

Time-to-Event Data: A Different Angle by Population Evolution Charts And New Statistical Tests

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the idea for population evolution charts came from.....

- **A practical question:** in a pool of studies with differing observation length, we observed a decrease in estimated hazard for CV disease in the later stage.
- Initially we had about 35% who reported preexisting heart disease at baseline – can we still compare this population with the one selected out by death after 3 years?
- After 3 years the hazard for CV disease seemed to be lower – may be all problematic patients already died?
- Finally we found that the proportion of patients reporting heart disease at baseline even slightly increased for the pool population after 3 years.

This was the beginning !

A slightly more formal setup ...

- Consider a random variable T for the event time
- Allow for censoring: a second rv C for time-to-censoring
- Assume covariates, known at baseline, either metric or categorical (binary)

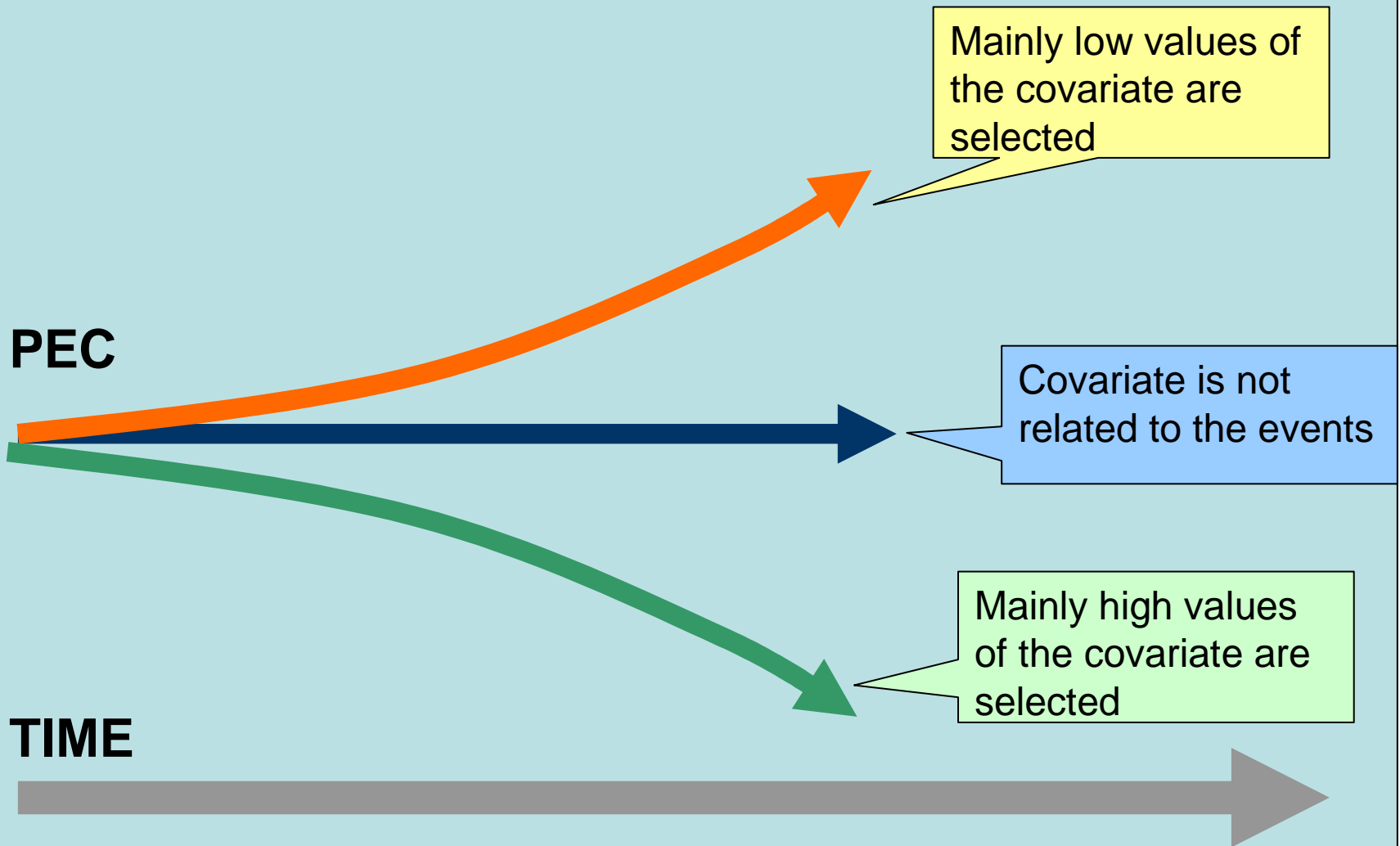
Selection Processes

- The following groups/cohorts are considered:
 - $G(0)$ the initial group at time zero; the full group
 - $E(t)$ the group of patients with an event $\leq t$
 - $Z(t)$ the group of patients with censoring $\leq t$
- Thus the available group $G(t)$ at time t can be written:
 $G(t) = G(0) \div E(t) \div Z(t)$, where \div denotes set subtraction.
- $G(t)$ can be conceived to arise from $G(0)$ by two selection processes, namely events and censoring.

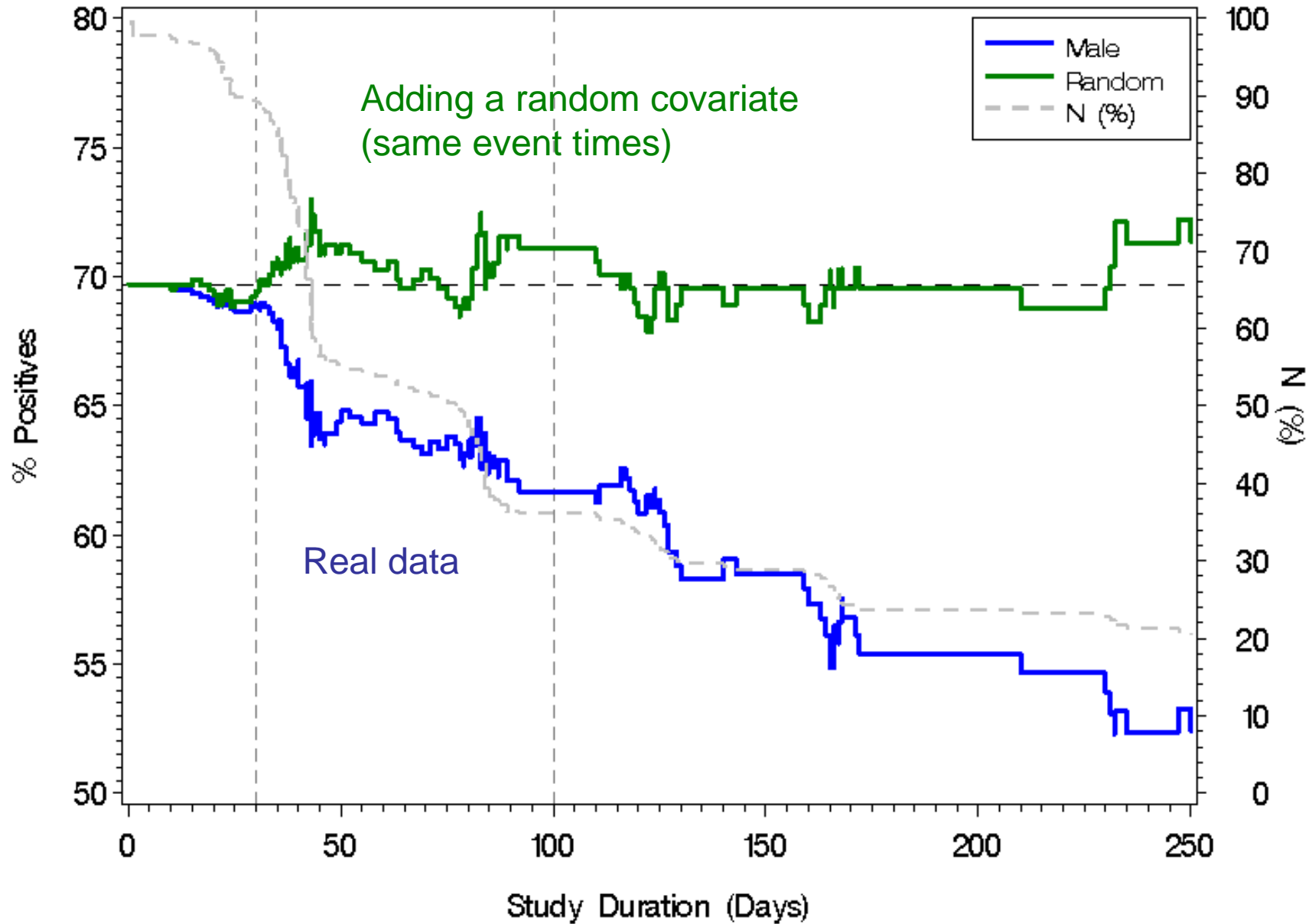
Selection Processes: binary

Simple idea behind Population Evolution Charts (PEC)	
Cohort at baseline: $G(0)$	This cohort is reduced through events/censoring in the course of the study
Cohort at time t : $G(t)$	This cohort is at risk after the time t has elapsed
Binary covariate X e.g. sex	$P(X = 1) = \pi_1(0)$ the expectation at $t = 0$
Suppose X not related to events/censoring	Expect $P(X = 1) = \pi_1(t) = \pi_1(0)$
Suppose X is related to events/censoring	Expect $\pi_1(t) < \pi_1(0)$ or $\pi_1(t) > \pi_1(0)$
Define a simple PEC	Course of $\hat{\pi}_1(t)$, i.e. the proportion of $X = 1$ for the cohorts $G(t)$.
PECs come as graphics	PECs do not need difficult assumptions

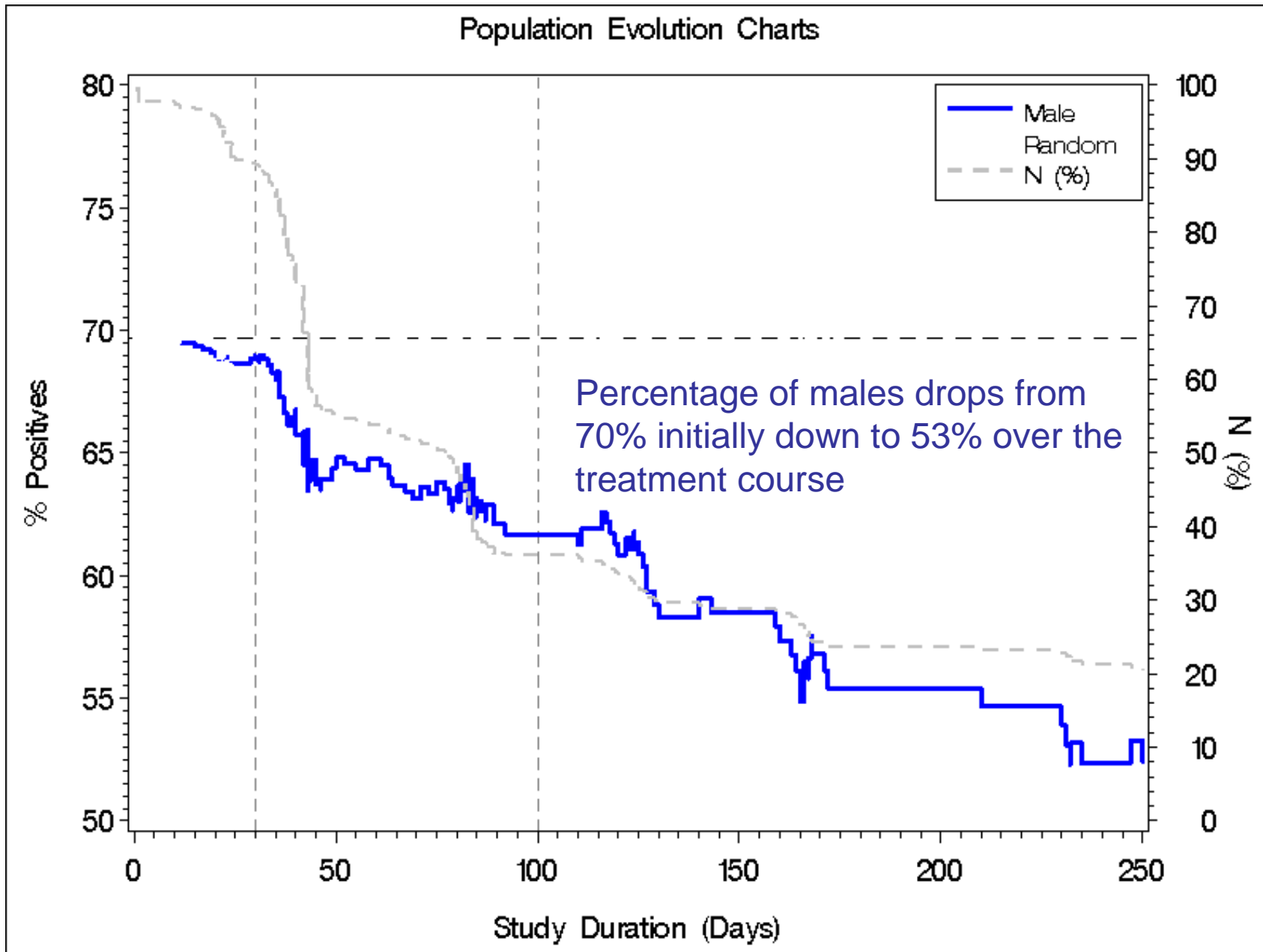
Selection Processes Interpretation



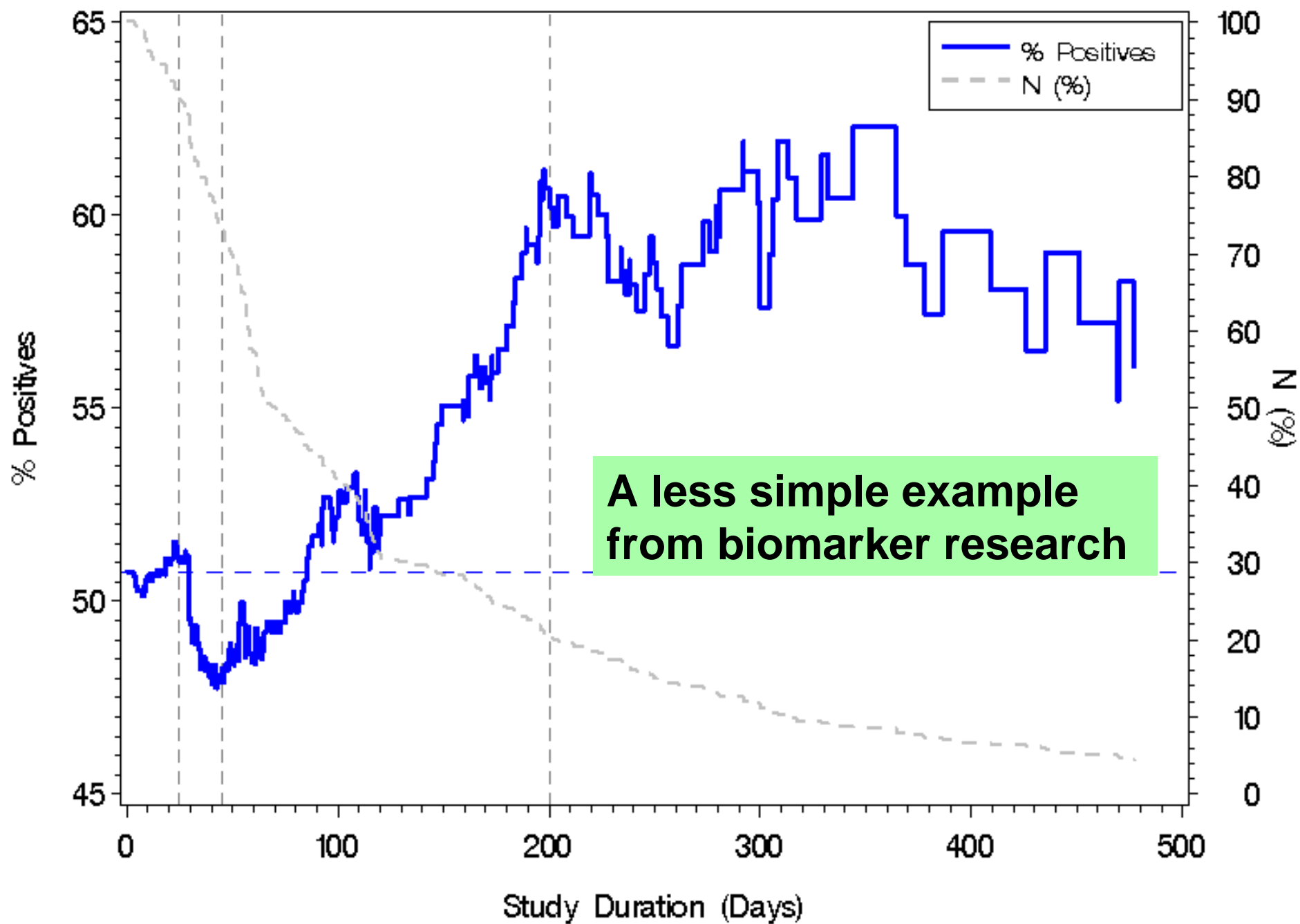
Population Evolution Charts



Population Evolution Charts



Population Evolution Chart



A less simple example
from biomarker research

Set up of the terminology for the case of binary covariates

Basic Notation

Basic Definitions	Random Variable T
Survivor Function $S(t)$	$S(t) = 1 - F(t) = P(T > t)$
Integrated Hazard $H(t)$	$H(t) = -\log(S(t))$
Density $f(t)$	$f(t) = \frac{d}{dt} F(t)$
Hazard $h(t)$	$h(t) = \frac{d}{dt} H(t) = -\frac{d}{dt} \log(S(t)) = \frac{f(t)}{S(t)}$

Define Population Evolution Chart (PEC)

PEC Base Definition	$P(X = 1 T > t)$
Usual regression considers	$P(T > t X)$

Define Population Evolution Chart (PEC)

Equivalent Definitions of Population Evolution Charts (PEC)	
$\Psi_X(t)$	
I. Base Definition	$P(X = 1 T > t)$
II. PEC as selection process	$\frac{1}{S(t)} [P(X = 1) - (1 - S(t)) \cdot P(X = 1 T \leq t)]$
III. PEC in terms of subgroup survivor function	$\frac{S_1(t)}{S(t)} P(X = 1)$
PECs come as graphics	PECs do not need difficult assumptions

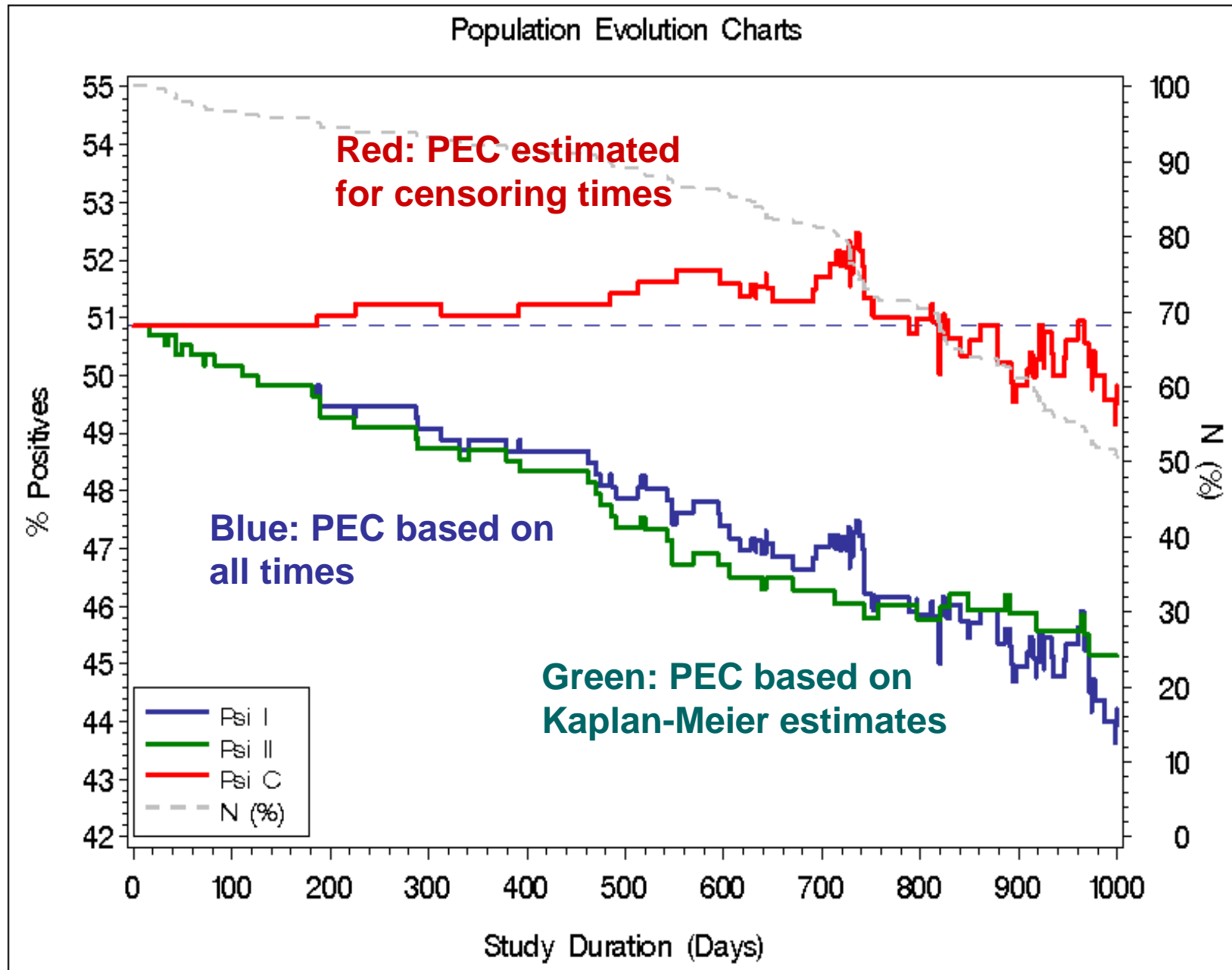
Wait a moment

- What about censoring?
- How to deal with it in the estimation of a PEC?

Estimate Population Evolution Chart

Estimation of Population Evolution Charts (PEC) The Censoring Problem	
Censoring C independent of T Consider all observed times $A = \min(T, C)$	$P(A > t) = P(T > t) \cdot P(C > t)$ KM-estimates $\hat{P}(A > t) = \hat{P}(T > t) \cdot \hat{P}(C > t)$
Assumption: Covariate not related to censoring	Since $P(X = 1 A > t) = P(X = 1 T > t)$ Estimate $\hat{\Psi}_I(t) = \frac{1}{ G(t) } \sum_{j \in G(t)} x_j$ from all times
Using only assumptions for Kaplan-Meier estimates	$\hat{\Psi}_{II}(t) = \frac{\hat{S}_1(t)}{\hat{S}(t)} \cdot \bar{x}(0)$ Provides proper dealing with censoring times
Define a PEC for the censoring process – to check for selective censoring	$\hat{\Psi}_C(t) = \frac{\hat{P}(C > t X = 1)}{\hat{P}(C > t)} \cdot \bar{x}(0)$
PECs come as graphics	PECs do not need difficult assumptions

All Estimates



For binary covariates Evolution Charts offer

- A simple graphical representation of dependencies
- Depicts time dynamics in an easy way
- Could also serve to check for selective censoring

Population Evolution Charts and Cox proportional hazard model

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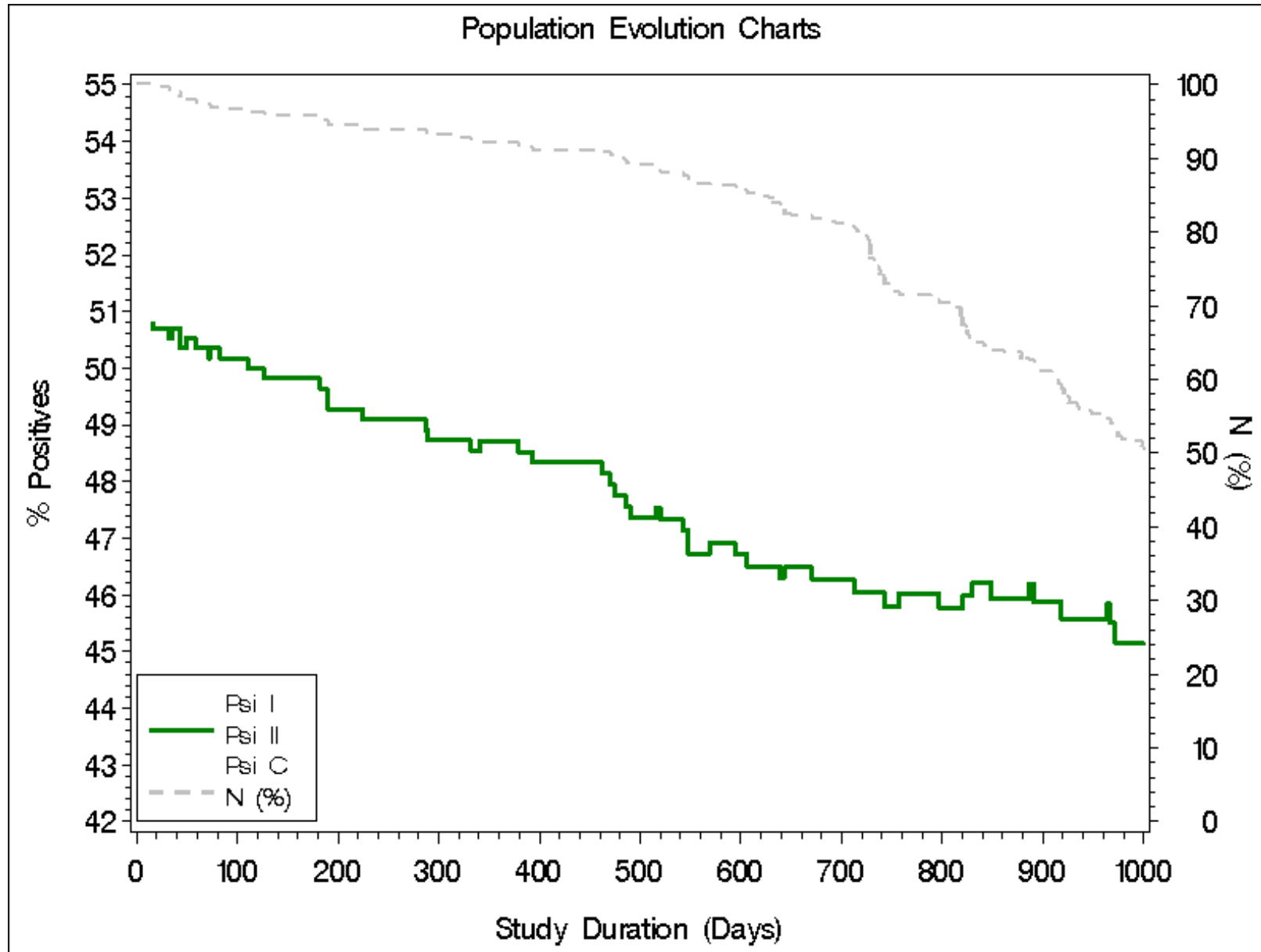
PEC & Cox

Studying Relation of a Binary Covariate with Event Times	
Standard Approach $P(T > t X)$	Cox model $S(t X = 1) = S(t X = 0)^\lambda$ Crucial proportionality $h_1(t) = \lambda h_0(t)$
Population Evolution Chart $P(X = 1 T > t)$	No assumptions regarding the hazards
PECs come as graphics	PECs do not need difficult assumptions

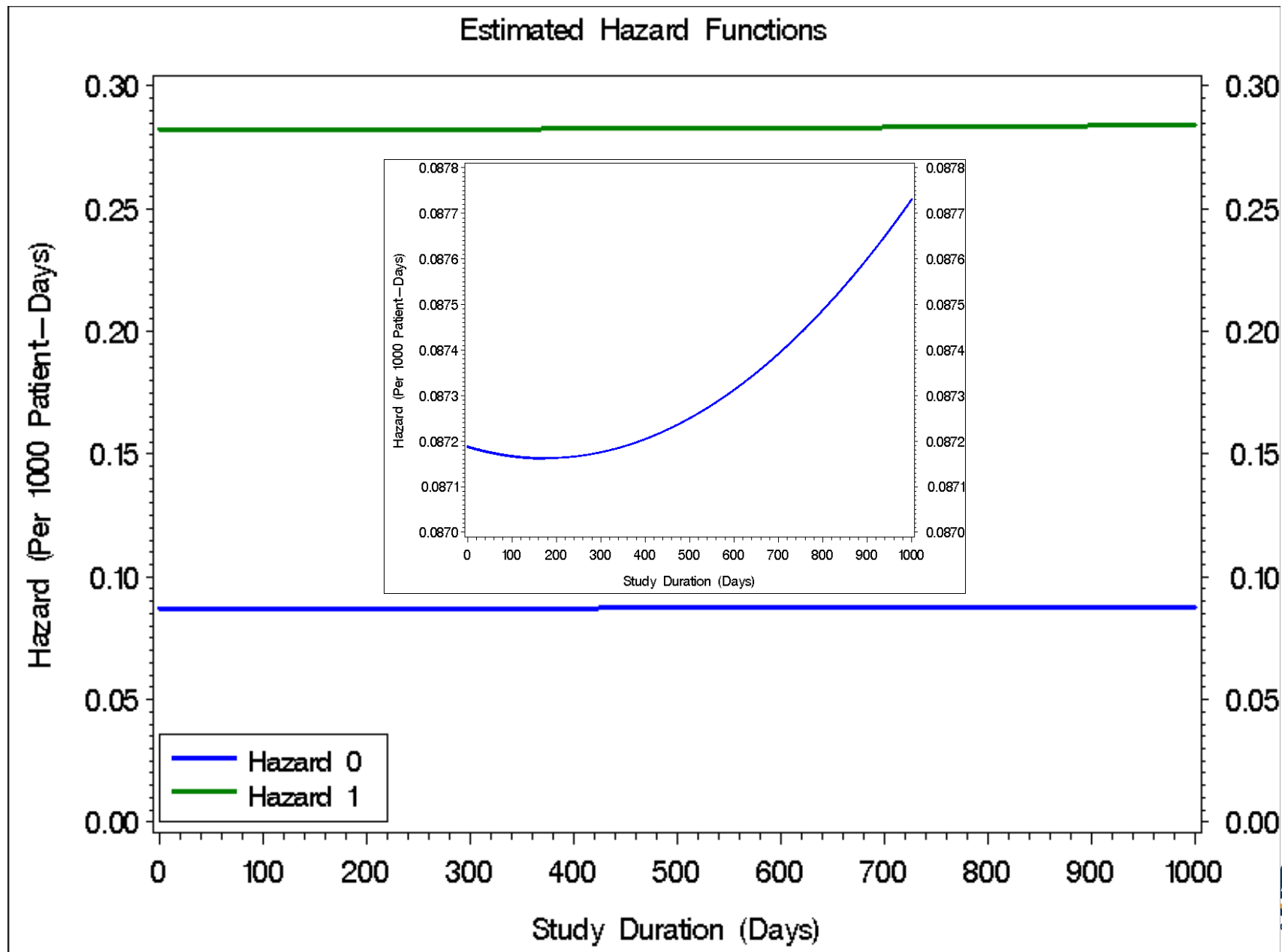
PEC & Cox

PEC monotonic	<p>A necessary condition for validity of the Cox model</p> $\Psi'_X(t) = (1 - \lambda) \cdot h_0(t) \cdot \Psi_X(t) \cdot (1 - \Psi_X(t))$
PEC monotonic & Cox Model holds	Explicit hazard estimate is possible
<p>PEC linear & Cox Model holds</p> $\Psi_X(t) = \alpha \cdot t + \pi_1$	$h_0(t) = \frac{1}{1 - \lambda} \cdot \frac{\alpha}{(\alpha \cdot t + \pi_1) \cdot (-\alpha \cdot t + \pi_0)}$
<p>results:</p> $\pi_1 = 50.86\%$ $\alpha = 4.88 \frac{\%}{1000d}$ $\lambda = 3.24$	<p>obtained from PEC</p> <p>obtained by linear regression PEC</p> <p>obtained from Cox regression</p>

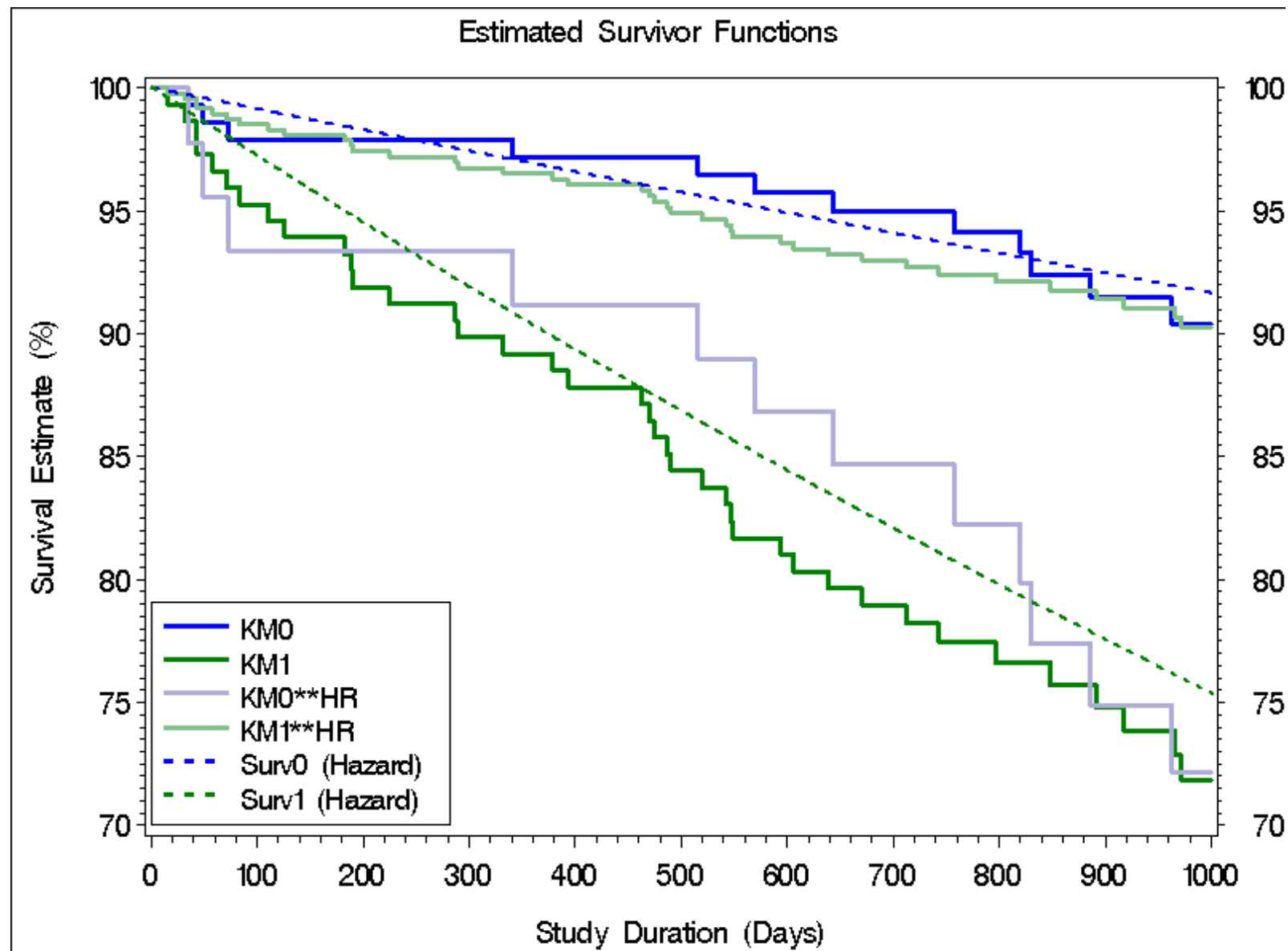
Estimate PEC by KM-method



Estimate explicit hazard



Survival Estimates



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For binary covariates Evolution Charts + Cox offer

- A simple check of the proportional hazard assumption (necessary condition: monotonic course of PEC)
- May be used to achieve explicit estimates of the hazard function

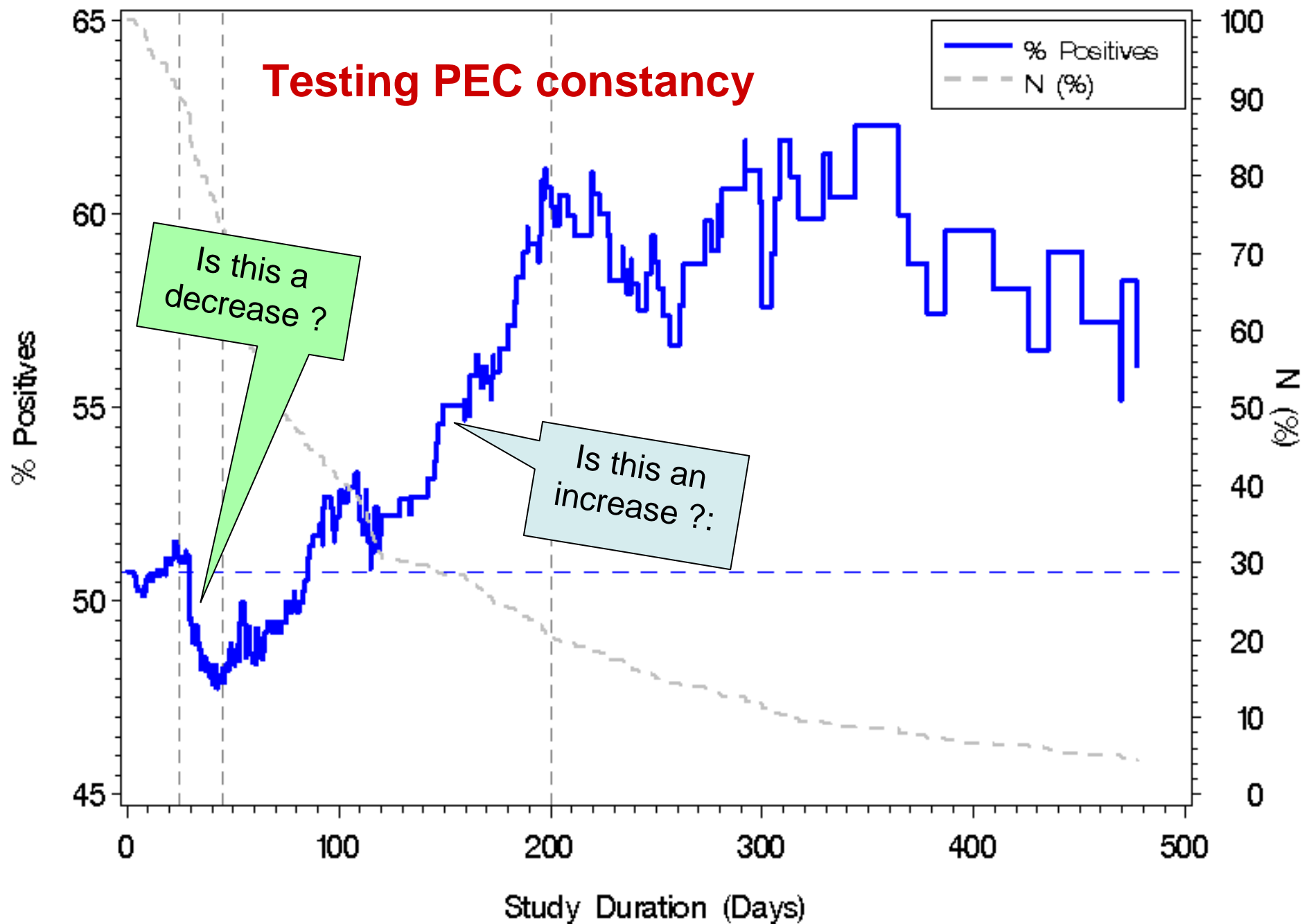
Testing options for Evolution Charts

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Global Tests

Testing on constancy of Population Evolution Charts (PEC) Binary covariate case : overall test	
H₀: PEC overall constant Test variant 1	H ₀ : $\Psi_X(t) := P(X = 1 T > t) = \text{const.}$ Equivalent to H ₀ : $S_1(t) = S_0(t)$ Testing as usual e.g. by logrank test
H₀: PEC overall constant Test variant 2	H ₀ : $\Psi_X(t) := P(X = 1 T > t) = \text{const.}$ Can be tested by the Wald-Wolfowitz runs test <u>Warning: doesn't work with ties in the time variable !!</u>
PECs come as graphics	PECs do not need difficult assumptions

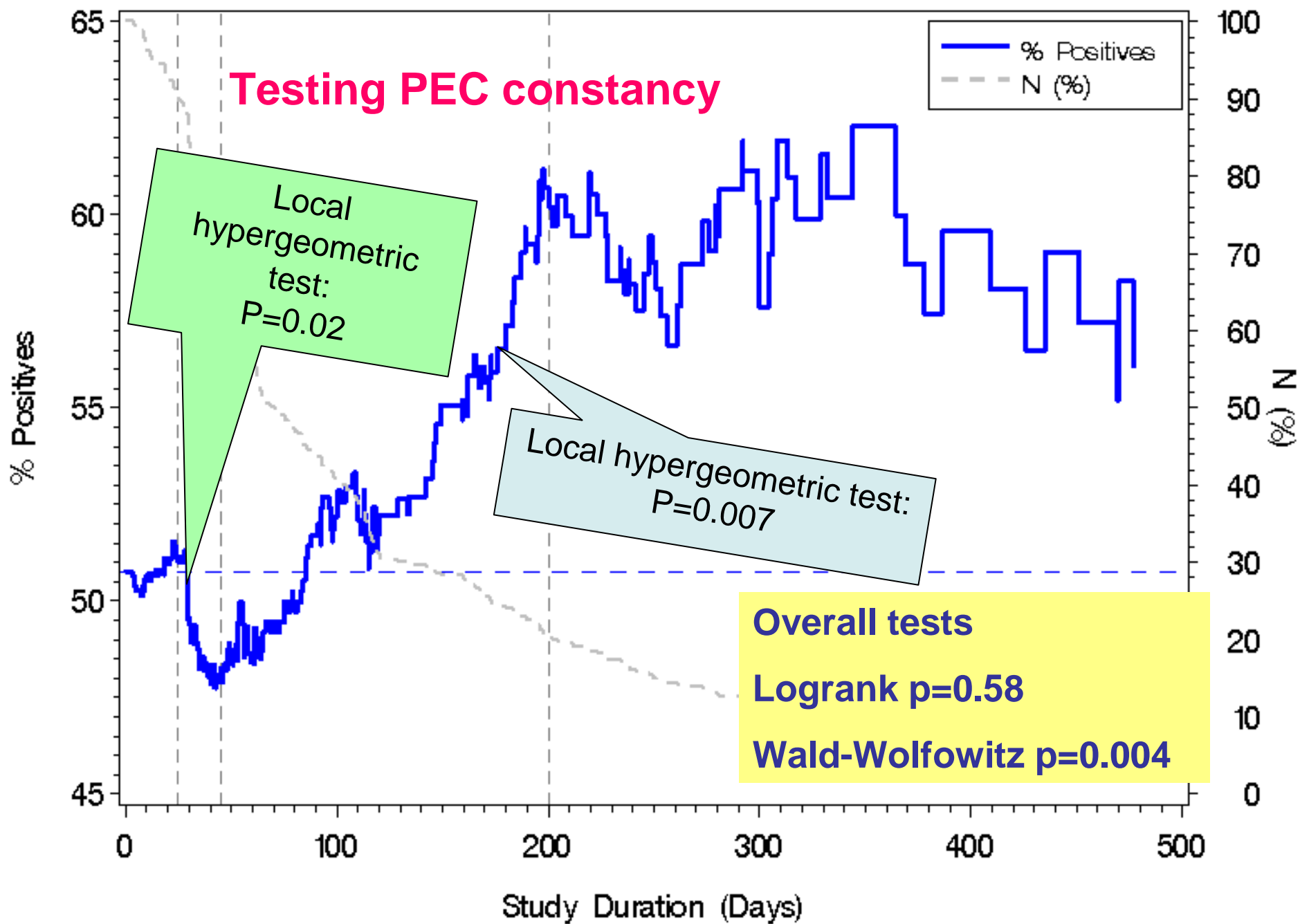
Population Evolution Chart



Local Tests

Testing on constancy of Population Evolution Charts (PEC) Binary covariate case: local hypotheses	
H₀: PEC constant between t₁ and t₂	H ₀ : $\Psi_X(t_1) = \Psi_X(t_2) = \Psi_X(t) \quad t_1 < t < t_2$
H₀: local	Testing: may be based on hypergeometric probability
H₀: PEC regarding CENSORING	Interchange the roles of event times T and censoring times C

Population Evolution Chart



Local Tests/Details

Testing on constancy of Population Evolution Charts (PEC) Binary covariate case: local hypotheses	
H₀: PEC constant between t₁ and t₂	H ₀ : $\Psi_X(t_1) = \Psi_X(t_2) = \Psi_X(t) \quad t_1 < t < t_2$
Assume: censoring and covariate independent	Testing: may be based on all times Use: hypergeometric probability But: compromised power
Exact approach	Testing: based on event times conditioned on the censored times in the interval t ₁ to t ₂ But: complicated algorithm
Approximate: Exact approach	Hyp. Prob1: #censor shifted to t ₁ . Hyp. Prob2: #censor shifted to t ₂ . Geom. mean of Prob1 and Prob2

For binary covariates Evolution Charts offer....

- Global and local test opportunities to assess changes in the time dynamic

What to do with metric covariates

Define Evolution Chart for metric covariates

Definitions of Evolution Charts $\Psi_X(t)$	
I. Density Base	$f(x T > t)$
II. Selection process	$\frac{1}{S(t)} [f(x T = 0) - (1 - S(t)) \cdot f(x T \leq t)]$
PECs come as graphics	PECs do not need difficult assumptions

Difficulties

- How to follow up a conditional distribution over time graphically?
- Choose some property of the distribution and follow up over time - choices

Evolution Chart for Metric Covariates

Choices	
Take one or several quantiles of the distribution	Reduces the problem to the binary case Quantile Evolution Chart (QuEC)
Choose a moment of the distribution e.g. the mean	Moment Evolution Chart (MEC)
Estimate a MEC based on the selection process view	$\frac{1}{\hat{S}(t)} \left[\hat{\mu}(x T = 0) - (1 - \hat{S}(t)) \cdot \hat{\mu}(x T \leq t) \right]$
MECs come as graphics	MECs do not need difficult assumptions

For metric covariates Evolution Charts offer

- Cutoff-free representation of dependencies
- Time-dynamic is made obvious
- Can suggest potential cutoffs

Published PEC

- [1] Moecks J., Franke W., Ehmer B., Quarder O. (1997): Analysis of Safety Database for Long-Term Epoetin- β Treatment: A Meta-Analysis Covering 3697 Patients. *In: Koch & Stein (editors) "Pathogenetic and Therapeutic Aspects of Chronic Renal Failure"*. Marce Dekker: New York, 163-179.
- [2] Moecks J., Köhler W., Scott M., Maurer J., Budde M., Givens S.(2002): Dealing with Selective Dropout in Clinical Trials. *Pharmaceutical Statistics, Vol 1: 119-130*
- [3] Moecks J., Koehler W.(2007): Population Evolution Charts: A Fresh Look to Time-To-Event Data. *Statistical Computing and Graphics, Vol. 18, No 2, Dec 2007, p. 12-19*

Drop me a line to receive a better copy
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thanks for your attention

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