

28.3: The t-test as a Linear Model

The t-test is often presented as a specialized tool for comparing means, but it can also be viewed as an application of the general linear model. In this case, the model would look like this:

$$\widehat{BP} = \widehat{\beta}_1 * Marijuana + \widehat{\beta}_0$$

However, smoking is a binary variable, so we treat it as a *dummy variable* like we discussed in the previous chapter, setting it to a value of 1 for smokers and zero for nonsmokers. In that case, $\widehat{\beta}_1$ is simply the difference in means between the two groups, and $\widehat{\beta}_0$ is the mean for the group that was coded as zero. We can fit this model using the `lm()` function, and see that it gives the same t statistic as the t-test above:

```
##
## Call:
## lm(formula = TVHrsNum ~ RegularMarij, data = NHANES_sample)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.293 -1.133 -0.133  0.867  2.867
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      2.133      0.119   17.87  <2e-16 ***
## RegularMarijYes    0.660      0.249    2.65  0.0086 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 1.5 on 198 degrees of freedom
## Multiple R-squared:  0.0343, Adjusted R-squared:  0.0295
## F-statistic: 7.04 on 1 and 198 DF, p-value: 0.00861
```

We can also view the linear model results graphically (see the right panel of Figure 28.1). In this case, the predicted value for nonsmokers is $\widehat{\beta}_0$ (2.13) and the predicted value for smokers is $\widehat{\beta}_0 + \widehat{\beta}_1$ (2.79).

To compute the standard errors for this analysis, we can use exactly the same equations that we used for linear regression – since this really is just another example of linear regression. In fact, if you compare the p-value from the t-test above with the p-value in the linear regression analysis for the marijuana use variable, you will see that the one from the linear regression analysis is exactly twice the one from the t-test, because the linear regression analysis is performing a two-tailed test.

28.3.1 Effect sizes for comparing two means

The most commonly used effect size for a comparison between two means is Cohen's d, which (as you may remember from Chapter 18) is an expression of the effect in terms of standard error units. For the t-test estimated using the general linear model outlined above (i.e. with a single dummy-coded variable), this is expressed as:

$$d = \frac{\widehat{\beta}_1}{SE_{\text{residual}}}$$

We can obtain these values from the analysis output above, giving us a $d = 0.45$, which we would generally interpret as a medium sized effect.

We can also compute R^2 for this analysis, which tells us how much variance in TV watching is accounted for. This value (which is reported in the summary of the `lm()` analysis) is 0.03, which tells us that while the effect may be statistically significant, it accounts for relatively little of the variance in TV watching.

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