



# Today's Key Topics

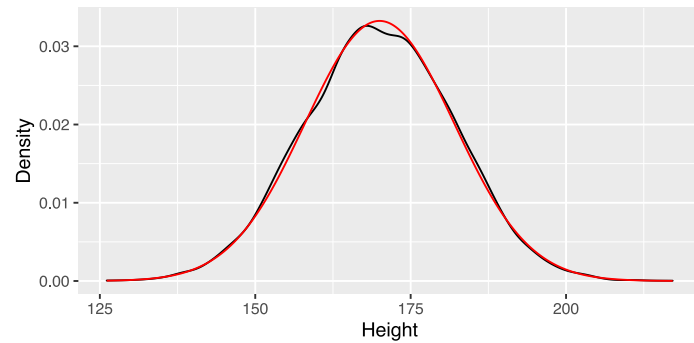
- One-tailed vs Two-tailed Hypotheses
- Null vs Alternative Hypotheses
- The Null Distribution
- z-Scores
- t-tests
- $\alpha$



# More about Height

- Last time we simulated the heights of a population of 10,000 people
  - $\bar{x} = 170$  cm
  - $\sigma = 12$  cm

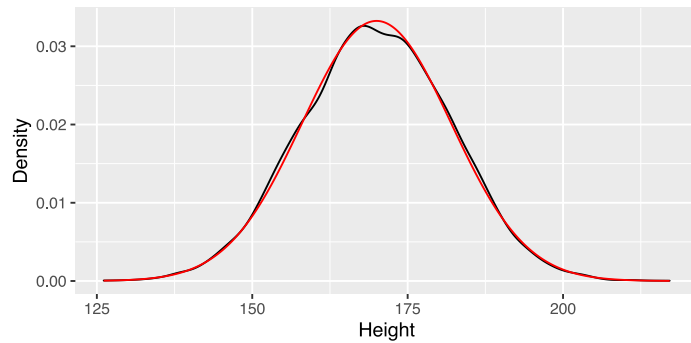
```
## height
## 1 183.4
## 2 168.6
## 3 179.1
## 4 170.1
```



# More about Height

- Last time we simulated the heights of a population of 10,000 people
  - $\bar{x} = 170$  cm
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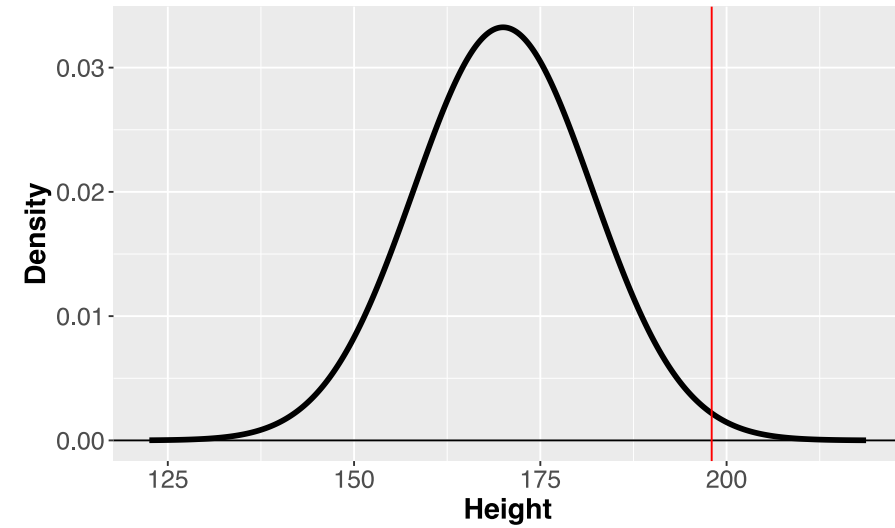


- This time, you'll learn how to compute the probability of randomly observing a specific value within the normal distribution



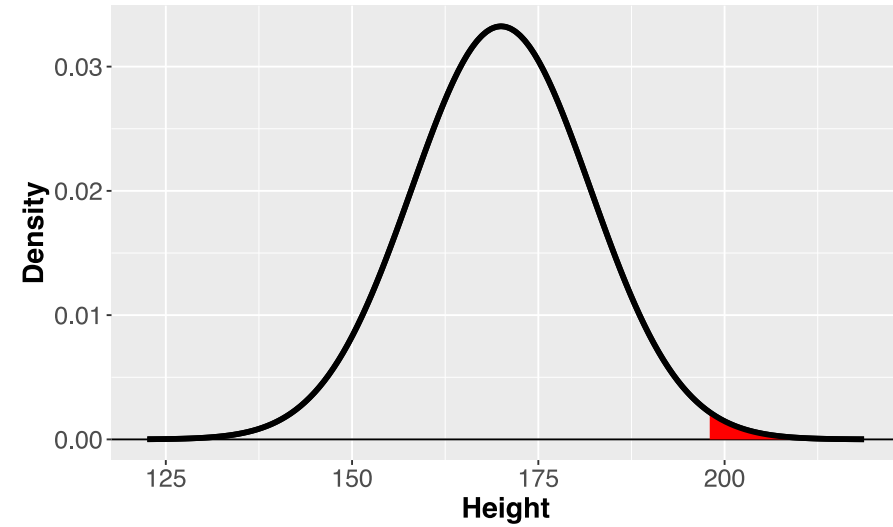
# How Unusual is Casper?

- In his socks, Casper is 198 cm tall
- How likely would we be to find someone Casper's height in our population?



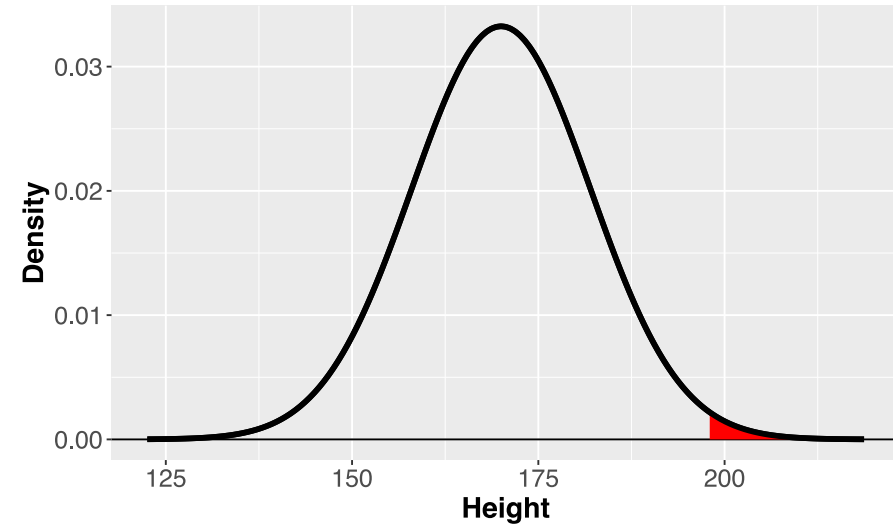
# How Unusual is Casper (Take 2)?

- In his socks, Casper is 198 cm tall
- How likely would we be to find someone Casper's height **or taller** in our population?



# How Unusual is Casper (Take 2)?

- In his socks, Casper is 198 cm tall
- How likely would we be to find someone Casper's height **or taller** in our population?
- The area is 0.0098
- So the probability of finding someone in the population of Casper's height or greater is 0.0098 (or,  $p = 0.0098$ )



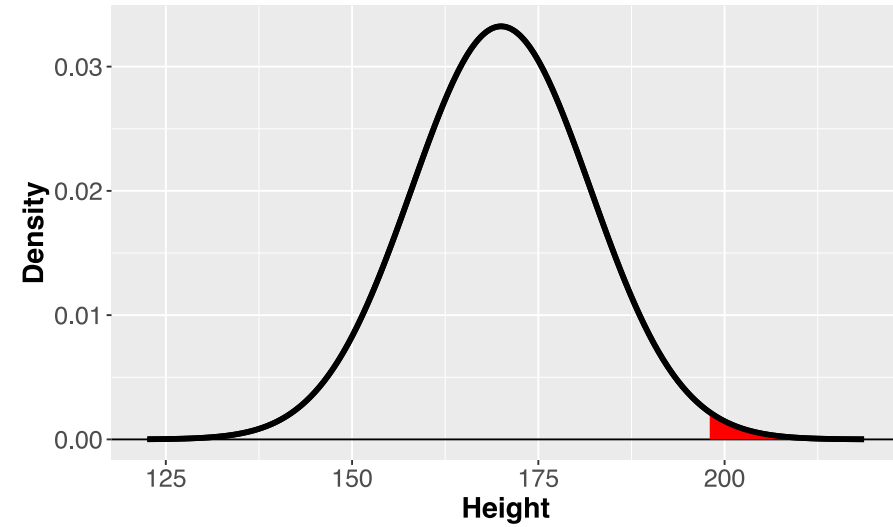


# Area under the Curve

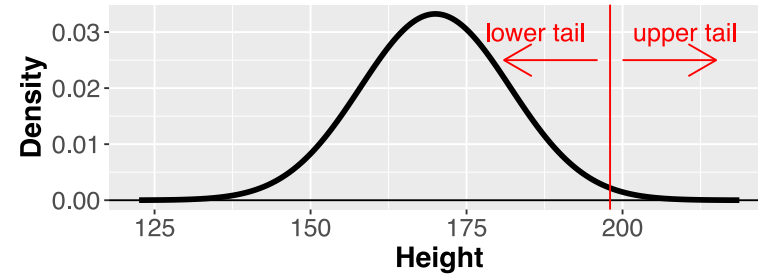
- So now we know that the area under the curve can be used to quantify **probability**
- But how do we calculate area under the curve?
- Luckily, R has us covered, using (in this case) the **pnorm()** function

```
pnorm(198, mean = 170, sd=12,  
      lower.tail = FALSE)
```

```
## [1] 0.009815
```



# Area under the Curve



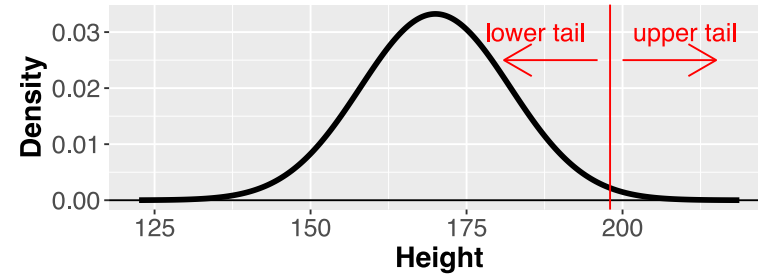
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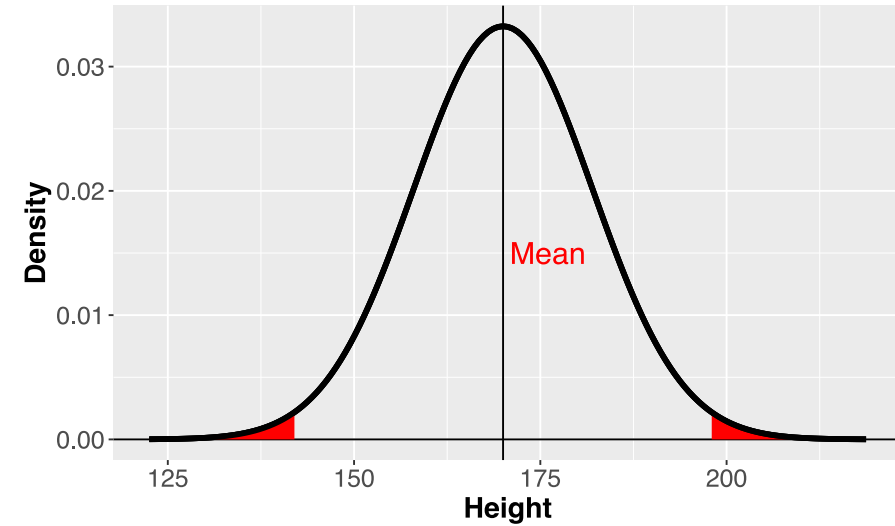
```
## [1] 1
```

# Tailedness

- In this example, we kind of knew that Casper was *tall*
  - It made sense to ask what the likelihood of finding someone 198 cm or greater was
  - This is called a **one-tailed hypothesis** (we're not expecting Casper to be well *below* average height!)

# Tailedness

- In this example, we kind of knew that Casper was *tall*
  - It made sense to ask what the likelihood of finding someone 198 cm or greater was
  - This is called a **one-tailed hypothesis**
- Often our hypothesis might be vaguer
  - We expect Casper to be "*different*", but we're not sure how
  - In this case, we would make a non-directional, or **two-tailed hypothesis**

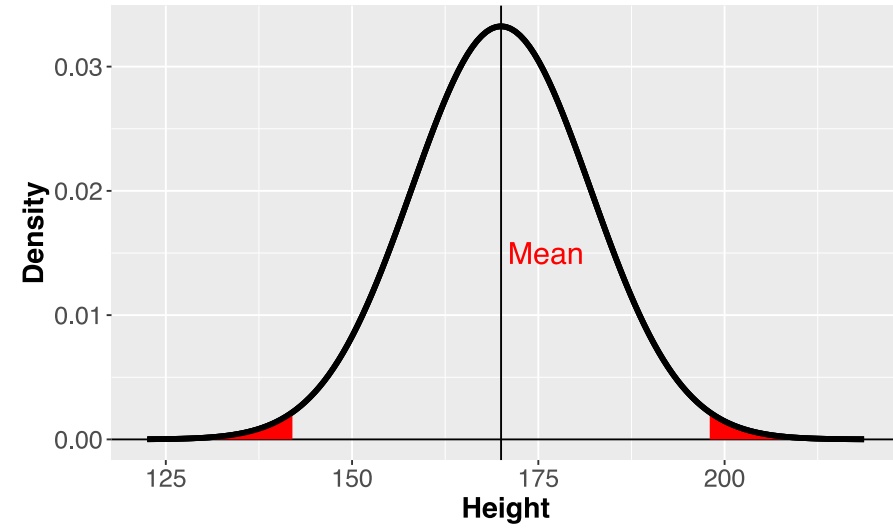


# Tailedness

- For a two-tailed hypothesis we need to sum the relevant upper and lower areas:

```
2 * pnorm(198, 170, 12, lower.tail = FALSE)
```

```
## [1] 0.01963
```



# So: Is Casper Special?

- How surprised should we be that Casper is 198 cm tall?
- Given the population he's in, the probability that he's 28cm or more taller than the mean of 170 is 0.0098
  - (Keep in mind, this is according to a *one-tailed hypothesis*)

# So: Is Casper Special?

- How surprised should we be that Casper is 198 cm tall?
- Given the population he's in, the probability that he's 28cm or more taller than the mean of 170 is 0.0098
  - (Keep in mind, this is according to a *one-tailed hypothesis*)
- A more accurate way of saying this is that 0.0098 is the probability of selecting him (or someone even taller than him) from the population at random
  - There is about a 1% chance of selecting someone Casper's size or taller from the population.



# A Judgement Call

We have to decide:

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If a 1% probability is *small enough*



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If a 1% chance doesn't impress us much



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Note that, in either case, we have nothing (mathematical) to say about the *reasons* for Casper's height



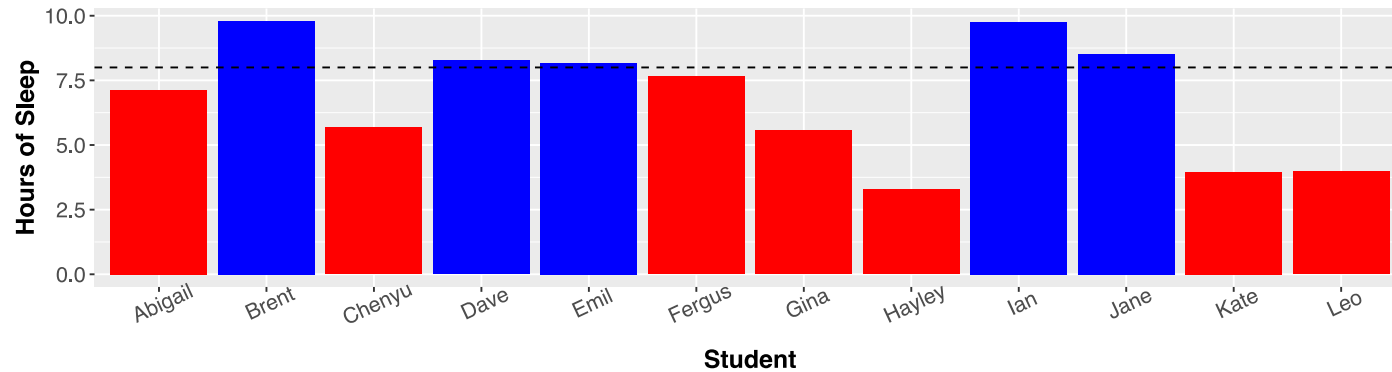


# Sleeping Guidelines

The USMR instructors are concerned that university students are not following the recommended sleep guidelines of 8 hours per night, and worry this could affect their academic performance. Is this idea worth further investigation?

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# Information About our Study

- **Null Hypothesis** ( $H_0$ )
  - Students are getting the recommended amount of sleep
- **Alternative Hypothesis** ( $H_1$ )
  - Students are getting less than the recommended amount of sleep
- There are 12 students

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summary(m)
```

```
##      sleep      names
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## 1st Qu.:5.16  Brent  :1
## Median :7.38  Chenyu :1
## Mean   :6.81  Dave   :1
## 3rd Qu.:8.33  Emil   :1
## Max.   :9.79  Fergus :1
##                               (Other):6
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sd(m$sleep)
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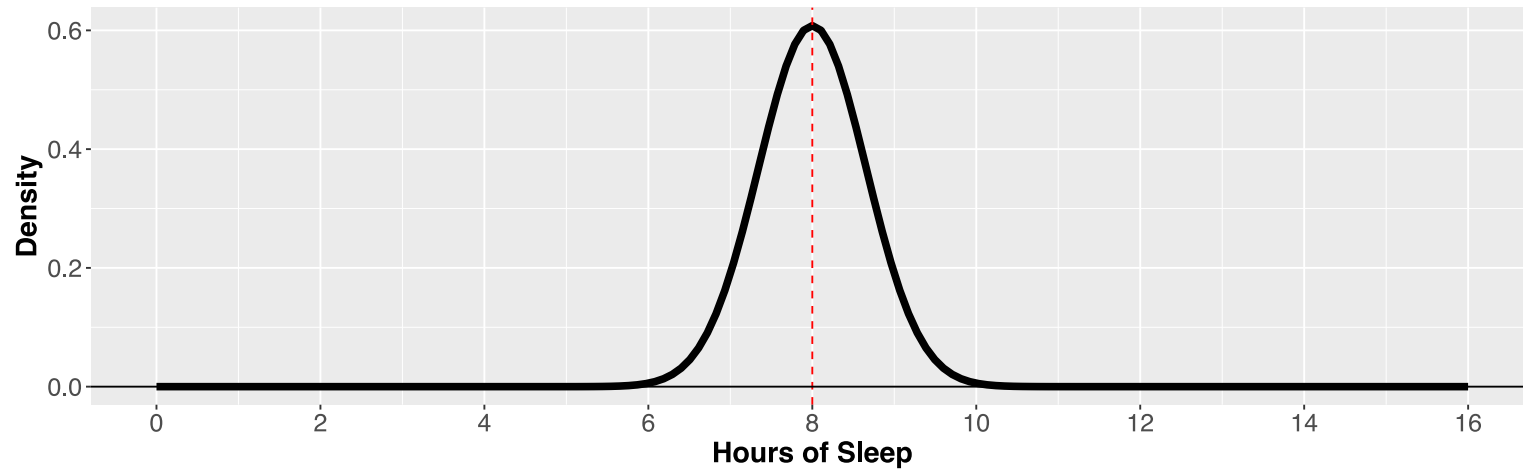
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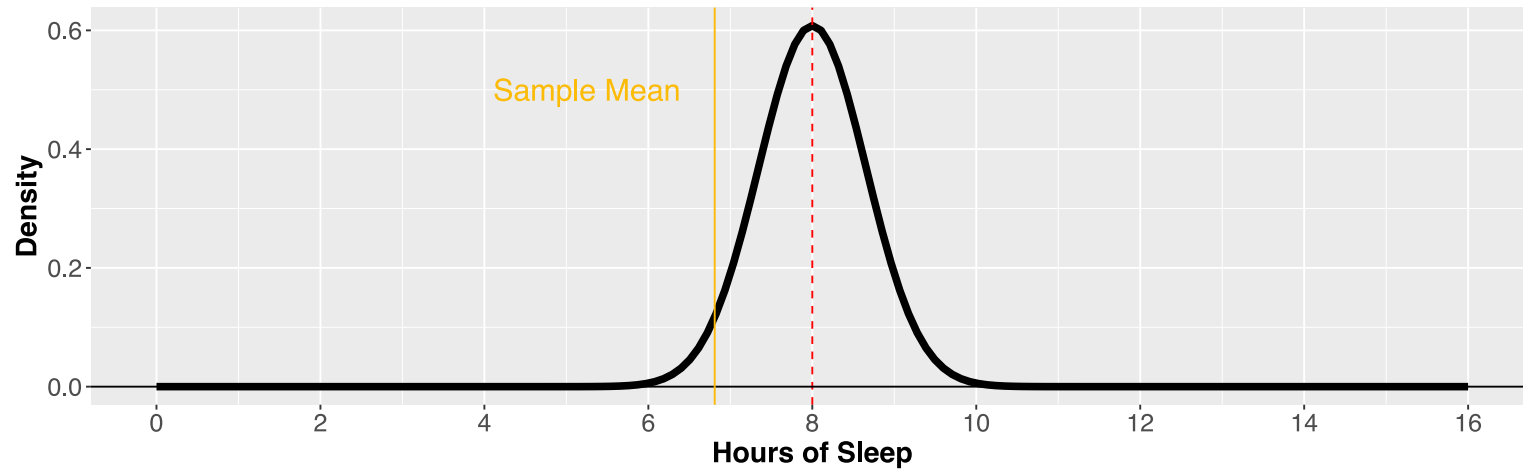
What is the probability of a group of 12 people getting a mean of 6.81 hours of sleep, given that most adults need around 8? (assuming they came from the same population)

# Back to the Normal Distribution



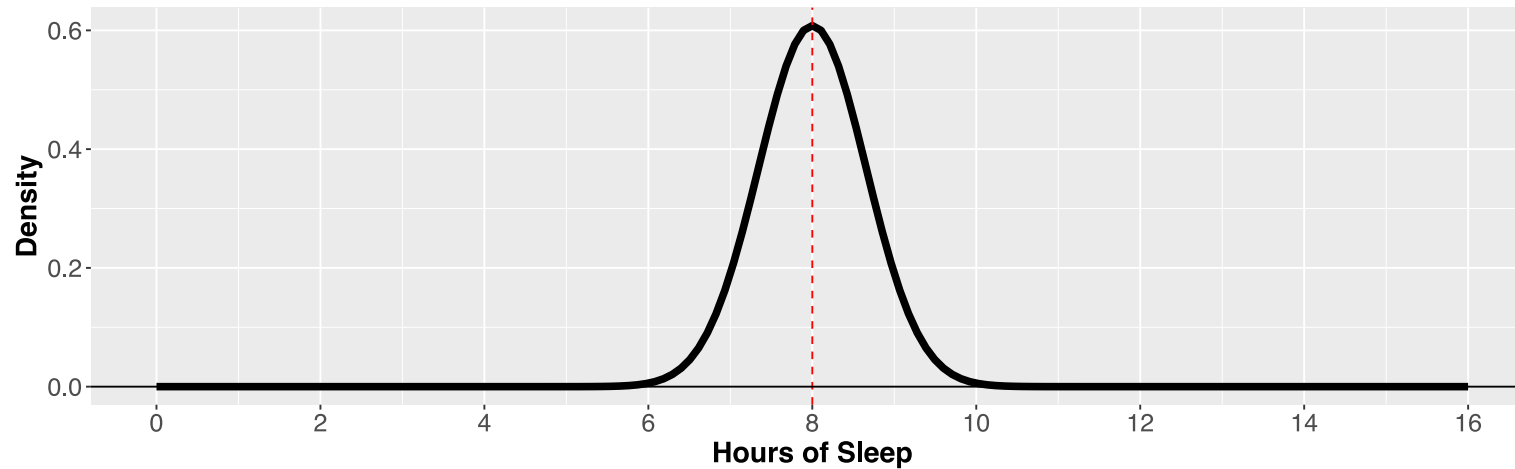
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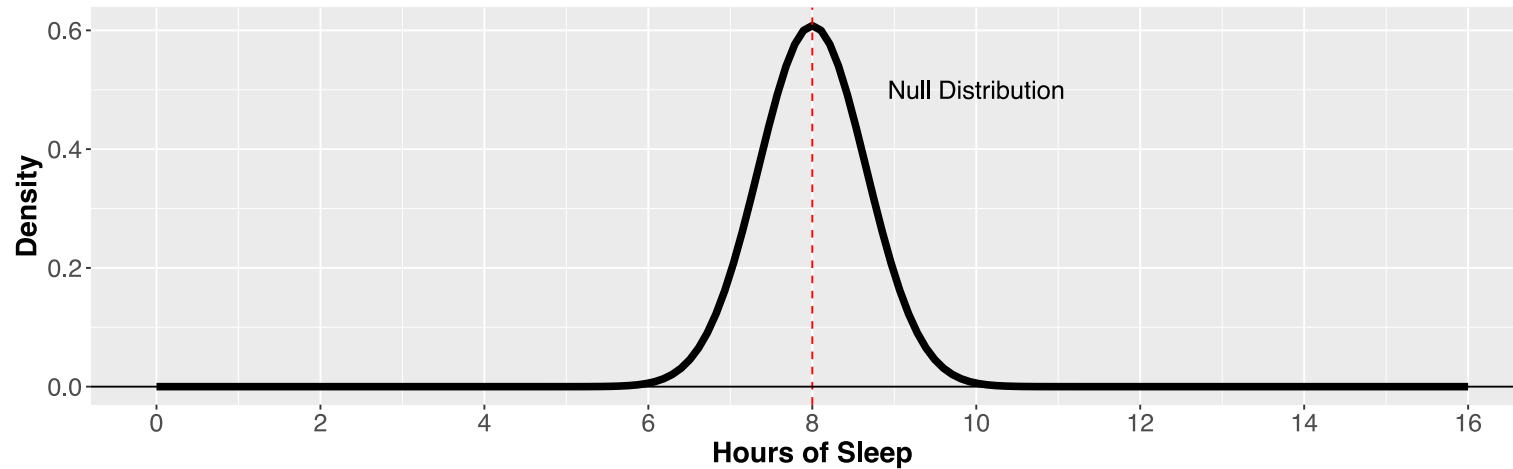
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# The Null Distribution



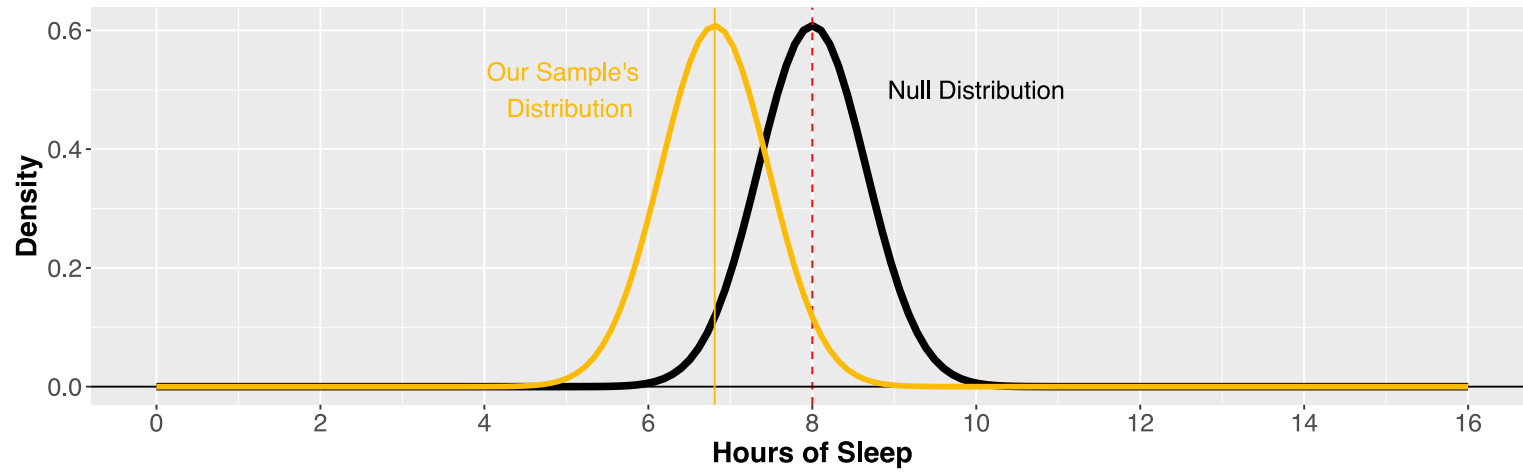
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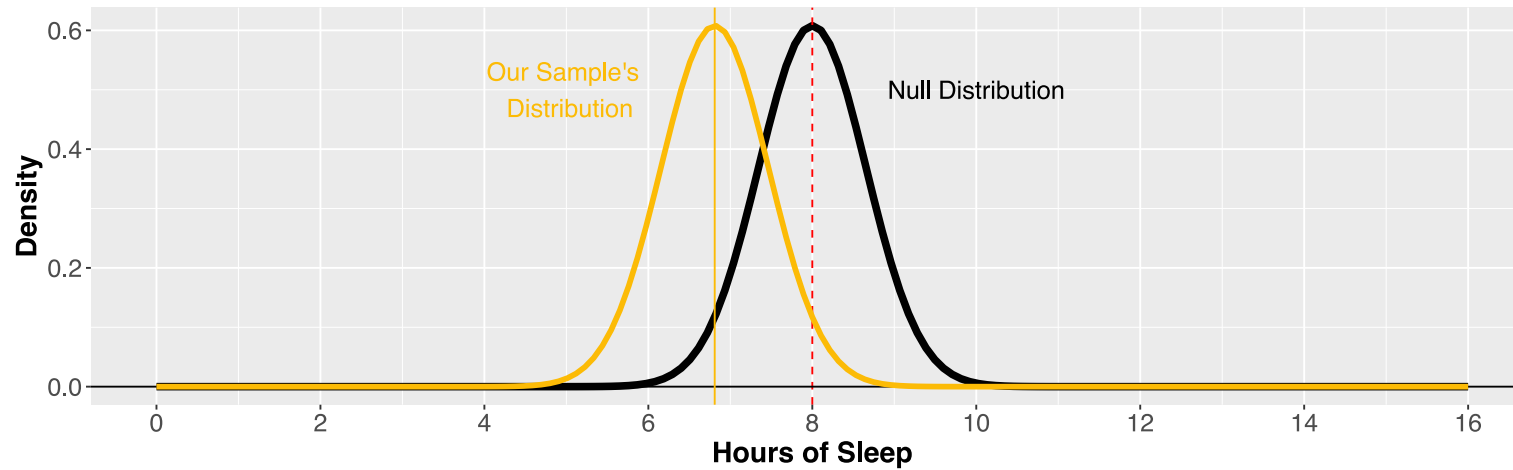


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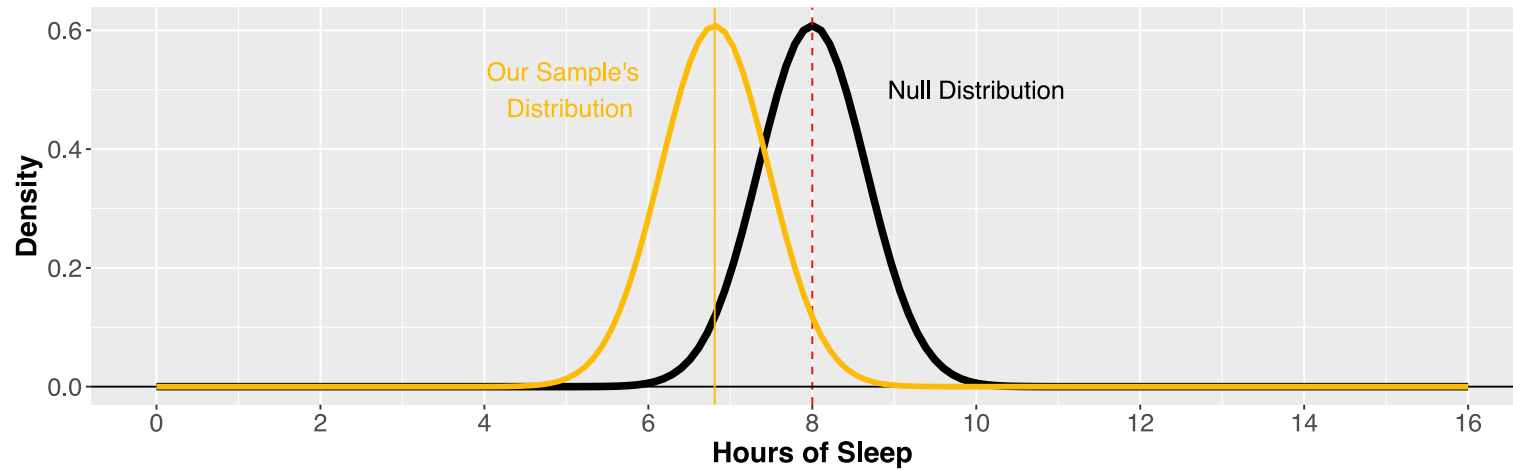
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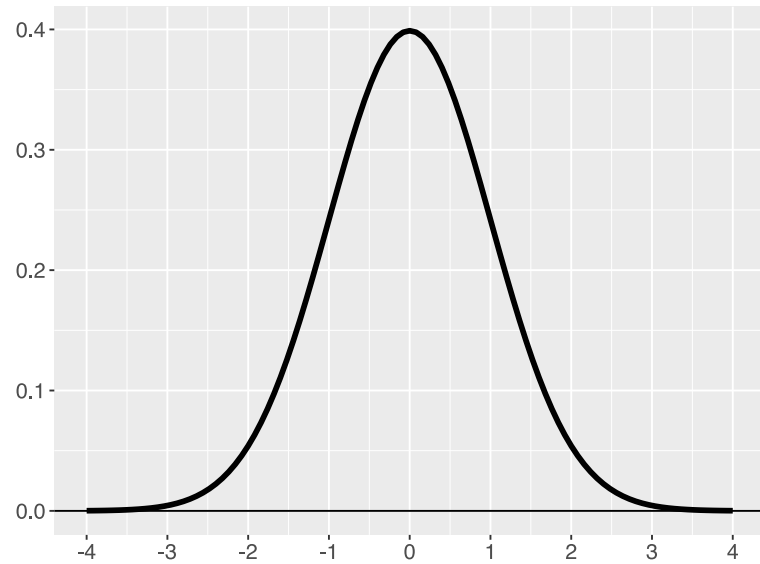
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- From this, it *looks* as though our students may be getting less sleep than they should.
- However, simply seeing a shift in the curves isn't enough evidence to make that claim with certainty.

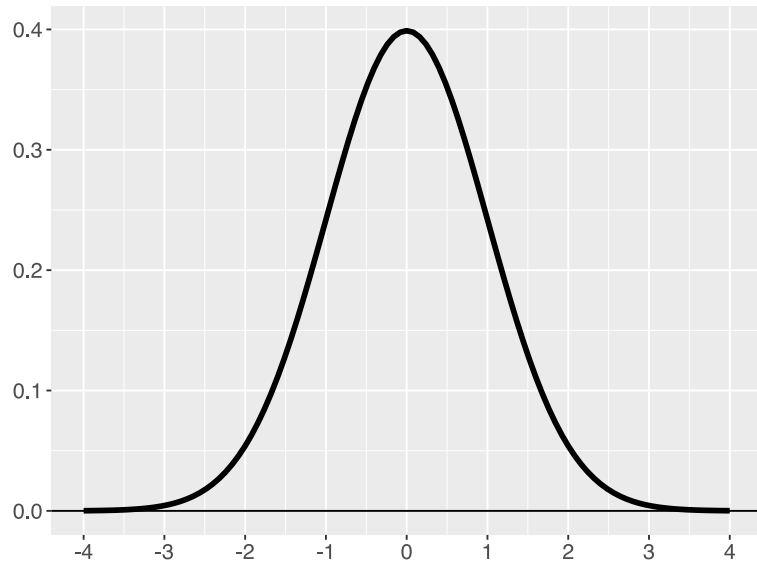
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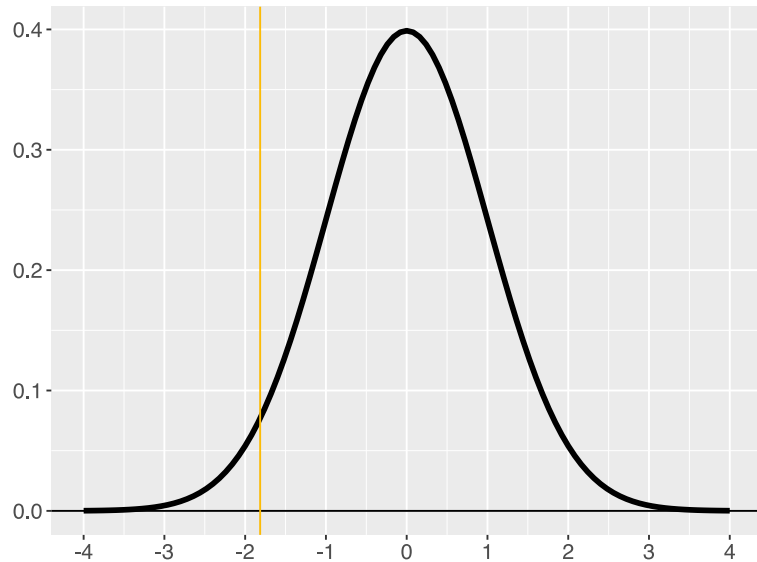


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$$z = \frac{\bar{x} - \mu}{(\sigma / \sqrt{n})}$$

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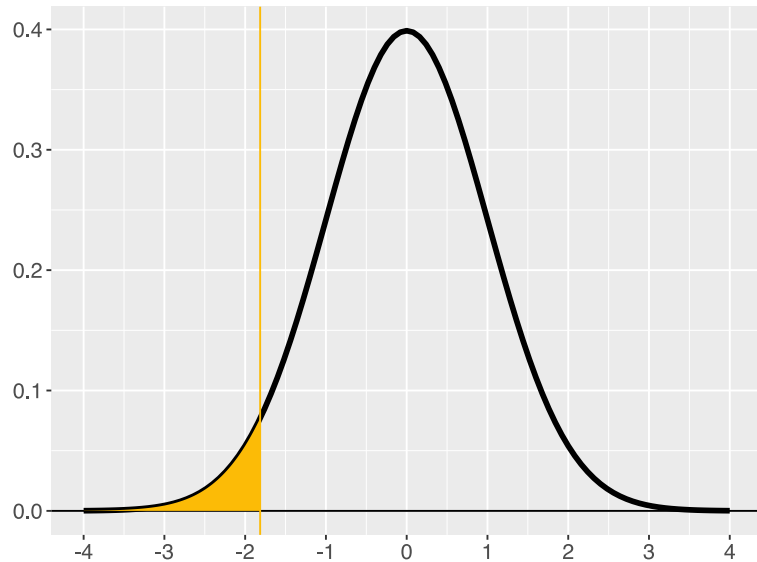
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```
(mean(m$sleep) - 8) / (sd(m$sleep) / sqrt(12))
```

```
## [1] -1.814
```

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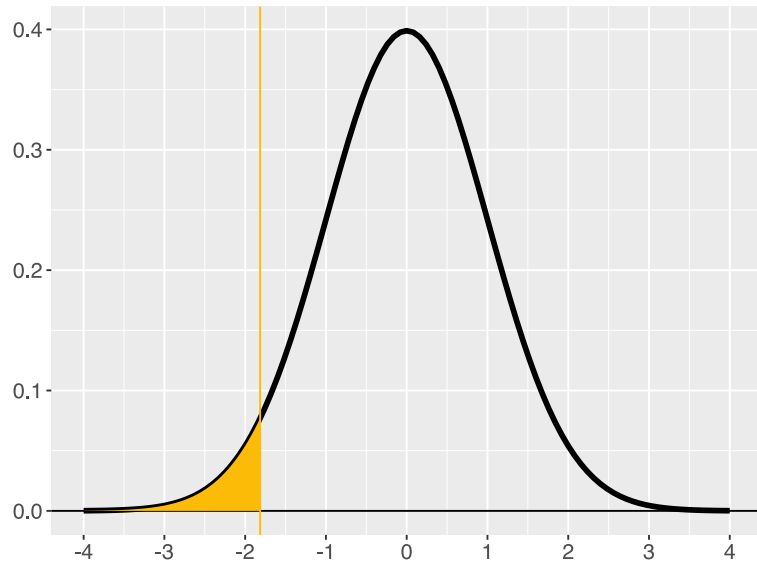
```
## [1] -1.814
```

```
pnorm(-1.814, mean = 0, sd = 1)
```

```
## [1] 0.03484
```

# Back to the Normal Distribution...Again

We can compute the probability of a score's occurrence if it is part of a standardized normal distribution ( $\mu = 0, \sigma = 1$ )



If you picked 12 people at random from a population of people who get the recommended number of hours of sleep, there would be a 3% chance that their average sleep would be 6.81 hours or less







# A Small Confession



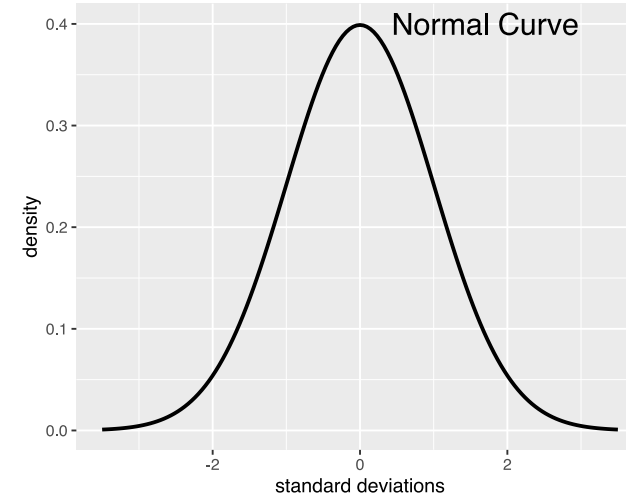
## Part Two wasn't entirely true

- All of the principles are correct, but for smaller  $n$  the normal curve isn't the best estimate
- For that we use the  $t$  distribution

# The $t$ Distribution



"A. Student", or William Sealy Gossett

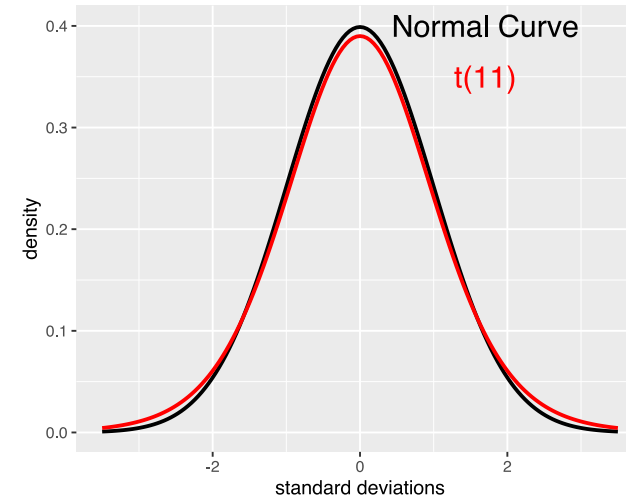


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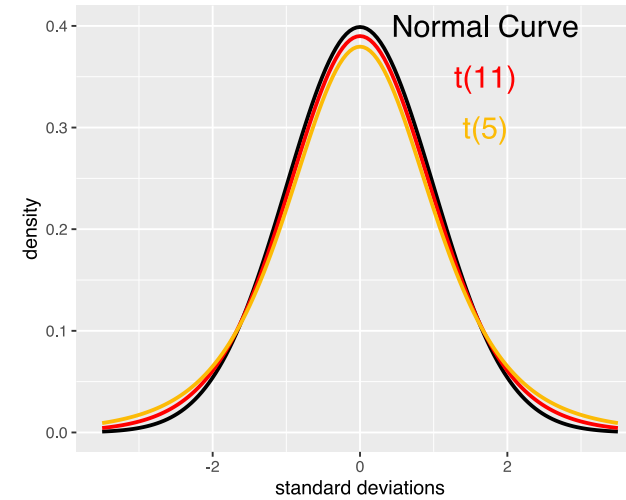


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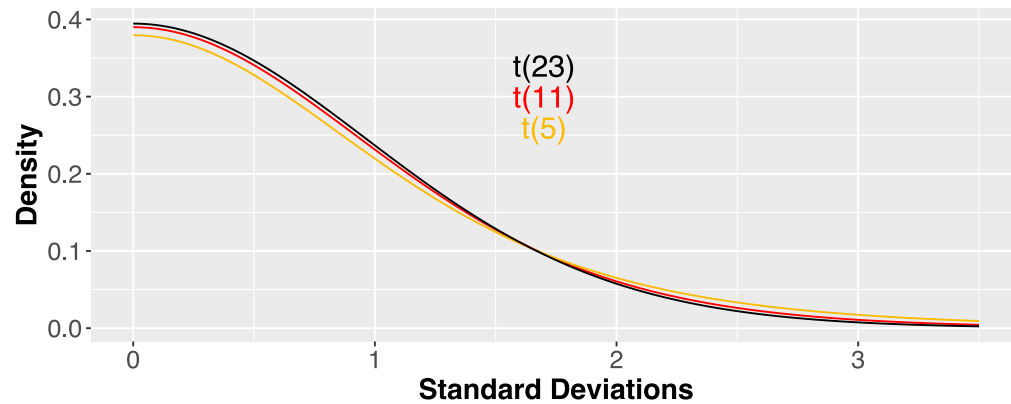
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# The $t$ Distribution

- Conceptually, the  $t$  distribution increases uncertainty when the sample is small
  - The probability of more extreme values is slightly higher
- Exact shape of distribution depends on sample size



# Using the t Distribution

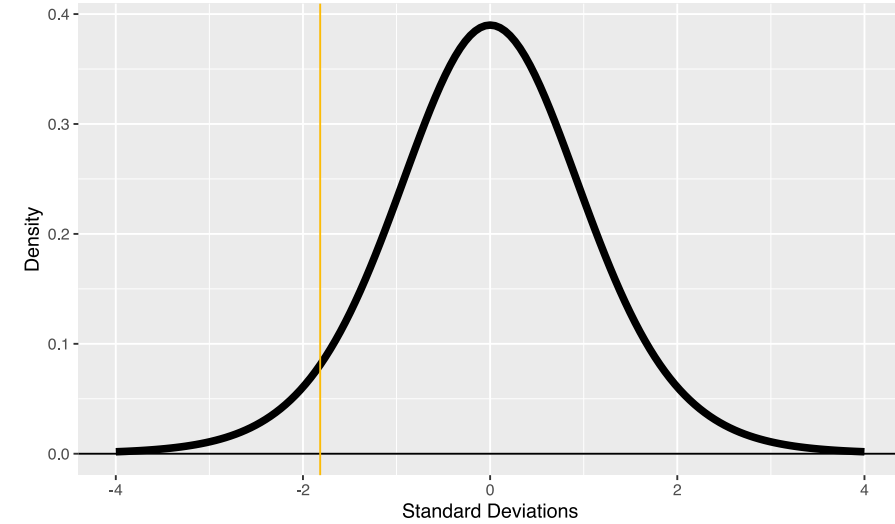
- In part 2, we calculated the mean hours of sleep for the group as 6.81
- We used the formula  $z = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$  to calculate  $z$ , and the standard normal curve to calculate probability

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- **The formula for a one-sample  $t$ -test is the same as the formula for  $z$** 
  - What differs is the *distribution we are using to calculate probability*
  - We need to know the degrees of freedom (to get the right  $t$ -curve)
- so  $t(\text{df}) = \frac{\bar{x} - \mu}{\sigma / \sqrt{n}}$

# Probability According to t

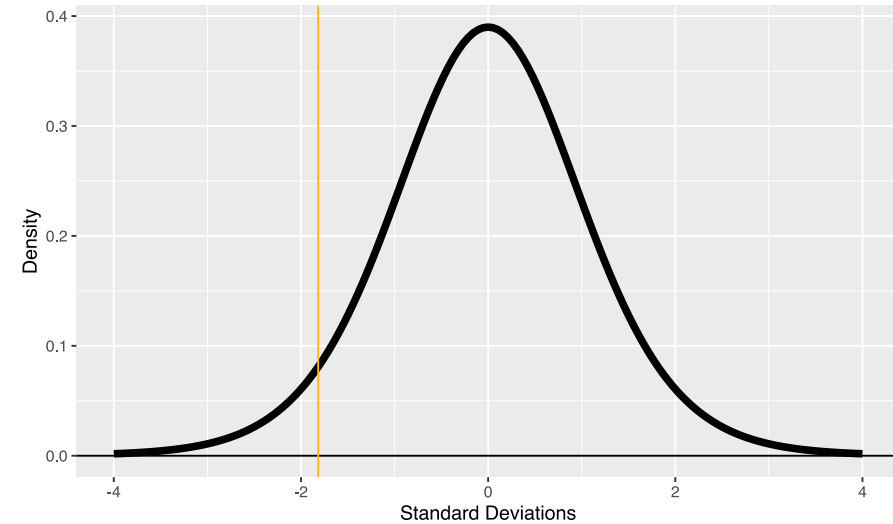
- for 12 people who got a mean 6.81 hours of sleep with a sd of 2.2732
- $t(11) = \frac{6.81-8}{2.27/\sqrt{12}} = -1.816$





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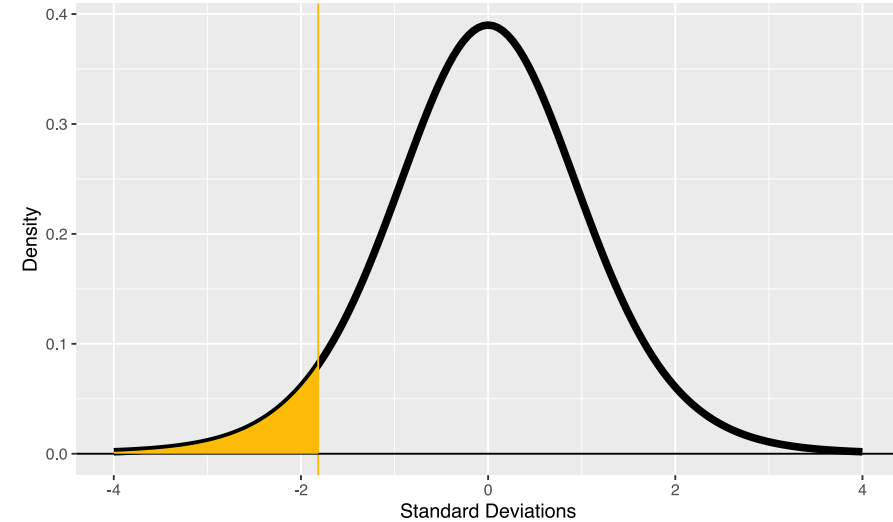


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- instead of `pnorm()` we use `pt()` for the  $t$  distribution
- `pt()` requires the degrees of freedom:

```
pt(-1.814, df=11, lower.tail = TRUE)
```

```
## [1] 0.04851
```



Did We Have to Do All That Work?

# Did We Have to Do All That Work?

No.

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```
head(m$sleep)
```

```
## [1] 7.122 9.791 5.674 8.264 8.176 7.647
```

```
t.test(m$sleep, mu=8, alternative = "less")
```

```
##  
##      One Sample t-test  
##  
## data:  m$sleep  
## t = -1.8, df = 11, p-value = 0.05  
## alternative hypothesis: true mean is less than 8  
## 95 percent confidence interval:  
##      -Inf 7.988  
## sample estimates:  
## mean of x  
##      6.809
```

- **One-sample  $t$ -test**
- Compares a single sample against a hypothetical mean ( $\mu$ )

# Types of Hypothesis

```
t.test(m$sleep, mu=0, alternative = "less")
```

- Note the use of `alternative="less"`
- This refers to the direction of our **alternative hypothesis,  $H_1$** 
  - $H_1$  is that our students would be getting *less* sleep than the average person.
- Can also have `alternative="greater"`...

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- Can also have `alternative="greater"`...
  - Our students are getting *more* sleep than the average person
- ...And `alternative="two.sided"`
  - Our students are getting *different* amounts of sleep than the average person

# Putting it Together

For  $t(11) = -1.816, p = 0.0483$ :

If you picked 12 people at random from a population of people who get the recommended number of hours of sleep, there would be a 5% chance that their average sleep would be 6.81 hours or less

# Putting it Together

For  $t(11) = -1.816, p = 0.0483$ :

If you picked 12 people at random from a population of people who get the recommended number of hours of sleep, there would be a 5% chance that their average sleep would be 6.81 hours or less

- Is 5% low enough for you to believe that the mean sleep probably wasn't due to chance?
- Perhaps we'd better face up to this question!

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- Typically, in Psychology,  $\alpha$  is set to .05
  - We're willing to take a 5% risk of incorrectly rejecting the null hypothesis.

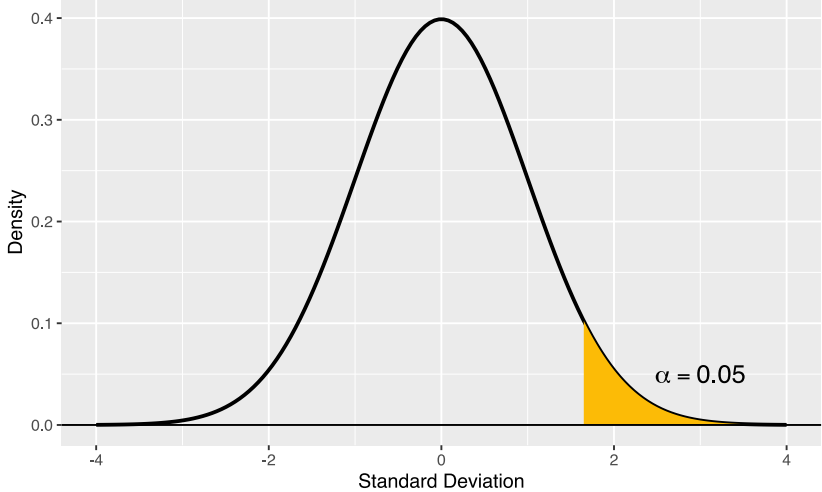
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- Typically, in Psychology,  $\alpha$  is set to .05
  - We're willing to take a 5% risk of incorrectly rejecting the null hypothesis.
- It's important to set  $\alpha$  *before* any statistical analysis

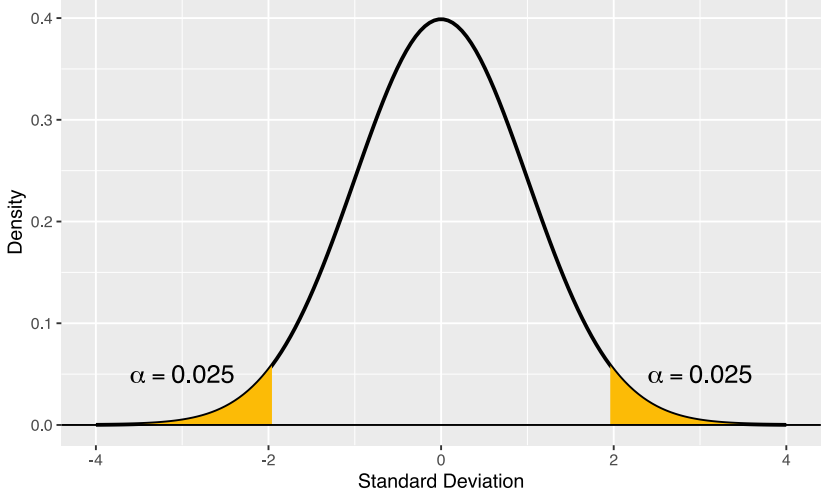


# Making a Decision

One-Tailed



Two-Tailed



$$p < .05$$

- The  $p$ -value is the probability of finding our results under  $H_0$ , the null hypothesis
- $H_0$  is essentially "🐛 happens"
- $\alpha$  is the maximum level of  $p$  at which we are prepared to conclude that  $H_0$  is false (and argue for  $H_1$ )

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there is a 5% probability of falsely rejecting  $H_0$

- Wrongly rejecting  $H_0$  (false positive) is a **type 1 error**
- Wrongly failing to reject  $H_0$  (false negative) is a **type 2 error**

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  - Compares the mean from a range of scores to a specific value

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  - Compares the means of two independent groups
  - Lets you examine whether two groups differ significantly from each other on the variable of interest
- **Paired-samples  $t$ -test**
  - Compares means that are paired in some way
  - Allows you to compare measures that come from the same individual, e.g.

