CSE 535

Project 3: Evaluation of IR Models

Information Retrieval



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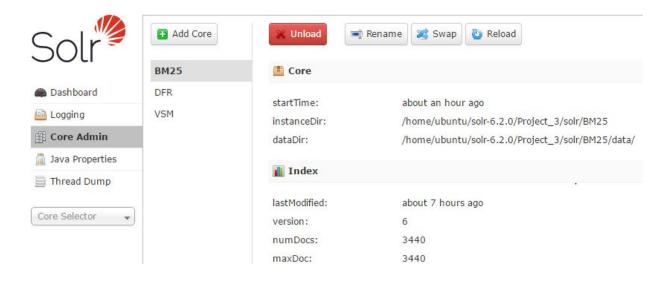
OVERVIEW

The objective of this IR project is to implement the three IR models namely Vector Space Model, BM25 and Divergence from Randomness model using Solr, evaluate the IR system and improve the search result based on the understanding of the models.

- BM25: The Best Matching (BM25) algorithm is a probabilistic Information Retrieval (IR) model. Its often called Okapi Weighting, which is a ranking function in Information Retrieval used by search engines to rank matching documents according to their relevance to given search query.
- Vector Space Model: Also known as Term Vector Model, is an algebraic model used for representing documents and queries as vectors in the term space. It allows computing a continuous degree of similarity between queries and documents, ranking documents according to their possible relevance, partial matching.
- □ Divergence from Randomness: The Divergence from Randomness (DFR) paradigm is a generalisation of one of the very first models of Information Retrieval, Harter's 2-Poisson indexing-model. It is based on the following components: Randomness Model, First Normalization and Term Frequency Normalization.

We successfully implemented the 3 models and created 3 cores for the same. Changes were made as we made progress on improvisation of the model

Please see the snapshot of the Solr UI displaying the 3 core named according to their models.



BM25 Model

<u>Initial Implementation</u>: This is the default model. In order to implement the model changes were made in the schema.xml file by inserting the following statement.

```
<similarity class="solr.BM25SimilarityFactory">
```

On calculating the MAP value through tree eval the result received was **0.6554**.

Please find below the screenshot for the same:

	100000000000000000000000000000000000000	armina and
P_1000	020	0.0070
runid	all	BM25
num_q	all	20
num ret	all	381
num_rel	all	305
num_rel_ret	all	159
map	all	0.6554
gm map	all	0.5831
Rprec	all	0.6400
bpref	all	0.6653
recip_rank	all	1.0000
iprec_at_recall_0.00	all	1.0000
iprec_at_recall_0.10	all	0.9846
iprec_at_recall_0.20	all	0.9500
iprec_at_recall_0.30	all	0.8657
iprec_at_recall_0.40	all	0.8323
iprec_at_recall_0.50	all	0.6871
iprec_at_recall_0.60	all	0.5222
iprec_at_recall_0.70	all	0.4233
iprec_at_recall_0.80		0.3633
iprec_at_recall_0.90	all	0.3133
iprec_at_recall_1.00		0.3133
P_5	all	0.8800
P_10	all	0.6750
P_15	all	0.4967

<u>Improvement</u>: In order to get a better MAP value, we played around the values of the solr.BM25SimilarityFactory class in order to get the improved score.

 K_1 : This parameter controls how quickly an increase in term frequency results in term-frequency saturation. The default value is 1.2 Lower values result in quicker saturation, and higher values in slower saturation.

b: This parameter controls how much effect field-length normalization should have. The default is 0.75.

Some of the value pair used in order to end up to a final value

```
\rightarrow (k<sub>1</sub>,b)=(1.0,1.0) =>MAP=0.6363
```

$$\rightarrow$$
 (k₁,b)=(1.0,0.7) =>MAP=0.6563

$$\rightarrow$$
 (k₁,b)=(1.25,0.7) => MAP=0.6573

Final value:

 $k_1 = 1.25$

b = 0.69

MAP = 0.6871

Below are the changes in the schema.xml file

On making further changes in values of b and k_1 there is negligible change in the MAP value so we decided to make changes in the **schema.xml** file by adding those filter classes that would enhance the performance and in turn the MAP value.

Please see the screenshot below:

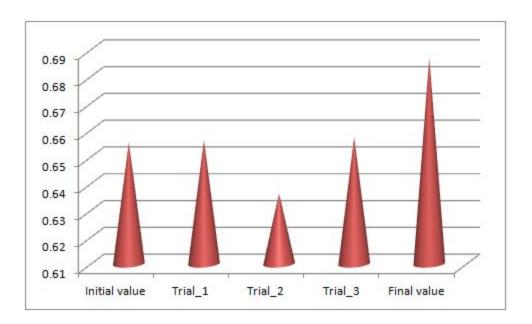
We have added following Filter classes in schema.xml:

- **Porter Stem Filter** This filter applies Porter Stem Algorithm for English, it returns the root word while dropping the suffix(es). It has been benchmarked as four times faster than the English Snowball stemmer, so can provide a performance enhancement.
- **Beider Morse Filter** This filter allows identification of similar names, even if they are spelled differently or in different languages.
- N Gram Filter It generates n-gram tokens of sizes in the given range of minGramSize and maxGramSize.

After doing the above changes we successfully landed to an improved MAP value.

P_1000	020	0.0070
runid	all	BM25
num q	all	20
num ret	all	651
num_rel	all	305
num rel ret	all	180
map	all	0.6871
gm map	all	0.6133
Rprec	all	0.6565
bpref	all	0.7086
recip rank	all	1.0000
iprec_at_recall_0.00	all	1.0000
iprec at recall 0.10	all	0.9846
iprec at recall 0.20	all	0.9500
iprec at recall 0.30	all	0.8657
iprec_at_recall_0.40	all	0.8315
iprec at recall 0.50	all	0.7190
iprec_at_recall_0.60	all	0.5928
iprec at recall 0.70	all	0.5125
iprec_at_recall_0.80	all	0.4844
iprec_at_recall_0.90	all	0.3312
iprec_at_recall_1.00	all	0.3312
P_5	all	0.8800
P_10	all	0.6750
P 15	all	0.5000
P 20	all	0.3975
P 30	all	0.2900
P 100	all	0.0900
P 200	all	0.0450
P_500	all	0.0180
0	111	0.0000





Divergence From Randomness (DFR model)

There are three components of DFR (strings).

<u>Initial Implementation:</u> To implement this model in the default mode, a similarity class was added to the schema.xml file as under with the default values with respect to the information given in the project guidelines.

On calculating the MAP value through trec eval the result received was **0.6468**

Improvement: In order to increase the map score, there was need to change the values of the 3 strings.

We have to put in the right combination as much possible of **basicModel**, **afterEffect** and **normalization** in order to improve the score.

We finally settled for the following:

1.basicModel:**Be**: Limiting form of Bose-Einstein

2. afterEffect:**B**: Ratio of two Bernoulli processes

3.normalization: **H2**: term frequency density inversely related to length

With Improved MAP value =0.6965

Some of the trial values used in order to end up to a final value were:

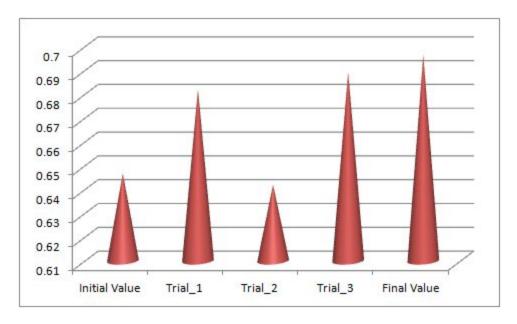
- → (basicModel, afterEffect, normalization) => (D: Divergence approximation of the Binomial,B: Ratio of two Bernoulli processes, H2: term frequency density inversely related to length) MAP =0.6822
- → (basicModel, afterEffect, normalization) =>(P: Poisson approximation of the Binomial, B: Ratio of two Bernoulli processes, H3: term frequency normalization provided by Dirichlet prior)

 MAP=0.6420
- → (basicModel, afterEffect, normalization)=> (Be: Limiting form of Bose-Einstein, L: Laplace's law of succession, H2: term frequency density inversely related to length) MAP=0.6894

Along with the additions in schema.xml (similar to that for BM25 model) we also modified the query and added The DisMax query parser to improve the score. It is designed to process simple phrases entered by users and to search for individual terms across several fields using different weighting (boosts) based on the significance of each field.

Similar changes were made to text en and text ru.

The Graphical Analysis of DFR model is:



Vector Space Model(VSM)

The vector space model procedure can be divided into three stages.

The first stage is the document indexing where content bearing terms are extracted from the document text.

The second stage is the weighting of the indexed terms to enhance retrieval of document relevant to the user.

The last stage ranks the document with respect to the query according to a similarity measure.

Similarity Class for VSM: ClassicSimilarityFactory was added to schema.xml

On running the default settings for VSM the result obtained for MAP value was **0.6469**.

```
P_200
                                        820
                                                     0.0350
  500
                                        020
                                                     0.0140
  1000
                                        820
                                                     0.0070
runtd
                                        all
                                                      NSM
num_q
                                        all
                                                      381
num_rel
                                                      305
num_rel_ret
                                                      0.6469
map
gm_map
                                                     0.5739
                                                     8.6573
Rprec
bpref
                                        all
                                                     0.6578
recip_rank
                                                     1.0000
iprec_at_recall_0.00
iprec_at_recall_0.10
iprec_at_recall_0.20
iprec_at_recall_0.30
                                        all
                                                     1.0000
                                        all
                                                     0.9846
                                        all
                                                     8.9464
                                        all
                                                     0.8561
iprec_at_recall_0.30
iprec_at_recall_0.40
iprec_at_recall_0.50
iprec_at_recall_0.60
iprec_at_recall_0.70
iprec_at_recall_0.80
iprec_at_recall_0.90
iprec_at_recall_1.00
P 5
                                        all
                                                     0.8373
                                                     0.6477
                                                      0.5020
                                                     0.4864
                                        all
                                                     0.3714
                                        all
                                                     0.3231
                                        all
                                                     0.3231
                                        all
P_5
P_10
                                                     0.8600
                                                     0.6600
                                        all
                                                     0.4900
                                        all
  28
                                                     0.3850
  38
                                                     0.2567
                                                      0.0770
   200
                                                      0.0385
                                                      0.0154
                                                      8.6877
```

In order to improve the performance of this model we added some more stopwords to **stopwords_en.txt**, **stopwords_de.txt**, **stopwords_ru.txt** so as to get better precision and recall values. Also we added a list of synonyms of common words to **synonyms.txt** so as to get better MAP result. These two changes can be seen in the screenshot below:

Following Filter classes have been added:

Synonym Filter - Each token is looked up in the list of synonyms and if a match is found, then the synonym is emitted in place of the token. The position value of the new tokens are set such they all occur at the same position as the original token.

☐ **Hyphenated Words Filter** - This filter is used to use hyphenated terms as one single terms in order to provide better matching results at the time of query searching and matching.

Both these filters have been added to provide better MAP results in the VSM model and it is done so by performing better matching of terms in the document with the query terms.

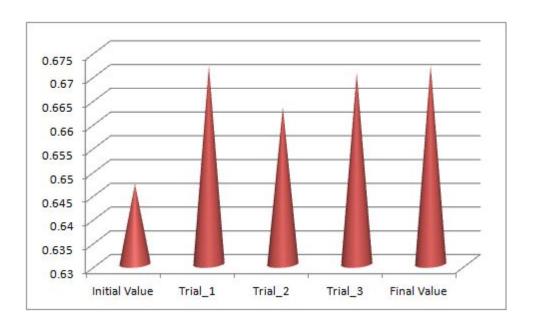
Before arriving to the final MAP value we had some intermediated values as

- \rightarrow MAP = 0.6728
- → MAP=0.6631
- → MAP=0.6704

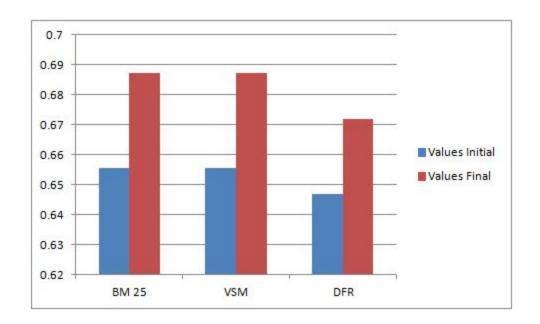
The final improved MAP value is **0.6719** which can be seen in the screenshot below:

num_ret	all	561
num rel	all	305
num rel ret	all	170
мар	all	0.6719
gm map	all	0.5978
Rprec	all	0.6738
bpref	all	0.6903
rectp rank	all	1.0000
iprec at recall 0.00	all	1.0800
iprec at recall 0.10	all	0.9846
iprec at recall 0.20	all	0.9464
iprec_at_recall_0.30	all	0.8578
iprec at recall 0.40	all	0.8391
iprec at recall 0.50	all	0.6682
iprec at recall 0.60	all	0.5570
iprec at recall 0.70	all	8.4926
iprec_at_recall_0.80	all	0.4445
iprec at recall 0.90	all	0.3435
iprec at recall 1.00	all	0.3435
P 5	all	0.8600
P 18	all	8.6688
P 15	all	8,4986
P 20	all	0.3925
P 30	all	0.2833
P 100	all	0.0850
P 200	all	0.0425
P 500	all	8.0170
P 1000	all	0.0085

The Graphical Analysis of VSM model is:



Collective Graphical Analysis of the three models



Conclusion

The three IR models have been implemented successfully and we tried to improve the performance of all the three models by using their properties and the features provided by Solr 6.2.0.

In BM25 model we settled for values of b and k_1 as 0.69 and 1.25 respectively, as better results were achieved through these values and also by adding N gram filter, Beider Morse filter and Porter Stem Filter.

In DFR model with the help of the 3 components we improved the MAP value along with some changes in the schema.xml.

In VSM model we added Synonym Filter, Hyphenated Filter as well added more stopwords to the stopwords.txt file of all three language fields.

All these provided some improvements in the system but there is a lot more scope of improvement to do in future prospects.