

```
import seaborn as sns
```

```
# "mpg" veri setini yükle
df = sns.load_dataset("mpg")
```

✓ Min max gibi degerleri tek tek yazdim. aksi durumda sadece en son verdigin komutu uyguluyor

Elimizde categorical variable'lar da oldugu icin her zaman fonksiyon halini kullanamayiz.

```
#1
```

```
df.info , df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 398 entries, 0 to 397
Data columns (total 9 columns):
#   Column      Non-Null Count  Dtype
---  ---
0   mpg         398 non-null    float64
1   cylinders   398 non-null    int64
2   displacement 398 non-null    float64
3   horsepower   392 non-null    float64
4   weight       398 non-null    int64
5   acceleration 398 non-null    float64
6   model_year   398 non-null    int64
7   origin       398 non-null    object
8   name         398 non-null    object
dtypes: float64(4), int64(3), object(2)
memory usage: 28.1+ KB
(<bound method DataFrame.info of      mpg cylinders displacement horsepower weight acceleration \
0   18.0      8      307.0    130.0   3504      12.0
1   15.0      8      350.0    165.0   3693      11.5
2   18.0      8      318.0    150.0   3436      11.0
3   16.0      8      304.0    150.0   3433      12.0
4   17.0      8      302.0    140.0   3449      10.5
..  ...  ...
393 27.0      4      140.0     86.0   2790      15.6
394 44.0      4      97.0     52.0   2130      24.6
395 32.0      4     135.0     84.0   2295      11.6
396 28.0      4     120.0     79.0   2625      18.6
397 31.0      4     119.0     82.0   2720      19.4

      model_year origin      name
0         70    usa  chevrolet chevelle malibu
1         70    usa    buick skylark 320
2         70    usa  plymouth satellite
3         70    usa    amc rebel sst
4         70    usa    ford torino
..  ...  ...
393        82    usa    ford mustang gl
394        82  europe      vw pickup
395        82    usa    dodge rampage
396        82    usa    ford ranger
397        82    usa    chevy s-10

[398 rows x 9 columns]>,
None)
```

```
#2
```

```
df.head(10)
```

| | mpg | cylinders | displacement | horsepower | weight | acceleration | model_year | origin | name |
|---|------|-----------|--------------|------------|--------|--------------|------------|--------|---------------------------|
| 0 | 18.0 | 8 | 307.0 | 130.0 | 3504 | 12.0 | 70 | usa | chevrolet chevelle malibu |

#3.1
df.max()

```
mpg          46.6
cylinders      8
displacement  455.0
horsepower    230.0
weight       5140
acceleration   24.8
model_year    82
origin        usa
name          vw rabbit custom
dtype: object
```

#3.2
df.min()

```
mpg          9.0
cylinders      3
displacement  68.0
horsepower    46.0
weight      1613
acceleration   8.0
model_year    70
origin        europe
name          amc ambassador brougham
dtype: object
```

#3.3
df.var

```
<bound method NDFrame._add_numeric_operations.<locals>.<var of      mpg cylinders displacement horsepower weight acceleration \
0  18.0      8      307.0      130.0   3504      12.0
1  15.0      8      350.0      165.0  3693      11.5
2  18.0      8      318.0      150.0  3436      11.0
3  16.0      8      304.0      150.0  3433      12.0
4  17.0      8      302.0      140.0  3449      10.5
..  ...  ...
393 27.0      4      140.0      86.0  2790      15.6
394 44.0      4      97.0      52.0  2130      24.6
395 32.0      4      135.0      84.0  2295      11.6
396 28.0      4      120.0      79.0  2625      18.6
397 31.0      4      119.0      82.0  2720      19.4

      model_year origin      name
0         70  usa  chevrolet chevelle malibu
1         70  usa    buick skylark 320
2         70  usa  plymouth satellite
3         70  usa    amc rebel sst
4         70  usa    ford torino
..  ...  ...
393        82  usa    ford mustang gl
394        82 europe    vw pickup
395        82  usa    dodge rampage
396        82  usa    ford ranger
397        82  usa    chevy s-10

[398 rows x 9 columns]>
```

```
df["horsepower"].df.weight

(0  130.0
1  165.0
2  150.0
3  150.0
4  140.0
...
393  86.0
394  52.0
395  84.0
396  79.0
397  82.0
Name: horsepower, Length: 398, dtype: float64,
0  3504
1  3693
```

```

2    3436
3    3433
4    3449
...
393   2790
394   2130
395   2295
396   2625
397   2720
Name: weight, Length: 398, dtype: int64)

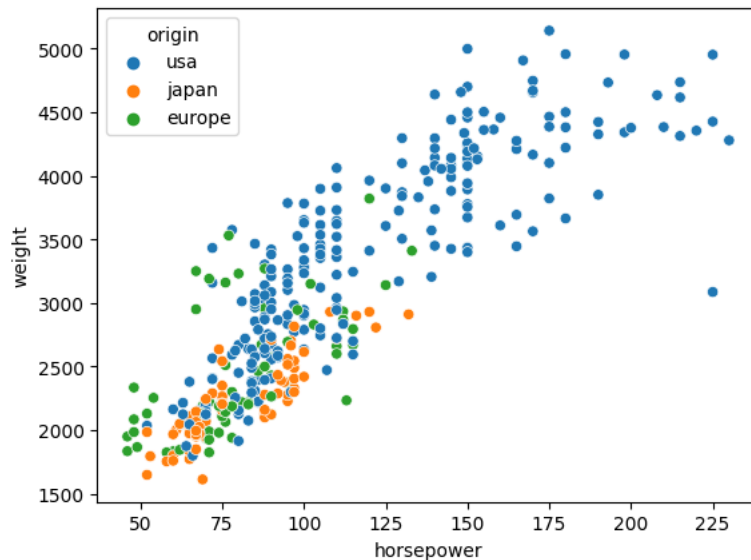
```

```

import seaborn as sns
sns.scatterplot(x="horsepower",y="weight",hue="origin",data=df)

```

<Axes: xlabel='horsepower', ylabel='weight'>



```

df_corr=df["mpg"].corr(df.acceleration)
print(df_corr)

```

0.42028891210165065

```

#17
df.isnull().sum()

```

```

mpg      0
cylinders 0
displacement 0
horsepower 6
weight    0
acceleration 0
model_year 0
origin    0
name      0
dtype: int64

```

```

#17
import pandas as pd
X_horse=df.iloc[:,3:4]
X_weight=df.iloc[:,4:5]
X_acc=df.iloc[:, 5:6]
y=df.iloc[:, :1]

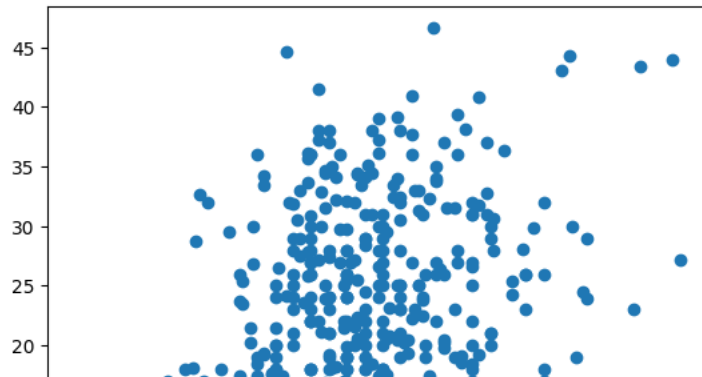
```

```

#18
import matplotlib.pyplot as plt
plt.scatter(X_acc,y)

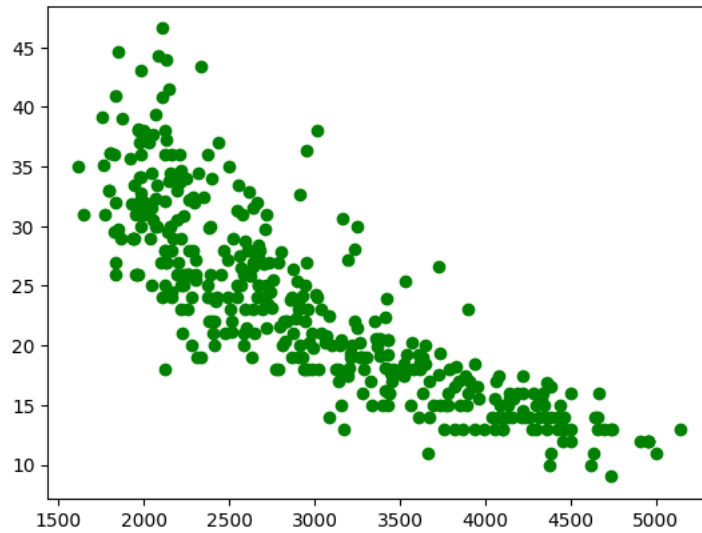
```

<matplotlib.collections.PathCollection at 0x7b04b990cb20>



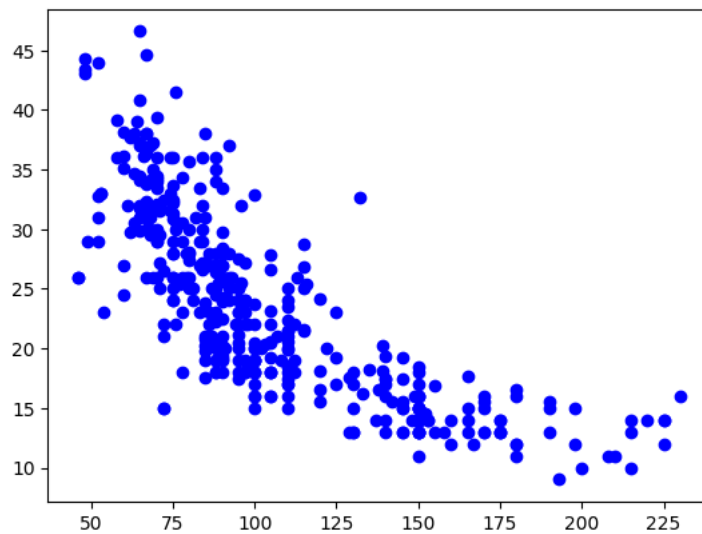
```
#18
import matplotlib.pyplot as plt
plt.scatter(X_weight,y,c="g")
```

<matplotlib.collections.PathCollection at 0x7b04b9826650>



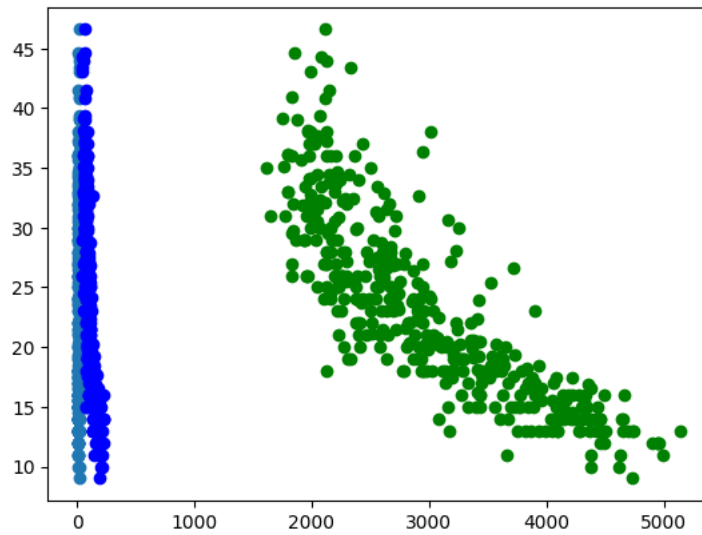
```
#18
import matplotlib.pyplot as plt
plt.scatter(X_horse,y,c="b")
```

<matplotlib.collections.PathCollection at 0x7b04b98a90f0>



```
#18
import matplotlib.pyplot as plt
plt.scatter(X_acc,y)
plt.scatter(X_horse,y,c="b")
plt.scatter(X_weight,y,c="g")
```

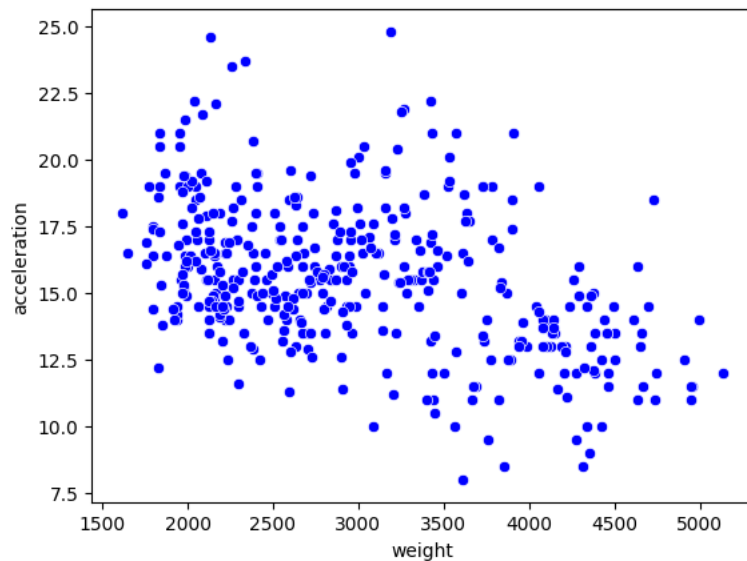
<matplotlib.collections.PathCollection at 0x7b04b973d450>



```
#18
from sklearn.linear_model import LinearRegression
from sklearn.metrics import r2_score, mean_absolute_error, mean_squared_error
from sklearn.preprocessing import StandardScaler
from sklearn.metrics.pairwise import cosine_similarity
mpg=df.mpg
```

```
#18
import pandas as pd
import numpy as np
import seaborn as sns
korelasyon=np.corrcoef(df.weight,df.acceleration)
sns.scatterplot(x="weight",y="acceleration",data=df,c="b")
```

<Axes: xlabel='weight', ylabel='acceleration'>



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

```
#20
import pandas as pd
from sklearn.linear_model import LinearRegression
from sklearn.model_selection import train_test_split
from sklearn.metrics import mean_squared_error
temiz_df=df.dropna(inplace=True)
X = df[['horsepower', 'acceleration', 'weight']]
y = df['mpg']
lr.fit(X,y)
temiz_df=df.dropna(inplace=True)
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_state=42)
```

```
#19
import pandas as pd
df_corr=df[['mpg']].corr(df.acceleration)
print(df_corr)
```

```
0.4233285369027874
```

```
#20
#MPG BIZIM Y YNAI TARGET COLONUMUZ DIGER DEGISKENLER ISE X COLONUMUZDUR.
from sklearn.linear_model import LinearRegression
lr = LinearRegression()
lr.fit(X_train,y_train)
target_tahmin=lr.predict(X_test)
mse = mean_squared_error(y_test, target_tahmin)
```

```
#20
import pandas as pd
from sklearn.linear_model import LinearRegression
intercept=lr.intercept_
coef=lr.coef_
```

```
#20
print(coef),print(intercept),print(lr.score(X_train,y_train))
```

```
[-0.05492359 -0.02881278 -0.00580125]
47.115174659445685
0.7150226018087609
(None, None, None)
```

```
eldeki_veri=[[130,13,3500]]
eldeki_df=pd.DataFrame(eldeki_veri)
eldeki_df.values.reshape(-1,1)
tahmini_mpg=lr.predict(eldeki_df)
##Tahmin yapti fakat gorsellestirmede eklentisiz kaldi
```

```
/usr/local/lib/python3.10/dist-packages/sklearn/base.py:439: UserWarning: X does not have valid feature names, but LinearRegression was fitted with feature names
warnings.warn(
```

```
#23
import pandas as pd
from sklearn.model_selection import train_test_split
from sklearn.ensemble import RandomForestClassifier
from sklearn.metrics import accuracy_score
```

```
veri = ['mpg', 'displacement', 'horsepower', 'weight', 'acceleration']
X = df[veri]
y = df['origin']
```

```
#23
from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.8, random_state=42)
```

```
#23
from sklearn.ensemble import RandomForestClassifier
rf=RandomForestClassifier(random_state=42)
```

```
#23
rf.fit(X_train,y_train)
```

```
RandomForestClassifier
RandomForestClassifier(random_state=42)
```

#24

```
y_pred=rf.predict(X_test)
dogruluk=accuracy_score(y_test, y_pred)
print(dogruluk)
```

```
0.7579617834394905
```

#25

```
onem_sirasi=rf.feature_importances_
print(onem_sirasi)
```

```
[0.18548811 0.3211235 0.15512013 0.18508231 0.15318596]
```