

In [ ]:

BIVARIATE ANALYSIS

In [ ]:

K.SASIKIRAN  
MCA(R)  
2019202049

In [1]:

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
import math
df=pd.read_csv("suv_data.csv")
df.head(5)
```

Out[1]:

	User ID	Gender	Age	EstimatedSalary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0

In [2]:

```
print("number of Users Purchased: "+str(len(df.index)))
```

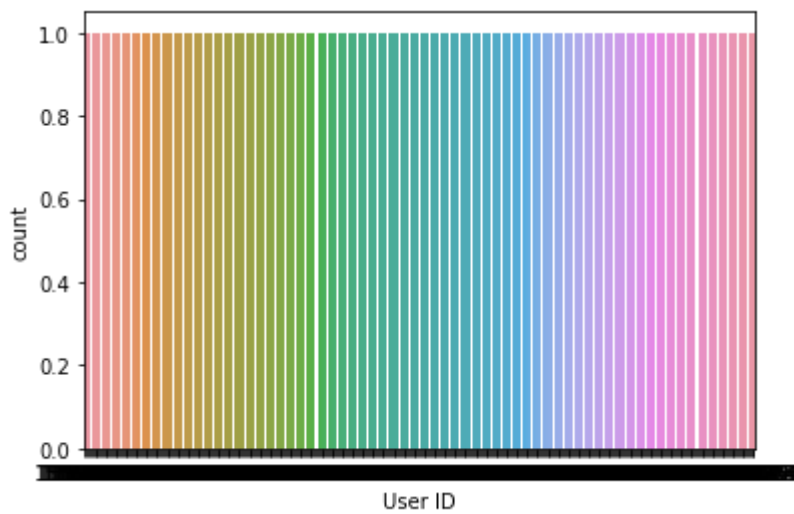
number of Users Purchased: 400

In [3]:

```
sns.countplot(x="User ID",data=df)
```

Out[3]:

<AxesSubplot:xlabel='User ID', ylabel='count'>

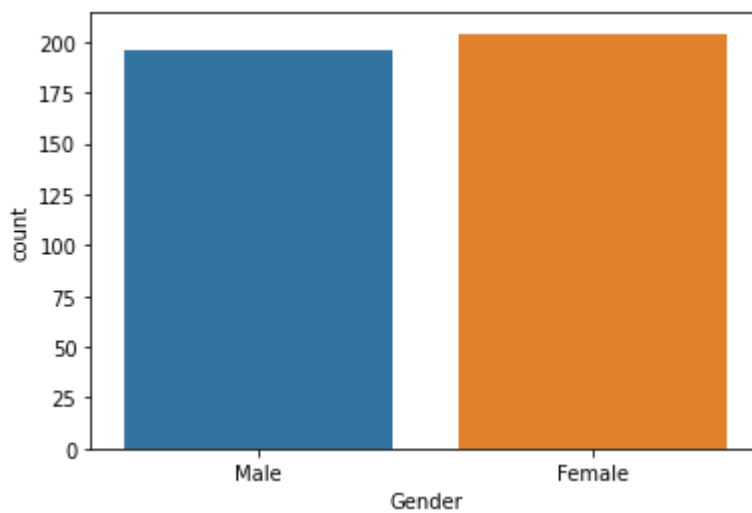


In [4]:

```
sns.countplot(x="Gender",data=df)
```

Out[4]:

<AxesSubplot:xlabel='Gender', ylabel='count'>

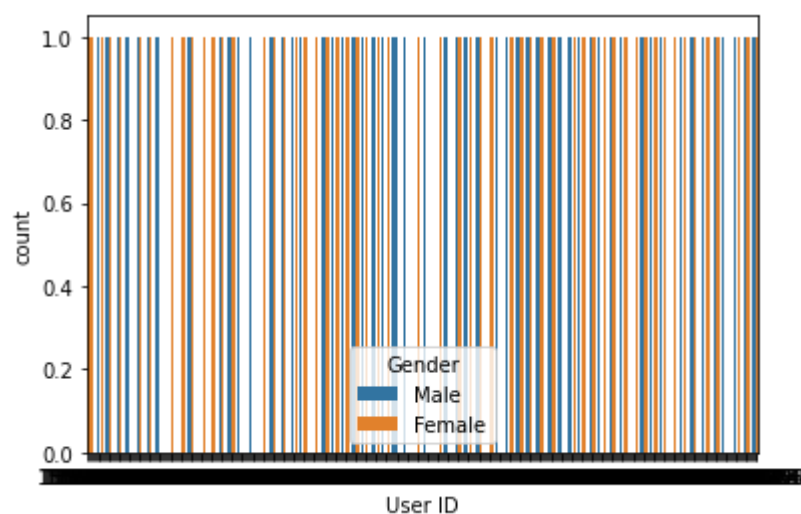


In [5]:

```
sns.countplot(x="User ID",hue="Gender",data=df)
```

Out[5]:

<AxesSubplot:xlabel='User ID', ylabel='count'>

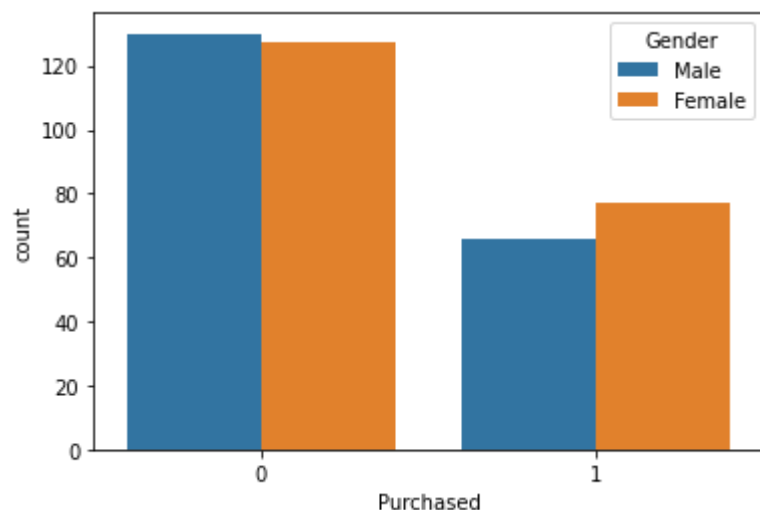


In [6]:

```
sns.countplot(x="Purchased",hue="Gender",data=df)
```

Out[6]:

<AxesSubplot:xlabel='Purchased', ylabel='count'>



In [7]:

```
df["Age"].plot.hist()
```

File "<ipython-input-7-2a3dfb671752>", line 1  
df["Age"].plot.hist()  
^

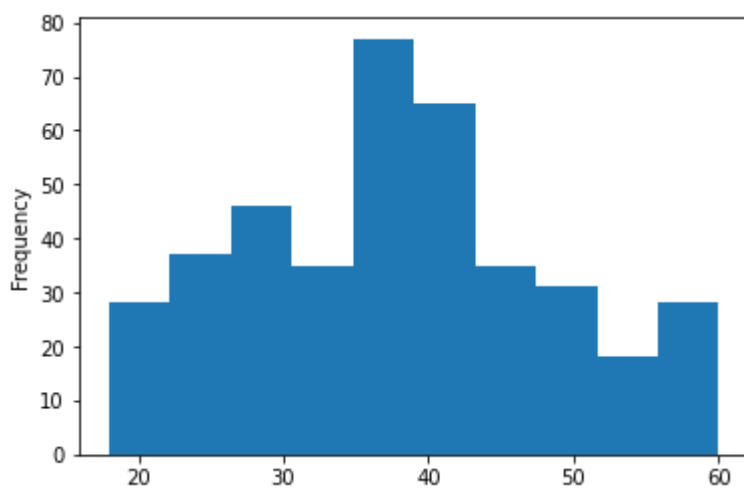
**SyntaxError:** invalid syntax

In [8]:

```
df["Age"].plot.hist()
```

Out[8]:

<AxesSubplot:ylabel='Frequency'>

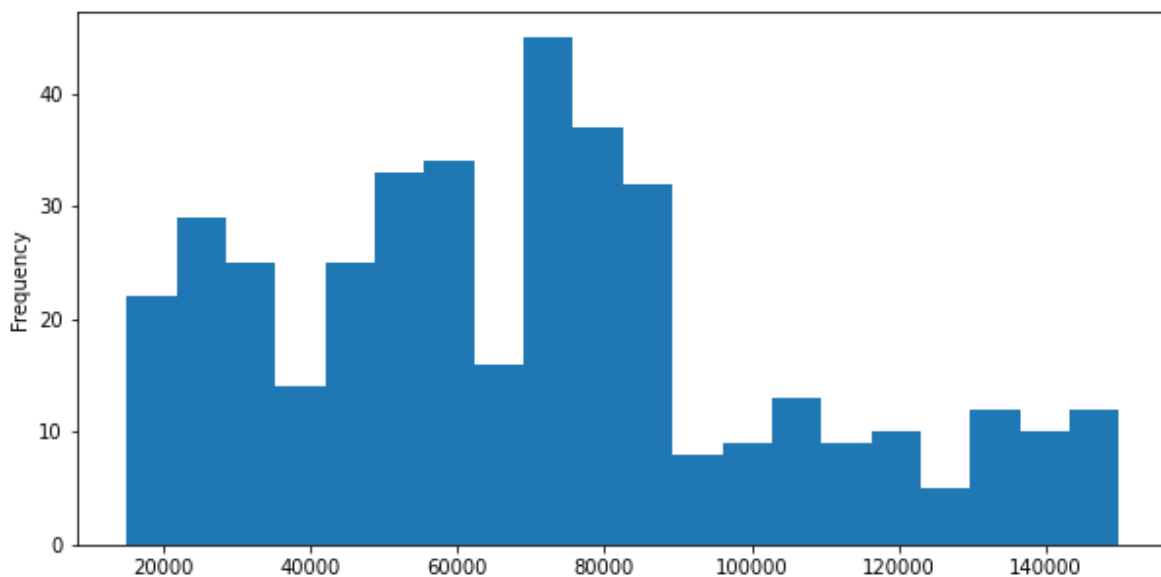


In [9]:

```
df["EstimatedSalary"].plot.hist(bins=20,figsize=(10,5))
```

Out[9]:

<AxesSubplot:ylabel='Frequency'>



In [10]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 400 entries, 0 to 399
Data columns (total 5 columns):
#   Column          Non-Null Count  Dtype
---  -
0   User ID         400 non-null    int64
1   Gender          400 non-null    object
2   Age             400 non-null    int64
3   EstimatedSalary 400 non-null    int64
4   Purchased       400 non-null    int64
dtypes: int64(4), object(1)
memory usage: 14.1+ KB
```

In [11]:

```
df.isnull()
```

Out[11]:

	User ID	Gender	Age	EstimatedSalary	Purchased
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
...	...	...	...	...	...
395	False	False	False	False	False
396	False	False	False	False	False
397	False	False	False	False	False
398	False	False	False	False	False
399	False	False	False	False	False

400 rows × 5 columns

In [13]:

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

Out[13]:

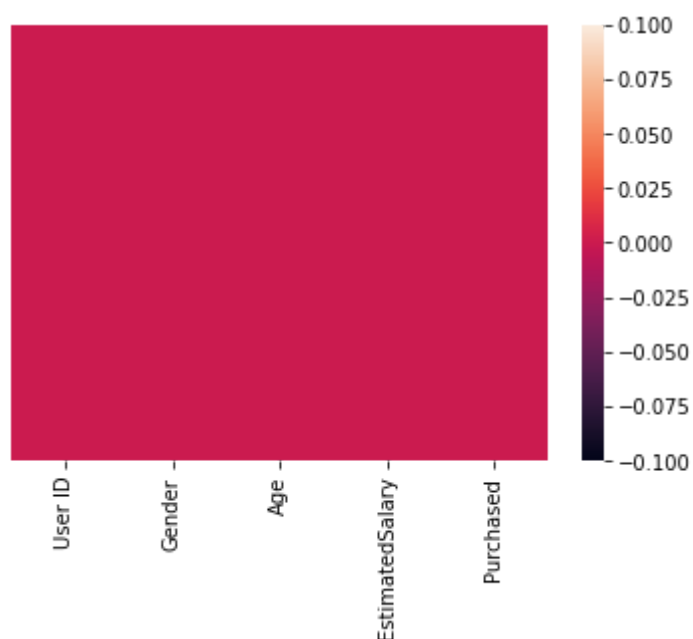
```
User ID      0
Gender       0
Age          0
EstimatedSalary  0
Purchased    0
dtype: int64
```

In [14]:

```
sns.heatmap(df.isnull() , yticklabels=False)
```

Out[14]:

<AxesSubplot:>

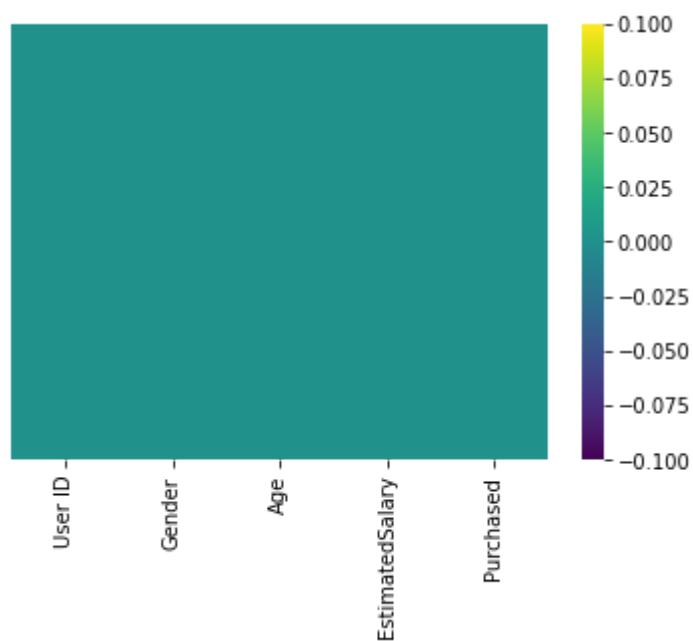


In [15]:

```
sns.heatmap(df.isnull() , yticklabels=False , cmap="viridis")
```

Out[15]:

<AxesSubplot:>

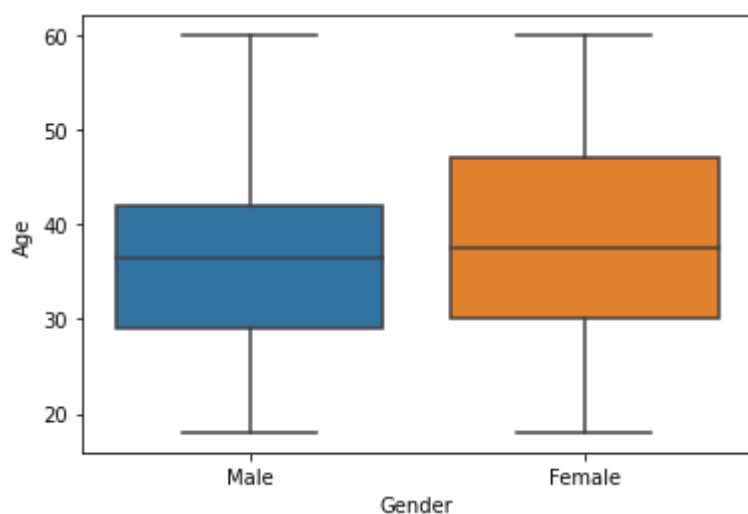


In [16]:

```
sns.boxplot(x="Gender",y="Age",data=df)
```

Out[16]:

<AxesSubplot:xlabel='Gender', ylabel='Age'>

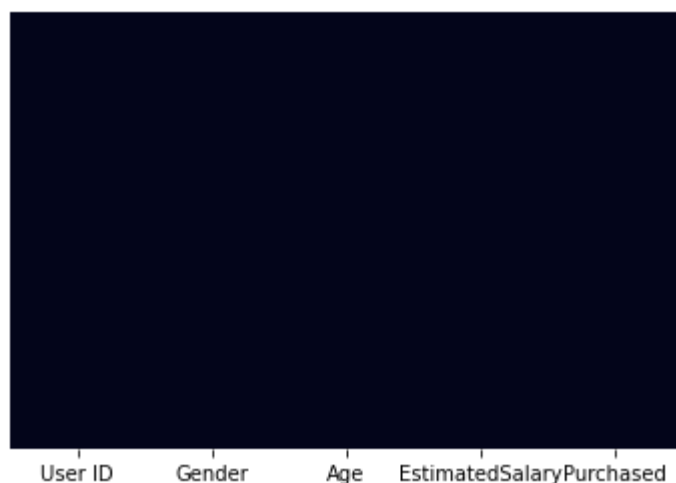


In [17]:

```
df.dropna(inplace=True)  
sns.heatmap(df.isnull(), yticklabels=False, cbar=False)
```

Out[17]:

<AxesSubplot:>



In [18]:

```
df.head(10)
```

Out[18]:

	User ID	Gender	Age	EstimatedSalary	Purchased
0	15624510	Male	19	19000	0
1	15810944	Male	35	20000	0
2	15668575	Female	26	43000	0
3	15603246	Female	27	57000	0
4	15804002	Male	19	76000	0
5	15728773	Male	27	58000	0
6	15598044	Female	27	84000	0
7	15694829	Female	32	150000	1
8	15600575	Male	25	33000	0
9	15727311	Female	35	65000	0

In [19]:

```
pd.get_dummies(df['Gender'], drop_first=True)
```

Out[19]:

	Male
0	1
1	1
2	0
3	0
4	1
...	...
395	0
396	1
397	0
398	1
399	0

400 rows × 1 columns



In [20]:

```
pd.get_dummies(df[ 'Gender' ])
```

Out[20]:

	Female	Male
0	0	1
1	0	1
2	1	0
3	1	0
4	0	1
...	...	...
395	1	0
396	0	1
397	1	0
398	0	1
399	1	0

400 rows × 2 columns

In [21]:

```
pd.get_dummies(df[ 'Gender' ] , drop_first=True)
```

Out[21]:

	Male
0	1
1	1
2	0
3	0
4	1
...	...
395	0
396	1
397	0
398	1
399	0

400 rows × 1 columns

In [23]:

```
pur=pd.get_dummies(df[ 'Purchased' ])
pur.head(10)
```

Out[23]:

	0	1
0	1	0
1	1	0
2	1	0
3	1	0
4	1	0
5	1	0
6	1	0
7	0	1
8	1	0
9	1	0

In [24]:

```
df=pd.concat([df,pur],axis=1)
```

In [25]:

```
df.head(10)
```

Out[25]:

	User ID	Gender	Age	EstimatedSalary	Purchased	0	1
0	15624510	Male	19	19000	0	1	0
1	15810944	Male	35	20000	0	1	0
2	15668575	Female	26	43000	0	1	0
3	15603246	Female	27	57000	0	1	0
4	15804002	Male	19	76000	0	1	0
5	15728773	Male	27	58000	0	1	0
6	15598044	Female	27	84000	0	1	0
7	15694829	Female	32	150000	1	0	1
8	15600575	Male	25	33000	0	1	0
9	15727311	Female	35	65000	0	1	0

In [44]:

```
X=df.drop("EstimatedSalary",axis=1)
y=df["EstimatedSalary"]
```

In [45]:

```
from sklearn.model_selection import train_test_split
```

In [46]:

```
X_train, X_test, y_train, y_test = train_test_split(X,y,test_size=0.3,random_state=1)
```

In [47]:

```
from sklearn.linear_model import LogisticRegression
```

In [48]:

```
logmodel=LogisticRegression()
```

In [51]:

```
from sklearn.metrics import confusion_matrix
```

In [ ]:

```
-----WITH MY DATASET FUTURE TOP 50 RESTAURANTS-----
```

In [54]:

```
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
%matplotlib inline
import math
df=pd.read_csv("Future50.csv")
df.head(5)
```

Out[54]:

	Rank	Restaurant	Location	Sales	YOY_Sales	Units	YOY_Units	Unit_Volume	Franchis
0	1	Evergreens	Seattle, Wash.	24	130.50%	26	116.70%	1150	N
1	2	Clean Juice	Charlotte, N.C.	44	121.90%	105	94.40%	560	Y
2	3	Slapfish	Huntington Beach, Calif.	21	81.00%	21	90.90%	1370	Y
3	4	Clean Eatz	Wilmington, N.C.	25	79.70%	46	58.60%	685	Y
4	5	Pokeworks	Irvine, Calif.	49	77.10%	50	56.30%	1210	Y

In [55]:

```
print("number of Restaurants : "+str(len(df.index)))
```

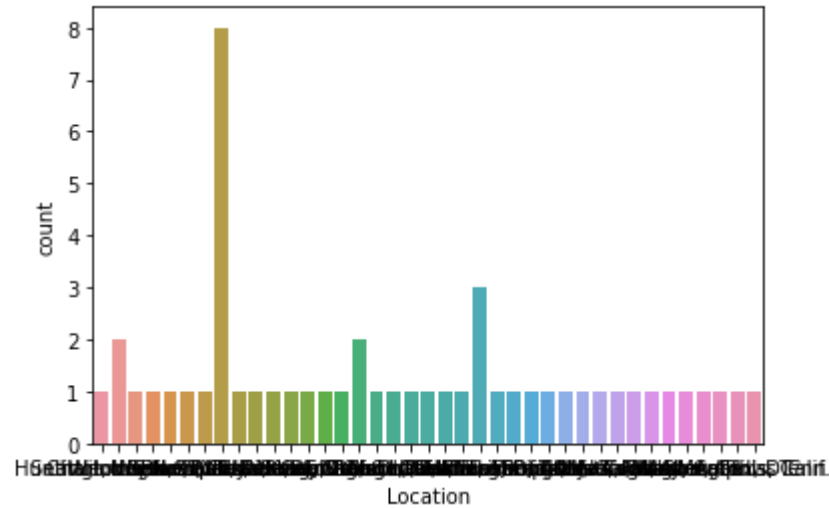
number of Restaurants : 50

In [56]:

```
sns.countplot(x="Location",data=df)
```

Out[56]:

<AxesSubplot:xlabel='Location', ylabel='count'>



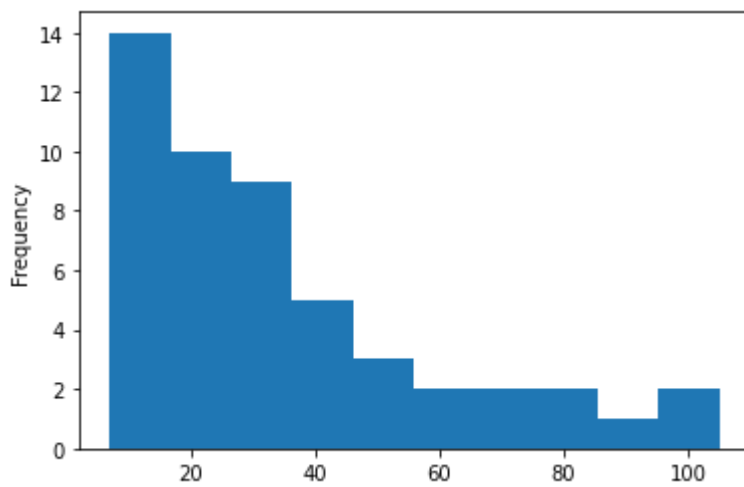


In [60]:

```
df["Units"].plot.hist()
```

Out[60]:

<AxesSubplot:ylabel='Frequency'>

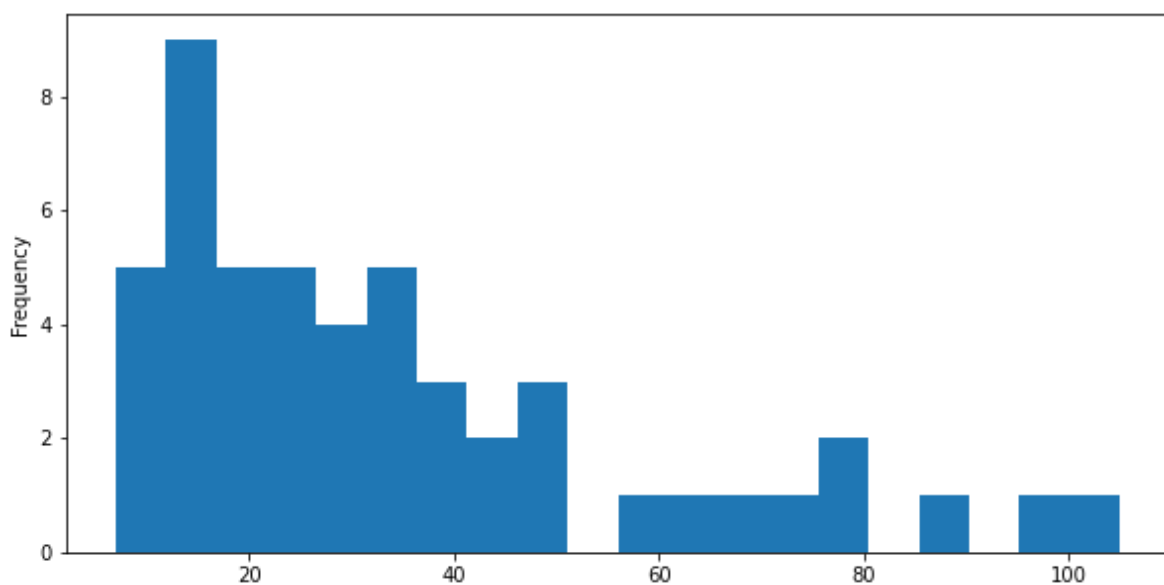


In [61]:

```
df["Units"].plot.hist(bins=20,figsize=(10,5))
```

Out[61]:

<AxesSubplot:ylabel='Frequency'>



In [62]:

```
df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 50 entries, 0 to 49
Data columns (total 9 columns):
#   Column          Non-Null Count  Dtype
---  -
0   Rank             50 non-null    int64
1   Restaurant       50 non-null    object
2   Location         50 non-null    object
3   Sales            50 non-null    int64
4   YOY_Sales        50 non-null    object
5   Units            50 non-null    int64
6   YOY_Units        50 non-null    object
7   Unit_Volume      50 non-null    int64
8   Franchising      50 non-null    object
dtypes: int64(4), object(5)
memory usage: 2.6+ KB
```

In [63]:

```
df.isnull()
```

Out[63]:

	Rank	Restaurant	Location	Sales	YOY_Sales	Units	YOY_Units	Unit_Volume	Franchising
0	False	False	False	False	False	False	False	False	False
1	False	False	False	False	False	False	False	False	False
2	False	False	False	False	False	False	False	False	False
3	False	False	False	False	False	False	False	False	False
4	False	False	False	False	False	False	False	False	False
5	False	False	False	False	False	False	False	False	False
6	False	False	False	False	False	False	False	False	False
7	False	False	False	False	False	False	False	False	False
8	False	False	False	False	False	False	False	False	False
9	False	False	False	False	False	False	False	False	False
10	False	False	False	False	False	False	False	False	False
11	False	False	False	False	False	False	False	False	False
12	False	False	False	False	False	False	False	False	False
13	False	False	False	False	False	False	False	False	False
14	False	False	False	False	False	False	False	False	False
15	False	False	False	False	False	False	False	False	False
16	False	False	False	False	False	False	False	False	False
17	False	False	False	False	False	False	False	False	False
18	False	False	False	False	False	False	False	False	False
19	False	False	False	False	False	False	False	False	False
20	False	False	False	False	False	False	False	False	False
21	False	False	False	False	False	False	False	False	False
22	False	False	False	False	False	False	False	False	False
23	False	False	False	False	False	False	False	False	False
24	False	False	False	False	False	False	False	False	False
25	False	False	False	False	False	False	False	False	False
26	False	False	False	False	False	False	False	False	False
27	False	False	False	False	False	False	False	False	False
28	False	False	False	False	False	False	False	False	False
29	False	False	False	False	False	False	False	False	False
30	False	False	False	False	False	False	False	False	False
31	False	False	False	False	False	False	False	False	False
32	False	False	False	False	False	False	False	False	False
33	False	False	False	False	False	False	False	False	False



	Rank	Restaurant	Location	Sales	YOY_Sales	Units	YOY_Units	Unit_Volume	Franchising
34	False	False	False	False	False	False	False	False	False
35	False	False	False	False	False	False	False	False	False
36	False	False	False	False	False	False	False	False	False
37	False	False	False	False	False	False	False	False	False
38	False	False	False	False	False	False	False	False	False
39	False	False	False	False	False	False	False	False	False
40	False	False	False	False	False	False	False	False	False
41	False	False	False	False	False	False	False	False	False
42	False	False	False	False	False	False	False	False	False
43	False	False	False	False	False	False	False	False	False
44	False	False	False	False	False	False	False	False	False
45	False	False	False	False	False	False	False	False	False
46	False	False	False	False	False	False	False	False	False
47	False	False	False	False	False	False	False	False	False
48	False	False	False	False	False	False	False	False	False
49	False	False	False	False	False	False	False	False	False

In [64]:

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

Out[64]:

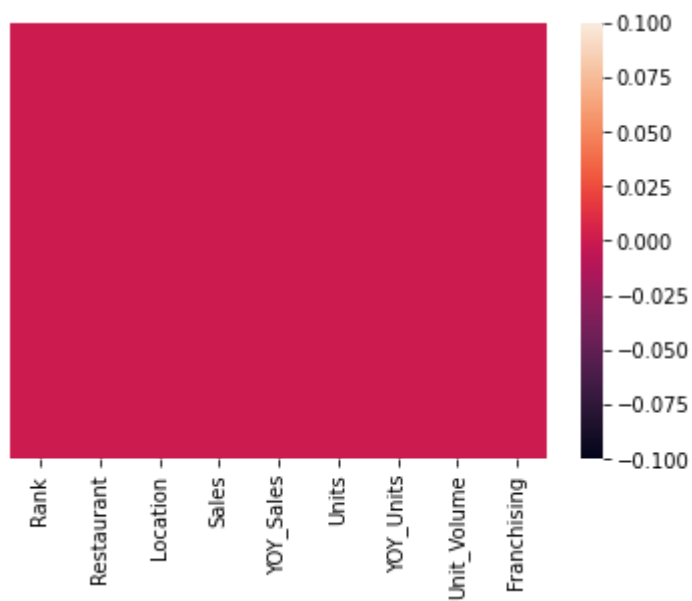
```
Rank      0
Restaurant 0
Location  0
Sales      0
YOY_Sales 0
Units      0
YOY_Units 0
Unit_Volume 0
Franchising 0
dtype: int64
```

In [65]:

```
sns.heatmap(df.isnull() , yticklabels=False)
```

Out[65]:

<AxesSubplot:>

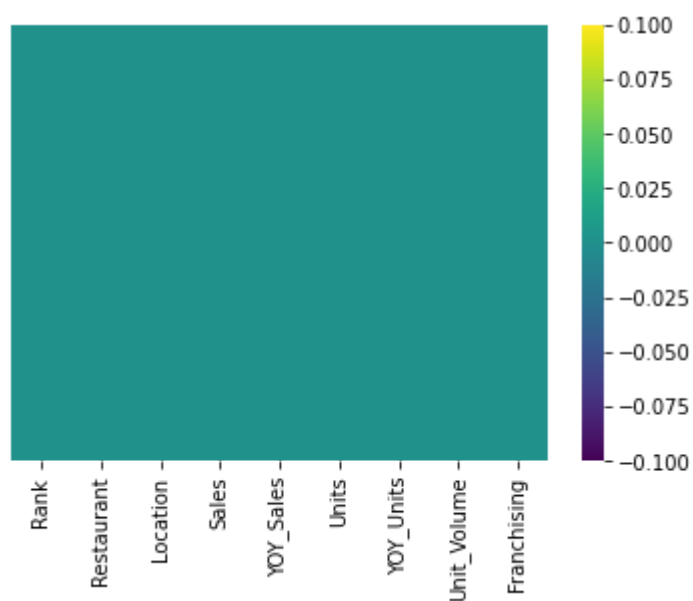


In [66]:

```
sns.heatmap(df.isnull() , yticklabels=False , cmap="viridis")
```

Out[66]:

<AxesSubplot:>

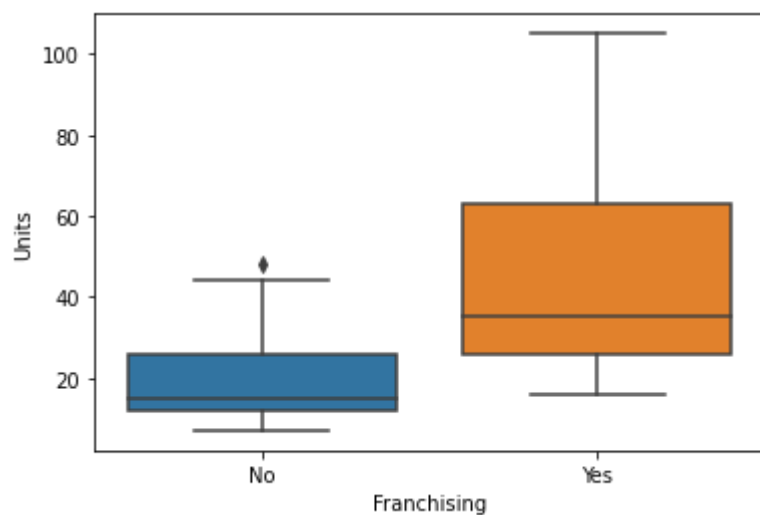


In [68]:

```
sns.boxplot(x="Franchising",y="Units",data=df)
```

Out[68]:

<AxesSubplot:xlabel='Franchising', ylabel='Units'>

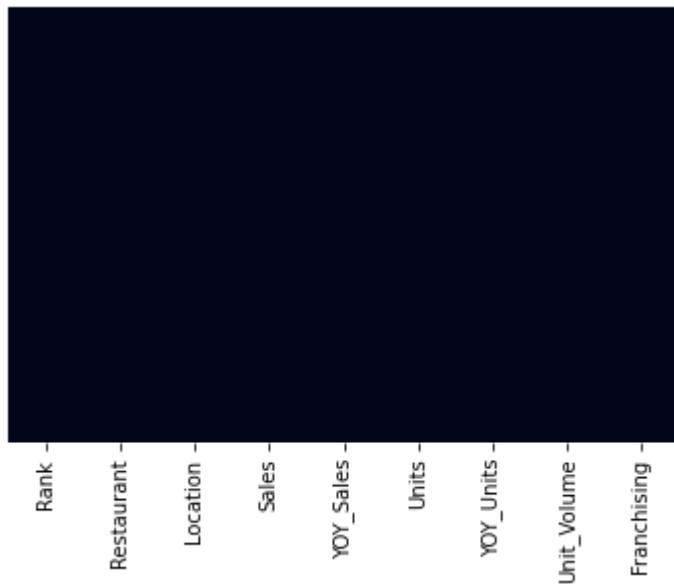


In [69]:

```
df.dropna(inplace=True)  
sns.heatmap(df.isnull() , yticklabels=False , cbar=False)
```

Out[69]:

<AxesSubplot:>

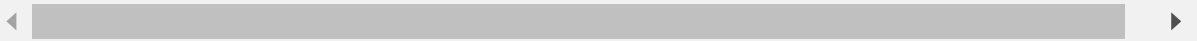


In [70]:

```
df.head(10)
```

Out[70]:

	Rank	Restaurant	Location	Sales	YOY_Sales	Units	YOY_Units	Unit_Volume	Franchisur
0	1	Evergreens	Seattle, Wash.	24	130.50%	26	116.70%	1150	N
1	2	Clean Juice	Charlotte, N.C.	44	121.90%	105	94.40%	560	Y€
2	3	Slapfish	Huntington Beach, Calif.	21	81.00%	21	90.90%	1370	Y€
3	4	Clean Eatz	Wilmington, N.C.	25	79.70%	46	58.60%	685	Y€
4	5	Pokeworks	Irvine, Calif.	49	77.10%	50	56.30%	1210	Y€
5	6	Playa Bowls	Belmar, N.J.	39	62.90%	76	28.80%	580	Y€
6	7	The Simple Greek	Blue Bell, Pa.	24	52.50%	36	33.30%	775	Y€
7	8	Melt Shop	New York, N.Y.	20	39.60%	19	35.70%	1260	Y€
8	9	Creamistry	Yorba Linda, Calif.	24	36.80%	60	27.70%	465	Y€
9	10	Joella's Hot Chicken	Louisville, Ky.	29	35.50%	17	30.80%	1930	N



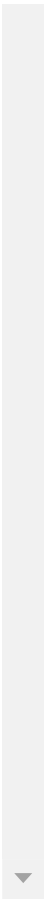
In [71]:

```
pd.get_dummies(df['Franchising'], drop_first=True)
```

Out[71]:

	Yes
0	0
1	1
2	1
3	1
4	1
5	1
6	1
7	1
8	1
9	0
10	1
11	1
12	1
13	0
14	0
15	1
16	1
17	1
18	0
19	1
20	0
21	1
22	0
23	1
24	0
25	0
26	0
27	1
28	0
29	1
30	1
31	0
32	0
33	0

Yes	
34	1
35	1
36	1
37	0
38	0
39	1
40	0
41	1
42	0
43	1
44	1
45	1
46	0
47	0
48	0
49	1



In [72]:

```
pd.get_dummies(df['Franchising'])
```

Out[72]:

	No	Yes
0	1	0
1	0	1
2	0	1
3	0	1
4	0	1
5	0	1
6	0	1
7	0	1
8	0	1
9	1	0
10	0	1
11	0	1
12	0	1
13	1	0
14	1	0
15	0	1
16	0	1
17	0	1
18	1	0
19	0	1
20	1	0
21	0	1
22	1	0
23	0	1
24	1	0
25	1	0
26	1	0
27	0	1
28	1	0
29	0	1
30	0	1
31	1	0
32	1	0
33	1	0



	No	Yes
34	0	1
35	0	1
36	0	1
37	1	0
38	1	0
39	0	1
40	1	0
41	0	1
42	1	0
43	0	1
44	0	1
45	0	1
46	1	0
47	1	0
48	1	0
49	0	1



In [73]:

```
pd.get_dummies(df['Franchising'] , drop_first=True)
```

Out[73]:

	Yes
0	0
1	1
2	1
3	1
4	1
5	1
6	1
7	1
8	1
9	0
10	1
11	1
12	1
13	0
14	0
15	1
16	1
17	1
18	0
19	1
20	0
21	1
22	0
23	1
24	0
25	0
26	0
27	1
28	0
29	1
30	1
31	0
32	0
33	0

Yes	
34	1
35	1
36	1
37	0
38	0
39	1
40	0
41	1
42	0
43	1
44	1
45	1
46	0
47	0
48	0
49	1

In [74]:

```
pur=pd.get_dummies(df['Location'])
pur.head(10)
```

Out[74]:

	Agoura Hills, Calif.	Anaheim, Calif.	Atlanta, Ga.	Belmar, N.J.	Blue Bell, Pa.	Charlotte, N.C.	Columbus, Ohio	Conway, Ark.	Denver, Colo.	Doral, Fla.
0	0	0	0	0	0	0	0	0	0	0
1	0	0	0	0	0	1	0	0	0	0
2	0	0	0	0	0	0	0	0	0	0
3	0	0	0	0	0	0	0	0	0	0
4	0	0	0	0	0	0	0	0	0	0
5	0	0	0	1	0	0	0	0	0	0
6	0	0	0	0	1	0	0	0	0	0
7	0	0	0	0	0	0	0	0	0	0
8	0	0	0	0	0	0	0	0	0	0
9	0	0	0	0	0	0	0	0	0	0

10 rows × 39 columns

In [75]:

```
df=pd.concat([df,pur],axis=1)
```

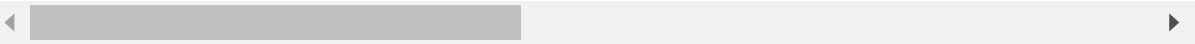
In [76]:

```
df.head(10)
```

Out[76]:

	Rank	Restaurant	Location	Sales	YOY_Sales	Units	YOY_Units	Unit_Volume	Franchisir
0	1	Evergreens	Seattle, Wash.	24	130.50%	26	116.70%	1150	N
1	2	Clean Juice	Charlotte, N.C.	44	121.90%	105	94.40%	560	Y
2	3	Slapfish	Huntington Beach, Calif.	21	81.00%	21	90.90%	1370	Y
3	4	Clean Eat	Wilmington, N.C.	25	79.70%	46	58.60%	685	Y
4	5	Pokeworks	Irvine, Calif.	49	77.10%	50	56.30%	1210	Y
5	6	Playa Bowls	Belmar, N.J.	39	62.90%	76	28.80%	580	Y
6	7	The Simple Greek	Blue Bell, Pa.	24	52.50%	36	33.30%	775	Y
7	8	Melt Shop	New York, N.Y.	20	39.60%	19	35.70%	1260	Y
8	9	Creamistry	Yorba Linda, Calif.	24	36.80%	60	27.70%	465	Y
9	10	Joella's Hot Chicken	Louisville, Ky.	29	35.50%	17	30.80%	1930	N

10 rows × 48 columns



In [ ]: