

```
In [1]: import numpy as np
import pandas as pd
```

```
In [4]: df=pd.read_csv("Bank_Customer_Churn.csv",encoding='latin1')
```

```
In [5]: df.head()
```

```
Out[5]:
```

	Unnamed: 0	age	gender	dependents	occupation	city	churn
0	0	66	Male	0.0	self_employed	187.0	0
1	1	35	Male	0.0	self_employed	NaN	0
2	2	31	Male	0.0	salaried	146.0	0
3	3	90	NaN	NaN	self_employed	1020.0	1
4	4	42	Male	2.0	self_employed	1494.0	1

```
In [7]: df.shape
```

```
Out[7]: (28382, 7)
```

```
In [8]: df.tail
```

```
Out[8]: <bound method NDFrame.tail of
occupation      city  churn
0              0    66    Male    0.0  self_employed    187.0    0
1              1    35    Male    0.0  self_employed     NaN    0
2              2    31    Male    0.0      salaried    146.0    0
3              3    90     NaN    NaN  self_employed   1020.0    1
4              4    42    Male    2.0  self_employed   1494.0    1
...          ...    ...    ...    ...    ...    ...    ...
28377         28377    10  Female    0.0      student   1020.0    0
28378         28378    34  Female    0.0  self_employed   1046.0    0
28379         28379    47    Male    0.0      salaried   1096.0    1
28380         28380    50    Male    3.0  self_employed   1219.0    0
28381         28381    18    Male    0.0      student   1232.0    1

[28382 rows x 7 columns]>
```

```
In [9]: df.isnull().sum()
```

```
Out[9]: Unnamed: 0      0
age              0
gender          525
dependents      2463
occupation       80
city            803
churn            0
dtype: int64
```

```
In [10]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 28382 entries, 0 to 28381
Data columns (total 7 columns):
 #   Column        Non-Null Count  Dtype  
---  -
 0   Unnamed: 0    28382 non-null  int64  
 1   age           28382 non-null  int64  
 2   gender        27857 non-null  object  
 3   dependents    25919 non-null  float64 
 4   occupation    28302 non-null  object  
 5   city          27579 non-null  float64 
 6   churn         28382 non-null  int64  
dtypes: float64(2), int64(3), object(2)
memory usage: 1.5+ MB
```

```
In [11]: city_mean=df['city'].mean()
```

```
In [12]: df['city'].replace(np.NaN,city_mean,inplace=True)
```

```
In [13]: df.isnull().sum()
```

```
Out[13]: Unnamed: 0      0
age          0
gender       525
dependents   2463
occupation   80
city         0
churn        0
dtype: int64
```

```
In [20]: df['occupation']=df['occupation'].dropna()
```

```
In [21]: df.isnull().sum()
```

```
Out[21]: Unnamed: 0      0
age          0
gender       525
dependents   2463
occupation    0
city         0
churn        0
dtype: int64
```

```
In [22]: gender_mode=df['gender'].mode()
```

```
In [23]: gender_mode
```

```
Out[23]: 0    Male
Name: gender, dtype: object
```

```
In [24]: df['gender']=df['gender'].fillna(value='gender_mode')
```

```
In [25]: df.isnull().sum()
```

```
Out[25]: Unnamed: 0      0
age          0
gender       0
dependents   2463
occupation   0
city         0
churn        0
dtype: int64
```

```
In [26]: dep_mean=df['dependents'].mean()
```

```
In [27]: df['dependents']=df['dependents'].fillna(value='dep_mean')
```

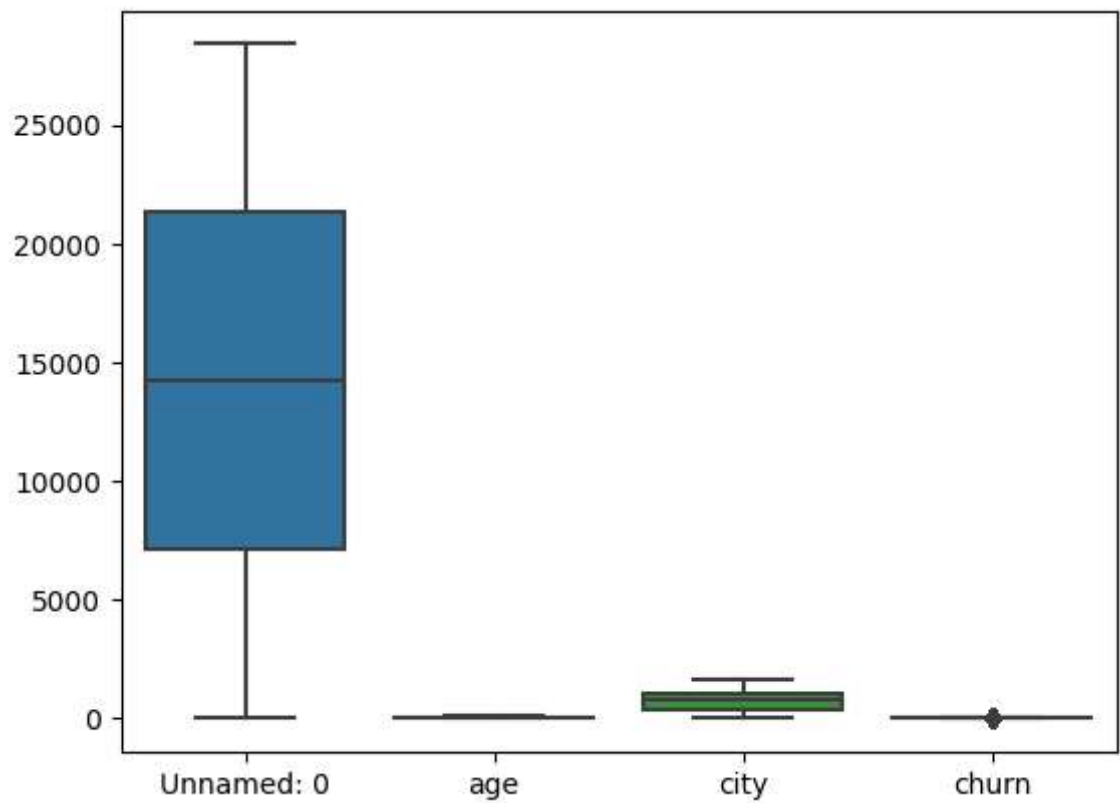
```
In [28]: df.isnull().sum()
```

```
Out[28]: Unnamed: 0      0
age          0
gender       0
dependents    0
occupation    0
city          0
churn         0
dtype: int64
```

```
In [29]: import matplotlib.pyplot as plt
import seaborn as sns
```

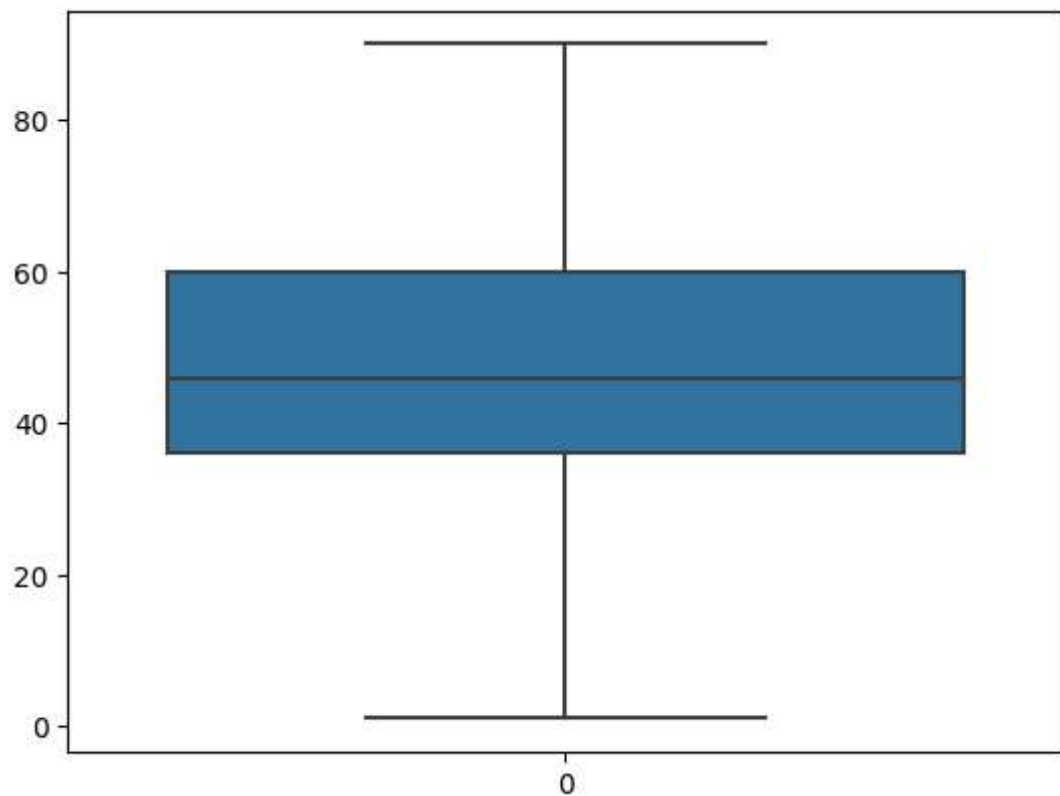
```
In [30]: sns.boxplot(df)
```

```
Out[30]: <AxesSubplot: >
```



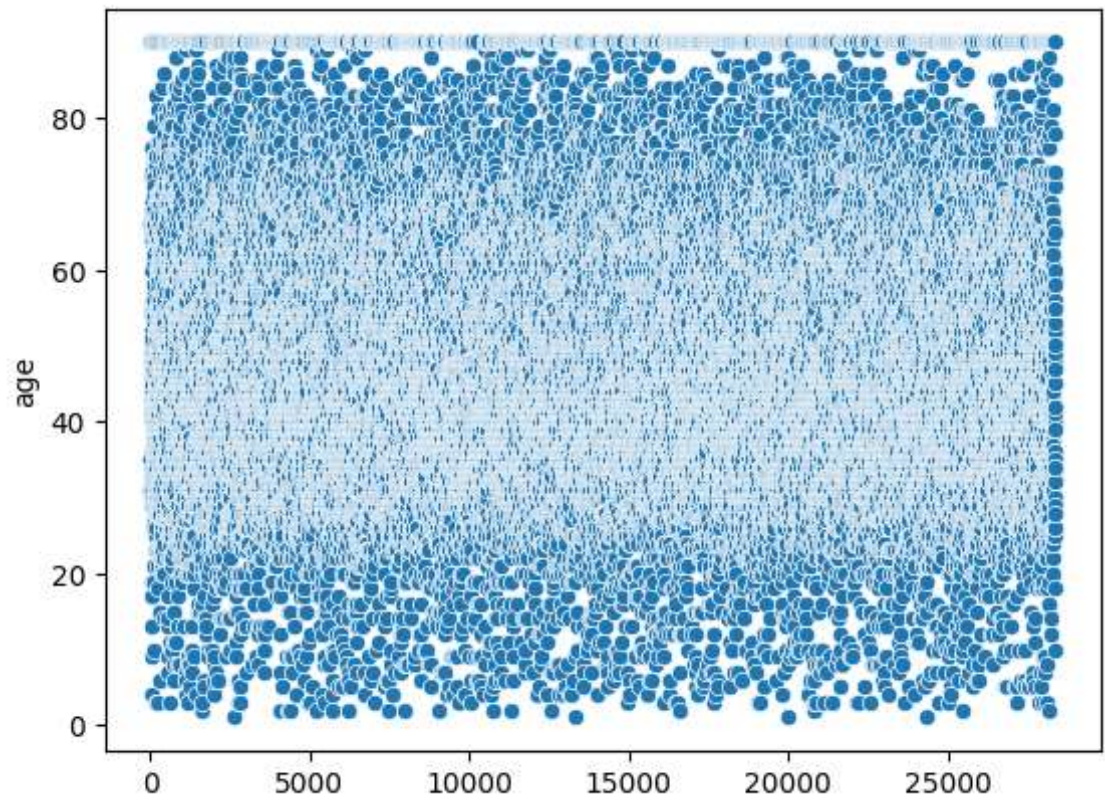
```
In [32]: sns.boxplot(df['age'])
```

```
Out[32]: <AxesSubplot: >
```



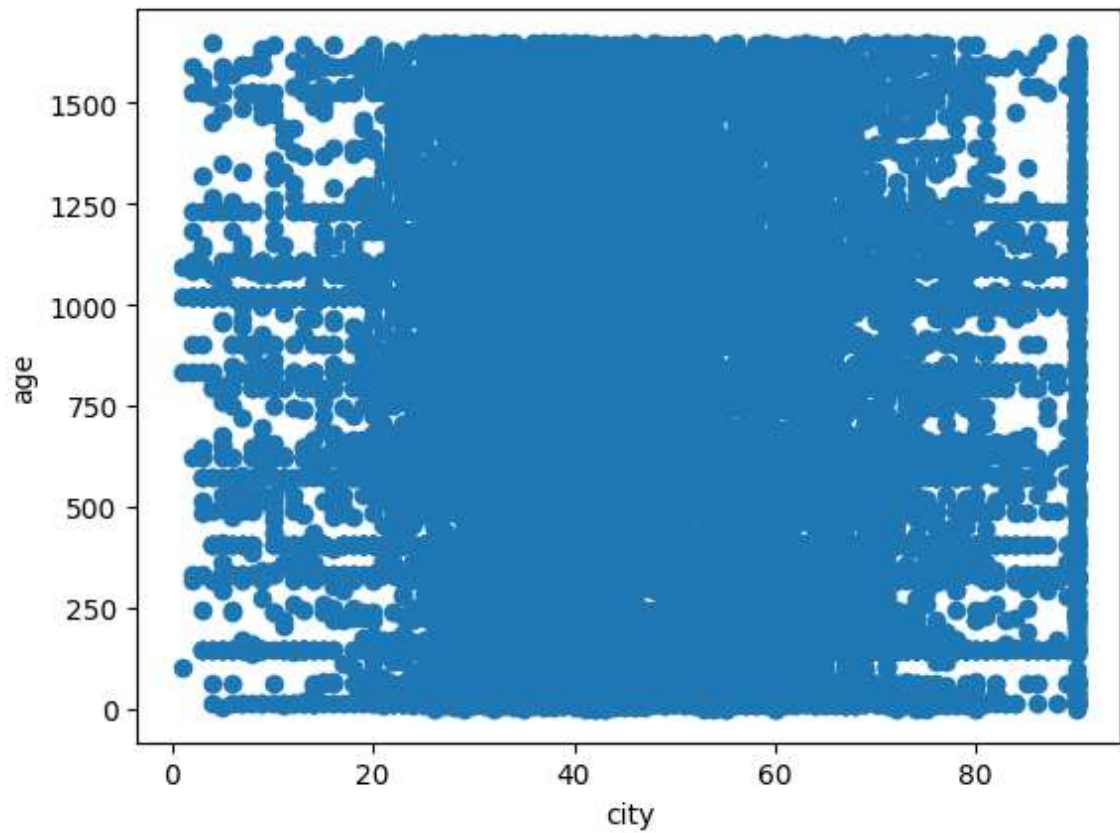
```
In [33]: sns.scatterplot(df['age'])
```

```
Out[33]: <AxesSubplot: ylabel='age'>
```



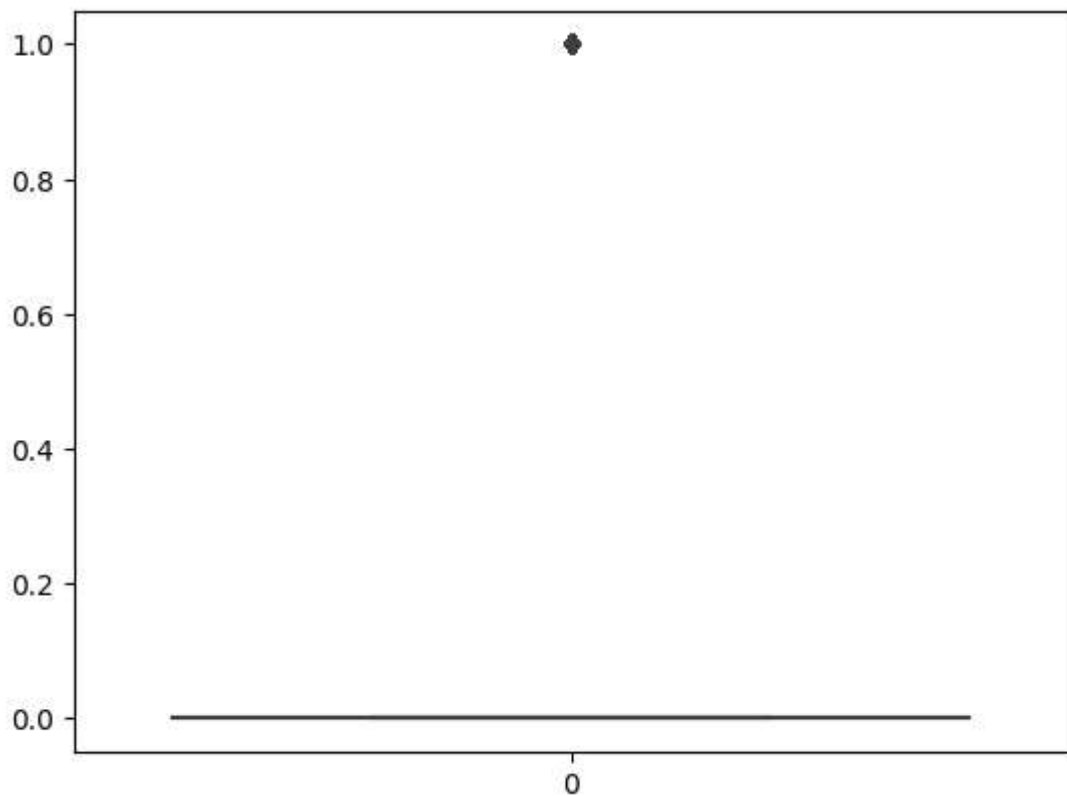
```
In [35]: plt.scatter(y=df['city'],x=df['age'])  
plt.xlabel("city")  
plt.ylabel("age")
```

```
Out[35]: Text(0, 0.5, 'age')
```



```
In [36]: sns.boxplot(df['churn'])
```

```
Out[36]: <AxesSubplot: >
```



```
In [39]: Q1=np.percentile(df['age'],25)
Q3=np.percentile(df['age'],75)
```

```
In [40]: Q1,Q3
```

```
Out[40]: (36.0, 60.0)
```

```
In [41]: IQR=Q3-Q1
```

```
In [42]: IQR
```

```
Out[42]: 24.0
```

```
In [43]: lower_bound=Q1-1.5*IQR
upper_bound=Q3+1.5*IQR
```

```
In [44]: lower_bound,upper_bound
```

```
Out[44]: (0.0, 96.0)
```

```
In [45]: outliers_rows=df[(df['age'] < lower_bound) | (df['age'] > upper_bound)]
```

```
In [46]: outliers_rows.shape
```

```
Out[46]: (0, 7)
```

```
In [48]: outliers_rows
```

```
Out[48]:
```

Unnamed: 0	age	gender	dependents	occupation	city	churn
------------	-----	--------	------------	------------	------	-------

```
In [49]: df['age']=np.where(
    df['age']>upper_bound,
    upper_bound,np.where(
    df['age']<lower_bound,
    lower_bound,
    df['age']
    )
)
```

```
In [50]: outliers_rows=df[(df['age'] < lower_bound) | (df['age'] > upper_bound)]
```

```
In [51]: outliers_rows
```

```
Out[51]:
```

Unnamed: 0	age	gender	dependents	occupation	city	churn
------------	-----	--------	------------	------------	------	-------

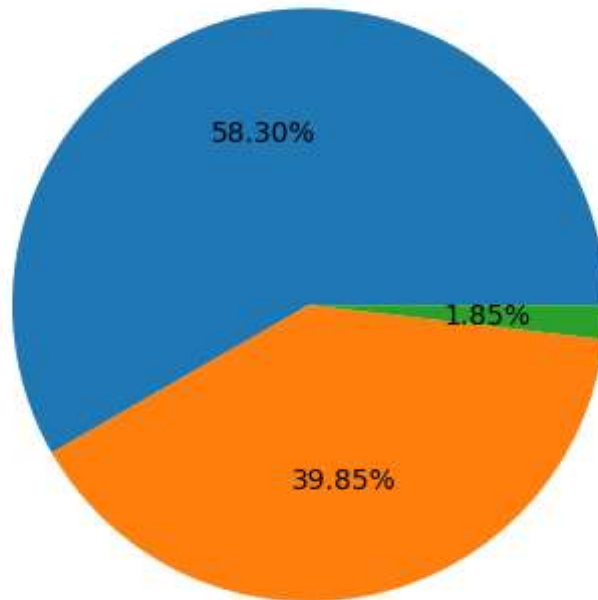
```
In [52]: df['gender'].value_counts()
```

```
Out[52]: Male          16548
Female          11309
gender_mode         525
Name: gender, dtype: int64
```



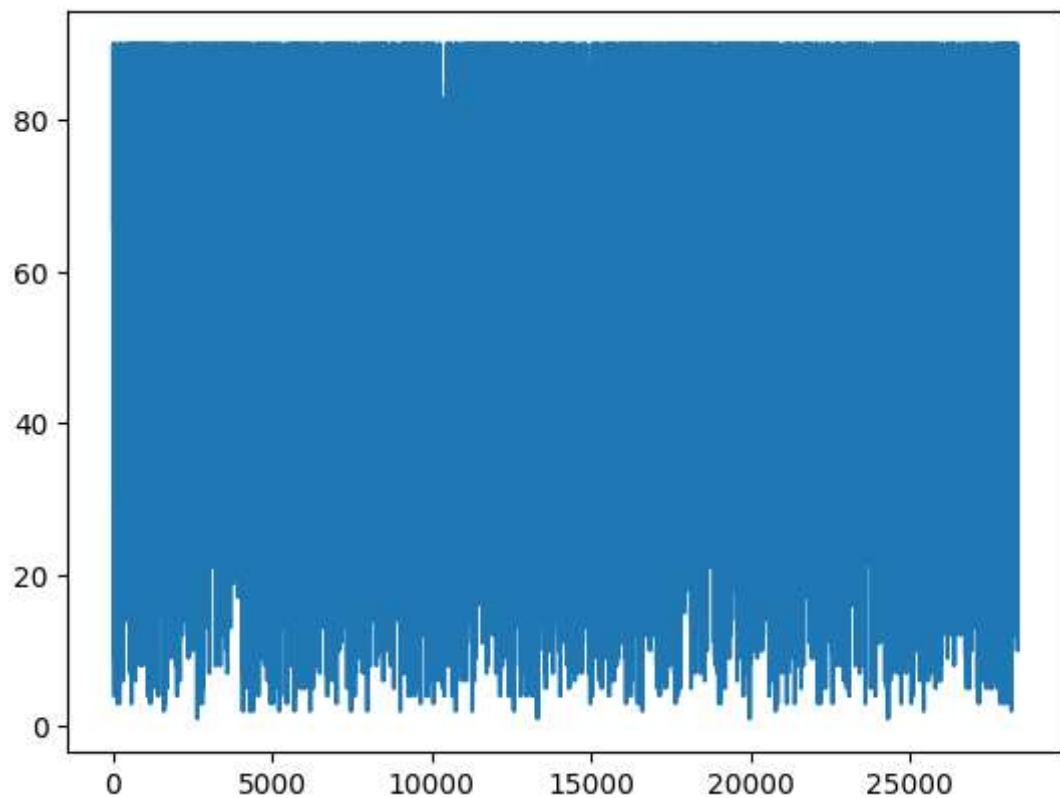
```
In [53]: plt.pie(df['gender'].value_counts(),autopct="%1.2f%%")  
plt.suptitle("Pie chart of Gender")  
plt.show()
```

Pie chart of Gender



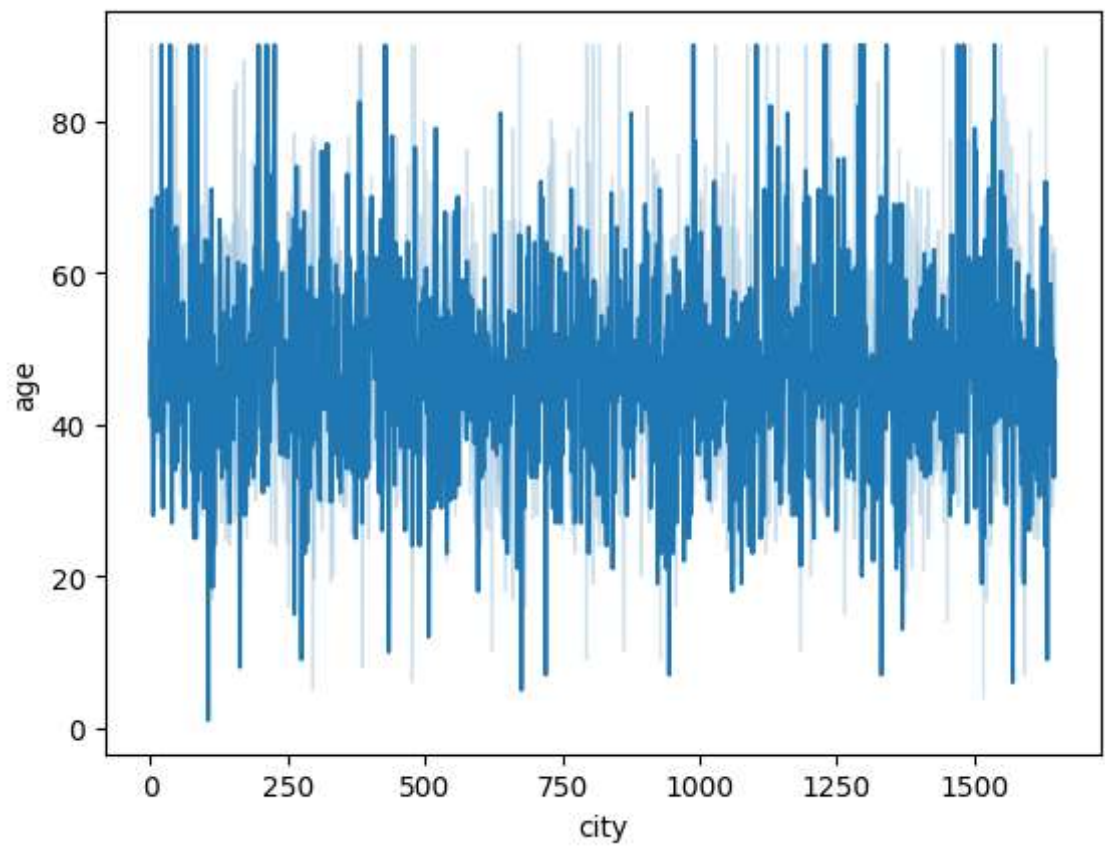
```
In [54]: plt.plot(df['age'])
```

```
Out[54]: [<matplotlib.lines.Line2D at 0x22a63dd2530>]
```

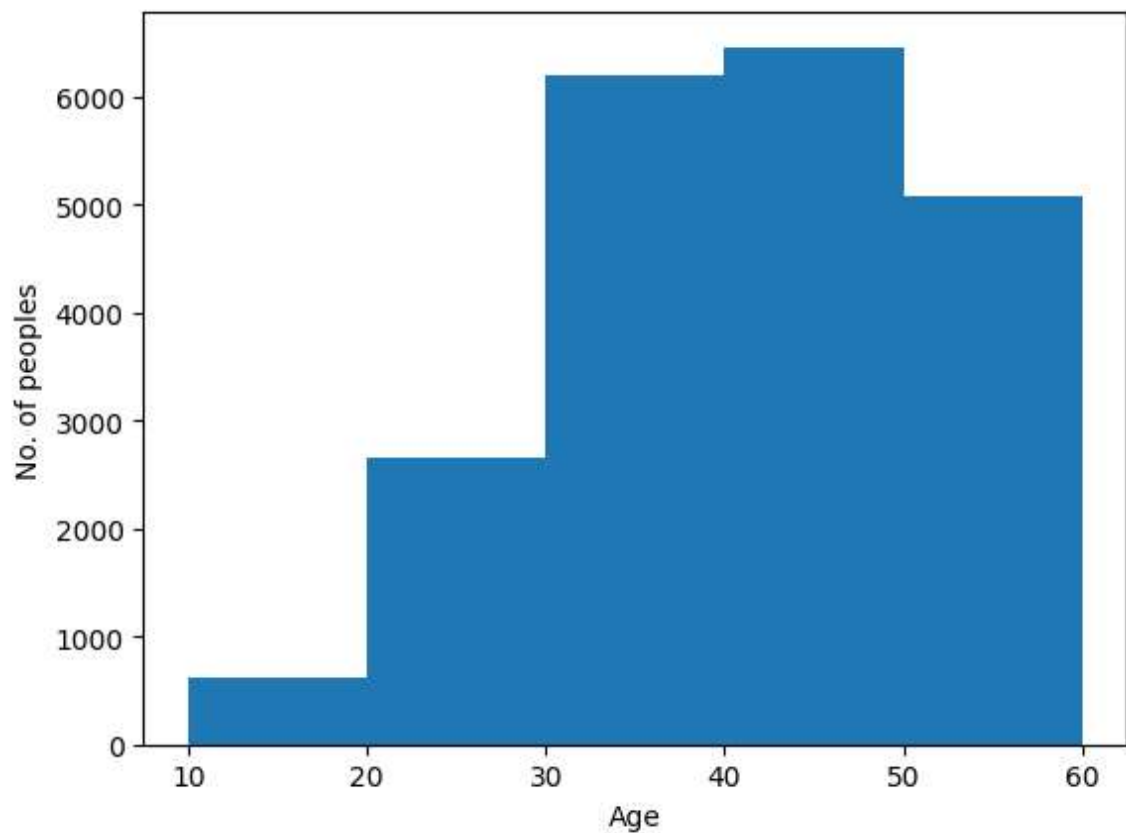


```
In [55]: sns.lineplot(y=df['age'],x=df['city'])
```

```
Out[55]: <AxesSubplot: xlabel='city', ylabel='age'>
```

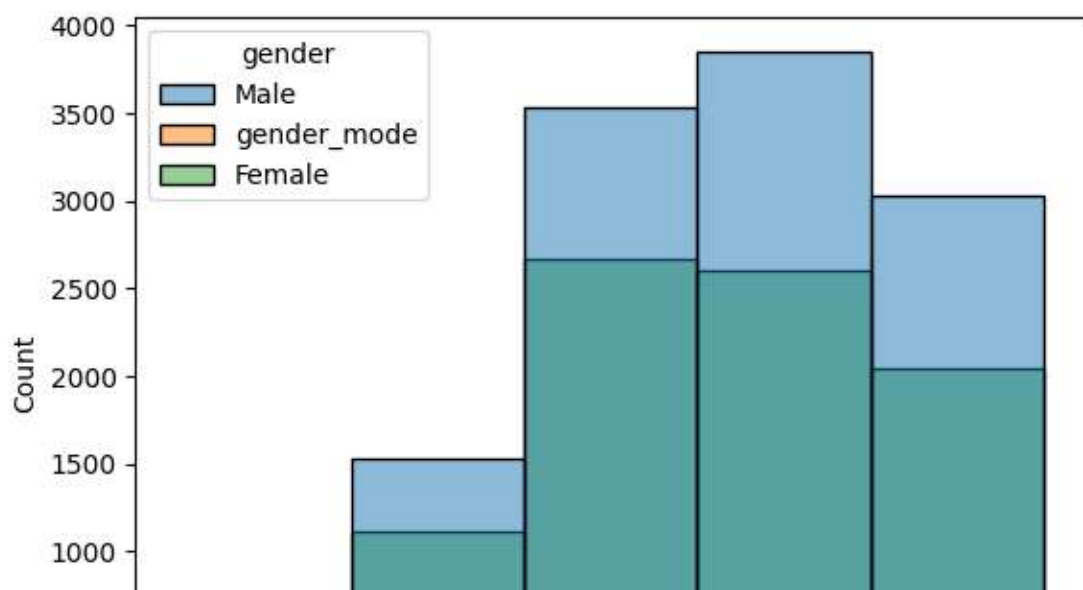


```
In [56]: bins=[10,20,30,40,50,60]
plt.hist(x=df["age"],bins=bins)
plt.xlabel("Age")
plt.ylabel("No. of peoples")
plt.show()
```



```
In [57]: sns.histplot(x=df['age'],bins=bins,hue=df['gender'])
```

Out[57]: <AxesSubplot: xlabel='age', ylabel='Count'>



```
In [ ]:
```

