```
In [261]: import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
```

In [262]: df\_new1=pd.read\_csv("aerofit\_treadmill.csv")
 df\_new1.head()

### Out[262]:

	Product	Age	Gender	Education	MaritalStatus	Usage	Fitness	Income	Miles
0	KP281	18	Male	14	Single	3	4	29562	112
1	KP281	19	Male	15	Single	2	3	31836	75
2	KP281	19	Female	14	Partnered	4	3	30699	66
3	KP281	19	Male	12	Single	3	3	32973	85
4	KP281	20	Male	13	Partnered	4	2	35247	47

In [263]: len(df\_new1)

Out[263]: 180

In [264]: df\_new1.info()

<class 'pandas.core.frame.DataFrame'>
RangeIndex: 180 entries, 0 to 179
Data columns (total 9 columns):

- 0. 0 0.	00-0		
#	Column	Non-Null Count	Dtype
0	Product	180 non-null	object
1	Age	180 non-null	int64
2	Gender	180 non-null	object
3	Education	180 non-null	int64
4	MaritalStatus	180 non-null	object
5	Usage	180 non-null	int64
6	Fitness	180 non-null	int64
7	Income	180 non-null	int64
8	Miles	180 non-null	int64

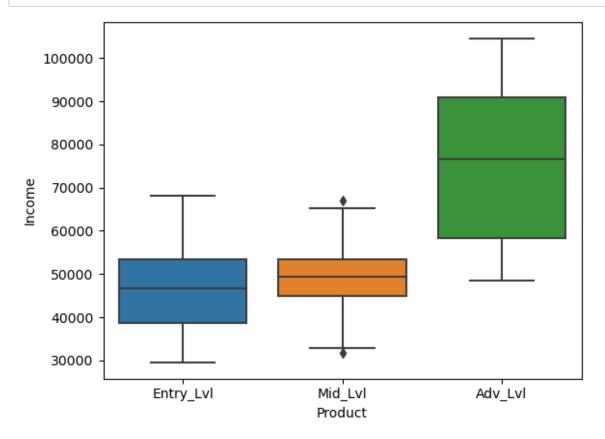
dtypes: int64(6), object(3)
memory usage: 12.8+ KB

```
In [265]: df_new1.describe()
Out[265]:
                                 Education
                                                Usage
                                                           Fitness
                                                                         Income
                                                                                       Miles
                          Age
             count 180.000000
                                180.000000
                                           180.000000
                                                       180.000000
                                                                      180.000000
                                                                                 180.000000
                     28.788889
                                 15.572222
                                             3.455556
                                                          3.311111
                                                                    53719.577778
                                                                                 103.194444
             mean
               std
                      6.943498
                                  1.617055
                                             1.084797
                                                         0.958869
                                                                    16506.684226
                                                                                   51.863605
                     18.000000
                                 12.000000
                                             2.000000
                                                         1.000000
                                                                    29562.000000
                                                                                   21.000000
               min
               25%
                     24.000000
                                 14.000000
                                             3.000000
                                                         3.000000
                                                                    44058.750000
                                                                                   66.000000
              50%
                     26.000000
                                 16.000000
                                             3.000000
                                                         3.000000
                                                                    50596.500000
                                                                                   94.000000
              75%
                     33.000000
                                 16.000000
                                             4.000000
                                                         4.000000
                                                                    58668.000000
                                                                                  114.750000
                     50.000000
                                 21.000000
                                             7.000000
                                                         5.000000
                                                                   104581.000000
                                                                                 360.000000
              max
In [266]:
            df_new1.isnull().sum()
Out[266]: Product
                                 0
                                 0
            Age
            Gender
                                 0
            Education
            MaritalStatus
                                 0
                                 0
            Usage
            Fitness
                                 0
            Income
                                 0
            Miles
                                 0
            dtype: int64
In [267]: df_new1["Product"]=df_new1["Product"].replace(["KP281","KP481","KP781"],["Entry_Lv1"
In [268]: df_new1.head()
Out[268]:
                 Product Age Gender
                                      Education MaritalStatus Usage Fitness Income Miles
             0 Entry_Lvl
                           18
                                  Male
                                              14
                                                                    3
                                                                             4
                                                                                 29562
                                                                                          112
                                                         Single
                Entry_LvI
                           19
                                  Male
                                              15
                                                                    2
                                                                             3
                                                                                 31836
                                                                                           75
                                                         Single
             2 Entry_LvI
                                              14
                                                      Partnered
                                                                    4
                                                                             3
                                                                                 30699
                           19
                               Female
                                                                                           66
               Entry_LvI
                           19
                                              12
                                                                    3
                                                                             3
                                                                                 32973
                                                                                           85
                                  Male
                                                         Single
                           20
                                              13
                                                                    4
                                                                             2
                                                                                 35247
                                                                                           47
               Entry_LvI
                                  Male
                                                      Partnered
```

In [269]: def boxView(df,x,y):

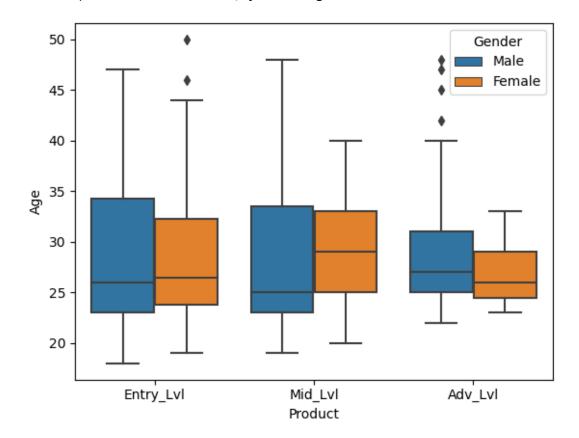
sns.boxplot(data=df\_new1,x=x,y=y)

In [270]: boxView(df\_new1,'Product','Income')



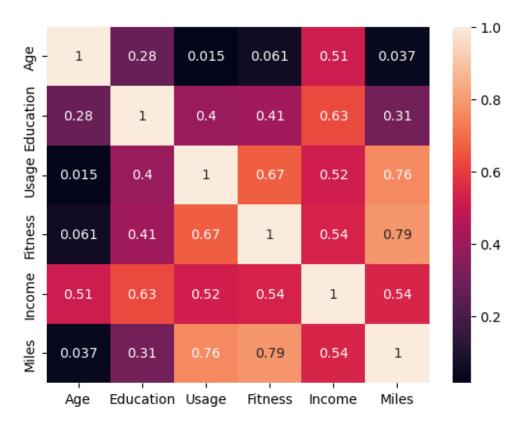
In [271]: sns.boxplot(data=df\_new1,hue='Gender',y='Age',x="Product")

Out[271]: <AxesSubplot:xlabel='Product', ylabel='Age'>

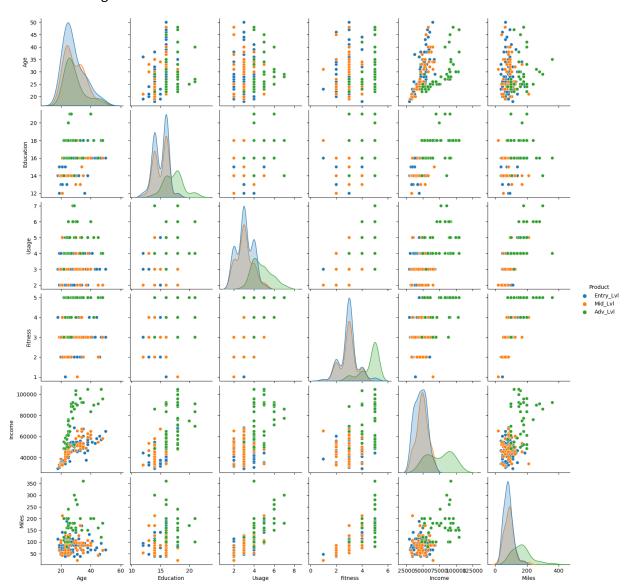


In [272]: sns.heatmap(df\_new1.corr(),annot=True)

Out[272]: <AxesSubplot:>

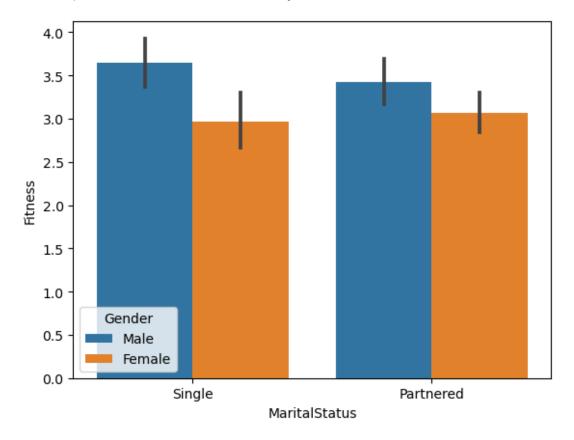


Out[273]: <seaborn.axisgrid.PairGrid at 0x2591de02b20>



In [274]: #Relation between Maritialstatus and fitness
sns.barplot(data=df\_new1,x='MaritalStatus',y='Fitness',hue='Gender')

Out[274]: <AxesSubplot:xlabel='MaritalStatus', ylabel='Fitness'>



In [275]: df\_new1

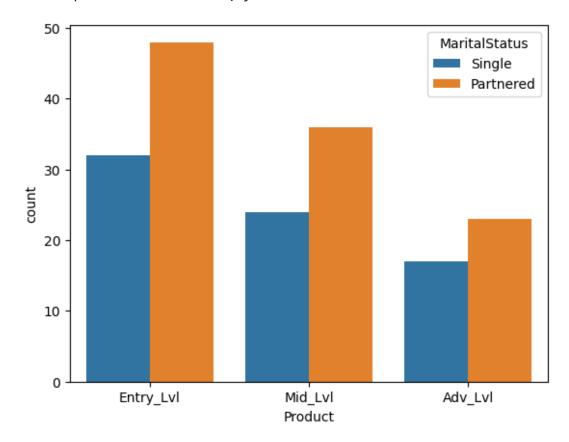
### Out[275]:

	Product	Age	Gender	Education	MaritalStatus	Usage	Fitness	Income	Miles
0	Entry_LvI	18	Male	14	Single	3	4	29562	112
1	Entry_LvI	19	Male	15	Single	2	3	31836	75
2	Entry_LvI	19	Female	14	Partnered	4	3	30699	66
3	Entry_LvI	19	Male	12	Single	3	3	32973	85
4	Entry_LvI	20	Male	13	Partnered	4	2	35247	47
175	Adv_LvI	40	Male	21	Single	6	5	83416	200
176	Adv_LvI	42	Male	18	Single	5	4	89641	200
177	Adv_LvI	45	Male	16	Single	5	5	90886	160
178	Adv_Lvl	47	Male	18	Partnered	4	5	104581	120
179	Adv_LvI	48	Male	18	Partnered	4	5	95508	180

180 rows × 9 columns

In [276]: #Relation between Product bought and maritial status
sns.countplot(data=df\_new1,x='Product',hue='MaritalStatus')

Out[276]: <AxesSubplot:xlabel='Product', ylabel='count'>



### Out[294]:

Gender	Product	Female	Male
0	Adv_LvI	7	33
1	Entry_LvI	40	40
2	Mid_Lvl	29	31

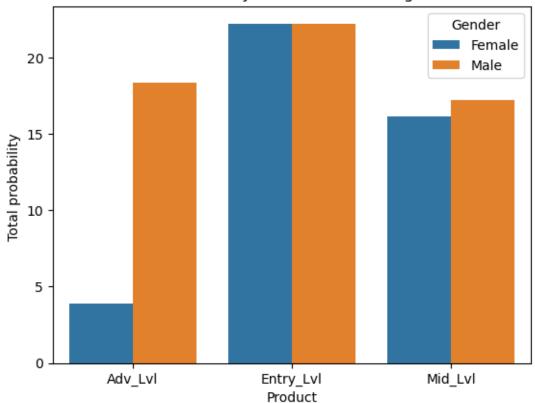
```
In [295]: #Probability of Total Sales based on gender
    prob_gender=(crsstb_g.iloc[:,1:]/len(df_new1))*100
    prob_gender
    prob_plot_data=pd.concat([crsstb_g['Product'],prob_gender],axis=1)
    print(prob_plot_data)

print()
    prob_plot_data_clean=pd.melt(prob_plot_data,id_vars=['Product'],var_name='Gender',vat_print(prob_plot_data_clean)

sns.barplot(data=prob_plot_data_clean,x='Product',y='Probability',hue='Gender')
    plt.title("Total Probability of Sales based on gender")
    plt.ylabel("Total probability ")
    plt.show()
```

```
Product
                Female
                             Male
0
    Adv_Lvl
              3.888889 18.333333
1 Entry_Lvl 22.22222 22.22222
2
    Mid Lvl 16.111111 17.222222
    Product Gender Probability
0
    Adv Lvl Female
                        3.888889
1 Entry_Lvl Female
                       22.22222
2
    Mid Lvl Female
                       16.111111
3
    Adv Lvl
               Male
                       18.333333
4
  Entry_Lvl
               Male
                       22.22222
    Mid_Lvl
               Male
                       17.222222
```

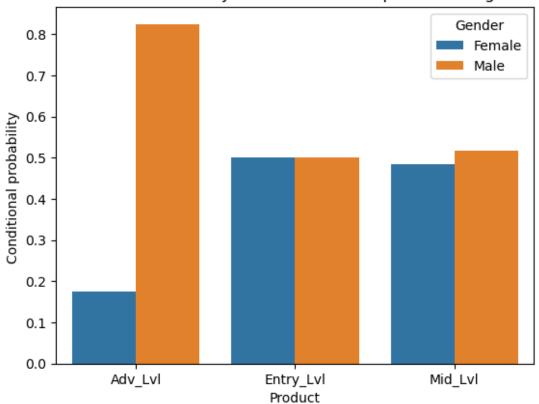
# Total Probability of Sales based on gender



```
In [299]: #Conditional probability for each product based on gender
    cond_crsstb_g=pd.crosstab(index=df_new1['Product'],columns=df_new1['Gender'],margins
    c_prob_plot_data_clean=pd.melt(cond_crsstb_g,id_vars=['Product'],var_name='Gender',v.
    print(cond_crsstb_g)
    print()
    sns.barplot(data=c_prob_plot_data_clean,x='Product',y='Probability',hue='Gender')
    plt.title("Conditional Probability of Sales based on product and gender")
    plt.ylabel("Conditional probability ")
    plt.show()
```

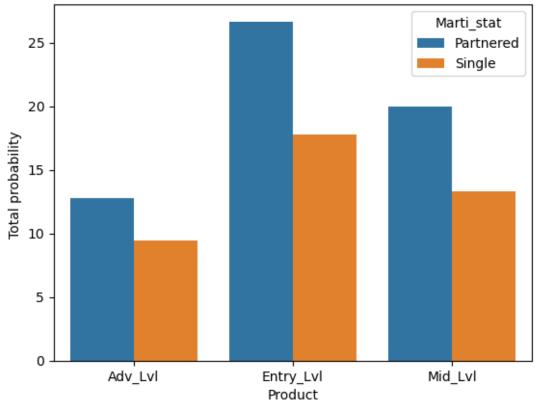
Gender	Product	Female	Male
0	Adv_Lvl	0.175000	0.825000
1	Entry_Lvl	0.500000	0.500000
2	${\sf Mid\_Lvl}$	0.483333	0.516667

# Conditional Probability of Sales based on product and gender



```
Product Partnered
                           Single
    Adv Lvl 12.777778
                         9.444444
1 Entry_Lvl 26.666667
                        17.77778
    Mid_Lvl 20.000000 13.333333
     Product Marti_stat Probability
0
    Adv_Lvl Partnered
                          12.777778
1 Entry Lvl Partnered
                          26.666667
2
    Mid_Lvl Partnered
                          20.000000
3
                Single
                          9.444444
    Adv_Lvl
4 Entry Lvl
                Single
                          17.777778
5
    Mid_Lvl
                Single
                          13.333333
```

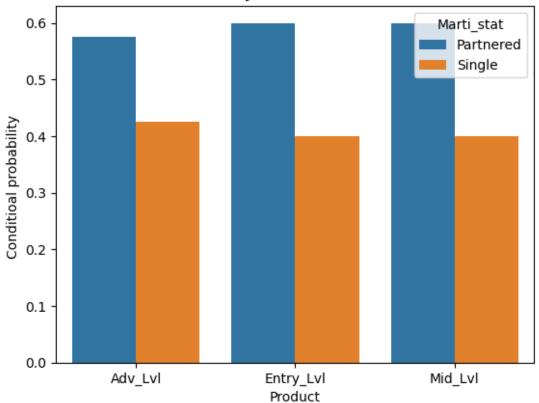
# Total Probability of Sales based on MaritalStatus



In [298]: #Conditional prob for each prodcut based on Martitals status
 cond\_crsstb\_g=pd.crosstab(index=df\_new1['Product'],columns=df\_new1['MaritalStatus'],c

MaritalStatus	Product	Partnered	Single
0	Adv_Lvl	0.575	0.425
1	Entry_Lvl	0.600	0.400
2	${\sf Mid\_Lvl}$	0.600	0.400

# Conditional Probability of Sales based on MaritialStatus



```
In [282]: #Categorizing based on age
bins=[13,19,29,39,49,59]
label=['Teens','20s','30s','40s','50s']
df_new1["Age_cat"]=pd.cut(df_new1["Age"],bins=bins,labels=label)
df_new1
```

# Out[282]:

	Product	Age	Gender	Education	MaritalStatus	Usage	Fitness	Income	Miles	Age_cat
0	Entry_LvI	18	Male	14	Single	3	4	29562	112	Teens
1	Entry_LvI	19	Male	15	Single	2	3	31836	75	Teens
2	Entry_LvI	19	Female	14	Partnered	4	3	30699	66	Teens
3	Entry_LvI	19	Male	12	Single	3	3	32973	85	Teens
4	Entry_LvI	20	Male	13	Partnered	4	2	35247	47	20s
175	Adv_LvI	40	Male	21	Single	6	5	83416	200	40s
176	Adv_LvI	42	Male	18	Single	5	4	89641	200	40s
177	Adv_LvI	45	Male	16	Single	5	5	90886	160	40s
178	Adv_LvI	47	Male	18	Partnered	4	5	104581	120	40s
179	Adv_LvI	48	Male	18	Partnered	4	5	95508	180	40s

180 rows × 10 columns

```
In [303]: #Probability of Sales based on Age_cat
          crsstb_a=pd.crosstab(index=df_new1['Product'],columns=df_new1['Age_cat'],margins=Fal
          prob_mstatus=crsstb_a.iloc[:,1:]/len(df_new1)*100
          prob_mstatus
          prob_plot_data=pd.concat([crsstb_a['Product'],prob_mstatus],axis=1)
          print(prob plot data)
          print()
          prob_plot_data_clean=pd.melt(prob_plot_data,id_vars=['Product'],var_name='Age_cat',v
          prob_plot_data_clean
          print(prob plot data clean)
          sns.barplot(data=prob_plot_data_clean,x='Product',y='Probability',hue='Age_cat')
          plt.title("Total Probability of Sales based on Age Category")
          plt.ylabel("Total probability ")
          plt.show()
               Product
                           Teens
                                        20s
                                                   30s
                                                             40s
                                                                      50s
          0
               Adv_Lvl 0.000000
                                 15.000000
                                              4.44444 2.777778 0.000000
          1 Entry Lvl 2.222222
                                  27.222222
                                            11.111111 3.333333 0.555556
               Mid_Lvl 0.555556 17.777778
                                            12.222222 2.777778 0.000000
                Product Age_cat Probability
          0
                Adv_Lvl
                         Teens
                                    0.000000
```

2

3

4

5

6

7

8

9

11

12

14

Entry\_Lvl

Entry\_Lvl

Entry\_Lvl

10 Entry\_Lvl

13 Entry\_Lvl

Mid Lvl

 $Adv_Lvl$ 

Mid Lvl

Adv\_Lvl

Mid\_Lvl

Adv Lvl

Mid\_Lvl

Adv\_Lvl

Mid\_Lvl

Teens

Teens

20s

20s

20s

30s

30s

30s

40s

40s

40s

50s

50s

50s

2.22222

0.555556

15.000000

27.222222

17.777778

4.44444 11.111111

12.222222

2.777778

3.333333

2.777778

0.000000

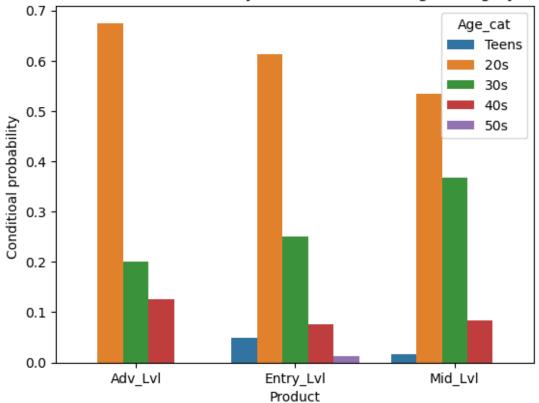
0.555556

0.000000

Total Probability of Sales based on Age Category Age\_cat Teens 25 20s 30s 40s 20 50s Total probability 15 10 5 0 Entry\_Lvl Product Mid\_Lvl Adv\_Lvl

Age_cat	Product	Teens	20s	30s	40s	50s
0	Adv_Lvl	0.000000	0.675000	0.200000	0.125000	0.0000
1	Entry_Lvl	0.050000	0.612500	0.250000	0.075000	0.0125
2	Mid_Lvl	0.016667	0.533333	0.366667	0.083333	0.0000





In [285]: #Categorizing based on Salary
bins=[20000,29000,39000,49000,59000,69000,79000,89000,99000,109000]
label=['20k-30k','30k-40k','40k-50k','50k-60k','60k-70k','70k-80k','80k-90k','90k-100
df\_new1["Income\_cat"]=pd.cut(df\_new1["Income"],bins=bins,labels=label)
df\_new1

### Out[285]:

	Product	Age	Gender	Education	MaritalStatus	Usage	Fitness	Income	Miles	Age_cat	Income_
0	Entry_LvI	18	Male	14	Single	3	4	29562	112	Teens	30k-
1	Entry_LvI	19	Male	15	Single	2	3	31836	75	Teens	30k-
2	Entry_LvI	19	Female	14	Partnered	4	3	30699	66	Teens	30k-
3	Entry_LvI	19	Male	12	Single	3	3	32973	85	Teens	30k-
4	Entry_LvI	20	Male	13	Partnered	4	2	35247	47	20s	30k-
175	Adv_LvI	40	Male	21	Single	6	5	83416	200	40s	80k-
176	Adv_LvI	42	Male	18	Single	5	4	89641	200	40s	90k-1
177	Adv_LvI	45	Male	16	Single	5	5	90886	160	40s	90k-1
178	Adv_LvI	47	Male	18	Partnered	4	5	104581	120	40s	100k-1
179	Adv_Lvl	48	Male	18	Partnered	4	5	95508	180	40s	90k-1

180 rows × 11 columns

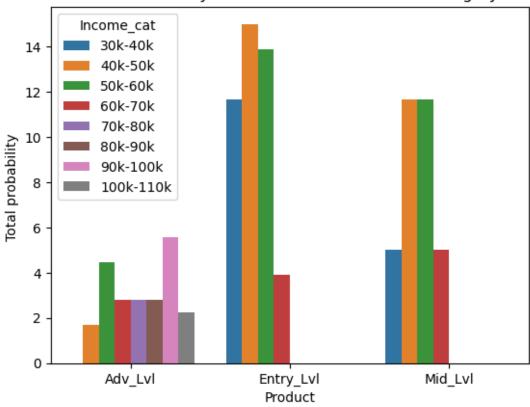
```
In [306]: #Probability of Sales based on Age_cat
          crsstb s=pd.crosstab(index=df new1['Product'],columns=df new1['Income cat'],margins=
          prob_mstatus=crsstb_s.iloc[:,1:]/len(df_new1)*100
          prob_mstatus
          prob_plot_data=pd.concat([crsstb_s['Product'],prob_mstatus],axis=1)
          print(crsstb s)
          print()
          prob_plot_data_clean=pd.melt(prob_plot_data,id_vars=['Product'],var_name='Income_cat
          prob_plot_data_clean
          print(prob plot data clean)
          sns.barplot(data=prob plot data clean,x='Product',y='Probability',hue='Income cat')
          plt.title("Total Probability of Sales based on Income Category")
          plt.ylabel("Total probability ")
          plt.show()
                                           40k-50k
                                                     50k-60k
                                                              60k-70k
                                                                        70k-80k
          Income cat
                         Product
                                  30k-40k
                                                                                 80k-90k
                         Adv_Lvl
                                        0
                                                  3
                                                           8
                                                                    5
                                                                              5
                                                                                       5
                                                                    7
          1
                       Entry_Lvl
                                       21
                                                 27
                                                          25
                                                                              0
                                                                                       0
          2
                         Mid Lvl
                                        9
                                                 21
                                                          21
                                                                    9
                                                                              0
                                                                                       0
          Income_cat 90k-100k
                                 100k-110k
          0
                             10
                                         4
          1
                              0
                                         0
          2
                              0
                                         0
                Product Income_cat
                                     Probability
          0
                Adv Lvl
                            30k-40k
                                        0.000000
          1
               Entry Lvl
                            30k-40k
                                       11.666667
          2
                Mid_Lvl
                            30k-40k
                                        5.000000
          3
                Adv Lvl
                            40k-50k
                                        1.666667
          4
              Entry_Lvl
                            40k-50k
                                       15.000000
                            40k-50k
          5
                Mid Lvl
                                       11.666667
          6
                Adv Lvl
                            50k-60k
                                        4.44444
          7
               Entry_Lvl
                            50k-60k
                                       13.888889
          8
                Mid_Lvl
                            50k-60k
                                       11.666667
          9
                Adv Lvl
                            60k-70k
                                        2.777778
          10
              Entry_Lvl
                            60k-70k
                                        3.888889
          11
                Mid_Lvl
                            60k-70k
                                        5.000000
          12
                Adv_Lvl
                            70k-80k
                                        2.777778
              Entry Lvl
          13
                            70k-80k
                                        0.000000
          14
                Mid Lvl
                            70k-80k
                                        0.000000
          15
                Adv_Lvl
                            80k-90k
                                        2.777778
              Entry_Lvl
                            80k-90k
          16
                                        0.000000
          17
                Mid Lvl
                            80k-90k
                                        0.000000
                Adv Lvl
          18
                           90k-100k
                                        5.55556
          19
              Entry_Lvl
                           90k-100k
                                        0.000000
                Mid_Lvl
                           90k-100k
          20
                                        0.000000
          21
                Adv Lvl
                          100k-110k
                                        2.22222
          22
              Entry Lvl
                          100k-110k
                                        0.000000
```

Mid\_Lvl

100k-110k

0.000000

Total Probability of Sales based on Income Category



<pre>Income_cat</pre>	Product	30k-40k	40k-50k	50k-60k	60k-70k	70k-80k	80k-90k	\	
0	Adv_Lvl	0.0000	0.0750	0.2000	0.1250	0.125	0.125		
1	Entry_Lvl	0.2625	0.3375	0.3125	0.0875	0.000	0.000		
2	Mid_Lvl	0.1500	0.3500	0.3500	0.1500	0.000	0.000		
<pre>Income_cat</pre>	90k-100k	100k-110k							
0	0.25	0.1							

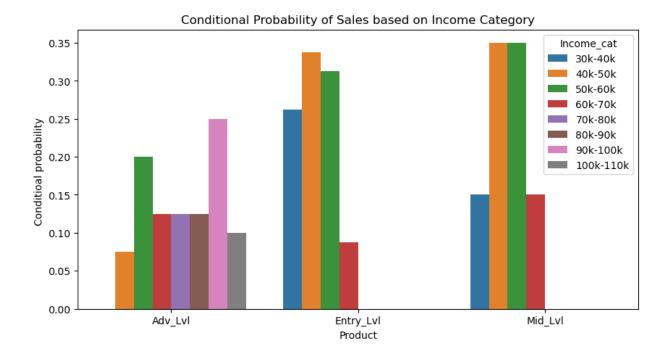
2

0.00

0.00

0.0

0.0



In [308]: df\_new1

Out[308]:

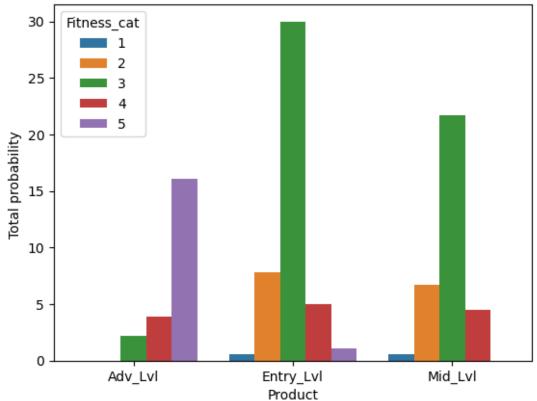
	Product	Age	Gender	Education	MaritalStatus	Usage	Fitness	Income	Miles	Age_cat	Income_
0	Entry_LvI	18	Male	14	Single	3	4	29562	112	Teens	30k-
1	Entry_LvI	19	Male	15	Single	2	3	31836	75	Teens	30k-
2	Entry_LvI	19	Female	14	Partnered	4	3	30699	66	Teens	30k-
3	Entry_LvI	19	Male	12	Single	3	3	32973	85	Teens	30k-
4	Entry_LvI	20	Male	13	Partnered	4	2	35247	47	20s	30k-
175	Adv_LvI	40	Male	21	Single	6	5	83416	200	40s	80k-
176	Adv_LvI	42	Male	18	Single	5	4	89641	200	40s	90k-1
177	Adv_LvI	45	Male	16	Single	5	5	90886	160	40s	90k-1
178	Adv_LvI	47	Male	18	Partnered	4	5	104581	120	40s	100k-1
179	Adv_LvI	48	Male	18	Partnered	4	5	95508	180	40s	90k-1

180 rows × 11 columns

In [ ]:

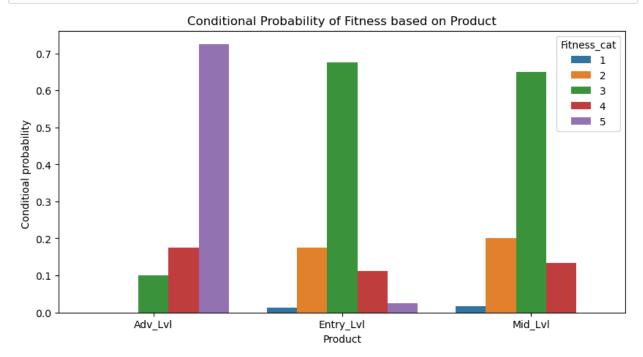
# In [309]: #Probability of Fitness based on product crsstb\_f=pd.crosstab(index=df\_new1['Product'],columns=df\_new1['Fitness'],margins=Fale prob\_mstatus=crsstb\_f.iloc[:,1:]/len(df\_new1)\*100 prob\_mstatus prob\_plot\_data=pd.concat([crsstb\_f['Product'],prob\_mstatus],axis=1) prob\_plot\_data\_clean=pd.melt(prob\_plot\_data,id\_vars=['Product'],var\_name='Fitness\_car prob\_plot\_data\_clean sns.barplot(data=prob\_plot\_data\_clean,x='Product',y='Probability',hue='Fitness\_cat') plt.title("Total Probability of Fitness based on Prodcut used") plt.ylabel("Total probability ") plt.show()

# Total Probability of Fitness based on Prodcut used



```
In [290]: #Conditional prob for fitness based on product
    cond_crsstb_s=pd.crosstab(index=df_new1['Product'],columns=df_new1['Fitness'],margin_c_prob_plot_data_clean=pd.melt(cond_crsstb_s,id_vars=['Product'],var_name='Fitness_cc_c_prob_plot_data_clean

    plt.figure(figsize=(10,5))
    sns.barplot(data=c_prob_plot_data_clean,x='Product',y='Probability',hue='Fitness_cat_plt.title("Conditional Probability of Fitness based on Product")
    plt.ylabel("Conditioal probability ")
    plt.show()
```





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