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requirements.txt
from sklearn.svm
import SVC
import optuna
import torch
import scipy.io
import os
import data
def svm_objective(trial, X_train, y_train):
 C = trial.suggest_loguniform('C', 1e-3, 1e3)
 kernel = trial.suggest_categorical('kernel', ['linear', 'poly', 'rbf', 'sigmoid'])
 model = SVC(C=C, kernel=kernel)
 return cross_val_score(model, X_train, y_train, n_jobs=-1, cv=3).mean()
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def get_best_svm_model(X_train, y_train):
 study = optuna.create_study(direction='maximize')
 study.optimize(lambda trial: svm_objective(trial, X_train, y_train), n_trials=50)
 best_params = study.best_trial.params
 best_model = SVC(**best_params)
 best_model.fit(X_train, y_train)
 return best_model, study.best_trial
from sklearn.svm
import SVC
import optuna
import data
def svm_objective(trial, X_train, y_train):
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 best_params = study.best_trial.params
 best_model = SVC(**best_params)
 best_model.fit(X_train, y_train)
 return best_model, study.best_trial
from sklearn.tree import DecisionTreeClassifier
import optuna
import data
def dt_objective(trial, X_train, y_train):
 max_depth = trial.suggest_int('max_depth', 1, 32)
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model = DecisionTreeClassifier(max_depth=max_depth)
 return cross_val_score(model, X_train, y_train, n_jobs=-1, cv=3).mean()
def get_best_dt_model(X_train, y_train):
 study = optuna.create_study(direction='maximize')
 study.optimize(lambda trial: dt_objective(trial, X_train, y_train), n_trials=50)
 best_params = study.best_trial.params
 best_model = DecisionTreeClassifier(**best_params)
 best_model.fit(X_train, y_train)
 return best_model, study.best_trial
from sklearn.discriminant_analysis import LinearDiscriminantAnalysis
def get_best_blda_model(X_train, y_train):
 model = LinearDiscriminantAnalysis()
 model.fit(X_train, y_train)
 return model
from sklearn.neighbors import KNeighborsClassifier
import optuna
import data
def knn_objective(trial, X_train, y_train):
 n_neighbors = trial.suggest_int('n_neighbors', 1, 50)
 model = KNeighborsClassifier(n_neighbors=n_neighbors)
 return cross_val_score(model, X_train, y_train, n_jobs=-1, cv=3).mean()
def get_best_knn_model(X_train, y_train):
 study = optuna.create_study(direction='maximize')
 study.optimize(lambda trial: knn_objective(trial, X_train, y_train), n_trials=50)
 best_params = study.best_trial.params
 best_model = KNeighborsClassifier(**best_params)
 best_model.fit(X_train, y_train)
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return best_model, study.best_trial
from xgboost import XGBClassifier
import optuna
import data
def xgb_objective(trial, X_train, y_train):
 params = {
    'n_estimators': trial.suggest_int('n_estimators', 50, 500),
    'learning_rate': trial.suggest_loguniform('learning_rate', 0.01, 0.3),
    'max_depth': trial.suggest_int('max_depth', 3, 10),
    'eval_metric': 'mlogloss'
 }
 model = XGBClassifier(**params, use_label_encoder=False)
 return cross_val_score(model, X_train, y_train, n_jobs=-1, cv=3).mean()
def get_best_xgb_model(X_train, y_train):
 study = optuna.create_study(direction='maximize')
 study.optimize(lambda trial: xgb_objective(trial, X_train, y_train), n_trials=50)
 best_params = study.best_trial.params
 best_model = XGBClassifier(**best_params, use_label_encoder=False)
 best_model.fit(X_train, y_train)
 return best_model, study.best_trial
import pandas as pd
from sklearn.preprocessing import StandardScaler, LabelEncoder
from sklearn.model_selection import train_test_split
def preprocess_data(file_path):
 df = pd.read_csv(file_path)
 df = df.dropna()
 label_encoders = {}
 for column in df.select_dtypes(include=['object']).columns:
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le = LabelEncoder()
   df[column] = le.fit_transform(df[column])
   label_encoders[column] = le
 X = df.drop('target', axis=1)
 y = df['target']
 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)
 return X_train, X_test, y_train, y_test, label_encoders
from preprocessing.preprocess import preprocess_data
from models.svm_model import get_best_svm_model
from models.dt_model import get_best_dt_model
from models.blda_model import get_best_blda_model
from models.knn_model import get_best_knn_model
from models.xgb_model import get_best_xgb_model
# Cargar y preprocesar datos
X_train, X_test, y_train, y_test, label_encoders = preprocess_data('data/data.csv')
# Obtener y evaluar el mejor modelo SVM
svm_model, svm_trial = get_best_svm_model(X_train, y_train)
print(f'SVM Best Params: {svm_trial.params}')
print(f'SVM Test Accuracy: {svm_model.score(X_test, y_test)}')
# Obtener y evaluar el mejor modelo DT
dt_model, dt_trial = get_best_dt_model(X_train, y_train)
print(f'DT Best Params: {dt_trial.params}')
print(f'DT Test Accuracy: {dt_model.score(X_test, y_test)}')
# Obtener y evaluar el mejor modelo BLDA
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blda_model = get_best_blda_model(X_train, y_train)

print(f'BLDA Test Accuracy: {blda_model.score(X_test, y_test)}')

# Obtener y evaluar el mejor modelo KNN

knn_model, knn_trial = get_best_knn_model(X_train, y_train)

print(f'KNN Best Params: {knn_trial.params}')

print(f'KNN Test Accuracy: {knn_model.score(X_test, y_test)}')

# Obtener y evaluar el mejor modelo XGB

xgb_model, xgb_trial = get_best_xgb_model(X_train, y_train)

print(f'XGB Best Params: {xgb_trial.params}')

print(f'XGB Test Accuracy: {xgb_model.score(X_test, y_test)}')

pandas

scikit-learn

xgboost

optuna
```