

# Lecture 25: Linear Regression Part II

Chapter 7.2-7.4

## Questions for Today: Example From Text

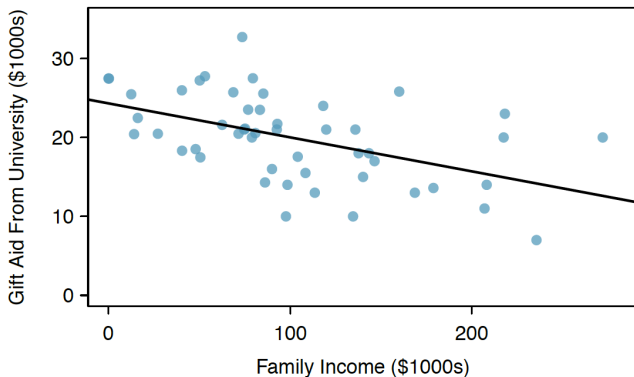
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- ▶ Outcome variable: gift aid



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Using these values,

	family income in \$1000's (x)	gift aid in \$1000's (y)
mean	$\bar{x} = 101.8$	$\bar{y} = 19.94$
sd	$s_x = 63.2$	$s_y = 5.46$
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they fit the **least-squares line**:

$$\hat{y} = b_0 + b_1x$$

$$\widehat{\text{aid}} = 24.3 - 0.0431 \times \text{family\_income}$$

What do 24.3 and  $-0.0431$  mean?

## Point Estimates of Intercept

Point estimate of intercept  $b_0$ : 24.3 (in \$1000's) describes the average aid if the family had no income.

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In this case it is relevant since some families make no income, but the intercept may have little or no practical value if there are no observations near  $x = 0$ .



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In example: for each additional \$1000 of family income, we expect a student to receive a difference of  $\$1000 \times (-0.0431) = -\$43.10$  in aid on average.

Even though we've labeled aid as the outcome variable, we are not positing a causal relationship; just an association.

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What would be the gift aid given to a family with \$1,000,000 (i.e.  $x = 1000$ ) in family income?

$$24.3 - 0.0431 \times 1000 = -18.8$$

The school will take \$18,800 dollars away from you?

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Ex: Ebay price for the video game Mario Kart. We convert the categorical  $x$  into a **indicator variable** `cond_new` which has 2 **levels**:

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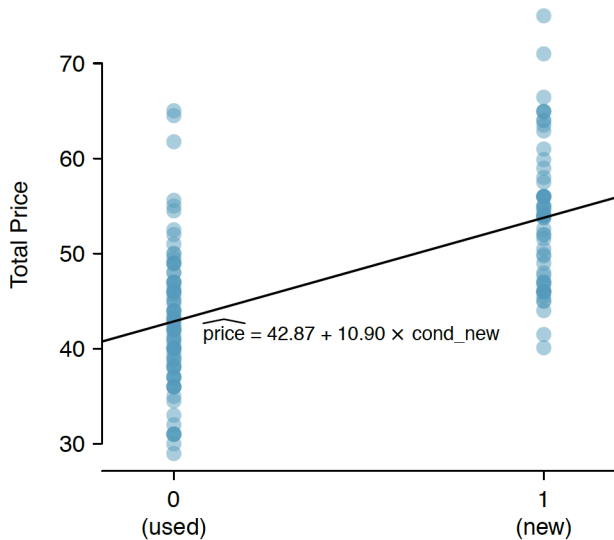
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The linear model is thus

$$\widehat{\text{price}} = b_0 + b_1 \times \text{cond\_new}$$



## Categorical Predictor x With Two Levels



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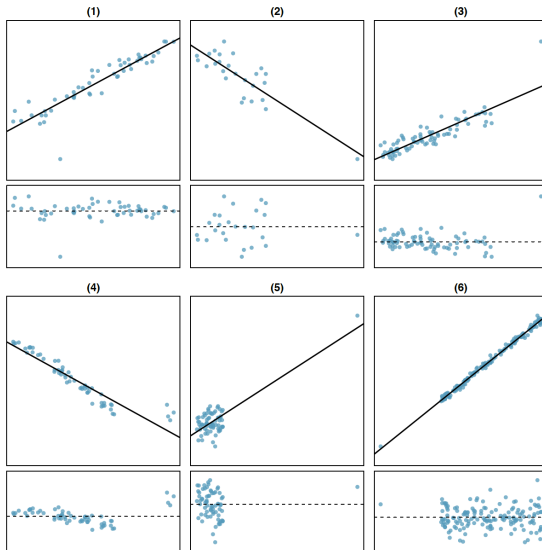
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This can be generalized for predictor variables  $x$  with more than two levels.

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Points that fall horizontally away from the center of the cloud tend to pull harder on the line, so we call them points with high **leverage**, i.e. large influence.

## Next Example

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But first, are the changes from

- ▶ 100 to 200
- ▶ 100,100 to 100,200

the same?

# Next Time

Multiple Regression: As opposed to **simple linear regression** where there is only one predictor/explanatory variable  $x$ , we now consider **many** variables  $x_1, x_2, \dots$