

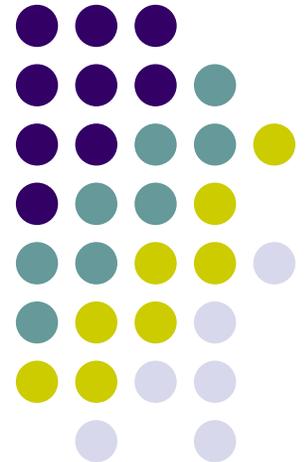
Cluster Analysis

K-means

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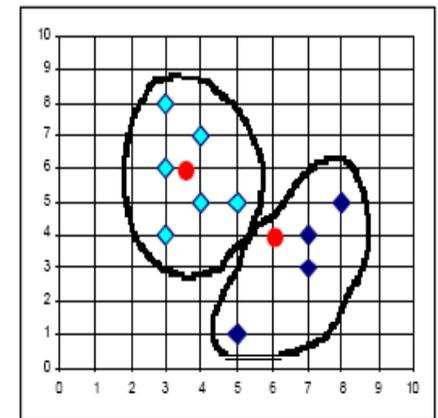
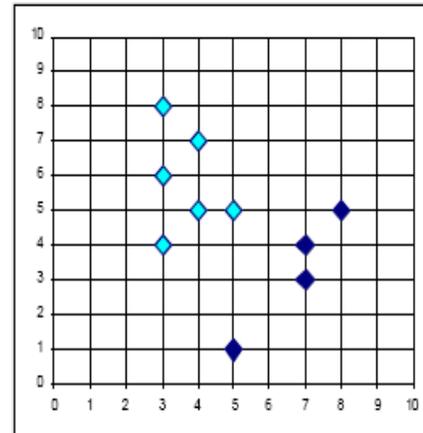
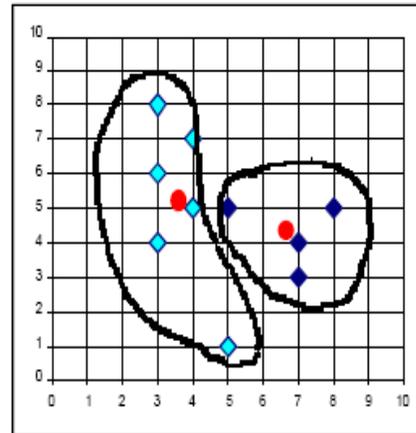
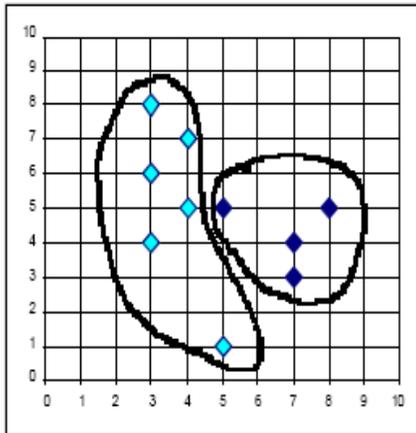
K-means : Overview

- isodata or c-Means
- Partitions the data into K mutually exclusive clusters
- Creates a single level of cluster
- Determine all clusters at once
- Assigns each observation to the cluster whose center is nearest
- The number of clusters (classes) are known a priori



The Algorithm

- i. Start with k random clusters
- ii. Move each observation to the nearest cluster center (centroids)
- iii. Recalculate the new centroids
- iv. Repeat until convergence criterion (minimum distance or max iterations) is met





Pros and Cons

- Advantages:
 - Simple & fast
 - Suitable for large dataset
- Disadvantages:
 - Unable to handle noisy data / outliers
 - Dependent on initial random clusters
 - Different classification results for different runs of the algorithm



K-means Example

- Classification of images extracted from the video of a speaker uttering 2 speech sounds (consonants)
- 40 images for each consonant
- Feature vector for each consonant = 6 values
- Squared Euclidean distance used to partition the data into 2 clusters
- Silhouette plot is shown to indicate the distance between the 2 clusters of features



/g/

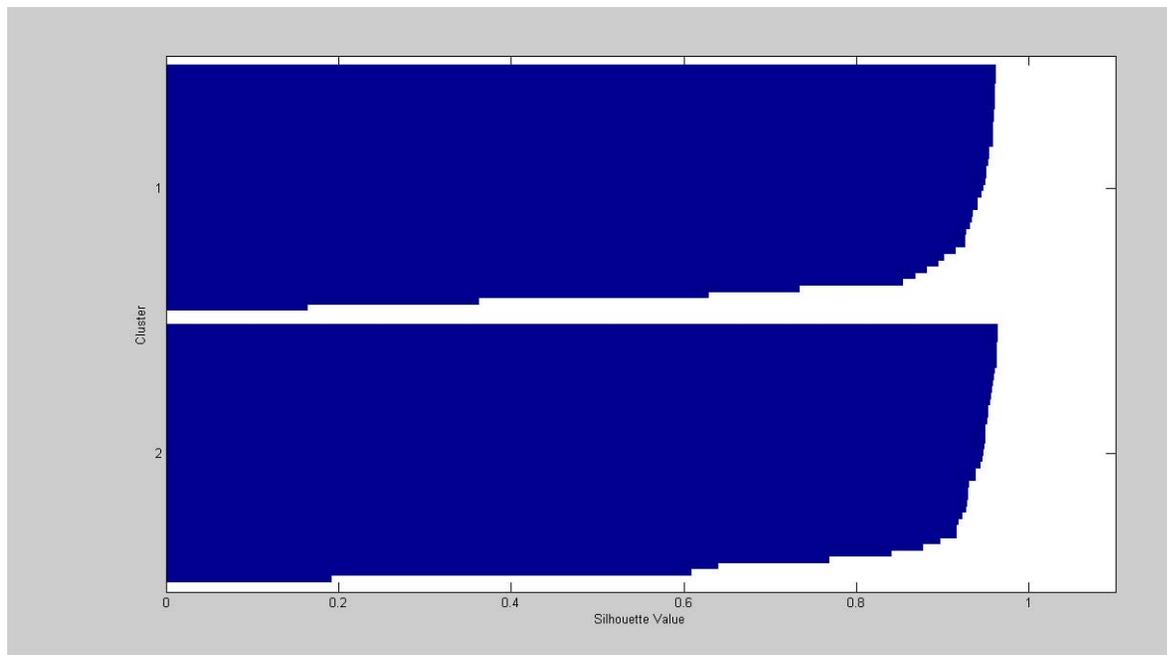


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Results – Silhouette Plot

- 79 out of the 80 features are correctly classify into the 2 clusters.
- Recognition rate = 98.75% for the classification of these 2 classes of images
- Silhouette plot for the 2 clusters of data shows that the distance between the two clusters are well separated (the silhouette value is near to 1)



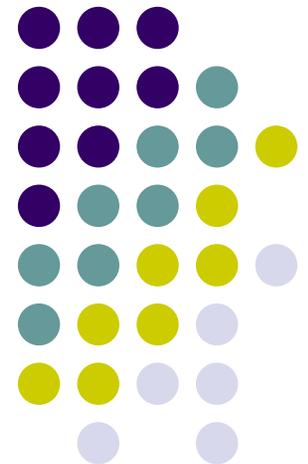


That's all for clustering

Next topic : Hidden Markov Models

Supervised Classifiers

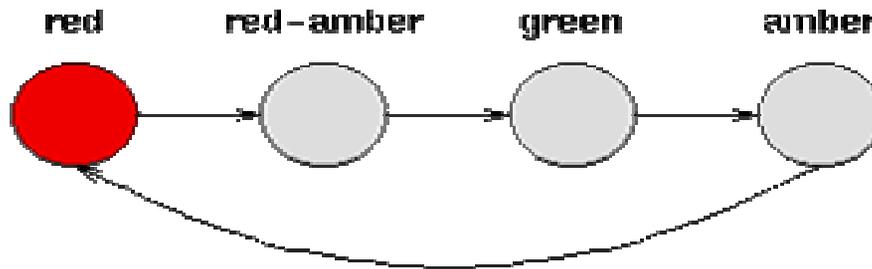
HIDDEN MARKOV MODELS (HMM)





HMM : Introduction

- A type of stochastic model that statistically model a system as Markov process (Markov chain) .
- A simple example of an observable Markov process : traffic light

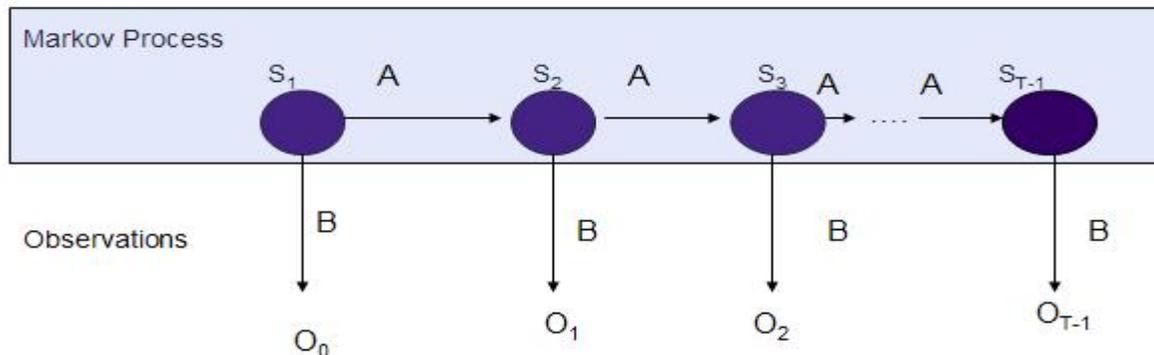


- Find patterns that appear over a space of time



What is Hidden in HMM?

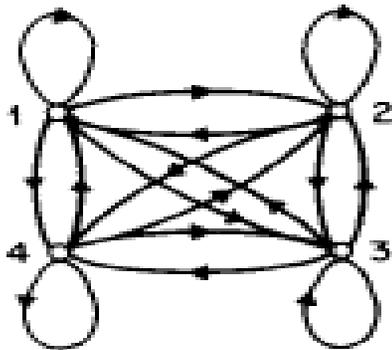
- Process where the states are not observable \rightarrow Hidden Markov Models
- HMM parameters (λ)
 - number of states
 - state transition probability distribution [A]
 - observation symbol probability distribution [B]
 - probabilities of the initial state
 - Num of observation symbols available (discrete)





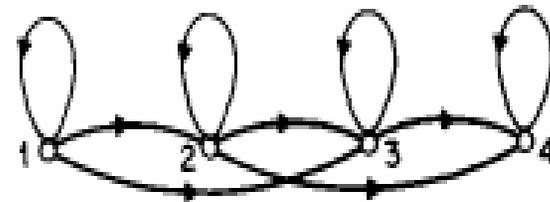
Types of HMM

Fully connected / Ergodic HMM



(a)

Left-right/ Bakis HMM



(b)

Design of HMM Classifiers



Basic Operations in Design of HMM classifier	Algorithm
<p>Learning /Training: Given an output sequence <i>or a set of such sequences</i> ($O=O_1, O_2, \dots, O_T$), find the <i>most likely set of HMM parameters λ to generate the training sequences</i></p>	Baum-Welch Algorithm
<p>Evaluation/ Recognition: Given the parameters of the model λ, compute the probability of a particular observation (test) sequence, O. $P(O \lambda)$</p>	Forward Algorithm



Why HMM?

- Suitable for modelling time-varying signals or classification tasks where patterns exist across a sequence of observations
- Applications:
 - Speech recognition
 - M. T. Chan, "HMM-based Audio Visual Speech Recognition Integrating Geometrics and Appearance-based Visual Features," presented at IEEE Workshop on Multimedia Signal Processing, Cannes, France, 2001
 - Bioinformatics / Genomics
 - Thomas Plötz, Gernot A. Fink, "Pattern recognition methods for advanced stochastic protein sequence analysis using HMMs", Pattern Recognition, Volume 39, 2006
 - Robotics (robot sensing)
 - Sven Koenig and Reid Simmons, *A Robot Navigation Architecture Based on Partially Observable Markov Decision Process Models*, in Artificial Intelligence Based Mobile Robotics: Case Studies of Successful Robot Systems, D. Kortenkamp, R. Bonasso, and R. Murphy, Eds., pp. 91-122. MIT press, 1998.
 - Character recognition
 - Agazzi, O.E., Kuo, S., 1993. *Hidden Markov model based optical character recognition in the presence of deterministic transformations*. Pattern Recognition 26 (12), 1813–1826
 - Economics / Finance
 - Rossi, Alessandro & Gallo, Giampiero M., 2006. "Volatility estimation via hidden Markov models," Journal of Empirical Finance, Elsevier, vol. 13(2), pages 203-230,



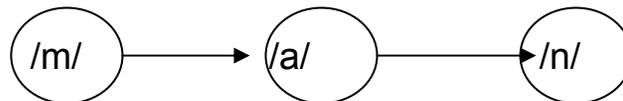
HMM - Speech Recognition

- Speech is a dynamic process (time varying)
- Usually left-right HMMs are used

Modelling Signal \rightarrow Phones



Modelling Phones \rightarrow Words



Modelling Words \rightarrow Sentence





HMM Example

- Classification of the 2 consonants again...
 - But we're using more frames now to model the mouth movement changes across time.
 - 35 videos of each consonant used for training HMM, 5 videos that are not used in training are used in testing the HMM
 - 3 states left-right HMM is used with 1 Gaussian mixture component & diagonal covariance
 - The features of middle frame and end frame forms one sequence of observation for the HMM
 - 1 HMM is created for each consonant (class)
 - Classification of 100% is obtained

Consonants	Start Frame	Middle Frame	End Frame
/m/			
/g/			



References & Toolboxes

1. Lawrence R. Rabiner, A Tutorial on Hidden Markov Models and Selected Applications in Speech Recognition. *Proceedings of the IEEE*, 77 (2), p. 257–286, February 1989.
2. S. Theodoridis and K. Koutroumbas, *Pattern Recognition*: Academic Press, 1999

HMM Toolboxes

1. Hidden Markov Models Toolkit (HTK)
<http://htk.eng.cam.ac.uk/>
2. BNT- HMM toolbox for MATLAB
<http://www.cs.ubc.ca/~murphyk/Software/HMM/hmm.html>
3. gpdsHMM- another HMM toolbox for MATLAB
<http://www.gpds.ulpgc.es/download/index.htm>
4. Jahmm- Java based HMM toolbox
<http://www.run.montefiore.ulg.ac.be/~francois/software/jahmm/>



Questions?