This is a closed book and closed notes quiz. No electronic aids or cheat sheets are allowed.

There are 2 pages, 3 parts of questions, and 20 total points in this quiz.

The questions are printed on the back of this paper!

Please carefully read the instructions and questions before you answer them.

Please pay special attention on your handwriting; if the answers are not recognizable by the instructor, the grading might be inaccurate (NO argument about this after the grading is done).

Try to keep your answers as concise as possible; grading is not by keyword matching.

Total /20
1 True/False Questions (3pts×2)

For the statement you believe it is False, please give your brief explanation of it (you do not need to explain anything when you believe it is True). Note the credit can only be granted if your explanation is correct.

1. Given a well-tuned unigram language model $p(w|\theta)$ estimated based on all the text books about the topic of “text mining”, we can safely conclude that $p(\text{“text mining”}|\theta) > p(\text{“mining text”}|\theta)$.
   **False, and Explain:** they should be equal, since in a unigram language model we cannot model the order of words.

2. Given a unigram language model and a bigram language model estimated on the same text collection **without smoothing**, perplexity of the unigram language model will be much larger than that of the bigram language model on this **same training corpus**.
   **True**

2 Multi-choice Questions (4pts×2)

1. Good “basic concepts” in a vector space model should be: (a) (c)
   (a) orthogonal to each other; (b) based on linguistic study;
   (c) able to automatically compute the weights in each document;
   (d) understandable by human.

2. Zipf’s law tells us: (b) (d)
   (a) head words take major portion in English vocabulary;
   (b) in a given French corpus, if the most frequent word’s frequency is 1, then the second frequent word’s frequency is around 0.5;
   (c) comparing to tail words, removing head words helps more to reduce the storage of documents represented by a vector space model when using a dense matrix data structure;
   (d) smoothing is necessary.

3 Short Questions (6 pts)

1. Let D be a document in a text collection. Suppose we add a copy of D to the collection. How would this affect the IDF values of all the words in the collection? Why?

   The new IDF is,
   $$IDF(w) = \begin{cases} 
   1 + \log\left(\frac{N+1}{DF(w)+1}\right) & \text{if } w \in D \\
   1 + \log\left(\frac{N+1}{DF(w)}\right) & \text{otherwise}
   \end{cases}$$

   where $N$ the original collection size, $DF(w)$ is the original document frequency of word $w$.

   Therefore, when $w$ occurs in $D$, its new IDF decreases (since $N \geq DF(w)$); otherwise, its new IDF increases.