Section 12, 11/18/19

t-Statistic (page 421)

Motivation

Suppose we have measurements of interest from two groups, n measurements X_1, \ldots, X_n from a control group and m measurements Y_1, \ldots, Y_m from a treatment group. The effect size is the difference $\mu_X - \mu_Y$ and may let us make inferences on the effect of the treatment. The natural estimate of this is $\overline{X} - \overline{Y}$. If we assume that the X's and Y's are independent and follow normal distributions $N(\mu_X, \sigma^2)$ and $N(\mu_Y, \sigma^2)$ respectively, then clearly

$$ar{X} - ar{Y} \sim N(\mu_X - \mu_Y, \sigma^2ig(rac{1}{n} + rac{1}{m}ig))$$

If we knew σ^2 , then we could construct a confidence interval using our knowledge of the normal distribution. However, we typically don't know σ^2 and so must estimate it. We combine information form the sample variances and define the **pooled sample variance** as

$$s_p^2 = rac{(n-1)s_X^2 + (m-1)s_Y^2}{m+n-2}$$

This is the weighted average of the sample variances. This leads us to the following theorem.

Theorem A

Suppose $X_1, \ldots, X_n \sim N(\mu_X, \sigma^2)$ and $Y_1, \ldots, Y_m \sim N(\mu_Y, \sigma^2)$ are independent of each other. Then the t-statistic

$$t = rac{(ar{X} - ar{Y}) - (\mu_X - \mu_Y)}{s_p \sqrt{rac{1}{n} + rac{1}{m}}}$$

follows a t-distribution with m + n - 2 degrees of freedom.

Proof

We know that a t-distribution is the ratio of a normal distribution and the square root of an independent chisquared distribution divided by its degrees of freedom.

First note that $(n-1)s_X^2/\sigma^2$ and $(m-1)s_Y^2/\sigma^2$ are chi-squared random variables with n-1 and m-1 degrees of freedom. They are also independent as the X's and Y's are. Thus their sum is also a chi squared distribution with m + n - 2 degrees of freedom. Thus

$$U = \frac{(\bar{X} - \bar{Y}) - (\mu_X - \mu_Y)}{\sigma \sqrt{\frac{1}{n} + \frac{1}{m}}}$$

is distributed according to a standard normal and

$$V^2 = ig(rac{(n-1)s_X^2}{\sigma^2} + rac{(m-1)s_Y^2}{\sigma^2}ig)rac{1}{m+n-2}$$

is a chi-squared distribution divided by its degrees of freedom. So, $t = U/\sqrt{V}$ follows a t-distribution.