



dummydu...

Two dummy variables without an interaction

As an example, we will use a version of the CPS data:

```
dat <- read.csv("https://rtgodwin.com/data/twodummies.csv")
```

In this example, the **university** variable is a **dummy** variable which equals to 1 if the individual has a university (BA) degree, and 0 otherwise. The other dummy variable in the data is **female**.

Variable	Description
wage	hourly wage of the worker
female	= 1 if the individual is female = 0 if male
university	= 1 if the individual has a university degree = 0 if no university degree
age	the age of the worker in years

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{female} + \beta_2 \text{university} + \beta_3 \text{age} + \epsilon$$

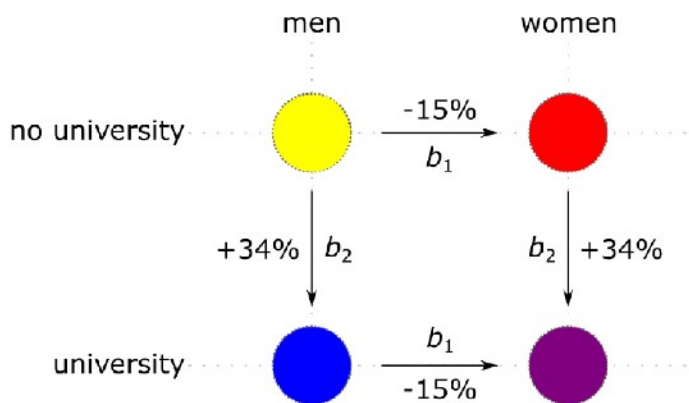
Estimate this in R:

```
summary(lm(log(wage) ~ female + university + age, data = dat))
```

	Estimate	Std. Error	t value	Pr(> t)	
(Intercept)	2.016700	0.043388	46.48	<2e-16	***
female	-0.152178	0.008503	-17.90	<2e-16	***
university	0.337940	0.008409	40.19	<2e-16	***
age	0.026435	0.001439	18.37	<2e-16	***

The interpretation of the results is that women make 15% less than men, and that a university degree increases wage by 34%. However, this model does not allow for the possibility that education has a different effect for women than it does for men. There is a difference between men and women, and there is a difference for a university degree, but there is no difference in the effect of university for men vs. women. See Figure 8.10.

Figure 8.10: University makes a difference, and gender makes a difference, but there is not a separate difference for university educated women.



3

$$\log(\text{wage}) = \beta_0 + \beta_1 \text{female} + \beta_2 \text{university} + \beta_3 (\text{female} \times \text{university}) + \beta_4 \text{age} + c$$

where β_3 is the additional percentage increase in wages for women with an education, versus men with an education. In R, we can do this by:

```
summary(lm(log(ahe) ~ female + bachelor + I(female * bachelor) + age,
           data = cps))
```

	Estimate	Std. Error	t value	Pr(> t)
(Intercept)	2.01896	0.04338	46.541	< 2e-16 ***
female	-0.17347	0.01173	-14.791	< 2e-16 ***
university	0.31895	0.01107	28.809	< 2e-16 ***
I(female * university)	0.04489	0.01704	2.635	0.00842 **
age	0.02662	0.00144	18.479	< 2e-16 ***

It is estimated that women make 17% less than men, that men with a degree make 32% more than men without a degree, and that women with a degree make $(32\% + 4.5\% \approx 36\%)$ more than women without a degree. There is a difference for men, a difference for

It is estimated that women make 17% less than men, that men with a degree make 32% more than men without a degree, and that women with a degree make (32% + 4.5% \approx 36%) more than women without a degree. There is a difference for men, a difference for women, and the difference between these two differences is β_3 (4.5%). See Figure 8.11.

Figure 8.11: University makes a difference, and gender makes a difference, but there is not a separate difference for university educated women.

