



# Addressing Alternative Explanations: Multiple Regression

17.871

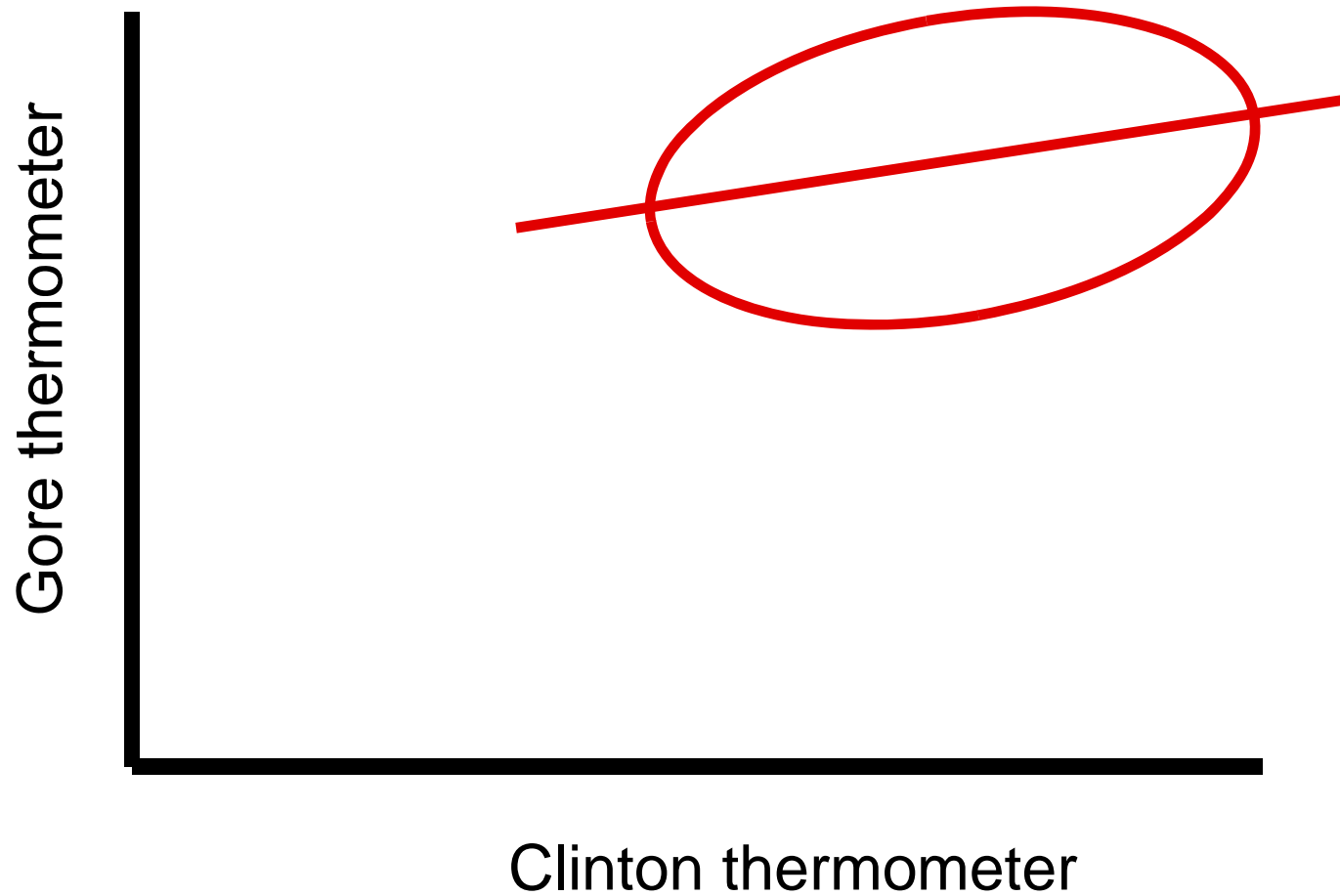


# Gore Likeability Example

- Did Clinton hurt Gore in the 2000 election?
- How would you test this?
- What alternative explanations would you need to address?
  
- Other examples of alternative explanations based on omitted variables?

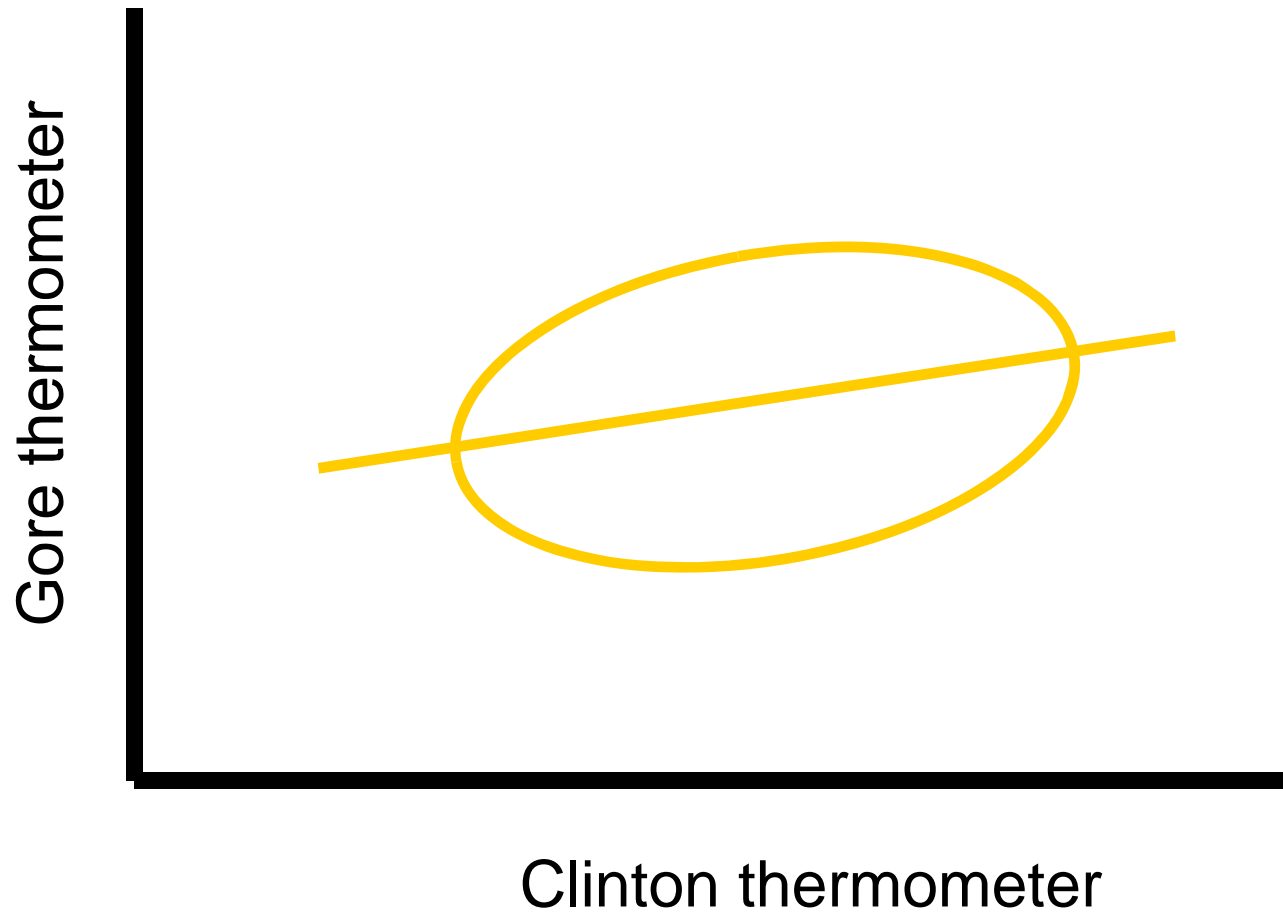


# Democratic picture

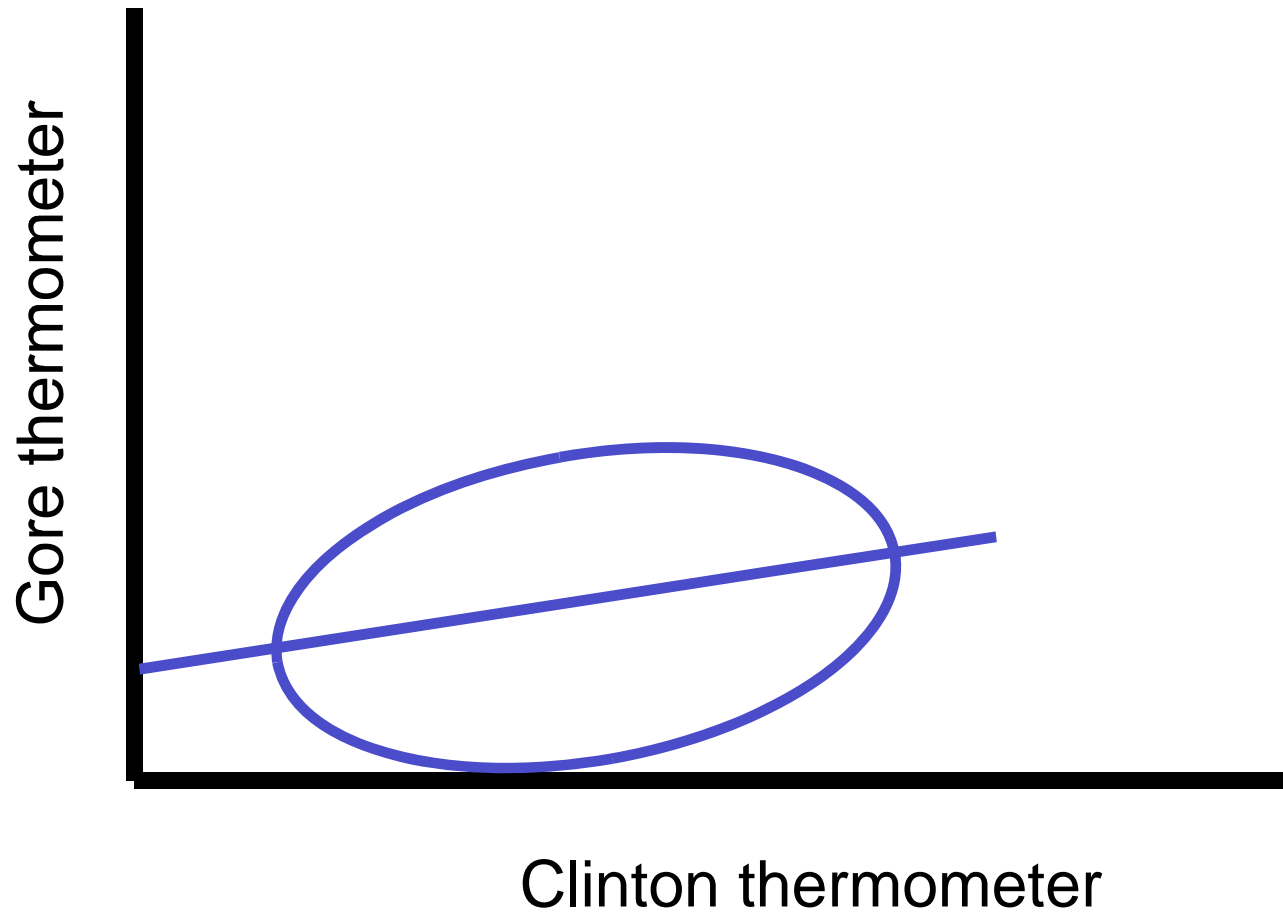




# Independent picture

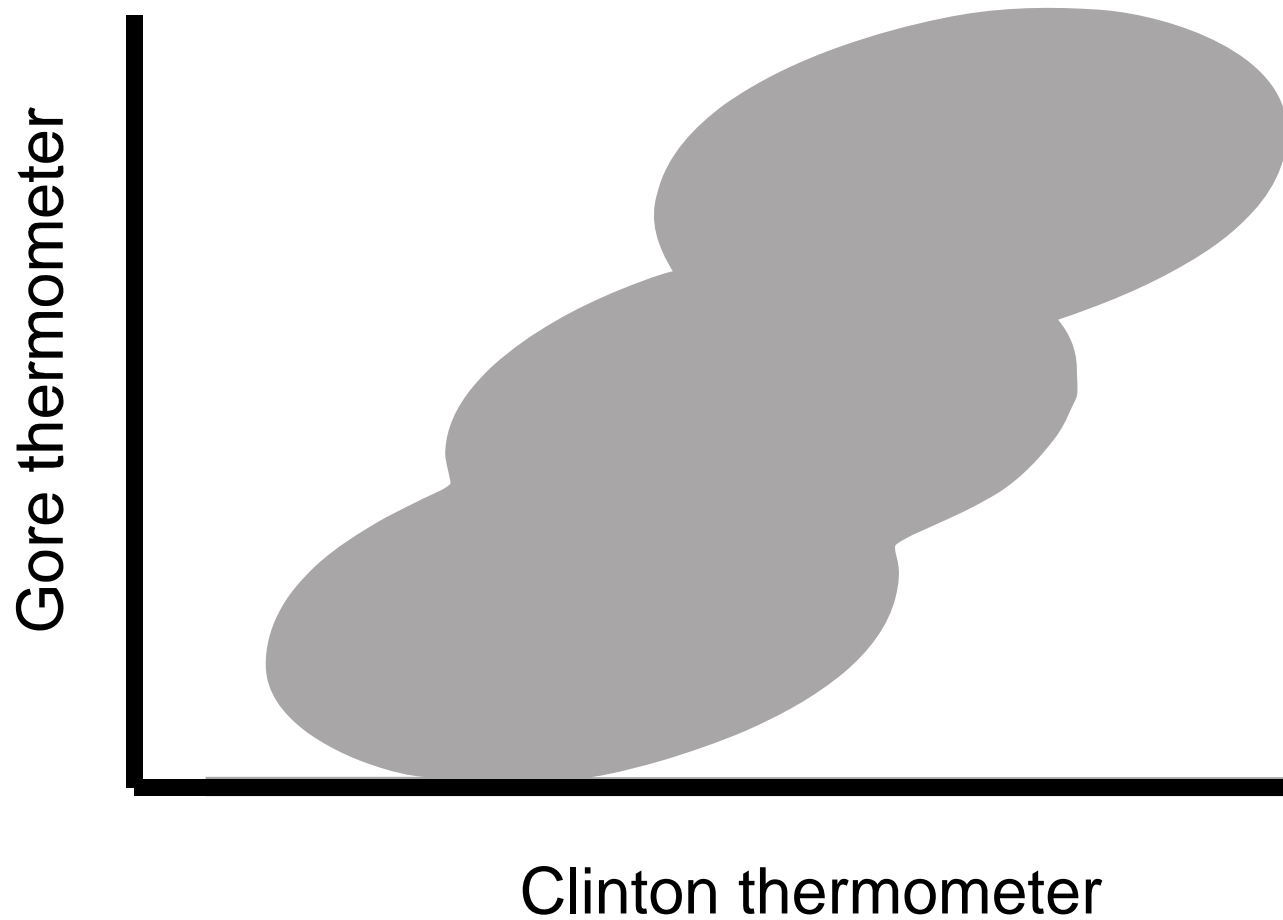


# Republican picture

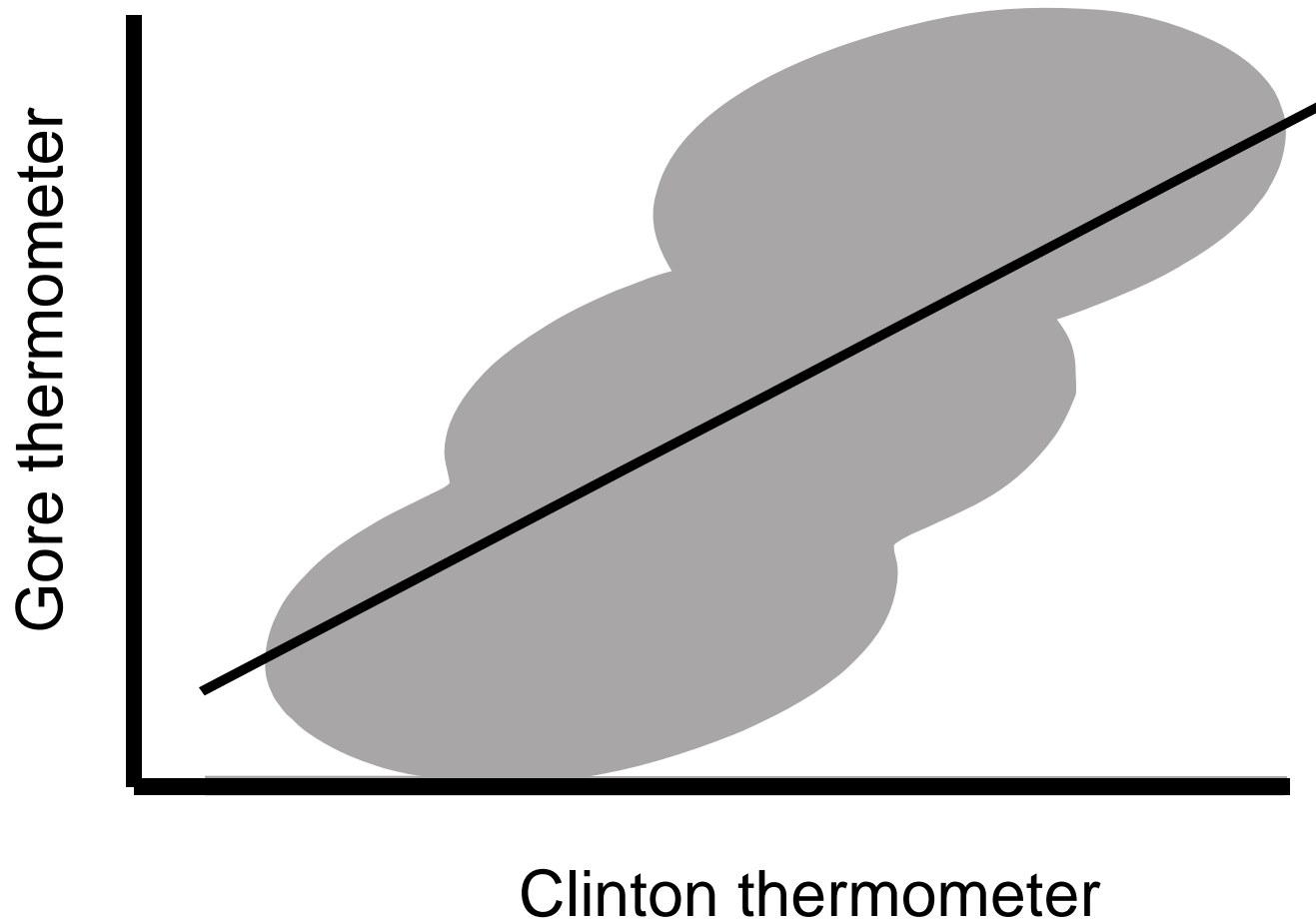




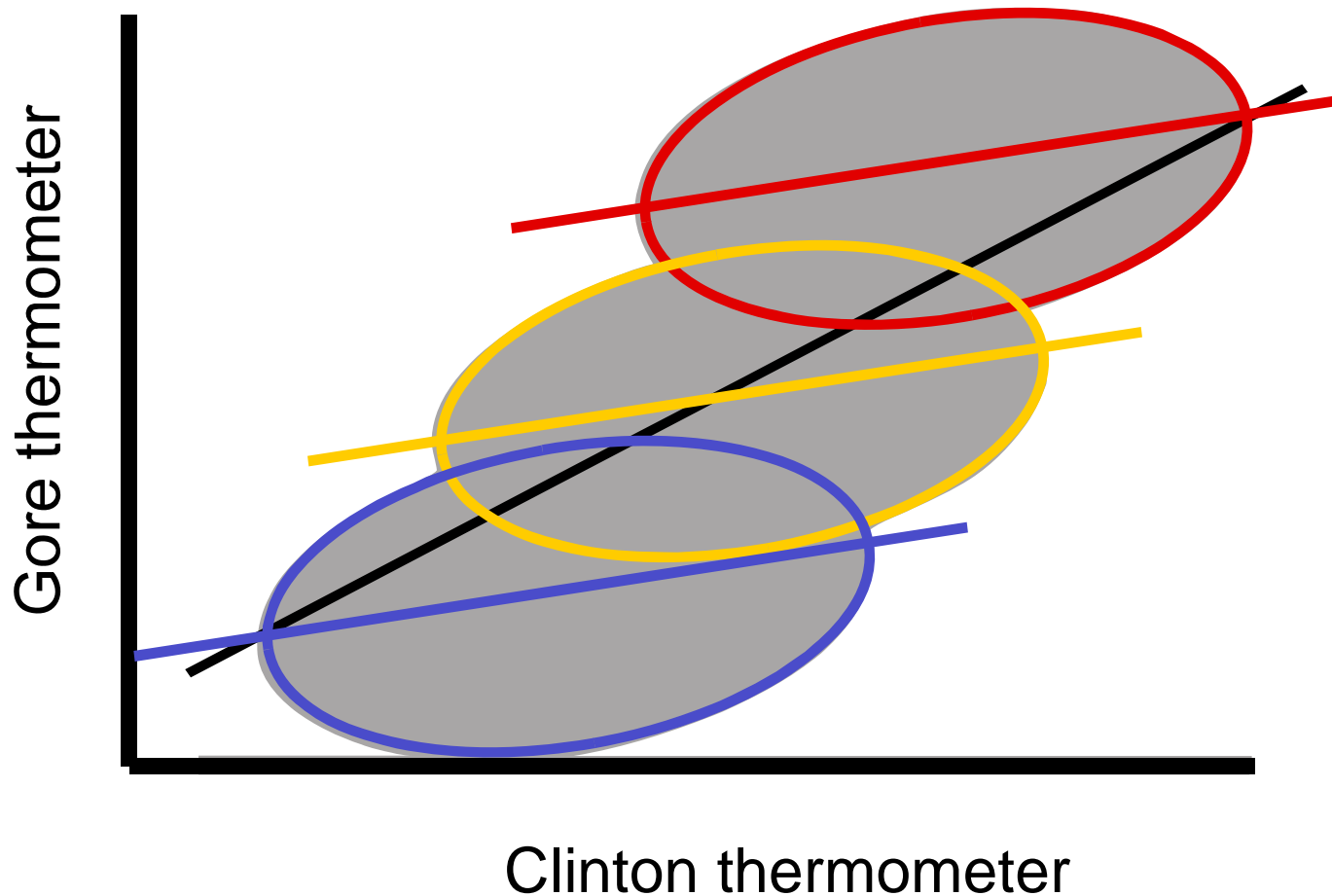
# Combined data picture



# Combined data picture with regression



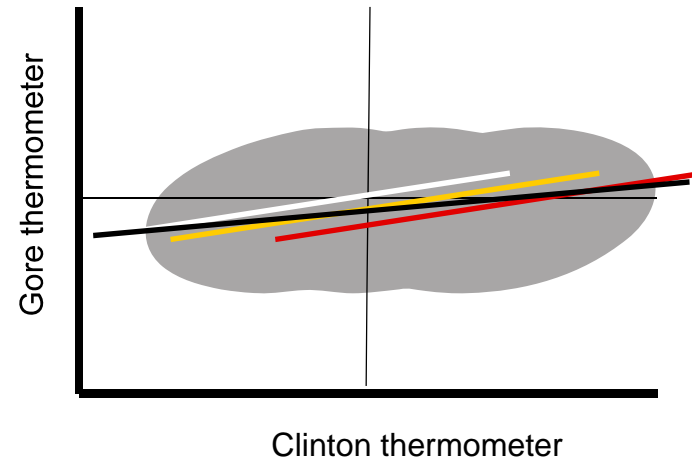
# Combined data picture with “true” regression lines overlaid



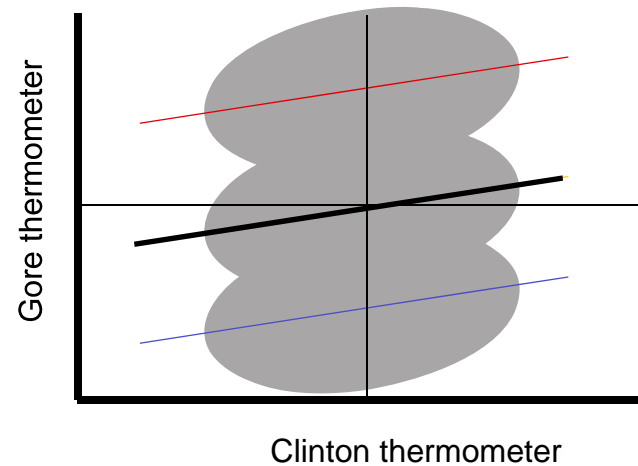


# Tempting yet wrong normalizations

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



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

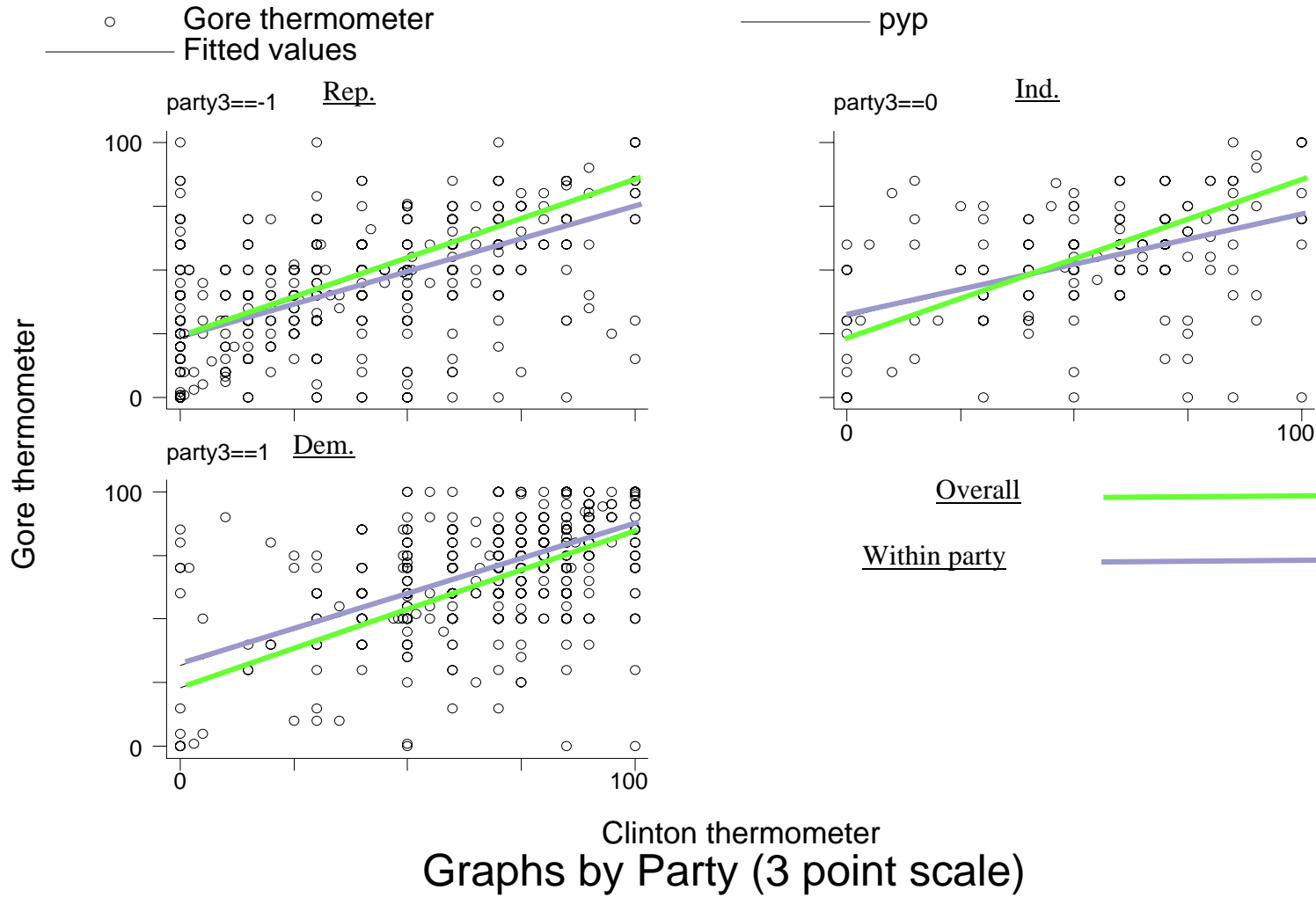




# Summary: Why we control

- Address alternative explanations by removing confounding effects
- Improve efficiency

# Gore vs. Clinton

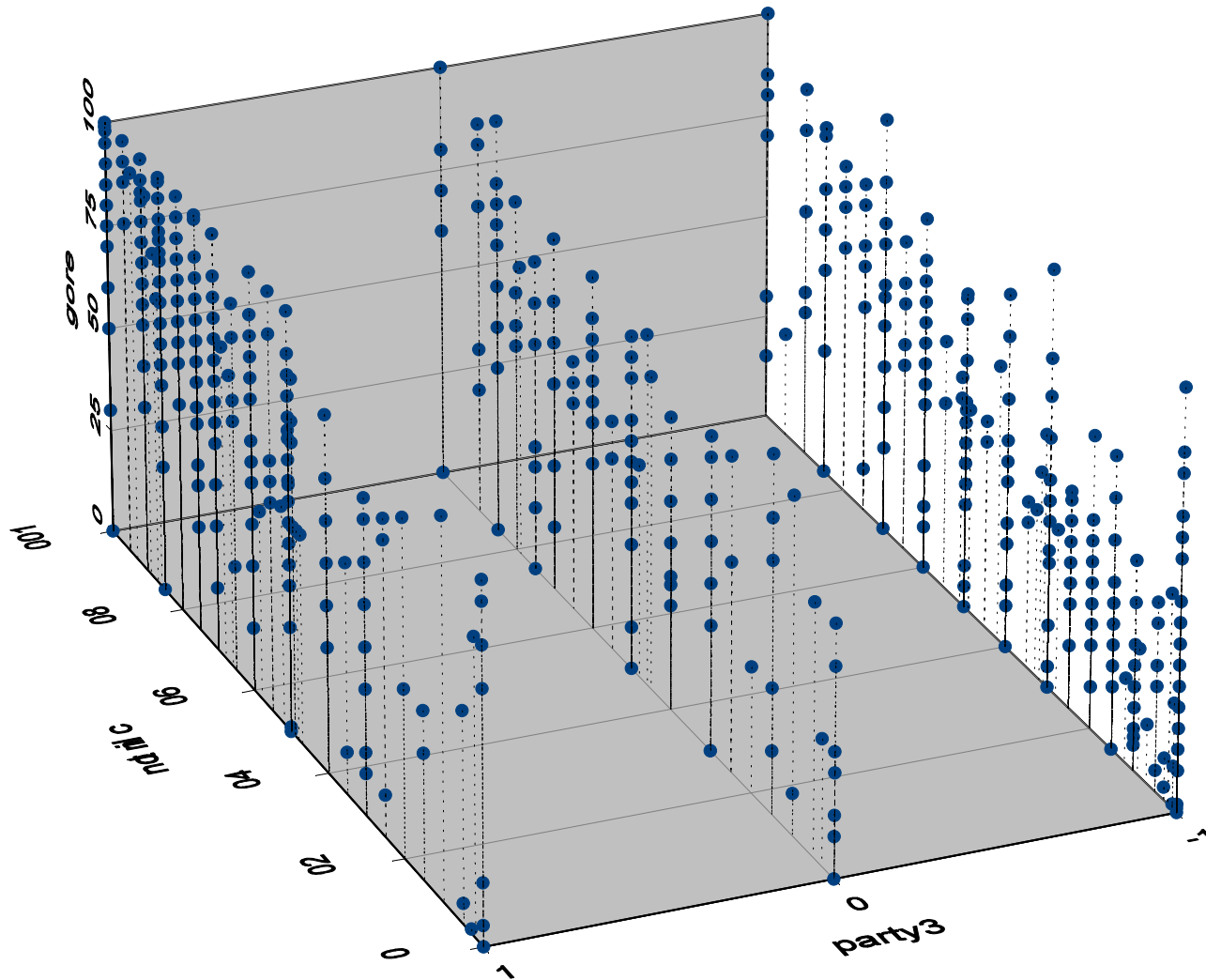




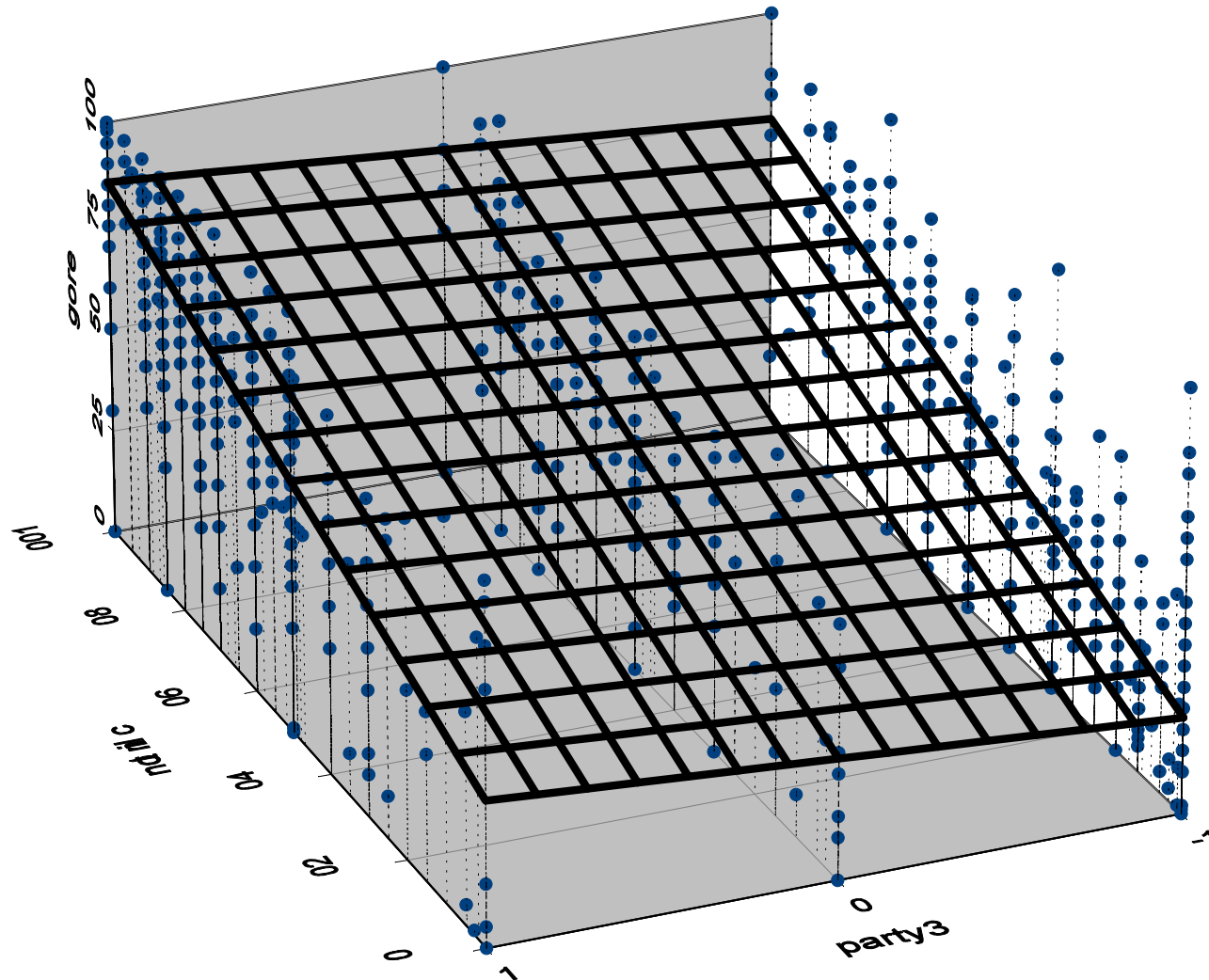
# The Linear Relationship between Three Variables

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

# 3D Relationship



# 3D Linear Relationship



# The Slope Coefficients

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

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



# The Slope Coefficients More Simply

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

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



# 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	$x_{1,n}$	$x_{2,n}$	...	$x_{k,n}$

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



# Consider two regression coefficients

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

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

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



# Separate regressions

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

# 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

# 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

# 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.))

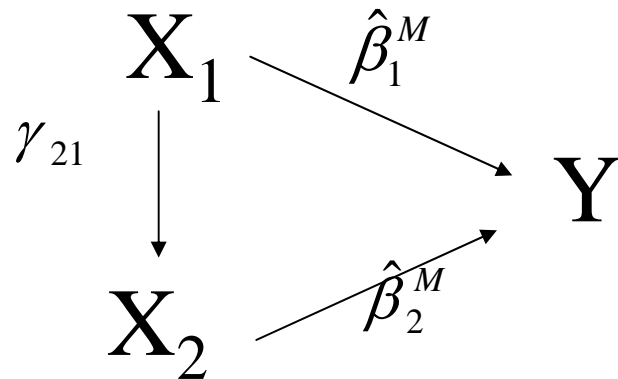
$$\hat{\beta}_1^M = \hat{\beta}_1^B - \hat{\beta}_2^M \gamma_{21}^M$$

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

# 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.)**

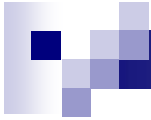
- |   |  |  |
|---|--|--|
| <input type="radio"/> Did not drink in the last 30 days | <input type="radio"/> 6 to 9 occasions   | <input type="radio"/> 20 to 39 occasions   |
| <input type="radio"/> 1 to 2 occasions                  | <input type="radio"/> 10 to 19 occasions | <input type="radio"/> 40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

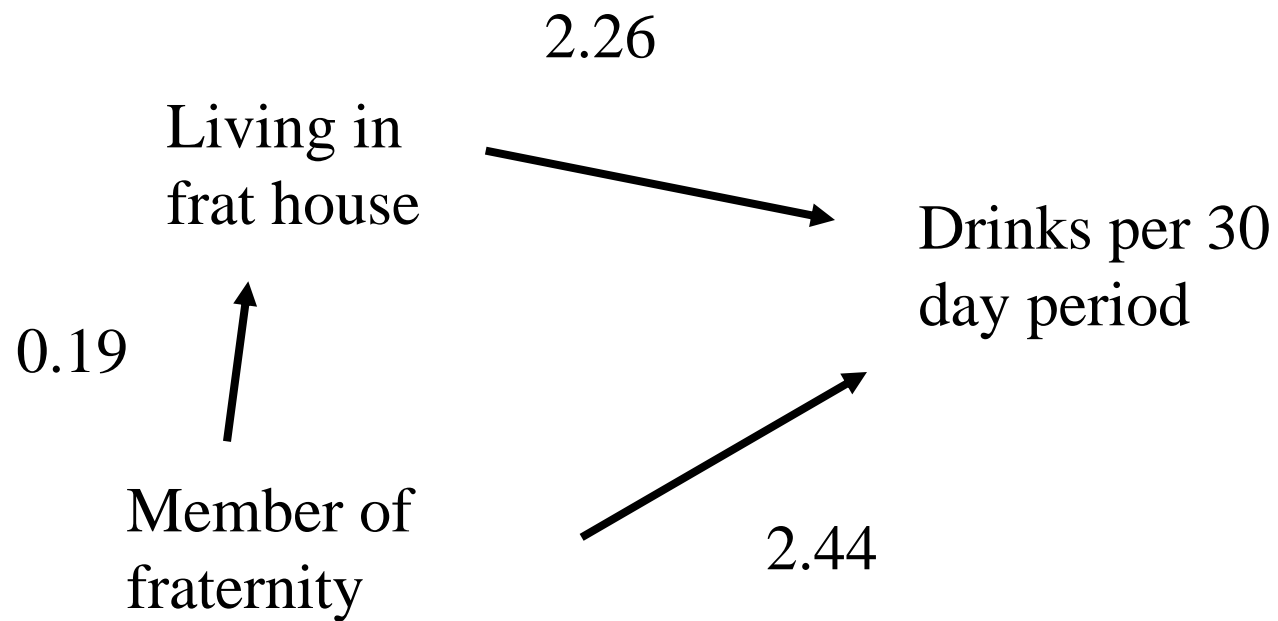


# Three Regressions


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%)