

Estimation of Treatment Effects Under Non-Compliances

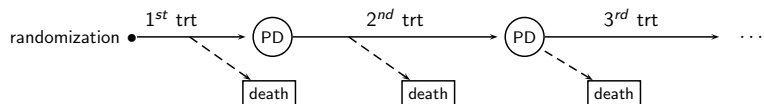
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in cooperation with Boehringer Ingelheim Pharma GmbH

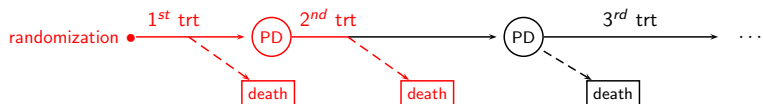
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- 2 Combination of Initial Models for PFS
- 3 Combination of Initial Models for PFS and OS
- 4 Numerical Examples

- Setting:
 - oncological trials
 - randomized
 - two-arm studies, (placebo-)controlled



PD: progressive disease

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 - oncological trials
 - randomized
 - two-arm studies, (placebo-)controlled



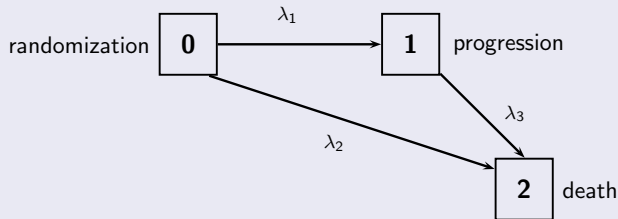
PD: progressive disease

- Setting:
 - oncological trials
 - randomized
 - two-arm studies, (placebo-)controlled
- Endpoints:
 - **Progression-free Survival (PFS):**
↪ *time from randomization to (first) progression or death from any cause, whatever comes first*
 - **Overall Survival (OS):**
↪ *time from randomization to death*

Issue:

- controlled trial design commonly only until progression, thereafter control group might receive (similar) treatment
 - non-ignorable non-compliances (e.g. because of severe side effects)
- ▶ this can lead to confounded estimates of treatment effect

Starting Point: Three-State Model



- assumption of constant transition rates between states gives exponentially distributed transition times
- PFS time is $T_{PFS} := T_0 \sim \text{Exp}(\lambda_1 + \lambda_2)$
- distribution function of OS with $t \in \mathbb{R}_+$

$$F_{T_{OS}}(t) = 1 - \frac{\lambda_1}{\lambda_1 + \lambda_2 - \lambda_3} e^{-\lambda_3 t} + \frac{\lambda_3 - \lambda_2}{\lambda_1 + \lambda_2 - \lambda_3} e^{-(\lambda_1 + \lambda_2)t}$$

Rank Preserving Structural Failure Time Model

The observed lifetime T_i of subject i may be linked to his/her latent baseline survival time U_i via the RPSFTM [Robins J M, Tsiatis A A (1991)]

$$U_i = \int_0^{T_i} e^{\beta_0 d_i(t)} dt, \quad (1)$$

with β_0 the causal effect and $d_i(t)$ a treatment indicator at time t .

- RPSFTMs are models for counterfactuals, since they presume that individuals can theoretically be observed under different conditions

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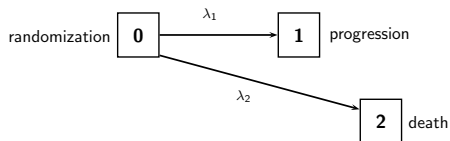
- RPSFTMs are models for counterfactuals, since they presume that individuals can theoretically be observed under different conditions
- Full compliance in treatment arm is given by $d_i(t) = 1$ for all t

$$U_i = e^{\beta_0} T_i$$

- e^{β_0} fraction of increase or decrease in survival time because of treatment

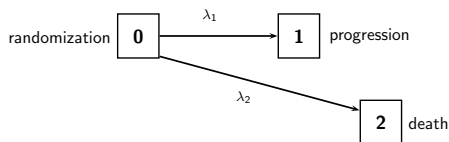
Combination of Initial Models for PFS

- restriction on model for PFS



Combination of Initial Models for PFS

- restriction on model for PFS
- assumption of full compliance in treatment arm



- in control arm:

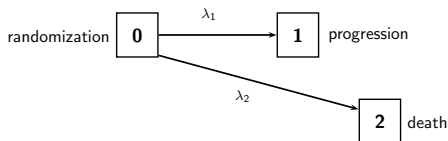
$$U = T_{PFS} \sim \text{Exp}(\lambda_1^C + \lambda_2^C)$$

- in treatment arm:

$$U = e^{\beta_0} \cdot T_{PFS} \quad \text{with} \quad T_{PFS} \sim \text{Exp}(\lambda_1^T + \lambda_2^T)$$

Combination of Initial Models for PFS

- restriction on model for PFS
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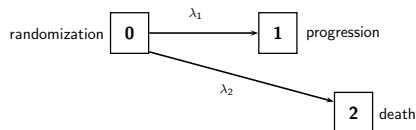
- in treatment arm:

$$U = e^{\beta_0} \cdot T_{PFS} \quad \text{with} \quad T_{PFS} \sim \text{Exp}(\lambda_1^T + \lambda_2^T)$$

⇒ causal parameter of the RPSFTM is $e^{\beta_0} = \frac{\lambda_1^T + \lambda_2^T}{\lambda_1^C + \lambda_2^C}$.

Combination of Initial Models for PFS

- assumption of partial compliance in treatment arm with time on treatment given by $\tau = p \cdot T_{PFS}$, $p \in [0, 1]$



- in control arm:

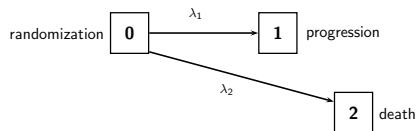
$$U = T_{PFS} \sim \text{Exp}(\lambda_1^C + \lambda_2^C)$$

- in treatment arm:

$$U = \int_0^\tau e^{\beta_0} dt + \int_\tau^{T_{PFS}} dt \quad \Rightarrow \quad T_{PFS} = U \cdot (1 + p(e^{\beta_0} - 1))^{-1}$$

Combination of Initial Models for PFS

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- in control arm:

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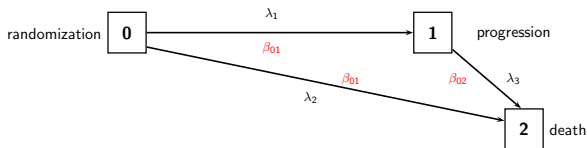
$$U = \int_0^\tau e^{\beta_0} dt + \int_\tau^{T_{PFS}} dt \Rightarrow T_{PFS} = U \cdot (1 + p(e^{\beta_0} - 1))^{-1}$$

⇒ distribution of PFS under partial compliance is

$$T_{PFS} \sim \text{Exp} \left((1 - p) \cdot (\lambda_1^C + \lambda_2^C) + p \cdot (\lambda_1^T + \lambda_2^T) \right)$$

Combination of Initial Models for PFS and OS

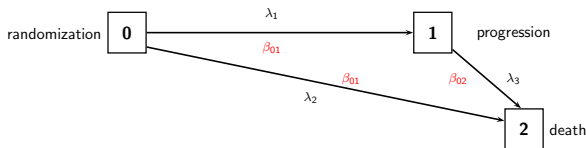
- assumption of full compliance in treatment arm
- 2nd line treatment after progression



- RPSFTM becomes $U = e^{\beta_{01}} \cdot T_{PFS} + e^{\beta_{02}} \cdot T_1$

Combination of Initial Models for PFS and OS

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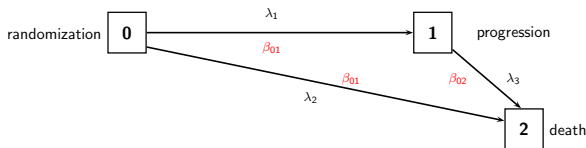


- RPSFTM becomes $U = e^{\beta_{01}} \cdot T_{PFS} + e^{\beta_{02}} \cdot T_1$
- assumption of identical progression rates in study arms

$$\frac{\lambda_1^T}{\lambda_1^T + \lambda_2^T} = \frac{\lambda_1^C}{\lambda_1^C + \lambda_2^C} \quad \text{and} \quad \frac{\lambda_2^T}{\lambda_1^T + \lambda_2^T} = \frac{\lambda_2^C}{\lambda_1^C + \lambda_2^C}$$

Combination of Initial Models for PFS and OS

- assumption of full compliance in treatment arm
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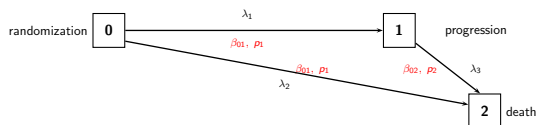
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⇒ causal parameters of the RPSFTM are

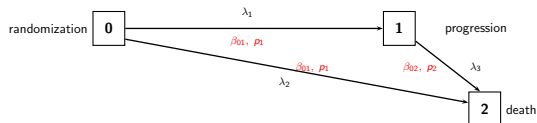
$$e^{\beta_{01}} = \frac{\lambda_1^T + \lambda_2^T}{\lambda_1^C + \lambda_2^C}, \quad e^{\beta_{02}} = \frac{\lambda_3^T}{\lambda_3^C}$$

Combination of Initial Models for PFS and OS



- assumption of partial compliance in treatment arm with
$$\tau_1 = p_1 \cdot T_{PFS} \quad \text{time on treatment until progression}$$
$$\tau_2 = p_2 \cdot T_1 \quad \text{time on treatment between progression and death}$$
- RPSFTM is
$$U = \int_0^{\tau_1} e^{\beta_{01}} dt + \int_{\tau_1}^{T_0} dt + \int_0^{\tau_2} e^{\beta_{02}} dt + \int_{\tau_2}^{T_1} dt$$

Combination of Initial Models for PFS and OS



- assumption of partial compliance in treatment arm with
 $\tau_1 = p_1 \cdot T_{PFS}$ time on treatment until progression
 $\tau_2 = p_2 \cdot T_1$ time on treatment between progression and death
- RPSFTM is $U = \int_0^{\tau_1} e^{\beta_{01}} dt + \int_{\tau_1}^{T_0} dt + \int_0^{\tau_2} e^{\beta_{02}} dt + \int_{\tau_2}^{T_1} dt$
- distribution of OS under partial compliance is given by

$$F_{T_{OS}}(t) = 1 - \frac{\lambda_1^P}{\lambda_1^P + \lambda_2^P - \lambda_3^P} e^{-\lambda_3^P t} + \frac{\lambda_3^P - \lambda_2^P}{\lambda_1^P + \lambda_2^P - \lambda_3^P} e^{-(\lambda_1^P + \lambda_2^P)t}$$

with $\lambda_1^P = (1 - p_1)\lambda_1^C + p_1\lambda_1^T$, $\lambda_2^P = (1 - p_1)\lambda_2^C + p_1\lambda_2^T$, $\lambda_3^P = (1 - p_2)\lambda_3^C + p_2\lambda_3^T$

- **G-Estimation**

- grid search over sequence of clinically relevant values β
- test for equality of baseline survival time distributions via log-rank statistic $G(\beta)$
- estimate is given by $\hat{\beta} = \operatorname{argmin}_{\beta} |G(\beta)|$

- for n -dimensional setting, with $n \geq 2$:

- grid search
- Nelder-Mead algorithm

Estimation for PFS alone:

True effect β_0	p	$\hat{\beta}$	MSE
$\log(1) = 0$	100%	0.0055	0.0039
	75%	0.0039	0.0071
	50%	-0.0027	0.0163
$\log(0.5) = -0.6931$	100%	-0.6862	0.0040
	75%	-0.6936	0.0612
	50%	-0.6957	0.2034
$\log(1/3) = -1.0986$	100%	-1.0949	0.0041
	75%	-1.0927	0.1757
	50%	-1.1105	0.5629

G-Estimation:

- N=1000 patients, 1 : 1 randomization
- 1000 runs
- grid search from -2 to 1, step size 0.01

- estimate $\hat{\beta}$ approximates true value β_0 very well
- absolute difference of estimate and true value less than 0.01 (stepsize of the grid search)
- estimation unaffected by non-compliances
- larger mean squared errors of $\hat{\beta}$ with rising non-compliances

Estimation for PFS and OS:

True Effect		Compliance Rate		Estimated Effect		Mean squared error	
β_{01}	β_{02}	p_1	p_2	$\hat{\beta}_{01}$	$\hat{\beta}_{02}$	$MSE(\hat{\beta}_{01})$	$MSE(\hat{\beta}_{02})$
log(0.5)	log(1)	100%	100%	-0.6829	-0.0168	0.0246	0.0161
		100%	75%	-0.6842	-0.0325	0.0224	0.0271
		75%	100%	-0.6742	-0.0211	0.0506	0.0141
		75%	75%	-0.6817	-0.0292	0.0502	0.0226
log(0.5)	log(0.5)	100%	100%	-0.6948	-0.7060	0.0266	0.5137
		100%	75%	-0.6909	-0.7233	0.0252	0.5652
		75%	100%	-0.6972	-0.7093	0.0769	0.5187
		75%	75%	-0.7101	-0.7174	0.0726	0.5551

G-Estimation:

- N=1000 patients, 1 : 1 randomization
- 1000 runs
- grid search from -2 to 1 for β_{01} and from $\beta_{01} - 1$ to $\beta_{01} + 1$ for β_{02} ,
- step size 0.1

- not as precise estimates as in one-parameter setting for PFS
- estimates of $\hat{\beta}_{02}$ are not as precise as those of $\hat{\beta}_{01}$
- Nelder-Mead algorithm saves computing time but gives less precise estimates than the grid search

- ▶ data from a non-small-cell lung cancer trial:

Trial arm	Number of patients	Number of progressions	Median PFS time [days]	Number of deaths	Median OS time [days]
Placebo	137	136	29	84	302
Treatment	263	248	85	163	298
Total	400	384	58	247	300

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- ▶ hazard ratio for placebo vs. treatment arm
 - via Cox-model
 - for OS: 1.043 (0.801, 1.358)
 - for PFS: 0.355 (0.285, 0.442)
 - via one-parameter RPSFTM
 - for OS: 1.051 (0.7866, 1.448)
 - for PFS: 0.395 (0.320, 0.472)
- ▶ so far used approaches failed to explain the different effect of the treatment on PFS and OS



- ▶ Trial Design:
 - until progression: placebo or test treatment
 - after progression: subsequent therapy (of same substance class as test treatment) in both study arms
- ▶ Modelling Approach:
two-parameter RPSFTM with varying compliance rates
 - until progression: assumption of full compliance to test treatment in treatment group
 - after progression: compliance to subsequent therapy in both treatment arms

Assumed compliance p_2		Hazard Ratio for	
treatment arm	control arm	PFS	OS
0.4	0.6	0.517	0.357
0.3	0.7	0.387	0.449
0.2	0.8	0.262	0.468
0.6	0.7	0.538	0.310

- ▶ Combination of three-state model with RPSFTM enables to adjust for confounding effects due to
 - subsequent treatment after progression
 - crossin / crossover
 - non-compliances in treatment arm
- ▶ Estimation of separate treatment effects before and after progression is possible

- ▶ **Outlook**
 - a model for random fraction of non-compliances
 - suspending the assumption of identical progression rates
 - extension to a four-state model with a transient state 'responded' (e.g. for hematologic indications)

For Further Reading

-  Fleischer F, Gaschler-Markefski B, Bluhmki E. 2009.
A statistical model for the dependence between progression-free survival and overall survival. *Statistics in Medicine* 28 (21): 2669-2686
-  Robins J M, Tsiatis A A. 1991.
Correcting for non-compliance in randomized trials using rank preserving structural failure time models. *Communications in Statistics*; **20**: 2609-2631.
-  Robins J M, Tsiatis A A. 1992.
Semiparametric estimation of an accelerated failure time model with time-dependent covariates. *Biometrika*; **79** (2): 311-319.
-  Korhonen P A, Laird N M, Palmgren J. 1999.
Correcting for non-compliance in randomized trials: an application to the ATBC study. *Statistics in Medicine* **18**, 2879–2897