

Lecture 7

2/19/15

Final size formula

$$X \rightarrow Y \rightarrow Z$$

Susceptible infectious removed

$$\frac{dx}{dt} = -\beta^* x y$$

$$\frac{dy}{dt} = \beta^* x y - \rho y$$

$$\frac{dz}{dt} = \rho y$$

$$1 - e^{-R_0 z}$$

$$= z$$

poisson
distributed
hits
→ prob of no hits

number
infected @
end of
epidemic

w/ at least
one hit

$$y = -\frac{1}{\beta^* x} \frac{dx}{dt}$$

$$dz = \rho y dt$$

$$dz = \rho \left(-\frac{1}{x \beta^*} \right) \frac{dx}{dt} dt$$

$$\int dz = \int -\frac{\rho}{x \beta^*} dx$$

$$z(T) - z(0) = -\frac{\rho}{\beta^*} \ln \frac{x(T)}{x(0)}$$

$$z(\infty) = -\frac{\rho}{\beta^*} \ln x(\infty)$$

$$-\frac{\beta^*}{\rho} z(\infty) = \ln x(\infty)$$

$$e^{-\frac{\beta^*}{\rho} z(\infty)} = x(\infty)$$

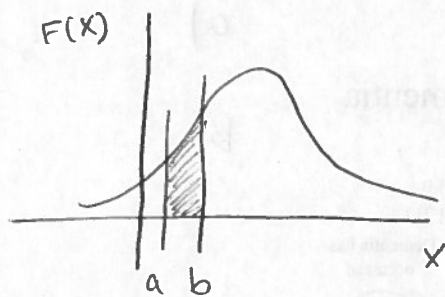
⇒ final size formula

Final Size Formula

Number of ppl who got infected is the number of ppl who could have infected that number

①

Continuous probability



$$P(X \in (a, b)) = \int_a^b f(x) dx$$

$f(x)$ probability density

$$P(X \in [a, a]) = 0 \Rightarrow P(X = a) = 0$$

Probability 0 does not mean impossible

The real numbers are uncountable

Discrete

$$EX = \sum_i x_i P(X = x_i)$$

Continuous

$$EX = \int_{-\infty}^{\infty} x f(x) dx$$

Normal - Standard Normal

$$\frac{e^{-x^2/2}}{\sqrt{2\pi}}$$

Waiting Time Variables

T

$$P(T < 0) = 0$$

No negative waiting times

Hazard

$$\lambda = \frac{P(T \in (t, t + \Delta t])}{P(T > t) \Delta t}$$

$$\lambda = \lim_{\Delta t \rightarrow 0} \frac{P(T \in (t, t + \Delta t])}{P(T > t) \Delta t}$$

(2)

$$\int_{-\infty}^t f(t) dt = P(T \leq t) = F(t)$$

Cumulative
Density
Function

$$F(-\infty) = 0 \quad - \text{Go far enough back \& the event hasn't happened}$$

$$F(\infty) = 1 \quad - \text{Go forward enough and the event has happened}$$

Monotone non-decreasing = F cannot decrease as t increases!
 F can be flat

$$\lambda(t) = \lim_{\Delta t \rightarrow 0} \frac{f(t) \Delta t}{(1 - F(t)) \Delta t}$$

$$\lambda(t) = \frac{f(t)}{1 - F(t)} = \frac{f(t)}{S(t)} \quad \frac{\text{density}}{\text{survival fn}}$$

$$\text{where } S(t) = 1 - F(t)$$

Example: Suppose I want a density fn w/ constant hazard

Facts:

$$\frac{dF}{dt} = f$$

$$1 - F(t) = S(t) \Rightarrow -\frac{dF}{dt} = \frac{dS}{dt}$$

$$\lambda = \frac{f(t)}{S(t)}$$

$$S(t) \lambda = f(t)$$

$$S(t) \lambda = -\frac{dS}{dt}$$

$$\lambda = -\frac{1}{S} \frac{dS}{dt}$$

$$-\lambda dt = \frac{dS}{S}$$

$$\begin{aligned} \int_0^{t'} \lambda dt &= \int_0^{t'} \frac{dS}{S} \\ e^{-\lambda t'} &= S(t') \end{aligned}$$

Constant Hazard, Exponential Survival!

(3)

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

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

$$\frac{dF}{dt} = -\frac{d}{dt} e^{-\lambda t} = -(-\lambda) e^{-\lambda t} = \lambda e^{-\lambda t}$$

$$f(t) = \lambda e^{-\lambda t}$$

$$ET = \int_{-\infty}^{\infty} t f(t) dt = \int_0^{\infty} t \lambda e^{-\lambda t} dt$$

$$u = t \quad dv = e^{-\lambda t} dt$$

$$du = dt \quad v = -\frac{1}{\lambda} e^{-\lambda t}$$

$$ET = \lambda \left\{ -\frac{1}{\lambda} e^{-\lambda t} t \Big|_0^{\infty} - \int_0^{\infty} \left(-\frac{1}{\lambda} e^{-\lambda t} \right) dt \right\}$$

$$= \lambda \int_0^{\infty} e^{-\lambda t} dt = -\frac{1}{\lambda} (0 - 1) = \frac{1}{\lambda}$$

Expected waiting time is the reciprocal of the rate for an exponential Survival function. (or constant hazard)

Example:

t^* - Survive to this - what is survival time after that?

$$P(T - t^* > u \mid T > t^*)$$

$$= \frac{P(T - t^* > u \cap T > t^*)}{P(T > t^*)}$$

$$= \frac{P(T - t^* > u)}{P(T > t^*)} = \frac{P(T > u + t^*)}{P(T > t^*)}$$

(4)

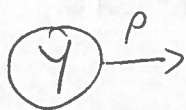
$$= \frac{e^{-\lambda(u+t^*)}}{e^{-\lambda t^*}}$$

$$= e^{-\lambda u}$$

If you wait t^* , the expected amount you have to wait is the same it was at the beginning.

Memorylessness

k-M model - Analyze very start of epidemic to get R_0



Drop a bunch of infectives into Y and watch them recover

$$\frac{dY}{dt} = -pY \quad Y = Y(0)e^{-pt}$$

Mean duration of infectivity: $\frac{1}{p}$

$$\frac{1}{p}$$

avg time infected

kM Model Incidence rate

$$\frac{dx}{dt} = -\beta XY$$

$$= -X(\beta Y)$$

hazard of removal

At beginning

$$X \cong N$$

$$\frac{dI}{dt} = \beta NY$$

Cumulative infections @ start

$$\frac{dI}{dt} = \beta N Y^{(0)}$$

$$\int_0^{t'} dI = \int_0^{t'} \beta N Y^{(u)} dt$$

$$I(t') - I(0) = \beta N \int_0^{t'} e^{-\rho t} Y^{(u)} dt$$

$$I(t) = \beta N Y_0(0) \int_0^{t'} e^{-\rho t'} dt'$$

$$I(\infty) = \frac{\beta N}{\rho} Y(0)$$

Total #
new
infectives $\rightarrow$ per initial
case

$$\frac{\beta N}{\rho} = R_0$$

Example:

Suppose 2 waiting times T_1, T_2

death
disease 1

death
disease 2

$$T = \min(T_1, T_2) = T_1 \wedge T_2 \quad \text{time of death?}$$

T_1, T_2 independent $\ddagger$ exponential

$$T > t^* = T_1 > t^* \cap T_2 > t^*$$

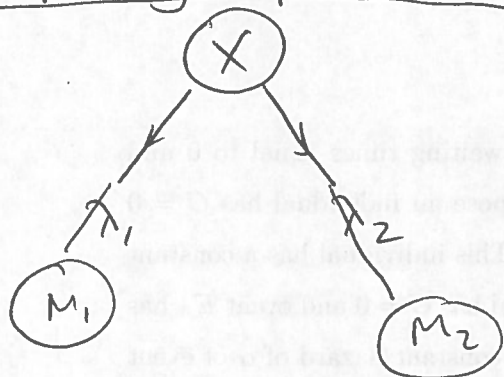
$$P(T > t^*) = P(T_1 > t^*) P(T_2 > t^*) = e^{-\lambda_1 t^*} e^{-\lambda_2 t^*}$$

$$P(T > t^*) = e^{-(\lambda_1 + \lambda_2) t^*}$$

Hazard $\lambda_1 + \lambda_2$

2 indep things $\rightarrow$ just add hazards!

Competing Exponential Risks



Independent processes

* You cannot add hazards unless you assume independence *

$$\left\{ \begin{array}{l} \frac{dX}{dt} = -(\lambda_1 + \lambda_2) X = -\lambda_1 X - \lambda_2 X \\ \frac{dM_1}{dt} = \lambda_1 X \\ \frac{dM_2}{dt} = \lambda_2 X \end{array} \right.$$

$$X(t) = e^{-(\lambda_1 + \lambda_2)t} X(0)$$

Aside! Assume only λ_1 (no λ_2)

Expected deaths = number @ risk $\times$ risk
in Δt

$$\Delta X = -X \lambda_1(t) \Delta t$$

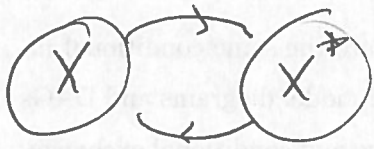
$$\frac{\Delta X}{\Delta t} = -\lambda_1 X$$

Adding the hazards $\Rightarrow$ independence

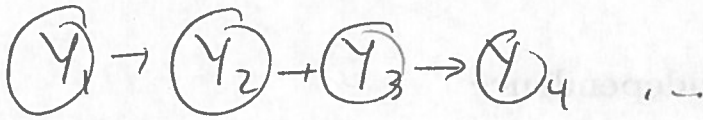
$$M_1(t) = \lambda_1 \int_0^t X(0) e^{-(\lambda_1 + \lambda_2)t} dt$$

$$M_1(\infty) = \frac{\lambda_1}{\lambda_1 + \lambda_2} X(0) \quad \left\{ \begin{array}{l} \text{Fraction in } M_1 \text{ is} \\ \frac{\lambda_1}{\lambda_1 + \lambda_2} \end{array} \right.$$

other models in Epidemiology



Loops



chains