

CT561: Systems Modelling & Simulation

Lecture 9: SIR Model Part 1

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<https://github.com/JimDuggan/SDMR>



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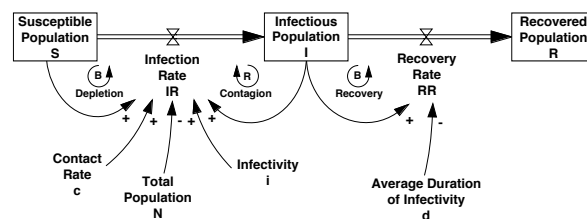
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Public Health: Modelling Infectious Disease Outbreaks

- Context
- SIR Model
- Related variables



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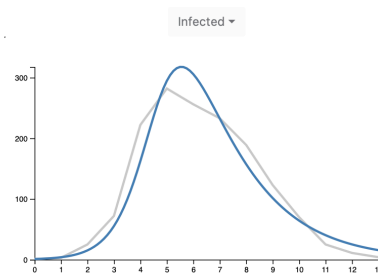
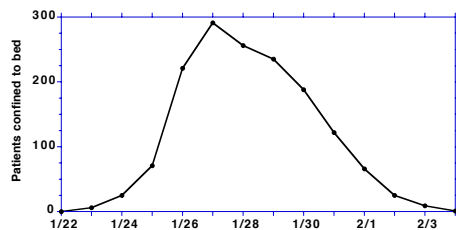
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(1) Context – Outbreak Dynamics

Figure 9-3
Dynamics of epidemic disease

Sources: *British Medical Journal*, 4 March 1978, p. 587;.



Influenza epidemic at an English boarding school, January 22-February 3, 1978. The data show the number of students Confined to bed for influenza at any time (the stock of symptomatic individuals).



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Examples of Infectious Agents – Microparasites (Vynnycky & White)

Type of Agent	Characteristics	Examples
Virus	Small, simple, obligatory parasites of larger cells	Measles, Mumps, Rubella, Ebola, Smallpox, SARS, Influenza
Bacteria	Larger and more complex than viruses- many are able to grow independently but some require a cell host	Bordella pertussis (whooping cough), Mycobacterium tuberculosis (tuberculosis), Salmonella typhi (typhoid fever)
Protozoa	Large single-celled organisms, more complex than bacteria- many are able to grow independently but some require a cell host	Plasmodium falciparum (Malaria)



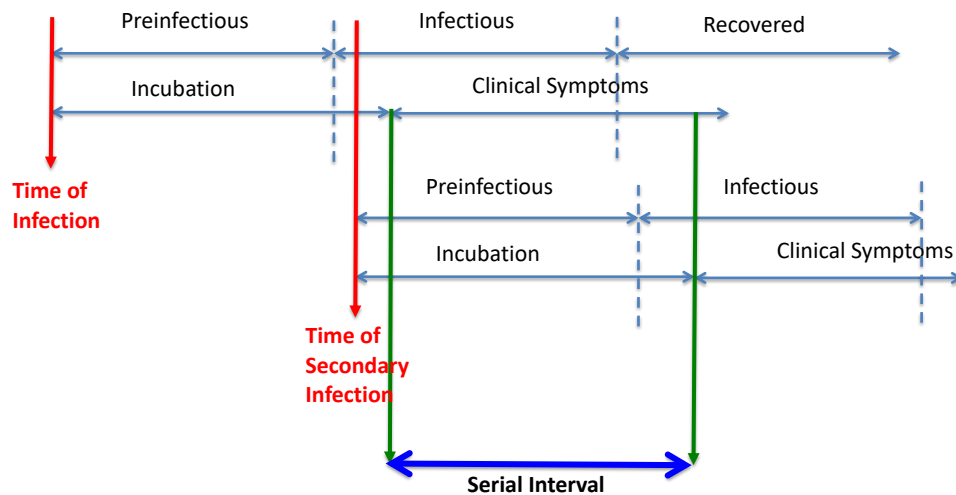
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Typical Life cycle of infection.



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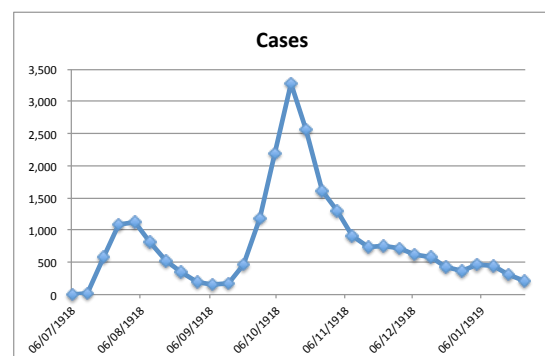
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Data from 1918/19

- The second wave of the 1918 pandemic was much deadlier than the first.
- The first wave had resembled typical flu epidemics; those most at risk were the sick and elderly, while younger, healthier people recovered easily.
- But in August, when the second wave began in France, Sierra Leone and the United States, the virus had mutated to a much deadlier form. This has been attributed to the circumstances of the First World War.

Gothenberg Data 1918/19



http://en.wikipedia.org/wiki/1918_flu_pandemic



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Challenge 9.1

<https://github.com/owid/covid-19-data/tree/master/public/data>

- Explore the data set our world in data (COVID-19)
- Focus on the epidemic curves (behaviour over time)

README.md

Data on COVID-19 (coronavirus) by Our World in Data

Our complete COVID-19 dataset is a collection of the COVID-19 data maintained by [Our World in Data](#). It is updated daily and includes data on confirmed cases, deaths, hospitalizations, and testing, as well as other variables of potential interest.

Download our complete COVID-19 dataset : [CSV](#) | [XLSX](#) | [JSON](#)

We will continue to publish up-to-date data on confirmed cases, deaths, hospitalizations, and testing, throughout the duration of the COVID-19 pandemic.

Our data sources

- Confirmed cases and deaths: our data comes from the [European Centre for Disease Prevention and Control \(ECDC\)](#). We discuss how and when the ECDC collects and publishes this data [here](#). The cases & deaths dataset is updated daily. Note: the number of cases or deaths reported by any institution—including the ECDC, the WHO, Johns Hopkins and others—on a given day does not necessarily represent the actual number on that date. This is because of the long reporting chain that exists between a new case/death and its inclusion in statistics. This also means that negative values in cases and deaths can sometimes appear when a country sends a correction to the ECDC, because it had previously overestimated the number of cases/deaths. Alternatively, large changes can sometimes (although rarely) be made to a country's entire time series if the ECDC decides (and has access to the necessary data) to correct values retrospectively.

A	B	C	D	E	F	G	H	I	J	K	L	M
iso_code	continent	location	date	total_cases	new_cases	cases_smoothed	total_deaths	new_deaths	deaths_smoothed	cases_per_100k	deaths_per_100k	recovered
ABW	North America	Aruba	2020-03-13	2	2	0.286		0		18.733	18.733	
ABW	North America	Aruba	2020-03-19			0.286		0				2.676
ABW	North America	Aruba	2020-03-20	4	2	0.286		0		37.465	18.733	2.676
ABW	North America	Aruba	2020-03-21			0.286		0				2.676
ABW	North America	Aruba	2020-03-22			0.286		0				2.676
ABW	North America	Aruba	2020-03-23			0.286		0				2.676
ABW	North America	Aruba	2020-03-24	12	8	1.429		0		112.395	74.93	13.38
ABW	North America	Aruba	2020-03-25	17	5	2.143		0		159.227	46.831	20.071
ABW	North America	Aruba	2020-03-26	19	2	2.429		0		177.959	18.733	22.747



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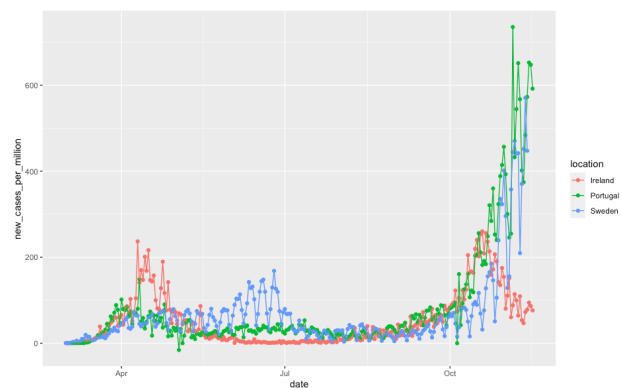
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(2) Susceptible-Infected-Recovered Model

- The total population of a region/community is divided into three categories:
 - Those susceptible to the disease (S)
 - Those who are infectious (I)
 - Those who have recovered (R)
- Births, deaths & migration ignored
- Population homogenous (no sub-groups)
- Once infected people recover after a time lag



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Stocks and Flows

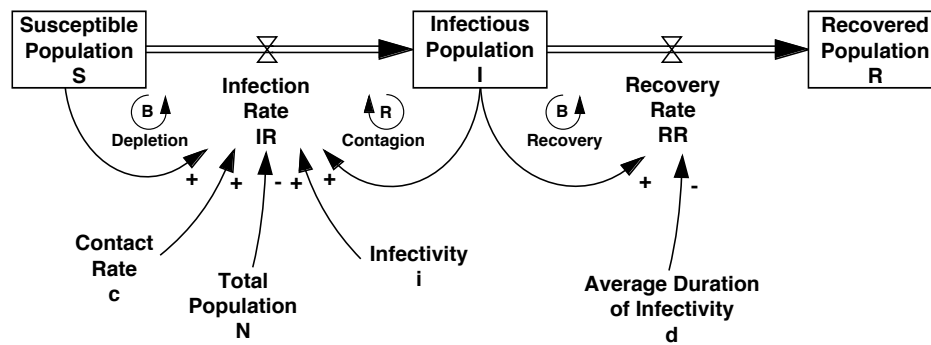


Figure 9-5 People remain infectious (and sick) for a limited time, then recover and develop immunity.



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Stock Equations

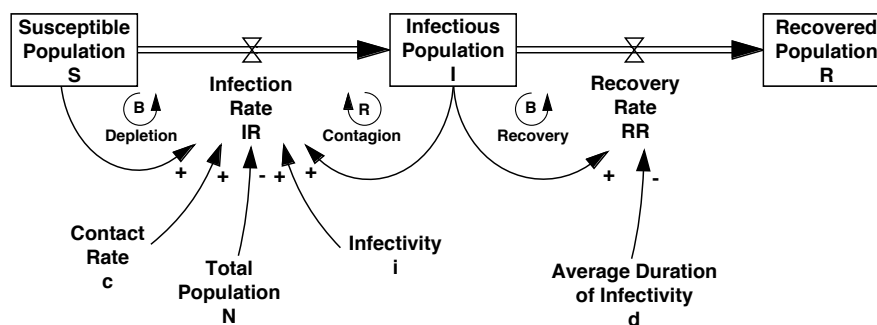


Figure 9-5 People remain infectious (and sick) for a limited time, then recover and develop immunity.

$$S = \text{INTEG}(-IR, 9999)$$

$$RR = I / d$$

$$I = \text{INTEG}(IR - RR, 1)$$

$$R = \text{INTEG}(RR, 0)$$



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Formulate the Infection Rate

- **Contact Rate** (people/person/time period)
- **Number of Infected**

Product is the Number of encounters for Infectious People per time period

- **Number of Susceptible**
- **Total Population**

Ratio is the chance of meeting a Susceptible Person

- **Infectivity**

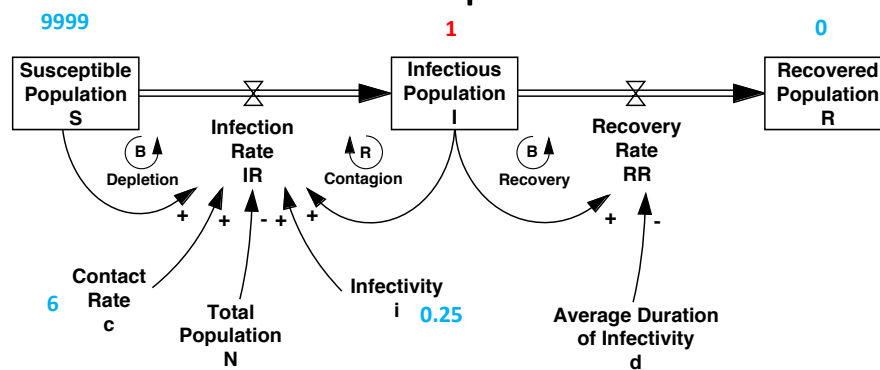
Probability that an encounter with a **Susceptible Person** will lead to an infection

$$IR = \text{Contact Rate} * \text{Infectious} * (\text{Susceptible}/N) * \text{Infectivity}$$



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Example



Number of Infectious Contacts = $1 * 6 = 6$ (#Encounters for infectious people)

Chance of Meeting Susceptible = $9999/10000 = 0.999$

Probability of Transmission = 0.25

$$IR = 6 * 0.9999 * 0.25 = 1.4985$$



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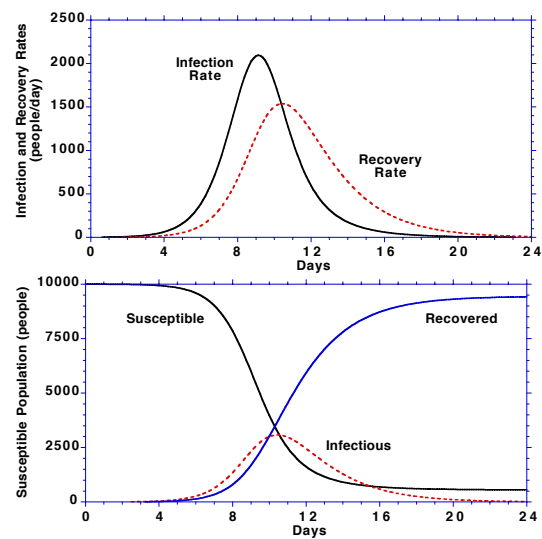
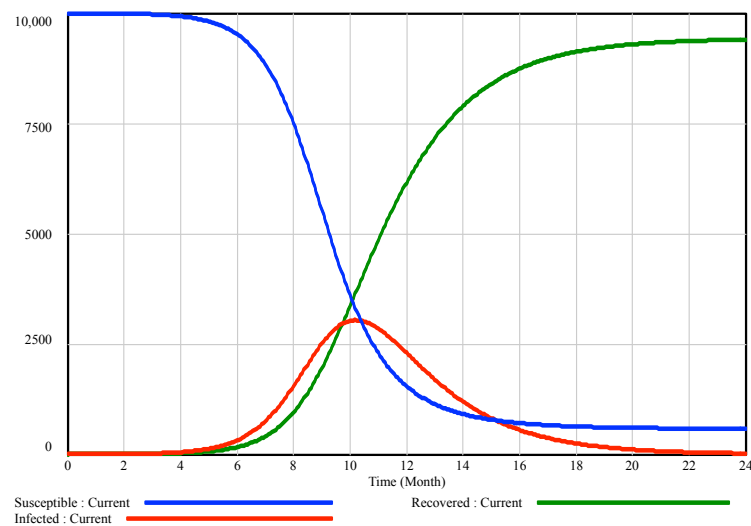


Figure 9-6 Simulation of an epidemic in the SIR model. The total population is 10,000. The contact rate is 6 per person per day, infectivity is 0.25, and average duration of infectivity is 2 days. The initial infective population is 1, and all others are initially susceptible.



Vensim Output



Epidemic dynamics for different contact rates

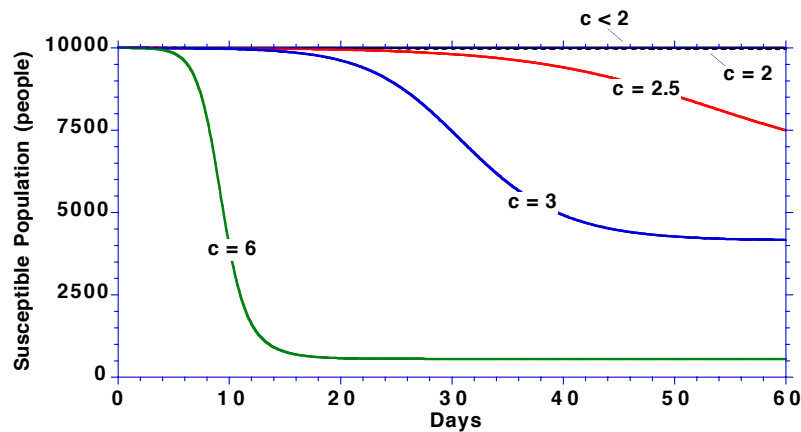


Figure 9-7 The contact rate is noted on each curve; all other parameters are as in Figure 9-6.



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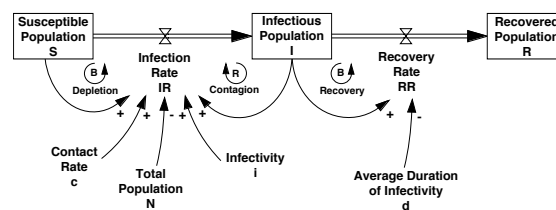
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Summary: *Key feature of SIR model*

- It captures a fundamental feature of infectious diseases
- The disease spreads through contact between susceptible and infectious
- Nonlinear equation



$$IR = \text{Contact Rate} * \text{Infectious} * (\text{Susceptible}/N) * \text{Infectivity}$$



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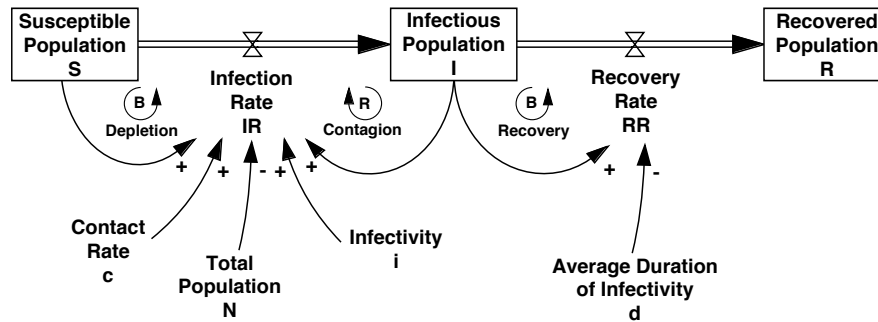
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Challenge 9.2



- Explore the SIR Model
- What three model conditions will halt disease spread?



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(3) Related variables

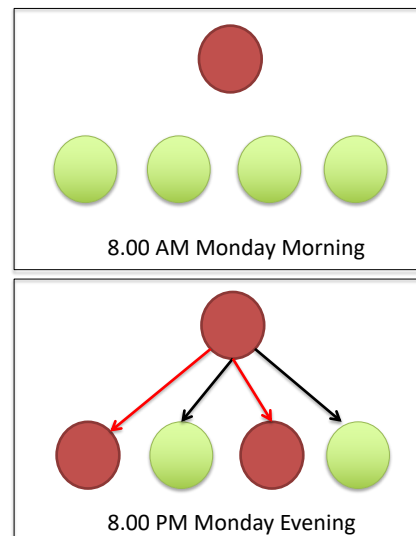
- Effective contacts
- Reproduction Number



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Effective Contacts (C_e) = $c * i$

- Defined as one which is sufficient to lead to infection, were it to occur between a **susceptible** and **infectious** individuals
- For example, if $C_e = 2$
 - An infectious person will infect two susceptible people in one day
 - They could meet 4 people, and pass on the virus with probability (0.50)



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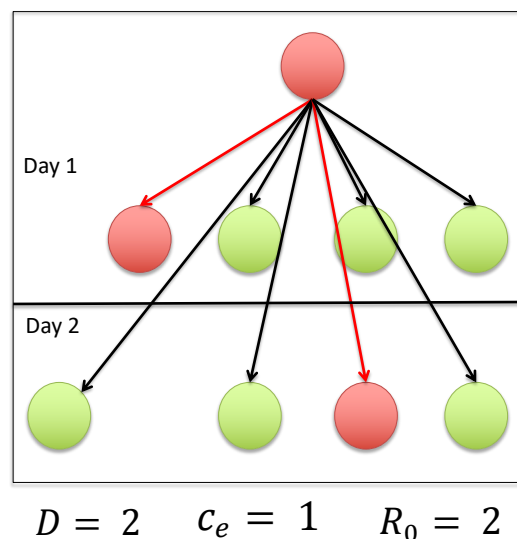
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Reproduction Number – R_0

- Formally defined as the average number of secondary infectious resulting from a typical infectious person being introduced to a totally susceptible population

$$R_0 = c_e D$$



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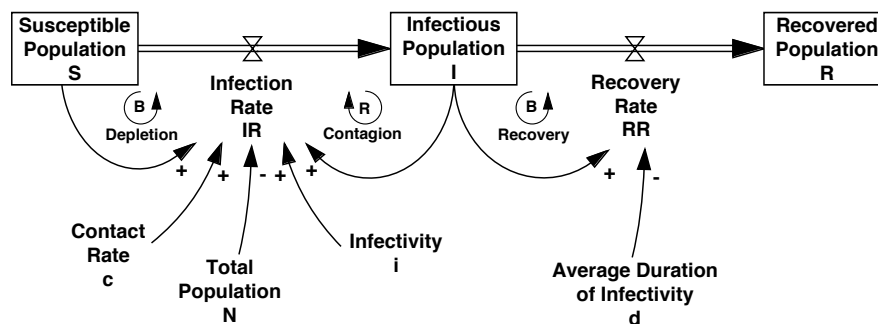
Approximate data for common potentially vaccine-preventable diseases

Infection	Serial Interval (Range)	R_0	Herd Immunity
Diphtheria	2-30 Days	6-7	85
Influenza	2-4 Days	2-4	50-75
Malaria	20 Days	5-100	80-99
Measles	7-16 Days	12-18	83-94
Pertussis	5-35 Days	12-17	92-94



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Challenge 9.3



- Add R_0 to the SIR Model



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Challenge 9.4 COVID-19 Model

- Build a population model of COVID-19
- Assume an SEIR basic structure
- Assume two streams:
 - Clinical Fraction (Pre-Clinical and Clinical)
 - Sub-clinical Fraction
- Assume subclinical are 50% as infectious as clinical
- Use Population ~ 5M
- 10 People infected at start
- Estimate R_0



Age-dependent effects in the transmission and control of COVID-19 epidemics

Nicholas G. Davies^{1,2}, Petra Klepac¹, Yang Liu¹, Kishor Prem¹, Mark Jit¹, CMMID COVID-19 working group and Rosalind M. Eggo^{1,2}

Parameter	Description	Value
Latent Time	Time Spent in Exposed Stock	3 Days
Pre-Infectious Clinical Time	Time Spent Pre-Infectious (and spreading)	2.1 Days
Infectious Time (Clinical)	Time spent infectious (and spreading)	2.9 Days
Infectious Time (Sub Clinical)	Time spent infectious	5 Days
Clinical Fraction	Proportion of infected who show symptoms	0.60
Contacts	Contacts/Person/Day	10
Infectivity	Probability of transmission	0.10



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