

## Lab2\_2

November 8, 2023

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[ ]: import numpy as np
import text_functions as tf
import nltk
import matplotlib.pyplot as plt
import matplotlib
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[ ]: def word_count(data_file_name):
    amount_dictionary = {}
    text_file = open(data_file_name, "r")
    for line in text_file:
        if line != "\n":
            words = line.split()
            for word in words:
                if amount_dictionary.get(word) is None:
                    amount_dictionary[word] = 1
                else:
                    amount_dictionary[word] += 1
    text_file.close()

    return amount_dictionary

def assign_random_vectors(data_file_name, dimension, ones_number, threshold):
    dictionary = {}
    amount_dictionary = word_count(data_file_name)
    text_file = open(data_file_name, "r")
    for line in text_file: #read line in the file
        words = line.split() # extract words from the line
        for word in words: # for each word
            if dictionary.get(word) is None: # If the word was not yed added to
↳ the vocabulary
                if amount_dictionary[word] < threshold:
                    dictionary[word] = tf.get_random_word_vector(dimension,
↳ ones_number) # assign a
                else:
                    dictionary[word] = np.zeros(dimension) # frequent words are
↳ assigned with empty vectors. In a way they will not contribute to the word
↳ embedding
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text_file.close()
return dictionary, amount_dictionary

def find_synonyms(test_name, lemmatizer, dimension):
    word_space = {} #embeddings
    number_of_tests = 0
    text_file = open(test_name, "r") #open TOEFL tasks
    for line in text_file:
        words = line.split()
        words = [lemmatizer.lemmatize(lemmatizer.lemmatize(lemmatizer.
↪lemmatize(word, 'v'), 'n'), 'a') for word in
            words] # lemmatize words in the current test
        word_space[words[0]] = np.zeros(dimension)
        word_space[words[1]] = np.zeros(dimension)
        word_space[words[2]] = np.zeros(dimension)
        word_space[words[3]] = np.zeros(dimension)
        word_space[words[4]] = np.zeros(dimension)
        number_of_tests += 1
    text_file.close()
    return word_space, number_of_tests

def create_embeddings(data_file_name, synonyms, words_dic, window_size):
    text_file = open(data_file_name, "r")
    lines = [[] for _ in range(2 * window_size)] # neighboring lines
    i = window_size
    while i < 2 * window_size:
        line = "\n"
        while line == "\n":
            line = text_file.readline()
        lines[i] = line.split()
        i += 1
    line = text_file.readline()
    while line != "":
        if line != "\n":
            lines.append(line.split())
            words = [item for sublist in lines for item in sublist]
            start_index = sum(len(line) for line in lines[:window_size])
            length = len(lines[window_size])
            i = 0
            while i < length:
                if not (synonyms.get(words[start_index+i]) is None):
                    k = 1

                    while (i - k >= 0) and (k <= window_size): #process left_
↪neighbors of the focus word

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        synonyms[words[start_index + i]] = np.
↪add(synonyms[words[start_index + i]], np.roll(words_dic[words[(start_index +
↪i) - k]], -k)) # The roll indicates the ammount of steps away from the focus
↪word

        k += 1

        k = 1
        while (i + k < length) and (k <= window_size): #process
↪right neighbors of the focus word
            synonyms[words[start_index + i]] = np.
↪add(synonyms[words[start_index + i]], np.roll(words_dic[words[start_index +
↪i + k]], k)) #update word embedding
            k += 1

        i += 1
        lines.pop(0)
        line = text_file.readline()
    return synonyms

def predict(test_name, lemmatizer, synonyms, amount_dictionary, dimension,
↪zero_vector, number_of_tests):
    i = 0
    text_file = open(test_name, 'r')
    right_answers = 0.0 # variable for correct answers
    number_skipped_tests = 0.0 # some tests could be skipped if there are no
↪corresponding words in the vocabulary extracted from the training corpus
    x = 0
    while i < number_of_tests:
        line = text_file.readline() #read line in the file
        words = line.split() # extract words from the line
        words = [lemmatizer.lemmatize(lemmatizer.lemmatize(lemmatizer.
↪lemmatize(word, 'v'), 'n'), 'a') for word in
            words] # lemmatize words in the current test

        try:
            if not(amount_dictionary.get(words[0]) is None): # check if
↪there word in the corpus for the query word
                k = 1
                while k < 5:
                    # if amount_dictionary.get(words[k]) is None:
                    #     word_space[words[k]] = np.random.randn(dimension)
                    if np.array_equal(synonyms[words[k]], zero_vector): #
↪if no representation was learnt assign a random vector
                        synonyms[words[k]] = np.random.randn(dimension)
                    k += 1
                right_answers += tf.
↪get_answer_mod([synonyms[words[0]], synonyms[words[1]], synonyms[words[2]],

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synonyms[words[3]], synonyms[words[4]]) #check
↳if word is predicted right
    except KeyError: # if there is no representation for the query
↳vector than skip
        number_skipped_tests += 1
        print("skipped test: " + str(i) + "; Line: " + str(words))
    except IndexError as e:
        print(e)
        print(f"Test {i} is out of range")
        print(f"On line {line}")
        break
    i += 1
text_file.close()
return 100 * right_answers / number_of_tests

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[ ]: # Base stuff
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def run_tests(threshold, dimensions, ones_number, window_size, number_of_tests):
    lemmatizer = nltk.WordNetLemmatizer() # create an instance of lemmatizer
    test_name = "new_toefl.txt" # file with TOEFL dataset
    data_file_name = "lemmatized.text" # file with the text corpus

    dimension_test_results = {}
    for dimension in dimensions:
        dimension_test_results[dimension] = []

    for i in range(number_of_tests):
        print(f"##### Test {i + 1} #####")
        for dimension in dimensions:
            zero_vector = np.zeros(dimension)
            random_vector_dic, amount_dic =
↳assign_random_vectors(data_file_name, dimension, ones_number, threshold)
            synonyms, number_of_tests = find_synonyms(test_name, lemmatizer,
↳dimension)
            processed_text = create_embeddings(data_file_name, synonyms,
↳random_vector_dic, window_size)
            accuracy = predict(test_name, lemmatizer, processed_text,
↳amount_dic, dimension, zero_vector, number_of_tests)

            print(f"Accuracy for dimension {dimension} is {accuracy}%")

            dimension_test_results[dimension].append(accuracy)

    return dimension_test_results

def plot_results(results):
    for result in list(results.keys()):

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results[result] = np.array(results[result])

points = len(next(iter(results.values()))))
colors = matplotlib.cm.rainbow(np.linspace(0, 1, points))
color_index = 0
for dimension in results.keys():
    for simulation in results[dimension]:
        plt.scatter(dimension, simulation, color=colors[color_index])
        color_index += 1
    color_index = 0
plt.title('Dimensionality vs Accuracy')
plt.ylabel('Accuracy')
plt.xlabel('Dimensions')
plt.ylim(0, 100)
plt.show()

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[ ]: threshold = 15000 # Frequency threshold in the corpus ??
dimensions = [100, 1000, 4000, 10000] # Dimensionality for high-dimensional
    ↪ vectors
test_number = 5 # number of tests to run

ones_number = 2 # number of nonzero elements in randomly generated
    ↪ high-dimensional vectors
window_size = 2 # number of neighboring words to consider both back and forth.
    ↪ In other words number of words before/after current word

# Run tests
results = run_tests(threshold, dimensions, ones_number, window_size,
    ↪ test_number)
plot_results(results)

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##### Test 1 #####
Accuracy for dimension 100 is 53.75%
Accuracy for dimension 1000 is 71.25%
Accuracy for dimension 4000 is 67.5%
Accuracy for dimension 10000 is 76.25%
##### Test 2 #####
Accuracy for dimension 100 is 62.5%
Accuracy for dimension 1000 is 68.75%
Accuracy for dimension 4000 is 75.0%
Accuracy for dimension 10000 is 71.25%
##### Test 3 #####
Accuracy for dimension 100 is 52.5%
Accuracy for dimension 1000 is 67.5%
Accuracy for dimension 4000 is 67.5%
Accuracy for dimension 10000 is 72.5%
##### Test 4 #####

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Accuracy for dimension 100 is 58.75%  
Accuracy for dimension 1000 is 68.75%  
Accuracy for dimension 4000 is 70.0%  
Accuracy for dimension 10000 is 76.25%  
##### Test 5 #####  
Accuracy for dimension 100 is 55.0%  
Accuracy for dimension 1000 is 57.5%  
Accuracy for dimension 4000 is 73.75%  
Accuracy for dimension 10000 is 71.25%

