index

c:\users\geral\documents\matlab\hw06\clustering.py

clustering

Author: Gerald Baulig

Modules

<u>kernel</u> <u>numpy</u>

Classes

builtins.object

Spectral

```
Spectral(X=None, gamma=1, epsilon=0, mode='default')
Container class for spectral clustering.
Computes the (W)eights, (D)egrees, (L)aplacian Matrix,
the eigenvalues and eigenvectors of a given dataset (X).
```

Spectral(X=None, gamma=1, epsilon=0, mode='default')

Methods defined here:

class Spectral(builtins.object)

```
init (self, X=None, gamma=1, epsilon=0, mode='default')
     Initialize self. See help(type(self)) for accurate signature.
cluster(self, K, mode='select')
     cluster(K, mode='select') -> kmeans generator
     Initialize and returns a KMeans generator.
     (See kmeans)
     Args:
         K: Number of cluster centers
         mode: The initialization mode (See init kmeans)
     Returns:
         Generator of KMeans, yields:
             Y: The labels
             means: The cluster centers
             delta: The update delta
             step: The update step counter
set(self, X=None, gamma=None, epsilon=None, mode=None)
     set(X=None, gamma=None, epsilon=None, mode=None)
     Use this function to change one or several properties.
     Recomputes the depending properties as required.
     Args:
         X: Set the dataset - updates the whole spectral information
         gamma: Set gamma - updates W, D, L, eigval and eigvec
         epsilon: Set epsilon - updates D, L, eigval and eigvec
         mode: Set the mode - updates L, eigval and eigvec
```

Data descriptors defined here:

D

The degree matrix.

 \mathbf{L}

```
The laplacian matrix.
             N
                  Number of datapoints.
             \mathbf{W}
                   The weight matrix.
             X
                   The inserted dataset.
               dict
                   dictionary for instance variables (if defined)
              weakref
                   list of weak references to the object (if defined)
             d
                   The dimension of the dataset.
             eigval
                   The eigenvalues extracted from L.
             eigvec
                   The eigenvectors extracted from L.
             epsilon
                   Epsilon for knn mode.
             gamma
                   The gamma value for the RBF kernel.
             mode
                   The KMeans initialization mode.
Functions
        dbscan(X, points, radius)
             dbscan(X, points, radius) -> yields Y, x, step
             Simple DBScan clustering with one path following agent.
             Takes long computation time, because almost every point needs to be checked.
             Otherwise it could not complete the pathfinding.
             The utilization of the radius is very low.
             Args:
                 X: The dataset.
                 points: Minimal number of points around core-points.
                 radius: The scan radius.
             Yields:
                 Y: The labels.
                 x: The current scan point.
                 step: The step counter.
        init kmeans(X, K, mode='mean')
```

X: The dataset.

Args:

init kmeans(X, K, mode='free') -> means
Initialize K cluster means on dataset X.

K: The number of cluster means.

mode: The mode how to initialize the cluster means.

```
mean = All means start (almost) at the mean of the dataset.
                 Pro: The result is determenistic.
                  Con: Needs more iterations.
             uniform = Uniform random distributed.
                 Pro: May work well on uniform distributed datasets.
                 Con: Ks may get lost in huge data gaps.
             normal = Normal random distributed.
                 Pro: May work well on normal distributed datasets.
                 Con: Ks may get lost in huge data gaps.
             select = Selects random points of the dataset.
                 Pro: Makes sure that each K has at least one point.
                 Con: Not determenistic like all the other random approaches.
             kmeans++ = Selects a random point and rearranges the
                 probabillities of selecting the next point according
                 to the distance.
                 Pro: May have an appropriate distributed of Ks.
                 Con: Great effort for a negligible improvment.
     Returns:
         means: The cluster means.
kernel trick(gram, C)
     kernel trick(gram, C) -> W
     Computes a Weight-Matrix of unsimilarity via "Kernel-Trick".
         gram: The Gram-Matrix of a Kernel-Function.
         C: An Association-Matrix with cluster assignments per column.
kmeans(X, means, epsilon=0.0, max it=1000, is kernel=False)
     kmeans(X, means, epsilon=0.0, max it=1000, isKernel=False) -
     > yields Y, means, delta, step
     A generator for simple KMeans clustering.
     Usage:
         See demo_kmeans.py
     Args:
         X: The data N-by-d, where
             N=num datapoints
             d=dimensions
         means: The cluster centers.
         epsilon: Convergence threshold.
              (default=0)
         max it:
         isKernel:
     Yields:
         Y: The labels or cluster association.
         means: The updated cluster centers.
         delta: The distance to the update step.
         step: The iteration step.
square_mag(u, v)
     square mag(a, b) -> squared magnitude
     Calculates the row-wise squared magnitude.
     This function is preferably used for distance comparisons,
     because taking a square root has a high computation time.
         np.sum((a-b)**2, axis=1)
     Args:
         u: numpy.ndarray
         v: numpy.ndarray
     Returns:
         The row-wise squared magnitude of a and b.
```

Data

 ${\bf LAPLACIAN_MODES} = ('default', 'shi', 'jordan')$