

DAY 3

Intelligent Audio Systems: A review of the foundations and applications of semantic audio analysis and music information retrieval



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These lecture notes contain hyperlinks to the CCRMA Wiki.

On these pages, you can find supplemental material for lectures - providing extra tutorials, support, references for further reading, or demonstration code snippets for those interested in a given topic .

Click on the  symbol on the lower-left corner of a slide to access additional resources.

WIKI REFERENCES...

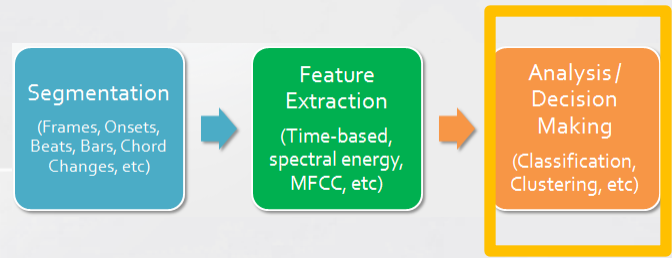


Review from Day 2

- BBQ Tomorrow evening: 6-9PM
- Let's talk lab...
- AdaBoost? Cross Validation?
- AdaBoost with too easy of a problem!
- Bug with CV
- Your feature vectors (e.g., only first 100ms)

Did you get to the Weka lab? Optional lab?

- How did the lab go?
- Did you try other audio files – other instrument recognizers?



ANALYSIS AND DECISION MAKING

Supervised vs. Unsupervised

- Unsupervised - “clustering”
- Supervised – binary classifiers (2 classes)
- Multiclass is derived from binary

Clustering

- Unsupervised learning – find pockets of data to group together
- Statistical analysis techniques

Clustering

- $K = \#$ of clusters
- Choosing the number of clusters – note that choosing the “best” number of clusters according to minimizing total squared distance will always result in same $\#$ of clusters as data points.

Clustering

The basic goal of clustering is to divide the data into groups such that the points within a group are close to each other, but far from items in other groups.

Hard clustering – each point is assigned to one and only one cluster.

K-Means

The key points relating to *k-means clustering* are:

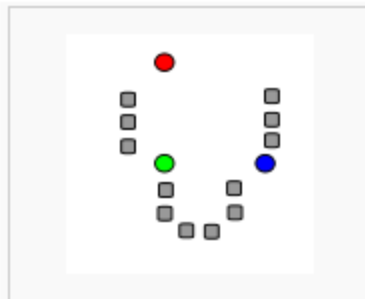
- k-means is an automatic procedure for clustering unlabelled data;
- it requires a pre-specified number of clusters;
- Clustering algorithm chooses a set of clusters with the minimum within-cluster variance
- Guaranteed to converge (eventually)
- Clustering solution is dependent on the initialization
(You get different results with each running)



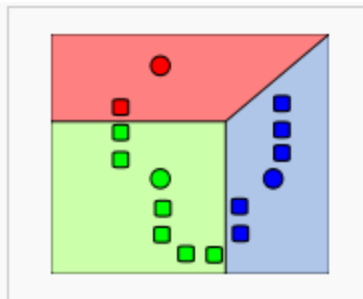
Demo

- http://home.dei.polimi.it/matteucc/Clustering/tutorial_html/AppletKM.html

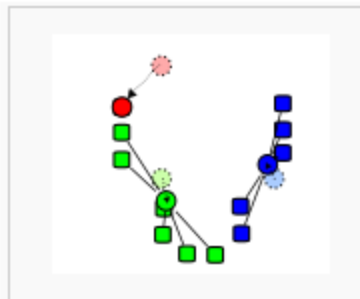
Demonstration of the standard algorithm



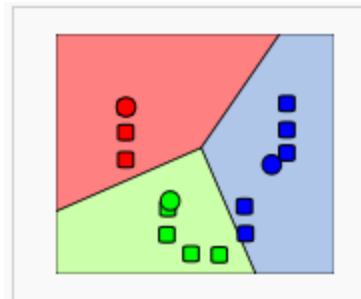
1) k initial "means" (in this case $k=3$) are randomly selected from the data set (shown in color).



2) k clusters are created by associating every observation with the nearest mean. The partitions here represent the [Voronoi diagram](#) generated by the means.



3) The [centroid](#) of each of the k clusters becomes the new means.



4) Steps 2 and 3 are repeated until convergence has been reached.

K-Means

The initialization method needs to be further specified.

There are several possible ways to initialize the cluster centers:

- *Choose random data points as cluster centers*
- *Randomly assign data points to K clusters and compute means as initial centers*
- *Choose data points with extreme values*
- *Find the mean for the whole data set then perturb into k means*
- *Find ground-truth for data*