k-nearestneighborsclassification

```
In []:
#matplotlib inline
Load data: XOR¶
In [5]:
from google.colab import drive
drive.mount('/content/gdrive')
!ls "/content/gdrive/My Drive/Colab Notebooks"
import pandas as pd
xor_data = pd.read_csv("/content/gdrive/My Drive/Colab Notebooks/XOR.csv",header=None)
print(xor_data)
Drive already mounted at /content/gdrive; to attempt to forcibly remount, call drive.mount(
butterfly_n2000.csv
                                     Tutorial_PyGRNN.ipynb
                                                            XOR.csv
'k-nearest neighbors classification.ipynb'
                                             WordCloud.ipynb
   0 1 2
0 0 0 0
1 0 1 1
2 1 0 1
3 1 1 0
In [6]:
X = xor_data[[0,1]].to_numpy()
y = xor_data[2].to_numpy()
print(X)
print(y)
[[0 0]]
 [0 1]
 [1 0]
 [1 1]]
[0 1 1 0]
```

Classify XOR dataset¶

```
In [15]:
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import ListedColormap
from sklearn import neighbors, datasets

n_neighbors = 3  # k parameter
weights = 'distance'  # 'uniform' or 'distance'

clf = neighbors.KNeighborsClassifier(n_neighbors, weights=weights)
clf.fit(X, y)

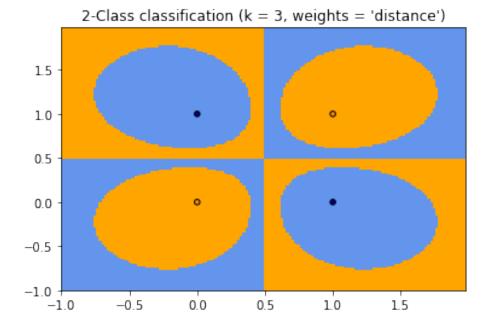
y_pred = clf.predict(X)

print(y_pred)
[0 1 1 0]
```

Plot classification surface: decision boundary¶

```
In [16]:
h = .02 # step size in the mesh
# Create color maps
cmap_light = ListedColormap(['orange', 'cyan', 'cornflowerblue'])
cmap_bold = ListedColormap(['darkorange', 'c', 'darkblue'])
# Plot the decision boundary. For that, we will assign a color to each
# point in the mesh [x_min, x_max]x[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 result into a color plot
Z = Z.reshape(xx.shape)
plt.figure()
plt.pcolormesh(xx, yy, Z, cmap=cmap_light)
# Plot also the training points
plt.scatter(X[:, 0], X[:, 1], c=y, cmap=cmap_bold,
                edgecolor='k', s=20)
plt.xlim(xx.min(), xx.max())
```

Text(0.5, 1.0, "2-Class classification (k = 3, weights = 'distance')")



Nearest Neighbors Classification: IRIS dataset¶

Sample usage of Nearest Neighbors classification. It will plot the decision boundaries for each class.

```
In []:
print(__doc__)
import numpy as np
import matplotlib.pyplot as plt
from matplotlib.colors import 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 = .02 # step size in the mesh
# Create color maps
cmap_light = ListedColormap(['orange', 'cyan', 'cornflowerblue'])
cmap_bold = ListedColormap(['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 the decision boundary. For that, we will assign a color to each
    # point in the mesh [x_min, x_max]x[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 result into a color plot
    Z = Z.reshape(xx.shape)
   plt.figure()
   plt.pcolormesh(xx, yy, Z, cmap=cmap_light)
   # Plot also the training points
    plt.scatter(X[:, 0], X[:, 1], c=y, cmap=cmap_bold,
                edgecolor='k', s=20)
    plt.xlim(xx.min(), xx.max())
    plt.ylim(yy.min(), yy.max())
    plt.title("3-Class classification (k = %i, weights = '%s')"
              % (n_neighbors, weights))
plt.show()
```

Automatically created module for IPython interactive environment

