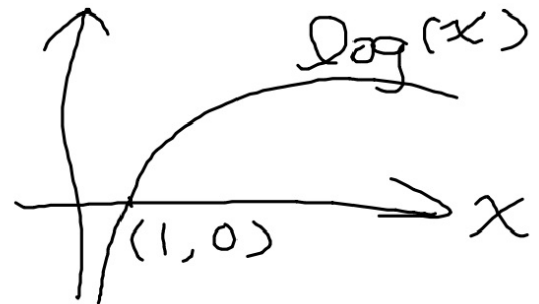


Computing in Epidemiology and Biostatistics
Find the maximum likelihood estimates for a Poisson regression
Wan-Yu Lin

Ex 20-2 : Data: seizure.csv please read it into R and call it "seizure" object :

- (1) Please use the Newton-Raphson method to find the maximum likelihood estimate (MLE) of the regression coefficients of Poisson regression ($\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2$), and compare your result with the R built-in function `glm(y~trt+age, data=seizure, offset=lweek, family=poisson)`
- (2) Please find the variance-covariance matrix for $\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2$, and compare your result with the R built-in function `vcov(model)`
- (3) Please find the log likelihood at $\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2$, and compare your result with the R built-in function `logLik(model)`



[Hint]

the average seizure count per week ≥ 0

$$\log\left(\frac{E(Y_i)}{\text{week}_i}\right) = \beta_0 + \beta_1 \text{trt}_i + \beta_2 \text{age}_i = \begin{bmatrix} 1 & \text{trt}_i & \text{age}_i \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{bmatrix} = \mathbf{x}_i' \boldsymbol{\beta}$$

$$\log(E(Y_i)) - \log(\text{week}_i) = \mathbf{x}_i' \boldsymbol{\beta}$$

$$\log(E(Y_i)) = \mathbf{x}_i' \boldsymbol{\beta} + \log(\text{week}_i)$$

$$\log(\mu_i) = \mathbf{x}_i' \boldsymbol{\beta} + \log(\text{week}_i) \quad \Rightarrow \quad \mu_i = e^{\mathbf{x}_i' \boldsymbol{\beta}} \text{week}_i = e^{\mathbf{x}_i' \boldsymbol{\beta}} e^{\log(\text{week}_i)} = e^{\mathbf{x}_i' \boldsymbol{\beta} + \log(\text{week}_i)}$$

$$L(\boldsymbol{\beta}) = \prod_{i=1}^n \frac{e^{-\mu_i} \mu_i^{y_i}}{y_i!}$$

$$\log L(\boldsymbol{\beta}) = \sum_{i=1}^n \{-\mu_i + y_i \log \mu_i - \log(y_i!)\}$$

$$= \sum_{i=1}^n \left\{ -e^{x_i' \boldsymbol{\beta}} \text{week}_i + y_i \left(\mathbf{x}_i' \boldsymbol{\beta} + \log(\text{week}_i) \right) - \log(y_i!) \right\}$$

$$= \sum_{i=1}^n \left\{ -e^{x_i' \boldsymbol{\beta}} \text{week}_i + y_i \mathbf{x}_i' \boldsymbol{\beta} + y_i \log(\text{week}_i) - \log(y_i!) \right\}$$

$$\frac{d \log L(\boldsymbol{\beta})}{d \boldsymbol{\beta}} = \sum_{i=1}^n \left\{ \mathbf{x}_{i_{3 \times 1}} y_{i_{1 \times 1}} - \mathbf{x}_{i_{3 \times 1}} e^{x_i' \boldsymbol{\beta}} \text{week}_{i_{1 \times 1}} \right\} = \sum_{i=1}^n \mathbf{x}_i \left\{ y_i - e^{x_i' \boldsymbol{\beta}} \text{week}_i \right\} = \sum_{i=1}^n \mathbf{x}_i \{y_i - \mu_i\}$$

$$= \mathbf{x}_1 \{y_1 - \mu_1\} + \mathbf{x}_2 \{y_2 - \mu_2\} + \cdots + \mathbf{x}_n \{y_n - \mu_n\}$$

$$= \begin{bmatrix} 1 \\ \text{trt}_1 \\ \text{age}_1 \end{bmatrix} \{y_1 - \mu_1\} + \begin{bmatrix} 1 \\ \text{trt}_2 \\ \text{age}_2 \end{bmatrix} \{y_2 - \mu_2\} + \cdots + \begin{bmatrix} 1 \\ \text{trt}_n \\ \text{age}_n \end{bmatrix} \{y_n - \mu_n\}$$

$$= \begin{bmatrix} 1 & 1 & \cdots & 1 \\ \text{trt}_1 & \text{trt}_2 & \cdots & \text{trt}_n \\ \text{age}_1 & \text{age}_2 & \cdots & \text{age}_n \end{bmatrix}_{3 \times n} \begin{bmatrix} y_1 - \mu_1 \\ y_2 - \mu_2 \\ \vdots \\ y_n - \mu_n \end{bmatrix}_{n \times 1}$$

$$\frac{d^2 \log L(\boldsymbol{\beta})}{d \boldsymbol{\beta}^2} = \sum_{i=1}^n \left\{ -\mathbf{x}_{i_{3 \times 1}} e^{x_i' \boldsymbol{\beta}} \text{week}_{i_{1 \times 1}} \mathbf{x}_{i_{1 \times 3}}' \right\} = \sum_{i=1}^n \left\{ -\mathbf{x}_i \mu_i \mathbf{x}_i' \right\}$$

$$= -\mathbf{x}_1 \mu_1 \mathbf{x}_1' - \mathbf{x}_2 \mu_2 \mathbf{x}_2' - \cdots - \mathbf{x}_n \mu_n \mathbf{x}_n'$$

$$= - \begin{bmatrix} 1 \\ \text{trt}_1 \\ \text{age}_1 \end{bmatrix} \mu_1 \begin{bmatrix} 1 & \text{trt}_1 & \text{age}_1 \end{bmatrix} - \begin{bmatrix} 1 \\ \text{trt}_2 \\ \text{age}_2 \end{bmatrix} \mu_2 \begin{bmatrix} 1 & \text{trt}_2 & \text{age}_2 \end{bmatrix} - \cdots$$

$$- \begin{bmatrix} 1 \\ \text{trt}_n \\ \text{age}_n \end{bmatrix} \mu_n \begin{bmatrix} 1 & \text{trt}_n & \text{age}_n \end{bmatrix}$$

$$= - \begin{bmatrix} 1 & 1 & \cdots & 1 \\ \text{trt}_1 & \text{trt}_2 & \cdots & \text{trt}_n \\ \text{age}_1 & \text{age}_2 & \cdots & \text{age}_n \end{bmatrix}_{3 \times n} \begin{bmatrix} \mu_1 & 0 & 0 & 0 \\ 0 & \mu_2 & 0 & 0 \\ 0 & 0 & \ddots & 0 \\ 0 & 0 & 0 & \mu_n \end{bmatrix}_{n \times n} \begin{bmatrix} 1 & \text{trt}_1 & \text{age}_1 \\ 1 & \text{trt}_2 & \text{age}_2 \\ \vdots & \vdots & \vdots \\ 1 & \text{trt}_n & \text{age}_n \end{bmatrix}_{n \times 3}$$

Ex 20-3 : Data: rate.csv

please read it into R and call it “rate” object :

- (1) Please use the Newton-Raphson method to find the maximum likelihood estimate (MLE) of the regression coefficients of Poisson regression ($\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2$), and compare your result with the R built-in function “glm(Death~Age+sex, offset=log(PY/100000), data=rate, family=poisson)”.
- (2) Please find the variance-covariance matrix for $\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2$, and compare your result with the R built-in function vcov(model).
- (3) Please find the log likelihood at $\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2$, and compare your result with the R built-in function logLik(model).

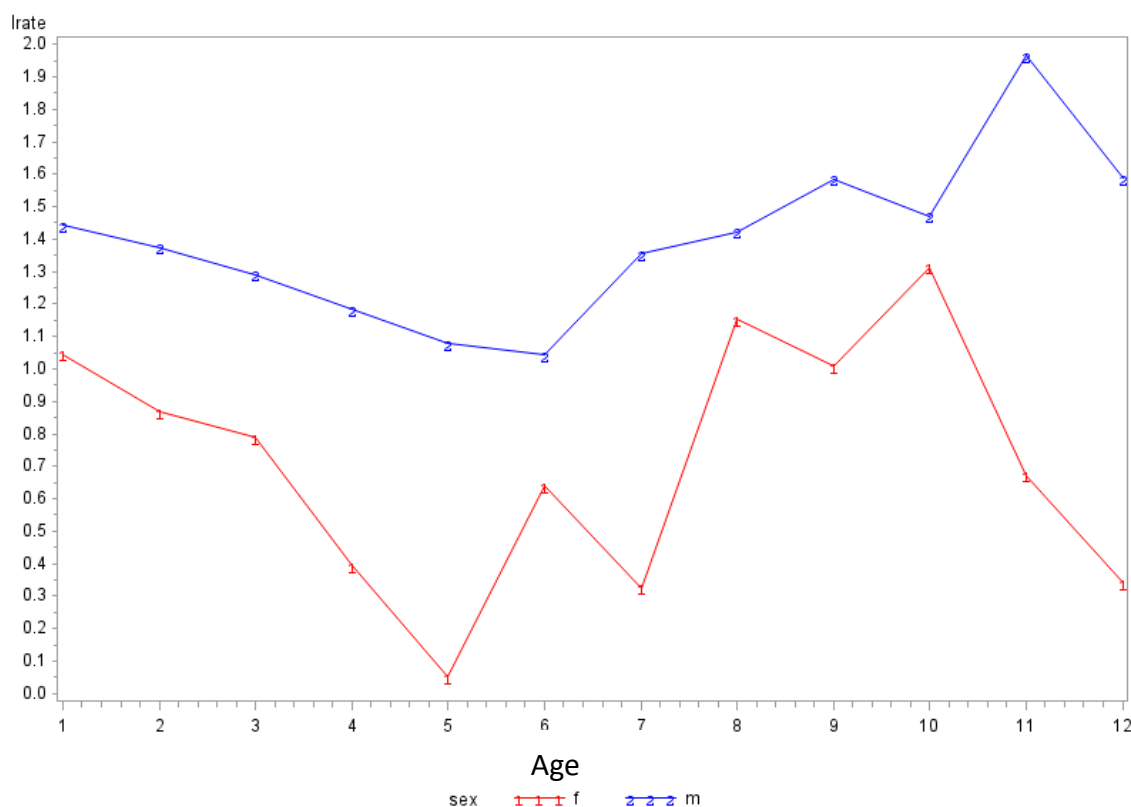
[Hint] **Death rate per 100,000 person years**

$$\log\left(\frac{E(Y_i)}{PY_i / 100000}\right) = \beta_0 + \beta_1 Age_i + \beta_2 sex_i = \begin{bmatrix} 1 & Age_i & sex_i \end{bmatrix} \begin{bmatrix} \beta_0 \\ \beta_1 \\ \beta_2 \end{bmatrix} = \mathbf{x}_i' \boldsymbol{\beta}$$

$$\log(E(Y_i)) - \log(PY_i / 100000) = \mathbf{x}_i' \boldsymbol{\beta}$$

$$\log(E(Y_i)) = \mathbf{x}_i' \boldsymbol{\beta} + \log(PY_i / 100000)$$

Standardized person years



1	30-34
2	35-39
3	40-44
4	45-49
5	50-54
6	55-59
7	60-64
8	65-69
9	70-74
10	75-79
11	80-84
12	85+

Homework: (8 points, please pay attention to all the words in this orange box):

Ex 20-4 : Continue Ex 20-3, because $\log(\text{Death rate per 100,000 person years})$ is not linear in Age, it will be more reasonable to recode Age as 11 dummy variables.

```
rate$Age.f <- factor(rate$Age)
```

- (1) Please use the Newton-Raphson method to find the maximum likelihood estimate (MLE) of the regression coefficients of Poisson regression ($\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2, \dots$) (4 points)

Please note: No need to make a plot for the initial value. Using the R built-in function `"glm(Death~Age.f+sex, offset=log(PY/100000), data=rate, family=poisson)"` to answer this homework will be scored as 0, although you may use it to check your own answer.

- (2) Please find the variance-covariance matrix for $\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2, \dots$ (2 points)

Please note: Using the R built-in function `"vcov(model)"` to answer this homework will be scored as 0, although you may use it to check your own answer.

- (3) Please find the log likelihood at $\hat{\beta}_0, \hat{\beta}_1, \hat{\beta}_2, \dots$ (2 points)

Please note: Using the R built-in function `"logLik(model)"` to answer this homework will be scored as 0, although you may use it to check your own answer.

[Hint] Recode Age as 11 dummy variables

	Age.f2	Age.f3	Age.f4	Age.f5	Age.f6	Age.f7	Age.f8	Age.f9	Age.f10	Age.f11	Age.f12
1	0	0	0	0	0	0	0	0	0	0	0
2	1	0	0	0	0	0	0	0	0	0	0
3	0	1	0	0	0	0	0	0	0	0	0
4	0	0	1	0	0	0	0	0	0	0	0
5	0	0	0	1	0	0	0	0	0	0	0
6	0	0	0	0	1	0	0	0	0	0	0
7	0	0	0	0	0	1	0	0	0	0	0
8	0	0	0	0	0	0	1	0	0	0	0
9	0	0	0	0	0	0	0	1	0	0	0
10	0	0	0	0	0	0	0	0	1	0	0
11	0	0	0	0	0	0	0	0	0	1	0
12	0	0	0	0	0	0	0	0	0	0	1