

Announcements

Make sure to sign in on the [google form](#) (I send a list of which section questions are useful for which pset questions afterwards)



Pset 0 due Friday 9/15

Introductions

- Name
- Year
- Previous stats courses
- One question or thought related to lecture last week

Goals each week

- Hand out and explain R code for the week. New relative to last year, we'll plan to not do any in-section coding questions. LLMs are good enough now to do most of your coding for you (and they're allowed in this class!).
- See similar examples to the homework (both in code and analysis).
- Learn something about the world.

Effective sample size

The following problems are intended as a review of Stat 110.

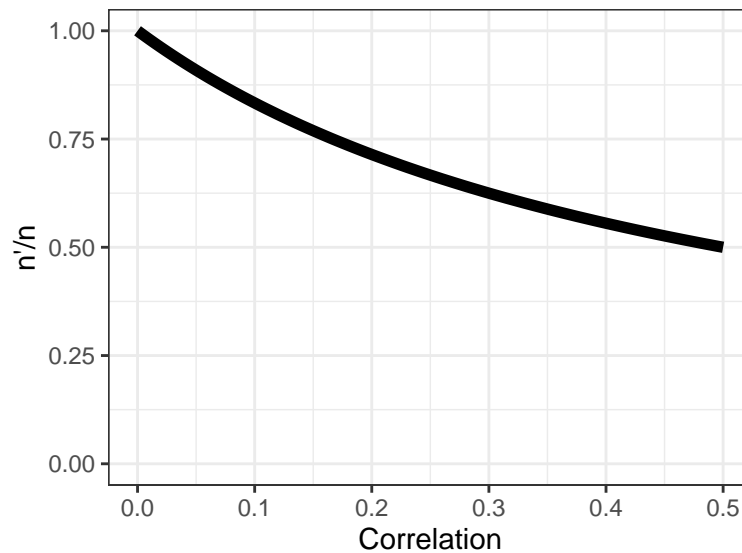
1. Suppose there is a gambler who goes to the casino for n days and makes $Z_1, Z_2, \dots, Z_n \sim \mathcal{N}(0, 1)$ each day where the winnings are independent of each other. (You can assume these are in thousands if the stakes aren't high enough.) What is the distribution of \bar{Z} ?
2. Now, suppose the gambler tends to win and lose in streaks. In particular, let $X_1, X_2, \dots, X_n \sim \mathcal{N}(0, 1)$ marginally be the winnings, but assume neighboring days have correlation ρ . That is,

$$\vec{X} \sim \text{MVN}(\vec{0}, \Sigma), \Sigma = \begin{bmatrix} 1 & \rho & 0 & 0 & \dots \\ \rho & 1 & \rho & 0 & \dots \\ 0 & \rho & 1 & \rho & \dots \\ 0 & 0 & \rho & 1 & \dots \\ \vdots & \vdots & \vdots & \vdots & \ddots \end{bmatrix}$$

Intuitively, should the variance of \bar{X} be higher or lower than the variance of \bar{Z} ?

3. What is the distribution of \bar{X} ?

4. What would the distribution be if the X_i had variance σ^2 instead of 1 but everything else remained the same?
5. Show that the variance can be written as $\vec{c}^T \Sigma \vec{c}$ where \vec{c} is a vector of $1/n$.
6. What is the approximate distribution for large n ?
7. By comparing the distributions in (1) and (6), determine the effective sample size n' when there are n random variables with the correlation structure of (2). That is, if you had n' independent Normals rather than n dependent Normals, what would n' have to be so that the variances of the sample means are the same?
8. Here is a plot of how the effective sample size changes with ρ .



We can test that our calculations are right by using a simulation. Explain what the following code does and whether the results agree with our expectations.

```

library(MASS) # For Multivariate Normal
set.seed(139)

nsim <- 10^5
n <- 70
p <- 0.2
n_eff <- as.integer(n / (1 + 2 * p))

Sigma = matrix(0, nrow = n, ncol = n)
diag(Sigma) <- 1
for (i in 2:n) {
  Sigma[i, i-1] <- p
  Sigma[i-1, i] <- p
}

outputs <- matrix(nrow = nsim, ncol = 3)
for (i in 1:nsim) {
  x <- rnorm(n, 0, 1)
  outputs[i,1] <- mean(x)

  x <- rnorm(n_eff, 0, 1)
  outputs[i,2] <- mean(x)

  x <- mvrnorm(n = 1, rep(0, n), Sigma)
  outputs[i,3] <- mean(x)
}

variances_out <- apply(outputs, 2, var) # Apply over columns
names(variances_out) <- c("Independent n", "Independent n'", "Dependent n")
variances_out

## Independent n Independent n'    Dependent n
##      0.01432340      0.01990641      0.01996290

```

9. You might have noticed that the plot of effective sample size versus correlation stops at a correlation of 0.5. Correlation ranges from -1 to 1, but our set-up actually doesn't work if $\rho > 0.5$ and n is large enough. To have a valid Σ matrix, it must satisfy the property that $\vec{x}^T \Sigma \vec{x} \geq 0$ for all $\vec{x} \in \mathbb{R}^n$ (that is, it must be positive, semi-definite). Show that for $\rho > 0.5$, choosing the vector $\vec{x} = (-1, 1, -1, \dots, -1)^T$ implies $\vec{x}^T \Sigma \vec{x} < 0$ if n is large enough, violating the requirements for Σ . (For simplicity, let n be odd.)

Student- t vs Normal

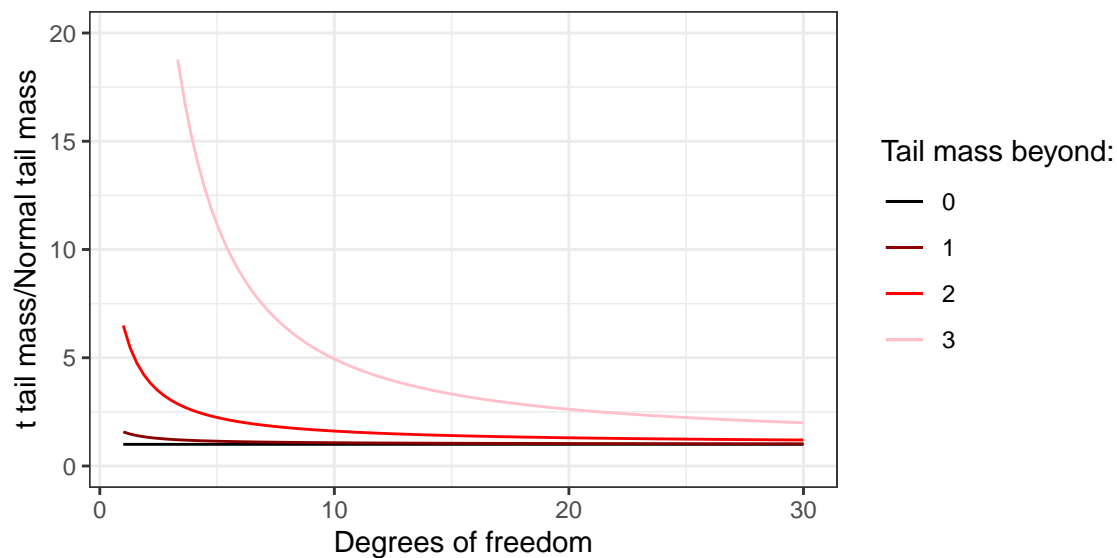
The following problems are intended as a review of Stat 111. We'll prove that the student- t distribution converges to the Normal distribution as its degrees of freedom increase and then analyze this convergence. This fact is useful for large n approximations.

1. Let $T_n \sim t_n$, so T_n can be represented as

$$T_n = \frac{Z}{\sqrt{V_n/n}}, Z \sim \mathcal{N}(0, 1), V_n \sim \chi_n^2$$

which also means V_n can be represented as $V_n = \sum_{i=1}^n Z_i^2$ for $Z_i \sim \mathcal{N}(0, 1)$. Show that $V_n/n \xrightarrow{p} 1$.

2. What tells us that if $V_n/n \xrightarrow{p} 1$, $\frac{1}{\sqrt{V_n/n}} \xrightarrow{p} 1$?
3. What tells us that if $Z \sim \mathcal{N}(0, 1)$ and $\frac{1}{\sqrt{V_n/n}} \xrightarrow{p} 1$, $\frac{Z}{\sqrt{V_n/n}} \xrightarrow{d} \mathcal{N}(0, 1)$?
4. What does this mean about the distribution of T_n as $n \rightarrow \infty$?
5. Do the centers or the tails converge faster?



6. What does this imply about generating p-values from a Normal approximation to the student- t distribution?

Country demographics

These problems will deal with a data set of country-level statistics from [UNdata](#) and [Varieties of Democracy](#).

1. Compare the following summary statistics for the 2010 populations (in millions of people) of Western African and Eastern African countries:

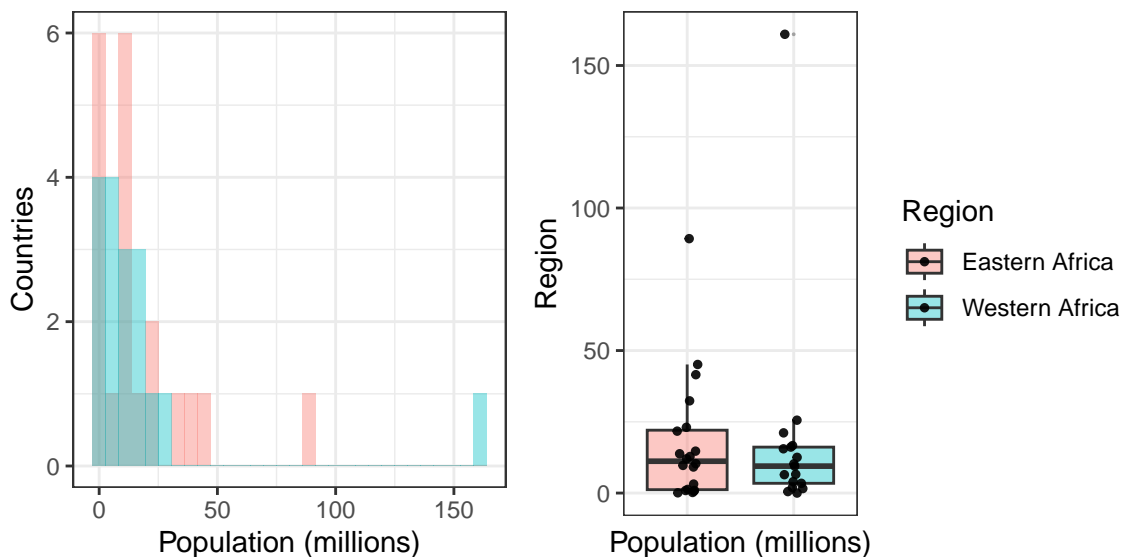
```
# Western Africa
pop1 <- countries[countries$Year == 2010 &
  countries$Region == "Western Africa",
  ]$`Population mid-year estimates (millions)`
round(c(summary(pop1), "SD" = sd(pop1)), 2)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.     SD
##      0.01   3.42   9.45   18.39  16.12  160.95   37.50
```

```
# Eastern Africa
pop2 <- countries[countries$Year == 2010 &
  countries$Region == "Eastern Africa",
  ]$`Population mid-year estimates (millions)`
round(c(summary(pop2), "SD" = sd(pop2)), 2)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.     SD
##      0.09   1.19  11.17   17.14  22.06   89.24   21.66
```

2. Compare the distributions. Would you expect to see a significant difference in a t -test?



3. Varieties of Democracy is a group of researchers that estimates a democracy score for each country each year based on a large compilation of data. Note any trends in the democracy index.

