



Contact during exam:
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EXAM IN TMA4315 GENERALIZED LINEAR MODELS

Friday December 10th, 2010
Time: 09:00 – 13:00

Permitted aids:

Tabeller og formler i statistikk, Tapir Forlag

K. Rottmann: Matematisk formelsamling

Calculator HP30S / CITIZEN SR-270X

Yellow, stamped A4-sheet with your own handwritten notes.

Examination results are due: December 28th 2010

Problem 1 Number of buss passengers

A bus driver wants to model how many passengers he gets from the bus stop close to the student home. He can think of three explanatory variables; which route it is (8 am or 9 am), if it is during the semester or not, and the temperature. He has data for 20 days, given in the table below. He consider three different models, all analyzed in R (see edited printout below); *model 1* gives `result1`, *model 2* gives `result2` and *model 3* gives `result3`.

- a) Set up the generalized linear model (GLM) used for *model 1* mathematically, specify assumptions, and specify the design matrix X for the first 6 observations. Also specify which strategy that is used to ensure identifiability, and discuss briefly alternative(s). Explain, mathematically and with words, what model the R notation `temp*semester` gives (as in *model 2*).

| | Passengers | route | semester | temp |
|----|------------|-------|-------------|------|
| 1 | 3 | 8am | semester | 8.8 |
| 2 | 1 | 9am | nonSemester | 11.5 |
| 3 | 1 | 8am | nonSemester | 12.0 |
| 4 | 3 | 8am | semester | 14.8 |
| 5 | 0 | 8am | nonSemester | -1.2 |
| 6 | 0 | 8am | nonSemester | 7.8 |
| 7 | 0 | 8am | nonSemester | 6.9 |
| 8 | 1 | 9am | nonSemester | 7.5 |
| 9 | 6 | 8am | semester | 7.7 |
| 10 | 2 | 8am | semester | 5.5 |
| 11 | 1 | 8am | nonSemester | 13.7 |
| 12 | 1 | 8am | nonSemester | 13.1 |
| 13 | 0 | 9am | nonSemester | 14.2 |
| 14 | 2 | 9am | nonSemester | 0.2 |
| 15 | 4 | 8am | nonSemester | -4.7 |
| 16 | 0 | 9am | nonSemester | 26.3 |
| 17 | 3 | 9am | semester | 3.1 |
| 18 | 2 | 8am | semester | -4.0 |
| 19 | 1 | 9am | nonSemester | 18.4 |
| 20 | 2 | 8am | nonSemester | -5.0 |

- b) Consider *model 1*. Based on the results from R:
 What is the expected number of passengers for the 9 am route, during the semester when it is 5.4 degrees C?
 What is the expected number of passengers for the 8 am route, during non-semester when it is -15.2 degrees C?
- c) We now want to compare models: Set up a hypothesis for testing *model 2* against *model 1* using the likelihood ratio test (i.e. based on deviance), and do the test.
 Which of the models, *model 1*, *model 2* or *model 3*, would you prefer. Why?
- d) Let Y_1, \dots, Y_N be independent responses with $Y_i \sim Po(\lambda_i)$. For the model of interest, with $p < N$ parameters, let \hat{y}_i be the fitted values based on the maximum likelihood estimates. Find an expression, based on y_i and \hat{y}_i , for the deviance in this case.

```
> result1 = glm(Passengers~temp+semester, family=poisson(link="log"))
> summary(result1)
```

Coefficients:

| | Estimate | Std. Error | z value | Pr(> z) |
|------------------|----------|------------|---------|------------|
| (Intercept) | 0.25406 | 0.30667 | 0.828 | 0.40741 |
| temp | -0.03451 | 0.02462 | -1.401 | 0.16107 |
| semestersemester | 1.08499 | 0.35365 | 3.068 | 0.00216 ** |

Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

Null deviance: 30.406 on 19 degrees of freedom
 Residual deviance: 17.677 on 17 degrees of freedom
 AIC: 62.03

```
> result2 = glm(Passengers~temp*semester, family=poisson(link="log"))
> summary(result2)
```

Coefficients:

| | Estimate | Std. Error | z value | Pr(> z) |
|-----------------------|----------|------------|---------|----------|
| (Intercept) | 0.44315 | 0.29124 | 1.522 | 0.1281 |
| temp | -0.07445 | 0.03384 | -2.200 | 0.0278 * |
| semestersemester | 0.54611 | 0.46383 | 1.177 | 0.2390 |
| temp:semestersemester | 0.10002 | 0.05316 | 1.881 | 0.0599 . |

Null deviance: 30.406 on 19 degrees of freedom
 Residual deviance: 13.981 on 16 degrees of freedom
 AIC: 60.334

```
> result3 = glm(Passengers~temp+semester+route, family=poisson(link="log"))
> summary(result3)
```

Coefficients:

| | Estimate | Std. Error | z value | Pr(> z) |
|------------------|----------|------------|---------|------------|
| (Intercept) | 0.28227 | 0.32780 | 0.861 | 0.38918 |
| temp | -0.03345 | 0.02501 | -1.338 | 0.18095 |
| semestersemester | 1.06849 | 0.36035 | 2.965 | 0.00303 ** |
| route9am | -0.09713 | 0.42224 | -0.230 | 0.81806 |

Null deviance: 30.406 on 19 degrees of freedom
 Residual deviance: 17.623 on 16 degrees of freedom
 AIC: 63.976

Problem 2 Negative binomial distribution

The probability density function for a negative binomial random variable is

$$f_y(y; \theta, r) = \frac{\Gamma(y+r)}{y!\Gamma(r)}(1-\theta)^r\theta^y$$

for $y = 0, 1, 2, \dots$, $r > 0$ and $\theta \in (0, 1)$, and where $\Gamma()$ denotes the gamma function. (There are also other parameterizations of the negative binomial distributions, but use this for now.)

- a) Show that the negative binomial distribution is a member of the exponential family. You can in this question consider r as a known constant.
- b) Use the general formulas for a exponential family to show that $E(Y) = \mu = r\frac{\theta}{1-\theta}$ and $Var(Y) = \mu\frac{1}{1-\theta}$
- c) Set up a GLM for the dataset in problem 1 with a negative binomial response function and a linear component similar to that in *model 1*.
Argue for your choice of link-function.
What role does r have?
In which situations could it be beneficial to use a negative binomial response function instead of a Poisson response function? Why?