

How do we calculate statistics of  $Z$ ?

Clever recursive argument.

Relate epidemic statistics for a given population to those of a population with a smaller # of initial susceptibles.

Let  $Z^{n'}$ ,  $\Phi^{n'}$  be the total epidemic size and infection pressure with  $n'$  initial susceptibles

but always normalize infection rate by  $\frac{1}{n+m}$   
↑ not  $n'$

$$\Phi^{n'} = \frac{\lambda}{n+m} \int_0^t I^{n'}(s) ds$$

$$S^{n'}(0) = n'$$

$$I^{n'}(0) = m$$

04/08/04 : HW 3 due 04/14 Wed at 5 PM

HW 4 (last) : look for over wknd

- due ~~at~~ Wednesday April 28

(maybe a little later)

≥ 360

A

No PK OH

280-359

B

on 04/13

200-279

C

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"Take Your Students to Lunch"

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Tedious calculation for PNF for  $Z$ :

$$P_k^{n'} = \text{Prob}(Z^{n'} = k)$$

$$\tilde{P}_k^{n'} = \text{Prob}(\text{The individuals infected, with } n' \text{ initial susceptibles are precisely those labeled } \{1, 2, \dots, k\})$$

Recursion: Choose  $k \leq l \leq n'$

$$\tilde{P}_k^{n'} = \tilde{P}_k^l \mathbb{E}(\exp(-\Phi^l(n-l)) | Z^l = k)$$

Why? LHS = Prob (Individuals  $1 \dots k$  out of  $n'$  get infected but inds.  $k+1 \dots n'$  don't)

$\tilde{P}_k^l$   
= Prob (Individuals  $1 \dots k$  out of  $l$  people get infected)

Prob (Individuals  $l+1, \dots, n'$  not infected | individuals  $1 \dots k$  out of  $1 \dots l$  are the ones infected)

Each individual  $l+1, \dots, n'$  has probability  $e^{-\Phi^l}$  not to get infected by the  $k$  out of  $l$ ,

Let us study the conditional expectation

$$\Phi^l = \frac{\lambda}{n+m} \sum_{j=-(m-1)}^{Z^l} T_j \quad T_j : \text{random sum}$$

$$Z^l = \min \left\{ i : Q_{(i+1)} > \frac{\lambda}{n+m} \sum_{j=-(m-1)}^i T_j \right\}$$

Can't evaluate the cond exp by just  
subbing in  $Z^l = k$  because  $Z^l$   
depends on the  $T_j$ , and knowing  
that  $Z^l$  attains some value biases  
the PDFs for  $T_j$ .

But:  $Z^l$  only depends on the

$T_j$  with  $j \leq Z^l$

So it's like a Markov time

w.r.t. the rvs  $\{T_j\}_{j=-(m-1)}^l$

Markov times are good when systems have "strong Markov property" or when one can construct martingales

Martingales (Karlin + Taylor Sec. 6.1-6.5)

Remember a filtration is an increasing family of  $\sigma$ -algebras  $\mathcal{A}_t$

$\mathcal{A}_t$  is like the collection of events which are determined up to and including time  $t$ .

Martingale w.r.t. filtration  $\mathcal{A}_t$   
 $Y(t)$

is a stoch. process such that

$$\underline{E(Y(t) | \mathcal{A}_s) = Y(s) \text{ for } s \leq t}$$

and

$$|E(Y(t))| < \infty$$

A more concrete way to write martingale property is that if  $A$  is an event  $A \in \mathcal{A}_s$  (only involving info about  $Y(t)$  up to and including  $t=s$ ) then

$$E(Y(s') | A \text{ and } Y(s)=i) = i$$

for  $s \leq s'$

Examples of martingales:

Let  $X_1, \dots, X_n$  be ~~id~~ independent  
 with  $\langle X_j \rangle < \infty$

Then 
$$Y_n = \sum_{j=1}^n (X_j - \langle X_j \rangle)$$

is a martingale w.r.t. the filtration generated by  $\{X_n\}$

What about ~~variance~~ <sup>square</sup> of the sum?

$$\text{Let } V_n = \left( \sum_{j=1}^n X_j \right)^2 - \sum_{j=1}^n \sigma_j^2$$

with  $\{X_j\}$  being ~~identical~~ independent

$$\langle X_j^2 \rangle = \sigma_j^2, \quad \langle X_j \rangle = 0$$

Then can check

$$\mathbb{E}(V_{n+1} | \mathcal{A}_n) = V_n$$

so  $V_n$  is a martingale w.r.t.

the filtration  $\mathcal{A}_n$  generated by

$\{X_n\}$ ,

$$\mathbb{E}(V_{n+1} | \mathcal{A}_n) = \mathbb{E}\left( \left( \sum_{j=1}^{n+1} X_j \right)^2 - \sum_{j=1}^{n+1} \sigma_j^2 \mid \mathcal{A}_n \right)$$

$$= \mathbb{E}\left[ \left( \sum_{j=1}^n X_j \right)^2 - \sum_{j=1}^n \sigma_j^2 + X_{n+1}^2 + 2X_{n+1} \sum_{j=1}^n X_j - \sigma_{n+1}^2 \mid \mathcal{A}_n \right]$$

$$= \mathbb{E}\left[ V_n + X_{n+1}^2 - \sigma_{n+1}^2 + 2X_{n+1} \sum_{j=1}^n X_j \mid \mathcal{A}_n \right]$$

$$= V_n + \cancel{\sigma_{n+1}^2} - \cancel{\sigma_{n+1}^2} + 2 \sum_{j=1}^n X_j \mathbb{E}(X_{n+1} \mid \mathcal{A}_n)$$

Branching process:  $X_{n+1} = \sum_{j=1}^{X_n} Z_{n,j}$

where  $Z_{n,j}$  are iid r.v.s,  $\langle Z_{n,j} \rangle = m$

Multiplicative normalization:

$$Y_n = \frac{X_n}{m^n}$$

$$\begin{aligned} E[Y_{n+1} | \mathcal{A}_n] &= E\left[\frac{\sum_{j=1}^{X_n} Z_{n,j}}{m^{n+1}} \mid \mathcal{A}_n\right] \\ &= \sum_{j=1}^{X_n} \frac{E[Z_{n,j} | \mathcal{A}_n]}{m^{n+1}} \\ &= \sum_{j=1}^{X_n} \frac{m}{m^{n+1}} \\ &= \frac{X_n}{m^n} = Y_n \end{aligned}$$

So  $Y_n$  is a martingale w.r.t.

filtration  $\mathcal{A}_n$  generated by the  $\{X_n\}$ .

What's the use of martingales?

Mathematical: control fluctuations and  
prove smoothness

control extreme values:

For example

$$\text{Prob} \left\{ \max_{0 \leq k \leq n} X_k > a \right\} \leq \frac{E(|X_n|)}{a}$$

for martingales

At long times martingales  
converge to random variables

Computational purposes: relation to Markov times

Optional Stopping Theorem:

If  $Y(t)$  is martingale and  $T$  is

Markov time with  $\text{Pr}(T < \infty) = 1$ ,

$E \left( \sup_{t \geq 0} |Y(t)| \right) < \infty$  then

$$E(Y(T)) = E(Y(0))$$

One important application is to  
 Wald using random sums where  
 the number of terms in the sum  
 is a Markov time.

Wald identity: (K&T Sec. 6.4)

Let  $\{X_n\}$  be iid, nonnegative r.v.s,

$$S_0 = 0, \quad S_n = \sum_{j=1}^n X_j \quad \text{and } T \text{ is}$$

a Markov time w.r.t. the filtration  $\mathcal{A}_n$   
 generated by  $\{X_n\}$  and which

satisfies  $T \leq N$  for some  $N < \infty$ ,

Let  $\mathcal{L}_X(\theta) = \langle e^{-\theta X} \rangle$  be moment  
 gen fun.

$$\text{Then } \mathbb{E} \left( \frac{\exp(-\theta S_T)}{\mathcal{L}_X(\theta)^T} \right) = 1$$

$$\text{Proof: } Y_n = \frac{e^{-\theta S_n}}{\mathcal{L}_X(\theta)^n} \quad \text{for } 0 \leq n \leq N, \quad Y_n = Y_N \quad \text{for } n \geq N$$

is a bounded martingale: use Opt Stop Thm.

Applying Wald identity to the recursion formula w/ the cond. exp., we'd get

$$E \left( \frac{\exp(-\Phi^l(n-l))}{\sum_I \left( \frac{\lambda(n-l)}{n+m} \right)^{z_l+m}} \right) = 1$$

$$\sum_{k=0}^{\infty} E \left( \frac{\exp(-\Phi^l(n-l))}{\left( \sum_I \left( \frac{\lambda(n-l)}{n+m} \right)^{z_l+m} \right)} \Big| Z^l = k \right)$$

$$\times \text{Prob}(Z^l = k) = 1$$

$$\sum_{k=0}^{\infty} \frac{E(\exp(-\Phi^l(n-l)) | Z^l = k) \text{Prob}(Z^l = k)}{\sum_I \left( \frac{\lambda(n-l)}{n+m} \right)^{k+m}} = 1 \quad \parallel p_k^l$$

$$\sum_{k=0}^{\infty} \frac{\tilde{p}_k^{n'}}{\tilde{p}_k^l} \frac{p_k^l}{\left( \sum_I \left( \frac{\lambda(n-l)}{n+m} \right)^{k+m} \right)} = 1$$

$$P_k^{n'} = \binom{n'}{k} \tilde{P}_k^{n'}$$

After some factorial reduction

$$\sum_{k=0}^{\ell} \frac{\binom{n'-k}{\ell-k}}{\binom{n'}{\ell}} P_k^{n'} \sum_{\ell} \left( \frac{\lambda(n'-\ell)}{n+m} \right)^{-k-m} = 1$$

for  $0 \leq \ell \leq n'$

Triangular system for  $P_k^{n'}$  ... recursively  
solvable

$$P_0^{n'}$$