

Introduction

This notebook was written in Google Colab. We used their GPU-time to train, evaluate and test our model.

This project was made by myself (Marinus v/d Ende & Matthijs Prinsen)

✓ Custom packages (Dependencies)

Since we curated our dataset on Roboflow, we need to pip install it first.

```
1 !pip install roboflow
```

 Show hidden output


✓ Initialization

Then we get to import os for directory handling, and torch for training.

CUDA is for GPU utilization during training. We will want this minimise training time.

We used Google's T4 GPU with 15GB of VRAM.

```
1 import os
2 import torch
3
4 if torch.cuda.is_available():
5     print("CUDA is available! 🎉")
6     print(f"CUDA version: {torch.version.cuda}")
7     print(f"Number of GPUs: {torch.cuda.device_count()}")
8 else:
9     print("CUDA is not available. 😞")
10
11 !nvidia-smi
```

 CUDA is not available. 😞
/bin/bash: line 1: nvidia-smi: command not found

You need to select a GPU on runtime in order to use CUDA.

✓ Clone YOLOv7

Next, we can clone the YOLOv7 repository and install dependencies.

```
1 # Download YOLOv7 repository and install requirements
2 if not os.path.isdir("yolov7"):
3     !git clone https://github.com/WongKinYiu/yolov7
4 %cd yolov7
5 !pip install -r requirements.txt
```


 Show hidden output

✓ Dataset from Roboflow

Since we curated our datasets on Roboflow, we can programically import and extract the data in the right format for the YOLO model.

This model requires a custom .yaml file with information about classes and the directories for the train, validation and test sets.

```
1 from roboflow import Roboflow
2 rf = Roboflow(api_key="bpTnKg4hhE850z3ox0Fg")
3 project = rf.workspace("endexspace").project("supermarket-items-yolov7")
4 version = project.version(2)
5 dataset = version.download("yolov7")
```

 loading Roboflow workspace...
loading Roboflow project...
Downloading Dataset Version Zip in Supermarket-Items-(YOLOv7)-2 to yolov7pytorch:: 100%|██████████| 419317/419317 [00:08<00:00, 521:

Extracting Dataset Version Zip to Supermarket-Items-(YOLOv7)-2 in yolov7pytorch:: 100%|██████████| 8264/8264 [00:06<00:00, 1321.5011

✓ Preparing model for training

Now that we have:

1. Initialized our environment and enabled GPU usage,
2. Cloned the model we want to use (YOLOv7)
3. Downloaded our dataset from Roboflow

We can get the starting weights for the tiny YOLOv7 model.

```
1 # Download the Tiny model weights.
2 !wget https://github.com/WongKinYiu/yolov7/releases/download/v0.1/yolov7-tiny.pt
```

 [Show hidden output](#)

Since we are using the tiny YOLOv7 model we are going to create our own configuration file for the model architecture.

All we're really changing is the class count. Originally it had a class count of 80, we are changing that to 31

```
1 %%writefile yolov7/cfg/training/yolov7_grocery-tiny.yaml
2 # parameters
3 nc: 31 # number of classes
4 depth_multiple: 1.0 # model depth multiple
5 width_multiple: 1.0 # layer channel multiple
6
7 # anchors
8 anchors:
9   - [10,13, 16,30, 33,23] # P3/8
10  - [30,61, 62,45, 59,119] # P4/16
11  - [116,90, 156,198, 373,326] # P5/32
12
13 # yolov7-tiny backbone
14 backbone:
15   # [from, number, module, args] c2, k=1, s=1, p=None, g=1, act=True
16   [[-1, 1, Conv, [32, 3, 2, None, 1, nn.LeakyReLU(0.1)]], # 0-P1/2
17
18   [-1, 1, Conv, [64, 3, 2, None, 1, nn.LeakyReLU(0.1)]], # 1-P2/4
19
20   [-1, 1, Conv, [32, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
21   [-2, 1, Conv, [32, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
22   [-1, 1, Conv, [32, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
23   [-1, 1, Conv, [32, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
24   [[-1, -2, -3, -4], 1, Concat, [1]],
25   [-1, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 7
26
27   [-1, 1, MP, []], # 8-P3/8
28   [-1, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
29   [-2, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
30   [-1, 1, Conv, [64, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
31   [-1, 1, Conv, [64, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
32   [[-1, -2, -3, -4], 1, Concat, [1]],
33   [-1, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 14
34
35   [-1, 1, MP, []], # 15-P4/16
36   [-1, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
37   [-2, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
38   [-1, 1, Conv, [128, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
39   [-1, 1, Conv, [128, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
40   [[-1, -2, -3, -4], 1, Concat, [1]],
41   [-1, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 21
42
43   [-1, 1, MP, []], # 22-P5/32
44   [-1, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
45   [-2, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
46   [-1, 1, Conv, [256, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
47   [-1, 1, Conv, [256, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
48   [[-1, -2, -3, -4], 1, Concat, [1]],
49   [-1, 1, Conv, [512, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 28
50 ]
51
52 # yolov7-tiny head
53 head:
54   [[-1, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
55   [-2, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
56   [-1, 1, SP, [5]],
57   [-2, 1, SP, [9]],
```

```

58  [-3, 1, SP, [13]],
59  [[-1, -2, -3, -4], 1, Concat, [1]],
60  [-1, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
61  [[-1, -7], 1, Concat, [1]],
62  [-1, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 37
63
64  [-1, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
65  [-1, 1, nn.Upsample, [None, 2, 'nearest']],
66  [21, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # route backbone P4
67  [[-1, -2], 1, Concat, [1]],
68
69  [-1, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
70  [-2, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
71  [-1, 1, Conv, [64, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
72  [-1, 1, Conv, [64, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
73  [[-1, -2, -3, -4], 1, Concat, [1]],
74  [-1, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 47
75
76  [-1, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
77  [-1, 1, nn.Upsample, [None, 2, 'nearest']],
78  [14, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # route backbone P3
79  [[-1, -2], 1, Concat, [1]],
80
81  [-1, 1, Conv, [32, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
82  [-2, 1, Conv, [32, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
83  [-1, 1, Conv, [32, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
84  [-1, 1, Conv, [32, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
85  [[-1, -2, -3, -4], 1, Concat, [1]],
86  [-1, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 57
87
88  [-1, 1, Conv, [128, 3, 2, None, 1, nn.LeakyReLU(0.1)]],
89  [[-1, 47], 1, Concat, [1]],
90
91  [-1, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
92  [-2, 1, Conv, [64, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
93  [-1, 1, Conv, [64, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
94  [-1, 1, Conv, [64, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
95  [[-1, -2, -3, -4], 1, Concat, [1]],
96  [-1, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 65
97
98  [-1, 1, Conv, [256, 3, 2, None, 1, nn.LeakyReLU(0.1)]],
99  [[-1, 37], 1, Concat, [1]],
100
101 [-1, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
102 [-2, 1, Conv, [128, 1, 1, None, 1, nn.LeakyReLU(0.1)]],
103 [-1, 1, Conv, [128, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
104 [-1, 1, Conv, [128, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
105 [[-1, -2, -3, -4], 1, Concat, [1]],
106 [-1, 1, Conv, [256, 1, 1, None, 1, nn.LeakyReLU(0.1)]], # 73
107
108 [57, 1, Conv, [128, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
109 [65, 1, Conv, [256, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
110 [73, 1, Conv, [512, 3, 1, None, 1, nn.LeakyReLU(0.1)]],
111
112 [[74,75,76], 1, IDetect, [nc, anchors]], # Detect(P3, P4, P5)
113 ]

```

➦ Writing yolov7/cfg/training/yolov7_grocery-tiny.yaml

✓ Training

We can now start with the training.

```

1 dataset_location: str = dataset.location
2 loc: str = f"{dataset_location}/data.yaml\"
3 print(loc)
4
5 !python yolov7/train.py --epochs 50 --workers 8 --device 0 --batch-size 32 \
6 --data {loc} --img 640 640 --cfg yolov7/cfg/training/yolov7_grocery-tiny.yaml \
7 --weights 'yolov7-tiny.pt' --name yolov7_tiny_grocery_fixed_res --hyp yolov7/data/hyp.scratch.tiny.yaml

```



```

61         -2 1 10512 models.common.Conv [256, 64, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
62         -1 1 36992 models.common.Conv [64, 64, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
63         -1 1 36992 models.common.Conv [64, 64, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
64 [-1, -2, -3, -4] 1 0 models.common.Concat [1]
65         -1 1 33024 models.common.Conv [256, 128, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
66         -1 1 295424 models.common.Conv [128, 256, 3, 2, None, 1, LeakyReLU(negative_slope=0.1)
67 [-1, 37] 1 0 models.common.Concat [1]
68         -1 1 65792 models.common.Conv [512, 128, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
69         -2 1 65792 models.common.Conv [512, 128, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
70         -1 1 147712 models.common.Conv [128, 128, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
71         -1 1 147712 models.common.Conv [128, 128, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
72 [-1, -2, -3, -4] 1 0 models.common.Concat [1]
73         -1 1 131584 models.common.Conv [512, 256, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
74         57 1 73984 models.common.Conv [64, 128, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
75         65 1 295424 models.common.Conv [128, 256, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
76         73 1 1180672 models.common.Conv [256, 512, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
77 [74, 75, 76] 1 95606 models.yolo.IDetect [30, [[10, 13, 16, 30, 33, 23], [30, 61, 62, 45, 59,
/usr/local/lib/python3.11/dist-packages/torch/functional.py:534: UserWarning: torch.meshgrid: in an upcoming release, it will be
  return _VF.meshgrid(tensors, **kwargs) # type: ignore[attr-defined]
Model Summary: 263 layers, 6093462 parameters, 6093462 gradients, 13.4 GFLOPS

```

```

Transferred 330/344 items from yolov7-tiny.pt
Scaled weight_decay = 0.0005
Optimizer groups: 58 .bias, 58 conv.weight, 61 other
train: Scanning 'Supermarket-Items-(YOLOv7)-2/train/labels' images and labels... 3645 found, 0 missing, 0 empty, 0 corrupted: 100%
train: New cache created: Supermarket-Items-(YOLOv7)-2/train/labels.cache
val: Scanning 'Supermarket-Items-(YOLOv7)-2/valid/labels' images and labels... 324 found, 0 missing, 0 empty, 0 corrupted: 100%
val: New cache created: Supermarket-Items-(YOLOv7)-2/valid/labels.cache

autoanchor: Analyzing anchors... anchors/target = 5.12, Best Possible Recall (BPR) = 1.0000
/content/yolov7/train.py:299: FutureWarning: `torch.cuda.amp.GradScaler(args...)` is deprecated. Please use `torch.amp.GradScaler
  scaler = amp.GradScaler(enabled=cuda)
Image sizes 640 train, 640 test
Using 2 dataloader workers
Logging results to runs/train/yolov7_tiny_grocery_fixed_res
Starting training for 50 epochs...

```

Epoch	gpu_mem	box	obj	cls	total	labels	img_size
0% 0/114	[00:00<?, ?it/s]	/content/yolov7/train.py:360: FutureWarning: `torch.cuda.amp.autocast(args...)` is deprecated. Please use `torch.amp.autocast(enabled=cuda):`					
0/49	1.4G	0.06751	0.04262	0.08366	0.1938	305	640: 100% 114/114 [02:35<00:00, 1.36s/it]
	Class	Images	Labels		P	R	mAP@.5 mAP@.5:.95: 100% 6/6 [00:08<00:00, 1.37s/it]
	all	324	1867		0.06	0.209	0.0689 0.0265

Epoch	gpu_mem	box	obj	cls	total	labels	img_size
1/49	5.82G	0.04783	0.05099	0.07148	0.1703	283	640: 100% 114/114 [02:19<00:00, 1.22s/it]
	Class	Images	Labels		P	R	mAP@.5 mAP@.5:.95: 100% 6/6 [00:04<00:00, 1.24it/s]
	all	324	1867		0.381	0.277	0.205 0.0801

We were averaging around 1.2 seconds per iteration and I wanted to try and optimise that.

I therefore stopped the training and adjusted the batch-size and workers to try and find an optimal combination of settings to utilize as much of the resources in colab as possible.

```

1 !python yolov7/train.py \
2 --resume \
3 --batch-size 64 \
4 --workers 8

```

```

2025-01-27 20:43:38.083653: E external/local_xla/xla/stream_executor/cuda/cuda_fft.cc:485] Unable to register cuFFT factory: Att
2025-01-27 20:43:38.103241: E external/local_xla/xla/stream_executor/cuda/cuda_dnn.cc:8454] Unable to register cuDNN factory: Att
2025-01-27 20:43:38.109266: E external/local_xla/xla/stream_executor/cuda/cuda_blas.cc:1452] Unable to register cuBLAS factory: A
2025-01-27 20:43:38.123405: I tensorflow/core/platform/cpu_feature_guard.cc:210] This TensorFlow binary is optimized to use avail
To enable the following instructions: AVX2 AVX512F FMA, in other operations, rebuild TensorFlow with the appropriate compiler fla
2025-01-27 20:43:39.195652: W tensorflow/compiler/tf2tensorrt/utils/py_utils.cc:38] TF-TRT Warning: Could not find TensorRT
Resuming training from ./runs/train/yolov7_tiny_grocery_fixed_res/weights/last.pt
YOLOR v0.1-128-ga207844 torch 2.5.1+cu121 CUDA:0 (Tesla T4, 15102.0625MB)

```

```

Namespace(weights='./runs/train/yolov7_tiny_grocery_fixed_res/weights/last.pt', cfg='', data='/content/Supermarket-Items-(YOLOv7)
tensorboard: Start with 'tensorboard --logdir runs/train', view at http://localhost:6006/
hyperparameters: lr0=0.01, lr1=0.01, momentum=0.937, weight_decay=0.0005, warmup_epochs=3.0, warmup_momentum=0.8, warmup_bias_lr=
/content/yolov7/train.py:71: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), whi
  run_id = torch.load(weights, map_location=device).get('wandb_id') if weights.endswith('.pt') and os.path.isfile(weights) else N
wandb: Currently logged in as: endex-space (endex-space-university-of-groningen). Use `wandb login --relogin` to force relogin
wandb: Using wandb-core as the SDK backend. Please refer to https://wandb.me/wandb-core for more information.
wandb: Tracking run with wandb version 0.19.4
wandb: Run data is saved locally in /content/wandb/run-20250127_204344-vqer9kz6
wandb: Run `wandb offline` to turn off syncing.
wandb: Resuming run yolov7_tiny_grocery_fixed_res
wandb: View project at https://wandb.ai/endex-space-university-of-groningen/YOLOR
wandb: View run at https://wandb.ai/endex-space-university-of-groningen/YOLOR/runs/vqer9kz6
/content/yolov7/train.py:87: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default value), whi
  ckpt = torch.load(weights, map_location=device) # load checkpoint

```

	from	n	params	module	arguments
0	-1	1	928	models.common.Conv	[3, 32, 3, 2, None, 1, LeakyReLU(negative_slope=0.1)]
1	-1	1	18560	models.common.Conv	[32, 64, 3, 2, None, 1, LeakyReLU(negative_slope=0.1)]

2	-1	1	2112	models.common.Conv	[64, 32, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
3	-2	1	2112	models.common.Conv	[64, 32, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
4	-1	1	9280	models.common.Conv	[32, 32, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
5	-1	1	9280	models.common.Conv	[32, 32, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
6	[-1, -2, -3, -4]	1	0	models.common.Concat	[1]
7	-1	1	8320	models.common.Conv	[128, 64, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
8	-1	1	0	models.common.MP	[]
9	-1	1	4224	models.common.Conv	[64, 64, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
10	-2	1	4224	models.common.Conv	[64, 64, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
11	-1	1	36992	models.common.Conv	[64, 64, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
12	-1	1	36992	models.common.Conv	[64, 64, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
13	[-1, -2, -3, -4]	1	0	models.common.Concat	[1]
14	-1	1	33024	models.common.Conv	[256, 128, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
15	-1	1	0	models.common.MP	[]
16	-1	1	16640	models.common.Conv	[128, 128, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
17	-2	1	16640	models.common.Conv	[128, 128, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
18	-1	1	147712	models.common.Conv	[128, 128, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
19	-1	1	147712	models.common.Conv	[128, 128, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
20	[-1, -2, -3, -4]	1	0	models.common.Concat	[1]
21	-1	1	131584	models.common.Conv	[512, 256, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
22	-1	1	0	models.common.MP	[]
23	-1	1	66048	models.common.Conv	[256, 256, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
24	-2	1	66048	models.common.Conv	[256, 256, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
25	-1	1	590336	models.common.Conv	[256, 256, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
26	-1	1	590336	models.common.Conv	[256, 256, 3, 1, None, 1, LeakyReLU(negative_slope=0.1)
27	[-1, -2, -3, -4]	1	0	models.common.Concat	[1]
28	-1	1	525312	models.common.Conv	[1024, 512, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)
29	-1	1	131584	models.common.Conv	[512, 256, 1, 1, None, 1, LeakyReLU(negative_slope=0.1)

Unfortunately, after some configuration I could not get the model to train faster than 1.2 s/it.

✓ Testing the trained Model

Now that we have trained the model, and can view the beautiful convergence of the training curves we get to test our model.

```
1 dataset_location = dataset.location
2 loc: str = f"{dataset_location}/test/images/"
3 print(loc)
4
5 # Run evaluation
6 !python yolov7/detect.py \
7   --weights best.pt \
8   --conf 0.1 \
9   --source {loc}

"/content/Supermarket-Items-(YOLOv7)-2/test/images"
Namespace(weights=['best.pt'], source='/content/Supermarket-Items-(YOLOv7)-2/test/images', img_size=640, conf_thres=0.1, iou_thres=0.45)
YOLOR v0.1-128-ga207844 torch 2.5.1+cu121 CPU

/content/yolov7/models/experimental.py:252: FutureWarning: You are using `torch.load` with `weights_only=False` (the current default)
  ckpt = torch.load(w, map_location=map_location) # load
Fusing layers...
IDetect.fuse
/usr/local/lib/python3.11/dist-packages/torch/functional.py:534: UserWarning: torch.meshgrid: in an upcoming release, it will be replaced
  return _VF.meshgrid(tensors, **kwargs) # type: ignore[attr-defined]
Model Summary: 208 layers, 6086070 parameters, 0 gradients, 13.3 GFLOPS
Convert model to Traced-model...
traced_script_module saved!
model is traced!

2 bananas, 1 bread, Done. (397.1ms) Inference, (1.4ms) NMS
Traceback (most recent call last):
  File "/content/yolov7/detect.py", line 196, in <module>
    detect()
  File "/content/yolov7/detect.py", line 136, in detect
    cv2.imshow(str(p), im0)
cv2.error: OpenCV(4.10.0) /io/opencv/modules/highgui/src/window.cpp:1301: error: (-2:Unspecified error) The function is not implemented
```

Below is some simple code to print out our model's predictions on our testing data.

The model performed spectacularly with a mAP of 95% on our testing data. It can be seen that the model sometimes misses an image but it rarely mislabels objects.

A success in our books.

```
1 #display inference on ALL test images
2
3 import glob
4 from IPython.display import Image, display
```

```
5
6 i = 0
7 limit = 10000 # max images to print
8 for imageName in glob.glob('/content/runs/detect/exp2/*.jpg'): #assuming JPG
9     if i < limit:
10         display(Image(filename=imageName))
11         print("\n")
12     i = i + 1
```