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In [ ]: from copy import copy
        import numpy as np
        from ultralytics.data import build_dataloader, build_yolo_dataset
        from ultralytics.engine.trainer import BaseTrainer
        from ultralytics.models import yolo
        from ultralytics.nn.tasks import DetectionModel
        from ultralytics.utils import LOGGER, RANK
        from ultralytics.utils.plotting import plot_images, plot_labels, plot_results
        from ultralytics.utils.torch_utils import de_parallel, torch_distributed_zero_first
        class DetectionTrainer(BaseTrainer):
            A class extending the BaseTrainer class for training based on a detection model.
                  `python
                from ultralytics.models.yolo.detect import DetectionTrainer
                args = dict(model='yolov8n.pt', data='coco8.yaml', epochs=3)
                trainer = DetectionTrainer(overrides=args)
                trainer.train()
            def build_dataset(self, img_path, mode='train', batch=None):
                Build YOLO Dataset.
                Args:
                    img path (str): Path to the folder containing images.
                    mode (str): `train` mode or `val` mode, users are able to customize different augmentations for each mode.
                    batch (int, optional): Size of batches, this is for `rect`. Defaults to None.
                gs = max(int(de_parallel(self.model).stride.max() if self.model else 0), 32)
                return build_yolo_dataset(self.args, img_path, batch, self.data, mode=mode, rect=mode == 'val', stride=gs)
            def get_dataloader(self, dataset_path, batch_size=16, rank=0, mode='train'):
                   "Construct and return dataloader."
                assert mode in ['train', 'val']
                with torch distributed zero first(rank): # init dataset *.cache only once if DDP
                    dataset = self.build_dataset(dataset_path, mode, batch_size)
                shuffle = mode == 'train'
                if getattr(dataset, 'rect', False) and shuffle:
                    LOGGER.warning("WARNING 🛕 'rect=True' is incompatible with DataLoader shuffle, setting shuffle=False")
                     shuffle = False
                workers = self.args.workers if mode == 'train' else self.args.workers * 2
                return build_dataloader(dataset, batch_size, workers, shuffle, rank) # return dataloader
            def preprocess_batch(self, batch):
    """Preprocesses a batch of images by scaling and converting to float."""
                batch['img'] = batch['img'].to(self.device, non_blocking=True).float() / 255
                return batch
            def set model attributes(self):
                """N1 = de_parallel(self.model).model[-1].nl # number of detection layers (to scale hyps)."""
                # self.args.box *= 3 / nl # scale to layers
                \# self.args.cls *= self.data["nc"] / 80 * 3 / nl \# scale to classes and layers
                # self.args.cls *= (self.args.imgsz / 640) ** 2 * 3 / nl # scale to image size and layers
                self.model.nc = self.data['nc'] # attach number of classes to model
                self.model.names = self.data['names'] # attach class names to model
                self.model.args = self.args # attach hyperparameters to model
                # TODO: self.model.class_weights = labels_to_class_weights(dataset.labels, nc).to(device) * nc
            def get_model(self, cfg=None, weights=None, verbose=True):
                  ""Return a YOLO detection model.""
                model = DetectionModel(cfg, nc=self.data['nc'], verbose=verbose and RANK == -1)
                if weights:
                    model.load(weights)
                return model
            def get_validator(self):
    """Returns a DetectionValidator for YOLO model validation."""
                self.loss names = 'box loss', 'cls loss', 'dfl loss'
                return yolo.detect.DetectionValidator(self.test_loader, save_dir=self.save_dir, args=copy(self.args))
            def label_loss_items(self, loss_items=None, prefix='train'):
                Returns a loss dict with labelled training loss items tensor.
                Not needed for classification but necessary for segmentation & detection
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keys = [f'{prefix}/{x}' for x in self.loss_names]
    if loss items is not None:
        loss_items = [round(float(x), 5) for x in loss_items] # convert tensors to 5 decimal place floats
        return dict(zip(keys, loss_items))
    else:
        return kevs
def progress_string(self):
    """Returns a formatted string of training progress with epoch, GPU memory, loss, instances and size."""
    return ('\n' + '%11s' *
            (4 + len(self.loss_names))) % ('Epoch', 'GPU_mem', *self.loss_names, 'Instances', 'Size')
def plot_training_samples(self, batch, ni):
     ""Plots training samples with their annotations."""
    plot_images(images=batch['img'],
                batch idx=batch['batch idx'],
                cls=batch['cls'].squeeze(-1),
                bboxes=batch['bboxes'],
                paths=batch['im_file'],
                fname=self.save_dir / f'train_batch{ni}.jpg',
                on_plot=self.on_plot)
def plot_metrics(self):
     """Plots metrics from a CSV file."""
    plot_results(file=self.csv, on_plot=self.on_plot) # save results.png
def plot_training_labels(self):
      "Create a labeled training plot of the YOLO model."""
    boxes = np.concatenate([lb['bboxes'] for lb in self.train_loader.dataset.labels], 0)
    cls = np.concatenate([lb['cls'] for lb in self.train loader.dataset.labels], 0)
    plot_labels(boxes, cls.squeeze(), names=self.data['names'], save_dir=self.save_dir, on_plot=self.on_plot)
```

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In [1]: from ultralytics import YOLO
        # Load a model
        model = YOLO('yolov8n.yaml') # build a new model from YAML
        model = YOLO('yolov8n.pt') # load a pretrained model (recommended for training)
        model = YOLO('yolov8n.yaml').load('yolov8n.pt') # build from YAML and transfer weights
        # Train the model
        results = model.train(data='C:/Users/muzam/Desktop/Files/IIT/Deep Learning/Final Project/ultralytics-main/data.yaml', epochs=2
                           a11
                                     1392
                                                1518
                                                          0.903
                                                                     0.802
                                                                                0.889
                                                                                           0.606
        25 epochs completed in 7.015 hours.
        Optimizer stripped from runs\detect\train4\weights\last.pt, 6.2MB
        Optimizer stripped from runs\detect\train4\weights\best.pt, 6.2MB
        Validating runs\detect\train4\weights\best.pt...
        Ultralytics YOLOv8.0.221 🚀 Python-3.9.13 torch-1.13.0+cpu CPU (AMD Ryzen 7 4800H with Radeon Graphics)
        YOLOv8n summary (fused): 168 layers, 3006038 parameters, 0 gradients, 8.1 GFLOPs
                         Class
                                                                                mAP50 mAP50-95): 100%| 44/44 [01:22
                                   Images Instances
                                                          Box(P
                                                                         R
                                     1392
                                                          0.902
                                                                                0.889
                           a11
                                                1518
                                                                     0.802
                                                                                           0.606
                        pistol
                                     1392
                                                 669
                                                          0.887
                                                                     0.786
                                                                                0.871
                                                                                           0.685
                                     1392
                                                 849
                                                          0.918
                                                                     0.818
                                                                                0.908
                         Knife
                                                                                           0.527
        Speed: 1.2ms preprocess, 53.1ms inference, 0.0ms loss, 0.4ms postprocess per image
        Results saved to runs\detect\train4
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