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\DELL E5490\Downloads\heart disease.csv")
df
```

Out[2]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalen
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	
4233	1	50	1.0	1	1.0	0.0	0	
4234	1	51	3.0	1	43.0	0.0	0	
4235	0	48	2.0	1	20.0	NaN	0	
4236	0	44	1.0	1	15.0	0.0	0	
4237	0	52	2.0	0	0.0	0.0	0	

4238 rows × 16 columns

→

In [3]:

df.head()

Out[3]:

	male	age	education	currentSmoker	cigsPerDay	BPMeds	prevalentStroke	prevalentHy _l
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 bpmeds<="" cigsperday="" er="" method="" ndframe.describe="" of="" th=""><th>le</th><th>age</th><th>educ</th><th>ation</th><th>current</th><th>:Smok</th></bound>								le	age	educ	ation	current	:Smok
er c 0	ıgsperu 1	4.0)		0			0.0	0.0	\			
1	1 39 0 46						0			0.0	0.0		
2	1	48		2.0 1.0			1			0.0	0.0		
3	0	61		3.0)		1		3	0.0	0.0		
4	0	46		3.0)		1		2	3.0	0.0		
							•••						
4233	1	50		1.0			1 1			1.0	0.0		
4234 4235	1 0	51 48		3.0 2.0			1			3.0 0.0	0.0 NaN		
4236	0	44		1.0			1			5.0	0.0		
4237	0	52		2.0			0			0.0	0.0		
MI	preval	entS	troke	pre	valer	ntHyp	diabe	tes	tot	Chol	sysBP	diaBP	В
97 \			0			0		0	1	95.0	106.0	70.0	26.
1 73			0			0		0	2	50.0	121.0	81.0	28.
2 34			0			0		0	2	45.0	127.5	80.0	25.
3 58			0			1		0	2	25.0	150.0	95.0	28.
4 10			0			0		0	2	85.0	130.0	84.0	23.
• • •			• • •			• • •		• • •		• • •	• • •	• • •	
4233 97			0			1		0	3	13.0	179.0	92.0	25.
4234			0			0		0	2	07.0	126.5	80.0	19.
71 4235			0			0		0	2	48.0	131.0	72.0	22.
00 4236			0			0		0	2	10.0	126.5	87.0	19.
16 4237			0			0		0	2	69.0	133.5	83.0	21.
47													
	heartR	ate	gluco	se	TenYe	earCHD							
0		0.0		.0		0							
1		5.0		.0		0							
2		5.0		.0		0							
3		5.0	103			1							
4	8	5.0		.0		0							
4233	6	6.0		.0		1							
4234		5.0		.0		0							
4235		4.0		.0		0							
4236		6.0		laN		0							
4237		0.0	107			0							

[4238 rows x 16 columns]>

```
In [7]:
```

```
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

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

TO FIND MISSING VALUES

In [8]:

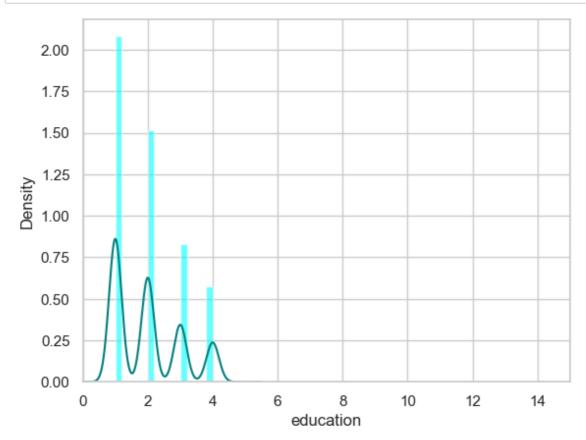
```
df.isnull().sum()
```

Out[8]:

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

In [12]:

```
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 [13]:

```
print(df["education"].mean(skipna=True))
print(df["education"].median(skipna=True))
```

1.9789499153157513

2.0

In [14]:

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

9.155261915998112

In [15]:

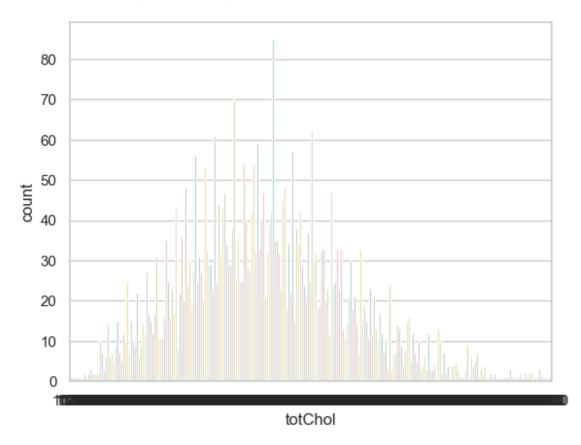
```
print(df['totChol'].isnull().sum()/df.shape[0]*100)
```

In [16]:

```
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 1

Name: count, Length: 248, dtype: int64



In [17]:

```
print(df['totChol'].value_counts().idxmax())
```

```
In [18]:
```

```
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 [19]:

```
data.isnull().sum()
```

Out[19]:

0 male 0 age 0 education 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 [20]:

```
pd.set_option('display.max_rows',4238)
pd.set_option('display.max_columns',16)
```

In [21]:

```
pd.set_option('display.width',50)
```

In [22]:

```
print('This DataFrame has %d Rows and %d Columns'%(df.shape))
```

This DataFrame has 4238 Rows and 16 Columns

In [23]:

```
features_matrix=df.iloc[:,0:15]
```

```
In [24]:
```

```
target_vector=df.iloc[:,-2]
```

In [25]:

```
print('The Features Matrix Has %d Rows And %d Column(s)'%(features_matrix.shape))
```

The Features Matrix Has 4238 Rows And 15 Column(s)

In [28]:

```
df["glucose"].fillna(df["glucose"].median(skipna=True),inplace=True)
df
  385
           1
                39
                           2.0
                                              0
                                                         0.0
                                                                    0.0
                                                                                       0
                                                                                                       0
  386
                           2.0
                                              0
                                                         0.0
                                                                    0.0
                                                                                       0
  387
           0
                55
                           3.0
                                              0
                                                         0.0
                                                                    0.0
                                                                                       0
                                                                                                       1
                           2.0
                                              0
  388
           0
                39
                                                         0.0
                                                                    0.0
                                                                                       0
                                                                                                       0
  389
           0
                39
                           2.0
                                              1
                                                        20.0
                                                                    0.0
                                                                                       0
                                                                                                       0
  390
           1
                51
                           1.0
                                              1
                                                        20.0
                                                                    0.0
                                                                                       0
                                                                                                       0
  391
           0
                54
                           1.0
                                              0
                                                         0.0
                                                                    0.0
                                                                                                       0
                                                                                       0
                                              0
  392
           0
                48
                           3.0
                                                         0.0
                                                                    0.0
                                                                                       0
                                                                                                       0
                                                        20.0
  393
                                                                    0.0
                                                                                                       0
           1
                51
                           1.0
                                              1
                                                                                       0
                           2.0
                                                                    0.0
  394
           0
                65
                                              0
                                                         0.0
                                                                                       0
                                                                                                       1
                           2.0
                                              0
  395
           0
                65
                                                         0.0
                                                                  NaN
                                                                                       0
                                                                                                       1
  396
                39
                           3.0
                                              0
                                                         0.0
                                                                    0.0
                                                                                       0
                                                                                                       0
           1
```

In [29]:

```
print(df["cigsPerDay"].mean(skipna=True))
print(df["cigsPerDay"].median(skipna=True))
```

9.003088619624615

0.0

In [30]:

```
print((df['BPMeds'].isnull().sum()/df.shape[0]*100))
```

```
In [31]:
```

```
print((df['BMI'].isnull().sum()/df.shape[0]*100))
```

0.4483246814535158

In [32]:

```
print((df['heartRate'].isnull().sum()/df.shape[0]*100))
```

0.023596035865974516

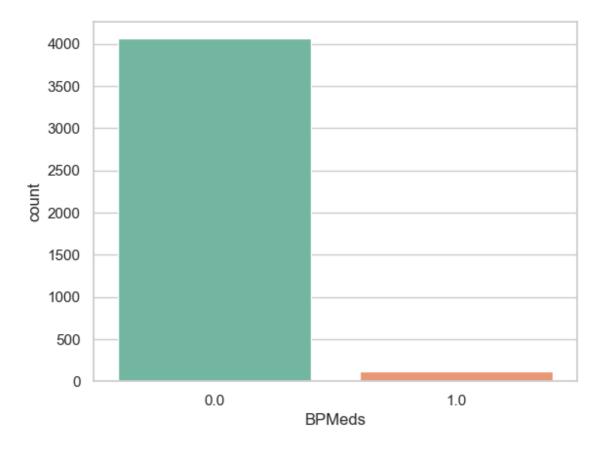
In [33]:

```
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

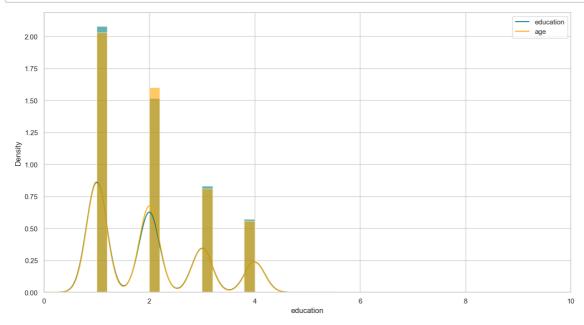


In [34]:

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

In [36]:

```
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=0.6)
data["education"].plot(kind='density',color='orange')
ax.legend(["education","age"])
ax.set(xlabel='education')
plt.xlim(-0,10)
plt.show()
```



In [37]:

```
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 [38]:

```
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[38]:

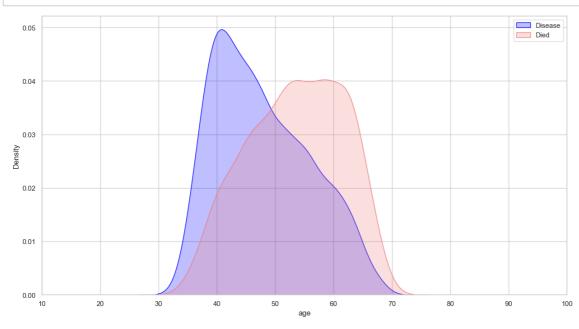
	age	education	cigsPerDay	BPMeds	diabetes	BMI	heartRate	Disease	 sysBP_220
0	39	4.0	0.0	0.0	0	26.97	80.0	1	 Fal
1	46	2.0	0.0	0.0	0	28.73	95.0	1	 Fal
2	48	1.0	20.0	0.0	0	25.34	75.0	1	 Fal
3	61	3.0	30.0	0.0	0	28.58	65.0	0	 Fal
4	46	3.0	23.0	0.0	0	23.10	85.0	1	 Fal

5 rows × 492 columns

EXPLORATORY DATA ANALYSIS

In [46]:

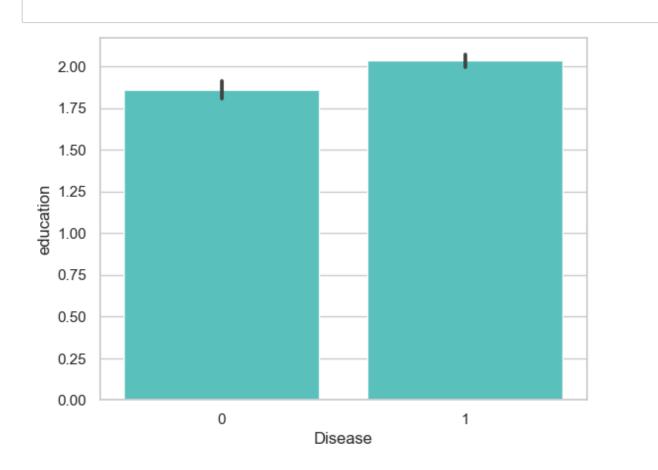
```
plt.figure(figsize=(15,8))
ax = sns.kdeplot(final_train["age"][final_train.Disease == 1],color="blue",shade="Black'
sns.kdeplot(final_train["age"][final_train.Disease == 0],color="lightcoral",shade="black
plt.legend(['Disease','Died'])
ax.set(xlabel='age')
plt.xlim(10,100)
plt.show()
```



In [51]:

```
final_train['IsMinor']=np.where(final_train['age']<=16,1,0)
print(final_train['IsMinor'])</pre>
           0
85
           0
86
           0
87
88
           0
89
           0
90
           0
91
           0
92
93
           0
94
           0
95
           0
96
           0
97
           0
98
99
100
101
           0
102
           0
           0
103
101
```

In [53]:



In [55]:

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

