HEART DISEASE

PROBLEM STATEMENT: TO PREDICT AND ANALYZE

```
In [1]: import pandas as pd
         import numpy as np
         from sklearn import preprocessing
         import matplotlib.pyplot as plt
         import seaborn as sns
         sns.set(style="white")#white background style for Seaborn plots
         sns.set(style="whitegrid",color_codes=True)
         import warnings
         warnings.simplefilter(action='ignore')
In [2]: df=pd.read_csv(r"C:\Users\manis\Downloads\framingham_heart_disease.csv")
Out[2]:
                male age
                           education currentSmoker cigsPerDay
                                                                  BPMeds prevalentStroke
             0
                   1
                       39
                                  4.0
                                                   0
                                                              0.0
                                                                       0.0
                                                                                         0
             1
                   0
                       46
                                  2.0
                                                   0
                                                              0.0
                                                                       0.0
                                                                                         0
             2
                       48
                                  1.0
                                                   1
                                                             20.0
                                                                       0.0
                                                                                         0
                   1
            3
                   0
                                  3.0
                                                             30.0
                                                                       0.0
                                                                                         0
                       61
                                                   1
                                                             23.0
                                                                                         0
             4
                   0
                       46
                                  3.0
                                                                       0.0
         4233
                   1
                       50
                                  1.0
                                                   1
                                                              1.0
                                                                       0.0
                                                                                         0
         4234
                   1
                       51
                                  3.0
                                                                       0.0
                                                                                         0
                                                             43.0
         4235
                                  2.0
                                                   1
                                                             20.0
                                                                                         0
                   0
                       48
                                                                      NaN
                                                   1
                                                             15.0
                                                                       0.0
         4236
                   0
                       44
                                  1.0
                                                                                         0
         4237
                   0
                                  2.0
                                                   0
                                                              0.0
                                                                                         0
                       52
                                                                       0.0
        4238 rows × 16 columns
```

In [3]: df.head()

Out[3]:		male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	preval			
	0	1	39	4.0	0	0.0	0.0	0				
	1	0	46	2.0	0	0.0	0.0	0				
	2	1	48	1.0	1	20.0	0.0	0				
	3	0	61	3.0	1	30.0	0.0	0				
	4	0	46	3.0	1	23.0	0.0	0				
4									•			
In [4]:	df.shape											
Out[4]:	(4238, 16)											
In [5]:	df.describe											

```
Out[5]: <bound method NDFrame.describe of
                                                   male age education currentSmoker ci
         gsPerDay BPMeds
                                                               0.0
         0
                  1
                       39
                                 4.0
                                                    0
                                                                       0.0 \
         1
                       46
                                 2.0
                                                    0
                                                              0.0
                                                                       0.0
                  0
         2
                  1
                      48
                                 1.0
                                                    1
                                                              20.0
                                                                       0.0
                  0
         3
                                 3.0
                                                    1
                                                             30.0
                                                                       0.0
                      61
         4
                  0
                      46
                                 3.0
                                                    1
                                                             23.0
                                                                       0.0
                                  . . .
                                                               . . .
                                                                       . . .
         . . .
                                                              1.0
                                                                       0.0
         4233
                  1
                      50
                                 1.0
                                                    1
         4234
                                 3.0
                                                    1
                                                             43.0
                                                                       0.0
                  1
                       51
         4235
                  0
                      48
                                 2.0
                                                    1
                                                             20.0
                                                                       NaN
                                                    1
         4236
                  0
                      44
                                 1.0
                                                             15.0
                                                                       0.0
         4237
                  0
                      52
                                 2.0
                                                    0
                                                              0.0
                                                                       0.0
                                 prevalentHyp diabetes totChol
                                                                    sysBP
               prevalentStroke
                                                                            diaBP
                                                                                      BMI
         0
                                             0
                                                        0
                                                             195.0 106.0
                                                                             70.0 26.97
         1
                              0
                                             0
                                                             250.0 121.0
                                                        0
                                                                             81.0
                                                                                    28.73
                                                                              80.0
         2
                              0
                                             0
                                                        0
                                                             245.0 127.5
                                                                                    25.34
         3
                              0
                                             1
                                                        0
                                                             225.0 150.0
                                                                             95.0 28.58
         4
                              0
                                             0
                                                        0
                                                             285.0 130.0
                                                                             84.0 23.10
         . . .
                            . . .
                                            . . .
                                                               . . .
                                                                       . . .
                                                                              . . .
                                                      . . .
         4233
                              0
                                                             313.0
                                                                    179.0
                                                                             92.0 25.97
                                             1
                                                        0
         4234
                              0
                                             0
                                                        0
                                                             207.0 126.5
                                                                             80.0
                                                                                   19.71
         4235
                              0
                                             0
                                                        0
                                                             248.0 131.0
                                                                             72.0
                                                                                   22.00
         4236
                              0
                                             0
                                                        0
                                                             210.0 126.5
                                                                             87.0 19.16
                              0
         4237
                                             0
                                                              269.0 133.5
                                                                              83.0 21.47
                                                        0
               heartRate glucose TenYearCHD
         0
                     80.0
                              77.0
                                              0
                    95.0
         1
                              76.0
                                              0
         2
                     75.0
                              70.0
                                              0
         3
                    65.0
                             103.0
                                              1
         4
                    85.0
                              85.0
                                              0
                     . . .
                               . . .
                                             . . .
         4233
                    66.0
                              86.0
                                              1
         4234
                    65.0
                              68.0
                                              0
         4235
                    84.0
                              86.0
                                              0
         4236
                     86.0
                               NaN
                                              0
         4237
                    80.0
                             107.0
         [4238 rows x 16 columns]>
```

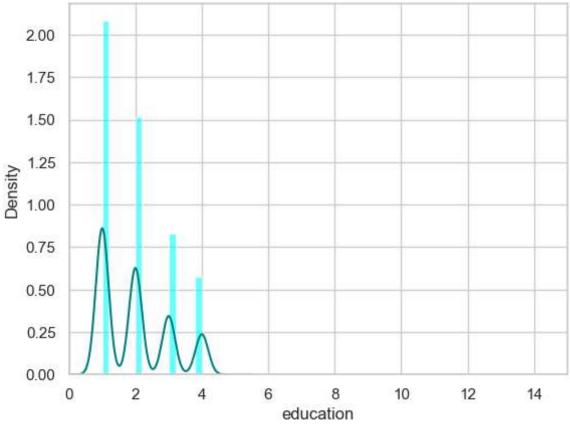
localhost:8888/nbconvert/html/python/HEARTDISEASE.ipynb?download=false

In [6]: df.info()

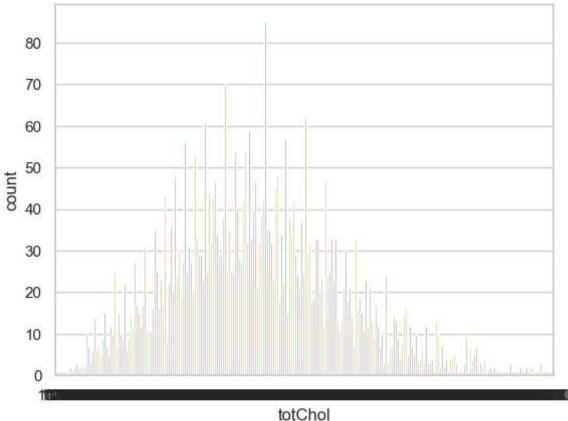
<class 'pandas.core.frame.DataFrame'> RangeIndex: 4238 entries, 0 to 4237 Data columns (total 16 columns): Column Non-Null Count Dtype ----------0 male 4238 non-null int64 int64 1 age 4238 non-null 2 4133 non-null float64 education currentSmoker 4238 non-null int64 4 cigsPerDay 4209 non-null float64 5 **BPMeds** float64 4185 non-null 6 prevalentStroke 4238 non-null int64 7 prevalentHyp 4238 non-null int64 int64 8 diabetes 4238 non-null 9 totChol 4188 non-null float64 10 sysBP 4238 non-null float64 11 diaBP 4238 non-null float64 float64 12 BMI 4219 non-null 4237 non-null float64 13 heartRate 14 glucose 3850 non-null float64 15 TenYearCHD 4238 non-null int64 dtypes: float64(9), int64(7) memory usage: 529.9 KB

TO FIND THE MISSING VALUES

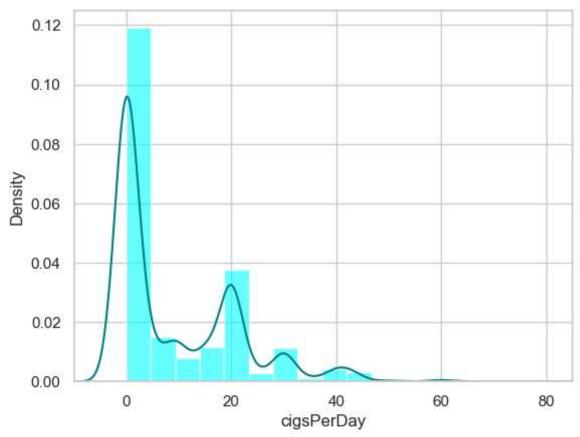
```
In [7]: df.isnull().sum()
                              0
Out[7]: male
        age
                              0
                            105
        education
        currentSmoker
                              0
                             29
        cigsPerDay
        BPMeds
                             53
        prevalentStroke
                              0
                              0
        prevalentHyp
        diabetes
                              0
        totChol
                             50
        sysBP
                              0
        diaBP
                              0
        BMI
                             19
        heartRate
                              1
                            388
        glucose
                              0
        TenYearCHD
        dtype: int64
In [8]: | ax=df["education"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.6
        df["education"].plot(kind='density',color='teal')
         ax.set(xlabel='education')
         plt.xlim(-0,15)
         plt.show()
```



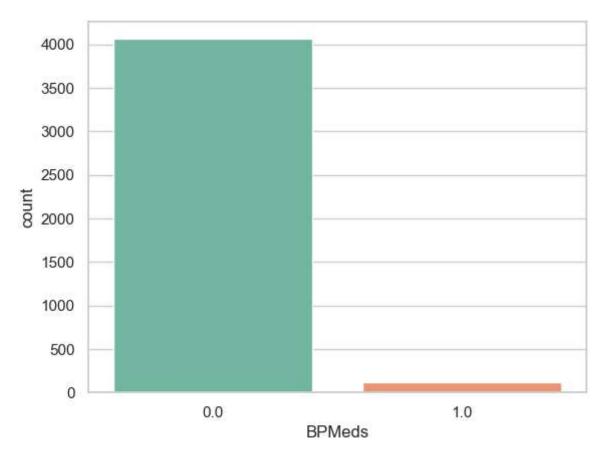
```
In [9]: print(df["education"].mean(skipna=True))
         print(df["education"].median(skipna=True))
       1.9789499153157513
       2.0
In [10]: print((df['glucose'].isnull().sum()/df.shape[0]*100))
       9.155261915998112
In [11]: print((df['totChol'].isnull().sum()/df.shape[0]*100))
       1.1798017932987257
In [12]: print(df['totChol'].value_counts())
         sns.countplot(x='totChol',data=df,palette='Set2')
         plt.show()
       totChol
       240.0
                 85
       220.0
                 70
       260.0
                 62
       210.0
                 61
       232.0
                 59
       392.0
                 1
       405.0
                 1
       359.0
                  1
       398.0
                  1
       119.0
       Name: count, Length: 248, dtype: int64
```



```
In [13]: print(df['totChol'].value_counts().idxmax())
        240.0
In [14]: data=df.copy()
         data["education"].fillna(df["education"].median(skipna=True),inplace=True)
         data["totChol"].fillna(df["totChol"].value_counts().idxmax(),inplace=True)
         data.drop('glucose',axis=1,inplace=True)
In [15]: data.isnull().sum()
Out[15]: male
                              0
                              0
         age
         education
                              0
         currentSmoker
                              0
         cigsPerDay
                             29
         BPMeds
                             53
         prevalentStroke
                              0
         prevalentHyp
                              0
         diabetes
                              0
         totChol
                              0
         sysBP
                              0
         diaBP
                              0
         BMI
                             19
         heartRate
                              1
         TenYearCHD
                              0
         dtype: int64
In [16]: ax=df["cigsPerDay"].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.
         df["cigsPerDay"].plot(kind='density',color='teal')
         ax.set(xlabel='cigsPerDay')
         plt.xlim(-10,85)
         plt.show()
```



```
In [17]: print(df["cigsPerDay"].mean(skipna=True))
         print(df["cigsPerDay"].median(skipna=True))
       9.003088619624615
       0.0
In [18]: print((df['BPMeds'].isnull().sum()/df.shape[0]*100))
       1.2505899008966492
In [19]: print((df['BMI'].isnull().sum()/df.shape[0]*100))
       0.4483246814535158
In [20]: print((df['heartRate'].isnull().sum()/df.shape[0]*100))
       0.023596035865974516
In [21]: print(df['BPMeds'].value_counts())
         sns.countplot(x='BPMeds',data=df,palette='Set2')
         plt.show()
       BPMeds
       0.0
               4061
       1.0
               124
       Name: count, dtype: int64
```



```
In [22]: print(df['heartRate'].value_counts().idxmax())
        75.0
In [23]: data=df.copy()
         data["cigsPerDay"].fillna(df["cigsPerDay"].median(skipna=True),inplace=True)
         data["BPMeds"].fillna(df["BPMeds"].value_counts().idxmax(),inplace=True)
         data["education"].fillna(df["education"].median(skipna=True),inplace=True)
         data["totChol"].fillna(df["totChol"].value_counts().idxmax(),inplace=True)
         data.drop('glucose',axis=1,inplace=True)
         data.drop('BMI',axis=1,inplace=True)
         data.drop('heartRate',axis=1,inplace=True)
In [24]: data.isnull().sum()
Out[24]: male
                             0
                             0
         age
                             0
         education
         currentSmoker
         cigsPerDay
                             0
         BPMeds
                             0
         prevalentStroke
                             0
         prevalentHyp
                             0
         diabetes
                             0
         totChol
                             0
         sysBP
                             0
         diaBP
                             0
         TenYearCHD
         dtype: int64
In [25]: data.head()
```

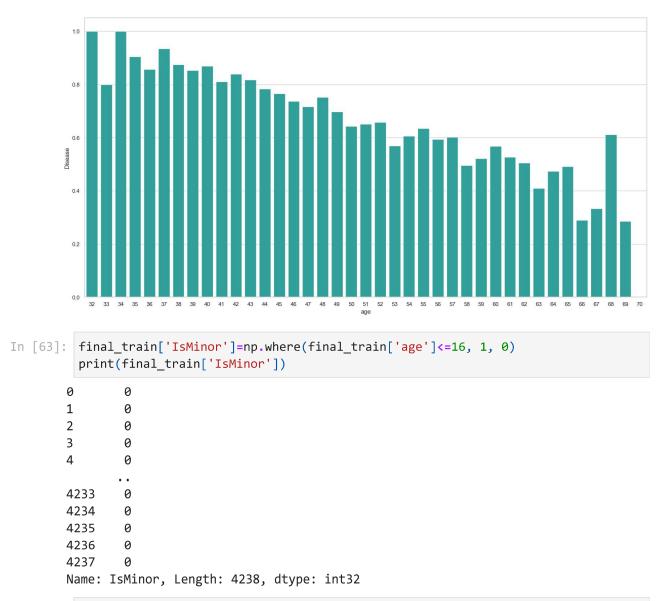
```
Out[25]:
             male age education currentSmoker cigsPerDay BPMeds prevalentStroke prevale
          0
                1
                     39
                               4.0
                                                0
                                                           0.0
                                                                    0.0
                                                                                      0
          1
                0
                     46
                               2.0
                                                           0.0
                                                                    0.0
                                                0
                                                                                      0
          2
                1
                     48
                               1.0
                                                 1
                                                          20.0
                                                                    0.0
                                                                                      0
          3
                               3.0
                                                          30.0
                                                                    0.0
                0
                     61
                                                                                      0
                0
                               3.0
                                                1
                                                          23.0
                                                                    0.0
                                                                                      0
          4
                     46
In [26]:
          plt.figure(figsize=(15,8))
          ax=df["education"].hist(bins=15,density=True,stacked=True,color='teal',alpha=0.6
          df["education"].plot(kind='density',color='teal')
          ax=data["education"].hist(bins=15,density=True,stacked=True,color='orange',alpha
          data["education"].plot(kind='density',color='orange')
          ax.legend(["education", "age"])
          ax.set(xlabel='education')
          plt.xlim(-0,10)
          plt.show()
                                                                                        education
         2.00
         1.75
         1.50
          1.25
        1.00
         0.75
         0.50
         0.25
         0.00
                                                   education
In [27]:
          data['Disease']=np.where((data["prevalentHyp"]+data["prevalentStroke"])>0,0,1)
          data.drop('prevalentHyp',axis=1,inplace=True)
          data.drop('prevalentStroke',axis=1,inplace=True)
          training=pd.get_dummies(data,columns=["currentSmoker","totChol","sysBP"])
In [28]:
          training.drop('TenYearCHD',axis=1,inplace=True)
          training.drop('male',axis=1,inplace=True)
          training.drop('diaBP',axis=1,inplace=True)
          final_train=training
          final_train.head()
```

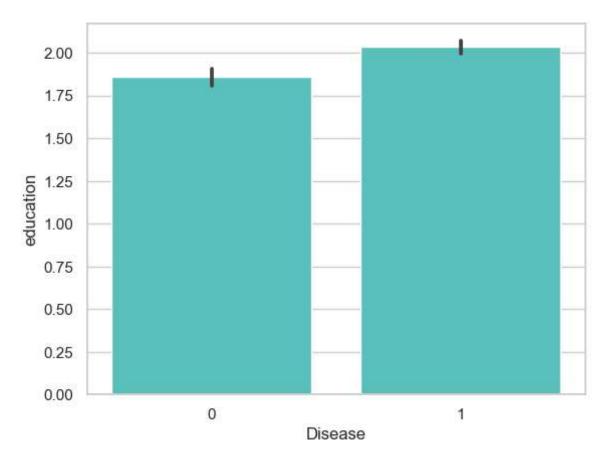
Out[28]:		age	education	cigsPerDay	BPMeds	diabetes	Disease	currentSmoker_0	currentSr
	0	39	4.0	0.0	0.0	0	1	True	
	1	46	2.0	0.0	0.0	0	1	True	
	2	48	1.0	20.0	0.0	0	1	False	
	3	61	3.0	30.0	0.0	0	0	False	
	4	46	3.0	23.0	0.0	0	1	False	

5 rows × 490 columns

EXPLORATORY DATA ANALYSIS

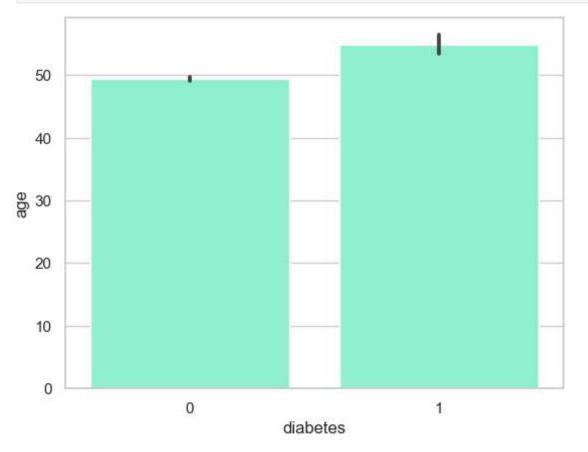
```
In [61]: plt.figure(figsize=(20,10))
    avg_survival_byage = final_train[["age", "Disease"]].groupby(['age'], as_index=F
    g = sns.barplot(x='age', y='Disease', data=avg_survival_byage, color="LightSeaGr
    plt.show()
```





import seaborn as sns
import matplotlib.pyplot as plt

Assuming 'train_df' is your DataFrame containing the data
sns.barplot(x='diabetes', y='age', data=df, color='aquamarine')
plt.show()



In []: