



Decoding optimal methods in treatment switching:
Recommendations from oncology-inspired simulation studies

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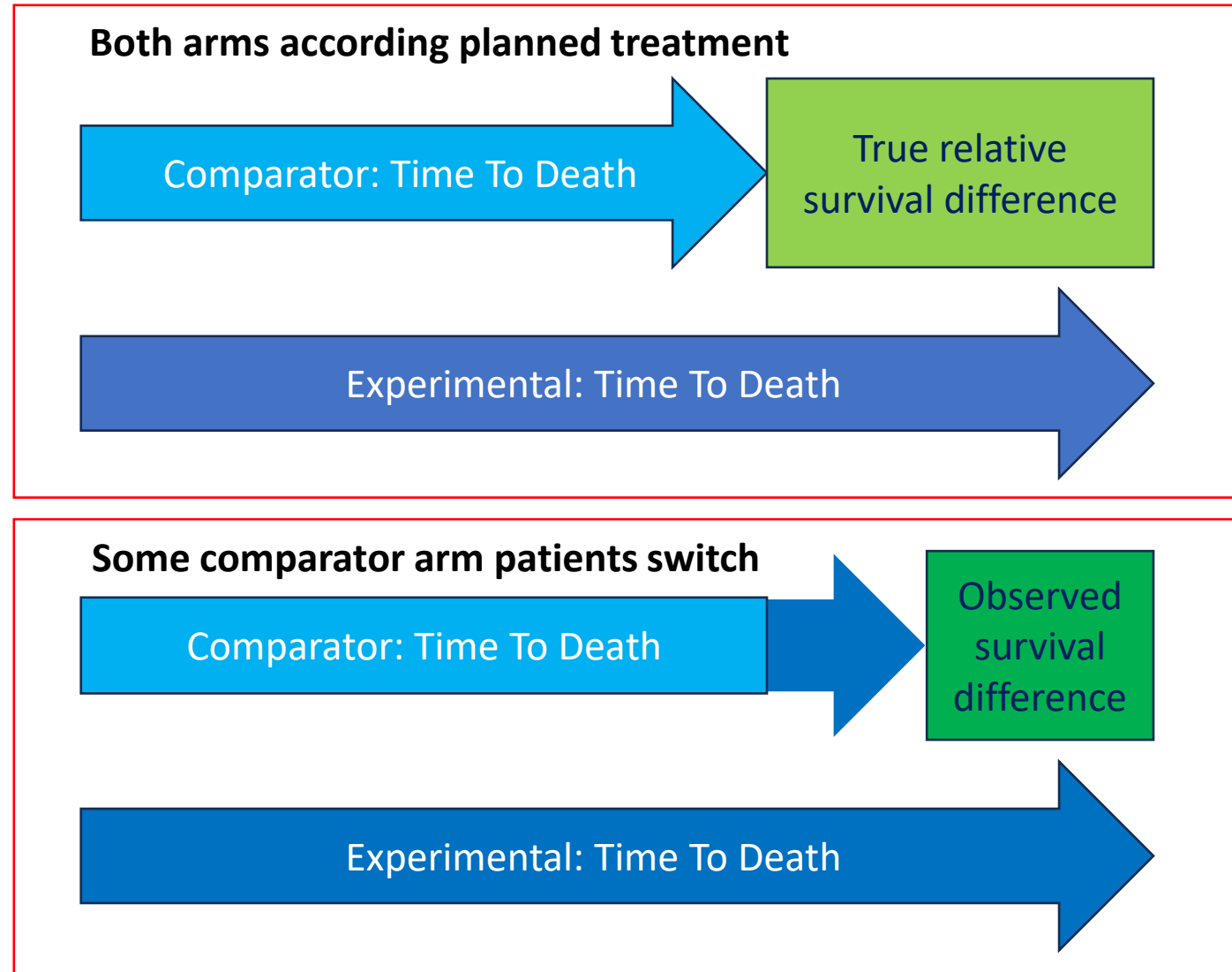
Agenda



- What is treatment switching?
- Why does treatment switching matter?
- The toolbox for treatment switching
 - IPCW
 - RPSFTM
 - TSE
- Factors influencing treatment switching
- Simulation study setups and limitations
- Factors influencing choice of adjustment method
- Conclusions

What is treatment switching?

- Patients receive subsequent therapy during RCT
 - If comparator arm patients switch, then OS is overestimated
 - Comparator OS benefits from partial contribution of more efficacious treatment
 - Survival difference is underestimated
- When to adjust
 - Sufficient numbers switch
 - Sufficient FU time after switch
 - Sufficient TE of subsequent therapy



Why does treatment switching matter?

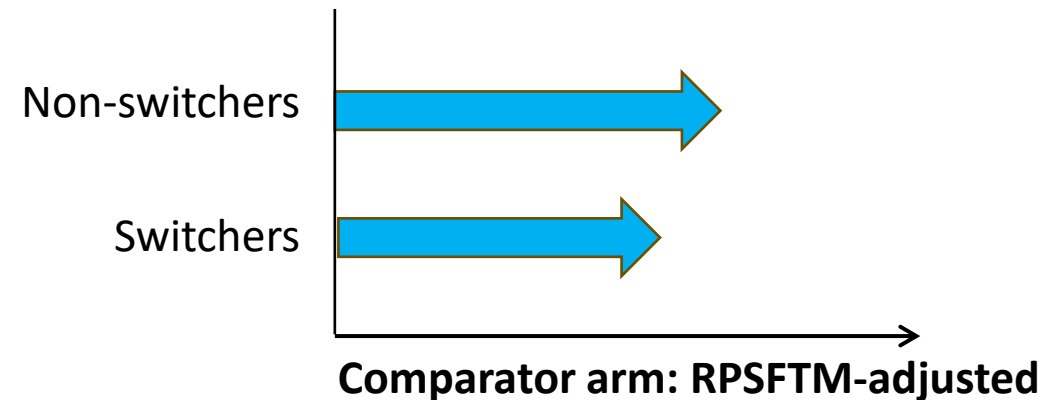
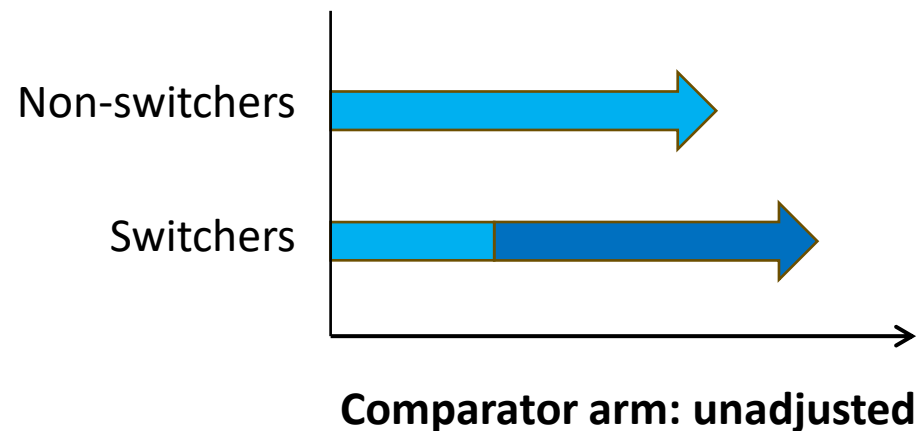
- Regulatory (EU): Conduct sensitivity analyses for impact of cross-over for TTE endpoints (*“Question and answer on adjustment for cross-over in estimating effects in oncology trials”, EMA/845963/2018*)
- HTA (UK): *Section 4.6 in Modelling methods, NICE HTA Manual (31st Jan 2022)*
- HTA: Benefit can be either over- or underestimated: Potential changes in cost-effectiveness assessments

The toolbox for treatment switching: IPCW

- IPCW censors switchers at the time of switch: Informative censoring
- IPCW corrects for this bias by weighting patients
 - Fit model of treatment censoring using baseline and time-varying covariates (no unmeasured confounders)
 - Weight is inverse of probability not being censored until certain time
- If two patients have similar covariates, the one who did not switch would receive higher weights
- Carry out weighted survival analysis, such as Cox PH model or KM estimator

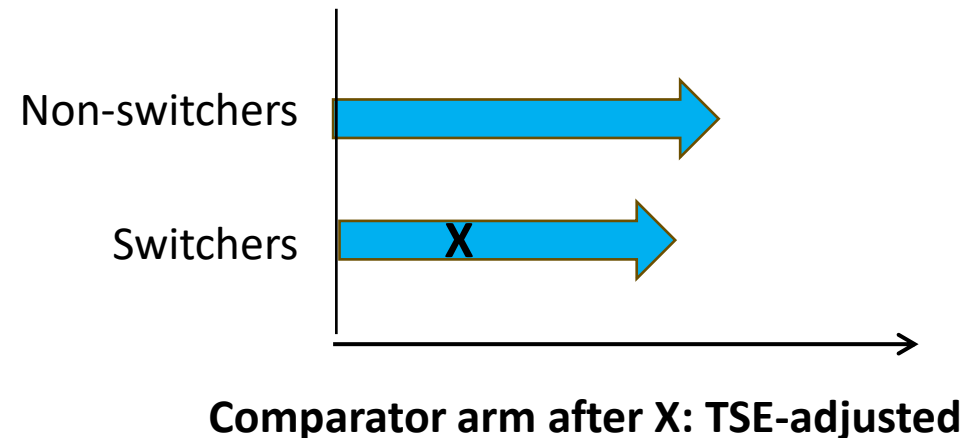
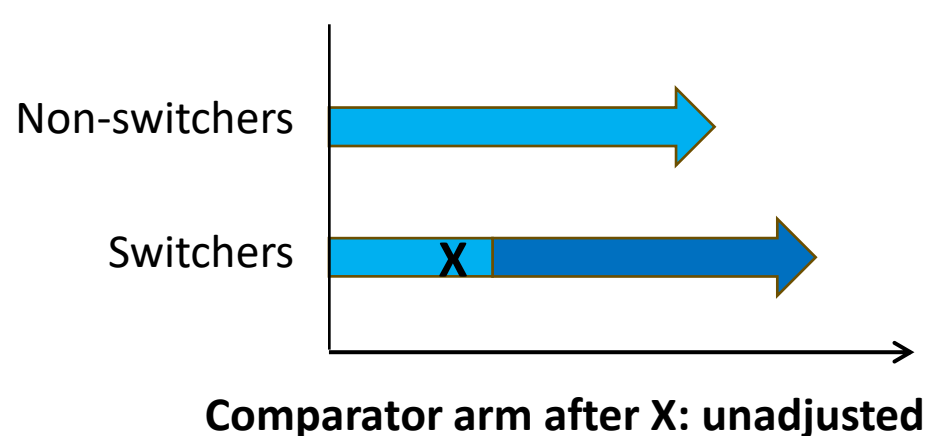
The toolbox for treatment switching: RPSFTM

- RPSFTM adjusts treatment effect that balances counterfactual survival
- Counterfactual survival time would have been observed in absence of treatment
- Assumes constant treatment effect irrespective of start of subsequent therapy
- Adjusts survival times that would have been observed in absence of switching
- Carry out survival analysis with shrunk survival times, such as Cox PH model or KM estimator



The toolbox for treatment switching: TSE

- TSE developed where switching only occurs after secondary baseline (denoted as **X**)
- Typically, this is disease progression (PD) in oncology trials
- Assumes that at secondary baseline, patients have comparable disease stage
- Switching soon after PD and no unmeasured confounders at PD
- Carry out survival analysis with shrunk survival times, such as Cox PH model or KM estimator



Factors influencing treatment switching

- Sample Size
- Switching proportion
- Hazard Ratio
- Censor Rate
- Time-dependent confounding
- Common Treatment Effect

Simulation study setups and limitations

- Joint survival and longitudinal model: Simultaneously generate time-dependent covariate and survival times
- Switching proportions: *Low, Moderate, High*
- Treatment effect: *Low, Medium, High*
- Censoring rate: *None, Low, High*
- Time-dependent confounding: *Yes, No*
- Common Treatment Effect: *Yes, No*
- Limitations: Limited scenarios, type of switching, correct model specification

Factors influencing choice of adjustment method

- Across scenarios RPSFTM has low bias
 - IPCW with even lower bias but only for moderate switching proportions
- Without time-dependent confounding:
 - For moderate and high switching proportions TSE is best
 - For low switching proportions RPSFTM is best
- With time-dependent confounding:
 - For moderate switching proportions all three methods work well
 - For high switching proportions RPSFTM works a bit better than TSE
- No censoring not relevant in clinical trials
- Size of treatment effect (HR) of limited impact

Conclusions

- Adjusting survival times for treatment switching is crucial
- Ground truth not known in clinical trials
- Diverse scenarios are suitable to different approaches
- Trial characteristics can indicate suitable approach

References

- Morden, James P., et al. "Assessing methods for dealing with treatment switching in randomised controlled trials: a simulation study." *BMC medical research methodology* 11 (2011): 1-20.
- Latimer, N. R., et al. "Adjusting for treatment switching in randomised controlled trials—a simulation study and a simplified two-stage method." *Statistical methods in medical research* 26.2 (2017): 724-751.
- Latimer, N. R., et al. "Assessing methods for dealing with treatment switching in clinical trials: a follow-up simulation study." *Statistical methods in medical research* 27.3 (2018): 765-784.
- Latimer, N. R., et al. "Improved two-stage estimation to adjust for treatment switching in randomised trials: g-estimation to address time-dependent confounding." *Statistical methods in medical research* 29.10 (2020): 2900-2918.
- Crowther, Michael J., and Paul C. Lambert. "Simulating biologically plausible complex survival data." *Statistics in medicine* 32.23 (2013): 4118-4134.
- Watkins, Claire, et al. "Further practical guidance on adjusting time-to-event outcomes for treatment switching." *Pharmaceutical statistics* (2025)



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