

# INTRODUCTION TO SAMPLE SIZE CALCULATIONS

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# Objectives

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- 1) Understand the concepts of sample size, power, effect size, and types of error
- 2) Understand the tradeoffs between these different parameters



# Sample Size & Power

# Why estimate sample size?

## Purpose

- Justify the cost and scope of the proposed research
- Define your assumptions about the expected effect size and the minimum relevant effect size
- Anticipate challenges (e.g., loss to follow-up)
- *Make sure your research is not doomed to fail (be underpowered to draw conclusions)*

# Example: Unbalanced Coin

- Flip a coin 4 times
  - ▣ 3 heads, 1 tail
  - ▣ Do you believe it is an unbalanced coin?
- Flip a coin 40 times
  - ▣ 30 heads, 10 tails
  - ▣ Now do you believe it is an unbalanced coin?
- Flip a coin 400 times
  - ▣ 300 heads, 100 tails
  - ▣ Are you convinced this is an unbalanced coin?

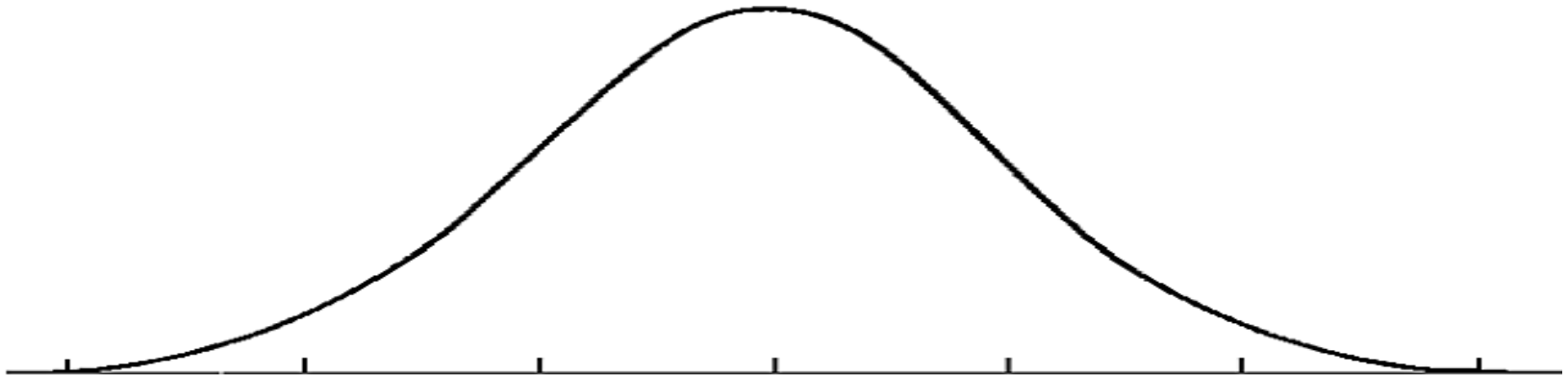


# Example: Unbalanced Coin

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With a larger sample size, we become more sure that the difference we **observe** is a **true** difference and not due to **chance**

# Example: Unbalanced Coin

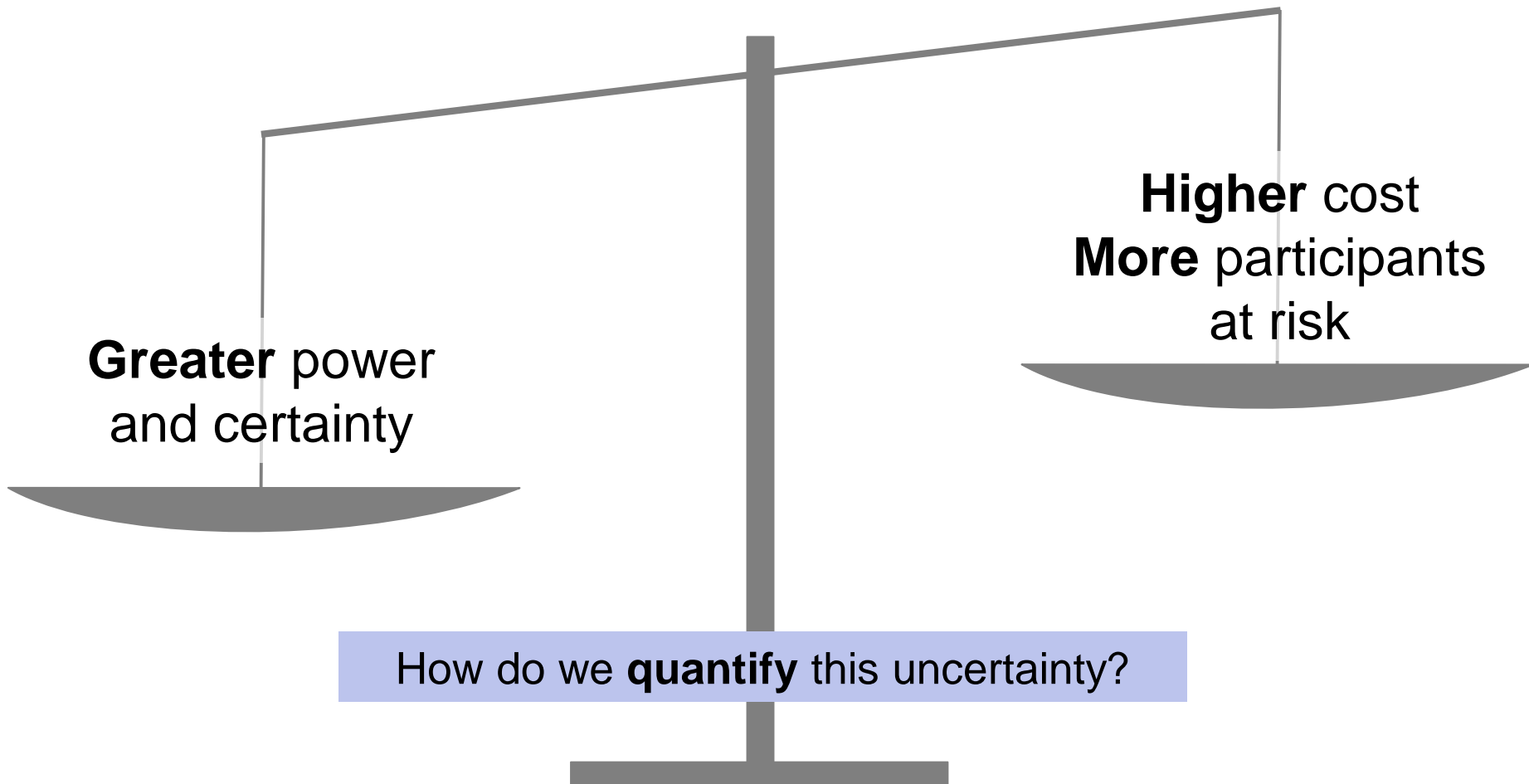






# How willing are we to risk making the **wrong** conclusion?

Increasing sample size has **tradeoffs...**



# P-values

- p-value tells us what the probability is that we would see the observed difference due to **chance** EVEN IF there was truly **no** difference
  - *How likely is it that the coin toss would show 30 heads and 10 tails if it was truly a balanced coin?*
- p-value **decreases** with greater number of subjects

Does a p-value tell us how **strong** an **association** is?

# Measures of association

- Calculated to explain *whether*, in *which direction* and *how strongly* an exposure is associated with an outcome
  - RR=1 means the exposure is ***not associated*** with the outcome
  - RR<1 means the exposure is ***negatively associated*** with the outcome and ***decreases*** risk
  - RR>1 means the exposure is ***positively associated*** with the outcome and ***increases*** risk
- RR = 0.8
  - “Subjects in the exposed group were 80% **as** likely / 20% **less** likely to have the outcome”
- RR = 1.5
  - “Subjects in the exposed group were 1.5 times **as** likely / 50% **more** likely to have the outcome.”

# P-value versus Effect Size

- My friend wants to improve her chances of getting pregnant; she can take one of two medications
- **Medication 1:** Increases her chance of getting pregnant by a factor of 1.1 and has a p-value of  $<0.0001$
- **Medication 2:** Increases her chance of getting pregnant by a factor of 4.5 and has a p-value of 0.1
- **Which medication would you choose?**



# What is Statistical Power?

- Power is a measure of our ability to convincingly **show** a difference when there **truly is** a difference
  - ▣ (REMEMBER: p-value is a measure of how often we would **observe** a difference when there **truly was no** difference)
- If I asked you to find out whether a coin was balanced but I only let you toss the coin 4 times, would you be able to make a convincing conclusion?
- **There is no point in doing a study unless we know that if there is a true difference, we will be able to show it**

# Components of Statistical Power

## Power is a function of four parameters

- Effect size
  - ▣ How **big** is the effect of your intervention?
- Sample size
  - ▣ How **many people** (or observations) do you have?
- Probability of false negative ( $\beta$  error)
  - ▣  $(1 - \beta)$  is **power**
- Probability of false positive ( $\alpha$  error)

# Components of Statistical Power

## Power is a function of four parameters

- Effect size
  - ▣ How **unbalanced** is your coin?
  - ▣ What if it only showed heads 60% of the time? (6/10)
- Sample size
  - ▣ How **many times** do you need to flip your coin?
- Probability of false negative ( $\beta$  error)
  - ▣ How willing are you to conclude that the coin is **BALANCED** when it is truly **UNBALANCED**
- Probability of false positive ( $\alpha$  error)
  - ▣ How willing are you to conclude that the coin is **UNBALANCED** when it is truly **BALANCED**?

# Alpha and Beta?

## Prisoner Example

		TRUTH	
		Guilty	Innocent
VERDICT	Prison	OK	<b>Problem</b>
	No prison	<b>Problem</b>	OK

**Which is worse?**

Problem A in **RED** or Problem B in **GREEN**?



# Alpha and Beta?

## Study Example

		TRUTH	
		Effect	No Effect
OBSERVED	Effect	OK	<b>Problem</b> ( $\alpha$ error)
	No Effect	<b>Problem</b> ( $\beta$ error)	OK

## Which is worse?

Problem A “**alpha –Type I error**” and Problem B “**beta-Type II error**”

Alpha usually 0.05 and beta usually 0.2 or 0.1

Willing to accept that 20% of the time we won't be able to convincingly show that there is a difference, even when there **truly is** a difference; and 5% of the time we will conclude that there **is** a difference, even when there truly is **NO** difference.

# What to solve for?

## Power is a function of four parameters

- Effect size
- Sample size
- Power ( $1-\beta$ )
- Probability of false positive ( $\alpha$  error)

**Like in algebra, set 3 equal to real numbers and solve for the 4th**

# What to solve for?

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## I have...

- Effect size
- Power ( $1-\beta$ )
- Probability of false positive ( $\alpha$  error)

## What do I solve for?

- Sample size

# What to solve for?

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## I have...

- Sample size
- Effect size
- Probability of false positive ( $\alpha$  error)

## What do I solve for?

- Power ( $1-\beta$ )

# What to solve for?

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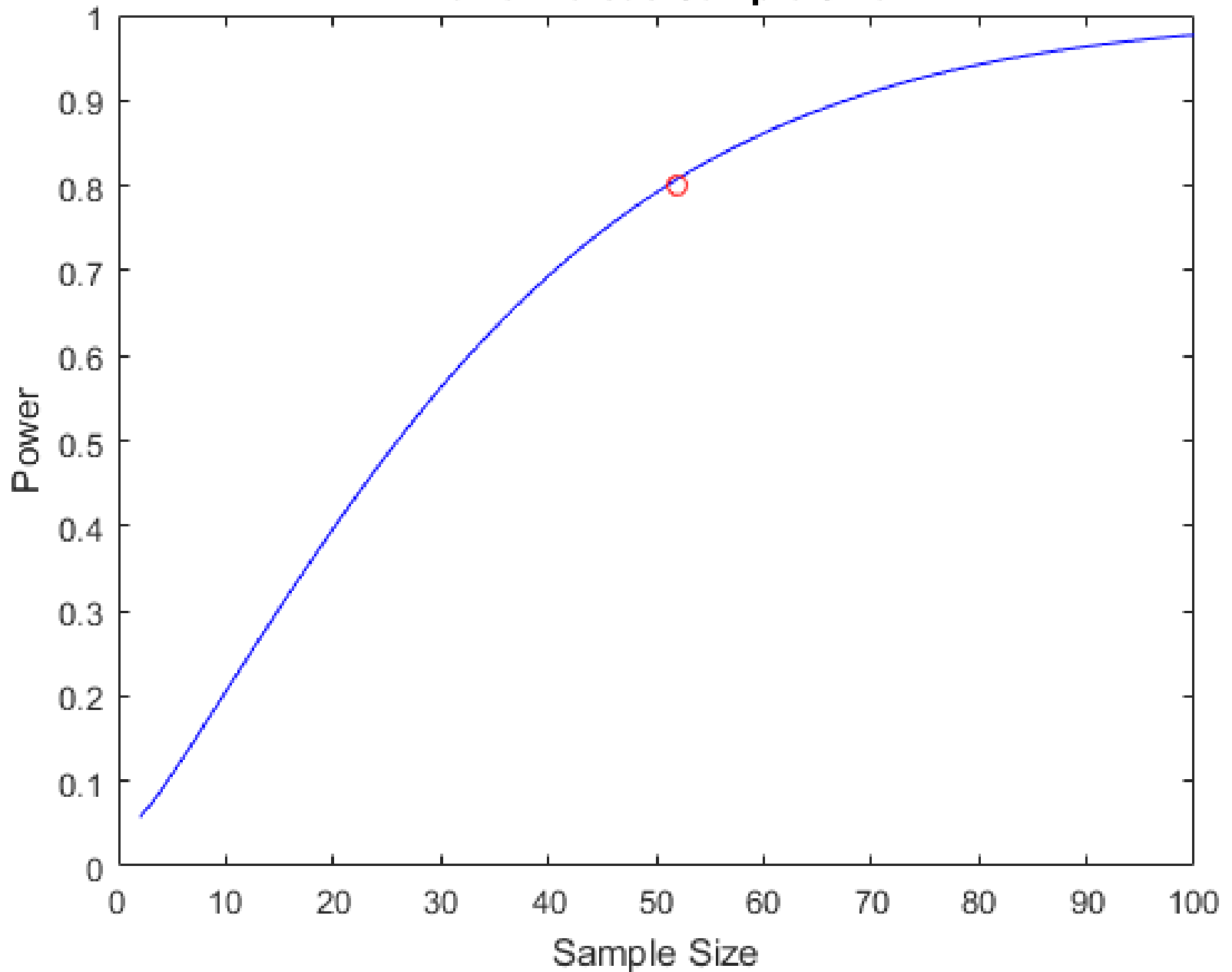
## I have...

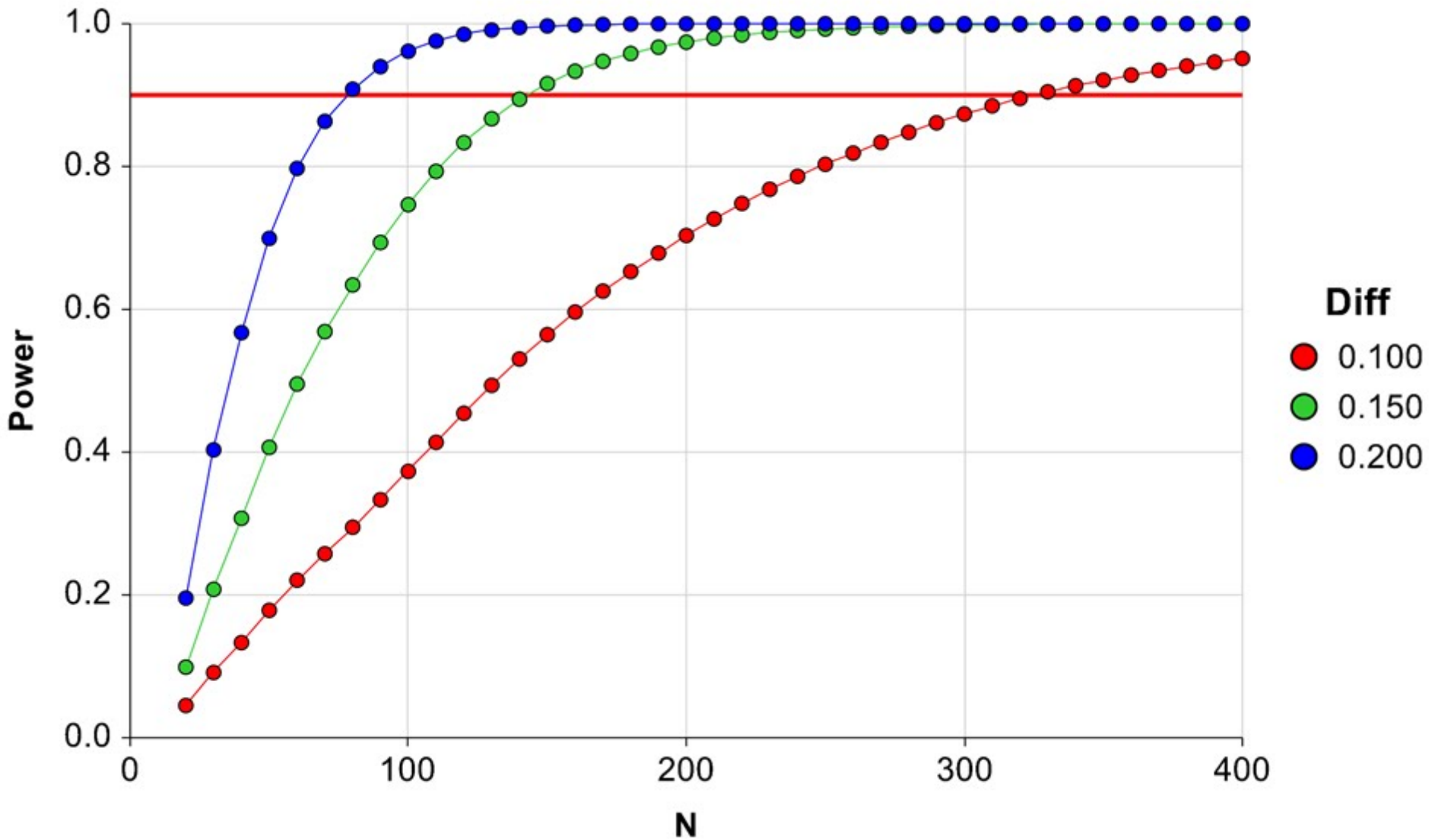
- Sample size
- Power ( $1-\beta$ )
- Probability of false positive ( $\alpha$  error)

## What do I solve for?

- Effect size

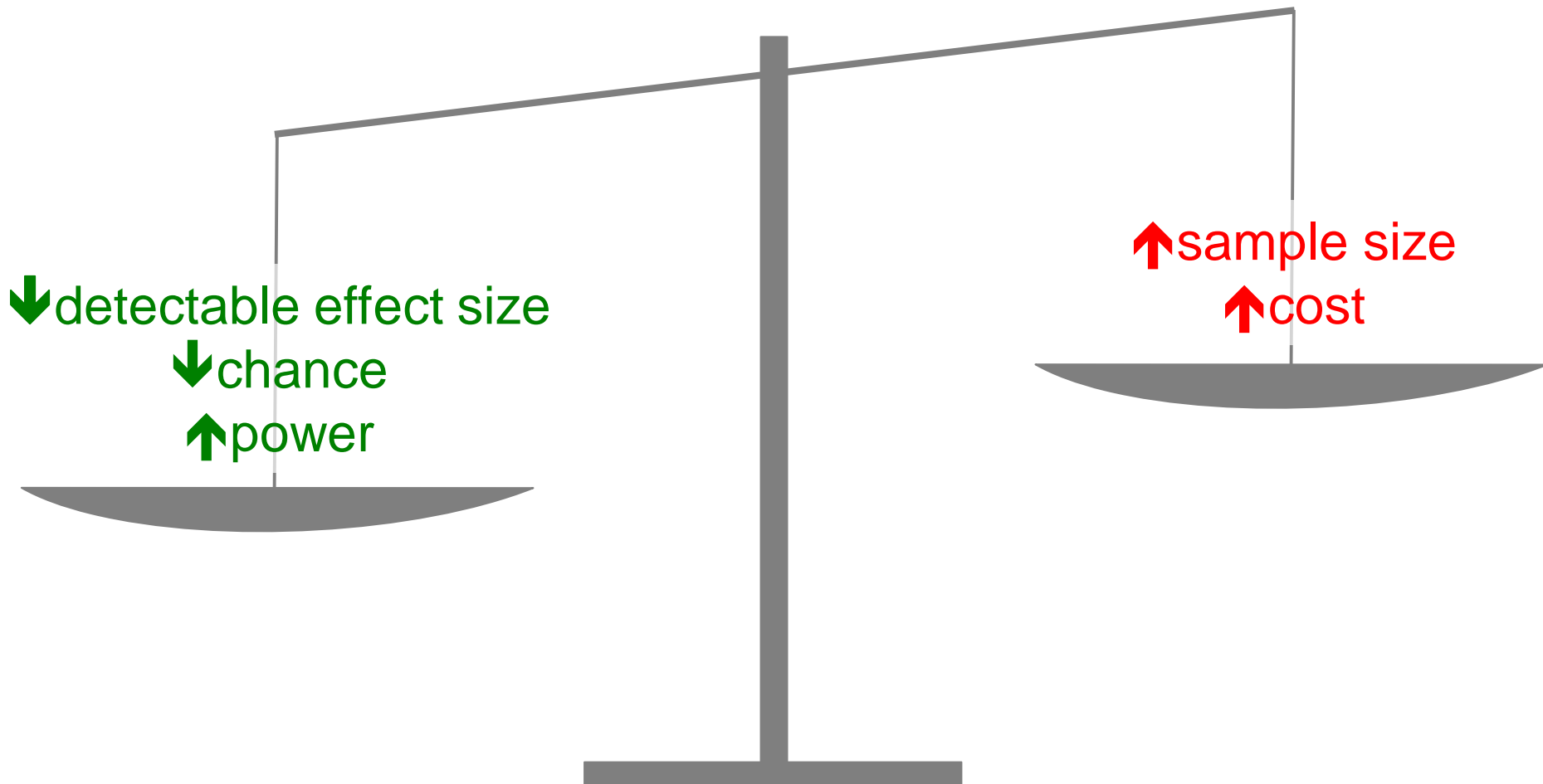
## Power versus Sample Size





# How are these concepts connected?

There are **tradeoffs**...





# Working with a statistician

- Statistician needs 3 of the 4 values:
  - ▣ Effect size
  - ▣ Sample size
  - ▣ Power ( $1-\beta$ )
  - ▣ Probability of false positive ( $\alpha$  error)

# Working with a statistician

- Statistician needs 3 of the 4 values:
  - ▣ Effect size – estimate from the literature or decide what effect size would be clinically meaningful
  - ▣ Sample size – may have a fixed number of participants or a limited budget
  - ▣ Power ( $1-\beta$ ) – usually 80% (sometimes 90% in clinical studies)
  - ▣ Probability of false positive ( $\alpha$  error) – usually 5%

# Working with a statistician

- Example: I want to test whether initiating HIV-positive children on a more palatable ART regimen reduces the chances of virologic failure after 6 months of treatment
  - ▣ Sample size: UNKNOWN
  - ▣ Effect size: Unsure, but anything less than a 10% reduction in virologic failure would be clinically irrelevant. Current regimens have probability of failure of 30%
  - ▣ Power: 80%
  - ▣ Alpha: 5%
- Statistician will tell you how many people you would need to enroll to detect a difference as small as a 10% reduction in virologic failure from the baseline of 30% failure

# Working with a statistician

- Example: I want to test whether viral loads of HIV-positive children who received an intensive ART adherence intervention are higher or lower than children who received standard of care adherence counseling. However, I only have funds to process a total of 50 samples
  - ▣ Sample size: 50 samples, 25 with the intervention, 25 SoC
  - ▣ Effect size: Unknown
  - ▣ Power: 80%
  - ▣ Alpha: 5%
- Statistician will tell you how much of a difference in viral load you would be able to detect with 50 samples

# Calculating your own sample size

- Manual formulas or statistical software both valid options
- Most important to understand the **inputs** well
- Many statistical software options (non-exhaustive list)
  - Stata, SPSS, R: all flexible and powerful, require knowledge of coding
  - EpiInfo (StatCalc): free and user-friendly, also have web-based and mobile-based options
  - Power Sample Size (Vanderbilt University) is free and easy to use, does not require coding:  
<http://biostat.mc.vanderbilt.edu/wiki/Main/PowerSampleSize>

# STATA Examples

```
. power twoproportions 0.8 0.9, alpha (0.05) power (0.80)
```

```
Performing iteration ...
```

```
Estimated sample sizes for a two-sample proportions test
```

```
Pearson's chi-squared test
```

```
Ho: p2 = p1 versus Ha: p2 != p1
```

```
Study parameters:
```

```
alpha = 0.0500  
power = 0.8000  
delta = 0.1000 (difference)  
p1 = 0.8000  
p2 = 0.9000
```

```
Estimated sample sizes:
```

```
          N = 398  
N per group = 199
```

```
.
```

# STATA Examples

```
. power twoproportions 0.6 0.9, alpha (0.05) power (0.80)
```

Performing iteration ...

Estimated sample sizes for a two-sample proportions test  
Pearson's chi-squared test  
Ho:  $p_2 = p_1$  versus Ha:  $p_2 \neq p_1$

Study parameters:

```
alpha =    0.0500  
power =    0.8000  
delta =    0.3000 (difference)  
p1 =      0.6000  
p2 =      0.9000
```

Estimated sample sizes:

```
      N =      64  
N per group =    32
```

```
.
```

# STATA Examples

```
. power twomeans 2.5 2.75, alpha (0.05) power (0.80)
```

```
Performing iteration ...
```

```
Estimated sample sizes for a two-sample means test  
t test assuming sd1 = sd2 = sd  
Ho: m2 = m1 versus Ha: m2 != m1
```

```
Study parameters:
```

```
alpha = 0.0500  
power = 0.8000  
delta = 0.2500  
m1 = 2.5000  
m2 = 2.7500  
sd = 1.0000
```

```
Estimated sample sizes:
```

```
N = 506  
N per group = 253
```

```
.
```



# 5-minute activity break

Using YOUR study question, imagine you are either working with a biostatistician to do power calculations OR calculating your own sample size. Define 3 of the following 4 parameters and write a 1 sentence description of what the biostatistician or software will tell you.

- Effect size
- Power
- Sample size
- Alpha error



# Summary



# Statistical power and sample size

- Components of statistical power:
  - ▣ Effect size
  - ▣ Sample size
  - ▣ Power
  - ▣ Alpha error

# Questions?

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