

In [1]:

```
print("hello world")
```

hello world

In [2]:

```
import ipywidgets as widgets
from IPython.display import display

slider = widgets.IntSlider(value=5, min=0, max=10, step=1)
display(slider)
```



In [4]:

```
import pandas as pd
df = pd.read_csv('C:\\Users\\TCS\\Downloads\\annual-enterprise-survey-2024-financial-year-provisional-size-bands (1).csv')
df.head()
```

Out[4]:

	year	industry_code_ANZSIC	industry_name_ANZSIC	rme_size_grp	variable	value	unit	Unnamed: 7	Unnamed: 8	Unnamed: 9	Unnamed: 10
0	2011	A	Agriculture, Forestry and Fishing	a_0	Activity unit	46134	COUNT	NaN	NaN	NaN	NaN
1	2011	A	Agriculture, Forestry and Fishing	a_0	Rolling mean employees	0	COUNT	NaN	NaN	NaN	NaN
2	2011	A	Agriculture, Forestry and Fishing	a_0	Salaries and wages paid	279	DOLLARS(millions)	NaN	NaN	NaN	NaN
3	2011	A	Agriculture, Forestry and Fishing	a_0	Sales, government funding, grants and subsidies	8187	DOLLARS(millions)	NaN	NaN	NaN	NaN
4	2011	A	Agriculture, Forestry and Fishing	a_0	Total income	8866	DOLLARS(millions)	NaN	NaN	NaN	NaN

In [9]:

```
import pandas as pd
df = pd.read_excel(r'C:\Users\TCS\Downloads\FSI-2023-DOWNLOAD (1).xlsx')
df.head()
```

Out[9]:

	Country	Year	Rank	Total	S1: Demographic Pressures	S2: Refugees and IDPs	C3: Group Grievance	E3: Human Flight and Brain Drain	E2: Economic Inequality	E1: Economy	P1: State Legitimacy	P2: Public Services	P3: Human Rights	C1: Security Apparatus	C Fractionalized Elit
0	Somalia	2023	1st	111.9	10.0	9.0	8.7	8.6	9.1	9.5	9.6	9.8	9.0	9.5	10.0
1	Yemen	2023	2nd	108.9	9.6	9.6	8.8	6.4	7.9	9.9	9.8	9.6	9.6	8.6	9.9
2	South Sudan	2023	3rd	108.5	9.7	10.0	8.6	6.5	8.6	8.6	9.8	9.7	8.7	9.9	9.9
3	Congo Democratic Republic	2023	4th	107.2	9.7	9.8	9.4	6.4	8.4	8.1	9.3	9.3	9.3	8.8	9.9
4	Syria	2023	5th	107.1	7.4	9.1	9.1	8.0	6.5	9.6	10.0	9.0	9.1	9.4	9.9

In [17]:

```
df.to_excel(r'C:\Users\TCS\Downloads\FSI-2023-DOWNLOAD (1).xlsx', index=False)
```

In [19]:

```
import pandas as pd

df = pd.DataFrame({
    'Name': ['Alice', 'Bob'],
    'Department': ['HR', 'IT'],
    'Salary': [50000, 60000]
})

df.to_excel(r'C:\Users\TCS\Downloads\FSI-2023-DOWNLOAD (1).xlsx', index=False)
```

In []:

In []:

EXP 3

CODE:

```
import pandas as pd
from sklearn.preprocessing import StandardScaler, MinMaxScaler
df = pd.read_csv('people-100.csv')
print("Missing values per column:")
print(df.isnull().sum())
for col in df.select_dtypes(include=['float64', 'int64']):
    df[col].fillna(df[col].mean(), inplace=True)
for col in df.select_dtypes(include=['object']):
    df[col].fillna(df[col].mode()[0], inplace=True)
df.drop_duplicates(inplace=True)
if 'UnnecessaryColumn' in df.columns:
    df.drop('UnnecessaryColumn', axis=1, inplace=True)
if 'Age' in df.columns:
    df['Age'] = df['Age'].astype(int)
numeric_cols = df.select_dtypes(include=['float64', 'int64']).columns
scaler = StandardScaler()
df[numeric_cols] = scaler.fit_transform(df[numeric_cols])
print("\nCleaned data preview:")
print(df.head())
print("\nData types after cleaning:")
print(df.dtypes)
```

OUTPUT:

Missing values per column:

Index	0
User Id	0
First Name	0
Last Name	0
Sex	0
Email	0
Phone	0
Date of birth	0
Job Title	0

dtype: int64

Cleaned data preview:

	Index	User Id	First Name	Last Name	Sex
0	-1.714816	88F7B33d2bcf9f5	Shelby	Terrell	Male
1	-1.680173	f90cD3E76f1A9b9	Phillip	Summers	Female
2	-1.645531	DbeAb8CcdfeFC2c	Kristine	Travis	Male
3	-1.610888	A31Bee3c201ef58	Yesenia	Martinez	Male
4	-1.576245	1bA7A3dc874da3c	Lori	Todd	Male

EXP 4

CODE:

```
import pandas as pd

df = pd.read_excel('/content/data.xlsx')

print("First 5 rows:\n", df.head(), "\n")
print("DataFrame Info:\n")
df.info()
print("\nBasic Statistics:\n", df.describe(), "\n")
print("Rows where Popularity > 1500:\n", df[df['Popularity'] > 1500], "\n")
print("Rows where Make is 'BMW':\n", df[df['Make'] == 'BMW'], "\n")
print("Rows where Popularity > 1500 and MSRP > 60000:\n", df[(df['Popularity'] > 1500) & (df['MSRP'] > 60000)], "\n")

mean_popularity = df['Popularity'].mean()
median_msrp = df['MSRP'].median()
mode_make = df['Make'].mode()[0]

print(f"Mean Popularity: {mean_popularity}")
print(f"Median MSRP: {median_msrp}")
print(f"Mode Make: {mode_make}")

popularity_range = df['Popularity'].max() - df['Popularity'].min()
msrp_variance = df['MSRP'].var()
popularity_std_dev = df['Popularity'].std()

print(f"Popularity Range: {popularity_range}")
print(f"MSRP Variance: {msrp_variance}")
print(f"Popularity Standard Deviation: {popularity_std_dev}")
```

OUTPUT:

First 5 rows:									
	Make	Model	Year	Engine	Fuel Type	Engine HP	\		
0	BMW	1 Series M	2011	premium	unleaded (required)	335.0			
1	BMW	1 Series	2011	premium	unleaded (required)	300.0			
2	BMW	1 Series	2011	premium	unleaded (required)	300.0			
3	BMW	1 Series	2011	premium	unleaded (required)	230.0			
4	BMW	1 Series	2011	premium	unleaded (required)	230.0			
	Engine	Cylinders	Transmission	Type	Driven	Wheels	Number of	Doors	\
0		6.0	MANUAL		rear	wheel drive		2.0	
1		6.0	MANUAL		rear	wheel drive		2.0	
2		6.0	MANUAL		rear	wheel drive		2.0	
3		6.0	MANUAL		rear	wheel drive		2.0	
4		6.0	MANUAL		rear	wheel drive		2.0	
	Market Category				Vehicle Size	Vehicle Style	\		
0	Factory Tuner,	Luxury,	High-Performance		Compact	Coupe			
1		Luxury,	Performance		Compact	Convertible			
2		Luxury,	High-Performance		Compact	Coupe			
3		Luxury,	Performance		Compact	Coupe			
4		Luxury			Compact	Convertible			
	highway	MPG	city	mpg	Popularity	MSRP			
0		26		19	3916	46135			
1		28		19	3916	40650			
2		28		20	3916	36350			
3		28		18	3916	29450			
4		28		18	3916	34500			
Data Engine Info:									