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import pandas as pd
import numpy as np
from sklearn.model selection import train test split, GridSearchCV
from sklearn.preprocessing import StandardScaler, OneHotEncoder
from sklearn.linear model import LinearRegression
from sklearn.tree import DecisionTreeRegressor
from sklearn.ensemble import RandomForestRegressor, GradientBoostingRegressor
from sklearn.svm import SVR
from sklearn.neighbors import KNeighborsRegressor
from sklearn.metrics import r2 score
# Load dataset
df = pd.read csv("C:/Users/KIIT/Downloads/ev charging patterns1.csv")
# Drop high-cardinality or irrelevant columns
df.drop(columns=['User ID'], inplace=True)
# Convert time features into numeric (hour, minute)
for col in ['Charging Start Time', 'Charging End Time']:
  df[col] = pd.to datetime(df[col])
  df[f'(col)] + our' = df[col].dt.hour
  df[f'{col}_minute'] = df[col].dt.minute
  df.drop(columns=[col], inplace=True)
# Separate categorical and numerical columns
categorical cols = ['Vehicle Model', 'Charging Station ID', 'Charging Station Location',
            'Time of Day', 'Day of Week', 'Charger Type', 'User Type']
numerical cols = [col for col in df.columns if col not in categorical cols + ['Charging Duration
(hours)']]
# One-Hot Encode categorical features
df = pd.get dummies(df, columns=categorical cols, drop first=True)
# Handle missing values (Fill numerical columns with median)
df.fillna(df.median(numeric only=True), inplace=True)
# Define features and target
target column = "Charging Duration (hours)"
X = df.drop(columns=[target_column])
y = df[target_column]
# Train-test split
X train, X test, y train, y test = train test split(X, y, test size=0.2, random state=42)
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# Define models
models = {
  "Linear Regression": LinearRegression(),
  "Decision Tree": DecisionTreeRegressor(random state=42),
  "Random Forest": RandomForestRegressor(n estimators=100, random state=42),
  "Gradient Boosting": GradientBoostingRegressor(n estimators=100, random state=42),
  "Support Vector Regressor": SVR(),
  "KNN Regressor": KNeighborsRegressor(n neighbors=5)
}
# Initialize results dictionary
results = {"Stage": [], "Model": [], "R2 Score": []}
# **Step 1: Before Normalization**
for name, model in models.items():
  model.fit(X_train, y_train)
  y_pred = model.predict(X_test)
  score = r2 score(y test, y pred)
  results["Stage"].append("Before Normalization")
  results["Model"].append(name)
  results["R2 Score"].append(score)
# **Step 2: After Normalization**
scaler = StandardScaler()
X_train[numerical_cols] = scaler.fit_transform(X_train[numerical_cols])
X test[numerical cols] = scaler.transform(X test[numerical cols])
for name, model in models.items():
  model.fit(X_train, y_train)
  y_pred = model.predict(X_test)
  score = r2_score(y_test, y_pred)
  results["Stage"].append("After Normalization")
  results["Model"].append(name)
  results["R2 Score"].append(score)
# **Step 3: After Tuning**
tuned models = {
  "Random Forest": GridSearchCV(RandomForestRegressor(random_state=42),
                     param_grid={"n_estimators": [100, 200], "max_depth": [10, None]},
                     cv=3, n jobs=-1),
  "Gradient Boosting": GridSearchCV(GradientBoostingRegressor(random_state=42),
                       param_grid={"n_estimators": [100, 200], "learning_rate": [0.05, 0.1]},
                       cv=3, n jobs=-1),
  "Support Vector Regressor": GridSearchCV(SVR(),
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param_grid={"C": [0.1, 1, 10], "gamma": ["scale", "auto"]},
                           cv=3, n_jobs=-1),
  "KNN Regressor": GridSearchCV(KNeighborsRegressor(),
                    param_grid={"n_neighbors": [3, 5, 7]},
                    cv=3, n jobs=-1),
  "Decision Tree": GridSearchCV(DecisionTreeRegressor(random state=42),
                    param_grid={"max_depth": [5, 10, None]},
                    cv=3, n_jobs=-1),
  "Linear Regression": LinearRegression()
}
for name, model in tuned_models.items():
  if isinstance(model, GridSearchCV):
    model.fit(X_train, y_train)
    best model = model.best estimator
  else:
    best_model = model
    best model.fit(X train, y train)
  y pred = best model.predict(X test)
  score = r2_score(y_test, y_pred)
  results["Stage"].append("After Tuning")
  results["Model"].append(name)
  results["R2 Score"].append(score)
# Convert results to DataFrame
results df = pd.DataFrame(results)
print(results_df)
code
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