Module 6: Exact-Mapping for Entity Resolution

Rebecca C. Steorts

joint with Brian Kundinger and Jerry Reiter

Reading

- ▶ Binette and Steorts (2020)
- ► Newcombe et al. (1959)
- ► Fellegi and Sunter (1969)
- ► Kundinger, Reiter, and Steorts (2024)

Assume two databases A and B (or sometimes called X_1 and X_2 .)

- ▶ The relative size of each database is N_1 and N_2 .
- Assume there are duplicates across the databases but not within them. This is called a bipartite record linkage assumption.
- Assume f = 1, ... F fields (or attributes).
- Let L_f denote the number of categories for field f.

Motivation

- ▶ Record pairs that refer to the same entity should be similar.
- ▶ Records pairs that refer to different entities should be dissimilar.

We can compare record pairs using similarity scores (or distance functions).

Examples: Jaccard, Edit, Jaro, Jaro-Winkler.

Comparison Vectors (or Data)

This motivates the comparison vector or comparison data.

Consider

$$\gamma_{ij} = (\gamma_{ij}^1, \dots, \gamma_{ij}^F),$$

where

 γ_{ij}^f compares field f for record $i \in A$ and $j \in B$.

Collect all the comparison vectors as

$$oldsymbol{\gamma} = \{oldsymbol{\gamma}_{ij}\}_{i=1,j=1}^{oldsymbol{N_1,N_2}}$$

Comparison Vectors (or Data)

The above notation is used in the literature as it is compact and short!

What do these vectors look like in practice?

Let

$$i = 1, 2, \ldots, N_1 \times N_2$$

enumerate the set of all record pairs in $A \times B$.

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▶ Each γ_i^j compares the *j*th field of the records.

How can we visualize the comparison vectors?

$$\gamma_1 = (\gamma_1^{(1)}, \gamma_1^{(2)}, \dots, \gamma_1^{(F)}) \tag{1}$$

$$\gamma_2 = (\gamma_2^{(1)}, \gamma_2^{(2)}, \dots, \gamma_2^{(F)}) \tag{2}$$

$$\gamma_{(N_1 \times N_2)} = (\gamma_{(N_1 \times N_2)}^{(1)}, \gamma_{(N_1 \times N_2)}^{(2)}, \dots, \gamma_{(N_1 \times N_2)}^{(F)}) \tag{4}$$

Let

$$\boldsymbol{\gamma} = (\gamma_1^{(1)}, \gamma_2^{(2)}, \dots, \gamma_{(N_1 \times N_2)}^{(F)})$$

Exact Mapping

We will walk through an approach to create a more efficient representation of the comparison vectors (but not do any dimension reduction).

- Let P be the number of exact agreement patterns in γ , which is bounded above by $\prod_{f=1}^{F} (L_f + 1)$.
- Consider the following function

$$h_f^{(i,j)} = I_{obs}(\gamma_{ij}^f) 2^{\gamma_{ij}^f + I(f > 1) \times \sum_{e=1}^{f-1} (L_e - 1))},$$
 (5)

which maps a record pair for a field f to a unique integer.

Summing over fields f = 1, ..., F for record pair (i, j) results in

$$h^{(i,j)} = \sum_{f=1}^{F} h_f^{(i,j)}.$$

- ▶ Enumerate unique hashed agreement patterns from 1 to P.
- Denote each unique mapped agreement pattern as $h_p = (h_p^1, \dots, h_p^F) \implies$ that when record pair (i, j) has agreement pattern p, we write $\gamma_{ij} = h_p$.
- ▶ Collect all the agreement patterns as $P = \{h_p \mid p \in [P]\}.$

- Nhen performing computation, it will be useful to represent a more representative version of h_p , known as a one hot encoding.
- ▶ $e(h_p)$ denotes the $\sum_{f=1}^F L_f$ length vector where the $\ell + \sum_{f=1}^F L_f$ component is 1 when $h_p = \ell$ and otherwise is 0.

Consider five fields with binary agreement patterns (and potential missingness). Suppose that records (5,7) in have agreement pattern (1,1,1,NA,2), which means

- agreement in the first three fields.
- missingness in the fourth field.
- complete disagreement in the fifth field.

- 1. Find $h(\gamma_{5,7})$
- 2. Find $e(h(\gamma_{5,7})$.

Recall that

$$h_f^{(i,j)} = I_{obs}(\gamma_{ij}^f) 2^{\gamma_{ij}^f + I(f>1) \times \sum_{e=1}^{f-1} (L_e-1))} \implies$$

$$h_1^{(5,7)} = 2^1 (6)$$

$$h_2^{(5,7)} = 2^3 (7)$$

$$h_3^{(5,7)} = 2^5 (8)$$

$$h_4^{(5,7)} = 0 (9)$$

$$h_4^{(5,7)} = 2^{10} (10)$$

This implies $h^{(5,7)} = 1066$. Assume this maps to unique integer 42.

How do we create the one hot encoding?

For a binary comparison, we have the following options:

- 1. (1,0): represents complete agreement of the record pairs.
- 2. (0,1): represents complete disagreement of the record pairs.
- 3. (0,0): represents missingness in the record pairs.

Given this,

$$e(h^{(5,7)}) = (1,0,1,0,1,0,0,0,0,1).$$