# On some practical issues in the analysis of survival data

On behalf of STRATOS TG8

Maja Pohar Perme

IBMI, University of Ljubljana

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### The plan of this talk

### TG8 members

- Michal Abrahamowicz
- Per Kragh Andersen
- Richard Cook
- Pierre Joly

Torben Martinussen

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- Maja Pohar Perme
- Jeremy Taylor
- Terry Therneau

#### Outline

- a few well known facts of survival analysis
- an outline of TG8 plan
- competing risks

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### Survival analysis

#### Data evolving in time

#### Target population and censoring

- Inference: parameters in the population, estimated on a sample
- Survival analysis: parameters in the population complete data (S(t), h(t))
- Sample: censored data (incomplete data)
- The goal: drawing inference for population parameters based on incomplete data.
- Assumption: independent censoring

### Independent censoring

- individuals censored at any given time *t* should not be a biased subsample of those who are at risk at time *t*.
- the extra information that the subject is not only alive, but also uncensored at time *t* does not change the hazard:

$$h(t) \approx \frac{P(T^* \le t + dt|T^* > t)}{dt} = \frac{P(T^* \le t + dt|T^* > t, C > t)}{dt}$$

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# Inference with independent censoring

 For independent observations (*T<sub>i</sub>*, *δ<sub>i</sub>*), where *T<sub>i</sub>* = min(*T<sub>i</sub>*<sup>\*</sup>, *C<sub>i</sub>*) and *δ<sub>i</sub>* = *I*(*T<sub>i</sub>*<sup>\*</sup> < *C<sub>i</sub>*), the likelihood can be expressed via the hazard (and cumulative hazard *H*) functions:

$$L(\theta) = \prod_i h_{\theta}(T_i)^{\delta_i} e^{-H_{\theta}(T_i)}$$

- This is the basis for the Nelson-Aalen estimator for *H* and Cox partial likelihood
- Using the relations between S and H leads to Kaplan-Meier estimator



### TG8 plan

### Level 2 papers

- Single endpoint
- Multiple endpoints

### Level 2 papers

- Avoiding pitfalls
- Checking assumptions
- Using state-of-the-art methods



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# TG8 plan

### Single endpoint

- The censoring assumption
- Cox model check PH, functional form
- time-varying covariates, time-dependent coefficients
- Alternatives to Cox models (AFT, cure models ...)

Patients on chemotherapy. Do treatment side effects improve the prognosis?

- Time 0: start of chemotherapy. With time, some patients develop side effects.
- Available data: patients followed for 5 years, some developed side effects, some did not (0/1 variable)



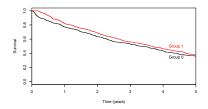
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# Immortal time bias

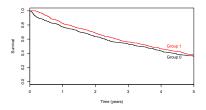
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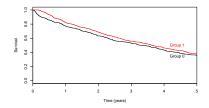




A time-varying covariate

#### Solution?

A hazard regression model with a time-varying covariate



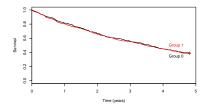


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A time-varying covariate

#### Solution?

- A hazard regression model with a time-varying covariate
- Conditional survival



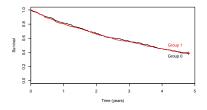


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A time-varying covariate

#### Solution?

- A hazard regression model with a time-varying covariate
- Conditional survival



#### It takes time to measure time ...

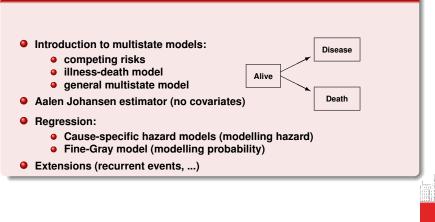
... and things can happen in between.



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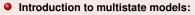
# TG8 plan

### Multiple endpoints

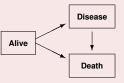


# TG8 plan

### Multiple endpoints



- competing risks
- illness-death model
- general multistate model
- Aalen Johansen estimator (no covariates)
- Regression:
  - Cause-specific hazard models (modelling hazard)
  - Fine-Gray model (modelling probability)
- Extensions (recurrent events, ...)

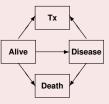


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# TG8 plan

### Multiple endpoints

- Introduction to multistate models:
  - competing risks
  - illness-death model
  - general multistate model
- Aalen Johansen estimator (no covariates)
- Regression:
  - Cause-specific hazard models (modelling hazard)
  - Fine-Gray model (modelling probability)
- Extensions (recurrent events, ...)



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# Competing risks, multistate models

#### Same ideas, but care needed

- Everything can be defined via hazard functions
- Cox model can still be used for modelling hazard functions
- No one-to-one relationship between hazard and survival
- More difficult to state what is of interest

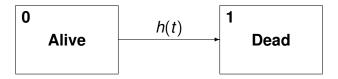
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# Competing risks example

### Survival of patients on dialysis

- Time 0: kidney failure, start of dialysis
- Events: death (the event of interest), kidney transplant
- Not everyone experiences the event of interest in the complete data
- Transplants are a competing risk, patients are not randomly transplanted
- Patients with best prognosis get transplanted
- If these patients are considered as censored, survival gets underestimated

# Single endpoint



#### In the population (complete observation):

- Every one ends up in state 1
- The probability of being in state 1 by time *t* is given uniquely from the hazard:

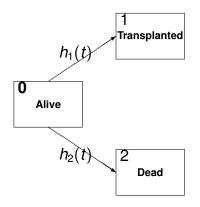
$$F(t) = 1 - S(t) = 1 - \exp(-H(t))$$

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Introd	uction

# Competing risks



- Several events can happen
- Even in the population, not every one experiences the same event
- Even if only one event is of interest, one cannot see others as 'independent censoring'

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# Relationship between rates and risks

Both rates are needed to compute one risk

Cause-specific hazards j = 1,2:

$$h_j(t) \approx \frac{P(\text{state } j \text{ by time } t + dt \mid \text{state } 0 \text{ time } t)}{dt}$$

Overall survival function:

 $S(t) = P(alive at time t) = e^{-[H_1(t) + H_2(t)]}$ 

Probability of experiencing event j at time u

 $P(\text{state } j \text{ at time } u) \approx S(u-)h_i(u)du$ 

Cumulative incidence function for event j

$$F_j(t) = P(\text{state } j \text{ by time } t) = \int_0^t S(u-)h_j(u)du$$

### Inference

#### The likelihood function

For independent observations (*T<sub>i</sub>*, *δ<sub>i</sub>* · *D<sub>i</sub>*), where *D<sub>i</sub>* = final state, *T<sub>i</sub>* = min(*T<sub>i</sub>*<sup>\*</sup>, *C<sub>i</sub>*) *δ<sub>i</sub>* = *I*(*T<sub>i</sub>*<sup>\*</sup> < *C<sub>i</sub>*), the likelihood may again be expressed via the hazard (and cumulative hazard *H*) functions:

$$L(\theta) = \prod_{i} h_{1\theta}(T_i)^{\delta_i l(D_i=1)} h_{2\theta}(T_i)^{\delta_i l(D_i=2)} e^{-H_{1\theta}(T_i) - H_{2\theta}(T_i)}$$

This likelihood can be factorized as:

$$L(\theta) = \prod_{i} h_{1\theta}(T_i)^{\delta_i l(D_i=1)} e^{-H_{1\theta}(T_i)} \prod_{i} h_{2\theta}(T_i)^{\delta_i l(D_i=2)} e^{-H_{2\theta}(T_i)}$$



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### Inference

### Consequence of the factorization

- If the model uses different parameters for different hazards hazard regression analysis can be performed by censoring the other cause
- Hazards can be modeled by censoring the other risk, for probabilities, both are needed
- Distinction between hazard rate and probability of an event (equal in single event case)



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# Back to dialysis example

#### What is the interest of the analysis?

- How many patients are still on dialysis after 5 years: overall survival
- Probability of dying in 5 years: cumulative incidence function
- Is one type of dialysis (PD, HD) safer than the other: model hazards

### In general

- To describe the fraction ending in a state: cumulative incidence function
- To understand the mechanisms by which subjects may fail: hazards
- Both can be useful for a complete description of the competing risks situation

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# Concluding remarks

### The TG8 plan

- Single endpoint and multistate models
- Avoiding the common pitfalls:
  - Mistaking competing risks for censoring
  - Not recognizing a time-varying covariate

### Distinction between hazard rates and probabilities in multistate models

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