

# **PSC7475: Varying Effects by Group**

## **Week 8: Lecture 13**

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Slides Updated: 2025-03-10

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- This week: finishing up regression!

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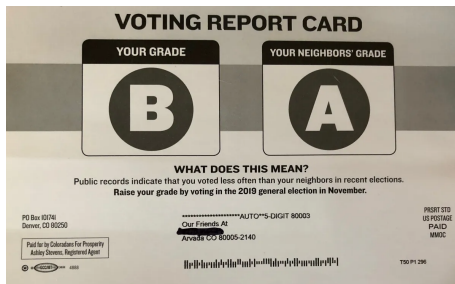
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  - Important questions for determining who should receive treatment.

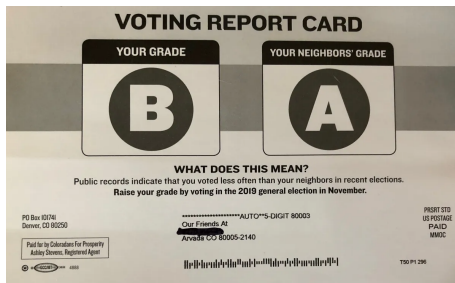
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# Social pressure experiment

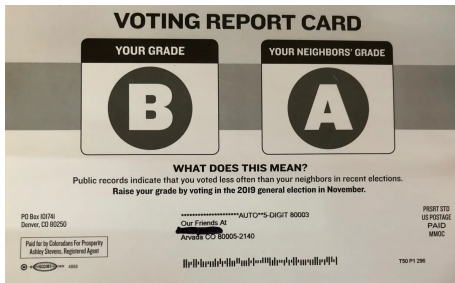


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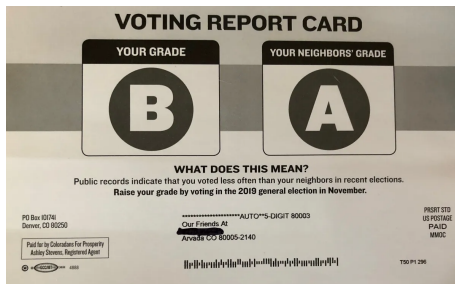
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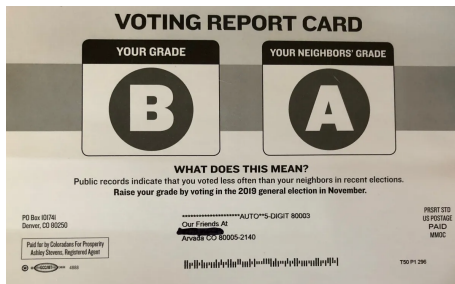
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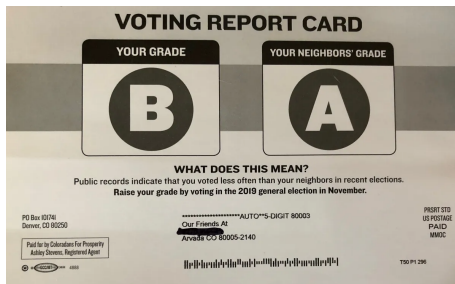


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  - Subsets, subsets, subsets.

# Social pressure experiment



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- Do 2004 voters react differently to social pressure mailer than nonvoters?
- Two approaches:
  - Subsets, subsets, subsets.
  - Interaction terms in regression.

# Subset approach

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- Easy way to estimate heterogeneous effects: our old friend, `filter()`, `group_by()`, and `summarize()`. Woo!

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- Easy way to estimate heterogeneous effects: our old friend, `filter()`, `group_by()`, and `summarize()`. Woo!
  - First, get the data

```
data(social, package="qss")
```

# Subset approach

- Now, estimate the ATE for the **voters**:

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```
VotersATE <- social %>%  
  filter(primary2004 == 1,  
         messages %in% c("Control", "Neighbors")) %>%  
  group_by(messages) %>%  
  summarize(primary2006_mean = mean(primary2006)) %>%  
  pivot_wider(names_from = "messages",  
              values_from = "primary2006_mean") %>%  
  mutate(ate_v = Neighbors - Control) %>%  
  select(ate_v)  
VotersATE
```

```
## # A tibble: 1 x 1  
##   ate_v  
##   <dbl>  
## 1 0.0965
```

# Filter approach

- Now, estimate the ATE for the **nonvoters**:



# Filter approach

- Now, estimate the ATE for the **nonvoters**:

```
NonvotersATE <- social %>%  
  filter(primary2004 == 0,  
         messages %in% c("Control", "Neighbors")) %>%  
  group_by(messages) %>%  
  summarize(primary2006_mean = mean(primary2006)) %>%  
  pivot_wider(names_from = "messages",  
              values_from = "primary2006_mean") %>%  
  mutate(ate_nv = Neighbors - Control) %>%  
  select(ate_nv)  
NonvotersATE
```

```
## # A tibble: 1 x 1  
##   ate_nv  
##   <dbl>  
## 1 0.0693
```

# Difference in effects

- How much does the estimated treatment effect differ between groups?

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```
VotersATE$ate_v - NonvotersATE$ate_nv
```

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## [1] 0.02722908
```

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- How much does the estimated treatment effect differ between groups?

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- Any easier way to allow for different effects of treatment by groups?

# Interaction terms

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$$\text{turnout}_i = \alpha + \beta_1 \text{primary2004}_i + \beta_2 \text{neighbors}_i + \beta_3 (\text{primary2004}_i \times \text{neighbors}_i) + \varepsilon_i$$

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- Primary 2004 variable multiplied by the neighbors variable.
  - Equal to 1 if voted in 2004 (`primary2004 == 1`) and received neighbors mailer (`neighbors == 1`)



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- Primary 2004 variable multiplied by the neighbors variable.
  - Equal to 1 if voted in 2004 ( $\text{primary2004} == 1$ ) and received neighbors mailer ( $\text{neighbors} == 1$ )
  - Easiest to understand by investigating predicted values.

# Predicted values from non-interacted model

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	Control ( $Z_i = 0$ )	Neighbors ( $Z_i = 1$ )
non-voter ( $X_i = 0$ )	$\hat{\alpha}$	$\hat{\alpha} + \hat{\beta}_2$
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- Effect of Neighbors for voters:  $(\hat{\alpha} + \hat{\beta}_1 + \hat{\beta}_2) - (\hat{\alpha} + \hat{\beta}_1) = \hat{\beta}_2$

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# Interpreting coefficients

$$\hat{Y}_i = \hat{\alpha} + \hat{\beta}_1 \text{primary2004}_i + \hat{\beta}_2 \text{neighbors}_i \\ + \hat{\beta}_3 (\text{primary2004}_i \times \text{neighbors}_i)$$

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- $\hat{\alpha}$ : turnout rate for 2004 nonvoters in control group.

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- $\hat{\alpha}$ : turnout rate for 2004 nonvoters in control group.
- $\hat{\beta}_1$ : avg difference in turnout between 2004 voters and nonvoters.
- $\hat{\beta}_2$ : effect of neighbors for 2004 nonvoters.
- $\hat{\beta}_3$ : difference in the effect of neighbors mailer between 2004 voters and nonvoters.

# Interactions in R

- You can include an interaction with var1:var2:

```
social.neighbor <- social %>%  
  mutate(neighbors = ifelse(messages=="Neighbors",1,  
                             ifelse(messages=="Control",0,NA)))  
  select(primary2006,primary2004,neighbors) %>%  
  drop_na()
```

```
fit <- lm(primary2006 ~ primary2004 + neighbors +  
          primary2004:neighbors, data = social.neighbor)
```

```
coef(fit)
```

```
##           (Intercept)           primary2004  
##           0.23710990           0.14869507  
##           neighbors primary2004:neighbors  
##           0.06020617           0.02722008
```

# Interactions in R

```
coef(fit)
```

```
##              (Intercept)              primary2004
##              0.23710990              0.14869507
##              neighbors primary2004:neighbors
##              0.06929617              0.02722908
```



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```

- Compare coefficients to earlier approach:

```
NonvotersATE$ate_nv
```

```
## [1] 0.06929617
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##              (Intercept)              primary2004
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```

```
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```

```
VotersATE$ate_v - NonvotersATE$ate_nv
```

```
## [1] 0.02722908
```