Word Hyphenation Correction and Query Expansion using Wikipedia for Book Retrieval

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INEX'10 Book Track, December 13-15

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 - Gutenberg
 - Google Books

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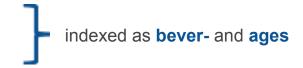
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 - Pseudo-relevance feedback
 - Wikipedia for selecting expansion words
 - One Wikipedia page for each query (Koolen et al., WSDM'09)

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- OMRF model (Metzler and Croft, SIGIR'05)
 Deveaud et al. LIA, University of Avignon

```
L1: Too often people are prepared to accept that alcoolic w1[bever]-L2:w2[ages] are thoroughly bad: a portal to doom, debauchery...
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• For each couple of lines (L1, L2) of each book

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 - Language modeling (LM), Dirichlet smoothing
 - 100 books retrieved for each query

Table 1. Book retrieval results on both initial and corrected INEX 2009 Book Track corpus in terms of Mean Average Precision (MAP) and precision at 10 (P@10).

Model	Uncorrec	cted data	Corrected data			
Wilde	MAP	P@10	MAP	P@10		
$\overline{\mathrm{LM},\mu=2500}$	0.302	0.486	0.304	0.507		
LM, $\mu = 1000$	0.299	0.493	0.302	0.507		
LM, $\mu = 0$	0.244	0.443	0.243	0.450		

• Two simple models

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- Bag of words
 - Language modeling approach
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- baseline_2 : content of <fact> fields





Queries Balle of Gellysburg New He herlands ...

http://en.wikipedia.org/wiki/Battle_of_Gatysburg

Battle of Gettysburg

The Bettle of Gettysburg (locally/ 'gdisharg' (listen), with an sesound), fought July 1-3, 1863, in and around the town of Gettysburg. Pennsylvania, was the battle with the largest number of casualties in the American Gvil War [6] and is often described as the war's turning point.[7] Union Maj. Gen. George Gordon Meade's Army of the Potomac defeated attacks by Confederate Gen. Robert E. Lee's Army of Northern Wriginia, ending Lee's invasion of the North. http://en.wikipedia.org/wiki/New_Netherlands

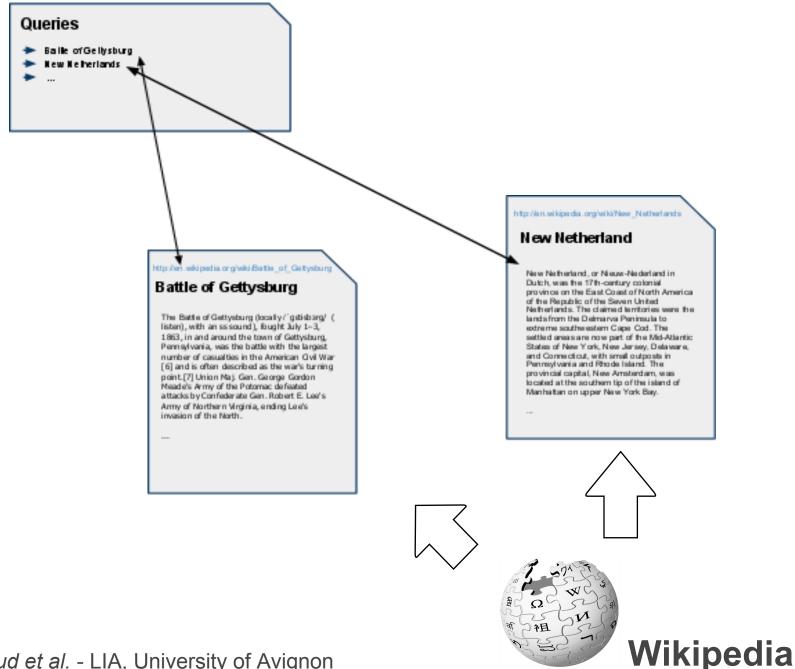
New Netherland

New Netherland, or Nieuw-Nederland in Dutch, was the 17th-cartury colonial province on the East Coast of North America of the Republic of the Seven United Netherlands. The claimed territories were the lands from the Definer we Peninsula to extreme southwestern Capa Cod. The settled areas are now part of the Mid-Allantic States of New York, New Jersey, Delavere, and Comedicat, with arreal outposts in Pennsylvaria and Rhode Island. The provincial capital, New Amsterdam, was located at the southern tip of the Island of Manhaitan on upper New York Bay.









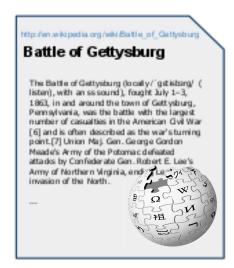
Koolen et al. (WSDM'09):

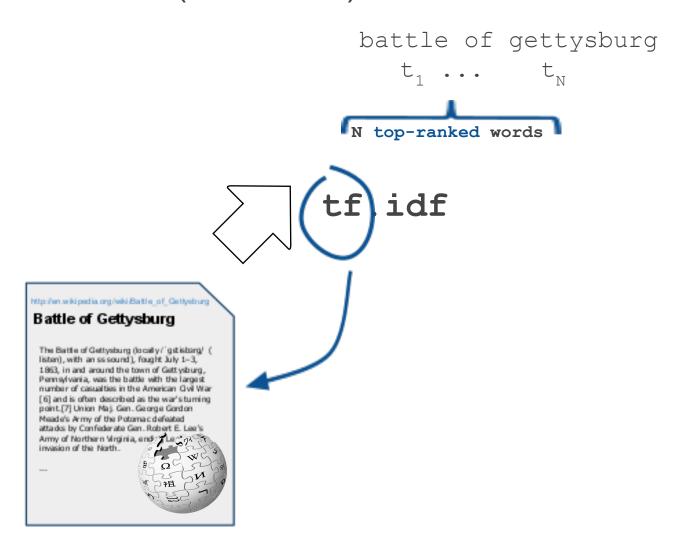
battle of gettysburg

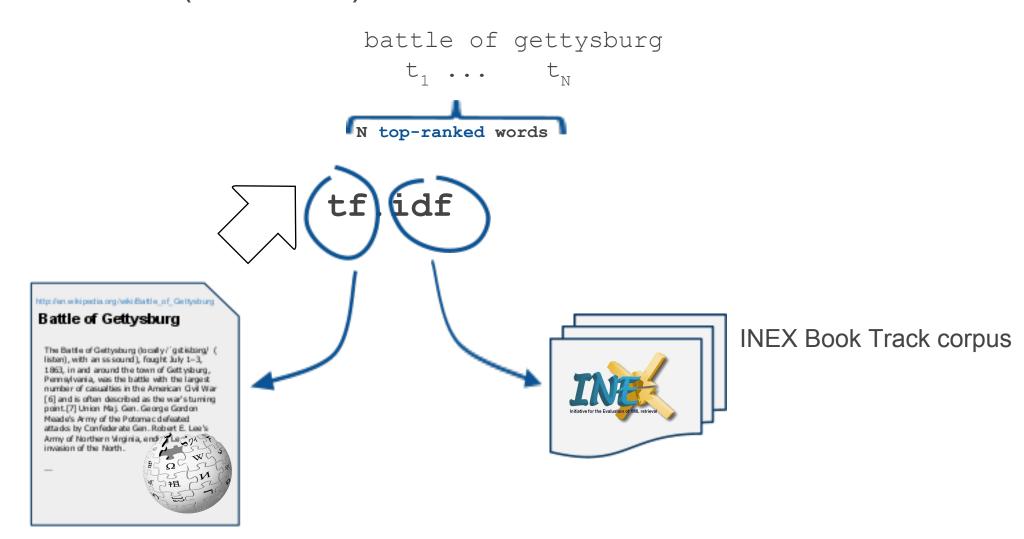
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battle of gettysburg t_1 \dots t_N
```

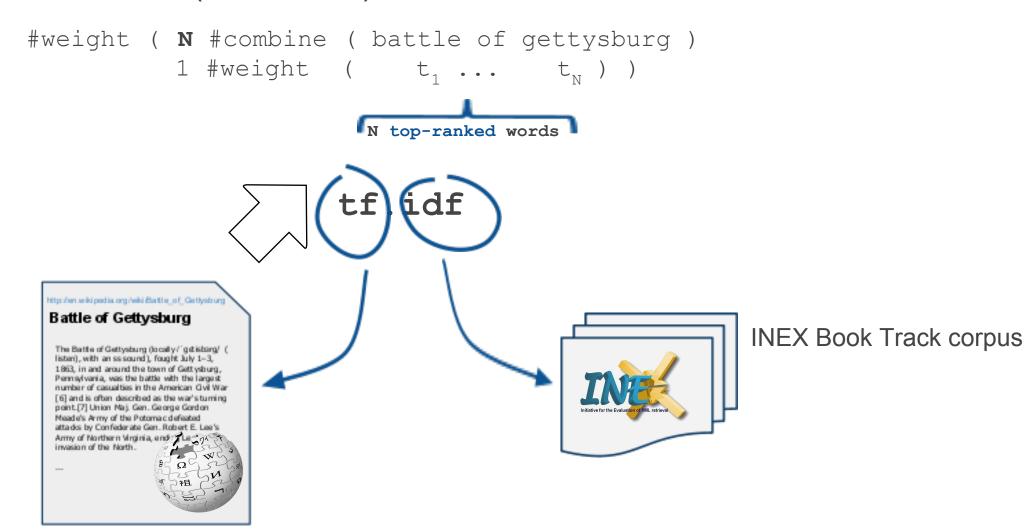


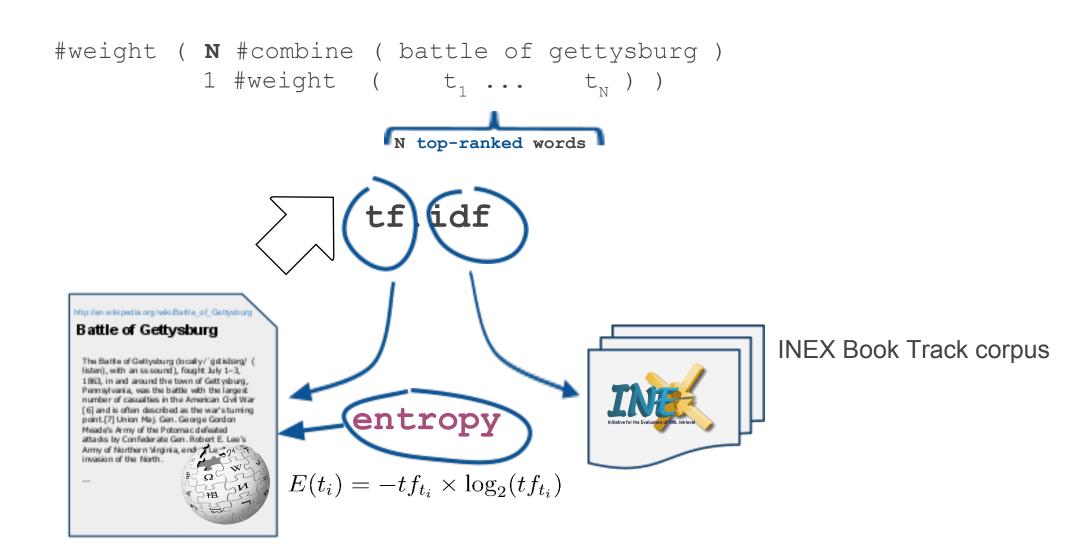


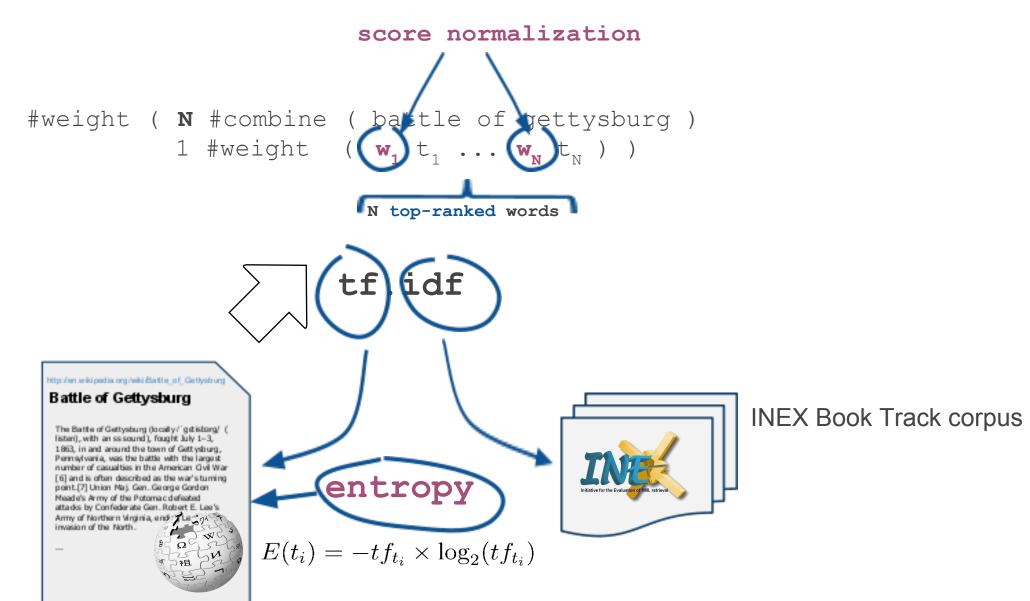


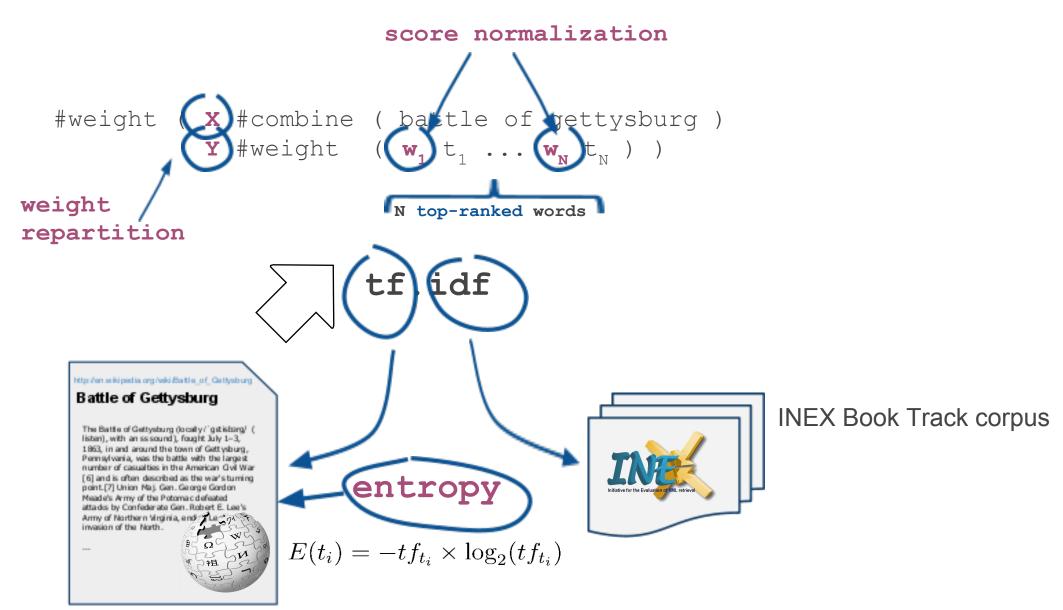


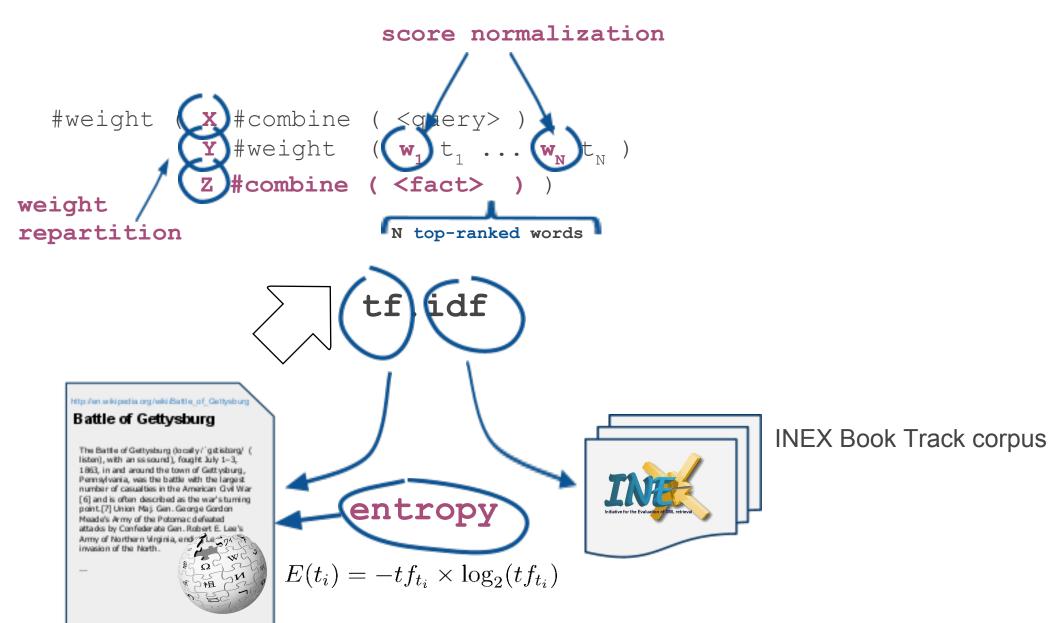




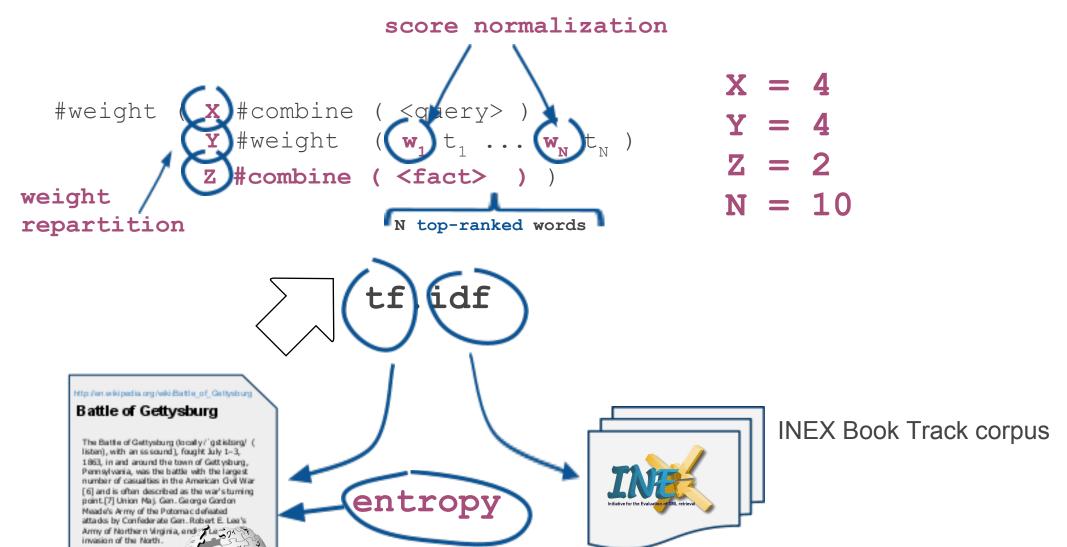








 $E(t_i) = -tf_{t_i} \times \log_2(tf_{t_i})$



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- We only use the <fact> fields

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Thank you for your attention

Appendix

Table 2. Query expansion performance at the top N words ranked by *tf.idf* or *entropy*, on the corrected and the uncorrected corpus, with (X:Y) weights († : t.test < 0.05; ‡ : t.test < 0.01).

Method	corpus	N = 5		N = 10		N = 20		N = 50	
		MAP	P@10	MAP	P@10	MAP	P@10	MAP	P@10
entropy (1:3)	Cor.	0.301	0.489	0.346	0.564	0.330	0.529	0.353	0.564
entropy (2:2)	Cor.	0.327	0.557	0.348	0.564	0.361	0.592^\dagger	0.363	0.593^{\ddagger}
entropy (2:2)	UnC.	0.323	0.542	0.346	0.579	0.358	0.586^\dagger	0.361	0.593^{\ddagger}
entropy (3:1)	Cor.	0.330	0.564	0.342^\dagger	0.564	0.349	0.564	0.347	0.557
tf.idf (1:3)	Cor.	0.245	0.479	0.249	0.450	0.257	0.464	0.246	0.486
tf.idf (2:2)	Cor.	0.277	0.486	0.290	0.521	0.289	0.140	0.295	0.514
tf.idf (3:1)	Cor.	0.310	0.536	0.311	0.543	0.317	0.557	0.314	0.536
tf.idf (3:1)	UnC.	0.307	0.529	0.311	0.543	0.314	0.557	0.313	0.529
Koolen et al.	Cor.	0.308	0.550	0.321	0.536	0.301	0.521	0.306	0.507
Koolen $et al.$	UnC.	0.308	0.550	0.315	0.521	0.300	0.514	0.304	0.500