

GOV 51 Section

Week 3: Matching, and Unit 1 (Causal Inf.) Review

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Topics

- ▶ Review of matching
- ▶ Causal inference (DiD, IV, Matching, & assumptions) review
- ▶ Project discussion

Motivating Matching

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- ▶ Essentially: how close can we get to a clone to get a true counter-factual
- ▶ Useful when DiD & IV aren't appropriate *and* we have quality covariate data.

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- ▶ What does that mean?
- ▶ Expected outcome for treated individuals had they not been treated (which is unobserved) can be estimated using the expected outcome of similar untreated individuals (from the matched set \mathcal{M}), making the untreated group a valid counterfactual.

Matching Example

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Unit	Pre-treatment Covariates		Treatment Indicator	Observed Treated	Outcomes Matched
	Treated	Matched			
1	X_1 ,	X_1^M	1, 0	$Y_1(1)$	$Y_1^M(0)$
2	X_2 ,	X_2^M	1, 0	$Y_2(1)$	$Y_2^M(0)$
3	X_3 ,	X_3^M	1, 0	$Y_3(1)$	$Y_3^M(0)$
4	X_4 ,	X_4^M	1, 0	$Y_4(1)$	$Y_4^M(0)$
\vdots	\vdots	\vdots	\vdots	\vdots	\vdots
N	X_N ,	X_N^M	1, 0	$Y_N(1)$	$Y_N^M(0)$

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How To Do Mahalanobis Matching?

Given a set of covariates X , the Mahalanobis distance between observations i and j is:

$$d_M(i, j) = \sqrt{(X_i - X_j)^T S^{-1} (X_i - X_j)} \quad (1)$$

where:

- ▶ X_i, X_j are covariate vectors.
- ▶ S is the empirical covariance matrix.
- ▶ S^{-1} accounts for correlations, ensuring scale-invariance.

How To Do Mahalanobis Matching?

1. Compute $d_M(i, j)$ for all treated-control pairs.
2. Perform matching:
 - ▶ **Greedy Matching:** Iteratively select closest pairs.
 - ▶ **Optimal Matching:** Minimize total distance.
3. Check balance using standardized mean differences.
4. Estimate treatment effects using matched pairs.

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```
library(MatchIt)
```

```
m.out <- matchit(mailer ~ age + voted96 + newvote +  
                dem,  
                data = gotv,  
                distance = "mahalanobis",  
                replace = TRUE,  
                exact = ~white)
```

```
matchsum <- summary(m.out)
```

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- ▶ So far, we have discussed at least three causal inference mechanisms:
 - ▶ Difference-in-Difference
 - ▶ Instrumental Variables
 - ▶ Matching
- ▶ Let's put your knowledge to the test!
- ▶ Jeopardy game: <https://jeopardylabs.com/play/gov-51-causal-inference-jeopard-review>

Final Project

Time to chat with others: what are you interested in studying?
what kind of data would you use / create? what kind of methods
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Problem Set I released, due by **February 20th, 2020 by 11:59pm
via Gradescope**. We will only grade your compiled PDF. Come to
OH if you need help!