

yolov5

November 12, 2020

1 Setup

Clone repo, install dependencies, %cd into ./yolov5 folder and check GPU.

```
[1]: !git clone https://github.com/ultralytics/yolov5 # clone repo
!pip install -qr yolov5/requirements.txt # install dependencies (ignore errors)
%cd yolov5

import torch
from IPython.display import Image, clear_output # to display images
from utils.google_utils import gdrive_download # to download models/datasets

clear_output()
print('Setup complete. Using torch %s %s' % (torch.__version__, torch.cuda.
↳ get_device_properties(0) if torch.cuda.is_available() else 'CPU'))
```

```
Setup complete. Using torch 1.6.0 _CudaDeviceProperties(name='Tesla
P100-PCIE-16GB', major=6, minor=0, total_memory=16280MB,
multi_processor_count=56)
```

2 2. Test

Test a model on COCO val or test-dev dataset to determine trained accuracy. Models are auto-downloaded from [Google Drive](#). To show results by class use the `--verbose` flag. Note that pycocotools metrics may be 1-2% better than the equivalent repo metrics, as is visible below, due to slight differences in mAP computation.

2.0.1 2.1 val2017

Download COCO val 2017 dataset, 1GB, 5000 images, and test model accuracy.

```
[ ]: # Download COCO val2017
gdrive_download('1Y6Kou6kEBOZEMCCpJSKStCor4KAReE43', 'coco2017val.zip') #_
↳ val2017 dataset
!mv ./coco ../ # move folder alongside /yolov5

[53]: # Run YOLOv5x on COCO val2017
!python test.py --weights runs/exp1/weights/best.pt --data kkctbn.yaml --img 416
```

```

Namespace(augment=False, batch_size=32, conf_thres=0.001,
data='./data/kkctbn.yaml', device='', img_size=416, iou_thres=0.65, merge=False,
save_json=False, save_txt=False, single_cls=False, task='val', verbose=False,
weights=['runs/exp1/weights/best.pt'])
Fusing layers...
Scanning labels ../kkctbn/labels.cache (402 found, 0 missing, 0 empty, 0
duplicate, for 402 images): 402it [00:00, 10086.38it/s]

```

	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.673	0.935	0.929

```

0.487
Speed: 1.8/1.8/3.6 ms inference/NMS/total per 416x416 image at batch-size 32

```

3 3. Train

```

[27]: !mkdir ../kkctbn
      !cp -r /kaggle/input/kkctbn ../kkctbn

```

```
mkdir: cannot create directory '../kkctbn': File exists
```

```

[19]: !mkdir ../kkctbn/images
      !mkdir ../kkctbn/labels
      !mv ../kkctbn/dataset/*.txt ../kkctbn/labels
      !mv ../kkctbn/dataset/*.jpg ../kkctbn/images

```

```

[30]: !cp /kaggle/input/kkctbn/kkctbn.yaml ./data
      !cp /kaggle/input/kkctbn/yolov5s.yaml ./models/yolov5s2.yaml

```

```

[33]: !ls ./models

```

```

__init__.py  experimental.py  yolo.py          yolov5s.yaml
__pycache__  export.py        yolov5l.yaml     yolov5s2.yaml
common.py    hub              yolov5m.yaml     yolov5x.yaml

```

```

[31]: !ls data

```

```

coco.yaml      hyp.finetune.yaml  kkctbn.yaml  voc.yaml
coco128.yaml   hyp.scratch.yaml  scripts

```

```

[ ]: # Start tensorboard (optional)
     %load_ext tensorboard
     %tensorboard --logdir runs

```

```

[42]: # Train YOLOv5s on coco128 for 3 epochs
      !python train.py --img 416 --batch 16 --epochs 50 --data kkctbn.yaml --cfg
      ↪ yolov5s2.yaml --weights yolov5s.pt --nosave --cache

```

```

2020-09-29 07:46:29.814952: I
tensorflow/stream_executor/platform/default/dso_loader.cc:48] Successfully

```

opened dynamic library libcudart.so.10.1
 Overriding model.yaml nc=80 with nc=2

Scanning labels ../kkctbn/labels.cache (402 found, 0 missing, 0 empty, 0 duplicate, for 402 images): 402it [00:00, 10746.67it/s]

Caching images (0.2GB): 100%| | 402/402 [00:01<00:00, 298.95it/s]

Scanning labels ../kkctbn/labels.cache (402 found, 0 missing, 0 empty, 0 duplicate, for 402 images): 402it [00:00, 2701.75it/s]

Caching images (0.2GB): 100%| | 402/402 [00:02<00:00, 183.98it/s]

Analyzing anchors... anchors/target = 5.08, Best Possible Recall (BPR) = 1.0000

0/49	0.818G	0.1241	0.05478	0.02647	0.2053	5	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0	0	0.00715
0.0022							
1/49	1.49G	0.1066	0.06273	0.01857	0.1879	3	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.963	0.0186	0.0948
0.0358							
2/49	1.49G	0.09701	0.0701	0.01275	0.1799	5	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.357	0.302	0.292
0.0926							
3/49	1.49G	0.08072	0.06094	0.007546	0.1492	1	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.0679	0.887	0.437
0.147							
4/49	1.49G	0.07168	0.05538	0.004227	0.1313	11	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.0768	0.915	0.495
0.173							
5/49	1.49G	0.06678	0.05039	0.002877	0.12	7	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.108	0.924	0.566
0.196							
6/49	1.49G	0.07032	0.0505	0.002292	0.1231	5	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.14	0.919	0.648
0.217							
7/49	1.49G	0.07096	0.04988	0.001966	0.1228	13	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.152	0.836	0.495
0.171							
8/49	1.49G	0.06969	0.04679	0.001945	0.1184	10	416
	Class	Images	Targets		P	R	mAP@0.5
	all	402	1.39e+03		0.172	0.914	0.588
0.2							
9/49	1.49G	0.06437	0.04825	0.001852	0.1145	8	416

		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.18	0.918	0.667
0.27							
	10/49	1.49G	0.06548	0.04574 0.001859	0.1131	4	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.221	0.909	0.672
0.252							
	11/49	1.49G	0.0634	0.05004 0.001818	0.1153	19	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.287	0.913	0.713
0.289							
	12/49	1.49G	0.06078	0.04642 0.001707	0.1089	6	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.167	0.91	0.575
0.206							
	13/49	1.49G	0.05666	0.04318 0.001522	0.1014	2	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.24	0.915	0.716
0.291							
	14/49	1.49G	0.05641	0.04423 0.001515	0.1022	9	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.296	0.912	0.786
0.335							
	15/49	1.49G	0.05405	0.04439 0.001552	0.09998	6	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.301	0.917	0.784
0.327							
	16/49	1.49G	0.0537	0.04673 0.001409	0.1018	5	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.359	0.912	0.735
0.302							
	17/49	1.49G	0.05039	0.04472 0.00128	0.09639	4	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.429	0.907	0.859
0.37							
	18/49	1.49G	0.05106	0.0435 0.001213	0.09577	7	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.357	0.899	0.74
0.276							
	19/49	1.49G	0.05793	0.04487 0.001323	0.1041	18	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.338	0.917	0.802
0.344							
	20/49	1.49G	0.05786	0.04459 0.00137	0.1038	6	416
		Class	Images	Targets	P	R	mAP@.5
		all	402	1.39e+03	0.345	0.923	0.732
0.32							
	21/49	1.49G	0.05307	0.04272 0.001318	0.09711	2	416

		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.419	0.902	0.811
0.335								
	22/49	1.49G	0.05167	0.04609	0.00132	0.09908	10	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.504	0.898	0.872
0.402								
	23/49	1.49G	0.04905	0.04325	0.001266	0.09356	12	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.401	0.904	0.839
0.367								
	24/49	1.49G	0.04886	0.04674	0.001146	0.09675	24	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.569	0.903	0.778
0.329								
	25/49	1.49G	0.04954	0.04529	0.001151	0.09598	22	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.549	0.912	0.896
0.426								
	26/49	1.49G	0.04926	0.04467	0.001114	0.09504	4	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.445	0.915	0.818
0.337								
	27/49	1.49G	0.04629	0.04426	0.001087	0.09164	6	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.466	0.92	0.832
0.372								
	28/49	1.49G	0.04446	0.04094	0.001016	0.08641	3	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.582	0.91	0.888
0.429								
	29/49	1.49G	0.04293	0.04005	0.0009504	0.08393	2	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.536	0.922	0.893
0.422								
	30/49	1.49G	0.04364	0.04214	0.0009038	0.08668	8	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.52	0.927	0.898
0.406								
	31/49	1.49G	0.04252	0.04248	0.0008888	0.08589	7	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.663	0.917	0.899
0.434								
	32/49	1.49G	0.04703	0.04387	0.0009314	0.09183	10	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.561	0.925	0.884
0.424								
	33/49	1.49G	0.04817	0.04416	0.0009875	0.09331	1	416

	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.617	0.912	0.9
0.424						
34/49	1.49G	0.04573	0.04302 0.0009362	0.08969	6	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.503	0.93	0.883
0.384						
35/49	1.49G	0.04273	0.04046 0.0009249	0.08411	10	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.633	0.925	0.906
0.45						
36/49	1.49G	0.04202	0.04391 0.0008563	0.08679	5	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.615	0.928	0.901
0.434						
37/49	1.49G	0.04016	0.03979 0.0008705	0.08082	12	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.639	0.921	0.905
0.445						
38/49	1.49G	0.03994	0.04174 0.0007848	0.08247	4	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.617	0.932	0.915
0.441						
39/49	1.49G	0.03955	0.04346 0.0007587	0.08377	13	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.671	0.912	0.894
0.431						
40/49	1.49G	0.03871	0.03958 0.0007388	0.07903	3	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.688	0.907	0.894
0.42						
41/49	1.49G	0.03822	0.04072 0.000698	0.07964	5	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.661	0.932	0.921
0.462						
42/49	1.49G	0.03847	0.04284 0.0006348	0.08194	1	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.597	0.927	0.911
0.452						
43/49	1.49G	0.03584	0.03823 0.0006067	0.07468	2	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.619	0.937	0.921
0.469						
44/49	1.49G	0.03671	0.04258 0.0005867	0.07988	6	416
	Class	Images	Targets	P	R	mAP@.5
	all	402	1.39e+03	0.67	0.931	0.92
0.464						
45/49	1.49G	0.03616	0.03944 0.0005387	0.07614	8	416

		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.722	0.928	0.918
0.463	46/49	1.49G	0.03723	0.0404	0.000551	0.07818	5	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.702	0.936	0.924
0.47	47/49	1.49G	0.03619	0.04299	0.0005315	0.07971	12	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.717	0.928	0.922
0.474	48/49	1.49G	0.03554	0.03967	0.0004738	0.07568	13	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.695	0.937	0.929
0.474	49/49	1.49G	0.03516	0.04059	0.0004458	0.07619	14	416
		Class	Images	Targets		P	R	mAP@.5
		all	402	1.39e+03		0.744	0.925	0.923

0.48

Optimizer stripped from runs/exp1/weights/last.pt, 14.8MB

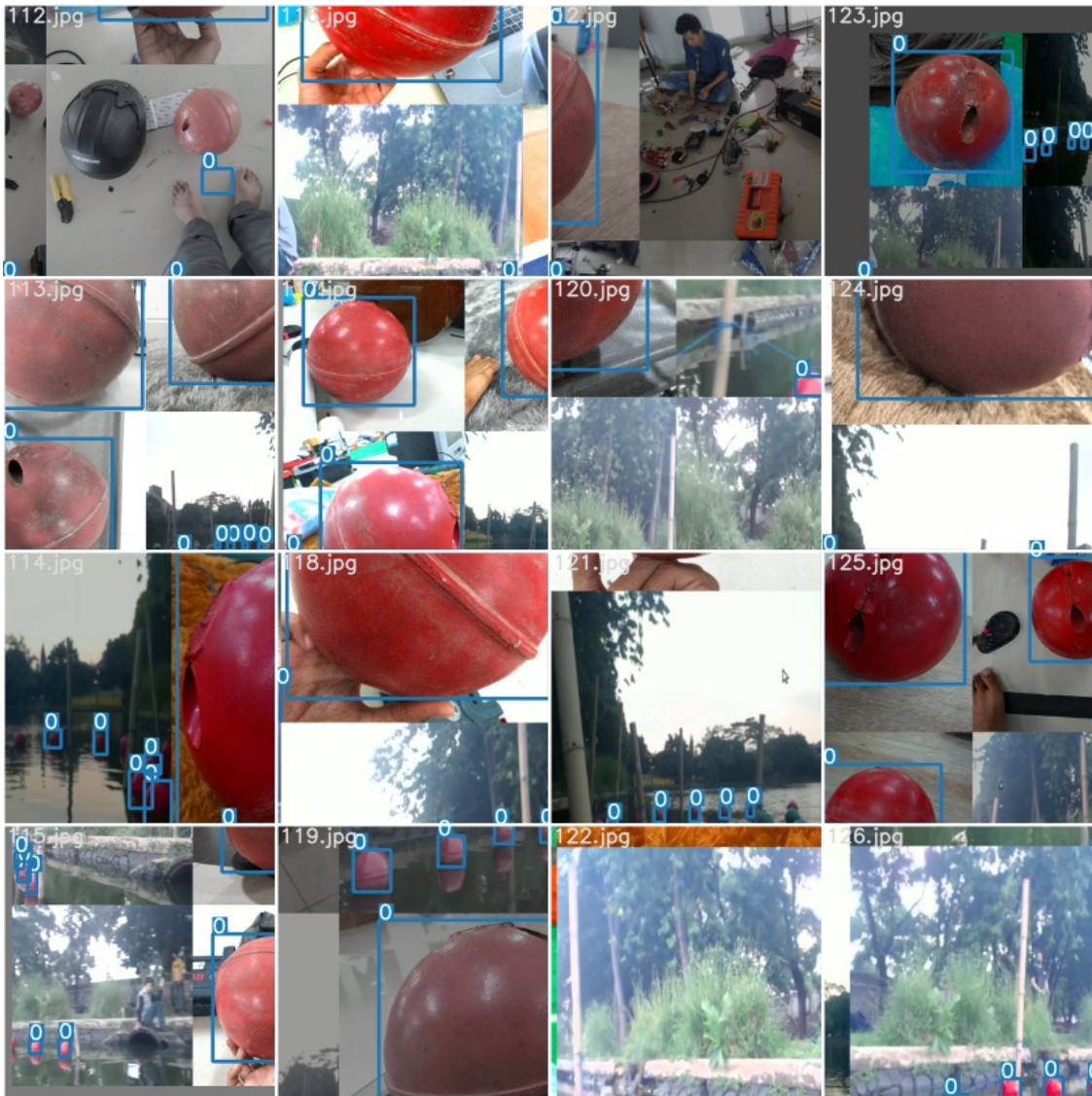
Optimizer stripped from runs/exp1/weights/best.pt, 14.8MB

4 4. Visualize

View runs/exp0/train*.jpg images to see training images, labels and augmentation effects. A **Mosaic Dataloader** is used for training (shown below), a new concept developed by Ultralytics and first featured in [YOLOv4](#).

```
[43]: Image(filename='runs/exp0/train_batch1.jpg', width=900) # view augmented
      ↪ training mosaics
```

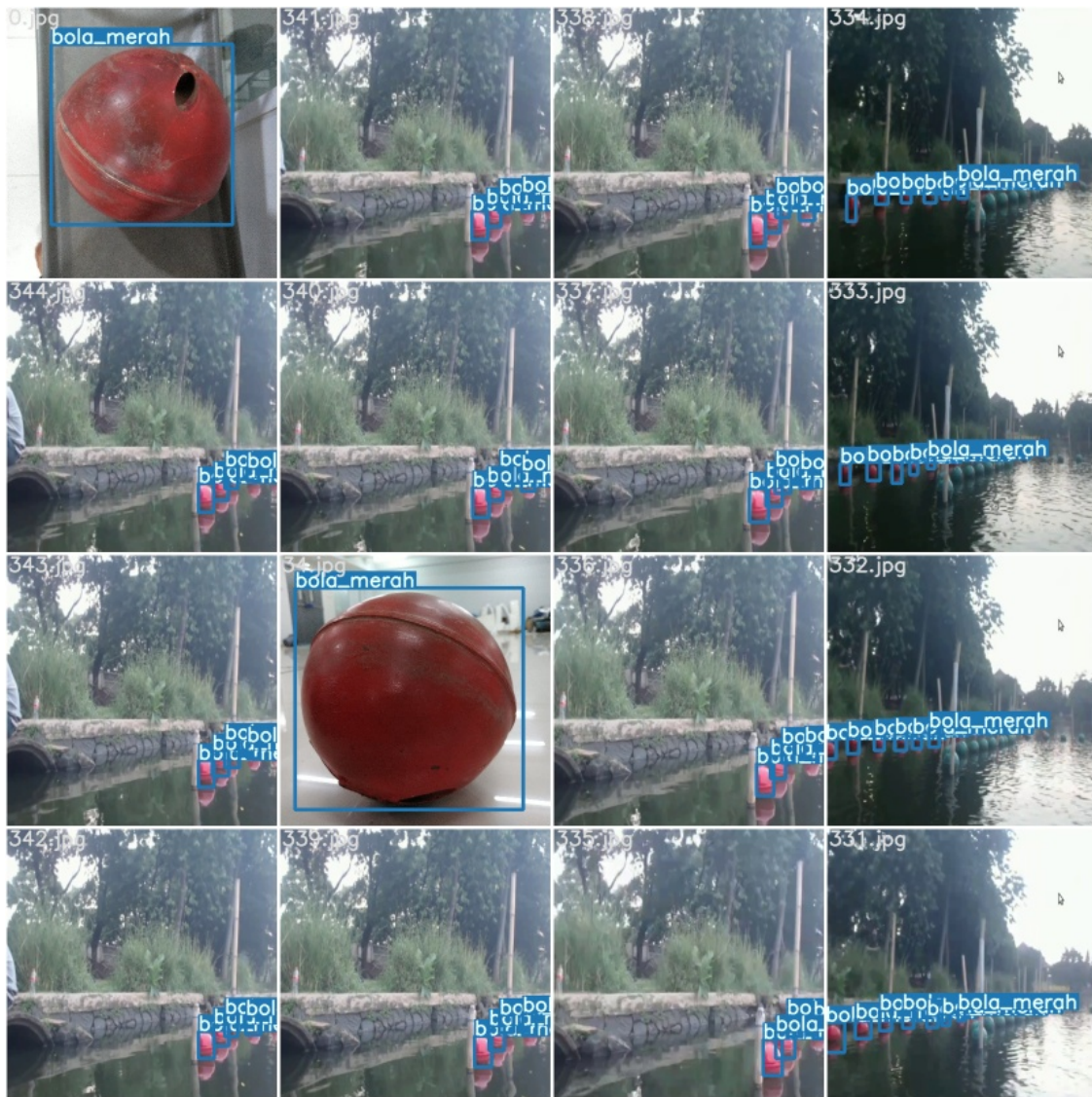
[43]:



View test_batch0_gt.jpg to see test batch 0 *ground truth* labels.

```
[44]: Image(filename='runs/exp0/test_batch0_gt.jpg', width=900) # view test image_
      ↪ labels
```

[44]:



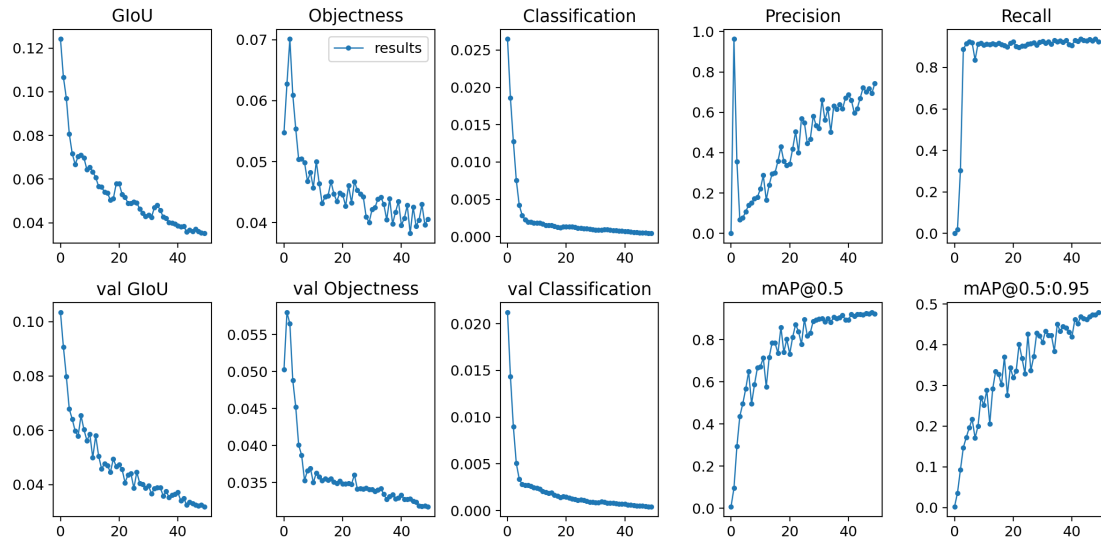
View test_batch0_pred.jpg to see test batch 0 *predictions*.

[48]: `!ls runs/exp1`

events.out.tfevents.1601365591.6416aa4de09b.1060.0	test_batch0_gt.jpg
hyp.yaml	test_batch0_pred.jpg
labels.png	train_batch0.jpg
labels_correlogram.png	train_batch1.jpg
opt.yaml	train_batch2.jpg
results.png	weights
results.txt	

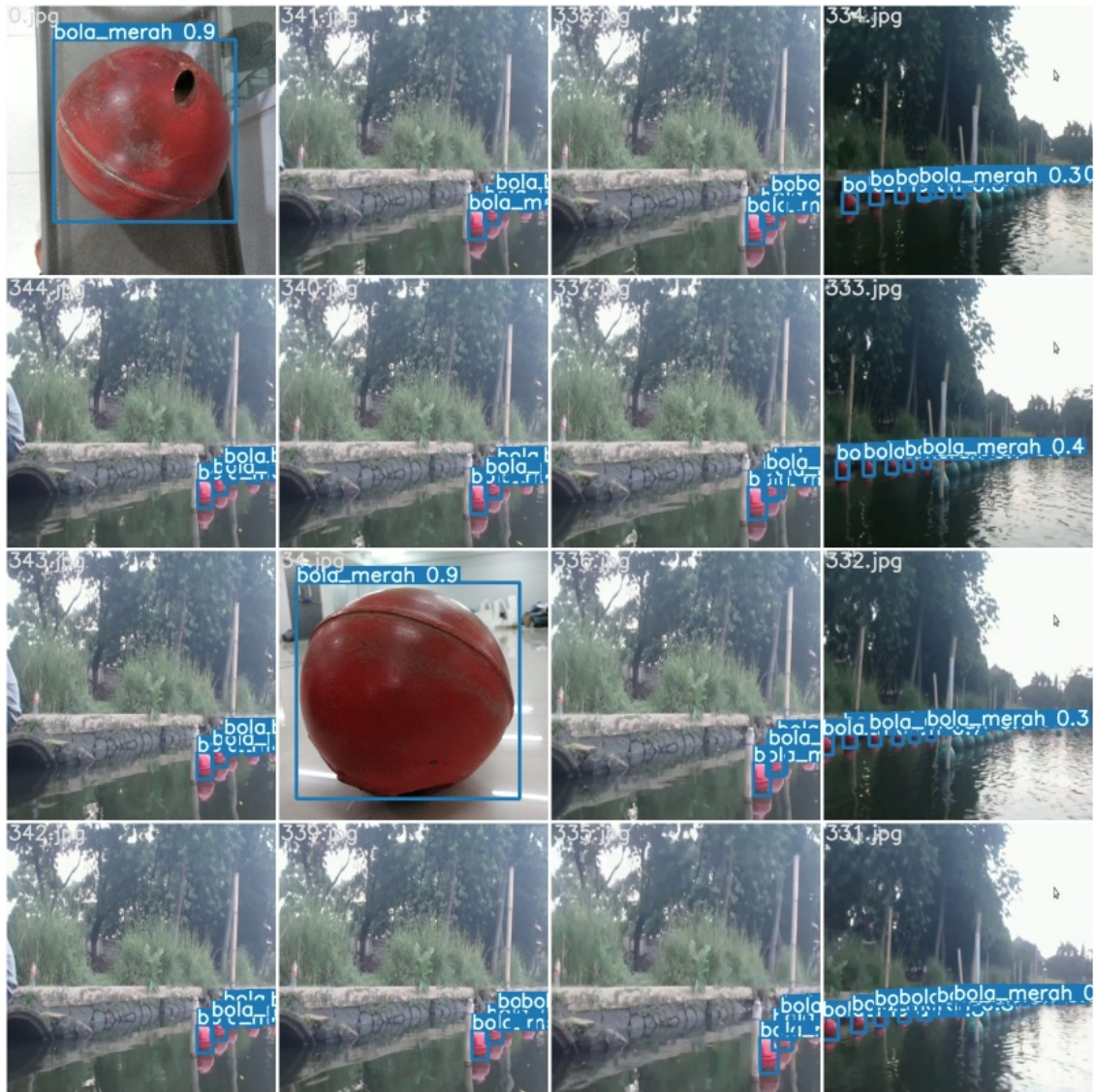
[52]: `Image(filename='runs/exp1/results.png', width=900)`

[52]:



```
[51]: Image(filename='runs/exp1/test_batch0_pred.jpg', width=900) # view test image_
      ↪ predictions
```

[51]:



Training losses and performance metrics are saved to Tensorboard and also to a runs/exp0/results.txt logfile. results.txt is plotted as results.png after training completes. Partially completed results.txt files can be plotted with from utils.utils import plot_results; plot_results(). Here we show YOLOv5s trained on coco128 to 300 epochs, starting from scratch (blue), and from pretrained yolov5s.pt (orange).