Lab #10 - Logistic Regression Part II

Econ 224
September 27th, 2018

Contaminated Wells in Bangladesh

Today we'll work with a dataset containing household-level information from Bangladesh: wells.csv. You can download the dataset from the course website at http://ditraglia.com/econ224/wells.csv.

Here is some background on the dataset from Gelman and Hill (2007):

Many of the wells used for drinking water in Bangladesh and other South Asian countries are contaminated with natural arsenic ... a research team from the United States and Bangladesh measured all the wells [in a small region] and labeled them with their arsenic level as well as a characterization of "safe" (below 0.5 in units of hundreds of micrograms per liter, the Bangladesh standard for arsenic in drinking water) or "unsafe" (above 0.5). People with unsafe wells were encouraged to switch to nearby private or community wells or to new wells of their own construction. A few years later, the researchers returned to find out who had switched wells.

Our goal is to predict which households will switch wells using the following information:

Name	Description
dist arsenic	Distance to closest known safe well (meters) Arsenic level of respondent's well (100s of micrograms/liter)
switch	Dummy variable: equals 1 if switched to a new well
assoc	Dummy variable: equals 1 if any member of the household is active in community organizations
educ	Education level of head of household (years)

To be clear, our dataset contains only information for households with an arsenic level of 0.5 or above, as these are the households that were encouraged to switch wells.

Exercises

- 1. Load the data and store it in a tibble called wells.
- 2. Use ggplot2 to make a histogram of arsenic. Be sure to label your plot appropriately. Comment on your findings.
- 3. Create a variable called dist100 that contains the same information as dist but measured in *hundreds* of meters rather than in meters.
- 4. Use ggplot2 to make a histogram of dist100. Be sure to label your plot appropriately. Comment on your findings.
- 5. Run a logistic regression using dist100 to predict switch and store the result in an object called fit1.
- 6. Use ggplot2 to plot the logistic regression function from fit1 along with the data, jittered appropriately.
- 7. Discuss your results from 5 and 6. In particular: based on fit1, is dist100 a statistically significant predictor of switch? Does the sign of its coefficient make sense? Explain.

- 8. Based on fit1, calculate the predicted probability of switching wells for a household for which the nearest safe well is zero meters away.
- 9. Based on fit1, calculate the predicted probability of switching wells for a household for which the nearest safe well is $\mathtt{dist}100$ hundred meters away, where a is the sample mean of $\mathtt{dist}100$.

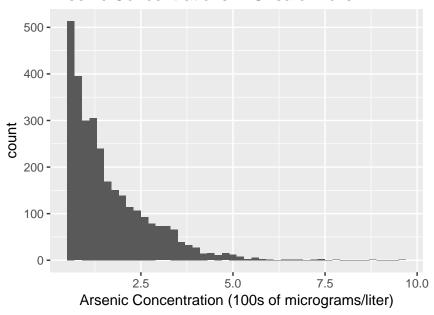
Solutions

1-2: Load Data, Histogram of arsenic

```
#----- Load data
library(tidyverse)
library(ggplot2)
wells <- read_csv('~/econ224/labs/wells.csv')

#----- Histogram of arsenic
ggplot(wells) +
   geom_histogram(aes(x = arsenic), binwidth = 0.2) +
   xlab('Arsenic Concentration (100s of micrograms/liter)') +
   ggtitle('Arsenic Concentrations in Unsafe Wells')</pre>
```

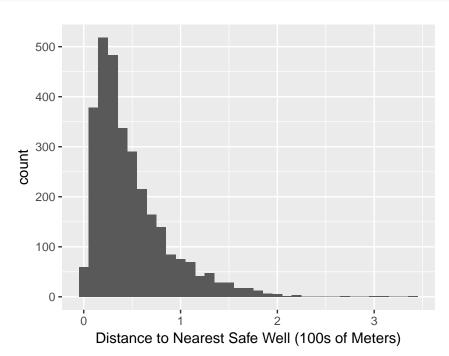
Arsenic Concentrations in Unsafe Wells



3-4: Histogram of dist100

```
#----- Create dist100
wells <- wells %>%
    mutate(dist100 = dist / 100)
#----- Plot histogram of dist100
```

```
ggplot(wells) +
  geom_histogram(aes(x = dist100), binwidth = 0.1) +
  xlab('Distance to Nearest Safe Well (100s of Meters)')
```



5-7: Generate, plot and interpret fit1

```
#----- Generate and summarize fit1
fit1 <- glm(switch ~ dist100, family = binomial(link = 'logit'), wells)</pre>
summary(fit1)
Call:
glm(formula = switch ~ dist100, family = binomial(link = "logit"),
   data = wells)
Deviance Residuals:
   Min
             1Q
                               3Q
                 Median
                                       Max
-1.4406 -1.3058
                 0.9669
                          1.0308
                                    1.6603
Coefficients:
            Estimate Std. Error z value Pr(>|z|)
                       0.06031 10.047 < 2e-16 ***
(Intercept) 0.60596
           -0.62188
                       0.09743 -6.383 1.74e-10 ***
dist100
Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
(Dispersion parameter for binomial family taken to be 1)
   Null deviance: 4118.1 on 3019 degrees of freedom
```

Residual deviance: 4076.2 on 3018 degrees of freedom

AIC: 4080.2

Number of Fisher Scoring iterations: 4

