

```
import pandas as pd
import matplotlib.pyplot as plt
import seaborn as sns
from statsmodels.graphics.regressionplots import influence_plot
import statsmodels.formula.api as smf
import numpy as np
```

```
import pandas as pd
```

```
#step2
```

```
from google.colab import files
uploaded = files.upload()
```

```
#step3
```

```
file_name = list(uploaded.keys())[0]
cars = pd.read_csv(file_name)
```

```
<IPython.core.display.HTML object>
```

```
Saving Cars.csv to Cars.csv
```

```
cars.shape
```

```
(81, 5)
```

```
isna = cars.isna().sum()
isna
```

```
HP      0
MPG      0
VOL      0
SP      0
WT      0
dtype: int64
```

```
#correlation
```

```
cars.corr()
```

```
{"summary":{"\n  \"name\": \"cars\",\n  \"rows\": 5,\n  \"fields\": [\n    {\n      \"column\": \"HP\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 0.7232487955377062,\n        \"min\": -0.7250383497637914,\n        \"max\": 1.0,\n        \"num_unique_values\": 5,\n        \"samples\": [\n          0.7250383497637914,\n          0.07651306534492211,\n          0.07745947360036072\n        ],\n        \"semantic_type\": \"\",\n        \"description\": \"\"\n      },\n      \"column\": \"MPG\",\n      \"properties\": {\n        \"dtype\": \"number\",\n        \"std\": 0.7287325628832262,\n        \"min\": -0.7250383497637914,\n        \"max\": 1.0,\n        \"num_unique_values\": 5,\n        \"samples\": [\n          1.0,\n          -0.5267590900278917,\n          0.5290565802560923\n        ],\n        \"semantic_type\": \"\",
```

```

"description\": {\n      }\n    },\n    {\n      \"column\":  

\"VOL\", \n      \"properties\": {\n        \"dtype\": \"number\", \n  

\"std\": 0.6615123170839469, \n        \"min\": -0.5290565802560923, \n  

\"max\": 1.0, \n        \"num_unique_values\": 5, \n        \"samples\":  

[\n          -0.5290565802560923, \n          0.999203080186856, \n  

1.0 \n        ], \n        \"semantic_type\": \"\", \n  

\"description\": {\n      }\n    },\n    {\n      \"column\":  

\"SP\", \n      \"properties\": {\n        \"dtype\": \"number\", \n  

\"std\": 0.7065122396814414, \n        \"min\": -0.6871246127261932, \n  

\"max\": 1.0, \n        \"num_unique_values\": 5, \n        \"samples\":  

[\n          -0.6871246127261932, \n          0.10243919098077894, \n  

0.1021700095142569 \n        ], \n        \"semantic_type\": \"\", \n  

\"description\": {\n      }\n    },\n    {\n      \"column\":  

\"WT\", \n      \"properties\": {\n        \"dtype\": \"number\", \n  

\"std\": 0.6608343322683942, \n        \"min\": -0.5267590900278917, \n  

\"max\": 1.0, \n        \"num_unique_values\": 5, \n        \"samples\":  

[\n          -0.5267590900278917, \n          1.0, \n  

0.999203080186856 \n        ], \n        \"semantic_type\": \"\", \n  

\"description\": {\n      }\n    }\n  ],\n  \"type\": \"dataframe\"}

```

#dataframes

```

cars_new = cars.iloc[:,1:]
cars_new.head()

```

```

{"summary": {\n  \"name\": \"cars_new\", \n  \"rows\": 81, \n  

\"fields\": [\n    {\n      \"column\": \"MPG\", \n  

\"properties\": {\n        \"dtype\": \"number\", \n        \"std\":  

9.131444731795982, \n        \"min\": 12.10126289, \n        \"max\":  

53.70068138, \n        \"num_unique_values\": 50, \n        \"samples\":  

[\n          38.31060597, \n          23.10317168, \n  

27.85625194 \n        ], \n        \"semantic_type\": \"\", \n  

\"description\": {\n      }\n    },\n    {\n      \"column\":  

\"VOL\", \n      \"properties\": {\n        \"dtype\": \"number\", \n  

\"std\": 22, \n        \"min\": 50, \n        \"max\": 160, \n  

\"num_unique_values\": 34, \n        \"samples\": [\n          98, \n  

102, \n          127 \n        ], \n        \"semantic_type\": \"\", \n  

\"description\": {\n      }\n    },\n    {\n      \"column\":  

\"SP\", \n      \"properties\": {\n        \"dtype\": \"number\", \n  

\"std\": 14.18143157452861, \n        \"min\": 99.56490661, \n  

\"max\": 169.5985128, \n        \"num_unique_values\": 68, \n  

\"samples\": [\n          115.5765794, \n          113.8291446, \n  

113.1853528 \n        ], \n        \"semantic_type\": \"\", \n  

\"description\": {\n      }\n    },\n    {\n      \"column\":  

\"WT\", \n      \"properties\": {\n        \"dtype\": \"number\", \n  

\"std\": 7.492812997393198, \n        \"min\": 15.71285853, \n  

\"max\": 52.99775236, \n        \"num_unique_values\": 81, \n  

\"samples\": [\n          37.04235003, \n          28.7620589, \n  

32.6758277 \n        ], \n        \"semantic_type\": \"\", \n  

\"description\": {\n      }\n    }\n  ],\n  \"type\": \"dataframe\", \"variable_name\": \"cars_new\"}

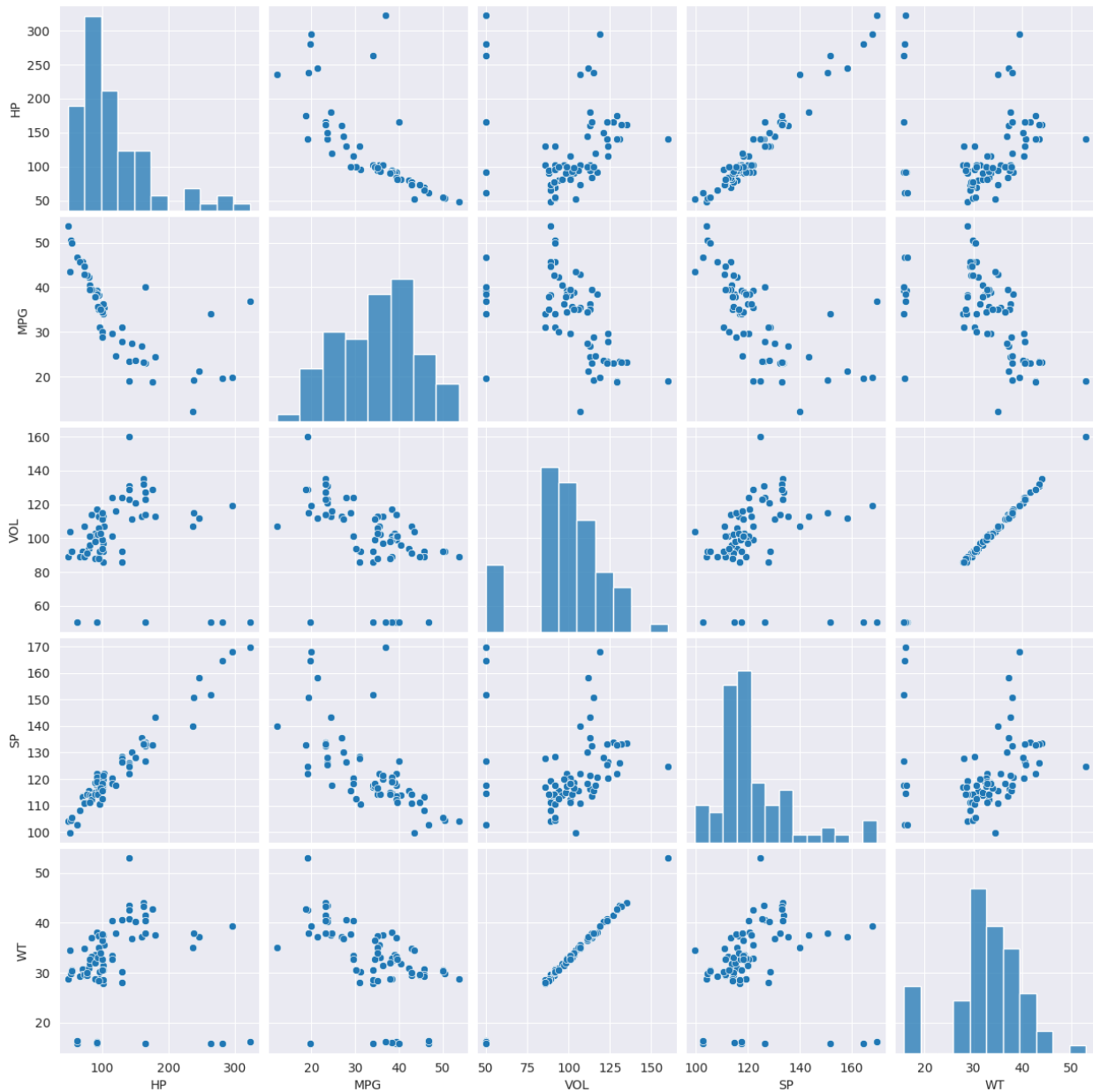
```

```
cars.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 81 entries, 0 to 80
Data columns (total 5 columns):
#   Column  Non-Null Count  Dtype  
---  -
0    HP      81 non-null     int64   
1    MPG      81 non-null     float64  
2    VOL      81 non-null     int64   
3    SP       81 non-null     float64  
4    WT       81 non-null     float64  
dtypes: float64(3), int64(2)
memory usage: 3.3 KB

sns.set_style(style='darkgrid')
sns.pairplot(cars)

<seaborn.axisgrid.PairGrid at 0x7f842a0f5350>
```



```
import statsmodels.formula.api as smf
model = smf.ols('MPG~WT+VOL+SP+HP',data=cars).fit()

model.params

Intercept    30.677336
WT           0.400574
VOL          -0.336051
SP           0.395627
HP           -0.205444
dtype: float64
```

