# 数据分析及实践-实验四

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#### 数据分析及实践-实验四

### Part 1 分类算法实践

#### 1.1 算法主要流程

Step 1. 将lab3中特征工程得到的DIY.csv转换为.npy文件,并进行5折交叉验证所需要的train/test数据集划分

```
data = np.load('./data/DIY.npy')

data_part = cross_validation(data, 5)

for i in range(5):
    print('part: %d' % (i+1))
    train, test = split_train_test(data_part, i)
```

Step 2. 用logistic回归进行二分类。

```
class MLP(object):
    # ...
for i in range(5):
    print('part: %d' % (i+1))
    train, test = split_train_test(data_part, i)
    model = MLP(train, test, train.shape[1]-1, 1, 0.1, 5001)
    model.train_model()
```

### 1.2 算法关键技术

Step 1: 将DIY.npy数据分为5个part,其中四个concatenate成为train,剩余一个为test。

Step 2: 构造class: MLP,实现sigmoid,sigmoid\_derivative(对sigmoid求导),forward(前向传播),judge(judge>0则代表predict结果正确),backward(反向传播),test\_model(在test集上计算正确率),train\_model(在train集上计算正确率)方法。

#### 1.3 算法实现

关键代码结构如下。完整代码见附录。

```
1
    import numpy as np
 2
    def cross_validation(data, k):
 3
4
        # split the data into k parts
 5
        # return the k parts
        return data.reshape(k, -1, data.shape[1])
 6
 7
8
    def split_train_test(data_part, k):
9
        # split the data_part into train and test
10
        # return the train and test
        test = data_part[k, :, :]
11
        train = np.delete(data_part, k, axis=0)
12
13
        train = np.concatenate(train, axis=0)
14
        return train, test
15
16
    class MLP(object):
        def __init__(self, train, test, n_in, n_out, lr, epoch):
17
18
            # ...
19
20
        def sigmoid(self, x):
21
             # ...
22
        def sigmoid_derivative(self, x):
23
24
            # ...
25
26
        def forward(self, x): # (bsz, n_in)
27
            # ...
28
29
        def judge(self, y_hat, y):
30
31
            if (y_hat > 0.5 \text{ and } y = 1) or (y_hat < 0.5 \text{ and } y = 0), then return positive:
    right predict!
32
            create a function to judge whether the prediction is correct
33
34
             # ...
35
36
        def backward(self, x, y): # (bsz, n_in) , (bsz, n_out)
37
            # ...
38
39
40
        def test_model(self, data, label):
41
            # ...
42
43
        def train_model(self):
44
            # ...
45
46
    def main():
        data = np.load('./data/DIY.npy')
47
48
49
        data_part = cross_validation(data, 5)
50
51
        for i in range(5):
             print('part: %d' % (i+1))
52
```

```
train, test = split_train_test(data_part, i)
model = MLP(train, test, train.shape[1]-1, 1, 0.1, 5001)
model.train_model()
print('\n\n')

if __name__ == '__main__':
main()
```

#### 1.4 实验记录

(k折交叉验证, 4:1比例, 共有5折)

在learning\_rate=0.1, epoch=5001时的结果如下:

```
C:\Users\x\Desktop\twodown\USTC_AD2022_Lab\lab4>python lab4_part1.py
part: 1
epoch: 5000, accs: 0.917989

part: 2
epoch: 5000, accs: 0.929038

part: 3
epoch: 5000, accs: 0.922502

part: 4
epoch: 5000, accs: 0.931684

part: 5
epoch: 5000, accs: 0.929194
```

k	ACC
1	0.917989
2	0.929038
3	0.922502
4	0.931684
5	0.929194
平均值	0.926081

## Part 2 预测算法实践

- 1. 尝试了lightgbm库
- 2. 尝试了sklearn库中的12种回归算法

#### 2.1 lightgbm

### 附录

#### part 1完整代码及实现细节

```
1
    import numpy as np
 2
    def cross_validation(data, k):
 4
        # split the data into k parts
 5
        # return the k parts
 6
        return data.reshape(k, -1, data.shape[1])
 7
 8
    def split_train_test(data_part, k):
 9
        # split the data_part into train and test
10
        # return the train and test
        test = data_part[k, :, :]
11
12
        train = np.delete(data_part, k, axis=0)
13
        train = np.concatenate(train, axis=0)
14
        return train, test
15
16
    class MLP(object):
17
        def __init__(self, train, test, n_in, n_out, lr, epoch):
18
            # self.bsz = train.shape[0]
19
            # self.train = train
20
            # self.test = test
21
            self.n_in = n_in
22
            self.n_out = n_out
            self.lr = lr
23
24
            self.epoch = epoch
25
            self.train_data = train[:, :-1]
26
            train_label = train[:, -1]
27
            self.train_label = train_label.reshape(train_label.shape[0], 1)
            self.test_data = test[:, :-1]
28
29
            test_label = test[:, -1]
30
            self.test_label = test_label.reshape(test_label.shape[0], 1)
31
            self.w = np.random.randn(self.n_in, self.n_out)
32
            self.b = np.random.randn(self.n_out)
33
        def sigmoid(self, x):
34
35
            return 1 / (1 + np.exp(-x))
36
37
        def sigmoid_derivative(self, x):
38
            return self.sigmoid(x) * (1 - self.sigmoid(x))
39
40
        def forward(self, x): # (bsz, n_in)
41
            o = np.dot(x, self.w) + self.b # (bsz, n_out)
42
            y_hat = self.sigmoid(o) # (bsz, n_out)
43
            return o, y_hat
44
        def judge(self, y_hat, y):
45
```

```
46
47
             if (y_hat > 0.5 \text{ and } y = 1) or (y_hat < 0.5 \text{ and } y = 0), then return positive:
     right predict!
             create a function to judge whether the prediction is correct
48
49
             s = (y_hat - 0.5) * (y - 0.5)
50
51
             return s
52
         def backward(self, x, y): # (bsz, n_in) , (bsz, n_out)
53
54
             bsz = x.shape[0]
             o, y_hat = self.forward(x) # (bsz, n_out)
55
56
57
             # 链式法则
58
             d_L_d_yhat = -y/y_hat + (np.ones_like(y)-y)/(np.ones_like(y)-y_hat) #
     (bsz, n_out)
59
             d_y_hat_d_o = self.sigmoid_derivative(y_hat) # (bsz, n_out)
60
             d_o_d_w = x \# (bsz, n_in)
             d_o_d_b = np.ones((bsz, 1)) # (bsz, 1)
61
62
             d_L_d_w = np.mean(d_L_d_y_hat * d_y_hat_d_o * d_o_d_w, axis=0) # (n_in_i)
63
             d_L_d_w = d_L_d_w.reshape(self.n_in, 1) # (n_in, 1)
64
65
             d_L_d_b = np.mean(d_L_d_y_hat * d_y_hat_d_o * d_o_d_b, axis=0) # (1,)
             self.w = self.w - self.lr * d_L_d_w
66
67
             self.b = self.b - self.lr * d_L_d_b
68
69
70
         def test_model(self, data, label):
71
             # pdb.set_trace()
72
             o, y_hat = self.forward(data)
73
             total = data.shape[0]
74
75
             correct are those whose prediction is correct, i.e. y_hat > 0.5 and y = 1
76
77
             correct = np.sum(self.judge(y_hat, label)>0)
78
             accs = correct / total
79
             return accs
80
81
         def train_model(self):
82
             for i in range(self.epoch):
83
                 self.backward(self.train_data, self.train_label)
                 if i!=0 and i%5000 == 0:
84
85
                      accs = self.test_model(self.test_data, self.test_label)
86
                     print('epoch: %d, accs: %f' % (i, accs))
87
88
     def main():
89
         data = np.load('./data/DIY.npy')
90
91
         data_part = cross_validation(data, 5)
92
93
         for i in range(5):
             print('part: %d' % (i+1))
94
95
             train, test = split_train_test(data_part, i)
             model = MLP(train, test, train.shape[1]-1, 1, 0.1, 5001)
96
97
             model.train_model()
98
             print('\n\n')
99
     if __name__ == '__main__':
100
101
         main()
```