

Mixed Models for Longitudinal Count Data

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Hedeker, D. & Gibbons, R.D. (2006). Longitudinal Data Analysis, chapter 12. Wiley.

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Dependent variable is a count

- number of hospitalizations
- number of service uses
- number of headaches (or some kind of disease symptom)
- number of times that an event occurs

Poisson distribution is often used to model count data

$$\Pr(y \mid \mu) = \frac{\exp(-\mu)\mu^y}{y!} \quad \text{for } y = 0, 1, 2, \dots$$

μ is the expected count (per unit of time)

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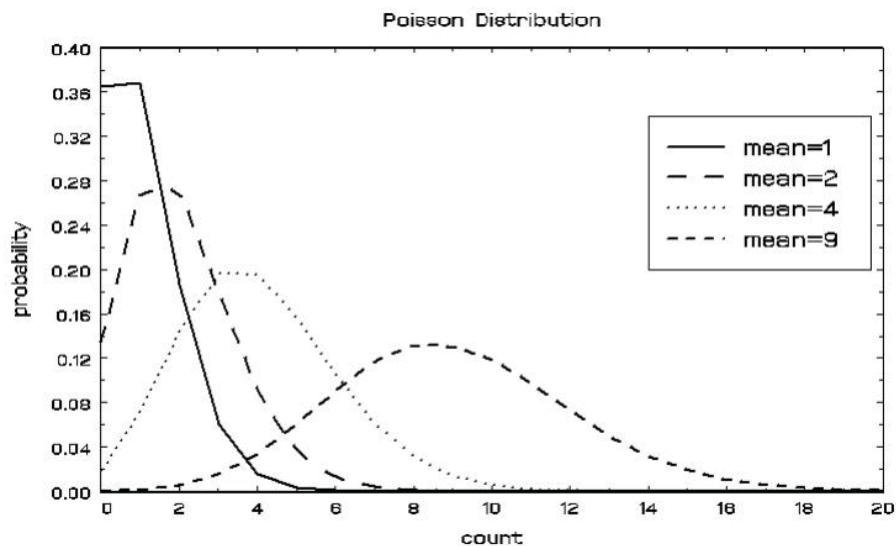
Can't I just analyze as continuous normal?

- count distribution is too skewed to satisfy normality (incorrect test results)
- normal model does not necessarily prevent negative estimated counts

Can't I just dichotomize count (0 vs >0) and analyze using logistic regression?

- loss of information resulting in under-powered tests
- is 1 event really equal to 100 events?

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- expected number of counts (per unit of time), strictly positive
- as mean increases, probability of 0s decreases, distribution approximates normal
- mean equals the variance (if variance is greater, then overdispersion)

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Reading materials and examples

- Cameron & Trivedi (1998) *Regression analysis of count data*, Cambridge Univ Press
- Long (1997) *Regression models for categorical and limited dependent variables*, Sage
- Elhai, Calhoun, & Ford (2008) Statistical procedures for analyzing mental health services data. *Psychiatry Research*, 160, 129-136.
- Walters (2007) Using Poisson Class Regression To Analyze Count Data in Correctional and Forensic Psychology. *Criminal Justice and Behavior*, 34, 1659-1674.
- Gagnon, Doron-LaMarca, Bell, O'Farrell, & Taft (2008) Poisson regression for modeling count and frequency outcomes in trauma research. *Journal of Traumatic Stress*, 21, 448-454.
- Supermix (for Windows):
 - www.ssicentral.com/index.php/products/supermix/downloads
 - www.ssicentral.com/index.php/products/supermix/examples
 - in Supermix (even the free student version), from Help menu, select “Contents,” “Examples from SMIX primer,” “Count outcomes”

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Notation is our friend!

- $i = 1, \dots, N$ level-2 units (clusters or subjects)
- $j = 1, \dots, n_i$ level-1 units (subjects or multiple observations)
- y_{ij} is the value of the count outcome, the number of events (y_{ij} can equal 0, 1, ...)
- t_{ij} is the length of time during which the events are recorded
 - can be equal ($t_{ij} = t$): all observations are based on the same period of time, and the number of events within that same time period is of interest
 - can vary (t_{ij}): observations are based on varying periods of time; this should be accounted for when modeling the number of events within the varying time periods

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Right-hand side of model

$$\mathbf{x}_{ij}'\boldsymbol{\beta} + \mathbf{z}_{ij}'\mathbf{v}_i$$

- \mathbf{x}_{ij} are covariates
 - at level-1, level-2, or cross-level interactions
 - can include polynomials, dummy variables, interactions, ...
- $\boldsymbol{\beta}$ are the regression coefficients for the covariates
- \mathbf{z}_{ij} are the random effect variable(s)
 - usually just an intercept for clustered data
 - often an intercept and time for longitudinal data
- \mathbf{v}_i are the random effects $\sim N(0, \boldsymbol{\Sigma}_v)$
 - how cluster i influences the observations within the cluster
 - how a subject starts and progresses across time

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Mixed-effects Poisson Regression Model

The mixed-effects Poisson regression model indicates the expected number of counts in t_{ij} as:

$$E(y_{ij}) = \mu_{ij} = t_{ij} \exp[\mathbf{x}_{ij}'\boldsymbol{\beta} + \mathbf{z}_{ij}'\mathbf{v}_i]$$

or

$$\log(\mu_{ij}) = \log(t_{ij}) + [\mathbf{x}_{ij}'\boldsymbol{\beta} + \mathbf{z}_{ij}'\mathbf{v}_i]$$

$$\log(\mu_{ij}) - \log(t_{ij}) = \mathbf{x}_{ij}'\boldsymbol{\beta} + \mathbf{z}_{ij}'\mathbf{v}_i$$

$$\log[\mu_{ij}/t_{ij}] = \mathbf{x}_{ij}'\boldsymbol{\beta} + \mathbf{z}_{ij}'\mathbf{v}_i$$

- link function for Poisson regression is the log link
- t_{ij} is sometimes called an offset or exposure variable
- $\exp \boldsymbol{\beta}$ = incidence or event rate ratio

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Simplest Poisson Regression example

(no random effects, no offset, dichotomous regressor)

Data and description: <https://stats.idre.ucla.edu/stata/dae/negative-binomial-regression/>

- School attendance data on 316 high school juniors
- Response variable is days absent (**daysabs**, range is 0 to 45)
- **male** is an indicator of student gender (0=F, 1=M)
 - **daysabs** mean for females = 6.6975
 - **daysabs** mean for males = 4.4877 (M to F ratio = .7281)
- $E(\text{daysabs}_i) = \mu_i = \exp(\beta_0 + \beta_1 \text{male}_i)$

$\exp \beta_0 = \text{mean for females} = 6.6975$ ($\hat{\beta}_0 = \log 6.6975 = 1.9017$)

$\exp(\beta_0 + \beta_1) = \exp(\beta_0) \times \exp(\beta_1) = \text{mean for males} = 4.4877$

$\exp(\hat{\beta}_1) = 4.4877 / \exp(\hat{\beta}_0) = 4.4877 / 6.6975 = .7281$ (M to F ratio)

($\hat{\beta}_1 = \log .7281 = -.3173$)

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Longitudinal example - Aspartame and headaches

- data from McKnight and Van Den Eeden (1993) *Statistics in Medicine*, also Van Den Eeden *et al.*, (1994) *Neurology*
- number of headaches in a two treatment, multiple period crossover trial
- number of headaches per week was repeatedly measured for 27 subjects
- Following a seven-day placebo run-in period, subjects received either aspartame or placebo in four seven-day treatment periods according to a double-blind crossover treatment design
- Each treatment period was separated by a washout day

Table 4. The number of headaches, treatment days, and belief about aspartame effects by subject and treatment period

Treatment order*	ID	Run-in		Treatment period†								Belief‡
		HA	Days	1		2		3		4		
				HA	Days	HA	Days	HA	Days	HA	Days	
APAP	2	0	7	5	7	2	7	—	—	—	—	V
	5	3	7	0	7	2	7	0	7	0	7	N
	13	7	7	7	7	7	7	6	7	7	7	N
	16	1	7	3	7	1	7	—	—	—	—	V
	19	0	7	1	7	1	7	0	7	1	7	V
	23	7	7	2	7	3	7	3	7	2	7	S
	25	1	7	6	7	1	7	7	7	0	7	V
APPA	1	3	7	0	7	3	7	1	7	0	7	N
	3	2	7	2	7	3	7	2	7	2	7	V
	6	1	7	1	7	0	7	3	7	1	7	S
	9	2	7	2	7	0	7	1	5	—	—	V
	17	4	7	1	1	—	—	—	—	—	—	S
	18	0	7	1	7	1	7	1	7	0	7	V
	21	1	7	2	7	3	7	3	7	6	7	V
	22	2	7	1	7	0	7	0	7	1	7	S
PAPA	7	1	7	1	7	4	7	2	7	3	7	N
	10	0	7	0	7	0	7	0	7	0	7	S
	11	0	7	3	7	1	7	0	7	1	3	S
	14	2	7	2	7	1	7	0	7	2	7	N
	24	1	7	0	7	1	7	0	7	2	7	V
	27	3	7	3	7	3	7	0	7	2	4	S
PAAP	4	0	7	0	7	0	7	0	7	0	7	V
	8	1	7	1	7	0	7	1	7	1	2	S
	12	0	7	0	7	5	7	0	7	0	7	S
	15	0	7	3	7	2	7	1	7	1	7	S
	20	1	7	6	7	1	2	—	—	—	—	V
	26	0	7	1	7	—	—	—	—	—	—	S

* A = aspartame; P = placebo. Run-in period used placebo capsules.
† HA = number of days on which subject reported a headache; Days = number of days subject participated during that period.
‡ Belief about how strongly subject felt aspartame caused headaches was asked prior to the start of the study. V = very sure; S = somewhat sure; N = not very sure or do not know.

ID	Headache	Period1	Period2	Period3	Period4	DrugAsp	NPeriods	NTDays
2	0	0	0	0	0	0	3	7
5	3	1	0	0	0	1	3	7
13	7	0	1	0	0	0	3	7
16	1	0	0	0	0	0	5	7
19	0	1	0	0	0	1	5	7
23	7	0	1	0	0	0	5	7
25	1	0	0	1	0	1	5	7
1	3	0	0	0	1	0	5	7
3	2	0	0	0	0	0	5	7
6	1	0	0	0	0	1	5	7
9	2	0	1	0	0	0	5	7
17	4	0	0	0	1	0	5	7
18	0	0	0	0	0	0	5	7
21	1	0	1	0	0	1	5	7
22	2	0	0	0	0	0	5	7
7	1	0	0	1	0	1	5	7
10	0	0	0	0	0	0	5	7
11	0	1	0	0	0	1	5	7
14	2	0	0	0	1	0	5	7
24	1	0	0	0	0	0	5	7
27	3	1	0	0	1	0	5	7
4	0	0	0	0	0	0	5	7
8	1	0	0	0	0	1	5	7
12	0	0	1	0	0	0	5	7
15	0	3	0	0	0	1	5	7
20	1	0	0	0	0	0	5	7
26	0	1	0	0	0	1	5	7

- ID = patient ID (27 patients in total)
- HeadAche = number of headaches during the week (0 to 7)
- Period1 = period 1 indicator (1 = first tx period, else 0)
- Period2 = period 2 indicator (1 = second tx period, else 0)
- Period3 = period 3 indicator (1 = third tx period, else 0)
- Period4 = period 4 indicator (1 = fourth tx period, else 0)
- DrugAsp = period-specific drug (0 = placebo, 1 = aspartame)
- Nperiods = number of periods person was observed (2 to 5)
- NTDays = number of treatment days in the period (1 to 7)

Stata example: aspart.do

```
cd "u:\Stata_long\"
log using aspart.log, replace
import excel using aspart.xlsx, sheet("aspartamin") firstrow clear

* summary statistics on all variables
summ

* Headache descriptives by DrugAsp
format Headache %8.4f
tabulate DrugAsp, summarize(Headache)

* some graphs
format Headache %2.0f
tway histogram Headache, discrete frequency
graph box Headache, over(DrugAsp, relabel(1 "Placebo" 2 "Aspartame"))
graph bar Headache, over(DrugAsp, relabel(1 "Placebo" 2 "Aspartame"))

* random intercept Poisson regression model
mepoisson Headache Period1 Period2 Period3 Period4 DrugAsp || ID:
```

```
. import excel using aspart.xlsx, sheet("aspartamin") firstrow clear
```

```
. summ
```

Variable	Obs	Mean	Std. Dev.	Min	Max
-----+					
ID	122	13.7377	7.783011	1	27
Headache	122	1.680328	1.886299	0	7
Period1	122	.2213115	.4168416	0	1
Period2	122	.204918	.4053062	0	1
Period3	122	.1803279	.3860457	0	1
-----+					
Period4	122	.1721311	.3790511	0	1
DrugAsp	122	.3852459	.4886602	0	1
NPeriods	122	4.721311	.7414783	2	5
NTDays	122	6.795082	.9529296	1	7

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```
. format Headache %8.4f
```

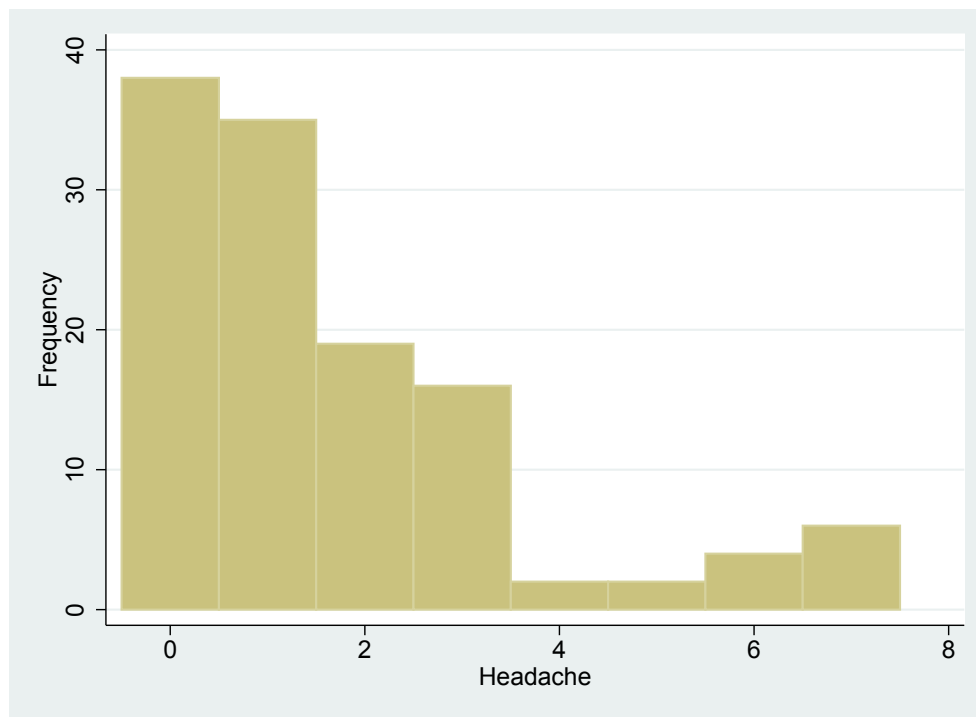
```
. tabulate DrugAsp, summarize(Headache)
```

		Summary of Headache		
DrugAsp		Mean	Std. Dev.	Freq.
-----+				
0		1.5333	1.8034	75
1		1.9149	2.0090	47
-----+				
Total		1.6803	1.8863	122

- Placebo mean = 1.5333
- Aspartame mean = 1.9149
- Rate ratio = $1.9149 / 1.5333 = 1.25$
- $\log(\text{Rate ratio}) = 0.223$

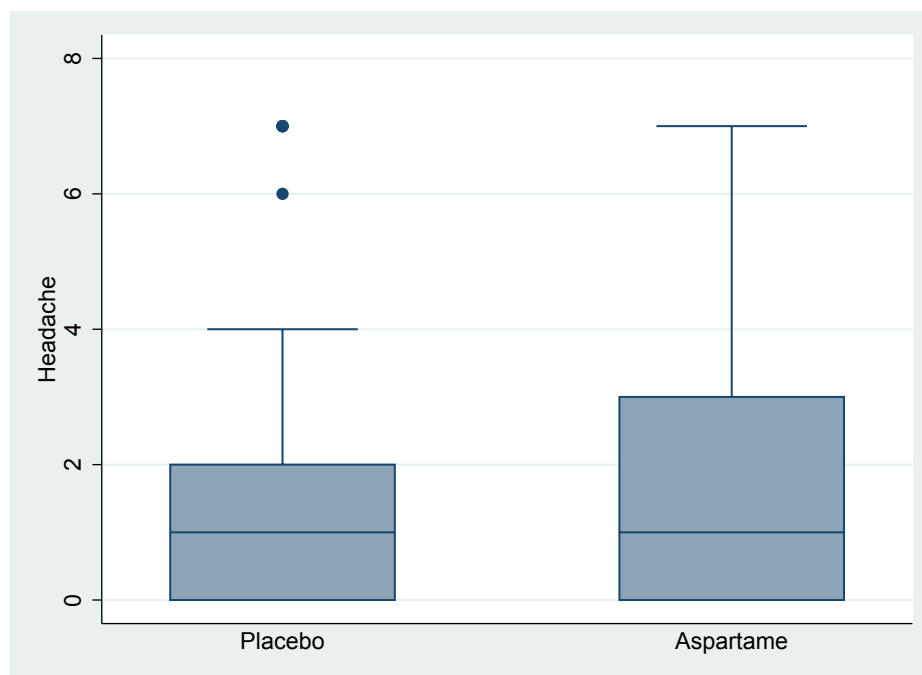
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```
. format Headache %2.0f
. twoway histogram Headache, discrete frequency
```



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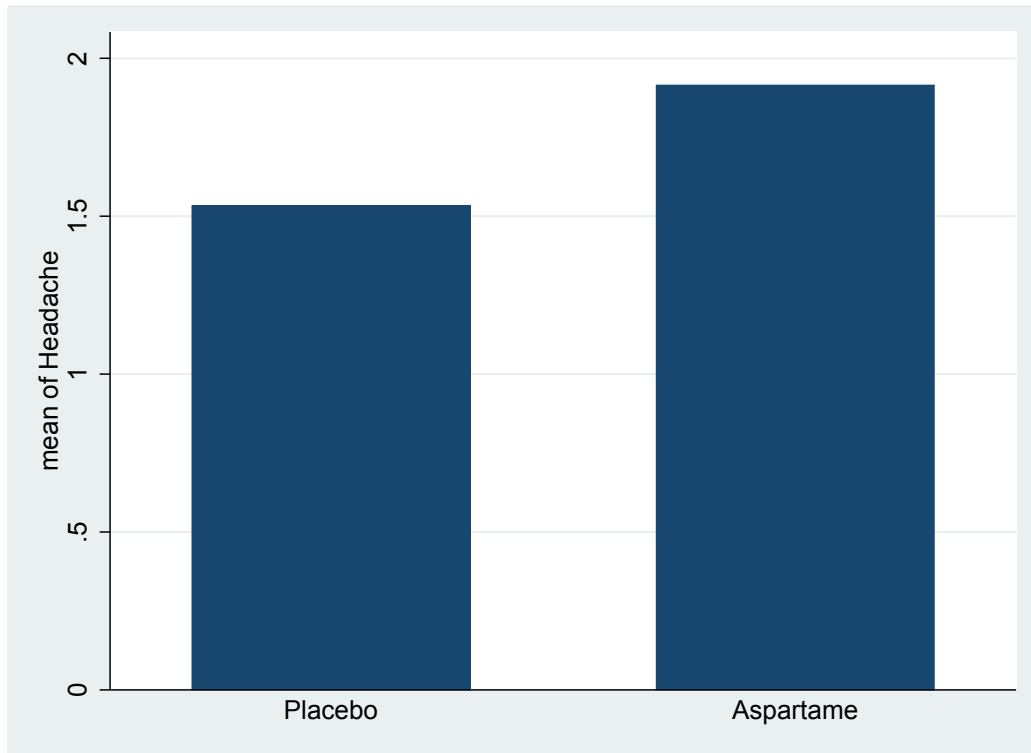
```
. graph box Headache, over(DrugAsp, relabel(1 "Placebo" 2 "Aspartame"))
```



note: 1 and 2 refer to 1st and 2nd levels of the variable DrugAsp (which are coded 0 and 1)

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```
. graph bar Headache, over(DrugAsp, relabel(1 "Placebo" 2 "Aspartame"))
```



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```
. mepoisson Headache Period1 Period2 Period3 Period4 DrugAsp || ID:
```

```
Mixed-effects Poisson regression
Group variable:          ID
Number of obs           =          122
Number of groups         =           27

Obs per group:
    min =           2
    avg =          4.5
    max =           5

Integration method: mvaghermite
Integration pts.        =           7

Wald chi2(5)            =           4.79
Prob > chi2              =          0.4426
Log likelihood = -203.17671
```

Headache	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
Period1	.0805749	.235016	0.34	0.732	-.380048 .5411979
Period2	.0344691	.2236416	0.15	0.878	-.4038603 .4727985
Period3	-.2272522	.2545615	-0.89	0.372	-.7261835 .2716791
Period4	-.159424	.2528384	-0.63	0.528	-.6549782 .3361301

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DrugAsp		.2150593	.1638877	1.31	0.189	-.1061546	.5362733
_cons		.2507236	.2050712	1.22	0.221	-.1512086	.6526557
-----+-----							
ID							
var(_cons)		.4330674	.1741208			.1969328	.9523419

LR test vs. Poisson model: chibar2(01) = 55.97				Prob >= chibar2 = 0.0000			

Comparing mixed Poisson to ordinary (fixed) Poisson regression

LR test vs. Poisson model: chibar2(01) = 55.97 Prob>=chibar2 = 0.0000

$H_0 : \sigma_v^2 = 0, \quad H_A : \sigma_v^2 > 0 \Rightarrow$ one-sided test

chibar2(01) refers to a 50:50 mixture of a χ_0^2 and a χ_1^2 distribution; chi-bar square distribution; p -value is obtained from χ_1^2 , but is halved

Interpretation of Drug Effect

- $\hat{\beta}_{\text{DrugAsp}} = .2151$
- $\exp(\hat{\beta}_{\text{DrugAsp}}) = 1.24$
- Aspartame increases the expected number of headaches (per week) by 24%, controlling for the period and random subject effects
- However, this is NOT a significant effect (p -value = .19)
- For Poisson random-intercept model, this is also the marginal effect (except for intercept β_0 , conditional $\beta =$ marginal β)

Aspartame increases the expected number of headaches by 24% controlling for the period effects

Observed means: Headaches across time by drug

drug	baseline	period 1	period 2	period 3	period 4
placebo	1.593 (n=27)	1.667 (n=12)	1.929 (n=14)	1.000 (n=13)	1.333 (n=9)
aspartame		2.267 (n=15)	1.636 (n=11)	2.000 (n=9)	1.667 (n=12)
	rate ratio	1.36	0.85	2.00	1.25

$$\text{Estimated means} = \exp(\mathbf{x}'\hat{\boldsymbol{\beta}} + \hat{\sigma}_v^2/2)$$

drug	baseline	period 1	period 2	period 3	period 4
placebo	1.603	1.737	1.659	1.278	1.367
aspartame		2.154	2.057	1.584	1.695
	rate ratio	1.24	1.24	1.24	1.24

SAS for mixed model analysis of count outcomes

PROC GLIMMIX (version 9.1.3 and thereafter)

- Multiple levels of nesting, crossed random effects
- Pseudo-likelihood estimation (by default)
 - Linearization to avoid integration over the random effects
 - Produces biased estimates if number of level-1 or level-2 units is small and/or ICC is large
- Full likelihood estimation using numerical quadrature for integration over the random effects **METHOD=QUAD**; however for 3-level models can only use **METHOD=QUAD(QPOINTS=1)** or **METHOD=LAPLACE** (these are equivalent)

SAS code: aspart.sas

```
PROC IMPORT DATAFILE = "U:\Stata_long\aspart.xlsx"
DBMS=xlsx REPLACE OUT = one;  GETNAMES=YES;

PROC MEANS;

PROC FORMAT;
VALUE Dgrp 0='Placebo' 1='Aspartame';

PROC SORT; BY DrugAsp;
/* summary stats for Headache by DrugAsp groups */
PROC MEANS; VAR Headache; BY DrugAsp; FORMAT DrugAsp Dgrp.;

/* Histogram of Headache */
PROC SGPLOT;
VBAR Headache;

/* Box-plots of Headache by DrugAsp groups */
PROC SGPLOT;
VBOX Headache / GROUP=DrugAsp;
FORMAT DrugAsp Dgrp.;
RUN;
```

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The MEANS Procedure

Variable	Label	N	Mean	Std Dev	Minimum	Maximum
ID	ID	122	13.7377049	7.7830110	1.0000000	27.0000000
Headache	Headache	122	1.6803279	1.8862985	0	7.0000000
Period1	Period1	122	0.2213115	0.4168416	0	1.0000000
Period2	Period2	122	0.2049180	0.4053062	0	1.0000000
Period3	Period3	122	0.1803279	0.3860457	0	1.0000000
Period4	Period4	122	0.1721311	0.3790511	0	1.0000000
DrugAsp	DrugAsp	122	0.3852459	0.4886602	0	1.0000000
NPeriods	NPeriods	122	4.7213115	0.7414783	2.0000000	5.0000000
NTDays	NTDays	122	6.7950820	0.9529296	1.0000000	7.0000000

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----- DrugAsp=Placebo -----

The MEANS Procedure

Analysis Variable : Headache Headache

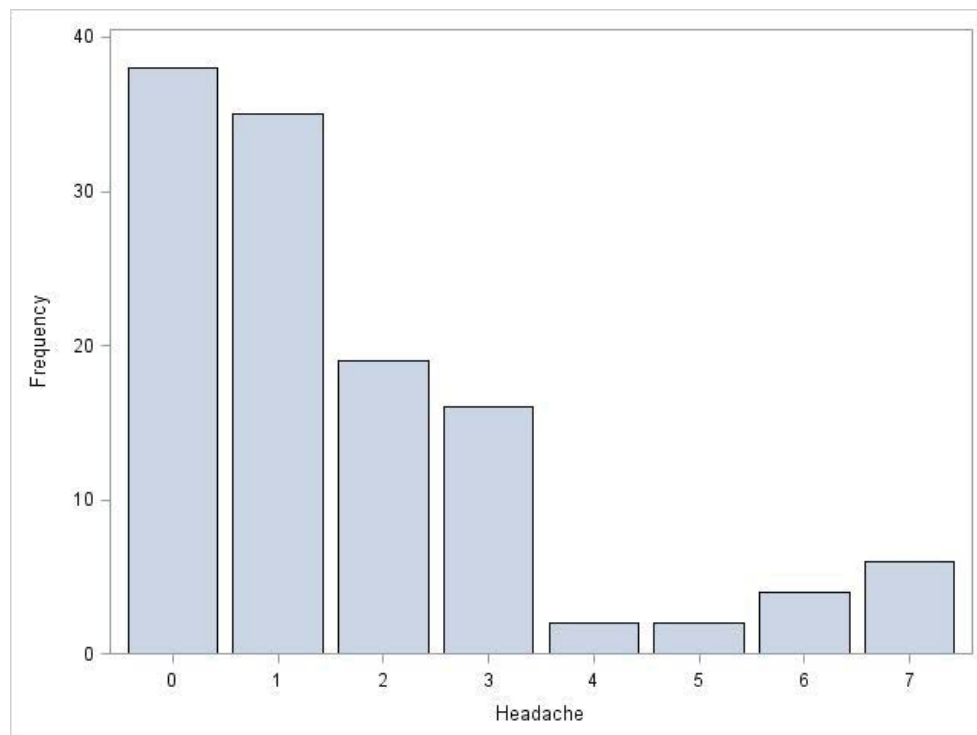
N	Mean	Std Dev	Minimum	Maximum
75	1.5333333	1.8034002	0	7.0000000

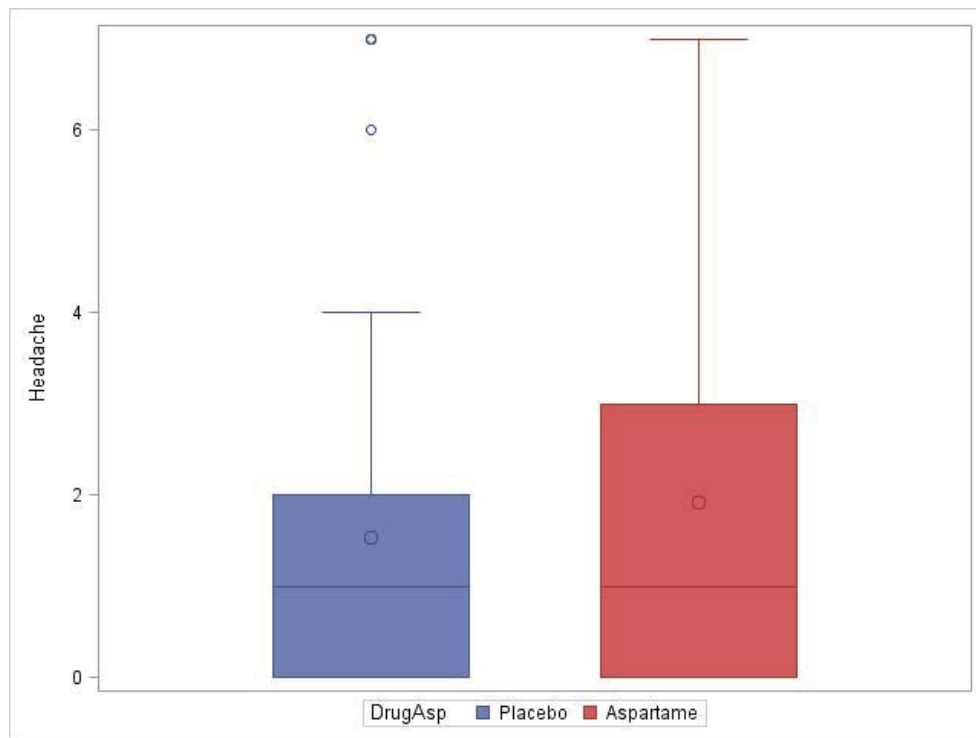
----- DrugAsp=Aspartame -----

Analysis Variable : Headache Headache

N	Mean	Std Dev	Minimum	Maximum
47	1.9148936	2.0089992	0	7.0000000

$$\Rightarrow \text{Rate ratio} = 1.915/1.533 = 1.249$$





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SAS code: aspart.sas (continued)

```
/* random intercept Poisson model */
PROC GLIMMIX METHOD=QUAD;
CLASS id;
MODEL Headache = Period1 Period2 Period3 Period4 DrugAsp
      / LINK=LOG DIST=POISSON;
RANDOM INT / SUBJECT=id;
COVTEST 'test of random intercept var' GLM;
RUN;
```

- **METHOD=QUAD** requests full-likelihood estimation (using numerical quadrature)
- **LINK=LOG** and **DIST=POISSON** for Poisson regression
- **COVTEST 'test of random intercept' GLM;**
statement yields a likelihood ratio test of

$$H_0 : \sigma_{v_0}^2 = 0, \quad H_A : \sigma_{v_0}^2 > 0 \Rightarrow \text{one-sided test}$$

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The GLIMMIX Procedure

Model Information

Data Set	WORK.ONE
Response Variable	Headache
Response Distribution	Poisson
Link Function	Log
Variance Function	Default
Variance Matrix Blocked By	ID
Estimation Technique	Maximum Likelihood
Likelihood Approximation	Gauss-Hermite Quadrature
Degrees of Freedom Method	Containment

Class Level Information

Class	Levels	Values
ID	27	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

Number of Observations Read	122
Number of Observations Used	122

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Dimensions

G-side Cov. Parameters	1
Columns in X	6
Columns in Z per Subject	1
Subjects (Blocks in V)	27
Max Obs per Subject	5

Optimization Information

Optimization Technique	Dual Quasi-Newton
Parameters in Optimization	7
Lower Boundaries	1
Upper Boundaries	0
Fixed Effects	Not Profiled
Starting From	GLM estimates
Quadrature Points	5

Iteration History

Convergence criterion (GCONV=1E-8) satisfied.

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Fit Statistics

-2 Log Likelihood	406.36
AIC (smaller is better)	420.36
AICC (smaller is better)	421.34
BIC (smaller is better)	429.43
CAIC (smaller is better)	436.43
HQIC (smaller is better)	423.05

Fit Statistics for Conditional Distribution

-2 log L(Headache r. effects)	351.67
Pearson Chi-Square	114.59
Pearson Chi-Square / DF	0.94

Covariance Parameter Estimates

Cov Parm	Subject	Estimate	Standard Error
Intercept	ID	0.4324	0.1734

Solutions for Fixed Effects

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Effect	Estimate	Standard Error	DF	t Value	Pr > t
Intercept	0.2507	0.2051	26	1.22	0.2324
Period1	0.08057	0.2350	90	0.34	0.7325
Period2	0.03446	0.2236	90	0.15	0.8779
Period3	-0.2273	0.2546	90	-0.89	0.3744
Period4	-0.1594	0.2528	90	-0.63	0.5299
DrugAsp	0.2151	0.1639	90	1.31	0.1927

Tests of Covariance Parameters Based on the Likelihood

Label	DF	-2 Log Like	ChiSq	Pr > ChiSq	Note
test of random intercept var	1	462.32	55.97	<.0001	MI

MI: P-value based on a mixture of chi-squares.

Spearman correlations of repeated counts :

aspart_corr.do

```
cd "u:\Stata_long\"
log using aspart_corr.log, replace
import excel using aspart.xlsx, sheet("aspartamin") firstrow clear
gen time=0
replace time=1 if Period1==1
replace time=2 if Period2==1
replace time=3 if Period3==1
replace time=4 if Period4==1
* summary statistics on all variables
summ
* reshape data to wide format
reshape wide Headache DrugAsp Period1 Period2 Period3 Period4 ///
          NTDays, i(ID) j(time)
* pairwise deleted Spearman correlation matrix
spearman Headache0 Headache1 Headache2 Headache3 Headache4, pw
```

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```
. spearman Headache0 Headache1 Headache2 Headache3 Headache4, pw
(obs=varies)
```

	Headac~0	Headac~1	Headac~2	Headac~3	Headac~4
Headache0	1.0000				
Headache1	0.1404	1.0000			
Headache2	0.3106	0.2066	1.0000		
Headache3	0.3029	0.4864	0.3013	1.0000	
Headache4	0.4420	0.5206	0.4304	0.3303	1.0000

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Spearman correlations of repeated counts :

aspart_corr.sas

```
PROC IMPORT DATAFILE = "U:\Stata_long\aspart.xlsx"
DBMS=xlsx REPLACE OUT = one;
GETNAMES=YES;
PROC MEANS;
RUN;

/* Get Spearman correlations of repeated counts */
PROC SORT; BY id;
DATA baseline; SET one;
if Period1=0 and Period2=0 and Period3=0 and Period4=0; Headache0=Headache;
DATA t1; SET one; if Period1=1; Headache1 = Headache;
DATA t2; SET one; if Period2=1; Headache2 = Headache;
DATA t3; SET one; if Period3=1; Headache3 = Headache;
DATA t4; SET one; if Period4=1; Headache4 = Headache;
DATA wide (KEEP=id Headache0-Headache4);
MERGE baseline t1 t2 t3 t4; BY id;

PROC CORR SPEARMAN; VAR Headache0-Headache4;
RUN;
```

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The CORR Procedure

5 Variables: Headache0 Headache1 Headache2 Headache3 Headache4

Simple Statistics

Variable	N	Mean	Std Dev	Median	Minimum	Maximum
Headache0	27	1.59259	1.92672	1.00000	0	7.00000
Headache1	27	2.00000	1.98068	1.00000	0	7.00000
Headache2	25	1.80000	1.75594	1.00000	0	7.00000
Headache3	22	1.40909	1.96781	1.00000	0	7.00000
Headache4	21	1.52381	1.88730	1.00000	0	7.00000

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Spearman Correlation Coefficients
 Prob > |r| under H0: Rho=0
 Number of Observations

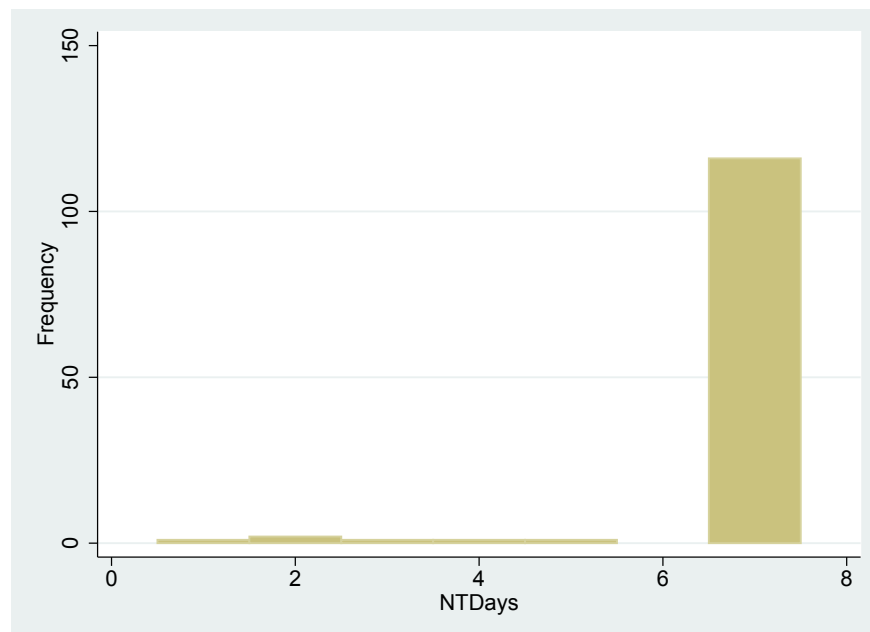
	Headache0	Headache1	Headache2	Headache3	Headache4
Headache0	1.00000	0.14039	0.31064	0.30294	0.44197
		0.4849	0.1307	0.1706	0.0449
	27	27	25	22	21
Headache1	0.14039	1.00000	0.20663	0.48640	0.52065
	0.4849		0.3217	0.0217	0.0155
	27	27	25	22	21
Headache2	0.31064	0.20663	1.00000	0.30127	0.43038
	0.1307	0.3217		0.1730	0.0515
	25	25	25	22	21
Headache3	0.30294	0.48640	0.30127	1.00000	0.33033
	0.1706	0.0217	0.1730		0.1436
	22	22	22	22	21
Headache4	0.44197	0.52065	0.43038	0.33033	1.00000
	0.0449	0.0155	0.0515	0.1436	
	21	21	21	21	21

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Prior analysis assumed that all subjects were assessed for 7 days for each period. Is this true? Does it matter?

From Aspart.do:

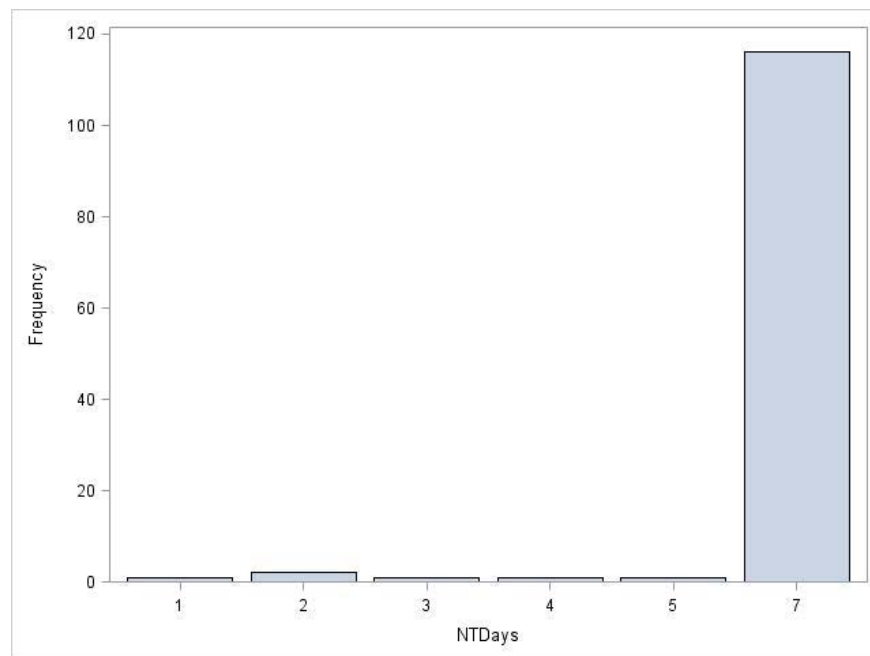
twoway histogram NTDays, discrete frequency



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From Aspart.sas:

```
PROC SGPLOT;  
VBAR NTDays; RUN;
```



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Observations with less than 7 days in a period

Period	ID	Headaches	NTDays	DrugAsp
1	17	1	1	1
2	20	1	2	1
3	9	1	5	0
4	11	1	3	1
4	27	2	4	1
4	8	1	2	0

Question

Including this information about NTDays in the model, will the drug effect be greater or smaller? or the same?

Hint: what did the prior analysis assume about NTDays?

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Mixed-effects Poisson Regression Models

The mixed-effects Poisson regression model without an offset or exposure variable:

$$\log(\mu_{ij}) = \mathbf{x}'_{ij}\boldsymbol{\beta} + \mathbf{z}'_{ij}\boldsymbol{v}_i$$

The mixed-effects Poisson regression model WITH an offset or exposure variable t_{ij} :

$$\log(\mu_{ij}) = \log(t_{ij}) + \mathbf{x}'_{ij}\boldsymbol{\beta} + \mathbf{z}'_{ij}\boldsymbol{v}_i$$

- the (log of the) offset is like a regressor with a slope=1
- for **Stata**, identify t_{ij} as the exposure variable (it will take the log of this variable internally)

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From `aspart.do`:

```
mepoisson Headache Period1 Period2 Period3 Period4 DrugAsp, ///
exposure(NTDays) || ID:
scalar m1=e(ll)
```

```
Mixed-effects Poisson regression      Number of obs      =      122
Group variable:           ID           Number of groups   =       27

                                   Obs per group:
                                   min =          2
                                   avg =         4.5
                                   max =          5

Integration method: mvaghermite        Integration pts.   =          7

                                   Wald chi2(5)          =          6.43
Log likelihood = -202.43392             Prob > chi2         =         0.2668
```

```
-----+-----
      Headache |      Coef.   Std. Err.      z    P>|z|     [95% Conf. Interval]
-----+-----
      Period1 |   .1001073   .2357201     0.42   0.671    - .3618957   .5621102
      Period2 |   .087958    .2249783     0.39   0.696    - .3529913   .5289074
```

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Period3		-.2115442	.2566767	-0.82	0.410	-.7146212	.2915328
Period4		-.0786684	.2544603	-0.31	0.757	-.5774014	.4200645
DrugAsp		.2796849	.1640922	1.70	0.088	-.0419299	.6012997
_cons		-1.712675	.2104049	-8.14	0.000	-2.125061	-1.300289
ln(NTDays)		1	(exposure)				
-----+-----							
ID							
var(_cons)		.4779805	.1930506			.2165808	1.054873
-----+-----							
LR test vs. Poisson model: chibar2(01) = 57.92				Prob >= chibar2 = 0.0000			

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Interpretation of Drug Effect

- $\hat{\beta}_{\text{DrugAsp}} = .2797$
- $\exp(\hat{\beta}_{\text{DrugAsp}}) = 1.32$
- Aspartame increases the expected rate of headaches (# of headaches per day) by 32%, controlling for the period and random subject effects
- 2-tailed non-significant (p -value = .088), but significant by a 1-tailed test (p -value = .088/2 = .044)
- For Poisson random-intercept model, this is also the marginal effect (except for intercept β_0 , conditional β = marginal β)

Aspartame increases the expected rate of headaches by 32% controlling for the period effects

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Observed means: Headaches/Day across time by drug

drug	baseline	period 1	period 2	period 3	period 4
placebo	.228 (n=27)	.238 (n=12)	.276 (n=14)	.147 (n=13)	.230 (n=9)
aspartame		.381 (n=15)	.266 (n=11)	.286 (n=9)	.272 (n=12)
rate ratio		1.60	0.96	1.95	1.18

$$\text{Estimated means} = \exp(\mathbf{x}'\hat{\boldsymbol{\beta}} + \hat{\sigma}_v^2/2)$$

drug	baseline	period 1	period 2	period 3	period 4
placebo	.228	.252	.249	.185	.211
aspartame		.334	.330	.244	.279
rate ratio		1.32	1.32	1.32	1.32

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SAS code: aspart.sas (continued)

```
/* random intercept Poisson model with offset */
PROC GLIMMIX METHOD=QUAD;
CLASS id;
logNTDays = log(NTDays);
MODEL Headache = Period1 Period2 Period3 Period4 DrugAsp
  / LINK=LOG S DIST=POISSON OFFSET=logNTDays;
RANDOM INT / SUBJECT=id;
COVTEST 'test of random intercept var' GLM;
RUN;
```

- User must take log of offset variable:
logNTDays = log(NTDays);
- Include MODEL option: OFFSET=logNTDays

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The GLIMMIX Procedure

Model Information

Data Set	WORK.ONE
Response Variable	Headache
Response Distribution	Poisson
Link Function	Log
Variance Function	Default
Offset Variable	logNTDays = log(NTDays);
Variance Matrix Blocked By	ID
Estimation Technique	Maximum Likelihood
Likelihood Approximation	Gauss-Hermite Quadrature
Degrees of Freedom Method	Containment

Fit Statistics

-2 Log Likelihood	404.87
AIC (smaller is better)	418.87
AICC (smaller is better)	419.86
BIC (smaller is better)	427.94
CAIC (smaller is better)	434.94
HQIC (smaller is better)	421.57

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Covariance Parameter Estimates

Cov Parm	Subject	Estimate	Standard Error
Intercept	ID	0.4770	0.1921

Solutions for Fixed Effects

Effect	Estimate	Standard Error	DF	t Value	Pr > t
Intercept	-1.7126	0.2104	26	-8.14	<.0001
Period1	0.1001	0.2357	90	0.42	0.6721
Period2	0.08792	0.2250	90	0.39	0.6969
Period3	-0.2116	0.2567	90	-0.82	0.4119
Period4	-0.07871	0.2545	90	-0.31	0.7578
DrugAsp	0.2797	0.1641	90	1.70	0.0918

Tests of Covariance Parameters Based on the Likelihood

Label	DF	-2 Log Like	ChiSq	Pr > ChiSq	Note
test of random intercept var	1	462.78	57.91	<.0001	MI

MI: P-value based on a mixture of chi-squares.

Empirical Bayes estimates of random effects

$$\log(\mu_{ij}) = \log(t_{ij}) + \mathbf{x}_{ij}'\boldsymbol{\beta} + v_i \quad \text{where } v_i \sim N(0, \sigma_v^2)$$

- Random effects v_i are also estimated
- can be of interest to indicate how particular subjects are doing
- can be used to rank, compare, or indicate unusual subjects
- From `Aspart.do` :

```
* get random effect estimates (and std errs)
predict u0, reffects reses(u0se)
* assign a value of 1 for one obs of each subject
* and rank the random effects
egen pick1obs = tag(ID)
egen u0rank = rank(u0) if pick1obs==1
list ID u0 u0se u0rank if pick1obs==1
* standard error bar chart of random effects
serrbar u0 u0se u0rank if pick1obs==1, scale(1.96) yline(0)
```

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```
. list ID u0 u0se u0rank if pick1obs==1
```

	ID	u0	u0se	u0rank
1.	2	.2936098	.3489556	18
4.	5	-.2885217	.3663934	9
9.	13	1.469566	.1751805	27
14.	16	.0260536	.3832188	17
17.	19	-.585127	.404517	4
22.	23	.7595504	.2416335	26
27.	25	.6363964	.2548456	23
32.	1	-.0479451	.3349817	14
37.	3	.3374991	.2880078	19
42.	6	-.1651728	.3497926	12
47.	9	-.0790354	.3773501	13
51.	17	.7314795	.430931	25
53.	18	-.5895116	.4042242	3
58.	21	.6306354	.2547467	22
63.	22	-.4340891	.3842416	6
68.	7	.3381366	.2880236	20

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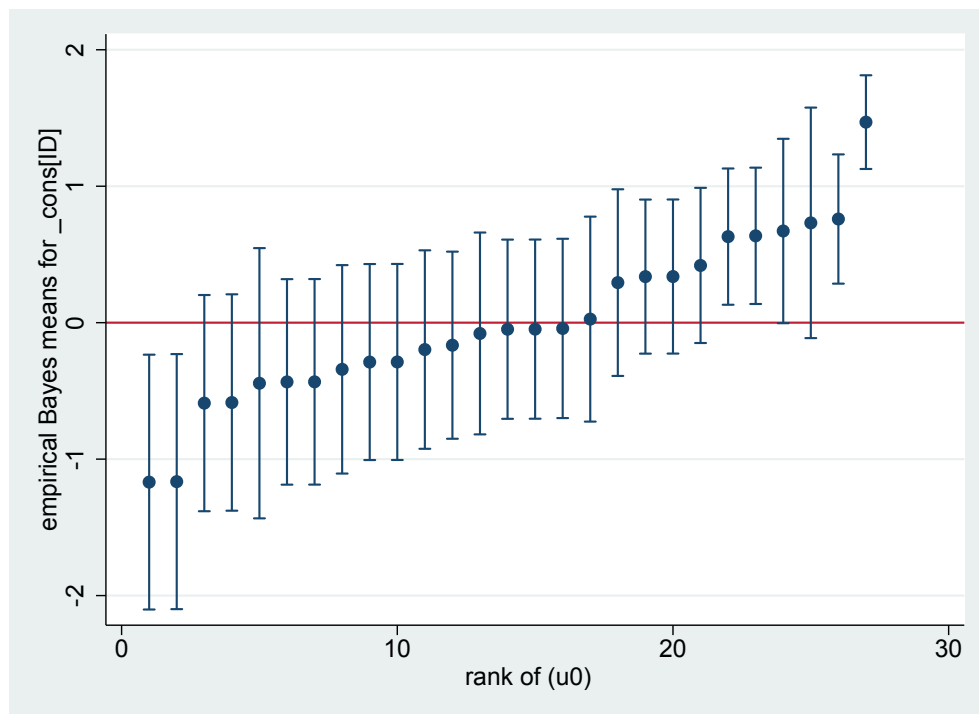
73.		10	-1.168355	.4763907	1	
78.		11	-.1970716	.371061	11	
83.		14	-.0473549	.3350045	15	
88.		24	-.4335562	.3842723	7	

93.		27	.4194811	.2900531	21	
98.		4	-1.164856	.4767669	2	
103.		8	-.3419288	.3896556	8	
108.		12	-.2879634	.3664214	10	
113.		15	-.0422512	.3352014	16	

118.		20	.6717975	.3446973	24	
121.		26	-.4439119	.5051993	5	
+-----+						

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```
. * standard error bar chart of random effects
. serrbar u0 u0se u0rank if pick1obs==1, scale(1.96) yline(0)
```



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From `aspart.sas`: include **SOLUTION** or **S** on **RANDOM** statement and use **ODS** to direct these estimates to a new dataset named **ebest**

```
PROC GLIMMIX METHOD=QUAD;
CLASS id;
logNTDays = log(NTDays);
MODEL Headache = Period1 Period2 Period3 Period4 DrugAsp
      / LINK=LOG S DIST=POISSON OFFSET=logNTDays;
RANDOM INT / SUBJECT=id S;
ODS OUTPUT SOLUTIONR=ebest;
COVTEST 'test of random intercept var' GLM;
RUN;

/* print out the estimated random effects */
PROC PRINT DATA=ebest;
RUN;
```

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Obs	Effect	Subject	Estimate	StdErr Pred	DF	tValue	Probt
1	Intercept	ID 1	-0.00585	0.3536	90	-0.02	0.9868
2	Intercept	ID 2	0.3379	0.3712	90	0.91	0.3651
3	Intercept	ID 3	0.3711	0.3179	90	1.17	0.2462
4	Intercept	ID 4	-1.1056	0.5204	90	-2.12	0.0364
5	Intercept	ID 5	-0.2411	0.3800	90	-0.63	0.5274
6	Intercept	ID 6	-0.1205	0.3657	90	-0.33	0.7426
7	Intercept	ID 7	0.3717	0.3180	90	1.17	0.2456
8	Intercept	ID 8	-0.2910	0.4015	90	-0.72	0.4704
9	Intercept	ID 9	-0.03007	0.3909	90	-0.08	0.9389
10	Intercept	ID 10	-1.1091	0.5202	90	-2.13	0.0357
11	Intercept	ID 11	-0.1490	0.3843	90	-0.39	0.6992
12	Intercept	ID 12	-0.2405	0.3801	90	-0.63	0.5285
13	Intercept	ID 13	1.4836	0.2383	90	6.23	<.0001
14	Intercept	ID 14	-0.00525	0.3537	90	-0.01	0.9882
15	Intercept	ID 15	-0.00012	0.3539	90	-0.00	0.9997
16	Intercept	ID 16	0.07582	0.3977	90	0.19	0.8492
17	Intercept	ID 17	0.7864	0.4676	90	1.68	0.0961
18	Intercept	ID 18	-0.5365	0.4172	90	-1.29	0.2018
19	Intercept	ID 19	-0.5321	0.4174	90	-1.27	0.2056
20	Intercept	ID 20	0.7152	0.3772	90	1.90	0.0611
21	Intercept	ID 21	0.6581	0.2939	90	2.24	0.0276
22	Intercept	ID 22	-0.3839	0.3967	90	-0.97	0.3358
23	Intercept	ID 23	0.7846	0.2845	90	2.76	0.0071
24	Intercept	ID 24	-0.3834	0.3968	90	-0.97	0.3366
25	Intercept	ID 25	0.6639	0.2939	90	2.26	0.0263
26	Intercept	ID 26	-0.3855	0.5172	90	-0.75	0.4580
27	Intercept	ID 27	0.4534	0.3205	90	1.41	0.1607

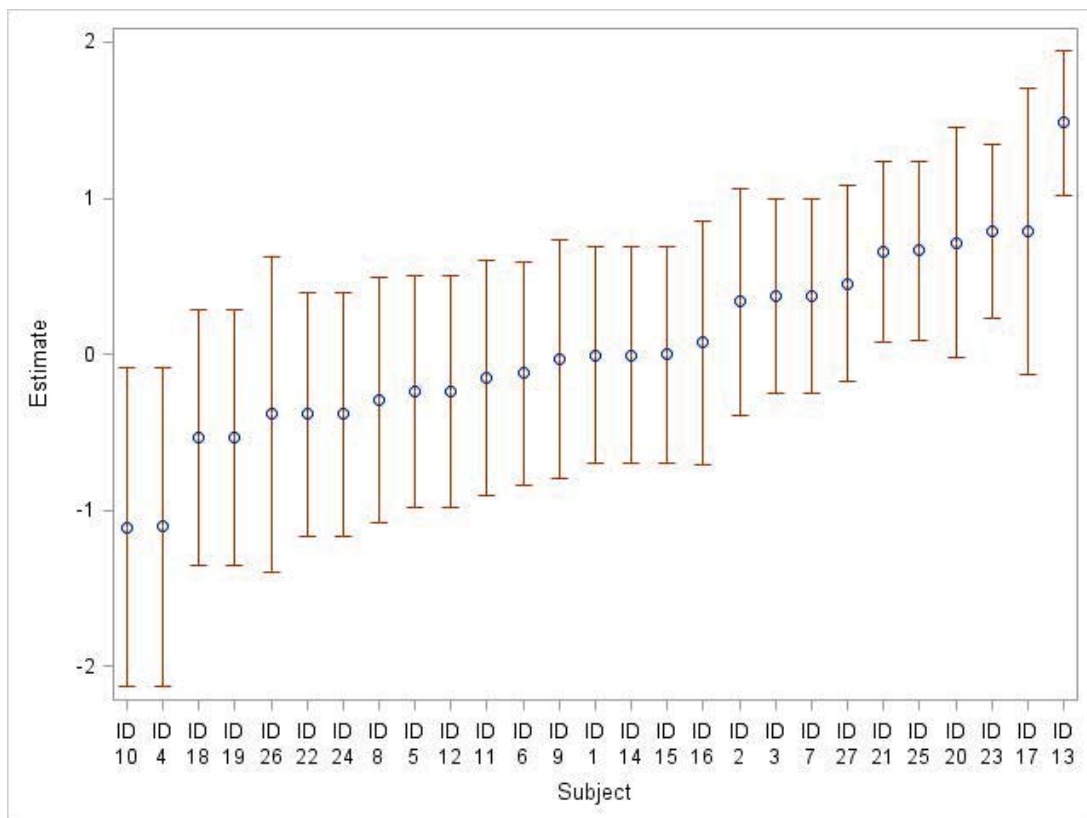
56

SAS code: aspart.sas (continued)

```
DATA ebestci; SET ebest;
lo = Estimate-PROBIT(.975)*StdErrPred;
hi = Estimate+PROBIT(.975)*StdErrPred;
RUN;

/* standard error plot with estimated random effects */
PROC SORT; BY Estimate;
PROC SGLOT DATA=ebestci;
SCATTER X=Subject Y=Estimate / YERRORLOWER=lo YERRORUPPER=hi;
RUN;
```

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Large empirical Bayes estimates of random effects

- ID 13 has large positive value (1.470) and has observed data: 7, 7, 7, 6, 7 headaches

note, also that this subject has the smallest standard error (0.175) due to number and consistency of responses

$$95\% \text{ C.I.} = 1.470 \pm 1.96 \times 0.175 = (1.127, 1.813)$$

- IDs 4 and 10 have large negative values (< -1) and have observed headaches of 0 for all periods

Random drug effect?

- In many studies, drug or treatment is a subject-level variable and doesn't vary across time
- In a crossover study, however, drug is a time-varying variable and DOES vary across time
- A time-varying variable can be considered as random at the subject level
 - Does the drug effect vary across subjects?
 - Is there subject heterogeneity in the number of headaches for aspartame relative to placebo?

Model in multilevel representation

$i = 1, \dots, 27$ subjects $j = 1, \dots, n_i$ periods (max = 5)

Level-1 model (within-subjects)

$$\log(\mu_{ij}) = \log(t_{ij}) + b_{0i} + b_{1i}P1_j + b_{2i}P2_j + b_{3i}P3_j + b_{4i}P4_j + b_{5i}Drug_{ij}$$

Level-2 model (between-subjects)

$$b_{0i} = \beta_0 + v_{0i}$$

$$b_{1i} = \beta_1$$

$$b_{2i} = \beta_2$$

$$b_{3i} = \beta_3$$

$$b_{4i} = \beta_4$$

$$b_{5i} = \beta_5 + v_{5i}$$

Does the effect of aspartame on headaches vary across subjects?
(is v_{5i} necessary?)

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From `Aspart.do` :

```
* random int & Drug Poisson regression model with varying exposure
mepoisson Headache Period1 Period2 Period3 Period4 DrugAsp, exposure(NTDays) ///
|| ID:DrugAsp, covariance(unstructured)
scalar m2=e(11)
```

```
Mixed-effects Poisson regression      Number of obs      =      122
Group variable:      ID                Number of groups   =      27
```

```
Obs per group:
      min =      2
      avg =      4.5
      max =      5
```

```
Integration method: mvaghermite      Integration pts.   =      7
```

```
Log likelihood = -200.08337      Wald chi2(5)      =      3.71
      Prob > chi2      =      0.5920
```

Headache	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]	
Period1	.1235719	.2479087	0.50	0.618	-.3623203	.609464
Period2	.0567582	.2318316	0.24	0.807	-.3976233	.5111397
Period3	-.2126376	.2655232	-0.80	0.423	-.7330535	.3077782
Period4	-.0662054	.2621424	-0.25	0.801	-.5799951	.4475843

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DrugAsp	.2468226	.2363062	1.04	0.296	-.216329	.7099742
_cons	-1.742829	.2224108	-7.84	0.000	-2.178746	-1.306912
ln(NTDays)	1	(exposure)				

ID						
var(DrugAsp)	.3938288	.2880716			.093904	1.651698
var(_cons)	.5281057	.2379992			.21833	1.277404

ID						
cov(_cons, DrugAsp)	-.144784	.2073222	-0.70	0.485	-.551128	.26156

LR test vs. Poisson model: chi2(3) = 62.62 Prob > chi2 = 0.0000

Note: LR test is conservative and provided only for reference.

Likelihood ratio test: comparing random DrugAsp to random intercept model

$$H_0 : \sigma_{v_5}^2 = \sigma_{v_{05}}^2 = 0$$

need `chibar2(12)`, 50:50 mixture of a χ_1^2 and a χ_2^2 distribution; p -value is obtained from the average of χ_1^2 and χ_2^2 (i.e., q and $q - 1$, where q is the number of (co)variance parameters in the null)

From `Aspart.do` :

```
. * get LR test for comparing models
. display "chibar2(12) = " 2*(m2-m1)
chibar2(12) = 4.7010964

. display "Prob > chibar2(12) = " .5*chi2tail(1, 2*(m2-m1)) + ///
> .5*chi2tail(2, 2*(m2-m1))
Prob > chibar2(12) = .06273014
```


Observed means: Headaches/Day across time by drug

drug	baseline	period 1	period 2	period 3	period 4
placebo	.228 (n=27)	.238 (n=12)	.276 (n=14)	.147 (n=13)	.230 (n=9)
aspartame		.381 (n=15)	.266 (n=11)	.286 (n=9)	.272 (n=12)
rate ratio		1.60	0.96	1.95	1.18

Estimated means = $\exp(\mathbf{x}'\hat{\boldsymbol{\beta}} + \hat{\sigma}_{v_0}^2/2)$ for placebo
= $\exp(\mathbf{x}'\hat{\boldsymbol{\beta}} + 1/2(\hat{\sigma}_{v_0}^2 + \hat{\sigma}_{v_5}^2 + 2\hat{\sigma}_{v_{05}}))$ for aspartame

drug	baseline	period 1	period 2	period 3	period 4
placebo	.228	.258	.242	.184	.217
aspartame		.348	.326	.249	.288
rate ratio		1.35	1.35	1.35	1.35

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From aspart.sas

```
/* random intercept & DrugAsp Poisson model with offset */
PROC GLIMMIX METHOD=QUAD DATA=one;
CLASS id;
logNTDays = log(NTDays);
MODEL Headache = Period1 Period2 Period3 Period4 DrugAsp
      / LINK=LOG S DIST=POISSON OFFSET=logNTDays;
RANDOM INT DrugAsp / SUBJECT=id TYPE=UN G GCORR S;
ODS OUTPUT SOLUTIONR=ebest2;
COVTEST 'test of random effects' GLM;
COVTEST 'test of random slope' . 0 0;
RUN;
```

- **COVTEST 'test of random effects' GLM;**
will compare this model to model without random effects using LR test:
 $H_0 : \sigma_{v_0}^2 = \sigma_{v_5}^2 = \sigma_{v_{05}} = 0$
- **COVTEST 'test of random slope' . 0 0;**
will compare this model to model without random slope and intercept
slope covariance using LR test: $H_0 : \sigma_{v_5}^2 = \sigma_{v_{05}} = 0$
- **SOLUTION** or **S** on **RANDOM** statement will list the estimates of the random effects; **ODS** statement saves these to dataset **ebest2**

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The GLIMMIX Procedure

Model Information

Data Set	WORK.ONE
Response Variable	Headache
Response Distribution	Poisson
Link Function	Log
Variance Function	Default
Offset Variable	logNTDays = log(NTDays);
Variance Matrix Blocked By	ID
Estimation Technique	Maximum Likelihood
Likelihood Approximation	Gauss-Hermite Quadrature
Degrees of Freedom Method	Containment

Class Level Information

Class	Levels	Values
ID	27	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

Number of Observations Read	122
Number of Observations Used	122

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Dimensions

G-side Cov. Parameters	3
Columns in X	6
Columns in Z per Subject	2
Subjects (Blocks in V)	27
Max Obs per Subject	5

Fit Statistics

-2 Log Likelihood	400.17
AIC (smaller is better)	418.17
AICC (smaller is better)	419.78
BIC (smaller is better)	429.83
CAIC (smaller is better)	438.83
HQIC (smaller is better)	421.64

Estimated G Matrix

Effect	Row	Col1	Col2
Intercept	1	0.5271	-0.1446
DrugAsp	2	-0.1446	0.3934

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Estimated G Correlation Matrix

Effect	Row	Col1	Col2
Intercept	1	1.0000	-0.3175
DrugAsp	2	-0.3175	1.0000

Covariance Parameter Estimates

Cov Parm	Subject	Estimate	Standard Error
UN(1,1)	ID	0.5271	0.2373
UN(2,1)	ID	-0.1446	0.2075
UN(2,2)	ID	0.3934	0.2876

Solutions for Fixed Effects

Effect	Estimate	Standard Error	DF	t Value	Pr > t
Intercept	-1.7428	0.2224	26	-7.83	<.0001
Period1	0.1235	0.2479	65	0.50	0.6200
Period2	0.05675	0.2318	65	0.24	0.8074
Period3	-0.2127	0.2655	65	-0.80	0.4261
Period4	-0.06624	0.2621	65	-0.25	0.8013
DrugAsp	0.2468	0.2364	25	1.04	0.3063

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Tests of Covariance Parameters Based on the Likelihood

Label	DF	-2 Log Like	ChiSq	Pr > ChiSq	Note
test of random effects	3	462.78	62.61	<.0001	--
test of random slope	2	404.87	4.70	0.0627	MI

MI: P-value based on a mixture of chi-squares.

--: Standard test with unadjusted p-values.

- **test of random effects** compares random trend model to ordinary Poisson regression using ordinary LR test (too conservative)
- **test of random slope** compares random trend to random intercept model

$$H_0 : \sigma_{v_5}^2 = \sigma_{v_{05}} = 0$$

uses **chibar2(12)**, 50:50 mixture of a χ_1^2 and a χ_2^2 distribution; p -value is obtained from the average of χ_1^2 and χ_2^2 (i.e., q and $q - 1$, where q is the number of (co)variance parameters in the null)

Graph of Random Drug Effects

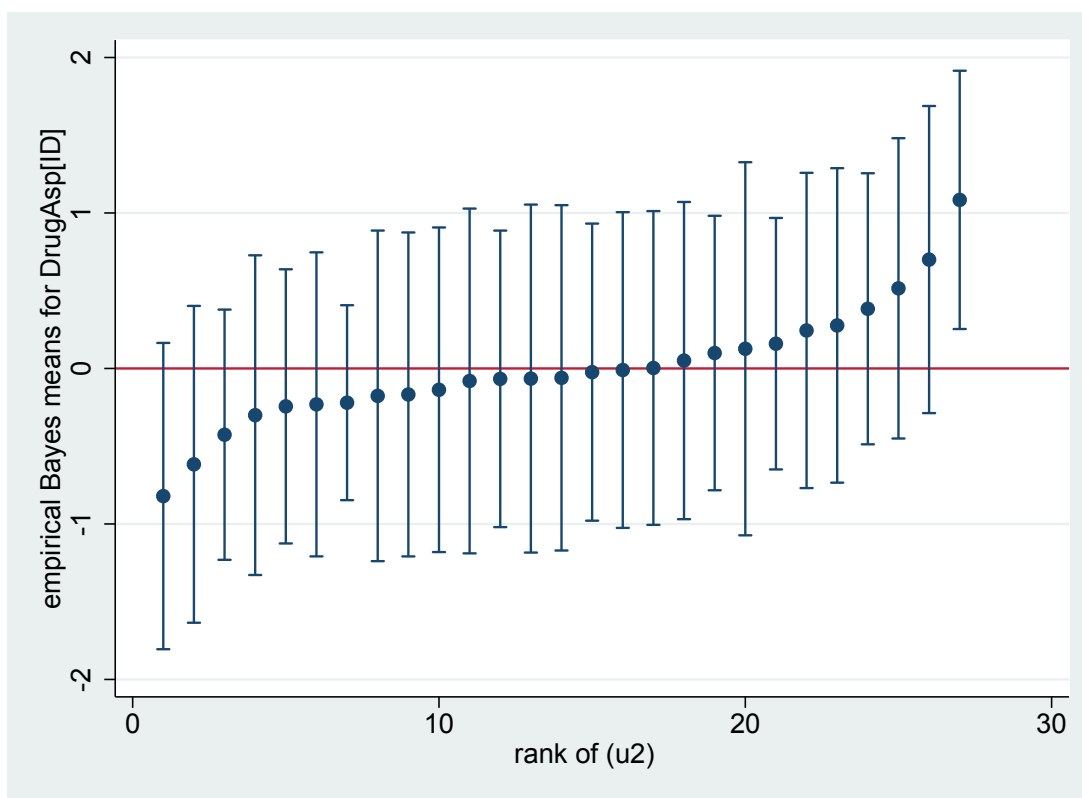
From Aspart.do :

```
* get random effect estimates (and std errs)
predict u2 u1, reffects reses(u2se u1se)

* rank the random DrugAsp effects
egen u2rank = rank(u2) if pick1obs==1
list ID u2 u2se u2rank if pick1obs==1

* standard error bar chart of random effects
serrbar u2 u2se u2rank if pick1obs==1, scale(1.96) yline(0)
```

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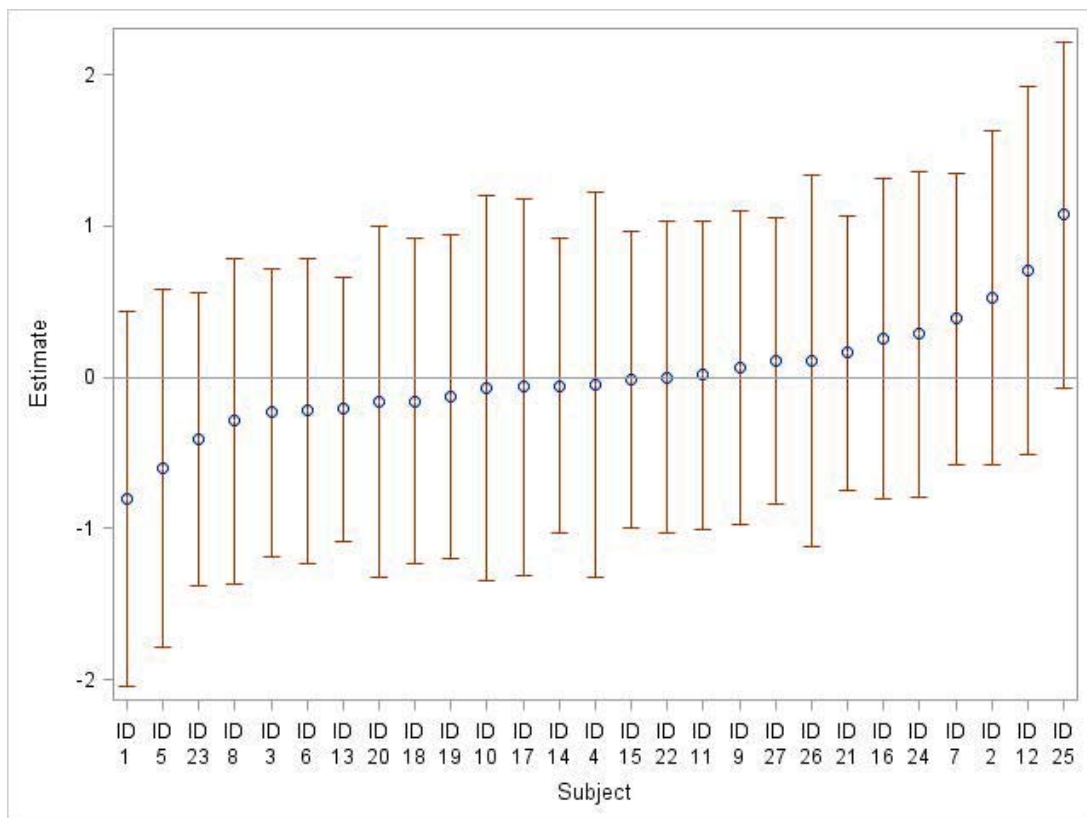
From aspart.sas

```
/* print out the estimated random effects */
PROC PRINT DATA=ebest2;
RUN;

/* select only the random slopes */
DATA ebest2ci; SET ebest2; if Effect = "DrugAsp";
lo = Estimate-PROBIT(.975)*StdErrPred;
hi = Estimate+PROBIT(.975)*StdErrPred;
RUN;

/* standard error plot with estimated random effects */
PROC SORT; BY Estimate;
PROC SGPLOT DATA=ebest2ci;
SCATTER X=Subject Y=Estimate / YERRORLOWER=lo YERRORUPPER=hi;
REFLINE 0 /AXIS=Y;
RUN;
```

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Very interesting subject

Empirical Bayes estimate of drug effect is very large (≈ 1) for ID=25

observed number of headaches:

1 (placebo), 6 (drug), 1 (placebo), 7 (drug), 0 (placebo)

Drug Effect Estimates

	model	estimate	std error	p-value
with ID=25	rand int	0.2797	0.1641	0.088
	rand drug	0.2468	0.2363	0.296
without ID=25	rand int	0.1384	0.1698	0.415
	rand drug	0.1462	0.1955	0.455

also, from random drug models, estimate of drug variance goes from 0.3938 (se = 0.288) to 0.0030 (se = 0.033) when this subject is removed

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Overdispersion

- Poisson model assumes that the mean equals the variance
- overdispersion occurs when the variance exceeds the mean
 - often present in real data, can change model estimates
 - inclusion of random effects, by accounting for individual differences, may decrease possibility of overdispersion
- Negative Binomial model relaxes this assumption by including an overdispersion parameter
 - Poisson model is a special case of Negative Binomial when this overdispersion parameter equals 0
- in **Stata** use **menbreg**
- in **SAS** use **MODEL** option **DIST=NEGBINOMIAL**

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From Aspart.do :

```
* random intercept Negative Binomial model with varying exposure
menbreg Headache Period1 Period2 Period3 Period4 DrugAsp, exposure(NTDays) ///
|| ID:
scalar m3=e(11)

* get LR test for comparing models
display "chibar2(01) = " 2*(m3-m1)
display "Prob > chibar2(01) = " .5*chi2tail(1, 2*(m3-m1))
```

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```
. menbreg Headache Period1 Period2 Period3 Period4 DrugAsp, exposure(NTDays) ///
> || ID:
```

Mixed-effects nbinomial regression	Number of obs	=	122
Overdispersion: mean			
Group variable: ID	Number of groups	=	27

Obs per group:		
min	=	2
avg	=	4.5
max	=	5

Integration method: mvaghermite	Integration pts.	=	7
---------------------------------	------------------	---	---

	Wald chi2(5)	=	5.76
Log likelihood = -202.1028	Prob > chi2	=	0.3298

Headache	Coef.	Std. Err.	z	P> z	[95% Conf. Interval]
-----+-----					
Period1	.1117396	.2539913	0.44	0.660	-.3860743 .6095534
Period2	.0869462	.2445365	0.36	0.722	-.3923365 .5662288

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Period3		-.2260488	.2760435	-0.82	0.413	-.767084	.3149865
Period4		-.0934193	.2769078	-0.34	0.736	-.6361486	.44931
DrugAsp		.2860314	.1773072	1.61	0.107	-.0614843	.6335472
_cons		-1.706765	.2180453	-7.83	0.000	-2.134126	-1.279404
ln(NTDays)		1	(exposure)				
-----+-----							
/lnalpha		-2.601152	1.389024			-5.323588	.1212846
-----+-----							
ID							
var(_cons)		.4614995	.193958			.2025023	1.05175
-----+-----							

LR test vs. nbinoial model: chibar2(01) = 23.17 Prob >= chibar2 = 0.0000

```

. scalar m3=e(l1)

. * get LR test for comparing models

. display "chibar2(01) = " 2*(m3-m1)
chibar2(01) = .66222995

. display "Prob > chibar2(01) = " .5*chi2tail(1, 2*(m3-m1))
Prob > chibar2(01) = .20788689

```

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From aspart.sas

```

/* random intercept Negative Binomial model with offset */
PROC GLIMMIX METHOD=QUAD;
CLASS id;
logNTDays = log(NTDays);
MODEL Headache = Period1 Period2 Period3 Period4 DrugAsp
/ LINK=LOG S DIST=NEGBINOMIAL OFFSET=logNTDays;
RANDOM INT / SUBJECT=id;
COVTEST 'test of random intercept var' GLM;
RUN;

```


The GLIMMIX Procedure

Model Information

Data Set	WORK.ONE
Response Variable	Headache
Response Distribution	Negative Binomial
Link Function	Log
Variance Function	Default
Offset Variable	logNTDays = log(NTDays);
Variance Matrix Blocked By	ID
Estimation Technique	Maximum Likelihood
Likelihood Approximation	Gauss-Hermite Quadrature
Degrees of Freedom Method	Containment

Class Level Information

Class	Levels	Values
ID	27	1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21 22 23 24 25 26 27

Number of Observations Read	122
Number of Observations Used	122

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Dimensions

G-side Cov. Parameters	1
R-side Cov. Parameters	1
Columns in X	6
Columns in Z per Subject	1
Subjects (Blocks in V)	27
Max Obs per Subject	5

Fit Statistics

-2 Log Likelihood	404.37
AIC (smaller is better)	420.37
AICC (smaller is better)	421.65
BIC (smaller is better)	430.74
CAIC (smaller is better)	438.74
HQIC (smaller is better)	423.45

Covariance Parameter Estimates

Cov Parm	Subject	Estimate	Standard Error
Intercept	ID	0.4541	0.1904
Scale		0.07969	0.1036

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Solutions for Fixed Effects

Effect	Estimate	Standard Error	DF	t Value	Pr > t
Intercept	-1.7047	0.2180	26	-7.82	<.0001
Period1	0.1123	0.2552	90	0.44	0.6610
Period2	0.08660	0.2458	90	0.35	0.7255
Period3	-0.2273	0.2773	90	-0.82	0.4145
Period4	-0.09455	0.2783	90	-0.34	0.7349
DrugAsp	0.2863	0.1782	90	1.61	0.1116

Tests of Covariance Parameters Based on the Likelihood

Label	DF	-2 Log Like	ChiSq	Pr > ChiSq	Note
test of random intercept var	1	427.37	23.00	<.0001	MI

MI: P-value based on a mixture of chi-squares.

Conclusions

- Van Den Eeden *et al.*, (1994), Aspartame ingestion and headaches: A randomized crossover trial, *Neurology*, 44, 1787-1793.
“the proportion of days subjects reported having a headache was higher during aspartame treatment compared with placebo treatment (aspartame = .33, placebo = .24, $p = .04$)”
- Levy, Hedeker, & Sanders (1995) To the editor: Aspartame and headache, *Neurology*, 45(8):1631-2; author reply 1632-3.
Increase of headaches by aspartame only for 1-tailed test in random intercept model; random drug model and model without subject 25 (very influential subject) shows no drug effect.
- Butchkoa & Stargelb (2001), Aspartame: Scientific evaluation in the postmarketing period, *Regulatory Toxicology and Pharmacology*, 34, 221-233.

“Evaluation of the anecdotal reports of adverse health effects, the first such system for a food additive, revealed that the reported effects were generally mild and also common in the general population and that there was no consistent or unique pattern of symptoms that could be causally linked to consumption of aspartame. Finally, the results of the extensive scientific research done to evaluate these allegations did not show a causal relationship between aspartame and adverse effects. Thus, the weight of scientific evidence confirms that, even in amounts many times what people typically consume, aspartame is safe for its intended uses as a sweetener and flavor enhancer.”

Summary

- Poisson model useful for modeling counts; $\exp \beta = \text{rate ratio}$
 - Poisson model assumes mean = variance
- Poisson overdispersion can be handled by
 - random effects
 - inclusion of overdispersion parameter (Negative Binomial regression)
 - random effects and overdispersion (mixed Negative Binomial regression)
- Zero-inflated models (ZIP, ZINB) are available in Stata and SAS without random effects; perhaps with random effects in the future (can use SAS PROC NLMIXED, but involves some programming)