

## Poisson Regression with R

### **glm** function output

- ▶ The output begins with echoing the function call. Then the information on deviance residuals is displayed.
- ▶ Deviance residuals are approximately normally distributed if the model is specified correctly.
- ▶ Here it shows a little bit of skeweness since median is not quite zero.

## **glm** function output

- ▶ The Poisson regression coefficients for each of the variables along with the standard errors, z-scores, p-values and 95% confidence intervals for the coefficients.
- ▶ The coefficient for math is 0.07.
- ▶ This means that the expected log count for a one-unit increase in math is 0.07.

# Poisson Regression with R

Coefficients:

Estimate Std. Error z value Pr(>|z|)

(Intercept)	-5.2471	0.6585	-7.97	1.6e-15 ***
progAcademic	1.0839	0.3583	3.03	0.0025 **
progVocational	0.3698	0.4411	0.84	0.4018
math	0.0702	0.0106	6.62	3.6e-11 ***

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

## Poisson Regression with R

### **glm** function output

- ▶ The indicator variable **progAcademic** compares between **prog = Academic** and **prog = "General"** , the expected log count for **prog = Academic** increases by about 1.1.
- ▶ The indicator variable **prog.Vocational** is the expected difference in log count ( $\approx 0.37$ ) between **prog = "Vocational"** and the reference group (**prog = "General"** ).

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- ▶ The output above indicates that the incident rate for **prog = “Academic”** is 2.96 times the incident rate for the reference group (**prog = “General”**).
- ▶ Likewise, the incident rate for **prog = “Vocational”** is 1.45 times the incident rate for the reference group holding the other variables at constant.

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- ▶ The percent change in the incident rate of **num\_awards** is by 7% for every unit increase in math.

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## Deviance

- ▶ In statistics, deviance is a quality of fit statistic for a model that is often used for statistical hypothesis testing.
- ▶ It is a generalization of the idea of using the sum of squares of residuals in ordinary least squares to cases where model-fitting is achieved by maximum likelihood.

## Poisson Regression with R

### **glm** function output

- ▶ The information on deviance is also provided.
- ▶ We can use the residual deviance to perform a goodness of fit test for the overall model.



## Poisson Regression with R

### **glm** function output

- ▶ The residual deviance is the difference between the deviance of the current model and the maximum deviance of the ideal model where the predicted values are identical to the observed.
- ▶ Therefore, if the residual difference is small enough, the goodness of fit test will not be significant, indicating that the model fits the data.

## Poisson Regression with R

### **glm** function output

- ▶ If the test had been statistically significant, it would indicate that the data do not fit the model well.
- ▶ We could try to determine if there are omitted predictor variables, if our linearity assumption holds and/or if there is an issue of over-dispersion.

## Comparing Candidate Models

- ▶ We can also test the overall effect of prog by comparing the deviance of the full model with the deviance of the model excluding prog.
- ▶ The two degree-of-freedom chi-square test indicates that prog, taken together, is a statistically significant predictor of **num\_awards**.

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## Comparing Models

```
# update m1 model dropping prog
m2 <- update(m1, . ~ . - prog)

# test model differences with chi square test
anova(m2, m1, test="Chisq")
```

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## Analysis of Deviance Table

Model 1: num\_awards ~ math

Model 2: num\_awards ~ prog + math

	Resid. Df	Resid. Dev	Df	Deviance	Pr(>Chi)
1	198	204			
2	196	189	2	14.6	0.00069 ***

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

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### Incident Rate Ratios

- ▶ Sometimes, we might want to present the regression results as **incident rate ratios** (IRRs) and their standard errors, together with the confidence interval.
- ▶ To compute the standard error for the incident rate ratios, we will use the **Delta method** (Numerical Computation Method).
- ▶ To this end, we make use the function `deltamethod` implemented in R package **msm**.

## Incident Rates

Incidence rate is the occurrence of an event over person-time, for example person-years.

$$\text{Incidence Rate} = \frac{\text{events}}{\text{Person Time}}$$

Note: the same time intervals must be used for both incidence rates.

### Incident Rate Ratios

A **rate ratio** (sometimes called an incidence density ratio) in epidemiology, is a relative difference measure used to compare the incidence rates of events occurring at any given point in time.

$$\text{Incidence Rate Ratio} = \frac{\text{Incidence Rate 1}}{\text{Incidence Rate 2}}$$



# Poisson Regression with R

## Delta Method

```
s <- deltamethod(list(~ exp(x1), ~ exp(x2), ~ exp(x3),  
  ~ exp(x4)), coef(m1), cov.m1)  
  
#exponentiate old estimates dropping the p values  
  
rexp.est <- exp(r.est[, -3])  
  
# replace SEs with estimates  
# for exponentiated coefficients  
  
rexp.est[, "Robust SE"] <- s
```

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rexp.est

	Estimate	Robust SE	LL	UL
(Intercept)	0.005263	0.00340	0.001484	0.01867
progAcademic	2.956065	0.94904	1.575551	5.54620
progVocational	1.447458	0.57959	0.660335	3.17284
math	1.072672	0.01119	1.050955	1.09484

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- ▶ Sometimes, we might want to look at the expected marginal means.
- ▶ For example, what are the expected counts for each program type holding math score at its overall mean?
- ▶ To answer this question, we can make use of the predict function.
- ▶ First off, we will make a small data set to apply the predict function to it.

## Poisson Regression with R

```
(s1 <- data.frame(math = mean(p$math),  
  prog = factor(1:3, levels = 1:3,  
  labels = levels(p$prog))))
```

	math	prog
1	52.65	General
2	52.65	Academic
3	52.65	Vocational

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```
predict(m1, s1, type="response", se.fit=TRUE)
```

```
$fit
```

1	2	3
0.2114	0.6249	0.3060

```
$se.fit
```

1	2	3
0.07050	0.08628	0.08834

```
$residual.scale
```

```
[1] 1
```

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- ▶ In the output above, we see that the predicted number of events for level 1 of prog is about 0.21, holding math at its mean.
- ▶ The predicted number of events for level 2 of prog is higher at 0.62, and the predicted number of events for level 3 of prog is about .31.
- ▶ The ratios of these predicted counts ( $\frac{0.625}{0.211} = 2.96$ ,  $\frac{0.306}{0.211} = 1.45$ ) match what we saw looking at the IRR.

