# Open lecture on statistics

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# Statistical notes for clinical researchers: A one-way repeated measures ANOVA for data with repeated observations

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### Repeated measures and correlated data

Repeated measurements can be obtained in many situations. A simple example is a pre-post design to assess the intervention effect. With respect to conditions, e.g., treatment A, B, and C, different treatments may be applied to the same group of subjects repeatedly according to the designated time schedule to compare the effects of treatments. In perspective of time, we may measure children's height repeatedly to know the growth pattern. In the field of dentistry, we measure the periodontal pocket depths of a tooth repeatedly at different surfaces: buccal, lingual, mesial, and distal. Common characteristics of these repeated measurements include correlations of the data points obtained by the same object, person, tooth, etc. The paired *t*-test is an analyzing method of correlated samples with two time points or occasions. For three or more time points or repeated conditions, we may use the repeated measures ANOVA which is equivalent to the one-way ANOVA for independent samples. Therefore, continuous outcome variables and categorical independent variables are the basic requirements.

### Comparison of one-way repeated measures ANOVA and classical one-way ANOVA

Table 1 shows a hypothetical data with repeated observations, P1, P2, and P3, which may represent scores measured in different conditions or different time points. If we disregard the correlated structure of three scores, the data can be analyzed using the one-way ANOVA. The classical one-way ANOVA decomposes the total variation of scores (SST, total sum of squares) into between-group variance (SSB, SS by conditions in Table 2) and within-group variance (SSW, SS residual in Table 2). The total variation is explained by SSB of different conditions (P1, P2, and P3), and the remainder is a substantial amount of unexplained within-group variance (Table 2). On the other hand, in the one-way repeated measures ANOVA, SSW is divided into SS by subject and unexplained SS (SSE, error sum of squares), which results in a significant difference among different conditions after considering subject effects, as shown in Table 3. In the classical ANOVA, different conditions do not make significant differences (p =0.1503), while the repeated measures ANOVA obtain significant differences (p < 0.0001). The variance proportion explained in ANOVA is used as an 'effect size' which represents the degree how the analysis is effective in explaining the variance of outcome variable. In result of the repeated measures ANOVA, 'the partial eta squared' is calculated by dividing the variance explained with the condition (SS by condition) by the total variance (SS by condition +SS residual), excluding variance by subjects. In Table 3, the partial eta squared is calculated as 0.69.

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Table 1. Hypothetical data with repeated observations

		Cond	Difference				
Subject	P1	P2	Р3	Subject mean	P2-P1	P3-P1	P3-P2
1	24	29	36	29.7	5	12	7
2	31	35	39	35.0	4	8	4
3	20	15	20	18.3	-5	0	5
4	17	21	24	20.7	4	7	3
5	22	22	25	23.0	0	3	3
6	25	30	35	30.0	5	10	5
7	27	27	30	28.0	0	3	3
8	19	20	24	21.0	1	5	4
Mean	23.1	24.9	29.1				
Grand Mean				25.7			
Variance					11.9	16.0	1.9

Table 2. Summary table for the classical on-way ANOVA (incorrect)

	Sum of squares	Degree of freedom	Mean squares	F	Effect size (eta squared)
Conditions	152.33 (SSB)	k - 1 = 2	76.16	2.06 (p = 0.1503)	152.33 / 926.96 = 0.16
Residual	774.63 (SSW)	(n-1)(k) = 21	36.89	(MS by condition / MS Residual)	(SS by condition / SST)
Total	926.96 (SST)	(n)(k) - 1 = 24 - 1 = 23			

Table 3. Summary table for the one-way repeated measures ANOVA

	Sum of squares	Degree of freedom	Mean squares	F	Effect size (partial eta-squared)
Subject	704.96	n - 1 = 7	100.78		
Conditions	152.33 (SSB)	K - 1 = 2	76.16	15.31 (p < 0.0001)	152.33 / (152.33 + 69.67) = 0.69
Residual	69.67 (SSE)	(n-1)(k-1) = 14	4.98	(MS by condition / MS Residual)	[SS by condition / (SS by condition + SS residual)]
Total	926.96 (SST)	(n)(k) - 1 = 24 - 1 = 23			

### Multivariate and univariate tests

The multivariate tests analyze multiple variables (P1, P2, P3) as one observation using a 'wide-form' data that one person has one record with P1 to P3. The multivariate approach requires a sufficient number of subjects (e.g., larger than 30) as the number of person is the number of information. Also a record with any missing value is deleted, therefore even small number of missing values may lead to substantial loss of information. When sample size is large, Pillai's trace and Wilk's lambda have similar power and show similar results (Please refer [g] of SPSS analysis results). The univariate tests are analyzing 'long form' data that one person has multiple records by occasions, e.g., three records per person in this example. The univariate tests have more power based on the increased number of information and have relative advantage in treating missing values that only the missing occasions are deleted. In case with many missing values, only the univariate approach works. However additional assumption, 'sphericity' is enquired to obtain reasonable analysis results by the univariate tests.

### Sphericity assumption of univariate tests

The univariate tests assume an equal correlation between the repeated measures which is called 'sphericity' condition. Sphericity refers to the condition where the variances of the differences between all possible pairs of groups (i.e., levels of the independent variable, 'conditions' here) are equal. Table 3 is the result based on the sphericity assumption. However, variances in differences between pairs of conditions, P2-P1, P3-P1, and P3-P2 are 11.9, 16.0, and 1.9, respectively, which seem fairly different (shown in the right side of Table 1). The Mauchly's test which evaluates the sphericity condition showed that the assumption of sphericity was unsatisfied (p = 0.025, [h] of SPSS results), thus additional remedy steps should be taken. The remedy is to adjust degrees of freedom of numerator and denominator by multiplying the adjustment factor, ε (epsilon). Two kinds of epsilon, Greenhouse-Geisser and Huynh-Feldt epsilon ([h] of SPSS results) are applied to the test of within-subject effect ([i] of SPSS results). If the epsilon is close to one, which means the sphericity assumption is met, then there is no need of adjustment. The quideline of epsilon is determined by the Greenhouse-Geisser epsilon, 0.75; if larger than 0.75 then apply the Huynh-Feldt epsilon, otherwise, apply the Greenhouse-Geisser epsilon.<sup>2</sup> In this example as the Greenhouse-Geisser epsilon was 0.586, we interpret the corrected results in (i) of the SPSS results by applying epsilon to degrees of freedom of 'repeat 1' and error portion (p = 0.003) with the heading Greenhouse-Geisser. The pairwise comparison of main effects of P1, P2, and P3 shows that significant differences are found in P3 and other two occasions by applying the Bonferroni correction which controls the type one error rate ([k] of SPSS results). We could see that mean of P3 is significantly different from those of P1 (p = 0.011) and P2 (p < 0.001), while means of P1 and P2 are similar (p = 0.585) as shown in table (k) of SPSS output.

# Other outputs

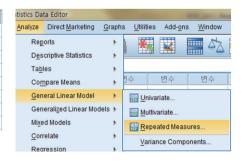
Assessment polynomial terms show that assuming linear relationship among repeated measurements is appropriate ([j] of SPSS results). Also the Profile plot ([l] of SPSS results) provides the trend of the repeated measurements. Tests of between-subjects effects traditionally provided by the repeated measures ANOVA SPSS output (not shown) represent the comparison of simple averaged values among groups without accounting for correlated structures; therefore the importance of the result is little in respect to the analysis of correlated data.

The procedure of the one-way repeated measures ANOVA using SPSS statistical package (SPSS Inc., Chicago, IL, USA):

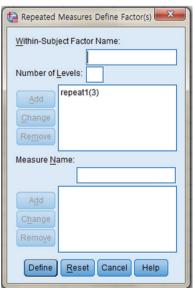
(a) Data

I	Subject	P1	P2	P3
ĺ	1	24	29	36
	2	31	35	39
İ	3	20	15	20
I	4	17	21	24
	5	22	22	25
	6	25	30	35
I	7	27	27	30
İ	8	19	20	24

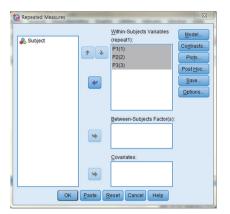
(b) Analyze-General Linear Model - Repeated measures



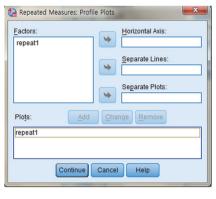
(c) Factor name (repeated measure) and number of levels (3) - Define



# (d) Define repeated variables



# (e) Request plots



# (f) Compare main effect -Bonferroni method



# (g) Result of multivariate tests

### Multivariate Tests<sup>a</sup>

Effect		Value	F	Hypothesis df	Error df	Sig.
repeat1	Pillai's Trace	.915	32.224 <sup>b</sup>	2.000	6.000	.001
	Wilks' Lambda	.085	32.224 <sup>b</sup>	2.000	6.000	.001
	Hotelling's Trace	10.741	32.224 <sup>b</sup>	2.000	6.000	.001
	Roy's Largest Root	10.741	32.224 <sup>b</sup>	2.000	6.000	.001

a. Design: Intercept Within Subjects Design: repeat1

# (h) Test of sphericity

### Mauchly's Test of Sphericity<sup>a</sup>

Measure: MEASURE\_1

					Epsilon <sup>b</sup>		
Within Subjects Effect	Mauchly's W	Approx. Chi- Square	df	Sig.	Greenhouse- Geisser	Huynh-Feldt	Lower-bound
repeat1	.294	7.340	2	.025	.586	.633	.500

Tests the null hypothesis that the error covariance matrix of the orthonormalized transformed dependent variables is proportional to an identity matrix.

# (i) Tests of within-subjects effects

# Tests of Within-Subjects Effects

Measure: MEASURE\_1

Source		Type III Sum of Squares	df	Mean Square	F	Sig.
repeat1	Sphericity Assumed	152.333	2	76.167	15.306	.000
	Greenhouse-Geisser	152.333	1.172	129.923	15.306	.003
	Huynh-Feldt	152.333	1.266	120.290	15.306	.003
	Lower-bound	152.333	1.000	152.333	15.306	.006
Error(repeat1)	Sphericity Assumed	69.667	14	4.976		
	Greenhouse-Geisser	69.667	8.207	8.488		
	Huynh-Feldt	69.667	8.865	7.859		
	Lower-bound	69.667	7.000	9.952		

b. Exact statistic

# (j) Assessment of polynomial terms

## **Tests of Within-Subjects Contrasts**

Measure: MEASURE\_1

Source	repeat1	Type III Sum of Squares	df	Mean Square	F	Sig.
repeat1	Linear	144.000	1	144.000	18.000	.004
	Quadratic	8.333	1	8.333	4.268	.078
Error(repeat1)	Linear	56.000	7	8.000		
	Quadratic	13.667	7	1.952		

# (k) Pairwise comparison

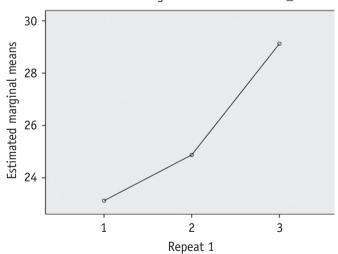
### **Tests of Within-Subjects Effects**

Measure: MEASURE\_1

Source		Type III Sum of Squares	df	Mean Square	F	Sig.
repeat1	Sphericity Assumed	152.333	2	76.167	15.306	.000
	Greenhouse-Geisser	152.333	1.172	129.923	15.306	.003
	Huynh-Feldt	152.333	1.266	120.290	15.306	.003
	Lower-bound	152.333	1.000	152.333	15.306	.006
Error(repeat1)	Sphericity Assumed	69.667	14	4.976		
	Greenhouse-Geisser	69.667	8.207	8.488		
	Huynh-Feldt	69.667	8.865	7.859		
	Lower-bound	69.667	7.000	9.952		

# (l) Profile plots

Estimated marginal means of MEASURE\_1



# References

- 1. Park E, Cho M, Ki CS. Correct use of repeated measures analysis of variance. Korean J Lab Med 2009;29:1-9.
- 2. Field A. Repeated measures designs: discovering statistics using SPSS. 3<sup>rd</sup> ed. London: SAGE Publications Ltd; 2009. Chapter 13.