

MIDS - w261 Machine Learning At Scale

Course Lead: Dr James G. Shanahan (**email** Jimi via James.Shanahan AT gmail.com)

Assignment - HW5

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Week: 5

NOTE: please replace 1234567 with your student id above

Due Time: HW is due the Tuesday of the following week by 8AM (West coast time). I.e., Tuesday, Feb 14, 2017 in the case of this homework.

- **HW5 Phase 1** This can be done on a local machine (with a unit test on the cloud such as AltaScale's PaaS or on AWS) and is due Tuesday, Week 6 by 8AM (West coast time). It will primarily focus on building a unit/systems and for pairwise similarity calculations pipeline (for stripe documents)
- **HW5 Phase 2** This will require the AltaScale cluster and will be due Tuesday, Feb 21 by 8AM (West coast time). The focus of HW5 Phase 2 will be to scale up the unit/systems tests to the Google 5 gram corpus. This will be a group exercise

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1 Instructions

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MIDS UC Berkeley, Machine Learning at Scale
DATSCIW261 ASSIGNMENT #5

Version 2017-9-2

IMPORTANT

This homework can be completed locally on your computer

=== INSTRUCTIONS for SUBMISSIONS ===

Follow the instructions for submissions carefully.

Each student has a HW-<user> repository for all assignments.

Click this link to enable you to create a github repo within the MIDS261 Classroom:

<https://classroom.github.com/assignment-invitations/3b1d6c8e58351209f9dd865537111ff8>
[\(https://classroom.github.com/assignment-invitations/3b1d6c8e58351209f9dd865537111ff8\)](https://classroom.github.com/assignment-invitations/3b1d6c8e58351209f9dd865537111ff8)

and follow the instructions to create a HW repo.

Push the following to your HW github repo into the master branch:

- Your local HW5 directory. Your repo file structure should look like this:

```
HW-<user>
  --HW3
    |__MIDS-W261-HW-03-<Student_id>.ipynb
    |__MIDS-W261-HW-03-<Student_id>.pdf
    |__some other hw3 file
  --HW4
    |__MIDS-W261-HW-04-<Student_id>.ipynb
    |__MIDS-W261-HW-04-<Student_id>.pdf
    |__some other hw4 file
  etc..
```

2 Useful References

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- See async and live lectures for this week

HW Problems

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3. HW5.0 data warehouse; star schema

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- What is a data warehouse? What is a Star schema? When is it used?

A Data Warehouse is a repository of data from a variety of sources and is typically used in an enterprise setting as the basis for Business Intelligence and other types of data analysis. DWs can contain a range of data from structured to unstructured.

Star Schema is an approach used in Data Warehouses where a fact table is used as a compound reference of multiple dimension tables. This is used to create a single table that can represent the relation between two dimension tables. For example, if we own a business with multiple locations, our data repository may contain a table for product inventory, customers, and store location. Each of these 3 tables would be dimensions. The fact table would contain foreign keys that reference the associated product/customer/store from their respective table.

3. HW5.1 Databases: 3NF; denormalized

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- In the database world What is 3NF? Does machine learning use data in 3NF? If so why?
- In what form does ML consume data?
- Why would one use log files that are denormalized?
- 3NF (Third Normal Form) is format that data is typically stored within a database where data from multiple tables is represented in smaller tables by their keys. 3NF does not usually contain all of the fields from the referenced tables and therefore in Machine Learning, we will often need to denormalize the data to get the complete picture. If the 3NF table contains all of the necessary information, it may be used, but this is not always the case.
- ML can technically consume data in 1st, 2nd, or 3rd normal form. As mentioned above, if the normalized data contains the required content, denormalization may not be necessary. In an ideal scenario, ML would have all of the required content within each record (3NF), but due to technologies such as Hadoop, we can perform massive joins much more efficiently than we could with traditional databases.
- Denormalized log files are beneficial because a typical log file can have a mass amount of rows. If we were required to reference other tables for additional data associated with each line item, we would be much less efficient than if we were to simply have all of the required content within each line.

3. HW5.2 Memory-backed map-side

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Using MRJob, implement a hashside join (memory-backed map-side) for left, right and inner joins. Use the following tables for this HW and join based on the country code (third column of the transactions table and the second column of the Countries table:

```
transactions.dat
Alice Bob|$10|US
Sam Sneed|$1|CA
Jon Sneed|$20|CA
Arnold Wesise|$400|UK
Henry Bob|$2|US
Yo Yo Ma|$2|CA
Jon York|$44|CA
Alex Ball|$5|UK
Jim Davis|$66|JA
```

```
Countries.dat
United States|US
Canada|CA
United Kingdom|UK
Italy|IT
```

Justify which table you chose as the Left table in this hashside join.

Please report the number of rows resulting from:

- (1) Left joining Table Left with Table Right
- (2) Right joining Table Left with Table Right
- (3) Inner joining Table Left with Table Right

```
In [251]: %%writefile transactions.txt
Alice Bob|$10|US
Sam Sneed|$1|CA
Jon Sneed|$20|CA
Arnold Wesise|$400|UK
Henry Bob|$2|US
Yo Yo Ma|$2|CA
Jon York|$44|CA
Alex Ball|$5|UK
Jim Davis|$66|JA
```

Overwriting transactions.txt

```
In [252]: %%writefile countries.txt
United States|US
Canada|CA
United Kingdom|UK
Italy|IT
```

Overwriting countries.txt

For the LEFT join, we output all of the rows from the TRANSACTIONS file, regardless of whether they have a match in the COUNTRIES file. If there is a match, we join the rows.

The Reason that I chose to use the TRANSACTIONS file as my LEFT table is because it logically made more sense. The table on the left will output ALL of its rows regardless of a match in the RIGHT table or not. In the scenario where we have a TRANSACTIONS file and a COUNTRIES file, it seemed to me that a typical use case would be more likely to be interested in joining the long description of the country onto each transaction, as opposed to the other way around.

Also, I chose to load my COUNTRIES file into memory as it was the smaller of the two tables.

```

In [14]: %%writefile leftHashJoin.py
#!/usr/bin/env python

from mrjob.job import MRJob
from mrjob.step import MRStep
from mrjob.compat import jobconf_from_env
import os

class leftJoin(MRJob):

    def steps(self):
        countries=[]
        return [
            MRStep(
                mapper_init = self.mapper_init,
                mapper = self.mapper,
                mapper_final = self.mapper_final
            )]

    def mapper_init(self):
        countries = open(str(os.path.dirname(os.path.realpath(__file__)))+"",
        self.left=0
        self.nomatch=0
        self.ct = {}
        for line in countries:
            ct_long, ct_short = line.split('|',1)
            self.ct[ct_short.strip()] = ct_long.strip()

    def mapper(self, _, line):
        cust,price,country = line.split('|',2)

        # In this case, we are left joining with the transactions file on t
        # Therefore, we will output 9 rows regardless of how many matches s

        if country in self.ct:
            self.left += 1
            yield None, country+", "+self.ct[country]+", "+cust+", "+price
        else:
            self.nomatch += 1
            yield None, country+", NULL, "+cust+", "+price

        # Mapper_Final will output the total joined rows
    def mapper_final(self):
        yield None, "left-joined "+str(self.left)+" row(s)."
        yield None, "No left match for "+str(self.nomatch)+" row(s)."
        yield None, "Total Rows: "+str(self.left + self.nomatch)

if __name__ == '__main__':
    leftJoin.run()

```

Overwriting leftHashJoin.py

For the RIGHT join, we output all of the rows from the COUNTRIES file, whether or not there is a match in the TRANSACTIONS file. If there is a match, we join them.


```

In [15]: %%writefile rightHashJoin.py
#!/usr/bin/env python

from mrjob.job import MRJob
from mrjob.step import MRStep
from mrjob.compat import jobconf_from_env
import os

class rightJoin(MRJob):

    def steps(self):
        return [
            MRStep(
                mapper_init = self.mapper_init,
                mapper = self.mapper,
                mapper_final = self.mapper_final
            )]

    def mapper_init(self):
        countries = open(str(os.path.dirname(os.path.realpath(__file__)))+"",
        self.right=0
        self.nomatch=0
        self.ct = {}
        self.matched_ct={}
        for line in countries:
            ct_long, ct_short = line.split('|',1)
            self.ct[ct_short.strip()] = ct_long.strip()
#            self.matched_ct[ct_short.strip()] = ct_long.strip()

    def mapper(self, _, line):
        cust,price,country = line.split('|',2)

        # In this case, we are right joining with the transactions file on t
        # Therefore, we will output all of the rows from the countries table
        # from the transactions table that have a country match

        if country in self.ct:
            self.right += 1
            yield None, country+", "+self.ct[country]+", "+cust+", "+price
            self.matched_ct[country] = self.ct[country]

# Mapper_Final will output the total joined rows
    def mapper_final(self):
        for key,value in self.ct.iteritems():
            if key not in self.matched_ct:
                self.nomatch += 1
                yield None, key+", "+value+", NULL, NULL"
        yield None, "right-joined "+str(self.right)+" row(s). "
        yield None, "No right match for "+str(self.nomatch)+" row(s). "
        yield None, "Total Rows: "+str(self.right + self.nomatch)

if __name__ == '__main__':
    rightJoin.run()

```

Overwriting rightHashJoin.py

For the INNER join, we output all of the records that match between the two files.

```
In [16]: %%writefile innerHashJoin.py
#!/usr/bin/env python

from mrjob.job import MRJob
from mrjob.step import MRStep
from mrjob.compat import jobconf_from_env
import os

class innerJoin(MRJob):

    def steps(self):
        countries=[]
        return [
            MRStep(
                mapper_init = self.mapper_init,
                mapper = self.mapper,
                mapper_final = self.mapper_final
            )]

    def mapper_init(self):
        countries = open(str(os.path.dirname(os.path.realpath(__file__)))+"",
self.inner=0
self.ct = {}
for line in countries:
    ct_long, ct_short = line.split('|',1)
    self.ct[ct_short.strip()] = ct_long.strip()

    def mapper(self, _, line):
        cust,price,country = line.split('|',2)
        if country in self.ct:
            self.inner += 1
            yield None, self.ct[country]+" "+cust+" "+price

    def mapper_final(self):
        yield None, "inner-joined "+str(self.inner)+" rows."

if __name__ == '__main__':
    innerJoin.run()
```

Overwriting innerHashJoin.py

```
In [17]: !chmod a+x leftHashJoin.py
!./leftHashJoin.py --jobconf mapred.map.tasks=1 transactions.txt
```

```
No configs found; falling back on auto-configuration
Creating temp directory /tmp/leftHashJoin.root.20170214.034404.240845
Running step 1 of 1...
Streaming final output from /tmp/leftHashJoin.root.20170214.034404.240845/output...
null      "US,United States,Alice Bob,$10"
null      "CA,Canada,Sam Sneed,$1"
null      "CA,Canada,Jon Sneed,$20"
null      "UK,United Kingdom,Arnold Wesise,$400"
null      "US,United States,Henry Bob,$2"
null      "CA,Canada,Yo Yo Ma,$2"
null      "CA,Canada,Jon York,$44"
null      "UK,United Kingdom,Alex Ball,$5"
null      "JA,NULL,Jim Davis,$66"
null      "left-joined 8 row(s)."
null      "No left match for 1 row(s)."
null      "Total Rows: 9"
Removing temp directory /tmp/leftHashJoin.root.20170214.034404.240845...
```

```
In [18]: !chmod a+x rightHashJoin.py
!./rightHashJoin.py --jobconf mapred.map.tasks=1 transactions.txt
```

```
No configs found; falling back on auto-configuration
Creating temp directory /tmp/rightHashJoin.root.20170214.034405.650385
Running step 1 of 1...
Streaming final output from /tmp/rightHashJoin.root.20170214.034405.650385/output...
null      "US,United States,Alice Bob,$10"
null      "CA,Canada,Sam Sneed,$1"
null      "CA,Canada,Jon Sneed,$20"
null      "UK,United Kingdom,Arnold Wesise,$400"
null      "US,United States,Henry Bob,$2"
null      "CA,Canada,Yo Yo Ma,$2"
null      "CA,Canada,Jon York,$44"
null      "UK,United Kingdom,Alex Ball,$5"
null      "IT,Italy,NULL,NULL"
null      "right-joined 8 row(s)."
null      "No right match for 1 row(s)."
null      "Total Rows: 9"
Removing temp directory /tmp/rightHashJoin.root.20170214.034405.650385...
```

```
In [19]: !chmod a+x innerHashJoin.py
!./innerHashJoin.py --jobconf mapred.map.tasks=1 transactions.txt
```

```
No configs found; falling back on auto-configuration
Creating temp directory /tmp/innerHashJoin.root.20170214.034407.159333
Running step 1 of 1...
Streaming final output from /tmp/innerHashJoin.root.20170214.034407.159333/output...
null      "United States,Alice Bob,$10"
null      "Canada,Sam Sneed,$1"
null      "Canada,Jon Sneed,$20"
null      "United Kingdom,Arnold Wesise,$400"
null      "United States,Henry Bob,$2"
null      "Canada,Yo Yo Ma,$2"
null      "Canada,Jon York,$44"
null      "United Kingdom,Alex Ball,$5"
null      "inner-joined 8 rows."
Removing temp directory /tmp/innerHashJoin.root.20170214.034407.159333...
```

3. HW5.2.1 (OPTIONAL) Almost stateless reducer-side join

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The following MRJob code, implements a reduce-side join for an inner join. The reducer is almost stateless, i.e., uses as little memory as possible. Use the tables from HW5.2 for this HW and join based on the country code (third column of the transactions table and the second column of the Countries table perform. Perform an left, right, inner joins using the code provided below and report the number of rows resulting from:

- (1) Left joining Table Left with Table Right
- (2) Right joining Table Left with Table Right
- (3) Inner joining Table Left with Table Right

Again make smart decisions about which table should be the left table (i.e., crosscheck the code).

Some notes on the code Here, the mapper receives its set of input splits either from the transaction table or from the countries table and makes the appropriate transformations: splitting the line into fields, and emitting a key/value. The key is the join key - in this case, the country code field of both sets of records. The mapper knows which file and type of record it is receiving based on the length of the fields. The records it emits contain the join field as the key, which acts as the partitioning key; We use the SORT_VALUES option, which ensures the values are sorted as well. Then, we employ a trick to ensure that for each join key, country records are seen always before transaction records. We achieve this by adding an arbitrary key to the front of the value: 'A' for countries, 'B' for customers. This makes countries sort before customers for each and every join/partition key. After that trick, the join is simply a matter of storing countries ('A' records) and crossing this array with each customer record.

```
In [ ]: import sys, os, re
        from mrjob.job import MRJob

        class MRJoin(MRJob):

            # Performs secondary sort
            SORT_VALUES = True

            def mapper(self, _, line):
                splits = line.rstrip("\n").split("|")

                if len(splits) == 2: # country data
                    symbol = 'A' # make country sort before transaction data
                    country2digit = splits[1]
                    yield country2digit, [symbol, splits]
                else: # person data
                    symbol = 'B'
                    country2digit = splits[2]
                    yield country2digit, [symbol, splits]

            def reducer(self, key, values):
                countries = [] # should come first, as they are sorted on artificial key
                for value in values:
                    if value[0] == 'A':
                        countries.append(value)
                    if value[0] == 'B':
                        for country in countries:
                            yield key, country[1:] + value[1:]

        if __name__ == '__main__':
            MRJoin.run()
```

5.3 Pairwise similarity - PHASE 1

In this part of the assignment we will focus on developing methods for detecting synonyms, using the Google 5-grams dataset. To accomplish this you must script two main tasks using MRJob:

(1) Using the systems tests data sets, write mrjob code to build the stripes

(2) Write mrjob code to build an inverted index from the stripes

(3) Using two (symmetric) comparison methods of your choice (e.g., correlations, distances, similarities), pairwise compare all stripes (vectors), and output to a file.

==Design notes for (1)==

For this task you will be able to modify the pattern we used in HW 3.2 (feel free to use the solution as reference). To total the word counts across the n-grams, output the support from the mappers using the total order inversion pattern:

<*word,count>

to ensure that the support arrives before the cooccurrences.

In addition to ensuring the determination of the total word counts, the mapper must also output co-occurrence counts for the pairs of words inside of each n-gram. Treat these words as a basket, as we have in HW 3, but count all stripes or pairs in both orders, i.e., count both orderings: (word1,word2), and (word2,word1), to preserve symmetry in our output for (2).

==Design notes for (3)==

For this task you will have to determine a method of comparison. Here are a few that you might consider:

- Jaccard
- Cosine similarity
- Spearman correlation
- Euclidean distance
- Taxicab (Manhattan) distance
- Shortest path graph distance (a graph, because our data is symmetric!)
- Pearson correlation
- Kendall correlation ...

However, be cautioned that some comparison methods are more difficult to parallelize than others, and do not perform more associations than is necessary, since your choice of association will be symmetric.

Please use the inverted index (discussed in live session #5) based pattern to compute the pairwise (term-by-term) similarity matrix.

Type *Markdown* and LaTeX: α^2

```

In [291]: %%writefile buildStripes.py
          #!/~/anaconda2/bin/python
          # -*- coding: utf-8 -*-

          from __future__ import division
          import re
          import mrjob
          import json
          from mrjob.protocol import RawProtocol
          from mrjob.job import MRJob
          from mrjob.step import MRStep

          class MRbuildStripes(MRJob):

              def mapper(self, _, line):
                  ngram,count,page_count,book_count = line.split('\t',3)
                  ngram = ngram.strip()
                  count = int(count)

                  # lowercase and parse out each word
                  words = ngram.lower().split()

                  d = {}

                  # Create a dictionary within a dictionary
                  # For example: d[biography] = {"a": 92, "of": 92, "george": 92, "ge

                  for term in sorted(words):
                      if term not in d.keys():
                          d[term] = {}

                      for term2 in sorted(words):
                          if term != term2:
                              if term2 in d[term]:
                                  d[term][term2] += count
                              else:
                                  d[term][term2] = count

                  # iterate through the dictionary and yield the top level term, the s
                  # Example: "biography, (general, 92)"

                  for k,v in d.iteritems():
                      for k2,v2 in d[k].iteritems():
                          yield k, (k2, v2)

              def reducer(self, key, line):

                  red_d = {}
                  term1 = key

                  # Combine the various term cooccurrence counts into a single dictio

                  for term,count in line:
                      count = int(count)
                      term2 = term

```

```
        if term1 not in red_d.keys():
            red_d[term1] = {}
        if term2 in red_d[term1]:
            red_d[term1][term2] += count
        else:
            red_d[term1][term2] = count

    for k,v in red_d.iteritems():
        yield k,v

#END SUDENT CODE531_STRIPES
if __name__ == '__main__':
    MRbuildStripes.run()
```

Overwriting buildStripes.py

```

In [292]: %%writefile invertedIndex.py
          #!/~/anaconda2/bin/python
          # -*- coding: utf-8 -*-

          from __future__ import division
          import collections
          import re
          import json
          import math
          # import numpy as np
          import itertools
          import mrjob
          from mrjob.protocol import RawProtocol
          from mrjob.job import MRJob
          from mrjob.step import MRStep
          import ast

          class MRinvertedIndex(MRJob):

          #START STUDENT CODE531_INV_INDEX

              def mapper(self,_,line):

                  line = line.strip()
                  key_term, words = line.split("\t")

                  # 'words' are coming in with the structure of a dictionary, but for
                  # ast.literal_eval converts it to the dictionary that it should be
                  words = ast.literal_eval(words)
                  _len = len(words)

                  # for each word, output the cooccurring terms and the number of assoc
                  for word in words:
                      yield word, (key_term, _len)

              def reducer(self,key,value):

                  d = collections.defaultdict(list)
                  for v in value:
                      d[key].append(v)
                  yield key,d[key]

          #END STUDENT CODE531_INV_INDEX

          if __name__ == '__main__':
              MRinvertedIndex.run()

```

Overwriting invertedIndex.py


```

In [311]: %%writefile similarity.py
          #!/~/anaconda2/bin/python
          # -*- coding: utf-8 -*-

          from __future__ import division
          import collections
          import re
          import json
          import math
          # import numpy as np
          import itertools
          import mrjob
          from mrjob.protocol import RawProtocol
          from mrjob.job import MRJob
          from mrjob.step import MRStep

          class MRsimilarity(MRJob):

              #START SUDENT CODE531_SIMILARITY

              MRJob.SORT_VALUES = True
              def steps(self):

                  JOBCONF_STEP1 = {}
                  JOBCONF_STEP2 = {
                      ##### IMPORTANT: THIS WILL HAVE NO EFFECT IN -r local MODE. MR
                      'mapreduce.job.output.key.comparator.class': 'org.apache.hadoop.
                      'mapreduce.partition.keycomparator.options': '-k1,lnr',

                  }

                  return [MRStep(jobconf=JOBCONF_STEP1,
                                mapper=self.mapper_pair_sim,
                                reducer=self.reducer_pair_sim)
                          ,
                          MRStep(jobconf=JOBCONF_STEP2,
                                mapper=None,
                                reducer=self.reducer_sort)
                          ]

              def mapper_pair_sim(self,_,line):
                  line = line.strip()
                  term,coterm = line.split("\t")
                  coterm = json.loads(coterm)

                  X = map(lambda x: x[0]+"."+str(x[1]) , coterm)

                  # taking advantage of symmetry, output only (a,b), but not (b,a)
                  # 'set' will output only the unique occurrences
                  for subset in itertools.combinations(sorted(set(X)), 2):
                      yield subset[0]+"."+subset[1], 1

              def reducer_pair_sim(self,key,value):
                  Doc1, Doc1_len, Doc2, Doc2_len = key.split(".")
                  doc1_len = int(Doc1_len)

```

```

doc2_len = int(Doc2_len)
t = sum(value)

# calculate the similarity values
jaccard = t / ( doc1_len + doc2_len - t )
cosine = t * ((1/math.sqrt(doc1_len)) * (1/math.sqrt(doc2_len)))
dice = (2*t) / (doc1_len + doc2_len)
overlap = t / min(doc1_len, doc2_len)

# Average the 4 similarities
avg = sum([jaccard,cosine,dice,overlap]) / 4

yield [avg,jaccard,cosine,overlap,dice], (Doc1+" - "+Doc2)

def reducer_sort(self,key,value):
    for v in value:
        yield key,v

#END STUDENT CODE531_SIMILARITY

if __name__ == '__main__':
    MRsimilarity.run()

```

Overwriting similarity.py

HW5.3.1 Run Systems tests locally on small datasets (PHASE1)

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Complete 5.3 and systems test using the below test datasets. Phase 2 will focus on the entire Ngram dataset.

To help you through these tasks please verify that your code gives the results below (for stripes, inverted index, and pairwise similarities).

Test datasets:

- googlebooks-eng-all-5gram-20090715-0-filtered.txt [see below]
- atlas-boon-test [see below]
- stripe-docs-test [see below]

A large subset of the Google n-grams dataset

<https://aws.amazon.com/datasets/google-books-ngrams/>
[\(https://aws.amazon.com/datasets/google-books-ngrams/\)](https://aws.amazon.com/datasets/google-books-ngrams/)

which we have placed in a bucket/folder on Dropbox and on s3:

<https://www.dropbox.com/sh/tmqpc4o0xswhkvz/AACUifrl6wrMrIK6a3X3lZ9Ea?dl=0>
[\(https://www.dropbox.com/sh/tmqpc4o0xswhkvz/AACUifrl6wrMrIK6a3X3lZ9Ea?dl=0\)](https://www.dropbox.com/sh/tmqpc4o0xswhkvz/AACUifrl6wrMrIK6a3X3lZ9Ea?dl=0)

s3://filtered-5grams/

In particular, this bucket contains (~200) files (10Meg each) in the format:

```
(ngram) \t (count) \t (pages_count) \t (books_count)
```

The next cell shows the first 10 lines of the googlebooks-eng-all-5gram-20090715-0-filtered.txt file.

DISCLAIMER: Each record is already a 5-gram. In real life, we would calculate the stripes cooccurrence data from the raw text by windowing over the raw text and not from the 5-gram preprocessed data (as we are doing here). Calculating pairs on this 5-gram is a little corrupt as we will be double counting cooccurrences. Having said that this exercise can still pull out some similar terms.

1: unit/systems first-10-lines

```
In [294]: %%writefile googlebooks-eng-all-5gram-20090715-0-filtered-first-10-lines.txt
A BILL FOR ESTABLISHING RELIGIOUS    59  59  54
A Biography of General George    92  90  74
A Case Study in Government    102 102  78
A Case Study of Female    447 447 327
A Case Study of Limited    55  55  43
A Child's Christmas in Wales    1099    1061    866
A Circumstantial Narrative of the    62  62  50
A City by the Sea    62  60  49
A Collection of Fairy Tales    123 117  80
A Collection of Forms of    116 103  82
```

Overwriting googlebooks-eng-all-5gram-20090715-0-filtered-first-10-lines.txt

2: unit/systems atlas-boon

```
In [295]: %%writefile atlas-boon-systems-test.txt
atlas boon    50  50  50
boon cava dipped    10  10  10
atlas dipped    15  15  15
```

Overwriting atlas-boon-systems-test.txt

3: unit/systems stripe-docs-test

Three terms, A,B,C and their corresponding stripe-docs of co-occurring terms

- DocA {X:20, Y:30, Z:5}
- DocB {X:100, Y:20}
- DocC {M:5, N:20, Z:5}

(1) build stripes for all the test data sets - run the commands and insure that your output matches the output below

```
In [296]: #####
# Make Stripes from ngrams for systems test 1
#####

!hdfs dfs rm --recursive systems_test_stripes_1
!python buildStripes.py -r local googlebooks-eng-all-5gram-20090715-0-filter
```

rm: Unknown command

Did you mean -rm? This command begins with a dash.

No configs found; falling back on auto-configuration

Creating temp directory /tmp/buildStripes.root.20170213.072821.757803

Running step 1 of 1...

Streaming final output from /tmp/buildStripes.root.20170213.072821.757803/output...

Removing temp directory /tmp/buildStripes.root.20170213.072821.757803...

```
In [297]: !cat systems_test_stripes_1
```

```
"a"      {"limited":55,"sea":62,"general":92,"female":447,"in":1201,"religious":59,"george":92,"biography":92,"city":62,"for":59,"tales":123,"government":102,"the":124,"forms":116,"wales":1099,"christmas":1099,"child's":1099,"collection":239,"by":62,"case":604,"circumstantial":62,"of":1011,"study":604,"bill":59,"establishing":59,"narrative":62,"fairy":123}
"bill"   {"a":59,"religious":59,"for":59,"establishing":59}
"biography" {"a":92,"of":92,"george":92,"general":92}
"by"     {"a":62,"city":62,"the":62,"sea":62}
"case"   {"a":604,"limited":55,"government":102,"of":502,"study":604,"female":447,"in":102}
"child's" {"a":1099,"wales":1099,"christmas":1099,"in":1099}
"christmas" {"a":1099,"wales":1099,"in":1099,"child's":1099}
"circumstantial" {"a":62,"of":62,"the":62,"narrative":62}
"city"    {"a":62,"the":62,"by":62,"sea":62}
"collection" {"a":239,"forms":116,"fairy":123,"tales":123,"of":355}
"establishing" {"a":59,"bill":59,"religious":59,"for":59}
"fairy"   {"a":123,"of":123,"tales":123,"collection":123}
"female"  {"a":447,"case":447,"study":447,"of":447}
"for"     {"a":59,"bill":59,"religious":59,"establishing":59}
"forms"   {"a":116,"of":232,"collection":116}
"general" {"a":92,"of":92,"george":92,"biography":92}
"george"  {"a":92,"of":92,"biography":92,"general":92}
"government" {"a":102,"case":102,"study":102,"in":102}
"in"      {"a":1201,"case":102,"government":102,"study":102,"child's":1099,"wales":1099,"christmas":1099}
"limited"  {"a":55,"case":55,"study":55,"of":55}
"narrative" {"a":62,"of":62,"the":62,"circumstantial":62}
"of"      {"a":1011,"case":502,"circumstantial":62,"limited":55,"the":62,"study":502,"collection":355,"general":92,"forms":232,"tales":123,"female":447,"narrative":62,"fairy":123,"george":92,"biography":92}
"religious" {"a":59,"bill":59,"for":59,"establishing":59}
"sea"     {"a":62,"city":62,"the":62,"by":62}
"study"   {"a":604,"case":604,"limited":55,"of":502,"government":102,"female":447,"in":102}
"tales"   {"a":123,"of":123,"fairy":123,"collection":123}
"the"     {"a":124,"city":62,"circumstantial":62,"of":62,"sea":62,"narrative":62,"by":62}
"wales"   {"a":1099,"in":1099,"christmas":1099,"child's":1099}
```

```

"a"      {"limited": 55, "sea": 62, "general": 92, "female": 447, "in": 1201, "religious": 59, "george": 92, "biography": 92, "city": 62, "for": 59, "tales": 123, "child's": 1099, "forms": 116, "wales": 1099, "christmas": 1099, "government": 102, "collection": 239, "by": 62, "case": 604, "circumstantial": 62, "fairy": 123, "of": 1011, "study": 604, "bill": 59, "establishing": 59, "narrative": 62, "the": 124}
"bill"    {"a": 59, "religious": 59, "for": 59, "establishing": 59}
"biography" {"a": 92, "of": 92, "george": 92, "general": 92}
"by"      {"a": 62, "city": 62, "the": 62, "sea": 62}
"case"    {"a": 604, "limited": 55, "government": 102, "of": 502, "study": 604, "female": 447, "in": 102}
"child's"  {"a": 1099, "wales": 1099, "christmas": 1099, "in": 1099}
"christmas" {"a": 1099, "wales": 1099, "in": 1099, "child's": 1099}
"circumstantial" {"a": 62, "of": 62, "the": 62, "narrative": 62}
"city"     {"a": 62, "the": 62, "by": 62, "sea": 62}
"collection" {"a": 239, "of": 355, "fairy": 123, "tales": 123, "forms": 116}
"establishing" {"a": 59, "bill": 59, "religious": 59, "for": 59}
"fairy"     {"a": 123, "of": 123, "tales": 123, "collection": 123}
"female"    {"a": 447, "case": 447, "study": 447, "of": 447}
"for"       {"a": 59, "bill": 59, "religious": 59, "establishing": 59}
"forms"     {"a": 116, "of": 232, "collection": 116}
"general"   {"a": 92, "of": 92, "george": 92, "biography": 92}
"george"    {"a": 92, "of": 92, "biography": 92, "general": 92}
"government" {"a": 102, "case": 102, "study": 102, "in": 102}
"in"        {"a": 1201, "case": 102, "government": 102, "study": 102, "child's": 1099, "wales": 1099, "christmas": 1099}
"limited"    {"a": 55, "case": 55, "study": 55, "of": 55}
"narrative" {"a": 62, "of": 62, "the": 62, "circumstantial": 62}
"of"        {"a": 1127, "case": 502, "circumstantial": 62, "george": 92, "limited": 55, "tales": 123, "collection": 471, "general": 92, "forms": 348, "female": 447, "narrative": 62, "study": 502, "fairy": 123, "the": 62, "biography": 92}
"religious"  {"a": 59, "bill": 59, "for": 59, "establishing": 59}
"sea"       {"a": 62, "city": 62, "the": 62, "by": 62}
"study"     {"a": 604, "case": 604, "limited": 55, "government": 102, "of": 502, "female": 447, "in": 102}
"tales"     {"a": 123, "of": 123, "fairy": 123, "collection": 123}
"the"       {"a": 124, "city": 62, "circumstantial": 62, "of": 62, "sea": 62, "narrative": 62, "by": 62}
"wales"     {"a": 1099, "in": 1099, "christmas": 1099, "child's": 1099}

```

```
In [298]: #####
# Make Stripes from ngrams for systems test 2
#####

!hdfs dfs rm --recursive systems_test_stripes_2
!python buildStripes.py -r local atlas-boon-systems-test.txt > systems_test_
```

```
rm: Unknown command
Did you mean -rm? This command begins with a dash.
No configs found; falling back on auto-configuration
Creating temp directory /tmp/buildStripes.root.20170213.072836.216996
Running step 1 of 1...
Streaming final output from /tmp/buildStripes.root.20170213.072836.216996/output...
Removing temp directory /tmp/buildStripes.root.20170213.072836.216996...
```

```
In [299]: !cat systems_test_stripes_2
```

```
"atlas" {"dipped":15,"boon":50}
"boon"  {"atlas":50,"dipped":10,"cava":10}
"cava"  {"dipped":10,"boon":10}
"dipped" {"atlas":15,"boon":10,"cava":10}
```

```
"atlas" {"dipped": 15, "boon": 50}
"boon"  {"atlas": 50, "dipped": 10, "cava": 10}
"cava"  {"dipped": 10, "boon": 10}
"dipped" {"atlas": 15, "boon": 10, "cava": 10}
```

```
In [300]: #####
# Stripes for systems test 3 (given, no need to build stripes)
#####

with open("systems_test_stripes_3", "w") as f:
    f.writelines([
        'DocA\t{"X":20, "Y":30, "Z":5}\n',
        'DocB\t{"X":100, "Y":20}\n',
        'DocC\t{"M":5, "N":20, "Z":5, "Y":1}\n'
    ])
!cat systems_test_stripes_3
```

```
"DocA" {"X":20, "Y":30, "Z":5}
"DocB" {"X":100, "Y":20}
"DocC" {"M":5, "N":20, "Z":5, "Y":1}
```

(2) Build Inverted Index - run the commands and insure that your output matches the output below

```
In [301]: !python invertedIndex.py -r local systems_test_stripes_1 > systems_test_index_1
```

```
No configs found; falling back on auto-configuration
Creating temp directory /tmp/invertedIndex.root.20170213.072857.238594
Running step 1 of 1...
Streaming final output from /tmp/invertedIndex.root.20170213.072857.238594/output...
Removing temp directory /tmp/invertedIndex.root.20170213.072857.238594...
```

```
In [302]: !python invertedIndex.py -r local systems_test_stripes_2 > systems_test_index_2
```

```
No configs found; falling back on auto-configuration
Creating temp directory /tmp/invertedIndex.root.20170213.072900.428320
Running step 1 of 1...
Streaming final output from /tmp/invertedIndex.root.20170213.072900.428320/output...
Removing temp directory /tmp/invertedIndex.root.20170213.072900.428320...
```

```
In [303]: !python invertedIndex.py -r local systems_test_stripes_3 > systems_test_index_3
```

```
No configs found; falling back on auto-configuration
Creating temp directory /tmp/invertedIndex.root.20170213.072903.458618
Running step 1 of 1...
Streaming final output from /tmp/invertedIndex.root.20170213.072903.458618/output...
Removing temp directory /tmp/invertedIndex.root.20170213.072903.458618...
```

```
In [306]: #####
# Pretty print systems tests for generating Inverted Index
#####

import json

for i in range(1,4):
    print "-"*100
    print "Systems test ",i," - Inverted Index"
    print "-"*100
    with open("systems_test_index_"+str(i),"r") as f:
        lines = f.readlines()
        for line in lines:
            line = line.strip()
            word,stripe = line.split("\t")
            stripe = json.loads(stripe)
            stripe.extend([["", ""] for _ in xrange(3 - len(stripe))])
            print "{0:>16} | {1:>16} | {2:>16} | {3:>16}".format(
                (word), stripe[0][0]+" "+str(stripe[0][1]), stripe[1][0]+" "
```

Systems test 1 - Inverted Index

"a"		"bill" 4		"biography" 4		"by" 4
"bill"		"a" 27		"establishing" 4		"for" 4
"biography"		"a" 27		"general" 4		"george" 4
"by"		"a" 27		"city" 4		"sea" 4
"case"		"a" 27		"female" 4		"government" 4
"child's"		"a" 27		"christmas" 4		"in" 7
"christmas"		"a" 27		"child's" 4		"in" 7
"circumstantial"		"a" 27		"narrative" 4		"of" 15
"city"		"a" 27		"by" 4		"sea" 4
"collection"		"a" 27		"fairy" 4		"forms" 3
"establishing"		"a" 27		"bill" 4		"for" 4
"fairy"		"a" 27		"collection" 5		"of" 15
"female"		"a" 27		"case" 7		"of" 15
"for"		"a" 27		"bill" 4		"establishing" 4
"forms"		"a" 27		"collection" 5		"of" 15
"general"		"a" 27		"biography" 4		"george" 4
"george"		"a" 27		"biography" 4		"general" 4
"government"		"a" 27		"case" 7		"in" 7
"in"		"a" 27		"case" 7		"child's" 4
"limited"		"a" 27		"case" 7		"of" 15
"narrative"		"a" 27		"circumstantial" 4		"of" 15
"of"		"a" 27		"biography" 4		"case" 7
"religious"		"a" 27		"bill" 4		"establishing" 4
"sea"		"a" 27		"by" 4		"city" 4
"study"		"a" 27		"case" 7		"female" 4
"tales"		"a" 27		"collection" 5		"fairy" 4
"the"		"a" 27		"by" 4		"circumstantial" 4
"wales"		"a" 27		"child's" 4		"christmas" 4

Systems test 2 - Inverted Index

"atlas"		"boon" 3		"dipped" 3		
"boon"		"atlas" 2		"cava" 2		"dipped" 3
"cava"		"boon" 3		"dipped" 3		
"dipped"		"atlas" 2		"boon" 3		"cava" 2

Systems test 3 - Inverted Index

"M"		"DocC" 4				
"N"		"DocC" 4				
"X"		"DocA" 3		"DocB" 2		
"Y"		"DocA" 3		"DocB" 2		"DocC" 4
"Z"		"DocA" 3		"DocC" 4		

Inverted Index

In []:

Systems test 1 - Inverted Index

"a"	bill 4	biography 4	by 4
"bill"	a 27	establishing 4	for 4
"biography"	a 27	general 4	george 4
"by"	a 27	city 4	sea 4
"case"	a 27	female 4	government 4
"child's"	a 27	christmas 4	in 7
"christmas"	a 27	child's 4	in 7
"circumstantial"	a 27	narrative 4	of 15
"city"	a 27	by 4	sea 4
"collection"	a 27	fairy 4	forms 3
"establishing"	a 27	bill 4	for 4
"fairy"	a 27	collection 5	of 15
"female"	a 27	case 7	of 15
"for"	a 27	bill 4	establishing 4
"forms"	a 27	collection 5	of 15
"general"	a 27	biography 4	george 4
"george"	a 27	biography 4	general 4
"government"	a 27	case 7	in 7
"in"	a 27	case 7	child's 4
"limited"	a 27	case 7	of 15
"narrative"	a 27	circumstantial 4	of 15
"of"	a 27	biography 4	case 7
"religious"	a 27	bill 4	establishing 4
"sea"	a 27	by 4	city 4
"study"	a 27	case 7	female 4
"tales"	a 27	collection 5	fairy 4
"the"	a 27	by 4	circumstantial 4
"wales"	a 27	child's 4	christmas 4

Systems test 2 - Inverted Index

"atlas"	boon 3	dipped 3	
"boon"	atlas 2	cava 2	dipped 3
"cava"	boon 3	dipped 3	
"dipped"	atlas 2	boon 3	cava 2

Systems test 3 - Inverted Index

"M"	DocC 4		
"N"	DocC 4		
"X"	DocA 3	DocB 2	
"Y"	DocA 3	DocB 2	DocC 4
"Z"	DocA 3	DocC 4	

(3) Calculate similarities - run the commands and insure that your output matches the output below

NOTE: you must run in hadoop mode to generate sorted similarities

```
In [307]: !python similarity.py -r hadoop systems_test_index_1 > systems_test_similarity_1
```

```
No configs found; falling back on auto-configuration
Looking for hadoop binary in $PATH...
Found hadoop binary: /usr/bin/hadoop
Using Hadoop version 2.6.0
Looking for Hadoop streaming jar in /home/hadoop/contrib...
Looking for Hadoop streaming jar in /usr/lib/hadoop-mapreduce...
Found Hadoop streaming jar: /usr/lib/hadoop-mapreduce/hadoop-streaming.jar
Creating temp directory /tmp/similarity.root.20170213.073031.946921
Copying local files to hdfs:///user/root/tmp/mrjob/similarity.root.20170213.073031.946921/files/...
Detected hadoop configuration property names that do not match hadoop version 2.6.0:
The have been translated as follows
  mapred.output.key.comparator.class: mapreduce.job.output.key.comparator.class
  mapred.text.key.comparator.options: mapreduce.partition.keycomparator.options
  mapred.text.key.partitioner.options: mapreduce.partition.keypartitioner.options
```

```
In [308]: !python similarity.py -r hadoop systems_test_index_2 > systems_test_similarity_2
```

```
No configs found; falling back on auto-configuration
Looking for hadoop binary in $PATH...
Found hadoop binary: /usr/bin/hadoop
Using Hadoop version 2.6.0
Looking for Hadoop streaming jar in /home/hadoop/contrib...
Looking for Hadoop streaming jar in /usr/lib/hadoop-mapreduce...
Found Hadoop streaming jar: /usr/lib/hadoop-mapreduce/hadoop-streaming.jar
Creating temp directory /tmp/similarity.root.20170213.073151.795311
Copying local files to hdfs:///user/root/tmp/mrjob/similarity.root.20170213.073151.795311/files/...
Detected hadoop configuration property names that do not match hadoop version 2.6.0:
The have been translated as follows
  mapred.output.key.comparator.class: mapreduce.job.output.key.comparator.class
  mapred.text.key.comparator.options: mapreduce.partition.keycomparator.options
  mapred.text.key.partitioner.options: mapreduce.partition.keypartitioner.options
```

```
In [309]: !python similarity.py -r hadoop systems_test_index_3 > systems_test_similarity
```

```
No configs found; falling back on auto-configuration
Looking for hadoop binary in $PATH...
Found hadoop binary: /usr/bin/hadoop
Using Hadoop version 2.6.0
Looking for Hadoop streaming jar in /home/hadoop/contrib...
Looking for Hadoop streaming jar in /usr/lib/hadoop-mapreduce...
Found Hadoop streaming jar: /usr/lib/hadoop-mapreduce/hadoop-streaming.jar
Creating temp directory /tmp/similarity.root.20170213.073316.826528
Copying local files to hdfs:///user/root/tmp/mrjob/similarity.root.20170213.073316.826528/files/...
Detected hadoop configuration property names that do not match hadoop version 2.6.0:
The have been translated as follows
  mapred.output.key.comparator.class: mapreduce.job.output.key.comparator.class
mapred.text.key.comparator.options: mapreduce.partition.keycomparator.options
mapred.text.key.partitionner.options: mapreduce.partition.keypartitioner.options
```


0.250000		0.181818			
0.223214		"religious" - "tales"		0.250000	0.142857
0.250000		0.250000			
0.180200		"religious" - "study"		0.188982	0.100000
0.250000		0.181818			
0.223214		"religious" - "sea"		0.250000	0.142857
0.250000		0.250000			
0.134980		"of" - "wales"		0.129099	0.055556
0.250000		0.105263			
0.287991		"of" - "the"		0.292770	0.157895
0.428571		0.272727			
0.410147		"of" - "tales"		0.387298	0.187500
0.750000		0.315789			
0.386912		"of" - "study"		0.390360	0.222222
0.571429		0.363636			
0.271593		"of" - "sea"		0.258199	0.117647
0.500000		0.210526			
0.134980		"of" - "religious"		0.129099	0.055556
0.250000		0.105263			
0.223214		"narrative" - "wales"		0.250000	0.142857
0.250000		0.250000			
0.559350		"narrative" - "the"		0.566947	0.375000
0.750000		0.545455			
0.458333		"narrative" - "tales"		0.500000	0.333333
0.500000		0.500000			
0.365956		"narrative" - "study"		0.377964	0.222222
0.500000		0.363636			
0.458333		"narrative" - "sea"		0.500000	0.333333
0.500000		0.500000			
0.223214		"narrative" - "religious"		0.250000	0.142
857		0.250000		0.250000	
0.410147		"narrative" - "of"		0.387298	0.187500
0.750000		0.315789			
0.223214		"limited" - "wales"		0.250000	0.142857
0.250000		0.250000			
0.365956		"limited" - "the"		0.377964	0.222222
0.500000		0.363636			
0.458333		"limited" - "tales"		0.500000	0.333333
0.500000		0.500000			
0.559350		"limited" - "study"		0.566947	0.375000
0.750000		0.545455			
0.223214		"limited" - "sea"		0.250000	0.142857
0.250000		0.250000			
0.223214		"limited" - "religious"		0.250000	0.14285
7		0.250000		0.250000	
0.410147		"limited" - "of"		0.387298	0.187500
0.750000		0.315789			
0.458333		"limited" - "narrative"		0.500000	0.33333
3		0.500000		0.500000	
0.559350		"in" - "wales"		0.566947	0.375000
0.750000		0.545455			
0.126374		"in" - "the"		0.142857	0.076923
0.142857		0.142857			
0.180200		"in" - "tales"		0.188982	0.100000
0.250000		0.181818			
0.389610		"in" - "study"		0.428571	0.272727
0.428571		0.428571			

	0.180200		"in" - "sea"		0.188982		0.100000	
0.250000		0.181818						
	0.180200		"in" - "religious"		0.188982		0.100000	
0.250000		0.181818						
	0.287991		"in" - "of"		0.292770		0.157895	
0.428571		0.272727						
	0.180200		"in" - "narrative"		0.188982		0.100000	
0.250000		0.181818						
	0.559350		"in" - "limited"		0.566947		0.375000	
0.750000		0.545455						
	0.458333		"government" - "wales"		0.500000		0.333333	
	0.500000		0.500000					
	0.180200		"government" - "the"		0.188982		0.100000	
0.250000		0.181818						
	0.223214		"government" - "tales"		0.250000		0.142857	
	0.250000		0.250000					
	0.559350		"government" - "study"		0.566947		0.375000	
	0.750000		0.545455					
	0.223214		"government" - "sea"		0.250000		0.142857	
0.250000		0.250000						
	0.223214		"government" - "religious"		0.250000		0.14	
2857		0.250000		0.250000				
	0.410147		"government" - "of"		0.387298		0.187500	
0.750000		0.315789						
	0.223214		"government" - "narrative"		0.250000		0.14	
2857		0.250000		0.250000				
	0.712500		"government" - "limited"		0.750000		0.6000	
00		0.750000		0.750000				
	0.559350		"government" - "in"		0.566947		0.375000	
0.750000		0.545455						
	0.223214		"george" - "wales"		0.250000		0.142857	
0.250000		0.250000						
	0.365956		"george" - "the"		0.377964		0.222222	
0.500000		0.363636						
	0.458333		"george" - "tales"		0.500000		0.333333	
0.500000		0.500000						
	0.365956		"george" - "study"		0.377964		0.222222	
0.500000		0.363636						
	0.223214		"george" - "sea"		0.250000		0.142857	
0.250000		0.250000						
	0.223214		"george" - "religious"		0.250000		0.142857	
	0.250000		0.250000					
	0.410147		"george" - "of"		0.387298		0.187500	
0.750000		0.315789						
	0.458333		"george" - "narrative"		0.500000		0.333333	
	0.500000		0.500000					
	0.458333		"george" - "limited"		0.500000		0.333333	
0.500000		0.500000						
	0.180200		"george" - "in"		0.188982		0.100000	
0.250000		0.181818						
	0.223214		"george" - "government"		0.250000		0.14285	
7		0.250000		0.250000				
	0.223214		"general" - "wales"		0.250000		0.142857	
0.250000		0.250000						
	0.365956		"general" - "the"		0.377964		0.222222	
0.500000		0.363636						
	0.458333		"general" - "tales"		0.500000		0.333333	

	0.500000		0.500000			
	0.365956		"general" - "study"		0.377964	0.222222
	0.500000		0.363636			
	0.223214		"general" - "sea"		0.250000	0.142857
	0.250000		0.250000			
	0.223214		"general" - "religious"		0.250000	0.14285
7		0.250000		0.250000		
	0.410147		"general" - "of"		0.387298	0.187500
	0.750000		0.315789			
	0.458333		"general" - "narrative"		0.500000	0.33333
3		0.500000		0.500000		
	0.458333		"general" - "limited"		0.500000	0.333333
		0.500000		0.500000		
	0.180200		"general" - "in"		0.188982	0.100000
	0.250000		0.181818			
	0.223214		"general" - "government"		0.250000	0.1428
57		0.250000		0.250000		
	0.712500		"general" - "george"		0.750000	0.600000
	0.750000		0.750000			
	0.268597		"forms" - "wales"		0.288675	0.166667
	0.333333		0.285714			
	0.438276		"forms" - "the"		0.436436	0.250000
	0.666667		0.400000			
	0.868292		"forms" - "tales"		0.866025	0.750000
	1.000000		0.857143			
	0.438276		"forms" - "study"		0.436436	0.250000
	0.666667		0.400000			
	0.268597		"forms" - "sea"		0.288675	0.166667
	0.333333		0.285714			
	0.268597		"forms" - "religious"		0.288675	0.166667
		0.333333		0.285714		
	0.328008		"forms" - "of"		0.298142	0.125000
	0.666667		0.222222			
	0.553861		"forms" - "narrative"		0.577350	0.400000
		0.666667		0.571429		
	0.553861		"forms" - "limited"		0.577350	0.400000
	0.666667		0.571429			
	0.215666		"forms" - "in"		0.218218	0.111111
	0.333333		0.200000			
	0.268597		"forms" - "government"		0.288675	0.166667
		0.333333		0.285714		
	0.553861		"forms" - "george"		0.577350	0.400000
	0.666667		0.571429			
	0.553861		"forms" - "general"		0.577350	0.400000
	0.666667		0.571429			
	0.223214		"for" - "wales"		0.250000	0.142857
	0.250000		0.250000			
	0.180200		"for" - "the"		0.188982	0.100000
	0.250000		0.181818			
	0.223214		"for" - "tales"		0.250000	0.142857
	0.250000		0.250000			
	0.180200		"for" - "study"		0.188982	0.100000
	0.250000		0.181818			
	0.223214		"for" - "sea"		0.250000	0.142857
	0.250000		0.250000			
	0.712500		"for" - "religious"		0.750000	0.600000
	0.750000		0.750000			

0.134980		"for" - "of"		0.129099		0.055556	
0.250000		0.105263					
0.223214		"for" - "narrative"		0.250000		0.142857	
0.250000		0.250000					
0.223214		"for" - "limited"		0.250000		0.142857	
0.250000		0.250000					
0.180200		"for" - "in"		0.188982		0.100000	
0.250000		0.181818					
0.223214		"for" - "government"		0.250000		0.142857	
0.250000		0.250000					
0.223214		"for" - "george"		0.250000		0.142857	
0.250000		0.250000					
0.223214		"for" - "general"		0.250000		0.142857	
0.250000		0.250000					
0.268597		"for" - "forms"		0.288675		0.166667	
0.333333		0.285714					
0.223214		"female" - "wales"		0.250000		0.142857	
0.250000		0.250000					
0.365956		"female" - "the"		0.377964		0.222222	
0.500000		0.363636					
0.458333		"female" - "tales"		0.500000		0.333333	
0.500000		0.500000					
0.559350		"female" - "study"		0.566947		0.375000	
0.750000		0.545455					
0.223214		"female" - "sea"		0.250000		0.142857	
0.250000		0.250000					
0.223214		"female" - "religious"		0.250000		0.142857	
0.250000		0.250000					
0.410147		"female" - "of"		0.387298		0.187500	
0.750000		0.315789					
0.458333		"female" - "narrative"		0.500000		0.333333	
0.500000		0.500000					
1.000000		"female" - "limited"		1.000000		1.000000	
1.000000		1.000000					
0.559350		"female" - "in"		0.566947		0.375000	
0.750000		0.545455					
0.712500		"female" - "government"		0.750000		0.600000	
0.750000		0.750000					
0.458333		"female" - "george"		0.500000		0.333333	
0.500000		0.500000					
0.458333		"female" - "general"		0.500000		0.333333	
0.500000		0.500000					
0.553861		"female" - "forms"		0.577350		0.400000	
0.666667		0.571429					
0.223214		"female" - "for"		0.250000		0.142857	
0.250000		0.250000					
0.223214		"fairy" - "wales"		0.250000		0.142857	
0.250000		0.250000					
0.365956		"fairy" - "the"		0.377964		0.222222	
0.500000		0.363636					
0.712500		"fairy" - "tales"		0.750000		0.600000	
0.750000		0.750000					
0.365956		"fairy" - "study"		0.377964		0.222222	
0.500000		0.363636					
0.223214		"fairy" - "sea"		0.250000		0.142857	
0.250000		0.250000					
0.223214		"fairy" - "religious"		0.250000		0.142857	

		0.250000		0.250000			
		0.410147		"fairy" - "of"		0.387298	0.187500
0.750000		0.315789					
		0.458333		"fairy" - "narrative"		0.500000	0.333333
		0.500000		0.500000			
		0.458333		"fairy" - "limited"		0.500000	0.333333
0.500000		0.500000					
		0.180200		"fairy" - "in"		0.188982	0.100000
0.250000		0.181818					
		0.223214		"fairy" - "government"		0.250000	0.142857
		0.250000		0.250000			
		0.458333		"fairy" - "george"		0.500000	0.333333
0.500000		0.500000					
		0.458333		"fairy" - "general"		0.500000	0.333333
0.500000		0.500000					
		0.868292		"fairy" - "forms"		0.866025	0.750000
1.000000		0.857143					
		0.223214		"fairy" - "for"		0.250000	0.142857
0.250000		0.250000					
		0.458333		"fairy" - "female"		0.500000	0.333333
0.500000		0.500000					
		0.223214		"establishing" - "wales"		0.250000	0.1428
57		0.250000		0.250000			
		0.180200		"establishing" - "the"		0.188982	0.100000
		0.250000		0.181818			
		0.223214		"establishing" - "tales"		0.250000	0.1428
57		0.250000		0.250000			
		0.180200		"establishing" - "study"		0.188982	0.1000
00		0.250000		0.181818			
		0.223214		"establishing" - "sea"		0.250000	0.142857
		0.250000		0.250000			
		0.712500		"establishing" - "religious"		0.750000	0.
600000		0.750000		0.750000			
		0.134980		"establishing" - "of"		0.129099	0.055556
		0.250000		0.105263			
		0.223214		"establishing" - "narrative"		0.250000	0.
142857		0.250000		0.250000			
		0.223214		"establishing" - "limited"		0.250000	0.14
2857		0.250000		0.250000			
		0.180200		"establishing" - "in"		0.188982	0.100000
		0.250000		0.181818			
		0.223214		"establishing" - "government"		0.250000	
0.142857		0.250000		0.250000			
		0.223214		"establishing" - "george"		0.250000	0.142
857		0.250000		0.250000			
		0.223214		"establishing" - "general"		0.250000	0.14
2857		0.250000		0.250000			
		0.268597		"establishing" - "forms"		0.288675	0.1666
67		0.333333		0.285714			
		0.712500		"establishing" - "for"		0.750000	0.600000
		0.750000		0.750000			
		0.223214		"establishing" - "female"		0.250000	0.142
857		0.250000		0.250000			
		0.223214		"establishing" - "fairy"		0.250000	0.1428
57		0.250000		0.250000			
		0.205207		"collection" - "wales"		0.223607	0.125000
		0.250000		0.222222			

	0.317849	"collection" - "the"	0.338062		0.200000	
	0.400000	0.333333				
	0.646872	"collection" - "tales"	0.670820		0.500000	
	0.750000	0.666667				
	0.317849	"collection" - "study"	0.338062		0.200000	
	0.400000	0.333333				
	0.205207	"collection" - "sea"	0.223607		0.125000	
	0.250000	0.222222				
	0.205207	"collection" - "religious"	0.223607		0.12	
5000		0.250000 0.222222				
	0.477970	"collection" - "of"	0.461880		0.250000	
	0.800000	0.400000				
	0.419343	"collection" - "narrative"	0.447214		0.28	
5714		0.500000 0.444444				
	0.419343	"collection" - "limited"	0.447214		0.2857	
14		0.500000 0.444444				
	0.156652	"collection" - "in"	0.169031		0.090909	
	0.200000	0.166667				
	0.205207	"collection" - "government"	0.223607		0.1	
25000		0.250000 0.222222				
	0.419343	"collection" - "george"	0.447214		0.28571	
4		0.500000 0.444444				
	0.419343	"collection" - "general"	0.447214		0.2857	
14		0.500000 0.444444				
	0.504099	"collection" - "forms"	0.516398		0.333333	
	0.666667	0.500000				
	0.205207	"collection" - "for"	0.223607		0.125000	
	0.250000	0.222222				
	0.419343	"collection" - "female"	0.447214		0.28571	
4		0.500000 0.444444				
	0.646872	"collection" - "fairy"	0.670820		0.500000	
	0.750000	0.666667				
	0.205207	"collection" - "establishing"	0.223607			
0.125000		0.250000 0.222222				
	0.223214	"city" - "wales"	0.250000		0.142857	
	0.250000	0.250000				
	0.559350	"city" - "the"	0.566947		0.375000	
0.750000		0.545455				
	0.223214	"city" - "tales"	0.250000		0.142857	
	0.250000	0.250000				
	0.180200	"city" - "study"	0.188982		0.100000	
	0.250000	0.181818				
	0.712500	"city" - "sea"	0.750000		0.600000	
0.750000		0.750000				
	0.223214	"city" - "religious"	0.250000		0.142857	
	0.250000	0.250000				
	0.271593	"city" - "of"	0.258199		0.117647	
0.500000		0.210526				
	0.458333	"city" - "narrative"	0.500000		0.333333	
	0.500000	0.500000				
	0.223214	"city" - "limited"	0.250000		0.142857	
	0.250000	0.250000				
	0.180200	"city" - "in"	0.188982		0.100000	
0.250000		0.181818				
	0.223214	"city" - "government"	0.250000		0.142857	
	0.250000	0.250000				
	0.223214	"city" - "george"	0.250000		0.142857	

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0.250000 | 0.250000
0.223214 | "city" - "general" | 0.250000 | 0.142857 |
0.250000 | 0.250000
0.268597 | "city" - "forms" | 0.288675 | 0.166667 |
0.333333 | 0.285714
0.223214 | "city" - "for" | 0.250000 | 0.142857 |
0.250000 | 0.250000
0.223214 | "city" - "female" | 0.250000 | 0.142857 |
0.250000 | 0.250000
0.223214 | "city" - "fairy" | 0.250000 | 0.142857 |
0.250000 | 0.250000
0.223214 | "city" - "establishing" | 0.250000 | 0.14285
7 | 0.250000 | 0.250000
0.205207 | "city" - "collection" | 0.223607 | 0.125000
| 0.250000 | 0.222222
0.223214 | "circumstantial" - "wales" | 0.250000 | 0.14
2857 | 0.250000 | 0.250000
0.559350 | "circumstantial" - "the" | 0.566947 | 0.3750
00 | 0.750000 | 0.545455
0.458333 | "circumstantial" - "tales" | 0.500000 | 0.33
3333 | 0.500000 | 0.500000
0.365956 | "circumstantial" - "study" | 0.377964 | 0.22
2222 | 0.500000 | 0.363636
0.458333 | "circumstantial" - "sea" | 0.500000 | 0.3333
33 | 0.500000 | 0.500000
0.223214 | "circumstantial" - "religious" | 0.250000 |
0.142857 | 0.250000 | 0.250000
0.410147 | "circumstantial" - "of" | 0.387298 | 0.18750
0 | 0.750000 | 0.315789
0.712500 | "circumstantial" - "narrative" | 0.750000 |
0.600000 | 0.750000 | 0.750000
0.458333 | "circumstantial" - "limited" | 0.500000 | 0.
333333 | 0.500000 | 0.500000
0.180200 | "circumstantial" - "in" | 0.188982 | 0.10000
0 | 0.250000 | 0.181818
0.223214 | "circumstantial" - "government" | 0.250000 |
0.142857 | 0.250000 | 0.250000
0.458333 | "circumstantial" - "george" | 0.500000 | 0.3
33333 | 0.500000 | 0.500000
0.458333 | "circumstantial" - "general" | 0.500000 | 0.
333333 | 0.500000 | 0.500000
0.553861 | "circumstantial" - "forms" | 0.577350 | 0.40
0000 | 0.666667 | 0.571429
0.223214 | "circumstantial" - "for" | 0.250000 | 0.1428
57 | 0.250000 | 0.250000
0.458333 | "circumstantial" - "female" | 0.500000 | 0.3
33333 | 0.500000 | 0.500000
0.458333 | "circumstantial" - "fairy" | 0.500000 | 0.33
3333 | 0.500000 | 0.500000
0.223214 | "circumstantial" - "establishing" | 0.250000 |
0.142857 | 0.250000 | 0.250000
0.419343 | "circumstantial" - "collection" | 0.447214 |
0.285714 | 0.500000 | 0.444444
0.458333 | "circumstantial" - "city" | 0.500000 | 0.333
333 | 0.500000 | 0.500000
0.712500 | "christmas" - "wales" | 0.750000 | 0.600000
| 0.750000 | 0.750000

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	0.180200	"christmas" - "the"	0.188982		0.100000	
	0.250000		0.181818			
	0.223214	"christmas" - "tales"	0.250000		0.142857	
	0.250000		0.250000			
	0.365956	"christmas" - "study"	0.377964		0.222222	
	0.500000		0.363636			
	0.223214	"christmas" - "sea"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"christmas" - "religious"	0.250000		0.142	
857	0.250000		0.250000			
	0.134980	"christmas" - "of"	0.129099		0.055556	
	0.250000		0.105263			
	0.223214	"christmas" - "narrative"	0.250000		0.142	
857	0.250000		0.250000			
	0.223214	"christmas" - "limited"	0.250000		0.14285	
7	0.250000		0.250000			
	0.559350	"christmas" - "in"	0.566947		0.375000	
	0.750000		0.545455			
	0.458333	"christmas" - "government"	0.500000		0.33	
3333	0.500000		0.500000			
	0.223214	"christmas" - "george"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"christmas" - "general"	0.250000		0.14285	
7	0.250000		0.250000			
	0.268597	"christmas" - "forms"	0.288675		0.166667	
	0.333333		0.285714			
	0.223214	"christmas" - "for"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"christmas" - "female"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"christmas" - "fairy"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"christmas" - "establishing"	0.250000		0.	
142857	0.250000		0.250000			
	0.205207	"christmas" - "collection"	0.223607		0.12	
5000	0.250000		0.222222			
	0.223214	"christmas" - "city"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"christmas" - "circumstantial"	0.250000			
0.142857	0.250000		0.250000			
	0.712500	"child's" - "wales"	0.750000		0.600000	
	0.750000		0.750000			
	0.180200	"child's" - "the"	0.188982		0.100000	
	0.250000		0.181818			
	0.223214	"child's" - "tales"	0.250000		0.142857	
	0.250000		0.250000			
	0.365956	"child's" - "study"	0.377964		0.222222	
	0.500000		0.363636			
	0.223214	"child's" - "sea"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"child's" - "religious"	0.250000		0.14285	
7	0.250000		0.250000			
	0.134980	"child's" - "of"	0.129099		0.055556	
	0.250000		0.105263			
	0.223214	"child's" - "narrative"	0.250000		0.14285	
7	0.250000		0.250000			
	0.223214	"child's" - "limited"	0.250000		0.142857	

		0.250000		0.250000		
		0.559350		"child's" - "in"		0.566947 0.375000
		0.750000		0.545455		
		0.458333		"child's" - "government"		0.500000 0.3333
33		0.500000		0.500000		
		0.223214		"child's" - "george"		0.250000 0.142857
		0.250000		0.250000		
		0.223214		"child's" - "general"		0.250000 0.142857
		0.250000		0.250000		
		0.268597		"child's" - "forms"		0.288675 0.166667
		0.333333		0.285714		
		0.223214		"child's" - "for"		0.250000 0.142857
		0.250000		0.250000		
		0.223214		"child's" - "female"		0.250000 0.142857
		0.250000		0.250000		
		0.223214		"child's" - "fairy"		0.250000 0.142857
		0.250000		0.250000		
		0.223214		"child's" - "establishing"		0.250000 0.14
2857		0.250000		0.250000		
		0.205207		"child's" - "collection"		0.223607 0.1250
00		0.250000		0.222222		
		0.223214		"child's" - "city"		0.250000 0.142857
		0.250000		0.250000		
		0.223214		"child's" - "circumstantial"		0.250000 0.
142857		0.250000		0.250000		
		0.712500		"child's" - "christmas"		0.750000 0.60000
0		0.750000		0.750000		
		0.365956		"case" - "wales"		0.377964 0.222222
		0.500000		0.363636		
		0.255952		"case" - "the"		0.285714 0.166667
		0.285714		0.285714		
		0.365956		"case" - "tales"		0.377964 0.222222
		0.500000		0.363636		
		0.830357		"case" - "study"		0.857143 0.750000
		0.857143		0.857143		
		0.180200		"case" - "sea"		0.188982 0.100000
		0.250000		0.181818		
		0.180200		"case" - "religious"		0.188982 0.100000
		0.250000		0.181818		
		0.386912		"case" - "of"		0.390360 0.222222
		0.571429		0.363636		
		0.365956		"case" - "narrative"		0.377964 0.222222
		0.500000		0.363636		
		0.559350		"case" - "limited"		0.566947 0.375000
		0.750000		0.545455		
		0.389610		"case" - "in"		0.428571 0.272727
		0.428571		0.428571		
		0.559350		"case" - "government"		0.566947 0.375000
		0.750000		0.545455		
		0.365956		"case" - "george"		0.377964 0.222222
		0.500000		0.363636		
		0.365956		"case" - "general"		0.377964 0.222222
		0.500000		0.363636		
		0.438276		"case" - "forms"		0.436436 0.250000
		0.666667		0.400000		
		0.180200		"case" - "for"		0.188982 0.100000
		0.250000		0.181818		

	0.559350	"case" - "female"	0.566947		0.375000	
	0.750000		0.545455			
	0.365956	"case" - "fairy"	0.377964		0.222222	
	0.500000		0.363636			
0	0.180200	"case" - "establishing"	0.188982		0.100000	
	0.250000		0.181818			
	0.317849	"case" - "collection"	0.338062		0.200000	
	0.400000		0.333333			
	0.180200	"case" - "city"	0.188982		0.100000	
	0.250000		0.181818			
222	0.365956	"case" - "circumstantial"	0.377964		0.222	
	0.500000		0.363636			
	0.365956	"case" - "christmas"	0.377964		0.222222	
	0.500000		0.363636			
	0.365956	"case" - "child's"	0.377964		0.222222	
	0.500000		0.363636			
	0.223214	"by" - "wales"	0.250000		0.142857	
0.250000		0.250000				
	0.559350	"by" - "the"	0.566947		0.375000	
0.750000		0.545455				
	0.223214	"by" - "tales"	0.250000		0.142857	
0.250000		0.250000				
	0.180200	"by" - "study"	0.188982		0.100000	
0.250000		0.181818				
	0.712500	"by" - "sea"	0.750000		0.600000	
0.750000		0.750000				
	0.223214	"by" - "religious"	0.250000		0.142857	
	0.250000		0.250000			
	0.271593	"by" - "of"	0.258199		0.117647	
0.500000		0.210526				
	0.458333	"by" - "narrative"	0.500000		0.333333	
	0.500000		0.500000			
	0.223214	"by" - "limited"	0.250000		0.142857	
0.250000		0.250000				
	0.180200	"by" - "in"	0.188982		0.100000	
0.250000		0.181818				
	0.223214	"by" - "government"	0.250000		0.142857	
	0.250000		0.250000			
	0.223214	"by" - "george"	0.250000		0.142857	
0.250000		0.250000				
	0.223214	"by" - "general"	0.250000		0.142857	
0.250000		0.250000				
	0.268597	"by" - "forms"	0.288675		0.166667	
0.333333		0.285714				
	0.223214	"by" - "for"	0.250000		0.142857	
0.250000		0.250000				
	0.223214	"by" - "female"	0.250000		0.142857	
0.250000		0.250000				
	0.223214	"by" - "fairy"	0.250000		0.142857	
0.250000		0.250000				
	0.223214	"by" - "establishing"	0.250000		0.142857	
	0.250000		0.250000			
	0.205207	"by" - "collection"	0.223607		0.125000	
	0.250000		0.222222			
	0.712500	"by" - "city"	0.750000		0.600000	
0.750000		0.750000				
	0.458333	"by" - "circumstantial"	0.500000		0.33333	

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3 |      0.500000 |      0.500000
    0.223214 | "by" - "christmas" |      0.250000 |      0.142857 |
    0.250000 |      0.250000
    0.223214 | "by" - "child's" |      0.250000 |      0.142857 |
    0.250000 |      0.250000
    0.180200 | "by" - "case" |      0.188982 |      0.100000 |
    0.250000 |      0.181818
    0.223214 | "biography" - "wales" |      0.250000 |      0.142857
|      0.250000 |      0.250000
    0.365956 | "biography" - "the" |      0.377964 |      0.222222 |
    0.500000 |      0.363636
    0.458333 | "biography" - "tales" |      0.500000 |      0.333333
|      0.500000 |      0.500000
    0.365956 | "biography" - "study" |      0.377964 |      0.222222
|      0.500000 |      0.363636
    0.223214 | "biography" - "sea" |      0.250000 |      0.142857 |
    0.250000 |      0.250000
    0.223214 | "biography" - "religious" |      0.250000 |      0.142
857 |      0.250000 |      0.250000
    0.410147 | "biography" - "of" |      0.387298 |      0.187500 |
    0.750000 |      0.315789
    0.458333 | "biography" - "narrative" |      0.500000 |      0.333
333 |      0.500000 |      0.500000
    0.458333 | "biography" - "limited" |      0.500000 |      0.33333
3 |      0.500000 |      0.500000
    0.180200 | "biography" - "in" |      0.188982 |      0.100000 |
    0.250000 |      0.181818
    0.223214 | "biography" - "government" |      0.250000 |      0.14
2857 |      0.250000 |      0.250000
    0.712500 | "biography" - "george" |      0.750000 |      0.600000
|      0.750000 |      0.750000
    0.712500 | "biography" - "general" |      0.750000 |      0.60000
0 |      0.750000 |      0.750000
    0.553861 | "biography" - "forms" |      0.577350 |      0.400000
|      0.666667 |      0.571429
    0.223214 | "biography" - "for" |      0.250000 |      0.142857 |
    0.250000 |      0.250000
    0.458333 | "biography" - "female" |      0.500000 |      0.333333
|      0.500000 |      0.500000
    0.458333 | "biography" - "fairy" |      0.500000 |      0.333333
|      0.500000 |      0.500000
    0.223214 | "biography" - "establishing" |      0.250000 |      0.
142857 |      0.250000 |      0.250000
    0.419343 | "biography" - "collection" |      0.447214 |      0.28
5714 |      0.500000 |      0.444444
    0.223214 | "biography" - "city" |      0.250000 |      0.142857 |
    0.250000 |      0.250000
    0.458333 | "biography" - "circumstantial" |      0.500000 |
0.333333 |      0.500000 |      0.500000
    0.223214 | "biography" - "christmas" |      0.250000 |      0.142
857 |      0.250000 |      0.250000
    0.223214 | "biography" - "child's" |      0.250000 |      0.14285
7 |      0.250000 |      0.250000
    0.365956 | "biography" - "case" |      0.377964 |      0.222222 |
    0.500000 |      0.363636
    0.223214 | "biography" - "by" |      0.250000 |      0.142857 |
    0.250000 |      0.250000

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0.223214	"bill" - "wales"	0.250000		0.142857	
0.250000		0.250000			
0.180200	"bill" - "the"	0.188982		0.100000	
0.250000		0.181818			
0.223214	"bill" - "tales"	0.250000		0.142857	
0.250000		0.250000			
0.180200	"bill" - "study"	0.188982		0.100000	
0.250000		0.181818			
0.223214	"bill" - "sea"	0.250000		0.142857	
0.250000		0.250000			
0.712500	"bill" - "religious"	0.750000		0.600000	
0.750000		0.750000			
0.134980	"bill" - "of"	0.129099		0.055556	
0.250000		0.105263			
0.223214	"bill" - "narrative"	0.250000		0.142857	
0.250000		0.250000			
0.223214	"bill" - "limited"	0.250000		0.142857	
0.250000		0.250000			
0.180200	"bill" - "in"	0.188982		0.100000	
0.250000		0.181818			
0.223214	"bill" - "government"	0.250000		0.142857	
0.250000		0.250000			
0.223214	"bill" - "george"	0.250000		0.142857	
0.250000		0.250000			
0.223214	"bill" - "general"	0.250000		0.142857	
0.250000		0.250000			
0.268597	"bill" - "forms"	0.288675		0.166667	
0.333333		0.285714			
0.712500	"bill" - "for"	0.750000		0.600000	
0.750000		0.750000			
0.223214	"bill" - "female"	0.250000		0.142857	
0.250000		0.250000			
0.223214	"bill" - "fairy"	0.250000		0.142857	
0.250000		0.250000			
0.712500	"bill" - "establishing"	0.750000		0.600000	
0.750000		0.750000			
0.205207	"bill" - "collection"	0.223607		0.125000	
0.250000		0.222222			
0.223214	"bill" - "city"	0.250000		0.142857	
0.250000		0.250000			
0.223214	"bill" - "circumstantial"	0.250000		0.142	
0.250000		0.250000			
0.223214	"bill" - "christmas"	0.250000		0.142857	
0.250000		0.250000			
0.223214	"bill" - "child's"	0.250000		0.142857	
0.250000		0.250000			
0.180200	"bill" - "case"	0.188982		0.100000	
0.250000		0.181818			
0.223214	"bill" - "by"	0.250000		0.142857	
0.250000		0.250000			
0.223214	"bill" - "biography"	0.250000		0.142857	
0.250000		0.250000			
0.334842	"a" - "wales"	0.288675		0.107143	
0.750000		0.193548			
0.465201	"a" - "the"	0.436436		0.214286	
0.857143		0.352941			
0.334842	"a" - "tales"	0.288675		0.107143	

0.750000	0.193548		
0.465201	"a" - "study"	0.436436	0.214286
0.857143	0.352941		
0.334842	"a" - "sea"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "religious"	0.288675	0.107143
0.750000	0.193548		
0.698916	"a" - "of"	0.695666	0.500000
0.933333	0.666667		
0.334842	"a" - "narrative"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "limited"	0.288675	0.107143
0.750000	0.193548		
0.465201	"a" - "in"	0.436436	0.214286
0.857143	0.352941		
0.334842	"a" - "government"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "george"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "general"	0.288675	0.107143
0.750000	0.193548		
0.273413	"a" - "forms"	0.222222	0.071429
0.666667	0.133333		
0.334842	"a" - "for"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "female"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "fairy"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "establishing"	0.288675	0.107143
0.750000	0.193548		
0.384281	"a" - "collection"	0.344265	0.142857
0.800000	0.250000		
0.334842	"a" - "city"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "circumstantial"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "christmas"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "child's"	0.288675	0.107143
0.750000	0.193548		
0.465201	"a" - "case"	0.436436	0.214286
0.857143	0.352941		
0.334842	"a" - "by"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "biography"	0.288675	0.107143
0.750000	0.193548		
0.334842	"a" - "bill"	0.288675	0.107143
0.750000	0.193548		

Systems test 2 - Similarity measures

average	pair	cosine	jaccard
overlap	dice		

0.389562	"cava" - "dipped"	0.408248		0.250000	
0.500000		0.400000			
0.625000	"boon" - "dipped"	0.666667		0.500000	
0.666667		0.666667			
0.389562	"boon" - "cava"	0.408248		0.250000	
0.500000		0.400000			
0.389562	"atlas" - "dipped"	0.408248		0.250000	
0.500000		0.400000			
1.000000	"atlas" - "cava"	1.000000		1.000000	
1.000000		1.000000			
0.389562	"atlas" - "boon"	0.408248		0.250000	
0.500000		0.400000			

Systems test 3 - Similarity measures

average	pair	cosine	jaccard
overlap	dice		

0.346722	"DocB" - "DocC"	0.353553	0.200000
0.500000	0.333333		
0.553861	"DocA" - "DocC"	0.577350	0.400000
0.666667	0.571429		
0.820791	"DocA" - "DocB"	0.816497	0.666667
1.000000	0.800000		

Pairwise Similarity

In []:

Systems test 1 - Similarity measures

average	pair	cosine	jaccard	
1.000000	female - limited	1.000000	1.000000	1
0.868292	fairy - forms	0.866025	0.750000	1
0.868292	forms - tales	0.866025	0.750000	1
0.830357	case - study	0.857143	0.750000	0
0.712500	bill - establishing	0.750000	0.600000	0
0.712500	christmas - wales	0.750000	0.600000	0
0.712500	circumstantial - narrative	0.750000	0.600000	0
0.712500	by - sea	0.750000	0.600000	0
0.712500	by - city	0.750000	0.600000	0
0.712500	child's - wales	0.750000	0.600000	0
0.712500	biography - george	0.750000	0.600000	0
0.712500	child's - christmas	0.750000	0.600000	0
...				

Systems test 2 - Similarity measures

average	pair	cosine	jaccard	
1.000000	atlas - cava	1.000000	1.000000	1
0.625000	boon - dipped	0.666667	0.500000	0
0.389562	cava - dipped	0.408248	0.250000	0
0.389562	boon - cava	0.408248	0.250000	0
0.389562	atlas - dipped	0.408248	0.250000	0
0.389562	atlas - boon	0.408248	0.250000	0

Systems test 3 - Similarity measures

average	pair	cosine	jaccard	
0.820791	DocA - DocB	0.816497	0.666667	1
0.553861	DocA - DocC	0.577350	0.400000	0
0.346722	DocB - DocC	0.353553	0.200000	0

=== END OF PHASE 1 ===

In []: