Lecture 12: Chapter 12

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UAB Mathematics

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§12.1 ANOVA (Analysis of Variance) Tests

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- The different samples are from populations that are categorized in only one way.

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 H_1 : At least one of the population means differs from the others. You can find "by hand" computations for a one-way ANOVA test, see page 605. The test is a right-tailed F test. The weights (in kg) of oak trees is given below for trees planted in the same plot but with different growth-enhancing methods applied. Use a 0.05 significance level to test the claim that the four treatment categories yield oak trees with the same mean weight. Does there appear to be a best method for enhancing growth of trees in this soil?

Control	Fert	Irr	Fert & Irr
0.24	0.92	0.96	1.07
1.69	0.07	1.43	1.63
1.23	0.56	1.26	1.39
0.99	1.74	1.57	0.49
1.80	1.13	0.72	0.95

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We use StatCrunch to conduct the test.

$\S12.2$ Example

What's our null and alternative hypotheses? Based on the output, do we reject or accept the null hypothesis? What does this mean? (Use $\alpha = 0.05$.)

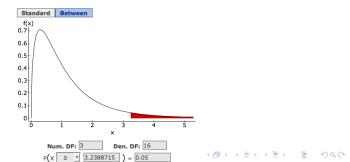
Options	1						8)
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ata store	d in	separate	columns.				
Column s	tati	stics					
Column a	e n	+ Mean +	Std. Dev.	¢	Std. Error	• •	
Control		5 1.19	0.6257395	6	0.2798392	24	
Fert		5 0.884	0.62492	4	0.279474	51	
Irr		5 1.188	0.3466554	5	0.1550290	03	
F & I		5 1.106	0.4359816	55	0.1949769	92	
NOVA ta Source	DF	SS	MS		F-Stat	P-val	
Source	DF	33	ms		r-Stat	P-vai	ue
Columns	3	0.3114	0.1038	0.	38010491	0.76	87
Error	16	4.36932	0.2730825				
Total	4.0	4.68072					

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You can also perform the previous test using a critical value test. The F critical value is computed using the significance, the F calculator, and the "degrees of freedom of the numerator" and the "degrees of freedom of the denominator" which are, respectively, one fewer than the number of columns and the difference between the total number of data points and the number of columns.

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- The samples must be independent not matched or paired.
- The different samples are from populations that are categorized in two ways.
- All cells have the same number of values.

The null hypothesis is always that there is **no effect due to interaction between the two factors**. The alternative is that there is an effect from the interaction between the two factors. The test statistic is always

$$F = \frac{MS(\text{interaction})}{MS(\text{error})}$$

Note: If you conclude that the there is an effect due to interaction, then you would not investigate the rows and columns separately.

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Below are pulse rates of men and women over and under the age of 30. Conduct a two-way ANOVA and state the results. Use a 0.05 significance level.

	< 30	≥ 30		
	78 104 78 64 60 98 82 98 90 96			
М	60 80 56 68 68 74 74 68 62 56	46 70 62 66 90 80 60 58 64 60		

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We get the following result from StatCrunch.

Analysis of Variance results:

Responses: Results Row factor: Row Factor Column factor: Column Factor

ANOVA table

Source	DF	SS	MS	F-Stat	P-value
Row Factor	1	1322.5	1322.5	10.862919	0.0022
Column Factor	1	592.9	592.9	4.8700374	0.0338
Interaction	1	448.9	448.9	3.6872319	0.0628
Error	36	4382.8	121.74444		
Total	39	6747.1			

So there does not appear that there is an effect due to the interaction of age and gender, but there does seem to be an effect due to age and a separate effect due to gender.

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We would like to perform a two-way ANOVA (with $\alpha = 0.05$ throughout) to test whether there is an effect due to interaction and due to the row and column factors separately.

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	< 30	30 to 50	> 50
G	575 600 600 620	600 620 620 645	620 645 650 650
NG	620 645 650 650	600 620 620 645	575 600 600 620

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NG	620 645 650 650	600 620 620 645	575 600 600 620

Well, there is certainly an effect due to interaction! So we don't perform the hypothesis tests for effects due to the row and column factors separately.

Say you conduct a one-way ANOVA on four sample means $\bar{x}_1 = 1.5$, $\bar{x}_2 = 1.6$, $\bar{x}_3 = 1.4$, and $\bar{x}_4 = 1.7$ and got that the *P*-value was 0.12. What would you conclude?

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