

```

# Latihan 1
# import library pandas
import pandas as pd

# Import library numpy
import numpy as np

# Import library matplotlib dan seaborn untuk visualisasi
import matplotlib.pyplot as plt
import seaborn as sns

# Import Module LinearRegression digunakan untuk memanggil algoritma
Linear Regression.
from sklearn.linear_model import LinearRegression

# import Module train_test_split digunakan untuk membagi data kita
menjadi training dan testing set.
from sklearn.model_selection import train_test_split

# import modul mean_absolute_error dari library sklearn
from sklearn.metrics import mean_absolute_error

#import math agar program dapat menggunakan semua fungsi yang ada pada
modul math.(ex:sqrt)
import math

# me-non aktifkan peringatan pada python
import warnings
warnings.filterwarnings('ignore')

#Panggil file (load file bernama CarPrice_Assignment.csv) dan simpan
dalam dataframe Lalu tampilkan 10 baris awal dataset dengan function
head()
data =pd.read_csv("CarPrice_Assignment.csv")
dataset = pd.DataFrame(data)
dataset.head(10)

```

	car_ID	symboling	CarName	...	citympg	highwaympg
price						
0	1	3	alfa-romero giulia	...	21	27
13495.000						
1	2	3	alfa-romero stelvio	...	21	27
16500.000						
2	3	1	alfa-romero Quadrifoglio	...	19	26
16500.000						
3	4	2	audi 100 ls	...	24	30
13950.000						
4	5	2	audi 100ls	...	18	22
17450.000						
5	6	2	audi fox	...	19	25

15250.000					
6	7	1	audi 100ls	...	19 25
17710.000					
7	8	1	audi 5000	...	19 25
18920.000					
8	9	1	audi 4000	...	17 20
23875.000					
9	10	0	audi 5000s (diesel)	...	16 22
17859.167					

[10 rows x 26 columns]

Latihan 2

melihat jumlah baris dan jumlah kolom (bentuk data) pada data df dengan fungsi .shape

dataset.shape

(205, 26)

Melihat Informasi lebih detail mengenai struktur DataFrame dapat dilihat menggunakan fungsi info()

dataset.info()

<class 'pandas.core.frame.DataFrame'>

RangeIndex: 205 entries, 0 to 204

Data columns (total 26 columns):

#	Column	Non-Null Count	Dtype
0	car_ID	205 non-null	int64
1	symboling	205 non-null	int64
2	CarName	205 non-null	object
3	fueltype	205 non-null	object
4	aspiration	205 non-null	object
5	doornumber	205 non-null	object
6	carbody	205 non-null	object
7	drivewheel	205 non-null	object
8	enginelocation	205 non-null	object
9	wheelbase	205 non-null	float64
10	carlength	205 non-null	float64
11	carwidth	205 non-null	float64
12	carheight	205 non-null	float64
13	curbweight	205 non-null	int64
14	enginetype	205 non-null	object
15	cylindernumber	205 non-null	object
16	enginesize	205 non-null	int64
17	fuelsystem	205 non-null	object
18	boreratio	205 non-null	float64
19	stroke	205 non-null	float64
20	compressionratio	205 non-null	float64
21	horsepower	205 non-null	int64
22	peakrpm	205 non-null	int64

```

23  citympg          205 non-null    int64
24  highwaympg       205 non-null    int64
25  price            205 non-null    float64
dtypes: float64(8), int64(8), object(10)
memory usage: 41.8+ KB

```

```

# melihat statistik data untuk data numeric seperti count, mean,
standard deviation, maximum, mininum, dan quartile.
dataset.describe()

```

	car_ID	symboling	wheelbase	...	citympg	highwaympg
price						
count	205.000000	205.000000	205.000000	...	205.000000	205.000000
mean	103.000000	0.834146	98.756585	...	25.219512	30.751220
std	59.322565	1.245307	6.021776	...	6.542142	6.886443
min	1.000000	-2.000000	86.600000	...	13.000000	16.000000
25%	52.000000	0.000000	94.500000	...	19.000000	25.000000
50%	103.000000	1.000000	97.000000	...	24.000000	30.000000
75%	154.000000	2.000000	102.400000	...	30.000000	34.000000
max	205.000000	3.000000	120.900000	...	49.000000	54.000000

```

[8 rows x 16 columns]

```

```

# cek nilai yang hilang / missing values di dalam data
dataset.isnull().sum()

```

car_ID	0
symboling	0
CarName	0
fueltype	0
aspiration	0
doornumber	0
carbody	0
drivewheel	0
enginelocation	0
wheelbase	0
carlength	0
carwidth	0
carheight	0
curbweight	0
enginetype	0
cylindernumber	0
enginesize	0

```

fuelsystem      0
boreratio       0
stroke          0
compressionratio 0
horsepower      0
peakrpm         0
citympg         0
highwaympg      0
price           0
dtype: int64

```

```

# Latihan 3
dataset.corr()

```

	car_ID	symboling	...	highwaympg	price
car_ID	1.000000	-0.151621	...	0.011255	-0.109093
symboling	-0.151621	1.000000	...	0.034606	-0.079978
wheelbase	0.129729	-0.531954	...	-0.544082	0.577816
carlength	0.170636	-0.357612	...	-0.704662	0.682920
carwidth	0.052387	-0.232919	...	-0.677218	0.759325
carheight	0.255960	-0.541038	...	-0.107358	0.119336
curbweight	0.071962	-0.227691	...	-0.797465	0.835305
enginesize	-0.033930	-0.105790	...	-0.677470	0.874145
boreratio	0.260064	-0.130051	...	-0.587012	0.553173
stroke	-0.160824	-0.008735	...	-0.043931	0.079443
compressionratio	0.150276	-0.178515	...	0.265201	0.067984
horsepower	-0.015006	0.070873	...	-0.770544	0.808139
peakrpm	-0.203789	0.273606	...	-0.054275	-0.085267
citympg	0.015940	-0.035823	...	0.971337	-0.685751
highwaympg	0.011255	0.034606	...	1.000000	-0.697599
price	-0.109093	-0.079978	...	-0.697599	1.000000

```

[16 rows x 16 columns]

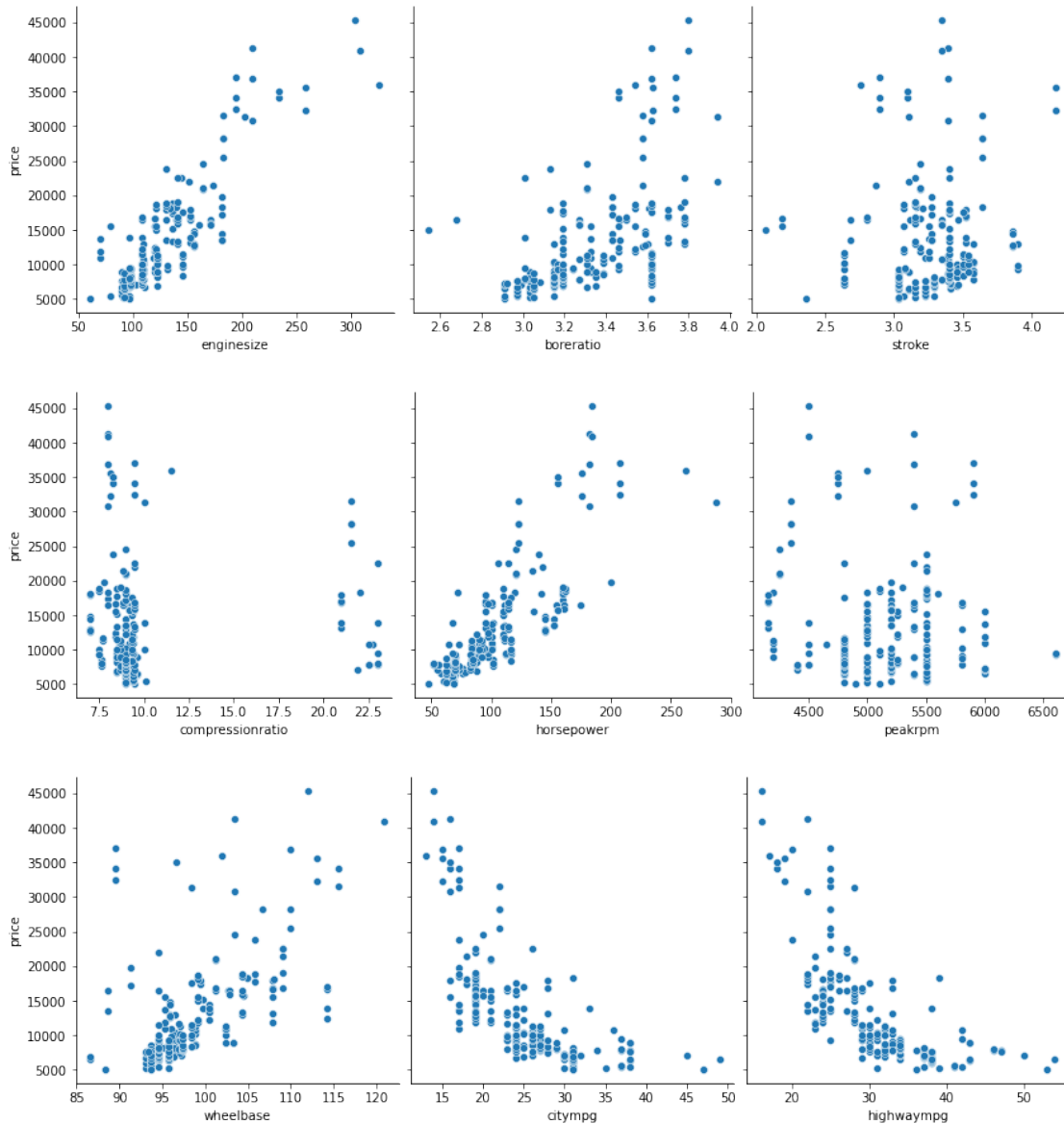
```

```

# Latihan 4
def pp(x,y,z):
    sns.pairplot(dataset, x_vars=[x,y,z], y_vars='price',size=4,
    aspect=1, kind="scatter")
    plt.show()

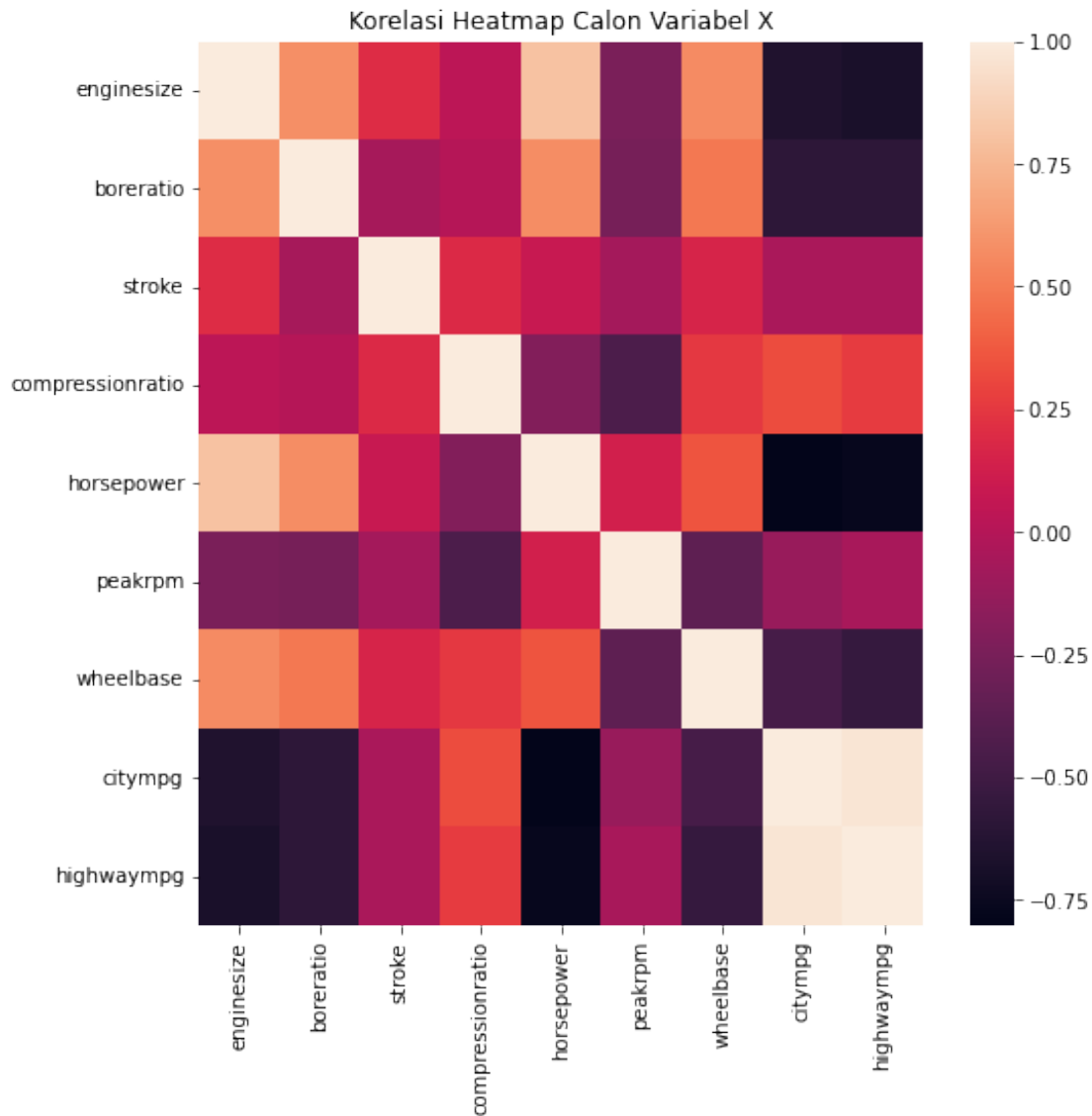
pp('enginesize', 'boreratio', 'stroke')
pp('compressionratio', 'horsepower', 'peakrpm')
pp('wheelbase', 'citympg', 'highwaympg')

```



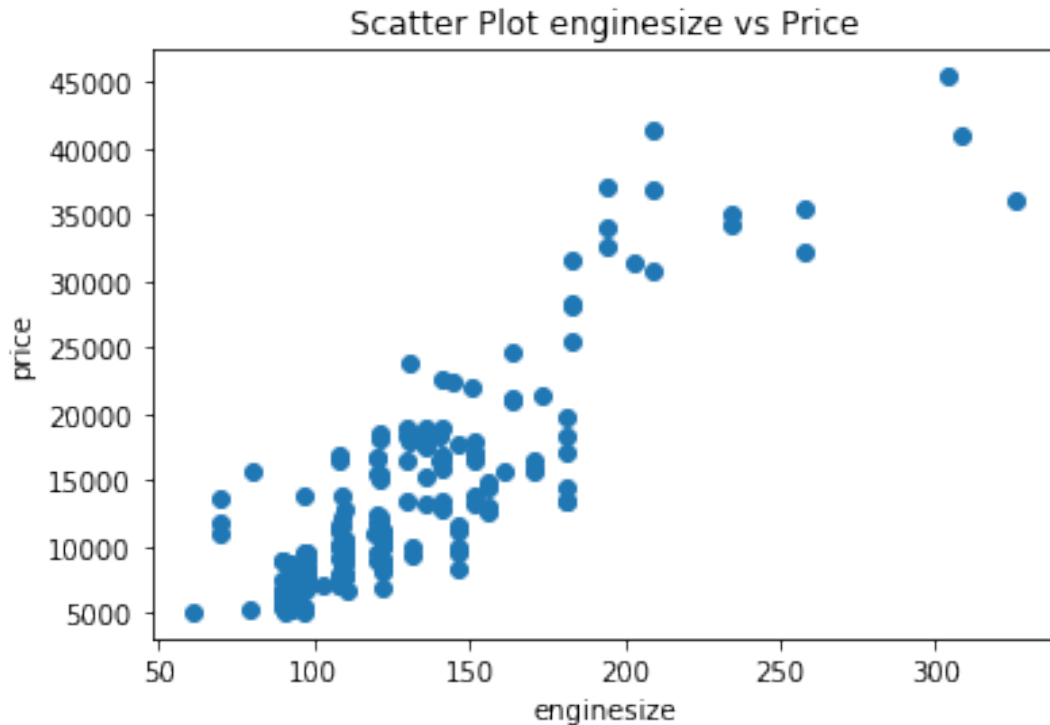
Latihan 5

```
plt.figure(figsize = (8,8))
data_fitur = dataset[['enginesize', 'boreratio',
'stroke','compressionratio', 'horsepower', 'peakrpm', 'wheelbase',
'citympg', 'highwaympg']]
sns.heatmap(data_fitur.corr(),annot=False,fmt="f").set_title("Korelasi
Heatmap Calon Variabel X")
plt.show()
```



Latihan 6

```
plt.scatter(dataset['enginesize'], dataset['price'])
plt.xlabel('enginesize')
plt.ylabel('price')
plt.title('Scatter Plot enginesize vs Price')
plt.show()
```



```
# Latihan 7
# Prepare data
# Pertama, buat variabel x dan y.
x = dataset['enginesize'].values.reshape(-1,1)
y = dataset['price'].values.reshape(-1,1)

# Latihan 8
x_mean = np.mean(x)
y_mean = np.mean(y)
print('nilai mean var x: ', x_mean, '\n'
      'nilai mean var y: ', y_mean)

nilai mean var x: 126.90731707317073
nilai mean var y: 13276.710570731706

# Latihan 9
atas = sum((x - x_mean)*(y - y_mean))
bawah = math.sqrt((sum((x - x_mean)**2)) * (sum((y - y_mean)**2)))
correlation = atas/bawah
print('Nilai Correlation Coefficient: ', correlation)

Nilai Correlation Coefficient: [0.8741448]

# Latihan 10
# slope
# Slope adalah tingkat kemiringan garis, intercept
# adalah jarak titik y pada garis dari titik 0
variance = sum((x - x_mean)**2)
```

```

covariance = sum((x - x_mean) * (y - y_mean))
theta_1 = covariance/variance
print('Nilai theta_1: ',theta_1)

Nilai theta_1:  [167.69841639]

# Latihan 11
# intercept
theta_0 = y_mean - (theta_1 * x_mean)
print('Nilai theta_1: ',theta_0)

Nilai theta_1:  [-8005.44553115]

# Latihan 12
# prediction manual
y_pred = theta_0 + (theta_1 * 130)

print(y_pred)

[13795.34859997]

# visualisasi prediksi dengan scatter plot
y_pred = theta_0 + (theta_1 * x)

plt.scatter(x,y)
plt.plot(x, y_pred, c='r')
plt.xlabel('enginesize')
plt.ylabel('Price')
plt.title('Plot enginesize vs Price')

Text(0.5, 1.0, 'Plot enginesize vs Price')

```




```
# Latihan 13
```

```
x_train, x_test, y_train, y_test = train_test_split(x, y, train_size =  
0.8, test_size = 0.2, random_state = 100)
```

```
# lathihan 14
```

```
regressor = LinearRegression()
```

```
# Latihan 15
```

```
regressor.fit(x_train, y_train)
```

```
LinearRegression()
```

```
# Latihan 16
```

```
print(regressor.coef_)
```

```
print(regressor.intercept_)
```

```
[[168.17363122]]
```

```
[-8037.06049611]
```

```
# Latihan 17
```

```
regressor.score(x_test, y_test)
```

```
0.8068161903454086
```

```
# Latihan 18
```

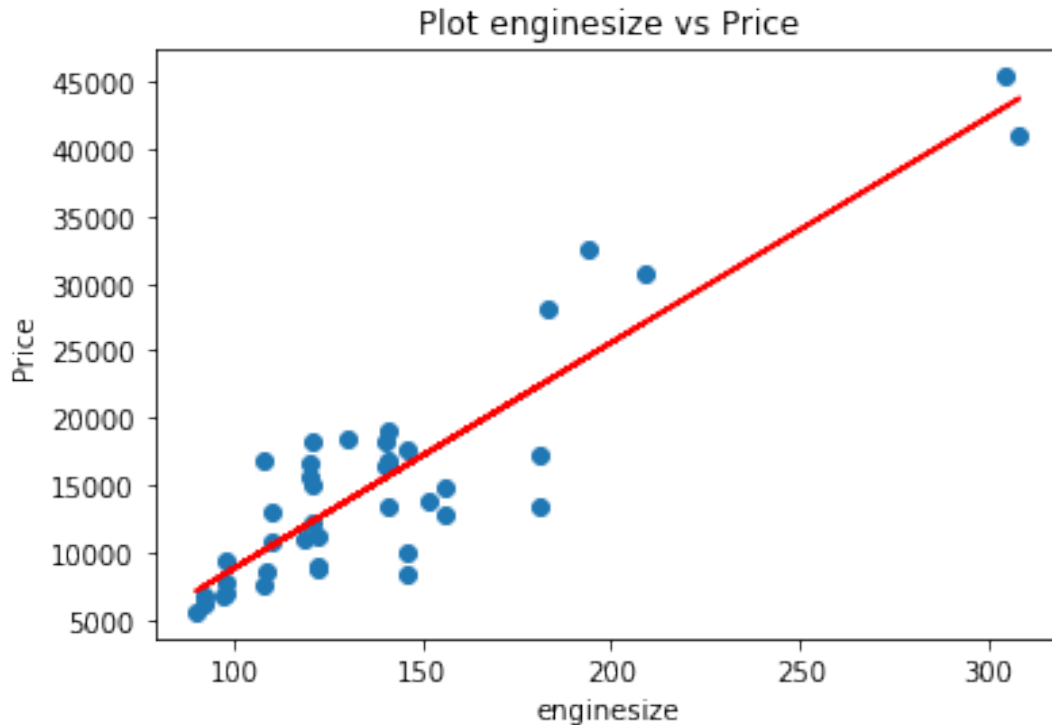
```
y_prediksi = regressor.predict(x_test)
```

```
plt.scatter(x_test, y_test)
```

```
plt.plot(x_test, y_prediksi, c='r')
```

```
plt.xlabel('enginesize')
plt.ylabel('Price')
plt.title('Plot enginesize vs Price')

Text(0.5, 1.0, 'Plot enginesize vs Price')
```



Latihan 19

#Prediksi harga mobil dengan enginesize 130.

```
print('nilai prediksi harga dengan enginesize 100 :
',regressor.predict([[100]]))
print('nilai prediksi harga dengan enginesize 150 :
',regressor.predict([[150]]))
print('nilai prediksi harga dengan enginesize 200 :
',regressor.predict([[200]]))

nilai prediksi harga dengan enginesize 100 : [[8780.30262568]]
nilai prediksi harga dengan enginesize 150 : [[17188.98418658]]
nilai prediksi harga dengan enginesize 200 : [[25597.66574748]]

np_table = np.concatenate((x_test,y_test,y_prediksi), axis=1)
new_dataframe = pd.DataFrame(data=np_table,
columns=['x_test','y_test','y_predict'])

new_dataframe
```

	x_test	y_test	y_predict
0	98.0	7738.0	8443.955363
1	109.0	8495.0	10293.865307

2	122.0	8845.0	12480.122512
3	98.0	9298.0	8443.955363
4	108.0	7603.0	10125.691675
5	122.0	11245.0	12480.122512
6	130.0	18420.0	13825.511562
7	140.0	16503.0	15507.247874
8	146.0	17669.0	16516.289662
9	181.0	17199.0	22402.366754
10	141.0	16845.0	15675.421506
11	121.0	18150.0	12311.948881
12	120.0	15580.0	12143.775250
13	110.0	12945.0	10462.038938
14	308.0	40960.0	43760.417919
15	92.0	6855.0	7434.913576
16	98.0	6938.0	8443.955363
17	121.0	12170.0	12311.948881
18	140.0	18280.0	15507.247874
19	156.0	14869.0	18198.025974
20	141.0	13415.0	15675.421506
21	141.0	16515.0	15675.421506
22	194.0	32528.0	24588.623960
23	90.0	5572.0	7098.566314
24	146.0	8449.0	16516.289662
25	181.0	13499.0	22402.366754
26	156.0	12764.0	18198.025974
27	183.0	28176.0	22738.714017
28	108.0	16925.0	10125.691675
29	119.0	11048.0	11975.601619
30	92.0	6189.0	7434.913576
31	209.0	30760.0	27111.228428
32	152.0	13860.0	17525.331449
33	141.0	19045.0	15675.421506
34	120.0	16630.0	12143.775250
35	110.0	10698.0	10462.038938
36	121.0	15040.0	12311.948881
37	146.0	9989.0	16516.289662
38	97.0	6849.0	8275.781732
39	122.0	8948.0	12480.122512
40	304.0	45400.0	43087.723394

Latihan 20

```

from sklearn import metrics
print('Mean Absolute Error:', metrics.mean_absolute_error(y_test,
y_prediksi))
print('Mean Squared Error:', metrics.mean_squared_error(y_test,
y_prediksi))
print('Root Mean Squared Error:',
np.sqrt(metrics.mean_squared_error(y_test, y_prediksi)))

```

Mean Absolute Error: 3123.611515387693
Mean Squared Error: 14882644.972928163
Root Mean Squared Error: 3857.8031278083854

```
plt.title('Comparison of Y values in test and the Predicted values')  
plt.ylabel('Test Set')  
plt.xlabel('Predicted values')  
plt.plot(y_prediksi, '.', y_test, 'x')  
plt.show()
```

