```
import pandas as pd
```

```
df = pd.read_excel('/content/car.xlsx')
₹
             brand
                    model transmission age fuel
                                                         price mileage
                                                                            power seats
        0
                18
                      244
                                       1
                                            4
                                                   1 1231000 0
                                                                   19 01 4 496471
                                                                                        5
        1
                10
                      263
                                       1
                                            6
                                                  4
                                                      786000.0
                                                                   19.01 4.496471
                                                                                        5
                                            2
        2
                31
                      123
                                       1
                                                   1 1489000.0
                                                                   19.01 4.496471
                                                                                        5
        3
                 9
                       55
                                       0
                                            1
                                                  4 1227000.0
                                                                   19.01 4.496471
                                                                                        5
        4
                 8
                       82
                                       1
                                            3
                                                   1
                                                      887000.0
                                                                   19.01 4.496471
                                                                                        5
      32009
                 5
                       199
                                       1
                                            6
                                                  4
                                                      292000.0
                                                                   19.01 4.496471
      32010
                32
                      295
                                       1
                                            6
                                                  4
                                                      534000.0
                                                                   19.01 4.496471
                                                                                        5
      32011
                33
                       25
                                            8
                                                  4
                                                      424000.0
                                                                   19.01 4.496471
                                                                                        5
                                       1
      32012
                10
                       120
                                       0
                                            5
                                                  4
                                                      685000.0
                                                                   19.01 4.496471
                                                                                        5
      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) 
from \ sklearn.neighbors \ import \ KNeighborsRegressor
from sklearn.metrics import mean_squared_error, r2_score
from sklearn.model_selection import cross_val_score
# Initialize model
knn = KNeighborsRegressor()
# Train the model
knn.fit(X_train, y_train)
# Predict
y_train_pred = knn.predict(X_train)
y_{test_pred} = knn.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 folds)
cv_scores = cross_val_score(knn, X, y, cv=5, scoring='r2')
print("Cross-validation R2 scores:", cv scores)
print("Mean CV R2 score:", cv_scores.mean())
→ Train R2: 0.8959997726812934
     Test R2: 0.8392815334132199
     Train MSE: 52445428707.97314
     Test MSE: 73072870702.13182
     Cross-validation R2 scores: [0.68830242 0.82024626 0.83714765 0.82468885 0.69242185]
     Mean CV R2 score: 0.7725614062124903
# Hyperparameter grid to search
param\_grid = \{
    'n_neighbors': [3, 5, 7, 9, 11],
    'weights': ['uniform', 'distance'],
    'metric': ['euclidean', 'manhattan']
```

```
{\tt from \ sklearn.model\_selection \ import \ GridSearchCV}
from sklearn.neighbors import KNeighborsRegressor
# Define model
knn = KNeighborsRegressor()
```

```
https://colab.research.google.com/drive/10ie5d0eTpwj7Goz624gx1pmrRdjVsZeD#printMode=true
```

```
# GridSearch with 5-fold cross-validation
grid_search = GridSearchCV(knn, param_grid, cv=5, scoring='r2', n_jobs=-1)
# Fit on training data
grid\_search.fit(X\_train, y\_train)
# Best model after tuning
best_knn = grid_search.best_estimator_
print("Best hyperparameters:", grid_search.best_params_)
# Predict and evaluate
y_train_pred = best_knn.predict(X_train)
y_test_pred = best_knn.predict(X_test)
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 scores with best model
cv_scores = cross_val_score(best_knn, X, y, cv=5, scoring='r2')
print("Cross-validation R2 scores:", cv_scores)
print("Mean CV R2 score:", cv_scores.mean())
Best hyperparameters: {'metric': 'manhattan', 'n_neighbors': 5, 'weights': 'distance'}
     Train R2: 0.9621322209006553
     Test R2: 0.8631164284201553
     Train MSE: 19096034309.597443
     Test MSE: 62236006475.95242
     Cross-validation R2 scores: [0.78612416 0.84699689 0.83975849 0.8644433 0.71163591]
     Mean CV R2 score: 0.8097917487900348
Start coding or \underline{\text{generate}} with AI.
```

https://colab.research.google.com/drive/1Oie5d0eTpwj7Goz624gx1pmrRdjVsZeD#printMode=true