# SEIR Graph Model

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## 1 SEIRS Model

Deterministic SEIR(S) Model:

$$\dot{S} = \frac{-\beta SI}{N} + \xi R + \nu N - \mu_0 S$$

$$\dot{E} = \frac{\beta SI}{N} - \sigma E - \mu_0 E$$

$$\dot{I} = \sigma E - \gamma I - \mu_I I - \mu_0 I$$

$$\dot{R} = \gamma I - \xi R - \mu_0 R$$

$$\dot{F} = \mu_I I$$

$$N = S + E + I + R$$

Variables & Parameters:

### Description

- S Number of susceptible individuals
- E Number of exposed individuals
- I Number of infectious individuals
- R Number of recovered individuals
- F Number of infection-related fatalities
- N Total number of living individuals
- $\beta$  Rate of transmission (exposure)
- $\sigma$  Rate of infection (upon exposure)
- $\gamma$  Rate of recovery (upon infection)
- $\xi$  Rate of re-susceptibility (upon recovery)
- $\mu_I$  Rate of infection-related death
- $\mu_0$  Rate of baseline death
- $\nu$  Rate of baseline birth

Setting  $\nu=\mu_0$  maintains constant population size (excluding infection-related fatalities) with corresponding rate of birth-death turnover introducing new susceptible individuals. Setting  $\nu$ ,  $\mu_0$ , and/or  $\xi$  to 0 removes vitality and/or re-susceptibility terms, respectively, and reduces the model to basic SEIR dynamics (with mortality).

## Stochastic SEIR(S) Model on a Graph:

Consider a graph G representing individuals (nodes) and their interactions (edges). Each individual is adjacent to a set of nodes that defines its set of "close contacts"  $C_G(i) = \{j : (i,j) \in G\}$ . At a given time, individual i makes contact with a random individual from their set of close contacts  $(j \in C_G(i))$  with probability 1-p or with a random individual from anywhere in the network with probability p. The latter global contacts represent individuals interacting with the population at large (i.e., individuals outside their social circle, such as on public transit, at an event, etc) with some probability. The parameter p defines the locality of the network: for p=0 an individual only interacts with their close contacts while p=1 represents a uniformly mixed population. Social distancing interventions may increase the locality of the network (i.e., decrease p) and/or decrease local connectivity of the network (i.e., decrease the degree of individuals,  $d_i = |C_G(i)|$ ).

Each individual i has a state  $X_i \in \{S, E, I, R, F\}$ . State transitions corresponding to the SEIRS dynamics are

$$\Pr(X_i = S \to E) = \left[ p rac{eta I}{N} + (1-p) rac{eta \sum_{j \in C_G(i)} \delta_{X_j = I}}{|C_G(i)|} 
ight] \delta_{X_i = S}$$
 $\Pr(X_i = E \to I) = \sigma \delta_{X_i = E}$ 
 $\Pr(X_i = I \to R) = \gamma \delta_{X_i = I}$ 
 $\Pr(X_i = I \to F) = \mu_I \delta_{X_i = I}$ 
 $\Pr(X_i = any \to S) = \xi \delta_{X_i = R} + \nu \delta_{X_i \neq F}$ 

where  $\delta_{X_i=A}=1$  if the state of  $X_i$  is A, or 0 if not. When p=1 and  $N\to\infty$ , this stochastic model approaches the same dynamics as the deterministic SEIR model.

References: Dottori, M. and Fabricius, G., 2015. SIR model on a dynamical network and the endemic state of an infectious disease. Physica A: Statistical Mechanics and its Applications, 434, pp.25-35.

## 2 SEIRS Model with Testing

Deterministic SEIR(S) Model with Testing:

$$\dot{S} = -\frac{\beta SI}{N} - q \frac{\beta_D SD_I}{N} + \xi R + \nu N - \mu_0 S$$

$$\dot{E} = \frac{\beta SI}{N} + q \frac{\beta_D SD_I}{N} - \sigma E - \theta_E \psi_E E - \mu_0 E$$

$$\dot{I} = \sigma E - \gamma I - \mu_I I - \theta_I \psi_I I - \mu_0 I$$

$$\dot{D}_E = \theta_E \psi_E E - \sigma D_E - \mu_0 D_E$$

$$\dot{D}_I = \theta_I \psi_I I + \sigma D_E - \gamma D_I - \mu_I D_I - \mu_0 D_I$$

$$\dot{R} = \gamma I + \gamma D_I - \xi R - \mu_0 R$$

$$\dot{F} = \mu_I I + \mu_I D_I$$

$$N = S + E + I + D_E + D_I + R$$

#### Variables & Parameters:

## Description

- S Number of susceptible individuals
- E Number of exposed individuals
- I Number of infectious individuals with undetected cases
- $D_E$  Number of exposed individuals with detected cases
- $D_I$  Number of exposed individuals with detected cases
- R Number of recovered individuals
- F Number of infection-related fatalities
- N Total number of living individuals
- β Rate of transmission (exposure, for individuals with undetected infections)
- $\beta_D$  Rate of transmission (exposure, for individuals with positively detected infections)
- $\sigma$  Rate of infection (upon exposure)
- $\gamma$  Rate of recovery (upon infection)
- $\xi$  Rate of re-susceptibility (upon recovery)
- $\mu_I$  Rate of infection-related death
- $\mu_0$  Rate of baseline death
- $\nu$  Rate of baseline birth
- $\theta_E$  Rate of baseline testing (for exposed individuals)
- $\theta_I$  Rate of baseline testing (for infectious individuals)
- $\phi_E$  Rate of testing when a close contact has tested positive (for exposed individuals)
- $\phi_I$  Rate of testing when a close contact has tested positive (for infectious individuals)
- $\psi_E$  Rate of positive test (given exposed state)
- $\psi_I$  Rate of positive test (given infectious state)
- q Rate of individuals with detected infection interacting with population

### Stochastic SEIR(S) Model with Testing on a Graph:

Consider a graph G representing individuals (nodes) and their interactions (edges) as before. Each individual i has a state  $X_i \in \{S, E, I, D_E, D_I R, F\}$ , where the new states  $D_E$  and  $D_I$  represent individuals who have tested positively for infection (exposed and infectious, respectively). Infectious individuals (state I) are tested at a baseline rate  $\theta$ , and may be tested at a higher rate  $\phi$  if another individual in their close contacts (i.e.,  $j \in C_G(i)$  has tested positive. Exposed and infectious individuals test positively rate  $\psi_E$  or  $psi_I$ , respectively, which moves them into the  $D_E$  or  $D_I$  state representing a detected case.

Now we also consider another graph Q which represents the interactions that each individual has if they test positively for the disease and enter into a form of quarantine. That is, individuals in the  $D_E$  or  $D_I$  states. The quarantine has the effect of dropping some fraction of the edges connecting the quarantined individual to others (according to some rule to be specified elsewhere, such as dropping a random fixed percentage of a node's edges). The edges of Q for each individual are then a subset of the edges of G for that individual. The set of nodes that are adjacent to a quarantined individual define their set of "quarantine contacts"  $C_Q(i) = \{j: (i,j) \in Q\}$ . At a given time, a quarantined individual i may come into contact with another individual in this quarantine contact set. A quarantined individual i may also be accessible to contact with a random individual from anywhere in the network with rate

State transitions corresponding to the SEIRS dynamics with testing are

$$\Pr(X_{i} = S \to E) = \left[ p \left( \frac{\beta I + q \beta_{D} D_{I}}{N} \right) + (1 - p) \left( \frac{\beta \left[ \sum_{j \in C_{G}(i)} \delta_{X_{j} = I} \right] + \beta_{D} \left[ \sum_{k \in C_{Q}(i)} \delta_{X_{k} = D_{I}} \right]}{|C_{G}(i)|} \right) \right] \delta_{X_{i} = S}$$

$$\Pr(X_{i} = E \to I) = \sigma \delta_{X_{i} = E}$$

$$\Pr(X_{i} = I \to R) = \gamma \delta_{X_{i} = I}$$

$$\Pr(X_{i} = I \to F) = \mu_{I} \delta_{X_{i} = I}$$

$$\Pr(X_{i} = E \to D_{E}) = \left( \theta_{E} + \phi_{E} \left[ \sum_{j \in C_{G}(i)} \delta_{X_{k} = D_{E}} + \delta_{X_{k} = D_{I}} \right] \right) \psi_{E} \delta_{X_{i} = E}$$

$$\Pr(X_{i} = I \to D_{I}) = \left( \theta_{I} + \phi_{I} \left[ \sum_{j \in C_{G}(i)} \delta_{X_{k} = D_{E}} + \delta_{X_{k} = D_{I}} \right] \right) \psi_{I} \delta_{X_{i} = I}$$

$$\Pr(X_{i} = D_{E} \to D_{I}) = \sigma \delta_{X_{i} = D_{E}}$$

$$\Pr(X_{i} = D_{I} \to R) = \gamma \delta_{X_{i} = D_{I}}$$

$$\Pr(X_{i} = D_{I} \to F) = \mu_{I} \delta_{X_{i} = D_{I}}$$

$$\Pr(X_{i} = D_{I} \to F) = \mu_{I} \delta_{X_{i} = D_{I}}$$

$$\Pr(X_{i} = any \to S) = \xi \delta_{X_{i} = R} + \nu \delta_{X_{i} \neq F}$$

where  $\delta_{X_i=A}=1$  if the state of  $X_i$  is A, or 0 if not.