

# Adaptive Sample Size Design

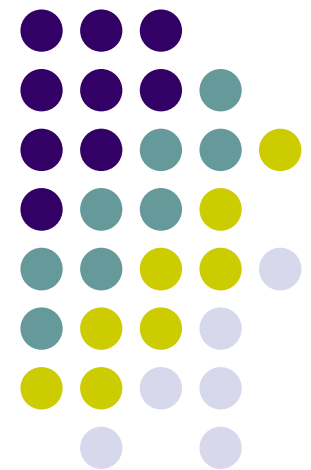
BASS 2008

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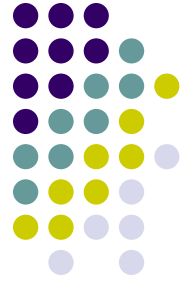
John Lawrence

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Food and Drug Administration



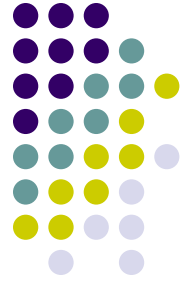
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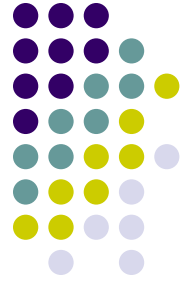
# Outline

- Overview
- Motivating example
- Estimates, confidence intervals and more
- Interpretations
- Discussions and Summary

# Introduction



- Sample size adjustment
  - Conditional power
    - Weighted test
    - Change of critical value
    - More



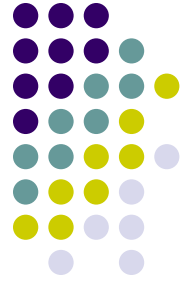
# Example

- Phase III trial for a cardiovascular indication
- Initial planned total sample size: 4000 subjects
- Expected number of events is 1600
- Two interim analyses were conducted
- By the time for third interim analysis, 830 events were collected from 3900 subjects



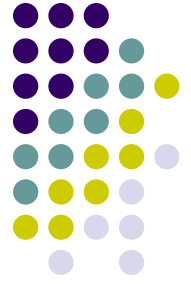
## Example (2)

- The sponsor proposed to adjust sample size at the third interim analysis based on paper by Chen, DeMets and Lan (CDL)
- Advantage of CDL method per Chen et al:
  - No inflation of type I error if the conditional power at interim analysis is larger than 50%
  - Ordinary test statistic can still be used



## Example (3)

- Some concerns:
  - Is the traditional point estimate and confidence interval still valid under this design?
  - The sample size adjustment is near the end of the trial enrollment
  - Potential operational bias



## CDL Method

- Chen, DeMets and Lan (2004) showed that if increasing sample size when the conditional power is greater than 50% at the interim look, the regular unweighted test statistic can still be used without inflating the type I error rate



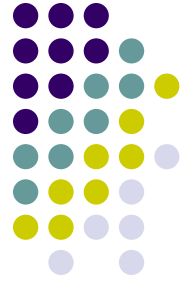
- In this presentation
  - Sample size increase only
    - No need for sample size reduction since there is efficacy boundary for early stopping
  - Sample size increase only at the last interim look prior to the final analysis
    - More data and better estimate of treatment effect
  - No more interim analysis after the sample size increase
- First, consider one interim analysis during a trial





# Notation

- $N_0$ : originally planned total sample size
- $N$ : new total sample size ( $N \geq N_0$ )
- $t_1$ : information time where the interim analysis is conducted
- $t^*$ : new maximum information time after sample size adjustment,  $t^* = N/N_0$
- $Z(t_1)$ : statistic computed at the interim analysis based on the first  $n$  subjects



## Notation (2)

- $Z(t_2)$ : final statistic based on the total  $N_0$  subjects
- $Z(t^*)$ : final statistic based on the total  $N$  subjects
- $c_k$ : critical value for  $k$ th interim analysis,  $k=1,2$

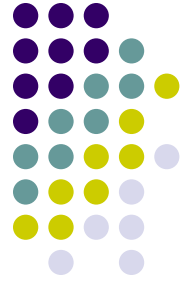


# CDL Method

- Under the null hypothesis  $H_0 : \theta = 0$ , the change of the conditional type I error is

$$\begin{aligned}\Delta &= P(Z(t^*) > c_2 \mid Z(t_1) = z, \theta = 0) - P(Z(t_2) > c_2 \mid Z(t_1) = z_1, \theta = 0) \\ &= \Phi\left(\frac{\sqrt{t_1} z_1 - \sqrt{t^*} c_2}{\sqrt{t^* - t_1}}\right) - \Phi\left(\frac{\sqrt{t_1} z_1 - c_2}{\sqrt{1 - t_1}}\right)\end{aligned}$$

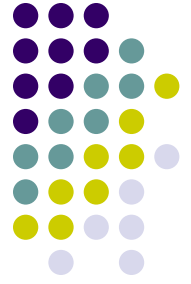
- Chen, DeMets and Lan showed that when the conditional power is greater than 50%, this change in conditional type I error decreases when N increase.



# CDL Method

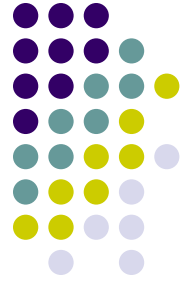
- the conditional power is greater than 50% if and only if the observed test statistic at the interim analysis is at least  $\sqrt{t_1}c_2$
- Under the condition  $z_1 \geq \sqrt{t_1}c_2$ , the change in type I error rate is

$$E\{P(Z(t^*) > c_2 \mid Z(t_1) = z_1, \theta = 0)\} - E\{P(Z(t_2) > c_2 \mid Z(t_1) = z_1, \theta = 0)\} \leq 0$$



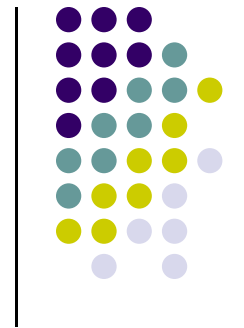
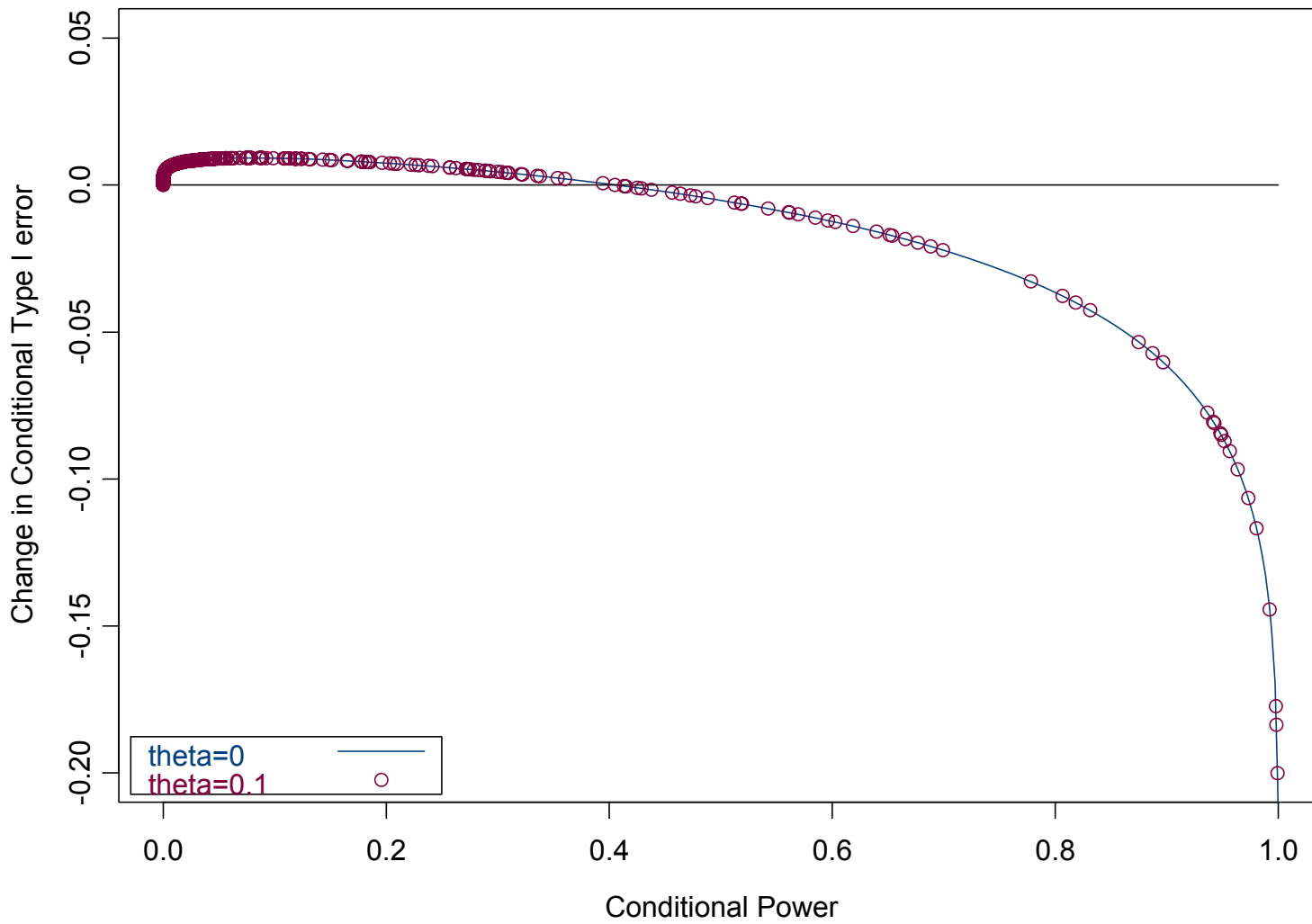
# CDL Method

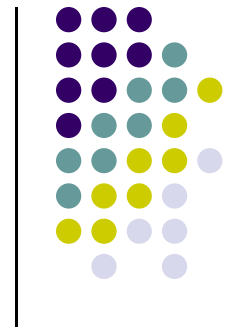
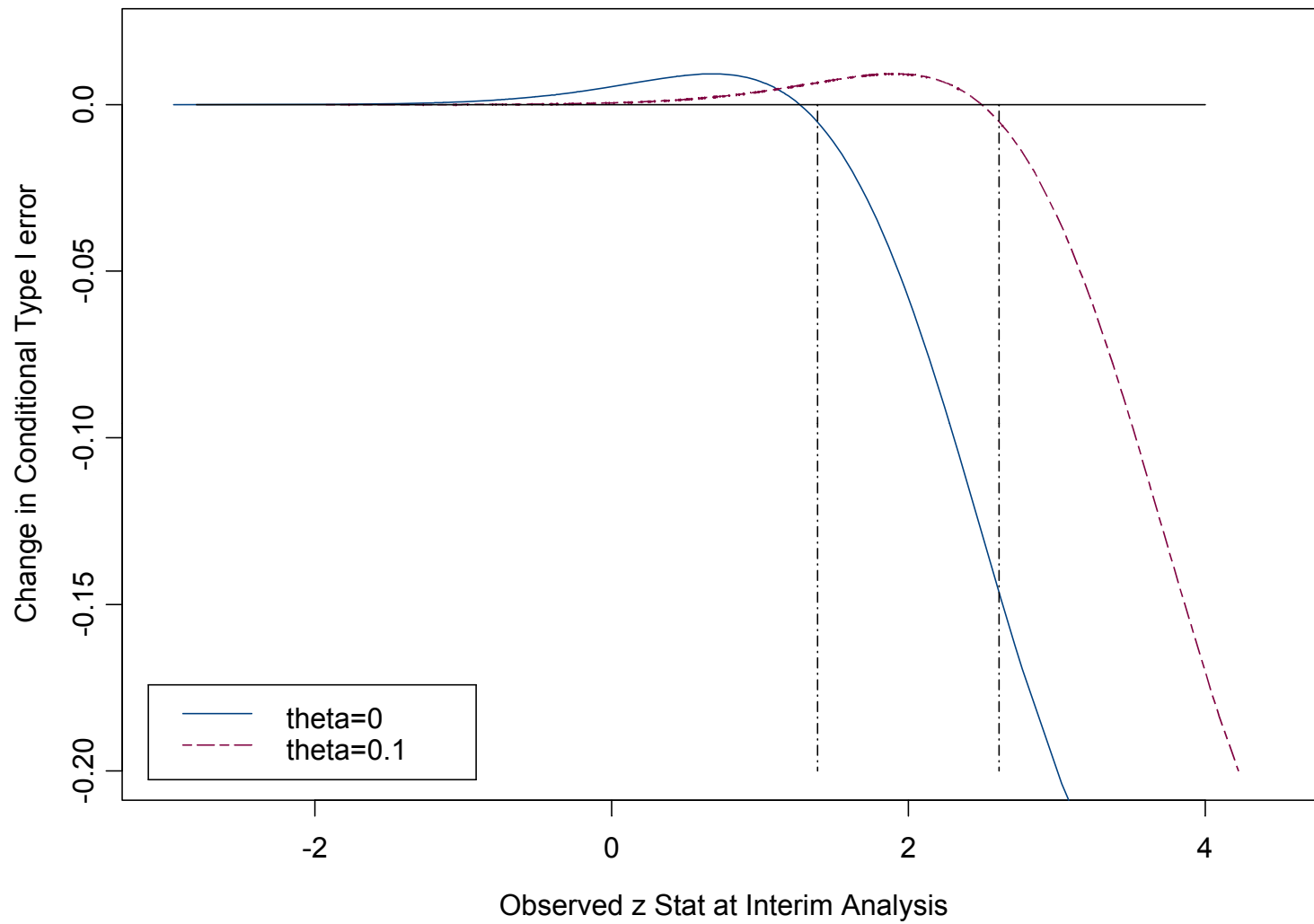
- Proposal: increase the sample size only when the first interim analysis test statistic is at least  $\sqrt{t_1}c_2$ . The unweighted test statistics can still be used and the critical values remain the same.
- The simulation in the paper also showed that the type I error is controlled if the CDL method applies to a trial with two or more interim analyses



# Extension

- Under null hypothesis  $H_0 : \theta = \theta_0$
- Increasing sample size when conditional power is no less than 50% will not inflate type I error.
- $CP \geq 50\%$  is equivalent to  $z_1 \geq \sqrt{t_1} c_2 + \sqrt{t_1} k \theta_0$  at the interim analysis
- $k$  depends on the total sample size  $N_0$  and  $\theta$  is the treatment effect





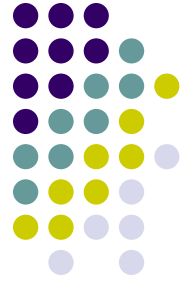




- In fact, increasing sample size at the interim look will not inflate type I error as long as the following condition is satisfied

$$z_1 > z^* = \sqrt{n}\theta_0 + \frac{\sqrt{t^* - t_1} - \sqrt{t^*(1 - t_1)}}{(\sqrt{t^* - t_1} - \sqrt{1 - t_1})\sqrt{t_1}} c_2$$

- $z^*$  is always smaller than  $\sqrt{t_1}c_2 + \sqrt{t_1}k\theta_0$  as long as  $N > N_0$

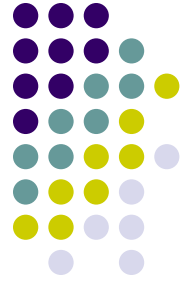


- Increasing the sample size may be conducted when the conditional power is less than 50% as long as the observed test statistic at interim look is larger than  $z^*$
- $z^*$  depends on the new information fraction  $t^*$  and cannot be decided until the interim look.



# Hypothesis Testing

- The total sample size will be increased if the conditional power exceeds  $L$  but below  $U$ 
  - In CDL method,  $L=50\%$
- The sample size remains the same
  - if the conditional power exceeds  $U$ , for example,  $U=80\%$ . The trial runs as expected
  - Or if the conditional power is below  $L$ . The trial is not promising enough to increase the sample size
- We did not take futility into consideration



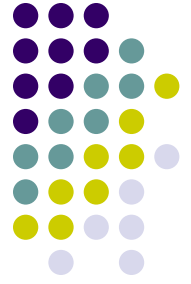
# Hypothesis Testing (2)

- Reject null hypothesis  $H_0 : \theta = \theta_0$  when
  - The total sample size is increased ( $50\% \leq CP < U$ ) and the final test statistic  $Z(t^*)$  based on the new sample size  $N$  exceeds critical value  $c_2$

$$C_1 : \frac{z - \sqrt{t_1} z_\alpha - \sqrt{t_1(1-t_1)} z_{1-\gamma}}{\sqrt{t_1} k} < \theta_0 \leq \frac{z - \sqrt{t_1} z_\alpha}{\sqrt{t_1} k} \quad \text{and} \quad Z^{(N)} - \sqrt{t^*} k \theta_0 > z_\alpha$$

- The total sample size remains the same ( $CP < 50\%$ , or  $CP \geq U$ ) and the final test statistic  $Z(t_2)$  based on  $N_0$  exceeds critical value  $c_2$

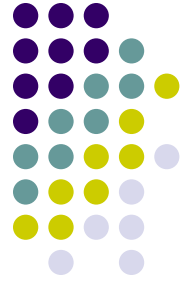
$$C_2 : \theta_0 > \frac{z - \sqrt{t_1} z_\alpha}{\sqrt{t_1} k} \quad \text{or} \quad \theta_0 \leq \frac{z - \sqrt{t_1} z_\alpha - \sqrt{t_1(1-t_1)} z_{1-\gamma}}{\sqrt{t_1} k} \quad \text{and} \quad Z^{(N_0)} - k \theta_0 > z_\alpha$$



## Hypothesis Testing (3)

- Reject hypothesis if one of the two conditions is satisfied:
  - Condition 1:  $b_2 < \theta_0 \leq b_1$  and  $\theta_0 < b_3$
  - Condition 2:  $(\theta_0 > b_1$  or  $\theta_0 \leq b_2)$  and  $\theta_0 < b_4$
- $b_1$ ,  $b_2$ ,  $b_3$ , and  $b_4$  are values depending on the interim test statistic, final test statistic, and information time  $t_1$

$$b_1 = \frac{z - \sqrt{t_1} z_\alpha}{\sqrt{t_1} k}, b_2 = \frac{z - \sqrt{t_1} z_\alpha - \sqrt{t_1(1-t_1)} z_{1-\gamma}}{\sqrt{t_1} k}, b_3 = \frac{z^{(N)} - z_\alpha}{\sqrt{t^*} k}, b_4 = \frac{z^{(N_0)} - z_\alpha}{k}$$



# Confidence Interval

- Depending on the relative positions of  $b_1$ ,  $b_2$ ,  $b_3$  and  $b_4$ , the one-sided confidence interval can be written as

$$\left\{ \begin{array}{ll}
 [b_4, \infty) & \text{if } b_3 = b_4 \text{ or } b_1 < \min(b_3, b_4) \text{ or } b_2 \geq \max(b_3, b_4) \\
 [b_3, \infty) & \text{if } b_2 < \min(b_3, b_4) \text{ and } b_1 \geq \max(b_3, b_4) \\
 (b_2, \infty) & \text{if } b_3 \leq b_2 < b_4 \leq b_1 \\
 (b_1, \infty) & \text{if } b_2 < b_4 \leq b_1 < b_3 \\
 [b_3, b_1] \cup [b_4, \infty) & \text{if } b_2 < b_3 \leq b_1 < b_4 \\
 (b_2, b_1] \cup [b_4, \infty) & \text{if } b_3 \leq b_2 < b_1 < b_4 \\
 [b_4, b_2] \cup (b_1, \infty) & \text{if } b_4 \leq b_2 < b_1 < b_3 \\
 [b_4, b_2] \cup [b_3, \infty) & \text{if } b_4 \leq b_2 < b_3 \leq b_1
 \end{array} \right.$$

<b>True Treatment difference</b>	<b>theta</b>	<b>COND 1</b>	<b>COND 2</b>	<b>COND 3</b>	<b>COND 4</b>	<b>COND 5</b>	<b>COND 6</b>	<b>COND 7</b>	<b>COND 8</b>
<b>0</b>	<b>0</b>	<b>77.0</b>	<b>21.9</b>	<b>0.4</b>	<b>0.1</b>	<b>0.1</b>	<b>0.1</b>	<b>0.0</b>	<b>0.4</b>
<b>0.02</b>	<b>0</b>	<b>76.4</b>	<b>21.5</b>	<b>0.6</b>	<b>0.2</b>	<b>0.3</b>	<b>0.1</b>	<b>0.1</b>	<b>0.8</b>
<b>0.05</b>	<b>0</b>	<b>75.1</b>	<b>20.8</b>	<b>0.8</b>	<b>0.8</b>	<b>0.6</b>	<b>0.1</b>	<b>0.4</b>	<b>1.5</b>
<b>0.1</b>	<b>0</b>	<b>73.9</b>	<b>19.2</b>	<b>0.6</b>	<b>2.6</b>	<b>0.8</b>	<b>0.1</b>	<b>0.9</b>	<b>1.9</b>
<b>0.15</b>	<b>0</b>	<b>75.3</b>	<b>19.3</b>	<b>0.2</b>	<b>3.1</b>	<b>0.4</b>	<b>0.0</b>	<b>0.9</b>	<b>0.9</b>
<b>0.2</b>	<b>0</b>	<b>77.0</b>	<b>20.9</b>	<b>0.0</b>	<b>1.5</b>	<b>0.1</b>	<b>0.0</b>	<b>0.3</b>	<b>0.2</b>
<b>0.02</b>	<b>0.02</b>	<b>77.0</b>	<b>21.9</b>	<b>0.4</b>	<b>0.1</b>	<b>0.1</b>	<b>0.1</b>	<b>0.0</b>	<b>0.4</b>
<b>0.05</b>	<b>0.02</b>	<b>76.0</b>	<b>21.3</b>	<b>0.7</b>	<b>0.4</b>	<b>0.3</b>	<b>0.1</b>	<b>0.2</b>	<b>1.0</b>
<b>0.1</b>	<b>0.02</b>	<b>74.0</b>	<b>19.9</b>	<b>0.7</b>	<b>1.8</b>	<b>0.8</b>	<b>0.1</b>	<b>0.7</b>	<b>1.9</b>
<b>0.15</b>	<b>0.02</b>	<b>74.6</b>	<b>18.9</b>	<b>0.3</b>	<b>3.1</b>	<b>0.6</b>	<b>0.0</b>	<b>1.0</b>	<b>1.3</b>
<b>0.2</b>	<b>0.02</b>	<b>76.5</b>	<b>20.1</b>	<b>0.1</b>	<b>2.2</b>	<b>0.2</b>	<b>0.0</b>	<b>0.5</b>	<b>0.4</b>
<b>0.15</b>	<b>0.15</b>	<b>77.0</b>	<b>21.9</b>	<b>0.4</b>	<b>0.1</b>	<b>0.1</b>	<b>0.1</b>	<b>0.0</b>	<b>0.4</b>
<b>0.2</b>	<b>0.15</b>	<b>75.2</b>	<b>20.7</b>	<b>0.8</b>	<b>0.8</b>	<b>0.5</b>	<b>0.1</b>	<b>0.4</b>	<b>1.5</b>

Based on 1,000,000 replications



## Confidence Interval (2)

- Simplified confidence interval

$$\begin{cases} [b_4, \infty) & \text{if } b_1 < \min(b_3, b_4) \text{ or } b_2 \geq \max(b_3, b_4) \text{ or } b_4 \leq b_3 \\ [b_3, \infty) & \text{if } (b_2 < \min(b_3, b_4) \text{ and } b_1 \geq \max(b_3, b_4)) \text{ or } b_4 > b_3 \end{cases}$$

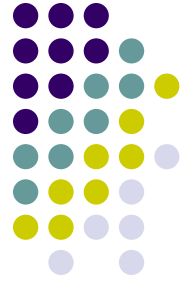
- The simplified confidence interval always has slightly larger probability coverage than the confidence interval shown previously





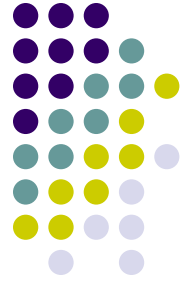
# Point Estimate

- The point estimate can be derived by changing the significance level in the two one-sided confidence intervals so that the upper and the lower confidence intervals intersect on a single point
- The intersection of the two confidence intervals is either  $\frac{z(t_2)}{k}$  or  $\frac{z(t^*) + z(t_2)}{(1 + \sqrt{t^*})k}$



# Practicality Issue

- Disjoint confidence interval
- Two-sided confidence interval
- Statistical property of the estimator

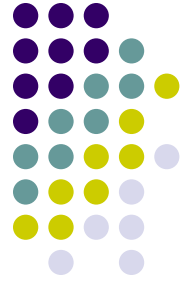


## Confidence Interval (3)

- The naive confidence interval may cause some problem.
- Naïve confidence interval:

$$\left[ \frac{\sqrt{t_1} z_1 + \sqrt{t^* - t_1} z_2 - c_2}{t^* \sqrt{N_0}}, \frac{\sqrt{t_1} z_1 + \sqrt{t^* - t_1} z_2 + c_2}{t^* \sqrt{N_0}} \right]$$

- $z_1, z_2$  are the before-interim and after-interim stage statistics



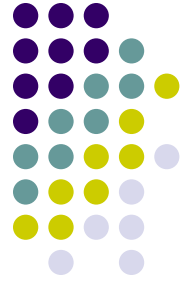
# Example

- Initial planned sample size: 600 subjects (300 per group)
- This can detect an effect size of 0.25 with 86% power
- True treatment effect is zero
- A single interim analysis at 50% information fraction (300 patients) is planned
- If the conditional power is greater than 50% but less than 80% in the interim analysis, the sample size can be increased.



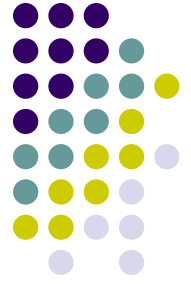
## Example (2)

- The test statistic in the interim analysis is 1.47 and the corresponding conditional power is 57%
- So the sample size is increased to 1090 subjects
- The confidence intervals and treatment effect estimates are computed based on the 1090 subjects



## Example (3)

- The one-sided naive confidence interval is  $[0.003, \infty)$
- The new confidence interval is  $[-0.08, \infty)$
- The naive confidence interval does not include zero



# CHW method

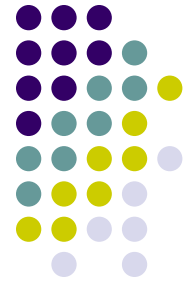
- CHW method: weighted statistics
- The usual point estimate and naïve confidence interval are not valid
- Lawrence and Hung (2003) proposed a point estimate and a confidence interval based on CHW method

# Comparison



- Compare CDL method with CHW method
  - Type I error
  - Power
  - Coverage probability





$\theta_0$	CHW	CDL
<b>0</b>	<b>0.0251</b>	<b>0.0243</b>
<b>0.02</b>	<b>0.0251</b>	<b>0.0243</b>
<b>0.05</b>	<b>0.0250</b>	<b>0.0242</b>
<b>0.1</b>	<b>0.0249</b>	<b>0.0240</b>
<b>0.15</b>	<b>0.0249</b>	<b>0.0240</b>
<b>0.25</b>	<b>0.0251</b>	<b>0.0244</b>

Table 2: Estimated Type I error based on 1,000,000 simulation



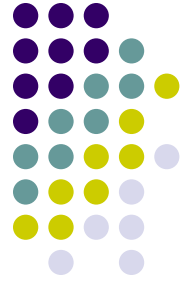
<b>True Treatment Difference</b>	$\theta_0$	<b>CHW</b>	<b>CDL</b>
<b>0</b>	<b>0</b>	<b>0.025</b>	<b>0.024</b>
<b>0.02</b>	<b>0</b>	<b>0.045</b>	<b>0.043</b>
<b>0.05</b>	<b>0</b>	<b>0.095</b>	<b>0.092</b>
<b>0.1</b>	<b>0</b>	<b>0.251</b>	<b>0.246</b>
<b>0.15</b>	<b>0</b>	<b>0.485</b>	<b>0.480</b>
<b>0.25</b>	<b>0</b>	<b>0.884</b>	<b>0.883</b>
<b>0.02</b>	<b>0.02</b>	<b>0.025</b>	<b>0.024</b>
<b>0.05</b>	<b>0.05</b>	<b>0.025</b>	<b>0.024</b>
<b>0.1</b>	<b>0.05</b>	<b>0.094</b>	<b>0.091</b>
<b>0.15</b>	<b>0.05</b>	<b>0.252</b>	<b>0.247</b>
<b>0.25</b>	<b>0.05</b>	<b>0.720</b>	<b>0.718</b>
<b>0.1</b>	<b>0.1</b>	<b>0.025</b>	<b>0.024</b>
<b>0.15</b>	<b>0.15</b>	<b>0.025</b>	<b>0.024</b>
<b>0.25</b>	<b>0.15</b>	<b>0.252</b>	<b>0.247</b>

Table 3: Power based on 1,000,000 simulation



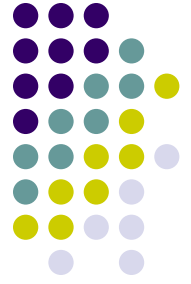
<b>True Treatment Difference</b>	$\theta_0$	<b>CHW</b>	<b>CDL</b>	<b>CDL simplified</b>	<b>Naïve</b>
<b>0</b>	<b>0</b>	<b>0.975</b>	<b>0.976</b>	<b>0.977</b>	<b>0.970</b>
<b>0.02</b>	<b>0</b>	<b>0.975</b>	<b>0.975</b>	<b>0.976</b>	<b>0.970</b>
<b>0.05</b>	<b>0</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.968</b>
<b>0.1</b>	<b>0</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.967</b>
<b>0.15</b>	<b>0</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.969</b>
<b>0.25</b>	<b>0</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.974</b>
<b>0.05</b>	<b>0.05</b>	<b>0.975</b>	<b>0.976</b>	<b>0.977</b>	<b>0.971</b>
<b>0.1</b>	<b>0.05</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.968</b>
<b>0.15</b>	<b>0.05</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.967</b>
<b>0.25</b>	<b>0.05</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.972</b>
<b>0.15</b>	<b>0.15</b>	<b>0.975</b>	<b>0.976</b>	<b>0.977</b>	<b>0.971</b>
<b>0.25</b>	<b>0.15</b>	<b>0.975</b>	<b>0.975</b>	<b>0.975</b>	<b>0.967</b>
<b>0.25</b>	<b>0.25</b>	<b>0.975</b>	<b>0.976</b>	<b>0.977</b>	<b>0.970</b>

Table 4: Coverage Probability of Confidence Interval



# Summary

- CDL method appears to provide a way to increase sample size during the interim analysis and still use the regular test statistic and critical values
- Without a valid confidence interval and point estimate, it is difficult to put the adaptive design into practice
- Statistical validity is not good enough. It has to be sensible in practice



# Reference

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