

Fecal Fat Example on p.254

The model (from page 256 equation (8.2)) is $\text{FECFAT}_{st} = \mu + \text{SUBJECT}_s + \text{PILLTYPE}_t + \varepsilon_{st}$, for $s = 1 \dots 6$ subjects, $t = 1 \dots 4$ pill-types, $\varepsilon_{st} \sim N(0, \sigma_\varepsilon^2)$ and $\text{SUBJECT}_s \sim N(0, \sigma_{\text{subj}}^2)$ - thus subjects are "random".

General Linear Model: fecfat versus subject, pilltype

```

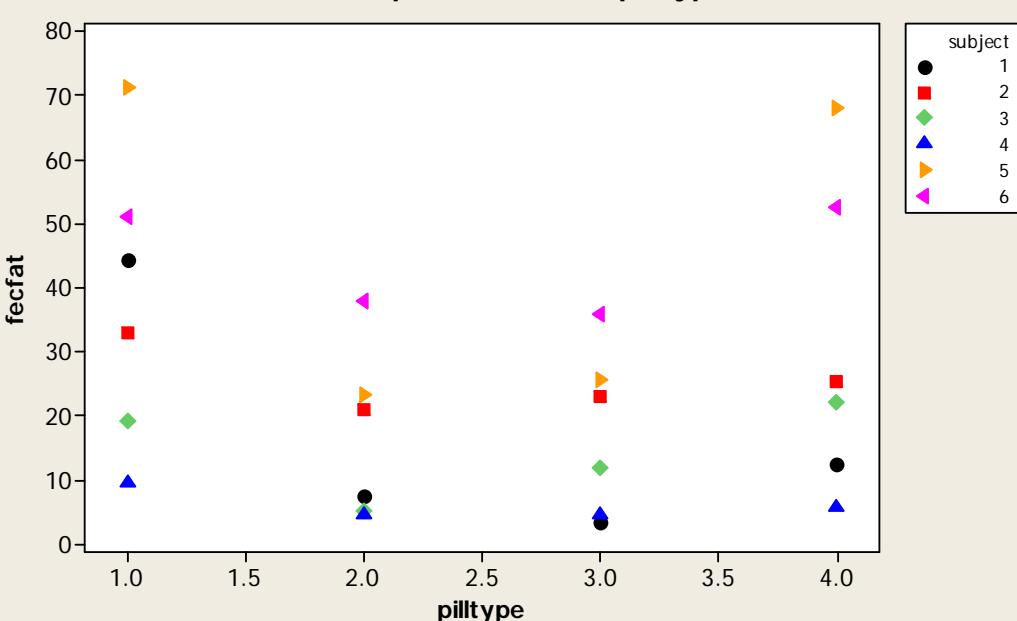
Factor      Type    Levels   Values
subject     random    6  1, 2, 3, 4, 5, 6
pilltype    fixed     4  1, 2, 3, 4

Analysis of Variance for fecfat, using Adjusted SS for Tests
Source   DF  Seq SS  Adj SS  Adj MS    F      P
subject   5  5588.4  5588.4  1117.7  10.45  0.000
pilltype  3  2008.6  2008.6  669.5   6.26  0.006
Error     15 1605.0  1605.0  107.0
Total     23 9202.0

S = 10.3440  R-Sq = 82.56%  R-Sq(adj) = 73.26%

Expected Mean Squares, using Adjusted SS
      Expected Mean Square
      Source for Each Term
1  subject (3) + 4.0000 (1)
2  pilltype (3) + Q[2]
3  Error   (3)

Variance Components, using Adjusted SS
      Estimated
Source      Value
subject     252.7
Error       107.0
  
```

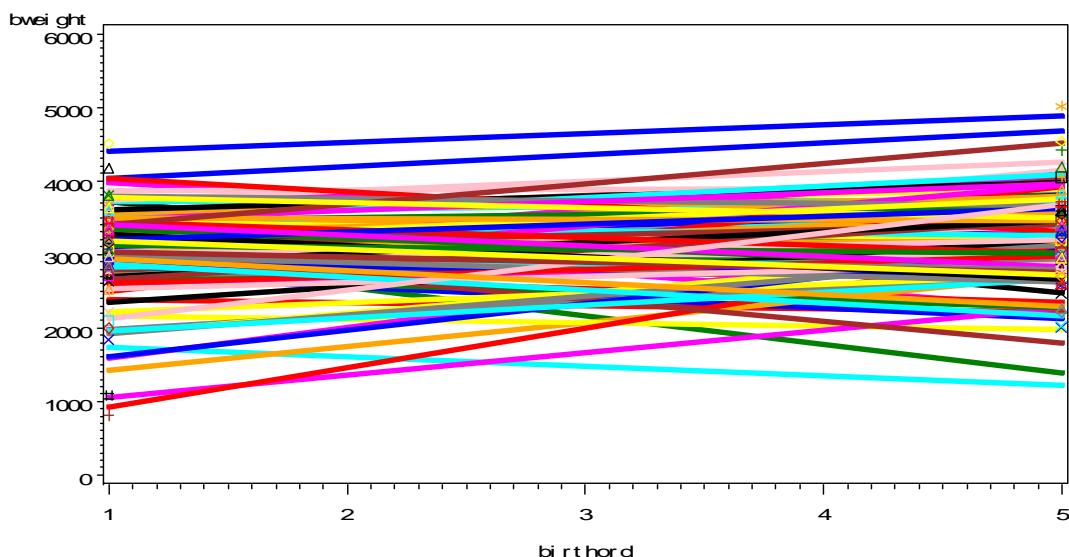
Scatterplot of fecfat vs pilltype

Now, "finish the story" by performing a MCP (MSP = mean separation procedure)

Fecal Fat Multiple Comparisons (using Bonferroni method)

Least Squares Means for fecfat					
pilltype Mean					
1	38.08				
2	16.53				
3	17.42				
4	31.07				
Bonferroni Simultaneous Tests					
Response Variable fecfat					
All Pairwise Comparisons among Levels of pilltype					
<u>pilltype = 1 subtracted from:</u>					
Difference SE of Adjusted					
pilltype of Means	Difference	SE of Difference	T-Value	P-Value	
2	-21.55	5.972	-3.608	0.0155	
3	-20.67	5.972	-3.461	0.0210	
4	-7.02	5.972	-1.175	1.0000	
<u>pilltype = 2 subtracted from:</u>					
Difference SE of Adjusted					
pilltype of Means	Difference	SE of Difference	T-Value	P-Value	
3	0.8833	5.972	0.1479	1.0000	
4	14.5333	5.972	2.4335	0.1676	
<u>pilltype = 3 subtracted from:</u>					
Difference SE of Adjusted					
pilltype of Means	Difference	SE of Difference	T-Value	P-Value	
4	13.65	5.972	2.286	0.2235	

GA Babies – p.262ff: only examining first- and last-born in families of 5



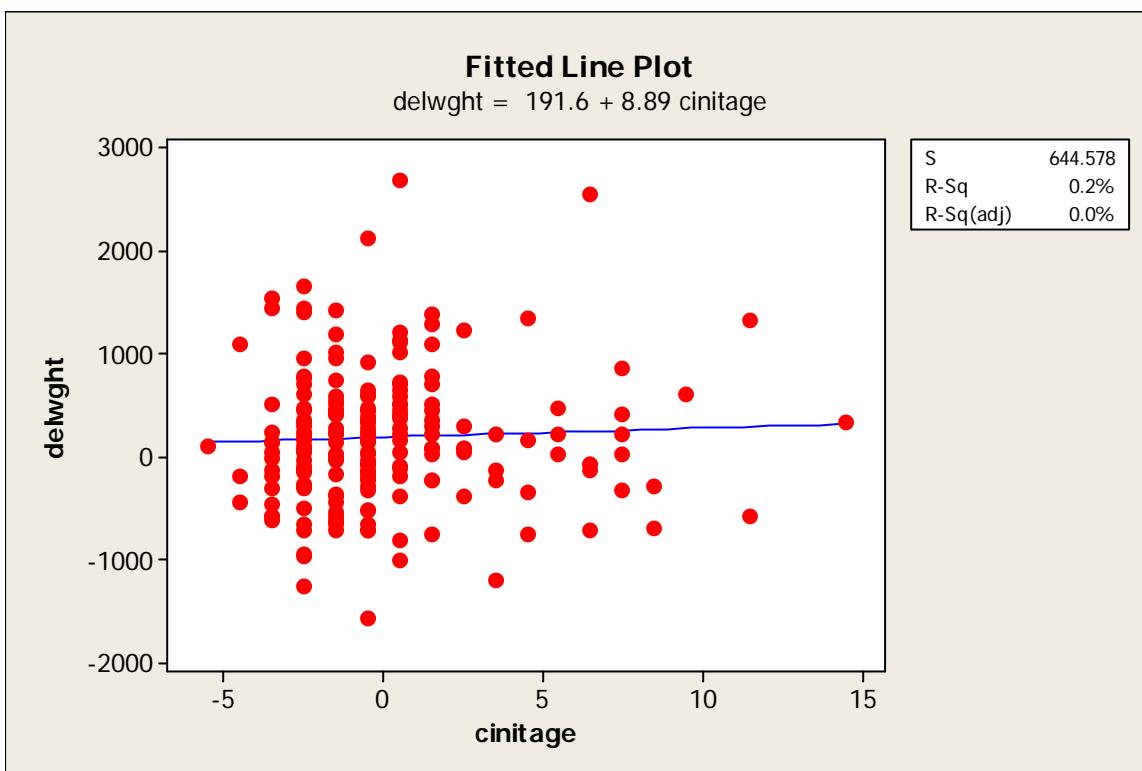
A. Paired t-test**One-Sample T: delwght**

Test of mu = 0 vs not = 0

Variable	N	Mean	StDev	SE Mean	95% CI	T	P
delwght	200	191.6	643.6	45.5	(101.9, 281.4)	4.21	0.000

B. Now take account of centered initial age (CINITAGE): initial age mean = 17.545 subtracted from each value of initial age

Model Function: DELWGHT = BW5 – BW1 = $\beta_0 + \beta_1 * CINITAGE + \epsilon$

**Regression Analysis: delwght versus cintage**

The regression equation is
delwght = 192 + 8.9 cintage

Predictor	Coef	SE Coef	T	P
Constant	191.64	45.58	4.20	0.000
cintage	8.89	14.16	0.63	0.531

S = 644.578 R-Sq = 0.2% R-Sq(adj) = 0.0%

Analysis of Variance					
Source	DF	SS	MS	F	P
Regression	1	163789	163789	0.39	0.531
Residual Error	198	82265157	415481		

- C. A variant of **Repeated Measures (RM) analysis** is performed in Table 8.6 on p.264. The key term is the interaction between birth order and centered initial age (called “_IbirXcinit~5”). That this term is non-significant ($p = 0.528$) tells us that the difference in birthweights between the first- and last-born is not related to (centered) initial age, exactly the same question answered in part B above. We get essentially the same results using SAS.

```
data three;
  set one; if birthord=1 or birthord=5;
  d5=(birthord=5); d5cinitage=d5*cinitage;
proc genmod data=three;
  class momid;
  model bweight=d5 cinitage d5*cinitage/dist=normal link=identity;
  repeated subject=momid / type=exch;
run;
```

The GENMOD Procedure

Model Information

Data Set	WORK.THREE
Distribution	Normal
Link Function	Identity
Dependent Variable	bweight

Number of Observations Read	400
Number of Observations Used	400

Class Level Information

Class	Levels	Values
momid	200	39 62 79 80 84 92 108 113 125 135 199 200 221 247 304 336 547 853 960 1232 1448 1638 1706 1785 1856 2083 2166 2292 2301 2383 2519 2598 2613 2647 2735 2822 2899 2906 2918 2928 3044 3168 3308 3377 3431 3438 3469 3477 3480 3504 3526 3668 3797 3838 3936 ...

Parameter Information

Parameter	Effect
Prm1	Intercept
Prm2	d5
Prm3	cinitage
Prm4	d5*cinitage

Criteria For Assessing Goodness Of Fit

Criterion	DF	Value	Value/DF
Deviance	396	129458162.97	326914.5530
Scaled Deviance	396	400.0000	1.0101
Pearson Chi-Square	396	129458162.97	326914.5530

Scaled Pearson X2	396	400.0000	1.0101																																										
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D. Finally, let's regress BW5 on CINITAGE and BW1 (see p.265) via

Model Function: $BW5 = \beta_0 + \beta_1 * CINITAGE + \beta_2 * BW1 + \epsilon$

Regression Analysis: bw5 versus cinitage, bw1

The regression equation is
bw5 = 2114 + 24.9 cinitage + 0.363 bw1

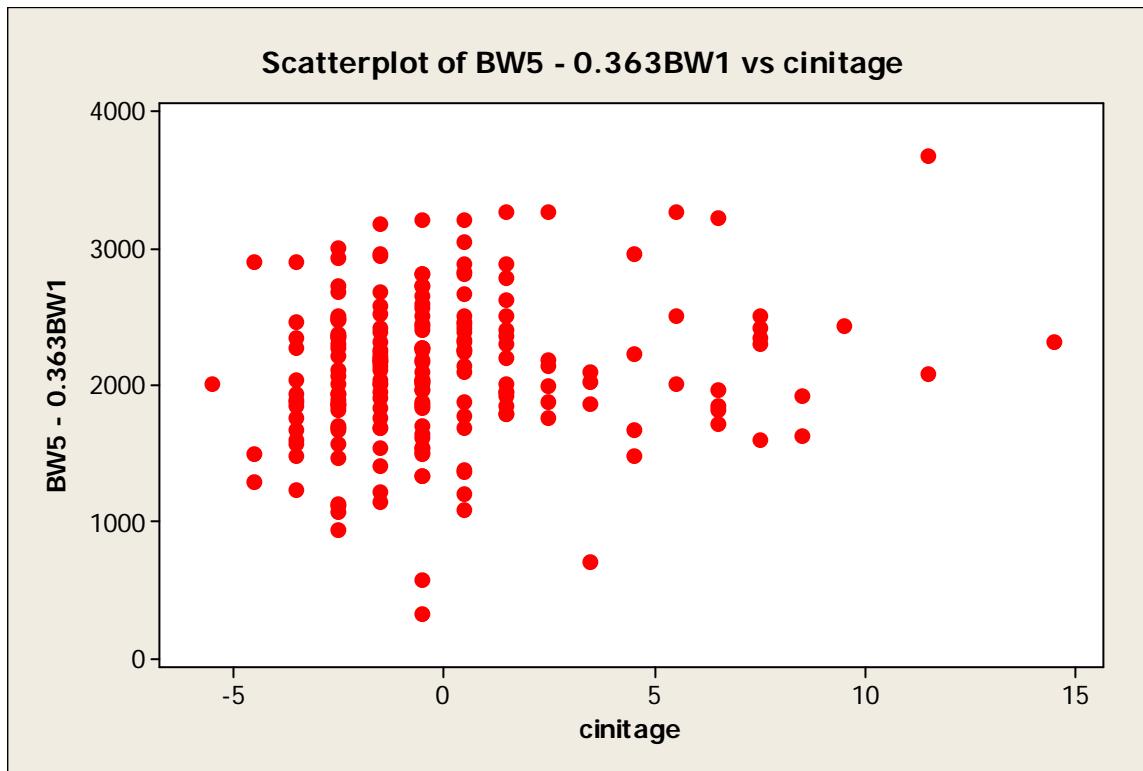
Predictor	Coef	SE Coef	T	P
Constant	2113.6	202.7	10.43	0.000
cinitage	24.91	11.82	2.11	0.036
bw1	0.36286	0.06604	5.49	0.000

S = 532.527 R-Sq = 16.4% R-Sq(adj) = 15.6%

Analysis of Variance

Source	DF	SS	MS	F	P
Regression	2	10961363	5480682	19.33	0.000
Residual Error	197	55866154	283585		
Total	199	66827517			

Note that β_2 above is estimated to be 0.363, so we let $Y = BW5 - 0.363 \cdot BW1$ in the following graph, and now note the significant relationship with CINITAGE ($p = 0.036$).



The above results notwithstanding, the authors recommend against using first-born birthweight as a covariate in observational studies such as this one (see discussion on p.265 center and bottom). In this case, we would conclude that the difference between first- and last-born birthweights does not depend upon the initial age of the mother. But, the conclusion is really secondary here – here, the authors are demonstrating that there are several methods to analyze repeated measures data. Obviously, the only one of these that is useful when we have more than two time-points is the third method (repeated measures) above; this method is called the GEE method, short for Generalized Estimating Equations. The authors claim that it is preferred as it is “robust”, that is, it assumes no specific variance-covariance pattern.