

4/13/2010 36-402/608 ADA-II H. Seltman
Breakout #22: Poisson Regression

In R, Poisson regression is performed using

```
result = glm(y ~ x..., data=my.dtf, family=poisson)
```

where "y" is a count, and "x..." is any prediction formula.

As usual `summary(result)` has the standard errors and p-values, as well as AIC (as `$aic`).

The `glm()` object has a `$deviance` component that can be used for the likelihood ratio test. E.g., to compare `glm` objects named "full" and "reduced" use:

```
p.val = 1 - pchisq(reduced$deviance - full$deviance, reduced$df.res - full$df.res)
```

Sometimes a reasonable alternative to Poisson regression is linear regression on a transformed outcome in the form of square root of counts. If the residual plots look OK, you can go with that model.

Use `family=quasipoisson` to check for under/over dispersion.

The analysis shown here is from problem 24 of chapter 22 of The Sleuth and represents valve characteristics and number of failures from a nuclear reactor. See the factor coding statements to get some idea of the valve characteristics. The number of failures is modeled as Poisson.

It is important to separate Poisson regression problems into those where the different units studied have equal exposure (in time or space) vs. those with unequal exposure. The latter can only be modeled if the extent of exposure is recorded also.

E.g., if $\log(\mu_i|x_i) = \beta_0 + \beta_1 x_i$ for "unit" exposure, then for exposure t_i we expect $\log(\mu_i/t_i|x_i) = \beta_0 + \beta_1 x_i$ which implies $\log(\mu_i) - \log(t_i) = \beta_0 + \beta_1 x_i$ which implies $\log(\mu_i) = \beta_0 + \beta_1 x_i + 1.0 * \log(t_i)$. In other words we can use the usual Poisson regression model if we include `log(exposure)` as an explanatory variable with a fixed, known coefficient of 1.0. This is done in R (and other programs) by setting the "offset" to `log(exposure)`.

```
valve=read.csv("ex2224.csv")
names(valve)=casemap(names(valve))
summary(valve)
#       system          operator        valve         size
# Min.   :1.000   Min.   :1.000   Min.   :1.000   Min.   :1.000
# 1st Qu.:3.000   1st Qu.:1.000   1st Qu.:3.000   1st Qu.:1.250
# Median :3.000   Median :2.500   Median :4.000   Median :2.000
# Mean   :3.422   Mean   :2.189   Mean   :3.856   Mean   :1.967
# 3rd Qu.:5.000   3rd Qu.:3.000   3rd Qu.:5.000   3rd Qu.:2.000
# Max.   :5.000   Max.   :4.000   Max.   :6.000   Max.   :3.000
```

```

#      mode        failures       time
# Min. :1.000   Min. : 0.000   Min. : 1.000
# 1st Qu.:1.000 1st Qu.: 0.000 1st Qu.: 1.000
# Median :2.000 Median : 0.000 Median : 2.500
# Mean   :1.578  Mean   : 1.611  Mean   : 4.344
# 3rd Qu.:2.000 3rd Qu.: 2.000 3rd Qu.: 4.000
# Max.   :2.000  Max.   :23.000  Max.   :36.000

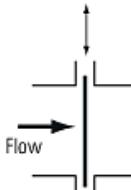
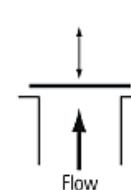
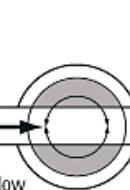
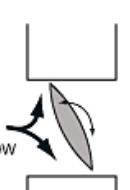
valve$operator=factor(valve$operator, labels=c("air", "solenoid", "motor", "Manual"))
valve$system=factor(valve$system, labels=c("contain", "nuclear", "power", "safety", "aux"))
valve$valve=factor(valve$valve, labels=c("ball", "Butterfly", "diaphragm", "gate",
                                         "Globe", "Dir"))

valve$mode=factor(valve$mode, labels=c("closed", "open"))
valve$sizeGroup=factor(valve$size, labels=c("small", "medium", "large"))

nrow(valve) # 90
valve[1:5,]

#      system operator      valve size mode failures time sizeGroup
# 1 contain motor          gate    3 closed     2    4    large
# 2 contain motor          gate    3 open      2    4    large
# 3 contain motor          Globe   1 closed     1    2    small
# 4 nuclear   air          Butterfly 2 open      0    2    medium
# 5 nuclear   air          diaphragm 2 closed     0    2    medium

```

| Valve movement | Linear | | Rotary | | |
|--|---|---|---|--|--|
| Operating motion of the closing device (obturator) | Straight line | | Rotating about an axis at right angles to the direction of flow | | |
| Direction of flow in the seating area | At right angles to the operating motion of the obturator | Longitudinal to the operating motion of the obturator | Through the obturator | Around the obturator | |
| Basic types | Gate valve | Globe valve | Ball valves | Butterfly valve | |
| Schematic |  |  |  |  | |

Question 1: What would have happened in our analyses if we hadn't used factor()?

```

v1=glm(failures~system+operator+valve+size+mode, data=valve,
       family=poisson, offset=log(time))
summary(v1)
# Coefficients:   Estimate Std. Error z value Pr(>|z|)
# (Intercept) -5.60059   0.90151 -6.212 5.22e-10 ***
# systemnuclear 0.84550   0.53347  1.585  0.11299
# systempower   0.81323   0.50904  1.598  0.11013
# systemsafety   0.88612   0.55155  1.607  0.10814
# systemaux    -0.05361   0.57345 -0.093  0.92552
# operatorsolenoid 0.74251   0.57904  1.282  0.19973
# operatormotor  -1.08856   0.25138 -4.330 1.49e-05 ***
# operatorManual -2.31326   0.47677 -4.852 1.22e-06 ***
# valveButterfly  0.64644   0.74999  0.862  0.38872
# valvediaphragm 0.57828   0.78207  0.739  0.45965
# valvegate     3.12242   0.59809  5.221 1.78e-07 ***
# valveGlobe     1.81486   0.60894  2.980  0.00288 **
# valveDir      1.04705   0.94094  1.113  0.26581
# size          1.03790   0.18381  5.647 1.64e-08 ***
# modeopen      -0.05197   0.18286 -0.284  0.77624
# (Dispersion parameter for poisson family taken to be 1)
# Null deviance: 385.53 on 89 degrees of freedom
# Residual deviance: 210.69 on 75 degrees of freedom
# AIC: 345.03

```

Question 2: What are our preliminary conclusions about valve failure? What specifically does the intercept tell us? What are some reasons that this model might be inadequate?

```

v1SG=glm(failures~system+operator+valve+sizeGroup+mode, data=valve,
          family=poisson, offset=log(time))
summary(v1SG)
# Coefficients:   Estimate Std. Error z value Pr(>|z|)
# (Intercept) -3.76867   0.81935 -4.600 4.23e-06 ***
# systemnuclear 0.91556   0.53184  1.721  0.08516 .
# systempower   1.01881   0.50548  2.016  0.04385 *
# systemsafety  1.22309   0.55518  2.203  0.02759 *
# systemaux     0.33292   0.58408  0.570  0.56869
# operatorsolenoid 0.70437   0.56669  1.243  0.21389
# operatormotor -1.19261   0.24851 -4.799 1.59e-06 ***
# operatorManual -2.47233   0.47660 -5.187 2.13e-07 ***
# valveButterfly  0.18533   0.76105  0.244  0.80761
# valvediaphragm 0.60674   0.78107  0.777  0.43727
# valvegate      2.95894   0.60010  4.931 8.19e-07 ***
# valveGlobe      1.79318   0.61040  2.938  0.00331 **
# valveDir        1.00891   0.93009  1.085  0.27803
# sizeGroupmedium -0.01219   0.28340 -0.043  0.96568
# sizeGrouplarge  1.61457   0.32104  5.029 4.93e-07 ***
# modeopen        -0.20934   0.19033 -1.100  0.27138
# (Dispersion parameter for poisson family taken to be 1)
# Null deviance: 385.53 on 89 degrees of freedom
# Residual deviance: 195.68 on 74 degrees of freedom
# AIC: 332.02

```

Question 3: What's different and which model is more appropriate?

```

v1q=glm(failures~system+operator+valve+sizeGroup+mode, data=valve,
         family=quasipoisson, offset=log(time))
summary(v1q)
# Coefficients:   Estimate Std. Error t value Pr(>|t|)
# (Intercept) -3.76867  1.74297 -2.162  0.0338 *
# systemnuclear  0.91556  1.13136  0.809  0.4210
# systempower    1.01881  1.07528  0.947  0.3465
# systemsafety   1.22309  1.18100  1.036  0.3037
# systemaux      0.33292  1.24248  0.268  0.7895
# operatorsolenoid 0.70437  1.20549  0.584  0.5608
# operatormotor  -1.19261  0.52864 -2.256  0.0270 *
# operatorManual -2.47233  1.01385 -2.439  0.0171 *
# valveButterfly  0.18533  1.61895  0.114  0.9092
# valvediaphragm 0.60674  1.66153  0.365  0.7160
# valvigate       2.95894  1.27657  2.318  0.0232 *
# valveGlobe      1.79318  1.29848  1.381  0.1714
# valveDir        1.00891  1.97853  0.510  0.6116
# sizeGroupmedium -0.01219  0.60286 -0.020  0.9839
# sizeGrouplarge   1.61457  0.68294  2.364  0.0207 *
# modeopen        -0.20934  0.40488 -0.517  0.6067
# (Dispersion parameter for quasipoisson family taken to be 4.525197)
# Null deviance: 385.53 on 89 degrees of freedom
# Residual deviance: 195.68 on 74 degrees of freedom
# AIC: NA

1 - pchisq(summary(v1q)$dispersion * v1SG$df.res, v1SG$df.res) # 0
exp(-2.47233) # 0.084

```

Question 3: How do we know that the Poisson (variance=mean) model is inadequate? What do you conclude about valve failure after adjusting for extra-Poisson variation?

```

valve$operMan = factor(as.numeric(valve$operator),
                      levels=c(4,1,2,3),
                      labels=c("Manual","air","solenoid","motor"))
v1qM=glm(failures~system+operMan+valve+sizeGroup+mode, data=valve,
          family=quasipoisson, offset=log(time))
summary(v1qM)
# ...
# operManair      2.47233   1.01385   2.439  0.01715 *
# operMansolenoid 3.17669   1.55816   2.039  0.04505 *
# operManmotor    1.27971   1.06563   1.201  0.23362

```

Question 4: How does the code work to change the baseline for operator?
 What different conclusions can we now justify?

```

v1q9=glm(failures~system+mode+valve+operMan*sizeGroup, data=valve,
          family=quasipoisson, offset=log(time))
summary(v1q9)
# Coefficients: (1 not defined because of singularities)
#                                     Estimate Std. Error t value Pr(>|t|)
# ...
# operManair                  -0.4566   1.3939  -0.328  0.7442
# operMansolenoid             -15.4342  2570.1681 -0.006  0.9952
# operManmotor                 -2.6733   1.9116  -1.398  0.1665
# sizeGroupmedium              -2.5676   2.3771  -1.080  0.2838
# sizeGrouplarge                -2.9967   2.4031  -1.247  0.2166
# operManair:sizeGroupmedium   2.3734   2.4639   0.963  0.3388
# operMansolenoid:sizeGroupmedium 18.8008  2570.1692  0.007  0.9942
# operManmotor:sizeGroupmedium   3.3524   2.8746   1.166  0.2475
# operManair:sizeGrouplarge     4.3476   2.4829   1.751  0.0844 .
# operMansolenoid:sizeGrouplarge NA        NA       NA       NA
# operManmotor:sizeGrouplarge    5.7416   2.8598   2.008  0.0486 *

```

Question 5: The above (partial) results are for the only significant 2-way interaction. Why does one line have NA? How could you explain the significant interaction to a client?