

Лабораторная работа №2. Введение в
проектирование нейронных сетей с помощью Python
По предмету "Киберфизические системы и
технологии"

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```
[1]: """
https://github.com/vedaant-varshney/ImageClassifierCNN/blob/master/
    →Image%20Classifier.ipynb

https://vitalflux.com/
    →different-types-of-cnn-architectures-explained-examples

https://sahiltinky94.medium.com/
    →know-about-googlenet-and-implementation-using-pytorch-92f827d675db
"""

from IPython.display import clear_output
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import os
import random
import scipy
import torch
import torch.nn as nn
import torch.nn.functional as F
import gc
import torch.backends.cudnn

num_workers = 0
batch_size = 100
valid_size = 0.2
learning_rate = 0.01
learning_momentum = 0.9
MODEL_NAME = "LeNet"

torch.cuda.synchronize()
torch.cuda.empty_cache()
gc.collect()

torch.use_deterministic_algorithms(False)
torch.backends.cudnn.benchmark = True

print("Is cuda available?", torch.cuda.is_available())
print("Is cuDNN version:", torch.backends.cudnn.version())
print("cuDNN enabled? ", torch.backends.cudnn.enabled)
```

```

print("cuDNN benchmark?", torch.backends.cudnn.benchmark)
print("is Use Deterministic Algorithms?", torch.backends.cudnn.
    ↪deterministic)

cuda = torch.device('cuda')
print(torch.cuda.get_device_properties(cuda))

```

```

Is cuda available? True
Is cuDNN version: 8302
cuDNN enabled? True
cuDNN benchmark? True
is Use Deterministic Algorithms? False
_CudaDeviceProperties(name='NVIDIA GeForce RTX 3080 Laptop GPU', major=8,
minor=6, total_memory=8191MB, multi_processor_count=48)

```

```

[2]: # mnist_train = np.genfromtxt(f"dataset/mnist_train.csv", delimiter=',', ↪
    ↪dtype=np.uint8)
# mnist_test = np.genfromtxt(f"dataset/mnist_test.csv", delimiter=',', ↪
    ↪dtype=np.uint8)

mnist_train = pd.read_csv(f"dataset/mnist_train.csv", header=None, ↪
    ↪engine="c", delimiter=',', dtype=np.uint8).to_numpy()
mnist_test = pd.read_csv(f"dataset/mnist_test.csv", header=None, ↪
    ↪engine="c", delimiter=',', dtype=np.uint8).to_numpy()

```

```

[3]: print(mnist_train.shape)
print(mnist_test.shape)

images_train = np.float32(np.reshape(mnist_train[:, 1:], (mnist_train.
    ↪shape[0], 28, 28, 1))) / 255
numbers_train = mnist_train[:, 0]
images_test = np.float32(np.reshape(mnist_test[:, 1:], (mnist_test.
    ↪shape[0], 28, 28, 1))) / 255
numbers_test = mnist_test[:, 0]

images_train = np.transpose(images_train, [0, 3, 1, 2])
images_test = np.transpose(images_test, [0, 3, 1, 2])

images_train = torch.tensor(images_train, device=cuda).
    ↪to(non_blocking=True)
numbers_train = torch.tensor(numbers_train, device=cuda).
    ↪to(non_blocking=True)
images_test = torch.tensor(images_test, device=cuda).to(non_blocking=True)

```

```

numbers_test = torch.tensor(numbers_test, device=cuda).
    ↳to(non_blocking=True)

print(images_train.shape)
print(numbers_train.shape)
print(images_test.shape)
print(numbers_test.shape)

print("Классы: ", np.unique(numbers_train.cpu()))

```

```

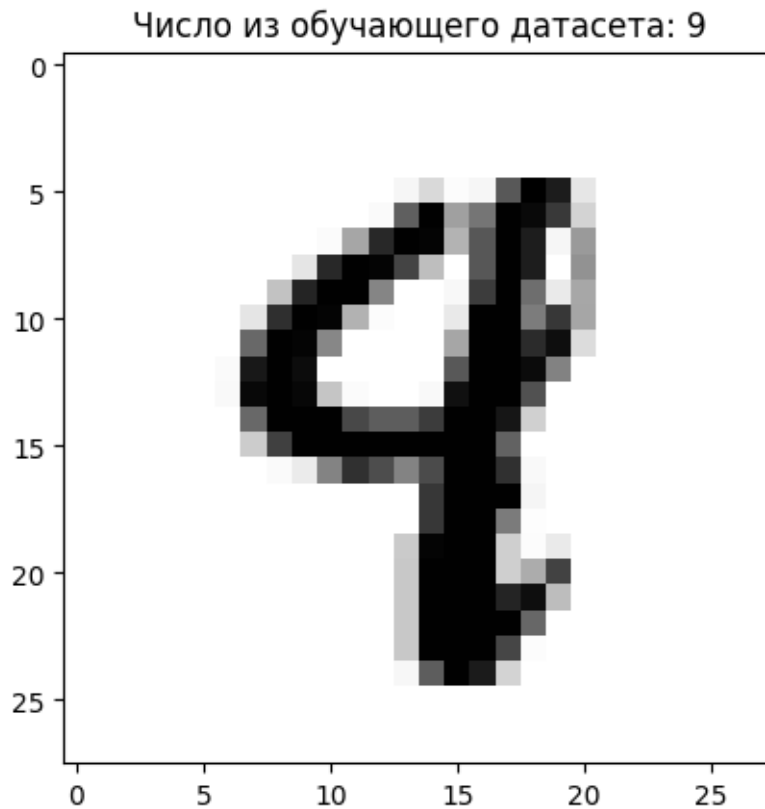
(60000, 785)
(10000, 785)
torch.Size([60000, 1, 28, 28])
torch.Size([60000])
torch.Size([10000, 1, 28, 28])
torch.Size([10000])
Классы:  [0 1 2 3 4 5 6 7 8 9]

```

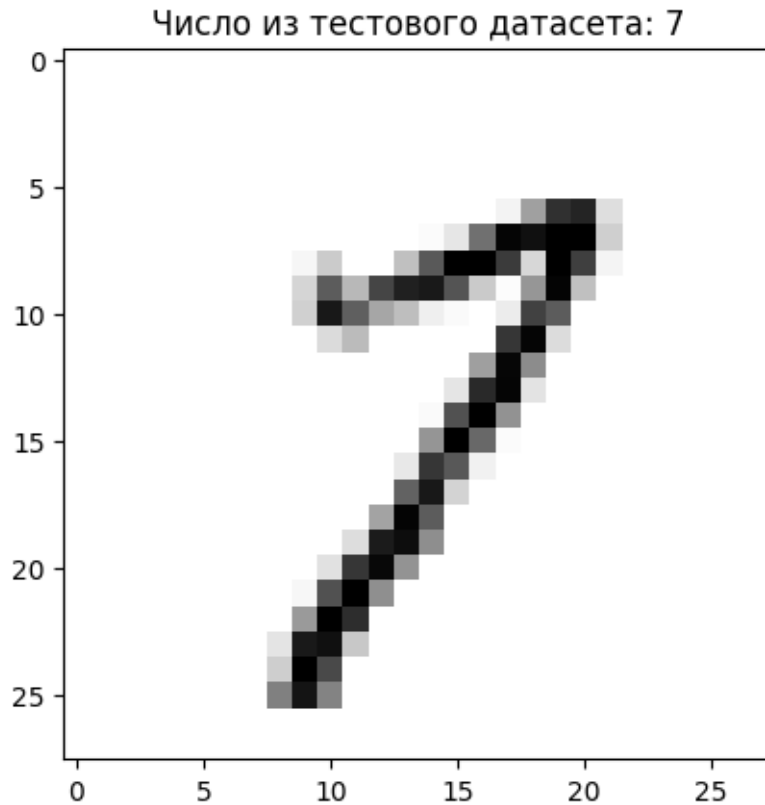
```

[4]: number_train = random.randint(0, images_test.cpu().shape[0])
plt.imshow(images_test.cpu()[number_train, 0, :, :], cmap=plt.cm.binary,
    ↳vmin=0, vmax=1)
plt.title(f"Число из обучающего датасета: {numbers_test.cpu().
    ↳numpy()[number_train]}")
print()

```



```
[5]: number_test = random.randint(0, images_test.cpu().shape[0])
plt.imshow(images_test.cpu()[number_test, 0, :, :], cmap=plt.cm.binary,
           ↪vmin=0, vmax=1)
plt.title(f"Число из тестового датасета: {numbers_test.cpu().
           ↪numpy()[number_test]}")
print()
```



```
[6]: print(images_train.shape)
      print(numbers_train.shape)
```

```
torch.Size([60000, 1, 28, 28])
torch.Size([60000])
```

```
[7]: from torchvision import datasets
      import torchvision.transforms as transforms
      from torch.utils.data.sampler import SubsetRandomSampler
      from torch.utils.data import TensorDataset, DataLoader

      from importlib import reload
      import LeNet

      reload(LeNet)

      # Picking Fashion-MNIST dataset

      train_transforms = transforms.Compose([
          transforms.RandomRotation(25),
```

```

        transforms.RandomInvert(p=0.5),
    ])
    # transforms.RandomInvert(p=0.5),
    normalize_transforms = transforms.Compose([
        transforms.Resize(size=LeNet.IMAGE_SIZE, interpolation=transforms.
        ↳InterpolationMode.BICUBIC),
        transforms.Normalize((0.5,), (0.5,)),
    ])
    # transforms.Normalize((0.5,), (0.5,)),

    train_data = TensorDataset(images_train, numbers_train)
    test_data = TensorDataset(images_test, numbers_test)

    # Finding indices for validation set
    num_train = len(train_data)
    indices = list(range(num_train))
    #Randomize indices
    np.random.shuffle(indices)

    split = int(np.floor(num_train * valid_size))
    train_index, test_index = indices[split:], indices[:split]

    # Making samplers for training and validation batches
    train_sampler = SubsetRandomSampler(train_index)
    valid_sampler = SubsetRandomSampler(test_index)

    # Creating data loaders
    train_loader = torch.utils.data.DataLoader(train_data, ↳
        ↳batch_size=batch_size, sampler=train_sampler,
                                                num_workers=num_workers)
    valid_loader = torch.utils.data.DataLoader(train_data, ↳
        ↳batch_size=batch_size, sampler=valid_sampler,
                                                num_workers=num_workers)
    test_loader = torch.utils.data.DataLoader(test_data, ↳
        ↳batch_size=batch_size, num_workers=num_workers)

    # Image classes
    classes = ['0', '1', '2', '3', '4', '5', '6', '7', '8', '9']

```

```

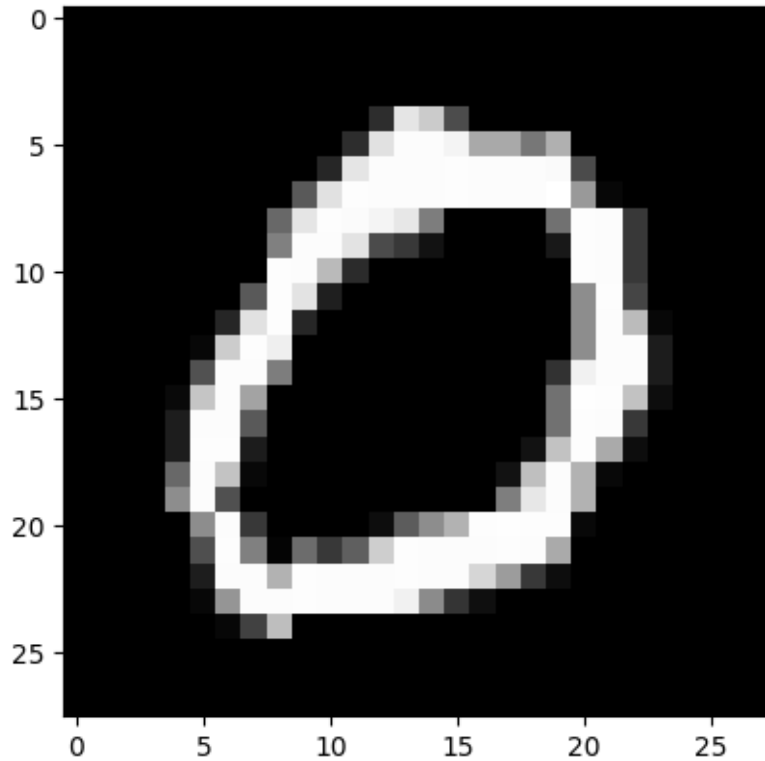
[8]: for batch_index, (data, target) in enumerate(train_loader):
        break
    print(data.shape, target.shape, classes[target[0]])
    image = normalize_transforms(train_transforms(data.cpu()))

```

```
plt.imshow(image[random.randint(0, image.shape[0]), 0, :, :], cmap=plt.cm.
→binary, vmin=-1, vmax=1)
```

```
torch.Size([100, 1, 28, 28]) torch.Size([100]) 7
```

```
[8]: <matplotlib.image.AxesImage at 0x1ea70355cf0>
```



```
[9]: from importlib import reload
import torchvision.models
import LeNet
import torch.optim as optim

reload(LeNet)

model = LeNet.LeNet()
model.cuda()
print(model)

# loss function (cross entropy loss)
criterion = nn.CrossEntropyLoss()
```



```

train_losses = []
valid_losses = []

# tracks validation loss change after each epoch
minimum_validation_loss = np.inf

# optimizer
optimizer = optim.SGD(model.parameters(), lr=learning_rate,
    ↪momentum=learning_momentum)

```

```

LeNet(
  (conv): Sequential(
    (0): Conv2d(1, 6, kernel_size=(5, 5), stride=(1, 1), padding=(2, 2))
    (1): Sigmoid()
    (2): AvgPool2d(kernel_size=(2, 2), stride=2, padding=0)
    (3): Conv2d(6, 16, kernel_size=(5, 5), stride=(1, 1))
    (4): Sigmoid()
    (5): AvgPool2d(kernel_size=(2, 2), stride=2, padding=0)
  )
  (fc): Sequential(
    (0): Linear(in_features=400, out_features=120, bias=True)
    (1): Sigmoid()
    (2): Linear(in_features=120, out_features=84, bias=True)
    (3): Sigmoid()
    (4): Linear(in_features=84, out_features=10, bias=True)
  )
)

```

```

[ ]: # epochs to train for
epochs = 1000

for epoch in range(1, epochs + 1):
    clear_output(wait=True)

    train_loss = 0
    valid_loss = 0

    # training steps
    model.train()

    index = 0
    for batch_index, (data, target) in enumerate(train_loader):
        index += 1
        # clears gradients
        optimizer.zero_grad()

```

```

    # forward pass
    output = model(normalize_transforms(train_transforms(data)))
    # loss in batch
    loss = criterion(output, target)
    # backward pass for loss gradient
    loss.backward()
    # update paremeters
    optimizer.step()
    # update training loss
    train_loss += loss.item() * data.size(0)

# validation steps
model.eval()
for batch_index, (data, target) in enumerate(valid_loader):
    # forward pass
    output = model(normalize_transforms(train_transforms(data)))
    # loss in batch
    loss = criterion(output, target)
    # update validation loss
    valid_loss += loss.item() * data.size(0)

# average loss calculations
train_loss = train_loss / len(train_loader.sampler)
valid_loss = valid_loss / len(valid_loader.sampler)
train_losses.append(train_loss)
valid_losses.append(valid_loss)

# Display loss statistics
print(
    f'Текущая Эпоха: {len(train_losses)}\nTraining Loss: {round(
→{train_loss, 6})}\nValidation Loss: {round(valid_loss, 6)}')

# Saving model every time validation loss decreases
if valid_loss <= minimum_validation_loss:
    print(f'Validation loss уменьшилась с {round(
→{minimum_validation_loss, 6})} до {round(valid_loss, 6)}')
    torch.save(model.state_dict(), f'{MODEL_NAME}.pt')
    minimum_validation_loss = valid_loss
    print('Сохранение новой модели')

plt.plot(train_losses, 'g')
plt.plot(valid_losses, 'r')
plt.ylim([0, max(np.max(np.array(train_losses)), np.max(np.
→array(train_losses)), 0.0) * 1.1])

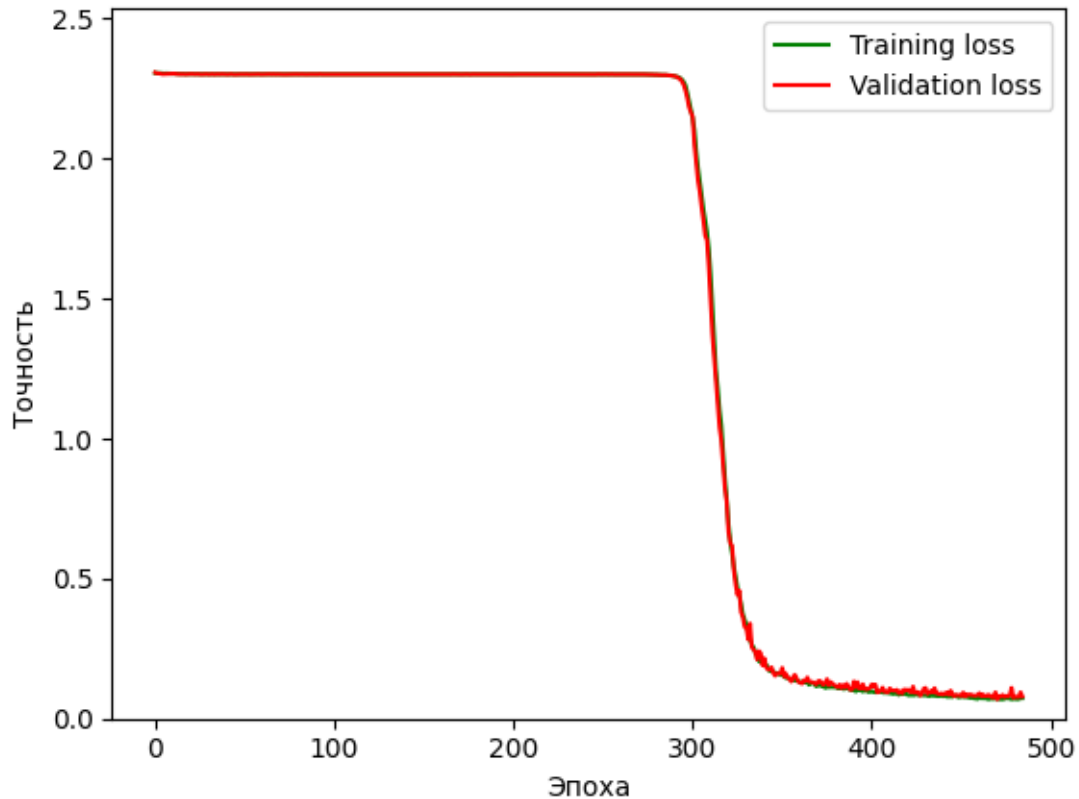
```

```
plt.xlabel("Эпоха")
plt.ylabel("Точность")
plt.legend(["Training loss", "Validation loss"])
plt.show()
```

Текущая Эпоха: 485

Training Loss: 0.073134

Validation Loss: 0.079139



```
[ ]: from importlib import reload
import LeNet

reload(LeNet)

model_new = LeNet.LeNet()
model_new.cuda()
model_new.load_state_dict(torch.load(f'{MODEL_NAME}.pt'))
```

```

[ ]: # tracking test loss
test_loss = 0.0
class_correct = list(0. for i in range(10))
class_total = list(0. for i in range(10))

model_new.eval()

for batch_idx, (data, target) in enumerate(test_loader):
    # forward pass
    output = model_new(normalize_transforms(train_transforms(data)))

    output_max = torch.max(output)
    output_min = torch.min(output)

    # batch loss
    loss = criterion(output, target)
    # test loss update
    test_loss += loss.item() * data.size(0)
    # convert output probabilities to predicted class
    _, pred = torch.max(output, 1)
    # compare predictions to true label
    correct_tensor = pred.eq(target.data.view_as(pred))
    correct = np.squeeze(correct_tensor.numpy()) if not torch.cuda.
    →is_available() else np.squeeze(
        correct_tensor.cpu().numpy())
    # calculate test accuracy for each object class
    for i in range(10):
        label = target.data[i]
        class_correct[label] += correct[i].item()
        class_total[label] += 1

# average test loss
test_loss = test_loss / len(test_loader.dataset)
print(f'Test Loss: {round(test_loss, 6)}')

for i in range(10):
    if class_total[i] > 0:
        print(f'Test Accuracy of {classes[i]}: {round(100 *
    →class_correct[i] / class_total[i], 2)}%')
    else:
        print(f'Test Accuracy of {classes[i]}s: N/A (no training
    →examples)')

print(

```

```

    f'Full Test Accuracy: {round(100. * np.sum(class_correct) / np.
→sum(class_total), 2)}% {np.sum(class_correct)} out of {np.
→sum(class_total)}')

```

```

output_min = output_min.cpu().detach().numpy()
output_max = output_max.cpu().detach().numpy()

```

```
[ ]: prediction_threshold_amount = 1000
```

```
[ ]: prediction_threshold_low = []
```

```

while len(prediction_threshold_low) < prediction_threshold_amount:
    for batch_idx, (data, target) in enumerate(train_loader):
        clear_output(wait=True)
        data = normalize_transforms(train_transforms(data))
        output = model_new(data[:1, :1, :, :]).cpu().detach()
        output = (output.numpy()[0] - output_min) / (output_max -
→output_min)
        if target[0] == np.argmax(output):
            continue
        prediction_threshold_low.append(np.max(output))
        print(f"calibration lower threshold {100 *
→len(prediction_threshold_low) / prediction_threshold_amount:.2f}%")
        print(f"real value = {target[0]}")
        print(f"prediction = {np.argmax(output)}")
        print(f"prediction % = {np.max(output)}")
        print(f"predictions = {output}")
        plt.imshow(data.cpu()[0, 0, :, :], cmap=plt.cm.binary, vmin=-1,
→vmax=1)
        plt.show()
        if len(prediction_threshold_low) >= prediction_threshold_amount:
            break

threshold_low = np.mean(np.array(prediction_threshold_low))
threshold_low_std = np.std(np.array(prediction_threshold_low))

clear_output(wait=True)
print(f"lower threshold = {threshold_low} with std = {threshold_low_std}")

```

```
[ ]: prediction_threshold_high = []
```

```

while len(prediction_threshold_high) < prediction_threshold_amount:
    for batch_idx, (data, target) in enumerate(train_loader):
        clear_output(wait=True)

```

```

        data = normalize_transforms(train_transforms(data))
        output = model_new(data[:1, :1, :, :]).cpu().detach()
        output = (output.numpy()[0] - output_min) / (output_max -
→output_min)
        prediction_threshold_high.append(np.max(output))
        print(f"calibration upper threshold {100 *
→len(prediction_threshold_high) / prediction_threshold_amount:.2f}%")
        print(f"real value = {target[0]}")
        print(f"prediction = {np.argmax(output)}")
        print(f"prediction % = {np.max(output)}")
        print(f"predictions = {output}")
        plt.imshow(data.cpu()[0, 0, :, :], cmap=plt.cm.binary, vmin=-1,
→vmax=1)
        plt.show()
        if len(prediction_threshold_high) >= prediction_threshold_amount:
            break

threshold_high = np.mean(np.array(prediction_threshold_high))
threshold_high_std = np.std(np.array(prediction_threshold_high))

clear_output(wait=True)
print(f"upper threshold = {threshold_high} with std =
→{threshold_high_std}")

```

```

[ ]: confidence_threshold = (threshold_high - threshold_high_std) / 2 +
→(threshold_low + threshold_low_std) / 2

plt.plot(prediction_threshold_high, "b")
plt.plot(prediction_threshold_low, "g")
plt.hlines([confidence_threshold, ], xmin=0,
→xmax=len(prediction_threshold_high), colors=["r", ])
plt.legend(["Верхний порог уверенности", "Нижний порог уверенности",
→"Линия раздела уверенности"])

print(f"final confidence threshold = {confidence_threshold}")
print(f"from {threshold_low} => to {threshold_high}")
print(f"window = {threshold_high - threshold_low}")

```