

Лабораторная работа 8

Вариант 13

Аминов С.С. М8О-408Б-19

Целью работы является исследование свойств некоторых динамических нейронных сетей, алгоритмов обучения, а также применение сетей в задаче распознавания динамических образов.

```
In [1]: import numpy as np

import torch
import torch.nn as nn
import torch.optim as optim

import matplotlib.pyplot as plt
from collections import deque
import tqdm
```

Зададим управляющий сигнал

```
In [14]: def u(k):
        return np.sin(k**2 - 2*k + np.pi)/7
```

```
In [15]: N = 500
t = np.linspace(0, 5, N)

x = u(t)
y = [0]
for i in range(len(t) - 1):
    y.append(y[-1] / (1 + y[-1]**2) + x[i])

y = np.array(y)
assert x.shape == y.shape
```

Сгенерируем датасет

```
In [16]: def gen_dataset(x, y, delay=5):
        return [(
            np.array(x[i:i+delay], dtype=np.float32),
            np.array(y[i+delay], dtype=np.float32)
        ) for i in range(len(x) - delay)]
```

```
In [17]: train_data = gen_dataset(x, y)
```

```
In [18]: data_loader = torch.utils.data.DataLoader(train_data, batch_size=1, shuffle=False)
```

Реализуем вспомогательный слой TDL и сеть NARX.

```
In [19]: class TDL(nn.Module):
        def __init__(self, in_features, delay=1):
            super(TDL, self).__init__()
            self.in_features = in_features
            self.delay = delay
            self.line = deque()
            self.clear()
```

```

def clear(self):
    self.line.clear()
    for i in range(self.delay):
        self.line.append(torch.zeros(self.in_features))

def push(self, input):
    self.line.appendleft(input)

def forward(self, input=None):
    return self.line.pop()

```

```

In [20]: class NARX(nn.Module):
    def __init__(self, in_features, hidden_features, out_features, delay1, delay2):
        super(NARX, self).__init__()

        self.in_features = in_features
        self.hidden_features = hidden_features
        self.out_features = out_features

        self.line1 = TDL(in_features, delay1)
        self.line2 = TDL(out_features, delay2)

        self.w1 = torch.nn.Parameter(torch.randn(in_features, hidden_features))
        self.w2 = torch.nn.Parameter(torch.randn(hidden_features, out_features))
        self.w3 = torch.nn.Parameter(torch.randn(out_features, hidden_features))

        self.b1 = torch.nn.Parameter(torch.randn(hidden_features))
        self.b2 = torch.nn.Parameter(torch.randn(out_features))

    def clear(self):
        self.line1.clear()
        self.line2.clear()

    def forward(self, input):
        res = torch.tanh(
            self.line1() @ self.w1 + self.line2() @ self.w3 + self.b1
        ) @ self.w2 + self.b2
        self.line1.push(input.clone().detach()) # сохранять будем копии
        self.line2.push(res.clone().detach())
        return res

```

```

In [21]: model = NARX(5, 10, 1, 3, 3)
optimizer = torch.optim.Adam(model.parameters(), lr=1e-3)
criterion = nn.MSELoss()
epochs = 40

```

```

In [22]: loss = []
model.train()
for epoch in tqdm.tqdm(range(epochs)):
    epoch_loss = []
    for X_batch, y_batch in data_loader:
        y_pred = model(X_batch)
        cur_loss = criterion(y_batch, y_pred)
        epoch_loss.append(cur_loss.item())
        cur_loss.backward()

    optimizer.step()
    optimizer.zero_grad()
    loss += [np.mean(epoch_loss)]

```


