```
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\n2
5      Single\n26      Single\n27         NaN\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      Sin
gle\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\n5
```

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-40-c7bae71de536>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained ass
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting

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

  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      Si
ngle\n17     Married\n18      Single\n19      Single\n20     Married\n21      Single\n22      Single\n23     Married\n24      Singl
e\n25      Single\n26      Single\n27     Married\n28     Married\n29      Single\n30     Married\n31     Married\n32      Single\n
33      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     Marr
ied\n50      Single\n51      Single\n52     Married\n53      Single\n54      Single\n55     Married\n56     Married\n57     Married
```

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        M
ale\n9        Male\n10     Female\n11     Female\n12       Male\n13       Male\n14       Male\n15     Female\n16       Male\n17
Female\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
```

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

```
0.011
```

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

```
<ipython-input-45-84999d5bb0b1>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained ass
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting

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

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

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

```
'0      Female\n1        Male\n2        Male\n3        Male\n4        Male\n5      Female\n6        Male\n7        Male\n8        M
ale\n9        Male\n10     Female\n11     Female\n12       Male\n13       Male\n14       Male\n15     Female\n16       Male\n17
Female\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
```

```
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      Ye
s\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      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      Yes\n69      No\n70      Yes\n71      No\n72      Yes\n73      No\n74      No\n75      Yes\n76
```

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

```
<ipython-input-50-1d8263e456ce>:2: FutureWarning: A value is trying to be set on a copy of a DataFrame or Series through chained ass
The behavior will change in pandas 3.0. This inplace method will never work because the intermediate object on which we are setting

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

  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      Ye
s\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      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      Yes\n69      No\n70      Yes\n71      No\n72      Yes\n73      No\n74      No\n75      Yes\n76
```

```
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	Male	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	Male	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.isnull().mean()
```

	0
ID	0.000
Marital Status	0.000
Gender	0.000
Income	0.006
Children	0.008
Education	0.000
Occupation	0.000