# TDT4117 - Assignment 1

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## Task 1

1

### Explain the main differences between Data retrieval and Information retrieval

IR deals with unstructured data and small errors are allowed. With Information retrieval we want to retrieve *information* rather than *data*.

Data retrieval deals with structured data, allows no errors. Either we get a precise result, or no result at all.

I like to think about data retrieval as retrieving data from a database and information retrieval as getting information about a subject.

2

#### Explain the main differences between Structured data and Unstructured data

Structured data is a pre-defined data model and straightforward and easy to analyse. Such data can for example be data retrieved with an SQL-query.

Unstructured data on the other hand does not have this pre-defined model, and are therefore harder to analyse. It can be for example text-heavy (such as a document), audio or video.

### Task 2

#### Explain:

- 1. Term frequency (*tf*)
  - Term frequency is a factor used in calculating the ranking of a document. It represents the frequency of a given term (word) in a document.
- 2. Document frequency (df)
  - Df is the number of documents that contains a specific term.

- 3. Inverse document frequency (*idf*)
  - The inverse document frequency is a factor that diminishes the weight of terms that occur often, such as the term "the".
- 4. Why *idf* is important for term weighting
  - *idf* is important because it introduces less weight to the less important terms, such as "the" and more weight to important terms such as "onomatopoeia". This is done to get a more correct ranking of documents

## Task 3

Given the following document collection containing words from the set O = Big, Cat,Small, Dog, answer the questions in subtasks 3.1 and 3.2

```
doc1 = \{Big \ Cat \ Small \ Dog\}
doc2 = \{Dog\}
doc3 = \{Cat \ Dog\}
doc4 = \{Big \ Cat \ Big \ Small \ Cat \ Dog\}
doc5 = \{Big \ Small\}
doc6 = \{Small \ Cat \ Dog \ Big\}
doc7 = \{Big \ Big \ Big\}
doc8 = \{Dog \ Cat \ Cat\}
doc9 = \{Cat \ Small\}
doc10 = \{Small \ Small \ Big \ Dog\}
```

#### Subtask 3.1

Given the following queries:

1

# Which of the documents will be returned as the result for the above queries using the Boolean model? Explain your answers and draw a figure to illustrate.

q/doc	1	2	3	4	5	6	7	8	9	10
1	Yes	No	Yes	Yes	No	Yes	No	Yes	No	No
2	Yes	No	No	Yes	No	Yes	No	No	Yes	No
3	Yes	No	Yes							
4	No	Yes	Yes	No	No	No	No	Yes	No	No
5	Yes	No	Yes	Yes	No	Yes	No	Yes	Yes	No

For query 1, I have checked if the documents contains both cat and dog, the ones that do get's retruned. The same for query 2. For query 3 I have returned all the documents that contains either dog or big. For query 4 the ones that has dog and not small and for query 5 all the documents that contains cat.

2

# What is the dimension of the vector space representing this document collection when you use the vector model and how is it obtained?

Each dimension represents a term. Since there are four terms in the collection vocabulary the dimension is 4.

Calculate the weights for the documents and the terms using tf and idf weighting. Put these values into a document-term-matrix. (Tip: use the equations in the book and state which one you used.)

term/doc	1	2	3	4	5	6	7	8	9	10
Big	1	0	0	2	1	1	2.58	0	0	1
Cat	1	0	1	2	0	1	0	2	1	0
Small	1	0	0	1	1	1	0	0	1	2
Dog	1	1	1	1	0	1	0	1	0	1

Term frequency

Term	Ni	$\log_2 \frac{N}{n_i}$
Big	6	0.74
Cat	6	0.74
Small	6	0.74
Dog	7	0.51

Inverse Document frequency

Term/doc	1	2	3	4	5	6	7	8	9	10
Big	0.74	0	0	0	0.74	0.74	1.9	0	0	0.74
Cat	0.74	0	0.74	1.48	0	0.74	0	1.48	0.74	0
Small	0.74	0	0	0	0.74	0.74	0	0	0.74	1.48
Dog	0.51	0.51	0.51	0.51	0	0.51	0	0.51	0	0.51

$$(1 + \log_2 t f_{i,j}) * \log_2 \frac{N}{n_i}$$

4

Study the documents 2, 3, 5 and 7 and compare them to document 9. Calculate the similarity between document 9 and these four documents according to Euclidean distance. (Use tf-idf weights for your computations).

The Euclidean distance is given by:

$$d(p,q) = \sqrt{(p_1 - q_1)^2 + (p_2 - q_2)^2 + (p_3 - q_3)^2 + (p_5 - q_5)^2}$$

Using the weights I get:

$$d(9,2) = \sqrt{(0-0)^2 + (0.74-0)^2 + (0.74-0)^2 + (0-0.51)^2} = \sqrt{0+0.5476+0.5476+0.2601} = 1.1641$$

$$d(9,3) = \sqrt{(0-0)^2 + (0.74-0.74)^2 + (0.74-0)^2 + (0-0.51)^2} = \sqrt{0+0+0.5476+0.2601} = 0.8987$$

$$d(9,5) = \sqrt{(0-0.74)^2 + (0.74-0)^2 + (0.74-0.74)^2 + (0-0)^2} = \sqrt{0.5476+0.5476+0+0} = 1.046$$

$$d(9,7) = \sqrt{(0-1.9)^2 + (0.74-0)^2 + (0.74-0)^2 + (0-0)^2} = \sqrt{0.5476+0.5476+0+0} = 2.169$$

5

For the vocabulary  $v = \{\text{Big Small Dog Cat}\}\ \text{i get:}$ 

$$d_1 = \{1, 1, 1, 1\}$$

$$q = \{0, 0, 0, 1\}$$

I'll just show the computation for  $d_1$ , then I'll summarize the rest in a table:

$$sim(d_1,q) = \frac{(1*0) + (1*0) + (1*0) + (1*1)}{\sqrt{1^2 + 1^2 + 1^1 + 1^1} * \sqrt{0^2 + 0^2 + 0^2 + 1^2}} = \frac{1}{2}$$

	Doc1	Doc2	Doc3	Doc4	Doc5	Doc6	Doc7	Doc8	Doc9	Dco10
Sim(doc,q)	0.5	0	0.707	0.63	0	0.5	0	0.707	0.707	0

When ranked, we get: [Doc3, Doc8, Doc9] [Doc4] [Doc1, Doc6] [Doc2, Doc5, Doc7, Doc10]

#### Subtask 3.2

Given the following queries: q1 = "Cat Dog", q2 = "Small"

1

# What are the main differences between BM25 model and the probabilistic model introduced by Robertson-Jones?

The main difference is that the Robertson-Jones model ws originally a framework for future models and that it does not have any weighted index terms like BM25. Also, it has no accurate estimate for the first probabilities.

2

I'll use the formula

$$RSV_d = \sum_{t \in q} \log_2 \left[ \frac{N}{dt_f} \right] \cdot \frac{(k1+1) \cdot t f_{td}}{k1((1-b) + b(\frac{L_d}{L_{ave}})) + t f_{td}}$$

for my calculations. I will show calculation for only one document and sumarize the rest in a table:

Generally in every calculation, I have:

$$L_{ave} = 3.1$$
 $N = 10$ 
 $dt_{cat} = 6$ 
 $dt_{dog} = 7$ 
 $dt_{small} = 6$ 

for document 1 and query 1 I have:

$$L_{d} = 4$$

$$tf_{cat} = 1$$

$$tf_{dog} = 1$$

$$tf_{small} = 1$$

which gives me:

$$\begin{split} RSV_d &= \sum_{t \in q} \log_2 \left[ \frac{N}{dt_f} \right] \cdot \frac{(k1+1) \cdot t f_{td}}{k1((1-b) + b(\frac{L_d}{L_{ave}})) + t f_{td}} \\ &= \log_2 \left[ \frac{N}{dt_{cat}} \right] \cdot \frac{(k1+1) \cdot t f_{cat}}{k1((1-b) + b(\frac{L_d}{L_{ave}})) + t f_{cat}} + \log_2 \left[ \frac{N}{dt_{dog}} \right] \cdot \frac{(k1+1) \cdot t f_{dog}}{k1((1-b) + b(\frac{L_d}{L_{ave}})) + t f_{dog}} \\ &= \log_2 \left[ \frac{10}{6} \right] \cdot \frac{(1.2+1) \cdot 1}{1.2((1-0.75) + 0.75(\frac{4}{3.1})) + 1} + \log_2 \left[ \frac{10}{7} \right] \cdot \frac{(1.2+1) \cdot 1}{1.2((1-0.75) + 0.75(\frac{4}{3.1})) + 1} \\ &= (\log_2 \frac{10}{6} + \log_2 \frac{10}{7}) \cdot \frac{682}{763} \\ &\approx 1.12 \end{split}$$

Below is a table summarizing:

q/doc	1	2	3	4	5	6	7	8	9	10
1	1.1186	0.7118	1.4640	1.1744	0.0	1.1186	0.0	1.5440	0.8621	0.4599
2	0.6587	0.0	0.0	0.5329	0.8621	0.6587	0.0	0.0	0.8621	0.9368

The ranking is now:

For Query 1: Doc8 Doc3 Doc4 [Doc1 Doc6] Doc9 Doc2 Doc10 [Doc5 Doc7]

For Query 2: Doc10 [Doc5 Doc9] [Doc1 Doc6] Doc4 [Doc2 Doc3 Doc7 Doc8]