cleaned_topic_modeling_and_similarities_with_W2Vec_tfidf_LDA-checkpoint

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1 Topic Modeling and Similarities

• Latent Dirichlet Allocation (LDA) with Python

```
[1]: import numpy as np
import sys, re
import nltk
import gensim

import glob, sys

data_dir = './datasets/bbc_sports/'
from nlp_utils import prepare_text
```

1.0.1 Load, Preprocess and Tokenize the document

```
[2]: def read_data_files(data_dir):
    all_data_files = glob.glob(data_dir+'*')
    raw_doc = []
    for file in all_data_files:
        # Use try and except method as some files may not be readable
        try:
            f = open(file, 'r', encoding='utf-8')
              raw_doc.append(f.read())
        except:
            print (f"skipping the unreadable file: {file}")
            pass
        return raw_doc

raw_doc = read_data_files(data_dir)
    print (f"Total # of documents: {len(raw_doc)}")
```

skipping the unreadable file: ./datasets/bbc_sports/199.txt Total # of documents: 510

```
[3]: def prepare_data(raw_doc):
         tokenized doc = [prepare_text(doc, TOKENIZE=True, STEM=True) for doc in_
      →raw_doc]
                     = gensim.corpora.Dictionary(tokenized doc)
         dictionary
         numerical_corpus = [dictionary.doc2bow(text) for text in tokenized_doc]
         return (dictionary, tokenized_doc, numerical_corpus)
     def model_lda(numerical_corpus, dictionary):
         return gensim.models.LdaModel(corpus=numerical_corpus, num_topics=10,__
     →id2word=dictionary)
     def model_tfidf(numerical_corpus, dictionary):
         tf idf = gensim.models.TfidfModel(numerical corpus, id2word=dictionary)
         similarity_object = gensim.similarities.Similarity(data_dir,
                                                             tf idf[numerical corpus],
      →num_features=len(dictionary))
         return tf_idf, similarity_object
     def model_w2v(tokenized_doc):
         w2v_model = gensim.models.Word2Vec(
                     sentences=tokenized_doc,
                     size=300, # The size of the dense vector to represent each
      \rightarrow token or word
                     window=10, # The maximum distance between the target word and
      \rightarrow its neighboring word.
                     min_count=5, # Minimium frequency count of words. The model
      →would ignore words with counts< min_count
                     workers=10) # How many threads to use behind the scenes
         w2v_model.train(tokenized_doc, total_examples=len(tokenized_doc), epochs=15)
         return w2v_model
     def save_model(model, model_name):
         import joblib
         joblib.dump(model, model_name)
         #model can be loaded as
         #joblib.load(model file joblib)
         print (f"model: {model} is saved to {model_name}")
         #import pickle
         #model_file_pickle = './datasets/tfidf_model.p'
         #pickle.dump(model, open(model_name, 'wb'))
```

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[4]: def sort_list(A, key=0):
         return sorted(A, reverse=True, key=lambda x: x[key] )
     def print_top_k(topics, all_topics, k=2):
         topics_sorted = sort_list(topics, key=1)
         for i, topics in enumerate(topics_sorted[:k]):
             idx = topics[0]
             print ("\n",i, all_topics[idx])
     def test_lda(raw_doc):
         (dictionary, tokenized_doc, numerical_corpus) = prepare_data(raw_doc)
         model = model_lda(numerical_corpus, dictionary)
         model_name = './models/lda_model'
         save_model(model, model_name)
         all_topics = model.print_topics()
         print( "\n first 10 most representative topics")
         for i in range(10):
             print(f"\nTopic #{i} : {model.print_topic(i, 5)}")
         doc_new = "My wife plans to go out tonight."
         doc_new_prepared = prepare_text(doc_new, TOKENIZE=True)
         print ( doc_new_prepared )
         doc bow = dictionary.doc2bow(doc new prepared)
         print (doc_bow)
         topics= model.get_document_topics( doc_bow )
         top_k_topics = print_top_k(topics, all_topics, k=2)
     test_lda(raw_doc)
    model: LdaModel(num_terms=10321, num_topics=10, decay=0.5, chunksize=2000) is
    saved to ./models/lda_model
     first 10 most representative topics
    Topic #0: 0.008*"play" + 0.007*"said" + 0.005*"cup" + 0.005*"last" +
    0.005*"world"
    Topic #1: 0.011*"said" + 0.007*"player" + 0.006*"england" + 0.006*"game" +
    0.006*"win"
    Topic #2: 0.008*"said" + 0.008*"win" + 0.006*"play" + 0.006*"game" +
    0.005*"two"
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0.005*"player"
    Topic #4: 0.010*"said" + 0.007*"play" + 0.007*"game" + 0.006*"win" + 0.005*"go"
    Topic #5 : 0.006*"first" + 0.005*"play" + 0.005*"champion" + 0.005*"said" +
    0.005*"win"
    Topic #6: 0.007*"said" + 0.006*"game" + 0.006*"back" + 0.006*"first" +
    0.005*"play"
    Topic #7: 0.009*"said" + 0.009*"game" + 0.006*"player" + 0.005*"play" +
    0.005*"year"
    Topic #8 : 0.007*"said" + 0.006*"england" + 0.005*"v" + 0.005*"last" +
    0.005*"world"
    Topic #9 : 0.009*"said" + 0.008*"play" + 0.007*"win" + 0.007*"game" +
    0.006*"would"
    ['wife', 'plans', 'go', 'tonight']
    [(57, 1), (2074, 1), (5826, 1)]
     0 (5, '0.006*"first" + 0.005*"play" + 0.005*"champion" + 0.005*"said" +
    0.005*"win" + 0.005*"year" + 0.004*"ireland" + 0.004*"world" + 0.004*"game" +
    0.004*"second"')
     1 (8, '0.007*"said" + 0.006*"england" + 0.005*"v" + 0.005*"last" +
    0.005*"world" + 0.004*"first" + 0.004*"win" + 0.004*"game" + 0.004*"ireland" +
    0.004*"year"')
[5]: def test_tfidf(raw_doc):
         (dictionary, tokenized_doc, numerical_corpus) = prepare_data(raw_doc)
         tf_idf, similarity_object = model_tfidf(numerical_corpus, dictionary)
        model_name = './models/tfidf_model'
        save_model(tf_idf, model_name)
        #query_text
        q_text = raw_doc[8]
        q_text_processed = prepare_text(q_text, TOKENIZE=True, STEM=True)
        print ( "\nfirst 10 tokens:\n",q_text_processed[:10])
        q_text_bow = dictionary.doc2bow(q_text_processed)
        print ( "\nfirst 10 bow:\n",q_text_bow[:10])
        q_text_tfidf = tf_idf[q_text_bow]
        print ( "\nfirst 10 tfidf:\n",q_text_tfidf[:10] )
         similarity_scores=list(similarity_object[q_text_tfidf])
         print ( "\nfirst 10 similarity scores:\n", similarity_scores[:10])
```

Topic #3: 0.012*"said" + 0.006*"win" + 0.006*"game" + 0.006*"play" +

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max_score = max(similarity_scores)
    max_score_index = similarity_scores.index(max_score)
    print (f"\nmax score {max_score} and max score index {max_score_index}")
    sorted_score = sorted(similarity_scores, reverse=True)
    indices = []
    for i in range(3):
        score = sorted score[i]
        indx = similarity_scores.index(score)
        print ( f"\nscore: {score} index:{indx}")
        indices.append(indx)
    print ("\nPritining the docs with very similar similarity score\n")
    print ("\n", raw_doc[indices[0]][:100] )
    print ("\n\n", raw_doc[indices[1]][:100] )
    print ("\n^{n}, raw_doc[indices[1]][:100])
test_tfidf(raw_doc)
model: TfidfModel(num_docs=510, num_nnz=67925) is saved to ./models/tfidf_model
first 10 tokens:
 ['robben', 'sidelin', 'broken', 'footchelsea', 'winger', 'arjen', 'robben',
'broken', 'two', 'metatars']
first 10 bow:
 1), (111, 1)]
first 10 tfidf:
 [(23, 0.060779464016390256), (45, 0.07075027144828656), (46,
0.014605368204018868), (50, 0.028797575313044728), (57, 0.015421487398221299),
(71, 0.0404114823434502), (94, 0.015008913499203016), (106,
0.004434193825346702), (110, 0.020478063934517673), (111, 0.026530003463721384)]
first 10 similarity scores:
 [0.024998276, 0.020571105, 0.009893991, 0.016575314, 0.014618116, 0.013003329,
0.020246074, 0.011348683, 1.0000001, 0.022904249]
max score 1.0000001192092896 and max score index 8
score: 1.0000001192092896 index:8
```

```
score: 0.2723110318183899 index:220
score: 0.1732272207736969 index:113
Pritining the docs with very similar similarity score
Robben sidelined with broken foot
Chelsea winger Arjen Robben has broken two metatarsal bones in hi
Robben plays down European return
Injured Chelsea winger Arjen Robben has insisted that he only has
```

Injured Chelsea winger Arjen Robben has insisted that he only has

1.1 Word2Vec Model

Robben plays down European return

```
[6]: def test word2vec(raw doc):
         (dictionary, tokenized_doc, numerical_corpus) = prepare_data(raw_doc)
         model = model_w2v(tokenized_doc)
         model_name = './models/w2v_model'
         save_model(model, model_name)
         words = list(model.wv.vocab)
         print (f"\nThere are {len(words)} words. First 10 words:\n{words[:10]}")
         vectors = np.array([model.wv[word] for word in words])
         print (f"\nvectors.shape:{vectors.shape}")
         my word = 'defend'
         sim_word = model.wv.most_similar(positive=my_word)
         print (f"\ntop 3 similar words to {my_word} are {sim_word[:3]}")
         v=model.wv[my_word]
         idx=dictionary.doc2idx([my_word])[0]
         print (f"\ndictionary index: {idx}")
         print (f"\nsanity check for word {my word} index: {idx}^th item in the__
     →dictionary: {dictionary[idx]}")
         word_pairs = [["celtic", "everton"],
```

```
["good", "bad"],
                   ["good", "celtic"],
                   ["good", "good"],
                   ["kid", "men"] ]
         for (w1, w2) in word_pairs:
             simi_score = model.wv.similarity(w1=w1, w2=w2)
             print (f"\nSimilarity score of ## Word2Vec Model## '{w1}' and '{w2}' :__
      →{simi score}")
     test_word2vec(raw_doc)
    model: Word2Vec(vocab=2817, size=300, alpha=0.025) is saved to
    ./models/w2v_model
    There are 2817 words. First 10 words:
    ['robinson', 'blast', 'coach', 'andi', 'insist', 'livid', 'side', 'deni', 'two',
    'tri']
    vectors.shape:(2817, 300)
    top 3 similar words to defend are [('rosenborg', 0.8802071809768677),
    ('burnley', 0.8630086779594421), ('milan', 0.8540078401565552)]
    dictionary index: 31
    sanity check for word defend index: 31 th item in the dictionary: defend
    Similarity score of ## Word2Vec Model## 'celtic' and 'everton' :
    0.8854888677597046
    Similarity score of ## Word2Vec Model## 'good' and 'bad': 0.711058497428894
    Similarity score of ## Word2Vec Model## 'good' and 'celtic' :
    0.11172105371952057
    Similarity score of ## Word2Vec Model## 'good' and 'good' : 1.0
    Similarity score of ## Word2Vec Model## 'kid' and 'men' : 0.16241569817066193
[]:
[]:
```