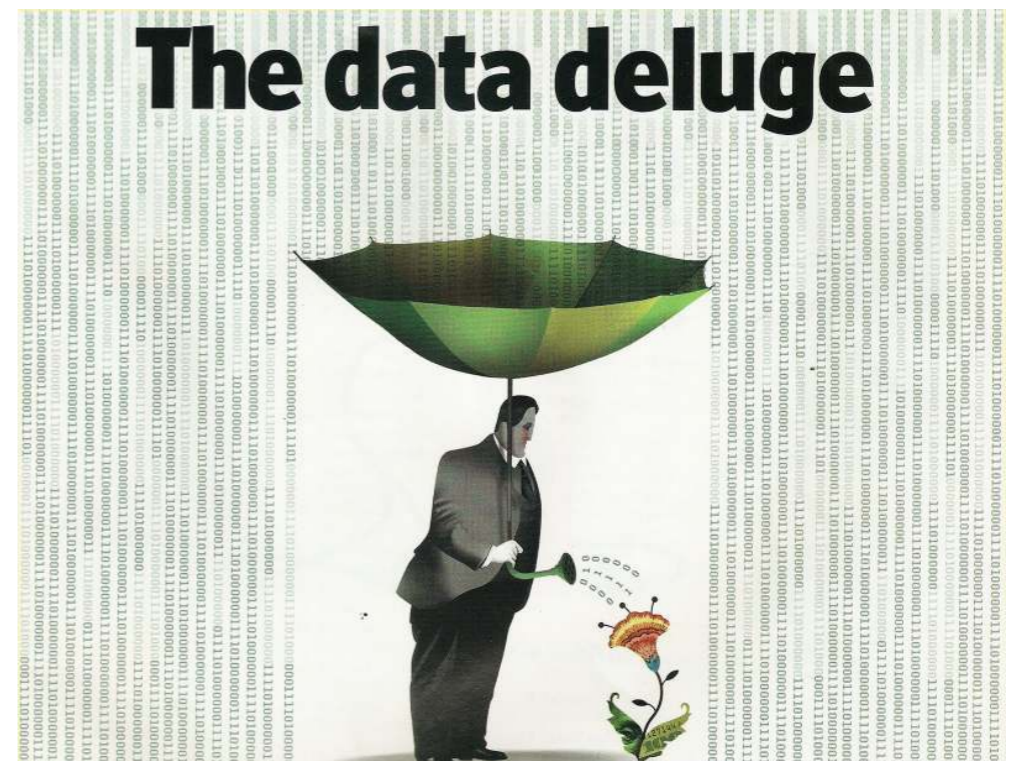
How do we do this???

# Big Data revolution



“It takes these very simple-minded instructions—‘Go fetch a number, add it to this number, put the result there, perceive if it’s greater than this other number’—but executes them at a rate of, let’s say, 1,000,000 per second. At 1,000,000 per second, the results appear to be magic.” [Playboy, Feb. 1, 1985]

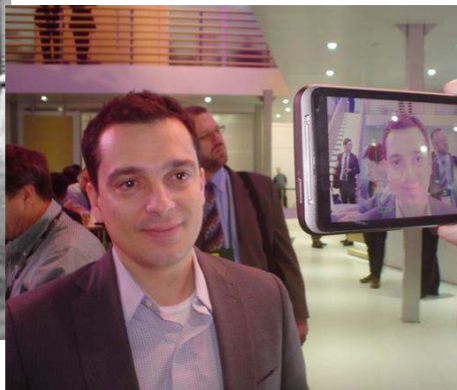
# DATA



# Computers



# +Machine Learning



### Autonomous Driving

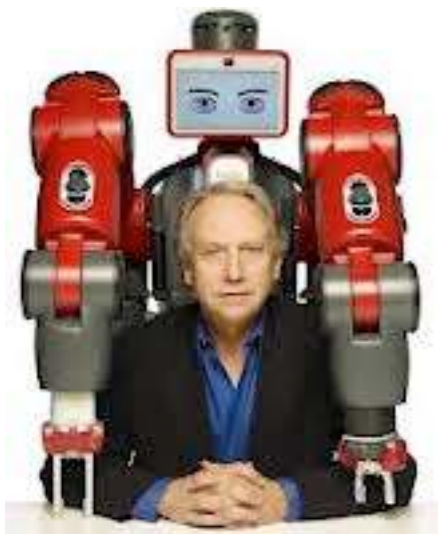
Google's modified Toyota Prius uses an array of sensors to navigate public roads without a human driver. Other components, not shown, include a GPS receiver and an inertial motion sensor.

**LiDAR**  
A rotating sensor on the roof scans more than 200 feet in all directions to generate a precise three-dimensional map of the car's surroundings.

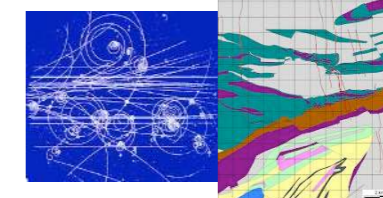
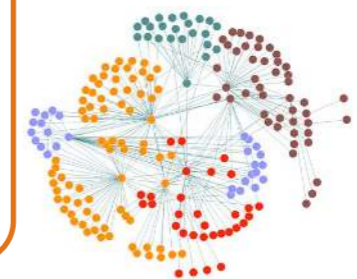
**VIDEO CAMERA**  
A camera mounted near the rear view mirror detects traffic lights and helps the car's onboard computer recognize moving obstacles like pedestrians and bicyclists.

**POSITION ESTIMATOR**  
A sensor mounted on the left rear wheel measures small movements made by the car and helps to accurately locate its position on the map.

**RADAR**  
Four standard automotive radar sensors, three in front and one in the rear, help determine the positions of distant objects.



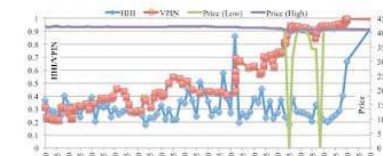
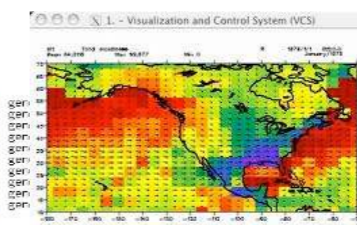
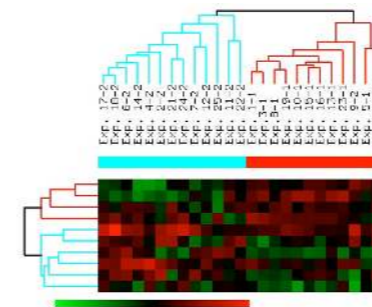
**Machine Learning**  
systems  
*learn from data*  
rather than being  
programmed



**Siri**  
Use your voice to send messages, set reminders, search for information, and more.



- 2007-05-22 18:33 - News in English  
**Soccer: Milan weighs attack options**  
Gilardino tipped to lead the line, Inzaghi late card
- 2007-05-22 17:29 - News in English  
**Wine wards off senile dementia**  
A glass a day stops mild impairment worsening, Italians say
- 2007-05-22 16:27 - News in English  
**Treasury may sell all of Alitalia**  
Formal bids for carrier due by July 2
- 2007-05-22 15:58 - News in English  
**Afghanistan: Rumsfeld on Bush call**  
'We respond to parliament, no one else,' FM says
- 2007-05-22 14:11 - News in English  
**Electronic nose sniffs asthma**  
Device developed by Italian researcher in Netherlands

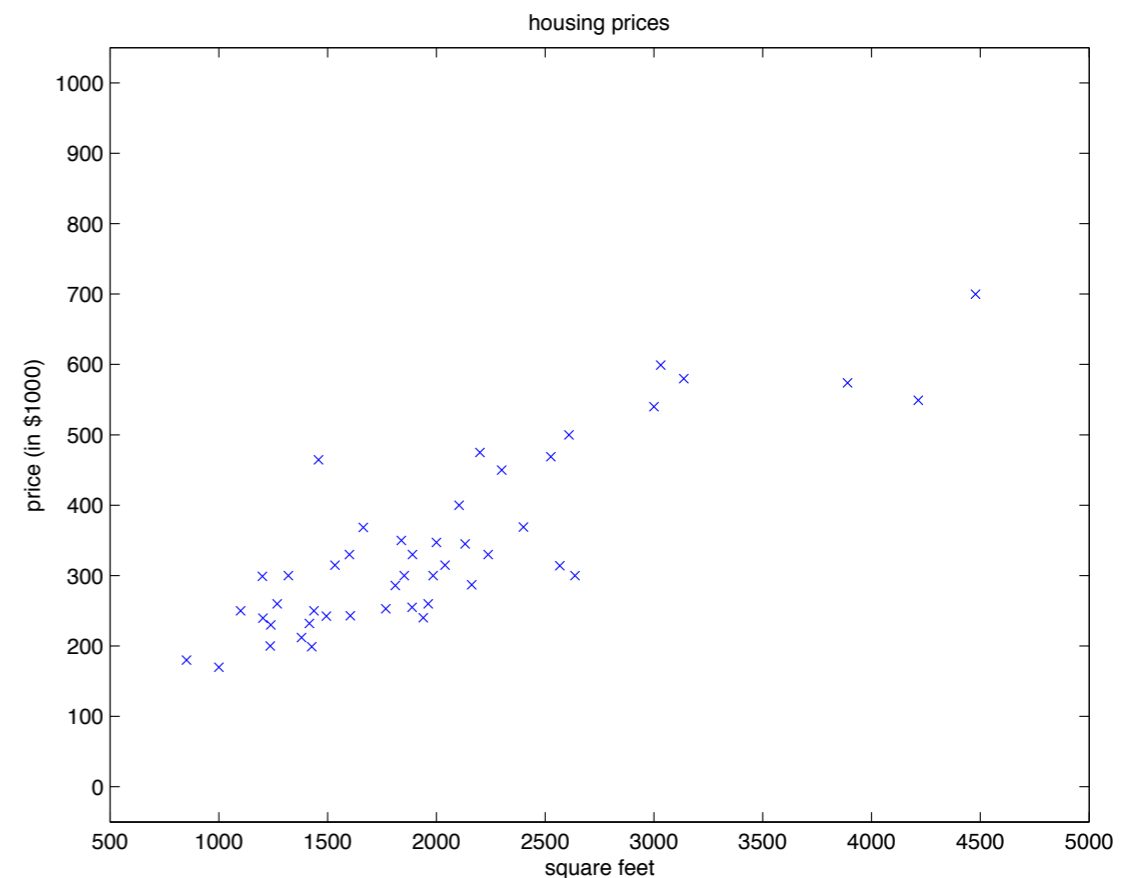


## DATA

Living area (feet <sup>2</sup> )	Price (1000\$)
2104	400
1600	330
2400	369
1416	232
3000	540
⋮	⋮

$$(x_1, y_1), \dots, (x_n, y_n)$$

Living area (feet <sup>2</sup> )	#bedrooms	Price (1000\$)
2104	3	400
1600	3	330
2400	3	369
1416	2	232
3000	4	540
⋮	⋮	⋮





# Text Classification



» 2007-05-22 18:33 - News in English  
 **Soccer: Milan weighs attack options**  
Gilardino tipped to lead the line, Inzaghi late card  
← Sport (football)

» 2007-05-22 17:29 - News in English  
 **Wine wards off senile dementia**  
A glass a day stops mild impairment worsening, Italians say  
← Health

» 2007-05-22 16:27 - News in English  
 **Treasury may sell all of Alitalia**  
Formal bids for carrier due by July 2  
← Politics (economy)

» 2007-05-22 15:58 - News in English  
 **Afghanistan: Rome cold on Bush call**  
'We respnd to parliament, no one else, ' FM says  
← Politics (foreign)

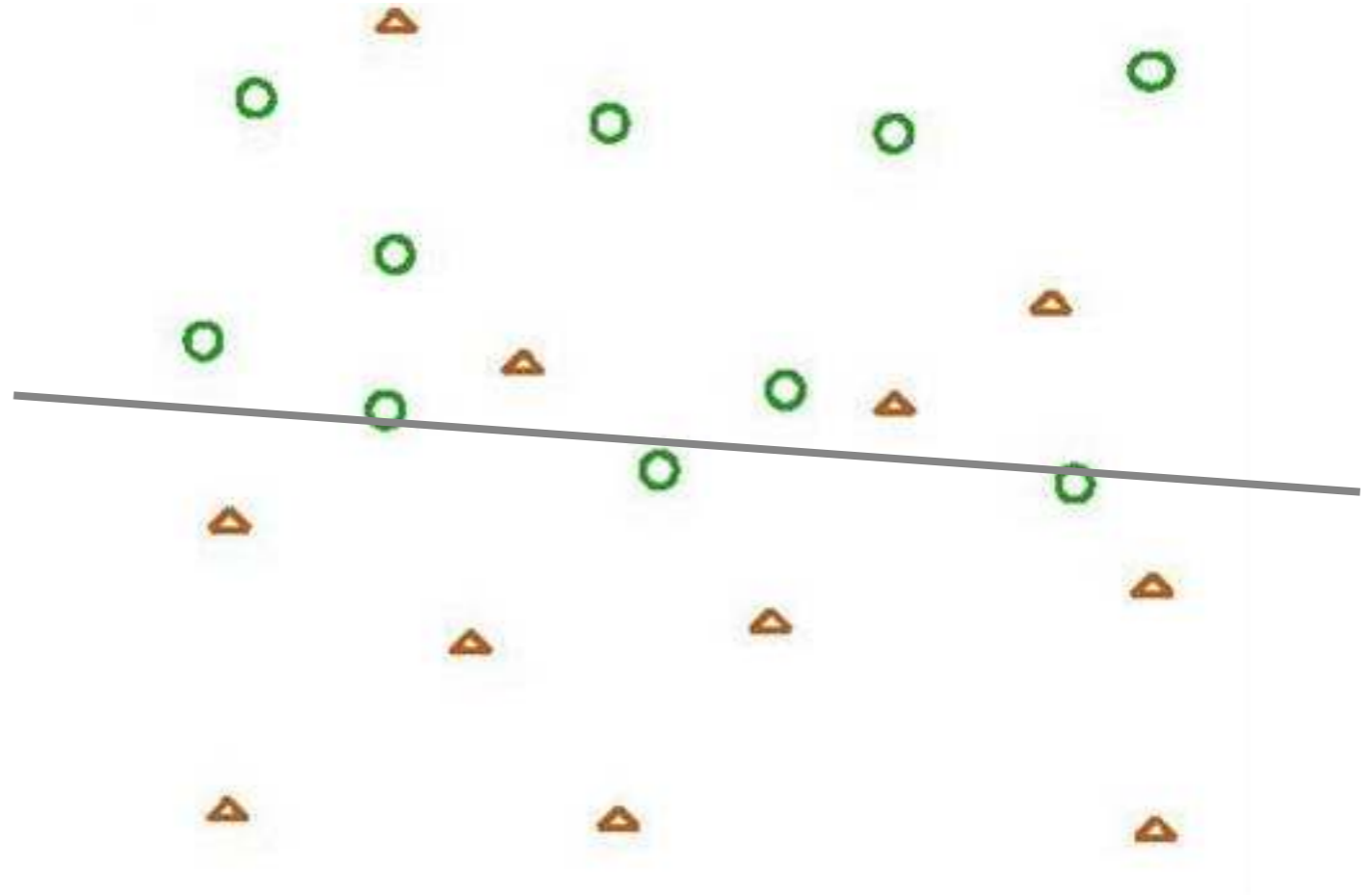
» 2007-05-22 14:11 - News in English  
 **Electronic nose sniffs asthma**  
Device developed by Itaian researcher in Netherlands  
← Technology

Subject	Date	Time	Body	Spam?
I has the viagra for you	03/12/1992	12:23 pm	Hi! I noticed that you are a software engineer so here's the pleasure you were looking for...	Yes
Important business	05/29/1995	01:24 pm	Give me your account number and you'll be rich. I'm totally serial	Yes
Business Plan	05/23/1996	07:19 pm	As per our conversation, here's the business plan for our new venture Warm regards...	No
Job Opportunity	02/29/1998	08:19 am	Hi !! am trying to fill a position for a PHP ...	Yes
[A few thousand rows ommitted]				
Call mom	05/23/2000	02:14 pm	Call mom. She's been trying to reach you for a few days now	No





# Basic Setting: Classification



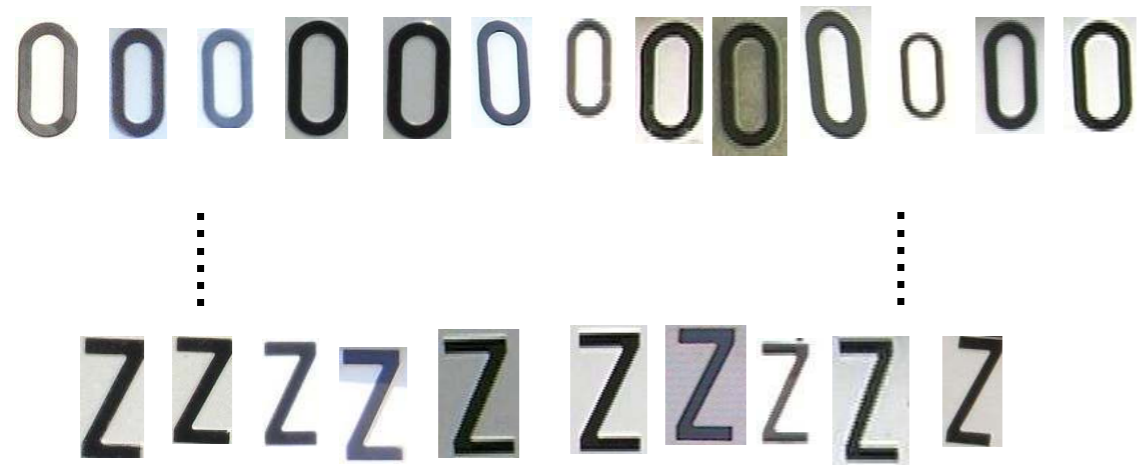
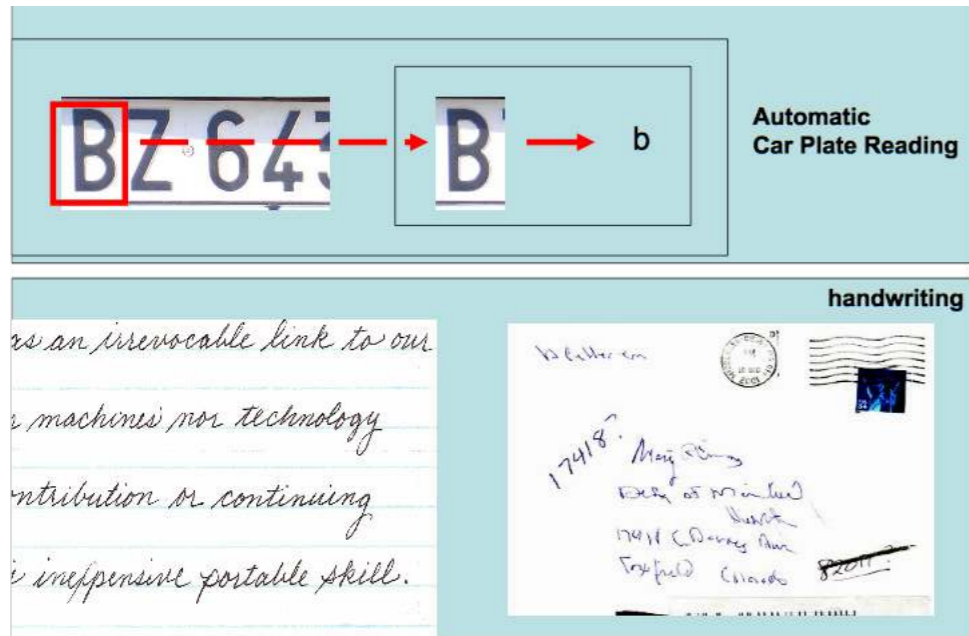
$(x_1, y_1), \dots, (x_n, y_n)$

$$X_n = \begin{pmatrix} x_1^1 & \dots & \dots & \dots & x_1^p \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ x_n^1 & \dots & \dots & \dots & x_n^p \end{pmatrix}$$

$$Y_n = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$



# Image Classification



(a) Sky

(b) Trees



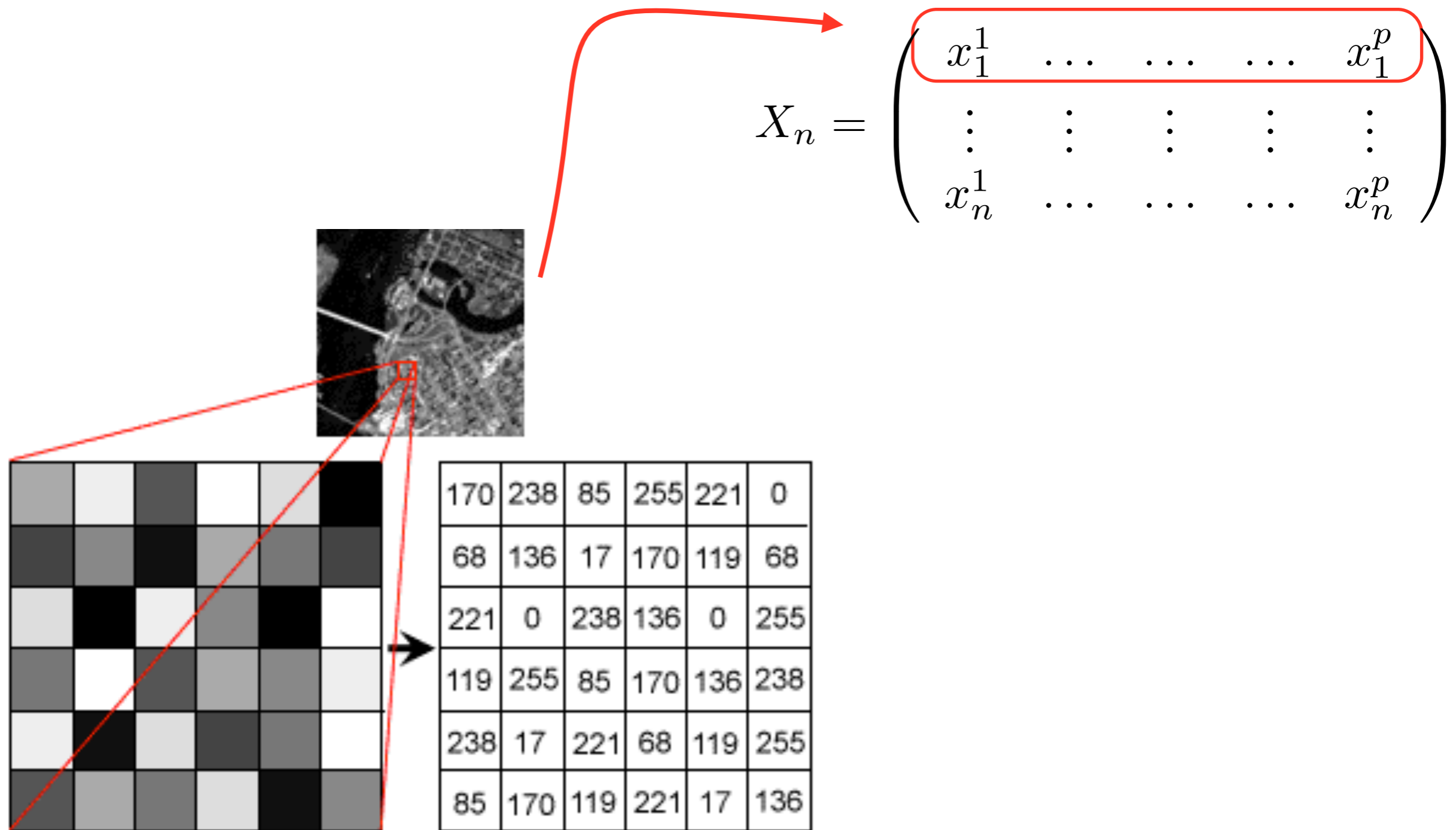
(c) Grass

(d) Snow





# Image Classification



$n$  patients  $p$  gene expression measurements



...



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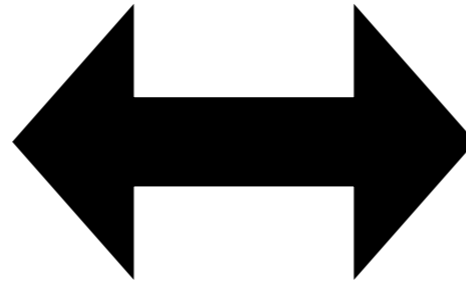
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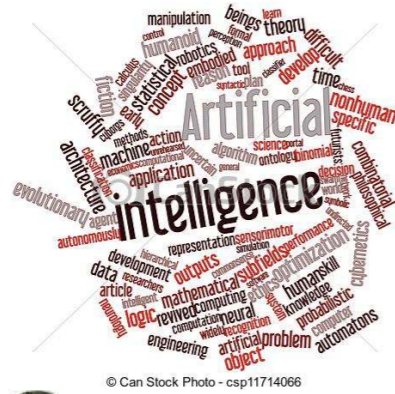
$$X_n = \begin{pmatrix} x_1^1 & \dots & \dots & \dots & x_1^p \\ \vdots & \vdots & \vdots & \vdots & \vdots \\ x_n^1 & \dots & \dots & \dots & x_n^p \end{pmatrix}; Y_n = \begin{pmatrix} y_1 \\ \vdots \\ y_n \end{pmatrix}$$

# Machine Learning

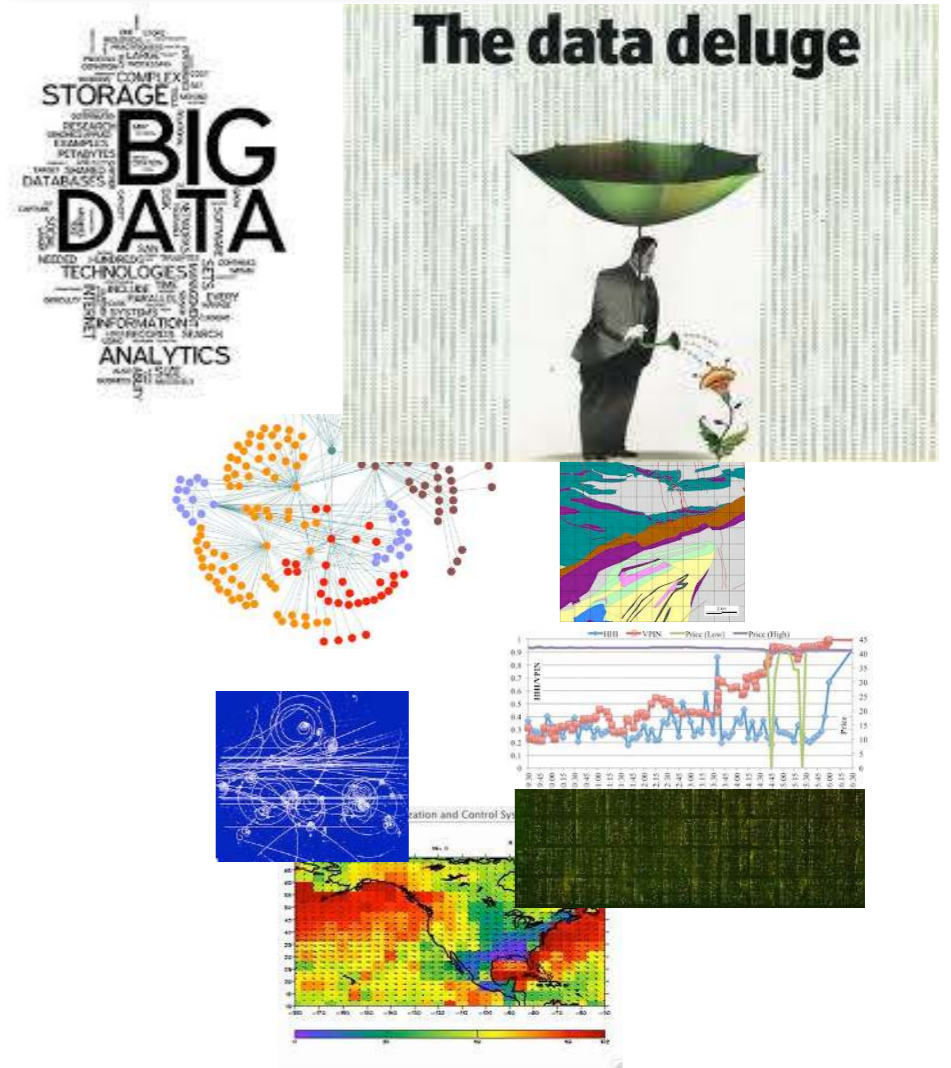
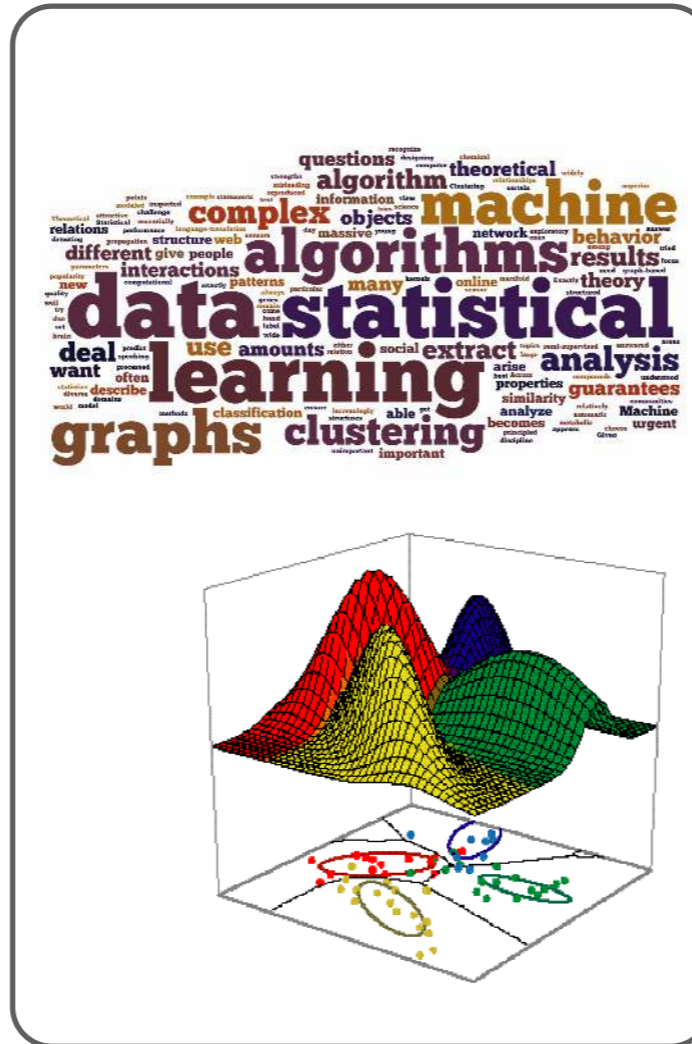
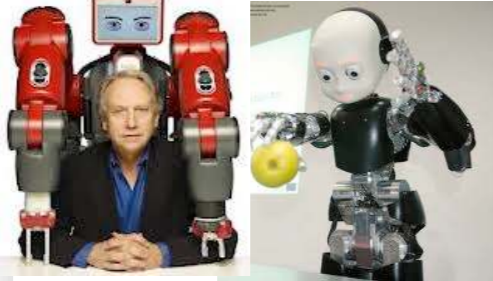
## Intelligent Systems



## Data Science



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- A quick tour of machine learning
- Basic statistical learning theory



- Local algorithms
- Model selection