## Kermack-McKendrick (SIR) Model of Infectious Diseases.

We suppose that the have a population $S$ of susceptible hosts for a disease, a population $I$ of infectious host and a population $R$ of recovered hosts. Roughly the situation may be flow charted as in the figure below:

$$
\begin{array}{|l|l|l|l|}
\hline S=\text { susceptible } & I=\text { infectious } \\
& \gamma I \\
\hline
\end{array}
$$

## Simplified SIR Model 01:

For our first simplified model we will assume:

- No births (or equivalently, birth rates = death rates);
- neither immigration nor emigration;
- removed $=0$;
- no loss of immunity.

This simplifies the flow chart to the following:

$$
S=\text { susceptible } \longrightarrow I=\text { Infective } \longrightarrow R=\text { Recovered }
$$

These assumptions imply $N(t)$ is constant, $N(t)=N(0)=N_{0}$. Each of the variables $S, I, R$ is a function of time $t$ where, in our cases, $t$ will be in days.

For some quantity $A$ let $\Delta A$ denote the change of quantity $A$ from day $m$ at midnight to day $m+1$ at midnight; so $\Delta A$ denotes the "(average) change of quantity $A$ per day".

$$
\begin{aligned}
\Delta S & =- \text { new infections } \\
\Delta I & =\text { new infections }- \text { new recovered } \\
\Delta R & =\text { new cures }
\end{aligned}
$$

We have:

$$
\Delta I=-\Delta S-\Delta R
$$

To set up the differential equations we consider an encounter between a susceptible and an infectious. We assume that there is a constant probability $\alpha$ of infection with each encounter. So assume an average of $n$ encounters per day between members of the population; a particular infectious individual will infect $n \alpha \frac{S}{N_{0}}$ individuals per day. (Note, there is the possibility of two infectious encountering the same susceptible, this is on the order of $\left(\frac{n}{N_{0}}\right)^{2}$ and for large populations $n \ll N_{0}$ this
quantity, along with higher order multiple encounters is negligible.) So the total number of individuals infected by the infectious group per day is

$$
n \alpha \frac{S}{N_{0}} I .
$$

Let $\lambda=\frac{n \alpha}{N_{0}}$. This gives us equation (1):

$$
\begin{equation*}
\frac{d S}{d t}=S^{\prime}=-\lambda I S \tag{1}
\end{equation*}
$$

Assume that infectious individuals become cured at a constant rate proportional to the number of infectious individuals; assume this happens with proportionality constant $\gamma$ so that gives us equation (2)

$$
\begin{equation*}
\frac{d R}{d t}=R^{\prime}=\gamma I \tag{2}
\end{equation*}
$$

Finally since the change of $I$ is the change of $S$ moving into $I$ and minus the change of $R$ coming from $I$ we have equation (3) and it's equivalent, equation (4):

$$
\begin{align*}
\frac{d I}{d t}=I^{\prime} & =-\frac{d S}{d t}-\frac{d R}{d t} \\
& =-S^{\prime}-R^{\prime} \\
& =\lambda I S-\gamma I  \tag{3}\\
& =(\lambda S-\gamma) I . \tag{4}
\end{align*}
$$

Following are some conclusions that can be made regarding this system of equations without solving them. Observe that by assumption $N(t)$ is constant so:

$$
\begin{aligned}
N(t) & =N(0)=N_{0} \\
S(t)+I(t)+R(t) & =N_{0} \\
S^{\prime}(t)+I^{\prime}(t)+R^{\prime}(t) & =0 .
\end{aligned}
$$

Since $S, I$ and $\lambda$ are all positive, equation (1) tells us that $S^{\prime}(t)<0$ so,

$$
\begin{aligned}
S^{\prime}(t) & <0 \\
\therefore \quad S(t) & <S_{0} \text { for all } t>0 \\
\text { So } \lambda S-\gamma & <\lambda S_{0}-\gamma \\
(\lambda S-\gamma) I & <\left(\lambda S_{0}-\gamma\right) I \\
I^{\prime} & <\left(\lambda S_{0}-\gamma\right) I .
\end{aligned}
$$

If $I^{\prime}<0$ then the epidemic fizzles out. So using this result with equation (3) from above we have that if $\lambda S_{0}-\gamma<0$ then the epidemic
fizzles; this happens when

$$
\begin{aligned}
\lambda S_{0}-\gamma & <0 \\
\lambda S_{0} & <\gamma \\
S_{0} & <\frac{\gamma}{\lambda} .
\end{aligned}
$$

For a fixed disease we have no control over the constant $\gamma$. But we have some control over the "constant" $\lambda$ : since we assume that $N(t)$ is the constant $N_{0}$ we have

$$
\lambda=\frac{n \alpha}{N_{0}}
$$

where, again, $\alpha$ is a constant related to the disease. But we can decrease $\alpha$ by making it less likely that when a susceptible meets an infectious that there will be a new infection by using masks and social distancing. Assuming we do this to obtain $\alpha^{\prime}$ a smaller $\alpha$, then since we want $I^{\prime}<0$. So we want:

$$
\begin{aligned}
S_{0} & <\frac{\gamma}{\lambda} \\
S_{0} & <\frac{\gamma N_{0}}{n \alpha^{\prime}} \\
n \alpha^{\prime} S_{0} & <\gamma N_{0} \\
n & <\frac{\gamma N_{0}}{\alpha^{\prime} S_{0}} .
\end{aligned}
$$

Assuming $S_{0} \approx N_{0}$ so that $\frac{N_{0}}{S_{0}} \approx 1$ this gives us a way to combat an epidemic. Obviously the smaller the value of $n$, the smaller the average number of interactions between members of our population, the faster the epidemic will end. But we must at least have $n<\frac{\gamma}{\alpha^{\prime}}$ in order to have any hope that the epidemic can be stopped. So the strategy of decreasing $n$ by sequestering enough individuals well below the $n<\frac{\gamma}{\alpha^{\prime}}$ threshold will eventually cause the epidemic to end.

