```
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")
    sns.set(style="whitegrid",color_codes=True)
    import warnings
    warnings.simplefilter(action='ignore')
```

In [2]: df=pd.read_csv(r"C:\Users\Lenovo\OneDrive\Desktop\Data Sets\framingham.csv")
 df

Out[2]:

		male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP
	0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	106.0
	1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	121.0
	2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	127.5
	3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	150.0
	4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	130.0
4:	233	1	50	1.0	1	1.0	0.0	0	1	0	313.0	179.0
4:	234	1	51	3.0	1	43.0	0.0	0	0	0	207.0	126.5
4:	235	0	48	2.0	1	20.0	NaN	0	0	0	248.0	131.0
4:	236	0	44	1.0	1	15.0	0.0	0	0	0	210.0	126.5
42	237	0	52	2.0	0	0.0	0.0	0	0	0	269.0	133.5

4238 rows × 16 columns

In [3]: | df.head()

Out[3]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP	dial
0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	106.0	7(
1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	121.0	8.
2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	127.5	8(
3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	150.0	9!
4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	130.0	84
4 (>

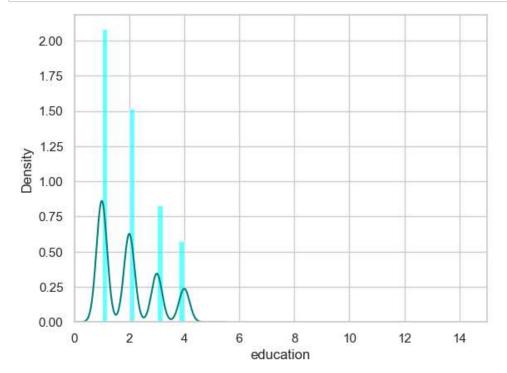
In [4]: df.tail()

Out[4]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBP
4233	1	50	1.0	1	1.0	0.0	0	1	0	313.0	179.0
4234	1	51	3.0	1	43.0	0.0	0	0	0	207.0	126.5
4235	0	48	2.0	1	20.0	NaN	0	0	0	248.0	131.0
4236	0	44	1.0	1	15.0	0.0	0	0	0	210.0	126.5
4237	0	52	2.0	0	0.0	0.0	0	0	0	269.0	133.5
4											•

```
In [7]: |df.shape
 Out[7]: (4238, 16)
 In [8]: df.describe()
 Out[8]:
                        male
                                    age
                                           education
                                                    currentSmoker
                                                                    cigsPerDay
                                                                                  BPMeds
                                                                                           prevalentStroke
                                                                                                         prevalentHyp
           count 4238.000000 4238.000000
                                                                                                                      4238
                                         4133.000000
                                                       4238.000000
                                                                   4209.000000
                                                                              4185.000000
                                                                                              4238.000000
                                                                                                          4238.000000
                    0.429212
                               49.584946
                                            1.978950
                                                          0.494101
                                                                      9.003089
                                                                                  0.029630
                                                                                                 0.005899
                                                                                                             0.310524
                                                                                                                         (
           mean
             std
                    0.495022
                                8.572160
                                            1.019791
                                                          0.500024
                                                                     11.920094
                                                                                  0.169584
                                                                                                 0.076587
                                                                                                             0.462763
                                                                                                                         (
                                                                                  0.000000
                    0.000000
                               32.000000
                                            1.000000
                                                          0.000000
                                                                      0.000000
                                                                                                 0.000000
                                                                                                             0.000000
                                                                                                                         (
            min
            25%
                    0.000000
                               42.000000
                                            1.000000
                                                          0.000000
                                                                      0.000000
                                                                                  0.000000
                                                                                                 0.000000
                                                                                                             0.000000
                                                                                                                         (
            50%
                    0.000000
                               49.000000
                                            2,000000
                                                          0.000000
                                                                      0.000000
                                                                                  0.000000
                                                                                                 0.000000
                                                                                                             0.000000
                                                                                                                         (
            75%
                    1.000000
                               56.000000
                                            3.000000
                                                          1.000000
                                                                     20.000000
                                                                                  0.000000
                                                                                                 0.000000
                                                                                                             1.000000
                                                                                                                         (
            max
                    1.000000
                               70.000000
                                            4.000000
                                                          1.000000
                                                                     70.000000
                                                                                  1.000000
                                                                                                 1.000000
                                                                                                             1.000000
 In [9]: 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
                                  4133 non-null
                                                    float64
               education
                                                    int64
           3
                                  4238 non-null
               currentSmoker
           4
                                  4209 non-null
                                                    float64
               cigsPerDay
           5
               BPMeds
                                  4185 non-null
                                                    float64
           6
               prevalentStroke 4238 non-null
                                                    int64
           7
                                  4238 non-null
                                                    int64
               prevalentHyp
           8
                                  4238 non-null
                                                    int64
               diabetes
           9
                                  4188 non-null
                                                    float64
                totChol
           10
                                  4238 non-null
                                                    float64
               sysBP
           11
               diaBP
                                  4238 non-null
                                                    float64
           12
               BMT
                                  4219 non-null
                                                    float64
           13
               heartRate
                                  4237 non-null
                                                    float64
                                  3850 non-null
                                                    float64
           14
               glucose
                                  4238 non-null
           15
               TenYearCHD
                                                    int64
          dtypes: float64(9), int64(7)
          memory usage: 529.9 KB
In [10]: df.isna().any()
Out[10]: male
                               False
          age
                               False
          education
                                True
          currentSmoker
                               False
          cigsPerDay
                                True
          BPMeds
                                True
          prevalentStroke
                               False
          prevalentHyp
                               False
          diabetes
                               False
          totChol
                                True
          sysBP
                               False
          diaBP
                               False
          BMI
                                True
          heartRate
                                True
          glucose
                                True
          TenYearCHD
                               False
          dtype: bool
```

```
In [11]: 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 [12]: print(df["education"].mean(skipna=True))
print(df["education"].median(skipna=True))
```

1.9789499153157513

2.0

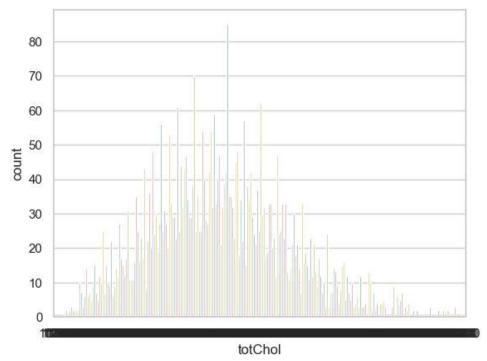
```
In [13]: print(df['glucose'].isnull().sum()/df.shape[0]*100)
```

9.155261915998112

```
In [14]: |print(df['totChol'].isnull().sum()/df.shape[0]*100)
```

1.1798017932987257

```
print(df['totChol'].value_counts())
In [15]:
         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
                    1
         Name: count, Length: 248, dtype: int64
```



```
In [18]: data.isna().sum()
Out[18]: 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 [19]: pd.set_option('display.max_rows',4238)
         pd.set_option('display.max_columns',16)
In [20]: pd.set option('display.width',50)
In [21]: print('This DataFrame has %d Rows and %d Columns'%(df.shape))
         This DataFrame has 4238 Rows and 16 Columns
In [22]: features matrix=df.iloc[:,0:15]
         target_vector=df.iloc[:,-2]
In [23]: print('The Features Matrix Has %d Rows And %d Column(s)'%(features_matrix.shape))
         print('The Target Matrix Has %d Rows And %d Column(s)'%(np.array(target_vector).reshape(-1,1).shape)
         The Features Matrix Has 4238 Rows And 15 Column(s)
         The Target Matrix Has 4238 Rows And 1 Column(s)
In [24]: |df['education'].mean()
Out[24]: 1.9789499153157513
In [25]: df['cigsPerDay'].mean()
Out[25]: 9.003088619624615
In [28]: |df['heartRate'].median()
Out[28]: 75.0
In [29]: df['BPMeds'].mean()
Out[29]: 0.02962962962963
```

```
In [30]: df["glucose"].fillna(df["glucose"].median(skipna=True),inplace=True)
df
```

Out[30]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHyp	diabetes	totChol	sysBf
0	1	39	4.0	0	0.0	0.0	0	0	0	195.0	106.(
1	0	46	2.0	0	0.0	0.0	0	0	0	250.0	121.(
2	1	48	1.0	1	20.0	0.0	0	0	0	245.0	127.
3	0	61	3.0	1	30.0	0.0	0	1	0	225.0	150.0
4	0	46	3.0	1	23.0	0.0	0	0	0	285.0	130.0
5	0	43	2.0	0	0.0	0.0	0	1	0	228.0	180.0
6	0	63	1.0	0	0.0	0.0	0	0	0	205.0	138.(
7	0	45	2.0	1	20.0	0.0	0	0	0	313.0	100.0
8	1	52	1.0	0	0.0	0.0	0	1	0	260.0	141.
9	1	43	1.0	1	30.0	0.0	0	1	0	225.0	162.0
10	0	50	1.0	0	0.0	0.0	0	0	0	254.0	133.(

```
In [31]: df.isnull().sum()
```

```
Out[31]: male
                                0
                                0
          age
         education
                              105
         currentSmoker
                                0
         cigsPerDay
                               29
         BPMeds
                               53
         prevalentStroke
                                0
         prevalentHyp
                                0
         diabetes
                                0
         totChol
                               50
         sysBP
                                0
         diaBP
                                0
         BMI
                               19
         {\tt heartRate}
                                1
         glucose
                                0
         TenYearCHD
                                0
         dtype: int64
```

```
In [32]: df['education'].fillna(df['education'].median(skipna=True),inplace=True)
    df['totChol'].fillna(df['totChol'].median(skipna=True),inplace=True)
    df['BMI'].fillna(df['BMI'].median(skipna=True),inplace=True)
    df['heartRate'].fillna(df['heartRate'].median(skipna=True),inplace=True)
    df['BPMeds'].fillna(df['BPMeds'].median(skipna=True),inplace=True)
    df['cigsPerDay'].fillna(df['cigsPerDay'].median(skipna=True),inplace=True)
```

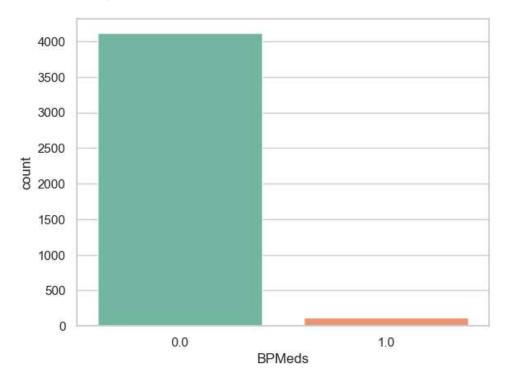
```
In [33]: df.isna().any()
Out[33]: male
                             False
                             False
         age
         education
                             False
         currentSmoker
                             False
         cigsPerDay
                             False
         BPMeds
                             False
         prevalentStroke
                             False
                             False
         prevalentHyp
         diabetes
                             False
         totChol
                             False
         sysBP
                             False
         diaBP
                             False
         BMI
                             False
         heartRate
                             False
                             False
         glucose
                             False
         TenYearCHD
         dtype: bool
In [34]: | df.drop('glucose',axis=1,inplace=True)
In [35]: df.isna().any()
Out[35]: male
                             False
         age
                             False
         education
                             False
         currentSmoker
                             False
         cigsPerDay
                             False
         BPMeds
                             False
         prevalentStroke
                             False
         prevalentHyp
                             False
         diabetes
                             False
         totChol
                             False
         sysBP
                             False
         diaBP
                             False
         BMI
                             False
         heartRate
                             False
         TenYearCHD
                             False
         dtype: bool
In [36]: print(df["cigsPerDay"].mean(skipna=True))
         print(df["cigsPerDay"].median(skipna=True))
         print((df['BPMeds'].isnull().sum()/df.shape[0]*100))
         print((df['BMI'].isnull().sum()/df.shape[0]*100))
         print((df['heartRate'].isnull().sum()/df.shape[0]*100))
         8.941481831052384
         0.0
         0.0
         0.0
         0.0
```

```
print(df['BPMeds'].value_counts())
sns.countplot(x='BPMeds',data=df,palette='Set2')
In [37]:
              plt.show()
```

BPMeds

0.0 4114 1.0 124

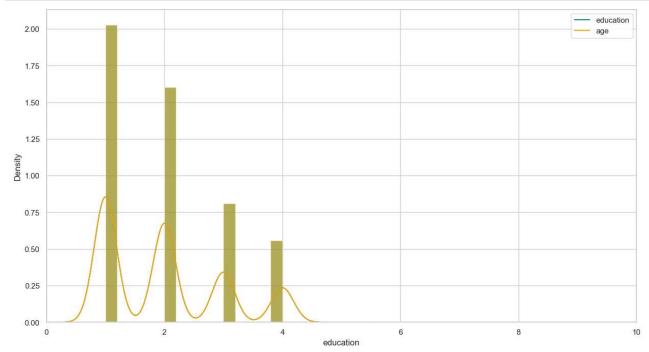
Name: count, dtype: int64



```
In [38]: print(df['heartRate'].value_counts().idxmax())
```

75.0

```
In [39]: 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=0.5)
    data["education"].plot(kind='density',color='orange')
    ax.legend(["education","age"])
    ax.set(xlabel='education')
    plt.xlim(-0,10)
    plt.show()
```



```
In [40]: 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)
```

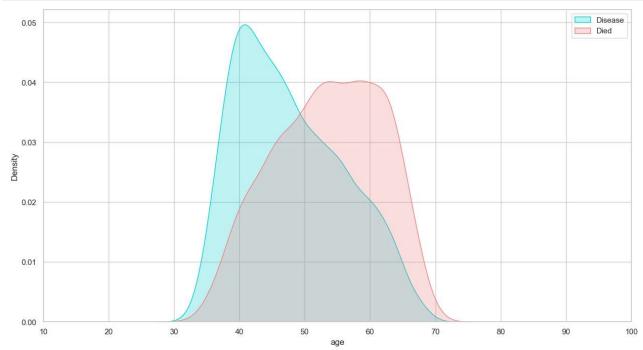
```
In [41]: training=pd.get_dummies(data,columns=["currentSmoker","totChol","sysBP"])
    training.drop('TenYearCHD',axis=1,inplace=True)
    training.drop('male',axis=1,inplace=True)
    training.drop('diaBP',axis=1,inplace=True)
    final_train=training
    final_train.head()
```

Out[41]:

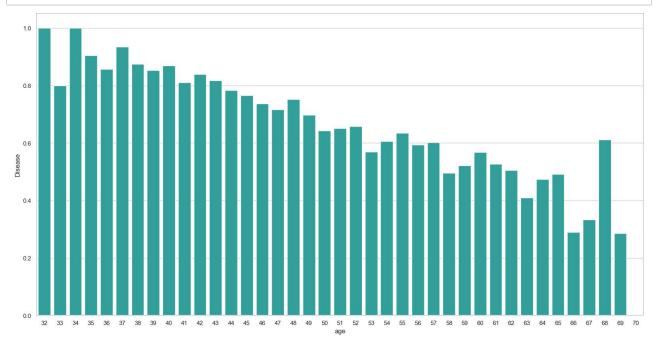
	age	education	cigsPerDay	BPMeds	diabetes	вмі	heartRate	Disease	 sysBP_220.0	sysBP_230.0	sysBP_232.0	s
0	39	4.0	0.0	0.0	0	26.97	80.0	1	 False	False	False	_
1	46	2.0	0.0	0.0	0	28.73	95.0	1	 False	False	False	
2	48	1.0	20.0	0.0	0	25.34	75.0	1	 False	False	False	
3	61	3.0	30.0	0.0	0	28.58	65.0	0	 False	False	False	
4	46	3.0	23.0	0.0	0	23.10	85.0	1	 False	False	False	

5 rows × 492 columns

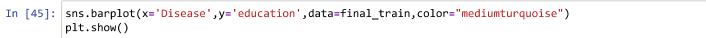
```
In [42]: plt.figure(figsize=(15,8))
    ax = sns.kdeplot(final_train["age"][final_train.Disease == 1],color="darkturquoise",shade=True)
    sns.kdeplot(final_train["age"][final_train.Disease == 0],color="lightcoral",shade=True)
    plt.legend(['Disease','Died'])
    ax.set(xlabel='age')
    plt.xlim(10,100)
    plt.show()
```

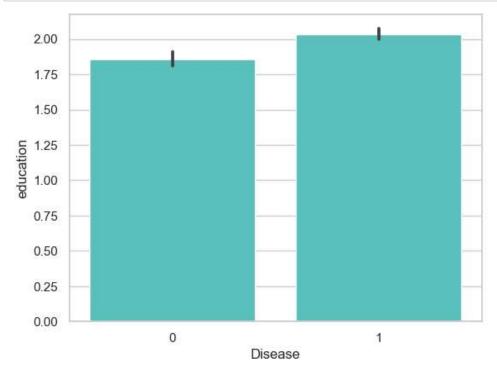


In [43]: plt.figure(figsize=(20,10))
 avg_survival_byage=final_train[["age","Disease"]].groupby(['age'],as_index=False).mean()
 g=sns.barplot(x='age',y='Disease',data=avg_survival_byage,color="LightSeaGreen")
 plt.show()



```
In [44]: | final_train['IsMinor']=np.where(final_train['age']<=16,1,0)</pre>
          print(final_train['IsMinor'])
          0
                   0
          1
                   0
          2
                   0
                   0
          3
                   0
                   0
                   0
                   0
                   0
                   0
          10
                   0
          11
                   0
          12
                   0
          13
                   0
          14
                   0
          15
                   0
          16
                   0
          17
                   0
                   0
          18
```





```
In [46]: import seaborn as sns
    import matplotlib.pyplot as plt
    sns.barplot(x='diabetes',y='age',data=df,color="aquamarine")
    plt.show()
```

