



Sciences Economiques et Sociales de la Santé
& Traitement de l'Information Médicale

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Competing risks, analysis and interpretation

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Competing risks, analysis and interpretation

SESSTIM

December 18, 2020

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Outline

Competing risks, what are they?

Standard survival analysis

Competing risks

Marginal versus competing risks

Two approaches to competing risks analysis

Multi-state approach

Subdistribution approach

Regression

Interpretation

Marginal versus competing risks

Competing risks: multi-state or subdistribution approach

Example

Rate and Risk

T time to event (e.g. death)

- **Risk**: $P(T \leq t)$ (or survival $P(T > t)$)
- **Rate** (hazard, incidence):

$$h(t) = P(T = t | T \geq t)$$

$$\lim_{\Delta t \downarrow 0} \frac{P(t \leq T < t + \Delta t | T \geq t)}{\Delta t}$$

discrete

continuous

Kaplan-Meier

year	0-1	1-2	2-3	3-4	4-5	5-6	6-7	Total
death	1	2	6	11	9	11	2	60
ensor	5	9	6	6	9	12	4	86

Kaplan-Meier

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$$\begin{aligned}
 P(> 6 \text{ year alive}) &= P(\text{year 0-1 alive}) \times \\
 &\quad P(\text{year 1-2 alive} \mid \text{alive until year 1}) \times \\
 &\quad \dots \times \\
 &\quad P(\text{year 5-6 alive} \mid \text{alive until year 5}) \\
 &= (1 - h_{0-1}) \times (1 - h_{1-2}) \times (1 - h_{2-3}) \times \dots \times (1 - h_{5-6})
 \end{aligned}$$

$$\text{Year 0-1: } h_{0-1} = 1/146 = 0.006849$$

$$\text{Year 1-2: } h_{1-2} = 2/140 = 0.014286$$

$$\text{Year 2-3: } h_{2-3} = 6/129 = 0.046512 \text{ etc.}$$

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Assumption: censored individuals represented by those at risk

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- One-to-one relation via

$$P(T > t) = \prod_{t_i \leq t} \{1 - h(t_i)\}$$

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$$P(T > t) = \exp\left\{-\int_0^t h(u) du\right\}$$

continuous

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- Kaplan-Meier
- Cox model $h(t) = h_0(t) \exp\{\beta_1 X_1 + \dots + \beta_p X_p\}$

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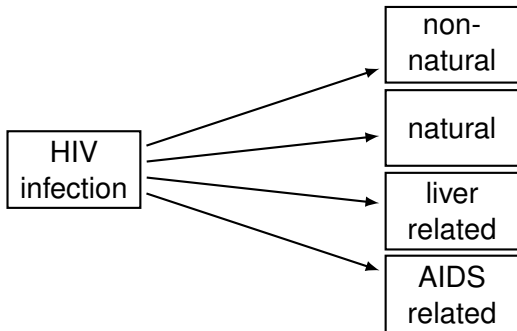
Competing risks: multi-state or subdistribution approach

Example

What and how?

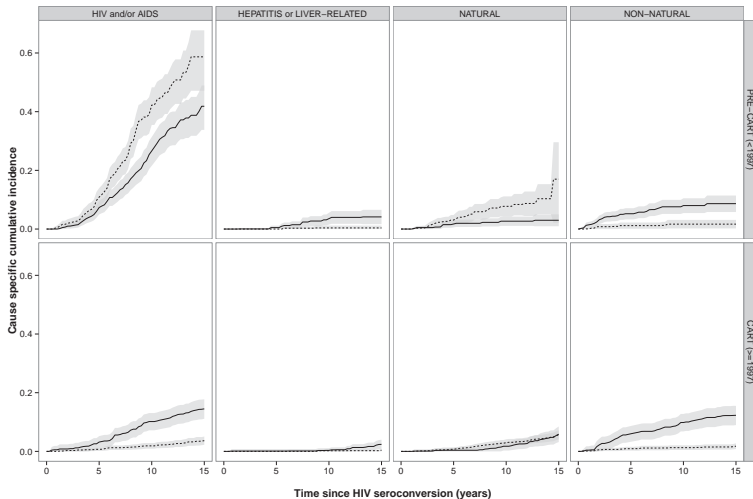
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 - or*
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Example I: Causes of death (COD) after HIV infection



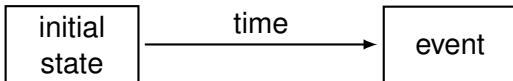
- Has the spectrum in causes of death changed after the introduction of cART (combination anti-retroviral therapy)

Cause-specific mortality by calendar period and hepatitis C status



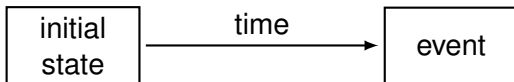
Beyond classical survival analysis

- Classical: transition between two states, one event type.
“We all die, but not all at the same age”



Beyond classical survival analysis

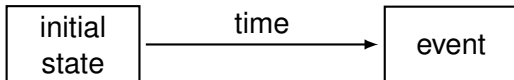
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- Life and **death** are richer than that
 1. Multiple causes of death. Competing risks:
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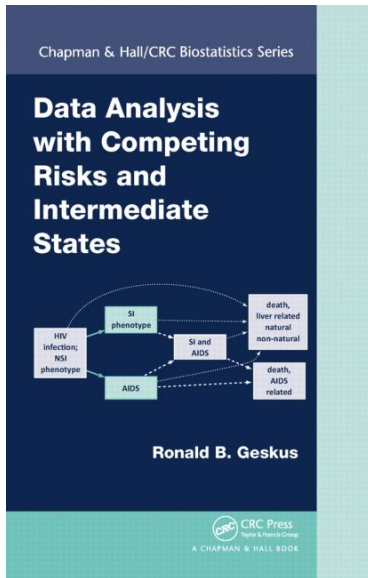
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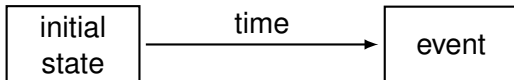
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Published by CRC Press, 2015



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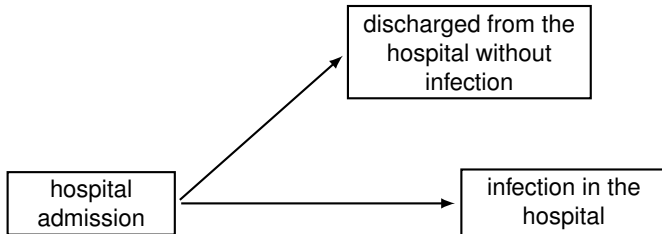
Interpretation

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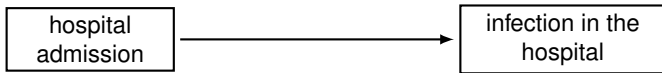
Competing risks: multi-state or subdistribution approach

Example

II: Time to staphylococcus infection during hospital stay

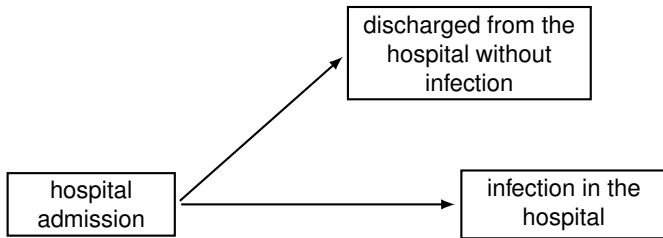


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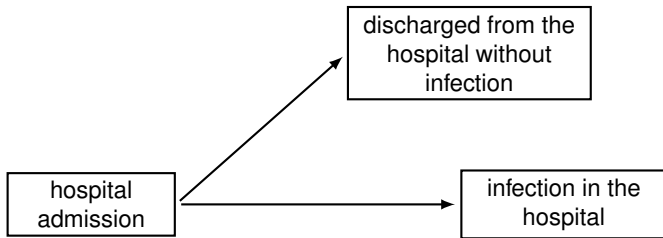
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Estimation with complete follow-up (artificial data)

week	0-1	1-2	2-3	3-4	4-5	5-6	6-7	> 7
infection	1	2	6	11	9	11	2	18
cumulative	1	3	9	20	29	40	42	60
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$$\hat{P}(\text{infection} \leq 6 \text{ weeks}) = 40/146$$

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Individuals with competing event remain in denominator,
competing event ignored in estimation

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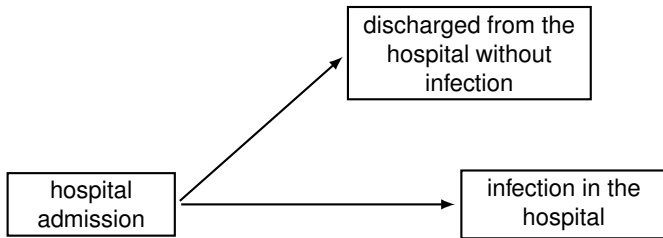
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- **Net risk.** Discharged individuals treated and interpreted as censored.

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 - Marginal distribution/net risk
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- Which of two hospitals has higher risk may depend on type of question

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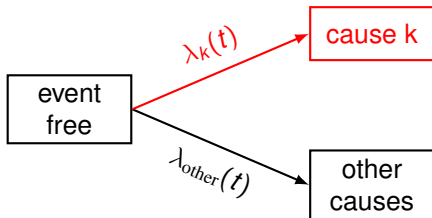
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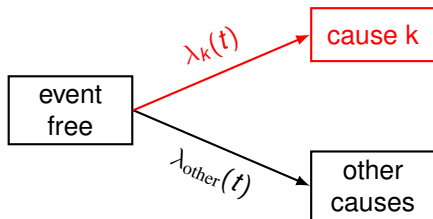
I: The multi-state approach: cause-specific hazard



- Transition rate to cause k:

$$\lambda_k(t) = P(T = t, E = k | T \geq t) = \frac{P(T=t, E=k)}{P(T \geq t)} \quad (\text{discrete})$$

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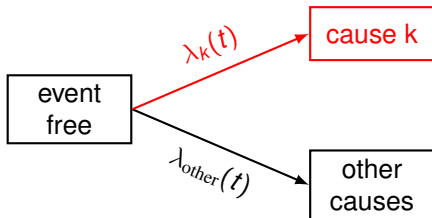


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- Risk $P(T \leq t, E = k)$: Aalen-Johansen estimator; combines $\widehat{\lambda}_k(t)$ with Kaplan-Meier based on overall hazard $h(t) = \sum_{e=1}^K \lambda_e(t)$ (e.g. $K = 4$)

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T_k^* time to event of type k : $T_k^* = T \times I\{E = k\} + \infty \times I\{E \neq k\}$

- $P(T_k^* \leq t) = P(T \leq t, E = k)$

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Denominator: event free **or with earlier competing event**

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- One-to-one relation with crude risk

$$\widehat{P}(T_k^* > t) = \prod_{t_i \leq t} \left\{ 1 - \widehat{h}_k(t_i) \right\}$$

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Regression on hazard

- Cause-specific hazard
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 - Standard Cox model
 - Interpretation: cause-specific event rate among event-free individuals
 - Not a marginal hazard, unless progression to competing risks unrelated
- Subdistribution hazard
 - Fine and Gray model: proportional subdistribution hazards
 - Interpretation: direct relation with cause-specific cumulative incidence

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Rates and risks in competing risks setting

	hazard		cumulative	
competing risks	marginal		net risk	
			marginal survival function	
			marginal cumulative incidence	
	cause-specific subdistribution	λ_k h_k	no corresponding quantity	
			crude risk	$P(T_k^* \leq t)$
			cause-specific cumulative incidence	
combined	overall	h	overall risk	$P(T \leq t)$
			overall survival function	
			overall cumulative incidence	

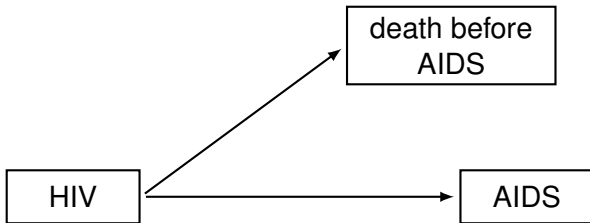
Marginal distribution

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- Standard hazard estimate. Basis for Kaplan-Meier estimate of cumulative incidence/net risk

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- **Assumption: Individuals that are censored can be represented by the ones that remain at risk.** Mechanisms unrelated
- Otherwise: Kaplan-Meier has no meaning.
- Extra information may allow to show informative/dependent censoring, but independence can never be tested for

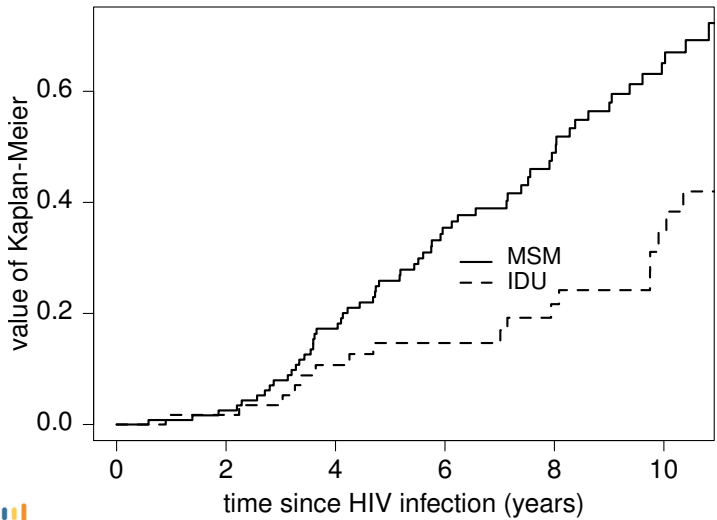
III: Time from HIV infection to AIDS



- Compare men who have sex with men (MSM) and injecting drug users (IDU)
- Interest in time to AIDS if there were no pre-AIDS death. Interest in etiology and marginal distribution
- Kaplan-Meier: censor at death (before AIDS)

Assumption: those alive without AIDS represent pre-AIDS deaths

Kaplan-Meier: IDU much slower progression ($p = 0.001$)



Explanation: dependent censoring

- Information on cause of death before AIDS

	IDU	MSM
Reason of death	Number	
HIV related infections	3	0
overdose/suicide	6	0
violence/accident	2	0
liver cirrhosis	2	0
cancer	0	1
heart attack	0	1
unknown	4	3

- Some causes of pre-AIDS death in IDU related to AIDS progression, i.e. close to AIDS. Net risk estimate for IDU biased downwards

Rates and risks in competing risks setting

	hazard		cumulative	
competing risks	marginal		net risk	
	cause-specific subdistribution	λ_k h_k	marginal survival function marginal cumulative incidence no corresponding quantity	
combined	overall	h	crude risk cause-specific cumulative incidence	$P(T_k^* \leq t)$
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 - If censoring due to competing event is independent, then marginal and cause-specific hazard are equal. Cumulative quantities different: Kaplan-Meier versus Aalen-Johansen

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- Subdistribution hazard: one-to-one relation with crude risk

Marginal or competing risks?

- Example I: spectrum in COD
Competing risks; marginal analysis completely hypothetical
No interest in change in AIDS-related death in world in which other COD's do not exist
- Example II: staphylococcus infection in hospital
 - **Marginal**: what if everyone would stay in hospital
 - **Competing risks**: how many infections are observed in hospital
- Example III: difference in natural history between IDU en MSM
Marginal analysis (but not feasible with ACS data)

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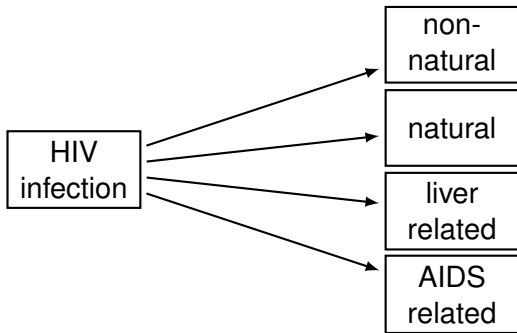
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Example

Which rate or risk?

- Both Cox model (cause-specific hazard) and Fine and Gray model (subdistribution hazard) make sense in presence of competing risks

Example I: Causes of death (COD) after HIV infection



- Has the spectrum in causes of death changed after the introduction of cART (combination anti-retroviral therapy)
- Different hazards can be chosen

Which rate or risk?

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- Example
 - AIDS-specific mortality reduced by cART
 - Other COD's: more frequent, even if cART has no side effects. No change in cause-specific hazard, but subdistribution hazard increases (“in the end we all die”)
 - Subdistribution hazard includes impact on other event types

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- Example
 - AIDS-specific mortality reduced by cART
 - Other COD's: more frequent, even if cART has no side effects. No change in cause-specific hazard, but subdistribution hazard increases (“in the end we all die”)
 - Subdistribution hazard includes impact on other event types
- Prediction: crude risk
Based on regression model for cause-specific or subdistribution hazard, but only latter has one-to-one relation with cause-specific cumulative incidence/crude risk

Outline

Competing risks, what are they?

Standard survival analysis

Competing risks

Marginal versus competing risks

Two approaches to competing risks analysis

Multi-state approach

Subdistribution approach

Regression

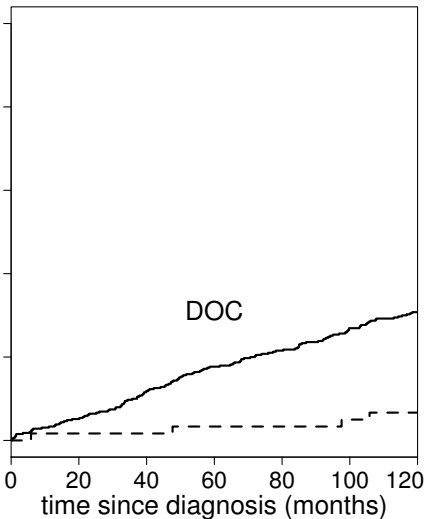
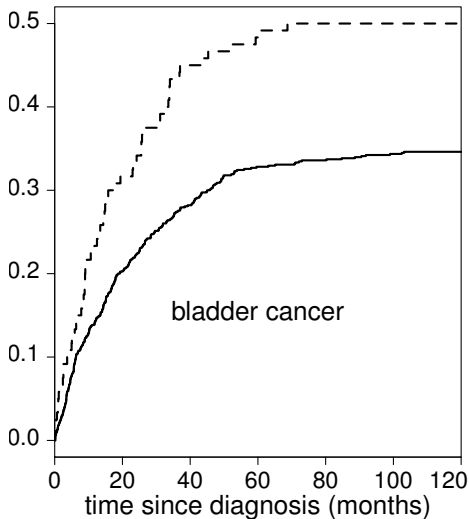
Interpretation

Marginal versus competing risks

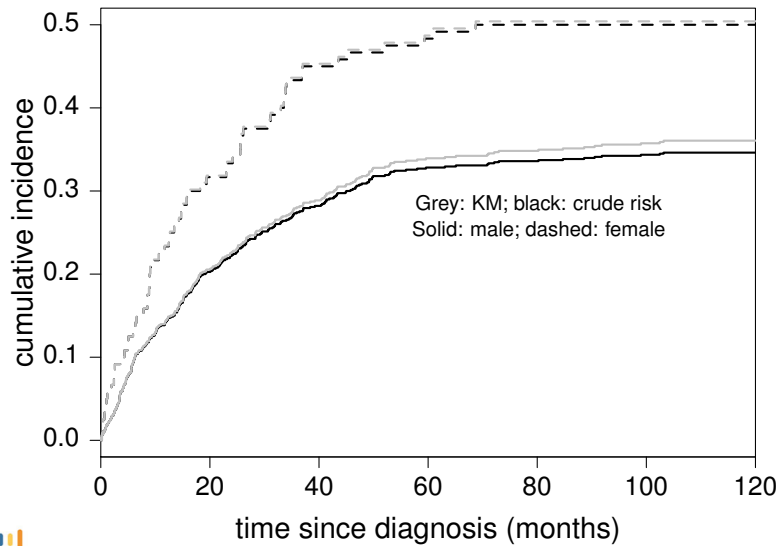
Competing risks: multi-state or subdistribution approach

Example

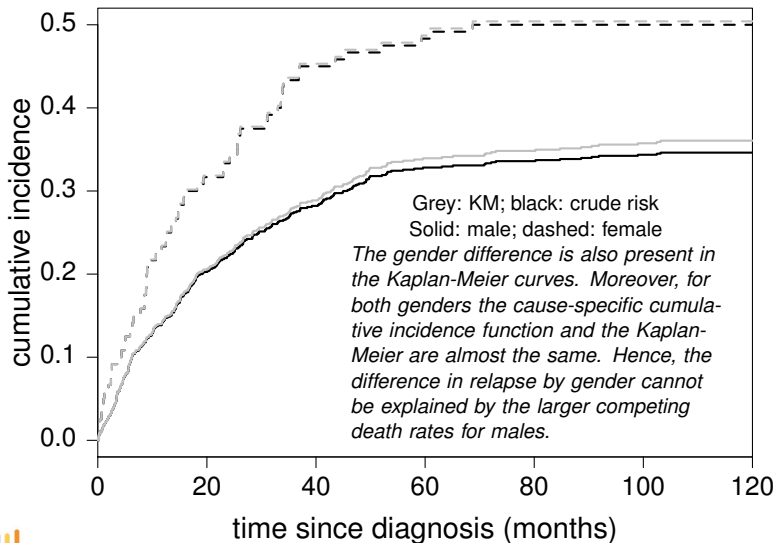
Bladder cancer; relapse, DOC competing



Bladder cancer; relapse, DOC competing



Bladder cancer; relapse, DOC competing



Answer

- The Kaplan-Meier tries to compare the marginal distribution of time to relapse for males and females. Only valid if DOC is noninformative for relapse.
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Answer

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- Estimates almost equal because there is little mortality due to other causes, at least during the first 40 months.
- If we combine both event times, the curves for males and females will become similar. Would estimate marginal hazard if every person that died would have progressed on the next day.
- All we can conclude is that females have a higher relapse-specific cumulative incidence than males. And females have a lower DOC-specific incidence than males.

THANKS!

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