### Zero-truncated negative binomial regression

Zero-truncated negative binomial regression is used to model count data for which the value zero cannot occur and for which over dispersion exists.

#### Zero-truncated negative binomial regression

- ➤ To fit the zero-truncated negative binomial model, we use the vglm function in the VGAM package.
- This function fits a very flexible class of models called vector generalized linear models to a wide range of assumed distributions.
- ▶ In our case, we believe the data come from the negative binomial distribution, but without zeros.
- Thus the values are strictly positive poisson, for which we use the positive negative binomial family via the posnegbinomial function passed to vglm.

# Zero-truncated negative binomial regression

#### Fitting the Model with R

We will use the hospitalstay data again.

```
m1 <- vglm(stay ~ age + hmo + died,
  family = posnegbinomial(),
  data = hospitalstay)</pre>
```

```
summary(m1)
##
## Call:
## vglm(formula = stay ~ age + hmo + died,
      family = posnegbinomial(),
##
##
      data = hospitalstay)
##
## Pearson Residuals:
##
              Min 10 Median 30 Max
## log(munb) -1.4 -0.70 -0.23 0.45 9.8
## log(size) -14.1 -0.27 0.45 0.76 1.0
```

Coefficients:

```
##
                 Estimate Std. Error z value
   (Intercept):1
                                       33.6
                   2.408
                               0.072
                              0.055
   (Intercept):2
                    0.569
                                        10.4
                  -0.016
                               0.013 - 1.2
## age
## hmo1
                  -0.147
                              0.059 - 2.5
                  -0.218
                              0.046 - 4.7
## died1
```

- ► The first intercept is what we know as the typical intercept.
- ▶ The second is the **over dispersion parameter**,  $\alpha$ .
- The number of linear predictors is two, one for the expected mean λ and one for the over dispersion.
- Next the dispersion parameter is printed, assumed to be one after accounting for overdispersion.

- ► The value of the coefficient for age, -0.0157 suggests that the log count of stay decreases by 0.0157 for each year increase in age.
- ► The coefficient for hmo, -0.1471 indicates that the log count of stay for HMO patient is 0.1471 less than for non-HMO patients.
- ► The log count of stay for patients who died while in the hospital was 0.2178 less than those patients who did not die.

- ► The value of the constant 2.4083 is the log count of the stay when all of the predictors equal zero.
- ▶ The value of the second intercept, the over dispersion parameter,  $\alpha$  is 0.5686.
- ► To test whether we need to estimate over dispersion, we could fit a zero-truncated Poisson model and compare the two. (Not Covered).