### Using Randomization Tests and IPUMS-USA\* to Investigate the Gender Wage Gap Laura Schultz, Rowan University

Historically, women have often earned less than men for performing the same job, a phenomenon known as the gender wage gap. Given that salary data are notoriously skewed, investigating the gender wage gap provides an engaging context for introducing students to modern randomization tests as an alternative to more traditional nonparametric tests. The first randomization test included in this classroom activity addresses whether the mean salary of male accountants and auditors is significantly greater than that of females with similar qualifications, and the second task investigates whether the ratio of the median salary of female accountants/auditors to the median salary of their male counterparts is significantly less than 1. (A ratio of 1 would indicate that the median female and male salaries are the same.) Both randomization tests utilize custom JMP scripts that I have written. The data set, a subset of a sample collected as part of the 2008 American Communities Survey that I downloaded from IPUMS-USA, consists of the total personal earnings for samples of full-time female and male accountants/auditors ages 25-34 with Bachelor's degrees in the Philadelphia metropolitan area. I chose this particular data set due to its appeal to the accounting and finance majors I teach at a university in the greater Philadelphia region. Given the flexibility of the IPUMS-USA interface, this classroom activity could be adapted to investigate the gender wage gap using samples from other professions, age groups, or geographical regions.

\*S. Ruggles, J.T. Alexander, K. Genadek, R. Goeken, M.B. Schroeder, and M. Sobek. *Integrated Public Use Microdata Series: Version 5.0* [Machine-readable database]. Minneapolis: University of Minnesota, 2010.



Sources: U.S. Census Bureau, American Community Survey, 2008; and Puerto Rico Community Survey, 2008.



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#### Total Personal Income for Full-Time Accountants/Auditors Age 25-34 with Bachelor's Degrees Working in the Philadelphia Metropolitan Area, 2008 ACS

#### SDA 3.5: Listcase

United States 2008

#### Mar 22, 2012 (Thu 08:05 AM CDT)

Variables							
Role	Name	Label	Range	MD	Dataset		
Variable	age	Age	0-95		1		
Variable	sex	Sex	1-2		1		
Variable	race	Race	1-9		1		
Variable	educd	Educational attainment	1-116		1		
Variable	occ	Occupation	0-9920		1		
Variable	incearn	Total personal earned income	-9999-1251000		1		
Filter	pwmetro(6160)	Place of work: metropolitan area(=Philadelphia, PA/NJ)	0-9360		1		
Filter	educd(101)	Educational attainment(=Bachelor's degree)	1-116		1		
Filter	occ(800)	Occupation	0-9920		1		
Filter	empstat(1)	Employment status(=Employed)	0-3		1		
Filter	age(25-34)	Age	0-95		1		
Filter	uhrswork(40-**)	Usual hours worked per week	0-99		1		

Cases listed: 68

#	age	sex	race	educd	occ	incearn
1	34	Male	White	Bachelor's degree	800	116,000

## The Salary Data



### The Randomization Test Statistics

Difference between two sample means

$$\left(\overline{x}_{Male} - \overline{x}_{Female} = \$13482\right)$$

Ratio of two sample medians

$$\left(\frac{\tilde{x}_{Female}}{\tilde{x}_{Male}}=0.867\right)$$

## The Randomization Procedure

- Compute the test statistic for the original data.
- Scramble the group labels and recalculate the test statistic for the permutation resample.
- Repeat many times; I generated 999 resamples.
- Construct the Monte Carlo permutation distribution and use it to estimate the *P*-value by locating the original test statistic.

# The Results: Comparing Means

#### Monte Carlo Permutation Distribution of $\overline{x}_{Male} - \overline{x}_{Female}$



• Eight resamples resulted in test statistics as large or larger than the original test statistic  $(\overline{x}_{Male} - \overline{x}_{Female} = \$13482)$ .

• Thus, the approximate right-sided *P*-value is  $P \approx \frac{8+1}{999+1} = .009$ 

# The Results: Comparing Medians

#### Monte Carlo Permutation Distribution of $\tilde{x}_{Female}$ / $\tilde{x}_{Male}$



• 86 resamples resulted in test statistics as small or smaller than the original test statistic  $\left(\frac{\tilde{x}_{Female}}{\tilde{x}_{Male}} = 0.867\right)$ .

• Thus, the approximate left-sided *P*-value is  $P \approx \frac{86+1}{999+1} = .087$ 

### Conclusions

- Randomization tests provide a modern nonparametric approach for analyzing notoriously skewed salary data.
- When comparing means, we find evidence of a gender wage gap between the salaries earned by male and female accountants/ auditors with Bachelor's degrees who work in the Philadelphia metropolitan area ( $\overline{x}_{Male} \overline{x}_{Female} = \$13482$ , P = .009, right-sided).
- However, there is not evidence of a significant wage gap when we consider the ratio of the median salaries of females and males in this same demographic ( $\tilde{x}_{Female} / \tilde{x}_{Male} = 0.867, P = .087$ , left-sided).