

```
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
!git clone https://github.com/ultralytics/yolov5
%cd yolov5
%pip install -qr requirements.txt
```

Cloning into 'yolov5'...

```
remote: Enumerating objects: 16836, done.
remote: Counting objects: 100% (11/11), done.
remote: Compressing objects: 100% (11/11), done.
remote: Total 16836 (delta 1), reused 6 (delta 0), pack-reused 16825
Receiving objects: 100% (16836/16836), 15.57 MiB | 12.92 MiB/s, done.
Resolving deltas: 100% (11541/11541), done.
/content/yolov5
```

```
41.3/41.3 kB 1.6 MB/s eta 0:00:00
207.3/207.3 kB 9.3 MB/s eta 0:00:00
4.5/4.5 MB 75.6 MB/s eta 0:00:00
64.9/64.9 kB 6.3 MB/s eta 0:00:00
865.5/865.5 kB 50.0 MB/s eta 0:00:00
62.7/62.7 kB 6.1 MB/s eta 0:00:00
```

ERROR: pip's dependency resolver does not currently take into account all the packages that are installed. This behaviour is the source of the following dependency conflicts.
google-colab 1.0.0 requires requests==2.31.0, but you have requests 2.32.3 which is incompatible.

```
import torch
import utils
display = utils.notebook_init()
```

YOLOv5 v7.0-350-g6096750f Python-3.10.12 torch-2.3.1+cu121 CUDA:0 (Tesla T4, 15102MiB)
Setup complete (2 CPUs, 12.7 GB RAM, 33.6/78.2 GB disk)

```
!unzip -q /content/drive/MyDrive/weeds.zip -d /content/yolov5/
```

```
import os
import shutil
from sklearn.model_selection import train_test_split

# Paths
data_dir = '/content/yolov5/agri_data/data'
train_images_dir = '/content/yolov5/agri_data/train/images'
val_images_dir = '/content/yolov5/agri_data/val/images'
train_labels_dir = '/content/yolov5/agri_data/train/labels'
val_labels_dir = '/content/yolov5/agri_data/val/labels'

# Create directories
os.makedirs(train_images_dir, exist_ok=True)
os.makedirs(val_images_dir, exist_ok=True)
os.makedirs(train_labels_dir, exist_ok=True)
os.makedirs(val_labels_dir, exist_ok=True)

# Get list of images and corresponding labels
images = [f for f in os.listdir(data_dir) if f.endswith('.jpeg')]
labels = [f for f in os.listdir(data_dir) if f.endswith('.txt')]

# Split data into train and validation sets
train_images, val_images, train_labels, val_labels = train_test_split(images, labels, test_size=0.2, random_state=42)

# Copy files to respective directories
for img, lbl in zip(train_images, train_labels):
    shutil.copy(os.path.join(data_dir, img), train_images_dir)
    shutil.copy(os.path.join(data_dir, lbl), train_labels_dir)

for img, lbl in zip(val_images, val_labels):
    shutil.copy(os.path.join(data_dir, img), val_images_dir)
    shutil.copy(os.path.join(data_dir, lbl), val_labels_dir)
```

```
data_yaml_content = """
train: /content/yolov5/agri_data/train/images
val: /content/yolov5/agri_data/val/images

nc: 2
names: ['crop', 'weed']
"""

with open('/content/yolov5/data.yaml', 'w') as f:
    f.write(data_yaml_content)
```

```
# For Conversion
# !python export.py --weights best.pt --img 512 --include tflite
```

```
!python train.py --img 320 --batch 32 --epochs 20 --data /content/yolov5/data.yaml --weights yolov5s.pt --cache --name weed_detection
```

Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Size
11/19	2.11G	0.0377	0.01709	0.004619	43	320: 100% 33/33 [00:06<00:00, 4.77it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 4.68it/s]
	all	260	48	0.139	0.561	0.156 0.107
12/19	2.11G	0.0363	0.01726	0.004269	32	320: 100% 33/33 [00:06<00:00, 5.31it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 3.15it/s]
	all	260	48	0.139	0.664	0.188 0.132
13/19	2.11G	0.03576	0.01724	0.003995	38	320: 100% 33/33 [00:08<00:00, 3.97it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 4.62it/s]
	all	260	48	0.133	0.575	0.131 0.0811
14/19	2.11G	0.03544	0.01718	0.004783	39	320: 100% 33/33 [00:05<00:00, 5.73it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 3.52it/s]
	all	260	48	0.124	0.733	0.169 0.124
15/19	2.11G	0.03516	0.01771	0.00392	35	320: 100% 33/33 [00:09<00:00, 3.62it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 4.73it/s]
	all	260	48	0.144	0.738	0.16 0.116
16/19	2.11G	0.03457	0.01743	0.003884	54	320: 100% 33/33 [00:05<00:00, 5.77it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 4.67it/s]
	all	260	48	0.148	0.7	0.167 0.114
17/19	2.11G	0.03233	0.01645	0.003944	38	320: 100% 33/33 [00:09<00:00, 3.54it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 3.66it/s]
	all	260	48	0.13	0.65	0.161 0.113
18/19	2.11G	0.03189	0.01661	0.003611	35	320: 100% 33/33 [00:05<00:00, 5.73it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 4.63it/s]
	all	260	48	0.147	0.621	0.176 0.124
19/19	2.11G	0.03266	0.01673	0.003681	44	320: 100% 33/33 [00:08<00:00, 4.07it/s]
	Class	Images	Instances	P	R	mAP50 mAP50-95: 100% 5/5 [00:01<00:00, 3.22it/s]
	all	260	48	0.149	0.611	0.193 0.132

20 epochs completed in 0.053 hours.

Optimizer stripped from runs/train/weed_detection/weights/last.pt, 14.3MB

Optimizer stripped from runs/train/weed_detection/weights/best.pt, 14.3MB

Validating runs/train/weed_detection/weights/best.pt...

Fusing layers...

Model summary: 157 layers, 7015519 parameters, 0 gradients, 15.8 GFLOPs

Class	Images	Instances	P	R	mAP50	mAP50-95: 100% 5/5 [00:02<00:00, 2.48it/s]
all	260	48	0.15	0.611	0.192	0.131
crop	260	20	0.174	0.9	0.22	0.174
weed	260	28	0.125	0.321	0.165	0.0886

Results saved to runs/train/weed_detection

```
!python val.py --weights runs/train/weed_detection/weights/best.pt --data /content/yolov5/data.yaml --img 320
```

```
val: data=/content/yolov5/data.yaml, weights=['runs/train/weed_detection/weights/best.pt'], batch_size=32, imgsz=320, conf_thres=0.6
YOLOv5 v7.0-350-g6096750f Python-3.10.12 torch-2.3.1+cu121 CUDA:0 (Tesla T4, 15102MiB)
```

Fusing layers...

Model summary: 157 layers, 7015519 parameters, 0 gradients, 15.8 GFLOPs

val: Scanning /content/yolov5/agri_data/val/labels.cache... 39 images, 221 backgrounds, 0 corrupt: 100% 260/260 [00:00<?, ?it/s]

Class	Images	Instances	P	R	mAP50	mAP50-95: 100%	9/9 [00:05<00:00, 1.71it/s]
all	260	48	0.15	0.611	0.192	0.132	
crop	260	20	0.174	0.9	0.22	0.174	
weed	260	28	0.125	0.321	0.165	0.0892	

Speed: 0.1ms pre-process, 3.3ms inference, 8.4ms NMS per image at shape (32, 3, 320, 320)

Results saved to runs/val/exp

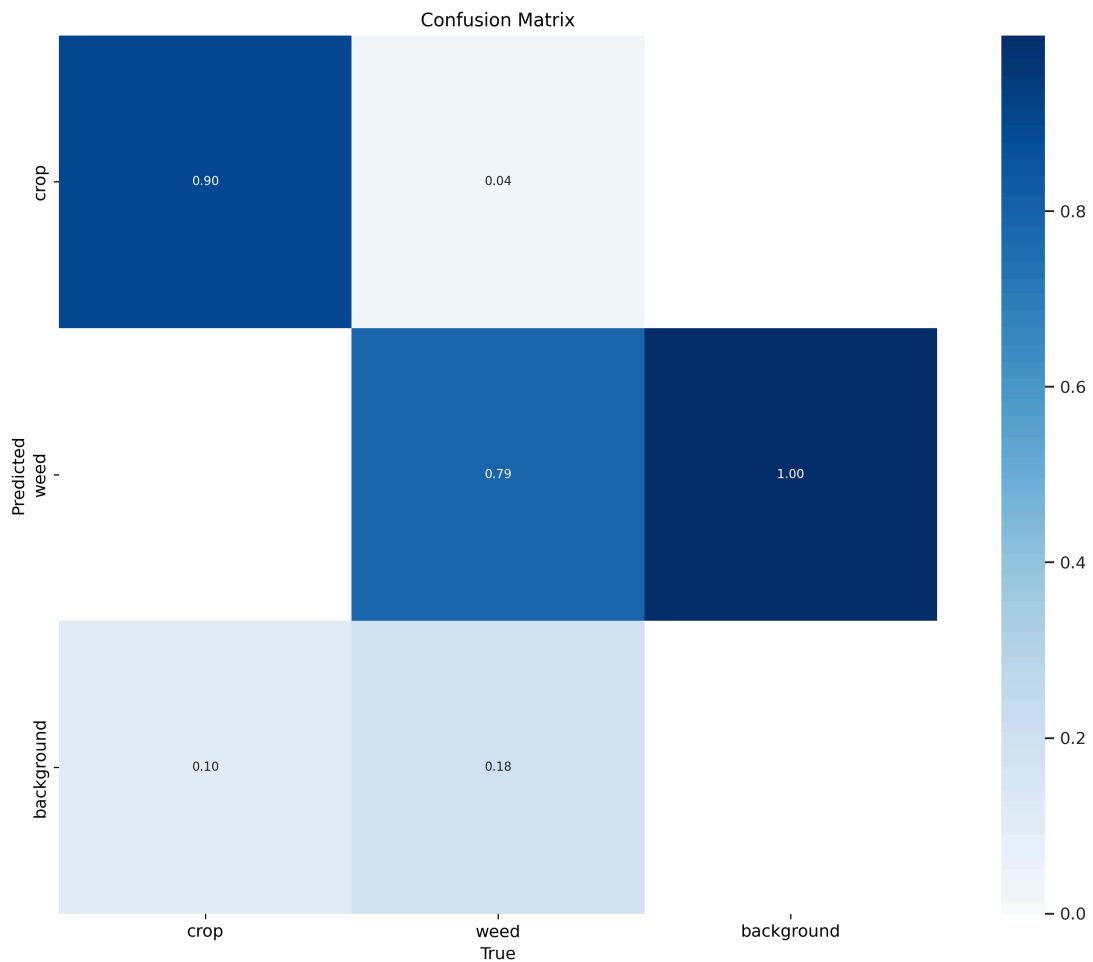
```

from IPython.display import Image, display

# Define the path to the confusion matrix image
confusion_matrix_path = 'runs/val/exp/confusion_matrix.png'

# Display the confusion matrix image
display(Image(filename=confusion_matrix_path))

```



> D

[] 3 cells hidden

