

A Bayesian Approach to Linking Historical Records of America's Enslaved

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Overview

Goal: Develop an effective probabilistic record linkage model that leverages the presence of alias information in historical slave records.

- Define Record Linkage
- Record Linkage for Historical Slave Records
- Model Specification & Method
- Simulation Performance
- Application & Results

What is Record Linkage?

A method used to identify **co-referent** entities from overlapping datasets.

Author	Title	Publ. Date	Author	Title
Stephen Jay Gould	The Mismeasure of Man	1996	Ruha Benjamin	Race After Technology
lan Hacking	Taming of Chance	1990	 lan Hackng	The Taming of Chance
Stephen T. Ziliak, Deirdre N. McCloskey	The Cult of Statistical Significance	2008	 Theodore M. Porter	The Rise of Statistical Thinking
Theodore Porter	Trust in Numbers	2020	 Michael E. Staub	The Mismeasure of Minds

Deterministic Record Linkage

- Define rules to match field values
- Eq: **JOIN** in SQL

Probabilistic Record Linkage

- Statistical model used to estimate matches between records
- More flexibility for error-prone data
- Eg: Bayesian Bipartite Record Linkage (BRL) (Sadinle 2017)

Comparison Data

 Calculate distances between values of a field with a chosen similarity metric

- Classify distances as levels of disagreement (Fellegi & Sunter, 1969)
- Level of disagreement | field ~ Multinomial Mixture

Assumption

 There are no duplicate records within a single file (Sadinle, 2017)

Publ. Date

2019

1990

2020

2018

Challenges of Linking Slave Records

1 entity, 1 alias

1 entity, 3 aliases

Data Availability

- Comparatively few digitized data sources
- · High rates of missing data
- Eg: Last names

Uncertainty

- Data collection is unregulated/nonstandardized
- · Falsified/contentious/subjective values
- Transcription errors
- Eg: Ages

Duplicate Records (Non-Erroneous)

- Many entities have multiple variable values for a field (Aliases)
- Potential violations of bipartite linkage structure

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		_		

	ID (LK)	Enslaved Name	Racial Descriptor	Age	Kin Relation	Kin Name
=	PELK100940	Cook, James	"mulatto"	25	<na></na>	<na></na>
	PELK100498	Lucy	"negro"	37	Mother of	Harrison
\dashv	PELK100498	Lucy	"negro"	37	Mother of	Arthennise
	PELK100498	Lucy	"negro"	37	Mother of	[unnamed child]

From: Oceans of Kinfolk (Kinfolkology)

ID (OOK)	Enslaved Name	Racial Descriptor	Age	Kin Relation	Kin Name
OOK104652	Cook, James	"yellow"	20	<na></na>	<na></na>
ООК106770	Lewis, Lucy	"black"	35	<na></na>	<na></na>
OOK109450	Lucy	"dark"	40	<na></na>	<na></na>
OOK115546	Lucy	"black"	38	Mother of	<na></na>

MARY EPPS

45

I cannot say much about the place as I have ben here but a short time but so far as I have seen I like very well. you will give my Respect to your lady, & Mr & Mrs Brown. If you have not written to Petersburg you will please to write as soon as can I have nothing More to Write at present but yours Respectfully EMMA BROWN (old name MARY EPPS).

- The Underground Railroad

- Louisiana Kindred, Kinfolkology

Existing Approaches

Record Linkage with Naïve Pre-Processing

- Remove alias values/records during data processing
- Perform record linkage as usual

	ID (LK)	Enslaved Name	Racial Descriptor	Age	Kin Relation	Kin Name
	PELK100940	Cook, James	"mulatto"	25	<na></na>	<na></na>
	PELK100498	Lucy	"negro"	37	Mother of	Harrison
	PELK100498	Lucy	"negro"	37	Mother of	Arthennise
_	PELK100498	Lucy	"negro"	37	Mother of	[unnamed child]

Record Linkage with Post-Processing

- Keep all records and values
- Perform record linkage as usual
- Resolve conflicts after estimation

ID (LK)		ID (OOK)	
PELK100940	Cook, James —	OOK104652	Cook, James
PELK100498	Lucy —	OOK106770	Lewis, Lucy
PELK100498	Lucy —	OOK109450	Lucy
PELK100498	Lucy —	OOK115546	Lucy

Bayesian

Bipartite Alias Record Linkage (BARL)

 Specify a Bayesian BRL model that accommodates alias records (BARL) without introducing conflict

Comparisons

$$oldsymbol{\gamma}_{ij}^f \mid oldsymbol{m}_f, oldsymbol{u}_f, oldsymbol{Z} \overset{ ext{iid}}{\sim} \operatorname{Mult}(1, K_f, oldsymbol{m}_f^{oldsymbol{Z}_i = j} oldsymbol{u}_f^{oldsymbol{Z}_i
eq j})
onumber \ \Gamma \mid oldsymbol{m}, oldsymbol{u}, oldsymbol{Z} \sim \prod_{i=1}^{N_1} \prod_{j=1}^{N_2} \prod_{f=1}^F \prod_{k=1}^{K_f} \left[(oldsymbol{m}_f(k))^{\mathbf{1}(oldsymbol{Z}_i = j)} (oldsymbol{u}_f(k))^{\mathbf{1}(oldsymbol{Z}_i
eq j)}
ight]^{\gamma_{ij}^f(k)} oldsymbol{V}_{ij}$$

Level of Disagreement Probabilities

$$m{m}_f \mid m{lpha}_f \sim \mathrm{Dirichlet}(m{lpha}_f) \qquad m{u}_f \mid m{eta}_f \sim \mathrm{Dirichlet}(m{eta}_f)$$

2. Choose a linkage prior for \boldsymbol{Z} that equally weights links between any two **records**

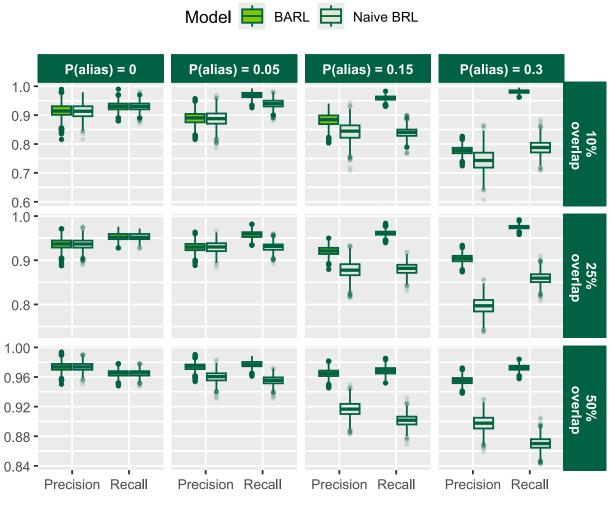
$$P(\boldsymbol{Z} \mid a, b) \propto \frac{\mathrm{B}\left(a + L(\boldsymbol{Z}), b + N_1 - L(\boldsymbol{Z})\right) \left(N_2 - L(\boldsymbol{Z})\right)!}{\mathrm{B}(a, b) N_2!} \mathbf{1}\left(\boldsymbol{Z}_i \neq \boldsymbol{Z}_{i'} \forall i \neq i'\right)$$

3. Use MCMC to generate posterior samples

- 1. Initialize agreement probabilities and links
- For each iteration, t = 1, . . ., T:
 Sample Z given data, m, u
 Sample m, u given data, Z
- 4. Estimate links directly from MCMC posterior samples or consolidate records to obtain point estimate

$$\widehat{\boldsymbol{Z}}_i = \begin{cases} j & \text{if } P(\boldsymbol{Z}_i = j \mid \boldsymbol{\Gamma}) > 1/2 \\ N_2 + i & \text{otherwise} \end{cases}$$

Model Performance Simulation



Data Simulation

- Records generated using GeCo
- Simulated overlapping datasets with aliases
- BARL compared to BRL with additional aliases removed (Naïve BRL)
- When no aliases are present, BARL = BRL

Results

 BARL: Greater average precision & recall than Naïve BRL

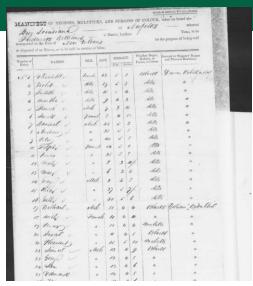
Applying BARL to Kinfolkology Data

Oceans of Kinfolk

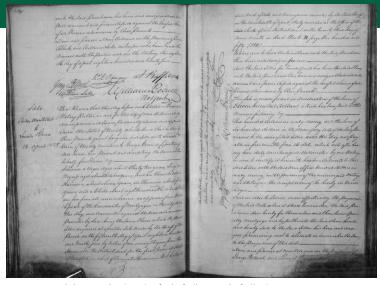
> 20,000 records of enslaved individuals trafficked by ship between domestic ports in the U.S. – transcribed & digitized

Louisiana Kindred

~ 1,500 records of individuals sold and bought in antebellum New Orleans – transcribed & digitized



From: Oceans of Kinfolk on Kinfolkology



From: Louisiana Kindred of Kinfolk on Kinfolkology

www.kinfolkology.org/data-overview



Aligned Variables

- Enslaved Name (first, last)
- Enslaved Age
- Enslaved Gender
- Enslaved Kin Name
- Enslaved Kin Relation
- Event Date
- Enslaver Name (first, last)
- Enslaver Location (city, county, state)



Thank You!

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Sources

- Fellegi, Ivan P., and Alan B. Sunter. "A Theory for Record Linkage." *Journal of the American Statistical Association* 64, no. 328 (1969): 1183–1210. https://doi.org/10.2307/2286061.
- Kinfolkology. "Kinfolkology." Accessed October 30, 2024. https://www.kinfolkology.org.
- National Center for Health Statistics. Division of Analysis and Epidemiology., "The Linkage of National Center for Health Statistics Survey Data to Centers for Medicare & Medicaid Services Transformed Medicaid Statistical Information System Claims Data (2014-2019): Matching Methodology and Analytic Considerations."
- Sadinle, Mauricio. "Bayesian Estimation of Bipartite Matchings for Record Linkage." *Journal of the American Statistical Association* 112, no. 518 (April 3, 2017): 600–612. https://doi.org/10.1080/01621459.2016.1148612.
- Still, William. The Underground Railroad: A Record of Facts, Authentic Narratives, Letters &c., Narrating the Hardships, Hair-Breadth Escapes and Death Struggles of the Slaves in Their Efforts for Freedom. Rev. ed. Philadelphia, Pa., Cincinnati, Ohio [etc.]: People's publishing company, 1879. https://www.loc.gov/resource/rbc0001.2019gen24984/.