

So succumbed due to decrement 1. 402  
27

$$\text{Soln} \quad \mu_x^{(1)}(10) = \frac{1}{40} = 0.025$$

$$\mu_x^{(2)}(10) = 0.01$$

$$\mu_x^{(3)}(10) = 0.0015(10) = 0.015$$

$$\text{and } \mu_{X(10)}^{(r)} = \sum_{j=1}^3 \mu_x^{(j)}(10) = 0.05$$

$$\text{so prob} = \frac{\mu_x^{(1)}(10)}{\mu_{X(10)}^{(r)}} = \frac{0.025}{0.05} = \underline{0.5}$$

38] To go from multiple to single  
decrement quantities

<sup>T</sup>  
primes

$$\text{i) calculate } tP_x^{(T)} = 1 - t\delta_x^{(T)} = 1 - \sum_{j=1}^m t\delta_x^{(j)}$$

$$\text{ii) calculate } \mu_{x+t}^{(j)} = \frac{d}{dt} t\delta_x^{(j)} \over tP_x^{(T)}$$

$$\text{iii) calculate } tP_x^{(j)} = \exp \left[ - \int_0^t \mu_{x+s}^{(j)} ds \right]$$

$$\text{As a check, } tP_x^{(r)} = \prod_{j=1}^m tP_x^{(j)}$$

$$\text{ex)} \quad m=3, \quad t g_x^{(1)} = 0.5 [1 - e^{-0.04t}] \quad (27.5)$$

$$t g_x^{(2)} = 0.4 [1 - e^{-0.04t}]$$

$$t g_x^{(3)} = 0.1 [1 - e^{-0.04t}]$$

Find  $t P_x^{(j)}$  for  $j=1, 2, 3$ .

$$\text{soln) } t P_x^{(r)} = 1 - \sum_{j=1}^m t g_x^{(j)} = 1 - [1 - e^{-0.04t}] = e^{-0.04t}$$

$$\mu_{x+t}^{(1)} = \frac{\frac{d}{dt} t g_x^{(1)}}{t P_x^{(r)}} = \frac{.5(.04) e^{-0.04t}}{e^{-0.04t}} = .02$$

$$\mu_{x+t}^{(2)} = \frac{\frac{d}{dt} t g_x^{(2)}}{t P_x^{(r)}} = .4(.04) = .016 \quad \left. \right\} \text{EXP}$$

$$\mu_{x+t}^{(3)} = \frac{\frac{d}{dt} t g_x^{(3)}}{t P_x^{(r)}} = .1 (.04) = .004$$

$$\text{so } t P_x^{(1)} = e^{-0.02t}$$

$$t P_x^{(2)} = e^{-0.016t}$$

$$t P_x^{(3)} = e^{-0.004t}$$

$\left. \right\} \text{EXP survival functions}$

39] To go from single to multiple decrement quantities

$$\text{i) calculate } \mu_{x+t}^{(j)} = \frac{-d}{dt} \log(t P_x^{(j)}) =$$

$$= \frac{-\frac{d}{dt} t P_x^{(1)}}{t P_x^{(1)}}$$

$$= \frac{\frac{d}{dt} t q_x^{(1)}}{t P_x^{(1)}}$$

M402 18

(28)

ii) Get  $t P_x^{(T)} = \prod_{j=1}^m t P_x^{(j)} = \exp \left[ - \int_0^t \mu_{x+s}^{(T)} ds \right]$

where  $\mu_{x+t}^{(T)} = \sum_{j=1}^m \mu_{x+j}^{(j)}$ .

iii)  $t q_x^{(j)} = \int_0^t s P_x^{(T)} \mu_{x+s}^{(j)} ds$

ex)  $m=2$   $t P_x^{(1)} = \frac{50-t}{50}$   $t < 50$  DeMoivre (50)

$$t P_x^{(2)} = \left( \frac{60-t}{60} \right)^2 \quad t < 60 \quad GD(2, 60)$$

Find  $t q_x^{(1)}$  for  $t < 50$

Sol(n)  $\mu_{x+t}^{(1)} = \frac{1}{50-t}$ ,  $\mu_{x+t}^{(2)} = \frac{2}{60-t}$

(See EI review 63)

$$t P_x^{(T)} = \prod_{j=1}^m t P_x^{(j)} = \frac{50-t}{50} \left( \frac{60-t}{60} \right)^2$$

$$t q_x^{(1)} = \int_0^t s P_x^{(T)} \mu_{x+s}^{(1)} ds =$$

$$\int_0^t \frac{50-s}{50} \left( \frac{60-s}{60} \right)^2 \frac{1}{50-s} ds = \int_0^t \frac{1}{(60-s)(50)} (60-s)^2 ds$$

$$= \frac{-1}{(60)^2(50)} \cdot \frac{(60-t)^3}{3} \Big|_0^t = \frac{-1}{(60)^2(50)} \left[ \frac{(60-t)^3 - 60^3}{3} \right]$$

$$= \frac{(60)^3 - (60-t)^3}{3(60)^2(50)}$$

• 10.7

(29)

1) often  $(X)$  and  $(Y)$  are dependent  
and are exposed to a common hazard,  
called a common shock.

ex) married people who travel together

2) Let  $\mu_{x+t}^*$  be the force of  
failure at time  $t$  that are specific to  
 $(X)$  but not  $(Y)$ . Similarly, let  
 $\mu_{y+t}^*$  be the force of failure at time  
 $t$  that are specific to  $(Y)$  but not  $(X)$ .

Let the common hazard = common shock

for both  $(X)$  and  $(Y)$  be constant

$$\mu_t^c \equiv \lambda, t \geq 0,$$

3) \* P320-1 The common shock model

assumes  $\mu_{x+t} = \mu_{x+t}^* + \lambda$

$\mu_{y+t} = \mu_{y+t}^* + \lambda$ , and

the total force of failure

M402

48

(30)

for the joint status  $(xy)$  is

$$\mu_{x+r, y+r}^* = \mu_{x+r}^* + \mu_{y+r}^* + \lambda \\ (= \mu_{x+r} + \mu_{y+r} - \lambda).$$

4) The 3 forces are additive because they represent hazard factors that are disjoint from each other. There is only one  $\lambda$  since failure of either  $(x)$  or  $(y)$  due to the common shock constitutes failure of the joint life status.

5)  $\overset{P321}{\times P_x^*} = \exp \left[ - \int_0^t (\mu_{x+r}^* + \lambda) dr \right] = e^{-\lambda t} e^{- \int_0^t \mu_{x+r}^* dr}$

$$= \times P_x^* e^{-\lambda t}$$

$\times P_y^* = \exp \left[ - \int_0^t (\mu_{y+r}^* + \lambda) dr \right] = e^{-\lambda t} e^{- \int_0^t \mu_{y+r}^* dr} =$

$$\times P_y^* e^{-\lambda t}$$

$\underbrace{\times P_{xy}}_{\text{joint life survival function}} = \exp \left[ - \int_0^t (\mu_{x+r}^* + \mu_{y+r}^* + \lambda) dr \right] =$

$$e^{- \int_0^t \mu_{x+r}^* dr} e^{- \int_0^t \mu_{y+r}^* dr} e^{-\lambda t} = \times P_x^* \times P_y^* e^{-\lambda t} \frac{1}{(e^{-\lambda t} e^{-\lambda t})}$$

$= \times P_x \times P_y e^{-\lambda t}$ .

6) \* P321 An equivalent way to develop the common shock model is to let  $T_x^*$  and  $T_y^*$  denote the future lifetime random variables for (x) and (y) without regard for the common shock hazard functions. Let  $\bar{W}$  denote the future lifetime of either (x) or (y) with regard to the common shock hazard factors only.

Assume  $T_x^* \perp\!\!\!\perp T_y^* \perp\!\!\!\perp \bar{W}$  with  $\bar{W} \sim \text{Exp}(\lambda)$ .

$$T_x = \min(T_x^*, \bar{W}), \quad T_y = \min(T_y^*, \bar{W})$$

and  $T_x \neq T_y$ .

$$\begin{aligned} \text{Since } T_x \perp\!\!\!\perp \bar{W}, \quad \# p_x &= S_{T_x}(t) = S_{T_x^*}(t) S_{\bar{W}}(t) \\ &= \# p_x^* e^{-\lambda t}. \end{aligned}$$

see top of 5)

$$\begin{aligned} \text{Since } T_y \perp\!\!\!\perp \bar{W}, \quad \# p_y &= S_{T_y}(t) = S_{T_y^*}(t) S_{\bar{W}}(t) = \\ &= \# p_y^* e^{-\lambda t}. \end{aligned}$$

M402 49 (3)

Note that  $x P_x = P(\min(T_x^*, W) > x) = P(T_x^* > x \text{ and } W > x)$

$$= P(T_x^* > x) P(W > x) \quad \text{since } T_x^* \perp\!\!\!\perp W.$$

$$x P_{xy} = P(T_x^* > x, T_y^* > x, \text{and } W > x) =$$

$$x P_x^* x P_y^* e^{-\lambda t} = x P_x x P_y e^{2\lambda t}$$

7)  $P(T_x = T_y) = \int_0^\infty x P_{xy} dt$

8) If  $\lambda = 0$ ,  $T_x^* = T_x \perp\!\!\!\perp T_y = T_y^*$ .

$$T_{xy} = \min(T_x, T_y), \quad T_{\bar{x}\bar{y}} = \max(T_x, T_y).$$

9) know Earlier chapter 10 formulas still hold

for the common shock model, but  $T_x$  is not independent of  $T_y$ . For the last survivor status ( $\bar{x}\bar{y}$ ),

i)  $T_{xy} + T_{\bar{x}\bar{y}} = T_x + T_y \text{ so}$

$T_{\bar{x}\bar{y}} = T_x + T_y - T_{xy}$ . If  $g(a,b) = g(b,a)$ , then  $g(T_{xy}, T_{\bar{x}\bar{y}}) = g(T_x, T_y)$ .

ii) survival function  $S_{T_{\bar{x}\bar{y}}}^*(t) = x P_{\bar{x}\bar{y}} = P(T_{\bar{x}\bar{y}} > t) =$

$$P_x + P_y - P_{xy}$$

(31.5)

iii) cdf  $F_{T_{xy}}(t) = P_{xy} = 1 - S_{T_{xy}}(t) = 1 - P_{\bar{xy}}$ .

iv)  $\bar{e}_{\bar{xy}} = E(T_{\bar{xy}}) = \bar{e}_x + \bar{e}_y - \bar{e}_{xy}$ .

v)  $\bar{A}_{\bar{xy}} = \bar{A}_x + \bar{A}_y - \bar{A}_{xy}$

vi)  $\bar{a}_{\bar{xy}} = \bar{a}_x + \bar{a}_y - \bar{a}_{xy}$

<sup>p322</sup> (Q) Know Suppose  $T_x^* \sim \text{EXP}(\mu_x^*)$ ,  $T_y^* \sim \text{EXP}(\mu_y^*)$

$(\tau \sim \text{EXP}(\lambda))$  in the common shock model.

Then  $T_x \sim \text{EXP}(\underbrace{\mu_x^* + \lambda}_{\mu_x})$ ,  $T_y \sim \text{EXP}(\underbrace{\mu_y^* + \lambda}_{\mu_y})$

$T_{xy} \sim \text{EXP}(\underbrace{\mu_x^* + \mu_y^* + \lambda}_{\mu_x + \mu_y - \lambda})$  since  $\mu_{xy}(t) = \mu_x^* + \mu_y^* + \lambda$ .

$$\bar{A}_x = \frac{\mu_x^* + \lambda}{\mu_x^* + \lambda + \delta} \quad \bar{A}_y = \frac{\mu_y^* + \lambda}{\mu_y^* + \lambda + \delta}, \quad \bar{A}_{xy} = \frac{\mu_x^* + \mu_y^* + \lambda}{\mu_x^* + \mu_y^* + \lambda + \delta}$$

$$\bar{a}_x = \frac{1}{\mu_x^* + \lambda + \delta}, \quad \bar{a}_y = \frac{1}{\mu_y^* + \lambda + \delta}, \quad \bar{a}_{xy} = \frac{1}{\mu_x^* + \mu_y^* + \lambda + \delta}.$$

$$\text{and } P(T_x = T_y) = \frac{\lambda}{\mu_x^* + \mu_y^* + \lambda}.$$

H3\*) In problems, see whether

$\mu_x$  ( $= \mu_x^* + \lambda$ ),  $\mu_y$  ( $= \mu_y^* + \lambda$ ) and  $\lambda$  are given or are  $\mu_x^*$ ,  $\mu_y^*$  and  $\lambda$  given.

For HW5, #4  $\mu_x = \mu_y = 0.06$  and  $\lambda = 0.02$  are given. So  $\mu_x^* = \mu_y^* = 0.04$ .

i) If you are told the force of mortality for  $(x) = v$ , the force of mortality for  $(y) = v$ , and common shock is incorporated into the forces of mortality  $v$  and  $v$ , then  $v = \mu_{x+t}$  and  $v = \mu_{y+t}$ .

ii) If you are told the non common forces of mortality for  $(x)$  and  $(y)$  are  $v$  and  $v$ , then  $v = \mu_{x+t}^*$  and  $v = \mu_{y+t}^*$ .

12) In 7]  $P(\tau_x = \tau_y) = \int_0^\infty t \rho_{xy} \lambda dt$  32.5

$$= \lambda \int_0^\infty e^{-t(\mu_x^* + \mu_y^* + \lambda)} dt = \frac{\lambda}{\mu_x^* + \mu_y^* + \lambda}.$$

end 610.7 material

# Ch 18 Estimating Survival Models (33)

Overlap with Math 473, 401, 403-404.

1)  $S(t) = P(T > t).$

2) <sup>P667</sup> Let  $T_1, \dots, T_n$  be independent and identically distributed (i.i.d.).

Let  $t_1 \leq t_2 \leq t_3 \leq \dots \leq t_n$  be the observed ordered  $T_i$ .

The empirical estimator

$$* \quad \hat{S}_E(t) = \frac{1}{n} \sum_{i=1}^n I(T_i > t) = \frac{\# T_i > t}{n} = \frac{N_t}{n}$$

$$\text{where } I(T_i > t) = \begin{cases} 1 & T_i > t \\ 0 & T_i \leq t, \end{cases}$$

3) <sup>P667</sup> Let  $N_t = \sum_{i=1}^n I(T_i > t)$ . So  $N_t$  is the observed value of  $N_t$ .

$$N_t \sim \text{bin}(n, S(t)).$$

$$E(\hat{S}_E(t)) = E\left(\frac{N_t}{n}\right) = n \frac{S(t)}{n} = S(t),$$

$$V(\hat{S}_E(t)) = \frac{1}{n^2} V(N_t) = \frac{n}{n^2} S(t)(1-S(t)) = \frac{S(t)(1-S(t))}{n}.$$

$$\hat{V}(\hat{S}_E(t)) = \frac{\hat{S}_E(t)(1-\hat{S}_E(t))}{n}$$

33.5

$$= \frac{\frac{nxt}{n}(1-\frac{nxt}{n})}{n} = \frac{nxt(n-nxt)}{n^3}$$

$$\frac{1}{\frac{nxt}{n}}$$

$$= \left(\frac{nxt}{n}\right)^2 \left(\frac{n-nxt}{nnxt}\right) = \left[\hat{S}_E(t)\right]^2 \left(\frac{1}{nxt} - \frac{1}{n}\right)$$

$$* \boxed{SE(\hat{S}_E(t))} = \hat{S}_E(t) \sqrt{\frac{1}{nxt} - \frac{1}{n}} = \boxed{\sqrt{\hat{V}(\hat{S}_E(t))}}$$

4) \* A linear or Wald 95%

confidence interval CI for  $\theta$

is  $\hat{\theta} \pm 1.96 SE(\hat{\theta})$ .

ex)  $\hat{\mu} = 762.76$ ,  $SE(\hat{\mu}) = 217.32$

A 95% CI for  $\mu$  is

$$\hat{\mu} \pm 1.96 SE(\hat{\mu}) = 762.76 \pm 1.96(217.32)$$

$$= 762.76 \pm 425.9472 = \boxed{[336.8128, 1188.7072]}$$

5) Know for exam 2)

Suppose 1000 white 71 year old

females buy a 1st year \$100000

life insurance policy. Actuaries

$$\text{use } 1 - \frac{s(t+x)}{s(x)} = 1 - p(y > t+x | y > x)$$

to estimate how many claims will  
be filed. Hence actuaries want ( $t=1$ )

$\frac{s(72)}{s(71)}$ , If  $\hat{s}(72) = 0.85$  and

$\hat{s}(71) = 0.87$ , about how many  
of the 1000 claims will be filed?

$$\text{Soln} \quad \left(1 - \frac{.85}{.87}\right) 1000 = 0.02299 (1000)$$

$$= \boxed{22.99} \quad \text{with } x=71 \text{ and } t=1.$$

Note  $t =$  term of the term insurance  
while  $x =$  age of insured at purchase.

34.5

6) know for exam 2

The following 20 survival times are listed from smallest to largest.

15, 22, 38, 49, 62, 71, 91, 102, 131,  
145, 177, 198, 247, 279, 319, 359, 468,  
526, 703, 790

a) Find  $\hat{S}_E(247)$ .

b) Find a 95% CI for  $\hat{S}(247)$

$$\text{sol(n)} \quad a) \hat{S}_E(247) = \frac{nt}{n} = \frac{\# > 247}{n}$$

$$= \frac{7}{20} = \boxed{0.35}$$

$$b) SE(\hat{S}_E(t)) = \hat{S}_E(t) \sqrt{\frac{1}{nt} - \frac{1}{n}}$$

$$= 0.35 \sqrt{\frac{1}{7} - \frac{1}{20}} = 0.1067$$

$$CI = \hat{S}_E(t) \pm 1.96 SE(\hat{S}_E(t))$$

$$= 0.35 \pm 1.96(0.1067) = 0.35 \pm .2090$$

$$\Rightarrow [0.1410, .5590].$$

Note: if the CI is  $[L, U]$

$[\max(0, L), \min(U, 1)]$  is a better CI  
since  $0 \leq \underbrace{s(t)}_{\text{prob}} \leq 1$ .

7) Suppose we observe  $n$  lives

from exact age  $x$  to  $x_K$

and we have deaths for the

K intervals:

interval		not given	interval length	deaths
1	$[x, x_1)$		$L_1$	$d_1$
2	$[x_1, x_2)$		$L_2$	$d_2$
⋮	⋮	⋮	⋮	⋮
K	$[x_{K-1}, x_K)$		$L_K$	$d_K$

$$\hat{s}_x(0) = 1$$

$$\hat{s}_x(L_1) = \frac{n-d_1}{n} = \hat{s}_x(0) - \frac{d_1}{n}$$

$$\hat{S}_X(L+d_2) = \frac{n - (d_1 + d_2)}{n} = \hat{S}_X(L_1) - \frac{d_2}{n} \quad (39.5)$$

$$\hat{S}_X\left(\sum_{j=1}^i L_j\right) = \frac{n - (d_1 + d_2 + \dots + d_i)}{n} = \hat{S}_X\left(\sum_{j=1}^{i-1} L_j\right) - \frac{d_i}{n}$$

$$\hat{S}_X\left(\sum_{j=1}^K L_j\right) = \frac{n - \sum_{j=1}^K d_j}{n} = \hat{S}_X\left(\sum_{j=1}^{K-1} L_j\right) - \frac{d_K}{n}$$

If  $L_i \equiv L$ , can find  $\hat{S}_X(0) = 1$ ,  $\hat{S}_X(L)$

$$\hat{S}_X(2L), \dots, \hat{S}_X(KL).$$

Linear interpolation is used to find

$\hat{S}_X(t)$  between endpoints, giving

the ogive empirical survival function.

Let  $t_L \leq t < t_U$ , then

$$\hat{S}_X(t) = \frac{(t_U - t) \hat{S}_X(t_L) + (t - t_L) \hat{S}(t_U)}{t_U - t_L},$$

ex) know for exam 2

100000 lives from exact age 50  
are observed for 30 years

age last birthday	deaths	$\hat{S}_{50}(t) = 1$	36)
50-59	1700	$\hat{S}_{50}(10) = 1 - \frac{1700}{100000} = 0.9830$	
60-69	4650	$\hat{S}_{50}(20) = 0.983 - \frac{4650}{100000} = .9365$	
70-74	5520	$\hat{S}_{50}(25) = .9365 - \frac{5520}{100000} = .8813$	
75-79	9680	$\hat{S}_{50}(30) = .8813 - \frac{9680}{100000} = .7845$	

given

Note  $\hat{S}_{50}(25) = \frac{100000 - (1700 + 4650 + 5520)}{100000}$

Go to 36 1/4

8) P662-3 The survival time of an individual is censored if the event of interest (death) has not been observed.

ex) 10000 65 year olds buy 10 year life insurance. During the 10 years, perhaps 20% die and 15% quit paying premiums. Then 80% of lifetimes are censored.

q) P663 36.9  
 Let  $y_i = \text{time until event (death)} \text{ for}$   
 the  $i$ th person. Let  $z_i = \text{time } i\text{th}$   
 person leaves "study" for any reason  
 other than event of interest = time  
 until person is censored. Then the  
 (right censored) survival time

$$T_i = \min(y_i, z_i).$$

Let  $\delta_i = \begin{cases} 0 & \text{if } T_i \text{ is censored } (T_i = z_i) \\ 1 & \text{if } T_i \text{ is not censored } (T_i = y_i) \end{cases}$

10) convention order the survival  
 times, if there are ties, put  
 censored cases after uncensored  
 cases. Denote a censored case  
 by  $T_i^* = T_i +$ .

$$T_1, T_2, T_3^*, \dots, T_{n-1}, T_n$$

11) Suppose  $T \geq 0$ ,  $t = \text{time}$

see ex above 8)

ex continued Find  $\hat{S}(t)$  at  $t =$  36  $\frac{1}{4}$

a) 2      b) 17

$$\text{Soln } \hat{S}_{50}(t) = \frac{(t_0 - t) \hat{S}_{50}(t_L) + (t - t_L) \hat{S}_{50}(t_0)}{t_0 - t_L}$$

a)  $\hat{S}_{50}(0) = 1$        $\hat{S}_{50}(10) = .9830$

$$\hat{S}(2) = \frac{(10-2) 1 + (2-0) .9830}{10-0}$$

$$= \frac{8 + 2(.983)}{10} = \boxed{.9966}$$

b)  $\hat{S}_{50}(10) = .9830$        $\hat{S}_{50}(20) = .9365$

$$\hat{S}(17) = \frac{(20-17) .9830 + (17-10) (.9365)}{20-10}$$

$$= \frac{3(.9830) + 7(.9365)}{10} = \boxed{.9505}$$

$$[0, \infty) = [t_0, t_1] \cup [t_1, t_2] \cup \dots \cup [t_{m-1}, t_m] \cup \{t_m\} \quad (37)$$

$\stackrel{\text{def}}{=} P(Y > t) \quad \text{as } t \rightarrow \infty$

want to estimate  $S(t)$ , but

there may be censoring,

Let  $\gamma_i = \text{time until event}$  and  $T_i = \min(\gamma_i, z_i)$ .

Let  $P_j = P(\text{surviving through } T_j | \text{alive at start of } T_j)$

$$= P(Y > T_j | Y > T_{j-1}) = \frac{P(Y > T_j, Y > T_{j-1})}{S(Y > T_{j-1})}$$

$$= \frac{S(T_j)}{S(T_{j-1})}, \quad \text{so} \quad P_1 = \frac{S(t_1)}{S(t_0)} = S(t_1)$$

Since  $S(0) = S(t_0) = 1, \dots,$

$$S(t_j) = S(t_1) \underbrace{\frac{S(t_2)}{S(t_1)} \frac{S(t_3)}{S(t_2)} \dots \frac{S(t_{j-1})}{S(t_{j-2})} \frac{S(t_j)}{S(t_{j-1})}}$$

telescoping product

$$= P_1 P_2 \cdots P_j = \prod_{i=1}^j P_i.$$

(37.5)

Let  $\hat{P}_i = 1 - \frac{\# \text{ dying in } I_i}{\# \text{ with potential to die in } I_i}$ .

(2) Let  $(t_{(j)}, \delta_j)$  correspond to ordered

$$T_1, T_2, T_3^*, \dots, T_{n-1}, T_n$$

$t_{(j)}$	$t_{(1)}$	$t_{(2)}$	$t_{(3)}$	$\dots$	$t_{(n-1)}$	$t_{(n)}$
$\delta_j$	1	1	0	$\dots$	0	1

(3) Let  $t_1, \dots, t_m$  be the ordered times where an event (death) occurred.

Let  $r_i = \# \text{ at risk at } t_i = \# \text{ alive and uncensored just before } t_i$ :  $r_i = \sum_{j=1}^n I(t_{(j)} \geq t_i)$ .

Let  $d_i = \# \text{ of events (deaths) at } t_i$ .

Then  $d_i \geq 1$  and it  $d_i = 1$  for  $i = 1, \dots, m$  there are no ties. If  $d_i > 1$  for some  $i$ , then there are ties.