Final Project: Childhood Bullying and Subsequent Drug Use

Shelley Facente, Steph Holm, Lizzy Kinnard, Veronica Pear

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Structural Causal Model

Causal Question: What is the effect of having been bullied prior to age 12 on incidence of drug use in adolescence or adulthood?

Target Causal Parameter:

▶ Difference in the counterfactual probability of drug use if all kids were bullied prior to age 12, and the counterfactual probability of drug use if all kids were not bullied prior to age 12:

$$\psi^{F}(P_{U,X}) = P_{U,X}(Y_1 = 1) - P_{U,X}(Y_0 = 1) = E_{U,X}(Y_1) - E_{U,X}(Y_0)$$

where Y_a denotes the counterfactual outcome under an intervention to set bullying status A=a.

Our Observed Data and Link to SCM

- Data are from the National Longitudinal Survey of Youth 1997
- ► A nationally representative cohort of youth age 12-16 (initial n=9000) in 1997, who were subsequently followed longitudinally.
- ► The target population is youth in the United States.

Data Description:

Key variables:

A: Bullying before the age of 12 (asked in 1997)

Y: Incident drug use ("cocaine or other hard drugs") after 1997

W: Race/ethnicity, age, sex, BMI, not living with both biological parents, mother's educational status (all Ws were measured at baseline)

Sample size: 7,703

We initially also considered: sexual orientation, learning/emotional disability, citizenship status

Table 1

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Covariate | Drug use (%) | No drug use (%) | 1330 (0.17266%) |
6373 (0.82734%) ——— | ———— Victim of bullying |
Yes | 319 (3.7%) | 1340 (17.4%) No | 1011 (2.9%) | 1788 (23.2%)
Mother's education | High school or less | 3867 (50.2%) | 732
(9.5%) Some college or more | 598 (7.8%) | 2506 (32.5%) Sex |
Female | 591 (7.7%) | 3218 (41.8%) Male | 739 (9.6%) | 3155
(41%) Race/ethnicity
Black | 227 (2.9%) | 1788 (23.2%) Hispanic | 288 (3.7%) | 1340
(17.4%) White | 815 (10.6%) | 3245 (42.1%)
Living with both biological parents | Lives with both biological
parents | 645 (8.4%) | 3176 (41.2%) Doesn't live with both
biological parents | 685 (8.9%) | 3197 (41.5%) BMI percentile |
0.5 (mean) | 1 (standard deviation)
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Marginal distribution of exposure and outcome:

Variable	No	Yes
Bullied < 12	6205	1498
Incident drug use	6373	1330

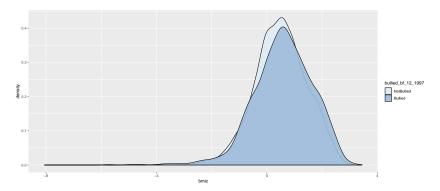


Estimation: Unadjusted ATE & SuperLearner

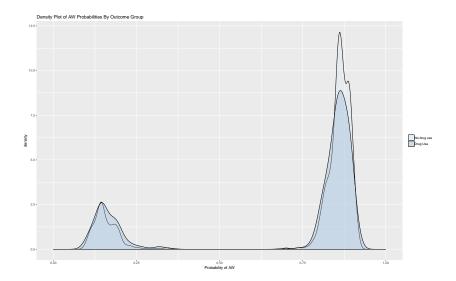
- ► The unadjusted ATE = mean(Y|A=1 Y|A=0) = 0.05
- ▶ We use SuperLearner for prediction in all models.
- ► Library: SL.glm, SL.glm.interaction, SL.glmnet, SL.bayesglm, SL.randomForest, SL.step, SL.mean, SL.loglinear
- 5-fold cross-validation

Checking for Practical Positivity Violations

- We tabulated exposure and outcome across all possible levels of our categorical variables
- Observations exist in every possible category of our variable set
- Our for our only continuous variable, BMI z-score (for age and sex), we looked at the distribution of BMI z scores in the two exposure categories

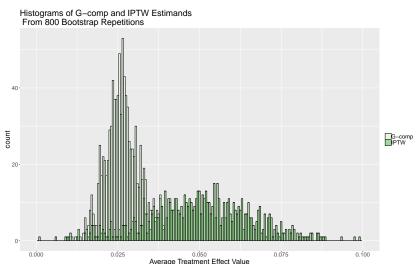


Positivity: Assessing the Model Weights



Confidence Intervals

- For TMLE, used the robust method built in to the ltmle package
- For G-comp and IPTW we performed a bootstrap



Estimation: G-comp, IPTW, & TMLE

Estimator	ATE (95% CI)	
G-computation	0.039 (0.017, 0.034)	
Stabilized IPTW	0.045 (0.017, 0.081)	
TMLE	0.044 (0.007, 0.08)	

Estimation: SuperLearner convex combinations

Algorithm	A Risk	A Coefficient	Y Risk	Y Coefficient
glm	0.1549655	0	0.1408945	0.4628552
glm.interaction	0.155051	0.2086733	0.141501	0
glmnet	0.1549798	0	0.1409052	0
bayesglm	0.1549653	0	0.1408937	0
random Forest	0.1903391	0.4607251	0.1707125	0.2239811
step	0.1549472	0.267622	0.1408951	0.2476689
mean	0.1566939	0.0629796	0.1428772	0.0654947
loglinear	0.1549381	0	0.1409062	0

Estimation: SuperLearner performance

CV. Super Learner

Algorithm	Avg Risk	SE
SuperLearner	0.1412062	0.0028693
Discrete SL	0.1407186	0.0027614
glm	0.1407131	0.0027616
glm.interaction	0.141278	0.0027675
glmnet	0.1407129	0.002763
bayesglm	0.1407127	0.0027616
random Forest	0.1678716	0.0041689
step	0.1407191	0.0027614
mean	0.1428802	0.0028201
loglinear	0.140726	0.0027623
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Conclusions/Interpretation of Results

Contributions of the Team Members

- Suggestion of a dataset and potential issues for exploration: Veronica
- Particular expertise that we each contributed:
 - 1. Shelley Project management
 - 2. Stephanie Pediatrics
 - 3. Veronica Social and Substance Use Epi
 - 4. Lizzy Social and Substance Use Epi
- Establishement of the causal model, delineation of the causal question and estimand of choice: Entire group
- ► Identifiability considerations: Entire group, with Lizzy and Shelley working on the DAG
- Creation of slides for causal question, SCM, background on our dataset: Lizzy and Shelley
- Coding of SuperLearner estimation: Veronica
- Coding of Practical Positivity Checks and Bootstrapping of Confidence Intervals: Stephanie
- Interpretation of Results: Entire Group