A study by Apex Enterprises investigated the ratings of potential employees by its personnel officers. Five personnel officers were selected at random, and four prospective employees were assigned at random to each selected officer.

options nodate nonumber ls=80 nocenter; data apex; input officer \$ rating @0; datalines;

A 76 A 65 A 85 A 74 B 59 B 75 B 81 B 67 C 49 C 63 C 61 C 46 D 74 D 71 D 85 D 89 E 66 E 84 E 80 E 79

proc varcomp data = apex; class officer;

model rating = officer; run;

Variance Component rating 80.41042 Var(officer) Var(Error) 73.28333

proc mixed data = apex cl alpha=0.1;

class officer;

model rating = / cl alpha=0.05 DDFM = SATTERTHWAITE; random officer; run;

The cl option after data=apex asks for the confidence limits.

The class statement lists all the categorical variables just as in glm.

The model rating =; line looks strange. In proc mixed, the model statement lists only the fixed effects. Then the random effects are listed separately in the random statement. In our example, there were no fixed effects, so we had no predictors on the model line. We had one random effect, so it went on the random line.

This is different from qlm, where all the factors (fixed and random) are listed on the model line, and then the random ones are repeated in the random statement.

Just in case you're not confused enough, proc varcomp assumes all factors are random effects unless they are specified as fixed ...

proc mixed gives a huge amount of output. Here are some pieces of it.

Cov Parm	Estimate	Alpha	Lower	Upper
officer	80.4104	0.1	29.5215	865.42
Residual	73.2833	0.1	43.9774	151.39

Solution fo	or Fixed Effec	ts				
Effect	Estimate	Standard Error	DF	t Value	Pr > t	Alpha
Intercept	71.4500	4.4437	4	16.08	<.0001	0.05
	Solution for	Fixed Effects	_	Effect Entercept	Lower 59.1124	Upper 83.7876

Example Pages 1082-1083 #'s 17.5, 17.6

One Factor in RCB - Random Factor & Random Block

options nodate nonumber ls=80 nocenter;

data dna;

inpu	t Sub	ject	Ana	lyst	DNAco	ntent	@@;	
data	lines	5 ;						
1	1	9.9000	2	1	10.6000	3	1	11.5000
4	1	11.3000	5	1	10.5000	6	1	8.0000
7	1	10.6000	8	1	12.2000	9	1	8.0000
10	1	9.7000	1	2	10.2000	2	2	10.6000
3	2	11.3000	4	2	11.6000	5	2	10.3000
6	2	8.2000	7	2	10.7000	8	2	12.8000
9	2	7.9000	10	2	9.6000	1	3	10.1000
2	3	10.5000	3	3	11.1000	4	3	11.3000
5	3	10.1000	6	3	7.9000	7	3	10.4000
8	3	12.6000	9	3	7.7000	10	3	9.3000
1	4	10.2000	2	4	10.5000	3	4	11.2000
4	4	11.3000	5	4	10.2000	6	4	7.9000
7	4	10.5000	8	4	12.7000	9	4	7.8000
10	4	9.4000	1	5	10.4000	2	5	10.9000
3	5	11.4000	4	5	11.6000	5	5	10.6000
6	5	8.4000	7	5	10.9000	8	5	12.5000
9	5	8.1000	10	5	9.5000			
;								

proc glm data = dna;
class subject analyst;
model DNAcontent = subject analyst;
random subject analyst / test;
run;

Source Type III Expected Mean Square

Subject Var(Error) + 5 Var(Subject)

Analyst Var(Error) + 10 Var(Analyst)

Tests of Hypotheses for Random Model Analysis of Variance

Dependent Variable: DNAcontent

Source	DF	Type III SS	Mean Square	F Value	Pr > F
Subject Analyst	9 4	94.457000 0.666000	10.495222 0.166500	538.22 8.54	<.0001 <.0001
Error: MS(Error)	36	0.702000	0.019500		

proc mixed data = dna cl;
class subject analyst;
model DNAcontent = / solution cl DDFM = SATTERTHWAITE;
random subject analyst;
run;

Cov Parm	Estimate	Alpha	Lower	Upper
Subject	2.0951	0.05	0.9901	7.0032
Analyst	0.01470	0.05	0.004788	0.1895
Residual	0.01950	0.05	0.01290	0.03290

Solution for Fixed Effects

		Standard				
Effect	Estimate	Error	DF	t Value	Pr > t	Alpha
Intercept	10.2500	0.4614	9.25	22.22	<.0001	0.05

Solution for Fixed Effects

Effect Lower Upper Intercept 9.2106 11.2894