



Research Applied Analytics & Statistics  
STATISTICS OF INCOME

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# Synthesizing the Supplemental Synthetic Public Use File

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October 24, 2024





# Transition to Synthetic PUF and Tiered Access

Tier	Access	To Whom
1	Tabular data and reports	Anybody – via website and published reports
2	Synthetic individual income information	Anybody who needs it – upon request to SOI
3	Validation server: Automated system allows researchers to access confidential tax return information in an environment that protects against disclosure	Researchers vetted by SOI with a research plan that could not be completed using tier 1 or tier 2 access.
4	Access to confidential microdata	Researchers approved for access through the Joint Statistical Research Program.



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# Taxpayer Privacy and Confidentiality

**Any publicly released tax data must protect the confidentiality of individual taxpayers.**

## Tabular released data

- Rule of 3
- Rule of 10
- Dominance Rule
- Associated Suppression
- Disclosure by subtraction
- Cross-cell disclosure
- Complimentary disclosure

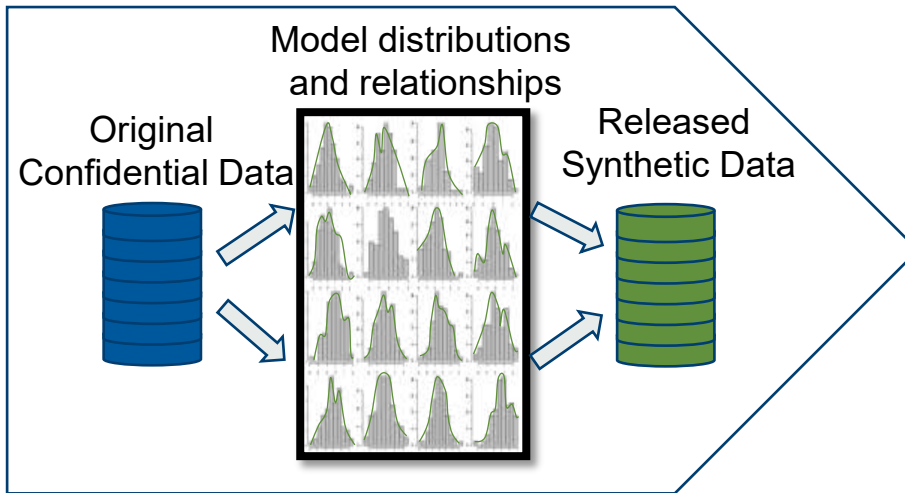
## Microdata release

- Subsampling
  - Reweighting
- Aggregation
- Top Coding
- Blurring
  - Multivariate
  - Univariate
  - Rebalancing
- Random Noise
  - Rounding
- Suppression

**As the scope of information on individuals that is publicly accessible increases, so too must SOI improve protection techniques.**

# Synthetic Data – General Approach

## General methodology



## No real observations are released

- Possibility of expanded demographic and/or tax information
- Possibility of multiple file releases targeting different population subsets

## Potential Pitfalls:

- Model overfitting may result in synthetic data too close to underlying data.
- Database Reconstruction Theorem (Dinur and Nissim, 2003): noisy subset sums can approximate individual records through solving a system of equations.
- Modeler may overcompensate for these concerns resulting in data without enough overlap to confidential data to be statistically useful.

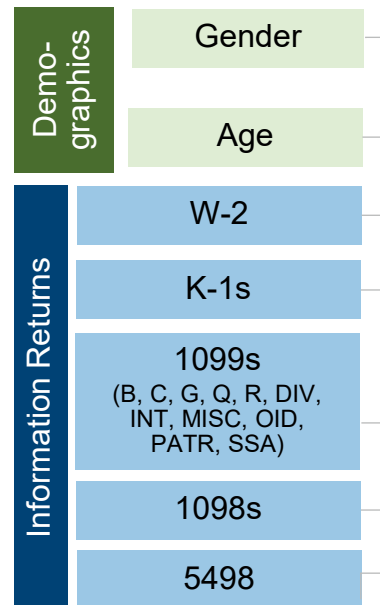
# Synthesis Process

## Sample Selection

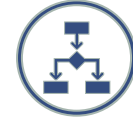


- Individuals receiving an information return in TY2012 with
  - No Form 1040
  - Income below filing threshold
- Based on Continuous Work History Sample (CWSHS)
  - A 1-in-1,000 sample
- Additional limitations:
  - Drop late filers
  - Deceased persons
  - Foreign residents
  - Missing or invalid age & gender
- Final sample size ~ 26,000

## Record Collection



## Synthesis



**Classification and Regression Tree (CART) Model**

- A decision tree algorithm structured as a sequence of decisions.
- Synthesized categorical variables first followed by continuous.



# Synthesis Process, cont.

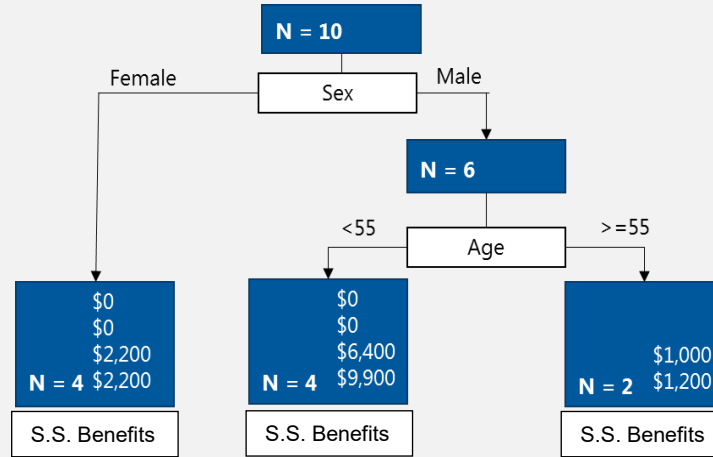
## Subdivide sample into 2 parts

- ① Those records with just demographic information
- ② Those records with at least one tax amount > 0

- ① • Randomly assigned *gender*, based on proportions of underlying data
- Synthesized *age* based on *gender*
- Assigned zeros to all tax variables

Synthesized zero records

- ② • Randomly assigned *gender*, based on proportions of underlying data
- Sequentially synthesize variables using CART starting with *age* conditional on previously synthesized outcome variables
  - Each point is randomly sampled with replacement
  - For continuous variables starting with *Social Security Benefits* then synthesized in order of linear correlation to *Social Security Benefits*.



Synthesized non-zero records

Obs	Value	Ntile	Optimal KDE Variance	Synthetic Value Distribution
1	\$0	1st	\$0	0
2	\$0	1st	\$0	0
3	\$6,400	66th	\$650	$\sim N(\mu=6,400, \sigma^2=650)$
4	\$9,900	98th	\$2,300	$\sim N(\mu=9,900, \sigma^2=2,300)$

- Then draw a value from a smoothed KDE distribution
  - $\sim N(\mu = \text{sampled value}, \sigma^2 = \text{"percentile variance"})$
  - Variance for a Kernel Density Estimator (KDE) of the percentile of the mean

Fully Synthetic File



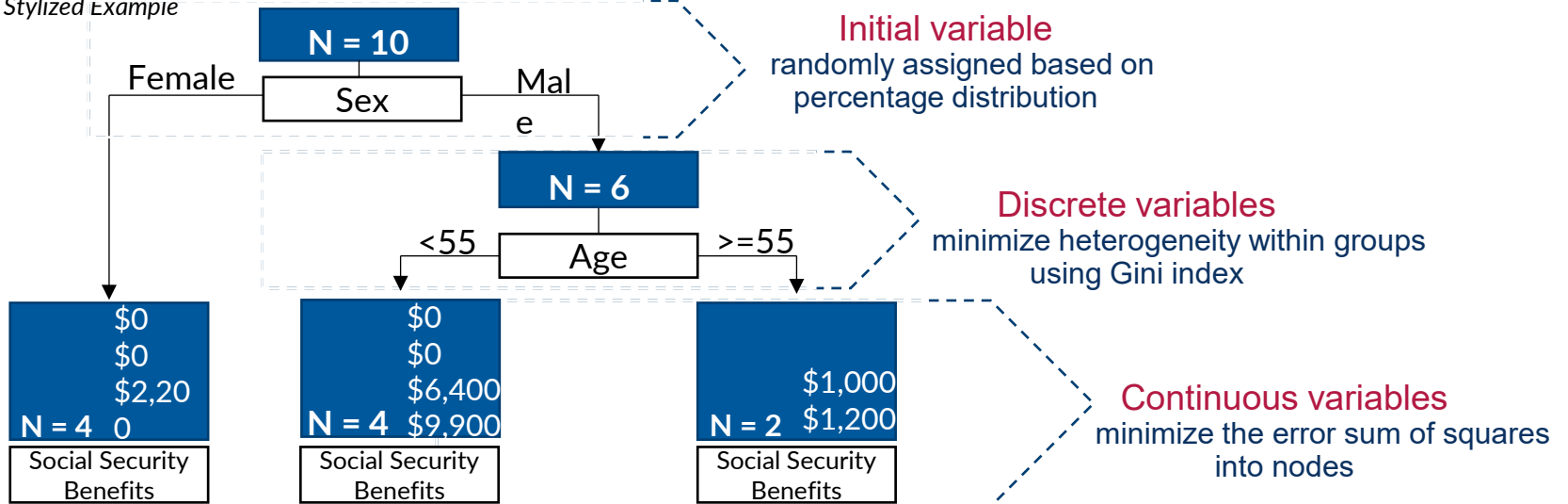
# CART - Methodology

## Process:

1. Assign *gender* based on the distribution of confidential data.
2. Predict *age* conditional on *gender*, minimizing heterogeneity within groups. Then randomly select value from within those final nodes.
3. Predict *Social Security Benefits* conditional on *gender* and *age*, to minimize Sum of Square Errors.
4. Predict next highest linearly correlated variable(s) conditional on *gender*, *age*, and *Social Security*

Synthesized non-zero records

Stylized Example







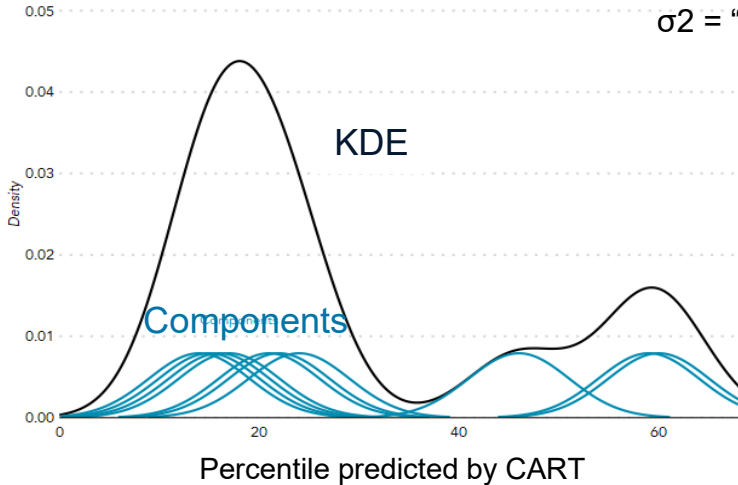
# CART – Methodology, cont.

Stylized Example, cont.

## Males, < 55

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Expanded Stylized Example



$\sim N(\mu = \text{sampled value}, \sigma^2 = \text{"percentile variance"})$

Draw a value from a smoothed Kernel Density function for each percentile of values predicted by CART.



# Ensuring Privacy

## Imposed protocols

- Sample of 1 in 1,000 observations
- Top code age at 85
- Terminal nodes limited to 50
- Kernel Density Estimator with variance  $\sigma^2$
- Run through simple tax calculator
- Round continuous variables

## Validation Metrics

- **Duplicates**
- **Unique-Donors**
- **Unique-Uniques**
- **Row-wise Squared Inverse Frequency**
- **$\ell$ -diversity of final nodes**



# Measuring Quality

**Summary statistics**

**Correlation fit**

**Kolmogorov-Smirnov (KS) test**

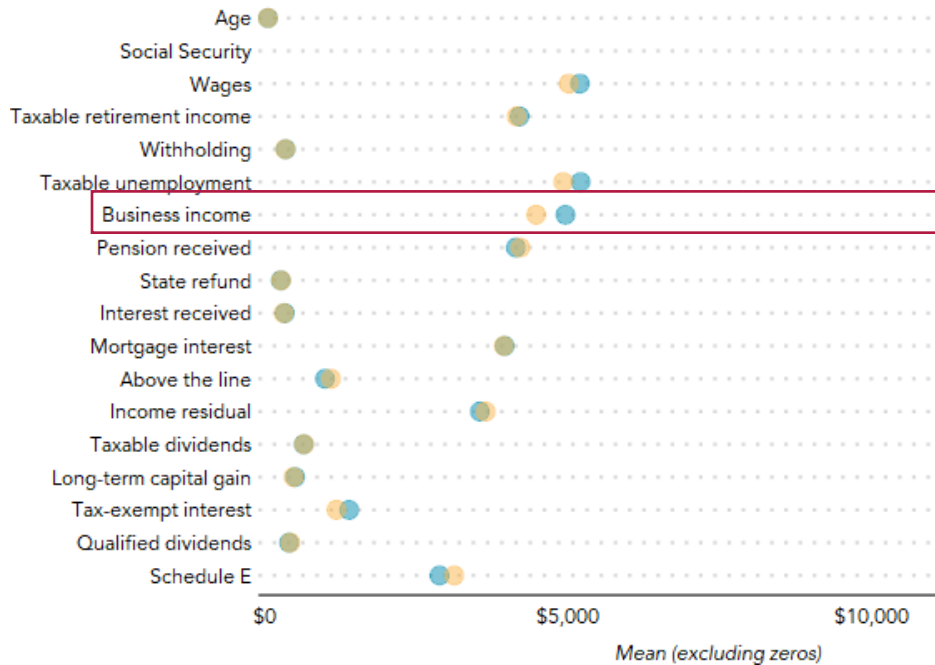
**Regression confidence interval overlap**



# Summary statistics

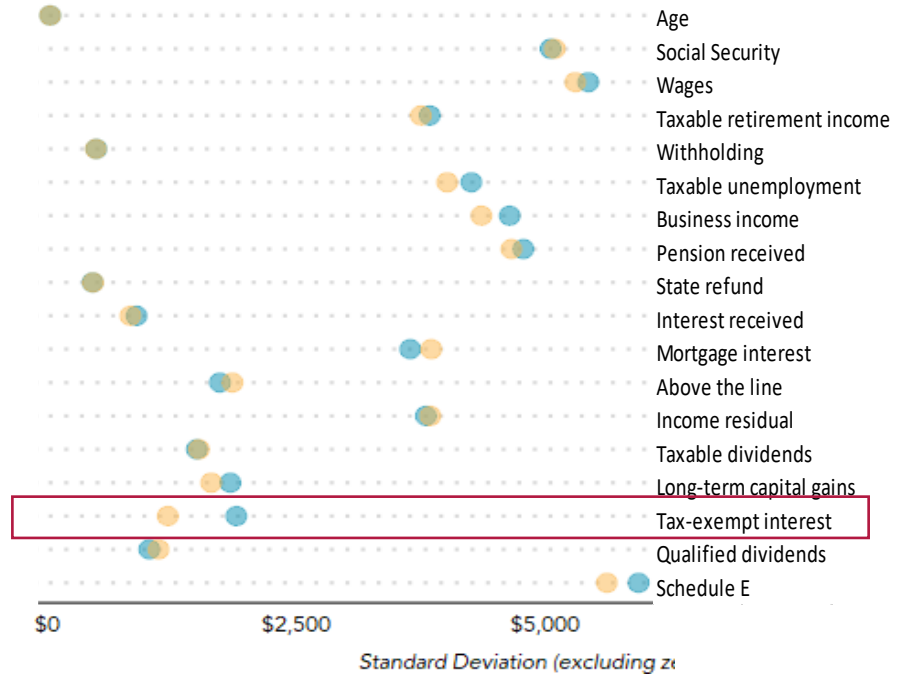
## Means

● Original ● Synthetic



## Standard deviations

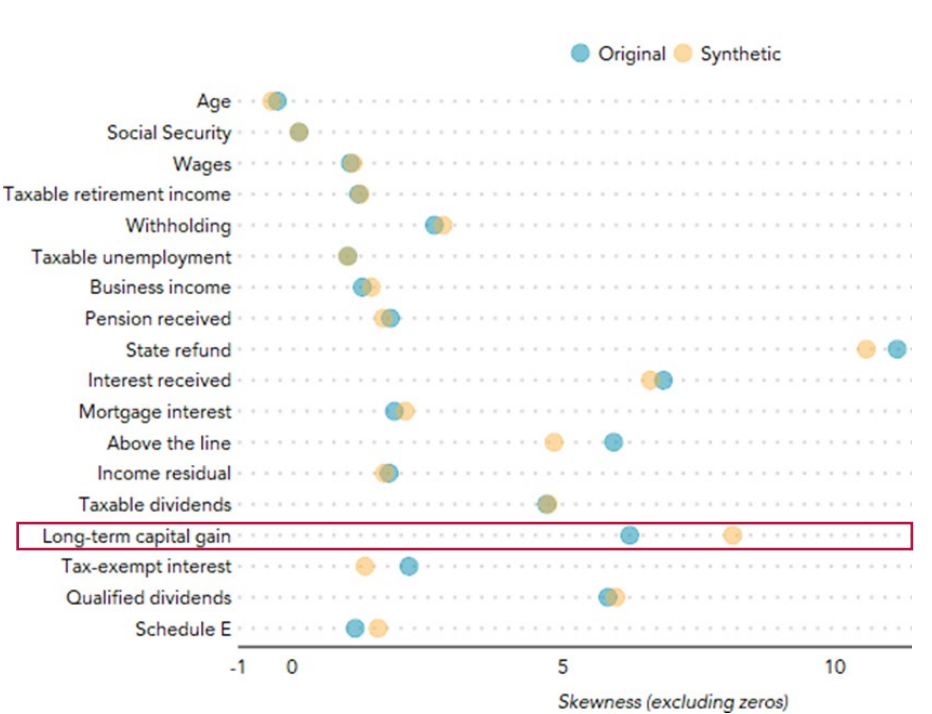
● Original ● Synthetic



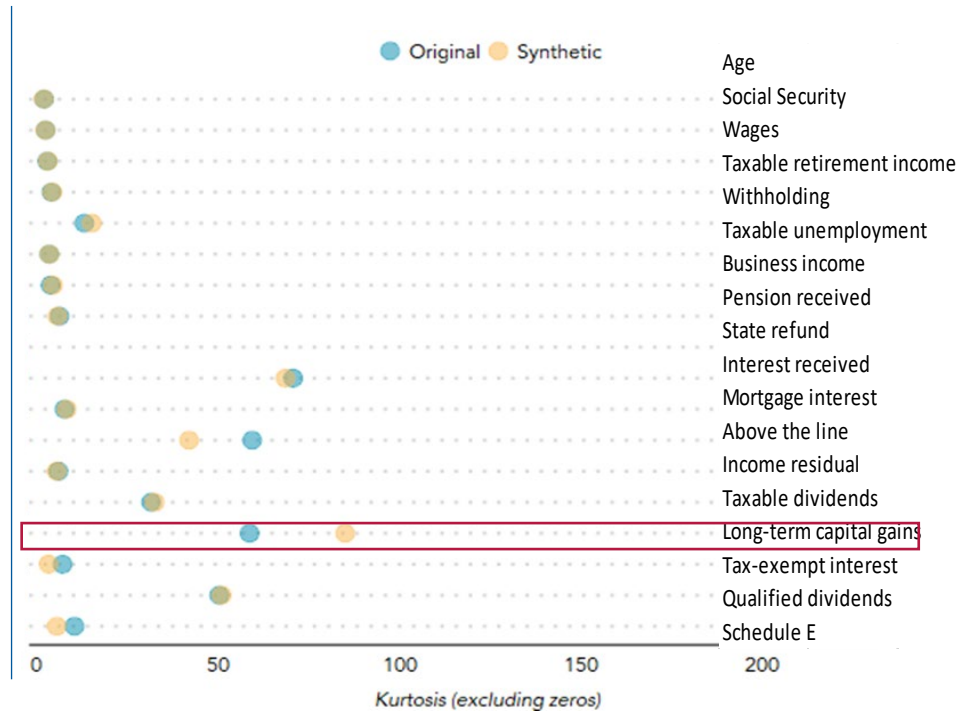


# Summary statistics, cont.

## Skewness



## Kurtosis



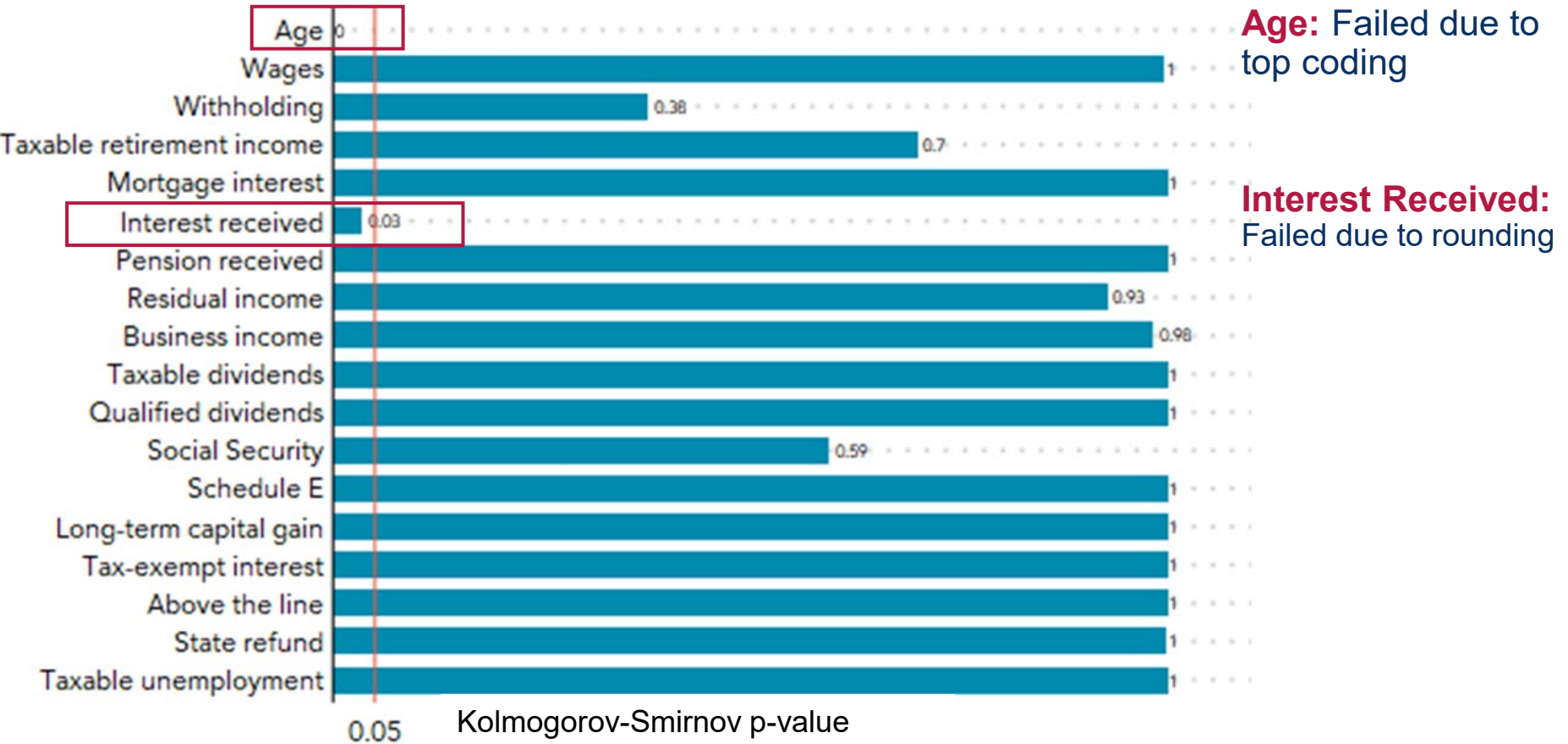




# Kolmogorov-Smirnov (KS) test

**Purpose:** Equivalence of univariate probability distributions

$H_0$  = samples come from the same underlying distribution







# Confidence interval overlap

**Purpose:** Average relative overlap between CIs for each coefficient in identical models.

Wages = f(all other vars)

## Interpretation:

1 = Perfect overlap

0 = No overlap, adjacent CIs

< 0 = The distance between CIs

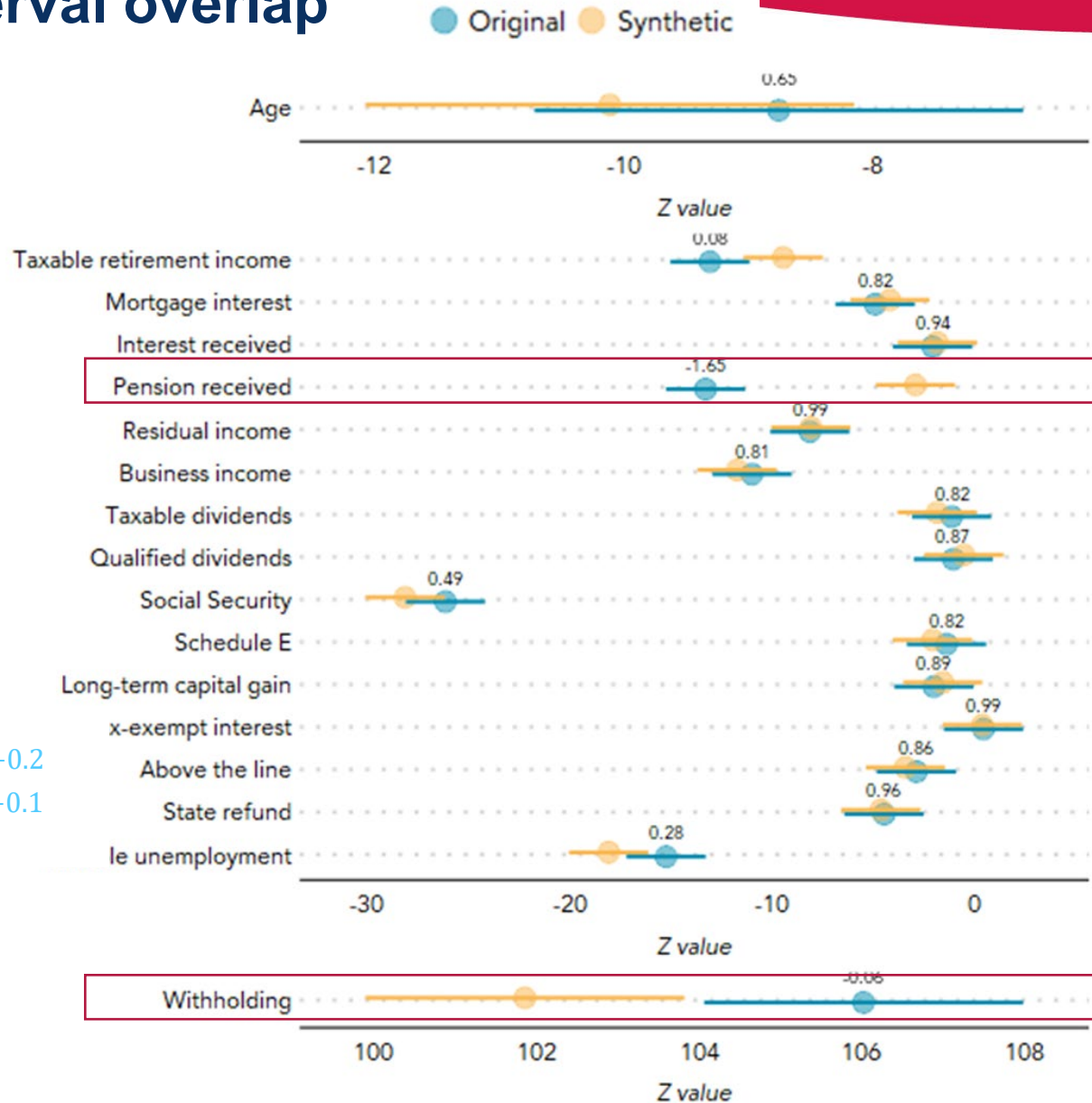
## Interval Overlap:

Pension Received = -1.65

$$\begin{aligned} \hat{\beta}_O &= -0.2 \\ \hat{\beta}_S &= -0.1 \end{aligned}$$

Withholding = -0.05

$$\begin{aligned} \hat{\beta}_O &= -5.6 \\ \hat{\beta}_S &= -5.1 \end{aligned}$$







# Thank you

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