

Alternative Models for IR

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Alternative Models for IR

- Set-Based Model
- Extended Boolean Model
- Fuzzy Set Model
- The Generalized Vector Model
- Latent Semantic Indexing
- Neural Network for IR
- Cluster Model

Cluster Model

- This model is an attempt to reduce the number of matches during retrieval.
- Hypothesis: **Closely associated documents tend to be relevant to the same clusters.**
- Instead of matching the query with every documents in the collection, it is matched with representatives of the class.

Cluster Model - Algorithm

- Let $D=\{d_1,d_2,\dots,d_m\}$ – finite set of documents
- Let $E=(e_{ij})_{n,n}$ – similarity matrix (or distance matrix)
- Let T be the threshold
- Any pair of documents d_i and d_j ($i \neq j$) whose similarity measures exceeds T or whose distance is less than T is grouped to form a cluster
- The remaining document form a single cluster.
- A representative vector of each class is constructed by computing the centroid of the document vector (mean)
- During retrieval, the query is compared with the cluster vectors based on similarity or distance

Steps with Examples

- Let term-document matrix

	d1	d2	d3
A =	1	1	0
	0	1	0
	1	1	1
	0	0	1
	1	1	0

- The similarity matrix

	d1	d2	d3
d1	1.0		
d2	0.9	1.0	
d3	0.4	0.4	1.0

Steps with Examples

- Threshold $T = 0.7$
- We get two clusters
 $C1 = \{d1, d2\}$ $C2 = \{d3\}$
- The cluster vectors (representatives) for C1 and C2
 $r1 = (1 \quad 0.5 \quad 1 \quad 0 \quad 1)$
 $r2 = (0 \quad 0 \quad 1 \quad 1 \quad 0)$
- Retrieval is performed by matching the query vector with $r1$ and $r2$

Exercise

- Given term-document matrix, find document cluster which is relevant for the query “to do”

To do is to be.
To be is to do.

d_1

To be or not to be.
I am what I am.

d_2

I think therefore I am.
Do be do be do.

d_3

Do do do, da da da.
Let it be, let it be.

d_4

		d_1	d_2	d_3	d_4
1	to	3	2	-	-
2	do	0.830	-	1.073	1.073
3	is	4	-	-	-
4	be	-	-	-	-
5	or	-	2	-	-
6	not	-	2	-	-
7	I	-	2	2	-
8	am	-	2	1	-
9	what	-	2	-	-
10	think	-	-	2	-
11	therefore	-	-	2	-
12	da	-	-	-	5.170
13	let	-	-	-	4
14	it	-	-	-	4

wt(to) – 1, wt(do) – 0.415

Fuzzy Model

- In fuzzy model, the document is represented as a fuzzy set of terms $[t_i, \mu(t_i)]$ where μ is the membership function
- μ assigns a membership degree to each term of the document
- The membership degree expresses the significance (weights) of term to the information contained in the documents

Fuzzy Model - Algorithm

- Each term t_i is represented by a fuzzy set f_i
- Fuzzy set operators are applied to obtain the desired result
- For single term query $q=tq$, documents from the fuzzy set f_q are retrieved
- For AND query $q=tq1 \wedge tq2$
 - Fuzzy sets f_{q1} and f_{q2} are obtained
 - Fuzzy intersection operator is used to obtain the resultant set
$$f_{q1} \wedge f_{q2} = \min\{(d_j, w_{q1}), (d_j, w_{q2})\}$$
 - The documents in this set are returned
- For OR query $q=tq1 \vee tq2$
 - Fuzzy sets f_{q1} and f_{q2} are obtained
 - Fuzzy intersection operator is used to obtain the resultant set
$$f_{q1} \vee f_{q2} = \max\{(d_j, w_{q1}), (d_j, w_{q2})\}$$
 - The documents in this set are returned

Example

- $D1 = \{\text{information, retrieval, query}\}$
- $D2 = \{\text{retrieval, query, model}\}$
- $D3 = \{\text{information, retrieval}\}$
- Vocabulary = $\{\text{information, model, query, retrieval}\}$
- The fuzzy sets induced by these terms are
 - $f1 = \{(d1, 1/3), (d2, 0), (d3, 1/2)\}$
 - $f2 = \{(d1, 0), (d2, 1/3), (d3, 0)\}$
 - $f3 = \{(d1, 1/3), (d2, 1/3), (d3, 0)\}$
 - $f4 = \{(d1, 1/3), (d2, 1/3), (d3, 1/2)\}$
- Query $q = t2 \wedge t4$ (model retrieval)
 - Fuzzy sets $f2$ and $f4$ are considered
 - $\text{Min}(f2(d1), f4(d1)), \text{Min}(f2(d2), f4(d2)), \text{Min}(f2(d3), f4(d3))$
 - $= \{(d1, 0), (d2, 1/3), (d3, 0)\}$
 - $d2$ is returned

Try : $q = t1 \vee t4$