

# K Means

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Latex code ref <https://www.overleaf.com/learn>

Algorithm content ref <https://stanford.edu/~cpiech/cs221/handouts/kmeans.html>

# 1 Basic Idea

Say you are given a data set where each observed example has a set of features, but has no labels. Labels are an essential ingredient to a supervised algorithm like Support Vector Machines, which learns a hypothesis function to predict labels given features. So we can't run supervised learning. What can we do?

One of the most straightforward tasks we can perform on a data set without labels is to find groups of data in our dataset which are similar to one another – what we call clusters.

K-Means is one of the most popular "clustering" algorithms. K-means stores  $k$  centroids that it uses to define clusters. A point is considered to be in a particular cluster if it is closer to that cluster's centroid than any other centroid.

K-Means finds the best centroids by alternating between (1) assigning data points to clusters based on the current centroids (2) choosing centroids (points which are the center of a cluster) based on the current assignment of data points to clusters.

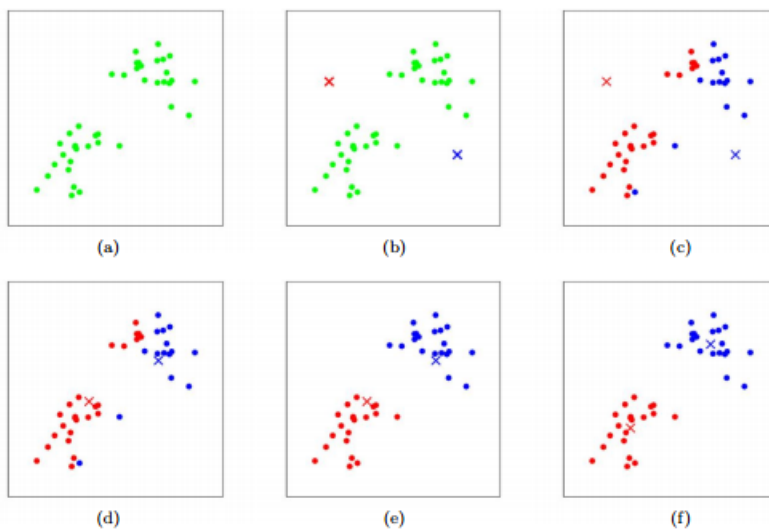


Figure 1: K-means algorithm. Training examples are shown as dots, and cluster centroids are shown as crosses. (a) Original dataset. (b) Random initial cluster centroids. (c-f) Illustration of running two iterations of k-means. In each iteration, we assign each training example to the closest cluster centroid (shown by "painting" the training examples the same color as the cluster centroid to which is assigned); then we move each cluster centroid to the mean of the points assigned to it. Images courtesy of Michael Jordan.

## 2 Algorithm

In the clustering problem, we are given a training set  $x^{(1)}, \dots, x^{(m)}$ , and want to group the data into a few cohesive "clusters." Here, we are given feature vectors for each data point  $x^{(i)} \in \mathbb{R}^n$  as usual; but no labels  $y^{(i)}$  (making this an unsupervised learning problem). Our goal is to predict  $k$  centroids and a label  $c^{(i)}$  for each datapoint. The k-means clustering algorithm is as follows:

- 1 Initialize **cluster centroids**  $\mu_1, \mu_2, \dots, \mu_k \in \mathbb{R}^n$  randomly.
- 2 Repeat until convergence:

For every  $i$ , set

$$c^{(i)} := \arg \min_j$$