$MuhammadHassanShah_20P-0025_C_Lab06$

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```
[1]: import numpy as np
  import pandas as pd
  from sklearn import preprocessing

df = pd.read_csv("titanic_train.csv")

# df.head()
  print(df.isna().sum())

# mean_age = df['age'].mean()
# mean_age

# label_encoder = preprocessing.LabelEncoder()
# for column in data:
# data[column] = label_encoder.fit_transform(data[column])

# data.head()
# data.head()
# data.head()
# data.head()
```

passenger_id	0			
pclass	0			
name	0			
sex	0			
age	174			
sibsp	0			
parch	0			
ticket	0			
fare	1			
cabin	659			
embarked	1			
boat	542			
body	777			
home.dest	386			
survived	0			
dtype: int64				

```
[2]: condition = df['body'].notnull()
     d2 = df[condition]
     d2.head()
[2]:
         passenger_id pclass
                                                            name
                                                                    sex
                                                                           age
                                                                                sibsp
                  1083
                                       Olsen, Mr. Henry Margido
                                                                   male
                                                                          28.0
     14
                   602
                             3
                                    Abbott, Mr. Rossmore Edward
                                                                   male
                                                                          16.0
                                                                                    1
     29
                   162
                             1
                                Holverson, Mr. Alexander Oskar
                                                                   male
                                                                          42.0
                                                                                    1
     40
                  1169
                             3
                                   Saether, Mr. Simon Sivertsen
                                                                   male
                                                                         38.5
                                                                                    0
     45
                  1239
                             3
                                   Theobald, Mr. Thomas Leonard
                                                                   male
                                                                         34.0
                                                                                    0
         parch
                             ticket
                                        fare cabin embarked boat
                                                                     body
                                      22.525
     5
             0
                             C 4001
                                                NaN
                                                           S
                                                               NaN
                                                                    173.0
     14
             1
                          C.A. 2673
                                      20.250
                                                NaN
                                                              NaN
                                                                    190.0
     29
                              113789
                                      52.000
                                                NaN
                                                           S
                                                              NaN
                                                                     38.0
     40
             0
                                       7.250
                                                              NaN
                                                                     32.0
                SOTON/O.Q. 3101262
                                                NaN
                                                           S
     45
             0
                             363294
                                       8.050
                                                NaN
                                                           S
                                                              NaN
                                                                    176.0
                    home.dest
                                survived
     5
                          NaN
                                       0
     14
         East Providence, RI
                                       0
     29
                New York, NY
                                       0
     40
                          NaN
                                       0
     45
                          NaN
                                       0
[3]: d2[d2['survived'] == 1].head()
[3]: Empty DataFrame
     Columns: [passenger_id, pclass, name, sex, age, sibsp, parch, ticket, fare,
     cabin, embarked, boat, body, home.dest, survived]
     Index: []
```

1 We can see if a body has some value, it means that the passenger did not survive. Making it Binary attribute

```
[4]: df['body'].fillna(1,inplace=True)
    df['body'] = df['body'].replace(np.nan,0)
    df.head()
```

```
[4]:
        passenger_id pclass
                                                                                name
     0
                 1216
                            3
                                                                 Smyth, Miss. Julia
                            3
     1
                  699
                                                                    Cacic, Mr. Luka
     2
                 1267
                            3
                                Van Impe, Mrs. Jean Baptiste (Rosalie Paula Go...
     3
                  449
                            2
                                             Hocking, Mrs. Elizabeth (Eliza Needs)
     4
                  576
                            2
                                                                    Veal, Mr. James
```

```
sibsp
                               parch
                                       ticket
                                                  fare cabin embarked boat
                                                                                body
           sex
                  age
                  NaN
                            0
                                    0
                                       335432
                                                 7.7333
                                                                       Q
                                                                           13
                                                                                 1.0
     0
        female
                                                           NaN
                                                                       S
                 38.0
                            0
     1
          male
                                    0
                                       315089
                                                 8.6625
                                                           NaN
                                                                          NaN
                                                                                 1.0
                                                                       S
        female
                 30.0
                                                24.1500
                                                                          NaN
                            1
                                    1
                                       345773
                                                           NaN
                                                                                 1.0
     3
        female
                 54.0
                            1
                                    3
                                        29105
                                                23.0000
                                                           NaN
                                                                       S
                                                                            4
                                                                                 1.0
                                                13.0000
          male
                 40.0
                            0
                                    0
                                        28221
                                                           NaN
                                                                          NaN
                                                                                 1.0
                         home.dest
                                    survived
     0
                               NaN
                                             1
     1
                           Croatia
                                             0
     2
                                            0
                               NaN
            Cornwall / Akron, OH
                                            1
        Barre, Co Washington, VT
[5]: condition = df['boat'].notnull()
     d3 = df[condition]
     d3.head(len(d3))
[5]:
          passenger_id pclass
                                                                                  name \
                               3
                                                                   Smyth, Miss. Julia
     0
                   1216
     3
                    449
                               2
                                             Hocking, Mrs. Elizabeth (Eliza Needs)
     7
                               2
                    560
                                                               Sinkkonen, Miss. Anna
     8
                   1079
                               3
                                                                   Ohman, Miss. Velin
                     43
                               1
                                   Bucknell, Mrs. William Robert (Emma Eliza Ward)
     11
     . .
                    •••
                               3
                                                     Touma, Master. Georges Youssef
     840
                   1256
                                   Minahan, Mrs. William Edward (Lillian E Thorpe)
     841
                    208
                               1
                                                           Carr, Miss. Helen "Ellen"
     842
                    709
                               3
     844
                    165
                               1
                                                        Hoyt, Mr. Frederick Maxfield
     847
                    467
                               2
                                                Kantor, Mrs. Sinai (Miriam Sternin)
                                 parch ticket
                                                     fare cabin embarked boat
              sex
                          sibsp
                                                                                  body
                    age
     0
          female
                    NaN
                              0
                                      0
                                         335432
                                                   7.7333
                                                             NaN
                                                                         Q
                                                                              13
                                                                                   1.0
     3
                              1
                                      3
                                                                         S
                                                                               4
          female
                   54.0
                                          29105
                                                  23.0000
                                                             NaN
                                                                                   1.0
     7
          female
                   30.0
                              0
                                      0
                                         250648
                                                                         S
                                                                              10
                                                                                   1.0
                                                  13.0000
                                                             NaN
     8
          female
                   22.0
                              0
                                         347085
                                                   7.7750
                                                             NaN
                                                                         S
                                                                               C
                                                                                   1.0
                                                                         С
          female 60.0
                              0
                                      0
                                          11813
                                                  76.2917
                                                                               8
                                                                                   1.0
     11
                                                             D15
     840
                    7.0
                                           2650
                                                  15.2458
                                                                         С
                                                                              С
                                                                                   1.0
            male
                              1
                                      1
                                                             NaN
     841
          female
                   37.0
                              1
                                      0
                                          19928
                                                  90.0000
                                                             C78
                                                                         Q
                                                                              14
                                                                                   1.0
     842
          female
                   16.0
                              0
                                      0
                                         367231
                                                   7.7500
                                                             NaN
                                                                         Q
                                                                              16
                                                                                   1.0
                                                                         S
     844
            male
                   38.0
                              1
                                      0
                                          19943
                                                  90.0000
                                                             C93
                                                                              D
                                                                                   1.0
          female
                                                  26.0000
                                                             NaN
                                                                         S
                                                                              12
                                                                                   1.0
     847
                   24.0
                                         244367
                                     home.dest
                                                 survived
     0
                                           NaN
                                                         1
     3
                         Cornwall / Akron, OH
                                                         1
```

```
7
              Finland / Washington, DC
                                                 1
8
                                     NaN
                                                  1
                       Philadelphia, PA
11
                                                  1
. .
840
                                     {\tt NaN}
                                                  1
841
                        Fond du Lac, WI
                                                  1
842 Co Longford, Ireland New York, NY
                                                  1
844
           New York, NY / Stamford CT
                                                  1
847
                    Moscow / Bronx, NY
                                                  1
```

[308 rows x 15 columns]

[6]: d3[d3['survived'] == 0].head(len(d3))

[6]:	passeng	er id	nclass							name	\
217	hanneng	19	_	1 Beattie, Mr. Thomso							`
219		1299		3 Yasbeck, Mr. Antoni							
272		968		3 Lindell, Mr. Edvard Bengtsson							
333		969		3 Lindell, Mrs. Edvard Bengtsson (Elin Gerda Per							
415		166	1	_							
443		544	_								
				Renouf, Mr. Peter Henry							
486		655	3	•							
644		921	3					Keere,	Mr. A	rtnur	
	sex	age	sibsp	parch	ticket	fare	cabin	embarked 1	boat	body	\
217	male	36.0	0	0	13050	75.2417	C6	С	Α	1.0	
219	male	27.0	1	0	2659	14.4542	NaN	С	С	1.0	
272	male	36.0	1	0	349910	15.5500	NaN	S	Α	1.0	
333	female	30.0	1	0	349910	15.5500	NaN	S	Α	1.0	
415	male	NaN	0	0	PC 17600	30.6958	NaN	C	14	1.0	
443	male	34.0	1	0	31027	21.0000	NaN	S	12	1.0	
486	male	32.0	1	0	3101278	15.8500	NaN	S	D	1.0	
644	male	NaN	0	0	323592	7.2500	NaN	S	Α	1.0	
home.dest survived											
217			Winnipeg, MN			0					
219			NaN			0					
272			NaN			0					
333					NaN	0					
415		New York, NY									
443	Elizabeth, NJ					0					
486	·										
644		1 3	•		NaN	0					

2 So If a boat was assigned, we still cannot be sure if the passenger survived. But maximum passengers did.

```
[7]: df['boat'] = df['boat'].replace(np.nan,'0')
     df.head()
        passenger_id pclass
[7]:
                                                                               name \
                 1216
                                                                Smyth, Miss. Julia
                 699
     1
                            3
                                                                    Cacic, Mr. Luka
     2
                 1267
                               Van Impe, Mrs. Jean Baptiste (Rosalie Paula Go...
     3
                  449
                            2
                                            Hocking, Mrs. Elizabeth (Eliza Needs)
                 576
                            2
                                                                    Veal, Mr. James
                              parch
                                                 fare cabin embarked boat
                                                                             body \
           sex
                       sibsp
                                     ticket
                 age
     0
        female
                 NaN
                           0
                                  0
                                     335432
                                               7.7333
                                                         NaN
                                                                              1.0
          male
                38.0
                           0
                                                                    S
                                                                          0
     1
                                  0
                                     315089
                                               8.6625
                                                         NaN
                                                                              1.0
                30.0
                                                                    S
        female
                                              24.1500
                                                         NaN
                                                                              1.0
                                  1
                                      345773
                54.0
                                              23.0000
                                                         NaN
                                                                    S
     3
        female
                           1
                                  3
                                       29105
                                                                              1.0
          male
                40.0
                                       28221
                                              13.0000
                                                         NaN
                                                                              1.0
                        home.dest survived
     0
                              NaN
                                           1
     1
                          Croatia
                                           0
     2
                              NaN
                                           0
            Cornwall / Akron, OH
     3
       Barre, Co Washington, VT
```

3 Home destination, passenger_id, p_class is irrelavant, so we drop it

```
[8]: df.drop('home.dest',axis = 1,inplace = True)
    df.drop('passenger_id',axis = 1,inplace = True)
    df.drop('name',axis = 1,inplace = True)
    df.head()
```

```
[8]:
        pclass
                                 sibsp
                                        parch
                                                ticket
                                                             fare cabin embarked boat
                     sex
                           age
                           NaN
                                     0
                                                335432
                                                          7.7333
                                                                    NaN
     0
              3
                 female
                                             0
                                                                                     13
     1
              3
                          38.0
                                     0
                                                315089
                                                          8.6625
                                                                    NaN
                                                                                 S
                                                                                      0
                   male
     2
                                                                    NaN
                                                                                 S
                 female
                          30.0
                                     1
                                             1
                                                345773
                                                         24.1500
                                                                                      0
     3
              2
                 female
                          54.0
                                     1
                                             3
                                                 29105
                                                         23.0000
                                                                    NaN
                                                                                 S
                                                                                      4
              2
                   male
                          40.0
                                     0
                                                 28221
                                                         13.0000
                                                                    {\tt NaN}
                                                                                      0
```

```
body survived
0 1.0 1
1 1.0 0
```

```
3
          1.0
                       1
          1.0
                       0
 [9]: print(df.isna().sum())
                    0
     pclass
                    0
     sex
     age
                  174
                    0
     sibsp
     parch
                    0
     ticket
                    0
     fare
                    1
     cabin
                  659
     embarked
                    1
     boat
                    0
     body
                    0
     survived
     dtype: int64
[10]: df['cabin'] = df['cabin'].replace(np.nan,'0')
      df['embarked'] = df['embarked'].replace(np.nan,'0')
      df['fare'] = df['fare'].replace(np.nan,0)
      df.head()
[10]:
         pclass
                                       parch ticket
                                                           fare cabin embarked boat
                     sex
                           age
                                sibsp
      0
              3
                           NaN
                                               335432
                                                         7.7333
                                                                     0
                 female
                                     0
                                                                               Q
                                                                                   13
              3
                                                         8.6625
                                                                               S
      1
                    male
                          38.0
                                     0
                                               315089
                                                                     0
                                                                                    0
      2
              3
                 female
                          30.0
                                     1
                                            1 345773
                                                        24.1500
                                                                     0
                                                                               S
                                                                                    0
      3
              2
                                                 29105
                                                        23.0000
                                                                     0
                                                                               S
                                                                                    4
                 female
                          54.0
                                     1
                                            3
              2
                          40.0
                                     0
                                            0
                                                 28221
                                                        13.0000
                                                                     0
                                                                               S
                                                                                    0
                    male
              survived
         body
          1.0
      0
                       1
          1.0
                       0
      1
      2
          1.0
                       0
      3
          1.0
                       1
                       0
          1.0
[11]: df['age'] = df['age'].replace(np.nan,df['age'].mean())
      df.head()
[11]:
         pclass
                                             parch
                                                     ticket
                                                                 fare cabin embarked \
                     sex
                                      sibsp
                                 age
      0
              3
                 female
                          29.519847
                                          0
                                                  0
                                                     335432
                                                               7.7333
                                                                          0
                                                                                    Q
      1
              3
                    male 38.000000
                                          0
                                                     315089
                                                               8.6625
                                                                          0
                                                                                    S
                                                  0
      2
              3
                female
                          30.000000
                                          1
                                                     345773
                                                             24.1500
                                                                          0
                                                                                    S
                                                  1
      3
              2
                 female 54.000000
                                          1
                                                  3
                                                      29105
                                                            23.0000
                                                                          0
                                                                                    S
```

1.0

0

2

```
4
              2
                   male 40.000000
                                        0
                                                    28221 13.0000
                                                                       0
                                                                                 S
        boat
              body survived
               1.0
      0
          13
                           1
      1
           0
               1.0
                           0
      2
                           0
           0
               1.0
      3
           4
               1.0
                           1
      4
               1.0
                           0
           0
[12]: print(df.isna().sum())
                 0
     pclass
     sex
                 0
                 0
     age
     sibsp
                 0
     parch
                 0
     ticket
                 0
     fare
                 0
     cabin
                 0
     embarked
                 0
     boat
                 0
     body
                 0
     survived
     dtype: int64
[13]: import matplotlib.pyplot as mp
      from sklearn.model_selection import train_test_split
      from sklearn.neural_network import MLPClassifier
      from sklearn.metrics import accuracy_score, precision_score, recall_score, u
       ⊶f1_score
      y = df['survived']
      df.drop('survived',axis = 1,inplace = True)
      X_test, X_train, y_test, y_train = train_test_split(df,y,test_size=0.2,__
       →random_state=42, shuffle=True)
      # X_train.hist(bins="auto",figsize=(9,7),grid=False)
      df.head()
[13]:
                                                              fare cabin embarked \
         pclass
                                    sibsp parch ticket
                    sex
                               age
                 female 29.519847
      0
              3
                                         0
                                                0
                                                   335432
                                                            7.7333
                                                                       0
                                                                                 Q
                                                                                 S
      1
              3
                                                                        0
                   male 38.000000
                                         0
                                                0 315089
                                                            8.6625
      2
              3 female 30.000000
                                         1
                                                1 345773 24.1500
                                                                        0
                                                                                 S
                                                                                 S
              2 female 54.000000
      3
                                         1
                                                3
                                                   29105 23.0000
                                                                        0
                                                                                 S
                   male 40.000000
                                                    28221
                                                           13.0000
                                                                        0
        boat body
```

```
1.0
      0
           13
      1
            0
                1.0
      2
            0
                1.0
      3
            4
                1.0
      4
            0
                1.0
[20]: label_encoder = preprocessing.LabelEncoder()
      for column in df:
           df[column] = label_encoder.fit_transform(df[column])
[21]:
      df.head(len(df))
[21]:
            pclass
                     sex
                          age
                                sibsp
                                        parch ticket
                                                         fare
                                                                cabin
                                                                        embarked
                                                                                   boat
                                                                                          body
                  2
                       0
                            40
                                                                    0
                                                                                2
                                                                                      5
      0
                                     0
                                             0
                                                   283
                                                           25
                                                                                             0
      1
                  2
                       1
                            51
                                     0
                                             0
                                                   257
                                                           48
                                                                    0
                                                                                3
                                                                                      0
                                                                                             0
      2
                  2
                       0
                            41
                                     1
                                             1
                                                   307
                                                          124
                                                                    0
                                                                                3
                                                                                      0
                                                                                             0
      3
                                                   237
                                                          120
                                                                                3
                  1
                       0
                            70
                                     1
                                             3
                                                                    0
                                                                                     14
                                                                                             0
      4
                  1
                       1
                            54
                                     0
                                             0
                                                   228
                                                           77
                                                                    0
                                                                                3
                                                                                      0
                                                                                             0
                                                           •••
                                                                                3
      845
                  0
                       1
                            71
                                     0
                                             0
                                                   488
                                                          172
                                                                   62
                                                                                      0
                                                                                             0
      846
                  0
                            75
                                     0
                                             0
                                                    58
                                                          143
                                                                   23
                                                                                      0
                                                                                            54
                       1
                                                                                1
      847
                       0
                            32
                                             0
                                                   145
                                                                    0
                                                                                3
                                                                                      4
                  1
                                     1
                                                          130
                                                                                             0
      848
                  2
                       0
                             8
                                     1
                                             1
                                                   630
                                                           80
                                                                    0
                                                                                3
                                                                                      0
                                                                                             0
      849
                                     0
                                             0
                                                   160
                                                           77
                                                                    0
                                                                                3
                                                                                      0
                                                                                             6
                  1
                       1
                            68
      [850 rows x 11 columns]
[22]: print(df.dtypes)
                   int64
      pclass
      sex
                   int64
      age
                   int64
                   int64
      sibsp
      parch
                   int64
     ticket
                   int64
      fare
                   int64
      cabin
                   int64
      embarked
                   int64
      boat
                   int64
      body
                   int64
      dtype: object
[23]: |mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes = u
       \hookrightarrow (3,3),random_state = 1)
      mlp.fit(X_train,y_train)
      y_pred = mlp.predict(X_test)
      y_score = mlp.score(X_test,y_test)
```

```
accuracy = accuracy_score(y_test, y_pred)
precision = precision_score(y_test, y_pred)
recall = recall_score(y_test, y_pred)
f1 = f1_score(y_test, y_pred)
print("3,3 neurons")
print('Accuracy:', accuracy)
print('Precision:', precision)
print('Recall:', recall)
print('F1 score:', f1)
```

```
ValueError
                                           Traceback (most recent call last)
/tmp/ipykernel 3169/1023893291.py in <module>
      1 mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes = 1
\hookrightarrow (3,3), random state = 1)
---> 2 mlp.fit(X_train,y_train)
      3 y_pred = mlp.predict(X_test)
      4 y_score = mlp.score(X_test,y_test)
      5 accuracy = accuracy_score(y_test, y_pred)
~/anaconda3/lib/python3.9/site-packages/sklearn/neural_network/
 →_multilayer_perceptron.py in fit(self, X, y)
    750
                    Returns a trained MLP model.
    751
--> 752
                return self._fit(X, y, incremental=False)
    753
    754
            def _check_solver(self):
~/anaconda3/lib/python3.9/site-packages/sklearn/neural_network/
 → multilayer perceptron.py in fit(self, X, y, incremental)
    391
                )
    392
--> 393
                X, y = self._validate_input(X, y, incremental, reset=first_pass
    394
    395
                n_samples, n_features = X.shape
~/anaconda3/lib/python3.9/site-packages/sklearn/neural_network/
 → multilayer_perceptron.py in _validate_input(self, X, y, incremental, reset)
   1098
   1099
            def _validate_input(self, X, y, incremental, reset):
-> 1100
                X, y = self. validate data(
                    Χ,
   1101
   1102
                    у,
~/anaconda3/lib/python3.9/site-packages/sklearn/base.py in _validate_data(self,
 →X, y, reset, validate_separately, **check_params)
```

```
579
                                                                      y = check_array(y, **check_y_params)
            580
                                                         else:
                                                                     X, y = check_X_y(X, y, **check_params)
 --> 581
            582
                                                          out = X, y
            583
~/anaconda3/lib/python3.9/site-packages/sklearn/utils/validation.py in__
   →check_X_y(X, y, accept_sparse, accept_large_sparse, dtype, order, copy, office_all_finite, ensure_2d, allow_nd, multi_output, ensure_min_samples, order, copy, office_all_finite, ensure_min_samples, order, copy, 
   ⇔ensure min features, y numeric, estimator)
            962
                                              raise ValueError("y cannot be None")
           963
--> 964
                                  X = check array(
            965
                                              Χ,
            966
                                              accept_sparse=accept_sparse,
~/anaconda3/lib/python3.9/site-packages/sklearn/utils/validation.py in_
   ⇔check_array(array, accept_sparse, accept_large_sparse, dtype, order, copy, u

⇔force_all_finite, ensure_2d, allow_nd, ensure_min_samples, u
   ⇔ensure min features, estimator)
           744
                                                                                 array = array.astype(dtype, casting="unsafe", __
   ⇔copy=False)
           745
                                                                     else:
 --> 746
                                                                                 array = np.asarray(array, order=order, dtype=dtype)
                                                          except ComplexWarning as complex_warning:
           747
           748
                                                                     raise ValueError(
~/anaconda3/lib/python3.9/site-packages/pandas/core/generic.py in_

    array_ (self, dtype)

        2062
         2063
                                   def __array__(self, dtype: npt.DTypeLike | None = None) -> np.
   →ndarray:
-> 2064
                                              return np.asarray(self._values, dtype=dtype)
         2065
         2066
                                   def __array_wrap__(
ValueError: could not convert string to float: 'male'
```

- 4 For some reasons, it gives this behaviour. I restarted the kernel and it becomes fine. But when I do it all over again it does the same thing. I'll take (3,4) as initial model and continue on.
- 5 Even the all the datatypes are int, it still does this. IDK why

```
[18]: print(df.dtypes)
     pclass
                  int64
     sex
                  int64
                  int64
     age
     sibsp
                  int64
     parch
                  int64
     ticket
                  int64
     fare
                  int64
                  int64
     cabin
                  int64
     embarked
     boat
                  int64
     body
                  int64
     dtype: object
[29]: |mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes = ___
      (3,4), random state = 1)
      mlp.fit(X_train,y_train)
      y_pred = mlp.predict(X_test)
      y_score = mlp.score(X_test,y_test)
      accuracy = accuracy_score(y_test, y_pred)
      precision = precision_score(y_test, y_pred)
      recall = recall_score(y_test, y_pred)
      f1 = f1_score(y_test, y_pred)
      print("3,4 neurons")
      print('Accuracy:', accuracy)
      print('Precision:', precision)
      print('Recall:', recall)
      print('F1 score:', f1)
     3,4 neurons
     Accuracy: 0.6529411764705882
     Precision: 0.5248618784530387
     Recall: 0.3877551020408163
     F1 score: 0.4460093896713615
 []:
[30]: |mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes = ___
       \hookrightarrow (4,4),random_state = 1)
```

```
mlp.fit(X_train,y_train)
      y_pred = mlp.predict(X_test)
      y_score = mlp.score(X_test,y_test)
      accuracy = accuracy_score(y_test, y_pred)
      precision = precision_score(y_test, y_pred)
      recall = recall_score(y_test, y_pred)
      f1 = f1_score(y_test, y_pred)
      print("4,4 neurons")
      print('Accuracy:', accuracy)
      print('Precision:', precision)
      print('Recall:', recall)
      print('F1 score:', f1)
     4,4 neurons
     Accuracy: 0.6397058823529411
     Precision: 0.0
     Recall: 0.0
     F1 score: 0.0
     /home/h/anaconda3/lib/python3.9/site-
     packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning:
     Precision is ill-defined and being set to 0.0 due to no predicted samples. Use
     `zero_division` parameter to control this behavior.
       _warn_prf(average, modifier, msg_start, len(result))
[32]: mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes = ___
       \hookrightarrow (4,5),random_state = 1)
      mlp.fit(X_train,y_train)
      y_pred = mlp.predict(X_test)
      y_score = mlp.score(X_test,y_test)
      accuracy = accuracy_score(y_test, y_pred)
      precision = precision_score(y_test, y_pred)
      recall = recall_score(y_test, y_pred)
      f1 = f1_score(y_test, y_pred)
      print("4,5 neurons")
      print('Accuracy:', accuracy)
      print('Precision:', precision)
      print('Recall:', recall)
      print('F1 score:', f1)
     4,5 neurons
     Accuracy: 0.6397058823529411
     Precision: 0.0
     Recall: 0.0
     F1 score: 0.0
```

packages/sklearn/metrics/_classification.py:1318: UndefinedMetricWarning:

/home/h/anaconda3/lib/python3.9/site-

```
Precision is ill-defined and being set to 0.0 due to no predicted samples. Use `zero_division` parameter to control this behavior.
_warn_prf(average, modifier, msg_start, len(result))
```

5,5 neurons

Accuracy: 0.6720588235294118 Precision: 0.6122448979591837 Recall: 0.24489795918367346 F1 score: 0.3498542274052478

6,6 neurons

Accuracy: 0.9720588235294118 Precision: 0.9669421487603306 Recall: 0.9551020408163265 F1 score: 0.9609856262833676

/home/h/anaconda3/lib/python3.9/sitepackages/sklearn/neural_network/_multilayer_perceptron.py:549:

```
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)
[36]: mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes =__
      (7,7), random state = 1)
      mlp.fit(X_train,y_train)
      y_pred = mlp.predict(X_test)
      y_score = mlp.score(X_test,y_test)
      accuracy = accuracy score(y test, y pred)
      precision = precision_score(y_test, y_pred)
      recall = recall_score(y_test, y_pred)
      f1 = f1_score(y_test, y_pred)
      print("7,7 neurons")
      print('Accuracy:', accuracy)
      print('Precision:', precision)
      print('Recall:', recall)
      print('F1 score:', f1)
     7,7 neurons
     Accuracy: 0.9705882352941176
     Precision: 0.9591836734693877
     Recall: 0.9591836734693877
     F1 score: 0.9591836734693877
     /home/h/anaconda3/lib/python3.9/site-
     packages/sklearn/neural_network/_multilayer_perceptron.py:549:
     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)
[37]: |mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes = u
       \hookrightarrow (8,8),random_state = 1)
      mlp.fit(X_train,y_train)
      y_pred = mlp.predict(X_test)
      y_score = mlp.score(X_test,y_test)
      accuracy = accuracy_score(y_test, y_pred)
      precision = precision score(y test, y pred)
      recall = recall_score(y_test, y_pred)
      f1 = f1_score(y_test, y_pred)
      print("8,8 neurons")
      print('Accuracy:', accuracy)
```

```
print('Precision:', precision)
      print('Recall:', recall)
      print('F1 score:', f1)
     8,8 neurons
     Accuracy: 0.9705882352941176
     Precision: 0.9629629629629
     Recall: 0.9551020408163265
     F1 score: 0.959016393442623
     /home/h/anaconda3/lib/python3.9/site-
     packages/sklearn/neural_network/_multilayer_perceptron.py:549:
     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)
[39]: |mlp = MLPClassifier(solver="lbfgs",alpha = 1e-5,hidden_layer_sizes = ___
      \hookrightarrow (10,10),random_state = 1)
      mlp.fit(X_train,y_train)
      y pred = mlp.predict(X test)
      y_score = mlp.score(X_test,y_test)
      accuracy = accuracy score(y test, y pred)
      precision = precision_score(y_test, y_pred)
      recall = recall_score(y_test, y_pred)
      f1 = f1_score(y_test, y_pred)
      print("10,10 neurons")
      print('Accuracy:', accuracy)
      print('Precision:', precision)
      print('Recall:', recall)
      print('F1 score:', f1)
     10,10 neurons
     Accuracy: 0.9485294117647058
     Precision: 0.926829268292683
     Recall: 0.9306122448979591
     F1 score: 0.9287169042769857
     /home/h/anaconda3/lib/python3.9/site-
     packages/sklearn/neural network/ multilayer perceptron.py:549:
     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)
```

4,4 neurons

Accuracy: 0.7573529411764706 Precision: 0.8225806451612904 Recall: 0.4163265306122449 F1 score: 0.5528455284552846

6 We are getting maximum accuracy at (8,8)

8,8 neurons

Accuracy: 0.9705882352941176 Precision: 0.9629629629629629 Recall: 0.9551020408163265 F1 score: 0.959016393442623

/home/h/anaconda3/lib/python3.9/site-

packages/sklearn/neural_network/ multilayer_perceptron.py:549:

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)
```

- 7 Initial = (3,4)
- 8 Fine Tuned = (8,8)

```
[26]: Accuracy Precision Recall F1_score Initial 0.652941 0.524862 0.387755 0.446009 Final 0.970588 0.962963 0.955102 0.959016
```

The initial model have .65 accuracy which is very low, after hit and trial using different layers, we can see improvement. But it doesn't always means the bigger the better. I did, and (9.9) and (10,10), we were decreasing in accuracy. So w