Gov 50: 8. Measurement: Summarizing Bivariate Relationships

Matthew Blackwell

Harvard University

Fall 2018

1. Today's agenda

- 2. Investigating fraud
- 3. Bivariate relationships

1/ Today's agenda

Logistics

- Problem set 2:
 - due Thursday by midnight.
 - remember to turn in Rmd and compiled pdf!
 - this time we start to take points off for Rmd files that don't compile.
- Midterm 1:
 - Next Tuesday.
 - Covers material through today.
 - Review session on Thursday.
- Mike's Monday section rescheduled to this Thursday (10/4) at 12pm in CGIS S020.
- Midterm course evaluations after the midterm.

- Talked about survey sampling, its problems
- How to summarize a single variable? Mean, median, range, SD.
- Now: how to summarize relationship *between* variables.
- Review 3.5–3.6
- Revisit the gay-marriage experiment:
 - LaCour and Green (2015). "When contact changes minds: An experiment of transmission of support for gay equality." *Science*, Vol. 346, No. 6215 pp. 1366–1369.
 - Broockman, Kalla, Aronow (2015). "Irregularities in LaCour (2014)"

2/ Investigating fraud

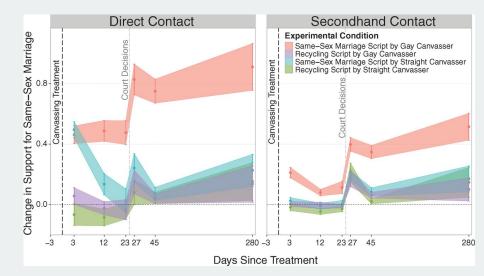
Changing minds on gay marriage

- Question: Can we effectively persuade people to change their minds?
- **Contact Hypothesis**: outgroup hostility diminishes when people from different groups interact with one another.
- Two randomized control trials in Los Angeles
- Target population: voters in Los Angeles.
- Sampling frame: registered voter list.
 - invited randomly selected voters to participate in an online baseline survey.
 - asked them to refer their friends and families with compensation.
 - those friends and family are also invited to participate in the online baseline survey.
 - panel data: baseline plus 6 waves.

Randomized treatment:

- gay vs. straight canvassers with similar characteristics
- same-sex marriage vs. recycling scripts (placebo)
- control group: no canvassing
- Persuasion scripts are the same except on important difference:
 - gay canvassers: they would like to get married but law prohibits it.
 - straight canvassers: their gay child, friend, or relative would to get married but the law prohibits it.
- Outcome measures:
 - support for same-sex marriage.
 - feeling toward gay people.

Big and lasting effects of persuasion



Reshaped data

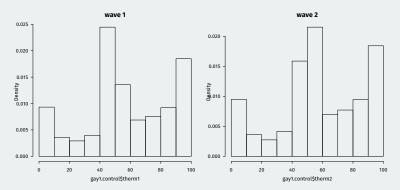
Name	Description	
study	Which study is the data from (1 = Study1, 2 = Study2)	
treatment	Five possible treatment assignment options	
therm1	Survey thermometer rating of feeling towards gay couples in waves 1 (0–100)	
therm2	Survey thermometer rating of feeling towards gay couples in waves 2 (0–100)	
therm3	Survey thermometer rating of feeling towards gay couples in waves 3 (0–100)	
therm4	Survey thermometer rating of feeling towards gay couples in waves 4 (0–100)	
gay.reshaped <- read.csv("data/gayreshaped.csv") names(gay.reshaped)		

##	[1]	"study"	"treatment"	"therm1"	"therm2"
##	[5]	"therm3"	"therm4"		

Comparison of gay thermometer across waves

• Compare between waves 1 and 2 for the control group in Study 1:



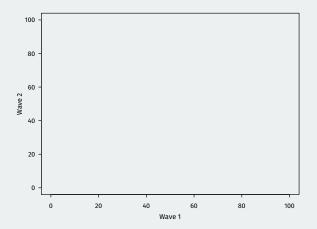


3/ Bivariate relationships

- Direct graphical comparison of two variables.
- Each point on the scatterplot (x_i, y_i)
- Use the plot() function

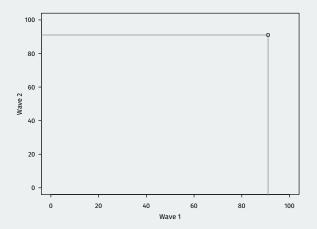
gay1.control[1, c("therm1", "therm2")]

therm1 therm2 ## 1 91 91



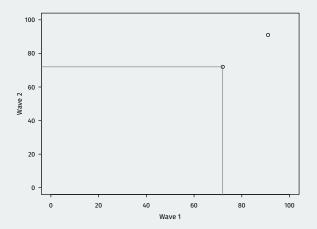
gay1.control[1, c("therm1", "therm2")]

therm1 therm2
1 91 91



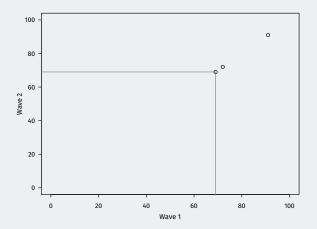
gay1.control[2, c("therm1", "therm2")]

therm1 therm2 ## 2 72 72



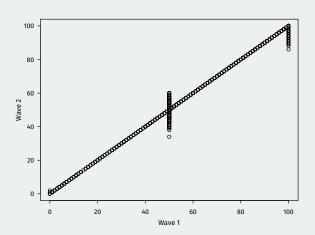
gay1.control[3, c("therm1", "therm2")]

##		therm1	therm2
##	3	69	69



gay1.control[1,c("therm1", "therm2")]

##		therm1	therm2
##	1	91	91



- Variables can be on different scales: makes it difficult to assess how well they "go together"
- Need a way to put any variable on common units.

z-score:

z-score of $x_i = \frac{x_i - \text{mean of } x}{\text{standard deviation of } x}$

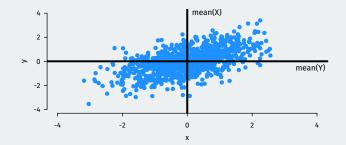
z-scores don't depend on units:

z-score of
$$(ax_i + b) = z$$
-score of x_i

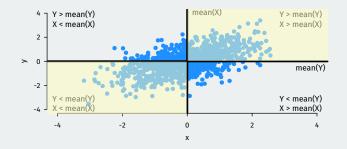
- How do variables move together on average?
- If I know one variable is big, does that tell me anything about how big the other variable is?
 - Positive correlation: when x is big, y is also big
 - Negative correlation: when x is big, y is small
 - High correlation: data cluster tightly around a line.
- The technical definition of the correlation coefficient:

$$\frac{1}{n-1} \sum_{i=1}^{n} \left[(z \text{-score for } x_i) \times (z \text{-score for } y_i) \right]$$

Correlation intuition



Correlation intuition



Large values of X tend to occur with large values of Y:

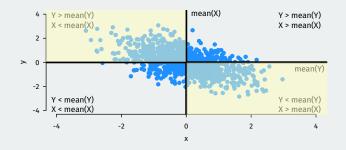
(z-score for x_i) × (z-score for y_i) = (pos. num.) × (pos. num) = +

• Small values of X tend to occur with small values of Y:

(z-score for x_i) × (z-score for y_i) = (neg. num.) × (neg. num) = +

If these dominate → positive correlation.

Correlation intuition



Large values of X tend to occur with small values of Y:

(z-score for x_i) × (z-score for y_i) = (pos. num.) × (neg. num) = -

Small values of X tend to occur with large values of Y:

(z-score for x_i) × (z-score for y_i) = (neg. num.) × (pos. num) = -

• If these dominate → negative correlation.

Properties of correlation coefficient

- Correlation measures **linear** association.
- Interpretation:
 - Correlation is between -1 and 1
 - Correlation of 0 means no linear association.
 - ▶ Positive correlations ~→ positive associations.
 - Negative correlations ~> negative associations.
 - Closer to -1 or 1 means stronger association.
- Order doesn't matter: cor(x,y) = cor(y,x)
- Not affected by changes of scale:
 - cor(x,y) = cor(ax+b, cy+d)
 - Celsius vs. Fahreneheit; dollars vs. pesos; cm vs. in.

- Use the **cor()** function
- Missing values: set the use = "pairwise" ~> available case analysis

cor(gay1.control\$therm1, gay1.control\$therm2, use = "pairwise")

[1] 0.998

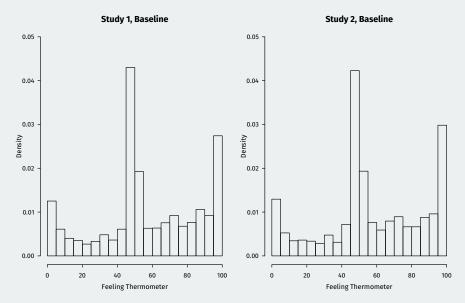
Extremely high correlation!

Comparisons between studies

- Cannot use plot() or cor(). Why?
- Different studies have different respondents.
- Start with histograms:

gay1 <- subset(gay.reshaped, (study == 1))
gay2 <- subset(gay.reshaped, (study == 2))</pre>

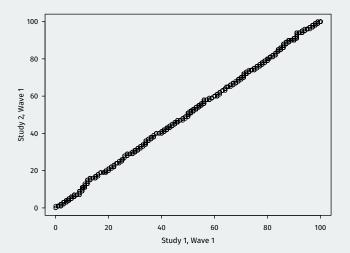
```
hist(gay1$therm1, freq = FALSE, breaks = 20,
ylim = c(0, 0.05), xlab = "Feeling Thermometer",
main = "Study 1, Baseline")
hist(gay2$therm1, freq = FALSE, breaks = 20,
ylim = c(0, 0.05), xlab = "Feeling Thermometer",
main = "Study 2, Baseline")
```



- Quantile-quantile plot (qq-plot): Plot the quantiles of each distribution against each other.
- Example points:
 - ► (min of *X*, min of *Y*)
 - (median of X, median of Y)
 - (25th percentile of X, 25th percentile of Y)
- 45 degree line indicates quality of the two distributions.

QQ-plot example

qqplot(gay1\$therm1, gay2\$therm1, xlab = "Study 1, Wave 1", ylab = "Study 2, Wave 1")



• Question wording of thermometer score attributed to 2012 Cooperative Campaign Analysis Project (CCAP):

Name	Description
caseid	unique respondent ID
gaytherm	Survey thermometer rating (0-100) of feeling to-
	wards gay couples

• CCAP has some missing data:

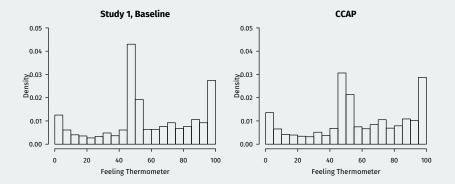
ccap <- read.csv("data/ccap2012.csv")
mean(is.na(ccap\$gaytherm))</pre>

[1] 0.0704

mean(is.na(gay1\$therm1))

[1] 0

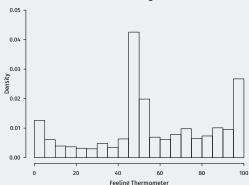
Comparison of CCAP and Study 1



- Suspiciously similar!
- What's the difference?

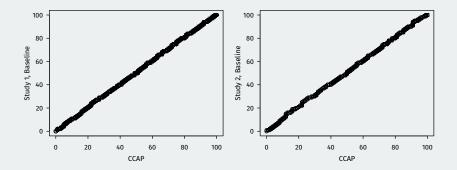
Recoding missing as 50s

ccap\$gaytherm[is.na(ccap\$gaytherm)] <- 50
hist(ccap\$gaytherm, freq = FALSE,
 ylim = c(0, 0.05), xlab = "Feeling Thermometer",
 main = "CCAP: with missing data as 50")</pre>



CCAP: with missing data as 50

QQ plots reveal extreme similarity



Retraction



- Scatterplots, correlation, and QQ-plots all help us visualize relationships between variables.
- With gay-marriage study, helped us detect fraud.
- After midterm: prediction!