

# Net time-dependent ROC curves: a solution for evaluating the accuracy of a marker to predict disease-related mortality

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## Prognostic markers of all-cause mortality, essential to :

- identify subjects at high-risk of death
  - optimize healthcare management
- ⇒ The capacity of a score to predict all-cause deaths is evaluated by using **time-dependent ROC curves**<sup>1</sup>

## Limits

- an important part of the mortality not related to the chronic disease
  - impossibility to identify excess deaths
- ⇒ Solution : distinguish between the expected mortality and the excess mortality, by using **additive net survival model**.

<sup>1</sup>Heagerty et al., Biometrics, 2000.

## Objective

- Evaluate the capacity of a marker to predict disease-specific mortality, deaths for which medical specialists can act

## Notations

- $X$  : Random variable representing the prognostic marker
- $x_j$  : Observation for the subject  $j$
- $n$  : Sample size,  $j = 1, \dots, n$
- $T_{Ej}$  : Time to death related to the disease
- $T_{Pj}$  : Time to expected death
- $T_j = \min(T_{Ej}, T_{Pj})$  : Time of the death
- $C_j$  : Time of the last follow-up point (right censoring)
- $H(t)$ ,  $H_E(t)$  and  $H_P(t)$  : Cumulative hazard functions of  $T$ ,  $T_E$  and  $T_P$  at time  $t$

## New estimator : net time-dependent ROC curve

- represents the **net sensitivity** plotted against 1 - the **net specificity** for all the thresholds  $c$  of a marker  $X$

By defining a binary test at the cut-off  $c$ ,

- **Net sensitivity** : proportion of positive test ( $X > c$ ) given that the disease-related death occurs before time  $t$  :

$$se_t(c) = Pr(X > c | T_E \leq t)$$

- **Net specificity** : proportion of negative test ( $X \leq c$ ) given that the disease-related death does not occur before time  $t$  :

$$sp_t(c) = Pr(X \leq c | T_E > t)$$

⇒ **Question** : How estimate the net sensitivity and the net specificity ?

**Lorent et al.** (submitted)

Estimation of the **cumulative  
cause-specific hazard**

$$\hat{H}_E(t)$$

**Pohar et al.** (2011)

**Heagerty et al.** (Biometrics, 2000)

Estimation of the **cumulative hazard**

$$\hat{H}(t)$$

Nearest-neighbor estimator  
**Akritas** (1994)

**Net time-dependent  
ROC curve**

**Time-dependent ROC  
curve**

By adapting the approach of Heagerty (Biometrics, 2000)

the two probabilities can be developed :

- $se_t(c) = \{(1 - G_X(c)) - S_{X,E}(c, t)\} / \{1 - S_{X,E}(-\infty, t)\}$
- $sp_t(c) = 1 - \{S_{X,E}(c, t) / S_{X,E}(-\infty, t)\}$

Estimation of  $S_{X,E}(c, t)$  : bivariate survival function of  $X$  and  $T_E$

⇒ implies to estimate  $H_E(t|X = x_j)$ , can be obtained from :

- $\hat{H}_E(t)^2$
- the calculation of the conditional at-risk and counting process (Use of Akritas estimator<sup>3</sup>)
  - $Y_{jl}^\pi(t) = I(T_l > t, C_l > t, |\hat{G}_X(x_j) - \hat{G}_X(x_l)| < \pi) / S_{Pl}(t)$
  - $N_{jl}^\pi(t) = I(T_l \leq t, C_l \geq T_j, |\hat{G}_X(x_j) - \hat{G}_X(x_l)| < \pi) / S_{Pl}(t)$

<sup>2</sup>Pohar et al., Biometrics, 2011.

<sup>3</sup>Akritas, Annals of Statistics, 1994.

Estimation of  $se_t(c)$  and  $sp_t(c)$ 

- the conditional cumulative excess hazard function is estimated by :

$$\hat{H}_E(t|X = x_j) = \int_0^t \frac{dN_{j\cdot}^\pi(u)}{Y_{j\cdot}^\pi(u)} - \int_0^t \frac{\sum_{l=1}^n Y_{jl}^\pi(u) dH_{Pj}(u)}{Y_{j\cdot}^\pi(u)}$$

⇒ Allows to obtain :

- an estimation of  $S_{X,E}(c, t)$
- an estimation of the net sensitivity and the net specificity for all the thresholds  $c$

⇒ Representation of the net time-dependent ROC curve  
Area under the curve = net AUC

## Objective : validate the proposed estimator

### 3 different scenarios were considered

- Expected ages of death in general population were simulated to establish life tables
  - Excess times-to-death were simulated
- ⇒ Distinction is possible between expected deaths and excess deaths
- ⇒ Calculation of the cause-specific AUC by censoring the expected deaths
- ⇒ Comparison between the traditional AUC, the cause-specific AUC and the net AUC for each sample



# Simulations (2)

Censoring rate	Effective	All-cause AUC <sub>t</sub>	Cause-specific AUC <sub>t</sub>	Net AUC
≈ 0.30	n=100	0.769 (0.049)	0.955 (0.015)	0.891 (0.089)
	n=250	0.774 (0.035)	0.963 (0.008)	0.906 (0.061)
	n=500	0.773 (0.024)	0.963 (0.006)	0.912 (0.049)
	n=1000	0.772 (0.017)	0.964 (0.004)	0.910 (0.038)
≈ 0.50	n=100	0.756 (0.056)	0.945 (0.017)	0.872 (0.094)
	n=250	0.766 (0.034)	0.954 (0.010)	0.886 (0.067)
	n=500	0.764 (0.024)	0.953 (0.007)	0.888 (0.051)
	n=1000	0.765 (0.018)	0.955 (0.005)	0.889 (0.037)
≈ 0.70	n=100	0.747 (0.063)	0.940 (0.020)	0.839 (0.105)
	n=250	0.754 (0.043)	0.941 (0.014)	0.850 (0.073)
	n=500	0.752 (0.032)	0.944 (0.009)	0.846 (0.057)
	n=1000	0.750 (0.019)	0.943 (0.006)	0.843 (0.034)

## Results

- ⇒ The net AUC provide significant correction of the all-cause AUC
- ⇒ The net AUC is closer to the cause-specific AUC

Introduction

Methods

Simulations

Applications

Interest

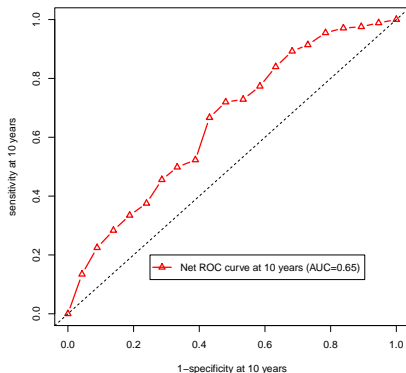
Limits

## Definition

- Disease associated = End stage renal disease
  - Choice treatment = Kidney transplantation
  - Possible trajectories after kidney graft :
    - return to dialysis
    - patient death related to the disease or not. Distinction is often impossible
- ⇒ Endpoint studied in the following applications : patient death related to the disease

Prognostic score of mortality of Hernandez<sup>4</sup>

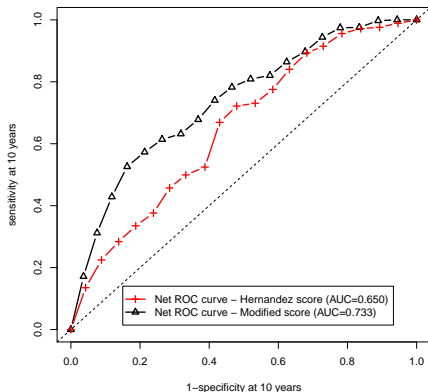
- tested by using DIVAT cohort from Nantes Hospital (n=1230)
  - 10 years prognostic, *net AUC* = 0,65,  $IC_{95}\% = [0,56 - 0,72]$
- ⇒ Difficult to validate the score in the prediction of excess deaths



<sup>4</sup>Hernandez, Transplantation, 2009.

## Other prognostic score of mortality, created from DIVAT cohort

- 10 years prognostic, *net AUC* = 0,73,  $IC_{95}\%$  = [0,64 - 0,80]
- ⇒ Capacity of this score to predict the disease-related mortality : acceptable



## Net time-dependent ROC curve

- useful when attribution of the deaths is impossible
- *net AUC* at time  $t$  is interpretable :  
for two patients randomly selected, probability that the patient with the higher value of the marker dies because of the disease, before the patient with the lower value.
- can be applied to others areas of medicine and biology
- implemented in an R package `ROct` available at :  
<http://www.divat.fr/en/software/roct>

## Limits

- When a distinction is feasible between deaths related to the disease and those that are not  $\Rightarrow$  competing risk model
- When all the observed mortality is related to the disease  $\Rightarrow$  time-dependent ROC curve **OR** net time-dependent ROC curve
- The use of the proposed estimator in the presence of informative censoring  $\Rightarrow$  noticeable effect on the results

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