Frequentist Hypothesis Testing: Intuitions

# Objectives

- Develop a robust intuition for the frequentist approach to hypothesis testing
- Distinguish a population distribution from a sampling distribution
- Relate the area under a sampling distribution to P-values
- Perform a Z-test
- Contrast the use-case of a t-test vs a Z-test

### Real or random?

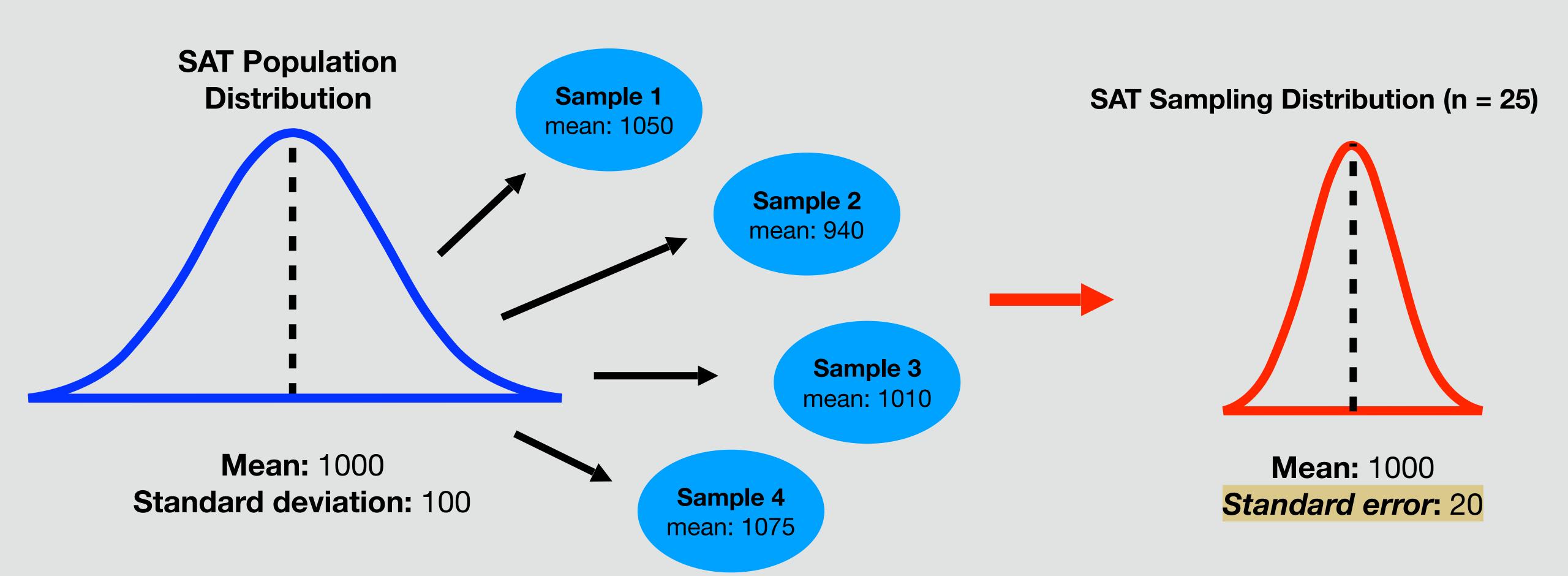
 The goal of hypothesis testing is to determine if an effect is statistically significant based on data

#### Example:

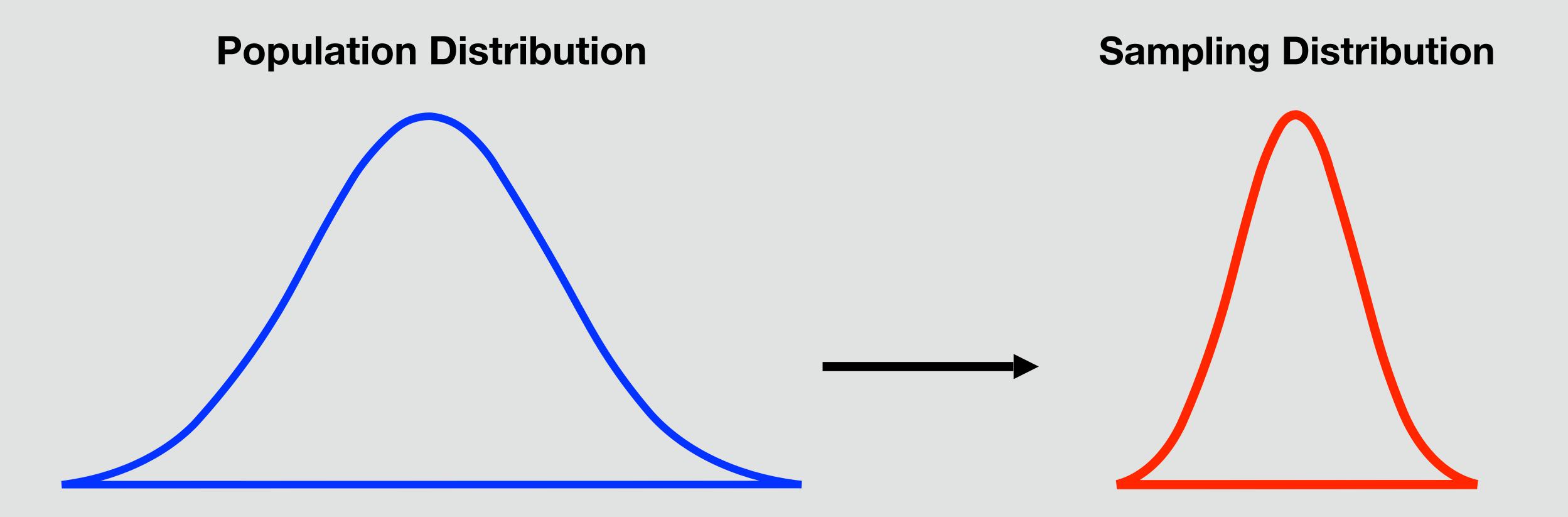
- Do SmartCards™ product actually improve SAT scores?
- Take a random sample of 25 students, have them study with the cards:
  - Population mean (without SmartCards): 1000
  - Experimental mean: 1050
- Can we quantify our certainty that this effect isn't due to randomness?

### Sampling Distributions

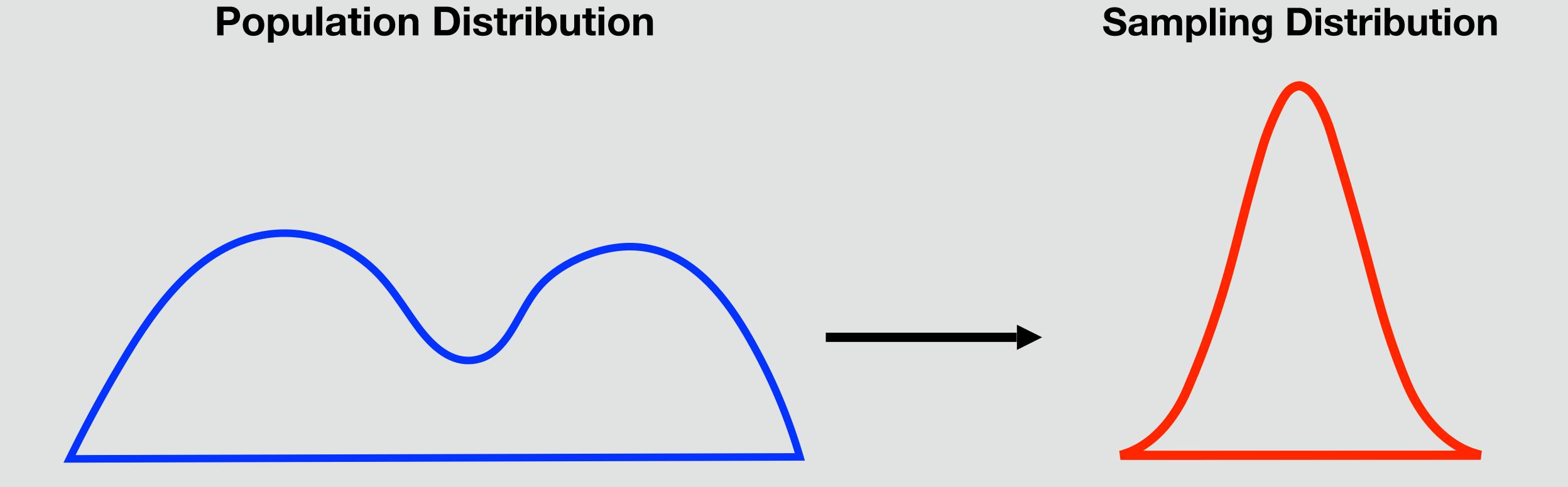
~~ Draw 25 person samples ~~



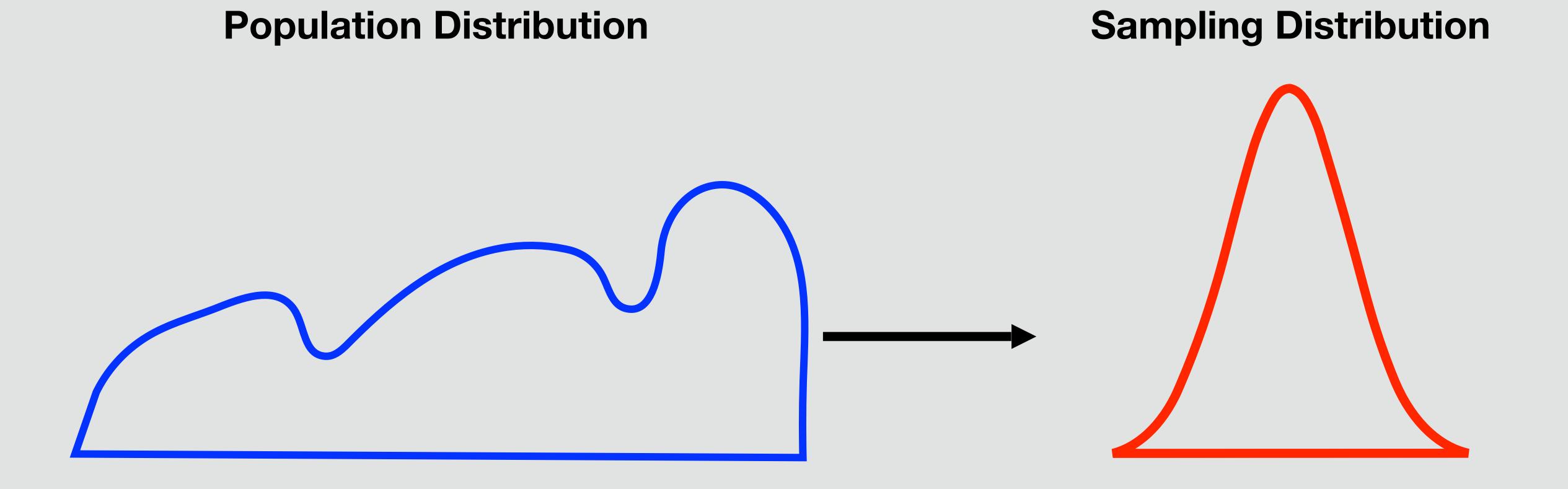
Nasty population distributions still have nice and normal sampling distributions



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Who do we have to thank for this?

THE CENTRAL LIMIT THEOREM

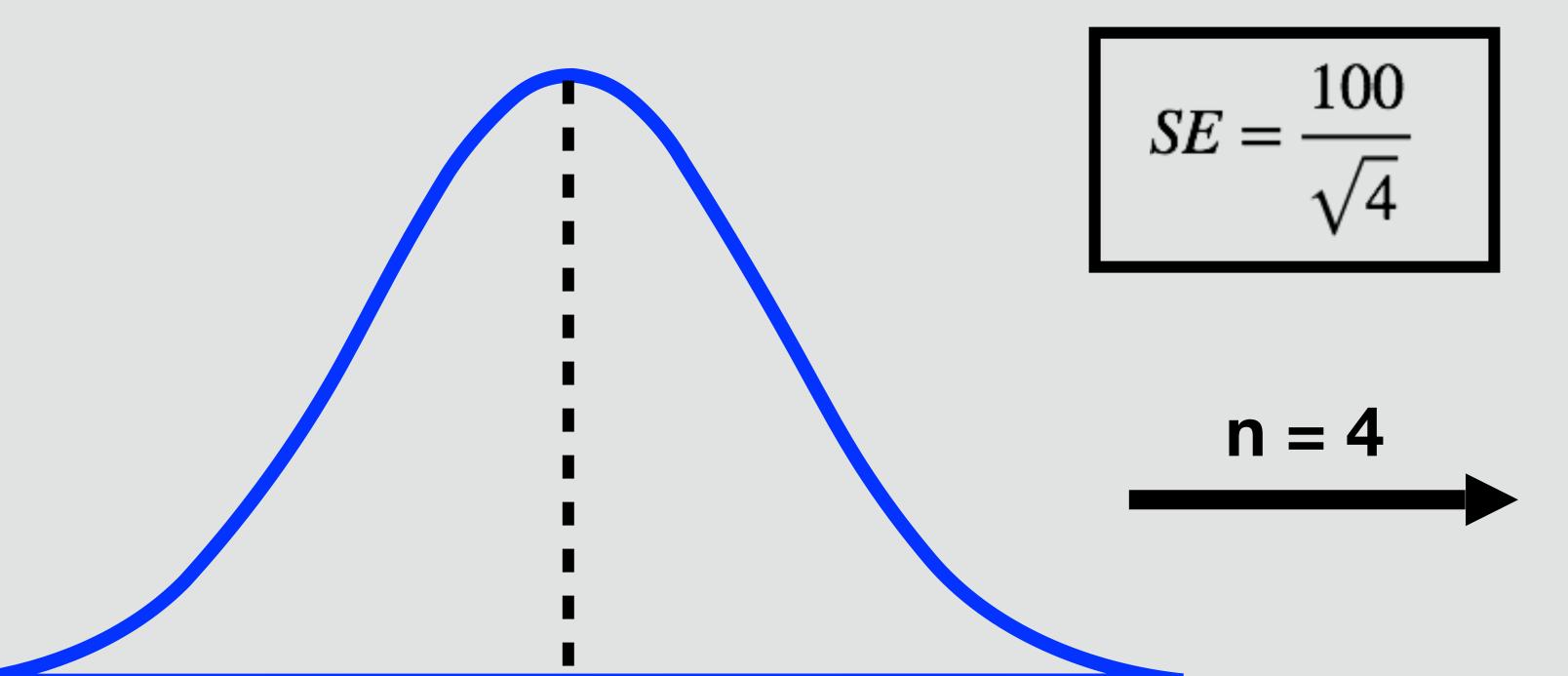
## Our friend, the standard error

- The standard deviation of sample means
- Measures the stability of sample means
- Depends on:
  - The population standard deviation
  - Sample size

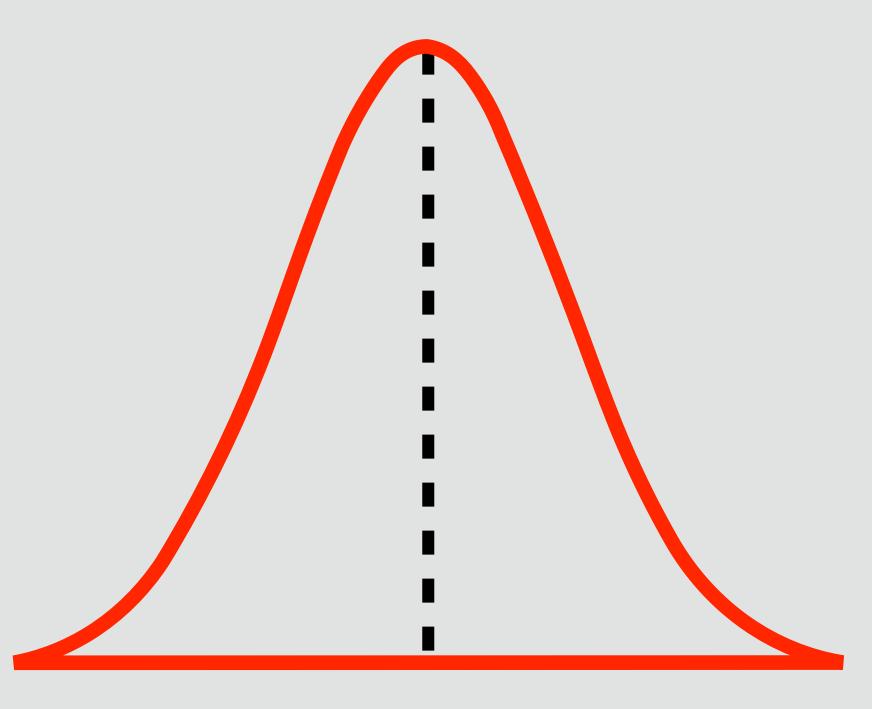
$$SE = \frac{\sigma}{\sqrt{n}}$$

### Sample Size affects the sampling distribution

#### **SAT Population Distribution**



**SAT Sampling Distribution (n = 4)** 



**Mean:** 1000

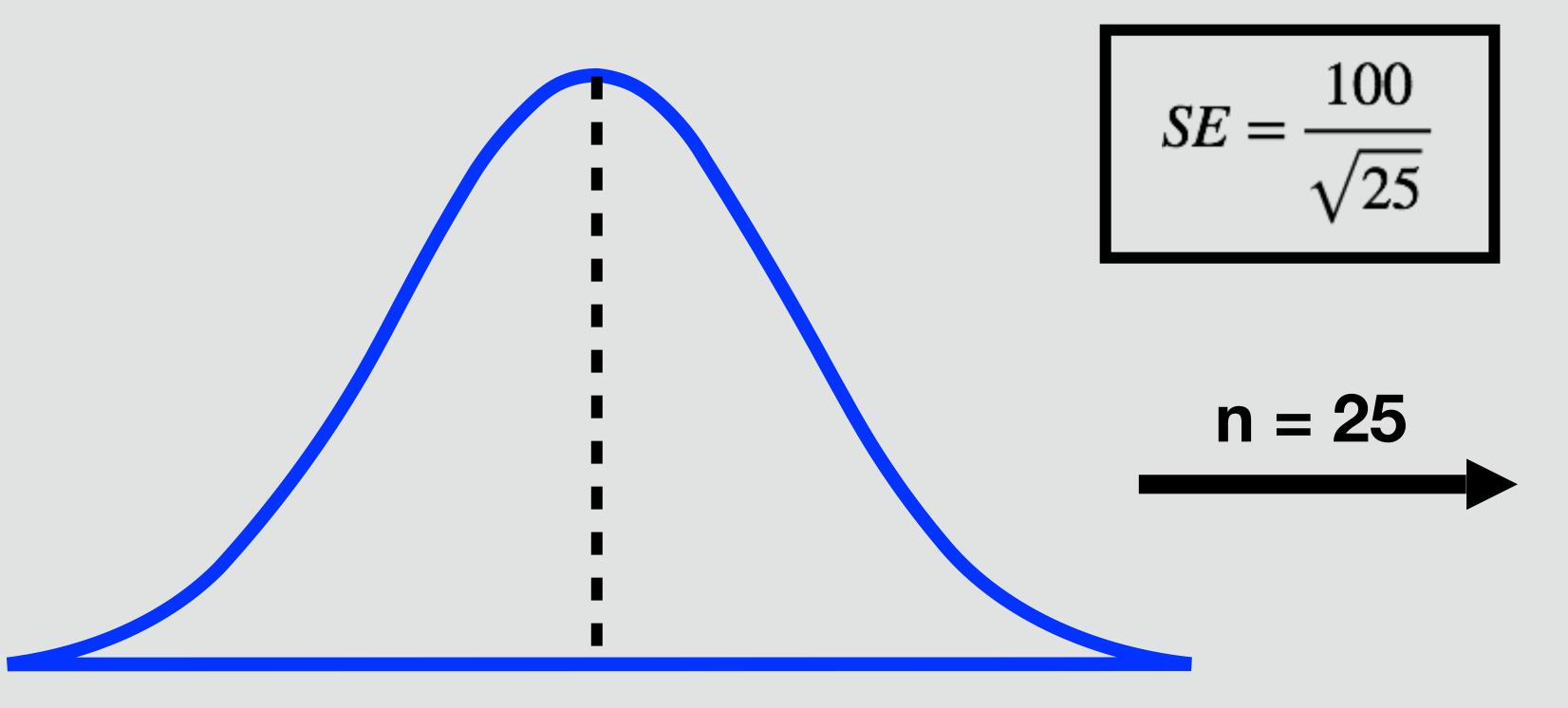
**Standard deviation: 100** 

**Mean:** 1000

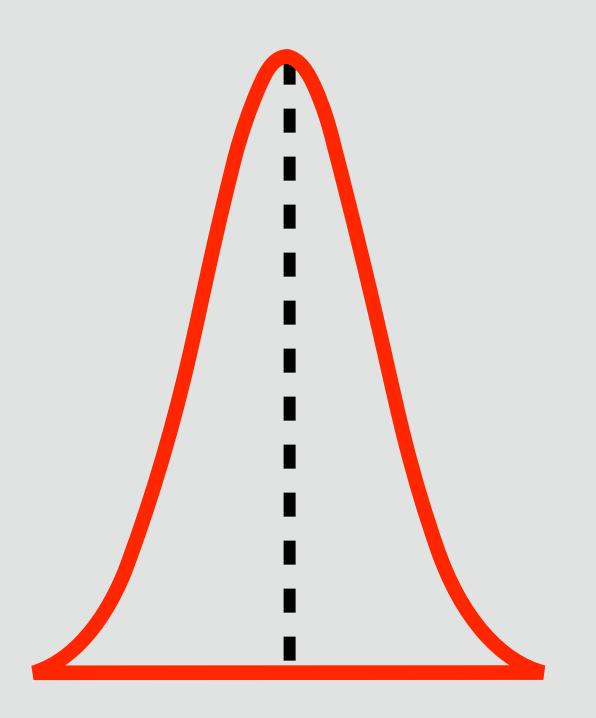
Standard error: 50

### Sample Size affects the sampling distribution

#### **SAT Population Distribution**



**SAT Sampling Distribution (n = 25)** 



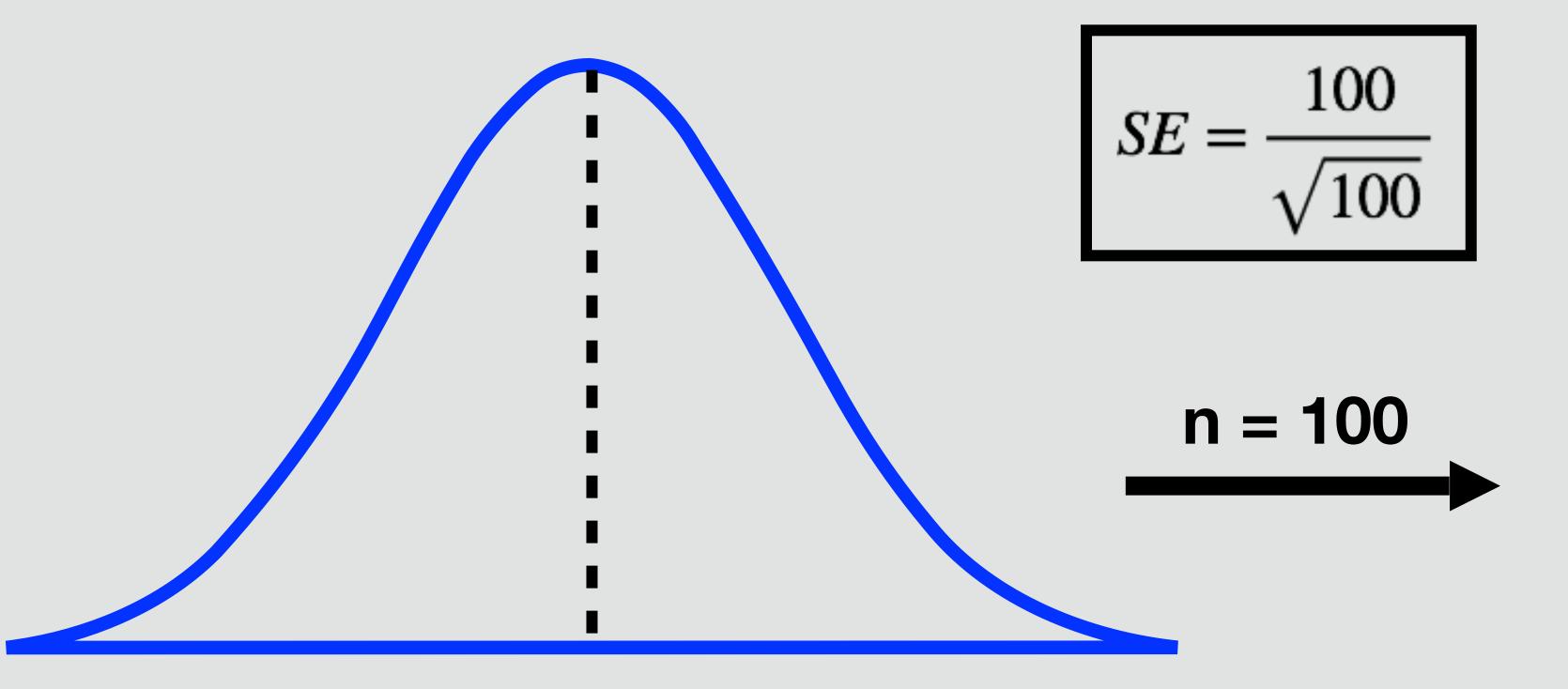
**Mean:** 1000

Standard error: 20

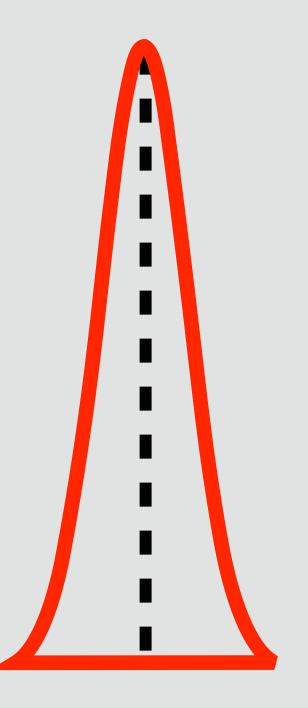
Mean: 1000 Standard deviation: 100

### Sample Size affects the sampling distribution

#### **SAT Population Distribution**



**SAT Sampling Distribution (n = 100)** 



**Mean:** 1000

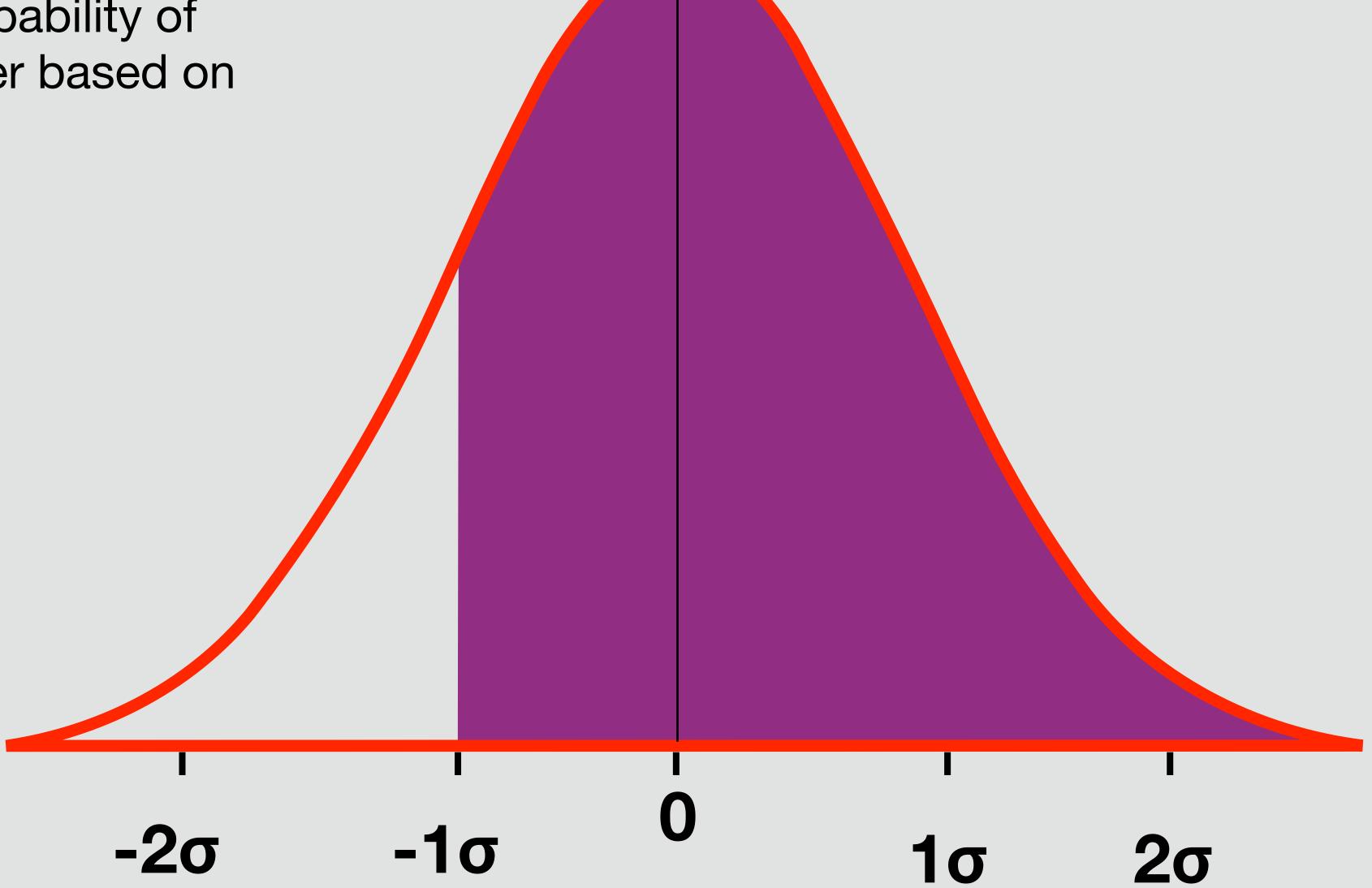
Standard error: 10

Mean: 1000 Standard deviation: 100

### Probability <> area under the curve

 We can calculate the probability of drawing a mean or greater based on area under the curve

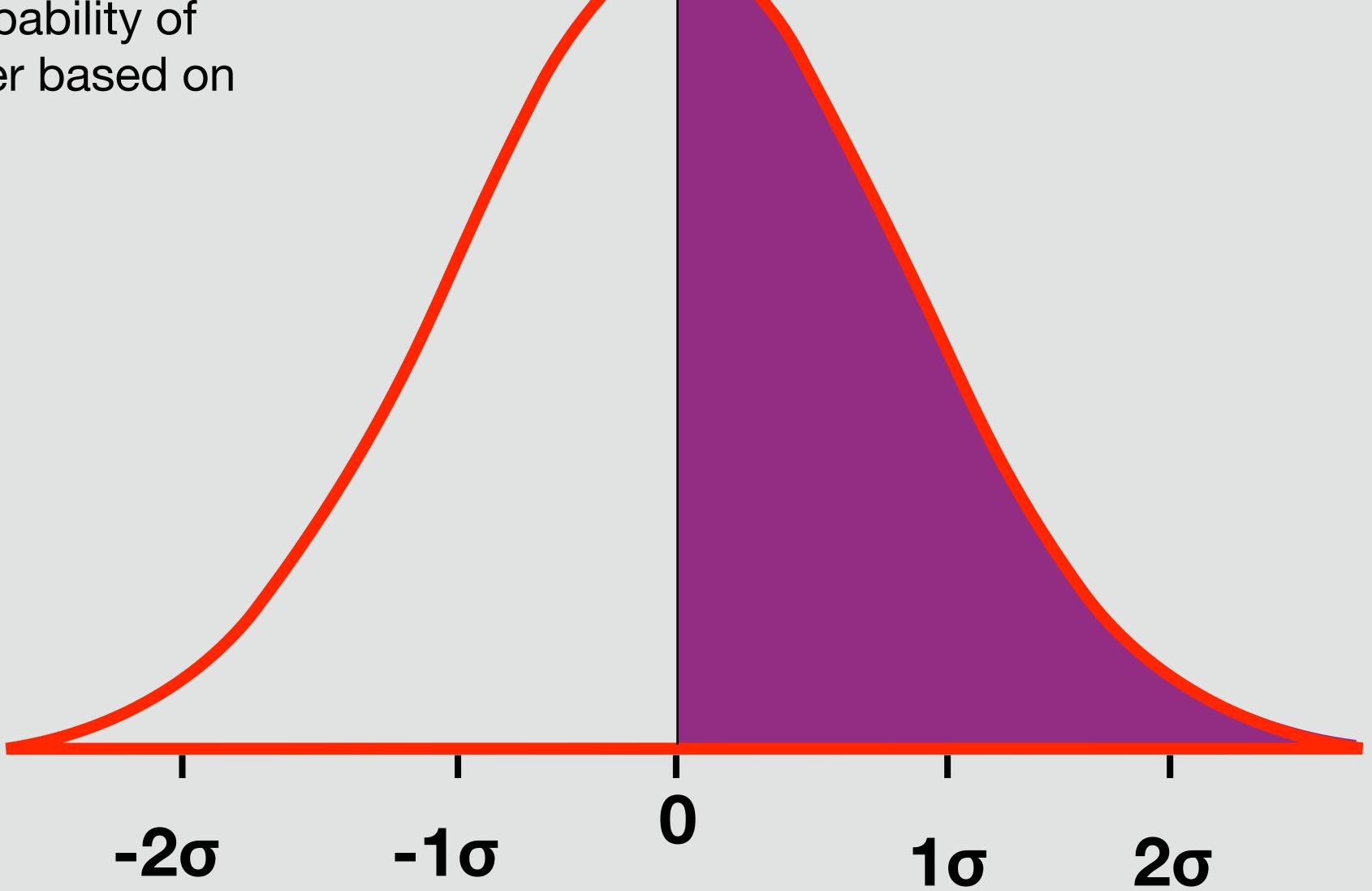
P = 0.85



### Probability <> area under the curve

 We can calculate the probability of drawing a mean or greater based on area under the curve

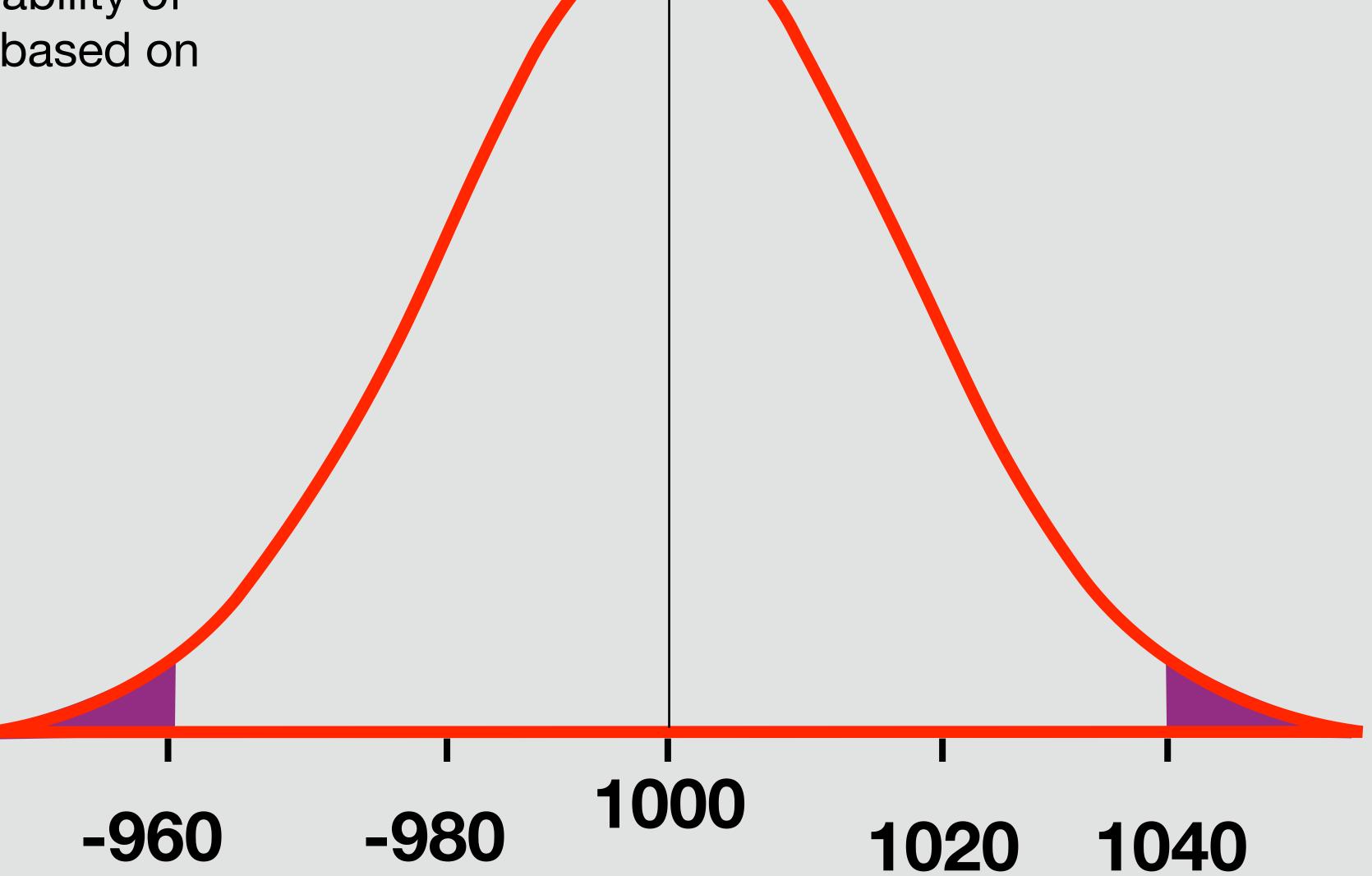
P = 0.50



### Probability <> area under the curve

 We can calculate the probability of drawing ranges of means based on area under the curve

P = 0.046



## Back to the SAT example

• Standard SAT population:

• Mean: 1000

• STD: 100

• Sampling distribution (n = 25):

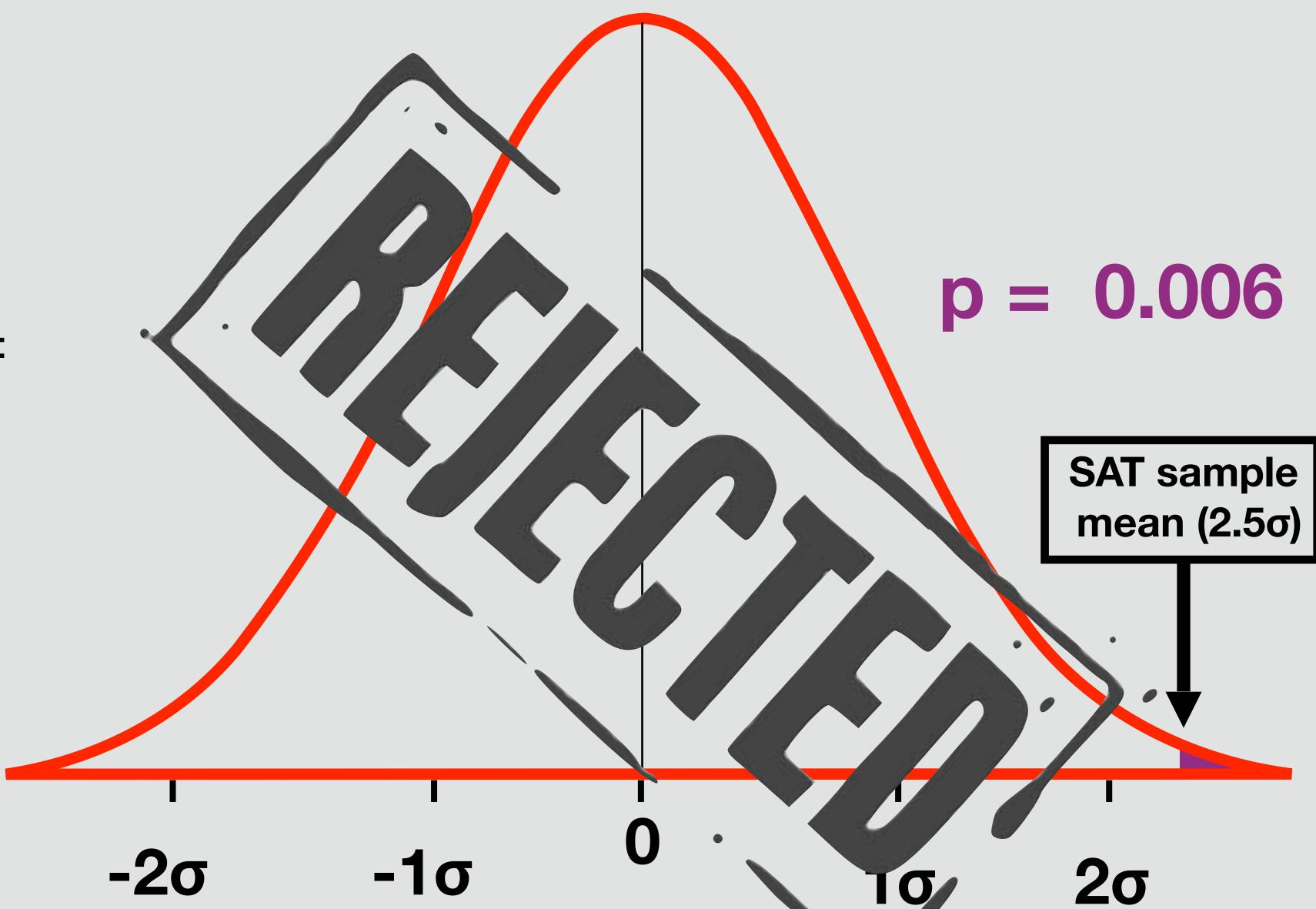
• Mean: 1000

Standard Error: 20

Experimental Results

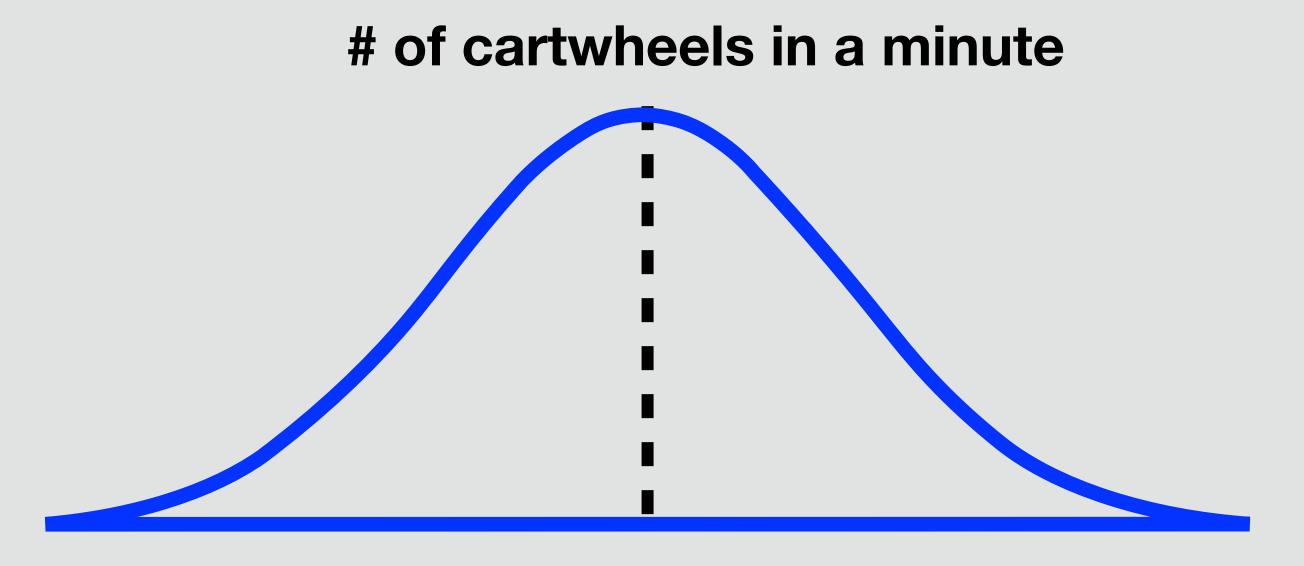
• Mean: 1050

$$Z = \frac{1050 - 1000}{20}$$



### Where the Z-test falters...

- We used *known* population parameters, mean and standard deviation to construct a sampling distribution
- It is rare to have that information



**Mean: ???** 

Standard deviation: ???

## ... the t-test prevails

- We don't know the exact population mean and standard deviation, so we conservatively estimate them based on samples themselves
- If we didn't know SAT population parameters we might:
  - Test one random sample of 25, untreated as the control
  - Test another random sample of 25 who studied the flash cards
  - Compute test statistic and p-value based on t-distribution, as opposed to the normal (z)
    distribution

$$t = \frac{M_1 - M_2}{SE} \qquad SE = \sqrt{\frac{s_1^2 + s_2^2}{n}}$$

## Summary

- The Central Limit Theorem makes sampling distributions predictable
- We construct a sampling distribution based on the null-hypothesis
- If it seems unlikely that the observed data came from that distribution, we reject the null hypothesis
- Use a t-test (or some other hypothesis test) when population parameters are unknown (which is most of the time). However

## Further reading

- Khan on T-distribution vs Z-distribution <a href="https://www.youtube.com/watch?v=5ABpqVSx331">https://www.youtube.com/watch?v=5ABpqVSx331</a>
- Free Udacity course on inferential statistics <a href="https://www.udacity.com/course/intro-to-inferential-statistics--ud201">https://www.udacity.com/course/intro-to-inferential-statistics--ud201</a>
- Wikipedia (has surprisingly good articles on these topics)
  - https://en.wikipedia.org/wiki/Sampling\_distribution
  - https://en.wikipedia.org/wiki/Student%27s\_t-test
  - https://en.wikipedia.org/wiki/Confidence\_interval
  - https://en.wikipedia.org/wiki/Analysis\_of\_variance