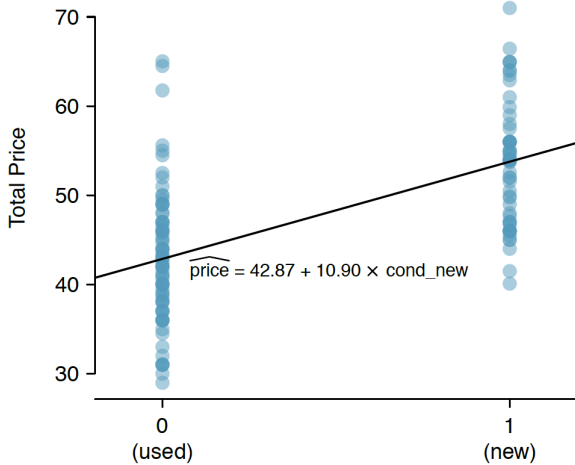


Lecture 26: Multiple Regression

Chapter 8.1

Categorical Predictor x With Two Levels



Simple Linear Regression Regression Table

eBay price of old vs new Mario Kart using $n = 141$. On page 355:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	42.87	0.81	52.67	0.0000
cond_new	10.90	1.26	8.66	0.0000
	$df = 139$			

Confidence Interval and Hypothesis Test for β_1

Questions for Today

Say on top of `cond_new` we are given three additional predictors:



Questions for Today

Say on top of `cond_new` we are given three additional predictors:



- ▶ `stock_photo`: is there a stock photo?
- ▶ `duration`: length of the auction in days (1 to 10)
- ▶ `wheels`: number of Wii wheels included

Questions for Today

How do we simultaneously incorporate all four predictors to model the eBay auction price?

Multiple Regression

Multiple Regression Results Table

On page 357:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	36.21	1.51	23.92	0.00
cond_new	5.13	1.05	4.88	0.00
stock_photo	1.08	1.06	1.02	0.31
duration	-0.03	0.19	-0.14	0.89
wheels	7.29	0.55	13.13	0.00
	$df = 136$			

where $df = n - k - 1 = 141 - 4 - 1 = 136$

Interpretation of Point Estimates

Comparison of Results

For simple linear regression:

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cond_new	10.90	1.26	8.66	0.00

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For simple linear regression:

	Estimate	Std. Error	t value	$\text{Pr}(> t)$
cond_new	10.90	1.26	8.66	0.00

For multiple regression:

	Estimate	Std. Error	t value	$\text{Pr}(> t)$
cond_new	5.13	1.05	4.88	0.00

Why the different point estimate?

Comparison of Result

Because `cond_new` is linearly correlated with `wheels`. We say that two predictor variables are **collinear** when they are correlated, and this complicates model estimation.

Comparison of Result

Because `cond_new` is linearly correlated with `wheels`. We say that two predictor variables are **collinear** when they are correlated, and this complicates model estimation.

In general we must be wary of predictor variables that are collinear, because the coefficient estimates may change erratically in response to small changes in the model or the data.

R^2 to Describe the Strength of Fit

Important Concept in Model Fitting

R^2_{adj} describes the strength of fit while adhering to the following:

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R_{adj}^2 describes the strength of fit while adhering to the following:

- ▶ **Parsimony**: Adoption of the simplest assumption in the formulation of a theory or in the interpretation of data.

Important Concept in Model Fitting

R_{adj}^2 describes the strength of fit while adhering to the following:

- ▶ **Parsimony**: Adoption of the simplest assumption in the formulation of a theory or in the interpretation of data.
- ▶ **Occam's Razor**: When you have two competing theories that make exactly the same predictions, the simpler one is the better.

Adjusted R^2_{adj}

Parsimony/Occam's Razor

Pared Down Mario Kart Regression Output

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	41.34	1.71	24.15	< 2e-16
condused	-5.13	1.05	-4.88	2.91e-06
stockPhotoyes	1.08	1.06	1.02	0.308
duration	-0.03	0.19	-0.14	0.888
wheels	7.30	0.55	13.13	< 2e-16

Residual standard error: 4.901 on 136 degrees of freedom
Multiple R-squared: 0.719, Adjusted R-squared: 0.7108

Pared Down Mario Kart Regression Output

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(Intercept)	41.34	1.71	24.15	< 2e-16
condused	-5.13	1.05	-4.88	2.91e-06
stockPhotoyes	1.08	1.06	1.02	0.308
duration	-0.03	0.19	-0.14	0.888
wheels	7.30	0.55	13.13	< 2e-16

Residual standard error: 4.901 on 136 degrees of freedom

Multiple R-squared: 0.719, Adjusted R-squared: 0.7108

Duration doesn't seem to be all that informative. Why not drop it?

Pared Down Mario Kart Regression Output

Coefficients:

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	41.22	1.49	27.65	< 2e-16
condused	-5.18	1.00	-5.20	7.21e-07
stockPhotoyes	1.12	1.02	1.10	0.275
wheels	7.30	0.54	13.40	< 2e-16

Residual standard error: 4.884 on 137 degrees of freedom

Multiple R-squared: 0.719, Adjusted R-squared: 0.7128

Next Time

Is there a systematic way to pick which predictor variables to include?

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Checking model assumptions as well.