

### Introduction

- Data collection is booming.
- Personal microdata are published after anonymization.
- Anonymized data are not truly private.
- Correlated public data can be exploited for de-anonymization!
- Database Matching

### **Applications of Database Matching**

- Data & network privacy
- Computer vision
- DNA sequencing
- Single-cell biological data alignment

#### System Model

- ▶ Unlabeled Database:  $\mathbf{D}^{(1)} \in \mathfrak{X}^{m_n \times n}$  with
- ► I.I.D. rows following a first-order stationary Markov process capturing correlation among attributes.
- $\blacktriangleright$  Probability transition matrix **P** over  $\mathfrak{X}$

$$\mathbf{P} = \gamma \mathbf{I} + (1 - \gamma) \mathbf{U}$$

$$U_{i,j} = u_j > 0, \forall (i,j) \in \mathfrak{X}^2$$

- $\blacktriangleright$   $\pi$ : stationary distribution of  $\mathbf{P}$ .
- Labeling Function: Uniform permutation  $\Theta_n$  of  $[m_n]$ .
- Synchronization Errors: Random column repetition pattern  $S^n \stackrel{\text{i.i.d.}}{\sim} p_S, \delta \triangleq p_S(0)$ .
- ► Labeled Database: Pair  $(\mathbf{D}^{(2)}, \Theta_n)$  with

$$D_{i,j}^{(2)} = egin{cases} E, & ext{if } S_j = 0 \ D_{\Theta_n^{-1}(i),j}^{(1)} \otimes \mathbbm{1}^{S_j} & ext{if } S_i \geq 1 \end{cases}$$

- ► Database Growth Rate:  $R = \lim_{n \to \infty} \frac{\log_2 m_n}{n}$ .
- Matching: Estimation of  $\Theta_n$ .

# Matching of Markov Databases Under **Random Column Repetitions**

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Objectives

- What are the sufficient and the necessary conditions on the database growth rate for successful matching?
- $\blacktriangleright$  Can we infer the repetition pattern  $S^n$  from  $(\mathbf{D}^{(1)}, \mathbf{D}^{(2)})$ ? If yes, how?

Main Result

Databases with growth rate R can be successfully matched if R < C where

$$egin{aligned} \mathcal{C} & \triangleq rac{(1-\delta)(1-\gamma)}{(1-\gamma\delta)}[\mathcal{H}(\pi) + \sum_{i\in\mathfrak{X}}u_i^2\log u_i] \ & -(1-\delta)^2\sum_{r=0}^\infty \delta^r\sum_{i\in\mathfrak{X}}u_i(\gamma^{r+1} + (1-\gamma^{r+1})u_i)\log(\gamma^{r+1} + (1-\gamma^{r+1})u_i) \end{aligned}$$

Furthermore, a necessary condition for the existence of a successful matching scheme is  $R \leq C$ .

### Achievability-I: Histogram-Based Repetition Detection

$\mathbf{D}^{(1)}$				
a b a c c b	$\mathbf{TT}(1)$	$\mathbf{D}^{(2)}$		
	$\mathbf{H}^{(+)}$			
	$\begin{bmatrix} 3 & 1 & 1 & 2 & 0 & 0 \\ 1 & 2 & 1 & 2 & 0 & 0 \end{bmatrix}$			
	$\rightarrow$ $\begin{bmatrix} 1 & 3 & 4 & 4 & 3 & 4 \\ 4 & 4 & 2 & 2 & 5 \end{bmatrix}$			
		$\begin{bmatrix} a & a & b & b \\ c & c & c & c & b & b \end{bmatrix}$		
	$\begin{array}{cccccccccccccccccccccccccccccccccccc$	c $c$ $b$ $b$ $b$ $b$ $c$		
0 0 0 <u>a</u> c c	( <b>1</b> )	$\frac{b \ b \ b \ b \ b \ c}{(2)}$		
	$\mathbf{H}^{(\perp)}$	$\mathbf{H}^{(2)}$		
	3 1 1 2 0 0	$3 \ 3 \ 1 \ 1 \ 1 \ 2 \ 0$		
	$1 \ 3 \ 4 \ 4 \ 3 \ 4$	$1 \ 1 \ 4 \ 4 \ 4 \ 4 \ 4$		
	4 4 3 2 5 3	4 4 3 3 3 2 3		
	$0 \ 0 \ 0 \ 0 \ 1$	$0 \ 0 \ 0 \ 0 \ 0 \ 1$		
	$\hat{S}^n = \begin{bmatrix} 2 & 0 & 3 & 1 & 0 & 1 \end{bmatrix}$			

	$\mathbf{D}^{(2)}$		
$D_{2,3}$	$D_{2,5}$	$D_{2,5}$	$D_{2,6}$
$D_{3,3}$	$D_{3,5}$	$D_{3,5}$	$D_{3,6}$
$D_{4,3}$	$D_{4,5}$	$D_{4,5}$	$D_{4,6}$
$D_{1,3}$	$D_{1,5}$	$D_{1,5}$	$D_{1,6}$
$D_{5,3}$	$D_{5,5}$	$D_{5,5}$	$D_{5,6}$





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### Lemma: Asymptotic Uniqueness of The Histograms

- As long as  $m_n = \omega(n^4)$
- column histograms are asymptotically
- histogram-based repetition detection is

### Achievability-II: Matching Scheme



### **Converse**

- ► A genie aided proof, assuming the repetition pattern  $S^n$ .
- Provides insight into privacy-preserving anonymized data sharing/publication

### Conclusion

- A wide range of applications of database
- Existence of an underlying structure
- Column histograms of the databases are asymptotically unique.
- Histograms help us infer the repetition
- A tight bound on the achievable database growth rates.