SVM Classification import required package import numpy as np import pandas as pd import seaborn as sns import matplotlib.pyplot as plt read data from source and describing df = pd.read csv('hearing test.csv') df.head() age physical_score test_result **0** 33.0 40.7 1 **1** 50.0 37.2 1 **2** 52.0 24.7 0 **3** 56.0 31.0 0 **4** 35.0 42.9 1 print(df.columns) Index(['age', 'physical_score', 'test_result'], dtype='object') In [4]: print(df.describe()) age physical_score test_result 5000 5000.000000 5000.000000 count 5000.000000 51.609000 32.760260 0.600000 mean 11.287001 8.169802 std 0.489947 -0.000000 0.000000 min 18.000000 25% 43.000000 26.700000 0.000000 35.300000 1.000000 50% 51.000000 75% 60.000000 38.900000 1.000000 max 90.000000 50.000000 1.000000 ## chec for data types print(df.info()) <class 'pandas.core.frame.DataFrame'> RangeIndex: 5000 entries, 0 to 4999 Data columns (total 3 columns): # Column Non-Null Count Dtype 5000 non-null float64 physical score 5000 non-null float64 2 test result 5000 non-null int64 dtypes: float64(2), int64(1)memory usage: 117.3 KB None sns.pairplot(df) Out[6]: <seaborn.axisgrid.PairGrid at 0x21c35280c10> 80 60 20 50 40 physical score 20 10 0 1.0 0.8 test result 0.6 0.4 0.2 0.0 0.00 0.25 0.50 20 60 20 age physical_score test_result check the relation between variables corr = df.corr() corr.style.background gradient(cmap = 'Greens') physical_score test_result 1.000000 age -0.782146 -0.683171 physical_score -0.782146 1.000000 0.792716 test_result -0.683171 0.792716 1.000000 select input and op variable x = df.drop(['test_result'], axis=1) y = df['test result'] splitting the data from sklearn.model selection import train test split x train, x test, y train, y test = train test split(x, y, random state= 123456, train creating a model from sklearn.svm import SVC model = SVC(C = 2)## fit the data model.fit(x_train, y_train) Out[11]: SVC(C=2) parameters to fine tune model print(model.get_params()) {'C': 2, 'break_ties': False, 'cache_size': 200, 'class_weight': None, 'coef0': 0.0,
'decision_function_shape': 'ovr', 'degree': 3, 'gamma': 'scale', 'kernel': 'rbf', 'max
_iter': -1, 'probability': False, 'random_state': None, 'shrinking': True, 'tol': 0.00 'break ties': False, 'cache sıze': ∠∪∪, 1, 'verbose': False} predict the values for unseen data y prediction = model.predict(x test) #print(y prediction) evaluation of classification model In [14]: from sklearn.metrics import confusion matrix cm = confusion_matrix(y_test, y_prediction) print(cm) [[339 54] [21 586]] correct = cm[0,0] + cm[1,1]wrong = cm[1,0] + cm[0,1]total = correct + wrong accuracy = correct/total print(accuracy) 0.925 from sklearn.metrics import accuracy score print(accuracy_score(y_test, y_prediction)) 0.925 precision value from sklearn.metrics import precision score print(precision score(y test, y prediction)) 0.915625 recal value from sklearn.metrics import recall_score print(recall_score(y_test, y_prediction)) 0.9654036243822076 F1 score from sklearn.metrics import f1 score print(f1 score(y test, y prediction)) 0.9398556535685646 classification report from sklearn.metrics import classification report print(classification_report(y_test, y_prediction)) precision recall f1-score support 0 0.94 0.86 0.90 393 0.92 0.97 0.94 607 1000 accuracy 0.93 0.93 0.91 0.92 1000 macro avg weighted avg 0.93 0.93 0.92 1000 from sklearn.metrics import roc curve, roc auc score, plot roc curve print(roc auc score(y test, y prediction)) 0.9139995221147679 fpr, tpr, threshold = roc_curve(y_test, y_prediction) print(fpr) print(tpr) print(threshold) plt.plot(fpr, tpr) 0.13740458 1. [0. 0.96540362 1. [2 1 0] Out[22]: [<matplotlib.lines.Line2D at 0x21c372fcb20>] 0.8 0.6 0.4 0.2 0.0 0.2 0.4 0.6 0.8 0.0 1.0 plot = plot roc curve(model, x test, y test) plt.plot([0,1], [0,1], linestyle = '--')Out[23]: [<matplotlib.lines.Line2D at 0x21c3735b850>] 1.0 True Positive Rate (Positive label: 1) 0.8 0.6 0.4 0.2 SVC (AUC = 0.97) 0.0 0.0 0.4 0.8 False Positive Rate (Positive label: 1)