

# Multiway Tipping Point Analyses in Longitudinal Clinical Trials with Missing Data

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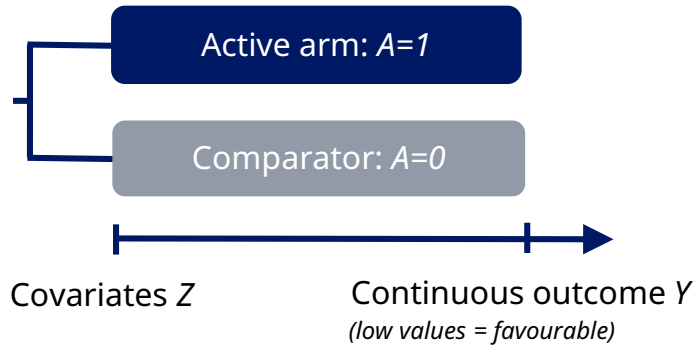
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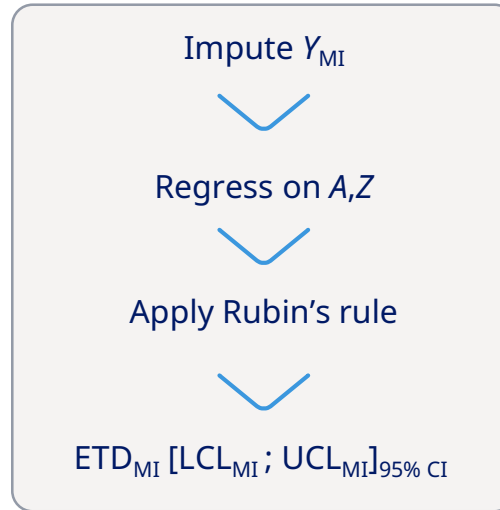
2020 ASA Biopharmaceutical Section  
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# Showing robustness to missing data assumptions

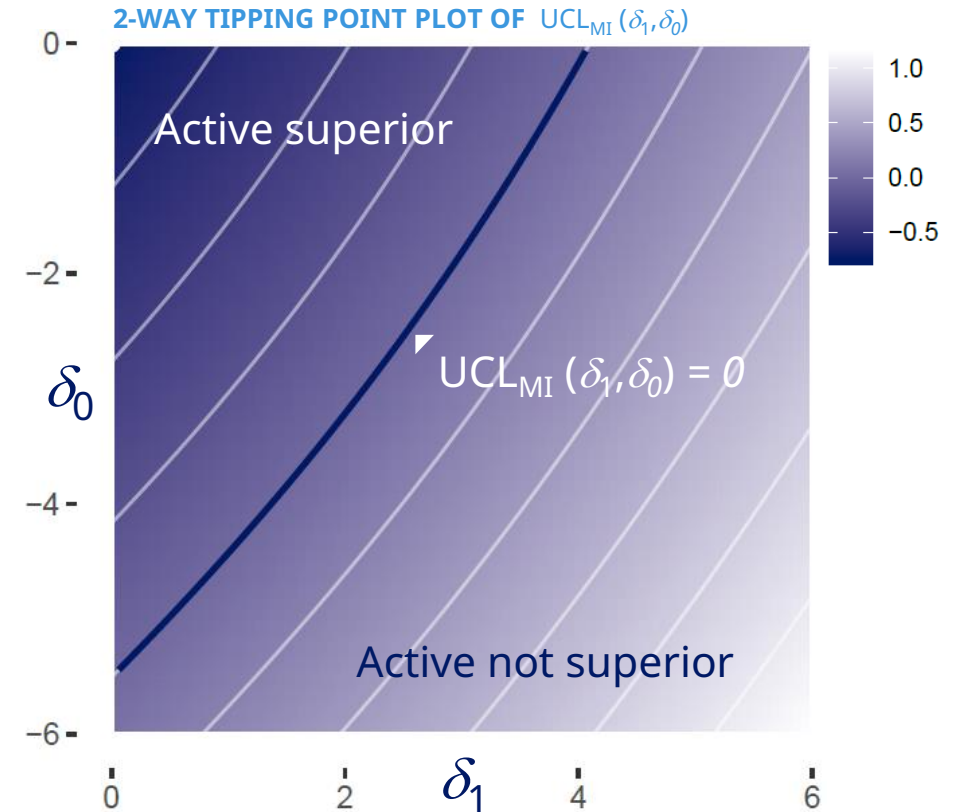
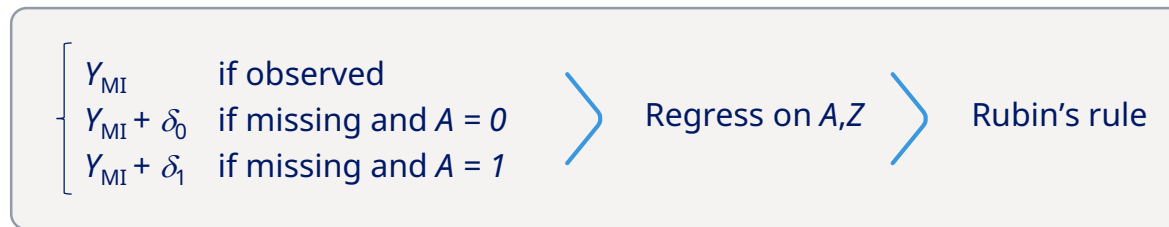
## 2-ARM PARALLEL GROUP RCT



## PRIMARY ANALYSIS (ANCOVA-MI)



## 2-WAY TIPPING POINT SENSITIVITY ANALYSIS



Re-estimating regression models and re-applying Rubin's rule in a 2D grid is a chore! **Can we tip faster?**

With little missing data, the information-to-ink ratio in a 2-way tipping point plot is small. **Can we tip smarter?**

Why stop at arm-specific penalisation – how about going subject-specific? **Can we tip more generally?**

# Can we tip faster?

*Yes! Utilising the simple form of ANCOVA-MI, we can evaluate the full 2-way tipping point surface essentially at the cost of 1 MI analysis + 2 regressions + some sums*

$$\text{UCL}(\delta_1, \delta_0) = \underbrace{\text{ETD}_{\text{MI}}(\delta_1, \delta_0)}_{\text{Linear term in } \delta_1, \delta_0} + 1.96 \cdot \underbrace{V_{\text{MI}}(\delta_1, \delta_0)^{1/2}}_{\text{Quadratic term in } \delta_1, \delta_0}$$

Linear term in  $\delta_1, \delta_0$

> Coefficients from primary ANCOVA-MI and linear regressions of arm-specific missingness indicators on  $A, Z$

Quadratic term in  $\delta_1, \delta_0$

> Coefficients from (co)variance expressions with residuals from previous regressions

## HIGH-LEVEL ALGORITHM FOR FAST 2-WAY TIPPING POINT

- 1** Do primary MI analysis with  $M$  imputations
- 2** Calculate average residual across  $M$  ANCOVAs from **1**
- 3** Regress arm-specific missingness indicators on  $A, Z$  and save residuals
- 4** Pre-calculate  $V_{\text{MI}}$  coefficients
- 5** Calculate  $\text{UCL}(\delta_1, \delta_0)$  in  $D \times D$  grid

# Can we tip smarter?

*Yes! Do the two 1-way tipping point analyses, draw a line between the points. If everything "south-east" of that line is clinically implausible then conclusions are robust to missing data*

$UCL_{MI}(\delta_1, \delta_0)$  is a convex function

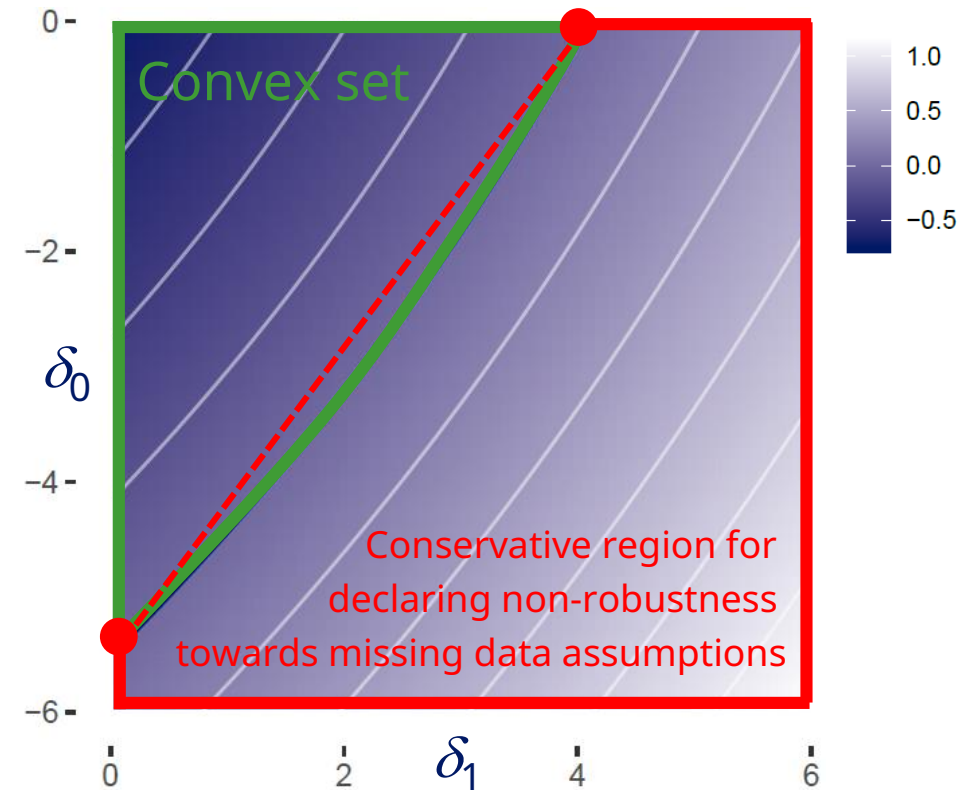
It's the sum of a linear function and the square root of a positive definite quadratic form

So  $\{(\delta_1, \delta_0) : UCL_{MI}(\delta_1, \delta_0) \leq 0\}$  is convex

Convex functions have convex levels sets

So it's conservative to "tip at the line" between the two 1-way tipping points

Lines between points in a convex set stay in that set



# Can we tip more generally?

*Yes! We can study UCL as a function of individual missing values – and investigate the most conservative configuration under (convex) clinically plausible constraints*

$$UCL(\mathbf{w}) = ETD_{MI}(\mathbf{w}) + 1.96 \cdot V_{MI}(\mathbf{w})^{1/2}$$

Consider the upper confidence limit as a real-valued function of the individual, unknown missing values

Maximize  $UCL(\mathbf{w})$  subject to  $\mathbf{w} \in F$

Given convex constraints  $F$  on missing values  $\mathbf{w}$ , what's the globally most conservative configuration possible?



SOLUTION  $\mathbf{w}_0$  FOR  $F = [a; b] \times \dots \times [a; b]$  (BOX-CONSTRAINTS)

$$w_{0i} = \begin{cases} a & \text{if } A_i = 1 \\ b & \text{if } A_i = 0 \end{cases}$$

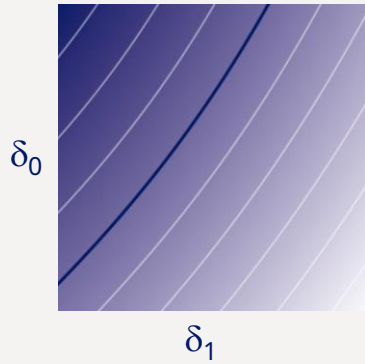
A SUFFICIENT CONDITION FOR ROBUSTNESS

- 1 Select a clinically plausible range  $[a; b]$
- 2 Calculate  $\mathbf{w}_0$  as per above
- 3 If  $UCL(\mathbf{w}_0) < 0$  then no configuration in the clinically plausible range can tip the conclusion

# Conclusion: 3 options to show robustness

*All 3 options comprehensively explore the space of plausible missing data assumptions – but options 2 and 3 are simpler*

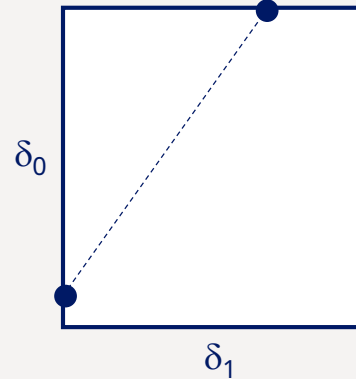
1



*"We calculated and plotted the 2-way tipping point surface and level curve  $UCL=0$  by [...long description ensues...]"*

*Since values below the level curve  $UCL=0$  are considered clinically implausible, conclusions are robust to missing data assumptions"*

2



*"We did two 1-way tipping point analyses and drew a line between results."*

*Since values below that line are considered clinically implausible, conclusions are robust to missing data assumptions"*

3

*"When tipping missing values to the extremes of the clinically plausible range by arm, we still observed  $UCL < 0$ "*

*Hence no configuration in the clinical plausible range will have  $UCL \geq 0$ , and so conclusions are robust to missing data assumptions"*