

# **Week 12 Lecture - Complex Experiments**

Undergraduate Research Methods in Psychology

Quinton Quagliano, M.S., C.S.P

Department of Psychology

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## **1** Learning Objectives

#### 1.1 Textbook Objectives

- Explain why researchers combine independent variables in a factorial design.
- Describe an interaction effect in both everyday terms and arithmetic terms.
- Identify and interpret the main effects and interactions from a factorial design.

## 1.2 Professor's Objectives

- Understand and produce examples of when factorial design would be useful and/or appropriate
- Discuss some basic statistics procedures that can be used with these designs

# 2 Chapter Overview

## 2.1 Chapter Overview

•	Up until now, we have only talked about experimental designs that deal with manipulated/independent variable and one measured/dependent
	dent variable.
•	However, we have designs that can look at two (or more) IVs at once and see their individual andimpact on the DV!
•	We refer to these asdesigns.

# 3 Experiments with Two IVS

#### 3.1 Overview

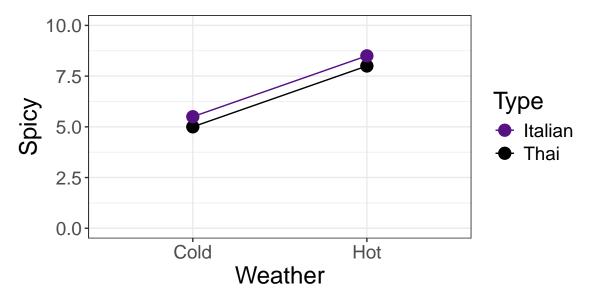
<ul> <li>We can add a second (and third)</li> </ul>	variable if we are curious
about more than one	

• In addition to the individual effects of both of the IVs, we also get an effect that describes how they change each other's relationship with the outcome.

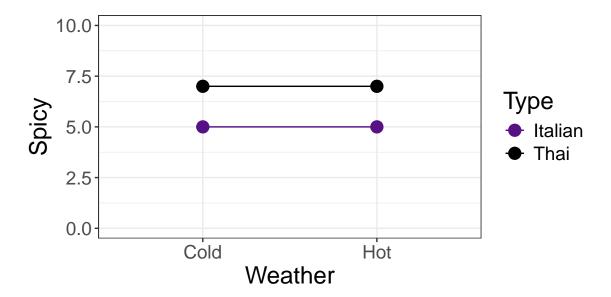
- Statistically, we might say this interaction is a " in differences"
  - Practically, this means that the differences between our groups may be different based on some other trait.
  - More on this later

#### 3.2 Intuitive Interactions

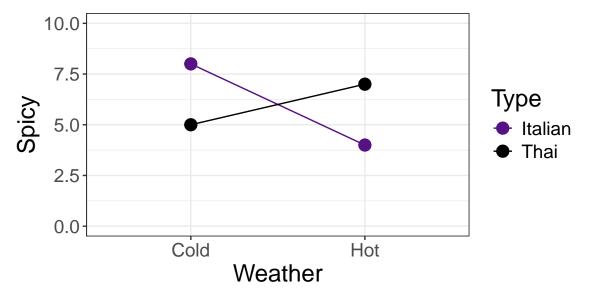
- When confronted with a causal relationship, sometimes we might say, "well it " what it depends on is the second (or third) IV
- We can see this even in our personal experiences, and many relationships do depend on factors
- Example: I am assessing how spicy I like my food (on a scale of 1 to 10; my outcome). First, is it cold or hot outside (IV 1)? Second, am I eating Thai or Italian (IV 2)? It is possible that my answer will be different based upon both of the IVs.
- · 4 Possible Outcomes:
- I like all of my food spicier when it is hot Weather effect, but not food



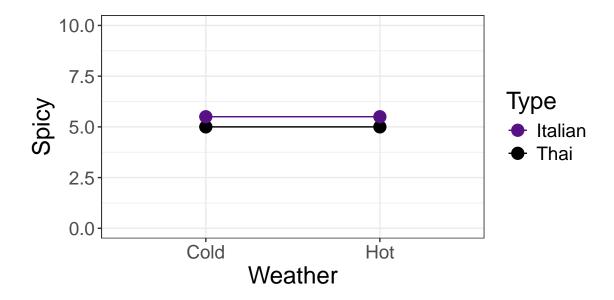
• I like Thai food spicier that Italian, regardless of weather - Food effect, but *not* temperature



- Whether I like by food spicy or not depends on both the weather, and type of food interaction effect
- Specifically, we are looking to see whether we have a \_\_\_\_\_interaction, like in the graph below:



 My preference for spice doesn't change, regardless of food type or weather - null findings



## 3.3 Study Two IVs

When we work with than one IV, we use a factorial design.
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- This creates more unique conditions = # of Conditions in IV 1 x
   # of Conditions in IV 2 = total number of conditions
- Both IVs do *not* have to be \_\_\_\_\_\_. Often, one will be some categorical, measured trait (e.g., gender, ethnicity, etc.)
- In addition to our statistics, we should show these differences in Interaction effects become especially clear with visual evidence.

## 3.4 Limit Testing

- Factorial designs can help us find whether outcomes are different for different of people.
- A strong intervention may not be as effective in a different group of people.
- This can be a \_\_\_\_\_ to our external validity, as we demonstrate findings in a more heterogeneous group.
- We also can establish whether one variable appears to \_\_\_\_\_ another on the relationship with the outcome variable.

#### 3.5 Test Theories

•	For some an effect differs based on some	reasons, we may have goo demographic variable.	od reason to believe that
•	Example: I have a new interve taking in new content. However, just lesser in general. Therefore for younger adults, than it will for	, I recognize that the neurop , I believe my intervention w	lasticity of older adults is
•	In essence, we may be able to hypotheses and investigate with		and "it depends" to ou

#### 3.6 Main Effects & Interactions

<ul> <li>Main Effects are those that come from each IV on the outcome.</li> </ul>				
<ul> <li>The main effect is</li> </ul>	as an average over the levels of the			
other IV. Similar to how we "cont - You have 1 main effect for each	rol" for other variable in multiple regression. IV			
<ul> <li>Marginal Means are the</li> </ul>	that we use to determine whether a			
main effect is present				
<ul> <li>We can test significance by taki</li> </ul>	ng the difference of the two marginal means,			
and calculating 95% CIs. If CIs	$0 \rightarrow non\text{-significant}$			

		IV <sub>1</sub> : Pho		
DV: Reaction time (ms)		Alcohol	Alcohol Plant	
IV <sub>2</sub> :	Aggressive	551	559	555 (average of 551 and 559)
Word type	Neutral	562	552	557 (average of 562 and 552)
Main effe Photo ty	ect for IV <sub>1</sub> : pe	556.5 (average of 551 and 562)	555.5 (average of 559 and 552)	

•	An interaction effect can be detected by looking	g at the differences of the main
	effect differences. If they are	different from one another, then
	we would say that there is an interaction effect	_
	<ul> <li>Interactions are often treated as</li> </ul>	important, theoretically,
	that main effects - when they are significant.	

- Conventional wisdom: If interaction is significant, focus on that mostly. If interaction is non-significant, focus on main effects of IVs.
  - Interpreting the main effects with a significant interaction can be leaving out important information!
- Stats sidebar: This type of analysis is usually done via Two-way ANOVA, which does all the work of calculating significance of interactions, and main effects for us.

# 4 Factorial Variations

#### 4.1 Overview

•	•	, we can lay -groups.	out a factorial design as being between-
•	But, we can designate to a total of 3 possible designs:		variable as between or within, leading

- Independent-Groups FactorialWithin-Groups Factorial
- Mixed Factorial

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4./	Independent-Groups	17621011
	macponacii Cicapo	2001911

•	This is when IVs are between-groups (i.e., participants are arranged into entirely separate groups)
•	One nuance is that this will likely require the largest sample ,
	as each group will have about 1/4th the total number of participants
4.3	Within-Groups Design
•	Much like with previous within-groups designs, this is when participants seepossible condition.
•	One thing to watch out for is the need for to prevent order effects Think about how many permutations of condition orders you may need!
4.4	Mixed Factorial Design
•	This is when one IV isgroups, and the other is within-group.
•	This is fairly common if we have one demographic variable (between-groups) and one manipulated variable that both demographics are exposed to each level (within-groups).
4.5	More Conditions
•	Many are going to naturally have more than one level
	- E.g., race, ethnicity, gender, etc.
•	We can use these in factorial designs all the same - and we write it as: $AxB$ Design.  - Where $A$ = Number of conditions in IV 1
	- Where $B$ = Number of conditions in IV 2 Statistics here get more to interpret - but a good starting point
·	is to use a line plot just like what we have done previously and see if lines cross or
	are parallel.

#### 4.6 More IVs

<ul> <li>Prof. P</li> </ul>	'aul Moes: '	'God himself	cannot	interpret a	a 4-way	' interaction	<ul> <li>neither</li> </ul>	can you'
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•	We can do 3 IVs, but with each additional variable the interpretation becomes expo	)-
	nentially more difficult and complicated.	

-	One popular alternative is to do this as a multiple	1	mode
	instead		

- Stat sidebar: ANOVA and linear regression are both types of the general model, so, in a roundabout way, these are actually equivalent!
- Remember to think carefully about what sorts of conclusions you can draw with a design before you use it, and whether an alternative provides a more conclusion.

# **5** Identify Factorial Designs

#### **5.1** Reading Empirical Articles

•	Look	for	words	like		
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- " ANOVA"
- "Factorial"
- "Interaction" or "Main Effects"
- You may also see phrasing like "2 x 2 design", referring to the two conditions of each IV.

## 5.2 In Popular Media

- Look for words like ...
  - " "
  - "Only when"
- You may also look for demographic variables ...
  - "For males this was the results, but for females..."