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A systematic review of using and reporting survival analyses in acute lymphoblastic leukemia literature

Chatree Chai-Adisaksopha^{1,2}, Alfonso Iorio^{1,2*}, Christopher Hillis^{2,3}, Wendy Lim¹ and Mark Crowther^{1,2}

Abstract

Backgrounds: Survival analysis is commonly used to determine the treatment effect among acute lymphoblastic leukemia (ALL) patients who undergo allogeneic stem cell transplantation (allo-SCT) or other treatments. The aim of this study was to evaluate the use and reporting of survival analyses in these articles.

Methods: We performed a systematic review by searching the MEDLINE, EMBASE and Cochrane library databases from inception to April 2015. Clinical trials of patients with ALL comparing allo-SCT compared to another treatment were included. We included only studies that used survival analysis as a part of the statistical methods.

Results: There were 14 studies included in the review. Sample size estimation was described in 4 (29 %) studies. Only 4 (29 %) studies reported the list of covariates assessed in the Cox regression and 6 (43 %) studies provided a description of censorship. All studies reported survival curves using the Kaplan-Meier method. The comparisons between groups were investigated using the log-rank test and Wilcoxon test. Crossing survival curves were observed in 11(79 %) studies. The Cox regression model was incorporated in 10 (71 %) studies. None of the studies assessed the Cox proportional hazards assumption or goodness-of-fit.

Conclusions: The use and reporting of survival analysis in adult ALL patients undergoing allo-SCT have significant limitations. Notably, the finding of crossing survival curves was common and none of the studies assessed for the proportional hazards assumption. We encourage authors, reviewers and editors to improve the quality of the use and reporting of survival analysis in the hematology literature.

Keywords: Acute lymphoblastic leukemia, Mortality, Systematic review, Regression analysis

Background

Survival analysis measures the time from a defined starting point to the occurrence of an interested event where the risk changes over time. The goals of survival analysis serve three purposes: (1) to estimate survival and hazard functions from survival data, (2) to compare survival and hazard functions between groups and (3) to assess the relationship between predictor variables and survival time. The essential components for survival analysis include the time to event and the binary event outcome (success or failure). The probability of survival can be represented generating a Kaplan-Meier (KM) curve from survival data. Indeed, the KM plot is based on the estimate of the conditional probability of the time to failure [1] calculated at each time point recording an event. The difference in survival between two or more groups (or the treatment effect if treatment is what defines the two groups) can be commonly compared using the log-rank test [2].

The Cox proportional hazard (PH) model is a widely used regression method for survival data. The Cox PH model estimates the effect of predictor variables using the hazard function which does not require specifying a baseline hazard rate [3]. The measure of the effect, unadjusted or adjusted for covariates, is demonstrated as a hazard ratio (HR) which is expressed as an exponent of a regression coefficient in the model. An important



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^{*} Correspondence: iorioa@mcmaster.ca

¹Departments of Clinical Epidemiology and Biostatistics, McMaster University, Hamilton, Canada

²Departments of Medicine, McMaster University, Hamilton, Canada Full list of author information is available at the end of the article

property of the Cox PH model is that the PH assumption requires the hazard ratio to be constant over time [4]. Therefore, the Cox PH model is considered to be a semi-parametric model. Other regression models that can be used for survival analysis include an extended Cox PH model or parametric survival model (Weibull, exponential, log-logistic, lognormal, etc.) [5].

A recent systematic review demonstrated that survival analysis was incorporated in only 29 % of internal medicine articles [6]. However, there has been an increasing trend to using survival analysis in all categories of medical journals [6].

Allogeneic stem cell transplantation (allo-SCT) is the most potent post-remission therapy in adult acute lymphoblastic leukemia (ALL). The benefit of allo-SCT in adult ALL remains controversial [7]. Survival analysis is generally used to determine the treatment effect among ALL patients who undergo allo-SCT or other treatments, both in terms of prolongation and increased likelihood of survival. Allo-SCT is associated with high treatment-related mortality. Patients who tolerate the treatment are more likely to have a prolong event free survival and overall survival. On the other hand, non allo-SCT is less intensive treatment but may be associated with lower long-term event free survival. ALL literature were chosen because we expected that the use and report of survival analysis in such articles are complicated. To investigate whether the heterogeneity in study results is at least in part explained by a more or less appropriate use of time to event analysis, we conducted a systematic review of clinical trials which investigated the efficacy and safety of allo-SCT in adult patients with acute ALL. The aim of this study was to evaluate the use and reporting of survival analyses in these articles.

Methods

Data sources

We performed a systematic review by searching in the MEDLINE, EMBASE and The Cochrane library (The Cochrane Register of Controlled Trials and Cochrane Database of Systematic Reviews) databases. The reference lists were searched from the retrieved articles. The search terms were: Bone Marrow Transplantation OR Hematopoietic Stem Cell Transplantation OR Peripheral Blood Stem Cell Transplantation AND nonmyeloblat* OR non-myeloblat* OR Precursor Cell Lymphoblastic Leukemia-Lymphoma OR lymphoblast* OR lymphoid. AND (random* OR RCT OR control* OR trial). The database search was performed from inception to April 2015 with no language restrictions.

Selection criteria

The studies were included if they met the following criteria; were a clinical trial, controlled clinical trial or randomized control trial with allo-SCT compared to autologous SCT or non-transplantation therapy in patients with ALL in first complete remission. We only included studies that used survival analysis as one of the statistical methods.

Study selection and data extraction

Two investigators (CC and CH) independently identified articles using predefined inclusion criteria. Disagreements were resolved by consensus. Two investigators (CC and CH) independently extracted the data using a standardized data extraction from. Disagreements were again resolved by consensus.

We collected the following data: study design, outcome of interest (death, relapse), number of patients and number of events, survival curves estimate, regression method to estimate the hazard rate (Cox PH model or parametric survival model), methods for comparing the survival curves, the shape of the survival curves, variable selection, model building strategy, censoring description, length of follow-up, sample size calculation, test of interaction between variables, test for time dependent covariates, test for proportionality assumption and test for goodness-of-fit.

Analytic criteria

To evaluate the quality of reporting survival analyses, we used the following list of criteria for the proper use and description of the survival analyses.

- 1. *Sample size:* We evaluated the methods that the investigators described for sample size calculation. In addition, in the studies that used multiple regression analysis we evaluated the number of the events and number of covariates in order to estimate the adequacy of power. According to Peduzzi et al., approximately ten events per covariate is appropriate in PH regression analysis [8].
- 2. *Censoring description:* We evaluated the description of censoring and whether the investigators reported this adequately, inadequately or there was no mention.
- 3. *Survival curves:* We evaluated the statistical methods used for generating survival curves. For the comparison of survival between the groups, we documented the reported methods (log-rank test or Wilcoxon test). We also noted the shape of the survival curves (evenly separated or crossing survival curves).
- 4. *Statistical significance:* The statistical test used to evaluate the difference between two survival curves is determined using a log-rank test or weighted

log-rank test (e.g. Wilcoxon test). The null hypothesis of the test is that there is no difference between the two survival curves. We documented the statistical test reported in the articles.

- 5. *Regression model:* The statistical methods used for survival regression analysis were evaluated. We were interested in the regression model that the investigators used for calculating the hazard ratio (e.g. Cox PH model, extended Cox PH model or parametric survival model). We were also interested in the other regression models (time-dependent variable, competing risk analysis or repeated event analysis). In addition, the test for interaction of the variable was checked. For the studies that used multivariate regression, we assessed the description of variable selection and the strategy used for model building.
- 6. *Check for the PH assumption:* We assessed the test for PH assumption described in the articles. The assessment included methods used for checking PH assumption (graphical approach or the goodness-of-fit testing approach).
- 7. *Model checking:* We evaluated whether the investigators assessed for goodness-of-fit measures. The residual-based diagnostics were also assessed (martingale residuals, Cox-Snell residuals, Schoenfeld residuals or deviance residuals).

Results

Study characteristics

A total of 881 citations were identified by the systematic search strategy. Of these, 325 studies were duplicates. After screening of the titles and abstracts using predefined inclusion criteria, 541 studies were excluded. The reasons for exclusion are summarised in Fig. 1. Of these, we identified 15 potential studies for full-text review. Two studies were identified following manual review of the references. We excluded three studies due to no clinical trials comparing allo-SCT with other treatments. Thus, 14 studies [9–22] were included in our systematic review.

The study characteristics are summarised in Table 1. All of the studies were clinical trials. Patients were randomized to receive either allo-SCT or other treatments (autologous SCT or consolidation chemotherapy). Patients were allocated to undergo allo-SCT if the patient had a human leukocyte antigen (HLA) matched sibling donor, otherwise, the patient received autologous SCT or consolidation chemotherapy according to the study protocols. The median follow-up ranged from 59 to 114 months. The time-to-event outcomes in the included studies were overall survival and disease-free survival.

Analytic criteria

1. Sample size

Sample size estimation was described in 4 of 14 studies (Table 2). The proportion of events per total patients ranged from 36 to 78 %. With respect to the sample size and number of covariates assessed in the regression analysis, only four studies reported the list of covariate assessed in the Cox regression model. Of these, two studies obtained more than ten events-per-covariate (event-per-covariate 20.4 and 23.2, respectively) [11, 13]. However, the other two studies had an event-per-covariate 8.3 [20] and 3.8 [22].

2. Censoring description

There were six studies that provided the censoring description.



Study	Sample size		Median Follow-up	Outcome	Design	
	Allo-SCT	Non-allo-SCT	(months)			
Attal [9]	41	64	30	DFS	Randomized trial	
Bernasconi [10]	11	29	48	DFS	Clinical trial	
Cornelissen [11]	96	161	65	OS, DFS	Clinical trial	
De Witte [12]	30	33	60	OS, DFS	Clinical trial	
Fielding [13]	81	77	98	OS, DFS	Clinical trial	
Goldstone [14]	443	588	59	OS, DFS	Clinical trial	
Hunault [15]	41	106	61	OS, DFS	Randomized [¶] trial	
lfrah [16]	18	32	60	DFS	Clinical trial	
Labar [17]	68	116	114	OS, DFS	Clinical trial	
Ribera [18]	84	98	70	OS, DFS	Clinical trial	
Sebban [19]	116	141	62	OS, DFS	Clinical trial	
Thomas [21]	100	159	62	OS, DFS	Clinical trial	
Takeuchi [20]	34	108	63	OS, DFS	Clinical trial	
Ueda [22]	17	40	62	OS, DFS	Clinical trial	

Table 1 Study characteristics

Genetically randomization

Abbreviation: Allo-SCT allogeneic-stem cell transplantation, DFS disease free survival, OS overall survival

3. Survival curves

All studies reported survival curves using the KM method. The comparisons between the groups were investigated using a log-rank test in all studies (two studies used both log-rank test and Wilcoxon test). With regards to the shape of the survival curves, 11 studies reported crossing survival curves [9–15, 17–20] whereas one study reported unevenly separate survival curves [21] and one study reported evenly separated survival curves [22]. The overlapping survival curves were observed in five studies [9, 12, 14, 17, 20]. We were not able to compare survival curves in one study where the graphs were plotted in the separately [16].

4. Statistical significance

All of the studies reported the statistical test used to measure the difference between survival curves. Of these, five studies reported statistical significance for the treatment effect between groups. However, eight studies reported non-statistical significance (one study did not report).

5. Regression model

The regression model was incorporated in 10 of 14 studies. All ten studies used the Cox PH model [9, 11, 14–18, 20–22]. There was no parametric survival analysis used in the included studies. One study mentioned the test for interaction and competing risk analysis [11]. None of the studies described variable selection. Only one study mentioned the strategy used for model building [18].

6. Check for the PH assumption

In studies that used Cox PH model, PH assumption checking was not mentioned in any of the studies that used Cox PH model.

7. Model checking

The summary measures of the regression diagnostic and goodness-of-fit were not mentioned in any of the studies.

Discussion

Our study demonstrates that survival analyses have been used extensively in the landmark trials evaluating all-SCT in adult patients with acute ALL. However, the majority of the trials poorly reported their statistical methods and results. Sample size estimation and censoring description were not routinely described. Almost all the presented survival curves crossed. Moreover, the Cox assumption was not assessed even if the investigators used the Cox PH model. In addition, goodness-of-fit or regression residual analysis were lacking in all of the trials.

Regarding the sample size estimation, according to Consolidated Standards of Reporting Trials (CONSORT), it is important that the authors indicate how sample size was determined [23]. The intent of the sample size estimation is to ensure that a particular study has sufficient statistical power to detect a difference in the treatment effect between groups. Our review demonstrates that only 4 of 14 (29 %) trials described a sample size estimation. With respect to the regression analysis, only four studies provided a full list of covariates. Of these, only two studies

Study	Sample size estimation	Event/total	Censoring	Survival curve	Shape of KM-curve	Significance	Regression model	Variable selection	Goodness of fit or PH assumption test
Attal [9]	Yes	86/135 (relapse)	NR	KM method, log-rank test	Crossing, overlapping	S	Cox-PH model	Informal	NR
Bernasconi [10]	NR	NR	NR	KM method, log-rank test	Crossing	NS	N/A	N/A	NR
Cornelissen [11]	NR	102/257 (death), 113/257 (relapse)	Yes	KM method, log-rank test	Crossing	NS	Cox-PH model, interaction, competing risk	Informal	NR
De Witte [12]	NR	32/66 (death)	NR	KM method, log-rank test, Wilcoxon test	Crossing, overlapping	S	N/A	N/A	NR
Fielding [13]	NR	116/165 (death)	NR	KM method, log-rank test	Crossing	NS	N/A	N/A	NR
Goldstone [14]	NR	531/1031 (death)	Yes	KM method, log-rank test	Crossing, overlapping	NS	Cox-PH model	N/A	NR
Hunault [15]	Yes	67/147 (death)	NR	KM method, log-rank test	Crossing	S	Cox-PH model	Informal	NR
lfrah [16]	Yes	37/50 (relapse)	NR	KM method, log-rank test	Curves were plotted in two different graphs	N/A	Cox-PH model	Informal	NR
Labar [17]	NR	245/340 (death), 131/340 (relapse)	Yes	KM method, log-rank test	Crossing, overlapping	NS	Cox-PH model	Informal	NR
Ribera [18]	Yes	144/182 (death), 88/182 (relapse)	Yes	KM method, log-rank test	Crossing	NS	Cox-PH model	Informal	NR
Sebban [19]	NR	146/255 (death)	NR	KM method, log-rank test	Crossing	NS	N/A	N/A	NR
Thomas [21]	NR	177/259 (relapse)	NR	KM method, log-rank test	Unevenly separate	S	Cox-PH model	Informal	NR
Takeuchi [20]	NR	83/142 (death)	Yes	KM method, log-rank test	Crossing, overlapping	NS	Cox-PH model	Informal	NR
Ueda [22]	NR	38/57 (death)	Yes	KM method, log-rank test, Wilcoxon test	Evenly separate	S	Cox-PH model	Informal	NR

Table 2 Reporting survival analyses in the included studies

Abbreviation: PH proportional hazard, NR not reported, N/A not applicable; S, statistical significance; NS, non-statistical significance

appeared to be sufficiently powered (event-to-covariate ratio more than 10).

In survival analysis, patients who do not experience the relevant outcome over the study period, patients who are lost to follow-up during the study period and patients who withdraw from the study are censored. There are three assumptions regarding censorship in survival analysis: independent, random and noninformative [4]. Thus, the description of censorship is an important aspect to report in publication. However, only 6 of 14 (43 %) trials described their censoring. More importantly, if relapse is the outcome of interest in these studies, patients who die from any cause will be censored. In this circumstance, censoring may be considered informative because patients may die from disease progression or treatment-related causes. Consequently, the results may change based on different censoring descriptions. Providing a definition of censorship is a critical component to reporting these trials in the literature.

All of the studies utilized survival curves. Not surprisingly, crossing survival curves were found in 10 of 14 studies. Allo-SCT is considered the most potent post remission therapy in adult ALL [24]. In long-term follow-up studies, the patients who underwent allo-SCT had a lower relapse rate due to a graft-versusleukemia effect [11]. However, these patients had a higher early mortality rate from the toxicity of myeloablative chemotherapy when compared with patients who received autologous SCT or consolidation chemotherapy [11, 13]. Therefore, survival curves comparing these two treatments may be expected to cross at some point. Early death from treatmentrelated complications (commonly found in allo-SCT) and late death from relapsed disease (commonly found in autologous SCT) should be taken into the account in the treatment of ALL. Crossing survival curves make the interpretation of the treatment effects from the interventions much more complicated.

The log-rank test is the most common method used to compare the difference between survival curves based on the chi-square test [25]. It is important to note that the log-rank test may be invalid if the survival curves cross because of an increase of the probability of type II error. Moreover, the log-rank test may lose power in the circumstance of crossing survival curves [26]. Our study reveals that, among ten analyses with crossed survival curves, eight were non-statistically significant and two were statistically significant. We found that five studies had overlapping survival curves that might be explainable for insignificant findings of the interventions. It was difficult to make a conclusion on the rest of the studies based on the log-rank test of crossing over survival curves. Strategies have been proposed to overcome the limitation of the log-rank test when the survival curves cross. The authors may consider analysing the survival curves at a fixed point in time [27]. Another alternative includes using a weighted log-rank (Harrington-Fleming) test which gives more weight to the later events [28]. Other weighted log-rank tests that may be useful are the methods developed by Gill et al. or Pepe and Fleming [29, 30]. Li et al. recently published a simulation study which investigated several statistical methods in the situation of crossing survival curves. This study showed that adaptive Neyman's smooth tests and the two-stage procedure provided greater stability and higher power as compared to the other methods [29].

Relapse disease and death are the most common outcomes in the ALL literature. Conventional KM method and Cox proportional hazard model convey no information regarding possible competing risks. Competing risk is an event that modifies the chance of the interested outcome [31]. For example, death from any cause is a competing risk for relapse disease. Using the competing risk analysis is therefore considered to be more appropriate in the treatment with high rate of complications. We observed only one study that used competing risk analysis [11]. We encouraged investigators to incorporate competing risk analysis, at least in the sensitivity analysis.

We found that the Cox PH model was commonly used in the collection of articles in our review. There was substantial inadequacy of the description of variable selection, the strategy used for fitting procedure and test for goodness-of-fit. As mentioned above, sample size estimation related to regression analysis was noted in only four studies. Of these, two studies were found to be underpowered based on low event-per-covariate ratio [8]. We strongly encourage authors to describe the process of variable selection, strategy of model building and provide evidence that the sample size is sufficient for regression analysis.

A lack of PH assumption checking may introduce bias to the regression analysis. Our review shows that none of the studies described an assessment of the PH assumption. The Cox PH model assumes that the hazard ratio for comparing any two groups of predictor variables is constant over time [4]. If this assumption is not met, the Cox PH model is not valid for the analysis. We observed that 11 of 14 (79 %) studies had crossing survival curves. A clear violation of the PH assumption occurs if survival curves cross [32, 33]. Therefore, a hazard ratio should not be used to compare the treatment effect between groups. We suggest that authors check for the PH assumption if the Cox PH model is incorporated in the analysis. When the PH assumption is violated, authors may consider using an alternative regression analysis, such as the extended Cox PH model or parametric survival analysis (Weibull, exponential, log-logistic or lognormal model).

Conclusions

Our systematic review evaluating reporting methods for survival analysis in adult ALL patients undergoing allo-SCT show significant shortcomings in the use and reporting of survival analysis. Sample size estimation was not routinely described and studies are frequently statistically underpowered. There was a lack of censoring description. Most notably, crossing survival curves were common and none of the studies checked for the PH assumption. Finally, the description of variable selection, fitting procedure and model checking were neglected.

Survival analysis has been used increasingly in medical research studies [6]. We raise awareness of these limitations and encourage authors, reviewers and editors to improve the quality of the use and reporting survival analysis in the literature.

Abbreviations

ALL, acute lymphoblastic leukemia; CONSORT, Consolidated Standards of Reporting Trials; HLA, human leukocyte antigen; HR, hazard ratio; KM, Kaplan-Meier; PH, proportional hazard; SCT, stem cell transplantation

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Availability of data and materials

Data will not be shared.

Authors' contributions

CC, AL, MC participated in the design of the study, CC, CH performed study selection, data abstraction, CC performed data analysis, CC drafted the manuscript, CH, WL, AI, MC read and approved the final manuscript.

Authors' information

Not applicable.

Competing interests

MC discloses having sat on advisory boards for Janssen, Leo Pharma, Portola, and AKP America. MC holds a Career Investigator award from the Heart and Stroke Foundation of Ontario, and the Leo Pharma Chair in Thromboembolism Research at McMaster University. MC's institution has received funding for research projects from Leo Pharma. MC has received funding for presentations from Leo Pharma, Bayer, Celgene, Shire and CSL Behring. CC, CH, AI and WL have no relevant conflicts of interest.

Consent for publication

Not applicable.

Ethics approval and consent to participate

Not applicable.

Author details

¹Departments of Clinical Epidemiology and Biostatistics, McMaster University, Hamilton, Canada. ²Departments of Medicine, McMaster University, Hamilton, Canada. ³Departments of Oncology, McMaster University, Hamilton, Canada. Received: 20 September 2015 Accepted: 21 May 2016 Published online: 08 June 2016

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