

DATA SCIENCE FOR ECONOMISTS

ECON 220 LAB

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Week 9, Handling IPUMS Data (Part 2) – 10/24/2025

Outline

01

IPUMS

02

Recoding and outlier
analysis

03

Basic sampling

What is IPUMS?

Integrated Public Use Microdata Series

Operated by the University of
Minnesota

Core mission: data harmonization

IPUMS takes datasets that were
originally collected with different
questions, codes, and variable
names and makes them consistent.
Lots of recoding!

Free access!

By providing access to detailed,
anonymized individual-level data
(microdata), IPUMS allows
researchers to ask complex
questions that can't be answered
with aggregated summary tables.

IPUMS provides census and survey data from around the world integrated across time and space. IPUMS integration and documentation makes it easy to study change, conduct comparative research, merge information across data types, and analyze individuals within family and community contexts. Data and services available free of charge.



U.S. Census and American Community Survey microdata from 1850 to the present. [Learn More](#)




[VISIT SITE](#)

Current Population Survey microdata including basic monthly surveys and supplements from 1962 to the present. [Learn More](#)

[VISIT SITE](#)

World's largest collection of census microdata covering over 100 countries, contemporary and historical. [Learn More](#)

[VISIT SITE](#)

Health survey data from around the world, including harmonized data collections for DHS , MICS , and PMA . [Learn More](#)

[VISIT SITE](#)

U.S. Census summary tables and GIS data from 1790 to the present. [Learn More](#)

[VISIT SITE](#)

Summary tables and GIS data from population, housing, and agricultural censuses around the world. [Learn More](#)

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— HELP POWER IPUMS —

Support our work to preserve and democratize access to the world's population data.

[DONATE](#)

— VIRTUAL OFFICE HOURS —

Tuesday, November 18
10:30am-12:00pm CT

[REGISTER FOR OFFICE HOURS](#)

— CALENDAR —

65th ISI World Statistics Congress

- How to register and extract data? Check guide on Canvas.

Importing required libraries

```
import pandas as pd
import numpy as np
import matplotlib.pyplot as plt
import os
path = os.getcwd()
```

Python

Load and explore the data

```
data = pd.read_csv('usa_00004.csv')
```

Python

	# YEAR	# SAMPLE	# SERIAL
count	3405809.0	3405809.0	340580
mean	2023.0	202301.0	758991.73689217
std	0.0	0.0	441473.568225181
min	2023.0	202301.0	
25%	2023.0	202301.0	37238
50%	2023.0	202301.0	75683
75%	2023.0	202301.0	114700
max	2023.0	202301.0	151901

8 rows x 20 cols

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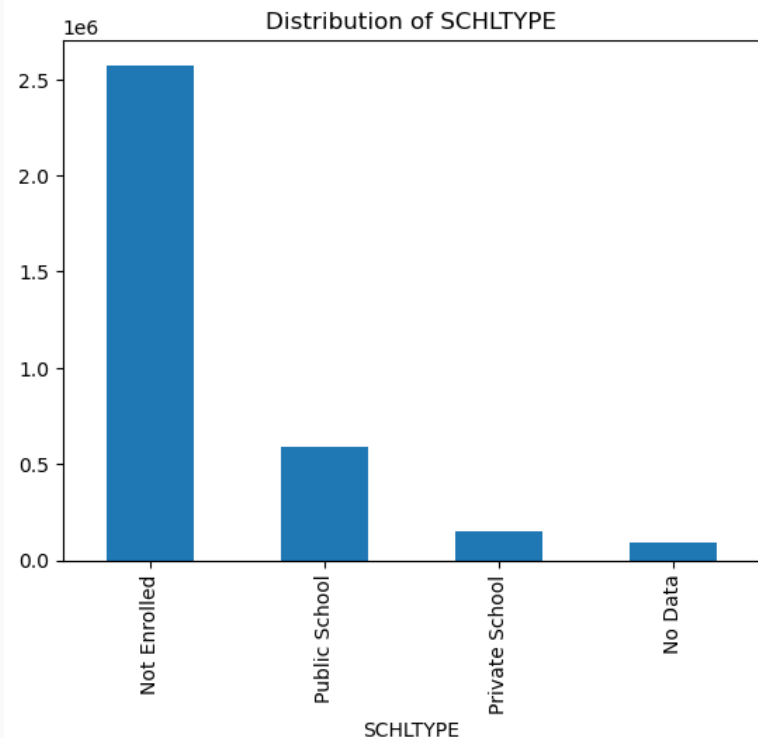
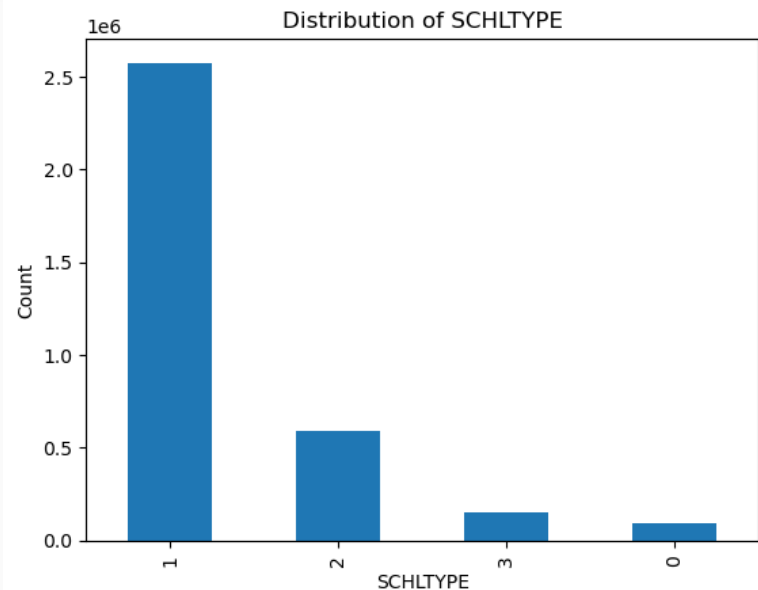


- Note the large number of observations!

Recoding a categorical variable

- IPUMS uses numeric codes for categorical variables.
- **Example:** SCHLTYPE (School Type).
- **Need the Codebook!** Tells us: 0=N/A, 1=Not Enrolled, 2=Public, 3=Private.
- **Recoding:** replace numbers with meaningful labels.

Recoding a categorical variable



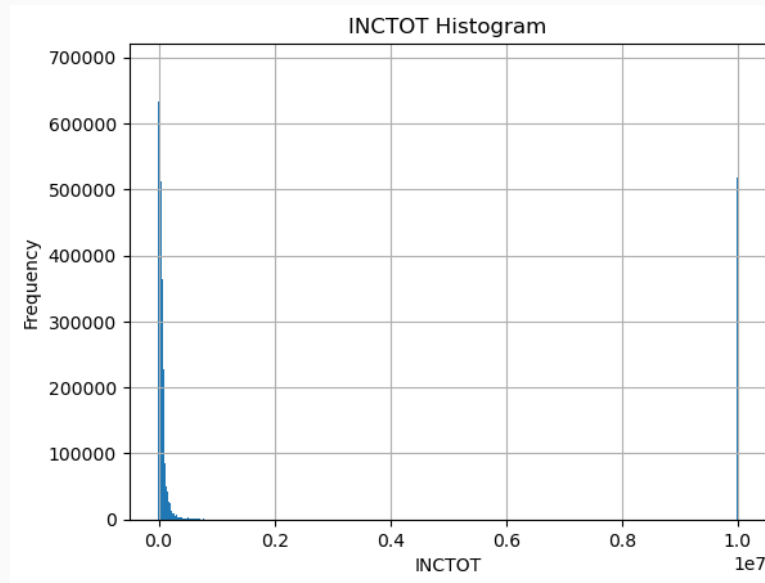
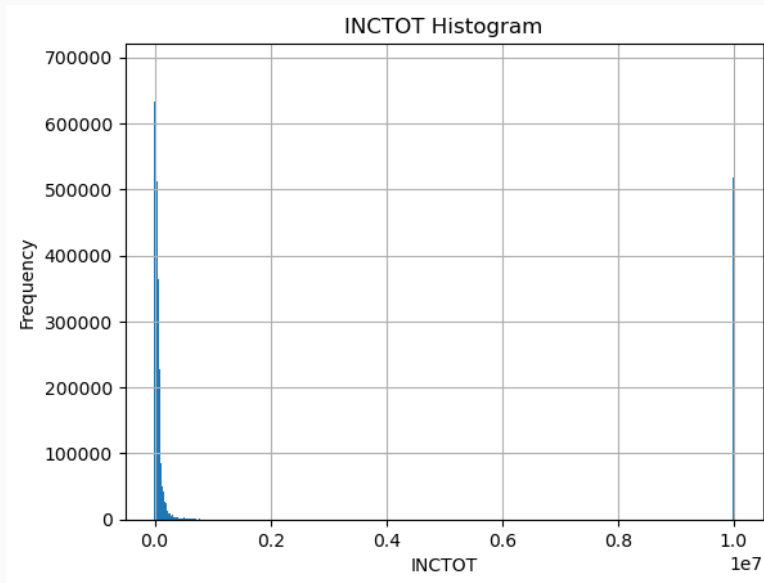
Outlier detection and treatment

- **Outliers:** Extreme values that might distort analysis.
- **Idea:**
 - Use a histogram (.hist()) to see the distribution.
 - Observe asymmetry and potential extreme high/low values (including negative).
- **Treatment:**
 - Interquartile range (IQR) = $Q3$ (75th percentile) - $Q1$ (25th percentile)
 - **Define Bounds:** Lower = $Q1 - 1.5 \times IQR$, Upper = $Q3 + 1.5 \times IQR$.
 - **Filter:** Keep data only within these bounds.

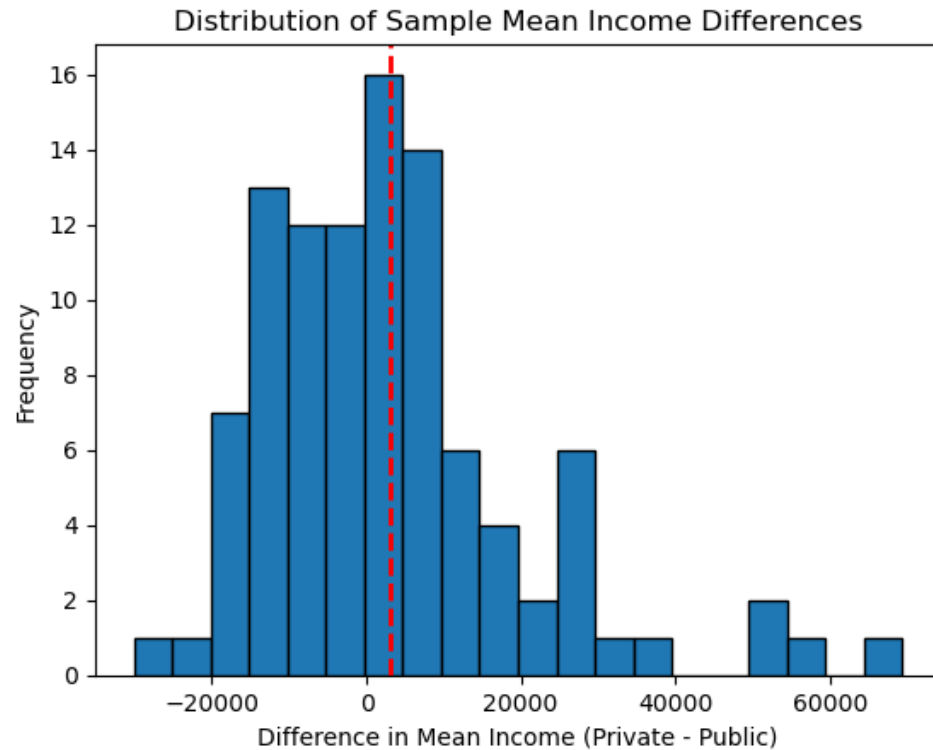
Outlier detection and treatment

- IQR might not catch all problematic values (e.g., 0 or negative income).
- **Domain knowledge:** Is zero/negative income plausible/useful for this analysis?
- **Filter:** Remove observations with $INCTOT \leq 0$.

Outlier detection and treatment



Random sample



Recap

- **Check the Codebook:** It's essential for understanding IPUMS numeric codes.
- **Recode Variables:** Use `.replace()` to make your data readable (e.g., 2 → "Public School").
- **Find Outliers:** Use `.hist()` to visualize data and spot extreme values.
- **Sample for Speed:** Use `.sample()` for quick analysis on large datasets, but be aware of sampling variability.

To-do list

- **Complete Data Exercise 6**
 - Upload Jupyter notebook (.ipynb file) and HTML file on **October 26**
- **Complete Data Exercise 7**
 - Upload Jupyter notebook (.ipynb file) and HTML file on **November 2**