

Part IV – Extensions: Competing Risks Endpoints and Non-Parametric AUC(t) Estimation



- Patrick J. Heagerty PhD
- Department of Biostatistics
- University of Washington

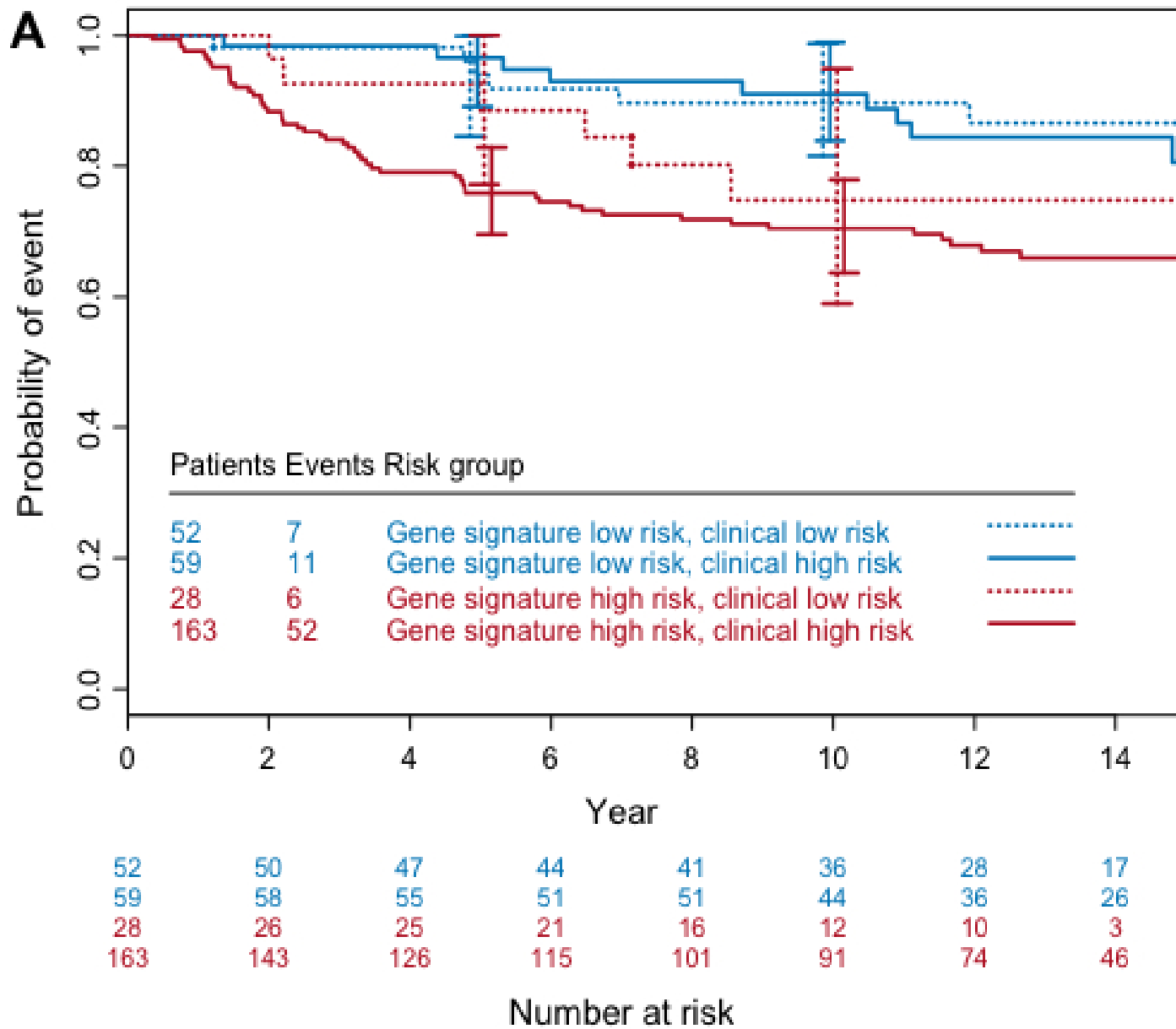
Session Four Outline

- Examples
 - ▷ Breast Cancer: 70 gene prediction / validation
 - ▷ HIV: markers of disease progression
- Competing Risks Data
- TP^C and cause-specific endpoints / Estimation (non-parametric)
- TP^I and cause-specific endpoints / Estimation (semi-parametric)
- Illustration / Software

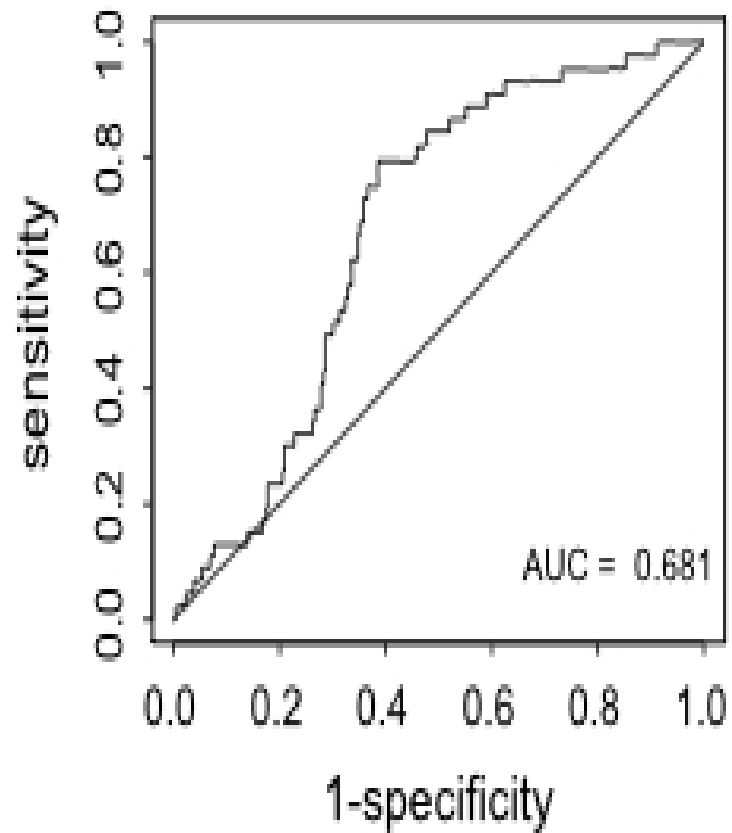
Example: BC and 70-gene Signature among Node-negative

Breast Cancer Prediction

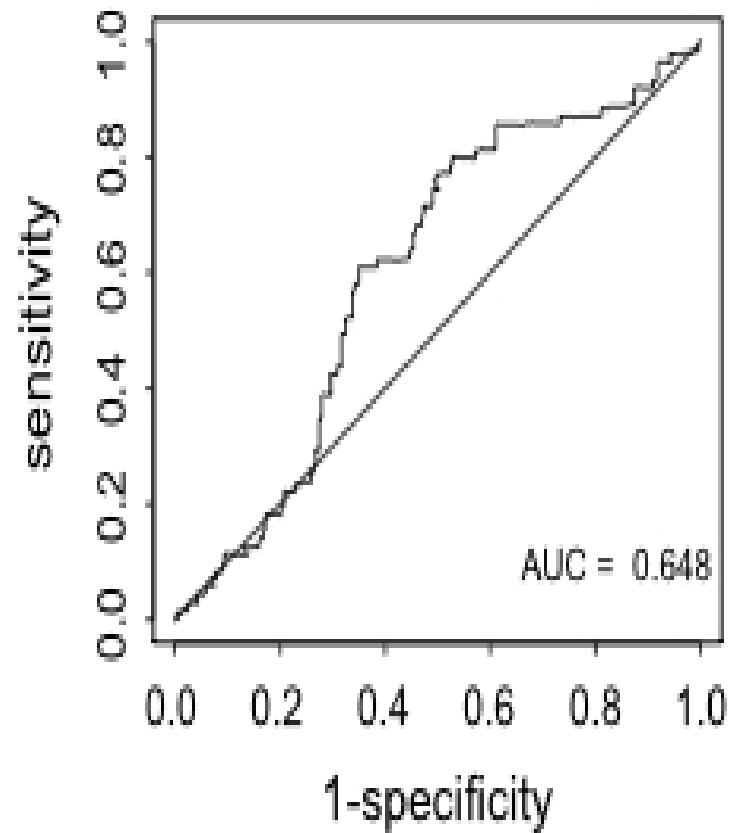
- $N = 307$ women from (5) Euro Centers
- Endpoint(s):
 - ▷ time-until-distant-metastases (next slide)
 - ▷ disease-free-survival
- Predictive measurements:
 - ▷ Clinicopathologic risk assessment
 - ▷ 70-gene Signature
- Goal: validate (added) utility of “signature”
- Buyse et al. (2006) *JNCI*



Gene signature score, for time to distant metastases at 5 years



Gene signature score, for overall survival at 10 years

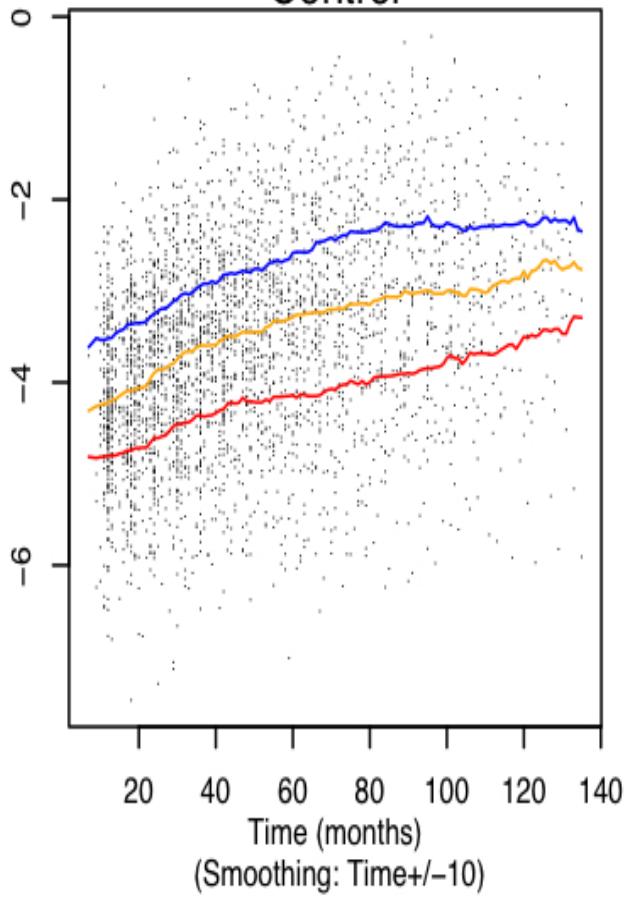


Example: Immune markers and disease progression

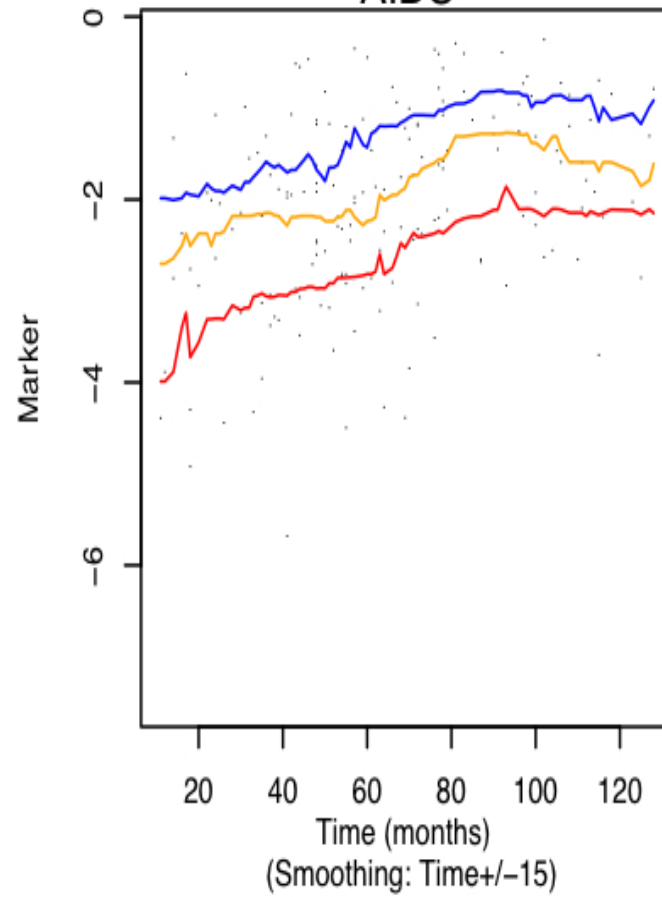
Multicenter AIDS Cohort Study

- $N = 447$ men observed to seroconvert
- Endpoint(s):
 - ▷ time-until-AIDS
 - ▷ time-until-death
- Predictive measurements:
 - ▷ CD4, CD8 at “baseline”
 - ▷ CD4, CD8 measured every 6 months
- Goal: evaluate markers as predictors of disease-progression
- Saha and Heagerty (2010)

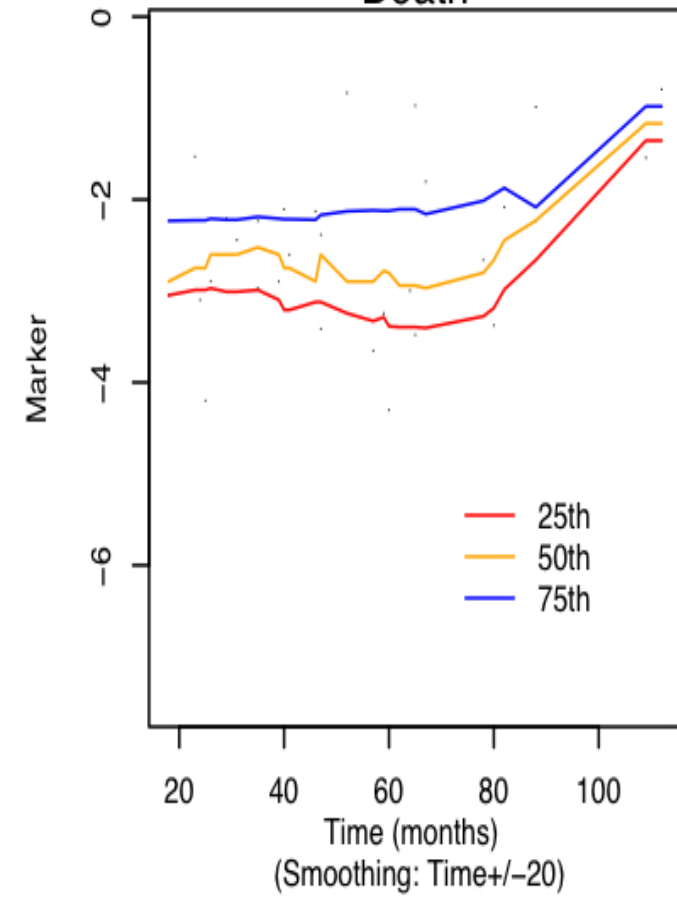
Control



AIDS



Death



Competing Risks Endpoints

- Observed time-until-event, and **type** of event.
 - ▷ Death, cause = (BC, other)
- “Derived” time-until-first-event, and type of event
 - ▷ Time until progression or Death (first event, type)
 - ▷ e.g. metastases, death (without metastases first)
 - ▷ e.g. AIDS, death (without AIDS first)
- **Representation**
 - ▷ (T_i^*, δ_i) where $\delta_i = 0, 1, 2, \dots, C$
 - ▷ δ_i : censored = 0; types = 1, 2, ... C

Sensitivity and Specificity for Survival (again!)

Let T denote the survival time, and let $N(t)$ denote the counting process for the uncensored outcome:

$$N(t) = 1(T \leq t)$$

Possible definitions:

$$\begin{aligned} \text{CASE}(t) &: \begin{cases} \text{Cumulative} & N(t) = 1 \\ \text{Incident} & dN(t) = 1 \end{cases} \\ \text{CONTROL}(t) &: \begin{cases} \text{Static} & N(t^*) = 0 \\ \text{Dynamic} & N(t) = 0 \end{cases} \end{aligned}$$

- Where t^* is a fixed “large” time, $t^* \gg t$.

Sensitivity and Specificity for Cause-specific Survival

Define:

$$\text{sensitivity}^{\mathbb{C}}(c, t; \mathbf{d}) : P(M > c \mid T \leq t; \delta = \mathbf{d})$$

$$\text{specificity}^{\mathbb{D}}(c, t) : P(M \leq c \mid T > t)$$

- “Cases” are broken into finer groups based on the **type** of case.
- e.g. high marker given metastases by time t ($d=1$)
- e.g. high marker given death w/o metastases by time t ($d=2$)

Sensitivity and Specificity for Cause-specific Survival

Example: $d=1, 2$

Case **1** : $T_i \leq t, \delta = \mathbf{1}$

Case **2** : $T_i \leq t, \delta = \mathbf{2}$

Control : $T_i > t, \delta = [1, 2]$

$$TP_t^{\mathbf{C}}(c, \mathbf{1}) = P(M > c \mid T_i \leq t, \delta = \mathbf{1})$$

$$TP_t^{\mathbf{C}}(c, \mathbf{2}) = P(M > c \mid T_i \leq t, \delta = \mathbf{2})$$

$$FP_t^{\mathbf{D}}(c) = P(M > c \mid T_i > t, \delta = [1, 2])$$

Estimation: Using “local” Cumulative Incidence

- **Cause-specific Cumulative Incidence**

- ▷ $C_d(t) = P(T \leq t; \delta = d)$

- ▷ Percent of population with event of type **d** by time t .

- Non-parametric estimation (K&P 1980, p. 168)

$$\hat{C}_d(t) = \sum_{s \leq t} \hat{S}(s-) \cdot \hat{\lambda}_d(s)$$

Estimation: Using “local” Cumulative Incidence

- Cumulative incidence estimator can handle censoring.
- Parallel the estimation of HLP(2000) using:

$$P(M > c \mid T \leq t, \delta = d) = \frac{P(M > c, T \leq t, \delta = d)}{C_d(t)}$$

$$\begin{aligned} \text{numerator} &= \int_c^\infty P(T \leq t, \delta = d \mid M = m) \cdot P(M = m) dm \\ &= \int_c^\infty C_d(t \mid M = m) \cdot P(M = m) dm \end{aligned}$$

Estimation: Using “local” Cumulative Incidence

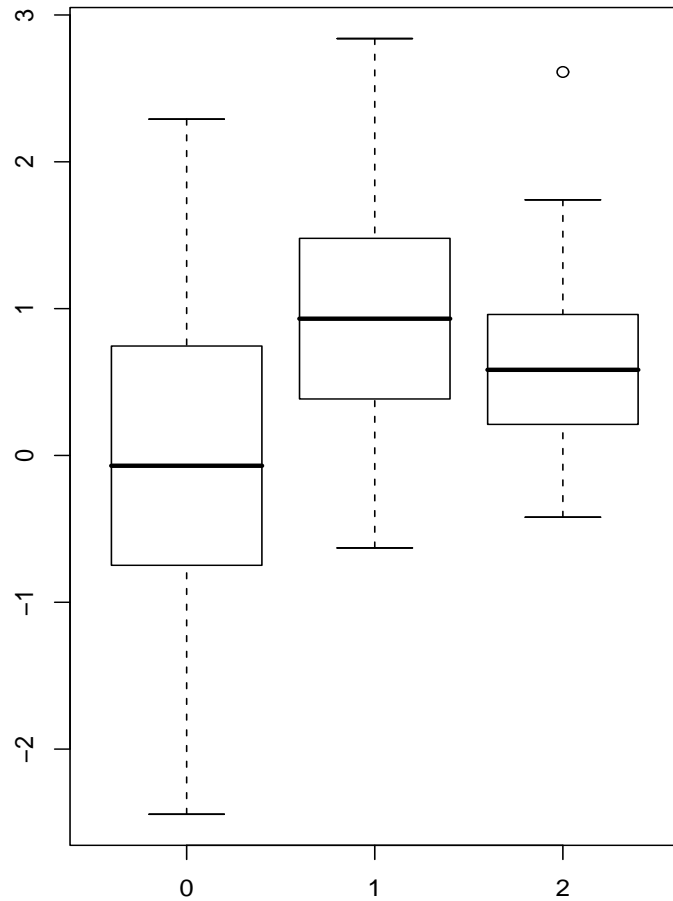
- Use local cause-specific cumulative incidence to estimate $C_d(t | M = m)$ and use empirical for $P(M = m)$.

- **Note:**

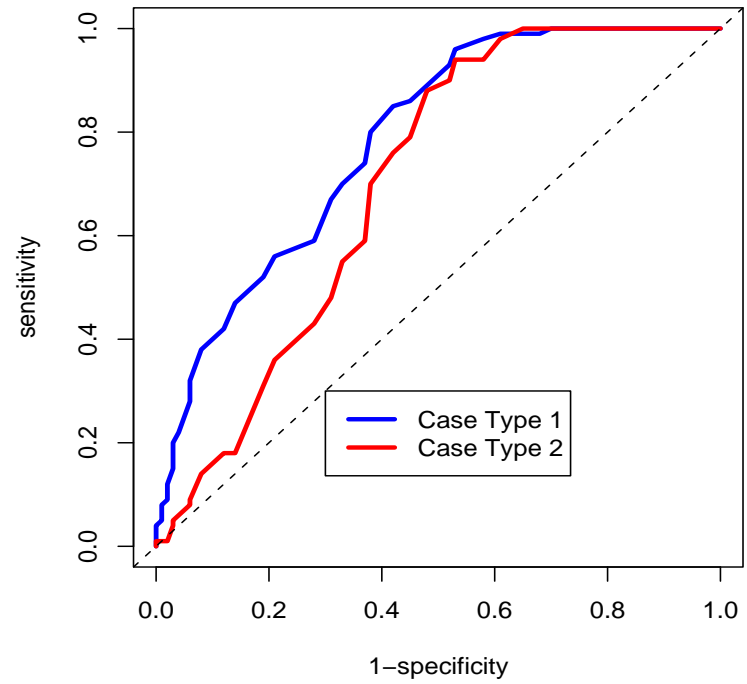
$$P(T > t | M = m) = 1 - \sum_d P(T \leq t, \delta = d | M = m)$$

- Use above to estimate $FP_t^{\mathbb{D}}(c)$ such that joint distribution is proper.

Marker versus Disease status

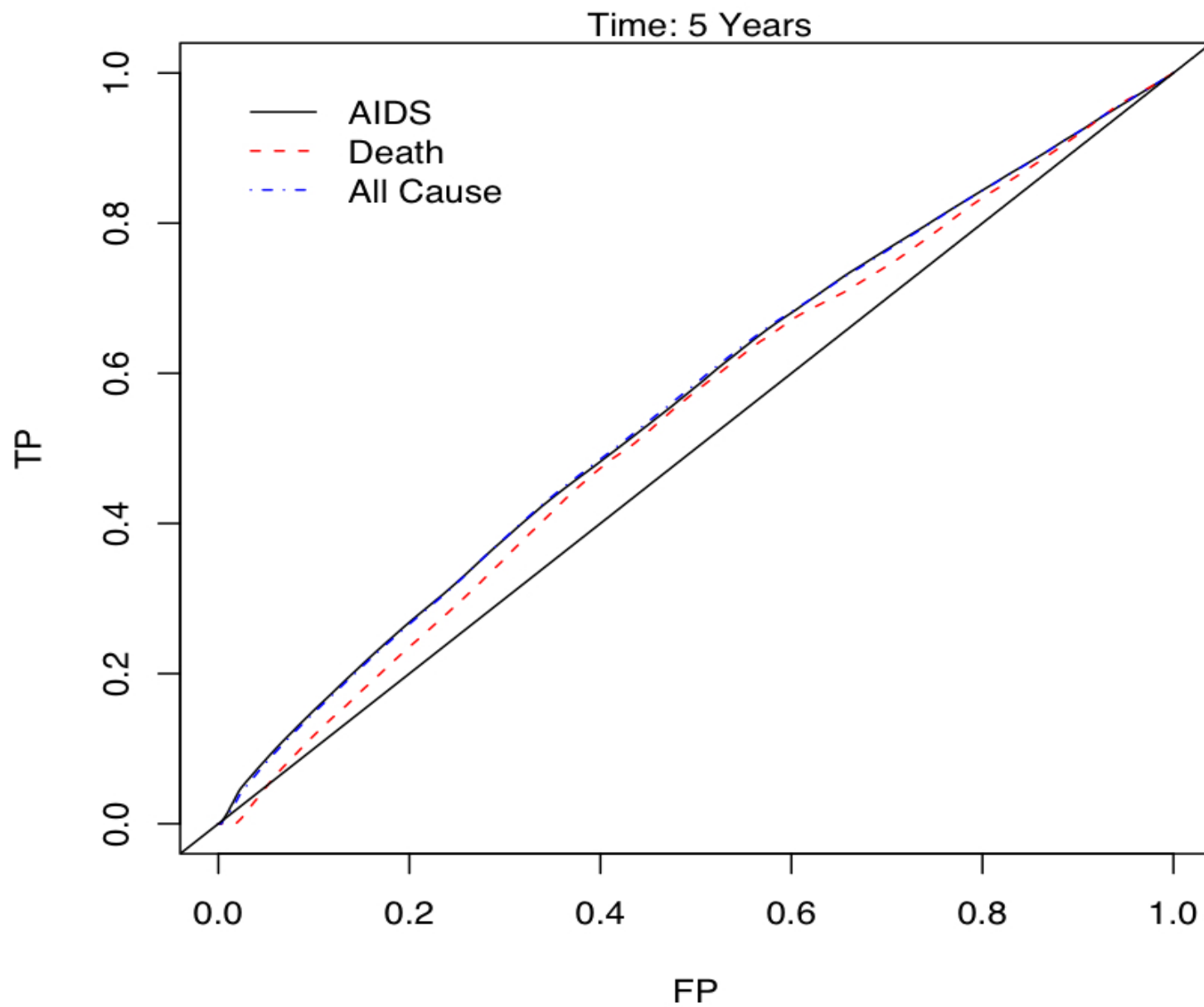


ROC curves



Software / Illustration

- **Software:** CRAN package for R called `survivalROC` – we have extended this to implement the competing risks calculations. (P. Saha)
- **MACS Data**
 - ▷ Baseline (e.g. seroconversion time) values of CD4 and CD8
 - ▷ Linear combination based on Cox regression
 - ▷ Case Type 1 = AIDS
 - ▷ Case Type 2 = death before AIDS
 - ▷ Time for cumulative case status = 5 years



Review: Sensitivity and Specificity for Survival

Define: Heagerty and Zheng (2005) / Saha and Heagerty (2010)

$$\begin{aligned} \text{sensitivity}^{\mathbb{I}}(c, t) &: P[\mathbf{M}(t) > c \mid T = t] \\ &P[\mathbf{M}(t) > c \mid dN(t) = 1] \end{aligned}$$

$$\begin{aligned} \text{specificity}^{\mathbb{D}}(c, t) &: P[\mathbf{M}(t) \leq c \mid T > t] \\ &P[\mathbf{M}(t) \leq c \mid N(t) = 0] \end{aligned}$$

$$TP_t^{\mathbb{I}}(c) = P[\mathbf{M}(t) > c \mid \mathbf{dN}(t)=1]$$

$$FP_t^{\mathbb{D}}(c) = P[\mathbf{M}(t) > c \mid N(t) = 0]$$

Sensitivity and Specificity for Cause-specific Survival

Define:

$$\text{sensitivity}^{\mathbb{I}}(c, t; \mathbf{d}) : P(M > c \mid T = t; \delta = \mathbf{d})$$

$$\text{specificity}^{\mathbb{D}}(c, t) : P(M \leq c \mid T > t)$$

- “Cases” are broken into finer groups based on the **type** of case.
- e.g. high marker given metastases **at** time t ($d=1$)
- e.g. high marker given death w/o metastases **at** time t ($d=2$)

Sensitivity and Specificity for Cause-specific Survival

Example: $d=1, 2$

Case **1** : $T_i = t, \delta = \mathbf{1}$

Case **2** : $T_i = t, \delta = \mathbf{2}$

Control : $T_i > t, \delta = [1, 2]$

$$TP_t^{\mathbb{I}}(c, \mathbf{1}) = P(M > c \mid T_i = t, \delta = \mathbf{1})$$

$$TP_t^{\mathbb{I}}(c, \mathbf{2}) = P(M > c \mid T_i = t, \delta = \mathbf{2})$$

$$FP_t^{\mathbb{D}}(c) = P(M > c \mid T_i > t, \delta = [1, 2])$$

Estimation: Hazard as Bridge

A general definition for the cause-specific hazard is

$$\lambda^{(d)}(t | M_i) = \frac{P(T_i = t, \delta_i = d | M_i)}{P(T_i \geq t | M_i)}$$

Then using a little algebra yields

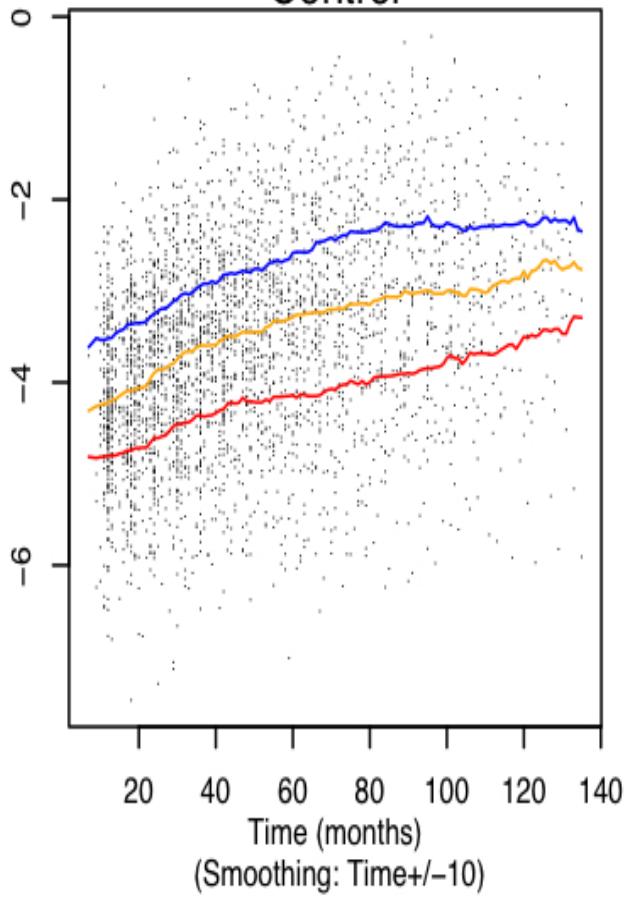
$$P(M_i = m | T_i = t, \delta_i = d) \propto \underbrace{\lambda^{(d)}(t | M_i = m)}_{\text{Estimate}} \cdot \underbrace{P(M_i = m | T_i \geq t)}_{\text{Smooth model} + \text{Empirical}}$$

Note: direct (easy) generalization of the HZ(2005) methods.

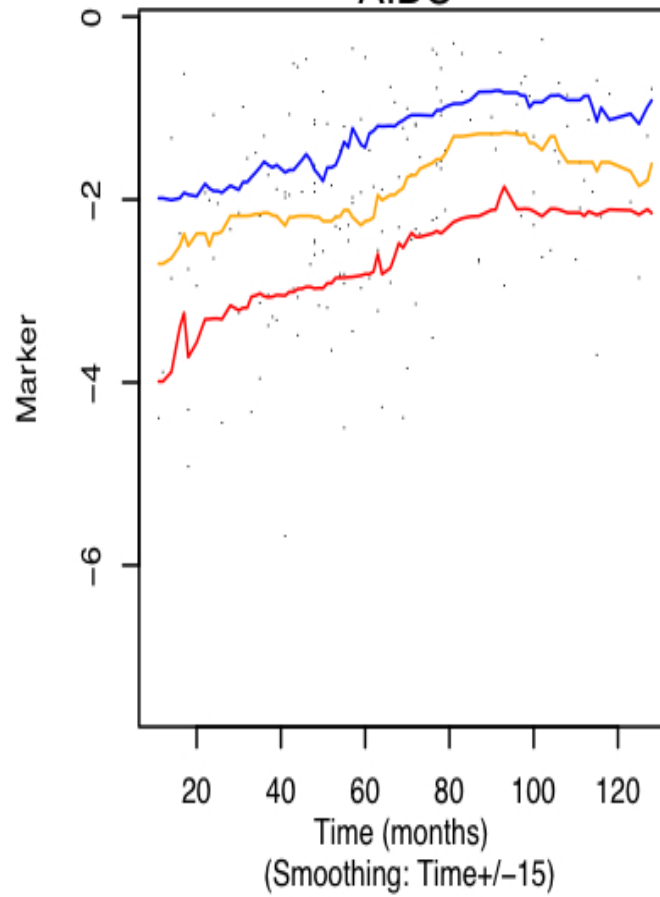
Software / Illustration

- **Software:** CRAN package for R called `risksetROC` – we have extended this to implement the competing risks calculations, and to handle time-dependent covariates. (P. Saha)
- **MACS Data**
 - ▷ Longitudinal values of CD4 and CD8
 - ▷ Linear combination based on Cox regression
 - ▷ Case Type 1 = AIDS (n=176)
 - ▷ Case Type 2 = death before AIDS (n=34)
 - ▷ ROC curve, and AUC versus time

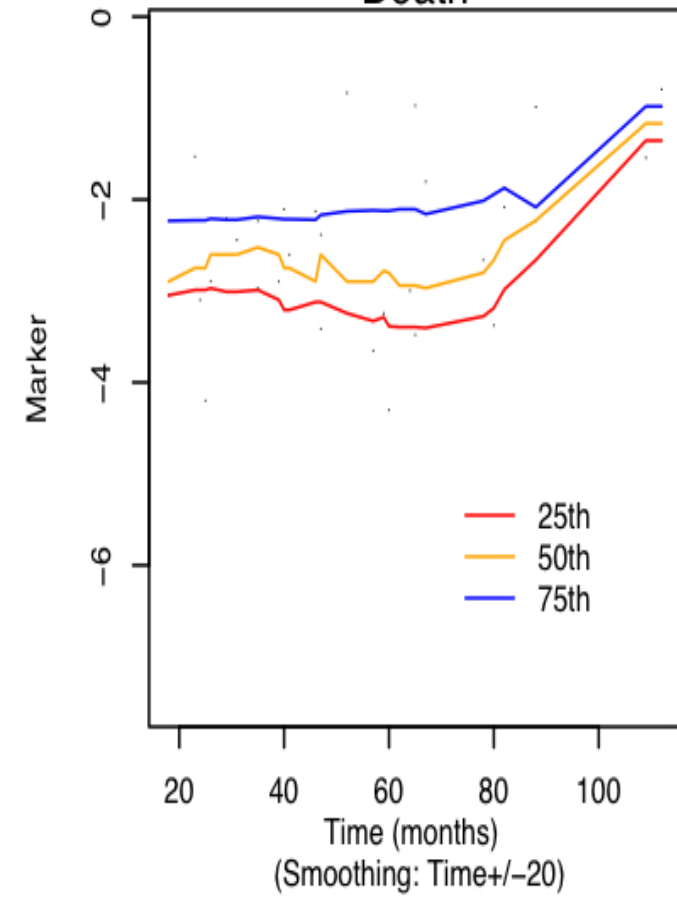
Control

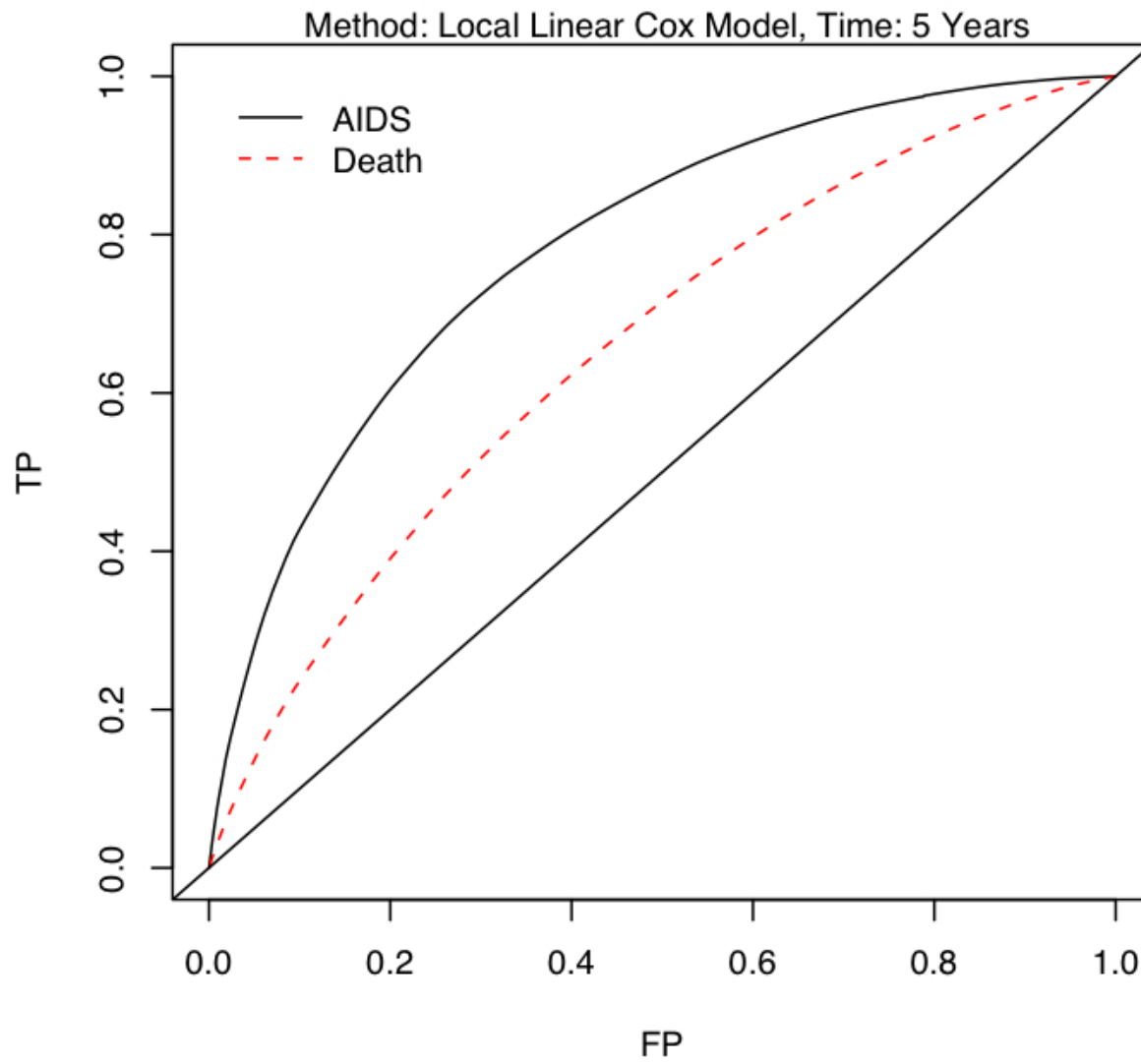


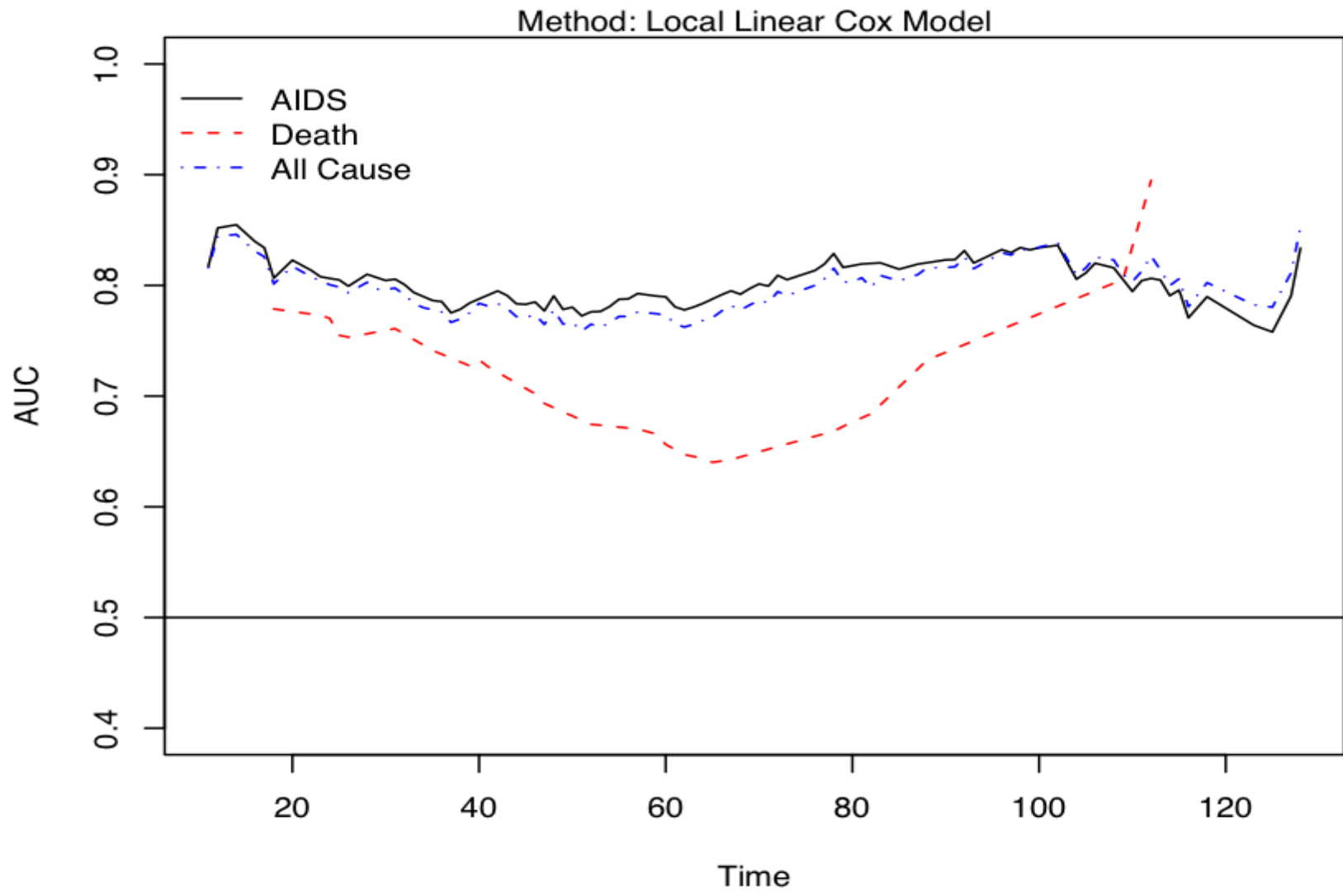
AIDS



Death







Summary

- Extension of time-dependent ROC methods to competing risks data.
- Cumulative Cases – uses non-parametric methods based on local cumulative incidence calculations.
- Incident Cases – uses semi-parametric methods that parallel those outlined in Heagerty and Zheng (2005).
- Time-dependent markers.

Thanks!



Nonparametric Estimation of $AUC(t)$

Risk set ranks can be used to nonparametrically estimate $AUC(t)$.

$$AUC(t) = P[M_j > M_k \mid dN_j(t) = 1, N_k(t) = 0]$$

$$M^*(t) = M_j \text{ for } dN_j(t) = 1 \quad \mathcal{C}(t) = \{k : N_k(t) = 0\}$$

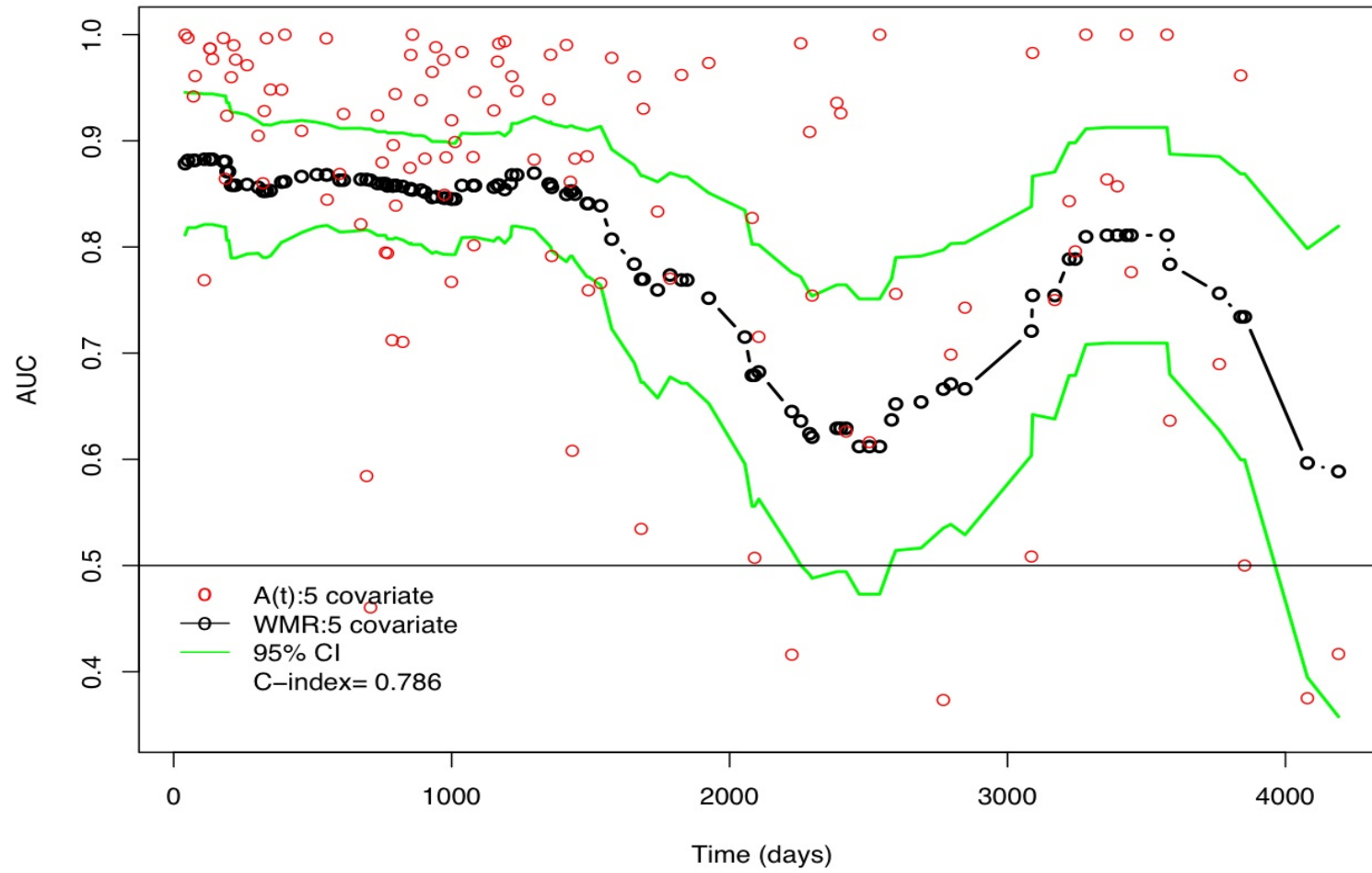
$$AUC(t) = E \{ 1[M^*(t) > M_k \mid k \in \mathcal{C}(t)] \}$$

$$\widehat{AUC}(t) = \frac{1}{n_t} \sum_{k \in \mathcal{C}(t)} 1[M^*(t) > M_k]$$

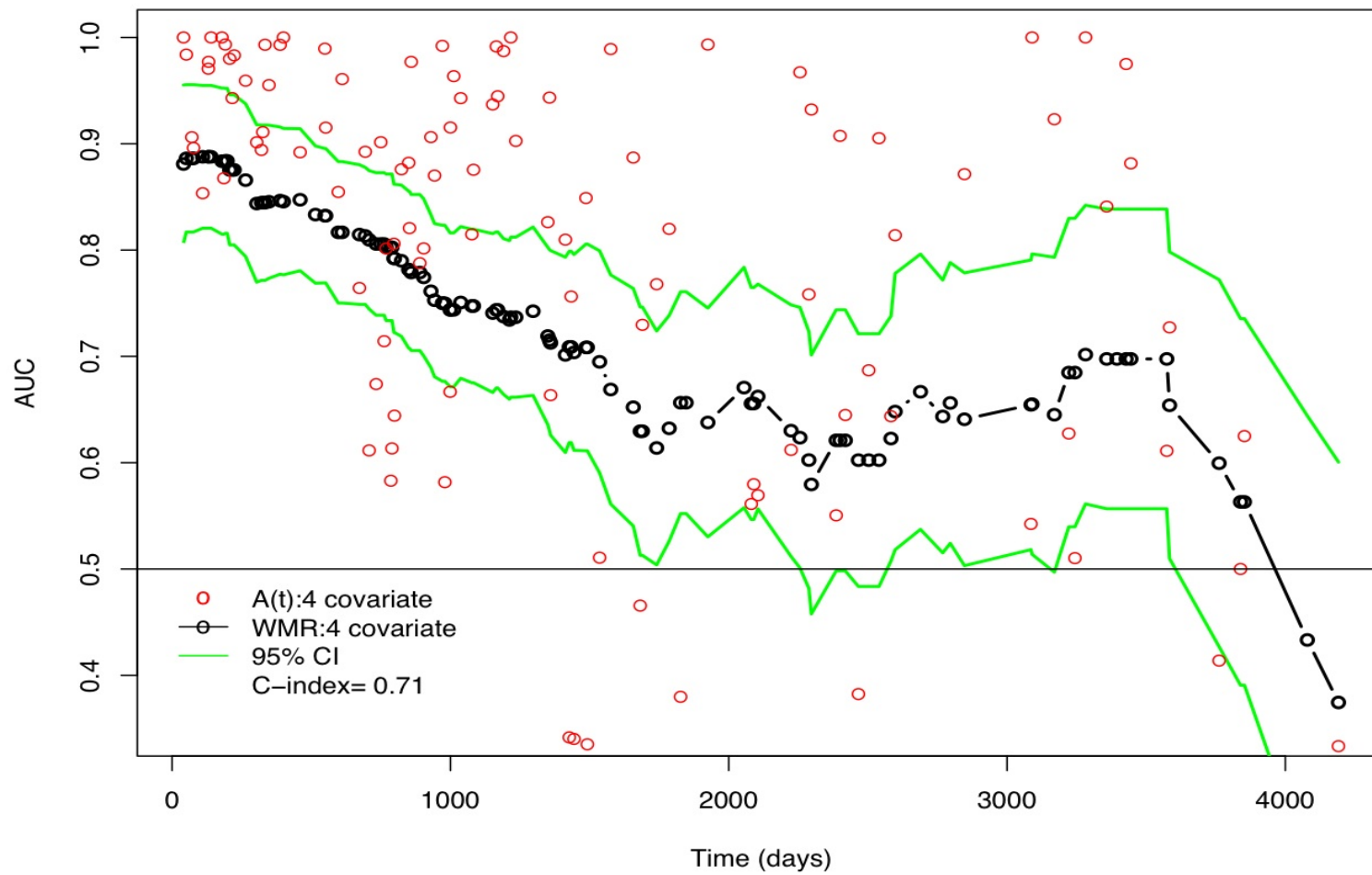
$$\widehat{AUC}(t) = \frac{[\text{risk set rank of } M^*(t)] - 1}{n_t}$$

Saha & Heagerty (submitted)

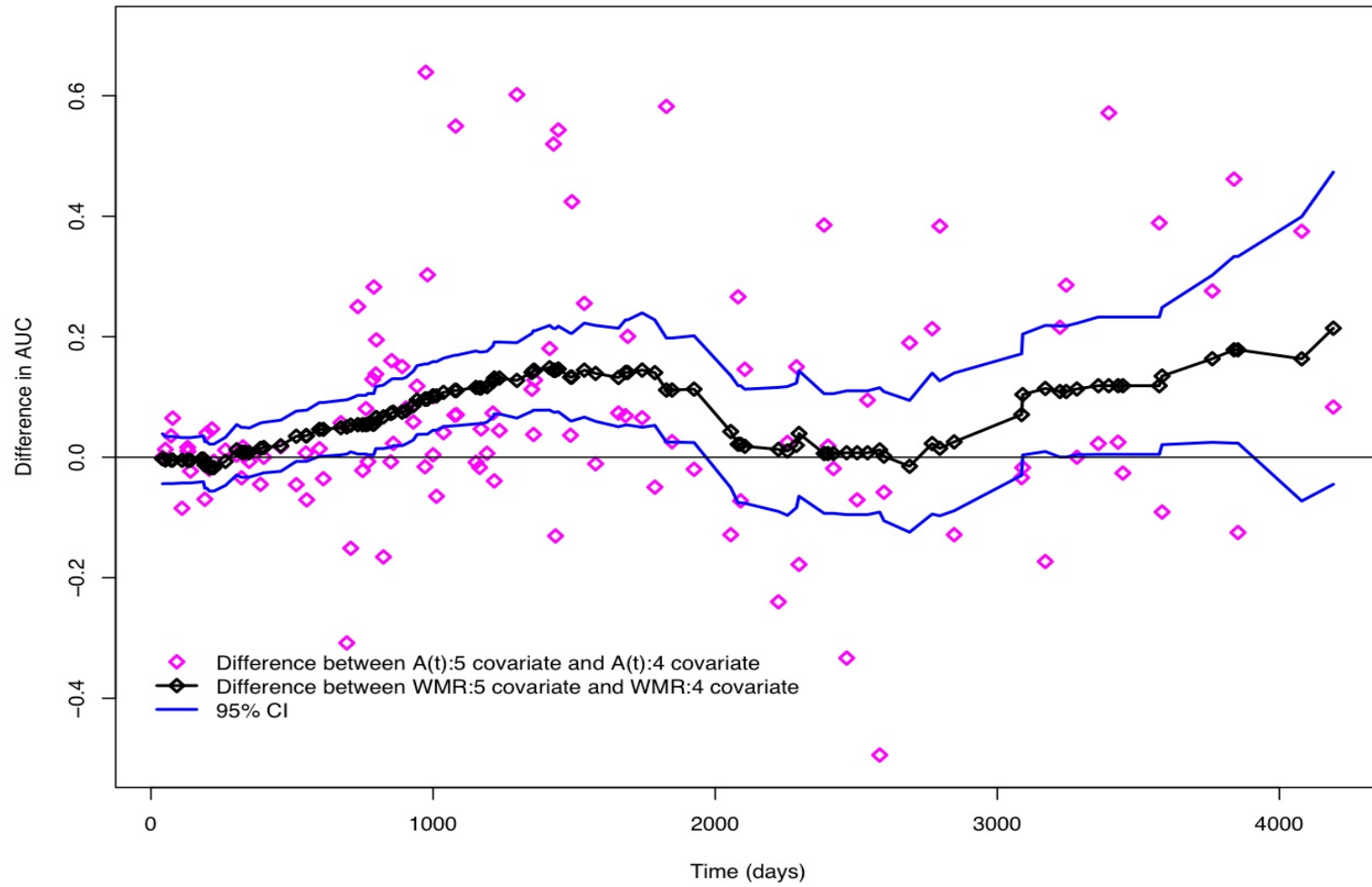
AUC based on five-covariate Model



AUC based on four-covariate Model



Difference in AUC between five-covariate model and four-covariate model



AUC Based on Risk Set Rank

