

Regression Analysis

Poisson Regression

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Statistical Inference: Data
Example

About This Lesson



Data Example 1: High School Awards

Objective: To model and predict the number of awards earned by students at one high school for multiple high schools.

Response Variable: The number of awards earned by students at a high school per year

Predicting Variables:

- The type of program in which the student was enrolled, with three levels:
1 = "General", 2 = "Academic" and 3 = "Vocational"; and
- The score on the final exam in math.

Acknowledgement: This data example was acquired from the Institute for Digital Research and Education at University of California, Los Angeles.

Data Example 1: Statistical Inference

```
m1 = glm(num_awards ~ prog + math, family="poisson", data=awardsdata)
summary(m1)
```

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-5.24712	0.65845	-7.969	1.60e-15 ***
progAcademic	1.08386	0.35825	3.025	0.00248 **
progVocational	0.36981	0.44107	0.838	0.40179
math	0.07015	0.01060	6.619	3.63e-11 ***

Null deviance: 287.67 on 199 degrees of freedom

Residual deviance: 189.45 on 196 degrees of freedom

```
1-pchisq((287.67-189.45),(199-196))
```

```
[1] 0
```

Test for significance β_{math} : p-value≈0 thus statistically significant

Test for overall regression: p-value ≈0 thus at least one predicting variables significantly explains the variability in the number of awards

Data Example 2: Insurance Claims

Objective: To explain factors that are associated to car insurance claims due to accidents or other events leading to car damage.

Response Variable: The number of car insurance claims per policyholder.

- Holders: numbers of policyholders; and
- Claims: numbers of claims

Predicting Variables:

- District of residence of policyholder (1 to 4): 4 is major cities.
- Classification of cars with levels <1 litre, 1–1.5 litre, 1.5–2 litre, >2 litre.
- Age group of the policyholder: <25, 25–29, 30–35, >35.

Data Example 2: Statistical Inference

```
m.ins = glm(Claims ~ District + Group + Age + offset(log(Holders)),  
data = Insurance, family = poisson)  
summary(m.ins)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-1.810508	0.032972	-54.910	< 2e-16 ***
.....				
Age.L	-0.394432	0.049404	-7.984	1.42e-15 ***
Age.Q	-0.000355	0.048918	-0.007	0.994210
Age.C	-0.016737	0.048478	-0.345	0.729910



What are these
Age variables?

Null deviance: 236.26 on 63 degrees of freedom

Residual deviance: 51.42 on 54 degrees of freedom

Data Example 2: Statistical Inference

```
library(MASS)
summary(Insurance)
```

```
Ins.dat = within(Insurance, {
  Age = factor(Age, ordered = F)
  Group = factor(Group, ordered = F)
  District = factor(District, ordered = F)})
```

```
m.ins = glm(Claims ~ District + Group + Age + offset(log(Holders)), data = Ins.dat, family = poisson)
summary(m.ins)
```

Data Example 2: Statistical Inference

```
m.ins = glm(Claims ~ District + Group + Age + offset(log(Holders)),  
data = Ins.dat, family = poisson)  
summary(m.ins)
```

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-1.82174	0.07679	-23.724	< 2e-16 ***
.....				
Age25-29	-0.19101	0.08286	-2.305	0.021149 *
Age30-35	-0.34495	0.08137	-4.239	2.24e-05 ***
Age>35	-0.53667	0.06996	-7.672	1.70e-04 ***

Test for significance

$$\beta_{age25-29} = -0.191 \text{ or } \exp(\beta_{age25-29}) = 0.826$$

$\beta_{age25-29}$, $\beta_{age30-35}$ & $\beta_{age>35}$: p-value < 0.05 thus statistically significant.

Null deviance: 236.26 on 63 degrees of freedom
Residual deviance: 51.42 on 54 degrees of freedom

Data Example 2: Statistical Inference (cont'd)

```
m.ins = glm(Claims ~ District + Group + Age + offset(log(Holders)),  
data = Ins.dat, family = poisson)  
summary(m.ins)
```

test for overall regression

```
1-pchisq((236.26-51.42),(63-54))
```

Test for overall regression: p-value ≈ 0 thus at least one predicting variables significantly explains the variability in the number of claims

Data Example 2: Statistical Inference (cont'd)

Is the district of residence of policyholder a statistically significant variable given all other predicting variables in the model?

Full model: District + Group + Age

Reduced model: Group + Age

```
library(aod)  
wald.test(b=coef(m.ins), Sigma=vcov(m.ins), Terms=2:4)
```

Wald test:

Chi-squared test:

X² = 14.6, df = 3, P(> X²) = 0.0022

Test for subsets of coefficients: p-value = 0.002 reject the null hypothesis and conclude that the District variable does have significant explanatory power

Summary

