

```
import pandas as pd
import numpy as np # Read the Data with Pandas
csv_data = pd.read_csv("https://storage.googleapis.com/dqlab-dataset/shopping_data.csv")
print(csv_data)
```

```
↩ CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
0           1    Male   19              15             39
1           2    Male   21              15             81
2           3  Female   20              16              6
3           4  Female   23              16             77
4           5  Female   31              17             40
..         ...    ...   ...              ...             ...
195        196  Female   35             120             79
196        197  Female   45             126             28
197        198    Male   32             126             74
198        199    Male   32             137             18
199        200    Male   30             137             83
```

[200 rows x 5 columns]

```
print(csv_data.head()) # show only first five rows
```

```
↩ CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
0           1    Male   19              15             39
1           2    Male   21              15             81
2           3  Female   20              16              6
3           4  Female   23              16             77
4           5  Female   31              17             40
5           6  Female   22              17             76
6           7  Female   35              18              6
7           8  Female   23              18             94
```

```
print(csv_data.head(10)) # show only first n rows
```

```
↩ CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
0           1    Male   19              15             39
1           2    Male   21              15             81
2           3  Female   20              16              6
3           4  Female   23              16             77
4           5  Female   31              17             40
5           6  Female   22              17             76
6           7  Female   35              18              6
7           8  Female   23              18             94
8           9    Male   64              19              3
9          10  Female   30              19             72
```

```
print(csv_data.tail()) # show only last five rows
```

```
↩ CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
195        196  Female   35             120             79
196        197  Female   45             126             28
197        198    Male   32             126             74
198        199    Male   32             137             18
199        200    Male   30             137             83
```

```
print(csv_data.tail(8)) # show only last n rows
```

```
↩ CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
192        193    Male   33             113              8
193        194  Female   38             113             91
194        195  Female   47             120             16
195        196  Female   35             120             79
196        197  Female   45             126             28
197        198    Male   32             126             74
198        199    Male   32             137             18
199        200    Male   30             137             83
```

```
print(csv_data.tail(-4)) #For negative values of n, this function returns all rows except the first |n| rows
```

```
↩ CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
4           5  Female   31              17             40
5           6  Female   22              17             76
6           7  Female   35              18              6
7           8  Female   23              18             94
8           9    Male   64              19              3
..         ...    ...   ...              ...             ...
195        196  Female   35             120             79
```

196	197	Female	45	126	28
197	198	Male	32	126	74
198	199	Male	32	137	18
199	200	Male	30	137	83

[196 rows x 5 columns]

```
print(csv_data.columns)           #.columns to access the column of the data source
```

```
Index(['CustomerID', 'Genre', 'Age', 'Annual Income (k$)',
      'Spending Score (1-100)'],
      dtype='object')
```

```
csv_data.index
```

```
RangeIndex(start=0, stop=200, step=1)
```

```
csv_data.dtypes           #Return the dtypes in the DataFrame
```

```

CustomerID    int64
Genre         object
Age           int64
Annual Income (k$)  int64
Spending Score (1-100)  int64
dtype: object

```

```
csv_data.info()
```

```

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 200 entries, 0 to 199
Data columns (total 5 columns):
#   Column                Non-Null Count  Dtype
---  -
0   CustomerID            200 non-null   int64
1   Genre                 200 non-null   object
2   Age                   200 non-null   int64
3   Annual Income (k$)    200 non-null   int64
4   Spending Score (1-100) 200 non-null   int64
dtypes: int64(4), object(1)
memory usage: 7.9+ KB

```

```
csv_data.values
```

```


```

```
[74, 'Female', 60, 50.0, 50],
[75, 'Male', 59, 54.0, 47],
[76, 'Male', 26, 54.0, 54],
[77, 'Female', 45, 54.0, 53],
[78, 'Male', 40, 54.0, 48],
[79, 'Female', 23, 54.0, 52],
[80, 'Female', 49, 54.0, 42],
[81, 'Male', 57, 54.0, 51],
[82, 'Male', 38, 54.0, 55],
[83, 'Male', 67, 54.0, 41],
[84, 'Female', 46, 54.0, 44],
[85, 'Female', 21, 54.0, 57],
[86, 'Male', 48, 54.0, 46],
[87, 'Female', 55, 57.0, 58],
[88, 'Female', 22, 57.0, 55],
[89, 'Female', 34, 58.0, 60],
[90, 'Female', 50, 58.0, 46],
[91, 'Female', 68, 59.0, 55],
[92, 'Male', 18, 59.0, 41],
[93, 'Male', 48, 60.0, 49],
[94, 'Female', 40, 60.0, 40],
[95, 'Female', 32, 60.0, 42],
[96, 'Male', 24, 60.0, 52],
[97, 'Female', 47, 60.0, 47],
[98, 'Female', 27, 60.0, 50],
[99, 'Male', 48, 61.0, 42],
[100, 'Male', 20, 61.0, 49],
[101, 'Female', 23, 62.0, 41],
[102, 'Female', 49, 62.0, 48]
```

```
csv_data.value_counts()
```



					count
CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)	
1	Male	19	15.0	39	1
138	Male	32	73.0	73	1
128	Male	40	71.0	95	1
129	Male	59	71.0	11	1
130	Male	38	71.0	75	1
...
70	Female	32	48.0	47	1
71	Male	70	49.0	55	1
72	Female	47	49.0	42	1
73	Female	60	50.0	49	1
200	Male	30	137.0	83	1

200 rows × 6 columns

```
dtype: int64
```

```
csv_data.ndim                                #Return an int representing the number of axes / array dimensions
```



```
2
```

```
csv_data.size                                #Return an int representing the number of elements in this object
```



```
1000
```

```
csv_data.shape                                #Return a tuple representing the dimensionality of the DataFrame
```



```
(200, 5)
```

```
csv_data.empty                                # Indicator whether Series/DataFrame is empty
```



```
False
```

```
print(csv_data["Age"])                        #access column using label
```

```

0      19
1      21
2      20
3      23
4      31
..
195    35
196    45
197    32
198    32
199    30
Name: Age, Length: 200, dtype: int64

```

```
csv_data.loc[5,'CustomerID'] # Access a group of rows and columns by label(s)
```

```
6
```

```
csv_data.loc[5]
```

```

5
CustomerID    6
Genre        Female
Age          22
Annual Income (k$)  17
Spending Score (1-100)  76

```

```
dtype: object
```

```
print(csv_data.loc[2:5])
```

```

CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
2          3  Female   20                16                6
3          4  Female   23                16               77
4          5  Female   31                17               40
5          6  Female   22                17               76

```

```
print(csv_data.iloc[5]) #Access row
```

```

CustomerID    6
Genre        Female
Age          22
Annual Income (k$)  17
Spending Score (1-100)  76
Name: 5, dtype: object

```

```
print(csv_data["Age"].iloc[1])
```

```
21
```

```
print("Shows data to 5th to less than 10th in a row:") #Data Based on Range
print(csv_data.iloc[5:10])
```

```

Shows data to 5th to less than 10th in a row:
CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
5          6  Female   22                17                76
6          7  Female   35                18                6
7          8  Female   23                18               94
8          9   Male   64                19                3
9         10  Female   30                19               72

```

```
# Data Filtering
print(csv_data[csv_data.Genre=='Female'])
```

```

CustomerID  Genre  Age  Annual Income (k$)  Spending Score (1-100)
2          3  Female   20                16                6
3          4  Female   23                16               77
4          5  Female   31                17               40
5          6  Female   22                17                76
6          7  Female   35                18                6
..         ...   ...   ...                ...                ...
191        192  Female   32               103               69
193        194  Female   38               113               91

```

194	195	Female	47	120	16
195	196	Female	35	120	79
196	197	Female	45	126	28

[112 rows x 5 columns]

```
print(csv_data[(csv_data.Genre=='Female') & (csv_data.Age>60)])
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
40	41	Female	65	38	35
62	63	Female	67	47	52
67	68	Female	68	48	48
90	91	Female	68	59	55
106	107	Female	66	63	50
116	117	Female	63	65	43

```
# str accessor to filter rows based on strings.
csv_data[csv_data.Genre.str.startswith('M')]
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
8	9	Male	64	19	3
10	11	Male	67	19	14
14	15	Male	37	20	13
...
187	188	Male	28	101	68
192	193	Male	33	113	8
197	198	Male	32	126	74
198	199	Male	32	137	18
199	200	Male	30	137	83

88 rows x 5 columns

```
# isin method to filter the names that exist in a given list
names = ['M','F','Male']
csv_data[csv_data.Genre.isin(names)]
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19	15	39
1	2	Male	21	15	81
8	9	Male	64	19	3
10	11	Male	67	19	14
14	15	Male	37	20	13
...
187	188	Male	28	101	68
192	193	Male	33	113	8
197	198	Male	32	126	74
198	199	Male	32	137	18
199	200	Male	30	137	83

88 rows x 5 columns

```
# query function can pass the conditions as a string
csv_data.query('Genre == "Female" and Age > 60')
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
40	41	Female	65	38	35
62	63	Female	67	47	52
67	68	Female	68	48	48
90	91	Female	68	59	55
106	107	Female	66	63	50
116	117	Female	63	65	43

Descriptive statistics include those that summarize the central tendency, dispersion and shape of a dataset's distribution, excluding NaN
 print(csv_data.describe())

	CustomerID	Age	Annual Income (k\$)	Spending Score (1-100)
count	200.000000	200.000000	200.000000	200.000000
mean	100.500000	38.850000	60.560000	50.200000
std	57.879185	13.969007	26.264721	25.823522
min	1.000000	18.000000	15.000000	1.000000
25%	50.750000	28.750000	41.500000	34.750000
50%	100.500000	36.000000	61.500000	50.000000
75%	150.250000	49.000000	78.000000	73.000000
max	200.000000	70.000000	137.000000	99.000000

For numeric data, the result's index will include count, mean, std, min, max as well as lower, 50 and upper percentiles
 # For object data (e.g. strings or timestamps), the result's index will include count, unique, top, and freq.
 # The top is the most common value. The freq is the most common value's frequency. Timestamps also include the first and last items
 print(csv_data.describe(include="all"))

	CustomerID	Genre	Age	Annual Income (k\$)	\
count	200.000000	200	200.000000	200.000000	
unique	NaN	2	NaN	NaN	
top	NaN	Female	NaN	NaN	
freq	NaN	112	NaN	NaN	
mean	100.500000	NaN	38.850000	60.560000	
std	57.879185	NaN	13.969007	26.264721	
min	1.000000	NaN	18.000000	15.000000	
25%	50.750000	NaN	28.750000	41.500000	
50%	100.500000	NaN	36.000000	61.500000	
75%	150.250000	NaN	49.000000	78.000000	
max	200.000000	NaN	70.000000	137.000000	

	Spending Score (1-100)
count	200.000000
unique	NaN
top	NaN
freq	NaN
mean	50.200000
std	25.823522
min	1.000000
25%	34.750000
50%	50.000000
75%	73.000000
max	99.000000

csv_data.describe(include=[np.number]) # Including only numeric columns

	CustomerID	Age	Annual Income (k\$)	Spending Score (1-100)
count	200.000000	200.000000	200.000000	200.000000
mean	100.500000	38.850000	60.560000	50.200000
std	57.879185	13.969007	26.264721	25.823522
min	1.000000	18.000000	15.000000	1.000000
25%	50.750000	28.750000	41.500000	34.750000
50%	100.500000	36.000000	61.500000	50.000000
75%	150.250000	49.000000	78.000000	73.000000
max	200.000000	70.000000	137.000000	99.000000

print(csv_data.describe(exclude=["0"])) # ignore non-numeric data for processing

```

↩ CustomerID      Age Annual Income (k$) Spending Score (1-100)
count  200.000000  200.000000      200.000000      200.000000
mean   100.500000   38.850000        60.560000       50.200000
std    57.879185   13.969007        26.264721       25.823522
min     1.000000   18.000000        15.000000        1.000000
25%    50.750000   28.750000        41.500000       34.750000
50%   100.500000   36.000000        61.500000       50.000000
75%   150.250000   49.000000        78.000000       73.000000
max    200.000000   70.000000       137.000000      99.000000

```

```
csv_data.isnull()
```

```

↩ CustomerID Genre Age Annual Income (k$) Spending Score (1-100)
0      False  False  False                False                False
1      False  False  False                False                False
2      False  False  False                False                False
3      False  False  False                False                False
4      False  False  False                False                False
...      ...    ...    ...                ...                ...
195     False  False  False                False                False
196     False  False  False                False                False
197     False  False  False                False                False
198     False  False  False                False                False
199     False  False  False                False                False

```

200 rows × 5 columns

```
csv_data.isna()
```

```

↩ CustomerID Genre Age Annual Income (k$) Spending Score (1-100)
0      False  False  False                False                False
1      False  False  False                False                False
2      False  False  False                False                False
3      False  False  False                False                False
4      False  False  False                False                False
...      ...    ...    ...                ...                ...
195     False  False  False                False                False
196     False  False  False                False                False
197     False  False  False                False                False
198     False  False  False                False                False
199     False  False  False                False                False

```

200 rows × 5 columns

```
print(csv_data.isnull().values.any())          # find missing value
```

```
↩ False
```

```
print(csv_data.isna().values.any())
```

```
↩ False
```

```
print(csv_data.duplicated().values.any())
```

```
↩ False
```

```

data_missing = pd.read_csv("https://storage.googleapis.com/dqlab-dataset/shopping_data_missingvalue.csv")
print(data_missing.isnull().values.any())

```

True

```
data_missing.dropna()
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19.0	15.0	39.0
3	4	Female	23.0	16.0	77.0
6	7	Female	35.0	18.0	6.0
7	8	Female	23.0	18.0	94.0
9	10	Female	30.0	19.0	72.0
...
195	196	Female	35.0	120.0	79.0
196	197	Female	45.0	126.0	28.0
197	198	Male	32.0	126.0	74.0
198	199	Male	32.0	137.0	18.0
199	200	Male	30.0	137.0	83.0

195 rows x 5 columns

```
print(data_missing.isnull().values.any())
```

True

```
data_missing=data_missing.dropna()
```

```
print(data_missing.isnull().values.any())
```

False

```
csv_data= csv_data.drop("Genre", axis=1) # drop column
print(csv_data)
```

	CustomerID	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	19	15	39
1	2	21	15	81
2	3	20	16	6
3	4	23	16	77
4	5	31	17	40
..
195	196	35	120	79
196	197	45	126	28
197	198	32	126	74
198	199	32	137	18
199	200	30	137	83

[200 rows x 4 columns]

```
csv_data.mean()
```

	0
CustomerID	100.50
Age	38.85
Annual Income (k\$)	60.56
Spending Score (1-100)	50.20

dtype: float64

```
csv_data.median()
```


	0
CustomerID	100.5
Age	36.0
Annual Income (k\$)	61.5
Spending Score (1-100)	50.0

dtype: float64

```
data_filling=data_missing.fillna(csv_data.mean())
```

```
print(data_filling.head(10))
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19.0	15.0	39.0
3	4	Female	23.0	16.0	77.0
6	7	Female	35.0	18.0	6.0
7	8	Female	23.0	18.0	94.0
9	10	Female	30.0	19.0	72.0
10	11	Male	67.0	19.0	14.0
11	12	Female	35.0	19.0	99.0
12	13	Female	58.0	20.0	15.0
13	14	Female	24.0	20.0	77.0
14	15	Male	37.0	20.0	13.0

```
data_filling=data_missing.fillna(csv_data.median())
```

```
print(data_filling.head(10))
```

	CustomerID	Genre	Age	Annual Income (k\$)	Spending Score (1-100)
0	1	Male	19.0	15.0	39.0
3	4	Female	23.0	16.0	77.0
6	7	Female	35.0	18.0	6.0
7	8	Female	23.0	18.0	94.0
9	10	Female	30.0	19.0	72.0
10	11	Male	67.0	19.0	14.0
11	12	Female	35.0	19.0	99.0
12	13	Female	58.0	20.0	15.0
13	14	Female	24.0	20.0	77.0
14	15	Male	37.0	20.0	13.0

#In order to convert data types in pandas, there are three basic options:

#Use astype() to force an appropriate dtype

#Create a custom function to convert the data

#Use pandas functions such as to_numeric() or to_datetime()

```
csv_data['Annual Income (k$)'].astype('float')
```

	Annual Income (k\$)
0	15.0
1	15.0
2	16.0
3	16.0
4	17.0
...	...
195	120.0
196	126.0
197	126.0
198	137.0
199	137.0

200 rows × 1 columns

dtype: float64

csv_data.dtypes



	0
CustomerID	int64
Genre	object
Age	int64
Annual Income (k\$)	int64
Spending Score (1-100)	int64

dtypes: object

csv_data['Annual Income (k\$)'] = csv_data['Annual Income (k\$)'].astype('float')
csv_data.dtypes



	0
CustomerID	int64
Genre	object
Age	int64
Annual Income (k\$)	float64
Spending Score (1-100)	int64

dtypes: object