Math 530/630 Final Replication Project

Josh Burkhart, Steve Chamberlin, Kristen Stevens December 2, 2015

Association of Arsenic Exposure with Lung Cancer Incidence Rates in the United States

Citation: Putila JJ, Guo NL (2001) Association of Arsenic Exposure with Lung Cancer Incidence Rates in the United States. PLOS ONE 6(10): e25886.

```
full <- read.csv("./File_S4.csv", row.names = 1) # change directory for FileS4.csv
```

The authors' data file (FileS4.csv) includes 757 observations of 20 variables. We have named this data.frame "full". The unit of analysis is County. The variables of interest are mean aresenic level in parts per million weighted by county population (Ascounty), median income of county (MedIncome), population (Population), county smoking prevalence calculated as a percent of respondents age 18 or older who reported having smoked more than 100 cigarettes in their lifetime (smkrate), and age-adjusted lung cancer incidence rates (AdjRate).

We first generated log transformed and centered variables from the variables described above.

```
full$lnAs <- log(full$Ascounty) - mean(na.omit(log(full$Ascounty)))
full$lnInc <- log(full$MedIncome) - mean(na.omit(log(full$MedIncome)))
full$Population <- as.numeric(as.character(full$Population))
full$lnsmk <- full$smkrate
full$lnar <- log(full$AdjRate)</pre>
```

Regression Analysis

"The first analysis sought to determine the influence of exposure levels of arsenic on lung cancer incidence in the U.S., and persistance of these effects controlling for possible confounders. The association between each contaminant and lung cancer incidence was assessed using Poisson regression in order to reflect the annual incidence rate as a counting measure."

For comparison, we performed a linear regression in addition to a Poisson regression for each analysis. The following models predict lung cancer incidence weighted by county population using the untransformed data.

```
# Poisson regression
glm1 <- glm(full$AdjRate ~ full$Ascounty, family = poisson, weights = as.numeric(full$Population))</pre>
summary(glm1)
Call:
glm(formula = full$AdjRate ~ full$Ascounty, family = poisson,
    weights = as.numeric(full$Population))
Deviance Residuals:
                  Median
   Min
              1Q
                                3Q
                                        Max
-6257.0 -103.2
                  161.1
                             444.0
                                     2459.2
```

```
Coefficients:
                Estimate Std. Error z value
(Intercept) 4.167391730 0.000023700 175836 <0.00000000000000000 ***
full$Ascounty 0.004479178 0.000001937
                                       2312 <0.0000000000000000 ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for poisson family taken to be 1)
   Null deviance: 317986702 on 741 degrees of freedom
Residual deviance: 312986732 on 740 degrees of freedom
  (15 observations deleted due to missingness)
AIC: 835273995
Number of Fisher Scoring iterations: 4
# odds ratio with CI
glm1_odds <- exp(cbind(OR = coef(glm1), confint(glm1)))</pre>
Waiting for profiling to be done...
# Linear regression
lm1 <- lm(full$lnar ~ full$Ascounty, weights = as.numeric(full$Population))</pre>
summary(lm1)
Call:
lm(formula = full$lnar ~ full$Ascounty, weights = as.numeric(full$Population))
Weighted Residuals:
   Min
            1Q Median
                            30
-738.43 -8.18 23.65 55.32 293.99
Coefficients:
             Estimate Std. Error t value
                                                    Pr(>|t|)
(Intercept) 4.14686 0.01727 240.168 <0.0000000000000000 ***
full$Ascounty 0.00373
                         0.00147
                                   2.537
                                                      0.0114 *
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 83.81 on 740 degrees of freedom
  (15 observations deleted due to missingness)
Multiple R-squared: 0.008621, Adjusted R-squared: 0.007282
F-statistic: 6.435 on 1 and 740 DF, p-value: 0.01139
# odds ratio with CI
lm1_odds <- exp(cbind(OR = coef(lm1), confint(lm1)))</pre>
# Poisson regression
SESassmk <- glm(full$AdjRate ~ full$smkrate + full$Ascounty + full$MedIncome,
   family = poisson, weights = as.numeric(full$Population))
```

summary(SESassmk)

```
Call:
glm(formula = full$AdjRate ~ full$smkrate + full$Ascounty + full$MedIncome,
   family = poisson, weights = as.numeric(full$Population))
Deviance Residuals:
   Min 10 Median
                                    Max
                             30
-3559.7 -240.9 46.1
                                  1949.6
                          347.7
Coefficients:
                    Estimate
                                 Std. Error z value
                                                             Pr(>|z|)
              3.520651941992 0.000121330940 29017 < 0.0000000000000002
(Intercept)
full$smkrate
             1.801926275227 0.000191317650 9419 < 0.0000000000000000
full$Ascounty 0.003931137793 0.000001929776 2037 <0.0000000000000002
(Intercept)
             ***
full$smkrate
full$Ascounty ***
full$MedIncome ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for poisson family taken to be 1)
   Null deviance: 305018305 on 584 degrees of freedom
Residual deviance: 180096592 on 581 degrees of freedom
  (172 observations deleted due to missingness)
AIC: 687702879
Number of Fisher Scoring iterations: 4
SESassmk_odds <- exp(cbind(OR = coef(SESassmk), confint(SESassmk)))
Waiting for profiling to be done...
# Linear regressions
SESassmklm <- lm(full$lnar ~ full$smkrate + full$Ascounty + full$MedIncome,
   weights = as.numeric(full$Population))
summary(SESassmklm)
Call:
lm(formula = full$lnar ~ full$smkrate + full$Ascounty + full$MedIncome,
   weights = as.numeric(full$Population))
Weighted Residuals:
   Min
           1Q Median
                          3Q
                                 Max
-442.09 -28.41 7.35 42.54 234.40
Coefficients:
                  Estimate
                             Std. Error t value
                                                          Pr(>|t|)
             3.2915838146 0.0692881380 47.506 < 0.0000000000000002
(Intercept)
```

```
full$smkrate
               2.1732197826  0.1129146604  19.247 < 0.0000000000000002
               0.0034245343 0.0012310863
                                          2.782
                                                             0.005583
full$Ascounty
full$MedIncome -0.0000023964 0.0000007225 -3.317
                                                             0.000968
(Intercept)
full$smkrate
full$Ascounty
full$MedIncome ***
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
Residual standard error: 69.48 on 581 degrees of freedom
  (172 observations deleted due to missingness)
                               Adjusted R-squared: 0.4404
Multiple R-squared: 0.4433,
F-statistic: 154.2 on 3 and 581 DF, p-value: < 0.0000000000000022
SESassmklm odds <- exp(cbind(OR = coef(SESassmklm), confint(SESassmklm)))
```

Table 1: Unadjusted Model

Summary of Poisson regressions of the effect of arsenic concentration (ppm) on county-level lung cancer incidence rates in the U.S. in an unadjusted model. (We broke up Table 1 into two parts: the Unadjusted Model and the Adjusted Model.)

```
Model_and_Variable = c("Arsenic")
Coefficient = c(round(summary(glm1)$coefficients[2], digits = 4))
Std.Error = c("1.9 x 10^-6") # summary(glm1)$coefficients[4]
Odds_Ratio_CI = c(paste(c(round(glm1_odds[2, 1], digits = 3), "(", round(glm1_odds[2, 2], digits = 3), "-", round(glm1_odds[2, 3], digits = 3), ")"), collapse = " "))
P_value = c("P<0.0001")
N = c(742)
df = data.frame(Model_and_Variable, Coefficient, Std.Error, Odds_Ratio_CI, P_value, N)
kable(df)</pre>
```

Model_and_Variable	Coefficient	Std.Error	Odds_Ratio_CI	P_value	N
Arsenic	0.0045	1.9×10^-6	1.004 (1.004 - 1.004)	P<0.0001	742

```
# should we add the P-value and N as variables? remove underscores from # table labels? table in knit pdf needs to be reformatted to fit the page # scientific notation? do we want to add a table for the results from the # linear model?
```

Table 1: Adjusted Model

Summary of Poisson regressions of the effect of arsenic concentration (ppm) on county-level lung cancer incidence rates in the U.S. in a model adjusted for both smoking and median county income.

```
Model_and_Variable = c("Arsenic", "Smoking", "Median Income")
Coefficient = c(round(summary(SESassmk)$coefficients[3], digits = 4), round(summary(SESassmk)$coefficients[4], digits = 4))
Std.Error = c("1.9 x 10^-6", "0.0002", "1.31 x 10^-9")
Odds_Ratio_CI = c(paste(c(round(SESassmk_odds[3], digits = 3), "(", round(SESassmk_odds[7], digits = 3), "-", round(SESassmk_odds[11], digits = 3), ")"), collapse = " "),
    paste(c(round(SESassmk_odds[12], digits = 3), "(", round(SESassmk_odds[6], digits = 3), "-", round(SESassmk_odds[10], digits = 3), ")"), collapse = " "),
    paste(c(round(SESassmk_odds[4], digits = 3), "(", round(SESassmk_odds[8], digits = 3), "-", round(SESassmk_odds[12], digits = 3), ")"), collapse = " "))
P_value = c("P<0.0001")
N = c(585)
df = data_frame(Model_and_Variable, Coefficient, Std.Error, Odds_Ratio_CI, P_value, N)
kable(df)</pre>
```

Model_and_Variable	Coefficient	Std.Error	Odds_Ratio_CI	P_value	N
Arsenic Smoking Median Income	1.8019	1.9 x 10 ⁻⁶ 0.0002 1.31 x 10 ⁻⁹	1.004 (1.004 - 1.004) 6.061 (6.059 - 6.064) 1 (1 - 1)	P<0.0001	585

```
# should we add the P-value and N as variables? remove underscores from # table labels? table in knit pdf needs to be reformatted to fit the page # scientific notation? do we want to add a table for the results from the # linear model?
```

Table 2

Difference in lung cancer incidence attributable to arsenic exposure alone for high and low-exposure areas in the U.S. based on the results of the adjusted Poisson models and the USGS survey quantiles in Figure 1.

```
Compound = c("Arsenic")
Low_ppm = c(1.477)
High_ppm = c(14.525)
B_Estimate = c(0.0039)
Lung_Cancer_Rate_Increase_Pct = c("5.3%")
df = data.frame(Compound, Low_ppm, High_ppm, B_Estimate, Lung_Cancer_Rate_Increase_Pct)
kable(df)
```

Compound	Low_ppm	$High_ppm$	B_Estimate	$Lung_Cancer_Rate_Increase_Pct$
Arsenic	1.477	14.525	0.0039	5.3%

```
# change values in table to variables, need to move this.
```

Figure 2

^{**}Description, explanation of variables, and code clean-up needed here.

```
## Estimate the 25, 50, amd 75% quartile points for each variable for the
## quartiles interaction models
AsCut <- NA
AsCut[1] <- as.numeric(summary(full$lnAs)[2])
AsCut[2] <- as.numeric(summary(full$lnAs)[3])</pre>
AsCut[3] <- as.numeric(summary(full$lnAs)[5])</pre>
SmkCut <- NA
SmkCut[1] <- as.numeric(summary(full$lnsmk)[2])</pre>
SmkCut[2] <- as.numeric(summary(full$lnsmk)[3])</pre>
SmkCut[3] <- as.numeric(summary(full$lnsmk)[5])</pre>
SESCut <- NA
SESCut[1] <- as.numeric(summary(full$lnInc)[2])</pre>
SESCut[2] <- as.numeric(summary(full$lnInc)[3])</pre>
SESCut[3] <- as.numeric(summary(full$lnInc)[5])</pre>
## Calculate Strat Groups ## Smoking Quartiles
smkgrp <- ifelse(is.na(full$lnsmk), NA, ifelse(full$lnsmk < SmkCut[1], 1, ifelse(full$lnsmk >=
    SmkCut[1] & full$lnsmk < SmkCut[2], 2, ifelse(full$lnsmk >= SmkCut[2] &
    full$lnsmk < SmkCut[3], 3, 4))))</pre>
## SES Low-Income Cutoffs
SESgrp <- ifelse(is.na(full$MedIncome), NA, ifelse(full$MedIncome < 24000 &
    !is.na(full$MedIncome), 1, ifelse(full$MedIncome >= 24000 & full$MedIncome <
    28700, 2, ifelse(full$MedIncome >= 28700 & full$MedIncome < 38300, 3, 4))))
## SES Quartiles SESgrp <- ifelse(full$lnInc< -0.158, 1,
## ifelse(full$lnInc>=-0.158 & full$lnInc< -0.00391, 2,
## ifelse(full$lnInc>=-0.00391 & full$lnInc <0.1478, 3, 4))) Arsenic
## Quartiles
AsQ <- ifelse(is.na(full$lnAs), NA, ifelse(full$lnAs < AsCut[1], 1, ifelse(full$lnAs >=
    AsCut[1] & full$lnAs < AsCut[2], 2, ifelse(full$lnAs >= AsCut[2] & full$lnAs <
    AsCut[3], 3, 4))))
full$smkgrp <- ifelse(is.na(full$lnsmk), NA, ifelse(full$lnsmk < SmkCut[1],</pre>
    1, ifelse(full$lnsmk >= SmkCut[1] & full$lnsmk < SmkCut[2], 2, ifelse(full$lnsmk >=
        SmkCut[2] & full$lnsmk < SmkCut[3], 3, 4))))</pre>
## SES Low-Income Cutoffs
full$SESgrp <- ifelse(is.na(full$MedIncome), NA, ifelse(full$MedIncome < 24000 &
    !is.na(full$MedIncome), 1, ifelse(full$MedIncome >= 24000 & full$MedIncome <
    28700, 2, ifelse(full$MedIncome >= 28700 & full$MedIncome < 38300, 3, 4))))
## SES Quartiles SESgrp <- ifelse(full$lnInc< -0.158, 1,
## ifelse(full$lnInc>=-0.158 & full$lnInc< -0.00391, 2,
## ifelse(full$lnInc>=-0.00391 & full$lnInc <0.1478, 3, 4))) Arsenic
## Quartiles
full$AsQ <- ifelse(is.na(full$lnAs), NA, ifelse(full$lnAs < AsCut[1], 1, ifelse(full$lnAs >=
    AsCut[1] & full$lnAs < AsCut[2], 2, ifelse(full$lnAs >= AsCut[2] & full$lnAs <
    AsCut[3], 3, 4))))
## Quartile-Based Interaction Models Convert quartiles to factors
AsQf <- as.factor(AsQ)
smkgrpf <- as.factor(smkgrp)</pre>
smkgrpfbak <- smkgrpf</pre>
SESgrpf <- as.factor(SESgrp)</pre>
```

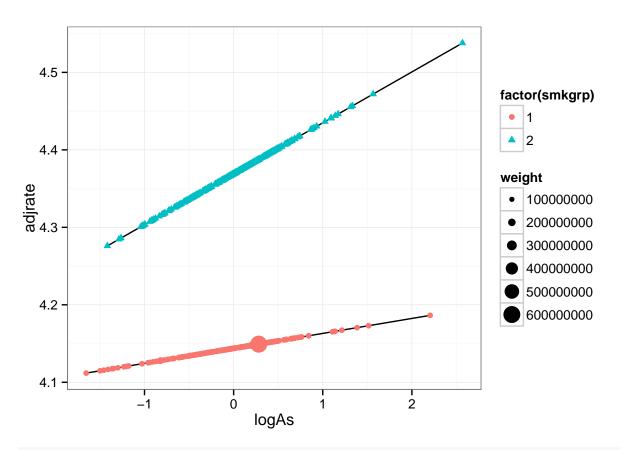
```
full$AsQf <- as.factor(full$AsQ)</pre>
full$smkgrpf <- as.factor(full$smkgrp)</pre>
full$smkgrpfbak <- full$smkgrpf</pre>
full$SESgrpf <- as.factor(full$SESgrp)</pre>
######### ARSENIC ## figure 2 Arsenic and Smoking This creates the data for the
######### first line of table 3 and the p value displayed on the right graph on
######## figure 2
smkgrpf <- smkgrpfbak</pre>
smkgrpf <- ifelse(is.na(smkgrpf), NA, ifelse(smkgrpf == 1 | smkgrpf == 2, 1,</pre>
intAsSmk <- aov(full$AdjRate ~ SESgrpf + AsQf * smkgrpf, weights = as.numeric(full$Population))
summary(intAsSmk)
                             Mean Sq F value
                                                             Pr(>F)
             Df
                     Sum Sq
              3 2695113615 898371205 37.943 < 0.0000000000000000 ***
SESgrpf
              3 1566705980 522235327 22.057
                                                  0.00000000000157 ***
AsQf
             1 2613664023 2613664023 110.390 < 0.00000000000000000 ***
smkgrpf
AsQf:smkgrpf 3
                 188901394 62967131
                                                             0.0475 *
                                         2.659
Residuals 574 13590400837
                              23676657
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
172 observations deleted due to missingness
## Without SES This creates the p value displayed on the left graph for
## figure 2
intAsSmk2 <- aov(full$AdjRate ~ AsQf * smkgrpf, weights = as.numeric(full$Population))</pre>
summary(intAsSmk2)
                     Sum Sq
                               Mean Sq F value
                                                             Pr(>F)
AsQf
              3 1890647672 630215891 24.392 0.00000000000000715 ***
              1 3665090147 3665090147 141.857 < 0.00000000000000000 ***
smkgrpf
AsQf:smkgrpf
                 191344179
                              63781393
                                         2.469
                                                             0.0611 .
              3
Residuals
          577 14907703852
                              25836575
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
172 observations deleted due to missingness
## Arsenic and SES this creates the statistics for table3 second model
intAsSES <- aov(full$AdjRate ~ smkgrpf + AsQf * SESgrpf, weights = as.numeric(full$Population))</pre>
summary(intAsSES)
             Df
                               Mean Sq F value
                                                             Pr(>F)
                     Sum Sq
smkgrpf
              1 4252222627 4252222627 180.22 < 0.0000000000000000 ***
              3 1303515192 434505064 18.42
                                                    0.000000000205 ***
AsQf
              3 1319745800 439915267 18.64
                                                    0.000000000151 ***
SESgrpf
AsQf:SESgrpf 9
                 377928730
                             41992081
                                          1.78
                                                             0.0691 .
Residuals
          568 13401373501
                              23593967
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
172 observations deleted due to missingness
```

```
## Plot the Interaction between Arsenic and Smoking WITHOUT SES
smkgrpf <- smkgrpfbak</pre>
smkgrpf <- ifelse(is.na(smkgrpf), NA, ifelse(smkgrpf==1 | smkgrpf==2,1,2))</pre>
# Two models are run to get the two lines for low and high smoking, unadjusted
r1 <- glm(full[smkgrpf==1,]$AdjRate ~ full[smkgrpf==1,]$lnAs, family=poisson, weights=as.numeric(full[smkgrpf==1,]$nAs, family=as.numeric(full[smkgrpf==1,]$nAs, family=a
summary(r1)
Call:
glm(formula = full[smkgrpf == 1, ]$AdjRate ~ full[smkgrpf ==
              1, ]$lnAs, family = poisson, weights = as.numeric(full[smkgrpf ==
             1, ]$Population))
Deviance Residuals:
             Min
                                                1Q
                                                             Median
                                                                                                              3Q
                                                                                                                                         Max
-4555.0
                                -105.5
                                                              172.0
                                                                                                   525.9
                                                                                                                                2252.1
Coefficients:
                                                                                                Estimate Std. Error z value
(Intercept)
                                                                                         4.14377600 0.00001604 258419.4
full[smkgrpf == 1, ]$lnAs 0.01931181 0.00003293
                                                                                                                               Pr(>|z|)
                                                                                          <0.0000000000000000 ***
(Intercept)
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for poisson family taken to be 1)
             Null deviance: 175916938 on 289 degrees of freedom
Residual deviance: 175572576 on 288 degrees of freedom
       (167 observations deleted due to missingness)
AIC: 543281859
Number of Fisher Scoring iterations: 4
r2 <- glm(full[smkgrpf==2,]$AdjRate ~ full[smkgrpf==2,]$lnAs, family=poisson, weights=as.numeric(full[smkgrpf==2,]$nAs, family=poiss
summary(r2)
glm(formula = full[smkgrpf == 2, ]$AdjRate ~ full[smkgrpf ==
             2, ]$lnAs, family = poisson, weights = as.numeric(full[smkgrpf ==
             2, ]$Population))
Deviance Residuals:
                                                                           Median
                                                                                                                            3Q
                Min 1Q
                                                                                                                                                           Max
-1773.74 -228.76
                                                                         53.34
                                                                                                              326.82
                                                                                                                                             1574.57
Coefficients:
                                                                                                Estimate Std. Error z value
                                                                                        4.36914074 0.00002369 184430
(Intercept)
```

```
Pr(>|z|)
                        <0.0000000000000000 ***
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for poisson family taken to be 1)
   Null deviance: 68053265 on 294 degrees of freedom
Residual deviance: 65713881 on 293 degrees of freedom
  (162 observations deleted due to missingness)
AIC: 205610885
Number of Fisher Scoring iterations: 4
   # this section builds the dataset from the model output for graphing
data1 <- cbind( c(t(r1$model[2]),t(r2$model[2])), # ln arsenic values</pre>
                      c(log(r1$fitted.values),log(r2$fitted.values)), # fitted dependent vars
                      c(r1$weights, r2$weights), # county populations
                      c(rep("1",dim(r1$model[2])[1]), rep("2",dim(r2$model[2])[1])))
                      # smoking group
data1 <- as.data.frame(data1, stringsAsFactors=FALSE)</pre>
names(data1) <- c("logAs","logRate", "weight", "smkgrp")</pre>
data1$logAs <- as.numeric(as.character(data1$logAs))</pre>
data1$logRate <- as.numeric(as.character(data1$logRate)) # these are the fitted values
data1$weight <- as.numeric(as.character(data1$weight))</pre>
data1$smkgrp <- as.numeric(as.character(data1$smkgrp))</pre>
# is zeroed out
data1$adjrate <- data1$adjinc+data1$logRate # the same as logRate</pre>
#This creates the graph but I'm not sure if you can have the fitted line without dots
assmkp <- ggplot(data1, aes(x=logAs, y=adjrate, shape=factor(smkgrp), color=factor(smkgrp)))
assmkp + stat_smooth(method = "glm", level=0.95, alpha=1, fill="grey80", color="black") +
        #scale_color_manual(values=c("grey50", "grey70")) +
        geom_point(aes(size=weight)) +
        geom point() +
        theme(legend.position = "right") +
        theme bw()
```

1534

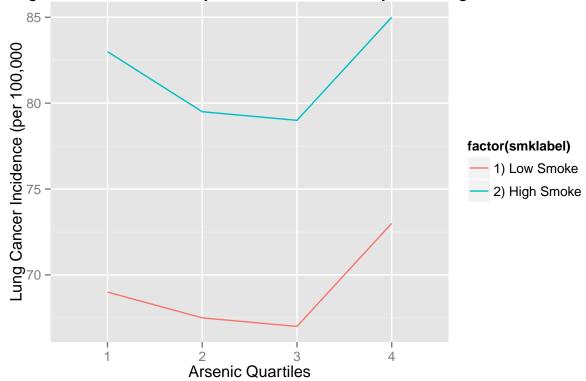
full[smkgrpf == 2,]\$lnAs 0.06571306 0.00004283



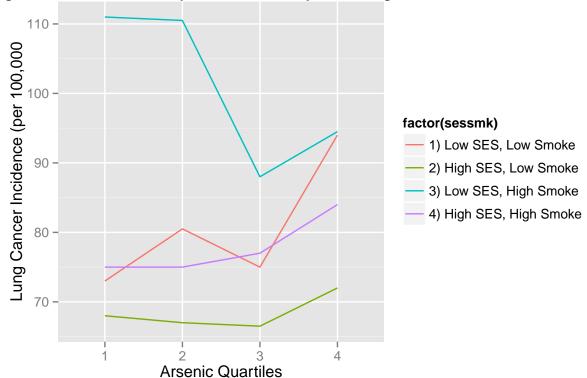
```
full$smkgrpf <- full$smkgrpfbak</pre>
full$smkgrpf <- ifelse(is.na(full$smkgrpf), NA, ifelse(full$smkgrpf==1 | full$smkgrpf==2, 1, 2))</pre>
full$SESgrp2f <- full$SESgrpf</pre>
full$SESgrp2f <- ifelse(is.na(full$SESgrp2f), NA, ifelse(full$SESgrp2f==1 | full$SESgrp2f==2, 1, 2))</pre>
# This creates the combined SES and smoking group into one variable, categories 1,2
# are the low smoking categories, important to note that the SES categories are not
# quartiles
full$sessmk = ifelse(full$SESgrp2f==1 & full$smkgrpf==1,"1) Low SES, Low Smoke",
              ifelse(full$SESgrp2f==2 & full$smkgrpf==1,"2) High SES, Low Smoke",
              ifelse(full$SESgrp2f==1 & full$smkgrpf==2,"3) Low SES, High Smoke",
            ifelse(full$SESgrp2f==2 & full$smkgrpf==2,"4) High SES, High Smoke",NA))))
# This is the graph with only two lines for the two smoking groups with actual data
full$smklabel <- ifelse(is.na(full$smkgrpf), NA, ifelse(full$smkgrpf==1,'1) Low Smoke', '2) High Smoke'
graphfile <- full %>%
             group_by(AsQf,smklabel) %>%
             summarise(meanrate = median(AdjRate) )
finalgraph <- graphfile[!is.na(graphfile$AsQf) & !is.na(graphfile$smklabel),]</pre>
ggplot(finalgraph, aes(x=AsQf, y=meanrate, color=factor(smklabel))) +
         geom line(aes(group=smklabel)) +
```

```
ggtitle("Lung Cancer Incidence by Arsenic Quartiles, by Smoking Cat") +
xlab("Arsenic Quartiles") +
ylab("Lung Cancer Incidence (per 100,000")
```

Lung Cancer Incidence by Arsenic Quartiles, by Smoking Cat



ng Cancer Incidence by Arsenic Qs, by Smoking/SES Cat



```
## GLMS for smoking levels WITH SES
smkgrpf <- smkgrpfbak
## Bottom 50% vs Top 50%
smkgrpf <- ifelse(is.na(smkgrpf), NA, ifelse(smkgrpf==1 | smkgrpf==2, 1, 2))

r1 <- glm(full[smkgrpf==1,]$AdjRate ~ full[smkgrpf==1,]$lnAs + full[smkgrpf==1,]$MedIncome, family=poiss
summary(r1)</pre>
```

Call:

```
glm(formula = full[smkgrpf == 1, ]$AdjRate ~ full[smkgrpf ==
    1, ]$lnAs + full[smkgrpf == 1, ]$MedIncome, family = poisson,
    weights = as.numeric(full[smkgrpf == 1, ]$Population))
```

Deviance Residuals:

```
Min 1Q Median 3Q Max -4941.6 -151.1 150.1 425.7 2249.4
```

Coefficients:

```
Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for poisson family taken to be 1)
   Null deviance: 175916938 on 289 degrees of freedom
Residual deviance: 168750465 on 287 degrees of freedom
  (167 observations deleted due to missingness)
AIC: 536459750
Number of Fisher Scoring iterations: 4
r2 <- glm(full[smkgrpf==2,]$AdjRate ~ full[smkgrpf==2,]$lnAs + full[smkgrpf==2,]$MedIncome, family=pois
summary(r2)
Call:
glm(formula = full[smkgrpf == 2, ]$AdjRate ~ full[smkgrpf ==
   2, ]$lnAs + full[smkgrpf == 2, ]$MedIncome, family = poisson,
   weights = as.numeric(full[smkgrpf == 2, ]$Population))
Deviance Residuals:
    Min
             10
                   Median
                                  3Q
                                           Max
-1769.12 -244.11
                    -9.96
                              270.72
                                       1591.53
Coefficients:
                                    Estimate
                                                 Std. Error z value
(Intercept)
                              4.688953956476 0.000113495701
                                                              41314
full[smkgrpf == 2, ]$lnAs
                              0.059353912842 0.000043115820
                                                               1377
full[smkgrpf == 2, ]$MedIncome -0.000007984707 0.00000002792
                                                              -2860
                                        Pr(>|z|)
                             <0.0000000000000000 ***
(Intercept)
full[smkgrpf == 2, ]$lnAs
                             Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for poisson family taken to be 1)
   Null deviance: 68053265 on 294 degrees of freedom
Residual deviance: 57437520 on 292 degrees of freedom
  (162 observations deleted due to missingness)
AIC: 197334525
Number of Fisher Scoring iterations: 4
## Plot the Interaction between Arsenic and Smoking with SES
data1 <- cbind( c(t(r1$model[2]),t(r2$model[2])), # ln arsenic values</pre>
               c(log(r1\fitted.values),log(r2\fitted.values)), # fitted dependent variables
               c(r1$weights, r2$weights), # county populations
               c(rep("1",dim(r1$model[2])[1]), rep("2",dim(r2$model[2])[1]))) # number of
data1 <- as.data.frame(data1, stringsAsFactors=FALSE)</pre>
names(data1) <- c("logAs","logRate", "weight", "smkgrp")</pre>
data1$logAs <- as.numeric(as.character(data1$logAs))</pre>
```

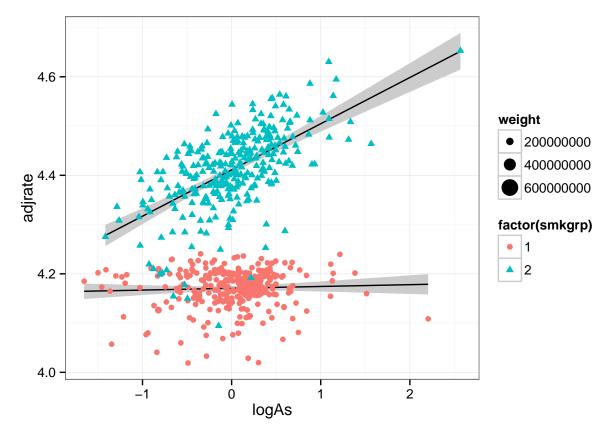


Table 3
Summary of ANOVA tests performed between Arsenic and covariates used in the regression analysis.

```
Interaction_Pair = c("Arsenic:Smoking", "Arsenic:MCI")
DF = c(3, 9)
F_value = c(2.6595, 1.7798)
P_value = c(0.04747, 0.06914)
df = data.frame(Interaction_Pair, DF, F_value, P_value)
kable(df)
```

Interaction_Pair	DF	F_value	P_value
Arsenic:Smoking Arsenic:MCI	3 9	2.6595 1.7798	$0.04747 \\ 0.06914$

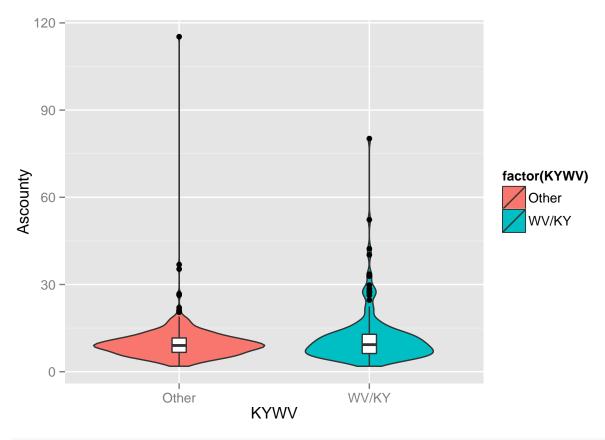
```
# change values in table to variables
```

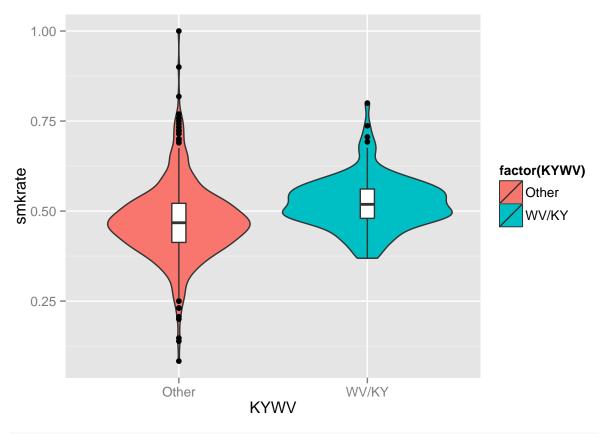
Figure 3

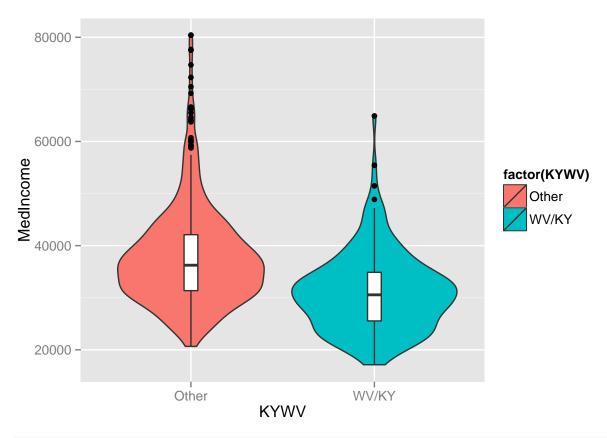
Combination violin and boxplots showing the average level of exposure and outcomes for counties in West Virginia or Kentucky comared with the remaining 10 states in the original sample.

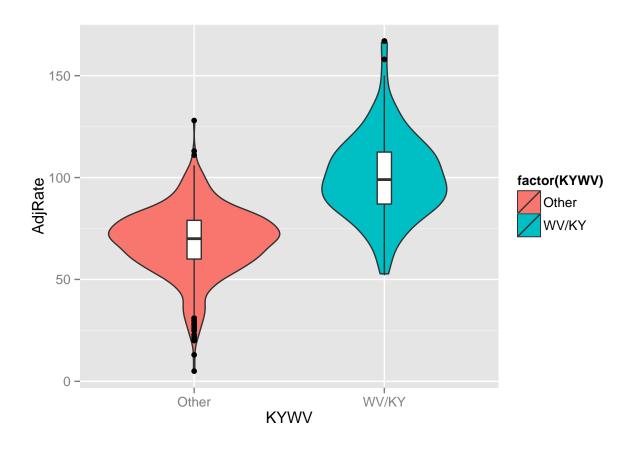
Instead of the bar plot, we represented the data in Figure 3 with a series of combination violin and boxplots. The authors of the original paper attempted to represent each of the variables of interest for this data on the same y-axis scale, including arsenic, income, lung cancer incidence rates, and smoking. Although the authors likely chose this visualization to save space, we decided that it was somewhat confusing and that additional information about the data distributions could be added if we instead used a combination of violin and boxplots.

Explanation of variables here? Add x label (location) and y label (arsenic, income, etc.)









Extension

^{**}Add extension here.