Final Project: Childhood Bullying and Subsequent Drug Use

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Causal Question: What is the effect of having

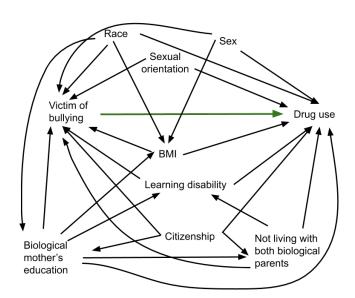
been bullied prior to age 12 on incidence of drug use in adolescence or adulthood?

Specify a Causal Model

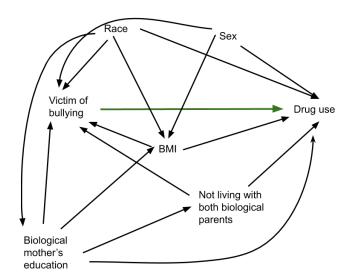
Our dataset:

- National Longitudinal Survey of Youth
- ▶ Nationally representative cohort of youth age 12-16
- ▶ Initial recruitment n = 9000 in 1997
- ► In our final dataset n = 7703

Original DAG



Final DAG



Structural Equations

Our endogenous nodes include: X = (W, A, Y), where $W = (W_1, W_2, W_3, W_4, W_5)$ is the set of baseline covariates, A is victim of bullying, and Y is incident drug use.

Our background variables (exogenous nodes) include: $U = (U_W, U_A, U_Y) \sim \mathbb{P}_U$.

We place no assumptions on the distribution \mathbb{P}_U . We have not placed any restrictions on the functional form.

Structural Equations

Our structural equations \mathcal{F} are:

$$W_{1} = f_{W_{1}}(U_{W_{1}}, W_{3})$$

$$W_{2} = f_{W_{2}}(U_{W_{2}})$$

$$W_{3} = f_{W_{3}}(U_{W_{3}})$$

$$W_{4} = f_{W_{4}}(U_{W_{4}}, W_{1})$$

$$W_{5} = f_{W_{5}}(U_{W_{5}}, W_{1}, W_{2}, W_{3})$$

$$A = f_{A}(U_{A}, W_{1}, W_{2}, W_{3}, W_{4}, W_{5})$$

$$Y = f_{Y}(U_{Y}, A, W_{1}, W_{2}, W_{3}, W_{4}, W_{5})$$

Where W_1 = mother's education; W_2 = sex; W_3 = race/ethnicity; W_4 = not living with both biological parents; W_5 = BMI z-score; A = bullied before the age of 12 (asked in 1997); Y = incident drug use ("cocaine or other hard drugs") after 1997.

Target Causal Parameter

Our target causal parameter is the 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:

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

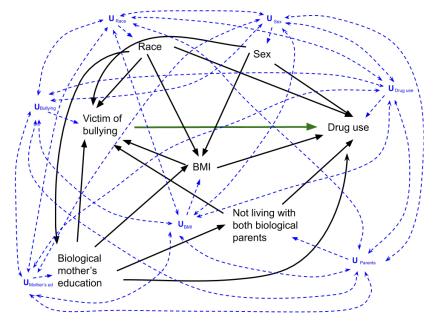
where Y_a denotes the counterfactual outcome under an intervention to set bullying status A=a. This target causal parameter is the average treatment effect (ATE), or causal risk difference.

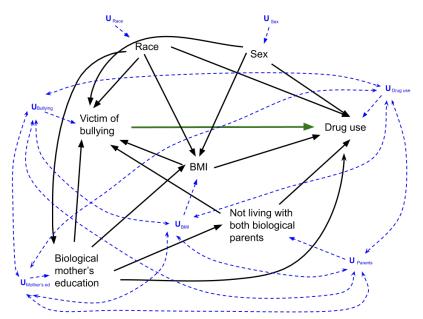
Link to the SCM

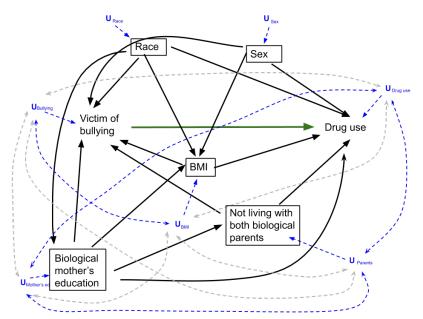
We assume that the observed data $O=(W,A,Y)\sim \mathbb{P}_0$ were generated by sampling n times from a data generating process described by the SCM. The statistical model $\mathcal M$ for the set of allowed distributions for the observed data is non-parametric.

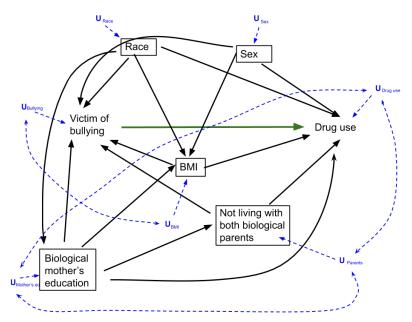
Table 1

Covariate	Drug use (%)	No drug use (%)
Drug use (Total)	1330 (17.3%)	6373 (82.7%)
Victim of bullying Yes No	319 (4.1%) 1011 (13.1%)	
Mother's education High school or less Some college or more	732 (9.5%) 598 (7.8%)	3867 (50.2%) 2506 (32.5%)
Sex Female Male	591 (7.7%) 739 (9.6%)	3218 (41.8%) 3155 (41%)
Race/ethnicity Black Hispanic Non-Black, Non-Hispanic	227 (2.9%) 288 (3.7%) 815 (10.6%)	1788 (23.2%) 1340 (17.4%) 3245 (42.1%)
Living with both biological parents Yes No	645 (8.4%) 685 (8.9%)	3176 (41.2%) 3197 (41.5%)
BMI z-score	0.513 (<i>mean</i>) 1.03 (<i>sd</i>)	0.505 (mean) 0.98 (sd)









Estimand and Statistical Model

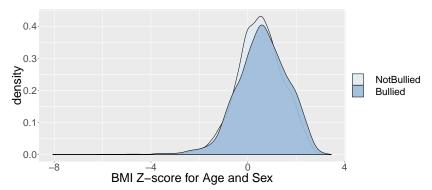
The target parameter of the observed data distribution (which equals the causal parameter in the augmented causal model $\mathcal{M}^{F\star}$) is the G-Computation formula:

$$\begin{split} \psi(\mathbb{P}_0) &= \mathbb{E}_0[\mathbb{E}_0(Y|A=1,W) - \mathbb{E}_0(Y|A=0,W)] = \\ \sum_{^{w1,w2,w3,w4,w5}} \left[\bar{Q}_0(1,W1=w1,W2=w2,W3=w3,W4=w4,W5=w5) - \\ \bar{Q}_0(0,W1=w1,W2=w2,W3=w3,W4=w4,W5=w5) \right] * \\ \mathbb{P}_0(W1=w1,W2=w2,W3=w3,W4=w4,W5=w5) \end{split}$$

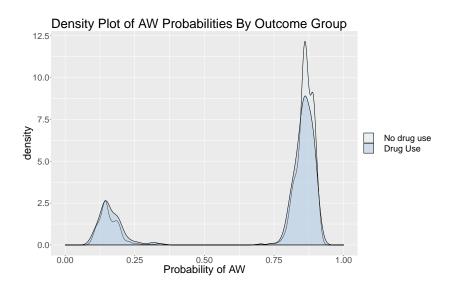
This is our statistical estimand.

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
- ► 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



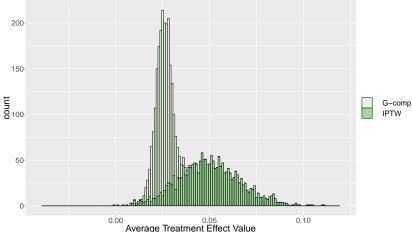
Estimation: Set-up

- We use SuperLearner for prediction in all models.
 - Library: SL.glm, SL.glm.interaction, SL.glmnet, SL.bayesglm, SL.randomForest, SL.step, SL.mean
 - 5-fold cross-validation
- ► We use the following estimators: G-computation (simple substitution estimator), stabilized IPTW, and TMLE

Confidence Intervals

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

Histograms of G-comp and IPTW Estimands From 2000 Bootstrap Repetitions



Estimation: Results

▶ The unadjusted ATE = mean(Y|A=1 - Y|A=0) = 0.05

Estimator	ATE (95% CI)
G-computation Stabilized IPTW TMLE	0.039 (0.017, 0.034) 0.045 (0.018, 0.084) 0.044 (0.007, 0.08)

Estimation: SuperLearner convex combinations

Algorithm	A Risk	A Coefficient	Y Risk	Y Coefficient
glm	0.15497	0	0.14089	0.463
glm.interaction	0.15505	0.209	0.1415	0
glmnet	0.15498	0	0.14091	0
bayesglm	0.15497	0	0.14089	0
randomForest	0.19034	0.461	0.17071	0.224
step	0.15495	0.268	0.1409	0.248
mean	0.15669	0.063	0.14288	0.065

Estimation: SuperLearner performance

CV.SuperLearner

Algorithm	Avg Risk	SE
SuperLearner	0.14121	0.00287
Discrete SL	0.14072	0.00276
glm	0.14071	0.00276
glm.interaction	0.14128	0.00277
glmnet	0.14071	0.00276
bayesglm	0.14071	0.00276
randomForest	0.16787	0.00417
step	0.14072	0.00276
mean	0.14288	0.00282

Interpretation

According to our analysis:

- ➤ **Statistical interpretation**: being bullied before the age of 12 is associated with a 4% absolute increase in drug use compared with not being bullied
- Causal interpretation: the difference between the average counterfactual risk of drug use if everyone was bullied versus if no one was bullied is 0.04

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Stabilized IPTW	0.045 (0.018, 0.084)
TMLE	0.044 (0.007, 0.08)

Limitations

- 1. Important exogenous variables
 - Parent drug use
 - Mental health
 - ► And more...
- 2. Necessary independence assumptions

Impacts

- Identify youth who are at risk for starting to use drugs as a result of bullying
- Supports use of anti-bullying interventions in schools

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
- Establishment 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
- ► Creation of SuperLearner library: Veronica
- ► Coding of ATE point estimates: Veronica
- Coding of Practical Positivity Checks and Bootstrapping of Confidence Intervals: Stephanie
- Interpretation of Results: Entire Group