

Лабораторная работа №1

Перцептроны. Процедура обучения Розенблатта

Целью работы является исследование свойств перцептрона Розенблатта и его применение для решения задачи распознавания образов.

```
[1]: import os
      os.environ["TF_CPP_MIN_LOG_LEVEL"] = "3"

      import numpy as np
      import tensorflow as tf
      import matplotlib.pyplot as plt
```

Вспомогательные функции

```
[2]: def make_xy(x, y):
      n = len(x)
      res = []
      for i in range(n):
          elem = (x[i], y[i])
          res.append(elem)
      return res
```

```
[3]: colour = ["-b", "-c"]

def draw_data(xy, labels, ncls, w = None, b = None, hist = None):
    n = len(xy)
    min_x = xy[0][0]
    max_x = xy[0][0]
    mpx, mpy = dict(), dict()
    for elem in labels:
        mpx[elem] = []
        mpy[elem] = []
    for i in range(n):
        [x0, y0] = xy[i]
        min_x = min(x0, min_x)
        max_x = max(x0, max_x)
        mpx[labels[i]].append(x0)
        mpy[labels[i]].append(y0)
    for i in mpx.keys():
        plt.scatter(mpx[i], mpy[i])
    if (w is not None):
        for i in range(w.shape[1]):
            line_x = np.linspace(min_x, max_x, 100)
            line_y = -(w[0][i] * line_x + b[i]) / w[1][i]
            plt.plot(line_x, line_y, colour[i])
    plt.show()
    if (hist is not None):
        plt.plot(hist)
```

```
plt.ylabel("mae")
plt.xlabel("Эпохи")
plt.show()
```

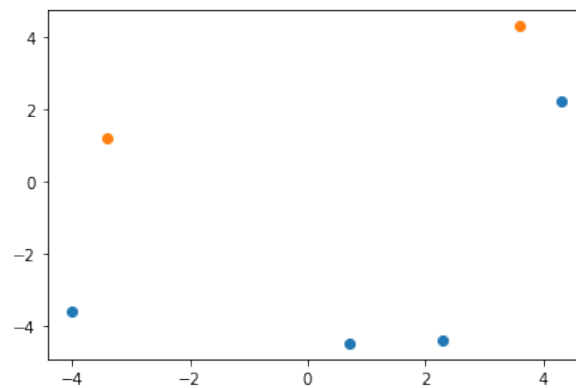
Вариант 4

Задание 1

```
[4]: x = [-4, -3.4, 0.7, 4.3, 2.3, 3.6]
      y = [-3.6, 1.2, -4.5, 2.2, -4.4, 4.3]
      labels = [0, 1, 0, 0, 0, 1]
```

Визуализация данных

```
[5]: xy = make_xy(x, y)
      draw_data(xy, labels, 2)
```



Обучение модели

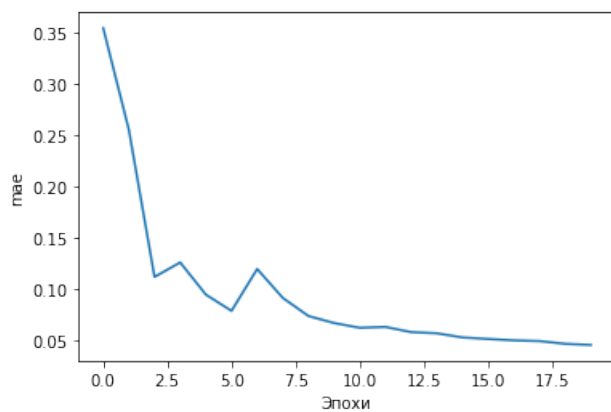
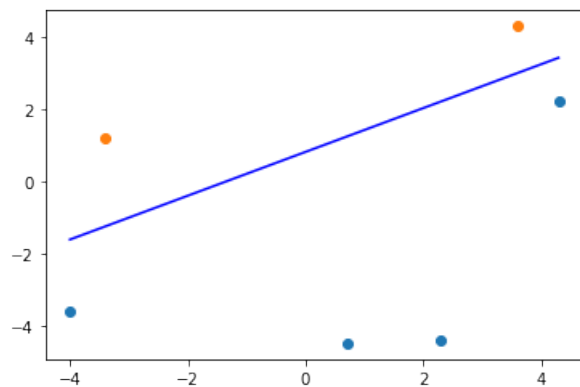
```
[6]: model = tf.keras.models.Sequential([
      tf.keras.layers.Dense(1, input_dim = 2, activation = "sigmoid")
    ])

    model.compile(
      optimizer = tf.keras.optimizers.Adam(learning_rate = 0.1),
      loss = "mse",
      metrics = ["mae"]
    )

    h = model.fit(x = xy, y = labels, batch_size = 1, epochs = 20, verbose = 0)

    w = model.get_weights()[0]
    b = model.get_weights()[1]
```

```
[7]: draw_data(xy, labels, 2, w, b, h.history["mae"])
```

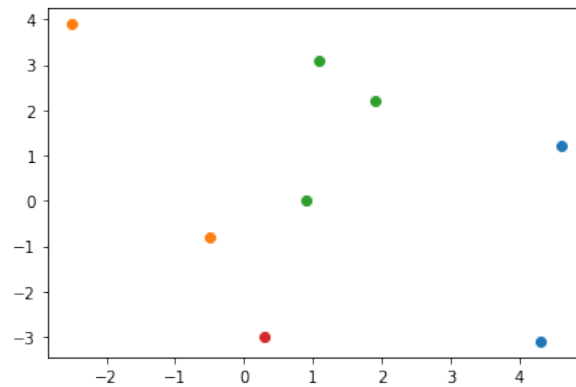


Задание 2

```
[8]: x = [4.3, -2.5, 0.9, 1.1, 0.3, -0.5, 4.6, 1.9]
      y = [-3.1, 3.9, 0, 3.1, -3, -0.8, 1.2, 2.2]
      labels = [(0, 0), (1, 1), (0, 1), (0, 1), (1, 0), (1, 1), (0, 0), (0, 1)]
```

Визуализация данных

```
[9]: xy = make_xy(x, y)
      draw_data(xy, labels, 4)
```



Обучение модели

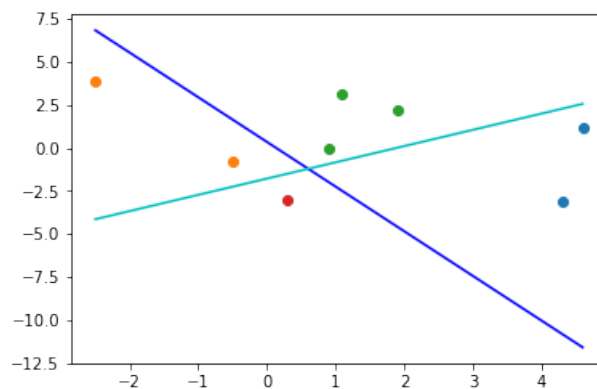
```
[10]: model = tf.keras.models.Sequential([
    tf.keras.layers.Dense(2, input_dim = 2, activation = "sigmoid")
])

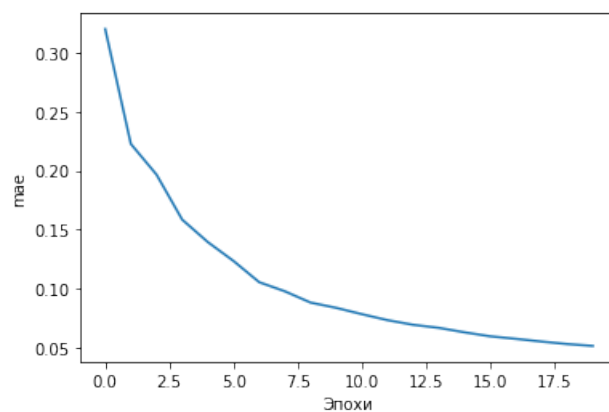
model.compile(
    optimizer = tf.keras.optimizers.Adam(learning_rate = 0.1),
    loss = "mse",
    metrics = ["mae"]
)

h = model.fit(x = xy, y = labels, batch_size = 1, epochs = 20, verbose = 0)

w = model.get_weights()[0]
b = model.get_weights()[1]
```

```
[11]: draw_data(xy, labels, 4, w, b, h.history["mae"])
```





Вывод

В ходе выполнения лабораторной работы я вспомнил основы использования библиотеки Tensorflow и Keras, реализовал однослойный персептрон. Данные получились линейно разделимыми.

[]: