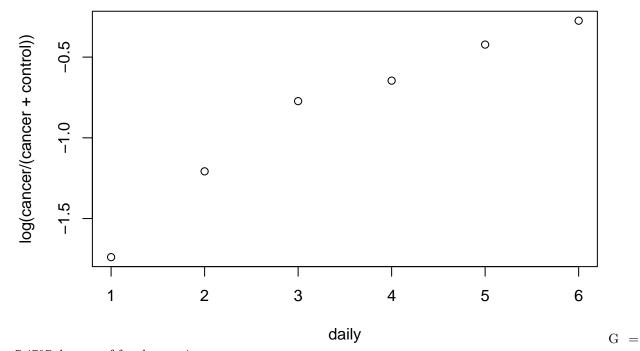
Homework 6

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1

a

```
cancer <- c(13, 55, 489, 475, 293, 38)
control <- c(61, 129, 570, 431, 154, 12)
y <- cbind(cancer, control)
daily = c(1,2,3,4,5,6)
lung <- data.frame(daily = c(1,2,3,4,5,6), rate = cancer / (cancer + control))
m1 <- glm(y ~ daily, family = binomial)
plot(daily, log(cancer/(cancer + control)))</pre>
```



7.4797 degrees of freedom = 4

 \mathbf{b}

```
phat <- fitted.values(m1)
ni = rowSums(y)
nphat = ni * phat
chis <- sum((y[,1] - nphat) ^ 2/ (nphat * (1-phat)))
chis</pre>
```

[1] 7.362

```
1-pchisq(chis,1)
## [1] 0.006661
We reject the null hypothesis.
\mathbf{c}
m0 \leftarrow glm(y \sim 1, family = binomial)
anova(m0, m1)
## Analysis of Deviance Table
## Model 1: y ~ 1
## Model 2: y ~ daily
   Resid. Df Resid. Dev Df Deviance
## 1
             5
                   130.9
## 2
             4
                     7.5 1
                                   123
1-pchisq(123.42,1)
## [1] 0
library(aod)
wald.test(b = coef(m1), Sigma = vcov(m1), Terms = 1)
## Wald test:
## -----
##
## Chi-squared test:
## X2 = 107.1, df = 1, P(> X2) = 0.0
LRT: 123.42, p=0 Wald: 107.1, p = 0 Both tests reject the null hypothesis.
\mathbf{d}
summary(m1)$coef
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) -1.6003
                            0.15466 -10.35 4.299e-25
## daily
                  0.4463
                            0.04159
                                     10.73 7.286e-27
confint(m1)
## Waiting for profiling to be done...
                 2.5 % 97.5 %
## (Intercept) -1.9059 -1.2994
               0.3654 0.5285
## daily
```

```
y2 <- matrix(c(68, 190, 964, 1001, 331, 166), ncol = 2)
m2 \leftarrow glm(y2 \sim c(1,2,3), family = binomial)
summary(m2)$coef
##
               Estimate Std. Error z value
                                               Pr(>|z|)
## (Intercept)
                 -4.980
                            0.17525 -28.41 1.365e-177
## c(1, 2, 3)
                   2.239
                            0.07379
                                     30.34 3.403e-202
confint(m2)
## Waiting for profiling to be done...
                2.5 % 97.5 %
## (Intercept) -5.331 -4.644
## c(1, 2, 3)
               2.097 2.386
estimate: 2.238741, confidence inteval: (2.096861 2.386239)
2
a
deaths \leftarrow c(2,32,12,104,28,206,28,186,31,102)
py <- c(18793,52407,10673,43248,5710,28612,2585,12663,1462,5317)
logpy \leftarrow log(py)
age <-gl(5,2)
smoke \leftarrow gl(2,1,length=10)
dataf <- data.frame(age=age,smoke=smoke,deaths=deaths,</pre>
                    py=py,logpy=logpy)
m1 <- glm(deaths ~ age+smoke,offset=logpy,family=poisson,</pre>
            data=dataf)
summary(m1)
##
## Call:
## glm(formula = deaths ~ age + smoke, family = poisson, data = dataf,
       offset = logpy)
##
##
## Deviance Residuals:
                  2
                            3
                                     4
                                               5
                                                        6
                                                                  7
## -2.1800
            0.9018 -1.3080 0.5104 -0.1379 0.0513
                                                             0.2289 -0.0873
## 1.9191 -0.9124
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
                            0.192 -41.30 < 2e-16 ***
## (Intercept) -7.919
```

```
0.195
                                    7.61 2.8e-14 ***
## age2
                 1.484
                 2.628
                            0.184
                                    14.30 < 2e-16 ***
## age3
## age4
                 3.351
                            0.185
                                    18.13 < 2e-16 ***
                 3.700
                                    19.25 < 2e-16 ***
                            0.192
## age5
## smoke2
                 0.355
                            0.107
                                    3.30 0.00096 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
      Null deviance: 935.091 on 9 degrees of freedom
## Residual deviance: 12.134 on 4 degrees of freedom
## AIC: 79.2
##
## Number of Fisher Scoring iterations: 4
datafinter <- c(0,1,0,2,0,3,0,4,0,5)
m2 <- glm(deaths~ age+smoke+inter,offset=logpy,family=poisson, data=dataf)
summary(m2)
##
## Call:
## glm(formula = deaths ~ age + smoke + inter, family = poisson,
##
      data = dataf, offset = logpy)
##
## Deviance Residuals:
##
        1
                 2
                          3
                                   4
                                            5
                                                    6
## -0.8775
            0.2712
                     0.2075 -0.0689
                                       0.6201 -0.2186 -0.4381
                                                                 0.1754
        9
## 0.0060 -0.0033
##
## Coefficients:
              Estimate Std. Error z value Pr(>|z|)
## (Intercept) -8.5856
                           0.3050 -28.15 < 2e-16 ***
                1.7346
                           0.2133
                                    8.13 4.2e-16 ***
## age2
                           0.2533
                                   12.43 < 2e-16 ***
## age3
                3.1484
## age4
                4.1420
                           0.3193
                                   12.97 < 2e-16 ***
                4.7310
                           0.3840
                                    12.32 < 2e-16 ***
## age5
                                    3.88 0.00011 ***
## smoke2
                1.4450
                           0.3729
## inter
               -0.3087
                           0.0973
                                   -3.17 0.00150 **
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for poisson family taken to be 1)
##
      Null deviance: 935.0909 on 9 degrees of freedom
## Residual deviance:
                       1.5464 on 3 degrees of freedom
## AIC: 70.61
##
## Number of Fisher Scoring iterations: 4
```

 \mathbf{b}

```
obsrate <- deaths/py
lam <- obsrate
eta <- log(lam)
z <- eta
w <- deaths
# do 5 Fisher scoring iterations
for (i in 1:4) {
m3 <- lm(z ~ age + smoke, weights=w) # regress z on X weights w
eta <- fitted(m3)
lam <- exp(eta)</pre>
z <- eta + (obsrate - lam)/lam
w <- py * lam }
coef(m3) # beta hat
## (Intercept)
                     age2
                                 age3
                                             age4
                                                        age5
                                                                   smoke2
##
      -7.9194
                   1.4840
                               2.6275
                                                       3.7001
                                                                   0.3545
                                           3.3505
varbeta <- summary(m3)$cov.unscaled # (X'WX)^-1</pre>
sqrt(diag(varbeta)) # standard errors
## (Intercept)
                                                                  smoke2
                     age2
                                             age4
                                                        age5
                                 age3
       0.1918
                   0.1951
                               0.1837 0.1848
##
                                                       0.1922
                                                                  0.1074
varbeta
              (Intercept)
                                age2
                                           age3
                                                      age4
## (Intercept) 0.036772 -0.0289194 -0.0286912 -0.0287139 -0.029050
                -0.028919 0.0380651 0.0294597 0.0294582 0.029436
## age2
## age3
                -0.028691 0.0294597 0.0337555 0.0294798 0.029447
              -0.028714 0.0294582 0.0294798 0.0341505 0.029446
## age4
                -0.029050 0.0294357 0.0294469 0.0294458 0.036948
## age5
                -0.009212 -0.0006159 -0.0009015 -0.0008731 -0.000453
## smoke2
##
                  smoke2
## (Intercept) -0.0092119
              -0.0006159
## age2
## age3
              -0.0009015
## age4
              -0.0008731
## age5
              -0.0004530
## smoke2
              0.0115292
vcov(m1)
               (Intercept)
                                age2
                                           age3
                                                      age4
                                                                age5
## (Intercept) 0.036772 -0.0289194 -0.0286913 -0.0287139 -0.029050
## age2
                -0.028919 0.0380651 0.0294597 0.0294582 0.029436
## age3
                -0.028691 0.0294597 0.0337555 0.0294798 0.029447
                -0.028714 0.0294582 0.0294798 0.0341506 0.029446
## age4
                -0.029050 0.0294358 0.0294470 0.0294459 0.036948
## age5
```