Analysis of time-to-event for observational studies: Guidance to the use of intensity models

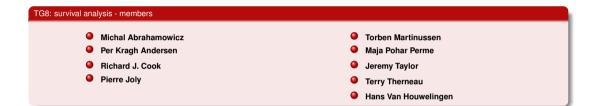
On behalf of STRATOS TG8

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STRATOS mini-symposium, ISCB, 2020

The plan of this talk



Submitted paper:

ANALYSIS OF TIME-TO-EVENT FOR OBSERVATIONAL STUDIES: GUIDANCE TO THE USE OF INTENSITY MODELS

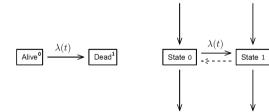
- Basic ideas and pitfalls of survival analysis, organized as checklists
- Hazard models and beyond
- Illustrative example patients with peripheral arterial disease

Survival analysis

Occurrence of a particular event in time

• $\lambda(t)$: intensity (hazard)

incomplete information: censoring or competing risk

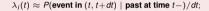


Introduction

Preliminary issues

PH model

Intensity or hazard function



- dynamic description of how events occur in time
- can be estimated directly (assuming independent censoring assumption)
- inclusion of time-dependent covariates
- taking account of delayed entry
- conditionally dependent censoring

May be of interest in its own right, insufficient for some questions - absolute risk

 $\lambda(t) = -\frac{d\log S(t)}{dt}$



Survival analysis - notation

| Standa | Standard notation | | | | | |
|--------|----------------------------------|--|--|--|--|--|
| | T _i : follow-up time | | | | | |
| 0 | δ_i : censoring indicator | | | | | |
| | V _i : entry time | | | | | |
| • | $Z_i(t)$: covariate vecor | | | | | |

Counting process notation

For each individual i

- $Y_i(t)$: at risk indicator. Drops from 1 to 0 in case of event or censoring. In case of delayed entry: can be 0 at t = 0
- $N_i(t)$: counting events. Jumps from 0 to 1 in case of event occurrence.
- \bigcirc $Z_i(t)$: covariate vector

Our data example

Peripheral arterial disease

- Common circulatory problem, narrowed arteries, sign of atherosclerosis, increased risk for CV (cardio-vascular) events
- 742 PAD patients and 713 controls, Slovenia, 5 years of follow up
- Baseline data, measurements at each visit, endpoints
- Goal: survival of patients with PAD (in comparison to controls) despite optimal treatment

Preliminary concepts and issues

In general:

- Time origin: unambigously defined, comparable, clinically relevant. Defines time axis, multiple time axes may be relevant
- Inclusion criteria: must be met by the time the patient enters the study (Y(t) first becomes 1) - danger of immortal time bias
- Event definition: Clearly defined, the definition should be clear at the time of event (when N(t) switches to 1) danger of immortal time bias
- Censoring: We wish to estimate a complete, uncensored, population. Independent censoring assumption. Why was a patient censored?

PAD example:

- Time origin: enrollment or birth, conditional survival in case of age as time axis.
- Inclusion criteria: PAD (and age-matched controls) at the time of enrollment. Ever or never PAD cannot be a criterium, time-varying covariate PAD could be
- Event definition: death (CV or non CV), major CV events(stroke, infarction), minor events (revascularization)
- Censoring: 5th visit after 5 years. Censored at 5 years. Non CV death as a competing risk.

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Age (years)

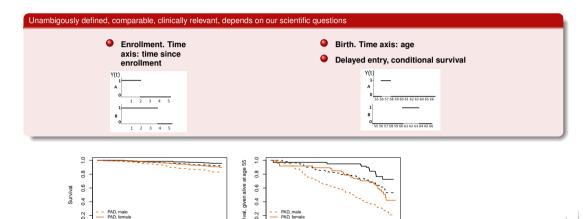
Preliminary concepts and issues - time origin

– Control, male

- Control female

Years since enrollment

0.0



- - Control, male

- Control female

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| Introduction | Preliminary issues | PH model | Competing risks |
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Proportional hazards models

Cox PH model

$\lambda(t|Z_i(t)) = \lambda_0(t) \exp(Z_i(t)^\top \beta)$

- Estimation: maximum partial likelihood
- Std. errors, tests as in classical likelihood
- Valid in simple and more general situation (factorization)

Alternatives

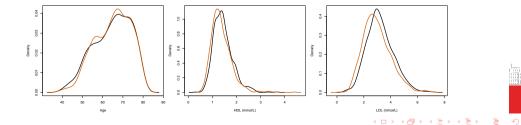
- Other PH models: parametric (constant, piecewise constant, Weibull, splines)
- Cox extensions: time-varying effects, stratified Cox
- Alternative models: additive hazards (Aalen), accelerated failure time (AFT) model

| 0000 | | PH model ○●○○ | |
|--------------------------|----|------------------|--|
| Cox PH mode | əl | | |
| Before fitting the model | | | |
| In general: | | PAD example: | |

- Check the covariates, check the dates
- ٠ Investigate covariate dependent censoring (Cox with censoring as the event): include such variables in the model
- Time-dependent covariates (extrapolation, external covariates, reverse causality bias)

PAD example:

- Covariates: PAD, Sex, Age, LDL, HDL
- Time-dependent covariates: carry last value forward



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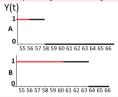
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Immortal time bias

The values of Z(t), N(t) and Y(t) should be defined so that they do not depend on N(s), Y(s) or Z(s) for s > t

Examples in PAD

Age axis: do not forget about delayed entry. Otherwise Y depending on N at a higher age.



Some controls are diagnosed with PAD at later visits. Do not exclude them from the control group. Options:

- PAD status can be time-fixed (value at enrollment)
- Time-dependent (current value)
- but NOT time-fixed at the value at the end of follow-up (ever PAD vs never-PAD). Example of Z depending on later values of itself

PH model ○○○● Competing risks

Fitting the Cox PH model - PAD example, part I

Event - death of any cause

The effect of PAD and sex (m vs f) - which time axis?

- Time since enrollment: add age (per 10 years, assume linearity)
- Age axis: add time since enrollment (FU, per year, assume linearity)
- Multiple axes: Poisson

| Time since enroll | | | Age axis | | | Both axes | | | | |
|-------------------|-----|------|--------------|-----|------|--------------|-----|------|--------------|--|
| | Cov | HR | 95% CI | Cov | HŘ | 95% CI | Cov | HR | 95% CI | |
| | PAD | 2.40 | (1.71, 3.37) | PAD | 2.40 | (1.70, 3.37) | PAD | 2.38 | (1.70, 3.35) | |
| | Sex | 2.00 | (1.40, 2.86) | Sex | 2.02 | (1.42, 2.90) | Sex | 2.01 | (1.41, 2.88) | |
| | Age | 1.93 | (1.57, 2.37) | FU | 1.18 | (1.05, 1.33) | | | | |



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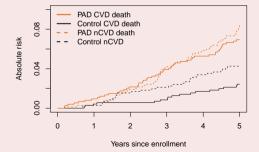
Preliminary issues

PH model

Competing risks analysis - PAD

Death of cardio-vascular reasons

- Non-CV cause: competing risk, not censoring (present in the complete population, elimination not of interest)
- Estimate probabilities: Aalen-Johansen





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Competing risks

Fitting the Cox model - PAD example, part II

Competing risks

- Non-CV cause: can be treated at censoring in the Cox model (factorization of the likelihood)
- Time fixed or time-dependent covariates
- All CV causes (death + stroke, infarction)

| | CV death | | | |
|----------------|----------|-------------|--|--|
| | Ti | me-fixed | | |
| | HR | 95 | | |
| PAD | 2.87 | (1.65-5) | | |
| Sex (m vs. f) | 1.67 | (0.97-2.88) | | |
| Age (per10yrs) | 1.93 | (1.40-2.66) | | |
| HDL (mmol/l) | 0.74 | (0.39-1.41) | | |
| LDL (mmol/l) | 0.92 | (0.72-1.18) | | |

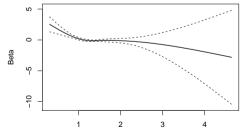


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After fitting the Cox model - PAD example, part III

Check assumptions

- Proportional hazards, linearity (continuous variables)
- Many methods available: Schoenfeld residuals, martingale residuals
- What to do if violated: confounder or the variable of interest (omission of strong predictors!)



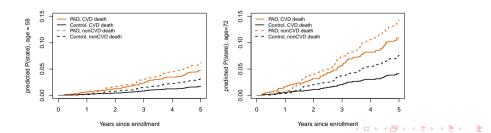
Competing risks

After fitting the Cox model - PAD example, part IV

Reporting and interpretation

- If only HRs are reported no absolute risks can be obtained
- Competing risks: hazard vs probability
- Absolute risks: prediction from t = 0 onwards

| | | xed, other cause |
|------------------|------|------------------|
| | HR | 95% CI |
| PAD | 2.04 | (1.31–3.19) |
| Sex (m vs. f) | 2.12 | (1.29 - 3.50) |
| Age (per 10 yrs) | 1.93 | (1.45-2.56) |
| HDL | 0.82 | (0.43 - 1.55) |
| LDL | 1.02 | (0.83-1.26) |
| | | |





Concluding remarks

The subtitles in the paper

- Preliminary concepts and issues
- The intensity
- Proportional hazard models and alternatives
- A check-list when fitting the Cox model
- Immortal time bias
- Prediction in the absence/presence of competing risks
- Issues in causal inference
- Illustrative applications + supplement with code

Analysis of time-to-event for observational studies: Guidance to the use of intensity models

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