### Introduction to survival analysis

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O jorgetendeiro/Seminar-2020-Survival-Analysis

## Plan for today

Gentle introduction to survival analysis.

Source:

Harrell, F. E., Jr. (2015). Regression Modeling strategies, 2nd edition.

Springer

Chapters:

17, 18, and 20.

### Survival analysis (SA)

#### Data:

For which the time until the event is of interest.

► This goes beyond *logistic regression*, which focuses on the *occurrence* of the event.

#### Outcome variable:

- ightharpoonup T = Time until the event.
- ▶ Often referred to as *failure time*, *survival time*, or *event time*.

### **Examples**

Survival time: Time until...

▶ death, desease, relapse.

Failure time: Time until...

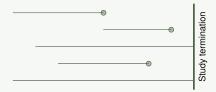
▶ product malfunction.

*Event time:* Time until...

▶ graduation, marriage, divorce.

### Advantages of SA over typical regression models

► SA allows modeling units that did not fail up to data collection (*censored on the right* data).



- Regression could be considered to model the expected survival time. But:
  - $\checkmark$  Survival time is often not normally distributed.
  - ✓ P(survival > t) is often more interesting than  $\mathbb{E}(\text{survival time})$ .

### Censoring

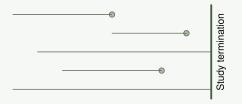
- ▶ For some subjects, the event did not occur up to the end of data collection.
- ► These data are right-censored.

Define random variables for the ith subject:

- $ightharpoonup T_i = \text{time to event}$
- $ightharpoonup C_i = censoring time$
- ▶  $e_i$  = event indicator =  $\begin{cases} 1 & \text{if event is observed } (T_i \leq C_i) \\ 0 & \text{if event is not observed } (T_i > C_i) \end{cases}$
- $ightharpoonup Y_i = \min(T_i, C_i) = \text{what occurred first (failure or censoring)}$

Variables  $\{Y_i, e_i\}$  include all the necessary information.

### Typical data set



$T_i$	$C_i$	$Y_i$	$e_i$
5	10	5	1
4	12	4	1
13+	13	13	0
5	10	5	1
15+	15	15	0

#### Observe the flexibility of SA data:

- ► Subjects may join the study at different moments.
- ► Censoring times may differ among subjects.

 $\{Y_i, e_i\}$  does include all the necessary information.

### Three main functions

Recall that the outcome variable is T = time until event.

► Survival function:

$$S(t) = P(T > t) = 1 - F(t),$$

where  $F = P(T \le t)$  is distribution function of T.

► Cumulative hazard function:

$$\Lambda(t) = -\log(S(t))$$

► Hazard function:

$$\lambda(t) = \Lambda'(t)$$

### Survival function

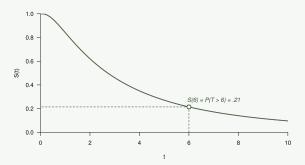
$$S(t) = P(T > t) = 1 - F(t)$$

#### Example:

If event = death, then S(t) = prob. that death occurs after time t.

### Properties:

- ►  $S(0) = 1, S(\infty) = 0.$
- ► Non-increasing function of *t*.



### Cumulative hazard function

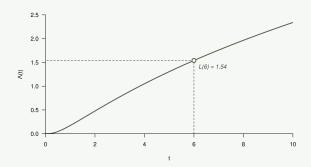
$$\Lambda(t) = -\log(S(t))$$

Idea:

Accumulated risk up until time t.

#### Properties:

- $\Lambda(0) = 0.$
- ► Non-decreasing function of *t*.

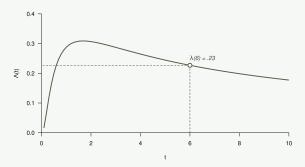


# Hazard function

$$\lambda(t) = \Lambda'(t)$$

Idea:

Instantaneous event rate at time t.



### Relation between the three functions

### All functions are related:

Any two functions can be derived from the third function.

▶ The three functions are equivalent ways of describing the same random variable (T = time until event).

More generally, all the following functions give mathematically equivalent specifications of the distribution of *T*:

- $\triangleright$  F(t): Distribution function
- $\blacktriangleright$  f(t): Density function
- $\triangleright$  S(t): Survival function
- $\blacktriangleright$   $\lambda(t)$ : Hazard function
- $\blacktriangleright$   $\Lambda(t)$ : Cumulative hazard function.

### **Examples**

Next are two primary examples of parametric survival distributions:

- ▶ the exponential distribution;
- ▶ the Weibull distribution.

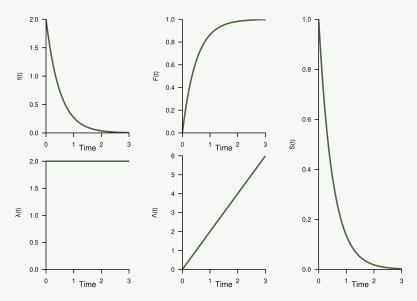
These models (still) include no covariates, thus:

► Each subject in the sample is assumed to have the same distribution of *T*.

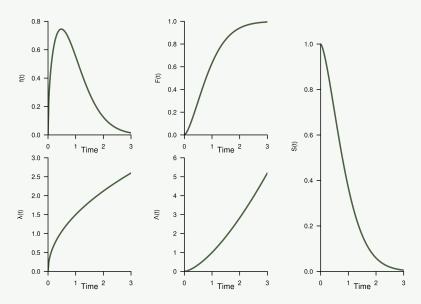
No formulas.

Instead: Let's plot.

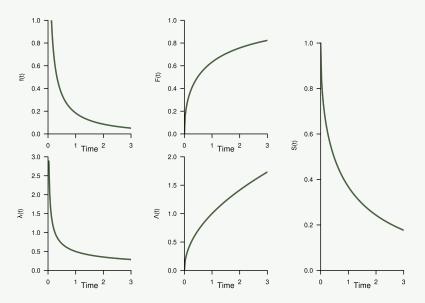
## Exponential survival distribution



### Weibull survival distribution (I)



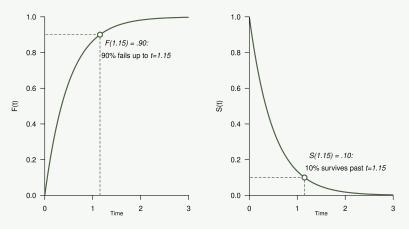
### Weibull survival distribution (II)



### **Quantiles**

Q: What is the time by which (100q)% of the population will fail?

*A*: Value  $t_q$  such that  $F(t_q) = q$ , or, equiv.,  $S(t_q) = 1 - q$ .



In particular, median survival time =  $t_{.50}$ .

### Expected failure time

(Note: *T* is skewed, so the mean is not the best summary. Better use median.)

Q: What is the expected failure time?

*A*: It is the area under the survival function.

