```
In [1]: import pandas as pd
   import numpy as np
   import statsmodels.api as sm
   import scipy.stats as st
   import matplotlib.pyplot as plt
   import seaborn as sns
   from sklearn.metrics import confusion_matrix
   import matplotlib.mlab as mlab
   %matplotlib inline
```

In [2]: df=pd.read_csv(r"C:\Users\LENOVO\Downloads\framingham.csv")
 df

Out[2]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp
0	1	39	4.0	0	0.0	0.0	0	0
1	0	46	2.0	0	0.0	0.0	0	0
2	1	48	1.0	1	20.0	0.0	0	0
3	0	61	3.0	1	30.0	0.0	0	1
4	0	46	3.0	1	23.0	0.0	0	0
4233	1	50	1.0	1	1.0	0.0	0	1
4234	1	51	3.0	1	43.0	0.0	0	0
4235	0	48	2.0	1	20.0	NaN	0	0
4236	0	44	1.0	1	15.0	0.0	0	0
4237	0	52	2.0	0	0.0	0.0	0	0

4238 rows × 16 columns

In [3]: df.head()

Out[3]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	dia
0	1	39	4.0	0	0.0	0.0	0	0	
1	0	46	2.0	0	0.0	0.0	0	0	
2	1	48	1.0	1	20.0	0.0	0	0	
3	0	61	3.0	1	30.0	0.0	0	1	
4	0	46	3.0	1	23.0	0.0	0	0	
4									•

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

```
localhost:8888/notebooks/HEARTDISEASES.ipynb
```

Out[5]: (4238, 16)

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	
1	age	4238 non-null	int64	
2	education	4133 non-null	float64	
3	currentSmoker	4238 non-null	int64	
4	cigsPerDay	4209 non-null	float64	
5	BPMeds	4185 non-null	float64	
6	prevalentStroke	4238 non-null	int64	
7	prevalentHyp	4238 non-null	int64	
8	diabetes	4238 non-null	int64	
9	totChol	4188 non-null	float64	
10	sysBP	4238 non-null	float64	
11	diaBP	4238 non-null	float64	
12	BMI	4219 non-null	float64	
13	heartRate	4237 non-null	float64	
14	glucose	3850 non-null	float64	
15	TenYearCHD	4238 non-null	int64	
d+vn	oc. float64(0) i	n+61(7)		

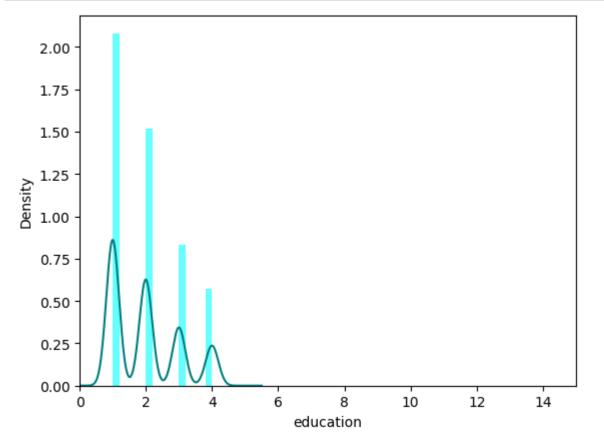
dtypes: float64(9), int64(7)
memory usage: 529.9 KB

In [7]: df.isnull().sum()

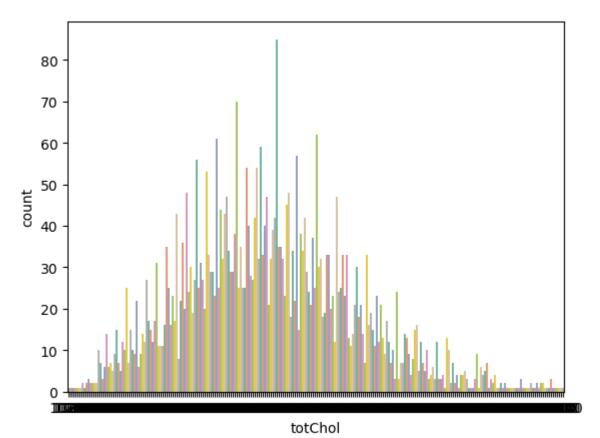
Out[7]: male

0 age 0 105 education currentSmoker 0 29 cigsPerDay **BPMeds** 53 prevalentStroke 0 prevalentHyp 0 diabetes 0 totChol 50 sysBP 0 diaBP 0 BMI 19 heartRate 1 glucose 388 TenYearCHD dtype: int64

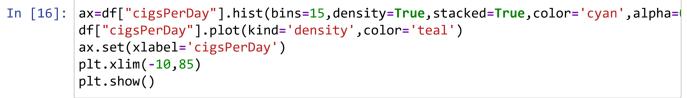
```
In [8]: a=df['education'].hist(bins=15,density=True,stacked=True,color='cyan',alpha=0.d
df['education'].plot(kind='density',color='teal')
a.set(xlabel='education')
plt.xlim(-0,15)
plt.show()
```

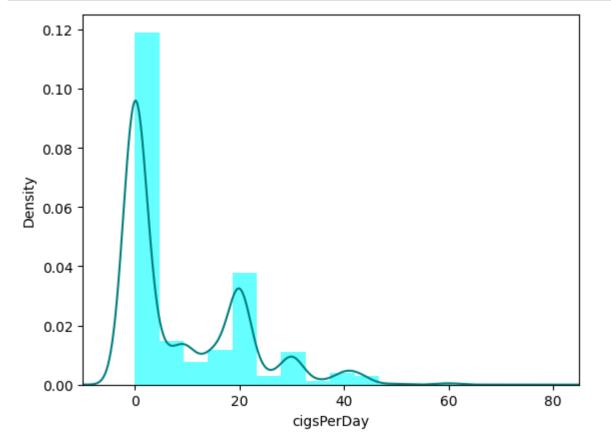


```
In [12]: print(df['totChol'].value_counts())
          sns.countplot(x='totChol',data=df,palette='Set2')
         plt.show()
         totChol
          240.0
                   85
          220.0
                   70
          260.0
                   62
          210.0
                   61
          232.0
                   59
                   . .
          392.0
                    1
         405.0
                    1
          359.0
                    1
          398.0
                    1
          119.0
         Name: count, Length: 248, dtype: int64
```



```
In [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
                               0
          TenYearCHD
          dtype: int64
```



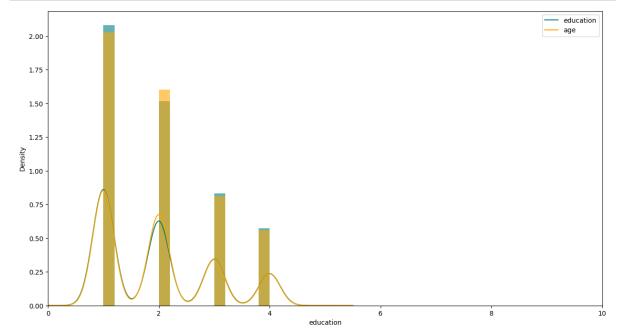


```
In [18]: print(df['cigsPerDay'].mean(skipna=True))
         print(df['cigsPerDay'].median(skipna=True))
         9.003088619624615
         0.0
In [19]:
         print((df['BPMeds'].isnull().sum()/df.shape[0]*100))
         1.2505899008966492
In [20]: print((df['BMI'].isnull().sum()/df.shape[0]*100))
         0.4483246814535158
In [21]: print((df['heartRate'].isnull().sum()/df.shape[0]*100))
         0.023596035865974516
         print(df['BPMeds'].value_counts())
In [22]:
         sns.countplot(x='BPMeds',data=df,palette='Set2')
         plt.show()
         BPMeds
         0.0
                4061
         1.0
                 124
         Name: count, dtype: int64
             4000
             3500
             3000
             2500
             2000
             1500
             1000
              500
                 0
                                  0.0
                                                                   1.0
                                                BPMeds
```

```
In [23]: print(df['heartRate'].value_counts().idxmax())
         75.0
In [24]:
         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 [25]: data.isnull().sum()
Out[25]: male
                             0
                             0
         age
                             0
         education
         currentSmoker
                             0
                             0
         cigsPerDay
                             0
         BPMeds
         prevalentStroke
                             0
                             0
         prevalentHyp
         diabetes
                             0
                             0
         totChol
                             0
         sysBP
         diaBP
                             0
         TenYearCHD
                             0
         dtype: int64
In [26]: data.head()
Out[26]:
```

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	dia
0	1	39	4.0	0	0.0	0.0	0	0	
1	0	46	2.0	0	0.0	0.0	0	0	
2	1	48	1.0	1	20.0	0.0	0	0	
3	0	61	3.0	1	30.0	0.0	0	1	
4	0	46	3.0	1	23.0	0.0	0	0	
4									•

```
In [27]: plt.figure(figsize=(15,8))
    ax=df["education"].hist(bins=15,density=True,stacked=True,color='teal',alpha=0
    df["education"].plot(kind='density',color='teal')
    ax=data["education"].hist(bins=15,density=True,stacked=True,color='orange',alpl
    data["education"].plot(kind='density',color='orange')
    ax.legend(["education","age"])
    ax.set(xlabel='education')
    plt.xlim(-0,10)
    plt.show()
```



In [29]: training=pd.get_dummies(data,columns=["currentSmoker","totChol","sysBP"])
 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[29]:

	age	education	cigsPerDay	BPMeds	diabetes	Disease	currentSmoker_0	currentSmoker_1	
0	39	4.0	0.0	0.0	0	1	True	False	
1	46	2.0	0.0	0.0	0	1	True	False	
2	48	1.0	20.0	0.0	0	1	False	True	
3	61	3.0	30.0	0.0	0	0	False	True	
4	46	3.0	23.0	0.0	0	1	False	True	

5 rows × 490 columns

localhost:8888/notebooks/HEARTDISEASES.ipynb

```
In [31]: plt.figure(figsize=(15,8))
    ax = sns.kdeplot(final_train["age"][final_train.Disease == 1],color="darkturque
    sns.kdeplot(final_train["age"][final_train.Disease == 0],color="lightcoral",sha
    plt.legend(['Disease','Died'])
    ax.set(xlabel='age')
    plt.xlim(10,100)
    plt.show()
```

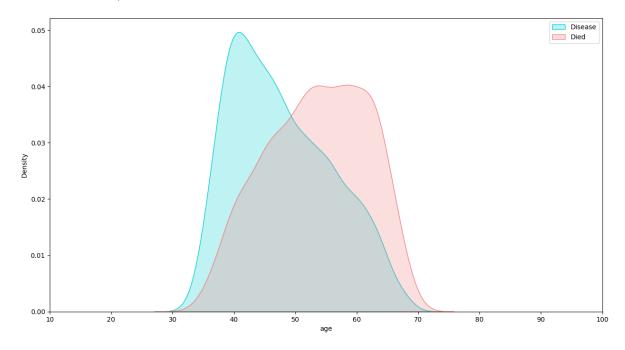
C:\Users\LENOVO\AppData\Local\Temp\ipykernel_6092\1255524709.py:2: FutureWarn
ing:

`shade` is now deprecated in favor of `fill`; setting `fill=True`. This will become an error in seaborn v0.14.0; please update your code.

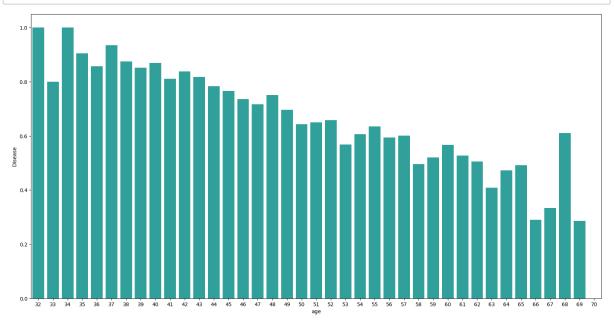
ax = sns.kdeplot(final_train["age"][final_train.Disease == 1],color="darktu
rquoise",shade=True)
C:\Users\LENOVO\AppData\Local\Temp\ipykernel_6092\1255524709.py:3: FutureWarn
ing:

`shade` is now deprecated in favor of `fill`; setting `fill=True`. This will become an error in seaborn v0.14.0; please update your code.

sns.kdeplot(final_train["age"][final_train.Disease == 0],color="lightcora
l",shade=True)



In [30]: plt.figure(figsize=(20,10))
 avg_survival_byage=final_train[["age","Disease"]].groupby(['age'],as_index=Fale
 g=sns.barplot(x='age',y='Disease',data=avg_survival_byage,color="LightSeaGreen
 plt.show()

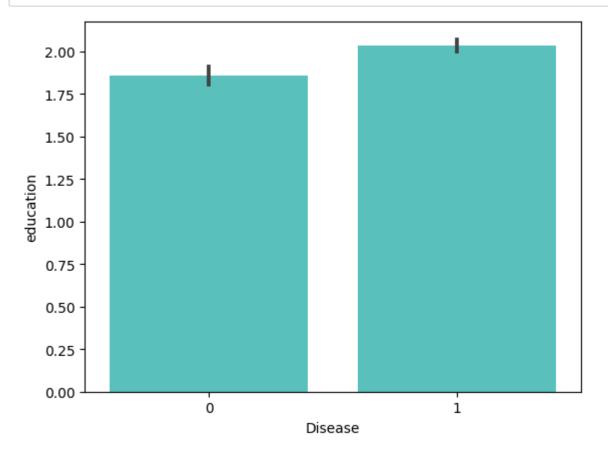


```
In [32]: final_train['IsMinor']=np.where(final_train['age']<=16,1,0)
    print(final_train['IsMinor'])</pre>
```

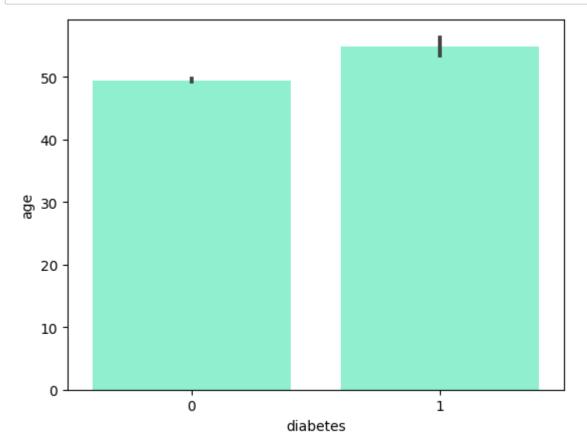
```
0
         0
1
         0
2
         0
         0
3
4
         0
4233
         0
4234
         0
4235
         0
4236
         0
4237
         0
```

Name: IsMinor, Length: 4238, dtype: int32

In [33]: sns.barplot(x='Disease',y='education',data=final_train,color="mediumturquoise"
 plt.show()



```
In [34]: import seaborn as sns
   import matplotlib.pyplot as plt
   sns.barplot(x='diabetes',y='age',data=df,color="aquamarine")
   plt.show()
```



In []: