

SOC-5811 Week 2: Quantitative research methods

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QUANTITATIVE METHODS

- ▶ Empiricism: social theories imply empirical claims that are falsifiable.
- ▶ Data doesn't speak for itself; it must be carefully interpreted, summarize, and analyzed.



QUANTITATIVE METHODS

- ▶ Statistics is about providing a concise summary of empirical data.
- ▶ This involves analyzing **samples** drawn from **populations**.



SAMPLES AND POPULATIONS

Say I have a bag filled with 50 marbles of different colors. I reach in the bag and pull out 10 marbles. I see that 6 marbles are blue and 4 are red.

What is the population and sample here?

What can I say about the sample? What can I say about the population?



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 - ▶ Measurement theory: very central to psychology, education, etc.



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 - ▶ *I measure the rate of a specific disease among a group that was vaccinated for that disease and one that was not. Was the vaccine effective at reducing the disease?*

GENERALIZATION: SUMMARY

1. Generalizing from sample to population.
2. Measurement.
3. Forecasting.
4. Causal inference.



QUANTITATIVE DATA

```
## # A tibble: 6 x 7
##   msa                population percapita
##   <chr>                <dbl>         <dbl>
## 1 Washington-Arlington-Alexand~ 5949.         47.4
## 2 San Jose-Santa Clara-Sunnyva~ 1919.         40.4
## 3 Atlanta, Georgia MSA          5545.         25.3
## 4 Austin-San Marcos, Texas MSA   2271.         24.5
## 5 Little Rock-North Little Roc~   724.         20.3
## 6 Bellingham, Washington MSA     167.         20.0
```

QUANTITATIVE DATA

```
## # A tibble: 5 x 2
##   population region
##   <dbl> <fct>
## 1    5949. 1_Northeast
## 2    1919. 4_West
## 3    5545. 3_South
## 4    2271. 3_South
## 5     724. 3_South
```



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LEVELS OF MEASUREMENT

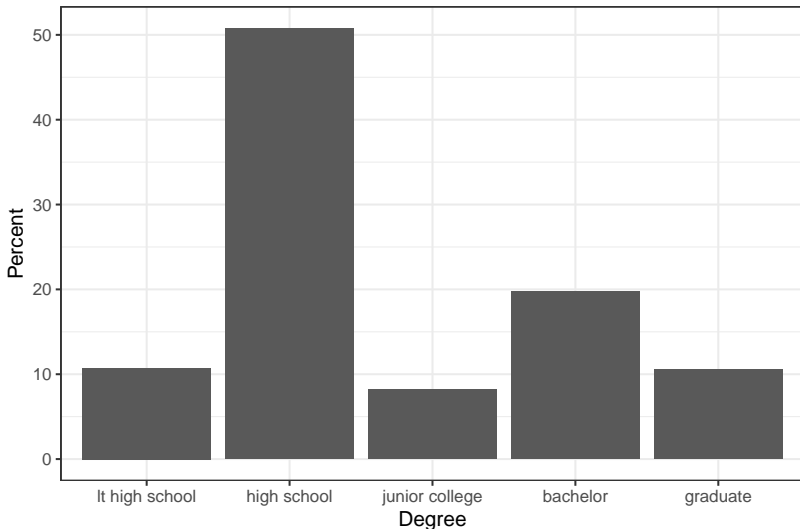
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 - ▶ Binary/dichotomous: special case of nominal variable that only has two values (e.g., coin flip)
- ▶ Measurement decisions are not always clear cut (e.g., age)

SUMMARIZING DATA

```
## # A tibble: 6 x 8
##   degree      agekdbrn  realrinc race      age pr
##   <fct>      <dbl+lbl> <dbl+lbl> <fct> <dbl+lb> <d
## 1 bachelor      35    45400 white      42
## 2 bachelor      32    54480 white      63
## 3 lt high scho~  17      908 white      62
## 4 high school    30    45400 white      55
## 5 graduate      30    54480 white      59
## 6 high school    20    8512. other      34
```



SUMMARIZING DATA



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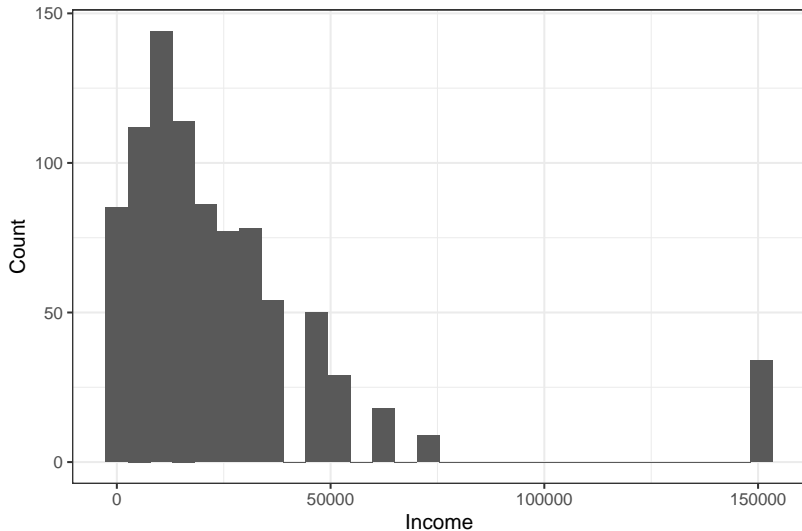


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Driven to DiscoverSM



SUMMARIZING DATA

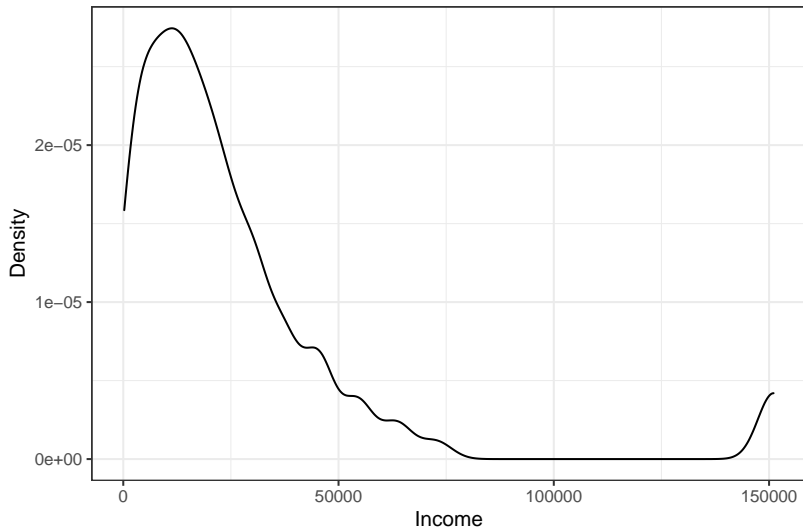


SUMMARIZING DATA

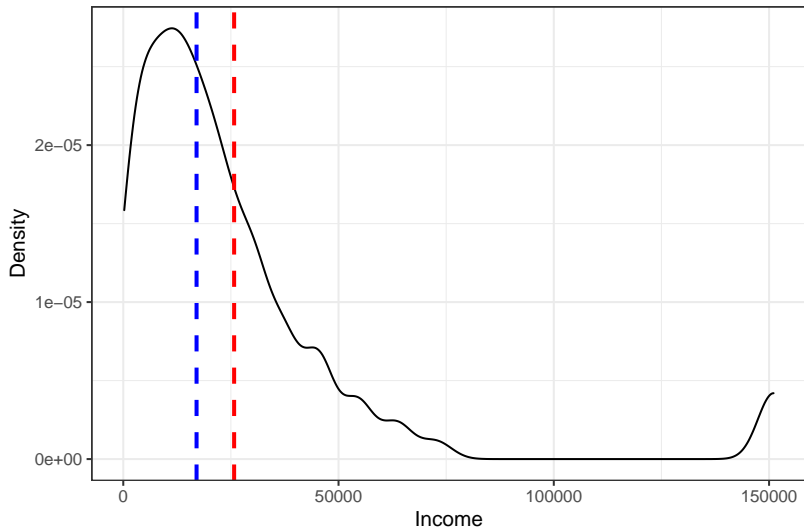
- ▶ Histograms of continuous data: why do we have to bin?
- ▶ How do we choose bins?



SUMMARIZING DATA



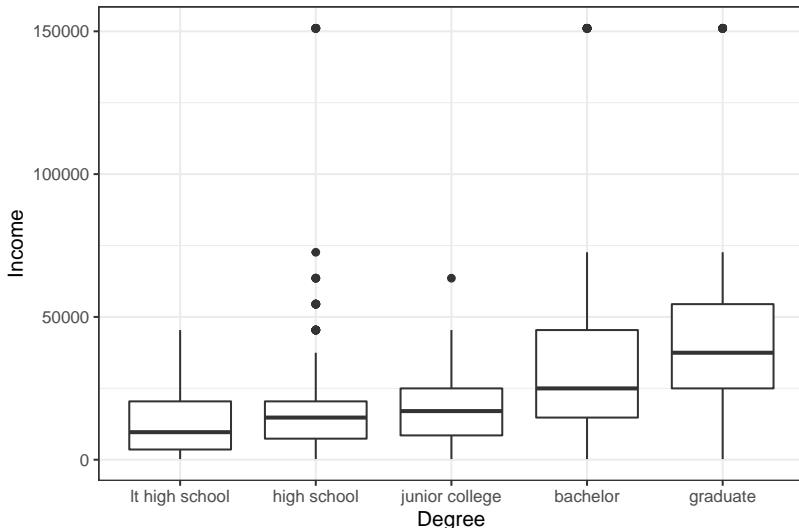
SUMMARIZING DATA



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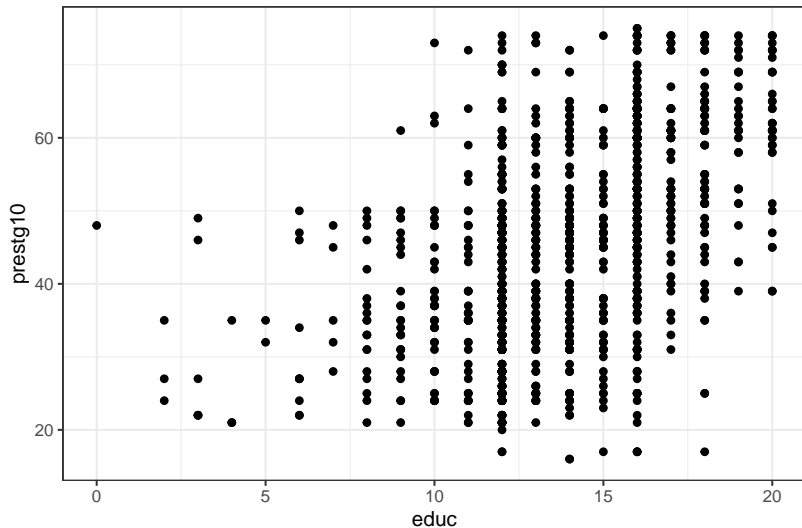
UNIVARIATE SUMMARY STATISTICS



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BIVARIATE RELATIONSHIPS



BIVARIATE RELATIONSHIPS

##		educ	prestg10	realrinc	agekdbrn
##	educ	1.00	0.49	0.35	0.42
##	prestg10	0.49	1.00	0.30	0.37
##	realrinc	0.35	0.30	1.00	0.22
##	agekdbrn	0.42	0.37	0.22	1.00

BIVARIATE RELATIONSHIPS

##		race		
##	degree	white	black	other
##	lt high school	0.06	0.02	0.03
##	high school	0.36	0.09	0.05
##	junior college	0.06	0.01	0.00
##	bachelor	0.15	0.03	0.02
##	graduate	0.08	0.01	0.01

How do we create models?

- ▶ Deterministic processes

How do we create models?

- ▶ Deterministic processes
- ▶ Stochastic processes