```
In [1]: # import the packages
# read the data
# divide into numerical and categorical
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns

file_path=r'C:\Users\omkar\OneDrive\Documents\Gen_AI\Data_files\Visadataset.csv'
visa_df=pd.read_csv(file_path)

cat=visa_df.select_dtypes(include='object').columns
num=visa_df.select_dtypes(exclude='object').columns
```

In [2]: visa_df[['continent','case_status']]

Out[2]: continent case_status 0 Asia Denied Asia Certified 2 Asia Denied Denied Asia 4 Africa Certified 25475 Asia Certified 25476 Certified Asia 25477 Certified Asia 25478 Asia Certified Certified 25479 Asia

25480 rows × 2 columns

```
In [3]: visa_df['continent'].value_counts()
Out[3]: continent
        Asia
                        16861
                        3732
        Europe
                      3292
        North America
        South America
                        852
        Africa
                          551
        Oceania
                          192
        Name: count, dtype: int64
In [4]: visa_df['case_status'].value_counts()
```

Out[4]: case_status

Certified 17018 Denied 8462

Name: count, dtype: int64

- How many members are certified from Asia
- How many members are denied from Asia

```
In [7]: con1=visa_df['continent']=='Asia'
         con2=visa_df['case_status']=='Certified'
         con3=visa_df['case_status']=='Denied'
         certi_con=con1&con2
         deni_con=con1&con3
         len(visa_df[certi_con]),len(visa_df[deni_con])
 Out[7]: (11012, 5849)
 In [8]: labels=visa_df['continent'].unique()
         certified,denied=[],[]
         for i in labels:
              con1=visa_df['continent']==i
              con2=visa_df['case_status']=='Certified'
             con3=visa_df['case_status']=='Denied'
             certi con=con1&con2
             deni_con=con1&con3
              certified.append(len(visa_df[certi_con]))
              denied.append(len(visa_df[deni_con]))
 In [9]:
         certified
 Out[9]: [11012, 397, 2037, 2957, 493, 122]
In [10]:
         case_labels=sorted(visa_df['case_status'].unique())
In [11]: pd.DataFrame(zip(certified,denied),
                      index=labels,
                      columns=case_labels)
Out[11]:
                         Certified Denied
                   Asia
                           11012
                                    5849
                  Africa
                             397
                                     154
          North America
                            2037
                                    1255
                 Europe
                            2957
                                     775
          South America
                             493
                                     359
                Oceania
                             122
                                      70
```

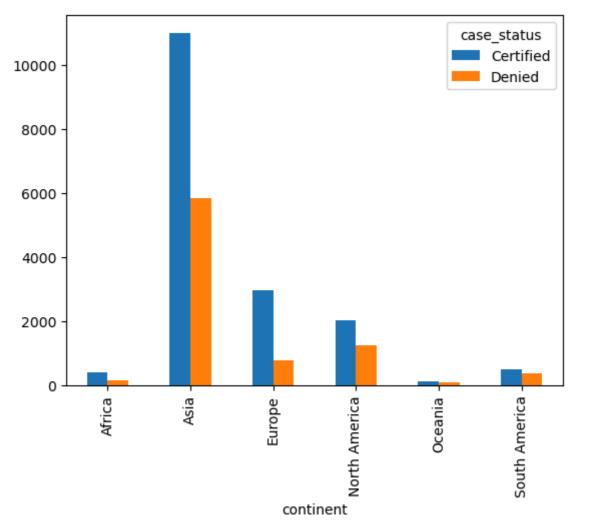
crosstab

```
In [13]: id_col=visa_df['continent']
    col=visa_df['case_status']
    r1=pd.crosstab(id_col,col)
    r1
```

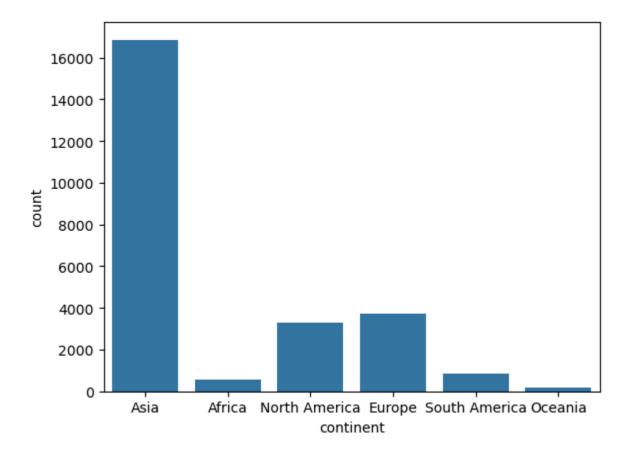
Out[13]: case_status Certified Denied continent 154 **Africa** 397 11012 Asia 5849 Europe 2957 775 **North America** 2037 1255 Oceania 122 70 **South America** 493 359

```
In [14]: r1.plot(kind='bar')
```

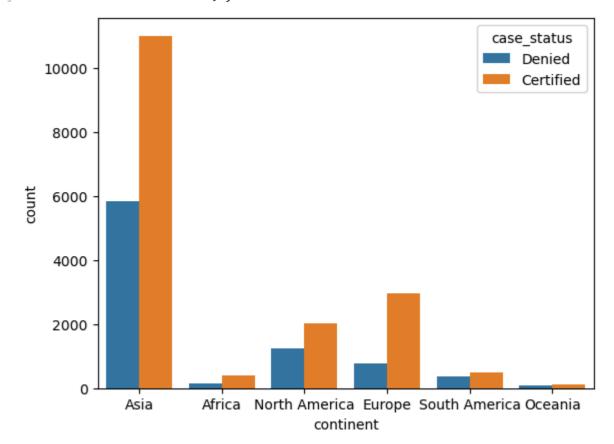
Out[14]: <Axes: xlabel='continent'>



Out[15]: <Axes: xlabel='continent', ylabel='count'>

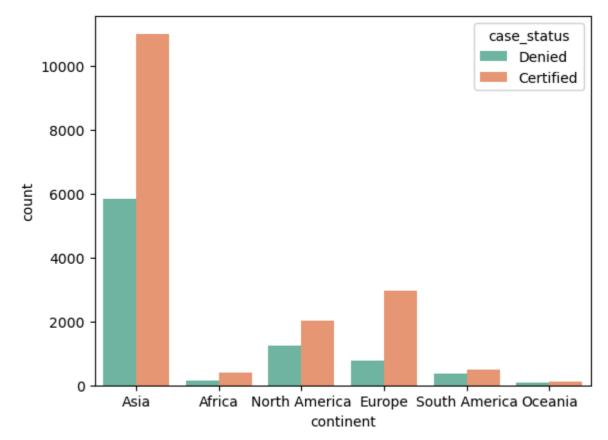


Out[16]: <Axes: xlabel='continent', ylabel='count'>



Out[17]: <Axes: xlabel='continent', ylabel='count'>

r1



```
In [18]: # Scatter plot
# correlation
# heatmaps

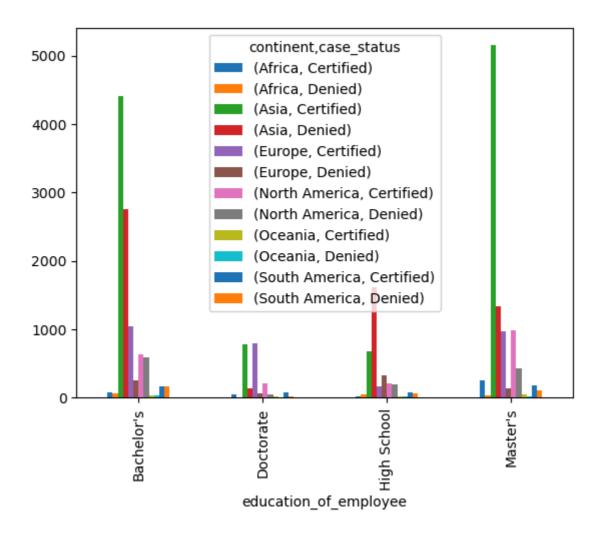
# StreamLit RAG

# Sunday Llma access api key 2hours 9 to 11

In [36]: id_col=visa_df['continent']
col=visa_df['case_status']
r1=pd.crosstab(id_col,col)
```

Out[36]:	case_status	Certified	Denied						
	continent								
	Africa	397	154						
	Asia	11012	5849						
	Europe	2957	775						
	North America	2037	1255						
	Oceania	122	70						
	South America	493	359						
In [38]:	<pre>col1=visa_df[' col2=visa_df[' col3=visa_df[' id_col=col1 cols=[col2,col pd.crosstab(col)</pre>	education case_stat	_of_empl	loyee']					
Out[38]:	education_of_er	nployee	Ва	chelor's	D	octorate	Hig	h School	
	cas	e_status (Certified	Denied	Certified	Denied	Certified	Denied	Certifie
		ontinent							
	CC	Africa	81	62	43	11	23	43	25
	cc		81 4407	62 2761	43 780	11 143	23 676	43 1614	25 514
	CC	Africa							
		Africa Asia	4407	2761	780	143	676	1614	514
	North A	Africa Asia Europe	4407 1040	2761 259	780 788	143 58	676 162	1614 328	514 96
	North A	Africa Asia Europe America	4407 1040 641	2761 259 584	780 788 207	143 58 51	676 162 210	1614 328 191	514 96 97
	North A	Africa Asia Europe America Oceania	4407 1040 641 38	2761 259 584 28	780 788 207 19	143 58 51 3	676 162 210 19	1614 328 191 17	514 96 97 4
In [42]:	North A	Africa Asia Europe America Oceania America continent education case_stat	4407 1040 641 38 160	2761 259 584 28 173	780 788 207 19	143 58 51 3	676 162 210 19	1614 328 191 17	514 96 97 4

Out[44]: <Axes: xlabel='education_of_employee'>



Scatter plots

bar chart : frequency table

• pie chart : relative frequency table

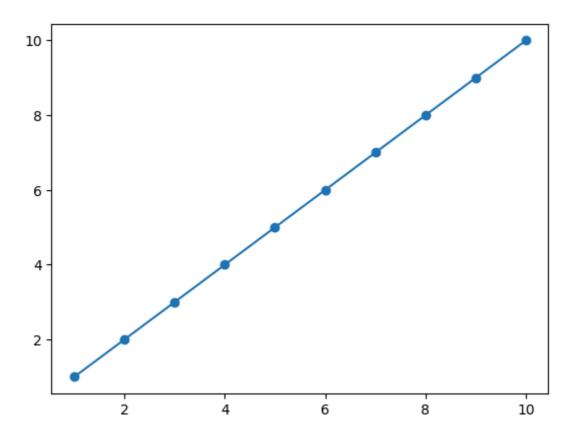
• histogram : frequency distribution plot

box plot : Outlier analysis

• scatterplot : Relation between two Numerical columns

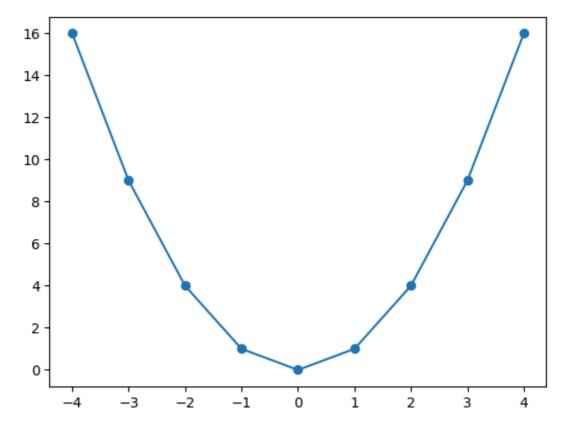
```
In [52]: x=list(range(1,11))
    y=list(range(1,11))
    plt.scatter(x,y)
    plt.plot(x,y)
```

Out[52]: [<matplotlib.lines.Line2D at 0x22f682929f0>]



```
In [54]: x=list(range(-4,5))
    y=[i*i for i in x]
    plt.scatter(x,y)
    plt.plot(x,y)
```

Out[54]: [<matplotlib.lines.Line2D at 0x22f679c5190>]



```
In [60]: col1=visa_df['prevailing_wage']
    col2=visa_df['no_of_employees']
    col3=visa_df['yr_of_estab']
```

```
plt.figure(figsize=(14,5))
plt.subplot(1,3,1).scatter(col1,col2)
plt.title('prevailing_wage and no_of_employees ')
plt.subplot(1,3,2).scatter(col1,col3)
plt.title('prevailing_wage and yr_of_estab ')
plt.subplot(1,3,3).scatter(col2,col3)
plt.title('no_of_employees and yr_of_estab ')
```

Out[60]: Text(0.5, 1.0, 'no_of_employees and yr_of_estab ')



Correlation

In [65]: visa_df.corr(numeric_only=True)

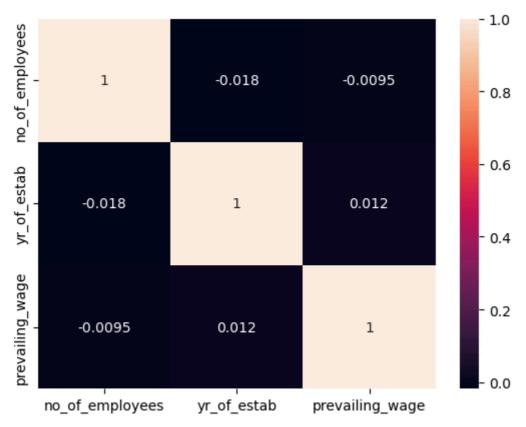
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\cup	иı	۱ <i>-</i>	U	ノン		۰

	no_or_employees	yr_ot_estab	prevailing_wage
no_of_employees	1.000000	-0.017770	-0.009523
yr_of_estab	-0.017770	1.000000	0.012342
prevailing_wage	-0.009523	0.012342	1.000000

Heatmap

In [70]: corr=visa_df.corr(numeric_only=True)
sns.heatmap(corr,annot=True)

Out[70]: <Axes: >



```
In []: # step-1: Wine quality dataset
    # Step-2: apply the correlation : 12x12=144 values
    # Step-3: By seeing this 144 human eyes can't identify which high and which low
    # Step-4: Get the column pair high post, high neg,no relation
    # Step-5: Draw the scatter plot

In [74]: file_path=r'C:\Users\omkar\OneDrive\Documents\Gen_AI\Data_files\winequality_red.
    wine_df=pd.read_csv(file_path)
    wine_df
```

()	7/1	
UULI	/ + 1	

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	sulfur dioxide	total sulfur dioxide	density	рН	sulph
	7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	
	7.8	0.880	0.00	2.6	0.098	25.0	67.0	0.99680	3.20	
- 2	7.8	0.760	0.04	2.3	0.092	15.0	54.0	0.99700	3.26	
3	3 11.2	0.280	0.56	1.9	0.075	17.0	60.0	0.99800	3.16	
	4 7.4	0.700	0.00	1.9	0.076	11.0	34.0	0.99780	3.51	
••	•									
1594	6.2	0.600	0.08	2.0	0.090	32.0	44.0	0.99490	3.45	
159	5.9	0.550	0.10	2.2	0.062	39.0	51.0	0.99512	3.52	
1590	6.3	0.510	0.13	2.3	0.076	29.0	40.0	0.99574	3.42	
1597	7 5.9	0.645	0.12	2.0	0.075	32.0	44.0	0.99547	3.57	
1598	6.0	0.310	0.47	3.6	0.067	18.0	42.0	0.99549	3.39	

1599 rows × 12 columns

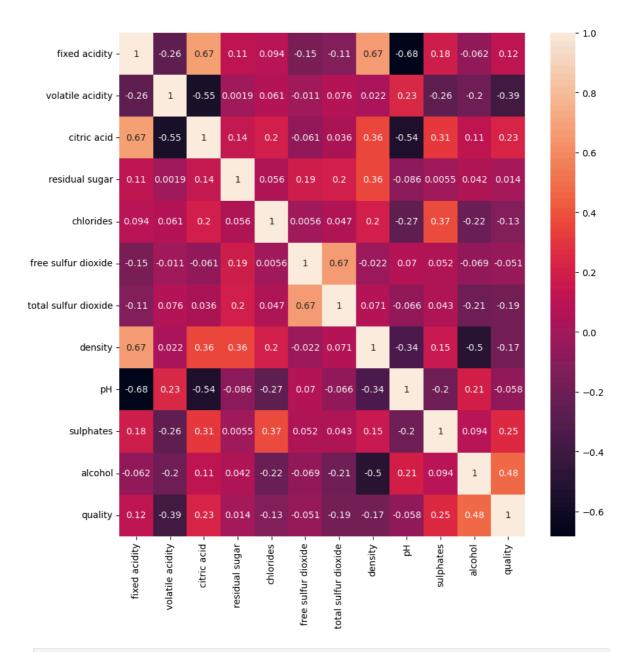
In [76]: wine_df.corr()

_		
() i i +	761	
UU L	1 / 0 1	

	fixed acidity	volatile acidity	citric acid	residual sugar	chlorides	free sulfur dioxide	total sulfur dioxide	
fixed acidity	1.000000	-0.256131	0.671703	0.114777	0.093705	-0.153794	-0.113181	(
volatile acidity	-0.256131	1.000000	-0.552496	0.001918	0.061298	-0.010504	0.076470	(
citric acid	0.671703	-0.552496	1.000000	0.143577	0.203823	-0.060978	0.035533	(
residual sugar	0.114777	0.001918	0.143577	1.000000	0.055610	0.187049	0.203028	(
chlorides	0.093705	0.061298	0.203823	0.055610	1.000000	0.005562	0.047400	(
free sulfur dioxide	-0.153794	-0.010504	-0.060978	0.187049	0.005562	1.000000	0.667666	-(
total sulfur dioxide	-0.113181	0.076470	0.035533	0.203028	0.047400	0.667666	1.000000	(
density	0.668047	0.022026	0.364947	0.355283	0.200632	-0.021946	0.071269	-
рН	-0.682978	0.234937	-0.541904	-0.085652	-0.265026	0.070377	-0.066495	-(
sulphates	0.183006	-0.260987	0.312770	0.005527	0.371260	0.051658	0.042947	(
alcohol	-0.061668	-0.202288	0.109903	0.042075	-0.221141	-0.069408	-0.205654	-(
quality	0.124052	-0.390558	0.226373	0.013732	-0.128907	-0.050656	-0.185100	-(
4								

In [80]: plt.figure(figsize=(10,10))
 sns.heatmap(wine_df.corr(),annot=True)

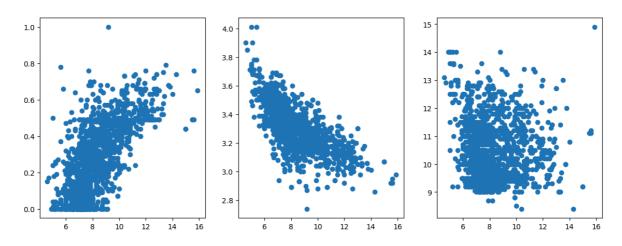
Out[80]: <Axes: >



In []: fixed acidity has 67% postive relation with citiric acide and 68% negative ph
no relation with alchol

```
In [84]: col1=wine_df['fixed acidity']
    col2=wine_df['citric acid']
    col3=wine_df['pH']
    col4=wine_df['alcohol']
    plt.figure(figsize=(14,5))
    plt.subplot(1,3,1).scatter(col1,col2)
    plt.subplot(1,3,2).scatter(col1,col3)
    plt.subplot(1,3,3).scatter(col1,col4)
```

Out[84]: <matplotlib.collections.PathCollection at 0x22f70615580>



In []: