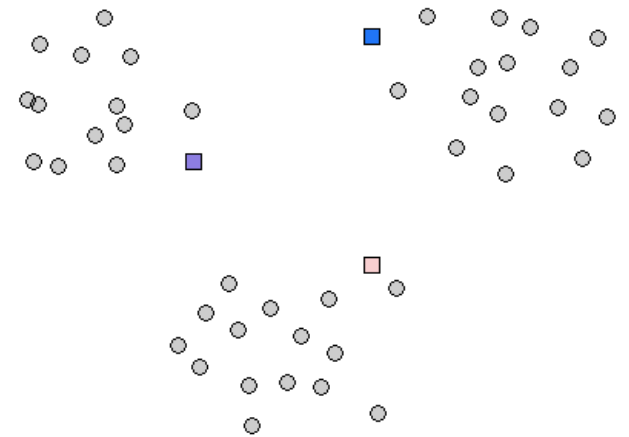


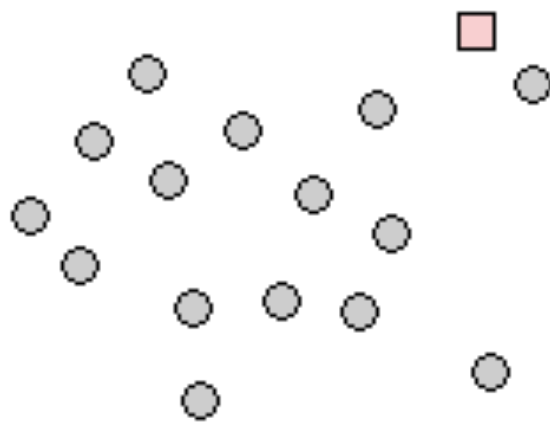
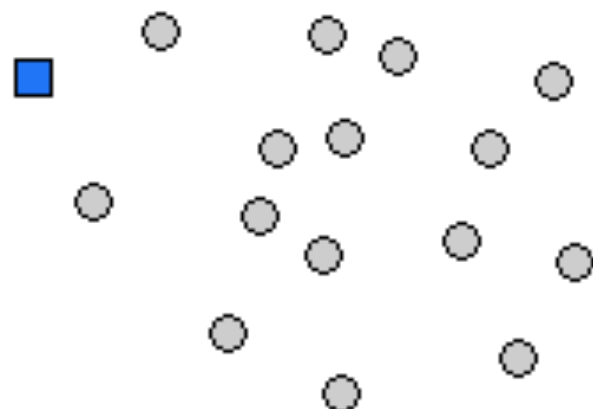
K-means

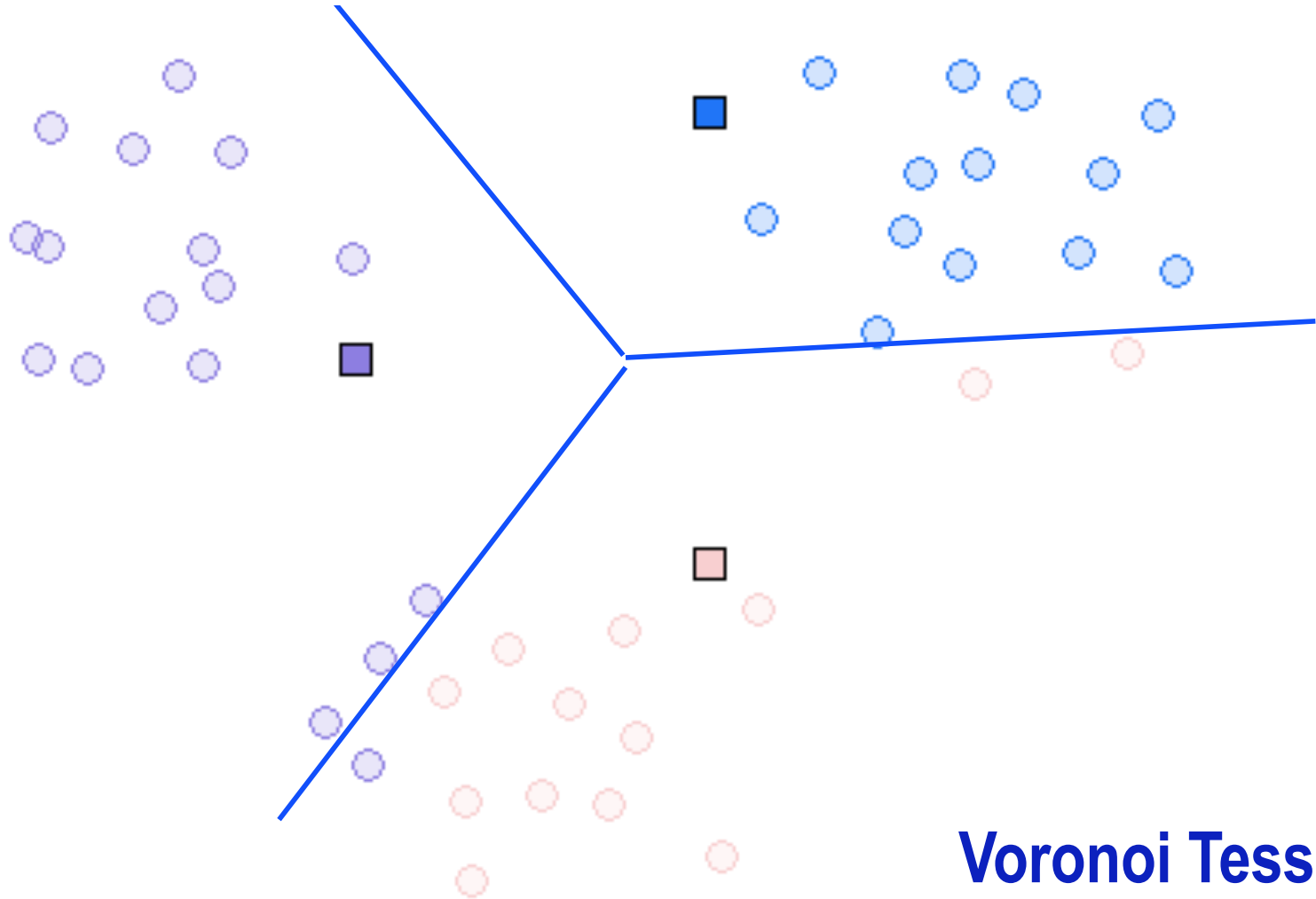
Lyle Ungar

Why cluster?

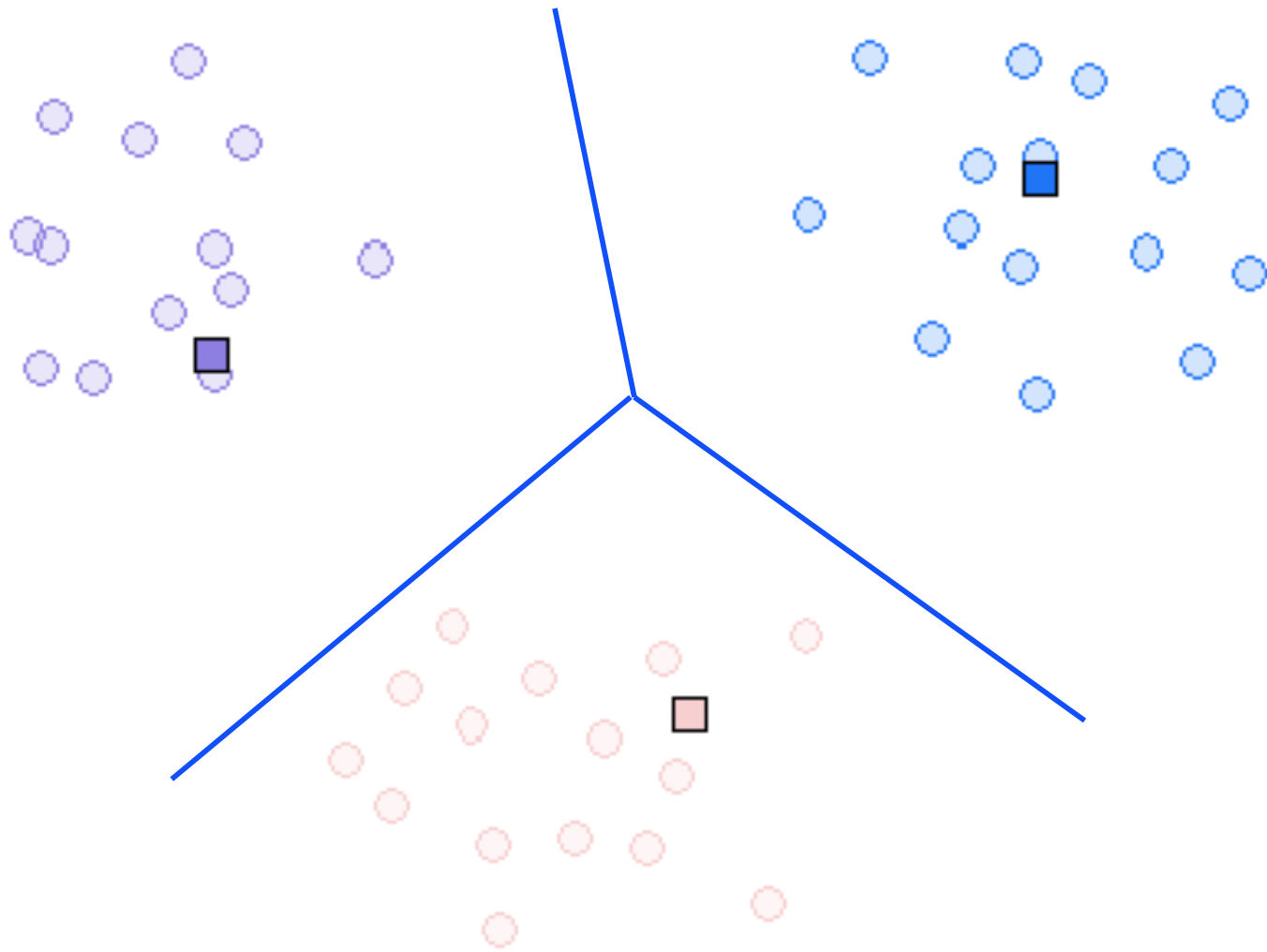
K-means: loss function, algorithm
Relation to EM and Gaussian Mixture
Models (GMM)

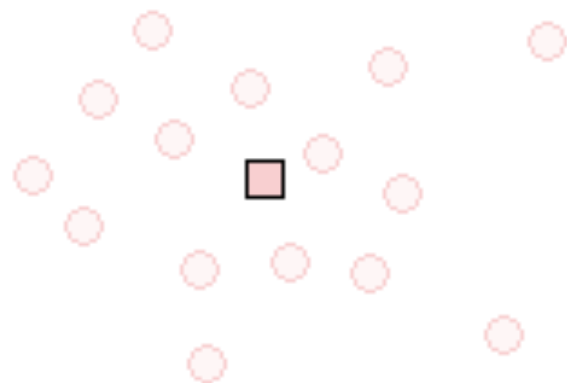
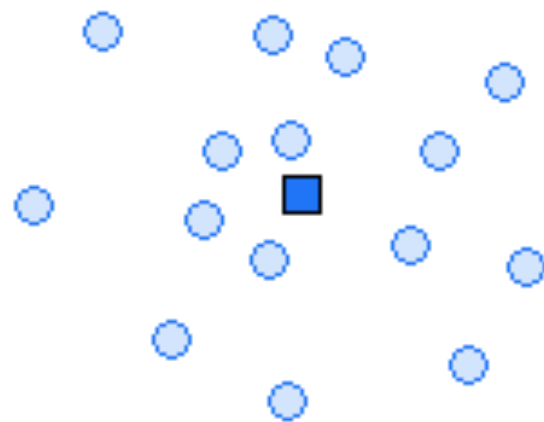
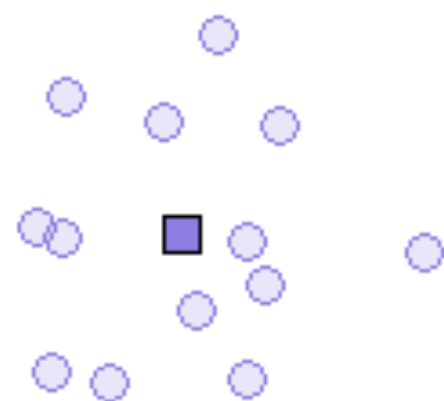






Voronoi Tessellation





K-Means algorithm

- ◆ Pick K cluster centroids at random
- ◆ Alternate until convergence
 - Assign points to nearest centroid
 - Set centroid to the mean of the examples assigned to it

K-means objective

Loss to be minimized

$$J(\mu, r) = \sum_{i=1}^n \sum_{k=1}^K r_{ik} \|\mu_k - \mathbf{x}_i\|_2^2$$

r_{ik} 1 iff point x_i in cluster k

μ_k centroid of cluster k

Reconstruction error of approximating every x_i by the center of the cluster it is in

K-means algorithm

$$J(\mu, r) = \sum_{i=1}^n \sum_{k=1}^K r_{ik} \|\mu_k - \mathbf{x}_i\|_2^2$$

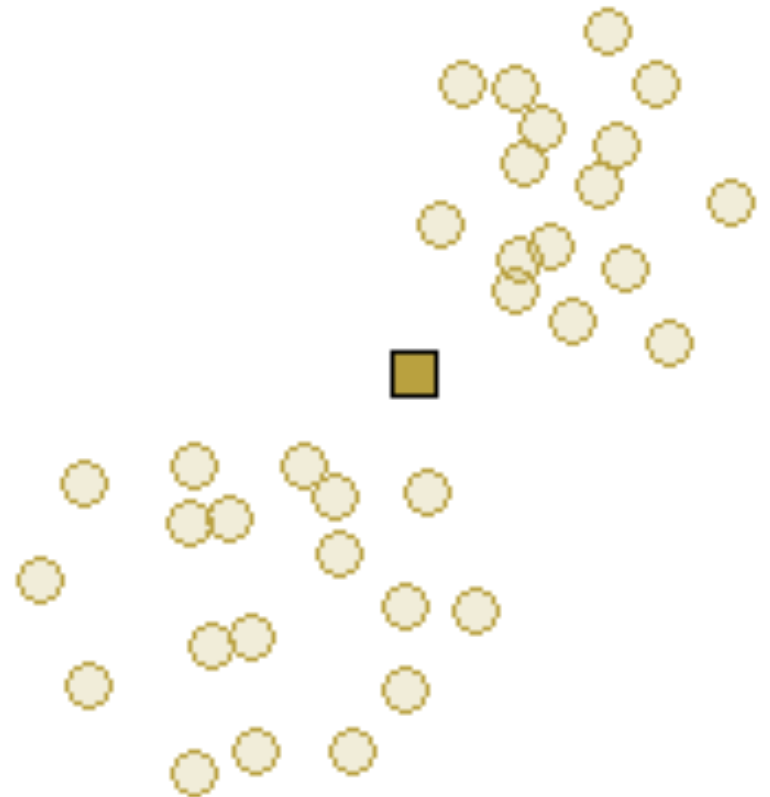
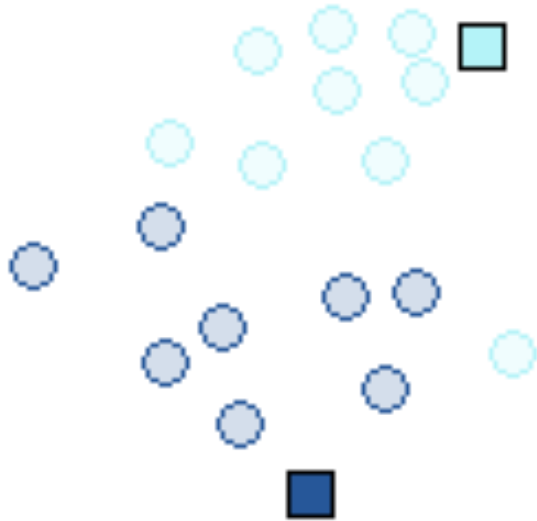
Assign point i to cluster k

$$\arg \min_r J(\mu, r) \rightarrow r_{ik} = \mathbf{1}(k = \arg \min_{k'} \|\mu_{k'} - \mathbf{x}_i\|_2^2)$$

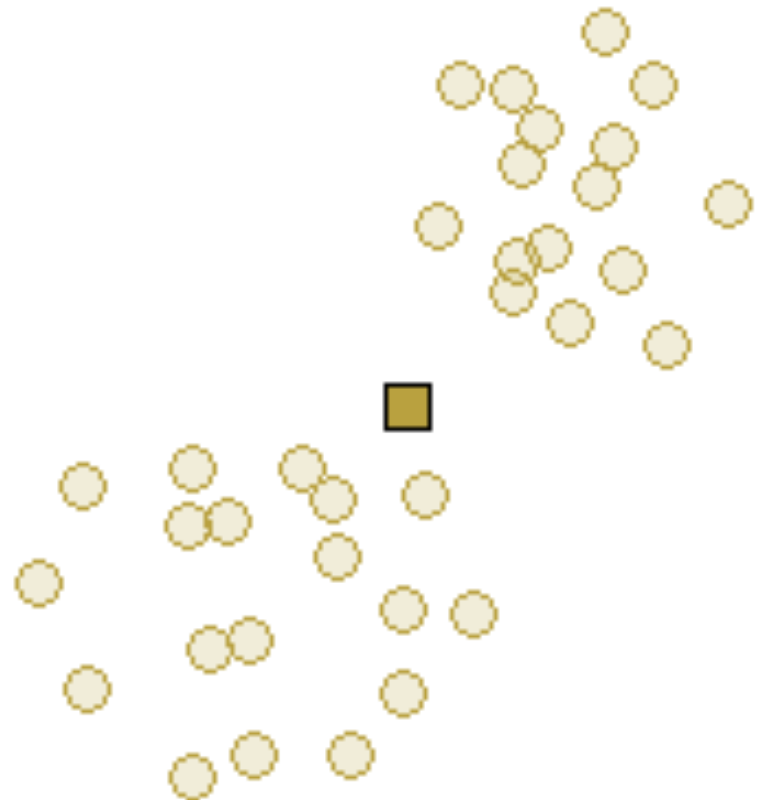
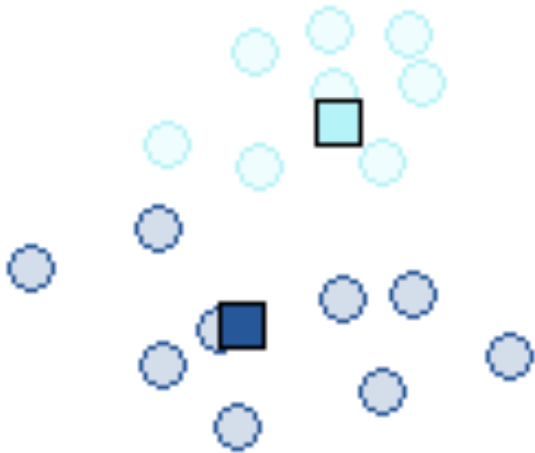
Compute centroid of cluster k

$$\arg \min_{\mu} J(\mu, r) \rightarrow \mu_k = \frac{\sum_i r_{ik} \mathbf{x}_i}{\sum_i r_{ik}}$$

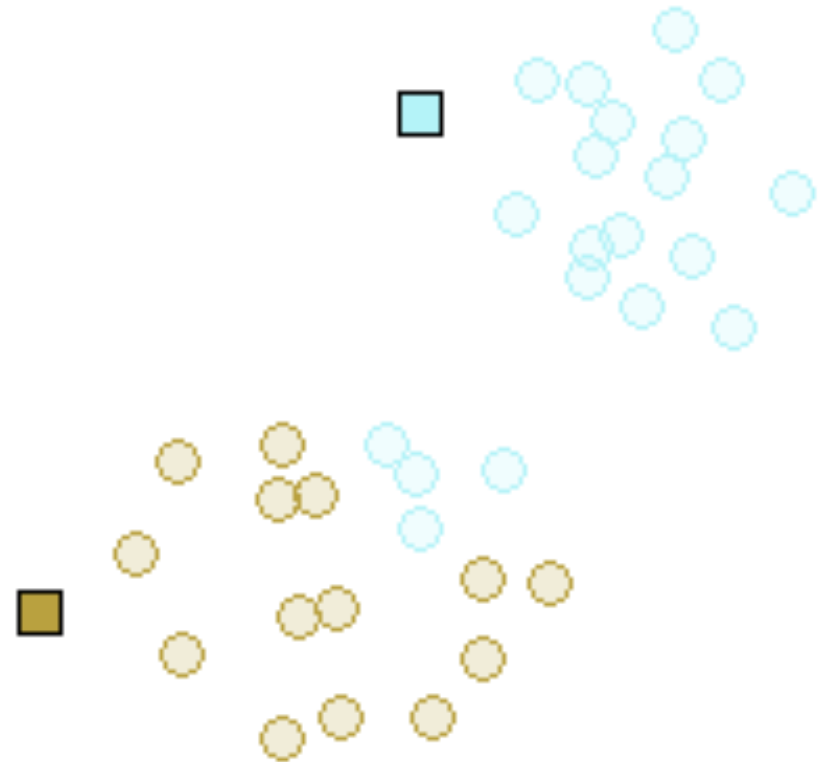
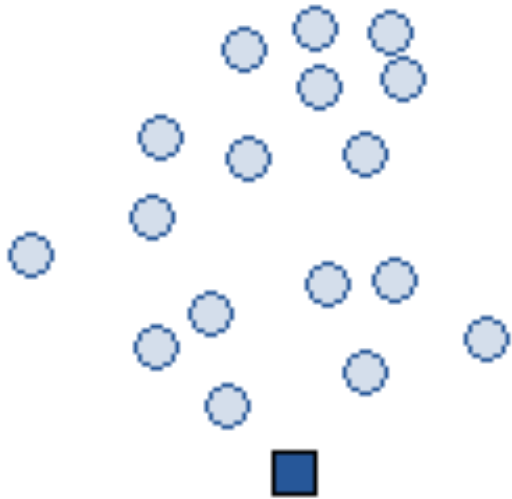
Bad start



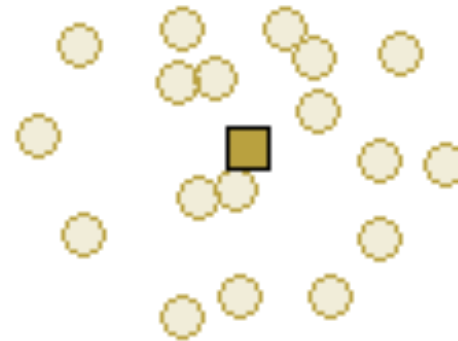
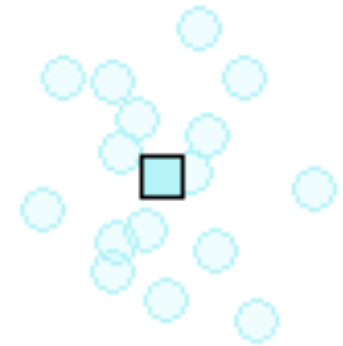
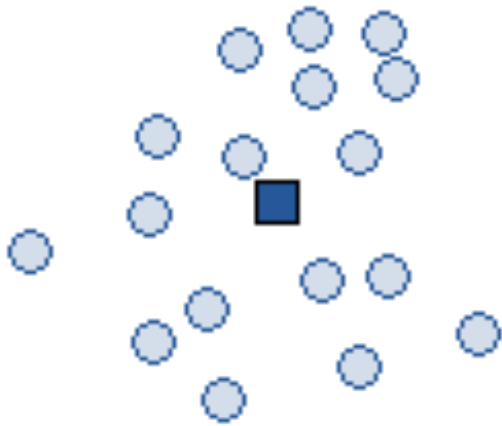
Bad start gives bad clusters



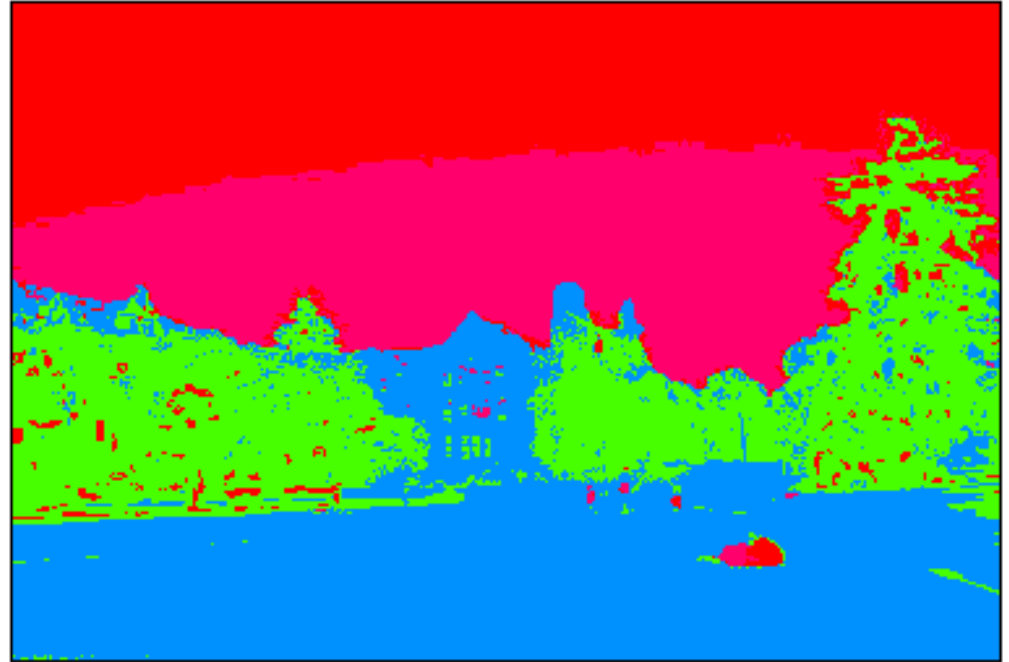
Better start



Better start gives better clusters



Example use - segmentation



What you should know

- ◆ **K-means uses alternating gradient descent**
 - estimates cluster membership and cluster centroids
 - minimizes reconstruction error
- ◆ **Usually initialized by selection of k random points, x_i**
 - But can also pick points that are spread out
 - Kmeans++ “chooses centers at random from the data points, but weighs the data points according to their squared distance squared from the closest center already chosen.”
- ◆ **Has a probabilistic interpretation**