Homework 4

Instructions

- This homework focuses on understanding and applying CoCoOp for CLIP prompt tuning. It
 consists of four questions designed to assess both theoretical understanding and
 practical application.
- Please organize your answers and results for the questions below and submit this jupyter notebook as a .pdf file.
- Deadline: 11/26 (Sat) 23:59

Preparation

- Run the code below before proceeding with the homework.
- If an error occurs, click 'Run Session Again' and then restart the runtime from the beginning.

```
!git clone https://github.com/mlvlab/ProMetaR.git
%cd ProMetaR/
!git clone https://github.com/KaiyangZhou/Dassl.pytorch.git
%cd Dassl.pytorch/
# Install dependencies
!pip install -r requirements.txt
!cp -r dassl ../
# Install this library (no need to re-build if the source code is modified)
# !python setup.py develop
%cd ..
!pip install -r requirements.txt
%mkdir outputs
%mkdir data
%cd data
%mkdir eurosat
!wget http://madm.dfki.de/files/sentinel/EuroSAT.zip -O EuroSAT.zip
!unzip -o EuroSAT.zip -d eurosat/
%cd eurosat
!gdown 1Ip7yaCWFi0eaOFUGga0lUdVi_DDQth1o
%cd ../../
import os.path as osp
from collections import OrderedDict
```

```
import math
import torch
import torch.nn as nn
from torch.nn import functional as F
from torch.cuda.amp import GradScaler, autocast
from PIL import Image
import torchvision.transforms as transforms
import torch
from clip import clip
from clip.simple tokenizer import SimpleTokenizer as Tokenizer
import time
from tgdm import tgdm
import datetime
import argparse
from dassl.utils import setup_logger, set_random_seed, collect_env_info
from dassl.config import get cfg default
from dassl.engine import build trainer
from dassl.engine import TRAINER_REGISTRY, TrainerX
from dassl.metrics import compute accuracy
from dassl.utils import load pretrained weights, load checkpoint
from dassl.optim import build optimizer, build lr scheduler
# custom
import datasets.oxford pets
import datasets.oxford flowers
import datasets.fgvc aircraft
import datasets.dtd
import datasets.eurosat
import datasets.stanford cars
import datasets.food101
import datasets.sun397
import datasets.caltech101
import datasets.ucf101
import datasets.imagenet
import datasets.imagenet_sketch
import datasets.imagenetv2
import datasets.imagenet_a
import datasets.imagenet_r
def print_args(args, cfg):
    print("**********")
    print("** Arguments **")
    print("***********")
    optkeys = list(args.__dict__.keys())
    optkeys.sort()
    for key in optkeys:
        print("{}: {}".format(key, args.__dict__[key]))
    print("*********")
    print("** Config **")
    print("********")
    print(cfg)
def reset_cfg(cfg, args):
    if args.root:
        cfg.DATASET.ROOT = args.root
```

```
if args.output_dir:
        cfg.OUTPUT_DIR = args.output_dir
    if args.seed:
        cfg.SEED = args.seed
    if args.trainer:
        cfg.TRAINER.NAME = args.trainer
    cfq.DATASET.NUM SHOTS = 16
    cfg.DATASET.SUBSAMPLE_CLASSES = args.subsample_classes
    cfg.DATALOADER.TRAIN X.BATCH SIZE = args.train batch size
    cfg.OPTIM.MAX EPOCH = args.epoch
def extend_cfg(cfg):
   Add new config variables.
   from yacs.config import CfgNode as CN
   cfg.TRAINER.COOP = CN()
    cfg.TRAINER.COOP.N_CTX = 16 # number of context vectors
   cfg.TRAINER.COOP.CSC = False # class-specific context
   cfg.TRAINER.COOP.CTX INIT = "" # initialization words
   cfg.TRAINER.COOP.PREC = "fp16" # fp16, fp32, amp
    cfg.TRAINER.COOP.CLASS_TOKEN_POSITION = "end" # 'middle' or 'end' or 'front'
    cfq.TRAINER.COCOOP = CN()
    cfg.TRAINER.COCOOP.N_CTX = 4 # number of context vectors
    cfg.TRAINER.COCOOP.CTX_INIT = "a photo of a" # initialization words
    cfg.TRAINER.COCOOP.PREC = "fp16" # fp16, fp32, amp
   cfg.TRAINER.PROMETAR = CN()
   cfg.TRAINER.PROMETAR.N_CTX_VISION = 4 # number of context vectors at the vis
    cfg.TRAINER.PROMETAR.N CTX TEXT = 4 # number of context vectors at the langu
    cfg.TRAINER.PROMETAR.CTX_INIT = "a photo of a" # initialization words
    cfg.TRAINER.PROMETAR.PREC = "fp16" # fp16, fp32, amp
    cfg.TRAINER.PROMETAR.PROMPT DEPTH VISION = 9 # Max 12, minimum 0, for 0 it w
    cfg.TRAINER.PROMETAR.PROMPT_DEPTH_TEXT = 9 # Max 12, minimum 0, for 0 it wil
    cfg.DATASET.SUBSAMPLE_CLASSES = "all" # all, base or new
    cfg.TRAINER.PROMETAR.ADAPT_LR = 0.0005
    cfg.TRAINER.PROMETAR.LR_RATIO = 0.0005
    cfg.TRAINER.PROMETAR.FAST_ADAPTATION = False
   cfg.TRAINER.PROMETAR.MIXUP_ALPHA = 0.5
   cfg.TRAINER.PROMETAR.MIXUP_BETA = 0.5
    cfg.TRAINER.PROMETAR.DIM_RATE=8
   cfg.OPTIM_VNET = CN()
    cfg.OPTIM_VNET.NAME = "adam"
    cfg.OPTIM_VNET.LR = 0.0003
    cfg.OPTIM_VNET.WEIGHT_DECAY = 5e-4
   cfg.OPTIM_VNET.MOMENTUM = 0.9
    cfg.OPTIM VNET.SGD DAMPNING = 0
    cfg.OPTIM_VNET.SGD_NESTEROV = False
    cfg.OPTIM_VNET.RMSPROP_ALPHA = 0.99
   cfg.OPTIM_VNET.ADAM_BETA1 = 0.9
   cfg.OPTIM_VNET.ADAM_BETA2 = 0.999
   cfg.OPTIM_VNET.STAGED_LR = False
   cfg.OPTIM_VNET.NEW_LAYERS = ()
   cfg.OPTIM_VNET.BASE_LR_MULT = 0.1
   # Learning rate scheduler
   cfg.OPTIM_VNET.LR_SCHEDULER = "single_step"
```

```
# -1 or 0 means the stepsize is equal to max_epoch
   cfg.OPTIM_VNET.STEPSIZE = (-1, )
   cfg.OPTIM VNET.GAMMA = 0.1
   cfg.OPTIM_VNET.MAX_EPOCH = 10
   # Set WARMUP_EPOCH larger than 0 to activate warmup training
   cfg.OPTIM VNET.WARMUP EPOCH = -1
   # Either linear or constant
   cfg.OPTIM_VNET.WARMUP_TYPE = "linear"
   # Constant learning rate when type=constant
   cfg.OPTIM_VNET.WARMUP_CONS_LR = 1e-5
   # Minimum learning rate when type=linear
   cfg.OPTIM_VNET.WARMUP_MIN_LR = 1e-5
   # Recount epoch for the next scheduler (last_epoch=-1)
   # Otherwise last_epoch=warmup_epoch
   cfg.OPTIM_VNET.WARMUP_RECOUNT = True
def setup_cfg(args):
   cfg = get_cfg_default()
   extend cfg(cfg)
   # 1. From the dataset config file
   if args.dataset_config_file:
        cfg.merge_from_file(args.dataset_config_file)
   # 2. From the method config file
    if args.config_file:
        cfg.merge_from_file(args.config_file)
   # 3. From input arguments
    reset_cfg(cfg, args)
   cfg.freeze()
    return cfg
_tokenizer = _Tokenizer()
def load_clip_to_cpu(cfg): # Load CLIP
   backbone_name = cfg.MODEL.BACKBONE.NAME
   url = clip._MODELS[backbone_name]
   model_path = clip._download(url)
   try:
       # loading JIT archive
       model = torch.jit.load(model_path, map_location="cpu").eval()
       state_dict = None
   except RuntimeError:
        state_dict = torch.load(model_path, map_location="cpu")
   if cfg.TRAINER.NAME == "":
     design_trainer = "CoOp"
   else:
      design_trainer = cfg.TRAINER.NAME
   design_details = {"trainer": design_trainer,
                      "vision_depth": 0,
                      "language_depth": 0, "vision_ctx": 0,
                      "language_ctx": 0}
   model = clip.build_model(state_dict or model.state_dict(), design_details)
```

return model

```
from dassl.config import get_cfg_default
cfg = get_cfg_default()
cfg.MODEL.BACKBONE.NAME = "ViT-B/16" # Set the vision encoder backbone of CLIP to
clip model = load clip to cpu(cfg)
class TextEncoder(nn.Module):
    def __init__(self, clip_model): # 초기화 하는 함수
        super().__init__()
        self.transformer = clip_model.transformer
        self.positional_embedding = clip_model.positional_embedding
        self.ln_final = clip_model.ln_final
        self.text projection = clip model.text projection
        self.dtype = clip_model.dtype
   def forward(self, prompts, tokenized prompts): # 모델 호출
        x = prompts + self.positional embedding.type(self.dtype)
        x = x.permute(1, 0, 2) \# NLD \rightarrow LND
        x = self.transformer(x)
        x = x.permute(1, 0, 2) \# LND \rightarrow NLD
        x = self.ln_final(x).type(self.dtype)
        # x.shape = [batch_size, n_ctx, transformer.width]
        # take features from the eot embedding (eot_token is the highest number i
        x = x[torch.arange(x.shape[0]), tokenized_prompts.argmax(dim=-1)] @ self.
        return x
@TRAINER_REGISTRY.register(force=True)
class CoCoOp(TrainerX):
   def check_cfg(self, cfg):
        assert cfg.TRAINER.COCOOP.PREC in ["fp16", "fp32", "amp"]
    def build_model(self):
        cfg = self.cfg
        classnames = self.dm.dataset.classnames
        print(f"Loading CLIP (backbone: {cfg.MODEL.BACKBONE.NAME})")
        clip_model = load_clip_to_cpu(cfg)
        if cfg.TRAINER.COCOOP.PREC == "fp32" or cfg.TRAINER.COCOOP.PREC == "amp":
            # CLIP's default precision is fp16
            clip_model.float()
        print("Building custom CLIP")
        self.model = CoCoOpCustomCLIP(cfg, classnames, clip_model)
        print("Turning off gradients in both the image and the text encoder")
        name_to_update = "prompt_learner"
        for name, param in self.model.named_parameters():
            if name_to_update not in name:
```

```
param.requires_grad_(False)
   # Double check
    enabled = set()
    for name, param in self.model.named_parameters():
        if param.requires grad:
            enabled.add(name)
    print(f"Parameters to be updated: {enabled}")
    if cfg.MODEL.INIT WEIGHTS:
        load_pretrained_weights(self.model.prompt_learner, cfg.MODEL.INIT_WEI
    self.model.to(self.device)
    # NOTE: only give prompt_learner to the optimizer
    self.optim = build_optimizer(self.model.prompt_learner, cfg.OPTIM)
    self.sched = build lr scheduler(self.optim, cfg.OPTIM)
    self.register_model("prompt_learner", self.model.prompt_learner, self.opt
    self.scaler = GradScaler() if cfg.TRAINER.COCOOP.PREC == "amp" else None
   # Note that multi-gpu training could be slow because CLIP's size is
   # big, which slows down the copy operation in DataParallel
   device count = torch.cuda.device count()
    if device count > 1:
        print(f"Multiple GPUs detected (n_gpus={device_count}), use all of th
        self.model = nn.DataParallel(self.model)
def before_train(self):
    directory = self.cfg.OUTPUT DIR
    if self.cfg.RESUME:
        directory = self.cfg.RESUME
    self.start epoch = self.resume model if exist(directory)
   # Remember the starting time (for computing the elapsed time)
    self.time_start = time.time()
def forward_backward(self, batch):
    image, label = self.parse_batch_train(batch)
   model = self.model
    optim = self.optim
    scaler = self.scaler
    prec = self.cfg.TRAINER.COCOOP.PREC
    loss = model(image, label) # Input image 모델 통과
    optim.zero_grad()
    loss.backward() # Backward (역전파)
    optim.step() # 모델 parameter update
    loss_summary = {"loss": loss.item()}
    if (self.batch_idx + 1) == self.num_batches:
        self.update_lr()
```

```
return loss_summary
def parse batch train(self, batch):
    input = batch["img"]
    label = batch["label"]
    input = input.to(self.device)
    label = label.to(self.device)
    return input, label
def load model(self, directory, epoch=None):
    if not directory:
        print("Note that load_model() is skipped as no pretrained model is gi
        return
    names = self.get_model_names()
    # By default, the best model is loaded
    model_file = "model-best.pth.tar"
    if epoch is not None:
        model_file = "model.pth.tar-" + str(epoch)
    for name in names:
        model_path = osp.join(directory, name, model_file)
        if not osp.exists(model path):
            raise FileNotFoundError('Model not found at "{}"'.format(model_pa
        checkpoint = load checkpoint(model path)
        state dict = checkpoint["state dict"]
        epoch = checkpoint["epoch"]
        # Ignore fixed token vectors
        if "token_prefix" in state_dict:
            del state_dict["token_prefix"]
        if "token_suffix" in state_dict:
            del state_dict["token_suffix"]
        print("Loading weights to {} " 'from "{}" (epoch = {})'.format(name,
        # set strict=False
        self._models[name].load_state_dict(state_dict, strict=False)
def after_train(self):
  print("Finish training")
  do_test = not self.cfg.TEST.NO_TEST
  if do_test:
      if self.cfg.TEST.FINAL_MODEL == "best_val":
          print("Deploy the model with the best val performance")
          self.load_model(self.output_dir)
      else:
          print("Deploy the last-epoch model")
      acc = self.test()
```

```
# Show elapsed time
     elapsed = round(time.time() - self.time_start)
      elapsed = str(datetime.timedelta(seconds=elapsed))
      print(f"Elapsed: {elapsed}")
     # Close writer
      self.close writer()
      return acc
   def train(self):
        """Generic training loops."""
        self.before_train()
        for self.epoch in range(self.start_epoch, self.max_epoch):
            self.before epoch()
            self.run_epoch()
            self.after epoch()
        acc = self.after_train()
        return acc
parser = argparse.ArgumentParser()
parser.add_argument("--root", type=str, default="data/", help="path to dataset")
parser.add_argument("--output-dir", type=str, default="outputs/cocoop3", help="ou
parser.add argument(
   "--seed", type=int, default=1, help="only positive value enables a fixed seed
parser.add_argument(
   "--config-file", type=str, default="configs/trainers/ProMetaR/vit_b16_c2_ep10
parser.add argument(
   "--dataset-config-file",
   type=str,
   default="configs/datasets/eurosat.yaml",
   help="path to config file for dataset setup",
)
parser.add_argument("--trainer", type=str, default="CoOp", help="name of trainer"
parser.add_argument("--eval-only", action="store_true", help="evaluation only")
parser.add_argument(
   "--model-dir",
   type=str,
   default="",
   help="load model from this directory for eval-only mode",
)
parser.add_argument("--train-batch-size", type=int, default=4)
parser.add_argument("--epoch", type=int, default=10)
parser.add_argument("--subsample-classes", type=str, default="base")
parser.add_argument(
   "--load-epoch", type=int, default=0, help="load model weights at this epoch f
args = parser.parse_args([])
def main(args):
   cfg = setup_cfg(args)
    if cfg.SEED >= 0:
        set_random_seed(cfg.SEED)
```

```
if torch.cuda.is_available() and cfg.USE_CUDA:
    torch.backends.cudnn.benchmark = True

trainer = build_trainer(cfg)
if args.eval_only:
    trainer.load_model(args.model_dir, epoch=args.load_epoch)
    acc = trainer.test()
    return acc

acc = trainer.train()
return acc
```

```
IIII Lating: eurosat/2/30/rermanenttrop/rermanenttrop_ooz.jpg
  inflating: eurosat/2750/PermanentCrop/PermanentCrop 357.jpg
  inflating: eurosat/2750/PermanentCrop/PermanentCrop 1.jpg
  inflating: eurosat/2750/PermanentCrop/PermanentCrop_65.jpg
  inflating: eurosat/2750/PermanentCrop/PermanentCrop 736.jpg
/content/ProMetaR/data/eurosat
Downloading...
From: https://drive.google.com/uc?id=1Ip7yaCWFi0eaOFUGga0lUdVi DDQth1o
To: /content/ProMetaR/data/eurosat/split_zhou_EuroSAT.json
100% 3.01M/3.01M [00:00<00:00, 23.8MB/s]
/content/ProMetaR
```

∨ Q1. Understanding and implementing CoCoOp

- We have learned how to define CoOp in Lab Session 4.
- The main difference between CoOp and CoCoOp is **meta network** to extract image tokens that is added to the text prompt.
- Based on the CoOp code given in Lab Session 4, fill-in-the-blank exercise (4 blanks!!) to test your understanding of critical parts of the CoCoOp.

```
import torch.nn as nn
class CoCoOpPromptLearner(nn.Module):
   def __init__(self, cfg, classnames, clip_model):
        super().__init__()
        n cls = len(classnames)
        n_ctx = cfg.TRAINER.COCOOP.N_CTX
        ctx_init = cfg.TRAINER.COCOOP.CTX_INIT
        dtype = clip_model.dtype
        ctx_dim = clip_model.ln_final.weight.shape[0]
        vis_dim = clip_model.visual.output_dim
        clip_imsize = clip_model.visual.input_resolution
        cfg imsize = cfg.INPUT.SIZE[0]
        assert cfg_imsize == clip_imsize, f"cfg_imsize ({cfg_imsize}) must
        if ctx init:
            # use given words to initialize context vectors
            ctx_init = ctx_init.replace("_", " ")
            n_ctx = len(ctx_init.split(" "))
            prompt = clip.tokenize(ctx_init)
            with torch.no_grad():
                embedding = clip_model.token_embedding(prompt).type(dtype)
            ctx_vectors = embedding[0, 1: 1 + n_ctx, :]
            prompt_prefix = ctx_init
        else:
            # random initialization
            ctx_vectors = torch.empty(n_ctx, ctx_dim, dtype=dtype)
            nn.init.normal_(ctx_vectors, std=0.02)
            prompt_prefix = " ".join(["X"] * n_ctx)
        print(f'Initial context: "{prompt_prefix}"')
        print(f"Number of context words (tokens): {n_ctx}")
```

```
self.ctx = nn.Parameter(ctx_vectors) # Wrap the initialized promp
   ### Tokenize ###
    classnames = [name.replace("_", " ") for name in classnames] # 예
    name_lens = [len(_tokenizer.encode(name)) for name in classnames]
    prompts = [prompt prefix + " " + name + "." for name in classnames
   tokenized_prompts = torch.cat([clip.tokenize(p) for p in prompts])
    ####### 01. Fill in the blank ######
   ######## Define Meta Net ########
    self.meta net = nn.Sequential(OrderedDict([
        ("linear1", nn.Linear(vis dim, vis dim // 16)),
        ("relu", nn.ReLU(inplace=True)),
        ("linear2", nn.Linear(vis_dim // 16, ctx_dim))
    1))
    ## Hint: meta network is composed to linear layer, relu activation
    if cfg.TRAINER.COCOOP.PREC == "fp16":
       self.meta net.half()
   with torch.no_grad():
       embedding = clip model.token embedding(tokenized prompts).type
   # These token vectors will be saved when in save_model(),
   # but they should be ignored in load_model() as we want to use
   # those computed using the current class names
    self.register_buffer("token_prefix", embedding[:, :1, :]) # SOS
    self.register_buffer("token_suffix", embedding[:, 1 + n_ctx:, :])
    self.n_cls = n_cls
    self.n_ctx = n_ctx
    self.tokenized_prompts = tokenized_prompts # torch.Tensor
    self.name lens = name lens
def construct_prompts(self, ctx, prefix, suffix, label=None):
   # dim0 is either batch_size (during training) or n_cls (during tes
   # ctx: context tokens, with shape of (dim0, n_ctx, ctx_dim)
   # prefix: the sos token, with shape of (n_cls, 1, ctx_dim)
   # suffix: remaining tokens, with shape of (n_cls, *, ctx_dim)
    if label is not None:
       prefix = prefix[label]
       suffix = suffix[label]
    prompts = torch.cat(
           prefix, # (dim0, 1, dim)
           ctx, # (dim0, n_ctx, dim)
           suffix # /dima ... dim/
```

```
SUITEX, # (UIIII), A, UIIII)
          ],
          dim=1,
       )
       return prompts
   def forward(self, im features):
       prefix = self.token prefix
       suffix = self.token suffix
       ctx = self.ctx # (n ctx, ctx dim)
       ######## Q2,3. Fill in the blank #######
       bias = self.meta_net(im_features) # (batch, ctx_dim)
       bias = bias.unsqueeze(1) # (batch, 1, ctx_dim)
       ctx = ctx.unsqueeze(0) # (1, n_ctx, ctx_dim)
       ctx_shifted = ctx + bias # (batch, n_ctx, ctx_dim)
       # Use instance-conditioned context tokens for all classes
       prompts = []
       for ctx shifted i in ctx shifted:
          ctx i = ctx shifted i.unsqueeze(0).expand(self.n cls, -1, -1)
          pts_i = self.construct_prompts(ctx_i, prefix, suffix) # (n_cl
          prompts.append(pts i)
       prompts = torch.stack(prompts)
       return prompts
class CoCoOpCustomCLIP(nn.Module):
   def __init__(self, cfg, classnames, clip_model):
       super().__init__()
       self.prompt_learner = CoCoOpPromptLearner(cfg, classnames, clip_model)
       self.tokenized_prompts = self.prompt_learner.tokenized_prompts
       self.image_encoder = clip_model.visual
       self.text_encoder = TextEncoder(clip_model)
       self.logit_scale = clip_model.logit_scale
       self.dtype = clip_model.dtype
   def forward(self, image, label=None):
       tokenized_prompts = self.tokenized_prompts
       logit_scale = self.logit_scale.exp()
       image_features = self.image_encoder(image.type(self.dtype))
       image_features = image_features / image_features.norm(dim=-1, keepdim=Tru
       ######### 04. Fill in the blank #######
```

∨ Q2. Training CoCoOp

In this task, you will train CoCoOp on the EuroSAT dataset. If your implementation of CoCoOp in Question 1 is correct, the following code should execute without errors. Please submit the execution file so we can evaluate whether your code runs without any issues.

```
# Train on the Base Classes Train split and evaluate accuracy on the Base
args.trainer = "CoCoOp"
args.epoch = 100
args.output_dir = "outputs/cocoop"

args.subsample_classes = "base"
args.eval_only = False
cocoop_base_acc = main(args)
```

```
epocn [/4/שש] שמנכו [עש/בש] נבווופ שיבוש (ש.בו14) data (ש.בובא) נועס. נועס. טיבובא (ש.בובא) נועס.
    epoch [75/100] batch [20/20] time 0.099 (0.115) data 0.000 (0.013) loss 0.057!
    epoch [76/100] batch [20/20] time 0.127 (0.119) data 0.000 (0.013) loss 0.009
    epoch [77/100] batch [20/20] time 0.099 (0.116) data 0.000 (0.013) loss 0.010.
    epoch [78/100] batch [20/20] time 0.098 (0.115) data 0.000 (0.014) loss 0.1649
    epoch [79/100] batch [20/20] time 0.099 (0.115) data 0.000 (0.013) loss 0.724:
    epoch [80/100] batch [20/20] time 0.098 (0.114) data 0.000 (0.012) loss 0.099
    epoch [81/100] batch [20/20] time 0.100 (0.116) data 0.000 (0.013) loss 0.1414
    epoch [82/100] batch [20/20] time 0.099 (0.115) data 0.000 (0.013) loss 0.510
    epoch [83/100] batch [20/20] time 0.098 (0.115) data 0.000 (0.014) loss 0.098
    epoch [84/100] batch [20/20] time 0.100 (0.114) data 0.000 (0.013) loss 0.107!
    epoch [85/100] batch [20/20] time 0.098 (0.116) data 0.000 (0.014) loss 0.030!
    epoch [86/100] batch [20/20] time 0.100 (0.116) data 0.000 (0.013) loss 0.0370
    epoch [87/100] batch [20/20] time 0.100 (0.115) data 0.000 (0.013) loss 0.154
    epoch [88/100] batch [20/20] time 0.097 (0.114) data 0.000 (0.012) loss 0.097
    epoch [89/100] batch [20/20] time 0.098 (0.115) data 0.000 (0.013) loss 0.032
    epoch [90/100] batch [20/20] time 0.100 (0.114) data 0.000 (0.013) loss 0.049
    epoch [91/100] batch [20/20] time 0.099 (0.115) data 0.000 (0.013) loss 0.3079
    epoch [92/100] batch [20/20] time 0.100 (0.115) data 0.000 (0.014) loss 0.040
    epoch [93/100] batch [20/20] time 0.101 (0.116) data 0.000 (0.013) loss 0.166
    epoch [94/100] batch [20/20] time 0.099 (0.114) data 0.000 (0.013) loss 0.319:
    epoch [95/100] batch [20/20] time 0.101 (0.116) data 0.000 (0.014) loss 0.1212
    epoch [96/100] batch [20/20] time 0.099 (0.117) data 0.000 (0.014) loss 0.266
    epoch [97/100] batch [20/20] time 0.099 (0.116) data 0.000 (0.015) loss 0.0498
    epoch [98/100] batch [20/20] time 0.100 (0.119) data 0.000 (0.014) loss 0.175
    epoch [99/100] batch [20/20] time 0.099 (0.115) data 0.000 (0.013) loss 0.062
    epoch [100/100] batch [20/20] time 0.098 (0.115) data 0.000 (0.013) loss 0.00.
    Checkpoint saved to outputs/cocoop/prompt learner/model.pth.tar-100
    Finish training
    Deploy the last-epoch model
    Evaluate on the *test* set
                  42/42 [00:47<00:00, 1.14s/it]=> result
    100%
    * total: 4,200
    * correct: 3,844
    * accuracy: 91.5%
    * error: 8.5%
    * macro_f1: 91.6%
    Elapsed: 0:04:47
# Accuracy on the New Classes.
args.model_dir = "outputs/cocoop"
args.output_dir = "outputs/cocoop/new_classes"
args.subsample_classes = "new"
args.load_epoch = 100
args.eval_only = True
coop_novel_acc = main(args)
→ Loading trainer: CoCoOp
    Loading dataset: EuroSAT
    Reading split from /content/ProMetaR/data/eurosat/split_zhou_EuroSAT.json
    Loading preprocessed few-shot data from /content/ProMetaR/data/eurosat/split_
    SUBSAMPLE NEW CLASSES!
    Building transform_train
    + random resized crop (size=(224, 224), scale=(0.08, 1.0))
    + random flip
    + to torch tensor of range [0, 1]
    + normalization (mean=[0.48145466, 0.4578275, 0.40821073], std=[0.26862954, 0
    Building transform_test
    + resize the smaller edge to 224
```

```
+ 224x224 center crop
+ to torch tensor of range [0, 1]
+ normalization (mean=[0.48145466, 0.4578275, 0.40821073], std=[0.26862954, 0
           EuroSAT
Dataset
# classes 5
# train_x 80
# val
           20
# test
           3,900
Loading CLIP (backbone: ViT-B/16)
/usr/local/lib/python3.10/dist-packages/torch/optim/lr_scheduler.py:62: UserWa
  warnings.warn(
/content/ProMetaR/dassl/utils/torchtools.py:102: FutureWarning: You are using
  checkpoint = torch.load(fpath, map location=map location)
Building custom CLIP
Initial context: "a photo of a"
Number of context words (tokens): 4
Turning off gradients in both the image and the text encoder
Parameters to be updated: {'prompt_learner.meta_net.linear2.weight', 'prompt_'
Loading evaluator: Classification
Loading weights to prompt_learner from "outputs/cocoop/prompt_learner/model.p"
Evaluate on the *test* set
100%| 39/39 [00:44<00:00, 1.13s/it]=> result
* total: 3,900
* correct: 1,758
* accuracy: 45.1%
* error: 54.9%
* macro_f1: 40.9%
```

∨ Q3. Analyzing the results of CoCoOp

Compare the results of CoCoOp with those of CoOp that we trained in Lab Session 4. Discuss possible reasons for the performance differences observed between CoCoOp and CoOp.

CoOp is optimized for base classes, but struggles to generalize to novel ones due to fixed prompts. In contrast, CoCoOp achieves a better balance between base and novel class performance using meta-network, by leveraging image features to adapt the context tokens dynamically.

But in this case, CoOp's accuracy score for base class = 91.4, novel class = 51.5, whereas CoCoOp's accuracy score for base class = 91.5, novel class = 45.1. This unexpected result maybe due to the overfitting in CoCoOp.

