



Logistic Regression With Replications

Aggregate data for Logistic Regression with repetitions

Fit a logistic regression model

model.agg = glm(cbind(Obesity,Total-Obesity)~agegr+gender+edu, data=obdata.agg, family=binomial)

Test for GOF: Using deviance residuals

c(deviance(model.agg), 1-pchisq(deviance(model.agg),40))
[1] 29.0640209 0.8996714

Georgia Tech

3

Logistic Regression With Replications

Aggregate data for Logistic Regression with repetitions

Fit a logistic regression model

model.agg = glm(cbind(Obesity,Total-Obesity)~agegr+gender+edu, data=obdata.agg, family=binomial)

Test for GOF: Using deviance residuals

c(deviance(model.agg), 1-pchisq(deviance(model.agg),40))
[1] 29.0640209 0.8996714

With replications, we can perform a goodness of fit test. *p-value* = **0.899** indicates a good fit.

Georgia Tech

4

Logistic Regression With Replications

summary(model.agg)

Coefficients:

	Estimate	Std. Error	z value	Pr(> z)
(Intercept)	-1.20581	0.15730	-7.666	1.78e-14 ***
agegr25to34	0.47271	0.14428	3.276	0.001052 **
agegr35to44	0.76486	0.14196	5.388	7.13e-08 ***
agegr45to64	0.84815	0.13240	6.406	1.49e-10 ***
agegr65+	0.60086	0.13751	4.370	1.24e-05 ***
genderFemale	0.23041	0.06363	3.621	0.000293 **
edu9to11Grade	0.05632	0.12229	0.461	0.645110
eduHighSchool	-0.03440	0.11436	-0.301	0.763579
eduSomeCollege	0.13947	0.11036	1.264	0.206301
eduCollege+	-0.40077	0.11757	-3.409	0.000653 **

- Regression coefficient output for estimation and statistical inference is the same with or without replications.
- Null and residual deviance output is different with replications. Why?

Null deviance: 127.701 on 49 degrees of freedom Residual deviance: 29.064 on 40 degrees of freedom

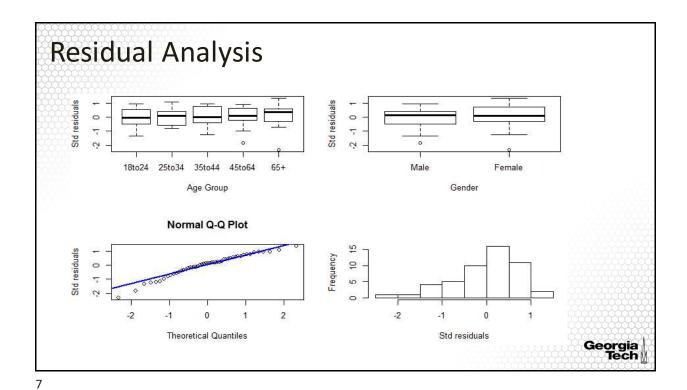
> Georgia Tech

5

Residual Analysis

hist(res, 10, xlab="Std residuals", main="")

Georgia Tech



Prediction of Adult Obesity: Results

- Both gender and age group factors are statistically significant factors in explaining the variability in the classification of adults by obesity.
 - But the fitted model with education, gender, and age group factors does not improve prediction.
- After factor aggregation, goodness of fit can be performed.
- The *p-value* of the deviance test for goodness of fit is high, indicating good fit.
 - But residual analysis suggests that there may be some departures from normality and thus from goodness of fit.
- Models with different link functions or including interaction terms have not shown improvement. (Results not shown in this lecture.)
- The sample size is large enough for reliable statistical inference.

Georgia Tech

Prediction of Adult Obesity: Results

- Both gender and age group factors are statistically significant factors in explaining the variability in the classification of adults by obesity.
 - But the fitted model with education, gender, and age group factors does not improve prediction.
- After factor aggregation, goodness of fit can be performed.
- The *p-value* of the deviance test for goodness of fit is high, indicating good fit.
 - But residual analysis suggests that there may be some departures from normality and thus from goodness of fit.
- Models with different link functions or including interaction terms have not shown improvement. (Results not shown in this lecture.)
- The sample size is large enough for reliable statistical inference.

What can be done to improve the model fit and the predictive power?

- Include other factors in the model, such as income level, unemployment, race, and ethnicity, among others.
- Consider interaction terms between age, education, and gender groups and other factors.

Georgia Tech

9

Summary



Georgia Tech