# **Asignment 2: Document Similarity and Hashing**

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#### **Overview**

In this assignment you will explore the use of k-grams, Jaccard distance, min hashing, and LSH in the context of document similarity.

You will use four text documents for this assignment:

- http://www.cs.utah.edu/~jeffp/teaching/cs5140/A2/D1.txt
- http://www.cs.utah.edu/~jeffp/teaching/cs5140/A2/D2.txt
- http://www.cs.utah.edu/~jeffp/teaching/cs5140/A2/D3.txt
- http://www.cs.utah.edu/~jeffp/teaching/cs5140/A2/D4.txt

As usual, it is highly recommended that you use LaTeX for this assignment. If you do not, you may lose points if your assignment is difficult to read or hard to follow. Find a sample form in this directory: http://www.cs.utah.edu/~jeffp/teaching/latex/

## 1 Creating *k*-Grams (40 points)

You will construct several types of k-grams for all documents. All documents only have at most 27 characters: all lower case letters and space. Yes, the space counts as a character in character k-grams.

- [G1] Construct 2-grams based on characters, for all documents.
- [G2] Construct 3-grams based on characters, for all documents.
- [G3] Construct 2-grams based on words, for all documents.

Remember, that you should only store each k-gram once, duplicates are ignored.

**A: (20 points)** How many distinct k-grams are there for each document with each type of k-gram? You should report  $4 \times 3 = 12$  different numbers.

| File   | $k_2$ -Character | $k_3$ -Character | $k_2$ -Word |
|--------|------------------|------------------|-------------|
| D1.txt | 331              | 1299             | 521         |
| D2.txt | 361              | 1516             | 632         |
| D3.txt | 354              | 1543             | 841         |
| D4.txt | 298              | 1025             | 413         |

**B: (20 points)** Compute the Jaccard similarity between all pairs of documents for each type of k-gram. You should report  $3 \times 6 = 18$  different numbers.

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Table 1:  $k_2$ -Character Jacard Similarities

|        | D1 LL  | D0 ++  | D2 ++  | D1 ++  |
|--------|--------|--------|--------|--------|
|        | D1.txt | D2.txt | D3.txt | D4.txt |
| D1.txt | 1.000  | 0.845  | 0.770  | 0.705  |
| D2.txt | 0.845  | 1.000  | 0.761  | 0.707  |
| D3.txt | 0.770  | 0.761  | 1.000  | 0.720  |
| D4.txt | 0.705  | 0.707  | 0.720  | 1.000  |

Table 2:  $k_3$ -Character Jacard Similarities

|        | D1.txt | D2.txt | D3.txt | D4.txt |
|--------|--------|--------|--------|--------|
| D1.txt | 1.000  | 0.639  | 0.460  | 0.327  |
| D2.txt | 0.639  | 1.000  | 0.440  | 0.312  |
| D3.txt | 0.460  | 0.440  | 1.000  | 0.362  |
| D4.txt | 0.327  | 0.312  | 0.362  | 1.000  |

Table 3:  $k_2$ -Word Jacard Similarities

|        | D1.txt | D2.txt | D3.txt | D4.txt |
|--------|--------|--------|--------|--------|
| D1.txt | 1.000  | 0.257  | 0.033  | 0.005  |
| D2.txt | 0.257  | 1.000  | 0.025  | 0.006  |
| D3.txt | 0.033  | 0.025  | 1.000  | 0.012  |
| D4.txt | 0.005  | 0.006  | 0.012  | 1.000  |

### 2 Min Hashing (30 points)

We will consider a hash family  $\mathcal{H}$  so that any hash function  $h \in \mathcal{H}$  maps from  $h : \{k\text{-grams}\} \to [m]$  for m large enough (To be extra cautious, I suggest over  $m \ge 10{,}000$ ).

**A:** (25 points) Using grams G2, build a min-hash signature for document D1 and D2 using  $t = \{20, 60, 150, 300, 600\}$  hash functions. For each value of t report the approximate Jaccard similarity between the pair of documents D1 and D2, estimating the Jaccard similarity:

$$\hat{\mathsf{JS}}_t(a,b) = \frac{1}{t} \sum_{i=1}^t \begin{cases} 1 & \text{if } a_i = b_i \\ 0 & \text{if } a_i \neq b_i. \end{cases}$$

You should report 5 numbers.

| $\overline{t}$ | $J_S$ |
|----------------|-------|
| 20             | 0.750 |
| 60             | 0.667 |
| 150            | 0.633 |
| 300            | 0.640 |
| 600            | 0.612 |
|                |       |

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**B:** (5 point) What seems to be a good value for t? You may run more experiments. Justify your answer in terms of both accuracy and time.

From the direct calculation, we know that the Jacard Estimate between documents 1 and 2 using 3-grams is 0.639. Looking at the estimated Jacard Similarities above t=150 seems to be the best bet. To check, the value of t was increased up to 1900 to check the variation of the Jacard Similarity Estimate. As the table below shows, there was limited variation in the values.

| t    | $J_S$ |
|------|-------|
| 500  | 0.642 |
| 700  | 0.639 |
| 900  | 0.651 |
| 1100 | 0.638 |
| 1300 | 0.632 |
| 1500 | 0.629 |
| 1700 | 0.668 |
| 1900 | 0.635 |

#### 3 LSH (30 points)

Consider computing an LSH using t=160 hash functions. We want to find all documents pairs which have Jaccard similarity above  $\tau=0.4$ .

**A:** (8 points) Use the trick mentioned in class and the notes to estimate the best values of hash functions b within each of r bands to provide the S-curve

$$f(s) = 1 - (1 - s^b)^r$$

with good separation at  $\tau$ . Report these values.

The function  $f(s)=1-\left(1-s^b\right)^r$  is steepest at the point of inflection  $\tau=(1/r)^{1/b}$ . If we want to use s as our cut where  $\tau=s=\alpha=1-\beta$  we can substitute and get  $\tau=(b/t)^{1/b}$  which can be approximated as  $b\approx -\log_{\tau}(t)$ . Therefore, with t=160 and  $\tau=0.4$ , we have  $b=5.539=-\log_{\tau}(t)=-\frac{\log_{10}(t)}{\log_{10}(\tau)}$ . From b=t/r we have r=t/b=160/5.539=28.886. Therefore, the S-curve is given by

$$f(s) = 1 - \left(1 - s^{5.539}\right)^{28.886}$$

**B:** (24 points) Using your choice of r and b and  $f(\cdot)$ , what is the probability of each pair of the four documents (using [G2]) for being estimated to having similarity greater that  $\tau$ ? Report 6 numbers.

The probability of documents being estimated for using 3-grams have a similarity greater than  $\tau=0.4$  is given by the above equation, where s is the Jacard Similarity between the two documents.

| $\overline{f(D_i, D_j)}$             | Probability/100 |
|--------------------------------------|-----------------|
| f(D1.txt, D2.txt)                    | 0.999           |
| $f(\mathtt{D1.txt},\mathtt{D3.txt})$ | 0.334           |
| $f(\mathtt{D1.txt},\mathtt{D4.txt})$ | 0.022           |
| f(D2.txt, D3.txt)                    | 0.141           |
| f(D2.txt, D4.txt)                    | 0.036           |
| f(D3.txt, D4.txt)                    | 0.040           |

### 4 Bonus (3 points)

Describe a scheme like Min-Hashing for the  $Andberg\ Similarity$ , defined  $Andb(A,B) = \frac{|A\cap B|}{|A\cup B|+|A\triangle B|}$ . So given two sets A and B and family of hash functions, then  $\Pr_{h\in\mathcal{H}}[h(A)=h(B)]=Andb(A,B)$ . Note the only randomness is in the choice of hash function h from the set  $\mathcal{H}$ , and  $h\in\mathcal{H}$  represents the process of choosing a hash function (randomly) from  $\mathcal{H}$ . The point of this question is to design this process, and show that it has the required property.

Or show that such a process cannot be done.

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