

Potential Outcomes Simulation Homework

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Objective

In this exercise, you will be tasked with simulating an intervention study with a pre-determined average treatment effect. The goal is for you to understand the **potential outcome framework**, and the properties of **completely randomized experiments** through simulation.

Problem Statement

The goal is to simulate a data set with a treatment effect of $\tau = 5$.

The setting for our hypothetical study is Professor Hill's Causal Inference class. After the first attempt at Quiz I, Professor Hill decides to give students an opportunity to take the quiz again. Before the second attempt of the quiz, Professor Hill randomly assigns half the class to attend an extra tutoring session to half of the class. The other half of the class does not receive any additional help. Consider the half of the class that receives tutors as the treated group. The goal is to estimate the effect of the extra tutoring session on average test scores for the retake of Quiz 1.

We are assuming that SUTVA is satisfied.

Question 1: Generating potential outcomes; Calculating ATE (all seeing/omniscient) For this section, you are a god of Statistics. That is, assume you are omniscient and know the potential outcome of $Y(0)$ and $Y(1)$ for everyone.

- (a) Please simulate a dataset consistent with the assumptions below while demonstrating an average treatment effect (ATE) of approximately 5.

Simulation assumptions The Data Generating Process (DGP) has the following features:

- * Population size N is 1000.
- * The pretest (Causal Quiz I score) is independent and identically distributed with a Normal distribution with mean of 65 and standard deviation of 3.
- * The potential outcomes for the corresponding to Causal Quiz II score should be linearly related to the pretest quiz score. In particular they should take the form:

$$Y(0) = \beta_0 + \beta_1 X + 0 + \epsilon$$

$$Y(1) = \beta_0 + \beta_1 X + \tau + \epsilon$$

where β_0 is the intercept taking the value of **10**. β_1 is set to **1.1**. τ is 5. ϵ should be drawn from a $N(0,1)$ distribution. Please also set the seed at 1234 before generating these draws.

```
# Population size
N = 1000
# Intercept
beta0 = 10
# Slope
```

```

beta1 = 1.1
# Population average treatment effect
tau = 5

# Setting seed for consistent draws
set.seed(1234)
# Drawing epsilon from N(0,1) distribution
epsilon = rnorm(1,0,1)
# Drawing for pretest scores
X = rnorm(N,65,3)

# Initializing all outcomes for no intervention
Y0 = rep(NA,N)
# Simulating no intervention posttest scores from pretest scores
Y0 = beta0 + beta1*X + 0 + epsilon

# Initializing all outcomes for intervention for all
Y1 = rep(NA,N)
# Simulating all intervention posttest scores from pretest scores
Y1 = beta0 + beta1*X + 5 + epsilon

# This results in three vectors of simulated data, X, Y0, Y1

```

- (b) Write a function to generate the data generating process (DGP) for pretest, Y0, and Y1 with arguments for sample size, the coefficient on the pretest, and the random seed. Then use this function to simulate a data set with sample size equal to 100, seed equal to 1234, and the coefficient on the covariate set to 1.1. The probability of being assigned to treatment should be equal to .5.

```

# Start function, with inputs N, beta1, and seed
DGPgod = function(n,beta1,seed){
  # Setting seed for consistent draws
  set.seed(seed)
  # Drawing epsilon from N(0,1) distribution
  epsilon = rnorm(n,0,1)
  # Drawing for pretest scores, reset the seed
  set.seed(seed)
  X = rnorm(n,65,3)
  # Drawing for treatment assignment, reset the seed
  set.seed(seed)
  z = rbinom(n,1,0.5)

  # Initializing all outcomes for no intervention
  Y0 = rep(NA,n)
  # Simulating no intervention posttest scores from pretest scores
  Y0 = beta0 + beta1*X + 0 + epsilon

  # Initializing all outcomes for intervention for all
  Y1 = rep(NA,N)
  # Simulating all intervention posttest scores from pretest scores
  Y1 = beta0 + beta1*X + 5 + epsilon

  df1 = data.frame(X, z, Y0, Y1)

  return(df1)
}

```

```

# Return data frame
}

data1 = DGPGod(100,1.1,1234)
data1

```

##		X	z	Y0	Y1
## 1	61.37880	0	76.30962	81.30962	
## 2	65.83229	1	82.69295	87.69295	
## 3	68.25332	1	86.16310	91.16310	
## 4	57.96291	1	71.41350	76.41350	
## 5	66.28737	1	83.34524	88.34524	
## 6	66.51817	1	83.67604	88.67604	
## 7	63.27578	0	79.02862	84.02862	
## 8	63.36010	0	79.14948	84.14948	
## 9	63.30664	1	79.07286	84.07286	
## 10	62.32989	1	77.67284	82.67284	
## 11	63.56842	1	79.44807	84.44807	
## 12	62.00484	1	77.20694	82.20694	
## 13	62.67124	0	78.16211	83.16211	
## 14	65.19338	1	81.77717	86.77717	
## 15	67.87848	0	85.62582	90.62582	
## 16	64.66914	1	81.02577	86.02577	
## 17	63.46697	0	79.30266	84.30266	
## 18	62.26641	0	77.58186	82.58186	
## 19	62.48848	0	77.90016	82.90016	
## 20	72.24751	0	91.88809	96.88809	
## 21	65.40226	0	82.07658	87.07658	
## 22	63.52794	0	79.39005	84.39005	
## 23	63.67836	0	79.60564	84.60564	
## 24	66.37877	0	83.47623	88.47623	
## 25	62.91884	0	78.51700	83.51700	
## 26	60.65539	1	75.27272	80.27272	
## 27	66.72427	1	83.97145	88.97145	
## 28	61.92903	1	77.09828	82.09828	
## 29	64.95459	1	81.43491	86.43491	
## 30	62.19215	0	77.47542	82.47542	
## 31	68.30689	0	86.23988	91.23988	
## 32	63.57322	0	79.45495	84.45495	
## 33	62.87168	0	78.44941	83.44941	
## 34	63.49623	1	79.34459	84.34459	
## 35	60.11272	0	74.49490	79.49490	
## 36	61.49714	1	76.47924	81.47924	
## 37	58.45988	0	72.12583	77.12583	
## 38	60.97702	0	75.73373	80.73373	
## 39	64.11712	1	80.23454	85.23454	
## 40	63.60231	1	79.49664	84.49664	
## 41	69.34849	1	87.73283	92.73283	
## 42	61.79407	1	76.90484	81.90484	
## 43	62.43391	0	77.82193	82.82193	
## 44	64.15813	1	80.29332	85.29332	
## 45	62.01698	0	77.22434	82.22434	
## 46	62.09446	1	77.33539	82.33539	
## 47	61.67805	1	76.73853	81.73853	

```

## 48 61.24404 0 76.11646 81.11646
## 49 63.42852 0 79.24754 84.24754
## 50 63.50945 1 79.36355 84.36355
## 51 59.58191 0 73.73407 78.73407
## 52 63.25377 0 78.99707 83.99707
## 53 61.67333 1 76.73177 81.73177
## 54 61.95511 1 77.13566 82.13566
## 55 64.51307 0 80.80207 85.80207
## 56 66.68917 1 83.92114 88.92114
## 57 69.94345 0 88.58562 93.58562
## 58 62.67994 1 78.17458 83.17458
## 59 69.81773 0 88.40541 93.40541
## 60 61.52657 1 76.52142 81.52142
## 61 66.96977 1 84.32333 89.32333
## 62 72.64697 0 92.46066 97.46066
## 63 64.89572 0 81.35053 86.35053
## 64 62.99110 0 78.62058 83.62058
## 65 64.97719 0 81.46730 86.46730
## 66 70.33125 1 89.14146 94.14146
## 67 61.58418 0 76.60399 81.60399
## 68 69.10348 1 87.38166 92.38166
## 69 68.98869 0 87.21713 92.21713
## 70 66.00942 1 82.94683 87.94683
## 71 65.02068 0 81.52964 86.52964
## 72 63.63359 1 79.54148 84.54148
## 73 63.90043 0 79.92395 84.92395
## 74 66.94486 1 84.28763 89.28763
## 75 71.21081 0 90.40216 95.40216
## 76 64.53980 1 80.84039 85.84039
## 77 60.82790 0 75.51999 80.51999
## 78 62.82925 0 78.38860 83.38860
## 79 65.77479 0 82.61053 87.61053
## 80 64.04882 1 80.13665 85.13665
## 81 64.46663 1 80.73550 85.73550
## 82 64.49002 0 80.76903 85.76903
## 83 60.88309 0 75.59910 80.59910
## 84 64.47864 1 80.75272 85.75272
## 85 67.55070 0 85.15600 90.15600
## 86 67.09283 1 84.49972 89.49972
## 87 66.64999 0 83.86499 88.86499
## 88 63.79180 0 79.76825 84.76825
## 89 64.42522 0 80.67615 85.67615
## 90 61.41642 1 76.36353 81.36353
## 91 64.84052 0 81.27142 86.27142
## 92 65.76559 1 82.59734 87.59734
## 93 70.11789 0 88.83565 93.83565
## 94 68.00454 0 85.80651 90.80651
## 95 63.51325 0 79.36899 84.36899
## 96 66.06665 1 83.02887 88.02887
## 97 61.59618 0 76.62119 81.62119
## 98 67.63461 0 85.27628 90.27628
## 99 67.91875 0 85.68354 90.68354
## 100 71.36335 1 90.62080 95.62080

```

```
# Use the function
```

Answer the following questions based on the DGP or using your simulated data set. Remember that you are still all-seeing.

- (a) What is your interpretation of τ ?

In a nonsensical world, given omnipotence, if a student receives tutoring, their post test score can be expected to be 5 points higher than if they were not to receive tutoring. Given the assumption of omnipotence, this τ is the population average treatment effect.

- (b) How would you interpret the intercept in the DGP for $Y(0)$ and $Y(1)$?

In a nonsensical world, given omnipotence, an intercept in the DGP of 10 means we can expect a student with pretest score of 0 to score $Y(0) = 10$ and $Y(1) = 15$.

- (c) Consider: How would you interpret the β_1 coefficient?

In a nonsensical world, given omnipotence, for any change of 1 in student pretest score, we can expect the post test score to increase by β_1 times the pretest score, holding all other variables constant

Question 2: Calculating ATE (all seeing/omniscient) Answer this question using the simulated dataset from above.

- (a) The Sample Average Treatment Effect (SATE) is the average of individual treatment effects in the sample. Calculate it for your sample.

```
# Calculating SATE with difference of means
SATEGod = (mean(subset(data1,data1[,2]==1)[,4])-mean(subset(data1,data1[,2]==0)[,3]))
SATEGod

## [1] 4.907522
```

Question 3: Estimating SATE (not all seeing/researchers'view) For Questions 3 and 4, you are a **mere** researcher! Return your god-vision goggles and use only the data available to the researcher (that is, you will not have access to the counterfactual outcomes for each student).

- (a) Using the same simulated dataset used in the previous case where $\tau = 5$, please randomly assign students to treatment and control groups (remember, this is something a research would do in practice). The probability of being assigned to treatment should be equal to .5. One way to do this is by using the following command to generate treatment assignment:

Note that an alternative method is the following. . . think about what difference this might make in practice. . . Next, create the observed data set which must include pretest scores, treatment assignment and observed Y .

```
# Start function, with inputs N, beta1, and seed
DGP = function(n,beta1,seed){
  # Setting seed for consistent draws
  set.seed(seed)
  # Drawing epsilon from N(0,1) distribution
  epsilon = rnorm(n,0,1)
  # Drawing for pretest scores, reset the seed
  set.seed(seed)
  X = rnorm(n,65,3)
  # Drawing for treatment assignment, reset the seed
  set.seed(seed)
  z = rbinom(n,1,0.5)
  # Make intermediate data frame to subset in the next step
```

```

Y = beta0 + beta1 * X + 5*z + epsilon

df2 = data.frame(X, z, Y)

return(df2)
# Return data frame
}

data2 = DGP(1000,1.1,1234)
data2

```

```

##           X z           Y
## 1    61.37880 0 76.30962
## 2    65.83229 1 87.69295
## 3    68.25332 1 91.16310
## 4    57.96291 1 76.41350
## 5    66.28737 1 88.34524
## 6    66.51817 1 88.67604
## 7    63.27578 0 79.02862
## 8    63.36010 0 79.14948
## 9    63.30664 1 84.07286
## 10   62.32989 1 82.67284
## 11   63.56842 1 84.44807
## 12   62.00484 1 82.20694
## 13   62.67124 0 78.16211
## 14   65.19338 1 86.77717
## 15   67.87848 0 85.62582
## 16   64.66914 1 86.02577
## 17   63.46697 0 79.30266
## 18   62.26641 0 77.58186
## 19   62.48848 0 77.90016
## 20   72.24751 0 91.88809
## 21   65.40226 0 82.07658
## 22   63.52794 0 79.39005
## 23   63.67836 0 79.60564
## 24   66.37877 0 83.47623
## 25   62.91884 0 78.51700
## 26   60.65539 1 80.27272
## 27   66.72427 1 88.97145
## 28   61.92903 1 82.09828
## 29   64.95459 1 86.43491
## 30   62.19215 0 77.47542
## 31   68.30689 0 86.23988
## 32   63.57322 0 79.45495
## 33   62.87168 0 78.44941
## 34   63.49623 1 84.34459
## 35   60.11272 0 74.49490
## 36   61.49714 1 81.47924
## 37   58.45988 0 72.12583
## 38   60.97702 0 75.73373
## 39   64.11712 1 85.23454
## 40   63.60231 1 84.49664
## 41   69.34849 1 92.73283

```

42 61.79407 1 81.90484
43 62.43391 0 77.82193
44 64.15813 1 85.29332
45 62.01698 0 77.22434
46 62.09446 1 82.33539
47 61.67805 1 81.73853
48 61.24404 0 76.11646
49 63.42852 0 79.24754
50 63.50945 1 84.36355
51 59.58191 0 73.73407
52 63.25377 0 78.99707
53 61.67333 1 81.73177
54 61.95511 1 82.13566
55 64.51307 0 80.80207
56 66.68917 1 88.92114
57 69.94345 0 88.58562
58 62.67994 1 83.17458
59 69.81773 0 88.40541
60 61.52657 1 81.52142
61 66.96977 1 89.32333
62 72.64697 0 92.46066
63 64.89572 0 81.35053
64 62.99110 0 78.62058
65 64.97719 0 81.46730
66 70.33125 1 94.14146
67 61.58418 0 76.60399
68 69.10348 1 92.38166
69 68.98869 0 87.21713
70 66.00942 1 87.94683
71 65.02068 0 81.52964
72 63.63359 1 84.54148
73 63.90043 0 79.92395
74 66.94486 1 89.28763
75 71.21081 0 90.40216
76 64.53980 1 85.84039
77 60.82790 0 75.51999
78 62.82925 0 78.38860
79 65.77479 0 82.61053
80 64.04882 1 85.13665
81 64.46663 1 85.73550
82 64.49002 0 80.76903
83 60.88309 0 75.59910
84 64.47864 1 85.75272
85 67.55070 0 85.15600
86 67.09283 1 89.49972
87 66.64999 0 83.86499
88 63.79180 0 79.76825
89 64.42522 0 80.67615
90 61.41642 1 81.36353
91 64.84052 0 81.27142
92 65.76559 1 87.59734
93 70.11789 0 88.83565
94 68.00454 0 85.80651
95 63.51325 0 79.36899

96 66.06665 1 88.02887
97 61.59618 0 76.62119
98 67.63461 0 85.27628
99 67.91875 0 85.68354
100 71.36335 1 95.62080
101 66.24357 0 83.28245
102 63.57584 1 84.45871
103 65.19798 0 81.78377
104 63.49257 0 79.33935
105 62.52200 0 77.94821
106 65.50097 0 82.21805
107 62.31121 0 77.64606
108 65.50456 0 82.22320
109 66.06490 0 83.02636
110 64.84368 0 81.27595
111 64.41220 1 85.65748
112 63.05279 0 78.70900
113 61.67070 1 81.72800
114 67.54782 0 85.15188
115 65.06709 0 81.59616
116 67.49342 1 90.07390
117 61.26714 1 81.14956
118 65.50708 0 82.22681
119 67.01950 0 84.39462
120 64.92117 1 86.38701
121 64.42582 1 85.67701
122 62.65428 1 83.13780
123 71.17449 1 95.35010
124 67.25150 1 89.72716
125 70.47262 0 89.34410
126 65.24018 0 81.84426
127 63.10577 0 78.78494
128 60.46014 1 79.99286
129 63.09170 0 78.76477
130 65.67890 0 82.47310
131 68.04107 1 90.85887
132 65.75825 1 87.58683
133 61.48416 0 76.46062
134 67.00614 0 84.37547
135 60.04970 1 79.40457
136 63.90244 0 79.92684
137 64.05165 1 85.14069
138 59.15526 1 78.12254
139 67.76017 1 90.45625
140 63.13139 1 83.82165
141 63.99789 1 85.06364
142 69.18544 1 92.49914
143 66.91002 0 84.23770
144 64.67470 0 81.03374
145 66.54129 0 83.70918
146 66.19782 1 88.21687
147 69.98857 1 93.65028
148 65.82768 1 87.68634
149 66.51882 1 88.67697

150 66.04266 1 87.99447
151 63.86829 0 79.87788
152 65.29286 0 81.91976
153 69.91623 0 88.54660
154 62.37322 1 82.73495
155 65.36528 0 82.02357
156 69.08639 1 92.35716
157 64.29614 1 85.49113
158 61.83985 1 81.97045
159 62.39065 0 77.75993
160 63.82962 1 84.82245
161 62.45795 0 77.85639
162 64.21808 1 85.37925
163 63.75674 0 79.71800
164 64.45085 0 80.71288
165 66.22117 0 83.25034
166 66.87390 1 89.18592
167 70.03462 0 88.71628
168 64.79392 0 81.20462
169 64.03748 1 85.12039
170 69.41302 0 87.82532
171 70.11299 1 93.82862
172 65.12973 1 86.68595
173 64.00203 1 85.06957
174 59.53329 0 73.66439
175 69.23379 0 87.56843
176 62.48725 1 82.89840
177 61.62871 0 76.66782
178 74.13130 1 99.58819
179 65.70506 0 82.51059
180 64.90022 0 81.35699
181 56.80334 0 69.75146
182 64.70063 0 81.07090
183 67.92810 1 90.69694
184 66.24161 0 83.27964
185 67.73697 1 90.42299
186 70.95120 0 90.03005
187 68.50733 1 91.52717
188 63.47379 1 84.31243
189 67.11254 0 84.52797
190 64.40475 1 85.64681
191 63.38579 1 84.18630
192 56.43272 1 74.22024
193 62.63106 0 78.10452
194 66.46344 1 88.59760
195 71.50410 1 95.82254
196 66.50208 1 88.65299
197 66.86063 1 89.16690
198 62.10229 1 82.34662
199 65.48796 1 87.19942
200 58.76529 1 77.56358
201 66.45568 1 88.58648
202 67.09031 1 89.49611
203 65.55654 0 82.29771

204 67.10220 1 89.51315
205 65.93504 1 87.84023
206 67.28139 1 89.76999
207 70.52739 0 89.42259
208 68.33709 0 86.28316
209 65.09799 0 81.64046
210 61.65665 1 81.70787
211 66.25417 1 88.29765
212 63.79929 0 79.77899
213 69.48048 0 87.92202
214 60.17876 1 79.58955
215 63.75274 1 84.71227
216 66.26603 1 88.31464
217 64.54479 1 85.84753
218 63.18155 1 83.89355
219 64.08584 0 80.18970
220 66.88861 1 89.20701
221 67.68552 0 85.34924
222 66.98064 0 84.33891
223 71.82045 0 91.27598
224 68.52049 0 86.54604
225 65.86313 0 82.73715
226 63.02069 0 78.66299
227 73.75742 1 99.05230
228 67.03225 0 84.41289
229 62.94704 1 83.55742
230 65.55948 0 82.30192
231 64.02682 1 85.10511
232 64.17589 1 85.31877
233 62.19949 0 77.48594
234 65.35054 1 87.00243
235 65.95748 1 87.87239
236 61.76737 0 76.86657
237 55.30054 1 72.59745
238 64.23538 0 80.40404
239 65.08855 0 81.62693
240 66.78282 1 89.05538
241 65.17741 0 81.75428
242 66.24020 0 83.27762
243 61.70668 0 76.77958
244 67.13353 1 89.55805
245 67.15667 1 89.59122
246 65.75495 0 82.58210
247 69.07182 1 92.33628
248 66.21341 1 88.23921
249 65.79309 1 87.63677
250 65.80413 1 87.65259
251 66.31079 1 88.37880
252 68.18037 1 91.05853
253 66.35657 0 83.44442
254 66.98960 0 84.35175
255 61.59088 0 76.61359
256 63.88851 1 84.90686
257 69.43091 0 87.85097

258 61.32829 0 76.23721
259 65.77421 1 87.60969
260 66.21501 1 88.24151
261 67.92741 1 90.69595
262 63.95337 0 79.99983
263 65.47588 1 87.18209
264 59.71023 0 73.91800
265 66.01579 0 82.95596
266 63.00030 0 78.63377
267 64.28406 1 85.47382
268 61.43670 0 76.39261
269 66.15481 0 83.15522
270 66.99974 1 89.36629
271 64.08616 0 80.19016
272 70.47503 1 94.34755
273 67.01168 0 84.38341
274 67.84590 1 90.57912
275 71.14821 0 90.31243
276 63.04666 1 83.70021
277 67.42586 1 89.97706
278 67.95974 1 90.74230
279 64.98149 0 81.47347
280 65.95716 0 82.87193
281 61.96453 0 77.14917
282 66.41050 0 83.52172
283 62.89709 1 83.48583
284 67.44105 1 89.99884
285 62.56571 1 83.01085
286 65.95819 0 82.87341
287 62.46043 0 77.85995
288 64.26271 0 80.44322
289 60.34142 1 79.82271
290 65.38530 1 87.05227
291 67.95633 1 90.73741
292 65.54974 0 82.28796
293 59.70131 1 78.90521
294 63.13840 0 78.83171
295 69.96813 1 93.62099
296 70.42942 1 94.28216
297 61.47489 1 81.44734
298 63.89989 0 79.92318
299 66.06088 1 88.02059
300 65.95747 0 82.87237
301 63.26013 0 79.00618
302 62.14016 1 82.40090
303 64.46171 1 85.72846
304 68.02942 0 85.84218
305 65.07088 1 86.60159
306 63.05292 0 78.70918
307 63.48688 0 79.33119
308 69.84317 1 93.44188
309 63.65912 1 84.57807
310 67.28953 1 89.78166
311 69.41516 1 92.82839

312 66.33099 1 88.40776
313 63.73483 0 79.68660
314 64.88000 0 81.32799
315 63.52316 0 79.38320
316 68.68315 0 86.77918
317 64.55134 0 80.85692
318 69.64995 0 88.16493
319 63.31516 0 79.08507
320 63.05865 1 83.71740
321 65.42940 0 82.11547
322 65.07257 1 86.60401
323 63.48665 0 79.33086
324 60.25581 0 74.69999
325 65.09020 0 81.62929
326 62.85027 0 78.41872
327 68.24783 0 86.15523
328 62.14194 0 77.40345
329 68.37945 1 91.34388
330 63.05287 1 83.70912
331 65.87741 1 87.75762
332 67.69611 1 90.36442
333 63.44377 1 84.26941
334 66.66332 0 83.88409
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## 990 64.50244 0 80.78683
## 991 65.76501 0 82.59652
## 992 62.30758 1 82.64087
## 993 69.75191 1 93.31107
## 994 69.17828 0 87.48886
## 995 59.56323 1 78.70729
## 996 64.35104 0 80.56983
## 997 66.64957 1 88.86438
## 998 66.44821 0 83.57577
## 999 67.28255 1 89.77165
## 1000 63.66303 0 79.58367
```

```
# Use the function
```

(b) Estimate SATE using a difference in mean outcomes.

```
# Calculating SATE with difference of means
```

```
SATE = (mean(subset(data2,data2[,2]==1)[,3])-mean(subset(data2,data2[,2]==0)[,3]))
SATE
```

```
## [1] 5.165079
```

(c) Is this estimate close to the true SATE? Divide the difference between SATE and estimated SATE by the standard deviation of the observed outcome, Y to express this conditional bias in standard deviation units. This helps you understand the practical significance of this difference.


```
condbias = (SATE-SATEGod)/sd(data2[,3])
condbias
```

```
## [1] 0.05145739
```

(d) Consider: Why is \hat{SATE} different from SATE and τ ?

Since we are no longer god, we cannot control the even spread of treatment assignments nor see the counterfactuals to get a perfectly true SATE.

Question 4: Use Linear Regression to estimate the treatment effect

- (a) Now we will use linear regression to estimate SATE for the observed data set created by Question 2. With this set up, we will begin to better understand some fundamental assumptions crucial for the later R homework assignments.

```
reg1 = lm(Y~X+z,data2)
summary(reg1)
```

```
##
## Call:
## lm(formula = Y ~ X + z, data = data2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.662e-14 -8.230e-15 -3.270e-15  2.340e-15  2.921e-12
##
## Coefficients:
##              Estimate Std. Error    t value Pr(>|t|)
## (Intercept) -1.167e+01  6.385e-14 -1.827e+14  <2e-16 ***
## X              1.433e+00  9.822e-16  1.459e+15  <2e-16 ***
## z              5.000e+00  5.879e-15  8.505e+14  <2e-16 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 9.287e-14 on 997 degrees of freedom
## Multiple R-squared:  1, Adjusted R-squared:  1
## F-statistic: 1.451e+30 on 2 and 997 DF, p-value: < 2.2e-16
```

- (b) Consider: What is gained by using linear regression to estimate ATE instead of the mean difference estimation from above?

The standard error for estimate ATE is smaller.

Challenge Question: Treatment Effect Heterogeneity

- (a) Based on the following function: Simulate the following “response surfaces” (relationship between the mean of each potential outcome and the covariate(s)), $E[Y(0) | X]$ and $E[Y(1) | X]$. Plot them on the same plot (that is make a plot with X on the x-axis and $Y(0)/Y(1)$ on the y-axis. Also simulate $Y(0)$ and $Y(1)$ (that is, the expected values plus “noise”).

Note: X is the same pretest score used before.

$$\begin{aligned}
E[Y(0) | X] &= \beta_0^0 + \beta_1^0 X \\
Y(0) &= E[Y(0) | X] + \epsilon^0 \\
Y(0) &= \beta_0^0 + \beta_1^0 X + \epsilon^0 \\
E[Y(1) | X] &= \beta_0^1 + \beta_1^1 X \\
Y(1) &= E[Y(1) | X] + \epsilon^1 \\
Y(1) &= \beta_0^1 + \beta_1^1 X + \epsilon^1
\end{aligned}$$

where β_0^0 is set to **35**, β_1^0 is set to .6, β_0^1 is set to **15**, β_1^1 is set to 1. First generate a vector of predicted $Y(0)$ and $Y(1)$ (that is $E[Y(1) | X]$). Then generate $Y(0)$ and $Y(1)$ with noise added as ϵ^0 or ϵ^1 from a distribution of $N(0,1)$. Again, please also set seed at 1234.

- (b) Comment on your findings. In particular, note that there is no longer a tau included in the DGP. Is there still a SATE? Can we calculate SATE? (Remember I have to be omniscient to do this!) What is it? Consider: How do we interpret the average treatment effect in this setting?
- (c) Is the treatment effect the same for all students? If not, is there a pattern to the way it varies? Consider: Why do we care about treatment effect heterogeneity?
- (d) Now generate a similar plot from the initial DGP in Question 1 to reinforce the differences between a setting with constant treatment effect and a setting with heterogeneous treatment effects.