

```
In [2]: import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
%matplotlib inline
```

```
In [9]: df = pd.read_csv("Heart.csv")
```

```
In [10]: df
```

...

```
In [11]: df.info()
```

```
<class 'pandas.core.frame.DataFrame'>
RangeIndex: 1025 entries, 0 to 1024
Data columns (total 14 columns):
#   Column      Non-Null Count  Dtype
---  -
0   age         1025 non-null   int64
1   sex         1025 non-null   int64
2   cp          1025 non-null   int64
3   trestbps    1025 non-null   int64
4   chol        1025 non-null   int64
5   fbs         1025 non-null   int64
6   restecg     1025 non-null   int64
7   thalach     1025 non-null   int64
8   exang       1025 non-null   int64
9   oldpeak     1025 non-null   float64
10  slope       1025 non-null   int64
11  ca          1025 non-null   int64
12  thal        1025 non-null   int64
13  target      1025 non-null   int64
dtypes: float64(1), int64(13)
memory usage: 112.2 KB
```

```
In [12]: df['sex'] = df['sex'].astype('object')
df['cp'] = df['cp'].astype('object')
df['fbs'] = df['fbs'].astype('object')
df['restecg'] = df['restecg'].astype('object')
df['exang'] = df['exang'].astype('object')
df['slope'] = df['slope'].astype('object')
df['ca'] = df['ca'].astype('object')
df['thal'] = df['thal'].astype('object')
```

```
In [13]: df.dtypes
```

...

```
In [14]: df.isnull().sum()
```

...

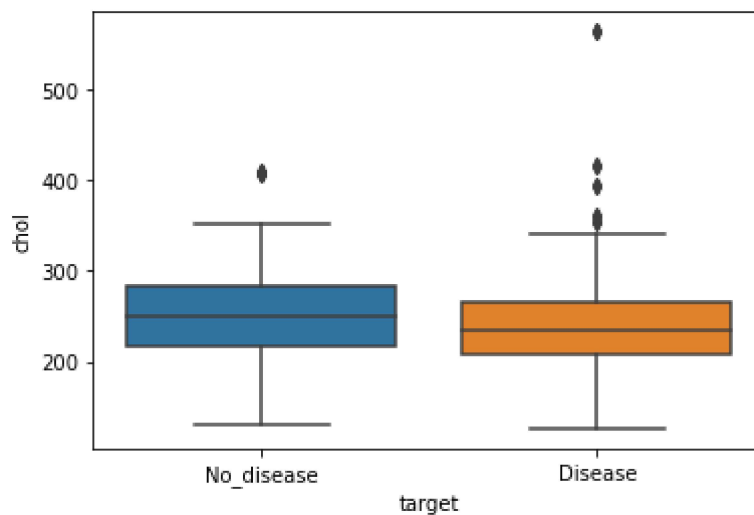
```
In [15]: df['target'] = df.target.replace({1: "Disease", 0: "No_disease"})
df['sex'] = df.sex.replace({1: "Male", 0: "Female"})
df['cp'] = df.cp.replace({0: "typical_angina", 1: "atypical_angina", 2: "non-anginal"})
df['exang'] = df.exang.replace({1: "Yes", 0: "No"})
df['fbs'] = df.fbs.replace({1: "True", 0: "False"})
df['slope'] = df.slope.replace({0: "upsloping", 1: "flat", 2: "downsloping"})
df['thal'] = df.thal.replace({1: "fixed_defect", 2: "reversable_defect", 3: "normal"})
```

```
In [16]: bxplt = sns.boxplot(df['target'], df['chol'])
plt.show
```

C:\ProgramData\Anaconda3\lib\site-packages\seaborn_decorators.py:36: FutureWarning: Pass the following variables as keyword args: x, y. From version 0.12, the only valid positional argument will be `data`, and passing other arguments without an explicit keyword will result in an error or misinterpretation.

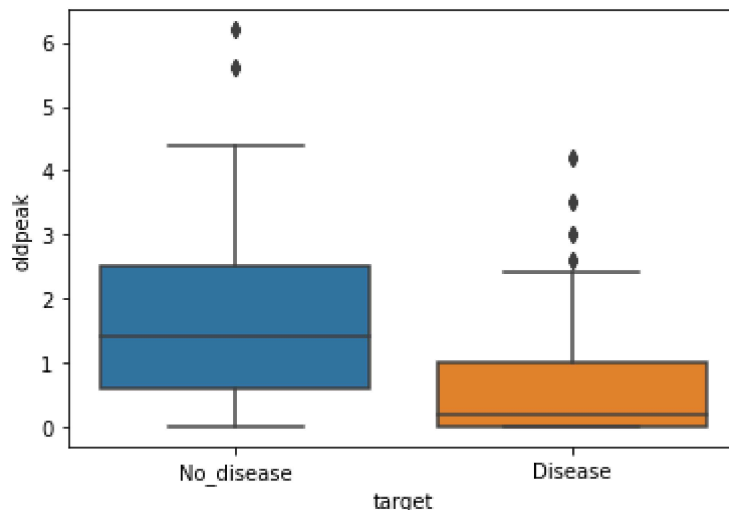
warnings.warn(

```
Out[16]: <function matplotlib.pyplot.show(close=None, block=None)>
```



```
In [19]: sns.boxplot(x="target", y="oldpeak", data=df)
```

```
Out[19]: <AxesSubplot:xlabel='target', ylabel='oldpeak'>
```



In [20]:

In [23]: `continous_features = ['age', 'trestbps', 'chol', 'thalach', 'oldpeak']`

In [24]:

Out[24]: `ModeResult(mode=array(['age']), dtype='<U8'), count=array([1]))`

```
In [26]: def outliers(df_out, drop = False):
    for each_feature in df_out.columns:
        feature_data = df_out[each_feature]
        Q1 = np.percentile(feature_data, 25.)
        Q3 = np.percentile(feature_data, 75.)
        IQR = Q3-Q1
        outlier_step = IQR * 1.5
        outliers = feature_data[~((feature_data >= Q1 - outlier_step) & (feature_

        if not drop:
            print('For the feature {}, No of Outliers is {}'.format(each_feature,
        if drop:
            df.drop(outliers, inplace = True, errors = 'ignore')
            print('Outliers from {} feature removed'.format(each_feature))
    outliers(df[continous_features])
```

For the feature age, No of Outliers is 0
For the feature trestbps, No of Outliers is 30
For the feature chol, No of Outliers is 16
For the feature thalach, No of Outliers is 4
For the feature oldpeak, No of Outliers is 7

In [27]: `outliers(df[continous_features], drop=True)`

Outliers from age feature removed
Outliers from trestbps feature removed
Outliers from chol feature removed
Outliers from thalach feature removed
Outliers from oldpeak feature removed

In [28]: `duplicated=df.duplicated().sum()`

In [31]: `duplicated`

Out[31]: 683

```
In [32]: if duplicated:
    print("Duplicated rows :{}".format(duplicated))
else:
    print("No duplicates")
```

Duplicated rows :683

```
In [33]: duplicates=df[df.duplicated(keep=False)]
duplicates.head()
```

Out[33]:

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope
0	52	Male	typical_angina	125	212	False	1	168	No	1.0	downsloping
1	53	Male	typical_angina	140	203	True	0	155	Yes	3.1	upsloping
2	70	Male	typical_angina	145	174	False	1	125	Yes	2.6	upsloping
3	61	Male	typical_angina	148	203	False	1	161	No	0.0	downsloping
4	62	Female	typical_angina	138	294	True	1	106	No	1.9	flat

```
In [34]: df.drop_duplicates()
```

Out[34]:

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slo
0	52	Male	typical_angina	125	212	False	1	168	No	1.0	downslop
1	53	Male	typical_angina	140	203	True	0	155	Yes	3.1	upslop
2	70	Male	typical_angina	145	174	False	1	125	Yes	2.6	upslop
3	61	Male	typical_angina	148	203	False	1	161	No	0.0	downslop
4	62	Female	typical_angina	138	294	True	1	106	No	1.9	
...	
723	68	Female	non-anginal pain	120	211	False	0	115	No	1.5	
733	44	Female	non-anginal pain	108	141	False	1	175	No	0.6	
739	52	Male	typical_angina	128	255	False	1	161	Yes	0.0	downslop
843	59	Male	asymtomatic	160	273	False	0	125	No	0.0	downslop
878	54	Male	typical_angina	120	188	False	1	113	No	1.4	

285 rows × 14 columns

```
In [35]: duplicates.head()
```

```
Out[35]:
```

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope
0	52	Male	typical_angina	125	212	False	1	168	No	1.0	downsloping
1	53	Male	typical_angina	140	203	True	0	155	Yes	3.1	upsloping
2	70	Male	typical_angina	145	174	False	1	125	Yes	2.6	upsloping
3	61	Male	typical_angina	148	203	False	1	161	No	0.0	downsloping
4	62	Female	typical_angina	138	294	True	1	106	No	1.9	flat

```
In [36]: df.drop_duplicates()
```

```
Out[36]:
```

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slo
0	52	Male	typical_angina	125	212	False	1	168	No	1.0	downslop
1	53	Male	typical_angina	140	203	True	0	155	Yes	3.1	upslop
2	70	Male	typical_angina	145	174	False	1	125	Yes	2.6	upslop
3	61	Male	typical_angina	148	203	False	1	161	No	0.0	downslop
4	62	Female	typical_angina	138	294	True	1	106	No	1.9	
...	
723	68	Female	non-anginal pain	120	211	False	0	115	No	1.5	
733	44	Female	non-anginal pain	108	141	False	1	175	No	0.6	
739	52	Male	typical_angina	128	255	False	1	161	Yes	0.0	downslop
843	59	Male	asymtomatic	160	273	False	0	125	No	0.0	downslop
878	54	Male	typical_angina	120	188	False	1	113	No	1.4	

285 rows × 14 columns

```
In [37]: duplicated=df.duplicated().sum()
```

```
In [38]: duplicated
```

```
Out[38]: 683
```

```
In [39]: df1 = pd.read_csv("student.csv", header = 0)
df2 = pd.read_csv("mark.csv", header = 0)
```

```
In [40]: df1.info()
```

...

```
In [41]: df2.info()
```

...

```
In [42]: df1.head()
```

...

```
In [43]: df2.head()
```

...

```
In [44]: df = pd.merge(df1, df2, on = 'Student_id')
df.head(10)
```

...

```
In [46]: df = pd.read_csv("Heart.csv")
df
```

...

```
In [47]: ddf =pd.read_csv("data.csv",encoding='cp1252')
```

```
C:\ProgramData\Anaconda3\lib\site-packages\IPython\core\interactiveshell.py:344: DtypeWarning: Columns (0) have mixed types.Specify dtype option on import or set low_memory=False.
exec(code_obj, self.user_global_ns, self.user_ns)
```

```
In [80]: ddf.head()
```

Out[80]:

	state	location	type	so2	no2	rspm	spm	pm2_5	date
0	0	Hyderabad	Residential, Rural and other Areas	4.8	17.4	108.833091	220.78348	40.791467	1990-02-01
1	0	Hyderabad	Industrial Area	3.1	7.0	108.833091	220.78348	40.791467	1990-02-01
2	0	Hyderabad	Residential, Rural and other Areas	6.2	28.5	108.833091	220.78348	40.791467	1990-02-01
3	0	Hyderabad	Residential, Rural and other Areas	6.3	14.7	108.833091	220.78348	40.791467	1990-03-01
4	0	Hyderabad	Industrial Area	4.7	7.5	108.833091	220.78348	40.791467	1990-03-01

```
In [50]: ddf=ddf.drop(['stn_code','agency','sampling_date','location_monitoring_station'],
```

```
In [51]: ddf=ddf.dropna(subset=['date'])
```

```
In [52]: COLS = ['so2', 'no2', 'rspm', 'spm', 'pm2_5']
from sklearn.impute import SimpleImputer
imputer = SimpleImputer(missing_values=np.nan, strategy='mean')
ddf[COLS] = imputer.fit_transform(ddf[COLS])
```

```
In [53]: ddf[COLS]
```

...

```
In [54]: ddf['type'].value_counts()
```

```
Out[54]: Residential, Rural and other Areas    179013
Industrial Area                               96089
Residential and others                        86791
Industrial Areas                             51747
Sensitive Area                               8979
Sensitive Areas                              5536
RIRUO                                         1304
Sensitive                                     495
Industrial                                   233
Residential                                  158
Name: type, dtype: int64
```

```
In [55]: ddf['type']
```

```
Out[55]: 0      Residential, Rural and other Areas
1      Industrial Area
2      Residential, Rural and other Areas
3      Residential, Rural and other Areas
4      Industrial Area
...
435734      RIRUO
435735      RIRUO
435736      RIRUO
435737      RIRUO
435738      RIRUO
Name: type, Length: 435735, dtype: object
```

```
In [56]: ddf['state'].value_counts()
```

...

```
In [57]: from sklearn.preprocessing import LabelEncoder
labelencoder=LabelEncoder()
ddf["state"]=labelencoder.fit_transform(ddf["state"])
ddf.head(5)
```

```
Out[57]:
```

	state	location	type	so2	no2	rspm	spm	pm2_5	date
0	0	Hyderabad	Residential, Rural and other Areas	4.8	17.4	108.833091	220.78348	40.791467	1990-02-01
1	0	Hyderabad	Industrial Area	3.1	7.0	108.833091	220.78348	40.791467	1990-02-01
2	0	Hyderabad	Residential, Rural and other Areas	6.2	28.5	108.833091	220.78348	40.791467	1990-02-01
3	0	Hyderabad	Residential, Rural and other Areas	6.3	14.7	108.833091	220.78348	40.791467	1990-03-01
4	0	Hyderabad	Industrial Area	4.7	7.5	108.833091	220.78348	40.791467	1990-03-01

```
In [58]: dfAndhra=ddf[(ddf['state']==0)]
dfAndhra
```

...

```
In [59]: dfAndhra['location'].value_counts()
```

...

```
In [60]: dfAndhra=ddf[(ddf['state']==0)]
dfAndhra
```

...

```
In [61]: dfAndhra['location'].value_counts()
```

```
Out[61]: Hyderabad      7764
Visakhapatnam    7108
Vijayawada      2093
Chittoor        1003
Tirupati        986
Kurnool         857
Patancheru      698
Guntur          629
Nalgonda        618
Ramagundam      554
Nellore         408
Khammam         385
Warangal        336
Ananthapur      324
Ongole          317
Kadapa          316
Srikakulam      315
Rajahmundry     311
Eluru           300
Visakhapatnam   207
```

```
In [62]: from sklearn.preprocessing import OneHotEncoder
onehotencoder=OneHotEncoder(sparse=False,handle_unknown='error',drop='first')
pd.DataFrame(onehotencoder.fit_transform(dfAndhra[["location"]]))
```

...

```
In [63]: df['ca'].unique()
```

```
Out[63]: array([2, 0, 1, 3, 4], dtype=int64)
```

```
In [64]: df[df['ca']==4]
```

...

```
In [65]: df.loc[df['ca']==4,'ca']=np.NaN
```



```
In [66]: df['thal'].nunique()
```

```
Out[66]: 4
```

```
In [67]: df['thal'].unique()
```

```
Out[67]: array([3, 2, 1, 0], dtype=int64)
```

```
In [68]: df[df['thal']==3]
```

...

```
In [69]: df.loc[df['thal']==3, 'thal']=np.NaN
```

```
In [70]: df.isna().sum()
```

...

```
In [71]: df = df.fillna(df.median())
df.isnull().sum()
```

...

```
In [72]: X = df.drop('target', axis=1)
```

```
In [73]: X.head()
```

```
Out[73]:
```

	age	sex	cp	trestbps	chol	fbs	restecg	thalach	exang	oldpeak	slope	ca	thal
0	52	1	0	125	212	0	1	168	0	1.0	2	2.0	2.0
1	53	1	0	140	203	1	0	155	1	3.1	0	0.0	2.0
2	70	1	0	145	174	0	1	125	1	2.6	0	0.0	2.0
3	61	1	0	148	203	0	1	161	0	0.0	2	1.0	2.0
4	62	0	0	138	294	1	1	106	0	1.9	1	3.0	2.0

```
In [74]: X.shape
```

```
Out[74]: (1025, 13)
```

```
In [75]: y = df['target']
y.head(10)
```

...

```
In [76]: from sklearn import preprocessing
df=df.apply(preprocessing.LabelEncoder().fit_transform)
```

```
In [77]: from sklearn.model_selection import train_test_split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size=0.2, random_st
```

```
In [78]: print("X_train : ",X_train.shape)
print("X_test : ",X_test.shape)
print("y_train : ",y_train.shape)
print("y_test : ",y_test.shape)
```

```
X_train : (820, 13)
X_test : (205, 13)
y_train : (820,)
y_test : (205,)
```

```
In [79]: from sklearn.tree import DecisionTreeClassifier
from sklearn import metrics
clf = DecisionTreeClassifier()
clf = clf.fit(X_train, y_train)
y_pred = clf.predict(X_test)
print(y_pred)
```

```
[1 0 0 1 0 0 0 0 0 0 0 0 0 1 0 1 1 1 0 1 0 1 1 1 1 1 1 1 1 0 1 1 1 0 0 1 0 0 0
 0 1 1 0 1 1 1 1 1 1 0 0 0 0 0 1 0 0 0 0 1 0 0 0 0 1 0 0 0 0 1 1 1 1 1 1 1 0
 0 0 0 1 0 0 0 0 1 0 1 0 1 0 1 0 0 0 1 1 1 0 0 0 1 1 1 1 1 0 0 1 0 1 0 1 1
 0 1 0 1 0 1 0 0 1 0 1 1 0 0 1 1 1 1 0 0 0 1 1 1 0 0 1 0 1 1 0 1 0 0 1 1 1
 1 0 1 1 1 0 0 0 1 1 0 0 0 1 1 0 1 1 1 1 1 1 1 1 1 1 1 1 1 0 1 1 0 0 1 0 0 1
 0 1 0 1 0 0 1 0 1 0 1 1 1 0 0 1 1 1 0 1]
```

```
In [ ]:
```