

## 21 Student's $t$ distribution

### 21.1 $N(\mu, 1)$

The following data comes from sampling 10 independent random variables, all  $N(\mu, 1)$ , where  $\mu$  needs to be estimated.

-0.68 1.69 1.49 1.88 2.75 1.58 2.20 2.32 1.28 1.69

$$n = 10, \quad \bar{X} = 1.62, \quad \sigma^2 = 1$$

The sample average  $\sim N(\mu, 1/10)$ . The ‘Z-score’ is

$$\sqrt{10}(\bar{X} - \mu)$$

We use this information to estimate  $\mu$ .

$$\sqrt{10}(1.62 - \mu)$$

If we take the interval  $[-1.96, 1.96]$ , and  $Z \sim N(0, 1)$ ,  $Z$  will be in this interval with 95% probability.

Where does 1.96 come from? From the tables, the probability  $Z \leq 1.96$  is 97.5%, so the probability  $Z \geq 1.96 = \text{probability } Z \leq -1.96 = 2.5\%$ .  $[-1.96, 1.96]$  is a symmetric interval with probability 95%, leaving 5% in the ‘tails.’

$$-1.96 \leq \sqrt{10}(\bar{X} - \mu) \leq 1.96$$

with 95% probability. Now

That means that  $\sqrt{10}(\bar{X} - \mu)$  will belong to the interval 19 times out of 20. We put this differently: knowing  $\bar{X}$  but not  $\mu$ , this will happen 19 times in 20, and

$$\bar{X} - \frac{1.96}{\sqrt{10}} \leq \mu \leq \bar{X} + \frac{1.96}{\sqrt{10}}$$

with 95% *confidence*.

$$1.00 \leq \mu \leq 2.24$$

with 95% confidence. This is 95% **confidence interval** for  $\mu$ .

### 21.2 Confidence intervals for $\mu$ , variance unknown

‘Student’ was the pseudonym of William Sealey Gosset, who was a statistician working in Guinness’s Brewery about a century ago. See Wikipedia for some background. His work involved testing fields of barley, etcetera; small samples, unlike much-repeatable laboratory measurements.

Here his methods are applied to estimating the mean of a normal distribution, *without knowing*  $\sigma$ . Except for that difference, confidence intervals can be constructed as previously (with  $\sigma$  known).

As usual, given  $X_1, \dots, X_n$ ,

$$\bar{X} = \frac{1}{n} \sum_{i=1}^n X_i, \quad S^2 = \frac{1}{n-1} \sum_{i=1}^n (X_i - \bar{X})^2$$

$\bar{X}$  is close to  $\mu$  and  $S^2$  is close to  $\sigma^2$ .

For the previous example,

-0.68 1.69 1.49 1.88 2.75 1.58 2.20 2.32 1.28 1.69

$$n = 10, \quad \bar{X} = 1.62, \quad S^2 = .845, \quad S = 0.92$$

$$t = \sqrt{n} \frac{\bar{X} - \mu}{S} \tag{21.1}$$

is approximately  $N(0, 1)$ : this is the Central Limit Theorem. Student's t-distribution is the *true* distribution of this quantity for given  $n$ .

As with the estimate of  $S$ , there is a small amount of cancellation, and the number of ‘degrees of freedom’ in the above random variable (21.1) is

$$n - 1.$$

**(21.2) Proposition** *When sample average and sample standard deviation are taken from a sampling of  $n$  iid random variables, all distributed  $N(\mu, \sigma^2)$ , The  $t$ -statistic (defined in Equation 21.1) follows*

Student's  $t$ -distribution with  
 $n - 1$  degrees of freedom.

*The  $P\%$  confidence interval for  $\mu$ , based on Student's  $t$ -distribution, is*

$$\left[ \bar{X} \mp \alpha \frac{S}{\sqrt{n}} \right]$$

*where  $\alpha$  depends on the percentage level  $P$  and on the number of degrees of freedom,  $n - 1$ .* ■

In this example, the  $t$ -statistic is

$$\sqrt{10} \left( \frac{1.62 - \mu}{0.92} \right)$$

To get a confidence interval with 95% confidence, we go to the official tables, page 40, noting that the table is organised differently from the Normal distribution. It gives the percentage points. The  $t$ -distribution is symmetric, so we can use the 2.5% percentage point. There are 9 degrees of freedom: the point is 2.262 (compare with 1.96!)

$$\sqrt{10} \left( \frac{1.62 - \mu}{0.92} \right) \in [-2.262, 2.262]$$

with 95% confidence.

$$\mu \in 1.62 \mp \frac{.92}{\sqrt{10}} \times 2.262$$

$$1.62 \mp 0.658$$

$$\mu \in [0.962, 2.278]$$