

# Introduction to Information Retrieval

<http://informationretrieval.org>

## IIR 14: Vector Space Classification

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(Based on slides by Hinrich Schütze at [informationretrieval.org](http://informationretrieval.org))

Spring 2017

# Overview

- 1 Intro vector space classification
- 2 Rocchio
- 3 kNN

# Take-away today

- **Vector space classification:** Basic idea of doing text classification for documents that are represented as vectors
- Rocchio classifier: Rocchio relevance feedback idea applied to text classification (briefly covered)
- $k$  nearest neighbor classification
- Linear classifiers (not covered)

# Outline

- 1 Intro vector space classification
- 2 Rocchio
- 3 kNN

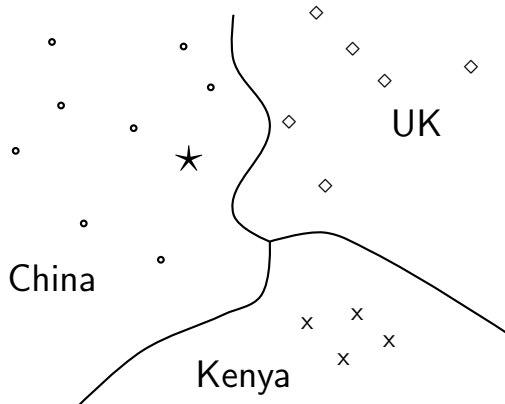
# Recall vector space representation

- Each document is a vector, one component for each term.
- Terms are axes.
- High dimensionality: 100,000s of dimensions
- Normalize vectors (documents) to unit length
- How can we do classification in this space?

# Vector space classification

- As before, the training set is a set of documents, each labeled with its class.
- In vector space classification, this set corresponds to a labeled set of points or vectors in the vector space.
- Premise 1: Documents in the same class form a **contiguous region**.
- Premise 2: Documents from different classes **don't overlap**.
- We define lines, surfaces, hypersurfaces to divide regions.

# Classes in the vector space



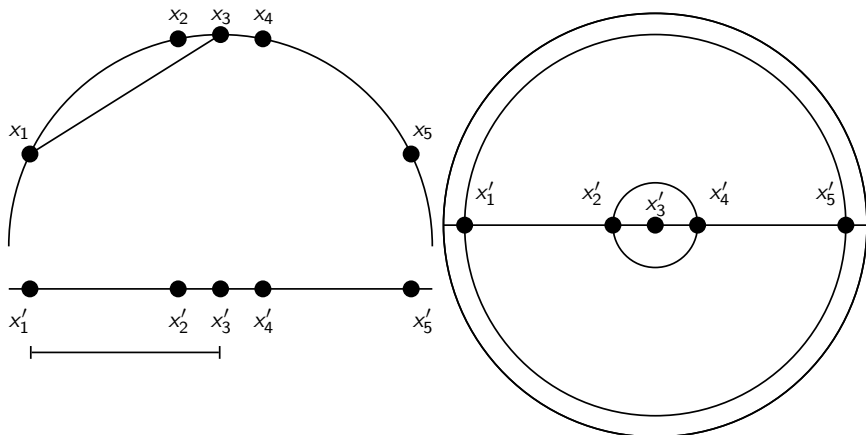
Should the document  $\star$  be assigned to *China*, *UK* or *Kenya*?

Find separators between the classes

Based on these separators:  $\star$  should be assigned to *China*

How do we find separators that do a good job at classifying new documents like  $\star$ ? – Main topic of today

## Aside: 2D/3D graphs can be misleading



*Left:* A projection of the 2D semicircle to 1D. For the points  $x_1, x_2, x_3, x_4, x_5$  at  $x$  coordinates  $-0.9, -0.2, 0, 0.2, 0.9$  the distance  $|x_2 x_3| \approx 0.201$  only differs by 0.5% from  $|x'_2 x'_3| = 0.2$ ; but  $|x_1 x_3|/|x'_1 x'_3| = d_{\text{true}}/d_{\text{projected}} \approx 1.06/0.9 \approx 1.18$  is an example of a large distortion (18%) when projecting a large area. *Right:* The corresponding projection of the 3D hemisphere to 2D.



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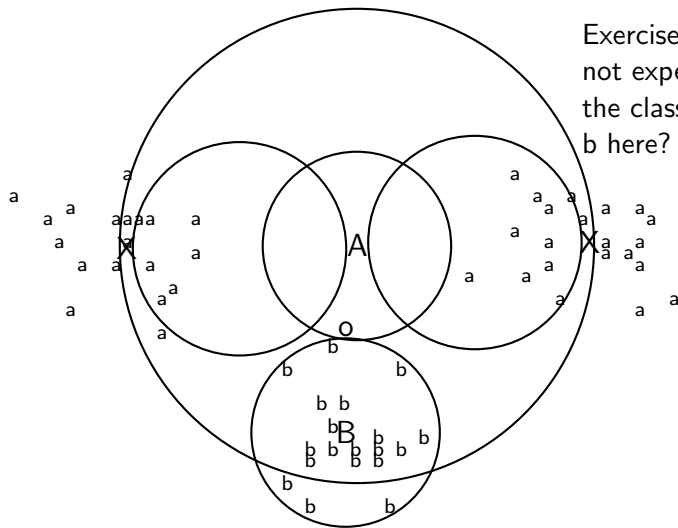
# Rocchio classification: Basic idea

- Compute a centroid for each class
  - The centroid is the average of all documents in the class.
- Assign each test document to the class of its closest centroid.

# Rocchio vs. Naive Bayes

- In many cases, Rocchio performs worse than Naive Bayes.
- One reason: Rocchio does not handle nonconvex, multimodal classes correctly.

# Rocchio cannot handle nonconvex, multimodal classes



Exercise: Why is Rocchio not expected to do well for the classification task a vs. b here?

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# kNN classification

- kNN classification is another vector space classification method.
- It also is very simple and easy to implement.
- kNN is more accurate (in most cases) than Naive Bayes and Rocchio.
- If you need to get a pretty accurate classifier up and running in a short time ...
- ...and you don't care about efficiency that much ...
- ...use kNN.

# kNN classification

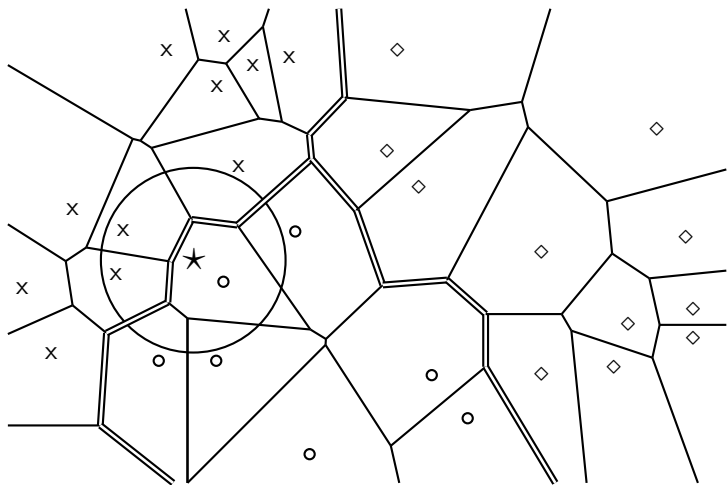
- $k\text{NN} = k$  nearest neighbors
- **kNN classification rule for  $k = 1$  (1NN):** Assign each test document to the class of its **nearest neighbor** in the training set.
- 1NN is not very robust – one document can be mislabeled or atypical.
- **kNN classification rule for  $k > 1$  (kNN):** Assign each test document to the **majority class of its  $k$  nearest neighbors** in the training set.
- Rationale of kNN: contiguity hypothesis
  - We expect a test document  $d$  to have the same label as the training documents located in the local region surrounding  $d$ .

# Probabilistic kNN

- Probabilistic version of kNN:  $P(c|d)$  = fraction of  $k$  neighbors of  $d$  that are in  $c$
- **kNN classification rule for probabilistic kNN:** Assign  $d$  to class  $c$  with highest  $P(c|d)$



kNN is based on Voronoi tessellation



kNN is a non-linear classifier! That is, separators are not “lines”.

# kNN algorithm

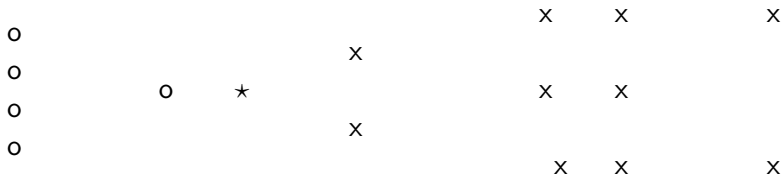
TRAIN-KNN( $\mathbb{C}, \mathbb{D}$ )

- 1  $\mathbb{D}' \leftarrow \text{PREPROCESS}(\mathbb{D})$
- 2  $k \leftarrow \text{SELECT-K}(\mathbb{C}, \mathbb{D}') // \textit{tuning}$
- 3 **return**  $\mathbb{D}', k$

APPLY-KNN( $\mathbb{D}', k, d$ )

- 1  $S_k \leftarrow \text{COMPUTENEARESTNEIGHBORS}(\mathbb{D}', k, d)$
- 2 **for each**  $c_j \in \mathbb{C}(\mathbb{D}')$
- 3 **do**  $p_j \leftarrow |S_k \cap c_j|/k$
- 4 **return**  $\arg \max_j p_j$

# Exercise



How is star classified by:

(i) 1-NN (ii) 3-NN (iii) 9-NN (iv) 15-NN (v) Rocchio?

# Software

- TiMBL: <http://ilk.uvt.nl/timbl/>
- Weka: <http://www.cs.waikato.ac.nz/ml/weka/> and <http://www.programcreek.com/2013/01/use-k-nearest-neighbors-knn-classifier-in-java/>

# Time complexity of kNN

## **kNN with preprocessing of training set**

training  $\Theta(|\mathbb{D}|L_{\text{ave}})$

testing  $\Theta(L_a + |\mathbb{D}|M_{\text{ave}}M_a) = \Theta(|\mathbb{D}|M_{\text{ave}}M_a)$

- kNN test time proportional to the size of the training set!
- The larger the training set, the longer it takes to classify a test document.
- kNN is inefficient for very large training sets.

# kNN: Discussion

- No training necessary
  - But linear preprocessing of documents is as expensive as training Naive Bayes.
  - We always preprocess the training set, so in reality training time of kNN is linear.
- kNN is very accurate if training set is large.
- But kNN can be very inaccurate if training set is small.

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