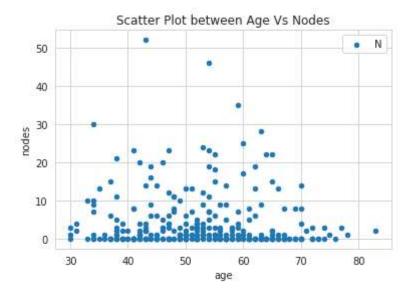
```
#Supressing Warnings
import warnings
warnings.filterwarnings("ignore")
# importing packages into the code
import pandas as pd
import numpy as np
import seaborn as sns
import matplotlib.pyplot as plt
#uploading files to Google colab
#This section is copied from the following link: https://colab.research.google.com/noteboo
from google.colab import files
uploaded = files.upload()
for fn in uploaded.keys():
 print('User uploaded file "{name}" with length {length} bytes'.format(
     name=fn, length=len(uploaded[fn])))
# reading the Haberman's cancer survival dataset
cancer_survival = pd.read_csv("haberman_survival.csv")
     Choose Files haberman survival.csv

    haberman survival.csv(application/vnd.ms-excel) - 3124 bytes, last modified: 10/2/2021 - 100%

    done
    Saving haberman survival.csv to haberman survival (1).csv
    User unloaded file "haberman survival csy" with length 3124 bytes
#printing dataset's rank
print(cancer_survival.shape)
    (306, 4)
# Printing the columns available in the dataset
print(cancer survival.columns)
    Index(['age', 'year', 'nodes', 'status'], dtype='object')
# printing the values of statuses available in the dataset
print(cancer survival['status'].value counts())
    1
         225
    2
          81
    Name: status, dtype: int64
# plotting a scatter plot between age and nodes
cancer_survival.plot(kind='scatter', x='age', y='nodes')
plt.title("Scatter Plot between Age Vs Nodes")
```

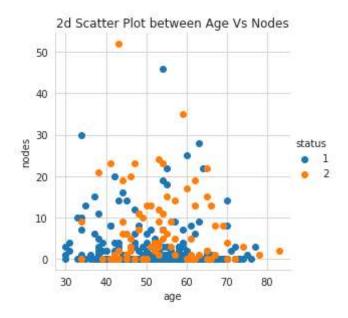
plt.legend("Nodes Vs Ages")
plt.show()



Observations:

- 1. Positive auxillary nodes are high for the patients age range between 40-65 years
- 2. Outliers(2 points) are present in the data
- 3. There is not much clarification between age vs nodes scatter plot

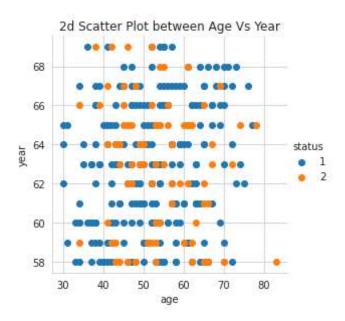
```
sns.set_style('whitegrid')
sns.FacetGrid(cancer_survival, hue='status', height=4).map(plt.scatter, "age", "nodes").ad
plt.title("2d Scatter Plot between Age Vs Nodes")
plt.show()
```



Observations:

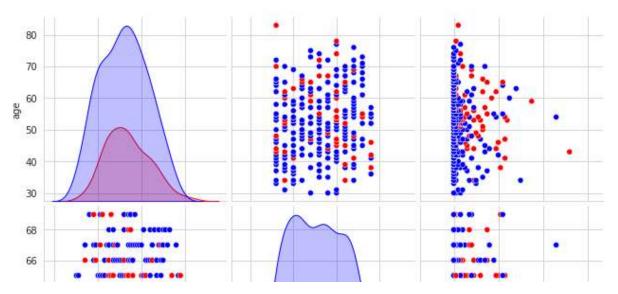
1. Unable to differentiate the survival status variable using this 2d scatter plot between age Vs positive auxiliary nodes

```
plt.close('all')
sns.set_style('whitegrid')
sns.FacetGrid(cancer_survival, hue='status', height=4).map(plt.scatter, "age", "year").add
plt.title("2d Scatter Plot between Age Vs Year")
plt.show()
```



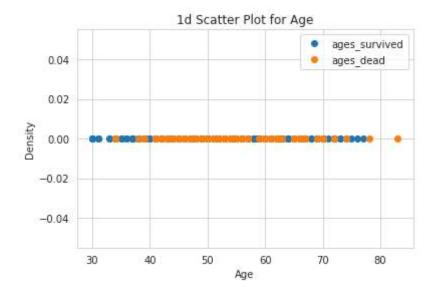
1. unable to differentiate status using 2d-plot between age Vs year

```
plt.close('all')
color_dict = {1:'blue', 2:'red'}
sns.set_style('whitegrid')
sns.pairplot(cancer_survival, hue='status', palette=color_dict, height=3)
plt.show()
```



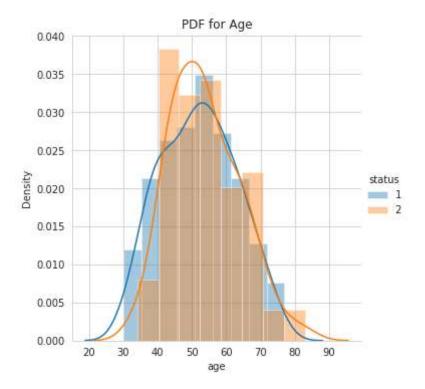
- 1. None of the 2d-scater plots have clear descrimination of status between the variables
- 2. Seems like variable 'nodes' follow the normal distribution
- 3. Patients with less number of nodes are more survived
- 4. Patient's operated in 1960's are more survived

```
# Plotting 1d scatter plots
plt.close('all')
status_survived = cancer_survival.loc[cancer_survival['status'] == 1]
status_dead = cancer_survival.loc[cancer_survival['status'] == 2]
plt.plot(status_survived['age'], np.zeros_like(status_survived['age']), 'o', label='ages_s
plt.plot(status_dead['age'], np.zeros_like(status_dead['age']), 'o', label='ages_dead')
plt.xlabel("Age")
plt.ylabel("Density")
plt.title("1d Scatter Plot for Age")
plt.legend()
plt.show()
```



Observations: Data points are over lapped, status can't be clearly seperated

```
plt.close('all')
sns.FacetGrid(cancer_survival, hue='status',height=5).map(sns.distplot, 'age').add_legend(
plt.title("PDF for Age")
plt.show()
```

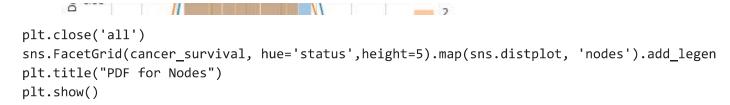


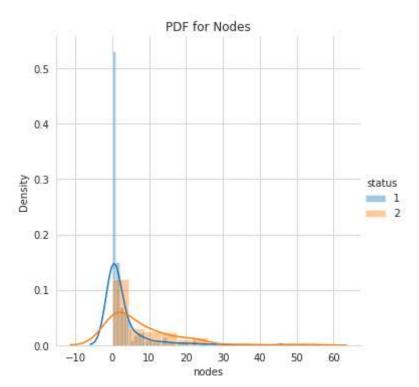
- 1. Seems like PDF of status (status=2) follows Guassian distribution
- 2. Patients aged between 40-60 years are more dead

```
plt.close('all')
sns.FacetGrid(cancer_survival, hue='status',height=5).map(sns.distplot, 'year').add_legend
plt.title("PDF for Year")
plt.show()
```



- 1. None of the distributions follow Normal distribution
- 2. Patients operated between 1963-1965 are more dead
- 3. Patients operated between 1959-1961 are more survived





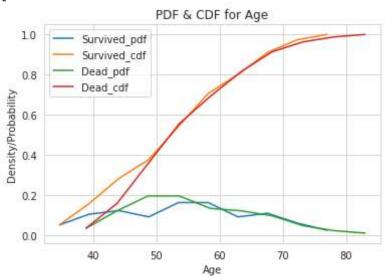
Observations:

- 1. Seems that status of survival and status of dead are follow normal distribution
- 2. Patients with 2 positive auxiliary nodes are more survived compared to dead
- 3. Patients with more than 20 positive auxiliary nodes are more dead

```
density = True)
```

```
pdf = counts/(sum(counts))
print(pdf);
print(bin_edges);
cdf = np.cumsum(pdf)
plt.plot(bin_edges[1:],pdf, label='Dead_pdf');
plt.plot(bin_edges[1:], cdf, label='Dead_cdf')
plt.xlabel("Age")
plt.ylabel("Density/Probability")
plt.legend()
plt.title("PDF & CDF for Age")
plt.show()
```

```
[0.05333333 0.10666667 0.12444444 0.09333333 0.16444444 0.16444444 0.09333333 0.16444444 0.16444444 0.09333333 0.11111111 0.06222222 0.02666667]
[30. 34.7 39.4 44.1 48.8 53.5 58.2 62.9 67.6 72.3 77. ]
[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. ]
```



1. Patients aged less than 38 years are always survived

```
# Calculating mean, Variance and standard deviation
print("Mean of Patients_Survived_nodes:")
print(np.mean(status_survived['nodes']))
print("Std-Dev of Patients_Survived_nodes:")
print(np.std(status_survived['nodes']))

print("Mean of Patients_Dead_nodes:")
print(np.mean(status_dead['nodes']))
print("Std-Dev of Patients_Dead_nodes:")
print(np.std(status_dead['nodes']))

Mean of Patients_Survived_nodes:
    2.791111111111113
    Std-Dev of Patients_Survived_nodes:
    5.857258449412131
```

```
Mean of Patients_Dead_nodes: 7.45679012345679
Std-Dev of Patients_Dead_nodes: 9.128776076761632
```

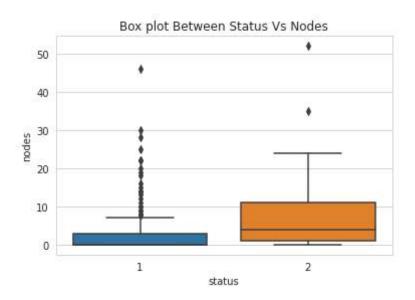
- 1. Patients with less than 3 nodes are more likely to survive
- 2. Patients with more than 7 nodes are more likely to dead

```
#Calculating Median, Quantiles, Percentiles, IQR
print("Median of nodes for survival patients:")
print(np.median(status survived['nodes']))
print("Median of nodes for dead patients:")
print(np.median(status dead['nodes']))
print("Quantiles of nodes for survival patients:")
print(np.percentile(status survived["nodes"],np.arange(0, 100, 25)))
print("Quantiles of nodes for dead patients:")
print(np.percentile(status_dead["nodes"],np.arange(0, 100, 25)))
print("90% survival patients nodes")
print(np.percentile(status_survived["nodes"],90))
print("90% dead patients nodes")
print(np.percentile(status dead["nodes"],90))
from statsmodels import robust
print("MAD of nodes for survived patients:")
print(robust.mad(status_survived["nodes"]))
print("MAD of nodes for dead patients:")
print(robust.mad(status_dead["nodes"]))
     Median of nodes for survival patients:
     0.0
     Median of nodes for dead patients:
     4.0
     Quantiles of nodes for survival patients:
     [0. 0. 0. 3.]
     Quantiles of nodes for dead patients:
     [ 0. 1. 4. 11.]
     90% survival patients nodes
     8.0
     90% dead patients nodes
     MAD of nodes for survived patients:
     0.0
     MAD of nodes for dead patients:
     5.930408874022408
```

Observations:

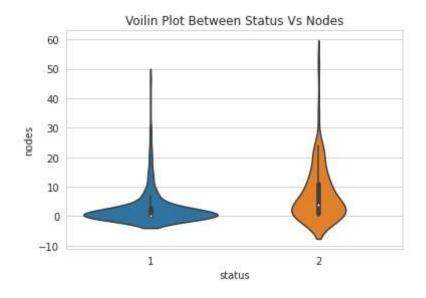
- 1. Patients with more than four nodes are dead with 75% probability
- 2. Patients with less than three nodes are survived with 100% probability

```
#Box plot
sns.boxplot(x='status',y='nodes', data=cancer_survival)
plt.title("Box plot Between Status Vs Nodes")
plt.show()
```



- 1. Inter quartile range of nodes for survived patients: 0-3
- 2. Inter quartile range of nodes for dead patients: 1-11

```
#Violin Plot
sns.violinplot(x="status", y="nodes", data=cancer_survival, size=8)
plt.title("Voilin Plot Between Status Vs Nodes")
plt.show()
```



Summary of the analysis:

1. Patients with less than 3 nodes have 100% probability to survive

- 2. Patients with less than 8 nodes have 90% probability to survive
- 3. Patients with less than 20 nodes have 90% probability to dead
- 4. Patients aged less than 38 years are 100% survived

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