

mvna, AN R-PACKAGE FOR THE MULTIVARIATE NELSON-AALEN ESTIMATOR IN MULTISTATE MODELS

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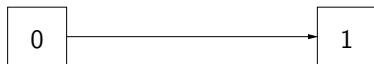


MULTISTATE MODEL FRAMEWORK

- ▶ Time-inhomogeneous Markovian multistate model
- ▶ Possible right-censoring and left-truncation
 - ▶ $(X_t)_{t \in [0, +\infty)}$ a stochastic process with state space $\{0, \dots, K\}$, and right-continuous sample paths

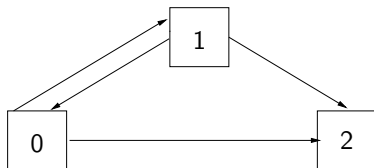
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$$\alpha_{ij}(t)dt = P(X_{t+dt} = j | X_t = i)$$

- ▶ Cumulative transition hazards:

$$A_{ij}(t) = \int_0^t \alpha_{ij}(u)du$$

THE NELSON-AALEN ESTIMATOR

- ▶ Nelson-Aalen estimator of the cumulative transition hazards

$$\hat{A}_{ij}(t) = \sum_{t_k \leq t} \frac{\Delta N_{ij}(t_k)}{Y_i(t_k)}$$

- ▶ $N_{ij}(t)$ number of transitions from state i to j by time t
- ▶ $Y_i(t)$ number of individuals in state i just before time t

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- ▶ $N_{ij}(t)$ number of transitions from state i to j by time t
 - ▶ $Y_i(t)$ number of individuals in state i just before time t
- ▶ The Nelson–Aalen estimator is simply a sum over empirical hazards/empirical conditional transition probabilities

VARIANCE ESTIMATION

- ▶ Aalen variance estimator

$$\hat{\sigma}_{ij}^2 = \sum_{t_k \leq t} \frac{dN_{ij}(t_k)}{Y_i(t_k)^2}$$

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$$\hat{\sigma}_{ij}^2 = \sum_{t_k \leq t} \frac{dN_{ij}(t_k)}{Y_i(t_k)^2}$$

- ▶ Greenwood variance estimator

$$\check{\sigma}_{ij}^2 = \sum_{t_k \leq t} \left\{ \frac{Y_i(t_k) - \Delta N_{ij}(t_k)}{Y_i(t_k)} \right\} \left\{ \frac{dN_{ij}(t_k)}{Y_i(t_k)^2} \right\}$$

RATIONALE

- ▶ The Nelson-Aalen estimator is the fundamental nonparametric estimator in event history analysis.
- ▶ No package available to compute it in multistate models
 - ▶ Univariate software tempts people to use objects that are meaningless in multistate framework, *e.g.*, Kaplan-Meier or single hazard estimates
- ▶ Cumulative hazard estimates give useful insights *e.g.*,
 - ▶ In competing risks analysis
 - ▶ With time-dependent covariates

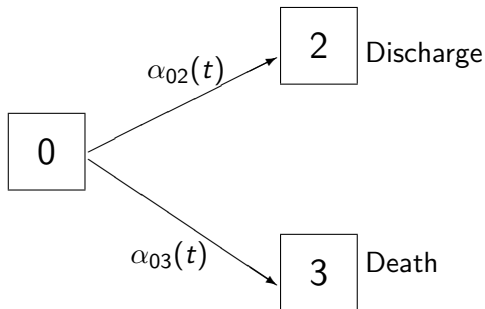
PACKAGE DESCRIPTION

- ▶ `mvna(data, state.numbers, tra, cens.name)`
- ▶ `xyplot.mvna(x, ...)`
- ▶ `plot.mvna(x, ...)`
- ▶ `print(x, ...)`
- ▶ `predict(object, times, ...)`

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- ▶ `print(x, ...)`
- ▶ `predict(object, times, ...)`
- ▶ 2 data sets:
 - ▶ Random samples from intensive care unit cohort data on hospital infections, with a minimum length of stay of 2 days
 - ▶ `sir.adm`:
 - ▶ Effect of pneumonia status on admission on the hazard of discharge and death, respectively
 - ▶ `sir.continuation`:
 - ▶ Effect of ventilation (time-dependent) on the hazard of end-of-stay, a combined discharge/death endpoint

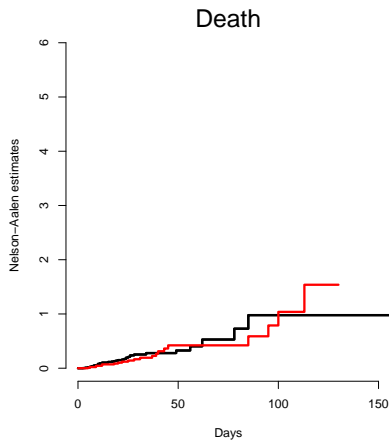
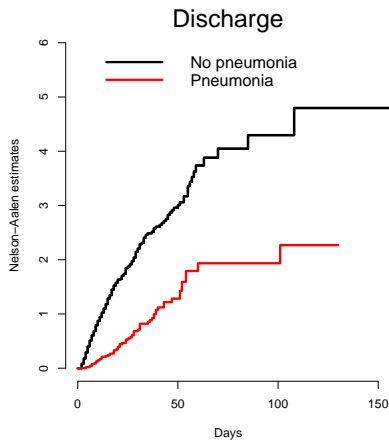
COMPETING RISKS EXAMPLE



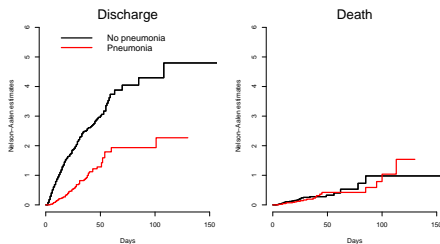
THE DATA SET

- ▶ 765 patients from medical and surgical ICUs
- ▶ 14 (2%) censored observations
- ▶ 97 (13%) patients with pneumonia on admission
 - ▶ 21 (22%) died
- ▶ 668 (87%) patients free of pneumonia
 - ▶ 56 (8%) died

NELSON-AALEN ESTIMATES

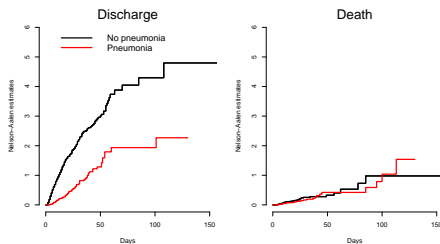


NELSON-AALEN ESTIMATES



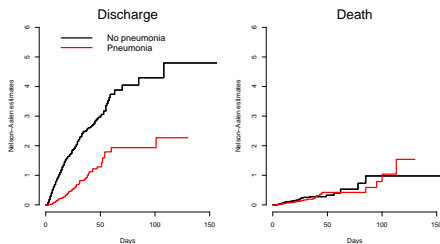
- ▶ More patients die after pneumonia on admission

NELSON-AALEN ESTIMATES



- ▶ More patients die after pneumonia on admission
 - ▶ Pneumonia prolongs hospital stay, as the all-cause hazard is reduced.
 - ▶ Patients with pneumonia stay longer in hospital, exposed to an unchanged death hazard

NELSON-AALEN ESTIMATES



- ▶ More patients die after pneumonia on admission
 - ▶ Pneumonia prolongs hospital stay, as the all-cause hazard is reduced.
 - ▶ Patients with pneumonia stay longer in hospital, exposed to an unchanged death hazard
- ⇒ **Pneumonia increases mortality**

HOW THIS PACKAGE SUPPLEMENTS WHAT ALREADY EXISTS

COMPETING RISKS

COX MODEL FOR THE CAUSE-SPECIFIC HAZARDS

- ▶ Pneumonia status as a baseline binary covariate

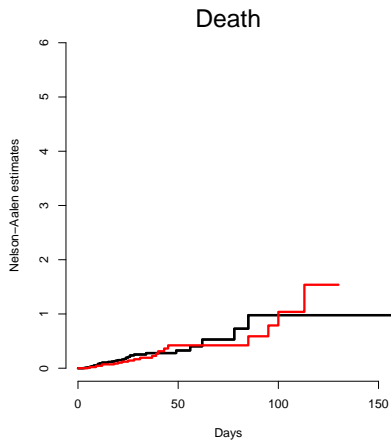
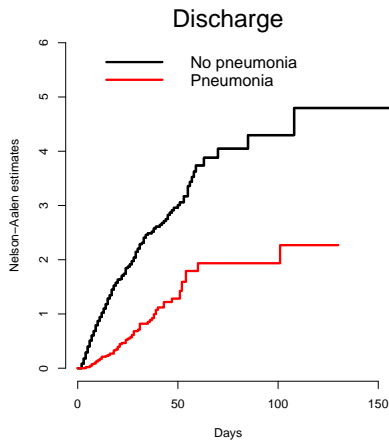
COMPETING RISKS

COX MODEL FOR THE CAUSE-SPECIFIC HAZARDS

- ▶ Pneumonia status as a baseline binary covariate
- ▶ Results:

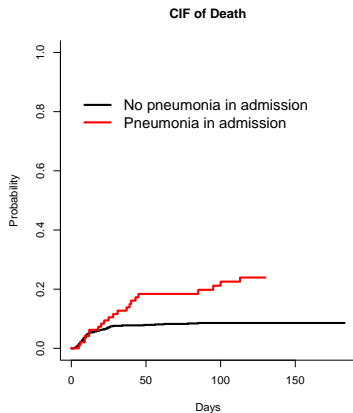
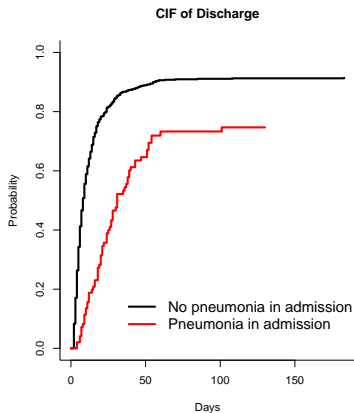
	CSHR	95% CI
Discharge	0.367	[0.258; 0.473]
Death	0.906	[0.537; 1.53]

NELSON-AALEN ESTIMATES



CUMULATIVE INCIDENCE FUNCTIONS

- ▶ Proportion of patients failing due to one risk as time progresses



NOTE ON THE VARIANCE ESTIMATORS

- ▶ Variance estimation is a more concerning problem with multistates

NOTE ON THE VARIANCE ESTIMATORS

- ▶ Variance estimation is a more concerning problem with multistates
- ▶ For standard survival data, Klein (1991) found that:
 - ▶ The Aalen estimator overestimates the true variance for risk sets ≤ 5
 - ▶ The Greenwood estimator underestimates the true variance, but has a smaller MSE
 - ▶ The 2 estimators coincide for risk sets ≥ 10
- ▶ Preliminary simulations in the multistate framework:
 - ▶ Comparable findings
- ▶ Recommendations:
 - ▶ Use of the Aalen estimator

SUMMARY

- ▶ The **mvna** package provides a way to easily estimate and display the cumulative transition hazards from multistate models
- ▶ Extremely useful in illustrating and understanding complex event history processes, *e.g.*,
 - ▶ with competing risks
 - ▶ with a time-dependent covariate
- ▶ Outlook:
 - ▶ **etm** package for computing the empirical transition matrix (transition probabilities)

BIBLIOGRAPHY



Aalen, O. (1978).

Nonparametric Inference for a family of Counting Processes.

The Annals of Statistics, 6:701–726.



Andersen, P. K., Borgan, O., Gill, R. D. and Keiding, N. (1993).

Statistical Models Based on Counting Processes.

Springer-Verlag, New-York.



Klein, J. P. (1991).

Small Sample Moments of Some Estimators of the Variance of the Kaplan-Meier and Nelson-Aalen Estimators.

Scandinavian Journal of Statistics, 18:333–340.

DATA & MODEL

- ▶ $(X_t)_{t \geq 0}$ the competing risks process
 - ▶ $X_t \in \{0, 2, 3\}$
- ▶ The failure time T at which patients leave the initial state 0 is
 - ▶ $T = \inf\{t \in [0, \infty) | X_t \neq 0\}$
 - ▶ X_T denotes the failure cause
- ▶ Cause-specific hazard:

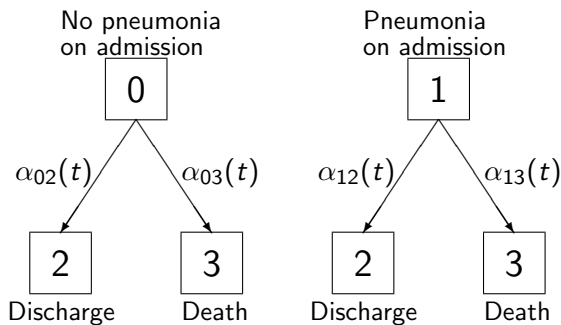
$$\alpha_{0i}(t) = \lim_{\Delta t \rightarrow 0} \frac{P(t \leq T < t + \Delta t, X_T = i | T \geq t)}{\Delta t}, i = 2, 3$$

HOW TO?

NEW DEFINITION OF THE COMPETING RISKS PROCESS

How to?

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How to?

```
> ### Matrix of logical indicating the possible transitions
> tra <- matrix(ncol=4,nrow=4,FALSE)
> tra[1:2,3:4] <- TRUE
> tra
      [,1] [,2] [,3] [,4]
[1,] FALSE FALSE TRUE  TRUE
[2,] FALSE FALSE TRUE  TRUE
[3,] FALSE FALSE FALSE FALSE
[4,] FALSE FALSE FALSE FALSE
```


How to?

```
> ### Nelson_Aalen estimates
> na.pneu <- mvna(data=dat.sir,
+               state.names=c("0","1","2","3"),
+               tra=tra,cens.name="cens")
```

How to?

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> ### Nelson_Aalen estimates
> na.pneu <- mvna(data=dat.sir,
+               state.names=c("0","1","2","3"),
+               tra=tra,cens.name="cens")

> na.pneu
Estimated cumulative hazard for transition 0 to 2

Time
[1]  1  16  30  46 183

Nelson-Aalen estimates
[1] 0.024 1.336 2.166 2.863 5.613

Variance estimates
[1] 0.000 0.005 0.014 0.033 1.411

Alternative variance estimates
[1] 0.000 0.004 0.013 0.031 0.266
```

How to?

```
> ### Plot
> xyplot(na.pneu,
+ tr.choice=c("0 2", "1 2", "0 3", "1 3"),
+ aspect=1, strip=strip.custom(bg="white",
+ factor.levels=
+   c("No pneumonia on admission - Discharge",
+     "Pneumonia on admission - Discharge",
+     "No pneumonia on admission - Death",
+     "Pneumonia on admission - Death"),
+ par.strip.text=list(cex=0.9)),
+ scales=list(alternating=1), xlab="Days",
+ ylab="Nelson-Aalen estimates")
```

CUMULATIVE INCIDENCE FUNCTIONS

- ▶ CIF of discharge

$$\begin{aligned}F_2(t) &= P(T \leq t, X_T = 2) \\ &= \int_0^t P(T > u-) \alpha_{02}(u) du\end{aligned}$$

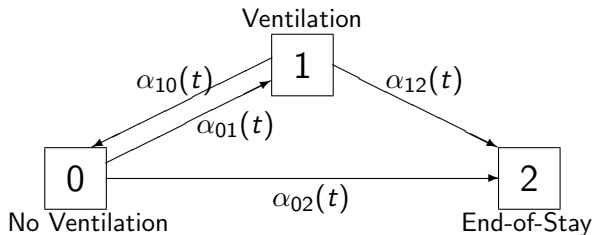
- ▶ Depends on both cause-specific hazards

$$P(T > t) = \exp\left(-\int_0^t \alpha_{02}(u) + \alpha_{03}(u) du\right)$$

- ▶ Loss of the one to one relationship between hazard and probability

TIME-DEPENDENT COVARIATE

AS A TRANSIENT STATE IN A MULTISTATE MODEL



MODEL

- ▶ $(V_t)_{t \geq 0}$ the stochastic process
 - ▶ $V_t \in \{0, 1, 2\}$

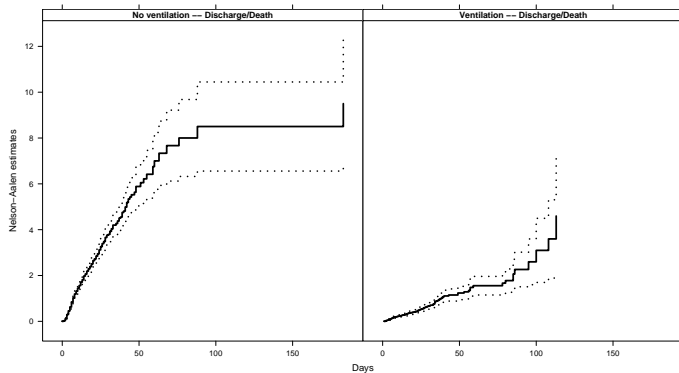
- ▶ Survival time:

$$T = \inf\{t \in [0, \infty) | X_t = 2\}$$

- ▶ The transition hazard is

$$\alpha_{ij}(t)dt = P(X_{t+dt} = j | X_t = i)$$

NELSON-AALEN ESTIMATES



TIME-DEPENDENT COVARIATE COX MODEL

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COX MODEL

- ▶ Cox proportional hazards model:

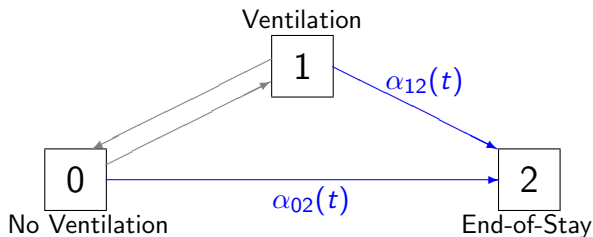
$$\alpha_{12}(t) = \alpha_{02}(t) \cdot \exp(\beta)$$

TIME-DEPENDENT COVARIATE

COX MODEL

- ▶ Cox proportional hazards model:

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0.179	[0.149; 0.214]

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