

July 8, 2013

Class goals

- ▶ This course provides an introduction to Generalized Linear Model (GLM) and Generalized Linear Mixed Model (GLMM)
 - ▶ Mathematical background
 - ▶ Intuition and conceptualization
 - ▶ Geometrical interpretation
 - ▶ Common issues and solutions for GLM and GLMM analyses
 - ▶ Relation to ANOVA
- ▶ We will learn
 - ▶ how to conduct, interpret and report GLM and GLMM analyses in R
 - ▶ how to visualize data in R
 - ▶ how to prepare data for visualization and analysis (transformation)
 - ▶ The course will be part lecture, part learning by doing.

Lecture 1:

- ▶ (re-)introducing Generalized Linear Models (GLM)
- ▶ relation between GLM and ANOVA
- ▶ example (linear) models and geometric interpretation
- ▶ Generalized Linear Mixed Models (GLMM)

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Lecture 2:

- ▶ relation between GLMM and ANOVA
- ▶ logistic regression
- ▶ common issues and solutions in GLMs and GLMMs
 - ▶ identifying and removing outliers
 - ▶ coding your predictors
- ▶ example models

Lecture 3: Beyond linear models

- ▶ Binomial models (logistic regression and mixed logit models)
- ▶ Empirical logit weighted linear regression
- ▶ Poisson models

Lecture 3: Beyond linear models

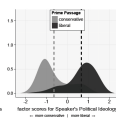
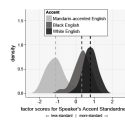
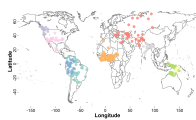
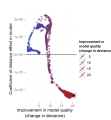
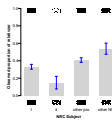
- ▶ Binomial models (logistic regression and mixed logit models)
- ▶ Empirical logit weighted linear regression
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Lecture 4: Tools for data analysis, exploration, and transformation

- ▶ `plyr`
- ▶ `reshape2`

Lecture 5: Visualizing and summarizing your data

► library ggplot2



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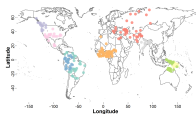
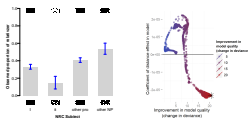


Table: Example Stargazer table generated from R

- ▶ library stargazer
- ▶ library knitr

	<i>Dependent variable:</i>		
	<i>(logged) RT</i>		<i>Correct response?</i>
	<i>OLS</i>		<i>logistic</i>
	<i>(1)</i>	<i>(2)</i>	<i>(3)</i>
Intercept	6.497*** (0.030)	6.466*** (0.028)	1.664** (0.666)
Word frequency (logged)	-0.031*** (0.006)	-0.031*** (0.006)	0.412*** (0.154)
Native language	0.285*** (0.042)	0.286*** (0.042)	-1.642* (0.886)
Trial position	-0.0003** (0.0001)		
Word frequency (logged):Native language	-0.027*** (0.009)	-0.027*** (0.009)	0.261 (0.212)
Observations	1, 659	1, 659	1, 659
R ²	0.161	0.158	
Adjusted R ²	0.159	0.157	
Akaike Inf. Crit.			520.100

Note: *p<0.1; **p<0.05; ***p<0.01

Lecture 6 and 7: Common issues and solutions in GLMs and GLMMs

- ▶ collinearity
- ▶ model evaluation
- ▶ random effect structure

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Lecture 8: Remaining issues and continued discussion

- ▶ Reporting GLMMs in your article

Oh, we are all so different ...

Folks in this class represent varied linguistic interest and varied degrees of expertise in statistics and R.

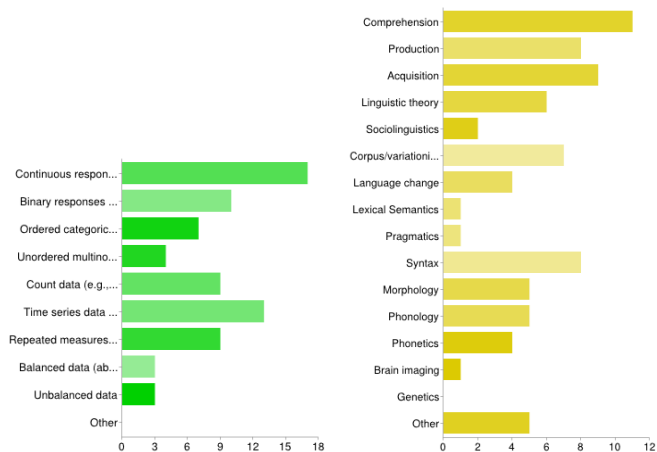


Figure: Background 7 and areas of interest fig-survey:subfig6

Oh, we are all so different ...

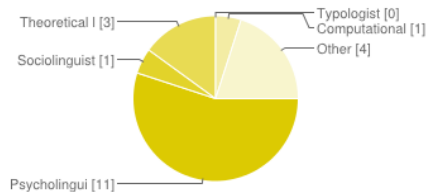


Figure: Background fig-survey:subfig4

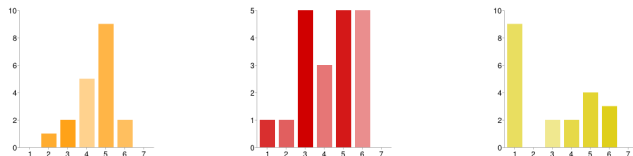


Figure: Expertise in regression fig-survey:subfig1, R fig-survey:subfig2, and lme4 fig-survey:subfig3

→ Please be patient and help each other out. (Change seating arrangement?)

Acknowledgments

- ▶ The slides for this class include (usually modified) materials prepared by:
 - ▶ Judith Degen (Rochester)
 - ▶ Maureen Gillespie (New Hampshire)
 - ▶ Dave Kleinschmidt (Rochester)
 - ▶ Victor Kuperman (Stanford)
 - ▶ Roger Levy (UCSD)
- ... with their permission
- ▶ I am also grateful for feedback from:
 - ▶ Austin Frank (Rochester)
 - ▶ Previous audiences to similar workshops at CUNY, Haskins, Rochester, Buffalo, UCSD, MIT, Iowa, and Groningen.