



# $z$ – or $t$ – distribution

## $z$ – or $t$ – distribution

*Elias Zintzaras, M.Sc., Ph.D.*

*Professor in Biomathematics-Biometry  
Department of Biomathematics  
**School of Medicine**  
**University of Thessaly***

*Institute for Clinical Research and Health Policy Studies  
Tufts University School of Medicine  
Boston, MA, USA*

*Theodoros Mprotsis, MSc, PhD  
Teacher & Research Fellow  
**(<http://biomath.med.uth.gr>)**  
**University of Thessaly**  
Email: [tmprotsis@uth.gr](mailto:tmprotsis@uth.gr)*



$z - t$  formulas

$$\bar{x} \pm z \frac{s}{\sqrt{n}}$$

$$\bar{x} \pm t \frac{s}{\sqrt{n}}$$



# Examples

## Example 1

- We randomly select ten people aged 65 and over (sample) of millions of people in this age group (population of interest)
- Find the mean diastolic blood pressure
- In our paper or report we say «The average diastolic blood pressure is ...»

## Example 2

- We randomly select five students from a high school class
- Find the mean height
- In our paper or report we say «The average height is ...»

These examples sound problematic



## The limits of data in research

- Our primary focus while conducting quantitative research or analysis is typically on large populations
  - The prevalence of depression among people aged 65 or older
  - The average height of students
- However, we almost always use sample data to represent the larger population due to cost and time limitations
- However, sample data is always an estimate or approximation of the whole population that it was taken from



## The limits of data in research

- We are less confident that our sample size  $n$  is representative of our entire population as it gets smaller. The risk of error is greater
- Additionally, we often know nothing about the population: its mean, variance, or standard deviation

Thus, we are using sample data from a larger population about which we have little or no information



## Sample Size

- Therefore, we understand that the larger the sample, the more confident we are that it represents the entire population
- In a large sample, we are more likely to capture the natural variation of the data in the population
- However, there is a point where increasing the sample size further offers no additional statistical benefits

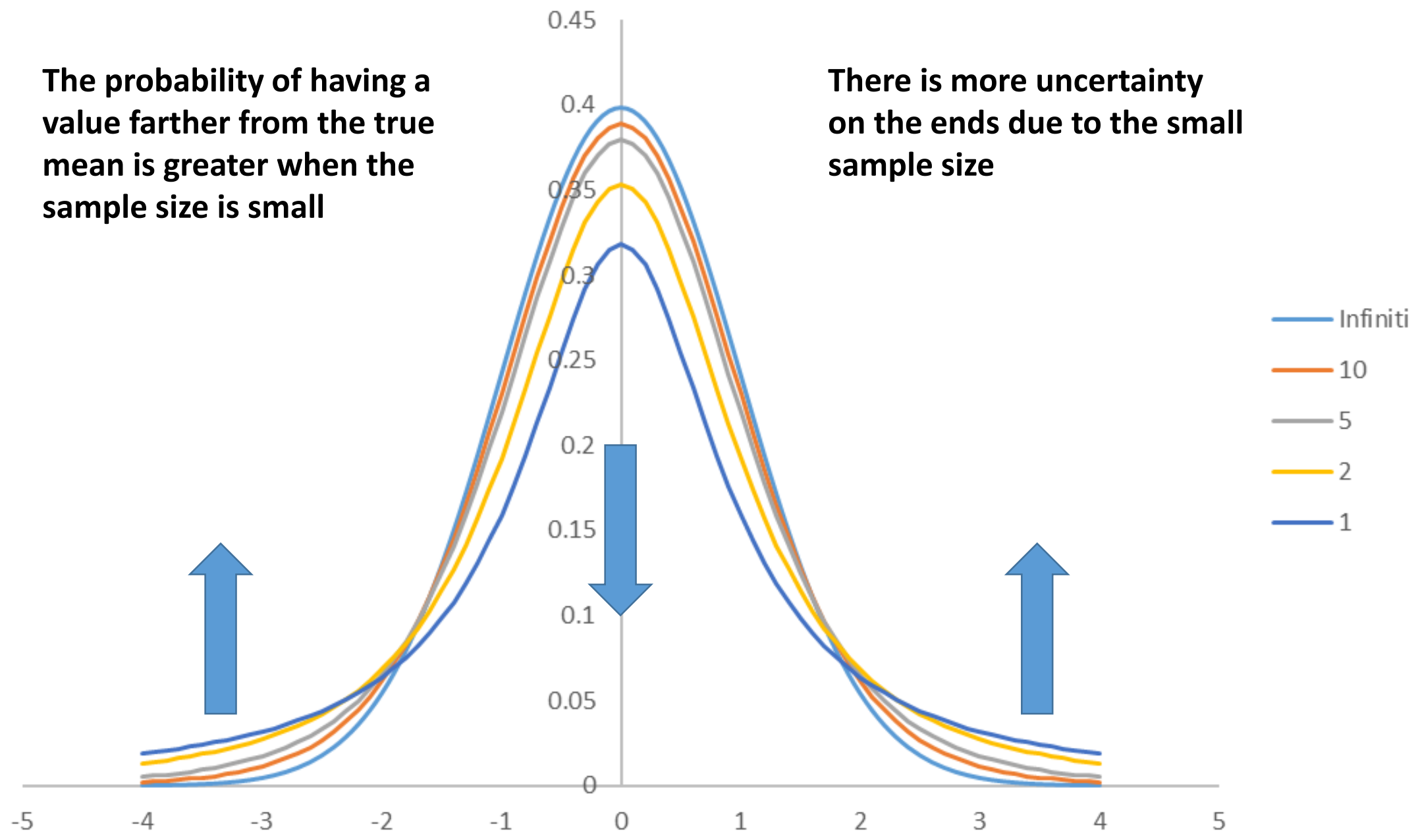


## *t – Distribution(s)*

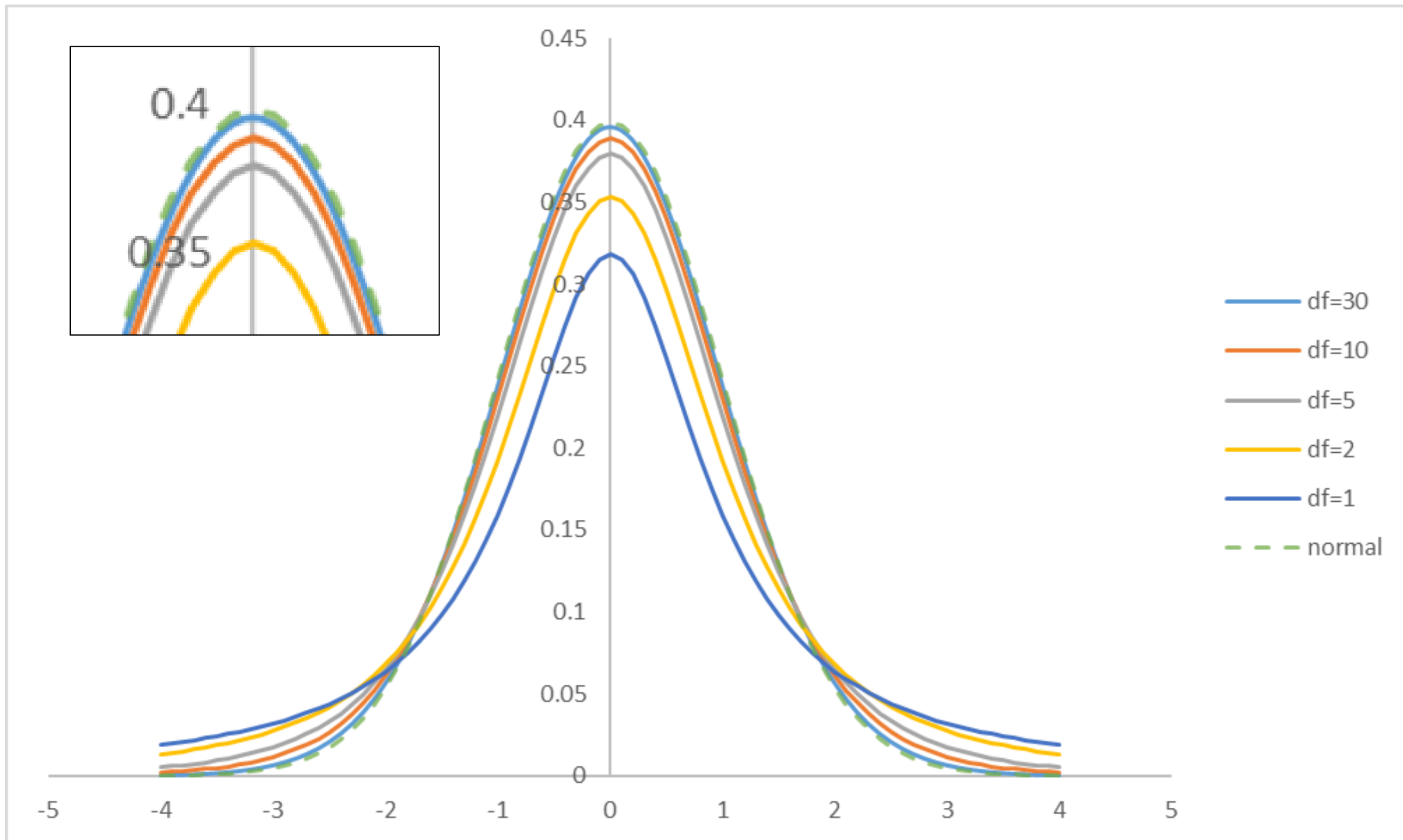
- When the sample size is  $n \leq 30$ , or when the variance or standard deviation of the population is unknown, we use the t-distribution instead of the z-distribution (standard normal distribution)
- What does this allow us to do?
  - The t-distribution allows us to work with small samples ( $n \leq 30$ )
  - However, we sacrifice some certainty in our calculations, as the margin of error increases
  - The t-distribution adjusts for the sample size by using  $n - 1$  degrees of freedom. The t-distribution is different for each sample size
  - The bell curve of the t-distribution is "pulled" down in the middle and "fatter" on the ends compared to the z-distribution. The smaller the sample, the more the distribution is pulled down in the center and fatter at the tails
  - As  $n$  increases  $n > 30$  and especially  $n \geq 100$ , the t-distribution and the z-distribution become similar and eventually hard to tell apart

The probability of having a value farther from the true mean is greater when the sample size is small

There is more uncertainty on the ends due to the small sample size







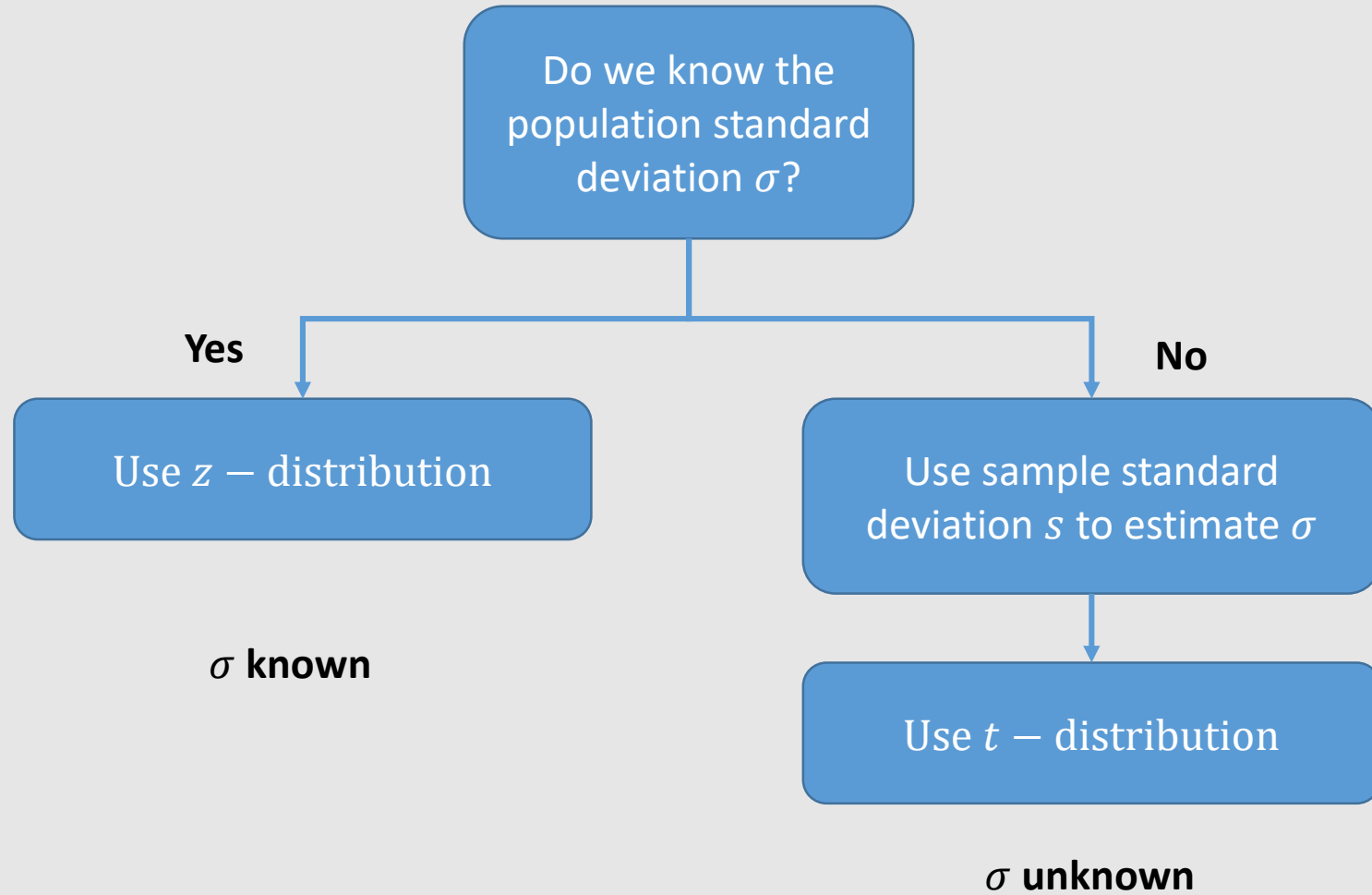


## Degree of Freedom

- Degrees of freedom are simply an adjustment to the sample size, typically  $(n - 1)$  or in some cases  $(n - 2)$  or more
- It is related to the idea that we estimate something for the larger population, most often the population variance or standard deviation
- It results in a slightly larger margin of error in the estimate
- It is a statistical adjustment



# Decision Tree





# Decision Tree

