In [1]: **import** numpy **as** np import pandas as pd import matplotlib.pyplot as plt import seaborn as sns In [2]: df=pd.read\_csv("areofit\_analysis.csv") Product Age Gender Education MaritalStatus Usage Fitness Income Miles Out[2]: **0** KP281 18 Male 14 Single 3 29562 112 KP281 19 75 Male 15 3 31836 Single 66 KP281 19 Female Partnered 30699 KP281 12 85 Male 32973 19 Single Partnered 4 KP281 20 Male 13 2 35247 47 21 83416 KP781 40 Male Single 5 200 KP781 18 89641 176 42 Male Single 200 KP781 Male 16 Single 5 5 90886 160 18 5 104581 178 KP781 47 Male Partnered 120 KP781 48 Male 18 Partnered 95508 180 180 rows × 9 columns In [3]: df.head() Out[3]: Product Age Gender Education MaritalStatus Usage Fitness Income Miles KP281 18 Male 29562 112 3 14 Single 3 31836 KP281 19 Male 15 Single 75 19 Female 3 30699 66 KP281 14 Partnered 4 KP281 19 Single 3 32973 85 **4** KP281 20 Male 13 Partnered 4 2 35247 47 In [4]: # Shape of Data df.shape (180, 9) Out[4]: # Describe of Data df.describe() Age Education Out[5]: Usage **Fitness** Income Miles **count** 180.000000 180.000000 180.000000 180.000000 180.000000 180.000000 mean 28.788889 15.572222 3.455556 3.311111 53719.577778 103.194444 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 14.000000 66.000000 **25**% 24.000000 3.000000 3.000000 44058.750000 26.000000 16.000000 3.000000 3.000000 50596.500000 94.000000 **50**% 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 [6]: # NULL Values df.isnull().sum() Out[6]: Age Gender 0 Education 0 MaritalStatus 0 0 Usage Fitness 0 0 Income 0 Miles dtype: int64 In [7]: # Value\_Counts df['Product'].value\_counts(normalize=True) Out[7]: KP281 0.444444 KP481 0.333333 KP781 0.222222 Name: proportion, dtype: float64 In [8]: # Unique Values df.nunique() Product 3 Out[8]: Age 32 Gender 2 Education 8 MaritalStatus 2 6 Usage 5 Fitness Income 62 Miles 37 dtype: int64 In [9]: # counting Male and Female ax=sns.countplot(data=df, x="Gender") for i in ax.containers: ax.bar\_label(i) 104 100 80 76 60 40 20 Female Male Gender In [10]: # Males Have More Numbers of Peoples In [11]: ax=sns.histplot(data=df, x="Age", kde=True) for i in ax.containers: ax.bar\_label(i) 40 30 20 10 20 25 30 35 40 50 Age In [12]: # Between 23 To 26 Have More Customers In [13]: ax=sns.barplot(data=df, x="Product", y="Income") for i in ax.containers: ax.bar\_label(i) 80000 75441.6 70000 60000 48973.7 50000 46418 40000 30000 20000 10000 0 -KP281 KP481 KP781 Product In [14]: # product KP781 More Income sns.boxplot(data=df, x="Gender", y="Income", hue="Product") <Axes: xlabel='Gender', ylabel='Income'> Product 100000 KP281 KP481 KP781 90000 80000 Income 70000 60000 50000 40000 30000 Male Female Gender In [16]: sns.boxplot(data=df, x="Gender", y="Miles", hue="Product") <Axes: xlabel='Gender', ylabel='Miles'> Product 350 KP281 KP481 300 KP781 250 Wiles 200 150 100 50 Male Female Gender In [35]: numeric\_data = df.select\_dtypes(include='number') sns.heatmap(numeric\_data.corr(), cmap='viridis', annot=True, fmt=".2f", linewidths=.5) plt.show() 1.0 Product -0.8 0.51 1.00 0.28 0.02 0.06 0.04 Age -0.28 Education -1.00 0.40 0.41 0.63 0.31 - 0.6 0.02 0.40 1.00 0.67 0.52 0.76 Usage -0.4 Fitness -0.06 0.41 0.67 1.00 0.54 0.79 0.54 0.51 0.63 0.52 0.54 Income -1.00 - 0.2 Miles -0.54 0.04 0.31 0.76 0.79 Product Age Education Usage Fitness Income Miles pd.crosstab(index=df['Gender'], columns=df['Product'], margins=True, normalize="index")\*100 Out[17]: Product KP281 **KP481** Gender Female 52.631579 38.157895 9.210526 **Male** 38.461538 29.807692 31.730769 **All** 44.44444 33.333333 22.222222 In [ ]: # KP281 Have High Percenatge Compare to Other Product # Female Have High Percentage In KP281 and KP481 and Male Have More In KP781 In [18]: | pd.crosstab(index=df["Income"], columns=df["Product"], margins=True) Out[18]: Product KP281 KP481 KP781 All Income 29562 0 0 1 30699 0 1 0 31836 0 2 32973 0 5 3 0 5 34110 95866 1 1 99601 1 1 103336 0 1 1 104581 2 2 0 40 180 80 60 63 rows × 4 columns # KP281 Have More Sales In [19]: pd.crosstab(index=df["Age"], columns=df["Product"], margins=True) Out[19]: Product KP281 KP481 KP781 All Age 18 0 0 1 1 19 20 2 3 0 5 21 0 7 22 3 7 23 3 18 24 5 3 4 12 11 7 25 26 3 2 12 28 29 30 2 3 31 32 2 0 4 33 34 6 35 36 0 1 0 2 1 7 38 40 1 42 0 1 1 0 44 1 0 1 45 46 0 1 48 0 1 2 ΑII 80 60 40 180 Recommendations 1.Female customers Have More We Have To Focus To Increase Female Customers Which Females Like The Most Female Customers Using KP281 Have Most 2.We Have To Focus Males Customers Which Product Males Using Most 3.KP781 Have More Income and KP281 Have More sales.