ex12

September 18, 2024

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[49]: from ucimlrepo import fetch_ucirepo
      import pandas as pd
      from sklearn.model_selection import train_test_split, GridSearchCV,_
       →RandomizedSearchCV
      from sklearn.preprocessing import StandardScaler
      from sklearn.ensemble import RandomForestClassifier
      from sklearn.svm import SVC
      from sklearn.metrics import accuracy_score
[50]: heart_disease = fetch_ucirepo(id=45)
      X = heart_disease.data.features
      y = heart_disease.data.targets
      # print(heart_disease.metadata)
      # print(heart_disease.variables)
      data = pd.DataFrame(X).join(pd.DataFrame(y))
      print(data.isnull().sum())
                 0
     age
                 0
     sex
     ср
     trestbps
                 0
     chol
                 0
     fbs
                 0
     restecg
     thalach
     exang
                 0
     oldpeak
     slope
                 0
                 4
     ca
                 2
     thal
     num
     dtype: int64
```

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[51]: data.dropna(inplace=True)
      print(data.isnull().sum())
                  0
     age
                  0
     sex
                  0
     ср
     trestbps
                  0
     chol
                  0
     fbs
                  0
     restecg
                  0
     thalach
                  0
     exang
     oldpeak
                  0
     slope
                  0
                  0
     ca
                  0
     thal
     num
                  0
     dtype: int64
[52]: data.head()
[52]:
                       trestbps
                                 chol fbs
                                             restecg thalach exang oldpeak slope
         age
              sex
                   ср
      0
          63
                    1
                             145
                                   233
                                          1
                                                    2
                                                           150
                                                                    0
                                                                            2.3
                                                                                     3
                1
          67
                    4
                             160
                                   286
                                          0
                                                    2
                                                           108
                                                                    1
                                                                            1.5
                                                                                     2
      1
                1
      2
                    4
                             120
                                   229
                                          0
                                                    2
                                                           129
                                                                            2.6
                                                                                     2
          67
                1
                                                                    1
                                                                                     3
      3
          37
                1
                    3
                             130
                                   250
                                          0
                                                    0
                                                           187
                                                                    0
                                                                            3.5
      4
          41
                0
                    2
                             130
                                   204
                                          0
                                                    2
                                                           172
                                                                    0
                                                                            1.4
                                                                                     1
          ca
             thal
                    num
               6.0
      0.0
      1 3.0
               3.0
                      2
      2 2.0
               7.0
                      1
      3 0.0
               3.0
                      0
      4 0.0
               3.0
                      0
[53]: X = data.drop('num', axis=1)
      y = data['num']
      X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2,_
       →random_state=42)
      scaler = StandardScaler()
      X_train = scaler.fit_transform(X_train)
      X_test = scaler.transform(X_test)
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[54]: rf_model = RandomForestClassifier(random_state=42)
      rf_model.fit(X_train, y_train)
      rf_pred = rf_model.predict(X_test)
      rf_accuracy = accuracy_score(y_test, rf_pred)
      print(f"Random Forest Accuracy: {rf_accuracy:.2f}")
      svc model = SVC(random state=42)
      svc_model.fit(X_train, y_train)
      svc_pred = svc_model.predict(X_test)
      svc_accuracy = accuracy_score(y_test, svc_pred)
      print(f"SVC Accuracy: {svc accuracy:.2f}")
     Random Forest Accuracy: 0.60
     SVC Accuracy: 0.65
[55]: rf_params = {
          'n_estimators': [50, 100, 200],
          'max_depth': [None, 5, 2],
          'min_samples_split': [2, 5, 3]
      }
      svc_params = {
          'C': [0.1, 1, 10],
          'kernel': ['linear', 'rbf'],
          'gamma': ['scale', 'auto']
      }
      rf_grid = GridSearchCV(estimator=rf_model, param_grid=rf_params,_
       ⇔scoring='accuracy', cv=5)
      rf_grid.fit(X_train, y_train)
      print("Best parameters for Random Forest (GridSearchCV):", rf_grid.best_params_)
      print("Best accuracy for Random Forest (GridSearchCV):", rf_grid.best_score_)
      svc_grid = GridSearchCV(estimator=svc_model, param_grid=svc_params,_
      ⇔scoring='accuracy', cv=5)
      svc_grid.fit(X_train, y_train)
      print("Best parameters for SVC (GridSearchCV):", svc_grid.best_params_)
      print("Best accuracy for SVC (GridSearchCV):", svc_grid.best_score_)
     Best parameters for Random Forest (GridSearchCV): { 'max depth': None,
     'min_samples_split': 3, 'n_estimators': 50}
     Best accuracy for Random Forest (GridSearchCV): 0.5613475177304965
     Best parameters for SVC (GridSearchCV): {'C': 0.1, 'gamma': 'scale', 'kernel':
     'linear'}
     Best accuracy for SVC (GridSearchCV): 0.5823581560283688
[56]:
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rf_random = RandomizedSearchCV(estimator=rf_model,__
       →param_distributions=rf_params, n_iter=10, scoring='accuracy', cv=5, 
       →random_state=42)
      rf random.fit(X train, y train)
      print("Best parameters for Random Forest (RandomizedSearchCV):", rf_random.
       ⇒best params )
      print("Best accuracy for Random Forest (RandomizedSearchCV):", rf_random.
       ⇔best_score_)
      svc random = RandomizedSearchCV(estimator=svc model,___
       →param_distributions=svc_params, n_iter=10, scoring='accuracy', cv=5,
       →random_state=42)
      svc_random.fit(X_train, y_train)
      print("Best parameters for SVC (RandomizedSearchCV):", svc_random.best_params_)
      print("Best accuracy for SVC (RandomizedSearchCV):", svc_random.best_score_)
     Best parameters for Random Forest (RandomizedSearchCV): {'n_estimators': 50,
     'min_samples_split': 5, 'max_depth': 5}
     Best accuracy for Random Forest (RandomizedSearchCV): 0.5611702127659575
     Best parameters for SVC (RandomizedSearchCV): {'kernel': 'linear', 'gamma':
     'scale', 'C': 0.1}
     Best accuracy for SVC (RandomizedSearchCV): 0.5823581560283688
[57]: print("\nComparison of GridSearchCV and RandomizedSearchCV:")
      print("Random Forest - GridSearchCV:", rf_grid.best_params_, "Accuracy:", __
       →rf_grid.best_score_)
      print("Random Forest - RandomizedSearchCV:", rf_random.best_params_, "Accuracy:

¬", rf_random.best_score_)
      print("SVC - GridSearchCV:", svc grid.best params , "Accuracy:", svc grid.
       ⇔best_score_)
      print("SVC - RandomizedSearchCV:", svc_random.best_params_, "Accuracy:", __

svc_random.best_score_)
     Comparison of GridSearchCV and RandomizedSearchCV:
     Random Forest - GridSearchCV: {'max depth': None, 'min_samples_split': 3,
     'n_estimators': 50} Accuracy: 0.5613475177304965
     Random Forest - RandomizedSearchCV: {'n_estimators': 50, 'min_samples_split': 5,
     'max_depth': 5} Accuracy: 0.5611702127659575
     SVC - GridSearchCV: {'C': 0.1, 'gamma': 'scale', 'kernel': 'linear'} Accuracy:
     0.5823581560283688
     SVC - RandomizedSearchCV: {'kernel': 'linear', 'gamma': 'scale', 'C': 0.1}
     Accuracy: 0.5823581560283688
```