

Survival Data Analysis

Dr Nick Fieller

Department of Probability & Statistics

University of Sheffield

visiting



16 April – 27 April 2012

 © NRJF, University of Sheffield, 2012

Survival Data Analysis

3

Contents

Preliminaries

0: Introduction

1: Background & Basic Concepts

2: Single Sample Methods

2.1 Non-parametric Methods

2.6 Parametric Models

3: Two-Sample Comparisons

4: Regression Models

4.2 Parametric Regression Models

4.4 Proportional Hazards models

 © NRJF, University of Sheffield, 2012

Survival Data Analysis

2

Contents

Preliminaries

0: Introduction

1: Background & Basic Concepts

2: Single Sample Methods

2.1 Non-parametric Methods

2.6 Parametric Models

3: Two-Sample Comparisons

4: Regression Models

4.2 Parametric Regression Models

4.4 Proportional Hazards models

 © NRJF, University of Sheffield, 2012

Survival Data Analysis

3

Course web page

<http://nickfieller.staff.shef.ac.uk/tampere12/index.html>

Lecture notes and other course material

 © NRJF, University of Sheffield, 2012

Survival Data Analysis

4

Purpose of Lectures

- There are 'complete' printed notes
 - ◆ These are not a textbook
 - some explanations are omitted
 - ◆ They are intended to allow you to concentrate on understanding & for me to cover some material very quickly
- Some lectures will be very close to the printed notes
 - ◆ This is **intended**
- Other lectures will fill in details & provide examples & R demos

 © NRJF, University of Sheffield, 2012

Survival Data Analysis

5

Books

◆ Campbell, M. J. (2001)

Statistics at Square Two. BMJ

◆ Collett, D. (2003)

Modelling Survival Data in Medical Research. (2nd Edition) Chapman & Hall

◆ Everitt, Brian & Rabe-Heskith, Sophia (2001) *Analyzing Medical Data Using S-PLUS*. Springer. Support material at

<http://web1.iop.kcl.ac.uk/loP/Departments/BioComp/splusBook.shtml>

 © NRJF, University of Sheffield, 2012, Semester 1

Survival Data Analysis

6



Contents

Preliminaries

0: Introduction

1: Background & Basic Concepts

2: Single Sample Methods

2.1 Non-parametric Methods

2.6 Parametric Models

3: Two-Sample Comparisons

4: Regression Models

4.2 Parametric Regression Models

4.4 Proportional Hazards models



Survival Data Analysis

7

Objectives

- ◆ statistical modelling & analysis of *lifetime data*.
- ◆ Lifetime data arise especially in medical statistics and in reliability studies.
 - *survival time*:
 - time from diagnosis to death of a patient
 - *time to recovery or remission* of a patient or
 - *time to failure* of an electronic component.



Survival Data Analysis

8

Outline

- ◆ Types of survival data
- ◆ Censoring
- ◆ Parametric & nonparametric approaches

Single Sample Methods

- ◆ survivor & hazard functions
- ◆ Lifetables
- ◆ Kaplan-Meier estimators



Survival Data Analysis

9

Two Sample Comparisons:

- ◆ log rank test
- ◆ maximum likelihood test
- ◆ likelihood ratio test
- ◆ proportional hazards



Survival Data Analysis

10

Regression Models

- ◆ parametric models
 - exponential & Weibull
- ◆ nonparametric methods
 - proportional hazards or Cox regression
 - partial likelihood



Survival Data Analysis

11

Aims of survival analysis

- ◆ describe/model a single sample
 - making inferences on a single population
- ◆ compare two or more groups
 - effect of treatments on survival time
- ◆ investigate relationship with covariates
 - effects on survival time of covariates
 - adjust for covariates in comparisons



Survival Data Analysis

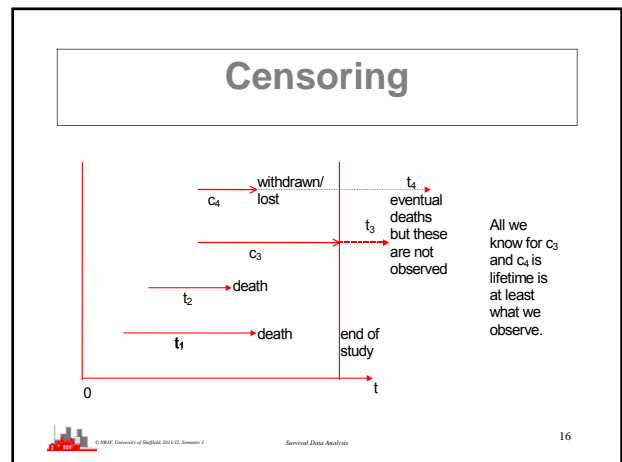
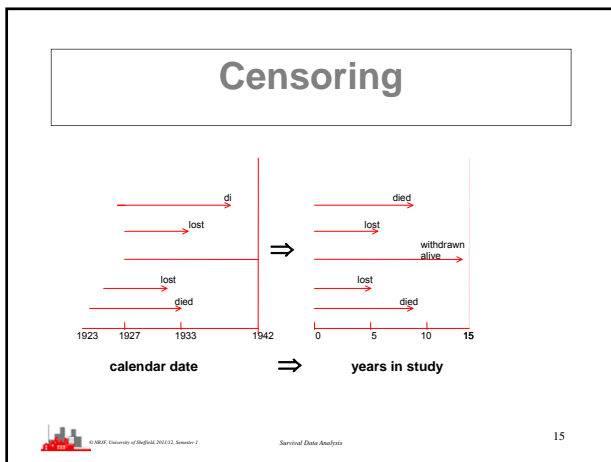
12



Contents

- Preliminaries
- 0: Introduction
- 1: **Background & Basic Concepts**
- 2: Single Sample Methods
 - 2.1 Non-parametric Methods
 - 2.6 Parametric Models
- 3: Two-Sample Comparisons
- 4: Regression Models
 - 4.2 Parametric Regression Models
 - 4.4 Proportional Hazards models

- **Censoring**
 - ◆ event of interest not yet occurred
 - ◆ observe only survival time is **at least** t
 - ◆ some observations t_i ($i=1, \dots, n$) are **censored**
 - ◆ censored observations provide information on survival so cannot be ignored




- ### Censoring
- **Right censoring**
 - ◆ lifetime exceeds some value
 - **Left censoring**
 - ◆ lifetime less than some value
 - **Interval censoring**
 - ◆ failure occurred during an interval

- **Aims:**
 - ◆ **estimate lifetime distributions**
 - \Rightarrow estimate properties of distribution
 - (median lifetime, prob of surviving > 5 years, ...)
 - ◆ **Censoring**
 - nonparametric — lifetables, Kaplan-Meier
 - parametric — exponential, Weibull.




- survival time is a random variable T
- $T > 0$, continuous variable
 - ◆ p.d.f. $f(t)$ ($t > 0$)
 - ◆ d.f. $F(t) = P[T \leq t]$
 - ◆ (so $f(t) = F'(t)$ and $F(t) = \int_0^t f(u) du$)




©NRJF, University of Sheffield, 2012/13, Semester 1 Survival Data Analysis 19

- **Survivor function**
 - ◆ $S(t) = P[T > t] = 1 - F(t)$
 - ◆ (so $S'(t) = -f(t)$, $S(t) = \int_t^\infty f(u) du$)
- **Hazard function**
 - ◆ $h(t) = \lim_{\delta t \rightarrow 0} \left[\frac{P[t \leq T < t + \delta t \mid T \geq t]}{\delta t} \right]$
 - ◆ “P[die at time T given survived until T]”



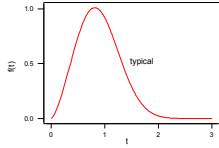
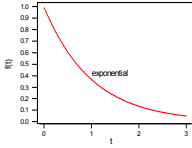
©NRJF, University of Sheffield, 2012/13, Semester 1 Survival Data Analysis 20


- **Cumulative hazard function**
 - ◆ $H(t) = \int_0^t h(u) du = -\log_e S(t)$
 - ◆ $f(t)$, $S(t)$, $H(t)$ and $h(t)$ are equivalent characterizations of a survival distribution
 - ◆ all interrelated



©NRJF, University of Sheffield, 2012/13, Semester 1 Survival Data Analysis 21

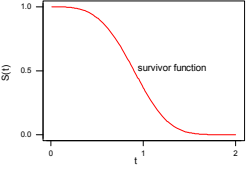

- **Typical patterns**



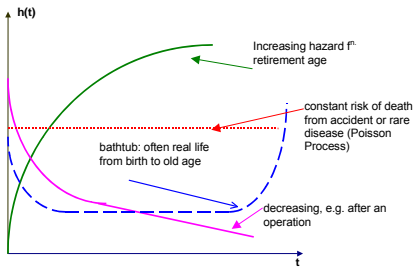
©NRJF, University of Sheffield, 2012/13, Semester 1 Survival Data Analysis 22

- **Survivor function**

©NRJF, University of Sheffield, 2012/13, Semester 1 Survival Data Analysis 23

Hazard functions:




Increasing hazard $h(t)$ retirement age

constant risk of death from accident or rare disease (Poisson Process)

bathtub: often real life from birth to old age

decreasing, e.g. after an operation



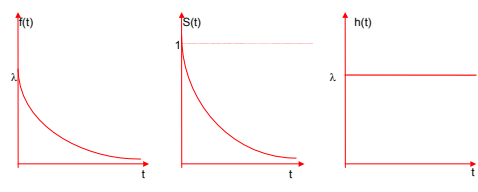
©NRJF, University of Sheffield, 2012/13, Semester 1 Survival Data Analysis 24



- Choose appropriate family of models by recognizing form of hazard function
 - ◆ practical situation
 - ◆ initial investigation of the data
- then can estimate parameters in model

- **Example:** exponential
 - ◆ $f(t) = \lambda e^{-\lambda t}$
 - ◆ $S(t) = e^{-\lambda t}$
 - ◆ $h(t) = \lambda$ (**NB** constant)
 - $[= f(t)/S(t)]$
- **only** constant for exponential survival distribution

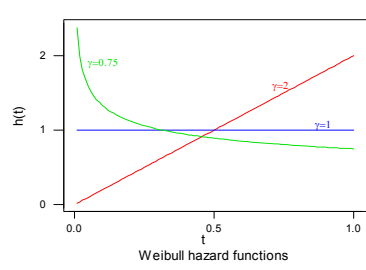
- exponential



- **Example:** Weibull

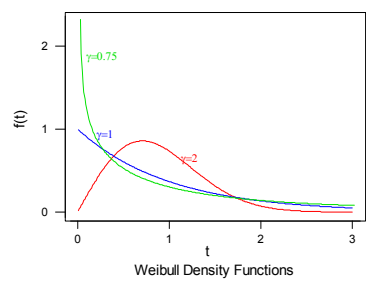
- ◆ $f(t) = \lambda \gamma t^{\gamma-1} \exp(-\lambda t^\gamma)$
- ◆ $S(t) = \exp(-\lambda t^\gamma)$
- ◆ $h(t) = \lambda \gamma t^{\gamma-1}$
 - $\gamma > 1$: increasing
 - $\gamma = 1$: constant (exponential)
 - $\gamma < 1$: decreasing

- e.g. $\lambda = 1$; Weibull hazard functions



Weibull hazard functions

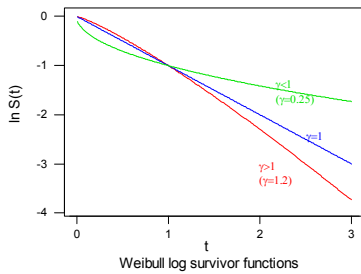
- e.g. $\lambda = 1$; Weibull density functions



Weibull Density Functions



- e.g. $\lambda = 1$; Weibull logsurvivor functions



- Weibull family
 - ♦ very flexible models
 - ♦ allows both increasing ($\gamma > 1$) & decreasing ($\gamma < 1$) hazard functions
- Difficult to estimate if γ close to 1
 - (numerical instability)



© NRJF, University of Sheffield, 2012. Semester 1

Survival Data Analysis

32



© NRJF, University of Sheffield, 2012

Survival Data Analysis

33



© NRJF, University of Sheffield, 2012

Survival Data Analysis

34



© NRJF, University of Sheffield, 2012

Survival Data Analysis

35



© NRJF, University of Sheffield, 2012

Survival Data Analysis

36

