## Evaluation of IR Models

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### 1 Introduction

The project deals with the evaluation of different IR models using a set of queries on a corpus indexed on Solr. The three different models used are:

- 1. Vector Space Model
- 2. BM25 Model
- 3. Divergence from Randomness Model

## 2 Implementing the default configurations of VSM

With the default configuration of similarity class in schema.xml, we can implement the vector space model.

```
<similarity class="solr.ClassicSimilarityFactory"/>
```

After implementing the model, and indexing the file: train.json, we evaluate the following trec value:

```
./\operatorname{trec\_eval} \ -q \ -c \ -M1000 \ \dots / \operatorname{qrel.txt} \ \dots / \operatorname{output\_vsm.txt} \ | \ \operatorname{grep} \ \operatorname{map}
```

# ${f 3}$ Implementing the default configurations of BM25

With the following configuration of similarity class in schema.xml, we can implement the BM25 model.

```
<similarity class="solr.BM25SimilarityFactory">
  <str name="b">0.80</str>
  <str name="k1">1.3</str>
  </similarity>\newline
```

```
      deboshree-Insptron-3558:~/Downloads/trec_eval.8.1$ ./trec_eval -q -c -M1000 ../qrel.txt ../output_vsm.txt | grep map

      map
      001
      0.3801

      map
      002
      0.3983

      map
      003
      0.6062

      map
      004
      0.5724

      map
      005
      0.5000

      map
      006
      0.5257

      map
      007
      1.0000

      map
      008
      1.0000

      map
      009
      0.7448

      map
      010
      1.0000

      map
      011
      1.0000

      map
      012
      0.4359

      map
      013
      0.1027

      map
      014
      0.7169

      map
      015
      0.7721

      map
      015
      0.7721

      map
      011
      0.6503
```

After implementing the model, and indexing the file: train.json, we evaluate the following trec value:

```
./\,\,\mathrm{trec\_eval}\,\,-\mathrm{q}\,\,-\mathrm{c}\,\,-\mathrm{M1000}\,\,\ldots/\,\,\mathrm{qrel.\,txt}\,\,\ldots/\,\,\mathrm{output\_bm25.\,txt}\,\,\mid\,\,\mathrm{grep}\,\,\,\mathrm{map}
```

```
      deboshree@deboshree=Inspiron-3558:~/Downloads/trec_eval.8.1$ ./trec_eval -q -c -M1000 ../qrel.txt ../output_bm25.txt | grep m

      map
      001
      0.3588

      map
      002
      0.4071

      map
      003
      0.5729

      map
      004
      0.5484

      map
      005
      0.5000

      map
      006
      0.4895

      map
      007
      0.8333

      map
      008
      0.4901

      map
      010
      0.9111

      map
      011
      1.0000

      map
      012
      0.7086

      map
      013
      0.0901

      map
      014
      0.5942

      map
      015
      0.8667

      map
      015
      0.8667
```

# 4 Implementing the default configurations of DFR

With the following configuration of similarity class in schema.xml, we can implement the DFR model.

```
<similarity class="solr.DFRSimilarityFactory">
  <str name="c">>3.0</str>\newline
  <str name="normalization">H2</str>
  <str name="afterEffect">B</str>
  <str name="basicModel">G</str>
  </similarity>\newline
```

After implementing the model, and indexing the file: train.json, we evaluate the following tree value:

```
./\operatorname{trec\_eval} \ -q \ -c \ -M1000 \ \dots / \operatorname{qrel.txt} \ \dots / \operatorname{output\_dfr.txt} \ | \ \operatorname{grep} \ \operatorname{map}
```

```
      deboshree@Inspiron-3558:~/Downloads/trec_eval.8.1$ ./trec_eval -q -c -M1000 ../qrel.txt ../output_dfr.txt | grep ma

      map
      001
      0.3735

      map
      002
      0.4160

      map
      003
      0.5471

      map
      004
      0.5804

      map
      005
      0.5000

      map
      006
      0.4914

      map
      007
      0.8333

      map
      008
      0.4901

      map
      009
      1.0000

      map
      010
      1.0000

      map
      011
      1.0000

      map
      012
      0.7303

      map
      012
      0.7303

      map
      014
      0.5942

      map
      014
      0.5942

      map
      015
      0.8667

      map
      010
      0.6343
```

## 5 Improvement of models

#### 5.1 VSM

#### Experiment1: Applying dismax query parser:

```
< request Handler name="/select" class="solr.Search Handler"> < lst name="defaults"> < str name="defType"> edismax</str> < str name="qf"> text_en^1.5 text_de^1.2 text_ru^0.2</str> < str name="echoParams"> explicit</str> < int name="rows"> 10</int> < !— < str name="df"> text</str> < </li>
```

The dismax query parser is used for query boosting and to provide different weights to different fields thorugh qf. By assigning different weights, we could see the following changes:

wt(text-en)	wt(text-de)	wt(text-ru)	MAP-initial	MAP-modified
0.8	1.2	0.2	0.6503	0.6641
1.5	1.2	0.2	0.6503	0.6951
1.5	1.2	1.2	0.6503	0.6541
1.0	1.7	0.2	0.6503	0.6832

#### Experiment2: Applying different filter class:

```
<analyzer type="index">
<charFilter class="solr.PatternReplaceCharFilterFactory"
pattern="([@#])" replacement=""/>
</analyzer type>
```

```
<analyzer type="query">
<charFilter class="solr.PatternReplaceCharFilterFactory"
pattern="([@#])" replacement=""/>
</analyzer type>
```

MAP value almost remained unchanged after applying the filter though it did show a little deflection.

Model	Initial	Optimized
MAP	0.6503	0.6560

#### 5.2 BM25

BM25 is a bag-of-words retrieval function that ranks a set of documents based on the query terms appearing in each document, regardless of the inter-relationship between the query terms within a document (e.g., their relative proximity). It is not a single function, but actually a whole family of scoring functions, with slightly different components and parameters.

#### Experiment1: Changing the values of parameter k1 and b:

k1 - Controls non-linear term frequency normalization (saturation).

b - Controls to what degree document length normalizes the tf values.

```
< similarity class="solr.BM25SimilarityFactory">\backslash \\ < str name="b">0.69</str> \backslash \\ < str name="k1">1.6</str> \backslash \\ </ similarity>
```

# Experiment 2: Using URLTokenizer instead of standard tokenizer for text-en for analyzer type query

```
<analyzer type='query'>
<tokenizer class="solr.UAX29URLEmailTokenizerFactory"/>
</analyzer>
```

Model	Initial	Optimized
MAP	0.6119	0.6261

#### Experiment 3: Using different filters and tokenizers

```
<analyzer type='query'>
<filter class="solr.PatternReplaceFilterFactory"
pattern="([^A Z][^a z])" replacement="" replace="all"/>
</analyzer>
```

The usage of this tokenizer decreased the value of MAP, therefore we refrain from using this. After testing with various tokenizers, we obtained better results with URL Email Tokenizer.

Model	Initial	Optimized
MAP	0.6261	0.6125

#### 5.3 DFR

The DFR has three parameters BasicModel which is the basic model of the information content, AfterEffect specifies the first normalization of information gain and Normalization refers to the second normalization. A parameter 'c' that controls the term frequency normalization with respect to the document length which is specified for normalization H1 and H2.

Experiment1: Tuning the parameters

Normalization	AfterEffect	Basic Model	С	MAP
H2	В	G	3	0.6463
H2	В	G	2	0.6263
H2	В	G	4	0.6342
H2	В	G	5	0.6333

From the above we notice that by decreasing value of c parameter for the H2 normalization parameter, the MAP value increases but when it touches 2.0, the value starts decreasing. Hence we adapt a smaller value of c but not too small.

Experiment2: Using URL tokenizer

```
<analyzer type= query >
<tokenizer class="solr.UAX29URLEmailTokenizerFactory"/>
</analyzer>
```

Model	Initial	Optimized
MAP	0.6463	0.6569

#### 6 Conclusion

After optimization of the default model settings, we obtain the following MAP values:

Model	Initial Value	Optimized Value
VSM	0.6503	0.6951
BM25	0.6117	0.6263
DFR	0.6463	0.6569

