Two-Way ANOVA Pt2

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Interpretation of Findings

- we're going to talk about what the findings mean for your factorial design
- we will have three things to check
 - main effect (relationship) between your first IV/factor and your
 DV (while ignoring everything else)
 - this will also be in the F test for this one IV
 - main effect 2 (relationship 2) between your second IV/factor and your DV (while ignoring everything else)
 - this will also be in the F test for this one IV
 - interaction (combination of IV1 and IV2) between your IVs on your DV (while ignoring the main effects)
 - this will also be in the F test for the interaction

Main Effect 1 Interpretation

- the F value for this main effect will tell us if there is a significant relationship between your first IV and your outcome
 - essentially is there a difference between the levels/conditions/groups in your first IV in their average outcome values
 - Ex: Is there a significant difference in average depression levels of your 3 BMI categories/groups?
- ▶ It is important to note that we are only looking at the relationship between BMI category and depression WHILE ignoring/controlling/adjusting for your second IV
 - Ex: There is a significant relationship between BMI category and depression while adjusting for race/ethnicity
 - Ex: There are differences in depression levels between BMI categories while adjusting for race/ethnicity

Main Effect 2 Interpretation

- same thing as the first main effect but now with your second IV
- ► F value tells you if there are differences in the levels/conditions of your second IV in its relationship with the DV
 - ► Ex: Are there differences in average depression levels between your participants based on their identified racial/ethnic groups?
- Similar here, we are ignoring/adjusting/controlling for IV1 and only talking about the unique relationship between IV2 and the DV

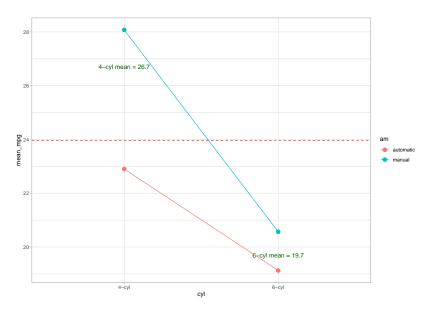
- this is what makes this a two-way ANOVA rather than a one-way ANOVA (one variable) or a ANCOVA (one variable and covariates [variables being controlled])
- we look at the relationship between IV1 and DV based on a specific condition/group from IV2
 - Ex:

vars n mean sd median trimmed mad min max range sl X1 1 18 23.97 4.98 22.15 23.74 4 17.8 33.9 16.1 0

n = 18, k = 4

residual sd = 3.32, R-Squared = 0.63

```
# A tibble: 4 x 4
cyl am mean_mpg sd_mpg
<fct> <fct> <fct> <dbl> <dbl>
1 4-cyl automatic 22.9 1.45
2 4-cyl manual 28.1 4.48
3 6-cyl automatic 19.1 1.63
4 6-cyl manual 20.6 0.751
```



Reporting Findings

- We first talk about the interaction
 - There was a significant interaction found between number of cylinders and transmission type on MPG; F(1, 14) = 1.22, p = XX
 - There was no evidence of a significant interaction between number of cylinders and transmission type on MPG; F(1, 14) = 1.22, p = .29
- ► Then the main effects
 - There was evidence of a significant main effect of number of cylinders on MPG; F(1, 14) = 18.62, p < .001
 - There was no evidence of a significant main effect of transmission type on MPG; F(1, 14) = 4.42, p = .05

Reporting Findings

- Post-hoc tests
 - ▶ A Tukey HSD test was run and found significant differences between some groups. For instance, there was a significant difference between XX and XX (p = XX)
- ► Simple-effects analysis
 - Simple effects analysis revealed that only in the 4-cylinder group, manual transmissions had higher average MPG than automatic transmissions; F(XX, XX) = XX, p = XX
 - if more than two groups, you'd talk to each of these differences