

cv-ppe-project

March 10, 2024

```
[3]: %%time
```

```
! pip install --upgrade ultralytics -qq
```

CPU times: user 30.7 ms, sys: 4.56 ms, total: 35.2 ms

Wall time: 5.92 s

```
[4]: import ultralytics
print(ultralytics.__version__)
```

8.1.10

```
[5]: import warnings
warnings.filterwarnings("ignore")

import os
import re
import glob
import random
import yaml

import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
from matplotlib.patches import Rectangle
import seaborn as sns

import IPython.display as display
from PIL import Image
import cv2

from ultralytics import YOLO
```

```
[6]: from google.colab import drive
drive.mount('/content/drive')
```

Drive already mounted at /content/drive; to attempt to forcibly remount, call drive.mount("/content/drive", force_remount=True).

```
[7]: class CFG:
    DEBUG = False
    FRACTION = 0.05 if DEBUG else 1.0
    SEED = 88

    # classes
    CLASSES = ['Hardhat', 'Mask', 'NO-Hardhat', 'NO-Mask',
               'NO-Safety Vest', 'Person', 'Safety Cone',
               'Safety Vest', 'machinery', 'vehicle']
    NUM_CLASSES_TO_TRAIN = len(CLASSES)

    # training
    EPOCHS = 3 if DEBUG else 50
    BATCH_SIZE = 16

    BASE_MODEL = 'yolov8x'
    BASE_MODEL_WEIGHTS = f'{BASE_MODEL}.pt'
    EXP_NAME = f'ppe_css_{EPOCHS}_epochs'

    OPTIMIZER = 'auto'
    LR = 1e-3
    LR_FACTOR = 0.01
    WEIGHT_DECAY = 5e-4
    DROPOUT = 0.0
    PATIENCE = 20
    PROFILE = False
    LABEL_SMOOTHING = 0.0

    # paths
    CUSTOM_DATASET_DIR = '/content/drive/MyDrive/archive/css-data/'
    OUTPUT_DIR = './'
```

```
[8]: import os
```

```
[9]: dict_file = {
    'train': os.path.join(CFG.CUSTOM_DATASET_DIR, 'train'),
    'val': os.path.join(CFG.CUSTOM_DATASET_DIR, 'valid'),
    'test': os.path.join(CFG.CUSTOM_DATASET_DIR, 'test'),
    'nc': CFG.NUM_CLASSES_TO_TRAIN,
    'names': CFG.CLASSES
}

with open(os.path.join(CFG.OUTPUT_DIR, 'data.yaml'), 'w+') as file:
    yaml.dump(dict_file, file)
```

```
[10]: ### read yaml file created
def read_yaml_file(file_path = CFG.CUSTOM_DATASET_DIR):
```

```

with open(file_path, 'r') as file:
    try:
        data = yaml.safe_load(file)
        return data
    except yaml.YAMLError as e:
        print("Error reading YAML:", e)
        return None

### print it with newlines
def print_yaml_data(data):
    formatted_yaml = yaml.dump(data, default_style=False)
    print(formatted_yaml)

file_path = os.path.join(CFG.OUTPUT_DIR, 'data.yaml')
yaml_data = read_yaml_file(file_path)

if yaml_data:
    print_yaml_data(yaml_data)

```

```

names:
- Hardhat
- Mask
- NO-Hardhat
- NO-Mask
- NO-Safety Vest
- Person
- Safety Cone
- Safety Vest
- machinery
- vehicle
nc: 10
test: /content/drive/MyDrive/archive/css-data/test
train: /content/drive/MyDrive/archive/css-data/train
val: /content/drive/MyDrive/archive/css-data/valid

```

```

[37]: def display_image(image, print_info = True, hide_axis = False):
    if isinstance(image, str):
        img = Image.open(image)
        plt.imshow(img)
    elif isinstance(image, np.ndarray):
        image = image[..., ::-1] # BGR to RGB
        img = Image.fromarray(image)
        plt.imshow(img)
    else:
        raise ValueError("Unsupported image format")

```

```

if print_info:
    print('Type: ', type(img), '\n')
    print('Shape: ', np.array(img).shape, '\n')

if hide_axis:
    plt.axis('off')

plt.show()

```

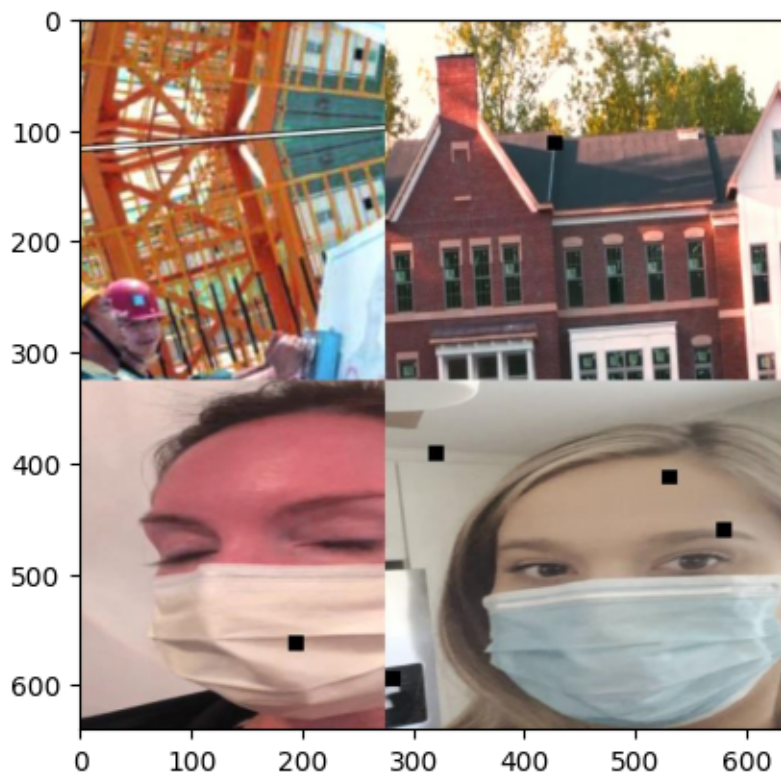
```

[12]: example_image_path = '/content/drive/MyDrive/archive/css-data/train/images/
      ↪-1670-_png_jpg.rf.0463edb430019e01ec79eed27a6349d6.jpg'
      display_image(example_image_path, print_info = True, hide_axis = False)

```

Type: <class 'PIL.JpegImagePlugin.JpegImageFile'>

Shape: (640, 640, 3)



```

[13]: def plot_random_images_from_folder(folder_path, num_images=20, seed=CFG.SEED):
      random.seed(seed)

```

```

image_files = [f for f in os.listdir(folder_path) if f.endswith((''.jpg', '.
png', '.jpeg', '.gif'))]

if len(image_files) < num_images:
    raise ValueError("Not enough images in the folder")

selected_files = random.sample(image_files, num_images)

num_cols = 5
num_rows = (num_images + num_cols - 1) // num_cols
fig, axes = plt.subplots(num_rows, num_cols, figsize=(12, 8))

for i, file_name in enumerate(selected_files):

    img = Image.open(os.path.join(folder_path, file_name))

    if num_rows == 1:
        ax = axes[i % num_cols]
    else:
        ax = axes[i // num_cols, i % num_cols]

    ax.imshow(img)
    ax.axis('off')

for i in range(num_images, num_rows * num_cols):
    if num_rows == 1:
        fig.delaxes(axes[i % num_cols])
    else:
        fig.delaxes(axes[i // num_cols, i % num_cols])

plt.tight_layout()
plt.show()

```

```

[14]: folder_path = CFG.CUSTOM_DATASET_DIR + 'train/images/'
      plot_random_images_from_folder(folder_path, num_images=20, seed=CFG.SEED)

```



```
[15]: def get_image_properties(image_path):
    # Read the image file
    img = cv2.imread(image_path)

    if img is None:
        raise ValueError("Could not read image file")

    properties = {
        "width": img.shape[1],
        "height": img.shape[0],
        "channels": img.shape[2] if len(img.shape) == 3 else 1,
        "dtype": img.dtype,
    }

    return properties
```

```
[16]: img_properties = get_image_properties(example_image_path)
img_properties
```

```
[16]: {'width': 640, 'height': 640, 'channels': 3, 'dtype': dtype('uint8')}
```

```
[17]: %%time
class_idx = {str(i): CFG.CLASSES[i] for i in range(CFG.NUM_CLASSES_TO_TRAIN)}

class_stat = {}
data_len = {}
class_info = []

for mode in ['train', 'valid', 'test']:
    class_count = {CFG.CLASSES[i]: 0 for i in range(CFG.NUM_CLASSES_TO_TRAIN)}

    path = os.path.join(CFG.CUSTOM_DATASET_DIR, mode, 'labels')

    for file in os.listdir(path):
        with open(os.path.join(path, file)) as f:
            lines = f.readlines()

            for cls in set([line[0] for line in lines]):
                class_count[class_idx[cls]] += 1

    data_len[mode] = len(os.listdir(path))
    class_stat[mode] = class_count

    class_info.append({'Mode': mode, **class_count, 'Data_Volume':
↳data_len[mode]})

dataset_stats_df = pd.DataFrame(class_info)
dataset_stats_df
```

CPU times: user 507 ms, sys: 302 ms, total: 809 ms

Wall time: 40.2 s

```
[17]:
```

	Mode	Hardhat	Mask	NO-Hardhat	NO-Mask	NO-Safety Vest	Person \
0	train	1314	1096	1380	1531	1864	2526
1	valid	42	19	37	44	56	84
2	test	30	16	25	30	36	59

	Safety Cone	Safety Vest	machinery	vehicle	Data_Volume
0	631	1319	2101	744	2605
1	13	28	26	16	114
2	8	22	22	15	82

```
[18]: fig, axes = plt.subplots(1, 3, figsize=(15, 5))

# Plot vertical bar plots for each mode in subplots
for i, mode in enumerate(['train', 'valid', 'test']):
    sns.barplot(
```

```

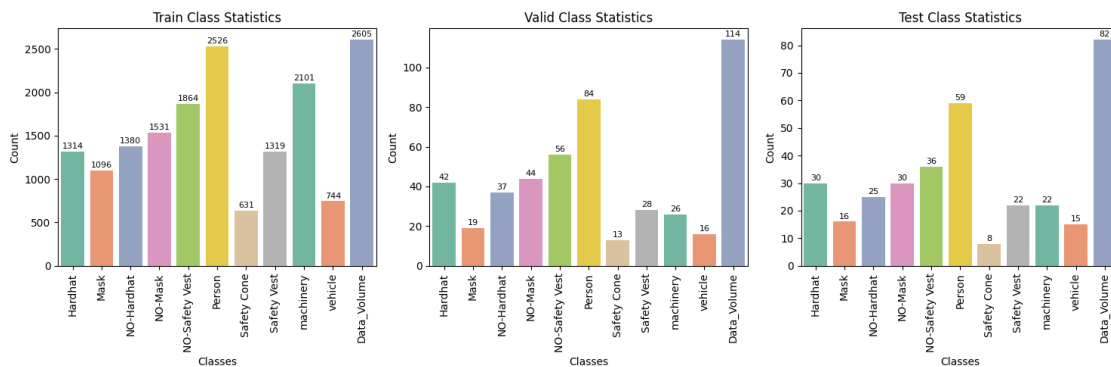
        data=dataset_stats_df[dataset_stats_df['Mode'] == mode].
        drop(columns='Mode'),
        orient='v',
        ax=axes[i],
        palette='Set2'
    )

    axes[i].set_title(f'{mode.capitalize()} Class Statistics')
    axes[i].set_xlabel('Classes')
    axes[i].set_ylabel('Count')
    axes[i].tick_params(axis='x', rotation=90)

    # Add annotations on top of each bar
    for p in axes[i].patches:
        axes[i].annotate(f"{int(p.get_height())}", (p.get_x() + p.get_width() / 2., p.get_height()),
            ha='center', va='center', fontsize=8, color='black',
            xytext=(0, 5),
            textcoords='offset points')

plt.tight_layout()
plt.show()

```



```

[19]: %%time

for mode in ['train', 'valid', 'test']:
    print(f'\nImage sizes in {mode} set:')

    img_size = 0
    for file in glob.glob(os.path.join(CFG.CUSTOM_DATASET_DIR, mode, 'images',
        '*')):

        image = Image.open(file)

```



```
if image.size != img_size:
    print(f'{image.size}')
    img_size = image.size
    print('\n')
```

Image sizes in train set:
(640, 640)

Image sizes in valid set:
(640, 640)

Image sizes in test set:
(640, 640)

CPU times: user 190 ms, sys: 101 ms, total: 291 ms
Wall time: 3.87 s

```
[20]: CFG.BASE_MODEL_WEIGHTS
```

```
[20]: 'yolov8x.pt'
```

```
[21]: import torch

# Define the device
device = torch.device('cuda' if torch.cuda.is_available() else 'cpu')
```

```
[22]: model = YOLO(CFG.BASE_MODEL_WEIGHTS)

results = model.predict(
    source = example_image_path,

    classes = [0],
    conf = 0.30,
    device = device, # inference with dual GPU
    imgsz = (img_properties['height'], img_properties['width']),

    save = True,
    save_txt = True,
    save_conf = True,
    exist_ok = True,
```

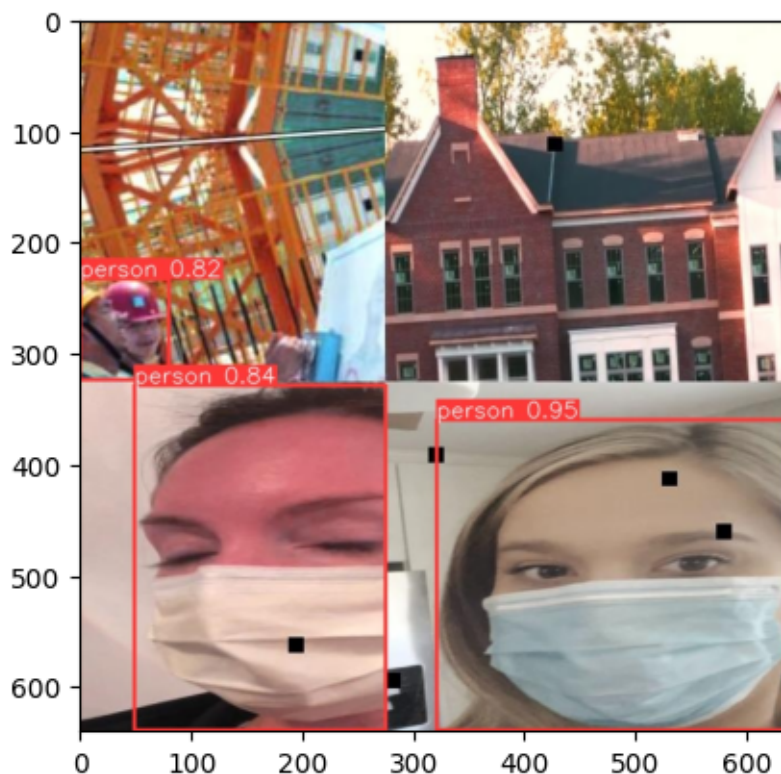
```
)
```

```
image 1/1 /content/drive/MyDrive/archive/css-  
data/train/images/-1670-_png_jpg.rf.0463edb430019e01ec79eed27a6349d6.jpg:  
640x640 3 persons, 97.5ms  
Speed: 4.8ms preprocess, 97.5ms inference, 870.8ms postprocess per image at  
shape (1, 3, 640, 640)  
Results saved to runs/detect/predict  
1 label saved to runs/detect/predict/labels
```

```
[23]: ### check predictions with base model  
example_image_inference_output = example_image_path.split('/')[-1]  
display_image(f'/content/runs/detect/predict/-1670-_png_jpg.rf.  
↪0463edb430019e01ec79eed27a6349d6.jpg')
```

```
Type: <class 'PIL.JpegImagePlugin.JpegImageFile'>
```

```
Shape: (640, 640, 3)
```



```
[24]: print('Model: ', CFG.BASE_MODEL_WEIGHTS)
      print('Epochs: ', CFG.EPOCHS)
      print('Batch: ', CFG.BATCH_SIZE)
```

```
Model:  yolov8x.pt
Epochs:  50
Batch:  16
```

```
[25]: ### Load pre-trained YOLO model
      model = YOLO(CFG.BASE_MODEL_WEIGHTS)
```

```
[26]: %%time

      ### train
      model.train(
          data=os.path.join(CFG.OUTPUT_DIR, 'data.yaml'),
          task='detect',
          imgsz=(img_properties['height'], img_properties['width']),
          epochs=CFG.EPOCHS,
          batch=CFG.BATCH_SIZE,
          optimizer=CFG.OPTIMIZER,
          lr0=CFG.LR,
          lrf=CFG.LR_FACTOR,
          weight_decay=CFG.WEIGHT_DECAY,
          dropout=CFG.DROPOUT,
          fraction=CFG.FRACTION,
          patience=CFG.PATIENCE,
          profile=CFG.PROFILE,
          label_smoothing=CFG.LABEL_SMOOTHING,
          name=f'{CFG.BASE_MODEL}_{CFG.EXP_NAME}',
          seed=CFG.SEED,
          val=True,
          amp=True,
          exist_ok=True,
          resume=False,
          device=0, # Specify CPU as the device
          verbose=False,
      )
```

Ultralytics YOLOv8.1.10 Python-3.10.12 torch-2.1.0+cu121 CUDA:0 (Tesla T4, 15102MiB)

```
engine/trainer: task=detect, mode=train, model=yolov8x.pt,
data=./data.yaml, epochs=50, time=None, patience=20, batch=16, imgsz=(640, 640),
save=True, save_period=-1, cache=False, device=0, workers=8, project=None,
name=yolov8x_ppe_css_50_epochs, exist_ok=True, pretrained=True, optimizer=auto,
verbose=False, seed=88, 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=0.001, 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,
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/yolov8x_ppe_css_50_epochs
Overriding model.yaml nc=80 with nc=10

```

	from	n	params	module
arguments				
0	-1	1	2320	ultralytics.nn.modules.conv.Conv
[3, 80, 3, 2]				
1	-1	1	115520	ultralytics.nn.modules.conv.Conv
[80, 160, 3, 2]				
2	-1	3	436800	ultralytics.nn.modules.block.C2f
[160, 160, 3, True]				
3	-1	1	461440	ultralytics.nn.modules.conv.Conv
[160, 320, 3, 2]				
4	-1	6	3281920	ultralytics.nn.modules.block.C2f
[320, 320, 6, True]				
5	-1	1	1844480	ultralytics.nn.modules.conv.Conv
[320, 640, 3, 2]				
6	-1	6	13117440	ultralytics.nn.modules.block.C2f
[640, 640, 6, True]				
7	-1	1	3687680	ultralytics.nn.modules.conv.Conv
[640, 640, 3, 2]				
8	-1	3	6969600	ultralytics.nn.modules.block.C2f
[640, 640, 3, True]				
9	-1	1	1025920	ultralytics.nn.modules.block.SPPF
[640, 640, 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	3	7379200	ultralytics.nn.modules.block.C2f
[1280, 640, 3]				
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  3   1948800  ultralytics.nn.modules.block.C2f
[960, 320, 3]
 16                -1  1    922240  ultralytics.nn.modules.conv.Conv
[320, 320, 3, 2]
 17            [-1, 12]  1          0  ultralytics.nn.modules.conv.Concat
[1]
 18                -1  3   7174400  ultralytics.nn.modules.block.C2f
[960, 640, 3]
 19                -1  1   3687680  ultralytics.nn.modules.conv.Conv
[640, 640, 3, 2]
 20            [-1, 9]  1          0  ultralytics.nn.modules.conv.Concat
[1]
 21                -1  3   7379200  ultralytics.nn.modules.block.C2f
[1280, 640, 3]
 22            [15, 18, 21]  1   8727598  ultralytics.nn.modules.head.Detect
[10, [320, 640, 640]]
Model summary: 365 layers, 68162238 parameters, 68162222 gradients, 258.2 GFLOPs

```

Transferred 589/595 items from pretrained weights

TensorBoard: Start with 'tensorboard --logdir runs/detect/yolov8x_ppe_css_50_epochs', view at <http://localhost:6006/>
Freezing layer 'model.22.dfl.conv.weight'

AMP: running Automatic Mixed Precision (AMP) checks with YOLOv8n...

Downloading

<https://github.com/ultralytics/assets/releases/download/v8.1.0/yolov8n.pt> to 'yolov8n.pt'...

100%| | 6.23M/6.23M [00:00<00:00, 95.1MB/s]

AMP: checks passed

WARNING updating to 'imgsz=640'. 'train' and 'val' imgsz must be an integer, while 'predict' and 'export' imgsz may be a [h, w] list or an integer, i.e. 'yolo export imgsz=640,480' or 'yolo export imgsz=640'

train: Scanning /content/drive/MyDrive/archive/css-data/train/labels.cache... 550 images, 1 backgrounds, 0 corrupt:
100%| | 550/550 [00:00<?, ?it/s]

train: WARNING /content/drive/MyDrive/archive/css-data/train/images/004720_jpg.rf.afc486560a4004c7cfd67910af31a29c.jpg: 1 duplicate labels removed

augmentations: 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))

val: Scanning /content/drive/MyDrive/archive/css-data/valid/labels.cache... 114 images, 10 backgrounds, 0 corrupt:

100%| | 114/114 [00:00<?, ?it/s]

Plotting labels to runs/detect/yolov8x_ppe_css_50_epochs/labels.jpg...

optimizer: 'optimizer=auto' found, ignoring 'lr0=0.001' and 'momentum=0.937' and determining best 'optimizer', 'lr0' and 'momentum' automatically...

optimizer: AdamW(lr=0.000714, momentum=0.9) with parameter groups 97 weight(decay=0.0), 104 weight(decay=0.0005), 103 bias(decay=0.0)

TensorBoard: model graph visualization added

Image sizes 640 train, 640 val

Using 2 dataloader workers

Logging results to runs/detect/yolov8x_ppe_css_50_epochs

Starting training for 50 epochs...

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
1/50	13.5G	1.424	2.344	1.582	190	640:

100%| | 35/35 [00:50<00:00, 1.43s/it]

Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:05<00:00, 1.44s/it]			
all	114	697	0.343	0.384	0.277

0.111

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
2/50	14.1G	1.365	1.82	1.524	156	640:

100%| | 35/35 [00:48<00:00, 1.38s/it]

Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.11s/it]			
all	114	697	0.21	0.32	0.164

0.0658

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
3/50	14.1G	1.41	1.813	1.548	93	640:

100%| | 35/35 [00:46<00:00, 1.32s/it]

Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.14s/it]			
all	114	697	0.288	0.268	0.202

0.089

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
-------	---------	----------	----------	----------	-----------	------

```

    4/50      14.1G      1.381      1.771      1.555      134      640:
100%|      | 35/35 [00:46<00:00, 1.34s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.16s/it]
      all      114      697      0.347      0.279      0.227
0.0867

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
    5/50      14G      1.352      1.738      1.539      106      640:
100%|      | 35/35 [00:46<00:00, 1.33s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.13s/it]
      all      114      697      0.464      0.342      0.318
0.129

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
    6/50      14G      1.334      1.68      1.53      100      640:
100%|      | 35/35 [00:47<00:00, 1.36s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.24s/it]
      all      114      697      0.366      0.345      0.279
0.12

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
    7/50      14.1G      1.321      1.624      1.483      83      640:
100%|      | 35/35 [00:46<00:00, 1.34s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.08s/it]
      all      114      697      0.451      0.35      0.322
0.13

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
    8/50      14G      1.259      1.516      1.472      136      640:
100%|      | 35/35 [00:48<00:00, 1.38s/it]
      Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.05s/it]

```

all 114 697 0.5 0.422 0.417
0.168

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
9/50	14.1G	1.265	1.474	1.465	148	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.15s/it]				
	all	114	697	0.489	0.437	0.372

0.16

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
10/50	14G	1.232	1.442	1.44	131	640:
100%	35/35	[00:47<00:00, 1.37s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.16s/it]				
	all	114	697	0.595	0.484	0.486

0.192

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
11/50	14G	1.238	1.426	1.432	155	640:
100%	35/35	[00:48<00:00, 1.38s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.05s/it]				
	all	114	697	0.644	0.442	0.496

0.216

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
12/50	14.1G	1.173	1.366	1.421	99	640:
100%	35/35	[00:48<00:00, 1.38s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.09s/it]				
	all	114	697	0.638	0.449	0.46

0.202

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
13/50	14.1G	1.157	1.275	1.389	153	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.07s/it]				
	all	114	697	0.712	0.43	0.478
0.194						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
14/50	14.1G	1.165	1.271	1.378	178	640:
100%	35/35	[00:47<00:00, 1.35s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.05s/it]				
	all	114	697	0.683	0.5	0.542
0.24						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
15/50	14G	1.123	1.217	1.354	143	640:
100%	35/35	[00:47<00:00, 1.35s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.09s/it]				
	all	114	697	0.678	0.51	0.531
0.237						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
16/50	14G	1.107	1.216	1.354	191	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.04s/it]				
	all	114	697	0.666	0.475	0.511
0.222						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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17/50      14G      1.102      1.172      1.347      110      640:
100%|      | 35/35 [00:46<00:00, 1.34s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.03s/it]
          all          114          697      0.673      0.523      0.553
0.235

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
18/50      14.1G          1.1          1.137          1.323          171          640:
100%|      | 35/35 [00:47<00:00, 1.35s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.10s/it]
          all          114          697      0.728      0.536      0.586
0.24

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
19/50      14G          1.06          1.085          1.309          203          640:
100%|      | 35/35 [00:47<00:00, 1.36s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.10s/it]
          all          114          697      0.727      0.541      0.596
0.251

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
20/50      14G          1.041          1.043          1.293          188          640:
100%|      | 35/35 [00:48<00:00, 1.37s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.11s/it]
          all          114          697      0.699      0.55      0.583
0.283

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
21/50      14G          1.033          1.001          1.275          163          640:
100%|      | 35/35 [00:46<00:00, 1.32s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.07s/it]

```

all 114 697 0.753 0.55 0.622
0.281

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
22/50	14.1G	1.045	1.054	1.303	146	640:
100%	35/35	[00:47<00:00, 1.37s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.05s/it]				
	all	114	697	0.72	0.542	0.587

0.279

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
23/50	14G	1.017	1.003	1.279	100	640:
100%	35/35	[00:46<00:00, 1.32s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.03s/it]				
	all	114	697	0.761	0.557	0.609

0.287

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
24/50	14G	0.979	0.9582	1.256	129	640:
100%	35/35	[00:48<00:00, 1.37s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.09s/it]				
	all	114	697	0.794	0.534	0.62

0.292

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
25/50	14G	0.9684	0.9333	1.251	143	640:
100%	35/35	[00:46<00:00, 1.34s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.04s/it]				
	all	114	697	0.809	0.573	0.655

0.296

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
26/50	14G	0.9663	0.9004	1.23	211	640:
100%	35/35	[00:46<00:00, 1.34s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.08s/it]				
	all	114	697	0.767	0.607	0.649
0.299						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
27/50	14.1G	0.946	0.9181	1.231	158	640:
100%	35/35	[00:47<00:00, 1.37s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.05s/it]				
	all	114	697	0.795	0.589	0.647
0.322						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
28/50	14.1G	0.9381	0.854	1.207	202	640:
100%	35/35	[00:47<00:00, 1.36s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.12s/it]				
	all	114	697	0.815	0.594	0.666
0.303						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
29/50	14G	0.9132	0.8375	1.201	90	640:
100%	35/35	[00:46<00:00, 1.32s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.07s/it]				
	all	114	697	0.821	0.613	0.677
0.326						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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```

30/50      14G      0.8989      0.7995      1.194      188      640:
100%|      | 35/35 [00:48<00:00, 1.37s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.11s/it]
          all      114      697      0.749      0.551      0.608
0.3

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
31/50      14G      0.8777      0.8079      1.179      110      640:
100%|      | 35/35 [00:46<00:00, 1.33s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.04s/it]
          all      114      697      0.788      0.596      0.647
0.319

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
32/50      13.8G      0.8616      0.7846      1.164      86      640:
100%|      | 35/35 [00:46<00:00, 1.34s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.02s/it]
          all      114      697      0.82      0.591      0.663
0.311

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
33/50      14.1G      0.8427      0.747      1.153      104      640:
100%|      | 35/35 [00:46<00:00, 1.33s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.03s/it]
          all      114      697      0.819      0.637      0.704
0.328

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
34/50      14.1G      0.8427      0.7387      1.158      132      640:
100%|      | 35/35 [00:47<00:00, 1.35s/it]
          Class      Images  Instances      Box(P      R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.08s/it]

```

	all	114	697	0.835	0.633	0.708
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0.337

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
35/50	14G	0.8299	0.7322	1.152	161	640:
100%	35/35	[00:47<00:00, 1.36s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.08s/it]				
	all	114	697	0.835	0.615	0.695

0.339

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
36/50	14.1G	0.801	0.7144	1.128	162	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.03s/it]				
	all	114	697	0.799	0.636	0.69

0.337

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
37/50	14G	0.7979	0.6967	1.132	149	640:
100%	35/35	[00:48<00:00, 1.37s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.02s/it]				
	all	114	697	0.829	0.634	0.695

0.334

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
38/50	14.1G	0.7841	0.6739	1.124	109	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.03s/it]				
	all	114	697	0.862	0.636	0.715

0.335

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
39/50	14.1G	0.7733	0.6561	1.108	208	640:
100%	35/35	[00:47<00:00, 1.35s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.04s/it]				
	all	114	697	0.838	0.624	0.691
0.337						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
40/50	14.1G	0.7543	0.6369	1.09	88	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.04s/it]				
	all	114	697	0.828	0.643	0.695
0.351						

Closing dataloader mosaic

`alumentations: 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))`

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
41/50	14G	0.7784	0.6011	1.112	64	640:
100%	35/35	[00:51<00:00, 1.47s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.06s/it]				
	all	114	697	0.871	0.602	0.687
0.333						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
42/50	14G	0.7627	0.5754	1.104	66	640:
100%	35/35	[00:45<00:00, 1.31s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.04s/it]				
	all	114	697	0.825	0.628	0.705
0.351						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
43/50	14G	0.7311	0.5336	1.074	88	640:
100%	35/35	[00:46<00:00, 1.34s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.09s/it]				
	all	114	697	0.867	0.623	0.712
0.365						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
44/50	14G	0.7036	0.5212	1.068	96	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.09s/it]				
	all	114	697	0.879	0.625	0.71
0.379						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
45/50	13.9G	0.6852	0.5022	1.053	66	640:
100%	35/35	[00:46<00:00, 1.33s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.04s/it]				
	all	114	697	0.855	0.651	0.72
0.36						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
46/50	14G	0.6757	0.479	1.044	95	640:
100%	35/35	[00:46<00:00, 1.32s/it]				
	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%	4/4	[00:04<00:00, 1.04s/it]				
	all	114	697	0.864	0.643	0.717
0.374						

Epoch	GPU_mem	box_loss	cls_loss	df1_loss	Instances	Size
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```

47/50      14G      0.663      0.4701      1.041      105      640:
100%|      | 35/35 [00:47<00:00, 1.35s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.03s/it]
          all          114          697      0.826      0.648      0.708
0.364

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
48/50      14G      0.6383      0.4542      1.025          71      640:
100%|      | 35/35 [00:46<00:00, 1.32s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.03s/it]
          all          114          697      0.816      0.658      0.711
0.374

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
49/50      14G      0.6243      0.4396      1.008          83      640:
100%|      | 35/35 [00:46<00:00, 1.33s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.04s/it]
          all          114          697      0.843      0.646      0.718
0.377

```

```

Epoch      GPU_mem      box_loss      cls_loss      dfl_loss      Instances      Size
50/50      14G      0.6281      0.4385      1.003          74      640:
100%|      | 35/35 [00:46<00:00, 1.32s/it]
          Class      Images  Instances      Box(P          R      mAP50
mAP50-95): 100%|      | 4/4 [00:04<00:00, 1.05s/it]
          all          114          697      0.86      0.654      0.722
0.375

```

50 epochs completed in 0.852 hours.

Optimizer stripped from runs/detect/yolov8x_ppe_css_50_epochs/weights/last.pt,
136.7MB

Optimizer stripped from runs/detect/yolov8x_ppe_css_50_epochs/weights/best.pt,
136.7MB

Validating runs/detect/yolov8x_ppe_css_50_epochs/weights/best.pt...
 Ultralytics YOLOv8.1.10 Python-3.10.12 torch-2.1.0+cu121 CUDA:0 (Tesla T4, 15102MiB)
 Model summary (fused): 268 layers, 68133198 parameters, 0 gradients, 257.4 GFLOPs

	Class	Images	Instances	Box(P	R	mAP50
mAP50-95): 100%		4/4	[00:07<00:00,	1.81s/it]		
	all	114	697	0.878	0.625	0.71

0.379

Speed: 0.5ms preprocess, 31.1ms inference, 0.0ms loss, 5.1ms postprocess per image

Results saved to runs/detect/yolov8x_ppe_css_50_epochs

CPU times: user 33min 21s, sys: 8min 17s, total: 41min 39s

Wall time: 51min 53s

[26]: ultralytics.utils.metrics.DetMetrics object with attributes:

```

ap_class_index: array([0, 1, 2, 3, 4, 5, 6, 7, 8, 9])
box: ultralytics.utils.metrics.Metric object
confusion_matrix: <ultralytics.utils.metrics.ConfusionMatrix object at 0x7c484fc99570>
curves: ['Precision-Recall(B)', 'F1-Confidence(B)', 'Precision-Confidence(B)', 'Recall-Confidence(B)']
curves_results: [[array([
0, 0.001001, 0.002002, 0.003003,
0.004004, 0.005005, 0.006006, 0.007007, 0.008008, 0.009009,
0.01001, 0.011011, 0.012012, 0.013013, 0.014014, 0.015015,
0.016016, 0.017017, 0.018018, 0.019019, 0.02002, 0.021021,
0.022022, 0.023023,
0.024024, 0.025025, 0.026026, 0.027027, 0.028028,
0.029029, 0.03003, 0.031031, 0.032032, 0.033033, 0.034034,
0.035035, 0.036036, 0.037037, 0.038038, 0.039039, 0.04004,
0.041041, 0.042042, 0.043043, 0.044044, 0.045045, 0.046046,
0.047047,
0.048048, 0.049049, 0.05005, 0.051051, 0.052052,
0.053053, 0.054054, 0.055055, 0.056056, 0.057057, 0.058058,
0.059059, 0.06006, 0.061061, 0.062062, 0.063063, 0.064064,
0.065065, 0.066066, 0.067067, 0.068068, 0.069069, 0.07007,
0.071071,
0.072072, 0.073073, 0.074074, 0.075075, 0.076076,
0.077077, 0.078078, 0.079079, 0.08008, 0.081081, 0.082082,
0.083083, 0.084084, 0.085085, 0.086086, 0.087087, 0.088088,
0.089089, 0.09009, 0.091091, 0.092092, 0.093093, 0.094094,
0.095095,
0.096096, 0.097097, 0.098098, 0.099099, 0.1001,
0.1011, 0.1021, 0.1031, 0.1041, 0.10511, 0.10611,

```

0.10711,	0.10811,	0.10911,	0.11011,	0.11111,	0.11211,
0.11311,	0.11411,	0.11512,	0.11612,	0.11712,	0.11812,
0.11912,					
	0.12012,	0.12112,	0.12212,	0.12312,	0.12412,
0.12513,	0.12613,	0.12713,	0.12813,	0.12913,	0.13013,
0.13113,	0.13213,	0.13313,	0.13413,	0.13514,	0.13614,
0.13714,	0.13814,	0.13914,	0.14014,	0.14114,	0.14214,
0.14314,					
	0.14414,	0.14515,	0.14615,	0.14715,	0.14815,
0.14915,	0.15015,	0.15115,	0.15215,	0.15315,	0.15415,
0.15516,	0.15616,	0.15716,	0.15816,	0.15916,	0.16016,
0.16116,	0.16216,	0.16316,	0.16416,	0.16517,	0.16617,
0.16717,					
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0.23924,					
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0.26326,					
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0.43744,	0.43844,	0.43944,	0.44044,	0.44144,	0.44244,
0.44344,	0.44444,	0.44545,	0.44645,	0.44745,	0.44845,
0.44945,	0.45045,	0.45145,	0.45245,	0.45345,	0.45445,
0.45546,					
	0.45646,	0.45746,	0.45846,	0.45946,	0.46046,
0.46146,	0.46246,	0.46346,	0.46446,	0.46547,	0.46647,
0.46747,	0.46847,	0.46947,	0.47047,	0.47147,	0.47247,
0.47347,	0.47447,	0.47548,	0.47648,	0.47748,	0.47848,
0.47948,					
	0.48048,	0.48148,	0.48248,	0.48348,	0.48448,
0.48549,	0.48649,	0.48749,	0.48849,	0.48949,	0.49049,
0.49149,	0.49249,	0.49349,	0.49449,	0.4955,	0.4965,
0.4975,	0.4985,	0.4995,	0.5005,	0.5015,	0.5025,
0.5035,					
	0.5045,	0.50551,	0.50651,	0.50751,	0.50851,
0.50951,	0.51051,	0.51151,	0.51251,	0.51351,	0.51451,
0.51552,	0.51652,	0.51752,	0.51852,	0.51952,	0.52052,
0.52152,	0.52252,	0.52352,	0.52452,	0.52553,	0.52653,
0.52753,					
	0.52853,	0.52953,	0.53053,	0.53153,	0.53253,
0.53353,	0.53453,	0.53554,	0.53654,	0.53754,	0.53854,
0.53954,	0.54054,	0.54154,	0.54254,	0.54354,	0.54454,
0.54555,	0.54655,	0.54755,	0.54855,	0.54955,	0.55055,
0.55155,					
	0.55255,	0.55355,	0.55455,	0.55556,	0.55656,
0.55756,	0.55856,	0.55956,	0.56056,	0.56156,	0.56256,
0.56356,	0.56456,	0.56557,	0.56657,	0.56757,	0.56857,
0.56957,	0.57057,	0.57157,	0.57257,	0.57357,	0.57457,
0.57558,					
	0.57658,	0.57758,	0.57858,	0.57958,	0.58058,
0.58158,	0.58258,	0.58358,	0.58458,	0.58559,	0.58659,
0.58759,	0.58859,	0.58959,	0.59059,	0.59159,	0.59259,
0.59359,	0.59459,	0.5956,	0.5966,	0.5976,	0.5986,
0.5996,					
	0.6006,	0.6016,	0.6026,	0.6036,	0.6046,
0.60561,	0.60661,	0.60761,	0.60861,	0.60961,	0.61061,
0.61161,	0.61261,	0.61361,	0.61461,	0.61562,	0.61662,
0.61762,	0.61862,	0.61962,	0.62062,	0.62162,	0.62262,
0.62362,					
	0.62462,	0.62563,	0.62663,	0.62763,	0.62863,
0.62963,	0.63063,	0.63163,	0.63263,	0.63363,	0.63463,

0.63564,	0.63664,	0.63764,	0.63864,	0.63964,	0.64064,
0.64164,	0.64264,	0.64364,	0.64464,	0.64565,	0.64665,
0.64765,					
	0.64865,	0.64965,	0.65065,	0.65165,	0.65265,
0.65365,	0.65465,	0.65566,	0.65666,	0.65766,	0.65866,
0.65966,	0.66066,	0.66166,	0.66266,	0.66366,	0.66466,
0.66567,	0.66667,	0.66767,	0.66867,	0.66967,	0.67067,
0.67167,					
	0.67267,	0.67367,	0.67467,	0.67568,	0.67668,
0.67768,	0.67868,	0.67968,	0.68068,	0.68168,	0.68268,
0.68368,	0.68468,	0.68569,	0.68669,	0.68769,	0.68869,
0.68969,	0.69069,	0.69169,	0.69269,	0.69369,	0.69469,
0.6957,					
	0.6967,	0.6977,	0.6987,	0.6997,	0.7007,
0.7017,	0.7027,	0.7037,	0.7047,	0.70571,	0.70671,
0.70771,	0.70871,	0.70971,	0.71071,	0.71171,	0.71271,
0.71371,	0.71471,	0.71572,	0.71672,	0.71772,	0.71872,
0.71972,					
	0.72072,	0.72172,	0.72272,	0.72372,	0.72472,
0.72573,	0.72673,	0.72773,	0.72873,	0.72973,	0.73073,
0.73173,	0.73273,	0.73373,	0.73473,	0.73574,	0.73674,
0.73774,	0.73874,	0.73974,	0.74074,	0.74174,	0.74274,
0.74374,					
	0.74474,	0.74575,	0.74675,	0.74775,	0.74875,
0.74975,	0.75075,	0.75175,	0.75275,	0.75375,	0.75475,
0.75576,	0.75676,	0.75776,	0.75876,	0.75976,	0.76076,
0.76176,	0.76276,	0.76376,	0.76476,	0.76577,	0.76677,
0.76777,					
	0.76877,	0.76977,	0.77077,	0.77177,	0.77277,
0.77377,	0.77477,	0.77578,	0.77678,	0.77778,	0.77878,
0.77978,	0.78078,	0.78178,	0.78278,	0.78378,	0.78478,
0.78579,	0.78679,	0.78779,	0.78879,	0.78979,	0.79079,
0.79179,					
	0.79279,	0.79379,	0.79479,	0.7958,	0.7968,
0.7978,	0.7988,	0.7998,	0.8008,	0.8018,	0.8028,
0.8038,	0.8048,	0.80581,	0.80681,	0.80781,	0.80881,
0.80981,	0.81081,	0.81181,	0.81281,	0.81381,	0.81481,
0.81582,					
	0.81682,	0.81782,	0.81882,	0.81982,	0.82082,
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0.82783,	0.82883,	0.82983,	0.83083,	0.83183,	0.83283,
0.83383,	0.83483,	0.83584,	0.83684,	0.83784,	0.83884,
0.83984,					
	0.84084,	0.84184,	0.84284,	0.84384,	0.84484,
0.84585,	0.84685,	0.84785,	0.84885,	0.84985,	0.85085,
0.85185,	0.85285,	0.85385,	0.85485,	0.85586,	0.85686,
0.85786,	0.85886,	0.85986,	0.86086,	0.86186,	0.86286,

0.86386,					
	0.86486,	0.86587,	0.86687,	0.86787,	0.86887,
0.86987,	0.87087,	0.87187,	0.87287,	0.87387,	0.87487,
0.87588,	0.87688,	0.87788,	0.87888,	0.87988,	0.88088,
0.88188,	0.88288,	0.88388,	0.88488,	0.88589,	0.88689,
0.88789,					
	0.88889,	0.88989,	0.89089,	0.89189,	0.89289,
0.89389,	0.89489,	0.8959,	0.8969,	0.8979,	0.8989,
0.8999,	0.9009,	0.9019,	0.9029,	0.9039,	0.9049,
0.90591,	0.90691,	0.90791,	0.90891,	0.90991,	0.91091,
0.91191,					
	0.91291,	0.91391,	0.91491,	0.91592,	0.91692,
0.91792,	0.91892,	0.91992,	0.92092,	0.92192,	0.92292,
0.92392,	0.92492,	0.92593,	0.92693,	0.92793,	0.92893,
0.92993,	0.93093,	0.93193,	0.93293,	0.93393,	0.93493,
0.93594,					
	0.93694,	0.93794,	0.93894,	0.93994,	0.94094,
0.94194,	0.94294,	0.94394,	0.94494,	0.94595,	0.94695,
0.94795,	0.94895,	0.94995,	0.95095,	0.95195,	0.95295,
0.95395,	0.95495,	0.95596,	0.95696,	0.95796,	0.95896,
0.95996,					
	0.96096,	0.96196,	0.96296,	0.96396,	0.96496,
0.96597,	0.96697,	0.96797,	0.96897,	0.96997,	0.97097,
0.97197,	0.97297,	0.97397,	0.97497,	0.97598,	0.97698,
0.97798,	0.97898,	0.97998,	0.98098,	0.98198,	0.98298,
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	0.024024,	0.025025,	0.026026,	0.027027,	0.028028,
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0.047047,	0.048048,	0.049049,	0.05005,	0.051051,	0.052052,
0.053053,	0.054054,	0.055055,	0.056056,	0.057057,	0.058058,
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0.077077,	0.078078,	0.079079,	0.08008,	0.081081,	0.082082,
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0.14915,	0.15015,	0.15115,	0.15215,	0.15315,	0.15415,
0.15516,	0.15616,	0.15716,	0.15816,	0.15916,	0.16016,
0.16116,	0.16216,	0.16316,	0.16416,	0.16517,	0.16617,
0.16717,	0.16817,	0.16917,	0.17017,	0.17117,	0.17217,
0.17317,	0.17417,	0.17518,	0.17618,	0.17718,	0.17818,
0.17918,	0.18018,	0.18118,	0.18218,	0.18318,	0.18418,
0.18519,	0.18619,	0.18719,	0.18819,	0.18919,	0.19019,
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0.1972,	0.1982,	0.1992,	0.2002,	0.2012,	0.2022,
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0.22122,	0.22222,	0.22322,	0.22422,	0.22523,	0.22623,
0.22723,	0.22823,	0.22923,	0.23023,	0.23123,	0.23223,
0.23323,	0.23423,	0.23524,	0.23624,	0.23724,	0.23824,
0.23924,	0.24024,	0.24124,	0.24224,	0.24324,	0.24424,
0.24525,	0.24625,	0.24725,	0.24825,	0.24925,	0.25025,
0.25125,	0.25225,	0.25325,	0.25425,	0.25526,	0.25626,
0.25726,	0.25826,	0.25926,	0.26026,	0.26126,	0.26226,

0.26326,					
	0.26426,	0.26527,	0.26627,	0.26727,	0.26827,
0.26927,	0.27027,	0.27127,	0.27227,	0.27327,	0.27427,
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0.28128,	0.28228,	0.28328,	0.28428,	0.28529,	0.28629,
0.28729,					
	0.28829,	0.28929,	0.29029,	0.29129,	0.29229,
0.29329,	0.29429,	0.2953,	0.2963,	0.2973,	0.2983,
0.2993,	0.3003,	0.3013,	0.3023,	0.3033,	0.3043,
0.30531,	0.30631,	0.30731,	0.30831,	0.30931,	0.31031,
0.31131,					
	0.31231,	0.31331,	0.31431,	0.31532,	0.31632,
0.31732,	0.31832,	0.31932,	0.32032,	0.32132,	0.32232,
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0.32933,	0.33033,	0.33133,	0.33233,	0.33333,	0.33433,
0.33534,					
	0.33634,	0.33734,	0.33834,	0.33934,	0.34034,
0.34134,	0.34234,	0.34334,	0.34434,	0.34535,	0.34635,
0.34735,	0.34835,	0.34935,	0.35035,	0.35135,	0.35235,
0.35335,	0.35435,	0.35536,	0.35636,	0.35736,	0.35836,
0.35936,					
	0.36036,	0.36136,	0.36236,	0.36336,	0.36436,
0.36537,	0.36637,	0.36737,	0.36837,	0.36937,	0.37037,
0.37137,	0.37237,	0.37337,	0.37437,	0.37538,	0.37638,
0.37738,	0.37838,	0.37938,	0.38038,	0.38138,	0.38238,
0.38338,					
	0.38438,	0.38539,	0.38639,	0.38739,	0.38839,
0.38939,	0.39039,	0.39139,	0.39239,	0.39339,	0.39439,
0.3954,	0.3964,	0.3974,	0.3984,	0.3994,	0.4004,
0.4014,	0.4024,	0.4034,	0.4044,	0.40541,	0.40641,
0.40741,					
	0.40841,	0.40941,	0.41041,	0.41141,	0.41241,
0.41341,	0.41441,	0.41542,	0.41642,	0.41742,	0.41842,
0.41942,	0.42042,	0.42142,	0.42242,	0.42342,	0.42442,
0.42543,	0.42643,	0.42743,	0.42843,	0.42943,	0.43043,
0.43143,					
	0.43243,	0.43343,	0.43443,	0.43544,	0.43644,
0.43744,	0.43844,	0.43944,	0.44044,	0.44144,	0.44244,
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0.44945,	0.45045,	0.45145,	0.45245,	0.45345,	0.45445,
0.45546,					
	0.45646,	0.45746,	0.45846,	0.45946,	0.46046,
0.46146,	0.46246,	0.46346,	0.46446,	0.46547,	0.46647,
0.46747,	0.46847,	0.46947,	0.47047,	0.47147,	0.47247,
0.47347,	0.47447,	0.47548,	0.47648,	0.47748,	0.47848,
0.47948,					
	0.48048,	0.48148,	0.48248,	0.48348,	0.48448,

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0.49149,	0.49249,	0.49349,	0.49449,	0.4955,	0.4965,
0.4975,	0.4985,	0.4995,	0.5005,	0.5015,	0.5025,
0.5035,					
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0.50951,	0.51051,	0.51151,	0.51251,	0.51351,	0.51451,
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0.52152,	0.52252,	0.52352,	0.52452,	0.52553,	0.52653,
0.52753,					
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0.53353,	0.53453,	0.53554,	0.53654,	0.53754,	0.53854,
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0.54555,	0.54655,	0.54755,	0.54855,	0.54955,	0.55055,
0.55155,					
	0.55255,	0.55355,	0.55455,	0.55556,	0.55656,
0.55756,	0.55856,	0.55956,	0.56056,	0.56156,	0.56256,
0.56356,	0.56456,	0.56557,	0.56657,	0.56757,	0.56857,
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0.57558,					
	0.57658,	0.57758,	0.57858,	0.57958,	0.58058,
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0.59359,	0.59459,	0.5956,	0.5966,	0.5976,	0.5986,
0.5996,					
	0.6006,	0.6016,	0.6026,	0.6036,	0.6046,
0.60561,	0.60661,	0.60761,	0.60861,	0.60961,	0.61061,
0.61161,	0.61261,	0.61361,	0.61461,	0.61562,	0.61662,
0.61762,	0.61862,	0.61962,	0.62062,	0.62162,	0.62262,
0.62362,					
	0.62462,	0.62563,	0.62663,	0.62763,	0.62863,
0.62963,	0.63063,	0.63163,	0.63263,	0.63363,	0.63463,
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0.64765,					
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0.66567,	0.66667,	0.66767,	0.66867,	0.66967,	0.67067,
0.67167,					
	0.67267,	0.67367,	0.67467,	0.67568,	0.67668,
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0.91191,					
	0.91291,	0.91391,	0.91491,	0.91592,	0.91692,
0.91792,	0.91892,	0.91992,	0.92092,	0.92192,	0.92292,
0.92392,	0.92492,	0.92593,	0.92693,	0.92793,	0.92893,
0.92993,	0.93093,	0.93193,	0.93293,	0.93393,	0.93493,
0.93594,					
	0.93694,	0.93794,	0.93894,	0.93994,	0.94094,
0.94194,	0.94294,	0.94394,	0.94494,	0.94595,	0.94695,
0.94795,	0.94895,	0.94995,	0.95095,	0.95195,	0.95295,
0.95395,	0.95495,	0.95596,	0.95696,	0.95796,	0.95896,
0.95996,					
	0.96096,	0.96196,	0.96296,	0.96396,	0.96496,
0.96597,	0.96697,	0.96797,	0.96897,	0.96997,	0.97097,
0.97197,	0.97297,	0.97397,	0.97497,	0.97598,	0.97698,
0.97798,	0.97898,	0.97998,	0.98098,	0.98198,	0.98298,
0.98398,					
	0.98498,	0.98599,	0.98699,	0.98799,	0.98899,
0.98999,	0.99099,	0.99199,	0.99299,	0.99399,	0.99499,
0.996,	0.997,	0.998,	0.999,	1]), array([[
0.86076,	0.86076,	0.86076, ...,	0,	0,	0],
[0.90476,	0.90476,	0.85714, ...,	0,	0,
0],					

```

[ 0.71014, 0.71014, 0.71014, ..., 0, 0,
0],
...,
[ 0.90244, 0.90244, 0.90244, ..., 0, 0,
0],
[ 0.96364, 0.96364, 0.94545, ..., 0, 0,
0],
[ 0.66667, 0.66667, 0.66667, ..., 0, 0,
0]]), 'Confidence', 'Recall']]
fitness: 0.41218191984872277
keys: ['metrics/precision(B)', 'metrics/recall(B)', 'metrics/mAP50(B)',
'metrics/mAP50-95(B)']
maps: array([ 0.47072, 0.53125, 0.29947, 0.27766, 0.31274,
0.3923, 0.34982, 0.3935, 0.48893, 0.27398])
names: {0: 'Hardhat', 1: 'Mask', 2: 'NO-Hardhat', 3: 'NO-Mask', 4: 'NO-Safety
Vest', 5: 'Person', 6: 'Safety Cone', 7: 'Safety Vest', 8: 'machinery', 9:
'vehicle'}
plot: True
results_dict: {'metrics/precision(B)': 0.878353021072615, 'metrics/recall(B)':
0.6251615993148584, 'metrics/mAP50(B)': 0.710474628943105,
'metrics/mAP50-95(B)': 0.3790382855049025, 'fitness': 0.41218191984872277}
save_dir: PosixPath('runs/detect/yolov8x_ppe_css_50_epochs')
speed: {'preprocess': 0.5319662261427495, 'inference': 31.13748316179242,
'loss': 0.0007570835581996985, 'postprocess': 5.107450903507702}
task: 'detect'

```

```
[27]: img_properties
```

```
[27]: {'width': 640, 'height': 640, 'channels': 3, 'dtype': dtype('uint8')}
```

```
[28]: model.export(
    format = 'openvino',
    imsz = (img_properties['height'], img_properties['width']),
    half = False,
    int8 = False,
    simplify = False,
    nms = False,
)
```

Ultralytics YOLOv8.1.10 Python-3.10.12 torch-2.1.0+cu121 CUDA:0 (Tesla T4, 15102MiB)

Model summary (fused): 268 layers, 68133198 parameters, 0 gradients, 257.4 GFLOPs

PyTorch: starting from

'runs/detect/yolov8x_ppe_css_50_epochs/weights/best.pt' with input shape (1, 3, 640, 640) BCHW and output shape(s) (1, 14, 8400) (130.4 MB)

```

requirements: Ultralytics requirement ['onnx>=1.12.0'] not found,
attempting AutoUpdate...
Collecting onnx>=1.12.0
  Downloading
onnx-1.15.0-cp310-cp310-manylinux_2_17_x86_64.manylinux2014_x86_64.whl (15.7 MB)
      15.7/15.7 MB 71.8 MB/s eta 0:00:00
Requirement already satisfied: numpy in /usr/local/lib/python3.10/dist-packages
(from onnx>=1.12.0) (1.23.5)
Requirement already satisfied: protobuf>=3.20.2 in
/usr/local/lib/python3.10/dist-packages (from onnx>=1.12.0) (3.20.3)
Installing collected packages: onnx
Successfully installed onnx-1.15.0

requirements: AutoUpdate success   9.9s, installed 1 package:
['onnx>=1.12.0']
requirements: Restart runtime or rerun command for updates
to take effect

```

```

ONNX: starting export with onnx 1.15.0 opset 17...
ONNX: export success   15.6s, saved as
'runs/detect/yolov8x_ppe_css_50_epochs/weights/best.onnx' (260.1 MB)
requirements: Ultralytics requirement ['openvino-dev>=2023.0'] not
found, attempting AutoUpdate...
Collecting openvino-dev>=2023.0
  Downloading openvino_dev-2023.3.0-13775-py3-none-any.whl (5.9 MB)
      5.9/5.9 MB 25.3 MB/s eta 0:00:00
Collecting addict>=2.4.0 (from openvino-dev>=2023.0)
  Downloading addict-2.4.0-py3-none-any.whl (3.8 kB)
Requirement already satisfied: defusedxml>=0.7.1 in
/usr/local/lib/python3.10/dist-packages (from openvino-dev>=2023.0) (0.7.1)
Collecting jstyleson>=0.0.2 (from openvino-dev>=2023.0)
  Downloading jstyleson-0.0.2.tar.gz (2.0 kB)
  Preparing metadata (setup.py): started
  Preparing metadata (setup.py): finished with status 'done'
Collecting networkx<=3.1.0 (from openvino-dev>=2023.0)
  Downloading networkx-3.1-py3-none-any.whl (2.1 MB)
      2.1/2.1 MB 64.0 MB/s eta 0:00:00
Requirement already satisfied: numpy>=1.16.6 in /usr/local/lib/python3.10/dist-
packages (from openvino-dev>=2023.0) (1.23.5)
Requirement already satisfied: opencv-python in /usr/local/lib/python3.10/dist-
packages (from openvino-dev>=2023.0) (4.8.0.76)
Collecting openvino-telemetry>=2022.1.0 (from openvino-dev>=2023.0)
  Downloading openvino_telemetry-2023.2.1-py3-none-any.whl (23 kB)
Requirement already satisfied: pillow>=8.1.2 in /usr/local/lib/python3.10/dist-
packages (from openvino-dev>=2023.0) (9.4.0)
Requirement already satisfied: pyyaml>=5.4.1 in /usr/local/lib/python3.10/dist-

```

```

packages (from openvino-dev>=2023.0) (6.0.1)
Requirement already satisfied: requests>=2.25.1 in
/usr/local/lib/python3.10/dist-packages (from openvino-dev>=2023.0) (2.31.0)
Requirement already satisfied: scipy>=1.8 in /usr/local/lib/python3.10/dist-
packages (from openvino-dev>=2023.0) (1.11.4)
Collecting texttable>=1.6.3 (from openvino-dev>=2023.0)
  Downloading texttable-1.7.0-py2.py3-none-any.whl (10 kB)
Requirement already satisfied: tqdm>=4.54.1 in /usr/local/lib/python3.10/dist-
packages (from openvino-dev>=2023.0) (4.66.1)
Collecting openvino==2023.3.0 (from openvino-dev>=2023.0)
  Downloading openvino-2023.3.0-13775-cp310-cp310-manylinux2014_x86_64.whl (38.3
MB)
                                     38.3/38.3 MB 187.8 MB/s eta
0:00:00
Requirement already satisfied: charset-normalizer<4,>=2 in
/usr/local/lib/python3.10/dist-packages (from requests>=2.25.1->openvino-
dev>=2023.0) (3.3.2)
Requirement already satisfied: idna<4,>=2.5 in /usr/local/lib/python3.10/dist-
packages (from requests>=2.25.1->openvino-dev>=2023.0) (3.6)
Requirement already satisfied: urllib3<3,>=1.21.1 in
/usr/local/lib/python3.10/dist-packages (from requests>=2.25.1->openvino-
dev>=2023.0) (2.0.7)
Requirement already satisfied: certifi>=2017.4.17 in
/usr/local/lib/python3.10/dist-packages (from requests>=2.25.1->openvino-
dev>=2023.0) (2024.2.2)
Building wheels for collected packages: jstyleson
  Building wheel for jstyleson (setup.py): started
  Building wheel for jstyleson (setup.py): finished with status 'done'
  Created wheel for jstyleson: filename=jstyleson-0.0.2-py3-none-any.whl
size=2385
sha256=4e03762859430a6cd617895ab1bb08a0b60fbdd4ac750a4ac8319fa44cbda335
  Stored in directory: /tmp/pip-ephem-wheel-cache-v84x3x33/wheels/12/51/c6/a1e75
1db88203e11c6d9ffe4683ca3d8c14b1479639bec1006
Successfully built jstyleson
Installing collected packages: texttable, openvino-telemetry, jstyleson, addict,
openvino, networkx, openvino-dev
  Attempting uninstall: networkx
    Found existing installation: networkx 3.2.1
    Uninstalling networkx-3.2.1:
      Successfully uninstalled networkx-3.2.1
Successfully installed addict-2.4.0 jstyleson-0.0.2 networkx-3.1
openvino-2023.3.0 openvino-dev-2023.3.0 openvino-telemetry-2023.2.1
texttable-1.7.0

requirements: AutoUpdate success 17.8s, installed 1 package:
['openvino-dev>=2023.0']

```

requirements: Restart runtime or rerun command for updates
to take effect

OpenVINO: starting export with openvino
2023.3.0-13775-ceeafaf64f3-releases/2023/3...
OpenVINO: export success 20.1s, saved as
'runs/detect/yolov8x_ppe_css_50_epochs/weights/best_openvino_model/' (260.3 MB)

Export complete (36.8s)
Results saved to /content/runs/detect/yolov8x_ppe_css_50_epochs/weights
Predict: yolo predict task=detect
model=runs/detect/yolov8x_ppe_css_50_epochs/weights/best_openvino_model
imgsz=640
Validate: yolo val task=detect
model=runs/detect/yolov8x_ppe_css_50_epochs/weights/best_openvino_model
imgsz=640 data=./data.yaml
Visualize: <https://netron.app>

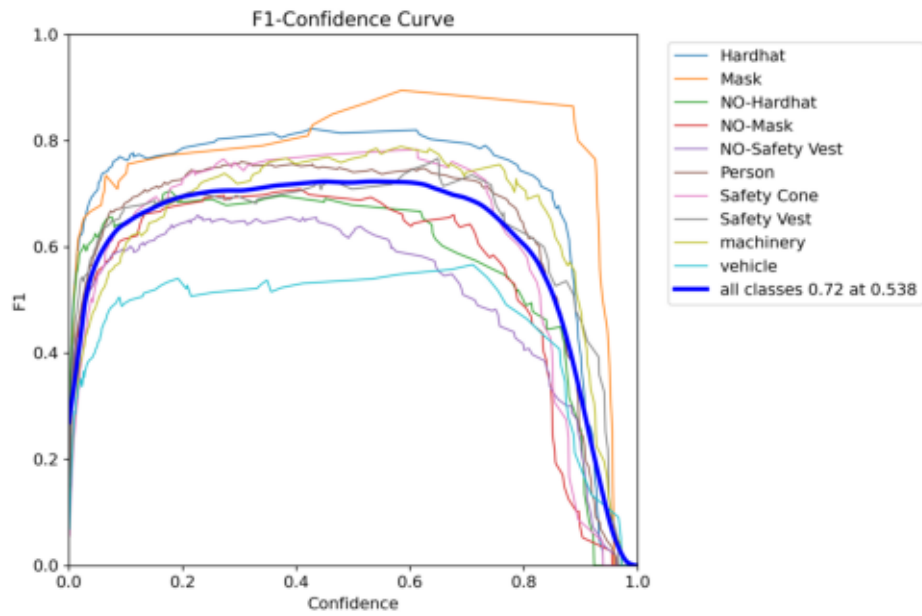
[28]: 'runs/detect/yolov8x_ppe_css_50_epochs/weights/best_openvino_model'

```
[29]: results_paths = [  
    i for i in  
        glob.glob(f'{CFG.OUTPUT_DIR}runs/detect/{CFG.BASE_MODEL}_{CFG.EXP_NAME}/*.  
↪png') +  
        glob.glob(f'{CFG.OUTPUT_DIR}runs/detect/{CFG.BASE_MODEL}_{CFG.EXP_NAME}/*.  
↪jpg')  
    if 'batch' not in i  
]  
  
results_paths
```

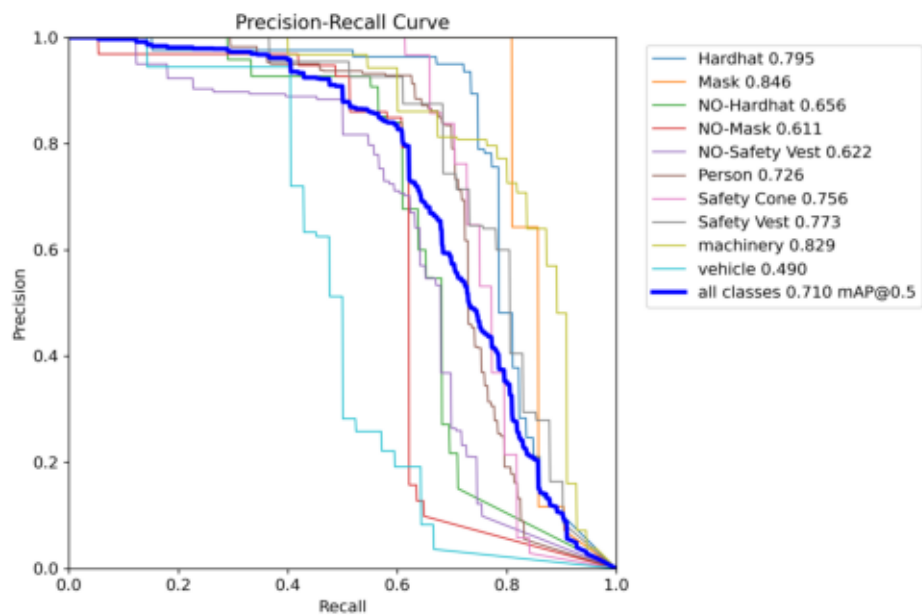
[29]: ['./runs/detect/yolov8x_ppe_css_50_epochs/confusion_matrix_normalized.png',
'./runs/detect/yolov8x_ppe_css_50_epochs/results.png',
'./runs/detect/yolov8x_ppe_css_50_epochs/confusion_matrix.png',
'./runs/detect/yolov8x_ppe_css_50_epochs/R_curve.png',
'./runs/detect/yolov8x_ppe_css_50_epochs/F1_curve.png',
'./runs/detect/yolov8x_ppe_css_50_epochs/P_curve.png',
'./runs/detect/yolov8x_ppe_css_50_epochs/PR_curve.png',
'./runs/detect/yolov8x_ppe_css_50_epochs/labels.jpg',
'./runs/detect/yolov8x_ppe_css_50_epochs/labels_correlogram.jpg']

```
[30]: for file in sorted(results_paths):  
    print(file)  
    display_image(file, print_info = False, hide_axis = True)  
    print('\n')
```

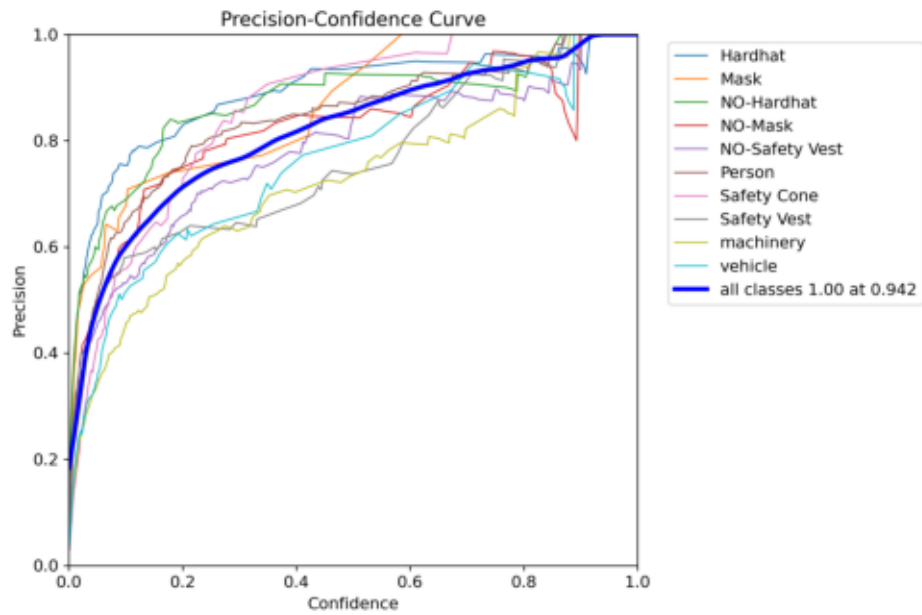

./runs/detect/yolov8x_ppe_css_50_epochs/F1_curve.png



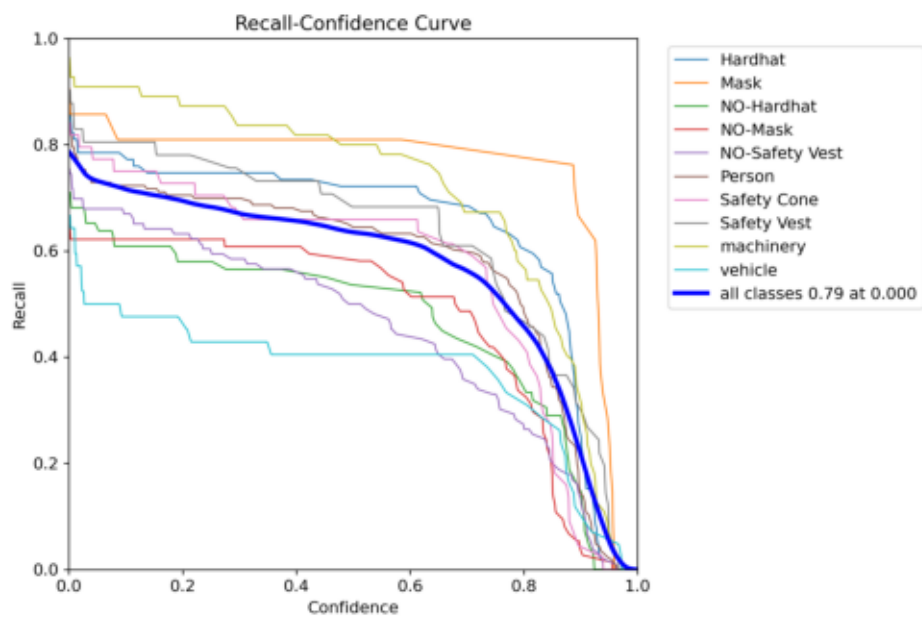
./runs/detect/yolov8x_ppe_css_50_epochs/PR_curve.png



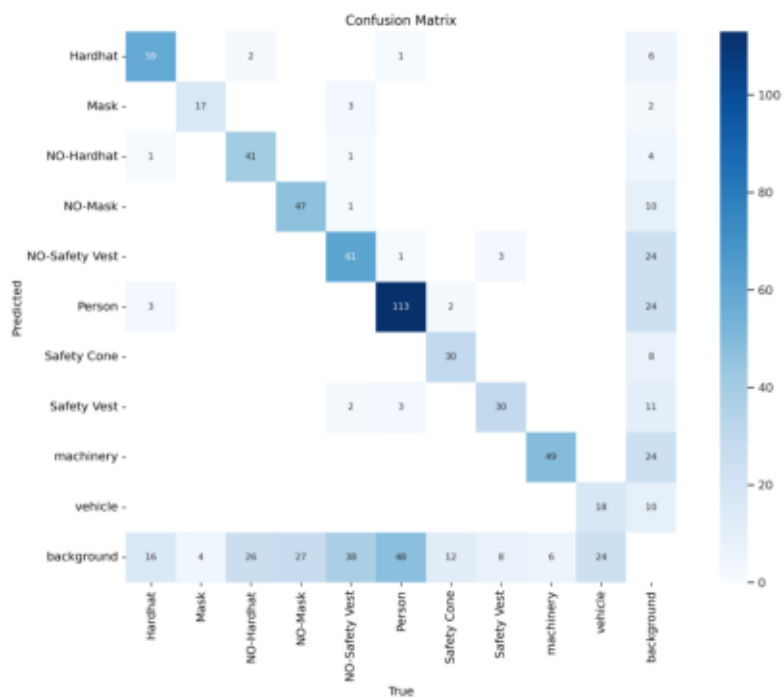
`./runs/detect/yolov8x_ppe_css_50_epochs/P_curve.png`



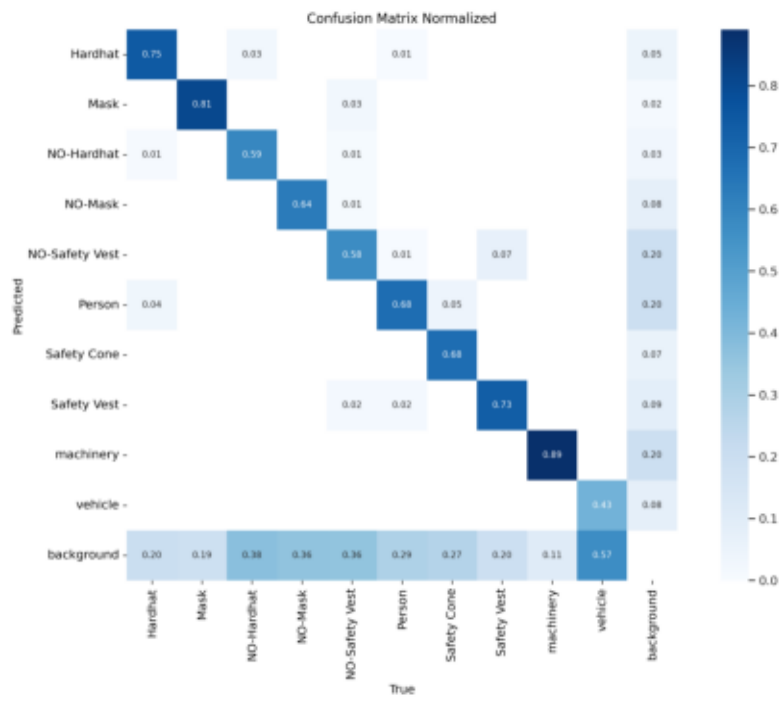
`./runs/detect/yolov8x_ppe_css_50_epochs/R_curve.png`



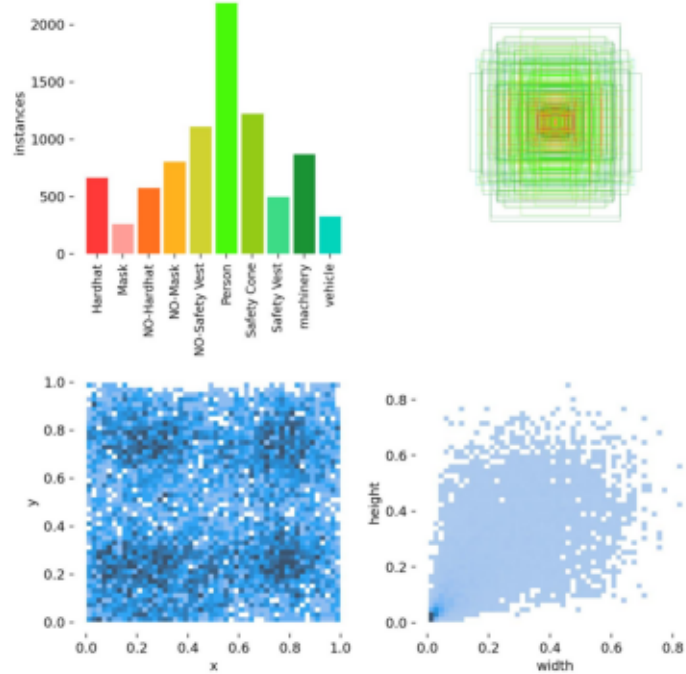
./runs/detect/yolov8x_ppe_css_50_epochs/confusion_matrix.png



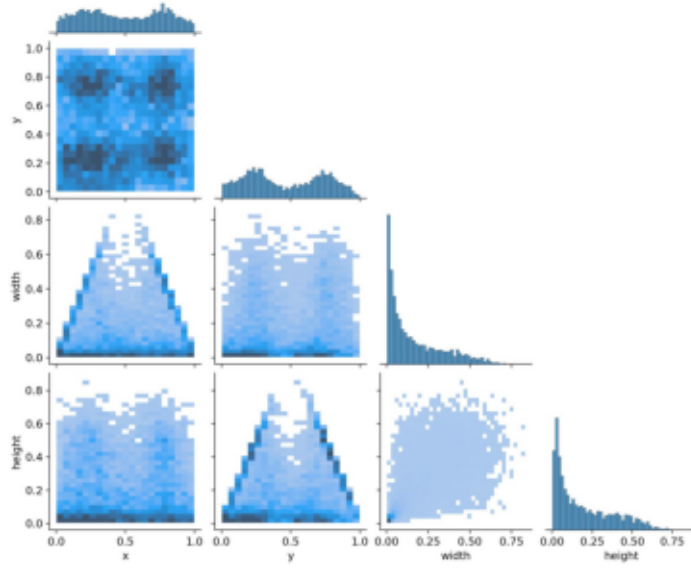
./runs/detect/yolov8x_ppe_css_50_epochs/confusion_matrix_normalized.png



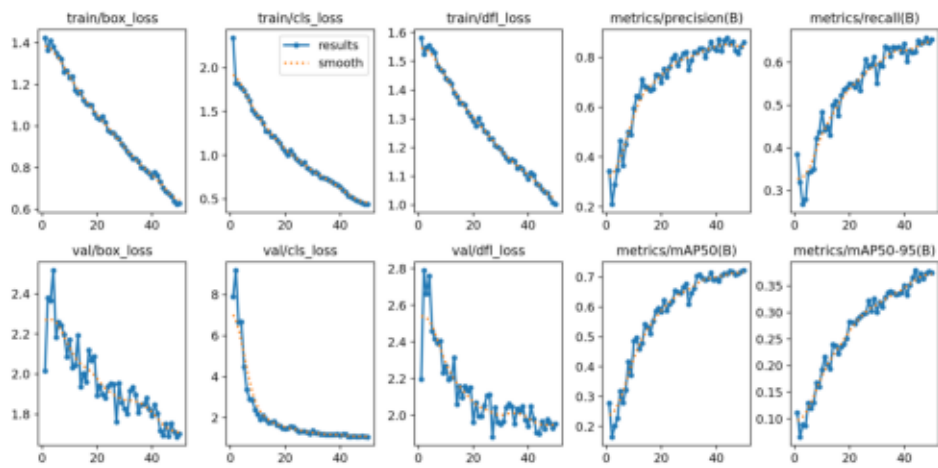
./runs/detect/yolov8x_ppe_css_50_epochs/labels.jpg



./runs/detect/yolov8x_ppe_css_50_epochs/labels_correlogram.jpg



./runs/detect/yolov8x_ppe_css_50_epochs/results.png



```
[31]: df = pd.read_csv(f'{CFG.OUTPUT_DIR}runs/detect/{CFG.BASE_MODEL}_{CFG.EXP_NAME}/
↳results.csv')
df = df.rename(columns=lambda x: x.replace(" ", ""))
df.to_csv(f'{CFG.OUTPUT_DIR}training_log_df.csv', index=False)
df
```

```
[31]:
```

	epoch	train/box_loss	train/cls_loss	train/df1_loss	\
0	1	1.42400	2.34370	1.5817	
1	2	1.36470	1.81960	1.5241	
2	3	1.41020	1.81300	1.5476	
3	4	1.38060	1.77140	1.5550	
4	5	1.35160	1.73770	1.5392	
5	6	1.33390	1.68040	1.5304	
6	7	1.32120	1.62450	1.4833	
7	8	1.25940	1.51610	1.4722	
8	9	1.26460	1.47450	1.4647	
9	10	1.23200	1.44220	1.4398	
10	11	1.23800	1.42610	1.4324	
11	12	1.17300	1.36560	1.4214	
12	13	1.15670	1.27530	1.3893	
13	14	1.16540	1.27080	1.3776	
14	15	1.12350	1.21700	1.3537	
15	16	1.10720	1.21620	1.3542	
16	17	1.10230	1.17200	1.3475	
17	18	1.09990	1.13750	1.3228	
18	19	1.06050	1.08460	1.3091	
19	20	1.04090	1.04270	1.2930	
20	21	1.03310	1.00120	1.2746	
21	22	1.04530	1.05370	1.3029	
22	23	1.01690	1.00320	1.2793	
23	24	0.97900	0.95822	1.2559	
24	25	0.96837	0.93330	1.2512	
25	26	0.96627	0.90041	1.2301	
26	27	0.94603	0.91813	1.2306	
27	28	0.93807	0.85403	1.2066	
28	29	0.91320	0.83754	1.2009	
29	30	0.89889	0.79952	1.1938	
30	31	0.87769	0.80790	1.1795	
31	32	0.86164	0.78464	1.1641	
32	33	0.84266	0.74699	1.1526	
33	34	0.84270	0.73874	1.1582	
34	35	0.82991	0.73224	1.1524	
35	36	0.80098	0.71440	1.1275	
36	37	0.79786	0.69670	1.1321	
37	38	0.78406	0.67392	1.1242	
38	39	0.77333	0.65614	1.1075	
39	40	0.75435	0.63694	1.0897	

40	41	0.77842	0.60114	1.1121
41	42	0.76274	0.57545	1.1038
42	43	0.73110	0.53361	1.0741
43	44	0.70358	0.52125	1.0678
44	45	0.68521	0.50217	1.0535
45	46	0.67569	0.47905	1.0440
46	47	0.66302	0.47008	1.0412
47	48	0.63831	0.45417	1.0246
48	49	0.62432	0.43957	1.0081
49	50	0.62814	0.43851	1.0025

	metrics/precision(B)	metrics/recall(B)	metrics/mAP50(B)	\
0	0.34346	0.38447	0.27705	
1	0.20989	0.31966	0.16448	
2	0.28798	0.26784	0.20156	
3	0.34736	0.27936	0.22656	
4	0.46445	0.34183	0.31786	
5	0.36602	0.34484	0.27855	
6	0.45069	0.35016	0.32186	
7	0.50019	0.42169	0.41671	
8	0.48880	0.43661	0.37168	
9	0.59518	0.48381	0.48595	
10	0.64430	0.44216	0.49623	
11	0.63793	0.44918	0.45970	
12	0.71179	0.42954	0.47844	
13	0.68305	0.49966	0.54233	
14	0.67757	0.50968	0.53143	
15	0.66626	0.47546	0.51072	
16	0.67309	0.52305	0.55288	
17	0.72820	0.53613	0.58613	
18	0.72673	0.54115	0.59597	
19	0.69929	0.54993	0.58344	
20	0.75260	0.54956	0.62166	
21	0.72033	0.54195	0.58699	
22	0.76111	0.55689	0.60912	
23	0.79446	0.53433	0.61972	
24	0.80913	0.57311	0.65484	
25	0.76696	0.60738	0.64854	
26	0.79508	0.58912	0.64692	
27	0.81486	0.59363	0.66584	
28	0.82096	0.61284	0.67684	
29	0.74940	0.55096	0.60839	
30	0.78770	0.59616	0.64679	
31	0.82005	0.59079	0.66296	
32	0.81890	0.63687	0.70358	
33	0.83464	0.63322	0.70826	
34	0.83498	0.61465	0.69521	

35	0.79885	0.63607	0.68989
36	0.82927	0.63422	0.69483
37	0.86230	0.63599	0.71450
38	0.83758	0.62425	0.69080
39	0.82789	0.64347	0.69454
40	0.87101	0.60212	0.68657
41	0.82541	0.62776	0.70464
42	0.86710	0.62254	0.71184
43	0.87864	0.62515	0.71044
44	0.85464	0.65119	0.72022
45	0.86426	0.64346	0.71741
46	0.82607	0.64759	0.70844
47	0.81603	0.65785	0.71145
48	0.84319	0.64606	0.71787
49	0.85958	0.65402	0.72241

	metrics/mAP50-95(B)	val/box_loss	val/cls_loss	val/df1_loss	lr/pg0 \
0	0.11102	2.0152	7.8957	2.1976	0.000231
1	0.06580	2.3786	9.1811	2.7896	0.000460
2	0.08897	2.3663	6.6512	2.6620	0.000679
3	0.08665	2.5165	6.6706	2.7594	0.000672
4	0.12914	2.1826	4.4810	2.4594	0.000672
5	0.11958	2.2584	3.3641	2.4151	0.000657
6	0.12982	2.2415	2.9022	2.3924	0.000643
7	0.16760	2.1972	2.8539	2.4043	0.000629
8	0.15997	2.0860	2.3836	2.2318	0.000615
9	0.19240	2.1700	2.1563	2.2686	0.000601
10	0.21630	2.0320	1.9025	2.1934	0.000587
11	0.20192	2.0441	2.1058	2.2045	0.000573
12	0.19368	2.1934	1.9139	2.3128	0.000558
13	0.23996	1.9361	1.7618	2.0598	0.000544
14	0.23714	2.0019	1.7653	2.1584	0.000530
15	0.22223	1.9603	1.8115	2.0970	0.000516
16	0.23509	2.1195	1.6430	2.1570	0.000502
17	0.24035	2.0681	1.5976	2.1344	0.000488
18	0.25075	2.0870	1.4815	2.1505	0.000474
19	0.28278	1.8934	1.4729	1.9615	0.000460
20	0.28106	1.9429	1.4731	2.0685	0.000445
21	0.27855	1.8989	1.5589	1.9945	0.000431
22	0.28680	1.8814	1.5362	1.9947	0.000417
23	0.29233	1.9387	1.4280	2.0528	0.000403
24	0.29649	1.9527	1.3053	2.1077	0.000389
25	0.29889	1.9489	1.3341	2.1112	0.000375
26	0.32235	1.7604	1.3252	1.8822	0.000361
27	0.30297	1.9539	1.2862	2.0418	0.000346
28	0.32603	1.8568	1.2129	1.9612	0.000332
29	0.30017	1.8268	1.3727	1.9500	0.000318

30	0.31928	1.8006	1.2481	1.9673	0.000304
31	0.31052	1.9173	1.2434	2.0422	0.000290
32	0.32806	1.9349	1.1861	2.0615	0.000276
33	0.33689	1.8971	1.1872	2.0469	0.000262
34	0.33942	1.8075	1.1726	1.9537	0.000247
35	0.33694	1.8470	1.1727	2.0363	0.000233
36	0.33368	1.8493	1.1799	2.0197	0.000219
37	0.33525	1.8810	1.1647	2.0488	0.000205
38	0.33701	1.8240	1.1879	1.9742	0.000191
39	0.35075	1.7913	1.1267	1.9412	0.000177
40	0.33280	1.8476	1.1831	2.0485	0.000163
41	0.35148	1.8027	1.1901	1.9787	0.000149
42	0.36540	1.7167	1.0667	1.9064	0.000134
43	0.37921	1.6930	1.0782	1.8980	0.000120
44	0.36002	1.7508	1.1016	1.9657	0.000106
45	0.37350	1.6908	1.0787	1.9249	0.000092
46	0.36381	1.7540	1.0839	1.9772	0.000078
47	0.37355	1.7117	1.0866	1.9578	0.000064
48	0.37677	1.6865	1.0742	1.9298	0.000050
49	0.37462	1.7014	1.0614	1.9525	0.000035

	lr/pg1	lr/pg2
0	0.000231	0.000231
1	0.000460	0.000460
2	0.000679	0.000679
3	0.000672	0.000672
4	0.000672	0.000672
5	0.000657	0.000657
6	0.000643	0.000643
7	0.000629	0.000629
8	0.000615	0.000615
9	0.000601	0.000601
10	0.000587	0.000587
11	0.000573	0.000573
12	0.000558	0.000558
13	0.000544	0.000544
14	0.000530	0.000530
15	0.000516	0.000516
16	0.000502	0.000502
17	0.000488	0.000488
18	0.000474	0.000474
19	0.000460	0.000460
20	0.000445	0.000445
21	0.000431	0.000431
22	0.000417	0.000417
23	0.000403	0.000403
24	0.000389	0.000389

```

25 0.000375 0.000375
26 0.000361 0.000361
27 0.000346 0.000346
28 0.000332 0.000332
29 0.000318 0.000318
30 0.000304 0.000304
31 0.000290 0.000290
32 0.000276 0.000276
33 0.000262 0.000262
34 0.000247 0.000247
35 0.000233 0.000233
36 0.000219 0.000219
37 0.000205 0.000205
38 0.000191 0.000191
39 0.000177 0.000177
40 0.000163 0.000163
41 0.000149 0.000149
42 0.000134 0.000134
43 0.000120 0.000120
44 0.000106 0.000106
45 0.000092 0.000092
46 0.000078 0.000078
47 0.000064 0.000064
48 0.000050 0.000050
49 0.000035 0.000035

```

```

[32]: print('='*50)
print('\nBest Training Box loss: ', df['train/box_loss'].min(), ', on epoch: ',
      df['train/box_loss'].argmin() + 1, '\n')
print('\nBest Validation Box loss: ', df['val/box_loss'].min(), ', on epoch: ',
      df['val/box_loss'].argmin() + 1, '\n')

print('='*50)
print('\nBest Training Cls loss: ', df['train/cls_loss'].min(), ', on epoch: ',
      df['train/box_loss'].argmin() + 1, '\n')
print('\nBest Validation Cls loss: ', df['val/cls_loss'].min(), ', on epoch: ',
      df['val/box_loss'].argmin() + 1, '\n')

print('='*50)
print('\nBest Training DFL loss: ', df['train/df1_loss'].min(), ', on epoch: ',
      df['train/box_loss'].argmin() + 1, '\n')
print('\nBest Validation DFL loss: ', df['val/df1_loss'].min(), ', on epoch: ',
      df['val/box_loss'].argmin() + 1, '\n')

```

Best Training Box loss: 0.62432 , on epoch: 49

Best Validation Box loss: 1.6865 , on epoch: 49

=====

Best Training Cls loss: 0.43851 , on epoch: 49

Best Validation Cls loss: 1.0614 , on epoch: 49

=====

Best Training DFL loss: 1.0025 , on epoch: 49

Best Validation DFL loss: 1.8822 , on epoch: 49

```
[33]: fig, (ax1, ax2, ax3) = plt.subplots(3, 1, figsize=(10, 15), sharex=True)
```

```
### Training and Validation Box Loss
```

```
ax1.set_title('Box Loss')
```

```
ax1.plot(df['epoch'], df['train/box_loss'], label='Training box_loss',  
        ↪marker='o', linestyle='-')
```

```
ax1.plot(df['epoch'], df['val/box_loss'], label='Validation box_loss',  
        ↪marker='o', linestyle='-')
```

```
ax1.set_ylabel('Box Loss')
```

```
ax1.legend()
```

```
ax1.grid(True)
```

```
### Training and Validation cls_loss
```

```
ax2.set_title('Cls Loss')
```

```
ax2.plot(df['epoch'], df['train/cls_loss'], label='Training cls_loss',  
        ↪marker='o', linestyle='-')
```

```
ax2.plot(df['epoch'], df['val/cls_loss'], label='Validation cls_loss',  
        ↪marker='o', linestyle='-')
```

```
ax2.set_ylabel('cls_loss')
```

```
ax2.legend()
```

```
ax2.grid(True)
```

```
### Training and Validation dfl_loss
```

```
ax3.set_title('DFL Loss')
```

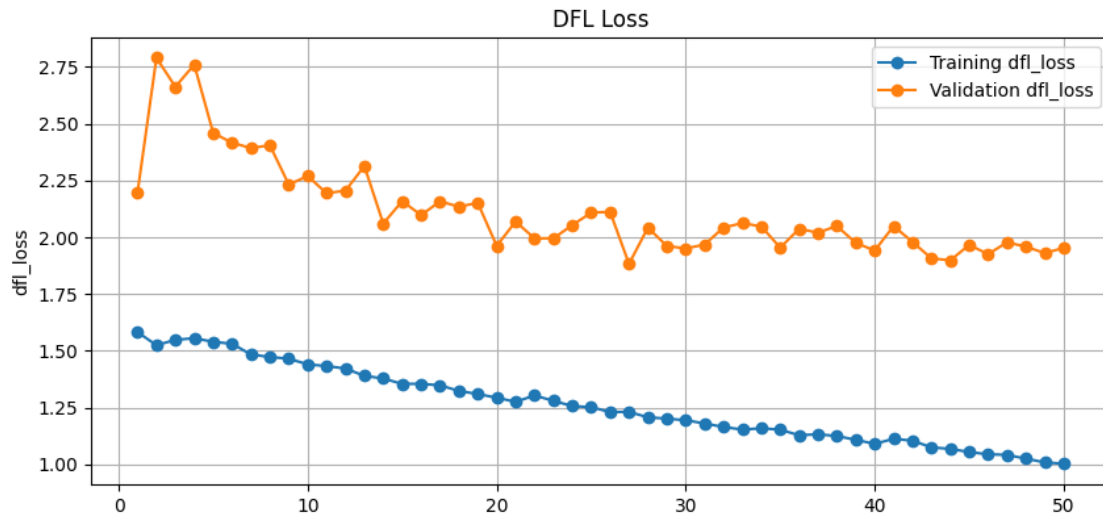
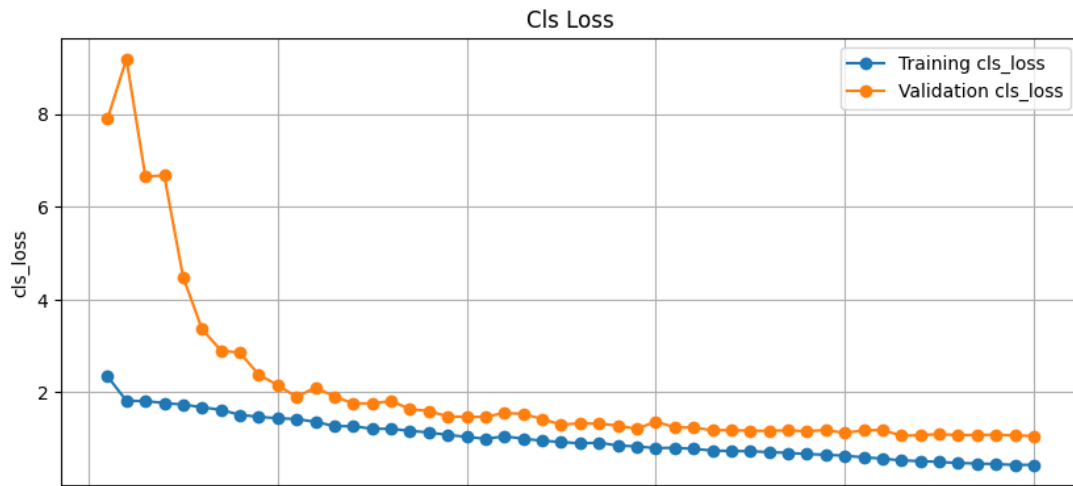
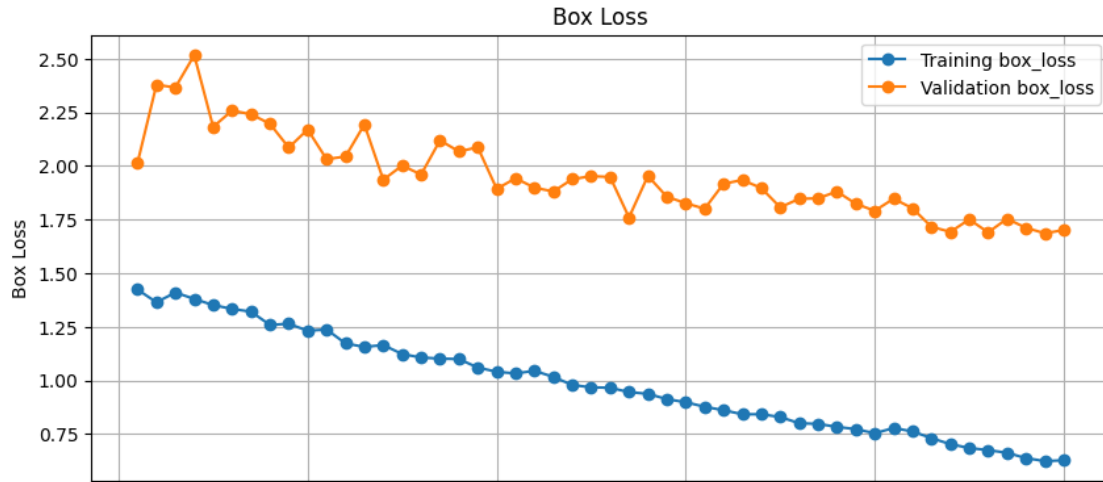
```
ax3.plot(df['epoch'], df['train/dfl_loss'], label='Training dfl_loss',  
        ↪marker='o', linestyle='-')
```

```
ax3.plot(df['epoch'], df['val/dfl_loss'], label='Validation dfl_loss',  
        ↪marker='o', linestyle='-')
```

```
ax3.set_xlabel('Epochs')
ax3.set_ylabel('dfl_loss')
ax3.legend()
ax3.grid(True)

plt.suptitle('Training Metrics vs. Epochs')
plt.show()
```

Training Metrics vs. Epochs



```
[34]: validation_results_paths = [
    i for i in
        glob.glob(f'{CFG.OUTPUT_DIR}runs/detect/{CFG.BASE_MODEL}_{CFG.EXP_NAME}/*.
        ↳png') +
        glob.glob(f'{CFG.OUTPUT_DIR}runs/detect/{CFG.BASE_MODEL}_{CFG.EXP_NAME}/*.
        ↳jpg')
    if 'val_batch' in i
]

len(validation_results_paths)
```

[34]: 6

```
[35]: if len(validation_results_paths) >= 1:
    print(validation_results_paths[-1])
```

./runs/detect/yolov8x_ppe_css_50_epochs/val_batch0_labels.jpg

```
[36]: ### check predictions or labels from a random validation batch
if len(validation_results_paths) >= 1:
    val_img_path = random.choice(validation_results_paths)
    print(val_img_path)
    display_image(val_img_path, print_info = False, hide_axis = True)
```

./runs/detect/yolov8x_ppe_css_50_epochs/val_batch2_labels.jpg

