



# Practical aspects of handling treatment switching in randomized clinical trials

BBS Spring Seminar

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# NVS Oncology Treatment Switching Guideline Team

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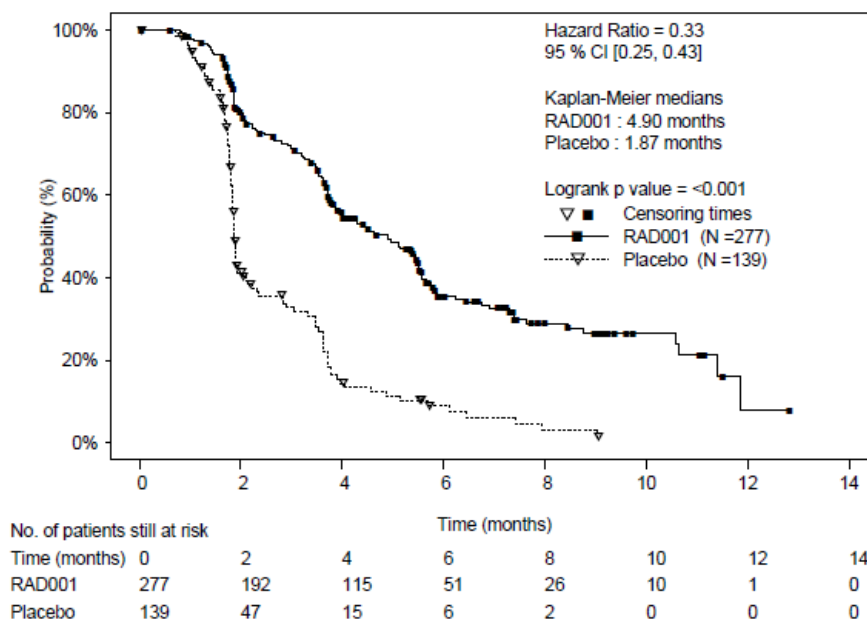
# Outline

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- Motivating example
- Time-dependent confounding
- Rank preserving structured failure time model
- Inverse probability weighting
- General recommendations
- Recent publications and existing software
- Acknowledgements
- References

# Motivating example

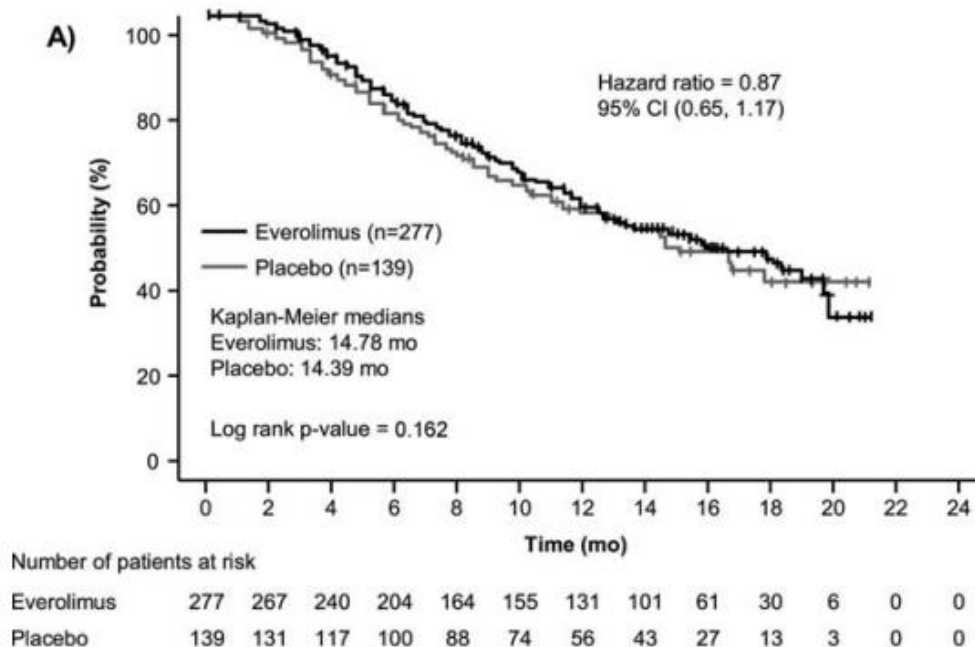
## *Everolimus in mRCC (RECORD-1)*



- Phase III study of everolimus in metastatic renal cell carcinoma Motzer et al (2008, 2010)
  - Double-blind, multicenter study with patients randomized to receive either everolimus (n = 277) or placebo (n = 139)
  - Primary endpoint PFS (HR=0.30, 95% CI 0.22-0.40, p<0.0001)
  - Regulatory approval based on PFS
- Protocol allowed crossover from placebo to everolimus upon progression

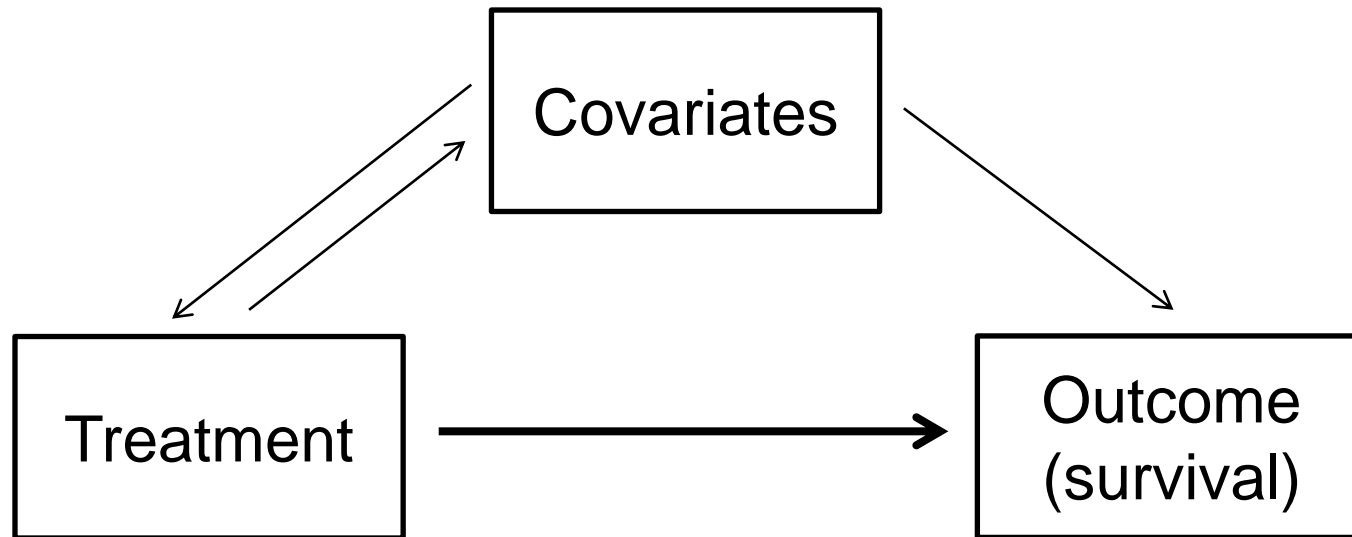
# Motivating example

## *Everolimus in mRCC (RECORD-1)*



- ~80% of placebo randomized patients switched to everolimus
- ITT analysis of OS: HR=0.87, 95% CI: 0.65-1.15, p-value=0.162
- ITT analysis provides a valid assessment of the **treatment policy**
- What about assessment of **treatment effect** of everolimus on OS if placebo patients never received everolimus?

# Time-dependent confounding

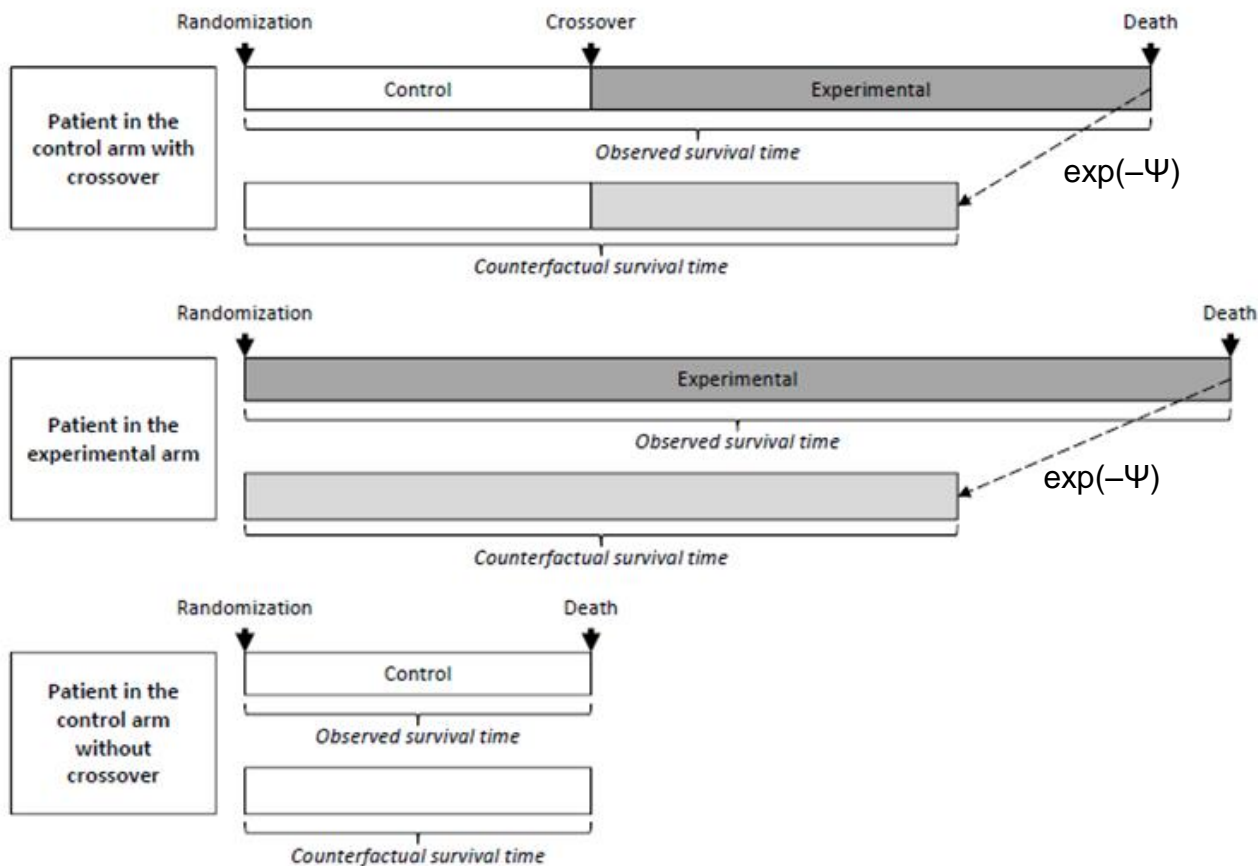


- 1 – Effect of interest
- 2 – Confounding effect on occurrence of the outcome
- 3 – Confounding effect on future exposure to treatment

# Rank preserving structural failure time (RPSFT) model

- Estimate the **survival time gained/lost by receiving active treatment** (i.e. either randomized or “cross-over” active treatment)
- Main assumption: **treatment is acting by multiplying survival time by a given factor once patient starts receiving active treatment** (**transparent but un-testable assumption**)
- Multiplicative factor interpreted as relative increase/decrease in survival if one took active treatment compared to taking control
- It works by reconstructing the survival duration of patients, as if they had never received active treatment

# RPSFT – acceleration factor



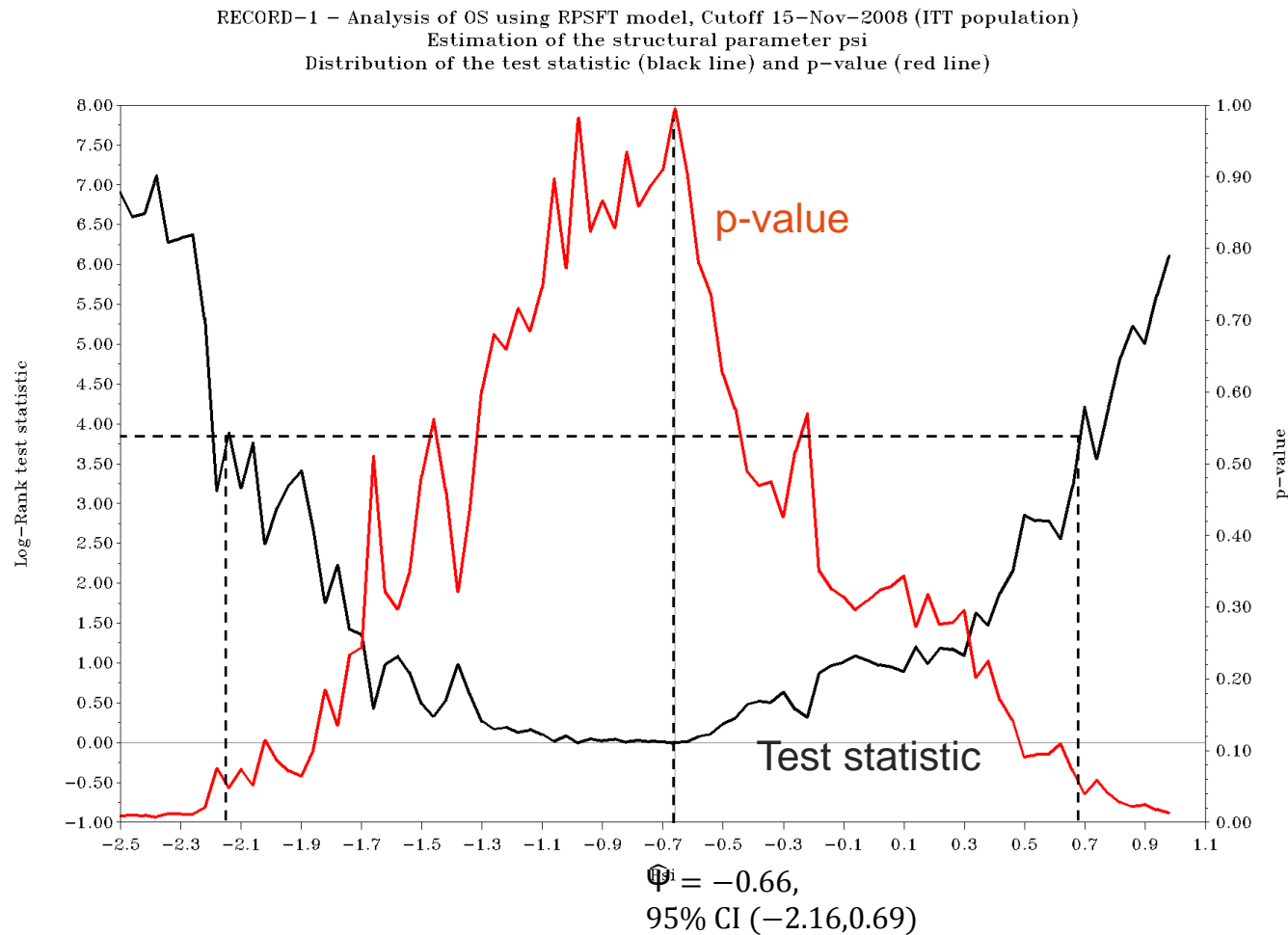
Treatment multiplies life by acceleration factor  $\exp(-\Psi)$

$\Psi < 0 \rightarrow$  time on experimental therapy extends life compared to control

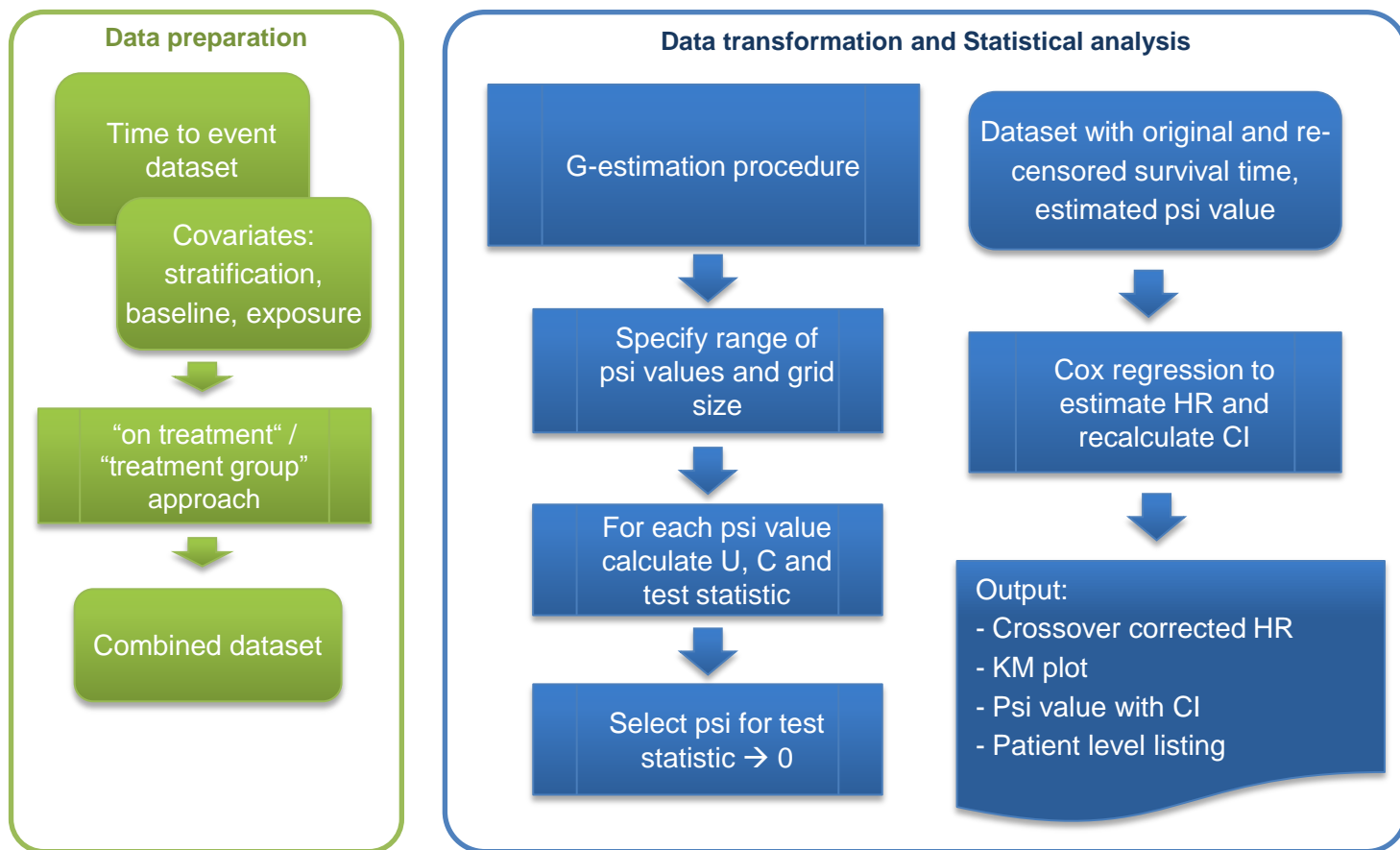
$\Psi > 0 \rightarrow$  time on control therapy extends life compared to experimental



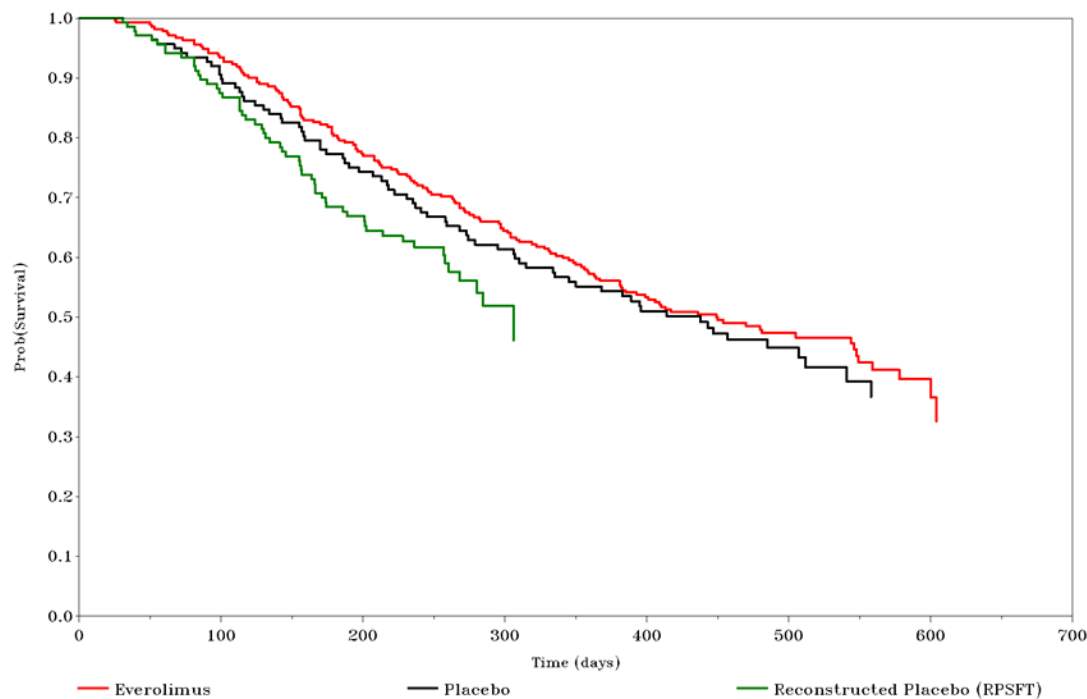
# RPSFT – G-estimation for psi (RECORD-1)



# Programming implementation for RPSFT



# RPSFT model results for RECORD-1



Type of analysis	Events/Censored (Placebo)	Events/Censored (Everolimus)	Hazard Ratio	95% CI Lower Limit	95% CI Higher Limit
ITT analysis	75/64	146/131	0.868	0.655	1.150
RPSFT analysis	57/82	146/131	0.603	0.220	1.651

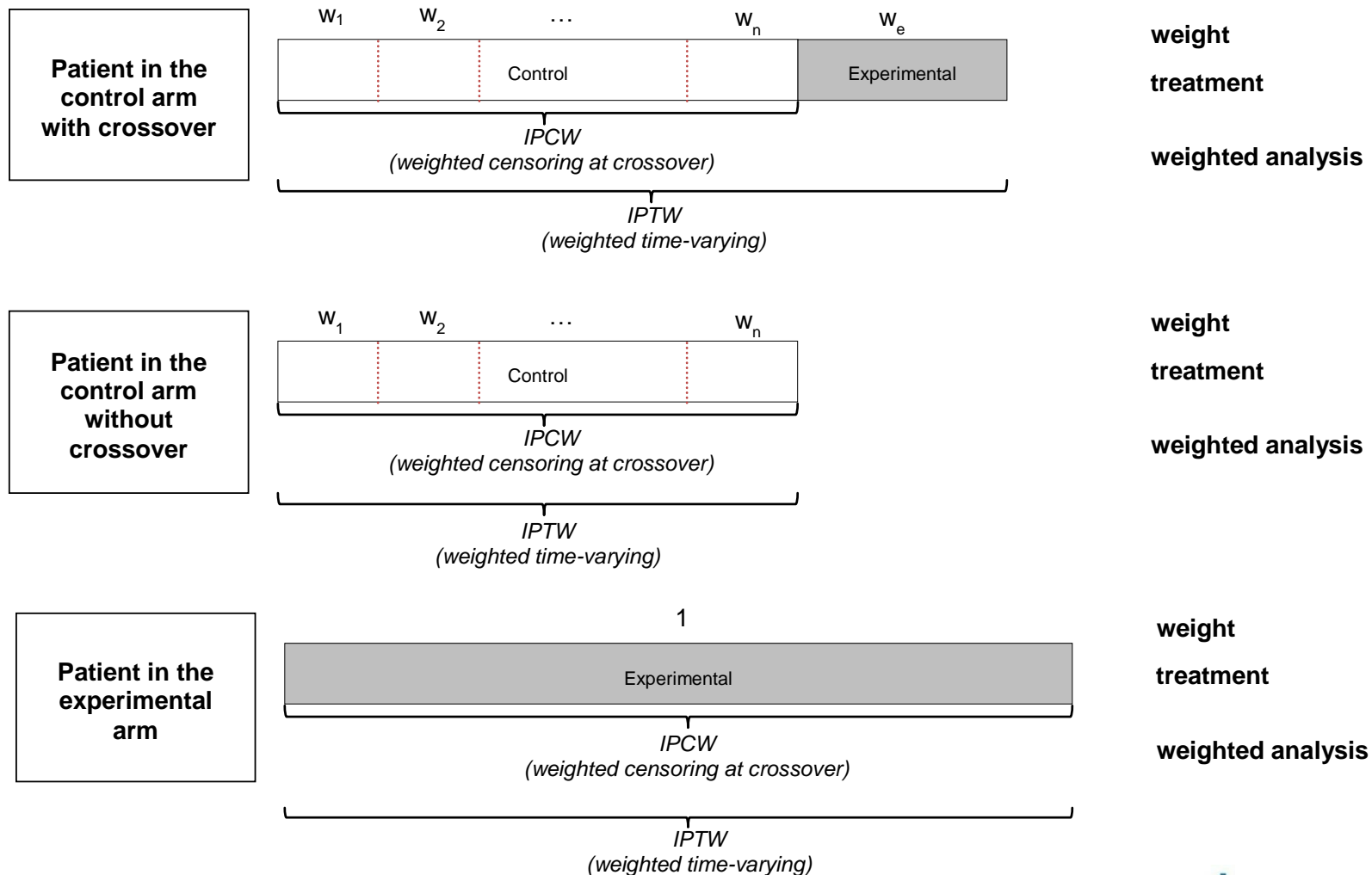
The corrected results used reconstructed survival time for all control arm patients

# Inverse probability weighting (IPW) approach

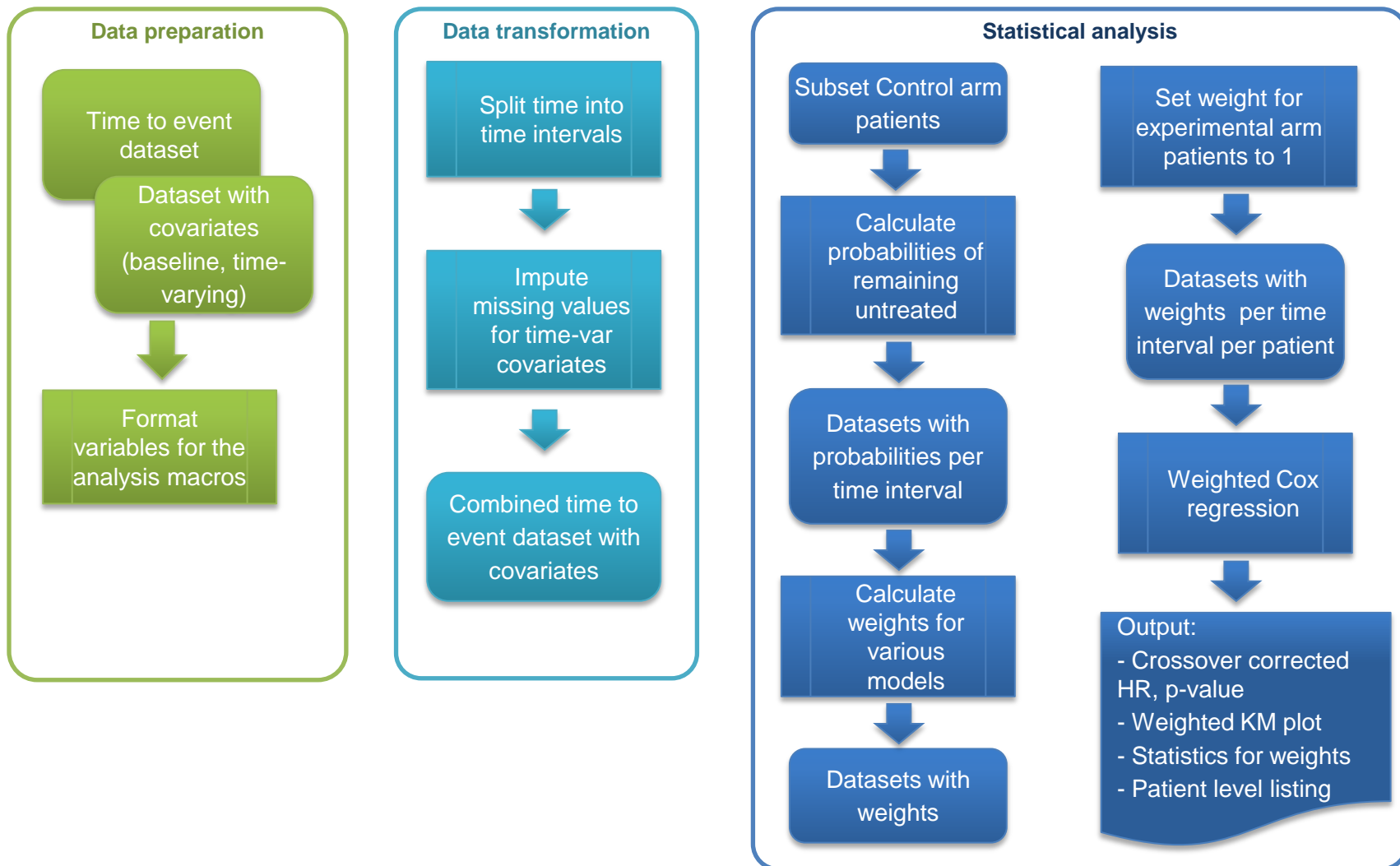
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- Model-based method that reweights control arm patients using propensity score methods.
- Treatment effect is expressed on the hazard ratio scale and estimated using weighted Cox model.
- Main assumption: no unmeasured confounders (all factors influencing crossover and survival are included in the model), non-testable

# Inverse probability weighted analysis (IPW)



# Programming implementation for IPW



# IPCW application to RECORD-1

**Table 1. Variables Included in All Cox Regressions Models Considered**

Description	Model														
	1	2	3	4	5	6	7	8	9	10	11	12	13	14	15
Age at baseline (y)	✓		✓	✓		✓								✓	✓
Country			✓		✓										
Sex	✓		✓	✓			✓							✓	✓
Race			✓	✓				✓						✓	✓
MSKCC prognostic score at baseline	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
KPS at baseline															✓
Prior treatment with sorafenib only	✓		✓	✓							✓	✓	✓		✓
Prior treatment with sunitinib only	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
Prior radiation treatment			✓	✓					✓				✓		✓
Prior nephrectomy			✓	✓						✓			✓		
Time since diagnosis			✓	✓											✓
Liver involvement		✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	
Bone involvement		✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	
Randomized treatment	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
Time period	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓	✓
HR	0.54	0.49	0.45	0.47	0.50	0.49	0.53	0.49	0.50	0.51	0.44	0.49	0.51	0.52	0.49
HR 95% CI	0.30, 1.01	0.26, 0.92	0.24, 0.84	0.27, 0.82	0.27, 0.94	0.26, 0.91	0.28, 1.00	0.26, 0.91	0.27, 0.93	0.27, 0.96	0.26, 0.76	0.26, 0.92	0.27, 0.96	0.28, 0.98	0.29, 0.83

CI=confidence interval; HR=hazard ratio; KPS=Karnofsky performance status; MSKCC=Memorial Sloan-Kettering Cancer Center

**Model 4: best model fit (AIC)**

# General recommendations

## *Points to consider*

- Describe the treatment switching mechanism
  - When was switching permitted? Patient-level or study-level
  - Why did switching occur? Post-progression switch, switch based on study milestone
- Quantify the extent of switching and characterize the population of switchers
  - Number and percentage of patients who switched, proportion of total exposure and/or follow up time that was affected by switch
  - Timing of treatment switching in a form of Kaplan-Meier plot estimated median, range
  - Baseline characteristics for switchers
- Select method to account for treatment switching, implementation and description of results
  - Consider feasibility of the underlying assumptions for each method
  - Identify potential sources of bias and describe them for the method selected
  - Ideally, provide a document with the details of the model(s)



# Recent publications and software

## Adjusting overall survival for treatment switch Recommendations of a cross-institutional statistical working group

Claire Watkins

BBS/EFSPi European Scientific Meeting  
Application of Methods for Health Technology Assessment  
23<sup>rd</sup> June 2015



### NICE DSU TECHNICAL SUPPORT DOCUMENT 16: ADJUSTING SURVIVAL TIME ESTIMATES IN THE PRESENCE OF TREATMENT SWITCHING

REPORT BY THE DECISION SUPPORT UNIT

July 2014

Nicholas R. Latimer<sup>1</sup>  
Keith R. Abrams<sup>2</sup>

Harvard School of Public Health – Software for MSM, SNM, etc (SAS)

<http://www.hsph.harvard.edu/causal/software/>

Causal inference book by Robins and Hernan – Parts of the book and the code available (in draft)

<http://www.hsph.harvard.edu/miguel-hernan/causal-inference-book/>

Medical Research Council – Software for RPSFT (STATA)

<http://www.mrc-bsu.cam.ac.uk/software/stata-software/>

# Acknowledgments

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- Mike Branson
- Bee Chen
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# References

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- Korhonen P, Zuber E, Branson M et al. *Correcting overall survival for the impact of crossover via a rank-preserving structural failure time (RPSFT) model in the RECORD-1 trial of everolimus in metastatic renal-cell carcinoma.* J Biopharm Stat 2012
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- Motzer, R., Escudier, B., Oudard, S., et al. (2010). Phase 3 trial of everolimus for metastatic renal cell carcinoma: Final results and analysis of prognostic factors. Cancer 116:4256– 4265.

# Back up

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# RPSFT – Artificial censoring algorithm

- An additional algorithm ('artificial-censoring') allows to **maintain the assumption of independent random censoring required for unbiased estimation**
- The artificial censoring algorithm works by shrinking the total follow-up time (time between randomization to analysis cut-off date) for all patients regardless of randomization group or treatment received
- Therefore every patient censored in the ITT analysis remains censored with duration equal or shorter to the original one; in addition, patients with an event in the original analysis may become censored via the artificial-censoring algorithm

