

1. When conducting the work of Lab 11, we conducted the test that uses the Central Limit Theorem even though the sample size was “small” (i.e., $n < 30$). It turns out, that how “far off” the t -test is can be computed using a first-order Edgeworth approximation for the error. Below, we will do this for the the further observations.

(a) Boos and Hughes-Oliver (2000) note that

$$P(T \leq t) \approx F_Z(t) + \underbrace{\frac{\text{skew}}{\sqrt{n}} \frac{(2t^2 + 1)}{6}}_{\text{error}} f_Z(t),$$

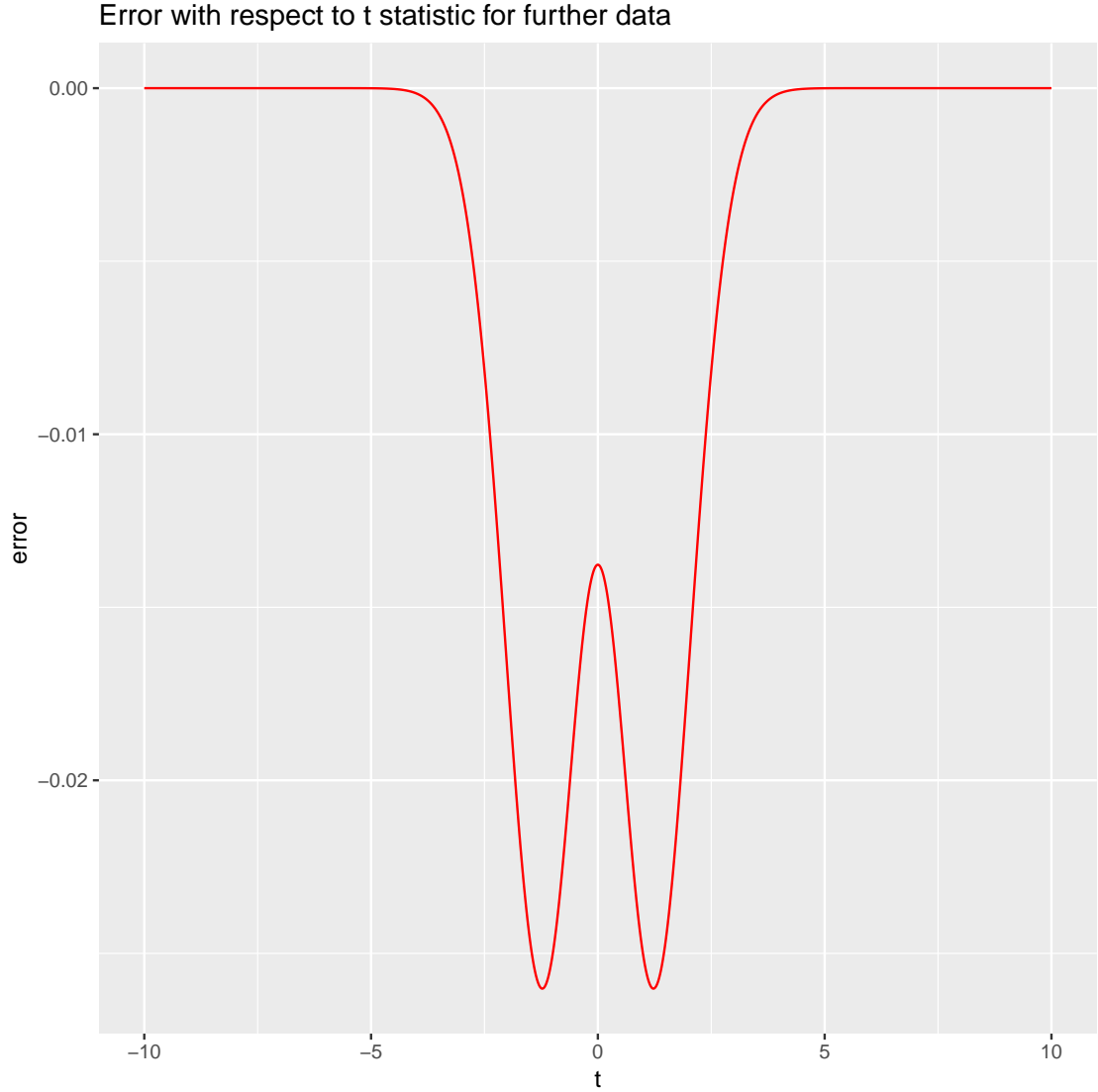
where $f_Z(\cdot)$ and $F_Z(\cdot)$ are the Gaussian PDF and CDF and skew is the skewness of the data. What is the potential error in the computation of the p -value when testing $H_0 : \mu_X = 0; H_a : \mu_X < 0$ using the zebra finch further data?

```
library(tidyverse)
library(e1071)
data = read_csv("zebrafinches.csv")
far.vec = data$farther #farther data
mu0 = 0
n = 25
far.t.stat = (mean(far.vec) - mu0)/(sd(far.vec)/sqrt(n)) #far t statistic
far.pdf = dnorm(far.t.stat) #pdf using the far t statistic
(far.error = (skewness(far.vec)/sqrt(n))*((2*far.t.stat^2 + 1)/6)*far.pdf) #potential error for
## [1] -1.226006e-13
```

As you can see, the potential error is essentially zero.

- (b) Compute the error for t statistics from -10 to 10 and plot a line that shows the error across t . Continue to use the skewness and the sample size for the zebra finch further data.

```
t.stat = seq(-10,10, by = 0.01)
error.vec = (skewness(far.vec)/sqrt(n))*((2*t.stat^2 + 1)/6)*dnorm(t.stat)
error.dat = data.frame(t = t.stat, error = error.vec)
error.plot = ggplot(data = error.dat, aes(x = t, y = error))+
  geom_line(color = "red") +
  labs(title = "Error with respect to t statistic for further data")
error.plot
```



- (c) Suppose we wanted to have a tail probability within 10% of the desired $\alpha = 0.05$. Recall we did a left-tailed test using the further data. How large of a sample size would we need? That is, we need to solve the error formula equal to 10% of the desired left-tail probability:

$$0.10\alpha \stackrel{\text{set}}{=} \underbrace{\frac{\text{skew}}{\sqrt{n}} \frac{(2t^2 + 1)}{6} f_Z(t)}_{\text{error}},$$

which yields

$$n = \left(\frac{\text{skew}}{6(0.10\alpha)} (2t^2 + 1) f_Z(t) \right)^2.$$

2. Complete the following steps to revisit the analyses from lab 11 using the bootstrap procedure.

- (a) Now, consider the zebra finch data. We do not know the generating distributions for the closer, further, and difference data, so perform resampling to approximate the sampling distribution of the T statistic:

$$T = \frac{\bar{x}_r - 0}{s/\sqrt{n}},$$

where \bar{x}_r is the mean computed on the r^{th} resample and s is the sample standard deviation from the original samples. At the end, create an object called `resamples.null.closer`, for example, and store the resamples shifted to ensure they are consistent with the null hypotheses at the average (i.e., here ensure the shifted resamples are 0 on average, corresponding to $t = 0$, for each case).

- (b) Compute the bootstrap p -value for each test using the shifted resamples. How do these compare to the t -test p -values?
 - (c) What is the 5th percentile of the shifted resamples under the null hypothesis? Note this value approximates $t_{0.05, n-1}$. Compare these values in each case.
 - (d) Compute the bootstrap confidence intervals using the resamples. How do these compare to the t -test confidence intervals?
3. Complete the following steps to revisit the analyses from lab 11 using the randomization procedure.
- (a) Now, consider the zebra finch data. We do not know the generating distributions for the closer, further, and difference data, so perform the randomization procedure
 - (b) Compute the randomization test p -value for each test.
 - (c) Compute the randomization confidence interval by iterating over values of μ_0 .
Hint: You can “search” for the lower bound from Q_1 and subtracting by 0.0001, and the upper bound using Q_3 and increasing by 0.0001. You will continue until you find the first value for which the two-sided p -value is greater than or equal to 0.05.
4. **Optional Challenge:** In this lab, you performed resampling to approximate the sampling distribution of the T statistic using

$$T = \frac{\bar{x}_r - 0}{s/\sqrt{n}}.$$

I’m curious whether it is better/worse/similar if we computed the statistics using the sample standard deviation of the resamples (s_r), instead of the original sample (s)

$$T = \frac{\bar{x}_r - 0}{s_r/\sqrt{n}}.$$

- (a) Perform a simulation study to evaluate the Type I error for conducting this hypothesis test both ways.
- (b) Using the same test case(s) as part (a), compute bootstrap confidence intervals and assess their coverage – how often do we ‘capture’ the parameter of interest?

References

Boos, D. D. and Hughes-Oliver, J. M. (2000). How large does n have to be for z and t intervals? *The American Statistician*, 54(2):121–128.