```
pip install torchviz
Collecting torchviz
  Downloading torchviz-0.0.2.tar.gz (4.9 kB)
  Preparing metadata (setup.py) ... ent already satisfied: torch in
/usr/local/lib/python3.10/dist-packages (from torchviz) (2.1.0+cu118)
Requirement already satisfied: graphviz in
/usr/local/lib/python3.10/dist-packages (from torchviz) (0.20.1)
Requirement already satisfied: filelock in
/usr/local/lib/python3.10/dist-packages (from torch->torchviz)
(3.13.1)
Requirement already satisfied: typing-extensions in
/usr/local/lib/python3.10/dist-packages (from torch->torchviz) (4.5.0)
Requirement already satisfied: sympy in
/usr/local/lib/python3.10/dist-packages (from torch->torchviz) (1.12)
Requirement already satisfied: networkx in
/usr/local/lib/python3.10/dist-packages (from torch->torchviz) (3.2.1)
Requirement already satisfied: jinja2 in
/usr/local/lib/python3.10/dist-packages (from torch->torchviz) (3.1.2)
Requirement already satisfied: fsspec in
/usr/local/lib/python3.10/dist-packages (from torch->torchviz)
(2023.6.0)
Requirement already satisfied: triton==2.1.0 in
/usr/local/lib/python3.10/dist-packages (from torch->torchviz) (2.1.0)
Requirement already satisfied: MarkupSafe>=2.0 in
/usr/local/lib/python3.10/dist-packages (from jinja2->torch->torchviz)
(2.1.3)
Requirement already satisfied: mpmath>=0.19 in
/usr/local/lib/python3.10/dist-packages (from sympy->torch->torchviz)
(1.3.0)
Building wheels for collected packages: torchviz
  Building wheel for torchviz (setup.py) ... e=torchviz-0.0.2-py3-
none-any.whl size=4133
sha256=b8d750eb8290d20c86e90cb55bc60b83bc41974fa622440a0e960df35c3acd4
  Stored in directory:
/root/.cache/pip/wheels/4c/97/88/a02973217949e0db0c9f4346d154085f4725f
99c4f15a87094
Successfully built torchviz
Installing collected packages: torchviz
Successfully installed torchviz-0.0.2
import multiprocessing as mp
import time
from typing import Any, Generator, Literal
import os
import torch
import torchvision
import torchvision.datasets
import torchvision.transforms
```

```
import torch.utils.data
import torch.nn
from torchvision import models as torch model
import seaborn as sns
from torchviz import make dot
from IPython.display import display
import time
from torchsummary import summary
from matplotlib import pyplot as plot
device = torch.device("cuda:0" if torch.cuda.is available() else
"cpu")
device
device(type='cuda', index=0)
CLASSES = ['plane', 'car', 'bird', 'cat', 'deer', 'dog', 'frog',
'horse', 'ship', 'truck']
N CLASSES = len(CLASSES)
def show images(images, title):
    num showed imgs x = 10
    num showed imgs y = 10
    figsize = (10, 10)
    fig, axes = plot.subplots(num showed imgs y, num showed imgs x,
figsize = figsize)
    fig.suptitle(title)
    plot.setp(plot.gcf().get axes(), xticks = [], yticks = [])
    dataiter = iter(images)
    for i, ax in enumerate(axes.flat):
        img = images[i][0].numpy().transpose(1, 2, 0)
        ax.imshow(img)
        ax.text(1, 1, CLASSES[images[i][1]], bbox=dict(fill=False,
edgecolor='red', linewidth=2))
def download data(transforms, batch size):
 dir_name = os.getcwd()
  train dataset = torchvision.datasets.CIFAR10(
      root = dir name, train = True, download = True,
      transform = transforms
  test dataset = torchvision.datasets.CIFAR10(
      root = dir name, train = False, download = True,
      transform = transforms
  )
  print('Number of train samples: {}'.format(len(train dataset)))
  show images(train dataset, 'Train samples')
```

```
print('Number of test samples: {}'.format(len(test dataset)))
  show_images(test_dataset, 'Test samples')
  train data loader = torch.utils.data.DataLoader(
      train dataset, batch size = batch size, shuffle = True
  test data loader = torch.utils.data.DataLoader(
      test dataset, batch size = batch size, shuffle = False
  return train data loader, test data loader
num epochs = 20
def get accuracy(data loader, model):
    tp = 0
    n = 0
    with torch.no grad():
        for images, labels in data_loader:
            labels = labels.to(device)
            images = images.to(device)
            outputs = model(images)
            _, predicted = torch.max(outputs.data, 1)
            n += labels.size(0)
            tp += (predicted == labels).sum()
    return tp / n
def train loop(model, train dataloader, optimizer, loss function):
  start all = time.time()
  for epoch in range(num epochs):
      start = time.time()
      for images, labels in train dataloader:
          images = images.to(device)
          labels = labels.to(device)
          outputs = model(images)
          loss = loss function(outputs.type(torch.float32),
torch.nn.functional.one hot(labels,
num classes=10).type(torch.float32))
          optimizer.zero grad()
          loss.backward()
          optimizer.step()
      end = time.time()
      print('Epoch[{}]: accuracy = {}, time = {}'.format(epoch,
get accuracy(train_dataloader, model), (end - start)))
  end all = time.time()
  print('train time = {}'.format((end all - start all)))
  # vield model
```

```
def test_model(model, test_dataloader):
   model.eval()
   print('Test accuracy: {}'.format(get_accuracy(test_dataloader,
model)))
   # yield model
```

Modifications:

```
def last_layer(in_features):
    return torch.nn.Linear(in_features, 10)

def last_sequental_layer(in_features):
    return torch.nn.Sequential(
        torch.nn.Linear(in_features, in_features//2),
        torch.nn.ReLU(),
        torch.nn.Linear(in_features//2, 10),
    )

def init_weights(model):
    print(model)
    if type(model) == torch.nn.Linear:
        torch.nn.init.xavier_uniform_(model.weight)
        print(model.weight)
```

- 1. ResNet18
- 1. Single layer

```
model = torch_model.resnet18(torch_model.ResNet18 Weights.DEFAULT)
/usr/local/lib/python3.10/dist-packages/torchvision/models/
utils.py:135: UserWarning: Using 'weights' as positional parameter(s)
is deprecated since 0.13 and may be removed in the future. Please use
keyword parameter(s) instead.
 warnings.warn(
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
Модель "CNN":
       Layer (type)
                                  Output Shape
           Conv2d-1
                              [-1, 64, 16, 16]
                                                        9,408
      BatchNorm2d-2
                              [-1, 64, 16, 16]
                                                         128
                              [-1, 64, 16, 16]
             ReLU-3
                                                             0
```

MaxPool2d-4	[-1, 64, 8, 8]	0
Conv2d-5	[-1, 64, 8, 8]	36,864
BatchNorm2d-6	[-1, 64, 8, 8]	128
ReLU-7	[-1, 64, 8, 8]	0
Conv2d-8	[-1, 64, 8, 8]	36,864
BatchNorm2d-9 ReLU-10	[-1, 64, 8, 8]	128
BasicBlock-11	[-1, 64, 8, 8]	0 0
Conv2d-12	[-1, 64, 8, 8] [-1, 64, 8, 8]	36,864
BatchNorm2d-13	[-1, 64, 8, 8]	128
ReLU-14	[-1, 64, 8, 8]	0
Conv2d - 15	[-1, 64, 8, 8]	36,864
BatchNorm2d-16	[-1, 64, 8, 8]	128
ReLU-17	[-1, 64, 8, 8]	0
BasicBlock-18	[-1, 64, 8, 8]	ő
Conv2d - 19	[-1, 128, 4, 4]	73,728
BatchNorm2d-20	[-1, 128, 4, 4]	256
ReLU-21	[-1, 128, 4, 4]	0
Conv2d-22	[-1, 128, 4, 4]	147,456
BatchNorm2d-23	[-1, 128, 4, 4]	256
Conv2d-24	[-1, 128, 4, 4]	8,192
BatchNorm2d-25	[-1, 128, 4, 4]	256
ReLU-26	[-1, 128, 4, 4]	0
BasicBlock-27	[-1, 128, 4, 4]	0
Conv2d-28	[-1, 128, 4, 4]	147,456
BatchNorm2d-29	[-1, 128, 4, 4]	256
ReLU-30	[-1, 128, 4, 4]	0
Conv2d-31	[-1, 128, 4, 4]	147,456
BatchNorm2d-32	[-1, 128, 4, 4]	256
ReLU-33	[-1, 128, 4, 4]	0
BasicBlock-34	[-1, 128, 4, 4]	204 012
Conv2d-35 BatchNorm2d-36	[-1, 256, 2, 2] [-1, 256, 2, 2]	294,912 512
ReLU-37	[-1, 250, 2, 2]	0
Conv2d-38	[-1, 256, 2, 2]	589,824
BatchNorm2d-39	[-1, 256, 2, 2]	512
Conv2d-40	[-1, 256, 2, 2]	32,768
BatchNorm2d-41	[-1, 256, 2, 2]	512
ReLU-42	[-1, 256, 2, 2]	0
BasicBlock-43	[-1, 256, 2, 2]	0
Conv2d-44	[-1, 256, 2, 2]	589,824
BatchNorm2d-45	[-1, 256, 2, 2]	512
ReLU-46	[-1, 256, 2, 2]	0
Conv2d-47	[-1, 256, 2, 2]	589,824
BatchNorm2d-48	[-1, 256, 2, 2]	512
ReLU-49	[-1, 256, 2, 2]	0
BasicBlock-50	[-1, 256, 2, 2]	0
Conv2d-51	[-1, 512, 1, 1]	1,179,648
BatchNorm2d-52	[-1, 512, 1, 1]	1,024

```
ReLU-53
                                 [-1, 512, 1, 1]
           Conv2d-54
                                 [-1, 512, 1, 1]
                                                        2,359,296
      BatchNorm2d-55
                                 [-1, 512, 1, 1]
                                                            1,024
                                 [-1, 512, 1, 1]
                                                          131,072
           Conv2d-56
                                                            1,024
      BatchNorm2d-57
                                 [-1, 512, 1, 1]
                                 [-1, 512, 1, 1]
             ReLU-58
                                                                0
                                 [-1, 512, 1, 1]
                                                                0
       BasicBlock-59
           Conv2d-60
                                 [-1, 512, 1, 1]
                                                        2,359,296
      BatchNorm2d-61
                                 [-1, 512, 1, 1]
                                                            1,024
             ReLU-62
                                 [-1, 512, 1, 1]
                                                                0
                                                       2,359,296
                                 [-1, 512, 1, 1]
           Conv2d-63
                                                           1,024
      BatchNorm2d-64
                                 [-1, 512, 1, 1]
                                 [-1, 512, 1, 1]
                                                                0
             ReLU-65
                                 [-1, 512, 1, 1]
                                                                0
       BasicBlock-66
AdaptiveAvgPool2d-67
                                 [-1, 512, 1, 1]
                                                                0
                                     [-1, 1000]
           Linear-68
                                                          513,000
Total params: 11,689,512
Trainable params: 11,689,512
Non-trainable params: 0
Input size (MB): 0.01
Forward/backward pass size (MB): 1.29
```

Params size (MB): 44.59

Визуализация модели:

Estimated Total Size (MB): 45.90

```
for param in model.parameters():
    param.requires_grad = False
model.fc = last_layer(512)
model = model.to(device)

transforms = torch_model.ResNet18_Weights.DEFAULT.transforms()
batch_size = 128
train_dataloader, test_dataloader = download_data(transforms, batch_size)

Files already downloaded and verified
Files already downloaded and verified
Number of train samples: 50000
```

WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Number of test samples: 10000

WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Train samples



Test samples



```
learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning_rate)
train_loop(model, train_dataloader, optimizer, loss_function)
Epoch[0]: accuracy = 0.7552199959754944, time = 134.36429405212402
Epoch[1]: accuracy = 0.7705399990081787, time = 133.38727378845215
```

```
Epoch[2]: accuracy = 0.7748799920082092, time = 134.6206178665161
Epoch[3]: accuracy = 0.7821799516677856, time = 133.35160636901855
Epoch[4]: accuracy = 0.7877799868583679, time = 132.4030101299286
Epoch[5]: accuracy = 0.7893799543380737, time = 130.85477805137634
Epoch[6]: accuracy = 0.7927199602127075, time = 132.90481090545654
Epoch[7]: accuracy = 0.7902599573135376, time = 133.37981700897217
Epoch[8]: accuracy = 0.792739987373352, time = 138.0912709236145
Epoch[9]: accuracy = 0.7948399782180786, time = 140.2856011390686
Epoch[10]: accuracy = 0.7905199527740479, time = 133.56772112846375
Epoch[11]: accuracy = 0.7960399985313416, time = 132.52500438690186
Epoch[12]: accuracy = 0.7979399561882019, time = 131.02950859069824
Epoch[13]: accuracy = 0.799519956111908, time = 132.790180683136
Epoch[14]: accuracy = 0.800279974937439, time = 132.8993923664093
Epoch[15]: accuracy = 0.800059974193573, time = 133.1646978855133
Epoch[16]: accuracy = 0.8001199960708618, time = 133.71345162391663
Epoch[17]: accuracy = 0.7939800024032593, time = 133.75645112991333
Epoch[18]: accuracy = 0.8032400012016296, time = 132.00247764587402
Epoch[19]: accuracy = 0.8007599711418152, time = 130.65841007232666
train time = 5328.962197542191
test model(model, test dataloader)
Test accuracy: 0.7849000096321106
print(model)
ResNet(
  (conv1): Conv2d(3, 64, kernel size=(7, 7), stride=(2, 2),
padding=(3, 3), bias=False)
  (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (relu): ReLU(inplace=True)
  (maxpool): MaxPool2d(kernel size=3, stride=2, padding=1, dilation=1,
ceil mode=False)
  (layer1): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
```

```
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (layer2): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 128, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (downsample): Sequential(
        (0): Conv2d(64, 128, \text{kernel size}=(1, 1), \text{stride}=(2, 2),
bias=False)
        (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (layer3): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(128, 256, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
```

```
(downsample): Sequential(
        (0): Conv2d(128, 256, kernel size=(1, 1), stride=(2, 2),
bias=False)
        (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (layer4): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(256, 512, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=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 stats=True)
      )
    (1): BasicBlock(
      (conv1): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(512, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  )
```

```
(avgpool): AdaptiveAvgPool2d(output_size=(1, 1))
  (fc): Linear(in_features=512, out_features=10, bias=True)
)
```

1. Sequental layer

```
model = torch_model.resnet18(torch_model.ResNet18_Weights.DEFAULT)

/usr/local/lib/python3.10/dist-packages/torchvision/models/
_utils.py:135: UserWarning: Using 'weights' as positional parameter(s)
is deprecated since 0.13 and may be removed in the future. Please use
keyword parameter(s) instead.
    warnings.warn(

# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input_size=(3, 32, 32))

print("Визуализация модели:")
display(make_dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named_parameters())))
```

Модель "CNN":

Output Shape	Param #
[-1, 64, 16, 16] [-1, 64, 16, 16] [-1, 64, 16, 16] [-1, 64, 8, 8]	Param #
[-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4]	73,726 256 0 147,456 256 8,192
	[-1, 64, 16, 16] [-1, 64, 16, 16] [-1, 64, 16, 16] [-1, 64, 8, 8] [-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4]

BatchNorm2d-25	[-1, 128, 4, 4]	256
ReLU-26	[-1, 128, 4, 4]	0
BasicBlock-27	[-1, 128, 4, 4]	0
Conv2d-28	[-1, 128, 4, 4]	147,456
BatchNorm2d-29	[-1, 128, 4, 4]	256
ReLU-30	[-1, 128, 4, 4]	147.456
Conv2d-31	[-1, 128, 4, 4]	147,456
BatchNorm2d-32 ReLU-33	[-1, 128, 4, 4]	256
BasicBlock-34	[-1, 128, 4, 4] [-1, 128, 4, 4]	0 0
Conv2d-35	[-1, 126, 4, 4]	294,912
BatchNorm2d-36	[-1, 256, 2, 2]	512
ReLU-37	[-1, 256, 2, 2]	0
Conv2d-38	[-1, 256, 2, 2]	589,824
BatchNorm2d-39	[-1, 256, 2, 2]	512
Conv2d-40	[-1, 256, 2, 2]	32,768
BatchNorm2d-41	[-1, 256, 2, 2]	512
ReLU-42	[-1, 256, 2, 2]	0
BasicBlock-43	[-1, 256, 2, 2]	0
Conv2d-44	[-1, 256, 2, 2]	589,824
BatchNorm2d-45	[-1, 256, 2, 2]	512
ReLU-46	[-1, 256, 2, 2]	0
Conv2d - 47	[-1, 256, 2, 2]	589,824
BatchNorm2d-48	[-1, 256, 2, 2]	512
ReLU-49	[-1, 256, 2, 2]	0
BasicBlock-50	[-1, 256, 2, 2]	1 170 640
Conv2d-51	[-1, 512, 1, 1]	1,179,648
BatchNorm2d-52 ReLU-53	[-1, 512, 1, 1]	1,024
Conv2d-54	[-1, 512, 1, 1] [-1, 512, 1, 1]	2,359,296
BatchNorm2d-55	[-1, 512, 1, 1]	1,024
Conv2d-56	[-1, 512, 1, 1]	131,072
BatchNorm2d-57	[-1, 512, 1, 1]	1,024
ReLU-58	[-1, 512, 1, 1]	0
BasicBlock-59	[-1, 512, 1, 1]	0
Conv2d-60	[-1, 512, 1, 1]	2,359,296
BatchNorm2d-61	[-1, 512, 1, 1]	1,024
ReLU-62	[-1, 512, 1, 1]	0
Conv2d-63	[-1, 512, 1, 1]	2,359,296
BatchNorm2d-64	[-1, 512, 1, 1]	1,024
ReLU-65	[-1, 512, 1, 1]	0
BasicBlock-66	[-1, 512, 1, 1]	0
AdaptiveAvgPool2d-67 Linear-68	[-1, 512, 1, 1] [-1, 1000]	0 513 000
Lilleal - 06	[-1, 1000]	513,000

Total params: 11,689,512
Trainable params: 11,689,512
Non-trainable params: 0

Input size (MB): 0.01

Forward/backward pass size (MB): 1.29

Params size (MB): 44.59

Estimated Total Size (MB): 45.90

Визуализация модели:

```
for param in model.parameters():
    param.requires grad = False
model.fc = last sequental layer(512)
model = model.to(device)
print(model)
ResNet(
  (conv1): Conv2d(3, 64, kernel_size=(7, 7), stride=(2, 2),
padding=(3, 3), bias=False)
  (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (relu): ReLU(inplace=True)
  (maxpool): MaxPool2d(kernel size=3, stride=2, padding=1, dilation=1,
ceil mode=False)
  (layer1): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
    )
  (layer2): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 128, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
```

```
(downsample): Sequential(
        (0): Conv2d(64, 128, \text{kernel size}=(1, 1), \text{stride}=(2, 2),
bias=False)
        (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (layer3): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(128, 256, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(256, 256, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (downsample): Sequential(
        (0): Conv2d(128, 256, kernel_size=(1, 1), stride=(2, 2),
bias=False)
        (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      )
    (1): BasicBlock(
      (conv1): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  )
```

```
(layer4): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(256, 512, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(512, 512, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=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 stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(512, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(512, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (avgpool): AdaptiveAvgPool2d(output size=(1, 1))
  (fc): Sequential(
    (0): Linear(in features=512, out features=256, bias=True)
    (1): ReLU()
    (2): Linear(in features=256, out features=10, bias=True)
  )
transforms = torch model.ResNet18 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
Files already downloaded and verified
Files already downloaded and verified
Number of train samples: 50000
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
```

WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Number of test samples: 10000

WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Train samples



Test samples



```
learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning_rate)
train_loop(model, train_dataloader, optimizer, loss_function)
Epoch[0]: accuracy = 0.7584599852561951, time = 133.47190737724304
Epoch[1]: accuracy = 0.7836199998855591, time = 134.56931257247925
```

```
Epoch[2]: accuracy = 0.7912200093269348, time = 131.76579809188843
Epoch[3]: accuracy = 0.7971199750900269, time = 132.62603068351746
Epoch[4]: accuracy = 0.7963399887084961, time = 132.72239804267883
Epoch[5]: accuracy = 0.8087999820709229, time = 132.0856158733368
Epoch[6]: accuracy = 0.817799985408783, time = 133.8889524936676
Epoch[7]: accuracy = 0.8195199966430664, time = 133.67858791351318
Epoch[8]: accuracy = 0.8260399699211121, time = 134.11023926734924
Epoch[9]: accuracy = 0.8386799693107605, time = 134.24782919883728
Epoch[10]: accuracy = 0.8505199551582336, time = 133.9661843776703
Epoch[11]: accuracy = 0.8570999503135681, time = 133.6306209564209
Epoch[12]: accuracy = 0.8524199724197388, time = 136.24574303627014
Epoch[13]: accuracy = 0.8709399700164795, time = 133.86288475990295
Epoch[14]: accuracy = 0.8695600032806396, time = 132.84558701515198
Epoch[15]: accuracy = 0.870959997177124, time = 134.21223258972168
Epoch[16]: accuracy = 0.8806599974632263, time = 133.45714831352234
Epoch[17]: accuracy = 0.878879964351654, time = 133.20917320251465
Epoch[18]: accuracy = 0.8874199986457825, time = 133.56253910064697
Epoch[19]: accuracy = 0.8967399597167969, time = 134.24799370765686
train time = 5326.537646770477
test model(model, test dataloader)
Test accuracy: 0.7937999963760376
```

1. Weights

```
model = torch model.resnet18(torch model.ResNet18 Weights.DEFAULT)
/usr/local/lib/python3.10/dist-packages/torchvision/models/
_utils.py:135: UserWarning: Using 'weights' as positional parameter(s)
is deprecated since 0.13 and may be removed in the future. Please use
keyword parameter(s) instead.
 warnings.warn(
Downloading: "https://download.pytorch.org/models/resnet18-
f37072fd.pth" to /root/.cache/torch/hub/checkpoints/resnet18-
f37072fd.pth
100%
         | 44.7M/44.7M [00:00<00:00, 85.8MB/s]
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
Модель "CNN":
                                Output Shape
       Layer (type)
                                                    Param #
______
                            [-1, 64, 16, 16]
           Conv2d-1
                                                     9,408
```

BatchNorm2d-2	[-1, 64, 16, 16]	128
ReLU-3	[-1, 64, 16, 16]	0
MaxPool2d-4	[-1, 64, 8, 8]	0
Conv2d-5	[-1, 64, 8, 8]	36,864
BatchNorm2d-6	[-1, 64, 8, 8]	128
ReLU-7	[-1, 64, 8, 8]	Θ
Conv2d-8	[-1, 64, 8, 8]	36,864
BatchNorm2d-9	[-1, 64, 8, 8]	128
ReLU-10	[-1, 64, 8, 8]	0
BasicBlock-11	[-1, 64, 8, 8]	0
Conv2d-12	[-1, 64, 8, 8]	36,864
BatchNorm2d-13	[-1, 64, 8, 8]	128
ReLU-14	[-1, 64, 8, 8]	0
Conv2d - 15	[-1, 64, 8, 8]	36,864
BatchNorm2d-16	[-1, 64, 8, 8]	128
ReLU-17	[-1, 64, 8, 8]	0
BasicBlock-18	[-1, 64, 8, 8]	0
Conv2d - 19	[-1, 128, 4, 4]	73,728
BatchNorm2d-20	[-1, 128, 4, 4]	256
ReLU-21	[-1, 128, 4, 4]	0
Conv2d - 22	[-1, 128, 4, 4]	147,456
BatchNorm2d-23	[-1, 128, 4, 4]	256
Conv2d - 24	[-1, 128, 4, 4]	8,192
BatchNorm2d-25	[-1, 128, 4, 4]	256
ReLU-26		0
BasicBlock-27	[-1, 128, 4, 4] [-1, 128, 4, 4]	0
Conv2d-28		
BatchNorm2d-29		147,456 256
ReLU-30	[-1, 128, 4, 4]	0
Conv2d-31	[-1, 128, 4, 4] [-1, 128, 4, 4]	_
BatchNorm2d-32		147,456 256
ReLU-33	[-1, 128, 4, 4]	
	[-1, 128, 4, 4]	0
BasicBlock-34	[-1, 128, 4, 4]	0
Conv2d - 35	[-1, 256, 2, 2]	294,912
BatchNorm2d-36	[-1, 256, 2, 2]	512
ReLU-37	[-1, 256, 2, 2]	0
Conv2d - 38	[-1, 256, 2, 2]	589,824
BatchNorm2d-39	[-1, 256, 2, 2]	512
Conv2d - 40	[-1, 256, 2, 2]	32,768
BatchNorm2d-41	[-1, 256, 2, 2]	512
ReLU-42	[-1, 256, 2, 2]	0
BasicBlock-43	[-1, 256, 2, 2]	0
Conv2d - 44	[-1, 256, 2, 2]	589,824
BatchNorm2d-45	[-1, 256, 2, 2]	512
ReLU-46	[-1, 256, 2, 2]	0
Conv2d - 47	[-1, 256, 2, 2]	589,824
BatchNorm2d-48	[-1, 256, 2, 2]	512
ReLU-49	[-1, 256, 2, 2]	0
BasicBlock-50	[-1, 256, 2, 2]	0

```
Conv2d-51
                                 [-1, 512, 1, 1]
                                                        1,179,648
                                 [-1, 512, 1, 1]
      BatchNorm2d-52
                                                            1,024
             ReLU-53
                                 [-1, 512, 1, 1]
                                                                0
                                                        2,359,296
                                 [-1, 512, 1, 1]
           Conv2d-54
      BatchNorm2d-55
                                 [-1, 512, 1, 1]
                                                            1,024
                                 [-1, 512, 1, 1]
           Conv2d-56
                                                          131,072
      BatchNorm2d-57
                                 [-1, 512, 1, 1]
                                                            1,024
             ReLU-58
                                 [-1, 512, 1, 1]
                                                                0
                                 [-1, 512, 1, 1]
       BasicBlock-59
                                                                0
           Conv2d-60
                                 [-1, 512, 1, 1]
                                                        2,359,296
                                 [-1, 512, 1, 1]
      BatchNorm2d-61
                                                            1,024
             ReLU-62
                                 [-1, 512, 1, 1]
                                 [-1, 512, 1, 1]
                                                        2,359,296
           Conv2d-63
      BatchNorm2d-64
                                 [-1, 512, 1, 1]
                                                           1,024
             ReLU-65
                                 [-1, 512, 1, 1]
                                                                0
       BasicBlock-66
                                 [-1, 512, 1, 1]
                                                                0
                                 [-1, 512, 1, 1]
                                                                0
AdaptiveAvgPool2d-67
                                 [-1, 1000]
           Linear-68
                                                          513,000
Total params: 11,689,512
Trainable params: 11,689,512
Non-trainable params: 0
Input size (MB): 0.01
Forward/backward pass size (MB): 1.29
```

Params size (MB): 44.59

Визуализация модели:

Estimated Total Size (MB): 45.90

```
for param in model.parameters():
    param.requires grad = False
model.fc = last layer(512)
model.fc.apply(init weights)
model = model.to(device)
Linear(in features=512, out features=10, bias=True)
Parameter containing:
                   0.0083, -0.1061,
tensor([[-0.1054,
                                     . . . ,
                                           0.0543, -0.0405,
                                                              0.02251,
                            0.0152,
        [ 0.0455,
                   0.0606,
                                           0.0355, -0.0918, -0.0505],
                                     . . . ,
                                     ..., -0.0400, -0.0598, 0.0342],
        [ 0.0220, -0.0252, 0.0451,
                                           0.0920, -0.0008, -0.1010],
                   0.0279, -0.0334,
        [-0.0165]
                                     . . . ,
        [ 0.0982, -0.0545,
                                           0.0712, -0.0628,
                            0.0046,
                                     . . . ,
                                                              0.0117],
                            0.0395,
        [ 0.0011, 0.0164,
                                           0.0753, -0.1057, 0.0034]],
       requires grad=True)
transforms = torch model.ResNet18 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
Downloading https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz to
/content/cifar-10-python.tar.gz
      | 170498071/170498071 [00:05<00:00, 31237559.94it/s]
100%||
Extracting /content/cifar-10-python.tar.gz to /content
Files already downloaded and verified
Number of train samples: 50000
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING:matplotlib.image:Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING:matplotlib.image:Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING:matplotlib.image:Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
```

imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Number of test samples: 10000

WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Train samples



Test samples



```
learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning_rate)
train_loop(model, train_dataloader, optimizer, loss_function)
Epoch[0]: accuracy = 0.7274599671363831, time = 134.01403093338013
Epoch[1]: accuracy = 0.7578799724578857, time = 134.17937111854553
```

```
Epoch[2]: accuracy = 0.7758399844169617, time = 136.39211130142212
Epoch[3]: accuracy = 0.7821199893951416, time = 133.69815516471863
Epoch[4]: accuracy = 0.7752799987792969, time = 136.0906012058258
Epoch[5]: accuracy = 0.7862199544906616, time = 136.01660108566284
Epoch[6]: accuracy = 0.7900999784469604, time = 133.74976420402527
Epoch[7]: accuracy = 0.7939800024032593, time = 134.50200629234314
Epoch[8]: accuracy = 0.7942599654197693, time = 134.31048798561096
Epoch[9]: accuracy = 0.7878999710083008, time = 135.3505744934082
Epoch[10]: accuracy = 0.7988199591636658, time = 135.93762755393982
Epoch[11]: accuracy = 0.79367995262146, time = 134.0378932952881
Epoch[12]: accuracy = 0.7895799875259399, time = 130.2157006263733
Epoch[13]: accuracy = 0.7952799797058105, time = 128.0634801387787
Epoch[14]: accuracy = 0.7955399751663208, time = 129.41853523254395
Epoch[15]: accuracy = 0.7999799847602844, time = 128.6872215270996
Epoch[16]: accuracy = 0.7971599698066711, time = 127.93745112419128
Epoch[17]: accuracy = 0.794160008430481, time = 129.3034646511078
Epoch[18]: accuracy = 0.7957800030708313, time = 129.42212867736816
Epoch[19]: accuracy = 0.7978799939155579, time = 129.11809086799622
train time = 5302.444882154465
test model(model, test dataloader)
Test accuracy: 0.7870999574661255
```

3*. Weights + Sequental

```
model = torch model.resnet18(torch model.ResNet18 Weights.DEFAULT)
/usr/local/lib/python3.10/dist-packages/torchvision/models/
utils.py:135: UserWarning: Using 'weights' as positional parameter(s)
is deprecated since 0.13 and may be removed in the future. Please use
keyword parameter(s) instead.
 warnings.warn(
Downloading: "https://download.pytorch.org/models/resnet18-
f37072fd.pth" to /root/.cache/torch/hub/checkpoints/resnet18-
f37072fd.pth
100%|
               | 44.7M/44.7M [00:00<00:00, 142MB/s]
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make_dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
Модель "CNN":
        Layer (type)
                                   Output Shape
```

Conv2d-1 BatchNorm2d-2 ReLU-3 MaxPool2d-4 Conv2d-5 BatchNorm2d-6 ReLU-7 Conv2d-8	[-1, 64, 16, 16] [-1, 64, 16, 16] [-1, 64, 16, 16] [-1, 64, 8, 8] [-1, 64, 8, 8] [-1, 64, 8, 8] [-1, 64, 8, 8]	9,408 128 0 0 36,864 128 0 36,864
BatchNorm2d-9	[-1, 64, 8, 8]	128
ReLU-10	[-1, 64, 8, 8]	0
BasicBlock-11	[-1, 64, 8, 8]	0
Conv2d-12	[-1, 64, 8, 8]	36,864
BatchNorm2d-13	[-1, 64, 8, 8]	128
ReLU-14 Conv2d-15 BatchNorm2d-16 ReLU-17 BasicBlock-18	[-1, 64, 8, 8] [-1, 64, 8, 8] [-1, 64, 8, 8] [-1, 64, 8, 8] [-1, 64, 8, 8]	36,864 128 0
Conv2d-19 BatchNorm2d-20 ReLU-21 Conv2d-22 BatchNorm2d-23	[-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4]	73,728 256 0 147,456 256
Conv2d-24 BatchNorm2d-25 ReLU-26 BasicBlock-27	[-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4] [-1, 128, 4, 4]	8,192 256 0 0
Conv2d-28	[-1, 128, 4, 4]	147,456
BatchNorm2d-29	[-1, 128, 4, 4]	256
ReLU-30	[-1, 128, 4, 4]	0
Conv2d-31	[-1, 128, 4, 4]	147,456
BatchNorm2d-32	[-1, 128, 4, 4]	256
ReLU-33	[-1, 128, 4, 4]	0
BasicBlock-34	[-1, 128, 4, 4]	0
Conv2d-35	[-1, 256, 2, 2]	294,912
BatchNorm2d-36	[-1, 256, 2, 2]	512
ReLU-37	[-1, 256, 2, 2]	0
Conv2d-38	[-1, 256, 2, 2]	589,824
BatchNorm2d-39	[-1, 256, 2, 2]	512
Conv2d-40	[-1, 256, 2, 2]	32,768
BatchNorm2d-41	[-1, 256, 2, 2]	512
ReLU-42	[-1, 256, 2, 2]	0
BasicBlock-43	[-1, 256, 2, 2]	0
Conv2d-44	[-1, 256, 2, 2]	589,824
BatchNorm2d-45	[-1, 256, 2, 2]	512
ReLU-46	[-1, 256, 2, 2]	0
Conv2d-47	[-1, 256, 2, 2]	589,824
BatchNorm2d-48	[-1, 256, 2, 2]	512
ReLU-49	[-1, 256, 2, 2]	0

```
BasicBlock-50
                              [-1, 256, 2, 2]
                              [-1, 512, 1, 1]
                                                  1,179,648
          Conv2d-51
     BatchNorm2d-52
                              [-1, 512, 1, 1]
                                                      1,024
                              [-1, 512, 1, 1]
            ReLU-53
                                                          0
          Conv2d-54
                              [-1, 512, 1, 1]
                                                  2,359,296
                              [-1, 512, 1, 1]
     BatchNorm2d-55
                                                     1,024
                                                    131,072
          Conv2d-56
                              [-1, 512, 1, 1]
     BatchNorm2d-57
                              [-1, 512, 1, 1]
                                                      1,024
                              [-1, 512, 1, 1]
            ReLU-58
                                                          0
      BasicBlock-59
                              [-1, 512, 1, 1]
                                                          0
                              [-1, 512, 1, 1]
                                                  2,359,296
          Conv2d-60
     BatchNorm2d-61
                              [-1, 512, 1, 1]
                                                      1,024
                              [-1, 512, 1, 1]
            ReLU-62
                                                          0
                              [-1, 512, 1, 1]
                                                  2,359,296
          Conv2d-63
     BatchNorm2d-64
                              [-1, 512, 1, 1]
                                                      1,024
                              [-1, 512, 1, 1]
                                                          0
            ReLU-65
                                                          0
      BasicBlock-66
                              [-1, 512, 1, 1]
                              [-1, 512, 1, 1]
AdaptiveAvgPool2d-67
                                                          0
                                  [-1, 1000]
          Linear-68
                                                    513,000
______
Total params: 11,689,512
Trainable params: 11,689,512
Non-trainable params: 0
                          Input size (MB): 0.01
Forward/backward pass size (MB): 1.29
Params size (MB): 44.59
Estimated Total Size (MB): 45.90
```

Визуализация модели:

```
for param in model.parameters():
    param.requires grad = False
model.fc = last sequental layer(512)
model.fc.apply(init weights)
model = model.to(device)
Linear(in features=512, out features=256, bias=True)
Parameter containing:
tensor([[ 0.0720, -0.0701, -0.0681,
                                     . . . ,
                                           0.0198, 0.0107, -0.0812],
        [-0.0230, 0.0588, 0.0868,
                                           0.0526, -0.0306, -0.0151],
                                    . . . ,
        [ 0.0261, -0.0610, -0.0360, ...,
                                           0.0334, -0.0109, 0.0788],
        [-0.0444, -0.0580, -0.0670,
                                                    0.0498, -0.07561,
                                     . . . ,
                                           0.0031,
        [-0.0387, 0.0448, -0.0629, ..., -0.0607,
                                                    0.0527,
                                                             0.0145],
        [-0.0198, -0.0583, 0.0566, \ldots, 0.0535, 0.0018, -0.0780]],
       requires grad=True)
ReLU()
Linear(in features=256, out features=10, bias=True)
Parameter containing:
tensor([[ 0.0628,  0.1211, -0.1041,  ..., -0.0775, -0.1007,
                                                             0.1224],
        [ 0.0815, -0.0602, -0.1055, ...,
                                           0.0397, -0.0264,
                                                             0.02041,
        [0.0301, -0.0678, 0.0719, \ldots, -0.1130, 0.1392, -0.1355],
        [-0.1405,
                  0.0262, 0.1393,
                                   . . . ,
                                           0.1376, -0.0812,
                                                             0.07041,
        [0.0861, 0.0051, -0.0794, \ldots, -0.0544, 0.0112, 0.1229],
        [0.1047, -0.0255, 0.1043, \ldots, -0.0666, -0.1122, 0.1334]],
       requires grad=True)
Sequential(
  (0): Linear(in features=512, out features=256, bias=True)
  (1): ReLU()
  (2): Linear(in features=256, out features=10, bias=True)
print(model)
ResNet(
  (conv1): Conv2d(3, 64, kernel size=(7, 7), stride=(2, 2),
padding=(3, 3), bias=False)
  (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (relu): ReLU(inplace=True)
  (maxpool): MaxPool2d(kernel size=3, stride=2, padding=1, dilation=1,
ceil mode=False)
  (layer1): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
```

```
(conv2): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(64, 64, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running_stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(64, 64, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (layer2): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(64, 128, kernel_size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (downsample): Sequential(
        (0): Conv2d(64, 128, kernel_size=(1, 1), stride=(2, 2),
bias=False)
        (1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(128, 128, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track_running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(128, 128, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(128, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (layer3): Sequential(
    (0): BasicBlock(
```

```
(conv1): Conv2d(128, 256, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (downsample): Sequential(
        (0): Conv2d(128, 256, kernel size=(1, 1), stride=(2, 2),
bias=False)
        (1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(256, 256, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track_running stats=True)
  (layer4): Sequential(
    (0): BasicBlock(
      (conv1): Conv2d(256, 512, kernel size=(3, 3), stride=(2, 2),
padding=(1, 1), bias=False)
      (bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(512, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=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 stats=True)
    (1): BasicBlock(
      (conv1): Conv2d(512, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
```

```
(bn1): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv2): Conv2d(512, 512, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (bn2): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (avgpool): AdaptiveAvgPool2d(output size=(1, 1))
  (fc): Sequential(
    (0): Linear(in features=512, out features=256, bias=True)
    (1): ReLU()
    (2): Linear(in features=256, out features=10, bias=True)
  )
)
transforms = torch model.ResNet18 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
Downloading https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz to
/content/cifar-10-python.tar.gz
     | 170498071/170498071 [00:08<00:00, 20688473.72it/s]
Extracting /content/cifar-10-python.tar.gz to /content
Files already downloaded and verified
Number of train samples: 50000
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING:matplotlib.image:Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING:matplotlib.image:Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
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imshow with RGB data ([0..1] for floats or [0..255] for integers).
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imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING:matplotlib.image:Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
```

WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Number of test samples: 10000

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Train samples



Test samples



```
learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning_rate)
train_loop(model, train_dataloader, optimizer, loss_function)
Epoch[0]: accuracy = 0.7594799995422363, time = 133.73124742507935
Epoch[1]: accuracy = 0.7796799540519714, time = 133.07196974754333
```

```
Epoch[2]: accuracy = 0.7849000096321106, time = 131.48094820976257
Epoch[3]: accuracy = 0.7990399599075317, time = 133.5985472202301
Epoch[4]: accuracy = 0.8082799911499023, time = 133.9399676322937
Epoch[5]: accuracy = 0.8157399892807007, time = 131.6291744709015
Epoch[6]: accuracy = 0.8211399912834167, time = 130.24699783325195
Epoch[7]: accuracy = 0.8364599943161011, time = 133.8600754737854
Epoch[8]: accuracy = 0.8285399675369263, time = 133.16687679290771
Epoch[9]: accuracy = 0.8524399995803833, time = 131.25341796875
Epoch[10]: accuracy = 0.8587200045585632, time = 133.86997056007385
Epoch[11]: accuracy = 0.8580799698829651, time = 133.42539286613464
Epoch[12]: accuracy = 0.8598399758338928, time = 133.3869035243988
Epoch[13]: accuracy = 0.8741199970245361, time = 134.33729028701782
Epoch[14]: accuracy = 0.8752399682998657, time = 132.73814916610718
Epoch[15]: accuracy = 0.8773799538612366, time = 132.74400520324707
Epoch[16]: accuracy = 0.8805599808692932, time = 134.41021251678467
Epoch[17]: accuracy = 0.8913599848747253, time = 135.32344675064087
Epoch[18]: accuracy = 0.8877599835395813, time = 133.90785479545593
Epoch[19]: accuracy = 0.8972399830818176, time = 132.90964651107788
train time = 5322.203461408615
test model(model, test dataloader)
Test accuracy: 0.7958999872207642
```

Densenet121

1. Single layer

```
model =
torch model.densenet121(torch model.DenseNet121 Weights.DEFAULT)
/usr/local/lib/python3.10/dist-packages/torchvision/models/
utils.py:135: UserWarning: Using 'weights' as positional parameter(s)
is deprecated since 0.13 and may be removed in the future. Please use
keyword parameter(s) instead.
  warnings.warn(
Downloading: "https://download.pytorch.org/models/densenet121-
a639ec97.pth" to /root/.cache/torch/hub/checkpoints/densenet121-
a639ec97.pth
100% | 30.8M/30.8M [00:00<00:00, 132MB/s]
print(model)
DenseNet(
  (features): Sequential(
    (conv0): Conv2d(3, 64, kernel size=(7, 7), stride=(2, 2),
padding=(3, 3), bias=False)
    (norm0): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (relu0): ReLU(inplace=True)
    (pool0): MaxPool2d(kernel size=3, stride=2, padding=1, dilation=1,
```

```
ceil mode=False)
    (denseblock1): DenseBlock(
      (denselayer1): DenseLayer(
        (norm1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(64, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(96, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(96, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(128, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer4): _DenseLayer(
        (norm1): BatchNorm2d(160, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(160, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
```

```
(denselayer5): DenseLayer(
        (norm1): BatchNorm2d(192, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(192, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(224, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(224, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (transition1): _Transition(
      (norm): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (pool): AvgPool2d(kernel size=2, stride=2, padding=0)
    (denseblock2): DenseBlock(
      (denselayer1): DenseLayer(
        (norm1): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(128, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(160, eps=1e-05, momentum=0.1,
```

```
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(160, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(192, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(192, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer4): DenseLayer(
        (norm1): BatchNorm2d(224, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(224, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer5): DenseLayer(
        (norm1): BatchNorm2d(256, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(288, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
```

```
(relu1): ReLU(inplace=True)
        (conv1): Conv2d(288, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer7): DenseLayer(
        (norm1): BatchNorm2d(320, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(320, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer8): DenseLayer(
        (norm1): BatchNorm2d(352, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(352, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer9): DenseLayer(
        (norm1): BatchNorm2d(384, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(384, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer10): DenseLayer(
        (norm1): BatchNorm2d(416, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
```

```
(conv1): Conv2d(416, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer11): DenseLayer(
        (norm1): BatchNorm2d(448, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(448, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer12): DenseLayer(
        (norm1): BatchNorm2d(480, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(480, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (transition2): Transition(
      (norm): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (pool): AvgPool2d(kernel size=2, stride=2, padding=0)
    (denseblock3): DenseBlock(
      (denselayer1): DenseLayer(
        (norm1): BatchNorm2d(256, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
```

```
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(288, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(288, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(320, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(320, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer4): DenseLayer(
        (norm1): BatchNorm2d(352, eps=1e-05, momentum=0.1,
affine=True, track running_stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(352, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer5): DenseLayer(
        (norm1): BatchNorm2d(384, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(384, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
```

```
(relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(416, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(416, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer7): DenseLayer(
        (norm1): BatchNorm2d(448, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(448, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer8): DenseLayer(
        (norm1): BatchNorm2d(480, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(480, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer9): DenseLayer(
        (norm1): BatchNorm2d(512, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
```

```
(conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer10): DenseLayer(
        (norm1): BatchNorm2d(544, eps=1e-05, momentum=0.1,
affine=True, track_running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(544, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer11): DenseLayer(
        (norm1): BatchNorm2d(576, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(576, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer12): DenseLayer(
        (norm1): BatchNorm2d(608, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(608, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer13): DenseLayer(
        (norm1): BatchNorm2d(640, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(640, 128, kernel size=(1, 1), stride=(1, 1),
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
```

```
padding=(1, 1), bias=False)
      (denselayer14): DenseLayer(
        (norm1): BatchNorm2d(672, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(672, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer15): DenseLayer(
        (norm1): BatchNorm2d(704, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(704, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer16): DenseLayer(
        (norm1): BatchNorm2d(736, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(736, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer17): DenseLayer(
        (norm1): BatchNorm2d(768, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(768, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
```

```
(denselayer18): DenseLayer(
        (norm1): BatchNorm2d(800, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(800, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer19): DenseLayer(
        (norm1): BatchNorm2d(832, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(832, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer20): DenseLayer(
        (norm1): BatchNorm2d(864, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(864, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer21): DenseLayer(
        (norm1): BatchNorm2d(896, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(896, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer22): DenseLayer(
```

```
(norm1): BatchNorm2d(928, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(928, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer23): DenseLayer(
        (norm1): BatchNorm2d(960, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(960, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer24): DenseLayer(
        (norm1): Batch\overline{N}orm2d(992, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(992, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (transition3): Transition(
      (norm): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (pool): AvgPool2d(kernel size=2, stride=2, padding=0)
    (denseblock4): DenseBlock(
      (denselayer1): _DenseLayer(
        (norm1): BatchNorm2d(512, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
```

```
(conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(544, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(544, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(576, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(576, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer4): DenseLayer(
        (norm1): BatchNorm2d(608, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(608, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer5): DenseLayer(
        (norm1): BatchNorm2d(640, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(640, 128, kernel size=(1, 1), stride=(1, 1),
```

```
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(672, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(672, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer7): DenseLayer(
        (norm1): BatchNorm2d(704, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(704, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer8): DenseLayer(
        (norm1): BatchNorm2d(736, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(736, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer9): DenseLayer(
        (norm1): BatchNorm2d(768, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(768, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
```

```
(norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer10): DenseLayer(
        (norm1): BatchNorm2d(800, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(800, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer11): DenseLayer(
        (norm1): BatchNorm2d(832, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(832, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer12): DenseLayer(
        (norm1): BatchNorm2d(864, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(864, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer13): DenseLayer(
        (norm1): BatchNorm2d(896, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(896, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
```

```
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer14): DenseLayer(
        (norm1): BatchNorm2d(928, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(928, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer15): DenseLayer(
        (norm1): BatchNorm2d(960, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(960, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer16): DenseLayer(
        (norm1): BatchNorm2d(992, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(992, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (norm5): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (classifier): Linear(in_features=1024, out features=1000, bias=True)
for param in model.parameters():
    param.requires grad = False
```

```
model.classifier = last_layer(1024)
model = model.to(device)
print(model)
DenseNet(
  (features): Sequential(
    (conv0): Conv2d(3, 64, kernel size=(7, 7), stride=(2, 2),
padding=(3, 3), bias=False)
    (norm0): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
    (relu0): ReLU(inplace=True)
    (pool0): MaxPool2d(kernel size=3, stride=2, padding=1, dilation=1,
ceil mode=False)
    (denseblock1): DenseBlock(
      (denselayer1): DenseLayer(
        (norm1): BatchNorm2d(64, eps=1e-05, momentum=0.1, affine=True,
track_running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(64, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(96, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(96, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(128, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
```

```
(denselayer4): DenseLayer(
        (norm1): BatchNorm2d(160, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(160, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer5): DenseLayer(
        (norm1): BatchNorm2d(192, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(192, 128, kernel_size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(224, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(224, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (transition1): Transition(
      (norm): BatchNorm2d(256, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (pool): AvgPool2d(kernel size=2, stride=2, padding=0)
    (denseblock2): DenseBlock(
      (denselayer1): DenseLayer(
        (norm1): BatchNorm2d(128, eps=1e-05, momentum=0.1,
```

```
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(128, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(160, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(160, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(192, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(192, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer4): DenseLayer(
        (norm1): BatchNorm2d(224, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(224, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer5): DenseLayer(
        (norm1): BatchNorm2d(256, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
```

```
(relu1): ReLU(inplace=True)
        (conv1): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(288, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(288, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer7): DenseLayer(
        (norm1): BatchNorm2d(320, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(320, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer8): DenseLayer(
        (norm1): BatchNorm2d(352, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(352, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer9): DenseLayer(
        (norm1): BatchNorm2d(384, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
```

```
(conv1): Conv2d(384, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer10): DenseLayer(
        (norm1): BatchNorm2d(416, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(416, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer11): DenseLayer(
        (norm1): BatchNorm2d(448, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(448, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer12): DenseLayer(
        (norm1): BatchNorm2d(480, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(480, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (transition2): _Transition(
      (norm): BatchNorm2d(512, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
```

```
(conv): Conv2d(512, 256, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (pool): AvgPool2d(kernel size=2, stride=2, padding=0)
    (denseblock3): DenseBlock(
      (denselayer1): DenseLayer(
        (norm1): BatchNorm2d(256, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(256, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(288, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(288, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(320, eps=1e-05, momentum=0.1,
affine=True, track running_stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(320, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer4): DenseLayer(
        (norm1): BatchNorm2d(352, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(352, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
```

```
(relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer5): DenseLayer(
        (norm1): BatchNorm2d(384, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(384, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(416, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(416, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer7): DenseLayer(
        (norm1): BatchNorm2d(448, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(448, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer8): DenseLayer(
        (norm1): BatchNorm2d(480, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(480, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
```

```
(conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer9): DenseLayer(
        (norm1): BatchNorm2d(512, eps=1e-05, momentum=0.1,
affine=True, track_running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer10): DenseLayer(
        (norm1): BatchNorm2d(544, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(544, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer11): DenseLayer(
        (norm1): BatchNorm2d(576, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(576, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer12): DenseLayer(
        (norm1): BatchNorm2d(608, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(608, 128, kernel size=(1, 1), stride=(1, 1),
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
```

```
padding=(1, 1), bias=False)
      (denselayer13): DenseLayer(
        (norm1): BatchNorm2d(640, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(640, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer14): DenseLayer(
        (norm1): BatchNorm2d(672, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(672, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer15): DenseLayer(
        (norm1): BatchNorm2d(704, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(704, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer16): DenseLayer(
        (norm1): BatchNorm2d(736, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(736, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
```

```
(denselayer17): DenseLayer(
        (norm1): BatchNorm2d(768, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(768, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer18): DenseLayer(
        (norm1): BatchNorm2d(800, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(800, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer19): DenseLayer(
        (norm1): BatchNorm2d(832, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(832, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer20): DenseLayer(
        (norm1): BatchNorm2d(864, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(864, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
```

```
(denselayer21): DenseLayer(
        (norm1): BatchNorm2d(896, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(896, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer22): DenseLayer(
        (norm1): BatchNorm2d(928, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(928, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer23): DenseLayer(
        (norm1): BatchNorm2d(960, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(960, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer24): DenseLayer(
        (norm1): BatchNorm2d(992, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(992, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (transition3): Transition(
```

```
(norm): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
      (relu): ReLU(inplace=True)
      (conv): Conv2d(1024, 512, kernel size=(1, 1), stride=(1, 1),
bias=False)
      (pool): AvgPool2d(kernel size=2, stride=2, padding=0)
    (denseblock4): DenseBlock(
      (denselayer1): DenseLayer(
        (norm1): BatchNorm2d(512, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(512, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer2): DenseLayer(
        (norm1): BatchNorm2d(544, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(544, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer3): DenseLayer(
        (norm1): BatchNorm2d(576, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(576, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer4): DenseLayer(
        (norm1): BatchNorm2d(608, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(608, 128, kernel size=(1, 1), stride=(1, 1),
```

```
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer5): DenseLayer(
        (norm1): BatchNorm2d(640, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(640, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel_size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer6): DenseLayer(
        (norm1): BatchNorm2d(672, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(672, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer7): DenseLayer(
        (norm1): BatchNorm2d(704, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(704, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer8): DenseLayer(
        (norm1): BatchNorm2d(736, eps=1e-05, momentum=0.1,
affine=True, track_running_stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(736, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
```

```
(norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer9): DenseLayer(
        (norm1): BatchNorm2d(768, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(768, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer10): DenseLayer(
        (norm1): BatchNorm2d(800, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(800, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer11): DenseLayer(
        (norm1): BatchNorm2d(832, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(832, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer12): DenseLayer(
        (norm1): BatchNorm2d(864, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(864, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
```

```
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer13): DenseLayer(
        (norm1): BatchNorm2d(896, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(896, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer14): DenseLayer(
        (norm1): BatchNorm2d(928, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(928, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer15): DenseLayer(
        (norm1): BatchNorm2d(960, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(960, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
      (denselayer16): DenseLayer(
        (norm1): BatchNorm2d(992, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
        (relu1): ReLU(inplace=True)
        (conv1): Conv2d(992, 128, kernel size=(1, 1), stride=(1, 1),
bias=False)
        (norm2): BatchNorm2d(128, eps=1e-05, momentum=0.1,
affine=True, track running stats=True)
```

```
(relu2): ReLU(inplace=True)
        (conv2): Conv2d(128, 32, kernel size=(3, 3), stride=(1, 1),
padding=(1, 1), bias=False)
    (norm5): BatchNorm2d(1024, eps=1e-05, momentum=0.1, affine=True,
track running stats=True)
  (classifier): Linear(in features=1024, out features=10, bias=True)
)
transforms = torch model.DenseNet121 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
Files already downloaded and verified
Files already downloaded and verified
Number of train samples: 50000
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING:matplotlib.image:Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
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imshow with RGB data ([0..1] for floats or [0..255] for integers).
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imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
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imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
WARNING: matplotlib.image: Clipping input data to the valid range for
imshow with RGB data ([0..1] for floats or [0..255] for integers).
```

imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Number of test samples: 10000

WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING:matplotlib.image:Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers). WARNING: matplotlib.image: Clipping input data to the valid range for imshow with RGB data ([0..1] for floats or [0..255] for integers).

Train samples



Test samples



```
learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.classifier.parameters(),
lr=learning_rate)
model = model.to(device)

train_loop(model, train_dataloader, optimizer, loss_function)
```

```
Epoch[0]: accuracy = 0.7708999514579773, time = 172.45439887046814
Epoch[1]: accuracy = 0.7917199730873108, time = 173.8807408809662
Epoch[2]: accuracy = 0.8006599545478821, time = 173.64650082588196
Epoch[3]: accuracy = 0.796999999940094, time = 173.86425352096558
Epoch[4]: accuracy = 0.8100199699401855, time = 173.65966701507568
Epoch[5]: accuracy = 0.8076399564743042, time = 174.50546216964722
Epoch[6]: accuracy = 0.8104000091552734, time = 173.82365036010742
Epoch[7]: accuracy = 0.8085799813270569, time = 173.53456020355225
Epoch[8]: accuracy = 0.811739981174469, time = 173.1445758342743
Epoch[9]: accuracy = 0.811199963092804, time = 174.12666630744934
Epoch[10]: accuracy = 0.8151999711990356, time = 173.08181023597717
Epoch[11]: accuracy = 0.8173199892044067, time = 173.460134267807
Epoch[12]: accuracy = 0.8162199854850769, time = 173.07563638687134
Epoch[13]: accuracy = 0.814300000667572, time = 173.42166662216187
Epoch[14]: accuracy = 0.8192600011825562, time = 174.47900819778442
Epoch[15]: accuracy = 0.8169800043106079, time = 171.43979167938232
Epoch[16]: accuracy = 0.8141999840736389, time = 171.9295723438263
Epoch[17]: accuracy = 0.8133400082588196, time = 172.0942211151123
Epoch[18]: accuracy = 0.8139399886131287, time = 172.2222945690155
Epoch[19]: accuracy = 0.8186999559402466, time = 172.34216904640198
train time = 6931.575216054916
test model(model, test dataloader)
Test accuracy: 0.8026999831199646
```

1. Sequental layer

```
model =
torch model.densenet121(torch model.DenseNet121 Weights.DEFAULT)
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
for param in model.parameters():
    param.requires_grad = False
model.classifier = last sequental_layer(1024)
model = model.to(device)
print(model)
transforms = torch model.DenseNet121 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
```

```
learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.classifier.parameters(),
lr=learning_rate)
train_loop(model, train_dataloader, optimizer, loss_function)
test_model(model, test_dataloader)
```

1. Weights

```
model =
torch model.densenet121(torch model.DenseNet121 Weights.DEFAULT)
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
for param in model.parameters():
    param.requires grad = False
model.classifier = last layer(1024)
model.classifier.apply(init weights)
model = model.to(device)
print(model)
transforms = torch model.DenseNet121 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
learning rate = 0.001
loss function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning rate)
train loop(model, train dataloader, optimizer, loss function)
test model(model, test dataloader)
```

Resnext50_32x4d

1. Single layer

```
model =
torch_model.resnext50_32x4d(torch_model.ResNeXt50_32X4D_Weights.DEFAUL
T)
```

```
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
for param in model.parameters():
    param.requires grad = False
model.fc = last layer(2048)
model = model.to(device)
print(model)
transforms = torch model.ResNeXt50 32X4D Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
learning rate = 0.001
loss function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning rate)
train loop(model, train dataloader, optimizer, loss function)
test model(model, test dataloader)
```

1. Sequental layer

```
model =
torch_model.resnext50_32x4d(torch_model.ResNeXt50_32X4D_Weights.DEFAUL
T)

# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input_size=(3, 32, 32))

print("Визуализация модели:")
display(make_dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named_parameters())))

for param in model.parameters():
    param.requires_grad = False
model.fc = last_sequental_layer(2048)
model = model.to(device)

print(model)

transforms = torch_model.ResNeXt50_32X4D_Weights.DEFAULT.transforms()
batch_size = 128
```

```
train_dataloader, test_dataloader = download_data(transforms,
batch_size)

learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning_rate)
train_loop(model, train_dataloader, optimizer, loss_function)
test_model(model, test_dataloader)
```

1. Weights

```
model =
torch model.resnext50 32x4d(torch model.ResNeXt50 32X4D Weights.DEFAUL
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
for param in model.parameters():
    param.requires grad = False
model.fc = last_layer(2048)
model.fc.apply(init weights)
model = model.to(device)
print(model)
transforms = torch model.ResNeXt50 32X4D Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
learning rate = 0.001
loss function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.fc.parameters(), lr=learning rate)
train loop(model, train dataloader, optimizer, loss function)
test model(model, test dataloader)
```

ViT_B_16

1. Single layer

```
model = torch model.vit b 16(torch model.ViT B 16 Weights.DEFAULT)
```

```
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
for param in model.parameters():
    param.requires_grad = False
model.heads = last layer(768)
model = model.to(\overline{device})
print(model)
transforms = torch_model.ViT_B_16_Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
learning rate = 0.001
loss function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.heads.parameters(),
lr=learning rate)
train loop(model, train dataloader, optimizer, loss function)
test model(model, test dataloader)
```

1. Sequental layer

```
model = torch model.vit b 16(torch model.ViT B 16 Weights.DEFAULT)
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
for param in model.parameters():
    param.requires grad = False
model.heads = last sequental layer(768)
model = model.to(device)
print(model)
transforms = torch model.ViT B 16 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
```

```
learning_rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.heads.parameters(),
lr=learning_rate)
train_loop(model, train_dataloader, optimizer, loss_function)
test_model(model, test_dataloader)
```

1. Weights

```
model = torch model.vit b 16(torch model.ViT B 16 Weights.DEFAULT)
# Информация об архитектуре и визуализация сети
print("Модель \"CNN\":")
summary(model.to(device), input_size=(3, 32, 32))
print("Визуализация модели:")
display(make dot(model(torch.randn(16, 3, 32, 32).to(device)),
params=dict(model.named parameters())))
for param in model.parameters():
    param.requires grad = False
model.heads = last layer(768)
model.heads.apply(init weights)
model = model.to(device)
print(model)
transforms = torch model.ViT B 16 Weights.DEFAULT.transforms()
batch size = 128
train dataloader, test dataloader = download data(transforms,
batch size)
learning rate = 0.001
loss_function = torch.nn.CrossEntropyLoss()
optimizer = torch.optim.Adam(model.heads.parameters(),
lr=learning rate)
train loop(model, train dataloader, optimizer, loss function)
test model(model, test dataloader)
```