

#### **FACULTY OF HEALTH SCIENCES - SCHOOL OF MEDICINE**

**MSc Health Statistics and Data Analytics** 

## Survival Analysis

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### Learning objectives

Understand what survival data is and how to handle censoring

Prepare survival curves using the life table and Kaplan-Meier methods

Estimate median survival times and survival rates at specified times

 Compare survival curves between two or more groups using the logrank test

# Survival Analysis

Also called "time to event analysis"

**Survival Time** is defined as the **time** starting from an already defined point to the occurrence of the event of interest (survival time).

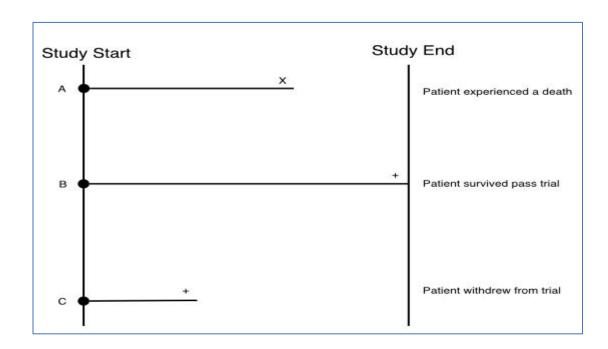
#### Examples:

- time to death
- time to cancer metastasis
- time to failure of a light bulb
- time to complete a PhD



- Survival time is almost never normally distributed.
- In many studies, some of the times to an event are not recorded (for various reasons) — These observations are called censored, and the corresponding phenomenon is called censoring.
- Thus, such data cannot be handled by already known parametric or non-parametric methods.

- **Censoring** is present when there is incomplete information about a subject's event time, but we don't know the exact event time.
- The most common case is having right censoring



- ✓ A person does not experience the event before the study ends
- ✓ A person is lost to follow-up during the study period or withdraws from the study

#### Survival Function *S(t)*=pr(T > t)

• Is the probability that a subject survives longer than time t.

Example: If t=100 years, S(t=100) = probability of surviving beyond 100 years.

• The graphical representation of S(t) is known as survival curve (Kaplan-Meier Curve).

#### **Assumptions**

- Observations should be independent
- Groups (if there are any) should be independent
- Accurate time calculation
- In the case of two or more groups, curves should not intersect with each other.



# Example for the construction of a Kaplan-Meier Curve

#### Example:

Survival analysis for patients with different types of cancer

Group A: Astrocytoma (n=20)

6,13,21,30, 31+,37,38, 47+,49,50,63,79,80+,82+,82+,86,98,149+,202,219 weeks

Group B: Glioblastoma (n=20)

10,10,12,15,16,20,24,25,28,30,34+,35,37,40+,48,70+,91,112,181,220 weeks

+ censored data

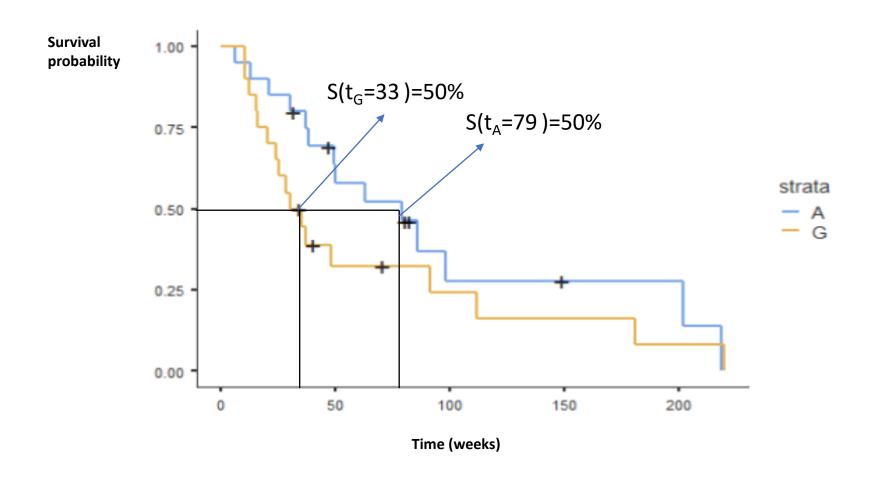


## Life table-Group A

Time (weeks) i	Live at start n <sub>i</sub>	Died d <sub>i</sub>	Cen sor ed	At risk	Probability of surviving (n <sub>i</sub> -d <sub>i</sub> )/n <sub>i</sub>	Cumulative Probability $S(t_i)=((n_i-d_i)/n_i)^*S(t_{i-1})$
1	20	О	О	20	1=(20-0)/20	1
6	20	1	О	19	0.950=(20-1)/20	0.950
13	19	1	0	18	0.947=(19-1)/19	0.947*0.950=0.90
21	18	1	0	17	0.944=(18-1)/18	0.944*0.90=0.85
30	17	1	0	16	0.941=(17-1)/17	0.941*0.85=0.80
31	16	О	1	16	1=(16-0)/16	1*0.80=0.80
37	15	1	О	14	0.933=(15-1)/15	0.933*0.80=0.75
••••						



## Comparison of two survival curves





# Log-Rank Test

We use log-rank test to compare survival curves

This is a non parametric test

 $H_0$ : There is no difference between the curves

H<sub>1</sub>: There is a difference between the curves

Log-rank test gives p=0.203>0.05 thus there is no significant difference between the two survival curves.

