Supplementary Materials for "Time-to-event analysis when the event is defined on a finite time interval"

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Introduction

We present additional results that are beyond what could be presented in the main manuscript. This document is organized as follows:

Appendix A: Description of the CIBMTR HCT data referenced in Section 2 of the main manuscript

Appendix B: Components of marginal log-likelihood function referenced in Section 4.2 of the main manuscript

Appendix C: Expression for joint density of (T_1, T_2) referenced in Section 3.4 of the main manuscript

Appendix D: Equivalence between standard transition probability expressions from the multi-state model literature and absolute risk profile contributions calculated via integration over the joint density, referenced in Section 3.5 of the main manuscript

Appendix E: Simulation results corresponding to the 'baseline' scenario enumerated in Section 5.1 and additional data scenarios described in Section 5.2 of the main manuscript

- E.1 baseline scenario
- **E.2** reducing sample size from n=5,000 to n=1,000
- **E.3** administrative censoring via a censoring variable simulated from an Exponential distribution with mean 100
- E.4 administrative censoring solely of the terminal event at 365 days
- **E.5** increasing the non-susceptibility fraction from 46% to 76%
- **E.6** modification of α_{23} and κ_{23} so that the hazard functions for the truncated and standard Weibull distributions are comparable on the interval $(0, \tau)$
- E.7 baseline scenario without frailty term, i.e. Gamma frailty variance equal to zero

Appendix F: Additional results from the analysis of stem cell transplantation data in Section 6 of the main manuscript

Appendix G: Details regarding the included R code

A Description of CIBMTR HCT data

Table A.1: Description of baseline characteristics of study population comprised of patients who underwent first HCT between 1999-2011 and distribution of observed outcomes (acute GHVD and/or death) by the end of 365 days post-transplantation.

			(Observed outcom	e category, %	
			Both	Acute GVHD	Death	Censored
			acute GVHD	& censored	without	for both
	N	%	& death	for death	acute GVHD	
Total subjects	9651		11.0	6.7	28.7	53.7
Gender						
Male	5366	55.6	11.7	6.9	28.3	53.1
Female	4285	44.4	10.1	6.4	29.2	54.4
Age, years						
<10	653	6.8	5.7	7.0	21.6	65.7
10-19	1162	12.0	8.5	8.3	21.8	61.4
20-29	1572	16.3	11.3	8.1	25.9	54.7
30-39	1581	16.4	11.6	6.3	26.5	55.6
40-49	2095	21.7	11.4	6.8	29.6	52.2
50-59	2008	20.8	12.6	5.6	35.2	46.5
60+	580	6.0	11.9	3.1	38.1	46.9
Disease type						
AML	4919	51.0	10.1	6.1	31.8	52.0
ALL	2071	21.5	10.4	6.6	29.3	53.7
CML	1525	15.8	12.9	8.3	18.0	60.9
MDS	1136	11.8	13.1	7.1	28.5	51.2
Disease status						
Early	4873	50.5	9.0	7.3	21.1	62.5
Intermediate	2316	24.0	11.2	7.0	28.4	53.5
Advanced	2462	25.5	14.6	5.1	43.9	36.4
HLA compatibility						
Identical sibling	3941	40.8	7.9	5.7	26.0	60.4
8/8	4100	42.5	11.8	7.3	29.7	51.1
7/8	1610	16.7	16.3	7.5	32.7	43.5

B Likelihood contributions and score equations

B.1 Marginal likelihood contributions

Let
$$\eta_{\omega} = X_{\omega}\beta_{\omega}$$
 and $\Lambda_{\omega,0}(y) = \int_{0}^{y} \lambda_{\omega,0}(s) ds$, for $\omega \in \{14, 24, 23, 34\}$.

$$f_{1}(y_{1}, y_{2}; X) = \int_{0}^{\infty} f_{1}(y_{1}, y_{2}; X, \gamma) f_{\gamma}(\gamma) d\gamma$$

$$= \int_{0}^{\infty} [(1 - \pi) \cdot \gamma \cdot \lambda_{23,0}(y_{1})e^{\eta_{23}} \cdot \exp\{-\gamma (\Lambda_{23,0}(y_{1})e^{\eta_{23}} + \Lambda_{24,0}(y_{1})e^{\eta_{24}})\} \cdot$$

$$\gamma \cdot \lambda_{34,0}(y_{2} - y_{1})e^{\eta_{34}} \cdot \exp\{-\gamma \Lambda_{34,0}(y_{2} - y_{1})e^{\eta_{34}}\} \cdot$$

$$[\Gamma(1/\theta)]^{-1} \theta^{-\frac{1}{\theta}} \gamma^{\frac{1}{\theta} - 1} \exp\{-\gamma/\theta\}] d\gamma$$

$$= \frac{(1 - \pi) \cdot \lambda_{23,0}(y_{1})e^{\eta_{23}} \cdot \lambda_{34,0}(y_{2} - y_{1})e^{\eta_{34}} \cdot (1 + \theta)}{[1 + \theta\{\Lambda_{23,0}(y_{1})e^{\eta_{23}} + \Lambda_{24,0}(y_{1})e^{\eta_{24}} + \Lambda_{34,0}(y_{2} - y_{1})e^{\eta_{34}}\}]^{1/\theta + 2}$$

$$\begin{split} f_2(y_1, y_2; X) &= \int_0^\infty f_2(y_1, y_2; X, \gamma) f_\gamma(\gamma) \, d\gamma \\ &= \int_0^\infty \left[(1 - \pi) \cdot \gamma \cdot \lambda_{23,0}(y_1) e^{\eta_{23}} \cdot \exp\left\{ -\gamma \left(\Lambda_{23,0}(y_1) e^{\eta_{23}} + \Lambda_{24,0}(y_1) e^{\eta_{24}} \right) \right\} \cdot \left[\exp\left\{ -\gamma \Lambda_{34,0}(y_2 - y_1) e^{\eta_{34}} \right\} \cdot \left[\Gamma \left(1/\theta \right) \right]^{-1} \theta^{-\frac{1}{\theta}} \gamma^{\frac{1}{\theta} - 1} \exp\left\{ -\gamma/\theta \right\} \right] \, d\gamma \\ &= (1 - \pi) \cdot \lambda_{23,0}(y_1) e^{\eta_{23}} \cdot \left[1 + \theta \left\{ \Lambda_{23,0}(y_1) e^{\eta_{23}} + \Lambda_{24,0}(y_1) e^{\eta_{24}} + \Lambda_{34,0}(y_2 - y_1) e^{\eta_{34}} \right\} \right]^{-1/\theta - 1} \end{split}$$

$$f_{3}(y_{1}, y_{2}; X) = \int_{0}^{\infty} f_{3}(y_{1}, y_{2}; X, \gamma) f_{\gamma}(\gamma) d\gamma$$

$$= \int_{0}^{\infty} [\pi \cdot \gamma \cdot \lambda_{14,0}(y_{2})e^{\eta_{14}} \cdot \exp\{-\gamma\Lambda_{14,0}(y_{2})e^{\eta_{14}}\} \cdot \left[\Gamma(1/\theta)\right]^{-1} \theta^{-\frac{1}{\theta}} \gamma^{\frac{1}{\theta}-1} \exp\{-\gamma/\theta\} d\gamma$$

$$= \pi \cdot \lambda_{14,0}(y_{2})e^{\eta_{14}} \cdot \left[\Gamma(1/\theta)\right]^{-1} \theta^{-\frac{1}{\theta}} \cdot \int_{0}^{\infty} \gamma^{1/\theta} \cdot \exp\left[-\gamma\left\{\frac{1}{\theta} + \Lambda_{14,0}(y_{2})e^{\eta_{14}}\right\}\right] d\gamma$$

$$= \pi \cdot \lambda_{14,0}(y_{2})e^{\eta_{14}} \{1 + \theta \cdot \Lambda_{14,0}(y_{2})e^{\eta_{14}}\}^{-1/\theta-1}$$

$$\begin{split} f_4(y_1, y_2; X) &= \int_0^\infty f_4(y_1, y_2; X, \gamma) f_\gamma(\gamma) \ d\gamma \\ &= \int_0^\infty \left[\pi \cdot \gamma \cdot \lambda_{14,0}(y_2) e^{\eta_{14}} \cdot \exp\left\{ -\gamma \Lambda_{14,0}(y_2) e^{\eta_{14}} \right\} + \\ &\quad (1 - \pi) \cdot \gamma \cdot \lambda_{24,0}(y_2) e^{\eta_{24}} \cdot \exp\left\{ -\gamma \left(\Lambda_{23,0}(y_2) e^{\eta_{23}} + \Lambda_{24,0}(y_2) e^{\eta_{24}} \right) \right\} \right] \cdot f_\gamma(\gamma) \ d\gamma \\ &= \int_0^\infty \left[\pi \cdot \gamma \cdot \lambda_{14,0}(y_2) e^{\eta_{14}} \cdot \exp\left\{ -\gamma \Lambda_{14,0}(y_2) e^{\eta_{14}} \right\} + \\ &\quad (1 - \pi) \cdot \gamma \cdot \lambda_{24,0}(y_2) e^{\eta_{24}} \cdot \exp\left\{ -\gamma \left(\Lambda_{23,0}(y_2) e^{\eta_{23}} + \Lambda_{24,0}(y_2) e^{\eta_{24}} \right) \right\} \right] \cdot \\ &\quad \left[\Gamma \left(1/\theta \right) \right]^{-1} \theta^{-\frac{1}{\theta}} \gamma^{\frac{1}{\theta} - 1} \exp\left\{ -\gamma/\theta \right\} \ d\gamma \end{split}$$

$$= \pi \cdot \lambda_{14,0}(y_2) e^{\eta_{14}} \left\{ 1 + \theta \cdot \Lambda_{14,0}(y_2) e^{\eta_{14}} \right\}^{-1/\theta - 1} + (1 - \pi) \cdot \lambda_{24,0}(y_2) e^{\eta_{24}} \left[1 + \theta \left\{ \Lambda_{23,0}(y_2) e^{\eta_{23}} + \Lambda_{24,0}(y_2) e^{\eta_{24}} \right\} \right]^{-1/\theta - 1}$$

$$\begin{split} f_5(y_1, y_2; X) &= \int_0^\infty f_5(y_1, y_2; X, \gamma) \cdot f_\gamma(\gamma) \, d\gamma \\ &= \int_0^\infty \pi \exp\left\{-\gamma \cdot \Lambda_{14,0}(y_2) e^{\eta_{14}}\right\} \cdot \left[\Gamma\left(1/\theta\right)\right]^{-1} \theta^{-\frac{1}{\theta}} \gamma^{\frac{1}{\theta}-1} \exp\left\{-\gamma/\theta\right\} \, d\gamma \\ &= \pi \left[\Gamma\left(1/\theta\right)\right]^{-1} \theta^{-\frac{1}{\theta}} \int_0^\infty \gamma^{\frac{1}{\theta}-1} \exp\left[-\gamma \left\{\frac{1}{\theta} + \Lambda_{14,0}(y_2) e^{\eta_{14}}(t)\right\}\right] \, d\gamma \\ &= \pi \left\{1 + \theta \cdot \Lambda_{14,0}(y_2) e^{\eta_{14}}\right\}^{-1/\theta} \end{split}$$

$$\begin{split} f_{6}(y_{1}, y_{2}; X) &= \int_{0}^{\infty} f_{6}(y_{1}, y_{2}; X, \gamma) f_{\gamma}(\gamma) \, d\gamma \\ &= \int_{0}^{\infty} \left[\pi \exp\left\{ -\gamma \cdot \Lambda_{14,0}(y_{2})e^{\eta_{14}} \right\} + \\ &\quad (1 - \pi) \cdot \exp\left\{ -\gamma \left(\Lambda_{23,0}(y_{2})e^{\eta_{23}} + \Lambda_{24,0}(y_{2})e^{\eta_{24}} \right) \right\} \right] \cdot f_{\gamma}(\gamma) \, d\gamma \\ &= \int_{0}^{\infty} \left[\pi \exp\left\{ -\gamma \cdot \Lambda_{14,0}(y_{2})e^{\eta_{14}} \right\} + \\ &\quad (1 - \pi) \cdot \exp\left\{ -\gamma \left(\Lambda_{23,0}(y_{2})e^{\eta_{23}} + \Lambda_{24,0}(y_{2})e^{\eta_{24}} \right) \right\} \right] \cdot \\ &\quad \left[\Gamma \left(1/\theta \right) \right]^{-1} \theta^{-\frac{1}{\theta}} \gamma^{\frac{1}{\theta} - 1} \exp\left\{ -\gamma/\theta \right\} \, d\gamma \\ &= \pi \left\{ 1 + \theta \cdot \Lambda_{14,0}(y_{2})e^{\eta_{14}} \right\}^{-1/\theta} + (1 - \pi) \left[1 + \theta \left\{ \Lambda_{23,0}(y_{2})e^{\eta_{23}} + \Lambda_{24,0}(y_{2})e^{\eta_{24}} \right\} \right]^{-1/\theta} \end{split}$$

C Joint density

As in the illness-death model with shared frailty, the transition hazards (6)-(9) determine the joint distribution of T_1 and T_2 . We denote the joint density on the observable region $\mathcal{U} = (0 < T_1 < T_2) \cap (0 < T_1 < \tau)$, by $g_{\mathcal{U}}(t_1, t_2)$, for $t_1 > t_2$. We assign the remaining probability mass along the line $T_1 = \tau$, with density denoted $g_{T_1=\tau,T_2}(t_2)$, representing subjects who never experience the non-terminal event T_1 . This implies that the probability mass for non-susceptible patients (L = 1) is distributed along the line $T_1 = \tau$. For patients who are susceptible (L = 0), the set of possible transitions coincides with the standard illness-death model in Xu et al. (2010) [1] and the corresponding density expressions align with the joint density, which we denote by f and can be found in the Supplementary Materials of Lee et al (2016) [2].

We derive the joint density corresponding to the model in Section 3 for the semi-Markov model, where
$$\begin{split} \Lambda_{\omega}(t) &= \int_{0}^{t} \lambda_{\omega}(s) \ ds. \\ g(t_{1}, t_{2} | \gamma) &= g(T_{1} = t_{1}, T_{2} = t_{2} | \gamma) \\ &= P(L = 1) \cdot g(t_{1}, t_{2} | L = 1, \gamma) + P(L = 0) \cdot g(t_{1}, t_{2} | L = 0, \gamma) \\ &= \pi \cdot g(t_{1}, t_{2} | L = 1, \gamma) + (1 - \pi) \cdot g(t_{1}, t_{2} | L = 0, \gamma) \\ &= \pi \cdot \lambda_{14}(t_{2} | \gamma) \cdot \exp\left\{-\Lambda_{14}(t_{2})\right\} \cdot 1(T_{1} = \tau) + (1 - \pi) \cdot f(t_{1}, t_{2} | \gamma) \\ &= \pi \cdot \lambda_{14}(t_{2} | \gamma) \cdot \exp\left\{-\Lambda_{14}(t_{2})\right\} \cdot 1(T_{1} = \tau) + (1 - \pi) \cdot f(t_{1}, t_{2} | \gamma) \\ &= \pi \cdot \lambda_{14}(t_{2} | \gamma) \cdot \exp\left\{-\Lambda_{14}(t_{2})\right\} \cdot 1(T_{1} = \tau) + (1 - \pi) \cdot f(t_{1}, \tau_{2}) + (1 - \pi) \cdot [\lambda_{23}(t_{1} | \gamma)\lambda_{34}(t_{2} - t_{1} | t_{1}, \gamma) \cdot \\ &\qquad \exp\left\{-\Lambda_{23}(t_{1} | \gamma) - \Lambda_{24}(t_{1} | \gamma) - \int_{0}^{t_{2} - t_{1}} \lambda_{34}(v | t_{1}, \gamma) \ dv\right\} \cdot 1(t_{1} < t_{2}) + \\ &\qquad \lambda_{24}(t_{2} | \gamma) \exp\left\{-\Lambda_{23}(t_{2} | \gamma) - \Lambda_{34}(t_{2} | \gamma)\right\} \cdot 1(T_{1} = \tau)] \,. \end{split}$$

Collecting like terms, we see that the probability density along the line $T_1 = \tau$ is given by

$$g_{T_1=\tau,T_2}(t_2|\gamma) = \pi \cdot \lambda_{14}(t_2|\gamma) \cdot \exp\left\{-\Lambda_{14}(t_2)\right\} + (1-\pi) \cdot \lambda_{24}(t_2|\gamma) \exp\left\{-\Lambda_{23}(t_2|\gamma) - \Lambda_{34}(t_2|\gamma)\right\}$$

and the probability density over the region \mathcal{U} is given by

$$g_{\mathcal{U}}(t_1, t_2|\gamma) = (1 - \pi) \cdot \lambda_{23}(t_1|\gamma)\lambda_{34}(t_2 - t_1|t_1, \gamma) \exp\left\{-\Lambda_{23}(t_1|\gamma) - \Lambda_{24}(t_1|\gamma) - \int_0^{t_2 - t_1} \lambda_{34}(v|t_1, \gamma) \, dv\right\}$$

D Equivalence of absolute risk profiles and transition probabilities

Showing the equivalence of the transition probabilities presented in Putter et al. [3], P_{12}^0 , P_{13}^2 , P_{13}^1 and P_{11} , and the absolute risk profiles presented in Section 3.5, which are calculated via integration of the joint density function. Note: The expressions in Putter et al. [3] are presented for the Markov model.

$$\begin{split} \text{Let } \Lambda_{\omega}(t) &= \int_{0}^{t} \lambda_{\omega}(s) \, ds. \\ p_{a\text{GVHD only}}(t|\gamma) &= \int_{t}^{\infty} \int_{0}^{t} g_{II}(u,v|\gamma) \, du \, dv \\ &= \int_{t}^{\infty} \int_{0}^{t} \lambda_{1}(u|\gamma) \cdot \lambda_{3}(v|\gamma) \exp\{-\Lambda_{1}(u|\gamma) - \Lambda_{2}(u|\gamma) - \Lambda_{3}(v|\gamma) + \Lambda_{3}(u|\gamma)\} \, du \, dv \\ &= \left[\int_{0}^{t} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(u|\gamma) - \Lambda_{2}(u|\gamma) + \Lambda_{3}(u|\gamma)\} \, du\right] \left[\int_{t}^{\infty} \lambda_{3}(v|\gamma) \exp\{-\Lambda_{3}(v|\gamma)\} \, dv\right] \\ &= \int_{0}^{t} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(u|\gamma) - \Lambda_{2}(u|\gamma) + \Lambda_{3}(u|\gamma)\} \cdot \exp\{-\Lambda_{3}(t|\gamma)\} \, du \\ &= P_{12}(0,t|\gamma) \\ &= \\ p_{a\text{GVHD and death}(t|\gamma) &= \int_{0}^{t} \int_{0}^{t} g_{II}(u,v|\gamma) \, dv \, du \\ &= \int_{0}^{t} \lambda_{1}(u|\gamma) \cdot \lambda_{3}(v|\gamma) \exp\{-\Lambda_{1}(u|\gamma) - \Lambda_{2}(u|\gamma) - \Lambda_{3}(v|\gamma) + \Lambda_{3}(u|\gamma)\} \, dv \, du \\ &= \int_{0}^{t} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(u|\gamma) - \Lambda_{2}(u|\gamma)\} \cdot \left[\int_{u}^{t} \lambda_{3}(v|\gamma) \cdot \exp\{-\Lambda_{3}(v|\gamma) + \Lambda_{3}(u|\gamma)\} \, dv\right] \, du \\ &= P_{13}^{*}(0,t|\gamma) \\ &= \\ p_{\text{death only}}(t|\gamma) &= \int_{0}^{t} g_{T-\tau,T_{2}}(v|\gamma) \, dv \\ &= \int_{0}^{t} \lambda_{2}(v|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \int_{u}^{\infty} g_{II}(u,v|\gamma) \, dv \, du + \int_{t}^{\infty} g_{T-\tau,T_{2}}(v|\gamma) \, dv \\ &= \int_{t}^{\infty} \int_{u}^{\infty} \lambda_{1}(u|\gamma) \cdot \lambda_{3}(v|\gamma) \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \end{aligned}$$

$$\begin{split} &\int_{t}^{\infty} \lambda_{2}(v|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \lambda_{1}(u|\gamma) \cdot \exp\{-\Lambda_{1}(u|\gamma) - \Lambda_{2}(u|\gamma)\} \, du + \int_{t}^{\infty} \lambda_{2}(v|\gamma) \cdot \exp\{-\Lambda_{1}(v|\gamma) - \Lambda_{2}(v|\gamma)\} \, dv \\ &= \int_{t}^{\infty} \left(\lambda_{1}(u|\gamma) + \lambda_{2}(u|\gamma)\right) \cdot \exp\{-\Lambda_{1}(u|\gamma) - \Lambda_{2}(u|\gamma)\} \, du \\ &= \exp\{-\Lambda_{1}(t|\gamma) - \Lambda_{2}(t|\gamma)\} \\ &= P_{11}(0, t|\gamma) \end{split}$$

E Additional simulation results

E.1 Baseline scenario

Table E.1: Empirical and analytical standard errors calculated using the standard deviations of the sampling distributions and means of the estimated analytical standard errors, based on 2,000 simulated datasets generated under the 'baseline' scenario with n=5,000.

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	$\operatorname{Conlon}^{\gamma}$	Conlon	ID^γ	Cure	MSM
	$\overline{SD \ SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$
Non-susceptible fra	ction						
$\beta_{s,0}$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.75 \ \ 0.05$	$0.05 \ 0.05$		$0.05 \ 0.05$	
$\beta_{s,x}$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.56 \ 0.06$	$0.06 \ \ 0.07$		$0.07 \ 0.07$	
Non-susceptible: 1	$\rightarrow 4$						
$\log(\kappa_{14})$	$0.29 \ \ 0.27$	$0.20 \ 0.18$	1.56 0.37	$0.22 \ \ 0.20$			
$\log(\alpha_{14})$	$0.03 \ 0.03$	$0.03 \ 0.02$	$0.13 \ 0.04$	$0.03 \ 0.02$			
$\beta_{14,x}$	$0.05 \ 0.05$	$0.05 \ 0.04$	$1.32 \ 0.06$	$0.05 \ 0.05$			
Susceptible: $2 \rightarrow 4$							
$\log(\kappa_{24})$	$0.15 \ 0.14$	$0.14 \ 0.14$	$0.74 \ \ 0.13$	$0.13 \ 0.14$	$0.10 \ 0.10$		
$\log(\alpha_{24})$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.43 \ 0.03$	$0.03 \ 0.03$	$0.04 \ 0.03$		
$\beta_{24,x}$	$0.08 \ \ 0.08$	$0.09 \ 0.08$	$0.60 \ 0.08$	$0.08 \ \ 0.08$	$0.06 \ 0.05$		$0.04 \ 0.04$
Susceptible: $2 \rightarrow 3$							
$\log(\kappa_{23})$	$0.12 \ 0.11$	$0.15 \ 0.14$	$0.76 \ 0.05$	$0.05 \ 0.05$	$0.04 \ 0.04$	$0.05 \ 0.05$	
$\log(\alpha_{23})$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.64 \ 0.02$	$0.02 \ \ 0.02$	$0.04 \ 0.03$	$0.02 \ \ 0.02$	
$\beta_{23,x}$	$0.06 \ 0.06$	$0.06 \ 0.06$	$2.90 \ 0.06$	$0.06 \ 0.06$	$0.06 \ 0.05$	$0.06 \ 0.06$	$0.05 \ 0.05$
Susceptible: $3 \rightarrow 4$							
$\log(\kappa_{34})$	$0.14 \ 0.14$	$0.12 \ 0.11$	$0.78 \ 0.15$	$0.12 \ 0.11$	$0.18 \ 0.17$		
$\log(\alpha_{34})$	$0.02 \ \ 0.02$	$0.02 \ \ 0.02$	$0.24 \ \ 0.02$	$0.02 \ \ 0.02$	$0.03 \ 0.02$		
$\beta_{34,x}$	$0.06 \ \ 0.05$	$0.05 \ 0.05$	$1.23 \ 0.05$	$0.05 \ 0.05$	$0.06 \ 0.06$		0.05 0.05
log-frailty variance							
$\log(\theta)$	$0.12 \ 0.12$		4.35 0.17		$0.16 \ 0.12$		

Table E.2:	Coverage	probabilities,	based on	2,000	simulated	datasets	generated	under the	e 'baseline'	' scenario
with $n=5,0$	000.									

	FI-ID^{γ}	FI-ID	$\operatorname{Conlon}^{\gamma}$	Conlon
Non-susceptible fra	ction			
$\beta_{s,0}$	0.94	0.44	0.46	0.95
$\beta_{s,x}$	0.95	0.95	0.91	0.90
Non-susceptible: 1	$\rightarrow 4$			
$\log(\kappa_{14})$	0.95	0.00	0.72	0.00
$\log(\alpha_{14})$	0.94	0.00	0.76	0.00
$\beta_{14,x}$	0.95	0.75	0.87	0.92
Susceptible: $2 \rightarrow 4$				
$\log(\kappa_{24})$	0.94	0.85	0.93	0.67
$\log(\alpha_{24})$	0.95	0.09	0.86	0.08
$\beta_{24,x}$	0.94	0.74	0.95	0.90
Susceptible: $2 \rightarrow 3$				
$\log(\kappa_{23})$	0.95	0.92	0.00	0.00
$\log(\alpha_{23})$	0.95	0.54	0.03	0.30
$\beta_{23,x}$	0.95	0.94	0.95	0.96
Susceptible: $3 \rightarrow 4$				
$\log(\kappa_{34})$	0.95	0.00	0.95	0.00
$\log(\alpha_{34})$	0.95	0.00	0.96	0.00
$\beta_{34,x}$	0.94	0.92	0.94	0.92
log-frailty variance				
$\log(\theta)$	0.95		0.96	

E.2 Reducing sample size from n=5,000 to n=1,000

Table E.3: Mean point estimates, based on 2,000 simulated datasets generated under the 'baseline' scenario with n=1,000.

	Truth	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	$\operatorname{Conlon}^{\gamma}$	Conlon	ID^{γ}	Cure	MSM
Non-susceptible frac	tion							
$\beta_{s,0}$	-0.41	-0.41	-0.32	-0.53	-0.41		0.14	
$\beta_{s,x}$	0.50	0.50	0.50	0.53	0.54		0.51	
Non-susceptible: 1 –	$\rightarrow 4$							
$\log(\kappa_{14})$	-8.52	-8.55	-7.13	-9.13	-7.49			
$\log(\alpha_{14})$	0.34	0.34	0.15	0.36	0.19			
$\beta_{14,x}$	0.25	0.25	0.19	0.29	0.23			
Susceptible: $2 \to 4$								
$\log(\kappa_{24})$	-5.30	-5.33	-5.23	-5.26	-5.12	-5.37		
$\log(\alpha_{24})$	0.34	0.34	0.23	0.30	0.25	0.09		
$\beta_{24,x}$	0.50	0.50	0.41	0.52	0.45	-0.08		0.08
Susceptible: $2 \rightarrow 3$								
$\log(\kappa_{23})$	-2.16	-2.21	-2.41	-1.80	-1.77	-1.98	-1.50	
$\log(\alpha_{23})$	-0.69	-0.69	-0.75	-0.63	-0.63	-1.10	-0.58	
$\beta_{23,x}$	0.25	0.25	0.24	0.25	0.24	-0.31	0.38	-0.26
Susceptible: $3 \rightarrow 4$								
$\log(\kappa_{34})$	-6.21	-6.25	-5.70	-6.26	-5.70	-7.04		
$\log(\alpha_{34})$	0.26	0.27	0.16	0.27	0.16	0.37		
$\beta_{34,x}$	0.15	0.15	0.14	0.16	0.14	0.11		0.15
log-frailty variance								
$\log(\theta)$	-1.71	-1.79		-2.06		-0.52		

Table E.4: Empirical and analytical standard errors calculated using the standard deviations of the sampling distributions and means of the estimated analytical standard errors, based on 2,000 simulated datasets generated under the 'baseline' scenario with n=1,000.

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	$\operatorname{Conlon}^{\gamma}$	Conlon	ID^γ	Cure	MSM
	$SD \widehat{SE}$	$SD \ \widehat{SE}$	$SD \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$
Non-susceptible fract	tion						
$\beta_{s,0}$	$0.10 \ 0.10$	$0.10 \ 0.11$	$0.55 \ 0.12$	$0.11 \ 0.11$		$0.11 \ 0.12$	
$\beta_{s,x}$	$0.14 \ 0.14$	$0.14 \ 0.14$	$0.26 \ 0.15$	$0.14 \ 0.15$		$0.14 \ 0.15$	
Non-susceptible: $1 -$	$\rightarrow 4$						
$\log(\kappa_{14})$	$0.63 \ 0.62$	$0.47 \ 0.42$	$1.65 \ 0.84$	$0.52 \ 0.46$			
$\log(\alpha_{14})$	$0.07 \ 0.07$	$0.06 \ 0.05$	$1.19 \ 0.09$	$0.06 \ 0.05$			
$\beta_{14,x}$	$0.12 \ \ 0.12$	$0.10 \ 0.10$	$0.71 \ 0.13$	$0.11 \ 0.10$			
Susceptible: $2 \to 4$							
$\log(\kappa_{24})$	$0.32 \ \ 0.31$	$0.31 \ \ 0.32$	$0.64 \ \ 0.30$	$0.30 \ \ 0.31$	$0.24 \ \ 0.23$		
$\log(\alpha_{24})$	$0.07 \ 0.07$	$0.07 \ 0.08$	$0.36 \ 0.06$	$0.06 \ \ 0.06$	$0.08 \ 0.06$		
$\beta_{24,x}$	$0.18 \ 0.18$	$0.21 \ 0.18$	$0.54 \ 0.18$	$0.19 \ 0.18$	$0.12 \ 0.12$		$0.08 \ \ 0.08$
Susceptible: $2 \rightarrow 3$							
$\log(\kappa_{23})$	$0.31 \ \ 0.30$	$0.68 \ 0.85$	$0.58 \ 0.11$	$0.10 \ 0.11$	$0.08 \ 0.10$	$0.12 \ 0.12$	
$\log(\alpha_{23})$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.61 \ 0.05$	$0.05 \ 0.05$	$0.09 \ 0.07$	$0.05 \ 0.05$	
$\beta_{23,x}$	$0.13 \ 0.13$	$0.13 \ 0.13$	$0.63 \ 0.13$	$0.12 \ 0.13$	$0.12 \ 0.12$	$0.14 \ 0.15$	$0.10 \ 0.11$
Susceptible: $3 \rightarrow 4$							
$\log(\kappa_{34})$	$0.31 \ \ 0.31$	$0.27 \ 0.26$	1.11 0.33	$0.27 \ \ 0.26$	$0.42 \ \ 0.37$		
$\log(\alpha_{34})$	$0.05 \ 0.05$	$0.04 \ 0.04$	$0.10 \ 0.05$	$0.04 \ 0.04$	$0.06 \ 0.05$		
$\beta_{34,x}$	$0.12 \ 0.12$	$0.11 \ 0.11$	$0.61 \ 0.12$	$0.11 \ 0.11$	$0.14 \ 0.14$		$0.11 \ 0.11$
log-frailty variance							
$\log(\theta)$	$0.37 \ \ 0.35$		$3.99 \ 0.66$		$0.38 \ \ 0.29$		

Table E.5: Coverage probabilities, based on 2,000 simulated datasets generated under the 'baseline' scenario with n=1,000.

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	Conlon^{γ}	Conlon
Non-susceptible fract	ion			
$\beta_{s,0}$	0.95	0.87	0.87	0.96
$\beta_{s,x}$	0.96	0.97	0.95	0.95
Non-susceptible: $1 - $	$\rightarrow 4$			
$\log(\kappa_{14})$	0.94	0.12	0.93	0.40
$\log(\alpha_{14})$	0.94	0.08	0.92	0.27
$\beta_{14,x}$	0.94	0.89	0.94	0.93
Susceptible: $2 \to 4$				
$\log(\kappa_{24})$	0.95	0.94	0.95	0.90
$\log(\alpha_{24})$	0.96	0.75	0.95	0.71
$\beta_{24,x}$	0.95	0.89	0.95	0.94
Susceptible: $2 \rightarrow 3$				
$\log(\kappa_{23})$	0.95	0.98	0.10	0.05
$\log(\alpha_{23})$	0.96	0.92	0.57	0.78
$\beta_{23,x}$	0.96	0.94	0.95	0.96
Susceptible: $3 \rightarrow 4$				
$\log(\kappa_{34})$	0.95	0.47	0.95	0.47
$\log(\alpha_{34})$	0.95	0.30	0.95	0.30
$\beta_{34,x}$	0.95	0.93	0.95	0.93
log-frailty variance				
$\log(\theta)$	0.97		0.97	

E.3 Administrative censoring via a censoring variable simulated from an Exponential distribution with mean 100

Table E.6: Mean point estimates, based on 2,000 simulated datasets of size n=5,000 generated with administrative censoring via a censoring variable simulated from an Exponential distribution with mean 100.

	Truth	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	$\operatorname{Conlon}^{\gamma}$	Conlon	ID^γ	Cure	MSM
Non-susceptible frac	tion							
$\beta_{s,0}$	-0.41	-0.40	-0.31	-0.60	-0.52		0.12	
$\beta_{s,x}$	0.50	0.50	0.48	0.57	0.56		0.51	
Non-susceptible: 1 -	$\rightarrow 4$							
$\log(\kappa_{14})$	-8.52	-8.52	-7.26	-10.46	-9.33			
$\log(\alpha_{14})$	0.34	0.33	0.17	0.52	0.40			
$\beta_{14,x}$	0.25	0.26	0.17	0.47	0.40			
Susceptible: $2 \rightarrow 4$								
$\log(\kappa_{24})$	-5.30	-5.32	-5.25	-5.23	-5.14	-5.20		
$\log(\alpha_{24})$	0.34	0.34	0.26	0.29	0.25	0.26		
$\beta_{24,x}$	0.50	0.50	0.43	0.52	0.49	-0.05		0.08
Susceptible: $2 \rightarrow 3$								
$\log(\kappa_{23})$	-2.16	-2.17	-2.27	-1.83	-1.81	-2.08	-1.50	
$\log(\alpha_{23})$	-0.69	-0.69	-0.74	-0.64	-0.67	-0.64	-0.59	
$\beta_{23,x}$	0.25	0.25	0.23	0.24	0.24	-0.18	0.37	-0.26
Susceptible: $3 \rightarrow 4$								
$\log(\kappa_{34})$	-6.21	-6.21	-5.95	-6.19	-5.95	-7.38		
$\log(\alpha_{34})$	0.26	0.26	0.21	0.25	0.21	0.40		
$\beta_{34,x}$	0.15	0.15	0.14	0.15	0.14	0.03		0.14
log-frailty variance								
$\log(\theta)$	-1.71	-1.75		-2.02		0.81		

Table E.7: Empirical and analytical standard errors calculated using the standard deviations of the sampling distributions and means of the estimated analytical standard errors, based on 2,000 simulated datasets of size n=5,000 generated with administrative censoring via a censoring variable simulated from an Exponential distribution with mean 100.

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	Conlon^{γ}	Conlon	ID^γ	Cure	MSM
	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$
Non-susceptible frac	tion						
$\beta_{s,0}$	$0.07 \ 0.06$	$0.06 \ 0.06$	$0.15 \ 0.08$	$0.07 \ 0.07$		$0.07 \ 0.07$	
$\beta_{s,x}$	$0.08 \ \ 0.08$	$0.08 \ \ 0.08$	$0.12 \ 0.08$	$0.08 \ \ 0.08$		$0.08 \ \ 0.08$	
Non-susceptible: 1 –	$\rightarrow 4$						
$\log(\kappa_{14})$	$0.91 \ 0.87$	$0.67 \ 0.69$	$1.17 \ 1.22$	$0.98 \ \ 0.99$			
$\log(\alpha_{14})$	$0.11 \ 0.11$	$0.10 \ 0.10$	$0.12 \ 0.12$	$0.11 \ 0.11$			
$\beta_{14,x}$	$0.16 \ 0.16$	$0.14 \ 0.14$	$0.53 \ 0.21$	$0.18 \ 0.18$			
Susceptible: $2 \to 4$							
$\log(\kappa_{24})$	$0.15 \ 0.15$	$0.15 \ 0.16$	$0.14 \ 0.14$	$0.14 \ 0.14$	$0.11 \ 0.12$		
$\log(\alpha_{24})$	$0.04 \ 0.03$	$0.04 \ 0.04$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$		
$\beta_{24,x}$	$0.10 \ 0.09$	$0.10 \ 0.09$	$0.10 \ 0.10$	$0.09 \ 0.09$	$0.09 \ 0.09$		$0.04 \ 0.04$
Susceptible: $2 \rightarrow 3$							
$\log(\kappa_{23})$	$0.14 \ 0.13$	$0.18 \ 0.17$	$0.06 \ 0.05$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.06 \ 0.06$	
$\log(\alpha_{23})$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.02 \ \ 0.02$	$0.03 \ 0.03$	$0.03 \ 0.03$	
$\beta_{23,x}$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.08 \ \ 0.08$	$0.05 \ 0.05$
Susceptible: $3 \rightarrow 4$							
$\log(\kappa_{34})$	$0.18 \ 0.18$	$0.16 \ 0.17$	$0.19 \ 0.19$	$0.16 \ 0.17$	$0.21 \ \ 0.20$		
$\log(\alpha_{34})$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$		
$\beta_{34,x}$	$0.08 \ \ 0.08$	$0.08 \ \ 0.07$	$0.08 \ \ 0.08$	$0.08 \ \ 0.07$	$0.10 \ 0.10$		$0.05 \ 0.05$
log-frailty variance							
$\log(\theta)$	$0.23 \ \ 0.23$		$0.67 \ 0.67$		$0.07 \ 0.07$		

Table E.8: Coverage probabilities, based on 2,000 simulated datasets of size n=5,000 generated with administrative censoring via a censoring variable simulated from an Exponential distribution with mean 100.

	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}$	FI-ID	Conlon^{γ}	Conlon
Non-susceptible frac	tion			
$\beta_{s,0}$	0.95	0.63	0.34	0.66
$\beta_{s,x}$	0.95	0.94	0.86	0.89
Non-susceptible: 1 –	$\rightarrow 4$			
$\log(\kappa_{14})$	0.93	0.54	0.67	0.89
$\log(\alpha_{14})$	0.94	0.62	0.66	0.89
$\beta_{14,x}$	0.95	0.90	0.82	0.89
Susceptible: $2 \to 4$				
$\log(\kappa_{24})$	0.95	0.92	0.92	0.78
$\log(\alpha_{24})$	0.95	0.45	0.72	0.17
$\beta_{24,x}$	0.95	0.87	0.94	0.95
Susceptible: $2 \rightarrow 3$				
$\log(\kappa_{23})$	0.95	0.97	0.00	0.00
$\log(\alpha_{23})$	0.95	0.71	0.50	0.83
$\beta_{23,x}$	0.94	0.93	0.95	0.94
Susceptible: $3 \rightarrow 4$				
$\log(\kappa_{34})$	0.96	0.63	0.95	0.63
$\log(\alpha_{34})$	0.96	0.56	0.95	0.56
$\beta_{34,x}$	0.95	0.94	0.95	0.94
log-frailty variance				
$\log(\theta)$	0.97		0.99	

E.4 Administrative censoring solely of the terminal event at 365 days

Table E.9: Mean point estimates, based on 2,000 simulated datasets of size n=5,000 with administrative censoring solely of the terminal event at 365 days post-transplantation. For the proposed finite interval illness-death model with frailty using B-spline parameterized baseline hazard functions (FI-ID γ_{BS}), we made the following specifications: for all transitions we placed two internal knots at the 33rd and 67th percentiles of the observed event times; we used quadratic B-splines for the $1\rightarrow 4$ and $3\rightarrow 4$ transitions and cubic B-splines for the $2\rightarrow 4$ and $2\rightarrow 3$ transitions.

	Truth	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}{}_{WB}$	$\text{FI-ID}^{\gamma}{}_{BS}$	FI-ID	Conlon^{γ}	Conlon	ID^γ	Cure	MSM
Non-susceptible fra	ction								
$\beta_{s,0}$	-0.41	-0.41	-0.38	-0.34	-0.54	-0.48		0.14	
$\beta_{s,x}$	0.50	0.50	0.49	0.50	0.54	0.53		0.50	
Non-susceptible: 1	$\rightarrow 4$								
$\log(\kappa_{14})$	-8.52	-8.51		-7.53	-9.66	-8.78			
$\log(\alpha_{14})$	0.34	0.33		0.21	0.45	0.35			
$\beta_{14,x}$	0.25	0.25	0.23	0.21	0.33	0.29			
Susceptible: $2 \to 4$									
$\log(\kappa_{24})$	-5.30	-5.30		-5.16	-5.24	-5.10	-5.50		
$\log(\alpha_{24})$	0.34	0.34		0.23	0.31	0.25	0.32		
$\beta_{24,x}$	0.50	0.50	0.48	0.41	0.49	0.46	-0.03		-0.05
Susceptible: $2 \rightarrow 3$									
$\log(\kappa_{23})$	-2.16	-2.17		-2.38	-1.80	-1.79	-2.02	-1.50	
$\log(\alpha_{23})$	-0.69	-0.69		-0.76	-0.62	-0.66	-0.71	-0.58	
$\beta_{23,x}$	0.25	0.25	0.25	0.23	0.22	0.22	-0.23	0.38	-0.26
Susceptible: $3 \rightarrow 4$									
$\log(\kappa_{34})$	-6.21	-6.22		-5.79	-6.18	-5.79	-7.87		
$\log(\alpha_{34})$	0.26	0.26		0.18	0.25	0.18	0.48		
$\beta_{34,x}$	0.15	0.15	0.15	0.14	0.15	0.14	0.01		0.14
log-frailty variance									
$\log(heta)$	-1.71	-1.73	-1.78		-1.89		0.80		

Table E.10: Empirical and analytical standard errors calculated using the standard deviations of the sampling distributions and means of the estimated analytical standard errors, based on 2,000 simulated datasets of size n=5,000 with administrative censoring solely of the terminal event at 365 days post-transplantation. For the proposed finite interval illness-death model with frailty using B-spline parameterized baseline hazard functions (FI-ID γ_{BS}), we made the following specifications: for all transitions we placed two internal knots at the 33rd and 67th percentiles of the observed event times; we used quadratic B-splines for the $1\rightarrow 4$ and $3\rightarrow 4$ transitions and cubic B-splines for the $2\rightarrow 4$ and $2\rightarrow 3$ transitions.

	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}{}_{WB}$	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}{}_{BS}$	FI-ID	$\operatorname{Conlon}^{\gamma}$	Conlon	ID^{γ}	Cure	MSM
	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$\overline{SD \ SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$
Non-susceptible fract	ion							
$\beta_{s,0}$	$0.05 \ \ 0.05$	$0.08 \ 0.06$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.05 \ 0.05$		$0.05 \ 0.05$	
$\beta_{s,x}$	$0.06 \ \ 0.06$	$0.07 \ 0.07$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.06 \ 0.06$		$0.07 \ 0.07$	
Non-susceptible: $1 \rightarrow$	→ 4							
$\log(\kappa_{14})$	$0.43 \ 0.42$		$0.37 \ \ 0.37$	$0.52 \ 0.55$	$0.44 \ 0.47$			
$\log(\alpha_{14})$	$0.05 \ 0.05$		$0.05 \ 0.05$	$0.05 \ 0.06$	$0.05 \ 0.05$			
$\beta_{14,x}$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.06 \ 0.06$	$0.07 \ 0.07$	$0.07 \ 0.07$			
Susceptible: $2 \rightarrow 4$								
$\log(\kappa_{24})$	$0.15 \ 0.14$		$0.13 \ 0.13$	$0.13 \ 0.13$	$0.13 \ 0.13$	$0.11 \ 0.11$		
$\log(\alpha_{24})$	$0.03 \ 0.03$		$0.03 \ 0.03$	$0.03 \ 0.03$	$0.02 \ 0.03$	$0.02 \ \ 0.02$		
$\beta_{24,x}$	$0.08 \ \ 0.08$	$0.09 \ 0.08$	$0.09 \ 0.08$	$0.08 \ \ 0.08$	$0.08 \ \ 0.08$	$0.08 \ \ 0.08$		$0.04 \ 0.04$
Susceptible: $2 \rightarrow 3$								
$\log(\kappa_{23})$	$0.12 \ 0.11$		$0.17 \ 0.16$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.05 \ 0.05$	
$\log(\alpha_{23})$	$0.03 \ 0.03$		$0.03 \ 0.03$	$0.02 \ \ 0.02$	$0.02 \ \ 0.02$	$0.03 \ 0.03$	$0.02 \ 0.02$	
$\beta_{23,x}$	$0.06 \ \ 0.06$	$0.07 \ 0.06$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.05 \ 0.06$	$0.07 \ 0.07$	$0.06 \ 0.06$	$0.05 \ 0.05$
Susceptible: $3 \rightarrow 4$								
$\log(\kappa_{34})$	$0.14 \ 0.14$		$0.12 \ 0.12$	$0.14 \ 0.15$	$0.12 \ 0.12$	$0.17 \ 0.16$		
$\log(\alpha_{34})$	$0.02 \ \ 0.02$		$0.02 \ \ 0.02$	$0.02 \ \ 0.02$	$0.02 \ \ 0.02$	$0.02 \ \ 0.02$		
$\beta_{34,x}$	$0.06 \ \ 0.05$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.06 \ 0.05$	$0.05 \ 0.05$	$0.08 \ \ 0.08$		$0.05 \ 0.05$
log-frailty variance								
$\log(\theta)$	$0.14 \ 0.14$	$0.22 \ \ 0.21$		$0.23 \ \ 0.24$		$0.05 \ 0.05$		

	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}{}_{WB}$	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}{}_{BS}$	FI-ID	Conlon^{γ}	Conlon
Non-susceptible fract	ion				
$\beta_{s,0}$	0.95	0.85	0.69	0.32	0.74
$\beta_{s,x}$	0.95	0.93	0.95	0.91	0.91
Non-susceptible: $1 \rightarrow$	• 4				
$\log(\kappa_{14})$	0.95		0.24	0.46	0.93
$\log(\alpha_{14})$	0.94		0.23	0.49	0.95
$\beta_{14,x}$	0.95	0.94	0.87	0.80	0.92
Susceptible: $2 \to 4$					
$\log(\kappa_{24})$	0.94		0.81	0.92	0.63
$\log(\alpha_{24})$	0.95		0.08	0.82	0.06
$\beta_{24,x}$	0.95	0.95	0.76	0.94	0.91
Susceptible: $2 \rightarrow 3$					
$\log(\kappa_{23})$	0.95		0.88	0.00	0.00
$\log(\alpha_{23})$	0.95		0.46	0.12	0.62
$\beta_{23,x}$	0.95	0.91	0.93	0.93	0.91
Susceptible: $3 \rightarrow 4$					
$\log(\kappa_{34})$	0.95		0.07	0.95	0.07
$\log(\alpha_{34})$	0.95		0.01	0.94	0.01
$\beta_{34,x}$	0.94	0.95	0.93	0.94	0.93
log-frailty variance					
$\log(\theta)$	0.96	0.96		0.96	

Table E.11: Coverage probabilities, based on 2,000 simulated datasets of size n=5,000 with administrative censoring solely of the terminal event at 365 days post-transplantation.

E.5 Increasing the non-susceptibility fraction from 46% to 76%

	Truth	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}$	FI-ID	Conlon^{γ}	Conlon	ID^{γ}	Cure	MSM
Non-susceptible fr	action							
$\beta_{s,0}$	0.84	0.84	0.96	0.74	0.91		1.15	
$\beta_{s,x}$	0.50	0.50	0.54	0.53	0.56		0.61	
Non-susceptible: 1	$l \rightarrow 4$							
$\log(\kappa_{14})$	-8.52	-8.50	-7.23	-8.86	-7.34			
$\log(\alpha_{14})$	0.34	0.33	0.16	0.37	0.18			
$\beta_{14,x}$	0.25	0.25	0.21	0.28	0.22			
Susceptible: $2 \rightarrow 4$	4							
$\log(\kappa_{24})$	-5.30	-5.31	-5.31	-5.33	-5.19	-6.44		
$\log(\alpha_{24})$	0.34	0.34	0.21	0.32	0.25	0.08		
$\beta_{24,x}$	0.50	0.50	0.33	0.52	0.41	0.14		0.15
Susceptible: $2 \rightarrow 3$	3							
$\log(\kappa_{23})$	-2.16	-2.18	-2.09	-1.83	-1.73	-2.76	-1.45	
$\log(\alpha_{23})$	-0.69	-0.69	-0.72	-0.62	-0.63	-1.45	-0.58	
$\beta_{23,x}$	0.25	0.25	0.30	0.23	0.30	-0.39	0.37	-0.40
Susceptible: $3 \rightarrow 4$	4							
$\log(\kappa_{34})$	-6.21	-6.22	-5.67	-6.26	-5.67	-5.91		
$\log(\alpha_{34})$	0.26	0.26	0.15	0.27	0.15	0.20		
$\beta_{34,x}$	0.15	0.15	0.14	0.15	0.14	0.14		0.14
log-frailty variance	е							
$\log(\theta)$	-1.71	-1.74		-1.68		-2.97		

Table E.12: Mean point estimates, based on 2,000 simulated datasets of size n=5,000 generated by increasing the non-susceptibility fraction from 46% to 76%.

Table E.13: Empirical and analytical standard errors calculated using the standard deviations of the sampling distributions and means of the estimated analytical standard errors, based on 2,000 simulated datasets of size n=5,000 generated by increasing the non-susceptibility fraction from 46% to 76%.

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	Conlon^{γ}	Conlon	ID^γ	Cure	MSM
	$SD \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$
Non-susceptible fract	ion						
$\beta_{s,0}$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.06 \ 0.06$	$0.06 \ 0.06$		1.08 0.05	
$\beta_{s,x}$	$0.08 \ \ 0.08$	$0.08 \ \ 0.08$	$0.08 \ \ 0.08$	$0.08 \ \ 0.08$		$0.89 \ 0.08$	
Non-susceptible: $1 \rightarrow$	→ 4						
$\log(\kappa_{14})$	$0.24 \ \ 0.24$	$0.16 \ 0.14$	$0.33 \ 0.33$	$0.17 \ 0.15$			
$\log(\alpha_{14})$	$0.03 \ 0.03$	$0.02 \ 0.02$	$0.04 \ 0.04$	$0.02 \ \ 0.02$			
$\beta_{14,x}$	$0.04 \ 0.04$	$0.04 \ 0.03$	$0.04 \ \ 0.27$	$0.04 \ 0.03$			
Susceptible: $2 \rightarrow 4$							
$\log(\kappa_{24})$	$0.22 \ \ 0.22$	$0.25 \ 0.26$	$0.26 \ 0.80$	$0.24 \ \ 0.24$	$0.11 \ 0.10$		
$\log(\alpha_{24})$	$0.05 \ 0.05$	$0.06 \ 0.06$	$0.13 \ 0.63$	$0.04 \ 0.05$	$0.02 \ 0.02$		
$\beta_{24,x}$	$0.12 \ 0.12$	$0.19 \ 0.16$	$0.14 \ \ 0.31$	$0.15 \ 0.15$	$0.03 \ 0.03$		$0.03 \ 0.03$
Susceptible: $2 \rightarrow 3$							
$\log(\kappa_{23})$	$0.18 \ 0.17$	$0.18 \ 0.17$	$0.08 \ \ 0.73$	$0.08 \ \ 0.07$	$0.05 \ 0.06$	$0.21 \ \ 0.07$	
$\log(\alpha_{23})$	$0.04 \ 0.05$	$0.04 \ 0.05$	$0.04 \ 0.03$	$0.03 \ 0.03$	$0.02 \ 0.03$	$0.06 \ 0.03$	
$\beta_{23,x}$	$0.10 \ 0.10$	$0.10 \ 0.10$	$0.10 \ 0.75$	$0.10 \ 0.10$	$0.07 \ 0.07$	$0.22 \ 0.09$	$0.07 \ 0.07$
Susceptible: $3 \rightarrow 4$							
$\log(\kappa_{34})$	$0.19 \ 0.19$	$0.17 \ 0.17$	$0.20 \ 0.86$	$0.17 \ 0.17$	$0.19 \ 0.19$		
$\log(\alpha_{34})$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$		
$\beta_{34,x}$	$0.08 \ \ 0.08$	$0.08 \ \ 0.07$	$0.09 \ 0.88$	$0.08 \ \ 0.07$	$0.08 \ 0.07$		$0.07 \ 0.07$
log-frailty variance							
$\log(\theta)$	$0.16 \ 0.16$		$0.92 \ 0.19$		$1.02 \ \ 0.54$		

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	$Conlon^{\gamma}$	Conlon
Non-susceptible f	raction			
$\beta_{s,0}$	0.95	0.37	0.63	0.81
$\beta_{s,x}$	0.95	0.93	0.93	0.90
Non-susceptible:	$1 \rightarrow 4$			
$\log(\kappa_{14})$	0.95	0.00	0.85	0.00
$\log(\alpha_{14})$	0.95	0.00	0.85	0.00
$\beta_{14,x}$	0.95	0.76	0.92	0.83
Susceptible: $2 \rightarrow$	4			
$\log(\kappa_{24})$	0.94	0.96	0.95	0.92
$\log(\alpha_{24})$	0.95	0.47	0.95	0.54
$\beta_{24,x}$	0.95	0.79	0.95	0.90
Susceptible: $2 \rightarrow$	3			
$\log(\kappa_{23})$	0.95	0.87	0.01	0.00
$\log(\alpha_{23})$	0.95	0.93	0.43	0.49
$\beta_{23,x}$	0.95	0.91	0.95	0.94
Susceptible: $3 \rightarrow$	4			
$\log(\kappa_{34})$	0.95	0.12	0.94	0.12
$\log(\alpha_{34})$	0.95	0.02	0.94	0.02
$\beta_{34,x}$	0.95	0.93	0.95	0.93
log-frailty variand	ce			
$\log(\theta)$	0.96		0.91	

Table E.14: Coverage probabilities, based on 2,000 simulated datasets of size n=5,000 generated by increasing the non-susceptibility fraction from 46% to 76%.

E.6 Truncated and standard Weibull distributions align on $(0, \tau)$

Table E.15: Mean point estimates, based on 2,000 simulated datasets of size n=5,000 under the modification of α_{23} and κ_{23} so that the hazard functions for the truncated and standard Weibull distributions are comparable on the interval $(0, \tau)$.

	Truth	$\mathrm{FI}\text{-}\mathrm{ID}^{\gamma}$	FI-ID	Conlon^{γ}	Conlon	ID^γ	Cure	MSM
Non-susceptible frac	tion							
$\beta_{s,0}$	-0.41	-0.41	-0.35	-0.38	-0.34		0.51	
$\beta_{s,x}$	0.50	0.50	0.51	0.50	0.51		0.66	
Non-susceptible: 1 -	$\rightarrow 4$							
$\log(\kappa_{14})$	-8.52	-8.51	-7.28	-8.22	-7.21			
$\log(\alpha_{14})$	0.34	0.34	0.17	0.30	0.16			
$\beta_{14,x}$	0.25	0.25	0.21	0.24	0.20			
Susceptible: $2 \to 4$								
$\log(\kappa_{24})$	-5.30	-5.30	-5.08	-5.29	-5.04	-6.05		
$\log(\alpha_{24})$	0.34	0.34	0.24	0.35	0.27	0.40		
$\beta_{24,x}$	0.50	0.50	0.45	0.49	0.46	0.02		0.10
Susceptible: $2 \rightarrow 3$								
$\log(\kappa_{23})$	-7.13	-7.14	-6.70	-7.17	-6.78	-4.49	-6.18	
$\log(\alpha_{23})$	0.64	0.64	0.55	0.65	0.57	-0.05	0.54	
$\beta_{23,x}$	0.25	0.25	0.22	0.25	0.23	-0.45	0.38	-0.46
Susceptible: $3 \rightarrow 4$								
$\log(\kappa_{34})$	-6.21	-6.23	-5.70	-6.16	-5.70	-7.73		
$\log(\alpha_{34})$	0.26	0.26	0.15	0.25	0.15	0.48		
$\beta_{34,x}$	0.15	0.15	0.13	0.15	0.13	-0.01		0.13
log-frailty variance								
$\log(\theta)$	-1.71	-1.73		-1.90		0.58		

Table E.16: Empirical and analytical standard errors calculated using the standard deviations of the sampling distributions and means of the estimated analytical standard errors, based on 2,000 simulated datasets of size n=5,000 under the modification of α_{23} and κ_{23} so that the hazard functions for the truncated and standard Weibull distributions are comparable on the interval $(0, \tau)$.

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	Conlon^{γ}	Conlon	ID^γ	Cure	MSM
	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$SD \ \widehat{SE}$	$\overline{SD \ SE}$	$SD \ \widehat{SE}$
Non-susceptible frac	tion						
$\beta_{s,0}$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.05 \ 0.05$	$0.05 \ 0.05$		$0.05 \ 0.05$	
$\beta_{s,x}$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.06 \ 0.06$		$0.07 \ 0.07$	
Non-susceptible: 1 –	$\rightarrow 4$						
$\log(\kappa_{14})$	$0.27 \ \ 0.27$	$0.20 \ 0.19$	$0.26 \ \ 0.26$	$0.20 \ \ 0.19$			
$\log(\alpha_{14})$	$0.03 \ 0.03$	$0.03 \ 0.02$	$0.03 \ 0.03$	$0.03 \ 0.02$			
$\beta_{14,x}$	$0.05 \ 0.05$	$0.05 \ 0.04$	$0.05 \ 0.05$	$0.05 \ 0.04$			
Susceptible: $2 \to 4$							
$\log(\kappa_{24})$	$0.13 \ 0.12$	$0.10 \ 0.11$	$0.12 \ 0.12$	$0.11 \ 0.11$	$0.25 \ 0.16$		
$\log(\alpha_{24})$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.02 \ \ 0.02$	$0.02 \ \ 0.02$	$0.06 \ 0.04$		
$\beta_{24,x}$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.06 \ 0.06$	$0.08 \ 0.07$		$0.03 \ 0.03$
Susceptible: $2 \rightarrow 3$							
$\log(\kappa_{23})$	$0.18 \ 0.17$	$0.15 \ 0.16$	$0.17 \ 0.17$	$0.15 \ 0.15$	$0.17 \ 0.13$	$0.15 \ 0.15$	
$\log(\alpha_{23})$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.03 \ 0.03$	$0.02 \ \ 0.02$	$0.07 \ 0.04$	$0.02 \ 0.02$	
$\beta_{23,x}$	$0.07 \ 0.08$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.07 \ 0.07$	$0.08 \ 0.08$	$0.07 \ 0.06$	$0.06 \ 0.06$
Susceptible: $3 \rightarrow 4$							
$\log(\kappa_{34})$	$0.16 \ 0.16$	$0.14 \ 0.14$	$0.16 \ 0.16$	$0.14 \ 0.14$	$0.21 \ 0.19$		
$\log(\alpha_{34})$	$0.02 \ \ 0.02$	$0.02 \ 0.02$	$0.02 \ 0.03$	$0.02 \ 0.02$	$0.03 \ 0.02$		
$\beta_{34,x}$	$0.07 \ 0.07$	$0.07 \ 0.06$	$0.07 \ 0.07$	$0.07 \ 0.06$	$0.10 \ 0.09$		$0.06 \ 0.06$
log-frailty variance							
$\log(\theta)$	$0.13 \ 0.13$		$0.15 \ 0.16$		$0.11 \ 0.07$		

Table E.17: Coverage probabilities, based on 2,000 simulated datasets of size n=5,000 under the modification of α_{23} and κ_{23} so that the hazard functions for the truncated and standard Weibull distributions are comparable on the interval $(0, \tau)$.

	$\mathrm{FI}\text{-}\mathrm{ID}^\gamma$	FI-ID	$\operatorname{Conlon}^{\gamma}$	Conlon
Non-susceptible fr	action			
$\beta_{s,0}$	0.95	0.78	0.91	0.70
$\beta_{s,x}$	0.95	0.95	0.95	0.95
Non-susceptible: 1	$\rightarrow 4$			
$\log(\kappa_{14})$	0.95	0.00	0.79	0.00
$\log(\alpha_{14})$	0.95	0.00	0.79	0.00
$\beta_{14,x}$	0.95	0.82	0.94	0.80
Susceptible: $2 \rightarrow 4$	1			
$\log(\kappa_{24})$	0.94	0.46	0.95	0.35
$\log(\alpha_{24})$	0.94	0.02	0.92	0.12
$\beta_{24,x}$	0.95	0.81	0.95	0.88
Susceptible: $2 \rightarrow 3$	3			
$\log(\kappa_{23})$	0.95	0.22	0.94	0.37
$\log(\alpha_{23})$	0.95	0.04	0.94	0.15
$\beta_{23,x}$	0.96	0.92	0.96	0.94
Susceptible: $3 \rightarrow 4$	1			
$\log(\kappa_{34})$	0.95	0.06	0.94	0.06
$\log(\alpha_{34})$	0.95	0.00	0.93	0.00
$\beta_{34,x}$	0.95	0.91	0.95	0.91
log-frailty variance	e			
$\log(\theta)$	0.95		0.84	

E.7 Baseline scenario without frailty, i.e. Gamma frailty variance equal to zero

Table E.18: Mean point estimates, based on 2,000 simulated datasets of size n=5,000 under the modification of no frailty.

	Truth	FI-ID	Conlon	ID	Cure	MSM
Non-susceptible frac	tion					
$\beta_{s,0}$	-0.41	-0.41	-0.46		0.18	
$\beta_{s,x}$	0.50	0.50	0.52		0.45	
Non-susceptible: 1 –	$\rightarrow 4$					
$\log(\kappa_{14})$	-8.52	-8.53	-8.84			
$\log(\alpha_{14})$	0.34	0.34	0.37			
$\beta_{14,x}$	0.25	0.25	0.27			
Susceptible: $2 \to 4$						
$\log(\kappa_{24})$	-5.30	-5.31	-5.26	-4.92		
$\log(\alpha_{24})$	0.34	0.34	0.33	-0.13		
$\beta_{24,x}$	0.50	0.50	0.50	0.06		0.10
Susceptible: $2\to 3$						
$\log(\kappa_{23})$	-2.16	-2.16	-1.77	-1.97	-1.47	
$\log(\alpha_{23})$	-0.69	-0.69	-0.60	-1.32	-0.51	
$\beta_{23,x}$	0.25	0.25	0.23	-0.28	0.35	-0.25
Susceptible: $3 \rightarrow 4$						
$\log(\kappa_{34})$	-6.21	-6.22	-6.22	-6.22		
$\log(\alpha_{34})$	0.26	0.26	0.26	0.26		
$\beta_{34,x}$	0.15	0.15	0.15	0.15		0.17

Table E.19: Empirical and analytical standard errors calculated using the standard deviations of the sampling distributions and means of the estimated analytical standard errors, based on 2,000 simulated datasets of size n=5,000 under the modification of no frailty.

	FI-ID	Conlon	ID	Cure	MSM
	$SD \ \widehat{SE}$				
Non-susceptible frac	tion				
$\beta_{s,0}$	0.05 0.04	0.05 0.05		0.05 0.05	
$\beta_{s,x}$	$0.06 \ 0.06$	$0.06 \ 0.06$		0.06 0.06	
Non-susceptible: 1 -	$\rightarrow 4$				
$\log(\kappa_{14})$	$0.21 \ 0.20$	$0.22 \ 0.22$			
$\log(\alpha_{14})$	$0.02 \ 0.02$	$0.02 \ 0.02$			
$\beta_{14,x}$	0.04 0.04	0.05 0.04			
Susceptible: $2 \to 4$					
$\log(\kappa_{24})$	$0.13 \ 0.14$	$0.13 \ 0.13$	0.08 0.08		
$\log(\alpha_{24})$	$0.03 \ 0.03$	$0.02 \ 0.03$	0.01 0.01		
$\beta_{24,x}$	0.08 0.08	0.08 0.08	$0.03 \ 0.04$		0.04 0.04
Susceptible: $2 \rightarrow 3$					
$\log(\kappa_{23})$	$0.10 \ 0.10$	0.05 0.05	$0.03 \ 0.04$	0.05 0.05	
$\log(\alpha_{23})$	$0.03 \ 0.03$	$0.02 \ 0.02$	$0.01 \ 0.02$	$0.02 \ 0.02$	
$\beta_{23,x}$	0.05 0.05	0.05 0.05	0.05 0.05	0.05 0.06	0.05 0.05
Susceptible: $3 \rightarrow 4$					
$\log(\kappa_{34})$	$0.12 \ 0.12$	$0.12 \ 0.12$	$0.12 \ 0.12$		
$\log(\alpha_{34})$	$0.02 \ 0.02$	$0.02 \ 0.02$	$0.02 \ 0.02$		
$\beta_{34,x}$	0.05 0.05	0.05 0.05	0.05 0.05		0.05 0.05

Table E.20: Coverage probabilities, based on 2,000 simulated datasets of size n=5,000 under the modification of no frailty.

	FI-ID	Conlon
Non-susceptible frac	ction	
$\beta_{s,0}$	0.95	0.78
$\beta_{s,x}$	0.95	0.94
Non-susceptible: 1 -	$\rightarrow 4$	
$\log(\kappa_{14})$	0.94	0.71
$\log(\alpha_{14})$	0.94	0.71
$\beta_{14,x}$	0.95	0.93
Susceptible: $2 \to 4$		
$\log(\kappa_{24})$	0.96	0.95
$\log(\alpha_{24})$	0.95	0.95
$\beta_{24,x}$	0.95	0.95
Susceptible: $2 \rightarrow 3$		
$\log(\kappa_{23})$	0.95	0
$\log(\alpha_{23})$	0.96	0.01
$\beta_{23,x}$	0.95	0.93
Susceptible: $3 \rightarrow 4$		
$\log(\kappa_{34})$	0.96	0.96
$\log(\alpha_{34})$	0.96	0.96
$\beta_{34,x}$	0.94	0.94

F Additional results from the analysis of stem cell transplantation data

	n	%
Acute GVHD, death within 100 days of transplantation	502	5.2
Acute GVHD, death 100 days post-transplantation	556	5.8
Acute GVHD, censored before 100 days	1	0.0
Acute GVHD, censored after 100 days post-transplantation	643	6.7
Death within 100 days without acute GVHD	983	10.2
Death after 100 days, without acute GVHD	1786	18.5
Censored after 100 days, without either	5159	53.5
Censored within 100 days, without either	21	0.2

Table F.1: Distribution of events for CIBMTR HCT data (n=9,651)

Table F.2: Estimates and standard errors for Weibull parameters from analyses of the CIBMTR HCT data (n=9,651) including: the proposed finite interval illness-death model with frailty (FI-ID^{γ}); the cure fraction illness-death model of Conlon et al. (2014) with an additional frailty (Conlon^{γ}); the illness-death model with frailty (ID^{γ}); and the cure fraction model for acute GVHD.

	FI-II	D^{γ}	$Conlon^{\gamma}$		$Conlon^{\gamma}$ ID^{γ}		Cu	ire
	Est	SE	Est	SE	Est	SE	Est	SE
Weibull parameter	s							
$\log(\kappa_{14})$	-7.98	0.18	-7.56	0.17				
$\log(\alpha_{14})$	0.14	0.03	0.08	0.03				
$\log(\kappa_{24})$	-16.16	0.03	-11.76	0.95	-7.43	0.17		
$\log(\alpha_{24})$	0.80	0.02	0.73	0.08	0.32	0.03		
$\log(\kappa_{23})$	-6.83	0.18	-6.58	0.18	-14.14	0.27	-5.93	0.13
$\log(\alpha_{23})$	0.66	0.02	0.62	0.03	1.05	0.02	0.48	0.02
$\log(\kappa_{34})$	-5.44	0.16	-5.33	0.15	-7.91	0.19		
$\log(\alpha_{34})$	-0.10	0.03	-0.13	0.03	0.07	0.03		



Figure F.1: Estimated baseline survivor functions from the CIBMTR HCT data for four transitions, based on: the proposed finite interval illness-death model with frailty (FI-ID^{γ}); the cure fraction illness-death model of Conlon et al. (2014) with an additional frailty (Conlon^{γ}); the illness-death model with frailty (ID^{γ}); the cure fraction model for acute GVHD; and the Cox model for acute GVHD. The marginal baseline survivor functions (with respect to the frailty, γ) are shown for models FI-ID^{γ} and Conlon^{γ} for the 'HCT, susc. \rightarrow Death' and 'HCT, susc. \rightarrow Acute GVHD' transitions.



Figure F.2: Estimated absolute risk profiles calculated for the first 150 post-transplantation corresponding to three patients (A, B, C) based on proposed finite-interval illness-death model with frailty (FI-ID $^{\gamma}$) in the first row, and the cure fraction illness-death model of Conlon et al. (2014) with an additional frailty (Conlon $^{\gamma}$) in the second row.



Figure F.3: For model: (a) proposed finite-interval illness-death model with frailty (FI-ID γ), and (b) Conlon et al.(2014) with an additional frailty (Conlon γ). Within each plot, the left panel presents the estimated absolute risk profiles (conditional on frailty, γ) for Patient C at 100 days post-transplantation for varying values of the gamma frailty, γ , and the right panel presents the estimated marginal absolute risk profiles for Patient C at 100 days post-transplantation.

G Sample R code

All statistical analyses were conducted in R [4]. In our implementation of the proposed finite interval illness-death model with both Weibull or truncated Weibull baseline hazard functions as well as B-spline parameterizations, separately, we make use of the optim function in the stats package for quasi-Newton non-linear optimization of the log-likelihood. To obtain starting values for the non-linear optimization, for the proposed model with Weibull and truncated Weibull baseline hazard, we utilized: the weibreg function in the eha package for univariate Weibull regression [5]; and FreqID_HReg function in the SemiCompRisks package [6] to fit an illness-death model with shared frailty. For the proposed model with B-spline parameterized baseline hazard functions, we used: the bSpline function in the splines2 package[7]; the coxph function in the survival package [8]; and the numdiff function in the pracma [9] package for numerical differentiation.

We provide R script and functions to run the proposed finite interval illness-death model with Weibull and truncated Weibull and B-spline baseline hazard parameterizations; please see files: sim-script-SemiCompRisksv3_1.R and sim-functions-SemiCompRisksv3_1.R.

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