Insolvency - (Quasi-)Poisson Model and Negative Binomial Model

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First the insolvency data are loaded:

```
library(catdata)
data(insolvency)
attach(insolvency)

## Das folgende Objekt ist maskiert encephalitis:
##
## year
```

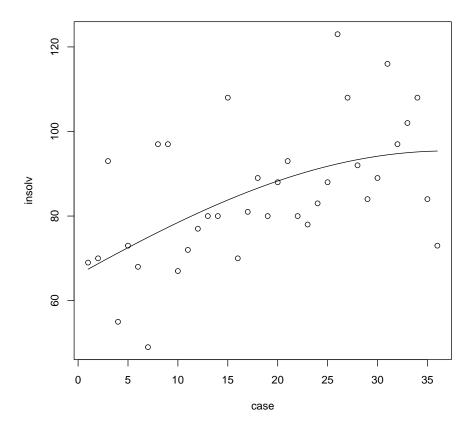
For the number of insolvent firms between 1994 and 1996 a Poisson model is fitted with time as predictor. Time is considered as a number from 1 to 36, denoting the month from January 1994 to December 1996.

```
ins1 <- glm(insolv ~ case + I(case^2), family=poisson(link=log), data=insolvency)
summary(ins1)
## Call:
## glm(formula = insolv ~ case + I(case^2), family = poisson(link = log),
      data = insolvency)
## Deviance Residuals:
   Min 1Q Median
                                3Q
                                         Max
## -3.2037 -0.9083 -0.2517 0.4880
                                      3.0340
##
## Coefficients:
               Estimate Std. Error z value Pr(>|z|)
##
## (Intercept) 4.1916952 0.0617994 67.827 < 2e-16 ***
             0.0197825 0.0073901
                                   2.677 0.00743 **
## I(case^2) -0.0002670 0.0001896 -1.409 0.15897
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for poisson family taken to be 1)
##
      Null deviance: 108.128 on 35 degrees of freedom
```

```
## Residual deviance: 75.287 on 33 degrees of freedom
## AIC: 306.82
##
## Number of Fisher Scoring iterations: 4
# plot(ins1)
```

Scatter-Plot of number of insolvent firms dependent of the month (1-36). With estimated curve of the log-linear model.

```
plot(case, insolv)
points(ins1\fitted.values, type="l")
```



In many real-world datasets the variance of count-data is higher than predicted by the Poisson distribution. So next a Poisson model with disperison parameter is fitted (Quasi-Poisson model).

```
ins2 <- glm(insolv ~ case + I(case^2), family=quasipoisson, data=insolvency)
summary(ins2)
##</pre>
```

```
## Call:
## glm(formula = insolv ~ case + I(case^2), family = quasipoisson,
    data = insolvency)
##
## Deviance Residuals:
   Min 1Q Median
##
                             3Q
                                       Max
## -3.2037 -0.9083 -0.2517 0.4880
                                     3.0340
##
## Coefficients:
##
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.1916952 0.0939826 44.601 <2e-16 ***
             0.0197825 0.0112387
                                   1.760
                                            0.0876 .
## I(case^2) -0.0002670 0.0002883 -0.926
                                            0.3611
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for quasipoisson family taken to be 2.312738)
##
      Null deviance: 108.128 on 35 degrees of freedom
## Residual deviance: 75.287 on 33 degrees of freedom
## AIC: NA
##
## Number of Fisher Scoring iterations: 4
# plot(ins2)
```

An alternative to a quasi-poisson model is to use the negative binomial distribution.

```
library(MASS)
ins3 <- glm.nb(insolv ~ case + I(case^2),data=insolvency)</pre>
summary(ins3)
##
## glm.nb(formula = insolv ~ case + I(case^2), data = insolvency,
     init.theta = 77.92952593, link = log)
##
##
## Deviance Residuals:
## Min 1Q Median 3Q
                                      Max
## -2.3666 -0.6333 -0.1710 0.3350
                                   2.0042
##
## Coefficients:
##
               Estimate Std. Error z value Pr(>|z|)
## (Intercept) 4.1953863 0.0861256 48.712 <2e-16 ***
## case 0.0192833 0.0105170 1.834
                                           0.0667 .
## I(case^2) -0.0002546 0.0002728 -0.933
                                          0.3506
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

```
## (Dispersion parameter for Negative Binomial(77.9295) family taken to be 1)
##
## Null deviance: 52.104 on 35 degrees of freedom
## Residual deviance: 36.312 on 33 degrees of freedom
## AIC: 296.27
##
## Number of Fisher Scoring iterations: 1
##
## Theta: 77.9
## Std. Err.: 35.5
##
## 2 x log-likelihood: -288.269
```

Since counts are rather large in addition a normal distribution model is fitted.

```
ins4 <- glm(insolv ~ case + I(case^2), family=gaussian(link=log), data=insolvency)</pre>
summary(ins4)
##
## Call:
## glm(formula = insolv ~ case + I(case^2), family = gaussian(link = log),
## data = insolvency)
## Deviance Residuals:
## Min 10 Median
                                30
                                        Max
## -25.809 -8.744 -2.374 4.560
                                      30.480
##
## Coefficients:
               Estimate Std. Error t value Pr(>|t|)
## (Intercept) 4.1836089 0.1005663 41.600 <2e-16 ***
             0.0208026 0.0115423 1.802 0.0806 .
## I(case^2) -0.0002915 0.0002896 -1.007 0.3214
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## (Dispersion parameter for gaussian family taken to be 193.2793)
##
      Null deviance: 9147.0 on 35 degrees of freedom
## Residual deviance: 6378.1 on 33 degrees of freedom
## AIC: 296.54
##
## Number of Fisher Scoring iterations: 4
```