

Computer Vision

Introduction

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DEPARTMENT OF COMPUTER SCIENCE,

FAST-NUCES, Peshawar

Course Details

LECTURES: Tuesday
& Wednesday

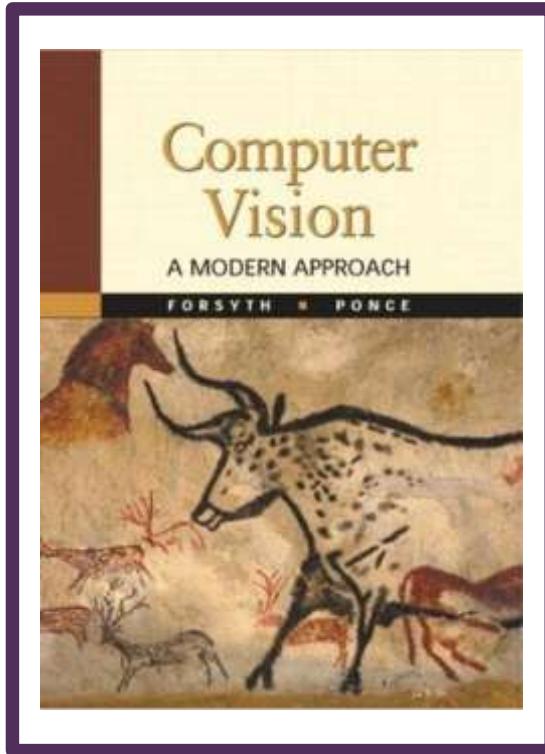
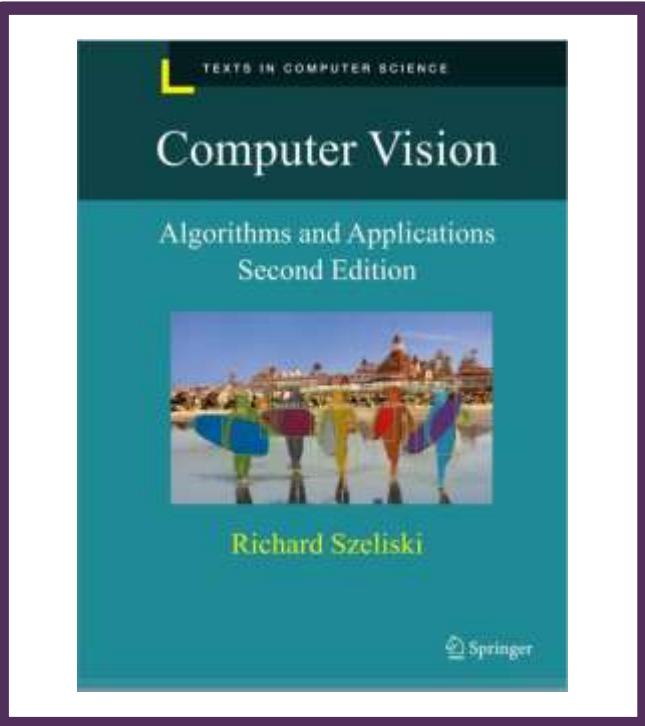
TIMINGS:

8:00 am – 9:30 am

MY OFFICE:

OFFICE HOURS:

EMAIL: m.tahir@nu.edu.pk



References

The material in these slides are based on:

1

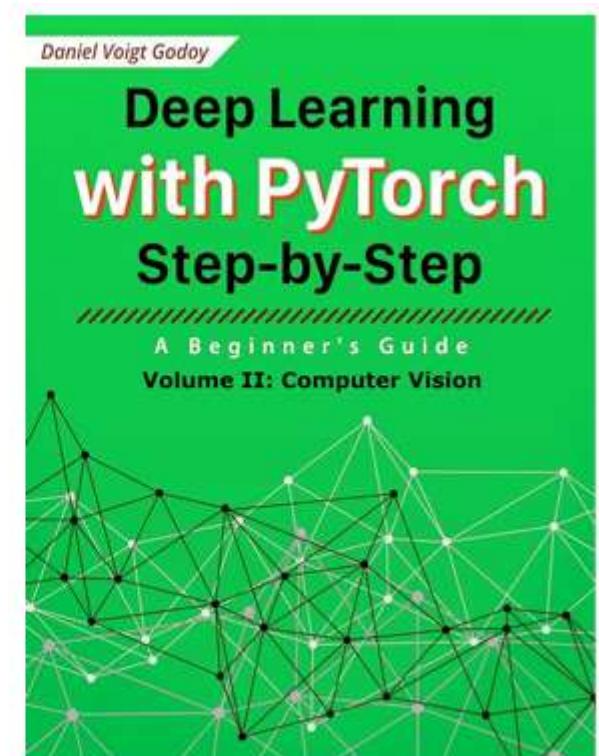
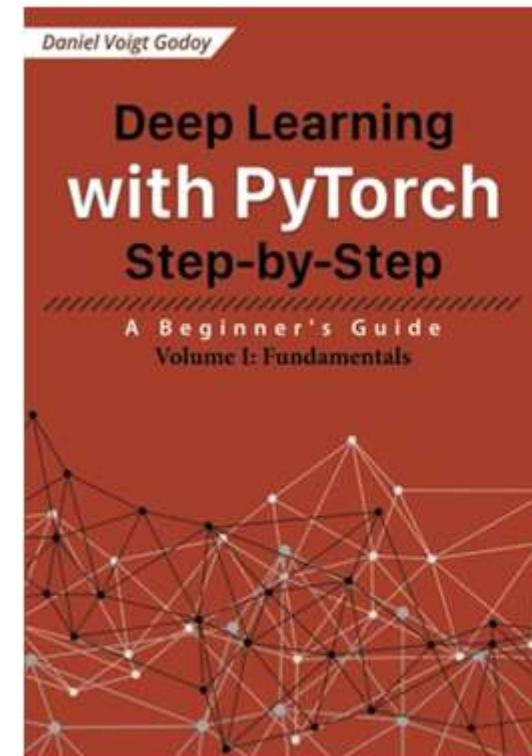
Rick Szeliski's book: [Computer Vision: Algorithms and Applications](#)

2

Forsythe and Ponce: [Computer Vision: A Modern Approach](#)

Recommended Books

Deep Learning with PyTorch Step-by-Step by Daniel Voigt Godoy



Course Administration

- Attendance Policy

- ✓ FAST attendance Policy?
- ✓ Students not making up to the 80% attendance will not be allowed to sit in the final exam
- ✓ Latecomers should be careful

- Assignment / Quiz Policy:

- ✓ Assignments will be hand-written EXCEPT in the case of coding task
- ✓ Deadlines must be followed.
- ✓ There will be no makeup of any quizzes
- ✓ Late assignment will be accepted with negative marking, -1 for each day
- ✓ A copied assignment/project will be simply marked zero.

- Lab Policy:

- ✓ Home Tasks

Course Project

- The course project gives students a chance to apply Computer Vision Algorithms discussed in class to a research oriented or industrial problem
- Working in Groups
- Evaluation of article reading and project
 - ✓ Article reading: Submit a survey of at least 5 articles you read and the list of the articles (2020 to 2024, TPAMI, IJCV, ICCV, CVPR, ECCV)
 - ✓ Submit an article including introduction, methods, experiments, results, and conclusions
 - ✓ Submit the project code, the readme document, and some testing samples (images, videos, etc.) for validation
- Presentation

Course Project

- Potential Topics

- ✓ Image-to-Image Translation (satellite images to maps, turning day scenes into night scenes, or transforming sketches into realistic images)
- ✓ Deepfake Technology (creating realistic-looking images or videos)
- ✓ 3D Object Recognition and Reconstruction (augmented reality)
- ✓ Event/Anomaly Recognition in Videos
- ✓ Multi-Modal Image Matching (optical, infrared)
- ✓ Multi Object Tracking (under occlusions, scale changes, and appearance variations)
- ✓ Remote Sensing and Satellite Image Analysis (land cover classification, disaster monitoring, and environmental assessment)
- ✓ Medical Image Analysis (Pathology slides, Radiology, Retinal)

Course Learning Outcomes

No	CLO (Tentative)	Domain	Taxonomy Level	PLO
1	Understanding basics of Computer Vision: algorithms, tools, and techniques	Cognitive	2	
2	Develop solutions for image/video understanding and recognition	Cognitive	3	
3	Design solutions to solve practical Computer Vision problems	Cognitive	3	



Outline

What is Computer Vision

Human Visual System

Computer Vision Challenges

Brief History of Computer Vision

Computer Vision Tasks

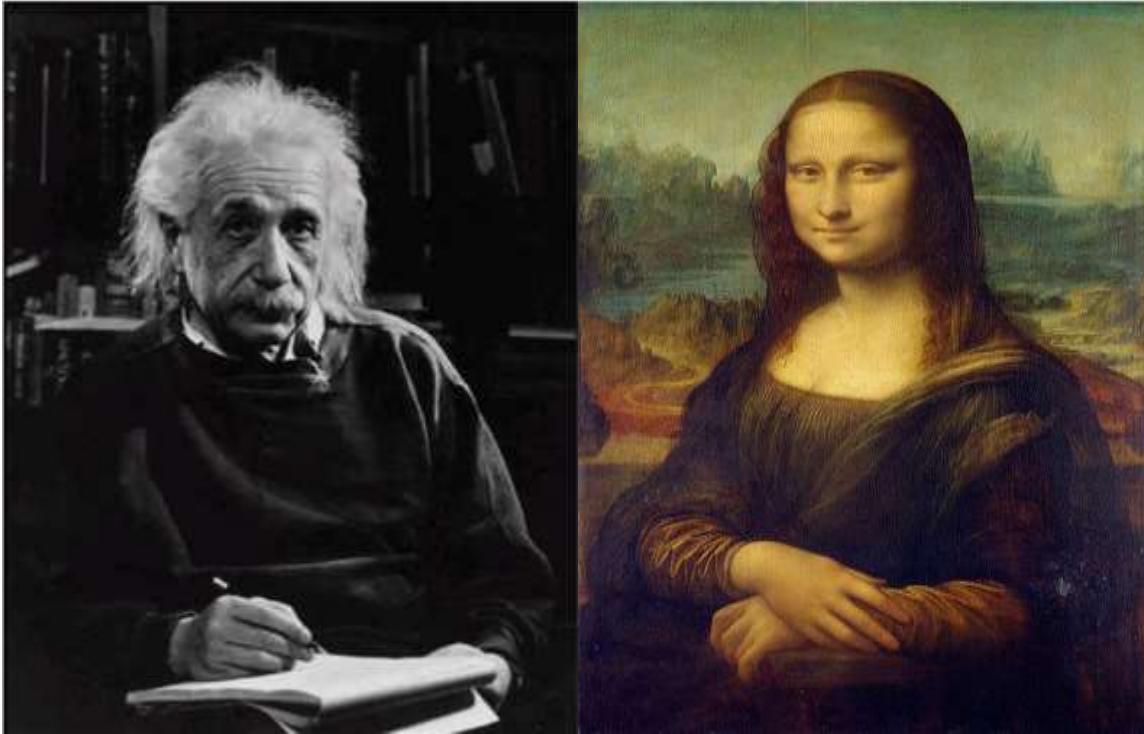
Computer Vision Applications



A picture
is worth a
100
words.



A picture is worth 10,000 words.



A picture
is worth a
100
words.



What is
computer vision?

What kind of
scene?

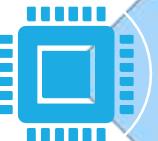
Where are the
cars?

How far is the
building?

What is computer vision?



Automatic understanding of images and video.



Giving the computer "eyes" to see and identify as humans would.



The study of understanding the world through visual perception.



It is also called:

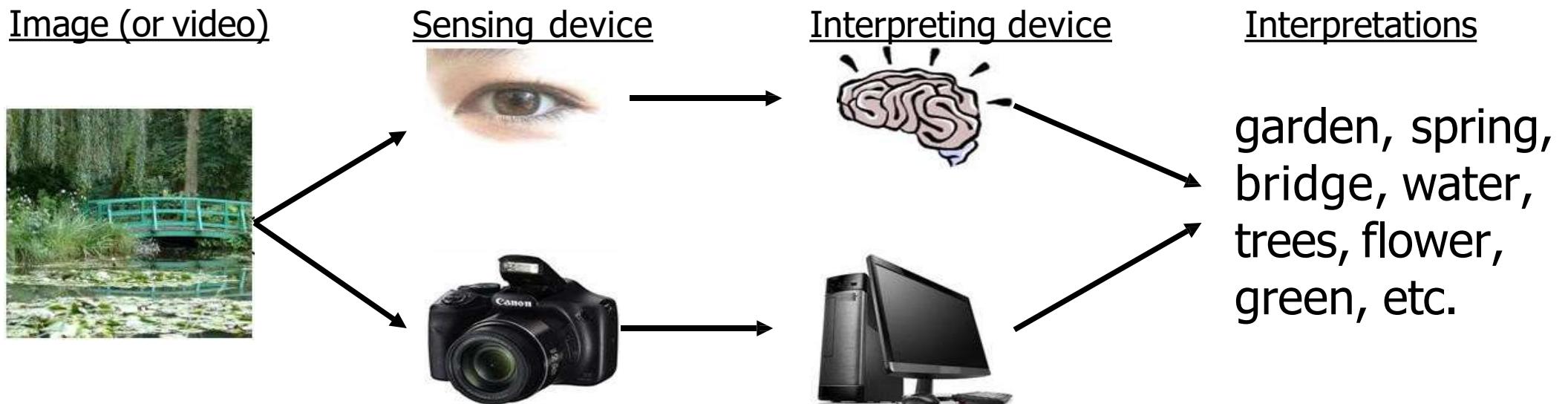
- *Image understanding*
- *Robot vision*
- *Image Analysis*

- What kind of scene?
- Where are the cars?
- How far is the building?



What is (computer) vision?

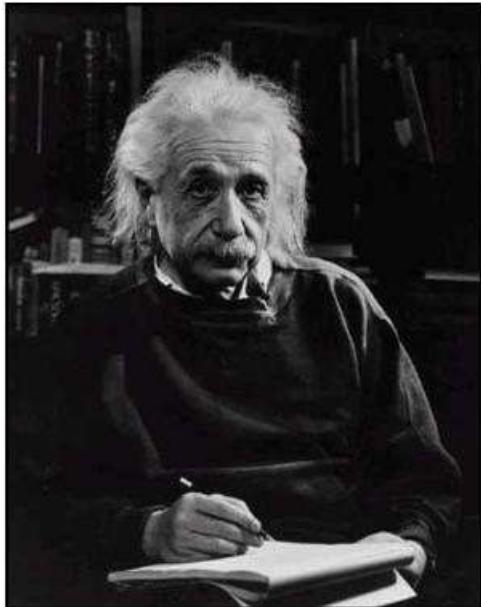
Image (or video) Sensing device Interpreting device
Interpretations



- Computer Vision emulates human vision
- Process images and extract information using context
- Make decisions using this information

What is computer vision?

- The goal of computer vision is to give computers human-like perception
 - ✓ bridge the gap between pixels and “meaning (semantic)”

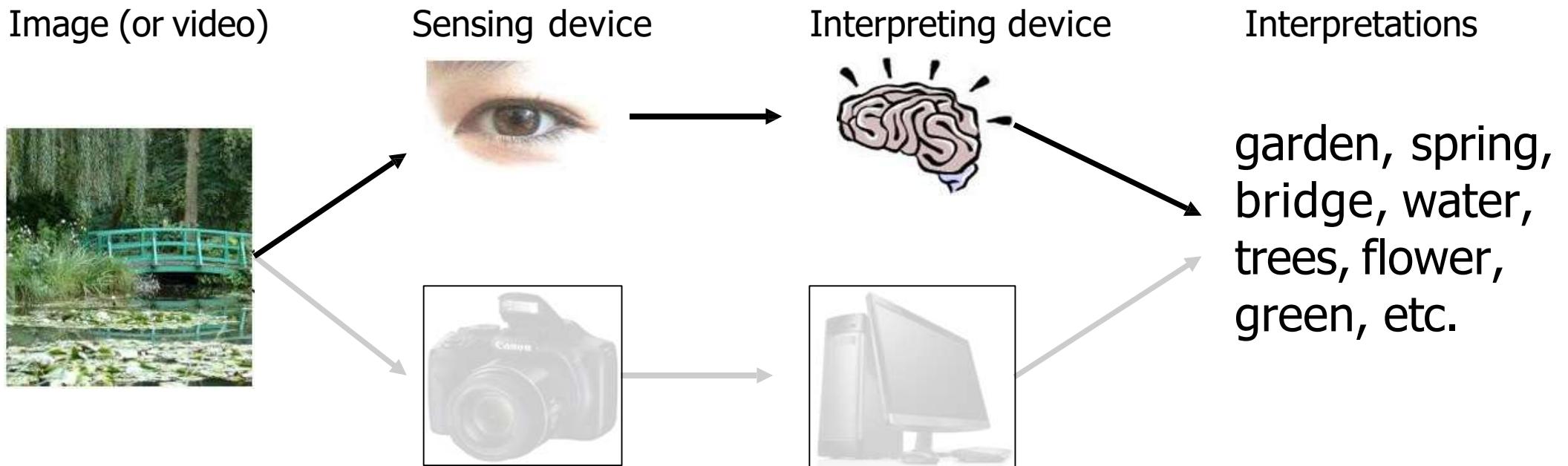


What we see

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

What is (computer) Vision?



Human vision

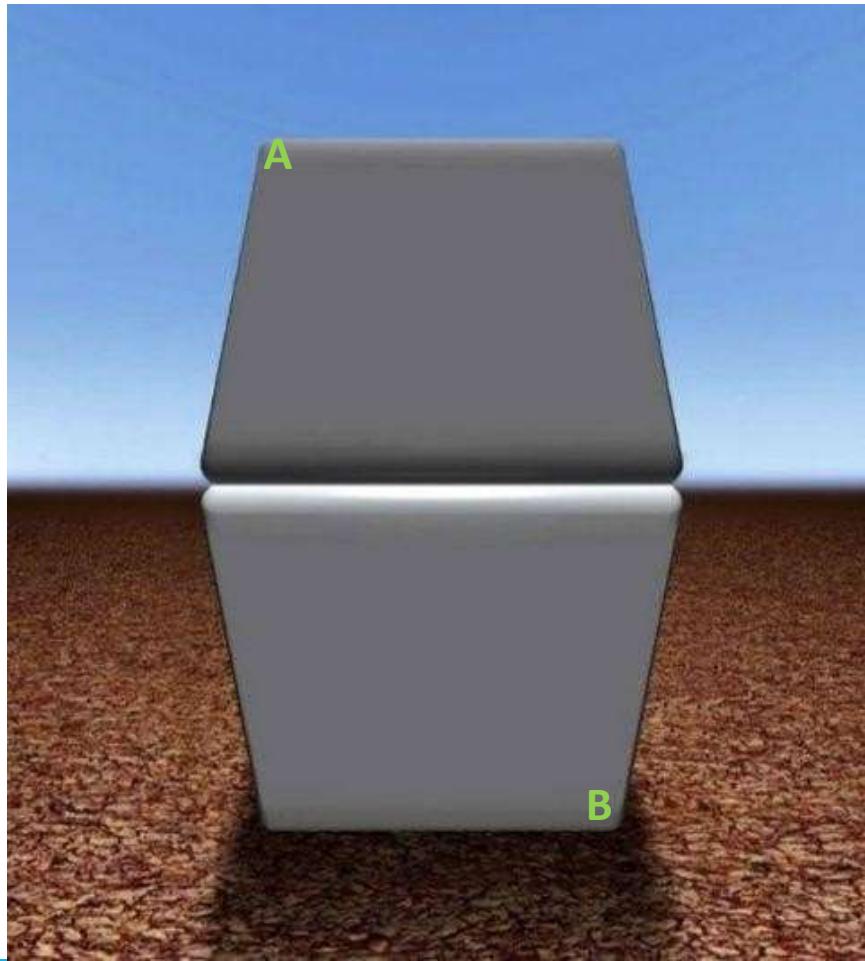
CAN DO AMAZING THINGS LIKE:

- ✓ Recognize people and objects
- ✓ Navigate through obstacles
- ✓ Understand mood in the scene
- ✓ Imagine stories

BUT STILL IS NOT PERFECT:

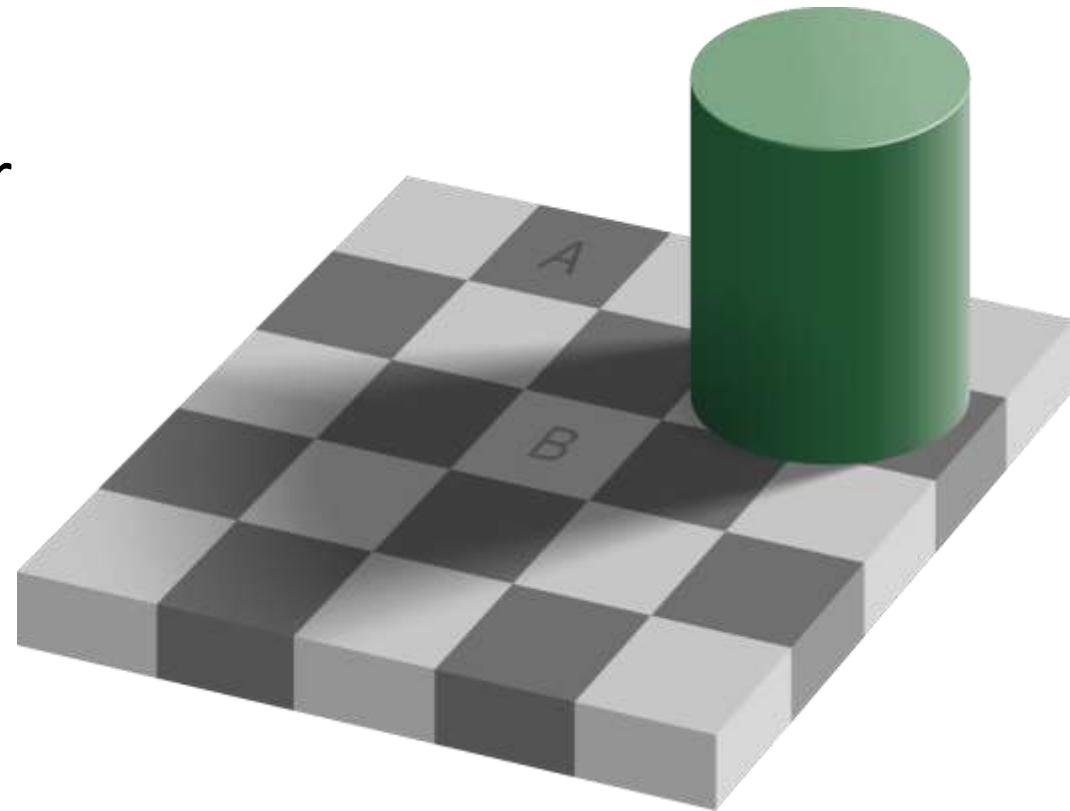
- ✓ Suffers from illusions
- ✓ Ignores many details
- ✓ Ambiguous description of the world

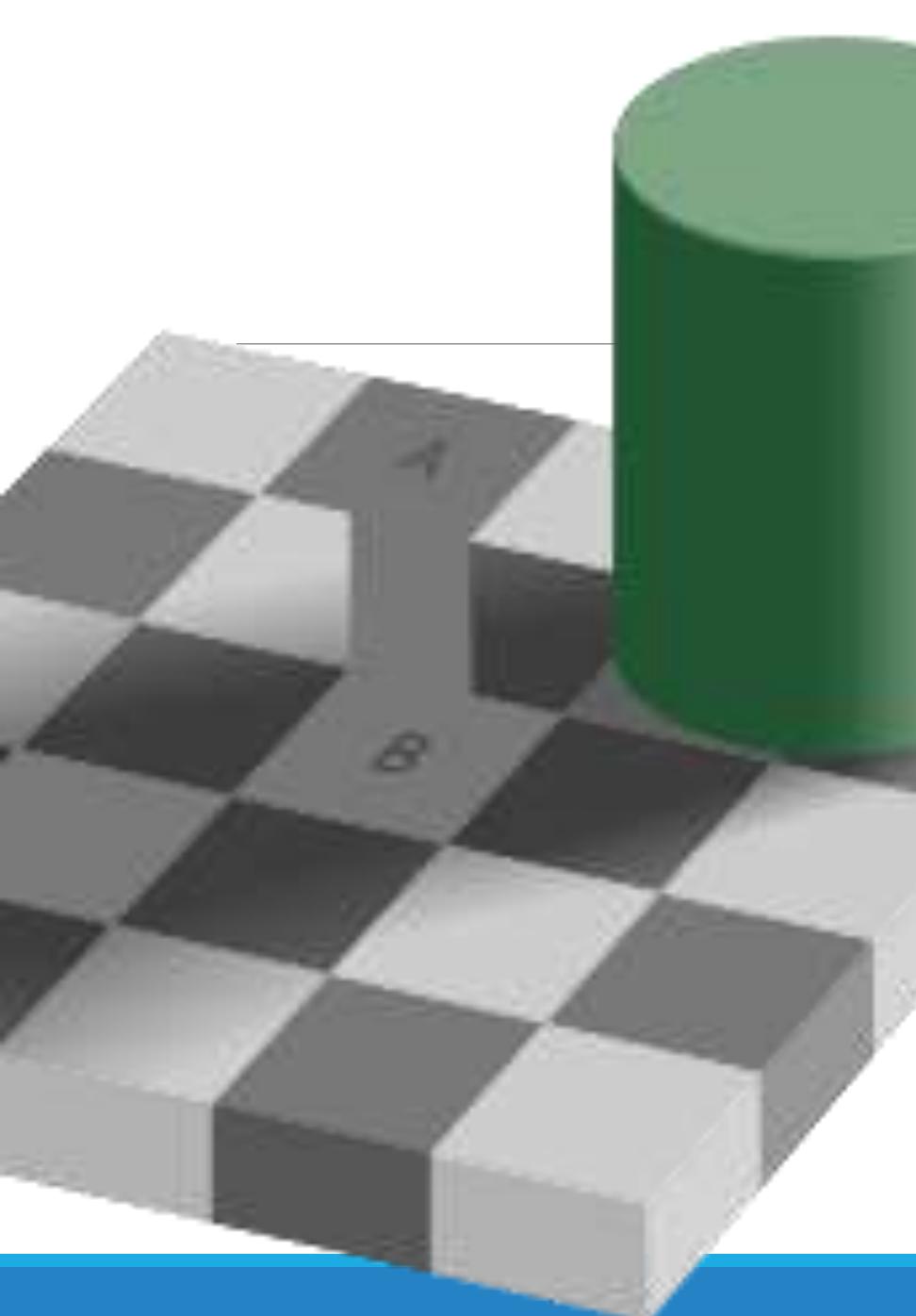
Measurement vs Perception



Same Color

A appears markedly darker than B, even though they are the exact same shade of gray.





Measurement vs Perception

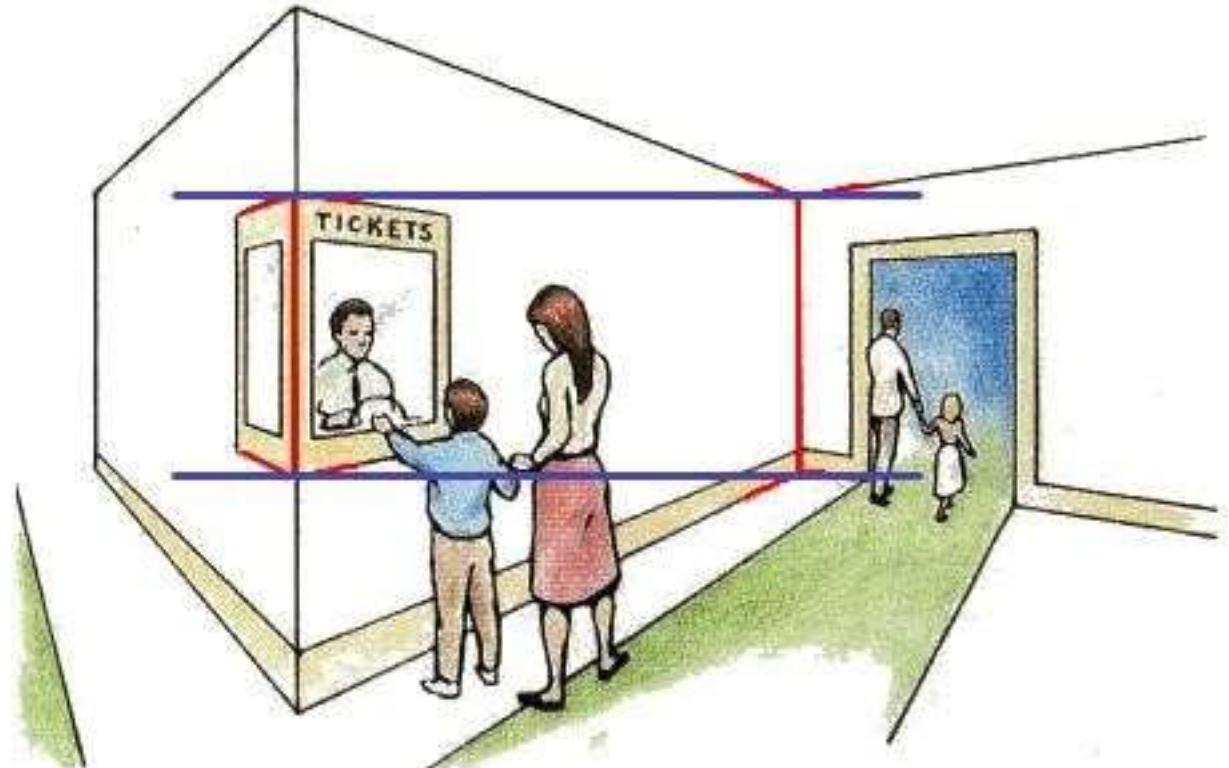
Lighter squares surrounded by darker squares appear lighter than average based on local contrast.

Conversely, darker squares surrounded by lighter squares appear darker.

Measurement vs Perception

- Length Illusion

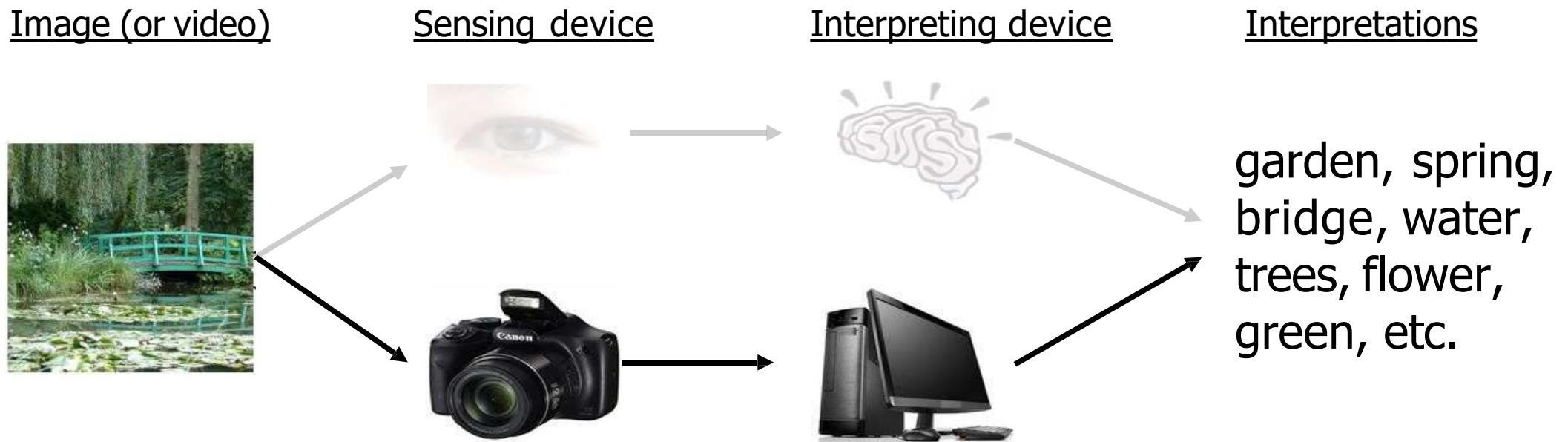
The brain assumes the “angles in” configuration to be closer, computes size constancy on it, and –given identical retinal size of the two angle arrangements – concludes that the “angle in”-line is shorter



Ref -

Gregory 1968: <https://michaelbach.de/ot/sze-muelue/index.html>

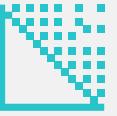
What is (computer) vision?



What is Computer Vision?



Information Extraction: features, 3D structure, motion etc.



Interpretation: Recognize Objects, scenes, actions, events.



No matter what your definition is ...

Vision is hard.
But is fun...

What is not Computer Vision?

Image Processing:

Takes an image and process is to produce new, more desirable image. Image enhancement, image compression, image restoration.

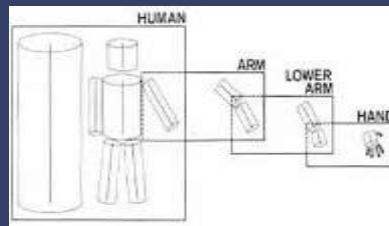
Pattern Recognition :

Takes a pattern and classifies it into one of predefined, finite set of classes.

Computer Graphics:

Synthesize images using powerful algorithms so that they correspond as close as possible to the real images.

Brief History of Computer Vision



1970's Example



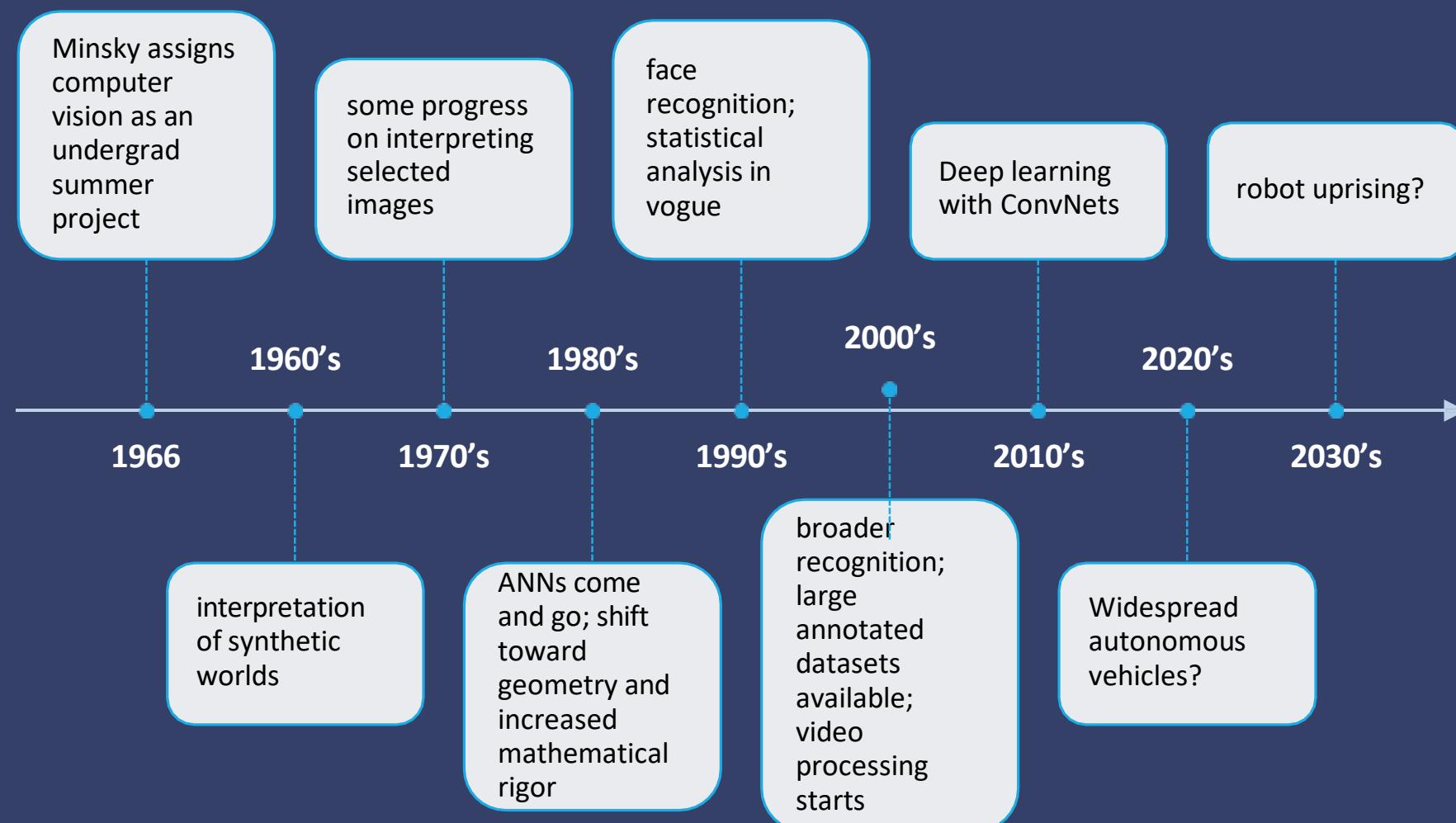
1980's Example



1990's Example



2010's Example

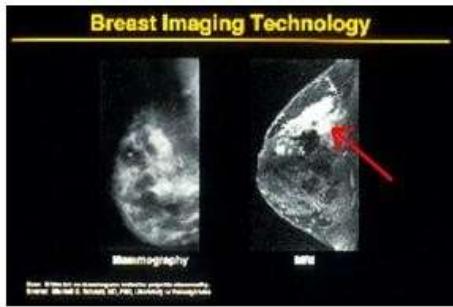


Why is Computer Vision so Difficult?

- It is an inverse problem, in which we seek to recover some unknowns given insufficient information to fully specify the solution.
- We resort to physics-based and probabilistic models, or machine learning from large sets of examples, to disambiguate between potential solutions.
- However, modeling the visual world in all its rich complexity is far more difficult than, say, modeling the vocal tract that produces spoken sounds



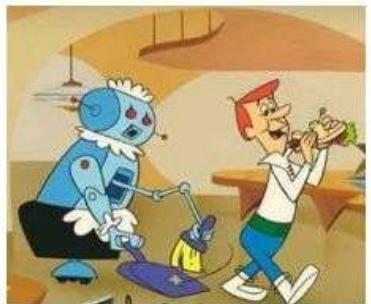
Safety



Health



Security



Comfort

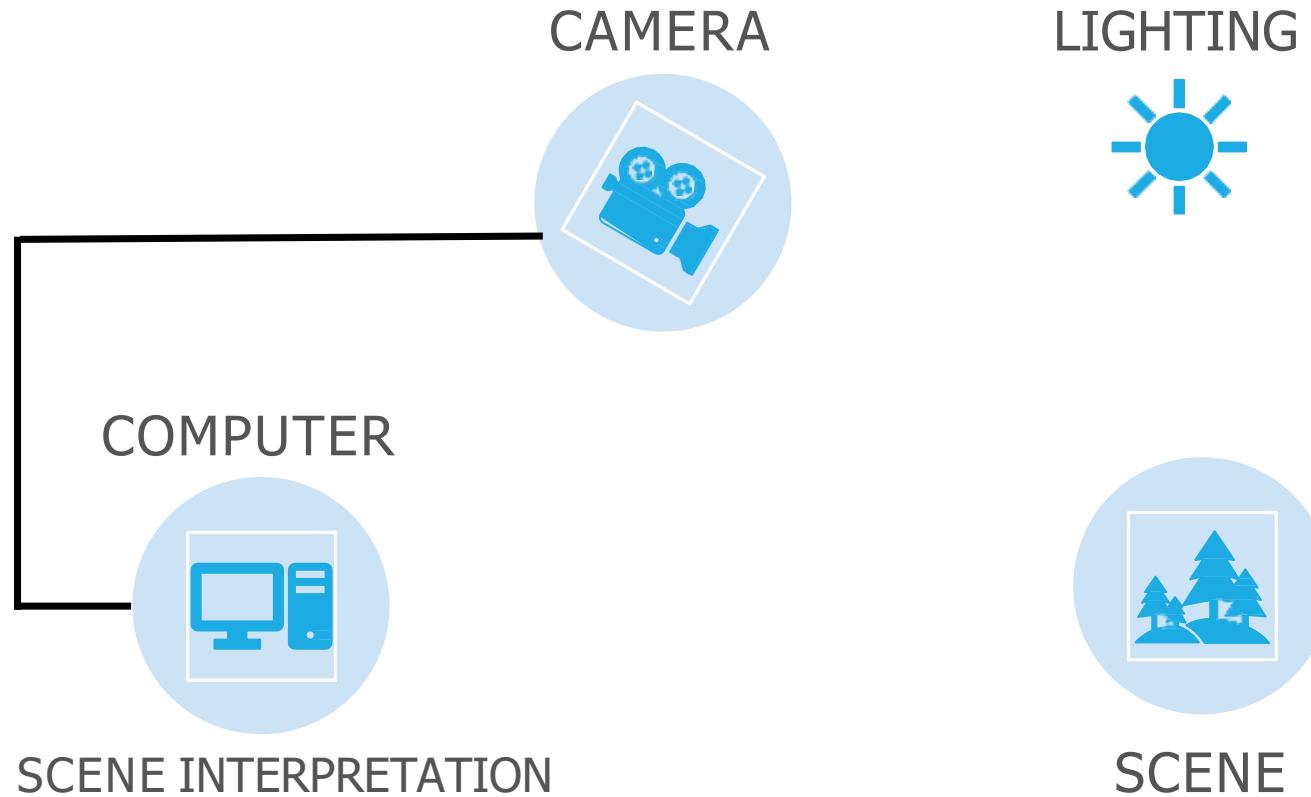


Fun



Access

Why
computer
vision
matters?



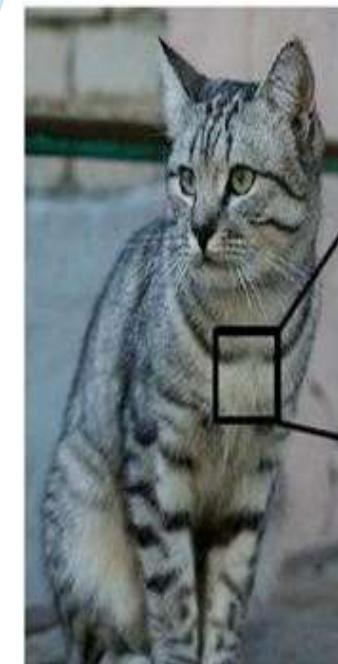
Components
of a
computer
vision system

Computer Vision Challenges

- Viewpoint Variation



- All pixels change when the camera moves



1105	512	380	515	164	16	246	59	96	183	112	215	184	87	83	851
112	90	202	206	284	79	90	203	99	249	122	126	128	200	94	851
116	89	98	109	128	189	47	94	98	99	124	112	109	109	93	851
119	81	93	93	120	131	127	189	93	98	182	99	99	92	181	841
126	91	81	64	80	93	88	89	181	187	188	99	75	84	94	852
134	580	85	35	55	89	64	24	66	87	112	119	88	76	81	913
135	337	147	185	65	81	88	65	93	94	34	84	142	93	85	821
138	537	149	540	189	15	88	78	62	45	63	68	73	85	1851	
125	330	148	137	159	121	117	54	95	79	86	111	54	84	72	845
127	525	131	147	133	127	126	121	115	96	99	75	61	64	72	841
130	218	169	123	218	189	233	510	112	189	206	96	74	66	72	783
136	99	98	97	149	147	110	118	113	114	113	180	186	96	77	845
132	77	84	82	77	79	242	222	157	115	117	129	129	129	119	971
142	88	83	89	78	31	88	183	134	126	116	191	187	154	131	182
143	85	73	88	89	71	62	81	129	138	135	185	81	88	119	183
137	85	75	87	165	98	68	45	76	138	126	187	97	84	185	1851
138	87	87	86	117	123	118	86	45	51	95	93	89	85	182	1873
134	546	153	80	82	128	524	584	76	48	45	68	88	181	187	1867
157	578	157	120	93	98	214	132	112	97	88	53	79	82	93	941
120	228	124	162	120	189	249	118	123	124	124	87	63	53	69	861
158	512	94	117	158	141	128	125	184	187	182	93	87	81	22	295
122	287	96	85	63	132	212	249	122	182	184	79	89	287	122	893
132	321	187	99	82	98	94	117	149	148	153	182	54	78	92	1871
133	364	549	193	75	58	78	83	93	183	139	189	182	81	89	8413



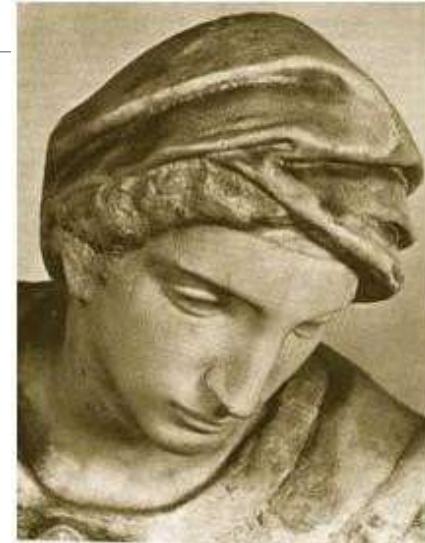
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Computer Vision Challenges

- Viewpoint Variation

Michelangelo 1475 – 1564

Adapted from L. Fei Fei, R. Fergus, A. Torralba



Computer Vision Challenges

- Background Clutter



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Computer Vision Challenges

- Illumination



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Computer Vision Challenges

- Occlusion



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Computer Vision Challenges

- Deformation



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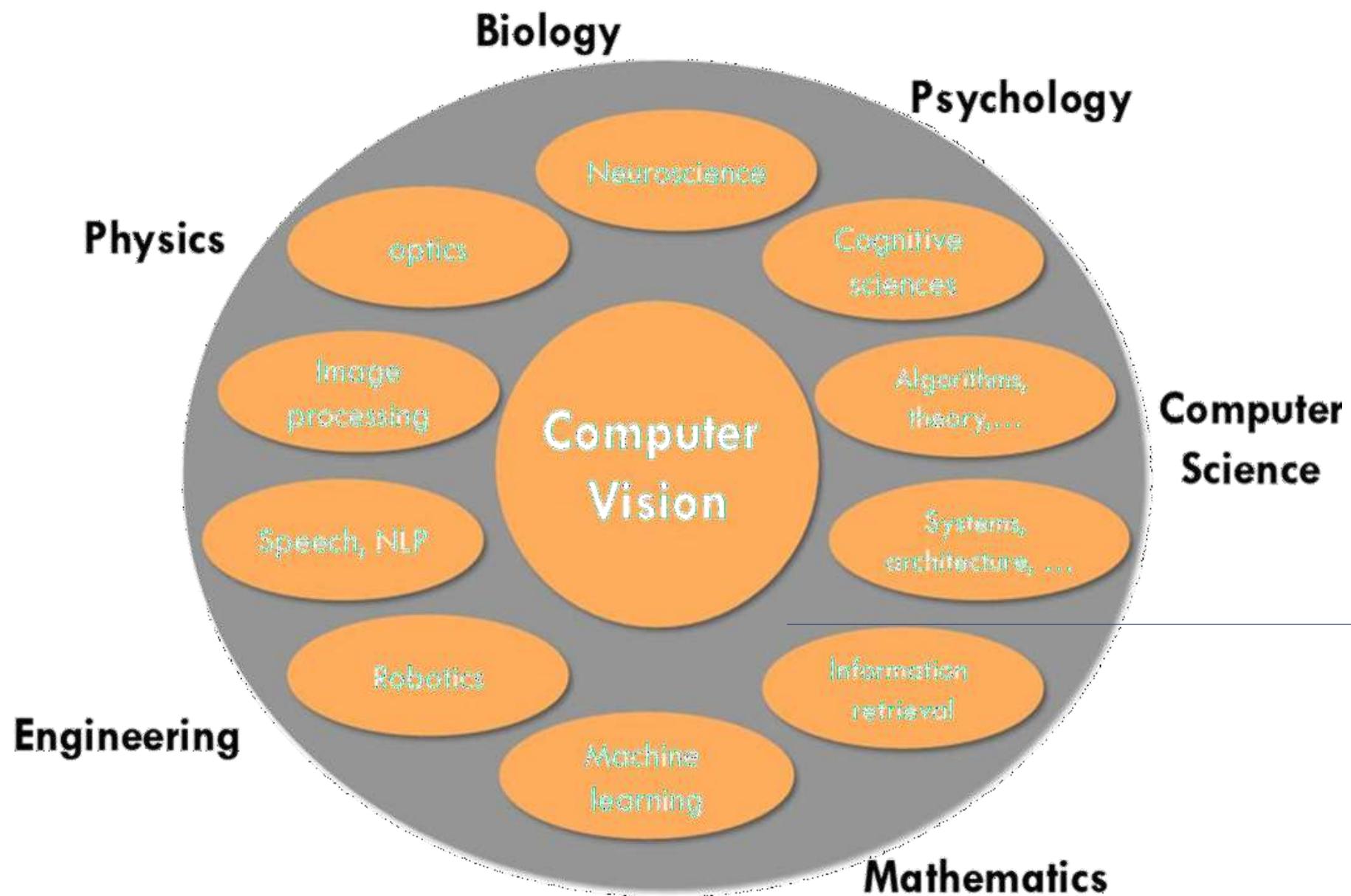
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Computer Vision Challenges

- Intra-class Variation



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Scope of Computer Vision

Computer Vision Tasks

Reorganization

- Segment (cluster) pixels with similar properties

Reconstruction

- Recover 3D information from data

Recognition

- Detect and identify objects

Understanding

- What is happening in the scene?

Computer Vision Tasks



Low Level Vision

Measurements

Enhancements

Region segmentation

Features



Mid Level Vision

Reconstruction

Depth

Motion Estimation



High Level Vision

Category detection

Activity recognition

Deep understandings

Pose estimation

Computer Vision Tasks

Low-Level or Early Vision

- Local image/shape properties



“There’s an edge!”

Mid-Level Vision

- Grouping and segmentation



“There’s an object and a background!”

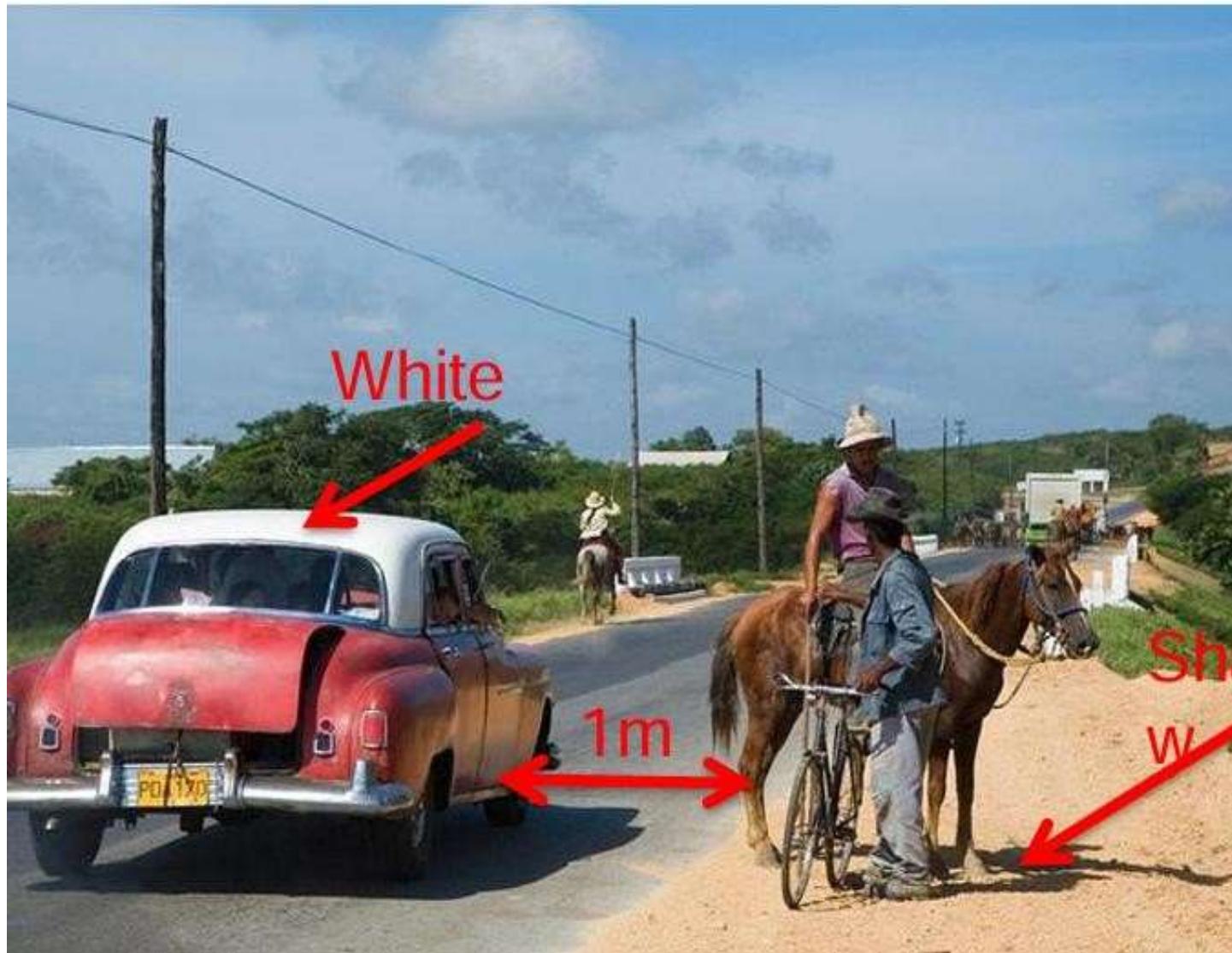
High Level Vision

- Recognition
- Classification



“It’s a chair! It’s in a room!”

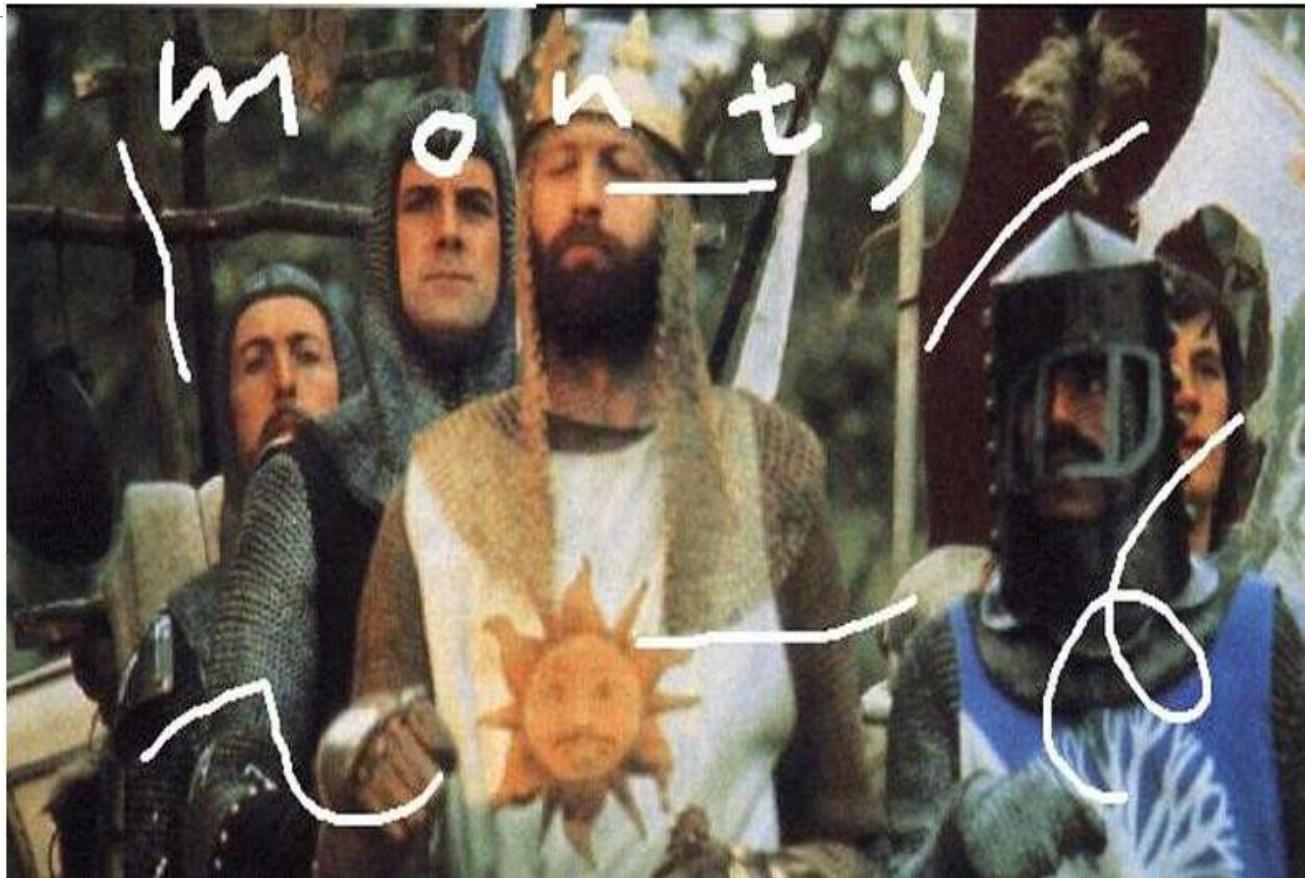
Computer Vision Tasks



- Low Level Vision
 - ✓ Measurements
 - ✓ Enhancements
 - ✓ Region segmentation
 - ✓ Features
- Mid Level Vision
 - ✓ Reconstruction
 - ✓ Depth
 - ✓ Motion Estimation
- High Level Vision
 - ✓ Category detection
 - ✓ Activity recognition
 - ✓ Deep understandings
 - ✓ Pose estimation

Image Enhancement

Replace those bad marks with its neighboring pixels so that it looks like the neighborhood



“Image Inpainting”, M. Bertalmío et al. 2000.

Image Resizing

less
important



Image Resizing

Traditional resizing uses and stretches the whole image.



Seam Carving:

Content-aware
resizing uses
important areas.

Extends in horizontal
direction and reduces
in vertical.



Image Segmentation

Ref -

<https://gts.ai/how-do-we-solve-the-challenges-faced-due-to-semantic-segmentation/>



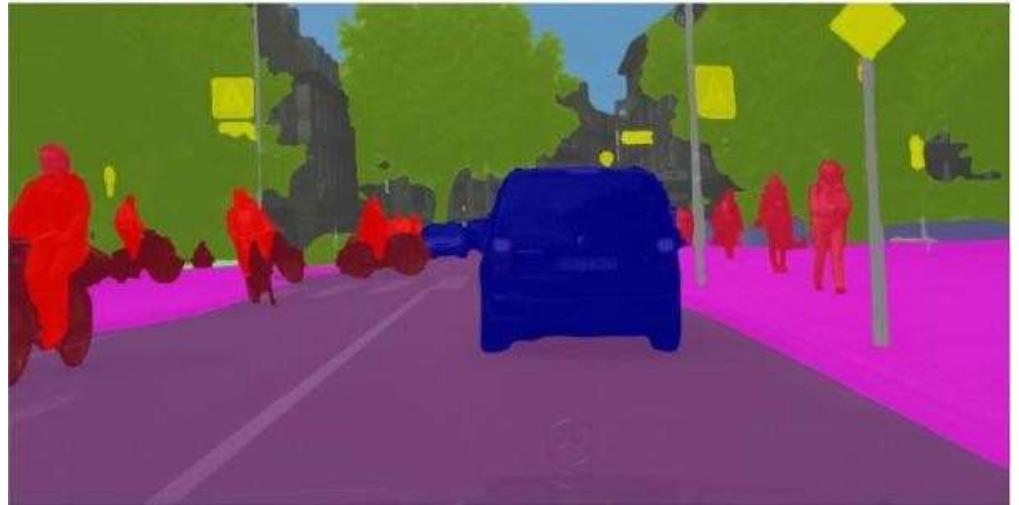
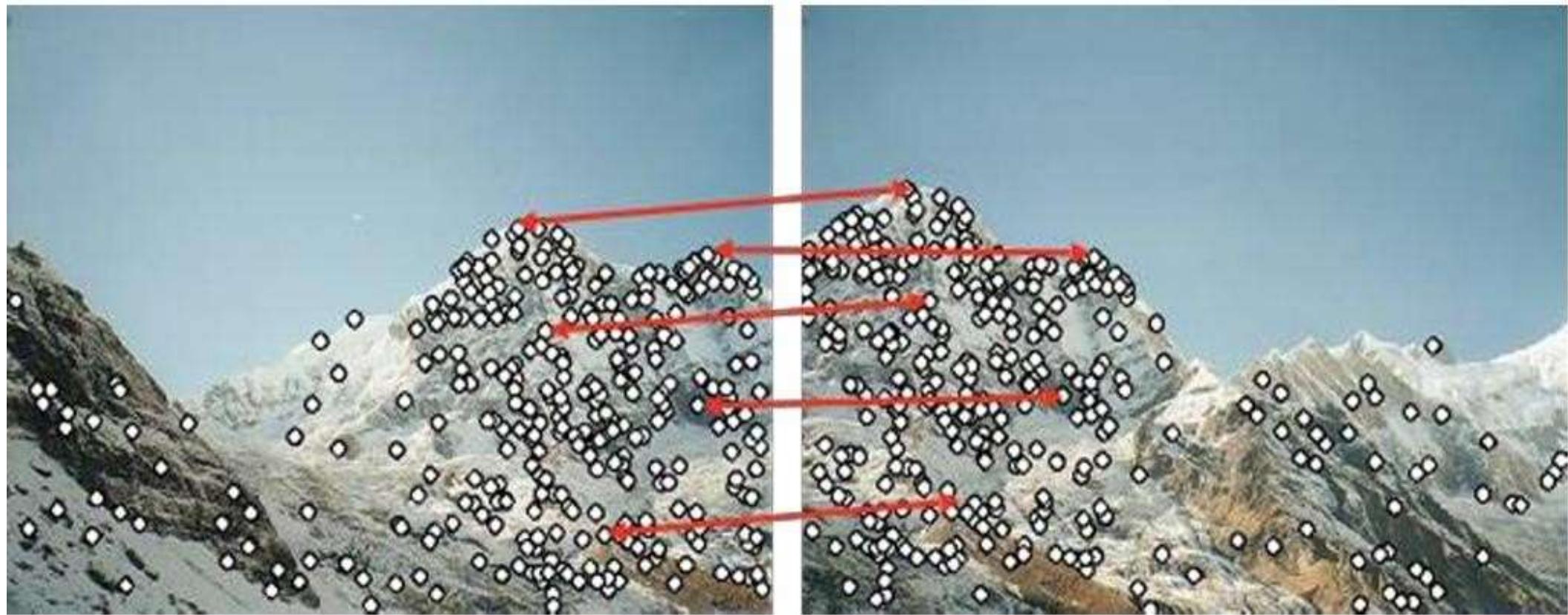
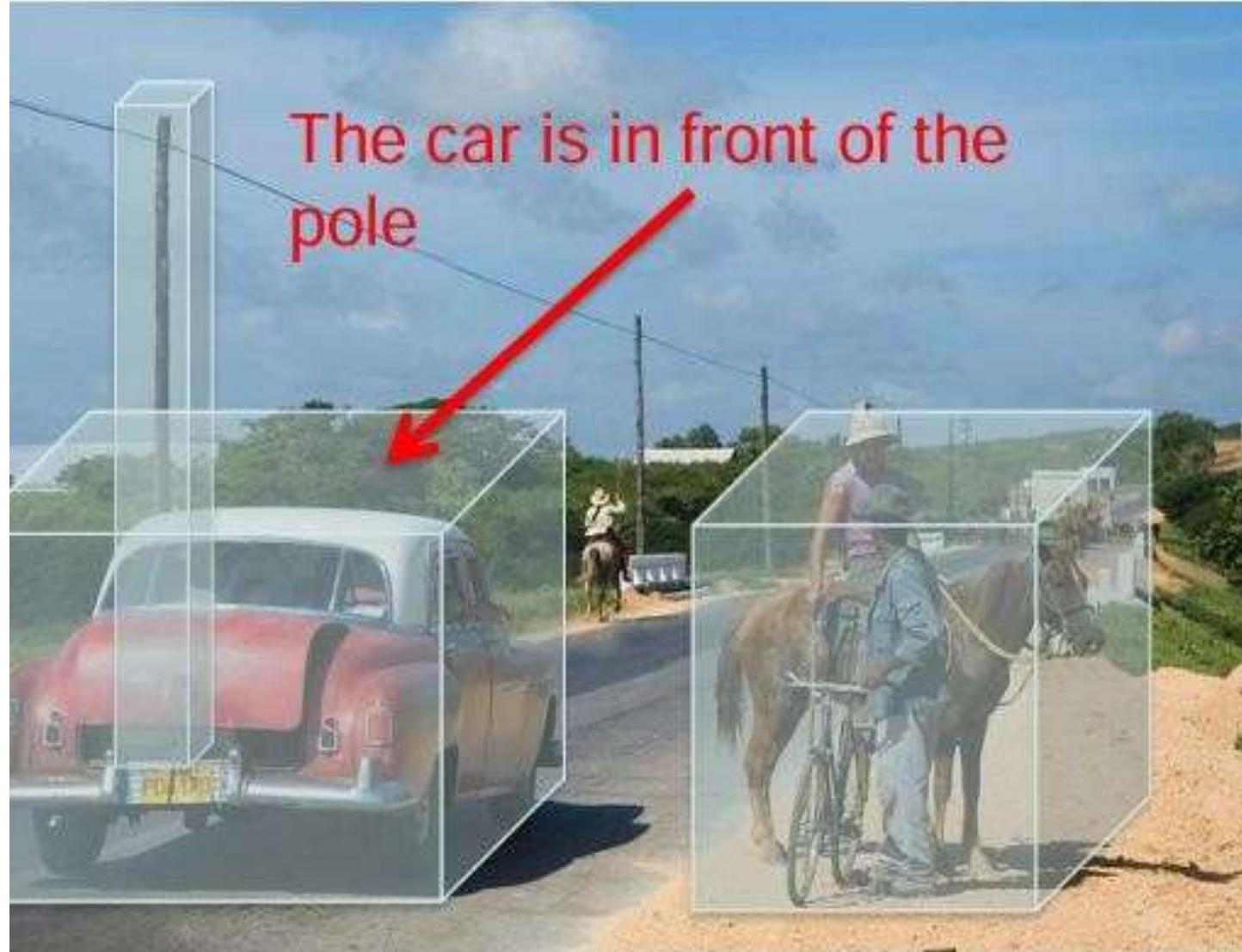


Image Segmentation

Image Features



Computer Vision Tasks



Low Level Vision

- Measurements
- Enhancements
- Region segmentation
- Features

Mid Level Vision

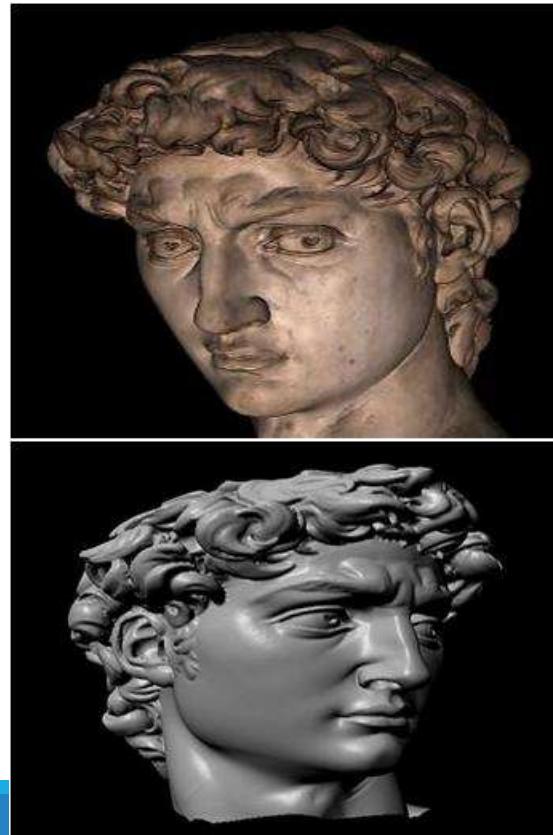
- Reconstruction
- Depth
- Motion Estimation

High Level Vision

- Category detection
- Activity recognition
- Deep understandings
- Pose estimation

Applications: 3D Scanning

- Combining multiple range and color images, allow to reliably and accurately digitize the external shape and surface characteristics of many physical objects



The Digital Michelangelo Project :-

<http://graphics.stanford.edu/projects/mich/>

✓ 2 BILLION polygons, accuracy to .29mm

Apple's 3D maps



Ref -

<http://www.youtube.com/watch?v=lnIVv-LsgZE>

Computer Vision Tasks

Low Level Vision

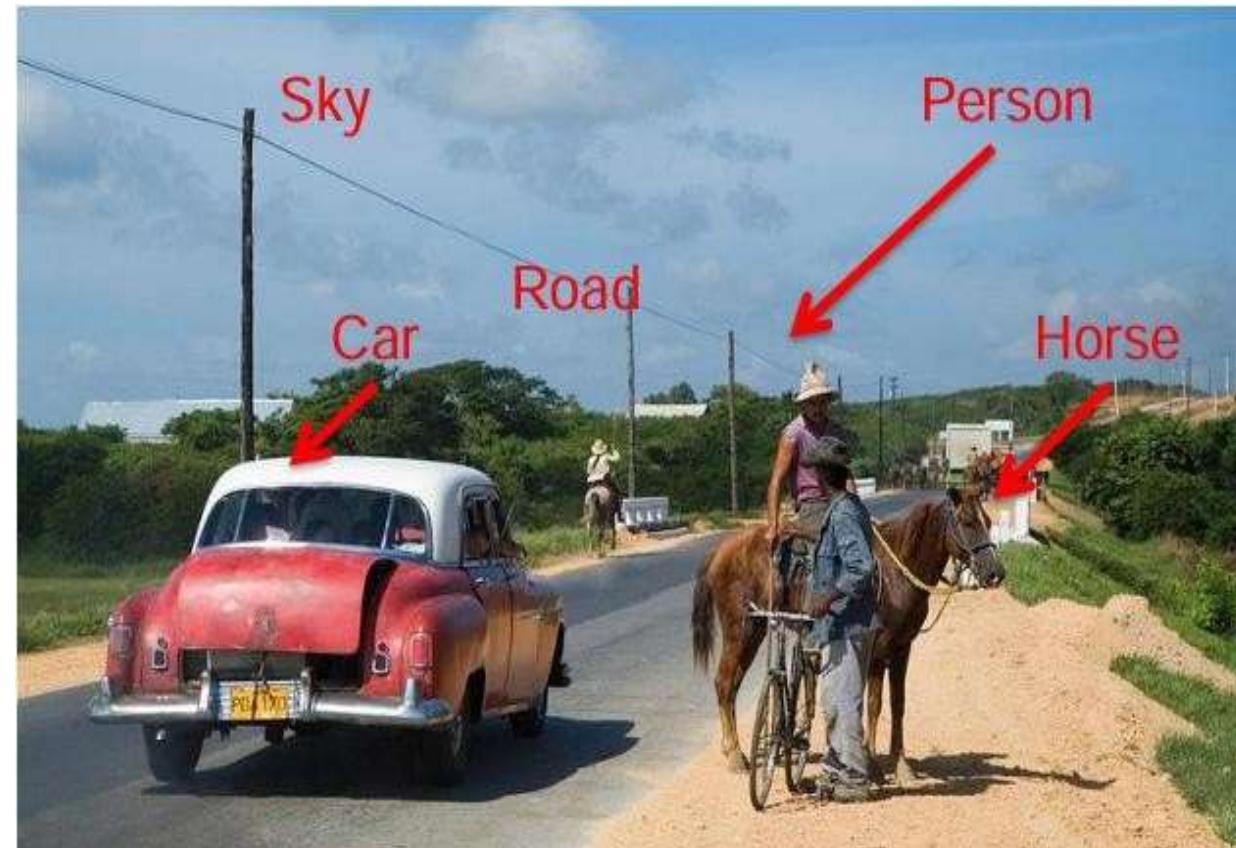
- Measurements
- Enhancements
- Region segmentation
- Features

Mid Level Vision

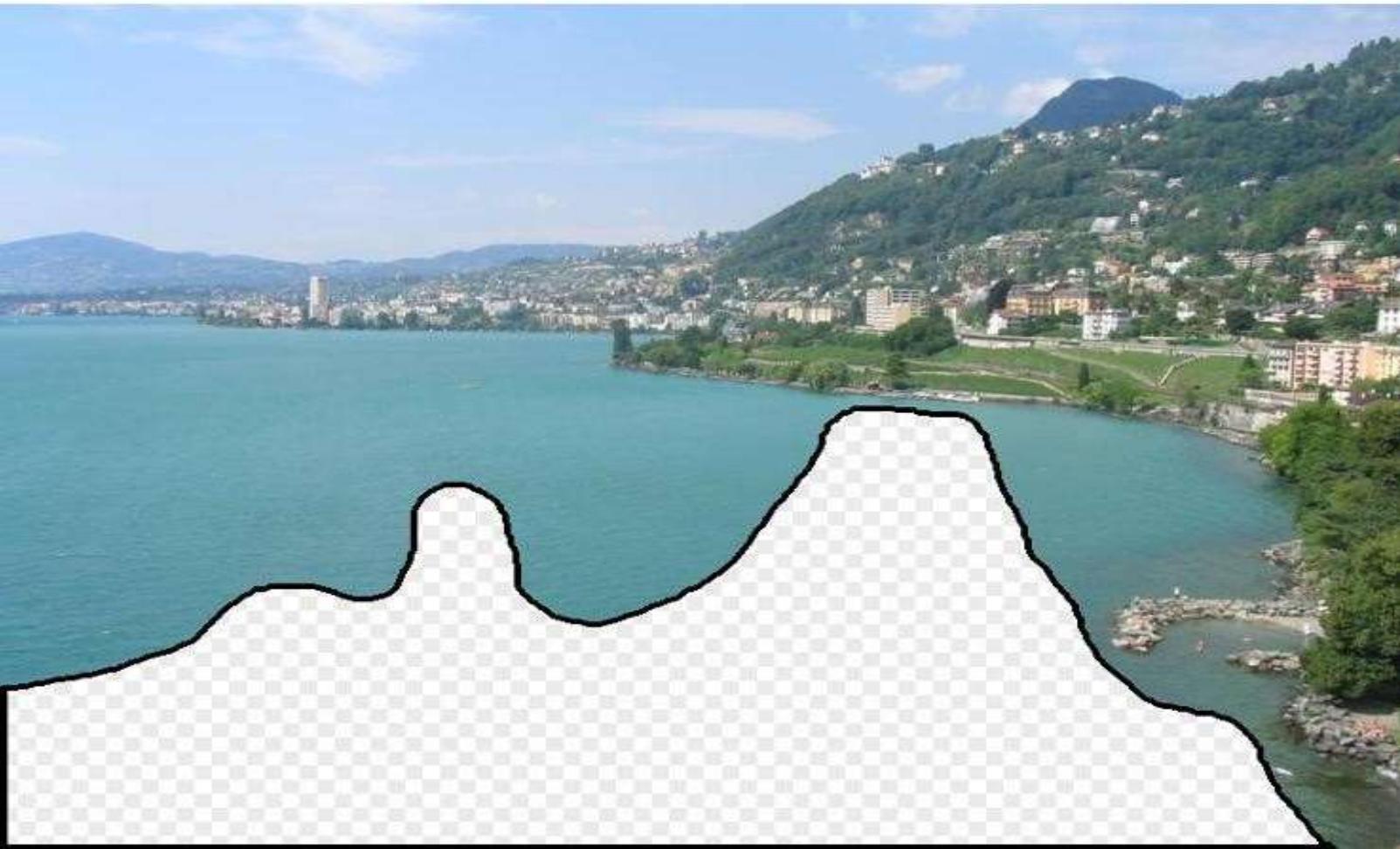
- Reconstruction
- Depth
- Motion Estimation

High Level Vision

- Category detection
- Activity recognition
- Deep understandings
- Pose estimation

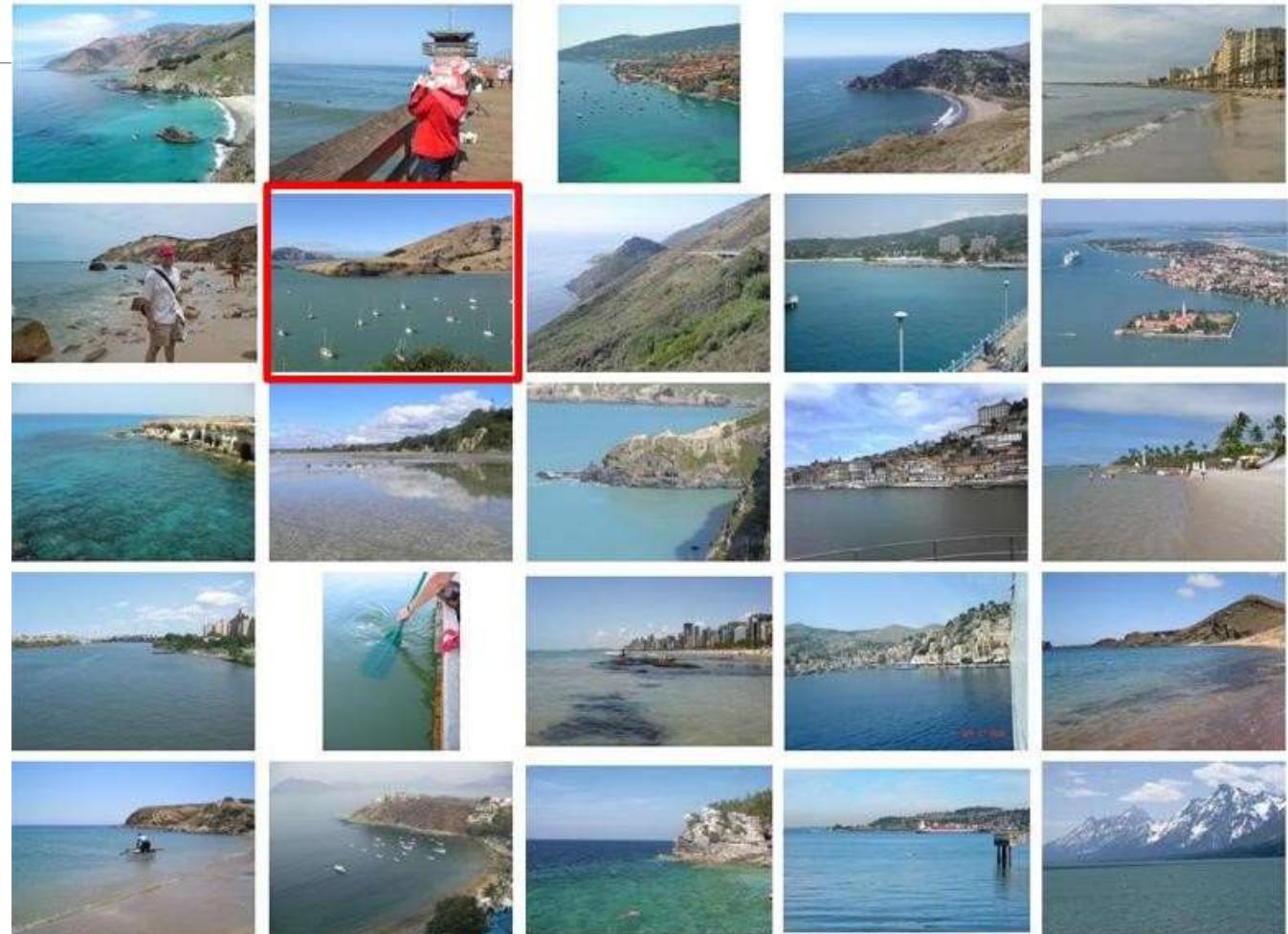
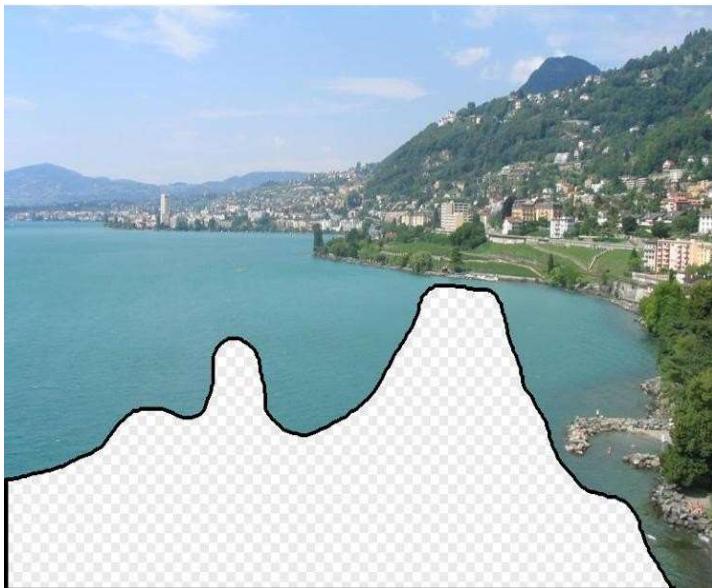


Scene Completion



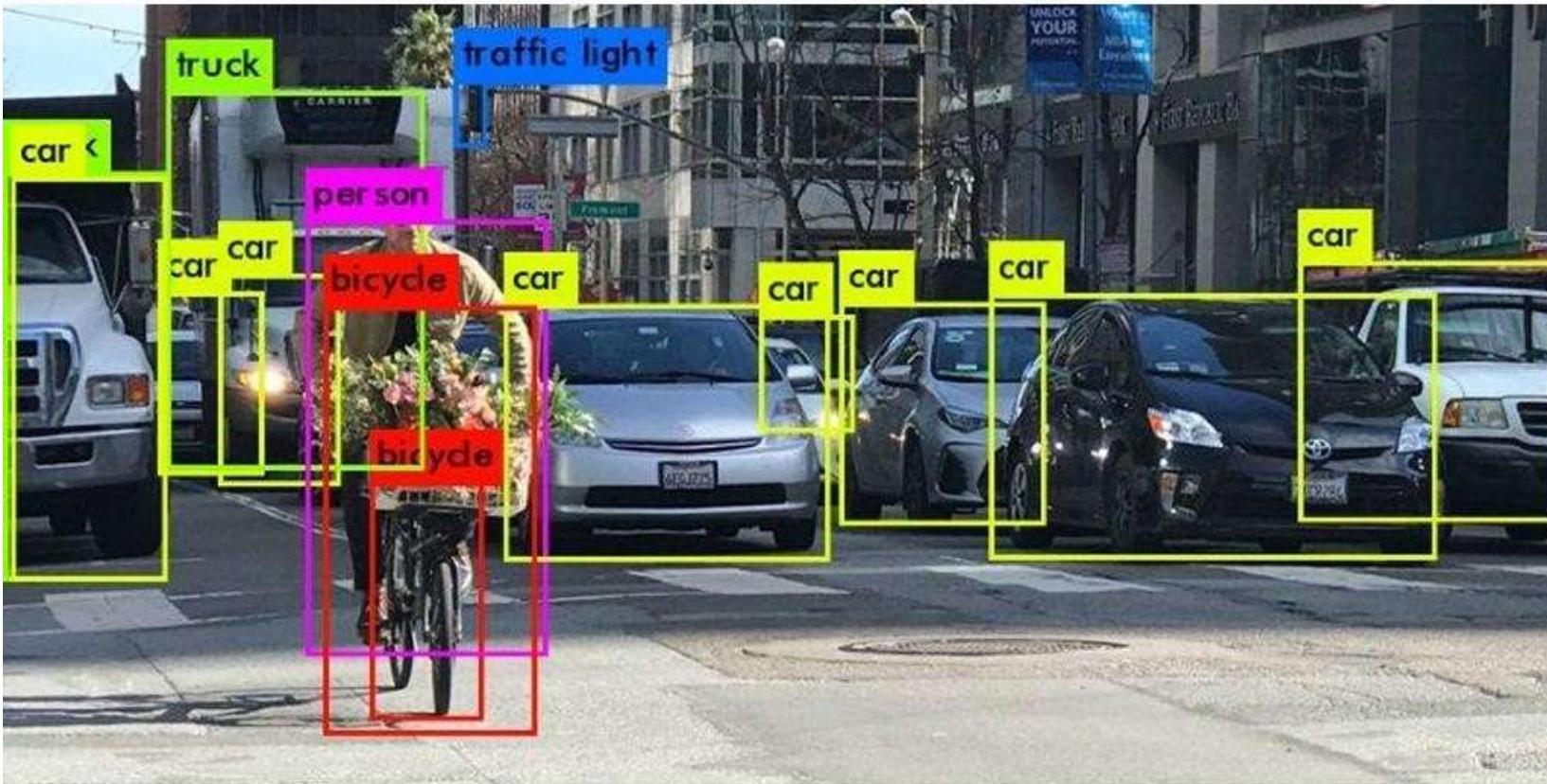
[Hays and Efros. Scene Completion Using Millions of Photographs. SIGGRAPH 2007 and CACM October 2008.]

Scene Completion



Nearest neighbor scenes from database of 2.3 million photos

Object Detection

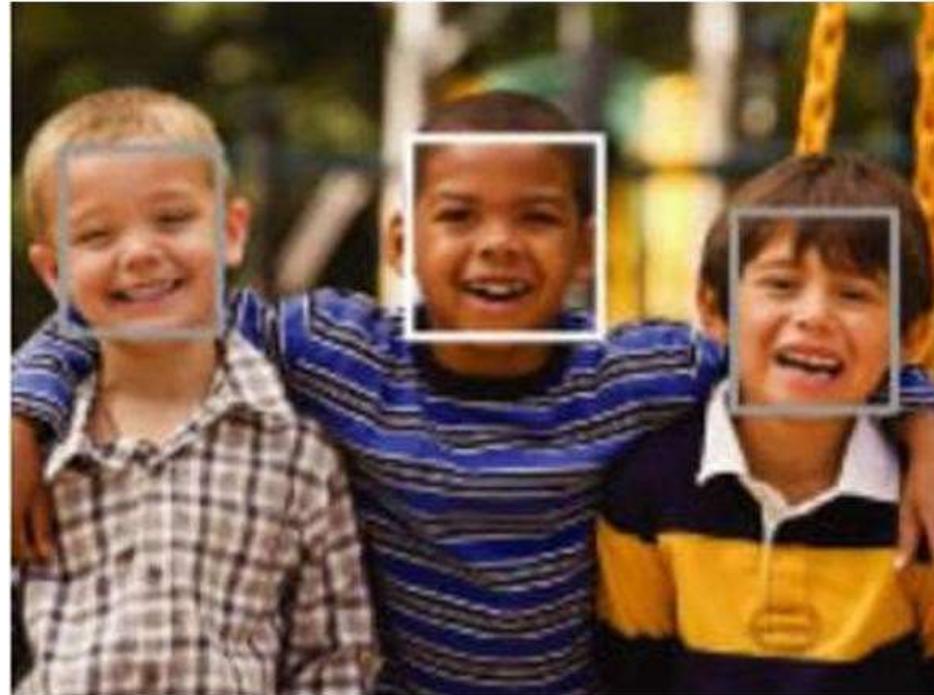


Ref -

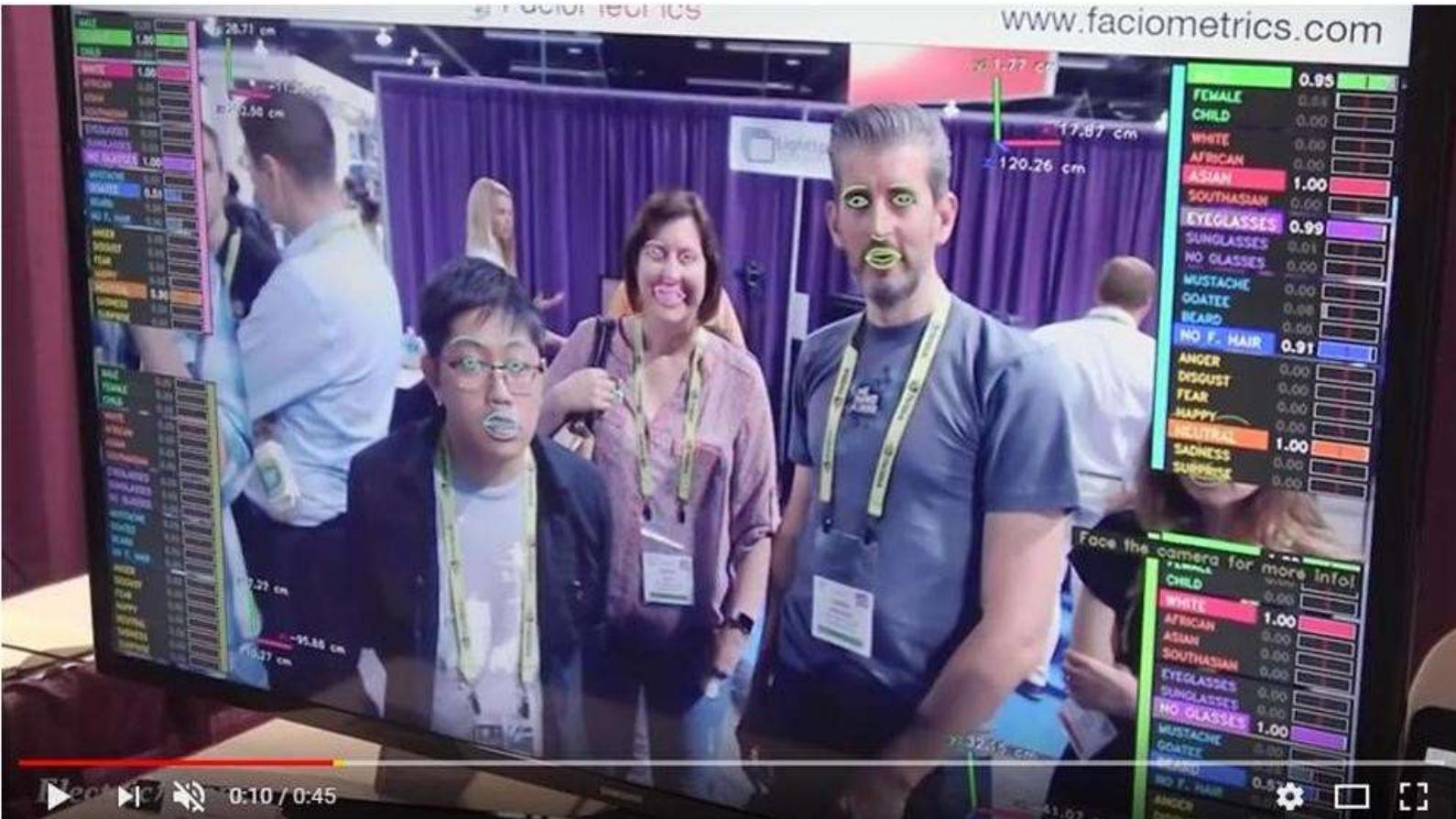
<https://heartbeat.fritz.ai/introduction-to-basic-object-detection-algorithms-b77295a95a63>

Face Detection

- Many new digital cameras now detect faces
 - ✓ Apple, Samsung, Canon, Sony, Fuji, ...



Face Detection



<https://youtu.be/Q0M90dbqs6M>

Smile Detection



Sony Cyber-shot® T70 Digital Still Camera

Object Recognition

Google Lens

Download

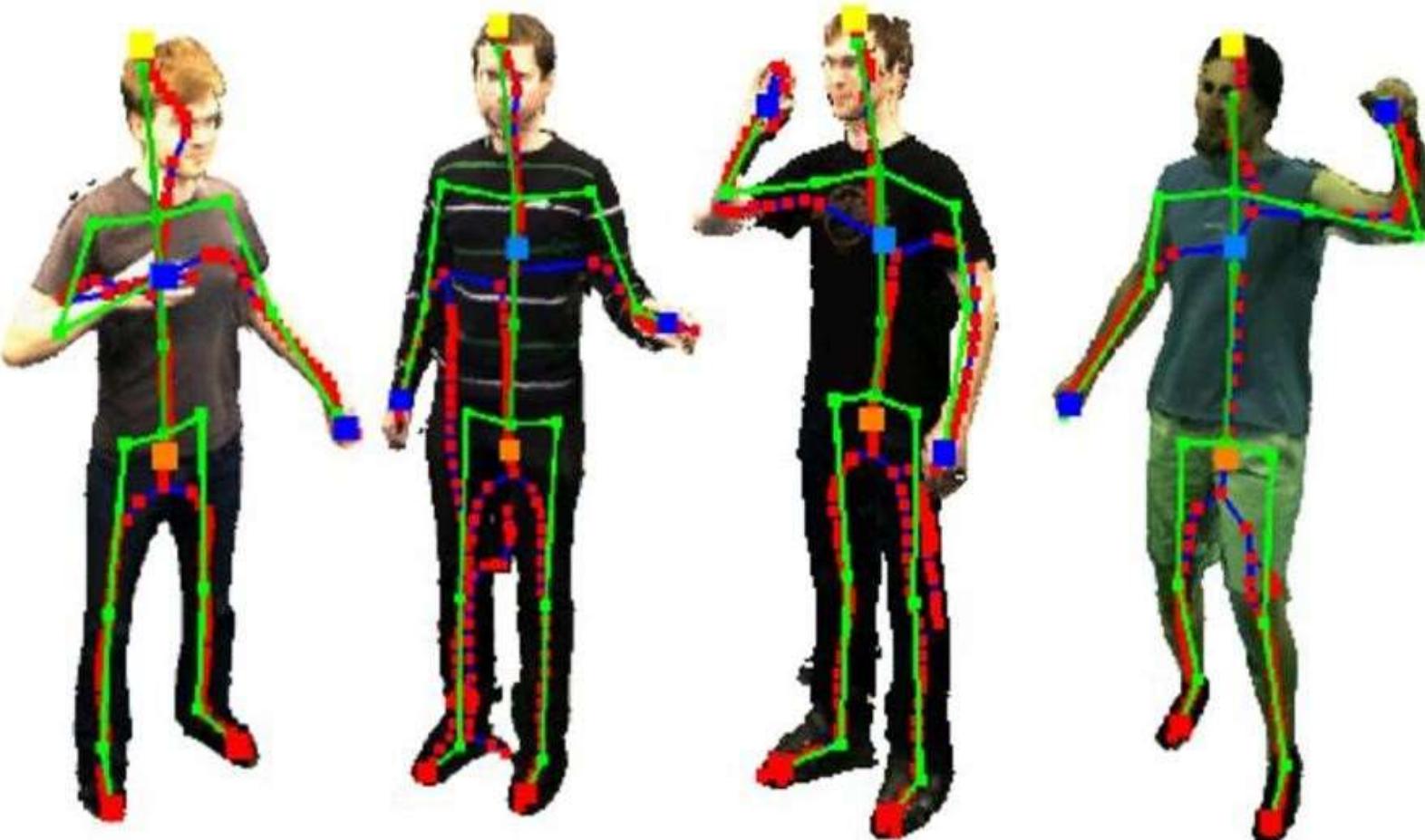
Search what you see

Explore what's around you in an entirely new way.

Tulip
Plants

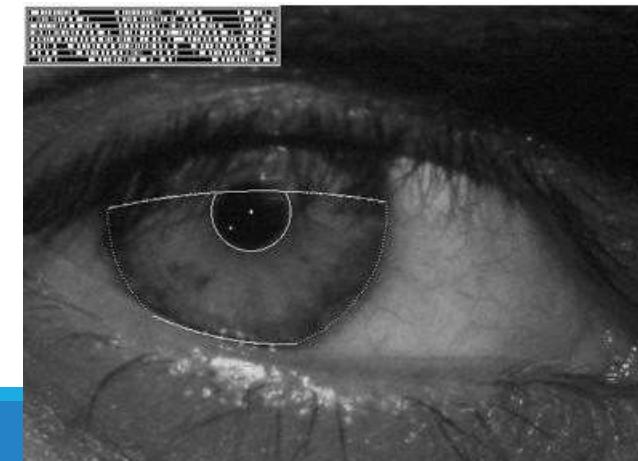
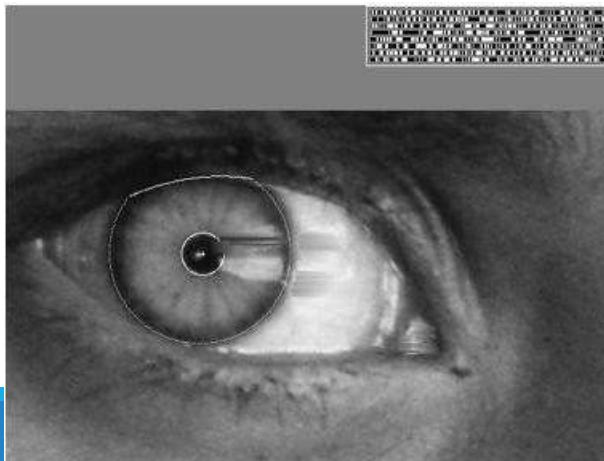
G Search

Pose Estimation



Vision-based biometrics

- “How the Afghan Girl was Identified by Her Iris Patterns” Read the [story wikipedia](#)



Vision-Based Interaction: Xbox Kinect



Interactive Games: Kinect

Object Recognition:

- <http://www.youtube.com/watch?feature=iv&v=fQ59dXOo63o>

Mario:

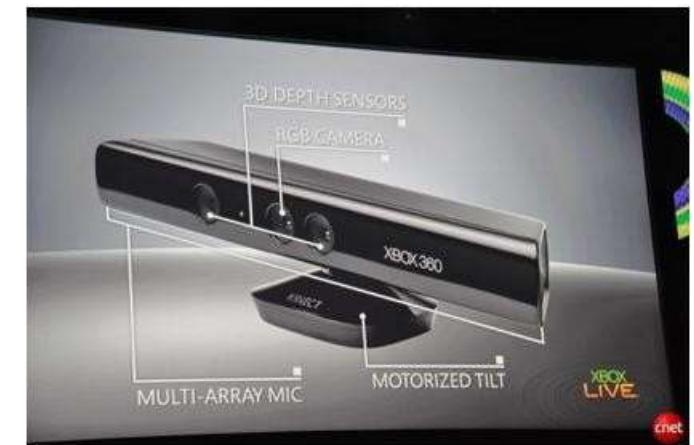
- <http://www.youtube.com/watch?v=8CTJL5IUjHg>

3D:

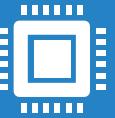
- <http://www.youtube.com/watch?v=7QrnwoO1-8A>

Robot:

- <http://www.youtube.com/watch?v=w8BmgtMKFbY>



Computer Vision Applications



Laptop: Biometrics auto-login (face recognition, 3D), OCR



Smartphones: QR codes, computational photography (Android Lens Blur, iPhone Portrait Mode), panorama construction (Google Photo Spheres), face detection, expression detection (smile), Snapchat filters (face tracking), Google Tango (3D reconstruction), Night Sight (Pixel)



Web: Image search, Google photos (face recognition, object recognition, scene recognition, geolocalization from vision), Facebook (image captioning), Google maps aerial imaging (image stitching), YouTube (content categorization)



VR/AR: Outside-in tracking (HTC VIVE), inside out tracking (simultaneous localization and mapping, HoloLens), object occlusion (dense depth estimation)

Computer Vision Applications



Motion: Kinect, full body tracking of skeleton, gesture recognition, virtual try-on



Medical imaging: CAT / MRI reconstruction, assisted diagnosis, automatic pathology, connectomics, endoscopic surgery



Industry: Vision-based robotics (marker-based), machine assisted router (jig), automated post, ANPR (number plates), surveillance, drones, shopping



Transportation: Assisted driving (everything), face tracking/iris dilation for drunkenness, drowsiness, automated distribution



Media: Visual effects for film, TV (reconstruction), virtual sports replay (reconstruction), semantics-based auto edits

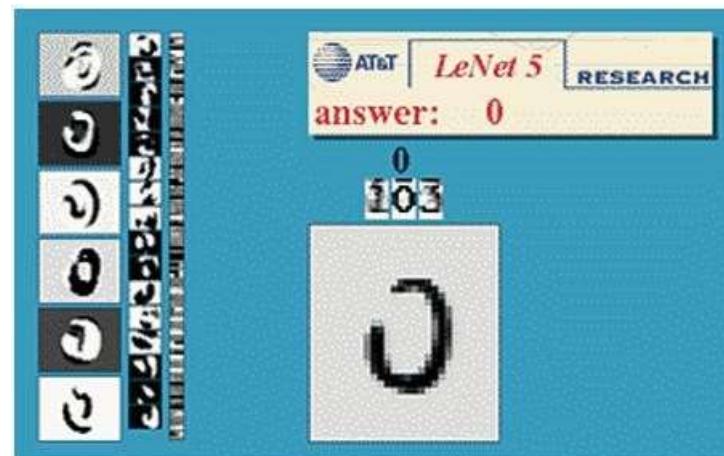
Computer Vision Applications

- Optical Character Recognition (OCR): Technology to convert images of text into text.
Scanners come with OCR software.



License plate readers

http://en.wikipedia.org/wiki/Automatic_number_plate_recognition



Mail digit recognition, AT&T labs



Live
Camera
Translation

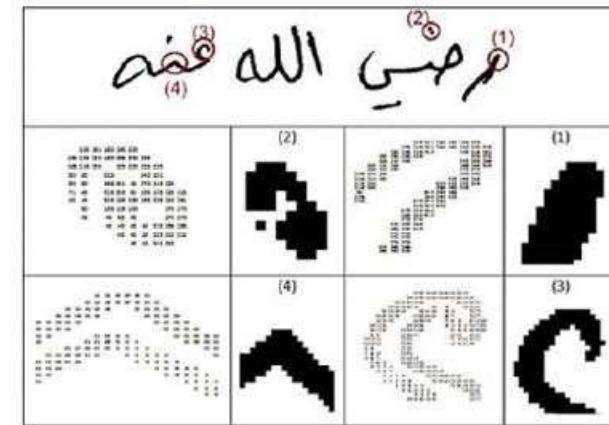


Computer Vision Applications

- Writer Identification

Sameh Awaida and Sabri Mahmoud, "Writer Identification of Arabic Text using Statistical and Structural Features", Cybernetics and Systems: An International Journal, Volume 44, Issue 1, pp. 57-76, 2012

Authors	Language	No. of writers	Top-1 (%)	Top-5 (%)	Top-10 (%)
Schomaker et al. (2003)	Dutch	250	88.0	98.0	99.0
B. Zhang (2003)	English	1,000	98.0	—	—
Bulacu and Schomaker (2005)	Dutch	250	78.0	—	92.6
Bulacu, Schomaker et al. (2007)	Arabic	350	88.0	—	99.0
He et al. (2008)	Chinese	500	39.2	62.4	77.2
Siddiqi et al. (Siddiqi and Vincent 2009)	English	200	86.0	—	—
Our approach	Arabic	100	88.0	96.0	98.5
Our approach (remaining 150 writers)	Arabic	150	84.0	96.3	97.3
Our approach	Arabic	250	75.0	91.8	95.4
Our approach (with rejection)	Arabic	250	80.5	94.0	97.0



السب الأسماء في المتن
المجتمع في التسلاج والنقل
من المعاشرة
أصوات
وخبرات الحياة الاجتماعية السامية.
الاحتفايات الخاصة مختلفة الممارسات والاتصالات والقيم
والمهارات التي تجعل لكتابه الإيجابية العالية
في مختلف أطيافه وفصالات الحياة الإنسانية إلى أقصى
حد توصله لهم إمكاناته وعمره القم انتهاء

Computer Vision Applications

- Automated Visual Inspection



Computer Vision Applications

- Object Recognition

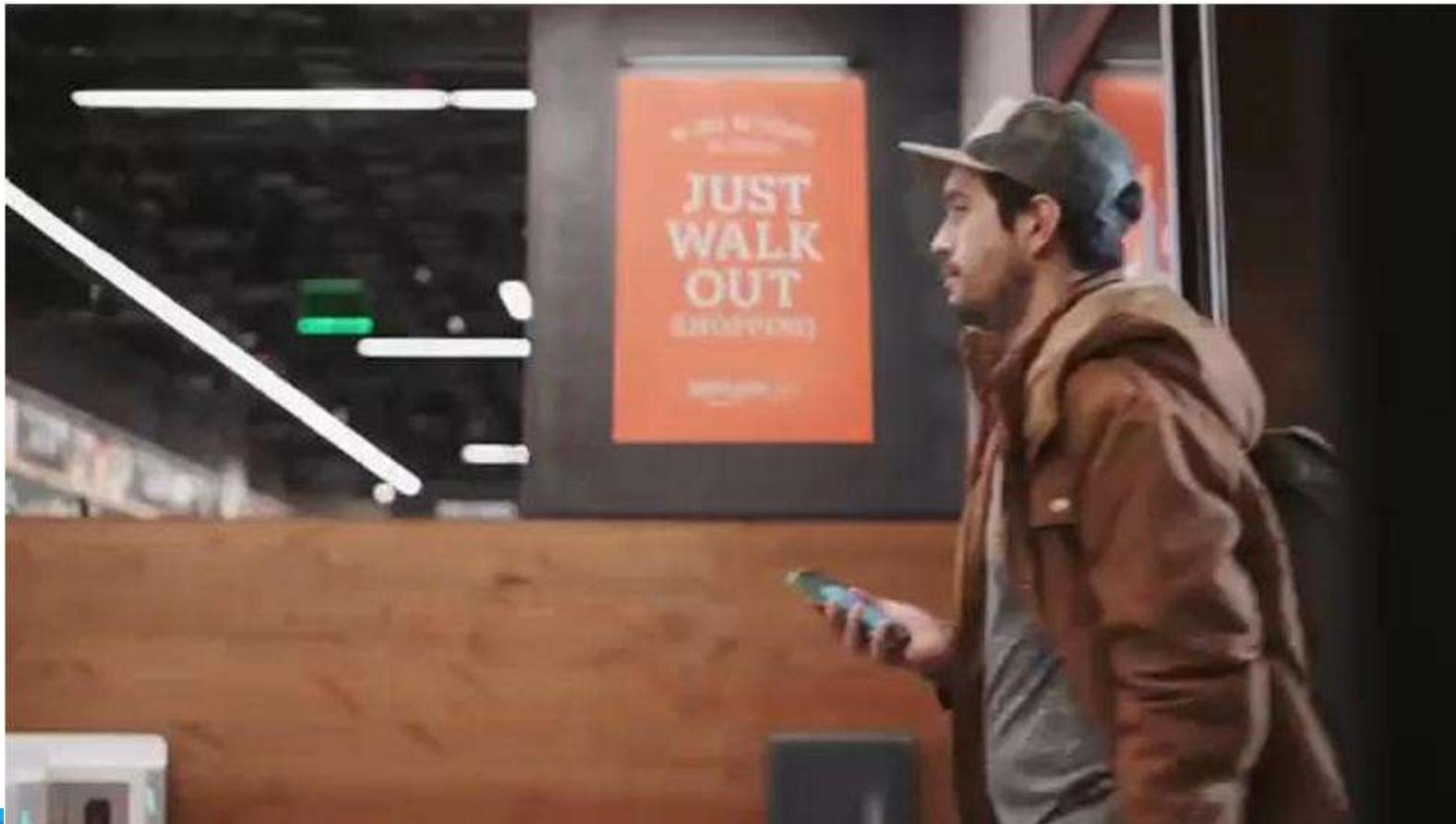


Toshiba Tech
IS-910T 2013

DataLogic
LaneHawk
LH4000 2012

Computer Vision Applications

- Amazon Go



Computer Vision Applications

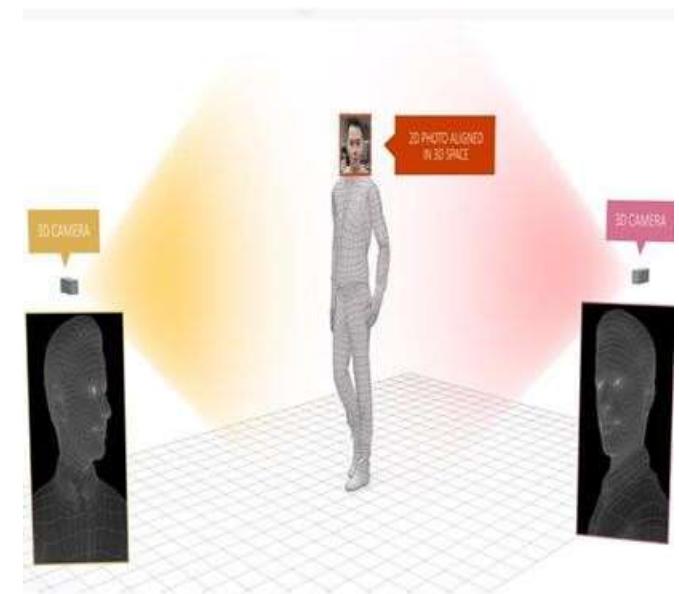
- Vision-based Biometrics



Fingerprint scanners on many new laptops, other devices



Apple Face ID



Face recognition systems now beginning to appear more widely
<http://www.sensiblevision.com/>

Computer Vision Applications

- Human Shape Capture



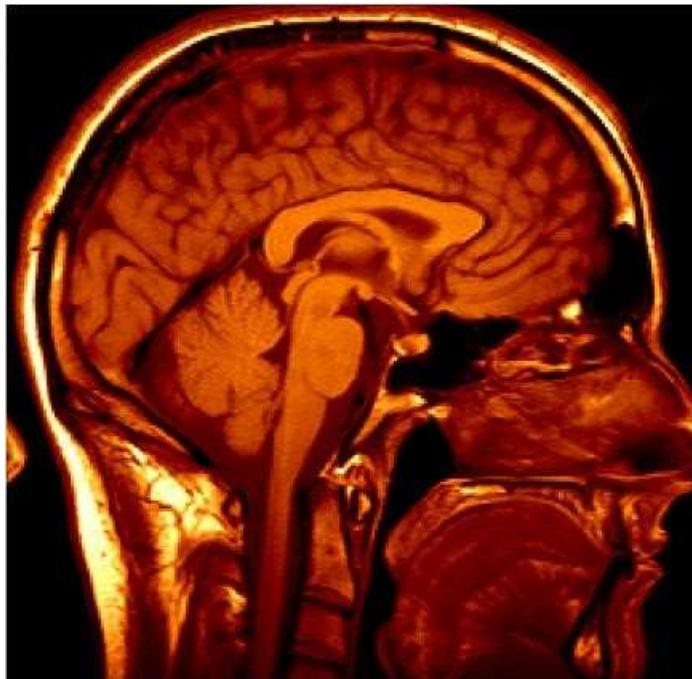
Computer Vision Applications

- Motion Capture



Computer Vision Applications

- Medical Imaging



3D imaging
MRI, CT



Image guided surgery

Computer Vision Applications

- Virtual Reality



Oculus Quest



Beat Saber

Computer Vision Applications

- Augmented Reality



Microsoft Hololens 2

Computer Vision Applications

- Body Tracking



ARKit 3

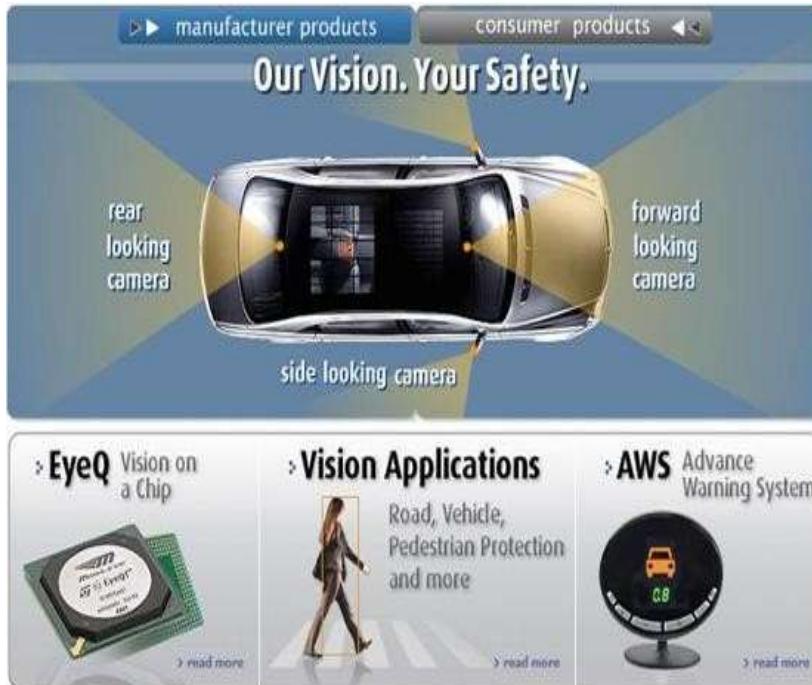


Computer Vision Applications

- Smart Cars

Mobileye

Vision systems currently in high end BMW, GM, Volvo models



BMW night vision

Computer Vision Applications

- AutoCars: Self Driving Cars



<https://waymo.com/tech/>

Computer Vision Applications

- Industrial Robots

- ✓ Vision-guided robots position nut runners on wheels



Computer Vision Applications

- Human Detection



Computer Vision Applications

- Video Surveillance and Monitoring



Computer Vision Applications

- Forest fire monitoring system



Early warning of forest fires

Adapted from Enis Cetin, Bilkent University

Computer Vision Applications

- Counting in Extremely Dense Crowd Images

Ground truth=634

Proposed Method by Idrees and Shah=640



Computer Vision Applications

- Counting in Extremely Dense Crowd Images

Ground Truth = 1428

Proposed Method = 1468



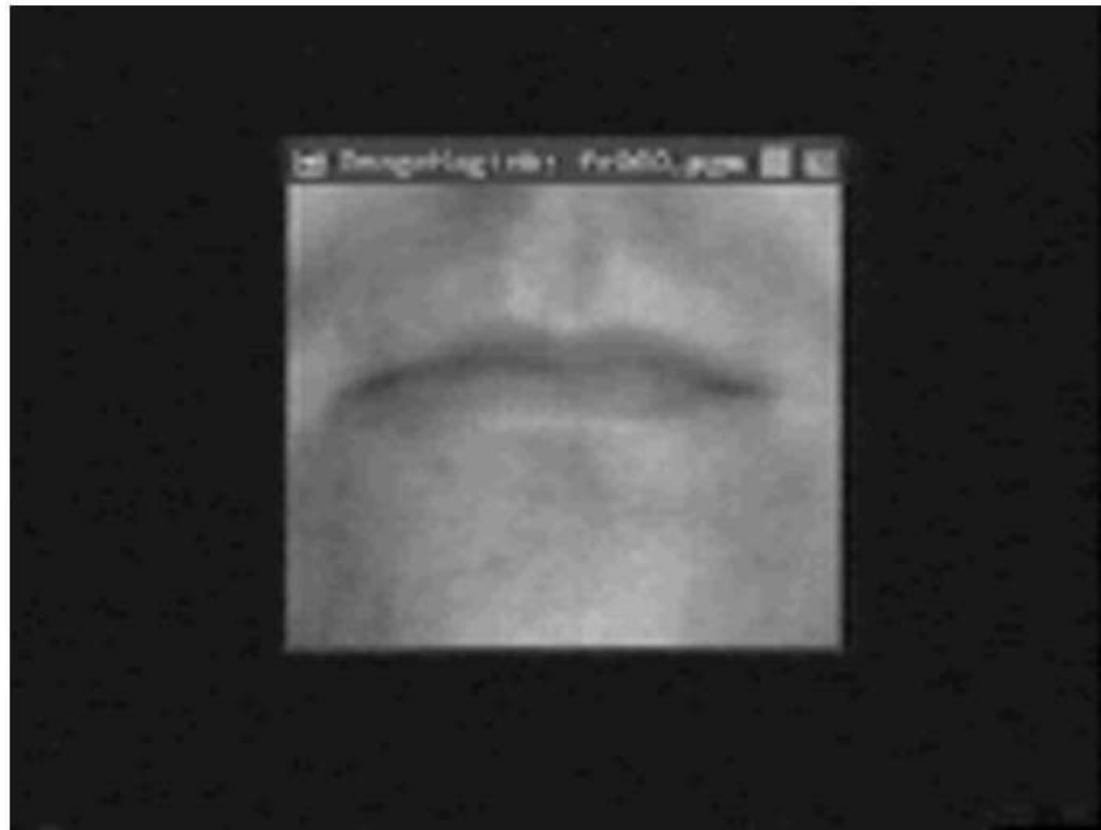
Computer Vision Applications

- Fatigue detection



Computer Vision Applications

- Lip reading



Computer Vision Applications

- Human Action Recognition

9 actions, 142 videos (UCF database, Shah).



Bench Swing



Dive



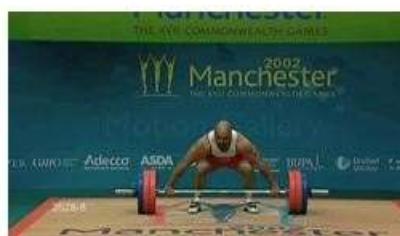
Swing



Run



Kick



Lift



Ride



Golf Swing



Skate

Computer Vision Applications

- Morphing

It is a special effect in motion pictures and animations that changes (or morphs) one image or shape into another through a seamless transition.



Python Image Processing Libraries

OpenCV: Widely used for computer vision tasks, provides functions for manipulation, filtering, feature detection, and object recognition.

Pillow: Easy-to-use for basic tasks like resizing, cropping, rotating, and applying filters; supports multiple image formats.

scikit-image: Collection of algorithms for tasks such as segmentation, feature extraction, and image restoration.

Mahotas: Offers functions for filtering, feature extraction, morphology, and more.

SimpleITK: Interface to ITK for medical image analysis, including registration and segmentation.

PyTesseract: Wrapper for Google's Tesseract-OCR Engine, used for optical character recognition.

imageio: Supports reading/writing various image formats and animated images.

matplotlib/seaborn: Notable for data visualization but can be utilized for basic image visualization and manipulation.

GDAL (Geospatial Data Abstraction Library): Powerful for working with raster and vector geospatial data

Working with Images

```
from PIL import Image, ImageFilter, ImageOps
import numpy as np
import matplotlib.pyplot as plt

# Open an image file
image = Image.open("test.jpg")

# Convert the image to a NumPy array
image_array = np.array(image)

# Manipulate the pixel values (e.g., invert colors)
manipulated_image_array = 255 - image_array # Inverts colors

# Convert the manipulated NumPy array back to an image
manipulated_image = Image.fromarray(manipulated_image_array)

# Rotate the image by 90 degrees clockwise
rotated_image = image.rotate(90)

# Resize the image to half its original size
resized_image = image.resize((image.width // 2, image.height // 2))

# Apply a Gaussian blur filter to the image
blurred_image = image.filter(ImageFilter.GaussianBlur(2))

# Apply a contour filter to the image
contour_image = ImageOps.invert(image.filter(ImageFilter.CONTOUR))
```

```
# Display all images
fig, axes = plt.subplots(2, 3, figsize=(15, 10))
fig.suptitle('Image Processing with Pillow', fontsize=16)

axes[0, 0].imshow(image)
axes[0, 0].set_title('Original Image')
axes[0, 0].axis('off')

axes[0, 1].imshow(rotated_image)
axes[0, 1].set_title('Rotated Image')
axes[0, 1].axis('off')

axes[0, 2].imshow(resized_image)
axes[0, 2].set_title('Resized Image')
axes[0, 2].axis('off')

axes[1, 0].imshow(blurred_image)
axes[1, 0].set_title('Blurred Image')
axes[1, 0].axis('off')

axes[1, 1].imshow(manipulated_image)
axes[1, 1].set_title('Manipulated Image')
axes[1, 1].axis('off')

axes[1, 2].imshow(contour_image)
axes[1, 2].set_title('Contour Image')
axes[1, 2].axis('off')

plt.show(block=True)
```

Image Processing

Original Image



Rotated Image



Resized Image



Blurred Image



Manipulated Image



Contour Image



OpenCV

```
import cv2
import numpy as np
import matplotlib.pyplot as plt
# Open an image file
image = cv2.imread("test2.jpg")
orig_image=image.copy()
# Convert the image to grayscale
gray_image = cv2.cvtColor(image, cv2.COLOR_BGR2GRAY)

# Initialize the SIFT detector
sift = cv2.SIFT_create()
# Detect keypoints in the image
keypoints = sift.detect(gray_image, None)
# Draw keypoints on the original image
image_with_keypoints = cv2.drawKeypoints(image, keypoints, None, flags=cv2.DRAW_MATCHES_FLAGS_DRAW_RICH_KEYPOINTS)

# Apply thresholding to the image
_, thresholded_image = cv2.threshold(gray_image, 127, 255, cv2.THRESH_BINARY)

# Rotate the image by 90 degrees clockwise
height, width = image.shape[:2]
rotation_matrix = cv2.getRotationMatrix2D((width / 2, height / 2), 90, 1)
rotated_image = cv2.warpAffine(image, rotation_matrix, (width, height))
# Add text on top of the rotated image
image_with_text = cv2.putText(rotated_image, 'Sample Text', (50, 50), cv2.FONT_HERSHEY_SIMPLEX, 1, (255, 255, 255), 2)

# Calculate histogram of the grayscale image
hist = cv2.calcHist([gray_image], [0], None, [256], [0, 256])

# Apply face detection to the image
# Load the pre-trained face detection cascade
face_cascade = cv2.CascadeClassifier(cv2.data.haarcascades + 'haarcascade_frontalface_default.xml')
faces = face_cascade.detectMultiScale(gray_image, scaleFactor=1.1, minNeighbors=5, minSize=(30, 30))
# Draw bounding boxes around detected faces
for (x, y, w, h) in faces:
    cv2.rectangle(image, (x, y), (x+w, y+h), (0, 255, 0), 2)
```

OpenCV

Original Image



Rotated Image

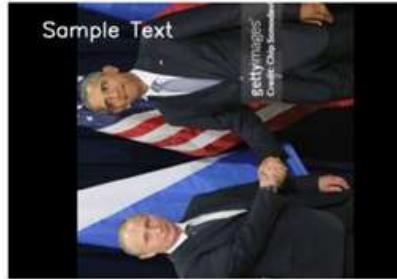
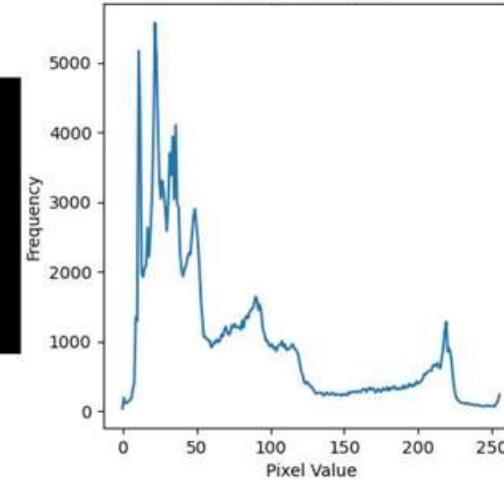


Image Histogram



Thresholded Image



Keypoint Detection (SIFT)



Face Detection



Pytesseract

- <https://github.com/UB-Mannheim/tesseract/wiki>
- pip install pytesseract

Noisyimage
to test
Tesseract OCR

Noisy image
to test
Tesseract OCR

```
import pytesseract
from PIL import Image
import matplotlib.pyplot as plt

# Path to the Tesseract executable (change this according to your system)
pytesseract.pytesseract.tesseract_cmd = r'C:\Program Files\Tesseract-OCR\tesseract.exe'

# Load the image
image_path = 'ocr.png'
image = Image.open(image_path)

# Perform OCR on the image
text = pytesseract.image_to_string(image)

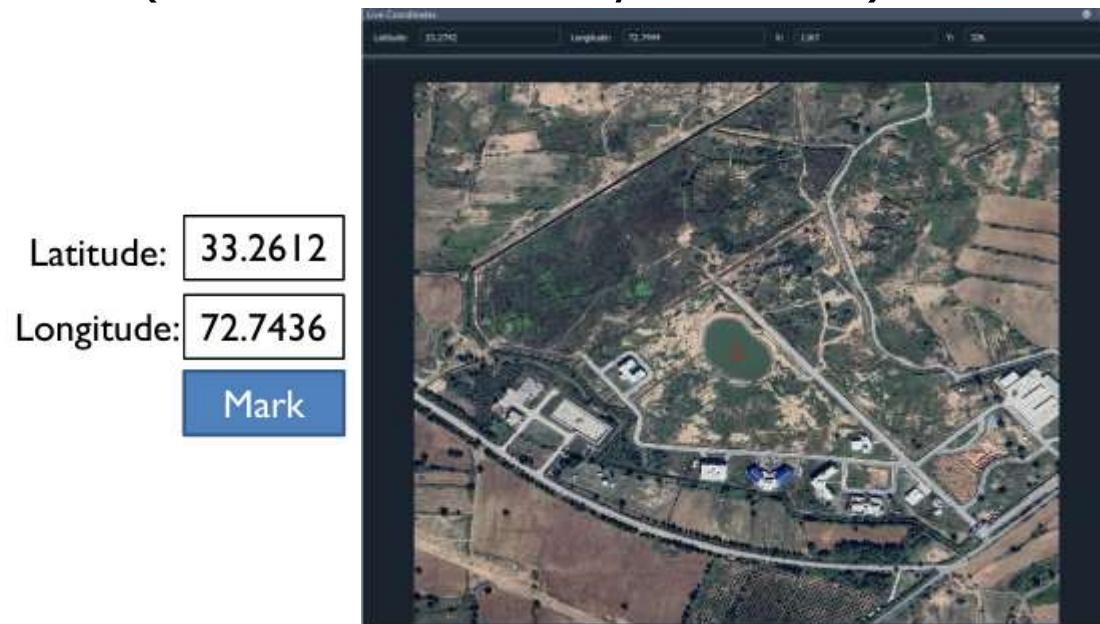
# Display the image
plt.figure(figsize=(8, 6))
plt.imshow(image)
plt.axis('off')

# Display extracted text on top of the image
plt.text(0, 0, text, color='red', fontsize=12, verticalalignment='top')

plt.show()
```

Assignment-1: Working with GeoTIFF Images Using GDAL

- GeoTIFF images encode raster data with geospatial metadata, enabling precise linking of each pixel to specific latitude and longitude coordinates. These images are widely used in GIS (Geographic Information System) applications for spatial analysis and visualization.
- You are provided with a sample GeoTIFF file (or download one of your choice).



Assignment-1: Working with GeoTIFF Images Using GDAL

- Your task is to perform the following steps using the GDAL library and a GUI framework like PyQt or Tkinter:

Task-A: Load and Display the Satellite Image

- Load the GeoTIFF file using the GDAL library.
- Display the satellite image in a window.

Task-B: Interactive Mouse Tracking

- Create a GUI that displays the image. As the user moves the mouse cursor over the image:
 - Show the pixel coordinates (X,Y) of the cursor in two text boxes.
 - Display the corresponding latitude and longitude of the pixel in two additional text boxes.
 - Update this information in real time.

Task-C: Marking a Specific Location

Add two input text boxes where users can enter latitude and longitude values.

Include a "Mark" button in the GUI:

When clicked, the button should plot a cross at the specified location on the displayed image based on the entered latitude and longitude.

Task-D (Bonus): Zoom and Pan Functionality

- Enhance the GUI with:
 - Zooming functionality using the mouse wheel.
 - Panning functionality via mouse drag.

Thank you