

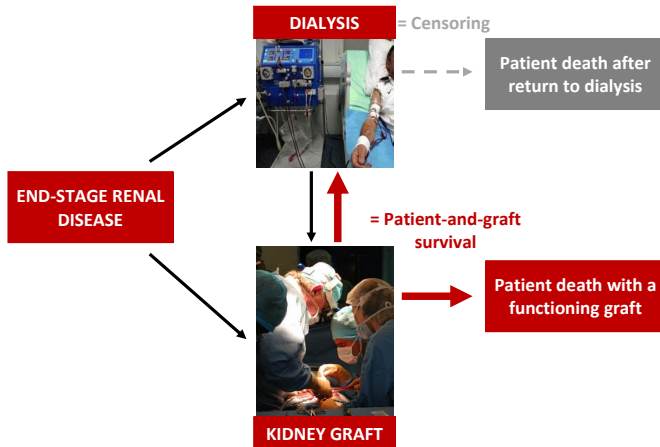
A multiplicative-regression model to compare the effect of factors associated with the time to graft failure between first and second renal transplant

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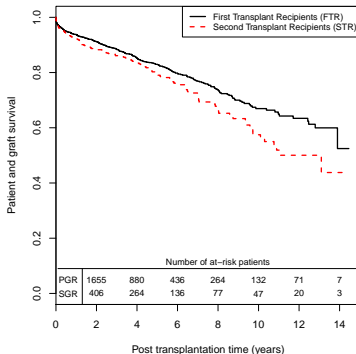
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Definition of the graft failure



Close patient-and-graft survival between first and second graft



Objective

- ⇒ Are risk factors associated with graft failure comparable between first and second grafts ?

Limits of classical survival models

- Test of interaction between each covariate and graft rank
- Only covariates common to first and second grafts

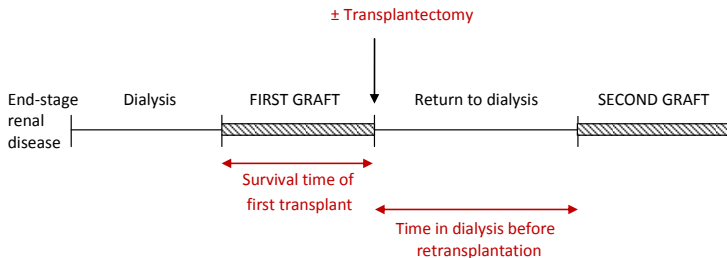


FIGURE 1: Clinical trajectory before second graft.

Classical approach

- Additive-regression model for relative survival
(Estève et al. Stat in Med 1990)
- Endpoint = mortality related to chronic diseases
- The expected mortality is based on general population

Proposed approach

- **Multiplicative**-regression model for relative survival
(Andersen et al. Stat in Med 1989)
- Endpoint = **graft failure** (return to dialysis or patient death)
- The expected graft failure hazard is **estimated** in a control group (first graft)

Inclusion criteria

French DIVAT database

- Centers : Nantes, Necker, Nancy, Toulouse, Montpellier, Lyon
- Adult recipients
- Transplanted from 1996 to 2010
- Under mycophenolate mofetil and steroids at transplantation



Group of interest

566 second transplant recipients (**STR**)



Control group

2206 first transplant recipients (**FTR**)

Multiplicative-regression models for relative survival

- The hazard function

$$h^{(o)}(t_i, z_i) = h^{(e)}(t_i, z_i^{(e)}) h^{(r)}(t_i, z_i^{(r)})$$

Observed hazard function in the STR group

z_i = covariates associated with the observed hazard

Expected hazard function in the FTR group

$z_i^{(e)}$ = subset of z_i , associated with the expected hazard

Relative hazard function in the STR group

$z_i^{(r)}$ = subset of z_i , associated with the relative hazard

STEP 1

STEP 2

STEP 1 :

Estimation of the expected hazard function ($N^{(e)} = 2206$ FTR)

- Parametric model and proportional hazards assumption

$$h^{(e)}(t_i, z_i^{(e)}) = h_0^{(e)}(t_i) \exp\left(\sum_{j=1}^{p^{(e)}} \beta_j^{(e)} z_{i,j}^{(e)}\right)$$

- $h_0^{(e)}(t_i)$ is a piecewise function
- Maximum-likelihood estimation

$$\log \mathcal{L} = \sum_{i=1}^{N^{(e)}} \{\delta_i \log(h^{(e)}(t_i, z_i^{(e)})) - H^{(e)}(t_i, z_i^{(e)})\}$$

with $\delta_i = 1$ if the graft failure is observed
 $\delta_i = 0$ if the event is right-censored

STEP 2 :

Estimation of the relative hazard function ($N^{(r)} = 566$ STR)

- Parametric model and proportional hazards assumption

$$h^{(r)}(t_i, z_i^{(r)}) = h_0^{(r)}(t_i) \exp\left(\sum_{j=1}^{p^{(r)}} \beta_j^{(r)} z_{i,j}^{(r)}\right)$$

- $h_0^{(r)}(t_i)$ is a piecewise function
- Maximum-likelihood estimation

$$\log \mathcal{L} = \sum_{i=1}^{N^{(o)}} \{ \delta_i \log(h^{(o)}(t_i, z_i)) - H^{(o)}(t_i, z_i) \}$$



$$h^{(e)}(t_i, z_i^{(e)}) h^{(r)}(t_i, z_i^{(r)}) \quad \int_0^{t_i} h^{(e)}(u, z_i^{(e)}) h^{(r)}(u, z_i^{(r)}) du$$

CASE 1 :

For $z_1^{(r)} \notin z_j^{(e)} \Rightarrow \exp(\beta_1^{(r)}) = \text{hazard ratio}$

$$\text{HR}_{z_1=1/z_1=0}^{(o)} = \frac{h^{(e)}(t_i, z_i^{(e)}) h_0^{(r)}(t_i) \exp(\sum_{j=1}^{p^{(r)}} \beta_j^{(r)} z_{i,j}^{(r)})}{h^{(e)}(t_i, z_i^{(e)}) h_0^{(r)}(t_i) \exp(\sum_{j=2}^{p^{(r)}} \beta_j^{(r)} z_{i,j}^{(r)})} = \exp(\beta_1^{(r)})$$

CASE 2 :

For $z_1^{(r)} \in z_j^{(e)} \Rightarrow \exp(\beta_1^{(r)}) = \text{weighting factor of HR}$

$$\begin{aligned} \text{HR}_{z_1=1/z_1=0}^{(o)} &= \frac{h_0^{(e)}(t_i) \exp(\sum_{j=1}^{p^{(e)}} \beta_j^{(e)} z_{i,j}^{(e)}) h_0^{(r)}(t_i) \exp(\sum_{j=1}^{p^{(r)}} \beta_j^{(r)} z_{i,j}^{(r)})}{h_0^{(e)}(t_i) \exp(\sum_{j=2}^{p^{(e)}} \beta_j^{(e)} z_{i,j}^{(e)}) h_0^{(r)}(t_i) \exp(\sum_{j=2}^{p^{(r)}} \beta_j^{(r)} z_{i,j}^{(r)})} \\ &= \exp(\beta_1^{(e)}) \exp(\beta_1^{(r)}) \end{aligned}$$

Introduction

Materials and
Methods

Results

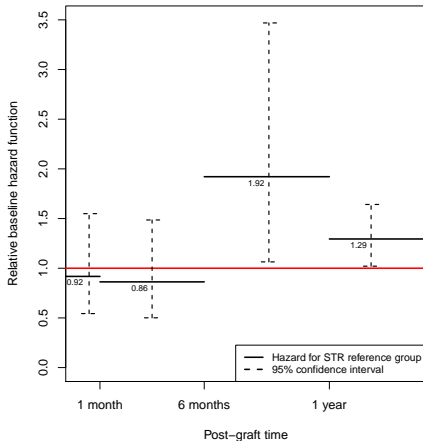
Conclusion

Covariates	Included in $z_j^{(e)}$
Recipient age (≥ 55 years / < 55 years)	✓
Recipient gender (male / female)	✓
Causal nephropathy (recurrent / non recurrent)	✓
History of comorbidities (positive / negative)	✓
Body mass index ($\geq 30 \text{ kg.m}^{-2}$ / $< 30 \text{ kg.m}^{-2}$)	✓
Anti-class I or II PRA (positive / negative)	✓
Dialysis prior transplantation (positive / negative)	
Recipient EBV or CMV serology (positive / négative)	
Type of donor (deceased donor / living donor)	✓
Donor age (≥ 55 years / < 55 years)	✓
Donor EBV serology (positive / négative)	✓
Donor gender (male / female)	
Cause of donor death (cerebro-vascular / other)	
Donor serum creatinine ($\geq 133 \mu\text{mol/l}$ / $< 133 \mu\text{mol/l}$)	
Donor CMV serology (positive / négative)	
Transplantation period (< 2005 / ≥ 2005)	✓
Number of HLA-A-B-DR mismatches (> 4 / ≤ 4)	✓
Induction therapy (depleting / non depleting)	✓
Cold ischemia time ($\geq 24\text{h}$ / $< 24\text{h}$)	✓
Survival time of the first transplant (< 1 year / ≥ 1 year)	
Time before retransplantation (> 3 years / ≤ 3 years)	
First graft transplantectomy (positive / négative)	

Results

Relative baseline hazard function $\Leftrightarrow \mathbf{z}^{(r)} = \mathbf{0}$

- Introduction
- Materials and Methods
- Results
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- Introduction
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Final multivariate model

Covariates	Expected (FTR)	Relative (STR)	Observed (STR)	95% CI	p value
Transplant period (< 2005/≥2005)	1.37	1.27	-	0.83 - 1.95	0.2604
Recipient gender (male / female)	1.19	0.68	-	0.45 - 1.02	0.0645
Recipient age (≥55 years/<55 years)	1.55	1.61	-	1.03 - 2.52	0.0387
Donor age (≥55 years/<55 years)	1.37	0.59	-	0.37 - 0.95	0.0294
Type of donor (deceased/living)	2.91	0.33	-	0.12 - 0.91	0.0332
Donor gender (male / female)	-	-	1.57	1.01 - 2.45	0.0443
Retransplant time (>3 years/≤3 years)	-	-	2.06	1.33 - 3.20	0.0012

xxx = forced covariates

CASE 1 : $z_1^{(r)} \notin z_j^{(e)}$

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$$HR^{(o)} = \exp(\beta_1^{(r)})$$

CASE 2 : $z_i^{(r)} \in z_j^{(e)}$

Covariates	Expected (FTR)	Relative (STR)	Observed (STR)	95% CI	p value
Transplant period (< 2005/≥2005)	1.37	1.27	-	0.83 - 1.95	0.2604
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Type of donor (deceased/living)	2.91	0.33	0.96	0.12 - 0.91	0.0332
Donor gender (male / female)	-	-	1.57	1.01 - 2.45	0.0443
Re transplant time (>3 years/≤3 years)	-	-	2.06	1.33 - 3.20	0.0012

xxx = forced covariates



$$HR^{(o)} = \exp(\beta_1^{(e)}) \exp(\beta_1^{(r)})$$

Clinical conclusions

- A particular attention to recipient age for clinical practice when faced a second transplantation should be paid
- A selection bias ?
Only transplants from "good quality" donors are proposed for STR when the donor is aged or deceased.
- An early effect of immunisation ?
The immunisation might take over the effect of other factors (donor age and donor type) for STR.

Statistical conclusion

- No necessity to test interactions between covariates and graft rank
- Possibility to take into account specific covariates for interest groups

Limits

- Parameters of the expected hazard function (STEP 1) were afterwards considered as constant when used for the estimation of the relative hazard (STEP 2)
⇒ A Monte-Carlo approach is in process
- Proportional hazard model with a piecewise baseline function were chosen for the estimation of both baseline hazards functions
⇒ A more flexible model for instance with spline functions
- Only time-invariant covariates were included in the model
⇒ A generalisation with time-dependent covariates

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Of note

- This presentation will be available online : <http://www.divat.fr/en>