```
In [1]:
```

```
import pandas as pd
import numpy as np
import random as rnd
import seaborn as sns
import matplotlib.pyplot as plt
from sklearn.linear model import LogisticRegression
from sklearn.svm import SVC, LinearSVC
from sklearn.ensemble import RandomForestClassifier
from sklearn.ensemble import VotingClassifier
from sklearn.neighbors import KNeighborsClassifier
from sklearn.naive bayes import GaussianNB
from sklearn.linear model import Perceptron
from sklearn.linear model import SGDClassifier
from sklearn.tree import DecisionTreeClassifier
from sklearn.calibration import CalibratedClassifierCV
from sklearn.metrics import accuracy score
from sklearn.neural network import MLPClassifier
from sklearn.preprocessing import MinMaxScaler, StandardScaler, OneHotEncoder, LabelEncod
er
from sklearn.model selection import train test split, KFold, cross val score, cross val p
redict, cross validate, GridSearchCV
import xgboost as xgb
train dataset = pd.read csv('train.csv')
test dataset = pd.read csv('test.csv')
combine = [train dataset, test dataset]
```

In [2]:

```
# train_dataset['isCabinNull'] = train_dataset['Cabin'].isnull()*1
# train_dataset.head()
```

In [3]:

```
# test_dataset['isCabinNull'] = test_dataset['Cabin'].isnull()*1
# test_dataset.head()
```

In [4]:

```
train_dataset=train_dataset.drop("Ticket",axis=1)
train_dataset=train_dataset.drop("Cabin",axis=1)
train_dataset.head()
```

Out[4]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Fare	Embarked
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	7.2500	s
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	remale	38.0	1	0	71.2833	С
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	7.9250	s
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	temale	35.0	1	0	53.1000	s
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	8.0500	s

In [5]:

```
test_dataset=test_dataset.drop("Ticket",axis=1)
test_dataset=test_dataset.drop("Cabin",axis=1)

test_dataset.head()
```

Out[5]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Fare	Embarked
0	892	3	Kelly, Mr. James	male	34.5	0	0	7.8292	Q
1	893	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	7.0000	s
2	894	2	Myles, Mr. Thomas Francis	male	62.0	0	0	9.6875	Q
3	895	3	Wirz, Mr. Albert	male	27.0	0	0	8.6625	s
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	12.2875	S

In [6]:

```
train_dataset['FamilySize'] = train_dataset['SibSp'] + train_dataset['Parch'] + 1
train_dataset.head()
```

Out[6]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Fare	Embarked	FamilySize
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	7.2500	s	2
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	71.2833	С	2
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	7.9250	s	1
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	53.1000	s	2
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	8.0500	s	1

In [7]:

```
test_dataset['FamilySize'] = test_dataset['SibSp'] + test_dataset['Parch'] + 1
train_dataset.head()
```

Out[7]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Fare	Embarked	FamilySize
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	7.2500	s	2
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38.0	1	0	71.2833	С	2
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	7.9250	s	1
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	53.1000	s	2
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	8.0500	s	1

In [8]:

```
def categorise(row):
    if row > 1 and row < 5:
        return 1
    return 0
train_dataset['NormalFamilySize'] = train_dataset['FamilySize'].apply(categorise)
train_dataset.head()</pre>
```

Out[8]:

Pas	sengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Fare	Embarked	FamilySize	NormalFamilySize
0	1	0	3	Braund, Mr. Owen Harris	male	22.0	1	0	7.2500	s	2	1

	Passengerld	Survived	Pclass	Cun niage Mrs. John	Sex	Age	SibSp	Parch	Fare	Embarked	FamilySize	NormalFamilySize
1	2	1	1	Bradley (Florence Briggs Th	female	38.0	1	0	71.2833	С	2	1
2	3	1	3	Heikkinen, Miss. Laina	female	26.0	0	0	7.9250	s	1	0
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35.0	1	0	53.1000	s	2	1
4	5	0	3	Allen, Mr. William Henry	male	35.0	0	0	8.0500	s	1	0

In [9]:

```
def categorise(row):
    if row > 1 and row < 5:
        return 1
    return 0
test_dataset['NormalFamilySize'] = test_dataset['FamilySize'].apply(categorise)
test_dataset.head()</pre>
```

Out[9]:

	Passengerld	Pclass	Name	Sex	Age	SibSp	Parch	Fare	Embarked	FamilySize	NormalFamilySize
0	892	3	Kelly, Mr. James	male	34.5	0	0	7.8292	Q	1	0
1	893	3	Wilkes, Mrs. James (Ellen Needs)	female	47.0	1	0	7.0000	s	2	1
2	894	2	Myles, Mr. Thomas Francis	male	62.0	0	0	9.6875	Q	1	0
3	895	3	Wirz, Mr. Albert	male	27.0	0	0	8.6625	s	1	0
4	896	3	Hirvonen, Mrs. Alexander (Helga E Lindqvist)	female	22.0	1	1	12.2875	s	3	1

In [10]:

28 40 38 18 34 33 28 19 20]

```
data = [train dataset, test dataset]
for each in data:
   meanAge = train dataset["Age"].mean()
    stdAge = test dataset["Age"].std()
    isNull = each["Age"].isnull().sum()
    # print(isNull)
    # rand age = np.random.randint(meanAge - stdAge, meanAge + stdAge)
   randomAgeGenerator = np.random.randint(meanAge - stdAge, meanAge + stdAge, size = is
Null)
   print(randomAgeGenerator)
   age copy = each["Age"].copy()
    age copy[np.isnan(age copy)] = randomAgeGenerator
    each["Age"] = age_copy
    each["Age"] = train dataset["Age"].astype(int)
[34 36 15 37 16 18 28 30 28 25 40 34 37 21 30 20 27 42 20 23 41 28 31 36
17 32 23 38 42 36 25 18 42 29 23 24 36 20 16 21 23 16 22 40 37 31 37 41
34 16 21 31 31 33 28 19 19 31 25 26 16 37 42 17 29 18 41 35 35 25 37 40
23 24 33 31 19 20 16 21 27 32 40 29 27 41 21 27 42 30 15 16 17 36 24 16
33 27 32 37 24 16 26 18 33 37 41 29 39 19 41 23 34 21 40 38 35 39 26 29
```

37 21 19 37 21 21 26 31 21 33 38 40 22 31 35 31 26 41 22 25 28 20 39 41 19 23 28 33 15 17 15 17 15 24 23 37 16 34 29 28 21 38 18 26 42 29 27 27

```
[17 35 19 37 20 42 36 23 32 20 21 37 34 31 32 36 18 24 24 35 32 30 38 33 27 27 21 26 30 27 18 19 24 36 40 33 34 28 28 36 38 38 21 30 17 38 37 28 28 15 33 29 30 30 38 15 16 22 36 26 25 22 32 28 41 36 21 36 37 37 36 37 21 28 23 37 23 39 32 25 20 36 16 27 24 18]
```

In [11]:

```
# train_dataset["Embarked"] = train_dataset["Embarked"].fillna('C')
train_dataset["Embarked"] = train_dataset["Embarked"].fillna('S')
train_dataset.head()
```

Out[11]:

	Passengerld	Survived	Pclass	Name	Sex	Age	SibSp	Parch	Fare	Embarked	FamilySize	NormalFamilySize
0	1	0	3	Braund, Mr. Owen Harris	male	22	1	0	7.2500	s	2	1
1	2	1	1	Cumings, Mrs. John Bradley (Florence Briggs Th	female	38	1	0	71.2833	С	2	1
2	3	1	3	Heikkinen, Miss. Laina	female	26	0	0	7.9250	S	1	0
3	4	1	1	Futrelle, Mrs. Jacques Heath (Lily May Peel)	female	35	1	0	53.1000	s	2	1
4	5	0	3	Allen, Mr. William Henry	male	35	0	0	8.0500	s	1	0

In [12]:

```
test_dataset = test_dataset.fillna(test_dataset['Fare'].mean())
```

In [13]:

```
train_dataset=train_dataset.drop("PassengerId",axis=1)
train_dataset=train_dataset.drop("Name",axis=1)
train_dataset.head()
```

Out[13]:

	Survived	Pclass	Sex	Age	SibSp	Parch	Fare	Embarked	FamilySize	NormalFamilySize
0	0	3	male	22	1	0	7.2500	s	2	1
1	1	1	female	38	1	0	71.2833	С	2	1
2	1	3	female	26	0	0	7.9250	s	1	0
3	1	1	female	35	1	0	53.1000	s	2	1
4	0	3	male	35	0	0	8.0500	s	1	0

In [14]:

```
test_dataset=test_dataset.drop("PassengerId",axis=1)
test_dataset=test_dataset.drop("Name",axis=1)
test_dataset.head()
```

Out[14]:

```
0
       3
           male
                 22
                        0
                              0
                                 7.8292
                                              Q
                                                                       0
                 38
                                 7.0000
                                              S
                                                        2
                                                                       1
1
       3 female
                        1
                              0
2
                 26
                        0
                                 9.6875
                                              Q
                                                                       0
       2
           male
                                                        1
3
                 35
                        0
                              0
                                 8.6625
                                              S
                                                        1
                                                                       0
       3
           male
                                              S
                                                        3
       3 female
                 35
                        1
                              1 12.2875
                                                                       1
In [15]:
le = LabelEncoder()
train dataset["Sex"] = le.fit_transform(train_dataset["Sex"])
print(train dataset["Sex"])
0
        1
1
        0
2
        0
3
        0
        1
886
       1
        0
887
888
        0
889
       1
890
       1
Name: Sex, Length: 891, dtype: int64
In [16]:
test dataset["Sex"] = le.fit transform(test dataset["Sex"])
print(test dataset["Sex"])
0
        1
1
        0
2
        1
3
        1
4
        0
413
       1
414
       0
415
       1
416
       1
417
       1
Name: Sex, Length: 418, dtype: int64
In [17]:
train dataset["Embarked"] = le.fit transform(train dataset["Embarked"])
print(train dataset["Embarked"])
0
        2
1
        0
2
        2
3
        2
        2
4
886
       2
       2
887
        2
888
        0
889
890
Name: Embarked, Length: 891, dtype: int64
In [18]:
test dataset["Embarked"] = le.fit_transform(test_dataset["Embarked"])
print(test dataset["Embarked"])
```

Fare Embarked FamilySize NormalFamilySize

Pclass

Sex Age SibSp Parch

```
2
       1
3
       2
       2
4
413
       2
414
      0
415
       2
416
       2
417
Name: Embarked, Length: 418, dtype: int64
In [19]:
X train = train dataset.drop("Survived", axis=1)
Y train = train dataset["Survived"]
X test = test dataset.copy()
In [20]:
sc = StandardScaler()
X train = sc.fit transform(X train)
X test = sc.transform(X test)
In [21]:
logreg = LogisticRegression()
logreg.fit(X train, Y train)
log pred = logreg.predict(X test)
acc log = round(logreg.score(X train, Y train) * 100, 2)
acc log
Out[21]:
81.14
In [22]:
knn = KNeighborsClassifier(algorithm='auto', leaf size=16, metric='minkowski',
                            metric_params=None, n_jobs=1, n_neighbors=6, p=2,
                            weights='uniform')
knn.fit(X train, Y train)
knn pred = knn.predict(X test)
acc knn = round(knn.score(X train, Y train) * 100, 2)
acc knn
Out[22]:
84.06
In [23]:
svc = SVC(probability = True, kernel = 'rbf', random state = 0)
svc.fit(X train, Y train)
svc pred = svc.predict(X test)
svc.score(X train, Y train)
acc svc = round(svc.score(X train, Y train) * 100, 2)
acc svc
Out[23]:
83.28
In [24]:
gaussian = GaussianNB()
gaussian.fit(X train, Y train)
gaussian pred = gaussian.predict(X test)
acc gaussian = round(gaussian.score(X train, Y train) * 100, 2)
acc_gaussian
Out[24]:
```

models = pd.DataFrame({

```
In [25]:
# decision tree = DecisionTreeClassifier(max depth = 6)
decision tree = DecisionTreeClassifier()
decision tree.fit(X train, Y train)
dt pred = decision tree.predict(X test)
acc decision tree = round(decision tree.score(X train, Y train) * 100, 2)
acc_decision tree
Out[25]:
98.2
In [26]:
# random forest = RandomForestClassifier(max depth=11, n estimators=50, n jobs=-1, random
random forest = RandomForestClassifier()
random forest.fit(X train, Y train)
rf pred = random forest.predict(X test)
random forest.score(X train, Y train)
acc random forest = round(random forest.score(X train, Y train) * 100, 2)
acc random forest
Out[26]:
98.2
In [27]:
xgb model = xgb.XGBClassifier()
xgb_model.fit(X_train, Y train)
xgb_pred = xgb_model.predict(X_test)
acc_xgb = round(xgb_model.score(X_train, Y_train) * 100, 2)
acc xgb
Out [27]:
96.52
In [28]:
mlp model = MLPClassifier(solver='lbfgs', hidden layer sizes=(11, 11, 11), random state=
1, \max iter = 1000)
mlp model.fit(X train, Y train)
mlp pred = xgb model.predict(X test)
acc mlp = round(mlp model.score(X train, Y train) * 100, 2)
acc mlp
/Library/Frameworks/Python.framework/Versions/3.10/lib/python3.10/site-packages/sklearn/n
eural network/ multilayer perceptron.py:559: ConvergenceWarning: lbfgs failed 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
  self.n_iter_ = _check_optimize_result("lbfgs", opt_res, self.max_iter)
Out[28]:
93.15
In [29]:
```

'Model': ['Support Vector Machines', 'KNN', 'Logistic Regression',

Out[29]:

```
Model Score
3
           Random Forest
                           98.20
6
             Decision Tree
                           98.20
5
                 XGBoost
                           96.52
7
                     MLP
                           93.15
                     KNN
                           84.06
0 Support Vector Machines
                           83.28
        Logistic Regression
                           81.14
4
              Naive Bayes 78.68
```

In [30]:

```
sft voting = VotingClassifier (estimators = [
    ('LogisticRegression', logreg),
    ('SVC', svc),
    ('KNN', knn),
    ('Gaussian', gaussian),
#
     ('perceptron', perceptron),
     ('Linear SVC', linear_svc),
    ('Decision Tree', decision_tree),
    ('RF', random forest),
    ('SGD', sgd),
    ('xgb_model', xgb_model),
    ('mlp model', mlp model)
], weights = [
   1,
    8,
    2,
    1,
    2,
    2,
    2,
], voting = 'soft')
sft voting.fit(X train, Y train)
soft pred = sft voting.predict(X test)
acc_soft = round(sft_voting.score(X_train, Y_train) * 100, 2)
acc soft
/Library/Frameworks/Python.framework/Versions/3.10/lib/python3.10/site-packages/sklearn/n
eural network/ multilayer perceptron.py:559: ConvergenceWarning: lbfgs failed 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
  self.n iter = check optimize result("lbfgs", opt res, self.max iter)
```

Out[30]:

90.91

In [33]:

```
#experiment section
import statistics
```

```
experiment_pred = []
for index in range (0,418):
   list_of_pred = [
        log_pred[index],
        knn pred[index],
        svc pred[index],
       gaussian_pred[index],
       dt pred[index],
       rf pred[index],
       xgb_pred[index],
       mlp pred[index]
         soft pred[index]
     list of pred.sort()
   median_pred = statistics.median(list of pred)
    print(list_of_pred)
    if median pred == 0.5:
       median_pred = knn_pred[index]
         median_pred = svc_pred[index]
    print(median_pred)
    experiment_pred.append(int(median_pred))
```

```
0.0
0.0
0.0
0.0
1.0
0.0
1
1.0
1.0
0.0
0.0
0.0
1.0
0.0
1.0
1.0
0.0
0.0
0.0
1.0
1.0
0.0
1.0
0
1.0
0.0
1.0
0.0
1.0
0.0
0.0
0.0
0
0.0
1.0
0.0
1
1
0.0
0.0
0.0
0.0
0.0
1.0
1.0
0.0
1.0
0.0
```

1.0

1.0 1.0 1.0 0.0 0.0 0.0 0.0 1.0 0.0 0.0 0.0 1.0 1.0 0.0 1.0 1.0 0.0 1.0 10.0 1 1.0 1.0 0.0 1.0 0.0 1.0 0 1 0.0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 1.0 0.0 1.0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 1.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1.0 1.0 1.0 0.0 0.0 1.0 1.0 0.0

1.0 0.0 0.0 1.0 0.0 1.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 1.0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 0.0 1 0.0 0.0 1.0 0.0 0.0 1.0 1.0 0.0 1.0 0.0 0.0 1.0 0.0 0.0 1.0 1.0 0.0 0.0 1.0 0.0 0.0 1.0 1.0 1.0 1.0 0.0 1.0 1.0 0.0 1.0 1.0 0.0 0.0 0.0 0.0

0.0 0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 1.0 1 0.0 1.0 0.0 0.0 0.0 1.0 0 0.0 1.0 0.0 1.0 1.0 1 1.0 0.0 1.0 1.0 0.0 1.0 0.0 0.0 0.0 1.0 0.0 0 0.0 0.0 0.0 1.0 1.0 1.0 1.0 0.0 0.0 0.0 0.0 1.0 0.0 1.0 1.0 0.0 0.0 0.0 0.0 0.0 1.0 0.0 0.0 0.0 1.0 0.0

0.0 0.0 0.0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 0.0 0 0.0 1 0.0 0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1.0 0 0.0 0.0 0.0 0.0 0.0 1.0 1.0 0.0 0.0 0 0.0 0.0 0.0 0.0 1.0 1.0 1.0 0.0 0.0 0.0 0.0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 1.0 0.0 0.0 1.0 0.0 0.0

0.0 0.0 0.0 0.0 0.0 1.0 0.0 1.0 0.0 1.0 0.0 1.0 1.0 0.0 0.0 0 0.0 1.0 0.0 0.0 0 0.0 1.0 1.0 1.0 0.0 0 1.0 0.0 0.0 0.0 1.0 0.0 0.0 1.0 0.0 0.0 0.0 0.0 0.0 0.0 1.0 1.0 0.0 1.0 0.0 0.0 0.0 0.0 0.0 1.0 0.0 0.0 1.0 0.0 1.0 0.0 0.0 1.0 0.0 1.0 0.0 0.0 0 0.0 1.0 1.0 1.0