# Use of Synthetic Data in Testing Administrative Records Systems

A presentation to the FCSM Tuesday, 10 Jan 2012



## Some Background on ADI, LLC

- ☐ Synthetic data from ADI was used in the 2010 Census for more cost-effective and precise testing of data capture
- ☐ This data was supplied in Digital Test Decks®, corresponding image files, and scripts for testing data capture modes other than paper
- ☐ Independently designed and developed a generic and powerful "Dynamic Data Generator™" (DDG) for creating synthetic test data
- ☐ Also doing medical (IBM) and intelligence (DARPA) synthetic data sets





## Security Aspects

- ☐ Program security around real data precludes engaging industry for scientific study, market research, and for consistent evaluation of multiple vendors
  - In Medical records, there are *HIPAA* laws
  - In Census records, there is *Title 13*
  - In IRS records, there is *Title 26*
  - In SSA records, there is the *Privacy Act of 1974 (5 U.S.C. § 552a)*
  - •
- ☐ Our synthetic data is realistic, but not real!





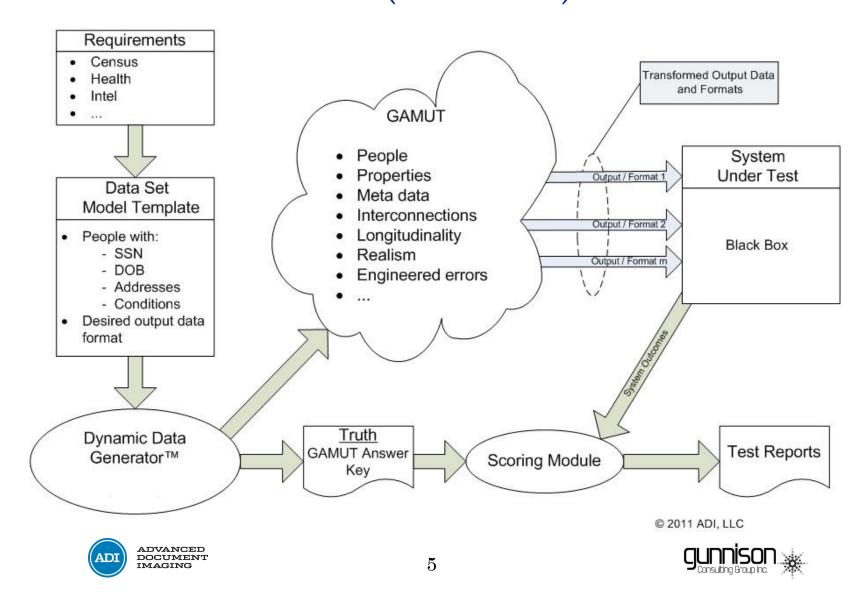
# Testing Administrative Records Systems with Synthetic Data

- ☐ Administrative Records will be very useful to Census, but testing the systems that are being developed to use them is extremely difficult
- ☐ Present testing approaches use large files of "real" data for which the "truth" is not known
- ☐ Synthetic, yet realistic data sets, <u>designed for test</u>, and for which the truth <u>is</u> known allows for quick, costeffective and precise testing and quantitative scoring
- ☐ Both true and false positives may be measured and used to improve systems in development

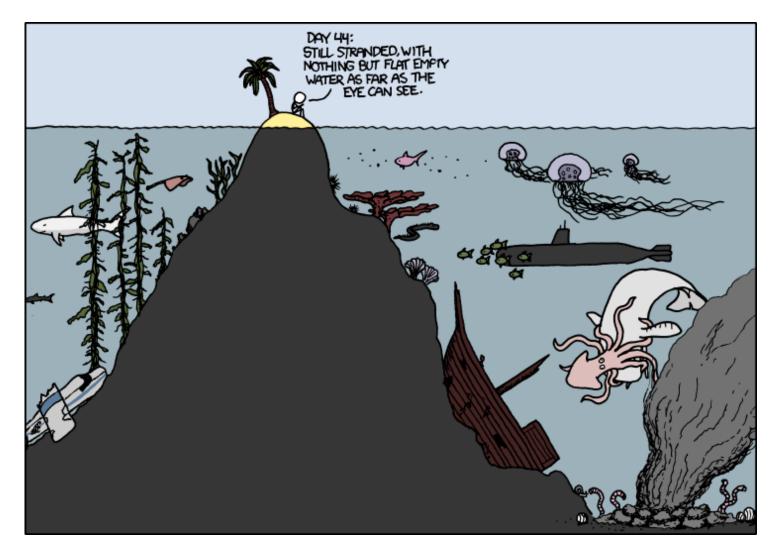




# Great Automated Model Universe for Test (GAMUT)



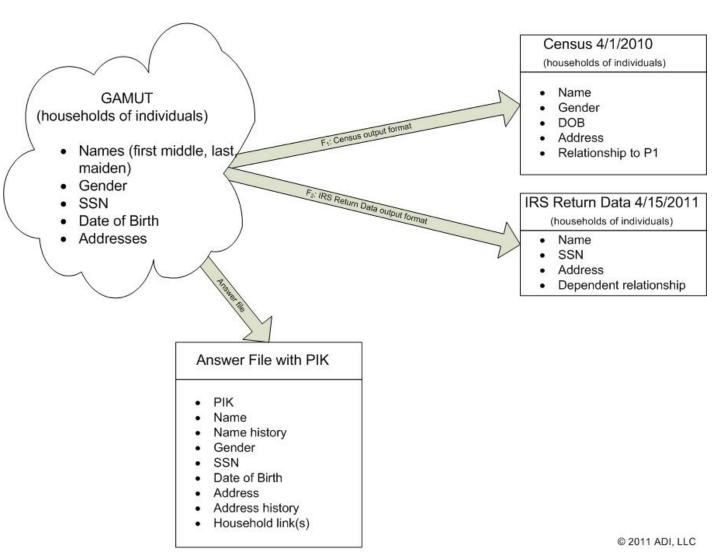
## A "Peek" at the GAMUT?







# Today's GAMUT Example







### Demo GAMUT Characteristics

- □(Only) about 1000 synthetic households generated for this demo GAMUT
- ☐ Two data feeds were made: Census and Tax (IRS)
- ☐Geographic scope:
  - DC, New Mexico, West Virginia





### Data Feed Characteristics

#### □Census Data Feed:

- Snapshot on 1 Apr 2010
- Names, DOB, Gender, Relationships
- Addresses
- PIK Numbers

#### □IRS Return Data Feed:

- Snapshot on 15 Apr 2011
- SSNs
- Names, Addresses
- Dependent Relationships





## Some GAMUT Demo "Features"

#### □ Census

- Dupes 2%
- Person 1 DOB missing or morphed (1-2%)
- Name morphing 2%
- Coverage 99%

#### $\Box$ Tax

- Filer SSN can be both husband and wife
- Filer name can be concatenation of both
- Moves 10%
- Coverage 85%





# Test Example: Person Matching

- Using this data, we explain how testing can be done using GAMUT and how to analyze the results with a classic Receiver Operating Characteristic (ROC) technique
- □ For this example, we are just looking at testing a hypothetical RL system that does matching of Census feed Person 1 to Tax Filers in Tax feed





## Test Plan: Person Matching

- $\square$ Output/Format 1 is  $F_1$  = Census Data
- $\square$ Output/Format 2 is  $F_2$  = IRS Tax Data
- □Say for each unique person in  $F_1$ , the System Under Test (SUT) is to predict the best person match(s) in  $F_2$  if any
- □Say there are N matches in the Truth, adding up both positive and negative matches
- ☐ The GAMUT Truth is M positive matches
  - $\triangleright$  Therefore  $M \le N$





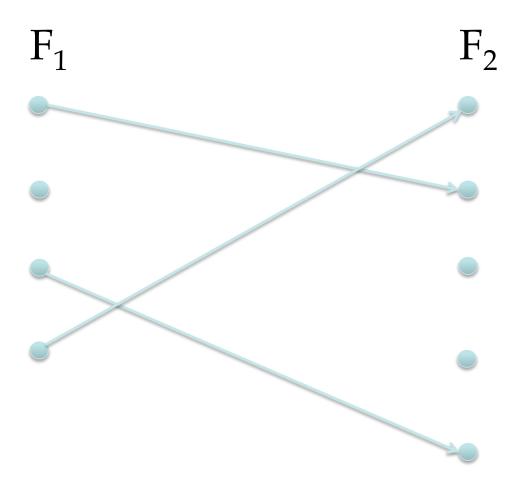
## Test Plan: Cont.

- □ The SUT predicts m matches  $(0 \le m \le N)$
- □Of the m matches, GAMUT Truth says cm of them are correct  $(0 \le c \le 1)$ : "True Positives"
- □ Therefore m cm = m(1 c) are "False Positives" (Type I errors)
- ☐Also, one can compute:
  - ➤"False Negatives" = M cm (Type II errors)
  - $\succ$  "True Negatives" = N M m(1 c)





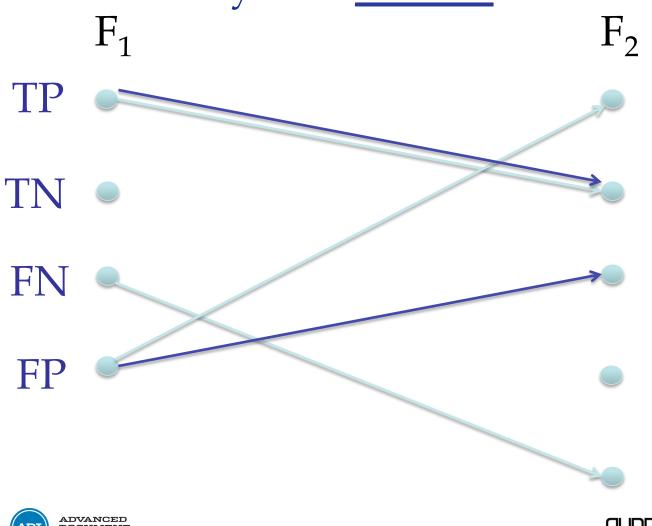
# Example of Test Truth







## Example of Test **Truth** with Classification System Results







## **Confusion Matrix**

|                |                   | SUT<br>Prediction  | <b>SUT Prediction</b>         | Row<br>Sums |
|----------------|-------------------|--------------------|-------------------------------|-------------|
|                |                   | Positive<br>Match  | Negative<br>Match             |             |
| Data<br>Truth  | Positive<br>Match | TP<br>cm           | FN<br>M - cm                  | M           |
| Data<br>Truth  | Negative<br>Match | <b>FP</b> m(1 - c) | <b>TN</b><br>N - M - m(1 - c) | N - M       |
| Column<br>Sums |                   | m                  | N - m                         | N           |

FP are Type I errors; FN are Type II





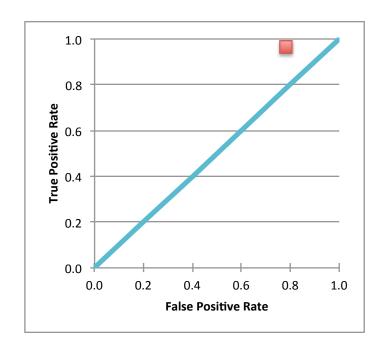
# Example Test - Case A

Generic ROC Plot and Confusion Matrix (Case A)

| N   | M   | m   | С      |
|-----|-----|-----|--------|
| 985 | 848 | 925 | 0.8843 |

|     | Prediction | _   |     |
|-----|------------|-----|-----|
|     | Pos        | Neg |     |
| Pos | 818        | 30  | 848 |
| Neg | 107        | 30  | 137 |
|     | 925        | 60  | 985 |

| TPR   | FPR   | Α     | f     |
|-------|-------|-------|-------|
| 0.965 | 0.781 | 0.861 | 0.923 |







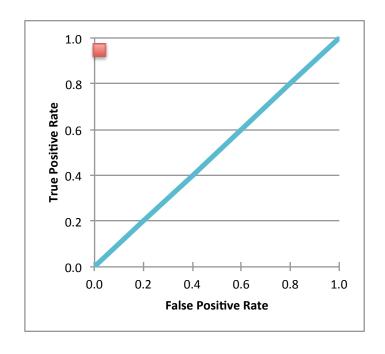
# Example Test - Case B

Generic ROC Plot and Confusion Matrix (Case B)

| N   | M   | m   | С      |
|-----|-----|-----|--------|
| 985 | 848 | 808 | 0.9963 |

|     | Prediction | _   |     |
|-----|------------|-----|-----|
|     | Pos        | Neg |     |
| Pos | 805        | 43  | 848 |
| Neg | 3          | 134 | 137 |
|     | 808        | 177 | 985 |

| TPR   | FPR   | Α     | f     |
|-------|-------|-------|-------|
| 0.949 | 0.022 | 0.953 | 0.972 |







## Conclusions

- ☐ The use of synthetic GAMUT testing data can significantly speed up and improve Administrative Records testing at Census, leading to improved system performance
- ☐ It can also help in other areas, for example:
  - Record Linkage Generally
  - Data Capture (all "modes")
  - Health Records Systems
  - Intelligence Systems
  - Census 2020 Research and Evaluations
- Remember, we don't aim to <u>replace</u> testing with "real" data, but rather to <u>supplement</u> it to speed up the development process to achieve quality software that's scalable and ready for production





## **Questions or Comments?**

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- □ADI Website:
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- ☐Sample data available on request



