## The Unigram Term Frequency Distribution

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The unigram posits that each word occurrence in a document is independent of all other word occurrences. I.e. we can think of the document generation process as a sequence of dice rolls, where there is a fixed probability of occurrence associated with each word. The chance observing a given document is simply the product of the word probabilities. To calculate the chance of observing a given set of word frequencies, we must count all the possible orderings that achieve that set of frequencies. Let  $\{x_i\}$  be the observed frequencies for a set of words. There are  $\frac{(\sum_i x_i)!}{\prod_i x_i!}$  word arrangements that achieve that set of word frequencies. Hence, the likelihood of generating a document with that set of frequencies is

$$P\left(\left\{x_i\right\}\Big|\sum_i x_i = l\right) = \frac{l!}{\prod_i x_i!} \prod_i \left(\frac{w_i}{\sum_i w_i}\right)^{x_i}.$$
 (1)

Note that the unigram is conditional on document length; the above gives the conditional likelihood of generating a particular set of frequencies given that their sum is l. The  $\{w_i\}$  are the unnormalized word occurrence probabilities.

To find maximum likelihood weights for a document set, it is easiest to consider minimization of negative log-likelihood. Let  $x_{ij}$  represent the number of times word j occurs in the  $i^{\text{th}}$  document; let  $l_i = \sum_j x_{ij}$ . Then, the negative log-likelihood is

$$J = \sum_{i,j} \log(x_{ij}!) - \sum_{i} \log(l_i!) + \sum_{i} l_i \log\left(\sum_{j} w_j\right) - \sum_{ij} x_{ij} \log w_j.$$
(2)

To minimize this quantity, we find settings which give us a zero gradient with respect to the weights. The partial derivative with respect to a weight is

$$\frac{\partial J}{\partial w_j} = \sum_i \frac{l_i}{\sum_j w_j} - \sum_i \frac{x_{ij}}{w_j}.$$
(3)

Note that the setting  $w_j = \sum_i x_{ij}$  gives us a zero gradient.