Simulation of Simpson's Paradox With Palmer Penguin Data

Andy Grogan-Kaylor

16 Feb 2021 08:01:33

Background

Simpson's paradox occurs when a bivariate association is reversed in a multivariate model. This example using the Palmer Penguins Data was inspired by a tweet by Andrew Heiss.



Figure 1: Palmer Penguins Illustration from @allison_horst

To begin, a little definition of penguin terminology is in order. Note the diagram defining culmen depth below.



Figure 2: Culmen Depth from @allison_horst

Setup

- . clear all
- . cd "/Users/agrogan/Desktop/newstuff/simpsons-paradox-palmer-penguins"/Users/agrogan/Desktop/newstuff/simpsons-paradox-palmer-penguins
- . use "penguins.dta"

Bivariate

. twoway (scatter culmen_depth_mm body_mass_g) ///

```
> (lfit culmen_depth_mm body_mass_g), ///
> title("Culmen Depth by Body Mass") ///
> caption("Palmer Penguin Data") ///
> scheme(michigan)
. graph export mygraph1.png, width(1000) replace
(file /Users/agrogan/Desktop/newstuff/simpsons-paradox-palmer-penguins/mygraph1.png
> written in PNG format)
```

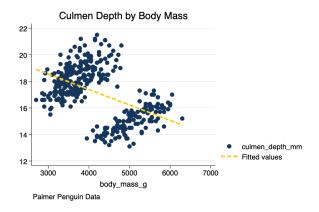


Figure 3: Scatterplot and Linear Fit

. regress cul	men_depth_mm b	ody_mass_g					
Source	SS	df	MS	Num	Number of obs		342
				F(1	, 340)	=	97.41
Model	296.15994	1	296.15994	Pro	b > F	=	0.0000
Residual	1033.67459	340	3.04021939	R-s	R-squared		0.2227
				Adj	R-squared	=	0.2204
Total	1329.83453	341	3.89980801	. Roo	Root MSE		1.7436
culmen_dep_m	Coef.	Std. Err.	t	P> t	[95% Co	onf.	Interval]
body_mass_g cons	0011621 22.03395	.0001177	-9.87 43.75	0.000	001393 21.0433		0009305 23.02455

Multivariate

- . twoway (scatter culmen_depth_mm body_mass_g) ///
- > (lfit culmen_depth_mm body_mass_g), ///
- > by(species, title("Culmen Depth by Body Mass") caption("Palmer Penguin Data")) ///
- > scheme(michigan)
- . graph export mygraph2.png, width(1000) replace
 (file /Users/agrogan/Desktop/newstuff/simpsons-paradox-palmer-penguins/mygraph2.png
 > written in PNG format)

The association is reversed when we take into account multiple variables.

. regress culmen_depth_mm body_mass_g species $\,$

	Source	SS	df	MS	Number of obs	=	342
					F(2, 339)	=	225.41
	Model	759.047284	2	379.523642	Prob > F	=	0.0000
	Residual	570.787248	339	1.6837382	R-squared	=	0.5708
_					Adj R-squared	=	0.5683

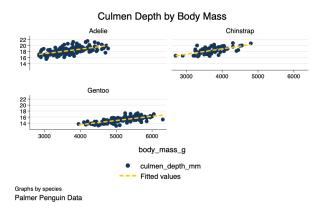


Figure 4: Scatterplot and Linear Fit

Total	1329.83453	341	3.8998080	1 Root	MSE =	1.2976
culmen_dep_m	Coef.	Std. Err.	t	P> t	[95% Conf.	Interval]
body_mass_g species _cons	.0004877 -1.974985 18.89014	.0001326 .1191142 .4200224	3.68 -16.58 44.97	0.000 0.000 0.000	.0002269 -2.209281 18.06396	.0007485 -1.740689 19.71631