YOLOv7 vs YOLOv8

(Compare performance between YOLOv7, YOLO 8l and YOLOv8m)

1. Approach:

Training YOLOv7, YOLOv8l and YOLOv8m (300 epochs)with COCO128 dataset until models converge and then compare (image size = 640, batch size = 8). All training was done in a Google Colab environment with nVidia Tesla T4 GPUs.

2. Object:

- Performace of 3 model: YOLOv7, YOLOv8l and YOLOv8m.

| | YOLOv7 | YOLOv8l | YOLOv8m |
|---------------|--------|---------|---------|
| Layer | 407 | 365 | 295 |
| Parameter (M) | 37.62 | 43.69 | 25.90 |
| GFLOPs | 106.5 | 165.7 | 79.3 |

3. Result:

3.1. Training time (200 epochs):

| | YOLOv7 | YOLOv8l | YOLOv8m | |
|---------------|-------------|-------------|-------------|--|
| Layer | 407 | 365 | 295 | |
| Parameter (M) | 37.62 | 43.69 | 25.9 | |
| Training time | 0.673 hours | 0.658 hours | 0.509 hours | |

In spite of the large-sized model of YOLOv8 (YOLOv8l) is bigger than YOLOv7, faster training time than YOLOv7. YOLOv8m size is the smallest so training time can be small

=> YOLOv8 model need training time less than YOLOv7 with the same size (number of parameters).

3.2. Inference time:

| | YOLOv7 | YOLOv8l | YOLOv8m |
|----------------|--------|---------|---------|
| Layer | 407 | 365 | 295 |
| Parameter (M) | 37.62 | 43.69 | 25.9 |
| Inference time | 16.9ms | 50.2 ms | 28.1 ms |

YOLOv7 has inference time less than YOLO8

3.3. Acccuracy: COCO128.yaml, 50 epochs, imgsize=640,batch=8

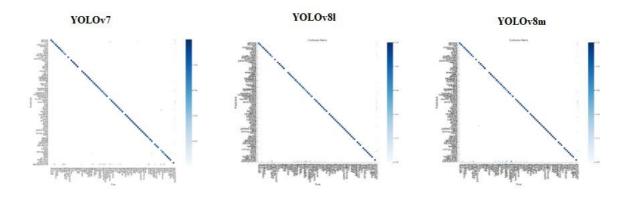
| | YOLOv7 | YOLOv8l | YOLOv8m | |
|---------------|--------|---------|---------|--|
| Layer | 407 | 365 | 295 | |
| Parameter (M) | 37.62 | 43.69 | 25.9 | |
| Presion | 0.936 | 0.948 | 0.962 | |
| Recall | 0.903 | 0.943 | 0.926 | |
| F1 score | 0.919 | 0.946 | 0.944 | |
| mAP@0.5 | 0.962 | 0.97 | 0.97 | |
| mAP@0.5:0.95 | 0.764 | 0.909 | 0.904 | |

^{=&}gt; YOLOv8 is better YOLOv7 when compare F1 score, mAP. YOLOv8 size 1 is better size m in this case.

3.4. Actual results:



YOLOv7 is better YOLOv8 with detecting distant objects



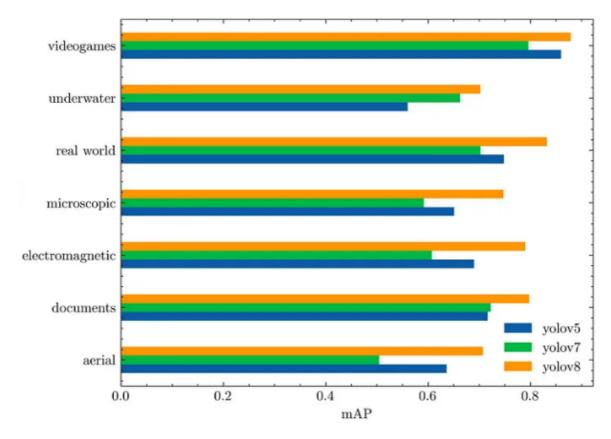
=> YOLOv7's multilayer false rate is greater than YOLOv8. The error rate in each wrong class of YOLOv8m is greater than YOLOv8l

4. Conclusion:

- YOLOv7 has the shortest inference time, then YOLOv8m and finally YOLOv8l has the largest inference time.
- YOLOv8m has less training time than YOLOv8l, YOLOv7 has the largest training time among the 3 models.
- YOLOv7 can detect distant objects better than YOLOv8l and YOLOv8m, but when detecting nearby overlapping objects, YOLOv7 is the worst compared to the other 2 models.
- Both YOLOv8l and YOLOv8m detect close and overlapping objects well. In this case (train with COCO128 data), YOLOv8m can detect many nearby objects and has higher accuracy than YOLOv8l

5. Inferences:

- 5.1. https://medium.com/@juneta.tao/yolov8-4b1c330f8c90
- Roboflow fine-tuned and compared YOLOv5, YOLOv7 and YOLOv8 on 100 different datasets. YOLOv8 performs the best over all datasets.



5.1. Research: A Full-Scale Reloading

| Method | Input Size | APval | APval 50 | FPS | FPS (bs=32) | Latency | Params | FLOPs |
|------------------|------------|--------|-------------|-----|-------------|---------|---------|---------|
| YOLOv7-Tiny [16] | 416 | 33.3%* | 49.9%* | 787 | 1196 | 1.3 ms | 6.2 M | 5.8 G |
| YOLOv7-Tiny [16] | 640 | 37.4%* | 55.2%* | 424 | 519 | 2.4 ms | 6.2 M | 13.7 G* |
| YOLOv7 [16] | 640 | 51.2% | 69.7%* | 110 | 122 | 9.0 ms | 36.9 M | 104.7 G |
| YOLOv7-E6E [16] | 1280 | 56.8% | 74.4%* | 16 | 17 | 59.6 ms | 151.7 M | 843.2 G |
| YOLOv8-N [6] | 640 | 37.3% | 52.6%* | 561 | 734 | 1.8 ms | 3.2 M | 8.7 G |
| YOLOv8-S [6] | 640 | 44.9% | 61.8%* | 311 | 387 | 3.2 ms | 11.2 M | 28.6 G |
| YOLOv8-M [6] | 640 | 50.2% | 67.2%* | 143 | 176 | 7.0 ms | 25.9 M | 78.9 G |
| YOLOv8-L [6] | 640 | 52.9% | 69.8%* | 91 | 105 | 11.0 ms | 43.7 M | 165.2 G |

Table 2: Comparisons with other YOLO-series detectors on COCO 2017 val. FPS and latency are measured in FP16-precision on a Tesla T4 in the same environment with TensorRT. All our models are trained for 300 epochs without pre-training or any external data. Both the accuracy and the speed performance of our models are evaluated with the input resolution of 640×640 .