

# Cours Systèmes Intelligents (Part: Computer Vision)

Corentin Domken

Contact: [corentin.domken@henallux.be](mailto:corentin.domken@henallux.be)

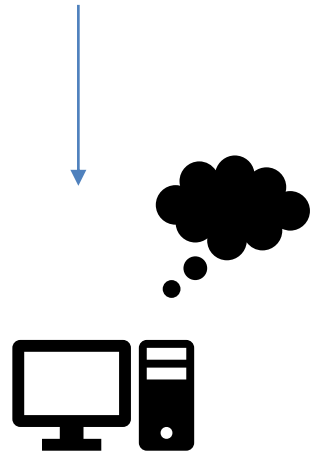


# Introduction

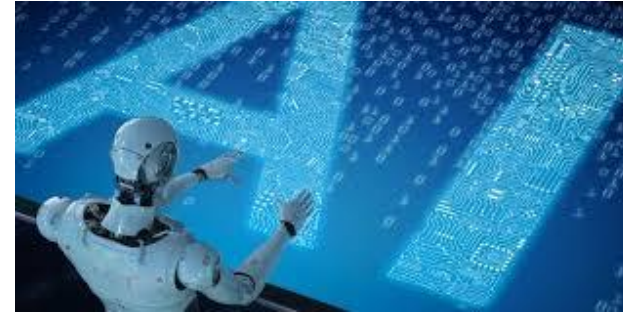
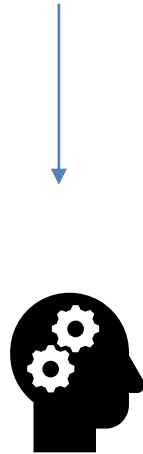
# What is Artificial Intelligence ?

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Artificial Intelligence



Human intelligence



- Making computers that think?
- The automation of activities we associate with human thinking, like decision making, learning ... ?

# What is Computer Vision?

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- Make computers understand images and videos.



What kind of scene?

Where are the cars?

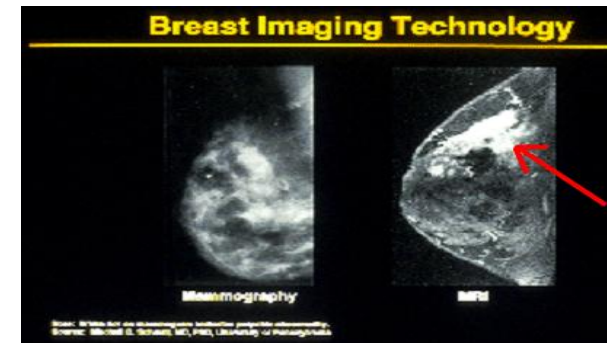
How far is the building?

...

# Computer Vision

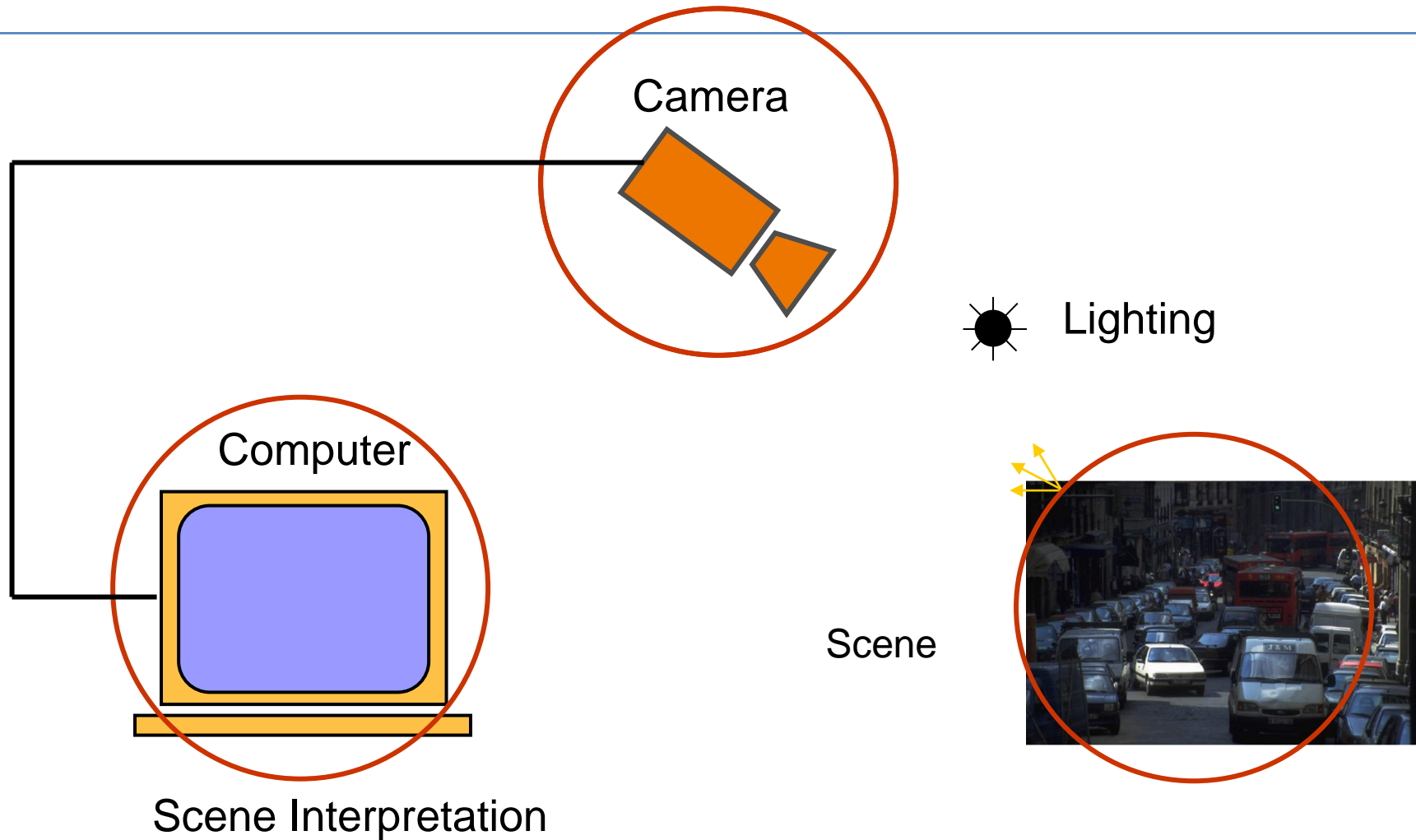
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- **Computer vision** is the science and technology of machines that **see**
- Concerned with the theory for building artificial systems that obtain information from images.
- The image **data** can take **many forms**, such as a **video** sequence, **depth images**, views from **multiple cameras**, or multi-dimensional data from a medical scanner



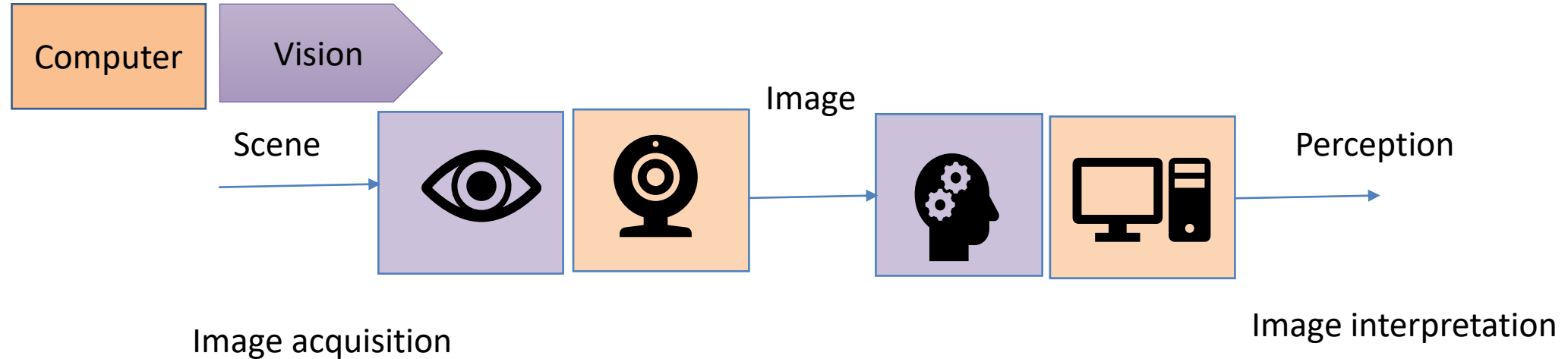
# Components of a computer vision system

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# Computer vision vs human vision

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# Computer vision vs human vision



What we see

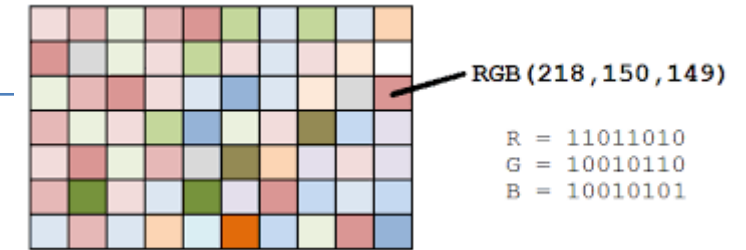
0	3	2	5	4	7	6	9	8
3	0	1	2	3	4	5	6	7
2	1	0	3	2	5	4	7	6
5	2	3	0	1	2	3	4	5
4	3	2	1	0	3	2	5	4
7	4	5	2	3	0	1	2	3
6	5	4	3	2	1	0	3	2
9	6	7	4	5	2	3	0	1
8	7	6	5	4	3	2	1	0



What a computer sees



# Computer vision vs human vision



- Take a **picture** with a **camera**, it is just a bunch of colored dots (**pixels**)
- Given an image or more, **extract properties** of the 3D world



-Traffic scene

- Number of vehicles
- Type of vehicles
- Location of closest obstacle
- Assessment of congestion
- Location of the scene captured
- ...



Need Machine Learning

# What is machine learning

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- ML is a **subset of the larger field of artificial intelligence (AI)** that “focuses on teaching computers how to learn without the need to be programmed for specific tasks,”
- **Automatically** learn to recognize complex patterns and make intelligent **decisions** based on data

# Computer Vision vs. Machine Learning

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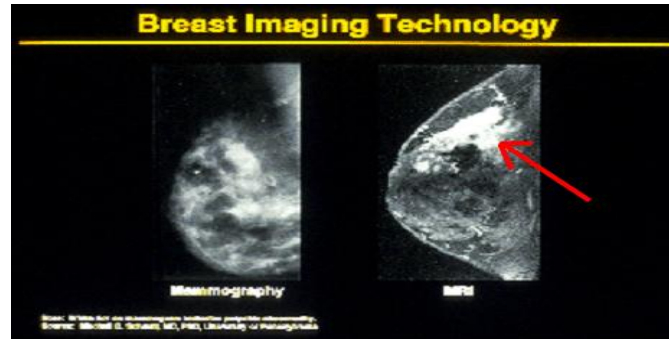
- In Machine Learning, it usually **does not care about how to obtain the data or sensors**
- Machine Learning is very useful for Computer Vision (e.g., learning for vision)

# Why computer vision matters

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Agriculture



Health



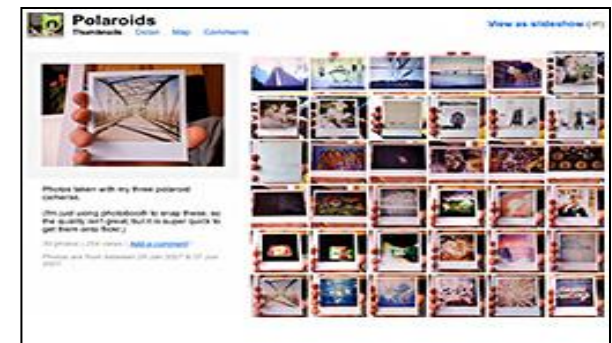
Security



Comfort



Fun

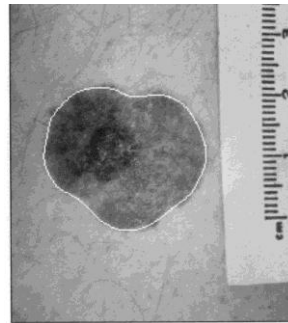


Access

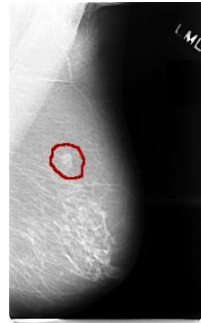
# Applications

demodulation  
demodulation

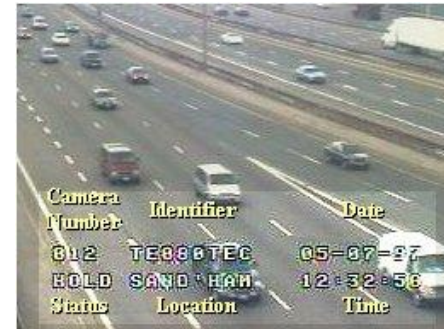
Character Recognition



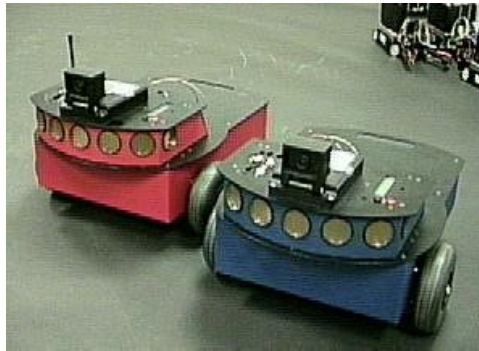
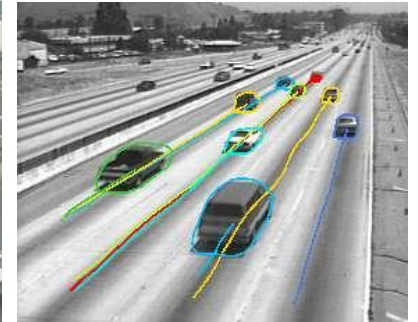
skin cancer



breast cancer



Traffic Monitoring



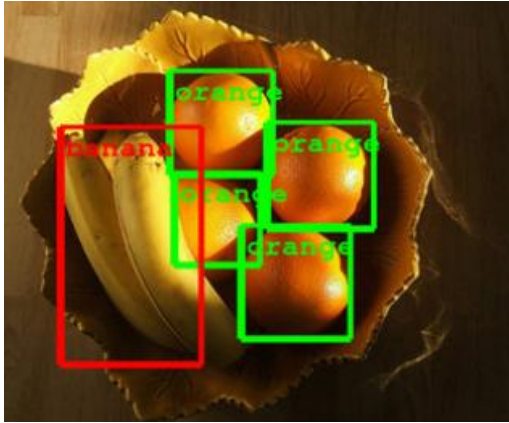
Autonomous Vehicles





# Applications

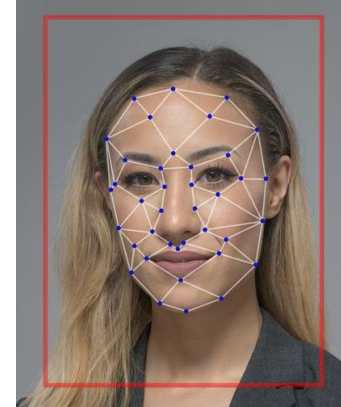
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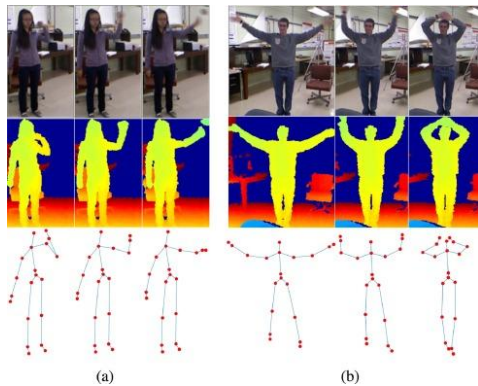
Object recognition



Face detection



Face recognition



Action recognition



Hand Gesture Recognition



Expression Recognition

# Basics on Computer vision



# Computer Vision is more than just learning!

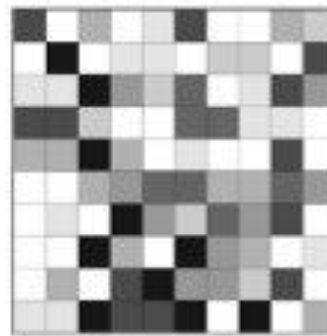
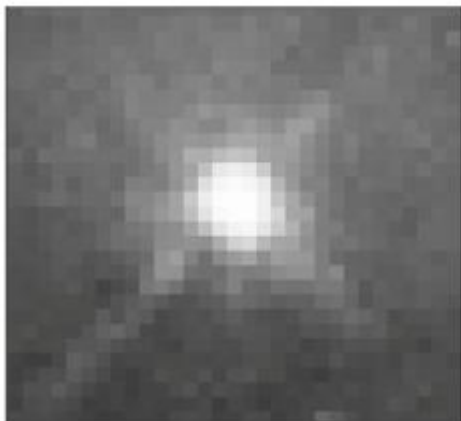
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- In Computer Vision, we care **how to obtain the visual data** (sensor design, active vision), **how to represent the visual data**, and others

# What is image?

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- An **image** is an array, or a **matrix**, of square pixels (picture elements) arranged in columns and rows
- For a grayscale image (Black and White), each pixel typically consists of 8 bits (1 byte)
- 8 bits represents  $2^8 = 256$  tonal levels (0-255)
- Each pixel has an assigned intensity that ranges from 0 to 255



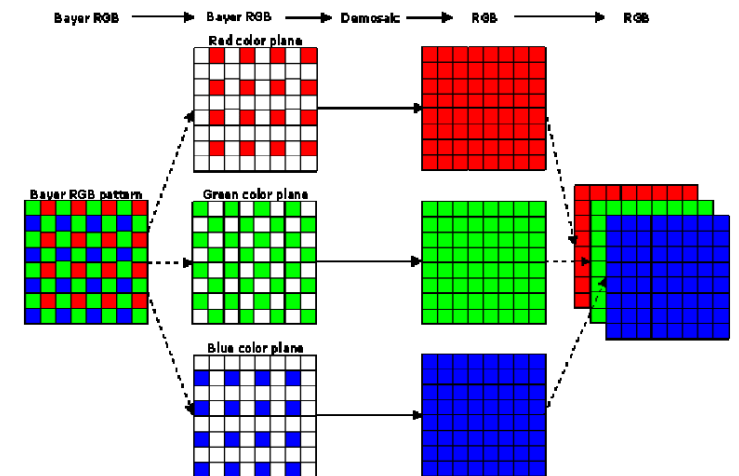
254	107
255	165

# What is image?

- A “true colour” image has 24 bits (3 bytes e.g., RGB)
- Each pixel is represented has 24 bits to encode color information =  $8 \times 8 \times 8$  bits =  $256 \times 256 \times 256$  colours = ~16 million colours



Consumer camera sensor



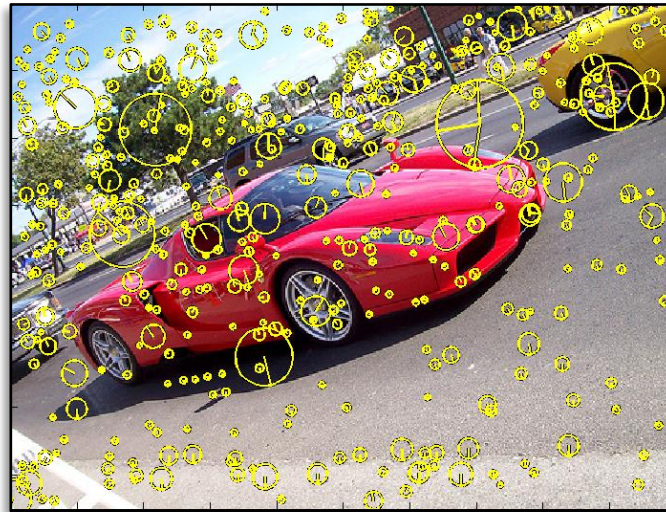
# Local Visual Features

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Images are high dimensional!

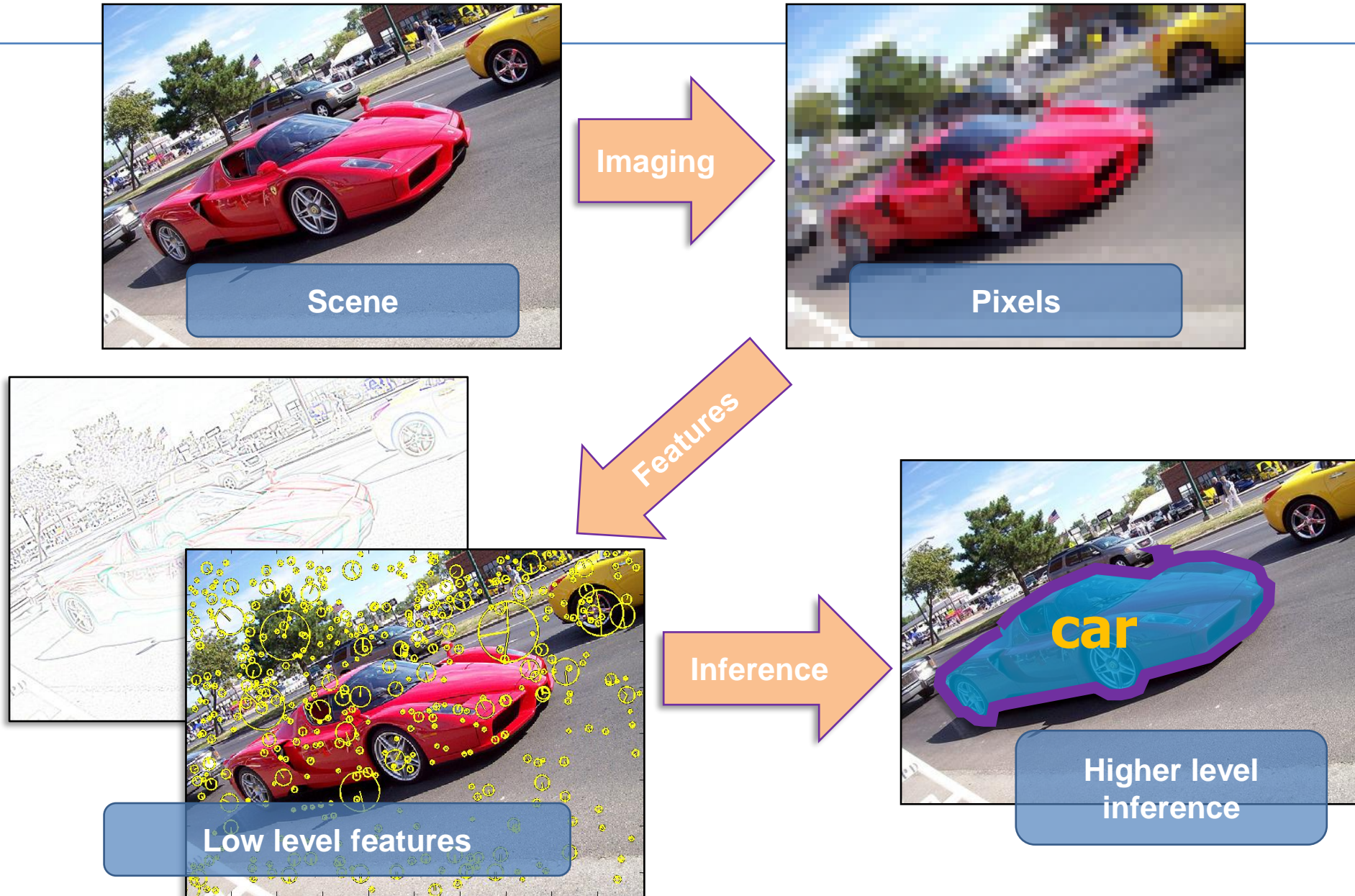
$$(640 \text{ width}) * (480 \text{ height}) = (307200 \text{ pixels})$$

Compute image statistics in a region (e.g., estimate the distribution of image gradient orientations)





# From Pixels to Visual features



# Some famous features

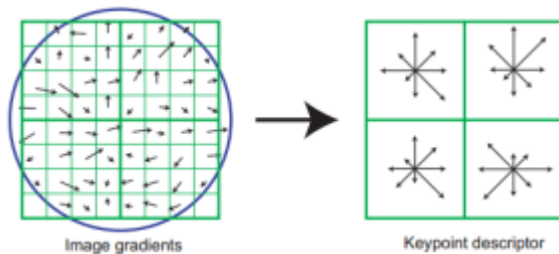
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- SIFT: Scale Invariant Feature Transform
- HOG: **Histogram of oriented gradients**
- LBP: Local binary pattern

# SIFT

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- **scale-invariant feature transform (SIFT)** is a [feature detection](#) algorithm in [computer vision](#) to detect and describe local features in images
- Published by David G. Lowe in 1999
- Invariant to scaling, rotation and translation
- Transforms an image into a large collection of local feature vectors (local descriptors called SIFT keys)



Object recognition



# HOG

- The **histogram of oriented gradients (HOG)** is a [feature descriptor](#) used in [computer vision](#) and [image processing](#) for the purpose of [object detection](#). The technique counts occurrences of gradient orientation in localized portions of an image
- Intuitively it tries to capture the shape of structures in the region by capturing information about gradients.

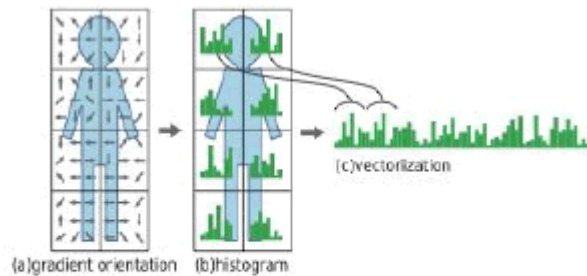


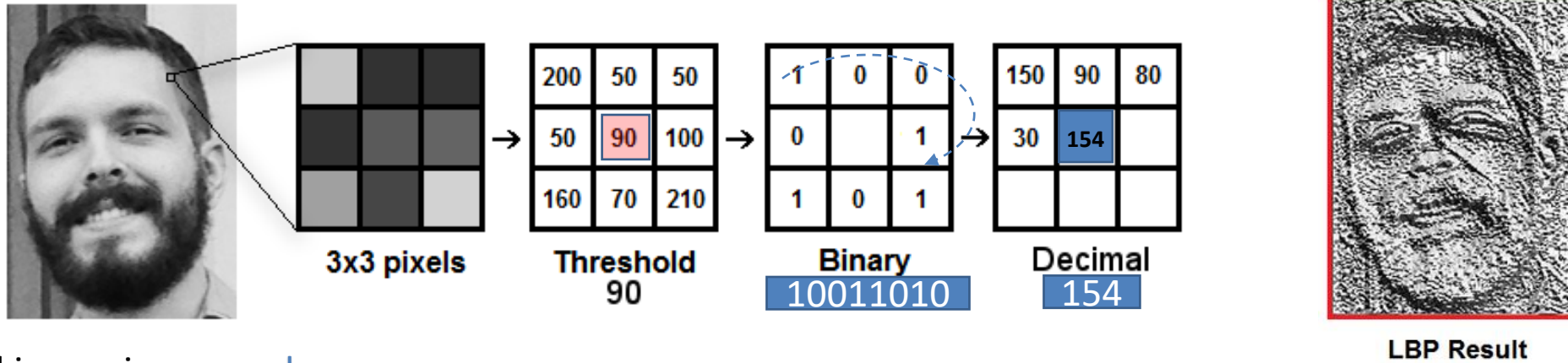
Figure 3.5 (a) (b) Overview of HOG calculation

Person following with a mobile robot



# LBP

## Local Binary Pattern

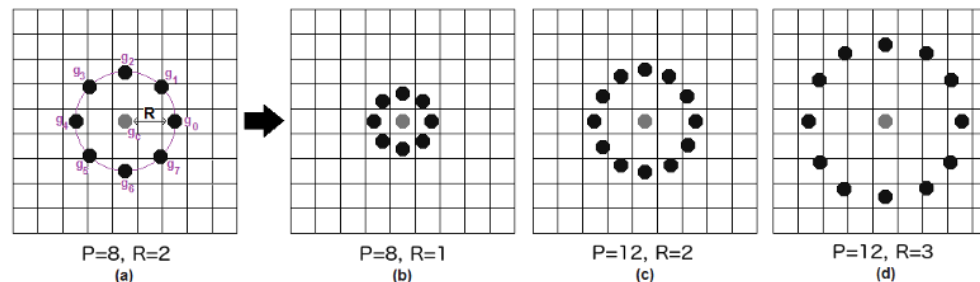


Facial image in **grayscale**

$$10011010 = (1 \times 2^7) + (0 \times 2^6) + (0 \times 2^5) + (1 \times 2^4) + (1 \times 2^3) + (0 \times 2^2) + (1 \times 2^1) + (0 \times 2^0) = 154$$


The intensity of each pixel (0~255)

Different number of radius and neighbors



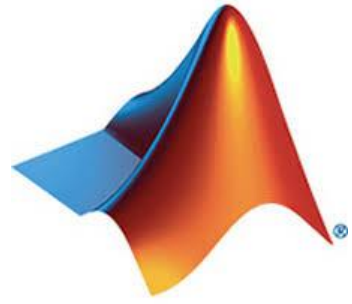
# Programming tools

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• [OpenCV](#) is a great performing computer vision tool and it works well with [C++](#) as well as [Python](#)

• OpenCV is prebuilt with all the necessary techniques and algorithms to perform several

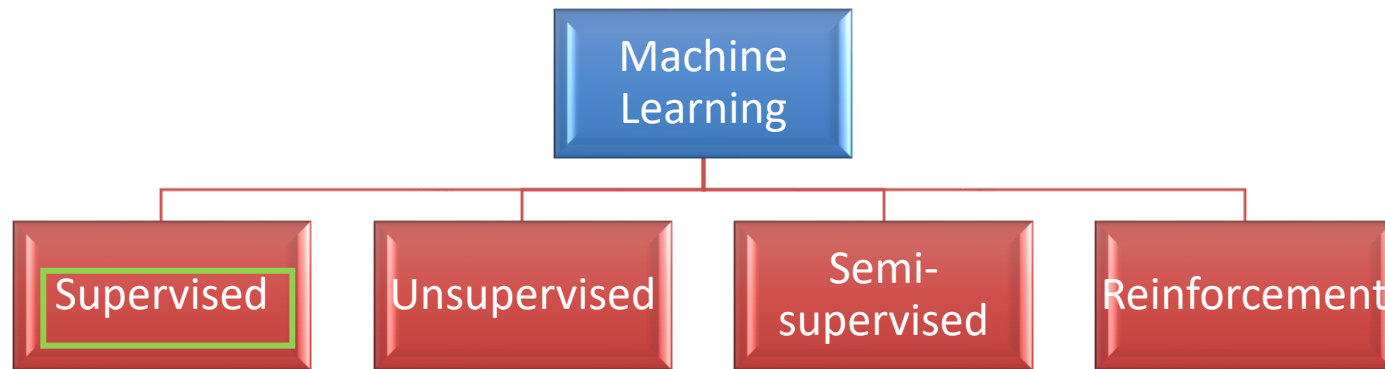


• [Matlab](#) is a great tool for creating image processing applications and is widely used in research

# Basics on Machine Learning

# Types of machine learning

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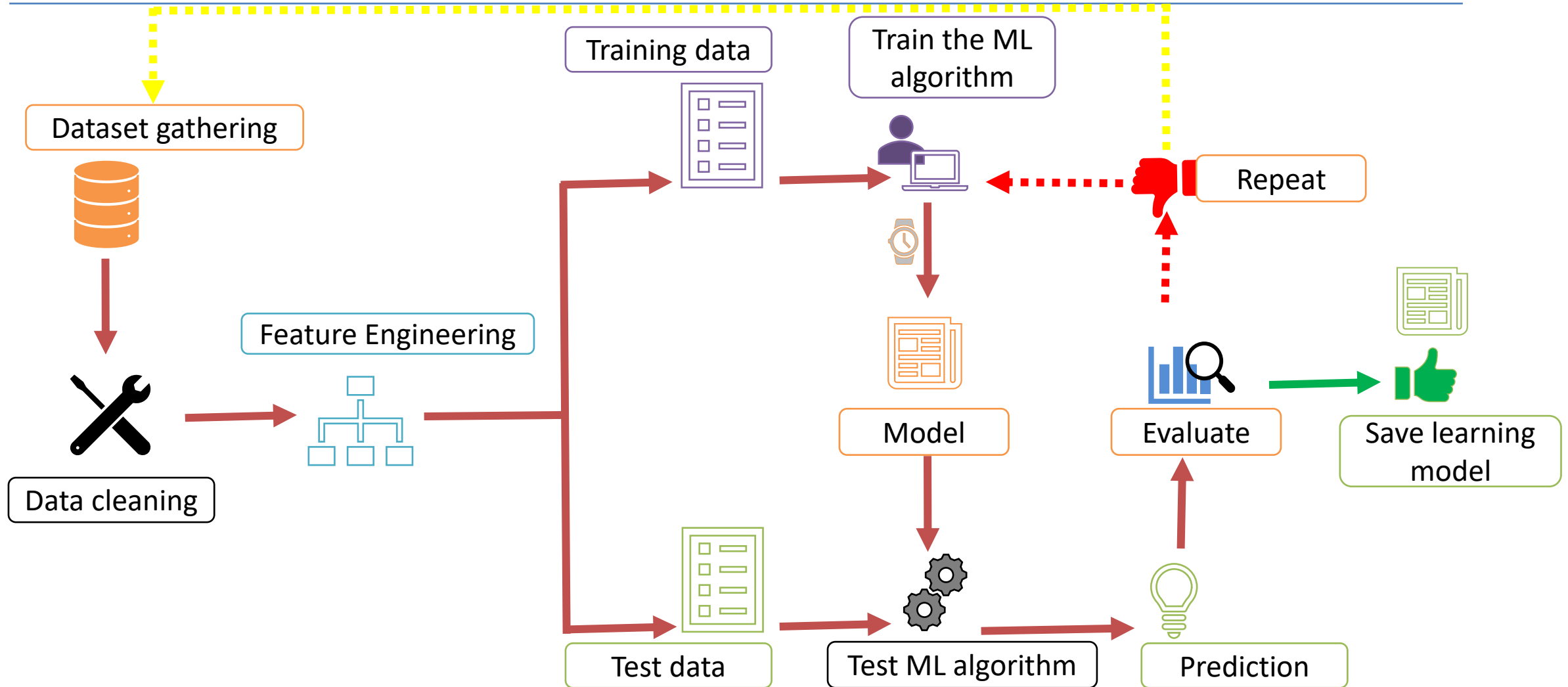


Task driven  
(regression/**classification**)

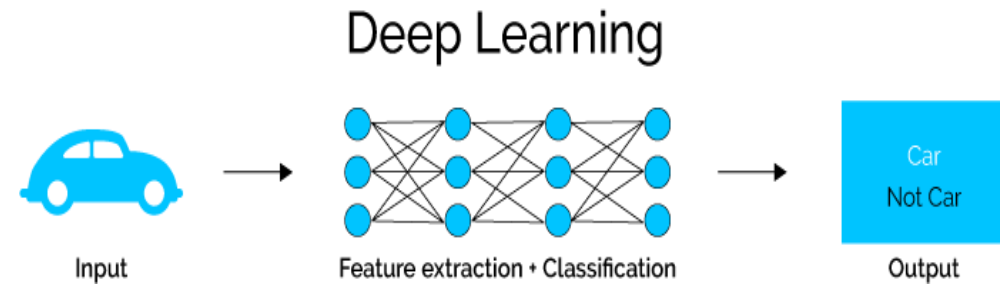
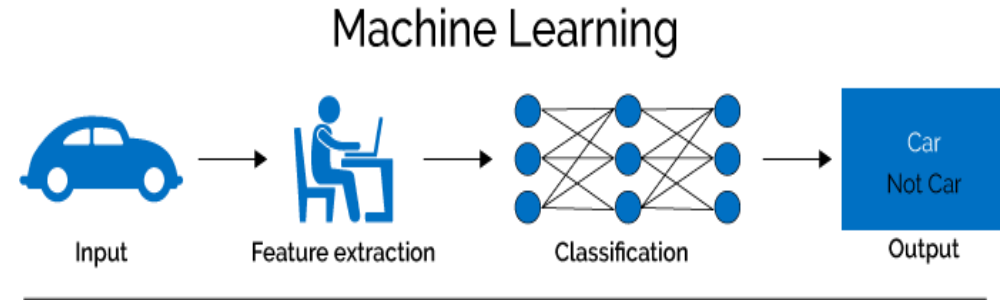
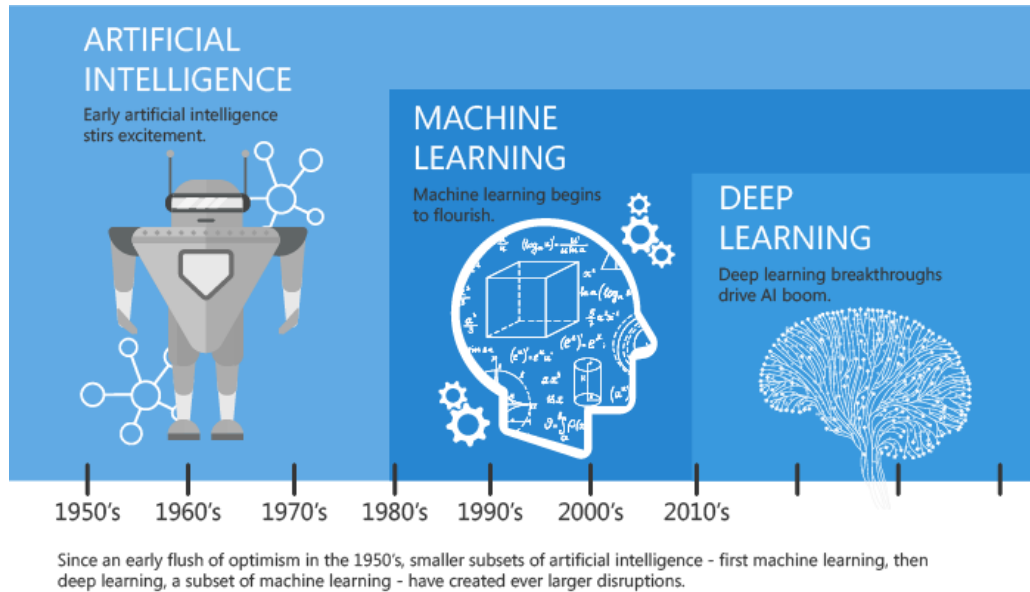
Data driven  
(Clustering)

Algorithm learns to react  
to environment

# Machine learning process



# Machine Learning VS Deep Learning





# Questions

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# References

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- <https://towardsdatascience.com/face-recognition-how-lbph-works-90ec258c3d6b>
- [https://docs.opencv.org/3.4.3/da/d60/tutorial face main.html](https://docs.opencv.org/3.4.3/da/d60/tutorial_face_main.html)
- <https://www.pyimagesearch.com/2015/12/07/local-binary-patterns-with-python-opencv/>
- <https://www.superdatascience.com/opencv-face-detection/>

# Object recognition on Raspberry

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- Tutorial on:  
<https://www.framboise314.fr/i-a-realisez-un-systeme-de-reconnaissance-dobjets-avec-raspberry-pi/>
- Use neural networks

# Face recognition project

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- Face recognition
- <https://github.com/informramiz/opencv-face-recognition-python/blob/master/README.md>
- Digit recognition
- <https://www.learnopencv.com/handwritten-digits-classification-an-opencv-c-python-tutorial/>
- Face recognition: HOG+KNN
- <https://github.com/emersonlaurentino/hog-face-recognition/blob/master/knn.py>
- <https://github.com/irfanhanif/FaceRecognition-HOG-PCA-SVM/blob/master/experiment.py>
- <https://github.com/vj18/HOG-SVM-Classfier/blob/master/makeSvmModel.py> ok
- [https://github.com/gsg213/Face-Recognition-using-LBP/blob/master/take\\_data\\_set.py](https://github.com/gsg213/Face-Recognition-using-LBP/blob/master/take_data_set.py)