

### Experiment - 3

Implement k - nearest neighbor's classification using python.

Source code:

```
# Import necessary modules
from sklearn.neighbors import KNeighborsClassifier
from sklearn.model_selection import train_test_split
from sklearn.datasets import load_iris

# Loading data
irisData = load_iris()

# Create feature and target arrays
X = irisData.data
y = irisData.target

# Split into training and test set
X_train, X_test, y_train, y_test = train_test_split(
    X, y, test_size=0.2, random_state=42)

knn = KNeighborsClassifier(n_neighbors=7)
knn.fit(X_train, y_train)

# predict on dataset which model has not
seen before
print(knn.predict(X_test))
```

Output:

[1 0 2 1 0 1 2 2 1 2 0 0 0 0 1 2 1 2 0 2 2 2 2 2 0 0]

# Example of Calculating Euclidean distance

from math import sqrt

# calculate the Euclidean distance between two vectors

def euclidean\_distance(row1, row2):

distance = 0.0

for i in range(len(row1)-1):

distance += (row1[i] - row2[i])\*\*2

return sqrt(distance)

# Test distance function

dataset =

[2.7810836, 2.550537003, 0],

[1.465489372, 2.362125076, 0],

[3.396561688, 4.400293529, 0],

[1.32807019, 1.850220317, 0],

[3.06407232, 3.005305973, 0],

[7.627531214, 2.088626775, 1],

[5.332441248, 2.088626775, 1],

[6.922596716, 1.77106367, 1],

[8.675418651, -0.242068655, 1],

[7.673756466, 3.508563011, 1]]

row0 = dataset[0]

for row in dataset: distance = euclidean\_

distance(row0, row)

Print(distance)

Output:

0.0

1.3290173915275787

1.9494646655653247

1.5591439385540549

0.5356280721938492

4.850940186986411

2.592833759950511

4.214227042632867

6.522409988228337

4.985585382449795



Knn example:

Implement knn algorithm and plot the result using python.

Source code:

```
import numpy as np
import matplotlib.pyplot as plt
import seaborn as sns
from matplotlib.colors import
```

~~Listed~~ ListedColormap

```
from sklearn import neighbors, datasets
n_neighbors = 15
```

```
# import some data to play with
```

```
iris = datasets.load_iris()
```

```
# we only take the first two features.
# we could avoid this ugly
```

```
# slicing by using a two-dim dataset
```

```
X = iris.data[:, :2]
```

```
y = iris.target
```

```
h = 0.02 # step size in the mesh
```

```
# create color maps
```

```
cmap_light = ListedColormap(["orange", "cyan",  
                             "cornflowerblue"])
```

```
cmap_bold = ["darkorange", "c", "darkblue"]
```

```
for weights in ["uniform", "distance"]:
```

# we create an instance of Neighbours classifier and fit the data.

```
clf = neighbors.KNeighborsClassifier(n_neighbors, weights = weights)
```

```
clf.fit(x, y)
```

# plot decision boundary. for that, we will assign a color to each.

# point in the mesh  $[x_{\min}, x_{\max}] \times [y_{\min}, y_{\max}]$ .

```
x_min, x_max = X[:, 0].min() - 1, X[:, 0].max() + 1
```

```
y_min, y_max = X[:, 1].min() - 1, X[:, 1].max() + 1
```

```
xx, yy = np.meshgrid(np.arange(x_min, x_max, h), np.arange(y_min, y_max, h))
```

```
z = clf.predict(np.c_[xx.ravel(), yy.ravel()])
```

# put the results into a color plot

```
Z = z.reshape(xx.shape)
```

```
plt.figure(figsize = (8, 6))
```

```
plt.contourf(xx, yy, Z, cmap = cmap_light)
```

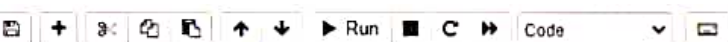
# plot also the training points

```
sns.scatterplot(  
    x = X[:, 0],  
    y = X[:, 1],  
    hue = iris.target_names[y],  
    palette = cmap_bold,  
    alpha = 1.0,  
    edgecolor = "black",  
)
```

```
plt.xlim(xx.min(), xx.max())  
plt.ylim(yy.min(), yy.max())  
plt.title("3 - class classification (k=1, weights = '1/s'
```

```
1. (n - neighbors, weights)  
)
```

```
plt.xlabel(iris.feature_names[0])  
plt.ylabel(iris.feature_names[1])  
plt.show()
```



```
plt.ylim(yy.min(), yy.max())
plt.title(
    "3-Class classification (k = %i, weights = '%s')" % (n_neighbors, weights)
)
plt.xlabel(iris.feature_names[0])
plt.ylabel(iris.feature_names[1])

plt.show()
```

