



# Move over ANOVA?

Aspects of analysing longitudinal data with  
Repeated Measures ANOVA and  
Mixed Models Regression

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# Outline

- Introduction:
  - Recap: Repeated Measures ANOVA
  - Recap: Mixed-effect Regression Models
- Categorical Data
- Akaike Information Criterion
- Comparison and Summary



# Repeated Measures ANOVA

- Same assumptions as regular ANOVA, only samples are not independent
- Particularly suited for longitudinal studies where change through time is the main subject of research
- Main difference with regular ANOVA is partitioning of variance
  - SSE is divided into a within-subjects (SSS) and among-subjects part
  - The within-subjects differences are removed from SSE



# Mixed Models

- Multiple regression models look for the best fitting formula, i.e. the combination of effects for each variable that best fits the observed data
- Mixed Models distinguish between random-effect and fixed-effect factors
- In longitudinal studies, time can be one of the independent variables

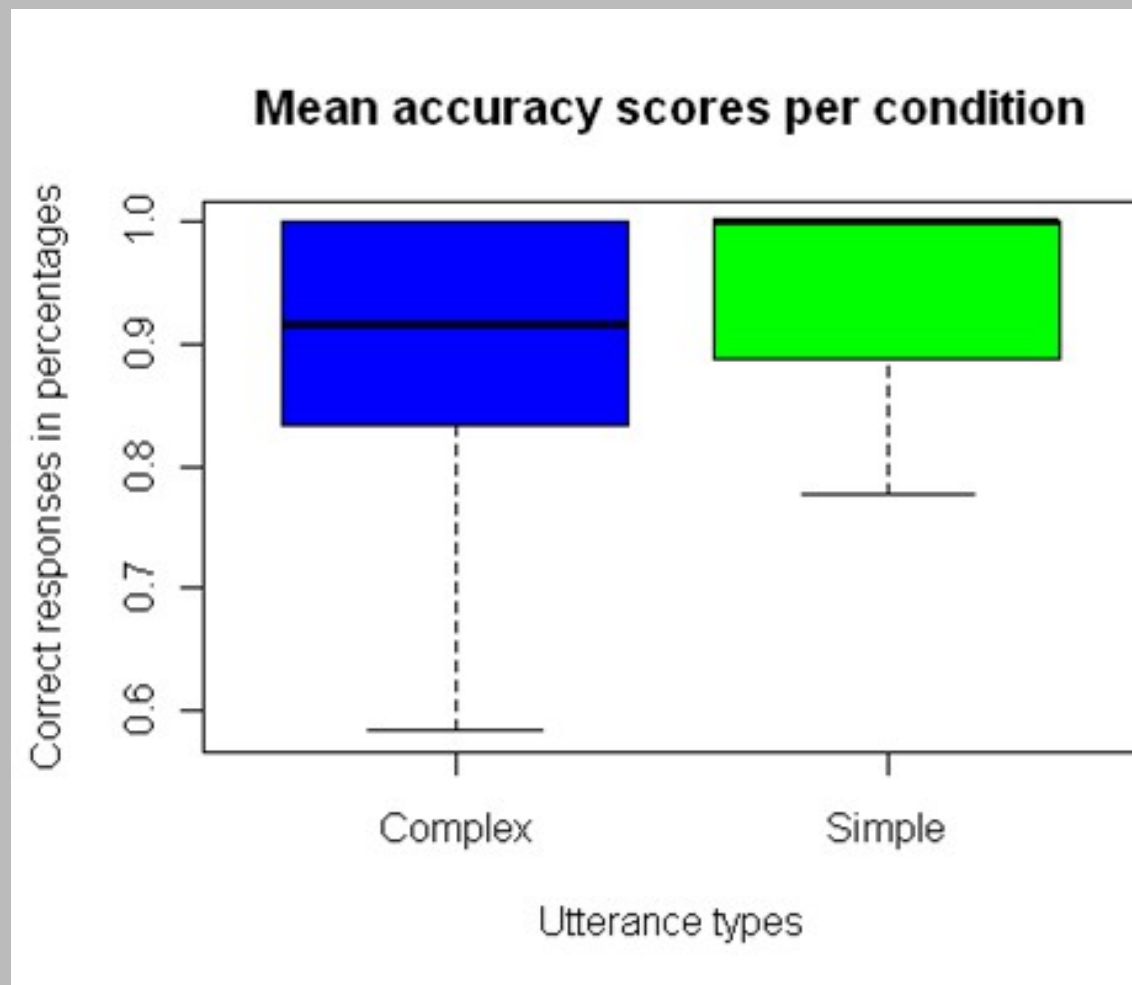


# Some previously mentioned issues

- Use of Repeated Measures ANOVA on categorical data
- Akaike Information Criterion

# Categorical Data

- Categorical data:
  - ANOVA Confidence intervals may extend into non-existent ranges
  - This underestimates the probability of existent values





# AIC

- Using Mixed-effects Regression Models, adding additional random factors is relatively easy
  - Will always lead to a better fit, but at the cost of adding complexity
- We can use the *Akaike Information Criterion* (AIC), to evaluate whether adding particular random effects is beneficial to the overall model

# AIC

- AIC is a function of the amount of parameters used, and the fit of the model
  - Lower result = better fit, given the amount of parameters
- In Ruggero's mixed models analysis, we can see this progression:





# AIC

	Df	AIC	BIC	logLik	Chisq	Chi	Df	Pr(>Chisq)
mmodel1	18	449.24	562.55	-206.62				
mmodel2	19	286.84	406.45	-124.42	164.3964		1	<2e-16 ***
mmodel4	21	290.79	422.98	-124.39	0.0567		2	0.9720
mmodel5	21	271.20	403.39	-114.60	19.5866		0	<2e-16 ***
mmodel3	28	275.32	451.58	-109.66	9.8796		7	0.1955

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mmodel2	19	286.84	406.45	-124.42				
mmodel5	21	271.20	403.39	-114.60	19.643		2	5.426e-05 ***



# AIC

	Df	AIC	BIC	logLik	Chisq	Chi	Df	Pr(>Chisq)
mmmodel5	21	271.2	403.39	-114.6				
mmmodel6	22	273.2	411.69	-114.6	0		1	0.9991
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mmmodel5	21	271.20	403.39	-114.60				
mmmodel7	22	249.23	387.72	-102.62	23.966		1	9.804e-07 ***



# AIC

	Df	AIC	BIC	logLik	Chisq	Chi	Df	Pr(>Chisq)
mmodel18	23	-31.656	113.13	38.828				
mmodel10	24	-51.621	99.46	49.810	21.965		1	2.777e-06 ***
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mmodel11	25	-52.813	104.56	51.407				
mmodel12	26	-54.469	109.20	53.234	3.6555		1	0.05588 .

# AIC

- In short: the AIC is one way to answer Ruggero's question: «Where is it convenient to stop?»
- There are also related measures for model fit:
  - Bayesian Information Criterion
  - AICc: takes into account sample size



# Comparison / Summary

- Aspects of Repeated Measures ANOVA:
  - Does not deal well with missing values
    - sometimes subjects have to be dropped due to this, which introduces the possibility of sample bias
  - Observations need to be made at the same time points for each subject
  - Requires tests for conditions of *sphericity*
  - Leads to spurious results for categorical outcomes
    - confidence intervals may extend beyond interpretable ranges



# Comparison / Summary

- Aspects of Mixed-effects regression:
  - Can use all available data points
    - not affected by *randomly* missing data
    - can model time effects
  - Can handle both covariates that change over time, and static covariates
  - Can be relatively difficult to implement
    - though see Wieling's presentation
  - Does not work very well on small samples



# Comparison / Summary

- Aspects of Mixed-effects regression:
  - Easy to add extra random factors
    - Testable how these factors affect the fit of the model using AIC



# When to use which method?

- Mixed Models:
  - missing data
  - irregular time points
  - modest to large samples
  - categorical outcomes
- RM ANOVA:
  - complete data
  - regular time points
  - small samples
  - assumptions of normality and sphericity satisfied





# When to use which method?

method	data missing?		time points		sample size		categorical data
	yes	no	regular	irregular	small	medium to large	
Mixed Models	green	green	green	green	red	green	green
RM ANOVA	red	green	green	red	green	green	red

# References

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