



```
import pandas as pd
import numpy as np
```

```
df=pd.read_csv("/content/bike_buyers.csv")
df
```



	ID	Marital Status	Gender	Income	Children	Education	Occupation	Home Owner	Cars	Commute Distance	Region	Age	Purchased Bike
0	12496	Married	Female	40000.0	1.0	Bachelors	Skilled Manual	Yes	0.0	0-1 Miles	Europe	42.0	No
1	24107	Married	Male	30000.0	3.0	Partial College	Clerical	Yes	1.0	0-1 Miles	Europe	43.0	No
2	14177	Married	Male	80000.0	5.0	Partial College	Professional	No	2.0	2-5 Miles	Europe	60.0	No
3	24381	Single	NaN	70000.0	0.0	Bachelors	Professional	Yes	1.0	5-10 Miles	Pacific	41.0	Yes
4	25597	Single	Male	30000.0	0.0	Bachelors	Clerical	No	0.0	0-1 Miles	Europe	36.0	Yes
...
995	23731	Married	Male	60000.0	2.0	High School	Professional	Yes	2.0	2-5 Miles	North America	54.0	Yes
996	28672	Single	Male	70000.0	4.0	Graduate Degree	Professional	Yes	0.0	2-5 Miles	North America	35.0	Yes
997	11809	Married	NaN	60000.0	2.0	Bachelors	Skilled Manual	Yes	0.0	0-1 Miles	North America	38.0	Yes
998	19664	Single	Male	100000.0	3.0	Bachelors	Management	No	3.0	1-2 Miles	North America	38.0	No
999	12121	Single	Male	60000.0	3.0	High School	Professional	Yes	2.0	10+ Miles	North America	53.0	Yes

```
df.head()
```



	ID	Marital Status	Gender	Income	Children	Education	Occupation	Home Owner	Cars	Commute Distance	Region	Age	Purchased Bike
0	12496	Married	Female	40000.0	1.0	Bachelors	Skilled Manual	Yes	0.0	0-1 Miles	Europe	42.0	No
1	24107	Married	Male	30000.0	3.0	Partial College	Clerical	Yes	1.0	0-1 Miles	Europe	43.0	No
2	14177	Married	Male	80000.0	5.0	Partial College	Professional	No	2.0	2-5 Miles	Europe	60.0	No
3	24381	Single	NaN	70000.0	0.0	Bachelors	Professional	Yes	1.0	5-10 Miles	Pacific	41.0	Yes
4	25597	Single	Male	30000.0	0.0	Bachelors	Clerical	No	0.0	0-1 Miles	Europe	36.0	Yes

```
df.tail()
```



	ID	Marital Status	Gender	Income	Children	Education	Occupation	Home Owner	Cars	Commute Distance	Region	Age	Purchased Bike
995	23731	Married	Male	60000.0	2.0	High School	Professional	Yes	2.0	2-5 Miles	North America	54.0	Yes
996	28672	Single	Male	70000.0	4.0	Graduate Degree	Professional	Yes	0.0	2-5 Miles	North America	35.0	Yes
997	11809	Married	NaN	60000.0	2.0	Bachelors	Skilled Manual	Yes	0.0	0-1 Miles	North America	38.0	Yes
998	19664	Single	Male	100000.0	3.0	Bachelors	Management	No	3.0	1-2 Miles	North America	38.0	No
999	12121	Single	Male	60000.0	3.0	High School	Professional	Yes	2.0	10+ Miles	North America	53.0	Yes

```
df.shape
```

```
(1000, 13)
```

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1000 entries, 0 to 999
Data columns (total 13 columns):
#   Column              Non-Null Count  Dtype
---  -
0   ID                   1000 non-null   int64
1   Marital Status       993 non-null    object
2   Gender               989 non-null    object
3   Income               994 non-null    float64
4   Children             992 non-null    float64
5   Education            1000 non-null    object
6   Occupation           1000 non-null    object
7   Home Owner           996 non-null    object
8   Cars                 991 non-null    float64
9   Commute Distance     1000 non-null    object
10  Region              1000 non-null    object
11  Age                  992 non-null    float64
12  Purchased Bike       1000 non-null    object
dtypes: float64(4), int64(1), object(8)
memory usage: 101.7+ KB
```


```
df.describe()
```

	ID	Income	Children	Cars	Age
count	1000.000000	994.000000	992.000000	991.000000	992.000000
mean	19965.992000	56267.605634	1.910282	1.455096	44.181452
std	5347.333948	31067.817462	1.626910	1.121755	11.362007
min	11000.000000	10000.000000	0.000000	0.000000	25.000000
25%	15290.750000	30000.000000	0.000000	1.000000	35.000000
50%	19744.000000	60000.000000	2.000000	1.000000	43.000000
75%	24470.750000	70000.000000	3.000000	2.000000	52.000000
max	29447.000000	170000.000000	5.000000	4.000000	89.000000

```
df.isnull()
```


	ID	Marital Status	Gender	Income	Children	Education	Occupation	Home Owner	Cars	Commute Distance	Region	Age	Purchased Bike
0	False	False	False	False	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False	False	False	False	False
3	False	False	True	False	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False	False	False	False	False
...
995	False	False	False	False	False	False	False	False	False	False	False	False	False
996	False	False	False	False	False	False	False	False	False	False	False	False	False
997	False	False	True	False	False	False	False	False	False	False	False	False	False
998	False	False	False	False	False	False	False	False	False	False	False	False	False
999	False	False	False	False	False	False	False	False	False	False	False	False	False

```
df.isnull().sum()
```




	0
ID	0
Marital Status	7
Gender	11
Income	6
Children	8
Education	0
Occupation	0
Home Owner	4
Cars	9
Commute Distance	0
Region	0
Age	8
Purchased Bike	0

df.isnull().mean()



	0
ID	0.000
Marital Status	0.007
Gender	0.011
Income	0.006
Children	0.008
Education	0.000
Occupation	0.000
Home Owner	0.004
Cars	0.009
Commute Distance	0.000
Region	0.000
Age	0.008
Purchased Bike	0.000

df["Marital Status"].to_string()



'0	Married\n1	Married\n2	Married\n3	Single\n4	Single\n5	Married\n6	Single\n7	Married\n8
NaN\n9	Married\n10	Married\n11	Single\n12	Married\n13	Married\n14	Single\n15	Single\n16	Single\n17
Married\n18	Single\n19	Single\n20	Married\n21	Single\n22	Single\n23	Married\n24	Single\n25	Single\n26
Single\n27	Single\n28	Married\n29	Single\n30	Married\n31	Married\n32	Single\n33	Single\n34	Single\n35
Single\n36	Married\n37	Single\n38	Single\n39	Single\n40	Single\n41	Single\n42	Married\n43	Married\n44
Married\n45	Married\n46	Married\n47	Single\n48	Married\n49	NaN\n50	Single\n51	Single\n52	Married\n53
Single\n54	Single\n55	Married\n56	Married\n57	Married\n58	Married\n59	Married		

✓ CATEGORICAL DATA FILLING

Filling Data by Mode method.

```
# Percentage of null values in Marital Status.  
df["Marital Status"].isnull().mean()
```

0.007

```
# Filling data
df["Marital Status"].fillna(df["Marital Status"].mode()[0],inplace=True)
```

<ipython-input-78-f2ee6d380762>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment. The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values is a copy. For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value, inplace=True) instead.

```
df["Marital Status"].fillna(df["Marital Status"].mode()[0],inplace=True)
```

```
df["Marital Status"].isnull().mean()
```

0.0

```
# After Fill the Marital Status column
df["Marital Status"].to_string()
```

```
'0      Married\n1      Married\n2      Married\n3      Single\n4      Single\n5      Married\n6      Single\n7      Married\n8      Married\n9      Married\n10     Married\n11     Single\n12     Married\n13     Married\n14     Single\n15     Single\n16     Single\n17     Married\n18     Single\n19     Single\n20     Married\n21     Single\n22     Single\n23     Married\n24     Single\n25     Single\n26     Single\n27     Married\n28     Married\n29     Single\n30     Married\n31     Married\n32     Single\n33     Single\n34     Single\n35     Single\n36     Married\n37     Single\n38     Single\n39     Single\n40     Single\n41     Single\n42     Married\n43     Married\n44     Married\n45     Married\n46     Married\n47     Single\n48     Married\n49     Married\n50     Single\n51     Single\n52     Married\n53     Single\n54     Single\n55     Married\n56     Married\n57     Married\n58     Married\n59
```

Filling Data by Bfill (BACKWARD FILL) Method.

```
# Show Gender column.
df["Gender"].to_string()
```

```
'0      Female\n1      Male\n2      Male\n3      NaN\n4      Male\n5      Female\n6      Male\n7      Male\n8      Male\n9      Male\n10     Female\n11     Female\n12     Male\n13     Male\n14     Male\n15     Female\n16     Female\n17     Male\n18     Male\n19     Male\n20     Female\n21     Female\n22     Male\n23     Female\n24     Male\n25     Male\n26     Male\n27     Female\n28     Male\n29     Female\n30     Female\n31     Male\n32     Female\n33     Male\n34     Male\n35     Female\n36     Female\n37     Female\n38     Male\n39     Female\n40     Female\n41     Female\n42     Female\n43     Female\n44     Female\n45     Female\n46     Female\n47     Female\n48     Male\n49     Male\n50     Female\n51     Male\n52     Female\n53     Female\n54     Female\n55     Male\n56     Male\n57     Male\n58     Female\n59     Male
```

```
df["Gender"].isnull().mean()
```

0.011

```
# Filling Gender column with Bfill method (Backward Fill)
df["Gender"].bfill(inplace=True)
```

<ipython-input-83-84999d5bb0b1>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment. The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values is a copy. For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value, inplace=True) instead.

```
df["Gender"].bfill(inplace=True)
```

```
# After fill Gender column
df["Gender"].to_string()
```

```
'0      Female\n1      Male\n2      Male\n3      Male\n4      Male\n5      Male\n6      Female\n7      Male\n8      Male\n9      Male\n10     Female\n11     Female\n12     Female\n13     Male\n14     Male\n15     Male\n16     Female\n17     Female\n18     Male\n19     Male\n20     Female\n21     Female\n22     Male\n23     Female\n24     Female\n25     Male\n26     Male\n27     Female\n28     Male\n29     Female\n30     Female\n31     Male\n32     Female\n33     Male\n34     Male\n35     Female\n36     Female\n37     Female\n38     Male\n39     Female\n40     Female\n41     Female\n42     Female\n43     Female\n44     Female\n45     Female\n46     Female\n47     Female\n48     Male\n49     Male\n50     Female\n51     Male\n52     Female\n53     Female\n54     Female\n55     Male\n56     Male\n57     Male\n58     Female\n59     Male
```

```
df["Gender"].isnull().mean()
```

0.0

Data filling by Ffill (FORWARD FILL) Method

```
# Show Home Owner null values
df["Home Owner"].isnull().mean()
```

0.004

```
df["Home Owner"].to_string()
```

'0	Yes\n1	Yes\n2	No\n3	Yes\n4	No\n5	Yes\n6	NaN\n7	Yes\n8	Yes\n9	Yes\n10	No\n11
No\n12	Yes\n13	Yes\n14	No\n15	Yes\n16	No\n17	Yes\n18	Yes\n19	Yes\n20	Yes\n21	Yes\n22	Yes
\n23	No\n24	No\n25	Yes\n26	No\n27	No\n28	Yes\n29	No\n30	Yes\n31	No\n32	No\n33	No\n34
4	No\n35	No\n36	Yes\n37	No\n38	No\n39	Yes\n40	No\n41	Yes\n42	Yes\n43	No\n44	Yes\n45
Yes\n46	Yes\n47	Yes\n48	No\n49	Yes\n50	No\n51	No\n52	Yes\n53	No\n54	No\n55	Yes\n56	Yes
\n57	No\n58	Yes\n59	Yes\n60	No\n61	Yes\n62	Yes\n63	Yes\n64	Yes\n65	Yes\n66	Yes\n67	Yes\n68
8	Yes\n69	No\n70	Yes\n71	No\n72	Yes\n73	No\n74	No\n75	Yes\n76	No\n77	Yes\n78	No\n79

```
# Filling Home Owner column
df["Home Owner"].ffill(inplace=True)
```

<ipython-input-88-bc4ad60ae0bd>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained assignment on the result of a filter operation. This behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting values is a copy. To avoid this warning, you can use df["Home Owner"] = df["Home Owner"].ffill(inplace=False) or df["Home Owner"] = df["Home Owner"].ffill().

For example, when doing 'df[col].method(value, inplace=True)', try using 'df.method({col: value}, inplace=True)' or df[col] = df[col].method(value, inplace=True).

```
df["Home Owner"].ffill(inplace=True)
```

```
df["Home Owner"].isnull().mean()
```

0.0

```
df["Home Owner"].to_string()
```

'0	Yes\n1	Yes\n2	No\n3	Yes\n4	No\n5	Yes\n6	Yes\n7	Yes\n8	Yes\n9	Yes\n10	No\n11
No\n12	Yes\n13	Yes\n14	No\n15	Yes\n16	No\n17	Yes\n18	Yes\n19	Yes\n20	Yes\n21	Yes\n22	Yes
\n23	No\n24	No\n25	Yes\n26	No\n27	No\n28	Yes\n29	No\n30	Yes\n31	No\n32	No\n33	No\n34
4	No\n35	No\n36	Yes\n37	No\n38	No\n39	Yes\n40	No\n41	Yes\n42	Yes\n43	No\n44	Yes\n45
Yes\n46	Yes\n47	Yes\n48	No\n49	Yes\n50	No\n51	No\n52	Yes\n53	No\n54	No\n55	Yes\n56	Yes
\n57	No\n58	Yes\n59	Yes\n60	No\n61	Yes\n62	Yes\n63	Yes\n64	Yes\n65	Yes\n66	Yes\n67	Yes\n68
8	Yes\n69	No\n70	Yes\n71	No\n72	Yes\n73	No\n74	No\n75	Yes\n76	No\n77	Yes\n78	No\n79