Лабораторная работа **3. MNIST** (Мигранов Денис Игоревич, группа **20225**).

```
In [ ]:
import torch
import torchvision
from torchvision import transforms
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
In [ ]:
batch size = 4
transform = transforms.Compose(
    [transforms.ToTensor()])
trainset = torchvision.datasets.MNIST(root='./data', train=True,
                                      download=True, transform=transform)
trainloader = torch.utils.data.DataLoader(trainset, batch size=batch size,
                                          shuffle=True, num workers=2)
testset = torchvision.datasets.MNIST(root='./data', train=False,
                                     download=True, transform=transform)
testloader = torch.utils.data.DataLoader(testset, batch size=batch size,
                                         shuffle=False, num workers=2)
classes = tuple(str(i) for i in range(10))
Downloading http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz
Downloading http://yann.lecun.com/exdb/mnist/train-images-idx3-ubyte.gz to ./data/MNIST/r
aw/train-images-idx3-ubyte.gz
Extracting ./data/MNIST/raw/train-images-idx3-ubyte.gz to ./data/MNIST/raw
Downloading http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz
Downloading http://yann.lecun.com/exdb/mnist/train-labels-idx1-ubyte.gz to ./data/MNIST/r
aw/train-labels-idx1-ubyte.gz
Extracting ./data/MNIST/raw/train-labels-idx1-ubyte.gz to ./data/MNIST/raw
Downloading http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz
Downloading http://yann.lecun.com/exdb/mnist/t10k-images-idx3-ubyte.gz to ./data/MNIST/ra
w/t10k-images-idx3-ubyte.gz
Extracting ./data/MNIST/raw/t10k-images-idx3-ubyte.gz to ./data/MNIST/raw
Downloading http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz
Downloading http://yann.lecun.com/exdb/mnist/t10k-labels-idx1-ubyte.gz to ./data/MNIST/ra
w/t10k-labels-idx1-ubyte.gz
Extracting ./data/MNIST/raw/t10k-labels-idx1-ubyte.gz to ./data/MNIST/raw
In [ ]:
import torch.nn as nn
import torch.nn.functional as F # Functional
```

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```
from tqdm import tqdm_notebook

In []:
import math

In []:
# класс наследуется от nn. Module
```

```
class SimpleConvNet1(nn.Module):
   def __init__(self, channels1, channels2, kernel_size1, kernel_size2, is max pool = T
rue):
        # вызов конструктора предка
        super(SimpleConvNet1, self).__init__()
        # необходмо заранее знать, сколько каналов у картинки (сейчас = 1),
        # которую будем подавать в сеть, больше ничего
        # про входящие картинки знать не нужно
        self.conv1 = nn.Conv2d(in channels=1, out channels=channels1, kernel size=5)
        new size = 28 - kernel size1 + 1
        if is max pool:
         self.pool = nn.MaxPool2d(kernel size=2, stride=2)
        else:
         self.pool = nn.AvgPool2d(kernel size=2, stride=2)
        new size = new size // 2
        self.conv2 = nn.Conv2d(in channels=channels1, out channels=channels2, kernel siz
e = 5)
       new size = new size - kernel size2 + 1
       new size = new size // 2
        self.fc1 size = new size * new size * channels2
        self.fc1 = nn.Linear(new size * new size * channels2, 120) # !!!
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)
    def forward(self, x):
        x = self.pool(F.relu(self.conv1(x)))
        #print(x.shape)
        x = self.pool(F.relu(self.conv2(x)))
        #print(x.shape)
        x = x.view(-1, self.fc1 size) # !!!
        x = F.relu(self.fc1(x))
        x = F.relu(self.fc2(x))
        x = self.fc3(x)
        return x
    def train(self, learning_rate = 1e-4, num_epochs = 3):
        loss fn = torch.nn.CrossEntropyLoss()
        optimizer = torch.optim.Adam(self.parameters(), lr=learning rate)
        # итерируемся
        for epoch in tqdm notebook(range(num epochs)):
            running loss = 0.0
            for i, batch in enumerate(tqdm notebook(trainloader)):
                # так получаем текущий батч
                X batch, y batch = batch
                # обнуляем веса
                optimizer.zero grad()
                # forward + backward + optimize
                y pred = self(X batch)
                loss = loss fn(y pred, y batch)
                loss.backward()
                optimizer.step()
                # выведем текущий loss
                running loss += loss.item()
                # выведем качество каждые 2000 батчей
                if i % 2000 == 1999:
                    print('[%d, %5d] loss: %.3f' %
```

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(epoch + 1, i + 1, running_loss / 2000))
                    running loss = 0.0
        print('Обучение закончено')
In [ ]:
net = SimpleConvNet1(6, 16, 5, 5) #как в примере
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
[1, 2000] loss: 1.071
   4000] loss: 0.427
[1,
[1,
   6000] loss: 0.324
[1, 8000] loss: 0.281
[1, 10000] loss: 0.224
[1, 12000] loss: 0.196
[1, 14000] loss: 0.175
[2, 2000] loss: 0.157
[2, 4000] loss: 0.130
[2, 6000] loss: 0.121
[2, 8000] loss: 0.122
[2, 10000] loss: 0.114
[2, 12000] loss: 0.115
[2, 14000] loss: 0.099
[3, 2000] loss: 0.088
    4000] loss: 0.097
[3,
    6000] loss: 0.082
[3,
    8000] loss: 0.084
[3,
[3, 10000] loss: 0.083
[3, 12000] loss: 0.081
[3, 14000] loss: 0.084
Обучение закончено
In [ ]:
def check network(net):
  class correct = list(0. for i in range(10))
  class total = list(0. for i in range(10))
  with torch.no grad():
      for data in testloader:
          images, labels = data
          y_pred = net(images)
          _, predicted = torch.max(y_pred, 1)
          c = (predicted == labels).squeeze()
          for i in range(4):
              label = labels[i]
              class correct[label] += c[i].item()
              class total[label] += 1
  for i in range(10):
      print('Accuracy of %2s : %2d %%' % (
          classes[i], 100 * class correct[i] / class total[i]))
  class correct t = sum(class correct)
  class total t = sum(class total)
  print('\nTotal accuracy:', (100. * class_correct_t / class_total_t))
```

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check network(net)
Accuracy of 0:99 %
Accuracy of 1:99 %
Accuracy of 2:99 %
Accuracy of 3:98%
Accuracy of 4:98%
Accuracy of
            5:96%
Accuracy of
            6:98%
Accuracy of
            7:96%
            8:97%
Accuracy of
Accuracy of 9: 97 %
Total accuracy: 98.17
Попробуем Average Pooling с теми же параметрами:
In [ ]:
net avg1 = SimpleConvNet1(6, 16, 5, 5, False) #как в примере, Но с Avg пулингом
net avg1.train()
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
[1, 2000] loss: 1.060
[1, 4000] loss: 0.473
[1, 6000] loss: 0.392
[1, 8000] loss: 0.342
[1, 10000] loss: 0.312
[1, 12000] loss: 0.278
[1, 14000] loss: 0.242
[2, 2000] loss: 0.211
[2, 4000] loss: 0.194
    6000] loss: 0.175
[2,
[2, 8000] loss: 0.161
[2, 10000] loss: 0.147
[2, 12000] loss: 0.152
[2, 14000] loss: 0.139
[3,
    2000] loss: 0.120
    4000] loss: 0.106
[3,
[3,
    6000] loss: 0.106
[3,
    8000] loss: 0.106
[3, 10000] loss: 0.101
[3, 12000] loss: 0.097
[3, 14000] loss: 0.102
Обучение закончено
In [ ]:
check network(net avg1)
Accuracy of 0 : 99 %
Accuracy of
            1:99%
Accuracy of
            2:98%
            3:96%
Accuracy of
Accuracy of
            4:96%
Accuracy of
            5: 97 %
Accuracy of
            6:97%
            7:97%
Accuracy of
Accuracy of
            8: 97 %
7 ~ ~ . . . . . . . . . . .
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Accuracy of 9:90 6

Total accuracy: 97.67
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Мах с остальными такими же параметрами был лучше...

Добавим в конструктор возможность регулировать размер fully connected слоёв:

```
In [ ]:
```

```
import math
# класс наследуется от nn.Module
class SimpleConvNet2(nn.Module):
   def init (self, channels1, channels2, kernel size1, kernel size2, fc1, fc2, is ma
x pool = True):
        # вызов конструктора предка
        super(SimpleConvNet2, self). init ()
        # необходмо заранее знать, сколько каналов у картинки (сейчас = 1),
        # которую будем подавать в сеть, больше ничего
        # про входящие картинки знать не нужно
        self.conv1 = nn.Conv2d(in channels=1, out channels=channels1, kernel size=kernel
size1)
       new size = 28 - \text{kernel size1} + 1
       if is max pool:
         self.pool = nn.MaxPool2d(kernel size=2, stride=2)
       else:
         self.pool = nn.AvgPool2d(kernel size=2, stride=2)
        new size = new size // 2
        self.conv2 = nn.Conv2d(in channels=channels1, out channels=channels2, kernel siz
e=kernel size2)
       new_size = new_size - kernel size2 + 1
        new size = new size // 2
        self.fc1_size = new_size * new_size * channels2
        self.fc1 = nn.Linear(new size * new size * channels2, fc1) # !!!
        self.fc2 = nn.Linear(fc1, fc2)
        self.fc3 = nn.Linear(fc2, 10)
    def forward(self, x):
        x = self.pool(F.relu(self.conv1(x)))
        #print(x.shape)
        x = self.pool(F.relu(self.conv2(x)))
        #print(x.shape)
        x = x.view(-1, self.fc1 size) # !!!
        x = F.relu(self.fcl(x))
        x = F.relu(self.fc2(x))
        x = self.fc3(x)
       return x
    def train(self, train loader, learning rate = 1e-4, num epochs = 3):
        loss fn = torch.nn.CrossEntropyLoss()
        optimizer = torch.optim.Adam(self.parameters(), lr=learning rate)
        # итерируемся
        for epoch in tqdm notebook(range(num epochs)):
            running loss = 0.0
            for i, batch in enumerate(tqdm notebook(train loader)):
                # так получаем текущий батч
                X batch, y batch = batch
                # обнуляем веса
                optimizer.zero grad()
                # forward + backward + optimize
                y pred = self(X batch)
                loss = loss_fn(y_pred, y_batch)
                loss.backward()
                optimizer.step()
                # выведем текущий loss
```

```
running loss += loss.item()
                # выведем качество каждые 2000 батчей
                if i % 2000 == 1999:
                    print('[%d, %5d] loss: %.3f' %
                          (epoch + 1, i + 1, running loss / 2000))
                    running loss = 0.0
        print('Обучение закончено')
    def predict(self, test loader):
        class correct = list(0. for i in range(10))
        class total = list(0. for i in range(10))
        with torch.no grad():
            for data in test loader:
                images, labels = data
                y pred = self(images)
                _, predicted = torch.max(y pred, 1)
                c = (predicted == labels).squeeze()
                for i in range(4):
                    label = labels[i]
                    class correct[label] += c[i].item()
                    class_total[label] += 1
        for i in range(10):
            print('Accuracy of %2s : %2d %%' % (
                classes[i], 100 * class correct[i] / class total[i]))
        class correct t = sum(class correct)
        class total t = sum(class total)
        print('\nTotal accuracy:', (100. * class correct t / class total t))
In [ ]:
net 2 = SimpleConvNet2(6, 16, 5, 5, 200, 100, False) #с Avg пулингом
net 2.train(trainloader, num epochs=4)
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
[1, 2000] loss: 0.985
[1, 4000] loss: 0.449
[1,
   60001 loss: 0.383
[1, 8000] loss: 0.339
[1, 10000] loss: 0.268
[1, 12000] loss: 0.252
[1, 14000] loss: 0.233
   2000] loss: 0.196
[2,
[2,
    4000] loss: 0.181
[2,
    6000] loss: 0.163
[2,
    8000] loss: 0.154
[2, 10000] loss: 0.143
[2, 12000] loss: 0.139
[2, 14000] loss: 0.127
[3, 2000] loss: 0.114
[3, 4000] loss: 0.116
[3, 6000] loss: 0.114
[3, 8000] loss: 0.096
[3, 10000] loss: 0.101
[3, 12000] loss: 0.093
[3, 14000] loss: 0.089
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2000] loss: 0.084
    4000] loss: 0.075
[4,
    6000] loss: 0.075
[4,
    8000] loss: 0.075
[4,
[4, 10000] loss: 0.072
[4, 12000] loss: 0.077
[4, 14000] loss: 0.075
Обучение закончено
In [ ]:
check_network(net_2)
Accuracy of
            0:99%
            1:99%
Accuracy of
Accuracy of 2:98 %
Accuracy of
            3:98%
Accuracy of
            4:98%
Accuracy of
            5:98%
            6:97%
Accuracy of
            7:96%
Accuracy of
Accuracy of
            8:95%
Accuracy of
            9:97%
Total accuracy: 97.86
In [ ]:
net 3 = SimpleConvNet2(6, 16, 5, 5, 200, 100) #с Мах пулингом
net 3.train(trainloader, num epochs=4)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
[1, 2000] loss: 0.907
    4000] loss: 0.374
[1,
    6000] loss: 0.287
[1,
    8000] loss: 0.232
[1,
[1, 10000] loss: 0.206
[1, 12000] loss: 0.174
[1, 14000] loss: 0.145
[2,
    2000] loss: 0.145
    4000] loss: 0.121
[2,
    6000] loss: 0.114
[2,
[2, 8000] loss: 0.108
[2, 10000] loss: 0.103
[2, 12000] loss: 0.091
[2, 14000] loss: 0.084
[3,
    2000] loss: 0.078
[3,
    4000] loss: 0.076
   6000] loss: 0.082
[3,
    8000] loss: 0.071
[3,
[3, 10000] loss: 0.072
[3, 12000] loss: 0.074
[3, 14000] loss: 0.063
[4,
    2000] loss: 0.056
[4,
    4000] loss: 0.064
[4,
    6000] loss: 0.054
[4, 8000] loss: 0.062
[4, 10000] loss: 0.050
[4, 12000] loss: 0.054
```

```
In [ ]:
net 3.predict(testloader)
Accuracy of 0:99 %
Accuracy of 1 : 99 %
Accuracy of 2 : 98 %
Accuracy of
            3:98%
            4:99%
Accuracy of
Accuracy of
            5 : 99 %
Accuracy of 6 : 98 %
Accuracy of 7 : 97 %
Accuracy of 8:99 %
Accuracy of 9 : 97 %
Total accuracy: 98.7
Попробуем добавить трансформы картинкам...
In [ ]:
batch size = 4
transform = transforms.Compose(
    [transforms.ToTensor(), transforms.RandomRotation(30)])
trainset t = torchvision.datasets.MNIST(root='./data', train=True,
                                      download=True, transform=transform)
trainset inc = torch.utils.data.ConcatDataset([trainset,trainset t])
trainloader_inc = torch.utils.data.DataLoader(trainset_inc, batch_size=batch_size,
                                          shuffle=True, num workers=2)
In [ ]:
net 4 = SimpleConvNet2(6, 16, 5, 5, 200, 100) #с Мах пулингом
net 4.train(trainloader inc, num epochs=2)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
[1, 2000] loss: 1.077
[1, 4000] loss: 0.446
    6000] loss: 0.342
[1,
    8000] loss: 0.312
[1,
[1, 10000] loss: 0.266
[1, 12000] loss: 0.231
[1, 14000] loss: 0.214
[1, 16000] loss: 0.199
[1, 18000] loss: 0.173
[1, 20000] loss: 0.177
[1, 22000] loss: 0.170
[1, 24000] loss: 0.170
[1, 26000] loss: 0.147
[1, 28000] loss: 0.143
[1, 30000] loss: 0.128
[2, 2000] loss: 0.131
[2, 4000] loss: 0.120
[2, 6000] loss: 0.119
[2, 8000] loss: 0.120
[2, 10000] loss: 0.124
[2, 12000] loss: 0.112
```

[4, 14000] loss: 0.056 Обучение закончено

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[2, 14000] loss: 0.098
[2, 16000] loss: 0.106
[2, 18000] loss: 0.100
[2, 20000] loss: 0.091
[2, 22000] loss: 0.095
[2, 24000] loss: 0.086
[2, 26000] loss: 0.089
[2, 28000] loss: 0.091
[2, 30000] loss: 0.087
Обучение закончено
In [ ]:
net 4.predict(testloader)
Accuracy of 0:98 %
            1:99%
Accuracy of
            2:98%
Accuracy of
Accuracy of
            3:99%
            4:98%
Accuracy of
            5:97
Accuracy of
Accuracy of
            6:98%
            7:97%
Accuracy of
            8:95%
Accuracy of
Accuracy of 9: 97 %
Total accuracy: 98.21
In [ ]:
net 4.train(trainloader inc, num epochs=1)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
    2000] loss: 0.080
[1,
[1,
    4000] loss: 0.086
    6000] loss: 0.086
[1,
[1,
   8000] loss: 0.073
[1, 10000] loss: 0.073
[1, 12000] loss: 0.070
[1, 14000] loss: 0.081
[1, 16000] loss: 0.075
[1, 18000] loss: 0.075
[1, 20000] loss: 0.069
[1, 22000] loss: 0.065
[1, 24000] loss: 0.074
[1, 26000] loss: 0.071
[1, 28000] loss: 0.068
[1, 30000] loss: 0.060
Обучение закончено
In [ ]:
net 4.predict(testloader)
Accuracy of 0:99 %
Accuracy of
            1:99%
Accuracy of
           2:98%
            3:98%
Accuracy of
Accuracy of
            4:98%
Accuracy of
            5: 97 %
            6:98%
Accuracy of
Accuracy of
            7:98%
Accuracy of
            8:98%
Accuracy of
            9:97%
```

```
Total accuracy: 98.53
In [ ]:
net 4.train(trainloader inc, num epochs=1)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
[1, 2000] loss: 0.059
[1, 4000] loss: 0.063
[1, 6000] loss: 0.058
[1, 8000] loss: 0.066
[1, 10000] loss: 0.056
[1, 12000] loss: 0.059
[1, 14000] loss: 0.060
[1, 16000] loss: 0.056
[1, 18000] loss: 0.056
[1, 20000] loss: 0.059
[1, 22000] loss: 0.057
[1, 24000] loss: 0.056
[1, 26000] loss: 0.054
[1, 28000] loss: 0.059
[1, 30000] loss: 0.055
Обучение закончено
In [ ]:
net 4.predict(testloader)
Accuracy of 0:99 %
Accuracy of 1 : 99 %
Accuracy of 2 : 99 %
Accuracy of 3 : 98 %
Accuracy of
            4:97%
Accuracy of
            5:98%
Accuracy of
            6:99%
            7:99%
Accuracy of
Accuracy of 8 : 97 %
Accuracy of 9: 98 %
Total accuracy: 98.87
In [ ]:
net 4.train(trainloader inc, num epochs=1)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
    2000] loss: 0.048
[1,
    4000] loss: 0.055
[1,
[1,
    6000] loss: 0.054
[1, 8000] loss: 0.045
[1, 10000] loss: 0.055
[1, 12000] loss: 0.049
[1, 14000] loss: 0.051
[1, 16000] loss: 0.050
[1, 18000] loss: 0.037
```

```
|1, 20000| loss: 0.044
[1, 22000] loss: 0.040
[1, 24000] loss: 0.051
[1, 26000] loss: 0.047
[1, 28000] loss: 0.045
[1, 30000] loss: 0.044
Обучение закончено
In [ ]:
net 4.predict(testloader) #итого 5 эпох
Accuracy of 0:99 %
Accuracy of 1 : 99 %
Accuracy of 2 : 99 %
Accuracy of 3:99 %
Accuracy of
            4:97%
Accuracy of
            5:97%
            6:99%
Accuracy of
            7:99%
Accuracy of
Accuracy of
            8:97%
Accuracy of
             9:98%
Total accuracy: 98.78
Попробуем другие размеры ядра, а потом увеличим число сверточных слоев
In [ ]:
net 5 = SimpleConvNet2(6, 16, 7, 3, 160, 80) #с Мах пулингом
net 5.train(trainloader inc, num epochs=3)
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:43: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
    2000] loss: 1.078
[1,
    4000] loss: 0.487
[1,
    6000] loss: 0.418
[1,
[1, 8000] loss: 0.367
[1, 10000] loss: 0.306
[1, 12000] loss: 0.278
[1, 14000] loss: 0.270
[1, 16000] loss: 0.229
[1, 18000] loss: 0.211
[1, 20000] loss: 0.194
[1, 22000] loss: 0.186
[1, 24000] loss: 0.168
[1, 26000] loss: 0.168
[1, 28000] loss: 0.155
[1, 30000] loss: 0.135
[2,
    2000] loss: 0.147
    4000] loss: 0.129
[2,
     6000] loss: 0.130
[2,
    8000] loss: 0.137
[2,
[2, 10000] loss: 0.125
[2, 12000] loss: 0.119
[2, 14000] loss: 0.114
[2, 16000] loss: 0.113
[2, 18000] loss: 0.110
[2, 20000] loss: 0.113
[2, 22000] loss: 0.109
[2, 24000] loss: 0.102
[2, 26000] loss: 0.110
[2, 28000] loss: 0.101
```

```
[3, 2000] loss: 0.083
    4000] loss: 0.087
[3,
    6000] loss: 0.088
[3,
[3,
    8000] loss: 0.086
[3, 10000] loss: 0.083
[3, 12000] loss: 0.087
[3, 14000] loss: 0.088
[3, 16000] loss: 0.085
[3, 18000] loss: 0.083
[3, 20000] loss: 0.085
[3, 22000] loss: 0.076
[3, 24000] loss: 0.075
[3, 26000] loss: 0.078
[3, 28000] loss: 0.075
[3, 30000] loss: 0.077
Обучение закончено
In [ ]:
net 5.predict(testloader)
Accuracy of 0:99 %
Accuracy of
            1:99%
Accuracy of
            2:98%
Accuracy of
            3:97%
Accuracy of
            4:98%
            5:96%
Accuracy of
Accuracy of 6 : 98 %
Accuracy of 7 : 98 %
Accuracy of 8 : 97 %
Accuracy of 9:96%
Total accuracy: 98.35
Добавим ещё сверточный слой
In [ ]:
# класс наследуется от nn.Module
class SimpleConvNet3C(nn.Module):
    def init (self, channels1, channels2, channels3, kernel size1, kernel size2, kern
el size3, fc1, fc2, is max pool = True):
        # вызов конструктора предка
        super(SimpleConvNet3C, self). init ()
        self.conv1 = nn.Conv2d(in channels=1, out channels=channels1, kernel size=kernel
_size1)
        new_size = 28 - kernel_size1 + 1
        if is max pool:
         self.pool = nn.MaxPool2d(kernel size=2, stride=2)
        else:
         self.pool = nn.AvgPool2d(kernel size=2, stride=2)
        new size = new size // 2
        self.conv2 = nn.Conv2d(in_channels=channels1, out_channels=channels2, kernel siz
e=kernel size2)
        new size = new size - kernel size2 + 1
        new size = new size // 2
        self.conv3 = nn.Conv2d(in channels=channels2, out channels=channels3, kernel siz
e=kernel size3)
        new size = new size - kernel size3 + 1
        #new size = new size // 2
        #print(new size)
        self.fc1 size = new size * new size * channels3
```

self.fcl = nn.Linear(new size * new size * channels3, fcl) # !!!

[2, 30000] loss: 0.101

```
self.fc2 = nn.Linear(fc1, fc2)
    self.fc3 = nn.Linear(fc2, 10)
def forward(self, x):
   x = self.pool(F.relu(self.conv1(x)))
    #print(x.shape)
   x = self.pool(F.relu(self.conv2(x)))
   #print(x.shape)
                                     \#x = self.pool(F.relu(self.conv3(x)))
   x = F.relu(self.conv3(x))
   #print(x.shape)
   x = x.view(-1, self.fc1 size) # !!!
   x = F.relu(self.fcl(x))
   x = F.relu(self.fc2(x))
   x = self.fc3(x)
   return x
def train(self, train loader, learning rate = 1e-4, num epochs = 3):
    loss fn = torch.nn.CrossEntropyLoss()
    optimizer = torch.optim.Adam(self.parameters(), lr=learning rate)
    # итерируемся
   for epoch in tqdm notebook(range(num epochs)):
        running_loss = 0.0
        for i, batch in enumerate(tqdm_notebook(train loader)):
            # так получаем текущий батч
            X_batch, y_batch = batch
            # обнуляем веса
            optimizer.zero grad()
            # forward + backward + optimize
            y pred = self(X batch)
            loss = loss_fn(y_pred, y_batch)
            loss.backward()
            optimizer.step()
            # выведем текущий loss
            running loss += loss.item()
            # выведем качество каждые 2000 батчей
            if i % 2000 == 1999:
                print('[%d, %5d] loss: %.3f' %
                      (epoch + 1, i + 1, running_loss / 2000))
                running loss = 0.0
   print('Обучение закончено')
def predict(self, test loader):
   class correct = list(0. for i in range(10))
   class total = list(0. for i in range(10))
   with torch.no grad():
        for data in test loader:
            images, labels = data
            y pred = self(images)
            _, predicted = torch.max(y pred, 1)
            c = (predicted == labels).squeeze()
            for i in range(4):
                label = labels[i]
                class correct[label] += c[i].item()
                class total[label] += 1
   for i in range(10):
        print('Accuracy of %2s : %2d %%' % (
            classes[i], 100 * class correct[i] / class total[i]))
    class correct t = sum(class correct)
   class total t = sum(class total)
   print('\nTotal accuracy:', (100. * class correct t / class total t))
```

```
net_6.train(trainloader_inc, num_epochs=3, learning_rate = 1e-3)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:47: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:49: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
    2000] loss: 0.734
    4000] loss: 0.295
[1,
    6000] loss: 0.241
[1,
[1,
    8000] loss: 0.201
[1, 10000] loss: 0.192
[1, 12000] loss: 0.172
[1, 14000] loss: 0.161
[1, 16000] loss: 0.156
[1, 18000] loss: 0.143
[1, 20000] loss: 0.148
[1, 22000] loss: 0.138
[1, 24000] loss: 0.129
[1, 26000] loss: 0.135
[1, 28000] loss: 0.124
[1, 30000] loss: 0.122
[2,
    2000] loss: 0.114
[2,
    4000] loss: 0.104
[2,
    6000] loss: 0.113
[2,
   8000] loss: 0.110
[2, 10000] loss: 0.109
[2, 12000] loss: 0.100
[2, 14000] loss: 0.107
[2, 16000] loss: 0.103
[2, 18000] loss: 0.095
[2, 20000] loss: 0.096
[2, 22000] loss: 0.106
[2, 24000] loss: 0.099
[2, 26000] loss: 0.091
[2, 28000] loss: 0.098
[2, 30000] loss: 0.112
    2000] loss: 0.089
[3,
[3,
    4000] loss: 0.086
    6000] loss: 0.087
[3,
    8000] loss: 0.092
[3,
[3, 10000] loss: 0.090
[3, 12000] loss: 0.091
[3, 14000] loss: 0.086
[3, 16000] loss: 0.099
[3, 18000] loss: 0.077
[3, 20000] loss: 0.095
[3, 22000] loss: 0.087
[3, 24000] loss: 0.083
[3, 26000] loss: 0.091
[3, 28000] loss: 0.089
[3, 30000] loss: 0.094
Обучение закончено
In [ ]:
net 6.predict(testloader)
```

```
Accuracy of 0:99 %
Accuracy of
           1:98%
Accuracy of
          2:98%
Accuracy of
           3:98%
Accuracy of
           4:99%
Accuracy of
           5:98%
Accuracy of
           6:98%
```

```
Accuracy of 7:98%
Accuracy of 8:97%
Accuracy of 9:96%
Total accuracy: 98.51
```

Улучшения нет... Вернемся к варианту с двумя слоями и добавим дропаут

```
In [ ]:
```

```
from torch.nn import Dropout
```

```
In [ ]:
```

```
import math
# класс наследуется от nn.Module
class SimpleConvNet2D(nn.Module):
   def init (self, channels1, channels2, kernel size1, kernel size2, fc1, dropout, i
s max pool = True):
        # вызов конструктора предка
       super(SimpleConvNet2D, self).__init__()
        # необходмо заранее знать, сколько каналов у картинки (сейчас = 1),
        # которую будем подавать в сеть, больше ничего
        # про входящие картинки знать не нужно
        self.conv1 = nn.Conv2d(in channels=1, out channels=channels1, kernel size=kernel
_size1)
       new size = 28 - \text{kernel size1} + 1
       if is max pool:
         self.pool = nn.MaxPool2d(kernel size=2, stride=2)
       else:
         self.pool = nn.AvgPool2d(kernel size=2, stride=2)
       new size = new size // 2
        self.conv2 = nn.Conv2d(in channels=channels1, out channels=channels2, kernel siz
e=kernel size2)
       new size = new size - kernel size2 + 1
       new size = new size // 2
       self.fc1 size = new_size * new_size * channels2
       self.fc1 = nn.Linear(new size * new size * channels2, fc1) # !!!
        #self.fc2 = nn.Linear(fc1, fc2)
       self.fc3 = nn.Linear(fc1, 10)
        self.dropout1 = Dropout(dropout)
       self.dropout2 = Dropout(dropout)
   def forward(self, x):
       x = self.pool(F.relu(self.conv1(x)))
        #print(x.shape)
       x = self.pool(F.relu(self.conv2(x)))
        #print(x.shape)
       x = x.view(-1, self.fc1 size) # !!!
       x = self.dropout1(F.relu(self.fc1(x)))
        \#x = self.dropout2(F.relu(self.fc2(x)))
       x = self.fc3(x)
       return x
   def train_network(self, train_loader, learning_rate = 1e-4, num_epochs = 3):
        loss fn = torch.nn.CrossEntropyLoss()
        optimizer = torch.optim.Adam(self.parameters(), lr=learning_rate)
        # итерируемся
        for epoch in tqdm notebook(range(num epochs)):
            running loss = 0.0
            for i, batch in enumerate(tqdm notebook(train loader)):
                # так получаем текущий батч
                X batch, y batch = batch
                # обнуляем веса
                optimizer.zero grad()
```

```
# forward + backward + optimize
                y_pred = self(X_batch)
                loss = loss fn(y pred, y batch)
                loss.backward()
                optimizer.step()
                # выведем текущий loss
                running loss += loss.item()
                # выведем качество каждые 2000 батчей
                if i % 2000 == 1999:
                    print('[%d, %5d] loss: %.3f' %
                           (epoch + 1, i + 1, running loss / 2000))
                    running_loss = 0.0
        print('Обучение закончено')
    def predict(self, test loader):
        class correct = list(0. for i in range(10))
        class total = list(0. for i in range(10))
        self.eval()
        with torch.no grad():
            for data in test loader:
                images, labels = data
                y pred = self(images)
                _, predicted = torch.max(y pred, 1)
                c = (predicted == labels).squeeze()
                for i in range(4):
                    label = labels[i]
                    class correct[label] += c[i].item()
                    class total[label] += 1
        for i in range(10):
            print('Accuracy of %2s : %2d %%' % (
                classes[i], 100 * class_correct[i] / class_total[i]))
        class_correct_t = sum(class_correct)
        class_total_t = sum(class_total)
        print('\nTotal accuracy:', (100. * class_correct_t / class_total_t))
        self.train()
In [ ]:
net 2D1 = SimpleConvNet2D(6, 16, 5, 5, 200, 0.5) #с Мах пулингом
net 2D1.train network(trainloader inc, num epochs=2, learning rate=1e-3)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:46: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:48: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
[1, 2000] loss: 0.590
[1, 4000] loss: 0.249
[1, 6000] loss: 0.215
[1, 8000] loss: 0.189
[1, 10000] loss: 0.181
[1, 12000] loss: 0.165
[1, 14000] loss: 0.155
[1, 16000] loss: 0.141
[1, 18000] loss: 0.135
[1, 20000] loss: 0.127
[1, 22000] loss: 0.137
[1, 24000] loss: 0.110
[1, 26000] loss: 0.124
[1, 28000] loss: 0.125
```

```
[2,
   2000] loss: 0.121
   4000] loss: 0.103
[2,
[2, 6000] loss: 0.107
[2, 8000] loss: 0.103
[2, 10000] loss: 0.103
[2, 12000] loss: 0.103
[2, 14000] loss: 0.098
[2, 16000] loss: 0.106
[2, 18000] loss: 0.094
[2, 20000] loss: 0.100
[2, 22000] loss: 0.098
[2, 24000] loss: 0.099
[2, 26000] loss: 0.091
[2, 28000] loss: 0.100
[2, 30000] loss: 0.103
Обучение закончено
In [ ]:
net 2D1.predict(testloader)
Accuracy of 0:99 %
Accuracy of 1:99 %
Accuracy of 2:98%
Accuracy of 3 : 98 %
Accuracy of 4: 99 %
Accuracy of 5 : 98 %
Accuracy of 6:99 %
Accuracy of 7:98%
Accuracy of 8:98%
Accuracy of 9:97%
Total accuracy: 98.75
In [ ]:
net 2D1.train network(trainloader inc, num epochs=1, learning rate=1e-3)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:46: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:48: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
[1, 2000] loss: 0.102
[1, 4000] loss: 0.101
[1, 6000] loss: 0.091
[1, 8000] loss: 0.098
[1, 10000] loss: 0.100
[1, 12000] loss: 0.085
[1, 14000] loss: 0.095
[1, 16000] loss: 0.090
[1, 18000] loss: 0.092
[1, 20000] loss: 0.096
[1, 22000] loss: 0.097
[1, 24000] loss: 0.092
[1, 26000] loss: 0.094
[1, 28000] loss: 0.089
[1, 30000] loss: 0.101
Обучение закончено
In [ ]:
net 2D1.predict(testloader)
Accuracy of 0:99 %
```

[1, 30000] loss: 0.116

Accuracy of 1:99 %

```
Accuracy of 2 : 98 %
Accuracy of 3 : 98 %
Accuracy of 4:98%
Accuracy of 5: 97 %
Accuracy of
            6:99%
            7:98%
Accuracy of
Accuracy of 8 : 98 %
Accuracy of 9: 98 %
Total accuracy: 98.67
In [ ]:
net 2D1.train network(trainloader inc, num epochs=1, learning rate=5e-4)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:46: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:48: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
[1,
    2000] loss: 0.127
[1,
    4000] loss: 0.119
[1,
    6000] loss: 0.121
[1,
   8000] loss: 0.107
[1, 10000] loss: 0.115
[1, 12000] loss: 0.098
[1, 14000] loss: 0.095
[1, 16000] loss: 0.109
[1, 18000] loss: 0.095
[1, 20000] loss: 0.097
[1, 22000] loss: 0.098
[1, 24000] loss: 0.097
[1, 26000] loss: 0.095
[1, 28000] loss: 0.086
[1, 30000] loss: 0.094
Обучение закончено
In [ ]:
net 2D1.predict(testloader)
Accuracy of 0:99 %
Accuracy of 1:99 %
Accuracy of 2 : 98 %
Accuracy of 3:99 %
Accuracy of 4 : 97 %
Accuracy of 5: 97 %
Accuracy of 6:98%
Accuracy of 7:98%
Accuracy of 8 : 97 %
Accuracy of 9:96%
Total accuracy: 98.28
In [ ]:
net 2D1.train network(trainloader inc, num epochs=2, learning rate=5e-4)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:46: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:48: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
```

[1,

20001 loss: 0.093

```
[1,
     4000] loss: 0.093
[1,
     6000] loss: 0.084
[1,
    8000] loss: 0.087
[1, 10000] loss: 0.092
[1, 12000] loss: 0.089
[1, 14000] loss: 0.083
[1, 16000] loss: 0.096
[1, 18000] loss: 0.082
[1, 20000] loss: 0.085
[1, 22000] loss: 0.098
[1, 24000] loss: 0.089
[1, 26000] loss: 0.089
[1, 28000] loss: 0.096
[1, 30000] loss: 0.080
[2,
    2000] loss: 0.082
[2,
    4000] loss: 0.077
    6000] loss: 0.084
[2,
   8000] loss: 0.074
[2, 10000] loss: 0.082
[2, 12000] loss: 0.098
[2, 14000] loss: 0.080
[2, 16000] loss: 0.085
[2, 18000] loss: 0.085
[2, 20000] loss: 0.085
[2, 22000] loss: 0.074
[2, 24000] loss: 0.092
[2, 26000] loss: 0.088
[2, 28000] loss: 0.080
[2, 30000] loss: 0.080
Обучение закончено
In [ ]:
net 2D1.predict(testloader)
Accuracy of
            0:99%
Accuracy of
            1:99%
Accuracy of
            2:98%
Accuracy of
             3:98%
Accuracy of
             4:97%
Accuracy of
             5:98%
             6:98%
Accuracy of
               : 99 %
             7
Accuracy of
             8:99%
Accuracy of
Accuracy of
             9:97%
Total accuracy: 98.88
In [ ]:
net_2D1.train_network(trainloader_inc, num_epochs=1, learning rate=1e-4)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:46: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:48: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
    2000] loss: 0.072
[1,
    4000] loss: 0.063
[1,
    6000] loss: 0.055
[1,
    8000] loss: 0.061
[1,
[1, 10000] loss: 0.069
[1, 12000] loss: 0.064
[1, 14000] loss: 0.065
```

[1, 16000] loss: 0.063 [1, 18000] loss: 0.064

```
[I, ZUUUU] LOSS: U.UJO
[1, 22000] loss: 0.059
[1, 24000] loss: 0.067
[1, 26000] loss: 0.066
[1, 28000] loss: 0.065
[1, 30000] loss: 0.061
Обучение закончено
In [ ]:
net 2D1.predict(testloader)
Accuracy of 0:99 %
Accuracy of
            1:99%
Accuracy of
            2:99%
Accuracy of
            3:99%
Accuracy of
            4:98%
           5:98%
Accuracy of
Accuracy of
           6:98%
           7:98%
Accuracy of
Accuracy of 8 : 99 %
Accuracy of 9:98%
Total accuracy: 98.91
In [ ]:
net 2D1.train network(trainloader inc, num epochs=1, learning rate=1e-4)
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:46: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:48: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
[1, 2000] loss: 0.061
[1, 4000] loss: 0.063
[1, 6000] loss: 0.064
[1, 8000] loss: 0.063
[1, 10000] loss: 0.057
[1, 12000] loss: 0.068
[1, 14000] loss: 0.067
[1, 16000] loss: 0.057
[1, 18000] loss: 0.055
[1, 20000] loss: 0.048
[1, 22000] loss: 0.063
[1, 24000] loss: 0.053
[1, 26000] loss: 0.055
[1, 28000] loss: 0.068
[1, 30000] loss: 0.062
Обучение закончено
In [ ]:
net 2D1.predict(testloader)
Accuracy of 0:99 %
Accuracy of 1:99 %
Accuracy of 2:98%
Accuracy of
           3:99%
Accuracy of
           4:98%
Accuracy of
            5:98%
            6:98%
Accuracy of
            7:99%
Accuracy of
            8: 98 %
Accuracy of
            9: 98 %
Accuracy of
Total accuracy: 98.97
```

```
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:46: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:48: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
[1, 2000] loss: 0.050
    4000] loss: 0.072
[1,
    6000] loss: 0.060
[1,
    8000] loss: 0.054
[1,
[1, 10000] loss: 0.054
[1, 12000] loss: 0.064
[1, 14000] loss: 0.058
[1, 16000] loss: 0.056
[1, 18000] loss: 0.056
[1, 20000] loss: 0.054
[1, 22000] loss: 0.060
[1, 24000] loss: 0.061
[1, 26000] loss: 0.061
[1, 28000] loss: 0.050
[1, 30000] loss: 0.066
Обучение закончено
In [ ]:
net 2D1.predict(testloader)
Accuracy of 0:99 %
Accuracy of
             1:99%
             2:98%
Accuracy of
            3:99%
Accuracy of
Accuracy of 4: 98 %
Accuracy of 5 : 98 %
Accuracy of 6 : 98 %
Accuracy of 7 : 99 %
Accuracy of 8:99 %
Accuracy of 9:98 %
Total accuracy: 99.07
Итак, добились точности 99.07%. Экспериментировали с разным количеством сверточных слоёв и разным
количеством полносвязных. Итого - два сверточных слоя и два полносвязных, между полносвязными - дропаут.
Ради интереса попробуем другую функцию активации
In [ ]:
import math
# класс наследуется от nn.Module
class SimpleConvNet2DLeaky(nn.Module):
         _init__(self, channels1, channels2, kernel_size1, kernel_size2, fc1, dropout, i
s max pool = True):
        # вызов конструктора предка
```

super(SimpleConvNet2DLeaky, self). init ()

про входящие картинки знать не нужно

new size = 28 - kernel size1 + 1

if is max pool:

else:

_size1)

которую будем подавать в сеть, больше ничего

self.pool = nn.MaxPool2d(kernel size=2, stride=2)

необходмо заранее знать, сколько каналов у картинки (сейчас = 1),

self.conv1 = nn.Conv2d(in channels=1, out channels=channels1, kernel size=kernel

net 2D1.train network(trainloader inc, num epochs=1, learning rate=1e-4)

In []:

```
self.pool = nn.AvgPool2d(kernel size=2, stride=2)
        new_size = new_size // 2
        self.conv2 = nn.Conv2d(in channels=channels1, out channels=channels2, kernel siz
e=kernel size2)
       new size = new size - kernel size2 + 1
       new size = new size // 2
        self.fc1 size = new size * new size * channels2
        self.fc1 = nn.Linear(new size * new size * channels2, fc1) # !!!
        #self.fc2 = nn.Linear(fc1, fc2)
        self.fc3 = nn.Linear(fc1, 10)
        self.dropout1 = Dropout(dropout)
        self.dropout2 = Dropout(dropout)
    def forward(self, x):
        x = self.pool(F.leaky_relu(self.conv1(x)))
        #print(x.shape)
        x = self.pool(F.leaky_relu(self.conv2(x)))
        #print(x.shape)
        x = x.view(-1, self.fc1 size) # !!!
        x = self.dropout1(F.leaky_relu(self.fc1(x)))
        \#x = self.dropout2(F.leaky relu(self.fc2(x)))
       x = self.fc3(x)
       print(x.shape)
       return x
    def train network(self, train loader, learning rate = 1e-4, num epochs = 3):
        loss fn = torch.nn.CrossEntropyLoss()
        optimizer = torch.optim.Adam(self.parameters(), lr=learning rate)
        # итерируемся
        for epoch in tqdm notebook(range(num epochs)):
            running loss = 0.0
            for i, batch in enumerate(tqdm notebook(train loader)):
                # так получаем текущий батч
                X batch, y batch = batch
                # обнуляем веса
                optimizer.zero grad()
                # forward + backward + optimize
                y pred = self(X batch)
                loss = loss fn(y pred, y_batch)
                loss.backward()
                optimizer.step()
                # выведем текущий loss
                running loss += loss.item()
                # выведем качество каждые 2000 батчей
                if i % 2000 == 1999:
                    print('[%d, %5d] loss: %.3f' %
                           (epoch + 1, i + 1, running loss / 2000))
                    running loss = 0.0
        print('Обучение закончено')
    def predict(self, test loader):
        class correct = list(0. for i in range(10))
        class_total = list(0. for i in range(10))
        self.eval()
        with torch.no grad():
            for data in test loader:
                images, labels = data
                y pred = self(images)
                _, predicted = torch.max(y pred, 1)
                c = (predicted == labels).squeeze()
                for i in range(4):
                    label = labels[i]
                    class correct[label] += c[i].item()
```

```
class total[label] += 1
        for i in range(10):
            print('Accuracy of %2s : %2d %%' % (
                classes[i], 100 * class correct[i] / class total[i]))
        class correct t = sum(class correct)
        class total t = sum(class total)
        print('\nTotal accuracy:', (100. * class correct t / class total t))
        self.train()
In [ ]:
net 2D1L = SimpleConvNet2DLeaky(6, 16, 5, 5, 200, 0.5) #с Мах пулингом
net 2D1L.train network(trainloader inc, num epochs=4, learning rate=1e-3)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:47: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
```

```
[1, 2000] loss: 0.564
[1, 4000] loss: 0.239
[1, 6000] loss: 0.177
[1, 8000] loss: 0.155
[1, 10000] loss: 0.145
[1, 12000] loss: 0.131
[1, 14000] loss: 0.127
[1, 16000] loss: 0.113
[1, 18000] loss: 0.111
[1, 20000] loss: 0.106
[1, 22000] loss: 0.122
[1, 24000] loss: 0.115
[1, 26000] loss: 0.106
[1, 28000] loss: 0.104
[1, 30000] loss: 0.096
[2, 2000] loss: 0.090
[2, 4000] loss: 0.097
[2, 6000] loss: 0.096
[2, 8000] loss: 0.110
[2, 10000] loss: 0.083
[2, 12000] loss: 0.091
[2, 14000] loss: 0.095
[2, 16000] loss: 0.087
[2, 18000] loss: 0.077
[2, 20000] loss: 0.088
[2, 22000] loss: 0.088
[2, 24000] loss: 0.097
[2, 26000] loss: 0.094
[2, 28000] loss: 0.089
[2, 30000] loss: 0.087
[3, 2000] loss: 0.091
[3, 4000] loss: 0.081
[3, 6000] loss: 0.079
[3, 8000] loss: 0.071
[3, 10000] loss: 0.087
[3, 12000] loss: 0.092
[3, 14000] loss: 0.087
[3, 16000] loss: 0.075
[3, 18000] loss: 0.077
[3, 20000] loss: 0.090
[3, 22000] loss: 0.083
[3, 24000] loss: 0.082
```

```
[3, 26000] loss: 0.083
[3, 28000] loss: 0.093
[3, 30000] loss: 0.085
    2000] loss: 0.067
[4,
    4000] loss: 0.078
[4,
[4,
    6000] loss: 0.082
[4, 8000] loss: 0.086
[4, 10000] loss: 0.083
[4, 12000] loss: 0.083
[4, 14000] loss: 0.081
[4, 16000] loss: 0.071
[4, 18000] loss: 0.090
[4, 20000] loss: 0.078
[4, 22000] loss: 0.078
[4, 24000] loss: 0.077
[4, 26000] loss: 0.096
[4, 28000] loss: 0.070
[4, 30000] loss: 0.073
Обучение закончено
In [ ]:
net 2D1L.predict(testloader)
Accuracy of 0:99 %
Accuracy of 1 : 99 %
Accuracy of 2 : 99 %
Accuracy of 3 : 99 %
Accuracy of 4: 98 %
Accuracy of 5 : 97 %
Accuracy of 6:98%
Accuracy of 7:99%
Accuracy of 8:98%
Accuracy of 9:97%
Total accuracy: 98.89
In [ ]:
net 2D1L.train network(trainloader inc, num epochs=1, learning rate=1e-4)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:45: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:47: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
   2000] loss: 0.073
[1, 4000] loss: 0.058
[1, 6000] loss: 0.052
[1, 8000] loss: 0.056
[1, 10000] loss: 0.050
[1, 12000] loss: 0.058
[1, 14000] loss: 0.051
[1, 16000] loss: 0.057
[1, 18000] loss: 0.049
[1, 20000] loss: 0.058
[1, 22000] loss: 0.049
[1, 24000] loss: 0.053
[1, 26000] loss: 0.051
[1, 28000] loss: 0.039
[1, 30000] loss: 0.046
Обучение закончено
In [ ]:
net 2D1L.predict(testloader)
```

```
Accuracy of 0 : 99 % Accuracy of 1 : 99 % Accuracy of 2 : 99 % Accuracy of 3 : 99 % Accuracy of 4 : 98 % Accuracy of 6 : 98 % Accuracy of 6 : 98 % Accuracy of 7 : 99 % Accuracy of 8 : 99 % Accuracy of 9 : 97 %
```

Total accuracy: 99.1

Улучшили ещё! Причём за меньшее число эпох, чем в случае с **ReLU** - всего **5**. Если продолжим дообучать, можем улучшить ещё.

Итак, лучший результат - 99.1%. Экспериментировал с разным количеством сверточных слоёв и разным количеством полносвязных. Итого - два сверточных слоя и два полносвязных, между полносвязными - дропаут. Увеличение количества свёрточных слоёв не помогло улучшить результат (к тому же там возникают проблемы с тем, что размера картинки не хватает - но можно исправить пэддингом и убрав пулинг). Экспериментировал также с разными функциями активациями - вместо ReLU брал LReLU, для которой и был получен лучший результат (за меньшее в сравнение с ReLU число эпох). Average pooling работал хуже, чем Мах.