ASSIGNMENT-3

SWAGATAM CHAKRABORTI(MT18146)

1. Tf-idf based document retrieval and relevance feedback

PREPROCESSING:

- 1. Each line of the file is pre processed individually
- 2. Word tokenization is done using nltk library using the regextokenization which handles the formation of the tokens and also the removal of the punctuations
- 3. Stopwords have been removed from the tokens formed.
- 4. Lemmatization have been performed over the tokens using the nltk library
- 5. If the line contains any numbers, it is converted into words using inject library and is stored in the vocab along with the number itself

METHODOLOGY:

- Traverse through all the documents, preprocess the data and maintain a vocab dictionary for each documents having word as the key and corresponding term frequency of the word in the document.
- 2. Create a list of dictionaries, having document name as the key for each dictionary and vocab dictionary of the corresponding calculated in the previous step as the value.
- 3. For the query entered by the user is pre-processed and a processed query is obtained.
- 4. For each words in the query, calculate the inverse document frequency, fot each documents, if the word matches in the vocab of the document, the tf-idf score is multiplied and appended in the dictionary with the document name as the key and tf-idf score as the value
- 5. Finally the dictionary is sorted based on the if-idf score.
- 6. For displaying the top k relevant documents, firstly the documents present in the match with the title displayed and remaining documents among the k documents is displayed from the sorted dictionary of document tf-idf score.

- 7. Based on the top K documents user provide the feedback for the K documents.
- 8. From the relevance documents we find the relevance centroid.
- 9. From the non relevance documents we find the non relevance centroid.
- 10. Taking alpha value as 1 and beta value as 0.7 we obtain the modify query of length of vocab based on the rochio algorithm. Repeat from step 5.

RESULT AND ANALYSIS:

Query: Navy engineering software systems

Top 10 documents:

ASSIGNMENT_1Q2_dataset\comp.graphics\38609 0.23634924248300748

ASSIGNMENT_1Q2_dataset\rec.motorcycles\104529 0.22736486159739

ASSIGNMENT_1Q2_dataset\comp.graphics\37261 0.21992594302269824

ASSIGNMENT_1Q2_dataset\comp.graphics\38947 0.1891285382086621

ASSIGNMENT_1Q2_dataset\comp.graphics\38625 0.1761849536586591

ASSIGNMENT_1Q2_dataset\comp.graphics\39738 0.1620484101023273

ASSIGNMENT_1Q2_dataset\comp.graphics\39655 0.15846910912665435

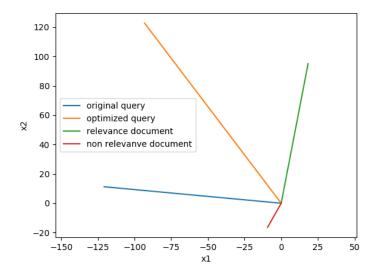
ASSIGNMENT_1Q2_dataset\comp.graphics\38976 0.1562327075167356

ASSIGNMENT_1Q2_dataset\comp.graphics\38241 0.1555031122478027

ASSIGNMENT_1Q2_dataset\comp.graphics\38855 0.15298591881092605

Relevance documents:

 $comp.graphics \ 38609, rec.motorcycles \ 104529, comp.graphics \ 37261, comp.graphics \ 38947, comp.graphics \ 38625$



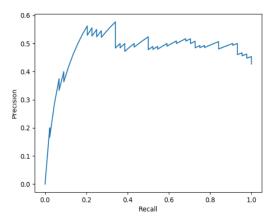
INFERENCE: From the 2D TSME plot it is evident that the modified query vector gets shifted away from the non-relevance document centroid vector.

2. PRECISION RECALL CURVE

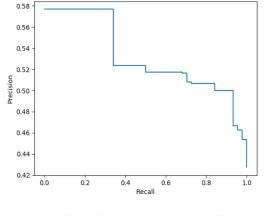
METHODOLOGY:

- 1. From the dataset obtain the queries having quid:4
- 2. From the quid:4 data obtain the feature 75 and the 0th feature i.e the relevance score.
- 3. Sort the data based on the 75th feature in descending order.
- 4. Calculate the precision and recall for each document retrieval.
- 5. Plot the precision and recall curve i.e Interpolated curve and 11 point precision and recall curve.

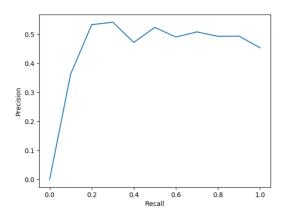
RESULT ANALYSIS:



Normal precision-recall curve



Interpolated precision-recall curve



11 point precision recall curve