



Towards polytomous mixed effects regression analysis and modeling

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The PROBLEM

- *Mixed effects* modeling is not explicitly implemented (in *R*) for *polytomous* (>2) *nominal* (unordered categorical) outcomes



One solution

- Apply the **Poisson** (loglinear/count) equivalent of the **multinomial** logit [logistic] model
 - Yields one simultaneously fit model
- Agresti 2002: 330-332 (cf. Faraway 2006: 101-102)
 - "*Loglinear* models treat categorical response variables symmetrically, focusing on associations and interactions in their joint distribution. *Logit* models [e.g. multinomial logistic], by contrast, describe how a single categorical response depends on explanatory variables. ...
Moreover, logit models with *categorical* explanatory variables have equivalent loglinear [count] models."₃



Alternative approach

- Apply the *one-vs-rest* heuristic using a set of *binary* mixed effects models (with the *lmer(..., family=binomial)* function in the *lme4* package)
 - Requires and yields *many & separately fit* models
- Agresti 2002: 515
 - Section 12.4.4 *Baseline-Category Logit Models with Random [Mixed] Effects*: "This requires using a vector or cluster-specific random effects $u_{j,j}$, one for each logit [model]. ... With nominal responses there is not reason to expect effects to be similar for different [outcomes] j ."



Formal specification of multinomial and Poisson models

- Multinomial

- $\text{logit}[P(Y)] = \log[P(Y)/P(\neg Y)] = \alpha + \beta X$

- Poisson

- $\log[n(Y)] = \alpha + \beta X$



Poisson-multinomial link – first with only *linguistic* predictor variables

- From multinomial variable representation to outcome-wise *counts* for each unique multinomial variable combination (context)

#	Proportion	Count	Lexeme	Person	Number	Agent	Observation
1	0.59598214	267	ajatella	First	NIL	Individual	1
2	0.08482143	38	harkita	First	NIL	Individual	1
3	0.27008929	121	mieltä	First	NIL	Individual	1
4	0.04910714	22	pohtia	First	NIL	Individual	1
5	1.00000000	1	ajatella	First	NIL	NIL	2
6	0.00000000	0	harkita	First	NIL	NIL	2
7	0.00000000	0	mieltä	First	NIL	NIL	2
8	0.00000000	0	pohtia	First	NIL	NIL	2



Comparing the equivalent multinomial/Poisson models

- Person=THIRD & Number=PLURAL & Agent=GROUP (altogether 23 cases in the data)

#	Proportion	Count	Lexeme	Person	Number	Agent	Observation
65	0.13043478	3	ajatella	Third	Plural	Group	17
66	0.26086957	6	harkita	Third	Plural	Group	17
67	0.08695652	2	miettiä	Third	Plural	Group	17
68	0.52173913	12	pohtia	Third	Plural	Group	17



Specifying and fitting equivalent multinomial/Poisson models

- Multinomial model (*baseline category*)
 - THINK.model_multinom.Person_Number_Agent.demo
 <- multinom(**Lexeme ~ Person + Number + Agent**,
 data=THINK.Lexeme_Person_Number_Agent_Section.multinomial_data_demo)
- Polytomous one-vs-rest
 - THINK.model_ovr.Person_Number_Agent.demo <- plr("Lexeme", "Person + Number + Agent",
 data=THINK.Lexeme_Person_Number_Agent_Section.multinomial_data_demo, heuristic="one.vs.rest", ...)
- Poisson model
 - THINK.model_poisson.Person_Number_Agent.demo
 <- glm(**Count ~ Observation + Lexeme + Lexeme:Person + Lexeme:Number + Lexeme:Agent**,
 data=THINK.Lexeme_Person_Number_Agent.counts_demo,
 family=**poisson**)



Comparing the equivalent multinomial/Poisson models

- Person=THIRD & Number=PLURAL & Agent=GROUP (altogether 23 cases in the data)

#	Proportion	Count	Lexeme	Person	Number	Agent	Observation
65	0.13043478	3	ajatella	Third	Plural	Group	17
66	0.26086957	6	harkita	Third	Plural	Group	17
67	0.08695652	2	miettiä	Third	Plural	Group	17
68	0.52173913	12	pohtia	Third	Plural	Group	17



Comparing the equivalent **multinomial/Poisson models**

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>mieltä</i>	<i>pohtia</i>	Σ
3	6	2	12	23

- Original proportions

<i>ajatella</i>	<i>harkita</i>	<i>mieltä</i>	<i>pohtia</i>
0.13043478	0.26086957	0.08695652	0.52173913

- Multinomial model (with *multinom/nnet*)

```
predict(THINK.model_multinom.Person_Number_Agent.demo,  
newdata=data.frame(matrix(c("Third", "Plural", "Group"),1,,  
dimnames=list(NULL,c("Person","Number","Agent")))),type="probs")
```

<i>ajatella</i>	<i>harkita</i>	<i>mieltä</i>	<i>pohtia</i>
0.1479454	0.2196239	0.1082725	0.5241583



Comparing the equivalent **polytomous** models

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>	Σ
3	6	2	12	23

- Original proportions

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.13043478	0.26086957	0.08695652	0.52173913

- One-vs-rest

```
predict.plr(THINK.model_ovr.Person_Number_Agent.demo,  
newdata=data.frame(matrix(c("Third","Plural","Group"),  
1,,dimnames=list(NULL,c("Person","Number","Agent")))),type="response")
```

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.1620709	0.2316961	0.1084190	0.5158476



Comparing the equivalent multinomial/**Poisson** models

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>miettiä</i>	<i>pohtia</i>	Σ
3	6	2	12	23

- Poisson model count estimates

```
fitted(THINK.model_poisson.Person_Number_Agent.demo)[65:68]
```

65	66	67	68	Σ
3.402750	5.051696	2.490234	12.055320	23

- Corresponding proportion estimates

```
fitted(THINK.model_poisson.Person_Number_Agent.demo)[65:68]/23
```

<i>[ajatella]</i>	<i>[harkita]</i>	<i>[miettiä]</i>	<i>[pohtia]</i>
0.1479456	0.2196390	0.1082711	0.5241444



Comparing the equivalent multinomial/ Poisson models – estimated probabilities

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>	Σ
3	6	2	12	23

- Original proportions

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.13043478	0.26086957	0.08695652	0.52173913

- Multinomial (baseline category) probability estimates

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.1479454	0.2196239	0.1082725	0.5241583

- One-vs-rest probability estimates

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.1620709	0.2316961	0.1084190	0.5158476

- Poisson-count based proportion estimates

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.1479456	0.2196390	0.1082711	0.5241444



Comparing model odds – *multinomial vs. one-vs-rest*

Multinomial (Baseline category)				Polytomous one-vs-rest				
VariableLevel	<i>harkita</i>	<i>mieltä</i>	<i>pohtia</i>	VariableLevel	<i>ajatella</i>	<i>harkita</i>	<i>mieltä</i>	<i>pohtia</i>
(Intercept)	0.35	0.29	0.32	(Intercept)	1.04	0.21	0.18	0.20
PersonFirst	0.72	1.35	0.20	PersonFirst	1.46	0.84	1.82	0.19
PersonSecond	0.28	2.34	0.38	PersonSecond	1.01	0.26	3.20	0.32
PersonThird	1.05	1.91	1.15	PersonThird	0.74	0.87	1.83	0.95
NumberPassive	0.74	1.41	2.51	NumberPassive	0.65	0.53	1.13	2.47
NumberPlural	0.82	0.62	1.06	NumberPlural	1.28	0.92	0.63	1.26
AgentGroup	4.96	2.09	9.03	AgentGroup	0.20	1.76	0.60	4.49
AgentIndividual	0.67	1.25	1.44	AgentIndividual	0.88	0.59	1.22	1.46

Multinomial model in comparison to the one-vs-rest exaggerates impact of variables – due to comparison with only baseline category (*ajatella*) instead all the three other categories (1 vs. 3) – though the effect directions are similar



Comparing model odds – *multinomial vs. Poisson*

Multinomial (Baseline category)				Poisson equivalent					
Variable	Level	<i>harkita mieltä pohtia</i>		Variable	<i>ajatella harkita mieltä pohtia</i>				
(Intercept)		0.35	0.29	0.32	(Intercept)	4.30	-	-	-
PersonFirst		0.72	1.35	0.20	Lexeme	-	0.35	0.29	0.32
PersonSecond		0.28	2.34	0.38	PersonFirst	4.99	3.59	6.71	NA
PersonThird		1.05	1.91	1.15	PersonSecond	2.65	0.75	6.21	NA
NumberPassive		0.74	1.41	2.51	PersonThird	0.87	0.92	1.67	NA
NumberPlural		0.82	0.62	1.06	NumberPassive	0.40	0.29	0.56	NA
AgentGroup		4.96	2.09	9.03	NumberPlural	0.94	0.78	0.59	NA
AgentIndividual		0.67	1.25	1.44	AgentGroup	0.11	0.55	0.23	NA
					AgentIndividual	0.70	0.47	0.87	NA

Not directly comparable odds → need to reorder baseline lexeme category for one of the two models



Comparing model odds – *multinomial vs. Poisson*

Multinomial (Baseline category)				Poisson equivalent				
VariableLevel	<i>ajatella</i>	<i>harkita</i>	<i>miettiä</i>	Variable	<i>ajatella</i>	<i>harkita</i>	<i>miettiä</i>	<i>pohtia</i>
(Intercept)	3.10	1.07	0.91	(Intercept)	4.30	-	-	-
				Lexeme	-	0.35	0.29	0.32
PersonFirst	4.99	3.59	6.71	PersonFirst	4.99	3.59	6.71	NA
PersonSecond	2.65	0.75	6.21	PersonSecond	2.65	0.75	6.21	NA
PersonThird	0.87	0.92	1.67	PersonThird	0.87	0.92	1.67	NA
NumberPassive	0.40	0.29	0.56	NumberPassive	0.40	0.29	0.56	NA
NumberPlural	0.94	0.78	0.59	NumberPlural	0.94	0.78	0.59	NA
AgentGroup	0.11	0.55	0.23	AgentGroup	0.11	0.55	0.23	NA
AgentIndividual	0.70	0.47	0.87	AgentIndividual	0.70	0.47	0.87	NA

AgentGroup	<i>ajatella</i>	<i>harkita</i>	<i>pohtia</i>
Multinomial	0.1107492	0.5492622	0.2308943
Poisson	0.1107363	0.5492630	0.2309020



Interim conclusion of the comparisons

- For all practical purposes, the multinomial and Poisson models of the same polytomous outcome setting are equivalent
 - (though the fitted coefficients/odds and outcome values are not *exactly* the same)



Adding an *extralinguistic* predictor variable – *Section*

- 23 sections in Helsingin Sanomat (1995)
 - Politics, letters to the editor, sports, economy/business, etc.
 - Where the article was printed
- 2 sections in SFNET (2002-2003)
 - Where the author decided to post the contributions
 - *sfnet.keskustelu.ihmissuhteet*: personal relationships
 - *sfnet.keskustelu.politiikka*: politics



Adding an *extralinguistic* predictor variable

- From a binary (dummy-coded logical TRUE/FALSE) to a multinomial variable representation
 - Person (1ST, 2ND, 3RD)
 - Number ([SINGULAR], PLURAL, PASSIVE)
 - Agent/Subject (INDIVIDUAL, GROUP)
 - Section (Newspaper departments, Newsgroups)

#	Lexeme	Person	Number	Agent	Section
1	pohtia	NIL	Passive	NIL	hs95_MN [opinion/excerpts ← other newspapers]
2	harkita	Third	NIL	Group	hs95_YO [letters to the editor]
3	miettä	Third	NIL	Individual	hs95_SP [sports]
4	miettä	Third	NIL	Individual	hs95_PO [(Finnish) domestic politics]
5	ajatella	Third	NIL	Individual	sfnet_politiikka [newsgroup on politics]
6	ajatella	Third	NIL	Individual	hs95_SP



Poisson-multinomial link with *linguistic* and *extralinguistic* predictor variables

- From multinomial variable representation to outcome-wise *counts* for each *unique* multinomial variable combination (*context*)

#	Proportion	Count	Lexeme	Person	Number	Agent	Section	Observation
1	0.5	1	ajatella	First	NIL	Individual	hs95_AE	1
2	0.0	0	harkita	First	NIL	Individual	hs95_AE	1
3	0.5	1	miettä	First	NIL	Individual	hs95_AE	1
4	0.0	0	pohtia	First	NIL	Individual	hs95_AE	1
5	0.8	4	ajatella	First	NIL	Individual	hs95_AK	2
6	0.0	0	harkita	First	NIL	Individual	hs95_AK	2
7	0.2	1	miettä	First	NIL	Individual	hs95_AK	2
8	0.0	0	pohtia	First	NIL	Individual	hs95_AK	2



Specifying and fitting equivalent multinomial/Poisson models

- Multinomial model (*baseline category*)
 - THINK.model_multinom.Person_Number_Agent_Section.demo
<- multinom(**Lexeme ~ Person + Number + Agent + Section**,
data=THINK.Lexeme_Person_Number_Agent_Section.multinomial_data_demo)
- Poisson model
 - THINK.model_poisson.Person_Number_Agent_Section.demo
<- glm(**Count ~ Observation + Lexeme + Lexeme:Person + Lexeme:Number + Lexeme:Agent + Lexeme:Section**,
data=THINK.Lexeme_Person_Number_Agent_Section.counts_demo,
family=**poisson**)



Comparing the equivalent multinomial/Poisson models

- Person=THIRD & Number=PLURAL & Agent=INDIVIDUAL & Section=sfnet_politiikka (altogether 41 cases in the data)

#	Proportion	Count	Lex.	Person	Number	Agent	Section	Obs.
973	0.78048780	32	ajatella	Third	Plural	Individual	sfnet_politiikka	244
974	0.04878049	2	harkita	Third	Plural	Individual	sfnet_politiikka	244
975	0.14634146	6	mieltä	Third	Plural	Individual	sfnet_politiikka	244
976	0.02439024	1	pohtia	Third	Plural	Individual	sfnet_politiikka	244



Comparing the equivalent multinomial/Poisson models

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>	Σ
32	2	6	1	41

- Original proportions

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.78048780	0.04878049	0.14634146	0.02439024

- Multinomial model (with *multinom/nnet*)

```
predict(THINK.model_multinom.Person_Number_Agent_Section.demo,  
newdata=data.frame(matrix(c("Third", "Plural", "Individual", "sfnet_politiikka"),1,,  
dimnames=list(NULL,c("Person","Number","Agent","Section")))), type="probs")
```

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.59814562	0.05051883	0.19773692	0.15359864



Comparing the equivalent multinomial/Poisson models

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>	Σ
32	2	6	1	41

- Poisson model count estimates

```
fitted(THINK.model_poisson.Person_Number_Agent_Section.demo)[973:976]
```

973	974	975	976
24.525218	2.070774	8.107170	6.296838

Corresponding proportion estimates

```
fitted(THINK.model_poisson.Person_Number_Agent_Section.demo)[973:976]/41
```

<i>[ajatella]</i>	<i>[harkita]</i>	<i>[mieltiä]</i>	<i>[pohtia]</i>
0.59817606	0.05050668	0.19773584	0.15358142



Specifying and fitting a Poisson equivalent mixed effects model

- ***lmer*(Count ~ (1|**Observation**) + Lexeme + Lexeme:Person + Lexeme:Number + Lexeme:Agent + (1|**Section**), data=THINK.Lexeme_Person_Number_Agent_Section.counts_demo, **family=poisson**)**
 - Observation as mixed effect in order to get coefficients properly estimated



Comparing overall fits between *fixed* and *mixed* poisson equivalent multinomial models

- Fixed model

- Section as fixed predictor (without interactions)

Residual Deviance	AIC
890.2	3237

- Mixed model

- Section (and Observation) as random predictor

AIC	BIC	logLik	Deviance
2131	2298	-1032	2063

- Conclusion

- Fixed → Mixed: Deviance:↑ – AIC:↓



Comparing the equivalent *fixed/mixed* Poisson models

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>	Σ
32	2	6	1	41

- Poisson model count estimates

```
fitted(THINK.model_poisson_lmer.Person_Number_Agent_Section.demo)[973:976]
```

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>	Σ
18.234855	3.621799	8.009096	10.226739	40.09249

- Corresponding proportion estimates

```
fitted(THINK.model_poisson.Person_Number_Agent_Section.demo)[973:976]/40.092
```

<i>ajatella</i>	<i>harkita</i>	<i>mieltiä</i>	<i>pohtia</i>
0.4548197	0.0903361	0.1997655	0.2550787

- Original observed proportions

0.78048780	0.04878049	0.14634146	0.02439024
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Comparing the equivalent *fixed/mixed* Poisson models

- Original counts

<i>ajatella</i>	<i>harkita</i>	<i>miettä</i>	<i>pohtia</i>	Σ
32	2	6	1	41

- Original observed proportions

<i>ajatella</i>	<i>harkita</i>	<i>miettä</i>	<i>pohtia</i>
0.78048780	0.04878049	0.14634146	0.02439024

- Fixed-model Poisson-count based proportion estimates

<i>ajatella</i>	<i>harkita</i>	<i>miettä</i>	<i>pohtia</i>
0.59817606	0.05050668	0.19773584	0.15358142

- Mixed-model Poisson-count based proportion estimates

<i>ajatella</i>	<i>harkita</i>	<i>miettä</i>	<i>pohtia</i>
0.4548197	0.0903361	0.1997655	0.2550787



Coefficients/odds for mixed effects model

VariableLevel	Estimate	Odds	Pr(> z)	Sign.
(Intercept)	1.02370	2.78	1.86e-05	***
Lexemeajatella:AgentGroup	-2.13995	0.12	6.22e-13	***
Lexemeajatella:AgentIndividual	0.11058	1.12	0.640741	
Lexemeajatella:NumberPassive	-0.06815	0.93	0.800813	
Lexemeajatella:NumberPlural	-0.85629	0.42	1.49e-05	***
Lexemeajatella:PersonFirst	-0.05036	0.95	0.856623	
Lexemeajatella:PersonSecond	-0.59566	0.55	0.086161	.
Lexemeajatella:PersonThird	0.43142	1.54	0.061095	.
Lexemeharkita	-1.05886	0.35	4.41e-16	***
Lexemeharkita:AgentGroup	-0.51877	0.60	0.088230	.
Lexemeharkita:AgentIndividual	-0.26797	0.76	0.350980	
Lexemeharkita:NumberPassive	-0.37416	0.69	0.227556	
Lexemeharkita:NumberPlural	-1.06220	0.35	3.19e-05	***
Lexemeharkita:PersonFirst	-0.39981	0.67	0.236232	
Lexemeharkita:PersonSecond	-1.87581	0.15	0.000111	***
Lexemeharkita:PersonThird	0.45837	1.58	0.085692	.
VariableLevel	Estimate	Odds	Pr(> z)	Sign.



Coefficients/odds for mixed effects model

VariableLevel	Estimate	Odds	Pr(> z)	Sign.
Lexememiettä	-1.23528	0.29	<2e-16	***
Lexememiettä:AgentGroup	-1.38180	0.25	2.05e-05	***
Lexememiettä:AgentIndividual	0.35441	1.43	0.192578	
Lexememiettä:NumberPassive	0.27640	1.32	0.354561	
Lexememiettä:NumberPlural	-1.32633	0.27	9.29e-10	***
Lexememiettä:PersonFirst	0.23689	1.27	0.446195	
Lexememiettä:PersonSecond	0.24092	1.27	0.516394	
Lexememiettä:PersonThird	1.07016	2.92	3.87e-05	***
Lexemepohtia	-1.08468	0.34	<2e-16	***
Lexemepohtia:AgentGroup	-0.05414	0.95	0.845087	
Lexemepohtia:AgentIndividual	0.35307	1.42	0.173795	
Lexemepohtia:NumberPassive	0.79151	2.21	0.004750	**
Lexemepohtia:NumberPlural	-0.80661	0.45	0.000211	***
Lexemepohtia:PersonFirst	-1.58039	0.21	3.59e-06	***
Lexemepohtia:PersonSecond	-1.49522	0.22	0.000158	***
Lexemepohtia:PersonThird	0.64560	1.91	0.007427	**



What about the full-scale model?

Variant of the prior model selected on the basis of extensive univariate analysis

→ Altogether 48 contextual feature variables:

- Morphological features pertaining to the node-verb or the entire verb-chain they are components of (10)
- semantic characterizations of verb-chains (6)
- syntactic argument types, without any subtypes (10)
- Syntactic arguments combined with their semantic and structural subtypes (20)
- extra-linguistic features (2)



Tentative overall mixed effects model

```
> summary(THINK.Lexeme_Person_Number_Agent_Patient_Manner_Location_Time_Modality_Section_Author.multinomial_data)
```

Lexeme	Person	Number	Agent	Patient
ajatella:1492	NIL :1056	NIL :2561	NIL : 897	NIL :1011
harkita : 387	Z_ANL_FIRST : 509	Z_ANL_PASS: 457	SX_AGE.SEM_GROUP : 256	SX_PAT.SEM_ABSTRACTION : 698
miettä : 812	Z_ANL_SECOND: 320	Z_ANL_PLUR: 386	SX_AGE.SEM_INDIVIDUAL:2251	SX_PAT.SEM_ACTIVITY : 489
pohtia : 713	Z_ANL_THIRD :1519			SX_PAT.INDIRECT_QUESTION : 438
				SX_LX_että_CS.SX_PAT : 396
				SX_PAT.SEM_INDIVIDUAL_GROUP: 109
				(Other) : 263

Manner	Location	Time	Modality1	Modality2
NIL :2902	NIL :3127	NIL :2763	NIL :2572	NIL :3107
SX_MAN.SEM_AGREEMENT: 48	SX_LOC: 277	SX_TMP.SEM_DEFINITE : 158	SX_VCH.SEM_NECESSITY : 487	SX_VCH.SEM_ACCIDENTAL: 44
SX_MAN.SEM_FRAME : 64		SX_TMP.SEM_INDEFINITE: 483	SX_VCH.SEM_POSSIBILITY: 345	SX_VCH.SEM_EXTERNAL : 79
SX_MAN.SEM_GENERIC : 113				SX_VCH.SEM_TEMPORAL : 119
SX_MAN.SEM_JOINT : 62				SX_VCH.SEM_VOLITION : 55
SX_MAN.SEM_NEGATIVE : 38				
SX_MAN.SEM_POSITIVE : 177				

Section	Author
Z_EXTRA_DE_ihmissuhteet:1028	Z_EXTRA_AU_sfnet_721 : 146
Z_EXTRA_DE_politiikka : 626	Z_EXTRA_AU_hs95_latomon_käyttäjäprofiili: 100
Z_EXTRA_DE_hs95_SP : 243	Z_EXTRA_AU_sfnet_331 : 99
Z_EXTRA_DE_hs95_KU : 224	Z_EXTRA_AU_sfnet_92 : 79
Z_EXTRA_DE_hs95_YO : 189	Z_EXTRA_AU_sfnet_966 : 77
Z_EXTRA_DE_hs95_UL : 114	Z_EXTRA_AU_sfnet_345 : 73
(Other) : 980	(Other) :2830



Tentative overall mixed effects model

```
lmer(Count ~ (1|Observation) + Lexeme +  
Lexeme:Person + Lexeme:Number +  
Lexeme:Agent + Lexeme:Patient +  
Lexeme:Manner + Lexeme:Location +  
Lexeme:Time + Lexeme:Modality1 +  
Lexeme:Modality2 + (1|Section),  
data=THINK.Lexeme_Person_..._Section.  
counts, family=poisson)
```



Tentative fitted results

- AIC BIC logLik deviance
5956 6857 -2848 5696

- Random effects:

Groups	Name	Variance	Std.Dev.
	Observation (Intercept)	0.150176	0.38753
	Section (Intercept)	0.075796	0.27531

- Number of obs: 7544, groups: Observation, 1886; Section, 25



Adjusted coefficients/odds

Predictor	Odds
(Intercept)	1.9
Lexemeharkita	0.12
Lexememiettiä	0.17
Lexemepohtia	0.12
Lexemeajatella:PersonZ_ANL_FIRST	(1.2)
Lexemeharkita:PersonZ_ANL_FIRST	(0.96)
Lexememiettiä:PersonZ ANL FIRST	1.6
Lexemepohtia:PersonZ_ANL_FIRST	0.24
Lexemeajatella:PersonZ_ANL_SECOND	(1.1)
Lexemeharkita:PersonZ_ANL_SECOND	0.39
Lexememiettiä:PersonZ_ANL_SECOND	2.5
Lexemepohtia:PersonZ_ANL_SECOND	0.47
Lexemeajatella:PersonZ_ANL_THIRD	1.3
Lexemeharkita:PersonZ_ANL_THIRD	(1.1)
Lexememiettiä:PersonZ_ANL_THIRD	1.7
Lexemepohtia:PersonZ_ANL_THIRD	(0.98)
Lexemeajatella:NumberZ_ANL_PASS	(0.86)
Lexemeharkita:NumberZ_ANL_PASS	0.64
Lexememiettiä:NumberZ_ANL_PASS	(1)
Lexemepohtia:NumberZ_ANL_PASS	(1.3)
Lexemeajatella:NumberZ_ANL_PLUR	0.66
Lexemeharkita:NumberZ_ANL_PLUR	(0.74)
Lexememiettiä:NumberZ_ANL_PLUR	0.5
Lexemepohtia:NumberZ ANL PLUR	(0.92)



Adjusted coefficients/odds ...

Lexemeajatella:AgentSX_AGE.SEM_GROUP	0.25
Lexemeharkita:AgentSX_AGE.SEM_GROUP	(0.89)
Lexememiettä:AgentSX_AGE.SEM_GROUP	0.45
Lexemepohtia:AgentSX_AGE.SEM_GROUP	1.8
Lexemeajatella:AgentSX_AGE.SEM_INDIVIDUAL	(0.79)
Lexemeharkita:AgentSX_AGE.SEM_INDIVIDUAL	(0.67)
Lexememiettä:AgentSX_AGE.SEM_INDIVIDUAL	(1)
Lexemepohtia:AgentSX_AGE.SEM_INDIVIDUAL	(1.2)



Adjusted coefficients/odds ...

Lexemeajatella:PatientSX_LX_että_CS.SX_PAT	(1.1)
Lexemeharkita:PatientSX_LX_että_CS.SX_PAT	0.27
Lexememiettiinä:PatientSX_LX_että_CS.SX_PAT	0.51
Lexemepohtia:PatientSX_LX_että_CS.SX_PAT	0.54
Lexemeajatella:PatientSX_PAT.DIRECT_QUOTE	0.039
Lexemeharkita:PatientSX_PAT.DIRECT_QUOTE	(0)
Lexememiettiinä:PatientSX_PAT.DIRECT_QUOTE	2.7
Lexemepohtia:PatientSX_PAT.DIRECT_QUOTE	7.7
Lexemeajatella:PatientSX_PAT.INDIRECT_QUESTION	0.13
Lexemeharkita:PatientSX_PAT.INDIRECT_QUESTION	(0.71)
Lexememiettiinä:PatientSX_PAT.INDIRECT_QUESTION	1.9
Lexemepohtia:PatientSX_PAT.INDIRECT_QUESTION	2.1
Lexemeajatella:PatientSX_PAT.INFINITIVE	0.66
Lexemeharkita:PatientSX_PAT.INFINITIVE	(0.64)
Lexememiettiinä:PatientSX_PAT.INFINITIVE	(0)
Lexemepohtia:PatientSX_PAT.INFINITIVE	(0.14)
Lexemeajatella:PatientSX_PAT.PARTICIPLE	(0.75)
Lexemeharkita:PatientSX_PAT.PARTICIPLE	(0.62)
Lexememiettiinä:PatientSX_PAT.PARTICIPLE	(0)
Lexemepohtia:PatientSX_PAT.PARTICIPLE	0.23



Adjusted coefficients/odds ...

Lexemeajatella:PatientSX_PAT.SEM_ABSTRACTION	0.41
Lexemeharkita:PatientSX_PAT.SEM_ABSTRACTION	(0.99)
Lexememiettiinä:PatientSX_PAT.SEM_ABSTRACTION	(1.2)
Lexemepohtia:PatientSX_PAT.SEM_ABSTRACTION	2.6
Lexemeajatella:PatientSX_PAT.SEM_ACTIVITY	0.25
Lexemeharkita:PatientSX_PAT.SEM_ACTIVITY	4.7
Lexememiettiinä:PatientSX_PAT.SEM_ACTIVITY	0.64
Lexemepohtia:PatientSX_PAT.SEM_ACTIVITY	1.5
Lexemeajatella:PatientSX_PAT.SEM_COMMUNICATION	0.11
Lexemeharkita:PatientSX_PAT.SEM_COMMUNICATION	(1)
Lexememiettiinä:PatientSX_PAT.SEM_COMMUNICATION	(0.91)
Lexemepohtia:PatientSX_PAT.SEM_COMMUNICATION	(1.2)
Lexemeajatella:PatientSX_PAT.SEM_INDIVIDUAL_GROUP	0.71
Lexemeharkita:PatientSX_PAT.SEM_INDIVIDUAL_GROUP	(0.57)
Lexememiettiinä:PatientSX_PAT.SEM_INDIVIDUAL_GROUP	0.28
Lexemepohtia:PatientSX_PAT.SEM_INDIVIDUAL_GROUP	0.22



Adjusted coefficients/odds ...

Lexemeajatella:MannerSX_MAN.SEM_AGREEMENT	(0.73)
Lexemeharkita:MannerSX_MAN.SEM_AGREEMENT	(0)
Lexememiettiä:MannerSX_MAN.SEM_AGREEMENT	0.051
Lexemepohtia:MannerSX_MAN.SEM_AGREEMENT	0.16
Lexemeajatella:MannerSX_MAN.SEM_FRAME	0.63
Lexemeharkita:MannerSX_MAN.SEM_FRAME	0.16
Lexememiettiä:MannerSX_MAN.SEM_FRAME	0.15
Lexemepohtia:MannerSX_MAN.SEM_FRAME	0.53
Lexemeajatella:MannerSX_MAN.SEM_GENERIC	(0.88)
Lexemeharkita:MannerSX_MAN.SEM_GENERIC	(0)
Lexememiettiä:MannerSX_MAN.SEM_GENERIC	0.095
Lexemepohtia:MannerSX_MAN.SEM_GENERIC	(0)
Lexemeajatella:MannerSX_MAN.SEM_JOINT	0.27
Lexemeharkita:MannerSX_MAN.SEM_JOINT	(0.66)
Lexememiettiä:MannerSX_MAN.SEM_JOINT	(0.69)
Lexemepohtia:MannerSX_MAN.SEM_JOINT	0.42
Lexemeajatella:MannerSX_MAN.SEM_NEGATIVE	0.46
Lexemeharkita:MannerSX_MAN.SEM_NEGATIVE	0.25
Lexememiettiä:MannerSX_MAN.SEM_NEGATIVE	0.19
Lexemepohtia:MannerSX_MAN.SEM_NEGATIVE	0.12
Lexemeajatella:MannerSX_MAN.SEM_POSITIVE	0.45
Lexemeharkita:MannerSX_MAN.SEM_POSITIVE	(0.8)
Lexememiettiä:MannerSX_MAN.SEM_POSITIVE	0.48
Lexemepohtia:MannerSX_MAN.SEM_POSITIVE	0.49



Adjusted coefficients/odds ...

Lexemeajatella:LocationSX_LOC	0.33
Lexemeharkita:LocationSX_LOC	0.4
Lexememiettiinä:LocationSX_LOC	0.67
Lexemepohtia:LocationSX_LOC	1.5
Lexemeajatella:TimeSX_TMP.SEM_DEFINITE	0.37
Lexemeharkita:TimeSX_TMP.SEM_DEFINITE	0.46
Lexememiettiinä:TimeSX_TMP.SEM_DEFINITE	0.57
Lexemepohtia:TimeSX_TMP.SEM_DEFINITE	(1.1)
Lexemeajatella:TimeSX_TMP.SEM_INDEFINITE	0.5
Lexemeharkita:TimeSX_TMP.SEM_INDEFINITE	0.67
Lexememiettiinä:TimeSX_TMP.SEM_INDEFINITE	0.74
Lexemepohtia:TimeSX_TMP.SEM_INDEFINITE	0.58



Adjusted coefficients/odds ...

Lexemeajatella:Modality1SX_VCH.SEM_NECESSITY	0.36
Lexemeharkita:Modality1SX_VCH.SEM_NECESSITY	(1.1)
Lexememiettiä:Modality1SX_VCH.SEM_NECESSITY	(1.1)
Lexemepohtia:Modality1SX_VCH.SEM_NECESSITY	0.64
Lexemeajatella:Modality1SX_VCH.SEM_POSSIBILITY	0.61
Lexemeharkita:Modality1SX_VCH.SEM_POSSIBILITY	(0.79)
Lexememiettiä:Modality1SX_VCH.SEM_POSSIBILITY	0.5
Lexemepohtia:Modality1SX_VCH.SEM_POSSIBILITY	0.43



Adjusted coefficients/odds ...

Lexemeajatella:Modality2SX_VCH.SEM_ACCIDENTAL	0.66
Lexemeharkita:Modality2SX_VCH.SEM_ACCIDENTAL	(0)
Lexememiettiinä:Modality2SX_VCH.SEM_ACCIDENTAL	0.19
Lexemepohtia:Modality2SX_VCH.SEM_ACCIDENTAL	0.17
Lexemeajatella:Modality2SX_VCH.SEM_EXTERNAL	(0.94)
Lexemeharkita:Modality2SX_VCH.SEM_EXTERNAL	(0.52)
Lexememiettiinä:Modality2SX_VCH.SEM_EXTERNAL	0.53
Lexemepohtia:Modality2SX_VCH.SEM_EXTERNAL	(0.63)
Lexemeajatella:Modality2SX_VCH.SEM_TEMPORAL	0.25
Lexemeharkita:Modality2SX_VCH.SEM_TEMPORAL	0.12
Lexememiettiinä:Modality2SX_VCH.SEM_TEMPORAL	(0.81)
Lexemepohtia:Modality2SX_VCH.SEM_TEMPORAL	(1)
Lexemeajatella:Modality2SX_VCH.SEM_VOLITION	0.32
Lexemeharkita:Modality2SX_VCH.SEM_VOLITION	0.36
Lexememiettiinä:Modality2SX_VCH.SEM_VOLITION	0.53
Lexemepohtia:Modality2SX_VCH.SEM_VOLITION	0.48



The End – Everything you never wanted to know about Finnish verbs of thinking

- Thank you!
- Questions, comments, suggestions?!?