## Lec19\_1

## R Markdown

```
aba <- read.csv("Abalone.csv")
head (aba)
##
   Sex Length Diameter Height Whole Shucked Viscera Shell Rings
## 1
     1 0.455 0.365 0.095 0.5140 0.2245 0.1010 0.150
                0.265 0.090 0.2255 0.0995 0.0485 0.070
     1 0.350
                0.420 0.135 0.6770 0.2565 0.1415 0.210
## 3
     2 0.530
                                                          9
                0.365 0.125 0.5160 0.2155 0.1140 0.155
     1 0.440
## 4
                                                          10
    0 0.330
               0.255 0.080 0.2050 0.0895 0.0395 0.055
## 5
    0 0.425 0.300 0.095 0.3515 0.1410 0.0775 0.120
## 6
```

```
library (dplyr)
library (mcprofile)
```

a.

```
x <- c(0.1, 0.2, 0.3, 0.4)
shell.fit <- glm(Rings ~ log(Shell), family = poisson(link = "log"), data = aba)
K.means <- cbind(1,log(x))
meanpro <- mcprofile(shell.fit, CM=K.means)
CI.est <- exp(confint(meanpro))
CI.est$estimate %>% unlist() -> s
CI.est$confint %>% unlist() -> ss
Means <- data.frame(Rings = s, lower = ss[1:4], ipper = ss[5:8])
Means <- `rownames<-`(Means, c("Weight 0.1", "Weight 0.2", "Weight 0.3", "Weight 0.4"))
K.means</pre>
```

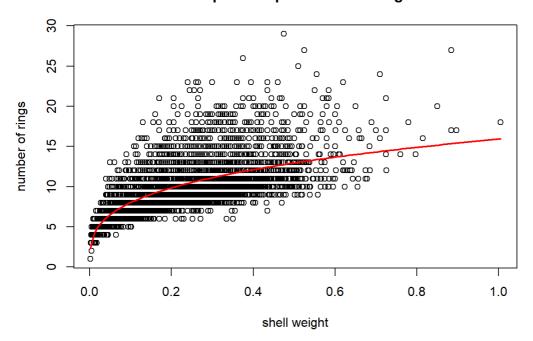
```
## [,1] [,2]
## [1,] 1 -2.3025851
## [2,] 1 -1.6094379
## [3,] 1 -1.2039728
## [4,] 1 -0.9162907
```

Means

```
## Weight 0.1 8.020632 7.881182 8.161429
## Weight 0.2 9.858468 9.744843 9.972957
## Weight 0.3 11.123016 10.984797 11.262369
## Weight 0.4 12.117423 11.936338 12.300537
```

```
plot(x = aba$Shell, y = aba$Rings, xlab = "shell weight", ylab = "number of rings", main = "Abalone poisson
predictor with log shell")
curve(expr = exp(shell.fit$coefficients[1] + log(x)*shell.fit$coefficients[2]), lwd = 2, add = TRUE, col = "
red")
```

## Abalone poisson predictor with log shell



Comparing to figure 5, 2 graphs are indentical.

b.

```
##log mean difference of a c-unit change is
\#\#beta_1*(log(x+c)-log(x))
x1 \leftarrow c(0.1, 0.2, 0.3)
c <- 0.1
change <- (\log(x1+c) - \log(x1))
K.ratio <- cbind(0, change)</pre>
mean.ratio <- mcprofile(shell.fit, CM = K.ratio)</pre>
CI.change <- exp(confint(mean.ratio))</pre>
##convert to percentage change
d <- unlist(CI.change$estimate) ##work on vectors so math manipulations are valid
f <- unlist(CI.change$confint)</pre>
100*(d-1) -> i ##Estiamte in percentage
100*(f-1) -> ii ##CIs in percentage
percent.change <- data.frame(estimate = i, lower = c(ii[1:3]), upper = c(ii[4:6]))
percent.change <- `row.names<-`(percent.change, c("change from 0.1 to 0.2", "change from 0.2 to 0.3", "change
e from 0.3 to 0.4"))
K.ratio
             change
## [1,] 0 0.6931472
```

```
## change
## [1,] 0 0.6931472
## [2,] 0 0.4054651
## [3,] 0 0.2876821
```

```
percent.change
```

```
## change from 0.1 to 0.2 22.913855 21.735715 24.109021
## change from 0.2 to 0.3 12.827024 12.193148 13.467488
## change from 0.3 to 0.4 8.940085 8.505481 9.378486
```

```
##check against part a
ests <- unlist(CI.est$estimate)
##convert to percentage change
m <- data.frame(manual = c((ests[2]-ests[1])/ests[1], (ests[3]-ests[2])/ests[2], (ests[4]-ests[3])/ests[3]))
cbind(m, percent.change)</pre>
```

```
## Estimate2 0.22913855 22.913855 21.735715 24.109021
## Estimate3 0.12827024 12.827024 12.193148 13.467488
## Estimate4 0.08940085 8.940085 8.505481 9.378486
```

From the comparison we can see percentage changes from mcprofile and rate of changes from manual calculation are identical.