version DBpedia[2]. In order to explore the potential for document expansion using general resources in SpokenDoc2, for our experiments we used the Japanese DBpedia[3] as an expansion source. As the provided dataset does not contain any segment boundaries, we consider each IPU as a document to be expanded. Afterwards the expanded IPUs can be segmented using different segmentation methods which allows us to examine the effect of the use of external knowledge on retrieval effectiveness.

This paper is structured as follows: Section 2 describes the methods we used to prepare, expand, and search the test collection, Section 3 gives details of the results achieved and analysis of the system performance, and finally Section 4 concludes and outlines directions for our future work.

2. RETRIEVAL METHODOLOGY

In this section we give an overview of the tools and methods we applied to perform the NTCIR-10 SpokenDoc passage retrieval task. Task participants were provided with several transcripts of the spoken content. We first extracted words from the transcripts and from the Japanese DBpedia collection. Since there are no topical boundaries marked within the transcripts, we applied alternative segmentation methods to define alternative target retrieval passages.

2.1 SDPWS Transcripts

Task participants were provided with both automatically and manually created transcripts of the oral presentations. Two types of transcripts were created using automatic speech recognition (ASR): n-best word-based and syllable-based [2]. Two forms of each of these transcript types were created using either matched or unmatched language models. For our participation in the task, we used both matched and unmatched 1-best word-based transcripts. For comparison we also used the manual transcript provided by the task organisers.

2.2 Japanese Data Preprocessing

In Japanese the individual morphemes of the sentences need to be recognized for further processing. We used the ChaSen system, version 2.4.0[4], based on the Japanese morphological analyzer JUMAN, version 2.0, with ipadic grammar, version 2.7.0, to extract the words from the sentences in ASR and manual transcripts, and in the external data collection - DBpedia. ChaSen provides both conjugated and base forms of the word, for later processing we used the latter since it avoids the need for stemming of different word forms.

2.3 IPU Expansion

We used the open-source Terrier information retrieval platform[5] to identify the expansion terms for each IPU. We represented each IPU as a query for the DBpedia document collection and carried out standard query expansion implemented in Terrier. We took 5 terms from the 10 top retrieved documents. The use of IPU expansion is marked in the run name by the addition of `_e` to the name of the run, in cases with no expansion `_ne` is added.

[2]http://dbpedia.org/
[3]http://ja.dbpedia.org/
[4]http://chasen-legacy.sourceforge.jp
[5]http://www.terrier.org

Table 1: Average length of relevant passages and passages containing relevant content per run (in IPUs).

Average Length of relevant passages: 7.27.

	manual	asr_matched	asr_unmatched
tt_ne	35.50	46.92	43.17
tt_e	20.78	27.75	23.92
Segm_5_3_ne	–	8.85	8.26
Segm_5_3_e	–	9.58	9.03
Segm_10_5_ne	–	23.86	19.92
Segm_10_5_e	–	23.69	21.48
Segm_15_7_ne	–	56.29	50.29
Segm_15_7_e	–	53.72	47.54

2.4 Text Segmentation

Our previous research on the Japanese data in the SpokenDoc Task at NTCIR-9 [4] showed that TextTiling [6] produces shorter segments than C99 segmentation [3], and that it achieved higher scores in retrieval. Thus we use only TextTiling lexically coherent segmentation for our SpokenDoc2 experiments. TextTiling uses cosine similarities between adjacent blocks of sentences in a text document to predict topical boundary points.

For fixed length segmentation we chose the values of 5, 10, 15 IPUs with a corresponding sliding window of 3, 5 and 7 IPUs. Our runs adopt the following naming convention: `Segm_<Segment_Length>_<Sliding_Step>`.

2.5 Retrieval Setup

For retrieval experiments we used the open-source Terrier information retrieval platform[6] with a standard language modelling method, with *lamda* equal to 0.35.

2.6 Post-editing of the results for fixed length segentation methods

The result list for the fixed length segmentation with sliding window approach contains overlapping segments. Previous research has shown that simple removal of overlapping segments further down in the retrieved ranked list is an effective method to improve retrieval effectiveness [9] [1]. However our assumption is that since our target segments might be longer than fixed length segments, it is reasonable to try to combine the segments instead of removing them. In cases where there are overlapping segments in the retrieved list, we put the overall segment at the highest of the rank positions. We carry out post-editing of the lists individually for each query, thus the new length of the segments varies depending on the search request.

3. RESULTS

In this section we give an overview of the results for the different runs according to the NTCIR-10 metrics. Figures 1, 2 and 3 show the results for the three metrics (uMAP, pwMAP and fMAP) respectively. Table 1 shows the average length of the actual relevant passages and passages containing relevant content per run (in IPUs), while Table 2 contains the average length of all the segments in the retrieved ranked lists.

[6]http://www.terrier.org

Table 2: Average length of all segments in the retrieved ranked lists (in IPUs).

	manual	asr_matched	asr_unmatched
tt_ne	39.78	52.03	49.80
tt_e	23.07	30.13	27.24
Segm_5_3_ne	–	7.01	7.05
Segm_5_3_e	–	7.26	7.44
Segm_10_5_ne	–	15.26	15.39
Segm_10_5_e	–	15.55	15.71
Segm_15_7_ne	–	27.69	27.99
Segm_15_7_e	–	27.66	28.29

Comparison of Tables 1 and 2 demonstrates an interesting trend: segments containing relevant content are shorter than average segments for lexical coherence based segmentation runs, while the fixed length segmentation runs follow the opposite trend.

Across all metrics the runs that use the language model (LM) that match the collection show better results than the runs which use the unmatched LM. Use of DBpedia for IPU expansion does not help the asr_unmatched runs to achieve the same scores as asr_matched. However the expansion consistently improves the results of longer segments (tt, Segm_10_5, Segm_15_7) for asr_unmatched runs according to the pwMAP score.

The pwMAP metric only counts as relevant segments for which the IPU in the centre of the segment is relevant. Since shorter segments have a greater likelihood of having the relevant content in the centre, runs Segm_5_3 achieve higher pwMAP scores for asr_matched transcripts. Since the fMAP metric is designed to capture the relevancy of segments, the Segm_5_3 runs receive higher fMAP scores as well.

4. CONCLUSION

This paper reports the methods and results for our participation in the NTCIR-10 SpokenDoc2 passage retrieval task. As could have been expected runs using the matched ASR transcript achieve better results than those using the unmatched transcripts. However performance of runs using the unmatched ASR transcript can sometimes be improved with the use of DBPedia as a general external knowledge source for document expansion. This improvement is only captured by one of the benchmark metrics. Further investigation of document expansion will focus on understanding how it modifies retrieval as measured by the other retrieval metrics, and will seek to develop methods to apply it more reliably to improve overall retrieval effectiveness.

5. ACKNOWLEDGMENTS

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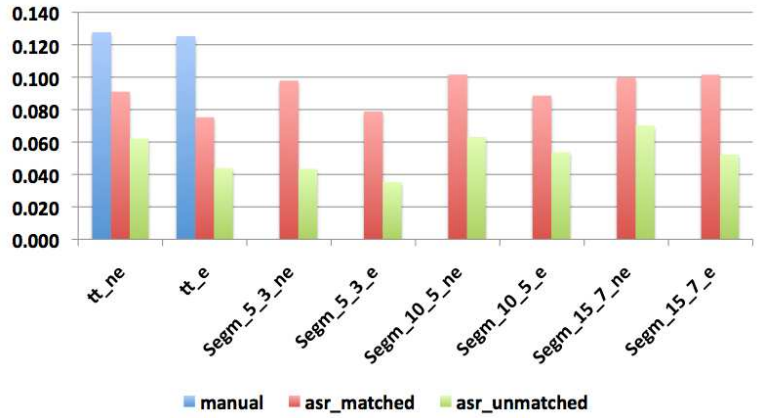

Figure 1: Scores for Utterance-based Measure (uMAP).

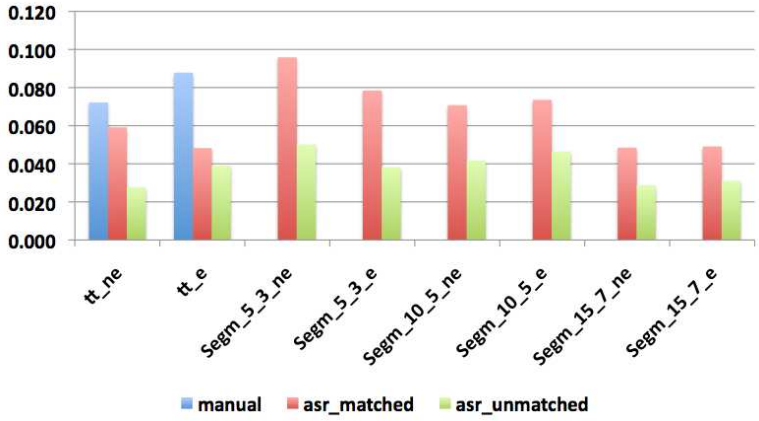

Figure 2: Scores for pointwise MAP (pwMAP).

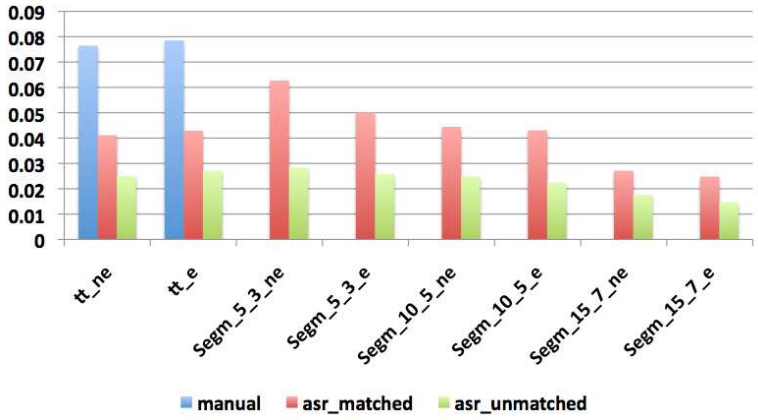

Figure 3: Scores for fraction MAP (fMAP).