pal solution

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[]: import re # preprocessing
     import math # PMI calculation
     import numpy as np # 2d arrays for clustering functions
     import codecs # file handle
     import argparse # parse command line arguments
     import Clustering_functions as cl # import clustering functions
     #fancy preprocessing
     import nltk.data, nltk.tag
     from collections import Counter
     from nltk.stem import WordNetLemmatizer
     from nltk.corpus import wordnet
     from nltk.tag.perceptron import PerceptronTagger
     # Class for calculating cosine similarity between words
     class SimTester():
         # Initialize with the needed variables
         def __init__(self):
             self.sum_all = 0 # used in PMI formula
             self.path_b = '' # path to the file B.txt
             self.path_t = '' # path to the file T.txt
             self.path_corpus = '' # path to the corpus file
             self.words = [] # list of all words from the raw text
             self.lemmatized_words = [] # lemmatized words (for fancy preprocessing)
             self.b = [] # list of B words
             self.t = [] # list of T words
             self.matrix = {} # feature matrix (raw frequencies)
             self.pmi_matrix = {} # feature matrix (with PMI weights)
             self.cos_sim_matrix = {} # 2d array cosine similarity
             self.cos_dist_matrix = {} # 2d array distance
             self.cos_sim = np.empty(shape=(1, 1)) # numpy 2d array for cosine_
      \rightarrowsimilarity
             self.cos_dist = np.empty(shape=(1, 1)) # numpy 2d array for distance
         # Lowercase and delete all punctuation
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def preprocess(self, line):
       new_line = re.sub(u'[^\w]+', '', line, flags=re.UNICODE) # remove_
\rightarrow punctuation
       new_line = new_line.lower() # lowercase
       self.words += new_line.split() # separate words by white spaces
   # Open file with a corpus, preprocess it
  def read_corpus(self):
       file = codecs.open(self.path_corpus, 'r', 'utf-8')
       for line in file:
           self.preprocess(line)
   # Read B and T set from text files
  def create_set(self, path_set, list_set):
       file = codecs.open(path_set, 'r', 'utf-8')
       for line in file:
           list_set.append(line.strip())
   # Initialize matrix
  def init_matrix(self, matrix, arr_length, init_value):
       for target in self.t:
           matrix[target] = self.init_vector(arr_length, init_value)
   # Initialize vector
  def init_vector(self, arr_length, init_value):
       vector = []
       i = 0
       while i <= arr_length-1:</pre>
           vector.append(init_value)
           i += 1
       return vector
   # Context windows size 5
  def find_windows(self, fancy_preprocessing):
       if not fancy_preprocessing:
           word_array = self.words
       else:
           self.improve()
           word_array = self.lemmatized_words
       for target in self.t:
           for i in range(len(word_array)):
               if target == word_array[i]:
                   # First word
                   if i == 0:
                       context = [word_array[i+1], word_array[i+2]]
                   # Second word
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elif i == 1:
                       context = [word_array[i-1], word_array[i+1],__
→word_array[i+2]]
                   # Penultimate word
                   elif i == len(word_array)-2:
                       context = [word_array[i-2], word_array[i-1],__
→word_array[i+1]]
                   # Last word
                   elif i == len(word_array)-1:
                       context = [word_array[i-2], word_array[i-1]]
                   # Normal case
                   else:
                       context = [word_array[i-2], word_array[i-1],__
→word_array[i+1], word_array[i+2]]
                   self.raw_freq(target, context)
       self.sum_values()
   # Raw frequencies
   def raw_freq(self, target, context):
       b_indices = []
       for word in context:
           if word in self.b:
               b index = self.b.index(word)
               b_indices.append(b_index)
       for key in self.matrix:
           if key == target:
               for b_index in b_indices:
                   self.matrix[key][b_index] += 1
   # Sum all values in the matrix
   def sum_values(self):
       for key in self.matrix:
           for value in self.matrix[key]:
               self.sum_all += value
   # Count PMI
   def count pmi(self):
       self.init_matrix(self.pmi_matrix, len(self.b), 0)
       for c in range(len(self.b)):
           # divisor: 2nd part
           p_j = 0
           for key in self.matrix:
               p_j += self.matrix[key][c]
           p_j = p_j / float(self.sum_all)
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for w in self.t:
               # dividend
               f_ij = self.matrix[w][c]
               p_ij = f_ij / float(self.sum_all)
               # divisor: 1st part
               p_i = 0
               for freq in self.matrix[w]:
                   p_i += freq
               p_i = p_i / float(self.sum_all)
               pmi_value = self.pmi(p_ij, f_ij, p_i, p_j)
               self.pmi_matrix[w][c] = round(pmi_value, 2)
   # PMI formula
   def pmi(self, p_ij, f_ij, p_i, p_j):
       if f_ij != 0 and p_i != 0 and p_j != 0:
           result = math.log(float(p_ij) / (p_i * p_j), 2)
           if result < 0:</pre>
               return 0
           else:
               return result
       else:
           return 0
   # Cosine similarity and distance
   def count_cos(self):
       self.init_matrix(self.cos_sim_matrix, len(self.t), 0)
       self.init_matrix(self.cos_dist_matrix, len(self.t), 0)
       for i in range(len(self.t)):
           for j in range(len(self.t)):
               vec_1 = self.pmi_matrix[self.t[i]]
               vec_2 = self.pmi_matrix[self.t[j]]
               dot_product = 0
               sqr_1 = 0
               sqr_2 = 0
               for val in range(len(vec_1)):
                   dot product += vec 1[val]*vec 2[val]
                   sqr_1 += vec_1[val]*vec_1[val]
                   sqr_2 += vec_2[val]*vec_2[val]
               if sqr_1 != 0 and sqr_2 != 0:
                   cos = float(dot_product) / (math.sqrt(sqr_1) * math.
→sqrt(sqr_2))
                   self.cos_sim_matrix[self.t[i]][j] = round(cos, 2)
                   if cos != 0:
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self.cos_dist_matrix[self.t[i]][j] = round(1/
 \rightarrowfloat(cos), 2)
                    else:
                        self.cos dist matrix[self.t[i]][j] = 0
    # Transform to numpy 2d arrays
    def np_array_transform(self):
        self.cos_sim = np.empty(shape=(len(self.t), len(self.t)))
        self.cos_dist = np.empty(shape=(len(self.t), len(self.t)))
        for i in range(len(self.t)):
            self.cos_sim[i] = self.cos_sim_matrix[self.t[i]]
            self.cos_dist[i] = self.cos_dist_matrix[self.t[i]]
        print('SIMILARITY MATRIX: ')
        print(self.cos_sim)
        print('\nDISTANCE MATRIX: ')
        print(self.cos_dist)
    # Improve: lemmatize, exclude determiners, conjunctions, prepositions
    def improve(self):
        wnl = WordNetLemmatizer()
        tagger = PerceptronTagger()
        tagset = None
        for word in self.words:
            pos = nltk.tag._pos_tag([word], tagset, tagger, lang='eng')
            stop_pos = ['DT', 'IN', 'CC', 'PRP']
            if pos[0][1] not in stop_pos:
                self.lemmatized_words.append(
                    wnl.lemmatize(word, self.get_pos(word)))
    # For improvement: get the top POS from the list of POS
    def get_pos(self, word):
        w synsets = wordnet.synsets(word)
        pos_counts = Counter()
        # noun, verb, adj, adv
        pos_counts["n"] = len([item for item in w_synsets if item.pos() == "n"])
        pos_counts["v"] = len([item for item in w_synsets if item.pos() == "v"])
        pos_counts["a"] = len([item for item in w_synsets if item.pos() == "a"])
        pos_counts["r"] = len([item for item in w_synsets if item.pos() == "r"])
        most_common_pos_list = pos_counts.most_common(3)
        return most_common_pos_list[0][0]
if __name__ == "__main__":
    # Example how to run: python pa1 solution.py -b B.txt -t T.txt -f text.txt
    # Example with additional preprocessing: python pa1_solution.py \neg b B.txt \neg t_{\sqcup}
\hookrightarrow T. txt - f text. txt - p
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# Command line arguments description
  parser = argparse.ArgumentParser(description="Similarity")
  parser.add_argument("-f", "--file", dest="path_corpus",
                       help="file path to the corpus to search for words" \sqcup
⇔similarity")
  parser.add_argument("-b", dest="path_b",
                       help="file path to the set B")
  parser.add_argument("-t", dest="path_t",
                       help="file path to the set T")
  parser.add_argument("-p", "--preprocess",
                       action="store_true", dest="fancy_preprocessing",
                       default=False,
                       help="perform extra preprocessing: lemmatization")
  args = parser.parse_args()
   # Initialize
  st = SimTester()
  st.path_corpus = args.path_corpus
  st.path_b = args.path_b
  st.path_t = args.path_t
   # Read corpus
  st.read_corpus()
  # Create sets B and T from files B.txt and T.txt
  st.create_set(st.path_b, st.b)
   st.create_set(st.path_t, st.t)
   # Initialize matrix with Add-2 smoothing
  laplace = 2
  st.init_matrix(st.matrix, len(st.b), laplace)
   # Find windows and create a feature matrix (raw frequency)
  st.find_windows(args.fancy_preprocessing)
   # Uncomment to see raw frequencies
   # print('\nRAW FREQUENCIES: ')
   # print(st.b)
   # for target in st.t:
        print(target, st.matrix[target])
   # Calculate PMI values and create a feature matrix with weights
  st.count_pmi()
   # Uncomment to see PMI values
   # print('\nPMI VALUES: ')
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# print(st.b)
# for target in st.t:
     print(target, st.pmi_matrix[target])
# Calculate cosine similarity and distance, print them
st.count_cos()
st.np_array_transform()
# Output plots and clusters
if not args.fancy_preprocessing:
   output_1 = 'plot_1.png'
   output_2 = 'plot_2.png'
else:
   output_1 = 'plot_improved_1.png'
   output_2 = 'plot_improved_2.png'
# Run hiersrchical clustering
cl.hierarchical_clusters_print(st.cos_sim, st.t)
cl.hierarchical_clusters_draw(st.cos_sim, st.t, output_filename=output_1)
# Run k-means clustering
cl.kmeans_clusters_print(st.cos_sim, st.t)
cl.pca_plot(st.cos_sim, st.t, output_filename=output_2)
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