

23/02/16

ALSM 2

Response Variable:

$y \in \{0,1\}$ Bernoulli

$y \in \{0,1,\dots,n\}$ Binomial

$y \in \mathbb{N}$ (integer) Poisson

$y \in \mathbb{R}$ Gaussian

$y \in \mathbb{R}^+$ exponential Weibull \rightarrow positive real numbers - time until failure

Life time example - response variable y is time, is placebo time same as drug time?

SURVIVAL ANALYSIS

- Modelling time until failure

- SA concerned with the statistical modelling of time to failure from a well defined origin of time or starting point

\rightarrow Time of hardware to fail since bought.

\rightarrow Time of patient to die from the time doctor has been diagnosed.

Distribution

- Times y are non-negative ($y \in \mathbb{R}^+$)

- Distributions are skewed with long tails

- Some subjects may survive beyond the study and their failure is not observed

\rightarrow In this case, the data is said to be 'censored'

Exponential Distribution

$$P(y|\theta) = \theta \exp[-\theta y] \quad y \in \mathbb{R}^+ \quad \theta \in \mathbb{R}^+ \quad \leftarrow \text{excludes 0.}$$

Expectation $E[y]$?

$$E[y] = \int_{y \in \mathbb{R}^+} y P(y|\theta) dy$$

$$= \int_0^\infty y \lambda \exp(-\lambda y) dy$$

$$= \int_0^\infty y \exp(-\lambda y) dy + \int_0^\infty \exp(-\lambda y) dy \quad (\text{integration by parts})$$

$$= [0-0] + \left[-\frac{1}{\lambda} \exp(-\lambda y) \right]_0^\infty$$

$$= 0 + 0 + \frac{1}{\lambda} = \frac{1}{\lambda} = \text{positive number}$$

Prove it is a function $\int_{-\infty}^{\infty} f(x) dx = \int_{-\infty}^{\infty} \lambda \exp(-\lambda y) dy$
 $= [\exp(-\lambda y)]_{-\infty}^{\infty}$
 $= 0 - (-1) = 1 \Rightarrow \text{Distribution}$

Best how function to map θ to positive real number \mathbb{R}^+ onto \mathbb{R} g should be exponential, g^{-1} should be log

Weibull Distribution

$$P(y|\theta, \lambda) = \lambda \theta y^{\theta-1} \exp[-\theta y^\lambda]$$

$y \in \mathbb{R}^+ \quad \theta \in \mathbb{R}^{++} \quad \lambda \in \mathbb{R}^{++}$

Exponential distribution is a special case of Weibull with $\lambda=1$

$$E[y] = \int_0^{\infty} y \lambda \theta y^{\theta-1} \exp[-\theta y^\lambda] dy$$

$$= \int_0^{\infty} \lambda \theta y^\theta \exp[-\theta y^\lambda] dy$$

Using substitution $y^\lambda = u^{\frac{1}{\lambda}}$ result in:

$$dy = \lambda y^{\lambda-1} du \quad dx = \frac{dy}{\lambda y^{\lambda-1}} = \frac{du}{\lambda u^{\frac{\lambda-1}{\lambda}}} = u^{\frac{1}{\lambda}-1} du$$

$$E[x^\lambda] = \int \lambda y^{\lambda\theta-1} e^{-y^\lambda} dy$$

$$= \int \lambda u^{\frac{\lambda\theta-1}{\lambda}} \frac{1}{\lambda} e^{-u} (u^{\frac{1}{\lambda}-1}) du$$

$$= \int u^{(\theta-1)+1} e^{-u} du$$

$$= \Gamma(\theta+1) \quad \Gamma(u) = \int_0^{\infty} s^u \exp(-s) ds$$

Survivor and Hazard Function

A. Probability of Failure $P(0 \leq t \leq y) = \int_0^y P(t|t) dt$

P can be exponential or Weibull (area between 0 and y)

Probability of Survival beyond time y

$$S(y) = 1 - F(y) \quad (\text{area to right of } y)$$

B. $F(y)$ and $S(y)$ of exponential Distribution

$$F(y) = \int_0^y \theta \exp(-\theta t) dt$$

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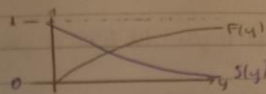
$$= [-\exp(-\theta t)]'$$

$$= 1 - \exp(-\theta y) = \text{probability of failure}$$

$$S(y) = 1 - 1 + \exp(-\theta y) = \exp(-\theta y)$$

- At 0.5 some probability of failure is reached

- When $F(y) = 0.5$ or $S(y) = 0.5$ called the median point of survival $F(y) = S(y) = 0.5$



C. Hazard Function

$$h(y) = \frac{P(y \leq t \leq y+dy)}{P(y \leq t)} \quad \begin{array}{l} \text{chance of failure between time } y \text{ and } y+dy \\ \text{Surviving beyond time } y \text{ } S(y) \end{array}$$

$P(y|t \leq y)$

*Chances of dying in the next minute given a survival rate and given you have not died before time y

$$h(y) = P(y|t) / S(y) \quad \text{derivative/function Weibull or exponential}$$

$$\text{Cumulative hazard function } H(y) = -\text{Log}[S(y)]$$

D. Link Function

Use Log

Exponential Distribution

$$F(t) = \lambda e^{-\lambda t} \quad \lambda > 0$$

$$F(t) = 1 - e^{-\lambda t} \quad S(t) = e^{-\lambda t}$$

$$h(t) = \lambda \quad \text{constant hazard function}$$

$$H(t) = \lambda t$$

Weibull Distribution

$$F(t) = p\lambda^p t^{p-1} e^{-(\lambda t)^p}$$

$$F(t) = 1 - e^{-(\lambda t)^p} \quad S(t) = e^{-(\lambda t)^p}$$

$$h(t) = p\lambda^p t^{p-1}$$

$$H(t) = (\lambda t)^p \quad \begin{array}{l} t > 0 \\ \lambda > 0 \end{array} \quad \begin{array}{l} p > 0 \text{ shape} \\ p > 0 \text{ scale} \end{array}$$

Drug Example Continued

$$L(\beta) = \prod_{i=1}^n P(y_i | \theta) \quad \text{Weibull or exponential}$$

- This is the definition of the likelihood when y_i is uncensored.

- When y_i is censored, use $S(y_i)$ instead of pdf: $P(y_i | \theta)$

- δ_i an indicator variable, 0 if y_i is censored, 1 if it is not $\delta_i = \begin{cases} 0 & \text{censored} \\ 1 & \text{observed} \end{cases}$

$$\hookrightarrow L(\beta) = \prod_{i=1}^n P(y_i | \theta)^{\delta_i} S(y_i)^{1-\delta_i} \quad \text{take either term depending on censored or not}$$

- Censor: if you lose track of patients or survives duration of study

Example: Exponential Distribution $L(t_0, t_1, \theta, x) = \prod_{i=1}^n [0.1 \exp(-0.1 x_i)]^{d_i} [\exp(-0.1 x_i)]^{1-d_i}$
 $L(\theta) = \prod_{i=1}^n \exp(-x_i \theta)$

- GMM not defined for use of exponential/Weibull use "survival" package

Is the drug better than the placebo effect?

- look at median survival time, $y_{med} = \frac{\log(2)}{\theta}$ for exponential dist.
 $= \frac{\log(2)}{\exp(\beta'x)}$ $\begin{cases} x=0 \text{ placebo} \\ x=1 \text{ drug (for } x=1) \end{cases}$

$\frac{y_{med}^{x=1}}{y_{med}^{x=0}} = \frac{1}{\exp \beta} \frac{\log(2)}{\exp(-1\beta)} = 1/4$ look at ratio between drug and placebo
 $\frac{y_{med}^{x=0}}{4} = y_{med}^{x=1}$

4 times higher than treatment group

Probability is 4 times higher placebo group - 4 times more likely to fail - placebo group

R gives p's estimates and corresponding S.E.'s

$2.72 \pm 2(0.01)$ 95% CI for $\beta_1 \Rightarrow \beta_1$ is not 0

\rightarrow relationship between time and drug exists

Exponential Distribution $E[y] = 1/\theta$

$\log[E[y]] = \log(1/\theta) = -\log \theta$

Map \mathbb{R}^+ onto \mathbb{R}

$\log \theta = \beta_0 + \beta_1 x_i$

$\theta = \exp[\beta_0 + \beta_1 x_i]$

- Look at hazard function or median to see if drug is working

Weibull Distribution

1. Show it is a distribution

α positive, θ positive, \exp is positive \Rightarrow positive function

Must show it integrates to 1

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ATSM 2

$$\int_0^{\infty} \lambda \theta y^{\lambda-1} \exp[-\theta y^{\lambda}] dy$$

$$= [-\exp[-\theta y^{\lambda}]]_0^{\infty} \quad \text{at } 0=1 \quad \text{at } \infty=0$$

$$1+0=1$$

Probability of Failure: $1 - \exp[-\theta y^{\lambda}]$ Probability of Survival: $1 - F(y)$

$$\text{Hazard function } h(y) = \frac{f(y)\lambda\theta}{S(y)} = \lambda \theta y^{\lambda-1} \quad (\text{accelerated failure time, function of } y)$$

Exponential case $\lambda=1$ $h(y)$ is $\theta \Rightarrow$ not a function of y \hookrightarrow Independent of age "Chance of survival same at 100 as at 1000" \hookrightarrow lack of memory $\rightarrow y$ not important, may be a limitation

2. Show it is a member of exponential family of distribution!

$$= \lambda \theta y^{\lambda-1} \exp[-\theta y^{\lambda}]$$

$$= \exp[-\theta y^{\lambda}] \exp[\log(\lambda \theta y^{\lambda-1})]$$

$$= \exp\left[\underbrace{-\theta y^{\lambda}}_{b(\theta)} \underbrace{\log(\lambda)}_{a(y)} + \underbrace{\log(\theta)}_{c(\theta)} + \underbrace{\log(y^{\lambda-1})}_{d(y)}\right]$$

3. Expectation

$$E[y] = \int_0^{\infty} y \lambda \theta y^{\lambda-1} \exp[-\theta y^{\lambda}] dy$$

Show $E[y] = (\frac{1}{\theta})^{1/\lambda} \Gamma(1 + \frac{1}{\lambda})$ with $\Gamma(u) = \int_0^{\infty} s^{u-1} \exp(-s) ds$

Substitution: $\mu = \theta y^{\lambda} \quad d\mu = \theta \lambda y^{\lambda-1} dy$

$$= \int_0^{\infty} y \lambda \mu \exp(-\mu) d\mu$$

$$E[y] = \int_0^{\infty} \left(\frac{\mu}{\theta}\right)^{1/\lambda} \exp(-\mu) d\mu$$

$$\log(\theta) = -3.07 - 1.731x \quad x \text{ is indicator for drug}$$

The log rescales the β values: $2.48 + 1.267$ (R output numbers)