Modules and Their Version

```
In [1]: import sys
        print("Python Version :",sys.version)
        Python Version: 3.8.1 (tags/v3.8.1:1b293b6, Dec 18 2019, 22:39:24) [MS
        C v.1916 32 bit (Intel)]
In [2]: import numpy as np
        print("Numpy Version :",np. version )
        Numpy Version: 1.19.1
In [3]: import pandas as pd
        print("Pandas version :",pd. version )
        Pandas version: 1.1.0
In [4]: import sklearn
        print("Sklearn Version :",sklearn. version )
        Sklearn Version: 0.23.2
In [5]: import matplotlib
        print("Matplotlib Version :", matplotlib. version )
        Matplotlib Version: 3.3.1
In [6]: import scipy as sc
        print("Scipy Version :",sc. version )
        Scipy Version: 1.5.2
```

```
In [7]: import matplotlib.pyplot as plt
    from pandas.plotting import scatter_matrix
    from sklearn.model_selection import train_test_split
    from sklearn.metrics import accuracy_score
    from sklearn.metrics import confusion_matrix
    from sklearn.metrics import classification_report
    from sklearn.neighbors import KNeighborsClassifier
```

Load The Dataset Using Pandas(titanic-dataset)

In [8]: pd.set_option('display.max_columns',10,'display.width',1000)
 titanic_dataset = pd.read_csv("E:\\java\\titanic.csv")
 titanic_dataset.head(5)

Out[8]:

	Passengerld	Survived	Pclass	Name	Sex	•••	Parch	Ticket	Fare	Cabin	Emba
0	1	0	3	Braund, Mr. Owen Harris	male		0	A/5 21171	7.2500	NaN	
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female		0	PC 17599	71.2833	C85	
2	3	1	3	Heikkinen, Miss. Laina	female		0	STON/O2. 3101282	7.9250	NaN	
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female		0	113803	53.1000	C123	
4	5	0	3	Allen, Mr. William Henry	male		0	373450	8.0500	NaN	

```
5 rows × 12 columns
 In [9]: titanic dataset.tail(5)
 Out[9]:
                Passengerld Survived Pclass
                                              Name
                                                      Sex ... Parch Ticket Fare Cabin Embarke
                                            Montvila,
           886
                       887
                                  0
                                               Rev.
                                                      male ...
                                                                  0 211536 13.00
                                                                                   NaN
                                             Juozas
                                            Graham,
                                              Miss.
           887
                       888
                                  1
                                                    female ...
                                                                  0 112053 30.00
                                                                                   B42
                                            Margaret
                                               Edith
                                           Johnston,
                                               Miss.
                                  0
                                        3 Catherine female ...
                                                                            23.45
           888
                       889
                                                                                   NaN
                                              Helen
                                             "Carrie"
                                            Behr, Mr.
           889
                       890
                                  1
                                                                  0 111369 30.00 C148
                                                Karl
                                                      male ...
                                              Howell
                                             Dooley,
           890
                       891
                                  0
                                                Mr.
                                                                  0 370376 7.75
                                                      male ...
                                                                                   NaN
                                              Patrick
          5 rows × 12 columns
In [10]: print("Shape of Data :",titanic dataset.shape)
          Shape of Data: (891, 12)
In [11]: print("Missing Values : \n", titanic dataset.isna().sum())
          Missing Values:
           PassengerId
                               0
          Survived
                              0
```

```
Pclass
                           0
         Name
                           0
         Sex
                           0
         Age
                         177
         SibSp
                           0
         Parch
                           0
         Ticket
                           0
         Fare
         Cabin
                         687
         Embarked
                           2
         dtype: int64
In [12]: print(titanic dataset.groupby('Embarked').size())
         Embarked
         C
              168
         0
               77
              644
         dtype: int64
In [13]:
         print(titanic dataset.describe())
                PassengerId
                                Survived
                                              Pclass
                                                                        SibSp
                                                              Age
           Parch
                         Fare
                                          891.000000
                 891.000000 891.000000
                                                       714.000000
                                                                   891.000000 89
         count
         1.000000 891.000000
                 446.000000
                                0.383838
                                            2.308642
                                                        29.699118
                                                                     0.523008
         mean
         0.381594
                     32,204208
                 257.353842
                                            0.836071
                                                        14.526497
                                0.486592
                                                                     1.102743
         std
         0.806057
                     49.693429
                   1.000000
                                0.000000
                                            1.000000
                                                         0.420000
                                                                     0.000000
         min
         0.000000
                      0.000000
                 223.500000
                                0.000000
                                            2.000000
                                                        20.125000
         25%
                                                                     0.000000
         0.000000
                      7.910400
                                            3.000000
         50%
                 446.000000
                                0.000000
                                                        28.000000
                                                                     0.000000
         0.000000
                     14.454200
         75%
                 668.500000
                                1.000000
                                            3.000000
                                                        38.000000
                                                                     1.000000
         0.000000
                     31.000000
```

```
891.000000
                                1.000000
                                            3.000000
                                                        80.000000
                                                                     8.000000
         max
         6.000000 512.329200
         Removing Cabin, Name and Ticket column
In [14]: titanic dataset=titanic dataset.drop(columns=['Cabin','Name','Ticket'])
In [15]: print("Missing Values : \n", titanic dataset.isna().sum())
         Missing Values :
          PassengerId
                            0
         Survived
                           0
         Pclass
                           0
         Sex
                         177
         Age
         SibSp
         Parch
         Fare
         Embarked
         dtype: int64
         Removing two missing Embarked row
In [16]: titanic dataset = titanic dataset.dropna(subset=['Embarked'])
         print("Missing Values : \n", titanic dataset.isna().sum())
In [17]:
         titanic dataset.shape
         Missing Values :
          PassengerId
                            0
         Survived
                           0
         Pclass
                           0
         Sex
                         177
         Age
         SibSp
                           0
         Parch
                           0
         Fare
                           0
```

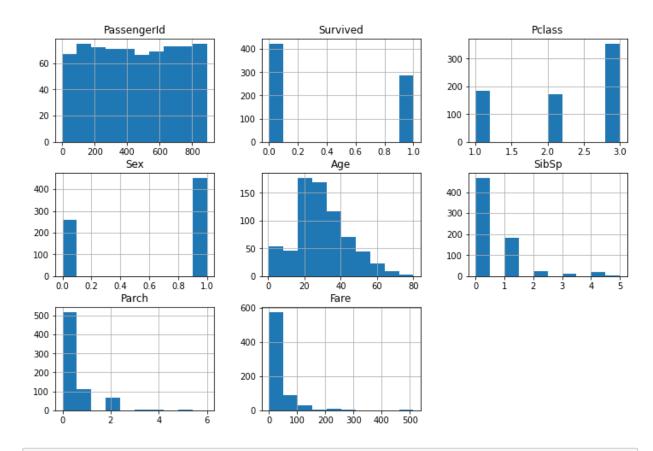
```
Embarked
                           0
         dtype: int64
Out[17]: (889, 9)
         Remove rows having NA Age value
In [49]: titanic dataset = titanic dataset.dropna()
In [21]: print("Missing Values : \n", titanic_dataset.isna().sum())
         titanic dataset.shape
         Missing Values :
          PassengerId
                          0
         Survived
                         0
         Pclass
         Sex
         Age
         SibSp
         Parch
         Fare
         Embarked
         dtype: int64
Out[21]: (712, 9)
         Preprocessing titanic_dataset
In [22]: df2=titanic dataset
In [23]: from sklearn.preprocessing import LabelEncoder
         labelencoder=LabelEncoder()
         df2.Sex=labelencoder.fit transform(df2.Sex)
In [24]: df2
Out[24]:
```

	Passengerld	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	0	3	1	22.0	1	0	7.2500	S
1	2	1	1	0	38.0	1	0	71.2833	С
2	3	1	3	0	26.0	0	0	7.9250	S
3	4	1	1	0	35.0	1	0	53.1000	S
4	5	0	3	1	35.0	0	0	8.0500	S
885	886	0	3	0	39.0	0	5	29.1250	Q
886	887	0	2	1	27.0	0	0	13.0000	S
887	888	1	1	0	19.0	0	0	30.0000	S
889	890	1	1	1	26.0	0	0	30.0000	С
890	891	0	3	1	32.0	0	0	7.7500	Q

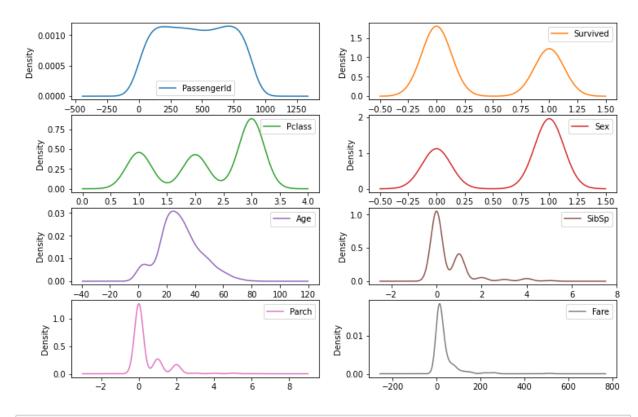
712 rows × 9 columns

Data Visulaization

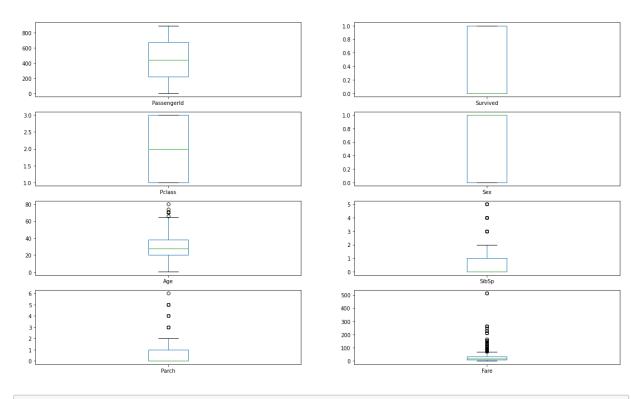
```
In [25]: titanic_dataset.hist(figsize=(12,8),sharex=False)
   plt.show()
```



In [26]: titanic_dataset.plot(kind='density',subplots=True,layout=(4,2),sharex=F
 alse,figsize=(12,8))
 plt.show()



In [27]: titanic_dataset.plot(kind='box',subplots=True,layout=(4,2),sharex=False
 , figsize=(20,12))
 plt.show()



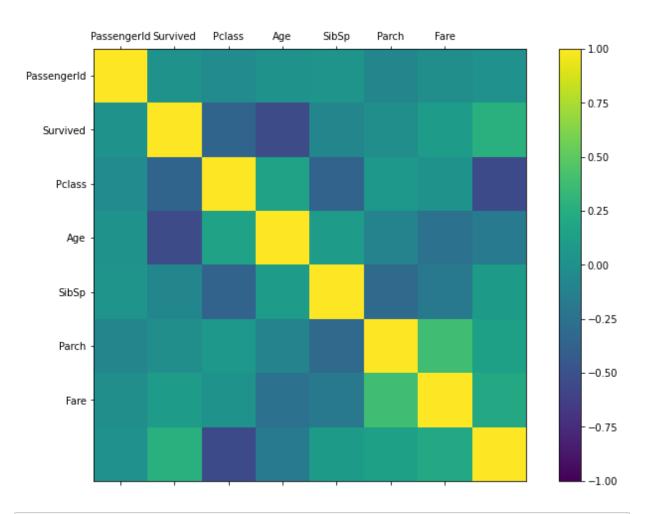
In [28]: correlation=titanic_dataset.corr()

In [29]: correlation

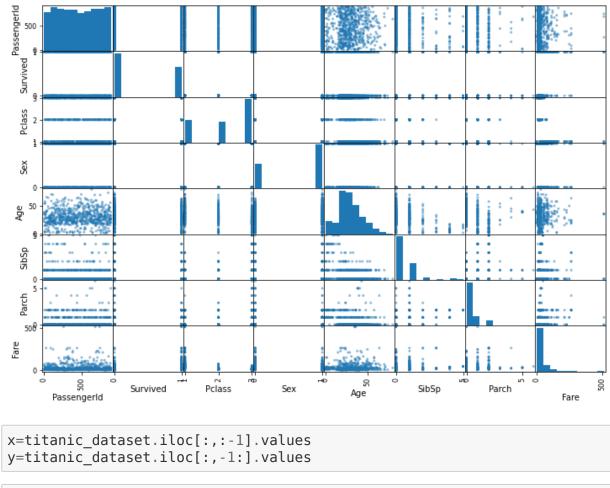
Out[29]:

	Passengerld	Survived	Pclass	Sex	Age	SibSp	Parch	F
Passengerld	1.000000	0.029526	-0.035609	0.024674	0.033681	-0.082704	-0.011672	0.009
Survived	0.029526	1.000000	-0.356462	-0.536762	-0.082446	-0.015523	0.095265	0.266
Pclass	-0.035609	-0.356462	1.000000	0.150826	-0.365902	0.065187	0.023666	-0.552
Sex	0.024674	-0.536762	0.150826	1.000000	0.099037	-0.106296	-0.249543	-0.182
Age	0.033681	-0.082446	-0.365902	0.099037	1.000000	-0.307351	-0.187896	0.093
SibSp	-0.082704	-0.015523	0.065187	-0.106296	-0.307351	1.000000	0.383338	0.139

		Passengerld	Survived	Pclass	Sex	Age	SibSp	Parch	F		
	Parch	-0.011672	0.095265	0.023666	-0.249543	-0.187896	0.383338	1.000000	0.206		
	Fare	0.009655	0.266100	-0.552893	-0.182457	0.093143	0.139860	0.206624	1.000		
	4								>		
In [30]:	<pre>fig=plt.figure(figsize=(12,8)) ax = fig.add_subplot(111) cx= ax.matshow(correlation, vmax=1, vmin=-1) ticks=np.arange(7) labels =['Passengerld','Survived','Pclass','Age','SibSp','Parch','Fare'] ax.set_xticks(ticks) ax.set_yticks(ticks) ax.set_yticklabels(labels) ax.set_yticklabels(labels) fig.colorbar(cx) plt.show()</pre>										



```
In [31]: scatter_matrix(titanic_dataset,figsize=(12,8))
plt.show()
```



```
In [32]: x=titanic_dataset.iloc[:,:-1].values
```

Logistic regression model

```
from sklearn.linear_model import LogisticRegression
In [34]:
         model5 = LogisticRegression()
         model5.fit(X_train,Y_train)
         Y pred5 = model5.predict(X test)
```

```
c:\users\hp\appdata\local\programs\python\python38-32\lib\site-packages
         \sklearn\utils\validation.py:72: DataConversionWarning: A column-vector
         y was passed when a 1d array was expected. Please change the shape of y
         to (n_samples, ), for example using ravel().
           return f(**kwargs)
         c:\users\hp\appdata\local\programs\python\python38-32\lib\site-packages
         \sklearn\linear model\ logistic.py:762: ConvergenceWarning: lbfgs faile
         d to converge (status=1):
         STOP: TOTAL NO. of ITERATIONS REACHED LIMIT.
         Increase the number of iterations (max_iter) or scale the data as shown
         in:
             https://scikit-learn.org/stable/modules/preprocessing.html
         Please also refer to the documentation for alternative solver options:
             https://scikit-learn.org/stable/modules/linear model.html#logistic-
         regression
           n iter i = check optimize result(
In [35]: print("Accuracy by LogisticRegression :",accuracy score(Y test,Y pred5
         Accuracy by LogisticRegression: 0.7710280373831776
In [ ]:
In [ ]:
         KNN model
In [36]: model=KNeighborsClassifier(n neighbors=10)
In [37]: model.fit(X train,Y train)
         <ipython-input-37-ffa49499a3bf>:1: DataConversionWarning: A column-vect
         or y was passed when a 1d array was expected. Please change the shape o
         f y to (n samples, ), for example using ravel().
           model.fit(X train,Y train)
```

```
Out[37]: KNeighborsClassifier(n_neighbors=10)
In [38]: Y pred=model.predict(X test)
         print("Accuracy by KNN :",accuracy score(Y_test,Y_pred))
         Accuracy by KNN : 0.7570093457943925
In [39]: confusion matrix(Y test,Y pred)
Out[39]: array([[ 0,
                        0, 37],
                [0, 0, 9],
                [ 6, 0, 162]], dtype=int64)
In [40]: print(classification report(Y test, Y pred))
                                    recall f1-score
                       precision
                                                       support
                                      0.00
                                                0.00
                    C
                            0.00
                                                            37
                            0.00
                                      0.00
                                                0.00
                    Q
                                                             9
                                      0.96
                                                0.86
                            0.78
                                                           168
                                                0.76
                                                           214
             accuracy
                            0.26
                                      0.32
                                                0.29
                                                           214
            macro avg
         weighted avg
                            0.61
                                      0.76
                                                0.68
                                                           214
         c:\users\hp\appdata\local\programs\python\python38-32\lib\site-packages
         \sklearn\metrics\ classification.py:1221: UndefinedMetricWarning: Preci
         sion and F-score are ill-defined and being set to 0.0 in labels with no
         predicted samples. Use `zero division` parameter to control this behavi
         or.
           warn prf(average, modifier, msg start, len(result))
         support vector model
In [41]:
        from sklearn.svm import SVC
         model2=SVC()
```

```
model2.fit(X train,Y train)
         Y pred2=model2.predict(X test)
         c:\users\hp\appdata\local\programs\python\python38-32\lib\site-packages
         \sklearn\utils\validation.py:72: DataConversionWarning: A column-vector
         y was passed when a 1d array was expected. Please change the shape of y
         to (n samples, ), for example using ravel().
           return f(**kwargs)
In [42]: print("Accuracy by SVM :",accuracy score(Y test,Y pred2))
         Accuracy by SVM : 0.7850467289719626
         Naive bayes model
In [43]: from sklearn.naive bayes import GaussianNB
         model3 = GaussianNB()
         model3.fit(X train,Y train)
         Y pred3=model3.predict(X test)
         c:\users\hp\appdata\local\programs\python\python38-32\lib\site-packages
         \sklearn\utils\validation.py:72: DataConversionWarning: A column-vector
         y was passed when a 1d array was expected. Please change the shape of y
         to (n samples, ), for example using ravel().
           return f(**kwargs)
In [44]: print("Accuracy score by Naive Bayes :",accuracy score(Y test,Y pred3))
         Accuracy score by Naive Bayes: 0.7570093457943925
In [45]: from sklearn.tree import DecisionTreeClassifier
         model4=DecisionTreeClassifier()
         model4.fit(X train,Y train)
         Y pred4 = model4.predict(X test)
In [46]: print("Accuracy score by Decision tree:",accuracy score(Y test,Y pred4
```

```
Accuracy score by Decision tree : 0.7570093457943925
----->>>>> Best accuracy I got by SVM <<<<<-----

In [47]: print("Accuracy by SVM :",accuracy_score(Y_test,Y_pred2))

Accuracy by SVM : 0.7850467289719626
```