

Negative binomial regression

Recap: inference with negative binomial models

```
...  
##           Estimate Std. Error z value Pr(>|z|)  
## (Intercept)  2.877771    0.123477  23.306 < 2e-16 ***  
## male        0.459148    0.027641  16.611 < 2e-16 ***  
## age        -0.007010    0.001731  -4.050 5.12e-05 ***  
## education2  0.024518    0.032534   0.754  0.451  
## education3  0.009252    0.040802   0.227  0.821  
## education4 -0.027732    0.044825  -0.619  0.536  
## diabetes   -0.010124    0.099126  -0.102  0.919  
## BMI        0.003693    0.003573   1.033  0.301  
...
```

Likelihood ratio test

```
m2 <- glm.nb(cigsPerDay ~ male + age + education +  
              diabetes + BMI, data = smokers)  
m3 <- glm.nb(cigsPerDay ~ male + age +  
              diabetes + BMI, data = smokers)  
m2$twologlik - m3$twologlik
```

```
## [1] 1.423055
```

```
pchisq(1.423, df=3, lower.tail=F)
```

```
## [1] 0.7001524
```

Likelihood ratio test

Why can I use the residual deviance to perform a likelihood ratio test for a Poisson regression model, but not for a negative binomial model?

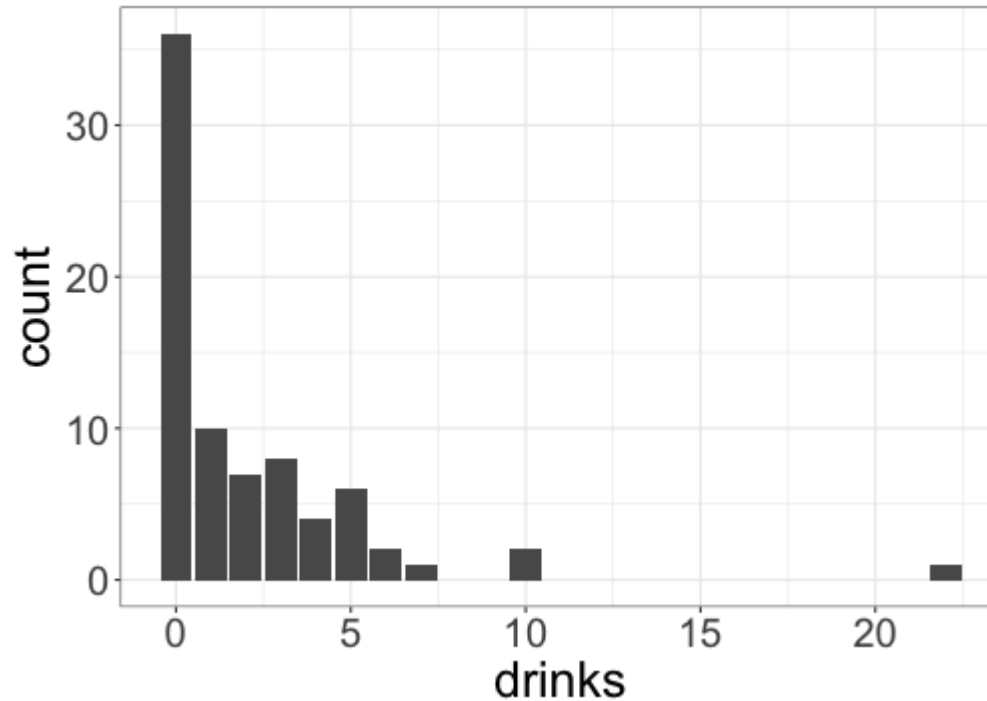
New data

Survey data from 77 college students on a dry campus (i.e., alcohol is prohibited) in the US. Survey asks students "How many alcoholic drinks did you consume last weekend?"

- + drinks: the number of drinks the student reports consuming
- + sex: an indicator for whether the student identifies as male
- + OffCampus: an indicator for whether the student lives off campus
- + FirstYear: an indicator for whether the student is a first-year student

Our goal: model the number of drinks students report consuming.

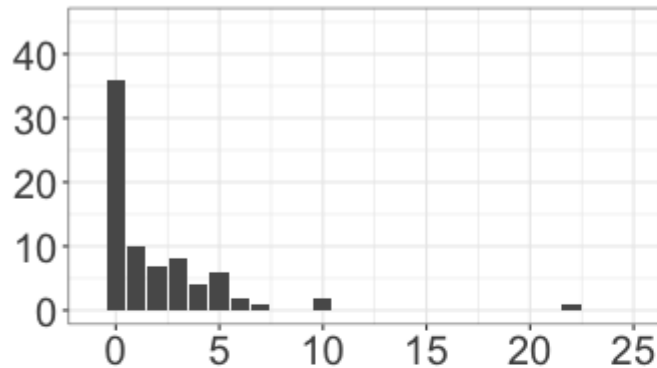
EDA: drinks



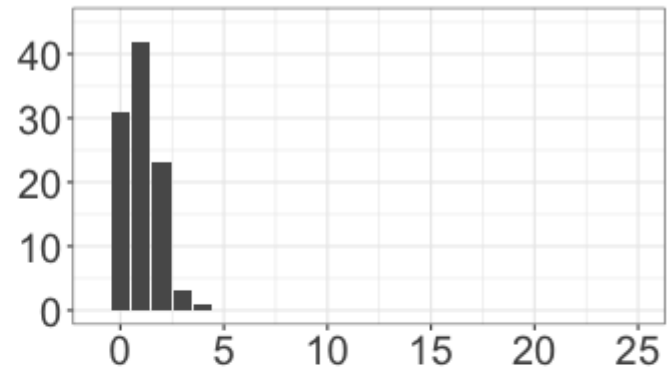
What do you notice about this distribution?

Comparisons with Poisson distributions

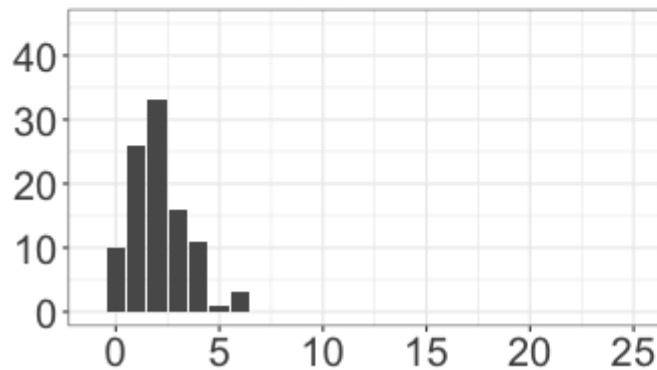
Observed data



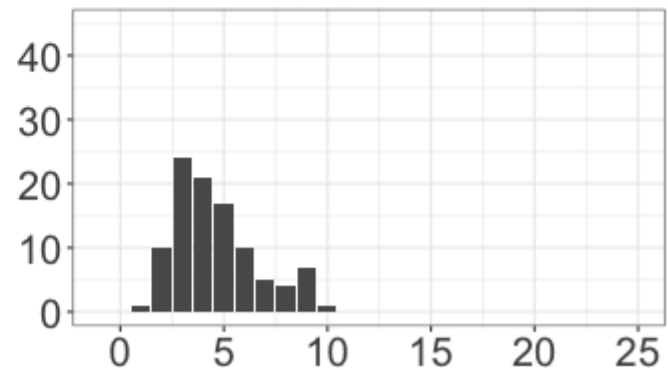
Poisson(1)



Poisson(2)

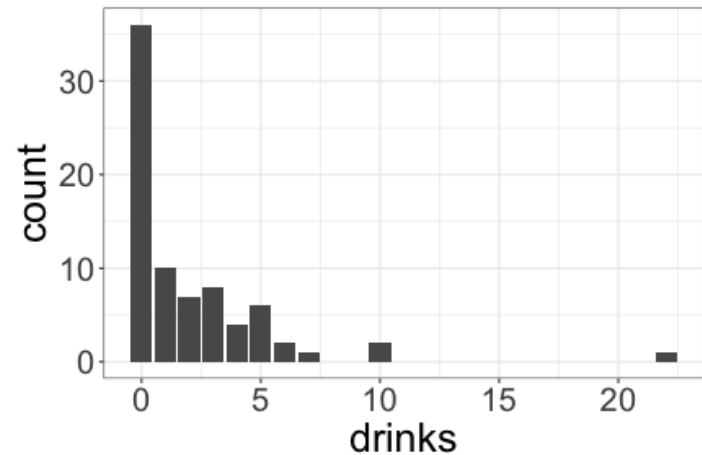


Poisson(5)



Excess zeros

Why might there be excess 0s in the data, and why is that a problem for modeling the number of drinks consumed?



Modeling

Zero-inflated Poisson (ZIP) model

$$P(Y_i = y) = \begin{cases} e^{-\lambda_i}(1 - \alpha_i) + \alpha_i & y = 0 \\ \frac{e^{-\lambda_i} \lambda_i^y}{y!} (1 - \alpha_i) & y > 0 \end{cases}$$

where

$$\log\left(\frac{\alpha_i}{1 - \alpha_i}\right) = \gamma_0 + \gamma_1 \textit{FirstYear}_i + \gamma_2 \textit{OffCampus}_i + \gamma_3 \textit{Male}_i$$

$$\log(\lambda_i) = \beta_0 + \beta_1 \textit{FirstYear}_i + \beta_2 \textit{OffCampus}_i + \beta_3 \textit{Male}_i$$

In R

```
library(pscl)
m1 <- zeroinfl(drinks ~ FirstYear + OffCampus + sex |
               FirstYear + OffCampus + sex,
               data = wdrinks)
summary(m1)
```

```
...
## Count model coefficients (poisson with log link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)      0.8010      0.1620   4.945 7.60e-07 ***
## FirstYearTRUE  -0.1619      0.3651  -0.444   0.6574
## OffCampusTRUE   0.3724      0.2135   1.744   0.0811 .
## sexm            0.9835      0.1889   5.205 1.94e-07 ***
##
## Zero-inflation model coefficients (binomial with logit link):
##               Estimate Std. Error z value Pr(>|z|)
## (Intercept)    -0.39618     0.39752  -0.997   0.319
## FirstYearTRUE   0.89197     0.65878   1.354   0.176
## OffCampusTRUE  -1.69137     1.47761  -1.145   0.252
## sexm           -0.07079     0.58846  -0.120   0.904
...
```

Interpretation

...

Zero-inflation model coefficients (binomial with logit link)

| ## | | Estimate | Std. Error | z value | Pr(> z) |
|----|---------------|----------|------------|---------|----------|
| ## | (Intercept) | -0.39618 | 0.39752 | -0.997 | 0.319 |
| ## | FirstYearTRUE | 0.89197 | 0.65878 | 1.354 | 0.176 |
| ## | OffCampusTRUE | -1.69137 | 1.47761 | -1.145 | 0.252 |
| ## | sexm | -0.07079 | 0.58846 | -0.120 | 0.904 |

...

How would I interpret the estimated coefficient 0.892 in the logistic regression component of the model?

Interpretation

```
...  
## Count model coefficients (poisson with log link):  
##           Estimate Std. Error z value Pr(>|z|)  
## (Intercept)    0.8010     0.1620   4.945 7.60e-07 ***  
## FirstYearTRUE -0.1619     0.3651  -0.444  0.6574  
## OffCampusTRUE  0.3724     0.2135   1.744  0.0811 .  
## sexm           0.9835     0.1889   5.205 1.94e-07 ***  
##  
...
```

How would I interpret the estimated coefficient 0.372 in the Poisson regression component of the model?