The Synthetic Longitudinal Business Database

Based on presentations by Kinney/Reiter/Jarmin/Miranda/Reznek²/Abowd on July 31, 2009 at the

Census-NSF-IRS Synthetic Data Workshop

[link] [link]

Kinney/Reiter/Jarmin/Miranda/Reznek/Abowd (2011) "Towards
Unrestricted Public Use Microdata: The Synthetic Longitudinal Business

Database.", CES-WP-11-04

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Overview

- LBD background
- Synthetic data generation
- Analytic validity
- Confidentiality protection
- Future plans

Elements

(Economic Surveys and Censuses)

Issue: (item) nonresponse

Solution: LBD

Match-merged and completed complex integrated data Issue: too much detail leads to disclosure issue Solution: Synthetic LBD

Public-use data With novel detail

(Business Register)

Issue: inexact link records

Solution: LBD

Novel analysis using Publicuse data with novel detail Issue: are the results right Solution: Early release/SDS

The ("Real") LBD

- Economic census covering nearly all private non-farm business establishments with paid employees
 - Contains: Annual payroll and Mar 12 employment (1976-2005), SIC/NAICS, Geography (down to county), Entry year, Exit year, Firm structure
- Used for looking at business dynamics, job flows, market volatility, international comparisons...

Longitudinal Business Database(LBD)

- Detailed description in <u>Jarmin and Miranda</u>
- Developed as a research dataset by the U.S.
 Census Bureau Center for Economic Studies
- Constructed by linking annual snapshot of the Census Bureau's Business Register (see Lecture 4)

Longitudinal Business Database(LBD)

- CES constructed
 - longitudinal linkages (using probabilistic matching, see <u>Lecture 10</u>),
 - re-timed multi-unit births and
 - dealt with missing data

Access to LBD data

- Different levels of access
 - Public use tabulations Business
 Dynamics Statistics
 http://www.ces.census.gov/index.php/bds
 - "Gold Standard" confidential microdata available through the Census Research Data Center Network
 - (LBD in RDC)
 - Most used dataset in the RDCs

Bridge between the two

- Synthetic data set
 - Available outside the Census RDC
 - Providing as much analytical validity as possible
 - Reduce the number of requests for special tabulations
 - Aid users requiring RDC access
- Experiment in public use business microdata

Why synthetic data?

- Concerns about confidentiality protection for census of establishments
 - LBD is a test case
- Criteria given for public release:
 - No actual values of confidential values could be released
 - Should provide valid inferences while protecting confidentiality

Generic structure

- Gold standard: given by internal LBD (already completed)
- Partially synthetic:
 - Unsynthesized:
 - County (but not released!) [x1]
 - SIC [x2]
 - Synthesized
 - Birth [y1] and death [y2] year:
 - Multi-unit status [y3]
 - Employment (March 12) [y4]
 - Payroll [y5]

Synthesis: General Approach

- Y=[y1|y2|y3|y4|y5]
- X=[x1|x2]
- Generate joint distribution of Y|X by sampling from conditionals
 - $-f(y1,y2,y3|X) = f(y1|X)\cdot f(y2|y1,X)\cdot f(y3|y1,y2,X)$
- Use SIC as "by" group

General approach to synthesis

- Drawing from $f(y_k|X,y_1,...,y_{k-1})$
 - Fit model using observed data
 - Draw new values of parameters from posterior distributions
 - Use new parameters to predict y_k from X and synthetic values of $y_1,...,y_{k-1}$

SRMI approach

- Calendar:
 - Step1: Impute y1 | X
 - Step 2: Impute y2 | [y1 | f(X)]
 - Where f(X) uses state [x1'] instead of county [x1]
- Type of firm
 - Step 3: Impute y3 | [y1|y2|X]
- Characteristics
 - Step 4: Impute y4(t) | [y1 | y2 | y3 | y4(t-1) | x2]
 - Step 5: Impute y5(t)|[y1|y2|y3|y4(t)|y5(t-1)|x2]

First Year

- Impute y1 (Firstyear) | SIC, County using variant of Dirichlet-Multinomial
 - "Prior" information is obtained by collapsing categories
 - Synthetic values obtained from sampling from multinomial distribution

Last Year

- Impute y2 (Last Year) | First Year, State, SIC
- Simple multinomial approach
 - Dirichlet-multinomial with flat prior
 - Sample from multinomial probabilities obtained from matching categories in observed data

Multi-unit Status

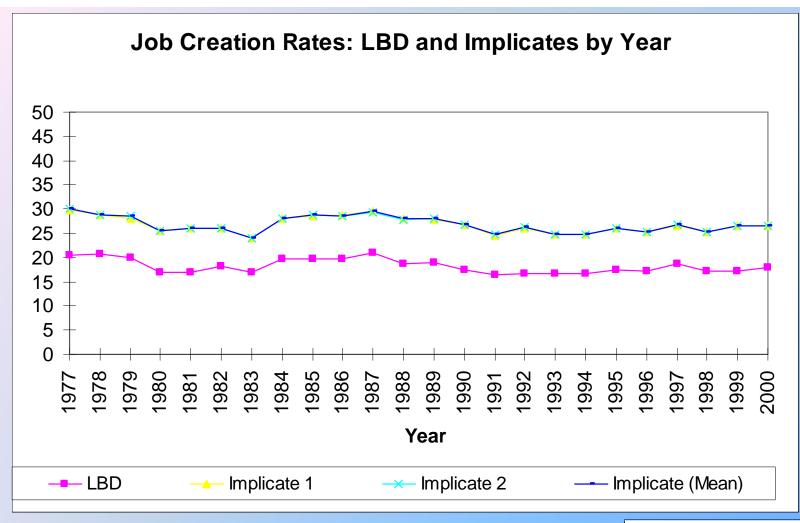
- Impute in two stages:
 - Categorical response: Always MU, sometimes MU, never MU
 - Imputed using simple multinomial approach
- Given change in status occurs, impute when change occurred (future)

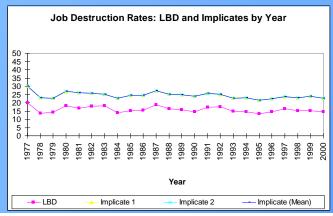
Employment and Payroll

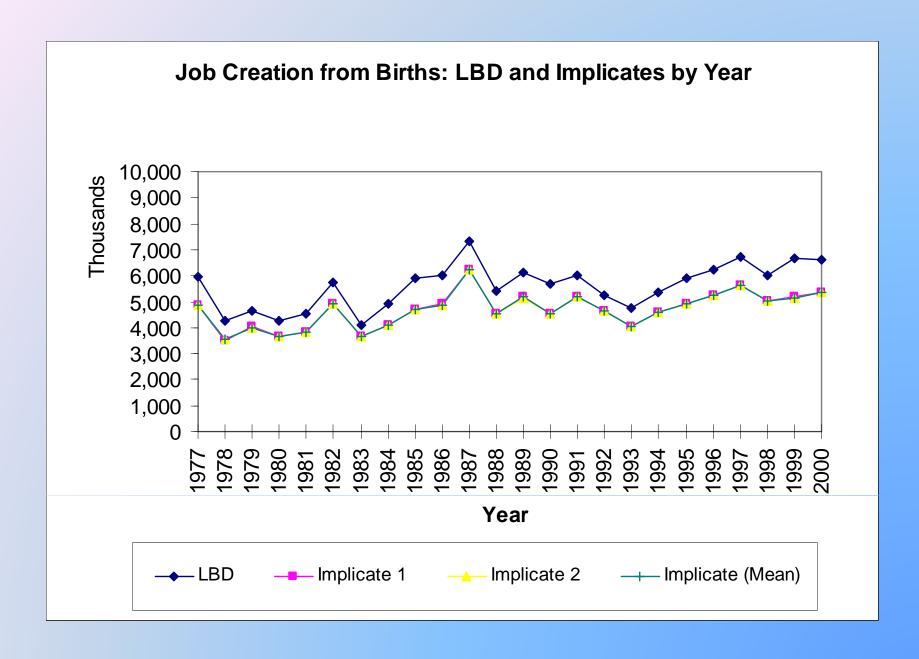
- Highly skewed longitudinal continuous variables
- Imputed using a set of normal linear models with kde transformation of response (Abowd and Woodcock, 2004)
- Impute year by year, employment and then payroll, based on groups
 - (3-digit SIC)
 - by (multiunit status)
 - by (continuer status)
 - by (top 5% status)
- If model too sparse, use 2-digit SIC as prior

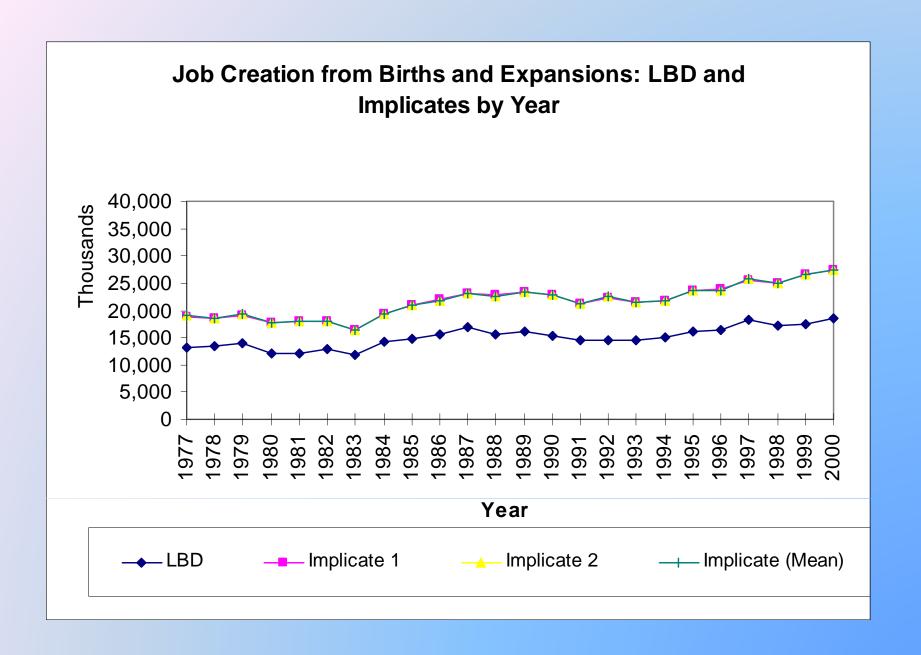
Analytic Validity Tests

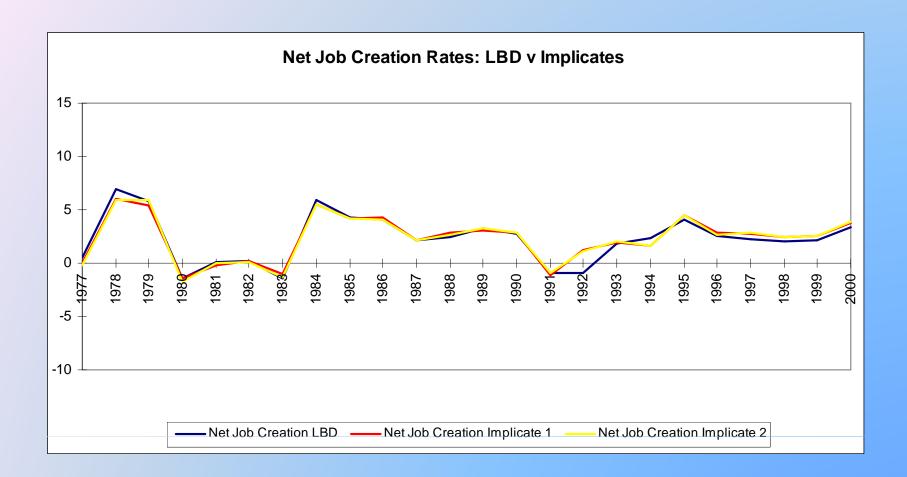
- Compare observed data and synthetic data for whole LBD
 - Job creation and destruction
 - Employment volatility
 - Gross employment levels

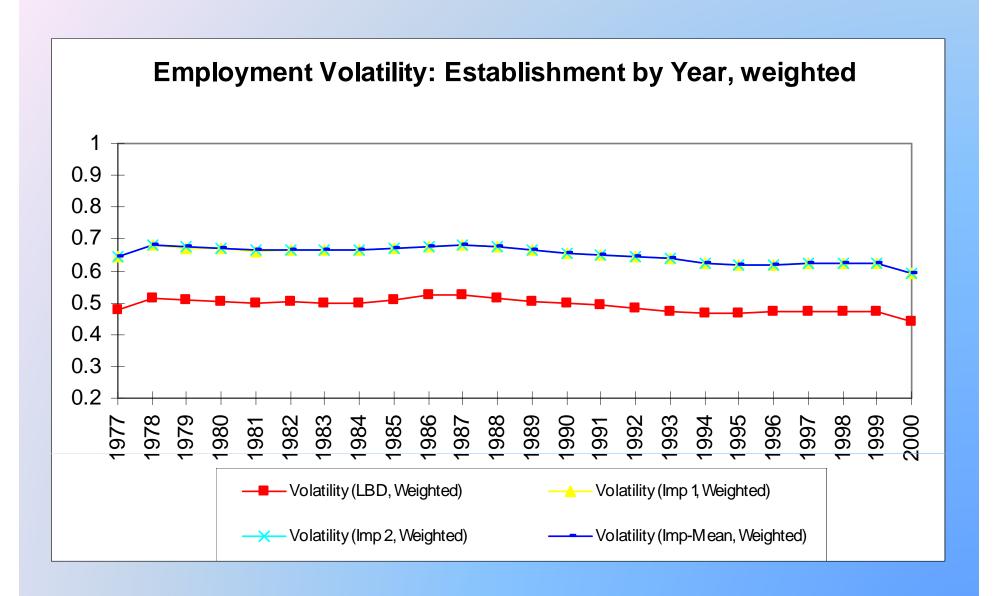


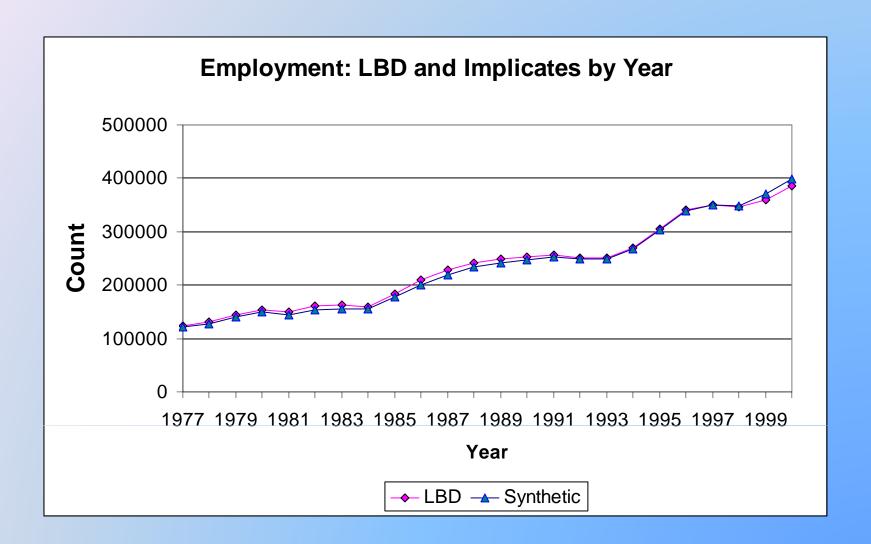












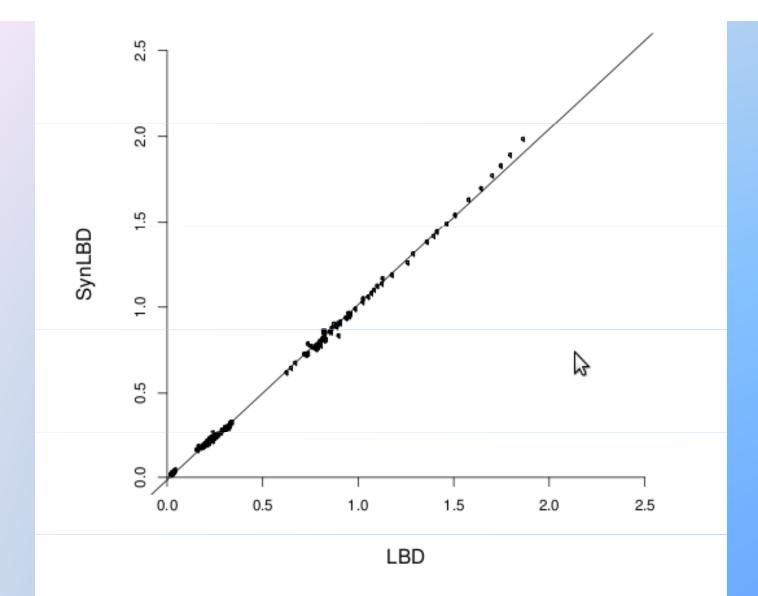


Figure 3: Share of Employment by Industry Sector and Year, 1976-2000

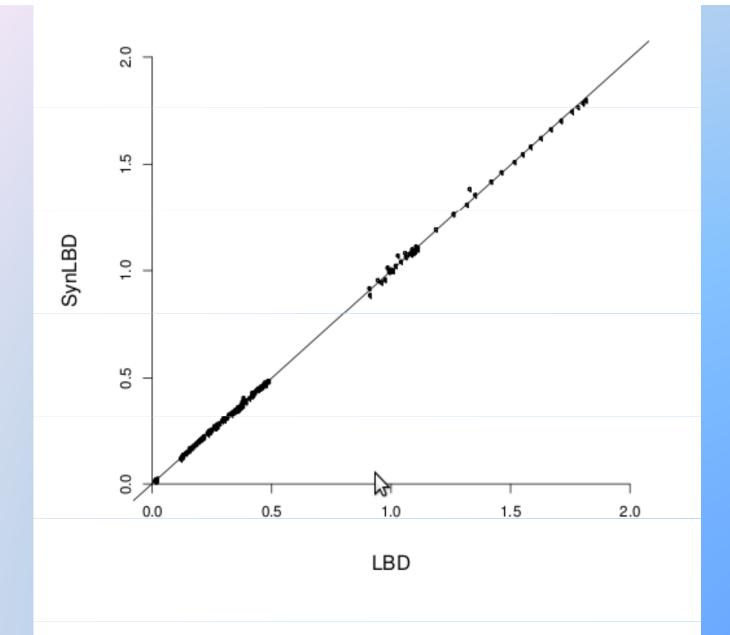


Figure 2: Share of Establishments by Industry Sector and Year, 1976-2000.

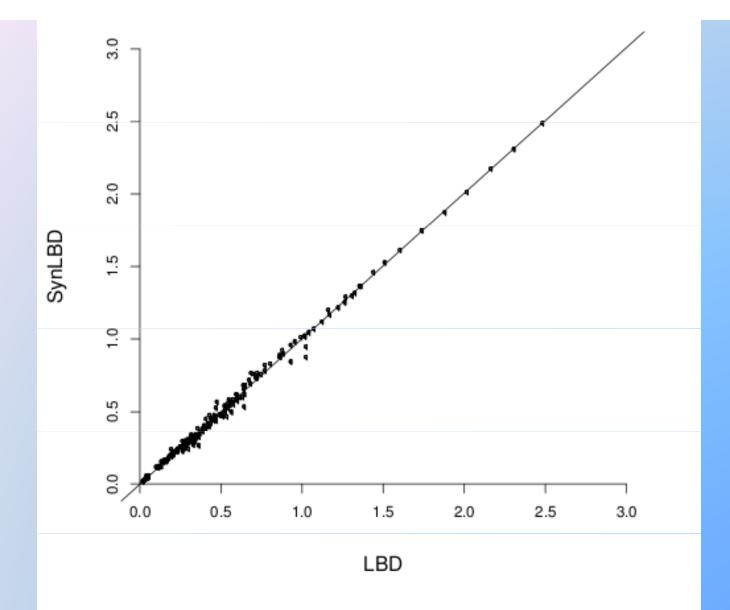


Figure 4: Share of Payroll by Industry Sector and Year, 1976-2000

$$EMP_i = \alpha + \beta EMP_{i-1} + \delta PAY_i + \theta IND_i + \psi STATE_i + \vartheta AGE_i + \gamma MU_i + \epsilon$$

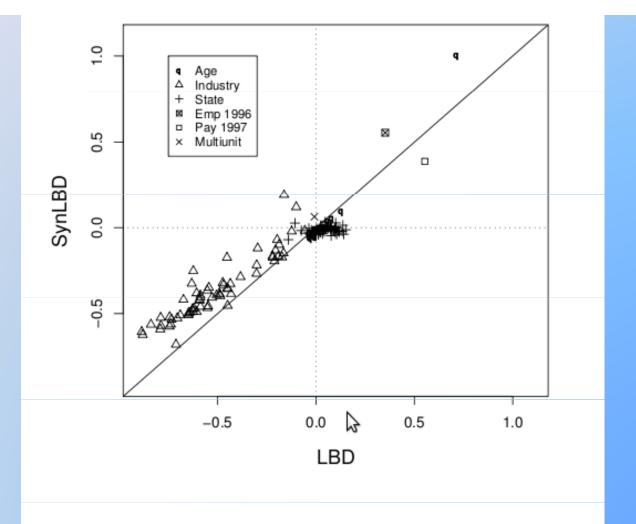


Figure 11: Regression Coefficients, LBD vs Synthetic

Confidentiality Protection

- Unavailable in SynLBD v2
 - Firm structure
 - firm linkages (across time, across implicates)
 - Geography
- Basic protection
 - replacing sensitive values of with draws from probability distributions

Disclosure analysis

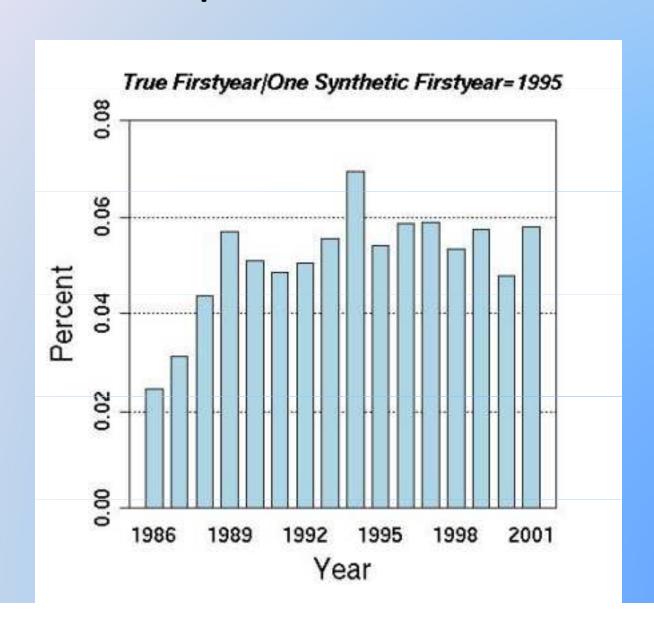
- High probability that an individual establishment's synthetic birth/death year is different from its actual birth/death year
- Synthetic maxima not necessarily near actual
- High between-imputation variability at establishment level

Synthesizing Firstyear (Birth) and Lastyear (Death)

- Positive probability exists of producing any feasible birth year, and substantial probability exists that synthesized firstyear is not the actual firstyear
- Table on next slide shows this: prob(actual birth year=synthetic birth year I synthetic birth year) is low
- Similar results hold for deaths
- Conclusions: establishment lifetimes are random, so users can't accurately attach establishment identifications to them

Summary Data: Observed Establishment Births Occuring in Same Year as Synthetic Births Synthetic Actual Minimum Mean Maximum 1975 25.41 88.89 1975 1.52 1976 1976 0.12 5.12 75.00 1977 5.09 71.43 1977 0.43 1978 1978 3.65 16.22 0.46 1979 1979 0.27 3.89 50.00 1980 1980 0.36 3.46 25.00 1981 0.26 3.91 50.00 1981 1982 1982 0.36 3.69 50.00 1983 0.39 50.00 1983 4.10 3.79 1984 1984 0.69 19.30 1985 23.73 1985 3.75 0.15 3.92 33.33 1986 1986 0.41 1987 1987 0.35 4.19 25.00 1988 1988 0.48 4.25 52.48 25.15 1989 1989 0.63 4.28 25.00 1990 1990 0.47 3.91 0.56 50.00 1991 1991 4.18 17.39 1992 1992 0.45 3.94 3.86 25.00 1993 1993 0.67 1994 1994 0.53 4.33 50.00 1995 1995 0.35 4.16 16.67 16.67 1996 1996 0.20 4.11 1997 1997 0.10 4.04 18.60 20.00 1998 1998 0.46 3.85 43.02 1999 1999 0.28 4.64 2000 2000 0.31 4.46 33.33 25.27 2001 2001 0.35 4.22

Example: Year of birth



Confidentiality Protection: Breaking Firm Links

- Firm characteristics not synthesized
- Firm characteristics more skewed than establishment characteristics
- Cannot link multi-unit establishments to their firms

Confidentiality Protection: Breaking Links Across Implicates

- Synthetic observations with the same LBDnum across implicates are not generated from the same LBD establishment
- Can't group (across implicates within year) observations generated from same establishment

Confidentiality Protection: Synthesizing Employment and Payroll

- Synthesis models are essentially regressions with transformed variables
- Synthesis captures low-dimensional relationships and sacrifices higher-dimensional ones
- Synthesized employment and payroll vary substantially around regression lines
- Synthesized employment and payroll vary significantly from observed values

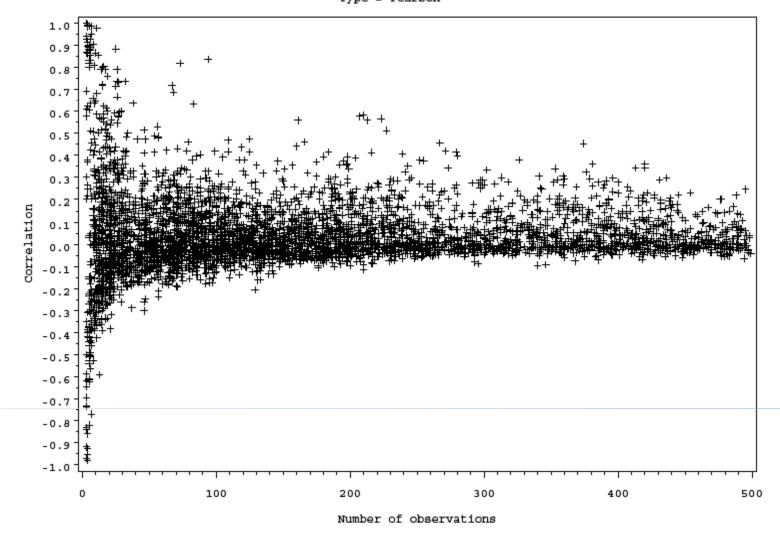
Example: Correlations Among Actual and Synthetic Data

SIC 573 - year 2000

Pearson Correlation Coefficients				
SIC 573				
Year: 2000				
		Synthetic		Synthetic
	Employment	Employment	Payroll	Payroll
Employment	1			
	41000			
Synthetic	0.003	1		
Employment	21100	41000		
Payroll	0.712	-0.012	1	
	41000	21100	41000	
Synthetic	0.007	0.444	0.004	1
Payroll	21100	41000	21100	41000

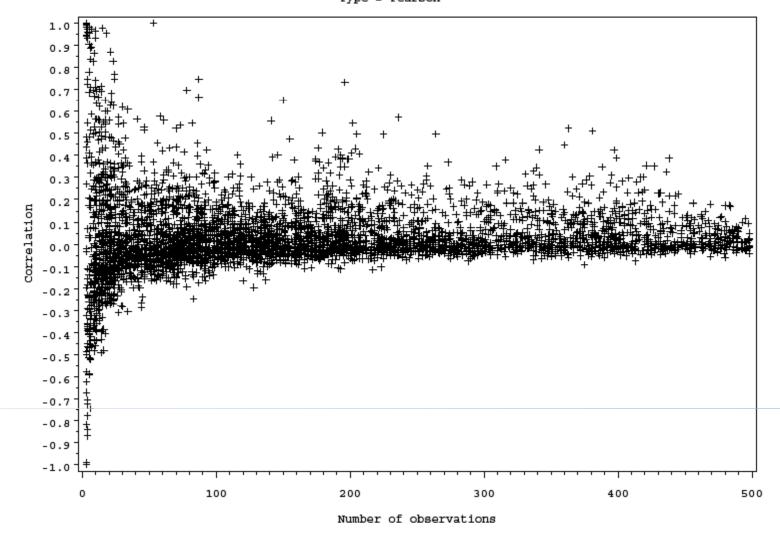
Correlations of observed vs synthetic Employment

Type = Pearson



Correlations of observed vs synthetic Payroll

Type = Pearson



Conclusions

- Analytical validity supported for broad analyses
 - Issues with some details
 - Obtain user feedback to inform future refinements
- Sufficient confidentiality protection
 - Basic metrics show strong protection
 - Differential privacy protection not yet verified

Ongoing work at Census

- Include NAICS, geography, changes in multiunit status, firm age & size
- Multiple Imputations for release
- Address bias in job creation/destruction
- Extend time series