

Synthetic Data Generation with library(tidysynthesis)

Aaron R. Williams

Confidential data

Sex	Age	Wages	Tax
Y ₁₁	y ₁₂	y ₁₃	y ₁₄
y ₂₁	y ₂₂	y ₂₃	y ₂₄



Goals

- Produce synthetic data file with the same record layout as administrative data that:
 - Protects the confidentiality of individual information
 - May be used for statistically valid analysis for certain research purposes
 - May be used as a "training dataset" to develop programs to run on confidential data or a formally private validation server

Estimating the multivariate distribution of the data

- Goal is to approximate the empirical multivariate distribution function for the data
- Joint multivariate probability distribution can be represented as the product of sequential, conditional probability distributions:

$$f(Y_1, Y_2, ..., Y_k | \theta_1, \theta_2, ..., \theta_k) =$$

$$f_1(Y_1 | \theta_1) \cdot f_2(Y_2 | Y_1, \theta_2) \cdots f_k(Y_k | Y_1, Y_2, \dots, Y_{k-1}, \theta_k)$$

• where Y_i the variables and θ_i are vectors of model parameters

Confidential data

Sex	Age	Wages	Tax
Y ₁₁	y ₁₂	y ₁₃	y ₁₄
y ₂₁	y ₂₂	y ₂₃	y ₂₄

Partially synthetic data (R. Little - 1993)

Sex	Age	Wages	Tax
Y ₁₁	y ₁₂	\hat{y}_{13}	$\hat{\mathcal{Y}}_{14}$
y ₂₁	y ₂₂	\hat{y}_{23}	\hat{y}_{24}

Confidential data

Sex	Age	Wages	Tax
Y ₁₁	y ₁₂	y ₁₃	y ₁₄
y ₂₁	y ₂₂	y ₂₃	y ₂₄

Partially synthetic data (R. Little - 1993)

Sex	Age	Wages	Tax
Y ₁₁	y ₁₂	$\hat{\mathcal{Y}}_{13}$	$\hat{\mathcal{Y}}_{14}$
y ₂₁	y ₂₂	\hat{y}_{23}	\hat{y}_{24}

Fully synthetic data (D. Rubin – 1993)

Sex	Age	Wages	Tax
\hat{y}_{11}	$\hat{\mathcal{Y}}_{12}$	$\hat{\mathcal{Y}}_{13}$	\hat{y}_{14}
$\widehat{\mathcal{Y}}_{11}$	\hat{y}_{12}	\hat{y}_{23}	\hat{y}_{24}

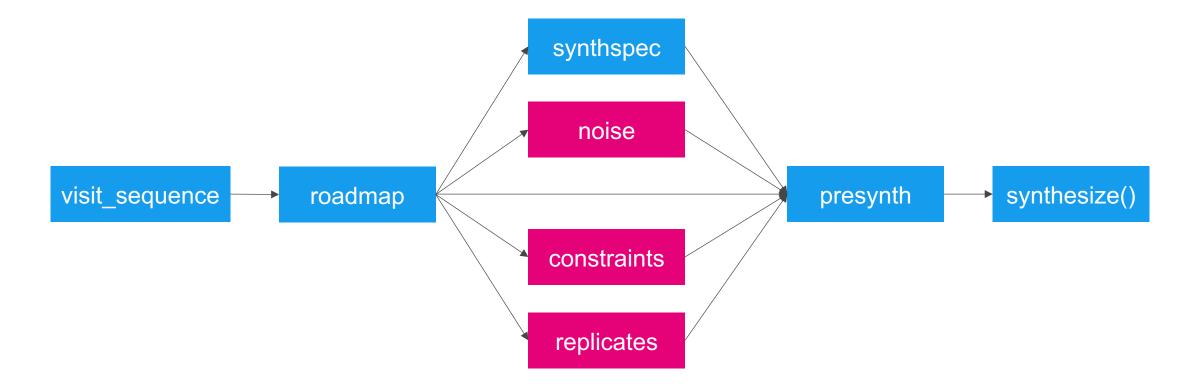
library(tidysynthesis)

New features

- Feature and target engineering with library(recipes)
- Model metrics
- Mid-synthesis constraints
- Weighted data

· URBAN·INSTITUTE ·

Workflow

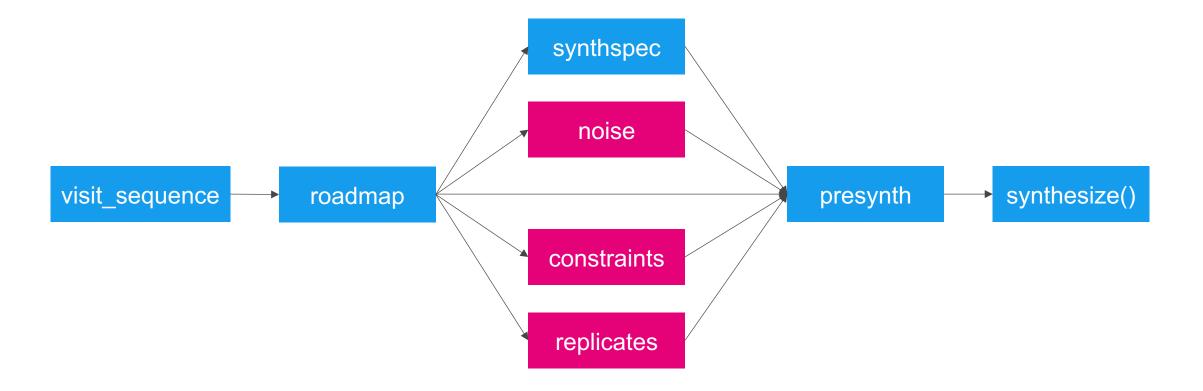


visit_sequence() and roadmap()

- Apply rules for determining a synthesis order
- Add starting data
- Add confidential data

· URBAN·INSTITUTE ·

Workflow



13

synthspec()

- Add model specifications
 - Feature and target engineering
 - Algorithms
 - Sampling methods
- Methods can vary from variable-to-variable

· URBAN·INSTITUTE·

14

noise()

Add noise to predictions beyond prediction error

15

constraints()

- Implement univariate and multivariate constraints
- Remedial measures:
 - Exclusions
 - Hard bounding
 - Z-bounding

· URBAN·INSTITUTE·

replicates()

- Create multiple synthetic data sets
- Run the synthesis process in parallel

Workflow

