

Untitled28.ipynb - Colab

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```
import pandas as pd
import Kagglehub
import os

# Ensure the dataset is downloaded and get its path
path = Kagglehub.dataset_download("afnansifafnan/electric-car-performance-and-battery-dataset")

# Construct the full file path
file_name = os.listdir(path)[0] # Assuming there's only one CSV file in the directory
full_file_path = os.path.join(path, file_name)

# Read the CSV file
data = pd.read_csv(full_file_path)
data.info()

... Downloading from https://www.kaggle.com/api/v1/datasets/download/afnansifafnan/electric-car-performance-and-battery-dataset?dataset_version_number=1...
100% [██████████] 16.2k/16.2k [00:00:00.00, 15.2MB/s]Extracting files...
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 478 entries, 0 to 477
Data columns (total 22 columns):
 #   Column           Non-Null Count  Dtype  
 --- 
 0   brand            478 non-null    object  
 1   model             477 non-null    object  
 2   top_speed_kmh    478 non-null    int64  
 3   battery_capacity_kWh 478 non-null    float64 
 4   battery_type     478 non-null    object  
 5   number_of_cells  276 non-null    float64 
 6   torque_nm        471 non-null    float64 
 7   efficiency_wh_per_km 478 non-null    int64  
 8   weight_kg         478 non-null    int64  
 9   acceleration_0_100_s 478 non-null    float64 
 10  fast_charging_power_kw_dc 477 non-null    float64 
 11  fast_charge_port 477 non-null    object  
 12  towing_capacity_kg 452 non-null    float64 
 13  cargo_volume_l   477 non-null    object  
 14  seats             478 non-null    int64  
 15  drivetrain        478 non-null    object  
 16  segment            478 non-null    object  
 17  length_mm          478 non-null    int64  
 18  width_mm           478 non-null    int64  
 19  height_mm          478 non-null    int64  
 20  car_body_type      478 non-null    object  
 21  source_url         478 non-null    object  
dtypes: float64(6), int64(7), object(9)
memory usage: 82.3+ KB
```

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```
x = data.drop('range_km', axis = 1)
y = data['range_km']
```

	brand	model	top_speed_kmh	battery_capacity_kWh	battery_type	number_of_cells	torque_nm	efficiency_wh_per_km	acceleration_0_100_s	fast_charging_power_kw_dc	towing_capacity_kg	cargo_volume_l	seats	drivetrain	segment
0	Abarth	500e Convertible	155	37.8	Lithium-ion	192.0	235.0	166	7.0	67.0	...	0.0	185	4	FWD B-Compact
1	Abarth	500e Hatchback	155	37.8	Lithium-ion	192.0	235.0	149	7.0	67.0	...	0.0	185	4	FWD B-Compact
2	Abarth	600e Scorpionissima	200	50.8	Lithium-ion	102.0	345.0	168	5.9	79.0	...	0.0	360	5	FWD JB-Compact
3	Abarth	600e Turismo	200	50.8	Lithium-ion	102.0	345.0	168	6.2	79.0	...	0.0	360	5	FWD JB-Compact
4	Always	U5	150	60.0	Lithium-ion	NaN	310.0	156	7.5	78.0	...	NaN	496	5	FWD JC-Medium
...	...	...	...	...	...	...	...	...	...	...	...	...	...	...	...
473	Zeekr	X Premium RWD	210	71.0	Lithium-ion	NaN	448.0	148	6.0	240.0	...	2000.0	539	5	RWD JD-Large
474	Zeekr	X Core RWD (MY25)	190	49.0	Lithium-ion	NaN	343.0	148	5.9	70.0	...	1600.0	362	5	RWD JB-Compact
475	Zeekr	X Long Range RWD (MY25)	190	65.0	Lithium-ion	NaN	343.0	146	5.6	114.0	...	1600.0	362	5	RWD JB-Compact
476	Zeekr	X Privilege AWD (MY25)	190	65.0	Lithium-ion	NaN	543.0	153	3.8	114.0	...	1600.0	362	5	AWD JB-Compact
477	Firefly	NaN	150	41.2	Lithium-ion	112.0	200.0	125	8.1	65.0	...	0.0	404	5	RWD B-Compact

Variables Terminal

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```
data.columns  
Index(['brand', 'model', 'top_speed_kmh', 'battery_capacity_kwh',  
       'battery_type', 'number_of_cells', 'torque_nm', 'efficiency_wh_per_kw',  
       'range_km', 'acceleration_0_100_s', 'fast_charging_power_kw_de',  
       'fast_charge_port', 'towing_capacity_kg', 'cargo_volume_l', 'seats',  
       'drivetrain', 'segment', 'length_mm', 'width_mm', 'height_mm',  
       'car_body_type', 'source_url'],  
      dtype='object')
```

```
import seaborn as sns  
import matplotlib.pyplot as plt
```

```
data = data[['range_km', 'seats']]
```

```
sns.heatmap(data.corr(numeric_only=True), annot = True)  
plt.show()
```

The heatmap displays the correlation coefficient between two variables: range\_km and seats. The diagonal elements are 1.0, indicating perfect correlation with themselves. The off-diagonal element shows a correlation of approximately -0.25. A color scale bar on the right indicates the correlation strength from -0.2 (dark blue) to 1.0 (dark red).

```
x = data[['top speed kmh', 'battery capacity kwh', 'torque nm']]
```

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```
x = data[['top_speed_kmh','battery_capacity_kWh','torque_nm']]
y = data[['range_km']]
```

x

	top_speed_kmh	battery_capacity_kWh	torque_nm
0	155	37.8	235.0
1	155	37.8	235.0
2	200	50.8	345.0
3	200	50.8	345.0
4	150	60.0	310.0
...	...	...	...
473	210	71.0	440.0
474	190	49.0	343.0
475	190	65.0	343.0
476	190	65.0	543.0
477	150	41.2	200.0

478 rows × 3 columns

```
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 = 42)
```

```
from sklearn.linear_model import LinearRegression
model = LinearRegression()
```

```
# Impute missing values in x_train with the mean of each column
x_train_imputed = x_train.fillna(x_train.mean())

# Train the model with the imputed data
model.fit(x_train_imputed, y_train)
```

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```
V_pred = model.predict(x_test_imputed)
array([[478.89714632],
[487.85798168],
[325.63986361],
[471.67746454],
[394.67969666],
[295.40386121],
[384.66462088],
[349.59692558],
[321.13432154],
[306.83179494],
[518.66933877],
[295.46396121],
[244.94141529],
[484.83115379],
[263.35082974],
[366.53252453],
[488.24432973],
[240.87377056],
[263.35082974],
[498.66334473],
[224.70927649],
[244.27839524],
[466.46230351],
[388.63801459],
[225.75821749],
[507.75857377],
[497.63460127],
[285.1320194 ],
[368.87976793],
[482.40808961],
[452.81298618],
[274.108086371],
[308.74996685],
[301.734142 ],
[349.24536797],
[502.17702563],
[380.57765803],
[394.67968666],
[467.99885764],
[324.31448892],
[379.42687337],
[450.8784849 ],
[358.176000221],
[478.65775986],
[394.67968666],
[371.36084945],
[501.78846181])
```

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```
[464.2419471],  
[380.0681319],  
[467.11528558],  
[480.17556325],  
[394.67968666],  
[380.57765880],  
[418.28731651],  
[408.35935587],  
[364.66395777],  
[448.68332815],  
[522.55376399],  
[334.56684119],  
[233.78422674],  
[340.66395777],  
[418.28731651],  
[438.85801021],  
[292.63184987],  
[524.3739515 ],  
[407.69885764],  
[437.59852082],  
[244.27839524],  
[334.56684119],  
[261.68414718]]
```

```
from sklearn.metrics import mean_squared_error  
mean_squared_error(y_test,Y_pred)
```

```
NameError: name 'Y_pred' is not defined.
```

```
from sklearn.metrics import mean_squared_error  
mean_squared_error(y_test, model.predict(x_test_imputed))
```

```
1903.0451228576375
```

```
from sklearn.metrics import r2_score  
r2_score(y_test, model.predict(x_test_imputed))
```

```
... 0.8125221961879764
```