Locality Sensitive Hashing

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Agenda

- Locality Sensitive Hashing (LSH)
- Hash functions
- Hashed shingles
- Signatures
- Characteristic Matrix
- Minhash (Jaccard Similarity Approximation)
- Back to LSH

```
cora_id unique_id
##
## 1
            1
                       1
            2
## 2
                       1
## 3
            3
                       1
            4
## 4
                       1
## 5
            5
                       1
            6
## 6
##
         cora_id unique_id
## 1694
            1874
                        135
## 1809
            1875
                        135
## 1695
            1876
                        136
## 1696
            1877
                        136
## 1697
            1878
                        136
## 1810
            1879
                        136
## [1] 1879
```

LSH

Locality sensitive hashing (LSH) is a fast method of blocking for record linkage that orginates from the computer science literature.

- LSH tries to preserve similarity after dimension reduction.
 - What kind of similarity? \leftrightarrow What kind of dimension reduction?

Data set

Consider the cora citation data set.

1. Shingle all records using a shingle size of 3. Then calculate the Jaccard similarity for all record pairs using the shingled records.

```
# get only the columns we want
# number of records
```

```
n <- nrow(cora)
# create id column
dat <- data.frame(id = seq_len(n))</pre>
# get columns we want
dat <- cbind(dat, cora[, c("title", "authors", "journal")])</pre>
shingles <- apply(dat, 1, function(x) {</pre>
  # tokenize strings
 tokenize character shingles(paste(x[-1], collapse=" "), n = 3)[[1]]
})
# empty holder for similarities
jaccard <- expand.grid(record1 = seq_len(n),</pre>
                       record2 = seq_len(n))
# don't need to compare the same things twice
jaccard <- jaccard[jaccard$record1 < jaccard$record2,]</pre>
time <- Sys.time() # for timing comparison</pre>
jaccard$similarity <- apply(jaccard, 1, function(pair) {</pre>
  # get jaccard similarity for each record pair
  jaccard_similarity(shingles[[pair[1]]], shingles[[pair[2]]])
})
# timing
time <- difftime(Sys.time(), time, units = "secs")</pre>
head(jaccard)
##
        record1 record2 similarity
## 1880
            1 2 0.8648649
## 3759
              1
                     3 0.8648649
## 3760
             2
                     3 1.0000000
## 5638
             1
                     4 0.8648649
## 5639
              2
                     4 1.0000000
## 5640
              3
                      4 1.0000000
  2. Visually plot the Jaccard similarity. What do you observe?
# plot the jaccard similarities for each pair of records
ggplot(jaccard) +
  geom_raster(aes(x = record1, y = record2,
                  fill=similarity)) +
  theme(aspect.ratio = 1) +
  scale fill gradient("Jaccard similarity") +
  xlab("Record id") + ylab("Record id")
```

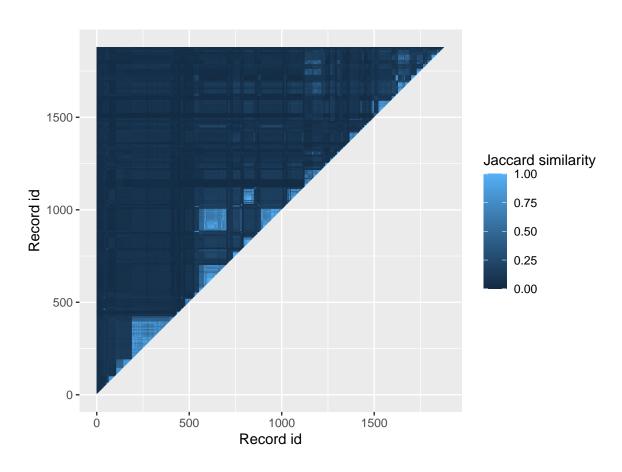


Figure 1: Jaccard similarity for each pair of records. Light blue indicates the two records are more similar and dark blue indicates less similar.

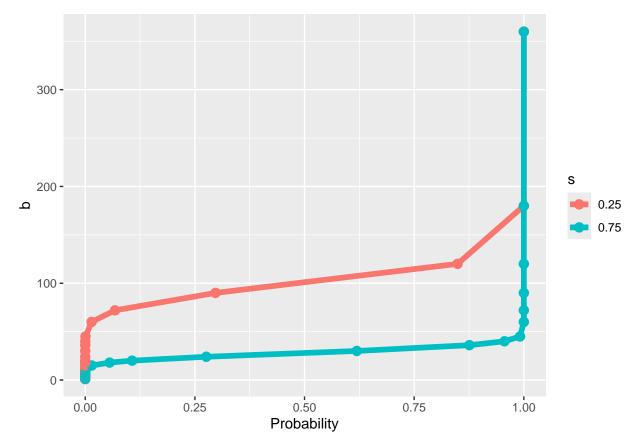
Perform LSH

3. To reduce the overall computational complexity, let's use the lsh approximation.

There an easy way to do LSH using the built in functions in the textreuse package via the functions minhash_generator and lsh (so we don't have to perform it by hand):

Find the number of buckets or bands to use

```
library(numbers)
m <- 360
bin_probs \leftarrow expand.grid(s = c(.25, .75), h = m, b = divisors(m))
#bin_probs
# choose appropriate num of bands and number of random permutations m (tuning parameters)
bin_probs$prob <- apply(bin_probs, 1, function(x) lsh_probability(x[["h"]], x[["b"]], x[["s"]]))
# plot as curves
ggplot(bin_probs) +
  geom_line(aes(x = prob, y = b, colour = factor(s), group = factor(s)), size = 2) +
  geom_point(aes(x = prob, y = b, colour = factor(s)), size = 3) +
  xlab("Probability") +
  scale_color_discrete("s")
## Warning: Using `size` aesthetic for lines was deprecated in ggplot2 3.4.0.
## i Please use `linewidth` instead.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
```



```
# create the minhash function
minhash <- minhash_generator(n = m, seed = 02082018)
b <- 90</pre>
```

Build corpus and perform shingling

```
head(dat)
##
     id
                  title
## 1 1 Inganas and M.R
## 2 2
                   <NA>
## 3 3
                   <NA>
## 4 4
                   <NA>
## 5 5
                   <NA>
## 6 6
                   <NA>
##
                                                                                            authors
## 1
                                   M. Ahlskog, J. Paloheimo, H. Stubb, P. Dyreklev, M. Fahlman, O
## 2 M. Ahlskog, J. Paloheimo, H. Stubb, P. Dyreklev, M. Fahlman, O. Inganas and M.R.
## 3 M. Ahlskog, J. Paloheimo, H. Stubb, P. Dyreklev, M. Fahlman, O. Inganas and M.R.
       M. Ahlskog, J. Paloheimo, H. Stubb, P. Dyreklev, M. Fahlman, O. Inganas and M.R. Andersson
     M. Ahlskog, J. Paloheimo, H. Stubb, P. Dyreklev, M. Fahlman, O. Inganas and M.R. Andersson
## 6
       M. Ahlskog, J. Paloheimo, H. Stubb, P. Dyreklev, M. Fahlman, O. Inganas and M.R. Andersson
##
                      journal
## 1 Andersson, J Appl. Phys.
## 2
                 JAppl. Phys.
## 3
                J Appl. Phys.
## 4
                 J Appl.Phys.
## 5
                J Appl. Phys.
                 J Appl.Phys.
# build the corpus using textreuse
docs <- apply(dat, 1, function(x) paste(x[-1], collapse = " ")) # get strings</pre>
names(docs) <- dat$id # add id as names in vector</pre>
corpus <- TextReuseCorpus(text = docs, # dataset</pre>
                          tokenizer = tokenize_character_shingles, n = 3,
                          simplify = TRUE, # shingles
                          progress = FALSE, # quietly
                          keep_tokens = TRUE, # store shingles
                          minhash_func = minhash) # use minhash
```

Find buckets, candidate records, and Jaccard similarity

Now, we find the buckets, candidates records, and calculate the Jaccard similarity for the candidate records (in the buckets)

```
# perform lsh to get buckets
buckets <- lsh(corpus, bands = b, progress = FALSE)

## Warning: `gather_()` was deprecated in tidyr 1.2.0.

## i Please use `gather()` instead.

## i The deprecated feature was likely used in the textreuse package.

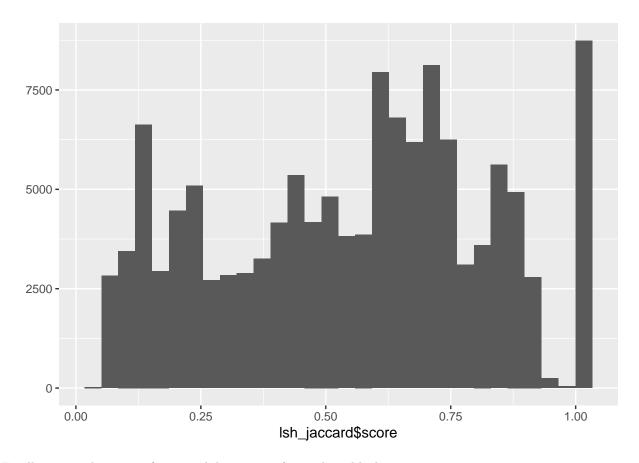
## Please report the issue at <https://github.com/ropensci/textreuse/issues>.

## This warning is displayed once every 8 hours.

## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was

## generated.
```

```
# grab candidate pairs
candidates <- lsh_candidates(buckets)</pre>
# get Jaccard similarities only for candidates
lsh_jaccard <- lsh_compare(candidates, corpus,</pre>
                            jaccard_similarity, progress = FALSE)
head(buckets)
## # A tibble: 6 x 2
    doc buckets
     <chr> <chr>
##
          fb93d6f4c56666ec8210570af8e8edd0
## 1 1
## 2 1
           cf942dbe840d4365cf182ea24d6951c8
## 3 1
           e9535be0f24e39103ba1f11442cc170e
## 4 1
           52a293069e3920a0f56e38a0f0c6af37
## 5 1
           b751d8b2d24bec53a78b3043dc017837
## 6 1
           0ce07d00e8c2e811204352b51229ed18
dim(buckets)
## [1] 169110
length(unique(buckets))
## [1] 2
We now plot the Jaccard similarities that are candidate pairs (under LSH)
## Warning: `qplot()` was deprecated in ggplot2 3.4.0.
## This warning is displayed once every 8 hours.
## Call `lifecycle::last_lifecycle_warnings()` to see where this warning was
## generated.
## `stat_bin()` using `bins = 30`. Pick better value with `binwidth`.
```



Finally, we need to move from candidate pairs of records to blocks.

library(igraph) #graph package

head(blocks)

```
## Warning: package 'igraph' was built under R version 4.3.2
##
## Attaching package: 'igraph'
## The following objects are masked from 'package:stats':
##
##
       decompose, spectrum
## The following object is masked from 'package:base':
##
##
       union
# think of each record as a node
# there is an edge between nodes if they are candidates
g <- make_empty_graph(n, directed = FALSE) # empty graph</pre>
g <- add_edges(g, is.vector((candidates[, 1:2]))) # candidate edges
g <- set_vertex_attr(g, "id", value = dat$id) # add id
# get custers, these are the blocks
clust <- components(g, "strong") # get clusters</pre>
blocks <- data.frame(id = V(g)$id, # record id
                     block = clust$membership) # block number
```

```
## id block
## 1 1 1
## 2 2
## 3 3 3
## 4 4 4
## 5 5 5
## 6 6 6
tail(blocks)
        id block
## 1874 1874 1874
## 1875 1875 1875
## 1876 1876 1876
## 1877 1877 1877
## 1878 1878 1878
## 1879 1879 1879
dim(blocks)
## [1] 1879
            2
```

Evaluation Metrics