



Addressing Alternative Explanations: Multiple Regression

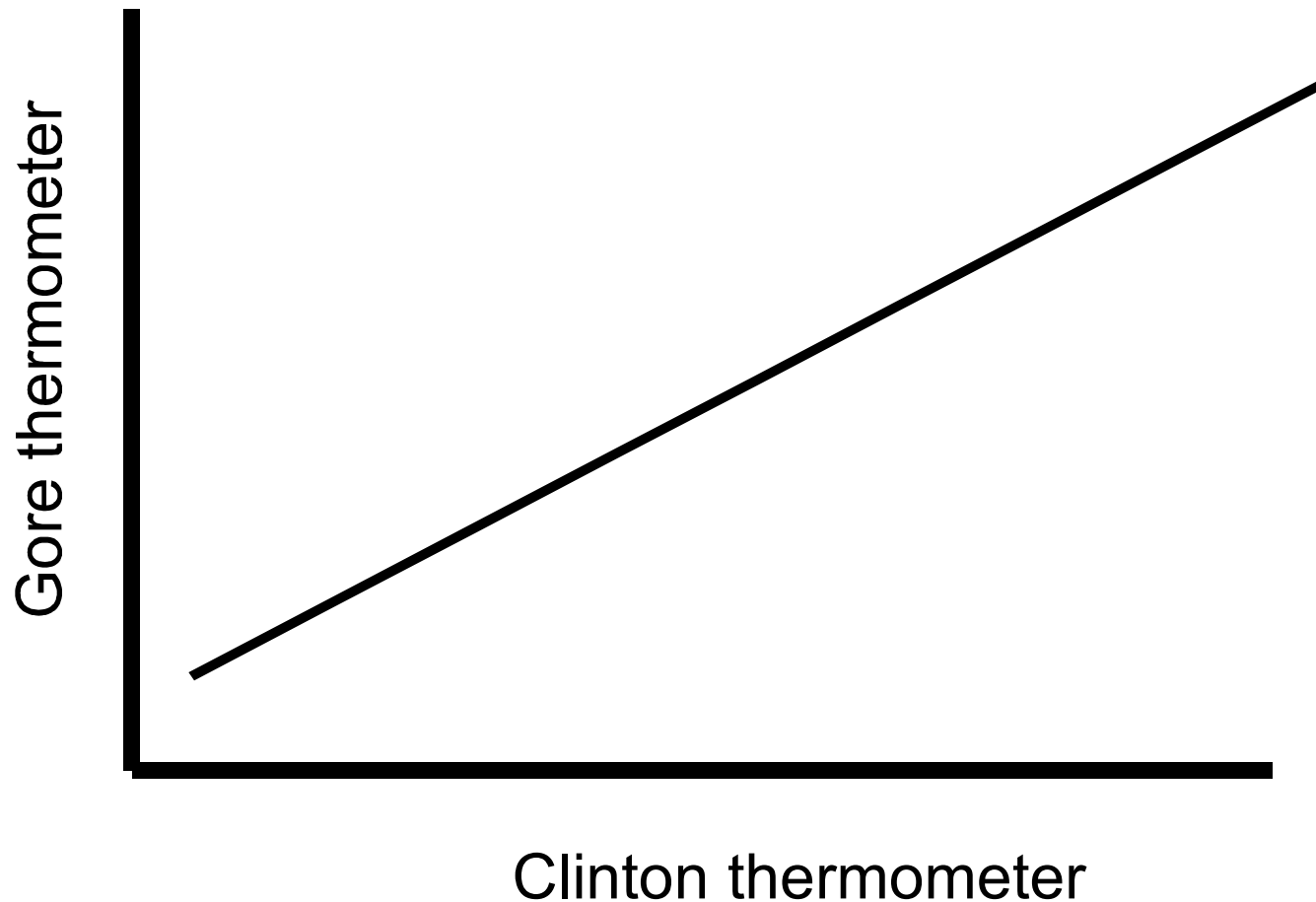
17.871



Did Clinton hurt Gore example

- Did Clinton hurt Gore in the 2000 election?
 - Treatment is not liking Bill Clinton
- How would you test this?

Bivariate regression of Gore thermometer on Clinton thermometer

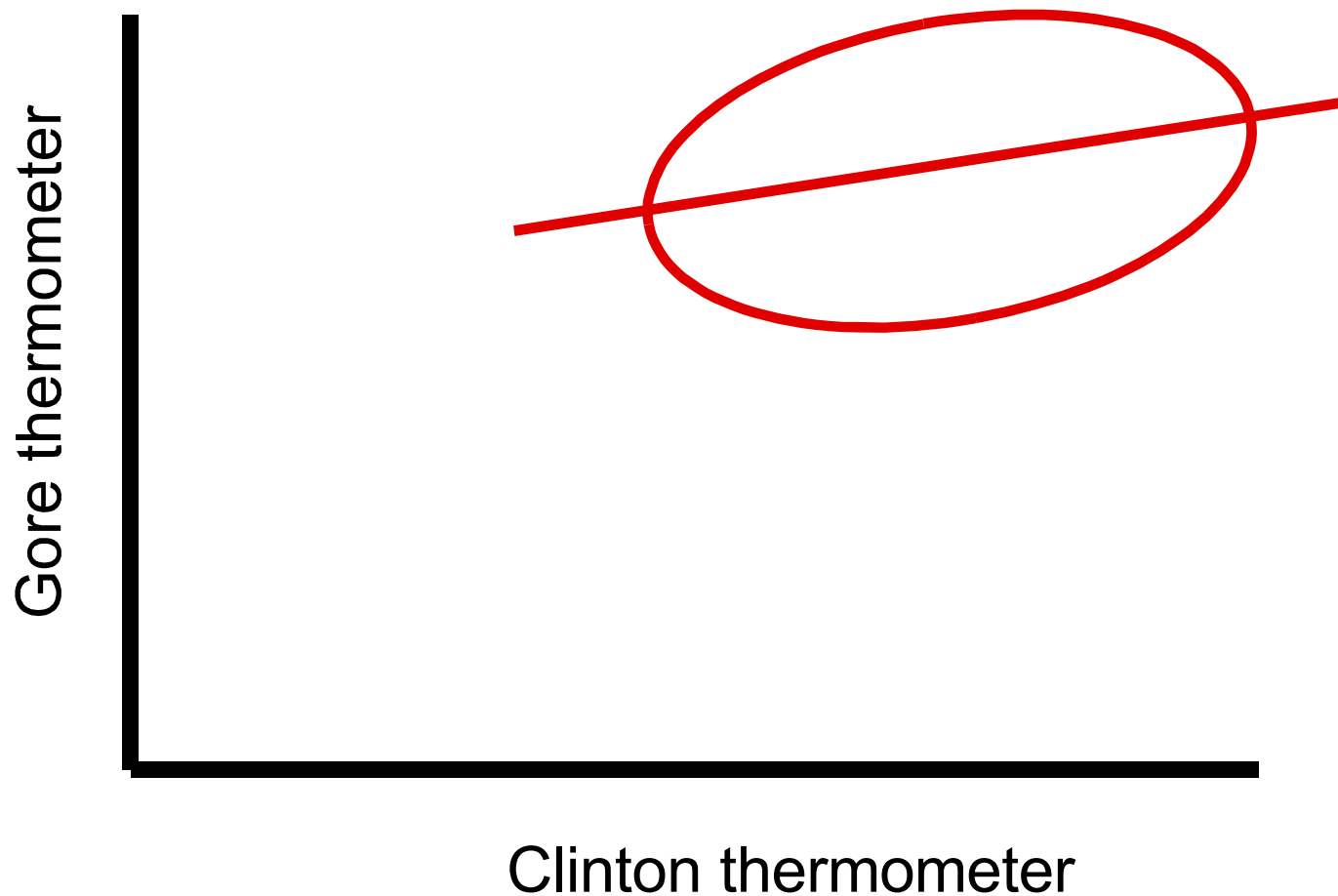




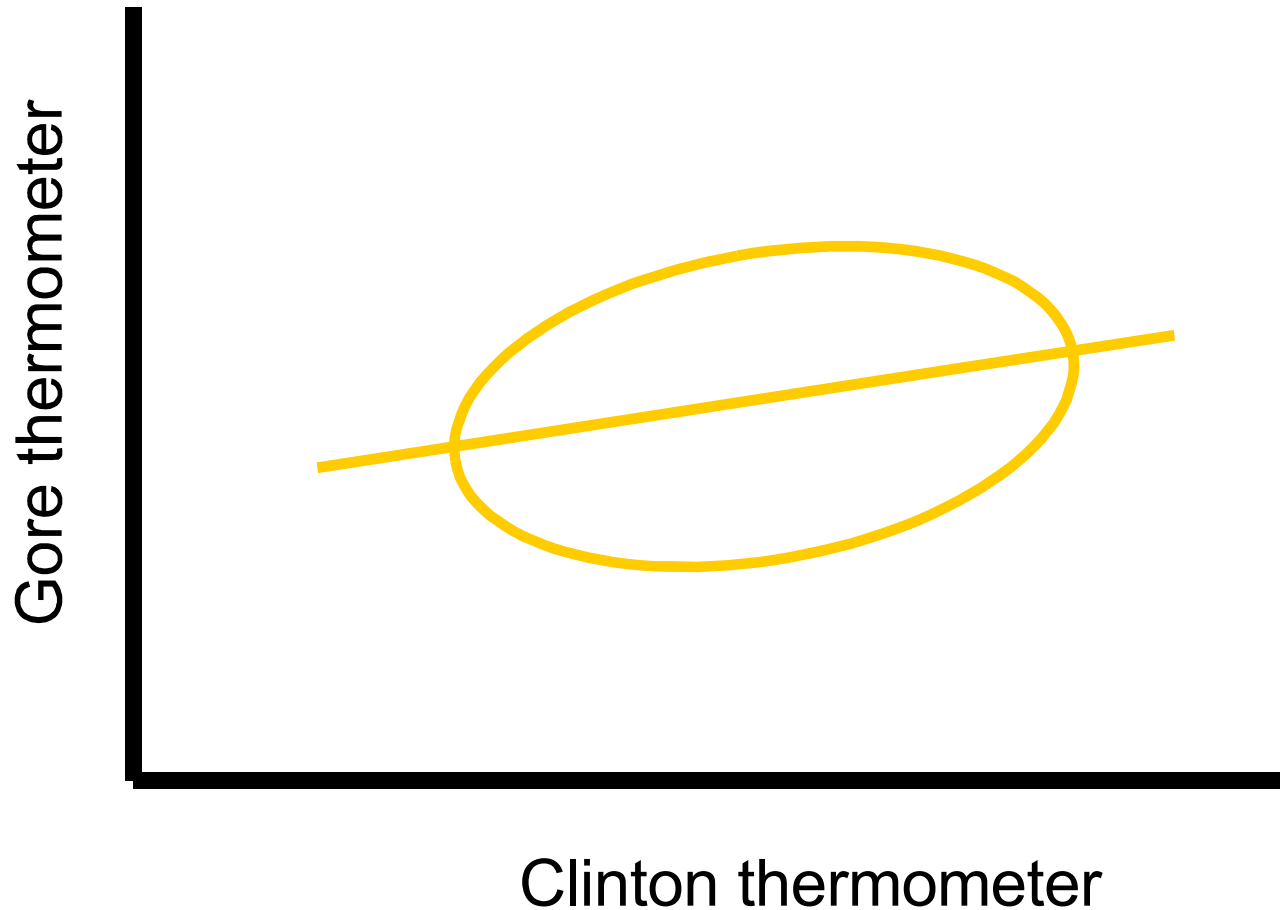
Did Clinton hurt Gore example

- What alternative explanations would you need to address?
- Nonrandom selection into the treatment group (disliking Clinton) from many sources
- Let's address one source: party identification
- How could we do this?
 - Matching: compare Democrats who like or don't like Clinton; do the same for Republicans and independents
 - Multivariate regression: control for partisanship statistically

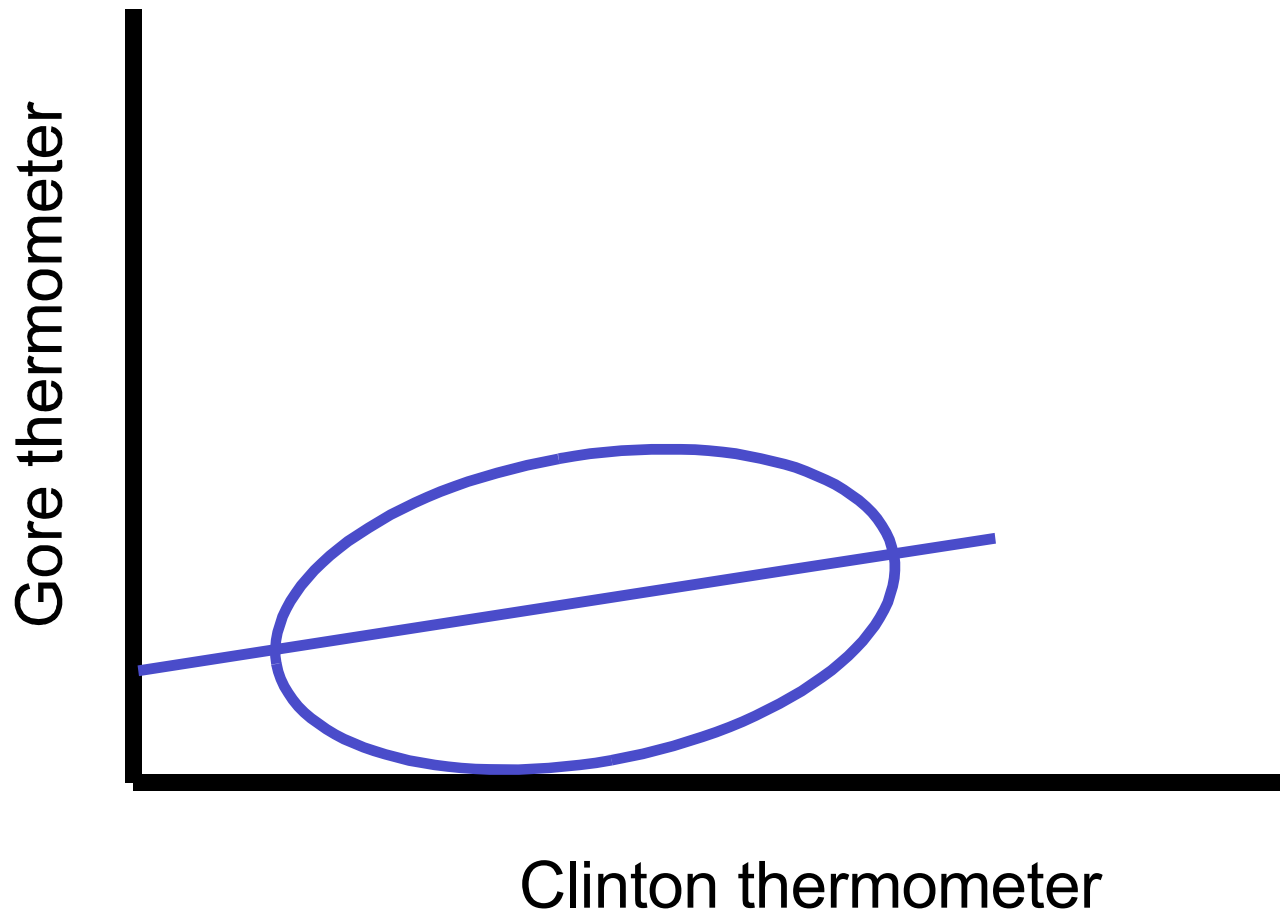
Democratic picture



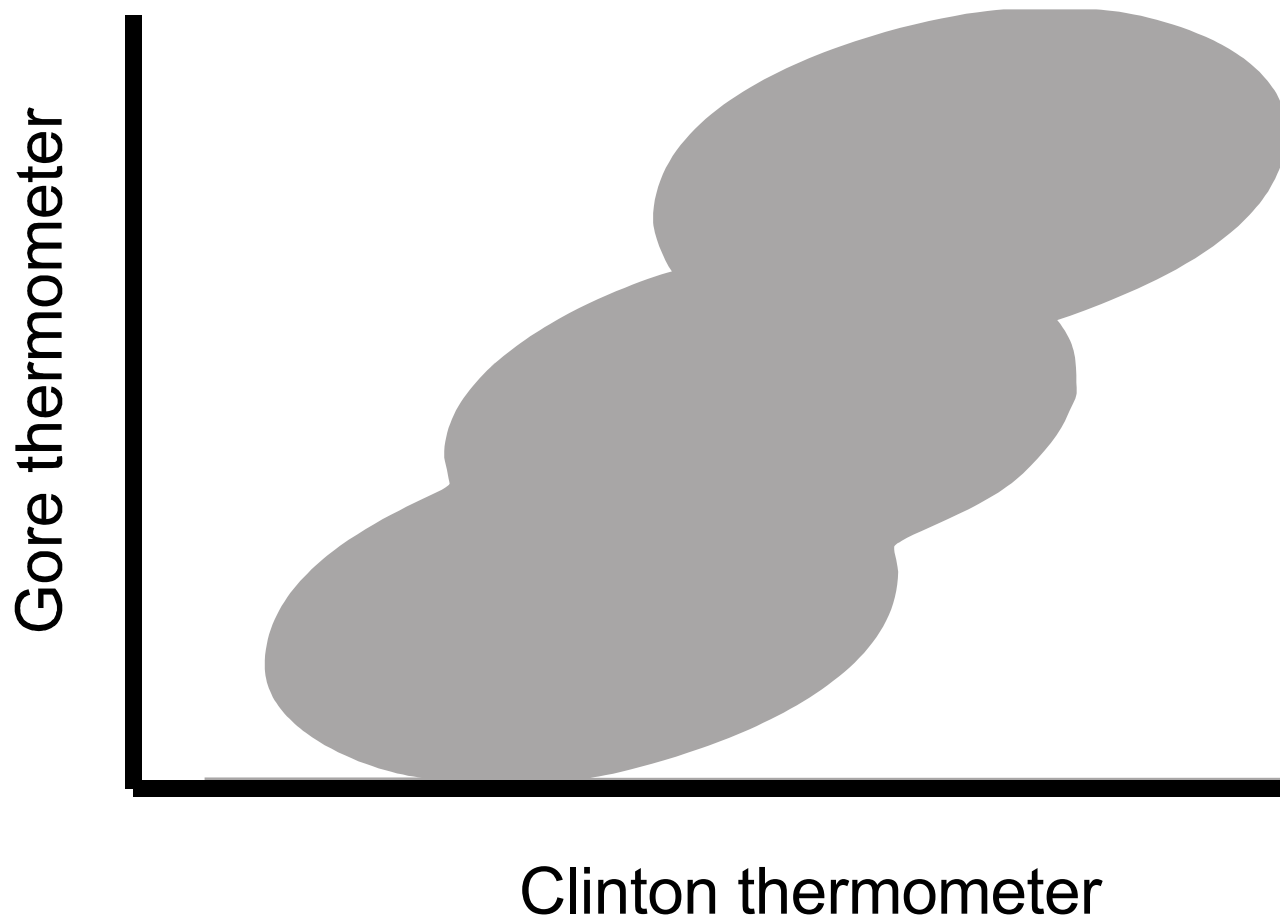
Independent picture



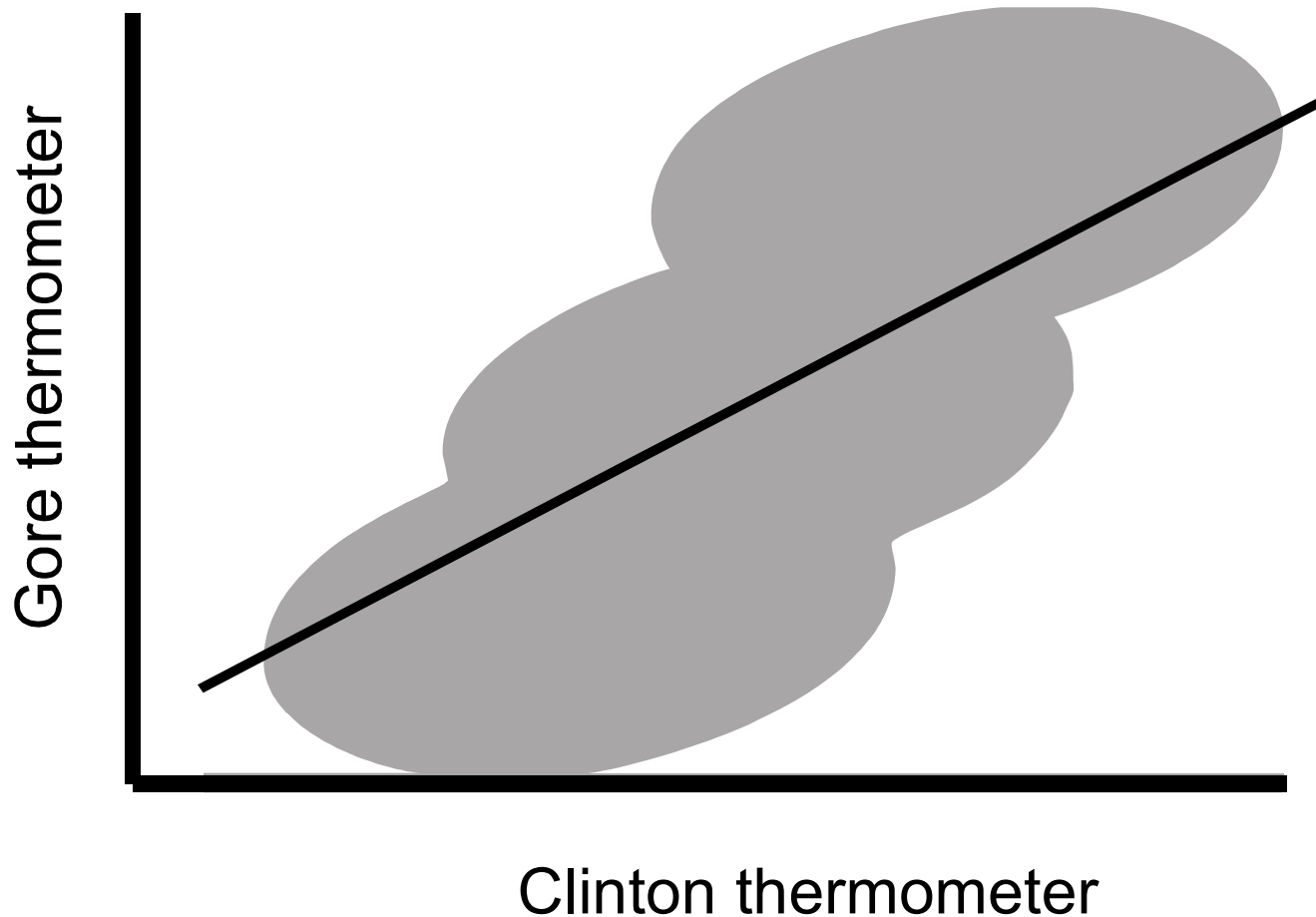
Republican picture



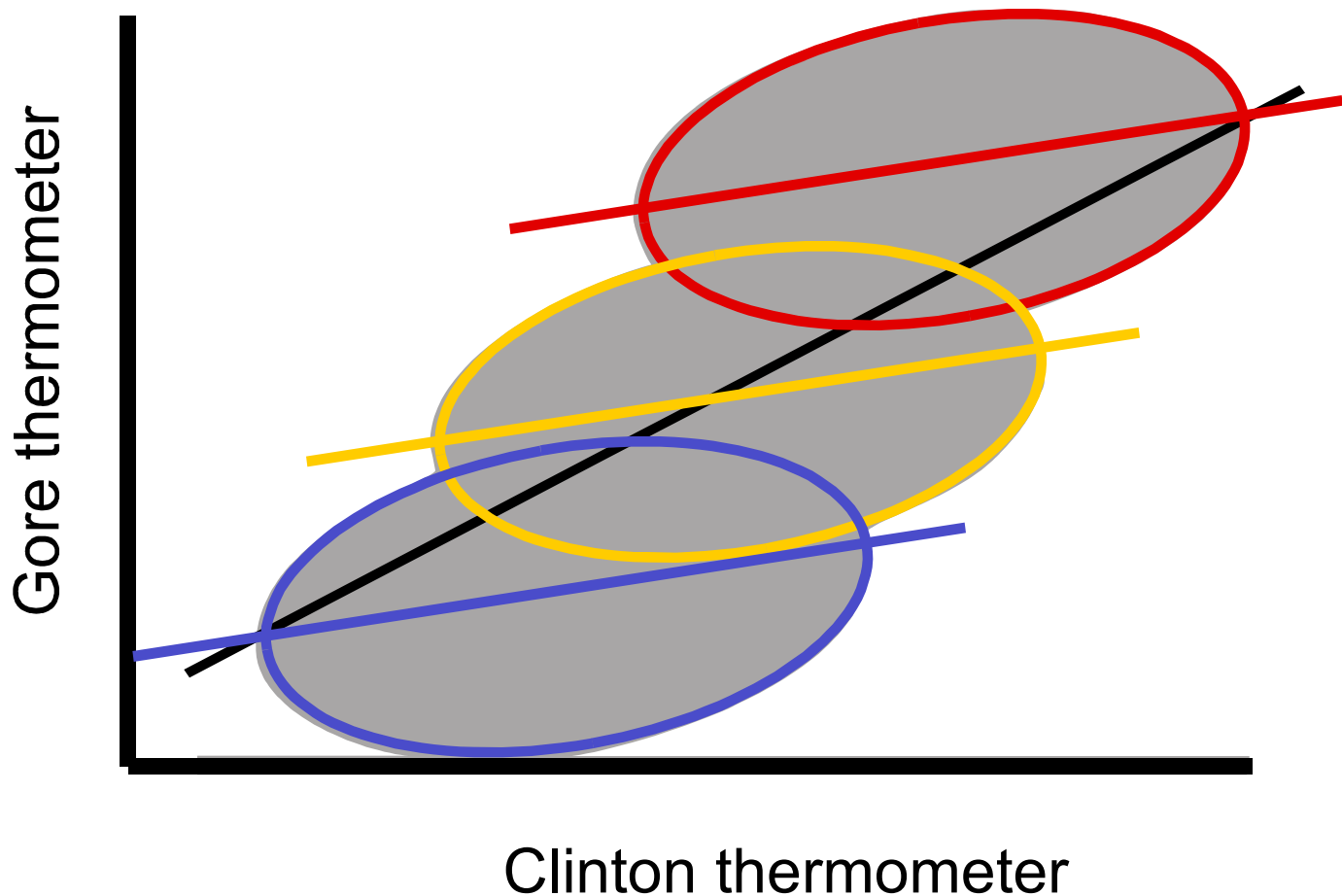
Combined data picture



Combined data picture with regression: bias!

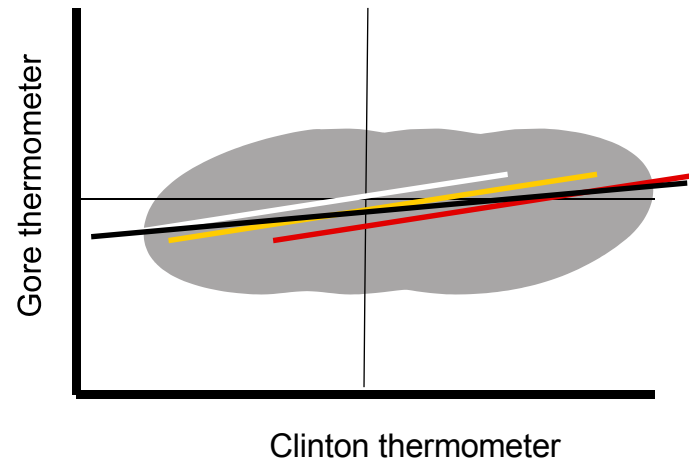


Combined data picture with “true” regression lines overlaid

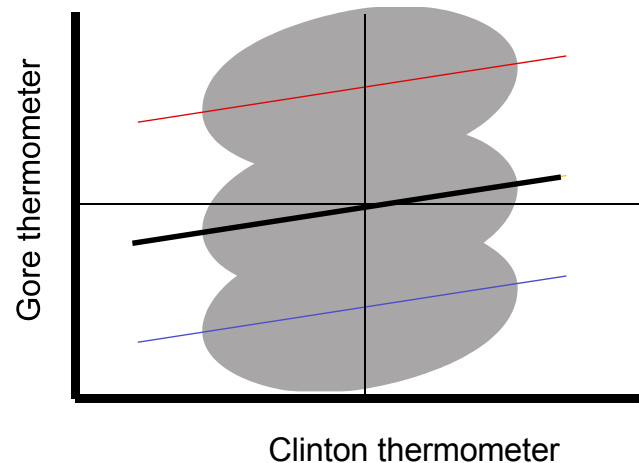


Tempting yet wrong normalizations

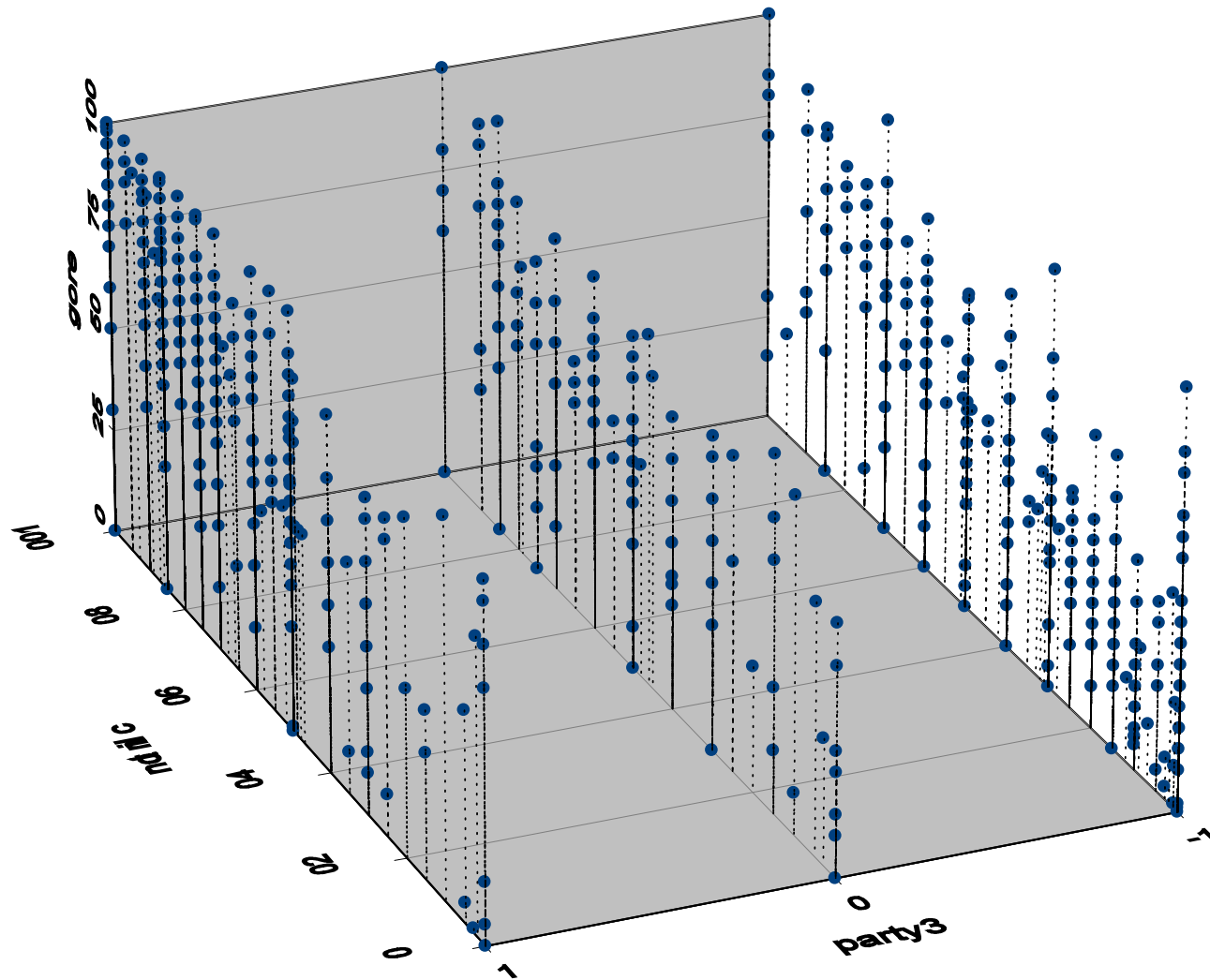
Subtract the Gore therm. from the avg. Gore therm. score



Subtract the Clinton therm. from the avg. Clinton therm. score



3D Relationship



The Linear Relationship between Three Variables

Gore
thermometer

Clinton
thermometer

Party ID

$$Y_i = \beta_0 + \beta_1 X_{1,i} + \beta_2 X_{2,i} + \varepsilon_i$$

Multivariate slope coefficients

Bivariate estimate:

$$\hat{\beta}_1^B = \frac{\text{cov}(X_1, Y)}{\text{var}(X_1)} \text{ vs.}$$

Clinton effect
(on Gore) in
bivariate (B)
regression

Multivariate estimate:

$$\hat{\beta}_1^M = \frac{\text{cov}(X_1, Y)}{\text{var}(X_1)} - \hat{\beta}_2^M \frac{\text{cov}(X_1, X_2)}{\text{var}(X_1)}$$

Party ID effect
(on Gore) in
multivariate (M)
regression

Clinton effect
(on Gore) in
multivariate (M)
regression

Clinton effect on
Party ID in
bivariate
regression

When does $\hat{\beta}_1^B = \hat{\beta}_1^M$? Obviously, when $\hat{\beta}_2^M \frac{\text{cov}(X_1, X_2)}{\text{var}(X_1)} = 0$

The Slope Coefficients

$$\hat{\beta}_1 = \frac{\sum_{i=1}^n (\bar{Y} - Y_i)(\bar{X}_1 - X_{1,i})}{\sum_{i=1}^n (\bar{X}_1 - X_{1,i})^2} - \hat{\beta}_2 \frac{\sum_{i=1}^n (\bar{X}_1 - X_{1,i})(\bar{X}_2 - X_{2,i})}{\sum_{i=1}^n (\bar{X}_1 - X_{1,i})^2} \text{ and}$$

$$\hat{\beta}_2 = \frac{\sum_{i=1}^n (\bar{Y} - Y_i)(\bar{X}_2 - X_{2,i})}{\sum_{i=1}^n (\bar{X}_2 - X_{2,i})^2} - \hat{\beta}_1 \frac{\sum_{i=1}^n (\bar{X}_1 - X_{1,i})(\bar{X}_2 - X_{2,i})}{\sum_{i=1}^n (\bar{X}_2 - X_{2,i})^2}$$

X_1 is Clinton thermometer, X_2 is PID, and Y is Gore thermometer

The Slope Coefficients More Simply

$$\hat{\beta}_1 = \frac{\text{cov}(X_1, Y)}{\text{var}(X_1)} - \hat{\beta}_2 \frac{\text{cov}(X_1, X_2)}{\text{var}(X_1)} \text{ and}$$

$$\hat{\beta}_2 = \frac{\text{cov}(X_2, Y)}{\text{var}(X_2)} - \hat{\beta}_1 \frac{\text{cov}(X_1, X_2)}{\text{var}(X_2)}$$

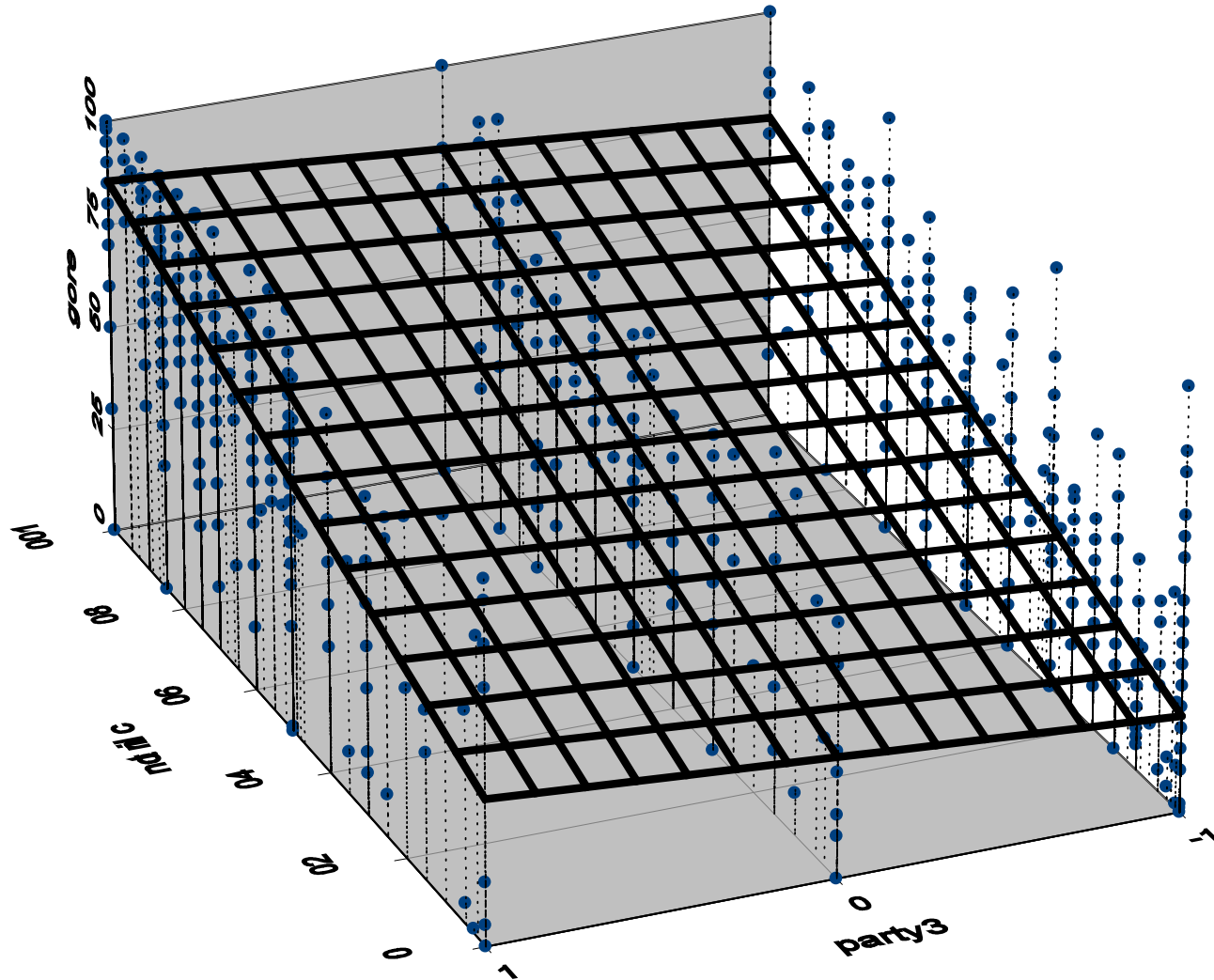
X_1 is Clinton thermometer, X_2 is PID, and Y is Gore thermometer

The Matrix form

y_1	1	$x_{1,1}$	$x_{2,1}$...	$x_{k,1}$
y_2	1	$x_{1,2}$	$x_{2,2}$...	$x_{k,2}$
...	1
y_n	1	$x_{1,n}$	$x_{2,n}$...	$x_{k,n}$

$$\beta = (X'X)^{-1} X'y$$

3D Linear Relationship



The Output

```
. reg gore clinton party3
```

Source	SS	df	MS			
Model	629261.91	2	314630.955	Number of obs =	1745	
Residual	522964.934	1742	300.209492	F(2, 1742) =	1048.04	
Total	1152226.84	1744	660.68053	Prob > F =	0.0000	
				R-squared =	0.5461	
				Adj R-squared =	0.5456	
				Root MSE =	17.327	

	gore	Coef.	Std. Err.	t	P> t	[95% Conf. Interval]	
clinton		.5122875	.0175952	29.12	0.000	.4777776	.5467975
party3		5.770523	.5594846	10.31	0.000	4.673191	6.867856
_cons		28.6299	1.025472	27.92	0.000	26.61862	30.64119

Interpretation of clinton effect: Holding constant party identification, a one-point increase in the Clinton feeling thermometer is associated with a .51 increase in the Gore thermometer.

Separate regressions

	(1)	(2)	(3)
Intercept	23.1	55.9	28.6
Clinton	0.62	--	0.51
Party	--	15.7	5.8

Is the Clinton effect causal?

- That is, should we be convinced that negative feelings about Clinton really hurt Gore?
- No!
 - The regression analysis has only ruled out nonrandom selection on party ID.
 - Nonrandom selection into the treatment could occur from
 - Variables other than party ID, or
 - Reverse causation, which is feelings about Gore influencing feelings about Clinton.
 - Additionally, the regression analysis may not have entirely ruled out nonrandom selection on party ID because it may have assumed the wrong functional form.
 - E.g., what if nonrandom selection on strong Republican/strong Democrat



Summary: Why we control

- Address alternative explanations by removing confounding effects
- Improve efficiency

Why did the Clinton Coefficient change from 0.62 to 0.51

```
. corr gore clinton party, cov  
(obs=1745)
```

```
-----+-----  
          |      gore   clinton   party3  
gore      |  660.681  
clinton   |  549.993   883.182  
party3    |  13.7008   16.905   .8735
```

The Calculations

$$\hat{\beta}_1^B = \frac{\text{cov}(gore, clinton)}{\text{var}(clinton)} = \frac{549.993}{883.182} = 0.6227$$

$$\hat{\beta}_1^M = \frac{\text{cov}(gore, clinton)}{\text{var}(clinton)} - \hat{\beta}_2^M \frac{\text{cov}(clinton, party)}{\text{var}(clinton)}$$

$$= \frac{549.993}{883.182} - 5.7705 \frac{16.905}{883.182}$$

$$= 0.6227 - 0.1105$$

$$= 0.5122$$

```
. corr gore clinton party, cov
(obs=1745)

      |      gore  clinton  party3
-----+-----
gore  |  660.681
clinton |  549.993  883.182
party3 |  13.7008  16.905   .8735
```


Accounting for total effects

$$\hat{\beta}_1^M = \frac{\text{cov}(X_1, Y)}{\text{var}(X_1)} - \hat{\beta}_2 \frac{\text{cov}(X_1, X_2)}{\text{var}(X_1)}$$

(i.e., regression coefficient
when we regress X_2 (as dep. var.)
on X_1 (as ind. var.)

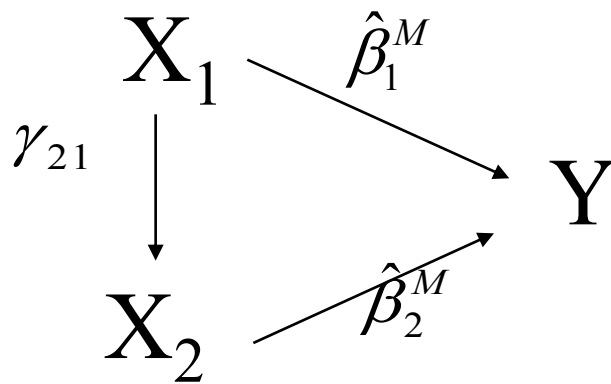
$$\hat{\beta}_1^M = \hat{\beta}_1^B - \hat{\beta}_2^M \gamma_{21}^M$$

$$\hat{\beta}_1^B = \hat{\beta}_1^M + \hat{\beta}_2^M \gamma_{21}^M$$

Accounting for the total effect

$$\hat{\beta}_1^B = \hat{\beta}_1^M + \hat{\beta}_2^M \gamma_{21}$$

Total effect = Direct effect + indirect effect



Accounting for the total effects in the Gore thermometer example

Effect	Total	Direct	Indirect
Clinton	0.62	0.51	0.11
Party	15.7	5.8	9.9

Other approaches to addressing confounding effects?

- Experiments
 - Difference-in-differences designs
 - Others?
-
- Is regression the best approach to addressing confounding effects?
 - Problems



Drinking and Greek Life Example

- Why is there a correlation between living in a fraternity/sorority house and drinking?
 - Greek organizations often emphasize social gatherings that have alcohol. The effect is being in the Greek organization itself, not the house.
 - There's something about the House environment itself.

Dependent variable: Times Drinking in Past 30 Days

C8. When did you last have a drink (that is more than just a few sips)?

- I have never had a drink → Skip to C22 (page 10)
- Not in the past year → Skip to C22 (page 10)
- More than 30 days ago, but in the past year → Skip to C17 (page 8)
- More than a week ago, but in the past 30 days → Go to C9
- Within the last week → Go to C9

C9. On how many occasions have you had a drink of alcohol in the past 30 days? (Choose one answer.)

- | | | |
|---|--|--|
| <input type="radio"/> Did not drink in the last 30 days | <input type="radio"/> 6 to 9 occasions | <input type="radio"/> 20 to 39 occasions |
| <input type="radio"/> 1 to 2 occasions | <input type="radio"/> 10 to 19 occasions | <input type="radio"/> 40 or more occasions |
| <input type="radio"/> 3 to 5 occasions | | |

```
. infix age 10-11 residence 16 greek 24 screen 102
timespast30 103 howmuchpast30 104 gpa 278-279 studying 281
timeshs 325 howmuchhs 326 socializing 283 stwgt_99 475-493
weight99 494-512 using da3818.dat,clear
(14138 observations read)
```

```
. recode timespast30 timeshs (1=0) (2=1.5) (3=4) (4=7.5)
(5=14.5) (6=29.5) (7=45)
(timespast30: 6571 changes made)
(timeshs: 10272 changes made)
```

```
. replace timespast30=0 if screen<=3
(4631 real changes made)
```

```
. tab timespast30
```

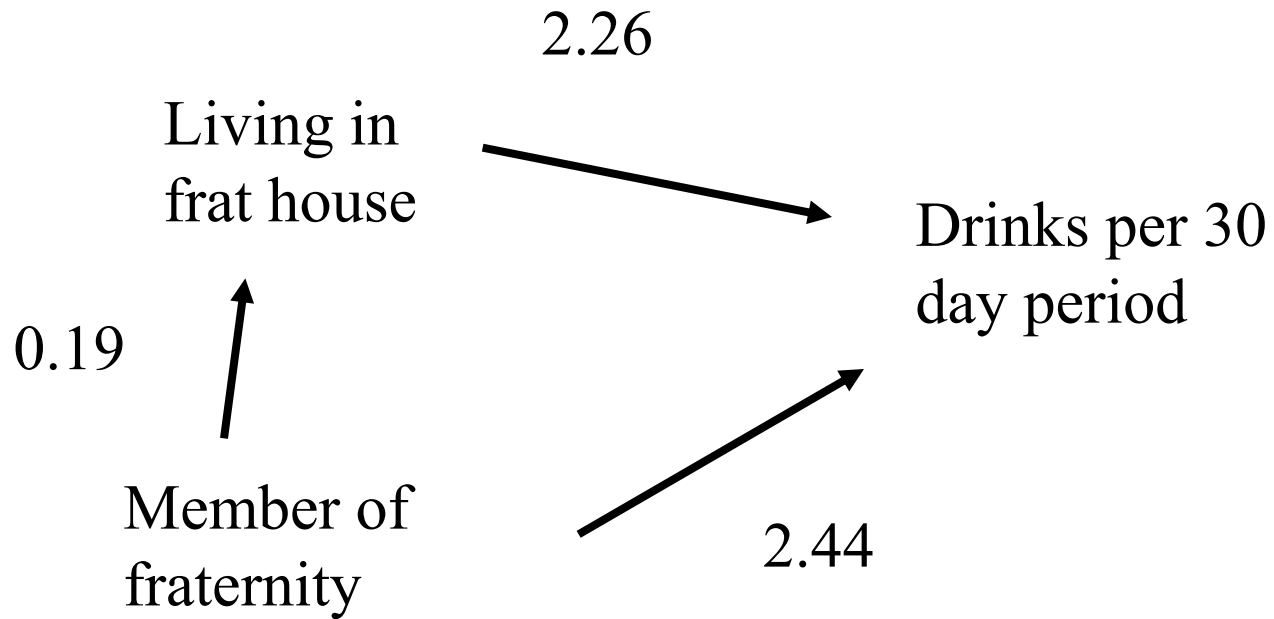
timespast30	Freq.	Percent	Cum.
0	4,652	33.37	33.37
1.5	2,737	19.64	53.01
4	2,653	19.03	72.04
7.5	1,854	13.30	85.34
14.5	1,648	11.82	97.17
29.5	350	2.51	99.68
45	45	0.32	100.00
Total	13,939	100.00	

Three Regressions

Dependent variable: number of times drinking in past 30 days			
Live in frat/sor house	4.44 (0.35)	---	2.26 (0.38)
Member of frat/sor	---	2.88 (0.16)	2.44 (0.18)
Intercept	4.54 (0.56)	4.27 (0.059)	4.27 (0.059)
R2	.011	.023	.025
N	13,876	13,876	13,876

Note: Corr. Between living in frat/sor house and being a member of a Greek organization is .42

The Picture



Accounting for the effects of frat house living and Greek membership on drinking

Effect	Total	Direct	Indirect
Member of Greek org.	2.88	2.44 (85%)	0.44 (15%)
Live in frat/ sor. house	4.44	2.26 (51%)	2.18 (49%)