Generalized Mixed Linear Models Practical 1

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December 3, 2014

Post-operative sore throat study

The aim of a study carried out at the Royal Berkshire Hospital, Reading, in 2004 was to investigate the incidence of sore throat in patients who had undergone orthopaedic, gynaecological, genitourinary or general surgery. Of particular interest was whether the occurrence of a sore throat was affected by the method used to deliver anaesthetic gas, and patients were allocated to one of three types of airway device, namely the laryngeal mask airway (LMA), the endo-tracheal tube (ETT), and the traditional face mask (FM). The decision on which of the three types of device to use for a particular patient was made by the consultant anaesthetist, and there were 12 anaesthetists involved.

The response variable was binary and concerned whether or not a patient experienced a sore throat in the 24 hour period following the operation. The values of certain explanatory variables were also recorded, including the age and sex of the patient, the duration of surgery, and, for LMA and ETT, whether or not the throat was lubricated before the airway was inserted.

The following eight variables are contained in the datafile **sorethroat.dta**.

PATIENT	Patient number (1 - 947)
AGE	Age of patient in years
SEX	Sex of patient $(0 = \text{male}, 1 = \text{female})$
DURATION	Duration of surgery in minutes
AIRWAY	Type of airway used (LMA, ETT or FM)
LUBRIC	Lubrication used in inserting mask $(0 = no, 1 = yes, . = n/a)$
CONSULT	Consultant anaesthetist $(1 - 12)$
SORE	Occurrence of sore throat $(0 = no, 1 = yes)$

- 1. How do the three types of airway compare in terms of the incidence of post-operative sore throat?
- 2. Is there any evidence that the probability that a consultant selects the face mask (FM) is dependent upon the age and sex of the patient or the duration of surgery?

Solution

Evidently, we need to evaluate the risks of FM(1), LMA(0) and ETT(2). We choose LMA as reference (arbitrary). We get the following. Clearly, FM has the highest preventive effect.

Logistic regre	ession		Numbe LR ch Prob	er of obs ui2(2)	=	947 26.23	
Log likelihood		Pseud	lo R2	=	0.0324		
sore	Odds Ratio	Std. Err.	z	P> z	[95%	Conf.	Interval]
airway							
1	.1008403	.0726669	-3.18	0.001	.0245	611	.4140202
2	1.681319	.4974374	1.76	0.079	.9414	826	3.002533
_cons	. 1944444	.0189319	-16.82	0.000	. 1606	643	.235327

But how is this influenced by other covariate such as gender and age?

Logistic regre	Number LR chi	of obs	=	947 47.52			
Log likelihood	= -381.6245	1		Prob > Pseudo	• Ch12	=	0.0000
sore	Odds Ratio	Std. Err.	z	P> z	[95%	Conf.	Interval]
airway							
1	.0941757	.0679809	-3.27	0.001	.0228	822	.3875968
2	1.830303	.5528838	2.00	0.045	1.012	508	3.308624
age	1.000994	.0061108	0.16	0.871	.9890	885	1.013043
sex	2.538882	.5415277	4.37	0.000	1.671	424	3.856546
_cons	.0977297	.0341282	-6.66	0.000	. 0492	922	.1937649

We see that gender is important, but not age. We also look at the effect of including duration:

Logistic regression Log likelihood = -381.26135				Numbe LR ch Prob Pseud	r of obs i2(4) > chi2 o R2	= = =	947 48.25 0.0000 0.0595
sore	Odds Ratio	Std. Err.	z	P> z	[95%	 Conf.	Interval]
airway							
1	.0872864	.0634804	-3.35	0.001	.0209	844	.3630753
2	1.97484	.6217158	2.16	0.031	1.065	511	3.660208
sex	2.554164	.5449203	4.40	0.000	1.681	313	3.880156
duration	.9966394	.003946	-0.85	0.395	.9889	354	1.004403
_cons	. 1151237	.0266415	-9.34	0.000	.0731	447	. 1811951

There is no effect of duration. Finally, we look at the consultant effect. This is more difficult as there are many consultants. Hence we need to perform a proper model evaluation.

Logistic regre Log likelihood	ession 1 = -377.4624	9		Numbo LR cl Prob Pseud	er of obs = hi2(13) = > chi2 = do R2 =	942 54.18 0.0000 0.0670
sore	Odds Ratio	Std. Err.	Z	P> z	05% Conf.	Interval
airway	 					
1	.1055685	.0774471	-3.06	0.002	.0250655	.4446233
2	1.909648	.6425182	1.92	0.055	.9875493	3.692731
sex	2.425342	. 5360854	4.01	0.000	1.572634	3.740402
consultant						
2	1.403994	.605088	0.79	0.431	.6032782	3.267479
3	.7801426	.490758	-0.39	0.693	.2273607	2.676902
4	1.103781	.4592449	0.24	0.812	.4883457	2.494814
5	.739868	.3498967	-0.64	0.524	.2928235	1.869401
6	1.05312	.4526899	0.12	0.904	.4535067	2.445524
7	1.760576	.7683945	1.30	0.195	.748432	4.141497
8	.7921757	.417397	-0.44	0.658	.2820485	2.224945
9	1.194777	.5105239	0.42	0.677	.5170962	2.760592
10	1.362303	.5826169	0.72	0.470	.589169	3.149978
11	1.002999	.6385044	0.00	0.996	.2880247	3.492781
12	1	(empty)				
_cons	.0937726	. 0343965	-6.45	0.000	.0456929	. 1924437

In this model, we see many specific consultant's effects, none of them is significant. But is there an overall consultant effect?

- We see that the log-likelihood is -377.46249 (AIC = 782.925, BIC = 850.797) for this model, whereas the model without consultant effect has log-likelihood of -381.63776 (AIC = 771.2755, BIC = 773.5353).
- The likelihood ratio statistic is -377.46249-(-381.63776) = 8.35062 which is not significant on a chi-square scale of 11 df (consultant 12 has no estimable effect).
- This is not the best way of dealing with categorical variables with many categories and we need to take this up again when discussing random effects.

Finally, we look at the question of FM selection and how this is affected by age and gender.

Logistic regre		Number	of obs	; =	947		
				LR chi	2(2)	=	3.40
				Prob >	chi2	=	0.1827
Log likelihood	1 = -326.09469	9		Pseudo	R2	=	0.0052
FM	Odds Ratio	Std. Err.	z	P> z	[95%	Conf.	Interval]
age	1.009157	.0068659	1.34	0.180	. 9957	895	1.022704
sex	1.344045	.2977323	1.33	0.182	.8706	5749	2.074778
_cons	.0658002	.025568	-7.00	0.000	. 0307	239	.1409221

We see that there is no age- or gender effect, but this result is heavily confounded by the consultant's effect as the next analysis shows.

Logistic regre	ssion			Numbe	er of obs =	651
				LR cl	hi2(8) =	91.45
				Prob	> chi2 =	0.0000
Log likelihood	= -240.2321	4		Pseud	do R2 =	0.1599
FM	Odds Ratio	Std. Err.	z	P> z	[95% Conf.	Interval]
age	1.020834	.0082594	2.55	0.011	1.004774	1.037152
sex	.7555842	.2001989	-1.06	0.290	.4495203	1.270037
consultant						
2	.6357229	.3307346	-0.87	0.384	.229314	1.762402
3	1	(empty)				
4	.3228287	.1981049	-1.84	0.065	.0969682	1.074768
5	4.703393	1.96275	3.71	0.000	2.075872	10.65668
6	1	(empty)				
7	1	(empty)				
8	1	(empty)				
9	4.219144	1.663601	3.65	0.000	1.948037	9.138008
10	.2630869	.1791228	-1.96	0.050	.0692721	.9991717
11	.2858109	.3071024	-1.17	0.244	.034791	2.34796
12	1	(empty)				
1						
_cons	.0518732	. 0296399	-5.18	0.000	.0169268	. 1589687

However, we see also see that this form of analysis has reached its limits since many individual consultant's effects are not estimable and, hence, this form of analysis suffers under loss of power.