

# Computer Vision: Part 1 (2D CV)

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# Lecture Content

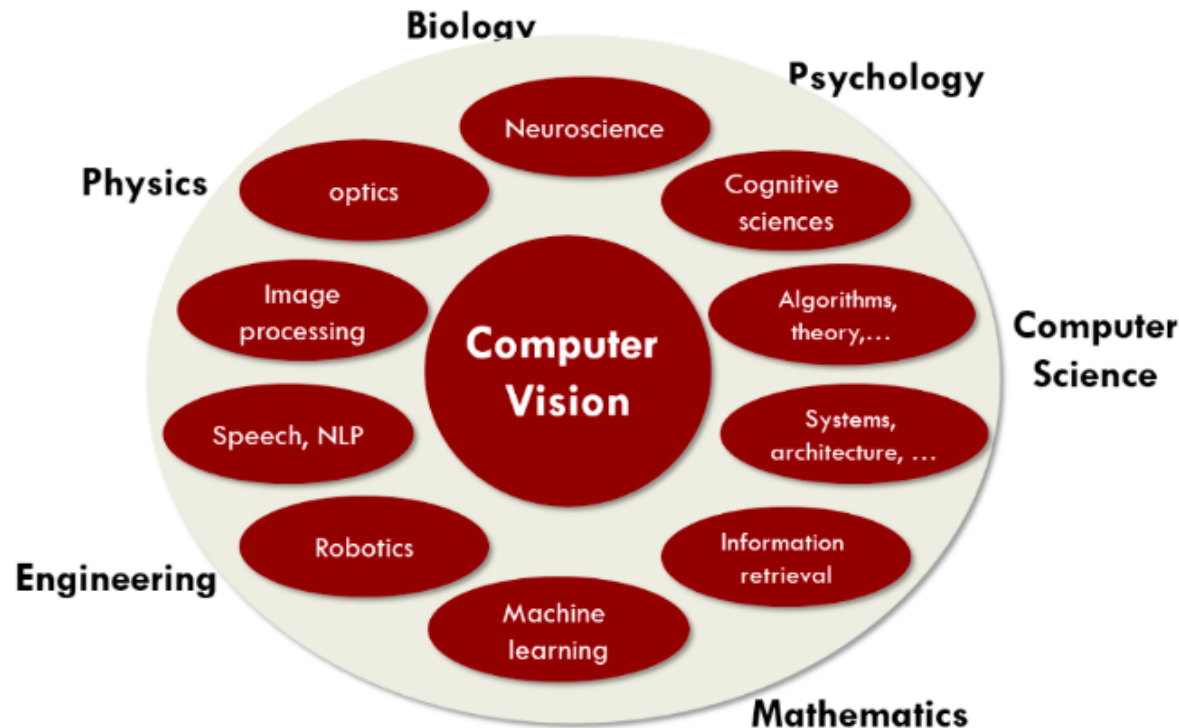
- ▶ What is Computer Vision?
- ▶ What is Image?
- ▶ Feature Detection and Matching
- ▶ Classification, Object Detection, Segmentation and Pose Estimation
- ▶ Visual Tracking
- ▶ Computer Vision Applications
- ▶ Conclusion
- ▶ References

# Session Outcomes

- ▶ Acquire basic knowledge of computer vision as an interdisciplinary field.
- ▶ Understand feature detection and matching, and its applications.
- ▶ Gain basic understanding of Computer vision tasks.

# What is Computer vision?

- **Computer vision** is an **interdisciplinary field** that deals with how computers can be made to gain high-level understanding from **digital images** or **videos**. It is a sub-field of AI.



# What is image?



What we see



```
0 2 15 0 0 11 10 0 0 0 0 9 9 0 0 0
0 0 0 4 60 157 236 255 255 177 95 61 32 0 0 29
0 10 16 119 238 255 244 245 243 250 249 255 222 103 10 0
0 14 170 255 255 244 254 255 253 245 255 249 253 251 124 1
2 98 255 228 255 251 254 211 141 116 122 215 251 238 255 49
13 217 243 255 155 33 226 52 2 0 10 13 232 255 255 36
16 229 252 254 49 12 0 0 7 7 0 70 237 252 235 62
6 141 245 255 212 25 11 9 3 0 115 236 243 255 137 0
0 87 252 250 248 215 60 0 1 121 252 255 248 144 6 0
0 13 113 255 255 245 255 182 181 248 252 242 208 36 0 19
1 0 5 117 251 255 241 255 247 255 241 162 17 0 7 0
0 0 0 4 58 251 255 246 254 253 255 120 11 0 1 0
0 0 4 97 255 255 255 248 252 255 244 255 182 10 0 4
0 22 206 252 246 251 241 100 24 113 255 245 255 194 9 0
0 111 255 242 255 158 24 0 0 6 39 255 232 230 56 0
0 218 251 250 137 7 11 0 0 0 2 62 255 250 125 3
0 173 255 255 101 9 20 0 13 3 13 182 251 245 61 0
0 107 251 241 255 230 98 55 19 118 217 248 253 255 52 4
0 18 146 250 255 247 255 255 255 249 255 240 255 129 0 5
0 0 23 113 215 255 250 248 255 255 248 248 116 14 12 0
0 0 6 1 0 52 153 233 255 252 147 37 0 0 4 1
0 0 5 5 0 0 0 0 0 14 1 0 6 6 0 0
```

What a computer sees

# What is image?

- ▶ Images are just **numbers**. i.e. a matrix of 2-dimensional numbers in this case (gray scale image).
- ▶ An image is made of **pixels**. Pixel value: 0 – 255.
- ▶ A **video** is a sequence of frames (images).



BW



Gray



RGB



# What is image?

- ▶ RGB color image has 3 channels:



Red



Green



Blue



# Feature Detection and Matching

- ▶ A pair of images to be matched. What kinds of **features** might one use to establish **a set of correspondences** between these images?





# Feature Detection and Matching

- ▶ A pair of images to be matched. What kinds of **features** might one use to establish **a set of correspondences** between these images?



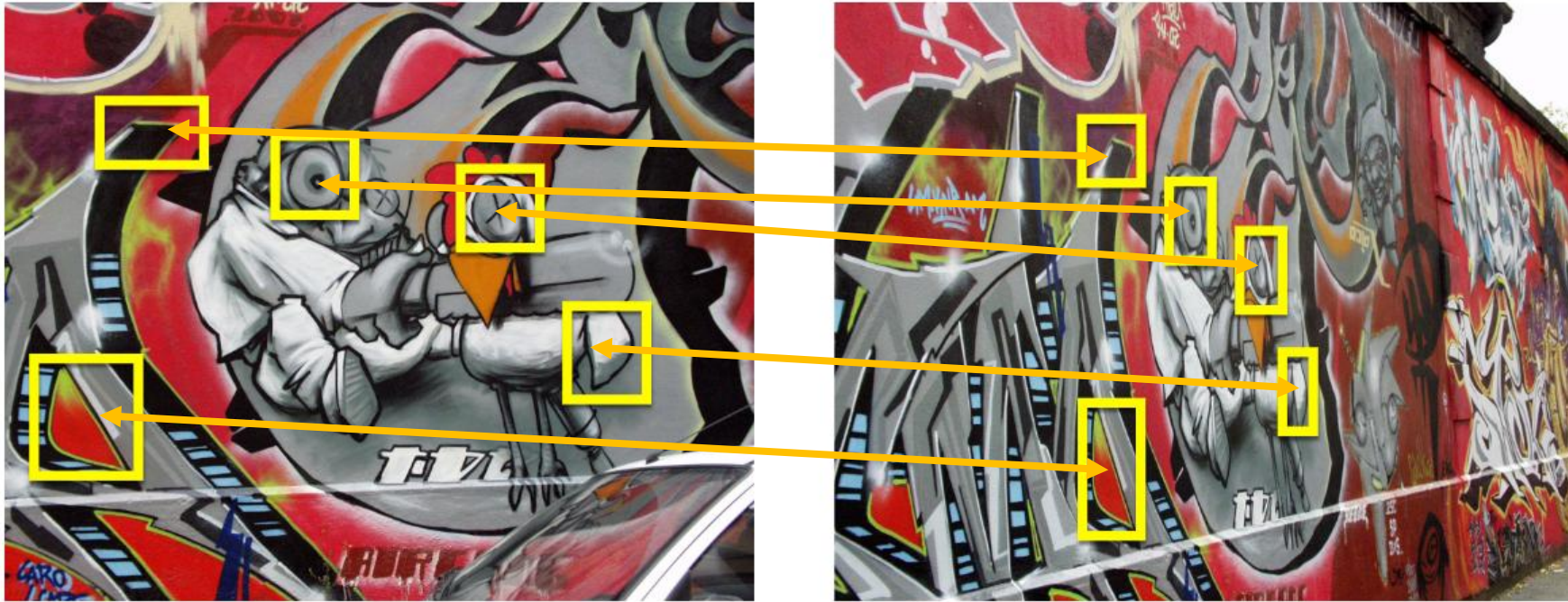
- ▶ The first kind of features that you may notice are **specific locations** in the images, such as building corners, doorways, etc.

# Feature Detection and Matching

- ▶ These kinds of **localized features** are often called **keypoint features** or **interest points** (or even **corners**) and are often described by the appearance of pixel patches surrounding the point location.
- ▶ **Four separate stages** in keypoint detection and matching pipeline:
  - Feature detection (extraction) - each image is searched for **locations** that are likely to match well in other images.
  - Feature description - each **region** around detected keypoint locations is converted into a more compact and stable (invariant) **descriptor** that can be matched against other descriptors. There is a **descriptor vector** for each **keypoint feature**.
  - Feature matching - efficiently searching for likely **matching candidates** in other images.
  - Feature tracking – is more suitable for **video processing** i.e. for video tracking applications.

# Feature Detection and Matching

- ▶ Feature detection and matching example:

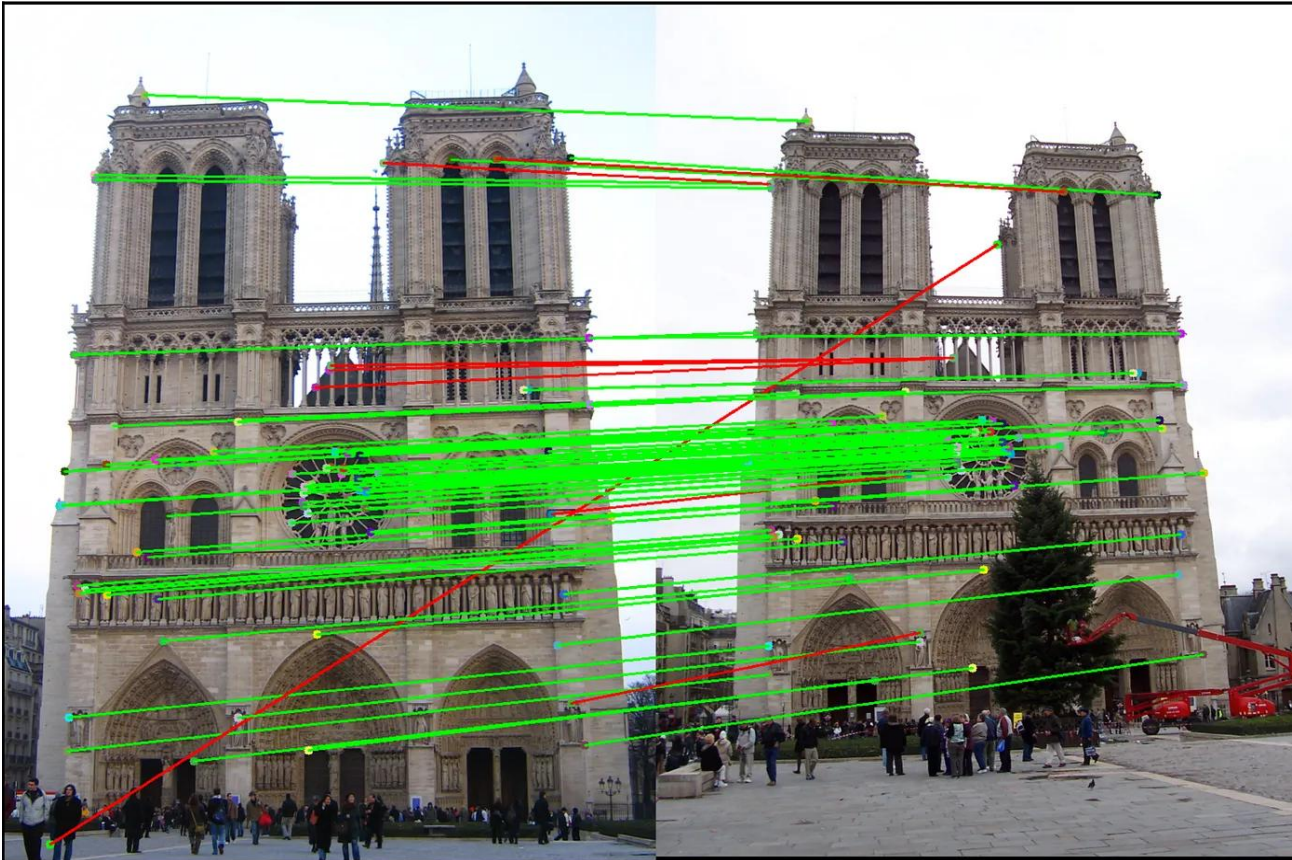


- ▶ False positive (outlier) matches can be detected and then removed using RANSAC.



# Feature Detection and Matching

- Feature detection and matching example:



What are the **red** matches?  
How do you remove them?

# Feature Detection and Matching

- ▶ Most descriptors are computed in a local manner, hence a description is obtained for every interest point or keypoint identified. These descriptors are (ideally) invariant under changes in illumination, translation, scale, and in-plane rotation, such that they can be reliably computed with a high degree of repeatability.
- ▶ Common feature descriptor algorithms:
  - Scale Invariant Feature Transform (SIFT) – length of each descriptor vector is 128 (128-D).
  - Speeded up robust features (SURF) – length of each descriptor is 64.
  - Features from Accelerated Segment Test (FAST).
  - Binary Robust Independent Elementary Features (BRIEF).
  - Oriented FAST and Rotated BRIEF (ORB).

# Feature Detection and Matching

- ▶ Common **feature matching** algorithms:
  - Brute-Force Matcher.
  - Fast Library for Approximate Nearest Neighbors (FLANN) Matcher.
- ▶ Applications of **feature matching** or **image matching** include:
  - Image registration
  - image mosaic or image stitching
  - Object recognition
  - Image retrieval
  - Camera calibration
  - 3D reconstruction
  - Visual SLAM e.g. <https://www.youtube.com/watch?v=G-5jesjNfLc>
  - Vision-based robot localization and navigation, etc.



# Classification, Object Detection, Segmentation and Pose Estimation

- **Image Classification:** A core task in Computer Vision.



(assume given a set of possible labels)  
{dog, cat, truck, plane, ...}



cat

# Classification, Object Detection, Segmentation and Pose Estimation

## ► Computer Vision Tasks:

**Classification**



**CAT**

No spatial extent

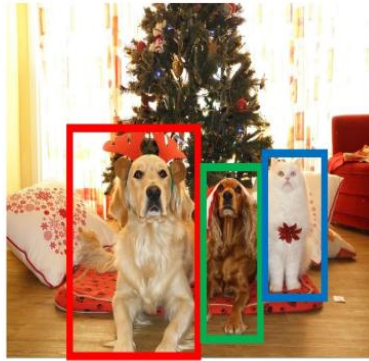
**Semantic Segmentation**



**GRASS, CAT,  
TREE, SKY**

No objects, just pixels

**Object Detection**



**DOG, DOG, CAT**

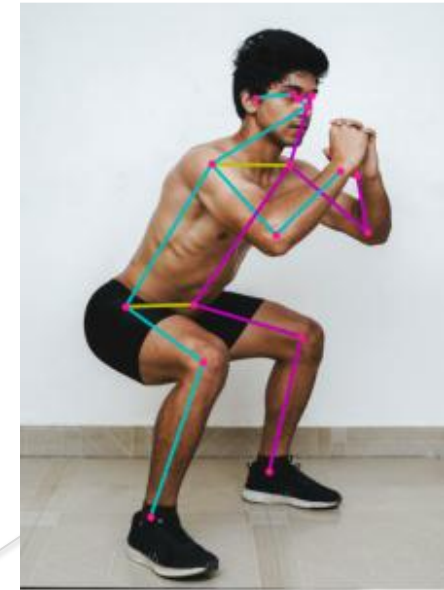
Multiple Object

**Instance Segmentation**



**DOG, DOG, CAT**

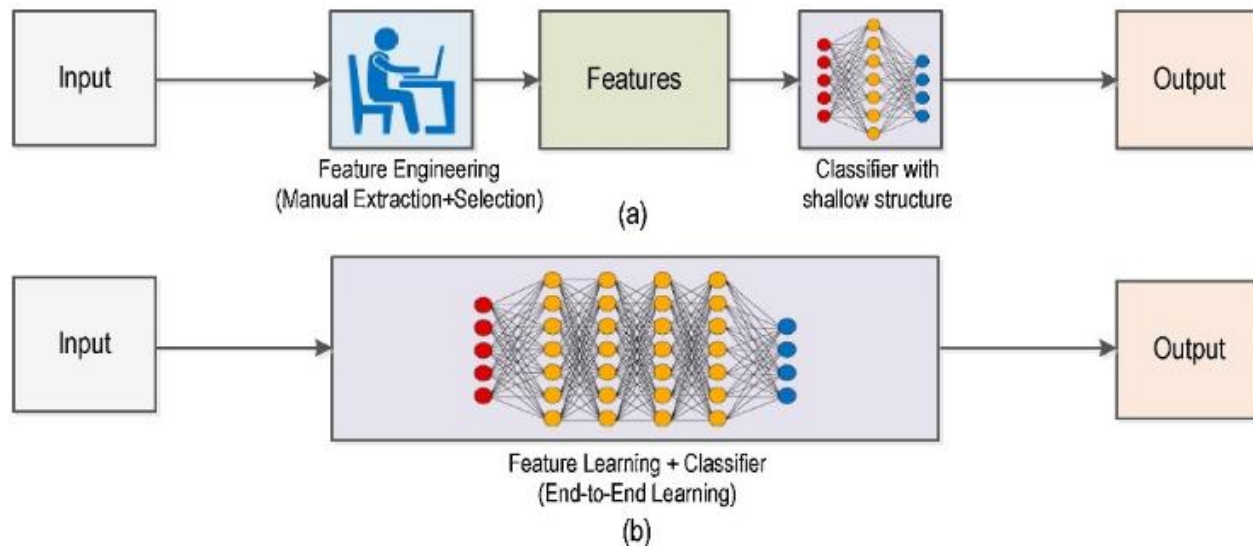
**Pose Estimation**



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# Classification, Object Detection, Segmentation and Pose Estimation

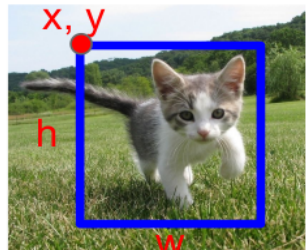
- ▶ Traditional computer vision workflow (a) vs deep learning computer vision workflow (b):



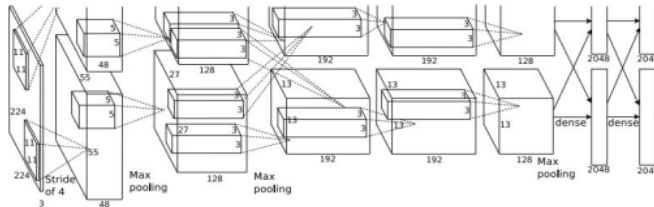
- ▶ Read more on: <https://arxiv.org/abs/1910.13796>

# Classification, Object Detection, Segmentation and Pose Estimation

## Object Detection: Single Object (Classification + Localization)



[This image is CC0 public domain](#)



Fully  
Connected:  
4096 to 1000

### Class Scores

Cat: 0.9  
Dog: 0.05  
Car: 0.01  
...

Correct label:  
Cat

Softmax  
Loss

Multitask Loss

+

Loss

Vector:  
4096  
Fully  
Connected:  
4096 to 4

Box  
Coordinates  
(x, y, w, h)

L2 Loss

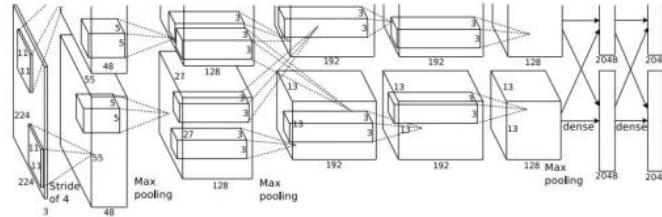
Correct box:  
(x', y', w', h')

Treat localization as a  
regression problem!

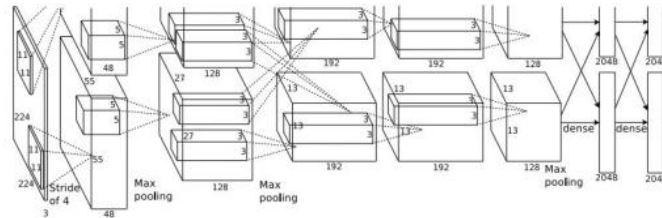


# Classification, Object Detection, Segmentation and Pose Estimation

## ► Object Detection: Multiple Objects



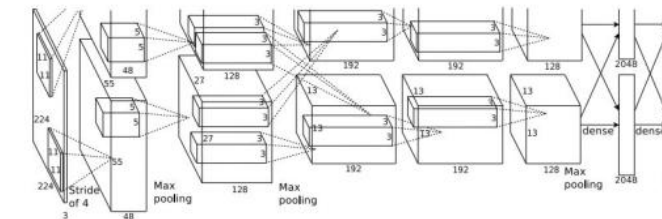
CAT: (x, y, w, h)



DOG: (x, y, w, h)

DOG: (x, y, w, h)

CAT: (x, y, w, h)



DUCK: (x, y, w, h)

DUCK: (x, y, w, h)

....

# Classification, Object Detection, Segmentation and Pose Estimation

- ▶ Types of object detectors:
  - Two-stage object detectors
    - Region proposals generation + object classification and bounding box prediction.
    - E.g. Faster RCNN
  - Single-stage object detectors
    - Removes the region proposals generation stage.
    - E.g. YOLO family (YOLOv1, ..., YOLOv8, ...)



# Visual Tracking

## ► Multi-Object Tracking (MOT):

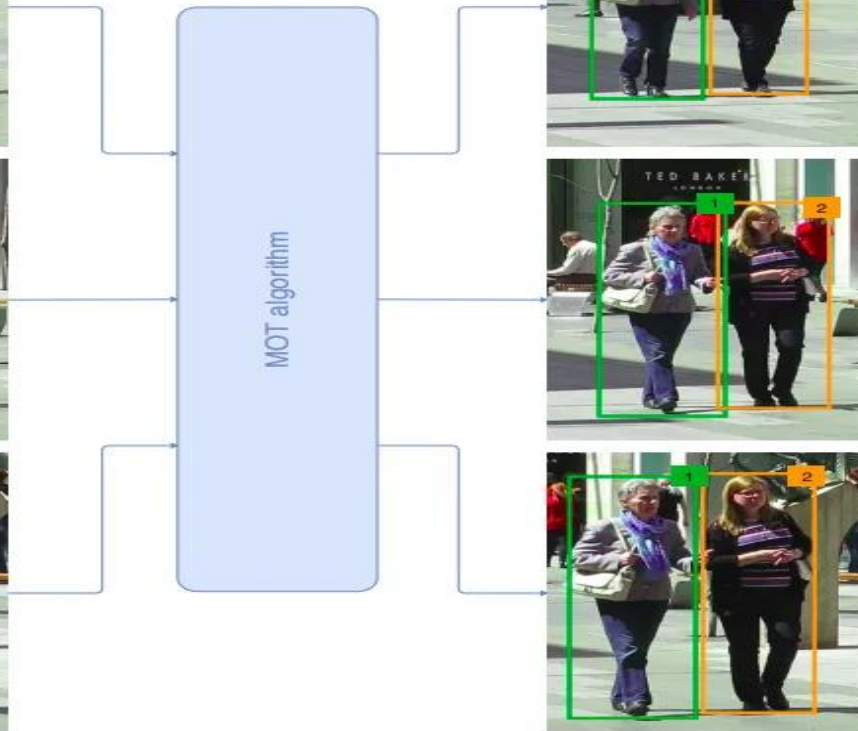
frame  $k = 1$



frame  $k = 5$



frame  $k = 10$



Adapted from:

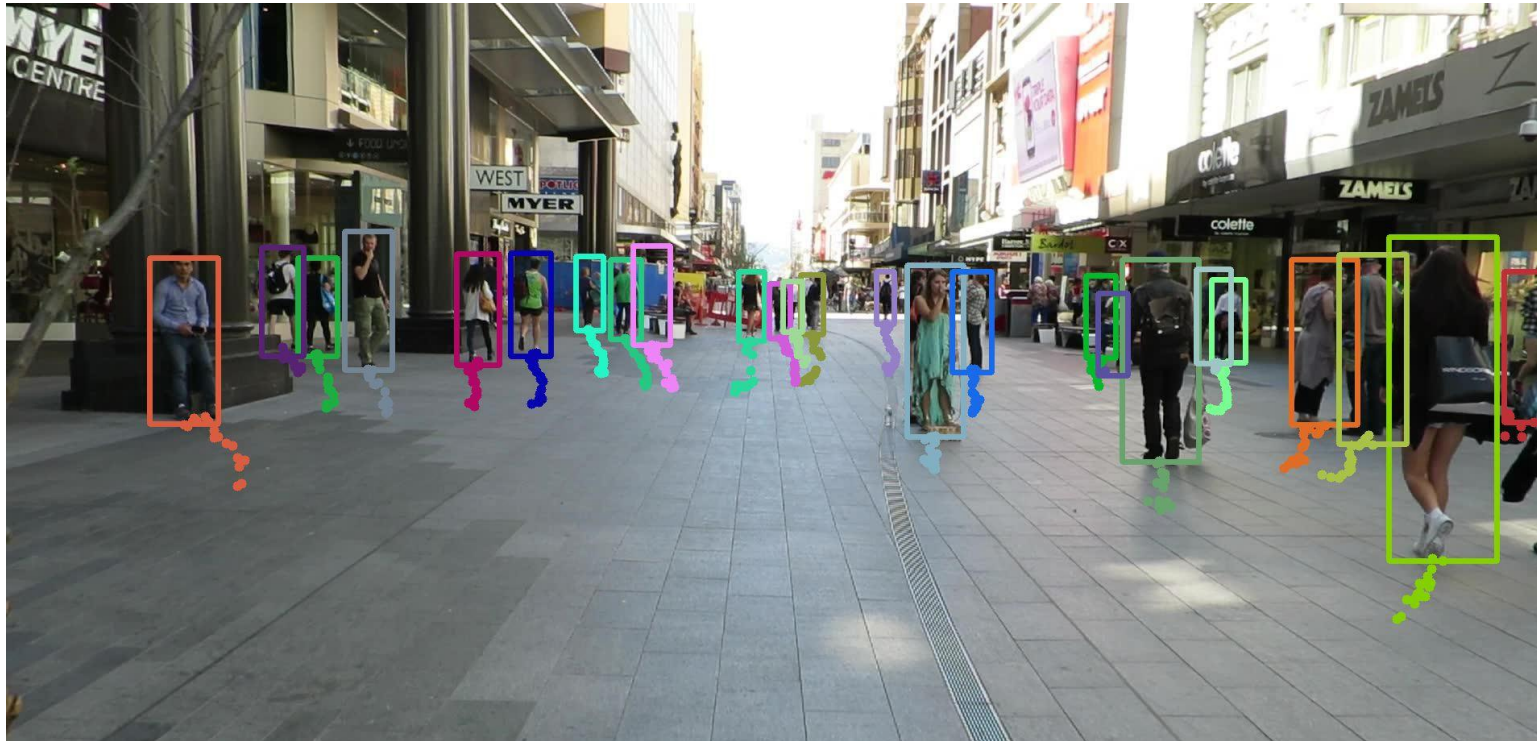
<https://arxiv.org/pdf/1907.12740>

# Visual Tracking

- ▶ **Key components** of MOT:
  - Object detection
    - YOLO family, FRCNN, etc.
  - Motion prediction
    - Kalman filter
    - GM-PHD filter
  - Data association
    - Hungarian algorithm using:
      - Motion information i.e. Intersection over Union (IoU), Mahalanobis distance, etc. and/or
      - Appearance information

# Visual Tracking

- Example of MOT:



- **Figure from:** Nathanael L. Baisa, "Occlusion-robust online multi-object visual tracking using a GM-PHD filter with CNN-based re-identification," *Journal of Visual Communication and Image Representation*, 2021.

# Computer Vision Applications

- ▶ **Security** e.g. Face Recognition:
  - [How does facial recognition work? \(youtube.com\)](#)
- ▶ **Security** e.g. Surveillance:
  - [How China is building an all-seeing surveillance state \(youtube.com\)](#)
- ▶ **Computer Vision (CV) in Robotics:**
  - [Computer vision applications in robotics \(youtube.com\)](#)
- ▶ **Computer Vision in Self Driving Cars and Autonomous Vehicles:**
  - [Self Driving Cars and Autonomous Vehicles Technology \(youtube.com\)](#)
- ▶ **Computer Vision in HealthCare:**
  - [What is Computer Vision in Healthcare Education \(youtube.com\)](#)

# Conclusion

- ▶ Image classification is a core task in computer vision.
- ▶ Feature detection and matching plays a crucial role in vision-based robot mapping, localization and visual SLAM.
- ▶ Understanding computer vision tasks is important to apply them to intelligent robot-based applications, security, healthcare, etc.

# References

- ▶ R. Szeliski, 'Computer Vision: Algorithms and Applications', Springer, 2021.  
[<https://szeliski.org/Book/>]
- ▶ <https://medium.com/data-breach/introduction-to-feature-detection-and-matching-65e27179885d>
- ▶ OpenCV: <https://opencv.org/>
- ▶ Open3D: <http://www.open3d.org/>
- ▶ <https://huggingface.co/learn/computer-vision-course/en/unit0/welcome/welcome>
- ▶ <https://github.com/polygon-software/python-visual-odometry>
- ▶ <https://github.com/luigifreda/pyslam>