

```
# This command displays the status of the installed GPUs, showing
# details like the driver version,
# GPU name, memory usage, and the processes using the GPU.
!nvidia-smi

Mon Jul 22 00:04:11 2024
+-----+
| NVIDIA-SMI 535.104.05                 Driver Version: 535.104.05 CUDA
Version: 12.2
+-----+
| GPU Name                  Persistence-M | Bus-Id      Disp.A |
Volatile Uncorr. ECC |
| Fan Temp   Perf          Pwr:Usage/Cap |           Memory-Usage |
GPU-Util Compute M. |
| MIG M. |
|
=====+=====+=====+=====
| 0  Tesla T4                Off | 00000000:00:04.0 Off |
0 |
| N/A  51C     P8            10W / 70W | 0MiB / 15360MiB |
0%       Default |
|
N/A |
+-----+
+-----+
| Processes:
|
| GPU  GI  CI      PID  Type  Process name
GPU Memory |
|          ID  ID
Usage      |
|
=====+
| No running processes found
|
+-----+
```

# Detecting Waldo with YOLOv8 and YOLOv5: A Hands-on Object Detection Training

## Introduction

Welcome to the hands-on training on detecting Waldo using the powerful object detection models YOLOv8 and YOLOv5. In this training, you will learn the complete workflow for creating a object detection system, including dataset preparation, model training, and evaluation.

1. Setup and Installation
2. Dataset Collection and Annotation
3. Preparing the Data
4. Training with YOLOv5
5. Training with YOLOv8
6. Evaluating the Models
7. Using SAHI technique to detect waldo

## Setup and Installation

To begin, we'll need to set up our environment and install the necessary dependencies.

```
import os
HOME = os.getcwd()
print(HOME)

/content

import pandas as pd
import matplotlib.pyplot as plt
import xml.etree.ElementTree as ET
import glob
from PIL import Image
```

## Installing and Checking Ultralytics

### Install Ultralytics

To install the specific version 8.0.196 of the `ultralytics` package, use the following pip command:

```
```bash !pip install ultralytics==8.0.196
```

```
# Pip install method (recommended)
```

```
!pip install ultralytics==8.0.196
from IPython import display
display.clear_output()

import ultralytics
ultralytics.checks()

Ultralytics YOLOv8.0.196 □ Python-3.10.12 torch-2.3.1+cu121 CUDA:0
(Tesla T4, 15102MiB)
Setup complete □ (2 CPUs, 12.7 GB RAM, 30.1/78.2 GB disk)
```

# Using Ultralytics YOLO

## Import YOLO from Ultralytics

To utilize the YOLO (You Only Look Once) model from the `ultralytics` package, import it as follows:

```
```python from ultralytics import YOLO
```

```
from ultralytics import YOLO

from IPython.display import display, Image
# display.clear_output()

from ultralytics import YOLO

from IPython.display import display, Image
# display.clear_output()
```

## Setting Up Dataset Directory and Installing Roboflow

To organize the datasets, first create a new directory named `datasets` in your home directory. You can do this using the following command:

```
!mkdir {HOME}/datasets  
%cd {HOME}/datasets  
  
!pip install roboflow --quiet  
  
!pip install roboflow  
  
/content/datasets
```

---

0:00:00 76.7/76.7 kB 2.5 MB/s eta  
0:00:00 178.7/178.7 kB 7.2 MB/s eta

————— 54.5/54.5 kB 7.0 MB/s eta  
0:00:00  
ent already satisfied: roboflow in /usr/local/lib/python3.10/dist-packages (1.1.36)  
Requirement already satisfied: certifi in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (2024.7.4)  
Requirement already satisfied: chardet==4.0.0 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (4.0.0)  
Requirement already satisfied: idna==3.7 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (3.7)  
Requirement already satisfied: cycler in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (0.12.1)  
Requirement already satisfied: kiwisolver>=1.3.1 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.4.5)  
Requirement already satisfied: matplotlib in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (3.7.1)  
Requirement already satisfied: numpy>=1.18.5 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.25.2)  
Requirement already satisfied: opencv-python-headless==4.10.0.84 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (4.10.0.84)  
Requirement already satisfied: Pillow>=7.1.2 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (9.4.0)  
Requirement already satisfied: python-dateutil in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (2.8.2)  
Requirement already satisfied: python-dotenv in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.0.1)  
Requirement already satisfied: requests in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (2.31.0)  
Requirement already satisfied: six in /usr/local/lib/python3.10/dist-packages (from roboflow) (1.16.0)  
Requirement already satisfied: urllib3>=1.26.6 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (2.0.7)  
Requirement already satisfied: tqdm>=4.41.0 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (4.66.4)  
Requirement already satisfied: PyYAML>=5.3.1 in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (6.0.1)  
Requirement already satisfied: requests-toolbelt in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.0.0)  
Requirement already satisfied: filetype in  
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.2.0)  
Requirement already satisfied: contourpy>=1.0.1 in  
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow) (1.2.1)  
Requirement already satisfied: fonttools>=4.22.0 in  
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow) (4.53.1)  
Requirement already satisfied: packaging>=20.0 in  
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow) (24.1)

```
Requirement already satisfied: pyparsing>=2.3.1 in
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow)
(3.1.2)
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.10/dist-packages (from requests->roboflow)
(3.3.2)
```

## Roboflow Dataset Download

The following Python code demonstrates how to use the Roboflow library to download a dataset for a specific project and version. Here's a step-by-step breakdown of the code:

```
from roboflow import Roboflow
rf = Roboflow(api_key="vaIQ9aYmyRqVHGw89lcc")
project = rf.workspace("bookdetection-hc8pr").project("waldo7")
version = project.version(1)
dataset = version.download("yolov8")

loading Roboflow workspace...
loading Roboflow project...

Downloading Dataset Version Zip in waldo7-1 to yolov8:: 100%|
██████████| 32780/32780 [00:01<00:00, 24938.49it/s]

Extracting Dataset Version Zip to waldo7-1 in yolov8:: 100%|
██████████| 447/447 [00:00<00:00, 2590.40it/s]
```

## Counting and Visualizing Image Files

The following code snippet performs the task of counting image files in specified directories and visualizing the results using a bar chart. Here's a step-by-step breakdown of each part:

```
train_images_dir = '/content/datasets/waldo7-1/train/images'
val_images_dir = '/content/datasets/waldo7-1/valid/images'

def count_images(directory):
    """Count the number of image files in the given directory."""
    return len([name for name in os.listdir(directory) if
os.path.isfile(os.path.join(directory, name)) and
name.lower().endswith('.png', '.jpg', '.jpeg')])

# Count the number of images in train and validation directories
num_train_images = count_images(train_images_dir)
num_val_images = count_images(val_images_dir)
```

```

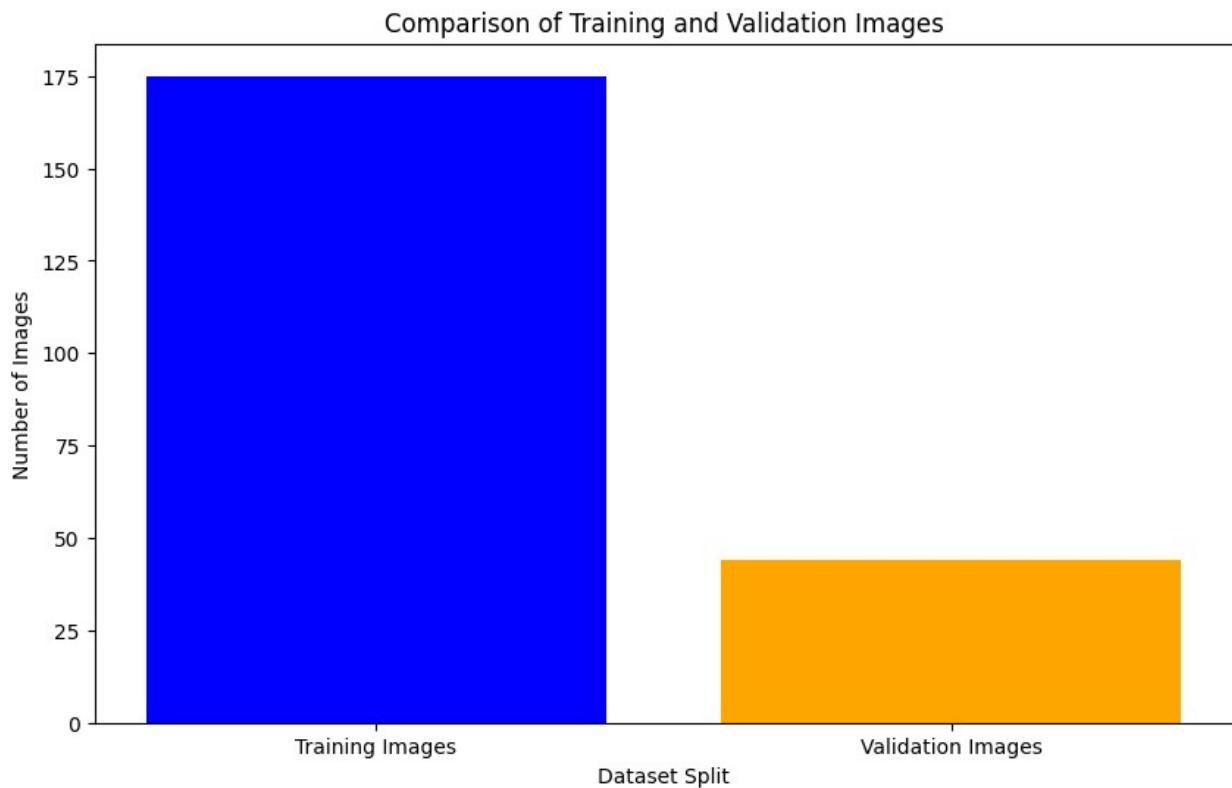
print(num_train_images,"images for training")
print(num_val_images,"images for validation")

175 images for training
44 images for validation

categories = ['Training Images', 'Validation Images']
num_images = [num_train_images, num_val_images]

plt.figure(figsize=(10, 6))
plt.bar(categories, num_images, color=['blue', 'orange'])
plt.xlabel('Dataset Split')
plt.ylabel('Number of Images')
plt.title('Comparison of Training and Validation Images')
plt.show()

```



## YOLOv8 Training Command Explained

### Command Arguments

task=detect:

- Purpose: Specifies the task as object detection. Details: YOLO (You Only Look Once) is designed for object detection. This argument tells YOLO to perform object detection on the provided dataset.

mode=train:

- Purpose: Indicates the mode of operation as training. Details: YOLO can be used for various purposes such as inference (making predictions) or evaluation. By setting mode=train, you are instructing YOLO to train the model

model=yolov8s.pt:

- Purpose: Specifies the model checkpoint file to use. Details: yolov8s.pt refers to the YOLOv8 small model checkpoint file. The .pt extension indicates a PyTorch model file. This file contains pre-trained weights that the training process will start from.

data={dataset.location}/data.yaml or data=/content/datasets/waldo7-1/data.yaml:

- Purpose: Provides the path to the dataset configuration file. Details: {dataset.location} is a placeholder for the actual path to your dataset. data.yaml is the configuration file that contains information about the dataset, such as class names and paths to training and validation images.

epochs=80:

- Purpose: Sets the number of training epochs. Details: An epoch is one complete pass through the training dataset. Setting epochs=80 means the model will be trained for 80 epochs. More epochs generally improve model performance but require more time.

imgsz=640:

- Purpose: Defines the image size for training. Details: imgsz=640 resizes each image to 640x640 pixels before feeding it into the model. This size impacts both the training speed and accuracy.

batch=16:

- Purpose: Specifies the batch size. Details: The batch size is the number of images processed together in one forward/backward pass. A batch size of 16 means that 16 images are processed simultaneously.

optimizer=SGD:

- Purpose: Chooses the optimizer for training. Details: SGD stands for Stochastic Gradient Descent, an optimization algorithm used to minimize the loss function and update the model weights.

```
!yolo task=detect mode=train model=yolov8s.pt  
data=/content/datasets/waldo7-1/data.yaml epochs=80 imgsz=640  
batch=16 optimizer=SGD plots=True
```

```
Ultralytics YOLOV8.2.61 □ Python-3.10.12 torch-2.3.1+cu121 CUDA:0  
(Tesla T4, 15102MiB)  
engine/trainer: task=detect, mode=train, model=yolov8s.pt,  
data=/content/datasets/waldo7-1/data.yaml, epochs=80, time=None,  
patience=100, batch=16, imgsz=640, save=True, save_period=-1,
```

```

cache=False, device=None, workers=8, project=None, name=train6,
exist_ok=False, pretrained=True, optimizer=SGD, verbose=True, seed=0,
deterministic=True, single_cls=False, rect=False, cos_lr=False,
close_mosaic=10, resume=False, amp=True, fraction=1.0, profile=False,
freeze=None, multi_scale=False, overlap_mask=True, mask_ratio=4,
dropout=0.0, val=True, split=val, save_json=False, save_hybrid=False,
conf=None, iou=0.7, max_det=300, half=False, dnn=False, plots=True,
source=None, vid_stride=1, stream_buffer=False, visualize=False,
augment=False, agnostic_nms=False, classes=None, retina_masks=False,
embed=None, show=False, save_frames=False, save_txt=False,
save_conf=False, save_crop=False, show_labels=True, show_conf=True,
show_boxes=True, line_width=None, format=torchscript, keras=False,
optimize=False, int8=False, dynamic=False, simplify=False, opset=None,
workspace=4, nms=False, lr0=0.01, lrf=0.01, momentum=0.937,
weight_decay=0.0005, warmup_epochs=3.0, warmup_momentum=0.8,
warmup_bias_lr=0.1, box=7.5, cls=0.5, dfl=1.5, pose=12.0, kobj=1.0,
label_smoothing=0.0, nbs=64, hsv_h=0.015, hsv_s=0.7, hsv_v=0.4,
degrees=0.0, translate=0.1, scale=0.5, shear=0.0, perspective=0.0,
flipud=0.0, fliplr=0.5, bgr=0.0, mosaic=1.0, mixup=0.0,
copy_paste=0.0, auto_augment=randaugment, erasing=0.4,
crop_fraction=1.0, cfg=None, tracker=botsort.yaml,
save_dir=runs/detect/train6
Overriding model.yaml nc=80 with nc=1

```

	from	n	params	module
arguments				
0	-1	1	928	ultralytics.nn.modules.conv.Conv
[3, 32, 3, 2]				
1	-1	1	18560	ultralytics.nn.modules.conv.Conv
[32, 64, 3, 2]				
2	-1	1	29056	ultralytics.nn.modules.block.C2f
[64, 64, 1, True]				
3	-1	1	73984	ultralytics.nn.modules.conv.Conv
[64, 128, 3, 2]				
4	-1	2	197632	ultralytics.nn.modules.block.C2f
[128, 128, 2, True]				
5	-1	1	295424	ultralytics.nn.modules.conv.Conv
[128, 256, 3, 2]				
6	-1	2	788480	ultralytics.nn.modules.block.C2f
[256, 256, 2, True]				
7	-1	1	1180672	ultralytics.nn.modules.conv.Conv
[256, 512, 3, 2]				
8	-1	1	1838080	ultralytics.nn.modules.block.C2f
[512, 512, 1, True]				
9	-1	1	656896	ultralytics.nn.modules.block.SPPF
				[512, 512, 5]
10	-1	1	0	torch.nn.modules.upsampling.Upsample
				[None, 2, 'nearest']

```

11           [-1, 6]  1      0
ultralytics.nn.modules.conv.Concat          [1]

12           -1  1    591360  ultralytics.nn.modules.block.C2f
[768, 256, 1]
13           -1  1      0
torch.nn.modules.upsampling.Upsample       [None, 2, 'nearest']

14           [-1, 4]  1      0
ultralytics.nn.modules.conv.Concat          [1]

15           -1  1    148224  ultralytics.nn.modules.block.C2f
[384, 128, 1]
16           -1  1    147712  ultralytics.nn.modules.conv.Conv
[128, 128, 3, 2]
17           [-1, 12] 1      0
ultralytics.nn.modules.conv.Concat          [1]

18           -1  1    493056  ultralytics.nn.modules.block.C2f
[384, 256, 1]
19           -1  1    590336  ultralytics.nn.modules.conv.Conv
[256, 256, 3, 2]
20           [-1, 9]  1      0
ultralytics.nn.modules.conv.Concat          [1]

21           -1  1    1969152  ultralytics.nn.modules.block.C2f
[768, 512, 1]
22           [15, 18, 21] 1    2116435
ultralytics.nn.modules.head.Detect        [1, [128, 256, 512]]

```

Model summary: 225 layers, 11,135,987 parameters, 11,135,971 gradients, 28.6 GFLOPs

```

Transferred 349/355 items from pretrained weights
TensorBoard: Start with 'tensorboard --logdir runs/detect/train6',
view at http://localhost:6006/
Freezing layer 'model.22.dfl.conv.weight'
AMP: running Automatic Mixed Precision (AMP) checks with YOLOv8n...
AMP: checks passed []
train: Scanning /content/datasets/waldo7-1/train/labels.cache... 175
images, 0 backgrounds, 0 corrupt: 100% 175/175 [00:00<?, ?it/s]
albumentations: Blur(p=0.01, blur_limit=(3, 7)), MedianBlur(p=0.01,
blur_limit=(3, 7)), ToGray(p=0.01), CLAHE(p=0.01, clip_limit=(1, 4.0),
tile_grid_size=(8, 8))
/usr/lib/python3.10/multiprocessing/popen_fork.py:66: RuntimeWarning:
os.fork() was called. os.fork() is incompatible with multithreaded
code, and JAX is multithreaded, so this will likely lead to a
deadlock.
    self.pid = os.fork()
val: Scanning /content/datasets/waldo7-1/valid/labels.cache... 44

```

```

images, 0 backgrounds, 0 corrupt: 100% 44/44 [00:00<?, ?it/s]
Plotting labels to runs/detect/train6/labels.jpg...
optimizer: SGD(lr=0.01, momentum=0.937) with parameter groups 57
weight(decay=0.0), 64 weight(decay=0.0005), 63 bias(decay=0.0)
TensorBoard: model graph visualization added □
Image sizes 640 train, 640 val
Using 2 dataloader workers
Logging results to runs/detect/train6
Starting training for 80 epochs...

```

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
1/80	4.17G	2.912	25.85	2.166	26
640: 100% 11/11 [00:07<00:00, 1.46it/s]					
	Class	Images	Instances	Box(P)	R
mAP50	mAP50-95): 100%	2/2	[00:01<00:00, 1.27it/s]		
	all	44	45	0.0014	0.0444
0.000701	0.00024				

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
2/80	4.07G	2.959	8.767	2.106	23
640: 100% 11/11 [00:03<00:00, 3.50it/s]					
	Class	Images	Instances	Box(P)	R
mAP50	mAP50-95): 100%	2/2	[00:00<00:00, 2.96it/s]		
	all	44	45	0.00167	0.489
0.00453	0.00134				

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
3/80	4.13G	2.536	4.089	1.95	20
640: 100% 11/11 [00:03<00:00, 3.56it/s]					
	Class	Images	Instances	Box(P)	R
mAP50	mAP50-95): 100%	2/2	[00:00<00:00, 3.45it/s]		
	all	44	45	0.189	0.0889
0.0742	0.03				

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
4/80	4.24G	2.268	3.122	1.648	26
640: 100% 11/11 [00:03<00:00, 2.92it/s]					
	Class	Images	Instances	Box(P)	R
mAP50	mAP50-95): 100%	2/2	[00:01<00:00, 1.82it/s]		
	all	44	45	0.351	0.267
0.198	0.0941				

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
5/80	4.13G	2.027	2.829	1.572	22
640: 100% 11/11 [00:04<00:00, 2.73it/s]					

		Class	Images	Instances	Box(P)	R
mAP50	mAP50-95):	100% all	2/2 [00:00<00:00, 2.97it/s] 44	45	0.457	0.311
0.291	0.15					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	6/80	4.13G	1.813	2.931	1.425	22
640: 100%	11/11	[00:03<00:00, 3.55it/s]				
mAP50	mAP50-95):	100% all	2/2 [00:00<00:00, 3.83it/s] 44	45	0.77	0.4
0.527	0.246					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	7/80	4.24G	1.621	2.154	1.299	23
640: 100%	11/11	[00:03<00:00, 3.58it/s]				
mAP50	mAP50-95):	100% all	2/2 [00:00<00:00, 3.43it/s] 44	45	0.868	0.422
0.557	0.28					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	8/80	4.1G	1.555	1.667	1.242	23
640: 100%	11/11	[00:04<00:00, 2.67it/s]				
mAP50	mAP50-95):	100% all	2/2 [00:01<00:00, 1.64it/s] 44	45	0.678	0.514
0.567	0.265					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	9/80	4.24G	1.628	2.225	1.242	22
640: 100%	11/11	[00:03<00:00, 3.55it/s]				
mAP50	mAP50-95):	100% all	2/2 [00:00<00:00, 3.69it/s] 44	45	0.834	0.489
0.705	0.396					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	10/80	4.1G	1.673	1.556	1.26	18
640: 100%	11/11	[00:03<00:00, 3.66it/s]				
mAP50	mAP50-95):	100% all	2/2 [00:00<00:00, 3.47it/s] 44	45	0.867	0.581
0.678	0.347					

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	11/80	4.09G	1.599	1.825	1.189	24
640:	100% 11/11 [00:03<00:00, 3.36it/s]	Class Images Instances		Box(P)		R
mAP50	mAP50-95): 100% 2/2 [00:00<00:00, 2.32it/s]	all	44	45	0.532	0.557
	0.598	0.322				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	12/80	4.13G	1.44	1.56	1.102	13
640:	100% 11/11 [00:04<00:00, 2.30it/s]	Class Images Instances		Box(P)		R
mAP50	mAP50-95): 100% 2/2 [00:01<00:00, 1.84it/s]	all	44	45	0.787	0.644
	0.703	0.347				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	13/80	4.24G	1.576	1.305	1.159	20
640:	100% 11/11 [00:02<00:00, 3.76it/s]	Class Images Instances		Box(P)		R
mAP50	mAP50-95): 100% 2/2 [00:00<00:00, 3.90it/s]	all	44	45	0.751	0.669
	0.746	0.401				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	14/80	4.1G	1.587	1.443	1.152	18
640:	100% 11/11 [00:02<00:00, 3.76it/s]	Class Images Instances		Box(P)		R
mAP50	mAP50-95): 100% 2/2 [00:00<00:00, 3.84it/s]	all	44	45	0.901	0.812
	0.898	0.395				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	15/80	4.09G	1.509	1.4	1.163	21
640:	100% 11/11 [00:03<00:00, 2.96it/s]	Class Images Instances		Box(P)		R
mAP50	mAP50-95): 100% 2/2 [00:01<00:00, 1.92it/s]	all	44	45	0.842	0.711
	0.85	0.429				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	16/80	4.13G	1.499	1.379	1.141	23
640:	100% 11/11 [00:04<00:00, 2.73it/s]	Class Images Instances		Box(P)		R

mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	2.83it/s]		
		all	44	45	0.858	0.733	
0.807	0.416						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	17/80	4.24G	1.337	1.279	1.09	22	
640:	100%	11/11	[00:02<00:00,	3.77it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	3.74it/s]		
		all	44	45	0.844	0.756	
0.825	0.411						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	18/80	4.09G	1.423	1.114	1.125	27	
640:	100%	11/11	[00:02<00:00,	3.74it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	4.00it/s]		
		all	44	45	0.852	0.867	
0.874	0.441						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	19/80	4.1G	1.427	1.039	1.107	21	
640:	100%	11/11	[00:03<00:00,	2.78it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:01<00:00,	1.80it/s]		
		all	44	45	0.782	0.889	
0.897	0.478						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	20/80	4.13G	1.456	1.154	1.092	23	
640:	100%	11/11	[00:03<00:00,	3.28it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	3.09it/s]		
		all	44	45	0.921	0.822	
0.892	0.458						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	21/80	4.24G	1.431	0.9806	1.095	16	
640:	100%	11/11	[00:02<00:00,	3.68it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	4.61it/s]		
		all	44	45	0.898	0.786	
0.854	0.464						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
22/80	4.1G	1.404	1.033	1.076	22	
640: 100% 11/11 [00:03<00:00, 3.60it/s]						
	Class	Images	Instances	Box(P)	R	
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 2.36it/s]						
	all	44	45	0.88	0.814	
0.89	0.495					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
23/80	4.1G	1.31	1.05	1.046	21	
640: 100% 11/11 [00:04<00:00, 2.48it/s]						
	Class	Images	Instances	Box(P)	R	
mAP50 mAP50-95): 100% 2/2 [00:01<00:00, 1.69it/s]						
	all	44	45	0.804	0.844	
0.904	0.449					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
24/80	4.13G	1.302	0.8991	1.051	14	
640: 100% 11/11 [00:02<00:00, 3.74it/s]						
	Class	Images	Instances	Box(P)	R	
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 4.11it/s]						
	all	44	45	0.804	0.911	
0.904	0.466					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
25/80	4.24G	1.29	0.8722	1.04	24	
640: 100% 11/11 [00:02<00:00, 3.74it/s]						
	Class	Images	Instances	Box(P)	R	
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 4.13it/s]						
	all	44	45	0.832	0.879	
0.919	0.45					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
26/80	4.09G	1.336	0.9296	1.036	19	
640: 100% 11/11 [00:03<00:00, 3.29it/s]						
	Class	Images	Instances	Box(P)	R	
mAP50 mAP50-95): 100% 2/2 [00:01<00:00, 1.94it/s]						
	all	44	45	0.906	0.859	
0.946	0.491					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
27/80	4.1G	1.309	0.8185	1.06	26	
640: 100% 11/11 [00:04<00:00, 2.73it/s]						
	Class	Images	Instances	Box(P)	R	
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 3.51it/s]						

		all	44	45	0.94	0.844
0.935	0.486					
<hr/>						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	28/80	4.13G	1.196	0.7939	1.009	24
640:	100% 11/11 [00:03<00:00, 3.56it/s]					
		Class	Images	Instances	Box(P)	R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 3.45it/s]			
		all	44	45	0.844	0.84
0.896	0.482					
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Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	29/80	4.24G	1.234	0.8059	1.036	20
640:	100% 11/11 [00:03<00:00, 3.63it/s]					
		Class	Images	Instances	Box(P)	R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 3.98it/s]			
		all	44	45	0.918	0.745
0.865	0.458					
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Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	30/80	4.1G	1.237	0.8184	1.014	21
640:	100% 11/11 [00:04<00:00, 2.57it/s]					
		Class	Images	Instances	Box(P)	R
mAP50	mAP50-95):	100%	2/2 [00:01<00:00, 1.74it/s]			
		all	44	45	0.941	0.822
0.828	0.404					
<hr/>						
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	31/80	4.1G	1.179	0.7388	0.9995	18
640:	100% 11/11 [00:03<00:00, 3.34it/s]					
		Class	Images	Instances	Box(P)	R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 3.64it/s]			
		all	44	45	0.874	0.773
0.837	0.433					
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Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	32/80	4.12G	1.118	0.7766	0.9865	18
640:	100% 11/11 [00:02<00:00, 3.75it/s]					
		Class	Images	Instances	Box(P)	R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 3.36it/s]			
		all	44	45	0.924	0.822
0.896	0.426					
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Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances

	33/80	4.08G	1.237	0.8058	1.017	15
640:	100%	11/11	[00:03<00:00, 3.48it/s]			
	Class	Images	Instances	Box(P)		R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 2.24it/s]			
	all	44	45	0.94	0.867	
0.913	0.412					
	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size						
	34/80	4.1G	1.192	0.7815	1.015	19
640:	100%	11/11	[00:04<00:00, 2.46it/s]			
	Class	Images	Instances	Box(P)		R
mAP50	mAP50-95):	100%	2/2 [00:01<00:00, 1.66it/s]			
	all	44	45	0.885	0.778	
0.901	0.469					
	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size						
	35/80	4.1G	1.159	0.7725	0.9947	26
640:	100%	11/11	[00:02<00:00, 3.74it/s]			
	Class	Images	Instances	Box(P)		R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 3.44it/s]			
	all	44	45	0.886	0.867	
0.914	0.469					
	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size						
	36/80	4.13G	1.159	0.7466	0.9983	18
640:	100%	11/11	[00:02<00:00, 3.68it/s]			
	Class	Images	Instances	Box(P)		R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 3.83it/s]			
	all	44	45	0.905	0.933	
0.922	0.472					
	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size						
	37/80	4.08G	1.114	0.7765	0.9809	12
640:	100%	11/11	[00:03<00:00, 3.32it/s]			
	Class	Images	Instances	Box(P)		R
mAP50	mAP50-95):	100%	2/2 [00:01<00:00, 1.89it/s]			
	all	44	45	0.968	0.889	
0.951	0.519					
	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size						
	38/80	4.11G	1.218	0.7692	0.9781	19
640:	100%	11/11	[00:03<00:00, 2.84it/s]			
	Class	Images	Instances	Box(P)		R
mAP50	mAP50-95):	100%	2/2 [00:00<00:00, 3.22it/s]			
	all	44	45	0.959	0.889	

0.937	0.519		
<hr/>			
Epoch	GPU_mem	box_loss	
Size	39/80	4.09G	
	640: 100% 11/11 [00:02<00:00, 3.82it/s]	1.144	
	Class	Images	
mAP50	mAP50-95): 100%	Instances	
	2/2 [00:00<00:00, 3.69it/s]	Box(P	
	all	44	R
		45	0.89
0.937	0.525	0.903	
<hr/>			
Epoch	GPU_mem	box_loss	
Size	40/80	4.13G	
	640: 100% 11/11 [00:03<00:00, 3.61it/s]	1.098	
	Class	Images	
mAP50	mAP50-95): 100%	Instances	
	2/2 [00:00<00:00, 2.59it/s]	Box(P	
	all	44	R
0.913	0.519	0.911	
<hr/>			
Epoch	GPU_mem	box_loss	
Size	41/80	4.23G	
	640: 100% 11/11 [00:04<00:00, 2.51it/s]	1.185	
	Class	Images	
mAP50	mAP50-95): 100%	Instances	
	2/2 [00:01<00:00, 1.77it/s]	Box(P	
	all	44	R
0.915	0.501	0.844	
<hr/>			
Epoch	GPU_mem	box_loss	
Size	42/80	4.1G	
	640: 100% 11/11 [00:03<00:00, 3.64it/s]	1.102	
	Class	Images	
mAP50	mAP50-95): 100%	Instances	
	2/2 [00:00<00:00, 3.71it/s]	Box(P	
	all	44	R
0.906	0.522	0.867	
<hr/>			
Epoch	GPU_mem	box_loss	
Size	43/80	4.09G	
	640: 100% 11/11 [00:02<00:00, 3.73it/s]	1.107	
	Class	Images	
mAP50	mAP50-95): 100%	Instances	
	2/2 [00:00<00:00, 3.95it/s]	Box(P	
	all	44	R
0.908	0.511	0.867	
<hr/>			
Epoch	GPU_mem	box_loss	
Size	44/80	4.12G	
		1.1	
		0.7004	
		0.9608	
		23	

640: 100% 11/11 [00:03<00:00, 3.32it/s]	Class	Images	Instances	Box(P)	R
mAP50 mAP50-95): 100% 2/2 [00:01<00:00, 1.97it/s]	all	44	45	0.944	0.911
0.947 0.5					

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
45/80	4.08G	1.079	0.6819	0.9648	18
640: 100% 11/11 [00:04<00:00, 2.61it/s]	Class	Images	Instances	Box(P)	R
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 2.36it/s]	all	44	45	0.91	0.911
0.939 0.514					

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
46/80	4.1G	1.056	0.6945	0.9464	18
640: 100% 11/11 [00:02<00:00, 3.72it/s]	Class	Images	Instances	Box(P)	R
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 3.85it/s]	all	44	45	0.94	0.911
0.939 0.513					

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
47/80	4.1G	1.007	0.6536	0.9483	20
640: 100% 11/11 [00:02<00:00, 3.78it/s]	Class	Images	Instances	Box(P)	R
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 3.88it/s]	all	44	45	0.911	0.911
0.915 0.542					

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
48/80	4.13G	1.096	0.6803	0.9831	17
640: 100% 11/11 [00:04<00:00, 2.73it/s]	Class	Images	Instances	Box(P)	R
mAP50 mAP50-95): 100% 2/2 [00:01<00:00, 1.72it/s]	all	44	45	0.971	0.911
0.936 0.537					

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
Size					
49/80	4.08G	1.042	0.6381	0.9435	15
640: 100% 11/11 [00:03<00:00, 3.25it/s]	Class	Images	Instances	Box(P)	R
mAP50 mAP50-95): 100% 2/2 [00:00<00:00, 3.22it/s]	all	44	45	0.953	0.911
0.941 0.553					

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	50/80	4.09G	1.065	0.6241	0.9351	25
640:	100% 11/11 [00:03<00:00, 3.59it/s]					
	Class	Images	Instances	Box(P)	R	
mAP50	mAP50-95): 100% 2/2 [00:00<00:00, 4.01it/s]					
	all	44	45	0.95	0.848	
0.937	0.565					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	51/80	4.09G	1.021	0.6433	0.9093	25
640:	100% 11/11 [00:03<00:00, 3.50it/s]					
	Class	Images	Instances	Box(P)	R	
mAP50	mAP50-95): 100% 2/2 [00:00<00:00, 2.41it/s]					
	all	44	45	0.976	0.867	
0.948	0.571					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	52/80	4.13G	1.021	0.6469	0.9203	22
640:	100% 11/11 [00:04<00:00, 2.52it/s]					
	Class	Images	Instances	Box(P)	R	
mAP50	mAP50-95): 100% 2/2 [00:01<00:00, 1.74it/s]					
	all	44	45	0.921	0.889	
0.931	0.56					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	53/80	4.24G	1.111	0.6877	0.9449	23
640:	100% 11/11 [00:02<00:00, 3.78it/s]					
	Class	Images	Instances	Box(P)	R	
mAP50	mAP50-95): 100% 2/2 [00:00<00:00, 4.31it/s]					
	all	44	45	0.928	0.86	
0.926	0.548					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	54/80	4.1G	0.9534	0.599	0.9298	21
640:	100% 11/11 [00:02<00:00, 3.67it/s]					
	Class	Images	Instances	Box(P)	R	
mAP50	mAP50-95): 100% 2/2 [00:00<00:00, 3.57it/s]					
	all	44	45	0.932	0.844	
0.928	0.537					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	55/80	4.1G	0.9681	0.5806	0.9012	20
640:	100% 11/11 [00:03<00:00, 3.37it/s]					

		Class	Images	Instances	Box(P)	R
mAP50	mAP50-95):	100% 2/2 [00:01<00:00, 1.71it/s]				
	all		44	45	0.907	0.844
0.913	0.494					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	56/80	4.12G	0.9951	0.6045	0.9115	26
640:	100% 11/11 [00:04<00:00, 2.52it/s]					
mAP50	mAP50-95):	100% 2/2 [00:00<00:00, 2.95it/s]				
	all		44	45	0.882	0.889
0.928	0.52					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	57/80	4.24G	0.9876	0.5857	0.9076	21
640:	100% 11/11 [00:02<00:00, 3.73it/s]					
mAP50	mAP50-95):	100% 2/2 [00:00<00:00, 3.44it/s]				
	all		44	45	0.921	0.911
0.945	0.553					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	58/80	4.1G	0.9642	0.5778	0.9117	17
640:	100% 11/11 [00:02<00:00, 3.71it/s]					
mAP50	mAP50-95):	100% 2/2 [00:00<00:00, 4.31it/s]				
	all		44	45	0.977	0.927
0.954	0.575					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	59/80	4.1G	0.9266	0.5756	0.9096	20
640:	100% 11/11 [00:04<00:00, 2.64it/s]					
mAP50	mAP50-95):	100% 2/2 [00:01<00:00, 1.74it/s]				
	all		44	45	0.977	0.933
0.953	0.553					
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
	60/80	4.12G	0.879	0.553	0.8916	18
640:	100% 11/11 [00:03<00:00, 3.17it/s]					
mAP50	mAP50-95):	100% 2/2 [00:00<00:00, 2.85it/s]				
	all		44	45	0.954	0.93
0.952	0.549					

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	61/80	4.24G	0.9212	0.5638	0.9012	24
mAP50	100% 11/11 [00:02<00:00, 3.78it/s]	Class Images Instances	Box(P)	R		
mAP50-95):	100% 2/2 [00:00<00:00, 3.82it/s]	all 44 45	0.977	0.933		
	0.951	0.535				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	62/80	4.1G	0.9006	0.529	0.9081	23
mAP50	100% 11/11 [00:03<00:00, 3.63it/s]	Class Images Instances	Box(P)	R		
mAP50-95):	100% 2/2 [00:00<00:00, 2.47it/s]	all 44 45	0.971	0.911		
	0.94	0.555				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	63/80	4.1G	0.8897	0.5531	0.888	21
mAP50	100% 11/11 [00:04<00:00, 2.59it/s]	Class Images Instances	Box(P)	R		
mAP50-95):	100% 2/2 [00:01<00:00, 1.66it/s]	all 44 45	0.948	0.911		
	0.948	0.524				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	64/80	4.12G	0.9292	0.5681	0.8886	26
mAP50	100% 11/11 [00:03<00:00, 3.40it/s]	Class Images Instances	Box(P)	R		
mAP50-95):	100% 2/2 [00:00<00:00, 3.87it/s]	all 44 45	0.952	0.873		
	0.942	0.53				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	65/80	4.24G	0.8459	0.509	0.9082	18
mAP50	100% 11/11 [00:02<00:00, 3.85it/s]	Class Images Instances	Box(P)	R		
mAP50-95):	100% 2/2 [00:00<00:00, 4.24it/s]	all 44 45	0.946	0.889		
	0.944	0.541				
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	66/80	4.1G	0.8918	0.5329	0.8999	18
	Class Images Instances	Box(P)	R			

mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	2.22it/s]		
		all	44	45	0.918	0.889	
0.945	0.551						

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	67/80	4.1G	0.8452	0.4951	0.8876	15	
640:	100%	11/11	[00:04<00:00,	2.44it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:01<00:00,	1.68it/s]		
		all	44	45	0.921	0.911	
0.95	0.572						

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	68/80	4.12G	0.8791	0.5226	0.9073	18	
640:	100%	11/11	[00:03<00:00,	3.55it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	3.66it/s]		
		all	44	45	0.936	0.911	
0.953	0.581						

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	69/80	4.24G	0.8648	0.521	0.8854	22	
640:	100%	11/11	[00:02<00:00,	3.71it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	4.25it/s]		
		all	44	45	0.948	0.911	
0.957	0.564						

Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
	70/80	4.1G	0.8835	0.5103	0.9012	25	
640:	100%	11/11	[00:03<00:00,	3.22it/s]			
		Class	Images	Instances	Box(P	R	
mAP50	mAP50-95):	100%	2/2	[00:01<00:00,	1.76it/s]		
		all	44	45	0.938	0.933	
0.957	0.555						

Closing dataloader mosaic

albumentations: Blur(p=0.01, blur\_limit=(3, 7)), MedianBlur(p=0.01, blur\_limit=(3, 7)), ToGray(p=0.01), CLAHE(p=0.01, clip\_limit=(1, 4.0), tile\_grid\_size=(8, 8))

/usr/lib/python3.10/multiprocessing/popen\_fork.py:66: RuntimeWarning: os.fork() was called. os.fork() is incompatible with multithreaded code, and JAX is multithreaded, so this will likely lead to a deadlock.

self.pid = os.fork()

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	
-------	---------	----------	----------	----------	-----------	--

Size	71/80	4.1G	0.8576	0.5109	0.899	15
640:	100%	11/11	[00:05<00:00,	1.91it/s]		
			Class	Images	Instances	Box(P)
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	3.31it/s]	R
			all	44	45	0.924
0.953		0.557				0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
72/80		4.13G	0.8544	0.5016	0.9008	14
640:	100%	11/11	[00:03<00:00,	3.61it/s]		
			Class	Images	Instances	Box(P)
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	4.25it/s]	R
			all	44	45	0.929
0.95		0.563				0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
73/80		4.24G	0.7731	0.455	0.8855	14
640:	100%	11/11	[00:03<00:00,	3.29it/s]		
			Class	Images	Instances	Box(P)
mAP50	mAP50-95):	100%	2/2	[00:01<00:00,	1.82it/s]	R
			all	44	45	0.941
0.947		0.573				0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
74/80		4.1G	0.7582	0.4819	0.8699	7
640:	100%	11/11	[00:03<00:00,	2.82it/s]		
			Class	Images	Instances	Box(P)
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	3.06it/s]	R
			all	44	45	0.93
0.947		0.552				0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
75/80		4.1G	0.7853	0.4224	0.869	15
640:	100%	11/11	[00:02<00:00,	3.74it/s]		
			Class	Images	Instances	Box(P)
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	4.92it/s]	R
			all	44	45	0.93
0.95		0.564				0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
76/80		4.13G	0.8135	0.4729	0.8924	15
640:	100%	11/11	[00:02<00:00,	3.80it/s]		
			Class	Images	Instances	Box(P)
mAP50	mAP50-95):	100%	2/2	[00:00<00:00,	4.81it/s]	R

0.95	0.543	all	44	45	0.933	0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	77/80	4.24G	0.736	0.4378	0.8495	14
mAP50	100% 11/11	[00:03<00:00, 3.14it/s]	Class	Images	Instances	Box(P)
	mAP50-95):	100% 2/2 [00:01<00:00, 1.97it/s]	all	44	45	R
0.952	0.55					0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	78/80	4.1G	0.7926	0.4547	0.8717	14
mAP50	100% 11/11	[00:03<00:00, 3.22it/s]	Class	Images	Instances	Box(P)
	mAP50-95):	100% 2/2 [00:00<00:00, 3.55it/s]	all	44	45	R
0.953	0.552					0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	79/80	4.09G	0.7255	0.425	0.8587	15
mAP50	100% 11/11	[00:02<00:00, 3.75it/s]	Class	Images	Instances	Box(P)
	mAP50-95):	100% 2/2 [00:00<00:00, 4.05it/s]	all	44	45	R
0.954	0.543					0.933
Size	Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances
640:	80/80	4.13G	0.7431	0.4406	0.8546	15
mAP50	100% 11/11	[00:02<00:00, 3.79it/s]	Class	Images	Instances	Box(P)
	mAP50-95):	100% 2/2 [00:00<00:00, 2.97it/s]	all	44	45	R
0.954	0.552					0.933
80 epochs completed in 0.119 hours.						
Optimizer stripped from runs/detect/train6/weights/last.pt, 22.5MB						
Optimizer stripped from runs/detect/train6/weights/best.pt, 22.5MB						
Validating runs/detect/train6/weights/best.pt...						
Ultralytics YOLOv8.2.61 □ Python-3.10.12 torch-2.3.1+cu121 CUDA:0						
(Tesla T4, 15102MiB)						
Model summary (fused): 168 layers, 11,125,971 parameters, 0 gradients,						
28.4 GFLOPs						
mAP50	100% 2/2	[00:00<00:00, 2.02it/s]	Class	Images	Instances	Box(P)
	mAP50-95):					R

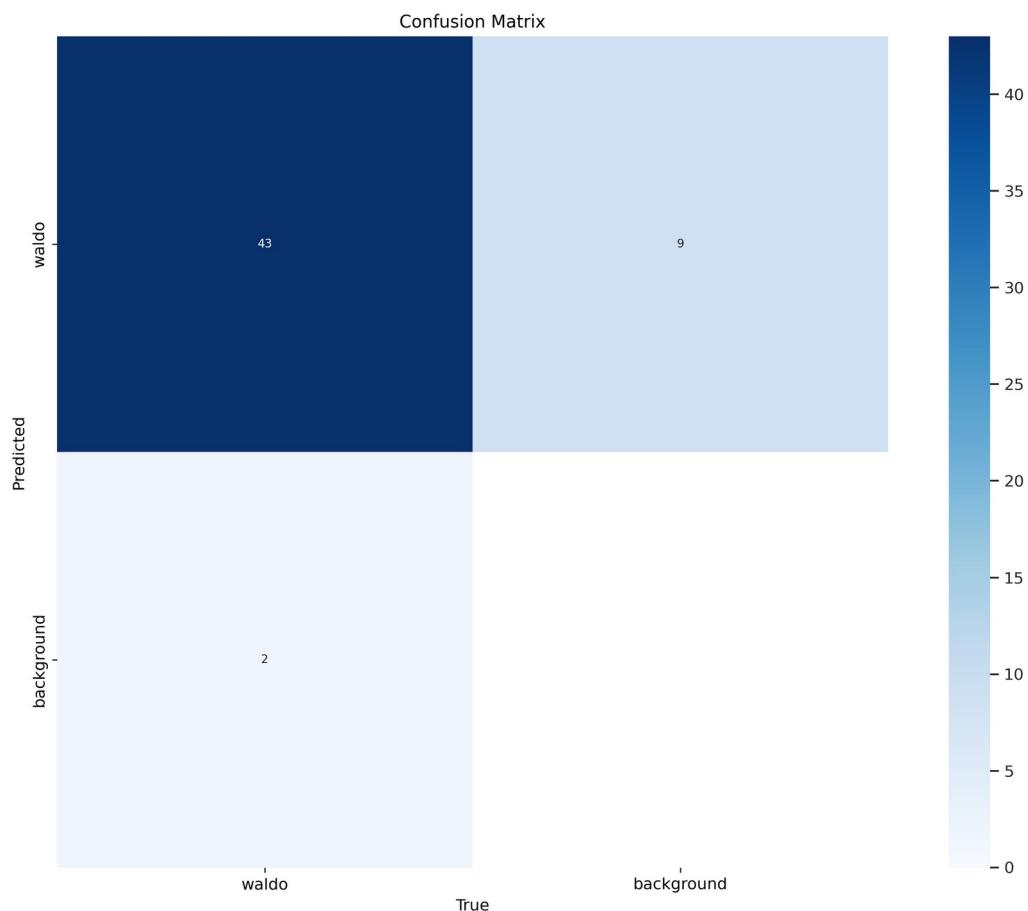
	all	44	45	0.936	0.911
0.953	0.581				
Speed: 0.2ms preprocess, 4.9ms inference, 0.0ms loss, 2.8ms postprocess per image					
Results saved to runs/detect/train6					
Learn more at <a href="https://docs.ultralytics.com/modes/train">https://docs.ultralytics.com/modes/train</a>					

## Display the Confusion Matrix

- The confusion matrix is a performance measurement tool for classification problems, showing how predictions match with actual classes. This visualization helps in evaluating the model's performance by illustrating the counts of true positives, false positives, true negatives, and false negatives.

```
%cd {HOME}
Image(filename=f'{HOME}/runs/detect/train6/confusion_matrix.png',
width=900)
```

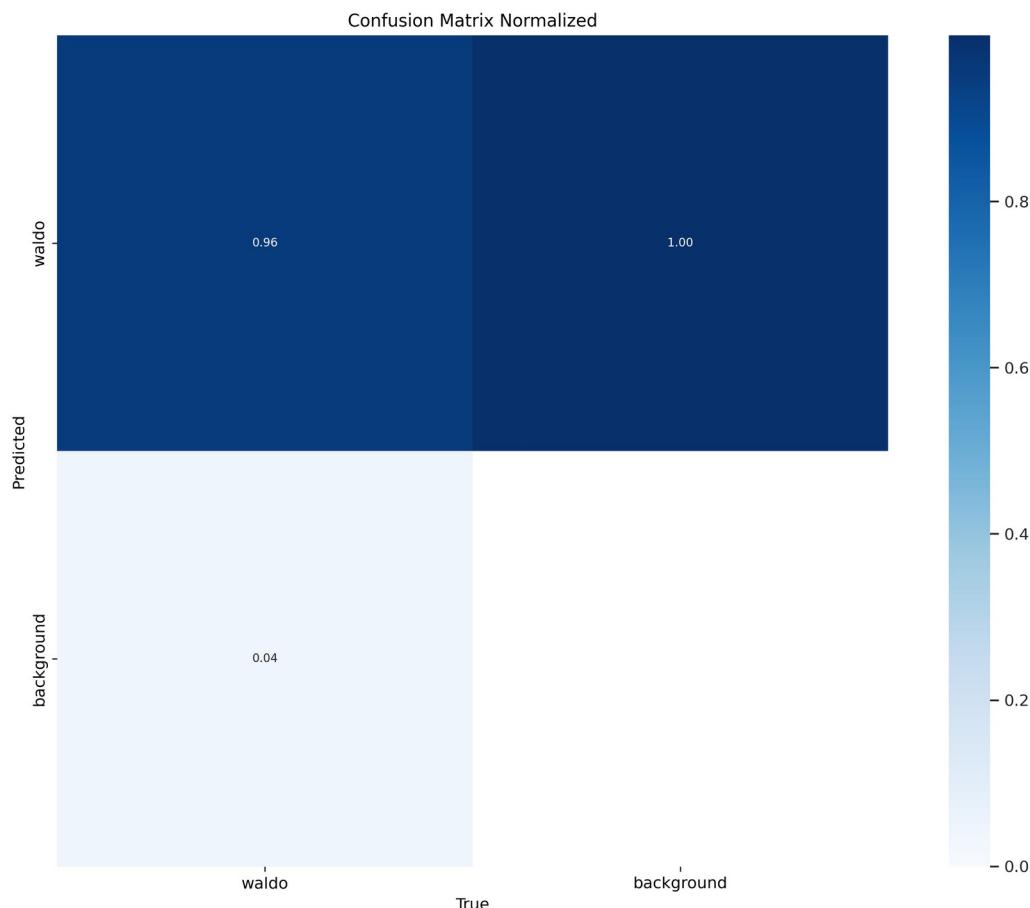
```
/content
```



## Display the Normalized Confusion Matrix

- The normalized confusion matrix provides a clearer view of the model's performance by showing the proportion of correctly and incorrectly classified samples relative to the total number of samples. It helps in evaluating the model's performance across different classes by normalizing the counts to percentages.

```
%cd {HOME}  
Image(filename=f'{HOME}/runs/detect/train6/confusion_matrix_normalized  
.png', width=900)  
  
/content
```

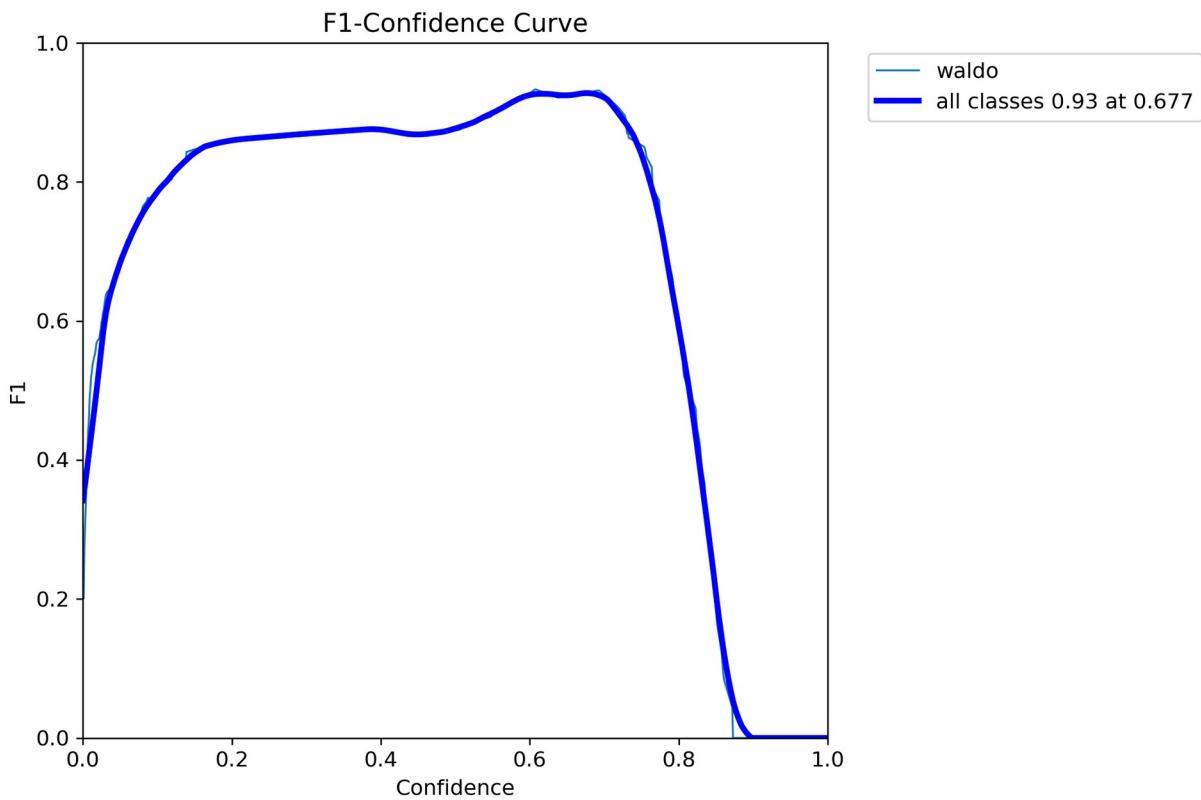


## Display the F1 Score Curve

- The F1 score is a metric that combines precision and recall, providing a single score to assess the model's performance, especially in imbalanced datasets. This curve helps in understanding how well the model balances precision and recall over different operating points.

```
%cd {HOME}  
Image(filename=f'{HOME}/runs/detect/train6/F1_curve.png', width=900)
```

```
/content
```

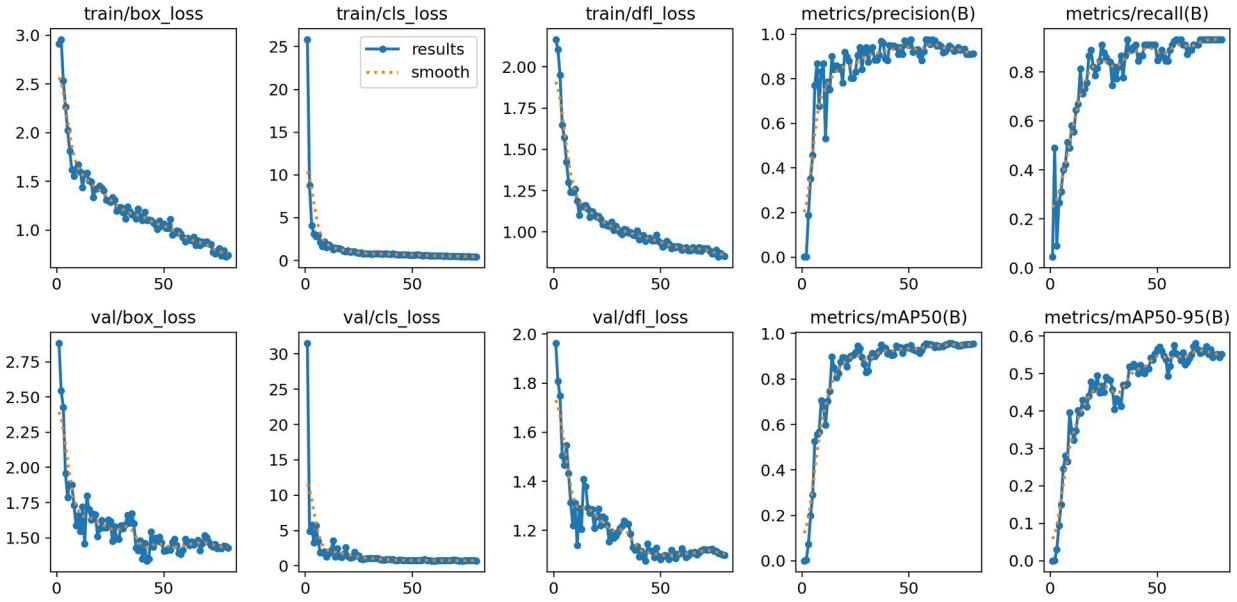


## Display the Results Image

the file typically includes visualizations of the training results, such as detection performance on validation images.

```
%cd {HOME}
Image(filename=f'{HOME}/runs/detect/train6/results.png', width=900)
#The training and validation loss graphs are printed as well as
precision and recall metrics
```

```
/content
```



## Display an Image with Predictions

- The val\_batch0\_pred.jpg file shows an image from the validation set with the model's predictions. Each detected object is annotated with its predicted class and the probability (confidence score) that the object belongs to that class. This visualization helps in evaluating how well the model predicts and classifies objects in the validation images.

```
%cd {HOME}
Image(filename=f'{HOME}/runs/detect/train6/val_batch0_pred.jpg',
width=600)
#Images from the training set is printed with the probability of each.
/content
```



## Run YOLOv8 for Validation

- The command runs the YOLOv8 CLI tool to validate the trained YOLOv8 model using the specified dataset.
- Precision (P): 0.936 - Indicates the proportion of true positive detections among all detections.
- Recall (R): 0.911 - Indicates the proportion of true positive detections among all actual objects.
- mAP50: 0.953 - Mean Average Precision at IoU threshold 0.50, a standard metric for object detection performance.

- mAP50-95: 0.58 - Mean Average Precision averaged over IoU thresholds from 0.50 to 0.95.

```
%cd {HOME}

!yolo task=detect mode=val
model=/content/runs/detect/train6/weights/best.pt
data=/content/datasets/waldo7-1/data.yaml

/content
Ultralytics YOLOv8.2.61 □ Python-3.10.12 torch-2.3.1+cu121 CUDA:0
(Tesla T4, 15102MiB)
Model summary (fused): 168 layers, 11,125,971 parameters, 0 gradients,
28.4 GFLOPs
val: Scanning /content/datasets/waldo7-1/valid/labels.cache... 44
images, 0 backgrounds, 0 corrupt: 100% 44/44 [00:00<?, ?it/s]
          Class      Images   Instances     Box(P)       R
mAP50  mAP50-95): 100% 3/3 [00:02<00:00,  1.21it/s]
                  all        44         45      0.936      0.911
0.953      0.58
Speed: 8.2ms preprocess, 17.0ms inference, 0.0ms loss, 14.7ms
postprocess per image
Results saved to runs/detect/val4
□ Learn more at https://docs.ultralytics.com/modes/val
```

## Run YOLOv8 for Prediction

- The command runs the YOLOv8 CLI tool to perform object detection on a specified image using a pre-trained model

```
%cd {HOME}
!yolo task=detect mode=predict
model={HOME}/runs/detect/train/weights/best.pt conf=0.30
source={HOME}/waldo_1.jpg save=True

/content
Ultralytics YOLOv8.2.61 □ Python-3.10.12 torch-2.3.1+cu121 CUDA:0
(Tesla T4, 15102MiB)
Model summary (fused): 168 layers, 11,125,971 parameters, 0 gradients,
28.4 GFLOPs

image 1/1 /content/waldo_1.jpg: 384x640 1 waldo, 69.8ms
Speed: 1.7ms preprocess, 69.8ms inference, 670.3ms postprocess per
image at shape (1, 3, 384, 640)
Results saved to runs/detect/predict2
□ Learn more at https://docs.ultralytics.com/modes/predict

%cd {HOME}
/content
```

## Yolov5 Training

- This command fetches the latest YOLOv5 code, models, and scripts from the Ultralytics GitHub repository, allowing you to use and modify YOLOv5 for object detection

```
#clone YOL0v5 and
!git clone https://github.com/ultralytics/yolov5 # clone repo
%cd yolov5
%pip install -qr requirements.txt # install dependencies
%pip install -q roboflow

import torch
import os
from IPython.display import Image, clear_output # to display images

print(f"Setup complete. Using torch {torch.__version__}
({torch.cuda.get_device_properties(0).name if
torch.cuda.is_available() else 'CPU'})")

Cloning into 'yolov5'...
remote: Enumerating objects: 16802, done. 0:00:00
remote: Counting objects: 100%
remote: (123/123), done. 0:00:00
remote: Compressing objects: 100% (97/97), done. 0:00:00
remote: Total 16802 (delta 54), reused 61 (delta 26), pack-reused
remote: 16679 207.3/207.3 kB 5.2 MB/s
eta 0:00:00
0:00:00 4.5/4.5 MB 18.6 MB/s eta
0:00:00 64.9/64.9 kB 8.2 MB/s eta
0:00:00 804.2/804.2 kB 28.1 MB/s eta
0:00:00 2.3/2.3 MB 39.0 MB/s eta
0:00:00 62.7/62.7 kB 8.9 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.
ipython 7.34.0 requires jedi>=0.16, which is not installed.
google-colab 1.0.0 requires requests==2.31.0, but you have requests
2.32.3 which is incompatible.
imageio 2.31.6 requires pillow<10.1.0,>=8.3.2, but you have pillow
10.4.0 which is incompatible.
Setup complete. Using torch 2.3.1+cu121 (Tesla T4)
```

## Setting Up Dataset Directory and Installing Roboflow

To organize the datasets, first create a new directory named `datasets` in your home directory. You can do this using the following command:

```
!pip install roboflow

from roboflow import Roboflow
rf = Roboflow(api_key="vaIQ9aYmyRqVHGw89lcc")
project = rf.workspace("bookdetection-hc8pr").project("waldo7")
version = project.version(1)
dataset = version.download("yolov5")

Requirement already satisfied: roboflow in
/usr/local/lib/python3.10/dist-packages (1.1.36)
Requirement already satisfied: certifi in
/usr/local/lib/python3.10/dist-packages (from roboflow) (2024.7.4)
Requirement already satisfied: chardet==4.0.0 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (4.0.0)
Requirement already satisfied: idna==3.7 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (3.7)
Requirement already satisfied: cycler in
/usr/local/lib/python3.10/dist-packages (from roboflow) (0.12.1)
Requirement already satisfied: kiwisolver>=1.3.1 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.4.5)
Requirement already satisfied: matplotlib in
/usr/local/lib/python3.10/dist-packages (from roboflow) (3.7.1)
Requirement already satisfied: numpy>=1.18.5 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.25.2)
Requirement already satisfied: opencv-python-headless==4.10.0.84 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (4.10.0.84)
Requirement already satisfied: Pillow>=7.1.2 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (10.4.0)
Requirement already satisfied: python-dateutil in
/usr/local/lib/python3.10/dist-packages (from roboflow) (2.8.2)
Requirement already satisfied: python-dotenv in
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.0.1)
Requirement already satisfied: requests in
/usr/local/lib/python3.10/dist-packages (from roboflow) (2.32.3)
Requirement already satisfied: six in /usr/local/lib/python3.10/dist-
packages (from roboflow) (1.16.0)
Requirement already satisfied: urllib3>=1.26.6 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (2.0.7)
Requirement already satisfied: tqdm>=4.41.0 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (4.66.4)
Requirement already satisfied: PyYAML>=5.3.1 in
/usr/local/lib/python3.10/dist-packages (from roboflow) (6.0.1)
Requirement already satisfied: requests-toolbelt in
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.0.0)
Requirement already satisfied: filetype in
/usr/local/lib/python3.10/dist-packages (from roboflow) (1.2.0)
Requirement already satisfied: contourpy>=1.0.1 in
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow)
```

```
(1.2.1)
Requirement already satisfied: fonttools>=4.22.0 in
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow)
(4.53.1)
Requirement already satisfied: packaging>=20.0 in
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow)
(24.1)
Requirement already satisfied: pyparsing>=2.3.1 in
/usr/local/lib/python3.10/dist-packages (from matplotlib->roboflow)
(3.1.2)
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.10/dist-packages (from requests->roboflow)
(3.3.2)
loading Roboflow workspace...
loading Roboflow project...

Downloading Dataset Version Zip in waldo7-1 to yolov5pytorch:: 100%|
██████████| 32779/32779 [00:01<00:00, 30092.26it/s]

Extracting Dataset Version Zip to waldo7-1 in yolov5pytorch:: 100%|
██████████| 447/447 [00:00<00:00, 3257.67it/s]
```

## Train YOLOv5 Model

Running the Python Script:

- !python /content/yolov5/train.py: What It Does: Executes the train.py script located in the yolov5 directory. Purpose: This script is responsible for training the YOLOv5 model on a specified dataset.

Setting Image Size --img 640:

- What It Does: Sets the input image size to 640x640 pixels. Purpose: Defines the resolution at which images will be processed during training. Higher resolutions can improve detection accuracy but require more computational resources.

Setting Batch Size --batch 16:

- What It Does: Sets the batch size to 16. Purpose: Determines the number of images processed together in one forward/backward pass through the model. A batch size of 16 balances memory usage and training speed.

Setting Number of Epochs --epochs 80:

- What It Does: Sets the number of training epochs to 80. Purpose: Defines how many times the entire training dataset will be passed through the model. More epochs typically improve the model's performance but increase training time.

Specifying Dataset --data {dataset.location}/data.yaml:

- What It Does: Points to the data.yaml file of the dataset. Purpose: This file contains information about the dataset, including class names and paths to training and validation images. {dataset.location} is a placeholder for the path where the dataset is downloaded.
- Setting Pre-trained Weights

--weights yolov5s.pt:

- What It Does: Specifies the pre-trained weights file yolov5s.pt. Purpose: Initializes the model with pre-trained weights, which can speed up training and improve performance by starting from a model that has already learned useful features.

Setting Optimizer --optimizer SGD:

- What It Does: Chooses Stochastic Gradient Descent (SGD) as the optimization algorithm. Purpose: Controls how the model weights are updated during training. SGD is a commonly used optimizer that can perform well for many types of neural network models.
- Enabling Cache

--cache:

- What It Does: Enables caching of images in memory. Purpose: Speeds up training by reducing the need to load images from disk repeatedly. This can significantly improve training time, especially for large datasets.

```
!python /content/yolov5/train.py --img 640 --batch 16 --epochs 80 --
data {dataset.location}/data.yaml --weights yolov5s.pt --optimizer
SGD --cache
```

```
2024-07-22 01:34:59.950305: E
external/local_xla/xla/stream_executor/cuda/cuda_dnn.cc:9261] Unable
to register cuDNN factory: Attempting to register factory for plugin
cuDNN when one has already been registered
2024-07-22 01:34:59.950376: E
external/local_xla/xla/stream_executor/cuda/cuda_fft.cc:607] Unable to
register cuFFT factory: Attempting to register factory for plugin
cuFFT when one has already been registered
2024-07-22 01:34:59.953043: E
external/local_xla/xla/stream_executor/cuda/cuda_blas.cc:1515] Unable
to register cuBLAS factory: Attempting to register factory for plugin
cuBLAS when one has already been registered
train: weights=yolov5s.pt, cfg=,
data=/content/yolov5/waldo7-1/data.yaml,
hyp=yolov5/data/hyps/hyp.scratch-low.yaml, epochs=80, batch_size=16,
imgsz=640, rect=False, resume=False, nosave=False, noval=False,
noautoanchor=False, noplots=False, evolve=None,
evolve_population=yolov5/data/hyps, resume_evolve=None, bucket=,
cache=ram, image_weights=False, device=, multi_scale=False,
single_cls=False, optimizer=SGD, sync_bn=False, workers=8,
project=yolov5/runs/train, name=exp, exist_ok=False, quad=False,
cos_lr=False, label_smoothing=0.0, patience=100, freeze=[0],
save_period=-1, seed=0, local_rank=-1, entity=None,
upload_dataset=False, bbox_interval=-1, artifact_alias=latest,
```

```

ndjson_console=False, ndjson_file=False
remote: Enumerating objects: 5, done.ote: Counting objects: 100%
(5/5), done.ote: Compressing objects: 100% (5/5), done.ote: Total 5
(delta 0), reused 0 (delta 0), pack-reused 0
https://github.com/ultralytics/yolov5
 * [new branch]          glenn-jocher-patch-1 -> origin/glen-jocher-
patch-1
github: up to date with https://github.com/ultralytics/yolov5 □
YOL0v5 □ v7.0-345-g8003649c Python-3.10.12 torch-2.3.1+cu121 CUDA:0
(Tesla T4, 15102MiB)

hyperparameters: lr0=0.01, lrf=0.01, momentum=0.937,
weight_decay=0.0005, warmup_epochs=3.0, warmup_momentum=0.8,
warmup_bias_lr=0.1, box=0.05, cls=0.5, cls_pw=1.0, obj=1.0,
obj_pw=1.0, iou_t=0.2, anchor_t=4.0, fl_gamma=0.0, hsv_h=0.015,
hsv_s=0.7, hsv_v=0.4, degrees=0.0, translate=0.1, scale=0.5,
shear=0.0, perspective=0.0, flipud=0.0, fliplr=0.5, mosaic=1.0,
mixup=0.0, copy_paste=0.0
Comet: run 'pip install comet_ml' to automatically track and visualize
YOL0v5 □ runs in Comet
TensorBoard: Start with 'tensorboard --logdir yolov5/runs/train', view
at http://localhost:6006/
Downloading
https://github.com/ultralytics/yolov5/releases/download/v7.0/yolov5s.p
t to yolov5s.pt...
100% 14.1M/14.1M [00:00<00:00, 209MB/s]

```

Overriding model.yaml nc=80 with nc=1

	from	n	params	module
arguments				
0		-1	1	3520 models.common.Conv
[3, 32, 6, 2, 2]				
1		-1	1	18560 models.common.Conv
[32, 64, 3, 2]				
2		-1	1	18816 models.common.C3
[64, 64, 1]				
3		-1	1	73984 models.common.Conv
[64, 128, 3, 2]				
4		-1	2	115712 models.common.C3
[128, 128, 2]				
5		-1	1	295424 models.common.Conv
[128, 256, 3, 2]				
6		-1	3	625152 models.common.C3
[256, 256, 3]				
7		-1	1	1180672 models.common.Conv
[256, 512, 3, 2]				
8		-1	1	1182720 models.common.C3
[512, 512, 1]				
9		-1	1	656896 models.common.SPPF

```

[512, 512, 5]
10           -1 1    131584  models.common.Conv
[512, 256, 1, 1]
11           -1 1      0
torch.nn.modules.upsampling.Upsample      [None, 2, 'nearest']
12           [-1, 6] 1      0  models.common.Concat
[1]
13           -1 1    361984  models.common.C3
[512, 256, 1, False]
14           -1 1    33024  models.common.Conv
[256, 128, 1, 1]
15           -1 1      0
torch.nn.modules.upsampling.Upsample      [None, 2, 'nearest']
16           [-1, 4] 1      0  models.common.Concat
[1]
17           -1 1    90880  models.common.C3
[256, 128, 1, False]
18           -1 1    147712  models.common.Conv
[128, 128, 3, 2]
19           [-1, 14] 1      0  models.common.Concat
[1]
20           -1 1    296448  models.common.C3
[256, 256, 1, False]
21           -1 1    590336  models.common.Conv
[256, 256, 3, 2]
22           [-1, 10] 1      0  models.common.Concat
[1]
23           -1 1    1182720  models.common.C3
[512, 512, 1, False]
24           [17, 20, 23] 1    16182  models.yolo.Detect
[1, [[10, 13, 16, 30, 33, 23], [30, 61, 62, 45, 59, 119], [116, 90,
156, 198, 373, 326]], [128, 256, 512]]
Model summary: 214 layers, 7022326 parameters, 7022326 gradients, 15.9
GFLOPs

```

```

Transferred 343/349 items from yolov5s.pt
AMP: checks passed ✅
optimizer: SGD(lr=0.01) with parameter groups 57 weight(decay=0.0), 60
weight(decay=0.0005), 60 bias
albumentations: Blur(p=0.01, blur_limit=(3, 7)), MedianBlur(p=0.01,
blur_limit=(3, 7)), ToGray(p=0.01), CLAHE(p=0.01, clip_limit=(1, 4.0),
tile_grid_size=(8, 8))
train: Scanning /content/yolov5/waldo7-1/train/labels.cache... 175
images, 0 backgrounds, 0 corrupt: 100% 175/175 [00:00<?, ?it/s]
train: Caching images (0.2GB ram): 100% 175/175 [00:01<00:00,
157.45it/s]
/usr/lib/python3.10/multiprocessing/popen_fork.py:66: RuntimeWarning:
os.fork() was called. os.fork() is incompatible with multithreaded
code, and JAX is multithreaded, so this will likely lead to a

```

```
deadlock.  
    self.pid = os.fork()  
val: Scanning /content/yolov5/waldo7-1/valid/labels.cache... 44  
images, 0 backgrounds, 0 corrupt: 100% 44/44 [00:00<?, ?it/s]  
val: Caching images (0.1GB ram): 100% 44/44 [00:00<00:00, 57.68it/s]
```

```
AutoAnchor: 5.86 anchors/target, 1.000 Best Possible Recall (BPR).  
Current anchors are a good fit to dataset □  
Plotting labels to yolov5/runs/train/exp6/labels.jpg...  
Image sizes 640 train, 640 val  
Using 2 dataloader workers  
Logging results to yolov5/runs/train/exp6  
Starting training for 80 epochs...
```

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640	0/79	3.66G	0.1205	0.02779	0	17
	1/79	4.48G	0.1138	0.02277	0	13
	2/79	4.48G	0.1118	0.01902	0	27
	3/79	4.48G	0.1045	0.01752	0	19
mAP50	mAP50-95:	100%	2/2	[00:02<00:00, 1.07s/it]	P	R
		all	44	45	0	0
0	0	0	0	0	0	0

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640	0/79	3.66G	0.1205	0.02779	0	17
	1/79	4.48G	0.1138	0.02277	0	13
	2/79	4.48G	0.1118	0.01902	0	27
	3/79	4.48G	0.1045	0.01752	0	19
mAP50	mAP50-95:	100%	2/2	[00:02<00:00, 1.07s/it]	P	R
		all	44	45	0.000227	0.0667
0.000135	1.35e-05	0.000135	1.35e-05	0.000135	1.35e-05	0.000135

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640	0/79	3.66G	0.1205	0.02779	0	17
	1/79	4.48G	0.1138	0.02277	0	13
	2/79	4.48G	0.1118	0.01902	0	27
	3/79	4.48G	0.1045	0.01752	0	19
mAP50	mAP50-95:	100%	2/2	[00:01<00:00, 1.92it/s]	P	R
		all	44	45	0.000227	0.0667
0.000126	3.3e-05	0.000126	3.3e-05	0.000126	3.3e-05	0.000126

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640	0/79	3.66G	0.1205	0.02779	0	17
	1/79	4.48G	0.1138	0.02277	0	13
	2/79	4.48G	0.1118	0.01902	0	27
	3/79	4.48G	0.1045	0.01752	0	19
mAP50	mAP50-95:	100%	2/2	[00:01<00:00, 1.82it/s]	P	R
		all	44	45	7.58e-05	0.0222
3.95e-05	2.37e-05	3.95e-05	2.37e-05	3.95e-05	2.37e-05	3.95e-05

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	4/79	4.48G	0.09953	0.01778	0	21
640:	100% 11/11 [00:02<00:00, 4.30it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 3.69it/s]				
		all	44	45	0.000455	0.133
0.000282			7.29e-05			
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	5/79	4.48G	0.09859	0.01605	0	18
640:	100% 11/11 [00:02<00:00, 4.14it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 3.84it/s]				
		all	44	45	0.000606	0.178
0.000389			7.39e-05			
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	6/79	4.48G	0.09478	0.01691	0	22
640:	100% 11/11 [00:02<00:00, 4.35it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 3.12it/s]				
		all	44	45	0.000606	0.178
0.000448			0.000196			
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	7/79	4.48G	0.09388	0.01731	0	23
640:	100% 11/11 [00:03<00:00, 2.85it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 2.35it/s]				
		all	44	45	0.000909	0.267
0.000794			0.000269			
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	8/79	4.48G	0.08823	0.01745	0	27
640:	100% 11/11 [00:02<00:00, 3.68it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 4.34it/s]				
		all	44	45	0.000985	0.289
0.00108			0.000294			
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	9/79	4.48G	0.08392	0.01651	0	25
640:	100% 11/11 [00:02<00:00, 4.34it/s]					
		Class	Images	Instances	P	R

mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.39it/s]		
		all		44	45	0.00121	0.356
0.00337	0.00085						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	10/79	4.48G	0.08101	0.01759		0	24
640:	100%	11/11	[00:02<00:00,	4.41it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	3.90it/s]		
		all		44	45	0.00209	0.6
0.00746	0.00227						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	11/79	4.48G	0.07687	0.01868		0	24
640:	100%	11/11	[00:03<00:00,	3.05it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	2.32it/s]		
		all		44	45	0.00275	0.778
0.0625	0.0152						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	12/79	4.48G	0.07223	0.01723		0	24
640:	100%	11/11	[00:03<00:00,	3.05it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:01<00:00,	1.68it/s]		
		all		44	45	0.00313	0.889
0.082	0.0169						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	13/79	4.48G	0.06836	0.01745		0	18
640:	100%	11/11	[00:02<00:00,	4.33it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.07it/s]		
		all		44	45	0.0887	0.222
0.0752	0.0145						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	14/79	4.48G	0.0658	0.01658		0	23
640:	100%	11/11	[00:02<00:00,	4.26it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.05it/s]		
		all		44	45	0.00624	0.867
0.172	0.0573						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
15/79	4.48G	0.06515	0.01781	0	18	
640:	100% 11/11 [00:02<00:00, 4.19it/s]					
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 3.24it/s]				
	all	44	45	0.00313	0.911	
0.153	0.0463					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
16/79	4.48G	0.06431	0.018	0	21	
640:	100% 11/11 [00:03<00:00, 3.10it/s]					
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:01<00:00, 1.70it/s]				
	all	44	45	0.00327	0.956	
0.187	0.057					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
17/79	4.48G	0.05875	0.01713	0	20	
640:	100% 11/11 [00:02<00:00, 3.79it/s]					
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 4.05it/s]				
	all	44	45	0.0119	0.867	
0.323	0.102					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
18/79	4.48G	0.05674	0.01602	0	22	
640:	100% 11/11 [00:02<00:00, 4.37it/s]					
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 4.19it/s]				
	all	44	45	0.648	0.245	
0.33	0.0934					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
19/79	4.48G	0.05681	0.017	0	18	
640:	100% 11/11 [00:02<00:00, 4.07it/s]					
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 3.76it/s]				
	all	44	45	0.629	0.226	
0.295	0.104					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
20/79	4.48G	0.05619	0.01614	0	22	
640:	100% 11/11 [00:03<00:00, 3.20it/s]					
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 2.05it/s]				

		all	44	45	0.583	0.178
0.249	0.103					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	21/79	4.48G	0.05291	0.01608	0	18
640:	100%	11/11	[00:03<00:00,	2.97it/s]		
			Class	Images	Instances	P R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00,	2.53it/s]		
			all	44	45	0.766 0.378
0.471	0.173					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	22/79	4.48G	0.04995	0.01632	0	21
640:	100%	11/11	[00:02<00:00,	4.24it/s]		
			Class	Images	Instances	P R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00,	3.63it/s]		
			all	44	45	0.413 0.222
0.21	0.0789					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	23/79	4.48G	0.05237	0.01637	0	20
640:	100%	11/11	[00:02<00:00,	4.11it/s]		
			Class	Images	Instances	P R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00,	3.92it/s]		
			all	44	45	0.776 0.333
0.446	0.171					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	24/79	4.48G	0.04981	0.01428	0	21
640:	100%	11/11	[00:02<00:00,	4.02it/s]		
			Class	Images	Instances	P R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00,	2.10it/s]		
			all	44	45	0.776 0.378
0.46	0.113					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	25/79	4.48G	0.05429	0.0129	0	18
640:	100%	11/11	[00:03<00:00,	2.97it/s]		
			Class	Images	Instances	P R
mAP50	mAP50-95:	100%	2/2 [00:01<00:00,	1.51it/s]		
			all	44	45	0.817 0.422
0.588	0.228					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances

	26/79	4.48G	0.0521	0.01432	0	23
640:	100%	11/11	[00:02<00:00,	3.98it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.05it/s]	
		all	44	45	0.769	0.4
0.462	0.171					
	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	27/79	4.48G	0.04834	0.01394	0	20
640:	100%	11/11	[00:02<00:00,	4.36it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.18it/s]	
		all	44	45	0.781	0.477
0.594	0.241					
	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	28/79	4.48G	0.04969	0.01395	0	16
640:	100%	11/11	[00:02<00:00,	4.26it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.06it/s]	
		all	44	45	0.842	0.511
0.626	0.23					
	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	29/79	4.48G	0.04956	0.01307	0	17
640:	100%	11/11	[00:03<00:00,	3.12it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	2.28it/s]	
		all	44	45	0.841	0.533
0.635	0.23					
	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	30/79	4.48G	0.04746	0.01246	0	20
640:	100%	11/11	[00:03<00:00,	3.10it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.15it/s]	
		all	44	45	0.922	0.524
0.663	0.251					
	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	31/79	4.48G	0.04727	0.01312	0	27
640:	100%	11/11	[00:02<00:00,	4.24it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.02it/s]	
		all	44	45	0.846	0.6

0.702	0.307						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	32/79	4.48G	0.04543	0.01272	0	26	
640:	100%	11/11	[00:02<00:00,	4.24it/s]			
		Class	Images	Instances	P	R	
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.19it/s]		
		all	44	45	0.825	0.63	
0.738	0.313						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	33/79	4.48G	0.04645	0.01218	0	20	
640:	100%	11/11	[00:03<00:00,	3.60it/s]			
		Class	Images	Instances	P	R	
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	2.33it/s]		
		all	44	45	0.82	0.622	
0.747	0.316						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	34/79	4.48G	0.04347	0.01137	0	18	
640:	100%	11/11	[00:03<00:00,	2.88it/s]			
		Class	Images	Instances	P	R	
mAP50	mAP50-95:	100%	2/2	[00:01<00:00,	1.83it/s]		
		all	44	45	0.749	0.664	
0.722	0.265						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	35/79	4.48G	0.04573	0.01151	0	21	
640:	100%	11/11	[00:02<00:00,	4.06it/s]			
		Class	Images	Instances	P	R	
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.11it/s]		
		all	44	45	0.721	0.622	
0.698	0.278						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	36/79	4.48G	0.04047	0.01184	0	13	
640:	100%	11/11	[00:02<00:00,	4.38it/s]			
		Class	Images	Instances	P	R	
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.21it/s]		
		all	44	45	0.704	0.644	
0.725	0.288						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	37/79	4.48G	0.04523	0.01111	0	15	

640: 100% 11/11 [00:02<00:00, 4.28it/s]	Class	Images	Instances	P	R
mAP50 mAP50-95: 100% 2/2 [00:00<00:00, 4.03it/s]	all	44	45	0.743	0.733
0.807 0.309					

Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size					
38/79	4.48G	0.0425	0.01205	0	19
640: 100% 11/11 [00:03<00:00, 3.15it/s]	Class	Images	Instances	P	R
mAP50 mAP50-95: 100% 2/2 [00:00<00:00, 2.42it/s]	all	44	45	0.751	0.667
0.75 0.343					

Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size					
39/79	4.48G	0.04213	0.009556	0	22
640: 100% 11/11 [00:03<00:00, 2.83it/s]	Class	Images	Instances	P	R
mAP50 mAP50-95: 100% 2/2 [00:00<00:00, 2.28it/s]	all	44	45	0.828	0.667
0.801 0.371					

Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size					
40/79	4.48G	0.04535	0.01037	0	27
640: 100% 11/11 [00:02<00:00, 4.34it/s]	Class	Images	Instances	P	R
mAP50 mAP50-95: 100% 2/2 [00:00<00:00, 3.64it/s]	all	44	45	0.888	0.707
0.823 0.366					

Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size					
41/79	4.48G	0.0404	0.01058	0	19
640: 100% 11/11 [00:02<00:00, 4.32it/s]	Class	Images	Instances	P	R
mAP50 mAP50-95: 100% 2/2 [00:00<00:00, 4.04it/s]	all	44	45	0.886	0.756
0.803 0.373					

Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size					
42/79	4.48G	0.03933	0.009997	0	24
640: 100% 11/11 [00:02<00:00, 3.95it/s]	Class	Images	Instances	P	R
mAP50 mAP50-95: 100% 2/2 [00:00<00:00, 2.87it/s]	all	44	45	0.847	0.741
0.811 0.378					

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	43/79	4.48G	0.04022	0.01001	0	16
	100% 11/11 [00:03<00:00, 2.88it/s]					
	Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2 [00:01<00:00, 1.87it/s]			
	all	44	45	0.823	0.733	
0.812	0.388					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	44/79	4.48G	0.03991	0.009386	0	18
	100% 11/11 [00:03<00:00, 3.43it/s]					
	Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 3.85it/s]			
	all	44	45	0.859	0.733	
0.825	0.385					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	45/79	4.48G	0.03839	0.01012	0	23
	100% 11/11 [00:02<00:00, 4.29it/s]					
	Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 4.01it/s]			
	all	44	45	0.886	0.733	
0.83	0.414					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	46/79	4.48G	0.03795	0.009762	0	21
	100% 11/11 [00:02<00:00, 4.23it/s]					
	Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 4.02it/s]			
	all	44	45	0.878	0.711	
0.847	0.395					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	47/79	4.48G	0.03833	0.009807	0	20
	100% 11/11 [00:03<00:00, 3.28it/s]					
	Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 2.12it/s]			
	all	44	45	0.894	0.711	
0.83	0.358					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	48/79	4.48G	0.03739	0.009665	0	24
	100% 11/11 [00:03<00:00, 2.90it/s]					

		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% all	2/2 [00:01<00:00, 1.83it/s] 44	45	0.907	0.711
0.823	0.349					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	49/79	4.48G	0.03846	0.009097	0	18
	100% 11/11	[00:02<00:00, 3.92it/s]				
mAP50	mAP50-95:	100% all	2/2 [00:00<00:00, 4.09it/s] 44	45	0.875	0.689
0.791	0.369					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	50/79	4.48G	0.0377	0.009421	0	17
	100% 11/11	[00:02<00:00, 4.26it/s]				
mAP50	mAP50-95:	100% all	2/2 [00:00<00:00, 4.10it/s] 44	45	0.851	0.759
0.836	0.44					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	51/79	4.48G	0.03658	0.009181	0	21
	100% 11/11	[00:02<00:00, 4.34it/s]				
mAP50	mAP50-95:	100% all	2/2 [00:00<00:00, 3.94it/s] 44	45	0.857	0.798
0.852	0.444					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	52/79	4.48G	0.03852	0.008489	0	19
	100% 11/11	[00:03<00:00, 3.08it/s]				
mAP50	mAP50-95:	100% all	2/2 [00:00<00:00, 2.15it/s] 44	45	0.896	0.766
0.854	0.411					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
640:	53/79	4.48G	0.03616	0.009133	0	19
	100% 11/11	[00:03<00:00, 2.94it/s]				
mAP50	mAP50-95:	100% all	2/2 [00:00<00:00, 3.49it/s] 44	45	0.871	0.778
0.856	0.419					

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	54/79	4.48G	0.03413	0.00892	0	13
640:	100% 11/11 [00:02<00:00, 4.34it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 4.08it/s]				
		all	44	45	0.887	0.756
0.864	0.434					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	55/79	4.48G	0.03776	0.008997	0	24
640:	100% 11/11 [00:02<00:00, 4.19it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 4.09it/s]				
		all	44	45	0.85	0.733
0.84	0.41					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	56/79	4.48G	0.03301	0.007689	0	19
640:	100% 11/11 [00:02<00:00, 4.13it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 2.18it/s]				
		all	44	45	0.869	0.733
0.867	0.446					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	57/79	4.48G	0.0351	0.009232	0	19
640:	100% 11/11 [00:03<00:00, 2.92it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:01<00:00, 1.87it/s]				
		all	44	45	0.874	0.778
0.874	0.434					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	58/79	4.48G	0.03736	0.009578	0	21
640:	100% 11/11 [00:02<00:00, 4.27it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 4.09it/s]				
		all	44	45	0.878	0.798
0.874	0.428					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	59/79	4.48G	0.03479	0.008936	0	18
640:	100% 11/11 [00:02<00:00, 4.30it/s]					
		Class	Images	Instances	P	R

mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.26it/s]		
		all		44	45	0.897	0.774
0.884	0.445						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	60/79	4.48G	0.03371	0.008347		0	24
640:	100%	11/11	[00:02<00:00,	4.38it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	3.04it/s]		
		all		44	45	0.842	0.826
0.881	0.43						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	61/79	4.48G	0.03239	0.009155		0	21
640:	100%	11/11	[00:03<00:00,	3.08it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:01<00:00,	1.71it/s]		
		all		44	45	0.88	0.814
0.883	0.456						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	62/79	4.48G	0.03071	0.008251		0	25
640:	100%	11/11	[00:03<00:00,	3.20it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.15it/s]		
		all		44	45	0.878	0.778
0.887	0.45						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	63/79	4.48G	0.03339	0.008987		0	24
640:	100%	11/11	[00:02<00:00,	4.34it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.19it/s]		
		all		44	45	0.856	0.791
0.886	0.446						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	
	64/79	4.48G	0.03114	0.007717		0	29
640:	100%	11/11	[00:02<00:00,	4.33it/s]			
		Class	Images	Instances		P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.13it/s]		
		all		44	45	0.931	0.733
0.885	0.457						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	

Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
65/79	4.48G	0.03384	0.007615	0	19	
640: 100% 11/11 [00:03<00:00, 3.61it/s]						
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:01<00:00, 1.85it/s]				
	all	44	45	0.93	0.733	
0.877	0.438					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
66/79	4.48G	0.03032	0.008065	0	15	
640: 100% 11/11 [00:04<00:00, 2.74it/s]						
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 2.24it/s]				
	all	44	45	0.915	0.711	
0.873	0.448					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
67/79	4.48G	0.0313	0.008261	0	24	
640: 100% 11/11 [00:02<00:00, 4.33it/s]						
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 4.09it/s]				
	all	44	45	0.92	0.711	
0.879	0.47					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
68/79	4.48G	0.02909	0.007898	0	23	
640: 100% 11/11 [00:02<00:00, 4.27it/s]						
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 3.92it/s]				
	all	44	45	0.919	0.711	
0.877	0.46					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
69/79	4.48G	0.02935	0.007453	0	16	
640: 100% 11/11 [00:02<00:00, 4.06it/s]						
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:00<00:00, 2.67it/s]				
	all	44	45	0.829	0.8	
0.879	0.472					
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
70/79	4.48G	0.02966	0.008425	0	19	
640: 100% 11/11 [00:03<00:00, 2.92it/s]						
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100% 2/2 [00:01<00:00, 1.74it/s]				

		all	44	45	0.942	0.756
0.883	0.498					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	71/79	4.48G	0.02827	0.008406	0	26
640:	100% 11/11 [00:02<00:00, 4.22it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 3.94it/s]			
		all	44	45	0.937	0.756
0.885	0.459					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	72/79	4.48G	0.02922	0.007721	0	23
640:	100% 11/11 [00:02<00:00, 4.30it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 4.04it/s]			
		all	44	45	0.931	0.756
0.888	0.476					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	73/79	4.48G	0.02967	0.008078	0	20
640:	100% 11/11 [00:02<00:00, 4.24it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 4.08it/s]			
		all	44	45	0.938	0.756
0.889	0.491					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	74/79	4.48G	0.02843	0.007629	0	19
640:	100% 11/11 [00:03<00:00, 3.01it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 2.02it/s]			
		all	44	45	0.934	0.756
0.893	0.487					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
	75/79	4.48G	0.02924	0.008114	0	26
640:	100% 11/11 [00:03<00:00, 2.93it/s]					
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2 [00:00<00:00, 2.15it/s]			
		all	44	45	0.935	0.756
0.894	0.47					
<hr/>						
Size	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances

	76/79	4.48G	0.02837	0.007847	0	29
640:	100%	11/11	[00:02<00:00,	4.33it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	3.95it/s]	
		all	44	45	0.938	0.756
0.894	0.481					

	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	77/79	4.48G	0.02834	0.007133	0	20
640:	100%	11/11	[00:02<00:00,	4.24it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	4.00it/s]	
		all	44	45	0.944	0.751
0.897	0.487					

	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	78/79	4.48G	0.02916	0.007699	0	26
640:	100%	11/11	[00:02<00:00,	4.28it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	2.95it/s]	
		all	44	45	0.907	0.756
0.887	0.462					

	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
Size						
	79/79	4.48G	0.02974	0.007398	0	15
640:	100%	11/11	[00:03<00:00,	3.01it/s]		
		Class	Images	Instances	P	R
mAP50	mAP50-95:	100%	2/2	[00:00<00:00,	2.11it/s]	
		all	44	45	0.902	0.756
0.888	0.467					

80 epochs completed in 0.097 hours.

Optimizer stripped from yolov5/runs/train/exp6/weights/last.pt, 14.4MB  
 Optimizer stripped from yolov5/runs/train/exp6/weights/best.pt, 14.4MB

Validating yolov5/runs/train/exp6/weights/best.pt...

Fusing layers...

Model summary: 157 layers, 7012822 parameters, 0 gradients, 15.8 GFLOPs

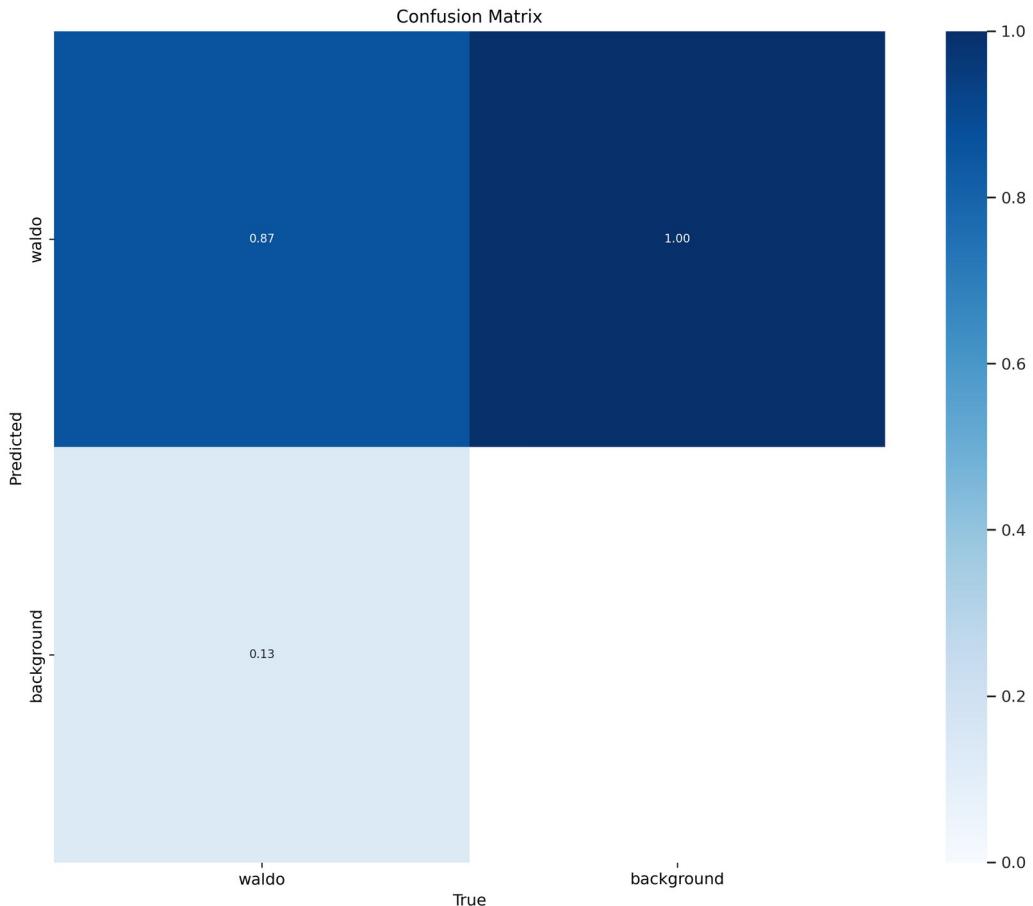
	Class	Images	Instances	P	R	
mAP50	mAP50-95:	100%	2/2	[00:01<00:00,	1.79it/s]	
		all	44	45	0.94	0.756
0.883	0.498					

Results saved to yolov5/runs/train/exp6

## Display the Confusion Matrix

- The confusion matrix is a performance measurement tool for classification problems, showing how predictions match with actual classes. This visualization helps in evaluating the model's performance by illustrating the counts of true positives, false positives, true negatives, and false negatives.

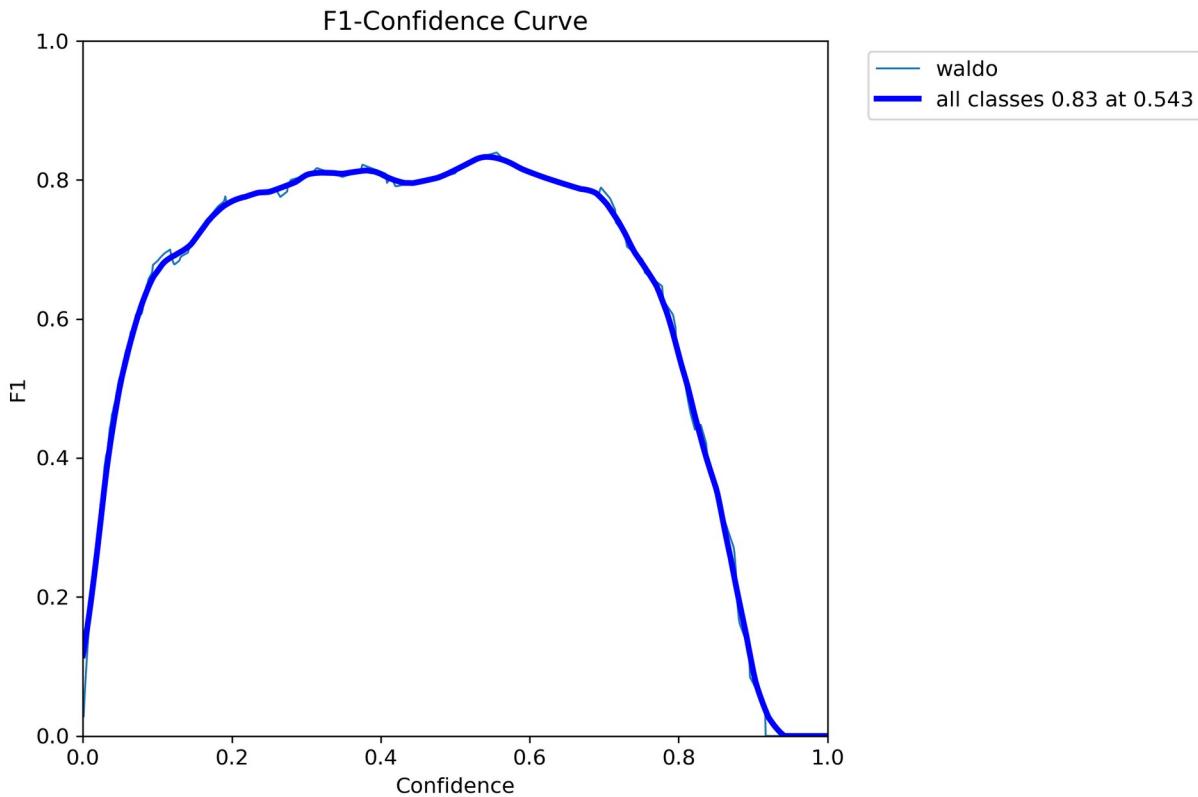
```
Image(filename=f'/content/yolov5/runs/train/exp6/confusion_matrix.png', width=900)
```



## Display the F1 Score Curve

- The F1 score is a metric that combines precision and recall, providing a single score to assess the model's performance, especially in imbalanced datasets. This curve helps in understanding how well the model balances precision and recall over different operating points.

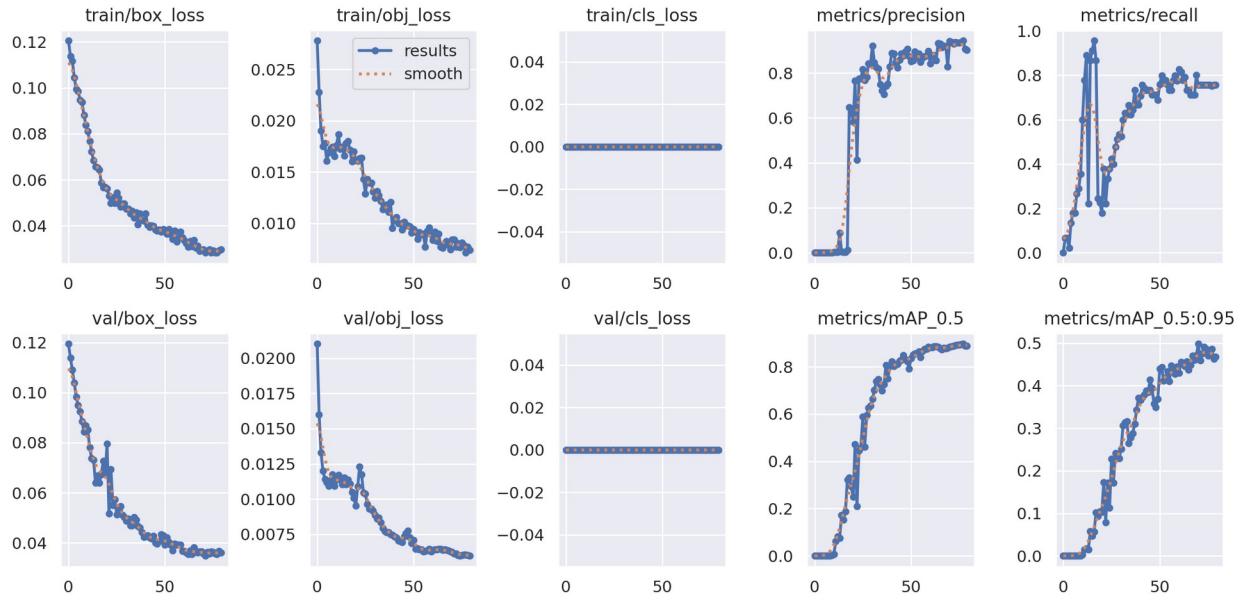
```
Image(filename=f'/content/yolov5/runs/train/exp6/F1_curve.png', width=900)
```



## Display the Results Image

the file typically includes visualizations of the training results, such as detection performance on validation images.

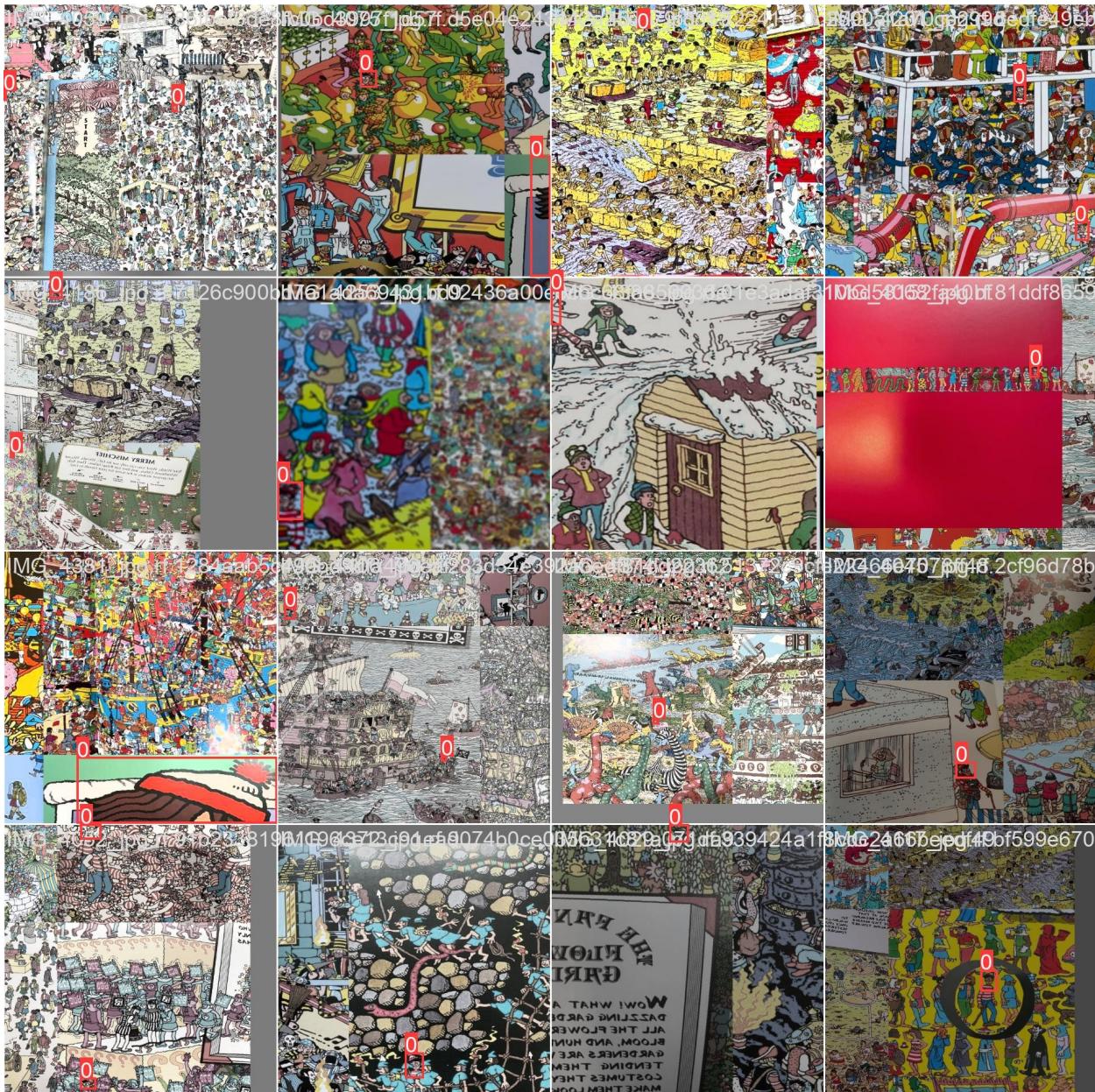
```
Image(filename=f'/content/yolov5/runs/train/exp6/results.png',  
width=900)
```



## Display an Image with Predictions

- The `val_batch0_pred.jpg` file shows an image from the validation set with the model's predictions. Each detected object is annotated with its predicted class. This visualization helps in evaluating how well the model predicts and classifies objects in the validation images.

```
Image(filename=f'/content/yolov5/runs/train/exp6/train_batch1.jpg',
      width=900)
```



## Run YOLOv8 for Prediction

- The command runs the YOLOv8 CLI tool to perform object detection on a specified image using a pre-trained model

```
!python /content/yolov5/detect.py --weights  
/content/yolov5/runs/train/exp6/weights/best.pt --img 640 --conf 0.3  
--source /content/waldo_1.jpg
```

```
detect: weights=['/content/yolov5/runs/train/exp6/weights/best.pt'],  
source=/content/waldo_1.jpg, data=yolov5/data/coco128.yaml,  
imgsz=[640, 640], conf_thres=0.1, iou_thres=0.45, max_det=1000,
```

```
device=, view_img=False, save_txt=False, save_csv=False,
save_conf=False, save_crop=False, nosave=False, classes=None,
agnostic_nms=False, augment=False, visualize=False, update=False,
project=yolov5/runs/detect, name=exp, exist_ok=False,
line_thickness=3, hide_labels=False, hide_conf=False, half=False,
dnn=False, vid_stride=1
YOLOv5 7.0-345-g8003649c Python-3.10.12 torch-2.3.1+cu121 CUDA:0
(Tesla T4, 15102MiB)

Fusing layers...
Model summary: 157 layers, 7012822 parameters, 0 gradients, 15.8
GFLOPs
image 1/1 /content/waldo_1.jpg: 384x640 4 waldos, 51.4ms
Speed: 0.5ms pre-process, 51.4ms inference, 530.9ms NMS per image at
shape (1, 3, 640, 640)
Results saved to yolov5/runs/detect/exp3
```

# SAHI (Slicing Aided Hyper Inference)

## Overview

SAHI, short for Slicing Aided Hyper Inference, is a method designed to improve the performance and accuracy of object detection models, especially on large images or images with densely packed objects. It works by slicing large images into smaller, more manageable pieces, performing inference on these smaller slices, and then combining the results.

## Key Features

1. **Slicing Large Images:**
  - SAHI slices large images into smaller tiles, enabling the model to process each tile independently. This approach is particularly useful for high-resolution images where objects might be small and spread across a large area.
2. **Handling Dense Object Regions:**
  - By breaking down the image, SAHI allows for better detection in densely populated object areas. Smaller slices mean the model can focus on detailed regions without being overwhelmed by the entire image's complexity.
3. **Combining Results:**
  - After processing the slices, SAHI merges the detection results to form a cohesive output. This ensures that objects detected in different slices are correctly mapped back to their original positions in the large image.
4. **Improving Detection Accuracy:**
  - The slicing mechanism helps in improving the overall detection accuracy by ensuring that objects at different scales and positions are effectively detected without being missed due to image size constraints.

# How SAHI Works

## 1. Image Slicing

The input image is divided into smaller overlapping or non-overlapping tiles. Overlapping tiles can help in ensuring that objects near the edges of slices are not missed.

## 2. Model Inference

Each tile is then fed independently to the object detection model (like YOLO). The model processes each tile and returns detection results for that specific region.

## 3. Post-Processing

The detection results from all the tiles are then combined. This step involves adjusting the coordinates of the detected objects to their original positions in the large image.

```
from sahi.utils.yolov8 import (
    download_yolov8s_model
)

from sahi import AutoDetectionModel
import torch
from sahi.utils.cv import read_image
from sahi.utils.file import download_from_url
from sahi.predict import get_prediction, get_sliced_prediction,
predict
from sahi.predict import visualize_object_predictions
from IPython.display import Image
from numpy import asarray
from ultralytics import YOLO

yolov8_model_path = "your_model_path"
download_yolov8s_model(yolov8_model_path)

device = torch.device("cuda" if torch.cuda.is_available() else "cpu")
print(f"Using device: {device}")

detection_model = AutoDetectionModel.from_pretrained(
    model_type='yolov8',
    model_path=yolov8_model_path,
    confidence_threshold=0.2,
    device=device, # or 'cuda:0'
)
image_path = 'your_image_path'

result = get_prediction(image_path, detection_model)
result.export_visuals(export_dir="demo_data/")
```

```

Image("demo_data/prediction_visual.png")

result = get_sliced_prediction(
    image_path,
    detection_model,
    slice_height = 256,
    slice_width = 256,
    overlap_height_ratio = 0.2,
    overlap_width_ratio = 0.2
)

result.export_visuals(export_dir="demo_data/")

Image("demo_data/prediction_visual.png")

```

## Conclusion: YOLOv8 vs YOLOv5 with SAHI Performance Comparison

While both YOLOv5 and YOLOv8 are highly effective object detection models, YOLOv8 has shown substantial improvements over YOLOv5 in various aspects, even when SAHI (Slicing Aided Hyper Inference) is applied. Here's a comparison highlighting why YOLOv8 often outperforms YOLOv5:

### 1. Architectural Advancements

- **YOLOv8:** Incorporates several architectural enhancements over YOLOv5, including more efficient backbone networks and improved neck designs. These upgrades lead to better feature extraction and more accurate object localization and classification.
- **YOLOv5:** Although effective, it lacks some of the architectural innovations present in YOLOv8, which can limit its performance on complex and diverse datasets.

### 2. Accuracy and Speed

- **YOLOv8:** Offers significant improvements in both accuracy and inference speed. YOLOv8's advancements in model architecture and optimization techniques result in higher mAP (mean Average Precision) and faster detection times compared to YOLOv5.
- **YOLOv5 with SAHI:** SAHI improves the accuracy of YOLOv5 by addressing challenges related to large images and dense object regions. However, the inherent limitations of YOLOv5's architecture still affect its overall performance relative to YOLOv8.

### 3. Handling of Large and Complex Images

- **YOLOv8:** Its optimized architecture enables it to handle large and complex images more effectively without requiring additional techniques like SAHI. YOLOv8's inherent capabilities make it more robust and efficient in processing high-resolution and densely packed images.

- **YOLOv5 with SAHI:** While SAHI enhances YOLOv5's ability to process large images by slicing them into manageable tiles, the underlying model architecture still imposes constraints. YOLOv5's performance, even with SAHI, may not match the efficiency and accuracy of YOLOv8.

## 4. Resource Efficiency

- **YOLOv8:** Optimized to deliver high performance while maintaining computational efficiency. It achieves better results with lower computational resources, making it suitable for deployment in resource-constrained environments.
- **YOLOv5 with SAHI:** The SAHI technique, while useful, adds computational overhead due to the need for slicing and merging images. YOLOv5 may require more computational resources and time to achieve comparable results to YOLOv8.

## Summary

In summary, while SAHI significantly enhances YOLOv5's performance on large and complex images by improving object detection accuracy and efficiency, YOLOv8 generally outperforms YOLOv5 even when SAHI is applied. YOLOv8's advanced architecture provides superior accuracy, faster inference times, and better handling of large and complex images, making it a more powerful and efficient choice for modern object detection tasks. The improvements in YOLOv8 make it a preferred option for applications requiring high performance and precision, even beyond the benefits of techniques like SAHI.