Poisson regression

- At this point, we are ready to perform our Poisson model analysis using the glm function.
- We fit the model and store it in the object model1 and get a summary of the model.

```
model1 <- glm(num_awards ~ prog + math,
family="poisson", data=p)
summary(model1)</pre>
```

```
Call:
glm(formula = num_awards ~ prog + math, family = "poi
Deviance Residuals:
Min 1Q Median 3Q Max
```

-2.204 -0.844 -0.511 0.256 2.680

```
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
              0.0702 0.0106 6.62 3.6e-11 *
math
             0 '*** 0.001 '** 0.01 '* 0.05
Signif. codes:
```

(Dispersion parameter for poisson family taken to be

Null deviance: 287.67 on 199 degrees of freedom Residual deviance: 189.45 on 196 degrees of freedom

ATC: 373.5

Number of Fisher Scoring iterations: 6

- ▶ It is recommended using robust standard errors for the parameter estimates to control for mild violation of the distribution assumption that the variance equals the mean.
- The R package sandwich can be used to obtain the robust standard errors and calculated the p-values accordingly.
- ➤ Together with the p-values, we have also calculated the 95% confidence interval using the parameter estimates and their robust standard errors.

sandwich R Package

- ► Robust Covariance Matrix Estimators
- Model-robust standard error estimators for cross-sectional, time series, and longitudinal data.

Robust Standard Errors

```
cov.model1 <- vcovHC(model1, type="HCO")</pre>
std.err <- sqrt(diag(cov.model1))</pre>
r.est <- cbind(Estimate= coef(model1),</pre>
"Robust SE" = std.err.
"Pr(>|z|)" = 2 * pnorm(abs(coef(model1)/std.err)|
lower.tail=FALSE),
LL = coef(model1) - 1.96 * std.err,
UL = coef(model1) + 1.96 * std.err)
```

r.est

SE Pr(> :	z)	LL U	JL	
-5.24712	0.64600	4.567e-16	-6.5133	-3
1.08386	0.32105	7.355e-04	0.4546	1
0.36981	0.40042	3.557e-01	-0.4150	1
0.07015	0.01044	1.784e-11	0.0497	0
	-5.24712 1.08386 0.36981	1.08386 0.32105 0.36981 0.40042	-5.24712	-5.24712