

# Лабораторная работа 5

## Вариант 13

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Целью работы является исследование свойств сетей Хопфилда, Хэмминга и Элмана, алгоритмов обучения, а также применение сетей в задачах распознавания статических и динамических образов.

```
In [1]: import matplotlib.pyplot as plt
import numpy as np
import torch
from torch import nn
from torch.utils.data import DataLoader
import tqdm
```

```
In [2]: def f1(k: float):
        return np.sin(4*np.pi*k)

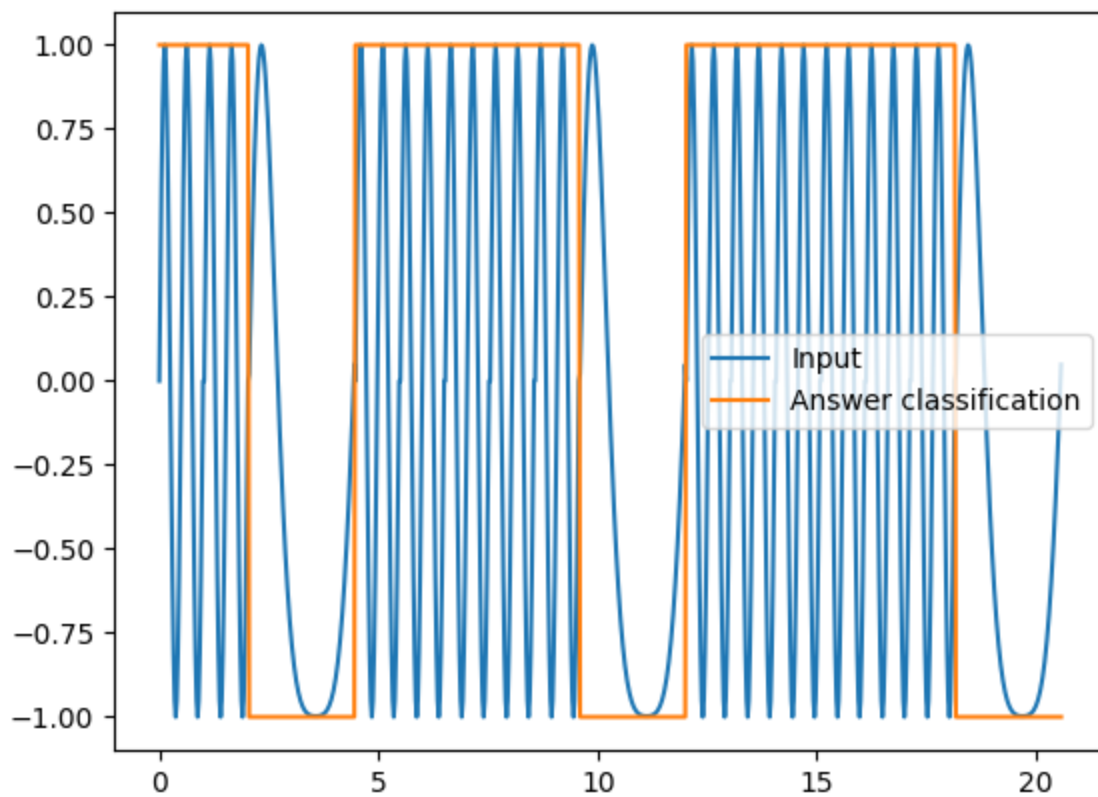
def f2(k: float):
    return np.sin(2*k**2 - 6*k + 3)
```

Классификация входной функции

```
In [3]: h = 0.025
k1 = np.arange(0, 1+h, h)
k2 = np.arange(-0.02, 2.36+h, h)
p1 = f1(k1)
p2 = f2(k2)
r = [2, 5, 6]
p = np.concatenate((np.tile(p1, r[0]),
                    p2,
                    np.tile(p1, r[1]),
                    p2,
                    np.tile(p1, r[2]),
                    p2))
t = np.concatenate((np.ones(len(p1)*r[0]),
                    -1*np.ones(len(p2)),
                    np.ones(len(p1)*r[1]),
                    -1*np.ones(len(p2)),
                    np.ones(len(p1)*r[2]),
                    -1*np.ones(len(p2))))
x = np.arange(len(p))*h

In [4]: plt.plot(x, p, label='Input')
plt.plot(x, t, label='Answer classification')
plt.legend()
```

```
Out[4]: <matplotlib.legend.Legend at 0x7f23743f9e80>
```



Создаем датасет

```
In [5]: w = 5
X = [p[i:i+w].astype('float32') for i in range(0, len(p) - w)]
y = [t[i:i+w].astype('float32') for i in range(0, len(p) - w)]
data = [(x,y) for x, y in zip(X,y)]
train_dataloader = DataLoader(data, batch_size=1, shuffle=False)
```

Класс слоя Элмана

```
In [6]: class ElmanLayer(nn.Module):
    def __init__(self, size_in, size_out):
        super().__init__()
        w1 = torch.randn(size_in, size_out)
        w2 = torch.randn(size_out, size_out)
        b = torch.randn(size_out)
        self.w1 = nn.Parameter(w1)
        self.w2 = nn.Parameter(w2)
        self.b = nn.Parameter(b)

    def forward(self, x):
        out = torch.matmul(x, self.w1)
        out = torch.add(out, self.b)
        if hasattr(self, "prev"):
            d = torch.matmul(self.prev, self.w2)
            out = torch.add(out, d)
        out = torch.tanh(out)
        self.prev = torch.tensor(out)
        return out

    def del_prev(self):
        if hasattr(self, "prev"):
            delattr(self, "prev")
```

Модель

```
In [7]: vectoriz = 8
        elman = ElmanLayer(w,vectoriz)
        linear = nn.Linear(vectoriz, w)
        model = nn.Sequential(elman,linear)
```

## Тренеруем модель

```
In [8]: optimizer = torch.optim.Adam(model.parameters(), lr=1e-3)
        loss_fn = nn.MSELoss()

        epoch = 300
        model.train()

        losses = []
        for ep in tqdm.tqdm(range(epoch)):
            model[0].del_prev()
            epoch_loss = []
            for (inp, out) in train_dataloader:
                pred = model(inp)

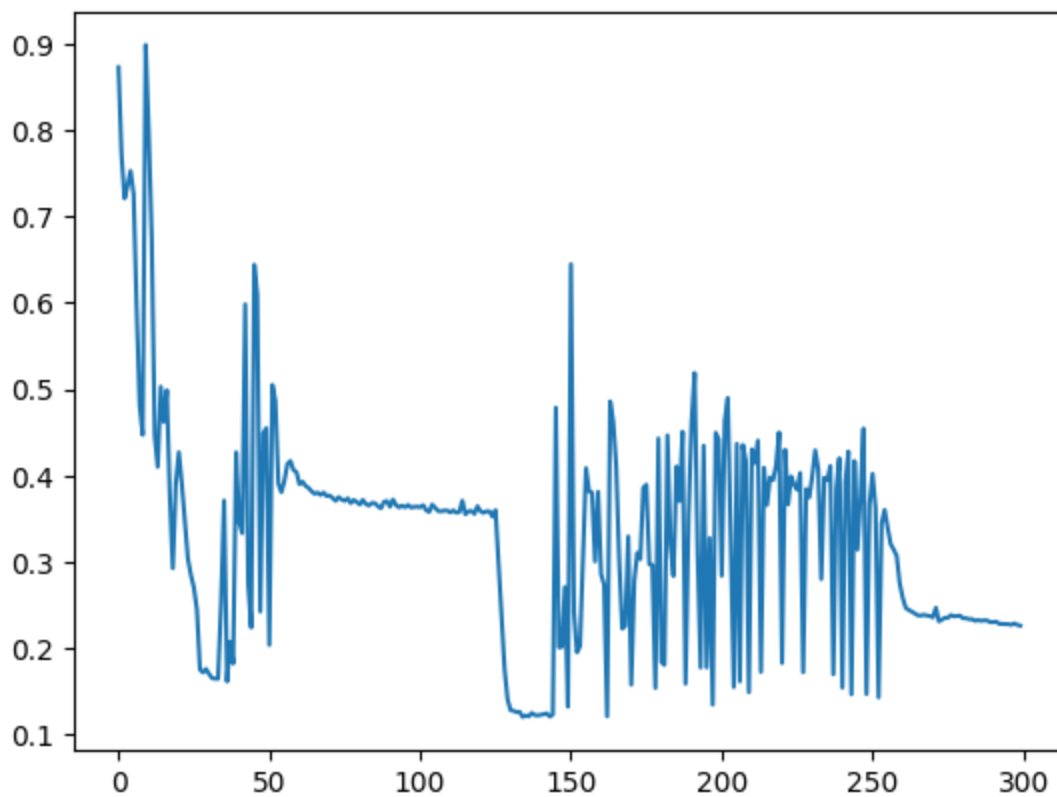
                loss = loss_fn(pred, out)
                loss = torch.sqrt_(loss)
                epoch_loss.append(loss.item())
                optimizer.zero_grad()
                loss.backward()
                optimizer.step()
            losses.append(np.mean(epoch_loss))
```

```
0%|          | 0/300 [00:00<?, ?it/s]/tmp/ipy
kernel_4843/370478990.py:18: UserWarning: To copy construct from a tensor, it is recomme
nded to use sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad
_(True), rather than torch.tensor(sourceTensor).
  self.prev = torch.tensor(out)
100%|██████████| 300/300 [02:40<00:00, 1.86it/s]
```

## График ошибки

```
In [9]: plt.plot(np.arange(epoch), losses)
```

```
Out[9]: [<matplotlib.lines.Line2D at 0x7f23712807c0>]
```



Делаем предсказание

```
In [10]: model.eval()
model[0].del_prev()
pred = []
for (inp, out) in tqdm.tqdm(train_dataloader):
    pred.append(model(inp).detach().numpy())
```

```
0%|          | 0/819 [00:00<?, ?it/s]/tmp/ipy
kernel_4843/370478990.py:18: UserWarning: To copy construct from a tensor, it is recomme
nded to use sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad
_(True), rather than torch.tensor(sourceTensor).
    self.prev = torch.tensor(out)
100%|██████████| 819/819 [00:00<00:00, 5324.74it/s]
```

Обрабатываем предсказание

```
In [11]: sum_ped = [0]*len(p)
for i, window in enumerate(pred):
    for j, item in enumerate(window[0]):
        sum_ped[i+j] = item

ans = [1 if elem > 0 else -1 for elem in sum_ped]
```

```
In [12]: plt.plot(x, p, label='Input')
plt.plot(x, t, label='Answer classification')
plt.plot(x, ans, label='Predict classification')
plt.legend()
```

```
Out[12]: <matplotlib.legend.Legend at 0x7f237121fdf0>
```

