```
import pandas as pd
import numpy as np
from sklearn import linear model
from sklearn.model selection import train test split
from sklearn import metrics
data = pd.read csv('iris.csv')
data.head(10)
data.shape
(150, 5)
X = data.drop('Name', axis=1)
y = data['Name']
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.20)
logr = linear model.LogisticRegression()
logr.fit(X train,y train)
#predict if tumor is cancerous where the size is 3.46mm:
predicted = logr.predict([[5.0,3.6,1.4,0.2]])
print(predicted)
y pred = logr.predict(X test)
print(X test,y pred )
['Iris-setosa']
                 SepalWidth PetalLength
                                             PetalWidth
     SepalLength
10
             5.4
                          3.7
                                       1.5
                                                    0.2
                                       4.1
67
             5.8
                          2.7
                                                    1.0
                                        5.3
111
             6.4
                          2.7
                                                    1.9
             5.6
                                       4.5
                                                    1.5
66
                          3.0
15
             5.7
                          4.4
                                       1.5
                                                    0.4
99
             5.7
                          2.8
                                       4.1
                                                    1.3
             5.6
                                       4.1
88
                          3.0
                                                    1.3
112
             6.8
                          3.0
                                       5.5
                                                    2.1
             6.0
                          3.0
                                       4.8
                                                    1.8
138
97
             6.2
                          2.9
                                       4.3
                                                    1.3
64
             5.6
                          2.9
                                       3.6
                                                    1.3
3
             4.6
                          3.1
                                       1.5
                                                    0.2
35
             5.0
                          3.2
                                        1.2
                                                    0.2
13
             4.3
                          3.0
                                        1.1
                                                    0.1
45
             4.8
                          3.0
                                                    0.3
                                        1.4
100
             6.3
                          3.3
                                       6.0
                                                    2.5
20
             5.4
                                                    0.2
                          3.4
                                       1.7
84
             5.4
                          3.0
                                       4.5
                                                    1.5
117
             7.7
                          3.8
                                       6.7
                                                    2.2
93
             5.0
                          2.3
                                       3.3
                                                    1.0
             5.1
39
                          3.4
                                       1.5
                                                    0.2
```

```
58
             6.6
                         2.9
                                       4.6
                                                   1.3
130
             7.4
                          2.8
                                       6.1
                                                   1.9
14
             5.8
                         4.0
                                       1.2
                                                   0.2
125
             7.2
                         3.2
                                       6.0
                                                   1.8
7
             5.0
                         3.4
                                       1.5
                                                   0.2
135
             7.7
                         3.0
                                       6.1
                                                   2.3
26
             5.0
                                                   0.4
                         3.4
                                       1.6
79
             5.7
                         2.6
                                       3.5
                                                   1.0
1
             4.9
                         3.0
                                       1.4
                                                   0.2 ['Iris-setosa'
'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor'
 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica'
 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica' 'Iris-
setosa'
 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor' 'Iris-setosa'
 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa' 'Iris-virginica'
 'Iris-setosa' 'Iris-virginica' 'Iris-setosa' 'Iris-versicolor'
 'Iris-setosa'l
C:\Users\MGM\anaconda3\Lib\site-packages\sklearn\base.py:493:
UserWarning: X does not have valid feature names, but
LogisticRegression was fitted with feature names
 warnings.warn(
from sklearn.metrics import classification report, confusion matrix
print(confusion matrix(y test, y pred))
print(classification_report(y_test, y_pred))
[[12 0 0]
 [0 10 0]
 [0 \quad 0 \quad 8]]
                 precision recall f1-score
                                                  support
    Iris-setosa
                      1.00
                                 1.00
                                           1.00
                                                        12
Iris-versicolor
                      1.00
                                 1.00
                                           1.00
                                                        10
                                                        8
 Iris-virginica
                      1.00
                                 1.00
                                           1.00
       accuracy
                                           1.00
                                                        30
      macro avg
                      1.00
                                 1.00
                                           1.00
                                                        30
   weighted avg
                      1.00
                                 1.00
                                           1.00
                                                       30
X = data.drop('Name', axis=1)
y = data['Name']
from sklearn.model selection import train test split
X train, X test, y train, y test = train test split(X, y, test size =
0.30)
logr = linear model.LogisticRegression()
logr.fit(X train,y train)
```

```
#predict if tumor is cancerous where the size is 3.46mm:
predicted = logr.predict([[5.0,3.6,1.4,0.2]])
print(predicted)
y pred = logr.predict(X test)
print(X_test,y_pred )
['Iris-setosa']
     SepalLength SepalWidth PetalLength
                                             PetalWidth
10
              5.4
                          3.7
                                        1.5
                                                     0.2
             4.9
34
                          3.1
                                        1.5
                                                     0.1
51
             6.4
                          3.2
                                        4.5
                                                     1.5
8
             4.4
                          2.9
                                        1.4
                                                     0.2
31
             5.4
                          3.4
                                        1.5
                                                     0.4
             7.7
                          2.6
                                                     2.3
118
                                        6.9
48
             5.3
                          3.7
                                        1.5
                                                     0.2
90
             5.5
                          2.6
                                        4.4
                                                     1.2
1
             4.9
                          3.0
                                        1.4
                                                     0.2
37
             4.9
                          3.1
                                        1.5
                                                     0.1
21
             5.1
                          3.7
                                        1.5
                                                     0.4
30
             4.8
                          3.1
                                        1.6
                                                     0.2
             7.3
107
                          2.9
                                        6.3
                                                     1.8
91
             6.1
                                                     1.4
                          3.0
                                        4.6
15
             5.7
                          4.4
                                        1.5
                                                     0.4
54
             6.5
                          2.8
                                        4.6
                                                     1.5
74
             6.4
                          2.9
                                        4.3
                                                     1.3
109
             7.2
                          3.6
                                        6.1
                                                     2.5
             5.4
                                                     0.4
16
                          3.9
                                        1.3
43
             5.0
                          3.5
                                        1.6
                                                     0.6
58
             6.6
                          2.9
                                        4.6
                                                     1.3
126
             6.2
                          2.8
                                        4.8
                                                     1.8
             5.8
67
                          2.7
                                        4.1
                                                     1.0
116
             6.5
                          3.0
                                        5.5
                                                     1.8
52
             6.9
                          3.1
                                        4.9
                                                     1.5
             7.9
                                        6.4
                                                     2.0
131
                          3.8
25
             5.0
                          3.0
                                        1.6
                                                     0.2
57
             4.9
                          2.4
                                        3.3
                                                     1.0
             5.5
                          2.4
                                        3.8
80
                                                     1.1
99
             5.7
                          2.8
                                        4.1
                                                     1.3
73
             6.1
                          2.8
                                        4.7
                                                     1.2
             5.2
                          4.1
32
                                        1.5
                                                     0.1
115
             6.4
                          3.2
                                        5.3
                                                     2.3
             5.4
                          3.0
                                        4.5
                                                     1.5
84
138
             6.0
                          3.0
                                        4.8
                                                     1.8
                                        5.0
77
             6.7
                          3.0
                                                     1.7
20
             5.4
                          3.4
                                        1.7
                                                     0.2
23
                          3.3
                                                     0.5
             5.1
                                        1.7
50
                          3.2
                                        4.7
             7.0
                                                     1.4
128
             6.4
                          2.8
                                        5.6
                                                     2.1
124
              6.7
                          3.3
                                        5.7
                                                     2.1
```

```
46
             5.1
                         3.8
                                                   0.2
                                       1.6
             7.7
135
                         3.0
                                       6.1
                                                   2.3
62
             6.0
                         2.2
                                       4.0
                                                   1.0
             7.7
122
                          2.8
                                       6.7
                                                   2.0 ['Iris-setosa'
'Iris-setosa' 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa'
 'Iris-virginica' 'Iris-setosa' 'Iris-versicolor' 'Iris-setosa'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica'
 'Iris-versicolor' 'Iris-setosa' 'Iris-versicolor' 'Īris-versicolor' 'Iris-versicolor' 'Iris-versicolor'
 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica' 'Iris-versicolor'
 'Iris-virginica' 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor'
 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica'
 'Iris-versicolor' 'Iris-virginica' 'Iris-virginica' 'Iris-setosa'
 'Iris-setosa' 'Iris-versicolor' 'Iris-virginica' 'Iris-virginica'
 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor' 'Iris-virginica']
C:\Users\MGM\anaconda3\Lib\site-packages\sklearn\base.py:493:
UserWarning: X does not have valid feature names, but
LogisticRegression was fitted with feature names
 warnings.warn(
from sklearn.metrics import classification report, confusion matrix
print(confusion matrix(y test, y pred))
print(classification report(y test, y pred))
[[17 0 0]
 [ 0 15 1]
 [ 0 0 12]]
                 precision recall f1-score support
                      1.00
                                 1.00
                                           1.00
                                                        17
    Iris-setosa
Iris-versicolor
                      1.00
                                 0.94
                                           0.97
                                                        16
Iris-virginica
                      0.92
                                 1.00
                                           0.96
                                                       12
                                           0.98
                                                       45
       accuracy
      macro avg
                      0.97
                                 0.98
                                           0.98
                                                        45
                      0.98
                                           0.98
                                                       45
   weighted avg
                                 0.98
X = data.drop('Name', axis=1)
y = data['Name']
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
logr = linear model.LogisticRegression()
logr.fit(X train,y train)
#predict if tumor is cancerous where the size is 3.46mm:
predicted = logr.predict([[5.0,3.6,1.4,0.2]])
```

```
print(predicted)
y pred = logr.predict(X test)
print(X_test,y_pred )
['Iris-setosa']
                   SepalWidth PetalLength
                                                PetalWidth
     SepalLength
38
              4.4
                            3.0
                                          1.3
                                                        0.2
122
              7.7
                            2.8
                                          6.7
                                                        2.0
127
              6.1
                            3.0
                                          4.9
                                                        1.8
95
              5.7
                            3.0
                                          4.2
                                                        1.2
149
              5.9
                            3.0
                                          5.1
                                                        1.8
              4.9
34
                            3.1
                                          1.5
                                                        0.1
7
              5.0
                                          1.5
                                                        0.2
                            3.4
67
              5.8
                            2.7
                                          4.1
                                                        1.0
2
              4.7
                            3.2
                                          1.3
                                                        0.2
14
              5.8
                           4.0
                                          1.2
                                                        0.2
28
              5.2
                            3.4
                                          1.4
                                                        0.2
97
              6.2
                            2.9
                                          4.3
                                                        1.3
31
              5.4
                                          1.5
                            3.4
                                                        0.4
30
              4.8
                            3.1
                                          1.6
                                                        0.2
73
                            2.8
                                          4.7
              6.1
                                                        1.2
146
              6.3
                            2.5
                                          5.0
                                                        1.9
17
              5.1
                            3.5
                                          1.4
                                                        0.3
72
              6.3
                            2.5
                                          4.9
                                                        1.5
62
                            2.2
                                          4.0
                                                        1.0
              6.0
45
              4.8
                            3.0
                                          1.4
                                                        0.3
142
              5.8
                            2.7
                                          5.1
                                                        1.9
78
                            2.9
                                          4.5
                                                        1.5
              6.0
52
              6.9
                            3.1
                                          4.9
                                                        1.5
91
              6.1
                            3.0
                                          4.6
                                                        1.4
43
                            3.5
              5.0
                                          1.6
                                                        0.6
1
              4.9
                            3.0
                                          1.4
                                                        0.2
22
              4.6
                            3.6
                                          1.0
                                                        0.2
9
              4.9
                            3.1
                                          1.5
                                                        0.1
133
              6.3
                            2.8
                                          5.1
                                                        1.5
              5.5
                            3.5
                                                        0.2
36
                                          1.3
                            3.1
141
              6.9
                                          5.1
                                                        2.3
48
              5.3
                            3.7
                                          1.5
                                                        0.2
39
              5.1
                            3.4
                                          1.5
                                                        0.2
80
              5.5
                            2.4
                                          3.8
                                                        1.1
53
              5.5
                            2.3
                                          4.0
                                                        1.3
18
              5.7
                            3.8
                                          1.7
                                                        0.3
145
              6.7
                                          5.2
                                                        2.3
                            3.0
88
              5.6
                            3.0
                                          4.1
                                                        1.3
94
              5.6
                            2.7
                                          4.2
                                                        1.3
47
              4.6
                            3.2
                                          1.4
                                                        0.2
40
              5.0
                            3.5
                                          1.3
                                                        0.3
59
              5.2
                            2.7
                                          3.9
                                                        1.4
54
              6.5
                            2.8
                                          4.6
                                                        1.5
13
              4.3
                            3.0
                                                        0.1
                                          1.1
```

```
83
             6.0
                          2.7
                                                   1.6
                                       5.1
                          3.3
49
             5.0
                                       1.4
                                                   0.2
111
             6.4
                          2.7
                                       5.3
                                                   1.9
35
             5.0
                          3.2
                                       1.2
                                                   0.2
5
             5.4
                         3.9
                                       1.7
                                                   0.4
118
             7.7
                         2.6
                                       6.9
                                                   2.3
12
             4.8
                         3.0
                                       1.4
                                                   0.1
32
             5.2
                         4.1
                                       1.5
                                                   0.1
10
             5.4
                         3.7
                                       1.5
                                                   0.2
102
             7.1
                         3.0
                                       5.9
                                                   2.1
                         2.5
113
             5.7
                                       5.0
                                                   2.0
                                       5.2
147
             6.5
                          3.0
                                                   2.0
             6.7
                          3.1
                                       5.6
140
                                                   2.4
79
             5.7
                          2.6
                                       3.5
                                                   1.0
114
             5.8
                          2.8
                                       5.1
                                                   2.4
             6.1
63
                         2.9
                                       4.7
                                                   1.4 ['Iris-setosa'
'Iris-virginica' 'Iris-virginica' 'Iris-versicolor'
 'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor' 'Iris-
setosa'
 'Iris-setosa' 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa'
 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica'
 'Iris-versicolor' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica' 'Iris-
setosa'
 'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor'
 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor'
 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa' 'Iris-versicolor'
 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-setosa'
 'Iris-virginica' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica'
 'Iris-virginica' 'Iris-virginica' 'Iris-virginica' 'Iris-versicolor'
 'Iris-virginica' 'Iris-versicolor']
C:\Users\MGM\anaconda3\Lib\site-packages\sklearn\base.py:493:
UserWarning: X does not have valid feature names, but
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 warnings.warn(
from sklearn.metrics import classification report, confusion matrix
print(confusion_matrix(y_test, y_pred))
print(classification report(y test, y pred))
[[27 0 0]
 [ 0 17 1]
 [0 \quad 0 \quad 15]
                 precision recall f1-score
                                                  support
    Iris-setosa
                      1.00
                                 1.00
                                           1.00
                                                       27
```

```
Iris-versicolor
                      1.00
                                 0.94
                                           0.97
                                                       18
                      0.94
                                 1.00
                                                       15
Iris-virginica
                                           0.97
                                           0.98
                                                       60
       accuracy
                      0.98
                                 0.98
                                           0.98
                                                       60
      macro avq
                      0.98
                                 0.98
                                           0.98
                                                       60
  weighted avg
X = data.drop('Name', axis=1)
v = data['Name']
from sklearn.model selection import train test split
X_train, X_test, y_train, y_test = train_test_split(X, y, test_size =
0.50)
logr = linear model.LogisticRegression()
logr.fit(X_train,y_train)
#predict if tumor is cancerous where the size is 3.46mm:
predicted = logr.predict([[5.0,3.6,1.4,0.2]])
print(predicted)
y pred = logr.predict(X test)
print(X test,y pred )
['Iris-setosa']
     SepalLength SepalWidth PetalLength
                                            PetalWidth
72
             6.3
                         2.5
                                       4.9
                                                   1.5
                         3.3
                                       5.7
124
             6.7
                                                   2.1
11
             4.8
                         3.4
                                       1.6
                                                   0.2
18
             5.7
                         3.8
                                       1.7
                                                   0.3
130
             7.4
                         2.8
                                       6.1
                                                   1.9
                         . . .
             . . .
                                       . . .
39
             5.1
                                       1.5
                                                   0.2
                         3.4
             5.7
                         2.9
                                       4.2
96
                                                   1.3
78
             6.0
                         2.9
                                       4.5
                                                   1.5
                                       3.9
82
             5.8
                         2.7
                                                   1.2
38
             4.4
                         3.0
                                       1.3
                                                   0.2
[75 rows x 4 columns] ['Iris-versicolor' 'Iris-virginica' 'Iris-
setosa' 'Iris-setosa'
 'Iris-virginica' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor'
 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa' 'Iris-setosa'
 'Iris-virginica' 'Iris-versicolor' 'Iris-versicolor' 'Iris-setosa'
 'Iris-virginica' 'Iris-setosa' 'Iris-versicolor' 'Iris-setosa'
 'Iris-setosa' 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa'
 'Iris-versicolor' 'Iris-setosa' 'Iris-setosa' 'Iris-virginica'
 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica' 'Iris-versicolor'
 'Iris-virginica' 'Iris-virginica' 'Iris-virginica' 'Iris-setosa'
 'Iris-virginica' 'Iris-virginica' 'Iris-virginica' 'Iris-setosa'
 'Iris-virginica' 'Iris-virginica' 'Iris-virginica' 'Iris-versicolor'
 'Iris-versicolor' 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor'
```

```
'Iris-virginica' 'Iris-virginica' 'Iris-versicolor' 'Iris-setosa'
 'Iris-setosa' 'Iris-versicolor' 'Iris-setosa' 'Iris-virginica'
 'Iris-versicolor' 'Iris-versicolor' 'Iris-virginica' 'Iris-setosa'
 'Iris-virginica' 'Iris-virginica' 'Iris-setosa' 'Iris-virginica'
 'Iris-virginica' 'Iris-virginica' 'Iris-setosa' 'Iris-virginica'
 'Iris-versicolor' 'Iris-setosa' 'Iris-versicolor' 'Iris-versicolor'
 'Iris-versicolor' 'Iris-setosa']
C:\Users\MGM\anaconda3\Lib\site-packages\sklearn\base.py:493:
UserWarning: X does not have valid feature names, but
LogisticRegression was fitted with feature names
 warnings.warn(
from sklearn.metrics import classification report, confusion matrix
print(confusion_matrix(y_test, y_pred))
print(classification report(y test, y pred))
[[27 0 0]
[ 0 21 3]
 [ 0 0 24]]
                              recall f1-score
                 precision
                                                 support
    Iris-setosa
                      1.00
                                1.00
                                          1.00
                                                      27
                      1.00
                                0.88
                                          0.93
                                                      24
Iris-versicolor
                                          0.94
                                                      24
Iris-virginica
                      0.89
                                1.00
                                                      75
       accuracy
                                          0.96
                      0.96
                                0.96
                                          0.96
                                                      75
      macro avq
  weighted avg
                      0.96
                                0.96
                                          0.96
                                                      75
```