# Frequentist Inference, pt. III

ECON 3640-001

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# Motivation

# Housekeeping

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Notes based on Keller (2009):
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• Chapter **12**, sections 12.1.

## Motivation

So far, we have assumed that the **population standard deviation** ( $\sigma$ ) was known when computing confidence intervals and hypothesis testing.

This assumption, however, is unrealistic.

Now, we **relax** such belief, and move on using a very similar approach.

Recall the formula for the z test:

$$z=rac{ar{x}-\mu}{\sigma/\sqrt{n}}$$

Now that the population standard deviation is *unknown*, what is the **best move**?

Replace it by its sample estimator, s!

$$t=rac{ar{x}-\mu}{s/\sqrt{n}}$$

Now, we do not know the population standard deviation anymore.

However, frequentist methods still assume that the population mean  $(\mu)$  is **normally distributed**.

In 1908, William S. Gosset came up with the **Student** *t* distribution, whose mean and variance are

• 
$$E(X) = 0$$

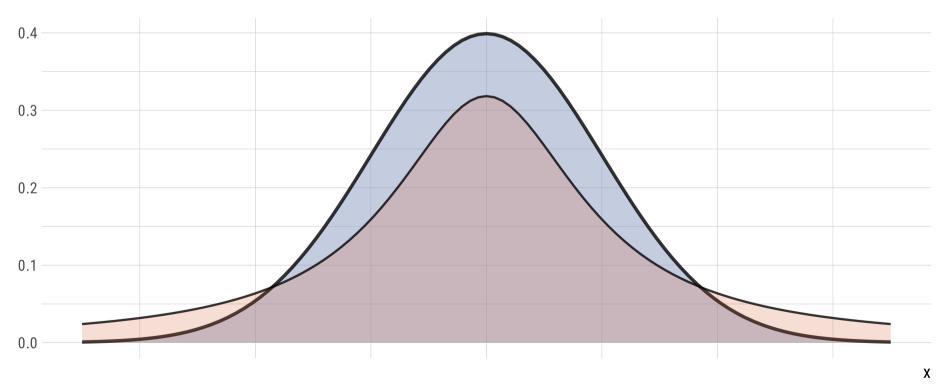
• 
$$\operatorname{Var}(X) = \frac{\nu}{\nu - 2}$$

where  $\nu = n-1$ .



The t distribution has a "mound-shaped" density curve, while the Normal's is bell-shaped.

#### **Standard Normal and t distributions**

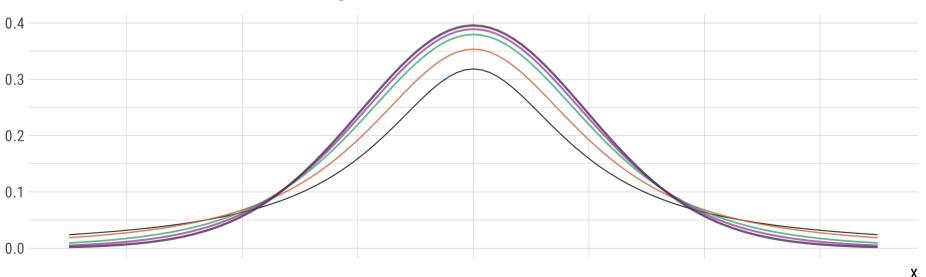


Its only parameter is  $\nu$ , the **degrees of freedom** of the distribution:

$$X \sim t(
u)$$

The larger the value of  $\nu$ , the more **similar** the t distribution is to the Normal.

#### t distributions with different degrees of freedom



When  $\sigma$  is unknown, we can define the **confidence interval** for the population mean  $(\mu)$  as:

$$ar{x}\pm t_{lpha/2,\,
u}igg(rac{s}{\sqrt{n}}igg)$$

Notice that, in addition to the significance level  $(\alpha)$ , we also must take the number of degrees of freedom  $(\nu = n - 1)$  into account to find the **standard error** for  $\bar{x}$ .

Furthermore, for **hypothesis testing**, the *t-statistic* is obtained with:

$$t=rac{ar{x}-\mu}{s/\sqrt{n}}$$

We **cannot** use the sampling distribution of  $\mu$  anymore, since the population standard deviation is unknown.

Such assumption considers an **infinitely large** sample, which is almost never the case in practice.

For **smaller** sample sizes, the *t* distribution is extremely useful, and its shape is conditional on this sample size.

But why (n-1) degrees of freedom?

#### An example:

Assuming that it can be profitable to recycle newspapers, a company's financial analyst has computed that the firm would make a profit if the mean weekly newspaper collection from each household exceeded 2 lbs. His study collected data from a sample of 148 households. The calculated sample average weight was 2.18 lbs.

Do these data provide sufficient evidence to allow the analyst to conclude that a recycling plant would be profitable? Assume a significance level of 1%, and a sample variance of .962 lbs<sup>2</sup>.

Next time: Review & practice problems