

Project Name : Vehicles Insurance Claim Data

Aim :

To describe and analyze and Draw the plots of all dataset

Libraries Info:

i will describe and analyze all the data with the help of Two libraries like....,

i used pandas library import pandas as pd

i used matplotlib.pyplot library import matplotlib.pyplot as plt

```
In [3]: import pandas as pd
import matplotlib.pyplot as plt
```

I used pd.read_excel for collect the excel file with thier link from file manager

```
In [51]: df = pd.read_excel("C:\Project Files (Shital)\Vehicle Insurance Claims Data.xlsx")
df
```

	AGE	CRUISE	DATE OF BIRTH	IN	PRICE	TYPE
2	29	687698	2000-09-06	OH	1413.14	430632
3	41	227811	1990-05-25	IL	1415.74	608117
4	44	367455	2014-06-06	IL	1583.91	610706
...
995	38	941851	1991-07-16	OH	1310.80	431289
996	41	186934	2014-01-05	IL	1436.79	608177
997	34	918516	2003-02-17	OH	1383.49	442797
998	62	533940	2011-11-18	IL	1356.92	441714
999	60	556080	1996-11-11	OH	766.19	612260

1000 rows × 22 columns

df.shape is used for show the shape of whole data such as 1000 rows and 24 columns are their

```
In [52]: df.shape
```

```
Out[52]: (1000, 22)
```

the len() function is used for length of whole data

```
In [53]: print(len(df))
```

1000

the replace() function is use to replace the values

```
In [55]: df['Policy_state'] = df['Policy_state'].replace({
    "OH": "Ohio",
    "IN": "in",
    "IL": "Illinois"

}, regex=True)

df.head()
```

Out[55]:

	Age	Policy_number	Policy_bind_date	Policy_state	Policy_annual_premium	Insured_zip	Insured
0	48	521585	2014-10-17	Ohio	1406.91	466132	
1	42	342868	2006-06-27	in	1197.22	468176	
2	29	687698	2000-09-06	Ohio	1413.14	430632	
3	41	227811	1990-05-25	Illinois	1415.74	608117	
4	44	367455	2014-06-06	Illinois	1583.91	610706	

5 rows × 22 columns



the replace() function is use to replace the multiple values at a time

```
In [56]: df['Incident_state'] = df['Incident_state'].replace({
    "OH": "Ohio",
    "IN": "in",
    "IL": "Illinois",
    "NY": "New York",
    "SC": "Sorth Carolina",
    "WV": "West Verginia",
    "VA": "Virginia",
    "NC": "North Carolina",
    "PA": "Pennsylvania",
    "OH": "Ohio",
    "IN": "in",
    "IL": "Illinois"

    }, regex=True)

df.head()
```

Out[56]:

	Age	Policy_number	Policy_bind_date	Policy_state	Policy_annual_premium	Insured_zip	Insured
0	48	521585	2014-10-17	Ohio	1406.91	466132	
1	42	342868	2006-06-27	in	1197.22	468176	
2	29	687698	2000-09-06	Ohio	1413.14	430632	
3	41	227811	1990-05-25	Illinois	1415.74	608117	
4	44	367455	2014-06-06	Illinois	1583.91	610706	

5 rows × 22 columns



the info() function is used for collect all info about data

```
In [57]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 22 columns):
#   Column                                          Non-Null Count  Dtype
---  -
0   Age                                           1000 non-null   int64
1   Policy_number                               1000 non-null   int64
2   Policy_bind_date                             1000 non-null   datetime64[ns]
3   Policy_state                                 1000 non-null   object
4   Policy_annual_premium                       1000 non-null   float64
5   Insured_zip                                  1000 non-null   int64
6   Insured_gender                              1000 non-null   object
7   Insured_occupation                          1000 non-null   object
8   Insured_relationship                        1000 non-null   object
9   Incident_date                               1000 non-null   datetime64[ns]
10  Incident_type                               1000 non-null   object
11  Collision_type                              1000 non-null   object
12  Incident_severity                           1000 non-null   object
13  Incident_state                              1000 non-null   object
14  Incident_city                               1000 non-null   object
15  Incident_location                           1000 non-null   object
16  Number_of_vehicles_involved                 1000 non-null   int64
17  Witnesses                                   1000 non-null   int64
18  Total_claim_amount                          1000 non-null   int64
19  Injury_claim                               1000 non-null   int64
20  Property_claim                             1000 non-null   int64
21  Vehicle_claim                              1000 non-null   int64
dtypes: datetime64[ns](2), float64(1), int64(9), object(10)
memory usage: 172.0+ KB
```

the column function is used for showing the number of column

```
In [58]: column_name = df.columns
```

```
print(column_name)
```

```
Index(['Age', 'Policy_number', 'Policy_bind_date', 'Policy_state',
       'Policy_annual_premium', 'Insured_zip', 'Insured_gender',
       'Insured_occupation', 'Insured_relationship', 'Incident_date',
       'Incident_type', 'Collision_type', 'Incident_severity',
       'Incident_state', 'Incident_city', 'Incident_location',
       'Number_of_vehicles_involved', 'Witnesses', 'Total_claim_amount',
       'Injury_claim', 'Property_claim', 'Vehicle_claim'],
      dtype='object')
```

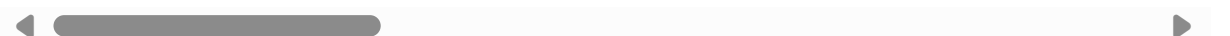
the head() function is used to shown initial information about data of excel file

```
In [59]: df.head(1)
```

```
Out[59]:
```

	Age	Policy_number	Policy_bind_date	Policy_state	Policy_annual_premium	Insured_zip	Insured_gender
0	48	521585	2014-10-17	Ohio	1406.91	466132	

1 rows × 22 columns



```
In [60]: df[['Insured_gender', 'Insured_relationship', 'Total_claim_amount']]
```

```
Out[60]:
```

	Insured_gender	Insured_relationship	Total_claim_amount
0	MALE	husband	71610
1	MALE	other-relative	5070
2	FEMALE	own-child	34650
3	FEMALE	unmarried	63400
4	MALE	unmarried	6500
...
995	FEMALE	unmarried	87200
996	FEMALE	wife	108480
997	FEMALE	other-relative	67500
998	MALE	wife	46980
999	FEMALE	husband	5060

1000 rows × 3 columns

the count() function is used to shown data of Insured gender such as,

there are 537 Female and 463 Male

```
In [61]: df['Insured_gender'].value_counts()
```

```
Out[61]: FEMALE    537  
        MALE      463  
        Name: Insured_gender, dtype: int64
```

the count() is used to shown data of all Policy State such as...,

the highest policy state of Ohio's count is 352

the lowest policy state of in's count is 310

```
In [62]: df['Policy_state'].value_counts()
```

```
Out[62]: Ohio        352  
        Illinois    338  
        in         310  
        Name: Policy_state, dtype: int64
```

The count() function is used to shown data of all Incident State such as...,

the highest Incidents occurs in New York state and it's count is 262

the lowest Incident occurs in Ohio state and it's count is 23

```
In [63]: df['Incident_state'].value_counts()
```

```
Out[63]: New York          262
         South Carolina    248
         West Virginia     217
         Virginia         110
         North Carolina    110
         Pennsylvania      30
         Ohio              23
         Name: Incident_state, dtype: int64
```

```
In [64]: Incident_severity= df.groupby("Incident_severity")
         Incident_severity.size()
```

```
Out[64]: Incident_severity
         Major Damage      276
         Minor Damage     354
         Total Loss       280
         Trivial Damage    90
         dtype: int64
```

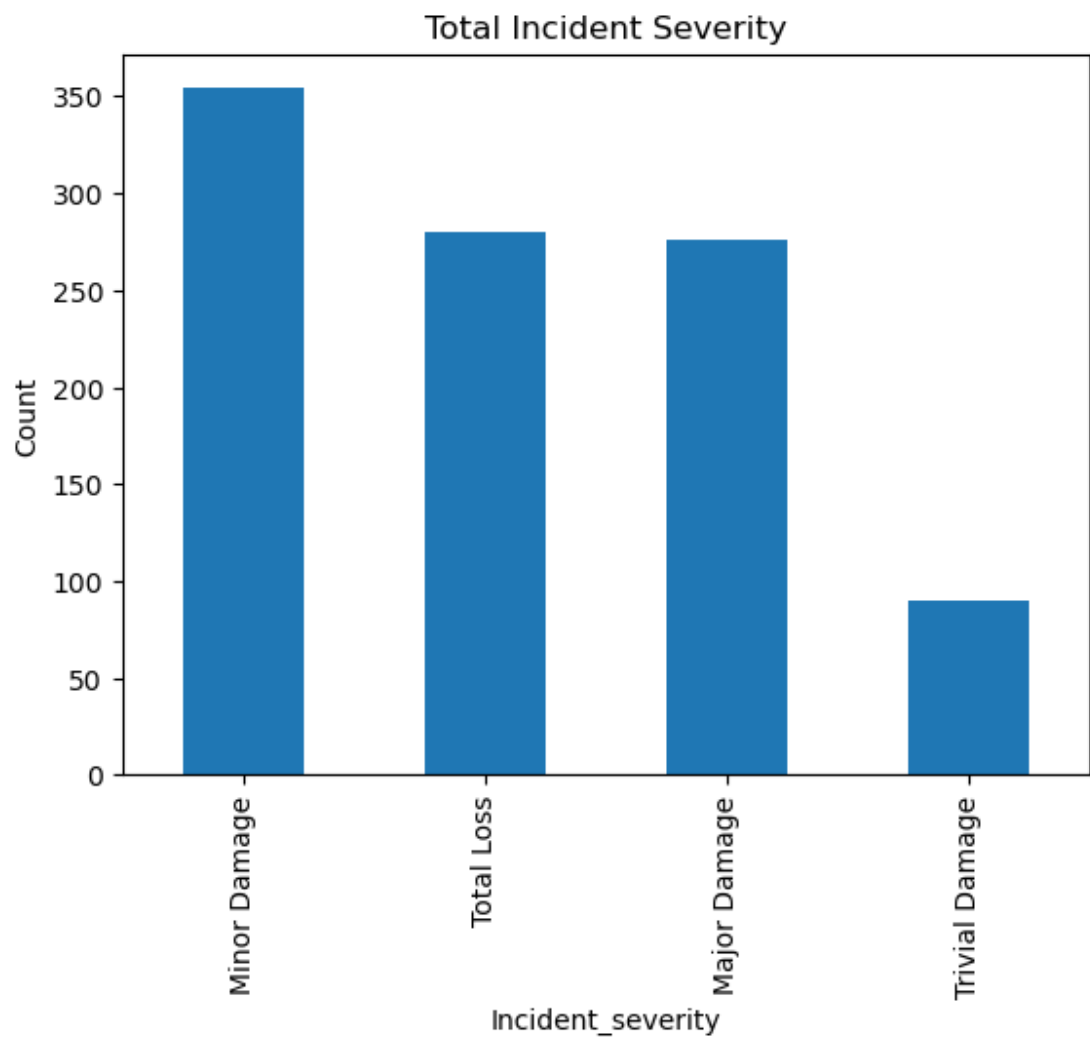
in this plot thier are Minor Damage vehicles are 350

the major Damage vehicles are 275

the total loss vehicles are 280

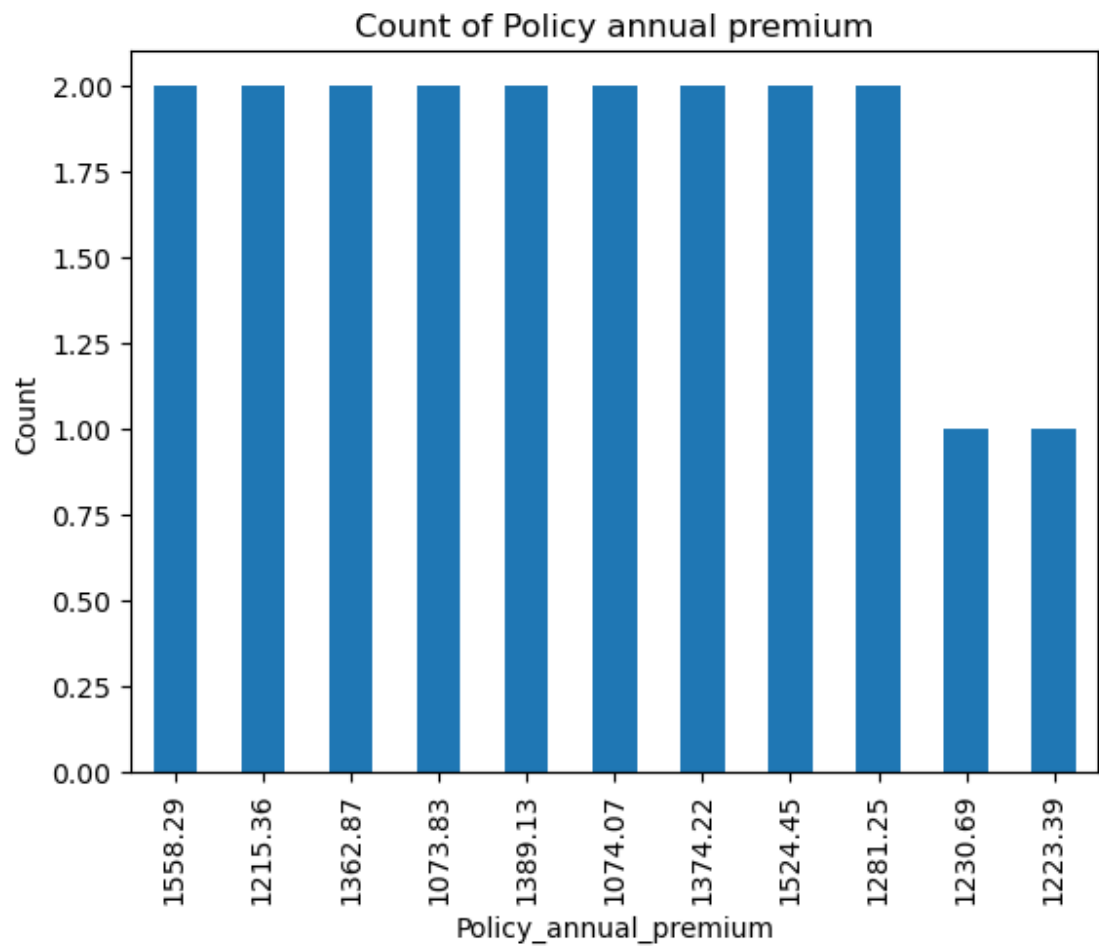
the trivial damage vehicles are 70

```
In [65]: df['Incident_severity'].value_counts().nlargest(11).plot(kind='bar')
plt.title('Total Incident Severity')
plt.xlabel('Incident_severity')
plt.ylabel('Count')
plt.show()
```



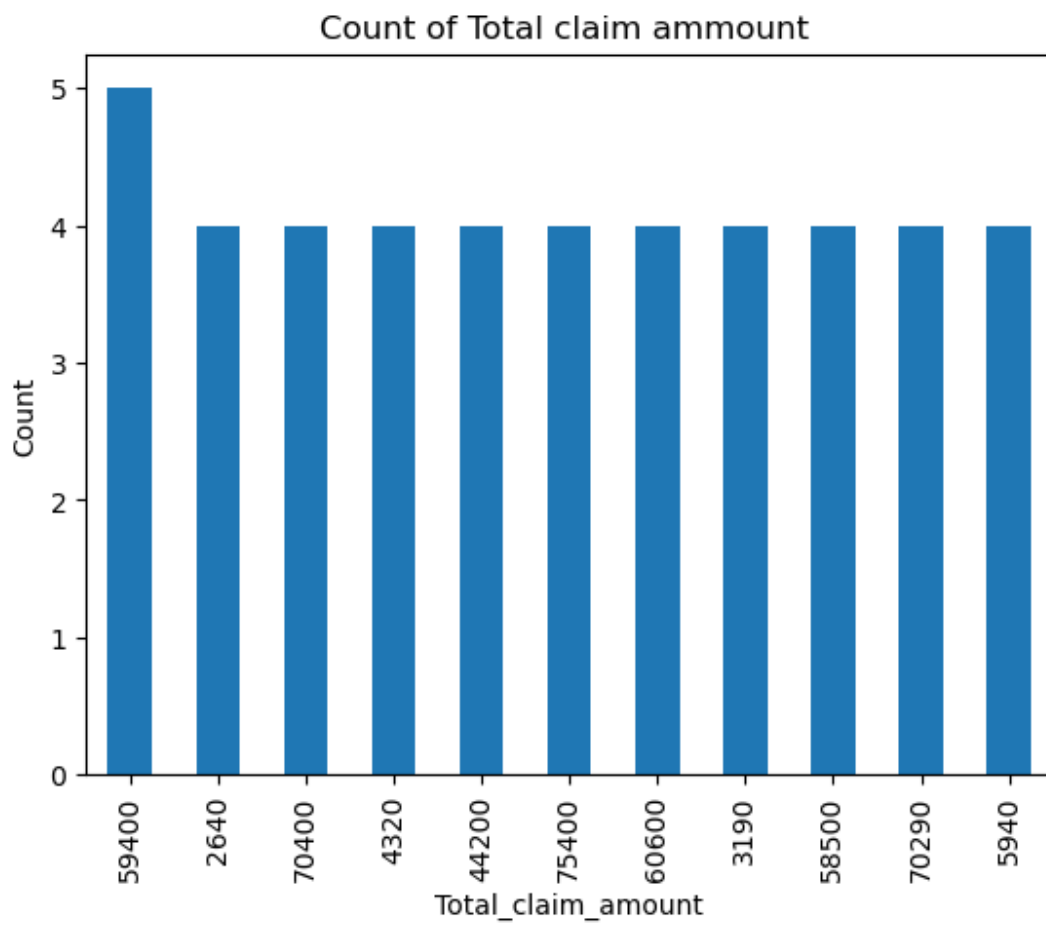
in this graph , the lowest Policy_annual_premium is 1230.69 and 1223.39 as compair to others

```
In [66]: df['Policy_annual_premium'].value_counts().nlargest(11).plot(kind='bar')
plt.title('Count of Policy annual premium')
plt.xlabel('Policy_annual_premium')
plt.ylabel('Count')
plt.show()
```



in this graph , the highest Total_claim_amount is 59400 among the others


```
In [67]: df['Total_claim_amount'].value_counts().nlargest(11).plot(kind='bar')
plt.title('Count of Total claim ammount ')
plt.xlabel('Total_claim_amount')
plt.ylabel('Count')
plt.show()
```



```
In [68]: age = df.groupby("Age")
```

```
age.size()
```

```
Out[68]: Age
19      1
20      1
21      6
22      1
23      7
24     10
25     14
26     26
27     24
28     30
29     35
30     42
31     42
32     38
33     39
34     44
35     32
36     32
37     41
38     42
39     48
40     38
41     45
42     32
43     49
44     32
45     26
46     33
47     24
48     25
49     14
50     13
51      9
52      4
53     13
54     10
55     14
56      8
57     16
58      8
59      5
60      9
61     10
62      4
63      2
64      2
dtype: int64
```

this is a group about Policy_state

in this group ...,

the highest policy state illinois is 338 and

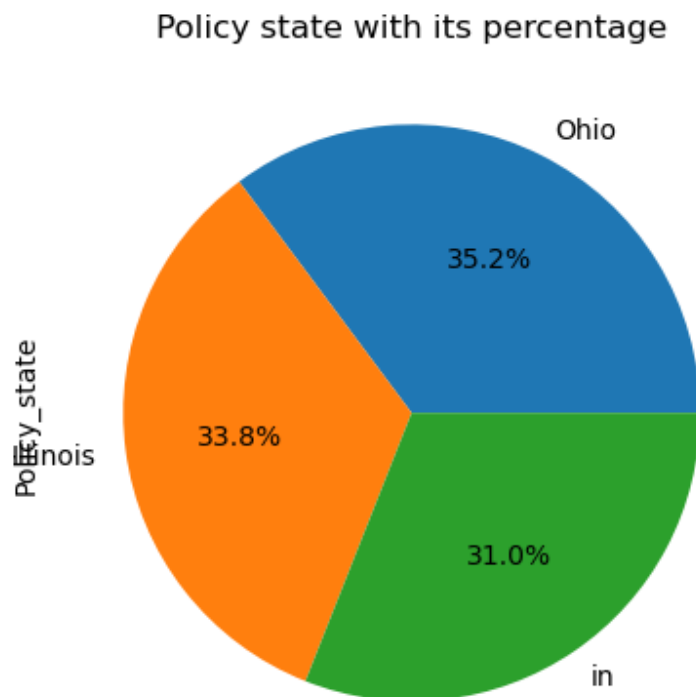
the middlest policy state Ohio is 352 ,

the lowest policy state in is 310

```
In [69]: Policy_state = df.groupby("Policy_state")  
  
Policy_state.size()
```

```
Out[69]: Policy_state  
Illinois    338  
Ohio        352  
in          310  
dtype: int64
```

```
In [70]: df['Policy_state'].value_counts().plot(kind='pie', autopct='%1.1f%%')  
plt.title('Policy state with its percentage')  
plt.show()
```



this is a group about Incident_state

in this group ...,

the highest Incident state New York is 262 and

the lowest Incident state West Verginia is 217

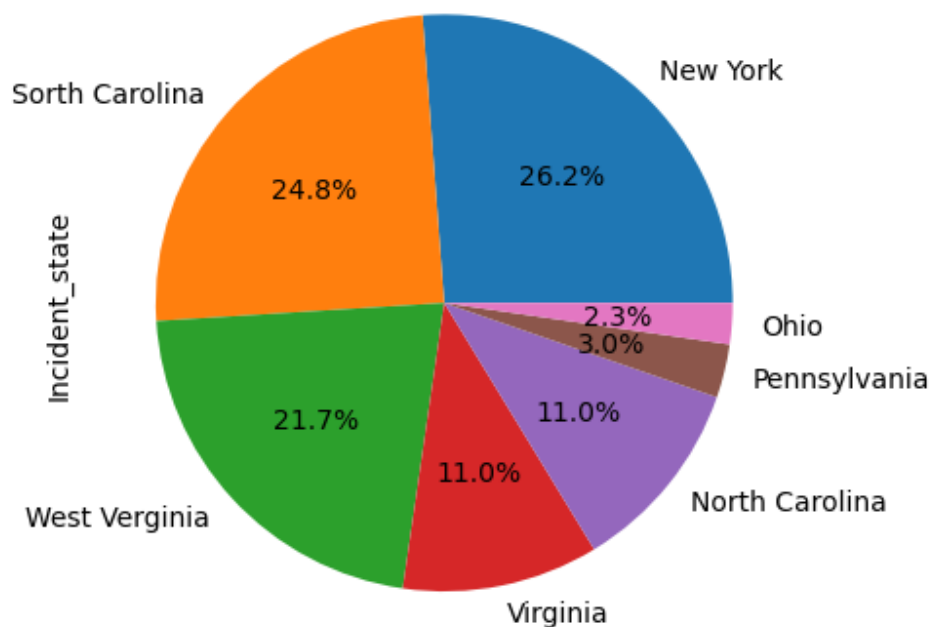
```
In [71]: Incident_state = df.groupby("Incident_state")
```

```
Incident_state.size()
```

```
Out[71]: Incident_state
New York      262
North Carolina 110
Ohio           23
Pennsylvania   30
South Carolina 248
Virginia       110
West Virginia  217
dtype: int64
```

```
In [72]: df['Incident_state'].value_counts().plot(kind='pie', autopct='%1.1f%%')
plt.title('Insident state with its percentage')
plt.show()
```

Insident state with its percentage



this is a group about Insured_gender

in this group ...,

the Insured genders are:

the female is 537

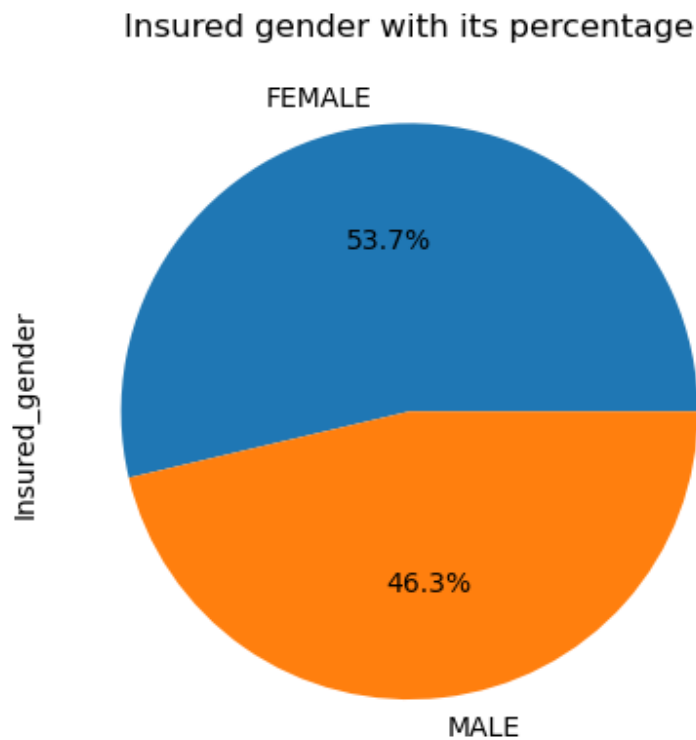
the male is 463

```
In [73]: Insured_gender= df.groupby("Insured_gender")
```

```
Insured_gender.size()
```

```
Out[73]: Insured_gender  
FEMALE    537  
MALE      463  
dtype: int64
```

```
In [74]: df['Insured_gender'].value_counts().plot(kind='pie', autopct='%1.1f%%')  
plt.title('Insured gender with its percentage')  
plt.show()
```



in this, the groupby is used for type of Collision

the ? is 178

the Front Collision is 254

the Rear Collision is 292

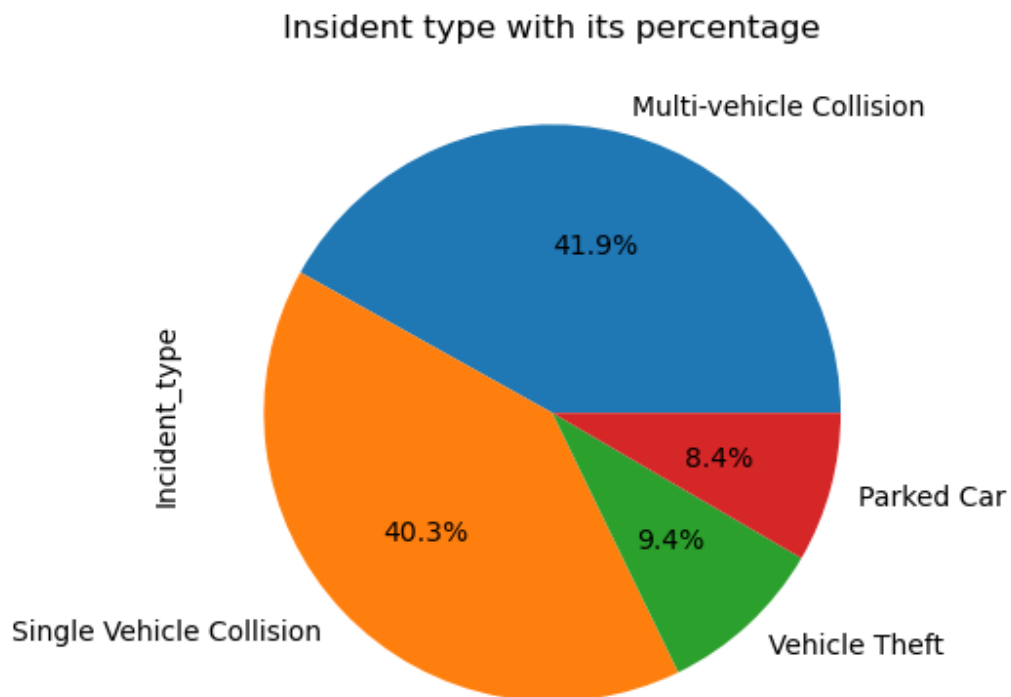
the Side Collision is 276

```
In [75]: Collision_type= df.groupby("Collision_type")
```

```
Collision_type.size()
```

```
Out[75]: Collision_type  
?                178  
Front Collision   254  
Rear Collision    292  
Side Collision    276  
dtype: int64
```

```
In [76]: df['Incident_type'].value_counts().plot(kind='pie', autopct='%1.1f%%')
plt.title('Insident type with its percentage')
plt.show()
```



in this, the groupby is used for type of incident

the Multi-vehicle Collision is 419

the Parked Cars are 84

the Single Vehicle Collision is 403

the Vehicle Theft is 94

```
In [77]: Incident_type= df.groupby("Incident_type")
Incident_type.size()
```

```
Out[77]: Incident_type
Multi-vehicle Collision    419
Parked Car                 84
Single Vehicle Collision   403
Vehicle Theft              94
dtype: int64
```

```
In [78]: Incident_city= df.groupby("Incident_city")
        Incident_city.size()
```

```
Out[78]: Incident_city
Arlington      152
Columbus       149
Hillsdale      141
Northbend      145
Northbrook     122
Riverwood      134
Springfield    157
dtype: int64
```

Conclusion:

In this project, I analyzed the all data like the highest Incidents occurs in Springfield city and it's count is 157 and also the lowest Incident occurs in Northbrook city and it's count is 122.

The Policy_annual_premium is 1230.69 and 1223.39 as compare to others

The highest Incident state New York is 262 and The lowest Incident state West Virginia is 217

The Insured genders are: Female is 537 and Male is 463

In this, type of incident The Multi-vehicle Collision is 419 , the Parked Cars are 84,the Single Vehicle Collision is 403,the Vehicle Theft is 94

In this, the groupby is used for type of Collision ,the ? is 178, the Front Collision is 254, the Rear Collision is 292, the Side Collision is 276