

36-309/749
Experimental Design for Behavioral
and Social Sciences

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Lecture 8: Contrasts and Multiple
Comparisons

Contrasts (in general)

- Context: An ANOVA rejects the “overall” null hypothesis that all k means of some factor are not equal, i.e., $H_0: \mu_1 = \dots = \mu_k$. When $k > 2$, this is not satisfying (scientifically).
- Contrasts let us ask “which are significantly different?”
- Terminology: Define a “**contrast**” or “analytic comparison” that is of scientific interest, e.g. compare μ_1 to μ_5 or compare μ_2 to the average of μ_1 , μ_3 and μ_5 , i.e., $\left(\frac{\mu_1 + \mu_3 + \mu_5}{3}\right)$
- **Contrast null hypothesis**: Express as something equal to zero. For the above examples, $\mu_1 = \mu_5 \rightarrow \mu_1 - \mu_5 = 0$ and $\mu_2 = \left(\frac{\mu_1 + \mu_3 + \mu_5}{3}\right) \rightarrow \mu_2 - \left(\frac{\mu_1 + \mu_3 + \mu_5}{3}\right) = 0$.

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Contrasts, cont.

- Computer package form: Re-express in **linear combination** form: $C_1\mu_1 + \dots + C_k\mu_k = 0$ which contains all of the parameters in the original “overall” null hypothesis **in order**. (The C values are called “weights” or “coefficients”. Some of the C ’s will be negative.) Enter just the coefficients into the computer.
 - Example 1: express $\mu_1 - \mu_5 = 0$ as $(1)\mu_1 + (0)\mu_2 + (0)\mu_3 + (0)\mu_4 + (-1)\mu_5 = 0$. Enter 1 0 0 0 -1 into SPSS or some other computer package.
 - Example 2: express $\mu_2 - \left(\frac{\mu_1 + \mu_3 + \mu_5}{3}\right) = 0$ as $(-1/3)\mu_1 + (1)\mu_2 + (-1/3)\mu_3 + (0)\mu_4 + (-1/3)\mu_5 = 0$. Enter into SPSS:
- Check your work: **Valid coefficient sets always add to zero.**

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Planned comparisons (contrasts)

- Planned comparisons **maintain type-1 error** at α (experiment-wise) only when:
 - They are chosen **in advance**, i.e., truly planned.
 - They are only used if the corresponding **overall** p -value is $\leq \alpha$.
 - They number no more than the F numerator df.
 - One-way (k level) ANOVA: $k-1$ planned contrasts
 - Two-way ($k \times m$) ANOVA:
 - Interaction expected: plan $(k-1)(m-1)$ contrasts like $\mu_{A_1B_1} - \mu_{A_2B_1} = 0$ or the often more informative form, such as $(\mu_{A_1B_1} - \mu_{A_2B_1}) - (\mu_{A_1B_2} - \mu_{A_2B_2}) = 0$.
 - Interaction not expected: $k-1$ planned contrasts for factor A; $m-1$ for factor B.
 - They are orthogonal (often ignored): the sum of products of corresponding coefficients equal zero, i.e., they ask **independent** questions.

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Planned comparisons, cont.

- **Optional** technical details: See gray boxes in textbook and/or ask Howard.
- In SPSS, comparisons made using the “Contrasts” button or the LMATRIX subcommand under GeneralLinearModel/Univariate are assumed to be planned, and the p-values are wrong otherwise.

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Multiple (post hoc, unplanned) comparisons

- Example 0: Darts game
- Example 1: In a study of twenty chocolate lovers vs. non-chocolate eaters (freaks of nature), researchers claimed that “higher levels of phenylacetylglutamine and citrate in the chocolate-desiring group suggest that these individuals may regulate the citric acid cycle slightly differently than those who don’t fancy a daily dose of chocolate.” (J Proteome Research, 6(11):4469-4477, 2007) Consider performing t-tests to see if the groups differ for each detectable compound. For 50 compounds (a low, but reasonable number), if they are all unrelated to chocolate, the chance of **avoiding** a false positive at a rate of 0.95 each is $0.95^{50}=0.077$. [Chance of getting 1 or 2 FP is 20% and 26% respectively.] Conclusion:

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Multiple Comparisons, cont.

- Example 2: In a study of the effect of magic beans on health, a carefully done, well powered, randomized clinical trial measures 12 health outcomes (BP, cholesterol, etc.). Assuming “magic beans” are useless, the chance of *avoiding* a false positive is:
 $0.95 * 0.95 * \dots * 0.95 = 0.95^{12} = 0.54$

So the chance of finding at least one (meaningless) finding is $1 - 0.54 = 46\%$.

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Multiple Comparisons, cont.

- Math: The number of ways you can choose 2 items from a list of k items is called “k choose 2”. We use the symbol $\binom{k}{2}$ and the answer is $k(k-1)/2$.
- Checking all $\binom{k}{2}$ pairs in a one-way ANOVA has exactly the same problem as for “magic beans”.
- Less obviously, when we pick out the smallest and largest sample means out of k means to compare, we are implicitly performing multiple comparisons, thus increasing the chance of making a type-1 error.
- The problem is also referred to as **post-hoc** testing, unplanned comparisons, and data snooping.

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Multiple Comparisons, cont.

- Most common goal: keep the **per-experiment type-1 error rate** at 0.05 (compare with FDR). The key to appropriate, honest post-hoc comparisons is to determine the size of the family of comparisons that you are considering, and handicap yourself (e.g., lower alpha or raise the p-value) to reduce the chances of a type 1 (FP) error, which, unfortunately, is at the expense of reduced power.
- Special example: In a two-way ANOVA with interaction, at least one of the three overall null hypotheses (two main effects plus interaction) are rejected at $p \leq 0.05$ about 14% of the time for null experiments if the “corrected model” p-value is not used as a “screen”.

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Multiple Comparisons, cont.

- Appropriate methods add a “penalty” for multiple comparisons
 - **Bonferroni procedure**: simplest and most general, but conservative (Holm’s-Bonferroni is a tiny bit better). Set $\alpha' = \alpha/m$ where m is the number of possible comparisons in the “family”, then compare the p-value to α' instead of α .
 - Based on the “degree of fishing”, choose one of these methods that gives adjusted p-values and/or adjusted CIs for some **specific situation** (generally more power than Bonferroni):
 - Tukey’s procedure: test all possible pairs for one factor.
 - Dunnett’s procedure: compare a control to all possible active treatments
 - Scheffé’s procedure: all possible simple and complex contrasts

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Contrasts in SPSS

- “Contrasts” button gives p-values assuming that the comparisons are an appropriate set of **planned** comparisons.
- “Post-hoc” button gives p-values assuming post-hoc comparisons within the “family” associated with the specific post-hoc procedure. In multi-way ANOVA, a no-interaction model is assumed. Tukey (all paires) and Dunnett (baseline vs. all others) are the most useful.

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Contrasts in SPSS, cont.

Multiple Comparisons

Tukey HSD	(I) Color	(J) Color	Mean Difference (I-J)	Std. Error	Sig.	95% Confidence Interval	
						Lower Bound	Upper Bound
white	red	green	-1.8000	.29804	.000	-2.6248	-.9752
		blue	-2.4571	.29804	.000	-3.2819	-1.6324
		white	.0952	.31021	.990	-.7632	-.9537
Red	white	green	1.8000	.29804	.000	.9752	2.6248
		blue	-.6571	.29804	.152	-1.4819	.1676
		red	1.8952	.31021	.000	1.0368	2.7537
green	white	blue	2.4571	.29804	.000	1.6324	3.2819
		red	.6571	.29804	.152	-.1676	1.4819
		blue	2.5524	.31021	.000	1.6939	3.4108
Blue	white	red	-.0952	.31021	.990	-.9537	.7632
		green	1.8952	.31021	.000	2.7537	1.0368
		red	-2.5524	.31021	.000	-3.4108	-1.6939

Homogeneous Subsets

Tukey HSD	Color	N	Subset	
			1	2
Tukey HSD	blue	6	2.3333	
	white	7	2.4286	
	red	7		4.2286
	green	7		4.8857
	Sig.			.989

Not necessarily non-overlapping!

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Contrasts in SPSS, cont.

➤ SPSS LMATRIX subcommand

- Requires syntax pasting and manual entry of contrast coefficients
- Very flexible: **any** valid contrast can be specified
- p-values are based on the comparisons being **planned**
- For post-hoc, calculate $t = \text{contr.} / \text{SE}(\text{contr.})$, $F = t^2$, and use, e.g., Scheffé procedure

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“Paste Syntax” in SPSS

- “Paste” instead of the final “OK” in SPSS causes the “syntax” to be displayed in the “Syntax Editor” instead of running the analysis.
- You can edit the syntax (following strict rules) and then “run” the syntax to run the analysis.
- Expert SPSS users often work mainly with “syntax”.
- For us, syntax is used to add features to an analysis for which there are no menu items.

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Contrasts in SPSS, cont.

➤ 1-way ANOVA example: factor name is “treatment”, level order is Placebo, Talk, Drug, Both. Paste:

```
UNIANOVA score BY treatment
/METHOD=SSTYPE(3)
/INTERCEPT=INCLUDE
/CRITERIA=ALPHA(0.05)
/DESIGN=treatment.
```

➤ Edit to:

```
UNIANOVA score BY treatment
/METHOD=SSTYPE(3)
/INTERCEPT=INCLUDE
/CRITERIA=ALPHA(0.05)
/LMATRIX "others - control" treatment -1 1/3 1/3 1/3
/LMATRIX "combo - (drug+talk)/2" treatment 0 1/2 1/2 -1
/LMATRIX "drug-talk" treatment 0 -1 1 0
/DESIGN=treatment.
```

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LMATRIX for 1-way ANOVA, cont.

```
/LMATRIX "others - control" treatment -1 1/3 1/3 1/3
```

Estimate of $\frac{\mu_D + \mu_T + \mu_B}{3} - \mu_P$

Custom Hypothesis Tests #1

Contrast Results (K Matrix)^a

Contrast		Dependent Variable
		score
L1	Contrast Estimate	16.333
	Hypothesized Value	0
	Difference (Estimate - Hypothesized)	16.333
	Std. Error	6.192
	Sig.	.017
	95% Confidence Interval for Difference	Lower Bound 3.324 Upper Bound 29.343

a. Based on the user-specified contrast coefficients (L) matrix others - placebo

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LMATRIX for 1-way ANOVA, cont.

/LMATRIX "combo - (drug+talk)/2" treatment 0 1/2 1/2 -1

$$\text{Estimate of } \mu_B - \frac{\mu_D + \mu_T}{2}$$

Custom Hypothesis Tests #2

Contrast Results (K Matrix) ^a		
Contrast		Dependent Variable score
L1	Contrast Estimate	20.400
	Hypothesized Value	0
	Difference (Estimate - Hypothesized)	20.400
	Std. Error	6.979
	Sig.	.009
	95% Confidence Interval for Difference	Lower Bound 5.738 Upper Bound 35.062

a. Based on the user-specified contrast coefficients (L) matrix: both - (drug+talk)/2

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LMATRIX for 1-way ANOVA, cont.

/LMATRIX "drug-talk" treatment 0 -1 1 0

$$\text{Estimate of } \mu_D - \mu_T$$

Custom Hypothesis Tests #3

Contrast Results (K Matrix) ^a		
Contrast		Dependent Variable score
L1	Contrast Estimate	16.400
	Hypothesized Value	0
	Difference (Estimate - Hypothesized)	16.400
	Std. Error	7.825
	Sig.	.050
	95% Confidence Interval for Difference	Lower Bound -0.039 Upper Bound 32.839

a. Based on the user-specified contrast coefficients (L) matrix: drug-talk

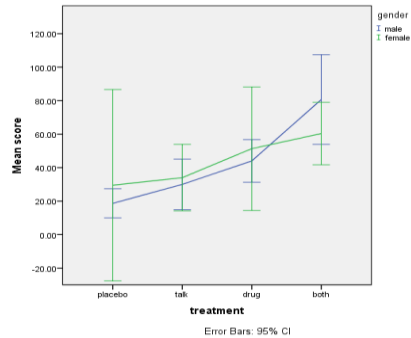
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Contrasts for 1-way ANOVA without interaction

- Additive model = parallel pattern in a graph of population means
- Valid questions: What are the effects of a specific change in level of one factor ignoring, fixing or averaging over the other factor?
- Conclusion: Analyze each factor separately as for 1-way ANOVA.

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Contrasts for 2-way with Interaction



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Interaction Contrasts, cont.

```
UNIANOVA score BY treatment gender
/METHOD=SSTYPE(3)
/INTERCEPT=INCLUDE
/EMMEANS=TABLES(treatment*gender)
/CRITERIA=ALPHA(.05)
/LMATRIX "M-F for placebo" gender 1 -1
      treatment*gender 1 -1  0 0  0 0  0 0
/LMATRIX "M-F for both" gender 1 -1
      treatment*gender 0 0  0 0  0 0  1 -1
/LMATRIX "M-F for (talk+drug)/2" gender 1 -1
      treatment*gender 0 0  1/2 -1/2  1/2 -1/2  0 0
/DESIGN=treatment gender treatment*gender.
```

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Interaction Contrasts, cont.

```
/LMATRIX "M-F for both" gender 1 -1
      treatment*gender 0 0  0 0  0 0  1 -1
```

$H_0: \mu_{BM} = \mu_{BF}$ Estimate: $\mu_{BM} - \mu_{BF}$

Contrast Results (K Matrix) ^a		Dependent Variable	
Contrast		score	
L1	Contrast Estimate	20.333	
	Hypothesized Value	0	
	Difference (Estimate - Hypothesized)	20.333	
	Std. Error	7.088	
	Sig.	.012	
	95% Confidence Interval for Difference	Lower Bound Upper Bound	5.130 35.537

a. Based on the user-specified contrast coefficients (L) matrix. M-F for both

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Interaction Contrasts, cont.

- Posthoc example: all pairs for the 2x4 ANOVA
- 8 groups = $\binom{8}{2} = 8*7/2 = 28$ pairs
- Write up to 28 LMATRIX commands
- Compute Bonferroni $\alpha' = 0.05/28 = 0.0018$
- Test $(2-1)*(4-1)=3$ planned contrasts using $\alpha=0.05$
- Reject any others if $p < 0.0018$

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Summary

- Contrasts allow more useful scientific conclusions when a rejected H_0 is vague, e.g., $H_0: \mu_1 = \dots = \mu_k$ (with $k > 2$) or H_0 : additive model is good enough.
- (Remember: main effect H_0 s in the presence of a significant interaction answer the wrong questions!)
- Running multiple tests increases the chance for false rejection. Beyond "df" tests, corrections must be used to "maintain the type-1 error rate at α ".
- Multiple comparisons corrections reduce power. Pre-planned contrasts should be selected before running the experiment to maximize power to where it is most needed.

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