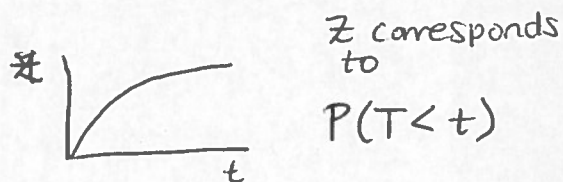
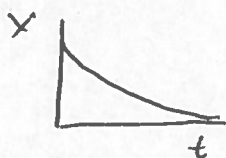
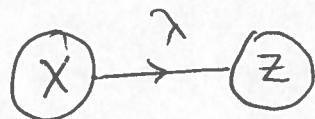


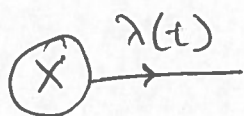
3/4/2016

Lecture 9



Exponential Distribution

- memoryless - addition waiting time does not depend on how long already waited
- constant hazard - hazard = probability per unit time given that the event hasn't already happened



$$\frac{dX}{dt} = -\lambda X$$

$$\frac{dX}{X} = -\lambda(t) dt$$

$$\int_{t=0}^{t=t'} \frac{dX}{X} = - \int_{t=0}^{t=t'} \lambda(t) dt$$

$$\ln X(t') - \ln X(0) = - \int_{t=0}^{t=t'} \lambda(t) dt$$

$$\frac{X(t')}{X(0)} = e^{-\int_0^{t'} \lambda(t) dt}$$

HW Problem -
tiny interval vs.
larger one

CONSTANT HAZARD - EXPONENTIAL WAITING TIME!

HIV

MSM in SF - 1981 HIV prev = 1%, 1983 HIV prev = 50%

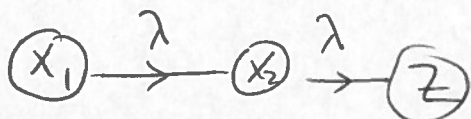
But since HIV → AIDS is not exponential!
(Incidence of AIDS was not highest in 1983)

Distributed Delays



* More realistic waiting time distributions can be approximated with chains of exponentials

$$\bar{T} = \frac{1}{\alpha_1} + \frac{1}{\alpha_2}$$



$$X_1(0)=1 \quad X_2(0)=0 \quad Z(0)=0$$

What is the distribution of ~~the waiting time~~ waiting times? That gives us the proportion in Z over time

$$\frac{dX_1}{dt} = \dot{X}_1 = -\lambda X_1$$

Z is the cumulative distribution fn

$$\frac{dX_2}{dt} = \lambda X_1 - \lambda X_2$$

What is the distribution of the sum of two independent identical exponential waiting times

$$\frac{dZ}{dt} = \lambda X_2$$

	<u>Continuous</u>		<u>Discrete</u>
<u>1</u>	Exponential	$\longleftrightarrow$	Negative Binomial Geometric
<u>Sum of 2</u>	??	$\longleftrightarrow$	Negative Binomial

$$X_1 = e^{-\lambda t}$$

$$\frac{dX_2}{dt} + \lambda X_2 = \lambda e^{-\lambda t}$$

$$\frac{dx_2}{dt} e^{\lambda t} + \lambda x_2 e^{\lambda t} = \lambda$$

$$\frac{d}{dt} (x_2 e^{\lambda t}) = \lambda \quad \int_0^{t'} d(x_2 e^{\lambda t}) = \int_0^{t'} \lambda dt$$

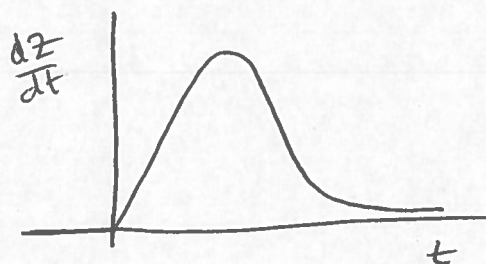
~~xxxxxx~~
$$x_2(t') e^{\lambda t'} - x_2(0) e^0 = \lambda t'$$

$$x_2(t) = \lambda t e^{-\lambda t}$$

$$\frac{dz}{dt} = \lambda x_2 = \lambda^2 t e^{-\lambda t}$$

z is the CDF

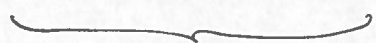
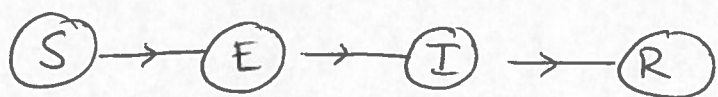
so $\frac{dz}{dt}$ gives the density fn of the arrival time



Gamma distribution with the
Shape parameter 2

Add n indep exponential, I get a ~~&~~ gamma w/
Shape parameter n .

χ^2 distribution is also a gamma distribution



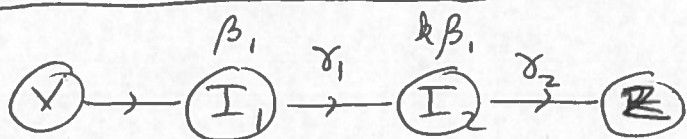
more realistic waiting time

$$S \rightarrow E_1 \rightarrow E_2 \rightarrow I \rightarrow R$$

$$S \rightarrow E \rightarrow I_1 \rightarrow I_2 \rightarrow R$$

$$S \rightarrow E_1 \rightarrow E_2 \rightarrow I_1 \rightarrow I_2 \rightarrow R \quad \underline{\text{etc.}}$$

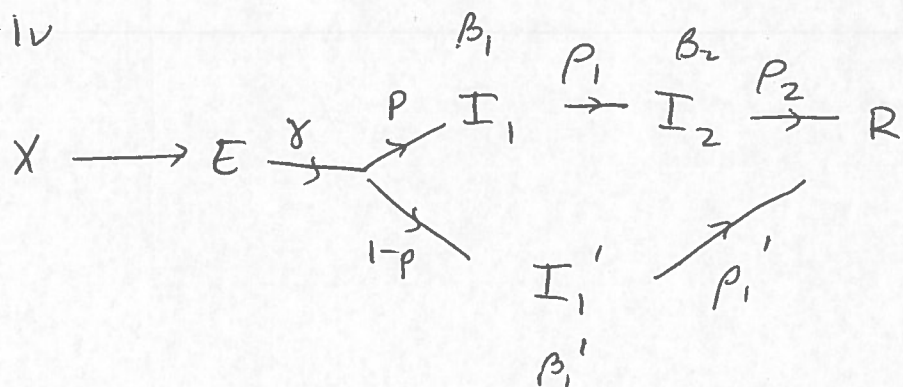
R_0 for chains and forks



What is R_0 ?

$$\frac{\beta_1}{\gamma_1} + \frac{k\beta_2}{\gamma_2}$$

FIV

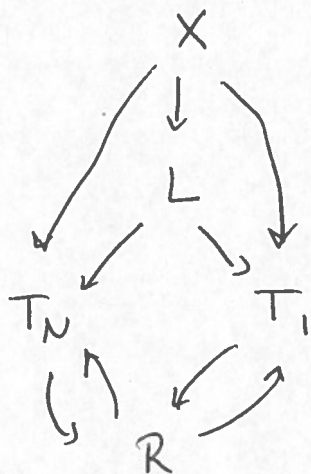


Symptomatic

asymptomatic

$$R_0 = p \left(\frac{\beta_1}{p_1} + \frac{\beta_2}{p_2} \right) + (1-p) \left(\frac{1}{p_1'} \beta_1' \right)$$

TB Model, Blower 1995



Most infected 3, remain latent

Small frxn become cases
- some infections 3, non-infections

Spontaneous remission

Immigration - Death



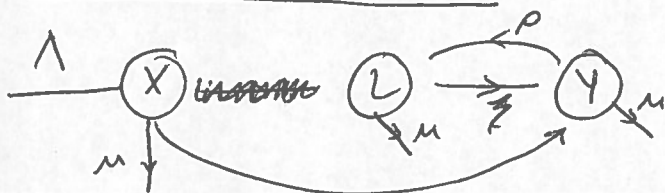
$$\dot{X} = \lambda - \mu X$$

$$\bar{X} = \lambda / \mu = \lambda \frac{1}{\mu}$$

Prevalence = incidence · duration

Modified SI process

HSV-2 Model - Loops



$$\dot{X} = \lambda - \mu X - \beta X \frac{Y}{N}$$

$$\dot{Y} = \beta X \frac{Y}{N} - \rho Y - \mu Y + \zeta L$$

$$\dot{L} = -\mu L + \rho Y - \zeta L$$

$$R_0 = \underbrace{\frac{1}{\rho + \mu} \beta}_{\text{cases per episode}} + k \frac{\beta}{\rho + \mu} + k^2 \frac{\beta}{\rho + \mu} + \dots$$

cases per episode

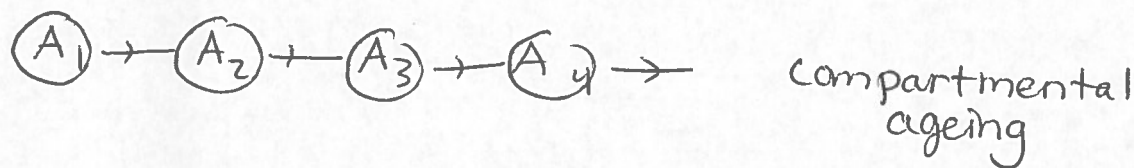
where k is the probability of each episode

$$\Rightarrow R_0 = \frac{\beta}{\rho + \mu} \cdot \frac{1}{1-k}$$

using sum of a geometric

$$k = \underbrace{\frac{\rho}{\rho + \mu}}_{\substack{\text{prob} \\ \text{you recover} \\ \text{(and not die)}}} \cdot \underbrace{\frac{\gamma}{\gamma + \mu}}_{\substack{\text{prob} \\ \text{you relapse} \\ \text{(and not die)}}$$

Modeling Ageing



Alternate approach:

$X(a, t)$ density of people in age at time t

$X(a, t) \Delta a$ # of ppl w/ age between a and $a + \Delta a$

Follow a cohort:

$$X(a + \Delta a, t + \Delta a) = X(a, t)(1 - \mu(a)\Delta a)$$

$$X(a + \Delta a, t + \Delta a) - X(a, t) = -X(a, t)\mu(a)\Delta a$$

$$X(a + \Delta a, t + \Delta a) - X(a, t + \Delta a) + X(a, t + \Delta a) - X(a, t) = -X(a, t)\mu(a)\Delta a$$

$$\frac{X(a+\Delta a, t+\Delta a) - X(a, t+\Delta a)}{\Delta a} + \frac{X(a, t+\Delta a) - X(a, t)}{\Delta t} = -X\mu$$

Take limit as $\Delta a \rightarrow 0$, $\Delta t \rightarrow 0$

$$\frac{\partial X}{\partial a} + \frac{\partial X}{\partial t} = -\mu(a)X \quad \text{PDE Age Structure Model}$$

McKendrick-von Foerster Equation