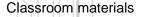


# Al for EDGE

**Object detection** 



- Computer vision
- Latest technological advances
- Object detection algorithms
- Object detection use cases



## **Computer vision**

- Scientific field that deals with how computers can gain high-level understanding from digital images
- Understanding in this context means the transformation of visual images into descriptions of the world
- Sub-domains of computer vision:
  - object detection,

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Classroom materials

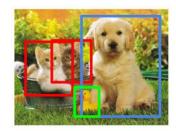
- image classification,
- video tracking,
- , motion estimation,
- scene reconstruction,
- 3D scene modeling, and
- image restoration

#### Classification



CAT

#### **Object Detection**



CAT, DOG, DUCK



Person Bicycle Background

- Object detection is a key field in artificial intelligence, allowing computer systems to "see" their environments by detecting objects in images.
- Object detection applications include
  - Animal detection

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Classroom materials

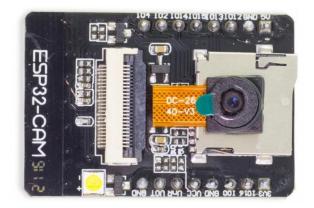
- Pedestrian detection
- People counting
- Vehicle detection
- Face detection
- Text detection
  - Number-plate recognition...



MediaPipe KNIFT: Template-based feature matching https://google.github.io/mediapipe/solutions/knift.html

- In the last few years, the rapid advances of deep learning techniques have greatly accelerated the evolution of object detection.
- Cameras are smaller, cheaper and of higher quality
- Computing platforms moved toward parallelization through multi-core processing and GPU
- As a result, numerous real-world applications, such as healthcare monitoring, autonomous driving, video surveillance, anomaly detection are based on deep learning object detection.

Those advances enabled a key architectural concept called Edge AI.



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Object detection can be performed using traditional(1) or modern(2) techniques:

- (1) image processing techniques don't require historical data for training and are unsupervised in nature
- (2) deep learning methods depend on supervised learning, the performance is limited by computation power (CPUs or GPUs)

Deep learning object detection is widely accepted by researchers and adopted to build commercial products.

Classroom materials



python detect.py --weights yolov7.pt --conf 0.25 --img-size 640 --source inference/images/horses.jpg



- Before 2014 traditional object detection
  - Viola-Jones detector (2001) the pioneering work that started the development of traditional detection methods
  - HOG detector (2006) a popular feature descriptor for object detection
  - DPM (2008) first introduction of bounding box regression
- After 2014 deep learning detection
  - Two-stage algorithms

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Classroom materials

- R-CNN and SPPNet (2014)
- Mask R-CNN (2017)
- Pyramid Networks/FPN (2017)
- G-RCNN (2021)

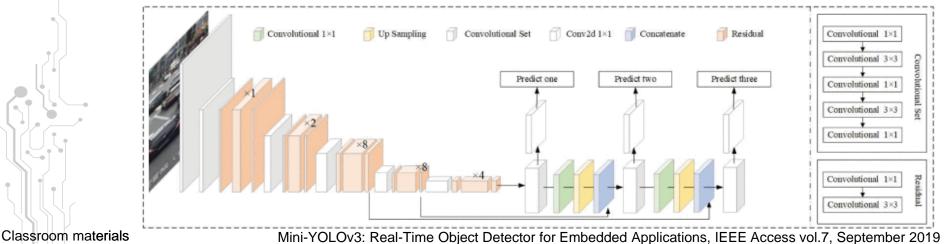
- One-stage algorithms
  - YOLO (2016)
  - SSD (2016)
  - RetinaNet (2017)
  - YOLOv3 (2018)
  - YOLOv4 (2020)
- YOLOR (2021)
- YOLOv7 (2022)

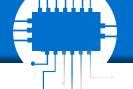
### **Object detection algorithms**

One-stage vs two-stage detectors

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- Object detector solves two subsequent tasks:
  - 1. Find an arbitrary number of objects
  - 2. Classify every single object and estimate its size with a bounding box
- Single-stage detectors combine both tasks into one step (higher performance at the cost of accuracy)
- The main advantage of single-stage is that those algorithms are generally faster than multi-stage detectors and structurally simpler
- The most popular one-stage detector is YOLO





### **Object detection algorithms**

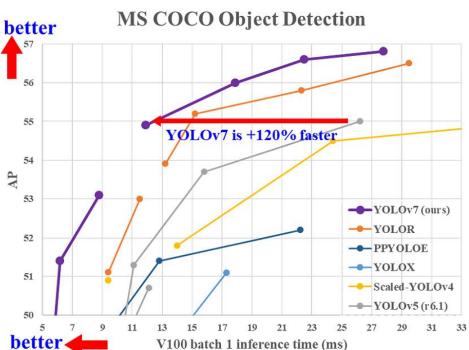
YOLOv7: Trainable bag-of-freebies sets new state-of-the-art for real-time object detectors Chien-Yao Wang, Alexey Bochkovskiy, Hong-Yuan Mark Liao

#### Performance (MS COCO)

|   |            | Model      | Test Size | Aptest | batch<br>[1fps] | batch 32<br>average time |     |
|---|------------|------------|-----------|--------|-----------------|--------------------------|-----|
| / |            | YOLOv7     | 640       | 51.4%  | 161 fps         | 2.8 ms                   | bet |
|   |            | YOLOv7-X   | 640       | 53.1%  | 114 fps         | 4.3 ms                   | 5   |
|   |            | YOLOv7-W6  | 1280      | 54.9%  | 84 fps          | 7.6 ms                   |     |
|   | ۱ <b>۲</b> | YOLOv7-E6  | 1280      | 56.0%  | 56 fps          | 12.3 ms                  | 5   |
|   |            | YOLOv7-D6  | 1280      | 56.6%  | 44 fps          | 15.0 ms                  |     |
|   |            | YOLOv7-E6E | 1280      | 56.8%  | 36 fps          | 18.7 ms                  | 5   |
|   |            |            |           |        |                 |                          |     |

| Model      | Parameters<br>(million) | FPS | AP test (%) |  |  |  |  |
|------------|-------------------------|-----|-------------|--|--|--|--|
| YOLO7-Tiny | 6.2                     | 286 | 38.7        |  |  |  |  |
| YOLOv7     | 36.9                    | 161 | 51.4        |  |  |  |  |
| YOLOv7-X   | 71.3                    | 114 | 53.1        |  |  |  |  |
| YOLOv7-W6  | 70.04                   | 84  | 54.9        |  |  |  |  |
| YOLOv7-E6  | 97.2                    | 56  | 56.0        |  |  |  |  |
| YOLOv7-D6  | 154.7                   | 44  | 56.6        |  |  |  |  |
| YOLOv7-E6E | YOLOv7-E6E 151.7        |     | 56.8        |  |  |  |  |

**Classroom materials** 

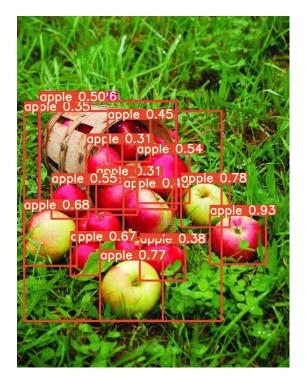




YOLOv7: Trainable bag-of-freebies sets new state-of-the-art for real-time object detectors

Chien-Yao Wang, Alexey Bochkovskiy, Hong-Yuan Mark Liao

|                      |         |        |      |                     | 5                        |                          |                       |                         |                       |
|----------------------|---------|--------|------|---------------------|--------------------------|--------------------------|-----------------------|-------------------------|-----------------------|
| Model                | #Param. | FLOPs  | Size | $\mathbf{AP}^{val}$ | $\mathbf{AP}^{val}_{50}$ | $\mathbf{AP}^{val}_{75}$ | $\mathbf{AP}^{val}_S$ | $\mathbf{AP}_{M}^{val}$ | $\mathbf{AP}_L^{val}$ |
| YOLOv4 [3]           | 64.4M   | 142.8G | 640  | 49.7%               | 68.2%                    | 54.3%                    | 32.9%                 | 54.8%                   | 63.7%                 |
| YOLOR-u5 (r6.1) [81] | 46.5M   | 109.1G | 640  | 50.2%               | 68.7%                    | 54.6%                    | 33.2%                 | 55.5%                   | 63.7%                 |
| YOLOv4-CSP [79]      | 52.9M   | 120.4G | 640  | 50.3%               | 68.6%                    | 54.9%                    | 34.2%                 | 55.6%                   | 65.1%                 |
| YOLOR-CSP [81]       | 52.9M   | 120.4G | 640  | 50.8%               | 69.5%                    | 55.3%                    | 33.7%                 | 56.0%                   | 65.4%                 |
| YOLOv7               | 36.9M   | 104.7G | 640  | 51.2%               | 69.7%                    | 55.5%                    | 35.2%                 | 56.0%                   | 66.7%                 |
| improvement          | -43%    | -15%   | -    | +0.4                | +0.2                     | +0.2                     | +1.5                  | =                       | +1.3                  |
| YOLOR-CSP-X [81]     | 96.9M   | 226.8G | 640  | 52.7%               | 71.3%                    | 57.4%                    | 36.3%                 | 57.5%                   | 68.3%                 |
| YOLOv7-X             | 71.3M   | 189.9G | 640  | 52.9%               | 71.1%                    | 57.5%                    | 36.9%                 | 57.7%                   | 68.6%                 |
| improvement          | -36%    | -19%   | -    | +0.2                | -0.2                     | +0.1                     | +0.6                  | +0.2                    | +0.3                  |
| YOLOv4-tiny [79]     | 6.1     | 6.9    | 416  | 24.9%               | 42.1%                    | 25.7%                    | 8.7%                  | 28.4%                   | 39.2%                 |
| YOLOv7-tiny          | 6.2     | 5.8    | 416  | 35.2%               | 52.8%                    | 37.3%                    | 15.7%                 | 38.0%                   | 53.4%                 |
| improvement          | +2%     | -19%   | -    | +10.3               | +10.7                    | +11.6                    | +7.0                  | +9.6                    | +14.2                 |
| YOLOv4-tiny-3l [79]  | 8.7     | 5.2    | 320  | 30.8%               | 47.3%                    | 32.2%                    | 10.9%                 | 31.9%                   | 51.5%                 |
| YOLOv7-tiny          | 6.2     | 3.5    | 320  | 30.8%               | 47.3%                    | 32.2%                    | 10.0%                 | 31.9%                   | 52.2%                 |
| improvement          | -39%    | -49%   | -    | =                   | =                        | =                        | -0.9                  | =                       | +0.7                  |
| YOLOR-E6 [81]        | 115.8M  | 683.2G | 1280 | 55.7%               | 73.2%                    | 60.7%                    | 40.1%                 | 60.4%                   | 69.2%                 |
| YOLOv7-E6            | 97.2M   | 515.2G | 1280 | 55.9%               | 73.5%                    | 61.1%                    | 40.6%                 | 60.3%                   | 70.0%                 |
| improvement          | -19%    | -33%   | -    | +0.2                | +0.3                     | +0.4                     | +0.5                  | -0.1                    | +0.8                  |
| YOLOR-D6 [81]        | 151.7M  | 935.6G | 1280 | 56.1%               | 73.9%                    | 61.2%                    | 42.4%                 | 60.5%                   | 69.9%                 |
| YOLOv7-D6            | 154.7M  | 806.8G | 1280 | 56.3%               | 73.8%                    | 61.4%                    | 41.3%                 | 60.6%                   | 70.1%                 |
| YOLOv7-E6E           | 151.7M  | 843.2G | 1280 | 56.8%               | 74.4%                    | 62.1%                    | 40.8%                 | 62.1%                   | 70.6%                 |
| improvement          | =       | -11%   | -    | +0.7                | +0.5                     | +0.9                     | -1.6                  | +1.6                    | +0.7                  |



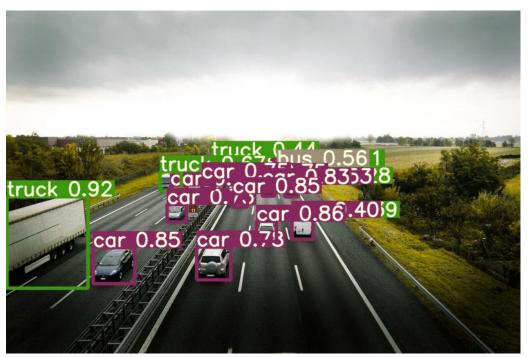


**Classroom materials** 

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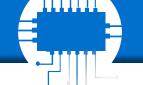


- Industrial PPE detection (personal protective equipment)
- Anomaly and defect detection in product assembly
- Autonomous driving
- Traffic monitoring and road maintenance
- People counting
- Parking occupancy
- Intrusion detection



Classroom materials

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### Solve the Mystery of Vehicle Detection Algorithm

https://www.mouser.mx/blog/mystery-of-vehicle-detection Trainable bag-of-freebies sets new state-of-the-art for real-time object detectors https://huggingface.co/spaces/akhaliq/yolov7

