

SNS COLLEGE OF ENGINEERING

(Autonomous)





Artificial Intelligence & Machine Learning

Unit 3 – Unsupervised Learning Expectation-Maximization Algorithm Prepared by, P.Ramya

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Expectation-Maximization Algorithm

The Expectation-Maximization Algorithm, or EM algorithm for short, is an approach for maximum likelihood estimation in the presence of latent variables. A general technique for finding maximum likelihood estimators in latent variable models is the expectation-maximization (EM) algorithm.



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The basic two steps of the EM algorithm

•E-step and M-step are often pretty easy for many of the machine learning problems in terms of implementation.



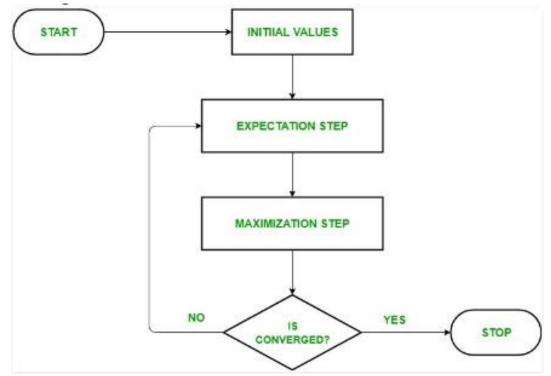
Algorithm:

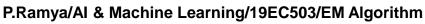
- •Given a set of incomplete data, consider a set of starting parameters.
- •Expectation step (E step): Using the observed available data of the dataset, estimate (guess) the values of the missing data.
- •Maximization step (M step): Complete data generated after the expectation (E) step is used in order to update the parameters.
- •Repeat step 2 and step 3 until convergence.



Elbow Method

In the Elbow method, we are actually varying the number of clusters (K) from 1 – 10. For each value of K, we are calculating WCSS (Within-Cluster Sum of Square). WCSS is the sum of squared distance between each point and the centroid in a cluster.







Implementation

Usage of EM algorithm -

- •It can be used to fill the missing data in a sample.
- •It can be used as the basis of unsupervised learning of clusters.
- •It can be used for the purpose of estimating the parameters of Hidden Markov Model (HMM).
- •It can be used for discovering the values of latent variables.



Advantages of EM algorithm

- •It is always guaranteed that likelihood will increase with each iteration.
- •The E-step and M-step are often pretty easy for many problems in terms of implementation.
- •Solutions to the M-steps often exist in the closed form.



Disadvantages of EM algorithm

- •It has slow convergence.
- •It makes convergence to the local optima only.
- •It requires both the probabilities, forward and backward (numerical optimization requires only forward probability).



