

The World's Largest Open Access Agricultural & Applied Economics Digital Library

# This document is discoverable and free to researchers across the globe due to the work of AgEcon Search.

# Help ensure our sustainability.

Give to AgEcon Search

AgEcon Search
<a href="http://ageconsearch.umn.edu">http://ageconsearch.umn.edu</a>
<a href="mailto:aesearch@umn.edu">aesearch@umn.edu</a>

Papers downloaded from **AgEcon Search** may be used for non-commercial purposes and personal study only. No other use, including posting to another Internet site, is permitted without permission from the copyright owner (not AgEcon Search), or as allowed under the provisions of Fair Use, U.S. Copyright Act, Title 17 U.S.C.

No endorsement of AgEcon Search or its fundraising activities by the author(s) of the following work or their employer(s) is intended or implied.



DOI: 10.1177/1536867X221141057

## Estimating the risk of events with stprisk

Matteo Bottai
Division of Biostatistics
Institute of Environmental Medicine
Karolinska Institutet
Stockholm, Sweden
matteo.bottai@ki.se

Abstract. Incidence rates are popular summary measures of the occurrence over time of events of interest. They are also called mortality rates or failure rates, depending on the context. The incidence rate is defined as the ratio between the total number of events and total follow-up time and can be estimated with the strate command. When the event of interest can occur multiple times on any given subject over a time period, like infections, the incidence rate represents an average count per unit of time, such as the average number of infections per year. When the event of interest can occur only once, such as death, an alternative summary measure is the risk, or probability, of occurrence per unit time, such as the risk of dying in one year. In this article, I present the stprisk command, which estimates risks, and illustrate its use and interpretation through a data example.

**Keywords:** st0698, stprisk, incidence rates, mortality rates, survival analysis

#### 1 Introduction

Scientific research often entails analyzing the occurrence over time of events of interest, such as death or a cancer diagnosis. This type of analysis is generally known as survival analysis, although different terms are used across different research areas. Stata has a comprehensive suite of commands for the analysis of survival data, and the help documentation offers an excellent description of the relevant methods and available commands (help st).

The strate command can estimate incidence rates, which are popular summary measures of the occurrence over time of events of interest. The incidence rate is defined as the ratio between the total number of events and total follow-up time. When the event of interest can occur multiple times on any given subject over a time period, like infections, the incidence rate represents an average count per unit of time, such as average number of infections per year. When the event of interest can occur only once, such as death, an alternative summary measure is the risk, or probability, of occurrence per unit of time, such as the risk of dying in one year.

The following section introduces the new stprisk command, which estimates the risk of occurrence of events of interest over time, and its relation to the popular incidence rate through a data example; section 3 shows the syntax of stprisk; section 4 provides some technical details about risks and the implementation of the stprisk command; and section 5 contains some final remarks.

#### 2 Incidence rates and incidence risks

I illustrate the basis of the new stprisk command with an example. I use the data from a fictitious clinical trial on survival in cancer patients available in Stata.

```
. sysuse cancer
(Patient survival in drug trial)
```

I summarize the content of the dataset with the describe command.

#### . describe

Contains data from /usr/local/stata17/ado/base/c/cancer.dta

Observations: 48 Patient survival in drug trial
Variables: 8 3 Mar 2020 16:09

Variable name	Storage type	Display format	Value label	Variable label		
studytime	byte	%8.0g		Months to death or end of exp.		
died	byte	%8.0g	diedlbl	Patient died		
drug	byte	%8.0g	type	Drug type		
age	byte	%8.0g	• •	Patient's age at start of exp.		
_st	byte	%8.0g		1 if record is to be used; 0 otherwise		
_d	byte	%8.0g		1 if failure; 0 if censored		
_ _t	byte	%10.0g		Analysis time when record ends		
_t0	byte	%10.0g		Analysis time when record begins		

Sorted by:

The interest of this study is in comparing survival with the three different treatment groups, which comprise two active drugs and a placebo. To help present the arguments contained in the remainder of this section, I set the unit of measurement of the time-to-death variable to be years with the stset command.

Time for analysis: time/12

48 total observations
0 exclusions

- 48 observations remaining, representing
- 31 failures in single-record/single-failure data
- 62 total analysis time at risk and under observation

At risk from t = 0
Earliest observed entry t = 0
Last observed exit t = 3.25

M. Bottai 971

Henceforth, I refer to the incidence rate as mortality rate, considering the event of interest is death. I estimate the mortality rate by treatment group with the **strate** command.

. strate drug

Failure \_d: died
Analysis time \_t: studytime/12

Estimated failure rates Number of records = 48

drug	D	Y	Rate	Lower	Upper
Placebo	19	15.0000	1.266667	0.807948	1.985827
Other	6	17.4167	0.344498	0.154769	0.766810
NA	6	29.5833	0.202817	0.091118	0.451446

Notes: Rate = D/Y = failures/person-time.

Lower and Upper are bounds of 95% confidence intervals.

From the above output, the mortality rate in the placebo group is 1.27. On its website, the Centers for Disease Control and Prevention defines the incidence rate as "a measure of the frequency with which new cases of illness, injury, or other health condition occur, expressed explicitly per a time frame [...]." A rate is not a risk. If one interpreted the above rate naïvely as a risk, one might conclude that any given patient is expected to die 1.27 times every year or, alternatively, that 100 patients are expected to report 127 deaths every year.

I now estimate the mortality risk with the stprisk command. Section 4 gives more details on its definition and interpretation.

#### . stprisk drug

Incidence risk

drug	Risk	[ 95% Conf. Int	. ]
Placebo	0.830235	0.578028 0.973	608
Other	0.381161	0.163559 0.724	
NA	0.206644	0.090824 0.430	

The interpretation of the above risks is simple. For example, in the placebo group, the probability for any given subject to die in a year is 0.83, or alternatively, we expect 83 deaths out of 100 subjects every year.

The mortality rate converges to the mortality risk as the latter tends to zero. This limit behavior is analogous to that of the mean of a Poisson distribution, which converges to the mean of the binomial distribution as the latter tends to zero. This explains why the mortality rate and the mortality risk are numerically closer in drug group 3 than they are in the placebo group 1.

## 3 The stprisk command

The syntax of the stprisk command, similar to that of strate, is

$$\mathtt{stprisk} \ \left[ \ \mathit{varlist} \ \right] \ \left[ \ \mathit{if} \ \right] \ \left[ \ \mathit{in} \ \right] \ \left[ \ \mathsf{,} \ \ \mathsf{level(\#)} \ \ \underline{\mathsf{graph}} \ \ \underline{\mathsf{now}} \\ \mathsf{hisker} \ \right]$$

stprisk tabulates rates by one or more categorical variables declared in *varlist*. When *varlist* is omitted, stprisk estimates the mortality risk for the entire dataset. The level() option specifies the level of the confidence intervals. The default is level(95) or as set by set level. The graph option plots rates against the groups defined by *varlist* when a *varlist* is specified. The nowhisker option omits the confidence intervals from the graph.

You must stset your data before using stprisk; see [ST] stset.

#### 4 Incidence risks

This section provides the basic definition and interpretation of incidence risks. Its content consists of slightly edited excerpts from published articles (Bottai 2017; Discacciati and Bottai 2017; Lagergren, Bottai, and Santoni 2021; and Bottai, Discacciati, and Santoni 2021). Let T represent a continuous time-to-event variable with support on the positive real half-line. Let  $S(t) = \Pr(T > t)$  and  $H(t) = -\log\{S(t)\}$  indicate the survival function and the cumulative hazard function, respectively. The function S(t) is defined over the entire real line,  $\mathbb{R}$ , while H(t) is defined over the set  $\{t \in \mathbb{R} : S(t) > 0\}$ .

The probability of occurrence of an event over the time interval  $[t_0, t_1]$ , with  $t_0 < t_1$ , conditional on  $T > t_0$ , is

$$\Pr(T \le t_1 \mid T > t_0) = 1 - S(t_1)/S(t_0)$$

defined over the set  $\{t \in \mathbb{R} : S(t) > 0\}$ . Bottai (2017) defined the geometric rate of the event over the interval  $[t_0, t_1]$ , conditional on  $T > t_0$ , as

$$G(t_0, t_1) = 1 - \left\{ S(t_1) / S(t_0) \right\}^{1/(t_1 - t_0)} \tag{1}$$

Bottai, Discacciati, and Santoni (2021) later referred to the above as the average probability of occurrence of the event. As explained in the articles, the word average indicates the geometric mean. The following example may help interpret  $G(t_0, t_1)$ . Suppose the event of interest is death and  $t_0 = 0$  and  $t_1 = 3$ . We split the time interval [0,3] into the three one-unit disjoint intervals [0,1], [1,2], and [2,3]. The mortality rate is the complement of the probability of surviving all three intervals conditional on being alive at time 0, which is S(3)/S(0). This is algebraically equal to the product of the probabilities of surviving each interval conditional on being alive at its start,  $S(3)/S(0) = \{S(1)/S(0)\}\{S(2)/S(1)\}\{S(3)/S(2)\}$ . The average probability per unit of time, therefore, is the geometric mean of the probabilities in the three intervals,  $\{S(3)/S(0)\}^{1/3}$ . Applying the geometric mean at each interval yields the probability of surviving the entire period [0,3].

M. Bottai 973

For example, the probability of surviving one year in the placebo group is 0.225, as evinced in the following output:

. sts list if drug==1, at(1)
 Failure \_d: died
Analysis time \_t: studytime/12
Kaplan--Meier survivor function

 Time	Beg. total	Fail	Survivor function	Std. error	[95% con	f. int.]
0	0	0	1.0000			
1	6	15	0.2250	0.0971	0.0721	0.4290
2	1	4				

Note: Survivor function is calculated over full data and evaluated at indicated times; it is not calculated from aggregates shown at left.

From (1), the monthly mortality risk over the first year is  $G(0, 12) = 1 - 0.225^{1/12} = 0.117$ .

The definition of incidence risk given in (1) can also be written as

$$G(t_0, t_1) = 1 - \exp\left\{\frac{H(t_0) - H(t_1)}{t_1 - t_0}\right\}$$
(2)

The stprisk command uses (2), with  $t_0$  set equal to the start of the follow-up time and  $t_1$  set equal to the largest observed time. The cumulative hazard  $H(t_0)$  is equal to zero. The cumulative hazard  $H(t_1)$  and its confidence interval are obtained from the sts list command with the Nelson-Aalen option; see [ST] sts list.

#### 5 Conclusions

The new stprisk command provides nonparametric estimates and confidence intervals for the risk of occurrence of an event of interest over time. The command is computationally efficient, and its syntax is patterned on that of the strate command.

The risk of occurrence of events is applicable to any event that can occur only once. In the example given in section 2, the event of interest is death, but risks can be assessed for other once-only events, such as first cancer diagnosis, hospital discharge, and first employment.

As the time period  $(t_0, t_1)$  tends to zero, the risk  $G(t_0, t_1)$  defined in (1) tends to the instantaneous risk, and the incidence rate tends to the instantaneous incidence rate, which is also known as the hazard. The similarities and differences between these quantities are expounded in the article by Bottai, Discacciati, and Santoni (2021).

## 6 Programs and supplemental materials

To install a snapshot of the corresponding software files as they existed at the time of publication of this article, type

```
net sj 22-4
net install st0698 (to install program files, if available)
net get st0698 (to install ancillary files, if available)
```

#### 7 References

Bottai, M. 2017. A regression method for modelling geometric rates. Statistical Methods in Medical Research 26: 2700–2707. https://doi.org/10.1177/0962280215606474.

Bottai, M., A. Discacciati, and G. Santoni. 2021. Modeling the probability of occurrence of events. Statistical Methods in Medical Research 30: 1976–1987. https://doi.org/10.1177/09622802211022403.

Discacciati, A., and M. Bottai. 2017. Instantaneous geometric rates via generalized linear models. Stata Journal 17: 358–371. https://doi.org/10.1177/1536867X1701700207.

Lagergren, J., M. Bottai, and G. Santoni. 2021. Patient age and survival after surgery for esophageal cancer. *Annals of Surgical Oncology* 28: 159–166. https://doi.org/10.1245/s10434-020-08653-w.

#### About the author

Matteo Bottai is a professor of biostatistics at the Division of Biostatistics, Institute of Environmental Medicine, Karolinska Institutet, Stockholm, Sweden.