

# Adaptive Sample Size Designs for Comparative Effectiveness Clinical Trials

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# Overview

1. Why Comparative Effectiveness Research?
2. Why Adaptive Designs?
3. Adapting Sample Size in  
Comparative Effectiveness Trials



# Comparative Effectiveness Research (CER)

American Recovery and Reinvestment Act<sup>1</sup>

- \$1.1 billion towards CER

Institute of Medicine Definition<sup>1</sup>

- *"... the generation and synthesis of evidence that compares the benefits and harms of alternative methods to prevent, diagnose, treat, and monitor or improve the delivery of care ... at both the individual and population levels"*



# Comparative Effectiveness Research

- Compares treatments that are *in practice*
- Focuses on the population as well as the individual
- Any *reliable* difference that is large enough to affect public behavior is important at the population level
- Many types of comparative effectiveness research
- We will be focusing on CE using randomized clinical trials



# Randomized Clinical Trials

- "Gold standard" for evidence-based practice in medicine
- How good is the "gold"?
  - Purpose: determine if the treatment is efficacious
    - Often do not make comparisons to other efficacious treatments
    - Compared to a standard treatment
  - Rigid inclusion/exclusion criteria



# Sample Size Calculation

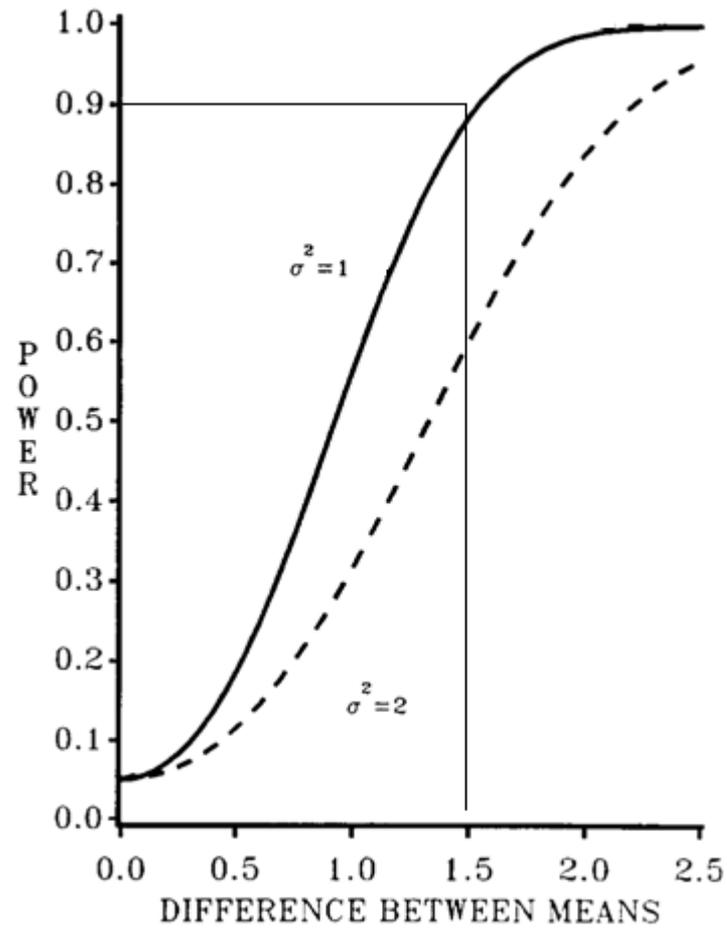
1. Choose study design, test, target test size  $\alpha_t$  and power  $P_t$
2. Determine planning variance,  $\sigma_0^2$ , and clinically important size of effect,  $\delta_0$ , to pick total sample size  $n_0$

From a linear models perspective, powers varies with increasing:

- Sample size (  $\uparrow$  )
- Size of Effect (  $\uparrow$  )
- Variance (  $\downarrow$  )



# Example: t-test Power



# Comparative Effectiveness Trials

- CE has been performed using randomized trials
- However, head-to-head comparisons of treatments are rare due to:
  - Lack of funding
  - Lack of methodology
- Standard clinical trial design may not be optimal for CER
- We will present a method for performing CE trials



# Comparative Effectiveness Trials

Unique aims for CE trials

- Compare two or more efficacious treatments
- Broader inclusion criteria reaching a wider group of individuals

Unique challenges for CE trials

- Small differences in treatments expected
- Large response variance expected

Solution: Use Adaptive Designs



# Adaptive Designs (AD)

## *Adaptive*

- Modifying study characteristics based on accumulating information

## *Design*

- Adaptations are *planned*
- Consistent with the FDA guidance<sup>2</sup> the PhRMA working group<sup>3</sup> stated:
  - "...modify aspects of the study as it continues, without undermining the validity and integrity of the trial"
  - "...changes are made by design, and not on an ad hoc basis"
  - "...not a remedy for inadequate planning."



# Adaptive Design Examples

## Learning Phase (Phase I,II)

- Adaptive Dose Response

## Combined Phases (Phase I,II,III)

- Seamless Phase I/II and II/III

## Confirmatory Phase (Phase III)

- Adaptive Randomization
- Sample Size Adjustment (Sample Size Re-estimation)



# Types of Sample Size Adjustment

- Group Sequential (GS)
  - Early stopping of trial at interim analysis
- Internal Pilot (IP)
  - Adjusts total sample size based on interim estimates of nuisance parameters
- Effect Size Adjustment
  - Adjusts total sample size based on interim estimates of the size of effect
- Combinations of types
  - Univariate Gaussian Linear Model with single interim analysis: Internal Pilot with Interim Analysis<sup>4</sup> (IPIA)



# Types of Sample Size Adjustment

Types	Theory	Software	Controversy
GS	✓	✓	No
IP	✓	✓	No
Effect Size	✓	✓	Yes
IPIA	✓	✓	No

- Fully capable of implementing any design
- Concerns over SSA based on observed size of effect:
  - Inflation of type I error rate
  - Bias
  - Inefficiency
- Effect size adjustment valid only if planned in advance



# Adaptive Comparative Effectiveness Trial

Recommendation for CE trials: use a two stage group sequential design with interim sample size adjustment

Setting:

- Compare the effectiveness of treatment A and treatment B
- A and B have similar cost and safety profiles
- Small differences in effectiveness would be population important



# Designing an Adaptive Comparative Effectiveness Trial

1. Specify primary and secondary sizes of effect and planning variance
  - 1.1 Primary ( $\delta_1$ ): reasonable size of effect that can be shown in small or moderate sized trial
  - 1.2 Secondary ( $\delta_2$ ): small size of effect that would affect public behavior if true
  - 1.3 Planning variance value for the test statistic ( $\sigma_0^2$ )
  - 1.4 Determine other study parameters,  $\alpha_t$  and  $P_t$
  - 1.5 Statisticians and clinicians must work together to determine  $\delta_1$  and  $\delta_2$



# Designing an Adaptive Comparative Effectiveness Trial

## 2. Choose decision rule and calculate initial sample size

### 2.1 Choose method and decision rule

- Univariate Gaussian Linear Model: use IPIA
- Methods need to be developed for binary data
- Pocock stopping bounds are appropriate
  - Prefer stopping early if there is a difference
  - O'Brien-Fleming bounds will save most of the alpha for the second stage

### 2.2 Determine first stage sample size, $n_1$ , assuming $\delta_1$ and $\sigma_0^2$ are true



# Designing an Adaptive Comparative Effectiveness Trial

## 3. Interim analysis

3.1 Collect  $n_1$  observations, compute  $\hat{\sigma}_1^2$

3.2 Decision:

- Enough evidence to conclude efficacy or futility
- Else continue to second stage

3.3 Second stage sample size calculation,  $n_2$ , based on  $\delta_2$  and  $\hat{\sigma}_1^2$  that achieves  $P_t$

## 4. Complete study

4.1 Collect  $n_2$  observations

4.2 Conduct analysis



# Advantages of this method

1. Small first stage that will detect large treatment differences
2. Larger second stage useful for detecting small, but important differences
3. Accounts for possible misspecification of nuisance parameters
4. Ensures a correctly powered study



# Conclusions

- Randomized clinical trials are an important tool for use in comparative effectiveness research
- Comparative effectiveness trials have unique challenges
- Our method of adaptive sample size adjustment appears to offer a statistically valid solution to these problems
- Further research is underway to better define the properties of the proposed method



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