

# Criticality in Epidemic Models

Epidemic models often exhibit the so-called threshold phenomena, or quantitatively different behavior above and below criticality. In general, below criticality the major epidemic is impossible or unlikely, whereas when the reproductive number is above one, a major epidemic is possible. We will describe the final outcome of the infection spread for several epidemic models in both subcritical and supercritical cases.

## Reed-Frost Model

$$S \longrightarrow I \longrightarrow R$$

- $(S_t, I_t)$  = state at time  $t = 0, 1, 2, \dots$

$S_t$  = # susceptible at time  $t$

$I_t$  = # infected at time  $t$

- 1 unit time  $\approx$  incubation period  
 $\approx$  recovery time

## Reed-Frost Model

- $p = 1 - q =$  probability of encounter for any  $(S, I)$  pair

$\Rightarrow$   $S$ -person at time  $t$  remains  $S$  at  $t + 1$  w/prob.  $q^{I_t}$

$\Rightarrow S_{t+1} \sim \mathbf{Bin}(S_t, q^{I_t})$   
 $I_{t+1} = S_t - S_{t+1}$

- $(S_0, I_0) = (n, m)$

epidemic continues until

$$\tau = \inf\{t : I_t = 0\}$$

- $S_\tau = \#$  susceptibles that avoid infection  
 $K = n + m - S_\tau$   
 $= I_0 + I_1 + I_2 + \dots =$  total damage

## Reed-Frost Model

Distribution of  $S_\tau$ ? ( $\Rightarrow K$ ?)

- Let  $m = I_0 = o(n)$ , and  
 $q = e^{-\lambda/n} \Rightarrow p = O(n^{-1})$   
 $\Rightarrow \mathbf{E}(\#\text{contacts}) \approx \text{fixed}$

$\Rightarrow$  in the beginning

$$\begin{aligned} I_{t+1} &\approx \mathbf{Bin}\left(S_t, 1 - e^{-\lambda I_t/n}\right) \\ &\approx \mathbf{Bin}(n, \lambda I_t/n) \\ &\approx \mathbf{Poisson}(\lambda I_t), \end{aligned}$$

so that

$$I_{t+1} \approx \text{sum of } I_t \text{ indep. } \mathbf{Poisson}(\lambda) \text{ r.v.'s}$$

## Reed-Frost Model

- Initially,  $I_0, I_1, I_2, \dots$  is  $\approx$  G-W process w/offspring  $\sim \mathbf{Poisson}(\lambda)$
- Such G-W process is
  - $\lambda \leq 1 \Rightarrow$  subcritical
    - $\Rightarrow$  progeny is finite (known Lap. tnsf.)
  - $\lambda > 1 \Rightarrow$  supercritical
    - $\Rightarrow$  w/prob.  $\sigma^m < 1$ 
      - progeny is finite (known Lap. tnsf.)
    - w/prob.  $1 - \sigma^m$ 
      - $I_t$  gets large, G-W approx. breaks

## Reed-Frost Model

When  $\lambda > 1$  w/prob.  $1 - \sigma^m$  both  $S_t$  and  $I_t$  are large, LLN gives

$$\begin{cases} S_{t+1} = S_t q^{I_t} \\ I_{t+1} = S_t(1 - q^{I_t}) \end{cases}$$

$$s_t = S_t/n, i_t = I_t/n \Rightarrow s_0 = 1, i_0 = \frac{m}{n} = \mu$$

$$\begin{cases} s_{t+1} = s_t e^{-\lambda i_t} \\ i_{t+1} = s_t(1 - e^{-\lambda i_t}) \end{cases}$$

and

$$s_\infty = 1 \cdot e^{-\lambda(i_0+i_1+\dots)} = e^{\lambda(1+\mu-s_\infty)}$$

$$I_\infty \approx n i_\infty = n(1 - s_\infty)$$

## Reed-Frost Model

$K = I_0 + I_1 + I_2 + \dots = \text{total epidemics}$

- If  $\frac{m}{n} \rightarrow \mu > 0 \Rightarrow \frac{K - n(1 - s_\infty)}{\sqrt{n(1 - s_\infty)}} \rightarrow N(0, 1)$
- If  $\frac{m}{n} \rightarrow \mu = 0 \Rightarrow$ 
  - $\lambda \leq 1 \Rightarrow K \sim \text{progeny of G-W process}$
  - $\lambda > 1 \Rightarrow$ 
    - \* w/prob.  $\sigma^m$   $K \sim \text{progeny of G-W}$
    - \* w/prob.  $1 - \sigma^m$   $\frac{K - n(1 - s_\infty)}{\sqrt{n(1 - s_\infty)}} \rightarrow N(0, 1)$

## General epidemic process

- Continuous time,  $S_0 = n, I_0 = m$
- Again  $S \rightarrow I \rightarrow R$  w/transition probabilities conditional on  $(S_t, I_t) = (S, I)$

$$\mathbf{P}_t \left( (S_{t+h}, I_{t+h}) = (S-1, I+1) \right) = \theta SIh + o(h)$$

$$\mathbf{P}_t \left( (S_{t+h}, I_{t+h}) = (S, I-1) \right) = \rho Ih + o(h)$$

$$\begin{aligned} \mathbf{P}_t \left( (S_{t+h}, I_{t+h}) = (S, I) \right) &= \\ &= 1 - \theta SIh - \rho Ih + o(h), \end{aligned}$$

all other transitions w/prob.  $o(h)$



# General epidemic process

Sellke's beautiful coupling argument

- wlog take  $\theta = 1$
- Index  $S$ -indiv. by  $i$  and  $I$ -indiv. by  $j$ ,  
so  $1 \leq i \leq n$  and  $1 \leq j \leq m$ .
- $\{l_i\}_{i=1}^n$  - "resistance to infection"  
 $l_{(1)} < l_{(2)} < \dots < l_{(n)}$  - order statistics  
i.i.d.  $\text{Exp}(1)$ .
- $\{\hat{r}_j\}_{j=1}^m$  - "infectious period" for  $I$ -indiv.  
 $\{r_i\}_{i=1}^n$  - "infectious period" for  $S$ -indiv.  
i.i.d.  $\text{Exp}(\rho)$ .  
 $r^{(k)} = r_i$  if  $l_{(k)} = l_i$

## General epidemic process

- Originally infected remain for  $\hat{r}_j$  time units
- S-indiv. indexed  $i$  accumulates “exposure” at rate  $I_t$  at time  $t$
- When “exposure”  $> l_i \Rightarrow S \rightarrow I$   
then stays infected for  $r_i$  time units

## General epidemic process

- $\nu$  = number of new infections
- If  $l_{(1)} > \sum_{j=1}^m \hat{r}_j$  then  $\nu = 0$   
else S-indiv. with  $l_{(1)}$  is infected and  $\nu \geq 1$
- Induction

$$\nu + 1 = \min \left\{ k : l_{(k)} > \sum_{j=1}^m \hat{r}_j + \sum_{i=1}^{k-1} \hat{r}^{(i)} \right\}$$

if inequality doesn't hold for  $\forall k$  then  $\nu = n$ .

## General epidemic process

Processes indexed by  $k$

**Theorem.** If  $n_k \rightarrow \infty$ ,  $\rho_k \rightarrow \infty$ , and

$$n_k \exp \left\{ -\frac{n_k + m_k}{\rho_k} \right\} \rightarrow b, \quad 0 < b < \infty,$$

then  $S_k(\infty) = n_k - \nu_k \rightarrow_{\mathcal{D}} \mathbf{Poisson}(b)$ .

Let  $R = \sum_{j=1}^m \hat{r}_j + \sum_{i=1}^{\nu} \hat{r}^{(i)}$  total exposure.

# General epidemic process

## Outline of the proof

- $\rho = o(n + m)$

- Almost birth-and-death process

- Death-rate to birth-rate is

$$\frac{\rho I_t}{S_t I_t} = \frac{\rho}{S_t} < \frac{\rho}{\epsilon(n + m)} \equiv q$$

on  $\{S_t > \epsilon(n + m)\}$ .

- Prob. of extinction with  $S(\infty) > \epsilon(n + m)$

< Prob. of extinction with D-B ratio  $q$

$$= q^m \rightarrow 0.$$

# General epidemic process

Outline of the proof (cont'd)

- Most people become infected, LLN gives

$$R \approx \sum_{j=1}^m \hat{r}_j + \sum_{i=1}^n \hat{r}^{(i)} \approx \frac{n+m}{\rho}$$

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$$\begin{aligned} S(\infty) = \# \text{ of } l_i \text{'s } > R &\approx \frac{n+m}{\rho} \\ &\approx \mathbf{Bin} \left( n, e^{-\frac{n+m}{\rho}} \right) \\ &\approx \mathbf{Poisson}(b). \end{aligned}$$

## General sampling procedure

- $(S_0, I_0) = (n, m)$ .
- I-person draws size of contacts  $\sim p$   
 $p = (p_0, p_1, p_2, \dots)$ , moments =  $(\lambda, \sigma^2)$
- $I_t$  contacts all at time  $t$ :  $\{S_t\} \cup \{I_t\}$
- $U_k = \#$  of susceptibles after  $k$  samplings.

$$U_0 = S_0 = n, I_0 = m$$

$$K_t = I_0 + I_1 + \dots + I_t = n + m - S_t$$

can recover

$$S_{t+1} = U_{K_t}, I_t = S_t - S_{t+1}$$

## General sampling procedure

- The process runs until  $I_t = 0$

$\Rightarrow S_t = S_{t+1}$ , and therefore

$$\begin{aligned} K_t &= n + m - S_t = n + m - S_{t+1} \\ &= n + m - U_{K_t} \end{aligned}$$

for the first time.

- Let  $Y_k \equiv U_k - nw^k$ , ( $w \equiv 1 - \lambda/n$ )

2 moments + tightness show that

$$\{n^{-1/3}Y_{tn^{2/3}}\} \Rightarrow \sigma W_t.$$



# General sampling procedure

The outcome of an almost critical epidemic  
Martin L of (1998)

- Total epidemic  $K = \sum_t I_t =$  first time when  $U_k$  hits  $n + m - k$ , or equivalently when  $n - U_k - k$  hits  $-m$

- Let  $\lambda = 1 + a/n^{1/3}$  and  $m = bn^{1/3}$ , same as

$$n^{1/3} \left( at - t^2/2 + \mathbf{O}(n^{-1/3}) \right) - Y_{tn^{2/3}}$$

hits  $-bn^{1/3}$ , simplifies to

$$W_t \text{ hits } b + at - t^2/2 \rightsquigarrow T.$$

- Finally, recall  $K = n^{2/3}T$ .