```
import pandas as pd
```

```
df = pd.read_excel('/content/car.xlsx')
df
```

```
₹
             brand model transmission age fuel
                                                        price mileage
                                                                           power seats
                                                  1 1231000.0
                                                                  19.01 4.496471
        0
                18
                      244
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        1
                10
                      263
                                           6
                                                 4
                                                     786000.0
                                                                  19.01 4.496471
                                                                                      5
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        2
                31
                      123
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                                                  1 1489000.0
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        3
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      32013
                31
                      247
                                      1
                                           2
                                                 4
                                                     392000.0
                                                                  19.01 4.496471
                                                                                      5
     32014 rows × 9 columns
X = df.drop('price',axis=1)
y = df['price']
from sklearn.model_selection import train_test_split
 X\_train, X\_test, y\_train, y\_test = train\_test\_split(X, y, test\_size=0.2, random\_state=True) 
{\tt from \ sklearn.tree \ import \ DecisionTreeRegressor}
from sklearn.model_selection import cross_val_score
from sklearn.metrics import mean_squared_error, r2_score
# Initialize the model with default params
dt_reg = DecisionTreeRegressor(random_state=42)
# Fit on training data
dt_reg.fit(X_train, y_train)
# Predict on train and test
y_train_pred = dt_reg.predict(X_train)
y_test_pred = dt_reg.predict(X_test)
# Evaluate
print('Train R2:', r2_score(y_train, y_train_pred))
print('Test R2:', r2_score(y_test, y_test_pred))
print('Train MSE:', mean_squared_error(y_train, y_train_pred))
print('Test MSE:', mean_squared_error(y_test, y_test_pred))
# Cross-validation (5-fold) on the entire dataset X, y
cv_scores = cross_val_score(dt_reg, X, y, cv=5, scoring='r2')
print('Cross-Validation R2 Scores:', cv scores)
print('Mean CV R2 Score:', cv_scores.mean())
→ Train R2: 0.9643323020743481
     Test R2: 0.8561125124893902
     Train MSE: 17986573269.737762
     Test MSE: 65420433593.05073
     Cross-Validation R2 Scores: [0.76931854 0.85220707 0.83311843 0.85433936 0.70270359]
     Mean CV R2 Score: 0.8023373965991863
from sklearn.tree import DecisionTreeRegressor
from sklearn.model_selection import GridSearchCV
from sklearn.metrics import mean_squared_error, r2_score
# Initialize base model
dt_reg = DecisionTreeRegressor(random_state=42)
# Define hyperparameter grid
param_grid = {
```

'criterion': ['squared_error', 'friedman_mse', 'absolute_error'],

'max_depth': range(1, 16),
'min_samples_split': [2, 5, 10],

```
'min_samples_leaf': [1, 2, 4]
# Setup GridSearch with 5-fold CV
grid_search = GridSearchCV(dt_reg, param_grid, cv=5, scoring='r2', n_jobs=-1)
# Fit on training data
grid_search.fit(X_train, y_train)
# Best estimator from grid search
best_dt_reg = grid_search.best_estimator_
print('Best Parameters:', grid_search.best_params_)
# Predict using the best model
y_train_pred = best_dt_reg.predict(X_train)
y_test_pred = best_dt_reg.predict(X_test)
# Evaluate
print('Train R2:', r2_score(y_train, y_train_pred))
print('Test R2:', r2_score(y_test, y_test_pred))
print('Train MSE:', mean_squared_error(y_train, y_train_pred))
print('Test MSE:', mean_squared_error(y_test, y_test_pred))
# Cross-validation on entire dataset with best model
cv_scores = cross_val_score(best_dt_reg, X, y, cv=5, scoring='r2')
print('Cross-Validation R2 Scores:', cv_scores)
print('Mean CV R2 Score:', cv_scores.mean())
Best Parameters: {'criterion': 'squared_error', 'max_depth': 15, 'min_samples_leaf': 4, 'min_samples_split': 2}
     Train R2: 0.9202757895790773
     Test R2: 0.8572464126246068
     Train MSE: 40203473605.0801
     Test MSE: 64904890234.96953
     Cross-Validation R2 Scores: [0.75314394 0.84872503 0.86234613 0.85558998 0.64100464]
     Mean CV R2 Score: 0.792161941708647
Start coding or generate with AI.
```

https://colab.research.google.com/drive/16BzbXFuqGaDOJZoMkF-kHNu5OClxpZ9n#printMode=true