

11/29/2022

Structured models

So far we have looked only at unstructured models which do not distinguish between age, sex, or other types. Here we will explore a type of model called risk structured model, which makes most sense when applied to STIs.

The most important question: how does this model change the predictions of the unstructured model?

Consider 2 classes: low risk and high risk

Assumptions:

- 1) Ignore demographics (birth/death)
- 2) Assume a SIS model (curable STIs)

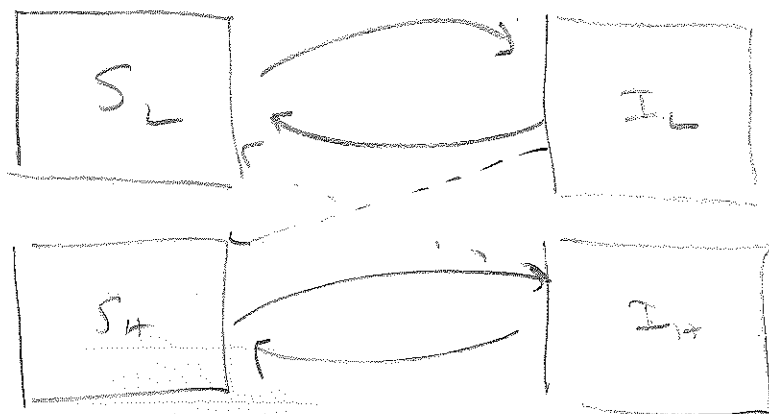
4 classes: $\hat{S}_L, \hat{S}_H, \hat{I}_L, \hat{I}_H$

Let $N = N_L + N_H$ = total population

$$S_L = \hat{S}_L / N \quad I_L = \hat{I}_L / N \quad S_H = \hat{S}_H / N \quad I_H = \hat{I}_H / N$$

which gives $S_L + I_L + S_H + I_H = 1$.

and let $n_L = S_L + I_L = N_L / N$ = fraction low risk and likewise for high risk



Let δ_H = recovery rate for high risk, δ_L = recovery rate for low risk.

β_{ij} = transmission rate from class j to class i
 (# contacts per time per susceptible in class i in class j)

ODEs

$$S'_L = -\beta_{LL} S_L I_L - \beta_{LH} S_L I_H + \delta_L I_L$$

$$I'_L = \beta_{LL} S_L I_L + \beta_{LH} S_L I_H - \delta_L I_L$$

$$S'_H = -\beta_{HH} S_H I_H - \beta_{HL} S_H I_L + \delta_H I_H$$

$$I'_H = \beta_{HH} S_H I_H - \beta_{HL} S_H I_L - \delta_H I_H$$

The force of infection $\lambda =$ per capita rate at which susceptible acquire infection. Two of them in the model (low risk and high risk).

$$\lambda_L = \beta_{LL} I_L + \beta_{LH} I_H$$

$$\lambda_H = \beta_{HH} I_H + \beta_{HL} I_L$$

Can write this in matrix form as

$$\begin{pmatrix} \lambda_L \\ \lambda_H \end{pmatrix} = \begin{pmatrix} \beta_{LL} & \beta_{LH} \\ \beta_{HL} & \beta_{HH} \end{pmatrix} \begin{pmatrix} I_L \\ I_H \end{pmatrix}$$

This matrix has a special name - who acquires infection from whom (WAZFW) Often useful to make assumptions about this matrix to determine values of β_{ij}

1) assortative mixing: high risk individuals are more likely to partner w/ high and low w/ low,

$$\beta_{LL}, \beta_{HH} > \beta_{LH}, \beta_{HL}$$

transmission highest among high risk $\beta_{HH} = \max(\beta_{ij})$

2) Sym. mix: ^{number of} interactions between high and low risk is the same as low and high $\beta_{LH} = \beta_{HL}$

Exercise Find the DFE

$$I_L = 0 \quad I_H = 0 \quad S_L = n_L \quad S_H = n_H$$

Possible to show

$$R_0 = \frac{\beta_{HH} n_H + \beta_{LH} n_L}{\gamma_H} + \frac{\beta_{LL} n_L + \beta_{HL} n_H}{\gamma_L}$$

$$= R_0^H + R_0^L$$

So the total R_0 is a weighted sum of the R_0 values for each risk class in the unstructured model, it's just β/γ . Suppose γ 's were the same

$$R_0 = \frac{(\beta_{HH} + \beta_{HL}) n_H + (\beta_{LL} + \beta_{LH}) n_L}{\gamma}$$

$$\text{Effective } \beta = (\beta_{HH} + \beta_{HL}) n_H + (\beta_{LL} + \beta_{LH}) n_L$$

Max. high risk $= \beta_{HH} n_H + \beta_{LL} n_L + \beta_{mix}$
 makes β higher \rightarrow