XOR-PNN-Python

Load PNN class

In [1]:

```
## VEROWULF
## Learn. Create. Play.
## vati27@gmail.com
## https://github.com/verowulf/PNN

import numpy as np
import math

# Probabilistic Neural Network with 4 layers
class PNN(object):
    def __init__(self):
        self.L2 = []    # Layer 2 that holds the patterns
        print('Empty PNN created.')

    def train(self, X, y, p=2):
        self.n_ = X.shape[1]  # num of features
        self.p_ = p           # num of classes

        # Layer 2 (Pattern): Set up empty lists for each class
        for k in range(self.p_):
            self.L2.append([])    # Using Python's basic lists because ndarray cannot appenc
                                  # Also perhaps we might have to input different data types

        # Enter patterns into Layer 2
        for i in range(X.shape[0]):
            self.L2[y[i]].append(X[i])

        self.L2 = np.array(self.L2)    # Change to ndarray for speed (Is this faster?)

        print('PNN with %d classes trained.' % self.p_)

    def crossValidate(self, X, y, sigma=0.5):
        result = self.predict(X, sigma)
```

```
            num_correct = sum(result[:, 0] == y)

            print('Cross validation accuracy with sigma %.2f: %.1f%%' % (sigma, num_correct/len(

        def predict(self, X, sigma=0.5):
            m = X.shape[0]
            accL3 = np.zeros((m, self.p_))
            accL4 = np.zeros(m)

            self.sigma_ = sigma    # smoothing parameter, not standard deviation
            self.C1_ = 2 * self.sigma_**2
            C2_ = (math.sqrt(2*math.pi) * self.sigma_) ** (- self.n_)

            # Layer 1 (Input): x
            for i in range(m):
                x = X[i]

                # Layer 3 (Averaging): for each class
                self.L3_ = np.zeros(self.p_)
                for k in range(self.p_):
                    for ki in range(len(self.L2[k])):
                        self.L3_[k] += self._activation(x, self.L2[k][ki])
                    self.L3_[k] /= len(self.L2[k])

                    # Multiply constant
                    self.L3_[k] *= C2_
                    accL3[i][k] = self.L3_[k]

                # Layer 4 (Output/Decision): Maxing
                self.L4_ = self.L3_.argmax()
                accL4[i] = self.L4_

            return np.column_stack((accL4, accL3))

        def _activation(self, x, w):
            diff = x - w
            return math.exp( - np.dot(diff, diff) / self.C1_ )


    # Normalize to unit length: [0, 1]
    # X must be ndarray
    def Normalize(X):
        x_max = X.max(axis=0)
        x_min = X.min(axis=0)
        return (X - x_min) / (x_max - x_min)
```

XOR example with PNN

In [2]:

```
###  XOR example

#import pandas as pd
#X = pd.DataFrame(np.array([[0, 0], [0, 1], [1, 0], [1, 1]]),
#                   columns=['x1', 'x2'])
#Y = pd.DataFrame(np.array([0, 1, 1, 0]),
#                   columns=['f'])

X = np.array([[0, 0], [0, 1], [1, 0], [1, 1]])
Y = np.array([0, 1, 1, 0])

pnn = PNN()

pnn.train(X, Y, p=2) # p is number of classes

X2 = np.array([[0.1, 0], [0, 0.9], [1, 0], [1, 1]])
result = pnn.predict(X2)

print(result)
```

```
Empty PNN created.
PNN with 2 classes trained.
[[0.         0.32053212 0.10521866]
 [1.         0.10521866 0.32053212]
 [1.         0.08615712 0.32413994]
 [0.         0.32413994 0.08615712]]
```

In [ ]: