

Threshold Regression Models for Time-to-Events Data

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Outline

- Cox model has been well used to analyzing time-to-event data. It has, however, limitations.
- An example to demonstrate the usefulness of the first-hitting time based **threshold regression (TR) model**.
- Brief Introduction of the **TR** model
- Tao Xiao will present Stata codes for the alternative model.

A non-proportional hazard example:

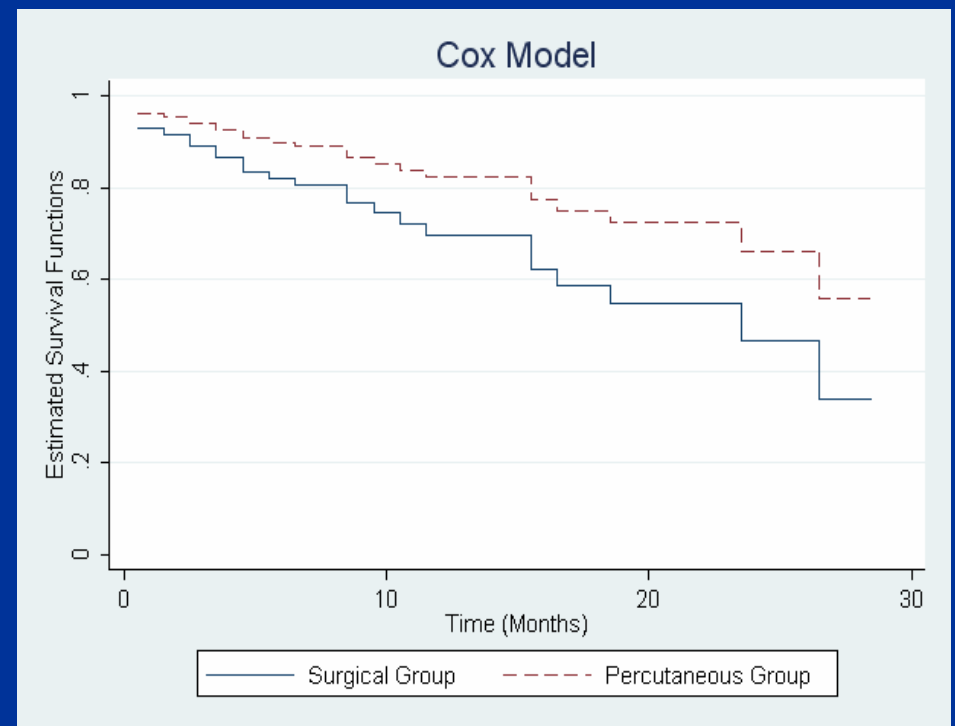
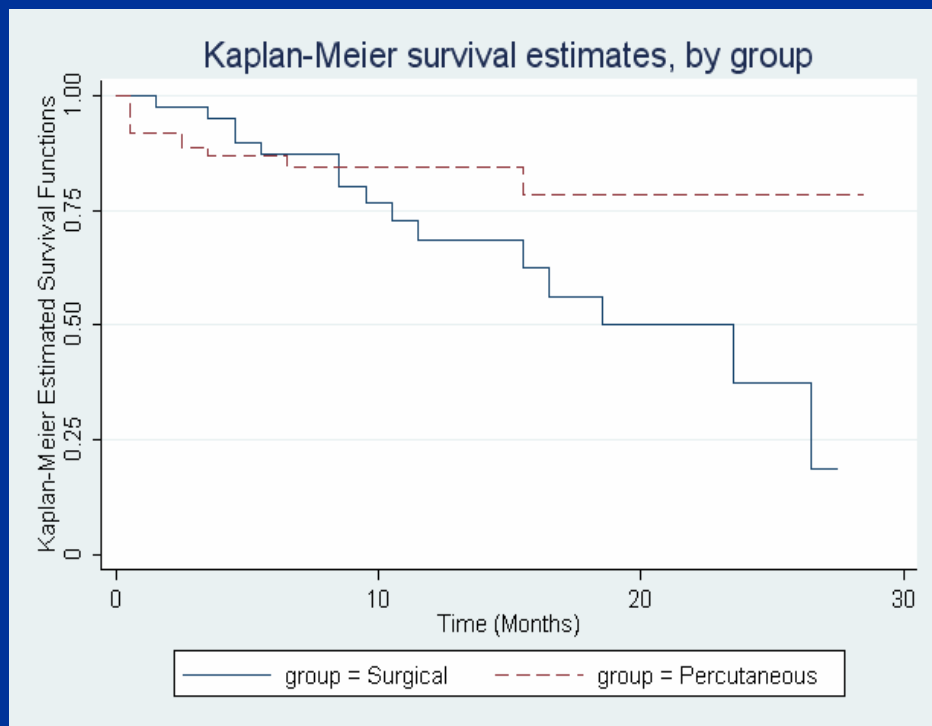
Time to infection of kidney dialysis patients with different catheterization procedures

(Nahman *et al* 1992, Klein & Moesberger 2003)

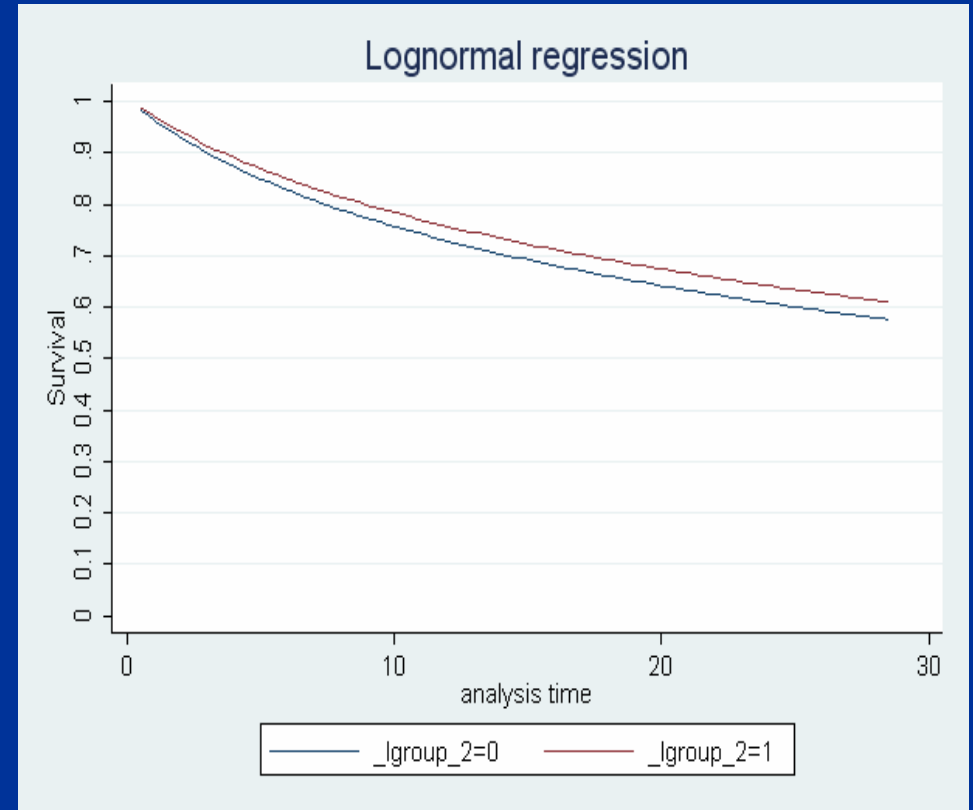
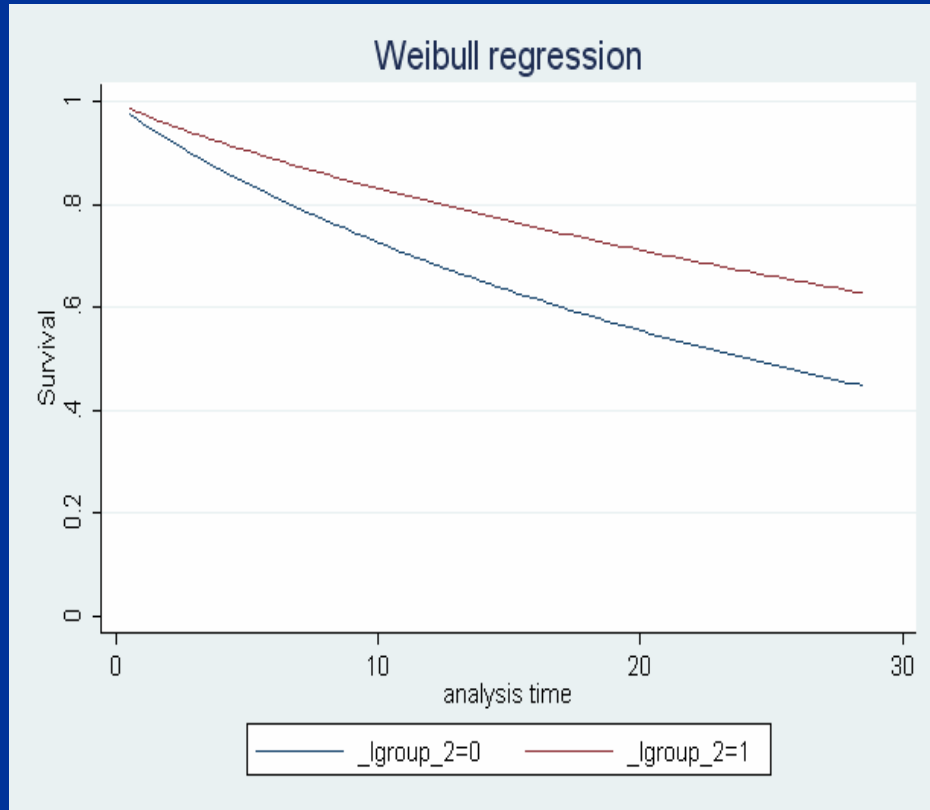
- Surgical group:
43 patients utilized a surgically placed catheter
- Percutaneous group:
76 patients utilized a percutaneous placement of their catheter

The survival time is defined by the time to cutaneous exit-site infection.

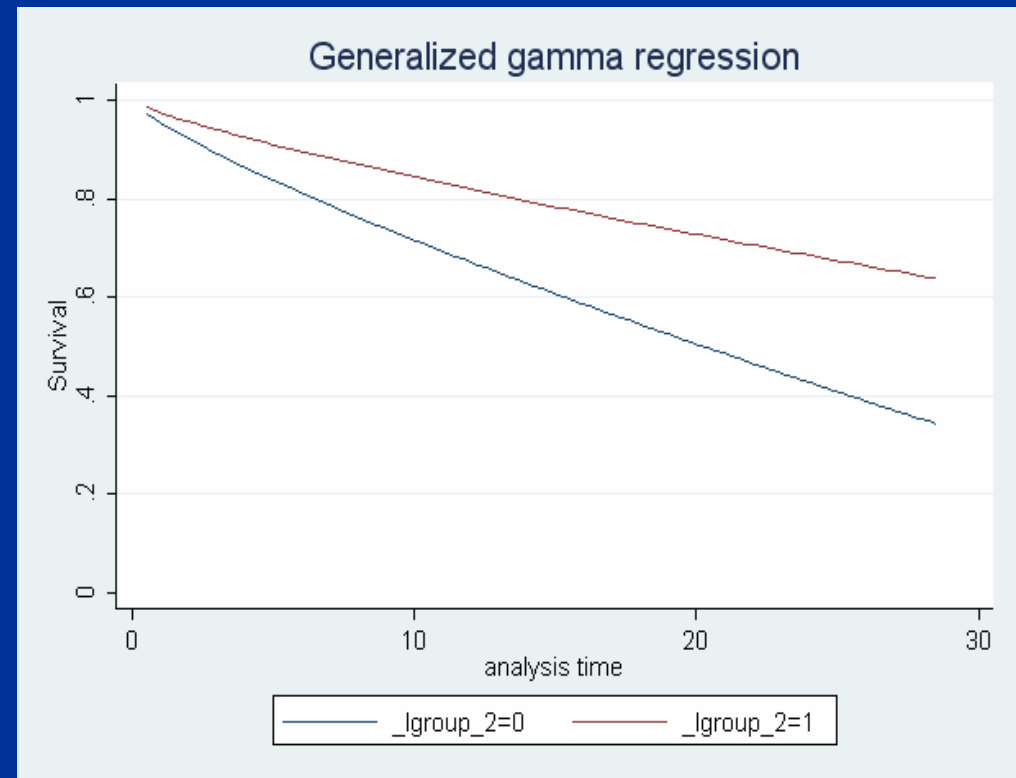
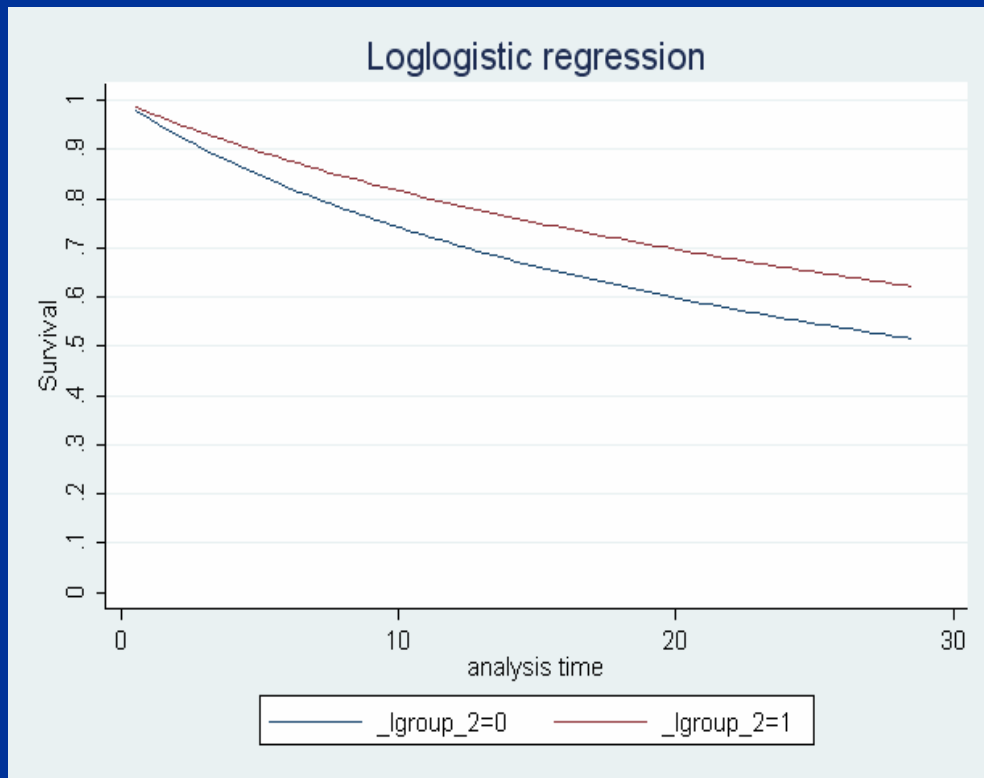
Kaplan-Meier Estimate versus PH Cox Model



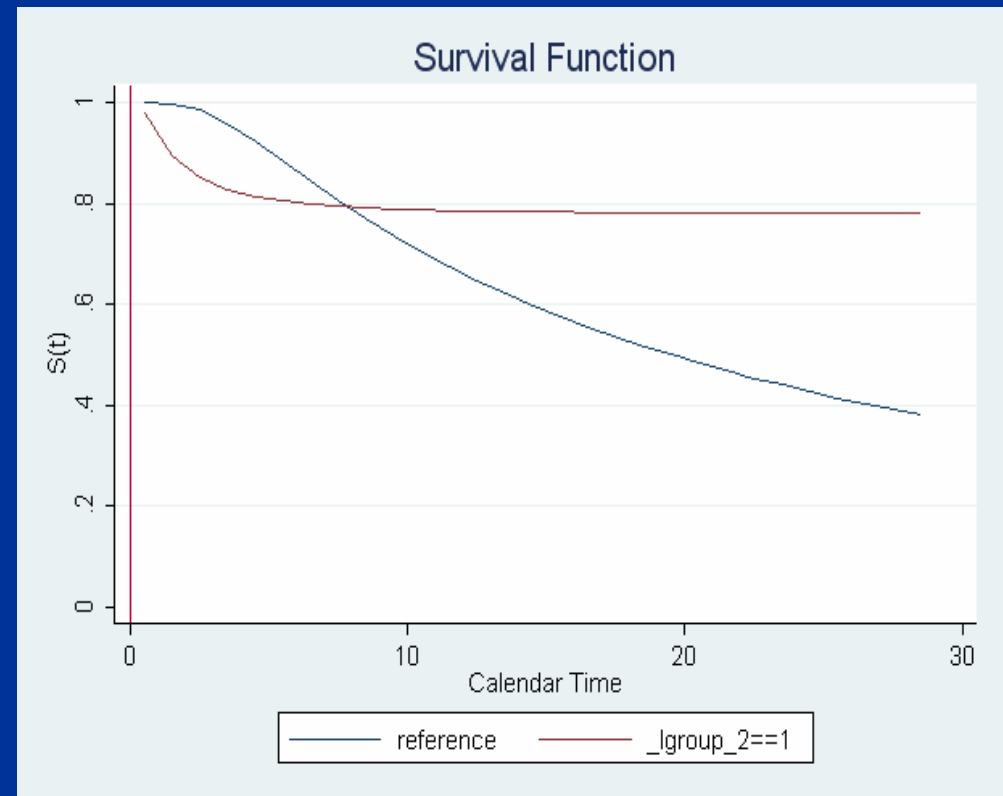
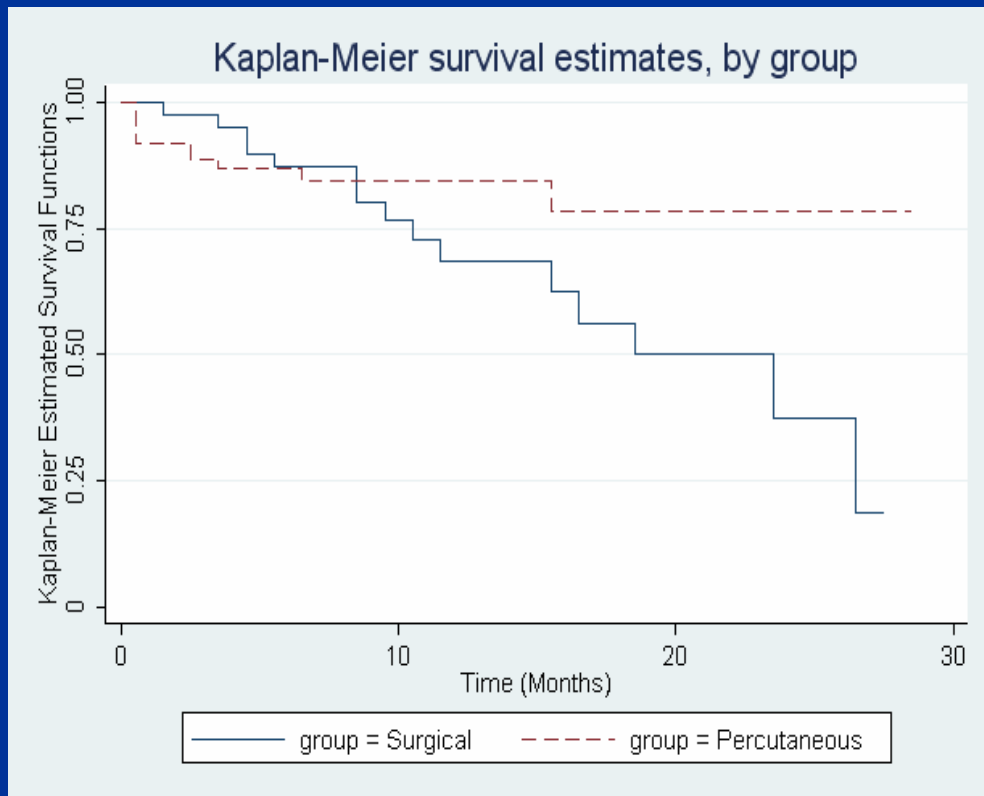
Weibull versus Lognormal



Loglogistic versus Gamma



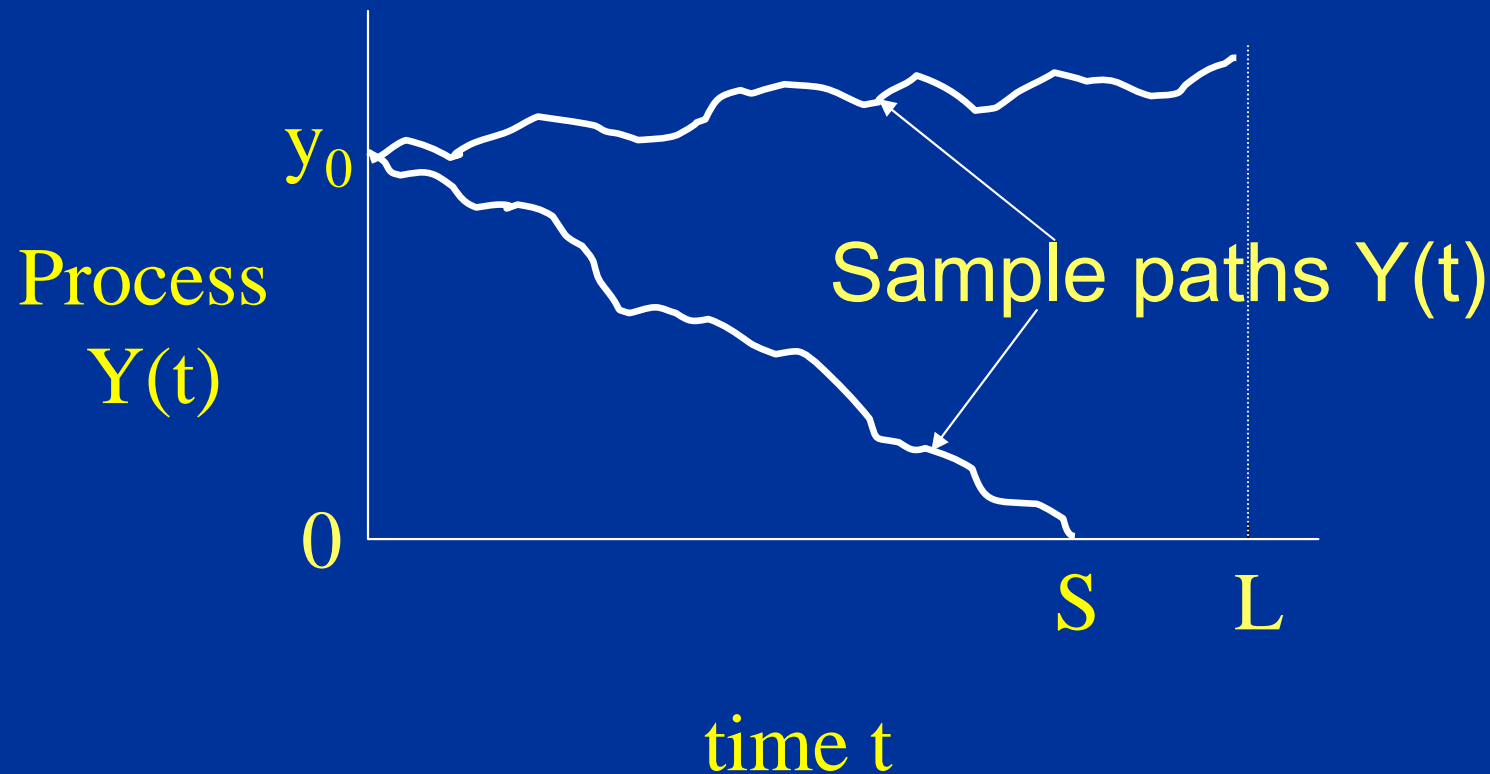
Kaplan-Meier Estimate versus First-hitting-time based Threshold Regression Model



First-hitting Time Based Threshold Regression:

Modeling Event Times by a Stochastic Process Reaching a Boundary (Lee & Whitmore 2006, *Statistical Sciences*)

- **Example: Equipment Failure**
Equipment fails when its cumulative wear **first** reaches a **failure threshold**.
- **Example: Health research**
People died at heart failure, lung failure, etc



Two sample paths of a stochastic process of interest:

- (1) One path experiences 'failure' at first hitting time S**
- (2) One path is 'surviving' at end of follow up at time L**

Parameters for the FHT Model

Model parameters for the latent process $Y(t)$:

- Process parameters: $\theta = (\mu, \sigma^2)$, where
 - μ is the mean drift and σ^2 is the variance
- Baseline level of process: $Y(0) = y_0$
- Because $Y(t)$ is latent, we set $\sigma^2 = 1$.

Likelihood Inference for the FHT Model

The likelihood contribution of each sample subject is as follows.

- If the subject fails at $S=s$:

$$f(s | y_0, \mu) = \Pr [\text{first-hitting-time in } (s, s+ds)]$$

- If the subject survives beyond time L :

$$1 - F(L | y_0, \mu) = \Pr [\text{no first-hitting-time before } L]$$

$$\ln L(\theta, x_0) = \sum_{i=1}^n \left\{ d_i \ln f(t_i | \theta, x_0) + (1 - d_i) \ln \bar{F}(t_i | \theta, x_0) \right\}.$$

where

d_i is the failure indicator for subject i

t_i is a censored survival time ($t_i = s_i$ if subject i fails)

f and \bar{F} denote the FHT p.d.f and complementary c.d.f.

Threshold Regression

Link Functions: parametric or semi-parametric

Possible Link functions for the baseline parameter

$Y(0)$ and drift parameter μ include

- Linear combinations of covariates X_1, \dots, X_p
- polynomial combinations of X_1, \dots, X_p
- Regression splines
- Penalized regression splines
- Random effects

Threshold regression (TR)

Regression estimates for parameters of:

1. **Process $Y(t)$:** Wiener process, gamma process, etc
2. **Boundary:** straight lines or curves
3. **Time scale:** calendar or running time, analytical time

References

- **Aalen O.O. and Gjessing H.K. (2001). Understanding the shape of the hazard rate: a process point of view. *Statistical Science*, 16: 1-22.**
- **Lawless, J. F. (2003). *Statistical Models and Methods for Lifetime Data*, Second Edition, Wiley.**
- **Lee, M.-L. T. and G. A. Whitmore (2006). Threshold regression for survival analysis: modeling event times by a stochastic process reaching a boundary. *Statistical Sciences*.**
- **Aalen O.O., Borgon O, and Gjessing H.K (2008). *Survival and Event History Analysis: A process Point of View*. Springer.**

