

Generalized Pairwise Comparison on Immuno-Oncology clinical trial data: a case study

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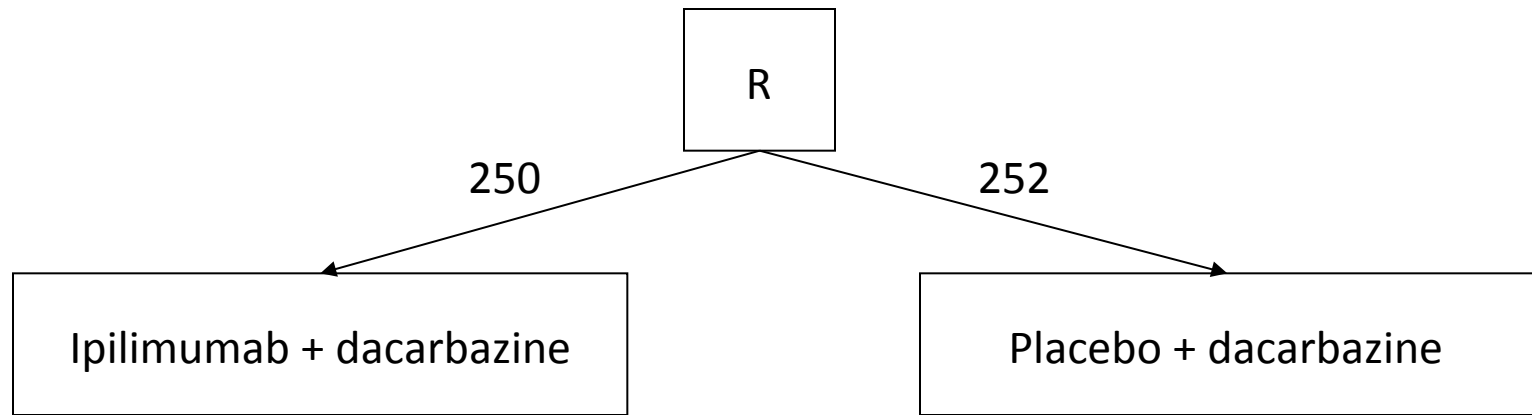
Department of Biostatistics HCL – LBBE UCBL
Department of Medical oncology HCL – LBBE UCBL



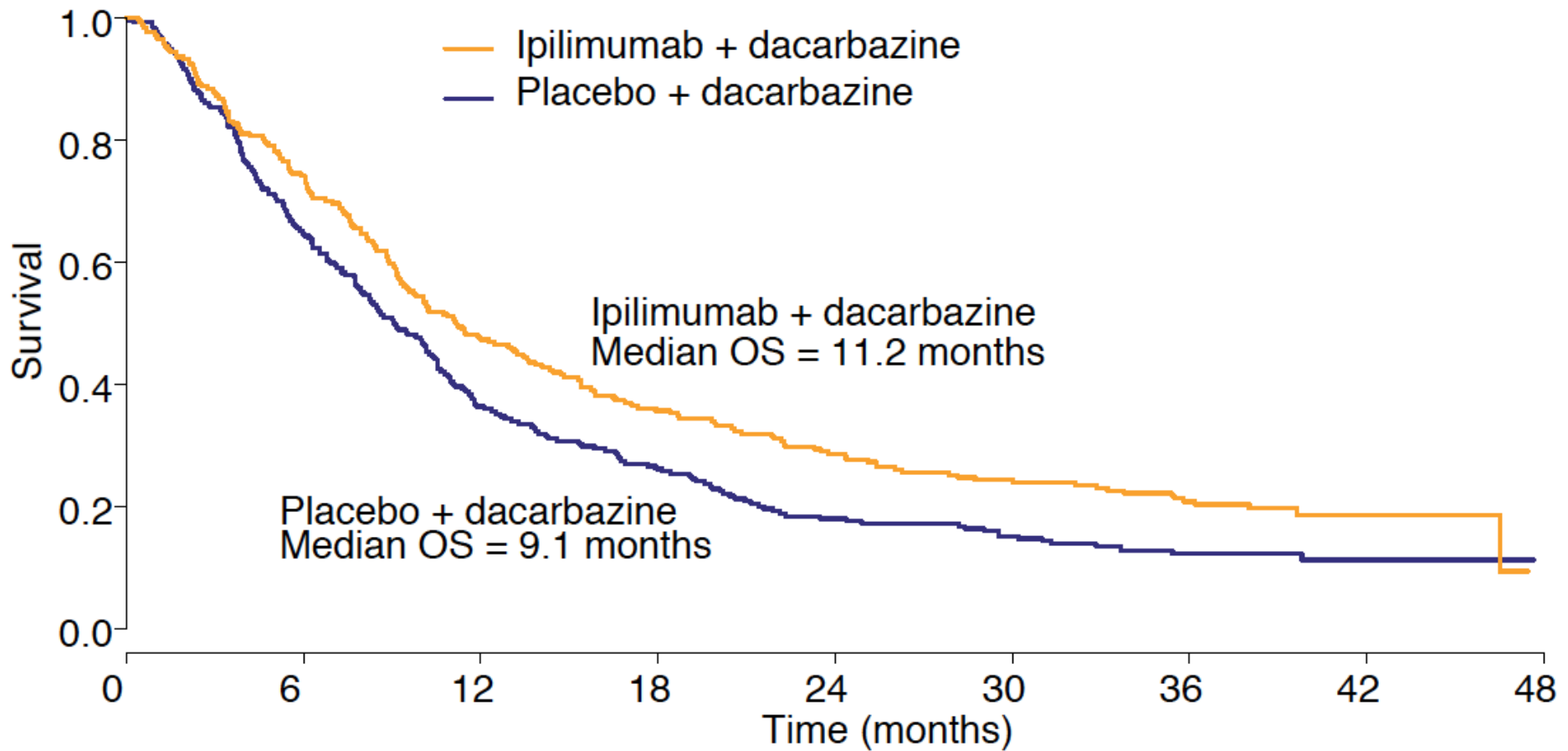
Case study

The CA184-024 trial

502 metastatic melanoma



OS results in the CA184-024 trial

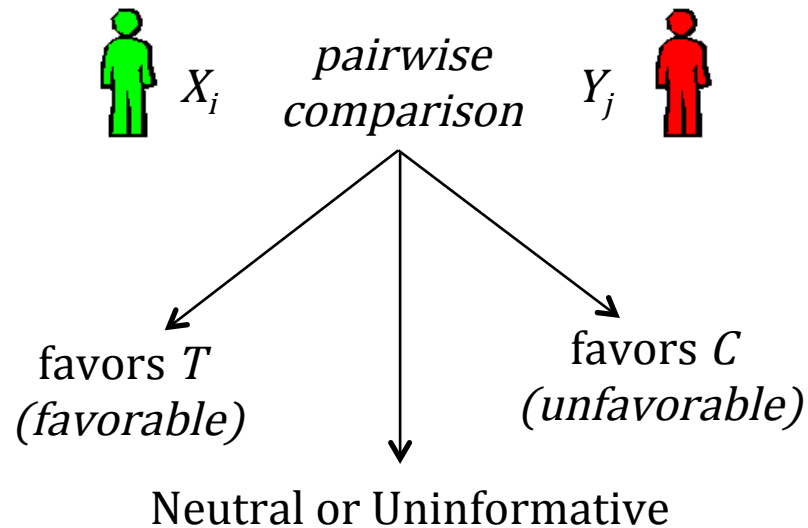
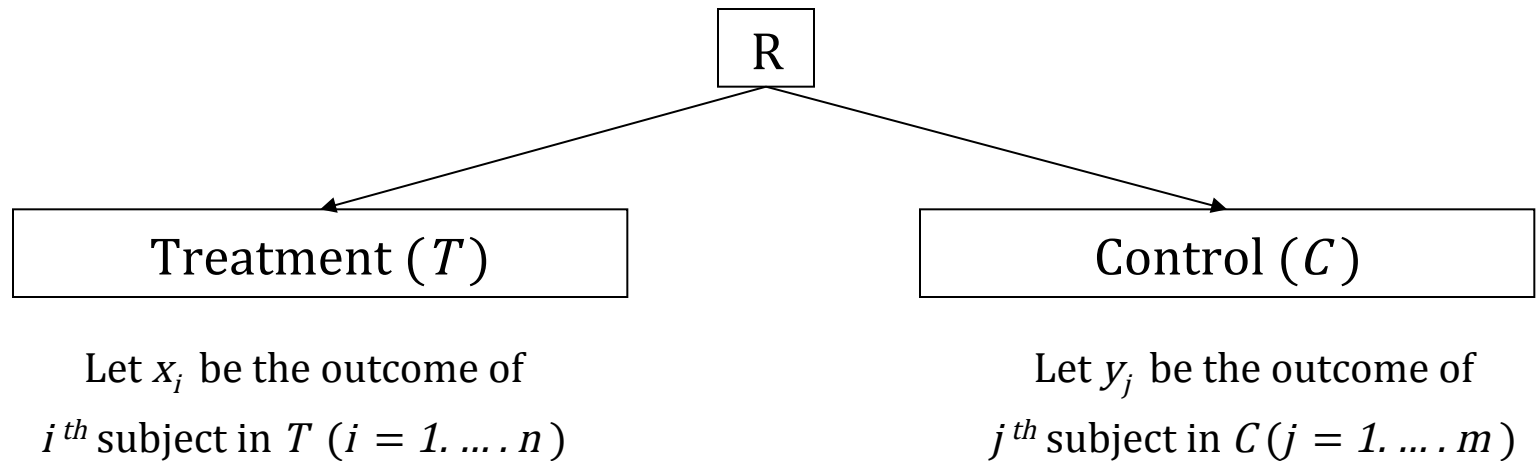


Pcb	252	160	89	64	44	37	26	7	0
Ipi	250	181	114	85	68	57	41	10	0

Outline

- The procedure of generalized pairwise comparisons
- A patient-oriented measure of treatment benefit
- Application on immuno-oncology trials
 - Simulation study
 - Illustration on an ipilimumab trial

Methods – Pairwise comparisons



Methods – Definition of thresholds

Continuous outcome

Pair	Rating
$x_i - y_j > \tau$	Favorable
$x_i - y_j < (-\tau)$	Unfavorable
$ x_i - y_j \leq \tau$	Neutral
x_i or y_j missing	Uninformative

Methods – Standard procedure for pairwise scoring

$$U_{ij} = \begin{cases} +1 & \text{when the pair } (X_i, Y_j) \text{ is favorable} \\ -1 & \text{when the pair } (X_i, Y_j) \text{ is unfavorable} \\ 0 & \text{otherwise} \end{cases}$$

$$\Delta = U = \frac{1}{m \cdot n} \sum_{i=1}^n \sum_{j=1}^m U_{ij}$$

Δ is named «net benefit»

An empirical distribution of Δ can be obtained by permutation

Some notations

- x_i^0 and y_j^0 : time-to-event
- x_i and y_j : time-to-observation

- Event indicator :

$$\delta_i = \begin{cases} 1 & \text{if } x_i = x_i^0 \\ 0 & \text{if } x_i < x_i^0 \end{cases} \quad \left. \vphantom{\begin{matrix} \delta_i \\ \varepsilon_j \end{matrix}} \right\} \text{in group T}$$
$$\varepsilon_j = \begin{cases} 1 & \text{if } y_j = y_j^0 \\ 0 & \text{if } y_j < y_j^0 \end{cases} \quad \left. \vphantom{\begin{matrix} \delta_i \\ \varepsilon_j \end{matrix}} \right\} \text{in group C}$$

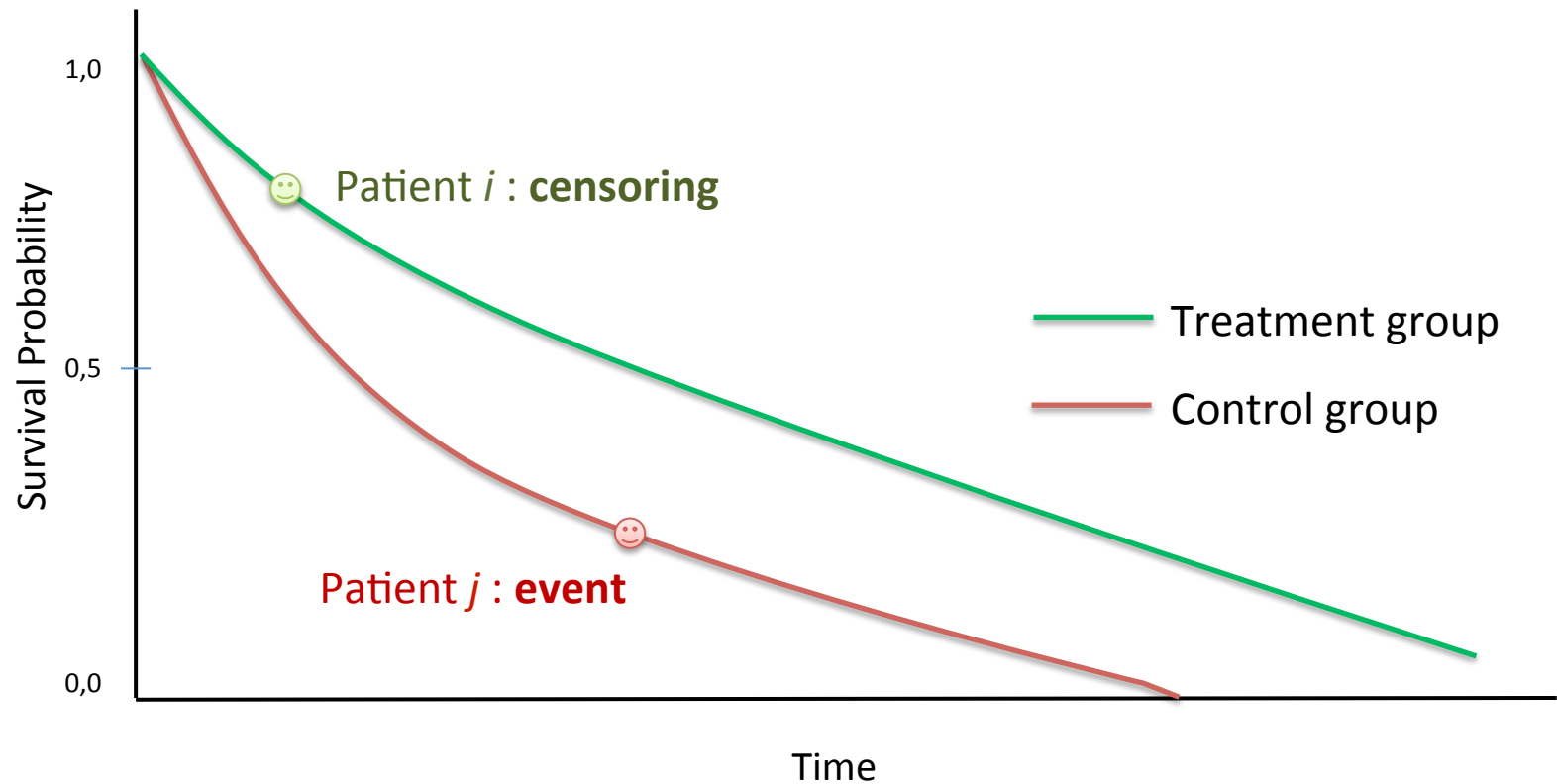
- Survival function:

$$S_T(t) = \mathbb{P} [x_i^0 \geq t] \quad \text{and} \quad S_C(t) = \mathbb{P} [y_j^0 \geq t]$$

The standard procedure to include time-to-event' outcome

$(\delta_i, \varepsilon_j)$	$x_i - y_j \geq \tau$	$x_i - y_j \leq -\tau$	$ x_i - y_j < \tau$
(1, 1)	Favorable	Unfavorable	Neutral
(0, 1)	Favorable	Uninformative	Uninformative
(1, 0)	Uninformative	Unfavorable	Uninformative
(0, 0)	Uninformative	Uninformative	Uninformative

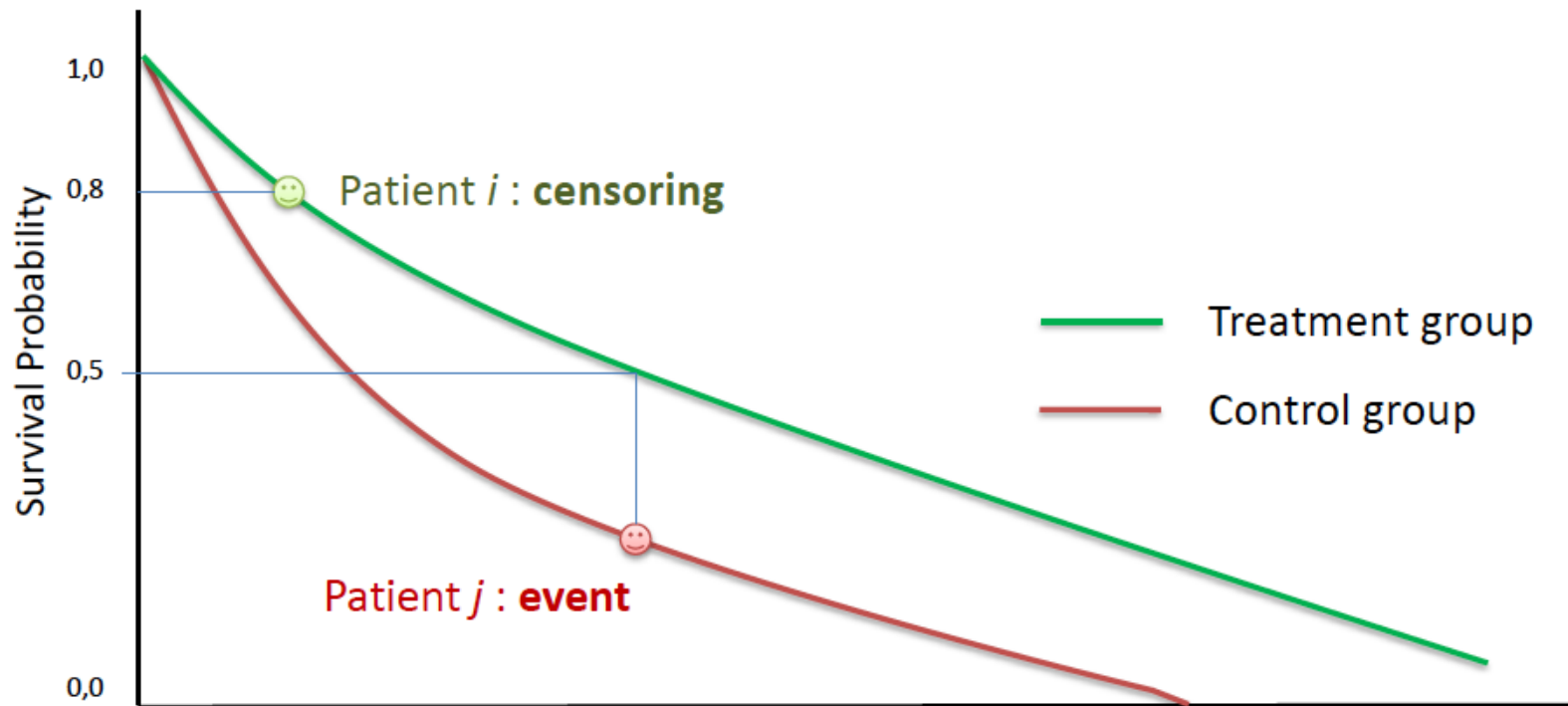
The standard procedure to include time-to-event outcome



The extended procedure taking into account 'non-informative' pairs

Based on the Kaplan-Meier estimate of the survival function

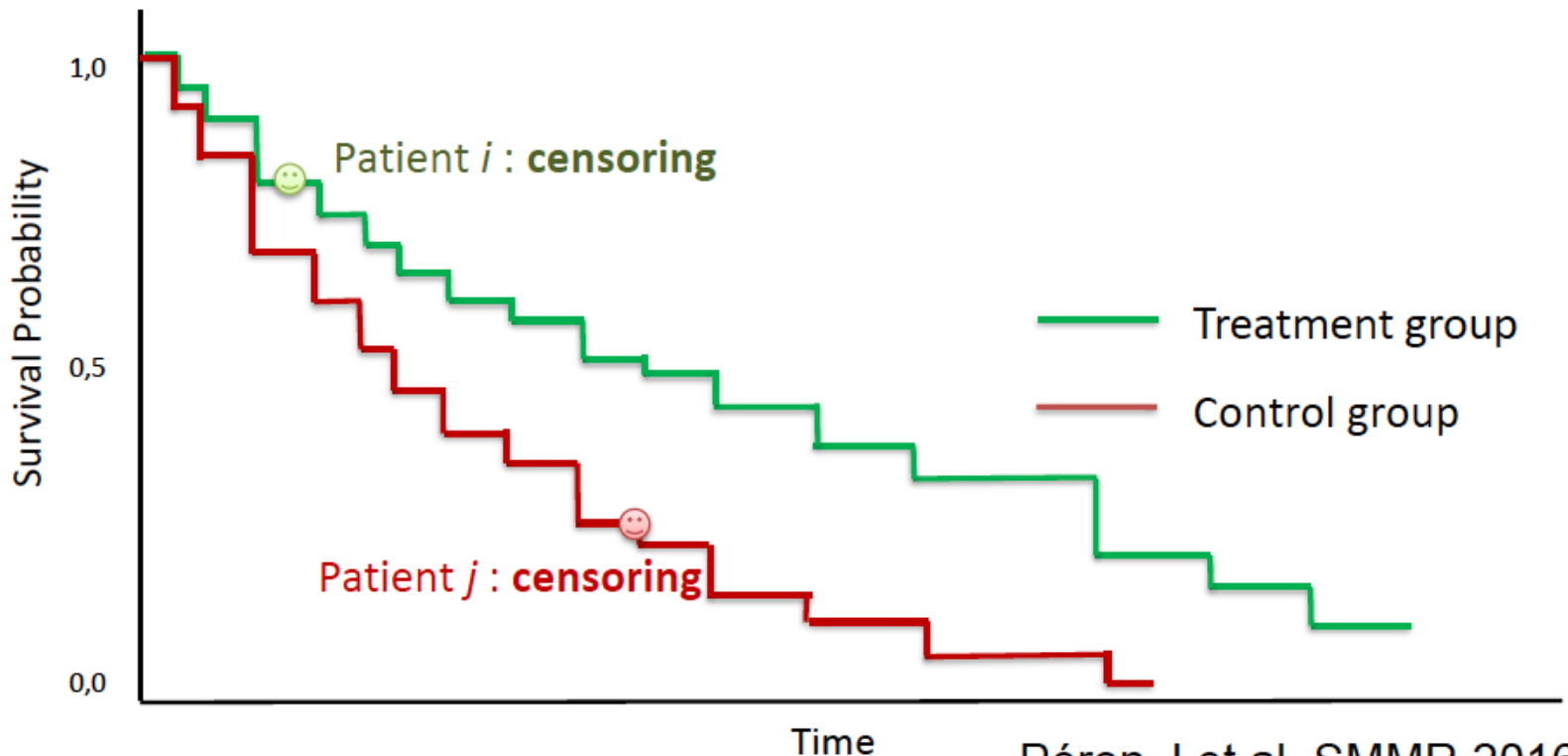
$$\mathbb{P}[(x_i^0 > y_j^0) | (x_i^0 > x_i)] = \frac{\hat{S}_{Ttt}(y_j)}{\hat{S}_{Ttt}(x_i)} = \frac{0,5}{0,8}$$



The extended procedure taking into account 'non-informative' pairs

When the estimation of the survival function is discontinue :

$$\mathbb{P}[(x_i^0 > y_j^0) | (x_i^0 > x_i), (y_j^0 > y_j)] = - \sum_{t > y_j}^{\infty} \frac{\hat{S}_{Ttt}(t)}{\hat{S}_{Ttt}(x_i) \hat{S}_{Ctrl}(y_j)} \cdot (\hat{S}_{Ctrl}(t^+) - \hat{S}_{Ctrl}(t^-))$$



The extended procedure taking into account 'non-informative' pairs

For pairs that can not be decisively classified because of censoring, we compute:

$$\mathbb{P}(x_i^0 > y_j^0 + \tau), \text{ et } \mathbb{P}(y_j^0 > x_i^0 + \tau), \text{ et } \mathbb{P}(|x_i^0 - y_j^0| < \tau)$$

The pairwise score is:

$$s_{ij} = \mathbb{P}(x_i^0 > y_j^0 + \tau) - \mathbb{P}(y_j^0 > x_i^0 + \tau)$$

The net benefit is then:

$$\hat{\Delta} = \frac{1}{m \cdot n} \sum_{i=1}^n \sum_{j=1}^m s_{ij}$$

Achievements of the extended procedure

- Reduction of the Bias of in the presence of censored observations
 - Correction available (proportional hazards)
- Increased power of the permutation test compared to standard procedure
 - Proportional hazards and administrative censoring < 67% (B Efron, Stanford Univ, 1967)
 - Late treatment effect

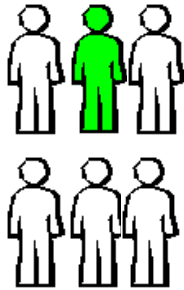
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The net benefit

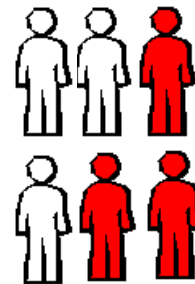
Probability for a random patient in the Treatment group to have a 'better outcome' than a random patient in the Control group ...

Treatment group



$$\Delta = \mathbb{P}(X > Y) - \mathbb{P}(Y > X)$$

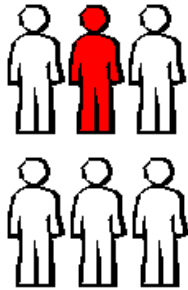
Control group



The net benefit

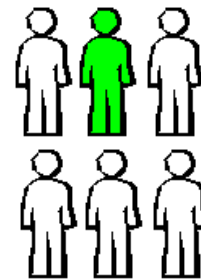
...minus the opposite probability.

Treatment group



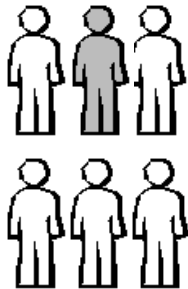
$$\Delta = \mathbb{P}(X > Y) - \mathbb{P}(Y > X)$$

Control group



The net benefit

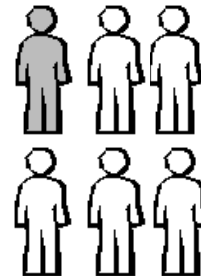
Treatment group



$$\mathbb{P}(Y=X)$$

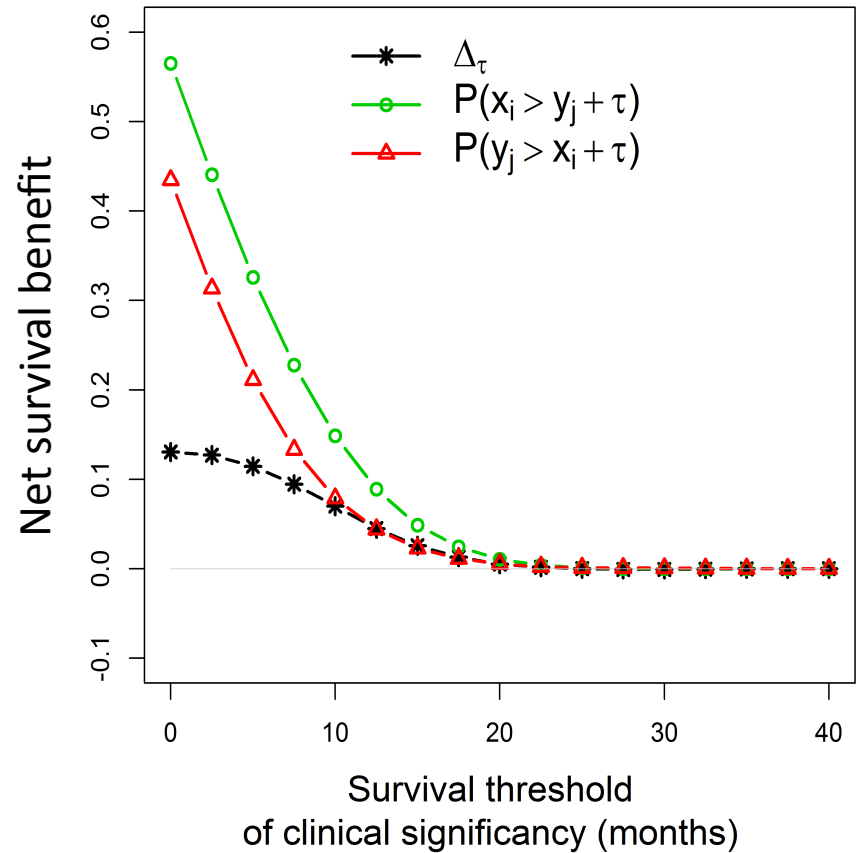
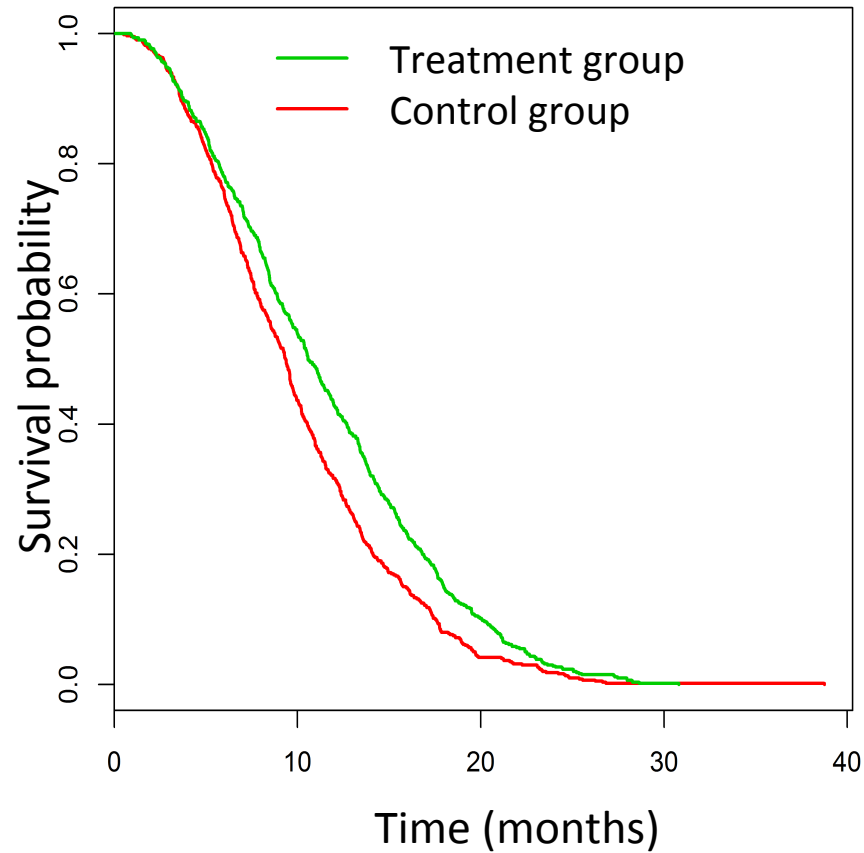
$$\Delta = \mathbb{P}(X>Y) - \mathbb{P}(Y>X)$$

Control group



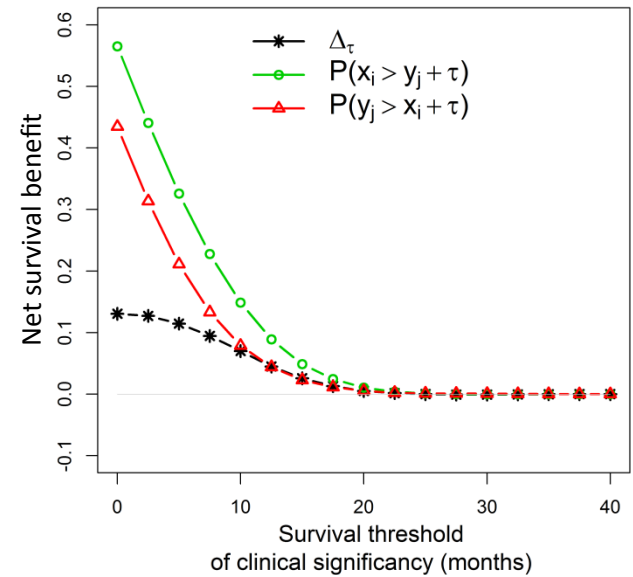
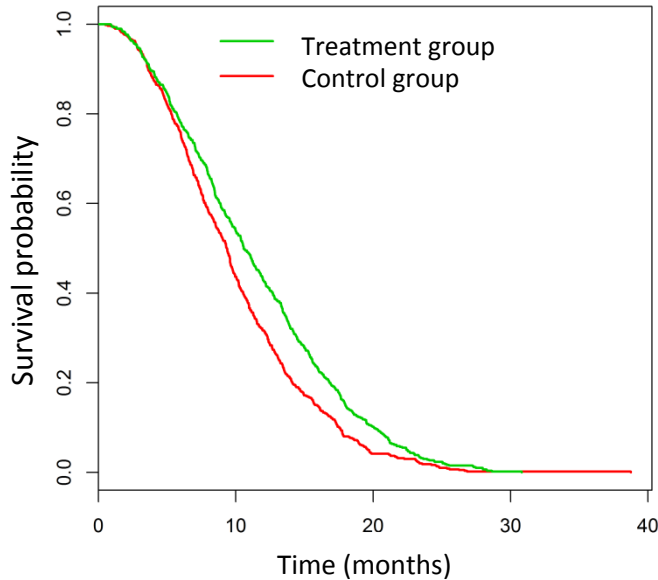
The net survival benefit

Proportional hazards

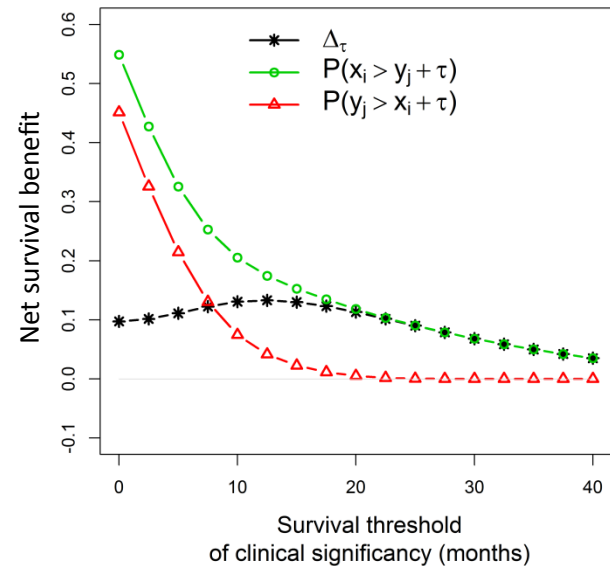
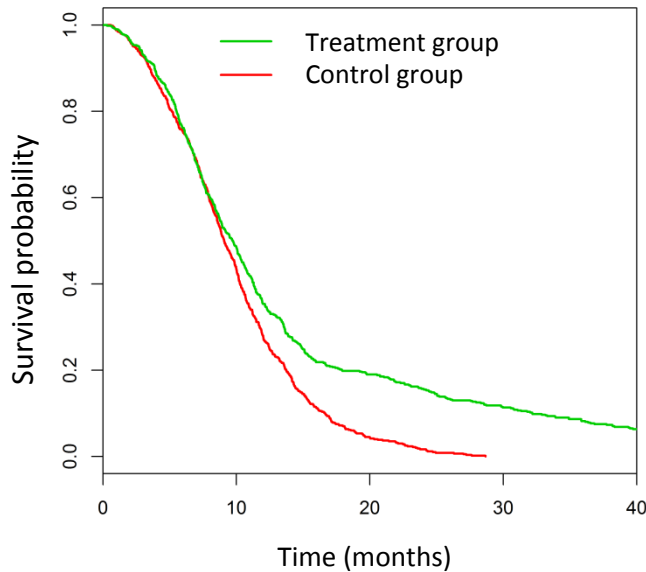


The net survival benefit

Proportional Hazards

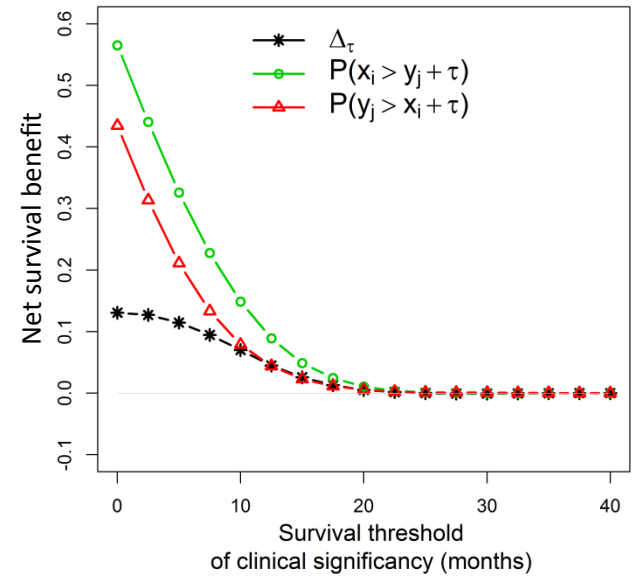
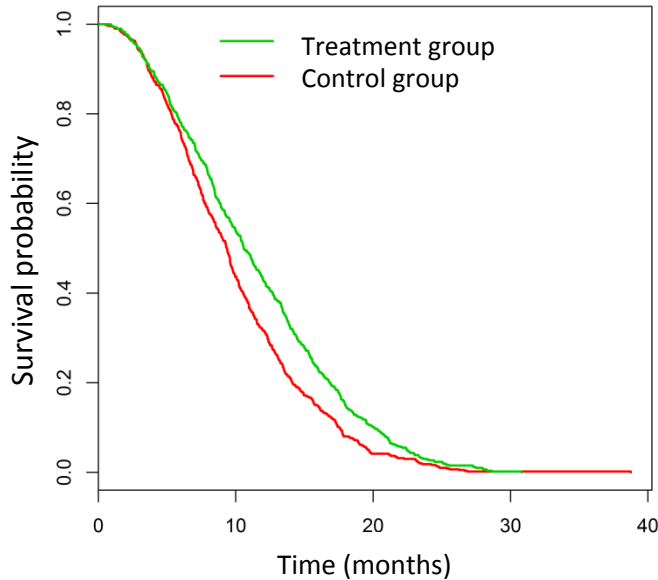


Delayed treatment effect

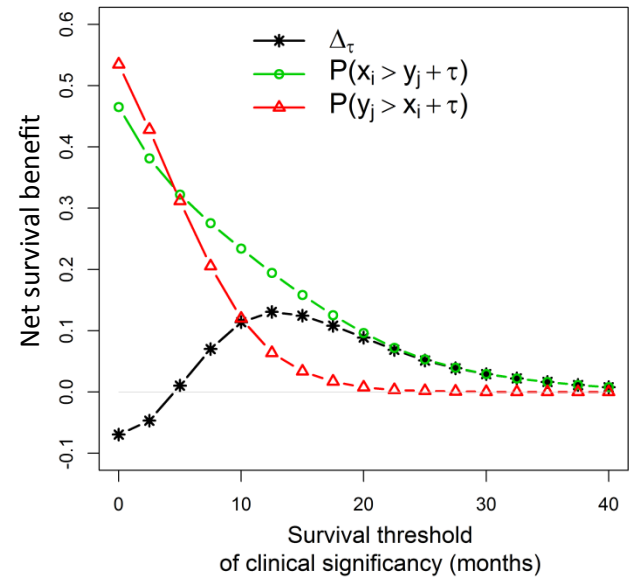
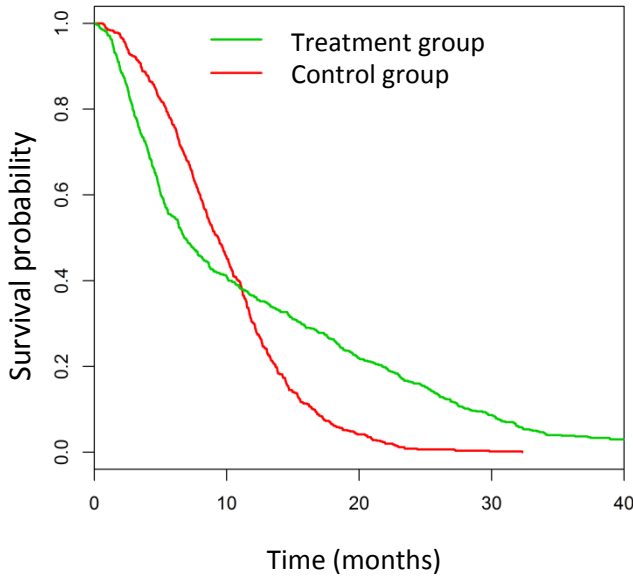


The net survival benefit

Proportional Hazards



Opposite hazards



Outline

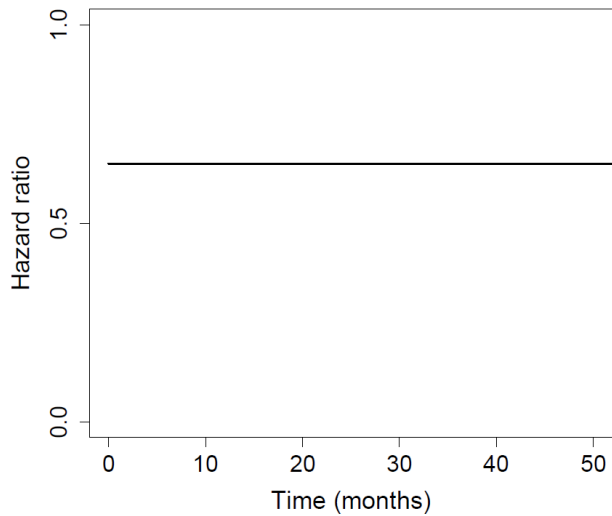
- The procedure of generalized pairwise comparisons
- A patient-oriented measure of treatment benefit
- Application on immuno-oncology trials
 - Simulation study
 - Illustration on an ipilimumab trial

Simulation study - Design

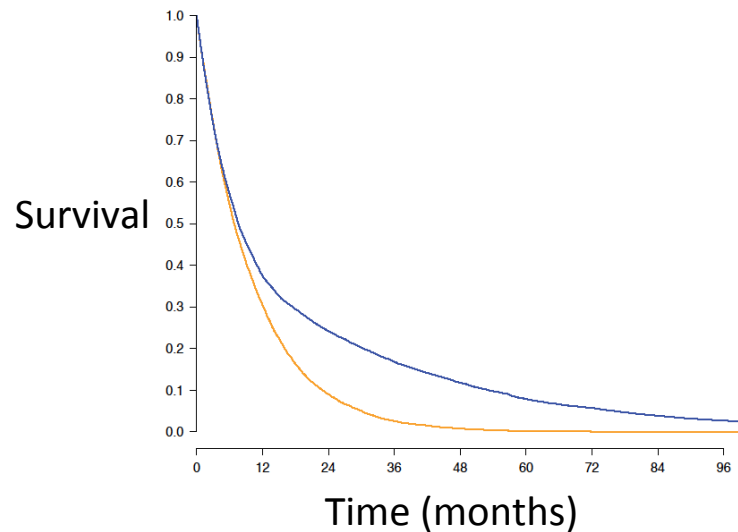
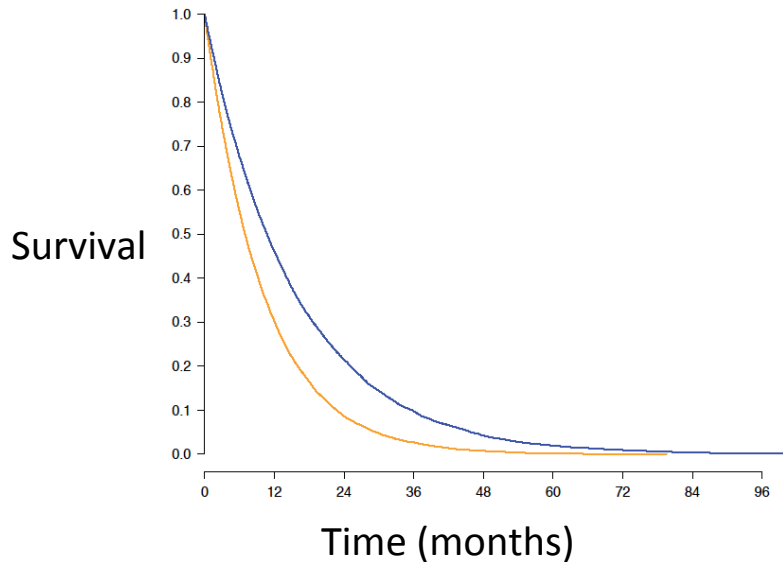
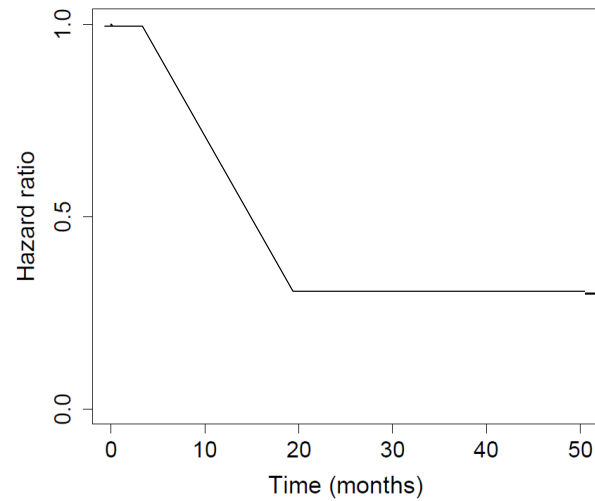
- **Objective:** To assess the power of tests based on generalized pairwise comparisons for delayed treatment effect
- Simulation of $M = 1000$ datasets with $N = 200$ patients
 - One time-to-event outcome

Simulation study - Design

Scenario 1 : Proportional hazards



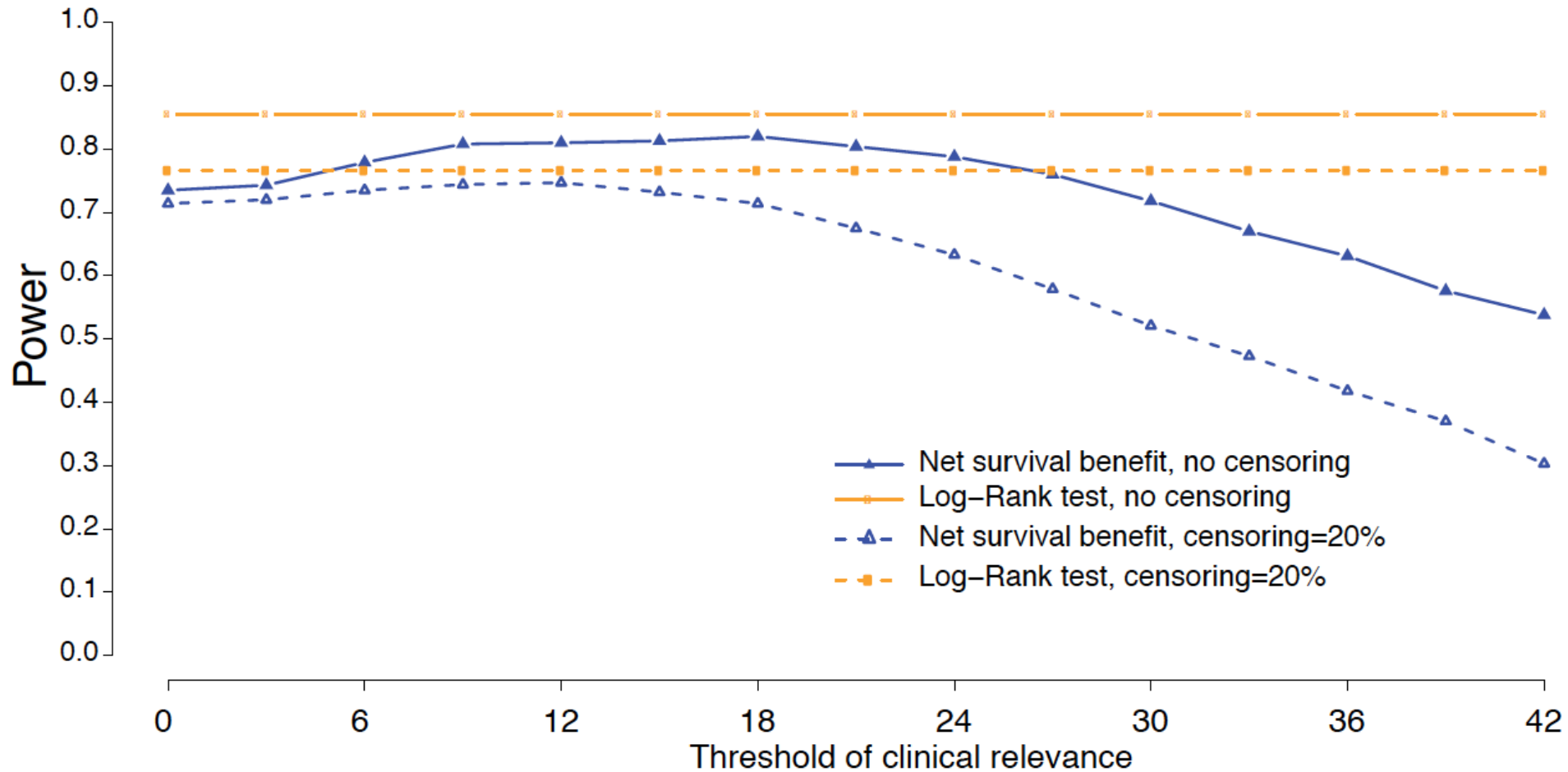
Scenario 2 : Late treatment effect



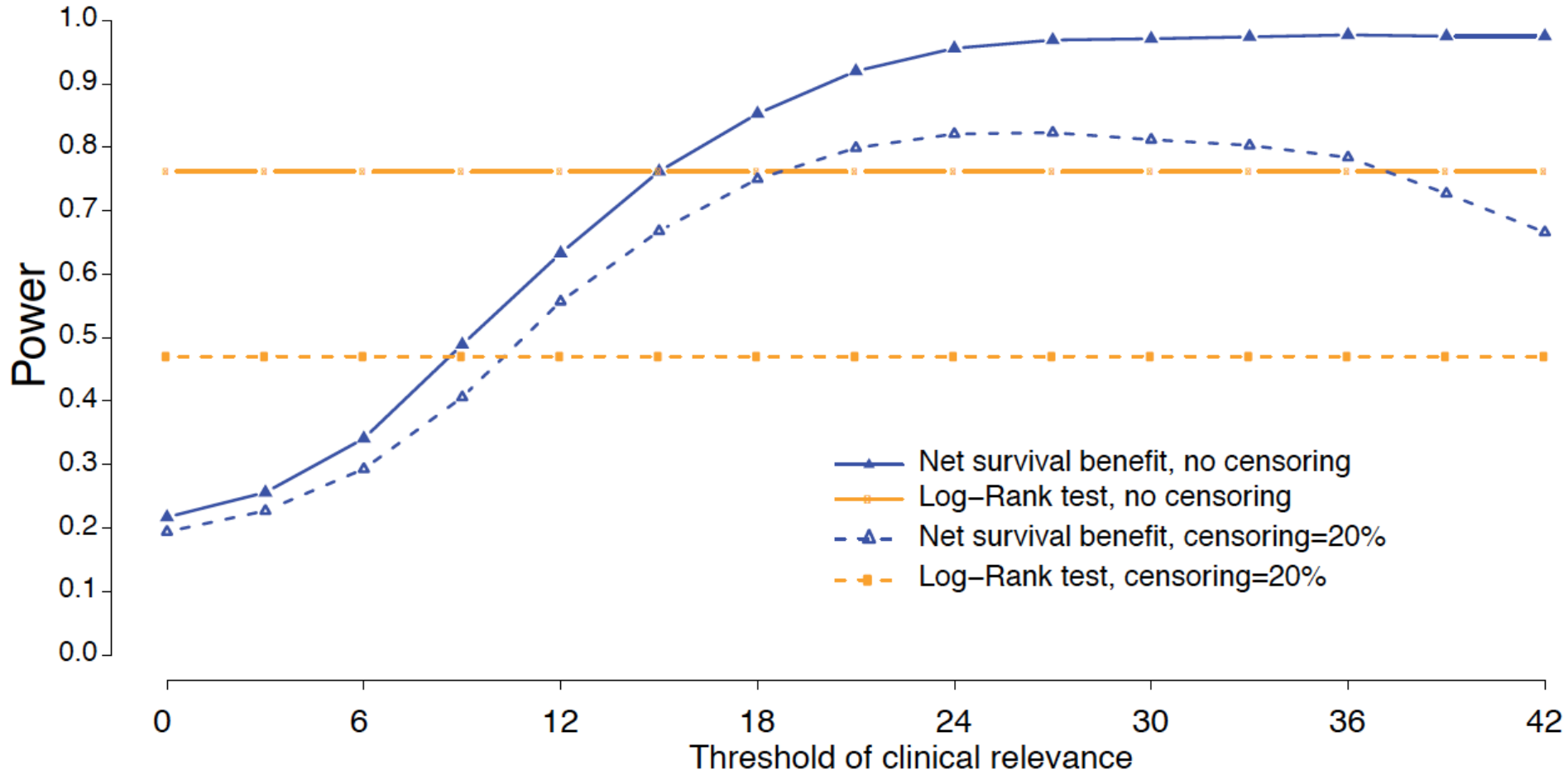
Simulation study - Design

- Administrative censoring proportion
 - Uniform distribution
 - Between 0% and 20%
- For each simulated dataset
 - Estimation of the net survival benefit of at least τ months [0 to 42 months] (extended procedure)
 - Test of the null hypothesis (Permutation test, Log-Rank test)

Proportional Hazards - POWER



Delayed treatment effect - POWER



Conclusions of the simulation study

When a long-term survival benefit is expected
(anticancer immune therapy)

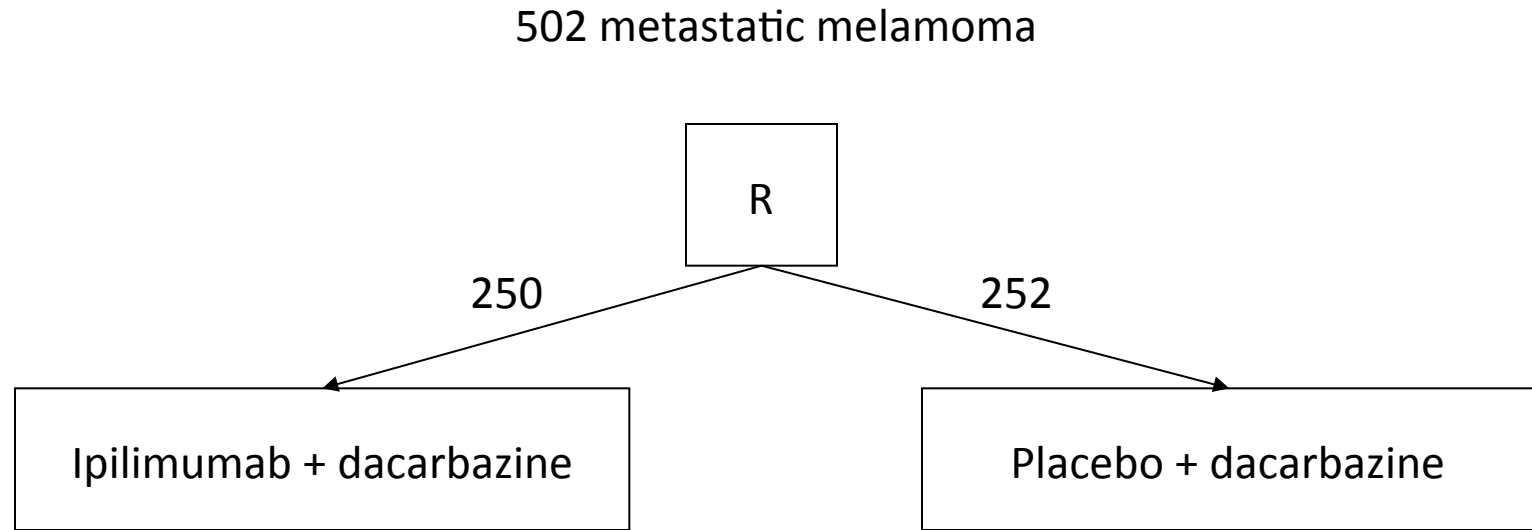
The net survival benefit is:

- Arguably more relevant than traditional methods → focus on long term survival differences
- More powerful than traditional method

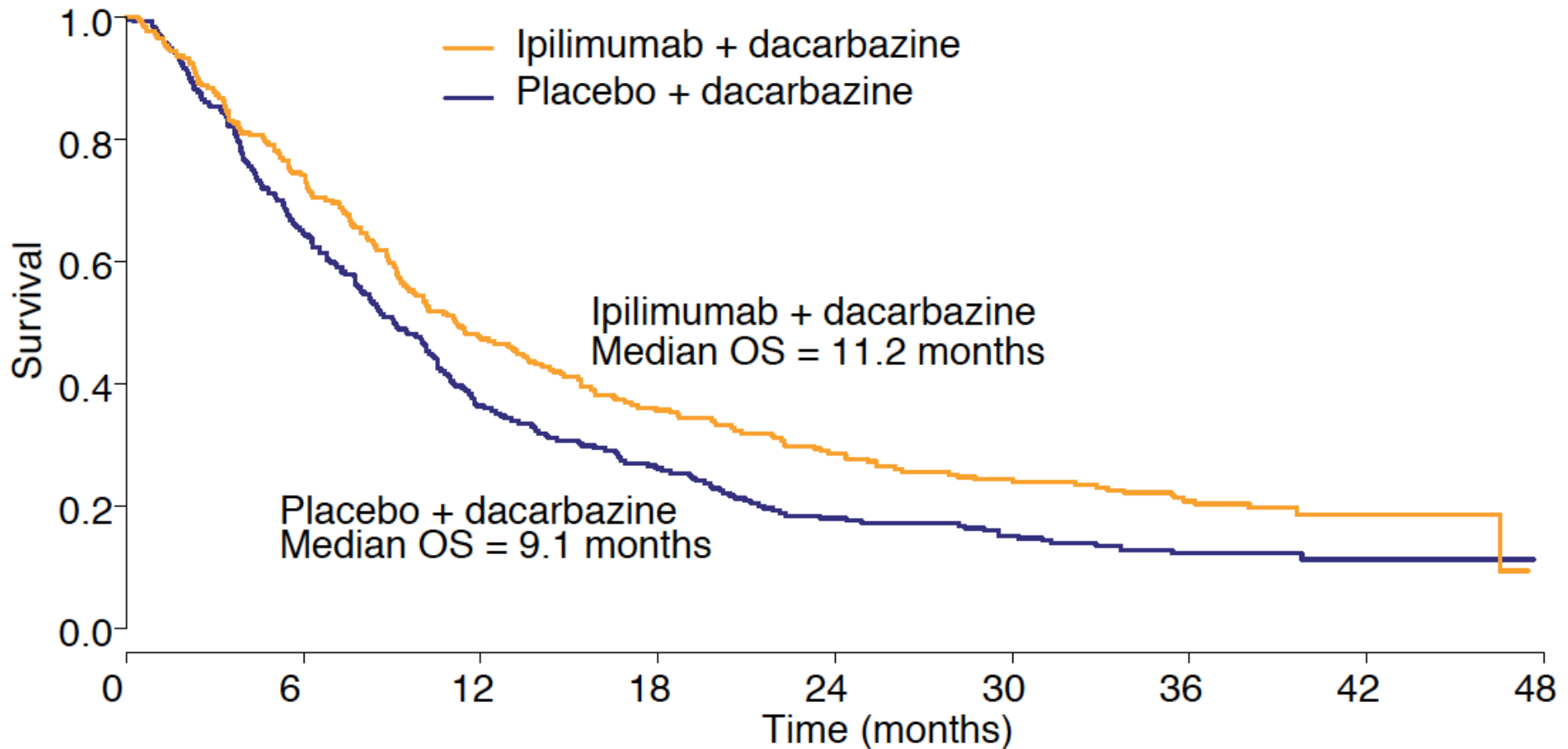
Outline

- The procedure of generalized pairwise comparisons
- A patient-oriented measure of treatment benefit
- **Application on immuno-oncology trials**
 - Simulation study
 - Illustration on an ipilimumab trial

The net survival benefit in the CA184-024 trial

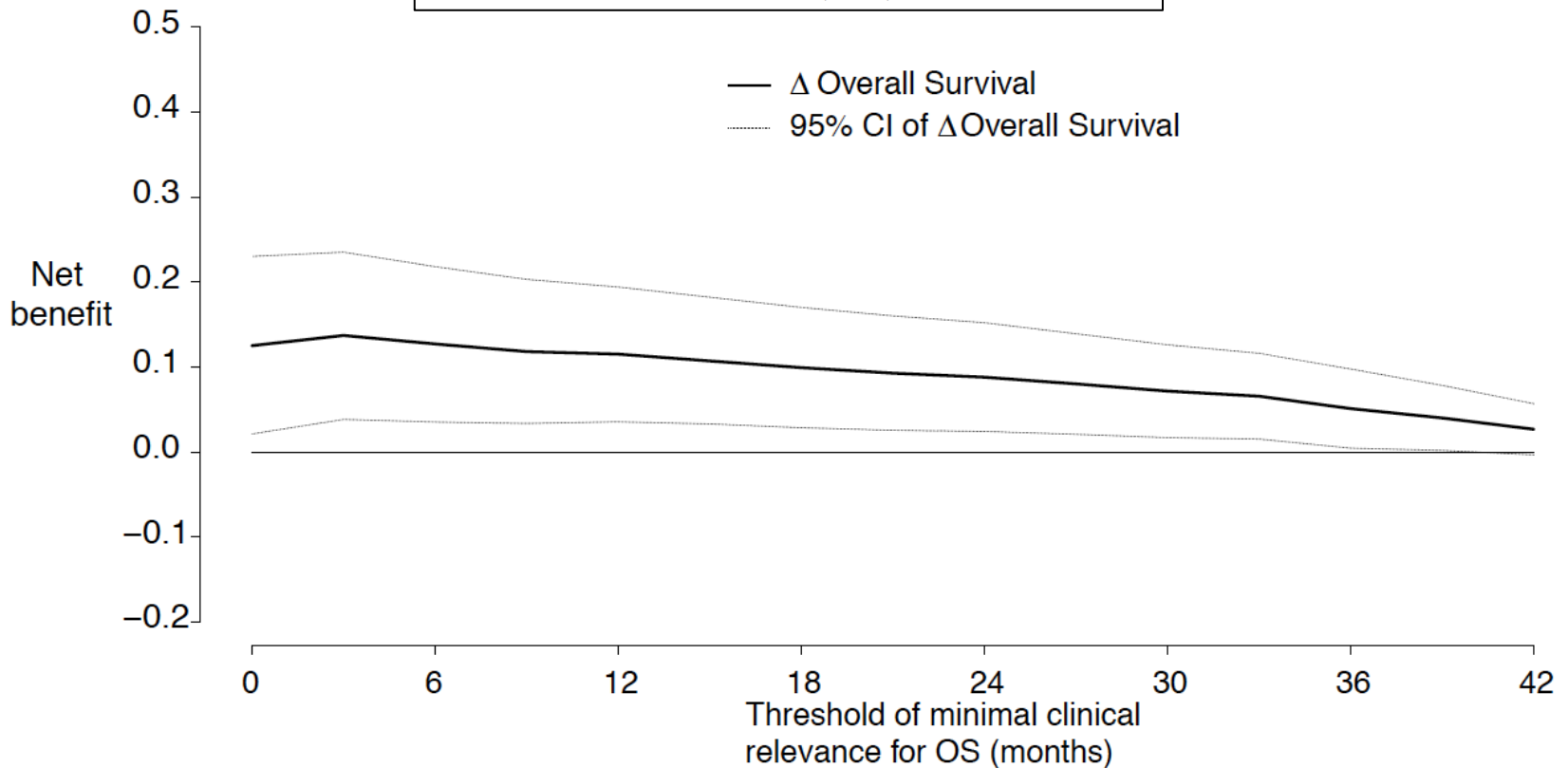
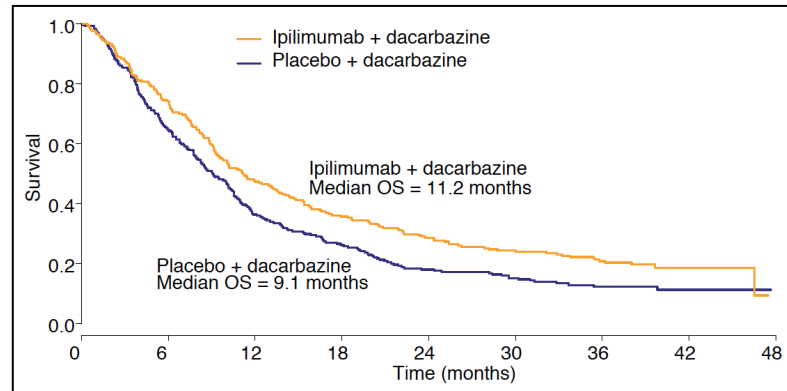


OS results in the CA184-024 trial

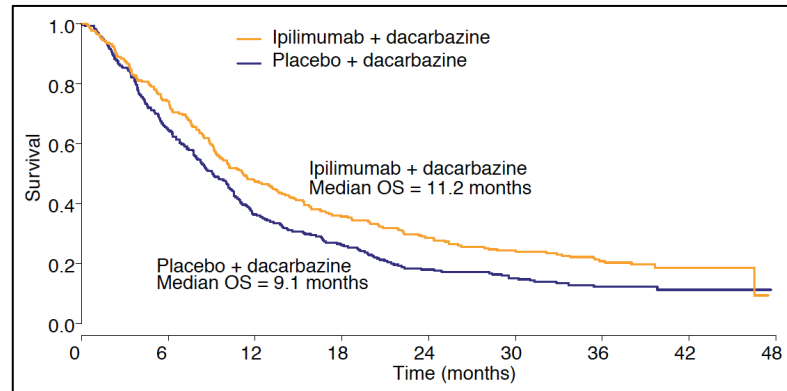


Pcb	252	160	89	64	44	37	26	7	0
Ipi	250	181	114	85	68	57	41	10	0

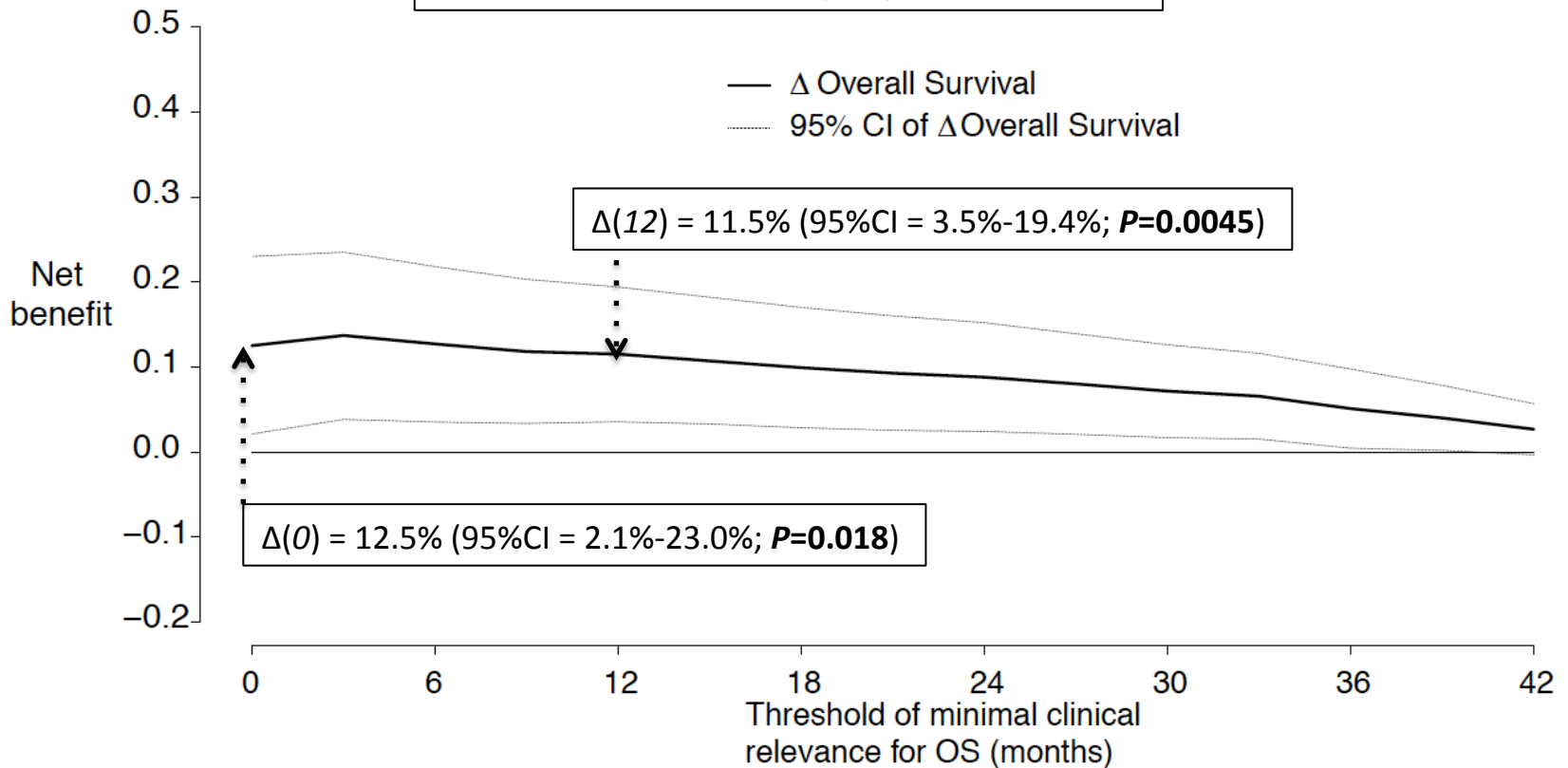
OS results in the CA184-024 trial



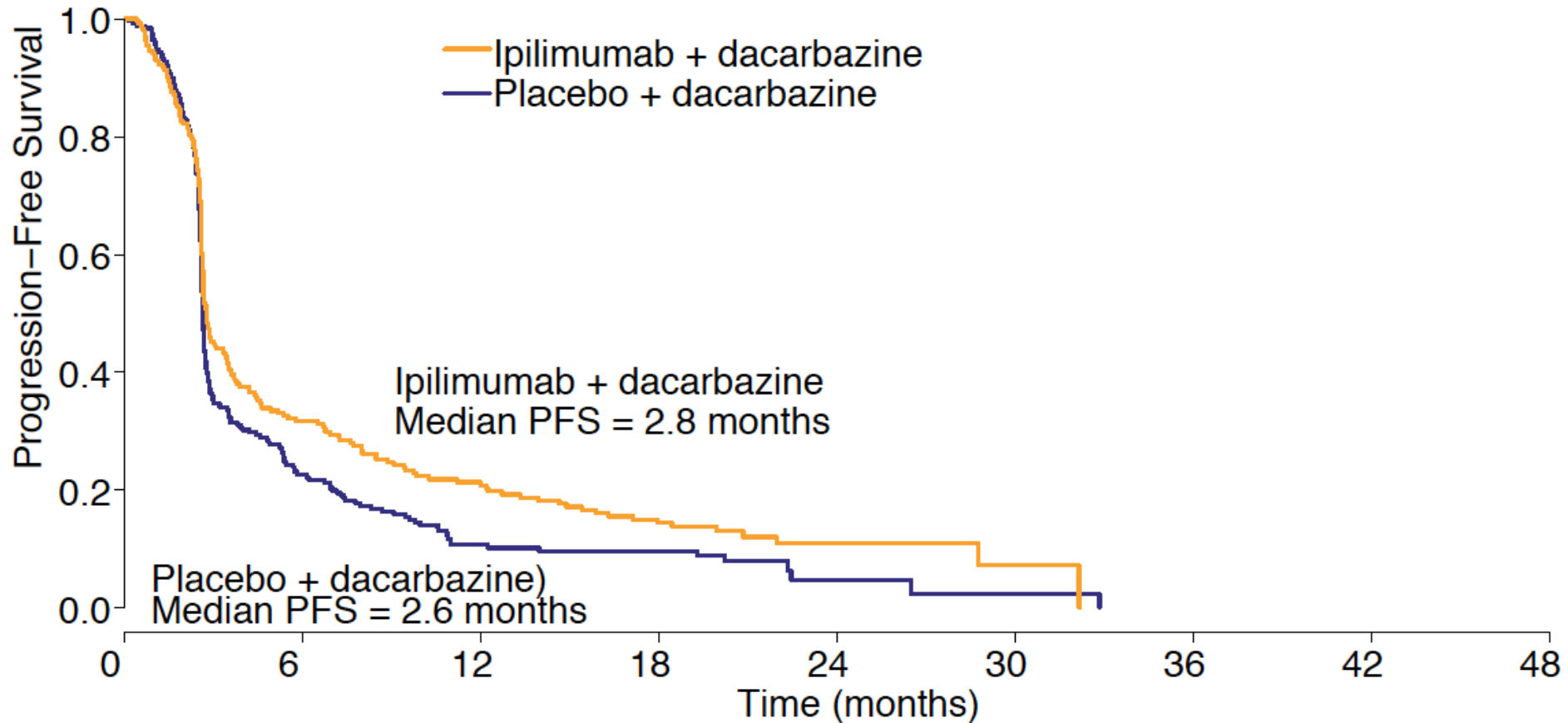
OS results in the CA184-024 trial



Log rank P = 0.0054

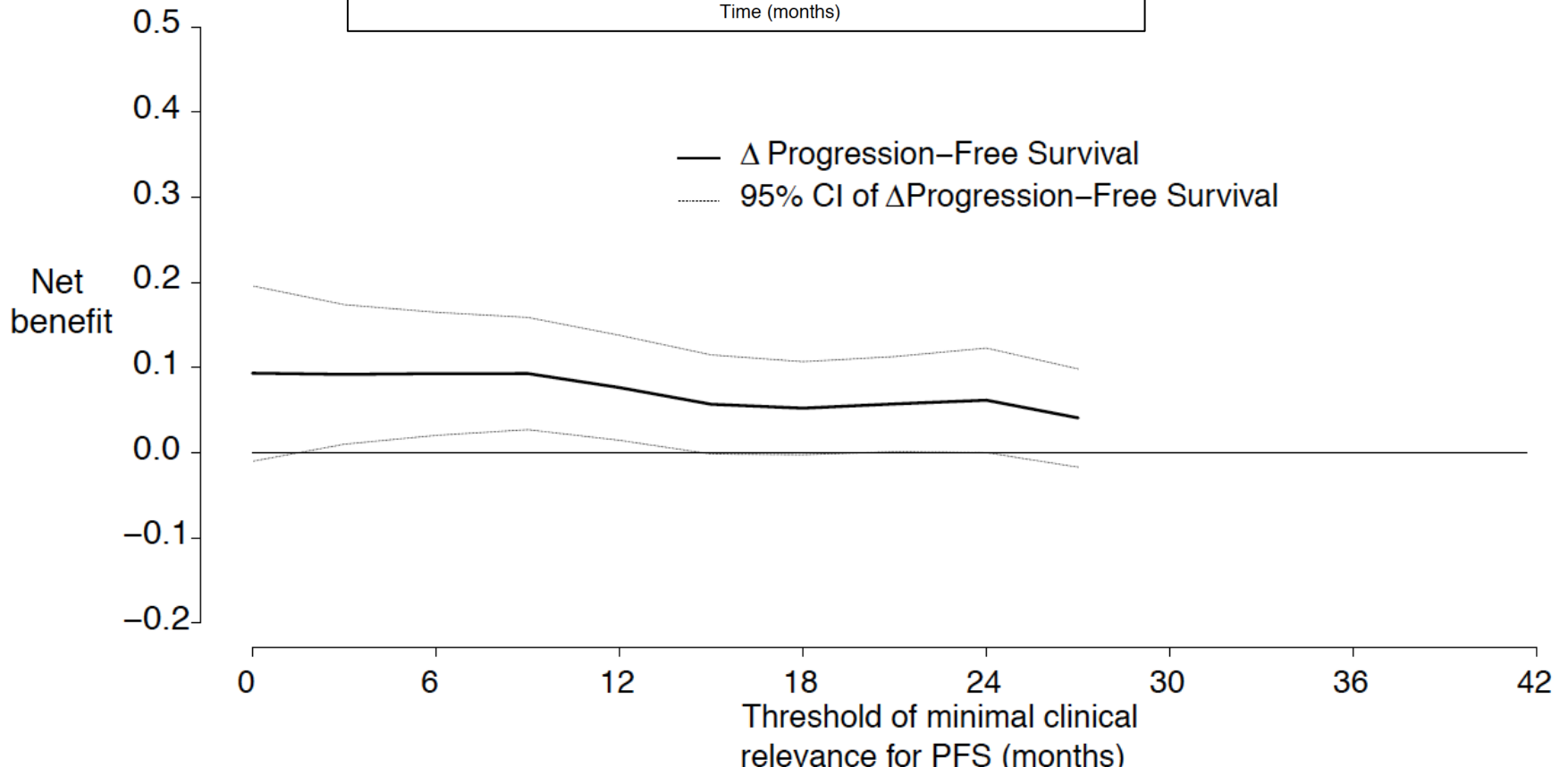
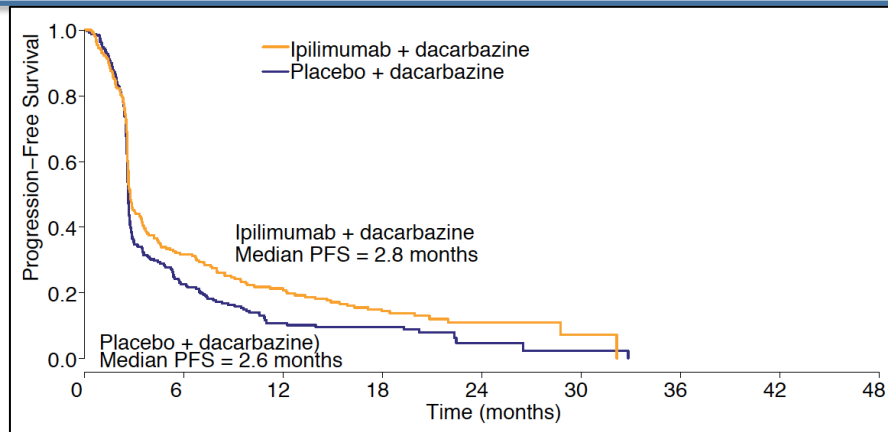


PFS results in the CA184-024 trial

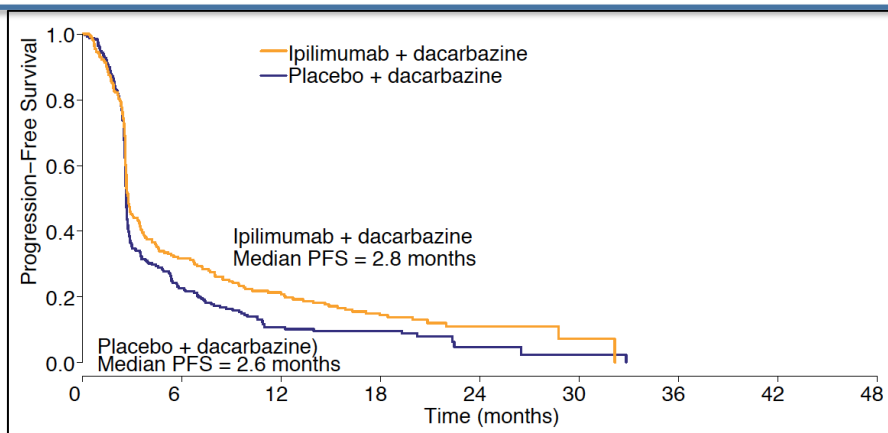


Pcb	252	52	20	13	2	1	0	0	0
Ipi	250	70	40	25	6	2	0	0	0

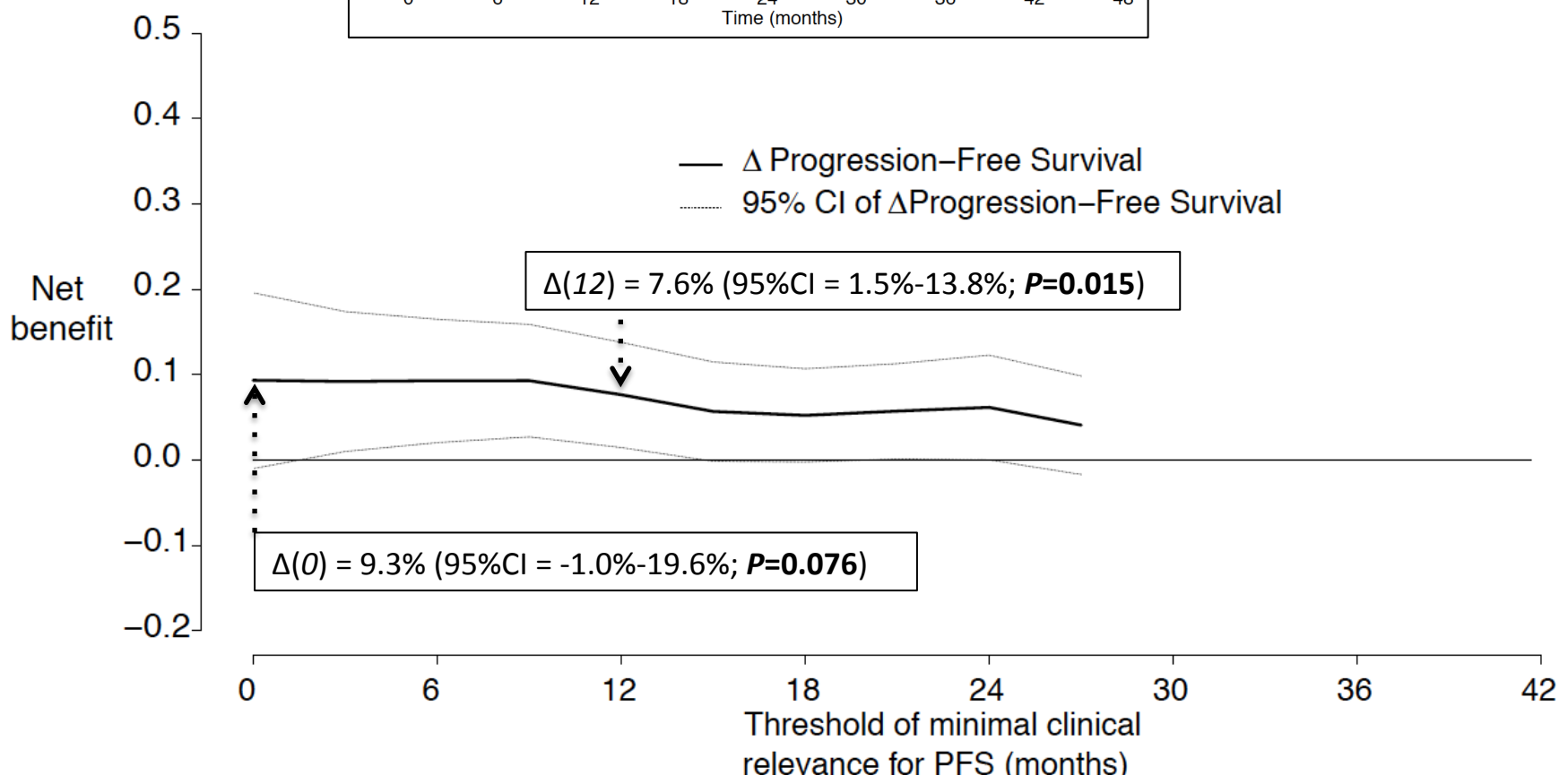
PFS results in the CA184-024 trial



PFS results in the CA184-024 trial



Log rank P = 0.022



Software implementation

A package 

- Available on CRAN (“BuyseTest”)
- Available on github (“<https://github.com/bozenne/BuyseTest>”)

Conclusions

The net benefit

- Is equivalent to standard non-parametric tests in simple cases
- Is **meaningful and patient-relevant**
- Can focus on **long-term survival differences**
- Allows **multicriteria analysis**
- May have better power than the logrank test (e.g. for delayed treatment effect)
- Is OK when hazards are **not proportionals**
- Is **available**

Thank you

References

- Buyse M. Reformulating the hazard ratio to enhance communication with clinical investigators. *Clin Trials* 5: 641-2, 2008.
- Buyse M. Generalized pairwise comparisons for prioritized outcomes in the two-sample problem. *Statist Med* 29: 3245-57, 2010.
- Péron J, Buyse M, Ozenne B, Roche L, Roy P. An extension of generalized pairwise comparisons for prioritized outcomes in the presence of censoring. *Statist Meth Med Res* DOI: 10.1177/0962280216658320, 2017.
- Péron J, Roy P, Ding K, Parulekar W, Roche L, Buyse M. Benefit-risk assessment of adding erlotinib to gemcitabine for the treatment of advanced pancreatic cancer. *Brit J Cancer* 112: 971-976, 2015.
- Péron J, Roy P, Ozenne B, Roche L, Buyse M. The net chance of a longer survival as a patient-oriented measure of benefit in randomized clinical trials. *JAMA Oncology* DOI: 10.1001/jamaoncol.2015. 6359, 2016.

Methods – Definition of priority

First priority outcome	Second priority outcome	Pair rating
Favorable	NA	Favorable
Unfavorable	NA	Unfavorable
Neutral/Uninf	Favorable	Favorable
Neutral/Uninf	Unfavorable	Unfavorable
Neutral/Uninf	Neutral/Uninf	Neutral/Uninf

Methods – Definition of priority

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Neutral/Uninf	Favorable	Favorable
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Neutral/Uninf	Neutral/Uninf	Neutral/Uninf

Methods – Definition of priority

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Neutral/Uninf	Favorable	Favorable
Neutral/Uninf	Unfavorable	Unfavorable
Neutral/Uninf	Neutral/Uninf	Neutral/Uninf

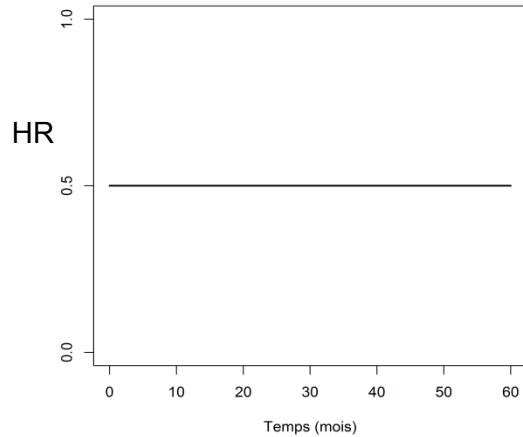
Simulation study - Design

- **Objective:** To compare the standard and the extended procedures of generalized pairwise comparison
- Simulation of $M = 1000$ datasets of with $N = 200$ patients
 - One time-to-event outcome
 - Threshold $\tau = 0$ months

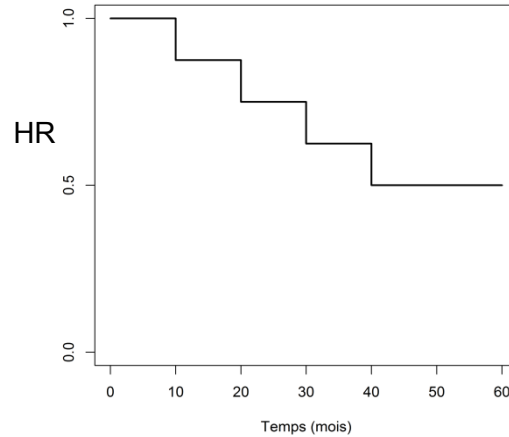
Simulation study - Design

- Survival time: exponential distributions

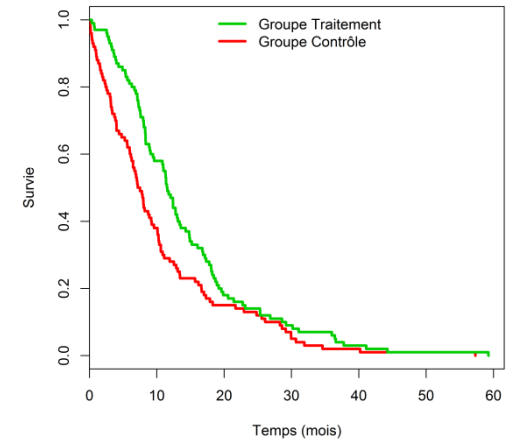
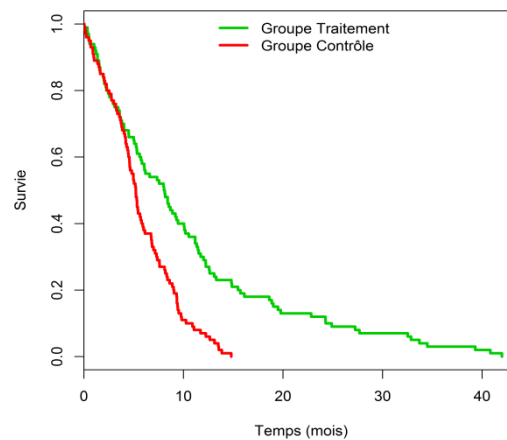
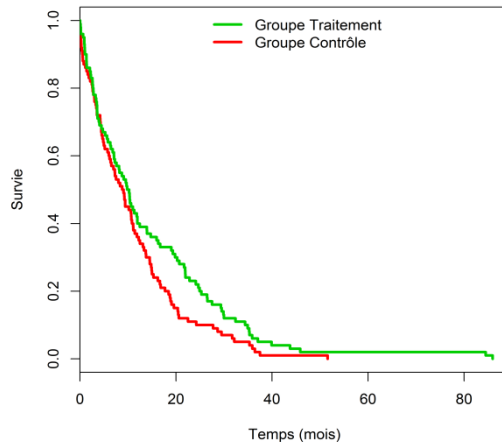
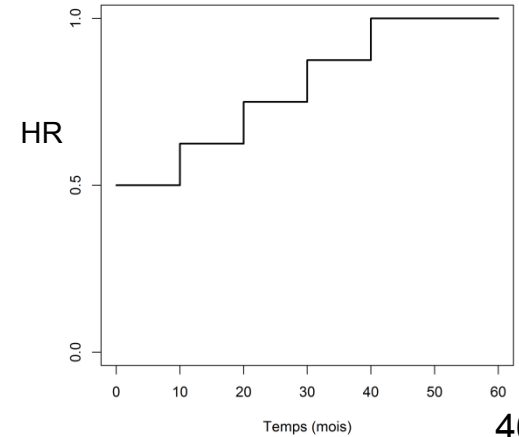
Scenario 1 :
Proportional hazards



Scenario 2 : Late
treatment effect



Scenario 3 : early
treatment effect



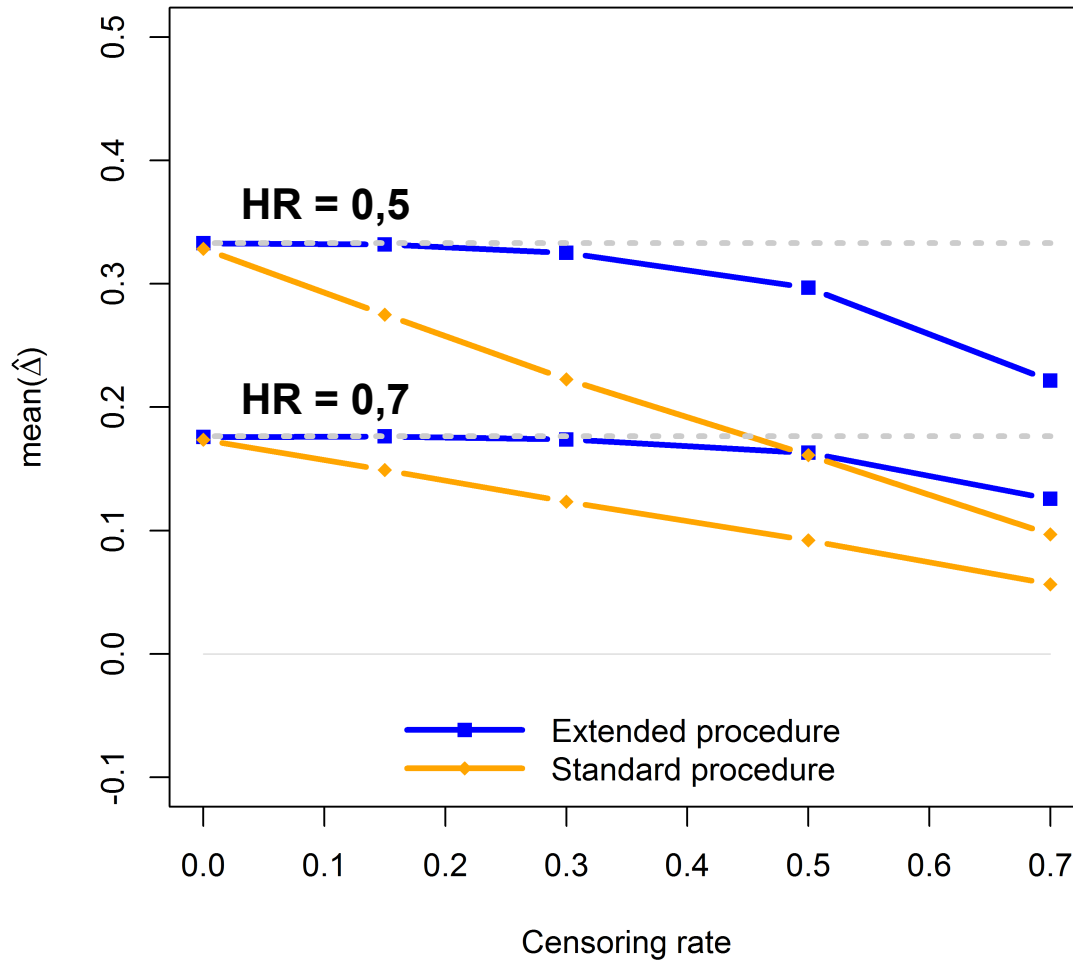
Simulation study - Design

- Several treatment effect size
 - Hazard ratio {0,5; 0,7; 1}
- Administrative censoring proportion
 - Uniform distribution
 - Between 0% and 70%

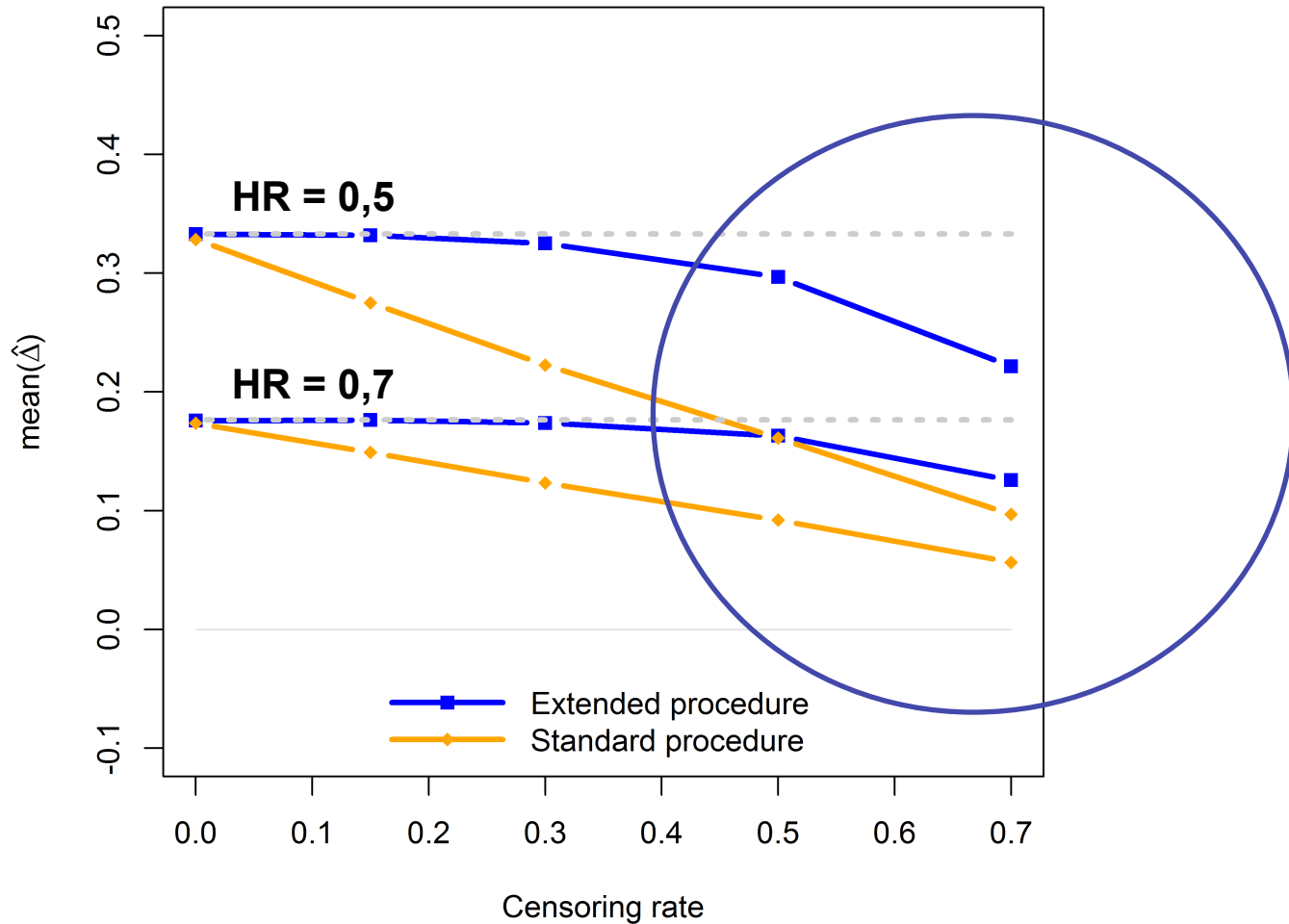
Simulation study - Design

- For each simulated dataset
 - Estimation of the net chance of a better outcome (standard and extended procedure)
 - Test of the null hypothesis (Permutation test, Log-Rank test)
- Endpoints
 - Bias
 - Power
 - Type 1 error

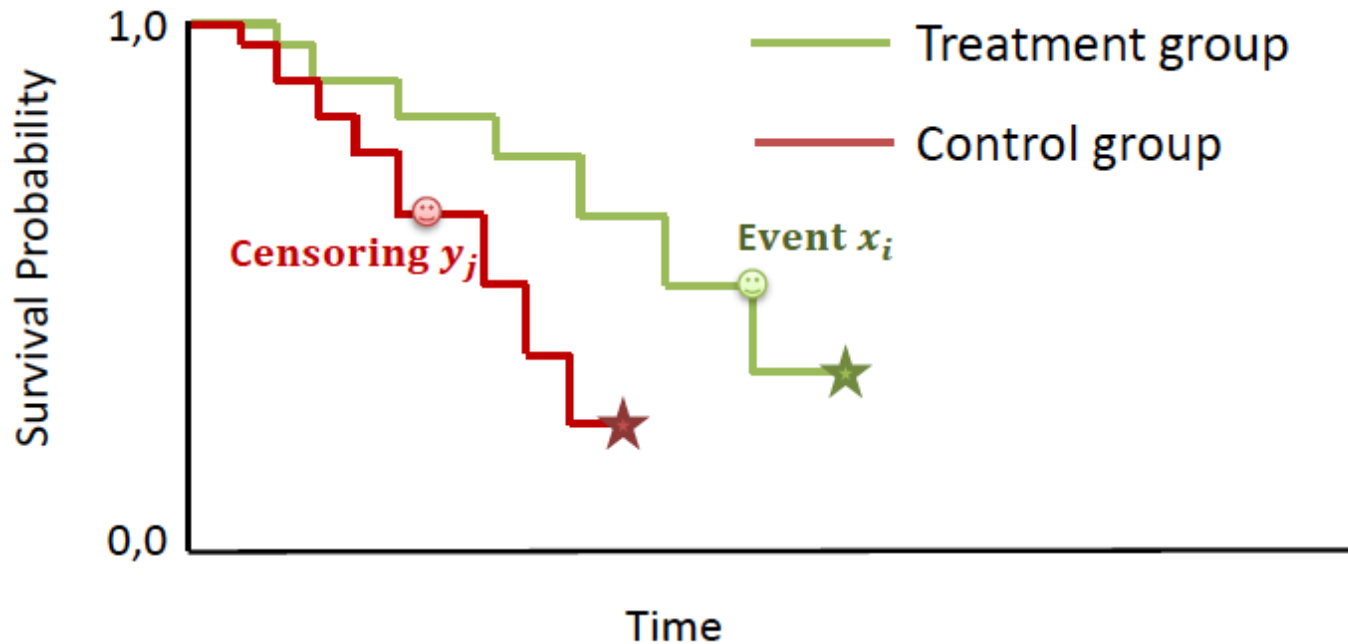
Scenario 1 – Proportional hazards



Scenario 1 – Proportional hazards



An explanation for this bias?



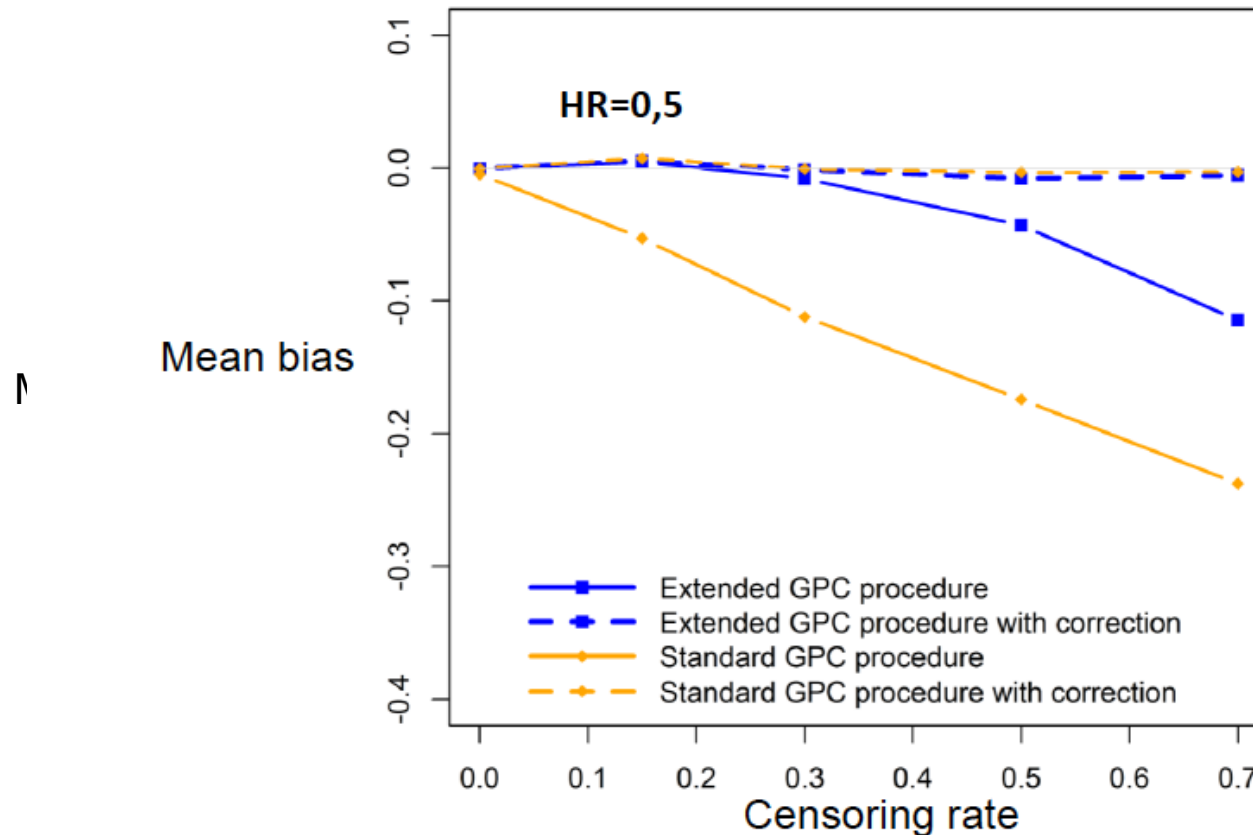
Standard procedure: **Uninformative** $\rightarrow p_{ij} = 0$

Extended procedure: $\begin{cases} \mathbb{P}[x_i^0 > y_j^0 + \tau] = 1 - \frac{\hat{S}_C(x_i - \tau)}{\hat{S}_C(y_j)} \\ \mathbb{P}[y_j^0 > x_i^0 + \tau] = \frac{\hat{S}_C(x_i + \tau)}{\hat{S}_C(y_j)} \end{cases}$ **Uninformative also** $\rightarrow p_{ij} = 0$

A correction for this bias

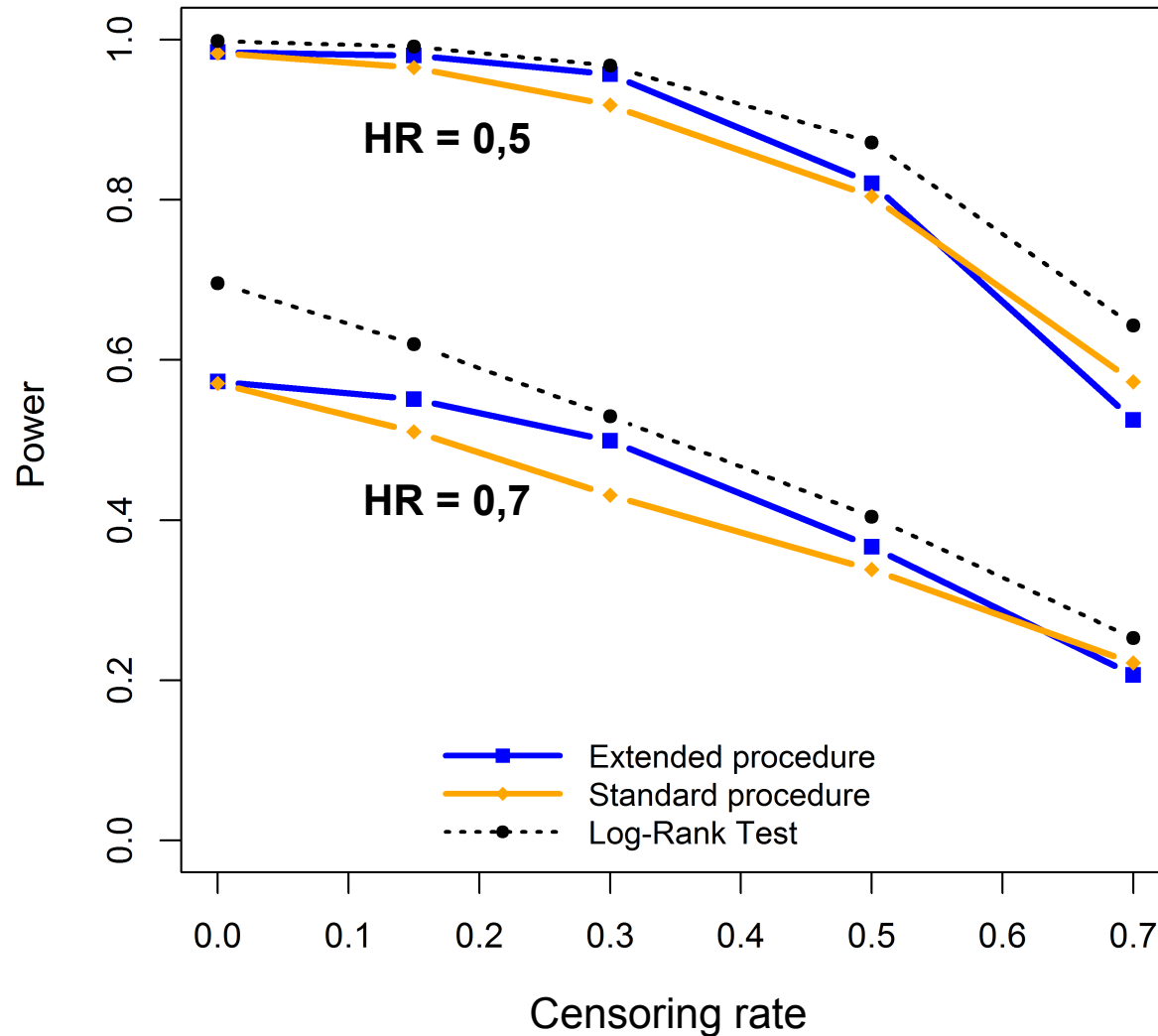
f_m : Proportion of informative pairs for the trial $m(m = 1, \dots, M)$

$\hat{\Delta}_{corr_m} = \frac{\hat{\Delta}_m}{f_m}$: corrected net chance of a better outcome



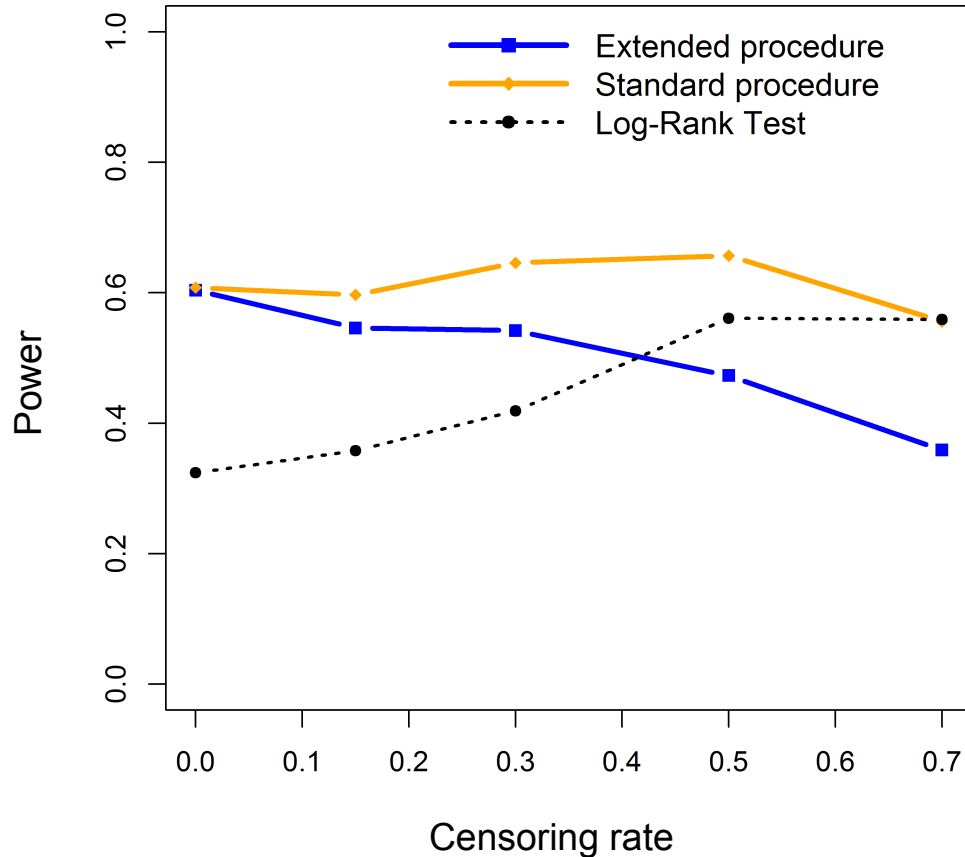
$$Bias = \mathbb{E}(\hat{\Delta} - \Delta)$$

Scenario 1 – Proportional hazards

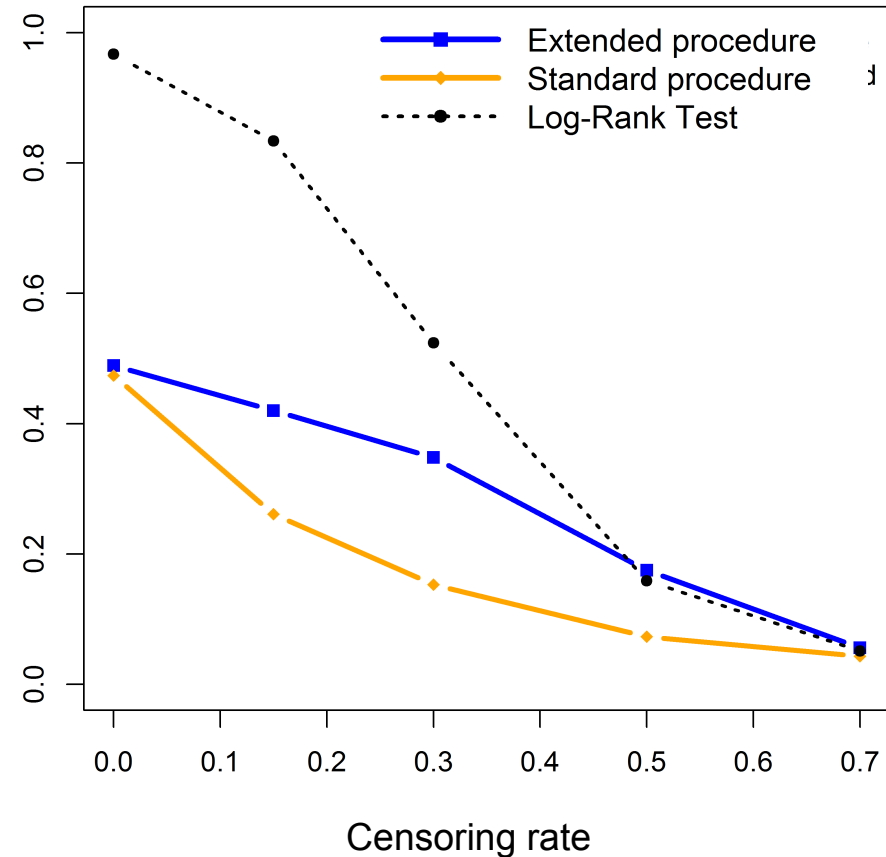


Scenario 2 et 3 – Non Proportional hazards

Early treatment effect



Late treatment effect



Type 1 error rate $\approx 5\%$