Module 4: Deterministic Blocking

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Reminders

- ► Homework 0 is due today at 5 PM EDT
- ► Homework 1 is ready to start!

Recap

Define blocking and relate it to the figure below.

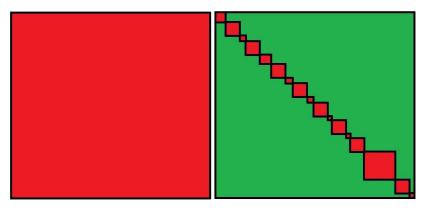


Figure 1: Left: All to all record comparison. Right: Example of resulting blocking partitions.

Traditional blocking

- What is traditional blocking?
- ➤ Consider the RLdata500 data set. What is a good blocking attribute to use? How can you validate this?
- What is probabilistic blocking?

Evaluation metrics

Explain the following evaluation metrics and why we use these:

- ► Reduction ratio
- Precision
- ► Recall
- Fscore

Reading

- ▶ Binette and Steorts (2020)
- Steorts, Ventura, Sadinle, Fienberg (2014)
- ► Murray (2016)
- ► Christen (2012), Chapter 4

Agenda

- ► Data Cleaning Pipeline
- Blocking
- ► Traditional Blocking
- Probabilistic Blocking
- Evaluation Metrics
- Examples

Load R packages and data

```
knitr::opts_chunk$set(echo = TRUE,
                       fig.width=4,
                       fig.height=3,
                       fig.align="center")
library(RecordLinkage)
library(blink)
library(italy)
library(tidyverse)
library(assert)
data(italy08)
data(italy10)
data(RLdata500)
```

Data Cleaning Pipeline

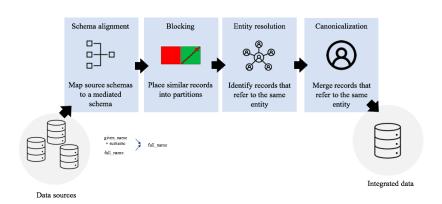


Figure 2: Data cleaning pipeline.

Blocking

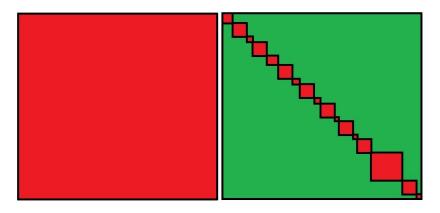


Figure 3: Left: All to all record comparison. Right: Example of resulting blocking partitions.

Blocking

- ▶ Blocking places similar records into partitions/blocks.
- ► ER (typically) is only performed within each block.

Traditional Blocking

- ▶ A deterministic (fixed) partition is formed based upon the data.
- ► A partition is created by treating certain fields that are thought to be nearly error-free as fixed.

Example: Blocking on date of birth year.

Traditional Blocking

- ▶ Benefits: simple, easy to understand, and fast to implement.
- Downsides: the blocks are treated as error free, which is not usually accurate and can lead to errors in the ER task that cannot be accounted for.

Probabilistic Blocking

► A probability model is used to cluster the data into blocks/partitions.

Example: Fellegi-Sunter (1969), or Locality Sensitive Hashing

Under both blocking approaches, record pairs that do not meet the blocking criteria are automatically classified as non-matches.

Evaluation Metrics

Evaluation metrics are important for ER as they help us evaluate our proposed methodology (as long as some notion of ground truth exists).

The three that we will focus on in this module are:

- reduction ratio
- precision
- recall
- f-measure

Reduction Ratio

The reduction ratio (RR) measures the relative reduction of the comparison space from the de-duplication or hashing technique.

See Christen (2012), Steorts, Ventura, Sadinle, Fienberg (2014) for a formal definition.

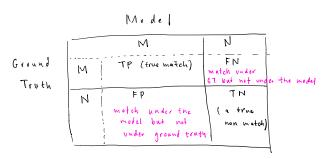
Pairwise Precision and Recall

Let's now turn to formally defining the pairwise precision and recall.

The confusion matrix

- 1. Pairs of data can be linked in both the handmatched training data (which we refer to as "truth") and under the estimated linked data. We refer to this situation as true positives (TP).
- 2. Pairs of data can be linked under the truth but not linked under the estimate, which are called false negatives (FN).
- 3. Pairs of data can be not linked under the truth but linked under the estimate, which are called false positives (FP).
- Pairs of data can be not linked under the truth and also not linked under the estimate, which we refer to as true negatives (TN).

The confusion matrix



Pairwise evaluation metrics

$$\mathsf{Recall} = \frac{\mathit{TP}}{\mathit{TP} + \mathit{FN}} = 1 - \mathsf{FNR}.$$

$$Precision = \frac{TP}{TP + FP} = 1 - FDR.$$

F-measure =
$$2 \times \frac{(precision \times recall)}{(precision + recall)}$$
.

Recall

- ▶ For blocking, it is critical the recall be as close a possible to 1.
- ➤ To think about why, what does it mean if we have a blocking criterion where our recall is 0.5?

See Shrivastava and Steorts (2018) and Chen, Shrivastava, Steorts (2018) for further regarding about blocking criterion using human rights data.

Example: RLdata500

Let's return to the RLdata500 data set, where we will block by last name initial.

Our goal are the following:

- visualize the blocks
- compute the evaluation metrics introduced

Example: RLdata500

head(RLdata500)

```
##
    fname c1 fname c2 lname c1 lname c2 by bm bd
     CARSTEN
                        MEIER
                                  <NA> 1949 7 22
## 1
                 <NA>
## 2
        GERD
                 <NA>
                        BAUER
                                  <NA> 1968 7 27
## 3
      ROBERT
                 <NA> HARTMANN
                                  <NA> 1930 4 30
## 4
      STEFAN
             <NA>
                        WOLFF
                                  <NA> 1957 9 2
## 5
        RALF
                 <NA>
                      KRUEGER
                                  <NA> 1966
                                            1 13
                 <NA> FRANKE
##
  6
     JUERGEN
                                  <NA> 1929
                                            7 4
```

Example: Traditional blocking

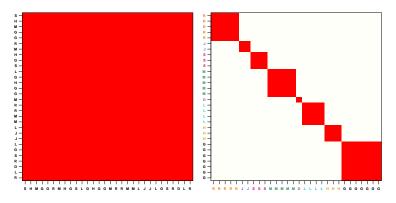


Figure 5: All-to-all record comparisons (left) versus partitioning records into blocks by lastname initial and comparing records only within each partition (right).

```
# Total number of all to all record comparisons choose(500,2)
```

[1] 124750

```
# Block by last name initial
last_init <- substr(RLdata500[,"lname_c1"], 1, 1)
head(last_init)

## [1] "M" "B" "H" "W" "K" "F"

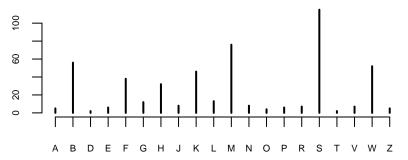
# Total number of blocks
length(unique(last_init))

## [1] 20</pre>
```

```
# Total number of records per block
recordsPerBlock <- table(last_init)
head(recordsPerBlock)

## last_init
## A B D E F G
## 5 56 2 6 38 12</pre>
```

Observe that the block sizes vary.



What is the overall dimension reduction form the original space to the reduced space induced by blocking?

Recall the total number of all-to-all record comparisons made was:

```
choose(500, 2)
```

[1] 124750

Using blocking, we have reduced the compison space to the following:

```
sum(choose(recordsPerBlock, 2))
```

[1] 14805

How do we calculate the reduction ratio (RR)?

The reduction ratio is

RR = % comparisons eliminated by blocking.

```
(choose(500, 2) - sum(choose(recordsPerBlock, 2))) /
choose(500, 2)
```

```
## [1] 0.8813226
```

How do we calculate the RR (via a function)?

[1] 0.8813226

```
reduction.ratio <- function(block.labels) {
   n_all_comp = choose(length(block.labels), 2)
   n_block_comp = sum(choose(table(block.labels), 2))

   (n_all_comp - n_block_comp) / n_all_comp
}
reduction.ratio(last_init)</pre>
```

Reduction Ratio

In summary, we have reduced the comparison space by roughly 88 percent.

Evaluation metrics

Let's now code up the evaluation metrics for pairwise precision and recall.

Pairwise Precision

```
precision <- function(block.labels, IDs) {
  ct = xtabs(~block.labels+IDs)

# Number of true positives
  TP = sum(choose(ct, 2))

# Number of positives = TP + FP
  P = sum(choose(rowSums(ct), 2))

return(TP/P)
}</pre>
```

Pairwise Recall

```
recall <- function(block.labels, IDs) {
  ct = xtabs(~IDs+block.labels)

# Number of true positives
  TP = sum(choose(ct, 2))

# Number of true links = TP + FN
  TL = sum(choose(rowSums(ct), 2))

return(TP/TL)
}</pre>
```

Pairwise Precision and Recall

```
precision(last_init, identity.RLdata500)

## [1] 0.003377237

recall(last_init, identity.RLdata500)

## [1] 1

precision(last_init, identity.RLdata500) ==
    recall(identity.RLdata500, last_init)

## [1] TRUE
```

Italian Survey on Household and Wealth (SHIW)

- We will now explore a case study to the Italian Survey on Household and Wealth (SHIW)
- ► The SHIW is a sample survey 383 households conducted by the Bank of Italy every two years (2008 and 2010).
- ► The data set is anonymized to remove first and last name (and other sensitive information).

SHIW

The following attribute information is available:

- ► PARENT (parental status)
- GENDER
- ANASC (year of birth)
- NASCREG (working status)
- CIT (employment status)
- ACOM4C (branch of activity)
- STUDIO (town size)
- Q (quality of life status)
- QUAL (whether or not Italian national)
- SETT (highest educational level obtained)
- ► IREG (region of italy)

Explore Data

head(italy08) # first year of SHIW ## id PARENT SEX ANASC NASCREG CIT ACOM4C STUDIO Q QUAL SETT IREG ## 1 1040021 1948 16 5 1 16 ## 2 1040022 10 2 1952 16 16 20 1 1 1972 5 1 1 20 ## 3 1110521 1 20 1 2 3 6 5 20 ## 4 1110522 3 1 1935 20 1 3 3 ## 5 1110523 2 1941 20 ## 6 119401 1 1941 4 3

Explore Data

```
head(italy10) # second year of SHIW
##
         id PARENT SEX ANASC NASCREG CIT ACOM4C STUDIO Q QUAL SETT IREG
## 1 1040021
                    2 1948
                               16
                                                               16
                    2 1952
## 2 1040022
               11
                               16
                                                               16
                   2 1941
                               20 1
                                                3 3 6
                                                               20
## 3 1110521
                             20 1
                                                2 3 6
                                                         5 20
## 4 1110522
                   1 1935
                               20 1
                                                5 1
## 5 1110523
                   1 1972
                                                               20
## 6 119721
                    2 1948
                               16
                                                2 2
                                                               17
```

Reformat Data

```
id08 <- italy08$id
id10 <- italy10$id
id <- c(italy08$id, italy10$id) # combine the id
italy08 <- italy08[-c(1)] # remove the id
italy10 <- italy10[-c(1)] # remove the id
italy <- rbind(italy08, italy10)
head(italy)</pre>
```

##		PARENT	SEX	ANASC	NASCREG	CIT	${\tt ACOM4C}$	STUDIO	Q	QUAL	SETT	IREG
##	1	1	2	1948	16	1	0	5	1	2	3	16
##	2	10	2	1952	16	1	0	7	1	2	3	16
##	3	1	1	1972	20	1	2	5	1	1	4	20
##	4	3	1	1935	20	1	2	2	3	6	5	20
##	5	3	2	1941	20	1	2	3	3	6	5	20
##	6	1	1	1941	7	1	0	4	3	6	5	7

Your turn

- Construct a blocking criterion for the SHIW data set
- Provide code to construct the blocks
- Are your blocks well balanced?
- What is the reduction ratio?
- What is the pairwise recall and precision?
- Would you recommend your blocking criterion for an ER task? Why or why not.

Hint: You might consider blocking on gender, regions (in Italy), or combinations of these. What do you find?

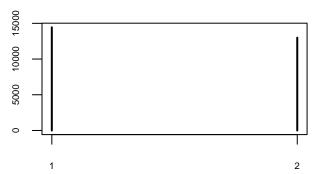
Let's block on gender.

```
# block by gender
blockByGender <- italy$SEX
recordsPerBlock <- table(blockByGender)
head(recordsPerBlock)</pre>
```

```
## blockByGender
## 1 2
## 14442 12993
```

The block sizes are similar. But note, they are still quite large.

```
# Checking block sizes
plot(recordsPerBlock,
cex.axis=0.6, xlab="", ylab="")
```



```
print(rr <- reduction.ratio(blockByGender))</pre>
```

[1] 0.4986234

We have reduced the overall space by rougly 50 percent.

```
precision(blockByGender, id)

## [1] 3.599727e-05

recall(blockByGender, id)

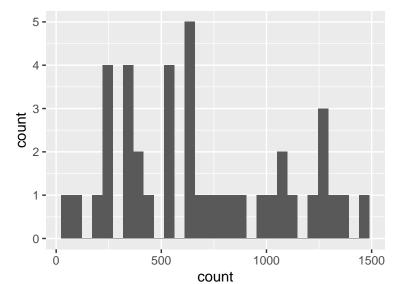
## [1] 0.9113109
```

Let's block on a combination of gender and region.

Hint: Use tidyverse.

```
Your turn colution ?

italy %>%
group.by(IREG, SEX) %>%
summarise(count = n(), .groups="drop") %>%
ggplot() +
geom_histogram(aes(count))
```



```
blockIDs = paste(italy$IREG, italy$SEX, sep="_")
table(blockIDs)

## blockIDs

## 1_1 1_2 10_1 10_2 11_1 11_2 12_1 12_2 13_1 13_2 14_1 14_2 15_1 15_2 16_1 16_2

## 1282 1248 536 534 624 611 772 671 368 339 249 229 1310 1019 846 615

## 17_1 17_2 18_1 18_2 19_1 19_2 2_1 2_2 20_1 20_2 3_1 3_2 4_1 4_2 5_1 5_2

## 246 209 333 262 969 709 91 73 634 659 1489 1386 324 321 1056 882

## 6_1 6_2 7_1 7_2 8_1 8_2 9_1 9_2

## 418 371 517 533 1254 1243 1124 1079

print(rr <- reduction.ratio(blockIDs))

## [1] 0.9667954
```

```
precision(blockIDs, id)

## [1] 0.0005429847

recall(blockIDs, id)

## [1] 0.9103717
```