

## Problem

Consider the following entry set:

0 1  
1 1  
5 6  
4 7  
6 7

Requirements:

Classify the above patterns into two distinct classes (you will have to find all the patterns that belongs to class 1 and all the patterns that belongs to class 2) using the dynamic kernels/clusters method

## Steps and hints

*"In various scientific areas (medicine, biology, archeology, economy, etc) vast sets of objects represented by a finite number of parameters frequently appear ; for the specialist, obtaining the groupings "natural and homogeneous" together with the most representative elements of such a set constitutes an important stage in the understanding of his data. A good approach for the solution of the problem is provided by clustering techniques which consist of finding a partition for a finite set such that each object resembles more to the objects within its group than the objects outside."*

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More information on dynamic kernels/clusters method:

[http://www.numdam.org/article/RO\\_1976\\_\\_10\\_2\\_75\\_0.pdf](http://www.numdam.org/article/RO_1976__10_2_75_0.pdf)

It is a convergent algorithm and it is used in data mining. It clusters the data into M groups (classes) where M is predefined.

Compared with other clustering algorithms, this algorithm requires less machine time and storage. It can be also referred as a generalization of the K-means algorithm.

The K-means algorithm is one of the most widely used clustering algorithm that uses an explicit distance measure to partition the data set into clusters.

## Usage and applications

It has been successfully used in various topics, including market segmentation, computer vision, geostatistics, astronomy and agriculture. It often is used as a preprocessing step for other algorithms, for example to find a starting configuration.

<http://dni-institute.in/blogs/k-means-clustering-examples-and-practical-applications/>

<https://arxiv.org/ftp/arxiv/papers/1002/1002.2425.pdf> - prediction of Students' Academic Performance

<https://pdfs.semanticscholar.org/2ba1/942a08f3b9da97afa3b55719b5005ae2e5d0.pdf>

**dynamicKernels (x[n][p], n, p, M, iclass[n])**

**\*) c[i] = x[i], i = 1,M // c[i] si x[i] - arrayes with p number of elements**

**\*) iclass[i] = 0, i=1,n**

**repeat**

**done = true**

**\*) g[k][j]=0, j=1,p si k=1,M // initialize center of gravity matrix**

**\*) miu[k]=0 // counter of number of patterns for class**

**for i = 1, n execute**

**{**

**dmin = MAX\_VALUE**

**for k =1, M execute**

**dik= calculateDistance(x[i], c[k])**

**if dik <dmin then**

**dmin = dik**

**kmin = k**

**miu[kmin] += 1**

**for j=1,p execute**

**g[kmin][j] += x[i][j]**

**if iclass[i] != kmin then**

**iclass [i] = kmin**

**done = false**

**}**

**if done ==false then**

**for k =1, M execute**

**for j=1,p execute**

$g[k][j] \neq \mu[k]$

\*)  $c[k]=x[i_0]$ , where  $x[i_0]$  is the closest pattern to  $g[k]$

While done== false

end