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Connected to miniconda3 (Python 3.12.2)
In [ ]: import nltk
        import random
        def generate_ngram(n, corpus):
            # ngram is a dictionary that stores the list of words and the count of the next_word occuring after certain list of words.
            ngram = \{\}
            for i in range(len(corpus) - n + 1):
                # words list framed as a tuple, as it should be immutable
               words_list = tuple(corpus[i : i + n - 1])
               next word = corpus[i + n - 1]
               if words_list not in ngram:
                   ngram[words_list] = {}
               if next_word not in ngram[words_list]:
                    ngram[words_list][next_word] = 0
               # if words_list and next_word are both in the dictionary, then increase the count
               ngram[words_list][next_word] += 1
            return ngram
        def predict_next_word(sentence, n, corpus, randomize=False, alpha=0.4):
            ngram = generate_ngram(n, corpus)
            highest word = None
            # n should not be larger than length of sentence. The judgement between n and length of sentence is done in the finish_sentence function.
            context = tuple(sentence[-(n - 1) :])
            # first determine if context exists in the ngram, and if there is any next_word existing for the context. If so, no need to trigger backoff
            if context in ngram and ngram[context]:
               possible_words = ngram[context]
               # possible_words is a dictionary, key is the next_word, and value is the count of next_word
               total_count = sum(possible_words.values())
                # create another new dictionary that summarizes probability of the next_word probability
               possible_words_probs = {}
               for word, num in possible_words.items():
                   prob = num / total_count
                    possible_words_probs[word] = prob
                    # when the probability is equal, choose the word that is first alphabetically
                    if randomize == False:
                       highest_word = max(
                           possible_words_probs,
                           key=lambda w: (possible_words_probs[w], -ord(w[0])),
                    # when randomize = True, we select randomly from the distribution, with weights being its respective probability
                    else:
                       highest word = random.choices(
                           list(possible words probs.keys()),
                           weights=possible_words_probs.values(),
                       ) [0]
            # if context did not exist in the ngram, trigger backoff, calculate stupid backoff all the way back to n=1 with the penalty alpha powered accordingly
            else:
                possible words probs = {}
               for i in range(n - 1, 0, -1):
                    context = tuple(sentence[-(i - 1) :])
                   new_gram = generate_ngram(i, corpus)
                    if context in new_gram and new_gram[context]:
                        possible_words = new_gram[context]
                       total_count = sum(possible_words.values())
                       for word, num in possible_words.items():
                           # if the words already appeared in the dictionary, no need to calculate the probability again
                           if word not in possible words probs:
                               prob = (num / total_count) * (alpha ** (n - i))
                               possible_words_probs[word] = prob
               # after iterating through all the (n-1)gram to unigram, find the highest probability when randomize = false
               if randomize == False:
                    highest_word = max(
                        possible_words_probs,
                        key=lambda w: (possible_words_probs[w], -ord(w[0])),
               else:
                   highest_word = random.choices(
                       list(possible_words_probs.keys()), weights=possible_words_probs.values()
                    [0]
            return highest_word
        def finish_sentence(sentence, n, corpus, randomize=False, alpha=0.4):
            "returns an extended sentence until"
            "the first ., ?, or ! is found OR until it has 10 total tokens"
            stop_tokens = {".", "?", "!"}
            # If n is larger than (length of sentence + 1), then we need to first cut n to the (length of sentence + 1). Use the whole sentence as context to predict next word.
            if n > (len(sentence) + 1):
               temp_n = len(sentence) + 1
               while len(sentence) < n - 1:</pre>
                    next_word = predict_next_word(sentence, temp_n, corpus, randomize, alpha)
                   if not next_word:
                       break
                    sentence.append(next_word)
                    if next_word in stop_tokens:
                       break
            while len(sentence) < 10:</pre>
               next word = predict next word(sentence, n, corpus, randomize, alpha)
               if not next word:
                    break
                sentence.append(next_word)
               if next_word in stop_tokens:
                    break
            return sentence
In [ ]: print("-
                    -----Testing-----
        corpus = tuple(
            nltk.word_tokenize(nltk.corpus.gutenberg.raw("austen-sense.txt").lower())
        print("-----Application #1 (Deterministic) -----")
        n = 4
        st = ["he", "did"]
       print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #2 (Deterministic) -----")
        n = 3
        st = ["he", "did"]
       print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #3 (Stochastic) -----")
        n = 3
        st = ["he", "did"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=True, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #4 (Deterministic) -----")
        n = 3
        st = ["it", "was"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #5 (Deterministic)-----")
        n = 4
        st = ["it", "was"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #6 (Stochastic)-----")
        n = 4
        st = ["it", "was"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=True, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #7 (Deterministic)-----")
        n = 4
        st = ["they", "would"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
                     -----Application #8 (Deterministic)----
        n = 3
        st = ["they", "would"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #9 (Stochastic)-----")
        n = 3
        st = ["they", "would"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=True, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #10 (Deterministic)----")
        n = 5
        st = ["they", "would", "ask", "him"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #11 (Deterministic)-----")
       n = 3
        st = ["they", "would", "ask", "him"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=False, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
        print("-----Application #12 (Stochastic)----")
        n = 3
        st = ["they", "would", "ask", "him"]
        print("Input:{}, N={}".format(" ".join(st), n))
        output = finish_sentence(st, n, corpus, randomize=True, alpha=0.4)
        print("Output:{}".format(" ".join(output)))
       -----Testing-----
       -----Application #1 (Deterministic) -----
       Input:he did, N=4
       Output: he did not feel the continuance of his existence secure
       -----Application #2 (Deterministic) -----
       Input:he did, N=3
       Output:he did not know what you are very much to
       -----Application #3 (Stochastic) -----
       Input:he did, N=3
       Output:he did nothing but only civility ?
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-----Application #4 (Deterministic) ------Input:it was, N=3 Output: it was not in the world , and the two -----Application #5 (Deterministic)-----Input: it was, N=4 Output: it was not a thing to be thought of; -----Application #6 (Stochastic)-----Input:it was, N=4 Output: it was a very great relief to you , in -----Application #7 (Deterministic)-----Input: they would, N=4 Output: they would all dine with lady middleton . -----Application #8 (Deterministic)-----Input: they would, N=3 Output: they would all be made for him , and the -----Application #9 (Stochastic)-----Input: they would, N=3 Output: they would soon , and then , '' cried sir -----Application #10 (Deterministic)-----Input: they would ask him, N=5 Output: they would ask him if there was any news . -----Application #11 (Deterministic)-----Input: they would ask him, N=3 Output: they would ask him if he had been in the -----Application #12 (Stochastic)-----Input: they would ask him, N=3 Output: they would ask him questions which his society bestowed on