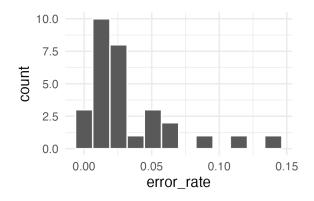
16: CONFIDENCE INTERVALS VIA BOOTSTRAP T DISTRIBUTIONS

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Goal: develop an **interval estimate** of a population parameter

Example: Suppose we observe $X_1, ..., X_n$, which represent the error rates of n different forensic examiners. I am interested in finding a confidence interval for μ , the average error rate in the population. We have a sample of size 30 shown below:



mean	sd	n
0.0325	0.0343	30

Let's review our confidence interval toolkit so far:

- 1. Construct a bootstrap distribution and find a percentile-based confidence interval
- 2. Either (a) assume a normal population or (b) use the CLT to construct a **formula t-based** confidence interval
- 3. Assume a different population distribution and construct a pivot-based confidence interval

Bootstrap percentile interval

```
error_rates <- examiner_error_rates$error_rate
n = length(error_rates)
N = 10^4
boot_means = numeric(N)

for(i in 1:N){
    x <- sample(error_rates, size = n, replace = TRUE)
    boot_means[i] <- mean(x, na.rm = TRUE)
}

quantile(boot_means, probs = c(.025, .975))</pre>
```

```
2.5% 97.5% 0.02147176 0.04576585
```

CLT-based t confidence interval

```
qt(.975, df = 29)
```

[1] 2.04523

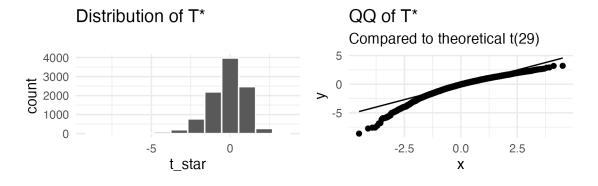
```
t.test(error_rates, conf.level = .95)$conf
```

```
[1] 0.01969500 0.04531505
attr(,"conf.level")
[1] 0.95
```

When do we trust this confidence interval?

```
t_star <- numeric(N)

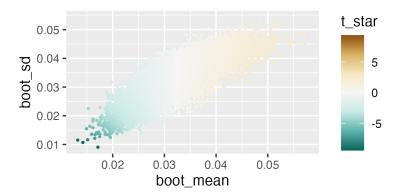
for(i in 1:N){
    x <- sample(error_rates, size = n, replace = TRUE)
    t_star[i] <- (mean(x) - mean(error_rates))/(sd(x)/sqrt(n))
}</pre>
```



• Distribution of *T** is slightly **left-skewed**

• Tails of T* don't seem to match the theoretical t

Why does this happen?



Bootstrap t confidence interval

The CLT-based confidence interval relies on:

$$1 - \alpha = P(t_{\alpha/2} < \frac{\bar{X} - \mu}{s/\sqrt{n}} < t_{1-\alpha/2})$$

Idea: generate "better" quantiles using the bootstrap distribution:

Then solve for μ :

to obtain the bootstrap t confidence interval:

Watch out!

Lower bound is computed from the **upper** percentile and upper bound is computed from the **lower** percentile

Bootstrap t algorithm

- 1. Repeat the bootstrap procedure many times:
 - Take bootstrap sample
 - Compute mean (\bar{X}^*) and sd (s^*) of each bootstrap sample
 - Compute t-statistic for each bootstrap sample: $T^* = \frac{\bar{X}^* \bar{X}}{s^* / \sqrt{n}}$
- 2. Find quantiles $Q_{\alpha/2}^*$ and $Q_{1-\alpha/2}^*$ using distribution of T^*
- 3. Compute confidence interval using the bootstrap t quantiles:

$$(\bar{x}-Q_{1-\alpha/2}^*\frac{s}{\sqrt{n}},\bar{x}-Q_{\alpha/2}^*\frac{s}{\sqrt{n}})$$

Example: Let's find a bootstrap t confidence interval for the 30 forensic examiners

```
n = length(error_rates)
N = 10^4
boot_means <- numeric(N)
boot_tstar <- numeric(N)
for(i in 1:N){
    x <- sample(error_rates, size = n, replace = TRUE)
    boot_means[i] <- mean(x)
    boot_tstar[i] <- (mean(x) - mean(error_rates))/(sd(x)/sqrt(n))
}
quantile(boot_tstar, probs = c(.025, .975))</pre>
```

```
2.5% 97.5%
-2.934144 1.717921

mean(error_rates) - quantile(boot_tstar, probs = c(.025, .975))*sd(error_rates)/sqrt(n)
```

```
2.5% 97.5%
```

Summary of results

0.05088264 0.02174506

	Lower	Upper
Percentile Bootstrap	.021	.0458
CLT t-based	.020	.0453
Bootstrap t	.022	.0509