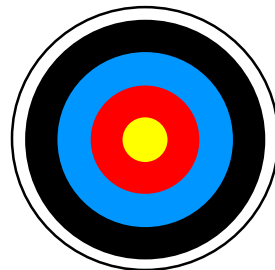


Statistical Techniques II

EXST7015

Post-ANOVA
More LSMeans



LSMeans

- **There is something else about the SAS LSMeans statement you should know.**
- **There are actually several "unusual" or unexpected behaviors of this statement. One we will discuss in connection with AnCova.**
- **However, there is another general behavior that we should see first.**

LSMeans (*continued*)

- What is the overall mean?

	Rep						Tmt
Tmt	1	2	3	4	5		Mean
1	2	4	6				4
2	2	6					4
3	3	3	7	8	9		6
4	4						4
5	3	4	6	7			5
6	5	6	7				6
7	3	5					4
				Sum	100		
				n	20		

LSMeans (*continued*)

	Rep					Tmt
Tmt	1	2	3	4	5	Mean
1	2	4	6			4
2	2	6				4
3	3	3	7	8	9	6
4	4					4
5	3	4	6	7		5
6	5	6	7			6
7	3	5				4
Mean	5			Sum	100	33
LSMean	4.71			n	20	7

LSMeans (*continued*)

- **LSMeans calculates means as the mean of means, not the raw mean of all observations.**
- **This is particularly important in unbalanced factorial designs.**
- **For one unbalanced 4 by 5 factorial the means and lsmeans are given below.**

LSMeans (*continued*)

- Raw data

	Tmt2				
Tmt1	1	2	3	4	5
1	2	3	1	2	2
	3	4	2	4	3
	4	5	3		3
					4
2	5	6	4	8	3
	9	6	5		3
			6		
3	4	6	4	3	8
	5	8	6	7	
	7				
	8				
4	7	6	5	4	5
	8	9	7	7	6
	8	9		7	7
	9				8

LSMeans (*continued*)

■ Comparison of Means & LSMeans.

	Tmt2						
Tmt1	1	2	3	4	5	LSMean	Raw Mean
1	3	4	2	3	3	3.00	3.00
2	7	6	5	8	3	5.80	5.50
3	6	7	5	5	8	6.20	6.00
4	8	8	6	6	6.5	6.90	7.00
LSMean	6.00	6.25	4.50	5.50	5.13	5.48	5.35
Mean	6.08	6.20	4.30	5.25	4.73		

LSMeans (*continued*)

- Which is better?
- This depends on the situation. Suppose we caught fish in the summer and in the winter, and wanted to express the average temperature at which fish were caught.
- The winter mean is 15c and the summer mean is 25c. What is the mean.

LSMeans (*continued*)

- **We do the calculations on the individual catches and find the mean is equal to 24.**
- **How can that be?**
- **Well we did 180 samples in the summer and only 20 samples in the winter. So the summer temperatures dominate our samples.**

LSMeans (*continued*)

- Perhaps the average temperature would be better expressed as 20, the mean of the means. That is LSMeans
- I generally use LSMeans.
- When testing hypotheses such as $H_0: \mu_1 = \mu_2 = \mu_3$ it is best that the overall mean not be dominated by some cell that has an unusually high number of observations.

LSMeans (*continued*)

- **On the other hand, cells with more observations are better estimates of the mean than cells with fewer estimates.**
- **If the null hypothesis is true, why lose power by treating the cells equally?**
- **Traditional ANOVA will use RAW means in its calculation.**
- **The choice is yours, except that PROC MIXED has only the LSMeans.**

Testing for differences between models

- **PROC MIXED provides several tools for comparing models**
- **The intent is to compare between full and reduced models. The statistics used differ from those used in regression.**
 - ▶ **Reduced models may be models with some terms omitted, or**
 - ▶ **Reduced models may be models with a simpler variance or covariance structure**

Testing for differences between models (*continued*)

- The test is called a likelihood ratio test.
 - ▶ It produces a Chi square statistic.
 - ▶ The degrees of freedom are the d.f. difference between the two models.

Testing for differences between models (*continued*)

- Homogeneous variance is tested automatically with some simple models
- Recall our Typhoid strain example, we requested separate variances for each group with the statement
 - ▶ `repeated / group=strain;`
- The resulting output was
 - ▶ Null Model Likelihood Ratio Test
 - ▶ DF Chi-Square Pr > ChiSq
 - ▶ 2 14.56 0.0007

Testing for differences between models (*continued*)

- Note that fitting 3 variances requires 3 d.f., while fitting a homogeneous variance model requires only 1 d.f.
- The 2 d.f. difference are the reason the test on the preceding page is a 2 d.f. model.
- This test is very similar to Bartlett's test of homogeneity of variance.

Testing for differences between models *(continued)*

- Suppose that for the baseball example you were told that the salaries of the some positions were highly variable, while others were more stable.
- Perhaps we should have tested for nonhomogeneous for this example.
- So we add the statement,
 - ▶ `repeated / group=strain;`

Testing for differences between models *(continued)*

- SAS fits the different variances for the positions, but does not provide a test.
- For some cases we will not get this test automatically. In that case we can calculate it ourselves.
- For the original fit we got the results,

► Covariance Parameter Estimates

►		Standard	Z				
► Cov Parm	Estimate	Error	Value	Pr > Z	Alpha	Lower	Upper
► team	3466.41	30458	0.11	0.4547	0.05	513.45	3.81E125
► Residual	1924296	145057	13.27	<.0001	0.05	1668871	2243534

Testing for differences between models *(continued)*

- When separate variances are requested we get the following results,

► Covariance Parameter Estimates

		Estimate	Standard Error	Z	Pr > Z	Alpha	Lower	Upper
► Cov Parm	Group							
► team		25008	35506	0.70	0.2406	0.05	4960.25	26828515
► Residual	Position 1b	3126672	0
► Residual	Position 2b	2276275	902599	2.52	0.0058	0.05	1189304	5985011
► Residual	Position 3b	1512066	600277	2.52	0.0059	0.05	789517	3981295
► Residual	Position c	759251	201637	3.77	<.0001	0.05	479387	1382686
► Residual	Position if	626561	240028	2.61	0.0045	0.05	333467	1582294
► Residual	Position of	2558744	407215	6.28	<.0001	0.05	1916409	3590143
► Residual	Position p	1875902	208345	9.00	<.0001	0.05	1526216	2361923
► Residual	Position ss	1384956	364052	3.80	<.0001	0.05	878092	2504484

- The first model estimated 2 parameters, while this model fits 9, a difference of 7.

Testing for differences between models *(continued)*

- **SAS reports the number of parameters fitted in the "Dimensions" section.**
 - ▶ **In order to do this 7 d.f. test we take the difference in the "-2 Res Log Likelihood" reported in the "Fit Statistics".**
 - ▶ **This value was 6346.8 for the reduced model and 6323.1 for the full model.**
 - ▶ **The difference is 23.7, a chi square value with 7 d.f.**

Testing for differences between models (*continued*)

- The probability of a greater chi square value is 0.001286226, a significant result.
- As with regression, when there is a difference in two models the larger model is better, since it presumably provides some information that the smaller model does not.
- If there is no significant difference we decide in favor of the simpler model.
- We just tested homogeneity of variance.

Other between model comparisons

- **SAS also provides some other statistics to compare between models. Also under the "Fit statistics" you will find**
 - ▶ **AIC (smaller is better) 6341.1**
 - ▶ **AICC (smaller is better) 6341.6**
 - ▶ **BIC (smaller is better) 6346.8**
- **And for the smaller model**
 - ▶ **AIC (smaller is better) 6350.8**
 - ▶ **AICC (smaller is better) 6350.8**
 - ▶ **BIC (smaller is better) 6352.1**

Other between model comparisons (*continued*)

- These are all penalized index values called "Information Criteria". As the note says, smaller is better for all 3.
- **AIC** is the Akaike Information Criteria
- **AICC** is the "Corrected AIC "
- **BIC** is the Bayesian Information Criterion
-
- and there are others.

Other between model comparisons (*continued*)

- These all work in a similar fashion. They provide an adjusted measure of goodness of fit.
- These are similar in concept to the "adjusted R^2 ", so they do not necessarily get smaller when the model gets larger.
- These results also indicate that the full model is better, but they do not provide a test with a probability value.