

Introduction

The project focuses on fine-tuning YOLOv8, a state-of-the-art object detection model, for the specific task of detecting potholes in road images. This is accomplished using a custom dataset derived from the RDD2022 dataset, which is specialized for road damage detection.

Dataset Description

- **Training and Validation Sets:** The final dataset, after several annotation corrections, includes 6,962 training images and 271 validation images. These images are crucial for training and evaluating the model's performance.
- **Data Preparation:** The dataset was unzipped and prepared for training. A significant step involved converting bounding box annotations from YOLO format to a standard format (xmin, ymin, xmax, ymax). The image dimensions were also used to denormalize the bounding box coordinates, a key step in data preprocessing.

Training Setup

- **Dataset YAML:** To train the model, a YAML file defining the paths to the images and the class names was necessary. This file serves as a guide for the model to understand the dataset structure and the classes involved.
- **Model Architecture:** YOLOv8 was chosen for its efficiency and accuracy in object detection tasks. Its ability to detect small and varied objects like potholes makes it a suitable choice for this application.



As there was no GPU available on the HPC, I ran the notebook on the runpod -
<https://www.runpod.io/console/gpu-secure-cloud>

NVIDIA GeForce RTX 4080, 16079MiB

Ran only 5 Epochs to save the money on runpod charges per hour.

```

# Sample training for 5 epoch.
EPOCHS = 5
yolo task=detect mode=train model=yolov8n.pt imgs=1280 data=pothole_v8.yaml epochs=(EPOCHS) batch=8 name=yolov8n_v8_50e

Downloading https://github.com/ultralytics/assets/releases/download/v0.3.0/yolov8n.pt to 'yolov8n.pt'...
100%|██████████| 755k/755k [100:00:00:00, 18.1MB/s]
100%|██████████| 755k/755k [100:00:00:00, 18.1MB/s]
Overriding model.yaml nc=88 with nc=1

      from n   params   module           arguments
0      -1 1     464  ultralytics.nn.modules.conv.Conv  [3, 16, 3, 2]
1      -1 1     4672 ultralytics.nn.modules.conv.Conv [16, 32, 3, 2]
2      -1 1     7368 ultralytics.nn.modules.block.C2f [32, 32, 1, True]
3      -1 1    18560 ultralytics.nn.modules.conv.Conv [32, 64, 3, 2]
4      -1 2    49664 ultralytics.nn.modules.block.C2f [64, 64, 2, True]
5      -1 1    73952 ultralytics.nn.modules.conv.Conv [64, 128, 3, 2]
6      -1 2    106328 ultralytics.nn.modules.block.C2f [128, 256, 2, True]
7      -1 1    205424 ultralytics.nn.modules.conv.Conv [128, 256, 1, 2]
8      -1 1    460288 ultralytics.nn.modules.block.C2f [256, 256, 1, True]
9      -1 1    164688 ultralytics.nn.modules.block.SPPF [256, 256, 5]
10     -1 1      0 torch.nn.modules.upsampling.Upsample [None, 2, 'nearest']
11     [-1, 6] 1      0 ultralytics.nn.modules.conv.Concat [1]
12     -1 1    148224 ultralytics.nn.modules.block.C2f [384, 128, 1]
13     -1 1      0 torch.nn.modules.upsampling.Upsample [None, 2, 'nearest']
14     [-1, 4] 1      0 ultralytics.nn.modules.conv.Concat [1]
15     -1 1    37248 ultralytics.nn.modules.block.C2f [192, 64, 1]
...
all      271    716  0.484   0.328   0.342   0.149
Speed: 0.3ms preprocess, 1.3ms inference, 0.0ms loss, 3.6ms postprocess per image
Results saved to runs/detect/yolov8n_v8_50e
💡 Learn more at https://docs.ultralytics.com/modes/train
Output is truncated. View as a scrollable element or open in a text editor. Adjust cell output settings.

```

Hyperparameter Choices

- It uses albumentation library for augmentations
- By default, it uses ADAM optimizer.

```
optimizer: AdamW(lr=0.002, momentum=0.9) with parameter groups 57 weight(decay=0.0), 64 weight(decay=0.0005), 63 bias(decay=0.0)
```

Results and Performance

- Training result:

Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
1/5	4.79G	2.153	4.056	1.814	9	1280: 1
	Class	Images	Instances	Box(P)	R	mAP50 m
	all	271	716	0.288	0.207	0.147 0.0512
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
2/5	4.42G	2.124	2.541	1.822	10	1280: 1
	Class	Images	Instances	Box(P)	R	mAP50 m
	all	271	716	0.359	0.203	0.188 0.08
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
3/5	4.47G	2.085	2.345	1.789	2	1280: 1
	Class	Images	Instances	Box(P)	R	mAP50 m
	all	271	716	0.474	0.258	0.243 0.103
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
4/5	4.44G	1.997	2.172	1.723	4	1280: 1
	Class	Images	Instances	Box(P)	R	mAP50 m
	all	271	716	0.482	0.328	0.341 0.149
Epoch	GPU_mem	box_loss	cls_loss	dfl_loss	Instances	Size
5/5	4.44G	1.964	2.068	1.688	8	1280: 1
	Class	Images	Instances	Box(P)	R	mAP50 m
	all	271	716	0.43	0.304	0.29 0.126

- Evaluation on Validation Set and Visualizations:

```
Validating runs/detect/yolov8n_v8_50e/weights/best.pt...
Ultralytics YOLOv8.0.208 🚀 Python-3.10.12 torch-2.1.0+cu118 CUDA:0 (NVIDIA GeForce RTX 4080, 16079MiB)
Model summary (fused): 168 layers, 3005843 parameters, 0 gradients, 8.1 GFLOPs
  Class   Images  Instances   Box(P      R      mAP50  m
    all      271       716      0.484     0.328     0.342     0.149
```



COCO evaluation:

Ground_truth_coco_format.json file contains ground truth annotation result and Predictions_coco_format.json file contains predictions in coco format.

I have used the below code but the result was giving the error while comparing the ground truth annotation json with prediction json

```

▶ from tqdm import tqdm
from sahi.utils.coco import Coco, CocoCategory, CocoImage, CocoAnnotation
from sahi.utils.file import save_json

# Define your class dictionary depending on how many classes you have.
# Here, I'm assuming '0' is for potholes. Adjust as per your dataset.
class_dic = {0: "pothole"}

def yolo_to_coco(x_center, y_center, w, h, image_w, image_h):
    # Convert YOLO format to COCO format
    w = w * image_w
    h = h * image_h
    x1 = ((2 * x_center * image_w) - w) / 2
    y1 = ((2 * y_center * image_h) - h) / 2
    return [x1, y1, w, h]

def convert_to_COCO(image_path, json_name):
    coco = Coco()
    for key, item in class_dic.items():
        coco.add_category(CocoCategory(id=key, name=item))

    # Adjust the glob pattern as per your folder structure
    image_path_list = list(Path(image_path).glob("*.jpg"))
    label_path_list = [Path(str(p).replace('images', 'labels')).replace('.jpg', '.txt') for p in image_path_list]

    print(f'The number of images: {len(image_path_list)}')
    print(f'The number of labels: {len(label_path_list)}')

    no_annotation = 0
    for img_path, l_path in tqdm(zip(image_path_list, label_path_list), total=len(image_path_list)):
        ImageHeight, ImageWidth = Image.open(img_path).size
        if os.path.isfile(l_path):
            coco_image = CocoImage(file_name=str(img_path.name), height=ImageHeight, width=ImageWidth)
            with open(str(l_path)) as f:
                lines = f.readlines()
                for line in lines:
                    line = line.strip()
                    obj_class, x_center, y_center, width_yolo, height_yolo = line.split(' ')
                    bbox = yolo_to_coco(float(x_center), float(y_center), float(width_yolo), float(height_yolo), ImageWidth, ImageHeight)
                    cat_name = class_dic[int(obj_class)]
                    coco_image.add_annotation(CocoAnnotation(bbox=bbox, category_id=int(obj_class), category_name=cat_name))
            coco.add_image(coco_image)
        else:
            no_annotation += 1

    print(f'The number of images that don't have a label: {no_annotation}')
    save_path = json_name
    save_json(data=coco.json, save_path=save_path)
    print('COCO Json File Created.')

# Specify your dataset paths
image_path = '/workspace/datasets/pothole_dataset_v8/valid/images/'
json_name_file = '/workspace/datasets/pothole_dataset_v8/valid/gt3.json'

```

```

▶ # COCO Evaluation Code
from pycocotools.coco import COCO
from pycocotools.cocoeval import COCOeval
import json

# Load ground truth annotations
ground_truth_annotations_path = '/workspace/datasets/pothole_dataset_v8/valid/gt5.json' # Correct path
cocoGt = COCO(ground_truth_annotations_path)

# Load YOLOv8 detections (predictions in COCO format)
detections_path = '/workspace/runs/detect/yolov8n_v8_50e_infer12803/labels/yolov8_predictions.json' # Correct path
cocoDt = cocoGt.loadRes(detections_path)

# Running COCO evaluation
cocoEval = COCOeval(cocoGt, cocoDt, 'bbox')
cocoEval.evaluate()
cocoEval.accumulate()
cocoEval.summarize()

⌚ loading annotations into memory...
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```

JSONDecodeError Traceback (most recent call last)

