Unit 3 Assignment

Sri Seshadri 11/5/2017

Chapter 6 Question 1

```
stress <- read.csv(file = "stress.csv")
knitr::kable(mosaic::fav_stats(stress$STRESS), caption = "Summary statistics")</pre>
```

Table 1: Summary statistics

min	Q1	median	Q3	max	mean	sd	n	missing
 0	0	1	3	9	1.729647	1.849082	651	0

a. Looking at the histogram and the summary statistic, where the mean and the standard deviaiton are alike, the most likely probability distribution in Poisson distribution.

```
qqnorm(y = stress$STRESS, main = "Normal Q-Q plot of stress")
qqline(y = stress$STRESS)
```

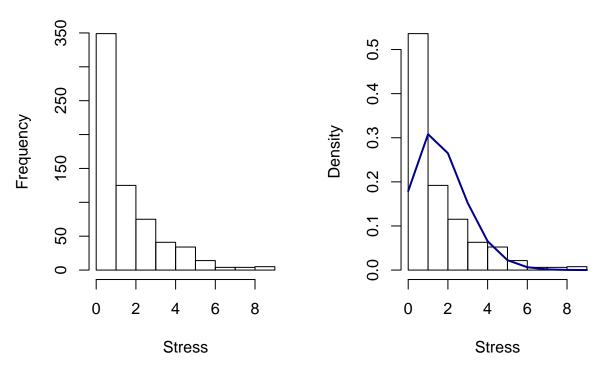
b. The variable "Stress" is a count variable and is discrete in nature. It does not align with a continuous distribution like normal. The evidence is seen in the clustering of points at different levels in the Q-Q plot.

Chapter 6 Question 2

```
options(width = 120)
# poisson regression
preg <- glm(formula = STRESS ~ COHES + ESTEEM + GRADES + SATTACH,
    family = poisson(link = "log"), data = stress)
preg.summary <- broom::tidy(preg)
knitr::kable(preg.summary, digits = 2, caption = "Poisson regression coefficients")</pre>
```

Histogram of stress – Frequency

Histogram of stress – Denisty



 $Figure \ 1: \ Histogram \ stress$

Normal Q-Q plot of stress

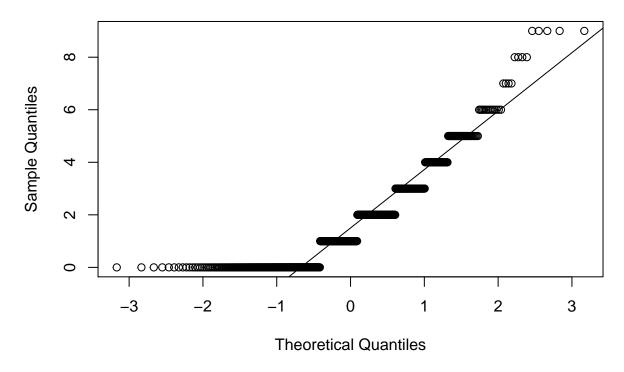


Figure 2: Normal Q-Q plot

Table 2: Poisson regression coefficients

term	estimate	$\operatorname{std.error}$	statistic	p.value
(Intercept)	2.73	0.23	11.68	0.00
COHES	-0.01	0.00	-4.47	0.00
ESTEEM	-0.02	0.01	-2.95	0.00
GRADES	-0.02	0.01	-2.38	0.02
SATTACH	-0.02	0.01	-2.85	0.00

Table 3: Overdispersed Poisson regression coefficients

term	estimate	std.error	statistic	p.value
(Intercept)	2.73	0.31	8.76	0.00
COHES	-0.01	0.00	-3.35	0.00
ESTEEM	-0.02	0.01	-2.21	0.03
GRADES	-0.02	0.01	-1.78	0.07
SATTACH	-0.02	0.01	-2.14	0.03

Table 4: Negative binomial regression coefficients

term	estimate	std.error	statistic	p.value
(Intercept)	2.76	0.34	8.08	0.00
COHES	-0.01	0.00	-3.24	0.00
ESTEEM	-0.02	0.01	-2.01	0.04
GRADES	-0.02	0.01	-1.74	0.08
SATTACH	-0.02	0.01	-2.02	0.04

a.

1. Poisson regression:

An increase in cohes by 1 unit reduces stress by 1.28%. An increase in sattach by 1 unit reduces stress by 1.63%

2. Negative binomial

An increase in cohes by 1 unit reduces stress by 1.33%.

An increase in sattach by 1 unit reduces stress by 1.66%

It is seen that the point estimates do not change much by the choice of distribution, because the mean and standard devation are not too far from eachother. (However the standard error does.)

b.

```
result <- stress %>% mutate(COHES.LEVEL = ifelse(COHES < mean(COHES) -
    sd(COHES), 0, ifelse(COHES > mean(COHES) + sd(COHES), 2,
    1))) %>% mutate(poisson.predict = preg$fitted.values) %>%
    mutate(nb.predict = nbreg$fitted.values) %>% group_by(COHES.LEVEL) %>%
    summarise(mean.poisson.prediction = mean(poisson.predict),
        mean.nb.prediction = mean(nb.predict))

knitr::kable(result, digits = 2, caption = "mean prediction for low (0), medium(1) , high(2) COHES")
```

Table 5: mean prediction for low (0), medium(1), high(2) COHES

COHES.LEVEL	mean.poisson.prediction	mean.nb.prediction
0	2.50	2.52
1	1.67	1.66
2	1.19	1.18

c.

Based on poisson regression the expected reduction in stress from high to low level of cohesion is : 52.52% Based on negative binomial regression the expected reduction on stress from high to low level of cohesion is : 53.33%

Chapter 6 Question 3

```
library(broom)
glances <- rbind(glance(preg), glance(overdisp.preg), glance(nbreg))
glances$model <- c("Poisson", "Overdispersed Poisson", "Negative Binomial")

output <- glances %>% select(model, AIC, BIC, logLik)
knitr::kable(output, digits = 2) # Notice no AICs for overdispersion
```

model	AIC	BIC	logLik
Poisson Overdispersed Poisson Negative Binomial	2417.22 NA 2283.59	NA	-1203.61 NA -1135.79
-			