

**STATISTICS CLASS**



welcome back!

PSYC 5143  
"Advanced Descriptive Statistics"  
Spring 2024  
January 17, 2024

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plus ça change, plus c'est la même chose

- the class will proceed more or less the same as last semester
- one substantive change
  - there is a published book from which you can do readings if you'd like (it's the same as the recommended book from last semester)
- drills meet on Thursdays at 940 & 1100, in Memorial Hall 314
  - these start next week

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getting ready for  
polynomial/power predictors  
("nonlinear" regression)

January 17, 2024

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## reminders about interactions/ moderation

- the presence of an interaction in a model is a theoretical claim that the slope of a predictor varies as a function of the value of another predictor
- to test this claim, we include a parameter that allows this slope to vary

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## a non-moderated model

$$Y = b_0 + b_1 * X + b_2 * Z$$

- both  $b_1$  and  $b_2$  have values that are the same regardless of the values of X and Z
- these slopes are constant
- the slopes have one (constant) value, a main or general effect

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## a moderated model: adding a parameter to allow slopes to vary

$$Y = b_0 + b_1 * X + (b_2 + b_3 * X) * Z$$

- in this equation,  $b_1$  is the slope of X and  $(b_2 + b_3 * X)$  is the slope of Z
- if we distribute Z over its slope, we can see why we use a product term as a new predictor in a moderated model

$$Y = b_0 + b_1 * X + b_2 * Z + b_3 * X * Z$$

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### a concrete example: non-moderated

- from Exam 2 last semester
- do TV watching and 'ability' predict achievement?
- summary (based on *M*-centered predictors)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	50.09600	0.27978	179.052	< 2e-16 ***
tv.c	-0.51000	0.16216	-3.145	0.00176 **
ability.c	0.61800	0.03003	20.582	< 2e-16 ***

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### a concrete example: moderated

- from Exam 2 last semester
- does TV watching predict achievement differently for different levels of ability?
- summary (based on *M*-centered predictors)

	Estimate	Std. Error	t value	Pr(> t )
(Intercept)	49.75889	0.27137	183.364	< 2e-16 ***
tv.c	-0.35629	0.15634	-2.279	0.0231 *
ability.c	0.60896	0.02869	21.223	< 2e-16 ***
tv.c:ability.c	-0.11548	0.01644	-7.026	7.06e-12 ***

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### interpretation of slopes changes when there is an interaction!

- slopes for predictors are no longer "main" or general effects
- they are only about a single specific value (zero!) of the other variable
- this is why the "pick-a-point" method of probing an interaction involves centering predictors at a bunch of values → changes the meaning of 0
- the change in the interpretation of a slope when its predictor is involved in an interaction is not well-understood by many (not just some students, but researchers in general!)

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### probing the TV × ability interaction

- center ability at its M and at  $M \pm 1SD$
- the slope of TV for low ability ( $M - 1SD$ ) is 0.74
- the slope of TV for average ability (M) is -0.36
- the slope of TV for high ability ( $M + 1SD$ ) is -1.45
- the slope of TV is not one thing anymore; it varies depending on the level of ability

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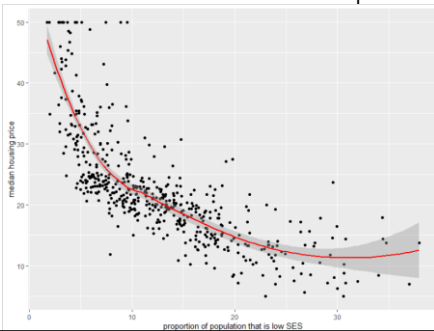
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what if there's only one predictor but it doesn't have a constant slope?



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### some terminology

- why is linear regression called *linear*?
- because the relationship between each predictor and the outcome is linear

$$Y = b_0 + b_1X$$

- as  $X$  increases,  $Y$  increases linearly
- the phrase often seen is that the model is "linear in the parameters"

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## some terminology

- how can a nonlinear relationship be handled by a linear model?
- when a predictor is transformed (e.g., by squaring), the relationship between the transformed predictor and the outcome is linear

$$Y = b_0 + b_1X + b_2X^2$$

- as  $X^2$  increases,  $Y$  increases linearly

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## how does this new parameter work?

- the idea is that the slope of  $X$  changes as the value of  $X$  changes
- so we can take our two-parameter model

$$Y = b_0 + b_1X$$

- and alter it so the slope of  $X$  depends on  $X$ 's value

$$Y = b_0 + (b_1 + b_2X)X$$

$$Y = b_0 + b_1X + b_2X^2$$

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## slope interpretation

- what is the slope of  $X$  when  $X = 0$ ?

$$Y = b_0 + b_1X + b_2X^2$$

$$Y = b_0 + (b_1 + b_2X)X$$

- it's  $b_1$ , **but only when  $X = 0$** ; this is the same kind of interpretation restriction that emerged with interactions
- slopes are (again) no longer main effects; they are simple/point slopes, interpretable at only a single value (of  $X$ , in this case)

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## summary

- to model a non-constant slope, we will add a parameter to allow the slope to vary
- that parameter will be the slope of a power ( $X$  raised to some power)
- we'll talk next week about how to approach these models

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