Свёрточные нейронные сети: CIFAR10</h3>

Выполнила: Трофимова Екатерина Александровна, 20223

Теория СNN

В этом ноутбке мы посмотрим, насколько хорошо **CNN** будут предсказывать классы на более сложном датасете картинок -- **CIFAR10**.

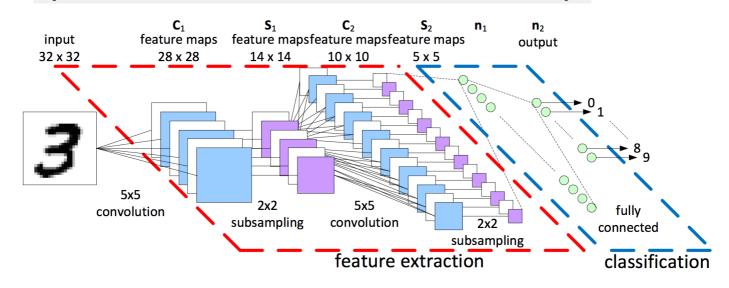
Внимание: Рассматривается задача классификации изображений.

Свёрточная нейросеть (Convolutional Neural Network, CNN) - это многослойная нейросеть, имеющая в своей архитектуре помимо полносвязных слоёв (а иногда их может и не быть) ещё и свёрточные слои (Conv Layers) и pooling-слои (Pool Layers).

Собственно, название такое эти сети получили потому, что в основе их работы лежит операция свёртки.

Сразу же стоит сказать, что свёрточные нейросети были придуманы прежде всего для задач, связанных с изображениями, следовательно, на вход они тоже "ожидают" изображение.

• Например, вот так выглядит неглубокая свёрточная нейросеть, имеющая такую архитектуру: Input -> Conv 5x5 -> Pool 2x2 -> Conv 5x5 -> Pool 2x2 -> FC -> Output



Свёрточные нейросети (простые, есть и намного более продвинутые) почти всегда строятся по следующему правилу:

INPUT -> [[CONV -> RELU]*N -> POOL?]*M -> [FC -> RELU]*L -> FC

то есть:

- 1). Входной слой: batch картинок -- тензор размера (batch_size, H, W, C) или (batch_size, C, H, W)
- **2).** M блоков (**M** \geq **0**) из свёрток и **pooling**-ов, причём именно в том порядке, как в формуле выше. Все эти M блоков вместе называют *feature extractor* свёрточной нейросети, потому что эта часть сети отвечает непосредственно за формирование новых, более сложных признаков поверх тех, которые подаются (то есть, по аналогии с **MI P**, мы опять же переходим к новому признаковому пространству, однако здесь оно строится

сложнее, чем в обычных многослойных сетях, поскольку используется операция свёртки)

3). L штук **FullyConnected-**слоёв (с активациями). Эту часть из L **FC-**слоёв называют **classificator**, поскольку эти слои отвечают непосредственно за предсказание нужно класса (сейчас рассматривается задача классификации изображений).

Свёрточная нейросеть на PyTorch

Ешё раз напомним про основные компоненты нейросети:

- непосредственно, сама архитектура нейросети (сюда входят типы функций активации у каждого нейрона);
- начальная инициализация весов каждого слоя;
- метод оптимизации нейросети (сюда ещё входит метод изменения learning rate);
- размер батчей (batch size);
- количетсво эпох обучения (num epochs);
- функция потерь (loss);
- тип регуляризации нейросети (weight decay, для каждого слоя можно свой);

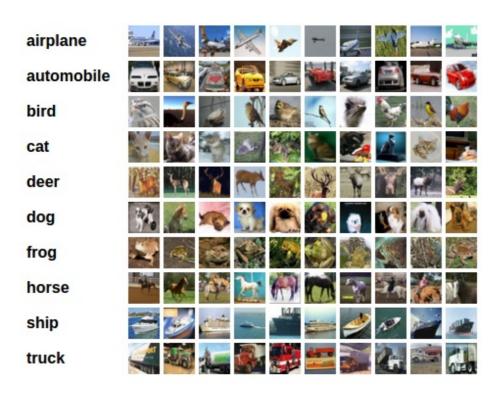
То, что связано с данными и задачей:

- само качество выборки (непротиворечивость, чистота, корректность постановки задачи);
- размер выборки;

Так как мы сейчас рассматриваем **архитектуру CNN**, то, помимо этих компонент, в свёрточной нейросети можно настроить следующие вещи:

- (в каждом ConvLayer) размер фильтров (окна свёртки) (kernel size)
- (в каждом ConvLayer) количество фильтров (out_channels)
- (в каждом ConvLayer) размер шага окна свёртки (stride) (stride)
- (в каждом ConvLayer) тип padding'a (padding)
- (в каждом PoolLayer) размер окна pooling'a (kernel size)
- (в каждом PoolLayer) шаг окна pooling'a (stride)
- (в каждом PoolLayer) тип pooling'a (pool type)
- (в каждом PoolLayer) тип padding'a (padding)

CIFAR10



```
CIFAR10: это набор из 60k картинок 32x32x3, 50k которых составляют обучающую выборку, и оставшиеся 10k -
тестовую. Классов в этом датасете 10: 'plane', 'car', 'bird', 'cat', 'deer', 'dog', 'frog',
'horse', 'ship', 'truck'.
In [2]:
# !pip install torch torchvision
In [3]:
import torch
import torchvision
from torchvision import transforms
from tqdm import tqdm_notebook
import numpy as np
import matplotlib.pyplot as plt
%matplotlib inline
In [4]:
transform = transforms.Compose(
    [transforms.ToTensor(),
     transforms.Normalize((0.5, 0.5, 0.5), (0.5, 0.5, 0.5))
trainset = torchvision.datasets.CIFAR10(root='../pytorch data', train=True,
                                         download=True, transform=transform)
trainloader = torch.utils.data.DataLoader(trainset, batch size=4,
                                           shuffle=True, num workers=2)
testset = torchvision.datasets.CIFAR10(root='../pytorch_data', train=False,
                                        download=True, transform=transform)
testloader = torch.utils.data.DataLoader(testset, batch size=4,
                                          shuffle=False, num workers=2)
classes = ('plane', 'car', 'bird', 'cat',
           'deer', 'dog', 'frog', 'horse', 'ship', 'truck')
Downloading https://www.cs.toronto.edu/~kriz/cifar-10-python.tar.gz to ../pytorch data/ci
far-10-python.tar.gz
Extracting ../pytorch data/cifar-10-python.tar.gz to ../pytorch data
Files already downloaded and verified
In [ ]:
trainset.data
In [6]:
trainloader.dataset.train list[0]
Out[6]:
['data batch 1', 'c99cafc152244af753f735de768cd75f']
In [7]:
# случайный индекс от 0 до размера тренировочной выборки
i = np.random.randint(low=0, high=50000)
plt.imshow(trainloader.dataset.data[i]);
 0
```

10

15

```
20 - 25 - 30 - 5 10 15 20 25 30
```

CNN для предсказания на CIFAR10.

Напишем свёрточную нейросеть для предсказания на CIFAR10

```
In [8]:
```

```
# Подключение зависимостей

import torch.nn as nn
import torch.nn.functional as F
from tqdm import tqdm_notebook
from torch.optim import lr_scheduler
```

Вспомогательные функции

```
In [9]:
```

```
# Попытка ускорить вычисления за счет дри
def get default_device():
    """Pick GPU if available, else CPU"""
    if torch.cuda.is available():
       return torch.device('cuda')
       return torch.device('cpu')
def to device(data, device):
    """Move tensor(s) to chosen device"""
    if isinstance(data, (list,tuple)):
        return [to device(x, device) for x in data]
    return data.to(device)
class DeviceDataLoader():
    """Wrap a dataloader to move data to a device"""
    def __init__(self, dl, device):
        self.dl = dl
        self.device = device
    def __iter__(self):
    """Yield a batch of data after moving it to device"""
        for b in self.dl:
            yield to device(b, self.device)
         len (self):
        """Number of batches"""
        return len(self.dl)
```

```
In [10]:
```

```
device = get_default_device()
device
Out[10]:
```

In [11]:

device(type='cpu')

```
#trainloader = DeviceDataLoader(trainloader, device)
```

```
#testloader = DeviceDataLoader(testloader, device)
```

In [12]:

```
# Функция для обучения модели
def train(net, epoch_num = 5, learning rate = 1e-3):
 loss fn = torch.nn.CrossEntropyLoss()
 optimizer = torch.optim.Adam(net.parameters(), lr=learning rate)
 # динамически изменяем LR
 scheduler = lr scheduler.CosineAnnealingLR(optimizer, T max=epoch num)
  # итерируемся
 for epoch in tqdm notebook(range(epoch num)):
   scheduler.step()
   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 = net(X batch)
       loss = loss_fn(y_pred, y_batch)
       loss.backward()
       optimizer.step()
       running loss += loss.item()
        # выводим 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')
```

In [13]:

```
# Функция для проверки качества
def check accuracy(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
 avg accuracy = 0
  for i in range(10):
   print('Accuracy of %5s: %2d %%' % (classes[i], 100 * class correct[i] / class total
[i]))
   avg_accuracy += 100 * class_correct[i] / class_total[i]
  print('Avg accuracy %2d %%' % (avg_accuracy / 10))
```

```
# Функция для визуальной проверки результата

def visualize_result(net, index):
    image = testloader.dataset.data[index]
    plt.imshow(image)

image = transform(image) # не забудем отмасштабировать!

y_pred = net(image.view(1, 3, 32, 32))
_, predicted = torch.max(y_pred, 1)

plt.title(f'Predicted: {classes[predicted.numpy()[0]]}')
```

Базовая архитектура

```
In [58]:
```

```
class SimpleConvNet(torch.nn.Module):
   def init (self):
       # вызов конструктора класса nn.Module()
       super(SimpleConvNet, self).__init__()
        # feature extractor
       self.pool = nn.MaxPool2d(kernel size=2, stride=2)
       self.conv1 = nn.Conv2d(in channels=3, out channels=6, kernel size=5)
       self.conv2 = nn.Conv2d(in channels=6, out channels=16, kernel size=5)
        # classificator
       self.fc1 = nn.Linear(5 * 5 * 16, 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)))
       x = self.pool(F.relu(self.conv2(x)))
       x = x.view(-1, 5 * 5 * 16)
       x = F.relu(self.fcl(x))
       x = F.relu(self.fc2(x))
       x = self.fc3(x)
       return x
```

In [59]:

```
net = SimpleConvNet()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0 Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1, 2000] loss: 1.893

[1, 4000] loss: 1.628

[1, 6000] loss: 1.506

[1, 8000] loss: 1.471

[1, 10000] loss: 1.431

[1, 12000] loss: 1.359
```

20001 1000 1 200

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```
ZUUU] 1055. 1.Z03
L 4 ,
    4000] loss: 1.288
[2,
[2, 6000] loss: 1.268
[2, 8000] loss: 1.272
[2, 10000] loss: 1.248
[2, 12000] loss: 1.236
    2000] loss: 1.152
[3,
    4000] loss: 1.146
[3,
     6000] loss: 1.132
    8000] loss: 1.153
[3,
[3, 10000] loss: 1.139
[3, 12000] loss: 1.120
     2000] loss: 1.031
[4,
    4000] loss: 1.037
[4,
     6000] loss: 1.026
[4,
    8000] loss: 1.052
[4,
[4, 10000] loss: 1.049
[4, 12000] loss: 1.039
    2000] loss: 0.947
[5,
[5,
    4000] loss: 0.950
    6000] loss: 0.949
[5,
[5, 8000] loss: 0.938
[5, 10000] loss: 0.947
[5, 12000] loss: 0.952
[6,
    2000] loss: 0.837
[6,
    4000] loss: 0.845
    6000] loss: 0.862
[6,
[6, 8000] loss: 0.884
[6, 10000] loss: 0.867
[6, 12000] loss: 0.878
     2000] loss: 0.798
[7,
    4000] loss: 0.780
[7,
    6000] loss: 0.796
[7,
    8000] loss: 0.778
[7, 10000] loss: 0.776
[7, 12000] loss: 0.806
    2000] loss: 0.717
[8,
[8, 4000] loss: 0.752
[8, 6000] loss: 0.720
[8, 8000] loss: 0.727
[8, 10000] loss: 0.732
[8, 12000] loss: 0.724
    2000] loss: 0.671
[9,
    40001 loss: 0.709
[9,
     60001 loss: 0.685
   8000] loss: 0.682
[9,
[9, 10000] loss: 0.712
[9, 12000] loss: 0.696
     2000] loss: 0.699
[10,
[10, 4000] loss: 0.677
[10, 6000] loss: 0.690
[10, 8000] loss: 0.686
[10, 10000] loss: 0.661
[10, 12000] loss: 0.669
Обучение закончено
```

Посмотрим на ассигасу на тестовом датасете:

```
Check_accuracy(net)

Accuracy of plane: 70 %
Accuracy of car: 77 %
Accuracy of bird: 50 %
Accuracy of cat: 44 %
```

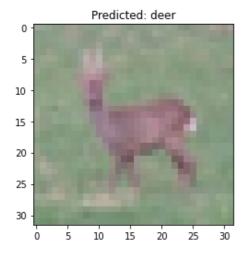
Accuracy of car : 77 % Accuracy of bird : 50 % Accuracy of cat : 44 % Accuracy of deer : 58 % Accuracy of dog : 51 % Accuracy of frog : 74 % Accuracy of horse : 71 % Accuracy of truck : 72 % Avg accuracy 64 %

При базовой архитектуре наблюдается средняя точность в районе 64% на 10 эпохах. Минимальная точность класса 44%. Среднее время вычисления на эпоху = 1:10

Проверим работу нейросети визуально (позапускайте ячейку несколько раз):

In [37]:

```
i = np.random.randint(low=0, high=10000)
visualize_result(net, i)
```



Эксперименты с числом сверточных слоев и каналов

Попробуем просто добавить новый сверточный слой

In [68]:

```
class ConvNet 3CL(nn.Module):
   def __init__(self):
        # вызов конструктора класса nn.Module()
       super(ConvNet 3CL, self). init ()
       self.pool = nn.MaxPool2d(kernel size=2, stride=2)
       self.conv1 = nn.Conv2d(in channels=3, out channels=6, kernel size=5)
       self.conv2 = nn.Conv2d(in_channels=6, out_channels=16, kernel_size=5)
       self.conv3 = nn.Conv2d(in channels=16, out channels=32, kernel size=5)
       self.fc1 = nn.Linear(3 * 3 * 32, 120)
       self.fc2 = nn.Linear(120, 84)
       self.fc3 = nn.Linear(84, 10)
   def forward(self, x):
       x = F.relu(self.conv1(x))
       x = self.pool(x)
       x = F.relu(self.conv2(x))
       x = F.relu(self.conv3(x))
       x = self.pool(x)
```

```
x = x.view(-1, 3 * 3 * 32)
x = F.relu(self.fc1(x))
x = F.relu(self.fc2(x))
x = self.fc3(x)
return x
```

In [69]:

```
net = ConvNet_3CL()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0 Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1, 2000] loss: 2.017
[1, 4000] loss: 1.828
[1, 6000] loss: 1.682
[1, 8000] loss: 1.572
[1, 10000] loss: 1.531
[1, 12000] loss: 1.507
[2, 2000] loss: 1.458
[2, 4000] loss: 1.427
    6000] loss: 1.412
[2,
[2,
   8000] loss: 1.393
[2, 10000] loss: 1.382
[2, 12000] loss: 1.362
[3,
   2000] loss: 1.302
   4000] loss: 1.314
[3,
[3, 6000] loss: 1.281
[3, 8000] loss: 1.275
[3, 10000] loss: 1.275
[3, 12000] loss: 1.268
[4, 2000] loss: 1.191
[4, 4000] loss: 1.183
[4, 6000] loss: 1.194
[4, 8000] loss: 1.202
[4, 10000] loss: 1.178
[4, 12000] loss: 1.187
[5, 2000] loss: 1.083
    4000] loss: 1.109
[5,
    6000] loss: 1.101
[5,
[5,
   8000] loss: 1.110
[5, 10000] loss: 1.115
[5, 12000] loss: 1.095
[6, 2000] loss: 1.014
[6,
   4000] loss: 1.024
   6000] loss: 1.003
[6,
   8000] loss: 1.015
[6,
```

[6, 10000] loss: 1.026 [6, 12000] loss: 1.029

```
[7, 2000] loss: 0.938
[7, 4000] loss: 0.932
    6000] loss: 0.966
[7,
   8000] loss: 0.935
[7,
[7, 10000] loss: 0.929
[7, 12000] loss: 0.946
[8, 2000] loss: 0.861
    4000] loss: 0.875
[8,
   6000] loss: 0.898
[8,
[8, 8000] loss: 0.867
[8, 10000] loss: 0.883
[8, 12000] loss: 0.875
   2000] loss: 0.829
[9,
    4000] loss: 0.839
[9,
   6000] loss: 0.834
[9,
[9, 8000] loss: 0.828
[9, 10000] loss: 0.844
[9, 12000] loss: 0.833
[10, 2000] loss: 0.800
[10, 4000] loss: 0.829
[10, 6000] loss: 0.820
[10, 8000] loss: 0.806
[10, 10000] loss: 0.832
[10, 12000] loss: 0.839
Обучение закончено
```

In [71]:

```
Check_accuracy(net)

Accuracy of plane: 64 %

Accuracy of car: 75 %

Accuracy of bird: 44 %
```

Accuracy of car: 75 %
Accuracy of bird: 44 %
Accuracy of cat: 44 %
Accuracy of deer: 54 %
Accuracy of dog: 46 %
Accuracy of frog: 72 %
Accuracy of horse: 66 %
Accuracy of ship: 75 %
Accuracy of truck: 70 %
Avg accuracy 61 %

Добавление еще одного сверточного слоя с малым количеством каналов отрицательно сказалось на качестве обучения (средний результат ухудшился до 61%) и времени обучения. Примерно на 9 эпохе процесс обучения застопорился. Попробуем теперь вернуться к 2м сверточным слоям, но увеличим число каналов

```
In [80]:
```

```
class ConvNet_2Cl(nn.Module):
    def __init__(self):
        # вызов конструктора класса nn.Module()
        super(ConvNet_2Cl, self).__init__()

        self.pool = nn.MaxPool2d(kernel_size=2, stride=2)

        self.conv1 = nn.Conv2d(in_channels=3, out_channels=64, kernel_size=5)
        self.conv2 = nn.Conv2d(in_channels=64, out_channels=128, kernel_size=5)

        self.fc1 = nn.Linear(5 * 5 * 128, 120)
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)

def forward(self, x):
        x = F.relu(self.conv1(x))
        x = self.pool(x)
```

```
x = F.relu(self.conv2(x))
x = self.pool(x)
x = x.view(-1, 5 * 5 * 128)
x = F.relu(self.fc1(x))
x = F.relu(self.fc2(x))
x = self.fc3(x)
return x
```

In [81]:

```
net = ConvNet_2Cl()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
  if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0 Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1, 2000] loss: 1.861
[1, 4000] loss: 1.567
[1, 6000] loss: 1.458
[1, 8000] loss: 1.388
[1, 10000] loss: 1.348
[1, 12000] loss: 1.317
[2, 2000] loss: 1.222
   4000] loss: 1.187
[2,
    6000] loss: 1.164
[2,
[2, 8000] loss: 1.159
[2, 10000] loss: 1.151
[2, 12000] loss: 1.104
[3,
   2000] loss: 1.024
   4000] loss: 1.002
[3,
[3, 6000] loss: 1.007
[3, 8000] loss: 0.995
[3, 10000] loss: 1.008
[3, 12000] loss: 1.017
[4, 2000] loss: 0.864
[4, 4000] loss: 0.889
[4, 6000] loss: 0.897
[4, 8000] loss: 0.899
[4, 10000] loss: 0.908
[4, 12000] loss: 0.894
[5, 2000] loss: 0.777
    4000] loss: 0.773
[5,
   6000] loss: 0.783
[5,
[5,
   8000] loss: 0.784
[5, 10000] loss: 0.789
[5, 12000] loss: 0.785
   2000] loss: 0.658
[6,
   4000] loss: 0.674
[6,
[6,
   6000] loss: 0.672
```

[6, 8000] loss: 0.669 [6, 10000] loss: 0.690

```
[6, 12000] loss: 0.668
   2000] loss: 0.566
[7,
    4000] loss: 0.575
    6000] loss: 0.579
[7,
    8000] loss: 0.581
[7, 10000] loss: 0.569
[7, 12000] loss: 0.551
[8, 2000] loss: 0.475
[8, 4000] loss: 0.489
[8, 6000] loss: 0.466
[8, 8000] loss: 0.501
[8, 10000] loss: 0.490
[8, 12000] loss: 0.495
    2000] loss: 0.431
[9,
    4000] loss: 0.418
    6000] loss: 0.420
[9, 8000] loss: 0.431
[9, 10000] loss: 0.427
[9, 12000] loss: 0.436
     2000] loss: 0.400
[10,
[10, 4000] loss: 0.420
[10, 6000] loss: 0.408
[10, 8000] loss: 0.403
[10, 10000] loss: 0.405
[10, 12000] loss: 0.416
Обучение закончено
```

In [82]:

```
check_accuracy(net)
```

```
Accuracy of plane: 76 %
Accuracy of car: 83 %
Accuracy of bird: 59 %
Accuracy of cat: 51 %
Accuracy of deer: 64 %
Accuracy of dog: 61 %
Accuracy of frog: 81 %
Accuracy of horse: 75 %
Accuracy of ship: 82 %
Accuracy of truck: 82 %
Avg accuracy 71 %
```

Увеличение числа каналов положительно сказалось на средней точности - 71% против базовых 64%. Но обучение в рамках эпохи теперь идет гораздо дольше. Попробуем одновременно увеличить число сверточных слоев и число каналов

In [24]:

```
class ConvNet_3CL_CH(nn.Module):
    def __init__(self):
        # BBB30B KOHCTPYKTOPA KJACCA nn.Module()
        super(ConvNet_3CL_CH, self).__init__()

        self.pool = nn.MaxPool2d(kernel_size=2, stride=2)

        self.conv1 = nn.Conv2d(in_channels=3, out_channels=64, kernel_size=5)
        self.conv2 = nn.Conv2d(in_channels=64, out_channels=128, kernel_size=5)
        self.conv3 = nn.Conv2d(in_channels=128, out_channels=256, kernel_size=5)

        self.fc1 = nn.Linear(256, 120) # 1 x 1 x 256
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)
```

```
def forward(self, x):
    x = F.relu(self.conv1(x))
    x = self.pool(x)
    x = F.relu(self.conv2(x))
    x = self.pool(x)
    x = F.relu(self.conv3(x))
    x = x.view(-1, 256)
    x = F.relu(self.fc1(x))
    x = F.relu(self.fc2(x))
    x = self.fc3(x)
```

In [25]:

```
net = ConvNet_3CL_CH()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0 Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
2000] loss: 1.913
    4000] loss: 1.647
[1,
    6000] loss: 1.544
[1,
    8000] loss: 1.488
[1,
[1, 10000] loss: 1.459
[1, 12000] loss: 1.390
[2, 2000] loss: 1.330
[2, 4000] loss: 1.299
[2, 6000] loss: 1.279
[2, 8000] loss: 1.257
[2, 10000] loss: 1.230
[2, 12000] loss: 1.227
[3, 2000] loss: 1.139
[3, 4000] loss: 1.139
[3, 6000] loss: 1.127
[3, 8000] loss: 1.101
[3, 10000] loss: 1.111
[3, 12000] loss: 1.075
[4, 2000] loss: 0.996
    4000] loss: 1.012
[4,
[4,
    6000] loss: 1.014
    8000] loss: 0.997
[4,
[4, 10000] loss: 0.991
[4, 12000] loss: 0.973
[5, 2000] loss: 0.875
[5, 4000] loss: 0.875
[5, 6000] loss: 0.890
[5, 8000] loss: 0.896
[5, 10000] loss: 0.892
[5, 12000] loss: 0.902
[6. 20001 loss: 0.777
```

```
4000] loss: 0.783
[6,
[6, 6000] loss: 0.794
[6, 8000] loss: 0.776
[6, 10000] loss: 0.798
[6, 12000] loss: 0.769
    2000] loss: 0.678
[7,
    4000] loss: 0.694
[7,
    6000] loss: 0.672
[7,
    8000] loss: 0.679
[7, 10000] loss: 0.680
[7, 12000] loss: 0.703
[8, 2000] loss: 0.607
[8, 4000] loss: 0.589
[8, 6000] loss: 0.600
[8, 8000] loss: 0.596
[8, 10000] loss: 0.588
[8, 12000] loss: 0.591
[9, 2000] loss: 0.529
[9, 4000] loss: 0.520
[9,
    6000] loss: 0.524
[9, 8000] loss: 0.544
[9, 10000] loss: 0.561
[9, 12000] loss: 0.528
[10, 2000] loss: 0.506
     4000] loss: 0.518
[10,
[10, 6000] loss: 0.514
[10, 8000] loss: 0.518
[10, 10000] loss: 0.534
[10, 12000] loss: 0.513
Обучение закончено
In [26]:
```

```
check_accuracy(net)
```

```
Accuracy of plane : 75 % Accuracy of car : 81 % Accuracy of bird : 54 % Accuracy of cat : 49 % Accuracy of deer : 67 % Accuracy of dog : 59 % Accuracy of frog : 79 % Accuracy of horse : 73 % Accuracy of ship : 81 % Accuracy of truck : 78 % Avg accuracy 70 %
```

Средняя точность стала чуть ниже - 70% против 71% на двух слоях. При этом сильно возросло время обучения. Теперь попробуем изменить размер ядра свертки для случая с двумя слоями

In [17]:

```
class ConvNet_2Cl_3KS(nn.Module):
    def __init__(self):
        # BBBOB KOHCTPYKTOPA KAACCA nn.Module()
        super(ConvNet_2Cl_3KS, self).__init__()

        self.pool = nn.MaxPool2d(kernel_size=2, stride=2)

        self.conv1 = nn.Conv2d(in_channels=3, out_channels=64, kernel_size=3)
        self.conv2 = nn.Conv2d(in_channels=64, out_channels=128, kernel_size=3)

        self.fc1 = nn.Linear(6 * 6 * 128, 120)
        self.fc2 = nn.Linear(120, 84)
```

```
def forward(self, x):
    x = F.relu(self.conv1(x))
    x = self.pool(x)
    x = F.relu(self.conv2(x))
    x = self.pool(x)
    x = self.pool(x)
    x = x.view(-1, 6 * 6 * 128)
    x = F.relu(self.fc1(x))
    x = F.relu(self.fc2(x))
    x = self.fc3(x)
    return x
```

In [18]:

```
net = ConvNet_2Cl_3KS()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
  if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0 Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1, 2000] loss: 1.801
    4000] loss: 1.463
[1,
    6000] loss: 1.395
[1,
[1,
    8000] loss: 1.290
[1, 10000] loss: 1.256
[1, 12000] loss: 1.211
[2, 2000] loss: 1.151
[2, 4000] loss: 1.105
[2, 6000] loss: 1.077
[2, 8000] loss: 1.052
[2, 10000] loss: 1.047
[2, 12000] loss: 1.039
[3, 2000] loss: 0.930
[3, 4000] loss: 0.924
[3, 6000] loss: 0.934
[3, 8000] loss: 0.945
[3, 10000] loss: 0.940
[3, 12000] loss: 0.934
[4, 2000] loss: 0.805
[4,
    4000] loss: 0.819
    6000] loss: 0.839
[4,
    8000] loss: 0.839
[4,
[4, 10000] loss: 0.847
[4, 12000] loss: 0.812
[5, 2000] loss: 0.714
[5, 4000] loss: 0.723
[5, 6000] loss: 0.730
```

[5, 8000] loss: 0.736 [5, 10000] loss: 0.749 [5, 12000] loss: 0.738

```
4000] loss: 0.627
[6,
[6,
     6000] loss: 0.637
    8000] loss: 0.647
[6,
[6, 10000] loss: 0.659
[6, 12000] loss: 0.649
[7,
     2000] loss: 0.557
     4000] loss: 0.558
[7,
     6000] loss: 0.559
[7,
[7,
    8000] loss: 0.555
[7, 10000] loss: 0.575
[7, 12000] loss: 0.564
[8,
    2000] loss: 0.511
[8, 4000] loss: 0.479
[8, 6000] loss: 0.505
[8, 8000] loss: 0.505
[8, 10000] loss: 0.491
[8, 12000] loss: 0.496
[9,
    2000] loss: 0.465
[9,
    4000] loss: 0.453
[9,
    6000] loss: 0.462
    8000] loss: 0.467
[9,
[9, 10000] loss: 0.473
[9, 12000] loss: 0.448
     2000] loss: 0.439
[10,
[10,
     4000] loss: 0.446
[10,
     6000] loss: 0.455
[10, 8000] loss: 0.442
[10, 10000] loss: 0.450
[10, 12000] loss: 0.457
Обучение закончено
In [19]:
check accuracy(net)
Accuracy of plane : 74 %
             car : 83 %
Accuracy of
Accuracy of
             bird : 57 %
Accuracy of
             cat : 52 %
             deer: 65
Accuracy of
Accuracy of
              dog : 63
Accuracy of
             frog : 77
Accuracy of horse : 77
Accuracy of ship: 81 %
Accuracy of truck : 81 %
Avg accuracy 71 %
```

Для двух слоев изменение размера ядра свертки не дало существенных изменений. Теперь посмотрим на **3**х слоях

```
In [27]:
```

[6,

2000] loss: 0.632

```
class ConvNet_3CL_CH_3KS(nn.Module):
    def __init__(self):
        # BBISOB ΚΟΗCΤΡΥΚΤΟΡΑ ΚΠΑCCA nn.Module()
        super(ConvNet_3CL_CH_3KS, self).__init__()

        self.pool = nn.MaxPool2d(kernel_size=2, stride=2)

        self.conv1 = nn.Conv2d(in_channels=3, out_channels=64, kernel_size=3)
        self.conv2 = nn.Conv2d(in_channels=64, out_channels=128, kernel_size=3)
        self.conv3 = nn.Conv2d(in_channels=128, out_channels=256, kernel_size=3)
```

```
self.fc1 = nn.Linear(2 * 2 * 256, 120)
self.fc2 = nn.Linear(120, 84)
self.fc3 = nn.Linear(84, 10)

def forward(self, x):
    x = F.relu(self.conv1(x))
    x = self.pool(x)
    x = F.relu(self.conv2(x))
    x = self.pool(x)
    x = F.relu(self.conv3(x))
    x = self.pool(x)
    x = x.view(-1, 2 * 2 * 256)
    x = F.relu(self.fc1(x))
    x = F.relu(self.fc2(x))
    x = self.fc3(x)
    return x
```

In [28]:

```
net = ConvNet_3CL_CH_3KS()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0

Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1,
   2000] loss: 1.869
[1, 4000] loss: 1.537
[1, 6000] loss: 1.431
[1, 8000] loss: 1.371
[1, 10000] loss: 1.279
[1, 12000] loss: 1.251
[2, 2000] loss: 1.146
[2, 4000] loss: 1.095
[2, 6000] loss: 1.078
[2, 8000] loss: 1.065
[2, 10000] loss: 1.071
[2, 12000] loss: 1.057
[3, 2000] loss: 0.934
[3, 4000] loss: 0.945
    6000] loss: 0.943
[3,
[3, 8000] loss: 0.919
[3, 10000] loss: 0.924
[3, 12000] loss: 0.914
[4, 2000] loss: 0.791
ſ4,
   4000] loss: 0.812
[4, 6000] loss: 0.798
[4,
   8000] loss: 0.824
[4, 10000] loss: 0.808
[4, 12000] loss: 0.813
[5, 2000] loss: 0.692
[5, 4000] loss: 0.697
[5, 6000] loss: 0.693
```

```
[5, 8000] loss: 0.722
[5, 10000] loss: 0.702
[5, 12000] loss: 0.689
     2000] loss: 0.589
[6,
    4000] loss: 0.567
[6,
    6000] loss: 0.565
[6,
    8000] loss: 0.579
[6,
[6, 10000] loss: 0.593
[6, 12000] loss: 0.611
[7,
    2000] loss: 0.471
    4000] loss: 0.479
[7,
    6000] loss: 0.489
[7,
    8000] loss: 0.469
[7,
[7, 10000] loss: 0.478
[7, 12000] loss: 0.490
    2000] loss: 0.375
[8,
    4000] loss: 0.401
[8, 6000] loss: 0.403
[8, 8000] loss: 0.379
[8, 10000] loss: 0.406
[8, 12000] loss: 0.399
     2000] loss: 0.351
[9,
     4000] loss: 0.339
[9,
     6000] loss: 0.323
[9,
[9,
    8000] loss: 0.336
[9, 10000] loss: 0.339
[9, 12000] loss: 0.343
[10,
    2000] loss: 0.318
[10,
     4000] loss: 0.321
[10, 6000] loss: 0.315
[10, 8000] loss: 0.324
[10, 10000] loss: 0.319
[10, 12000] loss: 0.321
Обучение закончено
```

In [29]:

check accuracy(net)

```
Accuracy of plane : 79 % Accuracy of car : 85 % Accuracy of bird : 60 % Accuracy of cat : 54 % Accuracy of deer : 70 % Accuracy of dog : 61 % Accuracy of frog : 79 % Accuracy of horse : 75 % Accuracy of ship : 82 % Accuracy of truck : 84 % Avg accuracy 73 %
```

А вот на **3**х слоях уже наблюдается небольшой прирост: **73**% против **70**%. Минимальная точность возросладо **54**%.

Добавление слоев и каналов позволило обогатить пространство признаков и улучшить тем самым результат классификации

Эксперименты с пулингом и нормализацией

Теперь попробуем поменять тип пулинга с max на avg

```
ın [∠U]:
class ConvNet 3Cl 3KS AvgPool(nn.Module):
    def __init__(self):
        # вызов конструктора класса nn.Module()
        super(ConvNet 3Cl 3KS AvgPool, self). init ()
        self.pool = nn.AvgPool2d(kernel size=2, stride=2)
        self.conv1 = nn.Conv2d(in channels=3, out channels=64, kernel size=3)
        self.conv2 = nn.Conv2d(in channels=64, out channels=128, kernel size=3)
        self.conv3 = nn.Conv2d(in channels=128, out channels=256, kernel size=3)
        self.fc1 = nn.Linear(2 * 2 * 256, 120)
        self.fc2 = nn.Linear(120, 84)
        self.fc3 = nn.Linear(84, 10)
    def forward(self, x):
        x = F.relu(self.conv1(x))
        x = self.pool(x)
        x = F.relu(self.conv2(x))
        x = self.pool(x)
        x = F.relu(self.conv3(x))
        x = self.pool(x)
        x = x.view(-1, 2 * 2 * 256)
       x = F.relu(self.fc1(x))
        x = F.relu(self.fc2(x))
        x = self.fc3(x)
        return x
In [21]:
net = ConvNet_3Cl_3KS_AvgPool()
train(net, learning rate=0.001, epoch num=10)
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
  if sys.path[0] == '':
/usr/local/lib/python3.7/dist-packages/torch/optim/lr scheduler.py:134: UserWarning: Dete
cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later,
you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step(
)`. Failure to do this will result in PyTorch skipping the first value of the learning r
ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjus
t-learning-rate
  "https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:17: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
[1, 2000] loss: 1.924
[1, 4000] loss: 1.625
[1, 6000] loss: 1.478
[1, 8000] loss: 1.390
[1, 10000] loss: 1.328
[1, 12000] loss: 1.277
[2, 2000] loss: 1.203
[2, 4000] loss: 1.143
   6000] loss: 1.126
[2,
[2, 8000] loss: 1.073
[2, 10000] loss: 1.062
[2, 12000] loss: 1.025
[3, 2000] loss: 0.942
[3, 4000] loss: 0.919
[3, 6000] loss: 0.919
[3, 8000] loss: 0.892
[3, 10000] loss: 0.890
[3. 120001 loss: 0.898
```

```
[4,
     2000] loss: 0.772
    4000] loss: 0.784
[4,
    6000] loss: 0.778
[4,
    8000] loss: 0.781
[4,
[4, 10000] loss: 0.763
[4, 12000] loss: 0.778
[5,
     2000] loss: 0.659
[5,
    4000] loss: 0.666
     6000] loss: 0.687
[5,
[5, 8000] loss: 0.645
[5, 10000] loss: 0.661
[5, 12000] loss: 0.657
    2000] loss: 0.552
[6,
[6,
    4000] loss: 0.547
[6,
    6000] loss: 0.556
[6,
    8000] loss: 0.566
[6, 10000] loss: 0.556
[6, 12000] loss: 0.574
    2000] loss: 0.466
    4000] loss: 0.458
[7,
    6000] loss: 0.473
[7,
[7, 8000] loss: 0.475
[7, 10000] loss: 0.464
[7, 12000] loss: 0.472
[8,
    2000] loss: 0.398
    4000] loss: 0.388
[8,
    6000] loss: 0.401
[8,
[8, 8000] loss: 0.375
[8, 10000] loss: 0.390
[8, 12000] loss: 0.400
    2000] loss: 0.347
[9,
    4000] loss: 0.355
[9,
[9,
    6000] loss: 0.355
[9, 8000] loss: 0.349
[9, 10000] loss: 0.340
[9, 12000] loss: 0.338
     2000] loss: 0.332
[10,
[10, 4000] loss: 0.332
[10, 6000] loss: 0.332
[10, 8000] loss: 0.330
[10, 10000] loss: 0.340
[10, 12000] loss: 0.334
Обучение закончено
In [22]:
check accuracy(net)
Accuracy of plane : 79 %
Accuracy of car: 84 %
Accuracy of
            bird : 64 %
Accuracy of
            cat : 57 %
Accuracy of
            deer : 74 %
Accuracy of
             dog : 64 %
Accuracy of frog: 82 %
Accuracy of horse: 78 %
Accuracy of ship: 85 %
Accuracy of truck : 83 %
Avg accuracy 75 %
```

минимальная - 57% против 54%. Т.е. положение признака оказалось немного важнее его нличия. Попробуем

добавить нормализацию

```
In [23]:
```

```
class ConvNet_3Cl_3KS_AvgPool_BN (nn.Module):
   def init (self):
       # вызов конструктора класса nn.Module()
       super(ConvNet_3Cl_3KS_AvgPool_BN, self).__init__()
       self.pool = nn.AvgPool2d(kernel size=2, stride=2)
       self.conv1 = nn.Conv2d(in channels=3, out channels=64, kernel size=3)
       self.bn1 = nn.BatchNorm2d(64)
       self.conv2 = nn.Conv2d(in channels=64, out channels=128, kernel size=3)
       self.bn2 = nn.BatchNorm2d(128)
       self.conv3 = nn.Conv2d(in channels=128, out channels=256, kernel size=3)
       self.bn3 = nn.BatchNorm2d(256)
       self.fc1 = nn.Linear(2 * 2 * 256, 120)
       self.fc2 = nn.Linear(120, 84)
       self.fc3 = nn.Linear(84, 10)
   def forward(self, x):
       x = self.bn1(F.relu(self.conv1(x)))
       x = self.pool(x)
       x = self.bn2(F.relu(self.conv2(x)))
       x = self.pool(x)
       x = self.bn3(F.relu(self.conv3(x)))
       x = self.pool(x)
       x = x.view(-1, 2 * 2 * 256)
       x = F.relu(self.fcl(x))
       x = F.relu(self.fc2(x))
       x = self.fc3(x)
       return x
```

In [24]:

```
net = ConvNet_3Cl_3KS_AvgPool_BN()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0 Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`

```
[1, 2000] loss: 1.862

[1, 4000] loss: 1.600

[1, 6000] loss: 1.479

[1, 8000] loss: 1.357

[1, 10000] loss: 1.289

[1, 12000] loss: 1.241

[2, 2000] loss: 1.110

[2, 4000] loss: 1.081

[2, 6000] loss: 1.094

[2, 8000] loss: 1.068

[2, 10000] loss: 1.066

[2, 12000] loss: 1.029
```

```
2000] loss: 0.943
    4000] loss: 0.903
[3,
[3, 6000] loss: 0.941
[3, 8000] loss: 0.920
[3, 10000] loss: 0.896
[3, 12000] loss: 0.898
     2000] loss: 0.770
[4,
     4000] loss: 0.777
[4,
     6000] loss: 0.787
[4,
    8000] loss: 0.769
[4,
[4, 10000] loss: 0.794
[4, 12000] loss: 0.782
[5,
     2000] loss: 0.660
     4000] loss: 0.668
[5,
[5,
     6000] loss: 0.651
    8000] loss: 0.677
[5,
[5, 10000] loss: 0.651
[5, 12000] loss: 0.649
    2000] loss: 0.547
[6, 4000] loss: 0.538
[6, 6000] loss: 0.551
[6, 8000] loss: 0.546
[6, 10000] loss: 0.535
[6, 12000] loss: 0.542
[7,
     2000] loss: 0.451
    4000] loss: 0.444
[7,
    6000] loss: 0.451
[7,
[7,
   8000] loss: 0.452
[7, 10000] loss: 0.446
[7, 12000] loss: 0.438
[8,
    2000] loss: 0.377
[8,
    4000] loss: 0.371
    6000] loss: 0.381
[8,
[8, 8000] loss: 0.376
[8, 10000] loss: 0.374
[8, 12000] loss: 0.366
[9,
    2000] loss: 0.321
[9, 4000] loss: 0.342
[9, 6000] loss: 0.330
[9, 8000] loss: 0.343
[9, 10000] loss: 0.323
[9, 12000] loss: 0.337
[10, 2000] loss: 0.322
[10, 4000] loss: 0.319
     6000] loss: 0.319
[10,
[10, 8000] loss: 0.331
[10, 10000] loss: 0.325
[10, 12000] loss: 0.329
Обучение закончено
```

In [25]:

check_accuracy(net)

Accuracy of plane : 81 % Accuracy of car : 85 % Accuracy of bird : 65 % Accuracy of cat : 57 % Accuracy of deer : 73 % Accuracy of dog : 64 %

```
Accuracy of frog : 81 % Accuracy of horse : 81 % Accuracy of ship : 83 % Accuracy of truck : 82 % Avg accuracy 75 %
```

Нормализация не повлияла на результат

Эксперимент с функцией активации

Обычно в качестве функции активации сверточных слоев используют функцию **ReLU**. Рассматривать функции активации вроде сигмоидной или тангенциальной мы не будем, т.к. они приводят к проблемам с затуханием или увеличением градиентов. Вместо этого попробуем использовать **ELU**, которая сохраняет преимущества **ReLU** и помогает избежать проблемы умирающего **ReLU**

In [27]:

```
class ConvNet 3Cl 3KS AvgPool ELU(nn.Module):
   def __init__(self):
        # вызов конструктора класса nn.Module()
       super(ConvNet 3Cl 3KS AvgPool ELU, self). init ()
       self.pool = nn.AvgPool2d(kernel size=2, stride=2)
       self.conv1 = nn.Conv2d(in channels=3, out channels=64, kernel size=3)
       self.conv2 = nn.Conv2d(in channels=64, out channels=128, kernel size=3)
       self.conv3 = nn.Conv2d(in channels=128, out channels=256, kernel size=3)
       self.fc1 = nn.Linear(2 * 2 * 256, 120)
       self.fc2 = nn.Linear(120, 84)
       self.fc3 = nn.Linear(84, 10)
   def forward(self, x):
       x = F.elu(self.conv1(x))
       x = self.pool(x)
       x = F.elu(self.conv2(x))
       x = self.pool(x)
       x = F.elu(self.conv3(x))
       x = self.pool(x)
       x = x.view(-1, 2 * 2 * 256)
       x = F.elu(self.fcl(x))
       x = F.elu(self.fc2(x))
       x = self.fc3(x)
       return x
```

In [30]:

```
net = ConvNet_3Cl_3KS_AvgPool_ELU()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0 Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1, 2000] loss: 1.932
[1, 4000] loss: 1.686
```

```
1000, 1000. 1.000
    6000] loss: 1.577
[1, 8000] loss: 1.484
[1, 10000] loss: 1.433
[1, 12000] loss: 1.373
[2, 2000] loss: 1.255
[2, 4000] loss: 1.227
[2, 6000] loss: 1.192
[2, 8000] loss: 1.178
[2, 10000] loss: 1.161
[2, 12000] loss: 1.134
[3, 2000] loss: 0.965
[3, 4000] loss: 0.994
[3, 6000] loss: 0.978
[3, 8000] loss: 0.978
[3, 10000] loss: 0.967
[3, 12000] loss: 0.973
[4, 2000] loss: 0.803
[4, 4000] loss: 0.784
[4, 6000] loss: 0.800
[4, 8000] loss: 0.815
[4, 10000] loss: 0.808
[4, 12000] loss: 0.828
[5,
   2000] loss: 0.639
[5, 4000] loss: 0.634
[5, 6000] loss: 0.651
[5, 8000] loss: 0.655
[5, 10000] loss: 0.642
[5, 12000] loss: 0.645
[6, 2000] loss: 0.480
[6, 4000] loss: 0.465
[6, 6000] loss: 0.501
[6, 8000] loss: 0.501
[6, 10000] loss: 0.508
[6, 12000] loss: 0.495
[7, 2000] loss: 0.340
[7, 4000] loss: 0.340
[7, 6000] loss: 0.340
[7, 8000] loss: 0.358
[7, 10000] loss: 0.359
[7, 12000] loss: 0.369
[8, 2000] loss: 0.235
[8, 4000] loss: 0.242
[8, 6000] loss: 0.251
[8, 8000] loss: 0.240
[8, 10000] loss: 0.253
[8, 12000] loss: 0.231
[9, 2000] loss: 0.190
[9, 4000] loss: 0.170
[9, 6000] loss: 0.178
[9, 8000] loss: 0.174
[9, 10000] loss: 0.180
[9, 12000] loss: 0.181
[10, 2000] loss: 0.163
[10, 4000] loss: 0.160
[10, 6000] loss: 0.145
[10, 8000] loss: 0.164
[10, 10000] loss: 0.156
[10, 12000] loss: 0.162
```

061711011140 namatitiatia

```
ООУЧЕНИЕ ЗАКОНЧЕНО
```

In [46]:

```
Check_accuracy(net)

Accuracy of plane : 78 %
Accuracy of car : 83 %
Accuracy of bird : 63 %
Accuracy of cat : 55 %
Accuracy of deer : 66 %
Accuracy of dog : 64 %
Accuracy of frog : 78 %
Accuracy of horse : 74 %
Accuracy of ship : 81 %
Accuracy of truck : 80 %
Avg accuracy 72 %
```

Несмотря на то, что значение функции потерь теперь меньше, точность тоже упала. **ELU** не дает нам выигрыша на текущей архитектуре

Эксперимент с числом полносвязных слоев

Начнем с простого - уберем 1 слой

```
In [14]:
```

```
class ConvNet 3Cl 3KS AvgPool 2Fl(nn.Module):
   def __init__(self):
        # вызов конструктора класса nn.Module()
        super(ConvNet_3Cl_3KS_AvgPool_2Fl, self).__init__()
       self.pool = nn.AvgPool2d(kernel size=2, stride=2)
       self.conv1 = nn.Conv2d(in channels=3, out channels=64, kernel size=3)
        self.conv2 = nn.Conv2d(in channels=64, out channels=128, kernel size=3)
        self.conv3 = nn.Conv2d(in channels=128, out channels=256, kernel size=3)
       self.fc1 = nn.Linear(2 * 2 * 256, 512)
       self.fc3 = nn.Linear(512, 10)
   def forward(self, x):
       x = F.elu(self.conv1(x))
       x = self.pool(x)
       x = F.elu(self.conv2(x))
       x = self.pool(x)
       x = F.elu(self.conv3(x))
       x = self.pool(x)
       x = x.view(-1, 2 * 2 * 256)
       x = F.elu(self.fc1(x))
       x = self.fc3(x)
       return x
```

In [15]:

```
net = ConvNet_3Cl_3KS_AvgPool_2Fl()
train(net, learning_rate=0.001, epoch_num=10)

/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
   if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr_scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning)
/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1, 2000] loss: 2.004
[1, 4000] loss: 1.805
[1, 6000] loss: 1.700
[1, 8000] loss: 1.636
[1, 10000] loss: 1.590
[1, 12000] loss: 1.529
[2, 2000] loss: 1.410
[2, 4000] loss: 1.411
    6000] loss: 1.376
[2,
[2, 8000] loss: 1.355
[2, 10000] loss: 1.306
[2, 12000] loss: 1.327
[3,
    2000] loss: 1.140
    4000] loss: 1.174
[3,
    6000] loss: 1.159
[3,
    8000] loss: 1.146
[3, 10000] loss: 1.167
[3, 12000] loss: 1.126
   2000] loss: 0.886
[4, 4000] loss: 0.953
[4, 6000] loss: 0.900
[4, 8000] loss: 0.969
[4, 10000] loss: 0.936
[4, 12000] loss: 0.958
[5,
    2000] loss: 0.654
[5, 4000] loss: 0.692
    6000] loss: 0.686
    8000] loss: 0.695
[5,
[5, 10000] loss: 0.737
[5, 12000] loss: 0.715
[6,
    2000] loss: 0.439
    4000] loss: 0.430
[6,
[6,
    6000] loss: 0.449
[6, 8000] loss: 0.449
[6, 10000] loss: 0.460
[6, 12000] loss: 0.463
    2000] loss: 0.232
[7, 4000] loss: 0.229
[7, 6000] loss: 0.235
[7, 8000] loss: 0.247
[7, 10000] loss: 0.247
[7, 12000] loss: 0.247
[8,
    2000] loss: 0.109
    4000] loss: 0.108
[8,
    6000] loss: 0.104
[8,
   8000] loss: 0.095
[8, 10000] loss: 0.104
[8, 12000] loss: 0.110
[9,
    2000] loss: 0.050
[9,
    4000] loss: 0.048
     6000] loss: 0.049
[9,
    8000] loss: 0.049
[9, 10000] loss: 0.050
[9, 12000] loss: 0.044
```

```
[10, 2000] loss: 0.036
[10, 4000] loss: 0.035
[10, 6000] loss: 0.035
[10, 8000] loss: 0.034
[10, 10000] loss: 0.029
[10, 12000] loss: 0.035
Обучение закончено
In [16]:
check accuracy(net)
Accuracy of plane : 74 %
            car : 80 %
Accuracy of
Accuracy of bird: 62 %
Accuracy of cat: 53 %
Accuracy of deer: 66 %
Accuracy of
            dog : 59 %
Accuracy of frog: 76 %
Accuracy of horse: 76 %
Accuracy of ship: 82 %
Accuracy of truck: 79 %
Avg accuracy 70 %
И наоборот - добавим 1 слой
In [17]:
class ConvNet 3Cl 3KS AvgPool 4Fl(nn.Module):
    def __init__(self):
        \# вызов конструктора класса nn.Module()
        super(ConvNet_3Cl_3KS_AvgPool_4Fl, self).__init__()
        self.pool = nn.AvgPool2d(kernel size=2, stride=2)
        self.conv1 = nn.Conv2d(in channels=3, out channels=64, kernel size=3)
        self.conv2 = nn.Conv2d(in channels=64, out channels=128, kernel size=3)
        self.conv3 = nn.Conv2d(in channels=128, out channels=256, kernel size=3)
        self.fc1 = nn.Linear(2 * 2 * 256, 512)
        self.fc2 = nn.Linear(512, 256)
        self.fc3 = nn.Linear(256, 128)
        self.fc4 = nn.Linear(128, 10)
    def forward(self, x):
        x = F.elu(self.conv1(x))
        x = self.pool(x)
        x = F.elu(self.conv2(x))
        x = self.pool(x)
        x = F.elu(self.conv3(x))
        x = self.pool(x)
        x = x.view(-1, 2 * 2 * 256)
        x = F.elu(self.fc1(x))
       x = F.elu(self.fc2(x))
        x = F.elu(self.fc3(x))
        x = self.fc4(x)
        return x
In [18]:
net = ConvNet 3Cl 3KS AvgPool 4Fl()
train(net, learning rate=0.001, epoch num=10)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm_notebook`
  if sys.path[0] == '':
/usr/local/lib/python3.7/dist-packages/torch/optim/lr scheduler.py:134: UserWarning: Dete
```

cted call of `lr scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later,

you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning)/usr/local/lib/python3.7/dist-packages/ipykernel_launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0

Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
[1, 2000] loss: 2.077
[1, 4000] loss: 1.844
[1, 6000] loss: 1.739
[1, 8000] loss: 1.624
[1, 10000] loss: 1.540
[1, 12000] loss: 1.471
[2, 2000] loss: 1.806
[2, 4000] loss: 1.525
[2, 6000] loss: 1.382
[2, 8000] loss: 1.323
[2, 10000] loss: 1.313
[2, 12000] loss: 1.270
   2000] loss: 1.257
[3,
[3,
    4000] loss: 1.153
   6000] loss: 1.158
[3, 8000] loss: 1.152
[3, 10000] loss: 1.090
[3, 12000] loss: 1.101
[4, 2000] loss: 1.003
[4, 4000] loss: 0.963
[4, 6000] loss: 0.961
[4, 8000] loss: 0.964
[4, 10000] loss: 1.006
[4, 12000] loss: 0.934
[5, 2000] loss: 0.785
[5, 4000] loss: 0.841
[5, 6000] loss: 0.800
[5, 8000] loss: 0.804
[5, 10000] loss: 0.811
[5, 12000] loss: 0.808
   2000] loss: 0.641
[6,
   4000] loss: 0.629
[6,
[6, 6000] loss: 0.633
[6, 8000] loss: 0.634
[6, 10000] loss: 0.647
[6, 12000] loss: 0.668
[7, 2000] loss: 0.515
[7, 4000] loss: 0.485
[7, 6000] loss: 0.472
[7, 8000] loss: 0.474
[7, 10000] loss: 0.471
[7, 12000] loss: 0.474
[8, 2000] loss: 0.355
[8, 4000] loss: 0.351
[8, 6000] loss: 0.352
[8, 8000] loss: 0.343
[8, 10000] loss: 0.360
[8, 12000] loss: 0.341
    2000] loss: 0.265
[9,
```

4000] loss: 0.266

60001 loss: 0.265

[9,

```
[9, 8000] loss: 0.260

[9, 10000] loss: 0.276

[9, 12000] loss: 0.268

[10, 2000] loss: 0.242

[10, 4000] loss: 0.244

[10, 6000] loss: 0.236

[10, 8000] loss: 0.237

[10, 10000] loss: 0.239

[10, 12000] loss: 0.244

Обучение закончено
```

```
check accuracy(net)
Accuracy of plane: 78 %
Accuracy of
             car : 82 %
Accuracy of bird: 60 %
Accuracy of
            cat : 56 %
Accuracy of
           deer : 68 %
Accuracy of
             dog : 55 %
Accuracy of frog: 79 %
Accuracy of horse : 74 \%
Accuracy of ship: 80 %
Accuracy of truck: 79 %
Avg accuracy 71 %
```

В обоих случаях точность классификации снизилась. Для текущей архитектуры оптимальным является наличие 3 линейных слоев

Сильная архитектура, которая уже была (!) в ноутбуке

Попробуем обучить ещё более сильную нейросеть:

In [15]:

```
class StrongConvNet(nn.Module):
        __init__(self):
        # вызов конструктора класса nn.Module()
       super(StrongConvNet, self). init ()
       self.pool = nn.MaxPool2d(kernel size=2, stride=2)
       self.dropout = nn.Dropout(p=0.2)
       self.conv1 = nn.Conv2d(in channels=3, out channels=8, kernel size=5)
       self.bn1 = nn.BatchNorm2d(8)
       self.conv2 = nn.Conv2d(in channels=8, out channels=16, kernel size=1)
       self.bn2 = nn.BatchNorm2d(16)
       self.conv3 = nn.Conv2d(in channels=16, out channels=16, kernel size=3)
       self.bn3 = nn.BatchNorm2d(16)
       self.conv4 = nn.Conv2d(in channels=16, out channels=32, kernel size=1)
       self.bn4 = nn.BatchNorm2d(32)
       self.conv5 = nn.Conv2d(in channels=32, out channels=32, kernel size=3)
       self.bn5 = nn.BatchNorm2d(32)
       self.fc1 = nn.Linear(4 * 4 * 32, 128)
       self.fc2 = nn.Linear(128, 10)
   def forward(self, x):
       x = self.bn1(F.relu(self.conv1(x)))
       x = self.pool(x)
       x = self.bn2(F.relu(self.conv2(x)))
       x = self.bn3(F.relu(self.conv3(x)))
       x = self.pool(x)
       x = self.bn4(F.relu(self.conv4(x)))
       x = self.bn5(F.relu(self.conv5(x)))
```

```
print(x.shape)
x = x.view(-1, 4 * 4 * 32)
x = F.relu(self.fc1(x))
x = self.dropout(x)
x = self.fc2(x)
return x
```

Обучим:

```
In [16]:
```

```
net = StrongConvNet()
train(net, learning rate=0.001, epoch num=10)
/usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:12: TqdmDeprecationWarning:
This function will be removed in tqdm==5.0.0
Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`
  if sys.path[0] == '':
```

/usr/local/lib/python3.7/dist-packages/torch/optim/lr scheduler.py:134: UserWarning: Dete cted call of `lr_scheduler.step()` before `optimizer.step()`. In PyTorch 1.1.0 and later, you should call them in the opposite order: `optimizer.step()` before `lr_scheduler.step()`. Failure to do this will result in PyTorch skipping the first value of the learning r ate schedule. See more details at https://pytorch.org/docs/stable/optim.html#how-to-adjus t-learning-rate

"https://pytorch.org/docs/stable/optim.html#how-to-adjust-learning-rate", UserWarning) /usr/local/lib/python3.7/dist-packages/ipykernel launcher.py:17: TqdmDeprecationWarning: This function will be removed in tqdm==5.0.0

Please use `tqdm.notebook.tqdm` instead of `tqdm.tqdm notebook`

```
2000] loss: 1.858
[1,
    4000] loss: 1.665
[1,
[1,
   6000] loss: 1.570
[1, 8000] loss: 1.522
[1, 10000] loss: 1.473
[1, 12000] loss: 1.398
   2000] loss: 1.300
[2,
   4000] loss: 1.293
[2,
[2,
   6000] loss: 1.291
[2, 8000] loss: 1.292
[2, 10000] loss: 1.263
[2, 12000] loss: 1.233
[3, 2000] loss: 1.156
[3, 4000] loss: 1.149
[3, 6000] loss: 1.156
   8000] loss: 1.148
[3,
[3, 10000] loss: 1.144
[3, 12000] loss: 1.148
[4, 2000] loss: 1.049
   4000] loss: 1.031
[4,
[4,
   6000] loss: 1.073
[4, 8000] loss: 1.057
[4, 10000] loss: 1.026
[4, 12000] loss: 1.039
   2000] loss: 0.974
[5,
    4000] loss: 0.984
[5,
[5,
   6000] loss: 0.998
[5,
   8000] loss: 0.966
[5, 10000] loss: 0.951
[5, 12000] loss: 0.955
   2000] loss: 0.905
[6,
    4000] loss: 0.910
[6,
    6000] loss: 0.907
[6,
```

```
[6, 8000] loss: 0.909
[6, 10000] loss: 0.898
[6, 12000] loss: 0.913
    2000] loss: 0.852
[7,
    4000] loss: 0.838
[7,
    6000] loss: 0.847
[7,
   8000] loss: 0.861
[7,
[7, 10000] loss: 0.830
[7, 12000] loss: 0.860
    2000] loss: 0.807
[8,
    4000] loss: 0.825
[8,
    60001 loss: 0.827
[8, 8000] loss: 0.797
[8, 10000] loss: 0.819
[8, 12000] loss: 0.827
[9,
    2000] loss: 0.775
    4000] loss: 0.797
[9,
    6000] loss: 0.779
[9,
    8000] loss: 0.781
[9,
[9, 10000] loss: 0.802
[9, 12000] loss: 0.791
[10, 2000] loss: 0.782
[10, 4000] loss: 0.793
[10, 6000] loss: 0.773
[10, 8000] loss: 0.795
[10, 10000] loss: 0.765
[10, 12000] loss: 0.764
Обучение закончено
```

In [17]:

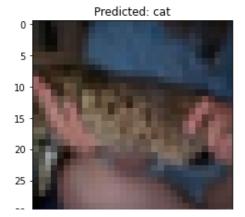
check_accuracy(net)

```
Accuracy of plane : 67 %
            car : 77 %
Accuracy of
            bird : 50 %
Accuracy of
Accuracy of
             cat : 48 %
Accuracy of
            deer : 60
Accuracy of
            dog : 51
Accuracy of frog: 72 %
Accuracy of horse : 69 %
Accuracy of ship: 75 %
Accuracy of truck: 73 %
Avg accuracy 64 %
```

Посмотрим визуально на работу нейросети:

In []:

```
i = np.random.randint(low=0, high=10000)
visualize result(i)
```



Лучшая архитектура

Итого: наилучший результат **75**% точности в среднем, максимальной точности в **85**% для отдельного класса и минимальной точности в **57**% для отдельного класса. Результат был достигнут при **3**х сверточных слоях с увеличенным числом каналов и меньшим ядром свертки для получения большего пространства признаков (что приводит также к сильному увеличению времени обучения), с **avg** пулингом, **ReLU** в качестве функции активации и **3**мя линейными полносвязными слоями в качестве классификатора.

Даже обучив более глубокую и прокаченную (BatchNorm, Dropout) нейросеть на этих данных мы видим, что качество нас всё ещё не устраивает, в реальной жизни необходимо ошибаться не больше, чем на 5%, а часто и это уже много. Как же быть, ведь свёрточные нейросети должны хорошо классифицировать изображения?

К сожалению, обучение нейросети с нуля на не очень большой выборке (а здесь она именно такая) часто приводит к переобучению, что плохо сказывается на тестовом качестве.

Для того, чтобы получить более качественную модель, часто **до**обучают сильную нейросеть, обученную на **ImageNet**, то есть используют технику **Transfer Learning**. О ней речь пойдёт далее в нашем курсе.

Полезные ссылки

- 1). Примеры написания нейросетей на **PyTorch** (официальные туториалы) (на английском): https://pytorch.org/tutorials/beginner/pytorch with examples. https://pytorch.org/tutorials/beginner/blitz/cifar10 tutorial.html
- **2).** Курс Стэнфорда: http://cs231n.github.io/
- 3). Практически исчерпывающая информация по основам свёрточных нейросетей (из cs231n) (на английском):

http://cs231n.github.io/convolutional-networks/ http://cs231n.github.io/understanding-cnn/ http://cs231n.github.io/transfer-learning/

4). Видео о Computer Vision от Andrej Karpathy: https://www.youtube.com/watch?v=u6aEYuemt0M