

# Lecture 13: Complex (Factorial) Designs

Monday, October 16, 2023

Your Teaching Fellows:

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Lectures: MWF 12:00 PM – 1:00 PM (003); 1:00 PM – 2:00 PM (004); 2:00 PM – 3:00 PM (010)

Office hours: Tuesdays 2:00 PM – 4:00 PM

# Factorial designs – doggy style

		Presence of Food		
		No Treat	Treat	<i>Marginal mean of command</i>
Command	No “Sit”	0.10	0.14	0.12
	“Sit”	0.30	0.90	0.60
	<i>Marginal mean of presence of food</i>			

+0.20

A few more  
dogs sit when  
commanded  
to sit

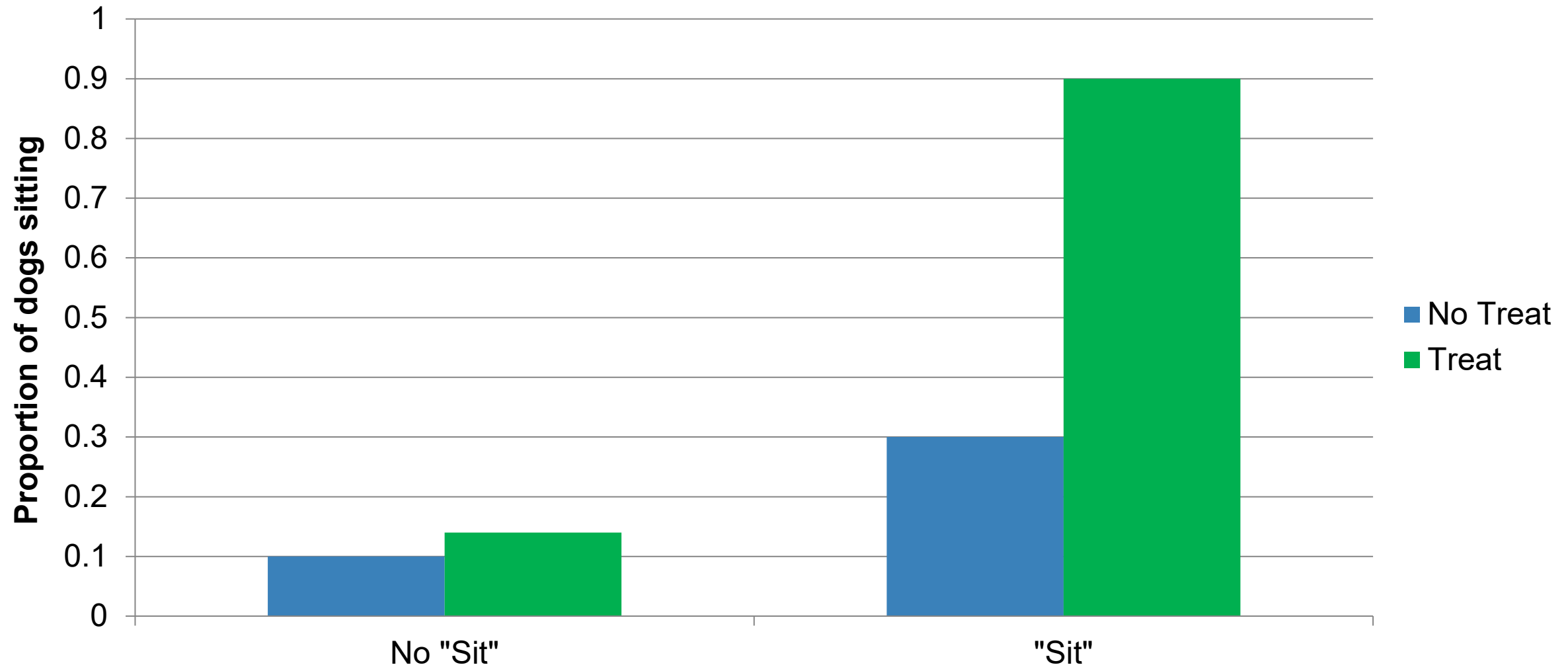
+0.76

Way more  
dogs sit  
when  
commanded  
to sit

+0.48

OVERALL:  
More dogs sit  
when  
commanded  
to sit

## Factorial designs – doggy style

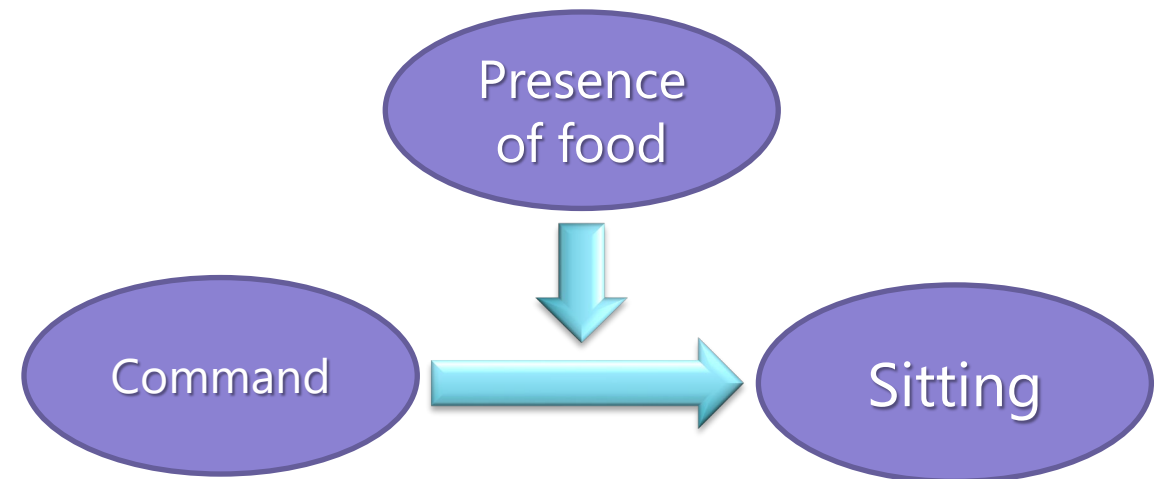
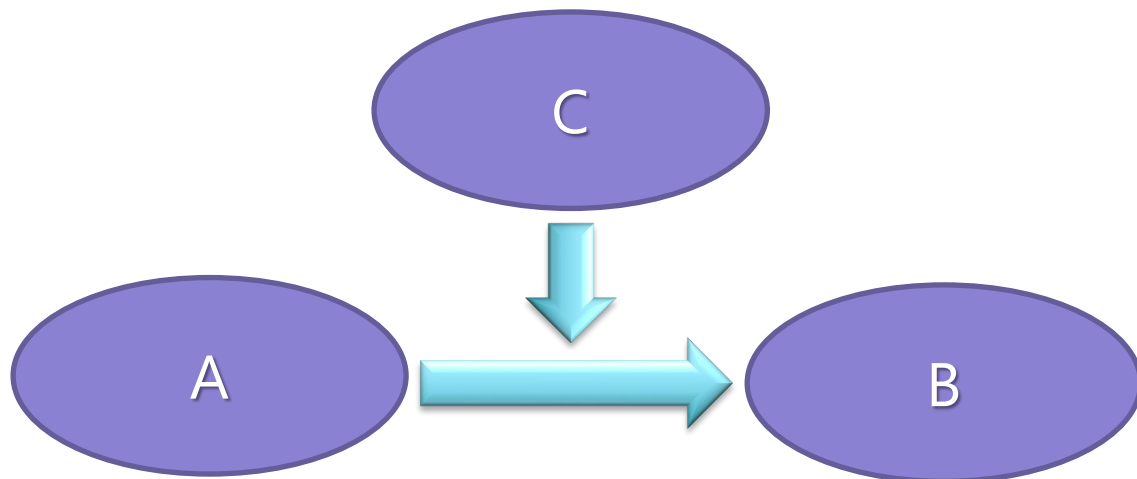


## Factorial designs – doggy style

- Results of doggy study is example of interaction
  - **The effect** of command on proportion of dogs sitting is **different** *depending on* whether food is present or not
  - When effect of the first IV on the DV changes *depending on* the level of the second IV
    - Second IV = “Moderator” or “moderator variable”
- Benefits:
  - Allow us to find interactions
  - More closely mimics real world situations as simple relationships are too unrealistic

## Factorial designs – doggy style

- Understanding interactions



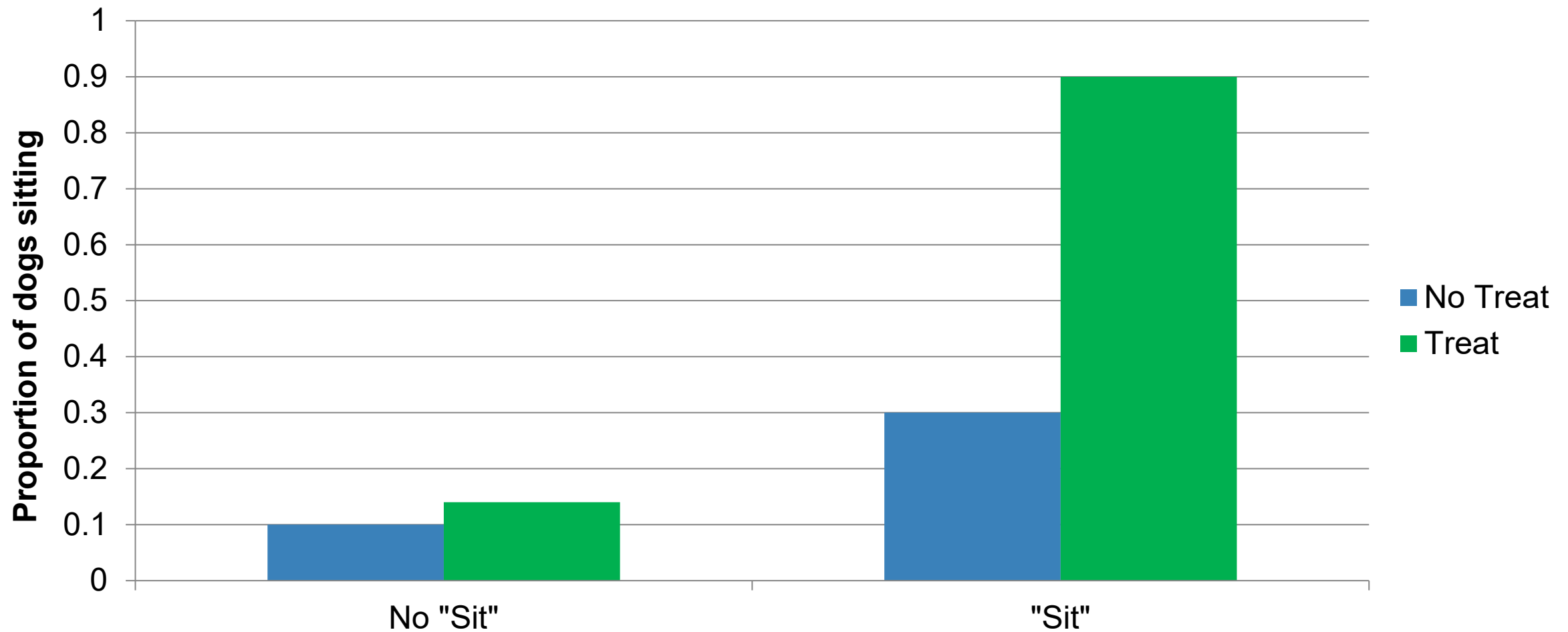
## Factorial designs – doggy style

- In this study, results support conclusion that...
  - Dogs respond best when you command them AND provide them with an incentive!

# Interpreting interaction graphs

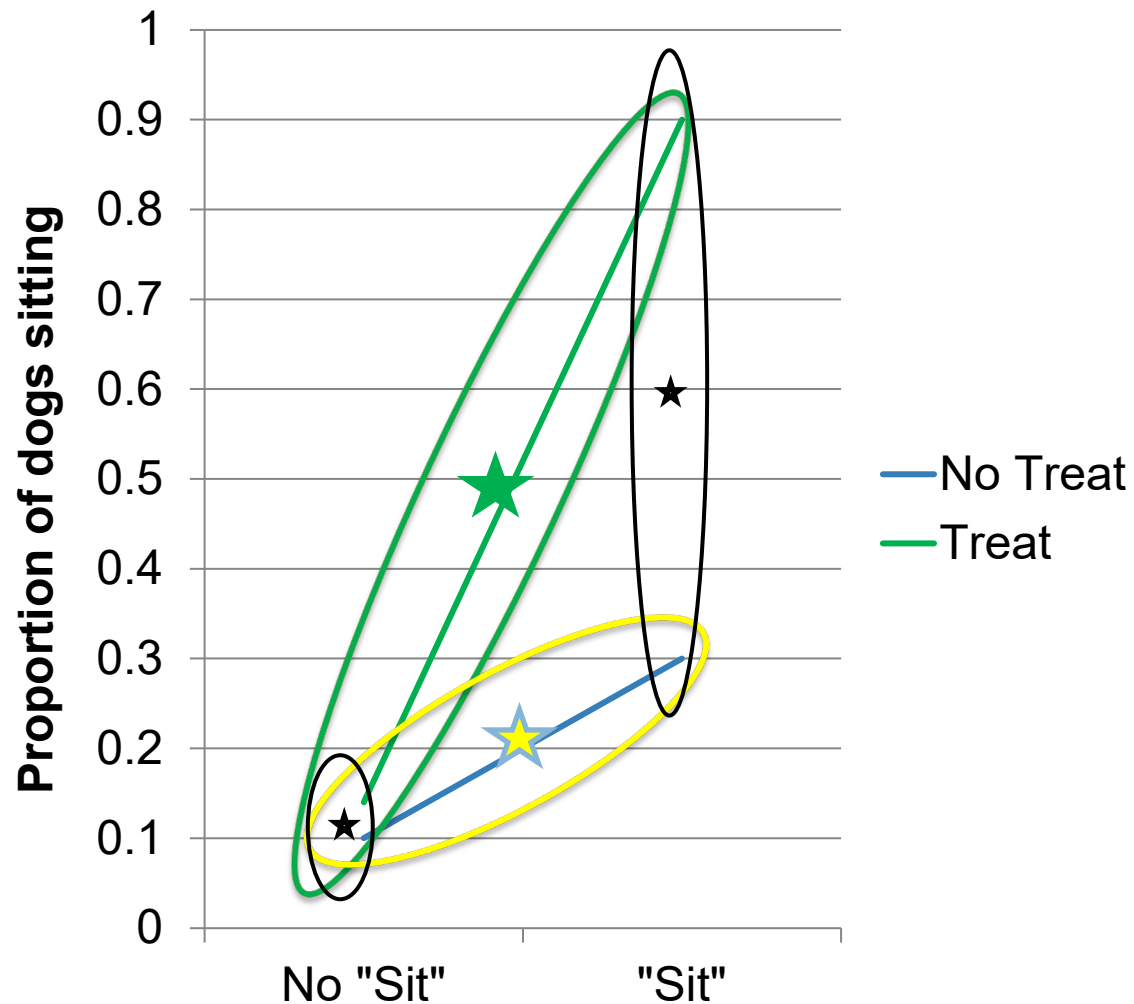
- Why?
  - To understand psychological phenomena more comprehensively, we need to understand interactions
  - Learning to visualise data can be tough at first; but makes interpreting data much easier once you get the hang of it
  - Easier to read journal articles

## Interpreting interaction graphs





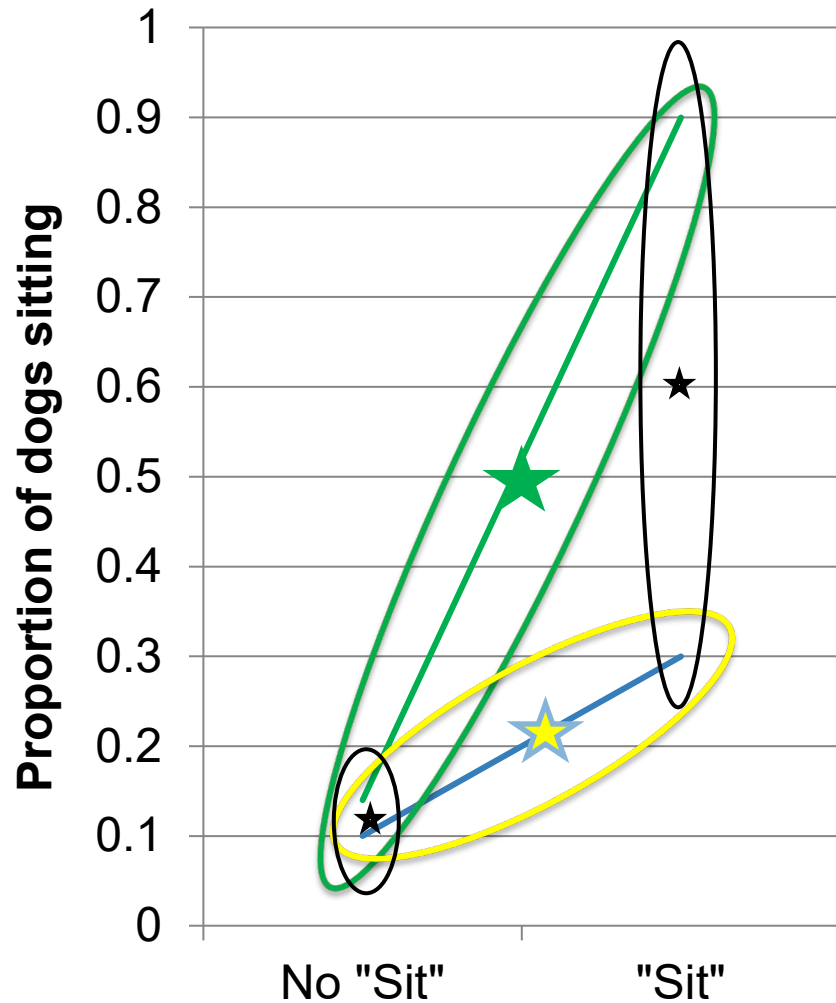
# Interpreting interaction graphs



## Key questions

- Are the lines parallel?
- Are the midpoints of the lines different?
- Are the average values of the DV at each level of the IV different?

# Interpreting interaction graphs

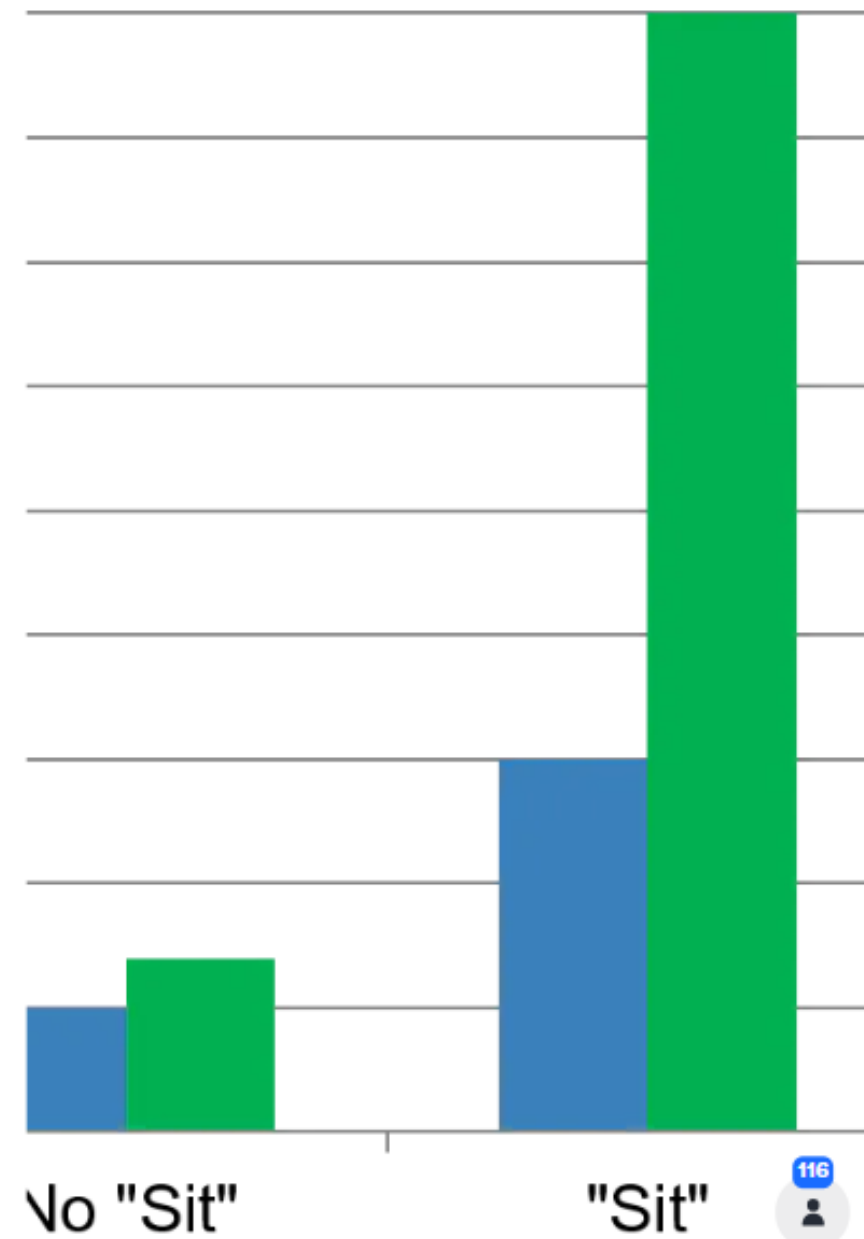
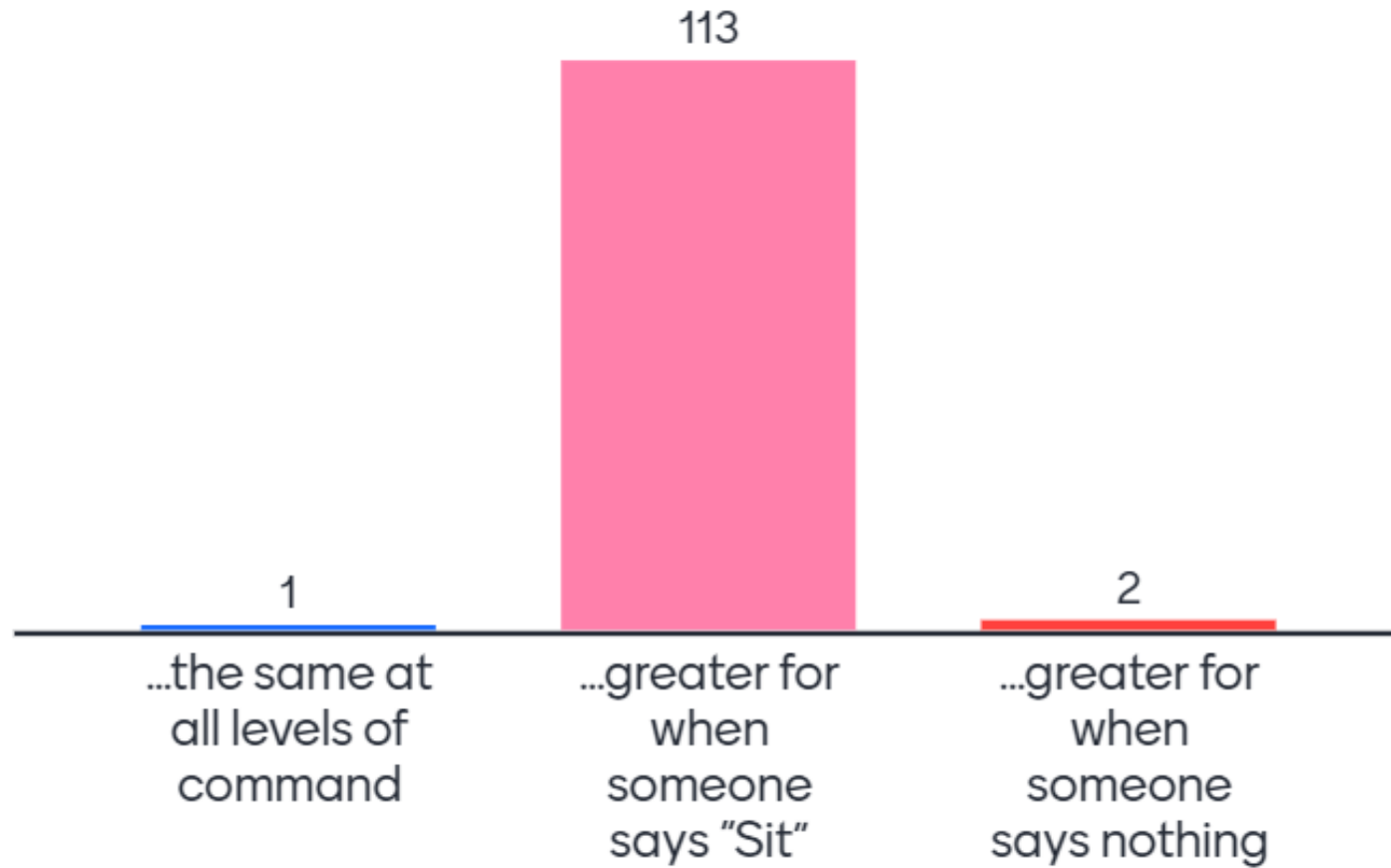


		Presence of Food		
		No Treat	Treat	<i>Marginal mean of command</i>
Command	No "Sit"	0.10	0.14	0.12
	"Sit"	0.30	0.90	0.60
<i>Marginal mean of presence of food</i>		0.20	0.52	

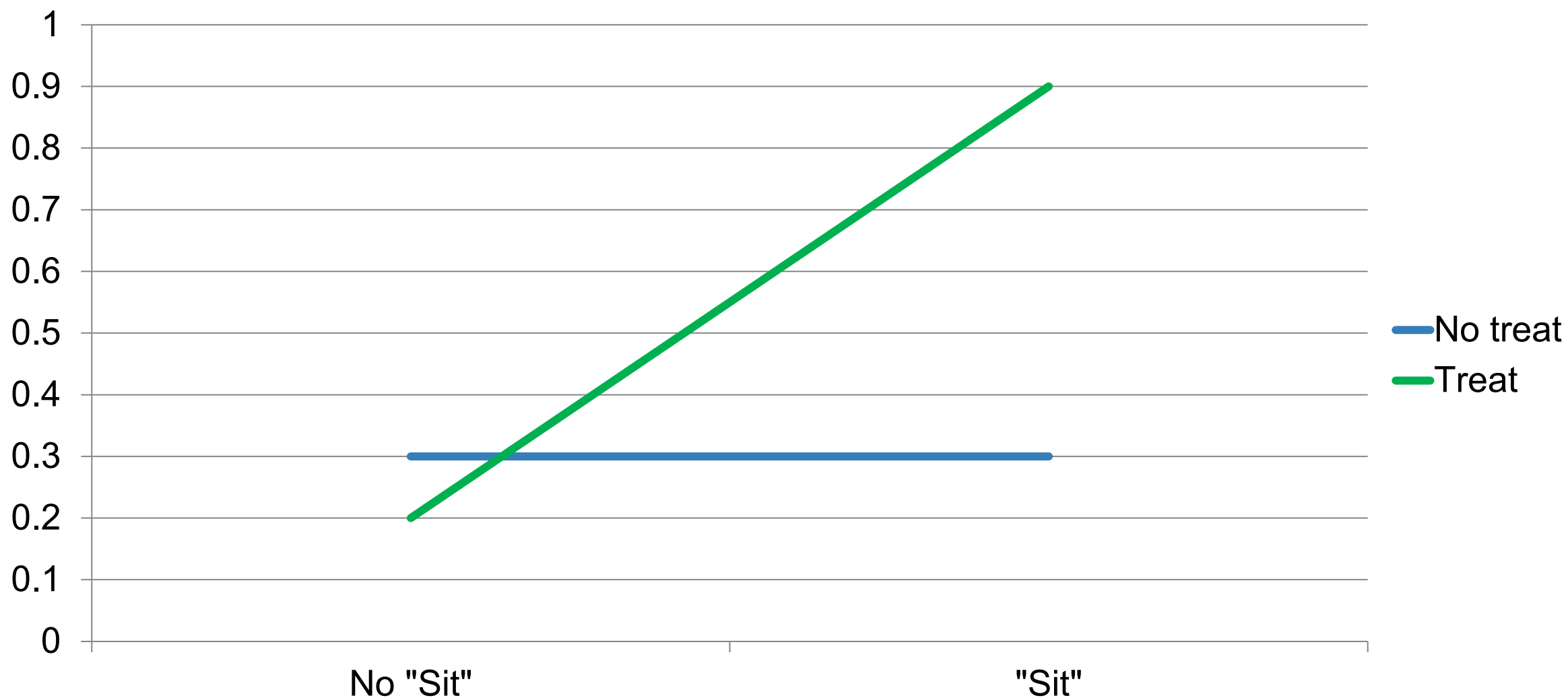
## Interpreting interaction graphs

- Strongly recommend that you look at Figure 11.4 in your text
  - 8 examples of different possible outcomes of 2 x 2 factorial designs
  - Try them, and see how many you get right
  - Come Monday with questions!

# The effect of presence of food is...



## Simple main effects



## Simple main effects

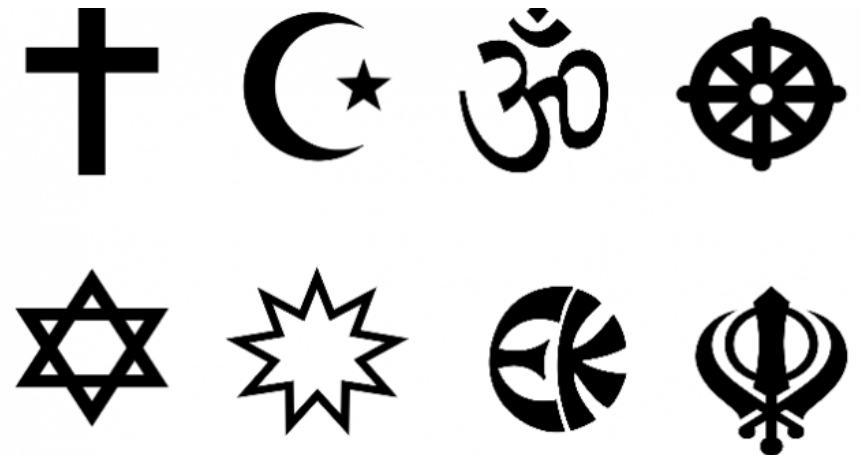
- No – main effects are *qualified* by an interaction between the two independent variables
- Simple main effects = effect of one IV on the DV *within a single level of the second IV*
  - More dogs will sit when you tell them to “sit” rather than not telling them to “sit” – IF you hold food in your hand
  - If you do not hold food in your hand, telling dogs to “sit” versus not telling them to “sit” does not affect the proportion of dogs that sit
- Decide on ONE WAY to split the data

## Learning objectives

- By the end of this class, you should be able to:
  - Explain IV x PV designs
  - Differentiate between crossed factorial designs and nested factorial designs

## IV x PV

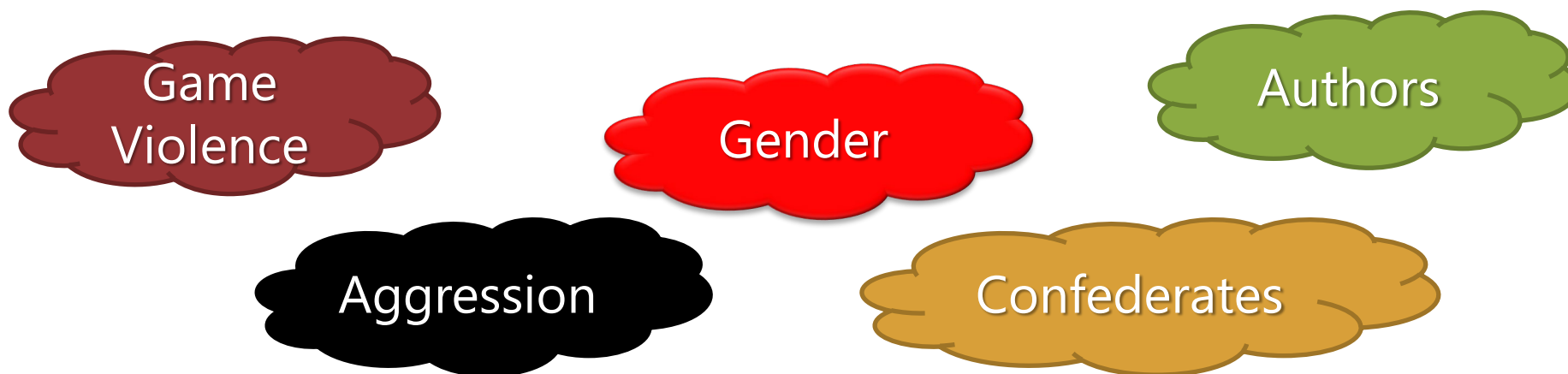
- Give me examples of participant variables:










## IV x PV

- Bartholomew and Anderson (2002):
  - Playing violent video games results in more aggression than playing non-violent video games
- Game violence:
  - Half played violent game, half played non-violent game
- Gender:
  - Half identified as men, half identified as women



# Factorial designs – Aggression

Factor #1: Game violence	Factor #2: Gender	DV: Aggression
		
		

# Factorial designs – Aggression

		Gender	
		Women	Men
Game violence	DBD		
	PGA Tour		

*Known as a 2-way factorial design  
Specifically, a 2x2 factorial design*

# Factorial designs – Aggression

		Gender		
		Women	Men	<i>Sample sizes</i>
Game violence	DBD	n = 10	n = 12	n = 22
	PGA Tour	n = 11	n = 10	n = 21
	<i>Sample sizes</i>	n = 21	n = 22	

# Factorial designs – Aggression

		Gender		
		Women	Men	<i>Marginal mean of violence</i>
Game violence	DBD	5.05	7.01	6.03
	PGA Tour	4.61	4.60	4.60
	<i>Marginal mean of gender</i>	4.83	5.80	

**Main Effect of Gender**

**Main Effect of Game Violence**

# Factorial designs – Aggression

		Gender		
		Women	Men	<i>Marginal mean of violence</i>
Game violence	DBD	5.05	7.01	6.03
	PGA Tour	4.61	4.60	4.60
	<i>Marginal mean of gender</i>	4.83	5.80	

+0.44

*A bit stronger*  
white noise

+2.41

*Way*  
stronger  
white noise

+1.43

OVERALL: Stronger  
white noise after  
violent game