

Introduction to Computer Vision for Robotics

AE640A Autonomous Navigation

5th March, 2019



Lecture Outline

- Introduction
 - What is CV?
 - Overview of the field
 - A look at history
 - Hard Problem?
- Human Vision System & the Machine
 - The human vision system
 - Fooling humans
 - The computer vision system
- Images as matrices.
 - How cameras work to produce these matrices
 - Meaning of Intensity, Color etc
 - Shoutout to Image Processing



Lecture Outline

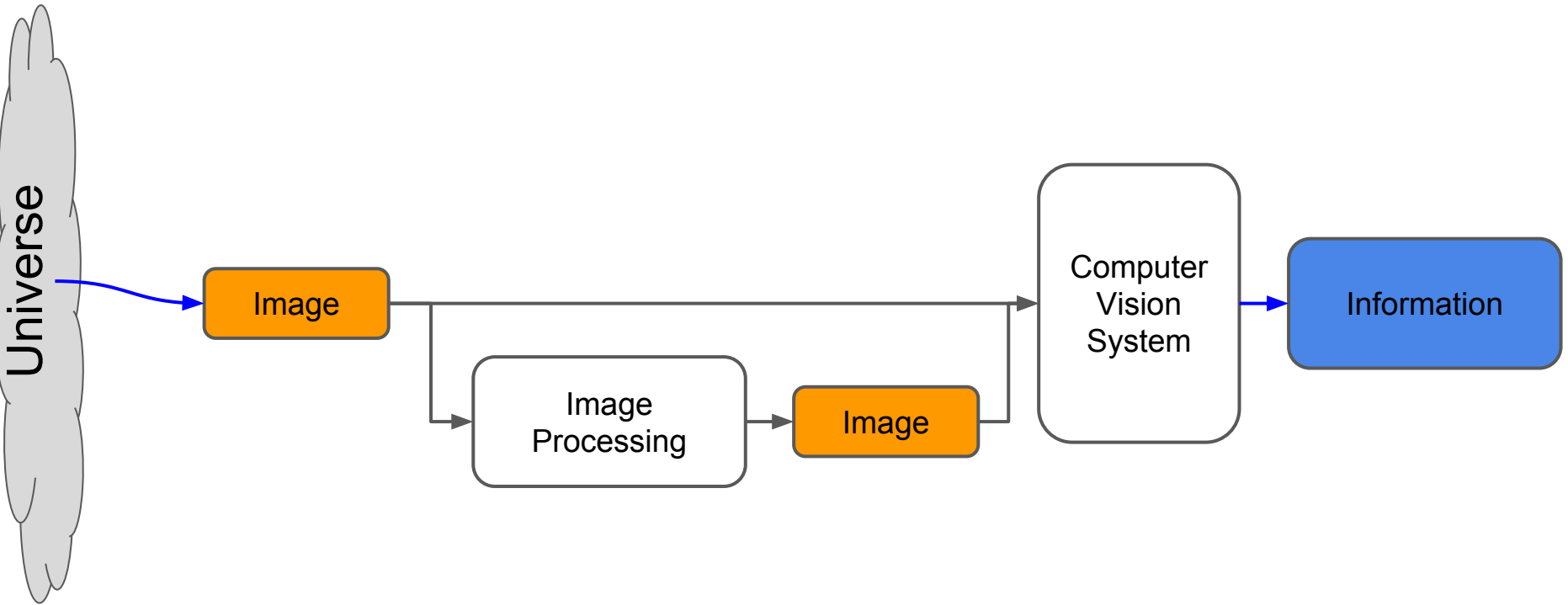
- Camera Model
 - Pinhole Camera Model
 - Intrinsic Camera Matrix
 - Camera Calibration



Introduction



What is Computer Vision?



What is Computer Vision?

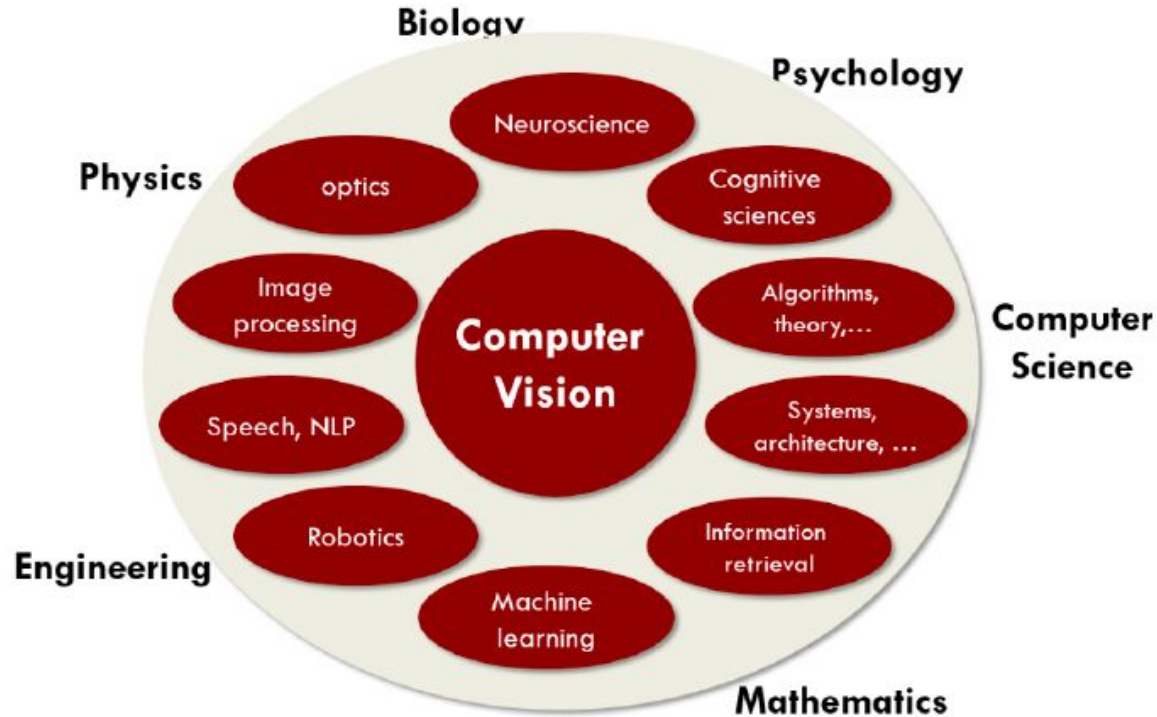


Image Credits: CS131, Fall '18, Stanford



What is Computer Vision?

- Computer Vision is deals with extracting information regarding the 3D world we live in using a single or a bunch of images.
- Computer Vision like most other fields today, is at the junction of numerous disciplines from Biology to Computer Science and has applications only limited by our imagination.



Overview of the field



Image Credits: XKCD, 1425, [2014](#)

Overview of the field

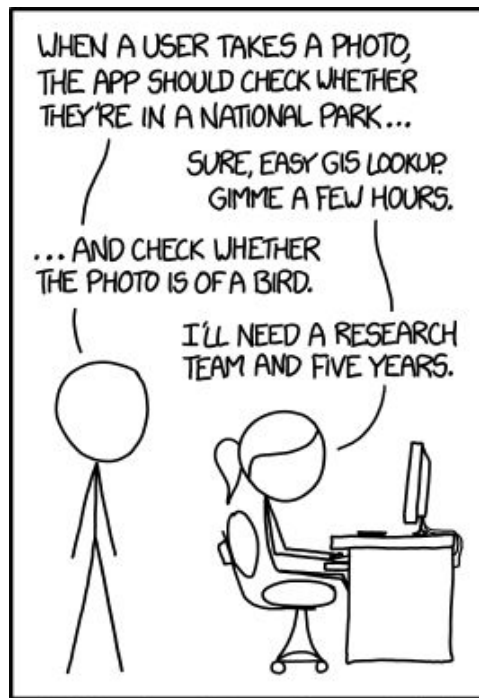
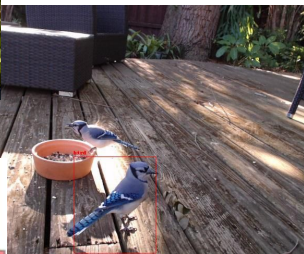
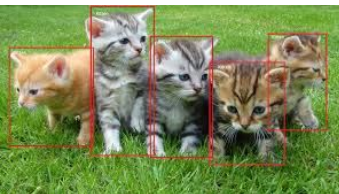
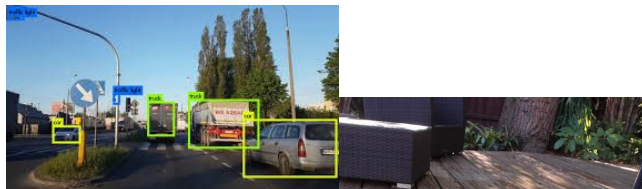


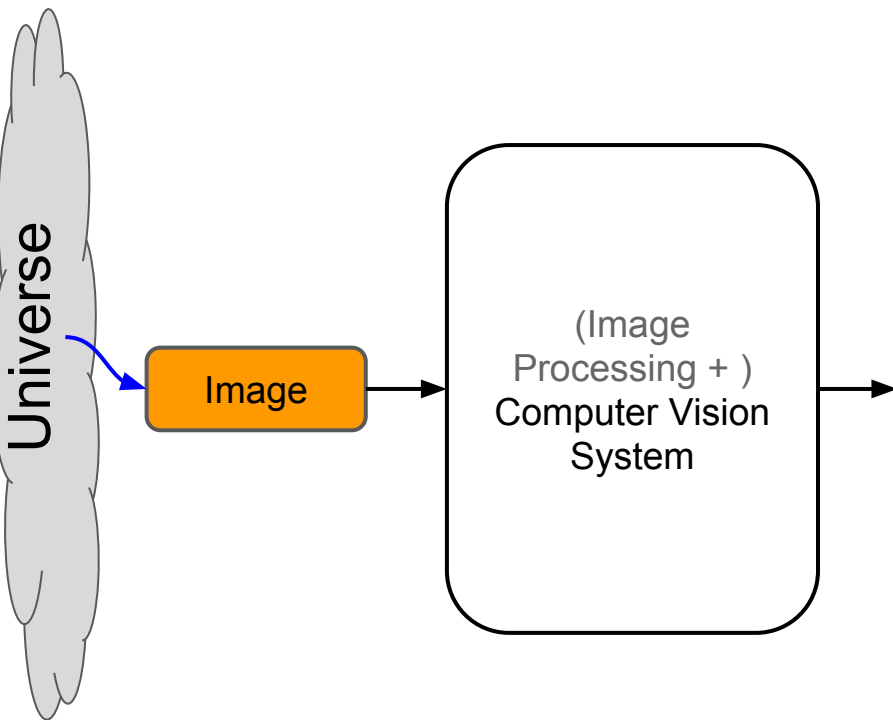
Image Credits: <https://tinyurl.com/y53by9pr>

Image Credits: XKCD, 1425, 2014

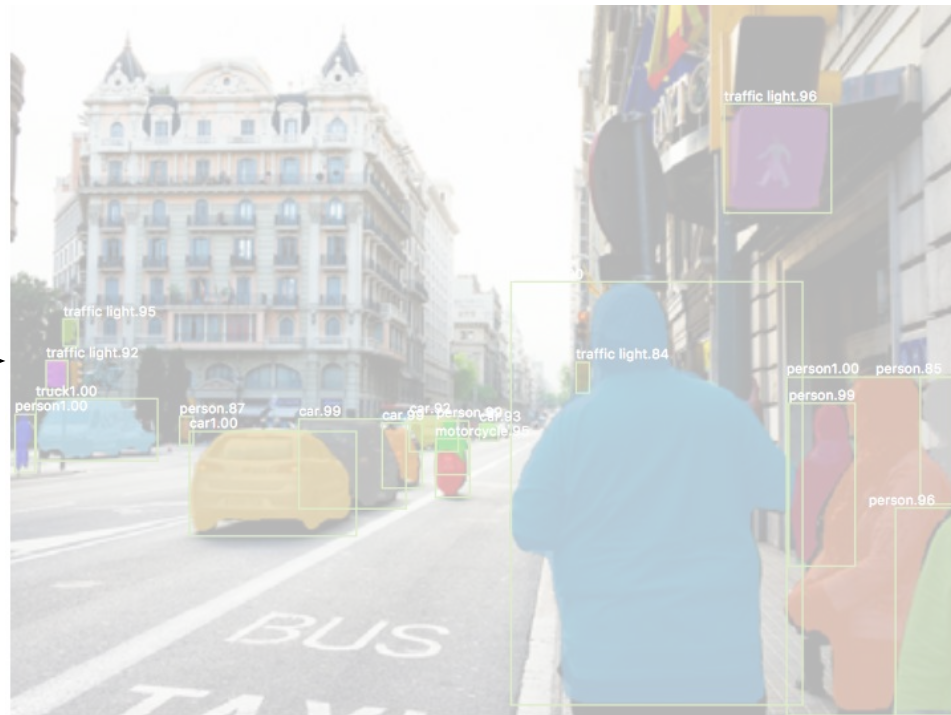
Image Credits: <https://tinyurl.com/y53by9pr>



Overview of the field



What kind of Information?



Overview of the field

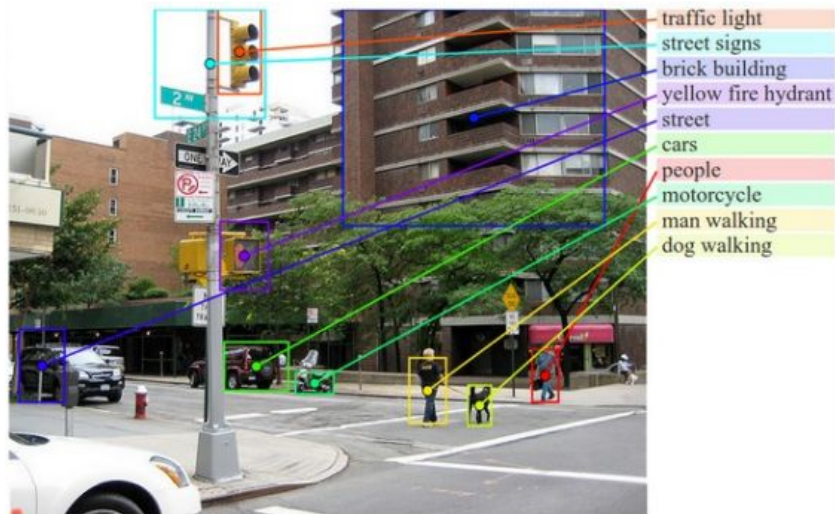


Image Credits: Karpathy, CVPR'15

What kind of Information?

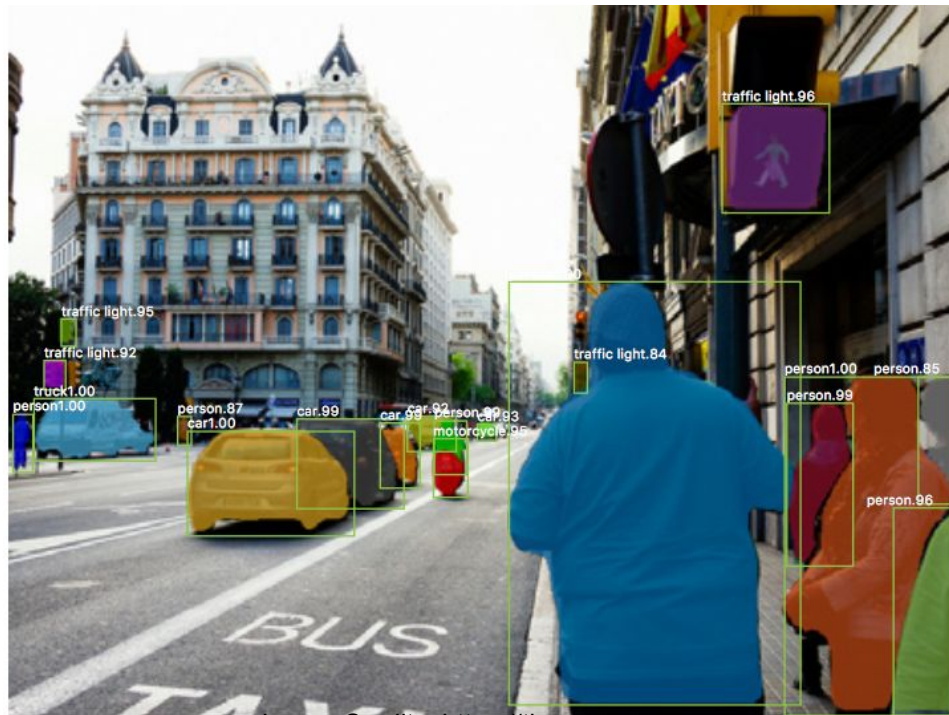


Image Credits: <https://tinyurl.com/lxuex6o>



Overview of the field

Primary themes in Computer Vision are:

1. Object Detection



Recognition: Cat?

Image: <https://tinyurl.com/yanp2o5e>

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection



Recognition: Cat?

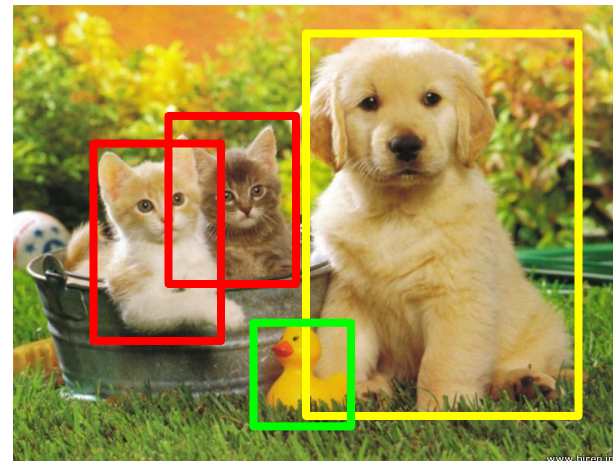
Localization: Where is the cat?

Image: <https://tinyurl.com/yanp2o5e>

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection



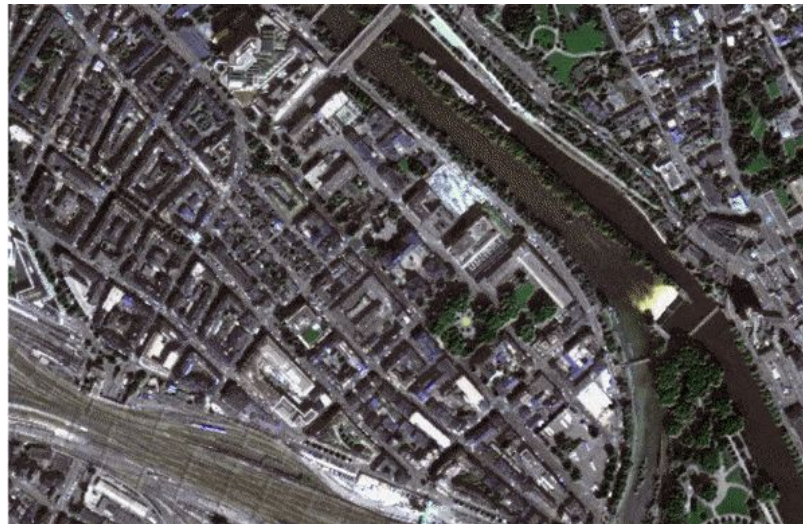
Object Detection: Which Objects
are here and where?

Image: <https://tinyurl.com/y4ly96rd>

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation



Segmentation: Which pixels belong to which object?

Credits: Own Work

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements

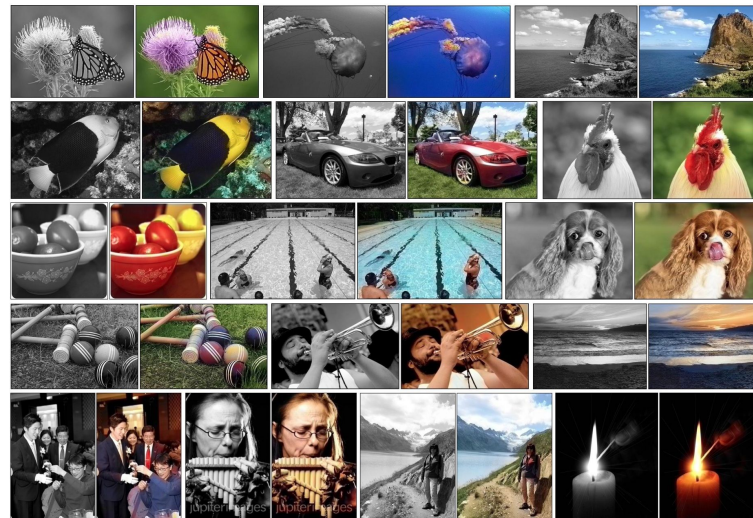


Image Colorization: From
Grayscale to Colored Images

Credits: Richard Zhang, CVPR 2016



Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements



Image Enhancement: Real Time
Image Enhancement

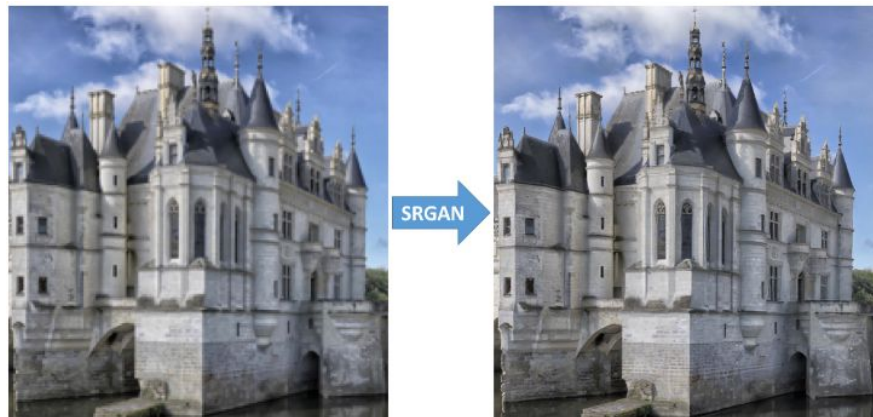
Credits: Michael Gharbi, ACM Graphics 2017



Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements



Super Resolution: Upsampling
Images while preserving quality

Credits: <https://github.com/tensorlayer/srgan>



Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements
4. Image to Text

Dataset of images and sentence descriptions

training image

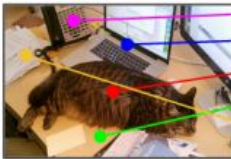


"A Tabby cat is leaning on a wooden table, with one paw on a laser mouse and the other on a black laptop"



Generative model

test image



"office telephone"

"shiny laptop"

"Tabby cat is sleeping"

"wooden office desk"

"messy pile of documents"

Image Description: Automatic semantic description for images

Credits: Karpathy, CVPR 2015

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements
4. Image to Text
5. Image Generation



Image Generation: A style based generator architecture for GANs

Credits: Tero Karras, arXiv 2018

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements
4. Image to Text
5. Image Generation
6. Motion Estimation



Optical Flow: Lucas Kanade
method for motion estimation

Credits: <https://tinyurl.com/y5rloh3g>



Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements
4. Image to Text
5. Image Generation
6. Motion Estimation
7. 3D reconstruction from Images

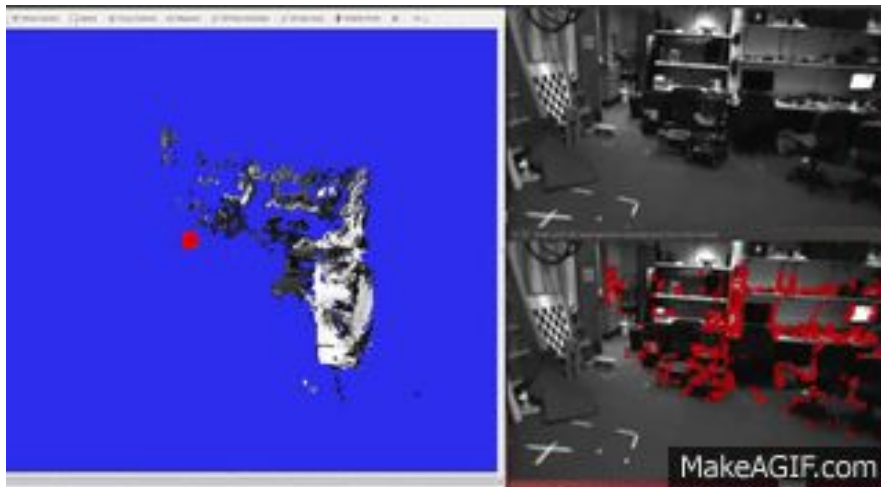


3D Reconstruction: REMODE,
Real Time Reconstruction
Credits: Matia Pizzoli, ICRA 2014

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements
4. Image to Text
5. Image Generation
6. Motion Estimation
7. 3D reconstruction from Images
8. Visual SLAM

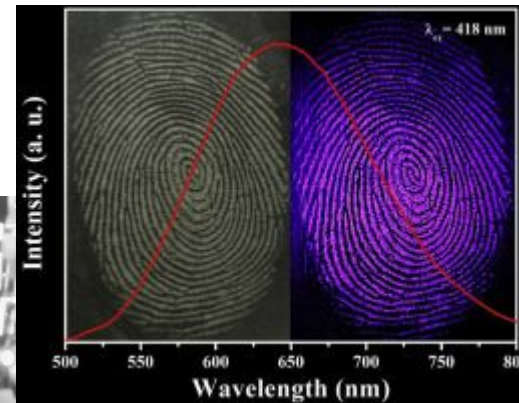
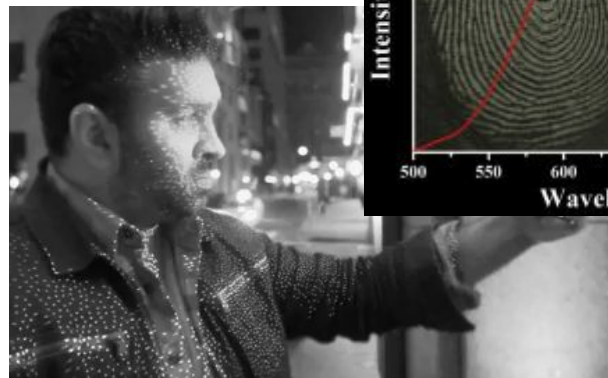


3D Reconstruction: REMODE,
Real Time Reconstruction
Credits: Matia Pizzoli, ICRA 2014

Overview of the field

Primary themes in Computer Vision are:

1. Object Detection
2. Segmentation
3. Image Modifications/Enhancements
4. Image to Text
5. Image Generation
6. Motion Estimation
7. 3D reconstruction from Images
8. Visual SLAM
9. Biometrics and more ...



Biometrics : Fingerprint
Detection, Apple Face ID

Credits:<https://tinyurl.com/y2a7wybz>,
TheVerge Youtube

A look at history

- **Robert Nathan** started writing computer programs for enhancing images from NASA's spacecraft's at Jet Propulsion Lab, NASA.
- The Summer Vision Project: Project at MIT to **solve a significant part of visual system**. Primary Objective was to divide the image into object, background and chaos regions, **over the course of a summer**.



Credits: EE604, nasa.gov

MASSACHUSETTS INSTITUTE OF TECHNOLOGY
PROJECT MAC

Artificial Intelligence Group
Vision Memo. No. 100.

July 7, 1966

THE SUMMER VISION PROJECT

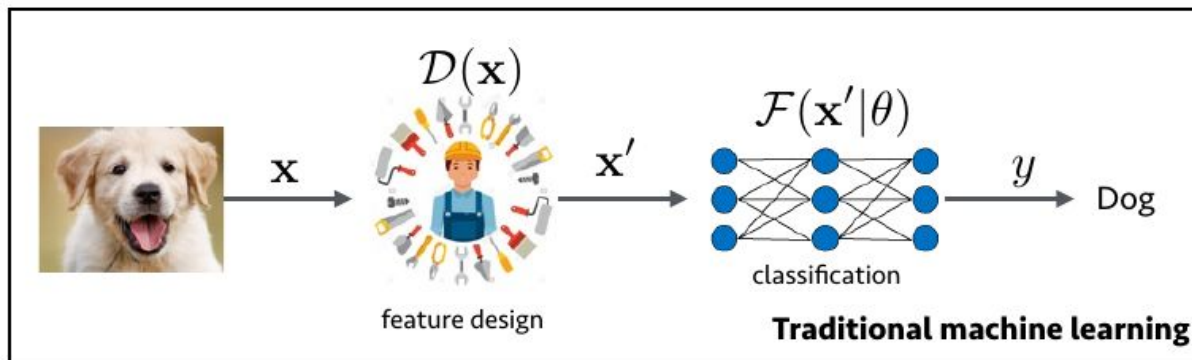
Seymour Papert

The summer vision project is an attempt to use our summer workers effectively in the construction of a significant part of a visual system. The particular task was chosen partly because it can be segmented into sub-problems which will allow individuals to work independently and yet participate in the construction of a system complex enough to be a real landmark in the development of "pattern recognition!!".

Credits: <https://tinyurl.com/y6bpo4nk>



A look at history



$\mathcal{D}(x)$

SIFT

SURF

TEXTON

STIP

HOG

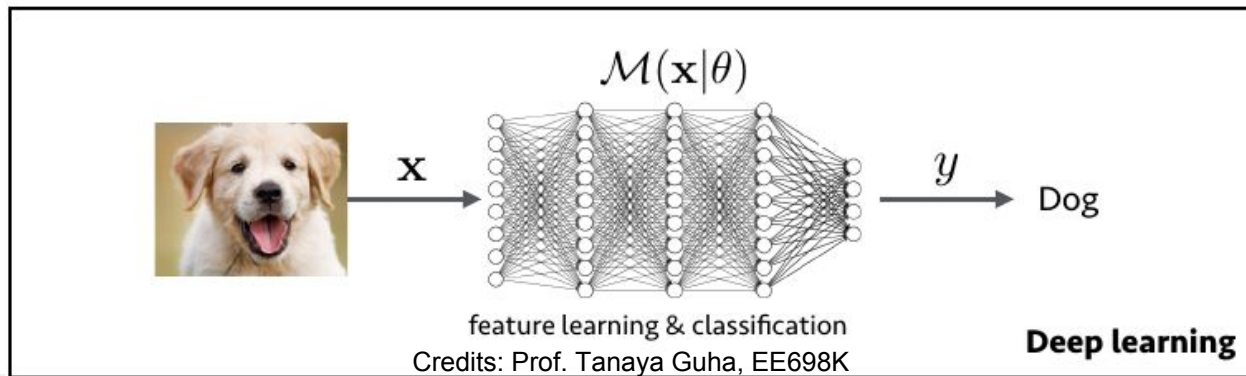
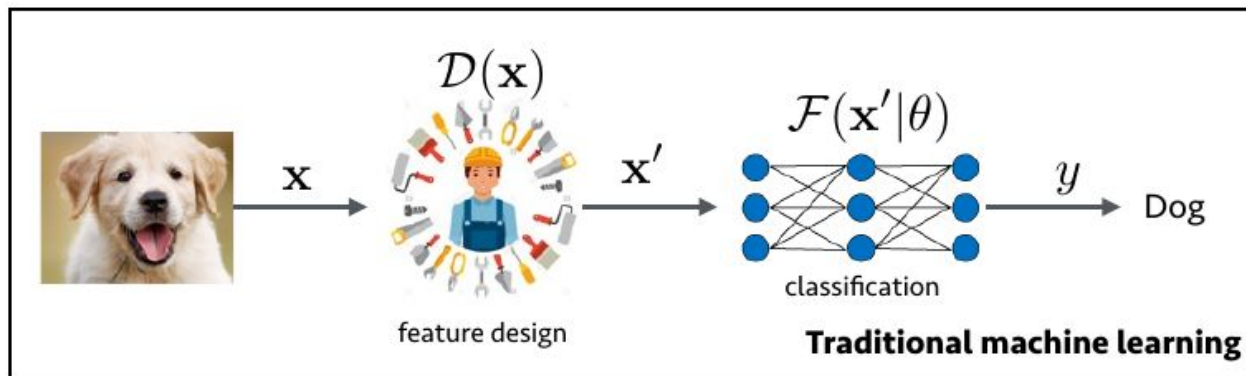
GIST

Many more ...

Credits: Prof. Tanaya Guha, EE698K



A look at history



Hard Problem?

- Why are we still working on roughly the same problem as the “summer vision project”?
- Why is it that creating 3D models of chairs is easier than identifying them?



Hard Problem?

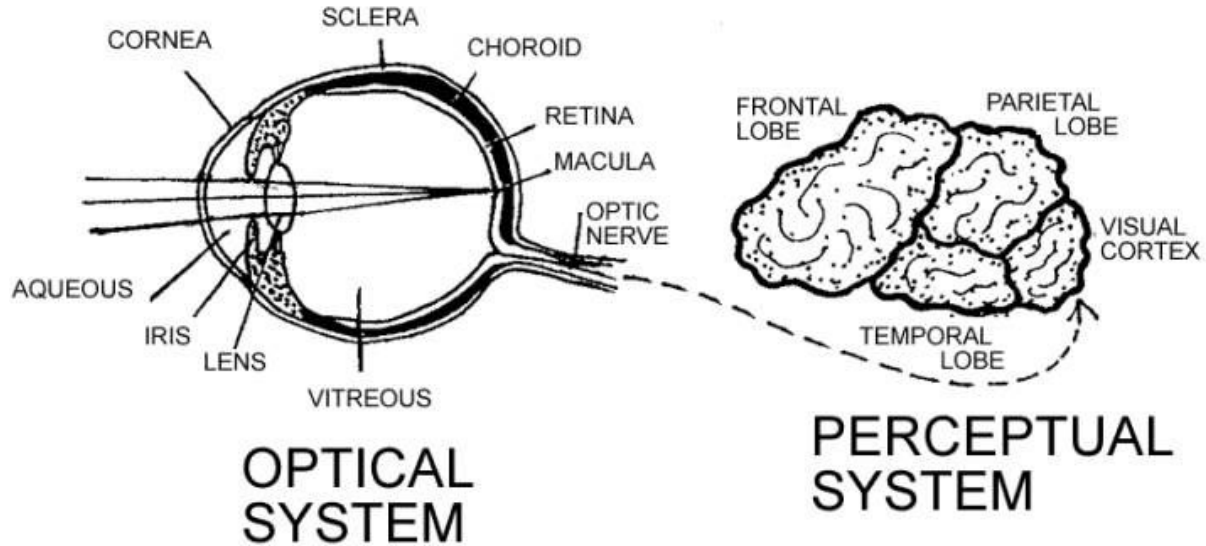
- Why are we still working on roughly the same problem as the “summer vision project”?
 - Why is it that creating 3D models of chairs is easier than identifying them?
- There is a large gap between some $\sim 1920 \times 1080 \times 3$ numbers and the high-level abstract meaning we associate with them.
- Images are **2D** representation of information from **3D** world.



Human Vision System & Computer Vision System



The human vision system



- VISUAL SYSTEM -

Credits: <https://tinyurl.com/y6bkhqqa>



The human vision system



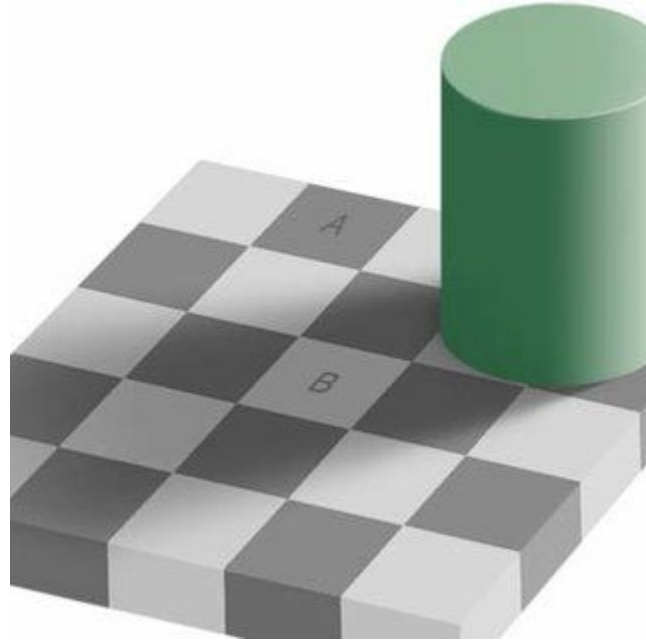
The human vision system



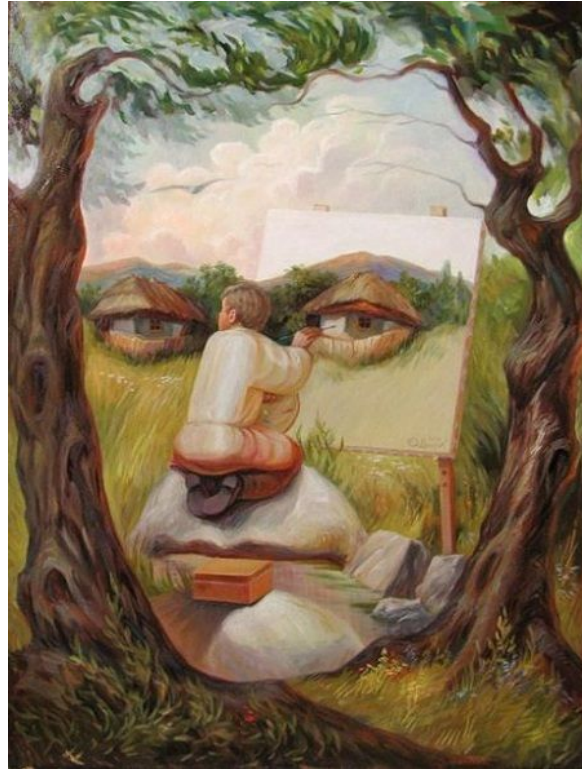
Credits: Ulas Bagci, UCF



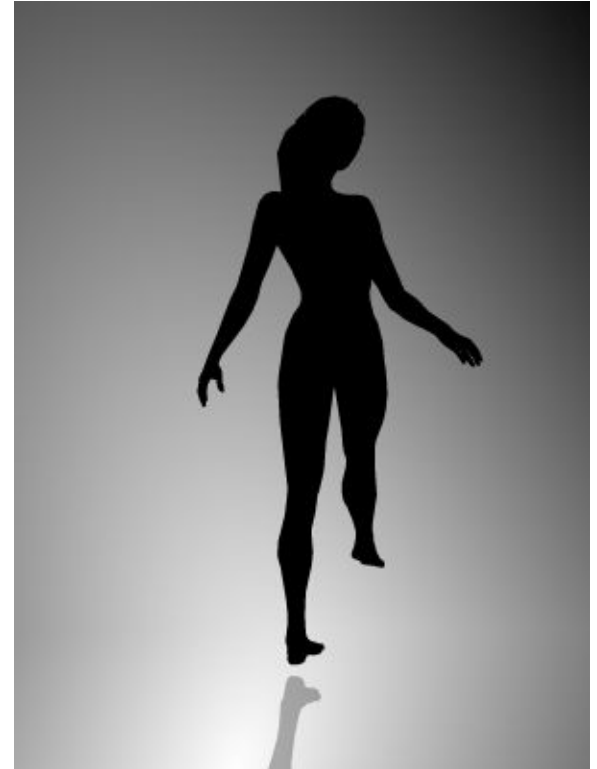
Fooling humans



Credits: <https://tinyurl.com/y49rp7sd>



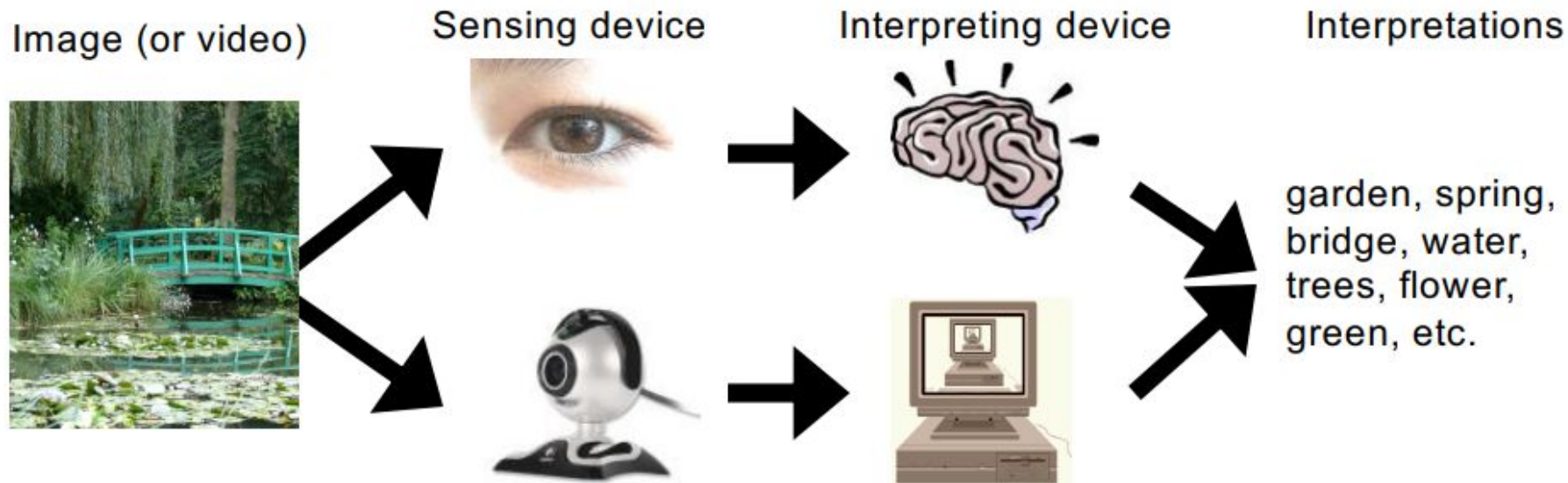
Credits: Oleg Shuplyak, Pinterest



Credits: Wikipedia, Spinning Dancer



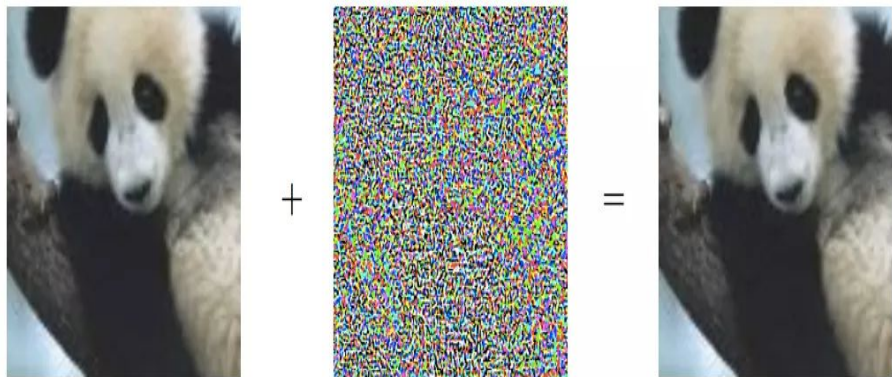
The ~~(human)~~ ~~(computer)~~ vision system



Credits: CS131, Stanford



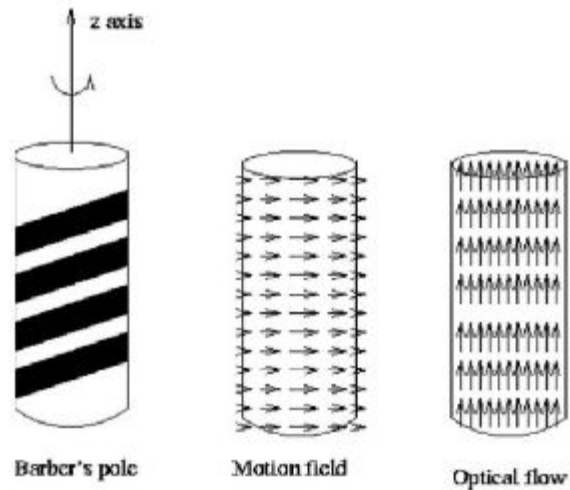
Fooling Computers



“panda”
57.7% confidence

“gibbon”
99.3 % confidence

Credits: <https://tinyurl.com/l5pwp6t>



Barber's pole

Motion field

Optical flow

Credits: Wikipedia, Barber Pole Illusion

Images as Matrices



Camera Models



Camera Models



Credits: <https://tinyurl.com/y6qen2vb>



Camera Models



Not this one but models as
in modelling a phenomena

Credits: <https://tinyurl.com/y6qen2vb>

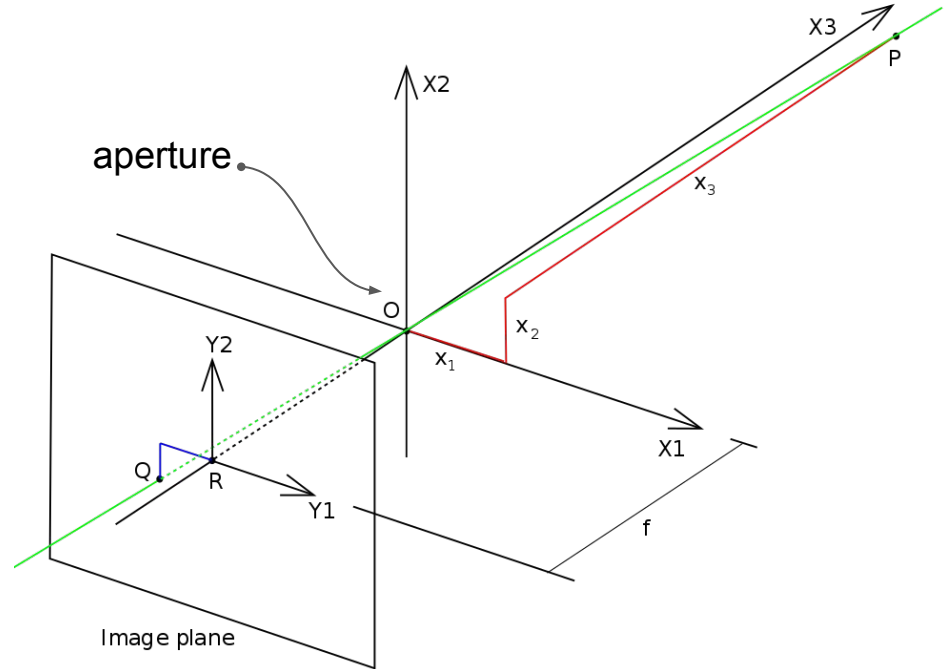
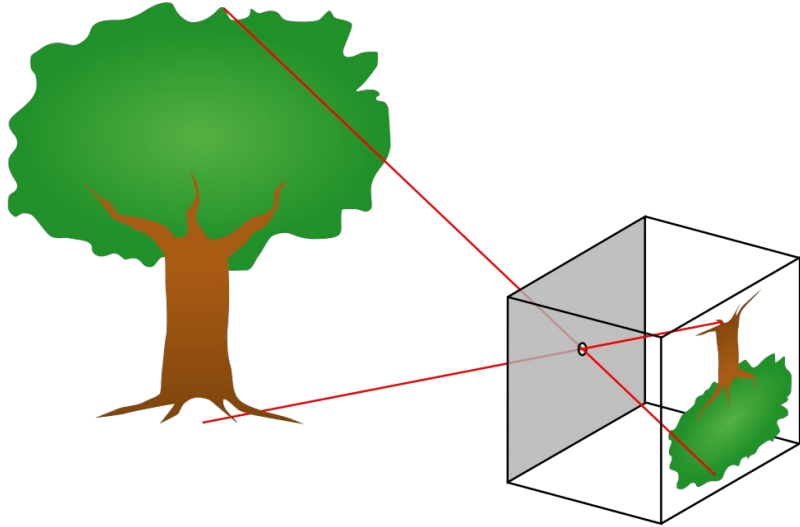


Camera Models

- Like so many things in engineering, we create a simple “model” of a camera to which is easy to understand and can approximate the actual functioning of a camera to a good degree.
- There are different models:
 - Pinhole camera model
 - Lens model
 - ...



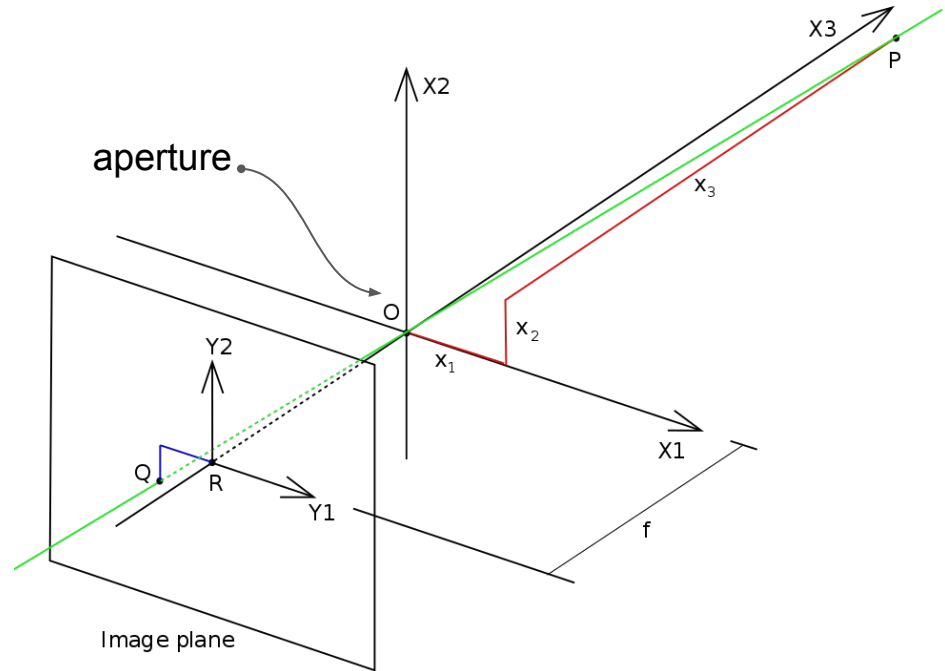
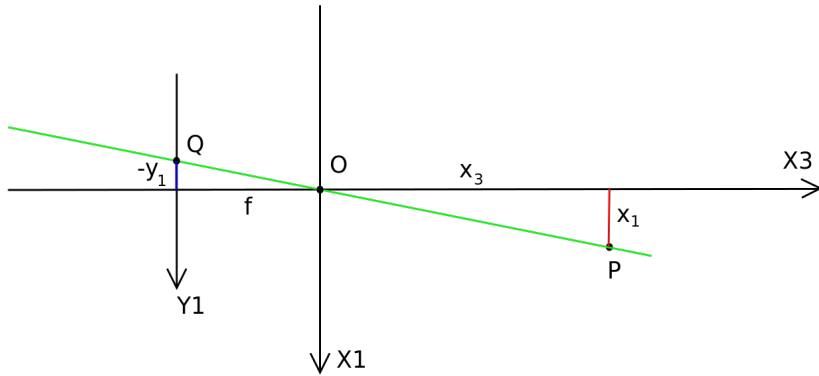
Pinhole camera model



Credits: Wikipedia, Pinhole Camera Model



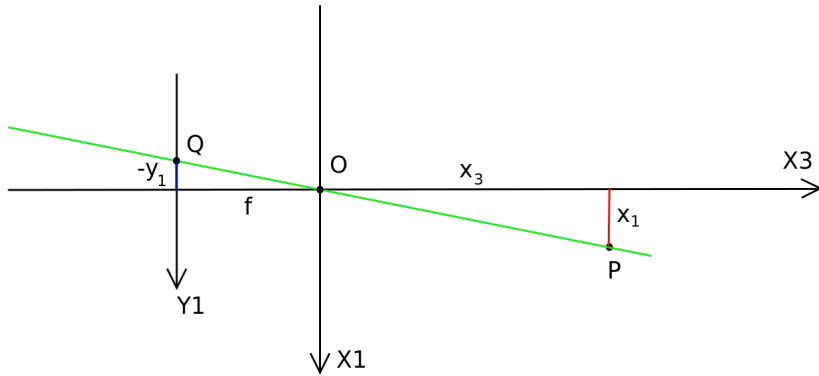
Pinhole camera model



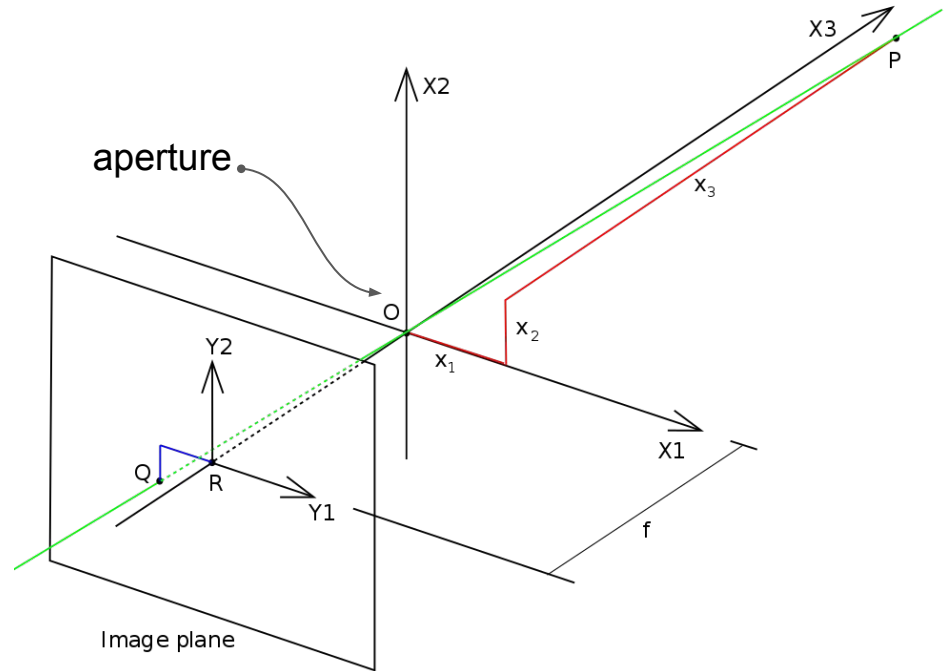
Credits: Wikipedia, Pinhole Camera Model



Pinhole camera model



$$y_1 = -\frac{f x_1}{x_3}$$



Credits: Wikipedia, Pinhole Camera Model



Pinhole camera model

$$x'_i = -xf/z \quad \text{where } x'_i = y_i \text{ and } z = x_3$$

$$x' = fx/z + c \quad \text{where } c \text{ is an offset in pixels}$$

$$p' = \begin{bmatrix} fx/z + c_x \\ fy/z + c_y \end{bmatrix} \quad p = \begin{bmatrix} x \\ y \\ z \end{bmatrix} \quad \text{Can we make this into a matrix multiplication of the form } \mathbf{p}' = \mathbf{M}\mathbf{p}?$$



Intrinsic camera matrix

- We'll introduce a new convention, *homogenous coordinates*.
- We write points just the way we did before, but add an extra row:
 - ▶ The extra row is a *scale factor* for the whole vector.

$$p = \begin{bmatrix} x \\ y \\ z \end{bmatrix} \quad \text{becomes} \quad p = \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$



Intrinsic camera matrix

- Eureka!

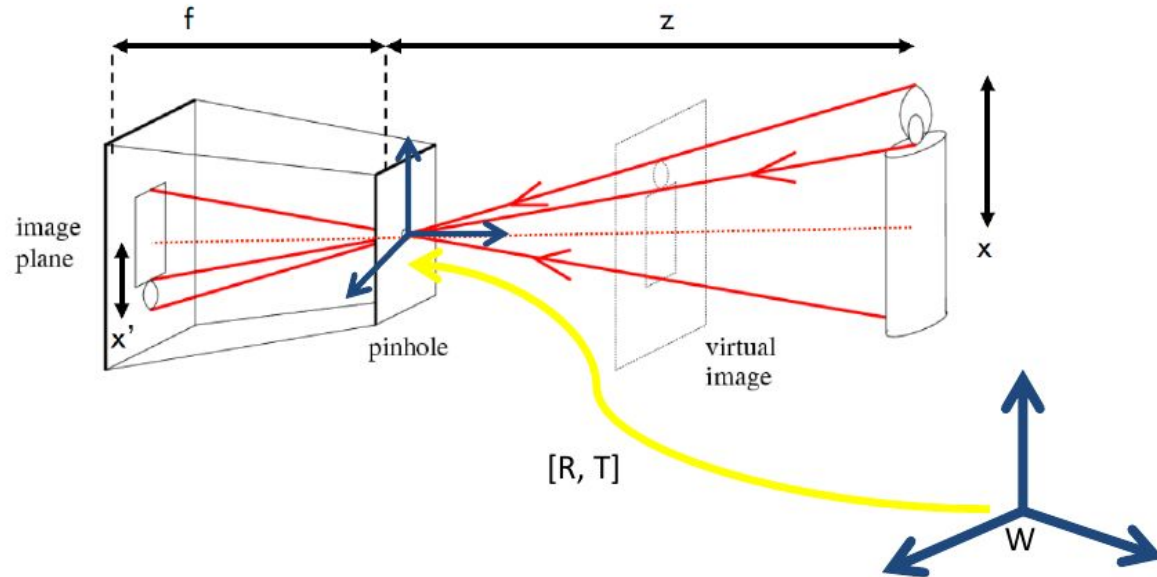
$$p' = \begin{bmatrix} fx/z + c_x \\ fy/z + c_y \\ 1 \end{bmatrix} = \begin{bmatrix} fx + c_x z \\ fy + c_y z \\ z \end{bmatrix} = \begin{bmatrix} f & 0 & c_x & 0 \\ 0 & f & c_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

2d homogeneous coordinates 2d homogeneous coordinates perspective transform 3d homogeneous coordinates

Credits: Edwin Olson, University of Michigan



Intrinsic camera matrix



Credits: Edwin Olson, University of Michigan



Intrinsic camera matrix

- The product of two rigid-body transformations is **always** another rigid-body transformation!
- So no matter how the object has been translated or rotated, we can describe its position with a single 4x4 matrix, which has the structure:

$$\begin{bmatrix} x' \\ y' \\ z' \\ 1 \end{bmatrix} = \begin{bmatrix} R_{00} & R_{01} & R_{02} & T_x \\ R_{10} & R_{11} & R_{12} & T_y \\ R_{20} & R_{21} & R_{22} & T_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

Credits: Edwin Olson, University of Michigan



Intrinsic camera matrix

$$\begin{bmatrix} x' \\ y' \\ s \end{bmatrix} = \begin{bmatrix} f & 0 & c_x & 0 \\ 0 & f & c_y & 0 \\ 0 & 0 & 1 & 0 \end{bmatrix} \begin{bmatrix} R_{00} & R_{01} & R_{02} & T_x \\ R_{10} & R_{11} & R_{12} & T_y \\ R_{20} & R_{21} & R_{22} & T_z \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ y \\ z \\ 1 \end{bmatrix}$$

2d
homogeneous
pixel (camera)
coordinates

Focal length and
focal center of camera

“Intrinsics”

Rigidly move every object in
the world to simulate the
camera's true position

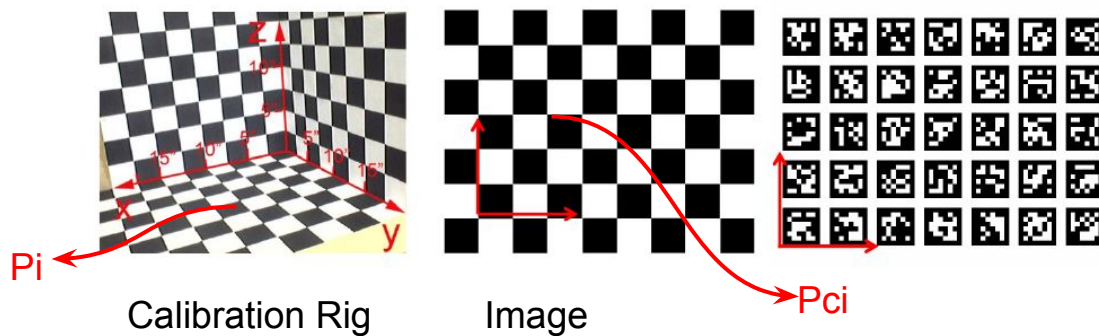
“Extrinsics”

3d
homogenous
(world)
coordinates

Credits: Edwin Olson, University of Michigan



Camera calibration



$$M = \arg \min_M \sum \|P_{ci} - MP_i\|^2$$

$$P_{ci} \Leftrightarrow P_i$$

Credits: Gaurav Pandey, Ford

