

Poisson Regression model Demo

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資料簡介 **DoctorVisits Dataset from package AER**

```
library(AER)
library(Hmisc)
library(ggplot2)
library(DataExplorer)
data(DoctorVisits)
str(DoctorVisits)
```

```
'data.frame':  5190 obs. of  12 variables:
 $ visits   : num  1 1 1 1 1 1 1 1 1 1 ...
 $ gender   : Factor w/ 2 levels "male","female": 2 2 1 1 1 2 2 2 1 ...
 $ age      : num  0.19 0.19 0.19 0.19 0.19 0.19 0.19 0.19 0.19 0.19 ...
 $ income   : num  0.55 0.45 0.9 0.15 0.45 0.35 0.55 0.15 0.65 0.15 ...
 $ illness  : num  1 1 3 1 2 5 4 3 2 1 ...
 $ reduced  : num  4 2 0 0 5 1 0 0 0 0 ...
 $ health   : num  1 1 0 0 1 9 2 6 5 0 ...
 $ private  : Factor w/ 2 levels "no","yes": 2 2 1 1 1 1 1 1 2 2 ...
 $ freepoor : Factor w/ 2 levels "no","yes": 1 1 1 1 1 1 1 1 1 1 ...
 $ freerepat: Factor w/ 2 levels "no","yes": 1 1 1 1 1 1 1 1 1 1 ...
 $ nchronic : Factor w/ 2 levels "no","yes": 1 1 1 1 2 2 1 1 1 1 ...
 $ lchronic : Factor w/ 2 levels "no","yes": 1 1 1 1 1 1 1 1 1 1 ...
```

```
sum(DoctorVisits$private=="yes" & DoctorVisits$freepoor=="yes" & DoctorVisits$freerepat=="yes")
```

```
[1] 0
```

```
sum(DoctorVisits$private=="yes" & DoctorVisits$freepoor=="yes")
```

```
[1] 0
```

```
sum(DoctorVisits$private=="yes" & DoctorVisits$freerepat=="yes")
```

```
[1] 0
```

```
sum(DoctorVisits$freepoor=="yes" & DoctorVisits$freerepat=="yes")
```

```
[1] 0
```

```
DoctorVisits$insurance <- as.factor(apply(DoctorVisits[,8:10], MARGIN = 1, function(row){
  return(ifelse(row[1]=="yes", "P",ifelse(row[2]=="yes", "GP", ifelse(row[3]=="yes", "GR", "N"))))
}))
DoctorVisits <- DoctorVisits[,c(1:7,13,11,12)]
sum(DoctorVisits$nchronic=="yes" & DoctorVisits$lchronic=="yes")
```

```
[1] 0
```

```
sum(DoctorVisits$lchronic=="yes")
```

```
[1] 605
```

```
sum(DoctorVisits$nchronic=="yes")
```

```
[1] 2092
```

```
sum(DoctorVisits$nchronic=="no" & DoctorVisits$lchronic=="no")
```

```
[1] 2493
```

```
DoctorVisits$chronDis <- as.factor(apply(DoctorVisits[,9:10], MARGIN = 1, function(row){
  return(ifelse(row[1]=="yes", "nch",ifelse(row[2]=="yes", "lch", "N")))
}))
DoctorVisits$isfemale <- ifelse(DoctorVisits$gender=="female",1,0)
DoctorVisits <- DoctorVisits[,c(1,12,3:8,11)]
head(DoctorVisits)
```

	visits	isfemale	age	income	illness	reduced	health	insurance	chronDis
1	1	1	0.19	0.55	1	4	1	P	N
2	1	1	0.19	0.45	1	2	1	P	N
3	1	0	0.19	0.90	3	0	0	N	N
4	1	0	0.19	0.15	1	0	0	N	N
5	1	0	0.19	0.45	2	5	1	N	nch
6	1	1	0.19	0.35	5	1	9	N	nch

將 private, freepoor, freerepat 三個變數合併成 insurance 類別變數: P=private, GP=freepoor, GR=freerepat, N= 沒有保險

nchronic,lchronic 合併成 chronDis 類別變數: nch= 有慢性疾病但不限制行動, lch= 有慢性疾病並且會限制行動, N= 沒有慢性疾病

Table 1: 變數解釋

變數	解釋	資料格式	備註
visits	過去兩週的看醫生 (諮詢) 的次數	num	counts:0~9

變數	解釋	資料格式	備註
isfemale	性別	num	1,0
age	年齡	num	years/100:0.19~0.72
income	年收入 (in 10,000 dollars)	num	income/10000:0.0~1.5
illness	過去兩週不舒服的次數	num	counts:0~5
reduced	過去兩週因生病或受傷的休養天數	num	counts:0~14
health	GHQ-12 心理健康問卷分數	num	mentally (healthy)0~12(unhealthy)
insurance	醫療保險種類	factor	P: 私人, GP: 政府低收, GR: 政府高齡與其他, N: 沒有保險
chronDis	慢性疾病種類	factor	nch: 不限制行動, lch: 限制行動, N: 沒有慢性疾病

```
latex(describe(DoctorVisits),title="",file="")
```

DoctorVisits 9 Variables 5190 Observations													
visits													
n	missing	distinct	Info	Mean	Gmd	.05	.10	.25	.50	.75	.90	.95	
5190	0	10	0.489	0.3017	0.5154	0	0	0	0	0	1	2	
Value	0	1	2	3	4	5	6	7	8	9			
Frequency	4141	782	174	30	24	9	12	12	5	1			
Proportion	0.798	0.151	0.034	0.006	0.005	0.002	0.002	0.002	0.001	0.000			
For the frequency table, variable is rounded to the nearest 0													
isfemale													
n	missing	distinct	Info	Sum	Mean	Gmd							
5190	0	2	0.749	2702	0.5206	0.4992							
age													
n	missing	distinct	Info	Mean	Gmd	.05	.10	.25	.50	.75	.90	.95	
5190	0	12	0.978	0.4064	0.2258	0.19	0.19	0.22	0.32	0.62	0.72	0.72	
Value	0.19	0.22	0.27	0.32	0.37	0.42	0.47	0.52	0.57	0.62	0.67	0.72	
Frequency	752	1213	523	301	146	126	181	222	273	316	315	822	
Proportion	0.145	0.234	0.101	0.058	0.028	0.024	0.035	0.043	0.053	0.061	0.061	0.158	
For the frequency table, variable is rounded to the nearest 0													
income													
n	missing	distinct	Info	Mean	Gmd	.05	.10	.25	.50	.75	.90	.95	
5190	0	14	0.983	0.5832	0.4085	0.15	0.25	0.25	0.55	0.90	1.10	1.30	
Value	0.00	0.01	0.06	0.15	0.25	0.35	0.45	0.55	0.65	0.75	0.90	1.10	1.30
Frequency	79	35	80	249	1195	462	400	467	455	441	589	361	162
Proportion	0.015	0.007	0.015	0.048	0.230	0.089	0.077	0.090	0.088	0.085	0.113	0.070	0.031
For the frequency table, variable is rounded to the nearest 0													
illness													
n	missing	distinct	Info	Mean	Gmd								
5190	0	6	0.934	1.432	1.481								
Value	0	1	2	3	4	5							
Frequency	1554	1638	946	542	274	236							
Proportion	0.299	0.316	0.182	0.104	0.053	0.045							
For the frequency table, variable is rounded to the nearest 0													

reduced

	n	missing	distinct	Info	Mean	Gmd	.05	.10	.25	.50	.75	.90	.95
5190		0	15	0.368	0.8618	1.592	0	0	0	0	0	2	7

Value	0	1	2	3	4	5	6	7	8	9	10	11	12	13
Frequency	4454	177	108	74	45	40	17	38	17	7	12	2	6	5
Proportion	0.858	0.034	0.021	0.014	0.009	0.008	0.003	0.007	0.003	0.001	0.002	0.000	0.001	0.001

Value	14
Frequency	188
Proportion	0.036

For the frequency table, variable is rounded to the nearest 0

health

	n	missing	distinct	Info	Mean	Gmd	.05	.10	.25	.50	.75	.90	.95
5190		0	13	0.797	1.218	1.84	0	0	0	0	2	4	6

Value	0	1	2	3	4	5	6	7	8	9	10	11	12
Frequency	3026	823	446	273	187	132	104	61	42	32	21	24	19
Proportion	0.583	0.159	0.086	0.053	0.036	0.025	0.020	0.012	0.008	0.006	0.004	0.005	0.004

For the frequency table, variable is rounded to the nearest 0

insurance

	n	missing	distinct
5190		0	4

Value	GP	GR	N	P
Frequency	222	1091	1579	2298
Proportion	0.043	0.210	0.304	0.443

chronDis

	n	missing	distinct
5190		0	3

Value	lch	N	nch
Frequency	605	2493	2092
Proportion	0.117	0.480	0.403

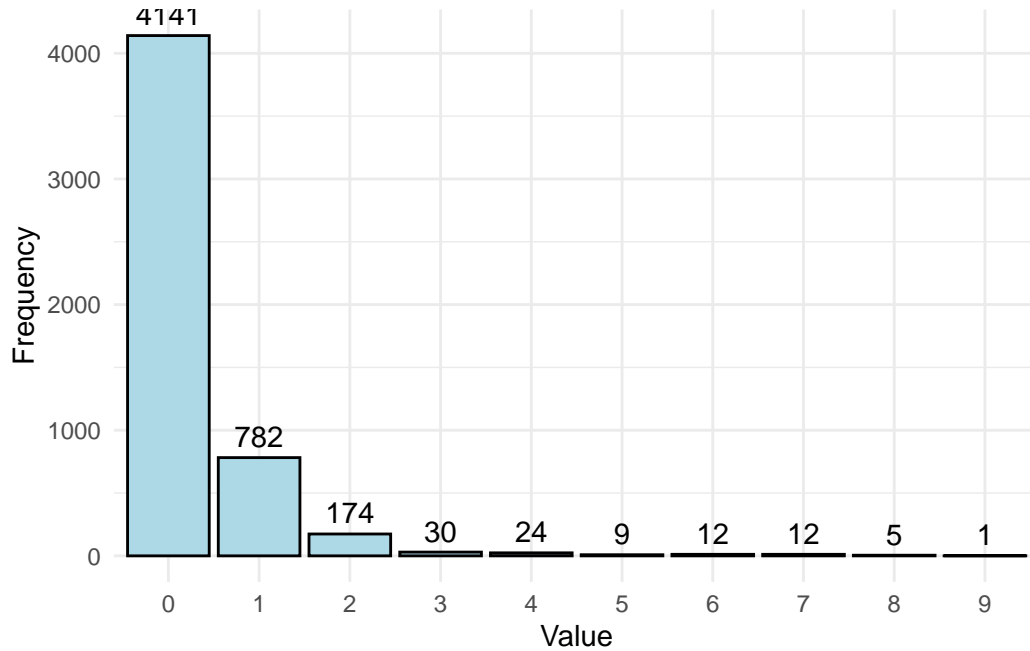
```
library(ggplot2)
mytable <- data.frame(table(DoctorVisits$visits))
ggplot(mytable, aes(x = factor(Var1), y = Freq)) +
  geom_bar(stat = "identity", fill = "lightblue", color = "black") +
  geom_text(aes(label = Freq), vjust = -0.5, hjust = 0.5) + # 調整 vjust 和 hjust
  labs(x = "Value", y = "Frequency") +
  theme_minimal()
```

建構模型

```
#build interaction terms
mydata <- model.matrix(~.*-1, data = DoctorVisits[, -1])
mydata <- data.frame(visits = DoctorVisits$visits, mydata)
fullmodel <- glm(visits~., data=mydata, family = poisson())
nullmodel <- glm(visits~1, data=mydata, family = poisson())
fit.step <- step(
  nullmodel,
  scope = list(lower = nullmodel, upper=fullmodel),
  direction = "both", k = log(5190), trace = FALSE)
cat("The number of variables selected is:", length(fit.step$coefficients)-1)
```

The number of variables selected is: 9

```
summary(fit.step)
```



Call:

```
glm(formula = visits ~ reduced + illness + illness.reduced +
    age + age.reduced + age.health + income.insuranceN + insuranceGP +
    reduced.health, family = poisson(), data = mydata)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-2.246650	0.087223	-25.757	< 2e-16 ***
reduced	0.248641	0.012224	20.341	< 2e-16 ***
illness	0.247100	0.021081	11.722	< 2e-16 ***
illness.reduced	-0.013566	0.002924	-4.639	3.50e-06 ***
age	0.693581	0.161334	4.299	1.72e-05 ***
age.reduced	-0.138728	0.021995	-6.307	2.84e-10 ***
age.health	0.126539	0.024041	5.264	1.41e-07 ***
income.insuranceN	-0.379462	0.094568	-4.013	6.01e-05 ***
insuranceGP	-0.586419	0.174576	-3.359	0.000782 ***
reduced.health	-0.005062	0.001455	-3.480	0.000501 ***

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

(Dispersion parameter for poisson family taken to be 1)

Null deviance: 5634.8 on 5189 degrees of freedom
 Residual deviance: 4294.3 on 5180 degrees of freedom
 AIC: 6645.9

Number of Fisher Scoring iterations: 6

按照慣例做法，將未選中的主效應項加入模型：

發現加入 health 會使得 age.health 與 health 不顯著，因此傾向保留主效應。

Table 2: fitted model

	Coefficients	Std Error	Z value	P value	Significance
(Intercept)	-2.349	0.088	-26.704	0	***
illness	0.242	0.021	11.296	0	***
insuranceGP	-0.610	0.175	-3.489	0	***
age	0.904	0.158	5.706	0	***
reduced	0.242	0.012	19.657	0	***
health	0.069	0.013	5.282	0	***
illness.reduced	-0.013	0.003	-4.486	0	***
age.reduced	-0.124	0.022	-5.745	0	***
income.insuranceN	-0.377	0.094	-3.988	0	***
reduced.health	-0.005	0.001	-3.653	0	***

```

oldformula <- fit.step$formula
newformula <- update(oldformula, ~.-age.health+health)
fit.step <- glm(newformula, data = mydata, family = poisson())
output <- summary(fit.step)$coefficients
output <- data.frame(
  coef = round(output[,1],3),
  sd = round(output[,2],3),
  z = round(output[,3],3),
  pvalue = round(output[,4],3)
)
output$sig <- sapply(output$pvalue, function(p) {
  ifelse(p < 0.001,"***",ifelse(p < 0.01,"**",ifelse(p < 0.05,"*",)))
})
colnames(output) <- c("Coefficients", "Std Error", "Z value", "P value", "Significance")
output <- output[c(1,3,8,5,2,10,4,6,7,9),]
latex(output, title="",file="",caption = "fitted model")

```

```

x <- seq(0, 0.8, by = 0.001)
curve(expr = exp(0.567 - 0.856 * x), from = 0, to = 0.8,
  xlab = "Age", ylab = "ratio",
  col = "blue", lwd = 2, main = expression(mu[female] / mu[male]))
grid()
abline(h=1, col="red")
points(x = 0.567/0.856, y = 1, col="black")
text(x = 0.567/0.856, y = 1.1,paste0("x is ",round(0.567/0.856,2)))

```

check for overdispersion

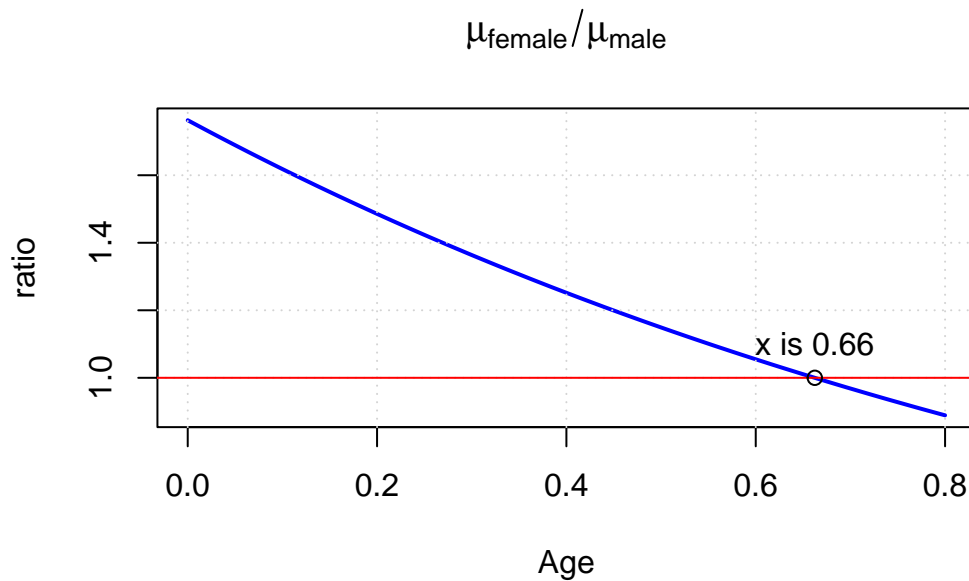
estimation of dispersion parameter

```

# Deviance Method
dispersion_deviance <- sum(resid(fit.step, type = "deviance")^2)/fit.step$df.residual
cat("estimated dispersion by deviance:", dispersion_deviance)

```

estimated dispersion by deviance: 0.8289777



```
# Pearson Method
dispersion_pearson <- sum(resid(fit.step, type = "pearson")^2)/fit.step$df.residual
cat("estimated dispersion by pearson:", dispersion_pearson)
```

estimated dispersion by pearson: 1.323534

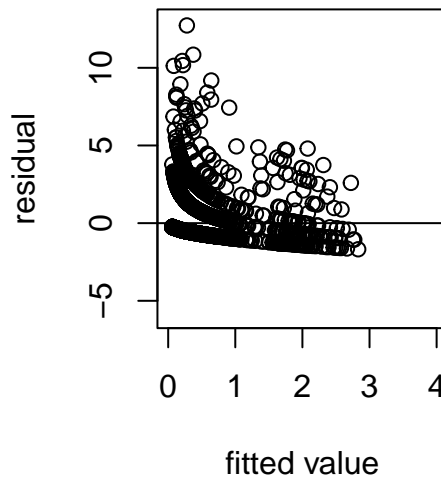
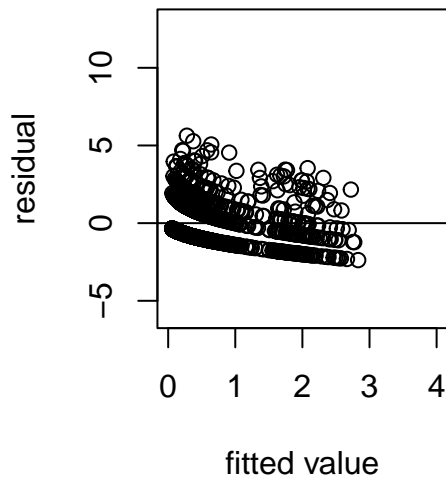
residual analysis

```
# 檢查 pearson residual 的變異數是否有等於 1
cat("Variance of Pearson residual:", var(resid(fit.step, type = "pearson")))
```

Variance of Pearson residual: 1.321134

```
# 殘差圖 (記得依 fitted value 大小排序)
par(mfrow = c(1, 2))
plot(sort(fit.step$fitted.values),
      resid(fit.step, type = "deviance")[order(fit.step$fitted.values, decreasing = FALSE)],
      main="Fig 2.1 Deviance residual plot", ylab = "residual", xlab="fitted value",
      ylim = c(-6,13), xlim=c(0,4))
abline(0,0)
plot(sort(fit.step$fitted.values),
      resid(fit.step, type = "pearson")[order(fit.step$fitted.values, decreasing = FALSE)],
      main="Fig 2.2 Pearson residual plot", ylab = "residual", xlab="fitted value",
      ylim = c(-6,13), xlim=c(0,4))
abline(0,0)
```

Fig 2.1 Deviance residual plot **Fig 2.2 Pearson residual plot**



dispersion test

```
dispersiontest(fit.step,alternative = "greater",
               trafo = 1)
```

Overdispersion test

```
data: fit.step
z = 6.5513, p-value = 2.852e-11
alternative hypothesis: true alpha is greater than 0
sample estimates:
alpha
0.3910117
```

```
dispersiontest(fit.step,alternative = "greater",
               trafo = function(mu) mu)
```

Overdispersion test

```
data: fit.step
z = 6.5513, p-value = 2.852e-11
alternative hypothesis: true alpha is greater than 0
sample estimates:
alpha
0.3910117
```

```
dispersiontest(fit.step,alternative = "greater",
               trafo = 2)
```

Overdispersion test


```
data: fit.step
z = 7.9487, p-value = 9.427e-16
alternative hypothesis: true alpha is greater than 0
sample estimates:
  alpha
0.9895823
```

```
dispersiontest(fit.step, alternative = "greater",
               trafo = function(mu) mu^2)
```

Overdispersion test

```
data: fit.step
z = 7.9487, p-value = 9.427e-16
alternative hypothesis: true alpha is greater than 0
sample estimates:
  alpha
0.9895823
```

```
dispersiontest(fit.step, alternative = "greater")
```

Overdispersion test

```
data: fit.step
z = 6.5513, p-value = 2.852e-11
alternative hypothesis: true dispersion is greater than 1
sample estimates:
dispersion
 1.391012
```

```
dispersiontest(fit.step, alternative = "two.sided")
```

Dispersion test

```
data: fit.step
z = 6.5513, p-value = 5.704e-11
alternative hypothesis: true dispersion is not equal to 1
sample estimates:
dispersion
 1.391012
```

```
dispersiontest(fit.step, alternative = "less")
```

Underdispersion test

```
data: fit.step
z = 6.5513, p-value = 1
alternative hypothesis: true dispersion is less than 1
sample estimates:
dispersion
 1.391012
```