Basic Analysis using Numpy and Pandas

import libraries

```
In [1]: import pandas as pd
import numpy as np
import matplotlib.pyplot as pp
```

import dataset

```
In [2]: data=pd.read_csv(r"E:\154\fiat500_VehicleSelection_Dataset - fiat500_VehicleSelection_
```

In [3]: display(data)

	ID	model	engine_power	age_in_days	km	previous_owners	lat	I
0	1.0	lounge	51.0	882.0	25000.0	1.0	44.907242	8.6115598
1	2.0	pop	51.0	1186.0	32500.0	1.0	45.666359	12.241889
2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.417
3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.634609
4	5.0	pop	73.0	3074.0	106880.0	1.0	41.903221	12.495650
				•••		•••		
1544	NaN	NaN	NaN	NaN	NaN	NaN	NaN	lenç
1545	NaN	NaN	NaN	NaN	NaN	NaN	NaN	con
1546	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Null valı
1547	NaN	NaN	NaN	NaN	NaN	NaN	NaN	fi
1548	NaN	NaN	NaN	NaN	NaN	NaN	NaN	sear
1549 r	ows ×	: 11 colu	ımns					

To display top 10 rows

In [4]:	da	ta.h	ead()						
Out[4]:		ID	model	engine_power	age_in_days	km	previous_owners	lat	lon
	0	1.0	lounge	51.0	882.0	25000.0	1.0	44.907242	8.611559868
	1	2.0	рор	51.0	1186.0	32500.0	1.0	45.666359	12.24188995
	2	3.0	sport	74.0	4658.0	142228.0	1.0	45.503300	11.41784
	3	4.0	lounge	51.0	2739.0	160000.0	1.0	40.633171	17.63460922
	4	5.0	рор	73.0	3074.0	106880.0	1.0	41.903221	12.49565029
	4								•

To display last 5 rows

[5]:	data.	tail()								
t[5]:		ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price	Ur
	1544	NaN	NaN	NaN	NaN	NaN	NaN	NaN	length	5	
	1545	NaN	NaN	NaN	NaN	NaN	NaN	NaN	concat	Ionprice	
	1546	NaN	NaN	NaN	NaN	NaN	NaN	NaN	Null values	NO	
	1547	NaN	NaN	NaN	NaN	NaN	NaN	NaN	find	1	
	1548	NaN	NaN	NaN	NaN	NaN	NaN	NaN	search	1	
	4										•
[6]:	data.	dtype	·S								
6]:	ID			float64							
	model			object							
	engin			float64							
	age_i	n_day	'S	float64							
	km	0116 6	unanc	float64 float64							
	lat	ous_c	wners	float64							
	lon			object							
	price			object							
	Unnam)	float64							
	Unnam			object							
	dtype	: obj	iect	-							

To view statistical summary

7]:	data.d	escribe()					
		ID	engine_power	age_in_days	km	previous_owners	lat
	count	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000	1538.000000
	mean	769.500000	51.904421	1650.980494	53396.011704	1.123537	43.541361
	std	444.126671	3.988023	1289.522278	40046.830723	0.416423	2.133518
	min	1.000000	51.000000	366.000000	1232.000000	1.000000	36.855839
	25%	385.250000	51.000000	670.000000	20006.250000	1.000000	41.802990
	50%	769.500000	51.000000	1035.000000	39031.000000	1.000000	44.394096
	75%	1153.750000	51.000000	2616.000000	79667.750000	1.000000	45.467960
	max	1538.000000	77.000000	4658.000000	235000.000000	4.000000	46.795612
	4						

To Print no of elements

```
In [8]: data.size
Out[8]: 17039
In [9]: data.ndim
Out[9]: 2
```

To print no of rows and columns

```
In [10]: data.shape
Out[10]: (1549, 11)
```

To find missing values

[11]:	data.	isna())								
t[11]:		ID	model	engine_power	age_in_days	km	previous_owners	lat	lon	price	Un
	0	False	False	False	False	False	False	False	False	False	
	1	False	False	False	False	False	False	False	False	False	
	2	False	False	False	False	False	False	False	False	False	
	3	False	False	False	False	False	False	False	False	False	
	4	False	False	False	False	False	False	False	False	False	
				•••	•••						
	1544	True	True	True	True	True	True	True	False	False	
	1545	True	True	True	True	True	True	True	False	False	
	1546	True	True	True	True	True	True	True	False	False	
	1547	True	True	True	True	True	True	True	False	False	
	1548	True	True	True	True	True	True	True	False	False	
	1549 r	ows ×	11 colur	mns							

To fill null values with constatns

lo	lat	previous_owners	km	age_in_days	engine_power	model	ID	
8.61155986	44.907242	1.0	25000.0	882.0	51.0	lounge	1.0	0
12.2418899	45.666359	1.0	32500.0	1186.0	51.0	рор	2.0	1
11.4178	45.503300	1.0	142228.0	4658.0	74.0	sport	3.0	2
17.6346092	40.633171	1.0	160000.0	2739.0	51.0	lounge	4.0	3
12.4956502	41.903221	1.0	106880.0	3074.0	73.0	pop	5.0	4
				•••				
lengt	5.000000	5.0	5.0	5.0	5.0	5	5.0	1544
conca	5.000000	5.0	5.0	5.0	5.0	5	5.0	1545
Null value	5.000000	5.0	5.0	5.0	5.0	5	5.0	1546
fin	5.000000	5.0	5.0	5.0	5.0	5	5.0	1547
searc	5.000000	5.0	5.0	5.0	5.0	5	5.0	1548
					umns	× 11 col	ows	1549 r
>					lumns	× 11 col	ows	1549 r
•		_	-					1
)					umns ,'price']]			1
•								1
•								data=
•					,'price']]	[['km'	data	data=
•					,'price']] price	[[' km'	data 250	data=
•					,'price']] price 8900	km	data 250 325	data= data 0
					<pre>price 8900 8800</pre>	km 000.0 500.0	250 328 1422	data= data 0 1
					price 8900 8800 4200	km 000.0 500.0 228.0	250 328 1422 1600	data= data 0 1 2
					price 8900 8800 4200 6000	km 000.0 500.0 228.0	250 328 1422 1600	data= data 0 1 2 3
					price 8900 8800 4200 6000 5700	km 000.0 500.0 228.0 000.0 880.0	250 328 1422 1600	data= data 0 1 2 3 4
					price]	km 000.0 500.0 228.0 000.0 880.0 	250 328 1422 1600	data= data 0 1 2 3 4
					price 8900 8800 4200 6000 5700 5	km 000.0 500.0 228.0 000.0 880.0 	250 328 1422 1600	data= data 0 1 2 3 4 1544
					price 8900 8800 4200 6000 5700 5	km 000.0 500.0 228.0 000.0 880.0 NaN	250 328 1422 1600	data= data 0 1 2 3 4 1544 1545

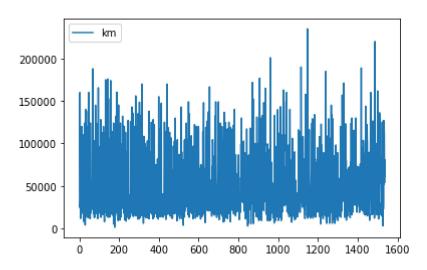
Type *Markdown* and LaTeX: α^2

```
In [15]: data=data[['km','price']]
```

Line Chart

```
In [22]: data.plot.line()
```

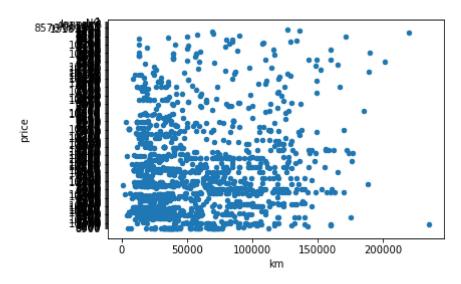
Out[22]: <AxesSubplot:>



Scatter Plot

```
In [27]: data.plot.scatter(x='km',y='price')
```

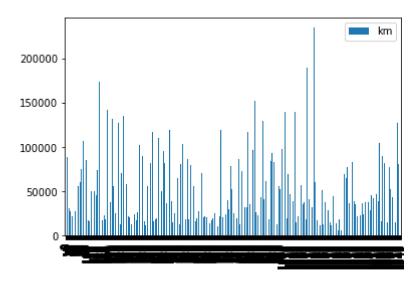
Out[27]: <AxesSubplot:xlabel='km', ylabel='price'>



Bar Chart

In [30]: data.plot.bar()

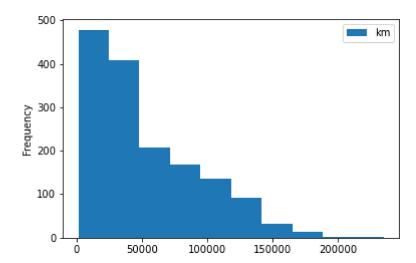
Out[30]: <AxesSubplot:>



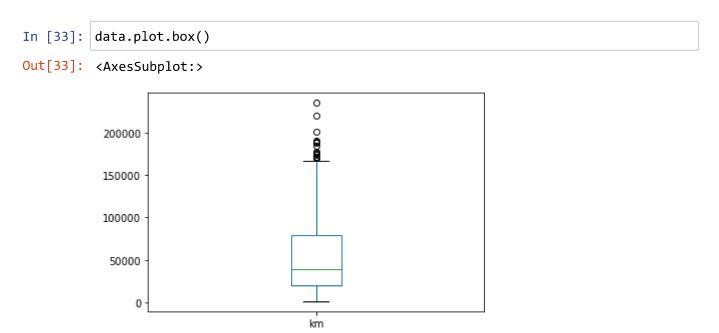
Histogram

In [31]: data.plot.hist()

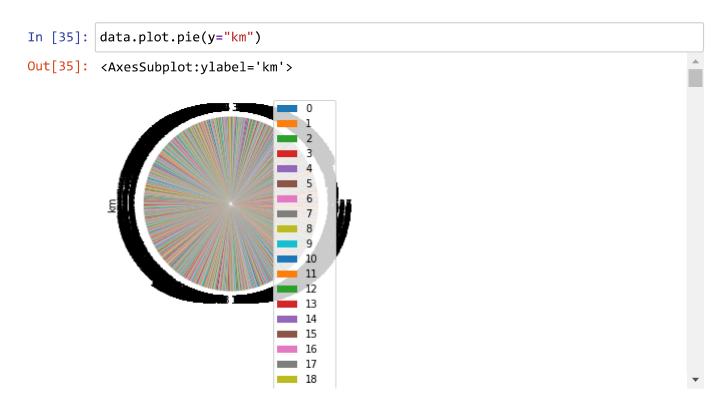
Out[31]: <AxesSubplot:ylabel='Frequency'>



Box Plot



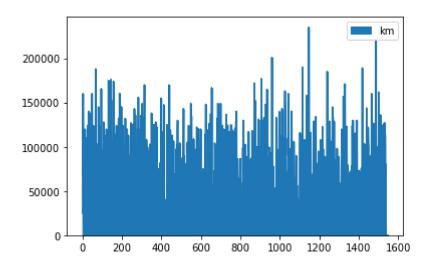
Pie Chart



Area

```
In [36]: data.plot.area()
```

Out[36]: <AxesSubplot:>



In []: