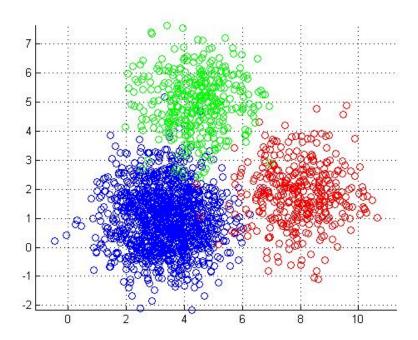
### k-means



mathworks.com

Note that this is not related to kNN.

### Explain:

- Choose k
- Choose k centroids
- Compute distances and assign each point to closest centroid
- Now re-compute centroids based on classes
- Now re-assign based on closest
- Converges because distances get smaller

Parameter sweep is a possibility for choosing k.

Assumes roughly spherical clusters.

### Assignment:

- Pick k random centroids from sample (Forgy method).
- Assign to k random clusters (random partitions method).

Uses

• Vector quantization (picking hopefully prototypical samples)

## **Examples**

# **Code time**

### **Discussion**

When should we use k-means vs logistic regression?

# 1 SVM

**SVM** 

### **Support Vector Machines**

• Goal: optimal separating hyperplane

• aka: Large Margin Classifier

Discussion:

Consider the example of dots at [[0, 1], [1, 0]].

### **Strategies**

At the beginning, one tends to do this:

- Transform data for SVM solver
- Randomly try a few kernels and parameters
- Test

#### A better strategy:

- Transform
- Scale data
- Consider linear, Gaussian, or RBF kernels
- Use cross validation to find the best C and  $\gamma$
- Test

Grid search often works well for C and  $\gamma$ . Try exponentially increasing C and decreasing  $\gamma$ . E.g.,  $2^{-5}, 2^{-3}, \ldots, 2^{15}$ . Categories usually work better as binary fields than as enums.

If there are lot of features, linear often works well.

n features, m training examples.

When n (e.g.,  $10^4$ ) is much larger m (e.g., 10 to  $10^3$ ), then linear regression or SVM with linear kernel tends to work well.

When n is small  $(n < 10^3)$  and m medium  $(m < 10^4)$ , then Gaussian kernel often works well.

When m is large ( $n<10^3$  and  $m>5\cdot 10^4$ ), add features and then use linear regression or SVM with a linear kernel.

# 2 Break

# **Break**

**Questions?**