

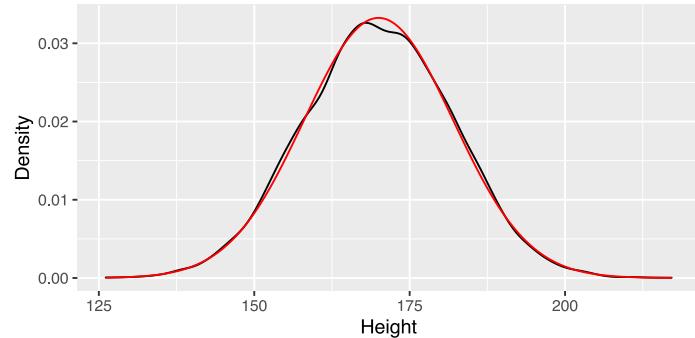
Today's Key Topics

- One-tailed vs Two-tailed Hypotheses
- Null vs Alternative Hypotheses
- The Null Distribution
- z-Scores
- t-tests
- α

More about Height

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

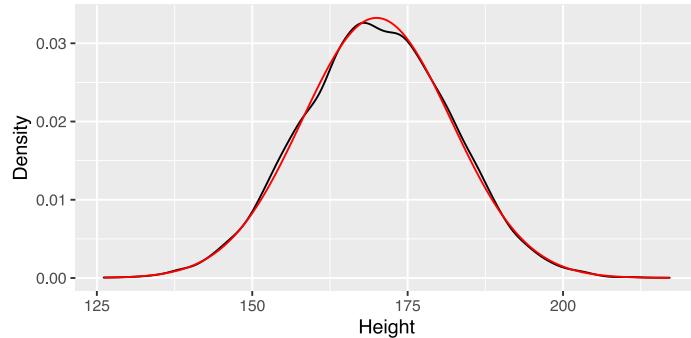
```
##   height
## 1 183.4
## 2 168.6
## 3 179.1
## 4 170.1
```



More about Height

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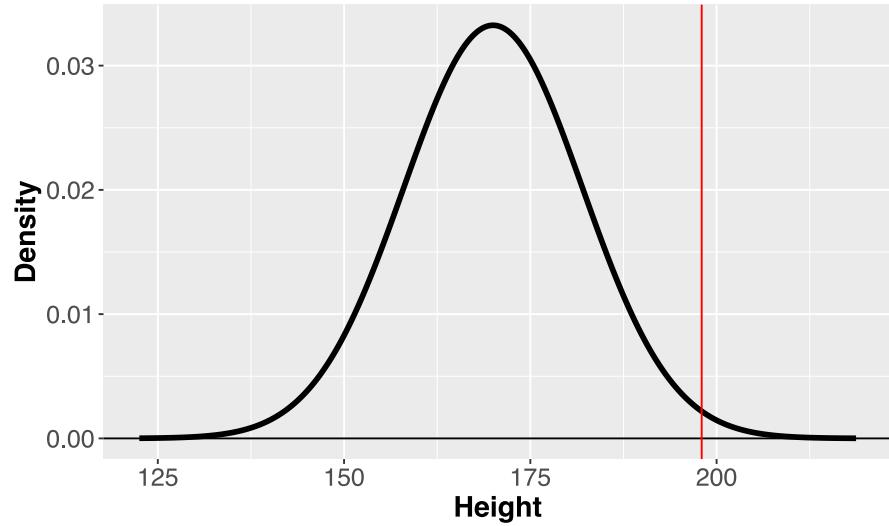


- This time, you'll learn how to compute the probability of randomly observing a specific value within the normal distribution



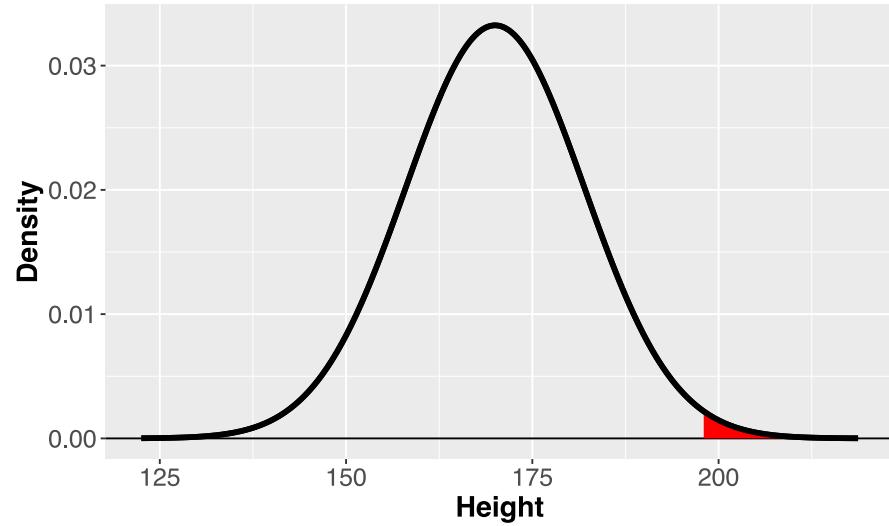
How Unusual is Casper?

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



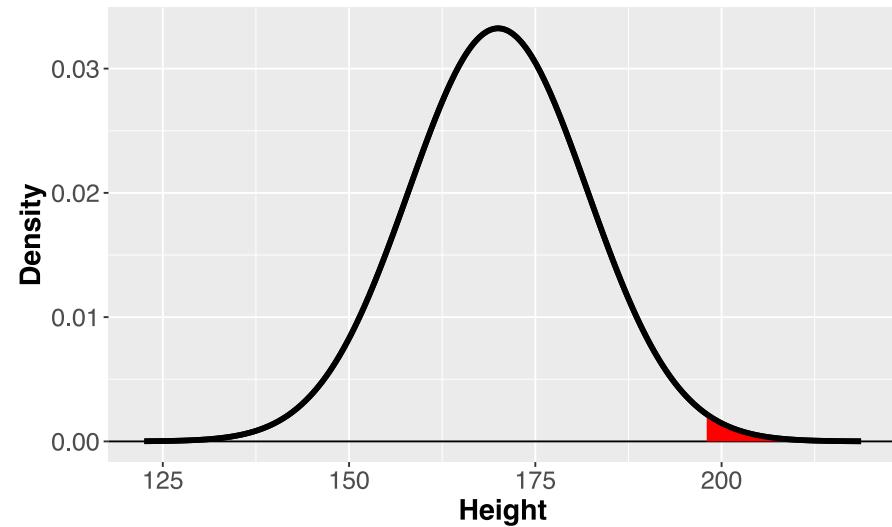
How Unusual is Casper (Take 2)?

- 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, $p = 0.0098$)

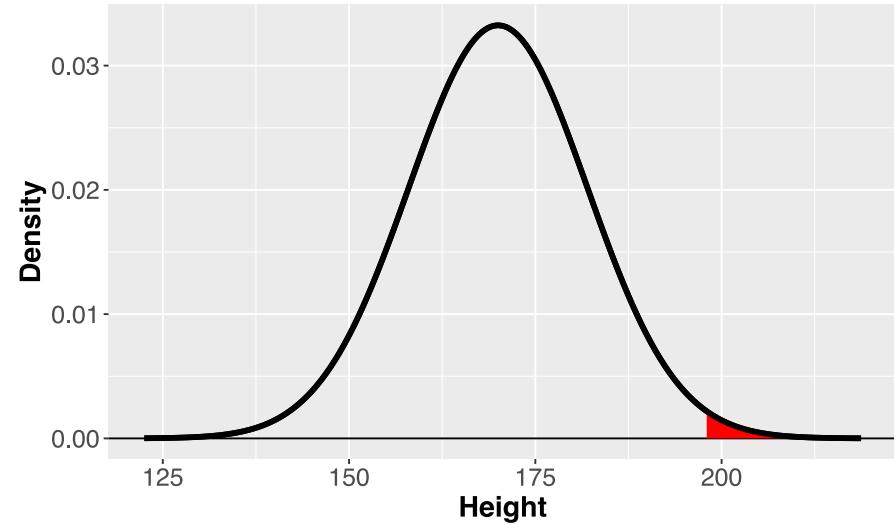


Area under the Curve

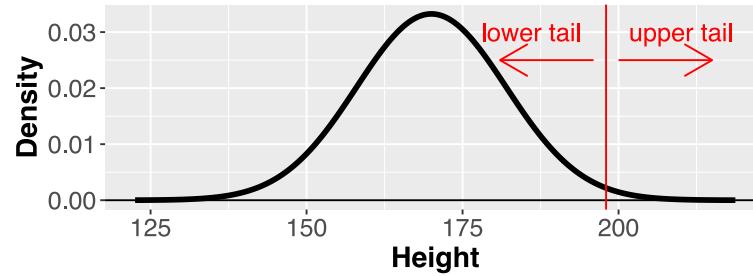
- So now we know that the area under the curve can be used to quantify **probability**
- But how do we calculate area under the curve?
- Luckily, R has us covered, using (in this case) the **pnorm()** function

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

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



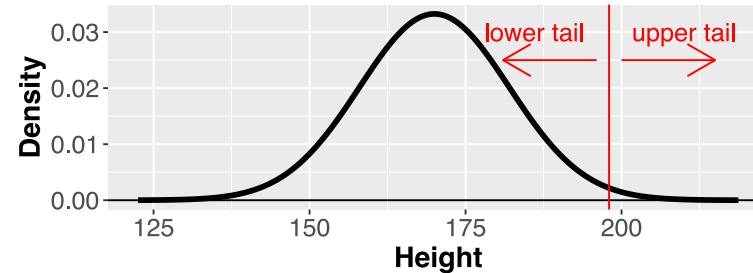
Area under the Curve



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pnorm(198, mean = 170, sd=12,  
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## [1] 0.9902
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Area under the Curve



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

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

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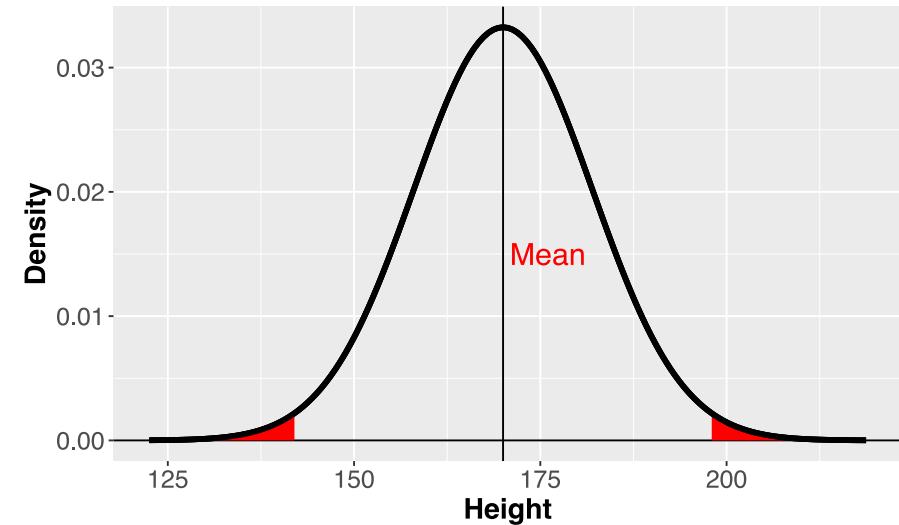
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Tailedness

- 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

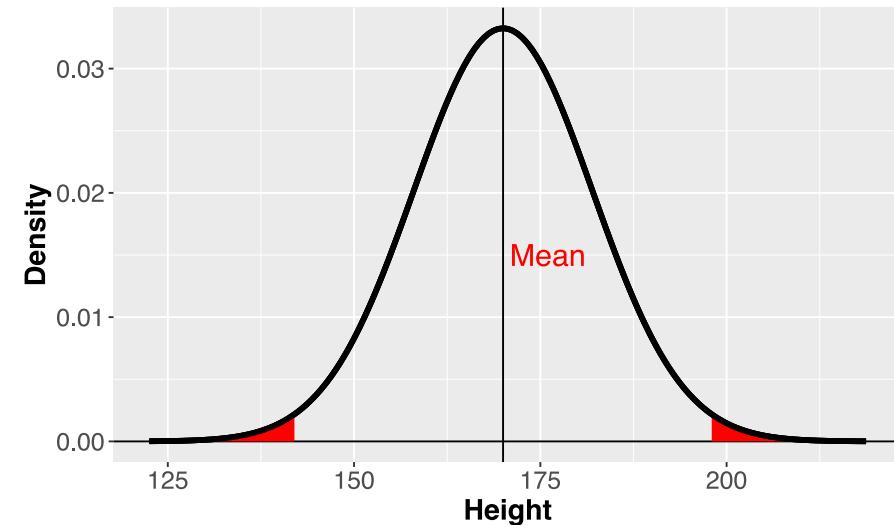
- In this example, we kind of knew that Casper was *tall*
 - It made sense to ask what the likelihood of finding someone 198 cm *or greater* was
 - This is called a **one-tailed hypothesis**
- Often our hypothesis might be vaguer
 - We expect Casper to be "*different*", but we're not sure how
 - In this 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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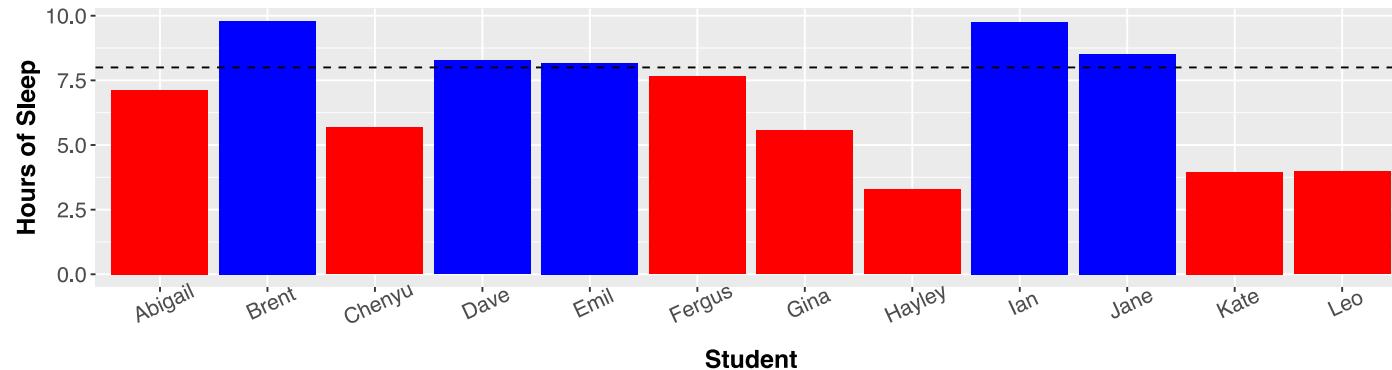
Note that, in either case, we have nothing (mathematical) to say about the *reasons* for Casper's height

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

```
summary(m)
```

```
##      sleep      names
##  Min.   :3.28   Abigail:1
##  1st Qu.:5.16  Brent   :1
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##  Mean   :6.81  Dave    :1
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##                (Other):6
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sd(m$sleep)
```

```
## [1] 2.273
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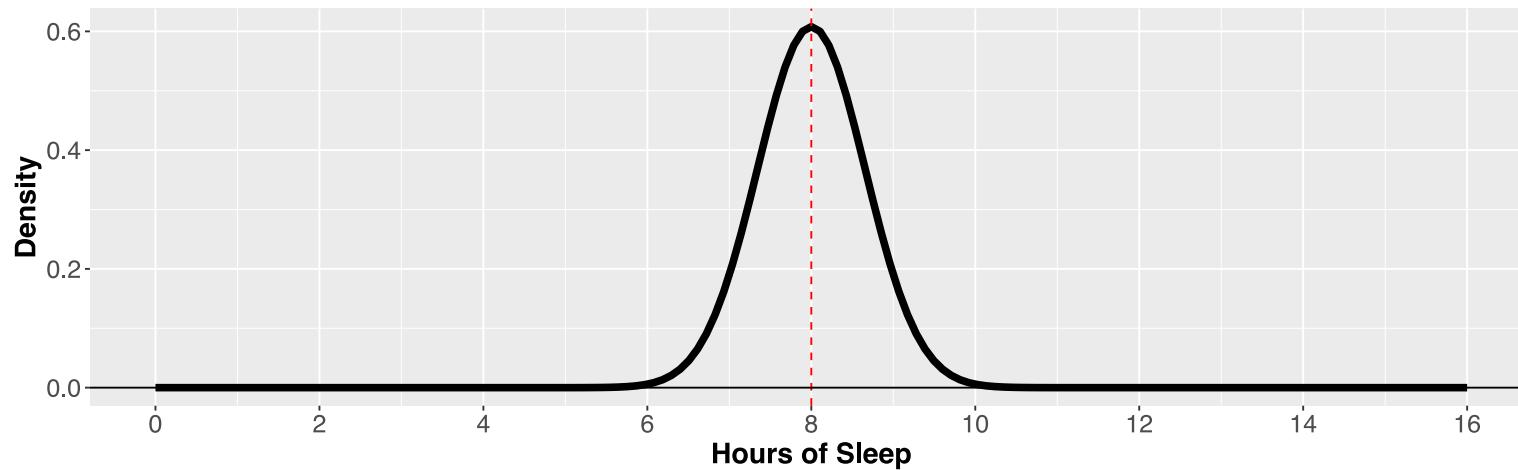
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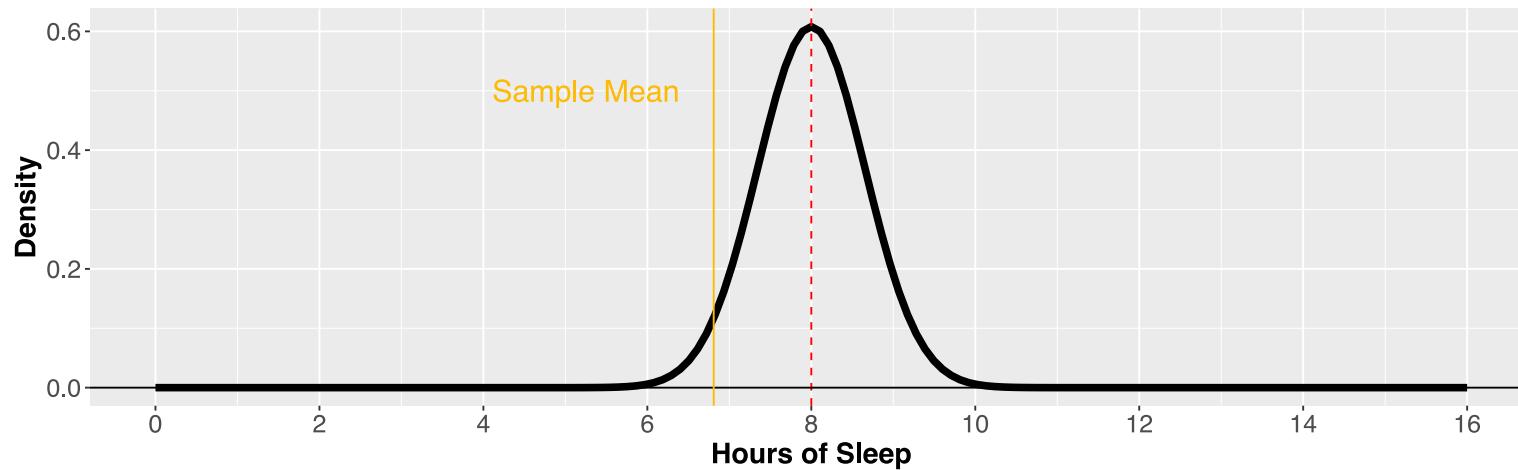
What is the probability of a group of 12 people getting a mean of 6.81 hours of sleep, given that most adults need around 8? (assuming they came from the same population)

Back to the Normal Distribution



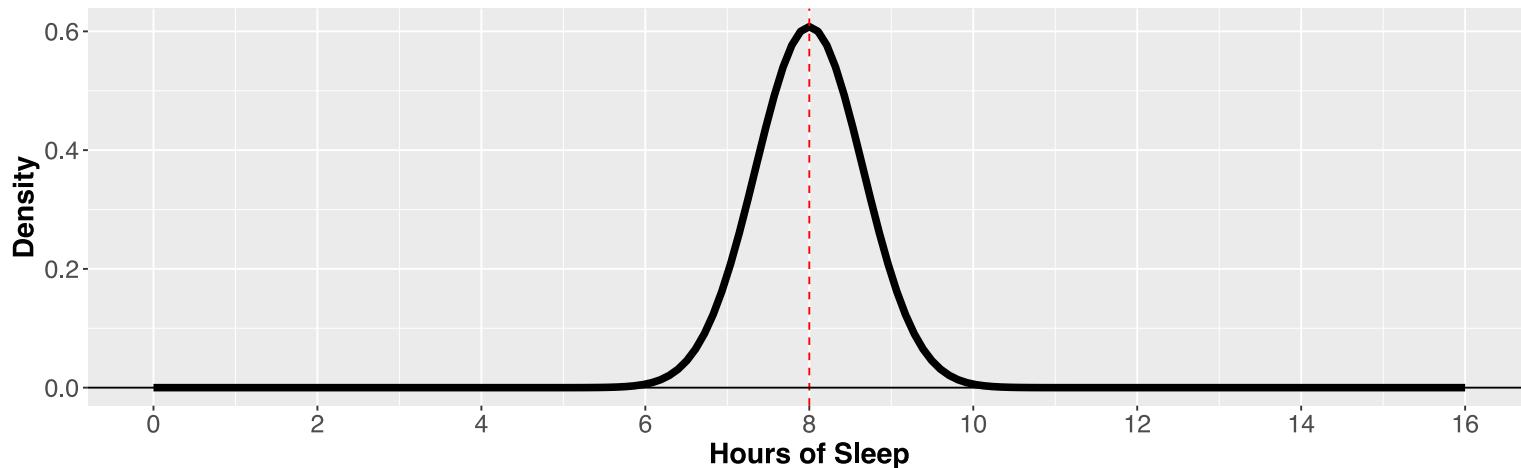
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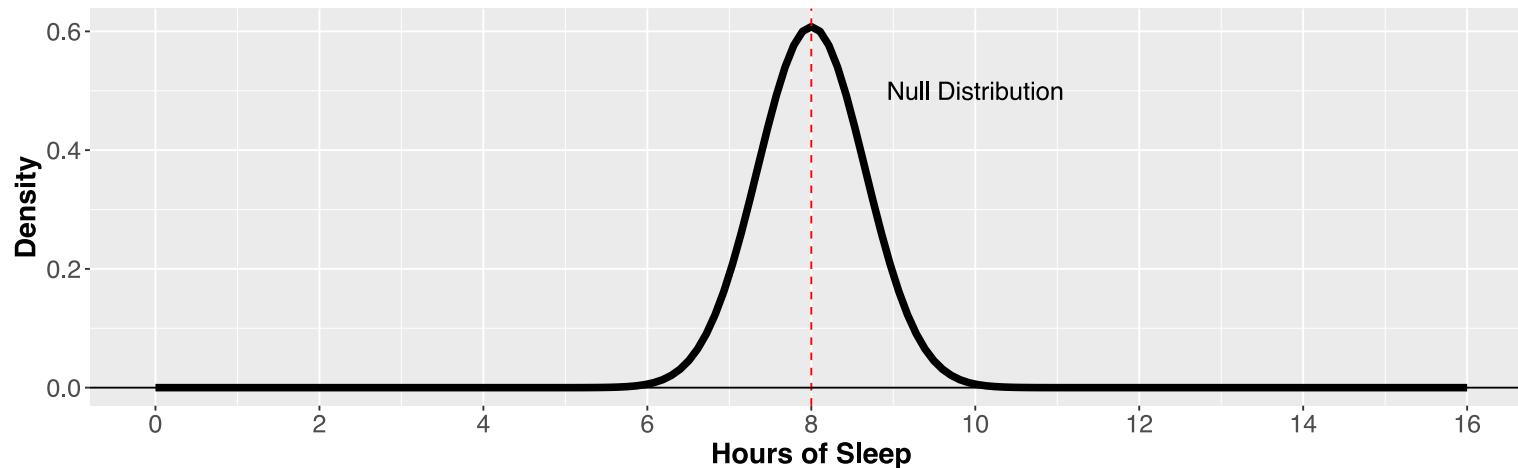
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The Null Distribution



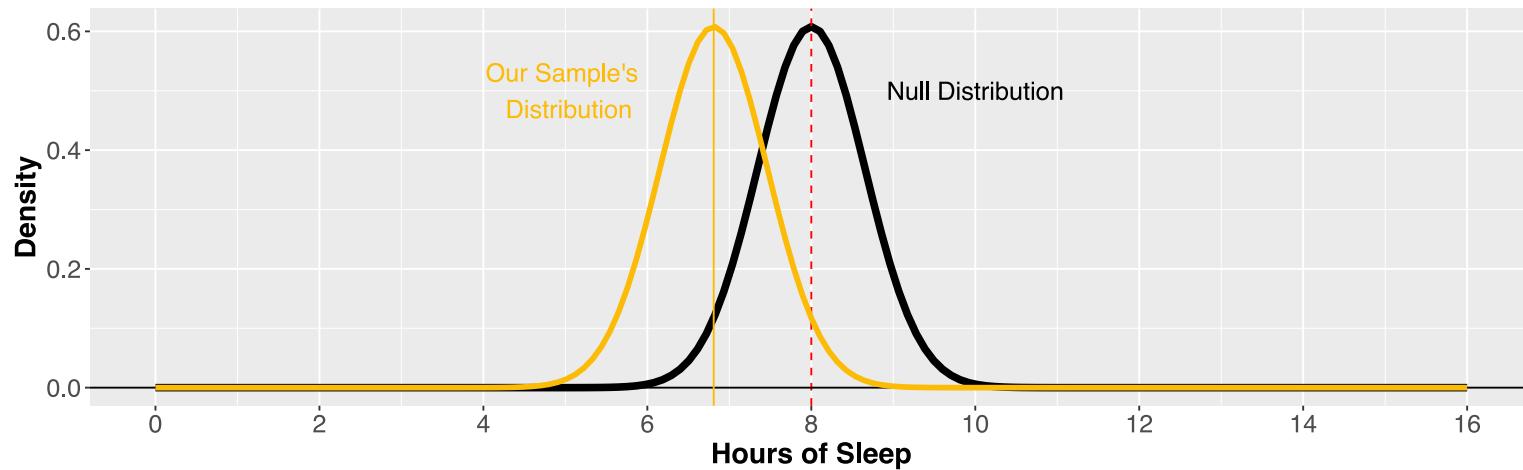
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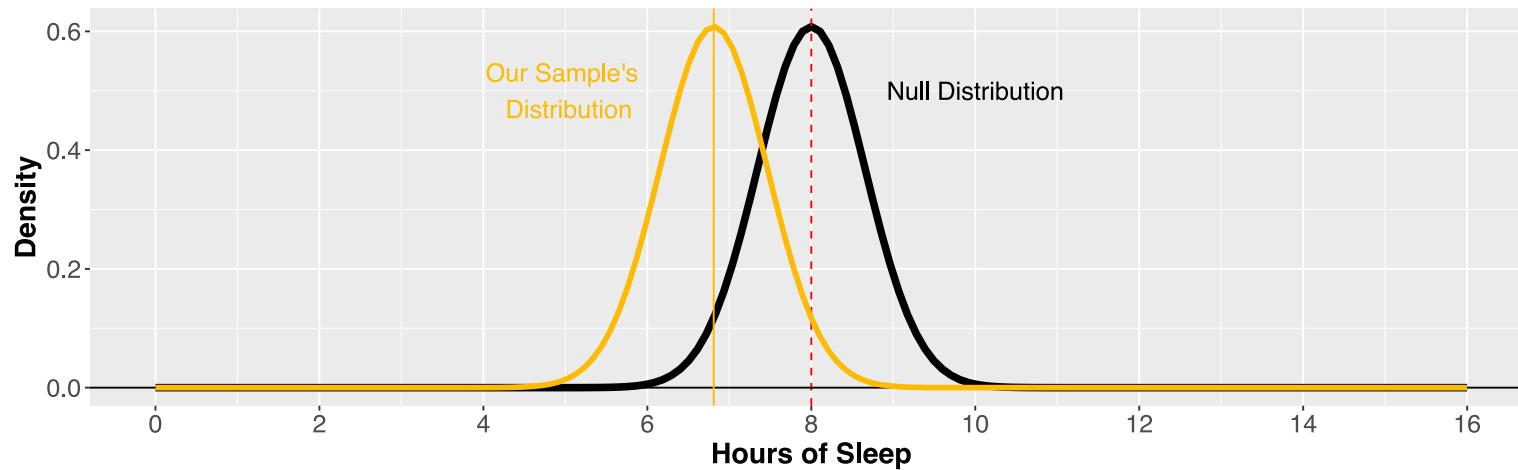


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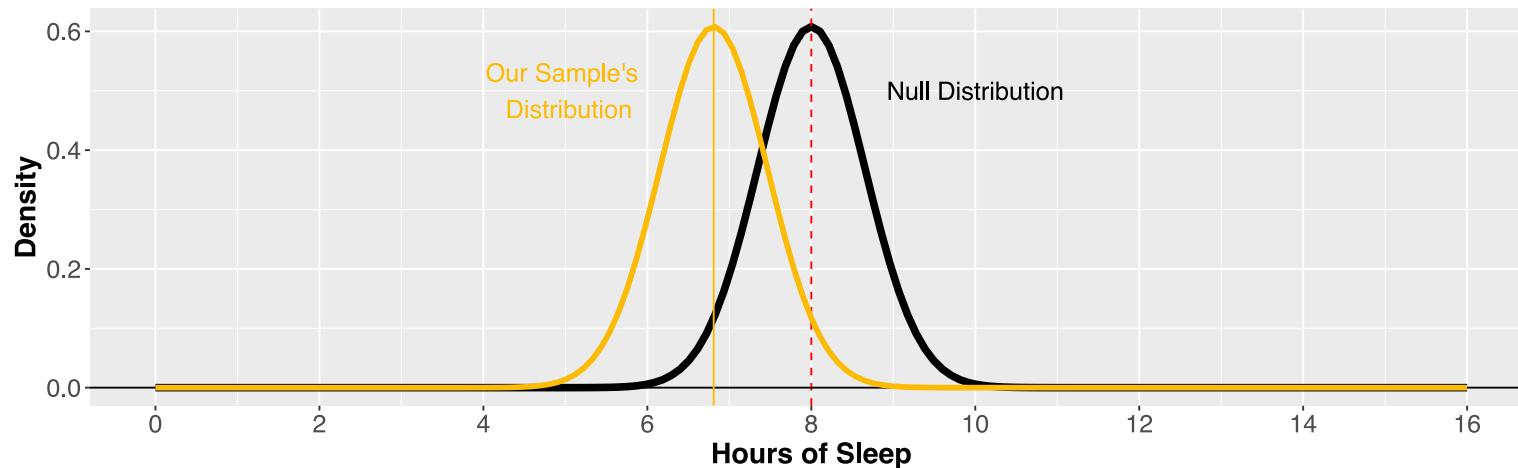


The Null Distribution



- From this, it *looks* as though our students may be getting less sleep than they should.

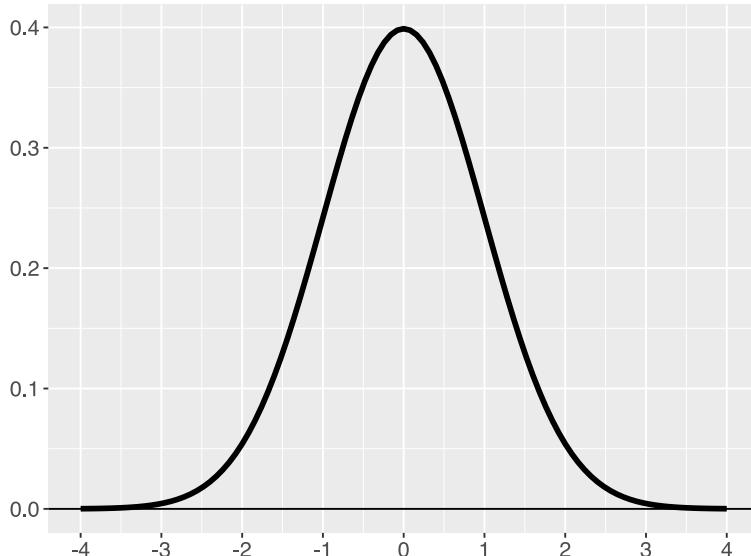
The Null Distribution



- 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.

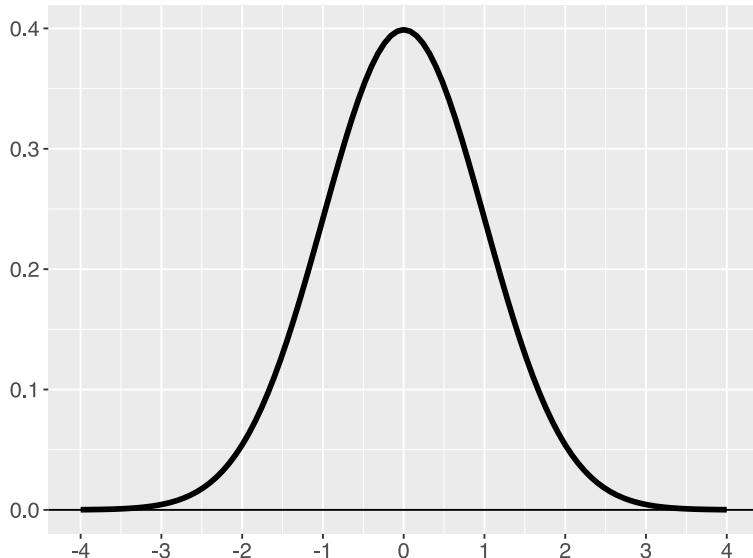
Back to the Normal Distribution...Again

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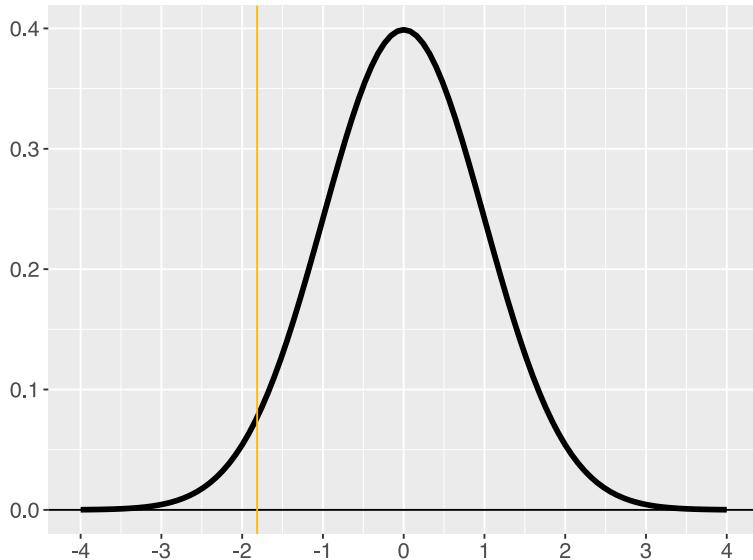


If our data are normally distributed, we can standardize the mean by converting it to a **z-score**

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

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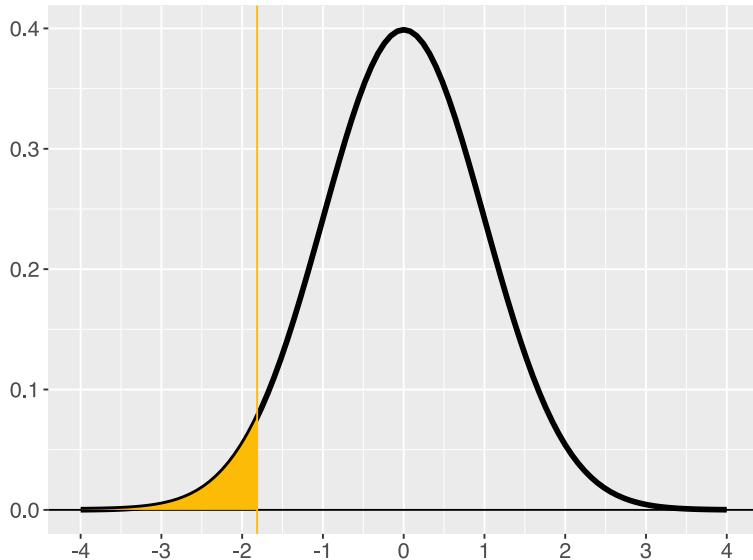
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```
(mean(m$sleep) - 8)/(sd(m$sleep)/sqrt(12))
```

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

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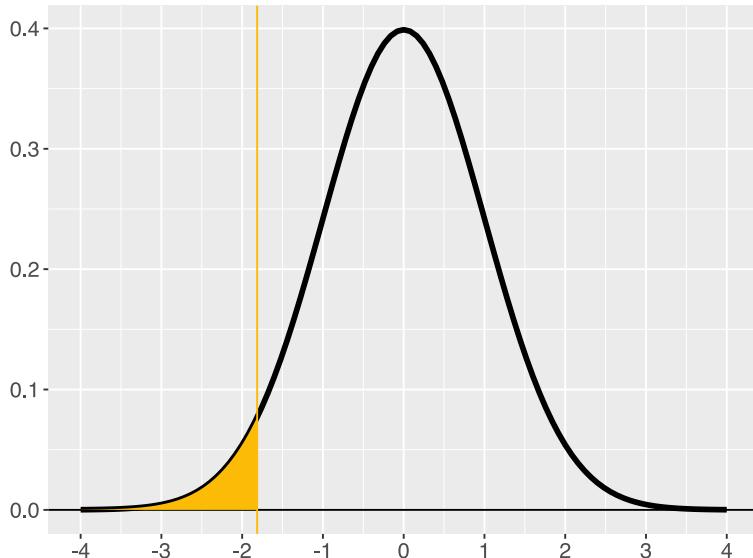
```
## [1] -1.814
```

```
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



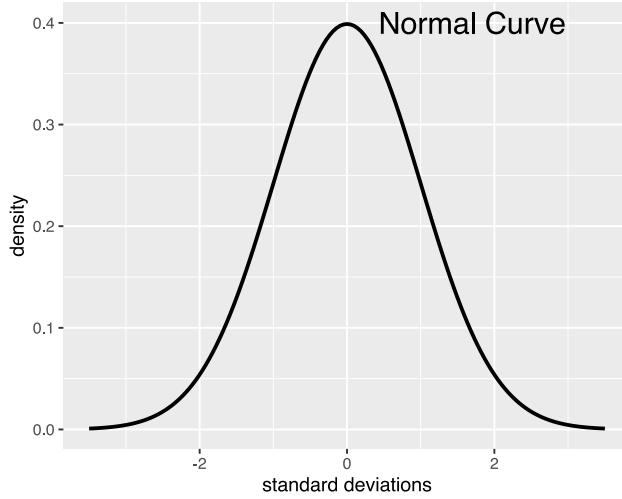
Part Two wasn't entirely true

- 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



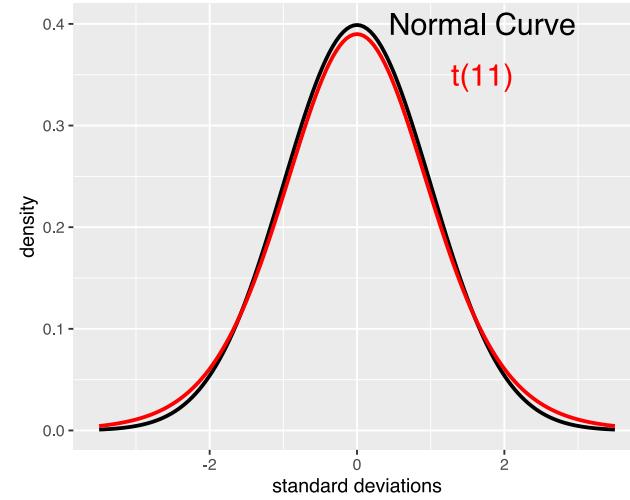
"A. Student", or William Sealy Gossett



The t Distribution



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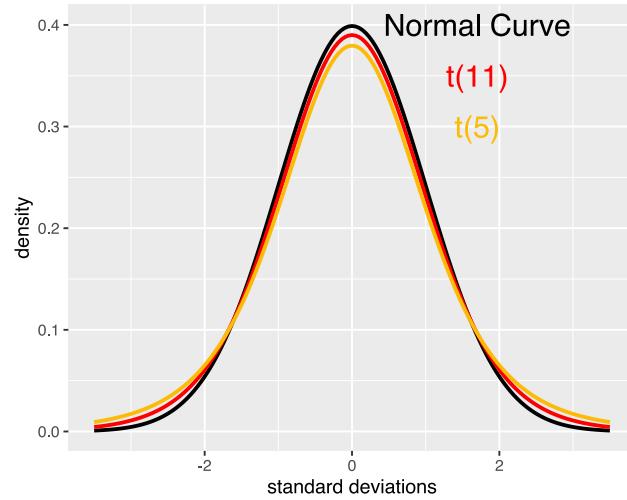


Note that the shape changes according to *degrees of freedom*

The t Distribution



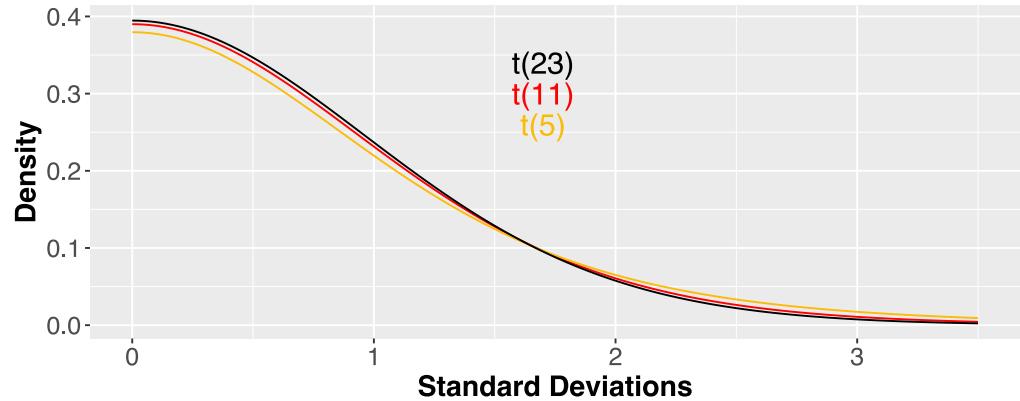
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Note that the shape changes according to *degrees of freedom*

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

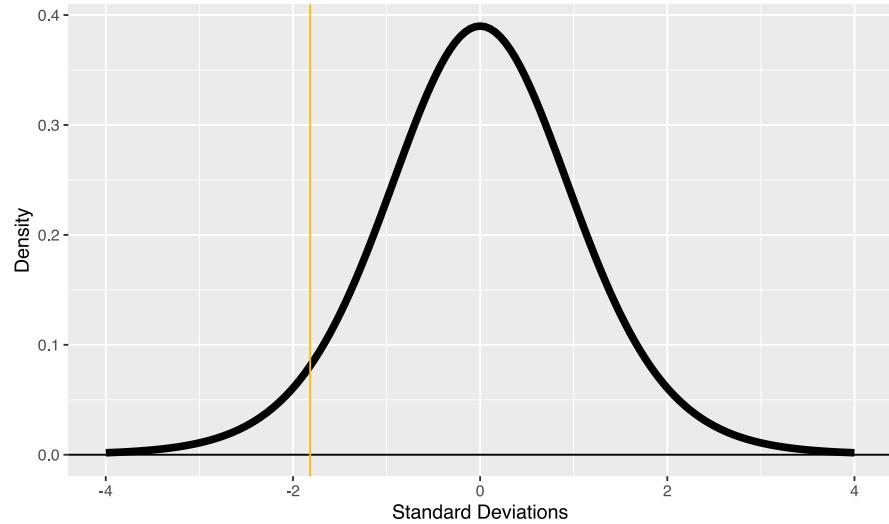
- In part 2, we calculated the mean hours of sleep for the group as 6.81
- We used the formula $z = \frac{\bar{x} - \mu}{\sigma/\sqrt{n}}$ to calculate z , and the standard normal curve to calculate probability

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

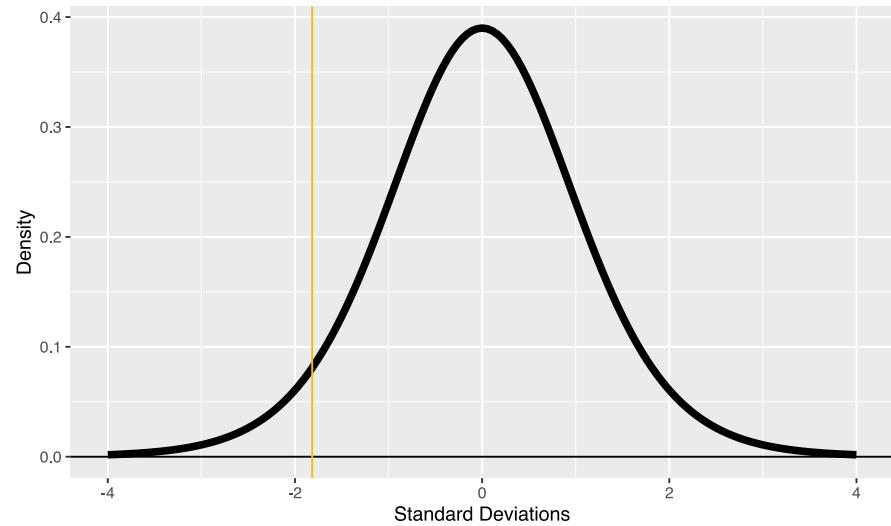
Probability According to t

- 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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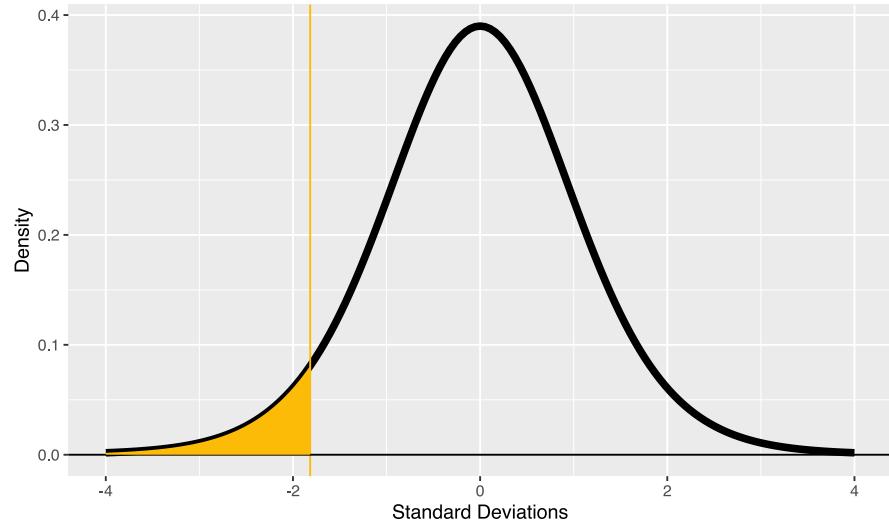


Probability According to t

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- $t(11) = \frac{6.81 - 8}{2.27/\sqrt{12}} = -1.816$
- 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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No.

```
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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- ...And `alternative="two.sided"`

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 - H_1 is that our students would be getting *less* sleep than the average person.
- 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!

Making a Decision

- To make this decision, we use a cut-off value for p called α

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 - If p is greater than α , we *fail to reject* H_0

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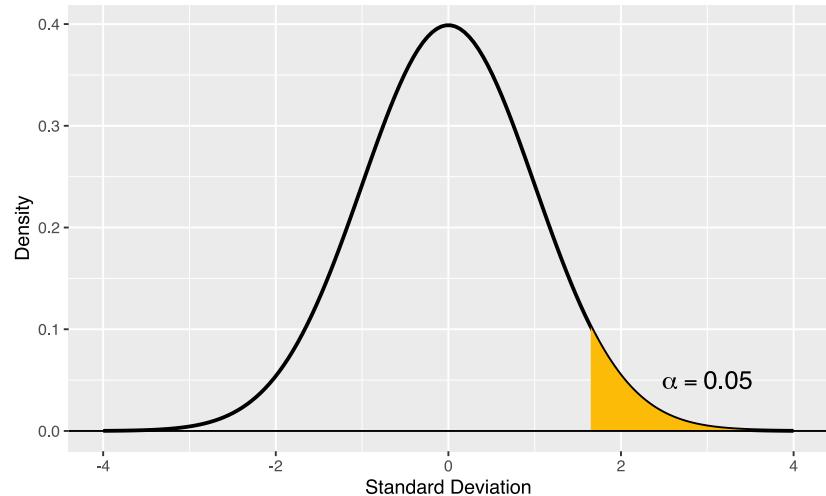
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- Typically, in Psychology, α is set to .05
 - We're willing to take a 5% risk of incorrectly rejecting the null hypothesis.

Making a Decision

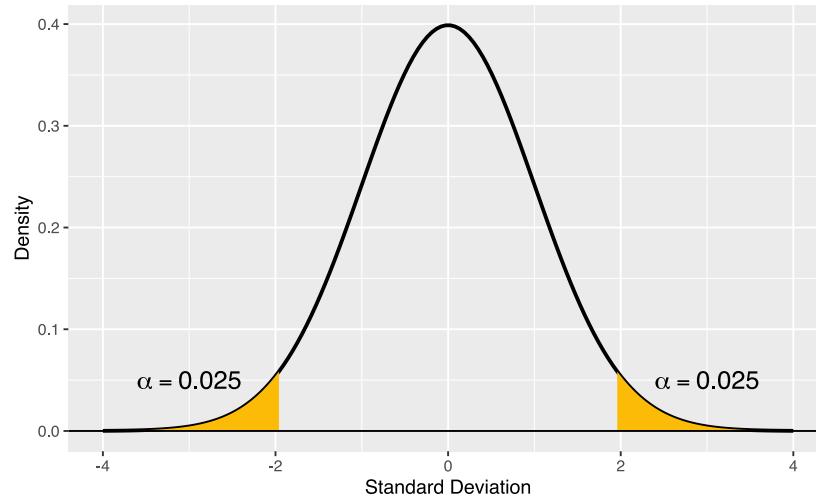
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- Typically, in Psychology, α is set to .05
 - We're willing to take a 5% risk of incorrectly rejecting the null hypothesis.
- It's important to set α before any statistical analysis

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

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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 - Allows you to compare measures that come from the same individual, e.g.

