

Лабораторная работа № 8 по Нейроинформатике

Динамические сети

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Вариант № 16

In [188]:

```
import numpy as np
import torch
import torch.nn as nn
from torch import optim
import matplotlib.pyplot as plt
from timeit import default_timer as timer
from tqdm import tqdm
from collections import deque
```

Создадим класс слоя с временной задержкой

In [189]:

```
class TDL(nn.Module):
    def __init__(self, in_features, delays=1):
        super(TDL, self).__init__()
        self.in_features = in_features
        self.delays = delays
        self.deque = deque()
        self.clear()
    def clear(self):
        self.deque.clear()
        for i in range(self.delays):
            self.deque.append(torch.zeros(self.in_features))
    def push(self, inputs):
        self.deque.appendleft(inputs)
    def forward(self, inputs=0):
        return self.deque.pop()
```

Создадим класс для **NARX** (нелинейная авторегрессионная сеть с внешними входами)

In [190]:

```
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.deque1 = TDL(in_features, delay1)
        self.deque2 = 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.ones(hidden_features))
        self.b2 = torch.nn.Parameter(torch.ones(out_features))
    def forward(self, inputs):
        out1 = torch.tanh(self.deque1() @ self.w1 + self.deque2() @ self.w3 + self.b1)
        out2 = out1 @ self.w2 + self.b2
        self.deque1.push(torch.tensor(inputs))
        self.deque2.push(torch.tensor(out2))
        return out2
    def clear(self):
        self.deque1.clear()
        self.deque2.clear()
```

Параметры модели

In [191]:

```
epochs = 200
n = 600
w = 5
```

Зададим управляющий и идеальный сигналы

In [192]:

```
t = np.linspace(0, 5, n)
ut = np.cos(t ** 2 - 15 * t + 3) - np.cos(t)
```

In [193]:

```
yt = [0]
for i in range(n - 1):
    yt += [yt[-1] / (1 + yt[-1] ** 2) + ut[i] ** 3]
```

In [194]:

```
train_data = [(np.array(ut[i:i + w], dtype=np.float32),
                  np.array(yt[i:i + w], dtype=np.float32)) for i in range(n - w)]
train_loader = torch.utils.data.DataLoader(dataset=train_data, batch_size=1, shuffle=False)
```

Создаём модель

In [195]:

```
model = NARX(5, 10, 5, 3, 3)
optimizer = optim.Adam(model.parameters(), lr=1e-3)
```

Обучаем

In [196]:

```
model.train()
```

Out[196]:

```
NARX(
  (deque1): TDL()
  (deque2): TDL()
)
```

In [197]:

```
train_loss = []
start_time = timer()
for i in range(epochs):
    pbar = tqdm(enumerate(train_loader))
    #model.clear()
    loss_per_epoch = []
    for _, (inputs, given_outputs) in pbar:
        outputs = model(inputs)
        crit = nn.MSELoss()
        loss = crit(given_outputs, outputs)
        loss_per_epoch += [loss.item()]
        optimizer.zero_grad()
        loss.backward()
        optimizer.step()
    train_loss += [np.mean(loss_per_epoch)]
    pbar.write('%d. loss = %f ' % (i + 1, train_loss[-1]))
end_time = timer()
```

Oit [00:00, ?it/s]C:\Users\volch\AppData\Local\Temp\ipykernel_5012\2841970211.py:17: User

Warning: To copy construct from a tensor, it is recommended to use sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad_(True), rather than torch.tensor(sourceTensor).

```
self.deque1.push(torch.tensor(inputs))
```

C:\Users\volch\AppData\Local\Temp\ipykernel_5012\2841970211.py:18: UserWarning: To copy construct from a tensor, it is recommended to use sourceTensor.clone().detach() or sourceTensor.clone().detach().requires_grad_(True), rather than torch.tensor(sourceTensor).

```
self.deque2.push(torch.tensor(out2))
```

595it [00:00, 1375.54it/s]

1. loss = 10.126014

595it [00:00, 1431.42it/s]

2. loss = 6.405377

595it [00:00, 1426.75it/s]

3. loss = 4.247128

595it [00:00, 1433.32it/s]

4. loss = 3.154549

595it [00:00, 1427.56it/s]

5. loss = 2.760195

595it [00:00, 1431.74it/s]

6. loss = 2.479308

595it [00:00, 1437.42it/s]

7. loss = 2.161671

595it [00:00, 1439.00it/s]

8. loss = 2.174271

595it [00:00, 1442.35it/s]

9. loss = 1.786616

595it [00:00, 1432.92it/s]

10. loss = 1.668497

595it [00:00, 1421.34it/s]

11. loss = 1.709817

595it [00:00, 1396.91it/s]

12. loss = 1.535436

595it [00:00, 1437.26it/s]

13. loss = 1.630287

595it [00:00, 1421.65it/s]

14. loss = 1.743165

595it [00:00, 1427.22it/s]

15. loss = 1.729573

595it [00:00, 1432.27it/s]

16. loss = 1.600381

595it [00:00, 1425.43it/s]

17. loss = 1.774336

595it [00:00, 1446.71it/s]

18. loss = 1.370347

595it [00:00, 1432.95it/s]

19. loss = 1.279798

595it [00:00, 1432.10it/s]

20. loss = 1.292278

595it [00:00, 1439.48it/s]

21. loss = 1.182380

595it [00:00, 1430.45it/s]

22. loss = 1.108096

595it [00:00, 1429.82it/s]

23. loss = 1.118440

595it [00:00, 1453.52it/s]

24. loss = 0.936224

595it [00:00, 1433.87it/s]

25. loss = 1.023491

595it [00:00, 1440.76it/s]

26. loss = 1.047860

595it [00:00, 1426.42it/s]

27. loss = 0.940208

595it [00:00, 1441.05it/s]

28. loss = 0.977149

595it [00:00, 1433.11it/s]

29. loss = 0.986639

595it [00:00, 1444.73it/s]

30. loss = 1.053385

595it [00:00, 1440.12it/s]

31. loss = 1.051814

595it [00:00, 1442.83it/s]

32. loss = 1.030685

595it [00:00, 1452.76it/s]

33. loss = 1.013026

595it [00:00, 1442.76it/s]

34. loss = 0.940240

595it [00:00, 1445.78it/s]

35. loss = 0.882799

595it [00:00, 1426.08it/s]

36. loss = 0.899881

595it [00:00, 1444.94it/s]

37. loss = 0.832506

595it [00:00, 1435.48it/s]

38. loss = 0.818019

595it [00:00, 1444.26it/s]

39. loss = 0.792344

595it [00:00, 1462.64it/s]

40. loss = 0.742686

595it [00:00, 1433.33it/s]

41. loss = 0.753732

595it [00:00, 1445.21it/s]

42. loss = 0.710376

595it [00:00, 1449.80it/s]

43. loss = 0.693317

595it [00:00, 1436.41it/s]

44. loss = 0.678962

595it [00:00, 1448.08it/s]

45. loss = 0.717800

595it [00:00, 1445.45it/s]

46. loss = 0.601143

595it [00:00, 1453.57it/s]

47. loss = 0.620551

595it [00:00, 1456.32it/s]

48. loss = 0.643751

595it [00:00, 1435.48it/s]

49. loss = 0.598219

595it [00:00, 1446.93it/s]

50. loss = 0.587942

595it [00:00, 1438.79it/s]

51. loss = 0.578667

595it [00:00, 1447.45it/s]

52. loss = 0.578387

595it [00:00, 1446.74it/s]

53. loss = 0.550301

595it [00:00, 1455.22it/s]

54. loss = 0.565554

595it [00:00, 1444.62it/s]

55. loss = 0.543771

595it [00:00, 1444.37it/s]

56. loss = 0.523524

595it [00:00, 1445.11it/s]

57. loss = 0.518379

595it [00:00, 1443.76it/s]

58. loss = 0.513059

595it [00:00, 1421.64it/s]

59. loss = 0.502928

595it [00:00, 1418.87it/s]

60. loss = 0.491748

595it [00:00, 1439.60it/s]

61. loss = 0.467623

595it [00:00, 1448.35it/s]

62. loss = 0.468435

595it [00:00, 1445.42it/s]

63. loss = 0.464246

595it [00:00, 1454.25it/s]

64. loss = 0.463490

595it [00:00, 1430.42it/s]

65. loss = 0.457314

595it [00:00, 1443.28it/s]

66. loss = 0.452185

595it [00:00, 1447.23it/s]

67. loss = 0.443755

595it [00:00, 1445.28it/s]

68. loss = 0.435647

595it [00:00, 1443.48it/s]

69. loss = 0.428068

595it [00:00, 1439.03it/s]

70. loss = 0.420788

595it [00:00, 1445.48it/s]

71. loss = 0.413721

595it [00:00, 1460.20it/s]

72. loss = 0.406513

595it [00:00, 1442.77it/s]

73. loss = 0.399040

595it [00:00, 1444.97it/s]

74. loss = 0.391951

595it [00:00, 1456.72it/s]

75. loss = 0.385554

595it [00:00, 1435.00it/s]

76. loss = 0.379742

595it [00:00, 1443.46it/s]

77. loss = 0.374996

595it [00:00, 1442.53it/s]

78. loss = 0.370609

595it [00:00, 1447.55it/s]

79. loss = 0.365966

595it [00:00, 1453.33it/s]

80. loss = 0.360876

595it [00:00, 1452.74it/s]

81. loss = 0.355497

595it [00:00, 1449.91it/s]

82. loss = 0.350227

595it [00:00, 1443.28it/s]

83. loss = 0.345443

595it [00:00, 1444.66it/s]

84. loss = 0.341321

595it [00:00, 1446.64it/s]

85. loss = 0.337837

595it [00:00, 1438.12it/s]

86. loss = 0.334837

595it [00:00, 1449.28it/s]

87. loss = 0.332063

595it [00:00, 1439.30it/s]

88. loss = 0.329220

595it [00:00, 1436.29it/s]

89. loss = 0.326131

595it [00:00, 1444.49it/s]

90. loss = 0.322838

595it [00:00, 1434.00it/s]

91. loss = 0.319631

595it [00:00, 1442.76it/s]

92. loss = 0.316743

595it [00:00, 1460.20it/s]

93. loss = 0.314088

595it [00:00, 1436.77it/s]

94. loss = 0.311299

595it [00:00, 1445.40it/s]

95. loss = 0.308181

595it [00:00, 1450.96it/s]

96. loss = 0.305032

595it [00:00, 1435.49it/s]

97. loss = 0.302472

595it [00:00, 1443.16it/s]

98. loss = 0.300764

595it [00:00, 1435.95it/s]

99. loss = 0.299122

595it [00:00, 1452.89it/s]

100. loss = 0.296436

595it [00:00, 1444.63it/s]

101. loss = 0.292136

595it [00:00, 1457.32it/s]

102. loss = 0.286436

595it [00:00, 1429.67it/s]

103. loss = 0.280150

595it [00:00, 1466.51it/s]

104. loss = 0.274600

595it [00:00, 1440.94it/s]

105. loss = 0.271094

595it [00:00, 1449.17it/s]

106. loss = 0.270220

595it [00:00, 1418.54it/s]

107. loss = 0.270262

595it [00:00, 1421.41it/s]

108. loss = 0.272679

595it [00:00, 1448.58it/s]

109. loss = 0.288829

595it [00:00, 1443.60it/s]

110. loss = 0.329095

595it [00:00, 1458.83it/s]

111. loss = 0.344066

595it [00:00, 1443.43it/s]

112. loss = 0.314018

595it [00:00, 1445.94it/s]

113. loss = 0.276153

595it [00:00, 1450.25it/s]

114. loss = 0.266861

595it [00:00, 1429.88it/s]

115. loss = 0.264573

595it [00:00, 1444.16it/s]

116. loss = 0.268092

595it [00:00, 1444.08it/s]

117. loss = 0.270422

595it [00:00, 1452.56it/s]

118. loss = 0.273680

595it [00:00, 1459.77it/s]

119. loss = 0.274282

595it [00:00, 1460.22it/s]

120. loss = 0.274388

595it [00:00, 1442.16it/s]

121. loss = 0.268124

595it [00:00, 1422.87it/s]

122. loss = 0.269410

595it [00:00, 1447.09it/s]

123. loss = 0.266370

595it [00:00, 1431.41it/s]

124. loss = 0.271464

595it [00:00, 1447.04it/s]

125. loss = 0.258507

595it [00:00, 1440.38it/s]

126. loss = 0.268037

595it [00:00, 1453.06it/s]

127. loss = 0.246643

595it [00:00, 1439.86it/s]

128. loss = 0.257732

595it [00:00, 1455.71it/s]

129. loss = 0.232777

595it [00:00, 1440.06it/s]

130. loss = 0.243915

595it [00:00, 1452.50it/s]

131. loss = 0.217265

595it [00:00, 1442.16it/s]

132. loss = 0.235778

595it [00:00, 1444.10it/s]

133. loss = 0.209310

595it [00:00, 1442.49it/s]

134. loss = 0.243735

595it [00:00, 1439.87it/s]

135. loss = 0.207337

595it [00:00, 1445.69it/s]

136. loss = 0.238532

595it [00:00, 1447.73it/s]

137. loss = 0.209820

595it [00:00, 1451.35it/s]

138. loss = 0.237038

595it [00:00, 1448.28it/s]

139. loss = 0.205297

595it [00:00, 1430.05it/s]

140. loss = 0.228404

595it [00:00, 1441.02it/s]

141. loss = 0.211902

595it [00:00, 1438.17it/s]

142. loss = 0.228456

595it [00:00, 1449.70it/s]

143. loss = 0.204086

595it [00:00, 1452.28it/s]

144. loss = 0.215419

595it [00:00, 1442.89it/s]

145. loss = 0.219498

595it [00:00, 1447.61it/s]

146. loss = 0.214190

595it [00:00, 1443.13it/s]

147. loss = 0.206538

595it [00:00, 1447.95it/s]

148. loss = 0.208043

595it [00:00, 1444.48it/s]

149. loss = 0.207606

595it [00:00, 1431.88it/s]

150. loss = 0.208671

595it [00:00, 1455.74it/s]

151. loss = 0.205331

595it [00:00, 1449.14it/s]

152. loss = 0.206893

595it [00:00, 1430.21it/s]

153. loss = 0.202990

595it [00:00, 1433.32it/s]

154. loss = 0.207107

595it [00:00, 1436.53it/s]

155. loss = 0.196437

595it [00:00, 1422.53it/s]

156. loss = 0.211698

595it [00:00, 1421.73it/s]

157. loss = 0.191082

595it [00:00, 1436.48it/s]

158. loss = 0.214614

595it [00:00, 1447.19it/s]

159. loss = 0.193516

595it [00:00, 1436.62it/s]

160. loss = 0.218641

595it [00:00, 1445.67it/s]

161. loss = 0.191112

595it [00:00, 1447.23it/s]

162. loss = 0.211363

595it [00:00, 1442.54it/s]

163. loss = 0.203206

595it [00:00, 1452.69it/s]

164. loss = 0.197787

595it [00:00, 1437.03it/s]

165. loss = 0.198272

595it [00:00, 1448.67it/s]

166. loss = 0.192917

595it [00:00, 1445.48it/s]

167. loss = 0.205881

595it [00:00, 1450.68it/s]

168. loss = 0.187253

595it [00:00, 1437.48it/s]

169. loss = 0.213052

595it [00:00, 1448.56it/s]

170. loss = 0.188132

595it [00:00, 1444.57it/s]

171. loss = 0.213710

595it [00:00, 1450.59it/s]

172. loss = 0.188331

595it [00:00, 1454.68it/s]

173. loss = 0.209259

595it [00:00, 1458.37it/s]

174. loss = 0.188208

595it [00:00, 1441.17it/s]

175. loss = 0.211784

595it [00:00, 1448.82it/s]

176. loss = 0.188130

595it [00:00, 1447.74it/s]

177. loss = 0.206827

595it [00:00, 1440.70it/s]

178. loss = 0.188124

595it [00:00, 1446.90it/s]

179. loss = 0.210105

595it [00:00, 1444.24it/s]

180. loss = 0.187953

595it [00:00, 1444.25it/s]

181. loss = 0.206255

595it [00:00, 1441.75it/s]

182. loss = 0.187681

595it [00:00, 1425.55it/s]

183. loss = 0.207856

595it [00:00, 1434.13it/s]

184. loss = 0.187910

595it [00:00, 1439.17it/s]

185. loss = 0.204547

595it [00:00, 1462.28it/s]

186. loss = 0.187847

595it [00:00, 1444.49it/s]

187. loss = 0.205093

595it [00:00, 1446.26it/s]

188. loss = 0.187989

595it [00:00, 1441.97it/s]

189. loss = 0.202763

595it [00:00, 1434.11it/s]

190. loss = 0.188210

595it [00:00, 1444.51it/s]

191. loss = 0.201781

595it [00:00, 1455.26it/s]

192. loss = 0.188466

595it [00:00, 1446.32it/s]

193. loss = 0.199997

595it [00:00, 1439.86it/s]

194. loss = 0.188809

595it [00:00, 1456.55it/s]

195. loss = 0.197154

595it [00:00, 1453.03it/s]

196. loss = 0.189057

595it [00:00, 1435.48it/s]

197. loss = 0.193807

595it [00:00, 1451.50it/s]

198. loss = 0.189796

```
595it [00:00, 1451.71it/s]
```

```
199. loss = 0.192564
```

```
595it [00:00, 1391.33it/s]
```

```
200. loss = 0.192022
```

```
In [198]:
```

```
model.eval()
```

```
Out[198]:
```

```
NARX(  
    (deque1): TDL()  
    (deque2): TDL()  
)
```

```
In [199]:
```

```
#model.clear()
```

```
In [200]:
```

```
predicted = []  
for x, _ in train_data:  
    predicted += [model(x).detach().numpy().item(-1)]
```

```
C:\Users\volch\AppData\Local\Temp\ipykernel_5012\2841970211.py:18: UserWarning: To copy c  
onstruct from a tensor, it is recommended to use sourceTensor.clone().detach() or sourceT  
ensor.clone().detach().requires_grad_(True), rather than torch.tensor(sourceTensor).  
    self.deque2.push(torch.tensor(out2))
```

Выводим результаты

```
In [201]:
```

```
print('Время обучения = {0} секунд'.format(int(end_time - start_time)))  
print('Количество эпох = {0}'.format(epochs))
```

```
Время обучения = 82 секунд  
Количество эпох = 200
```

```
In [202]:
```

```
fig, axes = plt.subplots(1, 2, figsize=(10, 6.5))  
fig.tight_layout(h_pad = 4, w_pad = 4)  
  
axes[0].set_title('Функция потерь')  
axes[0].set_xlabel('Эпоха')  
axes[0].set_ylabel('MSE')  
axes[0].plot(train_loss)  
  
axes[1].set_title('Результат работы сети')  
axes[1].set_xlabel('t')  
axes[1].set_ylabel('y')  
axes[1].plot(ut, label = 'Управляющий сигнал')  
axes[1].plot(yt, label = 'Идеальный сигнал')  
axes[1].plot(predicted, label = 'Предсказанный сигнал')  
axes[1].legend()
```

```
Out[202]:
```

```
<matplotlib.legend.Legend at 0x1c4d52ae7d0>
```



