# Experiment2: Ranked retrieval model

#### 任务:

- -在Experiment1的基础上实现最基本的Ranked retrieval model
  - Input: a query (like Ron Weasley birthday)
  - Output: Return the top K (e.g., K = 100) relevant tweets.
- Use SMART notation: Inc.ltc
  - Document: logarithmic tf (l as first character), no idf and cosine normalization
  - Query: logarithmic tf (I in leftmost column), idf (t in second column), no normalization
- 改进Inverted index
  - 在Dictionary中存储每个term的DF
  - 在posting list中存储term在每个doc中的TF with pairs (docID, tf)
- 选做
  - 支持所有的SMART Notations

Deadline: 2021.11.9

# Computing cosine scores

```
CosineScore(q)
  1 float Scores[N] = 0
  2 float Length[N]
  3 for each query term t
    do calculate w_{t,q} and fetch postings list for t
         for each pair(d, tf<sub>t,d</sub>) in postings list
         do Scores[d] + = w_{t,d} \times w_{t,a}
    Read the array Length
     for each d
     do Scores[d] = Scores[d]/Length[d]
     return Top K components of Scores[]
```

# tf-idf weighting has many variants

Term frequency		Docum	ent frequency	Normalization			
n (natural)	$tf_{t,d}$	n (no)	1	n (none)	1		
I (logarithm)	$1 + \log(tf_{t,d})$	t (idf)	$\log \frac{N}{\mathrm{df}_t}$	c (cosine)	$\frac{1}{\sqrt{w_1^2 + w_2^2 + \ldots + w_M^2}}$		
a (augmented)	$0.5 + \frac{0.5 \times tf_{t,d}}{max_t(tf_{t,d})}$	p (prob idf)	$\max\{0,\log \frac{N-\mathrm{df}_t}{\mathrm{df}_t}\}$	u (pivoted unique)	1/u		
b (boolean)	$egin{cases} 1 &  ext{if } \operatorname{tf}_{t,d} > 0 \ 0 &  ext{otherwise} \end{cases}$			b (byte size)	$1/\mathit{CharLength}^{lpha}$ , $lpha < 1$		
L (log ave)	$\frac{1 + \log(\operatorname{tf}_{t,d})}{1 + \log(\operatorname{ave}_{t \in d}(\operatorname{tf}_{t,d}))}$						

**SMART notation** for tf-idf variants.

Here *CharLength* is the number of characters in the document.

### Weighting may differ in queries vs documents

- Many search engines allow for different weightings for queries vs. documents
- SMART Notation: denotes the combination in use in an engine, with the notation ddd.qqq, using the acronyms from the previous table
- A very standard weighting scheme is: Inc.ltn
- Document: logarithmic tf (l as first character), no idf and cosine normalization

  A bad idea?
- Query: logarithmic tf (l in leftmost column), idf (t in second column), no normalization ...

## tf-idf example: Inc.ltc

Query: "best car insurance". Document: "car insurance auto insurance".

word	query				document				product	
	tf-raw	tf-wght	df	idf	weight	tf-raw	tf-wght	weight	n'lized	
auto	0	0	5000	2.3	0	1	1	1	0.52	0
best	1	1	50000	1.3	1.3	0	0	0	0	0
car	1	1	10000	2.0	2.0	1	1	1	0.52	1.04
insurance	1	1	1000	3.0	3.0	2	1.3	1.3	0.68	2.04

Key to columns: tf-raw: raw (unweighted) term frequency, tf-wght: logarithmically weighted term frequency, df: document frequency, idf: inverse document frequency, weight: the final weight of the term in the query or document, n'lized: document weights after cosine normalization, product: the product of final query weight and final document weight

$$\sqrt{1^2 + 0^2 + 1^2 + 1.3^2} \approx 1.92$$
  
 $1/1.92 \approx 0.52$   
 $1.3/1.92 \approx 0.68$ 

Final similarity score between query and document:  $\sum_{i} w_{qi} \cdot w_{di} = 0 + 0 + 1.04 + 2.04 = 3.08$