Project II

- Problem
 - Given m documents, compute the term-term relevance using MapReduce algorithm and Spark implementation
 - Input: A text file, each line represents a document
 - Output: A list of term-term pairs sorted by their similarity descending
 - t1 t2 s1 t3 t4 s2
- Sub-problems:
 - Compute Term Frequency Inverse Document Frequency (TF-IDF) for each term
 - Output: mxn matrix (m: #documents, n: #terms)
 - Computer and sort term-term relevance between a query term and all terms associated with the TF-IDF matrix
 - Input: a query term t
 - Output: term-term relevance between the query term and those terms in the tfidf matrix sorted by the relevance score (descending)

TF-IDF

- Term Frequency Inverse Document Frequency
 - Relevant to text processing
 - Common web analysis algorithm

The Algorithm, Formally

$$ext{tf}_{\mathbf{i}} = \frac{n_i}{\sum_k n_k}$$
 $ext{idf}_{\mathbf{i}} = \log \frac{|D|}{|\{d: t_i \in d\}|}$ $ext{tfidf} = ext{tf} \cdot ext{idf}$

- | D | : total number of documents in the corpus
- $|\{d: t_i \in d\}|$ humber of documents where the term t_i appears (that is $n_i \neq 0$.

Semantic Similarity

$$similarity = \cos(\theta) = \frac{A \cdot B}{\|A\| \|B\|} = \frac{\sum_{i=1}^{n} A_i B_i}{\sqrt{\sum_{i=1}^{n} A_i^2} \sqrt{\sum_{i=1}^{n} B_i^2}}$$

Example

| | | D1 | D2 | D3 |
|--|---------|----|----|----|
| D1: I like data science D2: I hate data D3: want A | 1 | 1 | 1 | 0 |
| | like | 1 | 0 | 0 |
| | data | 1 | 1 | 0 |
| | science | 1 | 0 | 0 |
| | | | | |
| | hate | 0 | 1 | 0 |
| | want | 0 | 0 | 1 |
| | Α | 0 | 0 | 1 |

Example

| | tf | | | |
|-------------------------------|---------|-----|-----|-----|
| | | D1 | D2 | D3 |
| | 1 | 1/4 | 1/3 | 0 |
| D1: I like data science | like | 1/4 | 0 | 0 |
| D2: I hate data D3: want A | data | 1/4 | 1/3 | 0 |
| 20 | science | 1/4 | 0 | 0 |
| | hate | 0 | 1/3 | 0 |
| | want | 0 | 0 | 1/2 |
| | Α | 0 | 0 | 1/2 |

Example

idf

| D1: I like data science | |
|-------------------------|--|
| D2: I hate data | |
| D2 | |

D3: want A

| | D1 | D2 | D3 |
|---------|----------|----------|----------|
| 1 | log(3/2) | log(3/2) | log(3/2) |
| like | log(3/1) | log(3/1) | log(3/1) |
| data | log(3/2) | log(3/2) | log(3/2) |
| science | log(3/1) | log(3/1) | log(3/1) |
| hate | log(3/1) | log(3/1) | log(3/1) |
| want | log(3/1) | log(3/1) | log(3/1) |
| Α | log(3/1) | log(3/1) | log(3/1) |

Example

tf*idf

| D1: I like data science |
|-------------------------|
| D2: I hate data |

D3: want A

| | D1 | D2 | D3 |
|---------|-------|-------|-------|
| I | 0.044 | 0.059 | 0.0 |
| like | 0.119 | 0.0 | 0.0 |
| data | 0.044 | 0.059 | 0.0 |
| science | 0.119 | 0.0 | 0.0 |
| hate | 0.0 | 0.159 | 0.0 |
| want | 0.0 | 0.0 | 0.238 |
| Α | 0.0 | 0.0 | 0.238 |

Example

I (0.044, 0.059, 0.0) A (0.0, 0.0, 0.238)

Similarity (I, A) = (0.044*0.0+0.059*0.0+0.0*0.238)

sqrt(0.044*0.044+0.059*0.059+0.0*0.0) x sqrt(0.0*0.0+0.0*0.0+0.238*0.238)