rk1

April 2, 2023

1 1

1.1

[ ]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import scipy.stats as stats
from mlxtend.feature_selection import ExhaustiveFeatureSelector as EFS
from sklearn.neighbors import KNeighborsClassifier
from sklearn.datasets import load_wine
```

1.2 №13

( ) “ - 1 / X”.

1.2.1

[ ]:
```
def diagnostic_plots(df, variable):
    plt.figure(figsize=(15,6))
    #
    plt.subplot(1, 2, 1)
    df[variable].hist(bins=30)
    ## Q-Q plot
    plt.subplot(1, 2, 2)
    stats.probplot(df[variable], dist="norm", plot=plt)
    plt.show()
```

[ ]:
```
#
data = pd.read_csv('cars.csv', sep=",")

data.head()
```

[ ]:

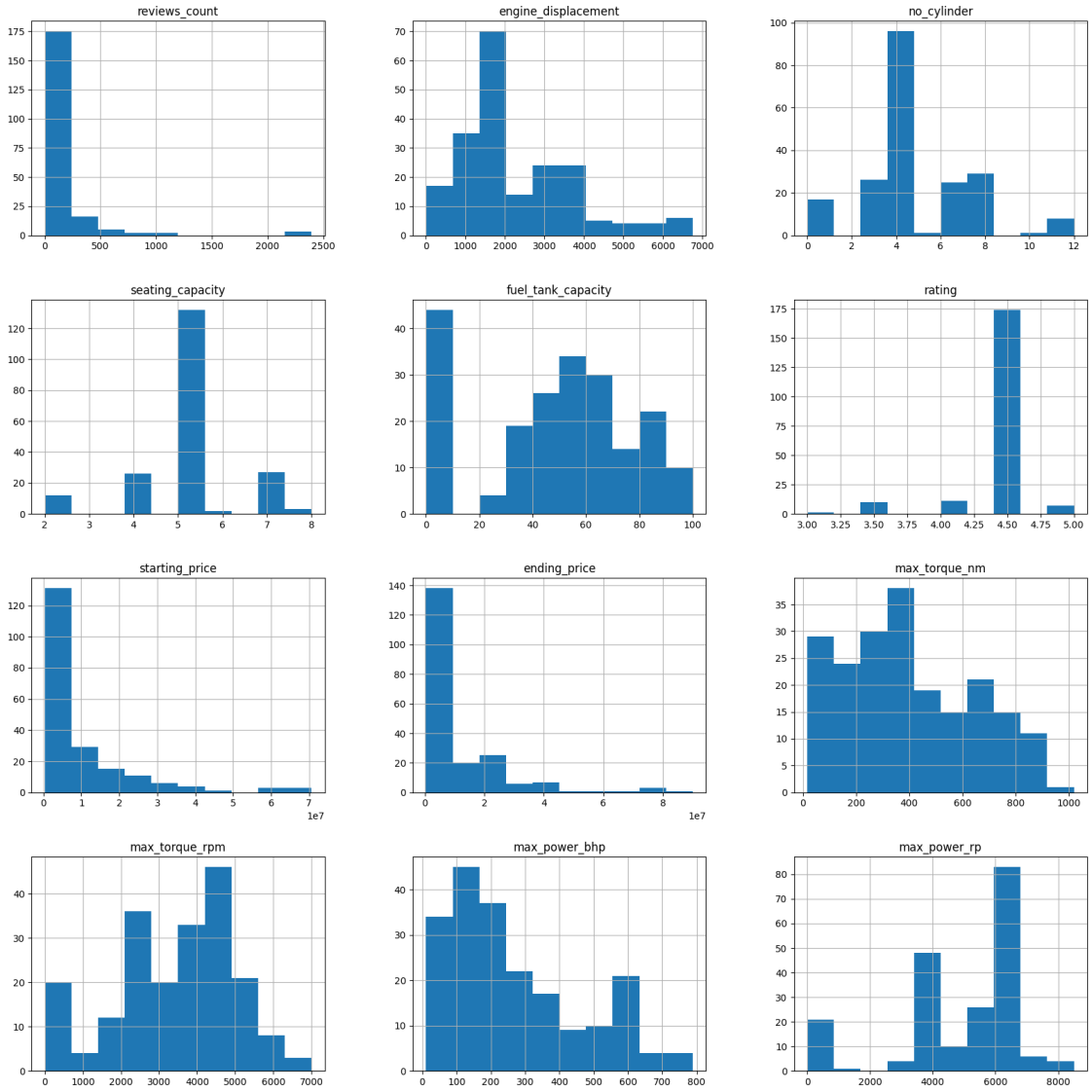
	car_name	reviews_count	fuel_type	engine_displacement	\
0	Maruti Alto K10	51	Petrol	998	
1	Maruti Brezza	86	Petrol	1462	
2	Mahindra Thar	242	Diesel	2184	
3	Mahindra XUV700	313	Diesel	2198	
4	Mahindra Scorpio-N	107	Diesel	2198	

	no_cylinder	seating_capacity	transmission_type	fuel_tank_capacity	\
0	3	5.0	Automatic	27.0	
1	4	5.0	Automatic	48.0	
2	4	4.0	Automatic	57.0	
3	4	7.0	Automatic	60.0	
4	4	7.0	Automatic	57.0	

	body_type	rating	starting_price	ending_price	max_torque_nm	\
0	Hatchback	4.5	399000	583000	89.0	
1	SUV	4.5	799000	1396000	136.8	
2	SUV	4.5	1353000	1603000	300.0	
3	SUV	4.5	1318000	2458000	450.0	
4	SUV	4.5	1199000	2390000	400.0	

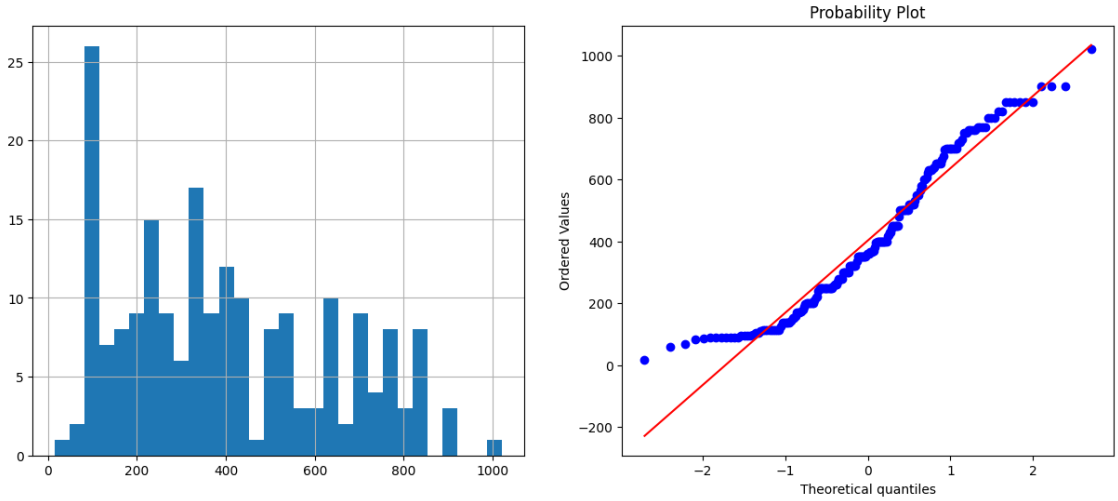
	max_torque_rpm	max_power_bhp	max_power_rp
0	3500	65.71	5500
1	4400	101.65	6000
2	2800	130.00	3750
3	2800	182.38	3500
4	2750	172.45	3500

[ ]:
```
data.hist(figsize=(20,20))
plt.show()
```



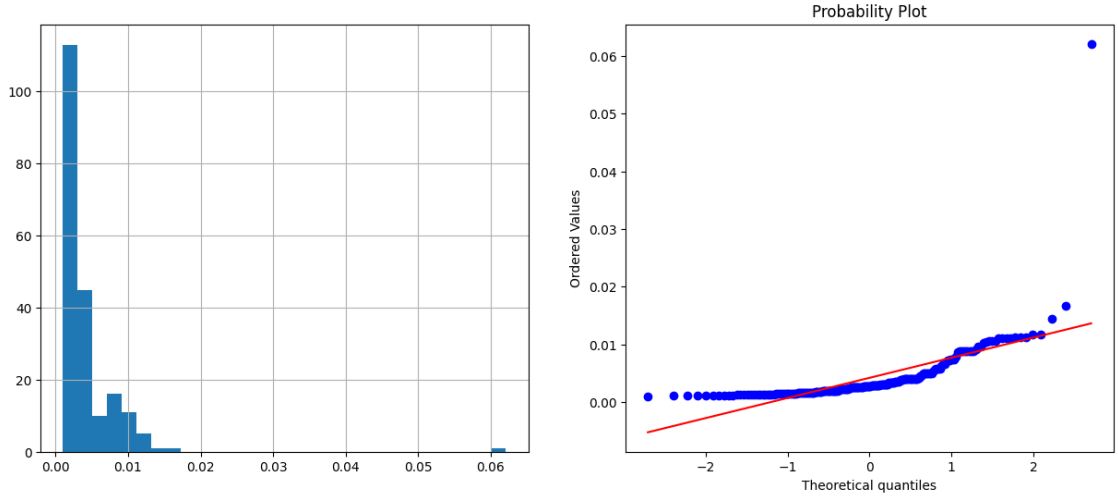
1.2.2

```
[ ]: diagnostic_plots(data, 'max_torque_nm')
```



1.2.3

```
[ ]: data['max_torque_nm_reciprocal'] = 1 / (data['max_torque_nm'])
diagnostic_plots(data, 'max_torque_nm_reciprocal')
```



1.3 №33

(feature selection). (wrapper method), (exhaustive feature selection).

1.3.1

[ ]:
```
wine = load_wine()
wine_X = wine.data
wine_y = wine.target
wine_feature_names = wine['feature_names']
wine_x_df = pd.DataFrame(data=wine['data'], columns=wine['feature_names'])
```

1.3.2 (wrapper methods)

[ ]:
```
knn = KNeighborsClassifier(n_neighbors=3)
```

[ ]:
```
efs1 = EFS(knn,
           min_features=2,
           max_features=4,
           scoring='accuracy',
           print_progress=True,
           cv=5)

efs1 = efs1.fit(wine_X, wine_y)
# efs1 = efs1.fit(iris_X, iris_y, custom_feature_names=iris_feature_names)

print('Best accuracy score: %.2f' % efs1.best_score_)
print('Best subset (indices):', efs1.best_idx_)
print('Best subset (corresponding names):', efs1.best_feature_names_)
```

```
Features: 1079/1079

Best accuracy score: 0.94
Best subset (indices): (0, 5, 6, 9)
Best subset (corresponding names): ('0', '5', '6', '9')
```

[ ]:
```
efs2 = EFS(knn,
           min_features=1,
           max_features=2,
           scoring='accuracy',
           print_progress=True,
           cv=5)

efs2 = efs2.fit(wine_X, wine_y)
# efs2 = efs2.fit(iris_X, iris_y, custom_feature_names=iris_feature_names)

print('Best accuracy score: %.2f' % efs2.best_score_)
print('Best subset (indices):', efs2.best_idx_)
print('Best subset (corresponding names):', efs2.best_feature_names_)
```

```
Features: 91/91

Best accuracy score: 0.93
Best subset (indices): (6, 9)
```

```
Best subset (corresponding names): ('6', '9')
```

1.3.3

5-22 , 5 -22 - .

[ ]:
```
data.hist('fuel_tank_capacity')
plt.show()
```