Binary Classification_Sk_NN

```
#import section
import sklearn
from sklearn.datasets import load_breast_cancer
from sklearn.model selection import train test split
from sklearn.neural network import MLPClassifier
from sklearn import metrics
#load database
data = load_breast_cancer()
#examine data
x_names = data['feature_names']
y_names = data['target_names']
x = data['data']
y = data['target']
print(y_names)
print(x_names)
     ['malignant' 'benign']
     ['mean radius' 'mean texture' 'mean perimeter' 'mean area'
      'mean smoothness' 'mean compactness' 'mean concavity'
      'mean concave points' 'mean symmetry' 'mean fractal dimension'
      'radius error' 'texture error' 'perimeter error' 'area error'
      'smoothness error' 'compactness error' 'concavity error'
      'concave points error' 'symmetry error' 'fractal dimension error'
      'worst radius' 'worst texture' 'worst perimeter' 'worst area'
      'worst smoothness' 'worst compactness' 'worst concavity'
      'worst concave points' 'worst symmetry' 'worst fractal dimension']
x_train,x_test,y_train,y_test = train_test_split(x,y,test_size=0.2,random_state=42)
mlp = MLPClassifier(hidden layer sizes=(200,), activation='logistic', alpha = 0.01, solve
mlp.fit(x_train,y_train)
```

Iteration 1, loss = 0.66423476 Iteration 2, loss = 0.62794413Iteration 3, loss = 0.60989763Iteration 4, loss = 0.59967550 Iteration 5, loss = 0.59427097Iteration 6, loss = 0.58521329 Iteration 7, loss = 0.55965895 Iteration 8, loss = 0.55196990 Iteration 9, loss = 0.53932907 Iteration 10, loss = 0.52514628 Iteration 11, loss = 0.51722911 Iteration 12, loss = 0.50711408 Iteration 13, loss = 0.49787412 Iteration 14, loss = 0.49050556 Iteration 15, loss = 0.48752876 Iteration 16, loss = 0.47673948 Iteration 17, loss = 0.47993544 Iteration 18, loss = 0.46797245 Iteration 19, loss = 0.46441103 Iteration 20, loss = 0.45171874 Iteration 21, loss = 0.44151621 Iteration 22, loss = 0.43654456 Iteration 23, loss = 0.42940712Iteration 24, loss = 0.42571736Iteration 25, loss = 0.42065474 Iteration 26, loss = 0.41780376 Iteration 27, loss = 0.41532171Iteration 28, loss = 0.40574464 Iteration 29, loss = 0.40541116 Iteration 30, loss = 0.40079164Iteration 31, loss = 0.39621897 Iteration 32, loss = 0.38990156 Iteration 33, loss = 0.38724137 Iteration 34, loss = 0.38653433 Iteration 35, loss = 0.38053886 Iteration 36, loss = 0.37144850 Iteration 37, loss = 0.37419820 Iteration 38, loss = 0.36781577 Iteration 39, loss = 0.36358413 Iteration 40, loss = 0.38588840 Iteration 41, loss = 0.35720477 Iteration 42, loss = 0.35326649Iteration 43, loss = 0.35139777 Iteration 44, loss = 0.35413033 Iteration 45, loss = 0.34662249Iteration 46, loss = 0.34465118 Iteration 47, loss = 0.35473041 Iteration 48, loss = 0.33575499 Iteration 49, loss = 0.32968173Iteration 50, loss = 0.33003552 Iteration 51, loss = 0.32563593Iteration 52, loss = 0.32142269 Iteration 53, loss = 0.33429541 Iteration 54, loss = 0.32404276Iteration 55, loss = 0.32392734Iteration 56, loss = 0.32152923 Iteration 57, loss = 0.31658237Iteration 58, loss = 0.30715736 Iteration 59, loss = 0.30706361 Iteration 60, loss = 0.31783154

```
Iteration 61, loss = 0.30688478
Iteration 62, loss = 0.30540081
Iteration 63, loss = 0.30700401
Iteration 64, loss = 0.30176876
Iteration 65, loss = 0.30186032
Iteration 66, loss = 0.30562930
Iteration 67, loss = 0.30408710
Iteration 68, loss = 0.29969768
Iteration 69, loss = 0.29255798
Iteration 70, loss = 0.30184832
Iteration 71, loss = 0.29473848
Iteration 72, loss = 0.30223574
Iteration 73, loss = 0.31478976
Iteration 74, loss = 0.28871198
Iteration 75, loss = 0.32356931
Iteration 76, loss = 0.29239020
Iteration 77, loss = 0.28854308
Iteration 78, loss = 0.31323790
Iteration 79, loss = 0.28649529
Iteration 80, loss = 0.28229284
Iteration 81, loss = 0.28659529
Iteration 82, loss = 0.29547102
Iteration 83, loss = 0.28898333
Iteration 84, loss = 0.27582272
Iteration 85, loss = 0.27657120
Iteration 86, loss = 0.27553380
Iteration 87, loss = 0.27425651
Iteration 88, loss = 0.27275920
Iteration 89, loss = 0.27612397
Iteration 90, loss = 0.27600615
Iteration 91, loss = 0.27009301
Iteration 92, loss = 0.27514190
Iteration 93, loss = 0.26935071
Iteration 94, loss = 0.26879087
Iteration 95, loss = 0.28248209
Iteration 96, loss = 0.27654794
Iteration 97, loss = 0.27641710
Iteration 98, loss = 0.26993826
Iteration 99, loss = 0.27454449
Iteration 100, loss = 0.27336509
Iteration 101, loss = 0.27683239
Iteration 102, loss = 0.26768910
Iteration 103, loss = 0.31394857
Iteration 104, loss = 0.27696649
Iteration 105, loss = 0.28943155
Iteration 106, loss = 0.26700490
Iteration 107, loss = 0.26183720
Iteration 108, loss = 0.26543810
Iteration 109, loss = 0.26998549
Iteration 110, loss = 0.26144614
Iteration 111, loss = 0.26681276
Iteration 112, loss = 0.26515997
Iteration 113, loss = 0.26467451
Iteration 114, loss = 0.25689668
Iteration 115, loss = 0.25912646
Iteration 116, loss = 0.27503798
Iteration 117, loss = 0.33879276
Iteration 118, loss = 0.27424440
Iteration 119, loss = 0.25668529
Iteration 120, loss = 0.29455947
```

```
Iteration 121, loss = 0.2//81186
     Iteration 122, loss = 0.31046773
     Iteration 123, loss = 0.27691942
     Iteration 124, loss = 0.26822355
     Iteration 125, loss = 0.25596092
     Iteration 126, loss = 0.25405997
     Iteration 127, loss = 0.30208047
     Iteration 128, loss = 0.26172731
     Iteration 129, loss = 0.25153075
     Iteration 130, loss = 0.27497685
     Iteration 131, loss = 0.26104151
     Iteration 132, loss = 0.26144125
     Iteration 133, loss = 0.25332222
     Iteration 134, loss = 0.27731017
predictions = mlp.predict(x_test)
```

ICEL. 4 CT 101 TO 1 TO 2 = 6. CO 20 20 TC /

metrics.accuracy_score(y_test,predictions)

0.9210526315789473

MLPClassifier

MLPClassifier(activation='logistic', alpha=0.01, hidden_layer_sizes=(200,), random state=1, solver='sgd', verbose=True)