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)

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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

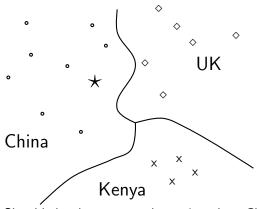
Recall vector space representation

- 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

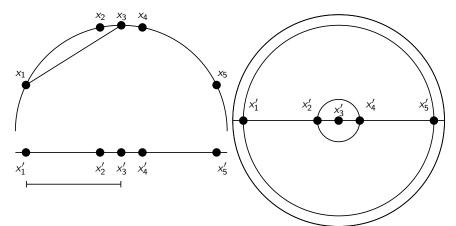
- As before, the training set is a set of documents, each labeled with its class.
- In vector space classification, this set corresponds to a labeled set of points or vectors in the vector space.
- Premise 1: Documents in the same class form a contiguous region.
- Premise 2: Documents from different 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 $x_1, \overline{x_2}, \overline{x_3}, x_4, x_5$ at x coordinates -0.9, -0.2, 0, 0.2, 0.9 the distance $|x_2x_3| \approx 0.201$ only differs by 0.5% from $|x_2'x_3'| = 0.2$; but $|x_1x_3|/|x_1'x_3'| = d_{\text{true}}/d_{\text{projected}} \approx 1.06/0.9 \approx 1.18$ 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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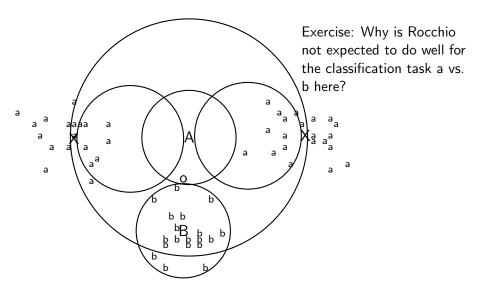
Rocchio classification: Basic idea

- 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



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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.

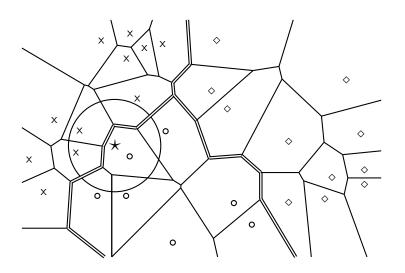
kNN classification

- kNN = 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}') / 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_i \in \mathbb{C}(\mathbb{D}')
 3 do p_i \leftarrow |S_k \cap c_i|/k
      return arg max_i p_i
```

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_{ave})
testing \Theta(L_a + |\mathbb{D}|M_{ave}M_a) = \Theta(|\mathbb{D}|M_{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.
- Question: Can we divide up the training set into regions, so that we only have to search in one region to do kNN classification for a given test document? (which perhaps would give us better than linear time complexity)

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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