

# Assignment 9b

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## Answer 1

```
df <- data.frame(
  i = c("Audrey", "Anna", "Bob", "Bill", "Caitlin", "Cara", "Dave", "Doug"),
  x = c(40, 45, 50, 55, 60, 65, 70, 75),
  z = c(0, 1, 1, 0, 0, 0, 1, 1),
  y0 = c(140, 140, 150, 150, 160, 160, 170, 170),
  y1 = c(135, 135, 140, 140, 155, 155, 160, 160)
)
```

(a) `audrey <- df$i == "Audrey"`  
`df$y1[audrey] - df$y0[audrey]`

```
## [1] -5
```

Thus, the treatment effect for Audrey is  $-5$ .

(b) `df$tau <- df$y1 - df$y0`  
`mean(df$tau)`

```
## [1] -7.5
```

Thus, the sample average treatment effect is  $-7.5$ .

(c) `df$y <- ifelse(df$z == 0, df$y0, df$y1)`  
`lm(y ~ z, data = df)`

```
##
## Call:
## lm(formula = y ~ z, data = df)
##
## Coefficients:
## (Intercept)          z
##      152.50         -3.75
```

Thus, the estimated treatment effect is  $-3.75$ .

For the sake of curiosity, consider

```
lm(y ~ z + x, data = df)
```

```
##
## Call:
## lm(formula = y ~ z + x, data = df)
##
```

```
## Coefficients:
## (Intercept)          z          x
##      104.375      -8.125       0.875
```

## Answer 2

It would seem that (b) is true; the prediction  $X\beta$  has a smoothing effect leading to the absence of more extreme values for income. In particular, the imputed data will be ‘missing’ extreme values produced by the error terms.