

How to calculate p-values!!!

How to calculate p-values!!!

But, where do we use it?

Two-Sided p-values are the most common



One-Sided and **Two-Sided**

Two-Sided

Draft: Sharing is strictly Prohibited!

In contrast, **One-Sided p-values** are rarely used

One-Sided and Two-Sided



With that said, let's
imagine I had a coin...



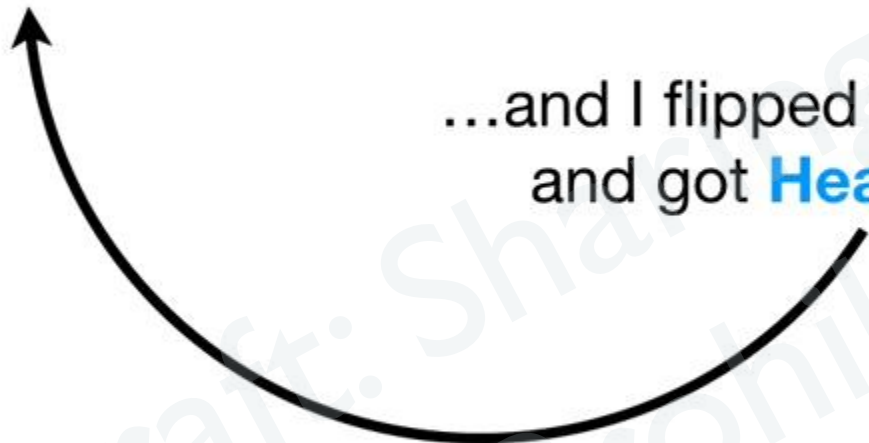
Draft: Sharing is strictly
Prohibited!



1st Flip



...and I flipped it once
and got **Heads**.



Draft: Sharing is strictly Prohibited!



1st Flip 2nd Flip



Then I flipped it again and got
Heads a second time.



Draft: Sharing is strictly Prohibited!



1st Flip 2nd Flip



Now, at this point, I might be tempted to think, “Wow! My coin is super special because it landed on **Heads** twice in a row!!!”



1st Flip



2nd Flip



Now, at this point, I might be tempted to think, "Wow!

My coin is super special because it landed on **Heads** twice in a row!!!"



This... ...is a hypothesis.



1st Flip 2nd Flip



Now, at this point, I might be tempted to think, "Wow!

My coin is super special because it landed on **Heads** twice in a row!!!"

However, in Statistics Lingo, the hypothesis is:
Even though I got 2 Heads in a row, my coin is no different from a normal coin.



1st Flip



2nd Flip



Statisticians call this the **Null Hypothesis**, and a small **p-value** will tell us to reject it.

My coin is super special because it landed on **Heads** twice in a row!!!”

However, in Statistics Lingo, the hypothesis is:

Even though I got 2 Heads in a row, my coin is no different from a normal coin.



1st Flip 2nd Flip



And if we reject this **Null Hypothesis**, we will know that our coin is special.

My coin is super special because it landed on **Heads** twice in a row!!!”

However, in Statistics Lingo, the hypothesis is:

Even though I got 2 Heads in a row, my coin is no different from a normal coin.



1st Flip 2nd Flip



So let's test this hypothesis by calculating a **p-value**.

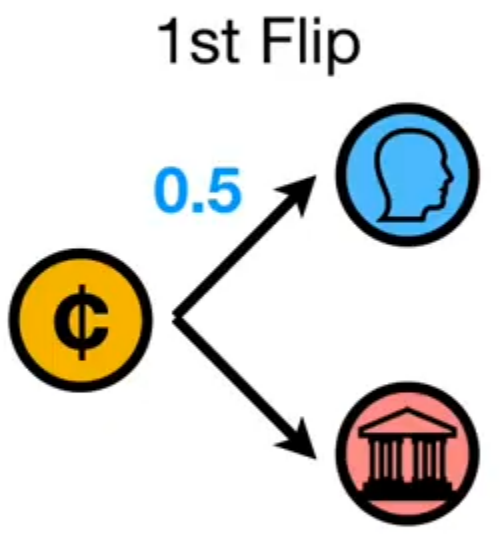
My coin is super special because it landed on **Heads** twice in a row!!!”

However, in Statistics Lingo, the hypothesis is:

Even though I got 2 Heads in a row, my coin is no different from a normal coin.

p-values are determined by adding up probabilities, so let's start by figuring out the **probability** of getting **2 Heads** in a row.

When we flip a normal,
everyday coin, there's a
50% chance we'll get
Heads...



Draft: Sharing is strictly
Prohibited!

When we flip a normal,
everyday coin, there's a
50% chance we'll get

Heads...

1st Flip

0.5



0.5



...and a **50%** chance we'll
get **Tails.**



Draft: Sharing is strictly
Prohibited!

1st Flip

0.5



0.5



Now, if we got **Heads** on the first flip...

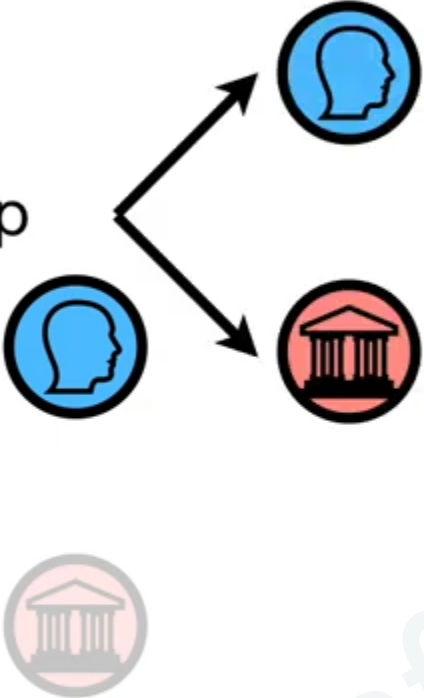
Draft. Sharing is strictly Prohibited!

2nd Flip

1st Flip

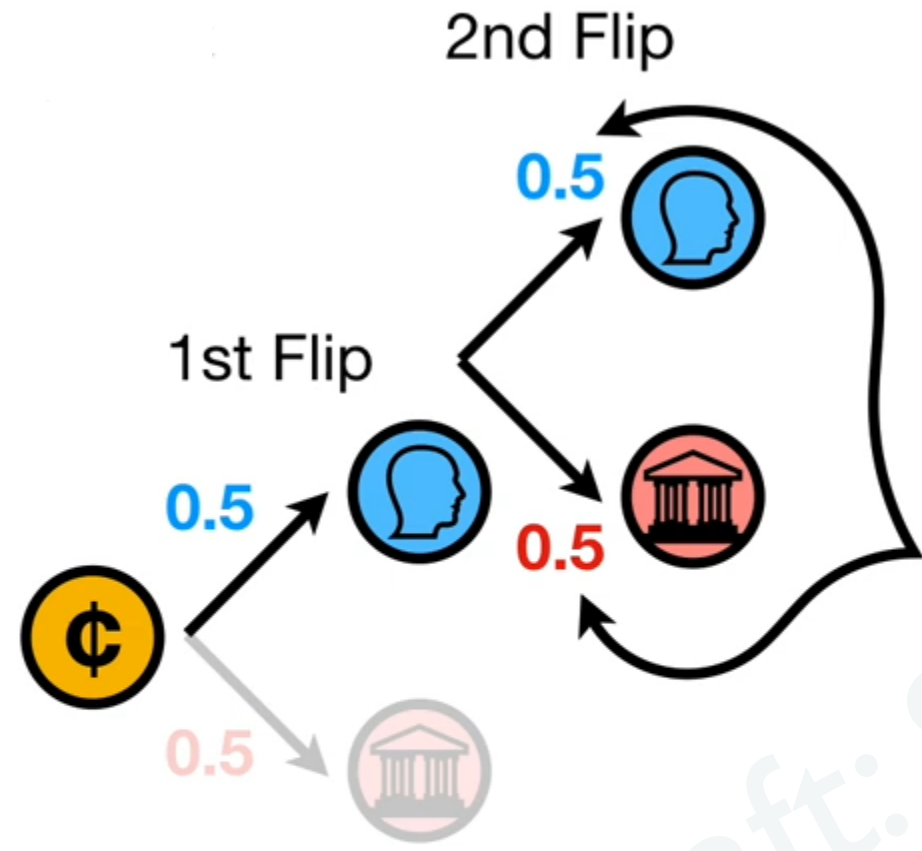
0.5

0.5



...and flipped the coin a second time...

Draft: Sharing is strictly Prohibited!

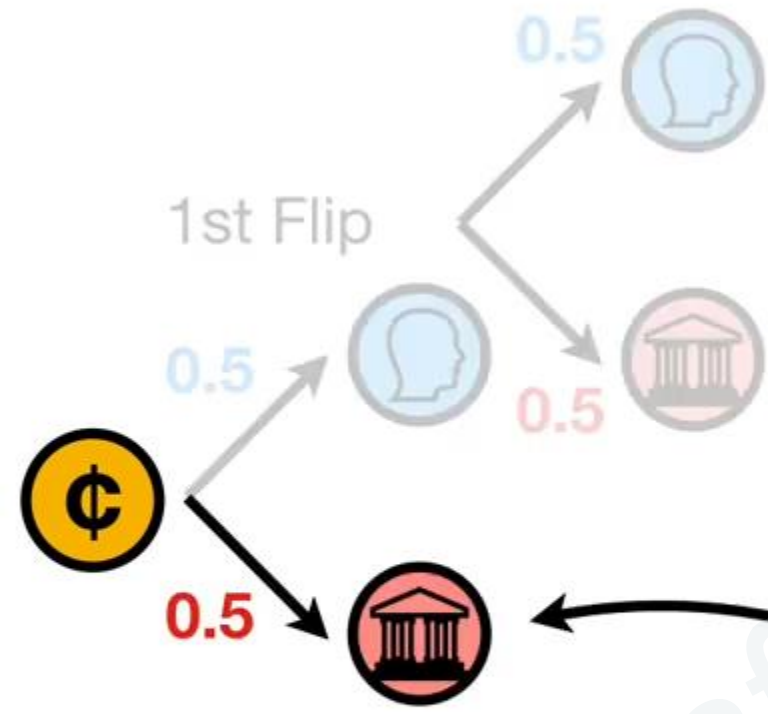


...then, just like before,
there's a **50%** chance we'll
get **Heads**, and a **50%**
chance we'll get **Tails**.

Draft: Share this strictly
Prohibited!

2nd Flip

1st Flip

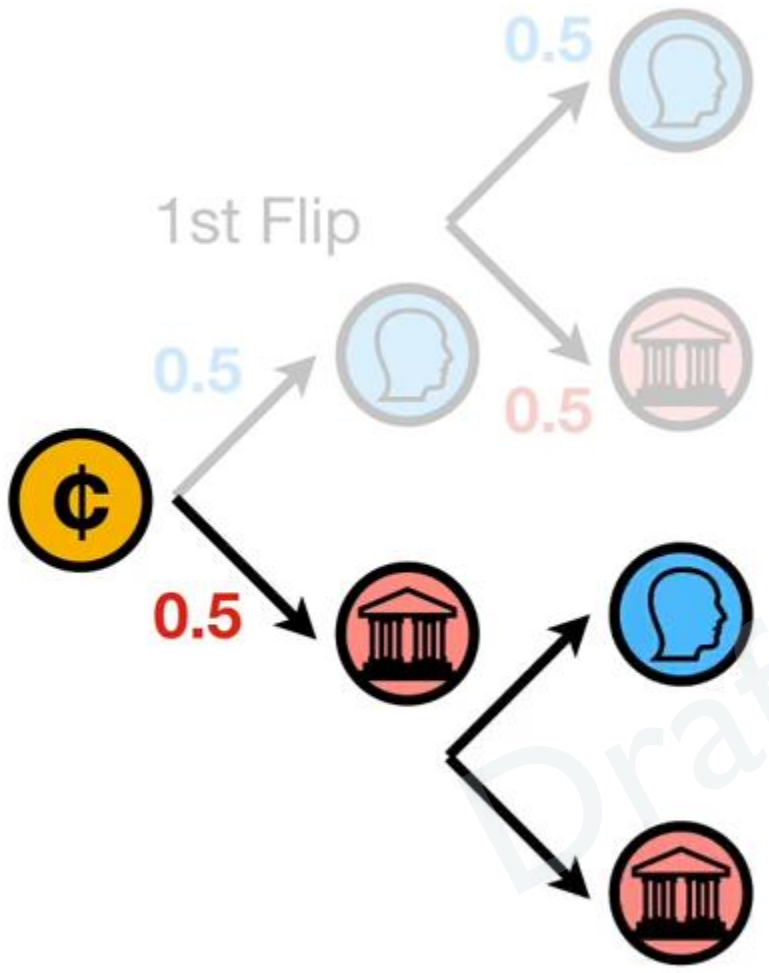


Likewise, if we got **Tails** on the first flip...

Draft: Sharing is strictly Prohibited!

2nd Flip

1st Flip

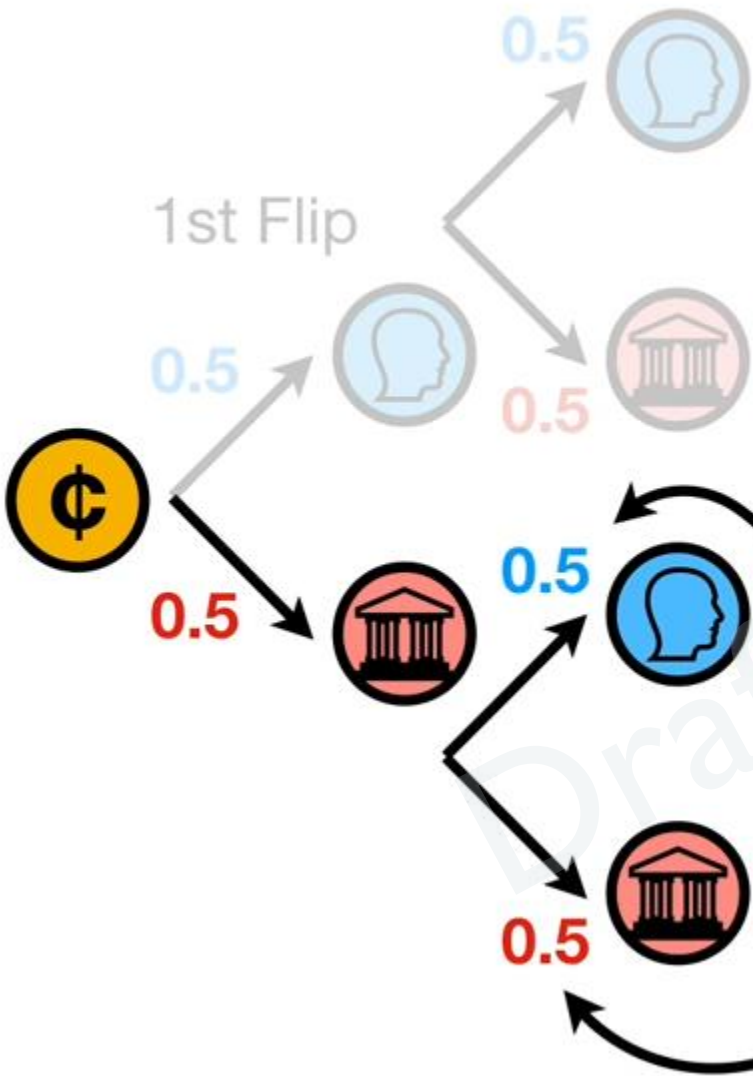


...and flipped the coin again...

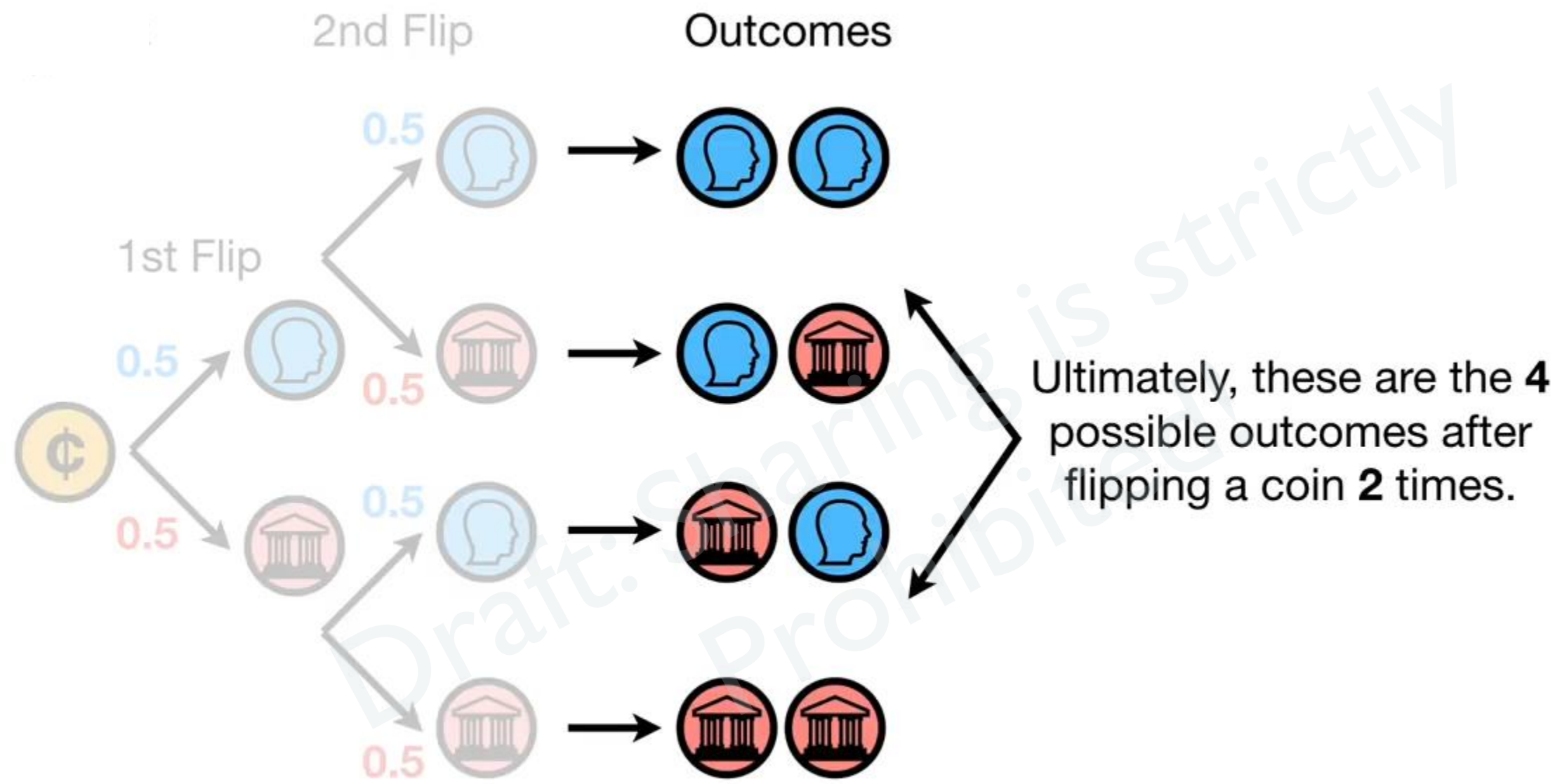
Draft: Sharing is strictly Prohibited!

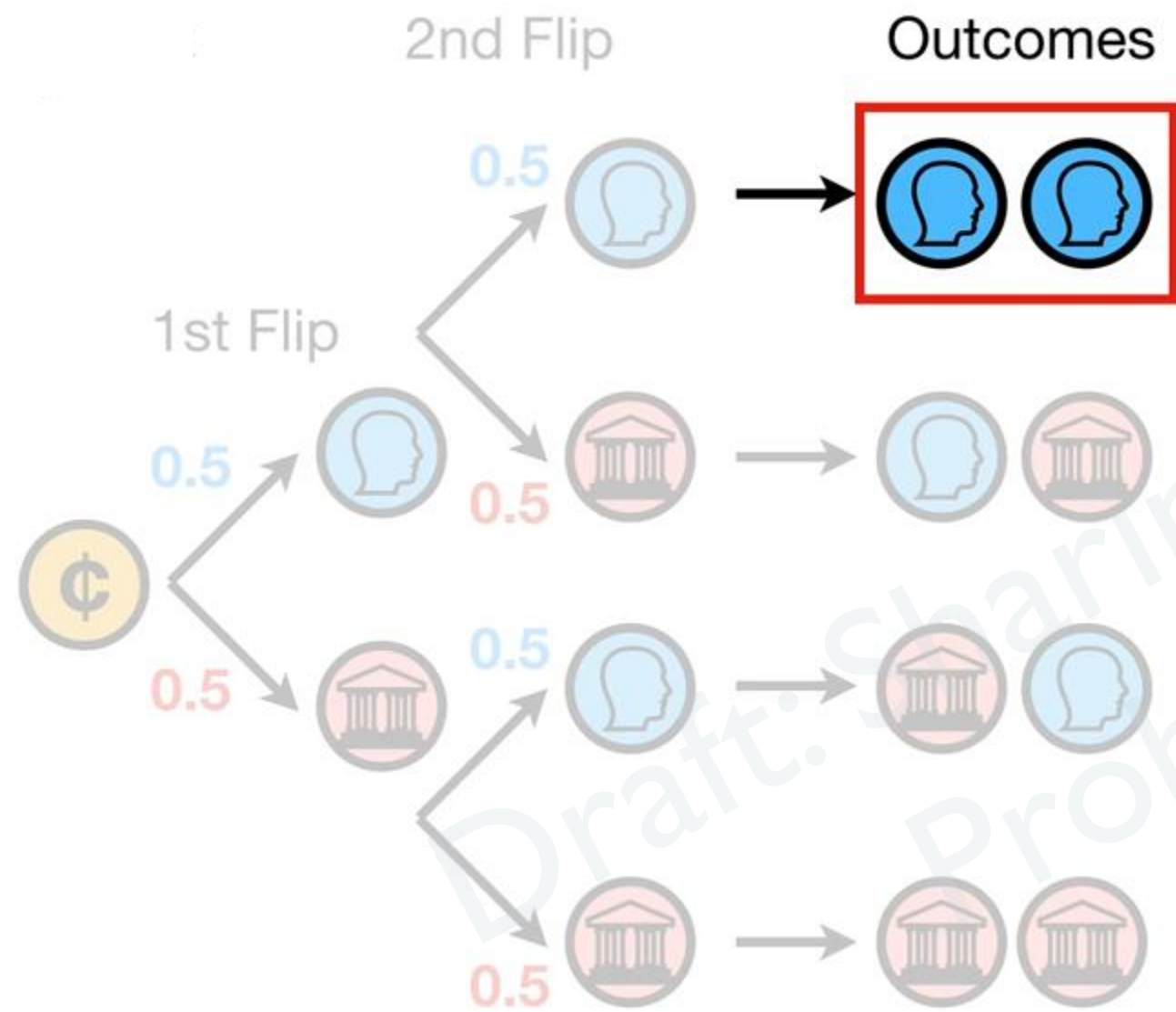
2nd Flip

1st Flip



...then, just like before,
there's a **50%** chance we'll
get **Heads**, and a **50%**
chance we'll get **Tails**.





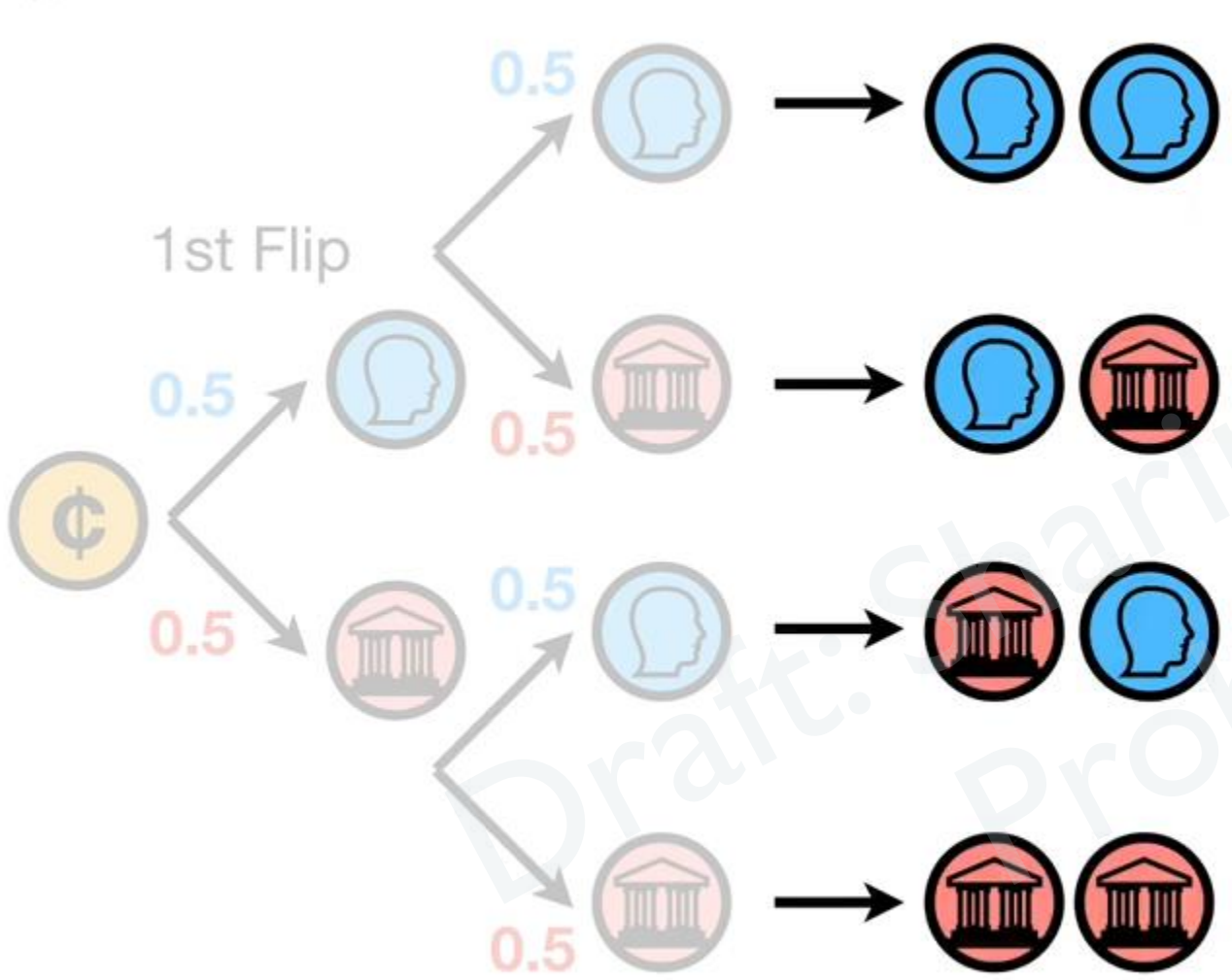
Thus, the *probability* of getting **2 Heads** is **0.25**.

The number of times we got **2 Heads**.

$$\frac{\text{The number of times we got 2 Heads.}}{\text{The total number of outcomes.}} = \frac{1}{4} = 0.25$$

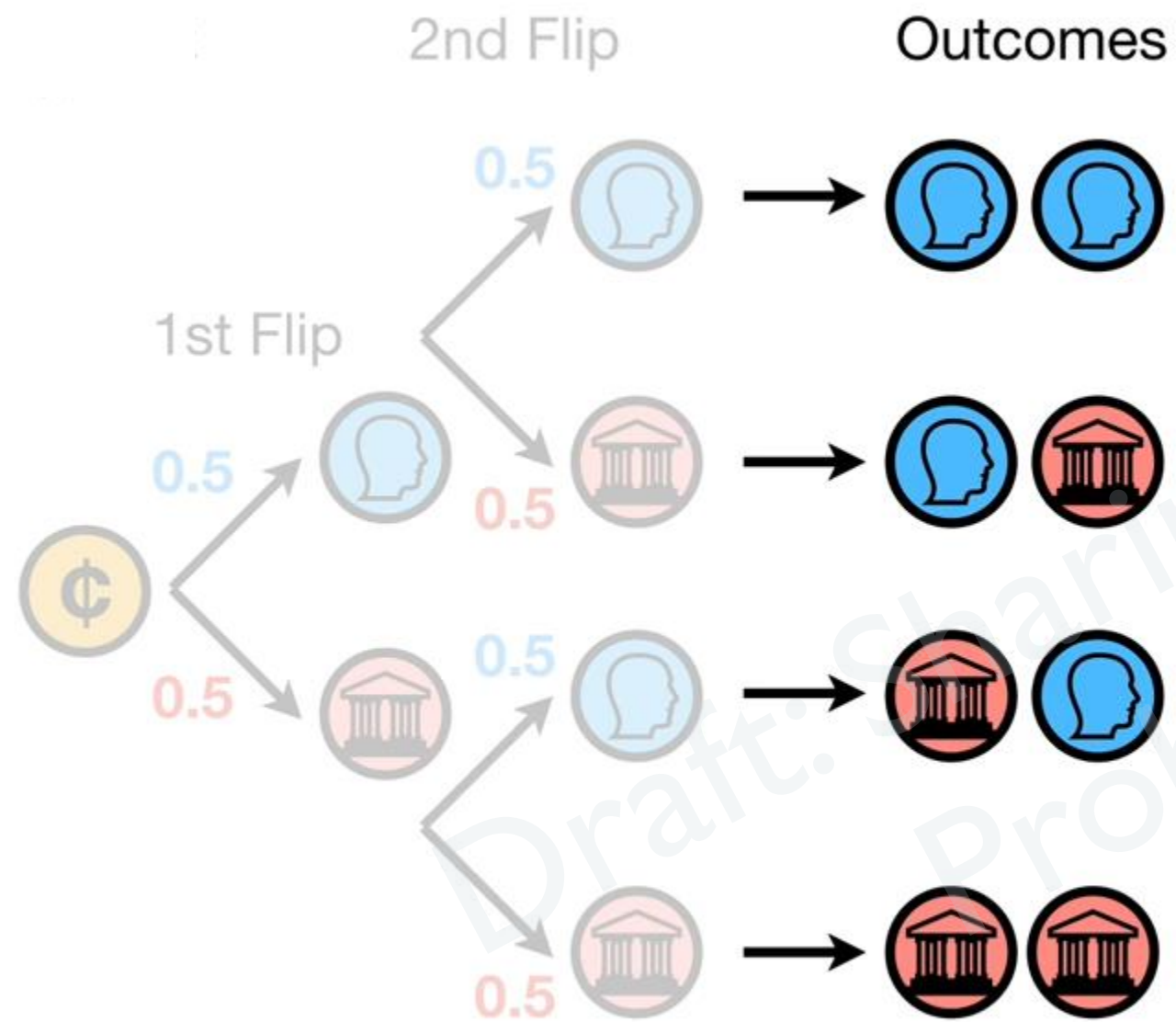
2nd Flip

Outcomes



Likewise, the probability of getting **2 Tails** is...

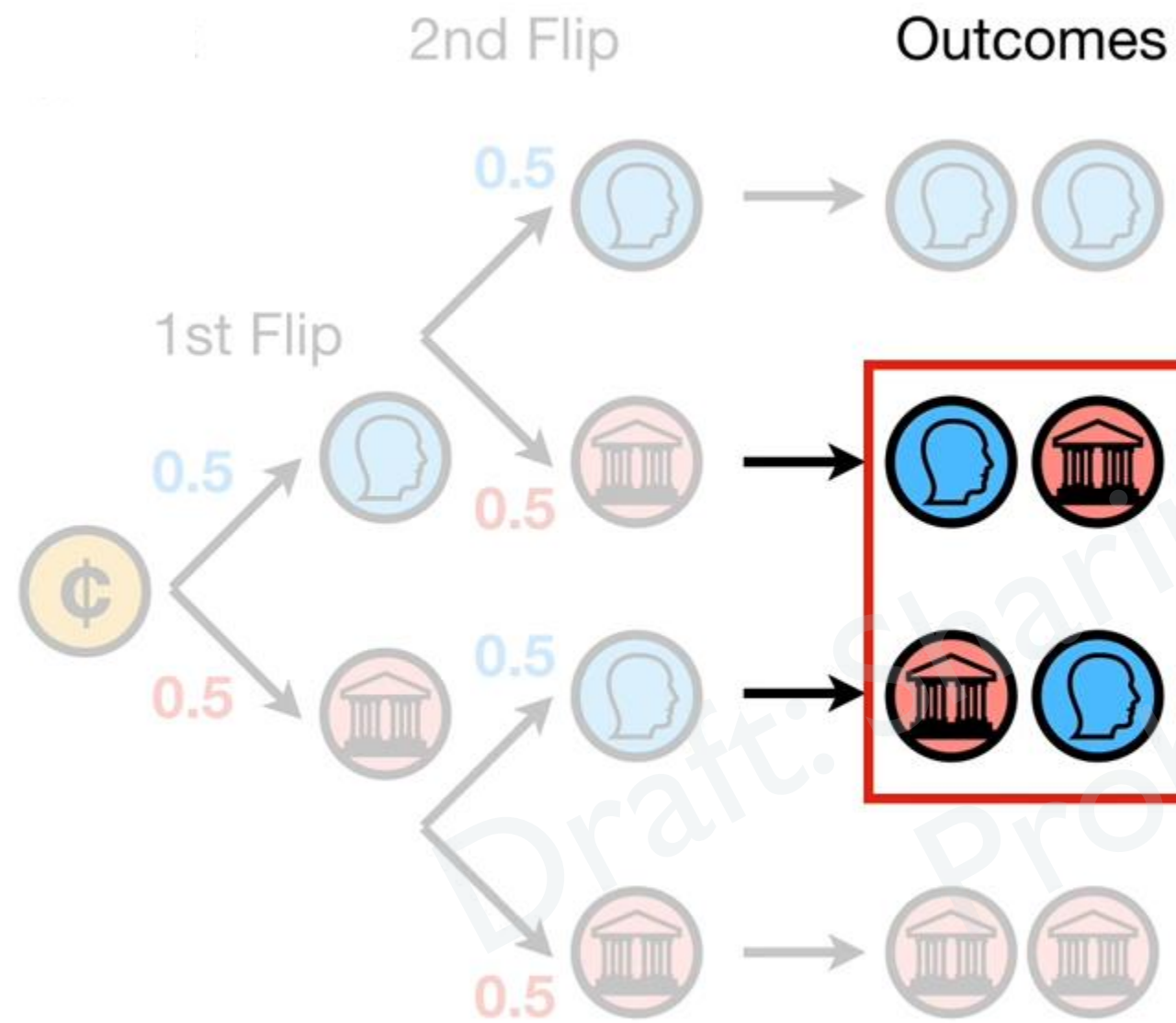
Draft: Sharing is strictly Prohibited!



Likewise, the probability of getting **2 Tails** is...

The number of times we got **2 Tails**.

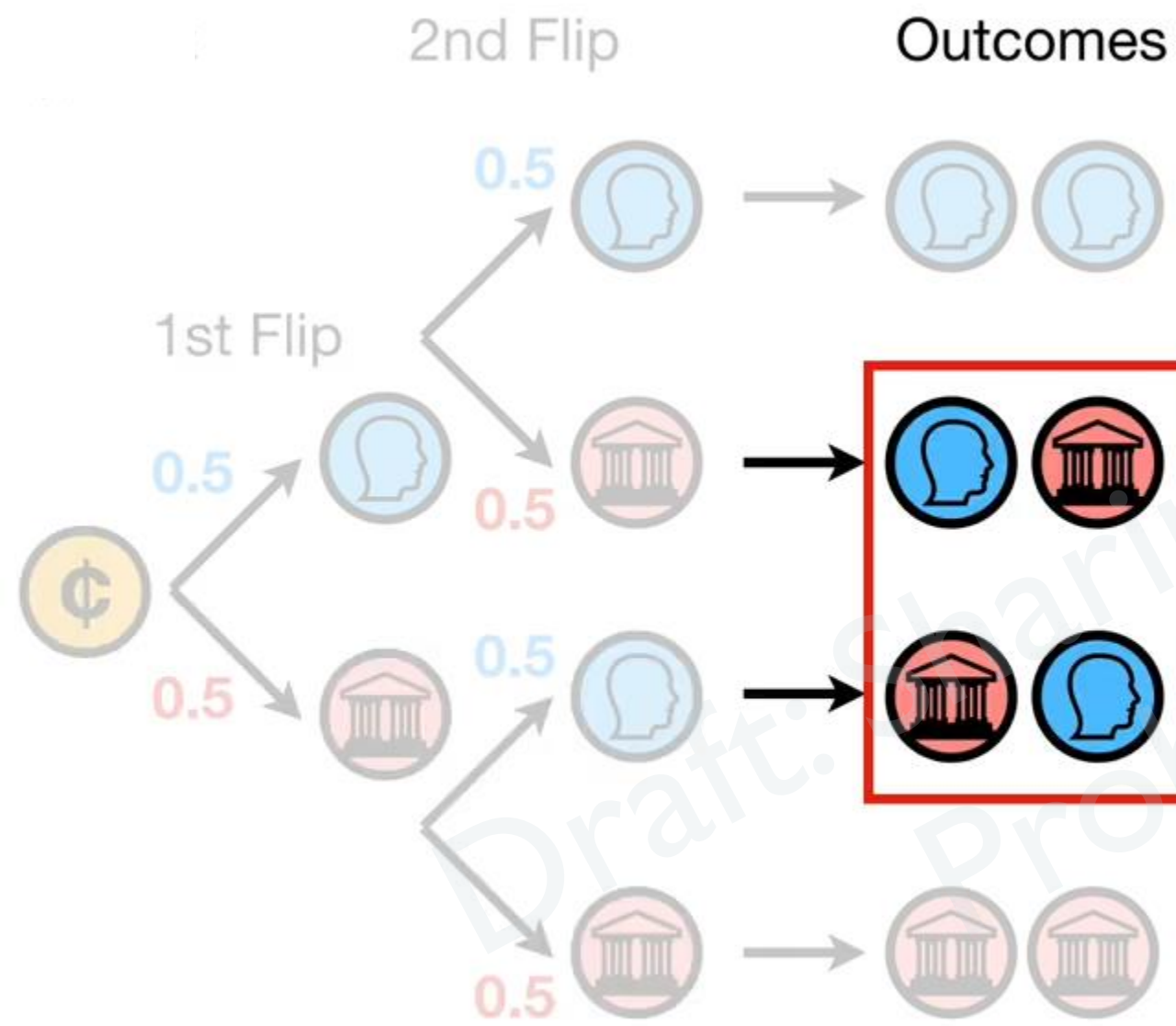
$$\frac{\text{The number of times we got 2 Tails.}}{\text{The total number of outcomes.}} = \frac{1}{4} = 0.25$$



Finally, the probability of getting **1 Heads** and **1 Tails**, regardless of the order, is...

The number of times **Heads** and **Tails** occurred

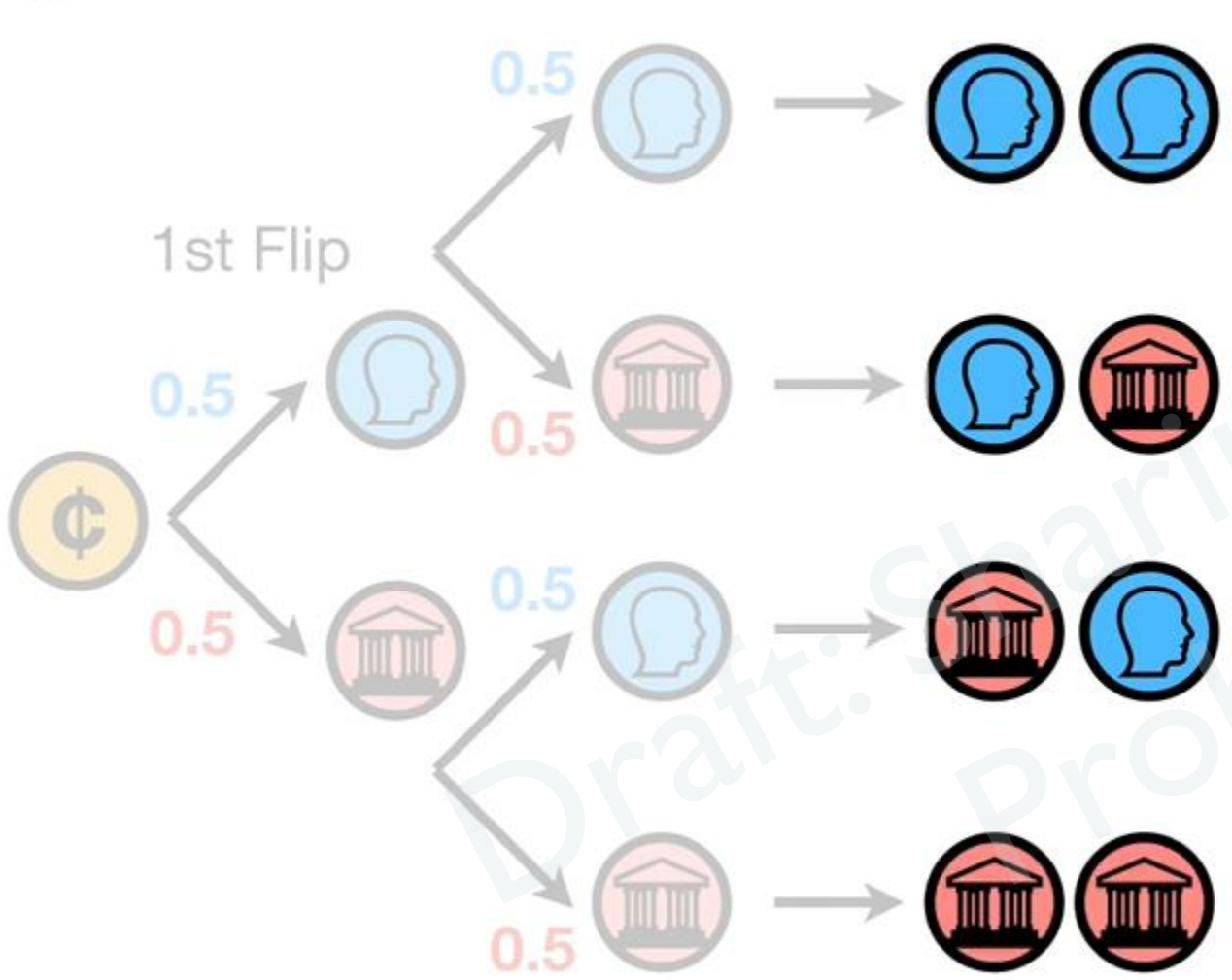
$$\frac{\text{The number of times } \mathbf{Heads} \text{ and } \mathbf{Tails} \text{ occurred}}{\text{The total number of outcomes.}} = \frac{2}{4} = 0.5$$



Because order does not effect the probabilities of getting **Heads** and **Tails**, we treat these outcomes as the same.





2nd Flip

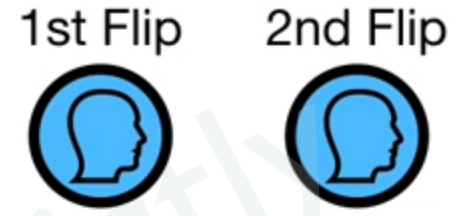
Outcomes



Now let's move the outcomes over to the left...

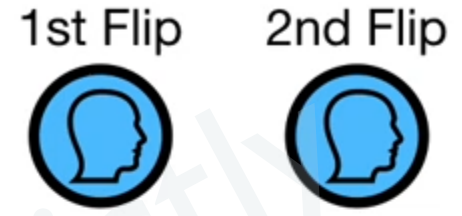
Draft: Sharing is strictly Prohibited!

<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	0.5
	0.25




...and calculate the **p-value** for getting two heads.

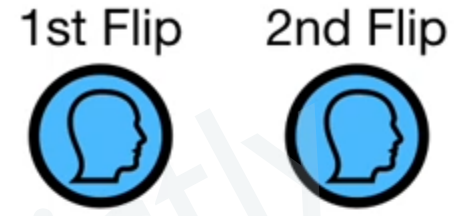
<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	
	0.25



A ***p-value*** is composed of three parts:

Draft: Sharing is strictly Prohibited!




<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	
	0.25



A ***p-value*** is composed of three parts:

- 1) The probability random chance would result in the observation.

p-value for **2 Heads** = 0.25

<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	0.25

1st Flip



2nd Flip



A *p-value* is composed of three parts:

- 1) The probability random chance would result in the observation.

In this case, the first part is just the probability that a normal coin would give us **2 Heads**, which is **0.25**.

p-value for **2 Heads** = 0.25

<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	
	0.25

1st Flip



2nd Flip

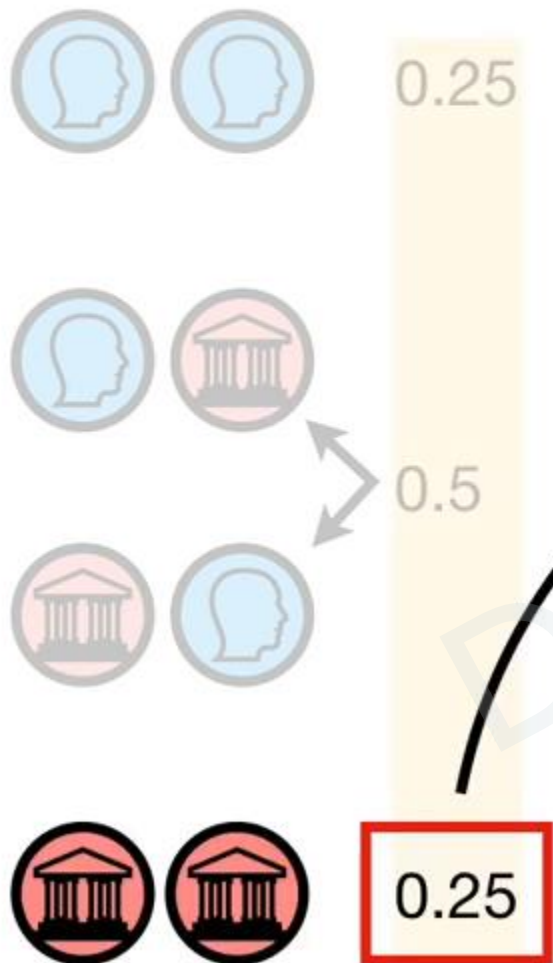


A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.

p-value for **2 Heads** = $0.25 + 0.25$

Outcomes Probability



In this case, getting **2 Tails** is as rare as **2 Heads**, so we add **0.25**.

1st Flip






2nd Flip



A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.

p-value for **2 Heads** = $0.25 + 0.25$




<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	
	0.25



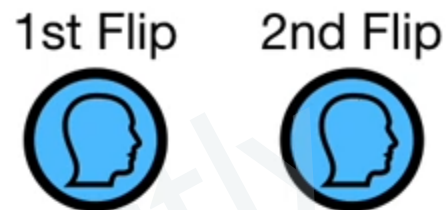
A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0$

<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	0.5
	0.25





In this case, the third part is **0**, because no other outcomes are rarer than **2 Heads** or **2 Tails**.



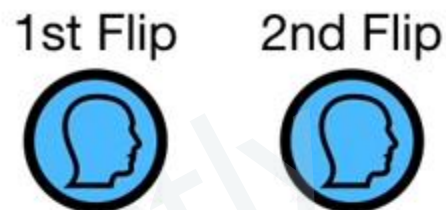
A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0$

<u>Outcomes</u>	<u>Probability</u>
	0.25
	0.5
	
	0.25

Now we just add everything together...



A *p-value* is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

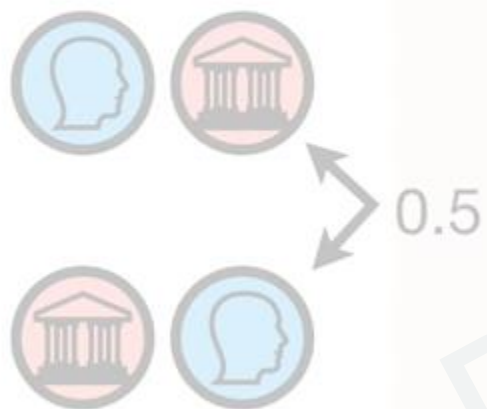
1st Flip



2nd Flip



Outcomes Probability



...and the **p-value** for getting **2 Heads** = **0.5**.

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

1st Flip



2nd Flip



Outcomes

Probability



0.25

Now remember, the reason we calculated the **p-value** was to test this hypothesis:



0.5



0.25

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

1st Flip



2nd Flip



Outcomes Probability



0.25

Now remember, the reason we calculated the **p-value** was to test this hypothesis:



0.5



0.25

Even though I got **2 Heads in a row, my coin is no different from a normal coin.**

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

1st Flip



2nd Flip



Outcomes Probability



0.25

Typically, we only reject a hypothesis if the **p-value** is less than **0.05**...



0.5

Even though I got 2 Heads in a row, my coin is no different from a normal coin.



0.25

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

1st Flip



2nd Flip



Outcomes Probability



0.25

...and since **0.5 > 0.05**, we fail to reject the hypothesis.



0.5

Even though I got 2 Heads in a row, my coin is no different from a normal coin.



0.25

A *p-value* is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

1st Flip



2nd Flip



Outcomes

Probability



0.25

In other words, the data, getting **2 Heads** in a row, failed to convince us that our coin is special.



0.5



0.25

A *p-value* is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

1st Flip



2nd Flip



Outcomes Probability



0.25



0.5



0.25

NOTE: The *probability* of getting **2 Heads**, **0.25**, is different from the **p-value** for getting **2 Heads**, **0.5**.

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

p-value for **2 Heads** = $0.25 + 0.25 + 0 = 0.5$

1st Flip



2nd Flip



Outcomes Probability



0.25



0.5



0.25

This is because the **p-value** is the sum of three parts...

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$

1st Flip



2nd Flip



Outcomes Probability



0.25



0.5



0.25

This is because the **p-value** is the sum of three parts...

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$

1st Flip



2nd Flip



Outcomes Probability



0.25



0.5



0.25

This is because the **p-value** is the sum of three parts...

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$

1st Flip



2nd Flip



Outcomes Probability



0.25



0.5



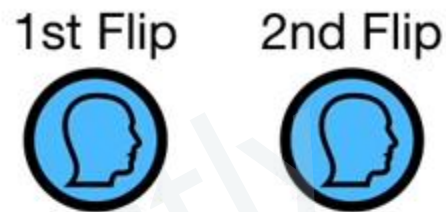
0.25

This is because the **p-value** is the sum of three parts...

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$



Outcomes Probability



Now the question is, **“Why do we care about things that are equally rare or more extreme?”**

A *p-value* is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$

1st Flip



2nd Flip



Outcomes Probability



In other words, why do we add **Parts 2** and **3** to the **p-value**?

A **p-value** is composed of three parts:

1) The probability random chance would result in the observation.

2) The probability of observing something else that is equally rare.

3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$

1st Flip



2nd Flip



Outcomes Probability



0.25



0.5



0.25

We add **Part 2**, the probability of something else that is equally rare, because although getting **2 Heads** might seem special, it doesn't seem as special when we know that other things are just as rare.

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

For example, imagine giving a loved one a flower and saying, “This is the rarest flower of this species, none are equally as rare.”



1st Flip



2nd Flip



A ***p-value*** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

For example, imagine giving a loved one a flower and saying, “This is the rarest flower of this species, none are equally as rare.”



Chances are, your loved one would think that the flower was super special.

1st Flip



2nd Flip



A ***p-value*** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

Now imagine saying to your loved one,
“This flower is equally as rare as all of
these other flowers.”



1st Flip



2nd Flip



A ***p-value*** is composed of
three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

Now imagine saying to your loved one,
“This flower is equally as rare as all of
these other flowers.”



In this case, your loved one might not
think the flower is very special.

1st Flip



2nd Flip



A ***p-value*** is composed of
three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.



NOTE: Even though these flowers are different colors, just knowing that they are equally rare would be a bummer.

1st Flip



2nd Flip



A ***p-value*** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$

1st Flip



2nd Flip



Outcomes

Probability



0.25



0.5



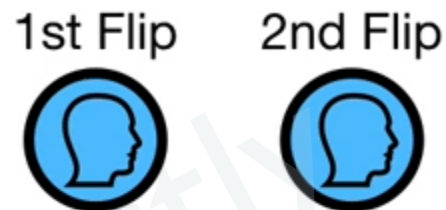
0.25





Because a lot of equally rare things would make something less special, we add **Part 2** to the **p-value**.

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

$$\text{p-value for 2 Heads} = 0.25 + 0.25 + 0 = 0.5$$



Outcomes	Probability
	0.25
	0.5
	
	0.25

And we add rarer things to the **p-value** for a similar reason.

A **p-value** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

OK, now that we know
that getting **2 Heads** in a
row is not very special or
statistically significant...



...what about getting **4**
Heads and **1 Tails**?

Draft: Sharing is strictly
Prohibited!



...what about getting **4**
Heads and **1 Tails**?

Would that suggest that our
coin is special?

Draft: Sharing is strictly
Prohibited!



Again, although we want to know if the coin is special, the **Null Hypothesis** focuses on a normal coin...

Even though I got **4 Heads** and **1 Tails**, my coin is no different from a normal coin.



...but if we get a small **p-value** and reject the **Null Hypothesis**, we will know that our coin *is* special.

Even though I got **4 Heads** and **1 Tails**, my coin is no different from a normal coin.



So let's calculate the **p-value** for getting **4 Heads** and **1 Tails**.

Even though I got **4 Heads** and **1 Tails**, my coin is no different from a normal coin.



All in all, when we flip a coin **5** times, there are **32** possible outcomes.

HHHHH	TTHHH	TTTHH	
	THTHH	TTHTH	
	THHTH	TTHHT	TTTTH
	THHHT	THTTH	TTTHT
TTHHH	HTTHH	THTHT	TTHTT
HTHHH	HTHTH	THHTT	THTTT
HHTHH	HTHHT	HTTTH	HTTTT
HHHTH	HHTTH	HTTHT	
HHHHT	HHTHT	HTHTT	TTTTT
	HHHTT	HHTTT	



The **p-value** for getting **4 Heads** and **1 Tails** is...

- 1) The probability we randomly get **4 Heads** and **1 Tails**:

HHHHH	TTHHH	TTHHH	
	THTHH	TTHTH	
	THTHH	TTHHT	TTTTH
	THTHT	THHTH	TTTHT
	HTTHH	THHTT	TTHTT
	HTHTH	THHTT	THTTT
	HTHHT	HTTTH	HTTTT
	HTHTH	HTTHT	
	HTHTT	HTHTT	TTTTT
	HTTTH	HTTTT	
	HTTHT		
	HTTHT		
	HTTTT		



The **p-value** for getting **4 Heads** and **1 Tails** is...

1) The probability we randomly get **4 Heads** and **1 Tails**:

$$\frac{5}{32}$$

Since **5** of the **32** outcomes had **4 Heads** and **1 Tails**.

HHHHH	TTHHH	TTHHH	
	THTHH	THTHH	
	THHHT	THHHT	TTTTH
	THTTH	THTTH	TTTHT
T HHHH	HHTHH	HHTHH	TTHTT
H THHH	HHTHH	HHTHH	TTHTT
HT HHH	HHTTH	HHTTH	TTHTT
HH T HH	HHTHT	HHTHT	TTHTT
HHH T H	HHHTT	HHHTT	TTHTT
HHHH T	HHHTT	HHHTT	TTTTT



The **p-value** for getting **4 Heads** and **1 Tails** is...

- 1) The probability we randomly get **4 Heads** and **1 Tails**:

$$\frac{5}{32} +$$

- 2) The probability we randomly get something else that is equally rare:

	TTHHH	TTTHH	
	THTHH	THTTH	
HHHHH	THTTH	THTHT	TTTTT
	THTHT	THTTT	
TTHHH	HHTHH	HTTTH	
HHTHH	HHTTH	HHTTT	
HTHHH	HHTHT	HTTTT	
HHHTH	HHHTT		
HHHTT			



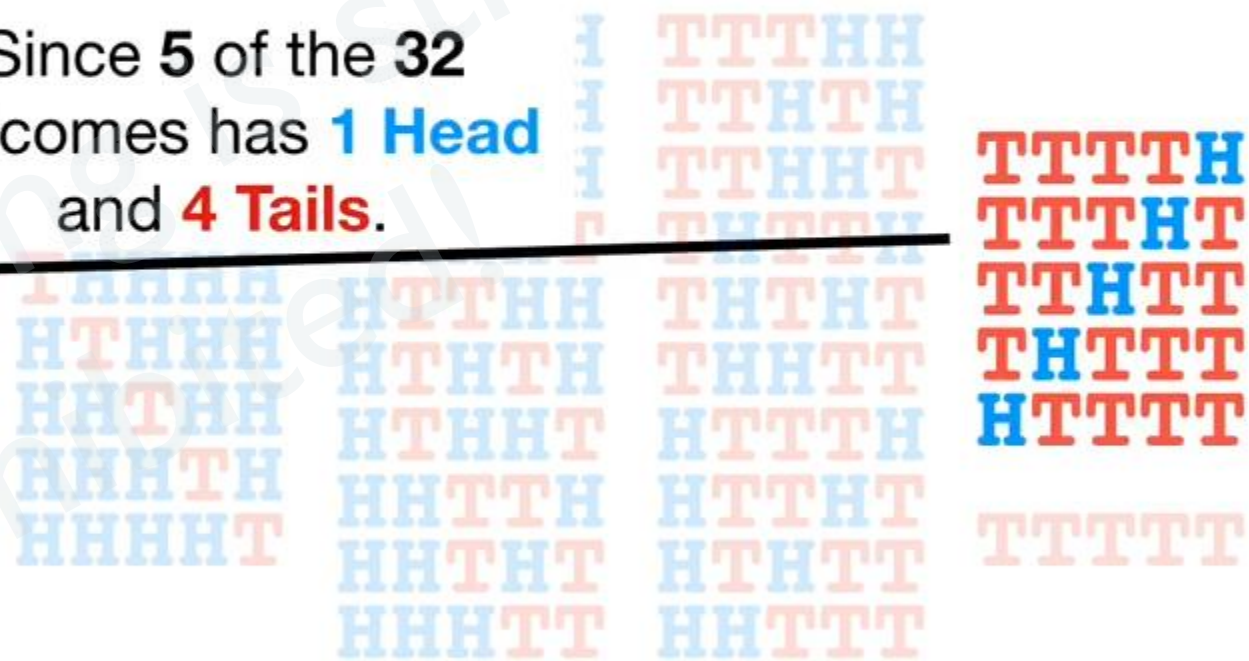
The **p-value** for getting **4 Heads** and **1 Tails** is...

- 1) The probability we randomly get **4 Heads** and **1 Tails**:

$$\frac{5}{32} + \frac{5}{32}$$

- 2) The probability we randomly get something else that is equally rare:

Since **5** of the **32** outcomes has **1 Head** and **4 Tails**.





The **p-value** for getting **4 Heads** and **1 Tails** is...

- 1) The probability we randomly get **4 Heads** and **1 Tails**:

$$\frac{5}{32} + \frac{5}{32} +$$

- 2) The probability we randomly get something else that is equally rare:

- 3) The probability we randomly get something rarer or more extreme:

	TTHHH	TTTHH	
	THTHH	TTHTH	
HHHHH	THHTH	TTHHT	TTTTH
	THHHT	THTTH	TTTHT
TTHHH	HHTHH	THTHT	THTTT
HHTHH	HHTHH	THHTT	HTTTT
HHHTH	HHTHT	HHTTT	
HHHHT	HHHTH	HHTHT	TTTTT
	HHHTH	HTHTT	
	HHHTT	HHTTT	



The **p-value** for getting **4 Heads** and **1 Tails** is...

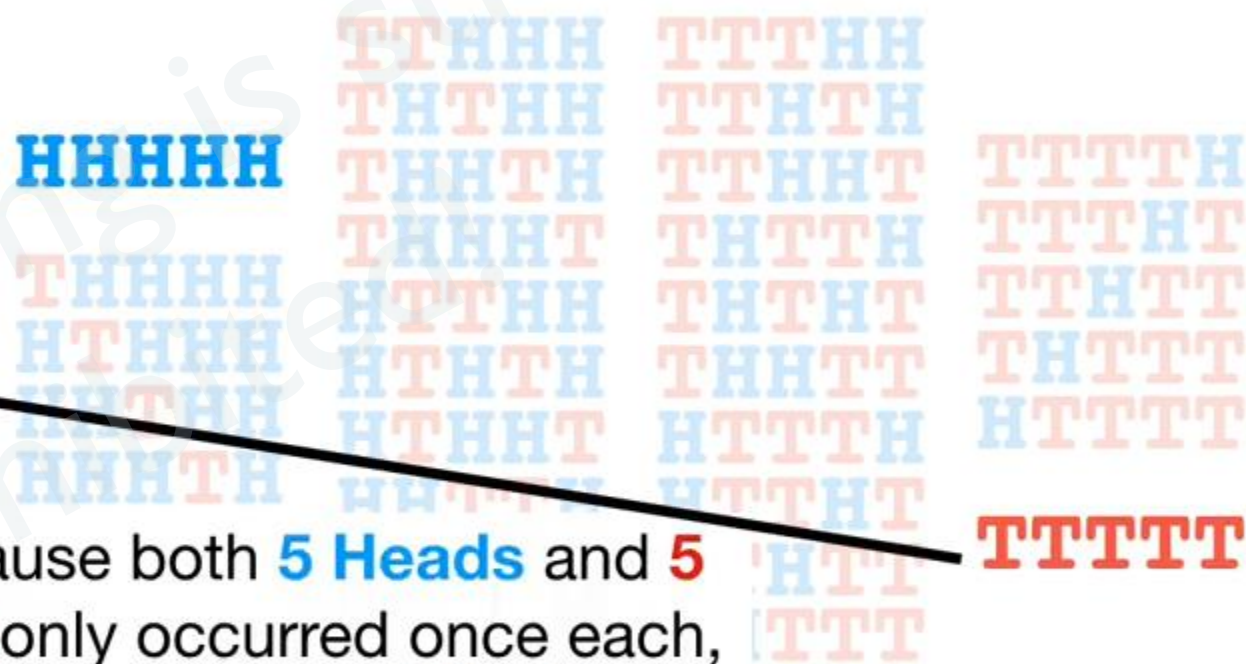
- 1) The probability we randomly get **4 Heads** and **1 Tails**:

$$\frac{5}{32} + \frac{5}{32} + \frac{2}{32}$$

- 2) The probability we randomly get something else that is equally rare:

- 3) The probability we randomly get something rarer or more extreme:

Because both **5 Heads** and **5 Tails** only occurred once each, they are rarer than **4 Heads** and **1 Tails**.

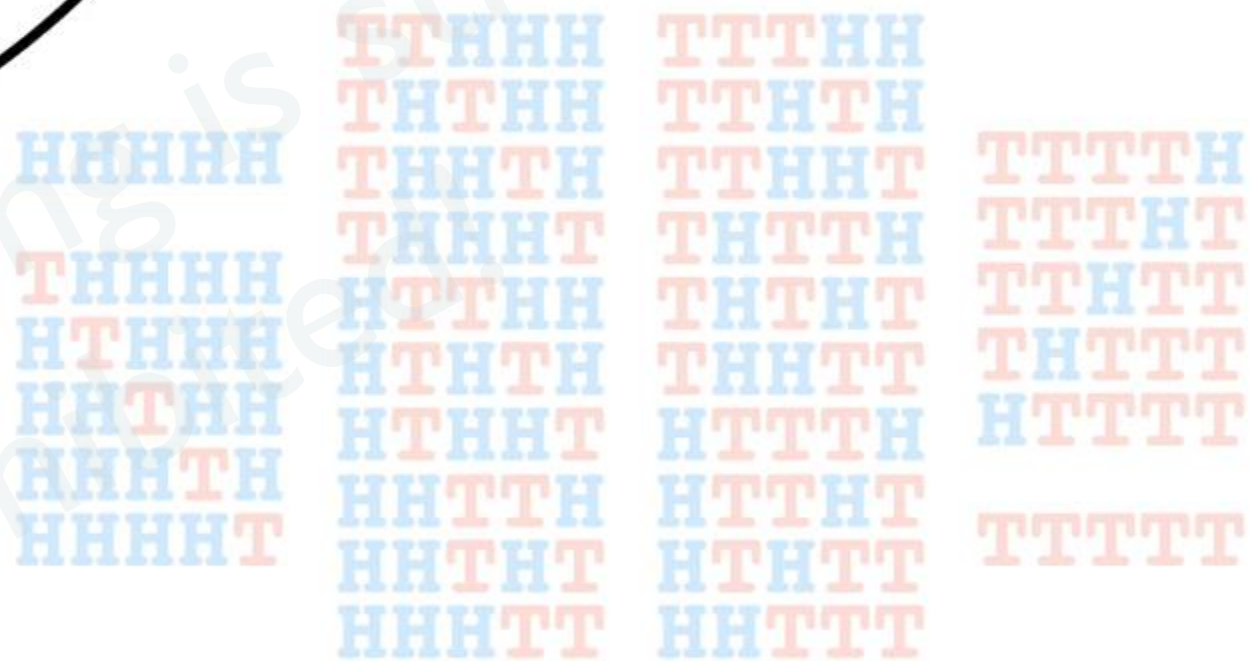


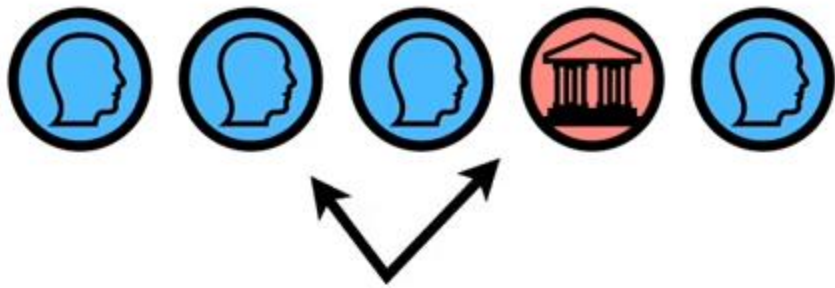


Even though I got **4 Heads**
and **1 Tails**, my coin is no
different from a normal coin.

Again, we typically only reject the
Null Hypothesis if the **p-value** is
less than **0.05...**

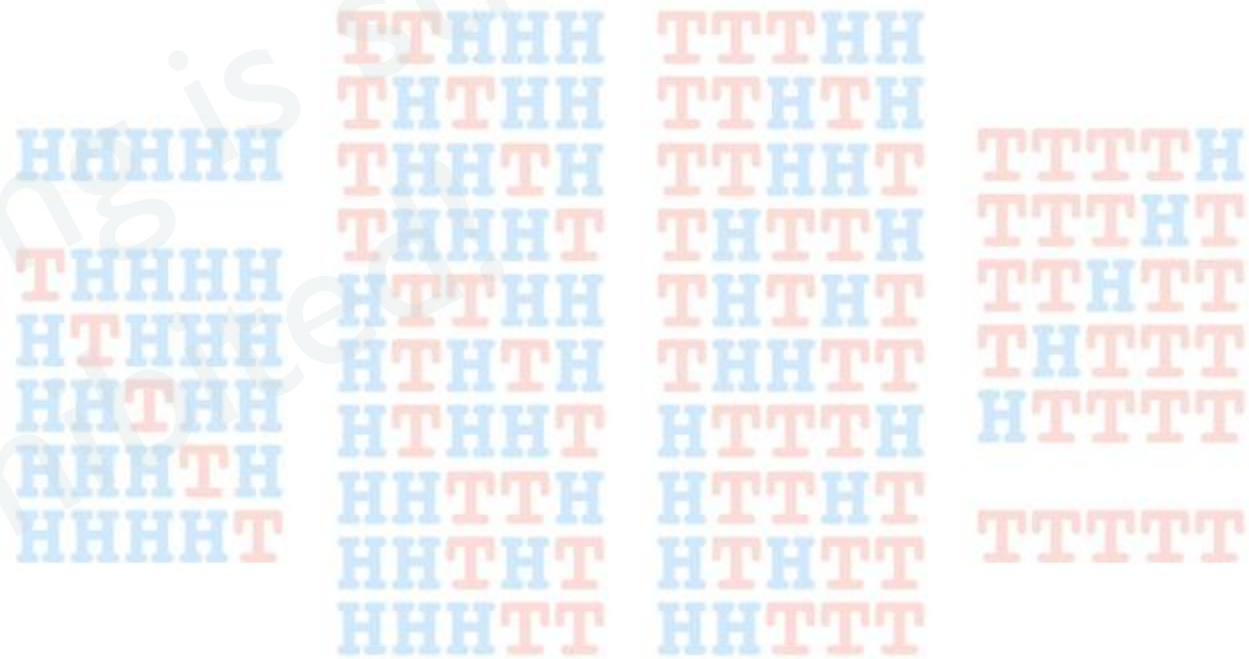
$$\frac{5}{32} + \frac{5}{32} + \frac{2}{32} = 0.375$$





In other words, the data, getting **4 Heads** and **1 Tail**, did not convince us that our coin was special.

$$\frac{5}{32} + \frac{5}{32} + \frac{2}{32} = 0.375$$



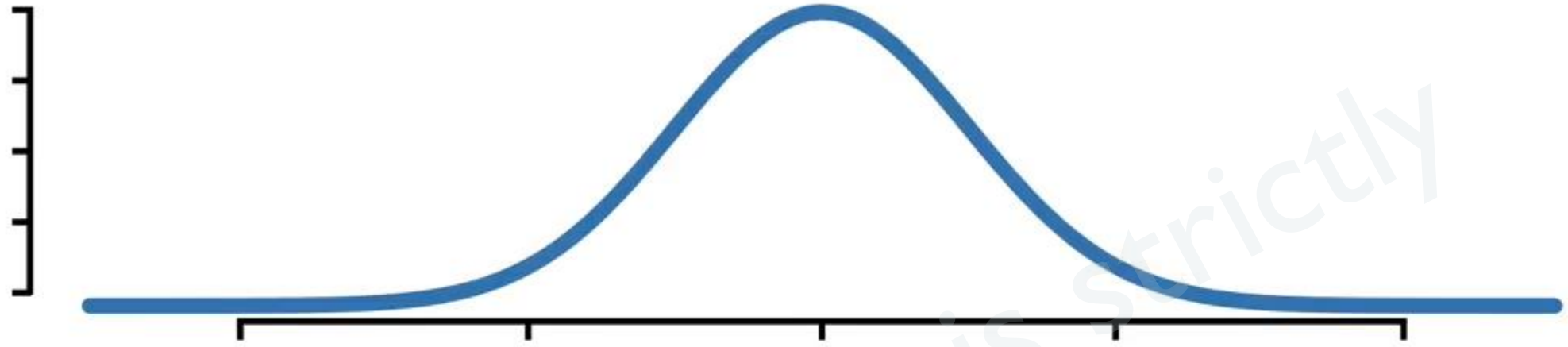
But what if we wanted to calculate **probabilities** and **p-values** for how tall or short people are?



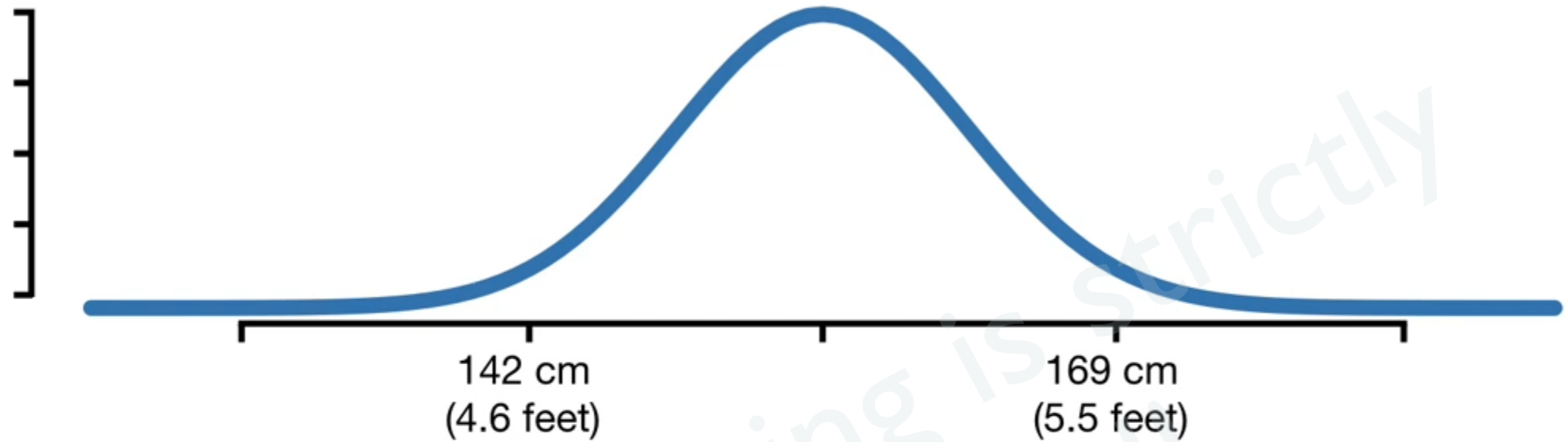
Draft: Sharing is strictly Prohibited!

In theory, we could try to list every single possible value for height.

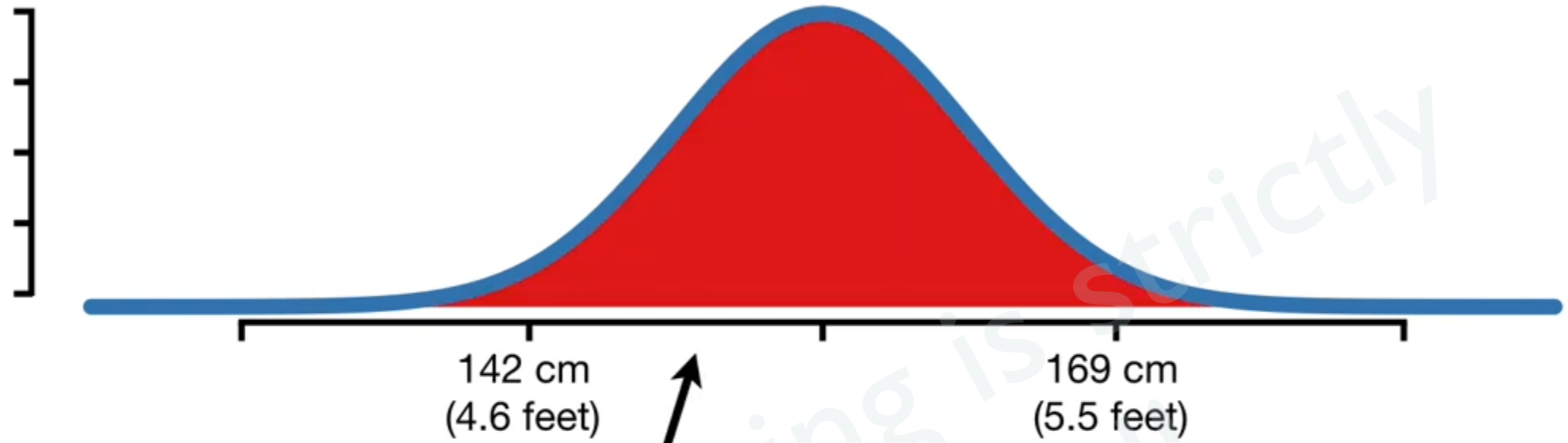
152.4 cm	152.9 cm	153.4 cm	etc...
152.5 cm	153.0 cm	153.5 cm	...
152.6 cm	153.1 cm	153.6 cm	etc...
152.7 cm	153.2 cm	153.6 cm	...
152.8 cm	153.3 cm	153.8 cm	etc...



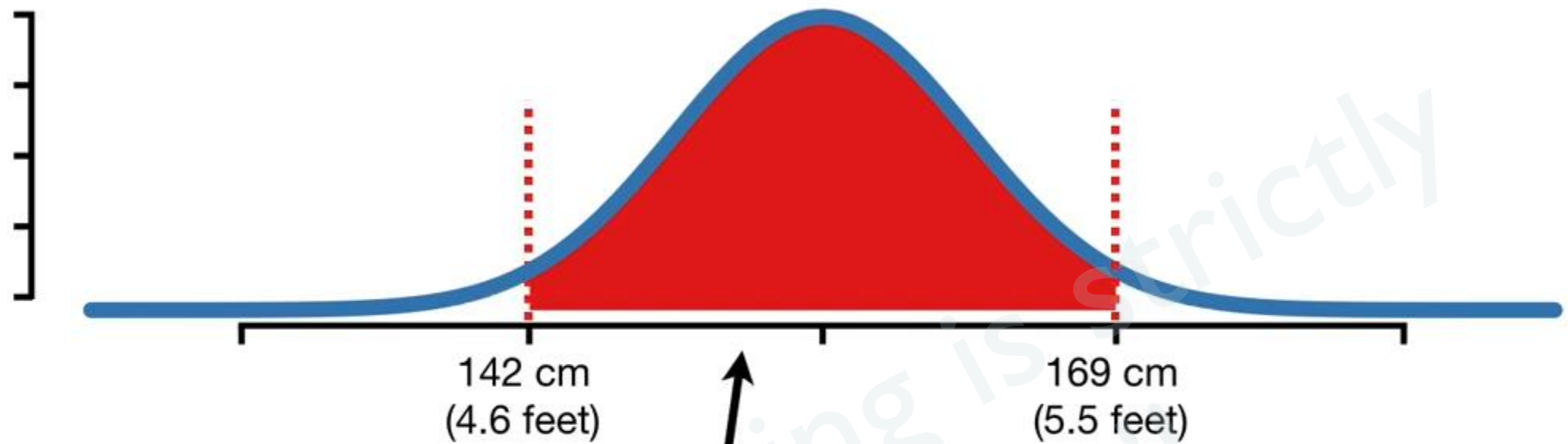
However, in practice, when we calculate **probabilities** and **p-values** for something continuous, like **Height**, we usually use something called a *statistical distribution*.



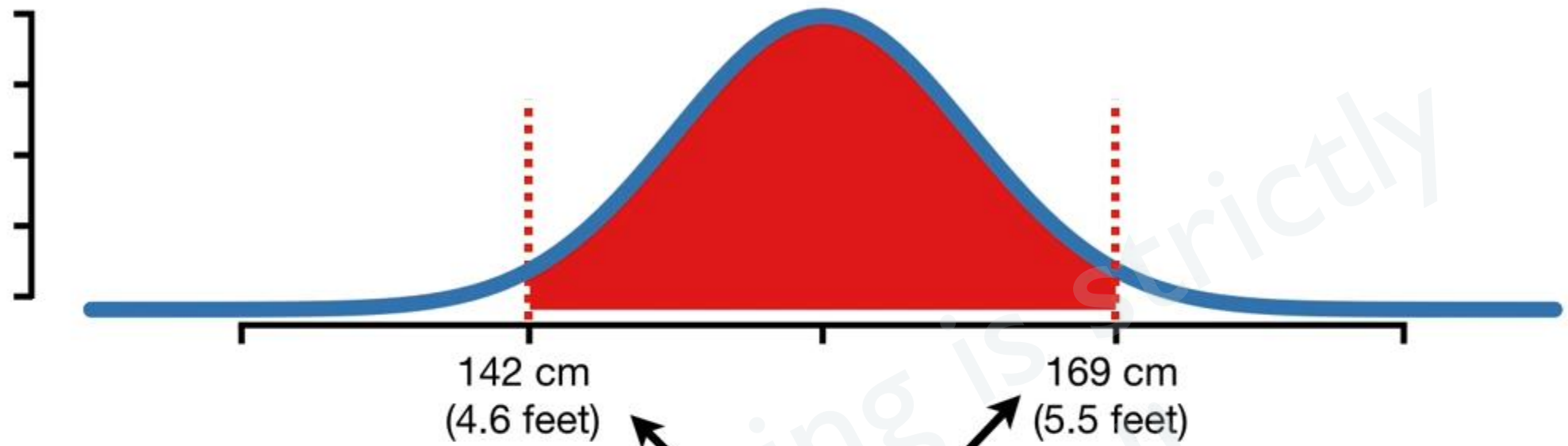
Here we have a distribution of height measurements from Brazilian women between **15** and **49** years old taken in **1996**.



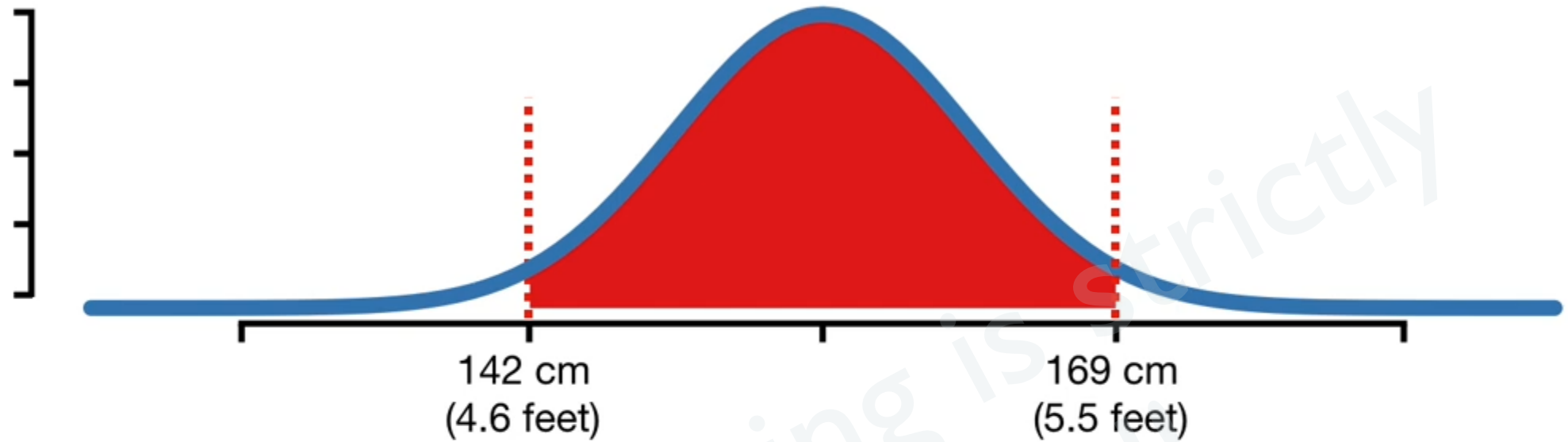
The **red area** under the curve indicates the probability that a person's height will be within a range of possible values.



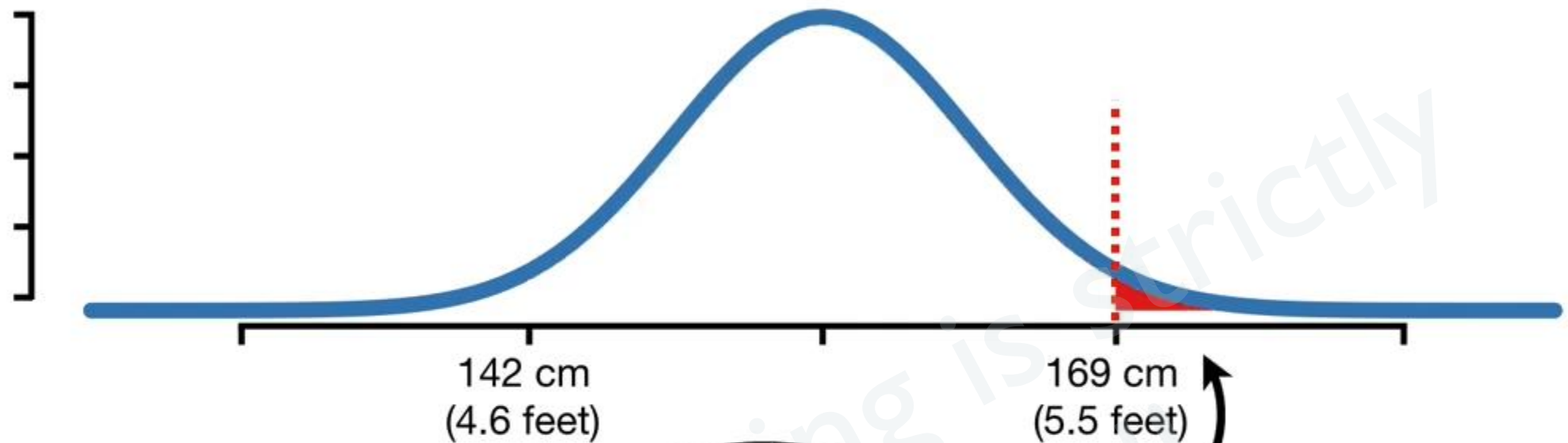
For example, **95%** of the area under the curve is between **142** and **169**...



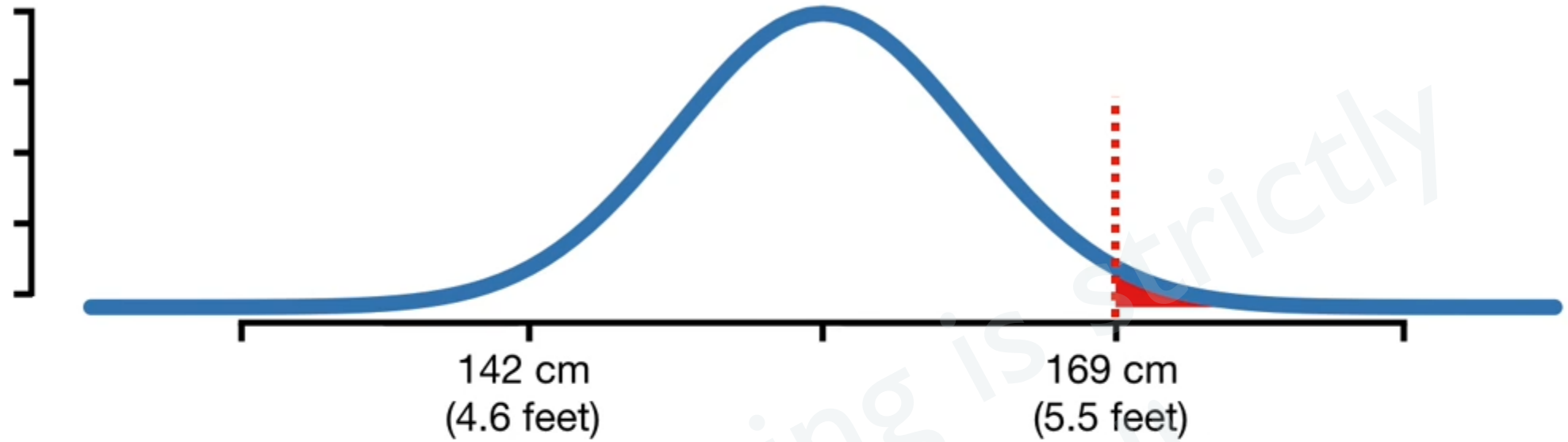
...and that means that **95%** of the Brazilian women were between **142** and **169** cm tall.



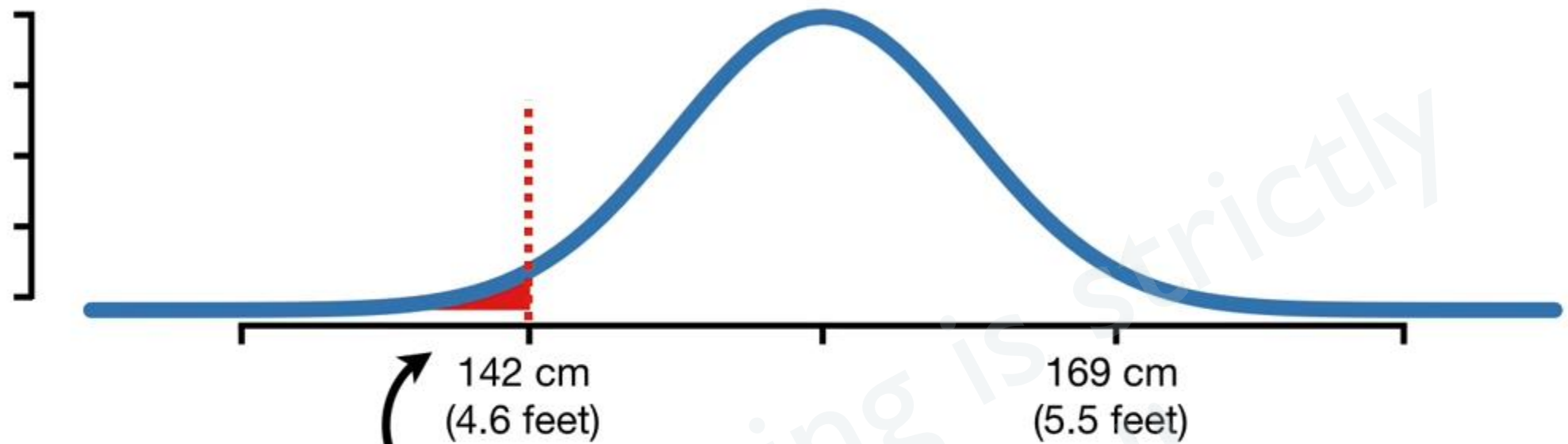
In other words, there is a **95%** probability that each time we measure a Brazilian woman, their height will be between **142** and **169** cm.



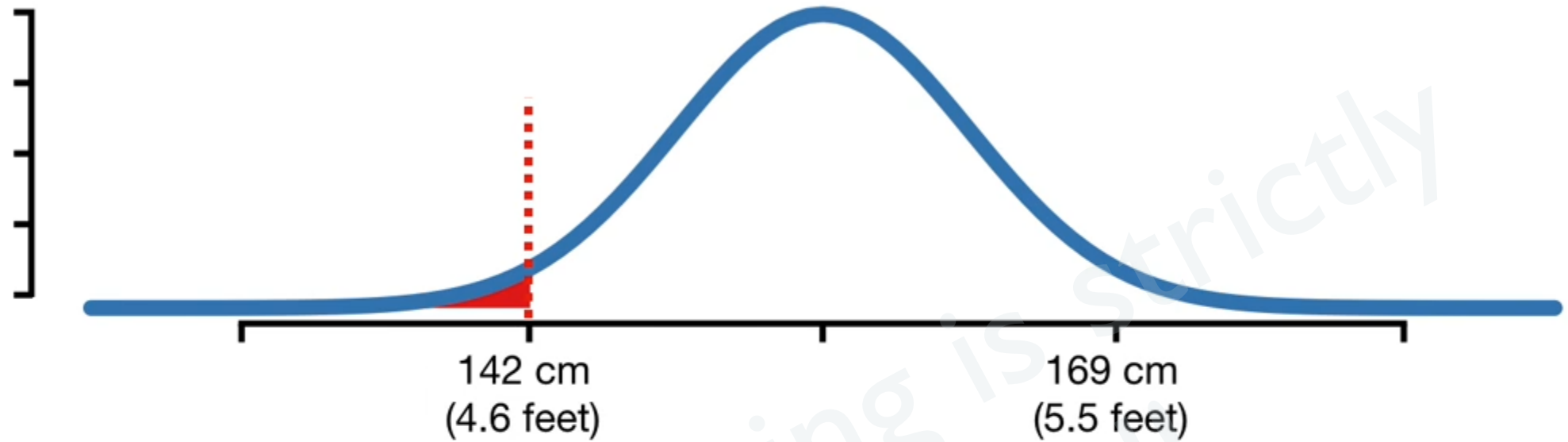
2.5% of the total area under the curve is greater than **169**.



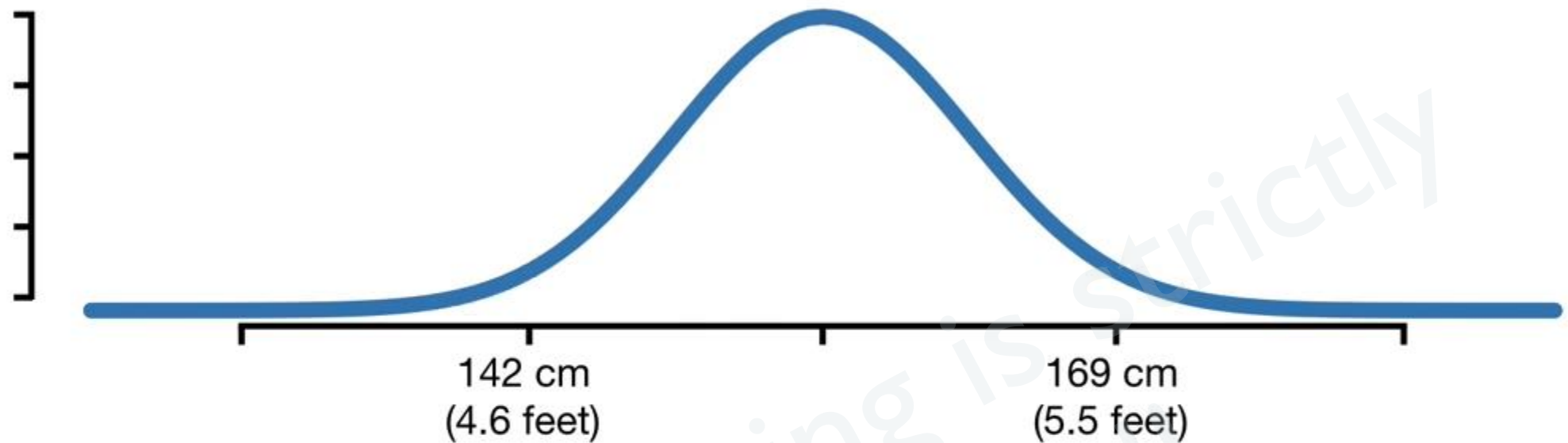
And that means there is a **2.5%** probability that each time we measure a Brazilian woman, their height will be *greater* than **169** cm.



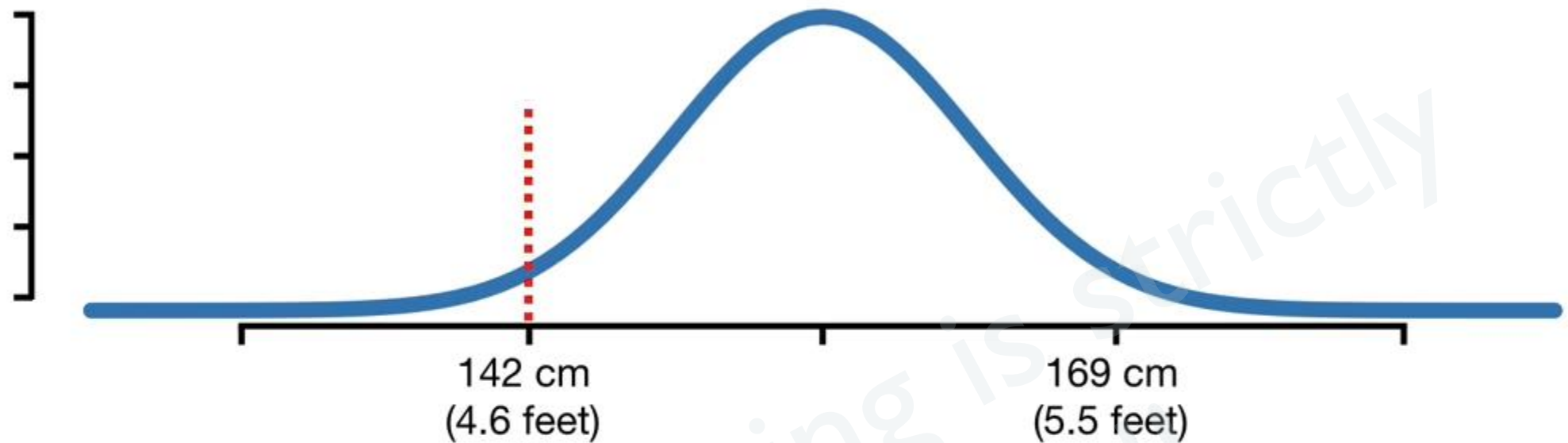
Likewise, **2.5%** of the total area under the curve is less than **142**.



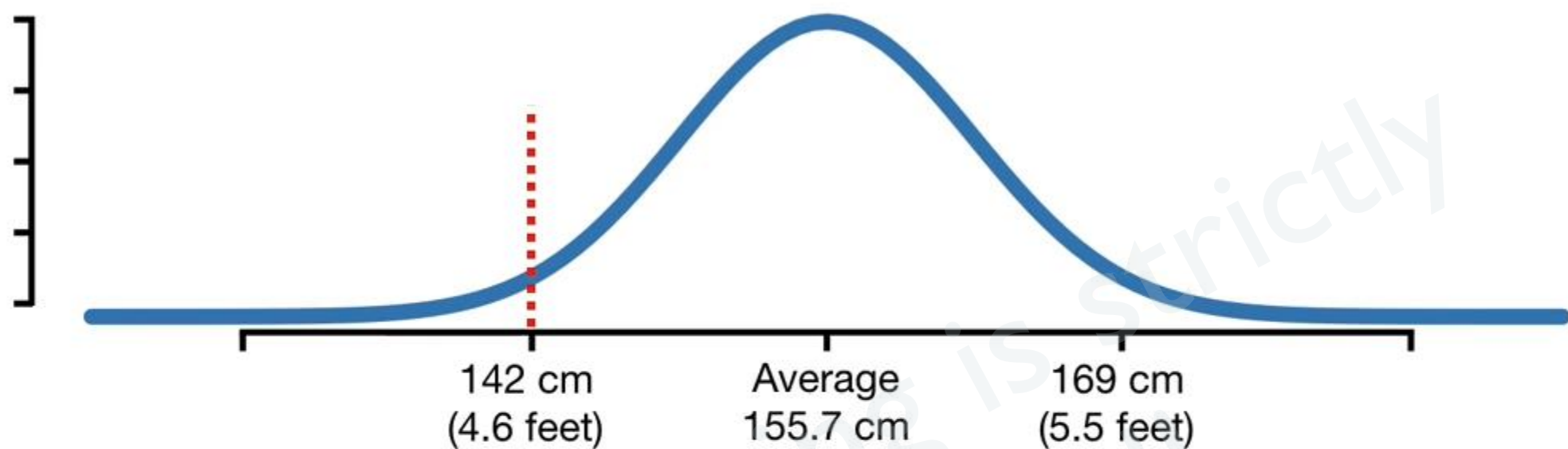
Thus, there is a **2.5%** probability that each time we measure a Brazilian woman, their height will be *less than* **142** cm.



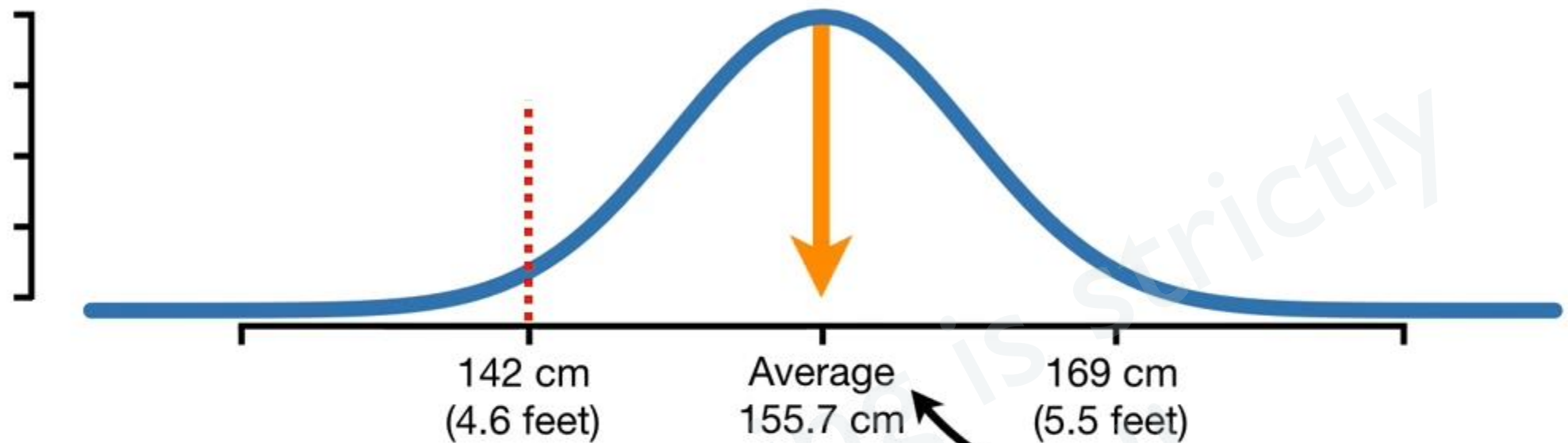
To calculate **p-values** with a distribution, you add up the percentages of area under the curve.



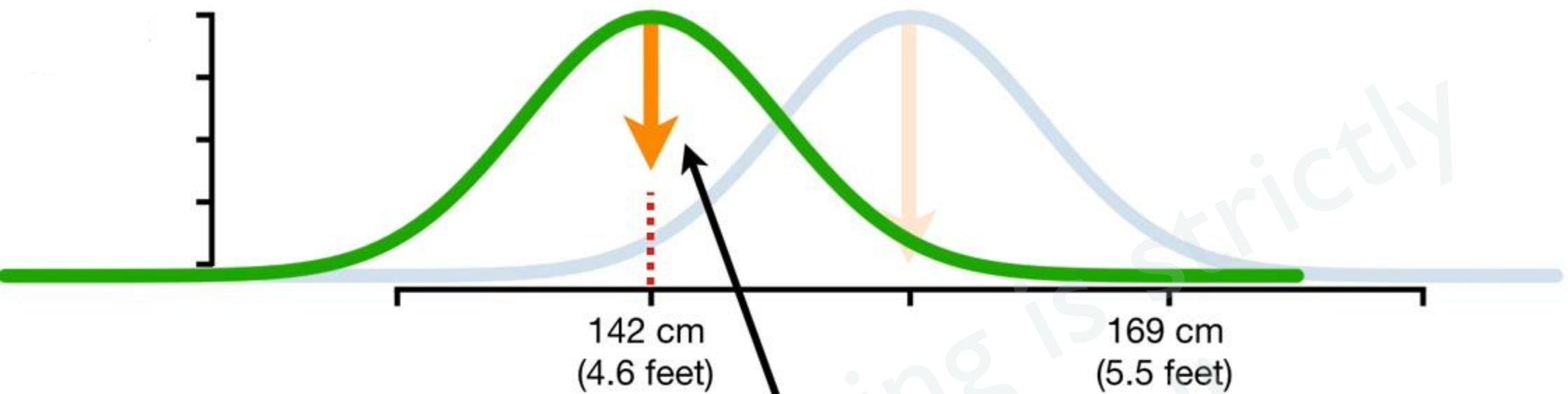
For example, imagine we measured someone who was **142** cm tall.



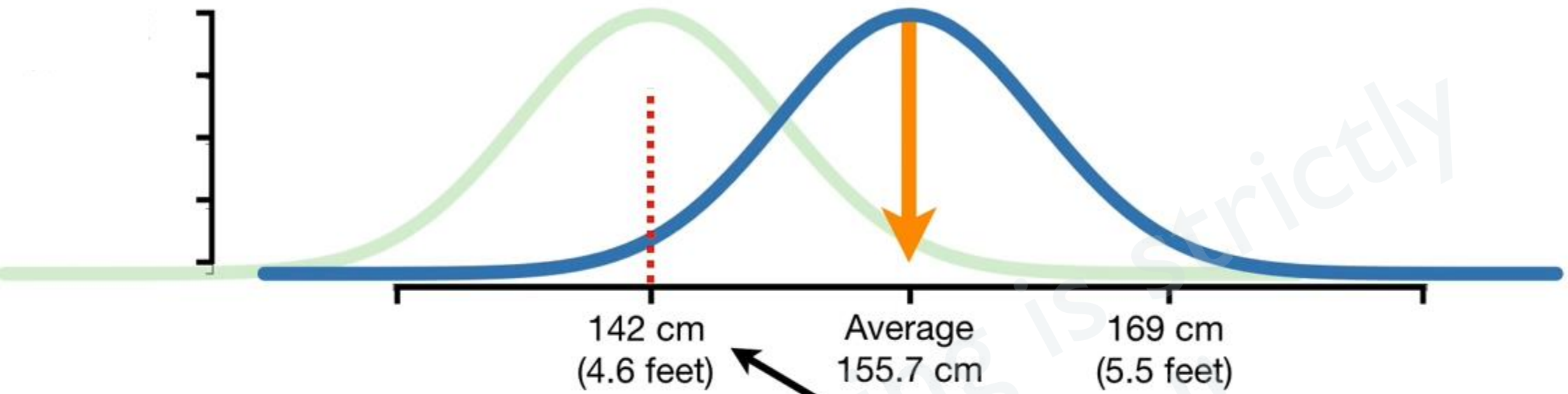
If we measured someone who was **142** cm tall, we might wonder if it came from this distribution heights, which has an average value of **155.7...**



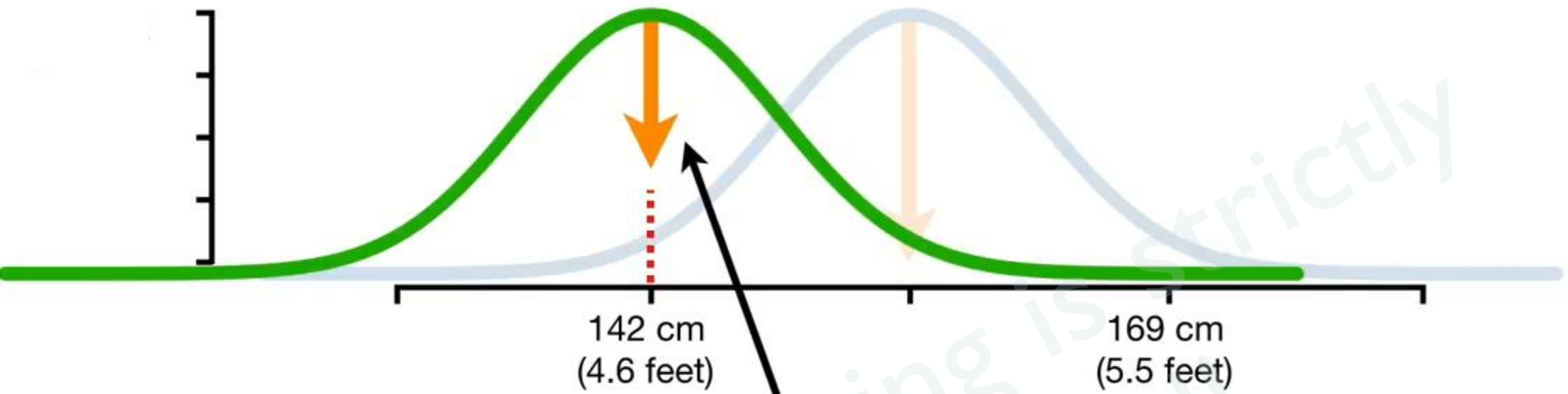
If we measured someone who was **142** cm tall, we might wonder if it came from this distribution heights, which has an average value of **155.7...**



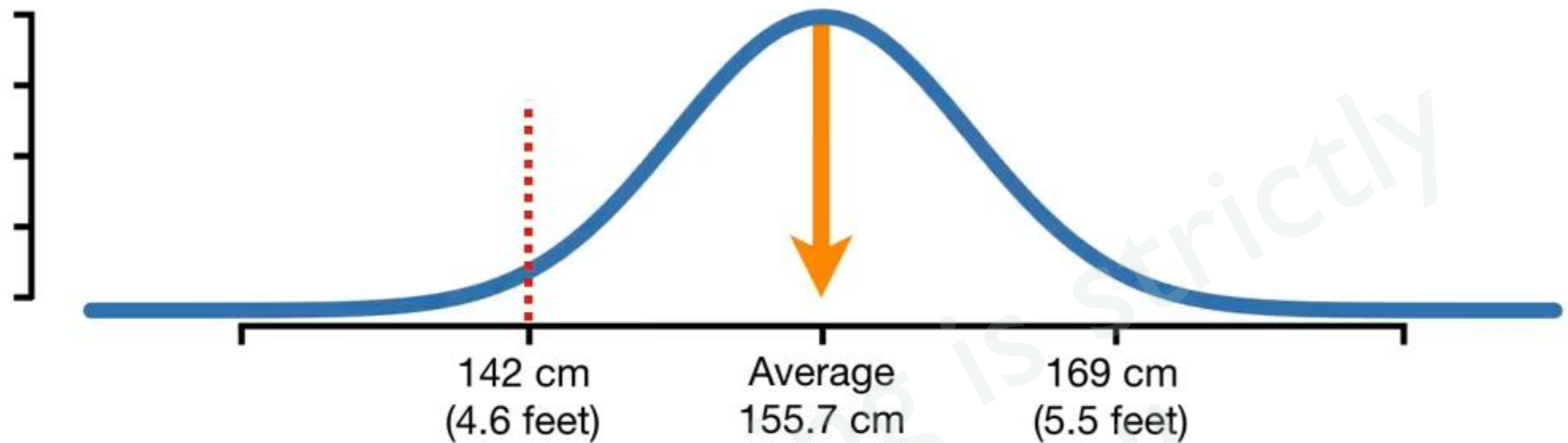
...or of it came from another distribution of heights, for example this **green distribution** has an average value of **142**.



So the question is, “is this measurement, **142 cm**, so far away from the mean of the **blue distribution (155.7 cm)** that we can reject the idea that it came from it?”

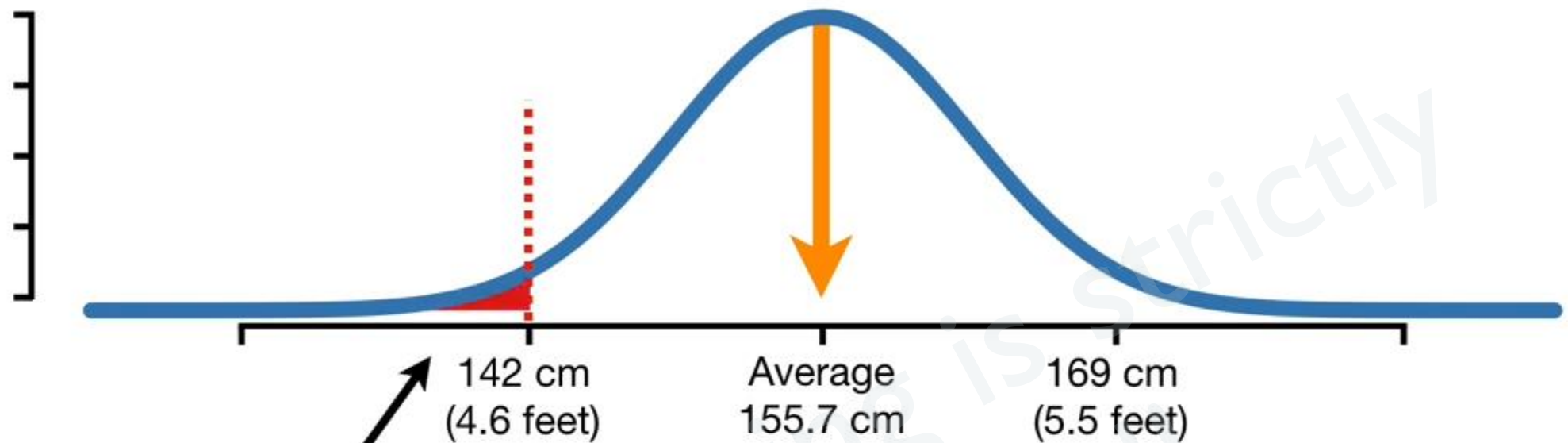


If so, then that would suggest that another distribution, like this **green one**, might do a better job explaining the data



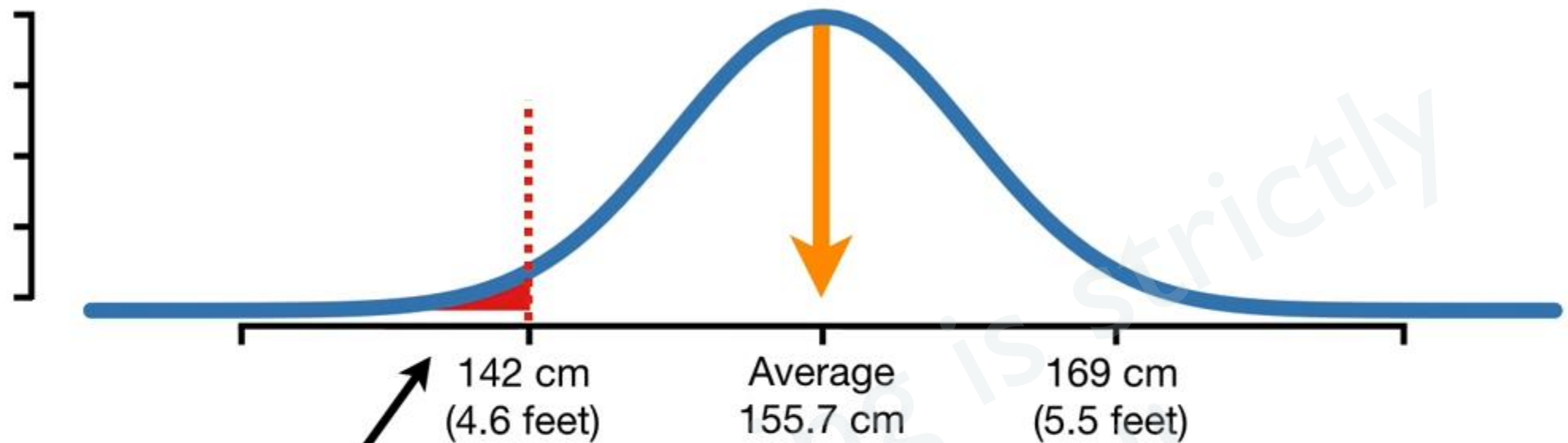
The **p-value** for the hypothesis “This measurement comes from the **blue distribution**” ...

p-value for **142 cm** given =
the **blue distribution**



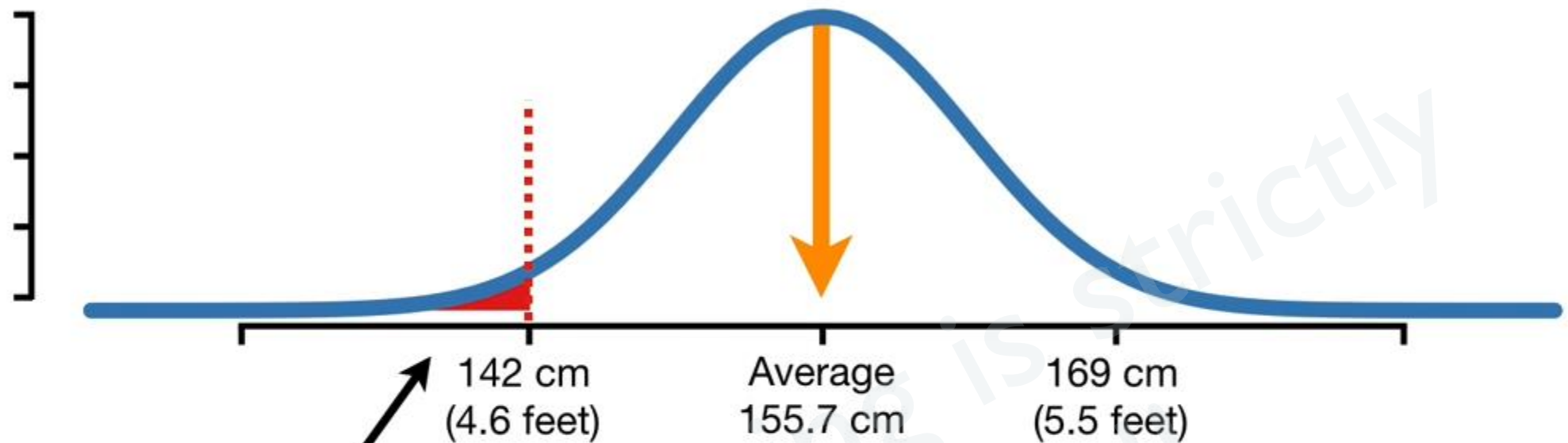
...starts with the **2.5%** of the area for people less than or equal to **142** cm.

p-value for **142** cm given the **blue distribution** = 0.025



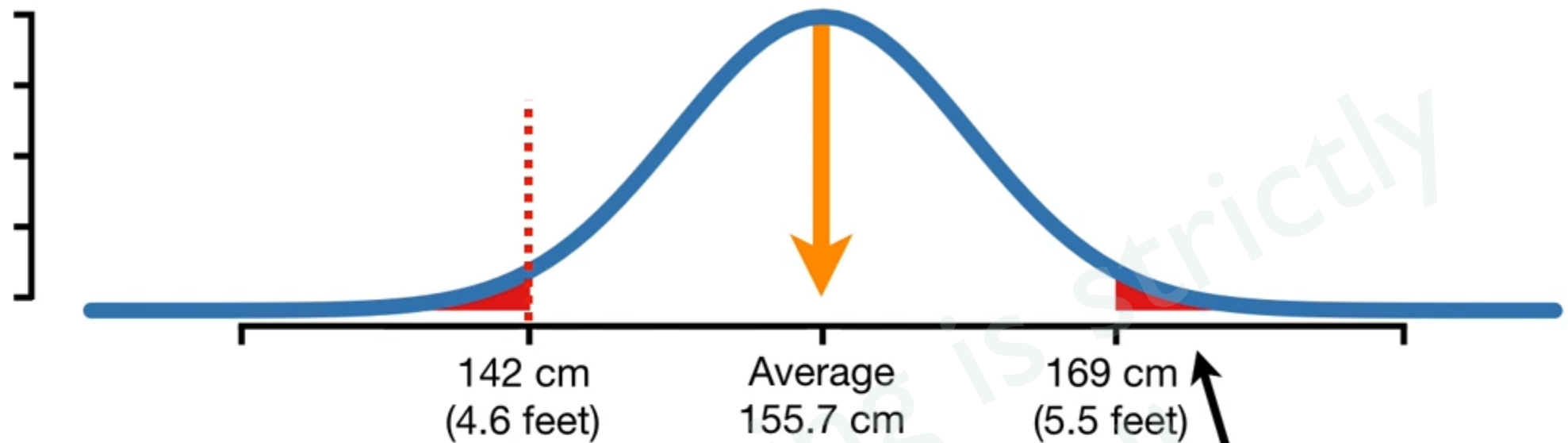
NOTE: When we are working with a distribution, we are interested in adding **more extreme** values to the **p-value** rather than **rarer** values.

p-value for 142 cm given the blue distribution = 0.025



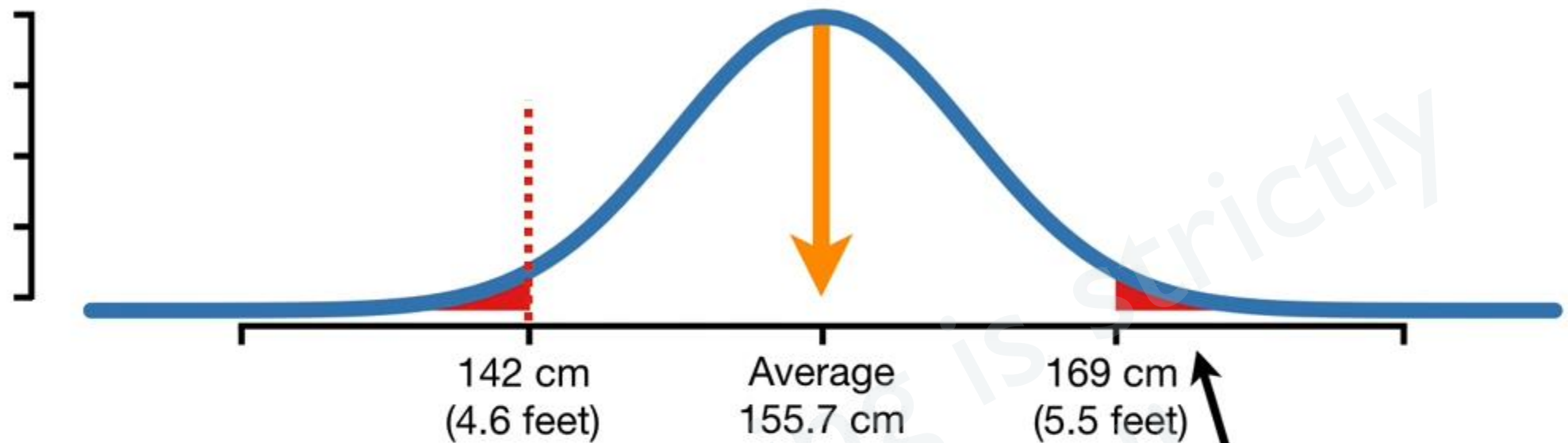
In this case, all heights further than **142** cm from the mean (**155.7**) are considered **more extreme** than what we observed.

p-value for 142 cm given the blue distribution = 0.025



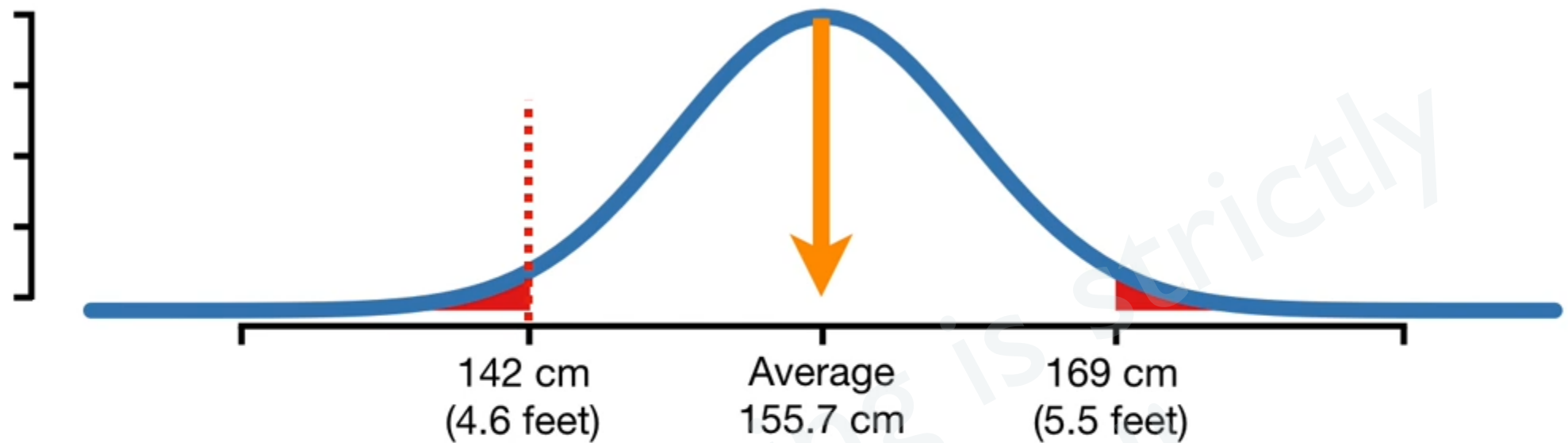
We also add the **2.5%** of the area for people **169** cm or taller.

p-value for **142** cm given the **blue distribution** = $0.025 + 0.025$



NOTE: Just like on the other side of the distribution, these values are considered **equal to or more extreme** because they are as far from the mean (**155.7**), or further.

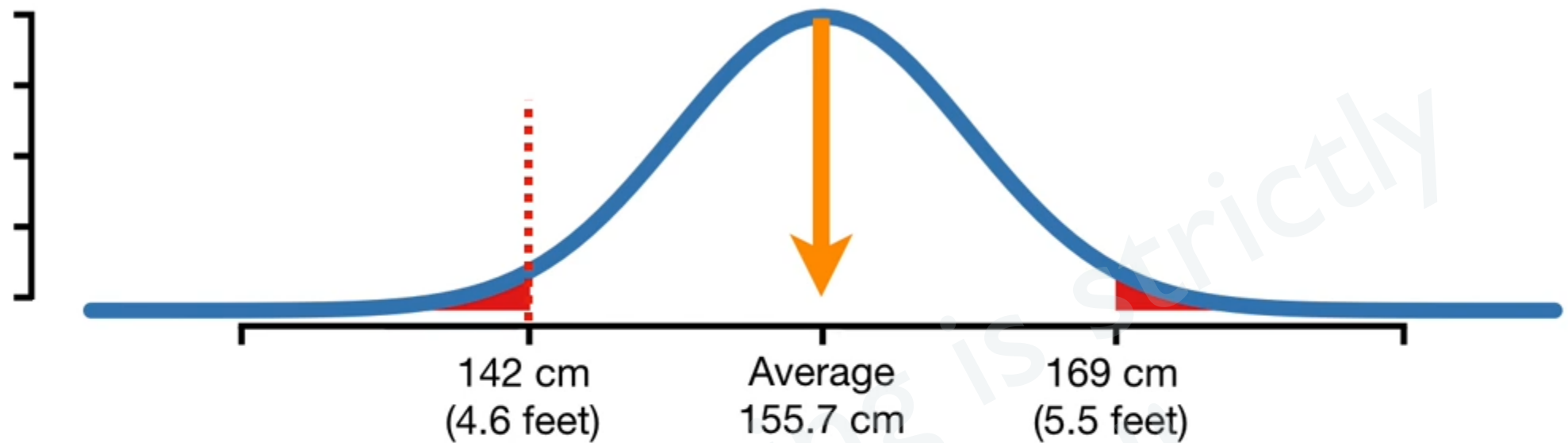
p-value for **142** cm given the **blue distribution** = $0.025 + 0.025$



Now we just do the math...

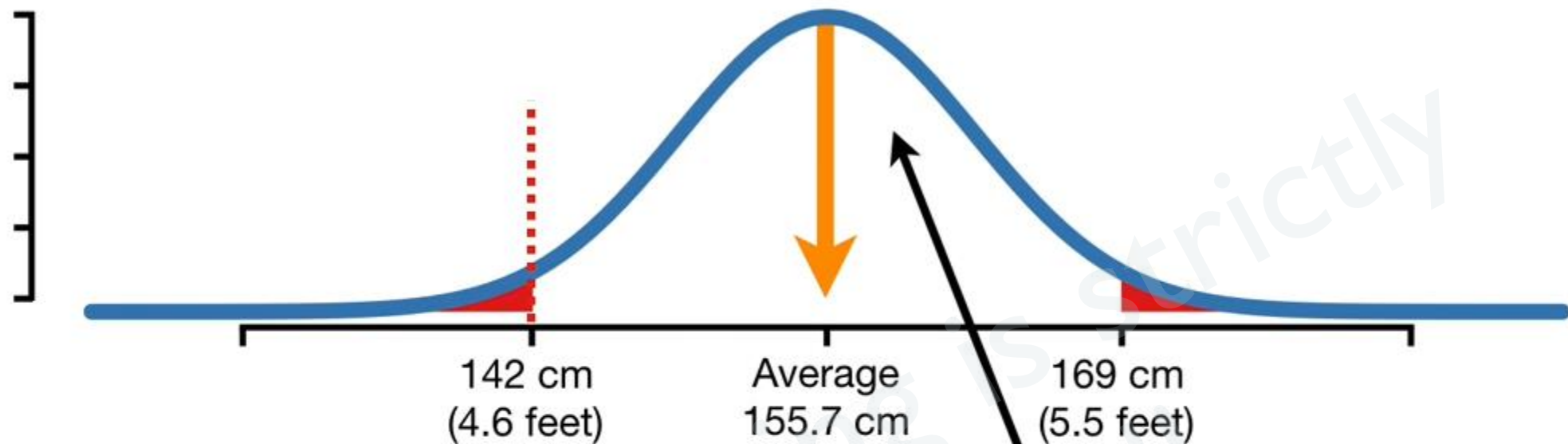


p-value for **142** cm given
the **blue distribution** = $0.025 + 0.025$



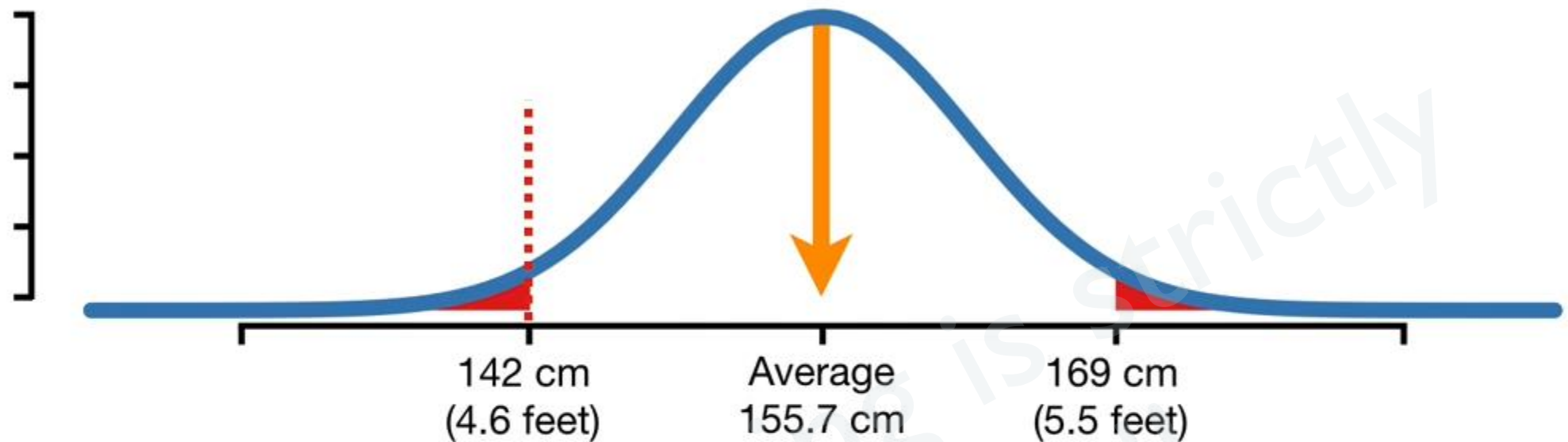
...and get **0.05**.

p-value for **142** cm given the **blue distribution** = $0.025 + 0.025 = 0.05$



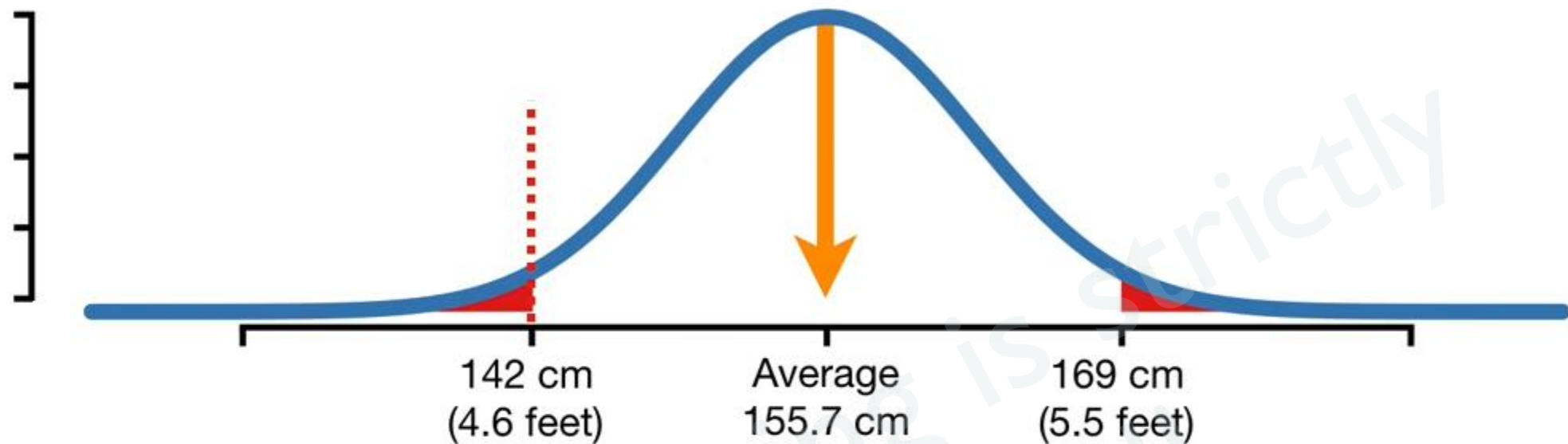
So the **p-value** for the hypothesis “Someone **142** cm tall could come from the **blue distribution**” is **0.05**.

p-value for **142** cm given the **blue distribution** = $0.025 + 0.025 = 0.05$



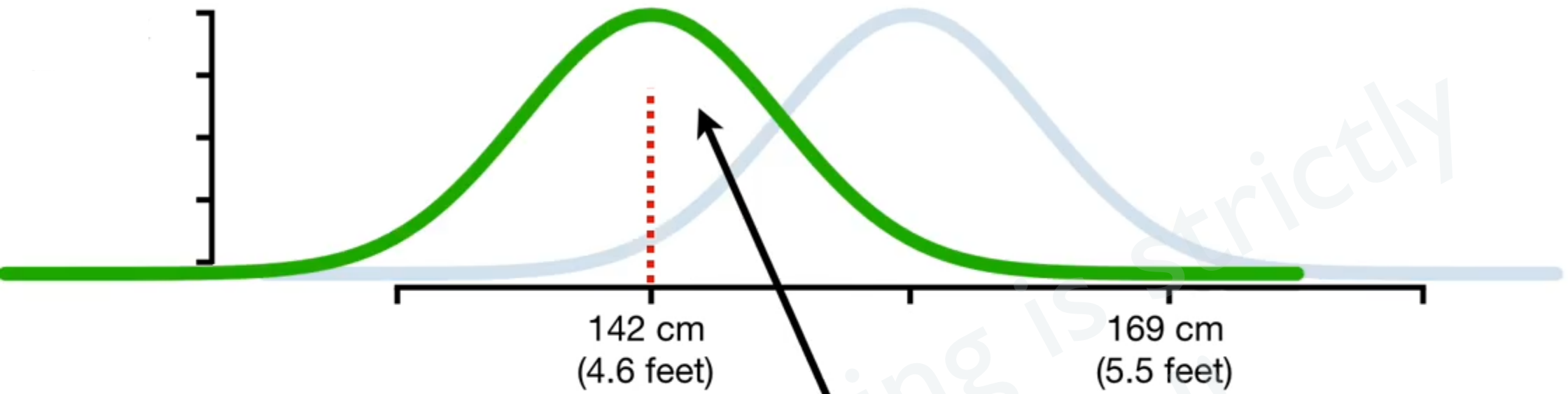
And since the cutoff for significance is usually **0.05**, we would say...

p-value for **142** cm given the **blue distribution** = $0.025 + 0.025 = 0.05$



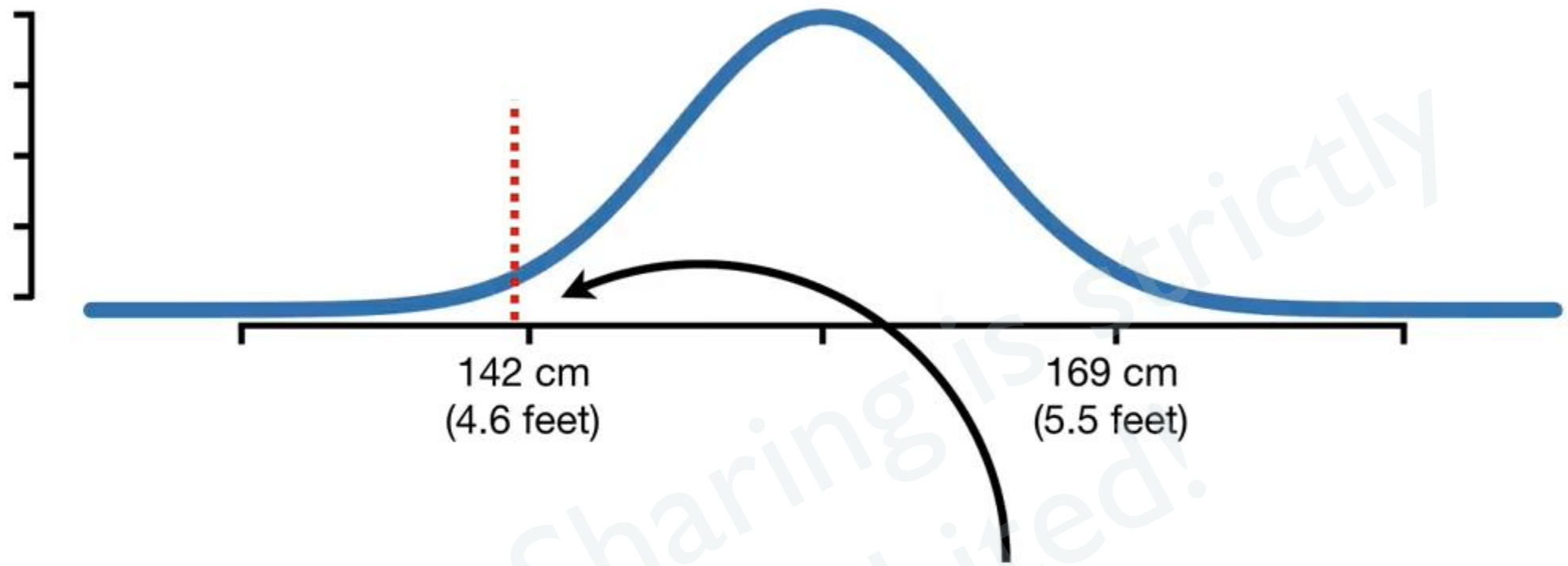
“Hmmm. Maybe it could come from this distribution, maybe not. It’s hard to tell since the **p-value** is right on the borderline.”

p-value for **142** cm given the **blue distribution** = $0.025 + 0.025 = 0.05$

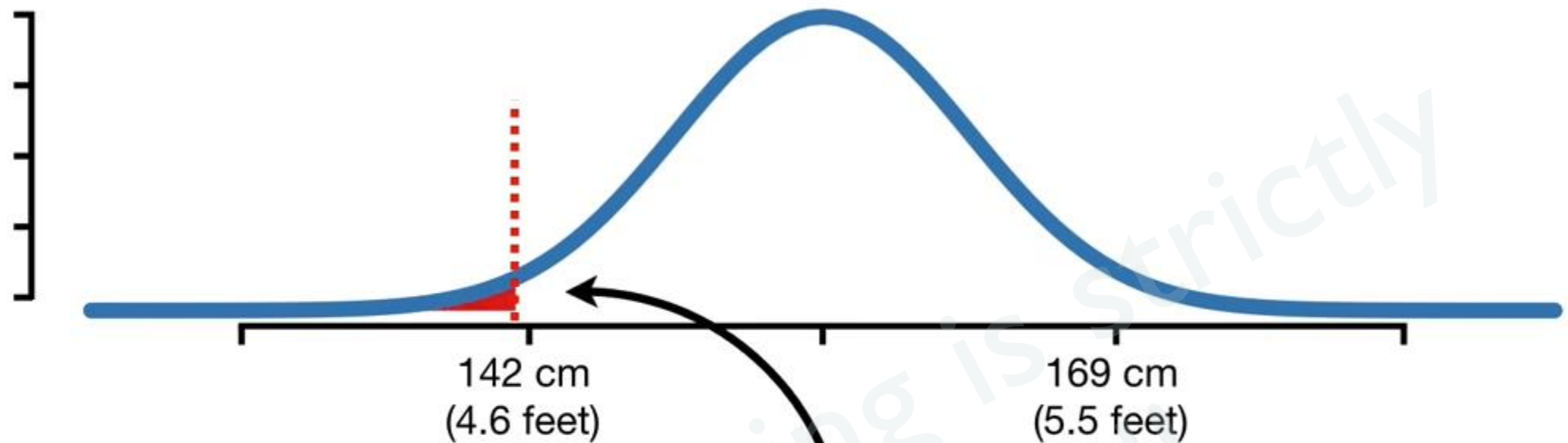


...or maybe they come from this distribution.
The data are inconclusive.

p-value for **142** cm given
the **blue distribution** = $0.025 + 0.025 = 0.05$

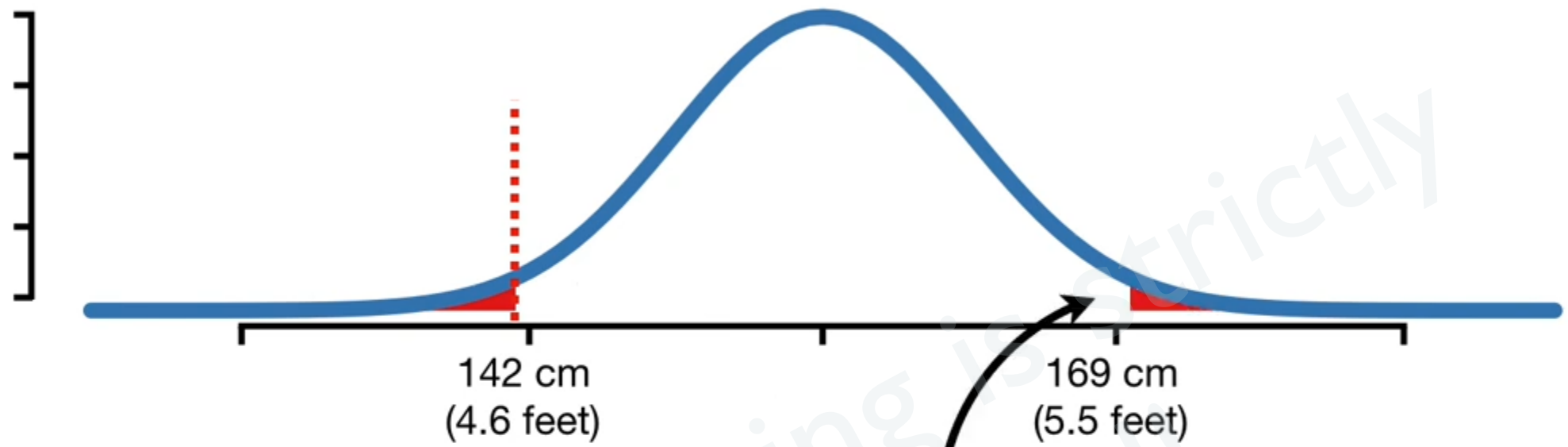


NOTE: If we had measured someone who was **141** cm tall, so just a little bit shorter than **142** cm...



...then the **p-value**
would be **0.016**...

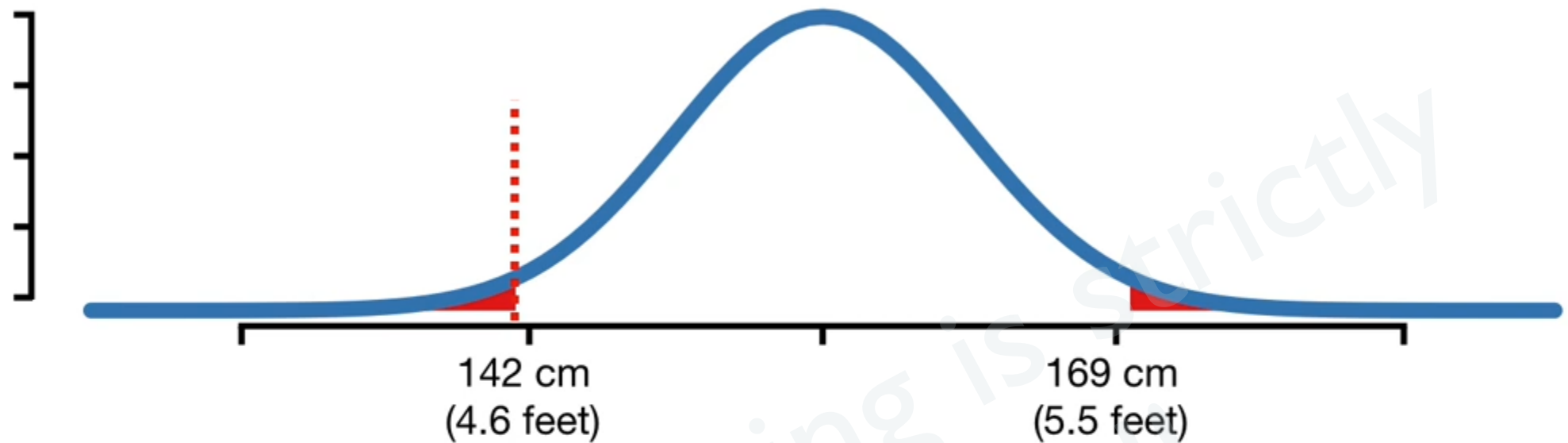
p-value for **141** cm given
the **blue distribution** = 0.016



Draft: Sharing is Strictly Prohibited!

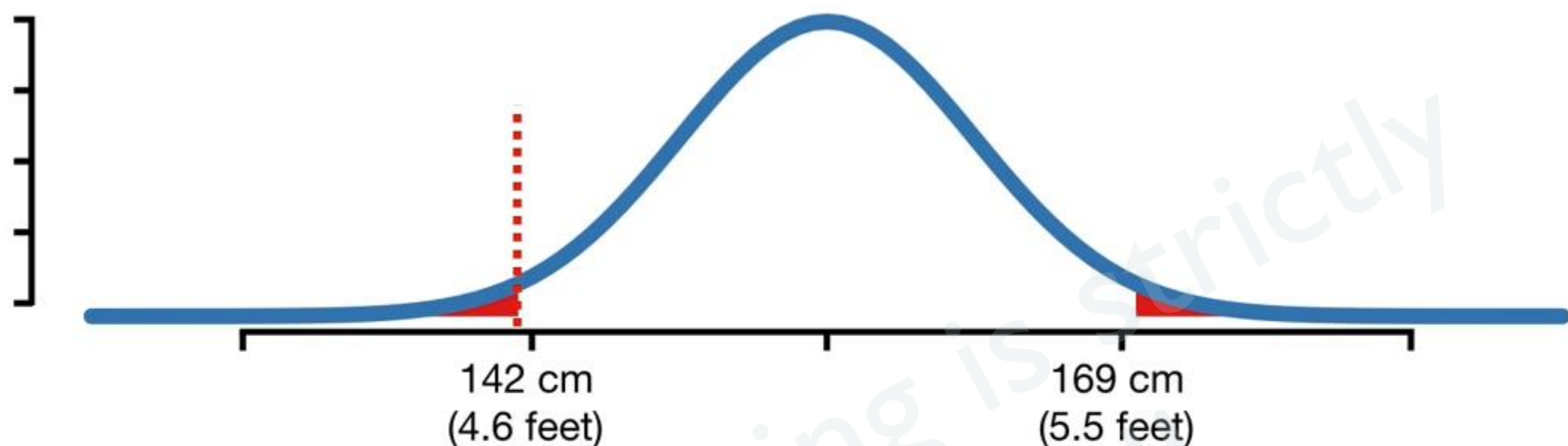
...plus **0.016**...

p-value for **141** cm given the **blue distribution** = $0.016 + 0.016$



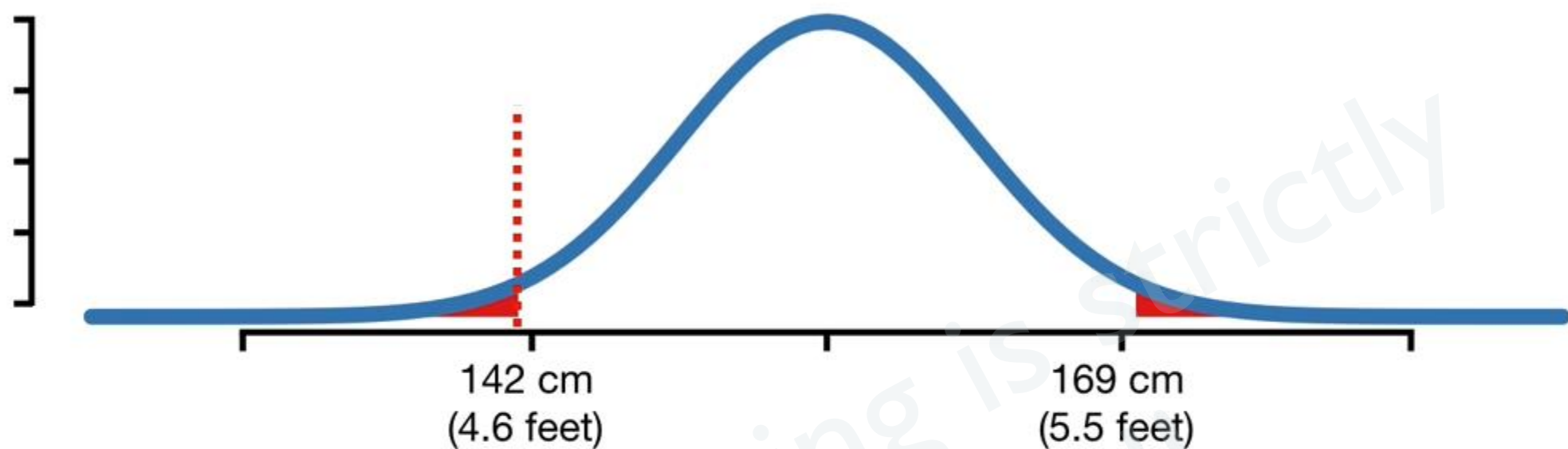
...which equals **0.03**.

p-value for **141** cm given
the **blue distribution** = $0.016 + 0.016 = 0.03$



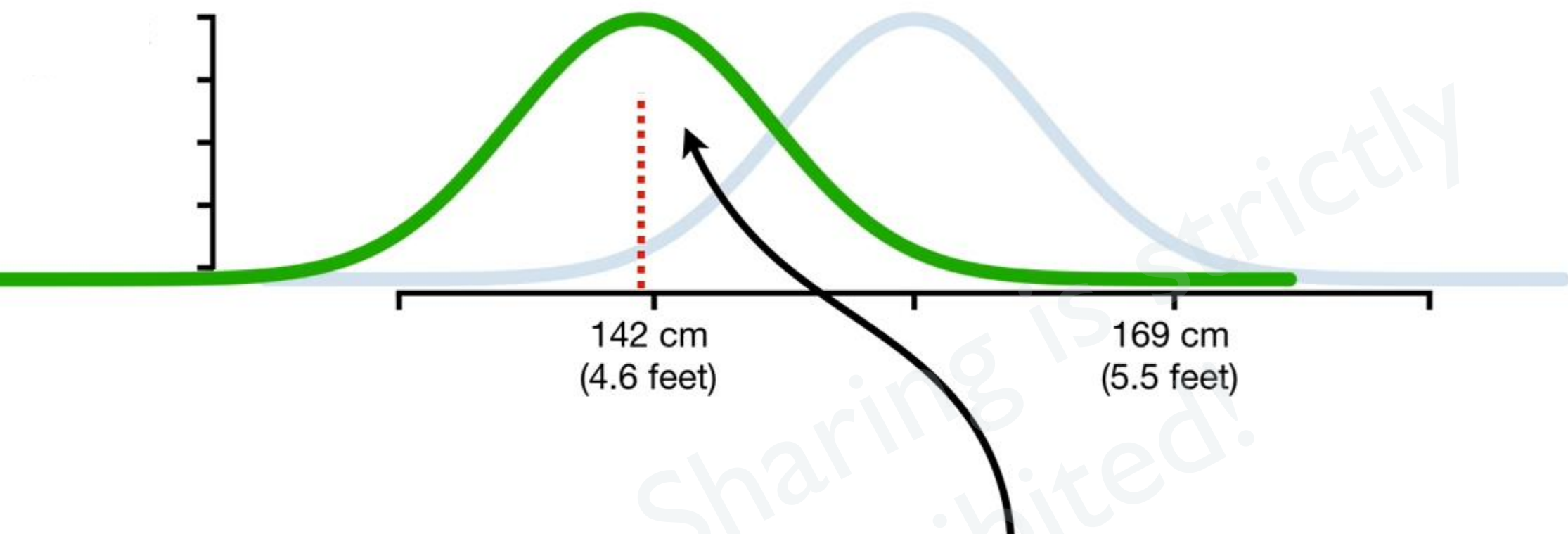
And since $0.03 < 0.05$, the standard threshold, we can reject the hypothesis that, given the **blue distribution**, it is normal to measure someone **141** cm tall.

p-value for **141** cm given the **blue distribution** = $0.016 + 0.016 = 0.03$



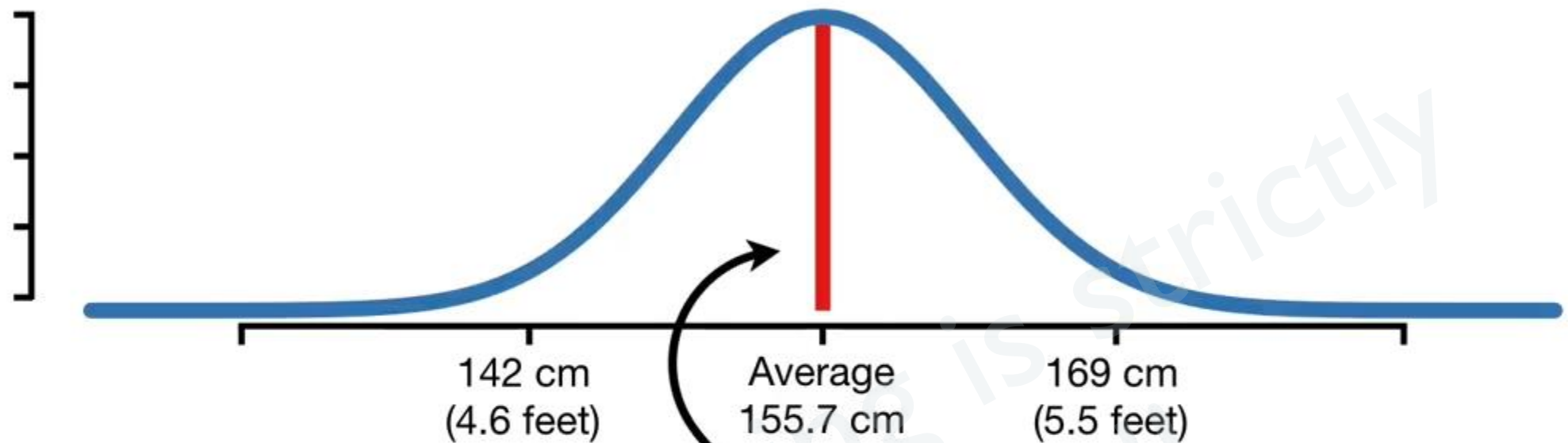
Thus, we will conclude that it's pretty special to measure someone that short.

p-value for **141** cm given the **blue distribution** = $0.016 + 0.016 = 0.03$

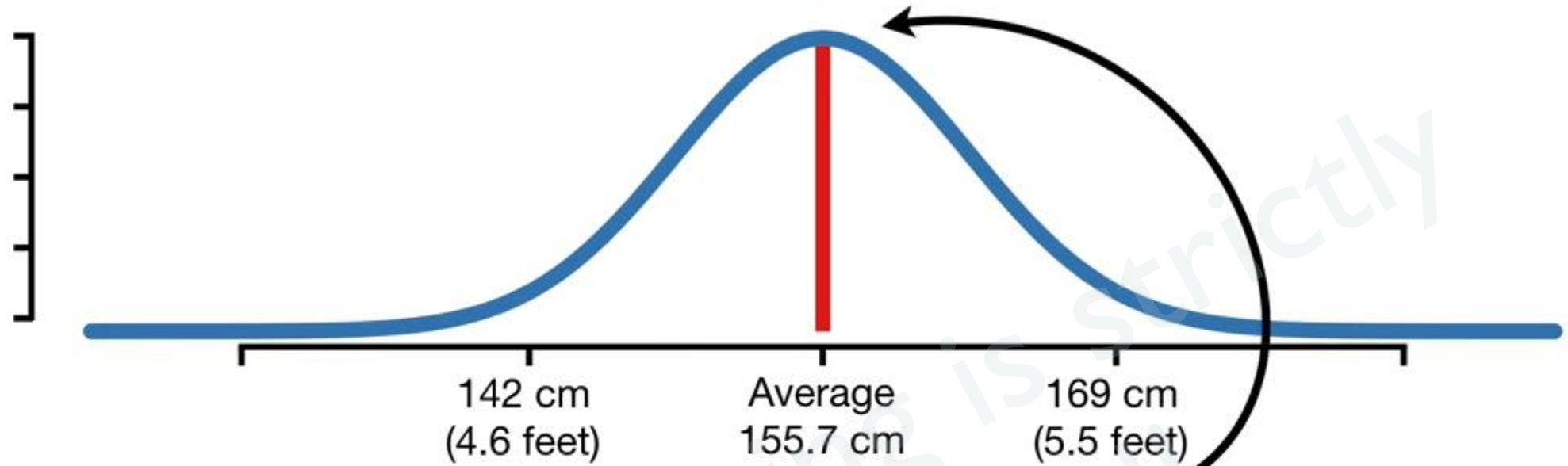


And that suggests that a different distribution of heights makes more sense.

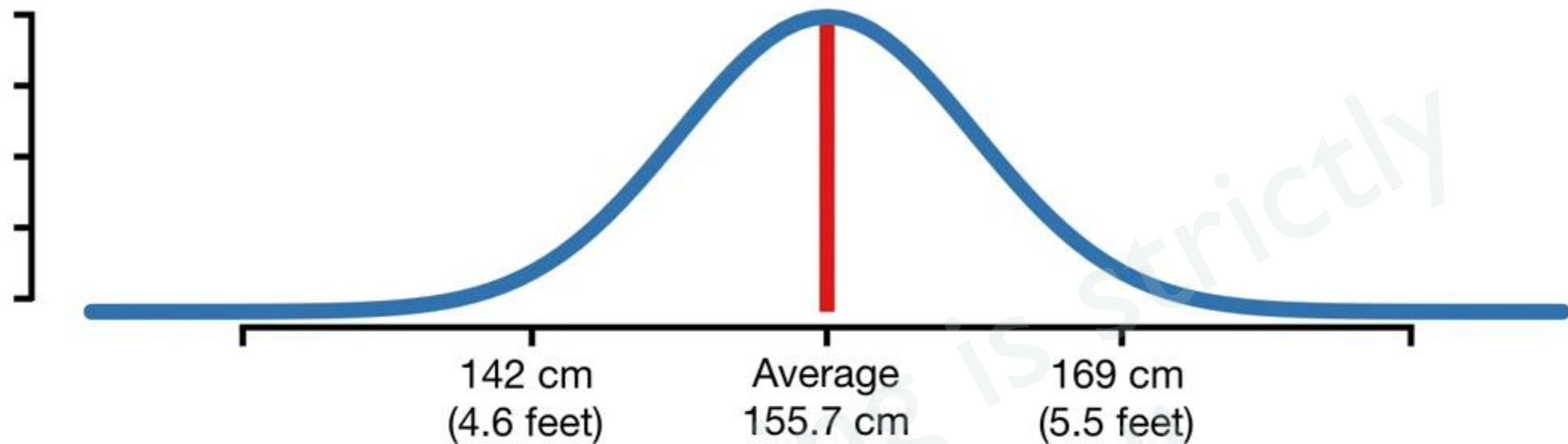
p-value for **141** cm given the **blue distribution** = $0.016 + 0.016 = 0.03$



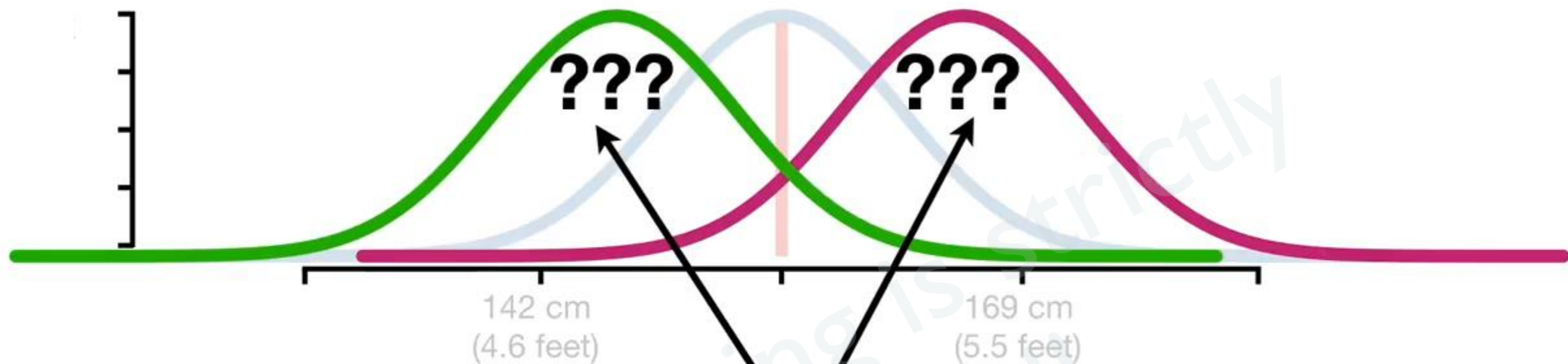
Now, what if we measured someone who is between **155.4** and **156** cm tall?



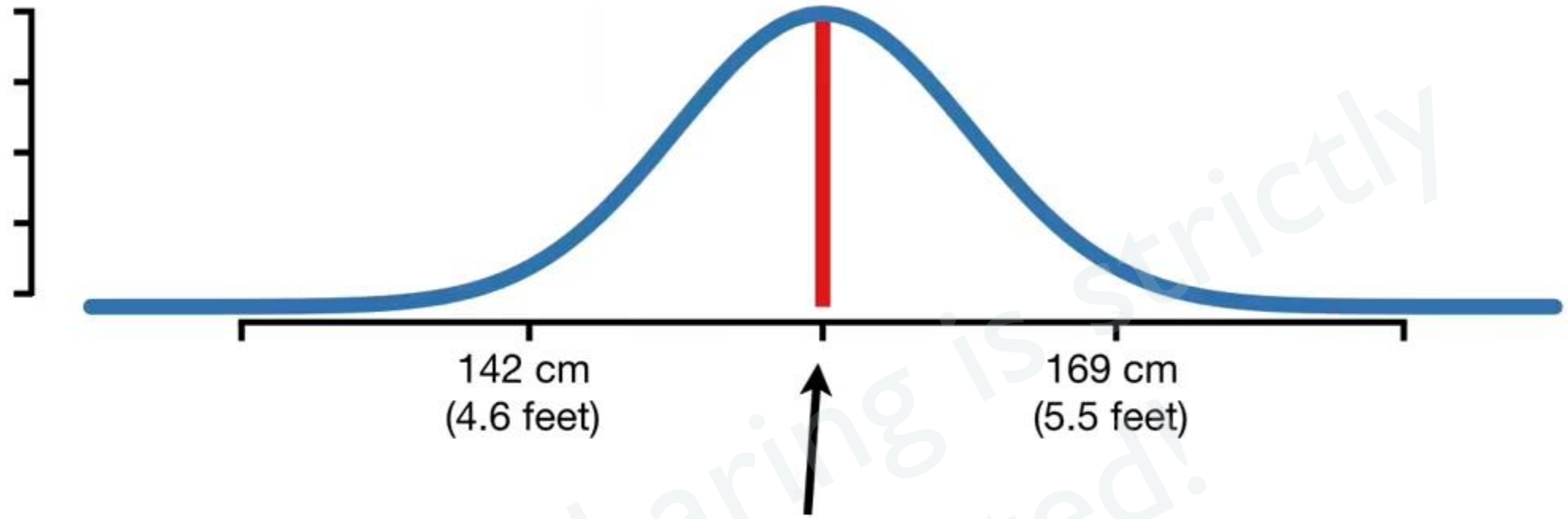
NOTE: The peak of the curve is right at the average height, so we are asking...



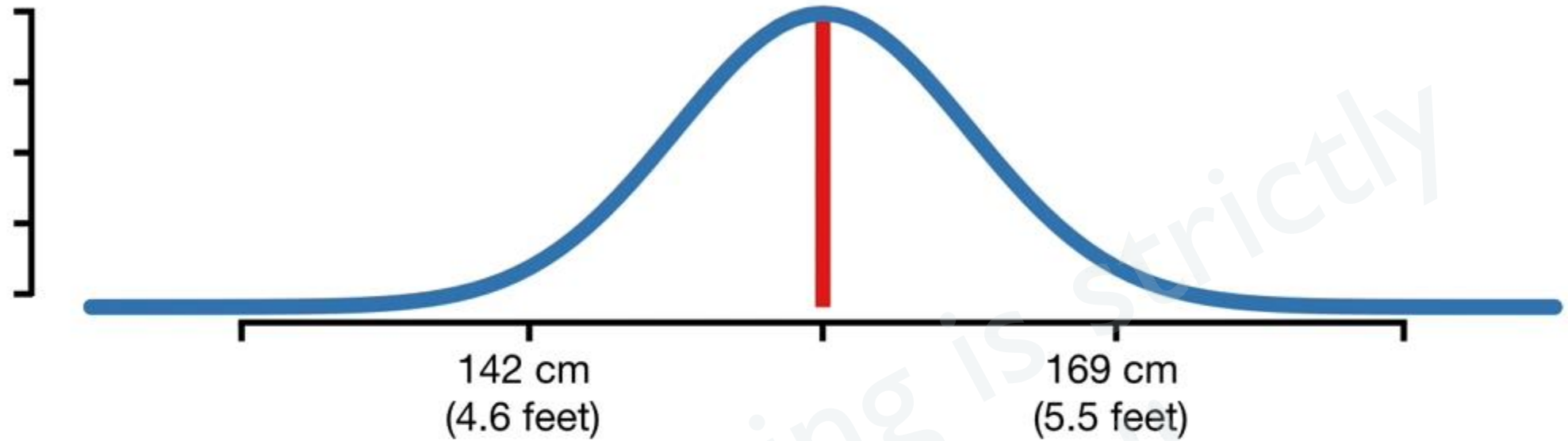
“Is a measurement between **155.4** and **156** so far away from the mean of the **blue distribution** (**155.5 cm**) that we can reject the idea that it came from it?”



If the **p-value** is small, then that suggests that some other distribution would do a better job explaining the data.

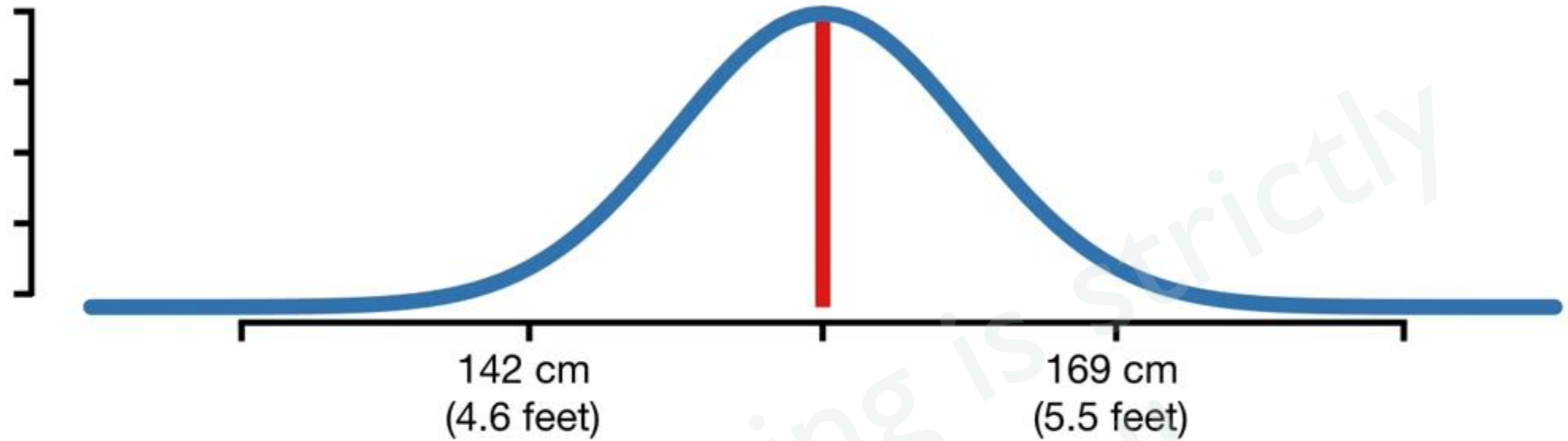


NOTE: The probability of someone being between **155.4** and **156** cm is only **0.04**. The **red area** is pretty small...barely a line!



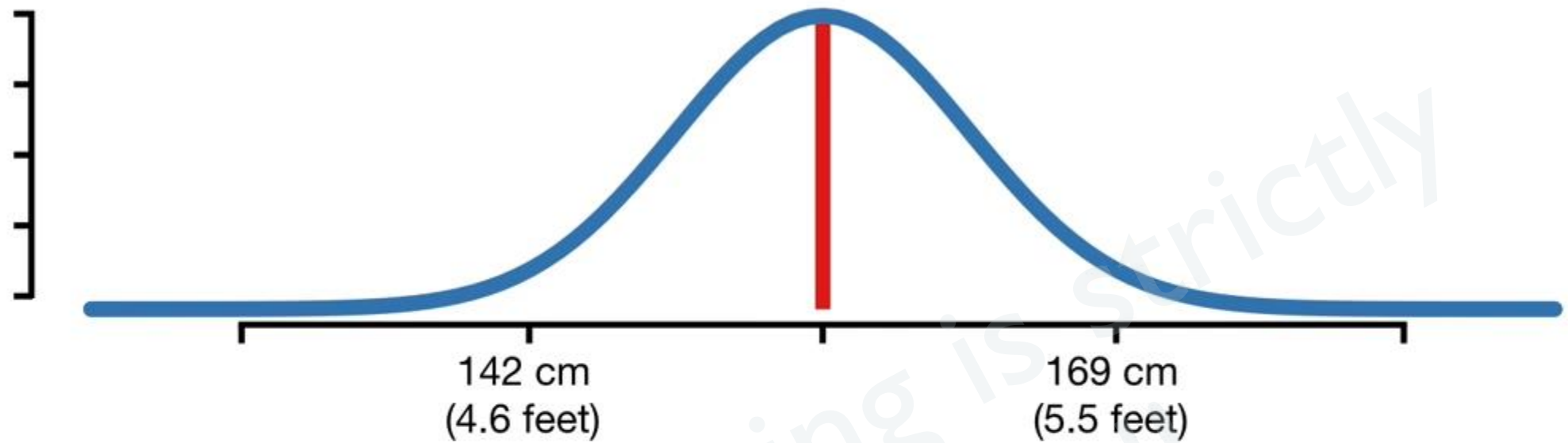
So **0.04** is the first part of calculating the **p-value**, since, given this distribution of heights, that is the probability that we would randomly measure someone in this range of values.

p-value for between
155.4 and **156** cm given = 0.04
the **blue distribution**



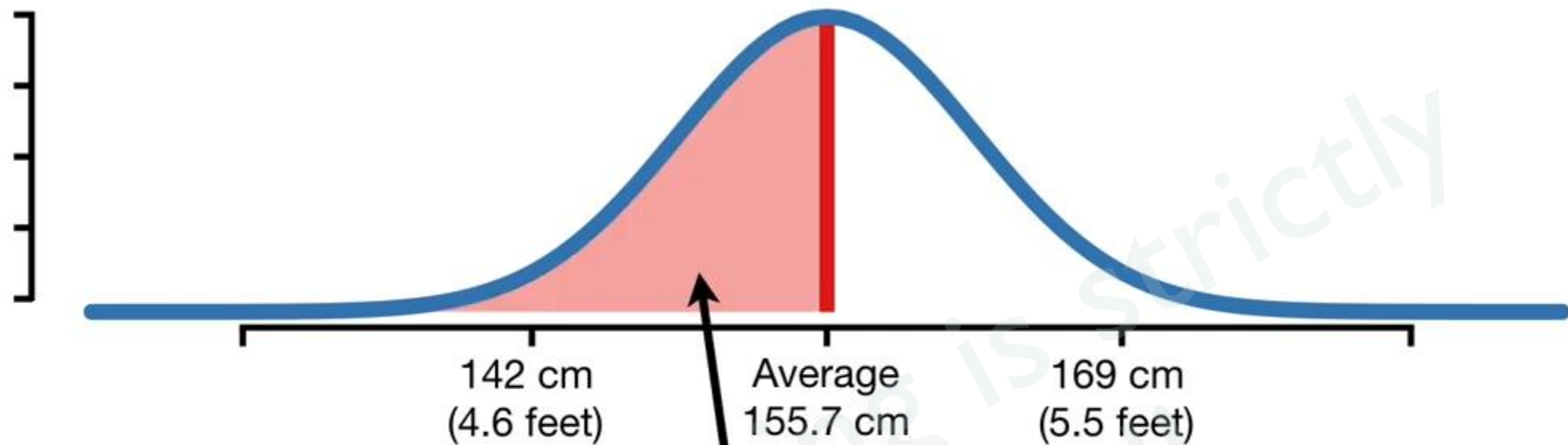
So **0.04** is the first part of calculating the **p-value**, since, given this distribution of heights, that is the probability that we would randomly measure someone in this range of values.

p-value for between
155.4 and **156** cm given = 0.04
the **blue distribution**



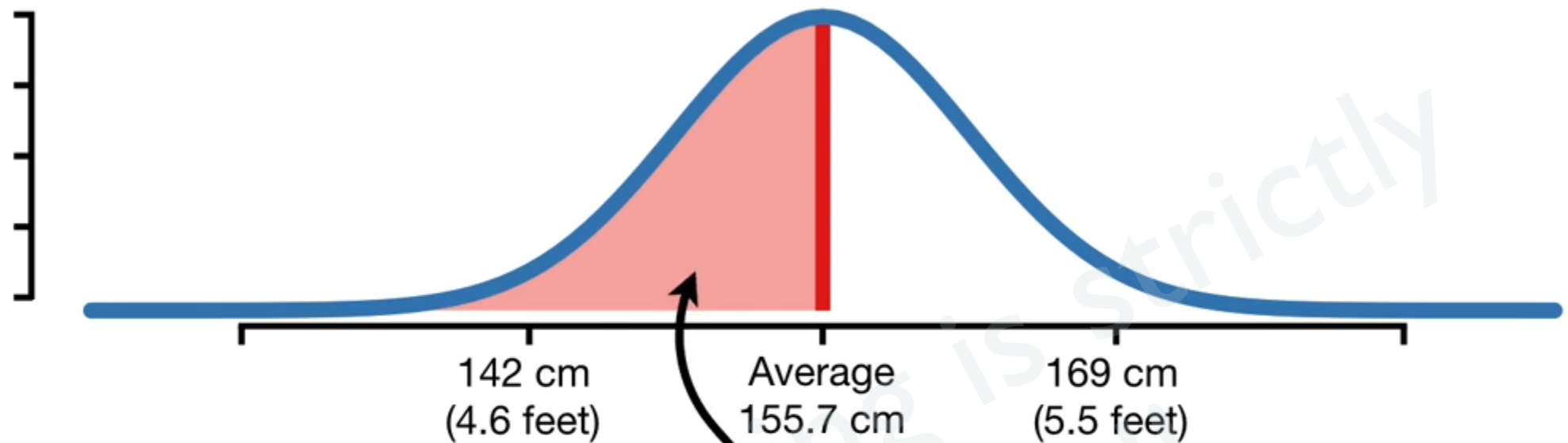
...now we need to figure out the **more extreme** parts.

p-value for between
155.4 and 156 cm given = 0.04
the **blue distribution**



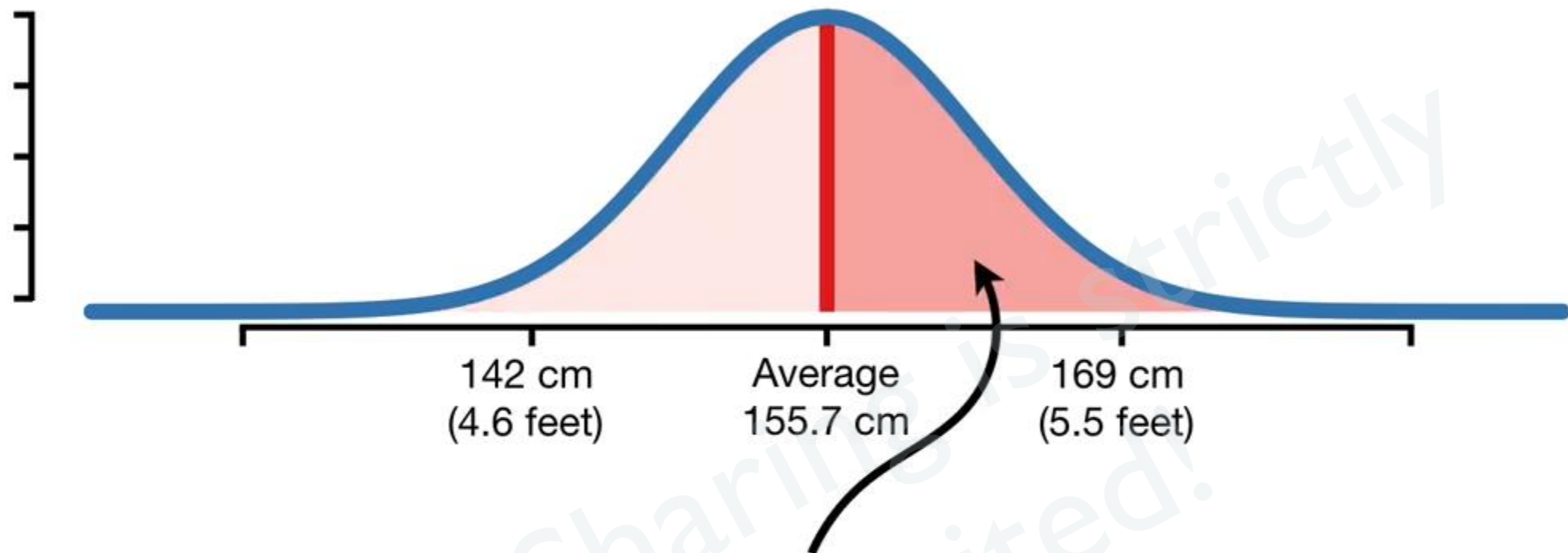
On the left side, all of the heights < 155.4 are further from the mean (**155.7**), thus, they are all **more extreme**.

p-value for between
155.4 and 156 cm given = 0.04
the **blue distribution**



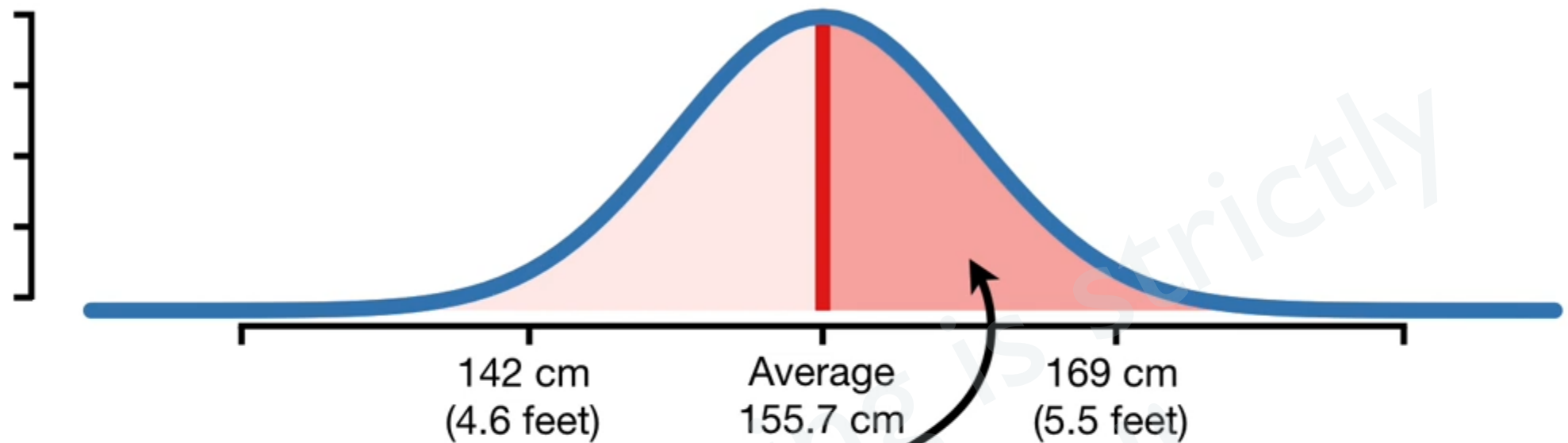
And because the **48%** of the area under the curve is for heights < 155.4 , we add **0.48** to the **p-value**.

p-value for between **155.4** and **156** cm given = 0.04
the **blue distribution**



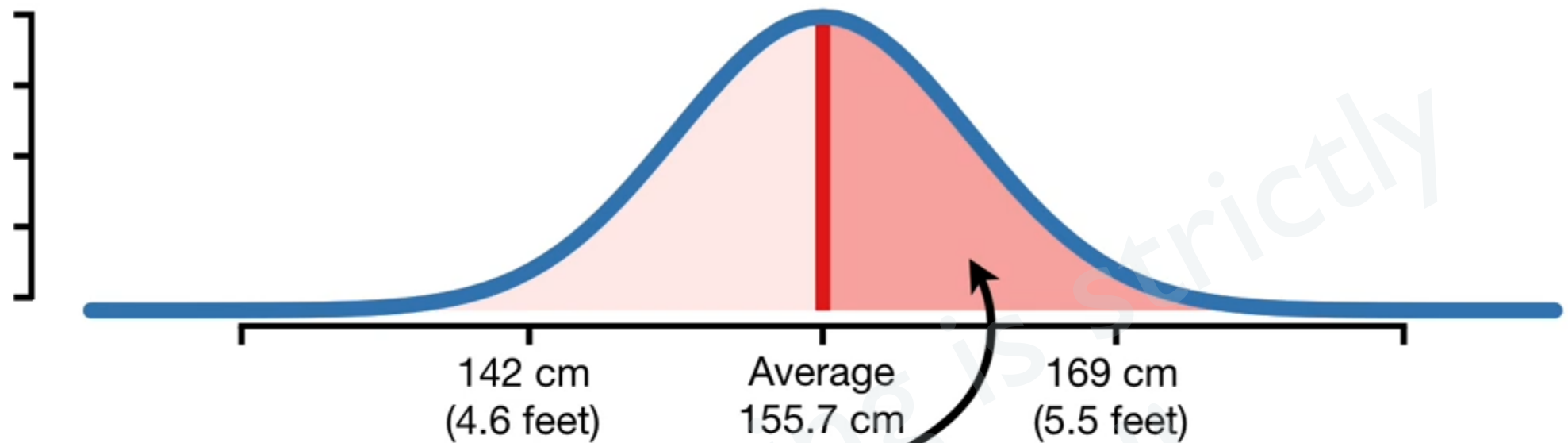
On the right side, all of the heights > 156 are further from the mean (**155.7**), thus, they are all **more extreme**.

p-value for between
155.4 and **156** cm given = $0.04 + 0.48$
the **blue distribution**



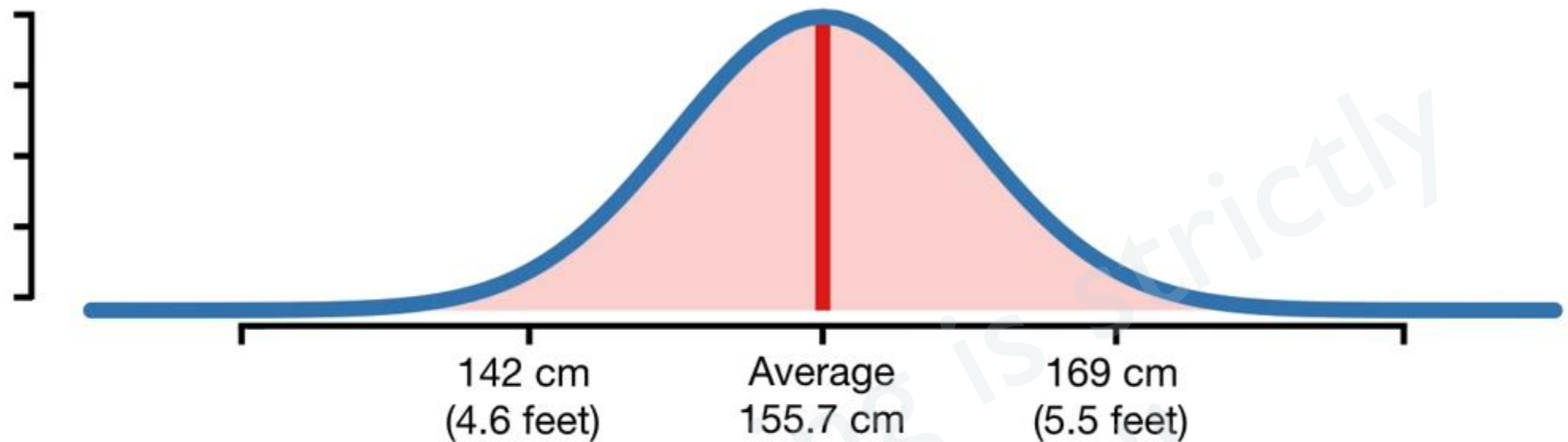
And because the **48%** of the area under the curve is for heights **> 156**, we add **0.48** to the **p-value**.

p-value for between
155.4 and **156** cm given = $0.04 + 0.48$
the **blue distribution**



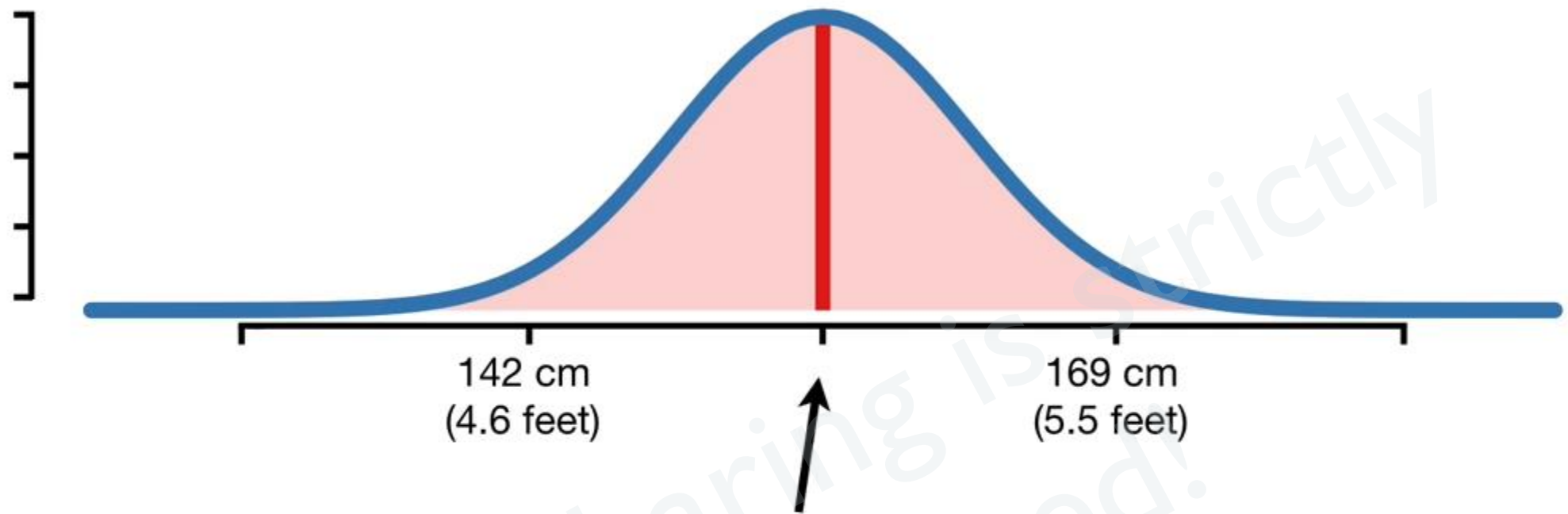
And because the **48%** of the area under the curve is for heights **> 156**, we add **0.48** to the **p-value**.

p-value for between **155.4** and **156** cm given = $0.04 + 0.48 + 0.48 =$ the **blue distribution**



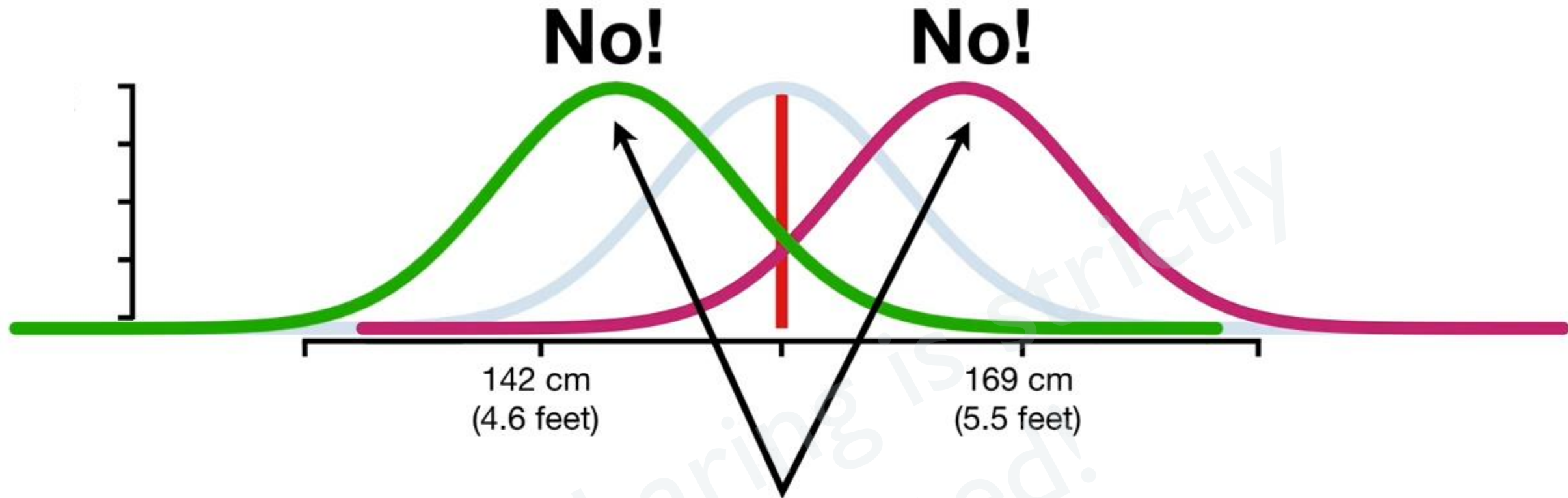
Ultimately, we end up adding all of the area under the curve, so the **p-value = 1**.

p-value for between **155.4** and **156** cm given = $0.04 + 0.48 + 0.48 = 1$
the **blue distribution**



So, this means that, given this distribution of heights, we would not find it unusual to measure someone who's height was close to the average, even though the probability is small (**0.04**).

p-value for between
155.4 and 156 cm given = $0.04 + 0.48 + 0.48 = 1$
the **blue distribution**

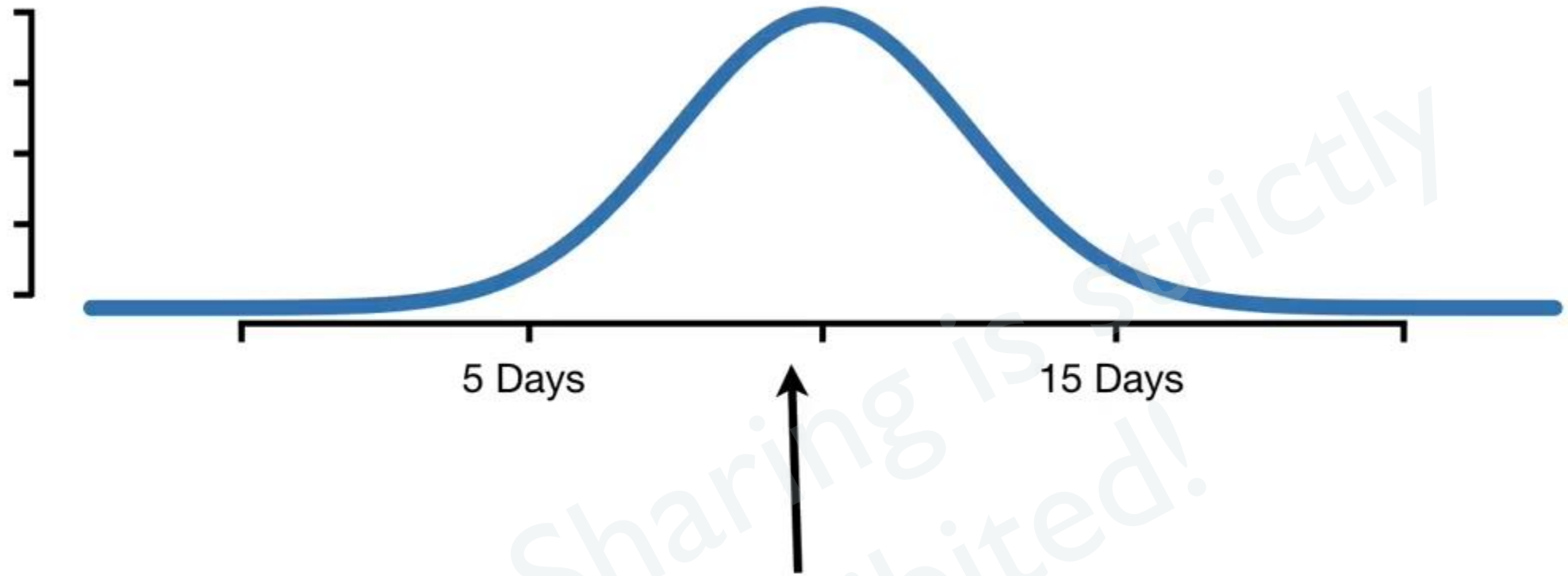


In other words, the data does not suggest that another distribution would do a better job explaining the data.

p-value for between
155.4 and 156 cm given = $0.04 + 0.48 + 0.48 = 1$
the **blue distribution**

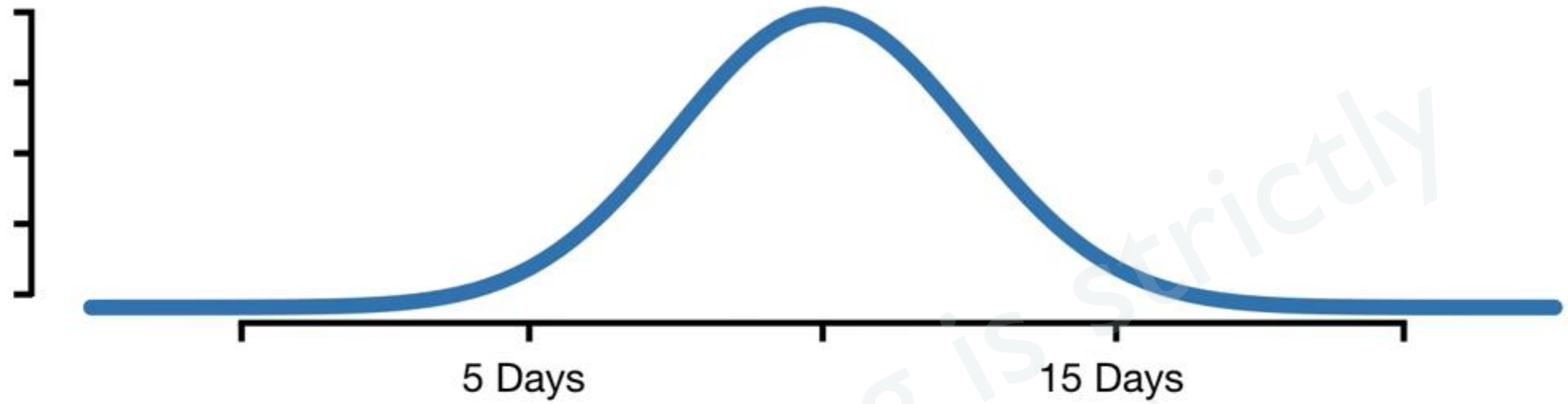
So far all we've only talked about **2-Sided p-values**.

Now I'll give you an example of a **One-Sided p-value** and tell you why it has the potential to be dangerous.

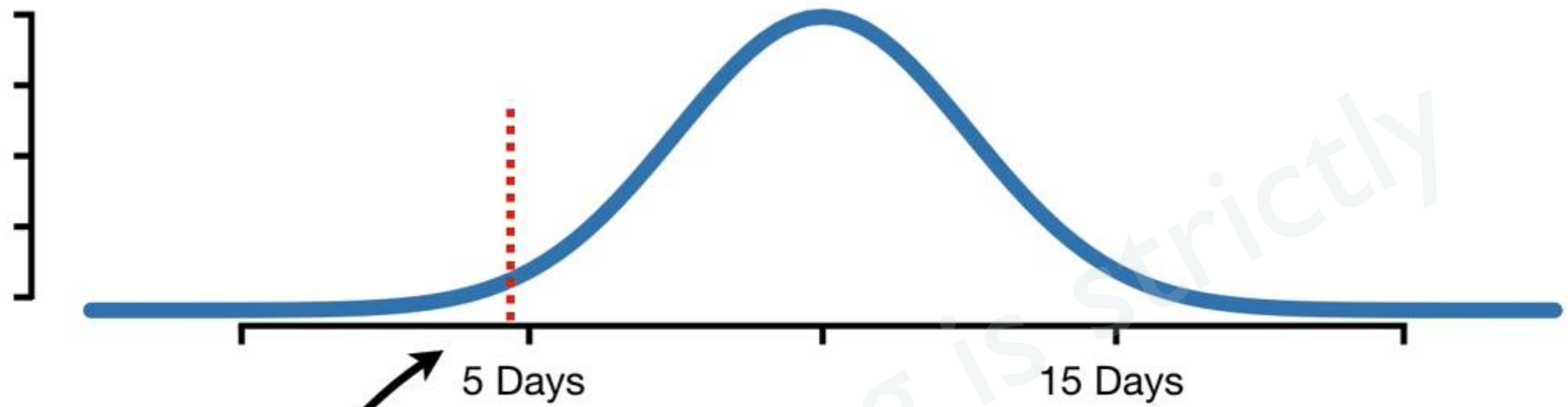


Imagine we measured how long it took a bunch of people to recover from an illness.



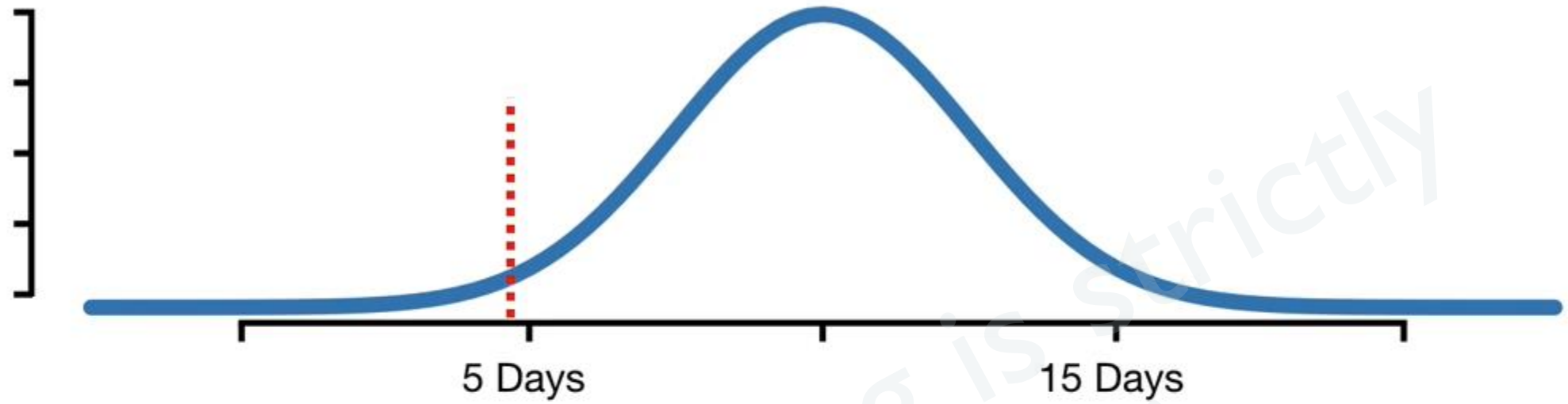


Now imagine we created a new drug, **SuperDrug**, and wanted to see if it helped people recover in fewer days.

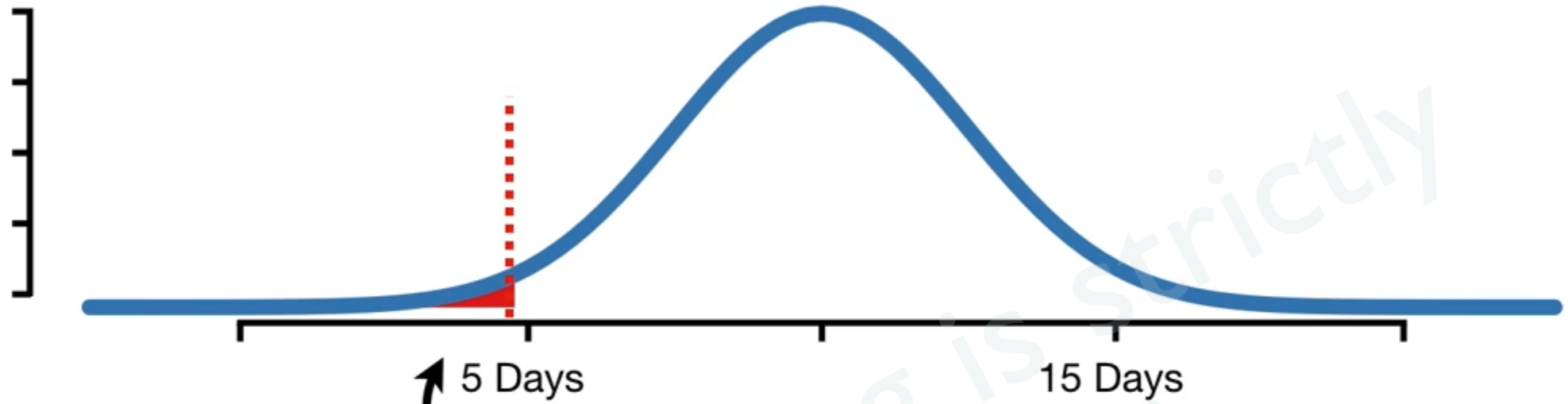


If we gave **SuperDrug** to a bunch of people and the average recovery was **4.5** days...

Sharing is strictly Prohibited!



...then a **Two-Sided p-value**, like the ones we've been computing all along, would be...

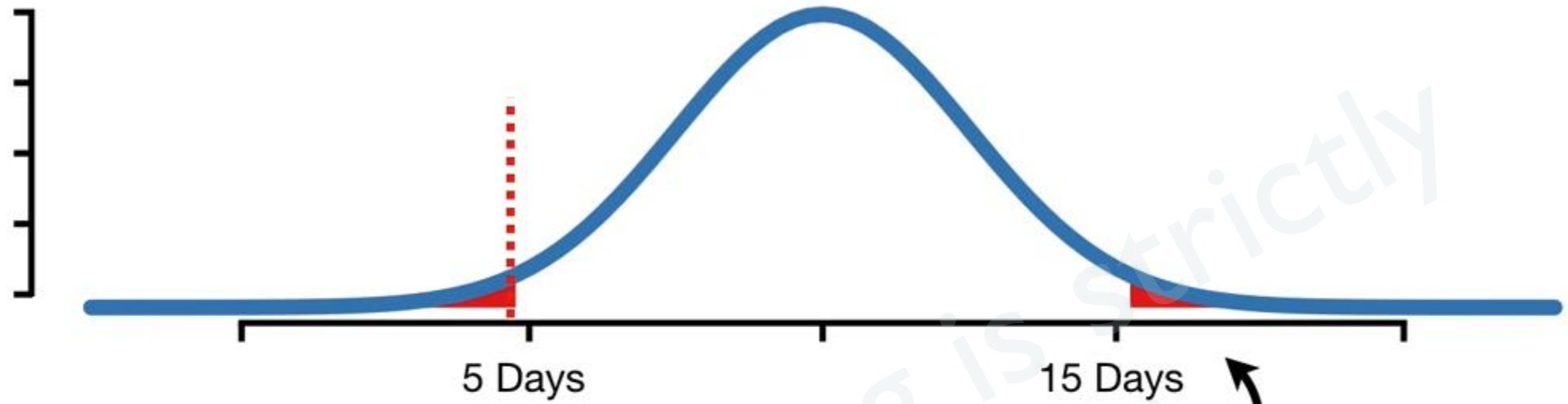


5 Days

15 Days

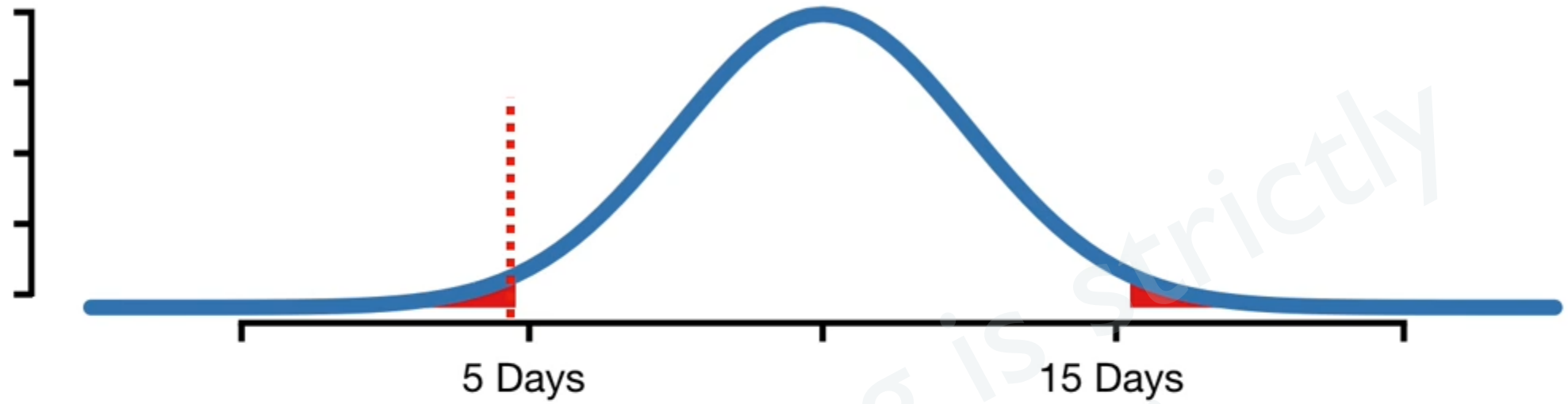
...the sum of *this* area under the curve, **0.016...**

Two-Sided p-value for 4.5 days = 0.016



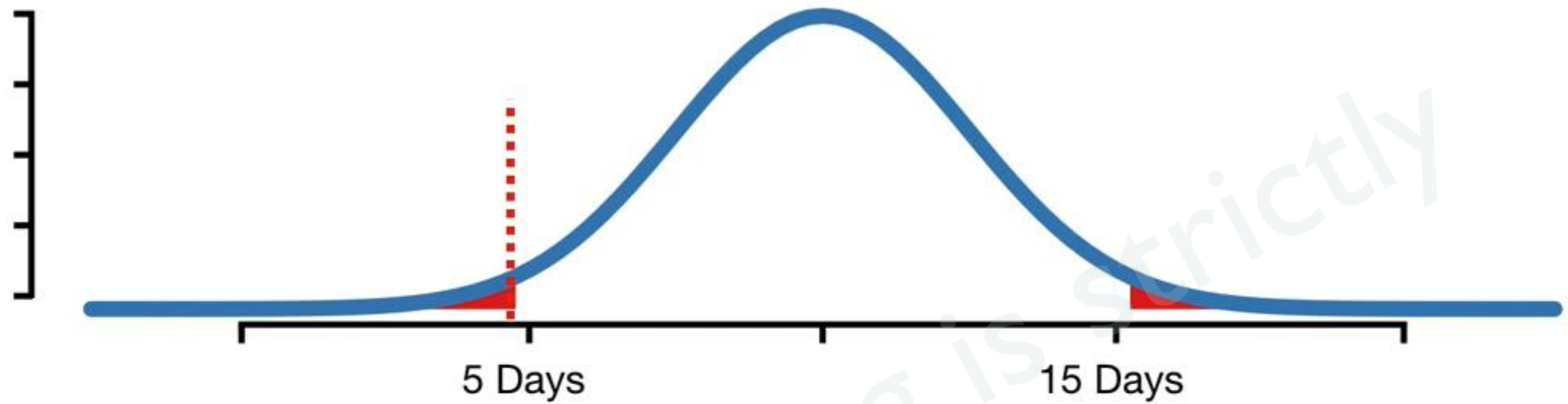
...plus *this* area under the curve, **0.016**...

Two-Sided p-value for **4.5** days = $0.016 + 0.016$



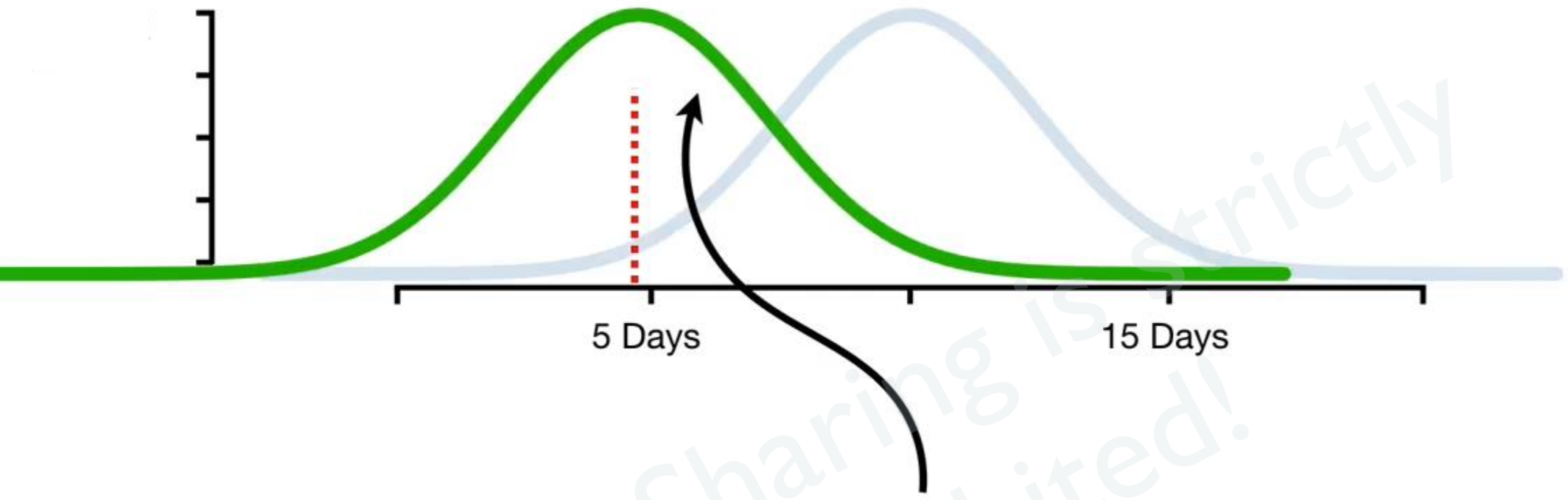
...and the total is **0.03**.

Two-Sided p-value for 4.5 days = $0.016 + 0.016 = 0.03$



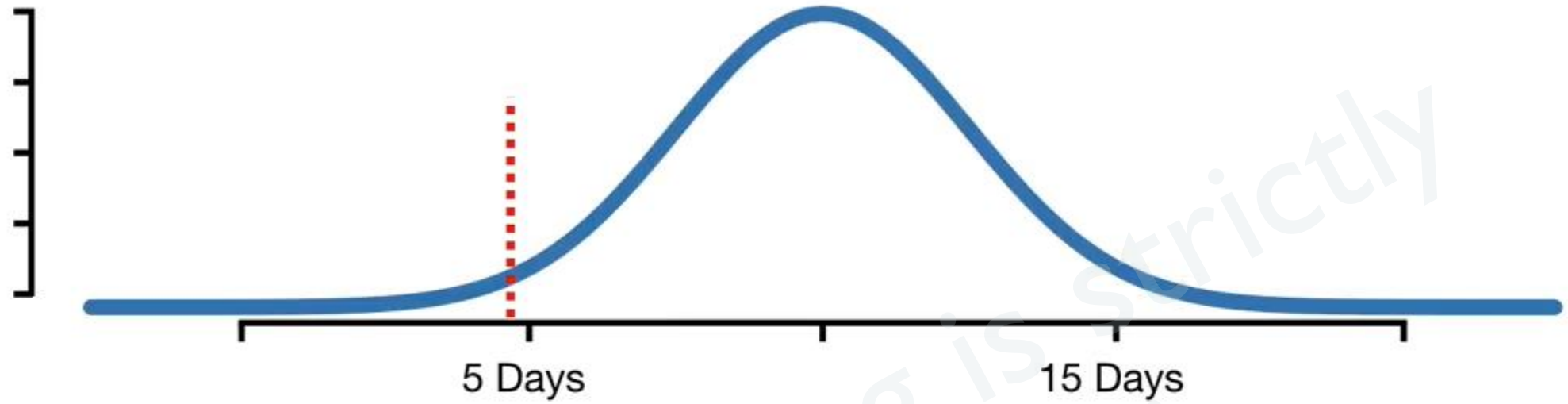
And since $0.03 < 0.05$, the **Two-Sided p-value** tells us that, given this distribution of recovery times, **SuperDrug** did something unusual.

Two-Sided p-value for 4.5 days = $0.016 + 0.016 = 0.03$

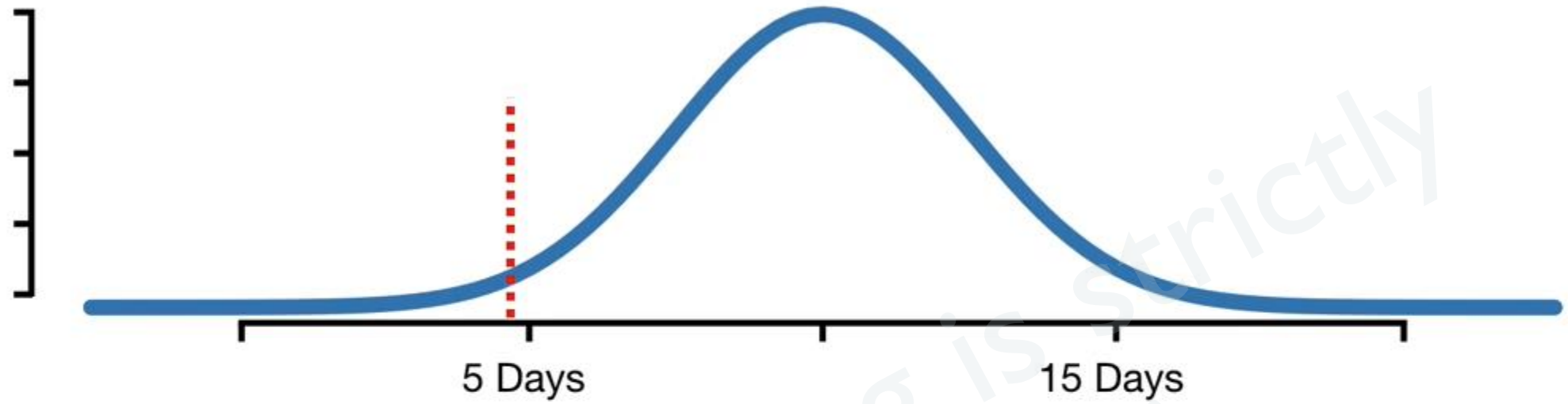


...and that suggests that some other distribution does a better job explaining the data.

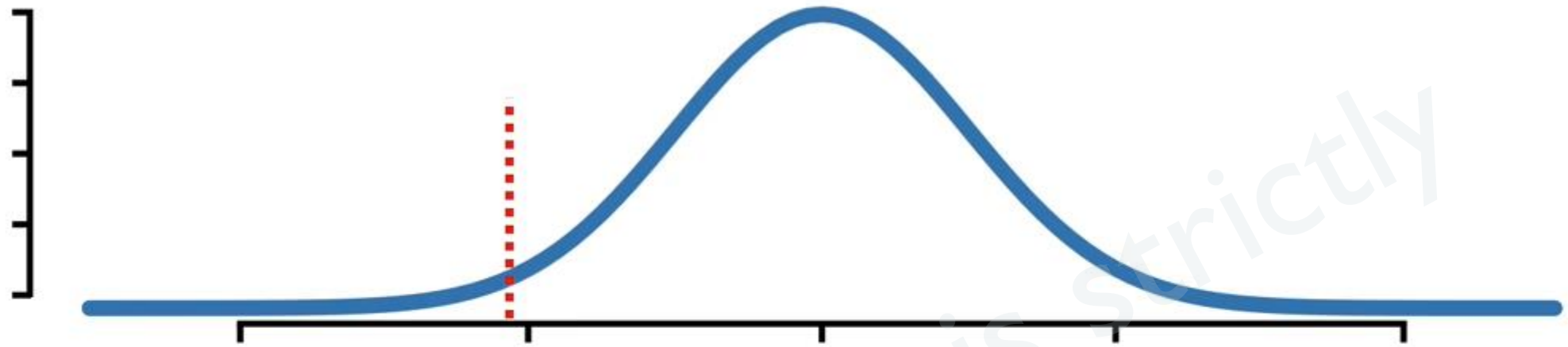
Two-Sided p-value for **4.5** days = $0.016 + 0.016 = 0.03$



For a **One-Sided p-value**, the first thing we do is decide which direction we want to see change in.



In this case, we'd like **SuperDrug** to shorten the time it takes to recover from the illness...



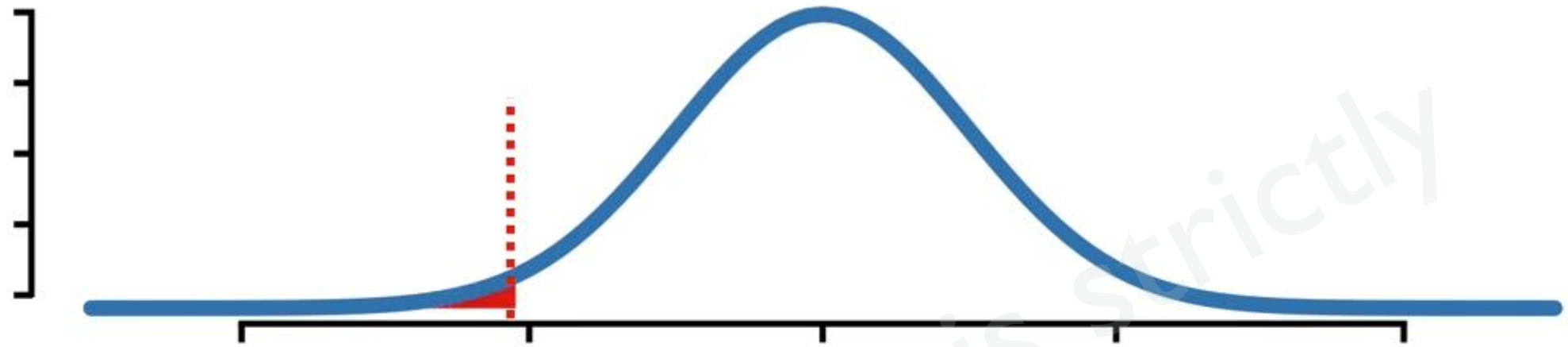
Shorter
Recovery
Times

5 Days

15 Days

...so that means we want to see if recovery times are shorter.

Draft. Sharing is strictly Prohibited!



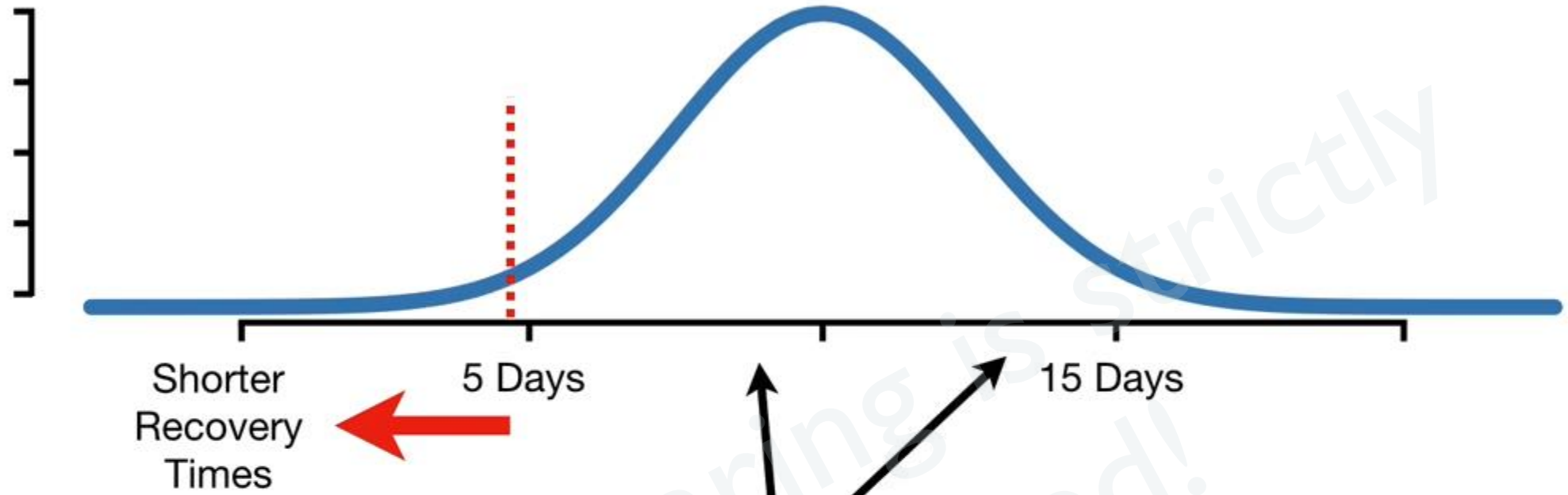
Shorter
Recovery
Times

5 Days

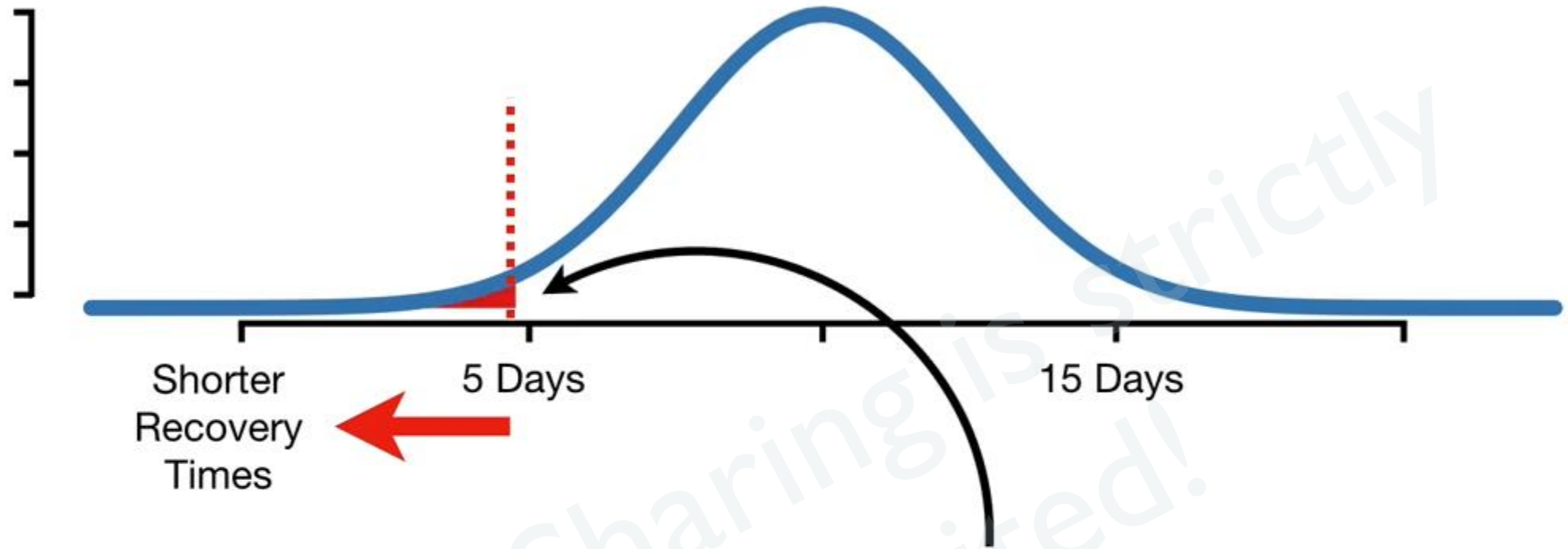
15 Days



Because we want to see change in
this direction, the only **more
extreme** values are **< 4.5** days.

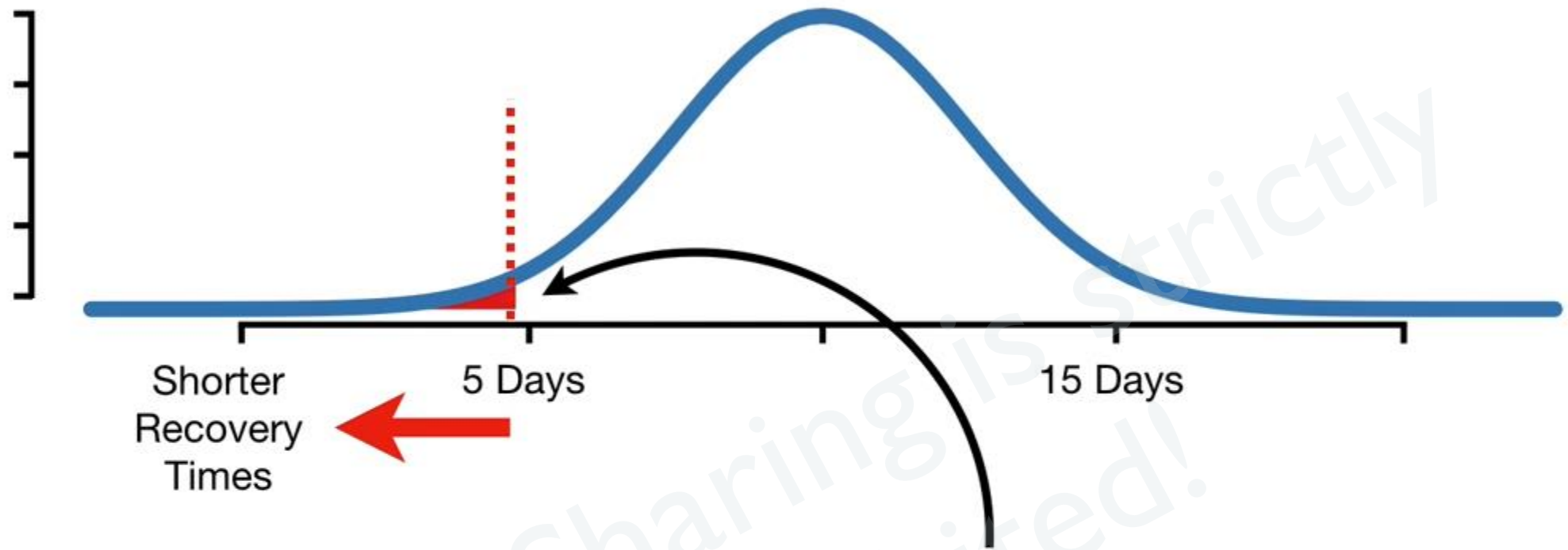


All of the values > 4.5 days are considered *less extreme*.



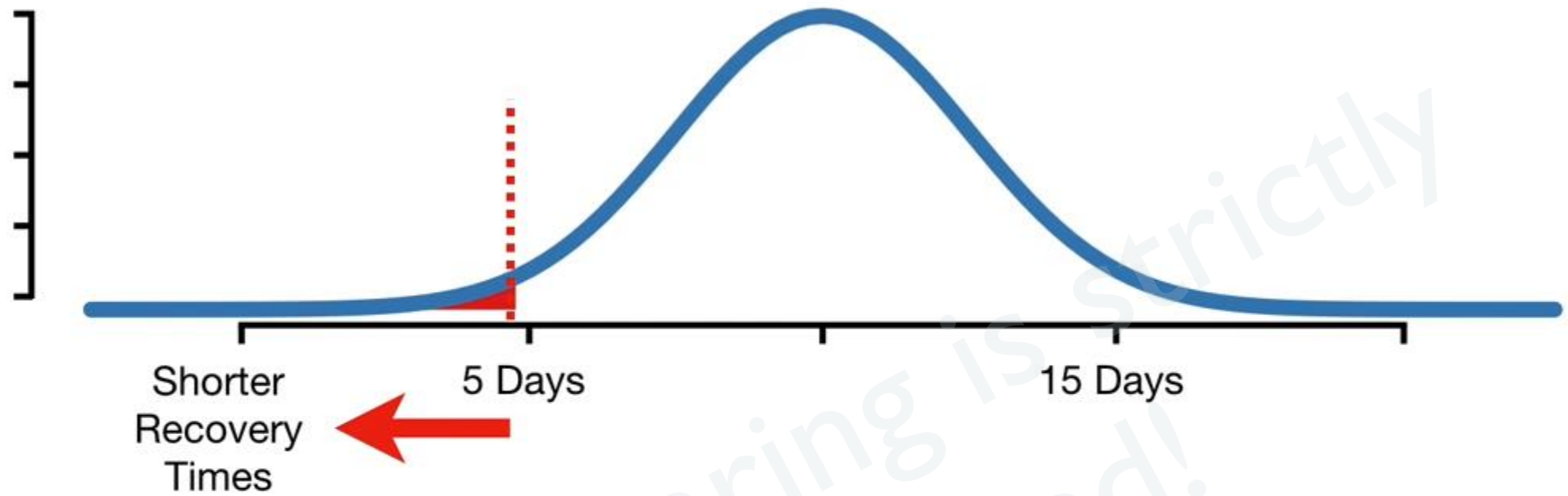
So, when we calculate a **One-Sided p-value**, we only use the area that is in the direction we want to see change, **0.016**.

One-Sided p-value for 4.5 days = 0.016



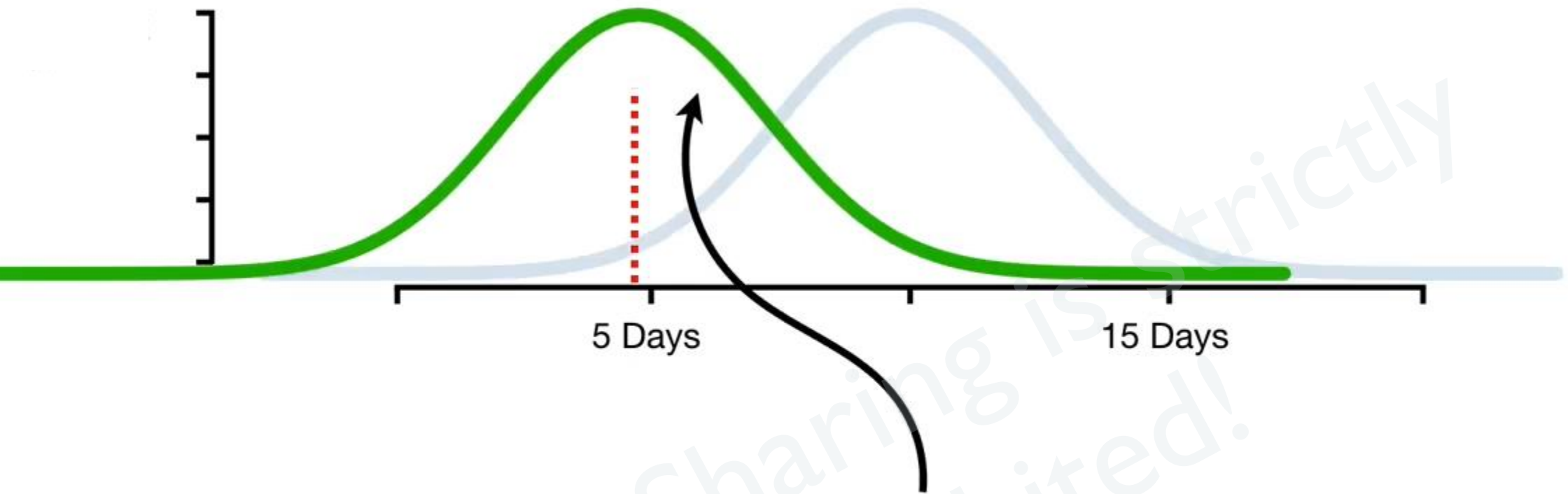
So, when we calculate a **One-Sided p-value**, we only use the area that is in the direction we want to see change, **0.016**.

One-Sided p-value for 4.5 days = 0.016



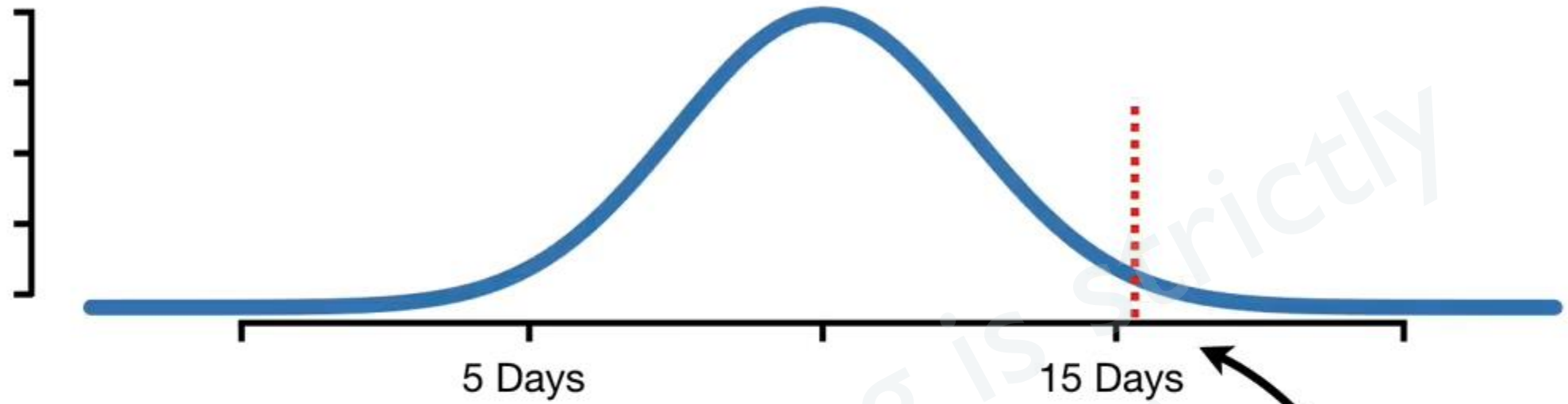
Again, since $0.016 < 0.05$, the **One-Sided p-value** would tell us that, given this distribution, **SuperDrug** did something unusual...

One-Sided p-value for 4.5 days = 0.016

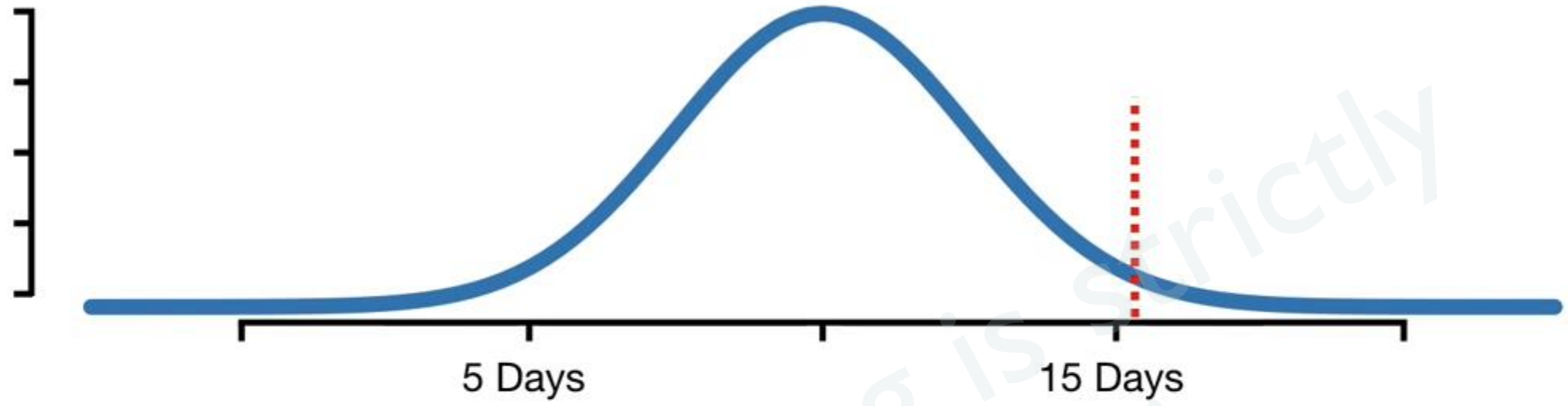


...and that some other distribution makes more sense.

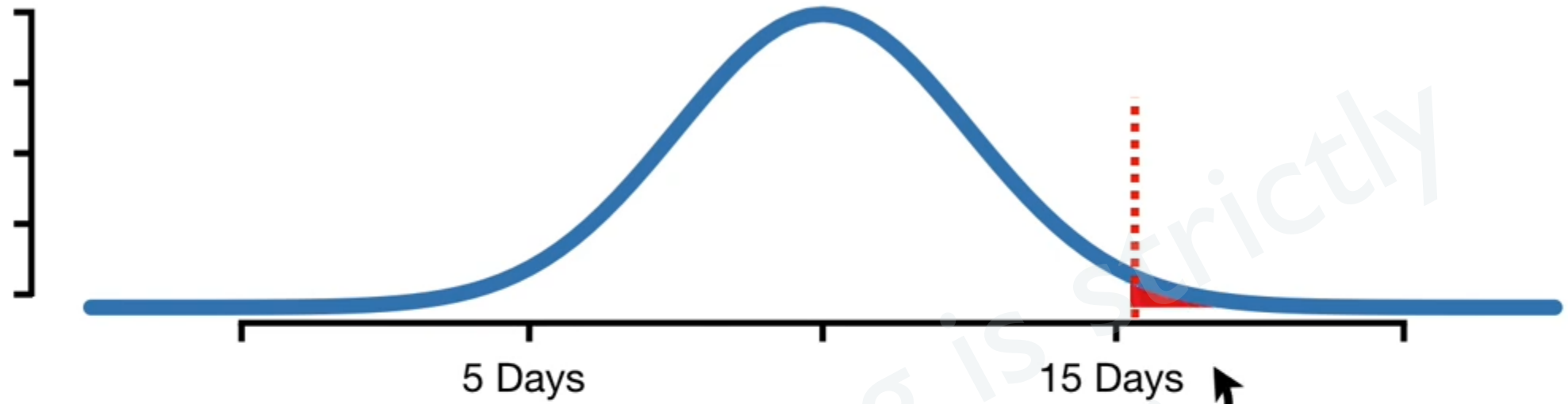
One-Sided p-value for 4.5 days = 0.016



Now, imagine that **SuperDrug** wasn't so super, and, on average, it took **15.5** days to recover.

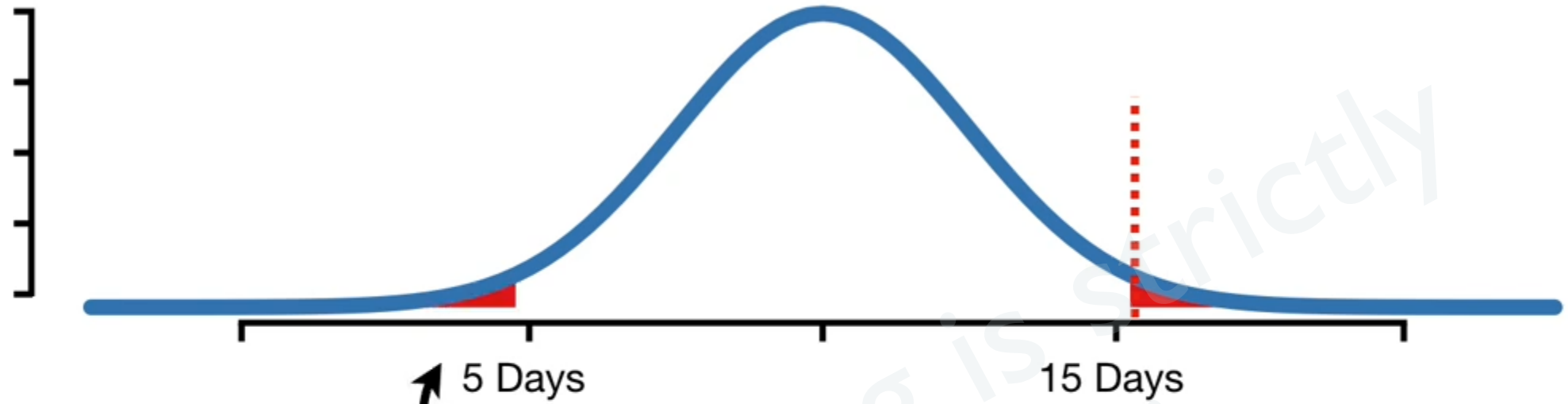


Just like before, the **Two-Sided p-value** would be...



...the sum of *this* area under the curve, **0.016...**

Two-Sided p-value for 15.5 days = 0.016

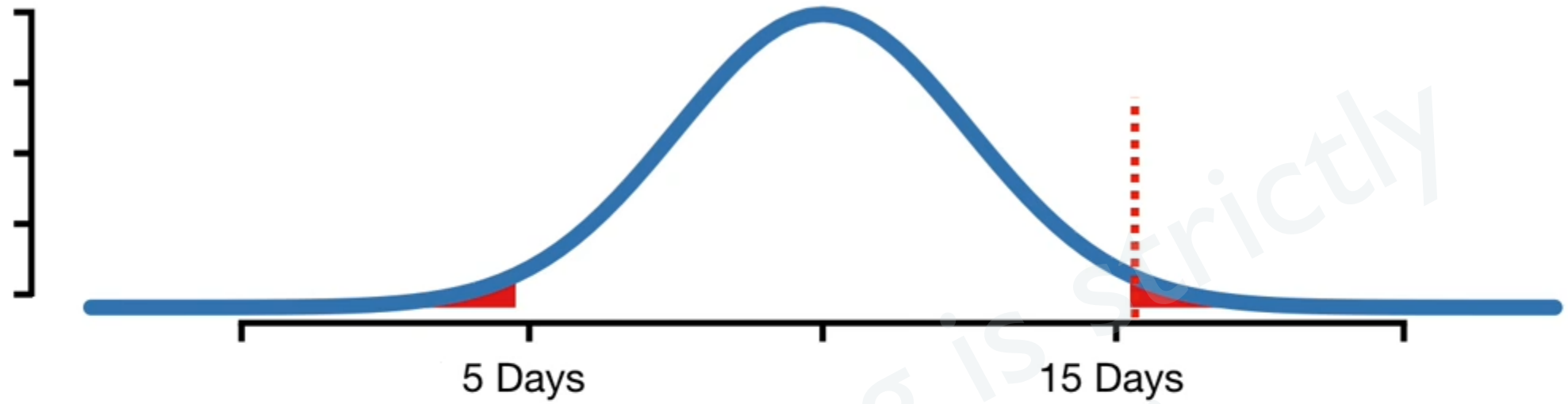


5 Days

15 Days

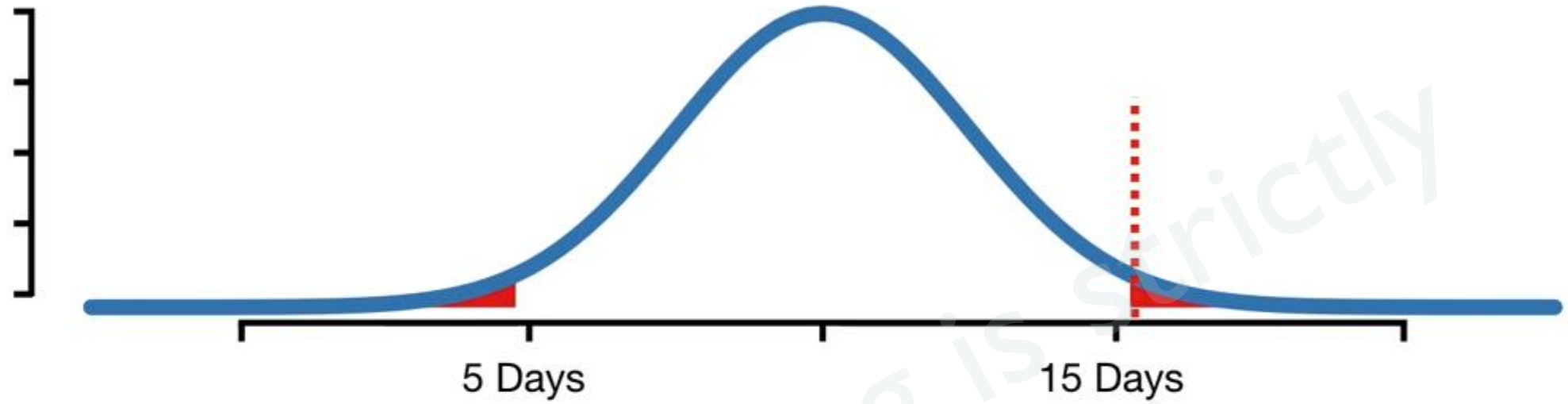
...plus *this* area under the curve, **0.016...**

Two-Sided p-value for **15.5** days = $0.016 + 0.016$

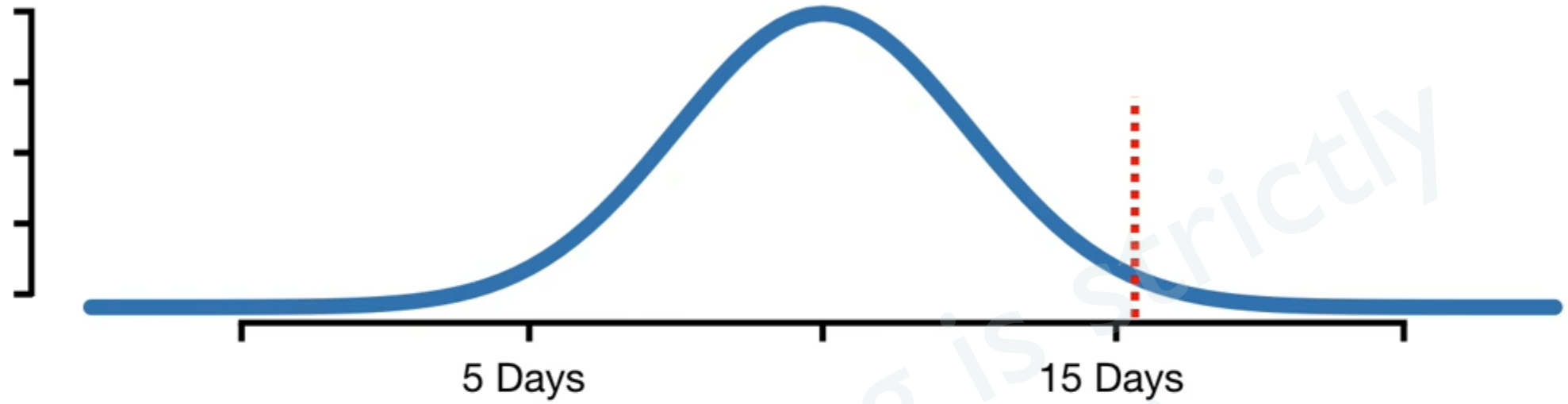


...and the total is **0.03**.

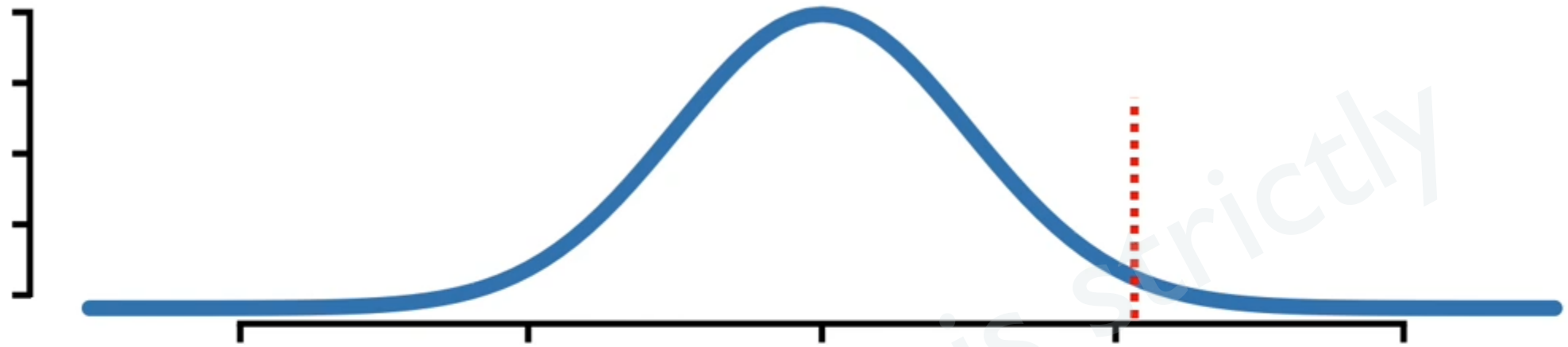
Two-Sided p-value for **15.5** days = $0.016 + 0.016 = 0.03$



In other words, regardless of whether **SuperDrug** is super and makes things better, or if is not so super and makes things worse, a **Two-Sided p-value** will detect something unusual happened.



For a **One-Sided p-value**, the first thing we do is decide which direction we want to see change in.

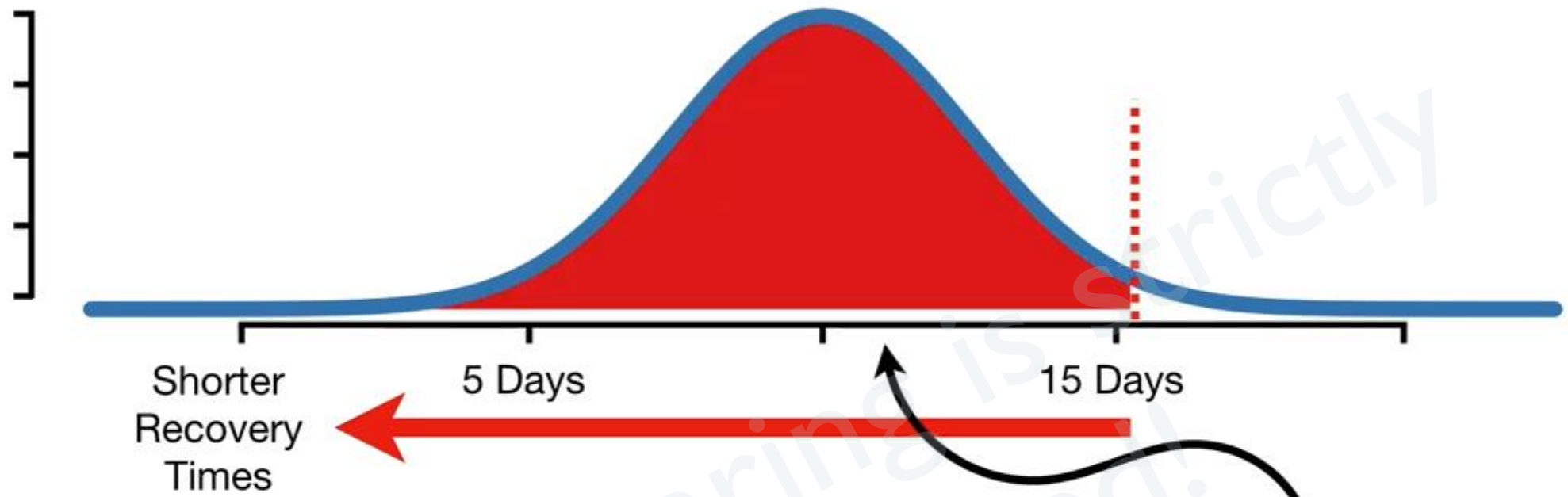


Shorter
Recovery
Times

5 Days

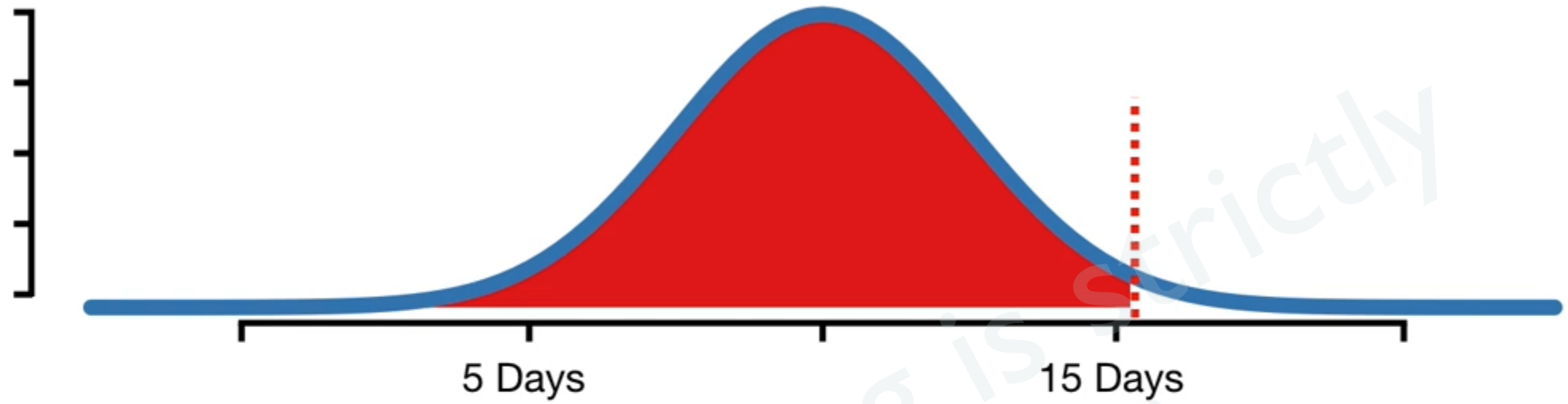
15 Days

...and just like before, that means we want
to see if recovery times are shorter.



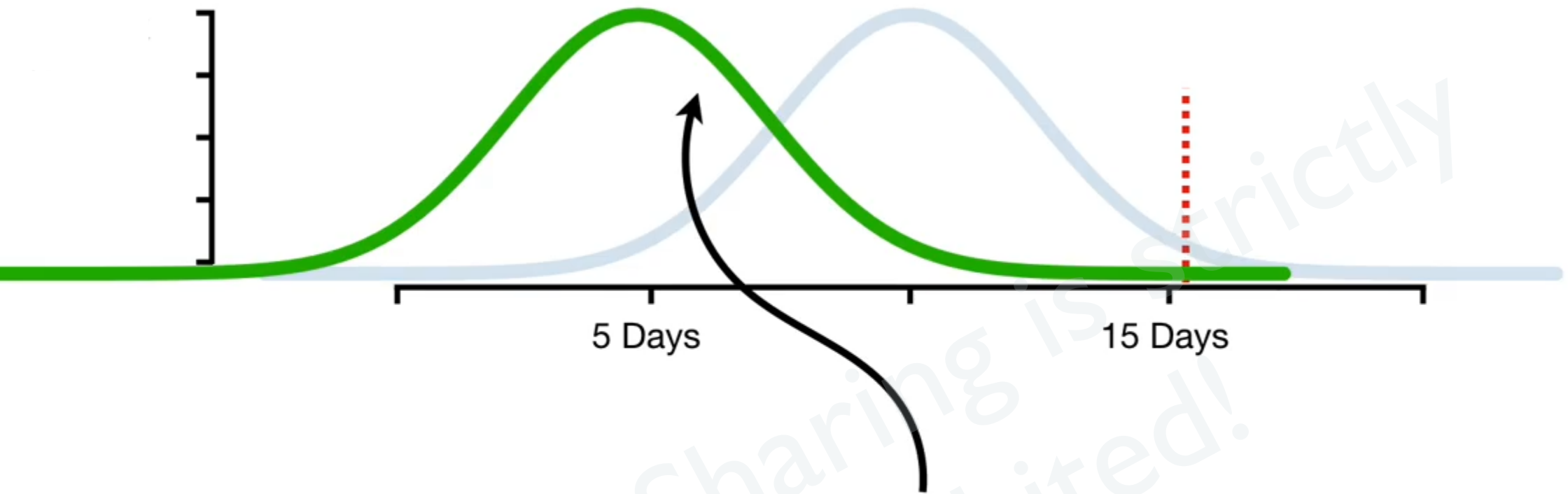
So the **One-Sided p-value** is this huge area, **0.98**, because it is **more extreme** in the direction we want to see change.

One-Sided p-value for 15.5 days = 0.98



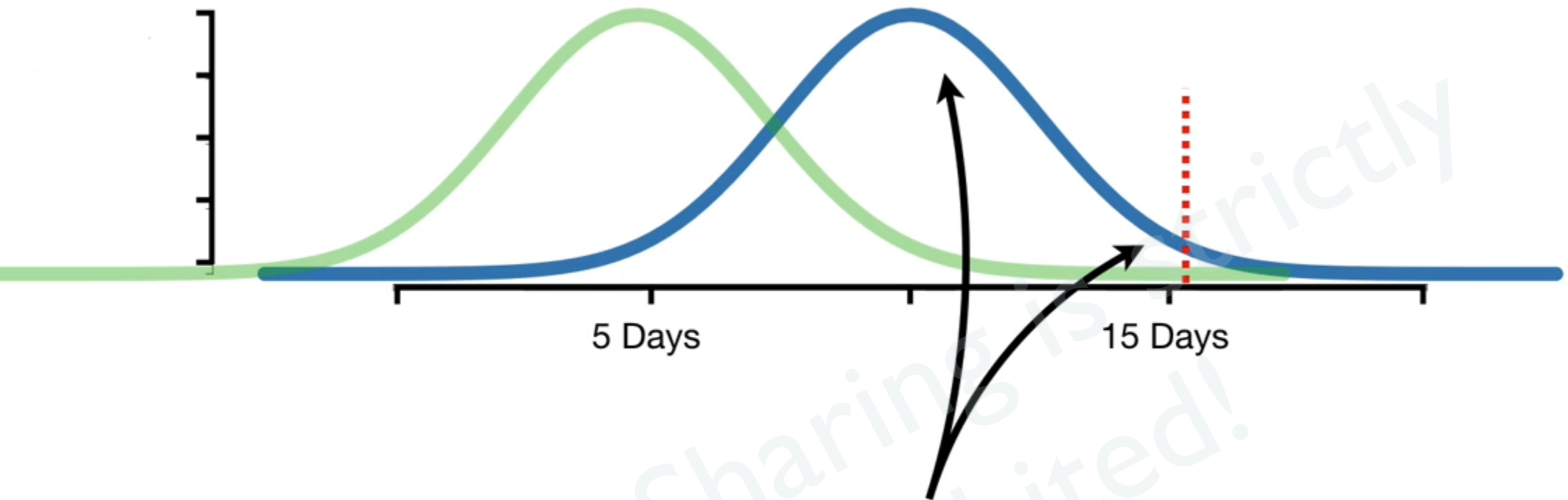
And since $0.98 > 0.05$, the **One-Sided p-value** would not detect that **SuperDrug** was doing anything unusual.

One-Sided p-value for **15.5** days = 0.98



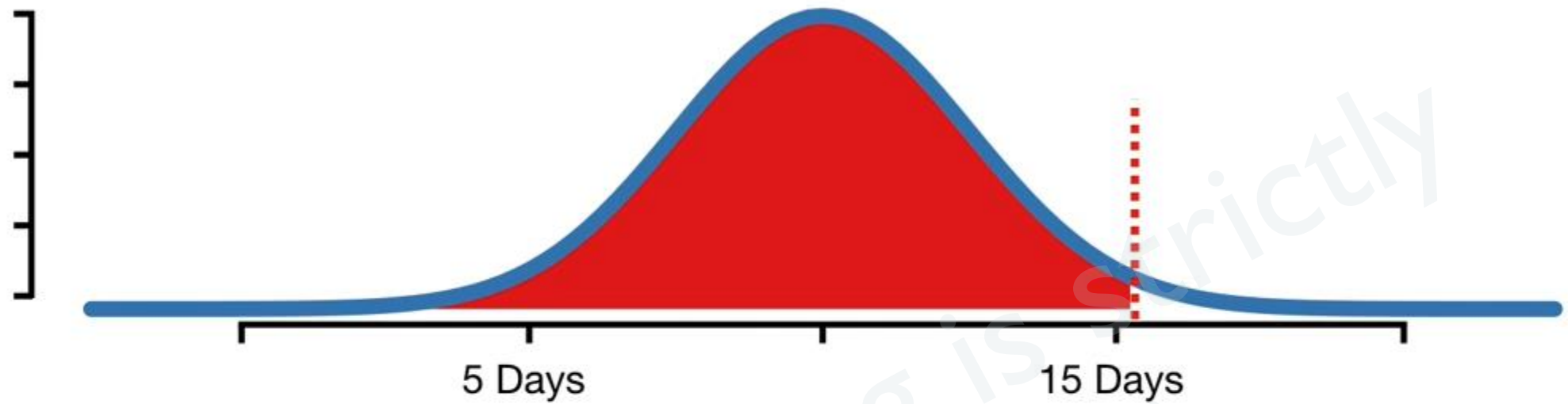
In other words, the **One-Sided p-value** is only looking to see if a distribution to the left of the original mean makes more sense...

One-Sided p-value for **15.5** days = 0.98



...and since the observation is on the right side of the mean, we fail to reject the hypothesis that the original distribution makes sense.

One-Sided p-value for 15.5 days = 0.98



And since failing to detect that **SuperDrug** is making things worse would be bad, **One-Sided p-values** are tricky and should be avoided, or only be used by experts who really know what they are doing.

In summary, a ***p-value*** is composed
of three parts:

Draft: Sharing is strictly
Prohibited!

In summary, a ***p-value*** is composed of three parts:

- 1) The probability random chance would result in the observation.

In summary, a ***p-value*** is composed of three parts:

1) The probability random chance would result in the observation.

2) The probability of observing something else that is equally rare.

In summary, a ***p-value*** is composed of three parts:

- 1) The probability random chance would result in the observation.
- 2) The probability of observing something else that is equally rare.
- 3) The probability of observing something rarer or more extreme.

THANK YOU!