

THE STRUCTURE OF EPIDEMIC MODELS

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http://www.hjones-sons.co.uk/fmdvideo.htm









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1 Building models

INTERESTS

• Invasion – threshold? R_0 ?

 R_0

The **basic reproductive ratio** of an epidemic is the mean number of new infections made by an infected individual in a mostly susceptible population

1 Building models

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- Invasion threshold? R_0 ?
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- Invasion threshold? R_0 ?
- Spread velocity / duration? final size?
- *Persistence?* pattern? control?

Types of model

- individual or continuous population?
- stochastic or deterministic?

Population space

- \bullet mean-field
- \bullet metapopulations
- \bullet spatial
- \bullet small-world

$Equilibrium\ formulae$

$$\pi_i = \tau_i / L$$

e.g.

$$\pi_S = S/N = A/L$$

for 'once-only' disease

Assuming homogeneous mixing,

$$R_0 = \pi_S = S/N,$$

so $R_0 = A/L.$

Ross (ca. 1911) $R_0 = Mb^2cd$ where M is # mosquitoes/human Ross (ca. 1911) $R_0 = Mb^2cd$ where M is # mosquitoes/human,

whence

 $R_0 < 1$ iff $M < M_c$.



2. Simple network models



Three outbreaks of measles (Enko 1889)

In each generation, Pr(escape infection) $= (1 - i)^{pN}$ (Enko 1889) $= (1 - p)^{iN}$ (Reed-Frost) where p = P(contact), i = proportion infected

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Simple network model (random graph)



Here $R_0 \equiv Np$ is < 1

Results for simple random graph:

Giant component exists iff $R_0 > 1$.

Diameter of giant, $T \sim \log N$.

Final size (and probability of a large outbreak) are both given by the largest solution of

$$z = 1 - \exp(-R_0 z)$$

Deterministic (continuous population) equivalent,

a differential equation model ('SIR'):

$$\dot{S} = -cSI \dot{I} = cSI - dI \dot{R} = dI$$

Results for 'SIR':

Large outbreak always occurs if $R_0 \equiv c/d > 1$, duration $T \sim \log N$, and the final size z is given by

$$z = 1 - \exp(-R_0 z)$$



Stochastic structures

"UNNECESSARY DETAILS"

Example:

Simple birth and death process

(1) $r_{n,n+1} = an$, $r_{n,n-1} = bn$

(2) Independent individuals, each with birth rate a and death rate b.

P(extinction) μ_n when initial pop. = n ?

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$$\mu_n = \mu_1^n$$

COUPLING



... is monotone with initial set

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... is monotone with initial set

... and with infection parameter

THINKING IN THE WRONG ORDER

LOOKING AT SIMPLER MODEL FIRST

. . .



3 Metapopulation models

Consider a population with local and global contacts



where the geography can be either mean-field
... or spatial



('great circle' or 'small world' model)

Consider first the process including only global contacts, with reproductive ratio $R_0 = Nq$.



Relative to this 'global-only' process, local contacts have an amplifying effect. Hence the overall reproductive ratio is

$$R_T = R_0 \mu$$

where μ is the mean size of a local outbreak.

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A key question for control is whether you can get local outbreaks below threshold (compare SARS and swine flu?)



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http://www.cybergeo.eu/index12803.html



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"swine flu map youtube"



4 Spatial models





Nearest-neighbour

... or more general dispersal distribution



Х



time



CALCULATION OF VELOCITIES

Provided the dispersal distribution falls off at least exponentially, deterministic models do provide reasonable approximations.

Many nonlinear spatial deterministic models have been studied, especially diffusion equations (KPP, Fisher, Skellam, ...)

Breakthrough in late 1980s: the approach of Diekmann (and others) shows how linear theory can find velocities for a wide range of nonlinear models.

All you need is the reproduction and dispersal kernel K that describes the space-time distribution of the infections made by an individual in a mostly susceptible population.

Can think of K as a space-time version of R_0

Three advantages of the R&D kernel approach:

- Much easier to calculate
- Not restricted to DEs and diffusion equations
- Can look at the broad dependence of the velocity on basic components
 (e.g. is it ~ log(R₀), ~ √R₀ or ~ R₀ ?)

.. but note

These calculated velocities are somewhat larger than those of the more realistic individual stochastic models - .. but note

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The problem is treating the population as continuous (atto-foxes) rather than determinism per se



5 Small worlds



Threshold: $R_T = R_0 \mu > 1$ (as for metapopulation model)

T reduces from $\sim N$ to $\sim \log N$ as the number of global links increases

'Small world' phenomenon:

The proportion of global links required to reduce T to $\sim \log N$ is surprisingly small. 'Scale-free' models:

A related study is of models with very high variability in the number of contacts per individual.



NOTE When considering different degree distributions

(a) the epidemic is run on a fixed network.(b) links from/to an individual are not independent

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(b) links from/to an individual are not independent.

Compare SRG / Reed-Frost where:

(a) doesn't matter

(b) they are independent.





T=2





Footnotes

The structuralist ethic

Only trust a model if you can take it apart and put it together again

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LIMITS OF PREDICTION

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LIMITS OF PREDICTION

Avian flu in humans currently has an R_0 of ~ 0.02 .



6 Pair approximations

Pair approximations try to use local correlations to capture spatial structure.



time

Reconsider the deterministic SIR:

$$\dot{S} = -cSI \dot{I} = cSI - dI \dot{R} = dI$$
More accurately

$$\dot{S} = -c[SI] \dot{I} = c[SI] - dI \dot{R} = dI$$

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$$\begin{split} & [\dot{SS}] = -2c[SSI] \\ & [\dot{SI}] = c([SSI] - [SI] - [ISI]) - d[SI] \\ & [\dot{SR}] = \cdots \\ & [\dot{II}] = \cdots \end{split}$$

For closure, use

$$[ABC] \approx (1 - \frac{1}{n}) \frac{[AB][BC]}{[B]} \times (1 - \phi + \phi \frac{[AC]}{[A][C]})$$

where the clustering parameter ϕ is

P(ac|ab & bc)

(Keeling 1999)

EXAMPLE hexagonal lattices (HBFs)



$$\phi = 6/15 = 0.4$$

We have seen that the SIR on this graph has $T \sim \sqrt{N}$.

How about the pair approximation SIR with $\phi=0.4$?

SIR (dashed line) and its pair approximation (solid line), for $\phi = 0, 0.2, 0.4$.

Also, spatial SIR ('S') and ordinary deterministic SIR (' \star ').



time

The pair approximation with $\phi = 0.4$ does fit well for epidemics on 'typical' graphs of degree 6 and clustering parameter $0.4 \dots$

... but **not** for the spatial (hexagonal) SIR.

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... but **not** for the spatial (hexagonal) SIR.

Is there a paradox here?

Just because a graph has degree 6 and clustering parameter 0.4, it doesn't have to be hexagonal. Just because a graph has degree 6 and clustering parameter 0.4, it doesn't have to be hexagonal.

In fact that's very very unlikely – we might say *Adams-improbable*.

'We are now cruising at a level of $2^{25,000}$ to 1 against and falling, and we will be restoring normality just as soon as we are sure what is normal anyway.'

(Adams 1979)

1 References

Adams, D (1979) The Hitchhiker's Guide to the Galaxy, Pan Books.

For other references see

http.ma.hw.ac.uk/~denis/

