

# *Introduction to survival analysis*

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🔗 [jorgetendeiro/Seminar-2020-Survival-Analysis](https://github.com/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:*

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

# Examples

*Survival time:* Time until...

- ▶ death, disease, 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:*
  - ✓ Survival time is often not normally distributed.
  - ✓  $P(\text{survival} > t)$  is often more interesting than  $\mathbb{E}(\text{survival time})$ .

# *Censoring*

We focus on

## Three main functions

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

- Survival function:

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

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

- Cumulative hazard function:

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

- Hazard function:

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

# Survival function

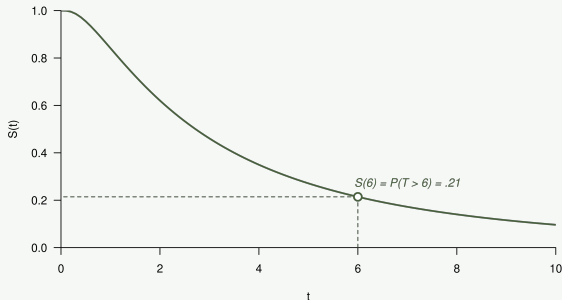
$$S(t) = P(T > t) = 1 - F(t)$$

*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

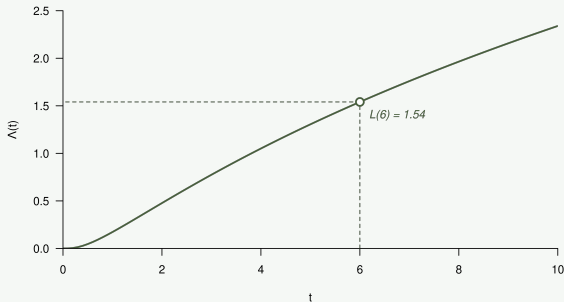
$$\Lambda(t) = -\log(S(t))$$

*Idea:*

Accumulated risk up until time  $t$ .

*Properties:*

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

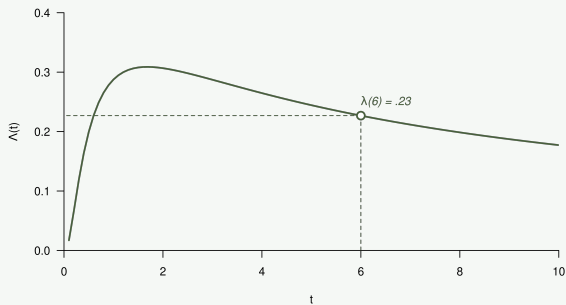


# Hazard function

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

*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  $T$ :

- ▶  $F(t)$ : Distribution function
- ▶  $f(t)$ : Density function
- ▶  $S(t)$ : Survival function
- ▶  $\lambda(t)$ : Hazard function
- ▶  $\Lambda(t)$ : 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

