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Overview:

As part of this projects, the aim is to implement various IR models, evaluate the IR system and improve the search results based on the understanding of the models, the implementation and the evaluation.

Twitter data in three languages - English, German and Russian, with 15 sample queries and the corresponding relevance judgements are given. We will index the given twitter data using Solr and implement the following three IR models:

- 1. Language Model,
- 2. BM25 and
- 3. Divergence from Randomness (DFR) Model.

The results from these three sets will be evaluated using the Trec_eval program. Based on the evaluation results, we are required to attempt to improve the performance in terms of Mean Average Precision (MAP).

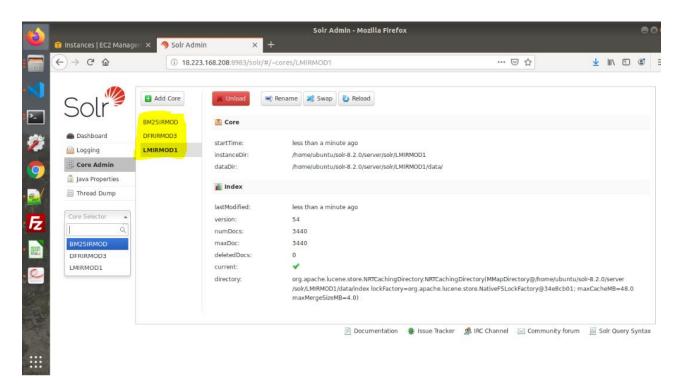
Dataset:

- 1. training tweet.json file has been provided.
- 2. The other required data and details have also been provided.

Implementation of IR Models:

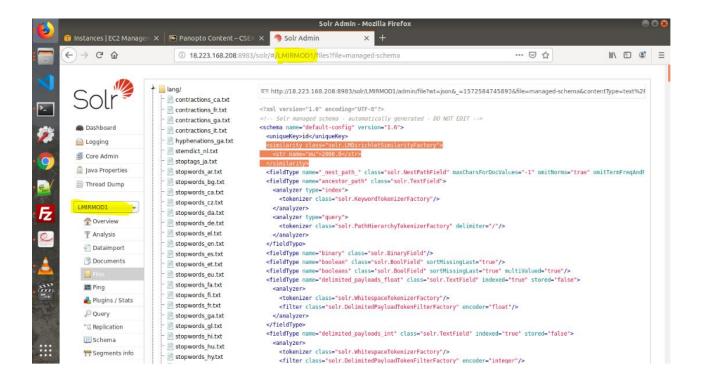
The 3 different IR models have been implemented in the Solr. Below steps are followed to create 3 different cores corresponding to each IR models with default schema in the solr.

- a) Created 3 different cores in the solr for each of the individual IR models.
- b) The default schema along with the default values of hyperparameters were used while creating the models at the first time.
- c) The given training_tweet.json dataset was posted to each of the different models in the solr.
- d) Based on the given set of training queries, the output files were generated in the Trec_eval format as mentioned in the project requirement documents.
- e) The trec_eval was run and the MAP(all) scores were obtained for each of the models separately on the training queries.
- f) Based on the MAP (all) scores for each of the IR models on default values, I tweaked few of the hyper-parameters values to achieve the best MAP(all) scores for each of the models. The observations are listed below in this report.

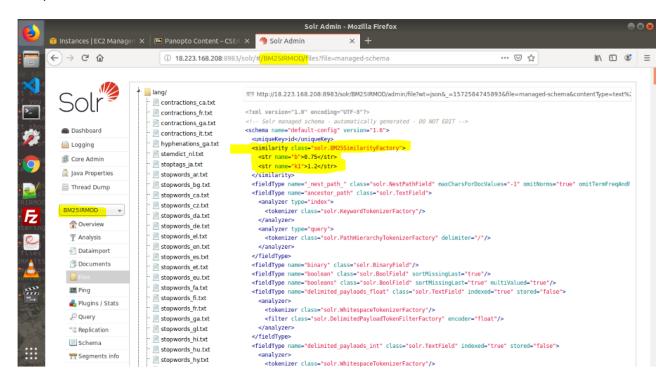


IR models with default parameters:

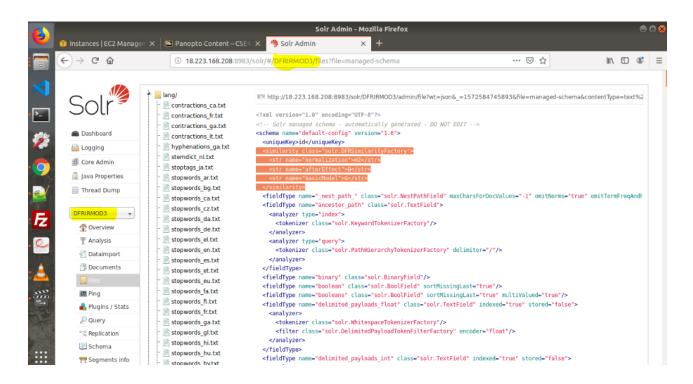
1) Language Model (LM):



2) BM25:



3) Divergence from Randomness (DFR) Model:



Improving the IR systems:

1) <u>BM25:</u>

There are two hyper-parameters available for BM25 model which are mainly "k1" and "b" values. As per understanding, the parameter "b" controls the length of the document and the parameter "k1" controls the frequency of the word.

Frequency is actually related to the length of the document and thus, if the document is large/small, the k1 value is increased/decreased to achieve good MAP score.

As per the same logic, I tried to tweak these two parameters values as shown below and obtained the MAP(all) scores respectively and compared the new MAP score with the MAP score obtained from the default values of these hyper-parameters. As shown below in the table, the MAP score corresponding to k1=1.1 & b=0.8 was good as compared to rest of the values.

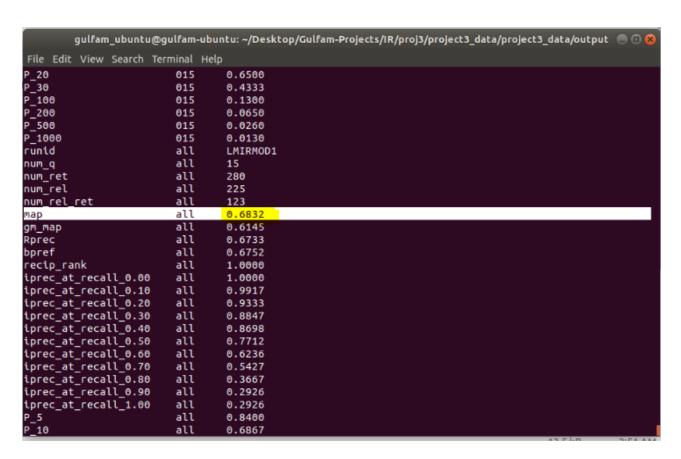
Model	Default Hyper-parameters	Default MAP(all) score	Tweaked Hyper-parameters values	New MAP(all) score	
BM25	<str name="b">0.75</str> <str name="k1">1.2</str>	0.6756	<str name="b">0.8</str>	0.6748	
			<str name="k1">1.5</str>		
			<str name="b">0.5</str>	0.6758	
			<str name="k1">1.7</str>	0.0738	
			<str name="b">0.75</str>	0.6758	
			<str name="k1">1.1</str>		
			<str name="b">0.5</str>	0.6747	
			<str name="k1">1</str>		
			<str name="b">0.7</str>	0.6733	
			<str name="k1">2</str>	0.0733	
			<str <mark="" name="b">>0.8tr></str>	0.6764	
			<str <mark="" name="k1">>1.1tr></str>	0.0704	

gulfam_ubuntu	@gulfam-u	ıbuntu: ~/Desktop/Gulfam-Projects/IR/proj3/project3_data/project3_data/output 🤎	0 8
File Edit View Search	Terminal H	Help	
P 20	015	0.6500	
P 30	015	0.4333	
P 100	015	0.1300	
P_200	015	0.0650	
P 500	015	0.0260	
P 1000	015	0.0130	
runid	all	BM25IRMOD	
num q	all	15	
num_ret	all	280	
num rel	all	225	
num rel ret	all	121	
map	all	0.6764	
gm map	all	0.6089	
Rprec	all	0.6474	
bpref	all	0.6712	
recip_rank	all	1.0000	
iprec_at_recall_0.00	all	1.0000	
iprec_at_recall_0.10		1.0000	
iprec_at_recall_0.20	all	0.9333	
iprec_at_recall_0.30	all	0.8847	
iprec_at_recall_0.40		0.8698	
iprec_at_recall_0.50		0.7365	
iprec_at_recall_0.60		0.6147	
iprec_at_recall_0.70		0.5322	
iprec_at_recall_0.80		0.3667	
iprec_at_recall_0.90		0.2815	
iprec_at_recall_1.00		0.2815	
P_5	all	0.8400	
P_10	all	0.6800	

2) <u>LM:</u>

There is a hyper-parameter called "mu" available for LM model. The default value of "mu" is used as 2000 and the corresponding default MAP(all) score 0.6135 was achieved. After analyzing and tweaking this parameter as shown in the below table, the good MAP(all) score 0.6832 was achieved at the "mu" value of 10. Increasing the "mu" value decreased the overall MAP score. Thus it depends on the document length and the frequency of the words. For a very dense document, may be the default value work fine but for smaller document, the small mu value can be considered to achieve good overall MAP score.

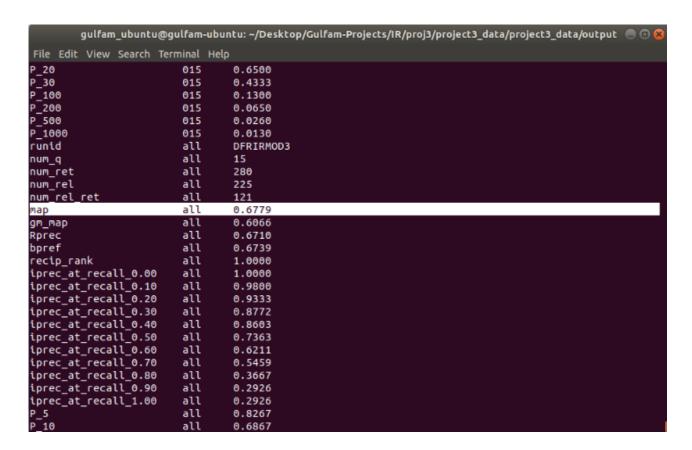
Model	Default Hyper-parameters	Default MAP(all) score	Tweaked Hyper-parameters values	New MAP(all) score
	<float name="mu">2000.0</float>	0.6135	<float name="mu">2200.0</float>	0.6105
			<float name="mu">1800.0</float>	0.6142
LM			<float name="mu">1500.0</float>	0.6178
			<float name="mu">1000.0</float>	0.6297
			<float name="mu">300.0</float>	0.6716
			<float name="mu">50.0</float>	0.6813
			<float name="mu<mark>">10.0<!--</mark-->float></float>	0.6832



3) <u>DFR:</u>

There is a hyper-parameter called "c" available for DFR model along with other parameters. The default value of "c" is used as 7 in most of the cases. After analyzing and tweaking this parameter as shown in the below table, the good MAP(all) score 0.6779 was achieved at the "c" value of 2.5. Decreasing the "c" value up to 1 increases the MAP score initially and got saturated at this point.

Model	Default Hyper-parameters	Default MAP(all) score	Tweaked Hyper-parameters values	New MAP(all) score	
	<str name="normalization">H2</str> <str name="afterEffect">B</str> <str name="basicModel">G</str>	0.675	<str name="normalization">H2</str>	0.6771	
			<str name="afterEffect">B</str>		
			<str name="basicModel">G</str>		
			<float name="c">5</float>		
			float name="c">7.0	0.677	
			<str name="normalization">H2</str>		
			<str name="afterEffect">B</str>		
			<str name="basicModel">G</str> >		
			<float name="c">2.0</float>	0.6776	
			<str name="normalization">H2</str>		
			<str name="afterEffect">B</str>		
DFR			<str name="basicModel">G</str>		
DIK			<float name="c">1.0</float>	0.675	
			<str name="normalization">H2</str>		
			<str name="afterEffect">B</str>		
			<str name="basicModel">G</str>		
			<float name="c">3.0</float>	0.6775	
			<str name="normalization">H2</str>		
			<str name="afterEffect">B</str>		
			<str name="basicModel">G</str>		
			<float name="c<mark>">2.5loat></float>		
			<str name="normalization">H2</str>	0.6779	
			<str name="afterEffect">B</str>		
			<str name="basicModel">G</str>		



Conclusion:

All the 3 different IR models were implemented successfully. Also, the improvements were done on each of the models separately by tweaking the different hyper-parameters and other factors and an increase in the overall MAP scores were observed which had been listed in this report.

References:

- 1. Class slides and project description documents
- 2. http://lucene.apache.org/solr/7 0 0/solr-core/org/apache/solr/search/similarities/package-summary.html
- 3. http://lucene.apache.org/solr/guide/7 5/other-schema-elements.html#OtherSchemaElements-Similarity
- 4. https://lucene.apache.org/solr/8 1 0//solrcore/org/apache/solr/search/similarities/LM
 DirichletSimilarityFactory.html
- 5. http://wiki.apache.org/solr/SchemaXml#Similarity
- 6. http://trec.nist.gov/trec_eval/