

# probabilities in multi-state survival data

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## aim

Present different approaches for the estimation of transition probabilities in multi-state survival data. The central question here is: what is the probability that a randomly selected person is in stage  $j$  at time  $t$ , conditionally on being in state  $h$  at time  $s$ :

$$p_{hij}(s, t) = P(X(t) = j | X(s) = h)$$

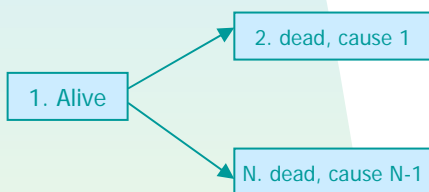
## Multi-state models (MSM)

A MSM is a model for a stochastic process in continuous time allowing individuals to move between a finite number of states. An individual moves from one state to another through time. The next state to which the individual moves, and the time of change, are specified through transition intensities that provide the instantaneous hazard for movement out of one state into another.

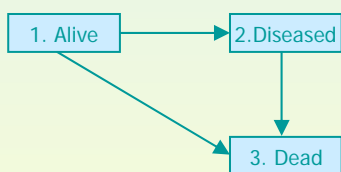
### Some MSM



The mortality model



The Competing risks model



The Illness-death model

## Nonparametric approaches

### Illness-death model

**Aalen–Johansen estimators of transition probabilities:** This approach can be thought of as the generalization of the Kaplan–Meier estimate of the simple mortality model (assumes the process to be Markovian):

$$\widehat{p}_{11}(s, t) = \prod_{s < t_{(i)} \leq t} \left( 1 - \frac{d_{12i} + d_{13i}}{n_{1i}} \right) \quad \widehat{p}_{22}(s, t) = \prod_{s < t_{(i)} \leq t} \left( 1 - \frac{d_{23i}}{n_{2i}} \right)$$

$$\widehat{p}_{12}(s, t) = \sum_{s < t_{(i)} \leq t} \widehat{p}_{11}(s, t_{(i-1)}) \frac{d_{12i}}{n_{1i}} \widehat{p}_{22}(t_{(i)}, t)$$

For a sample of  $n$  subjects,  $t_{(1)} < t_{(2)} < \dots < t_{(k)}$  denote the  $k$  event times;  $d_{hji}$  the  $n^\circ$  of transitions from state  $h$  to state  $j$  at time  $t_{(i)}$ ;  $n_{hi}$  the number of individuals at risk at time  $t_{(i)}$ .

**Markov-free estimator:** alternative estimators which do not rely on the Markov assumption are expressed as:

$$\widehat{p}_{11}(s, t) = \frac{1 - \widehat{H}(t)}{1 - \widehat{H}(s)}$$

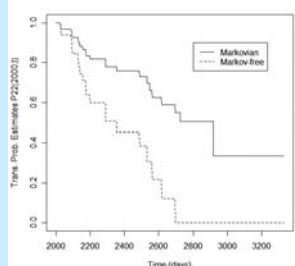
$$\widehat{p}_{12}(s, t) = \frac{1}{1 - \widehat{H}(s)} \sum_{i=1}^n W_i \phi_{s,x}(Z_{[i]}, Y_{(i)})$$

$$\widehat{p}_{22}(s, t) = \sum_{i=1}^n W_i \tilde{\phi}_{s,x}(Z_{[i]}, Y_{(i)}) / \sum_{i=1}^n W_i \tilde{\phi}_{s,s}(Z_{[i]}, Y_{(i)})$$

$Y = \min(T, C)$ ;  $T$  – survival time;  $C$  – right-censoring time  
 $Z$  is the sojourn time in state 1, with distrib. function  $H$ , and  $\widehat{H}$  its Kaplan–Meier estimator.  
 $\phi_{s,x}(u, v) = \mathbb{I}(s < u \leq t, v > t)$  and  $\tilde{\phi}_{s,x}(u, v) = \mathbb{I}(u \leq s, v > t)$   
 $W_i$  are the Kaplan–Meier weights attached to  $Y_{(i)}$ ;  
 $Y_{(1)} \leq \dots \leq Y_{(n)}$  denote the ordered sample of the  $Y_i$ 's, and  $Z_{[i]}$  for the pair attached (concomitant) to the  $Y_{(i)}$  value.

## Colon cancer data

For the Colon cancer data (Moertel et al. 2000) we may consider the recurrence as an associated state of risk, and use the illness-death model with states “alive and disease-free”, “alive with recurrence” and “dead”.



## Conclusions

- The “Markov-free” estimator, have fewer jump points but with bigger steps. The number of jump points and the size of the steps are related to censoring and to the sample size.
- With regard to the survival prognosis, we observe serious departures between both survival curves for individuals who have had recurrence.

## References

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Moertel CG et al. Levamisole and fluorouracil for adjuvant therapy of resected colon carcinoma. *New Eng. J. Med.* (2000) 352-358