In [1]: #importing libraries import pandas as pd import numpy as np import matplotlib.pyplot as plt import seaborn as sns import warnings warnings.filterwarnings('ignore') In [2]: df = pd.read\_csv('CVD\_cleaned.csv') In [3]: df.head() Out[3]: General\_Health Checkup Exercise Heart\_Disease Skin\_Cancer Other\_Cancer Depression Diabetes Arthritis Sex Age\_Category Height\_(cm) Weight\_(kg) BMI Smoking\_History Alcohol\_Consumption Fruit\_C Within 0 Poor the past No No No No No No Yes Female 70-74 150.0 32.66 14.54 Yes 0.0 2 years Within Very Good the past No Yes No No No Yes No Female 70-74 165.0 77.11 28.29 No 0.0 year Within Very Good the past Yes No No No No Yes No Female 60-64 163.0 88.45 33.47 No 4.0 year Within 75-79 180.0 93.44 28.73 0.0 3 Poor the past Yes Yes No No No Yes No Male No year Within No No No Male 80+ 191.0 88.45 24.37 Yes 0.0 Good the past No No No No year In [4]: df.shape (308854, 19)Out[4]: In [5]: df.describe().T 25% 50% 75% count mean std min max Height\_(cm) 308854.0 170.615249 10.658026 91.00 163.00 170.00 178.00 241.00 Weight\_(kg) 308854.0 83.588655 21.343210 24.95 68.04 81.65 95.25 293.02 **BMI** 308854.0 28.626211 6.522323 12.02 24.21 27.44 31.85 99.33 Alcohol\_Consumption 308854.0 5.096366 0.00 8.199763 0.00 1.00 6.00 30.00 Fruit\_Consumption 308854.0 29.835200 24.875735 0.00 12.00 30.00 30.00 120.00 **Green\_Vegetables\_Consumption** 308854.0 14.926238 0.00 4.00 12.00 20.00 128.00 FriedPotato\_Consumption 308854.0 6.296616 8.582954 0.00 2.00 8.00 128.00 In [6]: df.describe(exclude='number').T

	count	unique	top	freq
General_Health	308854	5	Very Good	110395
Checkup	308854	5	Within the past year	239371
Exercise	308854	2	Yes	239381
Heart_Disease	308854	2	No	283883
Skin_Cancer	308854	2	No	278860
Other_Cancer	308854	2	No	278976
Depression	308854	2	No	246953
Diabetes	308854	4	No	259141
Arthritis	308854	2	No	207783
Sex	308854	2	Female	160196
Age_Category	308854	13	65-69	33434
Smoking History	308854	2	No	183590

<class 'pandas.core.frame.DataFrame'>

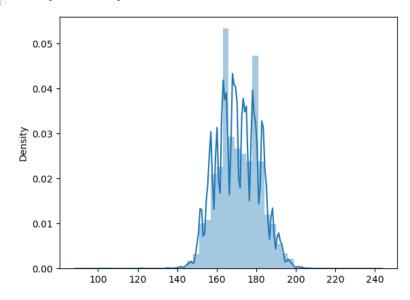
#### In [7]: df.info()

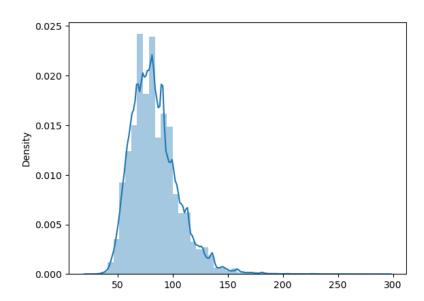
```
RangeIndex: 308854 entries, 0 to 308853
        Data columns (total 19 columns):
        # Column
                                         Non-Null Count Dtype
        --- -----
                                         -----
            General_Health
                                         308854 non-null object
            Checkup
                                         308854 non-null object
            Exercise
                                         308854 non-null object
            Heart_Disease
                                         308854 non-null object
        4 Skin_Cancer
                                         308854 non-null object
        5 Other_Cancer
                                         308854 non-null object
        6 Depression
                                         308854 non-null object
        7
            Diabetes
                                         308854 non-null object
        8 Arthritis
                                         308854 non-null object
                                         308854 non-null object
        9 Sex
                                         308854 non-null object
        10 Age_Category
                                         308854 non-null float64
        11 Height_(cm)
        12 Weight_(kg)
                                         308854 non-null float64
        13 BMI
                                         308854 non-null float64
                                         308854 non-null object
        14 Smoking_History
                                         308854 non-null float64
        15 Alcohol_Consumption
                                         308854 non-null float64
        16 Fruit_Consumption
         17 Green_Vegetables_Consumption 308854 non-null float64
         18 FriedPotato_Consumption
                                         308854 non-null float64
        dtypes: float64(7), object(12)
        memory usage: 44.8+ MB
In [8]: numeric = df.columns[df.dtypes != 'object']
        cat = df.columns[df.dtypes == 'object']
        print(numeric)
        print(cat)
        Index(['Height_(cm)', 'Weight_(kg)', 'BMI', 'Alcohol_Consumption',
               'Fruit_Consumption', 'Green_Vegetables_Consumption',
              'FriedPotato_Consumption'],
             dtype='object')
        Index(['General_Health', 'Checkup', 'Exercise', 'Heart_Disease', 'Skin_Cancer',
               'Other_Cancer', 'Depression', 'Diabetes', 'Arthritis', 'Sex',
               'Age_Category', 'Smoking_History'],
             dtype='object')
```

# **Analysing Dataset**

# **Univariate Analysis**

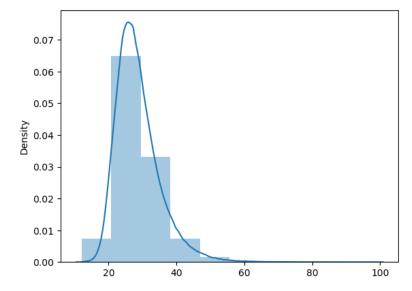
```
In [9]: sns.distplot(x = df['Height_(cm)'])
Out[9]: <Axes: ylabel='Density'>
```





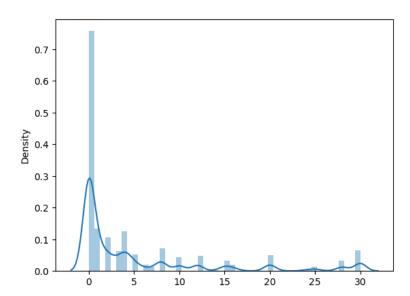
In [11]: sns.distplot(x = df['BMI'], bins=10)

Out[11]: <Axes: ylabel='Density'>



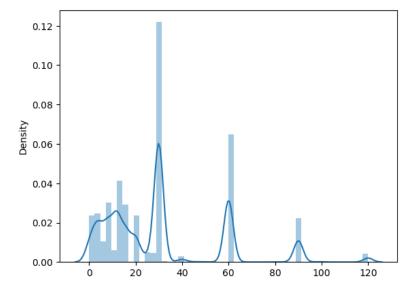
In [12]: sns.distplot(x = df['Alcohol\_Consumption'])

Out[12]: <Axes: ylabel='Density'>



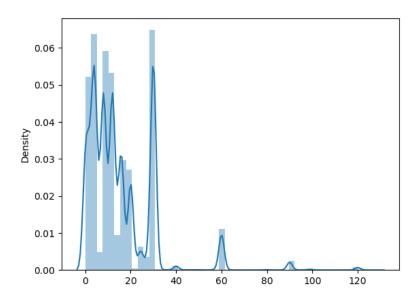
In [13]: sns.distplot(x = df['Fruit\_Consumption'])

Out[13]: <Axes: ylabel='Density'>



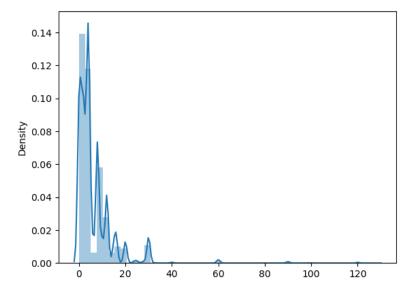
In [14]: sns.distplot(x = df['Green\_Vegetables\_Consumption'])

Out[14]: <Axes: ylabel='Density'>



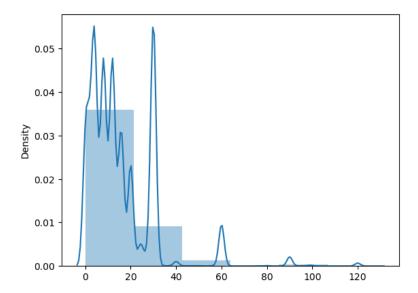
In [15]: sns.distplot(x = df['FriedPotato\_Consumption'])

Out[15]: <Axes: ylabel='Density'>



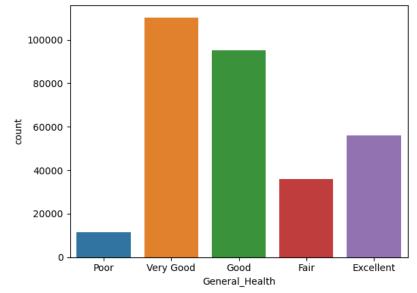
In [16]: sns.distplot(x = df['Green\_Vegetables\_Consumption'], bins=6)

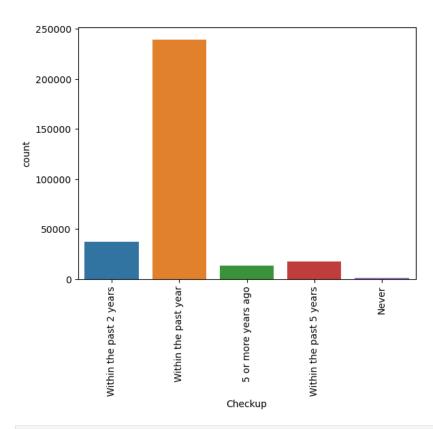
Out[16]: <Axes: ylabel='Density'>



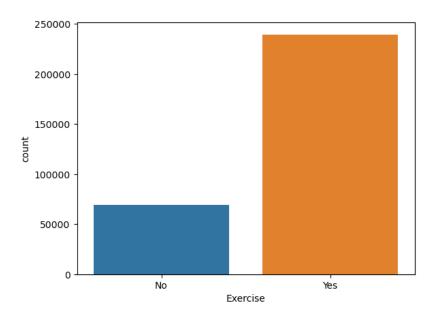
In [17]: sns.countplot(x=df['General\_Health'])

Out[17]: <Axes: xlabel='General\_Health', ylabel='count'>



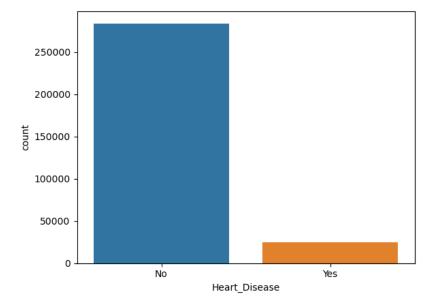


```
In [19]: sns.countplot(x=df['Exercise'])
Out[19]: <Axes: xlabel='Exercise', ylabel='count'>
```

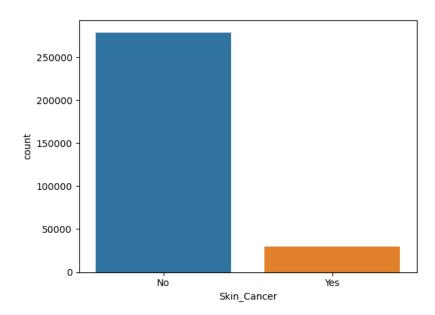


```
In [20]: sns.countplot(x=df['Heart_Disease'])
```

Out[20]: <Axes: xlabel='Heart\_Disease', ylabel='count'>

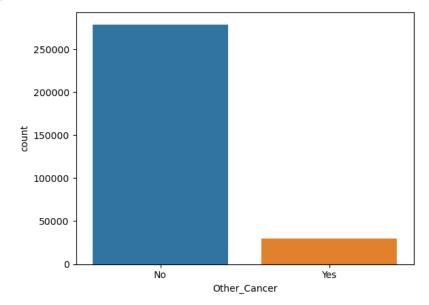


```
In [21]: sns.countplot(x=df['Skin_Cancer'])
Out[21]: <Axes: xlabel='Skin_Cancer', ylabel='count'>
```

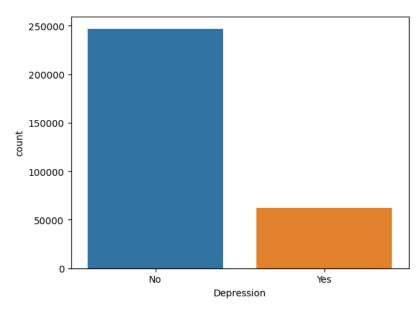


```
In [22]: sns.countplot(x=df['Other_Cancer'])
```

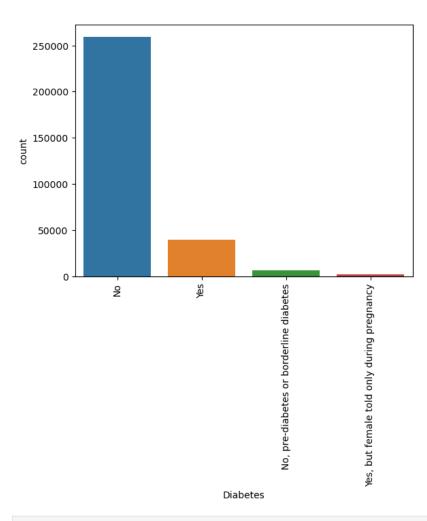
<Axes: xlabel='Other\_Cancer', ylabel='count'> Out[22]:



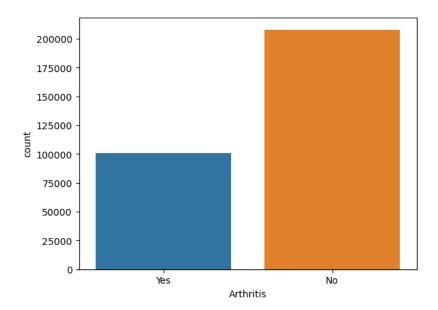
```
In [23]: sns.countplot(x=df['Depression'])
Out[23]: <Axes: xlabel='Depression', ylabel='count'>
```



Text(3, 0, 'Yes, but female told only during pregnancy')])

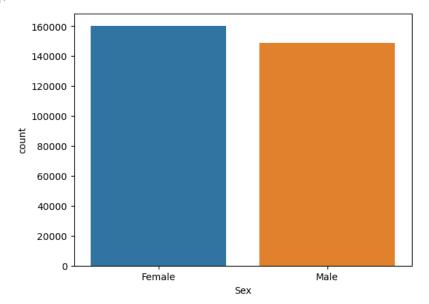


```
In [25]: sns.countplot(x=df['Arthritis'])
Out[25]: <Axes: xlabel='Arthritis', ylabel='count'>
```



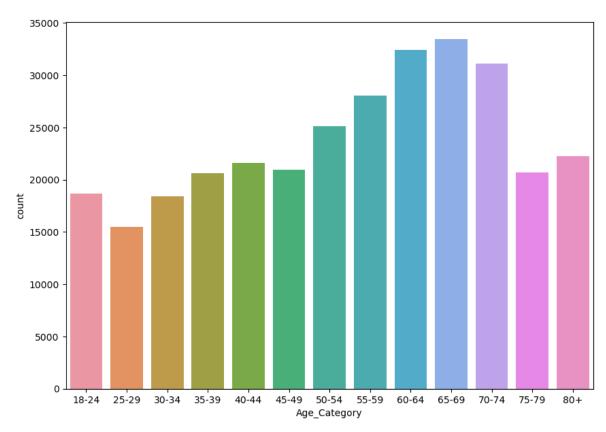
```
In [26]: sns.countplot(x=df['Sex'])
```

Out[26]: <Axes: xlabel='Sex', ylabel='count'>

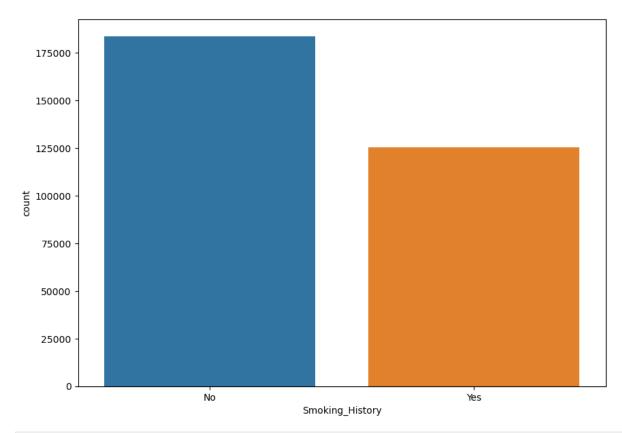


```
In [27]: plt.figure(figsize = (10,7))
sns.countplot(x=df['Age_Category'].sort_values())
```

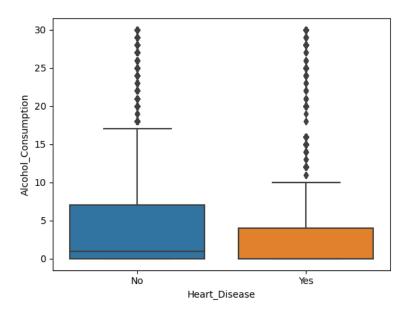
Out[27]: <Axes: xlabel='Age\_Category', ylabel='count'>



```
In [28]: plt.figure(figsize = (10,7))
sns.countplot(x=df['Smoking_History'].sort_values())
Out[28]: <Axes: xlabel='Smoking_History', ylabel='count'>
```

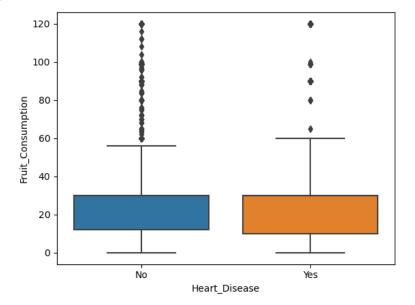


```
In [29]: sns.boxplot(x = df['Heart_Disease'], y= df ['Alcohol_Consumption'])
Out[29]: <Axes: xlabel='Heart_Disease', ylabel='Alcohol_Consumption'>
```



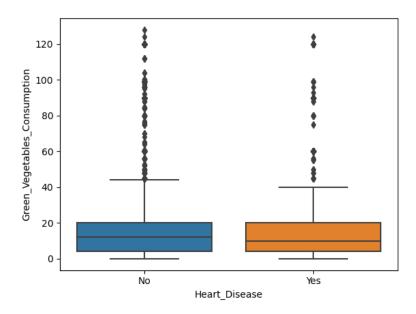
```
In [30]: sns.boxplot(x = df['Heart_Disease'], y= df ['Fruit_Consumption'])
```

Out[30]: <Axes: xlabel='Heart\_Disease', ylabel='Fruit\_Consumption'>



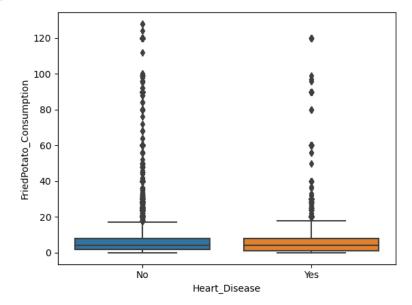
```
In [31]: sns.boxplot(x = df['Heart_Disease'], y= df ['Green_Vegetables_Consumption'])
```

Out[31]: <Axes: xlabel='Heart\_Disease', ylabel='Green\_Vegetables\_Consumption'>

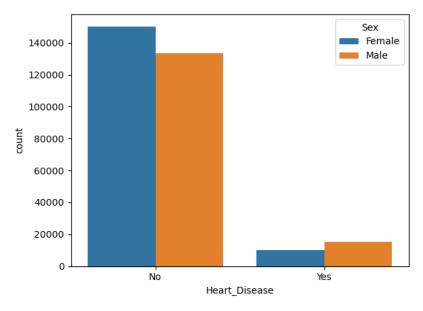


```
In [32]: sns.boxplot(x = df['Heart_Disease'], y= df ['FriedPotato_Consumption'])
```

Out[32]: <Axes: xlabel='Heart\_Disease', ylabel='FriedPotato\_Consumption'>

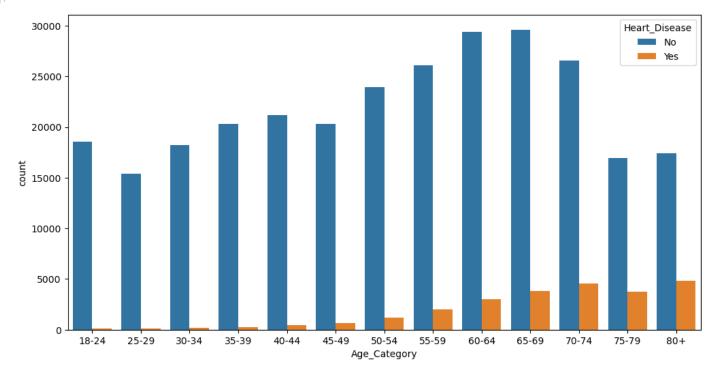


```
In [33]: sns.countplot(x = df['Heart_Disease'], hue= df ['Sex'])
Out[33]: <Axes: xlabel='Heart_Disease', ylabel='count'>
```



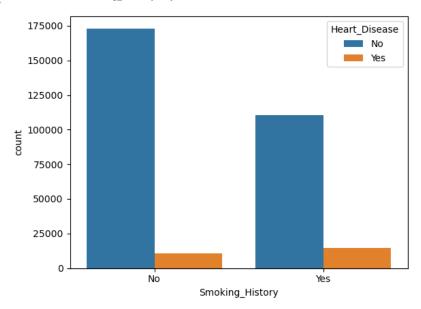
```
In [34]:
    plt.figure(figsize = (12,6))
    sns.countplot(x = df['Age_Category'].sort_values(), hue= df['Heart_Disease'])
```

Out[34]: <Axes: xlabel='Age\_Category', ylabel='count'>



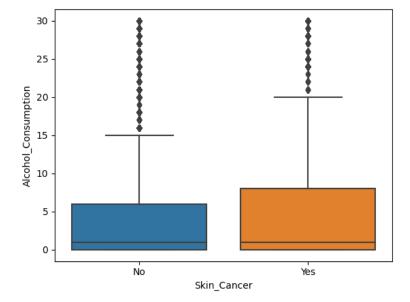
```
In [35]: sns.countplot(x = df['Smoking_History'].sort_values(), hue= df['Heart_Disease'])
```

Out[35]: <Axes: xlabel='Smoking\_History', ylabel='count'>



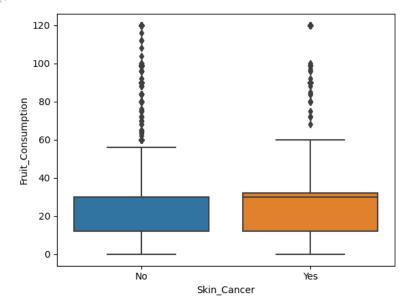
```
In [36]: sns.boxplot(x = df['Skin_Cancer'], y= df['Alcohol_Consumption'])
```

Out[36]: <Axes: xlabel='Skin\_Cancer', ylabel='Alcohol\_Consumption'>



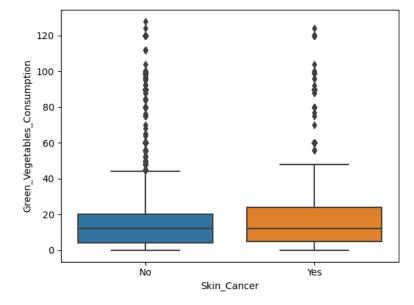
```
In [37]: sns.boxplot(x = df['Skin_Cancer'], y= df['Fruit_Consumption'])
```

```
Out[37]: <Axes: xlabel='Skin_Cancer', ylabel='Fruit_Consumption'>
```



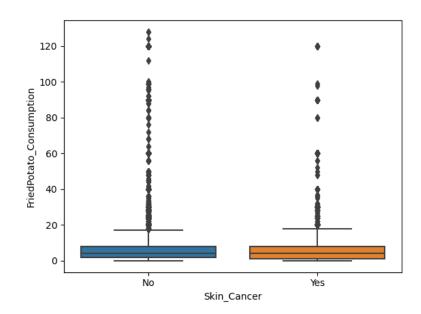
In [38]: sns.boxplot(x = df['Skin\_Cancer'], y= df['Green\_Vegetables\_Consumption'])

Out[38]: <Axes: xlabel='Skin\_Cancer', ylabel='Green\_Vegetables\_Consumption'>



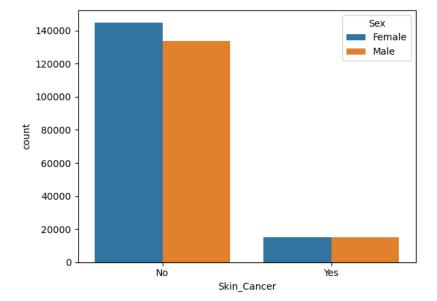
In [39]: sns.boxplot(x = df['Skin\_Cancer'], y= df['FriedPotato\_Consumption'])

Out[39]: <Axes: xlabel='Skin\_Cancer', ylabel='FriedPotato\_Consumption'>



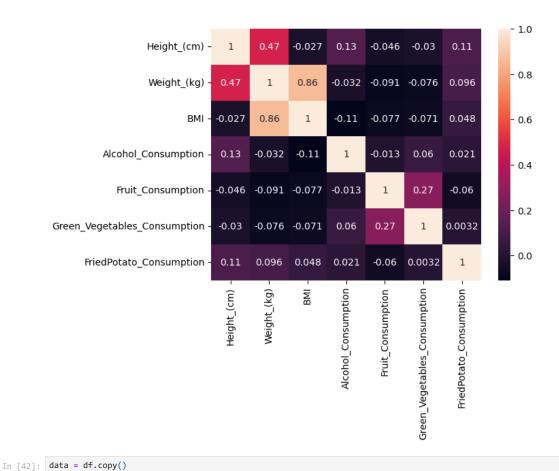
In [40]: sns.countplot(x = df['Skin\_Cancer'], hue= df['Sex'])

Out[40]: <Axes: xlabel='Skin\_Cancer', ylabel='count'>



In [41]: sns.heatmap(df.corr(method='pearson'), annot= True)

Out[41]: <Axes: >



### **Logistic Regression**

```
In [43]: from sklearn.model_selection import train_test_split
from sklearn.preprocessing import LabelEncoder
from imblearn.over_sampling import SMOTE
from sklearn.linear_model import LogisticRegression
from sklearn.tree import DecisionTreeClassifier , ExtraTreeClassifier
from sklearn.metrics import accuracy_score, classification_report, confusion_matrix, roc_auc_score

In [44]: lab = LabelEncoder()
for i in cat:
    data[i] = lab.fit_transform(data[i])

data.head()
```

```
Out[44]:
            General_Health Checkup Exercise Heart_Disease Skin_Cancer Other_Cancer Depression Diabetes Arthritis Sex Age_Category Height_(cm) Weight_(kg) BMI Smoking_History Alcohol_Consumption Fruit_Cons
         0
                       3
                               2
                                       0
                                                    0
                                                               0
                                                                           0
                                                                                      0
                                                                                              0
                                                                                                      1 0
                                                                                                                       10
                                                                                                                                150.0
                                                                                                                                           32.66 14.54
                                                                                                                                                                   1
                                                                                                                                                                                     0.0
                                                               0
                                                                           0
                                                                                                                       10
                                                                                                                                           77.11 28.29
                                                                                                                                                                                     0.0
                                                                                                      0
                                                                                                                                165.0
                                                                                                                                                                   0
                                                                                                                                           88.45 33.47
                       4
                                                    0
                                                               0
                                                                           0
                                                                                      0
                                                                                                      0
                                                                                                          0
                                                                                                                        8
                                                                                                                                163.0
                                                                                                                                                                   0
                                                                                                                                                                                     4.0
                       3
                                                               0
                                                                           0
                                                                                      0
                                                                                                      0 1
                                                                                                                       11
                                                                                                                                180.0
                                                                                                                                           93.44 28.73
                                                                                                                                                                                     0.0
                                                    0
                                                               0
                                                                           0
                                                                                      0
                                                                                                      0
                                                                                                                       12
                                                                                                                                191.0
                                                                                                                                           88.45 24.37
                                                                                                                                                                                     0.0
In [45]: x = data.drop('Heart_Disease', axis= 'columns')
         y= data['Heart_Disease']
In [46]: smo = SMOTE(random_state = 5)
In [47]: x_bal , y_bal = smo.fit_resample(x,y)
In [48]: X_train , x_test, Y_train, y_test = train_test_split(x_bal, y_bal, test_size = 0.3, random_state = 5)
In [49]: log = LogisticRegression()
In [50]: log.fit(X_train, Y_train)
Out[50]:
         ▼ LogisticRegression
         LogisticRegression()
In [51]: pred = log.predict(x_test)
In [52]: prob = log.predict_proba(x_test)
In [53]: print(confusion_matrix(y_test, pred))
          print('\n')
          print(classification_report(y_test, pred))
         print('\n')
         print(roc_auc_score(y_test, prob[:,1]))
         [[62040 23073]
          [20027 65190]]
                       precision
                                    recall f1-score support
                    0
                            0.76
                                      0.73
                                                0.74
                                                         85113
                            0.74
                                      0.76
                                                0.75
                                                         85217
                    1
                                                0.75
                                                        170330
             accuracy
                            0.75
                                      0.75
                                               0.75
                                                        170330
            macro avg
                            0.75
                                      0.75
                                               0.75
                                                       170330
         weighted avg
```

0.8230475424891888

#### **Decision Tree**

```
In [54]: dtree = DecisionTreeClassifier()
```

```
In [55]: dtree.fit(X_train,Y_train)
Out[55]: ▼ DecisionTreeClassifier
        DecisionTreeClassifier()
In [56]: tree_pred = dtree.predict(x_test)
        tree_prob = dtree.predict_proba(x_test)
In [57]: print(confusion_matrix(y_test, tree_pred))
        print('\n')
        print(classification_report(y_test, tree_pred))
        print(roc_auc_score(y_test, tree_prob[:,1]))
        [[73573 11540]
         [ 8744 76473]]
                     precision recall f1-score support
                  0
                         0.89
                               0.86 0.88
                                                  85113
                  1
                         0.87 0.90 0.88
                                                  85217
                                        0.88
            accuracy
                                                 170330
                         0.88 0.88 0.88
           macro avg
                                                 170330
        weighted avg
                         0.88 0.88 0.88 170330
        0.8809112107563303
```