

```
from google.colab import drive
drive.mount('/content/drive')
```

Mounted at /content/drive

```
ls drive/MyDrive/'Colab Notebooks'/bmi.csv
```

drive/MyDrive/Colab Notebooks/bmi.csv

```
ls drive/MyDrive/'Colab Notebooks'/'income(1) (1).csv'
```

drive/MyDrive/Colab Notebooks/income(1) (1).csv

```
import pandas as pd
import numpy as np
from scipy import stats
```

```
df=pd.read_csv("drive/MyDrive/Colab Notebooks/bmi.csv")
df
```

	Gender	Height	Weight	Index	
0	Male	174	96	4	
1	Male	189	87	2	
2	Female	185	110	4	
3	Female	195	104	3	
4	Male	149	61	3	
...	...	...	...	...	
495	Female	150	153	5	
496	Female	184	121	4	
497	Female	141	136	5	
498	Male	150	95	5	
499	Male	173	131	5	

500 rows × 4 columns

Next steps:

Generate code with df

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```
df.head()
```



	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3



Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)

df.dropna()



	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...	...	...	...	...
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5



500 rows x 4 columns

```
max_vals = np.max(np.abs(df[['Height', 'Weight']]))
max_vals
max_vals
```

 199

```
df1=pd.read_csv("drive/MyDrive/Colab Notebooks/bmi.csv")
df1
```

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...	...	...	...	...
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5

500 rows x 4 columns

Next steps: [Generate code with df1](#) [View recommended plots](#) [New interactive sheet](#)

```
from sklearn.preprocessing import StandardScaler
sc=StandardScaler()
df1[['Height','Weight']] = sc.fit_transform(df1[['Height','Weight']])
df.head(10)
```

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
5	Male	189	104	3
6	Male	147	92	5
7	Male	154	111	5
8	Male	174	90	3
9	Female	169	103	4

Next steps: [Generate code with df](#) [View recommended plots](#) [New interactive sheet](#)

```
from sklearn.preprocessing import MinMaxScaler

Scaler=MinMaxScaler()
df[['Height','Weight']]=Scaler.fit_transform(df[['Height','Weight']])

df.head(0)
```



	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...	...	...	...	...
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5

500 rows x 4 columns

Next steps: [Generate code with df2](#) [View recommended plots](#) [New interactive sheet](#)

```
df2=pd.read_csv("drive/MyDrive/Colab Notebooks/bmi.csv")
df2
```



	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...	...	...	...	...
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5

500 rows x 4 columns

Next steps: [Generate code with df2](#) [View recommended plots](#) [New interactive sheet](#)

```
from sklearn.preprocessing import Normalizer
Scaler=Normalizer()
df2[['Height','Weight']]=Scaler.fit_transform(df2[['Height','Weight']])
df2
```

	Gender	Height	Weight	Index
0	Male	0.875578	0.483077	4
1	Male	0.908381	0.418144	2
2	Female	0.859536	0.511075	4
3	Female	0.882353	0.470588	3
4	Male	0.925448	0.378875	3
...	...	...	...	...
495	Female	0.700071	0.714073	5
496	Female	0.835527	0.549450	4
497	Female	0.719753	0.694230	5
498	Male	0.844819	0.535052	5
499	Male	0.797227	0.603680	5

500 rows x 4 columns

Next steps: [Generate code with df2](#) [View recommended plots](#) [New interactive sheet](#)

```
df3=pd.read_csv("drive/MyDrive/Colab Notebooks/bmi.csv")
df3
```

	Gender	Height	Weight	Index
0	Male	174	96	4
1	Male	189	87	2
2	Female	185	110	4
3	Female	195	104	3
4	Male	149	61	3
...	...	...	...	...
495	Female	150	153	5
496	Female	184	121	4
497	Female	141	136	5
498	Male	150	95	5
499	Male	173	131	5

500 rows x 4 columns

Next steps: [Generate code with df3](#) [View recommended plots](#) [New interactive sheet](#)

```
from sklearn.preprocessing import MaxAbsScaler
Scaler=MaxAbsScaler()
df3[['Height','Weight']]=Scaler.fit_transform(df3[['Height','Weight']])
df3
```

	Gender	Height	Weight	Index	
0	Male	0.874372	0.60000	4	
1	Male	0.949749	0.54375	2	
2	Female	0.929648	0.68750	4	
3	Female	0.979899	0.65000	3	
4	Male	0.748744	0.38125	3	
...	...	...	...	...	
495	Female	0.753769	0.95625	5	
496	Female	0.924623	0.75625	4	
497	Female	0.708543	0.85000	5	
498	Male	0.753769	0.59375	5	
499	Male	0.869347	0.81875	5	

500 rows x 4 columns

Next steps: [Generate code with df3](#) [View recommended plots](#) [New interactive sheet](#)

```
df4=pd.read_csv("drive/MyDrive/Colab Notebooks/bmi.csv")

from sklearn.preprocessing import RobustScaler
Scaler=RobustScaler()
df4[['Height','Weight']]=Scaler.fit_transform(df4[['Height','Weight']])
df4.head()
```



	Gender	Height	Weight	Index	
0	Male	0.125000	-0.178571	4	
1	Male	0.660714	-0.339286	2	
2	Female	0.517857	0.071429	4	
3	Female	0.875000	-0.035714	3	
4	Male	-0.767857	-0.803571	3	

Next steps:

[Generate code with df4](#)[View recommended plots](#)[New interactive sheet](#)

```
import matplotlib
import matplotlib.pyplot as plt
import seaborn as sns
import statsmodels.api as sm
from sklearn.model_selection import train_test_split
from sklearn.linear_model import LinearRegression
from sklearn.feature_selection import RFE
from sklearn.linear_model import RidgeCV,LassoCV,Ridge,Lasso
from sklearn.feature_selection import SelectKBest
from sklearn.feature_selection import mutual_info_classif
from sklearn.feature_selection import mutual_info_regression
from sklearn.feature_selection import chi2
```

```
df=pd.read_csv('drive/MyDrive/Colab Notebooks/income(1) (1).csv')
df.columns
```



```
Index(['age', 'JobType', 'EdType', 'maritalstatus', 'occupation',
       'relationship', 'race', 'gender', 'capitalgain', 'capitalloss',
       'hoursperweek', 'nativecountry', 'SalStat'],
      dtype='object')
```

```
df1=df.drop(["Name","sex","Ticket","cabin","embarked"],axis=1)
```



```

-----
KeyError                                Traceback (most recent call last)
<ipython-input-24-957512aa0a29> in <cell line: 1>()
----> 1 df1=df.drop(["Name","sex","Ticket","cabin","embarked"],axis=1)

/usr/local/lib/python3.10/dist-packages/pandas/core/indexes/base.py in drop(self, labels, errors)
    7068         if mask.any():
    7069             if errors != "ignore":
-> 7070                 raise KeyError(f"{labels[mask].tolist()} not found in axis")
    7071             indexer = indexer[~mask]
    7072         return self.delete(indexer)

KeyError: "[ 'Name', 'sex', 'Ticket', 'cabin', 'embarked'] not found in axis"

```

Next steps: [Explain error](#)

df1.columns



```
Index(['Gender', 'Height', 'Weight', 'Index'], dtype='object')
```

```
import pandas as pd
```

```
from sklearn.feature_selection import SelectKBest
```

```
from sklearn.feature_selection import chi2
```

```
data=pd.read_csv('drive/MyDrive/Colab Notebooks/bmi.csv')
```

```
data=data.dropna()
```

df.columns



```
Index(['age', 'JobType', 'EdType', 'maritalstatus', 'occupation',
      'relationship', 'race', 'gender', 'capitalgain', 'capitalloss',
      'hoursperweek', 'nativecountry', 'SalStat'],
      dtype='object')
```

df



	age	JobType	EdType	maritalstatus	occupation	relationship	race	gender	capitalgain	capitalloss	hoursperweek	nativecountry	SalStat		
	0	45	Private	HS-grad	Divorced	Adm-clerical	Not-in-family	White	Female	0	0	28	United-States	less than or equal to 50,000	
	1	24	Federal-gov	HS-grad	Never-married	Armed-Forces	Own-child	White	Male	0	0	40	United-States	less than or equal to 50,000	
	2	44	Private	Some-college	Married-civ-spouse	Prof-specialty	Husband	White	Male	0	0	40	United-States	greater than 50,000	
	3	27	Private	9th	Never-married	Craft-repair	Other-relative	White	Male	0	0	40	Mexico	less than or equal to 50,000	
-----	4	20	Private	Some-college	Never-married	Sales	Not-in-family	White	Male	0	0	35	United-States	less than or equal to 50,000	-----
Next	.	Generate code with df			<input checked="" type="checkbox"/> View recommended plots			New interactive sheet							
	...	...	...	...	...	...	...	...	...	...	...	...	...	...	

Generate code with df

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```
import pandas as pd
import numpy as np
from scipy.stats import chi2_contingency
import seaborn as sns
tips=sns.load_dataset('tips')
tips.head()
```

	total_bill	tip	sex	smoker	day	time	size
0	16.99	1.01	Female	No	Sun	Dinner	2
1	10.34	1.66	Male	No	Sun	Dinner	3
2	21.01	3.50	Male	No	Sun	Dinner	3
3	23.68	3.31	Male	No	Sun	Dinner	2
4	24.59	3.61	Female	No	Sun	Dinner	4