

# Unsupervised Learning: K-Means Clustering

Some material adapted from slides by Andrew Moore, CMU.

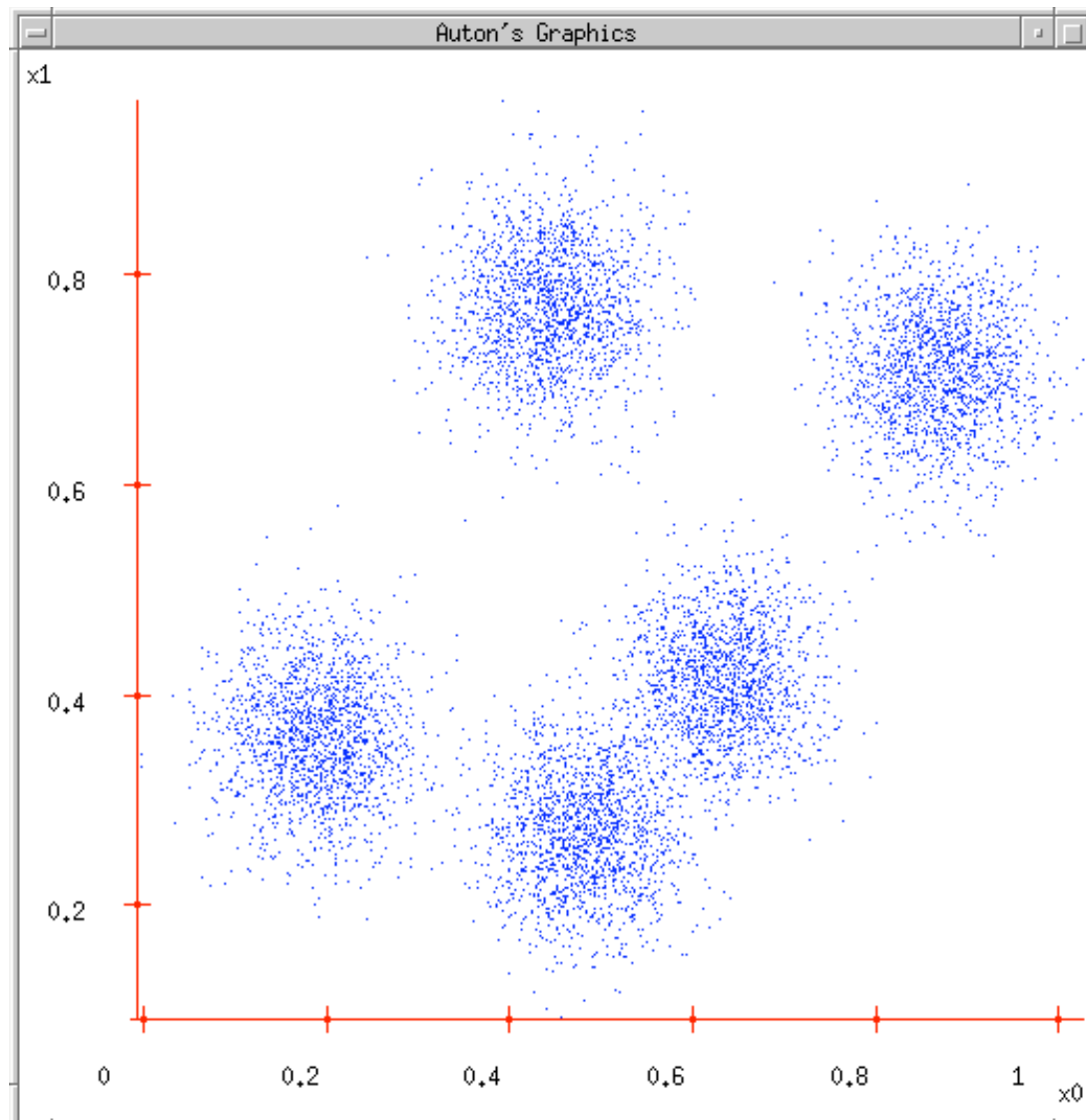
# Unsupervised Learning

case of when you don't have a target to reach a ground truth, unlike regression or classifier. no ground truth in unsupervised, purpose is other things than predictions, one purpose is clustering. how to group together based on similarity

- Supervised learning used labeled data pairs  $(x, y)$  to learn a function  $f : X \rightarrow Y$ .
- But, what if we don't have labels?
- No labels = **unsupervised learning**
  - Labels may be expensive to obtain, so we only get a few.
- **Clustering** is the unsupervised grouping of data points. It can be used for **knowledge discovery**.

classification vs clustering, classified already in past to learn. clustering just have random points in space, no name no categories, need to categorize ourselves.

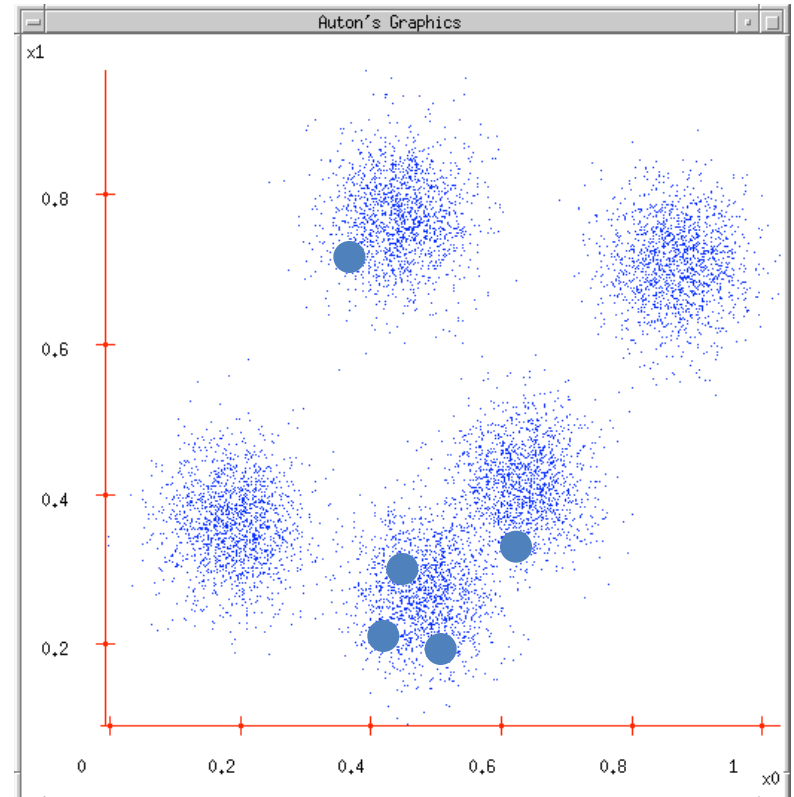
# Clustering Data



# K-Means Clustering

K-Means (  $k$  , data )

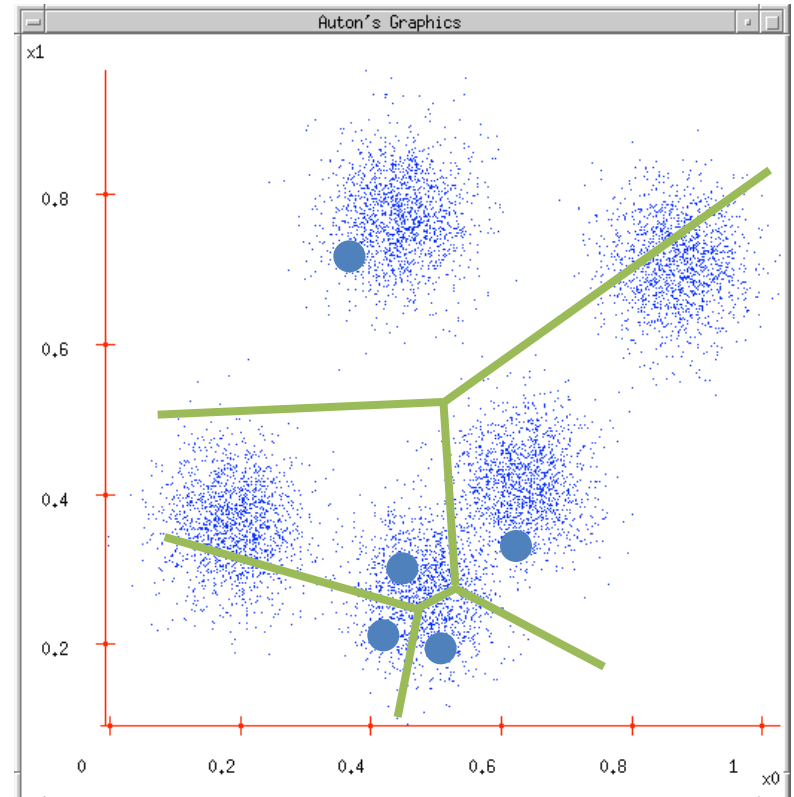
- Randomly choose  $k$  cluster center locations (centroids).
- Loop until convergence
  - Assign each point to the cluster of the closest centroid.
  - Reestimate the cluster centroids based on the data assigned to each.



# K-Means Clustering

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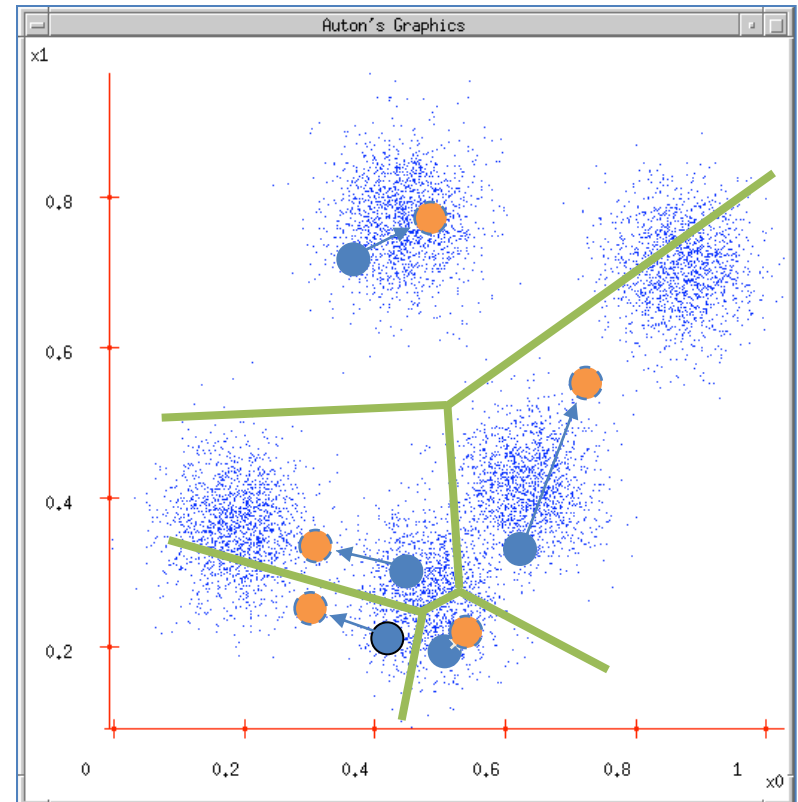
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# K-Means Clustering

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# K-means Algorithm

- For a current set of cluster means, assign each observation as:

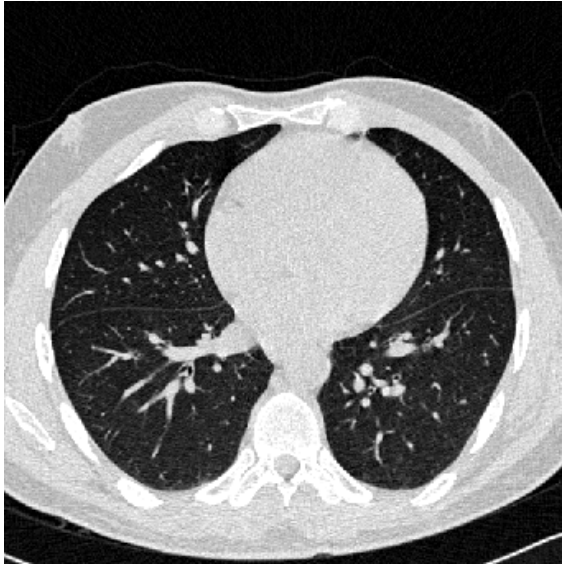
$$C(i) = \arg \min_{1 \leq k \leq K} \|x_i - m_k\|^2, \quad i = 1, \dots, N$$

- For a given assignment  $C$ , compute the cluster means  $m_k$ :

$$m_k = \frac{\sum_{i:C(i)=k} x_i}{N_k}, \quad k = 1, \dots, K.$$

- Iterate above two steps until convergence

# Image Segmentation Results



An image ( $I$ )

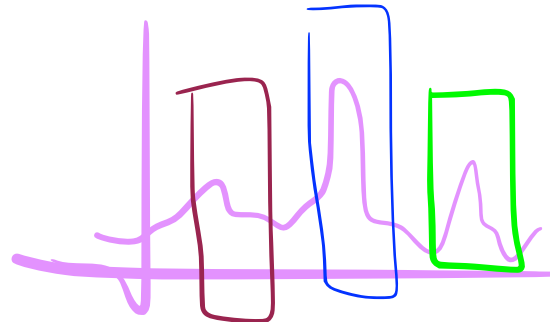


Three-cluster image ( $J$ ) on gray values of  $I$

Matlab code:

```
I = double(imread('...'));
```

```
J = reshape(kmeans(I(:),3),size(I));
```



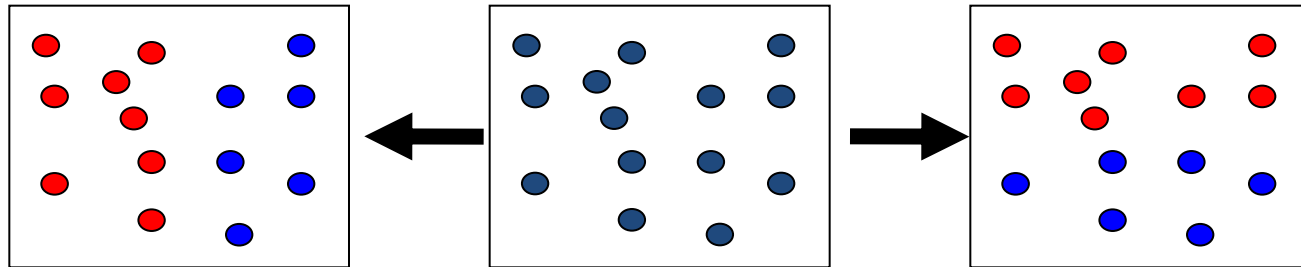


# Problems with K-Means

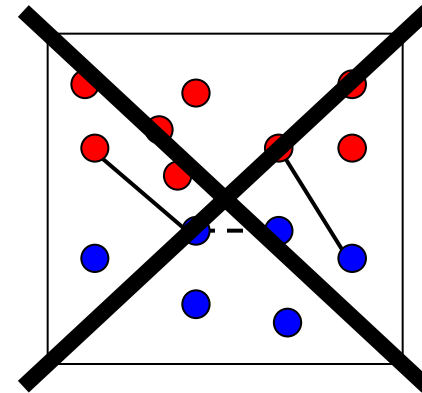
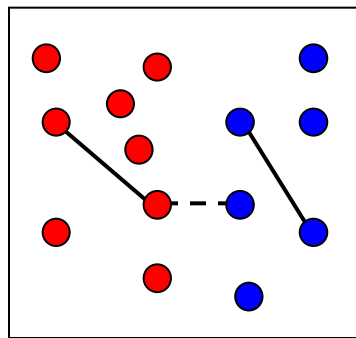
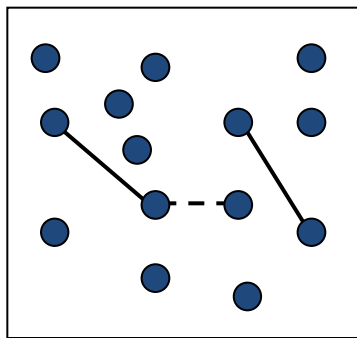
- **Very** sensitive to the initial points.
  - Do many runs of k-Means, each with different initial centroids.
  - Seed the centroids using a better method than random. (e.g. sampling point far apart)
- Must manually choose k.
  - Learn the optimal k for the clustering (meta-learning). (Note that this requires a performance measure.)

# Problems with K-Means

- How do you tell it which clustering you want?



- Constrained clustering techniques



— Same-cluster constraint  
(must-link)

- - - Different-cluster constraint  
(cannot-link)