Question 3

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December 3, 2020

Part A

```
fit1 = lm(iq ~ miles + first2y + miles * first2y, data = lead) fit1$coefficients  
## (Intercept) miles first2y miles:first2y  
## 98.4513331 0.5913399 -19.6116047 17.6546119  
Y_{iq} = \beta_{intercept} + \beta_{miles}X_1 + \beta_{first2y}X_2 + \beta_{interaction}X_1X_2 
Y = 98.4513331 + 0.5913399X1 + -19.6116047X2 + 17.6546119X1X2
```

The average IQ score for those who live 0 miles from the nearest smelter, and were not exposed during their first two years of life is 98.4513331. The difference in IQ score at 0 miles from the nearest smelter between those who were not exposed during their first two years of life, and those that were exposed during their first two years of life is 0.5913399. The slope for those who were not exposed during their first two years of life is -19.6116047. And the difference in slope between those exposed in their first two years, and those not exposed in their first two years is 17.6546119.

Part B

```
summary(fit1)
##
## Call:
## lm(formula = iq ~ miles + first2y + miles * first2y, data = lead)
##
## Residuals:
## Min 1Q Median 3Q Max
```

```
## -48.549 -9.405
                    0.154
                            9.926
                                   45.861
##
## Coefficients:
##
                Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 98.4513
                             4.3862
                                     22.446
                                             < 2e-16 ***
## miles
                                             0.77922
                  0.5913
                             2.1047
                                      0.281
## first2v
                -19.6116
                             7.8583
                                     -2.496
                                             0.01393 *
                                      3.221 0.00164 **
## miles:first2y 17.6546
                             5.4811
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 15.37 on 120 degrees of freedom
```

Multiple R-squared: 0.09873, Adjusted R-squared: 0.0762 ## F-statistic: 4.382 on 3 and 120 DF, p-value: 0.005783

$$H_0: \beta_{interaction} = 0$$

$$H_A: \beta_{interaction} \neq 0$$

There is evidence to suggest that the difference of of slopes between those who were exposed in their first two years, and those who were not exposed in their first two years is not 0 (p = 0.0016442, [6.8024443, 28.5067795]).

Part C

$$Y_{iq} = \beta_{intercept} + \beta_{miles} X_1$$
$$Y = 98.4513331 + 0.5913399X1$$

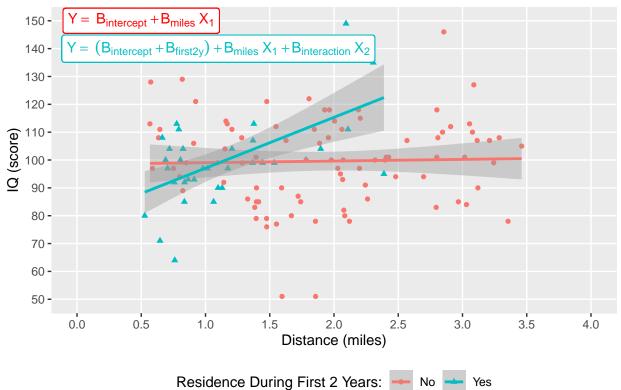
Part D

$$Y_{iq} = (\beta_{intercept} + \beta_{first2y}) + \beta_{miles} X_1 + \beta_{interaction} X_1$$

$$Y = 78.8397284 + 0.5913399X1 + 17.6546119X1$$

Part E

IQ Score by Distance from Smelter



Part F

$$H_0: \beta_{miles} = 0$$

 $H_A: \beta_{miles} \neq 0$

There is not enough evidence to suggest that the slope of those who were not exposed in their first two years is not $0 \text{ (p} = 0.7792212, \text{ ci} = [-3.5757711, 4.758451]).}$

Part G

```
vcov = vcov(fit1)
se = sqrt(vcov[2, 2] + vcov[4, 4] + 2 * vcov[2, 4])
alpha = 0.05

ci = c(
   fit1$coefficients[2] + fit1$coefficients[2] - qnorm(1 - (alpha / 2)) * se,
   fit1$coefficients[2] + fit1$coefficients[2] + qnorm(1 - (alpha / 2)) * se
)

t = (fit1$coefficients[2] + fit1$coefficients[2]) / se
p = 2 * pt(t, dim(lead)[1] - 1, lower.tail = FALSE)
```

```
H_0: \beta_{miles} + \beta_{interaction} = 0

H_A: \beta_{miles} + \beta_{interaction} \neq 0
```

There is not enough evidence to suggest that the slope of those who were exposed in their first two years is not $0 \text{ (p} = 0.8156142, ci = [-8.7364881, 11.1018479])}$.

Part H

Those who were not exposed during their first two years of life are accounting for a 0.5913399 increase in IQ score for each unit increase in miles from the smelter (95% Confidence Interval: [-3.5757711, 4.758451]). Those who were exposed in their first two years of life are accounting for a 1.1826799 increase in IQ score (95% Confidence Interval: [-8.7364881, 11.1018479]). Neither of which is a significant result.

Part I

```
fit_empty = lm(iq ~ miles, lead)
full = anova(fit1)
restricted = anova(fit_empty)

rss_restricted = restricted$`Sum Sq`[2]
rss_full = full$`Sum Sq`[4]
k_full = ( sum(full$Df)-full$Df[4] )
k_restricted = ( sum(restricted$Df) - restricted$Df[2] )
n = nrow(lead)

f = ( (rss_restricted - rss_full) / (k_full - k_restricted) ) / ( rss_full / (n - k_full) )
```

```
p = pf(f, k_full - k_restricted, n - k_full, lower.tail = FALSE)
q = qf(1 - alpha, k_full - k_restricted, n - k_full, lower.tail = TRUE)
```

Null Hypothesis: The full model with miles and first two years and the interaction between, does not account for the variability in IQ score

Alternative Hypothesis: The full model with miles and first two years and the interaction between, accounts for a significantly more amount of variability in IQ score.

$$F = \frac{(\frac{RSS_{restricted} - RSS_{full}}{k_{full} - k_{restricted}})}{(\frac{RSS_{full}}{n - k_{full}})}$$

F-statistic: 5.6806786 Critical value: 3.0711405 p-value: 0.0043849

We reject the null hypothesis that the full model is not better (F(2, 121) = 5.6806786, p(F > 3.0711405) = 0.0043849), in favor of using the full model. There is evidence to suggest that the full model is accounting for significantly more variability than the restricted model.