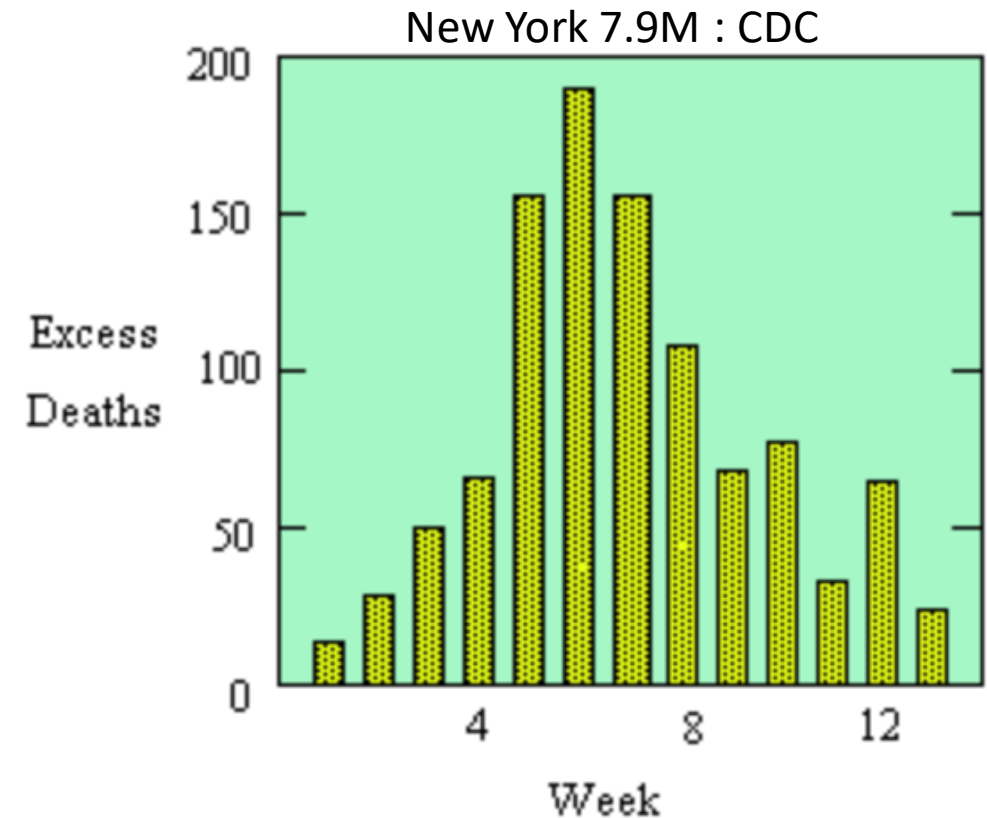


Hong Kong Flu (US, winter 1969-69)

virulent new strain of **influenza**, named *Hong Kong flu* for its place of discovery

Week	Flu-related deaths	Week	Flu-related deaths
1	14	8	108
2	28	9	68
3	50	10	77
4	66	11	33
5	156	12	65
6	190	13	24
7	156		

new cases \propto deaths 3 weeks ago



SIR MODEL - FOR SPREAD OF DISEASE

$S(t)$ = susceptible individuals
independent variable

$I(t)$ = infected " "
active cases

$R(t)$ = removed " "
total - active cases (recovered + dead)
dependent variables

$s(t), i(t), r(t)$ = fraction of population.

PROVE: $s(t) + i(t) + r(t) = 1$

assuming, births = 0 \Rightarrow susceptible can't increase
immigration = 0 \therefore susceptibles can only
become infected.

$\therefore s(t)$ is decreasing



assuming, each infected individual has "b" contacts per day.
not all of them are susceptible.

\therefore no. susceptible = $b \cdot s(t)$ per infected.
got infected.

\therefore total susceptible $\Delta S(t) = b \cdot s(t) \cdot I(t)$
got infected

$$\frac{dS}{dt} = -b \cdot s(t) \cdot I(t)$$

\nwarrow \nearrow
 contacts

$$\frac{dS}{dt} = -b \cdot s(t) \cdot i(t)$$

assuming, probability of an infected individual
recovering (or removed) is "k"

$$\Delta R(t) = k \cdot I(t)$$

$$\frac{dR}{dt} = k \cdot I(t)$$

$$\frac{dR}{dt} = k \cdot i(t)$$

\nwarrow probability of removal.

Since we assumed the total population to be constant,

$$\frac{ds}{dt} + \frac{di}{dt} + \frac{dr}{dt} = 0$$

or, equivalently, any increase in infected is due to decrease in susceptible and less than recovered.

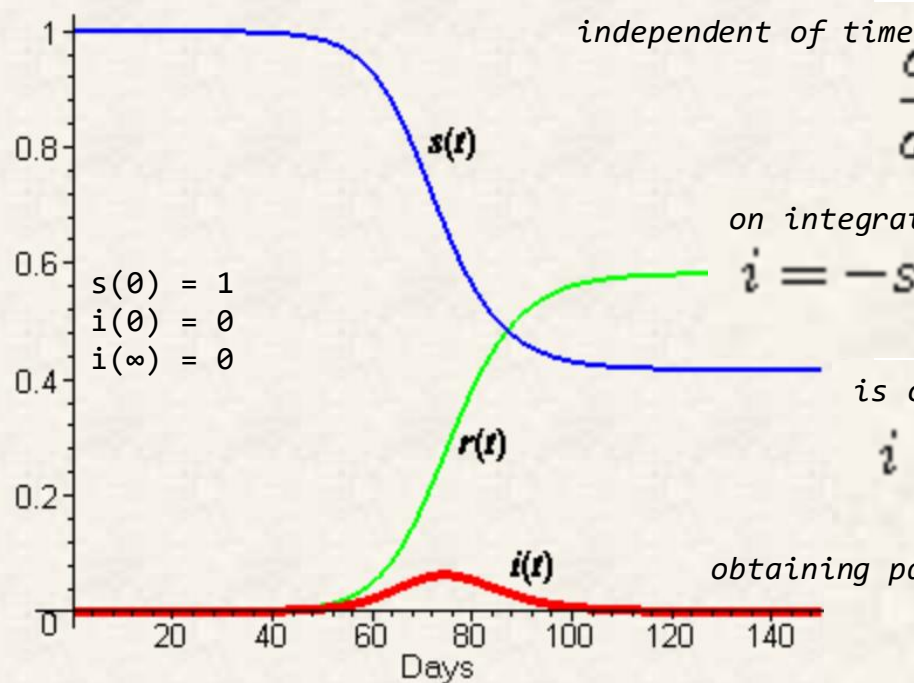
$$\frac{di}{dt} = -\frac{ds}{dt} - \frac{dr}{dt}$$

epidemic can't develop if:

$$-i'(t) < 0$$

$$-s(0) < 1/c$$

$$\frac{di}{dt} = b s(t) i(t) - k i(t) \quad i'(0) = (b s_0 - k) i_0$$



$$\frac{di}{ds} = -1 + \frac{1}{cs}$$

$$i = -s + \frac{1}{c} \ln s + q$$

$$i + s - \frac{1}{c} \ln s$$

$$c = \frac{\ln s_\infty}{s_\infty - 1}$$

what initial conditions we know.

$S(0)$ = population, everyone susceptible

$I(0)$ = some infected people

$R(0)$ = no one removed yet

$S(0)$	7,900,000
$I(0)$	10
$R(0)$	0

now we can guess the parameters

b : if an infecting contact happens once every 2 days,
 $b = 1/2$

k : if avg. period of infectiousness is 3 days D
 $k = 1/3$ (exponential decay)

contact number /infected

$$c = b.D$$

$$= b/k$$

basic reproductive number

$$R_0 = b.D$$

effective reproductive number

$$R = b.D.s(t)$$

R_0

1.4-1.6: 2009 flu H1N1, swine
1.4-2.8: 1918 flu H1N1, influenza
12.8: 1918-28 measles (US)
4.9: 1955 poliomyelitis (US)
5.7: COVID-19

For halting epidemic, $R \leq 1$

$$R = R_0.s(t)$$

$$\therefore s(t) = 1/R_0$$

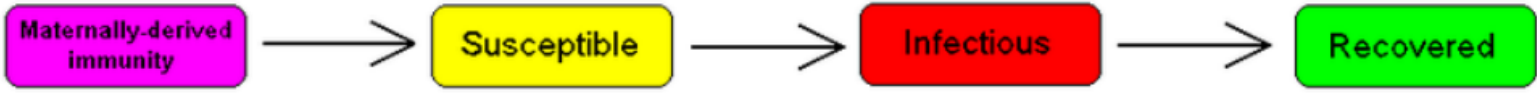
$$= 0.175$$

90% vaccine efficacy?

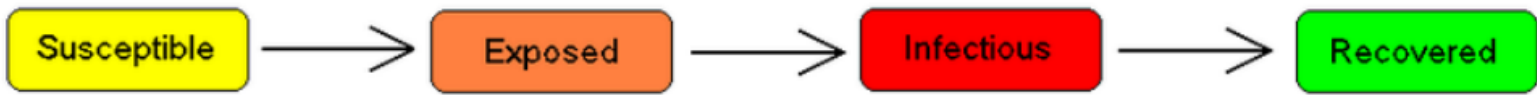
82.5% herd immunity needed

Time-Dependent R₀

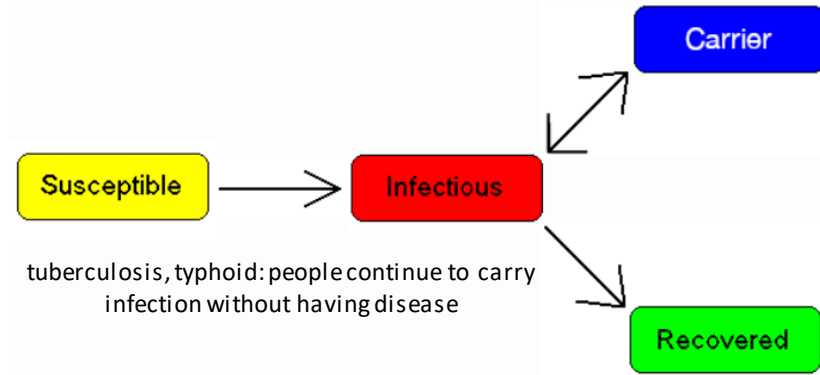
```
def R_0(t):
    return 5.0 if t < L else 0.9
```



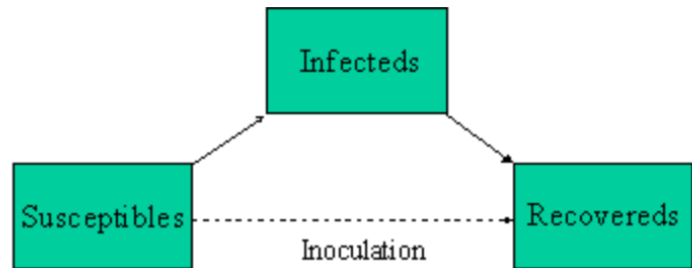
measles: babies are immune due to maternal antibodies



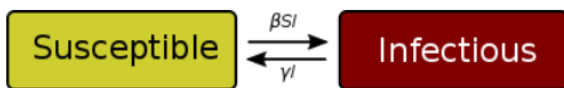
COVID-19: disease has significant incubation period



tuberculosis, typhoid: people continue to carry infection without having disease



immunization: for avoiding epidemic



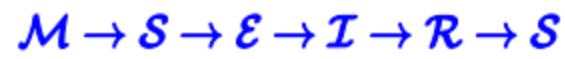
common cold, influenza: no long-lasting immunity



recovered people don't acquire immunity

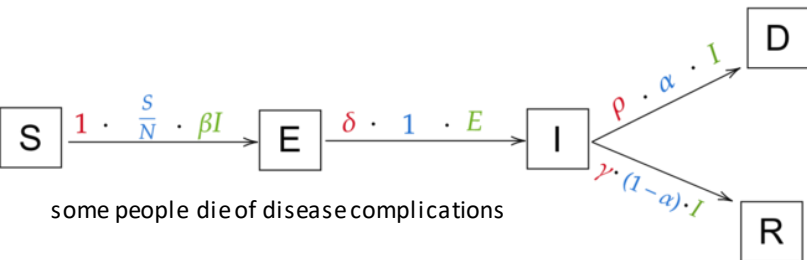
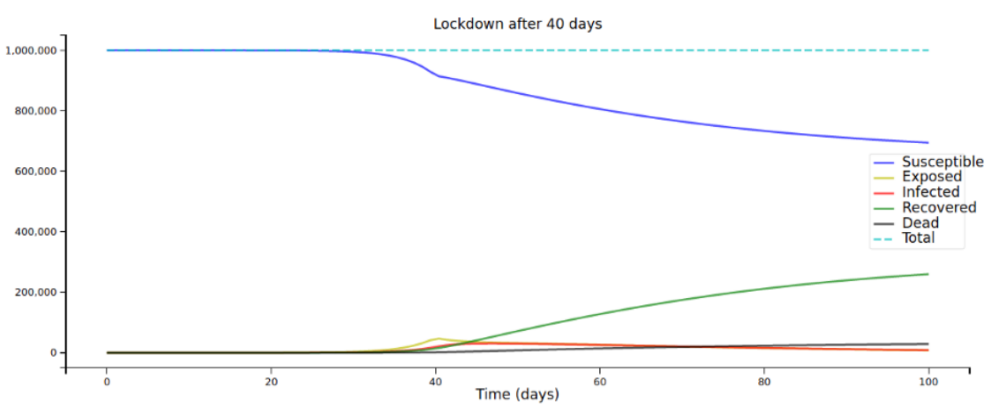
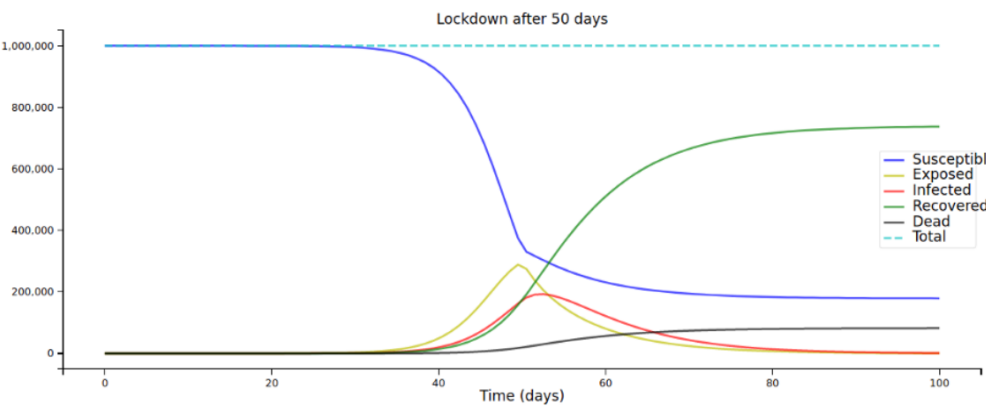
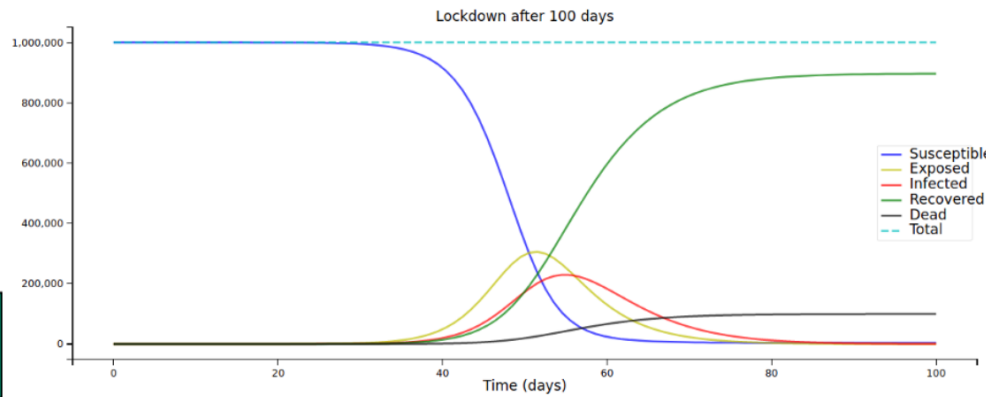


diseases with passive immunity and latency



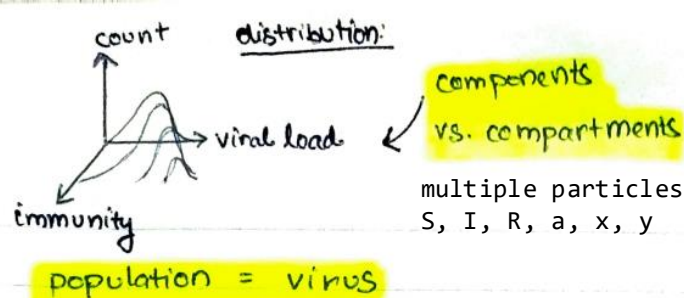
recovered people acquire temporary immunity

- resource and age dependent recovery and fatality rates
- no. of ICU beds, ventilators available
 - high risk groups, like elderly, diabetics
 - change in population structure due to fatality



some people die of disease complications

$$\begin{aligned} \frac{dS}{dt} &= -\beta \cdot I \cdot \frac{S}{N} \\ \frac{dE}{dt} &= \beta \cdot I \cdot \frac{S}{N} - \delta \cdot E \\ \frac{dI}{dt} &= \delta \cdot E - (1-\alpha) \cdot \gamma \cdot I - \alpha \cdot \rho \cdot I \\ \frac{dR}{dt} &= (1-\alpha) \cdot \gamma \cdot I \\ \frac{dD}{dt} &= \alpha \cdot \rho \cdot I \end{aligned}$$



viral load > immunity
 \rightarrow infected

viral load < I_0
 immunity > viral load
 \rightarrow recovered.

possible compartments:

damaged
 \rightarrow inactive
 within cell
 replicating proteins
 replicating capsule
 discharged

- 5 medicines affect different parts of life cycle.
- immune system effect
- medicine side effects.
- mutations (antiviral effect)
- presence of other viruses



IDM

Heterogeneous computing for epidemiological model fitting and simulation

Thomas Kovac^{1,2*}, Tom Haber^{2†}, Frank Van Reeth² and Niel Hens^{1,3}

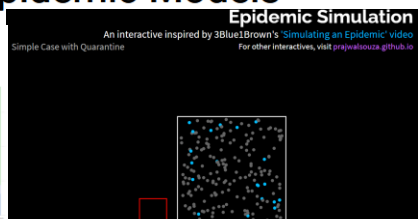
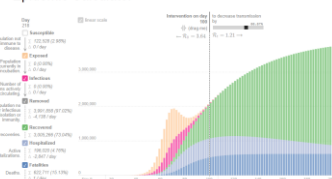
Spread of Infectious Disease Modeling and Analysis of Different Factors on Spread of Infectious Disease Based on Cellular Automata

Sheng Bin^{1,*}, Gengxin Sun¹ and Chih-Cheng Chen^{2,*}

Application Of Heterogeneous Computing Techniques To Compartmental Spatiotemporal Epidemic Models

Grant Donald Brown
 University of Iowa

Epidemic Calculator



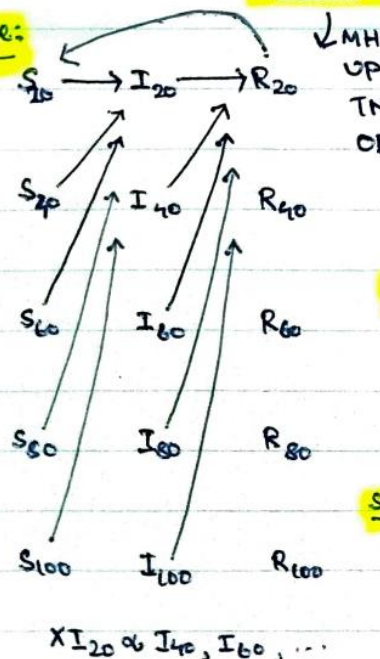
Compartments with

Other concerns:

- economics
- ethics

location?

age:



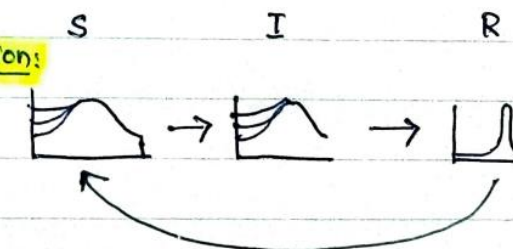
Separate compartments for hospital rooms?
 mall? road?

automatically generating model

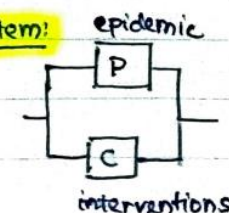
parameters / structure with state-space search (ML, GA)

finding optimal solution(s) through right interventions, by adjusting model parameters / structure.

distribution:



system:



Circuit simulation
 (reducing to simpler)

A Simulation of a COVID-19 Epidemic Based on a Deterministic SEIR Model

José M. Carcione¹, Juan E. Santos^{2,3,4}, Claudio Bagaini⁵ and Jing Ba^{2*}

Hybrid Multi-threaded Simulation of Agent-Based Pandemic Modeling using Multiple GPUs

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CAS: computer algebra system

function dependent on n variables
 $s(a, t), i(a, t), r(a, t)$
 $b(a, t), k(a, t)?$

levels of parallelism

- multiple parameters values (sir)
- separate graph nodes
- per-node distribution
- multiple time steps