```
# NOTE: The scaffolding code for this part of the assignment
# is adapted from https://github.com/pytorch/examples.
from future import print function
import argparse
import numpy as np
import torch
import torch.nn as nn
import torch.nn.functional as F
import torch.optim as optim
from torchvision import transforms
import torchvision
from torch.autograd import Variable
# NOTE: This is adapted from the torchvision datasets for
# CIFAR10 and CIFAR100, which can be found at
# https://github.com/pytorch/vision/blob/master/torchvision/datasets/cifar.py
# This version allows a validation dataset to be created.
from __future_ import print function
from PIL import Image
import os
import os.path
import errno
import numpy as np
import sys
if sys.version info[0] == 2:
    import cPickle as pickle
else:
    import pickle
import torch.utils.data as data
from torchvision.datasets.utils import download_url, check_integrity
from google.colab import drive
drive.mount('/content/gdrive')
class CIFAR10(data.Dataset):
    """`CIFAR10 <https://www.cs.toronto.edu/~kriz/cifar.html>`_ Dataset.
    Args:
        root (string): Root directory of dataset where directory
            ``cifar-10-batches-py`` exists.
        train (bool, optional): If True, creates dataset from training set, otherwise
            creates from test set.
        transform (callable, optional): A function/transform that takes in an PIL im
            and returns a transformed version. E.g, ``transforms.RandomCrop`
        target transform (callable, optional): A function/transform that takes in the
            target and transforms it.
        download (bool, optional): If true, downloads the dataset from the internet a
            puts it in root directory. If dataset is already downloaded, it is not
            downloaded again.
    base folder = 'cifar-10-batches-py'
```

```
url = "http://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz"
filename = "cifar-10-python.tar.gz"
tgz md5 = 'c58f30108f718f92721af3b95e74349a'
# validation examples will come from here
train list = [
    ['data batch 1', 'c99cafc152244af753f735de768cd75f'],
    ['data batch 2', 'd4bba439e000b95fd0a9bffe97cbabec'],
    ['data batch 3', '54ebc095f3ab1f0389bbae665268c751'],
    ['data_batch_4', '634d18415352ddfa80567beed471001a'],
    ['data batch 5', '482c414d41f54cd18b22e5b47cb7c3cb'],
]
test list = [
    ['test batch', '40351d587109b95175f43aff81a1287e'],
1
def __init__(self, root, split='train',
             transform=None, target transform=None,
             download=False, val samples=1000):
    self.root = os.path.expanduser(root)
    self.transform = transform
    self.target transform = target transform
    self.split = split # train, val, or test
    if download:
        self.download()
    if not self. check integrity():
        raise RuntimeError('Dataset not found or corrupted.' +
                            ' You can use download=True to download it')
    # now load the picked numpy arrays
    if self.split in ['train', 'val']:
        self.train data = []
        self.train labels = []
        for fentry in self.train list:
            f = fentry[0]
            file = os.path.join(self.root, self.base_folder, f)
            fo = open(file, 'rb')
            if sys.version info[0] == 2:
                entry = pickle.load(fo)
            else:
                entry = pickle.load(fo, encoding='latin1')
            self.train data.append(entry['data'])
            if 'labels' in entry:
                self.train labels += entry['labels']
                self.train_labels += entry['fine_labels']
            fo.close()
        self.train data = np.concatenate(self.train data)
        self.train data = self.train data.reshape((50000, 3, 32, 32))
```

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self.train_data = self.train_data.transpose((0, 2, 3, 1)) # convert to H
        self.val data = self.train_data[-val_samples:]
        self.val_labels = self.train_labels[-val_samples:]
        self.train_data = self.train_data[:-val_samples]
        self.train labels = self.train labels[:-val samples]
    elif self.split == 'test':
        f = self.test list[0][0]
        file = os.path.join(self.root, self.base_folder, f)
        fo = open(file, 'rb')
        if sys.version info[0] == 2:
            entry = pickle.load(fo)
        else:
            entry = pickle.load(fo, encoding='latin1')
        self.test_data = entry['data']
        if 'labels' in entry:
            self.test labels = entry['labels']
        else:
            self.test labels = entry['fine labels']
        fo.close()
        self.test data = self.test data.reshape((10000, 3, 32, 32))
        self.test_data = self.test_data.transpose((0, 2, 3, 1)) # convert to HWC
    else:
        raise Exception('Unkown split {}'.format(self.split))
def __getitem__(self, index):
    Args:
        index (int): Index
    Returns:
        tuple: (image, target) where target is index of the target class.
    if self.split == 'train':
        img, target = self.train data[index], self.train labels[index]
    elif self.split == 'val':
        img, target = self.val data[index], self.val labels[index]
    elif self.split == 'test':
        img, target = self.test data[index], self.test labels[index]
    # doing this so that it is consistent with all other datasets
    # to return a PIL Image
    img = Image.fromarray(img)
    if self.transform is not None:
        img = self.transform(img)
    if self.target_transform is not None:
        target = self.target transform(target)
    return img, target
def __len__(self):
```

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ir self.split == 'train':
            return len(self.train data)
        elif self.split == 'val':
            return len(self.val data)
        elif self.split == 'test':
            return len(self.test data)
    def _check_integrity(self):
        root = self.root
        for fentry in (self.train_list + self.test_list):
            filename, md5 = fentry[0], fentry[1]
            fpath = os.path.join(root, self.base folder, filename)
            if not check integrity(fpath, md5):
                return False
        return True
    def download(self):
        import tarfile
        if self._check_integrity():
            print('Files already downloaded and verified')
            return
        root = self.root
        download_url(self.url, root, self.filename, self.tgz_md5)
        # extract file
        cwd = os.getcwd()
        tar = tarfile.open(os.path.join(root, self.filename), "r:gz")
        os.chdir(root)
        tar.extractall()
        tar.close()
        os.chdir(cwd)
import matplotlib.pyplot as plt
# You should implement these (softmax.py, twolayernn.py, convnet.py)
description='CIFAR-10 Example'
# Hyperparameters
lr = 0.0001
momentum = 0.9
weight decay = 0.1
batch size = 256
epochs = 10
model = 'mymodel'
no cuda=False
seed=1
test batch size=1000
log interval=10
cifar10 dir = 'data'
cuda = True
```

```
torch.manual_seed(seed)
if cuda:
    torch.cuda.manual seed(seed)
# Load CIFAR10 using torch data paradigm
kwargs = {'num workers': 1, 'pin memory': True} if cuda else {}
# CIFAR10 meta data
n classes = 10
im size = (3, 32, 32)
# Subtract the mean color and divide by standard deviation. The mean image
# from part 1 of this homework was essentially a big gray blog, so
# subtracting the same color for all pixels doesn't make much difference.
# mean color of training images
cifar10 mean color = [0.49131522, 0.48209435, 0.44646862]
# std dev of color across training images
cifar10 std color = [0.01897398, 0.03039277, 0.03872553]
transform = transforms.Compose([
                 transforms.Resize(size=(224, 224)),
                 transforms.RandomHorizontalFlip(),
                 transforms.RandomCrop(size=(224,224), padding=4),
                 transforms.ToTensor(),
                 transforms.Normalize(cifar10 mean color, cifar10 std color),
            1)
# Datasets
train_dataset = CIFAR10(cifar10_dir, split='train', download=True,
                        transform=transform)
val dataset = CIFAR10(cifar10 dir, split='val', download=True,
                        transform=transform)
test dataset = CIFAR10(cifar10 dir, split='test', download=True,
                        transform=transform)
# DataLoaders
train loader = torch.utils.data.DataLoader(train dataset,
                 batch size=batch size, shuffle=True, **kwargs)
val loader = torch.utils.data.DataLoader(val dataset,
                 batch size=batch size, shuffle=True, **kwargs)
test loader = torch.utils.data.DataLoader(test dataset,
                 batch size=batch size, shuffle=True, **kwargs)
# def imshow(inp, title=None):
#
      inp = inp.numpy().transpose((1, 2, 0))
#
      mean = np.array(cifar10 mean color)
#
      std = np.array(cifar10_std_color)
      inp = std * inp + mean
#
      inp = np.clip(inp, 0, 1)
#
#
     plt.imshow(inp)
#
     if title is not None:
#
          plt.title(title)
      plt.pause(0.001) # Pause a bit so that plots are updated
if model == 'mymodel':
    # model = models.mymodel.MyModel(im_size, args.hidden_dim, args.hidden_dim2,
                                 arms karnal siza arms karnal siza?
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  model = torchvision.models.resnet152(pretrained=True)
  for param in model.parameters():
     param.requires grad = False
  print(model)
  # Modify the last layer
  # ReLU(inplace=True)
  # (5): Dropout(p=0.5, inplace=False)
  model.fc = torch.nn.Linear(2048, 10)
  print(model)
else:
   raise Exception('Unknown model {}'.format(model))
# cross-entropy loss function
criterion = F.cross entropy
model.cuda()
# TODO: Initialize an optimizer from the torch.optim package using the
# appropriate hyperparameters found in args. This only requires one line.
optimizer = optim.Adam(model.parameters(), lr=lr, weight decay=weight decay)
END OF YOUR CODE
def train(epoch):
  Train the model for one epoch.
  # Some models use slightly different forward passes and train and test
  # time (e.g., any model with Dropout). This puts the model in train mode
  # (as opposed to eval mode) so it knows which one to use.
  model.train()
  \max val loss = 1000
  # train loop
  for batch idx, batch in enumerate(train loader):
     # prepare data
     images, targets = Variable(batch[0]), Variable(batch[1])
     print(images.shape)
     # imshow(images[0])
     if cuda:
        images, targets = images.cuda(), targets.cuda()
     # TODO: Update the parameters in model using the optimizer from above.
     # This only requires a couple lines of code.
     loss = criterion(model(images), targets)
     optimizer.zero grad()
     loss.backward()
     optimizer.step()
```

END OF YOUR CODE

```
if batch idx % log interval == 0:
           val loss, val acc = evaluate('val', n batches=4)
           train loss = loss.data
           examples_this_epoch = batch_idx * len(images)
           epoch_progress = 100. * batch_idx / len(train_loader)
           print('Train Epoch: {} [{}/{} ({:.0f}%)]\t'
                 'Train Loss: {:.6f}\tVal Loss: {:.6f}\tVal Acc: {}'.format(
               epoch, examples_this_epoch, len(train_loader.dataset),
               epoch progress, train loss, val loss, val acc))
           if val loss < max_val_loss:</pre>
             model save name = 'transferLearning.pt'
             path = F"/content/gdrive/My Drive/transferLearning.pt"
             torch.save(model, path)
             max val loss = val loss
             print("Saved model this time!")
def evaluate(split, verbose=False, n batches=None):
   Compute loss on val or test data.
   model.eval()
   loss = 0
   correct = 0
   n = 0
   if split == 'val':
       loader = val loader
   elif split == 'test':
       loader = test_loader
   for batch i, batch in enumerate(loader):
       data, target = batch
       if cuda:
           data, target = data.cuda(), target.cuda()
       data, target = Variable(data, volatile=True), Variable(target)
       output = model(data)
       loss += criterion(output, target, size average=False).data
       # predict the argmax of the log-probabilities
       pred = output.data.max(1, keepdim=True)[1]
       correct += pred.eq(target.data.view as(pred)).cpu().sum()
       n_examples += pred.size(0)
       if n batches and (batch i >= n batches):
           break
   loss /= n_examples
   acc = 100. * correct / n_examples
   if verbose:
       print('\n{} set: Average loss: {:.4f}, Accuracy: {}/{} ({:.0f}%)\n'.format(
           split, loss, correct, n examples, acc))
    return loss, acc
```

```
# train the model one epoch at a time

for epoch in range(1, epochs + 1):
    train(epoch)
evaluate('test', verbose=True)

# Save the model (architecture and weights)

# Later you can call torch.load(file) to re-load the trained model into python
# See http://pytorch.org/docs/master/notes/serialization.html for more details
```

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```
Enter your authorization code:
        Mounted at /content/gdrive
        Downloading <a href="http://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz">http://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz</a> to data/cifar-
                                               170500096it [00:06, 26924132.12it/s]
        Files already downloaded and verified
        Files already downloaded and verified
        Downloading: "<a href="https://download.pytorch.org/models/resnet152-b121ed2d.pth" to /roc</a>
                                               100% 230M/230M [00:10<00:00, 23.1MB/s]
        ResNet(
          (conv1): Conv2d(3, 64, kernel_size=(7, 7), stride=(2, 2), padding=(3, 3), bias=
          (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running state
          (relu): ReLU(inplace=True)
          (maxpool): MaxPool2d(kernel size=3, stride=2, padding=1, dilation=1, ceil mode=
          (layer1): Sequential(
            (0): Bottleneck(
               (conv1): Conv2d(64, 64, kernel size=(1, 1), stride=(1, 1), bias=False)
               (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
               (conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
               (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
               (conv3): Conv2d(64, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
               (bn3): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
               (relu): ReLU(inplace=True)
               (downsample): Sequential(
                 (0): Conv2d(64, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
                 (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
              )
            (1): Bottleneck(
               (conv1): Conv2d(256, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
               (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
               (conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
               (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
               (conv3): Conv2d(64, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
               (bn3): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
               (relu): ReLU(inplace=True)
            )
            (2): Bottleneck(
               (conv1): Conv2d(256, 64, kernel size=(1, 1), stride=(1, 1), bias=False)
               (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
               (conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
               (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
               (conv3): Conv2d(64, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
               (bn3): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
              (relu): ReLU(inplace=True)
            )
          (layer2): Sequential(
            (0): Bottleneck(
               (conv1): Conv2d(256, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
               (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
               (cany)) · Cany)d(1)
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https://colab.research.google.com/drive/1em2z7zzx8s0tYgjr4Flvgp609Vls73D8#scrollTo=I5VovcO4I VW&printMode=true
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  (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(128, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
  (downsample): Sequential(
    (0): Conv2d(256, 512, kernel_size=(1, 1), stride=(2, 2), bias=False)
    (1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track_running
 )
(1): Bottleneck(
  (conv1): Conv2d(512, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(2): Bottleneck(
  (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track_running
 (relu): ReLU(inplace=True)
)
(3): Bottleneck(
  (conv1): Conv2d(512, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(4): Bottleneck(
  (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(5): Bottleneck(
  (conv1): Conv2d(512, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(128, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(6): Bottleneck(
  (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
```

```
(conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
    (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
    (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
    (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
    (relu): ReLU(inplace=True)
 )
 (7): Bottleneck(
    (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
    (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
    (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
    (conv3): Conv2d(128, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 )
)
(layer3): Sequential(
 (0): Bottleneck(
    (conv1): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
    (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(2, 2), padding=(1, 1)
    (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
    (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
    (relu): ReLU(inplace=True)
    (downsample): Sequential(
     (0): Conv2d(512, 1024, kernel size=(1, 1), stride=(2, 2), bias=False)
     (1): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track runnir
   )
 )
 (1): Bottleneck(
    (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
    (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
    (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
    (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
    (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 (2): Bottleneck(
    (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
    (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
    (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
    (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
    (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
    (relu): ReLU(inplace=True)
 )
 (3): Bottleneck(
    (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
    (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
    (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
    (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
    (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
    (relu): ReLU(inplace=True)
 (4): Bottleneck(
```

```
(conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(5): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(6): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(7): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(8): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(9): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(10): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
                          222-12 AE
```

```
(DNZ): Datchwormzu(ZDO, EPS=1E-DD, Momentum=D.1, alline=llue, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(11): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(12): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(13): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(14): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(15): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(16): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
```

```
(17): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(18): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(19): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
(20): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(21): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(22): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track runnir
  (relu): ReLU(inplace=True)
)
(23): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256. eps=1e-05. momentum=0.1. affine=True. track running
```

https://colab.research.google.com/drive/1em2z7zzx8s0tYgjr4Flvgp609Vls73D8#scrollTo=l5VovcO4l_VW&printMode=true 14/65

```
(conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track_runnir
  (relu): ReLU(inplace=True)
(24): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(25): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(26): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(27): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(28): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(29): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel\_size=(1, 1), stride=(1, 1), bias=False)
```

```
(DN3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, attine=Irue, track runnii
 (relu): ReLU(inplace=True)
)
(30): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(31): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
(32): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(33): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(34): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(35): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
```

```
(layer4): Sequential(
    (0): Bottleneck(
      (conv1): Conv2d(1024, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(512, 512, kernel size=(3, 3), stride=(2, 2), padding=(1, 1)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(512, 2048, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
      (downsample): Sequential(
        (0): Conv2d(1024, 2048, kernel size=(1, 1), stride=(2, 2), bias=False)
        (1): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      )
    )
    (1): Bottleneck(
      (conv1): Conv2d(2048, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(512, 2048, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
    (2): Bottleneck(
      (conv1): Conv2d(2048, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(512, 512, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(512, 2048, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
    )
  )
  (avgpool): AdaptiveAvgPool2d(output size=(1, 1))
  (fc): Linear(in features=2048, out features=1000, bias=True)
ResNet(
  (conv1): Conv2d(3, 64, kernel size=(7, 7), stride=(2, 2), padding=(3, 3), bias=
  (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running state
  (relu): ReLU(inplace=True)
  (maxpool): MaxPool2d(kernel size=3, stride=2, padding=1, dilation=1, ceil mode=
  (layer1): Sequential(
    (0): Bottleneck(
      (conv1): Conv2d(64, 64, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(64, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
      (downsample): Sequential(
        (0): Conv2d(64, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
        (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
      )
    (1): Bottleneck(
      (conv1): Conv2d(256, 64, kernel_size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(64 eps=1e-05 momentum=0 1 affine=True track running
```

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(conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
    (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(64, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 (2): Bottleneck(
    (conv1): Conv2d(256, 64, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1), padding=(1, 1),
   (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(64, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 )
)
(layer2): Sequential(
 (0): Bottleneck(
    (conv1): Conv2d(256, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(2, 2), padding=(1, 1)
   (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(128, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
   (downsample): Sequential(
     (0): Conv2d(256, 512, kernel size=(1, 1), stride=(2, 2), bias=False)
      (1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   )
 (1): Bottleneck(
   (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
   (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running
   (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 (2): Bottleneck(
   (conv1): Conv2d(512, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
   (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(128, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 )
 (3): Bottleneck(
    (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
    (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running
   (conv3): Conv2d(128, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
```

```
(4): BOTTLENECK(
   (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
   (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running
   (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 (5): Bottleneck(
   (conv1): Conv2d(512, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
   (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
  (6): Bottleneck(
   (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
    (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track_running
   (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track_running
   (relu): ReLU(inplace=True)
 )
 (7): Bottleneck(
    (conv1): Conv2d(512, 128, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
   (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(128, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
   (relu): ReLU(inplace=True)
 )
(layer3): Sequential(
 (0): Bottleneck(
   (conv1): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1)
   (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track runnir
   (relu): ReLU(inplace=True)
   (downsample): Sequential(
      (0): Conv2d(512, 1024, kernel size=(1, 1), stride=(2, 2), bias=False)
      (1): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
   )
 (1): Bottleneck(
   (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
   (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
   (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
   (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
   (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
```

```
(relu): ReLU(inplace=True)
(2): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(3): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(4): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(5): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(6): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(7): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track runnin⊄
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(8): Bottleneck(
```

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I/: CONVENTION, 200, NOTHICK_SIZE (I, I/, SCHICE (I, I/, DIAS-HAUSE)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
(9): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(10): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(11): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(12): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(13): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track runnin⊄
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
(14): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
```

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(conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(15): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(16): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(17): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(18): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(19): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
(20): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
```

```
(21): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
)
(22): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(23): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(24): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(25): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(26): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(27): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256 256 karnal ciza-13
                                            31 ctrido-11
                                                           1)
```

```
(CONVE), CONVER(EDO, EDO, REINEL_SIZE—(D, D), BUILUE—(I, I), pauding—(I, I,
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(28): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
(29): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(30): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
(31): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track_running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
  (relu): ReLU(inplace=True)
)
(32): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
 (relu): ReLU(inplace=True)
(33): Bottleneck(
  (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
  (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
  (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
  (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
  (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track runnir
```

```
(relu): ReLU(inplace=True)
    )
    (34): Bottleneck(
      (conv1): Conv2d(1024, 256, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1), padding=(1, 1)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
    )
    (35): Bottleneck(
      (conv1): Conv2d(1024, 256, kernel_size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(256, 1024, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
    )
  )
  (layer4): Sequential(
    (0): Bottleneck(
      (conv1): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(2, 2), padding=(1, 1)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(512, 2048, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
      (downsample): Sequential(
        (0): Conv2d(1024, 2048, kernel_size=(1, 1), stride=(2, 2), bias=False)
        (1): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      )
    )
    (1): Bottleneck(
      (conv1): Conv2d(2048, 512, kernel_size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(512, 2048, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
    (2): Bottleneck(
      (conv1): Conv2d(2048, 512, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1), padding=(1, 1)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True, track running
      (conv3): Conv2d(512, 2048, kernel size=(1, 1), stride=(1, 1), bias=False)
      (bn3): BatchNorm2d(2048, eps=1e-05, momentum=0.1, affine=True, track running
      (relu): ReLU(inplace=True)
    )
  )
  (avgpool): AdaptiveAvgPool2d(output size=(1, 1))
  (fc): Linear(in features=2048, out features=10, bias=True)
torch.Size([256, 3, 224, 224])
```

```
/usr/local/lib/python3.6/dist-packages/ipykernel_launcher.py:341: UserWarning: vc
/usr/local/lib/python3.6/dist-packages/torch/nn/ reduction.py:43: UserWarning: s:
  warnings.warn(warning.format(ret))
Train Epoch: 1 [0/49000 (0%)]
                                Train Loss: 2.368309
                                                        Val Loss: 2.380458
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [2560/49000 (5%)]
                                        Train Loss: 2.291346
                                                                Val Loss: 2.28630
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [5120/49000 (10%)]
                                        Train Loss: 2.225854 Val Loss: 2.22459
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [7680/49000 (16%)]
                                        Train Loss: 2.191221 Val Loss: 2.1933!
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [10240/49000 (21%)]
                                        Train Loss: 2.170921 Val Loss: 2.1646
Saved model this time!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
tarch Siza/[256
                 2 22/
                         22/11
```

```
LUICII.3128([230, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [12800/49000 (26%)]
                                        Train Loss: 2.158970 Val Loss: 2.12509
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [15360/49000 (31%)]
                                        Train Loss: 2.112028 Val Loss: 2.11732
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [17920/49000 (36%)]
                                        Train Loss: 2.142910 Val Loss: 2.08832
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [20480/49000 (42%)]
                                        Train Loss: 2.083063 Val Loss: 2.0558
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [23040/49000 (47%)]
                                        Train Loss: 2.026353 Val Loss: 2.04832
Saved model this time!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [25600/49000 (52%)]
                                        Train Loss: 2.074434 Val Loss: 2.01419
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [28160/49000 (57%)]
                                        Train Loss: 2.004257 Val Loss: 2.00240
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [30720/49000 (62%)]
                                        Train Loss: 1.993350 Val Loss: 1.9986
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [33280/49000 (68%)]
                                        Train Loss: 1.985238 Val Loss: 1.9697!
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [35840/49000 (73%)]
                                        Train Loss: 1.881185
                                                                 Val Loss: 1.95784
Saved model this time!
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [38400/49000 (78%)]
                                                                 Val Loss: 1.9365
                                        Train Loss: 1.940634
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [40960/49000 (83%)]
                                        Train Loss: 1.960648
                                                                Val Loss: 1.92324
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [43520/49000 (89%)]
                                        Train Loss: 1.910916 Val Loss: 1.90949
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 1 [46080/49000 (94%)]
                                        Train Loss: 1.876941
                                                                 Val Loss: 1.89832
Saved model this time!
torch.Size([256, 3, 224, 224])
tarch Siza/[256
                    224
```

```
LUICII.5128([230, 3, 224, 224])
Train Epoch: 1 [48640/49000 (99%)]
                                        Train Loss: 1.894295 Val Loss: 1.88310
Saved model this time!
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [0/49000 (0%)]
                                Train Loss: 2.264203 Val Loss: 1.868223
                                                                                ١
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [2560/49000 (5%)]
                                        Train Loss: 1.754707 Val Loss: 1.7538
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [5120/49000 (10%)]
                                        Train Loss: 1.718418 Val Loss: 1.7133
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [7680/49000 (16%)]
                                        Train Loss: 1.744086 Val Loss: 1.6954
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [10240/49000 (21%)]
                                        Train Loss: 1.659988 Val Loss: 1.66590
Saved model this time!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [12800/49000 (26%)]
                                        Train Loss: 1.633620
                                                              Val Loss: 1.6205(
Saved model this time!
torch.Size([256, 3, 224, 224])
                                        Train Loss: 1.580879 Val Loss: 1.60210
Train Epoch: 2 [15360/49000 (31%)]
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [17920/49000 (36%)]
                                        Train Loss: 1.543830 Val Loss: 1.58542
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [20480/49000 (42%)]
                                        Train Loss: 1.535140 Val Loss: 1.5521
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [23040/49000 (47%)]
                                        Train Loss: 1.539677 Val Loss: 1.5357
Saved model this time!
torch.Size([256. 3. 224. 2241)
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [25600/49000 (52%)]
                                       Train Loss: 1.514537 Val Loss: 1.52752
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [28160/49000 (57%)]
                                        Train Loss: 1.472236 Val Loss: 1.49409
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [30720/49000 (62%)]
                                       Train Loss: 1.480706
                                                                Val Loss: 1.4867
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [33280/49000 (68%)]
                                       Train Loss: 1.478842 Val Loss: 1.45784
Saved model this time!
torch.Size([256, 3, 224, 224])
             2 [25040/40000 /72011
                                        Train Lagg. 1 405777
```

```
irain Loss: 1.495///
                                                              val LOSS: 1.4533.
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [38400/49000 (78%)]
                                       Train Loss: 1.439593 Val Loss: 1.43089
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [40960/49000 (83%)]
                                       Train Loss: 1.453828 Val Loss: 1.41929
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [43520/49000 (89%)]
                                       Train Loss: 1.430559 Val Loss: 1.4013
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [46080/49000 (94%)]
                                       Train Loss: 1.419873 Val Loss: 1.4018.
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 2 [48640/49000 (99%)]
                                        Train Loss: 1.345147 Val Loss: 1.36536
Saved model this time!
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [0/49000 (0%)]
                                Train Loss: 1.710355 Val Loss: 1.200107
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [2560/49000 (5%)]
                                        Train Loss: 1.179942 Val Loss: 1.1566
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [5120/49000 (10%)]
                                        Train Loss: 1.180391 Val Loss: 1.14082
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [7680/49000 (16%)]
                                        Train Loss: 1.110277 Val Loss: 1.1199
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [10240/49000 (21%)]
                                        Train Loss: 1.192789 Val Loss: 1.11339
Saved model this time!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch Size([256
                 3
                   224
                         2241)
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [12800/49000 (26%)]
                                       Train Loss: 1.145720 Val Loss: 1.0953!
Saved model this time!
torch.Size([256, 3, 224, 224])
                                        Train Loss: 1.088721 Val Loss: 1.07940
Train Epoch: 3 [15360/49000 (31%)]
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [17920/49000 (36%)]
                                        Train Loss: 1.023585 Val Loss: 1.0729
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [20480/49000 (42%)]
                                        Train Loss: 1.083450 Val Loss: 1.0594
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [23040/49000 (47%)]
                                       Train Loss: 1.088979 Val Loss: 1.0256!
Saved model this time!
```

```
torcn.Size([250, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [25600/49000 (52%)]
                                        Train Loss: 1.013652 Val Loss: 1.03509
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [28160/49000 (57%)]
                                        Train Loss: 1.063842 Val Loss: 1.04013
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [30720/49000 (62%)]
                                        Train Loss: 1.063924
                                                              Val Loss: 1.01389
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [33280/49000 (68%)]
                                        Train Loss: 0.989091 Val Loss: 1.0125
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [35840/49000 (73%)]
                                        Train Loss: 1.070152
                                                                Val Loss: 0.9871
Saved model this time!
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [38400/49000 (78%)]
                                        Train Loss: 0.992967
                                                              Val Loss: 0.9872
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [40960/49000 (83%)]
                                        Train Loss: 1.004160 Val Loss: 0.97180
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [43520/49000 (89%)]
                                        Train Loss: 0.943324 Val Loss: 0.9828
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [46080/49000 (94%)]
                                                              Val Loss: 0.9669!
                                        Train Loss: 0.966776
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 3 [48640/49000 (99%)]
                                        Train Loss: 0.923507
                                                                 Val Loss: 0.96580
Saved model this time!
```

```
Javea model this time.
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
                                                        Val Loss: 0.854531
Train Epoch: 4 [0/49000 (0%)]
                                Train Loss: 1.451052
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [2560/49000 (5%)]
                                        Train Loss: 0.826237 Val Loss: 0.8443
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [5120/49000 (10%)]
                                        Train Loss: 0.785215 Val Loss: 0.83420
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [7680/49000 (16%)]
                                        Train Loss: 0.893330 Val Loss: 0.8137
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [10240/49000 (21%)]
                                        Train Loss: 0.800676 Val Loss: 0.81176
Saved model this time!
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [12800/49000 (26%)]
                                        Train Loss: 0.818147 Val Loss: 0.8023!
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [15360/49000 (31%)]
                                        Train Loss: 0.721139 Val Loss: 0.8035!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [17920/49000 (36%)]
                                        Train Loss: 0.843299 Val Loss: 0.80512
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [20480/49000 (42%)]
                                        Train Loss: 0.828044 Val Loss: 0.77600
Saved model this time!
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.773511 Val Loss: 0.79370
Train Epoch: 4 [23040/49000 (47%)]
torch.Size([256, 3, 224, 224])
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torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [25600/49000 (52%)]
                                        Train Loss: 0.799816
                                                              Val Loss: 0.7955(
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [28160/49000 (57%)]
                                        Train Loss: 0.800147
                                                                Val Loss: 0.78266
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [30720/49000 (62%)]
                                        Train Loss: 0.804919 Val Loss: 0.76832
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [33280/49000 (68%)]
                                        Train Loss: 0.748627 Val Loss: 0.77733
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [35840/49000 (73%)]
                                        Train Loss: 0.736430 Val Loss: 0.7603
Saved model this time!
torch.Size([256, 3, 224, 224])
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torch Siza/[256
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```

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LUICII.JIZE([ZJU, J, ZZ4, ZZ4])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [38400/49000 (78%)]
                                        Train Loss: 0.780497
                                                                 Val Loss: 0.76578
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [40960/49000 (83%)]
                                        Train Loss: 0.780525 Val Loss: 0.75198
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [43520/49000 (89%)]
                                        Train Loss: 0.789780
                                                              Val Loss: 0.7435{
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [46080/49000 (94%)]
                                        Train Loss: 0.755024 Val Loss: 0.74869
torch.Size([256, 3, 224, 224])
Train Epoch: 4 [48640/49000 (99%)]
                                        Train Loss: 0.786358 Val Loss: 0.74070
Saved model this time!
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
                                Train Loss: 1.286258 Val Loss: 0.671899
Train Epoch: 5 [0/49000 (0%)]
Saved model this time!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [2560/49000 (5%)]
                                        Train Loss: 0.704010 Val Loss: 0.68759
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [5120/49000 (10%)]
                                        Train Loss: 0.684448
                                                                Val Loss: 0.68493
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [7680/49000 (16%)]
                                        Train Loss: 0.670135 Val Loss: 0.67608
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [10240/49000 (21%)]
                                        Train Loss: 0.694678 Val Loss: 0.66630
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [12800/49000 (26%)]
                                        Train Loss: 0.667496
                                                                 Val Loss: 0.67908
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
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```
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [15360/49000 (31%)]
                                        Train Loss: 0.718447 Val Loss: 0.66069
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [17920/49000 (36%)]
                                        Train Loss: 0.660147
                                                                 Val Loss: 0.6608!
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [20480/49000 (42%)]
                                        Train Loss: 0.713746
                                                                 Val Loss: 0.65698
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [23040/49000 (47%)]
                                        Train Loss: 0.634169
                                                                 Val Loss: 0.65150
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [25600/49000 (52%)]
                                        Train Loss: 0.673367
                                                                 Val Loss: 0.66026
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
tarch Siza/[256
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                         22/11
```

```
LUICII.3128([230, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [28160/49000 (57%)]
                                        Train Loss: 0.605540 Val Loss: 0.6448!
Saved model this time!
torch.Size([256, 3, 224, 224])
                                                                 Val Loss: 0.64962
Train Epoch: 5 [30720/49000 (62%)]
                                        Train Loss: 0.685648
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [33280/49000 (68%)]
                                        Train Loss: 0.720873 Val Loss: 0.65030
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [35840/49000 (73%)]
                                        Train Loss: 0.626212 Val Loss: 0.65279
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [38400/49000 (78%)]
                                        Train Loss: 0.599460
                                                                 Val Loss: 0.62518
Saved model this time!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
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torch.Size([256, 3, 224, 224])
Train Epoch: 5 [40960/49000 (83%)]
                                        Train Loss: 0.600873 Val Loss: 0.6464!
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [43520/49000 (89%)]
                                        Train Loss: 0.662426
                                                                Val Loss: 0.64298
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [46080/49000 (94%)]
                                        Train Loss: 0.622890 Val Loss: 0.63738
torch.Size([256, 3, 224, 224])
Train Epoch: 5 [48640/49000 (99%)]
                                        Train Loss: 0.679195 Val Loss: 0.64100
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
                                Train Loss: 1.220950 Val Loss: 0.592006
                                                                                 ١
Train Epoch: 6 [0/49000 (0%)]
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [2560/49000 (5%)]
                                        Train Loss: 0.618742 Val Loss: 0.6148!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
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torch.Size([256, 3, 224, 224])
Train Epoch: 6 [5120/49000 (10%)]
                                        Train Loss: 0.610966
                                                                 Val Loss: 0.61408
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [7680/49000 (16%)]
                                        Train Loss: 0.638846
                                                                 Val Loss: 0.5872!
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [10240/49000 (21%)]
                                        Train Loss: 0.475878 Val Loss: 0.60216
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [12800/49000 (26%)]
                                        Train Loss: 0.501741
                                                                Val Loss: 0.58198
Saved model this time!
torch.Size([256, 3, 224, 224])
                                                                 Val Loss: 0.58920
Train Epoch: 6 [15360/49000 (31%)]
                                        Train Loss: 0.610225
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
tarch Ciza/[256
                    224
```

```
LUICII.3128([230, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [17920/49000 (36%)]
                                        Train Loss: 0.620539 Val Loss: 0.5863
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [20480/49000 (42%)]
                                        Train Loss: 0.625561 Val Loss: 0.57916
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [23040/49000 (47%)]
                                        Train Loss: 0.574829
                                                                 Val Loss: 0.59628
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [25600/49000 (52%)]
                                        Train Loss: 0.596519 Val Loss: 0.57720
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [28160/49000 (57%)]
                                        Train Loss: 0.610399 Val Loss: 0.57920
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
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torch.Size([256, 3, 224, 224])
Train Epoch: 6 [30720/49000 (62%)]
                                        Train Loss: 0.523238 Val Loss: 0.5825
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [33280/49000 (68%)]
                                        Train Loss: 0.611538 Val Loss: 0.5990
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [35840/49000 (73%)]
                                        Train Loss: 0.536688
                                                                Val Loss: 0.58468
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [38400/49000 (78%)]
                                        Train Loss: 0.528233 Val Loss: 0.57351
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [40960/49000 (83%)]
                                        Train Loss: 0.694261 Val Loss: 0.5877
torch.Size([256, 3, 224, 224])
torch.Size([256. 3. 224. 2241)
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torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [43520/49000 (89%)]
                                       Train Loss: 0.507718 Val Loss: 0.5786
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [46080/49000 (94%)]
                                       Train Loss: 0.633168 Val Loss: 0.58401
torch.Size([256, 3, 224, 224])
Train Epoch: 6 [48640/49000 (99%)]
                                                               Val Loss: 0.57694
                                       Train Loss: 0.534841
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [0/49000 (0%)]
                               Train Loss: 1.168081
                                                       Val Loss: 0.562900
                                                                               ١
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [2560/49000 (5%)]
                                       Train Loss: 0.509286 Val Loss: 0.5762
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [5120/49000 (10%)]
                                       Train Loss: 0.563745 Val Loss: 0.5723
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
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TOTCH.51Ze([250, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [7680/49000 (16%)]
                                        Train Loss: 0.609990
                                                                Val Loss: 0.5645
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [10240/49000 (21%)]
                                        Train Loss: 0.595339 Val Loss: 0.56383
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.555606 Val Loss: 0.55570
Train Epoch: 7 [12800/49000 (26%)]
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [15360/49000 (31%)]
                                        Train Loss: 0.531984 Val Loss: 0.55863
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [17920/49000 (36%)]
                                        Train Loss: 0.619223 Val Loss: 0.55799
torch.Size([256, 3, 224, 224])
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torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [20480/49000 (42%)]
                                        Train Loss: 0.518251 Val Loss: 0.5594!
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [23040/49000 (47%)]
                                        Train Loss: 0.515852 Val Loss: 0.56148
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.551437 Val Loss: 0.55326
Train Epoch: 7 [25600/49000 (52%)]
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [28160/49000 (57%)]
                                        Train Loss: 0.575369 Val Loss: 0.55309
Saved model this time!
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.571790 Val Loss: 0.56361
Train Epoch: 7 [30720/49000 (62%)]
torch.Size([256, 3, 224, 224])
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                         2241)
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```
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [33280/49000 (68%)]
                                       Train Loss: 0.585769 Val Loss: 0.56293
torch.Size([256, 3, 224, 224])
                                       Train Loss: 0.615460 Val Loss: 0.55574
Train Epoch: 7 [35840/49000 (73%)]
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [38400/49000 (78%)]
                                       Train Loss: 0.524053 Val Loss: 0.5522
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [40960/49000 (83%)]
                                       Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [43520/49000 (89%)]
                                       Train Loss: 0.556451 Val Loss: 0.5558!
torch.Size([256, 3, 224, 224])
```

```
torcn.Size([250, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.543320 Val Loss: 0.54632
Train Epoch: 7 [46080/49000 (94%)]
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 7 [48640/49000 (99%)]
                                        Train Loss: 0.552232
                                                                Val Loss: 0.55240
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
                                                        Val Loss: 0.550710
                                                                                 ١
Train Epoch: 8 [0/49000 (0%)]
                                Train Loss: 1.160033
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [2560/49000 (5%)]
                                        Train Loss: 0.578554 Val Loss: 0.55692
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [5120/49000 (10%)]
                                        Train Loss: 0.554731 Val Loss: 0.5386
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [7680/49000 (16%)]
                                        Train Loss: 0.531410 Val Loss: 0.5383
Saved model this time!
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
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torch.Size([256, 3, 224, 224])
Train Epoch: 8 [10240/49000 (21%)]
                                        Train Loss: 0.572418 Val Loss: 0.5340!
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [12800/49000 (26%)]
                                        Train Loss: 0.526520 Val Loss: 0.5384
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [15360/49000 (31%)]
                                        Train Loss: 0.440627 Val Loss: 0.5373!
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [17920/49000 (36%)]
                                        Train Loss: 0.531224 Val Loss: 0.53878
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [20480/49000 (42%)]
                                        Train Loss: 0.545627 Val Loss: 0.56376
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
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                         2241)
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COLCIIIOTEC([500, 0, 554, 554])
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [23040/49000 (47%)]
                                        Train Loss: 0.521682 Val Loss: 0.53179
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [25600/49000 (52%)]
                                        Train Loss: 0.598017 Val Loss: 0.54031
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [28160/49000 (57%)]
                                        Train Loss: 0.511356 Val Loss: 0.55503
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [30720/49000 (62%)]
                                        Train Loss: 0.539674 Val Loss: 0.5455
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [33280/49000 (68%)]
                                        Train Loss: 0.579209 Val Loss: 0.54890
torch.Size([256, 3, 224, 224])
```

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torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [35840/49000 (73%)]
                                        Train Loss: 0.570621 Val Loss: 0.53323
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [38400/49000 (78%)]
                                        Train Loss: 0.563569
                                                                 Val Loss: 0.53626
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [40960/49000 (83%)]
                                        Train Loss: 0.606820
                                                              Val Loss: 0.5348!
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [43520/49000 (89%)]
                                        Train Loss: 0.567834 Val Loss: 0.53423
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [46080/49000 (94%)]
                                        Train Loss: 0.460869 Val Loss: 0.54089
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 8 [48640/49000 (99%)]
                                        Train Loss: 0.557109 Val Loss: 0.5298.
Saved model this time!
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [0/49000 (0%)]
                                Train Loss: 1.024197
                                                        Val Loss: 0.540059
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [2560/49000 (5%)]
                                        Train Loss: 0.573959 Val Loss: 0.5425
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [5120/49000 (10%)]
                                        Train Loss: 0.588238 Val Loss: 0.52820
Saved model this time!
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.543430 Val Loss: 0.54201
Train Epoch: 9 [7680/49000 (16%)]
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [10240/49000 (21%)]
                                        Train Loss: 0.545229 Val Loss: 0.54300
torch.Size([256, 3, 224, 224])
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tarch Size/[256
                         22/11
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LUICII.JIZE([ZJU, J, ZZ4, ZZ4])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [12800/49000 (26%)]
                                        Train Loss: 0.584119 Val Loss: 0.53939
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [15360/49000 (31%)]
                                        Train Loss: 0.560484
                                                                Val Loss: 0.54623
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [17920/49000 (36%)]
                                        Train Loss: 0.567873 Val Loss: 0.53394
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [20480/49000 (42%)]
                                        Train Loss: 0.570251 Val Loss: 0.5319
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [23040/49000 (47%)]
                                        Train Loss: 0.517306 Val Loss: 0.52932
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [25600/49000 (52%)]
                                        Train Loss: 0.591225
                                                              Val Loss: 0.5248
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [28160/49000 (57%)]
                                        Train Loss: 0.517387 Val Loss: 0.5186
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [30720/49000 (62%)]
                                        Train Loss: 0.570756 Val Loss: 0.53143
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [33280/49000 (68%)]
                                        Train Loss: 0.525327 Val Loss: 0.52698
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [35840/49000 (73%)]
                                        Train Loss: 0.556059 Val Loss: 0.52970
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [38400/49000 (78%)]
                                        Train Loss: 0.467461
                                                               Val Loss: 0.53602
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [40960/49000 (83%)]
                                        Train Loss: 0.574416
                                                                Val Loss: 0.5218:
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [43520/49000 (89%)]
                                        Train Loss: 0.516058
                                                              Val Loss: 0.53232
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [46080/49000 (94%)]
                                        Train Loss: 0.546336
                                                                 Val Loss: 0.52500
torch.Size([256, 3, 224, 224])
Train Epoch: 9 [48640/49000 (99%)]
                                        Train Loss: 0.550028 Val Loss: 0.5336!
torch.Size([104, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [0/49000 (0%)]
                                Train Loss: 0.964872
                                                        Val Loss: 0.542323
Saved model this time!
torch.Size([256, 3, 224, 224])
tarch Siza/[256
                 2 22/
                         22/11
```

```
LUICII.3128([230, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [2560/49000 (5%)]
                                        Train Loss: 0.526727 Val Loss: 0.5401
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [5120/49000 (10%)]
                                        Train Loss: 0.501159 Val Loss: 0.5414!
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [7680/49000 (16%)]
                                        Train Loss: 0.560064 Val Loss: 0.5377!
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [10240/49000 (21%)]
                                        Train Loss: 0.537552 Val Loss: 0.52090
Saved model this time!
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [12800/49000 (26%)]
                                        Train Loss: 0.536036 Val Loss: 0.53203
torch.Size([256, 3, 224, 224])
```

```
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [15360/49000 (31%)]
                                        Train Loss: 0.552141 Val Loss: 0.5256
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [17920/49000 (36%)]
                                        Train Loss: 0.514192 Val Loss: 0.54114
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [20480/49000 (42%)]
                                        Train Loss: 0.568552 Val Loss: 0.5427.
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.523196 Val Loss: 0.53827
Train Epoch: 10 [23040/49000 (47%)]
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [25600/49000 (52%)]
                                        Train Loss: 0.595694 Val Loss: 0.53522
torch.Size([256, 3, 224, 224])
```

```
Train Epoch: 10 [28160/49000 (57%)]
                                        Train Loss: 0.527919
                                                              Val Loss: 0.52443
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [30720/49000 (62%)]
                                        Train Loss: 0.488037 Val Loss: 0.5217.
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [33280/49000 (68%)]
                                        Train Loss: 0.449341 Val Loss: 0.52978
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [35840/49000 (73%)]
                                        Train Loss: 0.530765
                                                              Val Loss: 0.53022
torch.Size([256, 3, 224, 224])
                                        Train Loss: 0.553016 Val Loss: 0.5390
Train Epoch: 10 [38400/49000 (78%)]
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [40960/49000 (83%)]
                                        Train Loss: 0.533984
                                                                Val Loss: 0.5328
torch.Size([256, 3, 224, 224])
tarch Ciza/[256
                 2
                    224
```

```
LUICII.3128([230, 3, 224, 224])
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [43520/49000 (89%)]
                                        Train Loss: 0.583151 Val Loss: 0.53158
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [46080/49000 (94%)]
                                        Train Loss: 0.548435 Val Loss: 0.53709
torch.Size([256, 3, 224, 224])
Train Epoch: 10 [48640/49000 (99%)]
                                        Train Loss: 0.649393 Val Loss: 0.52628
torch.Size([104, 3, 224, 224])
test set: Average loss: 0.5452, Accuracy: 8566/10000 (86%)
(tensor(0.5452, device='cuda:0'), tensor(85.6600))
```