

106-1 生物統計學二 實習課

R : Poisson Regression

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大綱

- Poisson Regression
 - Fit Poisson model

Poisson Regression

Fit Poisson model

Fit **g**eneralized **l**inear **m**odel

- Fit **Poisson** regression model

不同樣本之間需以相同的基準比較，
用來校正計數資料(count data)

```
model <- glm( Y ~ X1 + X2 + ... + Xp , offset = log(person-year) , family = poisson,  
              data = 資料檔名稱 )
```

- model配適結果的總結

```
summary( model )
```

- 告知R現在要建立的是 poisson regression
 - 也可使用 `family = poisson(link="log")`
- (查詢: `?glm.fit`、`?family`)

Example : Smoke lung data

Goal: 每日抽菸量如何影響肺癌死亡率

* 資料檔 : smokelung.csv (逗號分隔)

Coding Book	
變項名稱	變項描述
years.smoke	a factor giving the age that start smoking
cigarettes	a factor giving cigarette consumption/day
Time	man-years at risk
y	number of deaths

Example : Smoke lung data

Q:每日抽菸量如何影響肺癌死亡率

Variables

Y : y (number of deaths)

X_1 : years.smoke = { 15-19
20-24
25-29
30-34
35-39
40-44
45-49
50-54
55-59

X_2 : cigarettes = { 0
1-9
10-14
15-19
20-24
25-34
35+

Coding book

Reference →

years.smoke	$X_{1(1)}$	$X_{1(2)}$...	$X_{1(8)}$
15-19	0	0	...	0
20-24	1	0	...	0
25-29	0	1	...	0
⋮	⋮	⋮	⋱	⋮
55-59	0	0	...	1

Reference →

cigarettes	$X_{2(1)}$	$X_{2(2)}$...	$X_{2(6)}$
0	0	0	...	0
1-9	1	0	...	0
10-14	0	1	...	0
⋮	⋮	⋮	⋱	⋮
35+	0	0	...	1

利用 **contrasts**(變項名稱) 查看

Example : Smoke lung data

Model

offset term

$$E[Y|X] = m \cdot \lambda_X$$

$$\begin{cases} Y|X \sim \text{Poisson}(m \cdot \lambda_X) \\ \log(\lambda_X) = \beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{1(8)}X_{1(8)} + \beta_{2(1)}X_{2(1)} + \cdots + \beta_{2(6)}X_{2(6)} \end{cases}$$

$$\Rightarrow \lambda_X = e^{\beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{1(8)}X_{1(8)} + \beta_{2(1)}X_{2(1)} + \cdots + \beta_{2(6)}X_{2(6)}}$$

		cigarettes																	
		0		1-9		10-14		15-19		20-24		25-34		35+					
years.smoke	y	Time	y	Time	y	Time	y	Time	y	Time	y	Time	y	Time	y	Time			
15-19	1	10366	0	3121	0	3577	0	4317	0	5683	0	3042	0	670					
20-24	0	8162	0	2937	1	3286	0	4214	1	6385	1	4050	0	1166					
25-29	0	5969	0	2288	1	2546	0	3185	1	5483	4	4290	0	1482					
30-34	0	4496	0	2015	2	2219	4	2560	6	4687	9	4268	4	1580					
35-39	0	3512	1	1648	0	1826	0	1893	5	3646	9	3529	6	1336					
40-44	0	2201	2	1310	1	1386	2	1334	12	2411	11	2424	10	924					
45-49	0	1421	0	927	2	988	2	849	9	1567	10	1409	7	556					
50-54	0	1121	3	710	4	684	2	470	7	857	5	663	4	255					
55-59	2	826	0	606	3	449	5	280	7	416	3	284	1	104					

Example : Smoke lung data

```
> model.1 <- glm(y~ years.smoke + cigarettes, offset = log(Time), family = poisson, data = smokelung_data )  
> summary(model.1)
```

Call:

```
glm(formula = y ~ years.smoke + cigarettes, family = poisson,  
     data = smokelung_data, offset = log(Time))
```

Deviance Residuals:

Min	1Q	Median	3Q	Max
-1.8329	-0.8560	-0.3808	0.4241	2.1762

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)	
(Intercept)	-12.5784	1.1475	-10.961	< 2e-16	***
years.smoke20-24	0.9469	1.1548	0.820	0.412202	
years.smoke25-29	1.7016	1.0805	1.575	0.115284	
years.smoke30-34	3.2029	1.0204	3.139	0.001695	**
years.smoke35-39	3.2423	1.0242	3.166	0.001547	**
years.smoke40-44	4.2088	1.0137	4.152	3.30e-05	***
years.smoke45-49	4.4476	1.0171	4.373	1.23e-05	***
years.smoke50-54	4.9048	1.0201	4.808	1.52e-06	***
years.smoke55-59	5.4134	1.0239	5.287	1.24e-07	***
cigarettes1-9	1.2200	0.7073	1.725	0.084547	.
cigarettes10-14	2.0991	0.6363	3.299	0.000971	***
cigarettes15-19	2.3089	0.6327	3.649	0.000263	***
cigarettes20-24	2.9009	0.5956	4.870	1.11e-06	***
cigarettes25-34	3.1162	0.5947	5.240	1.61e-07	***
cigarettes35+	3.6059	0.6048	5.962	2.49e-09	***

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

(Dispersion parameter for poisson family taken to be 1)

Null deviance: 445.099 on 62 degrees of freedom
Residual deviance: 51.471 on 48 degrees of freedom
AIC: 201.31

Number of Fisher Scoring iterations: 6

在相同years.smoke下 (調整初次抽菸年齡的影響後),
每日抽35根菸以上者的肺癌死亡率為沒有抽菸者的
 $e^{3.6059} = 36.815$ 倍

$$H_0: \beta_{2(6)} = 0 \text{ vs. } H_1: \beta_{2(6)} \neq 0$$
$$Z = \frac{3.6059}{0.6048} = 5.962$$
$$p\text{-value} = 2.49 \times 10^{-9} < \alpha = 0.05$$

Meaning of $\beta_{2(6)}$

$$\log(\lambda_X) = \beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{1(8)}X_{1(8)} + \beta_{2(1)}X_{2(1)} + \cdots + \beta_{2(6)}X_{2(6)}$$

$$\Rightarrow \lambda_X = e^{\beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{1(8)}X_{1(8)} + \beta_{2(1)}X_{2(1)} + \cdots + \beta_{2(6)}X_{2(6)}}$$

While controlling $X_{1(1)}, \dots, X_{1(8)}$ & $X_{2(1)}, \dots, X_{2(5)}$,

➤ $X_{2(6)} = 1$

$$\Rightarrow E[Y|X_{1(1)}, \dots, X_{1(8)}, X_{2(1)}, \dots, X_{2(5)}, X_{2(6)} = 1] = m \cdot e^{\beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{1(8)}X_{1(8)} + \beta_{2(1)}X_{2(1)} + \cdots + \beta_{2(6)} \times (1)}$$

➤ $X_{2(6)} = 0$

$$\Rightarrow E[Y|X_{1(1)}, \dots, X_{1(8)}, X_{2(1)}, \dots, X_{2(5)}, X_{2(6)} = 0] = m \cdot e^{\beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{1(8)}X_{1(8)} + \beta_{2(1)}X_{2(1)} + \cdots + \beta_{2(6)} \times (0)}$$

$$\Rightarrow \text{Rate Ratio (RR) of } X_{2(6)} = \frac{m \cdot e^{\beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{2(6)} \times (1)}}{m \cdot e^{\beta_0 + \beta_{1(1)}X_{1(1)} + \cdots + \beta_{2(6)} \times (0)}} = e^{\beta_{2(6)}}$$

調整其他變項的影響後，
 $X_{2(6)} = 1$ 的RR為 $X_{2(6)} = 0$ 的 $e^{\beta_{2(6)}}$ 倍

Homework

example4：抽菸習慣與死亡率的關係

*請先匯入資料檔(需附上code)： example4.csv (逗號分隔)

- 利用R進行 Poisson regression，判斷調整年齡組別變項後，抽菸習慣和死亡率的關係：
 - 請寫下建立的迴歸模型(定義清楚符號代表的意思)
 - 請問抽菸習慣是否會影響死亡率?(請說明根據什麼結果判斷)
 - 校正年齡組別後，Smoke的Rate Ratio為?
 - 對於有抽菸習慣且年齡大於55歲的人，估計的死亡率為? ($\hat{\lambda}_x = ?$)

Coding Book	
變項名稱	變項描述
Agegroup	年齡組別 (1:<25歲/ 2:25-35歲 /3:35-45歲/ 4:45-55歲/ 5:>55歲)
Smoke	有無抽菸習慣(0: 沒有抽菸; 1:有抽菸習慣)
Death	死亡數
py	觀察到的人年(person-year)