

1. Download Haberman Cancer Survival dataset from Kaggle. You may have to create a Kaggle account to download data. (<https://www.kaggle.com/gilsousa/habermans-survival-data-set>)

```
In [32]: import warnings

warnings.filterwarnings("ignore")
```

2. Perform a similar analysis as above on this dataset with the following sections:

```
In [33]: import pandas as pd
import seaborn as sns
import matplotlib.pyplot as plt
import numpy as np
#Load Haberman.csv into a pandas DataFrame.
Haberman = pd.read_csv("C:/Users/aaa/Desktop/saraswati aaic/Haberman.csv")
Haberman.head()
```

Out[33]:

	age	year	nodes	status
0	30	64	1	1
1	30	62	3	1
2	30	65	0	1
3	31	59	2	1
4	31	65	4	1

High level statistics

* High level statistics of the dataset: number of points, number of features, number of classes, data-points per class.

```
In [34]: # (Q) how many data-points and features?
print ("No. of Datapoints",Haberman.shape[0])
print("No. of features",Haberman.shape[1])
```

No. of Datapoints 306
No. of features 4

```
In [35]: # print no of classes and data point of each class

Haberman["status"].value_counts()
```

Out[35]: 1 225
2 81
Name: status, dtype: int64

Observations

- 1)The dataset has 4 features and 305 data points.
- 2)The dataset has a collection of data of patient aged between 30-83 years those who had undergone cancer surgery in year 1958-1969.
- 3)Almost 75% of the patient had 0-4 nodes where 25% of them had 0 node and very few had up to 52 nodes.
- 4)The dataset has 224 datapoint labeled as "1" and 81 datapoint labeled as "2"
- 5)The dataset is an imbalance dataset.

Explain our objective.

Our objective is to classify a new patient belonging to status 1 or status 2 with the help of given data
i.e age years and nodes

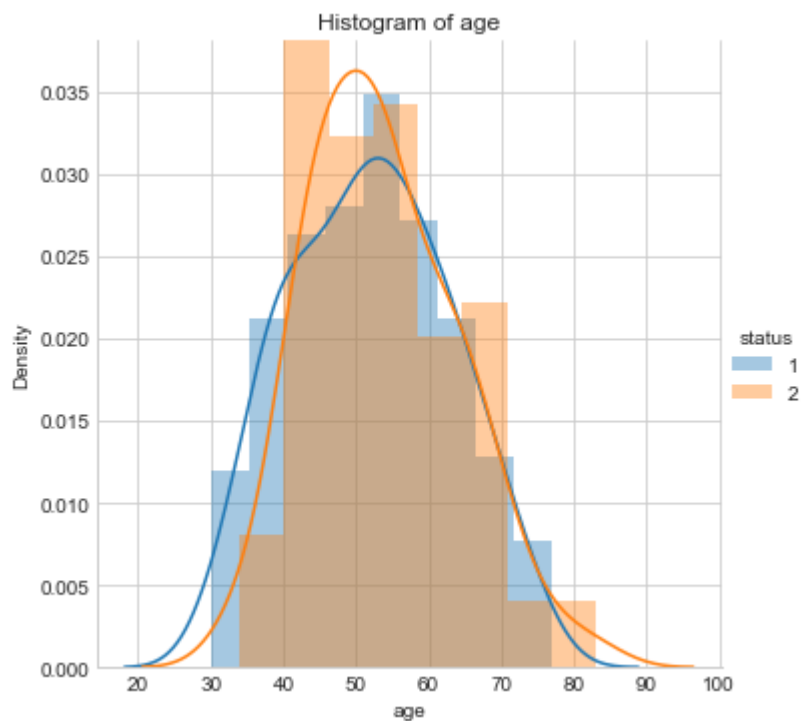
* Perform Univariate analysis(PDF, CDF, Boxplot, Violin plots) to understand which features are useful towards classification.

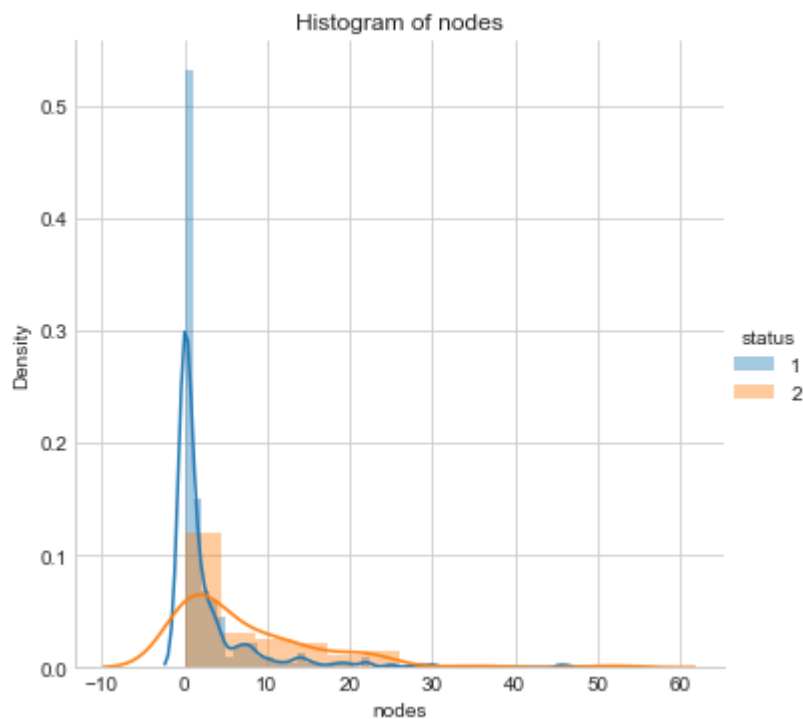
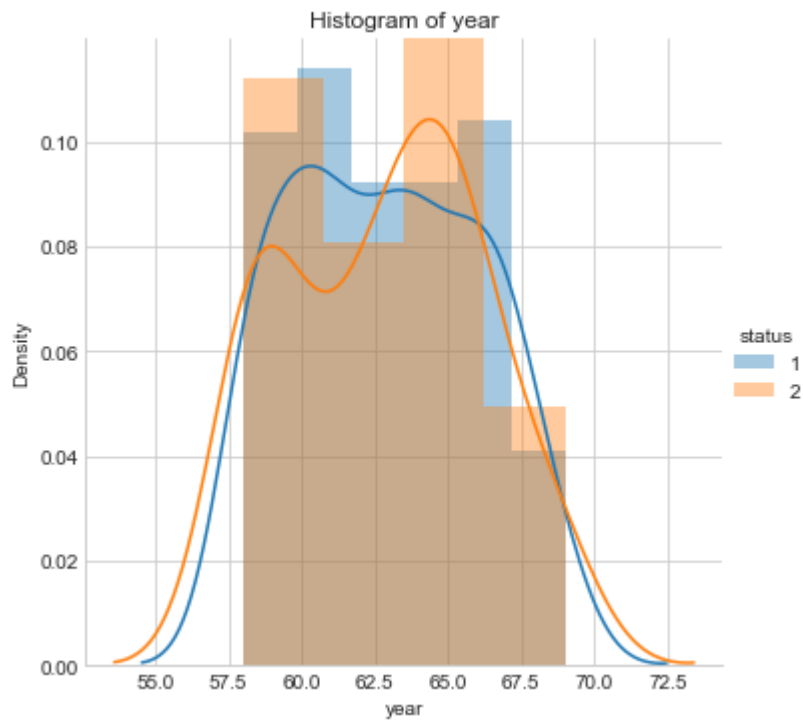
Histograms ,Pdf and Cdf

```
In [36]: # plot the pdf of given data
sns.FacetGrid(Haberman, hue="status", size=5) \
    .map(sns.distplot, "age") \
    .add_legend();
plt.title("Histogram of age")
plt.ylabel("Density")
plt.show();

sns.FacetGrid(Haberman, hue="status", size=5) \
    .map(sns.distplot, "year") \
    .add_legend();
plt.title("Histogram of year")
plt.ylabel("Density")
plt.show();

sns.FacetGrid(Haberman, hue="status", size=5) \
    .map(sns.distplot, "nodes") \
    .add_legend();
plt.title("Histogram of nodes")
plt.ylabel("Density")
plt.show();
```





Observations:

- 1) The survival status corresponding to operation year data points are overlapping, hence no conclusion about the survival status of the patient could be drawn based on the Year of operation.
- 2) The data is overlapping hence no major information could be gained. Patients with age less than 40 years have a higher chance to survive and patients with age more than 78 years are most likely to die within 5 years of surgery.

3) It is seen that 95% of the patient has nodes between 0 to 25.
Patient with 0-3 node had higher chances of survival.
Data is overlapping hence we can't find "point" and "if-else" conditions to build a simple model.

```
In [37]: # plot pdf and cdf for the same

counts, bin_edges = np.histogram(Haberman['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)

counts, bin_edges = np.histogram(Haberman['age'], bins=20,
                                  density = True)

pdf = counts/(sum(counts))
label=["pdf of age","cdf of age"]
plt.legend(label);
plt.title("pdf and cdf for age")
plt.xlabel("age")
plt.ylabel("% of person's")
plt.plot(bin_edges[1:],pdf);

plt.show();

counts, bin_edges = np.histogram(Haberman['year'], 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)

counts, bin_edges = np.histogram(Haberman['year'], bins=20,
                                  density = True)

pdf = counts/(sum(counts))
label=["pdf of year","cdf of year"]
plt.legend(label);
plt.title("pdf and cdf for year")
plt.xlabel("year")
plt.ylabel("% of person's")
plt.plot(bin_edges[1:],pdf);

plt.show();

counts, bin_edges = np.histogram(Haberman['nodes'], 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)
```

```

counts, bin_edges = np.histogram(Haberman['nodes'], bins=20,
                                density = True)

pdf = counts/(sum(counts))
label=["pdf of nodes","cdf of nodes"]
plt.legend(label);
plt.title("pdf and cdf for nodes")
plt.xlabel("nodes")
plt.ylabel("% of person's")
plt.plot(bin_edges[1:],pdf);

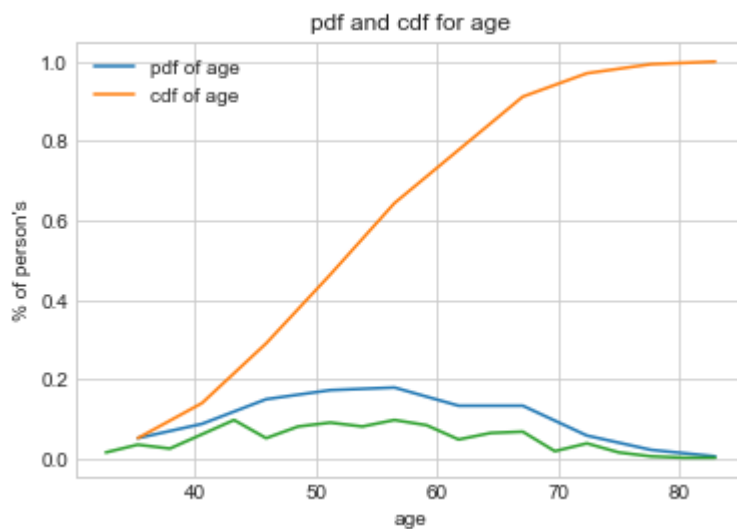
plt.show();

```

```

[0.05228758 0.08823529 0.1503268  0.17320261 0.17973856 0.13398693
 0.13398693 0.05882353 0.02287582 0.00653595]
[30.  35.3 40.6 45.9 51.2 56.5 61.8 67.1 72.4 77.7 83. ]

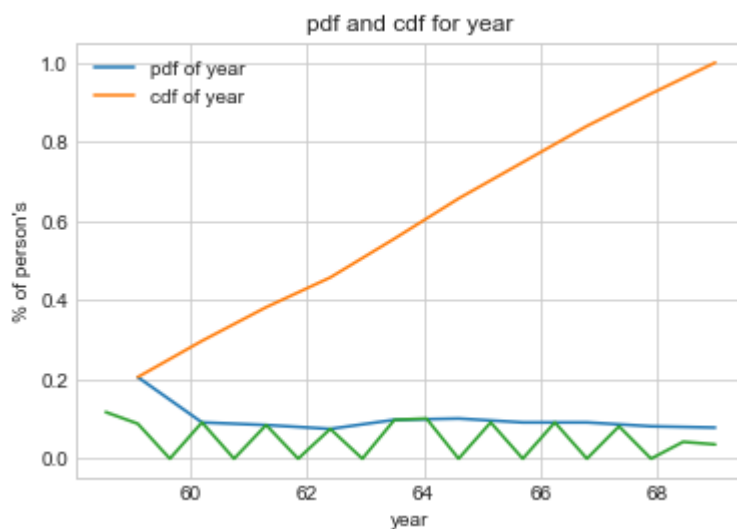
```



```

[0.20588235 0.09150327 0.08496732 0.0751634  0.09803922 0.10130719
 0.09150327 0.09150327 0.08169935 0.07843137]
[58.  59.1 60.2 61.3 62.4 63.5 64.6 65.7 66.8 67.9 69. ]

```

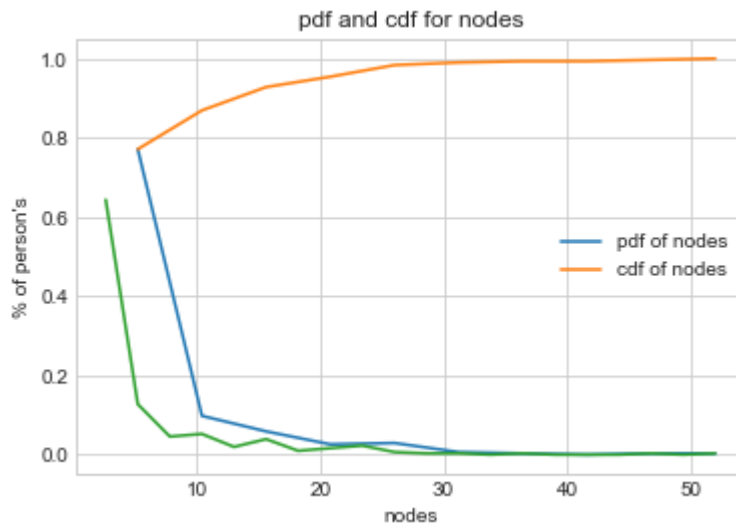


```

[0.77124183 0.09803922 0.05882353 0.02614379 0.02941176 0.00653595
 0.00326797 0.          0.00326797 0.00326797]

```

[0. 5.2 10.4 15.6 20.8 26. 31.2 36.4 41.6 46.8 52.]

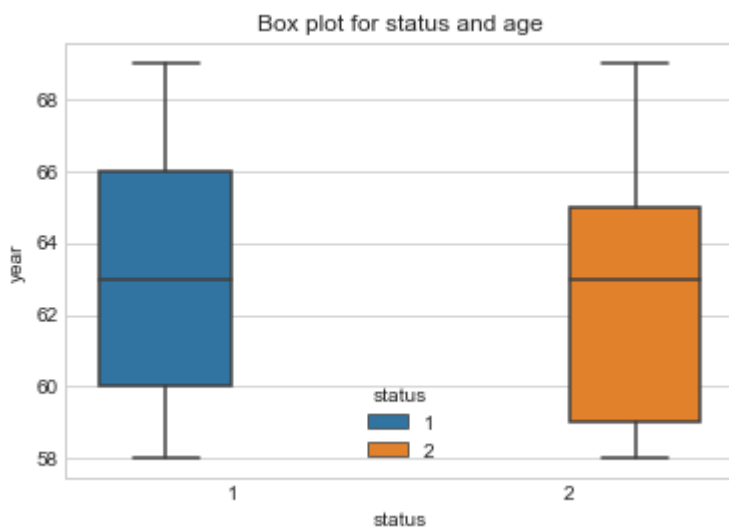
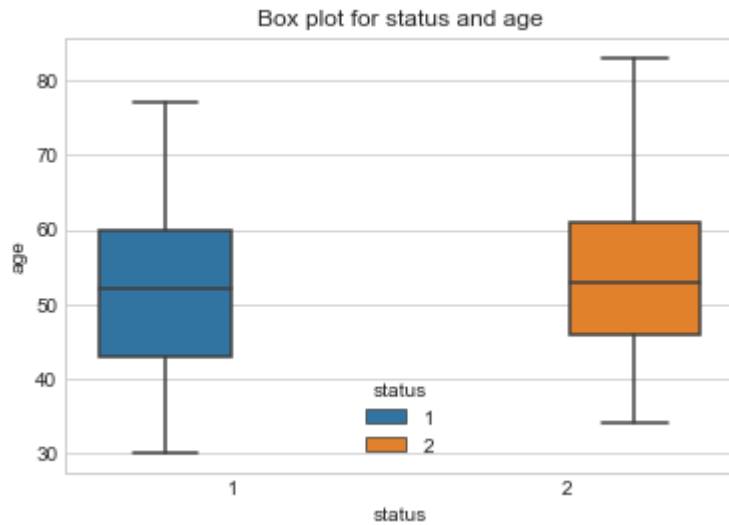


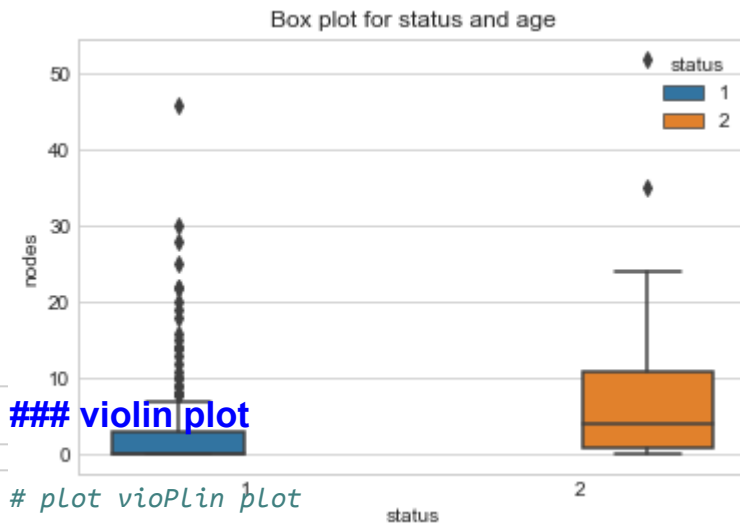
Observations:

- 1) Patient between age 32-36 has survived the operation and patient aged 77-85 has definitely not survived the operation.
- 2) as the data for both the case are evenly we can't draw the patient survival status from the year of operation. Except the patient who had surgery between 1961-1965 has high probability of survival.
- 3) patient with nodes <22 has better probability of survival.

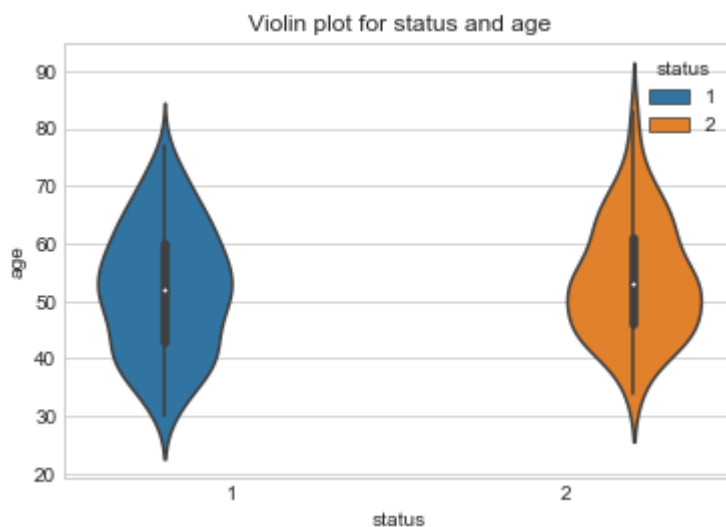
Box plot


```
In [48]: # plot box plot for the same
sns.boxplot(x='status',y='age',hue ="status", data=Haberman).set_title("Box plot
plt.show()
sns.boxplot(x='status',y='year',hue ="status", data=Haberman).set_title("Box plot
plt.show()
sns.boxplot(x='status',y='nodes',hue ="status", data=Haberman).set_title("Box plo
plt.show()
```





```
sns.violinplot(x="status", y="age", hue="status", data=Haberman, size=8)
plt.title("Violin plot for status and age")
plt.show()
sns.violinplot(x="status", y="year", hue="status", data=Haberman, size=8)
plt.title("Violin plot for status and year")
plt.show()
sns.violinplot(x="status", y="nodes", hue="status", data=Haberman, size=8)
plt.title("Violin plot for status and nodes")
plt.show()
```



observations:

- 1) As the data points are overlapping hence no major conclusion could be drawn from this plots
- 2) The number of node for survival is high or dense from 0-5.

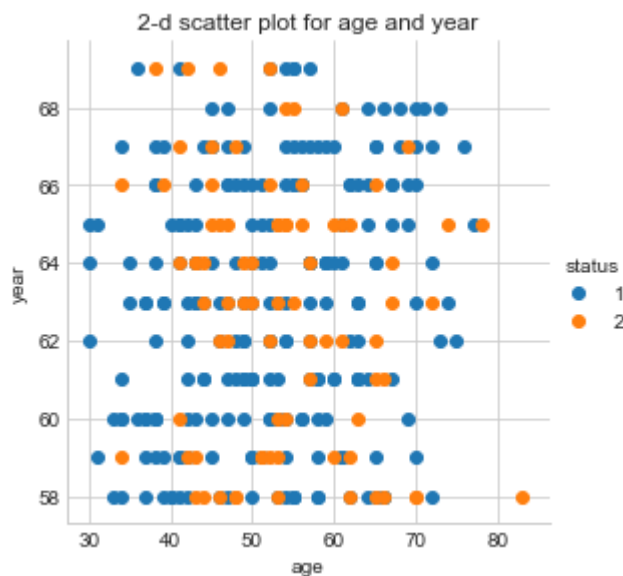
* Perform Bi-variate analysis (scatter plots, pair-plots) to see if combinations of features are useful in classification.

Scatter plot

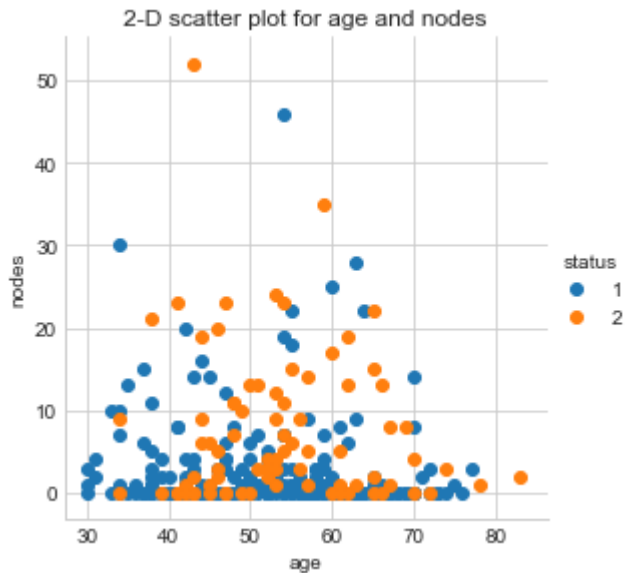
```
In [40]: # 2-D Scatter plot
Haberman.plot(kind = "scatter", x = "age", y = "year")
plt.title("2-D scatter plot of age")
plt.show()
```



```
In [41]: # 2-D Scatter plot with color-coding
sns.set_style("whitegrid");
sns.FacetGrid(Haberman, hue="status", size=4) \
    .map(plt.scatter, "age", "year") \
    .add_legend();
plt.title("2-d scatter plot for age and year")
plt.show();
```



```
In [42]: sns.set_style("whitegrid")
sns.FacetGrid(Haberman, hue = "status", size = 4).map(plt.scatter, "age", "nodes")
plt.title("2-D scatter plot for age and nodes")
plt.show()
```



observations

In the above 2d scatter plot class or status is not linearly seprable
0-5 node person survived and died as well but the died ratio is less than
survive ratio.

Pair plot

```
In [46]: # pairwise scatter plot: Pair-Plot
sns.set_style("whitegrid")
sns.pairplot(Haberman, hue = "status", vars = ["age", "year", "nodes"], size = 3)
plt.suptitle("pair plot of age, year and node")
plt.show()
```



Observations

The data are highly mixed up, hence we can't find "if-else" conditions to build a simple model to classify the survive status of the patient.