

Untitled14

May 26, 2023

```
[16]: import os
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
import warnings
import statsmodels.api as sm
import plotly.express as px
```

```
[17]: #define working directory
os.chdir("E:/Kurs/Data Glacier/Week 7")
```

```
[18]: # read and merge datasets
df1 = pd.read_csv('Age_Country.csv')
df2 = pd.read_csv('Education_Languages.csv')
df3 = pd.read_csv('Gender_SatRate.csv')
df4 = pd.read_csv('Marriage_Children.csv')

merged_df = pd.merge(df1, df2, on='Name', how='outer')
merged_df = pd.merge(merged_df, df3, on='Name', how='outer')
merged_df = pd.merge(merged_df, df4, on='Name', how='outer')

# write the merged dataframe to a new CSV file
merged_df.to_csv('master_data.csv', index=False)
```

```
C:\Users\ejot9\AppData\Local\Temp\ipykernel_19264\1445709634.py:9:
FutureWarning: Passing 'suffixes' which cause duplicate columns {'Email_x'} in
the result is deprecated and will raise a MergeError in a future version.
merged_df = pd.merge(merged_df, df4, on='Name', how='outer')
```

```
[19]: #Read the master data
df = pd.read_csv("master_data.csv")
```

```
[7]: # First 5 in our dataset
df.head()
```

```
[7]:
```

	Name	Email_x	Age	Country	\
0	Landon Patel	landonpatel@gmail.com	32	United States	
1	Cassidy Turner	cassidyturner@yahoo.com	24	Canada	
2	Axl Thompson	axl.thompson@hotmail.com	43	United States	
3	Ivy Wilson	ivywilson@gmail.com	58	United Kingdom	
4	Reece Scott	reece.scott@outlook.com	26	Australia	

	Email_y	Education Level	Number of Languages Spoken	\
0	landonpatel@gmail.com	Bachelor's Degree	2	
1	cassidyturner@yahoo.com	Master's Degree	1	
2	axl.thompson@hotmail.com	High School	3	
3	ivywilson@gmail.com	Bachelor's Degree	4	
4	reece.scott@outlook.com	Doctorate	4	

	Email_x.1	Gender	Satisfaction Rate	\
0	landonpatel@gmail.com	Male	8	
1	cassidyturner@yahoo.com	Non-binary	9	
2	axl.thompson@hotmail.com	Male	7	
3	ivywilson@gmail.com	Female	6	
4	reece.scott@outlook.com	Male	4	

	Email_y.1	Marital Status	Number of Children
0	landonpatel@gmail.com	Single	3
1	cassidyturner@yahoo.com	Married	4
2	axl.thompson@hotmail.com	Single	1
3	ivywilson@gmail.com	Divorced	0
4	reece.scott@outlook.com	Married	2

```
[10]: #Last 5 rows in our dataset
df.tail()
```

```
[10]:
```

	Name	Email_x	Age	Country	\
82	Ryann Davis	ryann.davis@gmail.com	24	Canada	
83	Lara Johnson	lara.johnson@yahoo.com	54	United States	
84	Paislee King	paislee_king@hotmail.com	61	United Kingdom	
85	Kole Scott	kole_scott@gmail.com	63	Australia	
86	Alaia Taylor	alaia_taylor@yahoo.com	23	United States	

	Email_y	Education Level	Number of Languages Spoken	\
82	ryann.davis@gmail.com	Bachelor's Degree	1	
83	lara.johnson@yahoo.com	Master's Degree	3	
84	paislee_king@hotmail.com	High School	1	
85	kole_scott@gmail.com	Bachelor's Degree	4	
86	alaia_taylor@yahoo.com	Doctorate	2	

	Email_x.1	Gender	Satisfaction Rate	\
82	ryann.davis@gmail.com	Female	10	

83	lara.johnson@yahoo.com	Female	3
84	paislee_king@hotmail.com	Female	4
85	kole_scott@gmail.com	Male	8
86	alaia_taylor@yahoo.com	Female	2

	Email_y.1	Marital Status	Number of Children
82	ryann.davis@gmail.com	Single	0
83	lara.johnson@yahoo.com	Married	2
84	paislee_king@hotmail.com	Single	1
85	kole_scott@gmail.com	Married	3
86	alaia_taylor@yahoo.com	Married	4

```
[11]: #Master Data General information
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 87 entries, 0 to 86
Data columns (total 13 columns):
#   Column                                Non-Null Count  Dtype
---  -
0   Name                                  87 non-null    object
1   Email_x                              87 non-null    object
2   Age                                   87 non-null    int64
3   Country                              87 non-null    object
4   Email_y                              87 non-null    object
5   Education Level                      87 non-null    object
6   Number of Languages Spoken           87 non-null    int64
7   Email_x.1                            87 non-null    object
8   Gender                               87 non-null    object
9   Satisfaction Rate                    87 non-null    int64
10  Email_y.1                            87 non-null    object
11  Marital Status                       87 non-null    object
12  Number of Children                   87 non-null    int64
dtypes: int64(4), object(9)
memory usage: 9.0+ KB
```

```
[12]: #Correlation between numerical columns
df.corr()
```

```
[12]:
```

	Age	Number of Languages Spoken	\
Age	1.000000		0.125291
Number of Languages Spoken	0.125291	1.000000	
Satisfaction Rate	0.071077		-0.090605
Number of Children	-0.081498		-0.074881

	Satisfaction Rate	Number of Children
Age	0.071077	-0.081498
Number of Languages Spoken	-0.090605	-0.074881

Satisfaction Rate	1.000000	-0.035048
Number of Children	-0.035048	1.000000

```
[15]: #Size of Data
df.size
```

```
[15]: 1131
```

```
[26]: #Import columns for duplication test
duplicates = df.duplicated()
```

```
[27]: # Check for Duplicates
duplicate_rows = df[duplicates]
print(duplicate_rows)
```

Empty DataFrame

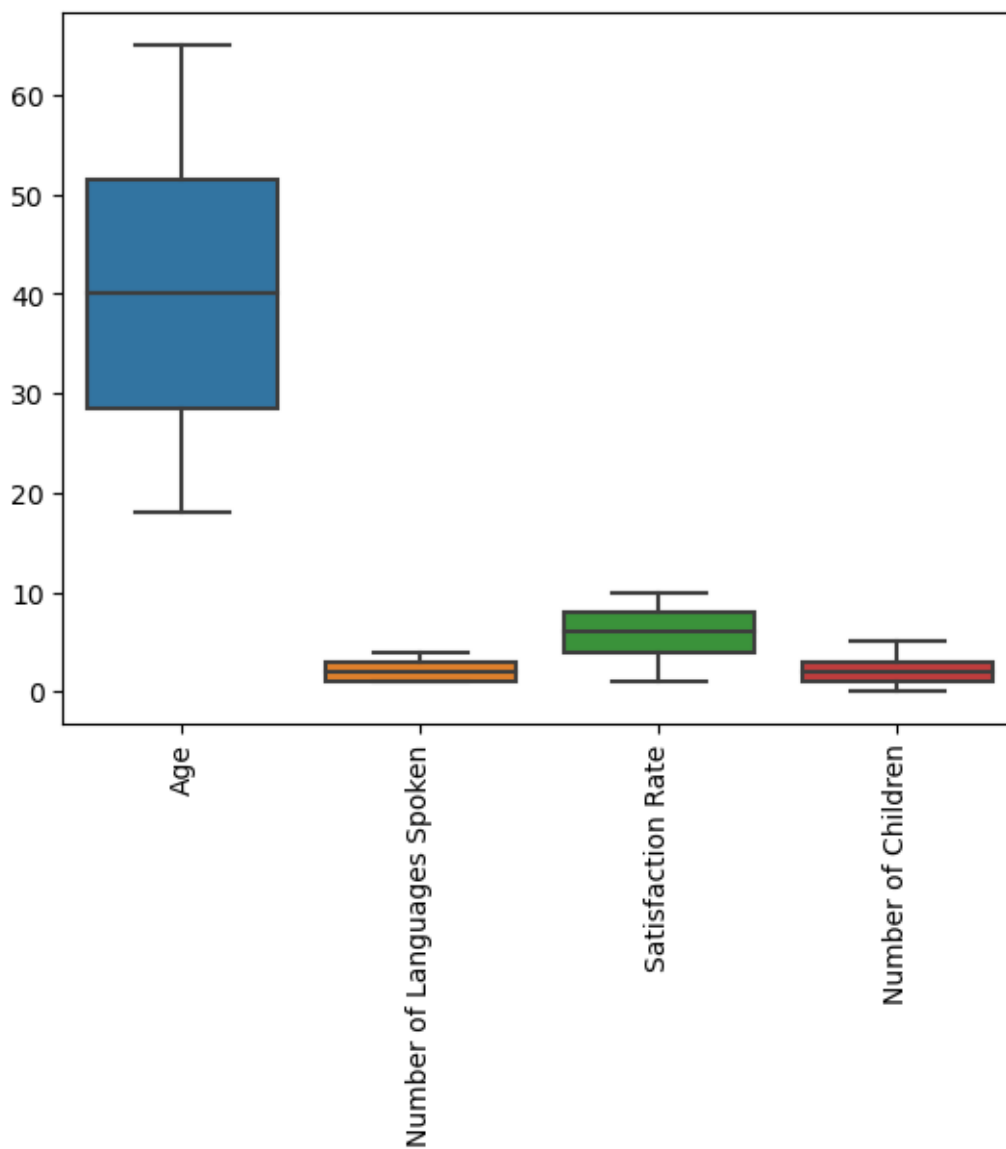
Columns: [Name, Email_x, Age, Country, Email_y, Education Level, Number of Languages Spoken, Email_x.1, Gender, Satisfaction Rate, Email_y.1, Marital Status, Number of Children]

Index: []

```
[38]: #Check for outliers
ax = sns.boxplot(data=df)

# Rotate the x-axis labels
ax.xaxis.set_ticklabels(ax.xaxis.get_ticklabels(), rotation=90)
```

```
[38]: [Text(0, 0, 'Age'),
Text(1, 0, 'Number of Languages Spoken'),
Text(2, 0, 'Satisfaction Rate'),
Text(3, 0, 'Number of Children')]
```



```
[39]: # Check for missing values
missing_values = df.isnull().sum()
```

```
[40]: # Display the count of missing values for each column
print(missing_values)
```

```
Name          0
Email_x       0
Age           0
Country       0
Email_y       0
Education Level 0
```

```

Number of Languages Spoken    0
Email_x.1                     0
Gender                         0
Satisfaction Rate             0
Email_y.1                     0
Marital Status                0
Number of Children            0
dtype: int64

```

```

[42]: # Select numeric columns
numeric_columns = df.select_dtypes(include='number')

```

```

[43]: # Calculate the skewness for each numeric column
skewness = numeric_columns.skew()

```

```

[44]: # Display the skewness values
print(skewness)

```

```

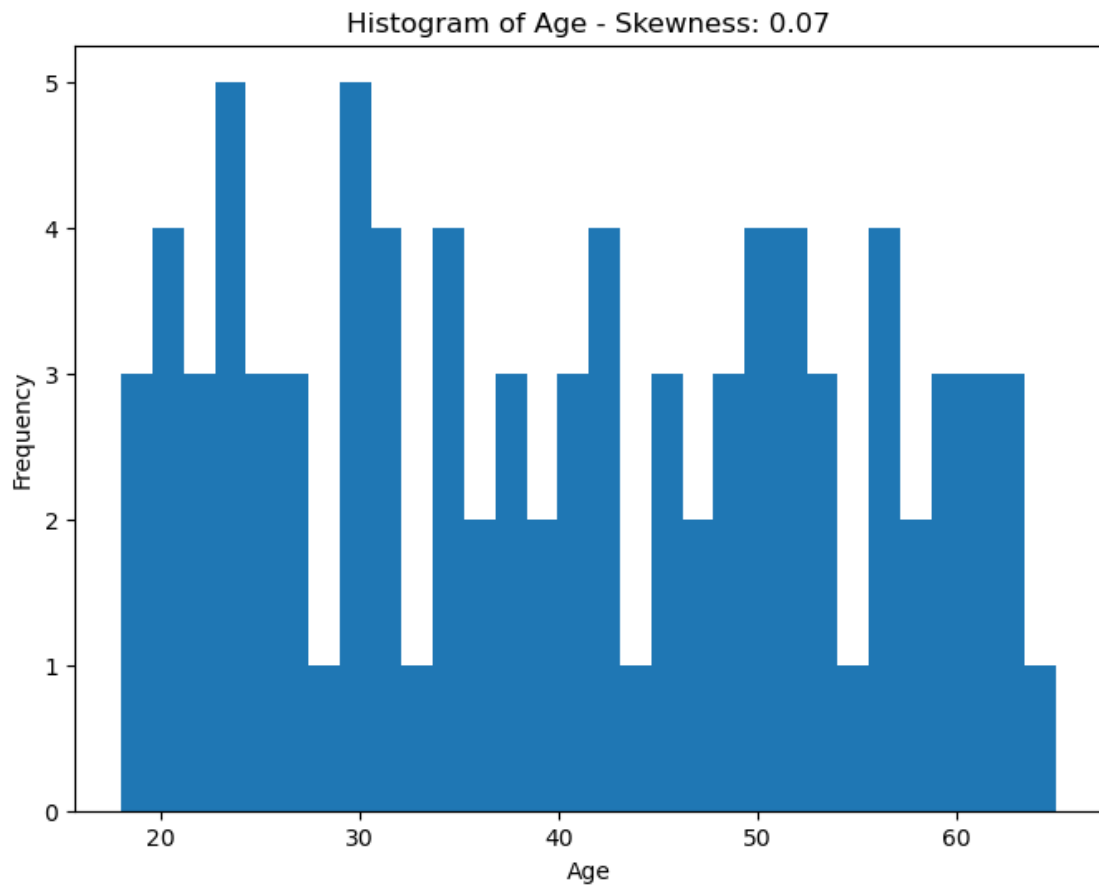
Age                           0.071673
Number of Languages Spoken    0.224759
Satisfaction Rate             -0.178100
Number of Children            0.330569
dtype: float64

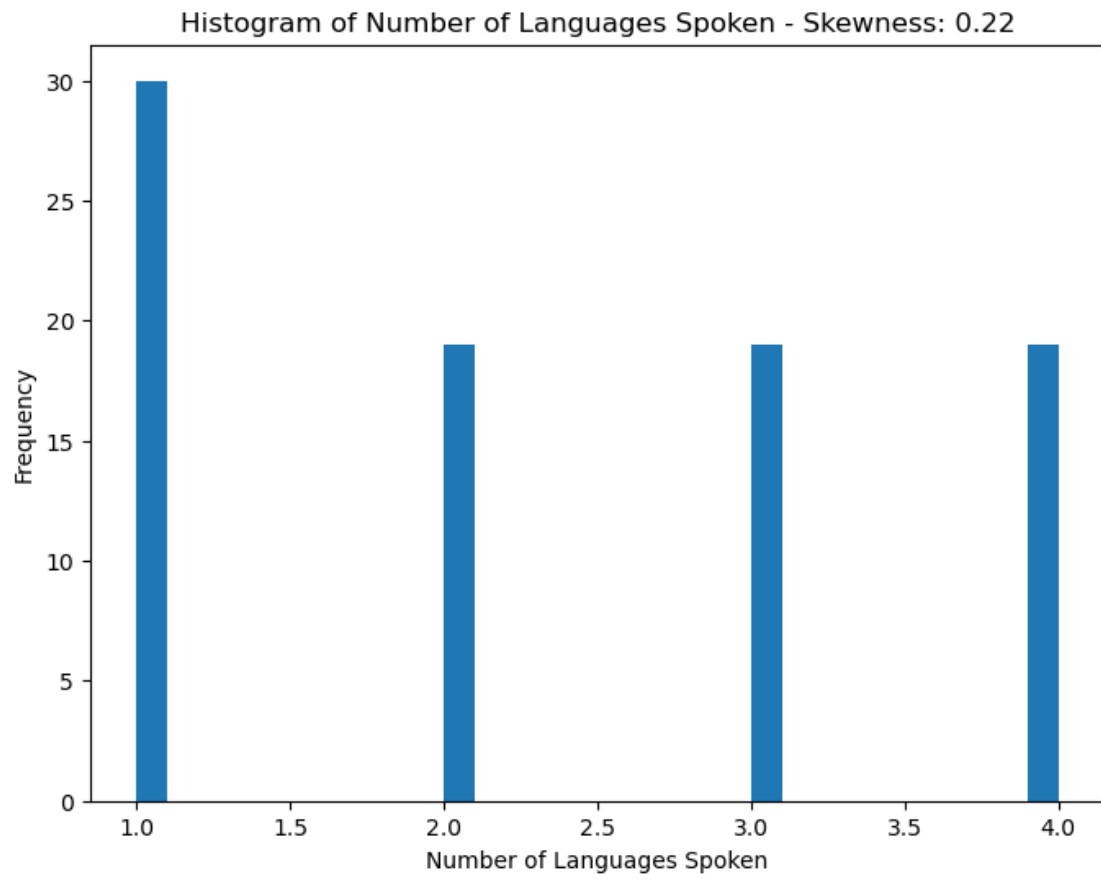
```

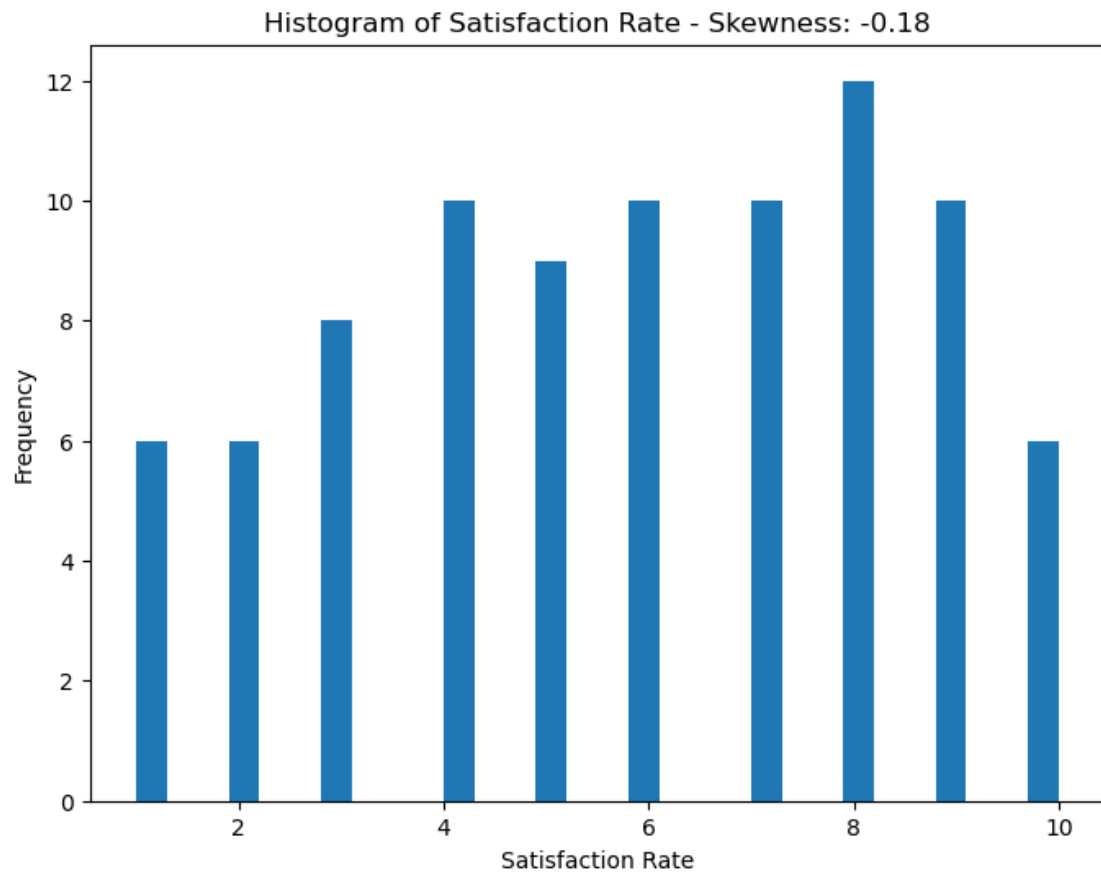
```

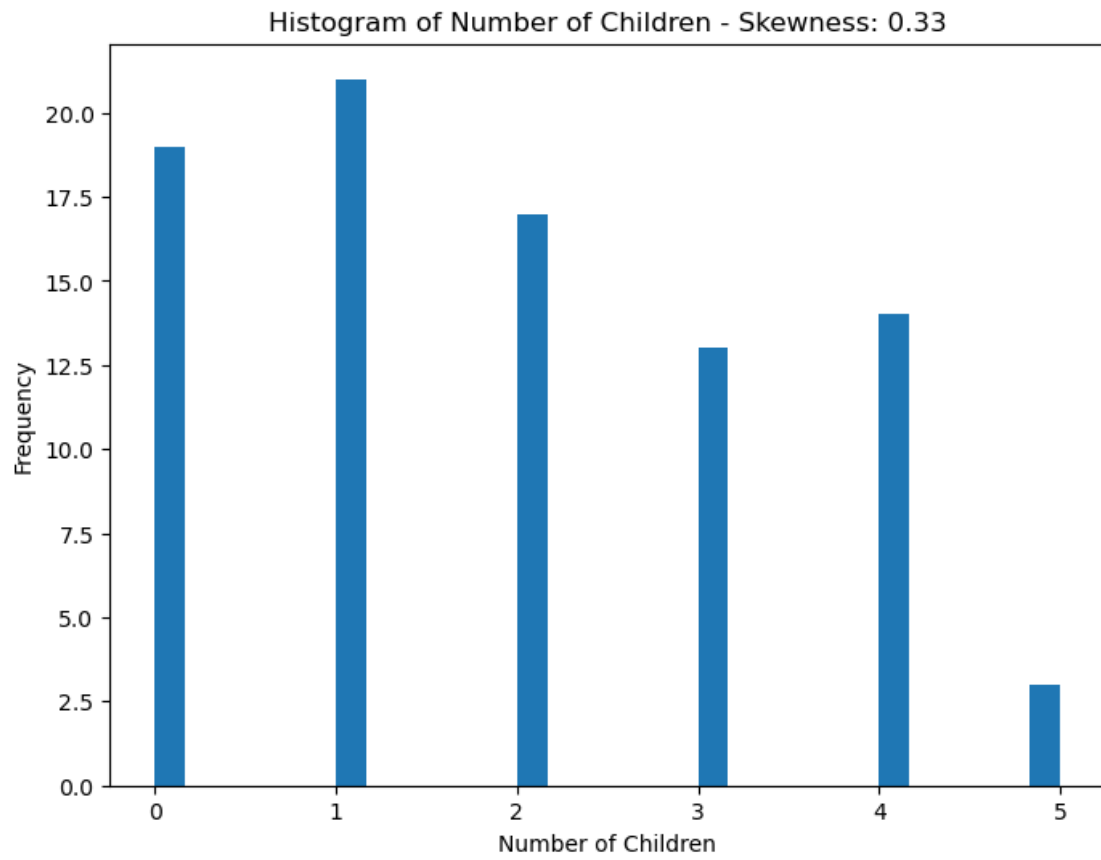
[52]: # Iterate over each numeric column and plot histogram
for column in numeric_columns:
    plt.figure(figsize=(8, 6))
    plt.hist(df[column], bins=30)
    plt.xlabel(column)
    plt.ylabel('Frequency')
    plt.title(f'Histogram of {column} - Skewness: {round(df[column].skew(),
↵2)})')

```









[]: