# Homework 2

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#### Exercise 1

The 4 main challenges of Entity Resolution are

- 1. Costly manual labelling
- 2. Scalability/computational efficiency
- 3. Limited treatment of uncertainty
- 4. Unreliable evaluation

#### Exercise 2

- a. For 10 records, there are  $10^2$  or 100 total brute-force comparisons
- b. For 100 records, there are  $100^2$  or 10,000 total brute-force comparisons For 1000 records, there are  $1000^2$  or 1,000,000 total brute-force comparisons For 10000 records, there are  $10000^2$  or 100,000,000 total brute-force comparisons
- c. The number of comparisons grows quadratically with the number of records.

## Exercise 3

Dataset with 1,000,000 entries, 500,000 are true matches, method found 600,000 as matches, and 400,000 of these are true matches. TP + FP + TN + FN = 50,000,000

```
a. TP = 400,000, FP = 200,000, TN = 49,300,000, FN = 100,000 b. Accuracy = \frac{400,000+49,300,000}{50,000,000} = 0.994 c. Precision = \frac{400,000}{400,000+200,000} = 2/3 d. Recall = \frac{400,000}{400,000+100,000} = 0.8 e. F-Measure = \frac{2*(2/3)*0.8}{(2/3)+0.8} = 0.7\overline{2}
```

f. Precision, recall, and f-measure are much better metrics than accuracy because there is a large number of true negatives which leads to a class imbalance.

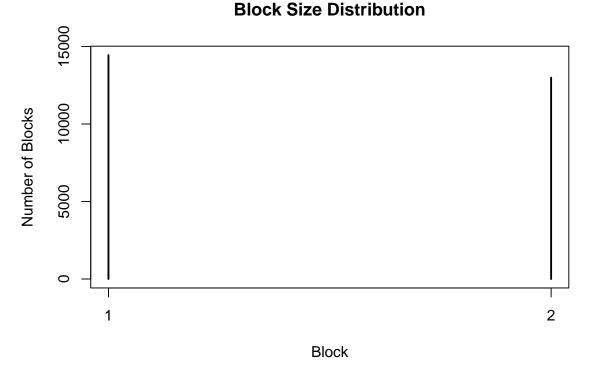
### Exercise 4

Italian Household Survey on Household and Wealth

```
a. # Load necessary packages
if (!require("pacman")) {
  install.packages("pacman")
  library(pacman)
}
```

```
## Loading required package: pacman
  p_load(RecordLinkage, blink, italy, tidyverse, assert)
  id08 <- italy08$id
  id10 <- italy10$id
  id <- c(italy08$id, italy10$id) # combine the id</pre>
  italy08 <- italy08[-c(1)] # remove the id</pre>
  italy10 <- italy10[-c(1)] # remove the id</pre>
  italy <- rbind(italy08, italy10)</pre>
  head(italy)
  ##
        PARENT SEX ANASC NASCREG CIT ACOM4C STUDIO Q QUAL SETT IREG
  ## 1
                  2
                    1948
                                             0
                                                     5 1
                                16
                                      1
             1
                  2
                                                             2
  ## 2
                                             0
                                                     7 1
                                                                  3
            10
                     1952
                                16
                                      1
                                                                       16
  ## 3
                  1 1972
                                20
                                             2
                                                     5 1
                                                             1
                                                                  4
                                                                       20
             1
                                      1
                                             2
                                                     2 3
  ## 4
              3
                  1 1935
                                20
                                      1
                                                                       20
  ## 5
              3
                  2
                    1941
                                20
                                      1
                                             2
                                                     3 3
                                                             6
                                                                  5
                                                                       20
  ## 6
              1
                  1
                     1941
                                             0
                                                     4 3
                                                                        7
  blockByGender <- italy$SEX</pre>
  recordsPerBlock <- table(blockByGender)</pre>
b. # Plot the blocks
```

plot(recordsPerBlock, xlab = "Block", ylab = "Number of Blocks", main = "Block Size Distribution")



There are only 2 blocks and Block 1 is slightly larger than Block 2 with 14442 and 12993 records respectively.

```
c. # Function to calculate reduction ratio
  ReductionRatio <- function(dataset) {</pre>
    n_all_comp = choose(length(dataset), 2)
    n block comp = sum(choose(table(dataset), 2))
    (n_all_comp - n_block_comp) / n_all_comp
  ReductionRatio(blockByGender)
  ## [1] 0.4986234
  The reduction ratio is 0.50. We reduced the comparison space by roughly 50%.
d. Precision: 3.6e-05
  .0036\% of the classified matches were true matches
  Recall: 0.91
  91% of the true matches are classified correctly
  # Precision Function
  precision <- function(block.labels, IDs) {</pre>
    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)
  }
  # Recall Function
  recall <- function(block.labels, IDs) {</pre>
    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)
  precision(blockByGender, id)
  ## [1] 3.599727e-05
```

```
recall(blockByGender, id)
```

## [1] 0.9113109

- e. This is an okay approach to blocking because 91% of the true matches are classified correctly (high recall)
- f. This is not a recommended approach to entity resolution because a very very minute percentage of the classified matches are true matches (very low precision)