

Лабораторная работа 2
по курсу
«Проектирование интеллектуальных систем»
Тема:
«Логистическая регрессия и полносвязная сеть»

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Проверил:

1. Задание на лабораторную работу

1. Создать логистическую регрессию для классификации набора данных MNIST. Функция логистической регрессии выглядит следующим образом:

$$f(x) = \frac{1}{1+e^{\omega \cdot x + b}}$$

2. Создать нейронную сеть с 5 полносвязными слоями для классификации набора данных MNIST с количеством нейронов в слоях от первого до пятого - (200,100,60,30,10)

1.1. 1. Логистическая регрессия

1.1.1. 1.1 Импорт библиотек

```
In [2]: import tensorflow as tf
        print("TensorFlow version is ", tf.__version__)
        from tensorflow.examples.tutorials.mnist import input_data
```

TensorFlow version is 1.15.0

1.1.2. 1.2 Placeholders для входных данных и истинных значений, Variables для весов и отступов

```
In [3]: x = tf.placeholder(tf.float32, [None, 784])
        W = tf.Variable(tf.zeros ([784, 10]))
        b = tf.Variable(tf.zeros ([10]), dtype=tf.float32 )
        y_true = tf.placeholder(tf.float32, [None, 10])
```

```
In [6]: y_pred = 1 / (1 + tf.exp(tf.matmul(x, W) + b))
```

```
In [13]: DATA_DIR = 'tmp/data'
        NUM_STEPS = 1000
        MINIBATCH_SIZE = 100
        LEARNING_RATE = 0.5
        data = input_data.read_data_sets(DATA_DIR, one_hot=True)
```

Extracting tmp/data/train-images-idx3-ubyte.gz
Extracting tmp/data/train-labels-idx1-ubyte.gz
Extracting tmp/data/t10k-images-idx3-ubyte.gz
Extracting tmp/data/t10k-labels-idx1-ubyte.gz

```
In [14]: cross_entropy = tf. reduce_mean (tf. nn . softmax_cross_entropy_with_logits(logits = y_pred , labels = y_true ))
        gd_step = tf. train . GradientDescentOptimizer ( LEARNING_RATE ). minimize (cross_entropy)
        correct_mask = tf. equal (tf. argmax ( y_pred , 1 ) , tf. argmax ( y_true , 1 ))
        accuracy = tf. reduce_mean (tf. cast ( correct_mask , tf. float32 ))
```

1.1.3. 1.3 Обучение модели

```
In [12]: with tf. Session () as sess :
        # Train
```

```

sess.run(tf.global_variables_initializer())
for i in range(NUM_STEPS):
    batch_x, batch_y = data.train.next_batch(MINIBATCH_SIZE)
    sess.run(gd_step, feed_dict={x: batch_x, y_true: batch_y})
    ans = sess.run(accuracy, feed_dict={x: data.test.images, y_true : data.test.labels})

print (" Accuracy : {:.4}% ". format ( ans *100))

```

Accuracy : 88.59%

Точность при обучении вышла 88,59%

1.2. 2. Нейронная сеть

1.2.1. 2.1 Количество нейронов в слоях

```

In [15]: # layers sizes
L1 = 200
L2 = 100
L3 = 60
L4 = 30
L5 = 10

```

1.2.2. 2.2 Данные и слои

```

In [17]: data = input_data.read_data_sets(DATA_DIR, one_hot=True)
x = tf.placeholder(tf.float32, [None, 784])
l1 = tf.layers.dense(x, L1, activation=tf.nn.relu, use_bias=True)
l2 = tf.layers.dense(l1, L2, activation=tf.nn.relu, use_bias=True)
l3 = tf.layers.dense(l2, L3, activation=tf.nn.relu, use_bias=True)
l4 = tf.layers.dense(l3, L4, activation=tf.nn.relu, use_bias=True)

y_pred = tf.layers.dense(l4, L5, activation=tf.nn.relu, use_bias=True)
y_true = tf.placeholder (tf.float32, [None, 10])

```

```

Extracting tmp/data/train-images-idx3-ubyte.gz
Extracting tmp/data/train-labels-idx1-ubyte.gz
Extracting tmp/data/t10k-images-idx3-ubyte.gz
Extracting tmp/data/t10k-labels-idx1-ubyte.gz

```

1.2.3. 2.3 Кросс-энтропия

```

In [18]: cross_entropy = tf.reduce_mean(tf.nn.softmax_cross_entropy_with_logits(logits=y_pred, labels=y_true))
gd_step = tf.train.GradientDescentOptimizer(LEARNING_RATE).minimize(cross_entropy)
correct_mask = tf.equal(tf.argmax(y_pred, 1), tf.argmax(y_true, 1))
accuracy = tf.reduce_mean(tf.cast(correct_mask, tf.float32))

```

1.2.4. 2.4 Обучение

```
In [20]: with tf.Session() as sess:
          sess.run(tf.global_variables_initializer())
          for i in range(NUM_STEPS):
              batch_x, batch_y = data.train.next_batch(MINIBATCH_SIZE)
              sess.run(gd_step, feed_dict={x: batch_x, y_true: batch_y})
              ans = sess.run(accuracy, feed_dict={x: data.test.images, y_true: data.test.labels})

          print("Accuracy : {:.4}% ".format(ans*100))
```

Accuracy : 86.44%

Точность обучения составила 86,44%

1.3. Список литературы

- [1] Google. Tensorflow. 2018. Feb. url - https://www.tensorflow.org/install/install_windows.
- [2] url - <https://virtualenv.pypa.io/en/stable/userguide/>.
- [3] Microsoft. about_Execution_Policies. 2018. url - <https://technet.microsoft.com/en-us/library/dd347641.aspx>.
- [4] Jupyter Project. Installing Jupyter. 2018. url - <http://jupyter.org/install>.