

## The Safety Net: Why the $t$ -distribution?

### 1. The Penalty for Guessing

Because we are using  $s$  (a sample guess) instead of  $\sigma$  (the population truth), our ruler is inherently less stable. To account for this uncertainty, we swap the Normal ( $z$ ) curve for the  **$t$ -distribution**.

The  $t$ -curve is indexed by **Degrees of Freedom** ( $df$ ). For a single sample mean, the formula is:

$$df = n - 1$$

**Task:** Calculate the  $df$  for the following Lead Engineer audits:

- You audit a single case ( $n = 12$ ):  $df =$  \_\_\_\_\_
- You audit a double-case batch ( $n = 24$ ):  $df =$  \_\_\_\_\_
- You audit a full pallet ( $n = 144$ ):  $df =$  \_\_\_\_\_

### 2. Identifying the “Safety Net”

Open `visualizingt.r` in RStudio and run **Lines 8 through 34**. This plot compares the “Steel Rod” (Normal) vs. the “Rubber Band” ( $t$ ,  $df = 2$ ).

**Visual Audit:** Look at the tails of the two curves (the areas far to the left and right).

- Which curve stays higher off the x-axis in the tails? \_\_\_\_\_
- In your own words, why does the  $t$ -distribution need these “thicker” tails when we only have a tiny amount of data?

### 3. The Convergence (Single Case vs. Pallet)

As our sample size ( $n$ ) increases, our estimate of the “noise” ( $s$ ) becomes more stable. The “Rubber Band” begins to tighten until it is as rigid as the “Steel Rod.”

**Task:** Run **Lines 37 through 58** in your script. This overlays three curves: the Normal curve, a  $t$ -curve for a single case ( $df = 11$ ), and a  $t$ -curve for a full pallet ( $df = 143$ ).

#### The Convergence Sketch

*(Sketch the three curves below. Use a solid line for the Normal curve and dashed/dotted lines for the two  $t$ -curves. Label the “Safety Net” gap between them.)*

**The “Critical” Audit:** In your R plot, notice where the curves cross the 5% threshold (the “tails”).

- For the **Normal curve**, the “Rare Event” boundary is at  $\pm 1.96$ .
- For the **Single Case** ( $df = 11$ ), the boundary is further out, at  $\pm 2.20$ .

**Question:** Why does the  $t$ -distribution move the “goalposts” further out when  $n$  is small? If we kept the goalposts at 1.96 for a sample of only 12 bottles, would we be *more* likely or *less* likely to accidentally claim a “Rare Event” that was actually just random noise?

#### 4. Final Reflection: The Law of Certainty

As our sample size ( $n$ ) grows, our “Best Guess” ( $s$ ) becomes so reliable that the  $t$ -distribution eventually turns back into the \_\_\_\_\_ distribution.

**Question:** Look at your sketch on the previous page. As  $n$  increases from 12 to 144, what happens to the “Safety Net” (the gap between the curves)? Why does having more data allow us to shrink this margin?

**Question:** Imagine you find a sample mean that is 3 standard errors away from the Null.

- With  $n = 3$ , the  $t$ -curve is so thick that 3 standard errors is still considered “common.”
- With  $n = 1000$ , the curve is so thin that 3 standard errors is a “Slam Dunk.”

Explain how the extra area in the  $t$ -distribution’s tails protects you from claiming a discovery when you actually just have a shaky estimate.