# exploratory\_data\_analysis\_exercise

#### September 24, 2021

#### 0.1 Exercise:

Download Haberman Cancer Survival dataset from Kaggle. You may have to create a Kaggle account Perform a similar alanlaysis as above on this dataset with the following sections:

High level statistics of the dataset: number of points, numer of features, number of classes, Explain our objective.

Perform Univaraite analysis (PDF, CDF, Boxplot, Voilin plots) to understand which features are Perform Bi-variate analysis (scatter plots, pair-plots) to see if combinations of features are Write your observations in english as crisply and unambigously as possible. Always quantify you

```
[1]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np

#Load Haberman Cancer Survival dataset
hcs = pd.read_csv("haberman.csv")
```

[2]: #How Many data-points and features print(hcs.shape)

(306, 4)

we have 306 rows and 4 columns

```
[3]: #number of classes
print(hcs.columns)
```

Index(['age', 'year', 'nodes', 'status'], dtype='object')

[4]: hcs.head()

```
[4]:
                    nodes
         age
              year
                             status
          30
                 64
                          1
                                   1
     0
     1
          30
                 62
                          3
                                   1
                          0
                                   1
     2
          30
                 65
     3
          31
                 59
                          2
                                   1
          31
                 65
                          4
                                   1
```

```
[5]: hcs.tail()
[5]:
          age year nodes
                             status
                  62
     301
           75
                          1
     302
                  67
                          0
           76
                                   1
     303
           77
                  65
                          3
                                   1
     304
           78
                  65
                          1
                                   2
     305
           83
                  58
                          2
                                   2
[6]: #data-points per class
     hcs['year'].value_counts()
[6]: 58
           36
           31
     64
           30
     63
     66
           28
     65
           28
     60
           28
     59
           27
     61
           26
     67
           25
     62
           23
     68
           13
     69
           11
     Name: year, dtype: int64
[7]: hcs['age'].value_counts()
[7]: 52
           14
     54
           13
     50
           12
     47
           11
     53
           11
     43
           11
     57
           11
     55
           10
     65
           10
     49
           10
     38
           10
     41
           10
     61
            9
     45
            9
     42
            9
     63
            8
     59
            8
     62
            7
     44
            7
```

```
58
            7
     56
            7
     46
            7
     70
            7
     34
            7
     48
            7
     37
            6
     67
            6
     60
            6
     51
            6
     39
            6
     66
            5
     64
            5
     72
            4
     69
            4
     40
            3
     30
            3
     68
            2
     73
            2
     74
            2
            2
     36
     35
            2
     33
            2
     31
            2
     78
            1
     71
            1
     75
            1
     76
            1
     77
             1
     83
            1
     Name: age, dtype: int64
[8]: hcs['nodes'].value_counts()
[8]: 0
           136
            41
     1
     2
            20
     3
            20
     4
             13
     6
             7
             7
     7
             7
     8
             6
     5
     9
             6
     13
             5
     14
             4
     11
             4
```

```
10
         3
15
         3
         3
19
         3
22
23
         3
12
         2
20
         2
46
         1
         1
16
17
         1
18
         1
21
         1
24
         1
25
         1
28
         1
30
         1
35
         1
52
         1
Name: nodes, dtype: int64
```

```
[9]: hcs['status'].value_counts()
```

[9]: 1 

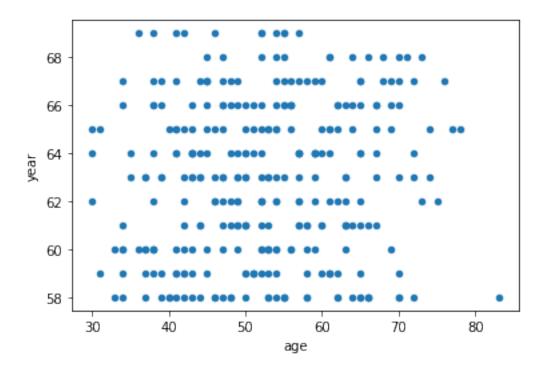
Name: status, dtype: int64

Analysis of survival of patients who had undergone surgery 225 patients survived more than 5 years and 81 patients survived less than 5 years

#### Bi-variate analysis

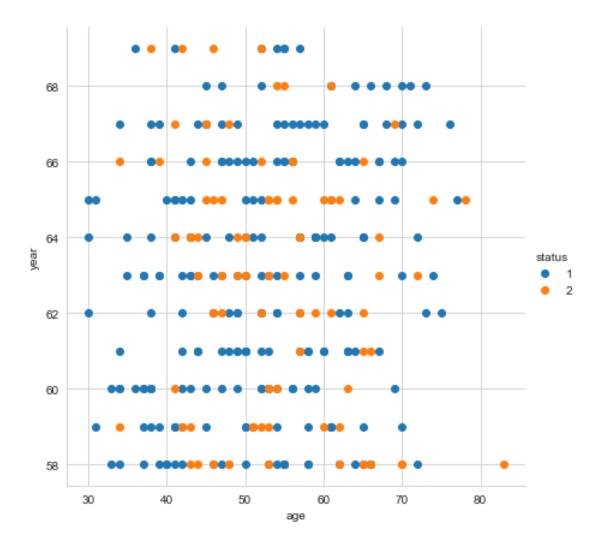
#### 1.1 2-D Scatter Plot

```
[10]: print(hcs.columns)
     Index(['age', 'year', 'nodes', 'status'], dtype='object')
[11]: hcs.plot(kind="scatter", x='age', y='year')
[11]: <AxesSubplot:xlabel='age', ylabel='year'>
```



C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\axisgrid.py:316: UserWarning: The `size` parameter has been renamed to `height`; please update your code.

warnings.warn(msg, UserWarning)

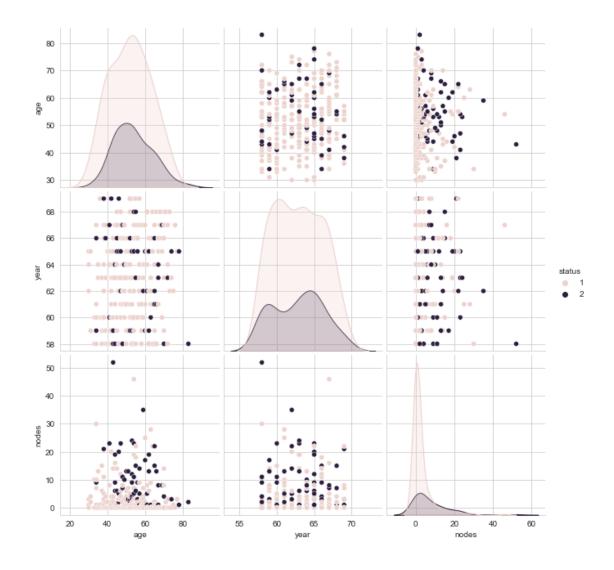


## 1.2 Pair-plot

```
[13]: plt.close();
    sns.set_style("whitegrid");
    sns.pairplot(hcs, hue="status", size=3).add_legend;
    plt.show()
```

C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\axisgrid.py:1912: UserWarning: The `size` parameter has been renamed to `height`; please update your code.

warnings.warn(msg, UserWarning)



#### 1.2.1 Observations:

Survival status is random spread but based on analysis

- There is high survival between age group 30 to 40
- $\bullet\,$  pateient have nodes between 0 and 1 has high survival and patients having node range 25 has very less survival

# 2 Univaraite analysis

#### 2.0.1 PDF

```
[14]: hcs_surv=hcs.loc[hcs["status"] == 1];
hcs_canc=hcs.loc[hcs["status"] == 2];
sns.FacetGrid(hcs, hue='status',height=5).map(sns.distplot,'age').add_legend();
```

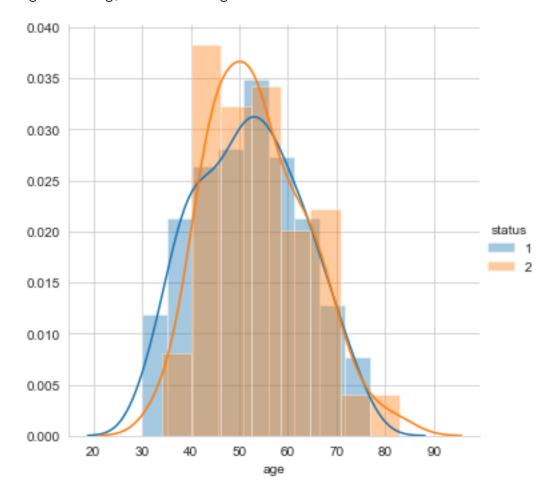
#### plt.show()

C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)



[15]: sns.FacetGrid(hcs, hue='status',height=5).map(sns.distplot,'year').add\_legend(); plt.show()

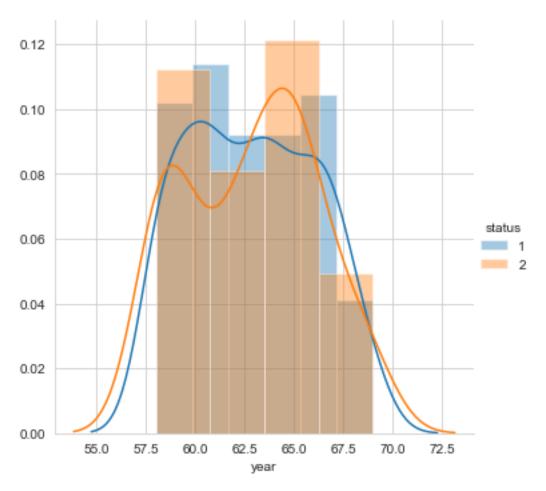
C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\distributions.py:2551:

FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)



```
[16]: sns.FacetGrid(hcs, hue='status',height=5).map(sns.distplot,'nodes').

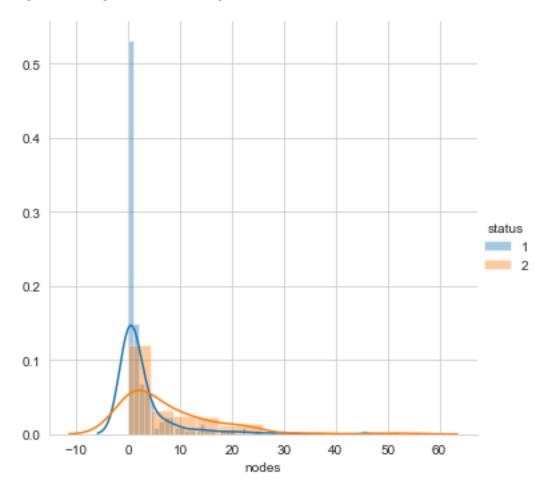
→add_legend();
plt.show()
```

C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)

C:\Users\Shadrach\anaconda3\lib\site-packages\seaborn\distributions.py:2551: FutureWarning: `distplot` is a deprecated function and will be removed in a future version. Please adapt your code to use either `displot` (a figure-level function with similar flexibility) or `histplot` (an axes-level function for histograms).

warnings.warn(msg, FutureWarning)



```
[17]: counts, bin_edges = np.histogram(hcs_canc['age'], bins=10, density= True);

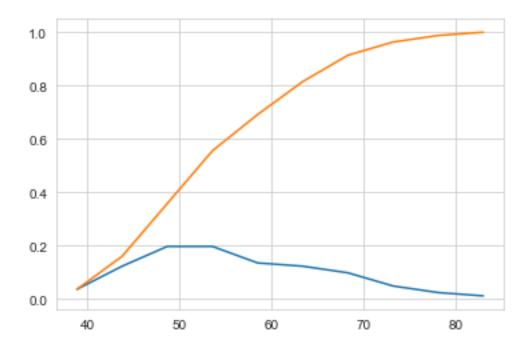
pdf = counts/(sum(counts));
 print(pdf);
 print(bin_edges);
 cdf=np.cumsum(pdf);

plt.plot(bin_edges[1:],pdf)
```

```
plt.plot(bin_edges[1:], cdf)
```

```
[0.03703704 0.12345679 0.19753086 0.19753086 0.13580247 0.12345679 0.09876543 0.04938272 0.02469136 0.01234568] [34. 38.9 43.8 48.7 53.6 58.5 63.4 68.3 73.2 78.1 83.]
```

#### [17]: [<matplotlib.lines.Line2D at 0x232cefebd90>]



```
[18]: #cancer
    counts1, bin_edges1 = np.histogram(hcs_canc['nodes'], bins=10, density= True);

pdf1 = counts1/(sum(counts1));
    print(pdf1);
    print(bin_edges1);
    cdf1=np.cumsum(pdf1);

sns.set_style("whitegrid");
    plt.plot(bin_edges1[1:],pdf1);
    plt.plot(bin_edges1[1:], cdf1, label = 'No');
    plt.xlabel('nodes')

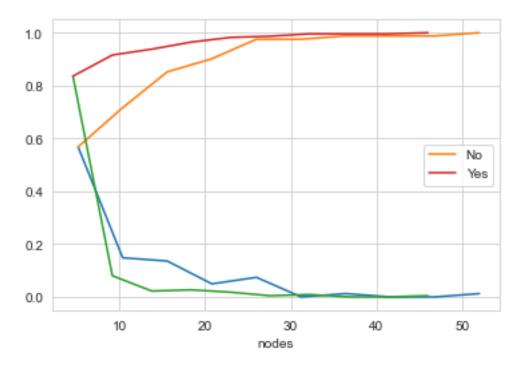
#survival
    counts2, bin_edges2 = np.histogram(hcs_surv['nodes'], bins=10, density= True);

pdf2 = counts2/(sum(counts2));
    print(pdf2);
```

```
print(bin_edges2);
cdf2=np.cumsum(pdf2);

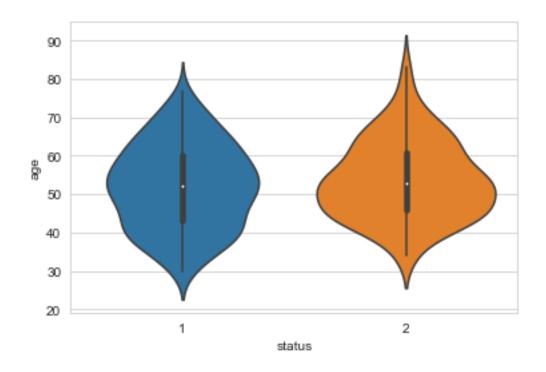
plt.plot(bin_edges2[1:],pdf2);
plt.plot(bin_edges2[1:], cdf2, label='Yes');
plt.xlabel('nodes');

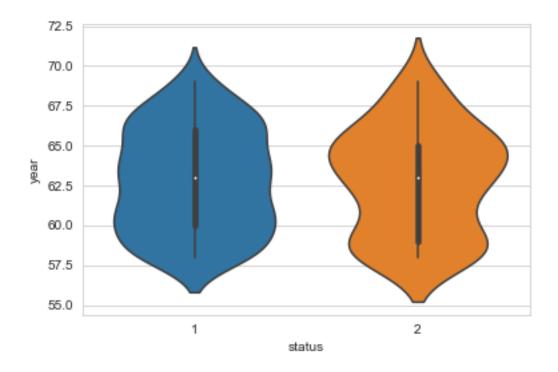
plt.legend();
plt.show();
```

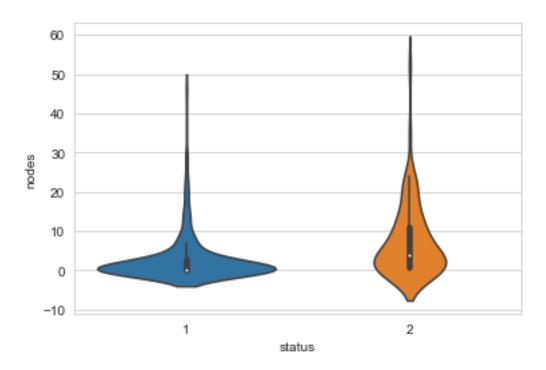


#### 2.1 Voilin Plot

```
[19]: sns.violinplot(x="status", y="age", data=hcs, size=8)
plt.show()
sns.violinplot(x="status", y="year", data=hcs, size=8)
plt.show()
sns.violinplot(x="status", y="nodes", data=hcs, size=8)
plt.show()
```

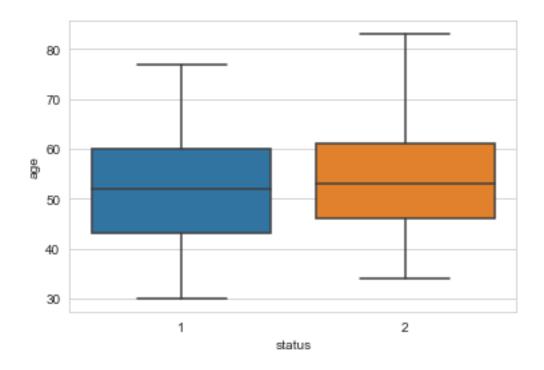


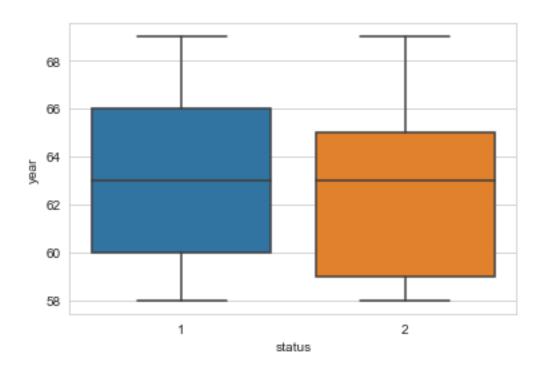


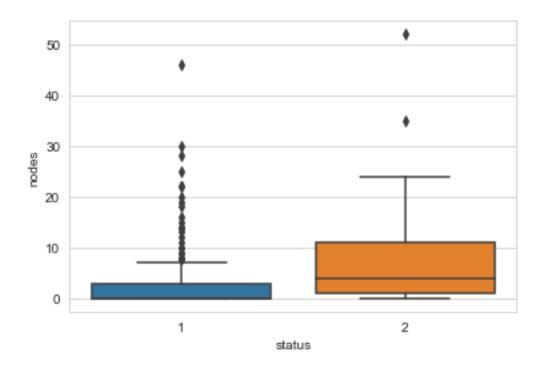


### 2.2 BOX PLOT

```
[20]: sns.boxplot(x="status", y="age", data=hcs)
plt.show()
sns.boxplot(x="status", y="year", data=hcs)
plt.show()
sns.boxplot(x="status", y="nodes", data=hcs)
plt.show()
```







#### 2.2.1 Observations:

- we have 306 rows and 4 columns
- data set has following columns ('age', 'year', 'nodes', 'status')
- Status is a feature which represent  $1 \rightarrow \text{survived}$ ,  $2 \rightarrow \text{no}$
- From the dataset we understood that 221 survived and 81 didn't

#### ###Post univariate and bivariate analysis

- There is high survival between/less age group 30 to 40
- patient having nodes between 0 and 1 has high survival and patients having node range >=25 has very less survival
- Between 1960 and 1965 there were more unsuccessful operations.

#### ###Conclusion

- These aren't the deciding factors but however there were non-survival cases between age group 30 to 40 but people less than 35 years have more chance of survival
- The objective of classifying the survival status of a new patient based on the given features is a difficult task as the data is imbalanced.