



## Analysis of 2-level repeated measures data

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This example illustrates how multilevel modeling may be used to recognize explicitly the hierarchical structure of repeated measurement data.

Five models will be fitted and discussed:

- A variance decomposition model
- Modeling linear growth
- Modeling non-linear growth
- Introducing a covariate when modeling non-linear growth
- A model with complex variation at level-1 of the hierarchy

### 1. Description of the data

The data set used contains repeated measurements on 82 striped mice and was obtained from the Department of Zoology at the University of Pretoria, South Africa (see du Toit, 1979). A number of male and female mice were released in an outdoor enclosure with nest boxes and sufficient food and water. They were allowed to multiply freely. Occurrence of birth was recorded daily and newborn mice were weighed weekly, from the end of the second week after birth until physical maturity was reached. The data set consists of the weights of 42 male and 40 female mice. For male mice, 9 repeated weight measurements are available and for the female mice 8 repeated measurements.

The first 11 observations from this data set, contained in **mouse.lsf**, and the variable names to be used are shown below.

	iden2	iden1	weight	constant	time	timesq	gender
1	1.00	1.00	15.00	1.00	1.00	1.00	1.00
2	1.00	2.00	17.00	1.00	2.00	4.00	1.00
3	1.00	3.00	23.00	1.00	3.00	9.00	1.00
4	1.00	4.00	24.00	1.00	4.00	16.00	1.00
5	1.00	5.00	26.00	1.00	5.00	25.00	1.00
6	1.00	6.00	31.00	1.00	6.00	36.00	1.00
7	1.00	7.00	37.00	1.00	7.00	49.00	1.00
8	1.00	8.00	42.00	1.00	8.00	64.00	1.00
9	1.00	9.00	46.00	1.00	9.00	81.00	1.00
10	2.00	1.00	11.00	1.00	1.00	1.00	1.00
11	2.00	2.00	14.00	1.00	2.00	4.00	1.00

The response variable WEIGHT contains the weight measurements (in grams) for all mice at the different times of measurement. The explanatory variables which may be used are the time points at which measurements were made (TIME), the squared values of these time points (TIMESQ), and the gender of the mice (GENDER). It is also assumed that the growth of the mice during this period can be adequately described with a parabolic function.

A hierarchical level-2 structure is incorporated where the individual mice are the level-2 units. Unique numbers identifying the mice are contained in the variable IDEN2, which will be used as the level-2 ID for the analysis. The variable IDEN1 identifies the occasions on which measurements for a particular mouse were made and will be used as the level-1 ID. From the description of the data set as given above, it follows that there are 82 level-2 units, with either 8 or 9 measurements nested within each level-2 unit.

The variable CONSTANT consists of a column of 1s and may be used as alternative to the intercept term automatically included in the model.

## 2. Variance decomposition

The simplest multilevel model is equivalent to a one-way ANOVA with random effects. Although this model is not interesting in itself, it is useful as a preliminary step in a multilevel analysis as it provides important information about the outcome variability at each of the levels of the hierarchy. It may also function as a baseline with which more sophisticated models may be compared.

Let the subscript  $i$  denote the  $i$ -th level-2 unit, in this case the  $i$ -th mouse. The subscript  $j$  refers to the  $j$ -th weight measurement for the  $i$ -th mouse. Using this notation, the one-way ANOVA model can be written as:

$$y_{ij} = \beta_0 + u_{ij} + e_{ij}$$

where  $u_{ij}$  denotes the random component on level 2 of the model. It is assumed that  $u_{ij}$  has an expected value of 0 and a variance of  $\Phi_{(2)}$ . The variance  $\Phi_{(2)}$  may be interpreted as the “between-group” variability. Likewise, it is assumed that  $e_{ij}$  is  $N(0, \Phi_{(1)})$  distributed. Thus  $\Phi_{(1)}$  may be interpreted as the “within-group” variability.

This model is also known as a fully unconditional model (Bryk & Raudenbush, 1992), as no predictors are specified at either level of the hierarchy.

To specify the model, open the data file **mouse.lsf** and then select the **Multilevel** option from the main menu bar. On the **Title and Options** dialog box, enter a title for this analysis and leave all other options default. Click **Next** to move to the next dialog box.

**Title and Options** [X]

Title (Maximum 70 characters):

Maximum Number of Iterations:

Convergence Criterion:

Missing Data Value:  Nfree:

Missing Dep Value:  Deviance:

Use OLS for starting values     Calculate effect sizes

Additional Output

Asymptotic Covariances     Residuals

Empirical Bayes Estimates     No Data Summary

Between and Within Covariance Matrices

To build Syntax, proceed to the Random Variables screen and click the Finish Button

Indicate that IDEN2 will serve as level-2 ID for this analysis on the **Identification Variables** dialog box. As there are no weights in these data, click **Next** twice to move on to the **Select Response and Fixed Variables** dialog box.

**Identification Variables** [X]

Variables in data

- iden2
- iden1
- weight
- constant
- time
- timesq
- gender

   Level 5 ID Variable

   Level 4 ID Variable

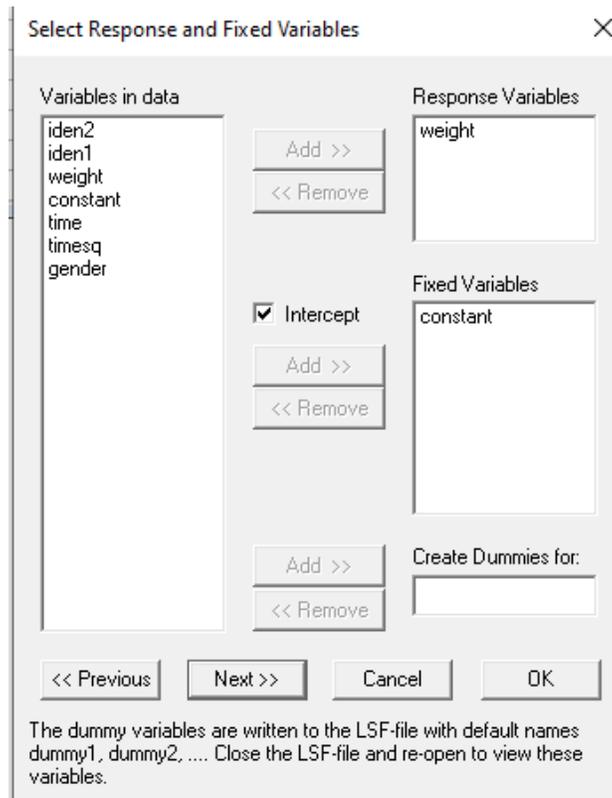
   Level 3 ID Variable

   Level 2 ID Variable

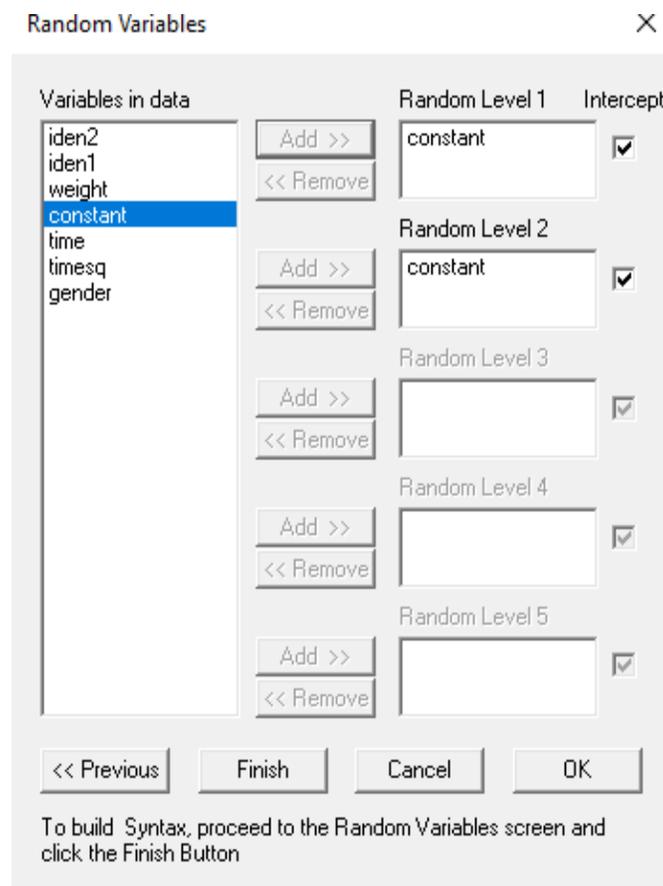
          

To build Syntax, proceed to the Random Variables screen and click the Finish Button

For the fully unconditional model we select WEIGHT as response variable.



Note that the box for Intercept is checked on the dialog box shown below. If it is unclicked, the variable CONSTANT should be selected as random at both levels on the **Random Variables** dialog box.



Click **Finish** to return to the main LISREL window where the completed syntax file **MOUSE.PRL** is now displayed.

```

L MOUSE.PRL
OPTIONS OLS=YES CONVERGE=0.001000 MAXITER=10 OUTPUT=STANDARD ;
TITLE=Mouse data: Variance decomposition;
SY='MOUSE.lsf';
ID2=iden2;
RESPONSE=weight;
FIXED=intcept;
RANDOM1=intcept;
RANDOM2=intcept;

```

With the exception of the use of the optional TITLE command, this input file is the most basic one which can be used for the analysis of a level-2 model. Note that in the OPTIONS command the default values of the options MAXITER, CONVERGE, and OUTPUT are used.

Run PRELIS with this input file. Convergence is achieved after 3 iterations, and the details of the last iteration, as given in the output file, are shown below.

```

NUMBER OF LEVEL 2 UNITS :      82
NUMBER OF LEVEL 1 UNITS :      698

```

N2 :	1	2	3	4	5	6	7	8
N1 :	9	9	9	9	9	9	9	9
N2 :	9	10	11	12	13	14	15	16
N1 :	9	9	9	9	9	9	9	9
N2 :	17	18	19	20	21	22	23	24
N1 :	9	9	9	9	9	9	9	9
N2 :	25	26	27	28	29	30	31	32
N1 :	9	9	9	9	9	9	9	9
N2 :	33	34	35	36	37	38	39	40
N1 :	9	9	9	9	9	9	9	9
N2 :	41	42	43	44	45	46	47	48
N1 :	9	9	8	8	8	8	8	8
N2 :	49	50	51	52	53	54	55	56
N1 :	8	8	8	8	8	8	8	8
N2 :	57	58	59	60	61	62	63	64
N1 :	8	8	8	8	8	8	8	8
N2 :	65	66	67	68	69	70	71	72
N1 :	8	8	8	8	8	8	8	8
N2 :	73	74	75	76	77	78	79	80
N1 :	8	8	8	8	8	8	8	8
N2 :	81	82						
N1 :	8	8						

Mouse data: Variance decomposition

-----  
CONVERGENCE REACHED IN 3 ITERATIONS  
-----

Mouse data: Variance decomposition

ITERATION NUMBER 3

+-----+  
| FIXED PART OF MODEL |  
+-----+

-----  
COEFFICIENTS                      BETA-HAT                      STD.ERR.                      Z-VALUE                      PR > |Z|  
-----  
intcept                              28.63410                      0.57021                      50.21634                      0.00000

+-----+  
| -2 LOG-LIKELIHOOD |  
+-----+

DEVIANCE= -2\*LOG(LIKELIHOOD) = 5425.49001592990  
NUMBER OF FREE PARAMETERS = 3

+-----+  
| RANDOM PART OF MODEL |  
+-----+

-----  
LEVEL 2                              TAU-HAT                      STD.ERR.                      Z-VALUE                      PR > |Z|  
-----  
intcept /intcept                      11.32910                      4.25185                      2.66451                      0.00771

-----  
LEVEL 1                              TAU-HAT                      STD.ERR.                      Z-VALUE                      PR > |Z|  
-----  
intcept /intcept                      130.32083                      7.42514                      17.55130                      0.00000

LEVEL 2 COVARIANCE MATRIX

intcept  
intcept      11.32910

#### LEVEL 2 CORRELATION MATRIX

```
intcept
intcept 1.0000
```

#### LEVEL 1 COVARIANCE MATRIX

```
intcept
intcept 130.32083
```

#### LEVEL 1 CORRELATION MATRIX

```
intcept
intcept 1.0000
```

In the first part of the abbreviated output file shown here, the data summary for the hierarchical structure is given. The first 42 level-2 units are the male mice and the last 40 the female mice.

From the random part of the output it can be seen that the variation over measurements (level 1) is large and overwhelms the variation between the mice (level 2). The so-called intraclass correlation can be calculated as:

$$\hat{\rho} = \frac{\hat{\Phi}_{(2)}}{\hat{\Phi}_{(2)} + \hat{\Phi}_{(1)}} = \frac{11.32910}{11.32910 + 130.32083} = 0.0799$$

indicating that about 8 percent of the variance in weight measurements is between mice. The value of  $-2\ln L$  (likelihood function) at convergence is 5425.4900.

### 3. Modeling linear growth

The variance decomposition model may now be extended by including the variable `TIME` as a fixed effect in the model. The model thus becomes

$$y_{ij} = \beta_0 + \beta_1 \text{TIME}_{ij} + u_{ij} + e_{ij}$$

with the variable `TIME` used as predictor of the response measurements.

The only change to the input file (`mouse2.prl`) is in the `FIXED` command, which now includes the variable `TIME`.



LEVEL 2 COVARIANCE MATRIX

	intcept
intcept	20.69397

LEVEL 2 CORRELATION MATRIX

	intcept
intcept	1.0000

LEVEL 1 COVARIANCE MATRIX

	intcept
intcept	16.46288

LEVEL 1 CORRELATION MATRIX

	intcept
intcept	1.0000

Both the fixed effects are highly significant, indicating significant variation in the intercepts and effect of time of measurement on the response variable over the different mice. An expected increase of 4.0922 grams in weight is expected for each increase of a week in TIME.

The log-likelihood value for this model is 4137.5788, compared to the value of 5424.4900 for the fully unconditional model. This reduction in the log-likelihood value indicates considerable variation between mice in their linear growth rates and also that the model fitted here explains more of the variation in the data than the previous one.

The random variation on level 2 of the model is higher and that on level-1 lower than in the fully unconditional model. The amount of variation in weight between mice is now calculated as

$$\hat{\rho} = \frac{\hat{\Phi}_{(2)}}{\hat{\Phi}_{(2)} + \hat{\Phi}_{(1)}} = \frac{20.69379}{20.69379 + 16.46288} = 0.5569,$$

that is 5 percent.

From the reduction in the level-1 variance component it can be seen that the variable TIME accounts for a considerable part of the variance previously noted on this level.

It is expected that the linear growth rate may vary from mouse to mouse around its mean value, rather than be fixed. The random component on level-2 of the hierarchy is thus extended to include TIME as well (**mouse3.prl**).



```

+-----+
| RANDOM PART OF MODEL |
+-----+

```

LEVEL 2		TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	/intcept	12.85320	2.80949	4.57492	0.00000
time	/intcept	-1.64430	0.59236	-2.77584	0.00551
time	/time	1.07389	0.19582	5.48415	0.00000

LEVEL 1		TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	/intcept	8.90088	0.54469	16.34126	0.00000

LEVEL 2 COVARIANCE MATRIX

	intcept	time
intcept	12.85320	
time	-1.64430	1.07389

LEVEL 2 CORRELATION MATRIX

	intcept	time
intcept	1.0000	
time	-0.4426	1.0000

LEVEL 1 COVARIANCE MATRIX

	intcept
intcept	8.90088

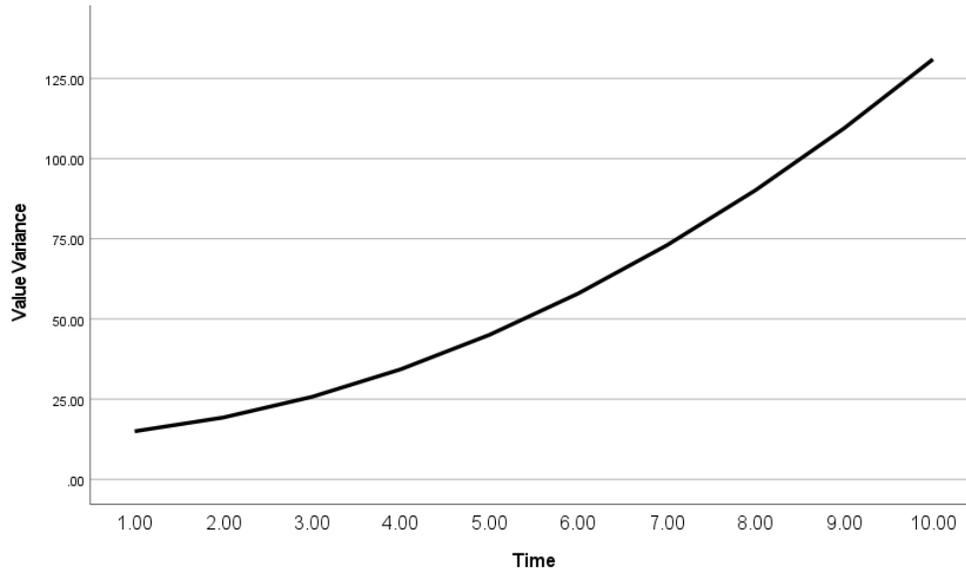
LEVEL 1 CORRELATION MATRIX

	intcept
intcept	1.0000

Once again a considerable reduction in the value of the function  $-2\ln L$  is noted. The estimates for the fixed effects in the model stayed fairly constant. On level 2 of the model we see that all three elements of the covariance matrix of random parameters  $\Phi_{(2)}$  are significant at a 5% level. The correlation between the intercept term and the TIME term is given as -0.4426. The level-1 or error variance is further reduced to 8.90088. It can thus be concluded that the inclusion of the variable TIME significantly reduced the variation between measurements, i.e., the level- units. The total variation on a particular level of the hierarchy may also be calculated. In this case, the total variance on level 2 is the variance of the sum of the two random coefficients associated with the intercept and TIME and may be written as:

$$\begin{aligned} \text{Var}\left(\text{Intercept}_{ij} + \text{TIME}_{ij}\Phi_{(2)\text{TIME},\text{TIME}}\right) &= \Phi_{(2)\text{INTERCEPT},\text{INTERCEPT}} + 2\Phi_{(2)\text{TIME},\text{INTERCEPT}}\left(\text{TIME}_{ij}\right) + \Phi_{(2)\text{TIME},\text{TIME}}\left(\text{TIME}_{ij}\right)^2 \\ &= 12.85320 + 2(-1.64430)\left(\text{TIME}_{ij}\right) + 1.07389\left(\text{TIME}_{ij}\right)^2 \end{aligned}$$

We can thus write the total variation at level 2 as a quadratic function of the variable TIME. A graph of this total variance against the nine time points is given in the figure below. The increase in variance over time is to be expected with data of this nature. It could also be an indication that the assumption that a parabolic function can adequately describe this phase in the development of the mice may not be valid and that other functions should be considered.



#### 4. Modeling non-linear growth

In data of this nature, it is unlikely that the increase in weight measurement will be linear for all mice over the time period concerned. A nonlinear component may be introduced in the model discussed in the previous section by adding a quadratic term (the variable TIMESQ) to the model. The model previously given is thus extended to:

$$y_{ij} = \beta_0 + \beta_1\text{TIME}_{ij} + \beta_2\text{TIMESQ}_{ij} + u_{0j} + u_{1j}\text{TIME}_{ij} + u_{2j}\text{TIMESQ}_{ij} + e_{ij}$$

The addition of the variable TIMESQ in this case leads to the following changes in the FIXED and RANDOM2 commands contained in the input file:

```
FIXED=intcept time timesq;
RANDOM2=intcept time timesq;
```

In order to obtain the empirical Bayes residuals for the level-2 models and the fitted values for each observation, the option OUTPUT = ALL is added to the OPTIONS command. The complete input file **mouse4.prl** is now:

```

L MOUSE4.prl
OPTIONS OUTPUT=ALL ;
TITLE=Mouse data: non-linear growth model;
SY=MOUSE.LSF;
ID2=iden2;
RESPONSE=weight;
FIXED=intcept time timesq;
RANDOM1=intcept;
RANDOM2=intcept time timesq;

```

Convergence of the iterative procedure was reached in five iterations, producing the following output.

ITERATION NUMBER 5

```

+-----+
| FIXED PART OF MODEL |
+-----+

```

COEFFICIENTS	BETA-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	4.16213	0.45748	9.09788	0.00000
time	6.90560	0.30980	22.29056	0.00000
timesq	-0.29629	0.02960	-10.01009	0.00000

```

+-----+
| -2 LOG-LIKELIHOOD |
+-----+

```

DEVIANCE= -2\*LOG(LIKELIHOOD) = 3400.93248789783  
NUMBER OF FREE PARAMETERS = 10

```

+-----+
| RANDOM PART OF MODEL |
+-----+

```

LEVEL 2	TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept /intcept	11.72906	2.70296	4.33933	0.00001
time /intcept	-5.86552	1.58228	-3.70699	0.00021
time /time	6.59372	1.23130	5.35509	0.00000
timesq /intcept	0.38927	0.14067	2.76733	0.00565
timesq /time	-0.55909	0.11275	-4.95880	0.00000
timesq /timesq	0.05814	0.01124	5.17424	0.00000

LEVEL 1	TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept /intcept	3.07863	0.20465	15.04328	0.00000

LEVEL 2 COVARIANCE MATRIX

	intcept	time	timesq
intcept	11.72906		
time	-5.86552	6.59372	
timesq	0.38927	-0.55909	0.05814

LEVEL 2 CORRELATION MATRIX

	intcept	time	timesq
intcept	1.0000		
time	-0.6670	1.0000	
timesq	0.4714	-0.9030	1.0000

LEVEL 1 COVARIANCE MATRIX

	intcept
intcept	3.07863

LEVEL 1 CORRELATION MATRIX

	intcept
intcept	1.0000

The fixed effects are all highly significant. There is an expected decrease of 0.2963 grams for each unit increase in the squared value of the time points. On the other hand, there is an estimated increase of 6.9056 grams with every increase of one week in time.

The expected value of the weights of the mice at time point number 2 may thus be calculated as

$$\begin{aligned} \text{Expected WEIGHT}_{i_2} &= 4.1621 + 2.00(6.9056) + 4.00(0.2963) \\ &= 16.7881 \text{ grams.} \end{aligned}$$

From the random part of the model it can be seen that all the estimates of the random coefficients at level 2 of the model are significant. This also holds for all the interaction terms at this level of the model. The correlation between TIME and TIMESQ is rather high, at -0.9030.

Variation over measurements, on level 1 of the model, has been drastically reduced through the inclusion of the variable TIMESQ in the analysis. When comparing  $-2\ln L$  for this model with that obtained for the linear growth model, a reduction of 473.7280 is noted, indicating that the inclusion of the variable TIMESQ significantly improved the fit of the model.

The empirical Bayes residuals and their variances for the first five mice, as given in the file **mouse4.ba2**, is given in the table below.

Mouse no.	Fixed effect	Empirical Bayes estimates
1	Intercept	$4.1621 + 5.9047 = 10.0668$
	TIME	$6.9056 - 3.6104 = 3.2952$
	TIMESQ	$-0.2963 + 0.3673 = 0.0710$
2	Intercept	$4.1621 + 0.1630 = 4.3251$
	TIME	$6.9056 - 1.1345 = 5.7711$
	TIMESQ	$-0.2963 + 0.1371 = 0.1592$
3	Intercept	$4.1621 - 3.1320 = 1.0301$
	TIME	$6.9056 + 2.3674 = 9.2730$
	TIMESQ	$-0.2963 - 0.1399 = -0.4362$
4	Intercept	$4.1621 + 0.1676 = 4.3297$
	TIME	$6.9056 - 0.1809 = 6.7247$
	TIMESQ	$-0.2963 + 0.0652 = 0.2311$
5	Intercept	$4.1621 - 5.5456 = 01.3835$
	TIME	$6.9056 + 0.9635 = 7.8691$
	TIMESQ	$-0.2963 + 0.0859 = 0.2104$

From this information the empirical Bayes estimates for any of the level-2 units may be computed. For the first five mice, these estimates are given in the following table.

Mouse no.	Fixed effect	Empirical Bayes estimates
1	Intercept	$4.1621 + 5.9047 = 10.0668$
	TIME	$6.9056 - 3.6104 = 3.2952$
	TIMESQ	$-0.2963 + 0.3673 = 0.0710$
2	Intercept	$4.1621 + 0.1630 = 4.3251$
	TIME	$6.9056 - 1.1345 = 5.7711$
	TIMESQ	$-0.2963 + 0.1371 = 0.1592$
3	Intercept	$4.1621 - 3.1320 = 1.0301$
	TIME	$6.9056 + 2.3674 = 9.2730$
	TIMESQ	$-0.2963 - 0.1399 = -0.4362$
4	Intercept	$4.1621 + 0.1676 = 4.3297$
	TIME	$6.9056 - 0.1809 = 6.7247$
	TIMESQ	$-0.2963 + 0.0652 = 0.2311$
5	Intercept	$4.1621 - 5.5456 = 01.3835$
	TIME	$6.9056 + 0.9635 = 7.8691$
	TIMESQ	$-0.2963 + 0.0859 = 0.2104$

The expected value of the weight of the first five mice at time point number 2 using the empirical Bayes estimates may thus be calculated as:

$$\text{WEIGHT}_{1_2} = 10.0668 + 2.00(3.2952) + 4.00(0.0710) = 16.9412 \text{ grams}$$

$$\text{WEIGHT}_{2_2} = 4.3251 + 2.00(5.7711) + 4.00(0.1592) = 15.2305 \text{ grams}$$

$$\text{WEIGHT}_{3_2} = 1.0301 + 2.00(9.2730) + 4.00(-0.4362) = 17.8313 \text{ grams}$$

$$\text{WEIGHT}_{4_2} = 4.3297 + 2.00(6.7247) + 4.00(0.2311) = 16.8547 \text{ grams}$$

$$\text{WEIGHT}_{5_2} = 9.7077 + 2.00(7.8961) + 4.00(-1.3835) = 19.9659 \text{ grams}$$

In the case of mice numbers 1, 3, 4, and 5, the estimated weights thus obtained are higher than previously calculated, while mouse number 2 is below the previously calculated value of 16.7881 units.

Finally, the residuals for the first 45 observations, that is the first five male mice, are considered. The following is an extract from the output file **mouse4.res**:

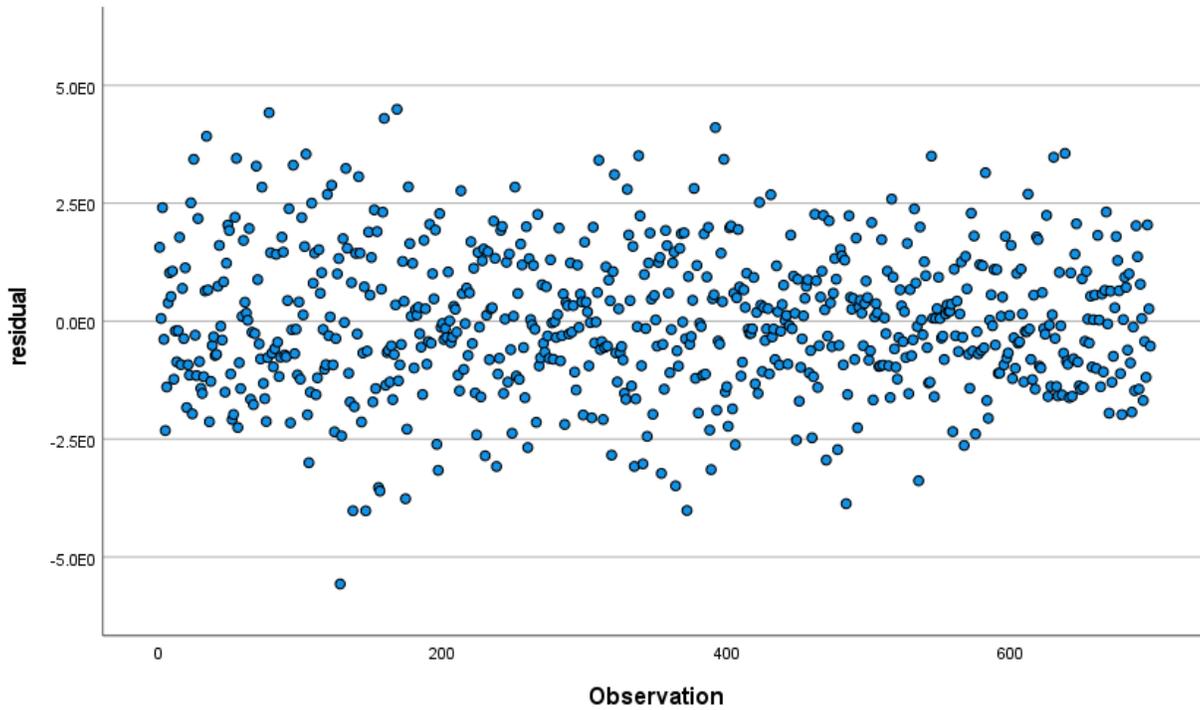
File	Edit	Format	View	Help			
					1	1	15.000
					2	1	17.000
					3	1	23.000
					4	1	24.000
					5	1	26.000
					6	1	31.000
					7	1	37.000
					8	1	42.000
					9	1	46.000
					10	2	11.000
					11	2	14.000
					12	2	20.000
					13	2	24.000
					14	2	29.000
					15	2	35.000
					16	2	36.000
					17	2	41.000
					18	2	43.000
					19	3	11.000
					20	3	16.000
					21	3	24.000
					22	3	30.000
					23	3	39.000
					24	3	39.000
					25	3	48.000
					26	3	47.000
					27	3	48.000
					28	4	13.000
					29	4	16.000
					30	4	21.000

					13.433	1.5670
					16.941	0.58731E-01
					20.592	2.4085
					24.384	-0.38379
					28.318	-2.3181
					32.394	-1.3944
					36.613	0.38726
					40.973	1.0269
					45.475	0.52453
					9.9370	1.0630
					15.230	-1.2305
					20.205	-0.20547
					24.862	-0.86203
					29.200	-0.20016
					33.220	1.7802
					36.921	-0.92109
					40.304	0.69611
					43.368	-0.36826
					9.8669	1.1331
					17.831	-1.8312
					24.923	-0.92315
					31.143	-1.1426
					36.490	2.5104
					40.964	-1.9642
					44.566	3.4336
					47.296	-0.29607
					49.153	-1.1533
					10.823	2.1767
					16.855	-0.85468
					22.424	-1.4238

The largest residual for the male mice is for observation number 25 which represents the fifth measurement for mouse 4, where a residual of 3.4336 is encountered.

A plot of the residuals against the observation numbers is shown below. For a more detailed discussion of the analysis of residuals in a multilevel context, the user is referred to Goldstein (1987, pp. 21-26).



## 5. Introducing a covariate while modeling non-linear growth

In this example we want to determine whether there is a significant difference between the growth pattern of the male and female mice, as modeled in the non-linear growth model discussed previously. This can be determined by adding the gender of the mice as covariate to the model fitted in the previous section.

The variable GENDER is introduced as covariate by modifying the FIXED command to the following:

```
FIXED=intcept time timesq gender gender*time gender*timesq;
```

Edit the previous file accordingly and save it under a different name (**mouse5.prl**). Run PRELIS with this input file to fit the model to the data. Convergence is reached after 4 iterations and the following output is obtained.

```
ITERATION NUMBER      4
```

```
+-----+
| FIXED PART OF MODEL |
+-----+
```

COEFFICIENTS	BETA-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	4.19133	0.44972	9.31996	0.00000
time	6.87771	0.29532	23.28885	0.00000
timesq	-0.29399	0.02900	-10.13902	0.00000
gender	-0.81492	0.44972	-1.81207	0.06998
gender *time	0.87235	0.29532	2.95389	0.00314
gender *timesq	-0.05947	0.02900	-2.05087	0.04028

```

+-----+
|  -2 LOG-LIKELIHOOD  |
+-----+

```

DEVIANCE= -2\*LOG(LIKELIHOOD) = 3389.53831478832  
NUMBER OF FREE PARAMETERS = 13

```

+-----+
|  RANDOM PART OF MODEL  |
+-----+

```

LEVEL 2		TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	/intcept	11.07339	2.60146	4.25660	0.00002
time	/intcept	-5.16047	1.46703	-3.51763	0.00044
time	/time	5.83687	1.11334	5.24265	0.00000
timesq	/intcept	0.34138	0.13349	2.55735	0.01055
timesq	/time	-0.50761	0.10447	-4.85897	0.00000
timesq	/timesq	0.05464	0.01069	5.11098	0.00000

LEVEL 1		TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	/intcept	3.07844	0.20464	15.04338	0.00000

LEVEL 2 COVARIANCE MATRIX

	intcept	time	timesq
intcept	11.07339		
time	-5.16047	5.83687	
timesq	0.34138	-0.50761	0.05464

LEVEL 2 CORRELATION MATRIX

	intcept	time	timesq
intcept	1.0000		
time	-0.6419	1.0000	
timesq	0.4389	-0.8989	1.0000

LEVEL 1 COVARIANCE MATRIX

	intcept
intcept	3.07844

LEVEL 1 CORRELATION MATRIX

	intcept
intcept	1.0000

The coefficients of GENDER\*TIME and GENDER\*TIMESQ are significant at a 5 percent level, but the coefficient for GENDER\*Intercept = GENDER is not. Only small changes are noticeable when the random part of the model fitted is compared to the corresponding section of the output obtained in the third example.

When comparing the values of  $-2\ln L$  for the two models, a reduction of 11.39 is noted. It can thus be concluded that, although the gender of the mice has no significance on the intercept, there are significant differences between the growth patterns of male and female mice prior to physical maturity.

**6. Complex variation at level 1 of the model**

In the final example of a level 2 model, the linear growth model fitted previously is extended to include complex variation on both levels of the hierarchy. The term “complex variation” refers to the existence of two or more random variables at the same level of the hierarchy. We include the variable TIME in this model to illustrate such a model.

We modify the RANDOM1 command previously used to include TIME as shown in **mouse6.prl**:

```

OPTIONS MAXITER=25 ;
TITLE=Mouse data: complex variation ;
SY=MOUSE.LSF;
ID2=iden2;
RESPONSE=weight;
FIXED=intcept time;
RANDOM1=intcept time;
RANDOM2=intcept time;
    
```

The output obtained for this model is given below. This model needed 18 iterations for convergence. The MAXITER option of the OPTIONS command was used to increase the number of iterations from the default of 10 to 25.

```

+-----+
| FIXED PART OF MODEL |
+-----+
    
```

COEFFICIENTS	BETA-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	8.67635	0.46342	18.72259	0.00000
time	4.30965	0.13067	32.98166	0.00000

```

+-----+
| -2 LOG-LIKELIHOOD |
+-----+
    
```

DEVIANCE=  $-2*\text{LOG}(\text{LIKELIHOOD}) = 3822.83149832815$   
 NUMBER OF FREE PARAMETERS = 8

```

+-----+
| RANDOM PART OF MODEL |
+-----+

```

LEVEL 2		TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	/intcept	13.35035	2.77510	4.81076	0.00000
time	/intcept	-1.61832	0.61801	-2.61861	0.00883
time	/time	1.17651	0.21925	5.36601	0.00000

LEVEL 1		TAU-HAT	STD.ERR.	Z-VALUE	PR >  Z
intcept	/intcept	13.73386	2.13539	6.43156	0.00000
time	/intcept	-2.91233	0.54471	-5.34654	0.00000
time	/time	0.83668	0.12967	6.45252	0.00000

LEVEL 2 COVARIANCE MATRIX

	intcept	time
intcept	13.35035	
time	-1.61832	1.17651

LEVEL 2 CORRELATION MATRIX

	intcept	time
intcept	1.0000	
time	-0.4083	1.0000

LEVEL 1 COVARIANCE MATRIX

	intcept	time
intcept	13.73386	
time	-2.91233	0.83668

LEVEL 1 CORRELATION MATRIX

	intcept	time
intcept	1.0000	
time	-0.8591	1.0000

From the random part of the output it can be seen that there is an increase in the variance of the intercept term on level 1 of the model when the coefficient for the variable TIME is also allowed to vary randomly over level 1 of the model. The error 1 variance, however, is also a function of the covariance between the intercept and TIME. All coefficients are highly significant.

When the two values of  $-2\ln L$  are compared for these models, a decrease of 50.43 is noted. The addition of the coefficient for the variable TIME on level 1 of the model thus seems to lead to an improved fit compared with the linear growth model. This implies that the level-1 error variances are heteroscedastic.

## 7. Conclusions

In the five examples discussed here, various models were considered for the analysis of repeated measurement data with a level-2 hierarchical structure. These models included a variance decomposition model, two linear growth models and a nonlinear growth model. The inclusion of a covariate and the possibility of complex level-1 variation were also considered.

When the respective  $-2\ln L$  values of these models are compared, the nonlinear model with a covariate included had the lowest value, namely, 3389.5383. It would thus appear that the growth of the 82 mice up to physical maturity can best be described by a parabola with the gender of the mice as a covariate. From the scatterplot shown previously, however, it seems as if other nonlinear functions for the modeling of the growth of the mice can also be considered.