```
In [1]:
import numpy as np
import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import warnings
warnings.filterwarnings('ignore')

In [2]:
data = pd.read_csv("heart_desease.csv")

In [3]:
data.head()
```

#### Out[3]:

	HeartDisease	BMI	Smoking	AlcoholDrinking	Stroke	PhysicalHealth	MentalHealth	<b>DiffWalk</b> i
0	No	16.60	Yes	No	No	3	30	
1	No	20.34	No	No	Yes	0	0	
2	No	26.58	Yes	No	No	20	30	
3	No	24.21	No	No	No	0	0	
4	No	23.71	No	No	No	28	0	Y
4								•

In [4]:
data.tail()

### Out[4]:

	HeartDisease	BMI	Smoking	AlcoholDrinking	Stroke	PhysicalHealth	MentalHealth	Dif
319790	Yes	27.41	Yes	No	No	7	0	
319791	No	29.84	Yes	No	No	0	0	
319792	No	24.24	No	No	No	0	0	
319793	No	32.81	No	No	No	0	0	
319794	No	46.56	No	No	No	0	0	
4								•

In [5]:

data.shape

### Out[5]:

(319795, 18)

In [6]:

```
data.columns
```

## Out[6]:

```
у',
    'Race', 'Diabetic', 'PhysicalActivity', 'GenHealth', 'SleepTime',
    'Asthma', 'KidneyDisease', 'SkinCancer'],
   dtype='object')
```

In [7]: H

```
data.info()
```

<class 'pandas.core.frame.DataFrame'> RangeIndex: 319795 entries, 0 to 319794 Data columns (total 18 columns):

#	Column	Non-Nu	Dtype	
0	HeartDisease	319795	non-null	object
1	BMI	319795	non-null	float64
2	Smoking	319795	non-null	object
3	AlcoholDrinking	319795	non-null	object
4	Stroke	319795	non-null	object
5	PhysicalHealth	319795	non-null	int64
6	MentalHealth	319795	non-null	int64
7	DiffWalking	319795	non-null	object
8	Sex	319795	non-null	object
9	AgeCategory	319795	non-null	object
10	Race	319795	non-null	object
11	Diabetic	319795	non-null	object
12	PhysicalActivity	319795	non-null	object
13	GenHealth	319795	non-null	object
14	SleepTime	319795	non-null	int64
15	Asthma	319795	non-null	object
16	KidneyDisease	319795	non-null	object
17	SkinCancer	319795	non-null	object
dtype	es: float64(1), in	t64(3),	object(14)	)

dtypes: float64(1), int64(3), object(14)

memory usage: 43.9+ MB

In [8]: ▶

```
data.describe()
```

## Out[8]:

	ВМІ	PhysicalHealth	MentalHealth	SleepTime
count	319795.000000	319795.00000	319795.000000	319795.000000
mean	28.325399	3.37171	3.898366	7.097075
std	6.356100	7.95085	7.955235	1.436007
min	12.020000	0.00000	0.000000	1.000000
25%	24.030000	0.00000	0.000000	6.000000
50%	27.340000	0.00000	0.000000	7.000000
75%	31.420000	2.00000	3.000000	8.000000
max	94.850000	30.00000	30.000000	24.000000

In [9]: ▶

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

### Out[9]:

HeartDisease	0
BMI	0
Smoking	0
AlcoholDrinking	0
Stroke	0
PhysicalHealth	0
MentalHealth	0
DiffWalking	0
Sex	0
AgeCategory	0
Race	0
Diabetic	0
PhysicalActivity	0
GenHealth	0
SleepTime	0
Asthma	0
KidneyDisease	0
SkinCancer	0
dtype: int64	

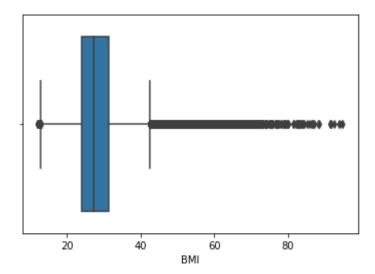
In [11]:

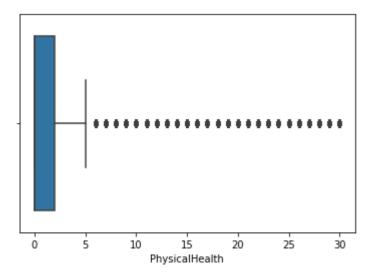
In [12]: ▶

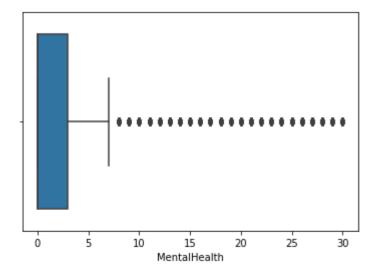
data\_num = data[['BMI', 'PhysicalHealth', 'MentalHealth', 'SleepTime']]

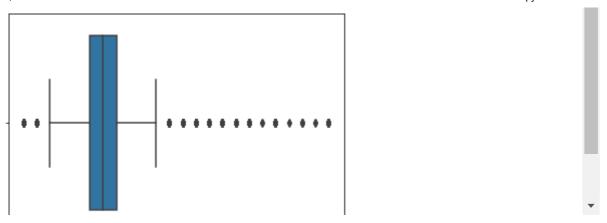
In [13]: ▶

```
for i in data_num.columns:
    sns.boxplot(x=data_num[i])
    plt.show()
plt.show()
```



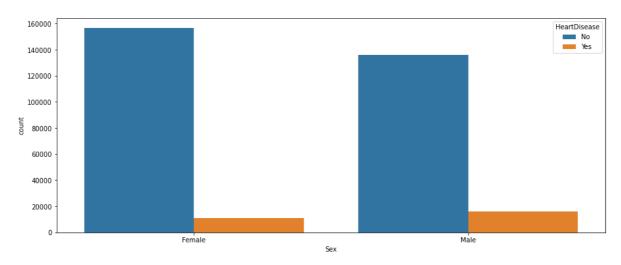






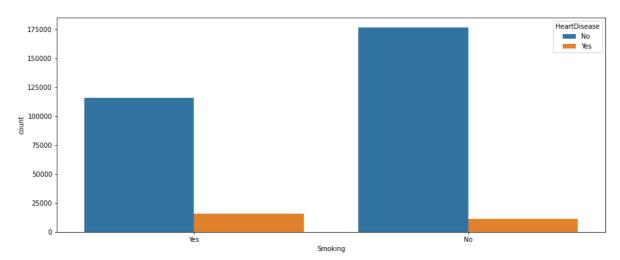
# In [27]: ▶

```
plt.figure(figsize=(15,6))
sns.countplot('Sex',hue='HeartDisease', data = data)
plt.xticks(rotation = 0)
plt.show()
```



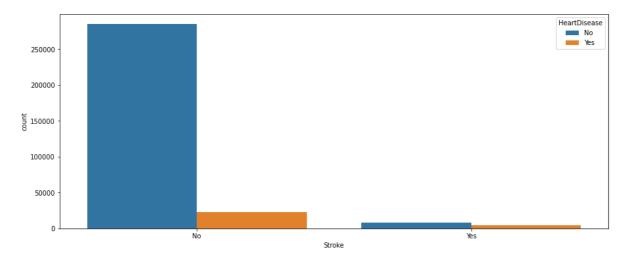
### In [26]:

```
plt.figure(figsize=(15,6))
sns.countplot('Smoking',hue='HeartDisease',data=data)
plt.xticks(rotation = 0)
plt.show()
```



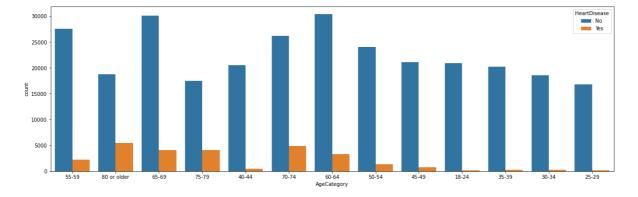
## In [25]: ▶

```
plt.figure(figsize=(15,6))
sns.countplot('Stroke',hue='HeartDisease',data=data)
plt.xticks(rotation = 0)
plt.show()
```



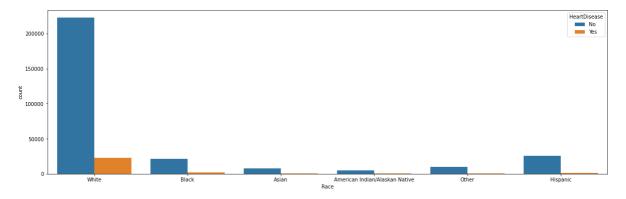
```
In [23]: ▶
```

```
plt.figure(figsize=(20,6))
sns.countplot('AgeCategory',hue='HeartDisease',data=data)
plt.xticks(rotation = 0)
plt.show()
```



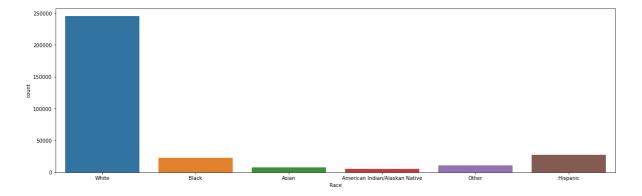
```
In [22]:
```

```
plt.figure(figsize=(20,6))
sns.countplot('Race',hue='HeartDisease',data=data)
plt.xticks(rotation = 0)
plt.show()
```



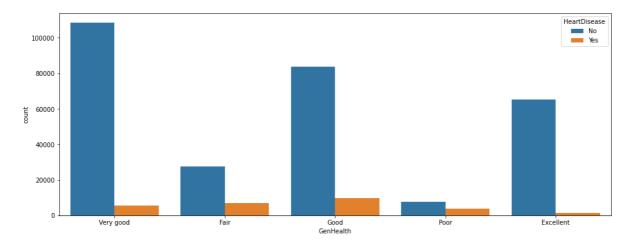
### In [28]:

```
plt.figure(figsize=(20,6))
sns.countplot('Race',data=data)
plt.xticks(rotation = 0)
plt.show()
```



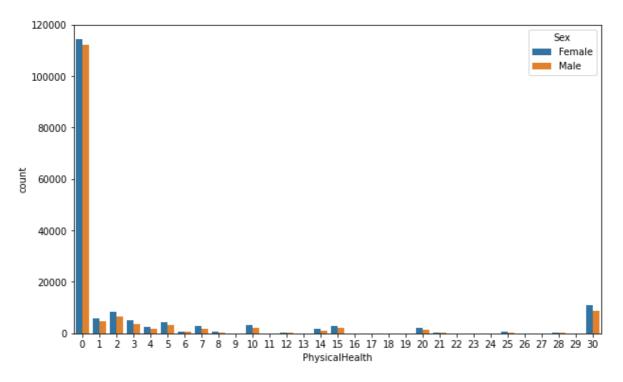
#### In [29]:

```
plt.figure(figsize=(16,6))
sns.countplot('GenHealth', hue='HeartDisease', data=data)
plt.xticks(rotation = 0)
plt.show()
```



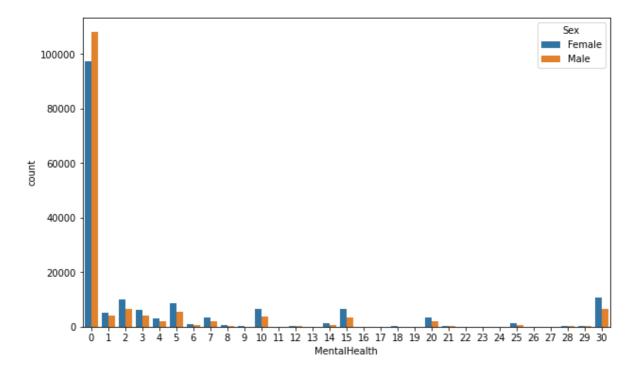
```
In [34]:
```

```
plt.figure(figsize=(10,6))
sns.countplot('PhysicalHealth',hue='Sex',data=data)
plt.xticks(rotation = 0)
plt.show()
```



```
In [35]:
```

```
plt.figure(figsize=(10,6))
sns.countplot('MentalHealth',hue='Sex',data=data)
plt.xticks(rotation = 0)
plt.show()
```



```
In [37]:
from sklearn.preprocessing import LabelEncoder
In [38]:
```

for i in data\_cat.columns:
 le=LabelEncoder()
 label=le.fit\_transform(data\_cat[i])
 data\_cat[i]=label

In [39]: ▶

data\_cat.head()

### Out[39]:

	HeartDisease	Smoking	AlcoholDrinking	Stroke	DiffWalking	Sex	AgeCategory	Race	Diabeti
0	0	1	0	0	0	0	7	5	
1	0	0	0	1	0	0	12	5	
2	0	1	0	0	0	1	9	5	
3	0	0	0	0	0	0	11	5	
4	0	0	0	0	1	0	4	5	
4									•

In [40]:

data1=pd.concat([data\_cat,data\_num],axis=1)

In [41]: ▶

data1.head()

#### Out[41]:

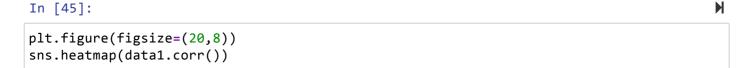
	HeartDisease	Smoking	AlcoholDrinking	Stroke	DiffWalking	Sex	AgeCategory	Race	Diabeti
0	0	1	0	0	0	0	7	5	
1	0	0	0	1	0	0	12	5	
2	0	1	0	0	0	1	9	5	
3	0	0	0	0	0	0	11	5	
4	0	0	0	0	1	0	4	5	
4									<b>&gt;</b>

In [42]: ▶

data1.corr()

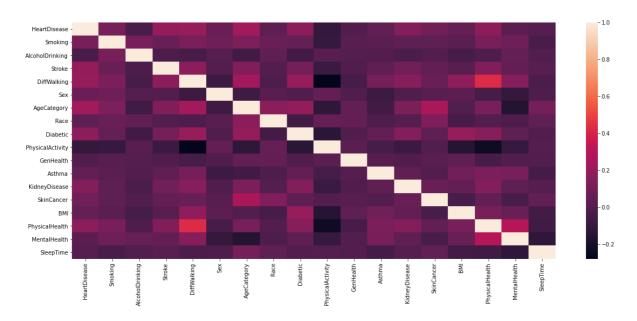
## Out[42]:

	HeartDisease	Smoking	AlcoholDrinking	Stroke	DiffWalking	Sex	Age
HeartDisease	1.000000	0.107764	-0.032080	0.196835	0.201258	0.070040	(
Smoking	0.107764	1.000000	0.111768	0.061226	0.120074	0.085052	(
AlcoholDrinking	-0.032080	0.111768	1.000000	-0.019858	-0.035328	0.004200	-(
Stroke	0.196835	0.061226	-0.019858	1.000000	0.174143	-0.003091	(
DiffWalking	0.201258	0.120074	-0.035328	0.174143	1.000000	-0.068860	(
Sex	0.070040	0.085052	0.004200	-0.003091	-0.068860	1.000000	-(
AgeCategory	0.233432	0.128331	-0.059528	0.137822	0.243263	-0.067478	
Race	0.034854	0.065499	0.036702	-0.003956	-0.015831	0.018855	(
Diabetic	0.168553	0.053847	-0.057372	0.101518	0.205502	-0.013456	(
PhysicalActivity	-0.100030	-0.097174	0.017487	-0.079455	-0.278524	0.048247	-(
GenHealth	-0.011062	0.020625	0.001629	-0.009335	-0.043552	-0.010283	(
Asthma	0.041444	0.024149	-0.002202	0.038866	0.103222	-0.069191	-(
KidneyDisease	0.145197	0.034920	-0.028280	0.091167	0.153064	-0.009084	(
SkinCancer	0.093317	0.033977	-0.005702	0.048116	0.064840	0.013434	(
ВМІ	0.051803	0.023118	-0.038816	0.019733	0.181678	0.026940	-(
PhysicalHealth	0.170721	0.115352	-0.017254	0.137014	0.428373	-0.040904	(
MentalHealth	0.028591	0.085157	0.051282	0.046467	0.152235	-0.100058	-(
SleepTime	0.008327	-0.030336	-0.005065	0.011900	-0.022216	-0.015704	(
4							•



#### Out[45]:

#### <AxesSubplot:>



```
In [46]:
```

```
data1.drop(['Race','BMI'],axis=1,inplace=True)
```

```
In [47]:

X=data1.iloc[:.1:]
```

```
X=data1.iloc[:,1:]
y=data1.iloc[:,0]
```

```
In [48]: ▶
```

X.shape

#### Out[48]:

(319795, 15)

In [49]:

y.shape

#### Out[49]:

(319795,)

In [50]:

```
from sklearn.model_selection import train_test_split
x_train,x_test,y_train,y_test=train_test_split(X,y,test_size=0.20)
```

```
In [51]:

from sklearn.linear_model import LogisticRegression

In [52]:

model=LogisticRegression()
model.fit(x_train,y_train)

Out[52]:
LogisticRegression()

In [54]:

y_pred = model.predict(x_test)

In [55]:

print("Training Accuracy:", model.score(x_train, y_train))
print("Testing Accuracy:", model.score(x_test, y_test))
```

Training Accuracy: 0.9146758079394612 Testing Accuracy: 0.916055598117544