

在 [] 中:

```
import torch
a = torch.cuda.is_available()
```

实验一 参考代码

在 [] 中:

```
from sklearn.datasets import load_breast_cancer
from sklearn.linear_model import LogisticRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import recall_score
from sklearn.metrics import precision_score
from sklearn.metrics import classification_report
from sklearn.metrics import accuracy_score
import warnings
warnings.filterwarnings("ignore")
# 这个方法只是解决了表面，没有根治
```

在 [] 中:

```
cancer=load_breast_cancer()#加载乳腺癌数据
X_train,X_test,y_train,y_test=train_test_split(cancer.data,cancer.target,test_size=0.2)
model = LogisticRegression()
model.fit(X_train,y_train)
train_score=model.score(X_train,y_train)
test_score=model.score(X_test,y_test)
print(f'train score:{train_score:.6f};testscore:{test_score:.6f}')

y_pred=model.predict(X_test)
accuracy_score_value= accuracy_score(y_test,y_pred)
recall_score_value = recall_score(y_test,y_pred)
precision_score_value=precision_score(y_test,y_pred)
classification_report_value=classification_report(y_test,y_pred)
print(f"准确率: {accuracy_score_value}")
print(f"召回率: {recall_score_value}")
print(f"精确率: {precision_score_value}")
print(classification_report_value)
```

train score:0.962637;testscore:0.929825

准确率: 0.9298245614035088

召回率: 0.9705882352941176

精确率:0.9166666666666666

	precision	recall	f1-score	support
0	0.95	0.87	0.91	46
1	0.92	0.97	0.94	68
accuracy			0.93	114
macro avg	0.93	0.92	0.93	114
weighted avg	0.93	0.93	0.93	114

实验二参考代码

在 [] 中:

```
import numpy as np
from sklearn.datasets import load_breast_cancer
from sklearn.model_selection import train_test_split
from sklearn.metrics import recall_score
from sklearn.metrics import precision_score
from sklearn.metrics import classification_report
from sklearn.metrics import accuracy_score
```

在 [] 中:

```
def loadTrainData():
    cancer=load_breast_cancer()#加载乳腺癌数据
    X=cancer.data
    #加载乳腺癌判别特征
    y=cancer.target#两个TAG, y=0时为阴性, y=1时为阳性
    #将数据集划分为训练集和测试集, 测试集占比为8, 2
    X_train,X_test,y_train,y_test = train_test_split(X, y, test_size=0.2)
    X_train = X_train.T
    X_test = X_test.T
    return X_train,X_test,y_train,y_test
```

在 [] 中:

```
def sigmoid(inx):
    from numpy import exp
    return 1.0/(1.0 + exp(-inx))
```

初始化参数

在 [] 中:

```
def initialize_para(dim):
    mu = 0
    sigma =0.1
    np.random.seed()
    w = np.random.normal(mu, sigma, dim)
    w = np.reshape(w, (dim,1))
    b=0
    return w, b
```

前向传播

在 [] 中:

```
def propagate(w, b, X, Y):  
    #eps防止log运算遇到e  
    eps = 1e-5  
    m = X.shape[1]  
    #计算初步运算结果  
    A = sigmoid(np.dot(w.T, x) + b)  
    #计算损失函数值大小  
    cost = -1 / m * np.sum(np.multiply(Y, np.log(A + eps)) + np.multiply(1 - Y, np.log(1 - A + eps)))  
    #计算梯度值  
    dw = 1 / m * np.dot(X, (A - Y).T)  
    db = 1 / m * np.sum(A - Y)  
    cost = np.squeeze(cost)  
  
    grads = {"dw": dw,  
            "db": db}  
    #返回损失函数大小以及反向传播的梯度值  
    return grads, cost, A
```

num1 terations**梯度下降次数**

learning_rate**学习率**

在 [] 中:

```
def optimize(w, b, X, Y, num_iterations, learning_rate):
    costs=[] #记录损失函数值
    #循环进行梯度下降
    for i in range(num_iterations):
        print(i)
        grads, cost, pre_y = propagate(w, b, X, Y)
        dw = grads["dw"]
        db = grads["db"]
        w=w - learning_rate * dw
        b=b - learning_rate * db
        #每100次循环记录一次损失函数大小并打印
        if i%100==0:
            costs.append(cost)
        if i%100==0:
            pre_Y[pre_Y >=0.5]=1
            pre_Y[pre_Y< 0.5]=0
            pre_Y=pre_Y.astype(np. int)
            acc =1 - np. sum(pre_Y^Y)/len(Y)
            print("Iteration: {} Loss = {}, Acc = {}".format(i, cost, acc))
    #最终参数值
    params = {"w":w,
              "b":b}
    return params, costs
```

在 [] 中:

```
def predict(w, b, X):  
    #样本个数  
    m = X.shape[1]  
    #初始化预测输出  
    Y_prediction = np.zeros((1, m))  
    #转置参数向量w  
    w=w.reshape(X.shape[0], 1)  
    #预测结果  
    Y_hat = sigmoid(np.dot(w.T, X)+b)  
    #将结果按照0.5的阈值转化为0/1  
    for i in range(Y_hat.shape[1]):  
        if Y_hat[:, i]>0.5:  
            Y_prediction[:, i]=1  
        else:  
            Y_prediction[:, i]=0  
    return Y_prediction
```

在 [] 中:

```
#训练以及预测
def Logisticmodel(X_train,Y_train,X_test,Y_test,num_iterations=1000,learning_rate=0.1):
    #初始化参数w,b
    w,b = initialize_para(X_train.shape[0])
    #梯度下降找到最优参数
    parameters, costs = optimize(w,b,X_train,Y_train,num_iterations,learning_rate)
    w = parameters["w"]
    b = parameters["b"]
    #训练集测试集的预测结果
    Y_prediction_train = predict(w,b,X_train)
    Y_prediction_test = predict(w,b,X_test)
    Y_prediction_test = Y_prediction_test.T
    #模型评价
    accuracy_score_value = accuracy_score(Y_test,Y_prediction_test)
    recall_score_value = recall_score(Y_test,Y_prediction_test)
    precision_score_value = precision_score(Y_test,Y_prediction_test)
    classification_report_value = classification_report(Y_test,Y_prediction_test)
    print("准确率: ",accuracy_score_value)
    print("召回率: ",recall_score_value)
    print("精确率: ",precision_score_value)
    print(classification_report_value)
    d = {"costs":costs,
        "Y_prediction_test":Y_prediction_test,
        "Y_prediction_train":Y_prediction_train,
        "w":w,"b":b,
        "learning_rate":learning_rate,
        "num_iterations":num_iterations}
    return d
```

在 [] 中:

```
if __name__=='_main_':
    X_train,X_test,y_train,y_test = loadTrainData()
    Logisticmodel(X_train,y_train,X_test,y_test)
```