

## Introduction

1. Survival func:  $S(t) = 1 - F(t) = P(T > t) = \int_t^\infty f(u)du$   
 $f(u) = -\frac{d}{du}S(u)$ ,  $S(0) = 1$ ,  $S(\infty) = 0$
2. Empirical survival func:  $\hat{S}_n(t) = \frac{1}{n} \sum_{i=1}^n I_{(t, \infty)}(T_i)$ ,  $n\hat{S}_n(t) \sim \text{Bin}(n, S(t))$
3. Hazard func:  $h(t) = \frac{f(t)}{S(t)}$ ,  $S(t) = \exp\left(-\int_0^t h(u)du\right) = \exp(-H(x))$ ,  $E[t] = \int_0^\infty S(t)dt$

## Mortality tables

4. Prob of dying:  $q_x = \frac{d_x}{l_x} = \frac{l_x - l_{x+1}}{l_x}$
5. Prob of surviving:  $p_x = 1 - q_x = \frac{l_{x+1}}{l_x}$
6.  $e_x = \sum_{i=1}^{\infty} i P[\text{death in year } (x+i)] = \frac{\sum_{n=x+1}^{\infty} l_n}{l_x}$
7. Force of motality:  $\mu_x = -\frac{1}{l} \frac{dl_x}{dx} = -\log(p_x)$
8. Mortality rate:  $m_x = \frac{\text{number of deaths}}{\text{person-time at risk}} = \frac{d_x}{\int_x^{x+1} l_t dt} = \frac{d_x}{\int_0^1 l_{x+t} dt}$   
 Central mortality rate:  $m_x = \frac{d_x}{l_x - \frac{d_x}{2}}$
9. Prob that die within  $n$  yrs:  ${}_nq_x = \frac{l_x - l_{x+n}}{l_x}$
10. Prob of surviving  $n$  yrs:  ${}_np_x = \frac{l_{x+n}}{l_x} = 1 - {}_nq_x$
11. Prob that die in age  $x+m \sim x+m+n$ :  ${}_m|{}_nq_x = \frac{l_{x+m} - l_{x+m+n}}{l_x} = {}_mp_x \cdot {}_nq_{x+m}$
12.  ${}_np_x = \exp\left(-\int_0^n \mu_{x+t} dt\right)$ ,  $l_x = \exp\left(-\int_0^x \mu_t dt\right)$
13. Linear interpolation (increasing  $\mu$ , deaths Uniformly distributed)  
 ${}_tl_x = l_x + t(l_{x+1} - l_x) = l_x - td_x$ ,  ${}_tq_x = tq_x$ ,  $\mu_{x+t} = \frac{q_x}{1-tq_x}$
14. Exponential interpolation (constant  $\mu$ )  
 ${}_tl_x = l_x \left(\frac{l_{x+1}}{l_x}\right)^t = l_x p_x^t$ ,  $\mu_{x+t} = -\log\left(\frac{l_{x+1}}{l_x}\right) = -\log(p_x)$
15. Hyperbolic interpolation (decreasing  $\mu$ )  
 ${}_tl_x = \frac{l_x p_x}{p_x + tq_x}$ ,  ${}_tp_x = \frac{p_x}{p_x + tq_x}$ ,  $\mu_{x+t} = \frac{q_x}{p_x + tq_x} = \frac{q_x}{1 - q_x(1-t)}$
16. Select table:  $l_{[x+1]} = l_x \cdot p'_x$  (forward, then calc all new  $l$ s),  $l_{[x]}$  (for backward only)

## Parametric models and estimation

17. de Moivre hazard:  $h(t) = \frac{1}{\omega - t}$ , where  $l_\omega = 0$  or negligible  
 $f(t) = \frac{1}{\omega}$  (deaths Uniformly distributed)

18. Weibull hazard:  $h(t) = \gamma t^{(\gamma-1)}$ ;  $t > 0, \gamma > 0, H(t) = t^\gamma$   
 $\gamma < 1$ : decreasing;  $\gamma = 1$ : constant,  $\gamma > 1$ : increasing  
 Generalised with a scale parameter  $\alpha$  and a location parameter  $\mu$ :  $h(t) = \frac{\gamma}{\alpha} \left( \frac{t - \mu}{\alpha} \right)^{\gamma-1}$
19. Gompertz hazard:  $h(t) = Bc^t$ , logarithm is linear in  $t$   
 $H(t) = -(c^t - 1) \ln g$ , where  $\ln g = \frac{-B}{\ln c}$   
 Gompertz-Makeham hazard:  $h(t) = A + Bc^t$  (adult mortality in developed countries)
20. Log-linear models:  $\log(T) = \alpha + \sigma\varepsilon$
21. 1

## Non-par and dist-free approaches

22. Kaplan-Meier estimator:  $\hat{S}(t) = \prod_{\{j:t_j \leq t\}} \left( 1 - \frac{d_j}{n_j} \right)$   
 Variance:
23. Nelson-Aalen estimator:  $\hat{H}(t) = \sum_{\{j:t_j \leq t\}} \left( \frac{d_j}{n_j} \right)$   
 Breslow estimator:  $\hat{S}(t) = \exp(-\hat{H}(t))$   
 (greater than K-M estimator by Taylor series, asymptotically equivalent)
24. Logrank test:

## Semi-par surv modelling

25. Cox PH model:  $h(t, X) = h_0(t)\psi(X; \beta)$ , commonly  $\psi(X; \beta) = \exp(\beta^T X)$  log-linear form  
 Likelihood  $L(\beta) = \prod_{j=1}^d \frac{\psi(i_j)}{\sum_{k \in R(\tau_j)} \psi(k)}$ ,  $I(\hat{\beta}) = -\frac{\partial^2 \ln L(\hat{\beta})}{\partial \beta^2}$ ,  
 95% confidence interval for  $\beta$ :  $\hat{\beta} \pm z_{\alpha/2} \sqrt{\frac{1}{I(\hat{\beta})}}$