

Adding interactions to regression models

Consider two predictors: $Y = \beta_0 + \beta_1^{(hs)} X_1 + \beta_2^{(ig)} X_2 + \epsilon$

\Rightarrow association between ig scores does not depend on hs

How can we relax this assumption?

$$\underline{E[Y | X_1=1, X_2] - E[Y | X_1=0, X_2] = \beta_1 + \beta_2 X_2}$$

want this to depend on X_2

assume linear dependence \nearrow

$$\begin{aligned} \text{New model: } Y &= \beta_0 + (\beta_1 + \beta_2 X_2) X_1 + \beta_2 X_2 + \epsilon \\ &= \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_{1,2} X_1 X_2 + \epsilon \end{aligned}$$

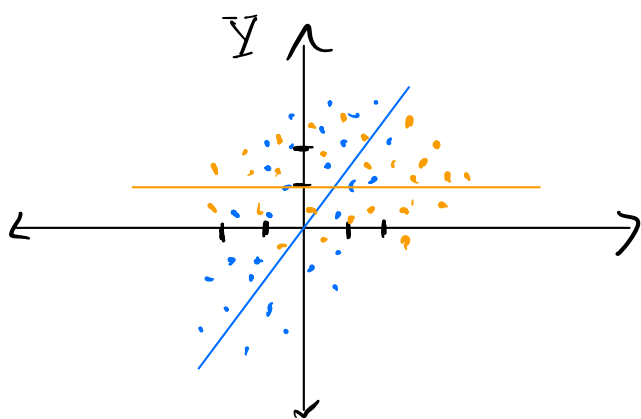
In order to fit coefficients w/ least squares write $X_3 = X_1 X_2$

$$Y = \beta_0 + \beta_1 X_1 + \beta_2 X_2 + \beta_3 X_3 + \epsilon$$

(interaction)

Example Suppose $X_1 \sim \text{Bernoulli}(1/2)$ $\beta_0 = 0, \beta_1 = 1$
 $X_2 \sim \text{Normal}(0,1)$ $\beta_2 = 1, \beta_{1,2} = -1$

Sketch data consistent with these parameters



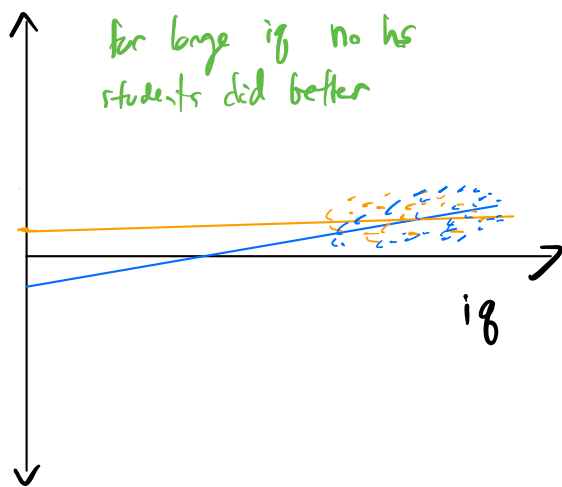
$$\text{when } X_1 = 0, Y = \beta_0 + \beta_2 X_2 + \epsilon = X_2$$

$$\begin{aligned} \text{when } X_1 = 1, Y &= \beta_0 + \beta_1 + (\beta_1 + \beta_2) X_2 + \epsilon \\ &= 1 + 0 X_2 + \epsilon \end{aligned}$$

Example (test scores)

	$\hat{\beta}$	p-val
Cost	-11.5	0.4
hs	51.2	0.001
ig	0.96	0.000
hs-ig	-0.48	0.003

Sketch data that is consistent w/ these values, focus on relation between regression lines, not exact #s



when $X_{hs} = 0$

$$Y = -11.5 + 0.96 X_{ig} + \epsilon$$

when $X_{hs} = 1$

$$Y = -11.5 + 51.2 + (0.96 - 0.48) X_{ig} + \epsilon$$

$$\approx 41 + 0.48 X_{ig} + \epsilon$$

typical ig $\approx 80 \Rightarrow \mu_Y \approx$

fourier models - see typed notes section 4/5