Import packages

In [1]:
```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
```

Read the data

In [2]:
```
visa_df=pd.read_csv(r"C:\Users\omkar\OneDrive\Documents\Data science\Naresh IT\N
visa_df.head(2)
```

Out[2]:

	case_id	continent	education_of_employee	has_job_experience	requires_job_training
0	EZYV01	Asia	High School	N	N
1	EZYV02	Asia	Master's	Y	N

Select the numerical coulmns

In [3]:
```
visa_df.select_dtypes(exclude='object').columns
```

Out[3]:
```
Index(['no_of_employees', 'yr_of_estab', 'prevailing_wage'], dtype='object')
```

prevailing_wage

- len
- max
- min
- mean
- median
- std
- 75%
- 50%
- 25%

count or len

In [5]:
```
len(visa_df['prevailing_wage'])
```

Out[5]:
```
25480
```

max

In [6]:
```
max(visa_df['prevailing_wage']) # Keyword
```

Out[6]: 319210.27

In [7]:
```
visa_df['prevailing_wage'].max() # Pandas
```

Out[7]: 319210.27

In [8]:
```
np.max(visa_df['prevailing_wage']) # numpy
```

Out[8]: 319210.27

min

In [9]:
```
min(visa_df['prevailing_wage'])
```

Out[9]: 2.1367

In [10]:
```
visa_df['prevailing_wage'].min()
```

Out[10]: 2.1367

In [11]:
```
np.min(visa_df['prevailing_wage'])
```

Out[11]: 2.1367

In [ ]:
```
#instead of len can we use nunique ?
#how many uniques values different
#how many total values different
```

mean

In [12]:
```
visa_df['prevailing_wage'].mean()
```

Out[12]: 74455.81459209183

In [13]:
```
np.mean(visa_df['prevailing_wage'])
```

Out[13]: 74455.81459209183

median

In [14]:
```
visa_df['prevailing_wage'].median()
```

Out[14]: 70308.20999999999

In [15]:
```
np.median(visa_df['prevailing_wage'])
```

Out[15]: 70308.20999999999

std

In [19]:
```
visa_df['prevailing_wage'].std()
```

Out[19]: 52815.94232687357

In [20]:
```
np.std(visa_df['prevailing_wage'])
```

Out[20]: 52814.90589711402

Mode is not good option because it is numerical variable

In [24]:
```
##All together
wage_count=round(len(visa_df['prevailing_wage']),2)
wage_min=round(visa_df['prevailing_wage'].min(),2)
wage_max=round(visa_df['prevailing_wage'].max(),2)
wage_mean=round(visa_df['prevailing_wage'].mean(),2)
wage_median=round(visa_df['prevailing_wage'].median(),2)
wage_std=round(visa_df['prevailing_wage'].std(),2)
list_values=[wage_count,wage_min,wage_max,
             wage_mean,wage_median,wage_std]
index_val=['count','min','max','mean','median','std']
pd.DataFrame(list_values,
             columns=['prevailing_wage'],
             index=index_val)
```

Out[24]:

	prevailing_wage
count	25480.00
min	2.14
max	319210.27
mean	74455.81
median	70308.21
std	52815.94

Percentile and Quantile

- Percentile:
  - np.percentile()
  - It will take two arguments
    - data :a
    - percentile: q the values varies from 0 to 100
    - if you want 50P data q=50
- Quantile:
  - np.quantile()
  - It will take two arguments
    - data :a

- percentile: q the values varies from 0 to 1
- if you want 50p q=0.5

25p-50p-75p

In [28]:
```
wage_25p=round(np.percentile(visa_df['prevailing_wage'],25),2)
wage_50p=round(np.percentile(visa_df['prevailing_wage'],50),2)
wage_75p=round(np.percentile(visa_df['prevailing_wage'],75),2)

print(f"the 25% data is {wage_25p}")
print(f"the 50% data is {wage_50p}")
print(f"the 75% data is {wage_75p}")
```

```
the 25% data is 34015.48
the 50% data is 70308.21
the 75% data is 107735.51
```

In [29]:
```
345.89678
```

Out[29]: 345.89678

In [30]:
```
round(345.89678,2)
```

Out[30]: 345.9

In [33]:
```
wage_25p=round(np.quantile(visa_df['prevailing_wage'],0.25),2)
wage_50p=round(np.quantile(visa_df['prevailing_wage'],0.50),2)
wage_75p=round(np.quantile(visa_df['prevailing_wage'],0.75),2)

print(f"the 25% data is {wage_25p}")
print(f"the 50% data is {wage_50p}")
print(f"the 75% data is {wage_75p}")
```

```
the 25% data is 34015.48
the 50% data is 70308.21
the 75% data is 107735.51
```

Understand the percentiles

- defination of 25percentile
  - there 25% of employees has salary less than 34015
  - total employees= 25480
  - 25% of employees= 25*25480/100= 6370
  - 6370 employees salary less than 34015

In [38]:
```
con=visa_df['prevailing_wage']<34015
len(visa_df[con])
```

Out[38]: 6370

In [39]:
```
con=visa_df['prevailing_wage']<wage_25p
len(visa_df[con])
```

Out[39]: 6370

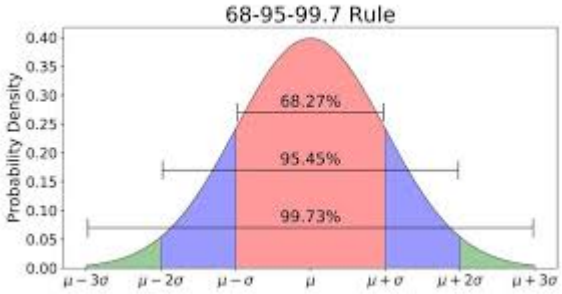
In [41]:
```
con=visa_df['prevailing_wage']<wage_50p
len(visa_df[con])

#50*25480/100
```

Out[41]: 12740.0

In [42]:
```
con=visa_df['prevailing_wage']<wage_75p
len(visa_df[con])
```

Out[42]: 19110

Emperical rule (68-95-99.7)



- First calculate mean value
- Second calculate std value
- Con1: mean-1*std
- Con2: mean+1*std
- If you apply above conditions on wage data , the output count should be equal to 68percentile data
- 68% of total employees: 17326

In [60]:
```
v1=wage_mean-1*wage_std
v2=wage_mean+1*wage_std
con1=visa_df['prevailing_wage']>v1
con2=visa_df['prevailing_wage']<v2

count1=len(visa_df[con1 & con2])
#####################################################################
count1 ,68*25480/100
```

Out[60]: (17171, 17326.4)

In [62]:
```
v1=wage_mean-2*wage_std
v2=wage_mean+2*wage_std
con1=visa_df['prevailing_wage']>v1
con2=visa_df['prevailing_wage']<v2
```

```
count1=len(visa_df[con1 & con2])
#################################################################
count1 ,95*25480/100
```

Out[62]: (24582, 24206.0)

In [63]:
```
v1=wage_mean-3*wage_std
v2=wage_mean+3*wage_std
con1=visa_df['prevailing_wage']>v1
con2=visa_df['prevailing_wage']<v2

count1=len(visa_df[con1 & con2])
#################################################################
count1 ,99.7*25480/100
```

Out[63]: (25186, 25403.56)

In [ ]: