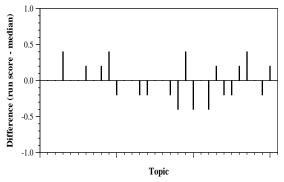
## Round 1 results — Run ir\_covid19\_cle\_ib submitted from IR\_COVID19\_CLE

## **Run Description**

We have used the data set with all the documents from corpus Commercial use subset, Non-commercial use subset, Custom license subset and bioRxiv/medRxiv subsets. We used "Paper\_id", "Title Id" and "Abstract" to index all the documents using Apache Lucene. We have indexed every document for all tokens present with in the document. However, in a collection of documents these tokens can be repeating in multiple documents as well. Here, we use inverted index to store tokens repeating in multiple indexes, so that when searched for a specific token, we can narrow down the search documents specifically all documents that token is present. We have used the query of the topic for querying the index. We parsed the query with English Analyzer and searched on the abstract text field of the index. For each query, We have retrieved the Top 100 documents and the relevance scores using Information based (IB Similarity) which models rely on normalized values of occurrence of a word in documents. Information Models are characterized by three elements that are normalization function, probability distribution and retrieval function. Reference Paper: Clinchant , S., & Gaussier, E. (2010). Information-based models for ad hoc IR. In Proceeding of the 33rd international ACM SIGIR conference on Research and development in information retrieval (SIGIR '10).

Summary Statistics			Overall measures Number of topics 30	
	19_cle_ib		Number of topics	
	utomatic		l number retrieved	2559
Contributed to judgment sets?	no		l relevant	2352
			l relevant retrieved	361
		MAF		0.0814
			n Bpref	0.1517
		Mear	n NDCG@10	0.3342
		_		
1.0	,		nent Counts at Rank 50	,
		50 -		
Document Level Averages 0.8-				
Precision		40 -		
At 5 docs 0.4867		Number of Dcuments		
At 10 docs         0.3867         5           At 15 docs         0.3533         2		30	n n	
At 15 docs 0.3533		0 J0		<ul> <li>Judged</li> <li>Relevant</li> </ul>
At 20 docs $0.3083$			III. III IA AIAAAA	
At 30 docs 0.2600				
R-Precision 0.2				
Exact 0.1560				
	2 0.4 0.6 0.8 1. Recall	0 0 <del>1 0 0 10</del> 0 10		4
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Торіс		Торіс		
Per-topic difference from median bpref for all Round 1 runs		Per-topic difference from median NDCG@10 for all Round 1 runs		

 $\underline{Round \ 1 \ results} - \underline{Run \ ir\_covid19\_cle\_ib \ submitted \ from \ IR\_COVID19\_CLE}$ 



Per-topic difference from median P@5 for all Round 1 runs