

Example 3 Explained

Yu-Chen Xue

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Example 3 – UCBA Admission Explained

1. UCBA Admissions 資料集的 Overview

UCBA Admissions 是儲存為 3D array 格式的資料集, 這裡先將 UCBA Admissions 轉換為 dataframe, 並命名為 `ucb_df`, 以直觀地展示這個資料集

```
ucb_df <- data.frame(UCBA Admissions)
ucb_df
```

```
##      Admit Gender Dept Freq
## 1  Admitted   Male    A  512
## 2  Rejected   Male    A  313
## 3  Admitted Female    A   89
## 4  Rejected Female    A   19
## 5  Admitted   Male    B  353
## 6  Rejected   Male    B  207
## 7  Admitted Female    B   17
## 8  Rejected Female    B    8
## 9  Admitted   Male    C  120
## 10 Rejected   Male    C  205
## 11 Admitted Female    C  202
## 12 Rejected Female    C  391
## 13 Admitted   Male    D  138
## 14 Rejected   Male    D  279
## 15 Admitted Female    D  131
## 16 Rejected Female    D  244
## 17 Admitted   Male    E   53
## 18 Rejected   Male    E  138
## 19 Admitted Female    E   94
## 20 Rejected Female    E  299
## 21 Admitted   Male    F   22
## 22 Rejected   Male    F  351
```

```
## 23 Admitted Female      F    24
## 24 Rejected Female      F   317
```

製作 UCBAmissions 資料集的 ‘flat’ contingency tables, 並指定欄 “Admit” 為比較的目標欄

```
ftable(UCBAmissions, col.vars="Admit")
```

```
##           Admit Admitted Rejected
## Gender Dept
## Male  A           512       313
##       B           353       207
##       C           120       205
##       D           138       279
##       E            53       138
##       F            22       351
## Female A            89        19
##       B            17         8
##       C           202       391
##       D           131       244
##       E            94       299
##       F            24       317
```

顯示 UCBAmissions 資料集中解釋變數的組成

```
dimnames(UCBAmissions)
```

```
## $Admit
## [1] "Admitted" "Rejected"
##
## $Gender
## [1] "Male"     "Female"
##
## $Dept
## [1] "A" "B" "C" "D" "E" "F"
```

顯示 UCBAmissions 資料集的 margin table (一個顯示某個欄位的各個數值的個數的表格), 這裡選取 UCBAmissions 的第 1 個解釋變數 (Admit) 相對第 2 個解釋變數 (Gender) 的個數

```
margin.table(UCBAdmissions, c(2,1))
```

```
##           Admit
## Gender   Admitted Rejected
##   Male      1198      1493
##   Female     557      1278
```

顯示 UCBAdmissions 資料集的 margin table, 這裡選取 UCBAdmissions 的第 1 個解釋變數 (Admit) 相對第 3 個解釋變數 (Dept) 的個數

```
margin.table(UCBAdmissions, c(3,1))
```

```
##           Admit
## Dept   Admitted Rejected
##   A         601      332
##   B         370      215
##   C         322      596
##   D         269      523
##   E         147      437
##   F          46      668
```

顯示 UCBAdmissions 資料集的 margin table, 這裡選取 UCBAdmissions 的第 3 個解釋變數 (Dept) 相對第 2 個解釋變數 (Gender) 的個數

```
margin.table(UCBAdmissions, c(2,3))
```

```
##           Dept
## Gender      A   B   C   D   E   F
##   Male    825 560 325 417 191 373
##   Female  108  25 593 375 393 341
```

2. 對 UCBAdmissions 資料集進行建模、分析

以另一種方法展現這個資料集, 將 Admit 拆分為 yes 和 no 兩種情況

```
### begin copying here
ucb.df = data.frame(gender=rep(c("Male", "Female"), c(6,6)),
                    dept=rep(LETTERS[1:6], 2),
                    yes=c(512, 353, 120, 138, 53, 22, 89, 17, 202, 131, 94, 24),
```

```

no=c(313,207,205,279,138,351,19,8,391,244,299,317))
### end copying here and paste into the R Console

ucb.df

```

```

##   gender dept yes  no
## 1   Male    A 512 313
## 2   Male    B 353 207
## 3   Male    C 120 205
## 4   Male    D 138 279
## 5   Male    E  53 138
## 6   Male    F  22 351
## 7 Female    A  89  19
## 8 Female    B  17   8
## 9 Female    C 202 391
## 10 Female   D 131 244
## 11 Female   E  94 299
## 12 Female   F  24 317

```

將 yes/no 作為響應變數，Gender * Dept 作為解釋變數，對 UCBA Admission 資料集構建廣義線性模型，其中 family=binomial(logit) 表示指定使用邏輯回歸。

```

mod.form = "cbind(yes,no) ~ gender * dept"      # mind the quotes here!
glm.out = glm(mod.form, family=binomial(logit), data=ucb.df)

```

對模型作 anova 表格，test="Chisq" 表示在輸出的表格中加上 Pr(>Chi) 一欄

可以看到 "gender" 和 "dept" 對學生是否被錄取都很有關係

```

anova(glm.out, test="Chisq")

```

```

## Analysis of Deviance Table
##
## Model: binomial, link: logit
##
## Response: cbind(yes, no)
##
## Terms added sequentially (first to last)
##

```

```
##
##              Df Deviance Resid. Df Resid. Dev  Pr(>Chi)
## NULL                      11      877.06
## gender          1      93.45          10      783.61 < 2.2e-16 ***
## dept            5     763.40           5       20.20 < 2.2e-16 ***
## gender:dept     5      20.20           0        0.00 0.001144 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

對模型作 summary 表格

其中 z-value 為 Critical value, “genderMale:deptB” 的期望值表示若同時為 genderMale 和 deptB, 則計算這種情況的 $\ln(\text{odds})$ 的式子需要額外考慮這一項, 即 $1.5442 - 1.0521 - 0.7904 + 0.8321$ “Residual deviance: 1.0791e-13 on 0 degrees of freedom” 說明該模型很完善地解釋了這個資料集

```
summary(glm.out)
```

```
##
## Call:
## glm(formula = mod.form, family = binomial(logit), data = ucb.df)
##
## Deviance Residuals:
##  [1]  0  0  0  0  0  0  0  0  0  0  0  0  0
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)      1.5442     0.2527   6.110 9.94e-10 ***
## genderMale       -1.0521     0.2627  -4.005 6.21e-05 ***
## deptB            -0.7904     0.4977  -1.588  0.11224
## deptC            -2.2046     0.2672  -8.252 < 2e-16 ***
## deptD            -2.1662     0.2750  -7.878 3.32e-15 ***
## deptE            -2.7013     0.2790  -9.682 < 2e-16 ***
## deptF            -4.1250     0.3297 -12.512 < 2e-16 ***
## genderMale:deptB  0.8321     0.5104   1.630  0.10306
## genderMale:deptC  1.1770     0.2996   3.929 8.53e-05 ***
## genderMale:deptD  0.9701     0.3026   3.206  0.00135 **
## genderMale:deptE  1.2523     0.3303   3.791  0.00015 ***
## genderMale:deptF  0.8632     0.4027   2.144  0.03206 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

```
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 8.7706e+02  on 11  degrees of freedom
## Residual deviance: 1.0791e-13  on  0  degrees of freedom
## AIC: 92.94
##
## Number of Fisher Scoring iterations: 3
```

計算 genderMale 的期望值的 exponential

```
exp(-1.0521)
```

```
## [1] 0.3492037
```

計算 genderMale 的期望值的 exponential 的相反數

```
1/exp(-1.0521)
```

```
## [1] 2.863658
```

計算 deptC 的期望值的 exponential

```
exp(-2.2046)
```

```
## [1] 0.1102946
```

將上面的兩個 exponential 數值相除，得到 genderMale:deptD 的近似值

```
exp(-2.2046) / exp(-2.1662)      # C:A / D:A leaves C:D
```

```
## [1] 0.9623279
```

將 yes/no 作為響應變數，Gender * Dept 作為解釋變數，對 UCBA Admission 資料集構建邏輯回歸模型，並對該模型作 anova 表格進行分析

在這個模型下，“gender”對結果沒有什麼貢獻

```
mod.form="cbind(yes,no) ~ dept + gender"
glm.out=glm(mod.form, family=binomial(logit), data=ucb.df)
anova(glm.out, test="Chisq")
```

```
## Analysis of Deviance Table
##
## Model: binomial, link: logit
##
## Response: cbind(yes, no)
##
## Terms added sequentially (first to last)
##
##
##          Df Deviance Resid. Df Resid. Dev Pr(>Chi)
## NULL                                11      877.06
## dept      5   855.32          6       21.74  <2e-16 ***
## gender    1     1.53          5       20.20   0.2159
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
```

對模型作 `summary` 表格進行分析

“genderMale” 對結果沒什麼貢獻 “Residual deviance: 20.204 on 5 degrees of freedom” 說明這個模型的解釋能力不如前面的模型

```
summary(glm.out)
```

```
##
## Call:
## glm(formula = mod.form, family = binomial(logit), data = ucb.df)
##
## Deviance Residuals:
##      1       2       3       4       5       6       7       8
## -1.2487 -0.0560  1.2533  0.0826  1.2205 -0.2076  3.7189  0.2706
##      9      10      11      12
## -0.9243 -0.0858 -0.8509  0.2052
##
## Coefficients:
##              Estimate Std. Error z value Pr(>|z|)
## (Intercept)  0.68192    0.09911   6.880 5.97e-12 ***
```

```

## deptB      -0.04340    0.10984  -0.395    0.693
## deptC      -1.26260    0.10663 -11.841 < 2e-16 ***
## deptD      -1.29461    0.10582 -12.234 < 2e-16 ***
## deptE      -1.73931    0.12611 -13.792 < 2e-16 ***
## deptF      -3.30648    0.16998 -19.452 < 2e-16 ***
## genderMale -0.09987    0.08085  -1.235    0.217
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## (Dispersion parameter for binomial family taken to be 1)
##
##      Null deviance: 877.056  on 11  degrees of freedom
## Residual deviance:  20.204  on  5  degrees of freedom
## AIC: 103.14
##
## Number of Fisher Scoring iterations: 4

```