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Accelerated Failure Time Model	
Introduction	
introduction	
	1
	1
Example: Kidney transplant	
<ul> <li>total of 863 Kidney transplant patients were</li> </ul>	
followed until death or censoring.	
Patients were censored if they moved away or still	
<ul><li>alive at the end of the study.</li><li>Goal: Investigate the effects of risk factors such as</li></ul>	
Gender, Race and Age on survival of Kidney transplant patients.	
Introduction	
In linear regression model, we model the	
relationship between predictors to the outcome through	
$Y_i = \beta_0 + \beta_1 x_{i1} + \dots + \beta_1 x_{ip} + \varepsilon_i.$	
<ul> <li>In generalized linear regression model, for example, logistic regression, we assume</li> </ul>	
$\operatorname{logit} \pi_{i1} = \beta_0 + \beta_1 x_{i1} + \dots + \beta_1 x_{ip}$	

#### Introduction

Similarly, we will focus the following two approaches to relate the outcome to predictors in survival analysis.

- Accelerated failure time model (Parametric regression)
- Cox's proportional hazard model (Semiparametric regression)

### Accelerated Failure Time (AFT) model

- Let T be the true survival time,  $\mathbf{x} = (x_1, \dots, x_p)^T$  be explanatory variables.
- The accelerated failure time model assumes  $Y = \log(T) = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \sigma \varepsilon,$  where
  - $(\beta_0,\ldots,\beta_p)$  are regression coefficients,
  - $-\sigma$  i a scale parameter
  - $-\ \epsilon$  is random error with mean 0 and standard deviation 1.
- It is also referred to as log-linear model.

#### Accelerated Failure Time (AFT) model

• The AFT model assumes that

$$Y = \log(T) = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \sigma \varepsilon,$$

- It models the survival time T directly, which makes it more appealing compared with other regression methods because of this quite direct physical interpretation.
- It is often used in engineering for modeling reliability, but relatively uncommon in medical sciences.

## Log-normal model

• The AFT model assumes that

$$Y = \log(T) = \beta_0 + \beta_1 x_1 + \dots + \beta_p x_p + \sigma \varepsilon.$$

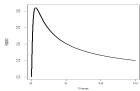
- A different distribution of  $\varepsilon$  gives different regression model.
- For example, an obvious choice is to assume that

$$\varepsilon \sim N(0,1)$$
.

• Then T follows a lognormal distribution.

## Log-normal model

• For lognormal distribution, the hazard is of form



May not be accurate for some real data applications.

# Some popular AFT models

# $\begin{array}{ll} \mbox{Distribution} & \mbox{Distribution of } T \\ \mbox{of } \varepsilon \\ \mbox{Extreme} & \mbox{Weibull } (\sigma = 1 \mbox{ reduce to} \end{array}$

values exponential)
Log-Gamma Gamma
Logistic Log-logistic
Normal Log-normal

