

The use of Administrative Records is increasing in many agencies, and is being studied at Census for use in 2020.

It can save a lot of money, but it's real value ("goodness") depends on matching quality, and further, how it is tuned for optimal use – both requiring testing.



Think of F_1 as Census data, and F_2 as tax data, with duplicates removed...

Typically, N_1 and N_2 are comparable, but they don't have to be the same.

Think of "entities" as persons or households, typically.

If a Census record is correctly matched with a tax record, say, then improved Census data can result; however, if the linkage is incorrect, it could be made worse.

System Positive Predictions	System Negative Predictions	Row Sums
True Positives	False Negatives (Type II Error)	True Positive Matches
False Positives (Type I	True Negatives	True Negative Matches
Predicted Positive Matches	Predicted Negative Matches	
	System Positive Predictions True Positives False Positives (Type I Predicted Positive Matches	System Positive PredictionsSystem Negative PredictionsTrue PositivesFalse Negatives (Type II Error)False Positives (Type ITrue Negatives (Type II Error)False Positives (Type IPredicted Negatives (Negative)Predicted Positive Negative MatchesPredicted Negative

If you run a test and can't estimate <u>all</u> the numbers in the black box, you haven't run a good enough test and can't optimize your RL system!

It is not sufficient to just say one has more "matches", as that does not tell you how many of your Predicted Positive Matches are False Positives.

Digging deeper, you need to know how many matches "escaped" your system, (the False Negatives).

If you are able to test your RL system in a given state and determine all the elements in the matrix, than you can optimize RL system performance and maximize your return on investment.

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Both of these approaches are very cost-effective because they replace a lot of human effort with automation.

Basic Idea for Today

- The GAMUT technology is best for system <u>development</u> testing, when good data, designed for test, is generally private or unavailable or both (Ref.1)
- The PDQ technology is best for <u>production</u> testing with real data to determine how good your matching really is and learn how to improve it (Ref.2)
- They can overlap to provide enriched testing and optimization benefits as you phase out of development and get into production

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Actually, synthetic data can be **better** for development testing than real data!!!

This is because you know the truth and can introduce engineered errors or variations that need to be tested.



I prefer to think about the SUT first, and then consider what kind of data is needed for a particular test plan.

The job of the SUT is to ingest various data streams from the GAMUT and infer some facts about the GAMUT that are not apparent, like does a person in one data stream match a person in another data stream.

Usually, looking at these data streams gives you only a little "peek" at what's really in the GAMUT.



Four modes of data capture are: paper (self-administered questionnaire), internet, telephone (CATI), and enumerator (CAPI).

Actually, because synthetic data is DESIGNED FOR TEST, it is actually better for testing than real data, especially in the development stages.

Basic RLPDQ Concept

- By using an independent RL system that has fundamentally different characteristics and approaches than the production RL system, one can bring automation to bear on this difficult and costly testing problem (Ref.3)
- The <u>key</u> is to cost-effectively get from "comparison space" which is of order N² to "entity match space" which is of order N

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The independent RL system can, say, use different technology to estimate matches, weight data fields differently, and perhaps use different blocking techniques.

Using automation greatly reduces the testing workload and cost.



The data source supplies the two files – census and tax data, say.

Matches both systems agree upon are highly likely to be correct, and this is the bulk of the answers you seek.

Even if the two engines don't agree, most of the time ONE got it right!

Arbitration on what's left involves humans looking at entity pairs, using automation to reduce effort.



So, say both files are roughly only a thousand or so records (N); even then, the number of record pairs that must be examined is about a million (N^2).

The primary PDQ job is to quickly and efficiently get you focused on "Entity Match Space" (order N), rather than "Comparison Space" (order N²).

This is an actual example result "to scale" - It's a little hard to see the small blue blob, so...



Here is a close-up of our little blue blob...

This is "entity matching space" detailed in Reference 3.

NOT to scale – usually the overlap region is most of it.

Getting a handle on False Negatives can help tune your system for maximum value.



So, now suppose you use both of these test technologies -



Ideally, you can use both techniques as the project progresses from development through production.

In the overlap region, as production ramps up, you can learn more about your RL System by comparing both sets of test outputs, as they tend to discover different types of errors.

In particular, RLPDQ is effective at uncovering "escapes" (False Negatives) in your real production data.



Since the two methods get at False Negatives in different ways, that increases the chances that you uncover these "hard to find" errors.

Thank you!

1.	Paxton, K. Bradley, and Hager, Thomas, <i>Use of Synthetic Data in Testing Administrative Records Systems</i> , Proceedings, Federal Committee on Statistical Methodology (FCSM), Washington, DC, 2012		
1.	Paxton, K. Bradley, Spiwak, Steven P., Huang, Douglass, and McGarity, Jame K., <i>Testing Production Data Capture Quality</i> , Proceedings, Federal Committe on Statistical Methodology (FCSM), Washington, DC, 2012		
1.	Paxton, K. Bradley, <i>Testing Record Linkage Production Data Quality</i> , . In JSM Proceedings, Government Statistics Section. Montreal, Canada: American Statistical Association. Pgs. 1157-1171, 2013.		
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Shoot me an e-mail or give me your business card, and I'll send you these references.