index.py

```
Authors: Manas Gaur, Amanuel Alambo
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Index structure:
   The Index class contains a list of IndexItems, stored in a dictionary
type for easier access
   each IndexItem contains the term and a set of PostingItems
    each PostingItem contains a document ID and a list of positions that
the term occurs
import util
import doc
from collections import OrderedDict,defaultdict #for dealing with sorting
import math
import cran
import json
import sys
class Posting:
   def __init__(self, docID):
       self.docID = docID
        self.positions = []
    def append(self, pos):
        self.positions.append(pos)
    def sort(self):
        ''' sort positions'''
        self.positions.sort()
```

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```
A Simple Search Engine
    def merge(self, positions):
        self.positions.extend(positions)
    def term_freq(self):
        ''' return the term frequency in the document'''
       #ToDo
        return len(self.positions) #term frequency in a document is the
class IndexItem:
   def __init__(self, term):
        self.term = term
        self.posting = {}  #postings are stored in a python dict for
        self.sorted_postings= [] # may sort them by docID for easier
    def add(self, docid, pos):
        ''' add a posting'''
        if not self.posting.has_key(docid):
            self.posting[docid] = Posting(docid)
        self.posting[docid].append(pos)
    def sort(self):
        ''' sort by document ID for more efficient merging. For each
document also sort the positions'''
        self.posting = OrderedDict(sorted(self.posting.items(), key=lambda
t:t[0]))
        for docID,pos_list in self.posting.iteritems():
            self.posting[docID] = sorted(pos_list, reverse=False)
class InvertedIndex:
    def __init__(self):
       self.items = defaultdict(list) # list of IndexItems--this
corresponds to a structure of the form
\{term-1: \{doc-1: [pos-1,...,pos-n]\},....\}
```

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```
self.nDocs = ∅ # the number of indexed documents
   def indexDoc(self, doc): # indexing a Document object
        ''' indexing a docuemnt, using the simple SPIMI algorithm, but no
need to store blocks due to the small collection we are handling. Using
save/load the whole index instead'''
       # ToDo: indexing only title and body; use some functions defined in
util.py
       # (2) remove stopwords,
       doc_tb = doc.title + doc.body #concatenating the title and the
       words = util.splitDoc(doc_tb) #call to util's splitDoc method
which performs calls tokenizing, stemming, stopword removal, lowercasing
       for term in list(set(words)): #iterating through the set of words
           term = str(term) #conversion into a string
           term_positions = [i for i,val in enumerate(words) if val==term]
           doc_dict = {} #dictionary to store a document's docID as
           doc_dict[doc.docID] = term_positions #term positions in a
document
           if term not in list(self.items.keys()):
                                                    #condition to check
presence of a key in a dict before appending a value
               self.items[term]=list()
                                                  #intializes a
dictionary for a term if the term doesn't appear as a key in the dictionary
               self.items[term].append(doc_dict) #a dictionary of
items
           else:
               self.items[term].append(doc_dict) #if term is present
```

```
in the items dictionary the docID and term positions are appended
       self.nDocs += 1 #increments with each call to the method
       return self.items
   def sort(self):
        ''' sort all posting lists by docID'''
   def find(self, term):
       return self.items[term]
   def save(self, filename):
       ''' save to disk'''
       # ToDo: using your preferred method to serialize/deserialize the
       with open(filename, 'w') as f:
           json.dump(self.items, f, indent=4) #json used for
   def load(self, filename):
       ''' load from disk'''
       with open(filename) as f:
           indexed_file = json.load(f)
       return indexed_file
   def idf(self, term):
       ''' compute the inverted document frequency for a given term'''
       #ToDo: return the IDF of the term
       idf_score = math.log(self.nDocs, len(self.items[term]))
       return idf_score
   # more methods if needed
```

```
def test():
   ''' test your code thoroughly. put the testing cases here'''
   II = InvertedIndex() #InvertedIndex class instantiated
   index_loaded = II.load('test.json') #loading an indexed file---name
of the file passed should be the same name used in indexing and saving the
file
   print type(index_loaded['four']) #testing for example entry 'four'
   print (index_loaded['four'][0])
   print 'Pass'
def indexingCranfield():
   # command line usage: "python index.py cran.all index_file"
   # the index is saved to index_file
   cran_file = sys.argv[1]
   index_file = sys.argv[2]
   cf = cran.CranFile(cran_file) #instantiation of CranFile class
   II = InvertedIndex() #InvertedIndex class instantiated
   dump=list()
                     #to store list of each doc to be indexed
   for doc in cf.docs:
       dump.append(II.indexDoc(doc))
   II.save(index_file) #saves to an output file----call to 'save'
method of 'InvertedIndex' class
   print 'Done'
if __name__ == '__main__':
   indexingCranfield() #call to indexingCranfield method
   test() #call to method 'test'
```

util.py

```
Authors: Manas Gaur, Amanuel Alambo
Instructor: Dr. keke Chen
   utility functions for processing terms
   shared by both indexing and query processing
import cran
import nltk
from nltk.tokenize import word_tokenize
from nltk.stem.snowball import SnowballStemmer
def isStopWord(word):
    ''' using the NLTK functions, return true/false'''
   with open('stopwords', 'r') as f:
         stop_words = f.read()
   return word in stop_words
def stemming(word):
    ''' return the stem, using a NLTK stemmer. check the project
description for installing and using it'''
```

```
##SnowballStemmer
   stemmer = SnowballStemmer("english", ignore_stopwords=True)
   return stemmer.stem(word)
   #######
def splitDoc(doc):
   tokens = word_tokenize(doc)
   words = [word for word in tokens if word.isalpha()] #punctuation
   words = [word for word in words if not isStopWord(word)] #stopwords
removal
   words = [stemming(word) for word in words] #word stemming
   words = [word.lower() for word in words] #lowercasing words
   return words
###### test snippet
if __name__ == '__main__':
   print isStopWord('ffff')
   print stemming('athletes')
   docs_list = []
   cf = CranFile ('../CranfieldDataset/cran.all') #preprocess all docs
in the collection
   for doc in cf.docs:
       docs_list.append(splitDoc(doc))
#####
```

query.py

```
Authors: Manas Gaur, Amanuel Alambo
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query processing
import util
import norvig_spell
import index
import math  #for cosine similarity
from itertools import izip
import json
from collections import defaultdict, OrderedDict
import cran
import cranqry
import numpy as np
import operator
import pandas as pd
import random
import metrics
from scipy.stats import wilcoxon
import sys
```

```
import random
import copy
class QueryProcessor:
   def __init__(self, query, index, collection):
        ''' index is the inverted index; collection is the document
collection'''
       self.raw query = query
       self.index = index
       self.docs_fname = collection
       self.cf = cran.CranFile ('../CranfieldDataset/cran.all')
       self.nDocs=len(self.cf.docs)
   def preprocessing(self):
        ''' apply the same preprocessing steps used by indexing,
           also use the provided spelling corrector. Note that
            spelling corrector should be applied before stopword
           removal and stemming (why?)'''
       corrected_terms_list = list()
       for term in self.raw_query.split(' '): #splitting on white space
           corrected_terms_list.append(norvig_spell.correction(term))
       try:
                 corrected_terms_list.remove('gw') #since we used the
Cranfield dataset for spelling correction, 'gw' appeared and we remove here
       except:
       corrected_terms_text = ' '.join(corrected_terms_list)
       terms = util.splitDoc(corrected_terms_text)
       return terms #list of terms
   def booleanQuery(self):
        ''' boolean query processing; note that a query like "A B C" is
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transformed to "A AND B AND C" for retrieving posting lists and merge
them'''
       #ToDo: return a list of docIDs
       terms = self.preprocessing()
       inv_index = self.index
                           #list of docIDs for all terms
       doc list = list()
       for term in terms:
           print term
           term_doc_list = list() #list of docIDs for a single term
           if inv_index.has_key(term):
                 for i in inv_index[term]:
                       key = ''.join(i.keys())
                       term_doc_list.append(key)
           else:
                 continue
           if len(doc_list) == 0: #a workaround to deal with an initially
empty list---as intersection with an empty list is an empty list
                 doc_list = term_doc_list
           else:
                 doc_list = list(set(doc_list).union(set(term_doc_list)))
             if len(doc list) == 0:
                 doc_list = term_doc_list #doc_list is first initialized
with a list of docIDs where the first term appears in
           else:
                #doc list =
list(set(doc_list).intersection(set(term_doc_list))) #list of terms' docIDs
combined by a logical AND---intersection done on the fly
                doc_list = list(set(doc_list).union(set(term_doc_list)))
               #print len(doc_list)
       doc_list = [str(doc_id) for doc_id in list(set(doc_list))]
       return doc list #list of docID containing all terms in a query
```

```
are returned to calling program---since sets support ordered list, when
needed they help in retrieving ranked results
   def vectorQuery(self, k):
        ''' vector query processing, using the cosine similarity. '''
       #ToDo: return top k pairs of (docID, similarity), ranked by their
cosine similarity with the query in the descending order
       terms = self.preprocessing()
       scores = defaultdict(list) #stores scores for each of the
documents----follow algorithm---docID is the key
       inv index = self.index #loads the index file which is in json
form
       qterm_weights = list() #this is for one query term
       docID = None
       throwawaylist = []
       for term in terms: #iterating through the query terms
           w_tq = terms.count(term) #frequency of a term in a query
           qterm_weights.append(w_tq/float(len(terms))) #normalize term
           try:
                 postings_list = inv_index[term] #postings list for a term
           except:
                 continue
           term_docID_list = [int(dict_ID.keys()[0]) for dict_ID in
postings_list]
           for doc in self.cf.docs:
               doc ID = doc.docID
               if int(doc_ID) in term_docID_list:
                   len_docID = len(util.splitDoc(doc.body)) #total
                   len positions = 0
                   for term_dict in postings_list:
                       if term_dict.keys()[0] == doc_ID:
                           len_positions = len(term_dict.values())
                           break
```

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```
tf = len_positions/float(len_docID)
                                                         #division of
                    tf idf = tf * self.idf(self.nDocs,len(postings list))
                    if doc_ID not in list(scores.keys()):
                        scores[doc_ID]=list()
                        scores[doc_ID].append(tf_idf)
                    else:
                        scores[doc_ID].append(tf_idf)
                else:
                    if doc_ID not in list(scores.keys()):
                        scores[doc_ID]=list()
                        scores[doc_ID].append(0.0)
                        scores[doc_ID].append(0.0)
       doc similarity = {}
       for docid, scores_list in scores.iteritems():
doc_similarity[docid]=self.cosine_similarity(qterm_weights,scores_list)
        sorted_scores = sorted(doc_similarity.items(),
key=operator.itemgetter(1), reverse=True)
       top_k = [(i[0], i[1]) for i in sorted_scores[:k]]
       return top_k
   def idf(self, nDocs, nDocs_term):
        ''' compute the inverted document frequency for a given term'''
       #ToDo: return the IDF of the term
       try:
           idf_score = math.log(nDocs, nDocs_term)
       except:
           idf_score = 1.0
       return idf_score
   def dot_product(self, v1, v2):
       return sum(map(lambda x: x[0] * x[1], izip(v1, v2)))
```

```
def cosine_similarity(self, vec1, vec2):
       prod = sum(map(lambda x: x[0] * x[1], izip(vec1, vec2)))
       vec1_len = math.sqrt(self.dot_product(vec1, vec1))
       vec2_len = math.sqrt(self.dot_product(vec2, vec2))
       if vec1_len == 0.0 or vec2_len == 0.0:
           return 0.0
       else:
           return round(prod / ((vec1_len * vec2_len)), 2)
def to_ndcg(qrels, q_text, idx_file, tk=10, n=2):
    column_names =['qid', 'docid', 'bool_rel', 'vec_rel'] #for creating a
   #df_qrels = pd.read_csv('.../CranfieldDataset/qrels.text',
names=column names, sep=' ') #can test by hard-coding
   df_qrels = pd.read_csv('.../CranfieldDataset/qrels.sample',
names=column_names, sep=' ') #can test by hard-coding
   #print df_qrels
   unique_qids = list(set(list(df_qrels.qid.values)))
   random.shuffle(unique_qids)
   random_qids = unique_qids[0:n]
   qrys = cranqry.loadCranQry('.../CranfieldDataset/query.text') #qrys is
   #qrys = cranqry.loadCranQry(q_text) #qrys is a dict
   qrys_ids = [key for key,val in qrys.iteritems()]
   II = index.InvertedIndex()
   index_file = II.load("index_file.json") #for hard-coded testing
   #index_file = II.load(idx_file)
   vec_agg_ndcg, bool_agg_ndcg = list(), list() #for storing aggregate
ndcg scores
   for qid in random_qids:
     print qid
```

```
df_qid = df_qrels[df_qrels["qid"] == qid] #dataframe for one
query id---comparison of an integer qid in a string qid
     qid_docids = list(df_qid['docid']) #list of doc ids for a randomly
chosen query id from qrels.text---to be used for ndcg_score
     print qid_docids
       st_qid = str(qid) #very important----the decimal number in
random_qids should be matched the octal numbers in the cranfield dataset
       if len(st_qid) == 1: #for handing decimal to octal qid
conversion
           st_qid = "00" + st_qid
       elif len(st_qid) == 2:
           st_qid = "0" + st_qid
       else:
           st_qid = st_qid
       if st_qid in qrys_ids:
           qp = QueryProcessor(qrys[st_qid].text, index_file, 'cran.all')
           bool_array = qp.booleanQuery()
           vec_array = qp.vectorQuery(10) #change back to 'tk'
           print bool array
           bool_array = [int(v) for v in bool_array]
           print bool_array
           bool_list = [(0,0)] * 10 #change back to tk
           idx = 0
           for doc_id in bool_array:
                 if doc_id in qid_docids: #iteratively check if a docid
returned by the vector model is present in grels.text for the specific
query(qid)
                       #y true[idx] = 1
                       bool_list[idx] = (1,1)
                       idx += 1
                 else:
                       bool_list[idx] = (0,1)
                 if idx == 10:
```

```
break
           #print bool list
           y_true = [int(bool_id[0]) for bool_id in bool list]
           y_score = [int(bool_id[1]) for bool_id in bool_list]
           print "bool", y_true
           print "bool", y_score
           bool_agg_ndcg.append(metrics.ndcg_score(y_true, y_score, 10))
           print vec_array
           y_score = [vec_id[1] for vec_id in vec_array] #y_score--to be
           vec_ids = [int(vec_id[0]) for vec_id in vec_array] #list of
docids from the list of tuples of the form (docid, similarity_score)
           #print vec ids
           y true = [0] * 10  ##added on 0317---change back to tk
           idx = 0
           for doc_id in vec ids:
                 if doc_id in qid_docids: #iteratively check if a docid
returned by the vector model is present in grels.text for the specific
query(qid)
                       y_{true}[idx] = 1
                       idx += 1
           print "vec", y_true
           print "vec", y_score
           vec_agg_ndcg.append(metrics.ndcg_score(y_true, y_score, 10))
           del qp
                     ##garbage collection
   return bool_agg_ndcg, vec_agg_ndcg
def test():
   ''' test your code thoroughly. put the testing cases here'''
   print 'Pass'
```

```
def query():
    ''' the main query processing program, using QueryProcessor'''
   II = index.InvertedIndex()
   index_file = sys.argv[1]
   index file = II.load(index file)
   proc_alg = sys.argv[2]
   proc_alg = proc_alg
   q_text = sys.argv[3]
   q_text = q_text
   qid = sys.argv[4]
   qid = qid
   qrys = cranqry.loadCranQry(q_text) #qrys is a dict
   #qrys = cranqry.loadCranQry('.../CranfieldDataset/query.text') #can
also be hard-coded like this one
   qp = QueryProcessor(qrys[qid].text, index_file, 'cran.all') #qid, and
index_file are to be passed by the user
   #print qp.booleanQuery()
   if proc_alg == '0':
       qp.booleanQuery()
       print qp.booleanQuery()
   elif proc_alg == '1':
       qp.vectorQuery(3)
       print qp.vectorQuery(3)
   # ToDo: the commandline usage: "echo query_string | python query.py
index_file processing_algorithm"
   # processing_algorithm: 0 for booleanQuery and 1 for vectorQuery
   # for booleanQuery, the program will print the total number of
documents and the list of document IDs
```

```
# for vectorQuery, the program will output the top 3 most similar
documents

if __name__ == '__main__':
    #test()
    query()
```

Batch_eval.py

```
Authors: Manas Gaur, Amanuel Alambo
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a program for evaluating the quality of search algorithms using the vector
```

```
model
it runs over all queries in query.text and get the top 10 results,
and then qrels.text is used to compute the NDCG metric
usage:
    python batch eval.py index file query.text grels.text n
   output is the average NDCG over all the queries for boolean model and
vector model respectively.
      also compute the p-value of the two ranking results.
import metrics
import query
from scipy.stats import ttest_ind
import sys
import math
def eval():
      idx_file = sys.argv[1] #index file
      idx_file = idx_file
      q_text = sys.argv[2] #query text
      q_text = q_text
      qrels = sys.argv[3] #qrels file
      qrels = qrels
      n = sys.argv[4] #qrels file
      n = int(n) #typecasting into int, so it can be processed by
      #vec_ndcg_scores = list()
      for i in range(1):
            bool_ndcg_score, vec_ndcg_score = query.to_ndcg(qrels, q_text,
idx_file, 10, n)
```