

#### **Statistical Principles for Clinical Development**

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# Learning Objectives



- To understand basic statistical principles relevant to clinical studies
- To understand the concepts of bias and variability
- To correctly understand p-values and hypothesis testing
- To understand issues around multiplicity

Design and Conduct are more important than Analysis



 In other words: Analysis cannot make up for poor design and conduct

Focus on good design and conduct, and analysis will be straightforward

# Stages of a Study



- Design: The conception, planning, and specification of the study
- Conduct: The running of the study
- Analysis: The analysis of the study (Number crunching)
- Reporting

# Adequate and Well-Controlled Study 21CFR314.126



- Clear objectives, summary of methods and results
- Design permits a valid comparison with a control
- Adequate selection of patients
- Assigning patients to treatment and control groups minimizes bias
- Adequate measures to minimize biases on subjects, observers, and analysts
- Well-defined and reliable assessment of subjects' responses
- Adequate analysis to assess drug results



# Randomized v. Observational Studies

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 Randomized study: characteristics of patients receiving drug A are similar to characteristics of patients receiving drug B

 Observational study: characteristics of patients receiving drug A may not be similar to characteristics of patients receiving drug B

# Confounding



• Without randomization there may be systematic differences (bias) when comparing people getting Drug A and people getting Drug B

This is known as confounding

Example: Drug A may be given to older sicker people. Even if there was no differences between the effects of Drug A and Drug B, the comparison may show Drug A has worse outcomes

Note: There are other sources of biases to be concerned about that may exist even with randomization. Example: Bias on part of observers (lack of blinding) PERSONAL HEALTH

#### The Health Benefits of Coffee

Drinking coffee has been linked to a reduced risk of all kinds of ailments, including Parkinson's disease, melanoma, prostate cancer, even suicide.





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## Variability v. Bias





#### Bias is worse than Variability



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# Variability v. Bias



- Statistics helps to quantify variability (in the analysis stage)
- Design and conduct generally reduce bias Examples: randomization, blinding
- Note: Statistics helps reduce bias and variability at each stage, design, conduct, and analysis

Reducing Variability with More Sample Size



• Sample size: number of people in study

• What is a better estimate of the average age of this session's attendees?

A. Pick a random sample of 5 attendees and calculate their average age

B. Pick a random sample of 20 attendees and calculate their average age

## Hypothesis Testing



- Null Hypothesis: Typically, what you are trying to show is not true
  - Ex: Drug A and placebo have the same effect (nothing)

• Alternative Hypothesis: Typically, what you are trying to show

#### Ex: Drug A has a better effect than placebo

## **P-Values**



• Probability of observing the effect or something more extreme, if the null hypothesis is true

Example: Study estimated drug effect is a reduction of 5mm of diastolic blood pressure. P-value: probability of observing 5 or 6 or 7 or 8 ..., if the drug had no effect

## **P-Values**



Small p-values are evidence against the null hypothesis (no drug effect)

Example: p-value = 0.02. If there was no drug effect, the chance of seeing what we saw or more extreme is 0.02. This is small. Leading us to doubt that there is no drug effect

P-value is **not** the probability that the null is true (no drug effect)

## **Decision Errors**



- Type 1 error: Concluding the drug has an effect on when it does not (FDA and society's problem) (and drug company's)
- Type 2 error: Concluding drug does not have an effect when it does (Drug company's problem) (and FDA and society's)

# Hypothesis Testing



Reject null hypothesis if p-value < alpha.</li>
Equivalent to saying the drug has an effect.
Example: alpha = 0.05

(Recall small p-values are evidence against the null hypothesis.)



### Sample Size and Power

- We can set the Type 1 error by choosing alpha
- We can limit the Type 2 error, by having larger sample size (more patients)
- More sample size = less variability = less likely to conclude drug has no effect when it really does (Type 2 error)

#### Multiplicity



#### (AKA: Multiple Bites from the Apple)

Example: Determining if a drug has effect

Drug has no effect

Study: Flip coin 4 times.

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H = a good outcome, T = a bad outcome
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If get 1, 2, or 3 H's, conclude drug has no effect. If get 4 H's, conclude drug has effect.

Study 1: HHTH Repeat study: Study 2: HTTT Study 3: THTH

Study 12: HHHH

# Multiplicity



- If you do enough studies or if you look at data in many ways, you will see an effect (even when there is no effect)
- This is known as data dredging or p-value hacking. It is a known problem with science

# Multiplicity



• Multiplicity can show up with multiple subgroups or endpoints

Subgroups: effect on males, effect on females, effect on people over 65

Endpoints: effect on blood pressure, effect on life expectancy, effect on happiness

• Prespecification: Tell the world ahead of time, what you will primarily look at Protocols and Statistical Analysis Plan (SAP) are how that is done.

# Challenge Question #1



#### Which of the following does not reduce bias?

- A. A larger sample size
- B. Randomization
- C. Blinding the knowledge of the drug from study participants and investigators
- D. Prespecification in the protocol and statistical analysis plan

# Challenge Question #2

#### FDA

#### Which of the following addresses multiplicity?

- A. A larger sample size
- B. Randomization
- C. Blinding the knowledge of the drug from study participants and investigators
- D. Prespecification in the protocol and statistical analysis plan

#### Resources



- Demonstrating Substantial Evidence of Effectiveness
- ICH E8(R1) General Considerations for Clinical Studies
- ICH E9 Statistical Principles for Clinical Trials
- ICH E9(R1) Estimands and Sensitivity Analysis in Clinical Trials
- <u>Multiple Endpoints in Clinical Trials</u>
- Adaptive Designs for Clinical Trials
- Adjusting for Covariates in Randomized Clinical Trials



# **Thank You!**

