PROJECT NAME = TITANIC SURVIVAL PREDICTION

Importing Library ¶

```
In [1]: import pandas as pd
    import numpy as np
    import seaborn as sns
    import matplotlib.pyplot as plt
    import warnings
    warnings.filterwarnings("ignore")
```

Data

In [3]: ▶ df

Out[3]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	
	0	892	0	3	Kelly, Mr. James	male	34.5	0	0	330911	
	1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	
	2	894	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	
	3	895	0	3	Wirz, Mr. Albert	ma l e	27.0	0	0	315154	
	4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	
		•••	•••								
	413	1305	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	
	414	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	1(
	415	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	
	416	1308	0	3	Ware, Mr. Frederick	ma l e	NaN	0	0	359309	
	417	1309	0	3	Peter, Master. Michael J	male	NaN	1	1	2668	1
	418 r	ows × 12 colu	ımns								



In [4]: ► df.head()

Out[4]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	Fare
	0	892	0	3	Kelly, Mr. James	male	34.5	0	0	330911	7.8292
	1	893	1	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	363272	7.0000
	2	894	0	2	Myles, Mr. Thomas Francis	male	62.0	0	0	240276	9.6875
	3	895	0	3	Wirz, Mr. Albert	male	27.0	0	0	315154	8.6625
	4	896	1	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	3101298	12.2875

In [5]: ► df.tail()

Out[5]:		Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Ticket	
	413	1305	0	3	Spector, Mr. Woolf	male	NaN	0	0	A.5. 3236	
	414	1306	1	1	Oliva y Ocana, Dona. Fermina	female	39.0	0	0	PC 17758	10
	415	1307	0	3	Saether, Mr. Simon Sivertsen	male	38.5	0	0	SOTON/O.Q. 3101262	
	416	1308	0	3	Ware, Mr. Frederick	male	NaN	0	0	359309	
	417	1309	0	3	Peter, Master. Michael J	male	NaN	1	1	2668	2
	4 4		_	_	_	_	_	_			

```
In [6]:
              df.describe()
     Out[6]:
                      PassengerId
                                    Survived
                                                Pclass
                                                                       SibSp
                                                                                  Parch
                                                                                              Far
                                                             Age
               count
                       418.000000
                                  418.000000 418.000000
                                                       332.000000
                                                                  418.000000
                                                                             418.000000
                                                                                        417.00000
                      1100.500000
                                              2.265550
               mean
                                    0.363636
                                                        30.272590
                                                                    0.447368
                                                                               0.392344
                                                                                         35.62718
                       120.810458
                                    0.481622
                                              0.841838
                                                                    0.896760
                 std
                                                         14.181209
                                                                               0.981429
                                                                                         55.90757
                       892.000000
                                    0.000000
                                               1.000000
                                                                    0.000000
                                                                               0.000000
                                                                                          0.00000
                 min
                                                          0.170000
                25%
                       996.250000
                                    0.000000
                                               1.000000
                                                        21.000000
                                                                    0.000000
                                                                               0.000000
                                                                                          7.89580
                50%
                      1100.500000
                                    0.000000
                                              3.000000
                                                        27.000000
                                                                    0.000000
                                                                               0.000000
                                                                                         14.45420
                      1204.750000
                                    1.000000
                                               3.000000
                                                        39.000000
                                                                    1.000000
                75%
                                                                               0.000000
                                                                                         31.50000
                     1309.000000
                                    1.000000
                                                        76.000000
                                               3.000000
                                                                    8.000000
                                                                               9.000000 512.32920
                max
 In [7]:
              type(df)
     Out[7]: pandas.core.frame.DataFrame
 In [8]:
              df.columns
              Index(['PassengerId', 'Survived', 'Pclass', 'Name', 'Sex', 'Age', 'SibS
              р',
                      'Parch', 'Ticket', 'Fare', 'Cabin', 'Embarked'],
                     dtype='object')
 In [9]:
              df.shape
     Out[9]: (418, 12)
In [10]:
              df.info()
              <class 'pandas.core.frame.DataFrame'>
              RangeIndex: 418 entries, 0 to 417
              Data columns (total 12 columns):
                #
                    Column
                                  Non-Null Count
                                                    Dtype
                                  -----
                                                    ----
                0
                    PassengerId 418 non-null
                                                    int64
                1
                    Survived
                                  418 non-null
                                                    int64
                2
                    Pclass
                                  418 non-null
                                                    int64
                3
                    Name
                                  418 non-null
                                                    object
                4
                    Sex
                                  418 non-null
                                                    object
                5
                                                    float64
                    Age
                                  332 non-null
                6
                    SibSp
                                  418 non-null
                                                    int64
                7
                    Parch
                                  418 non-null
                                                    int64
                8
                    Ticket
                                  418 non-null
                                                    object
                9
                    Fare
                                  417 non-null
                                                    float64
                10
                                                    object
                   Cabin
                                  91 non-null
                                  418 non-null
                11
                    Embarked
                                                    object
              dtypes: float64(2), int64(5), object(5)
              memory usage: 39.3+ KB
```

Data Cleaning

Checking Null Value and Duplicates In [11]: df.isna().sum() Out[11]: PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 Age 86 SibSp 0 Parch 0 Ticket 0 Fare 1 Cabin 327 Embarked dtype: int64 In [12]: df['Age'] = df['Age'].fillna(df['Age'].mean()) df['Fare'] = df['Fare'].fillna(df['Fare'].mean()) Could not convert x to numeric of Cabin Column. df.isna().sum() In [13]: Out[13]: PassengerId 0 Survived 0 Pclass 0 Name 0 Sex 0 Age SibSp 0 Parch 0 Ticket 0 Fare 0 Cabin 327 Embarked

dtype: int64

```
In [14]:

    df.duplicated()
    Out[14]: 0
                      False
              1
                      False
                      False
              2
              3
                      False
                      False
                      . . .
              413
                      False
              414
                      False
              415
                      False
              416
                      False
              417
                      False
              Length: 418, dtype: bool

    df.duplicated().sum()

In [15]:
    Out[15]: 0
```

Convert object to numeric

```
In [16]:
             df.info()
             <class 'pandas.core.frame.DataFrame'>
             RangeIndex: 418 entries, 0 to 417
             Data columns (total 12 columns):
                  Column
                               Non-Null Count
                                               Dtype
                               -----
                  -----
                                               ____
                  PassengerId 418 non-null
              0
                                               int64
                  Survived
              1
                               418 non-null
                                               int64
              2
                  Pclass
                               418 non-null
                                               int64
              3
                  Name
                               418 non-null
                                               object
              4
                  Sex
                               418 non-null
                                               object
              5
                               418 non-null
                                               float64
                  Age
              6
                  SibSp
                               418 non-null
                                               int64
              7
                  Parch
                               418 non-null
                                               int64
              8
                  Ticket
                               418 non-null
                                               object
              9
                  Fare
                               418 non-null
                                               float64
              10
                 Cabin
                               91 non-null
                                               object
              11 Embarked
                               418 non-null
                                               object
             dtypes: float64(2), int64(5), object(5)
             memory usage: 39.3+ KB
In [17]:
             Embarked = df['Embarked'].unique()
             for Embarkeds in Embarked:
                 print(Embarkeds)
             Q
             S
```

C

```
In [18]:
             df['Embarked'] = df['Embarked'].map( {'Q': 0,'S':1,'C':2}).astype(int)
             df['Sex'] = df['Sex'].map( {'female': 1, 'male':0}).astype(int)
In [72]:
         M df['Age'] = df['Age'].astype(int)
             df['Fare'] = df['Fare'].astype(int)
In [73]:
          M
             data = df.drop(['PassengerId','Name','Cabin','Ticket'], axis =1, inplace=T
In [74]:
             df.info()
             <class 'pandas.core.frame.DataFrame'>
             RangeIndex: 418 entries, 0 to 417
             Data columns (total 8 columns):
                  Column
                            Non-Null Count Dtype
                  Survived 418 non-null
              0
                                             int64
                  Pclass
              1
                            418 non-null
                                             int64
              2
                  Sex
                            418 non-null
                                             int32
              3
                  Age
                            418 non-null
                                             int32
              4
                            418 non-null
                                             int64
                  SibSp
              5
                  Parch
                            418 non-null
                                             int64
                  Fare
                            418 non-null
                                             int32
                  Embarked 418 non-null
                                             int32
             dtypes: int32(4), int64(4)
             memory usage: 19.7 KB
```

In [75]: ► df

Out[75]:

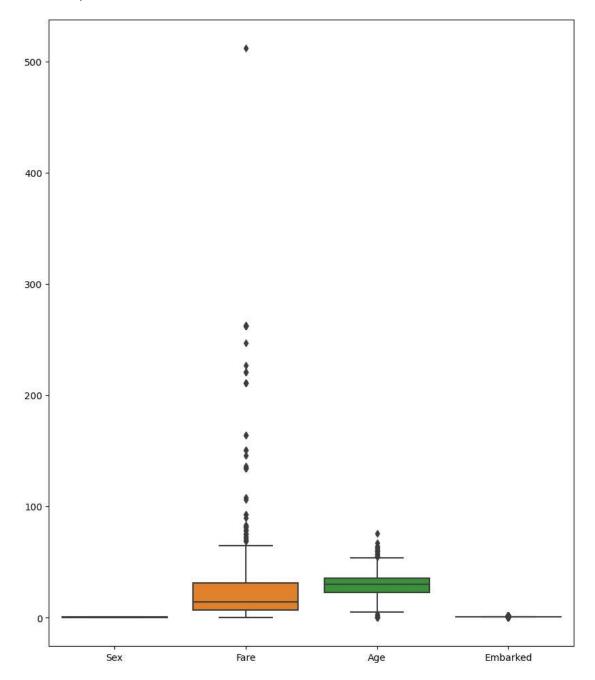
	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked
0	0	3	0	34	0	0	7	0
1	1	3	1	47	1	0	7	1
2	0	2	0	62	0	0	9	0
3	0	3	0	27	0	0	8	1
4	1	3	1	22	1	1	12	1
413	0	3	0	30	0	0	8	1
414	1	1	1	39	0	0	108	2
415	0	3	0	38	0	0	7	1
416	0	3	0	30	0	0	8	1
417	0	3	0	30	1	1	22	2

418 rows × 8 columns

Data Visualization

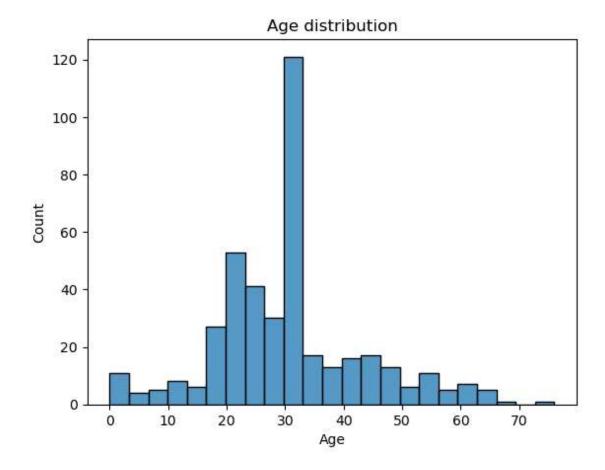
Outlier

Out[76]: <AxesSubplot:>

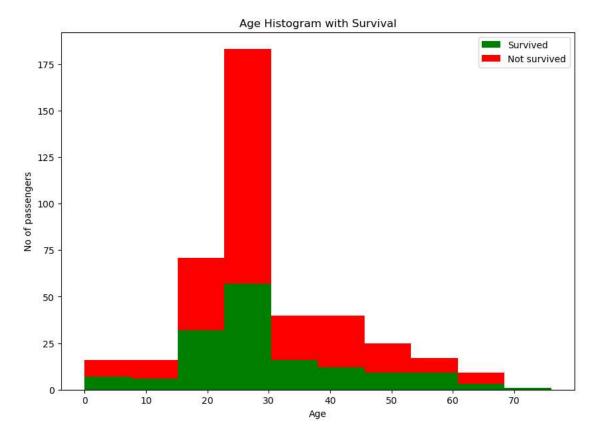


```
In [77]: In sns.histplot(df.Age)
plt.title('Age distribution')
```

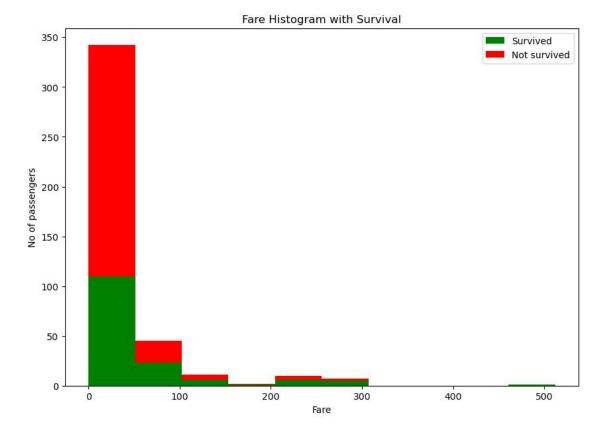
Out[77]: Text(0.5, 1.0, 'Age distribution')



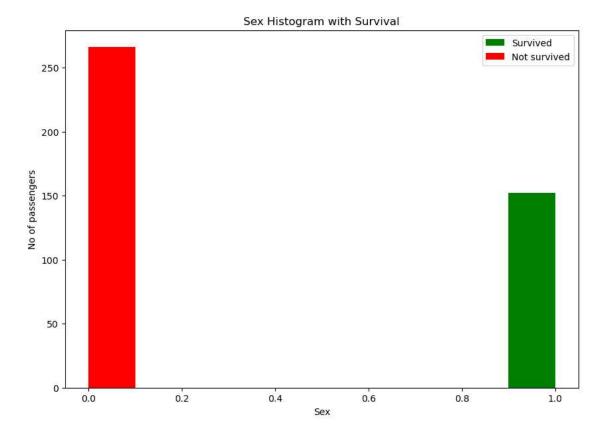
Out[78]: <matplotlib.legend.Legend at 0x2ada26fb880>



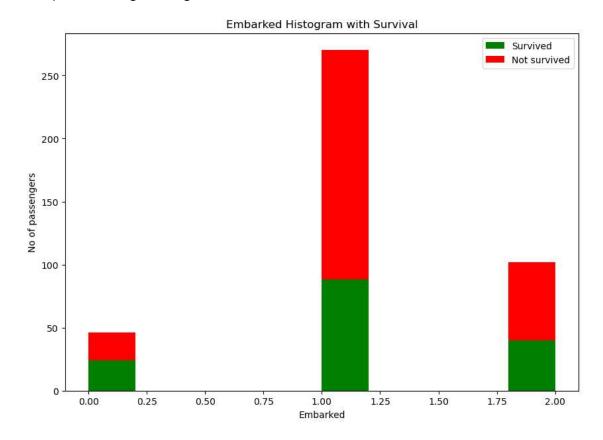
Out[79]: <matplotlib.legend.Legend at 0x2ada275b550>



Out[80]: <matplotlib.legend.Legend at 0x2ada2a76370>



Out[81]: <matplotlib.legend.Legend at 0x2ada30b38b0>



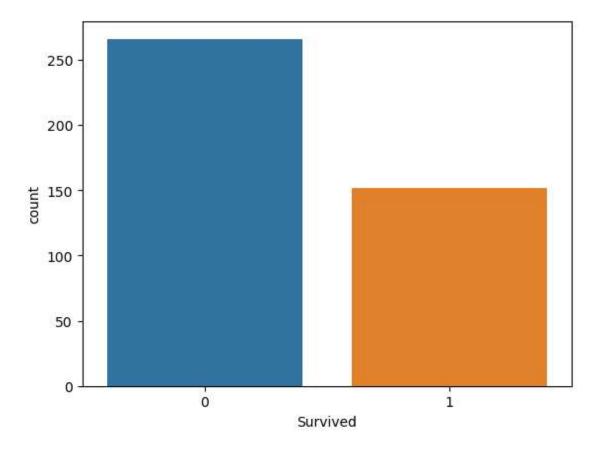
In [82]: ▶ df.Survived.value_counts()

Out[82]: 0 266 1 152

Name: Survived, dtype: int64

```
In [83]:  sns.countplot('Survived',data=df)
```

Out[83]: <AxesSubplot:xlabel='Survived', ylabel='count'>



Splitting The data

```
In [106]:
           ▶ | from sklearn.metrics import accuracy score
              from sklearn.metrics import classification report
In [107]:
              LR = LogisticRegression(solver='liblinear', max_iter=200)
              LR.fit(x_train, y_train)
              y_pred = LR.predict(x_test)
              LR1 = accuracy_score(y_pred,y_test)
              print('Logistic regression accuracy: {:.2f}%'.format(LR1*100))
              Logistic regression accuracy: 92.86%
In [108]:
           ▶ svc_model=SVC()
              svc_model.fit(x_train,y_train)
              pred=svc model.predict(x test)
           N SVC = accuracy_score(y_pred,y_test)
In [109]:
              print('SVC accuracy: {:.2f}%'.format(LR1*100))
              SVC accuracy: 92.86%
In [111]:
           ▶ RF1=RandomForestClassifier()
              RF1.fit(x train,y train)
              pred=RF1.predict(x_test)
In [113]:
           ▶ Rf1= accuracy_score(y_pred,y_test)
              print('RandomForestClassifier accuracy: {:.2f}%'.format(LR1*100))
              RandomForestClassifier accuracy: 92.86%
  In [ ]:
  In [ ]:
```