

Advances in Time-to-Event Analyses in Clinical Trials University of Pennsylvania, 2023

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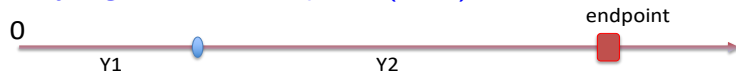
Congratulations to Devan, Lu, Zhenzhen and Fan!

This discussion:

- ▶ will focus on use of multiple endpoints in clinical trials.
- ▶ Comments mainly related to presentations of Devan Mehrotra and Lu Tian.

Lu Tian's work

Analyzing duration of response (DOR)



Y_1 : time to response; Y_2 : DOR

If $Y_1 < T$ then $T = Y_1 + Y_2$. If $Y_1 \geq T$ then $Y_1 = \infty$.

T : time to endpoint

C : censoring time

▶ **Induced informative censoring:**

The censoring time for Y_2 is $\max\{C - Y_1, 0\}$. If Y_1 is correlated with Y_2 then the observation of Y_2 is subject to informative censoring.

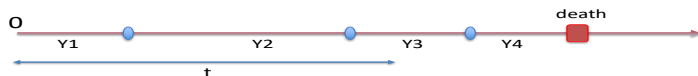
▶ Standard Kaplan-Meier estimation is biased. Lu Tian presented bias-corrected estimation of $S_{Y_2}(t)$ (same as Lin-Ying estimator.).

▶ Huang and Louis (1998, Biometrika): Joint estimation of a survival function and mark variables.

- **Mark variable could be DOR, medical cost, biomarker measurement, etc.**

Devan Mehrotra's work

Analyzing composite or multiple endpoints



Y_j : the j th gap time

$T_j = Y_1 + \dots + Y_j$: time to j th endpoint

t : time since 0

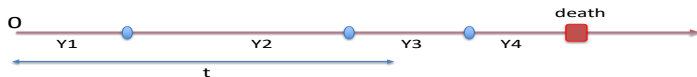
Data example: CV death, Non-fatal stroke, Non-fatal MI, Coronary revasc ≥ 30 days after randomization, Unstable angina

- Prentice, Williams & Peterson model (1981) assume the j th endpoint occurs with the hazard

$$\lambda_j(t - t_{j-1} \mid X, N^H(t^-)) = \lambda_{0j}(t - t_{j-1}) \exp\{\beta_j' X + \phi(N^H(t^-))\}$$

$\lambda_{0j}(\cdot)$: baseline hazard function. $N^H(t^-)$: history of events prior to t

Can PWP Model be used in clinical trials?



Y_j : the j th gap time; $T_j = Y_1 + \dots + Y_j$: time to j th endpoint

- ▶ For the purpose of testing, need simplification:

$$\lambda_j(t - t_{j-1} | X, N^H(t^-)) = \lambda_{0j}(t - t_{j-1}) \exp\{\beta_j' X + \phi(N^H(t^-))\}$$

$$\lambda_j(t - t_{j-1} | X, N^H(t^-)) = \lambda_{0j}(t - t_{j-1}) \exp\{\beta_j' X\}$$

– it requires that **each gap time carries NO memory from the event history!**

- ▶ PWP considered ordered endpoints so λ_j has concrete meaning; e.g., HIV - AIDS -Death
What does λ_j mean if multiple endpoints have arbitrary order? (CV death, Non-fatal stroke, Non-fatal MI, etc.)
- ▶ Competing risks issues: Since 'death' must be part of censoring for other endpoints, what type of hazards is PWP model estimating? - **Cause-specific hazard**
- ▶ If PWP model estimates cause-specific hazards, does the hazard provide proper causal interpretation for clinical trials?

Statistical approaches

- ▶ **Anderson-Gill approach:** The AG model assumes constant intensity ratio; the multiple event process does not carry memory:
$$\lambda(t | N^H(t^-), X) = \lambda_0(t) \exp\{\beta' X\}$$
 - AG model is similar to a Poisson process; 'memoryless' is a very strong assumption!
 - Does not handle absorbing event such as death.
- ▶ **Wei-Lachin, Brown, Kost-McDermott, Stouffer, Lachin-Bebu approaches:** Cox PH model for each component, average resulting HR estimates with differential weights.
 - Conceptually, better than PWP and AG approaches.
 - In the presence of CV-death, the hazard for other multiple events (Non-fatal stroke, Non-fatal MI, etc.) are cause-specific hazards.
 - Are we comfortable with the use of cause-specific hazards to evaluate trial effects?

Statistical approaches

- ▶ **Win ratio statistic:** Aggregate pairwise subject-level between-treatment comparison of survival times based on sequential order of endpoint importance (Pocock et al., 2012).
 - The win ratio statistic is simple in its format but hard to understand.
 - Fortunately, in semi-competing risks setting, the aim of hypothesis testing has been clarified to test on the **observable part of the hazards of latent variables**; see Luo et al. (Biometrics, 2015) and Mao Lu (2021).
 - Analysis results possess causal interpretation!
- ▶ **Claggett et al. approach:** Quantify mean cumulative count of events over time using AUC to compare treatments. AUC is an overall summary measure of all the events.
 - Conceptually, this is an appropriate approach in the presence of multiple endpoints (including death).
 - Analysis results possess causal interpretation!
 - Why worst performance in simulation?