

IMPORTANT PLOT IN DATA ANALYSIS



DATA ANALYSIS

Seaborn is a data visualization library based on MATPLOTLIB. it provides a high-level interface for drawing attractive and informative statistical graphics.

1.Importance plot in data visualization every data science student should know this:

- A) Count plot
- B) Bar plot
- C) Box plot
- D) Violin plot
- E) Strip plot
- F) Swarm plot
- G) Factor plot
- H) Dis plot
- I) Join plot
- K) Pair plot
- L) Rug plot
- M) Heatmap
- N) Scatter plot

In []:

2.Importing all the required libraries

In [1]:

```
import seaborn as sns
import numpy as np
import pandas as pd
import matplotlib.pyplot as plt
```

Import the data

In [2]:

```
mart = pd.read_excel(r"supermarket_sales.xlsx")
mart.shape
```

Out[2]:

(1000, 17)

In [3]:

```
mart.columns
```

Out[3]:

```
Index(['Invoice ID', 'Branch', 'City', 'Customer type', 'Gender',
       'Product line', 'Unit price', 'Quantity', 'Tax 5%', 'Total',
       'Date', 'Time', 'Payment', 'cogs', 'gross margin percentage',
       'gross income', 'Rating'],
      dtype='object')
```

In [4]:

```
mart.head()
```

Out[4]:

	Invoice ID	Branch	City	Customer type	Gender	Product line	Rating
0	692-92-5582	B	Mandalay	Member	Female	Food_and_beverages	5
1	351-62-0822	B	Mandalay	Member	Female	Fashion_accessories	1
2	529-56-3974	B	Mandalay	Member	Male	Electronic_accessories	2
3	299-46-1805	B	Mandalay	Member	Female	Sports_and_travel	9
4	319-50-3348	B	Mandalay	Normal	Female	Home_and_lifestyle	4



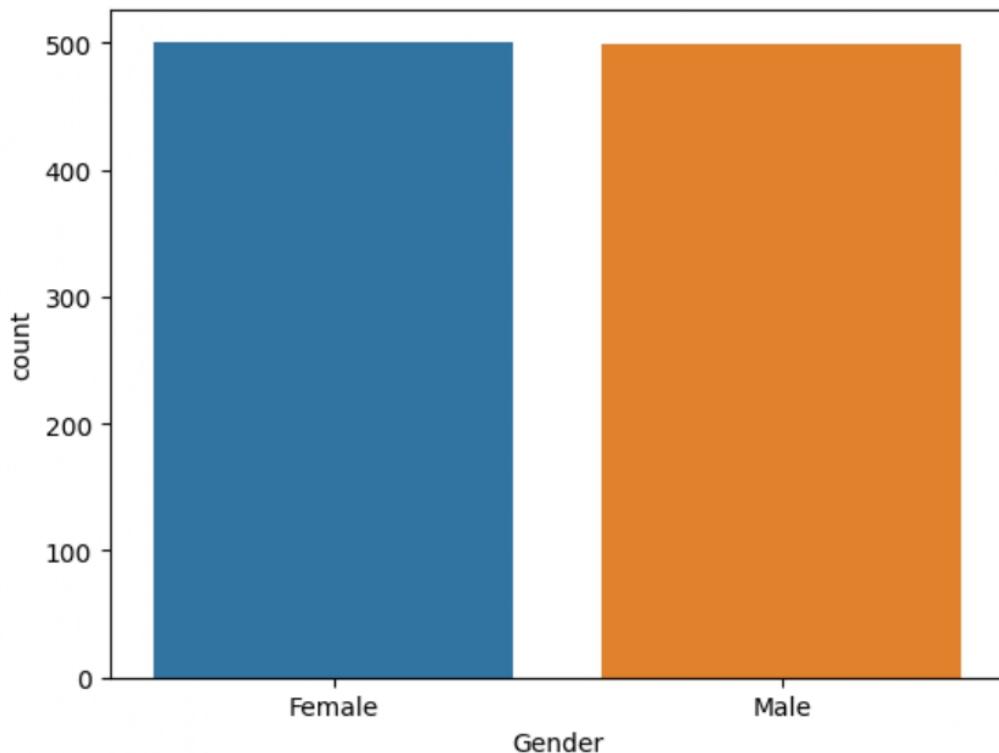
4.create a countplot

In [5]:

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

Out[5]:

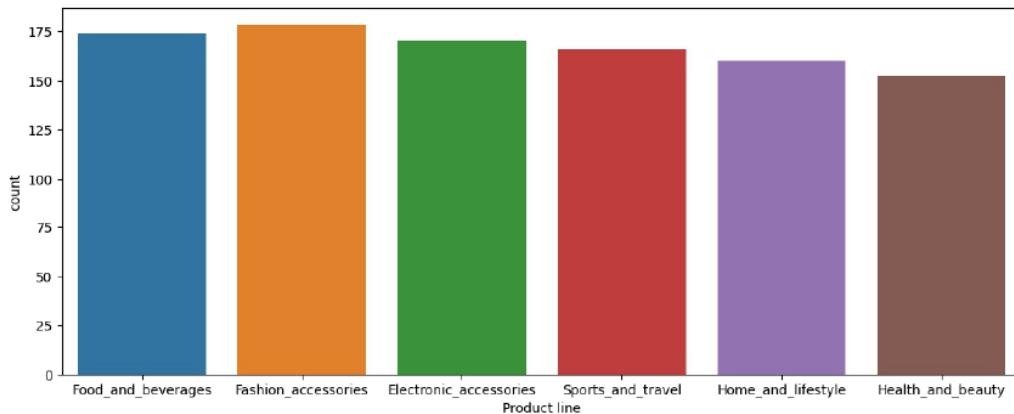
```
<AxesSubplot: xlabel='Gender', ylabel='count'>
```



5.create a basic COUNTPLOT to get number of transaction for each of the product line and change the figure size

In [6]:

```
plt.figure(figsize = (13,5))
sns.countplot(x = "Product line",data = mart)
plt.show()
```



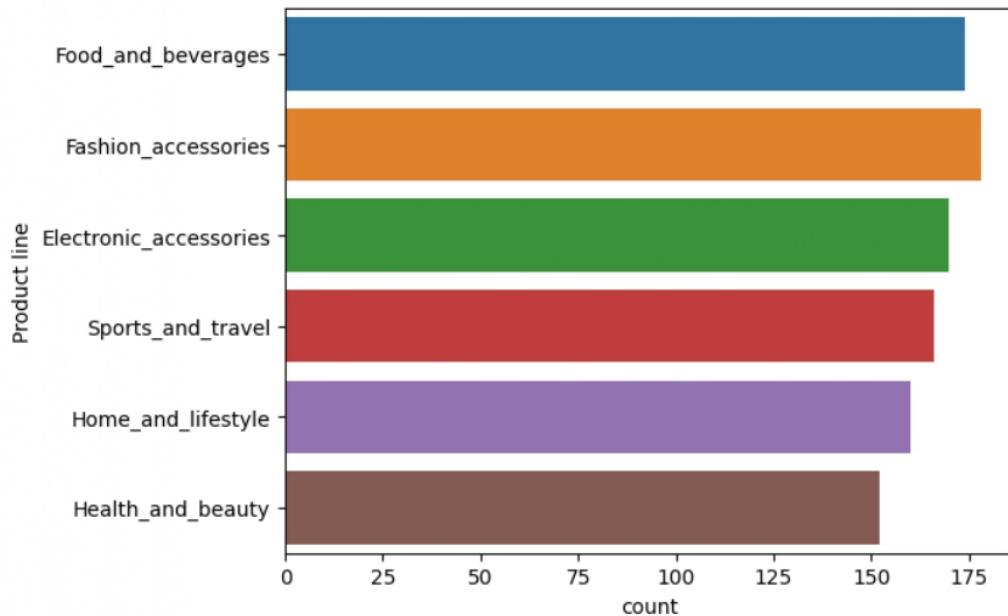
6. Make it horizontal bar plot

In [7]:

```
sns.countplot(y = "Product line",data = mart)
```

Out[7]:

```
<AxesSubplot: xlabel='count', ylabel='Product line'>
```



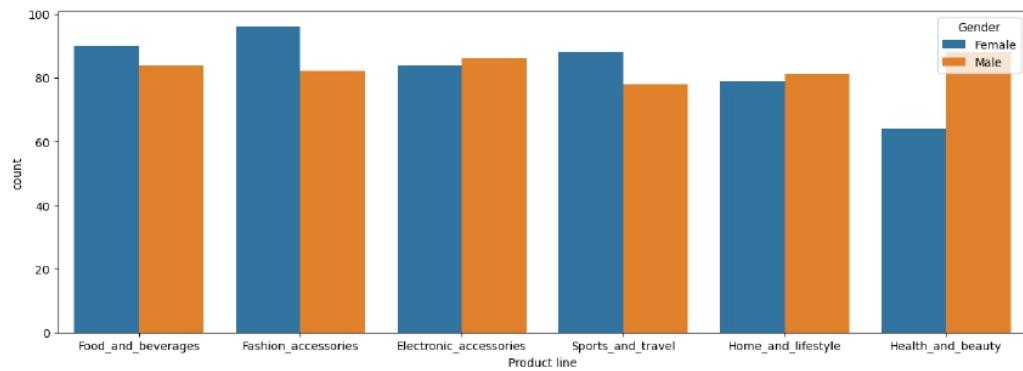
7.Add hue get the count on two catagories that is product line and gender

In [8]:

```
plt.figure(figsize = (15,5))
sns.countplot(x = "Product line",hue = "Gender",data = mart)
```

Out[8]:

```
<AxesSubplot: xlabel='Product line', ylabel='count'>
```



B)1.Bar plot

In [9]:

```
sns.get_dataset_names()
```

Out[9]:

```
['anagrams',
 'anscombe',
 'attention',
 'brain_networks',
 'car_crashes',
 'diamonds',
 'dots',
 'dowjones',
 'exercise',
 'flights',
 'fmri',
 'geyser',
 'glue',
 'healthexp',
 'iris',
 'mpg',
 'penguins',
 'planets',
 'seoice',
 'taxis',
 'tips',
 'titanic']
```

2.Import file

In [10]:

```
titanic = pd.read_csv("train_and_test2.csv")
```

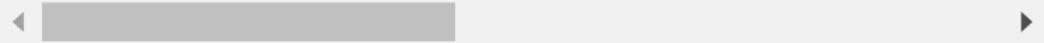
In [11]:

```
titanic.head()
```

Out[11]:

	Passengerid	Age	Fare	Sex	sibsp	zero	zero.1	zero.2	zero.3	z
0	1	22.0	7.2500	0	1	0	0	0	0	0
1	2	38.0	71.2833	1	1	0	0	0	0	0
2	3	26.0	7.9250	1	0	0	0	0	0	0
3	4	35.0	53.1000	1	1	0	0	0	0	0
4	5	35.0	8.0500	0	0	0	0	0	0	0

5 rows × 28 columns



In [12]:

```
titanic.columns
```

Out[12]:

```
Index(['Passengerid', 'Age', 'Fare', 'Sex', 'sibsp', 'zero', 'zero.1',
       'zero.2', 'zero.3', 'zero.4', 'zero.5', 'zero.6',
       'Parch', 'zero.7',
       'zero.8', 'zero.9', 'zero.10', 'zero.11', 'zero.12',
       'zero.13',
       'zero.14', 'Pclass', 'zero.15', 'zero.16', 'Embarked',
       'zero.17',
       'zero.18', '2urvived'],
      dtype='object')
```

In [13]:

```
titanic.index
```

Out[13]:

```
RangeIndex(start=0, stop=1309, step=1)
```

In [14]:

```
mart = pd.read_excel(r"supermarket_sales.xlsx")
```

In [15]:

```
mart.head()
```

Out[15]:

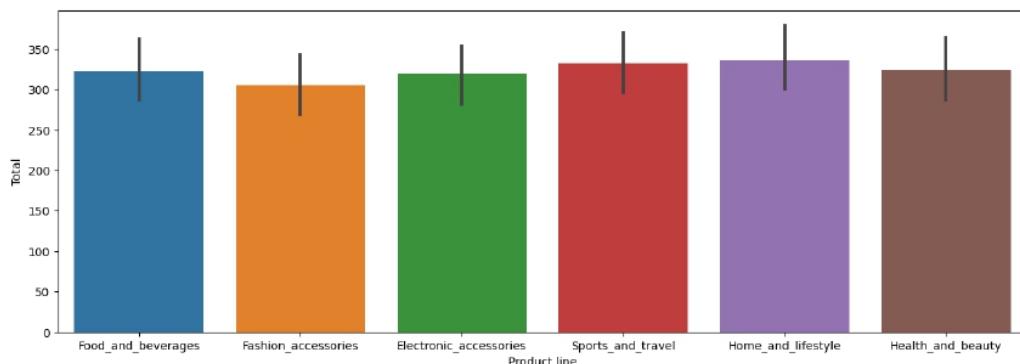
	Invoice ID	Branch	City	Customer type	Gender	Product line	Total
0	692-92-5582	B	Mandalay	Member	Female	Food_and_beverages	5
1	351-62-0822	B	Mandalay	Member	Female	Fashion_accessories	1
2	529-56-3974	B	Mandalay	Member	Male	Electronic_accessories	2
3	299-46-1805	B	Mandalay	Member	Female	Sports_and_travel	9
4	319-50-3348	B	Mandalay	Normal	Female	Home_and_lifestyle	4

In [16]:

```
plt.figure(figsize =(15,5))
sns.barplot(x = "Product line",y = "Total",data = mart)
```

Out[16]:

```
<AxesSubplot: xlabel='Product line', ylabel='Total'>
```



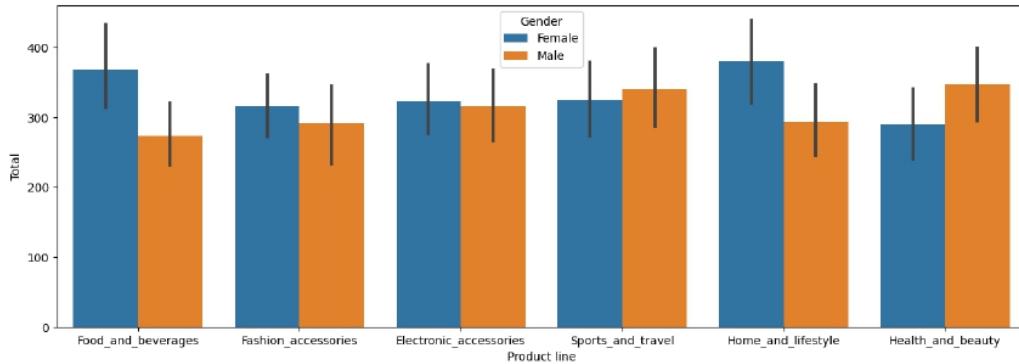
3.Add the HUE the barplot

In [17]:

```
plt.figure(figsize =(15,5))
sns.barplot(x = "Product line",y = "Total",hue ="Gender" ,data = mart)
```

Out[17]:

```
<AxesSubplot: xlabel='Product line', ylabel='Total'>
```



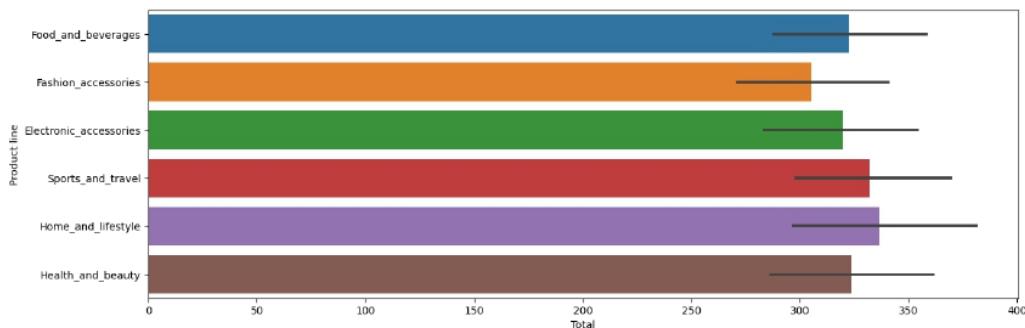
4. Make the barplot HORIZONTAL

In [18]:

```
plt.figure(figsize =(15,5))
sns.barplot(y = "Product line",x= "Total",data = mart)
```

Out[18]:

```
<AxesSubplot: xlabel='Total', ylabel='Product line'>
```



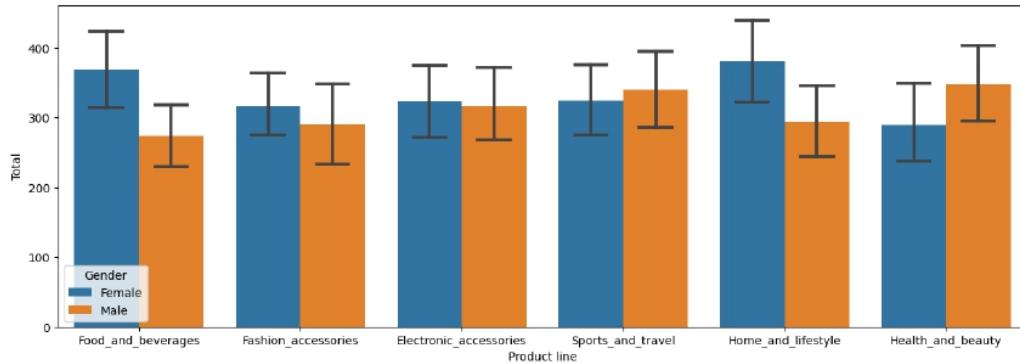
5. Add cap on the ERROR BAR

In [19]:

```
plt.figure(figsize =(15,5))
sns.barplot(x = "Product line",y = "Total",hue ="Gender" ,capsize = 0.2,c
```

Out[19]:

```
<AxesSubplot: xlabel='Product line', ylabel='Total'>
```



C) BOXPLOT

In []:

1.create a basic BOX plot one NUMERIC variable

In [20]:

```
mart = pd.read_excel(r"supermarket_sales.xlsx")
```

In [21]:

```
mart.head()
```

Out[21]:

	Invoice ID	Branch	City	Customer type	Gender	Product line	Rating
0	692-92-5582	B	Mandalay	Member	Female	Food_and_beverages	5
1	351-62-0822	B	Mandalay	Member	Female	Fashion_accessories	1
2	529-56-3974	B	Mandalay	Member	Male	Electronic_accessories	2
3	299-46-1805	B	Mandalay	Member	Female	Sports_and_travel	9
4	319-50-3348	B	Mandalay	Normal	Female	Home_and_lifestyle	4

In [22]:

```
mart.columns
```

Out[22]:

```
Index(['Invoice ID', 'Branch', 'City', 'Customer type', 'Gender',
       'Product line', 'Unit price', 'Quantity', 'Tax 5%', 'Total',
       'Date', 'Time', 'Payment', 'cogs', 'gross margin percentage',
       'gross income', 'Rating'],
      dtype='object')
```

In [23]:

```
mart.index
```

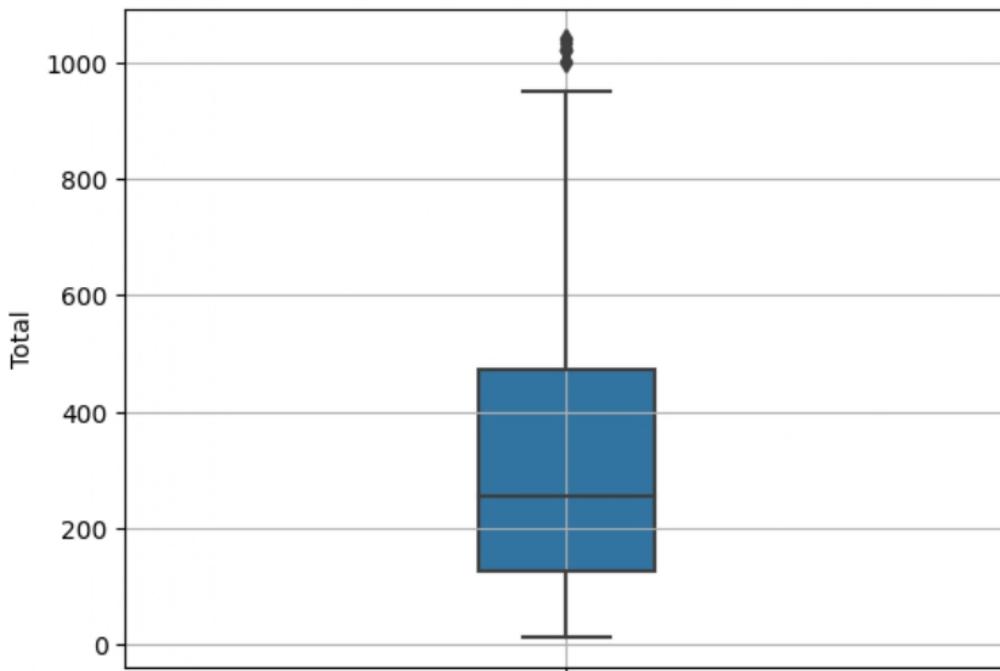
Out[23]:

```
RangeIndex(start=0, stop=1000, step=1)
```

In []:

In [24]:

```
sns.boxplot(y= "Total",data = mart,width = 0.2)  
plt.grid()
```



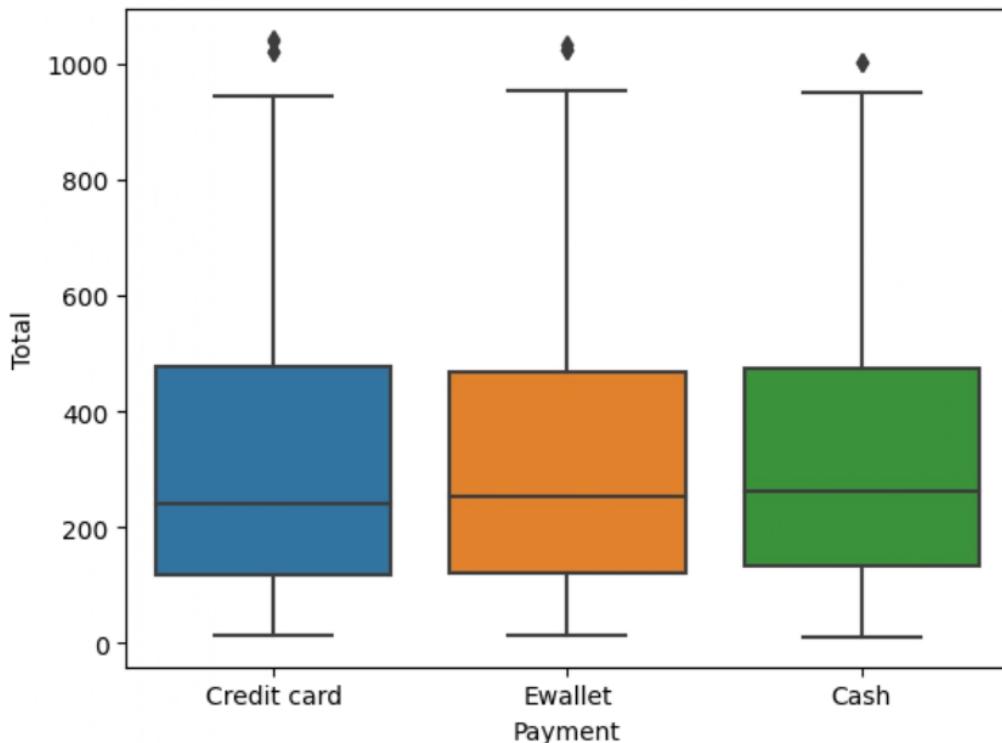
2.create a basic BOX plot one NUMERIC variable by CATEGORICAL variable

In [25]:

```
sns.boxplot(x = "Payment",y = "Total",data = mart)
```

Out[25]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```



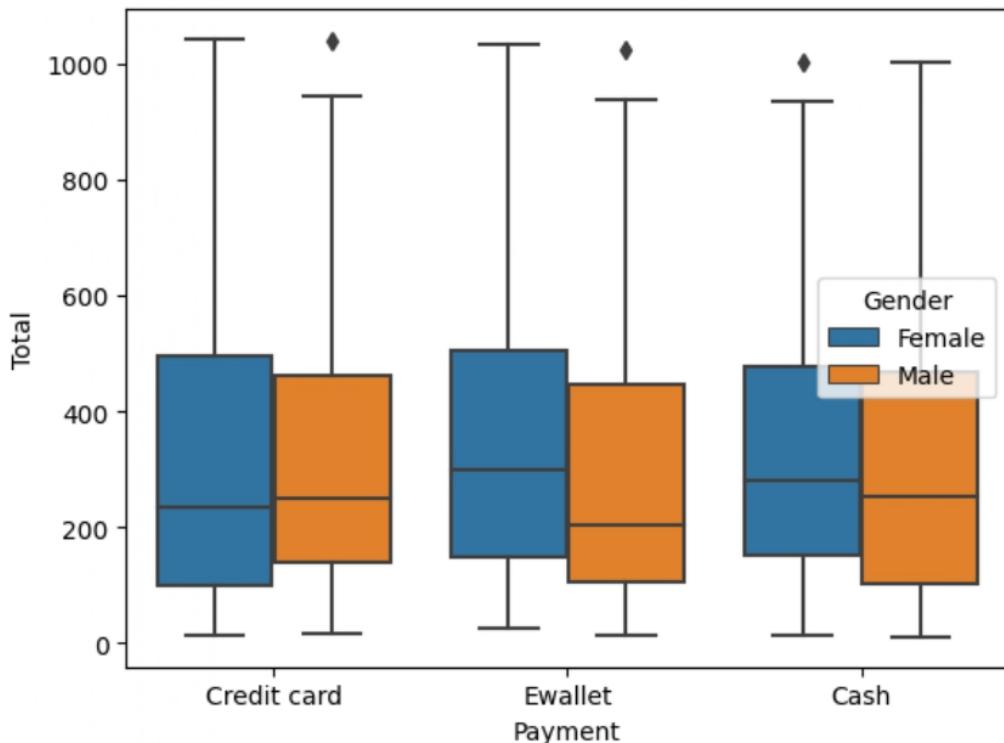
3.create a boxplot on one numeric variacle by two catogorical variable using HUE attribute

In [26]:

```
sns.boxplot(x = "Payment",y = "Total",hue = "Gender",data = mart)
```

Out[26]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```



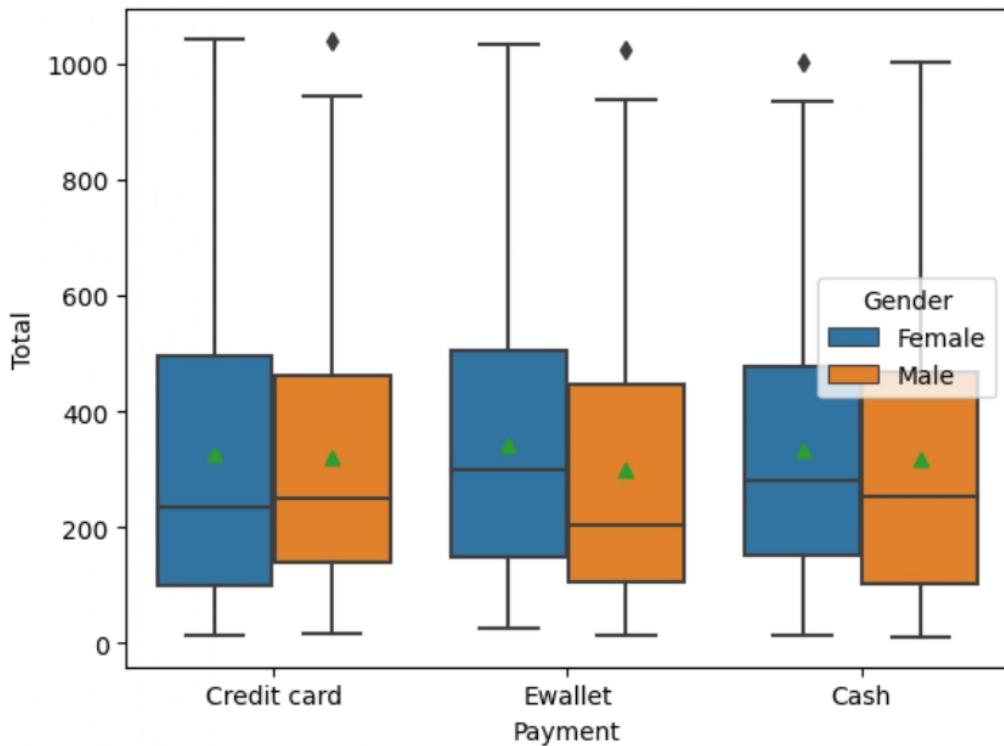
4.Add mean marker in the box plot using showmeans attribute and change its style using meanprops

In [27]:

```
sns.boxplot(x = "Payment",y = "Total",hue = "Gender",data = mart,showmear
```

Out[27]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```

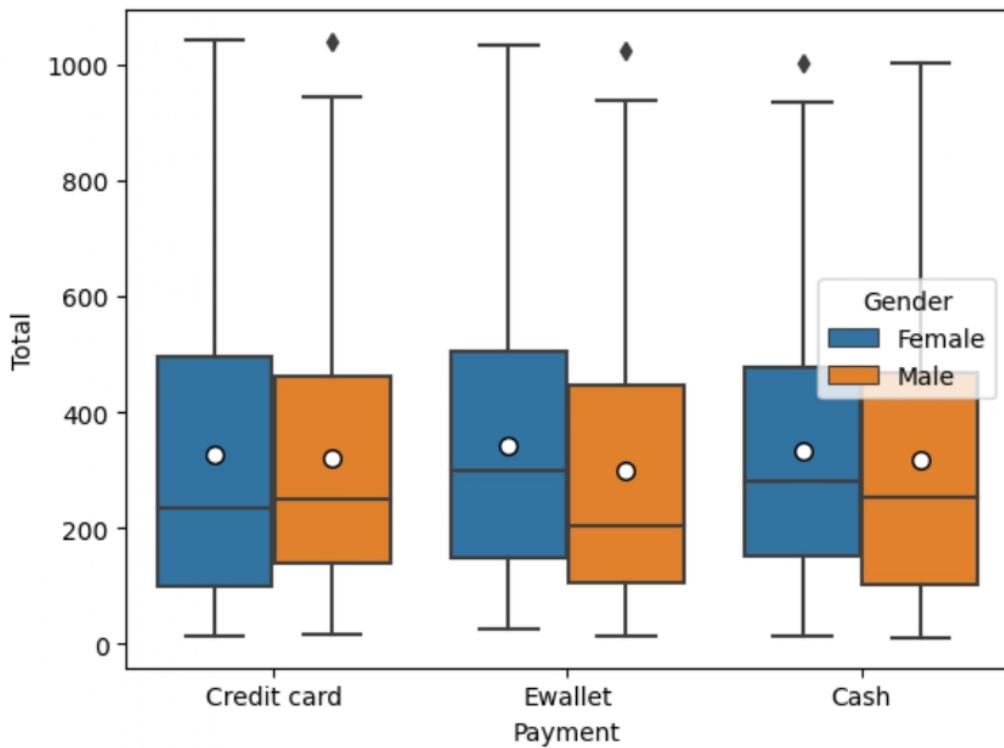


In [28]:

```
sns.boxplot(x = "Payment",y = "Total",hue = "Gender",data = mart,showmean=True)
```

Out[28]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```



D) VIOLIN PLOT

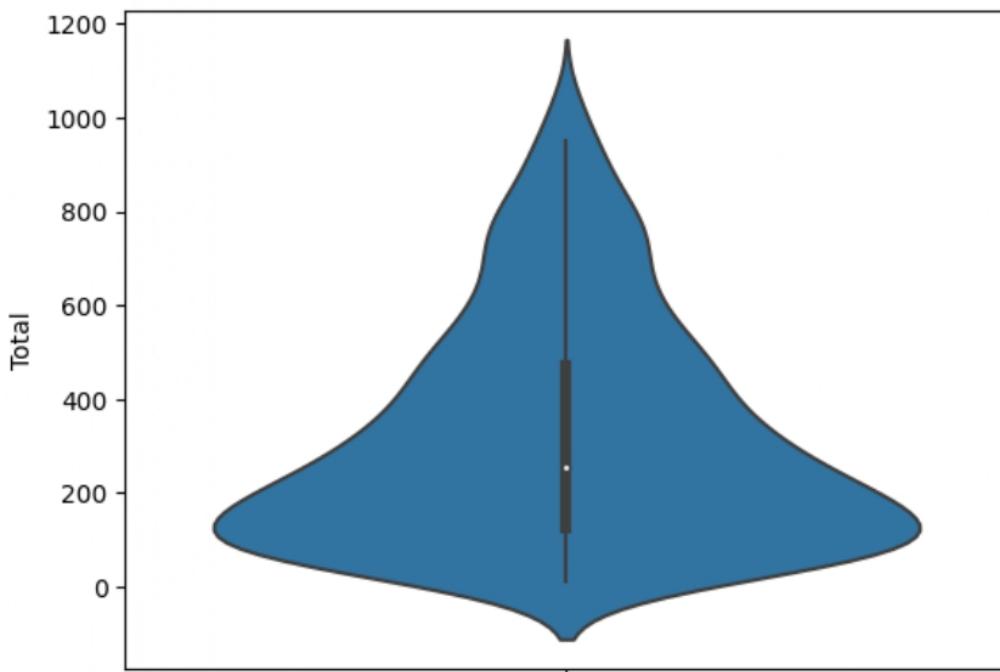
1.create a basic violinplot

In [29]:

```
sns.violinplot(y = "Total", data = mart)
```

Out[29]:

```
<AxesSubplot: ylabel='Total'>
```



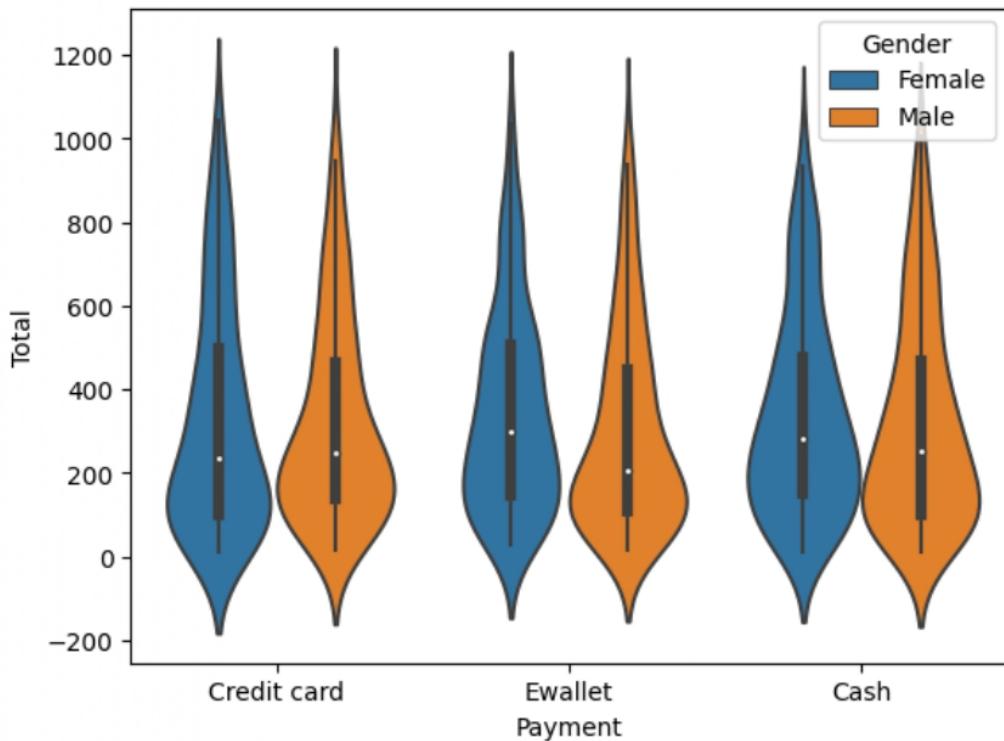
2.create a violinplot on two categorical and numeric variable and use split

In [30]:

```
sns.violinplot(x = "Payment",y ="Total",hue = "Gender",data = mart)
```

Out[30]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```

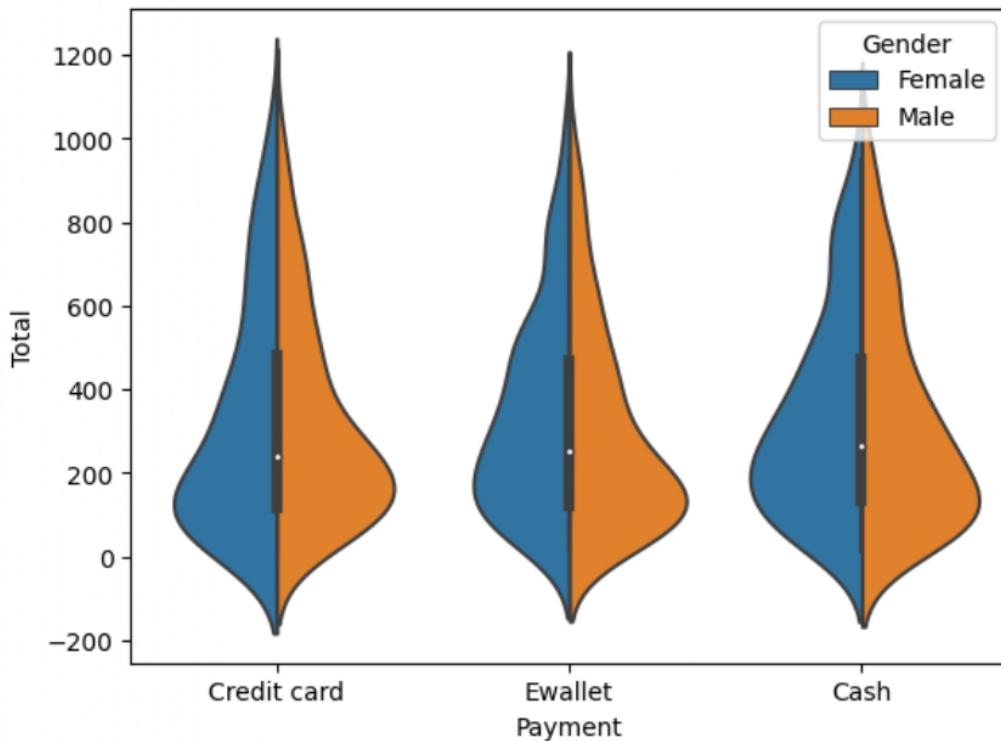


In [31]:

```
sns.violinplot(x = "Payment",y ="Total",hue = "Gender",data = mart,split
```

Out[31]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```



In []:

E) Strip plot

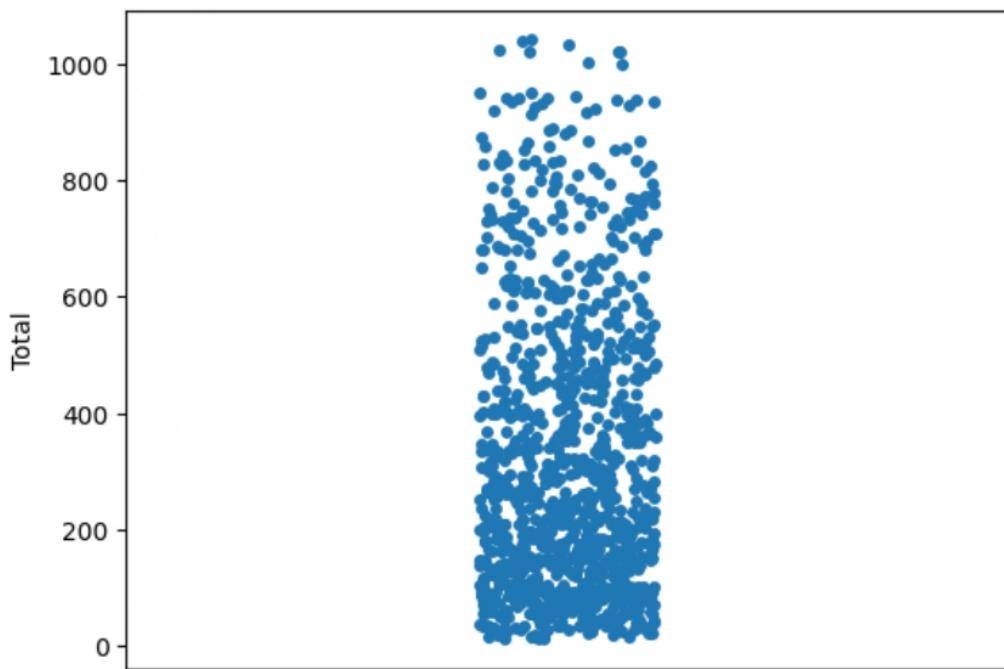
1.create a basic strip plot

In [32]:

```
sns.stripplot(y = "Total", data = mart)
```

Out[32]:

```
<AxesSubplot: ylabel='Total'>
```



2. Expand markers in strip plot using jitter

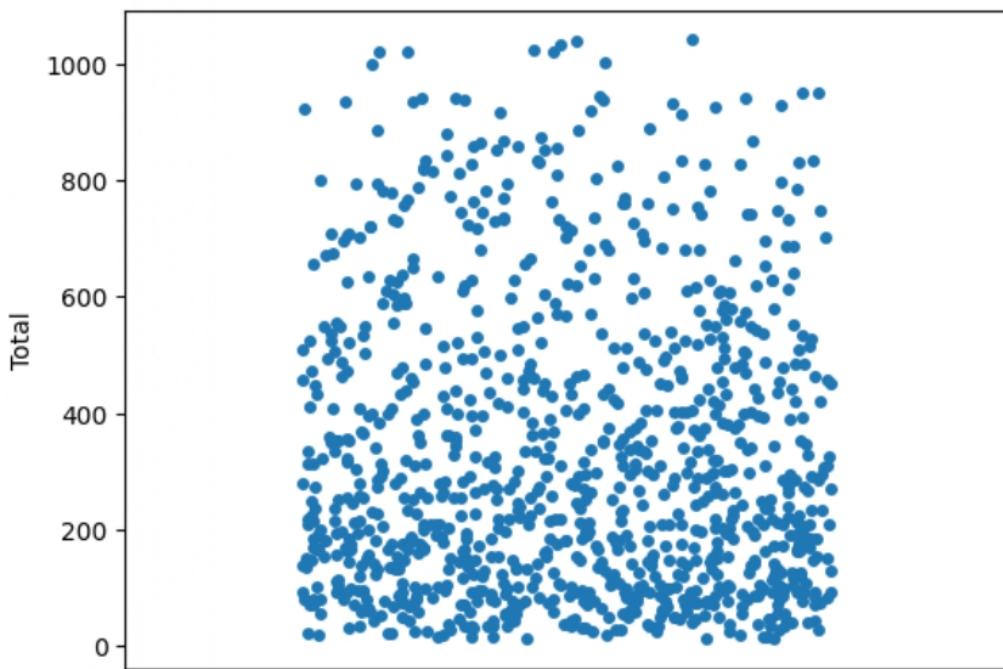


In [33]:

```
sns.stripplot(y = "Total", data = mart, jitter = 0.3)
```

Out[33]:

```
<AxesSubplot: ylabel='Total'>
```



In [34]:

```
# name = sns.load_dataset("titanic")
```

In [35]:

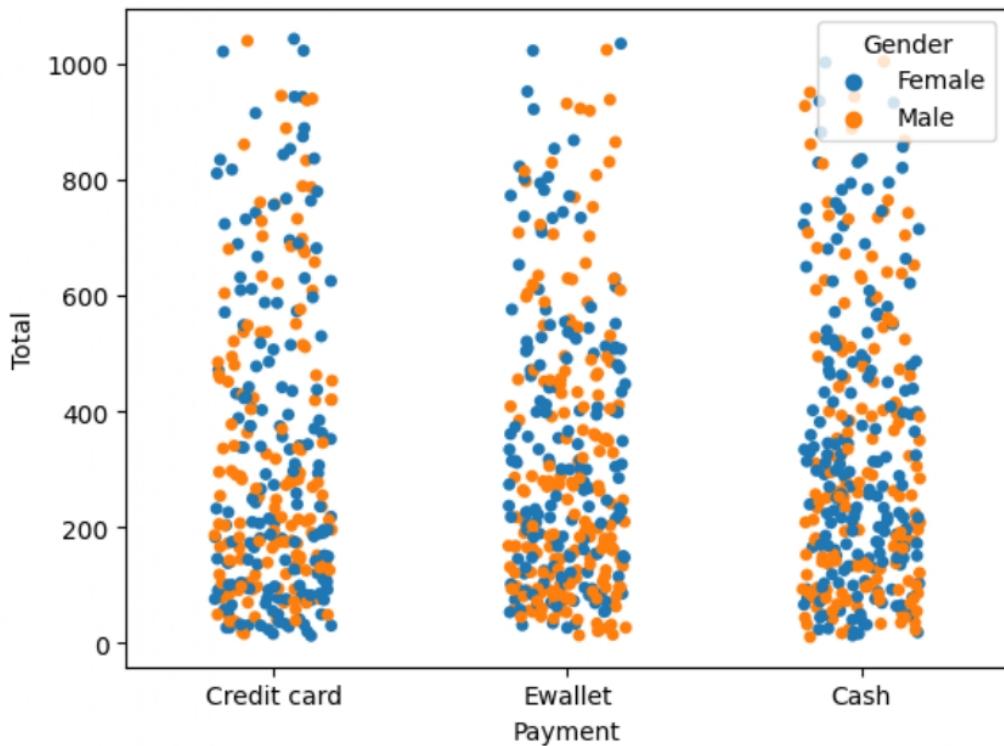
```
# name.head()
```

In [36]:

```
sns.stripplot(data = mart,y = 'Total',x = 'Payment',hue= 'Gender',jitter
```

Out[36]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```



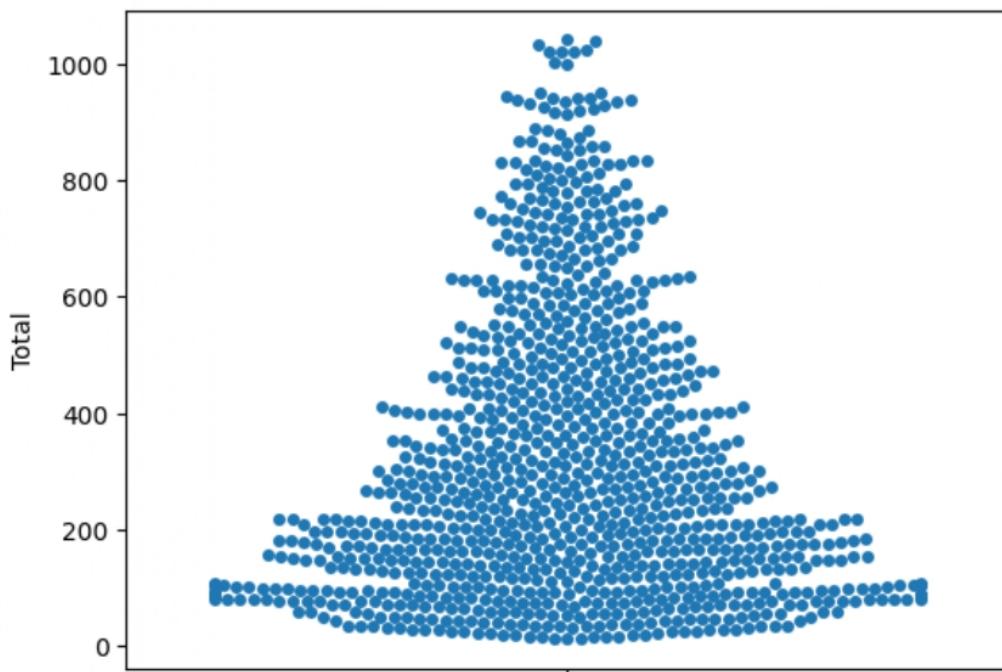
F) Swarm plot

In [37]:

```
sns.swarmplot(data = mart,y = "Total")
```

Out[37]:

```
<AxesSubplot: ylabel='Total'>
```



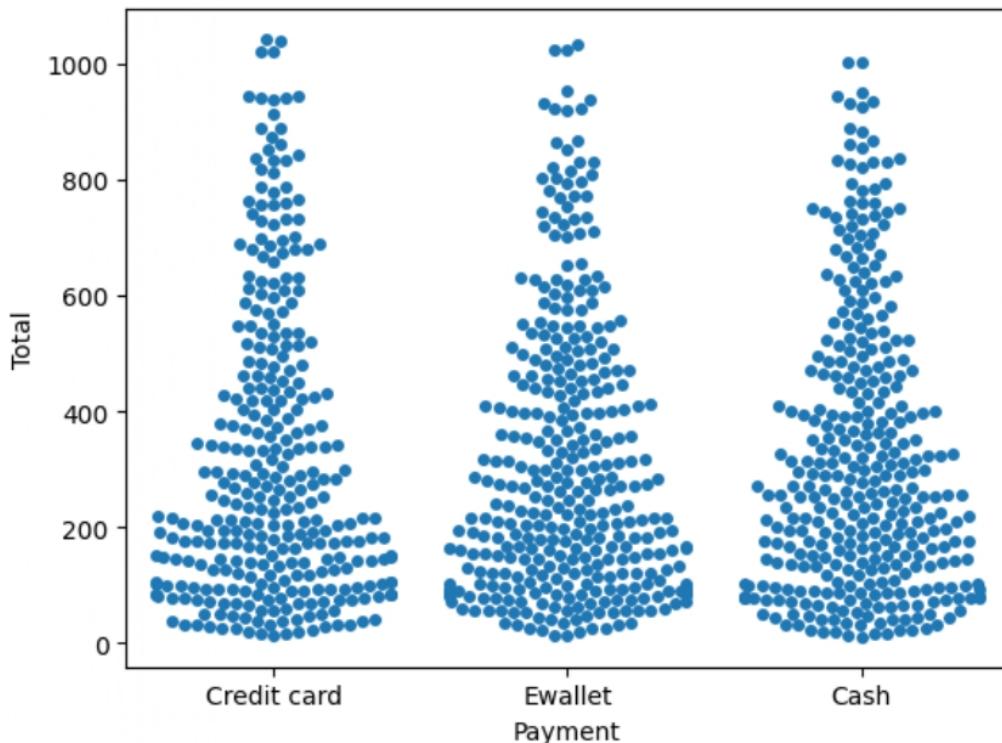
2.create swarm plot group by categories

In [38]:

```
sns.swarmplot(data = mart,y= "Total",x = "Payment")
```

Out[38]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```

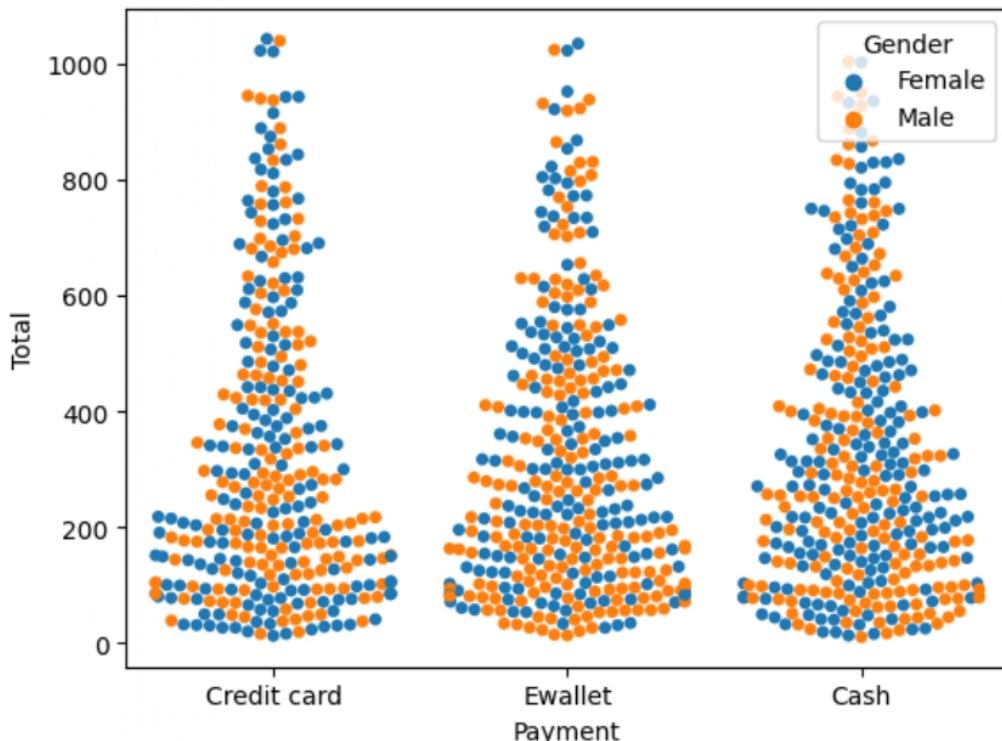


In [39]:

```
sns.swarmplot(data = mart,y= "Total",x = "Payment",hue = "Gender")
```

Out[39]:

```
<AxesSubplot: xlabel='Payment', ylabel='Total'>
```



3.showing hues separately on categorical axis

In []:

DISTRIBUTION

G) Histogram plot

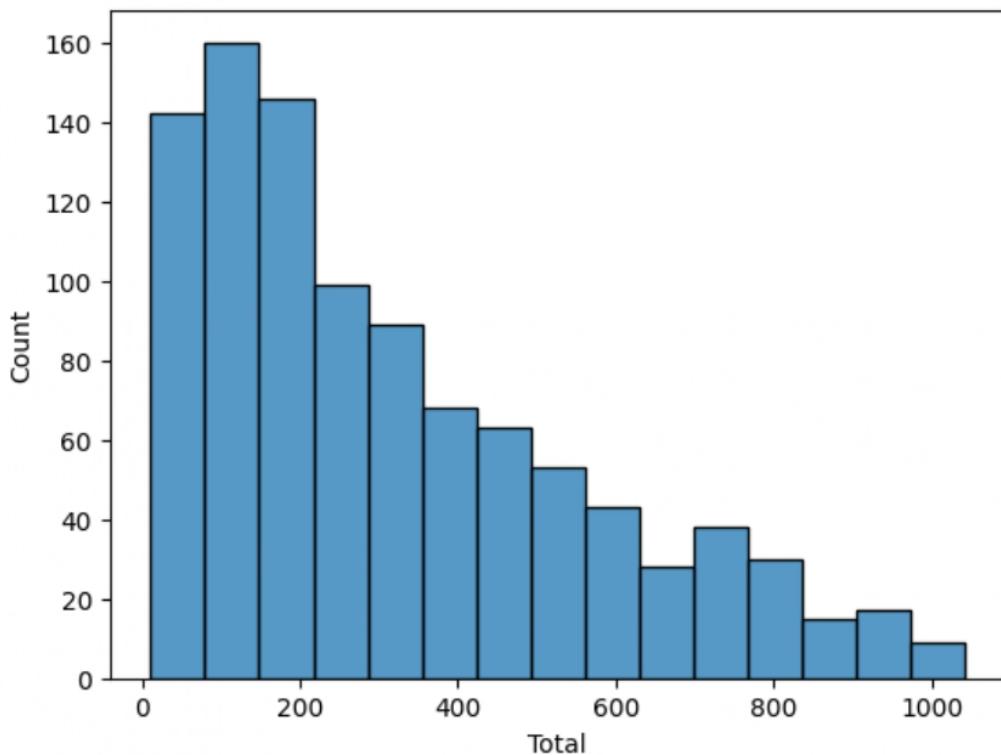
1.create a basic histogram plot

In [40]:

```
sns.histplot(data = mart,x = "Total")
```

Out[40]:

```
<AxesSubplot: xlabel='Total', ylabel='Count'>
```



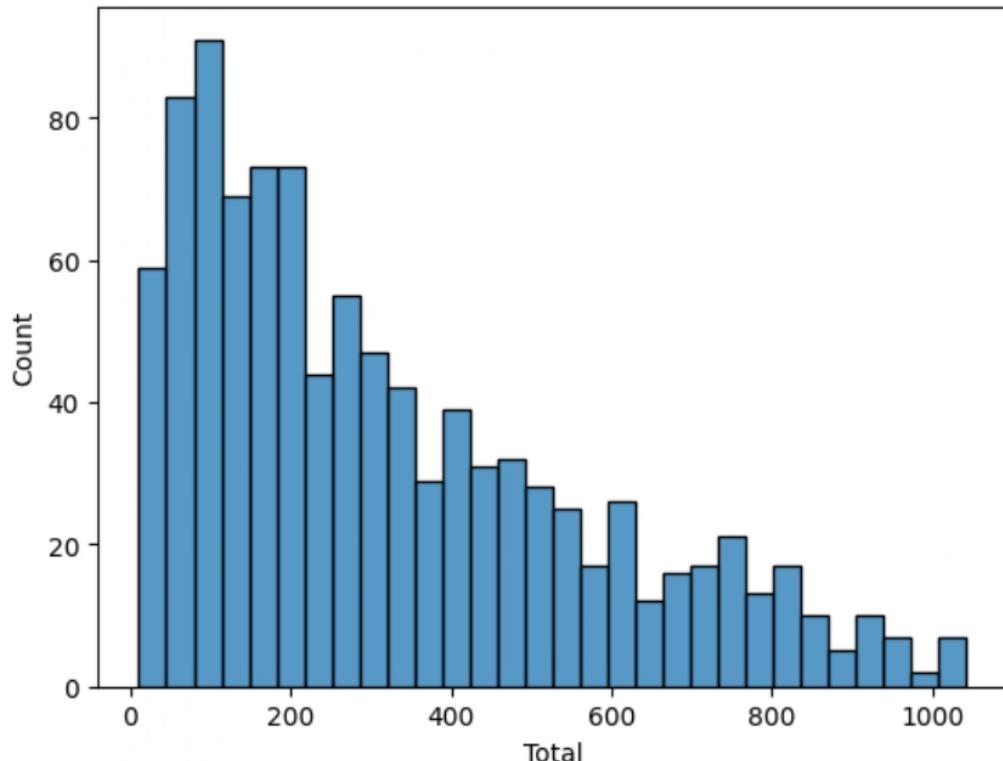
2.change number of bins and interval

In [41]:

```
sns.histplot(data = mart,x = "Total",bins = 30)
```

Out[41]:

```
<AxesSubplot: xlabel='Total', ylabel='Count'>
```



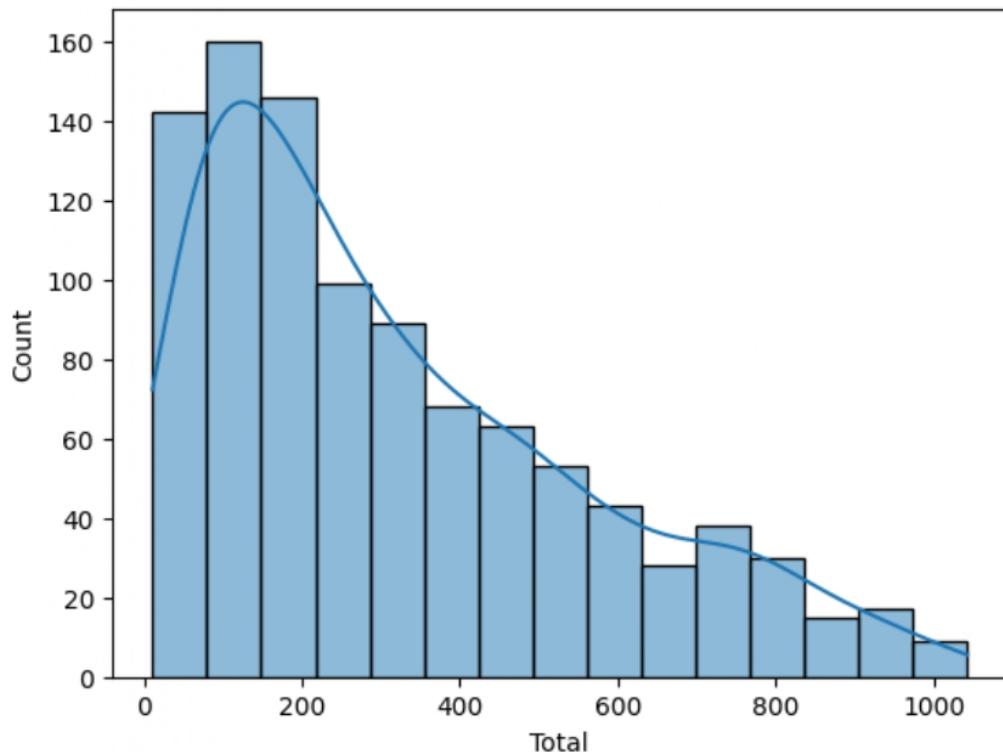
3.combine with KDE

In [42]:

```
sns.histplot(data = mart,x = "Total",kde = True)
```

Out[42]:

```
<AxesSubplot: xlabel='Total', ylabel='Count'>
```



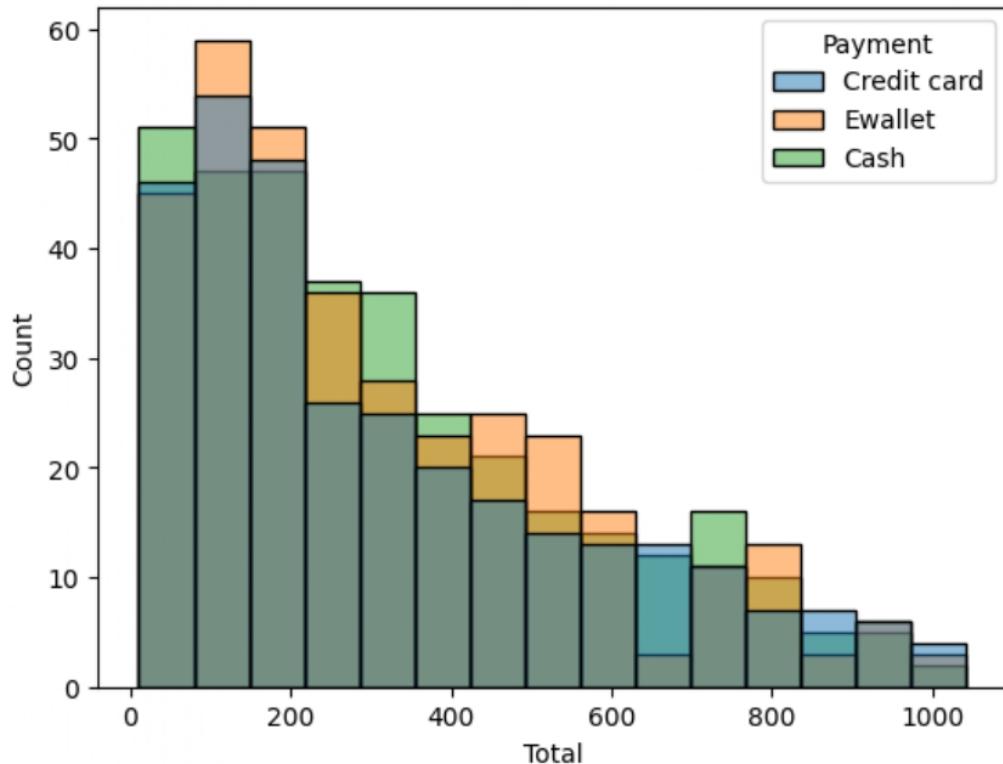
4.use a catagorical variable in HUE and STICK it MULTIPLE argument

In [43]:

```
sns.histplot(data = mart,x = "Total",hue = "Payment")
```

Out[43]:

```
<AxesSubplot: xlabel='Total', ylabel='Count'>
```

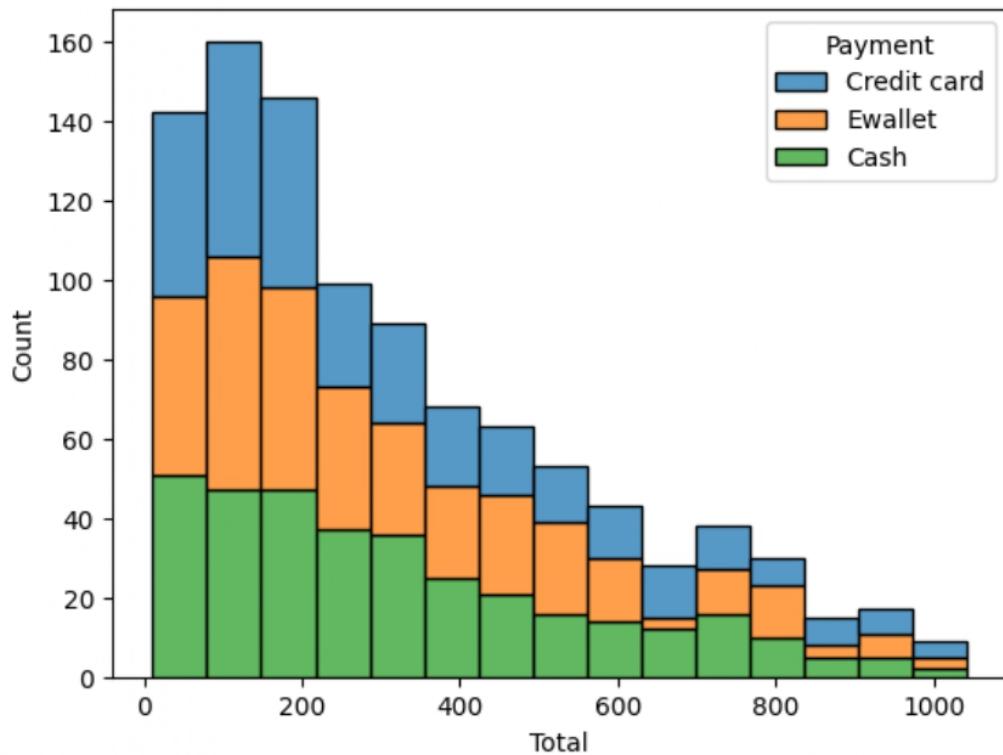


In [44]:

```
sns.histplot(data = mart,x = "Total",hue = "Payment",multiple = "stack")
```

Out[44]:

```
<AxesSubplot: xlabel='Total', ylabel='Count'>
```

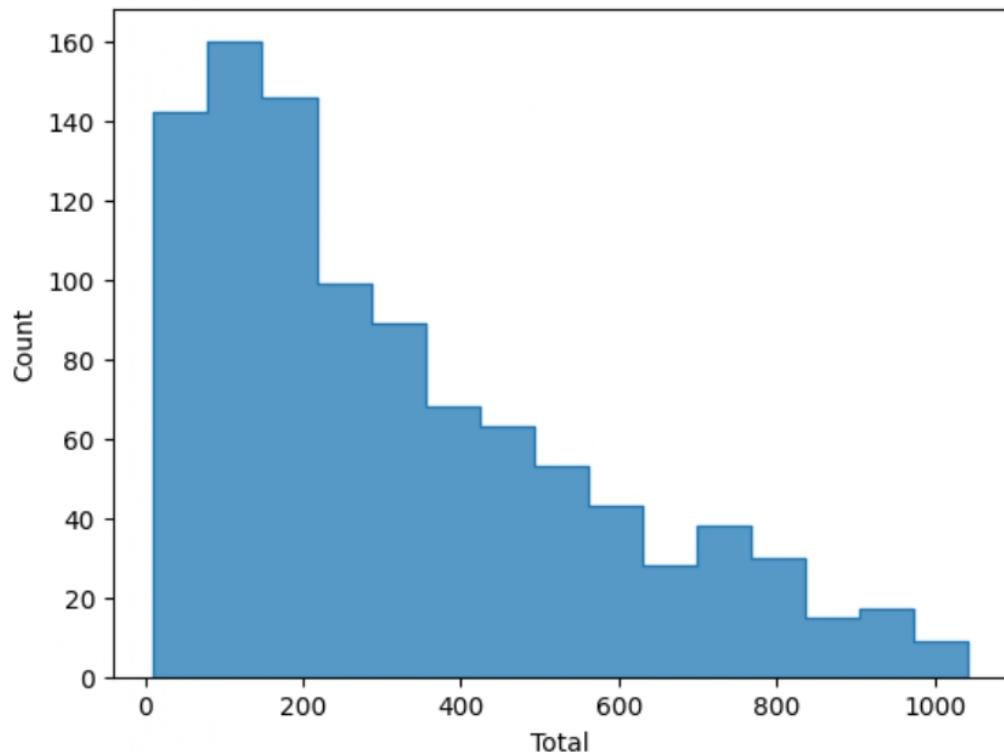


In [45]:

```
sns.histplot(data = mart,x = "Total",element = "step")
```

Out[45]:

```
<AxesSubplot: xlabel='Total', ylabel='Count'>
```

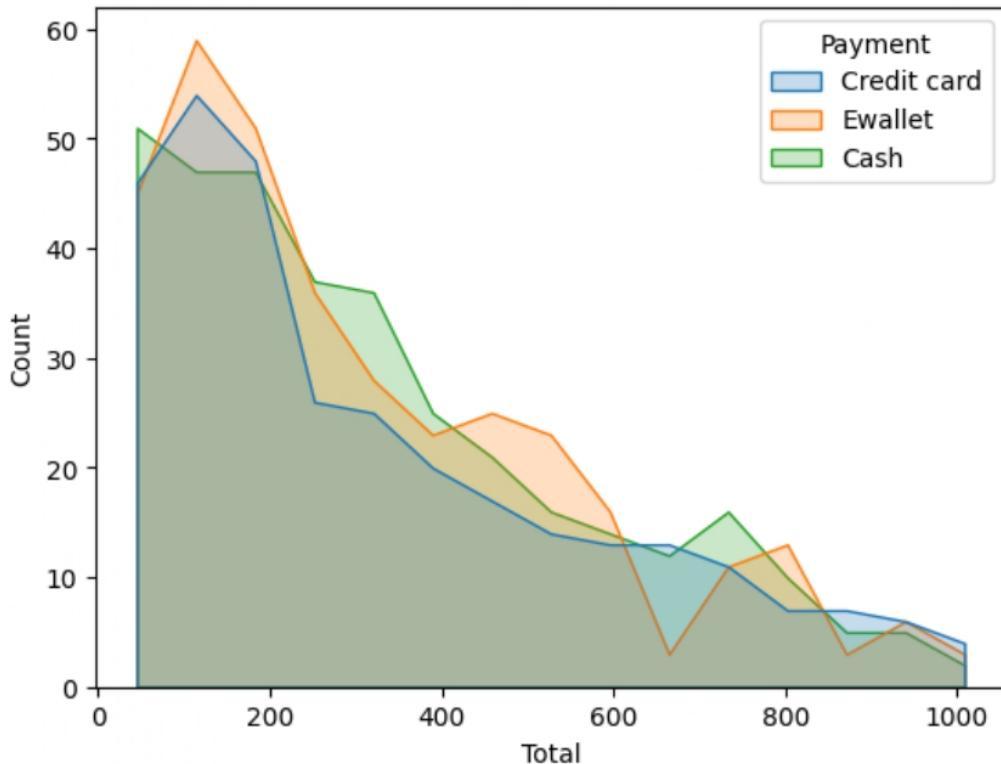


In [46]:

```
sns.histplot(data = mart,x = "Total",hue = "Payment",element = "poly")
```

Out[46]:

```
<AxesSubplot: xlabel='Total', ylabel='Count'>
```



H) KDEplot

In [47]:

```
# KDE means Kernel Density Estimation plot
```

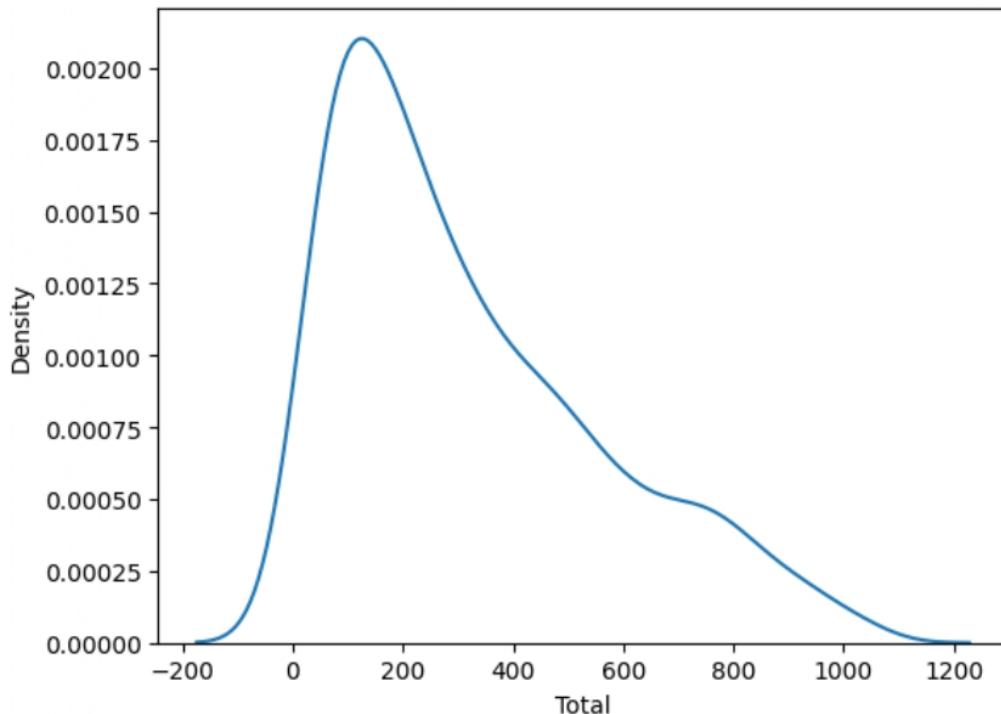
1.create a KDE plot for total column

In [48]:

```
sns.kdeplot(data = mart,x = "Total")
```

Out[48]:

```
<AxesSubplot: xlabel='Total', ylabel='Density'>
```



2.create a KDE plot for all the numeric variables in dataframe

In [49]:

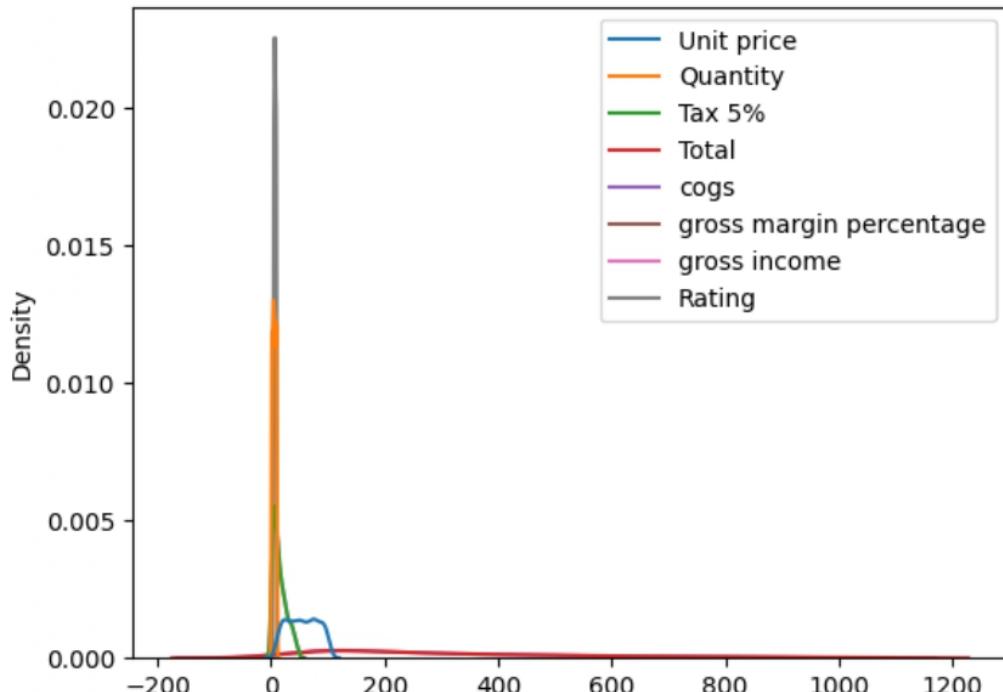
```
sns.kdeplot(data = mart)
```

C:\Users\joypa\AppData\Local\Temp\ipykernel_7164\124168617
3.py:1: UserWarning: Dataset has 0 variance; skipping density estimate. Pass `warn_singular=False` to disable this warning.

```
sns.kdeplot(data = mart)
```

Out[49]:

```
<AxesSubplot: ylabel='Density'>
```



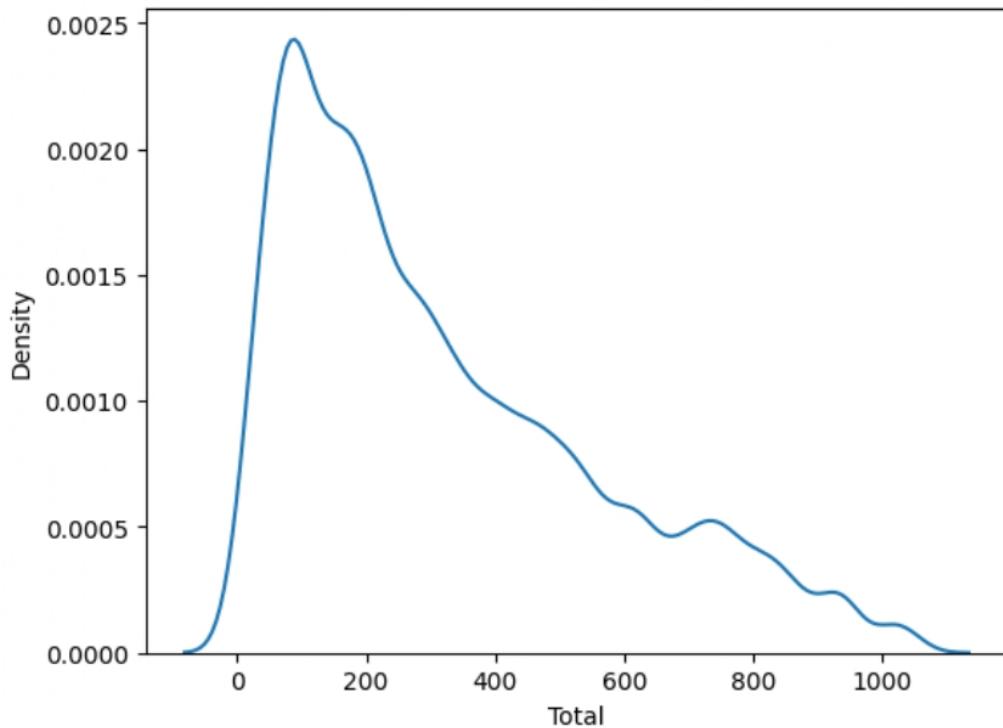
3.adjust the smoothing using bw_adjust

In [50]:

```
sns.kdeplot(data = mart,x = "Total",bw_adjust = 0.5)
```

Out[50]:

```
<AxesSubplot: xlabel='Total', ylabel='Density'>
```



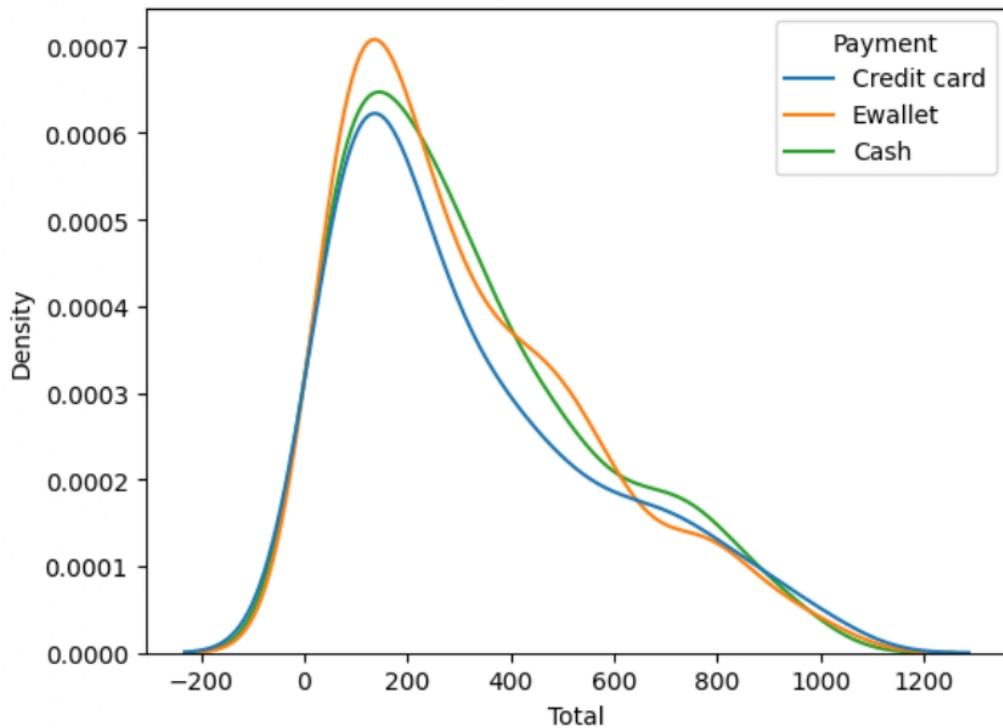
4.group the KDE on categorical variable

In [51]:

```
sns.kdeplot(data = mart,x = "Total",hue = "Payment")
```

Out[51]:

```
<AxesSubplot: xlabel='Total', ylabel='Density'>
```



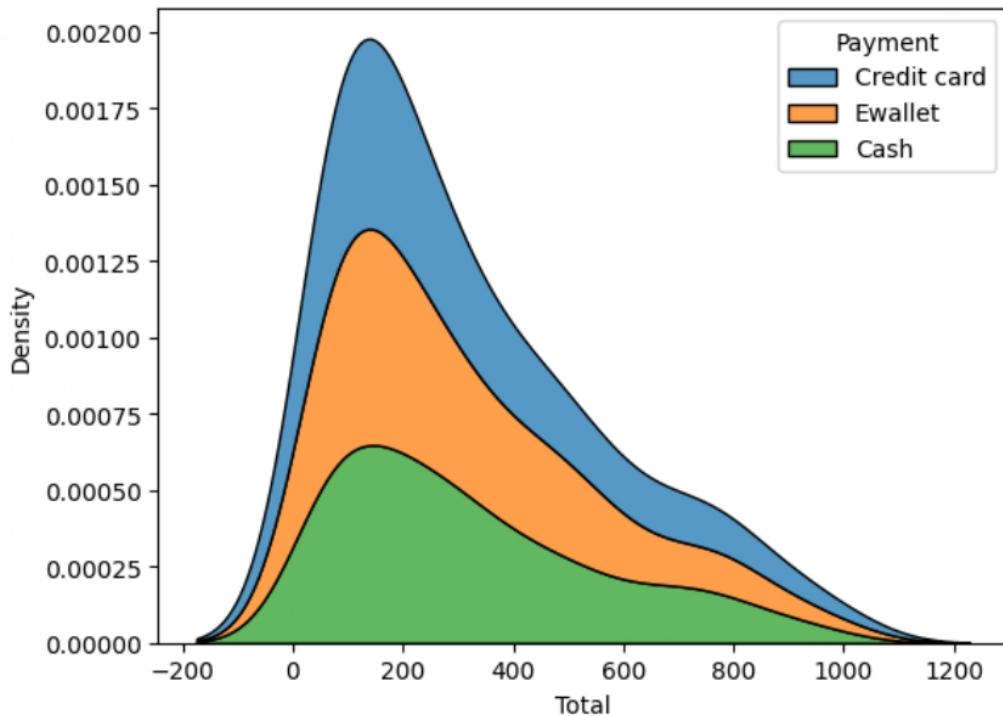
5.stack KDE on a categorical using MULTIPLE argument

In [52]:

```
sns.kdeplot(data = mart,x = "Total",hue = "Payment",multiple = "stack")
```

Out[52]:

```
<AxesSubplot: xlabel='Total', ylabel='Density'>
```



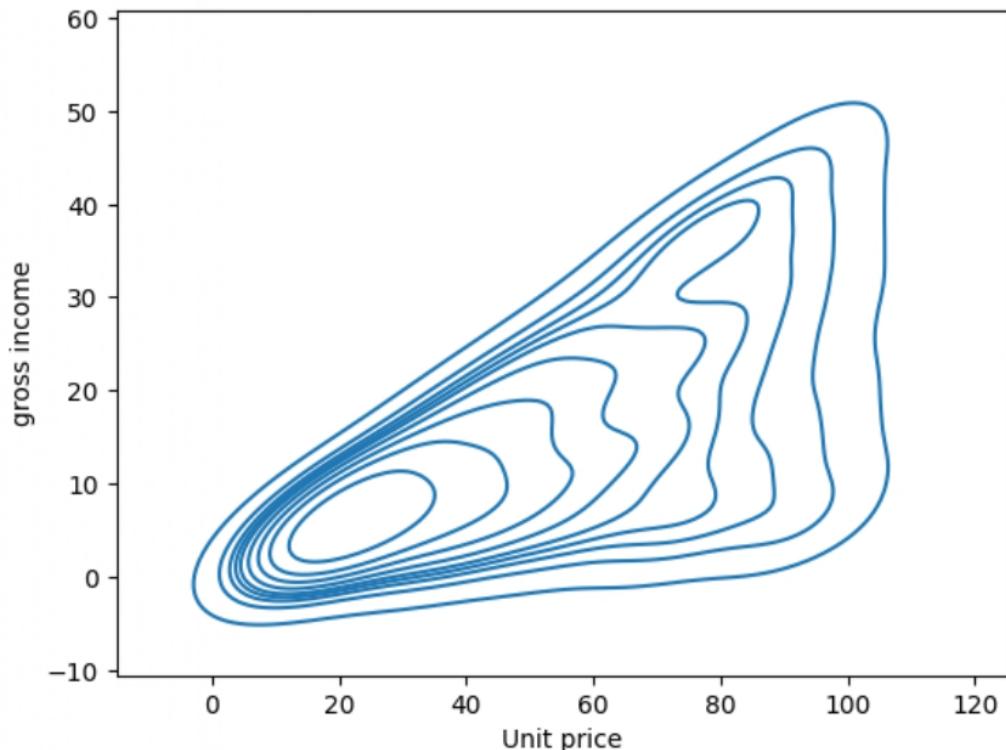
6.create a bivariate KDE

In [53]:

```
sns.kdeplot(data = mart,x = "Unit price",y = "gross income")
```

Out[53]:

```
<AxesSubplot: xlabel='Unit price', ylabel='gross income'>
```



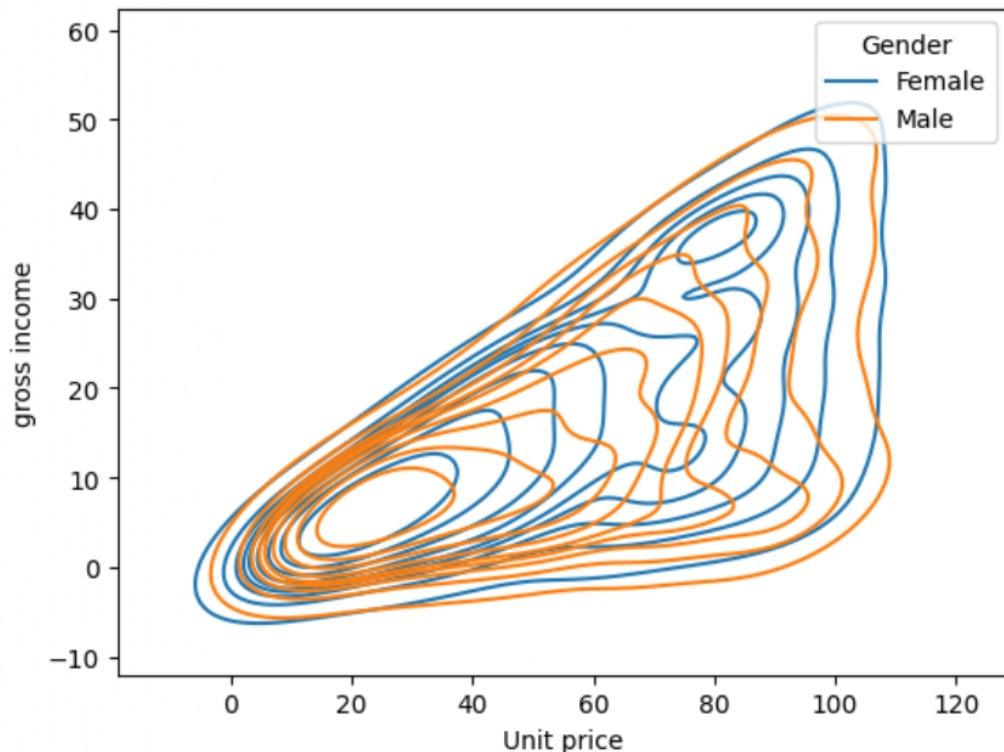
In []:

In [54]:

```
sns.kdeplot(data = mart,x = "Unit price",y = "gross income",hue = "Gender")
```

Out[54]:

```
<AxesSubplot: xlabel='Unit price', ylabel='gross income'>
```

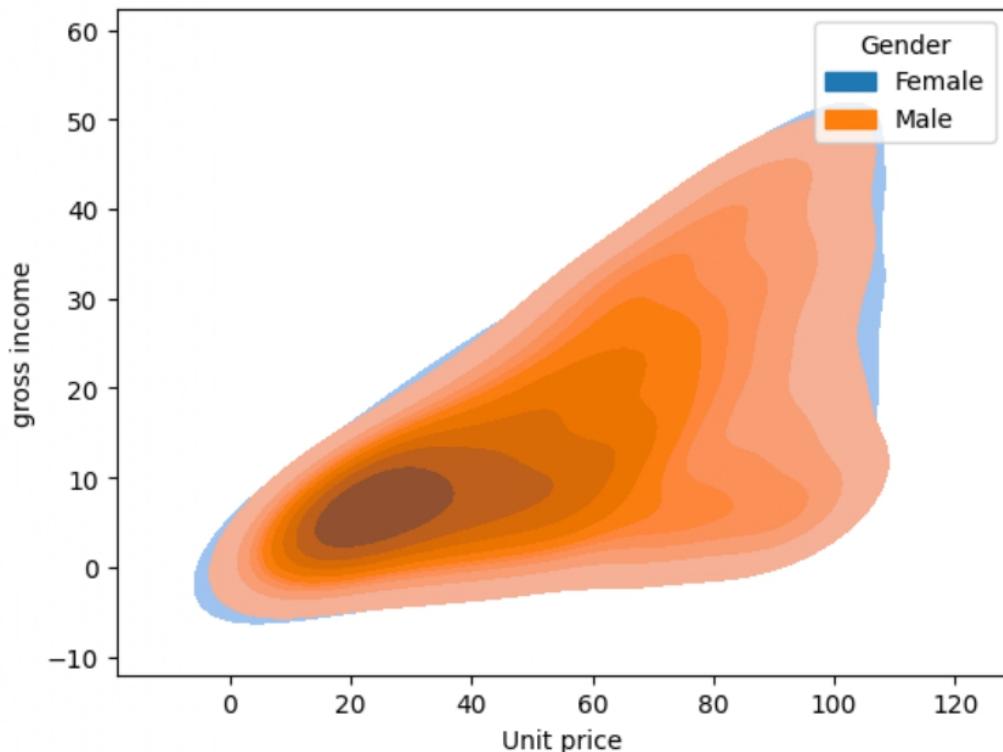


In [55]:

```
sns.kdeplot(data = mart,x = "Unit price",y = "gross income",hue = "Gender")
```

Out[55]:

```
<AxesSubplot: xlabel='Unit price', ylabel='gross income'>
```



I)RUG plot

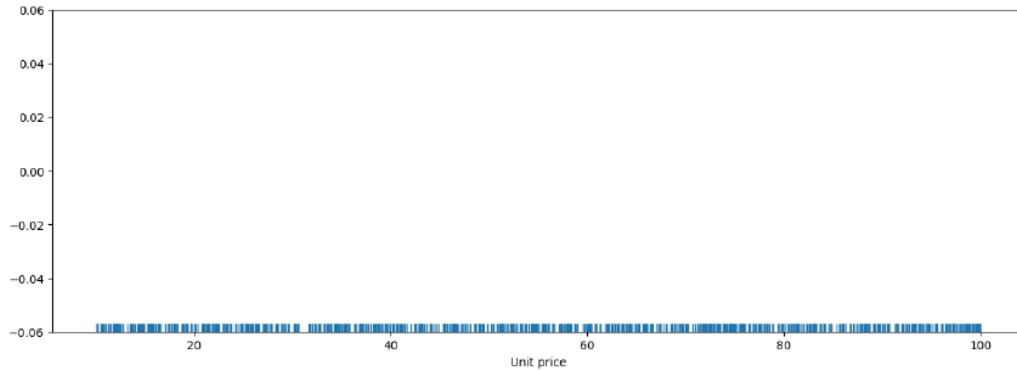
1.create a basic rug plot for one variable

In [56]:

```
plt.figure(figsize= (15,5))
sns.rugplot(data = mart,x = "Unit price")
```

Out[56]:

```
<AxesSubplot: xlabel='Unit price'>
```



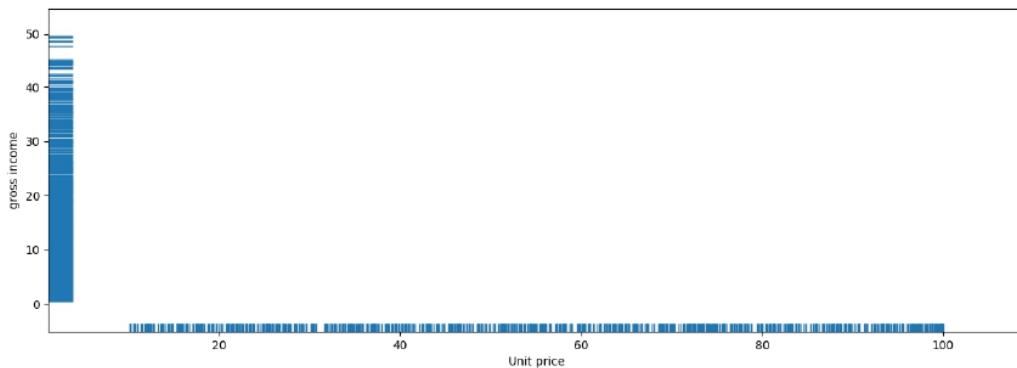
2. create a rug plot for two variable

In [57]:

```
plt.figure(figsize= (15,5))
sns.rugplot(data = mart,x = "Unit price",y = "gross income")
```

Out[57]:

```
<AxesSubplot: xlabel='Unit price', ylabel='gross income'>
```



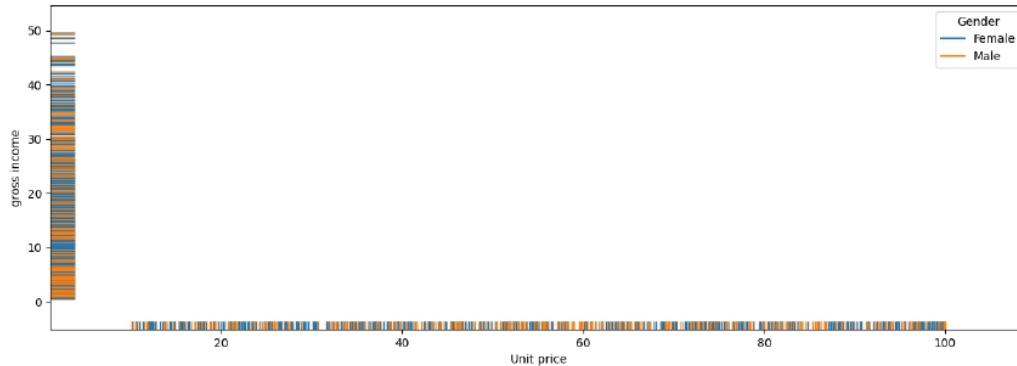
3. group it by a catagorical variable using HUE

In [58]:

```
plt.figure(figsize= (15,5))
sns.rugplot(data = mart,x = "Unit price",y = "gross income",hue = "Gender")
```

Out[58]:

```
<AxesSubplot: xlabel='Unit price', ylabel='gross income'>
```



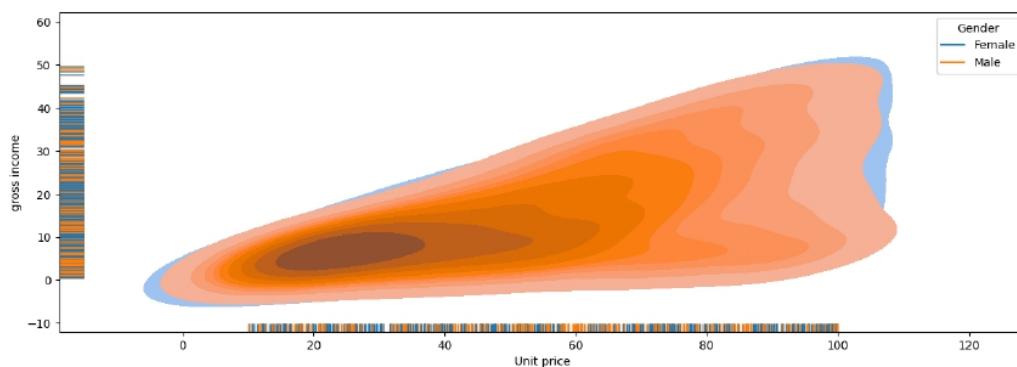
4. combine this with a KDE plot

In [59]:

```
plt.figure(figsize= (15,5))
sns.kdeplot(data = mart,x = "Unit price",y = "gross income",hue = "Gender")
sns.rugplot(data = mart,x = "Unit price",y = "gross income",hue = "Gender")
```

Out[59]:

```
<AxesSubplot: xlabel='Unit price', ylabel='gross income'>
```



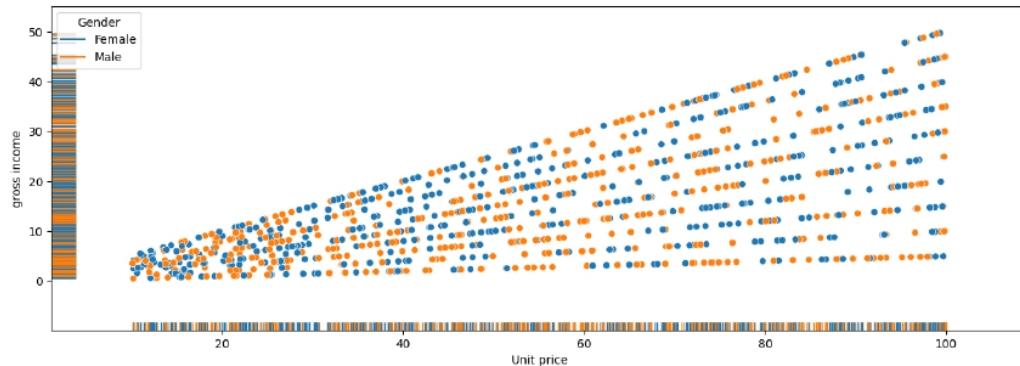
5. combine with scatter plot

In [60]:

```
plt.figure(figsize= (15,5))
sns.scatterplot(data = mart,x = "Unit price",y = "gross income",hue = "Gender")
sns.rugplot(data = mart,x = "Unit price",y = "gross income",hue = "Gender")
```

Out[60]:

<AxesSubplot: xlabel='Unit price', ylabel='gross income'>



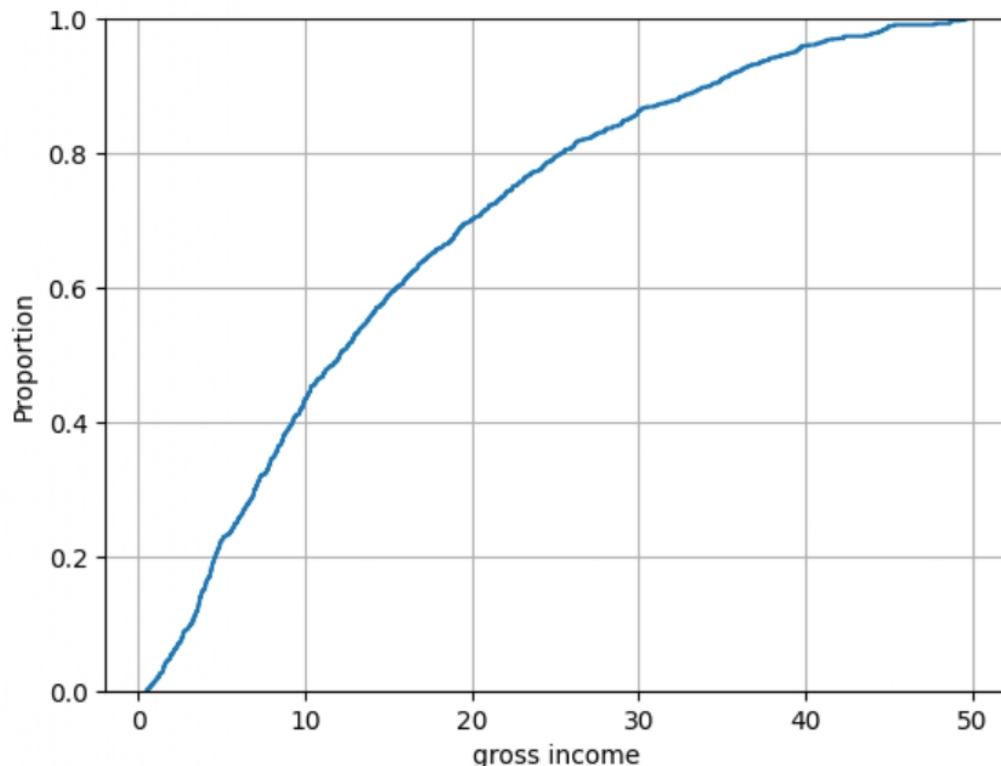
J) ECDF plot

In [61]:

```
# ECDF = Empirical Cumulative Distribution Function
```

In [62]:

```
sns.ecdfplot(data = mart,x = "gross income")
plt.grid()
```

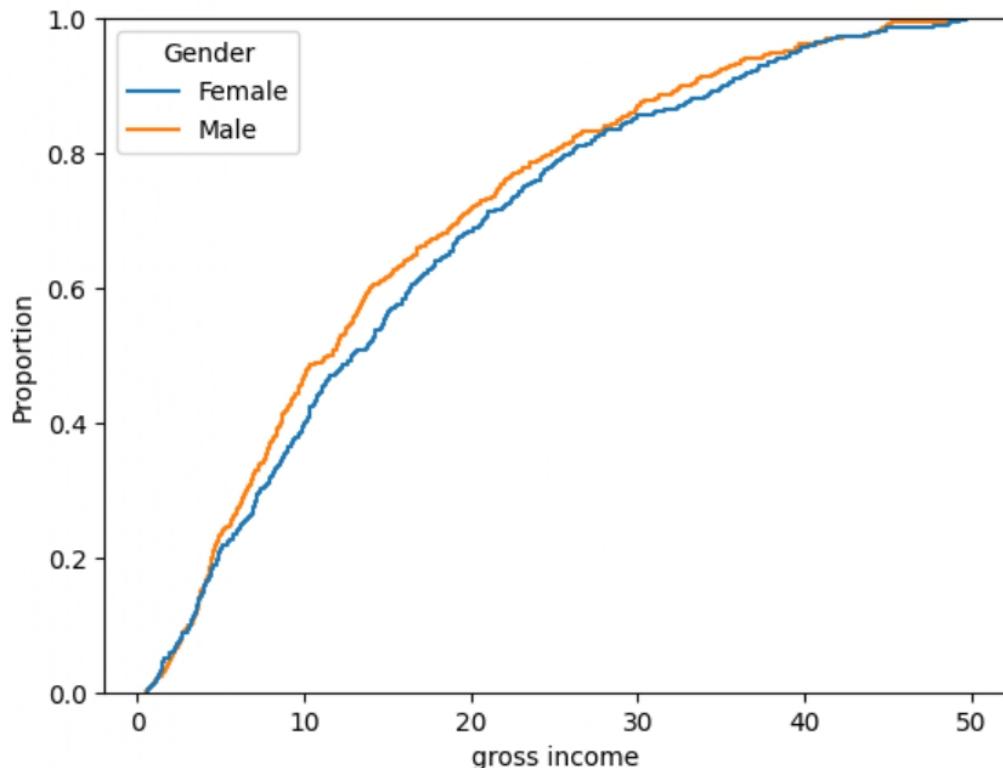


In [63]:

```
sns.ecdfplot(data = mart,x = "gross income",hue = "Gender",stat = "propo
```

Out[63]:

```
<AxesSubplot: xlabel='gross income', ylabel='Proportion'>
```

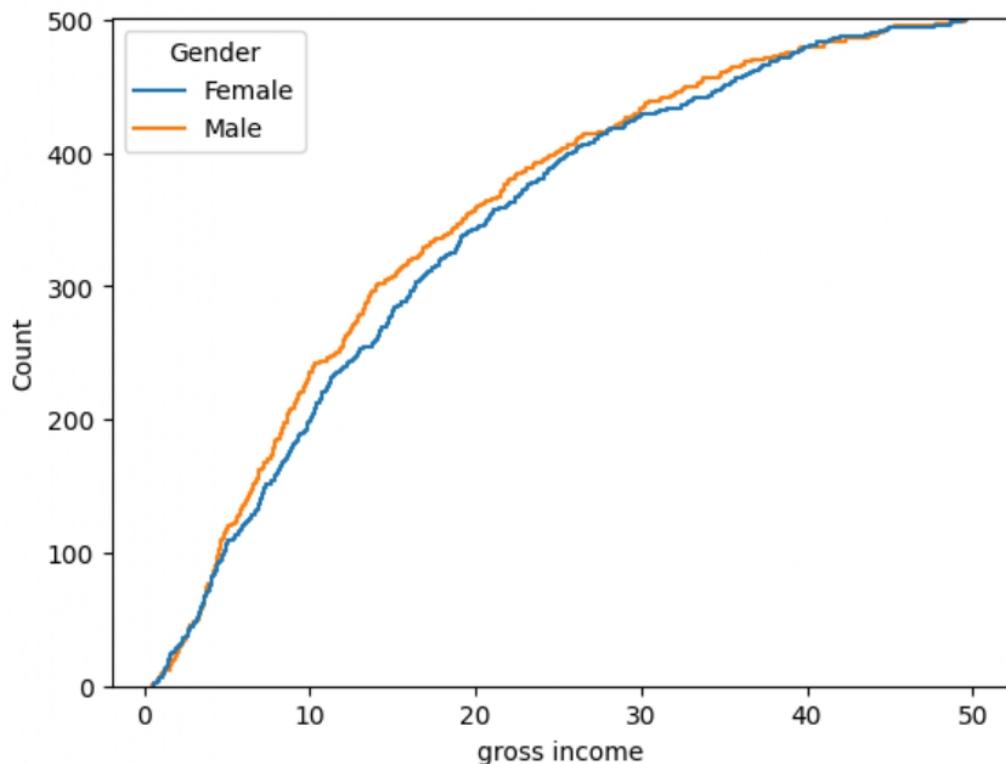


In [64]:

```
sns.ecdfplot(data = mart,x = "gross income",hue = "Gender",stat = "count")
```

Out[64]:

```
<AxesSubplot: xlabel='gross income', ylabel='Count'>
```



K) DIS plot

1. create a basic dis plot

In [65]:

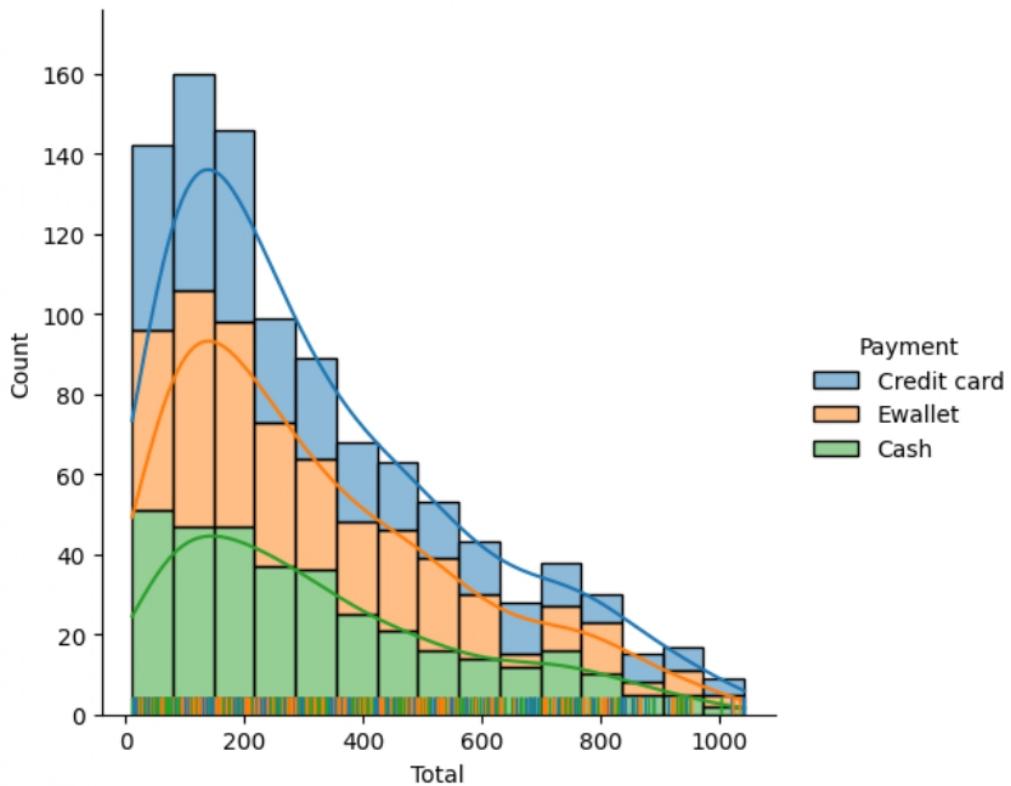
```
# DIS plot = Distribution plot
```

In [66]:

```
sns.displot(data = mart,x = "Total",kde = True,rug = True,hue = "Payment")
```

Out[66]:

```
<seaborn.axisgrid.FacetGrid at 0x2427e6f7280>
```



L) JOIN plot

In [67]:

```
irs= sns.load_dataset("iris")
```

In [68]:

```
iris.head()
```

Out[68]:

	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

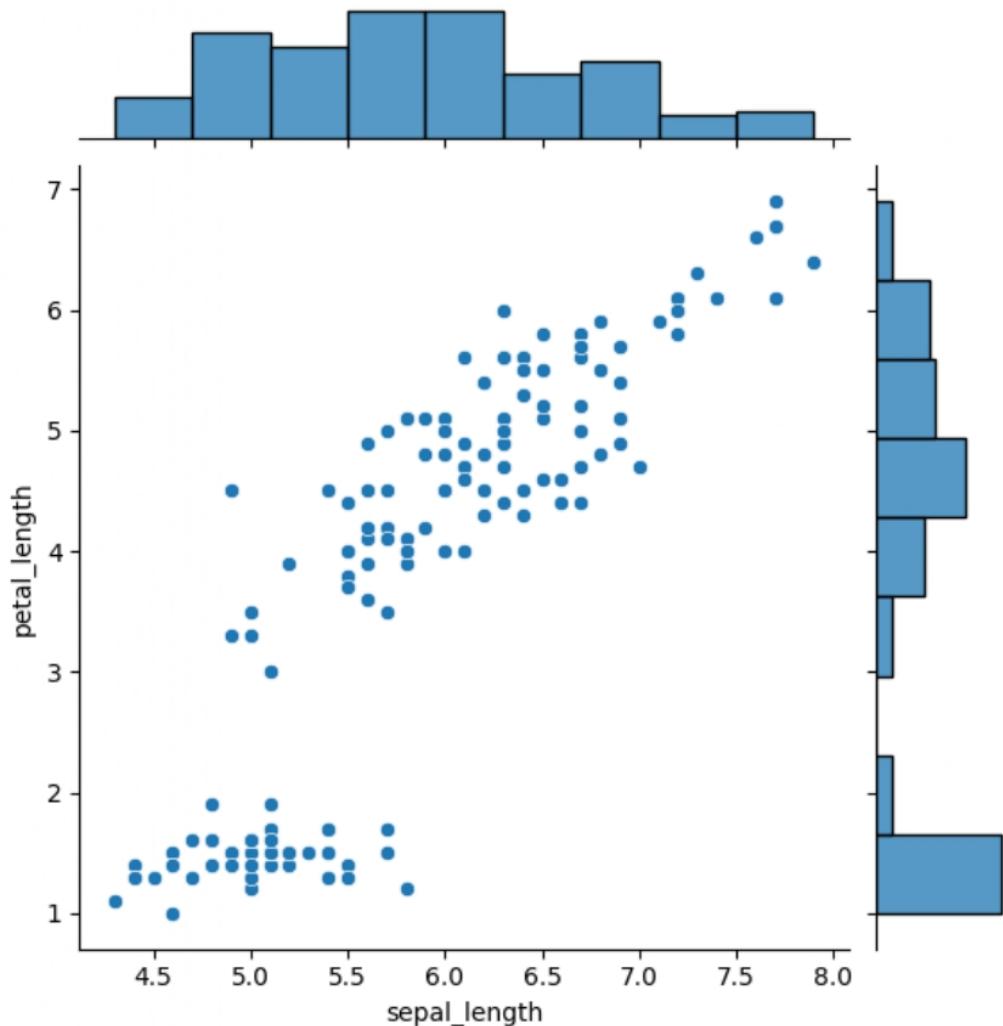
1. Draw a first basic JOIN plot

In [69]:

```
sns.jointplot(data = iris,x = "sepal_length",y = "petal_length")
```

Out[69]:

```
<seaborn.axisgrid.JointGrid at 0x2427e341090>
```

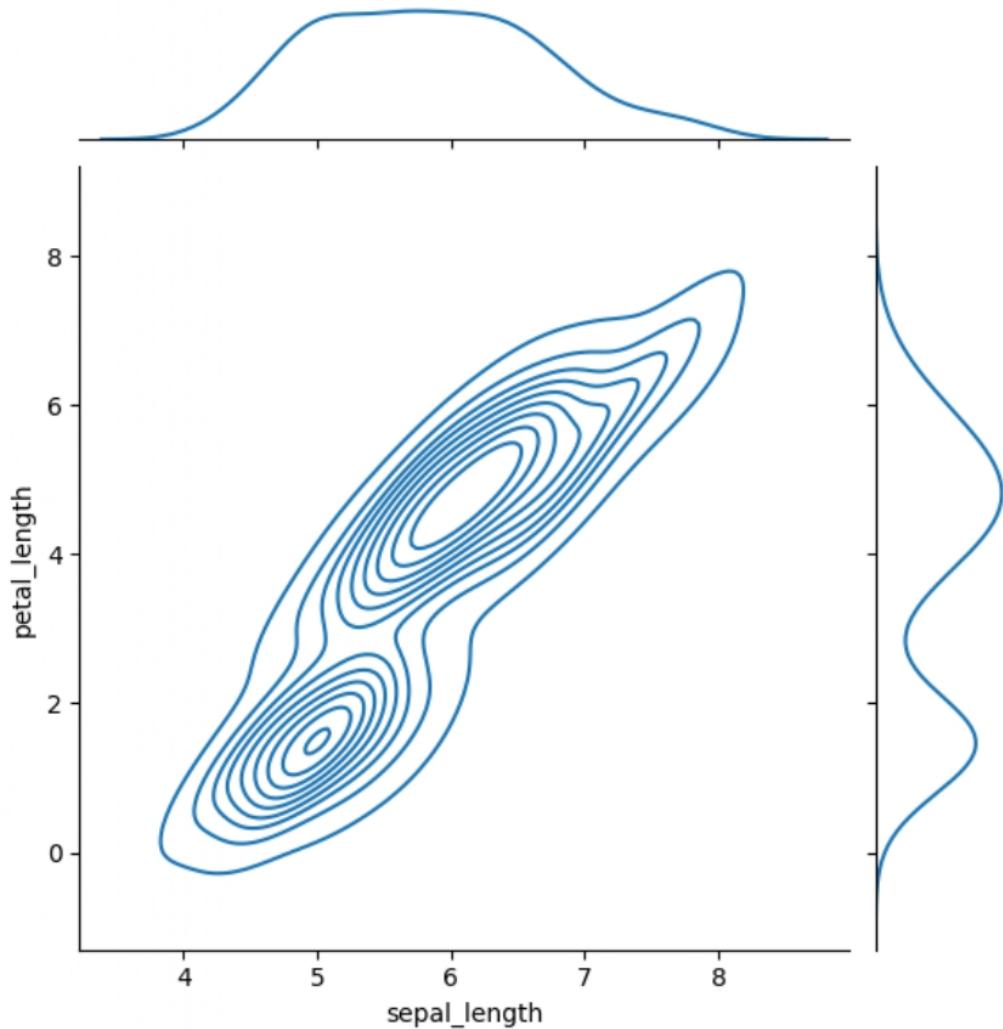


In [70]:

```
sns.jointplot(data = iris,x = "sepal_length",y = "petal_length",  
               kind = "kde")
```

Out[70]:

```
<seaborn.axisgrid.JointGrid at 0x2427e9c1ea0>
```

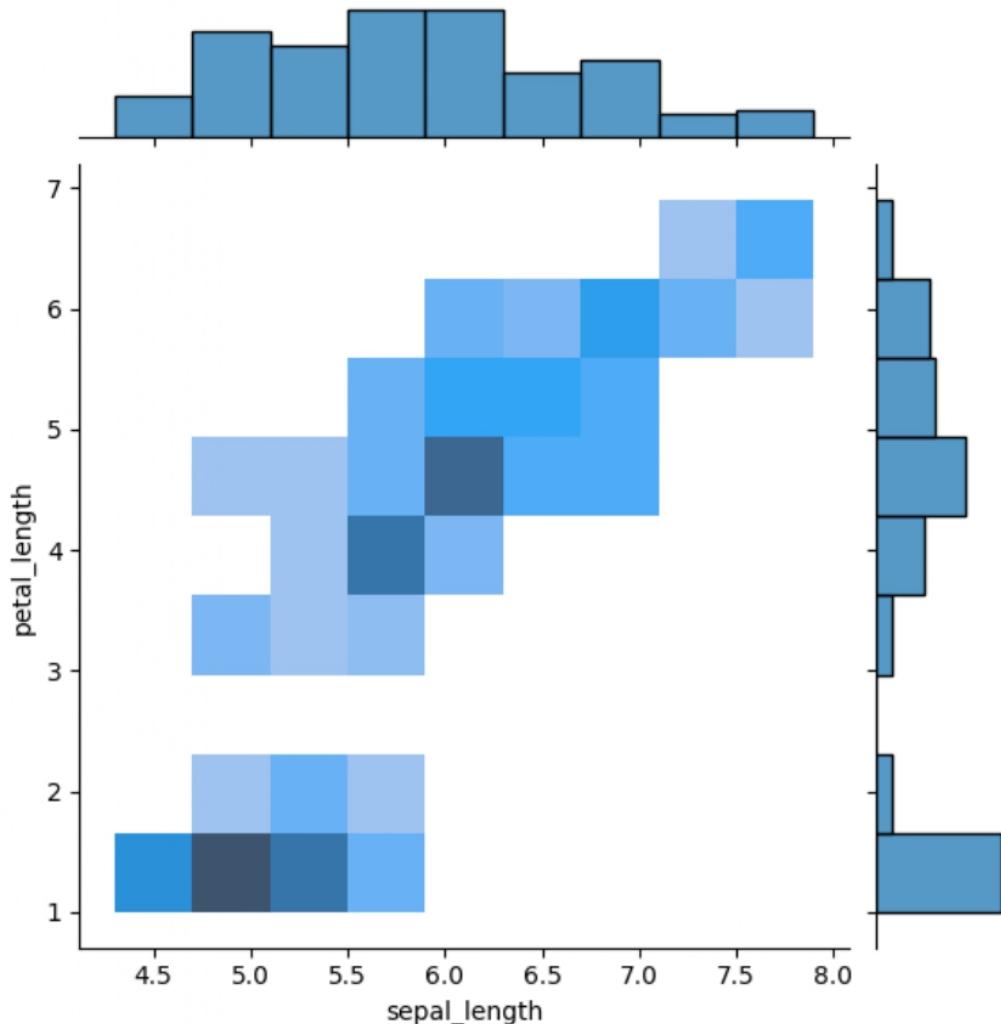


In [71]:

```
sns.jointplot(data = iris,x = "sepal_length",y = "petal_length",  
               kind = "hist")
```

Out[71]:

```
<seaborn.axisgrid.JointGrid at 0x2427bf5dc90>
```

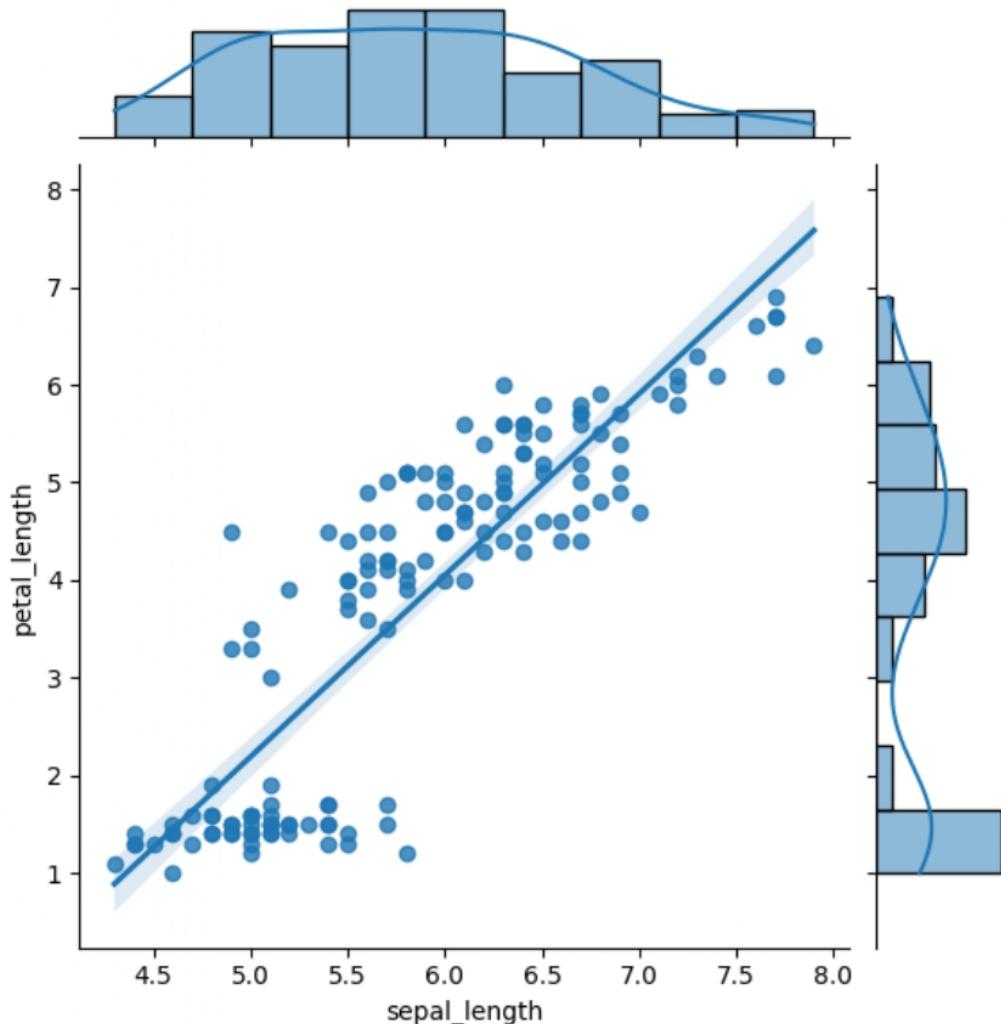


In [72]:

```
sns.jointplot(data = iris,x = "sepal_length",y = "petal_length",  
               kind = "reg")
```

Out[72]:

```
<seaborn.axisgrid.JointGrid at 0x2427c0cca00>
```

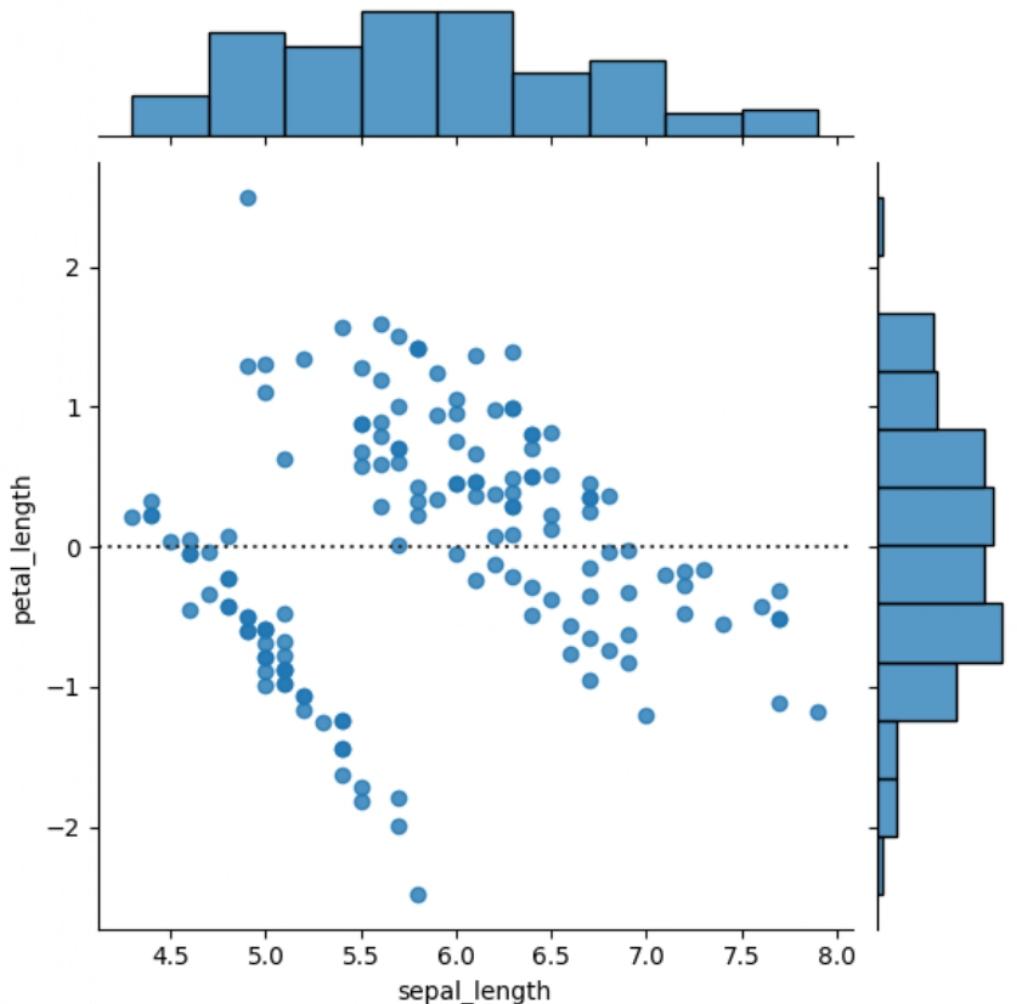


In [73]:

```
sns.jointplot(data = iris,x = "sepal_length",y = "petal_length",  
               kind = "resid")
```

Out[73]:

```
<seaborn.axisgrid.JointGrid at 0x2427f84ad10>
```



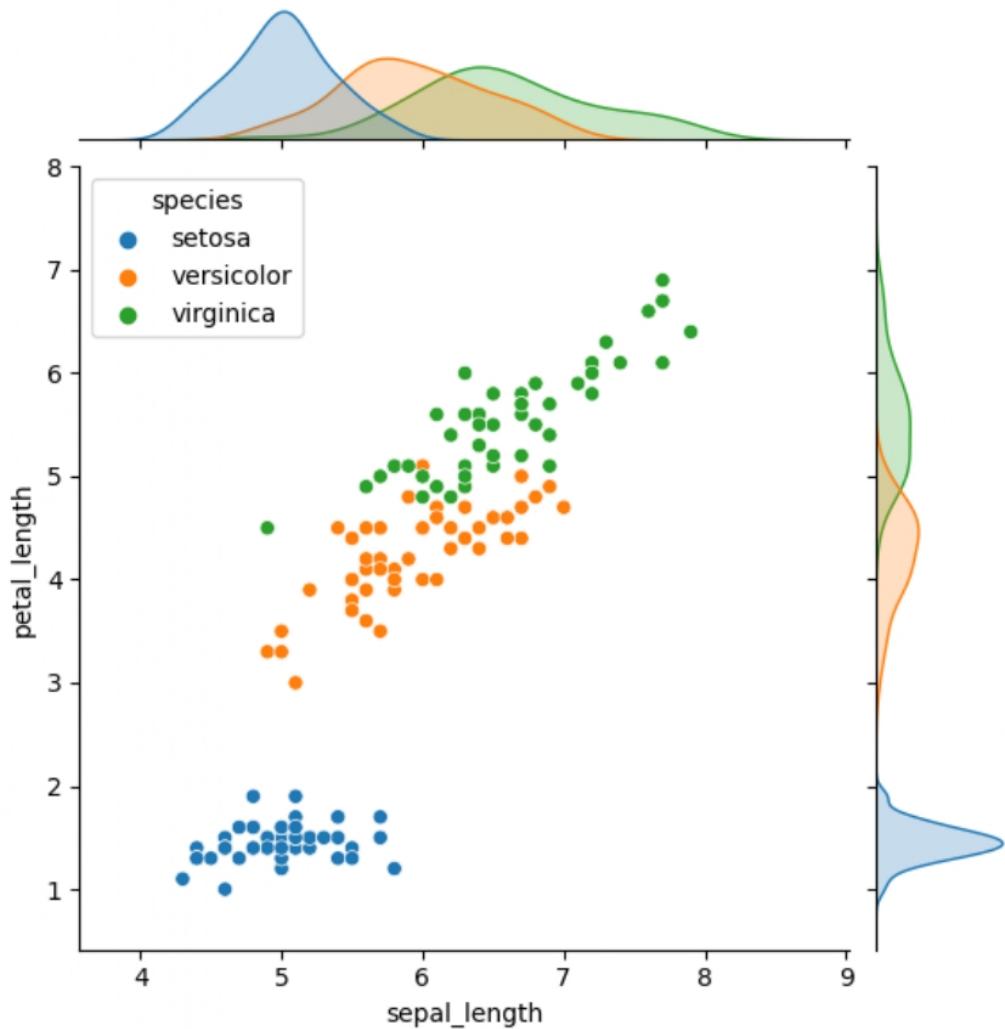
2. grouping on the based on catagorical value

In [74]:

```
sns.jointplot(data = iris,x = "sepal_length",y = "petal_length",  
               hue = "species")
```

Out[74]:

```
<seaborn.axisgrid.JointGrid at 0x2427fdec90>
```



M) PAIR plot

1. create a basic pair plot

In [75]:

```
irs.head()
```

Out[75]:

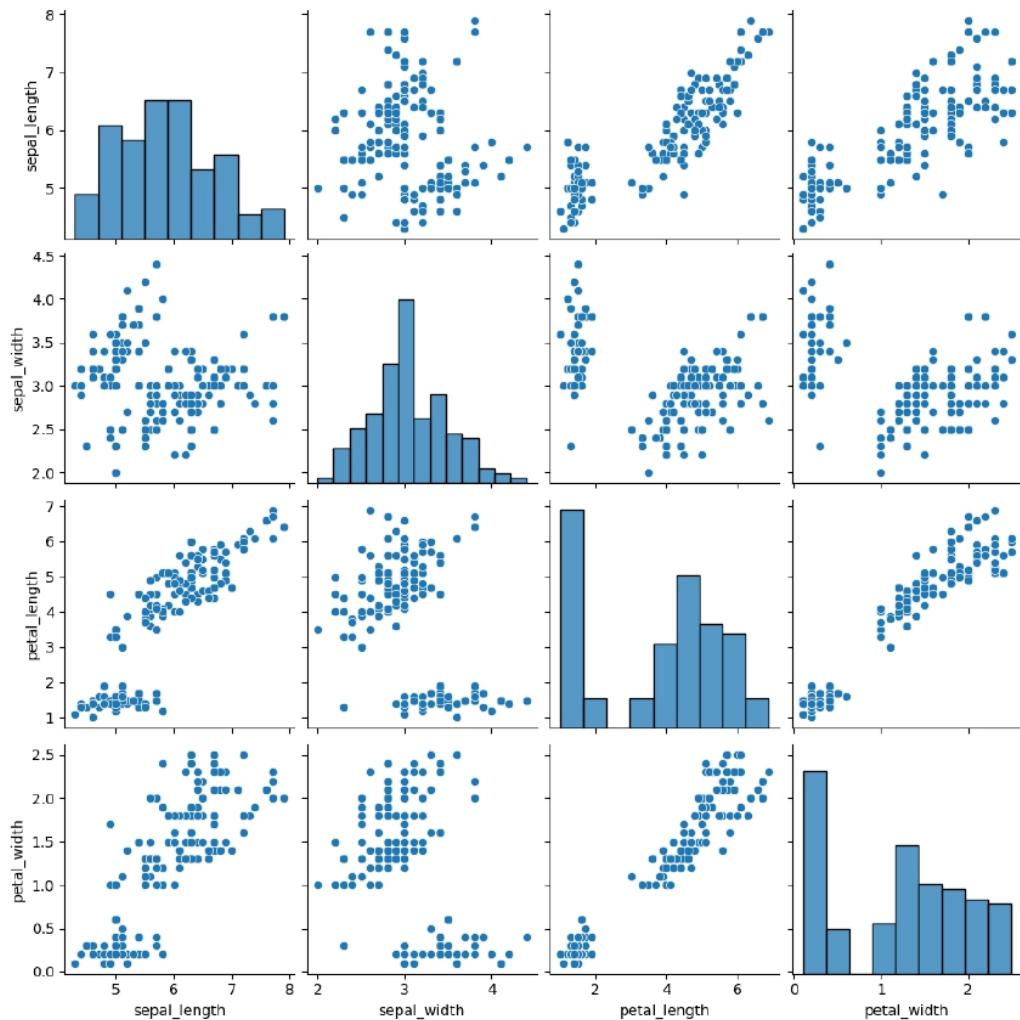
	sepal_length	sepal_width	petal_length	petal_width	species
0	5.1	3.5	1.4	0.2	setosa
1	4.9	3.0	1.4	0.2	setosa
2	4.7	3.2	1.3	0.2	setosa
3	4.6	3.1	1.5	0.2	setosa
4	5.0	3.6	1.4	0.2	setosa

In [76]:

```
sns.pairplot(data = irs)
```

Out[76]:

```
<seaborn.axisgrid.PairGrid at 0x242001f2ce0>
```



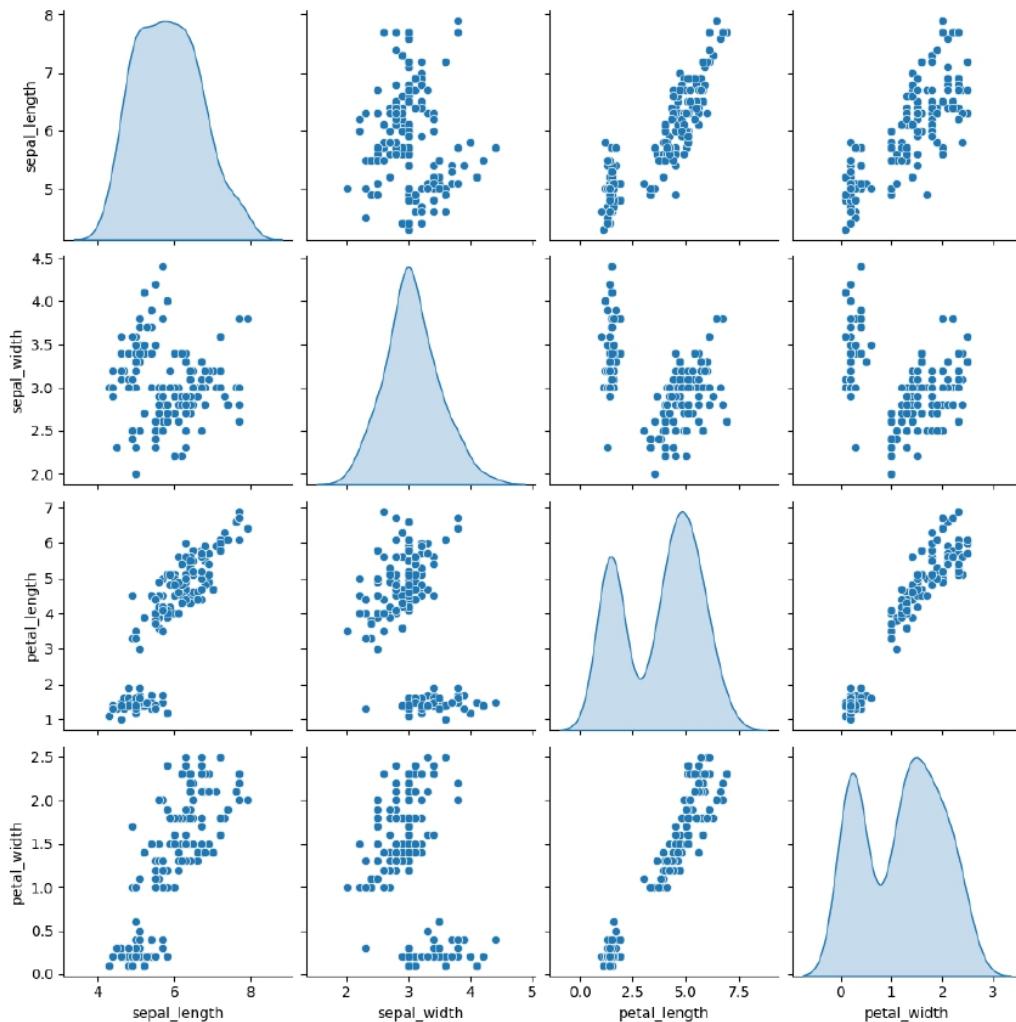
2.changing the diagonal plot KIND to KDE,HIST or None

In [77]:

```
sns.pairplot(data = irs,diag_kind = "kde")
```

Out[77]:

```
<seaborn.axisgrid.PairGrid at 0x24201fd9d50>
```



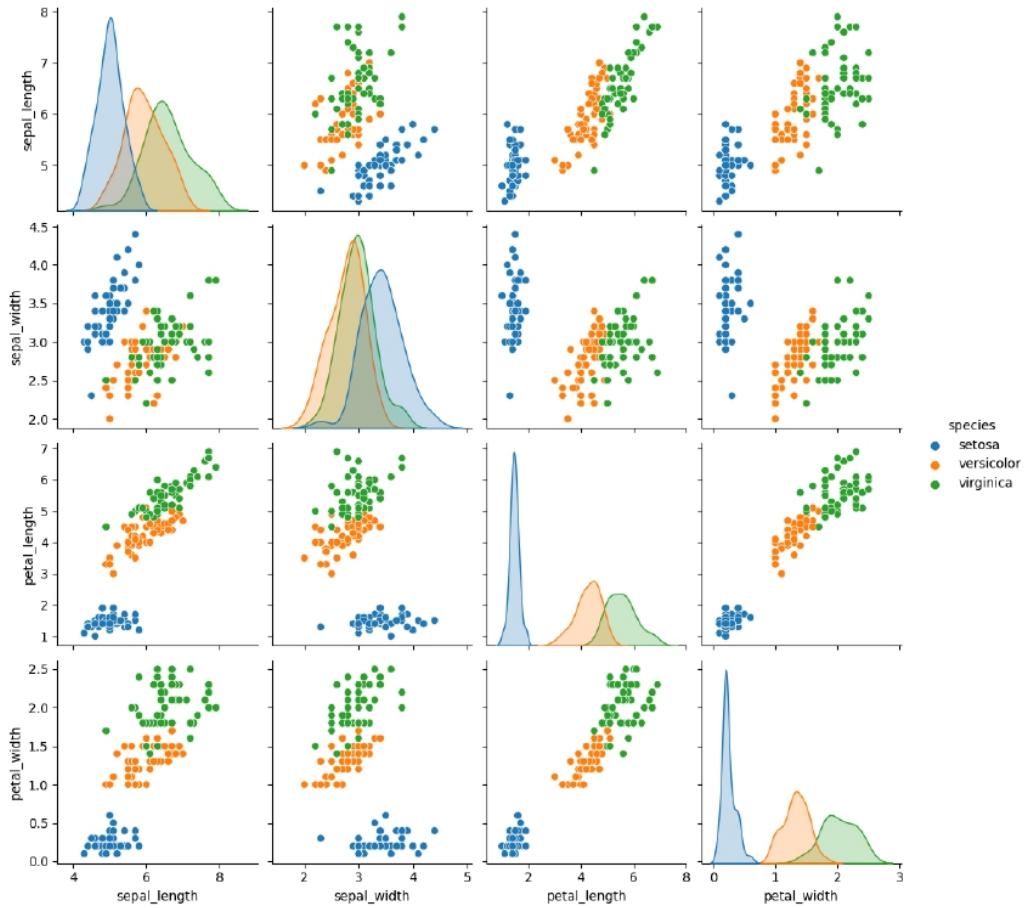
3. adding hue in the pair plot

In [78]:

```
sns.pairplot(data = irs,diag_kind = "kde",hue = 'species')
```

Out[78]:

```
<seaborn.axisgrid.PairGrid at 0x24202ec3eb0>
```



In [79]:

```
# Relational plot  
# 1. scatter plot  
# 2. line plot  
# rel plot
```

N) Scatter plot

1. create a basic seaborn plot

In [80]:

```
m = pd.read_excel("mart_train_sample.xlsx")
```

In [81]:

```
m.head()
```

Out[81]:

	Item_ID	Item_W	Item_Type	Item_MRP	Outlet_ID	Outlet_Year	Out
0	FDN27	9.437510	Snack Foods	195.313244	OUT046	1997	Outlets
1	FDW35	9.033762	Fruits and Vegetables	36.371449	OUT035	2003	2004
2	FDS28	20.884849	Frozen Foods	70.318167	OUT013	2008	2009
3	FDC09	19.976300	Snack Foods	121.723264	OUT018	2009	2010
4	FDX04	10.824812	Snack Foods	264.673302	OUT035	2004	2011

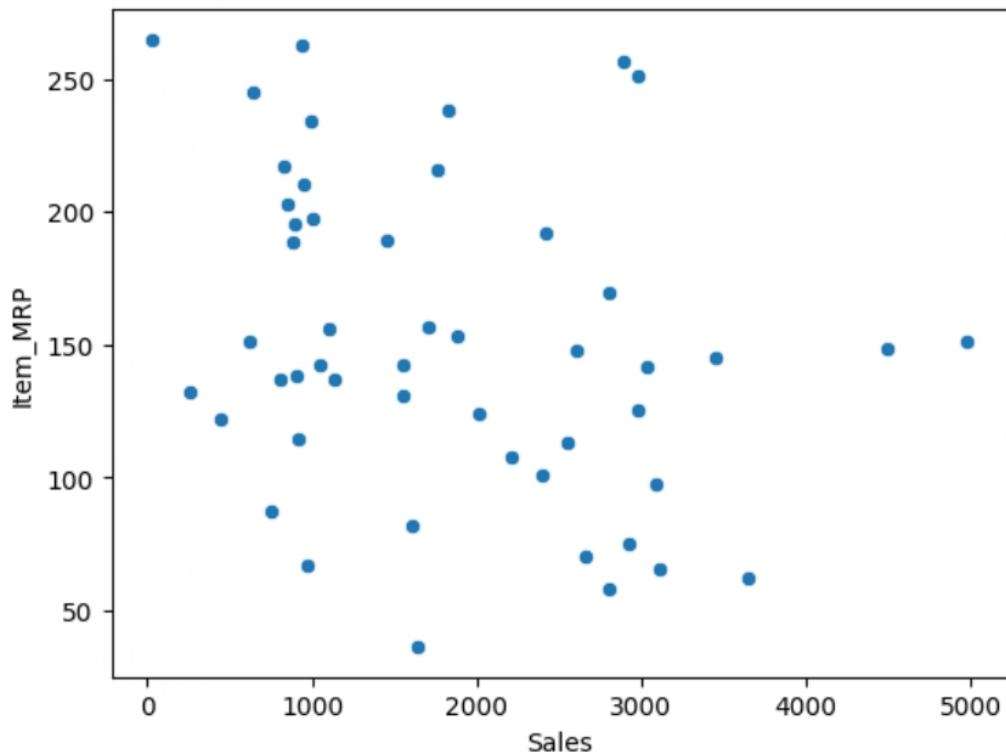
1.create a scatter plot

In [82]:

```
sns.scatterplot(data = m,x = "Sales",y = "Item_MRP")
```

Out[82]:

```
<AxesSubplot: xlabel='Sales', ylabel='Item_MRP'>
```



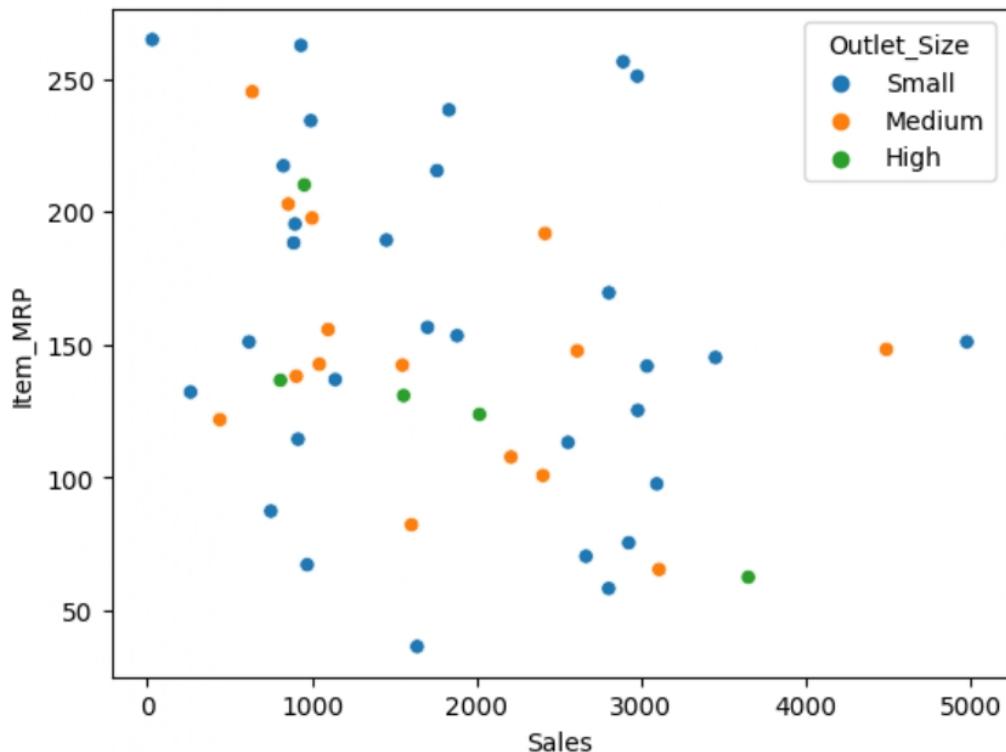
2. group basis on a catagorical variable using HUE

In [83]:

```
sns.scatterplot(data = m,x = "Sales",y = "Item_MRP",hue ="Outlet_Size" )
```

Out[83]:

```
<AxesSubplot: xlabel='Sales', ylabel='Item_MRP'>
```



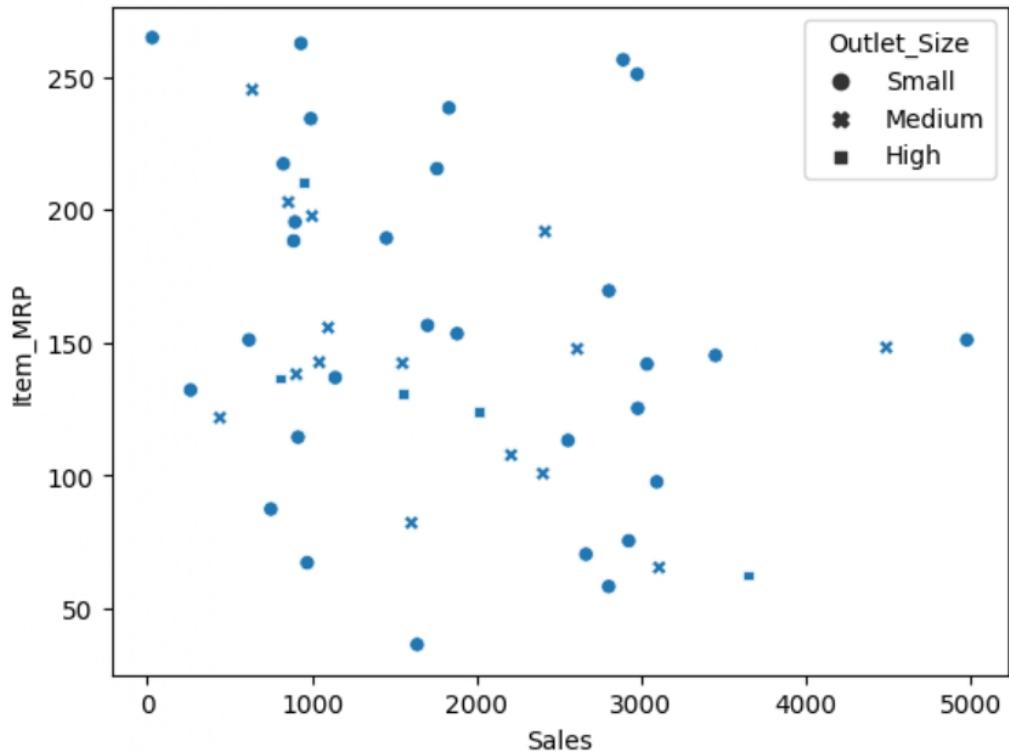
3. group basis on a catagorical variable using Style

In [84]:

```
sns.scatterplot(data = m,x = "Sales",y = "Item_MRP",style="Outlet_Size" )
```

Out[84]:

```
<AxesSubplot: xlabel='Sales', ylabel='Item_MRP'>
```



o) LINE plot

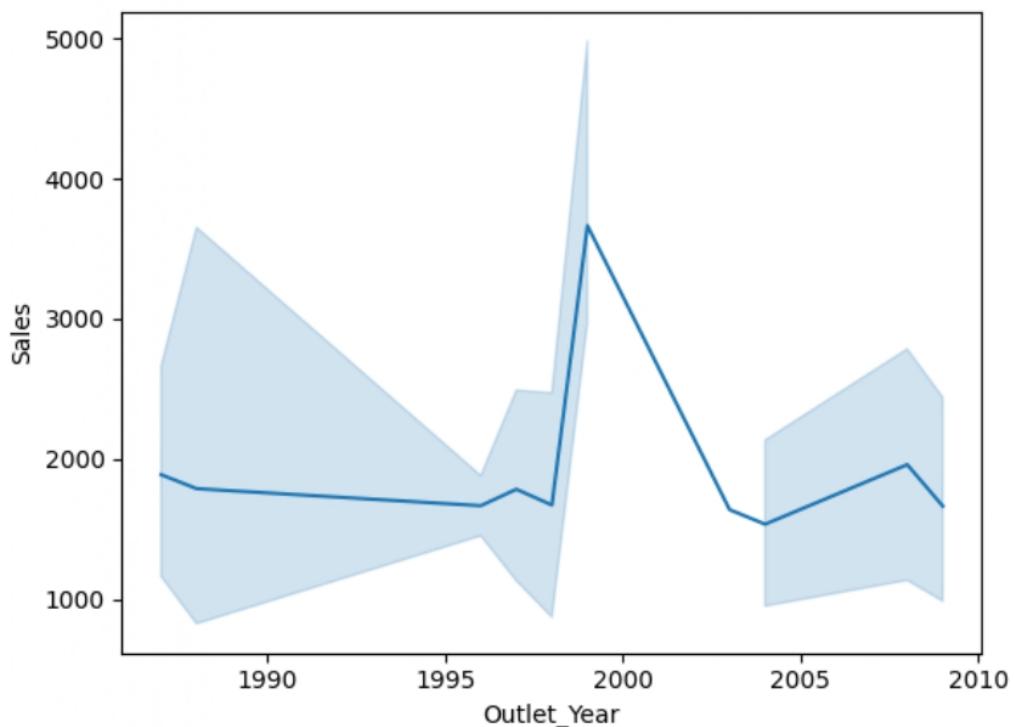
1.create a basic line plot

In [85]:

```
sns.lineplot(data = m,x = "Outlet_Year",y = "Sales")
```

Out[85]:

```
<AxesSubplot: xlabel='Outlet_Year', ylabel='Sales'>
```



In [86]:

```
sns.lineplot(data = m,x = "Outlet_Year",y = "Sales",ci = None)
```

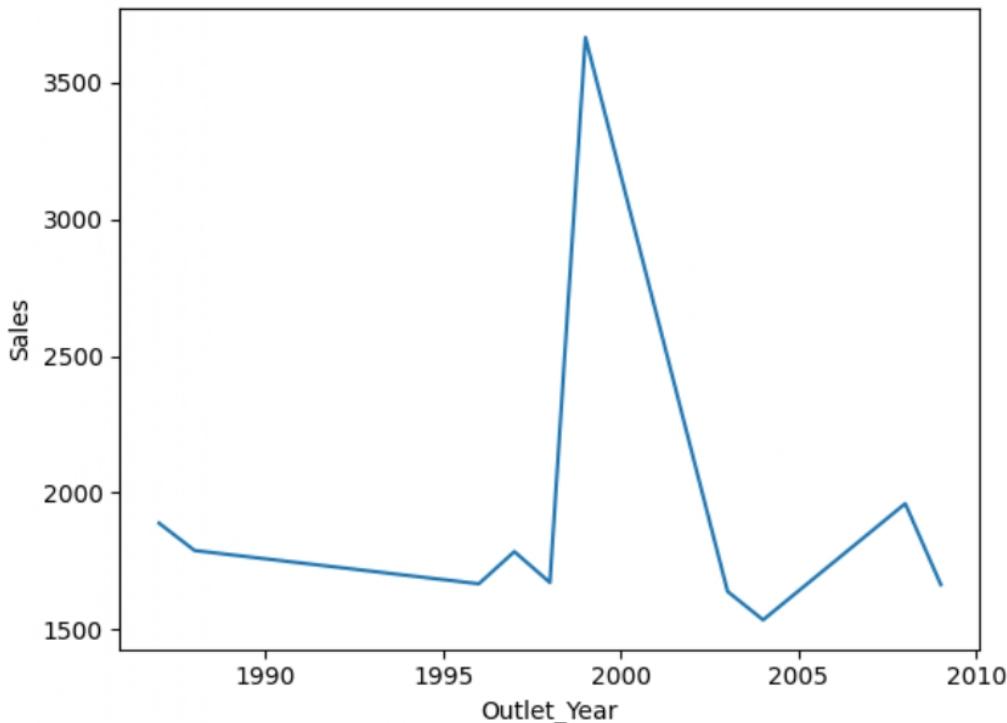
C:\Users\joypa\AppData\Local\Temp\ipykernel_7164\340264817
6.py:1: FutureWarning:

The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

```
sns.lineplot(data = m,x = "Outlet_Year",y = "Sales",ci = None)
```

Out[86]:

```
<AxesSubplot: xlabel='Outlet_Year', ylabel='Sales'>
```



In [87]:

```
m.columns
```

Out[87]:

```
Index(['Item_ID', 'Item_W', 'Item_Type', 'Item_MRP', 'Outlet_ID',
       'Outlet_Year', 'Outlet_Size', 'Outlet_Location_Type', 'Sales', 'Tier'],
      dtype='object')
```

In [88]:

```
sns.lineplot(data = m,x = "Outlet_Year",y = "Sales",ci = None,hue = "Outl
```

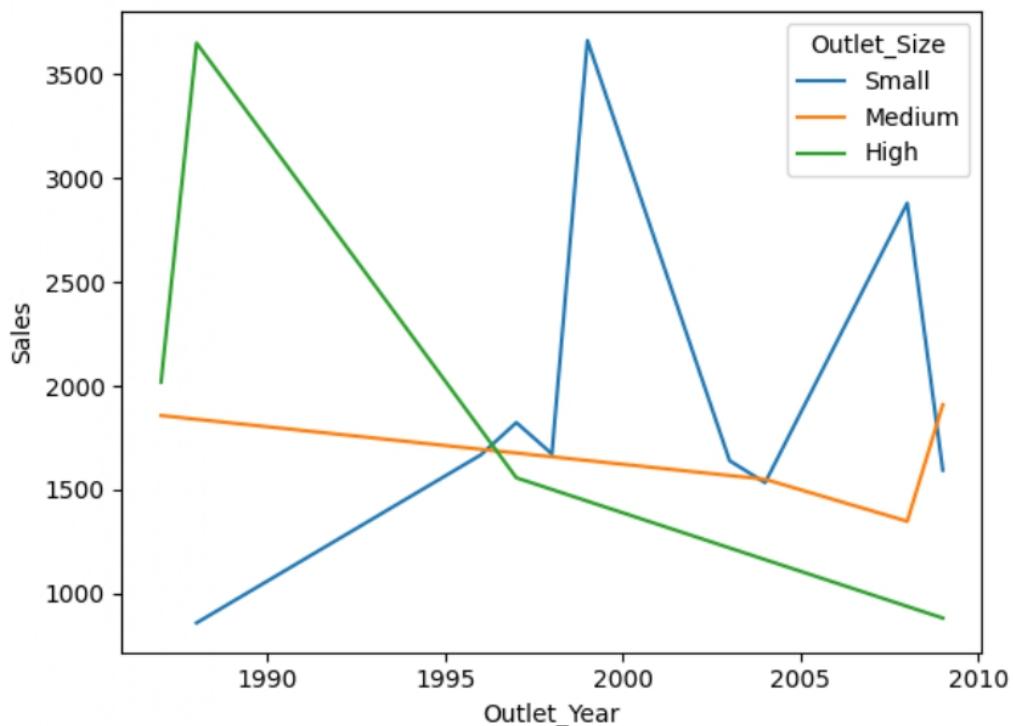
C:\Users\joypa\AppData\Local\Temp\ipykernel_7164\97926062
9.py:1: FutureWarning:

The `ci` parameter is deprecated. Use `errorbar=None` for the same effect.

```
sns.lineplot(data = m,x = "Outlet_Year",y = "Sales",ci = None,hue = "Outlet_Size")
```

Out[88]:

```
<AxesSubplot: xlabel='Outlet_Year', ylabel='Sales'>
```



P) REL plot

In [89]:

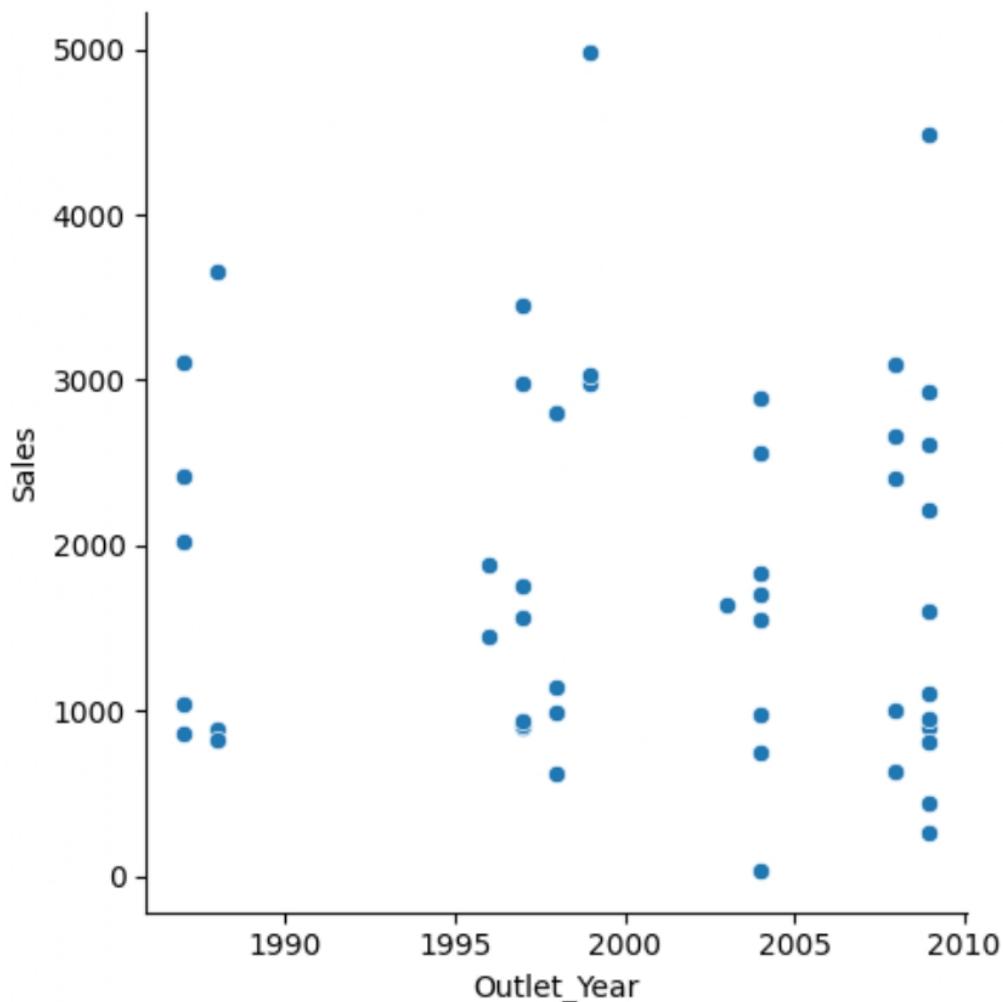
```
# REL plot  = Relational Plot
```

In [90]:

```
sns.relplot(data = m,x = "Outlet_Year",y = "Sales")
```

Out[90]:

```
<seaborn.axisgrid.FacetGrid at 0x2420400075b0>
```

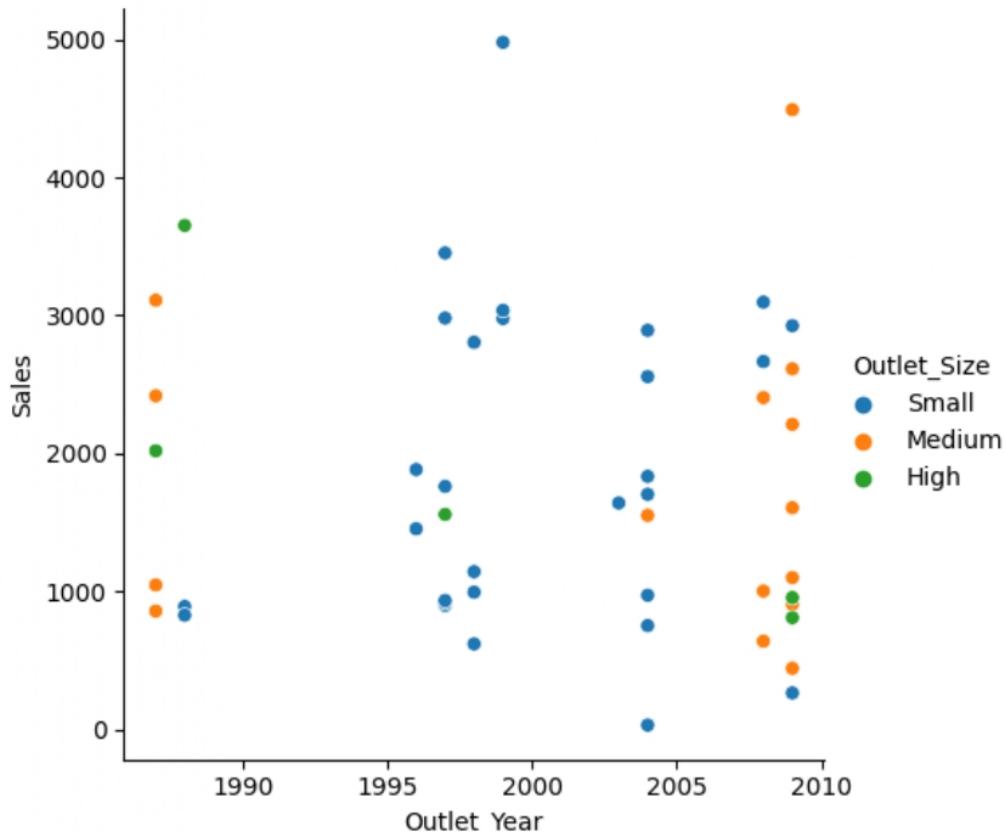


In [91]:

```
sns.relplot(data = m,x = "Outlet_Year",y = "Sales",hue = "Outlet_Size")
```

Out[91]:

```
<seaborn.axisgrid.FacetGrid at 0x24202dad9c0>
```

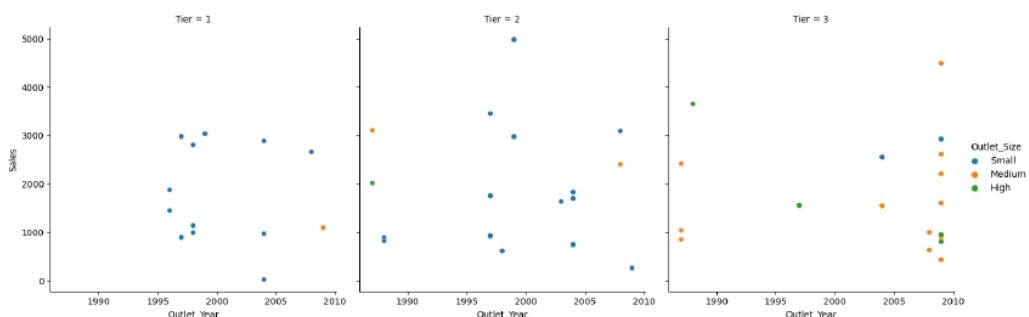


In [92]:

```
sns.relplot(data = m,x = "Outlet_Year",y = "Sales",hue = "Outlet_Size",col = "Tier")
```

Out[92]:

```
<seaborn.axisgrid.FacetGrid at 0x24202c96800>
```



P) REG plot

In [93]:

```
# reg plot = Regression plot
```

In [94]:

```
tip = sns.load_dataset("tips")
```

In [95]:

```
tip.head()
```

Out[95]:

	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

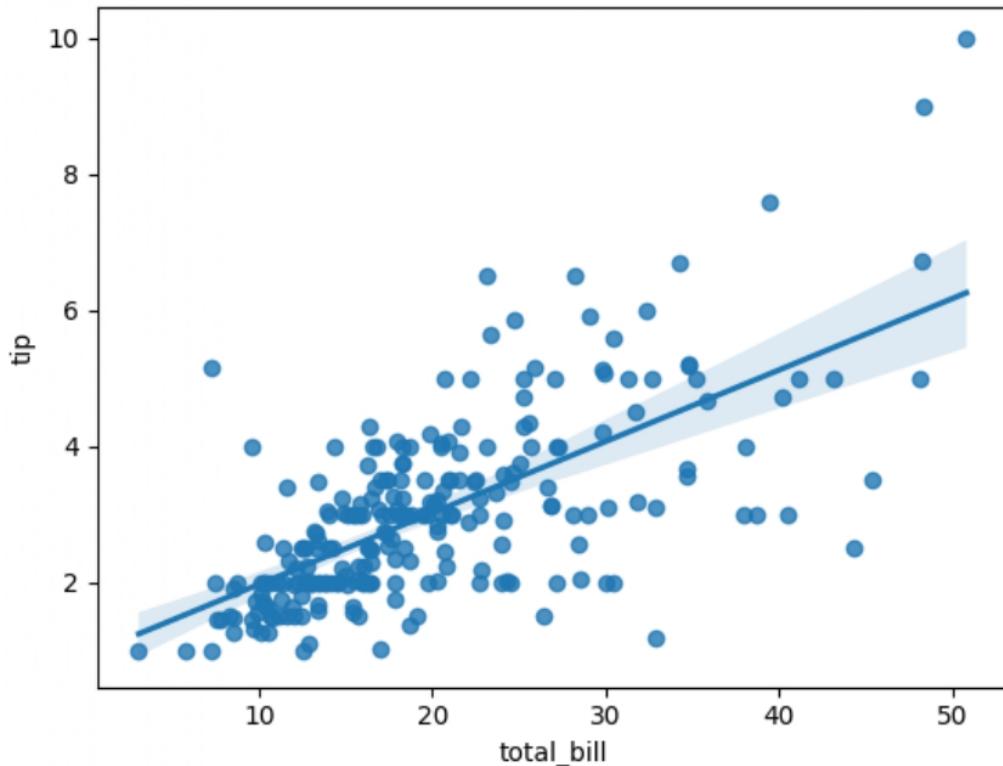
1. create a first ref plot\

In [96]:

```
sns.regplot(data = tip,x = "total_bill",y = "tip")
```

Out[96]:

```
<AxesSubplot: xlabel='total_bill', ylabel='tip'>
```

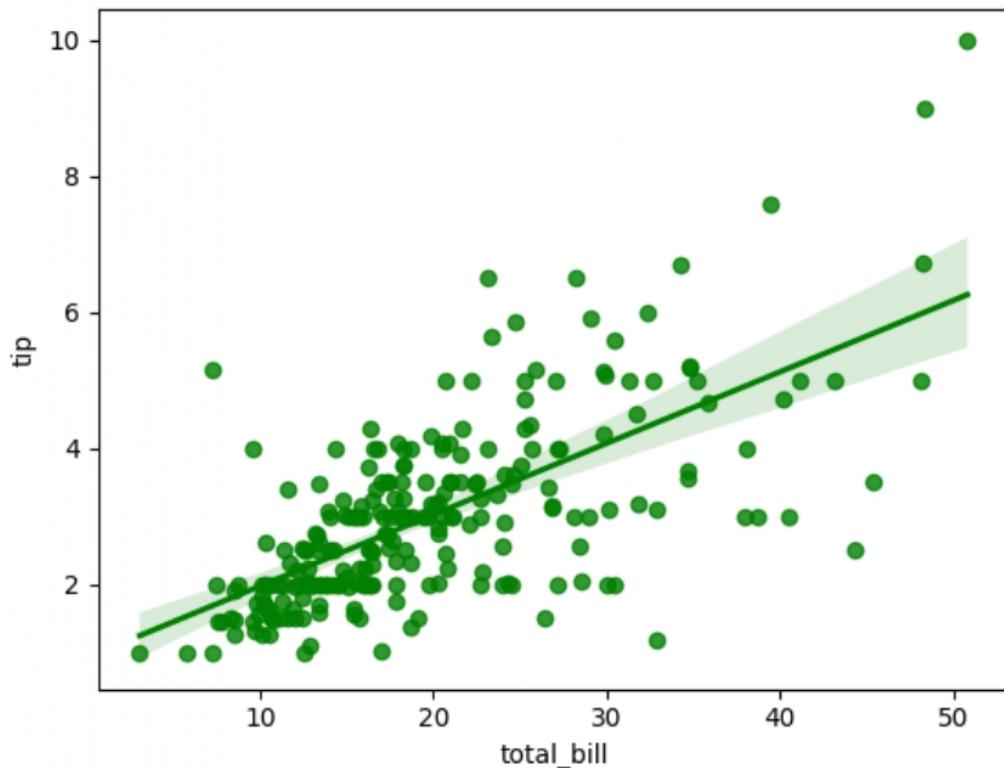


In [97]:

```
sns.regplot(data = tip,x = "total_bill",y = "tip",color= "green")
```

Out[97]:

```
<AxesSubplot: xlabel='total_bill', ylabel='tip'>
```

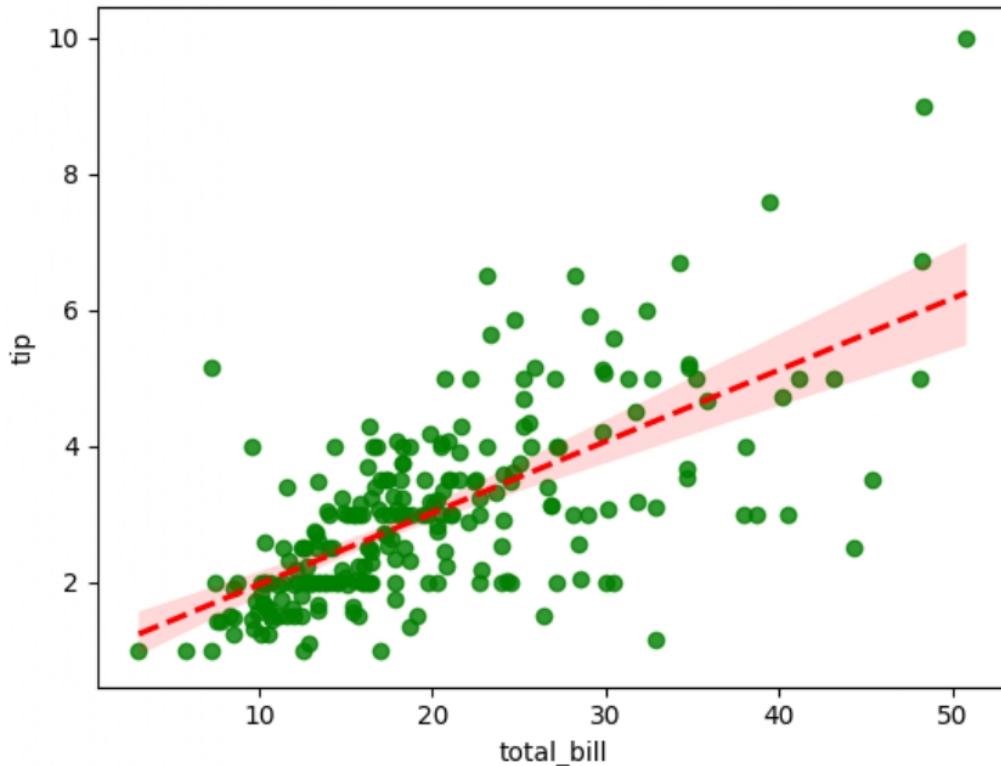


In [98]:

```
sns.regplot(data = tip,x = "total_bill",y = "tip",color= "green"  
            ,line_kws = {"color":"red","linestyle":"--"})
```

Out[98]:

```
<AxesSubplot: xlabel='total_bill', ylabel='tip'>
```



Q) HEATMAPS

In [99]:

```
# This plot only apply for numeric value not catagorical value
```

In []: