

```
In [1]: # Import required libraries
import pandas as pd
import numpy as np
import os
import glob
from datetime import datetime
import xml.etree.ElementTree as ET
import cv2
import matplotlib.pyplot as plt
import seaborn as sns
import warnings
warnings.filterwarnings('ignore')
```

```
/opt/conda/lib/python3.10/site-packages/scipy/_init__.py:146: UserWarning:
A NumPy version >=1.16.5 and <1.23.0 is required for this version of SciPy
(detected version 1.23.5
    warnings.warn(f"A NumPy version >={np_minversion} and <{np_maxversion}"
```

```
In [2]: !git clone https://github.com/ultralytics/yolov5
%cd yolov5
!pip install -qr requirements.txt
```

```
Cloning into 'yolov5'...
remote: Enumerating objects: 15994, done.
remote: Counting objects: 100% (27/27), done.
remote: Compressing objects: 100% (15/15), done.
remote: Total 15994 (delta 18), reused 18 (delta 12), pack-reused 15967
Receiving objects: 100% (15994/15994), 14.58 MiB | 13.15 MiB/s, done.
Resolving deltas: 100% (10984/10984), done.
/kaggle/working/yolov5
```

```
In [3]: input_data = '/kaggle/input/face-mask-detection'
```

```
In [4]: output_data = '/kaggle/working'
```

```
In [5]: annotations_path = "/kaggle/input/face-mask-detection/annotations"
```

```
In [6]: images_path = "/kaggle/input/face-mask-detection/images"
```

```
In [7]: dataset = {
    "file": [],
    "name": [],
    "width": [],
    "height": [],
    "xmin": [],
    "ymin": [],
    "xmax": [],
    "ymax": []
}

for anno in glob.glob(annotations_path+"*.xml"):
    tree = ET.parse(anno)

    for elem in tree.iter():
        if 'size' in elem.tag:
            for attr in list(elem):
                if 'width' in attr.tag:
                    width = int(round(float(attr.text)))
                if 'height' in attr.tag:
                    height = int(round(float(attr.text)))

        if 'object' in elem.tag:
            for attr in list(elem):

                if 'name' in attr.tag:
                    name = attr.text
                    dataset['name']+=[name]
                    dataset['width']+=[width]
                    dataset['height']+=[height]
                    dataset['file']+=[anno.split('/')[-1][-1][0:-4]]

                if 'bndbox' in attr.tag:
                    for dim in list(attr):
                        if 'xmin' in dim.tag:
                            xmin = int(round(float(dim.text)))
                            dataset['xmin']+=[xmin]
                        if 'ymin' in dim.tag:
                            ymin = int(round(float(dim.text)))
                            dataset['ymin']+=[ymin]
                        if 'xmax' in dim.tag:
                            xmax = int(round(float(dim.text)))
                            dataset['xmax']+=[xmax]
                        if 'ymax' in dim.tag:
                            ymax = int(round(float(dim.text)))
                            dataset['ymax']+=[ymax]
```

```
In [8]: df=pd.DataFrame(dataset)
df.head()
```

Out[8]:	file		name	width	height	xmin	ymin	xmax	ymax
0	maksssksksss737		with_mask	400	226	28	55	46	71
1	maksssksksss737		with_mask	400	226	98	62	111	78
2	maksssksksss737	mask_weared_incorrect		400	226	159	50	193	90
3	maksssksksss737		with_mask	400	226	293	59	313	80
4	maksssksksss737		with_mask	400	226	352	51	372	72

```
In [9]: name_dict = {
    'with_mask': 0,
    'mask_weared_incorrect': 1,
    'without_mask': 2
}
```

```
df['class'] = df['name'].map(name_dict)
```

```
In [10]: np.sort(df.name.unique())
```

```
Out[10]: array(['mask_weared_incorrect', 'with_mask', 'without_mask'], dtype=object)
```

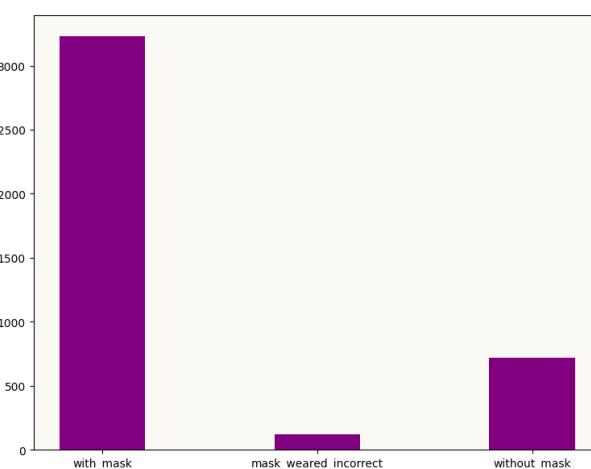
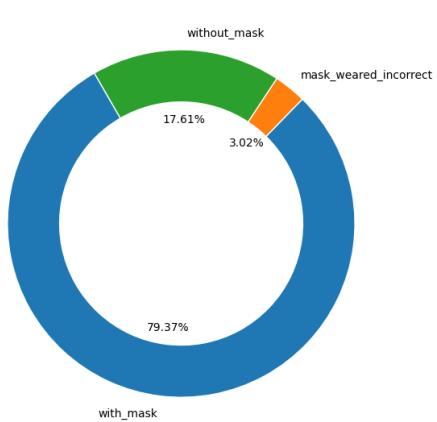
```
In [11]: labels = df.name.unique()
labels_list = [df['name'].value_counts()["with_mask"], df['name'].value_counts()["mask_weared_incorrect"], df['name'].value_counts()["without_mask"]]
print(labels_list)
print(labels)

[3232, 123, 717]
['with_mask' 'mask_weared_incorrect' 'without_mask']
```

```
In [12]: fig, (ax1, ax2) = plt.subplots(1, 2, figsize=(20,7))
background_color = '#faf9f4'
ax1.set_facecolor(background_color)
ax2.set_facecolor(background_color)
ax1.pie(labels_list, wedgeprops=dict(width=0.3, edgecolor='w'),
         labels=labels, radius=1, startangle=120, autopct='%1.2f%%')

ax2 = plt.bar(labels, list(labels_list),
               color ='purple', width = 0.4)

plt.show()
```



```
In [13]: fileNames = [*os.listdir("/kaggle/input/face-mask-detection/images")]
print('There are {} images in the dataset'.format(len(fileNames)))
```

There are 853 images in the dataset

```
In [14]: from sklearn.model_selection import train_test_split
train, test = train_test_split(fileNames, test_size=0.1, random_state=22)
test, val = train_test_split(test, test_size=0.7, random_state=22)
print("Length of Train =", len(train))
print("=*30")
print("Length of Valid =", len(val))
print("=*30")
print("Length of test =", len(test))
```

Length of Train = 767
=====
Length of Valid = 61
=====
Length of test = 25

```
In [15]: os.chdir('/kaggle/working/')
os.mkdir('./yolov5/data/train')
os.mkdir('./yolov5/data/val')
os.mkdir('./yolov5/data/test')
os.mkdir('./yolov5/data/train/images')
os.mkdir('./yolov5/data/train/labels')
os.mkdir('./yolov5/data/test/images')
os.mkdir('./yolov5/data/test/labels')
os.mkdir('./yolov5/data/val/images')
os.mkdir('./yolov5/data/val/labels')
```

```
In [16]: from PIL import Image

def copyImages(imageList, folder_Name):
    for image in imageList:
        img = Image.open(input_data+"/images/"+image)
        img1 = img.resize((640, 480))
        _ = img1.save(output_data+"/yolov5/data/"+folder_Name+"/"+image)

copyImages(train, "train")
copyImages(val, "val")
copyImages(test, "test")
```

```
In [17]: df['xmax'] = (640/df['width'])*df['xmax']
df['ymax'] = (480/df['height'])*df['ymax']
df['xmin'] = (640/df['width'])*df['xmin']
df['ymin'] = (480/df['height'])*df['ymin']
```

```
In [18]: df['x_center'] = (df['xmax']+df['xmin'])/(2*640)
df['y_center'] = (df['ymax']+df['ymin'])/(2*480)
df['box_height'] = (df['xmax']-df['xmin'])/(640)
df['box_width'] = (df['ymax']-df['ymin'])/(480)
```

```
In [19]: df = df.astype('string')
```

```
In [20]: def create_labels(image_list, data_name):
    fileNames = [x.split(".")[0] for x in image_list]

    for name in fileNames:
        data = df[df.file==name]
        box_list = []

        for index in range(len(data)):
            row = data.iloc[index]
            box_list.append(row['class']+ " "+row["x_center"]+ " "+row["y_center"]
                            + " "+row["box_height"]+ " "+row["box_width"])

        text = "\n".join(box_list)
        with open(output_data+"/yolov5/data/"+data_name+"/labels/"+name+".txt", "w") as file:
            file.write(text)

create_labels(train, "train")
create_labels(val, "val")
create_labels(test, "test")
```

```
In [21]: %cd yolov5
from IPython.display import Image, clear_output
import torch
from yolov5 import utils
display = utils.notebook_init()
```

YOLOv5 🚀 v7.0-224-g6262c7f Python-3.10.12 torch-2.0.0 CUDA:0 (Tesla P100-P
CIE-16GB, 16281MiB)
Setup complete 🎉 (2 CPUs, 15.6 GB RAM, 5016.7/8062.4 GB disk)

```
In [22]: yaml_text = """train: data/train/images
val: data/train/images

nc: 3
names: ['with_mask', 'mask_weared_incorrect', 'without_mask']"""

with open("data/data.yaml", 'w') as file:
    file.write(yaml_text)

%cat data/data.yaml

train: data/train/images
val: data/train/images

nc: 3
names: ['with_mask', 'mask_weared_incorrect', 'without_mask']
```

```
In [23]: # Customize iPython writeline for writing variables
from IPython.core.magic import register_line_cell_magic

@register_line_cell_magic
def writetemplate(line, cell):
    with open(line, 'w') as f:
        f.write(cell.format(**globals()))
```

```
In [24]: %%writetemplate models/custom_yolov5s.yaml

# parameters
nc: 3 # number of classes
depth_multiple: 0.33 # model depth multiple
width_multiple: 0.50 # layer channel multiple

# anchors
anchors:
    - [10,13, 16,30, 33,23] # P3/8
    - [30,61, 62,45, 59,119] # P4/16
    - [116,90, 156,198, 373,326] # P5/32

# YOLOv5 backbone
backbone:
    # [from, number, module, args]
    [[-1, 1, Focus, [64, 3]], # 0-P1/2
     [-1, 1, Conv, [128, 3, 2]], # 1-P2/4
     [-1, 3, BottleneckCSP, [128]],
     [-1, 1, Conv, [256, 3, 2]], # 3-P3/8
     [-1, 9, BottleneckCSP, [256]],
     [-1, 1, Conv, [512, 3, 2]], # 5-P4/16
     [-1, 9, BottleneckCSP, [512]],
     [-1, 1, Conv, [1024, 3, 2]], # 7-P5/32
     [-1, 1, SPP, [1024, [5, 9, 13]]],
     [-1, 3, BottleneckCSP, [1024, False]], # 9
    ]

# YOLOv5 head
head:
    [[-1, 1, Conv, [512, 1, 1]],
     [-1, 1, nn.Upsample, [None, 2, 'nearest']],
     [[-1, 6], 1, Concat, [1]], # cat backbone P4
     [-1, 3, BottleneckCSP, [512, False]], # 13

     [-1, 1, Conv, [256, 1, 1]],
     [-1, 1, nn.Upsample, [None, 2, 'nearest']],
     [[-1, 4], 1, Concat, [1]], # cat backbone P3
     [-1, 3, BottleneckCSP, [256, False]], # 17 (P3/8-small)

     [-1, 1, Conv, [256, 3, 2]],
     [[-1, 14], 1, Concat, [1]], # cat head P4
     [-1, 3, BottleneckCSP, [512, False]], # 20 (P4/16-medium)

     [-1, 1, Conv, [512, 3, 2]],
     [[-1, 10], 1, Concat, [1]], # cat head P5
     [-1, 3, BottleneckCSP, [1024, False]], # 23 (P5/32-large)

     [[17, 20, 23], 1, Detect, [nc, anchors]], # Detect(P3, P4, P5)
    ]
```

```
In [25]: start = datetime.now()
!python train.py --img 640 --batch 32 --epochs 50 --data data/data.yaml --cf
end = datetime.now()
```

WARNING 🚨 invalid check_version(3.1.31,) requested, please check values.
train: weights=yolov5s.pt, cfg=models/custom_yolov5s.yaml, data=data/data.yaml, hyp=data/hyps/hyp.scratch-low.yaml, epochs=50, batch_size=32, imgsz=640, rect=False, resume=False, nosave=False, noval=False, noautoanchor=False, noplots=False, evolve=None, bucket=, cache=ram, image_weights=False, device=, multi_scale=False, single_cls=False, optimizer=SGD, sync_bn=False, workers=8, project=runs/train, name=yolov5s_results, 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
github: up to date with <https://github.com/ultralytics/yolov5> ✅
 WARNING 🚨 invalid check_version(5.9.5,) requested, please check values.
 YOLOv5 🚀 v7.0-224-g6262c7f Python-3.10.12 torch-2.0.0 CUDA:0 (Tesla P100-P CIE-16GB, 16281MiB)

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 YOLO v5 🚀 runs in Comet
TensorBoard: Start with 'tensorboard --logdir runs/train', view at <http://localhost:6006/>
 Downloading <https://github.com/ultralytics/yolov5/releases/download/v7.0/yolov5s.pt> to yolov5s.pt...
 100% |██████████| 14.1M/14.1M [00:00<00:00, 173 MB/s]

	from	n	params	module
arguments				
0		-1	1	3520 models.common.Focus
[3, 32, 3]				
1		-1	1	18560 models.common.Conv
[32, 64, 3, 2]				
2		-1	1	19904 models.common.BottleneckCSP
[64, 64, 1]				
3		-1	1	73984 models.common.Conv
[64, 128, 3, 2]				
4		-1	3	161152 models.common.BottleneckCSP
[128, 128, 3]				
5		-1	1	295424 models.common.Conv
[128, 256, 3, 2]				
6		-1	3	641792 models.common.BottleneckCSP
[256, 256, 3]				
7		-1	1	1180672 models.common.Conv
[256, 512, 3, 2]				
8		-1	1	656896 models.common.SPP
[512, 512, [5, 9, 13]]				
9		-1	1	1248768 models.common.BottleneckCSP
[512, 512, 1, False]				
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    378624  models.common.BottleneckCSP
[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    95104  models.common.BottleneckCSP
[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   313088  models.common.BottleneckCSP
[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   1248768  models.common.BottleneckCSP
[512, 512, 1, False]
24   [[17, 20, 23] 1    21576  models.yolo.Detect
[3, [[10, 13, 16, 30, 33, 23], [30, 61, 62, 45, 59, 119], [116, 90, 156, 19
8, 373, 326]], [128, 256, 512]]
custom_YOL0v5s summary: 233 layers, 7260488 parameters, 7260488 gradients

```

Transferred 223/369 items from yolov5s.pt

AMP: checks passed ✓

optimizer: SGD(lr=0.01) with parameter groups 59 weight(decay=0.0), 70 weig
ht(decay=0.0005), 62 bias

albumentations: Blur(p=0.01, blur_limit=(3, 7)), MedianBlur(p=0.01, blur_li
mit=(3, 7)), ToGray(p=0.01), CLAHE(p=0.01, clip_limit=(1, 4.0), tile_grid_s
ize=(8, 8))

train: Scanning /kaggle/working/yolov5/data/train/labels... 767 images, 0 b
ackgr

train: New cache created: /kaggle/working/yolov5/data/train/labels.cache

train: Caching images (0.1GB ram): 17%|██████| 129/767 [00:01<00:08, 7
6.63libpng warning: iCCP: Not recognizing known sRGB profile that has been
edited

train: Caching images (0.7GB ram): 100%|██████████| 767/767 [00:10<00:00, 7
5.83i

val: Scanning /kaggle/working/yolov5/data/train/labels.cache... 767 images,
0 ba

val: Caching images (0.3GB ram): 52%|██████| 396/767 [00:06<00:04, 75.
37it/libpng warning: iCCP: Not recognizing known sRGB profile that has been
edited

val: Caching images (0.7GB ram): 100%|██████████| 767/767 [00:11<00:00, 65.
12it/

AutoAnchor: 5.66 anchors/target, 0.999 Best Possible Recall (BPR). Current
anchors are a good fit to dataset ✓

Plotting labels to runs/train/yolov5s_results/labels.jpg...

Image sizes 640 train, 640 val

Using 2 dataloader workers

Logging results to **runs/train/yolov5s_results**

Starting training for 50 epochs...

Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
-------	---------	----------	----------	----------	-----------	----

ze							
40: 1	0/49	7.83G	0.1062	0.06427	0.03557	296	6
50		Class	Images	Instances	P	R	mAP
46	0.000119	all	767	3650	0.000748	0.0591	0.000
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	1/49	7.83G	0.0951	0.07114	0.02986	211	6
50		Class	Images	Instances	P	R	mAP
11	0.000112	all	767	3650	0.000666	0.0527	0.0004
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	2/49	7.83G	0.08738	0.07262	0.0233	218	6
50		Class	Images	Instances	P	R	mAP
46	0.000115	all	767	3650	0.000619	0.0489	0.0004
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	3/49	7.83G	0.08087	0.06514	0.02023	261	6
50		Class	Images	Instances	P	R	mAP
13	0.00144	all	767	3650	0.00253	0.166	0.005
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	4/49	7.83G	0.07618	0.0596	0.01768	316	6
50		Class	Images	Instances	P	R	mAP
52	0.0226	all	767	3650	0.763	0.0921	0.06
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	5/49	7.83G	0.06956	0.05571	0.01743	234	6
50		Class	Images	Instances	P	R	mAP
0.1	0.0397	all	767	3650	0.787	0.124	
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	6/49	7.83G	0.06349	0.05206	0.01817	285	6
50		Class	Images	Instances	P	R	mAP

			all	767	3650	0.8	0.188	0.1
52	0.0559							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si	
40: 1	7/49	7.83G	0.06066	0.04855	0.0169	271	6	
50		Class	Images	Instances	P	R	mAP	
28	0.0962		all	767	3650	0.852	0.207	0.2
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si	
40: 1	8/49	7.83G	0.05565	0.05029	0.01578	192	6	
50		Class	Images	Instances	P	R	mAP	
82	0.113		all	767	3650	0.863	0.23	0.2
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si	
40: 1	9/49	7.83G	0.05314	0.04418	0.01509	204	6	
50		Class	Images	Instances	P	R	mAP	
74	0.154		all	767	3650	0.679	0.38	0.3
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si	
40: 1	10/49	7.83G	0.0525	0.04349	0.01426	280	6	
50		Class	Images	Instances	P	R	mAP	
08	0.161		all	767	3650	0.691	0.409	0.4
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si	
40: 1	11/49	7.83G	0.05055	0.04324	0.01274	201	6	
50		Class	Images	Instances	P	R	mAP	
28	0.166		all	767	3650	0.787	0.391	0.4
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si	
40: 1	12/49	7.83G	0.04817	0.04238	0.01117	197	6	
50		Class	Images	Instances	P	R	mAP	
72	0.207		all	767	3650	0.768	0.431	0.4
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si	

	13/49	7.83G	0.04831	0.04345	0.01102	205	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.858	0.466	0.5
26	0.251						
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
14/49	7.83G	0.04635	0.04059	0.009693		236	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.84	0.451	0.5
04	0.246						
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
15/49	7.83G	0.04454	0.04075	0.0091		207	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.88	0.449	0.5
39	0.279						
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
16/49	7.83G	0.04458	0.04079	0.009346		199	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.89	0.48	0.5
73	0.29						
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
17/49	7.83G	0.04354	0.04018	0.008952		222	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.861	0.477	0.
56	0.283						
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
18/49	7.83G	0.04268	0.04	0.00848		219	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.896	0.493	0.6
16	0.314						
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
19/49	7.83G	0.04154	0.0368	0.00827		258	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.889	0.495	0.6

07 0.317							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	20/49	7.83G	0.04062	0.03701	0.008281	220	6
50		Class	Images	Instances	P	R	mAP
03		all	767	3650	0.888	0.495	0.6
0.319							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	21/49	7.83G	0.03983	0.03638	0.008555	169	6
50		Class	Images	Instances	P	R	mAP
07		all	767	3650	0.862	0.496	0.6
0.326							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	22/49	7.83G	0.03952	0.03642	0.007434	196	6
50		Class	Images	Instances	P	R	mAP
86		all	767	3650	0.894	0.504	0.5
0.317							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	23/49	7.83G	0.03904	0.03705	0.007864	196	6
50		Class	Images	Instances	P	R	mAP
19		all	767	3650	0.907	0.528	0.6
0.333							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	24/49	7.83G	0.03819	0.03703	0.007761	265	6
50		Class	Images	Instances	P	R	mAP
03		all	767	3650	0.891	0.514	0.6
0.336							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	25/49	7.83G	0.03724	0.03665	0.007299	219	6
50		Class	Images	Instances	P	R	mAP
22		all	767	3650	0.904	0.507	0.6
0.357							
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
	26/49	7.83G	0.0371	0.03622	0.007698	233	6

40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.898	0.527	0.6
34	0.361	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	27/49	7.83G	0.03695	0.03845	0.007553	264	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.913	0.534	0.6
61	0.378	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	28/49	7.83G	0.0366	0.0355	0.006999	239	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.917	0.526	0.6
76	0.389	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	29/49	7.83G	0.03549	0.03472	0.006836	250	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.912	0.523	0.6
42	0.361	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	30/49	7.83G	0.03499	0.03367	0.006699	192	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.903	0.533	0.6
21	0.359	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	31/49	7.83G	0.03487	0.03573	0.006753	252	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.909	0.547	0.6
78	0.403	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	32/49	7.83G	0.03498	0.03465	0.006885	294	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.92	0.546	0.6
76	0.4	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances

ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	33/49	7.83G	0.03389	0.03316	0.006297	311	6
50		Class	Images	Instances	P	R	mAP
99	0.417	all	767	3650	0.915	0.555	0.6
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	34/49	7.83G	0.03302	0.03391	0.006165	315	6
50		Class	Images	Instances	P	R	mAP
62	0.403	all	767	3650	0.916	0.545	0.6
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	35/49	7.83G	0.03369	0.03283	0.006287	239	6
50		Class	Images	Instances	P	R	mAP
03	0.433	all	767	3650	0.622	0.682	0.7
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	36/49	7.83G	0.03303	0.03324	0.006188	219	6
50		Class	Images	Instances	P	R	mAP
94	0.433	all	767	3650	0.941	0.544	0.6
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	37/49	7.83G	0.0332	0.03491	0.006124	281	6
50		Class	Images	Instances	P	R	mAP
72	0.407	all	767	3650	0.94	0.557	0.6
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	38/49	7.83G	0.0324	0.03211	0.005959	204	6
50		Class	Images	Instances	P	R	mAP
13	0.446	all	767	3650	0.582	0.685	0.7
ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	39/49	7.83G	0.03216	0.03106	0.006064	261	6

		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.63	0.724	0.7
37	0.467	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	40/49	7.83G	0.03098	0.03222	0.005405	183	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.583	0.66	0.6
91	0.436	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	41/49	7.83G	0.03147	0.03236	0.005858	229	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.597	0.678	0.7
13	0.45	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	42/49	7.83G	0.03182	0.03281	0.00608	325	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.671	0.719	0.7
56	0.483	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	43/49	7.83G	0.03003	0.03055	0.00517	241	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.601	0.695	0.7
33	0.472	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	44/49	7.83G	0.031	0.03291	0.0059	225	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.643	0.738	0.7
67	0.497	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances
ze	45/49	7.83G	0.03106	0.03376	0.005378	386	6
40: 1		Class	Images	Instances	P	R	mAP
50		all	767	3650	0.587	0.709	0.7
35	0.473						

ze	Epoch	GPU_mem	box_loss	obj_loss	cls_loss	Instances	Si
40: 1	46/49	7.83G	0.03017	0.03169	0.005146	227	6
50		Class	Images	Instances	P	R	mAP
41	0.482		all	767	3650	0.615	0.719
69	0.504						0.7
76	0.495						
40: 1	48/49	7.83G	0.03037	0.03084	0.005805	223	6
50		Class	Images	Instances	P	R	mAP
76	0.495		all	767	3650	0.622	0.717
66	0.503						0.

50 epochs completed in 0.279 hours.

Optimizer stripped from runs/train/yolov5s_results/weights/last.pt, 14.9MB
 Optimizer stripped from runs/train/yolov5s_results/weights/best.pt, 14.9MB

Validating runs/train/yolov5s_results/weights/best.pt...

Fusing layers...

custom_YOL0v5s summary: 182 layers, 7251912 parameters, 0 gradients

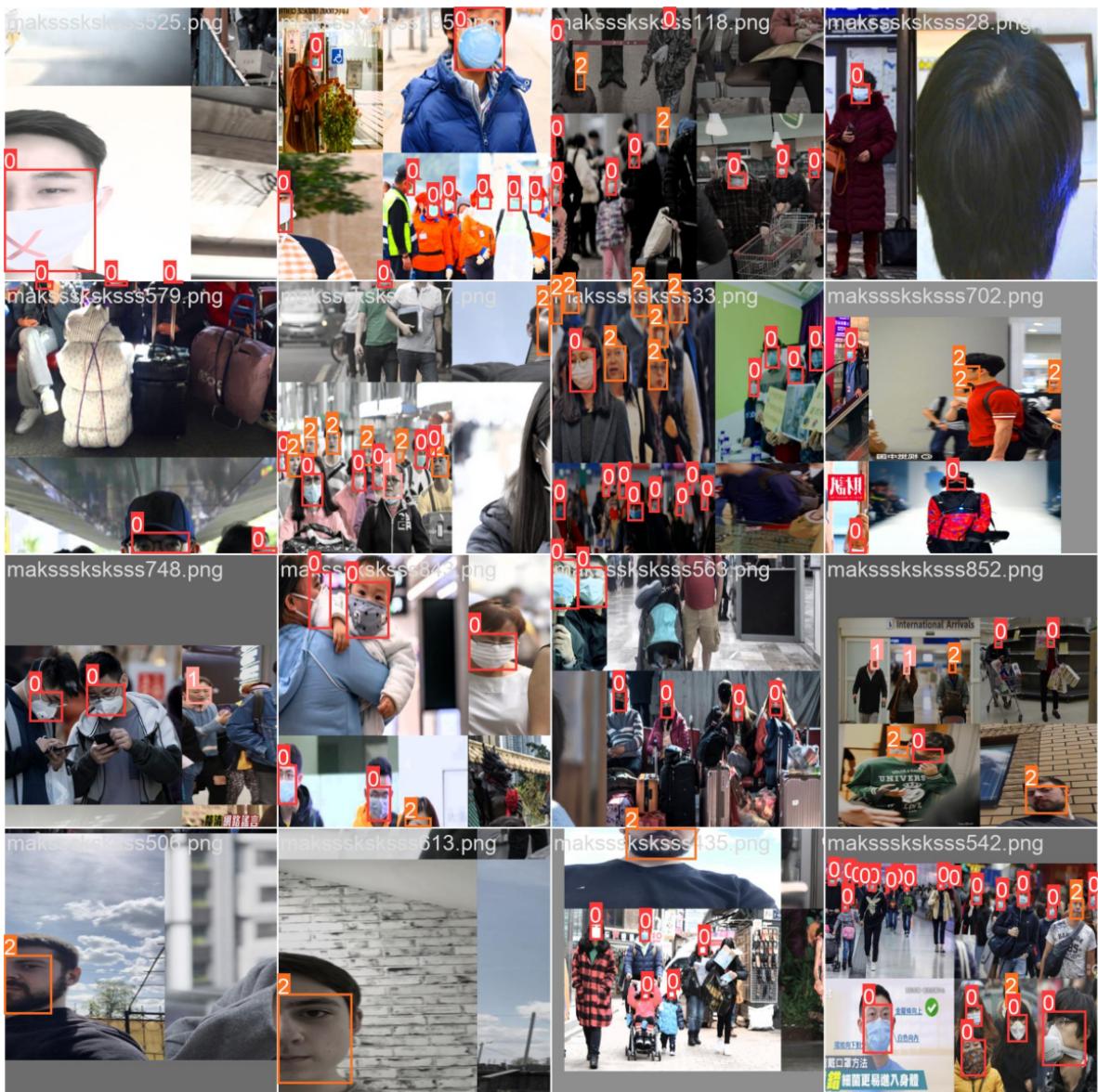
50	Class	Images	Instances	P	R	mAP
69	all	767	3650	0.673	0.735	0.7
68	with_mask	767	2911	0.798	0.962	0.9
55	mask_weared_incorrect	767	112	0.648	0.328	0.4
83	without_mask	767	627	0.572	0.914	0.8
83	0.53					

Results saved to runs/train/yolov5s_results

In [26]: `print("Runtime =", end=start)`

Runtime = 0:18:04.920269

```
In [27]: img = plt.imread('runs/train/yolov5s_results/train_batch2.jpg')
plt.figure(figsize=(30,15))
plt.imshow(img)
plt.axis('off')
plt.show()
```



```
In [ ]:
```

```
In [28]: results_df = pd.read_csv("/kaggle/working/yolov5/runs/train/yolov5s_results")
columns_list = []

for name in results_df.columns:
    if name == 'epoch':
        continue
    else:
        columns_list.append(results_df[name].tolist())

def plot_graphs_yolov5(values, titles, color):
    plt.figure(figsize = (30, 10))

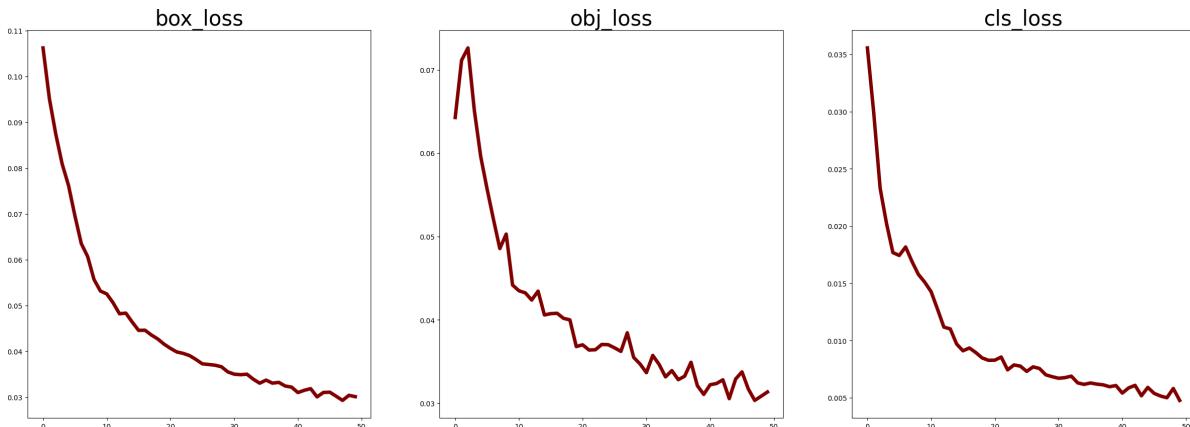
    for i, title in enumerate(titles):
        plt.subplot(1, len(titles), i + 1)
        plt.plot(values[i], linewidth=5, color=color)
        plt.title(title, size=30)

    plt.show()
```

```
In [29]: losses = columns_list[:3]

titles = ['box_loss', 'obj_loss', 'cls_loss']

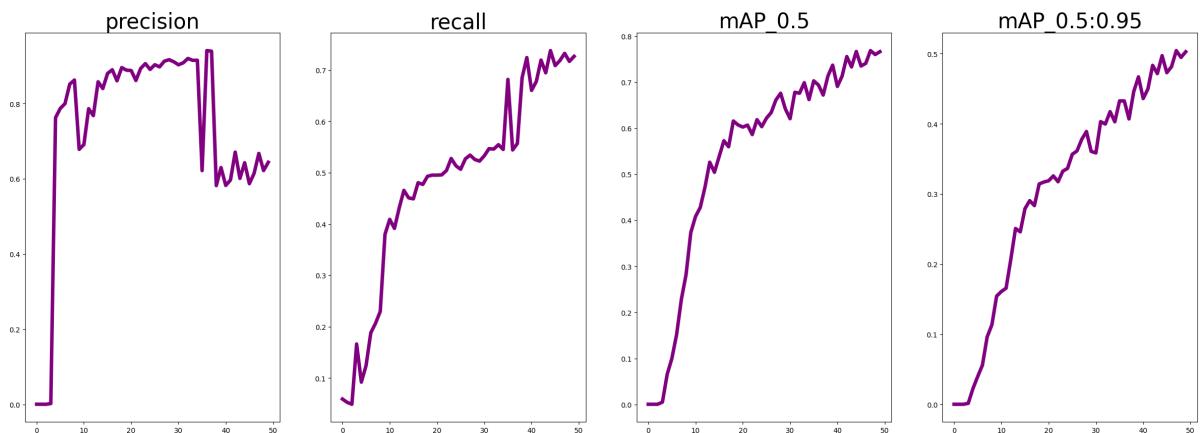
plot_graphs_yolov5(losses, titles, 'maroon')
```



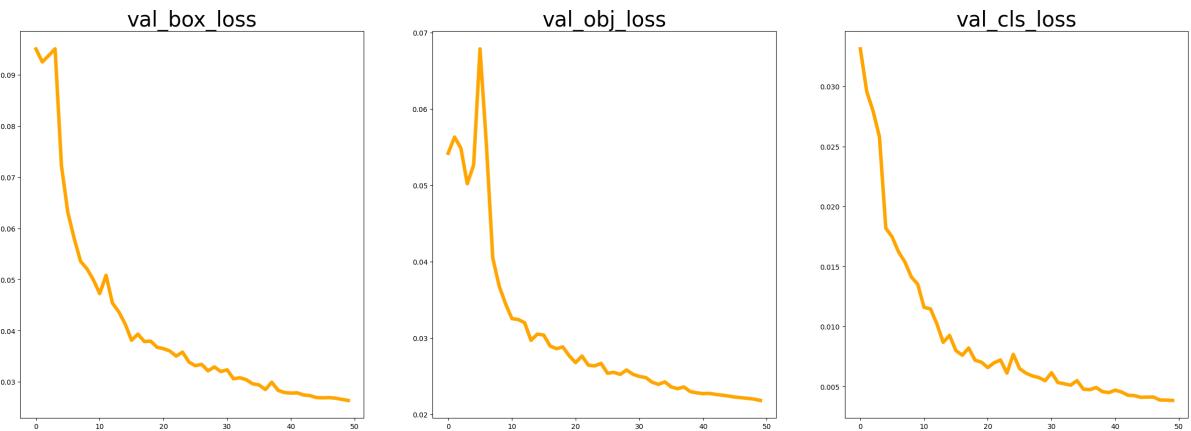
```
In [30]: metrics = columns_list[3:7]

titles = ['precision', 'recall', 'mAP_0.5', 'mAP_0.5:0.95']

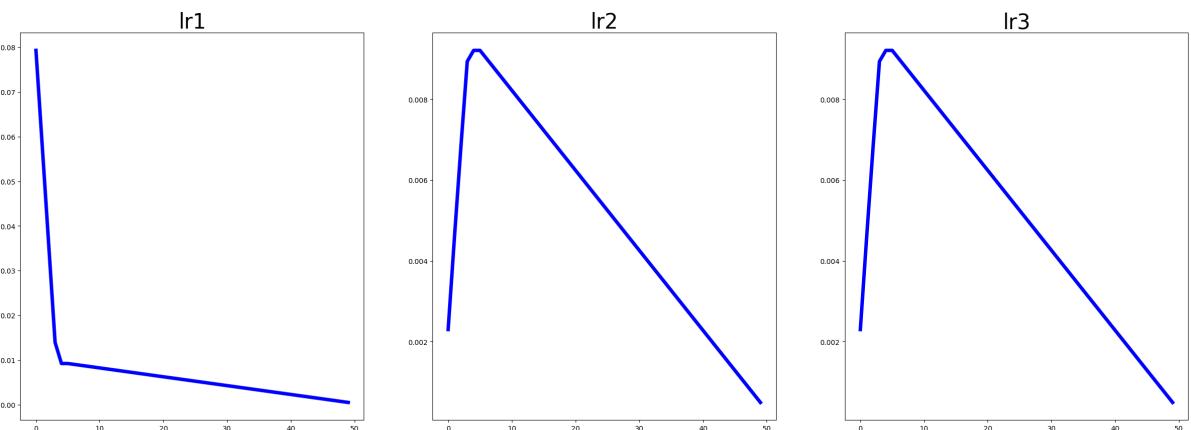
plot_graphs_yolov5(metrics, titles, 'purple')
```



```
In [31]: validation = columns_list[7:10]
titles = ['val_box_loss', 'val_obj_loss', 'val_cls_loss']
plot_graphs_yolov5(validation, titles, 'orange')
```



```
In [32]: learning_rates = columns_list[10:13]
titles = ['lr1', 'lr2', 'lr3']
plot_graphs_yolov5(learning_rates, titles, 'blue')
```



```
In [33]: !python detect.py --source data/test/images/ --weight runs/train/yolov5s_res
```

```
detect: weights=['runs/train/yolov5s_results/weights/best.pt'], source=data
/test/images/, data=data/coco128.yaml, imgsz=[640, 640], conf_thres=0.4, io
u_thres=0.45, max_det=1000, device=, view_img=False, save_txt=False, save_c
sv=False, save_conf=False, save_crop=False, nosave=False, classes=None, agn
ostic_nms=False, augment=False, visualize=False, update=False, project=runs
/detect, name=expTestImage, exist_ok=False, line_thickness=3, hide_labels=F
alse, hide_conf=False, half=False, dnn=False, vid_stride=1
WARNING 🚧 invalid check_version(5.9.5, ) requested, please check values.
YOLOv5 🚶 v7.0-224-g6262c7f Python-3.10.12 torch-2.0.0 CUDA:0 (Tesla P100-P
CIE-16GB, 16281MiB)
```

Fusing layers...

```
custom_YOL0v5s summary: 182 layers, 7251912 parameters, 0 gradients
image 1/25 /kaggle/working/yolov5/data/test/images/maksssksksss151.png: 480
x640 9 with_masks, 46.6ms
image 2/25 /kaggle/working/yolov5/data/test/images/maksssksksss164.png: 480
x640 6 with_masks, 3 without_masks, 7.0ms
image 3/25 /kaggle/working/yolov5/data/test/images/maksssksksss173.png: 480
x640 3 with_masks, 6.5ms
image 4/25 /kaggle/working/yolov5/data/test/images/maksssksksss193.png: 480
x640 3 with_masks, 6.8ms
image 5/25 /kaggle/working/yolov5/data/test/images/maksssksksss236.png: 480
x640 1 without_mask, 6.8ms
image 6/25 /kaggle/working/yolov5/data/test/images/maksssksksss24.png: 480x
640 (no detections), 8.2ms
image 7/25 /kaggle/working/yolov5/data/test/images/maksssksksss323.png: 480
x640 1 with_mask, 7.0ms
image 8/25 /kaggle/working/yolov5/data/test/images/maksssksksss381.png: 480
x640 4 with_masks, 6.5ms
image 9/25 /kaggle/working/yolov5/data/test/images/maksssksksss433.png: 480
x640 1 with_mask, 6.5ms
image 10/25 /kaggle/working/yolov5/data/test/images/maksssksksss466.png: 48
0x640 4 with_masks, 6.6ms
image 11/25 /kaggle/working/yolov5/data/test/images/maksssksksss497.png: 48
0x640 1 with_mask, 6.6ms
image 12/25 /kaggle/working/yolov5/data/test/images/maksssksksss607.png: 48
0x640 2 with_masks, 7.9ms
image 13/25 /kaggle/working/yolov5/data/test/images/maksssksksss611.png: 48
0x640 5 with_masks, 7.2ms
image 14/25 /kaggle/working/yolov5/data/test/images/maksssksksss630.png: 48
0x640 1 with_mask, 6.6ms
image 15/25 /kaggle/working/yolov5/data/test/images/maksssksksss640.png: 48
0x640 9 with_masks, 6.4ms
image 16/25 /kaggle/working/yolov5/data/test/images/maksssksksss65.png: 480
x640 2 with_masks, 6.7ms
image 17/25 /kaggle/working/yolov5/data/test/images/maksssksksss658.png: 48
0x640 (no detections), 6.5ms
image 18/25 /kaggle/working/yolov5/data/test/images/maksssksksss747.png: 48
0x640 1 with_mask, 8.5ms
image 19/25 /kaggle/working/yolov5/data/test/images/maksssksksss750.png: 48
0x640 7 without_masks, 7.2ms
image 20/25 /kaggle/working/yolov5/data/test/images/maksssksksss755.png: 48
0x640 5 with_masks, 7.4ms
image 21/25 /kaggle/working/yolov5/data/test/images/maksssksksss770.png: 48
0x640 4 with_masks, 2 without_masks, 6.7ms
image 22/25 /kaggle/working/yolov5/data/test/images/maksssksksss772.png: 48
0x640 2 without_masks, 6.7ms
image 23/25 /kaggle/working/yolov5/data/test/images/maksssksksss796.png: 48
0x640 13 with_masks, 6.6ms
```

```
image 24/25 /kaggle/working/yolov5/data/test/images/makssksksksss797.png: 48
0x640 6 with_masks, 6.6ms
image 25/25 /kaggle/working/yolov5/data/test/images/makssksksksss848.png: 48
0x640 3 with_masks, 7.1ms
Speed: 0.5ms pre-process, 8.5ms inference, 1.0ms NMS per image at shape (1,
3, 640, 640)
Results saved to runs/detect/expTestImage
```

```
In [34]: color_dict = {
    'with_mask': (0, 255, 0),
    'mask_weared_incorrect': (0, 0, 255),
    'without_mask': (255, 0, 0)
}
```

```
In [35]: def show_image(img_id):
    df_image = df[df.file==img_id]
    df_image[['xmin', 'ymin', 'xmax', 'ymax']] = df_image[['xmin', 'ymin',
    path = 'data/test/images/' + img_id + '.png'
    img = plt.imread(path)

    imge = img.copy()

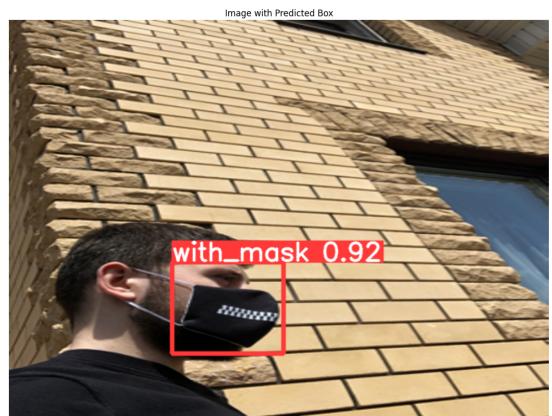
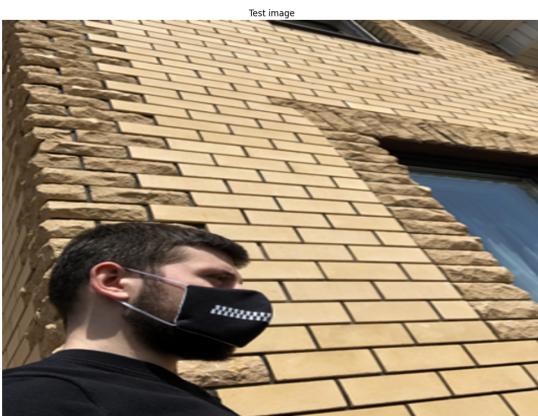
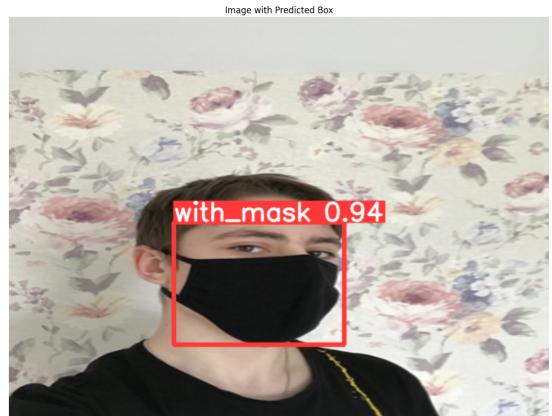
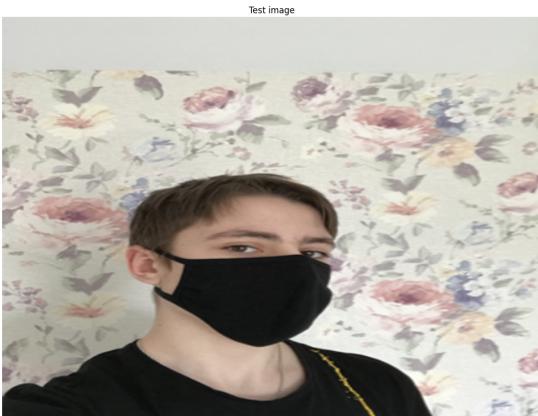
    for index in range(len(df_image)):
        row = df_image.iloc[index]
        cv2.rectangle(imge,
                      (row['xmin'], row['ymin']),
                      (row['xmax'], row['ymax']),
                      color=color_dict[row['name']],
                      thickness=2)

    # results are saved inside run/detect/expTestImage dir
    img_pred = plt.imread('runs/detect/expTestImage/' + img_id)
    # =====
    plt.figure(figsize=(30,20))

    plt.subplot(1,2,1)
    plt.imshow(imge)
    plt.axis('off')
    plt.title('Test image')

    plt.subplot(1,2,2)
    plt.imshow(img_pred)
    plt.axis('off')
    plt.title('Image with Predicted Box')
```

```
In [36]: import os, random
for i in range(5):
    show_image(random.choice(os.listdir("data/test/images/")))
    plt.show()
```





In [1]: