

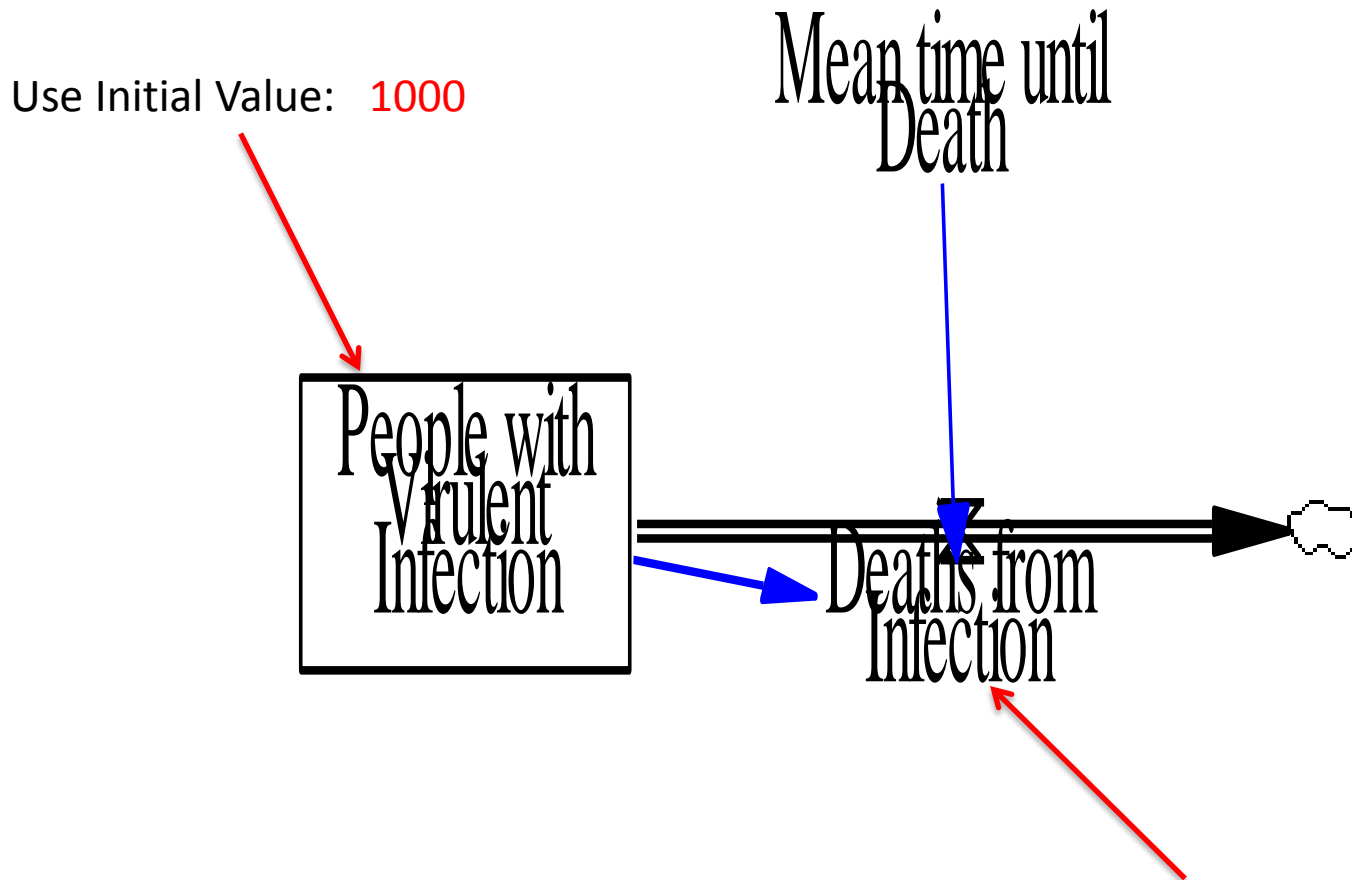
First Order Delays

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CMPT 858

2-11-2010

Simple First-Order Decay (Create this in Vensim!)

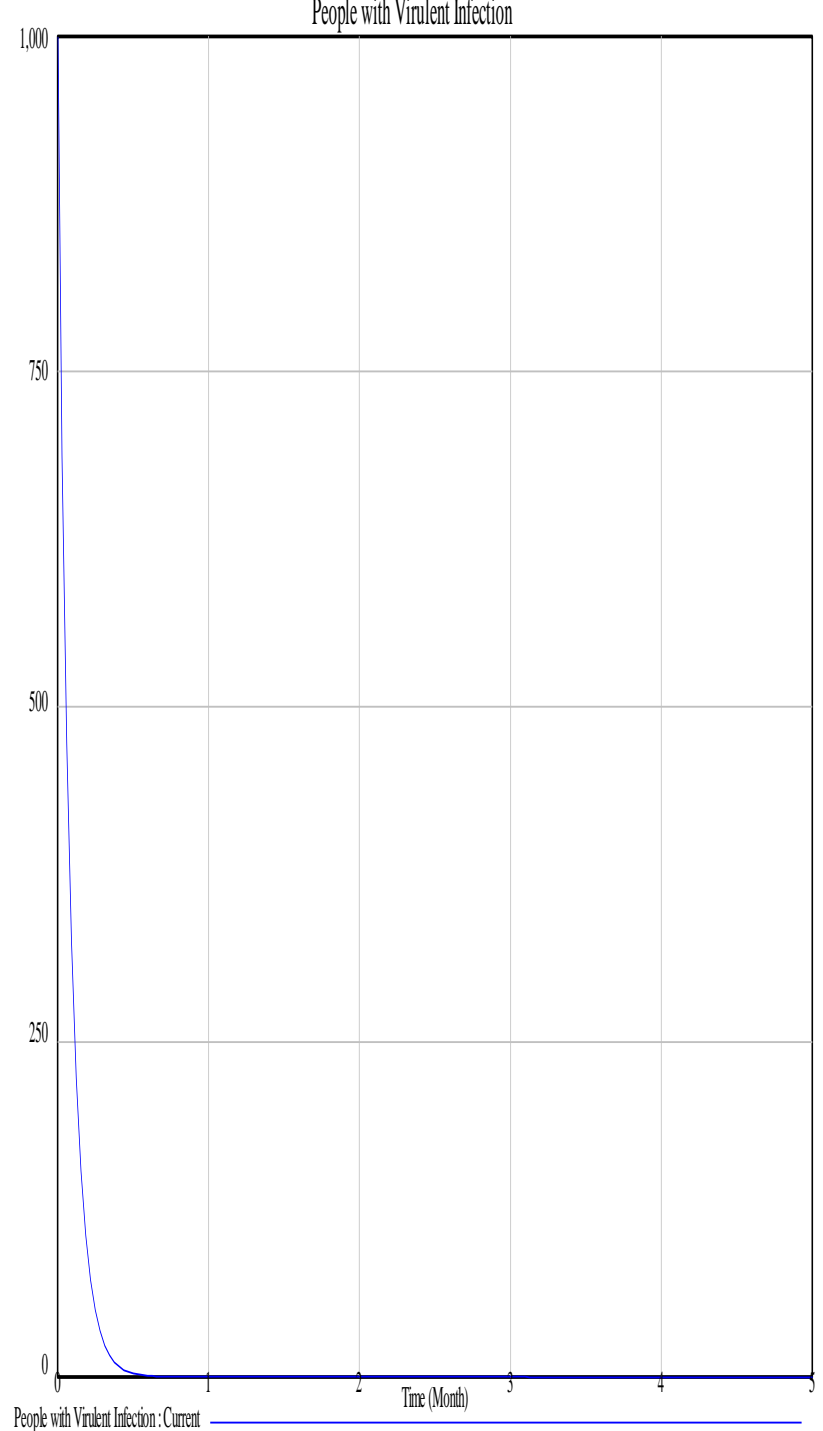


Use Formula: $\text{People with Virulent Infection} / \text{Mean time until Death}$

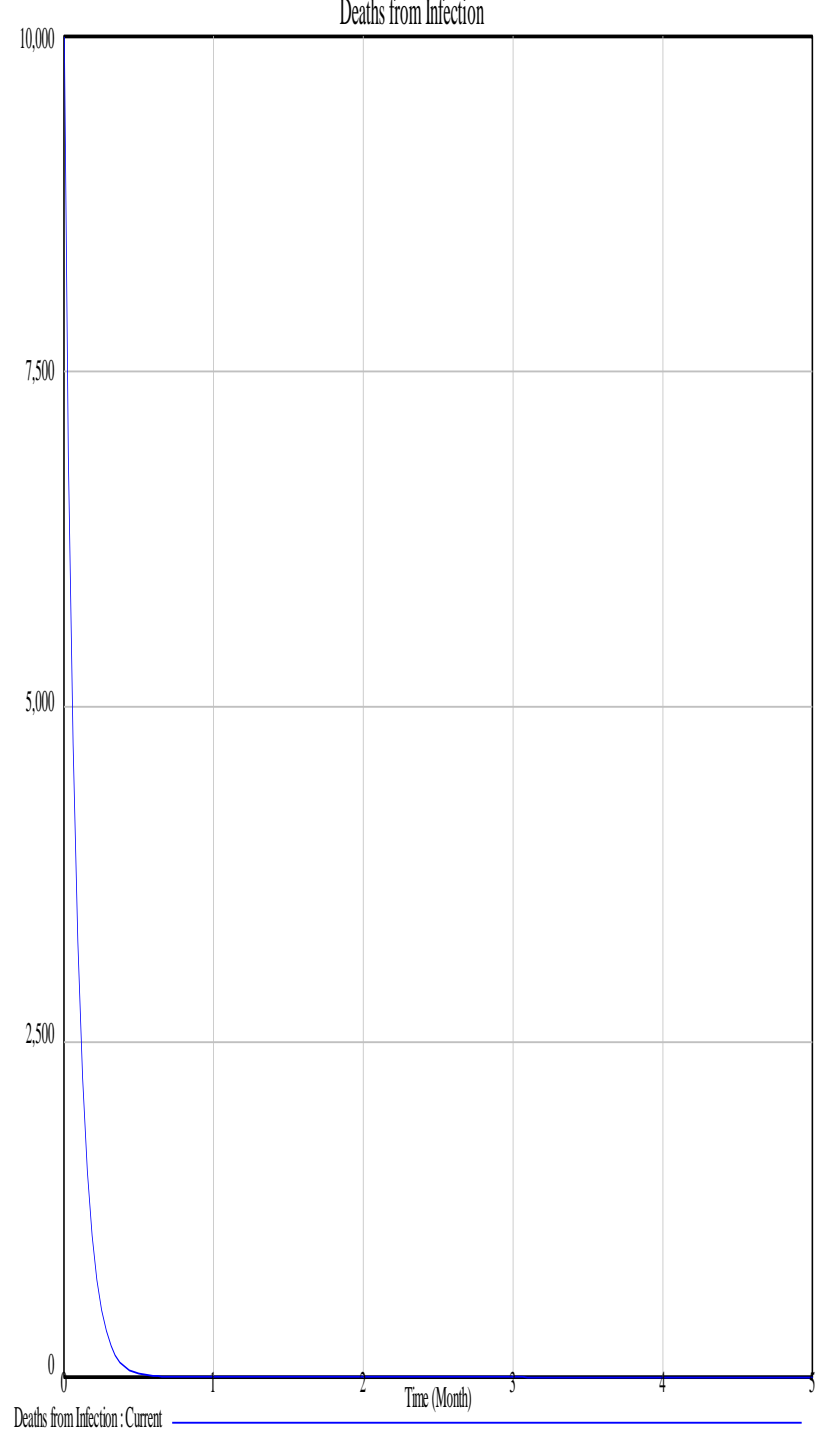
First Order Delays and Transition Processes

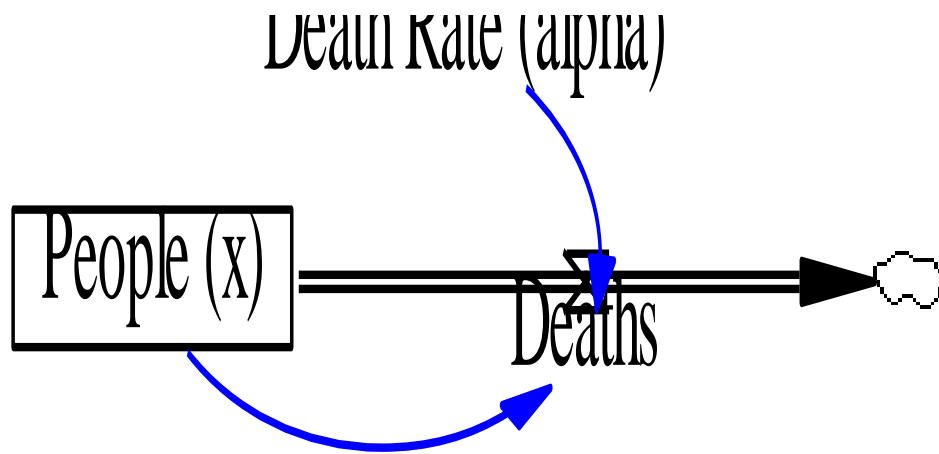
- We can think of first order delays as representing a deterministic approximation to a population experiencing a memoryless (Poisson) stochastic transition process
- The system is “memoryless” because the chance of e.g. a person leaving in the next unit of time is independent of how long they’ve been there!
- The probability distribution of residence time in the stock is exponentially distributed

Dynamics of Stock?



Dynamics of (Rate of) Death Flow?





- Alpha is per-time-unit likelihood of death
 - Chance of death over small Δt is $\alpha\Delta t$
 - If x people are at risk, # dying over Δt is $x \cdot (\text{Likelihood of death over } \Delta t) = x(\alpha\Delta t) = x\alpha\Delta t$
 - When people die, they flow out \Rightarrow cause a negative change in x .
 - We denote the change in x over the time Δt as Δx

Thus $\Delta x = -x\alpha\Delta t$

- As x is depleted (becomes smaller), Δx becomes smaller as well (for a fixed Δt)

Approximate Dynamics

Suppose

$$x(0)=1000$$

$$\Delta t=1$$

$$\alpha=.2$$

Time (t)	Stock Value (x)	Change in stock (Δx) $-x * \text{Alpha} * \text{DeltaT}$
0	1000	-200
1	800	-160
2	640	-128
3	512	-102.4
4	409.6	-81.92
5	327.68	-65.536

Flow Rate Dynamics

- The total change in x over the time Δt is Δx

Thus $\Delta x = -x\alpha\Delta t$

– This might be 10 people over a timeframe of .1 year (~36.5 days)

- The *rate of change* of x over given time Δt is $\Delta x/\Delta t$

This is just the sum of all of the flows

For system, $\Delta x/\Delta t = (-x\alpha\Delta t)/\Delta t = -x\alpha = -\text{People} * \text{DeathRate}$

Because x (People) changes, this flow rate changes over the course of the time we are observing

Suppose time is measured in years; then for our example above, $\Delta x/\Delta t = 10/.1 = 100$ people per year

Approximate Dynamics: Net Flow Rate

Reminder: Suppose

Initial $x=1000$

$\Delta t=1$

$\alpha=.2$

Time (t)	Stock Value (x)	Change in stock (Δx) $-x * \text{Alpha} * \text{DeltaT}$	Net Flow Rate= $\Delta x / \Delta t$ Here, $\Delta t=1$, so $\Delta x / \Delta t = \Delta x / 1 = \Delta x$
0	1000	-200	-200
1	800	-160	-160
2	640	-128	-128
3	512	-102.4	-102.4
4	409.6	-81.92	-81.92
5	327.68	-65.536	-65.536

Why is This Approximate?

- Our previous graphs used a value of $\Delta t=1$
- In calculating the change (Δx) from t to $t+\Delta t$ (here, $t+1$), we are assuming that the flow rate (people/year) *stays constant in that time*
 - Recall: In general, this flow rate will be determined by the value of stocks
 - So in assuming that the flow rate remains constant, we were basically assuming that the values of the stocks stay constant over time Δt
 - For our system, given that the value of the stock x (People) declines by around 20% per time unit, this is not a very good assumption!

How Can We Reduce the Error?

Try a Smaller Δt

- Let's work forward for $\frac{1}{2}$ of a year at a time instead of for a full year

$$x(0)=1000$$

$$\Delta t=.5$$

$$\alpha=.1$$

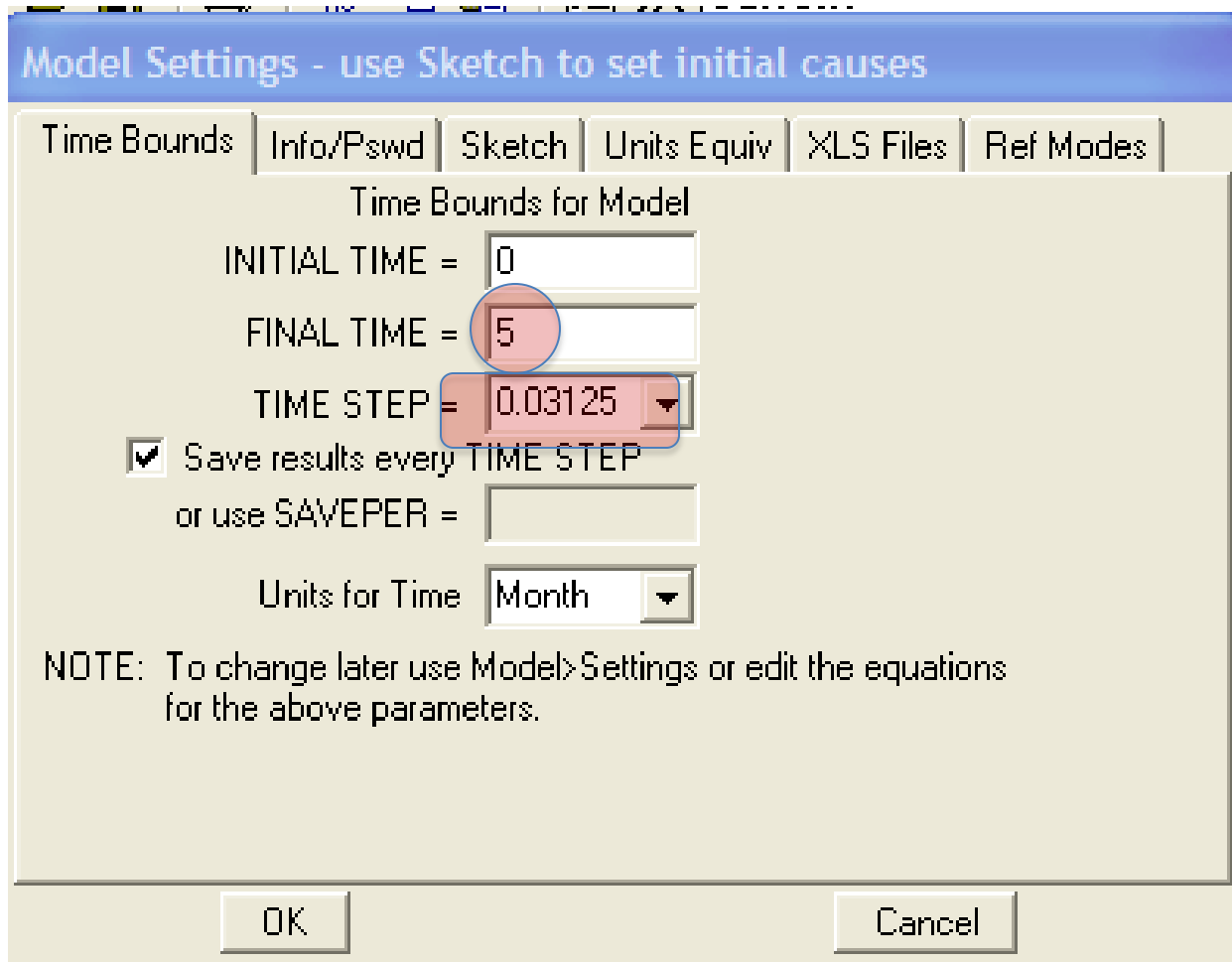
Time (t)	Stock Value (x)	Change in stock (Δx) $-x*\text{Alpha}*\text{DeltaT}$	Net Flow Rate= $\Delta x/\Delta t$ Here, $\Delta t=1$, so $\Delta x/\Delta t=\Delta x/1=\Delta x$
0	1000	-100	-200
0.5	900	-90	-180
1	810	-81	-162
1.5	729	-72.9	-145.8
2	656.1	-65.6	-131.2
2.5	590.5	-59.0	-118.1
3	531.4	-53.1	-106.3
3.5	478.3	-47.8	-95.7
4	430.5	-43.0	-86.1
4.5	387.4	-38.7	-77.5
5	348.7	-34.9	-69.7

Approximate Dynamics: Net Flow Rate

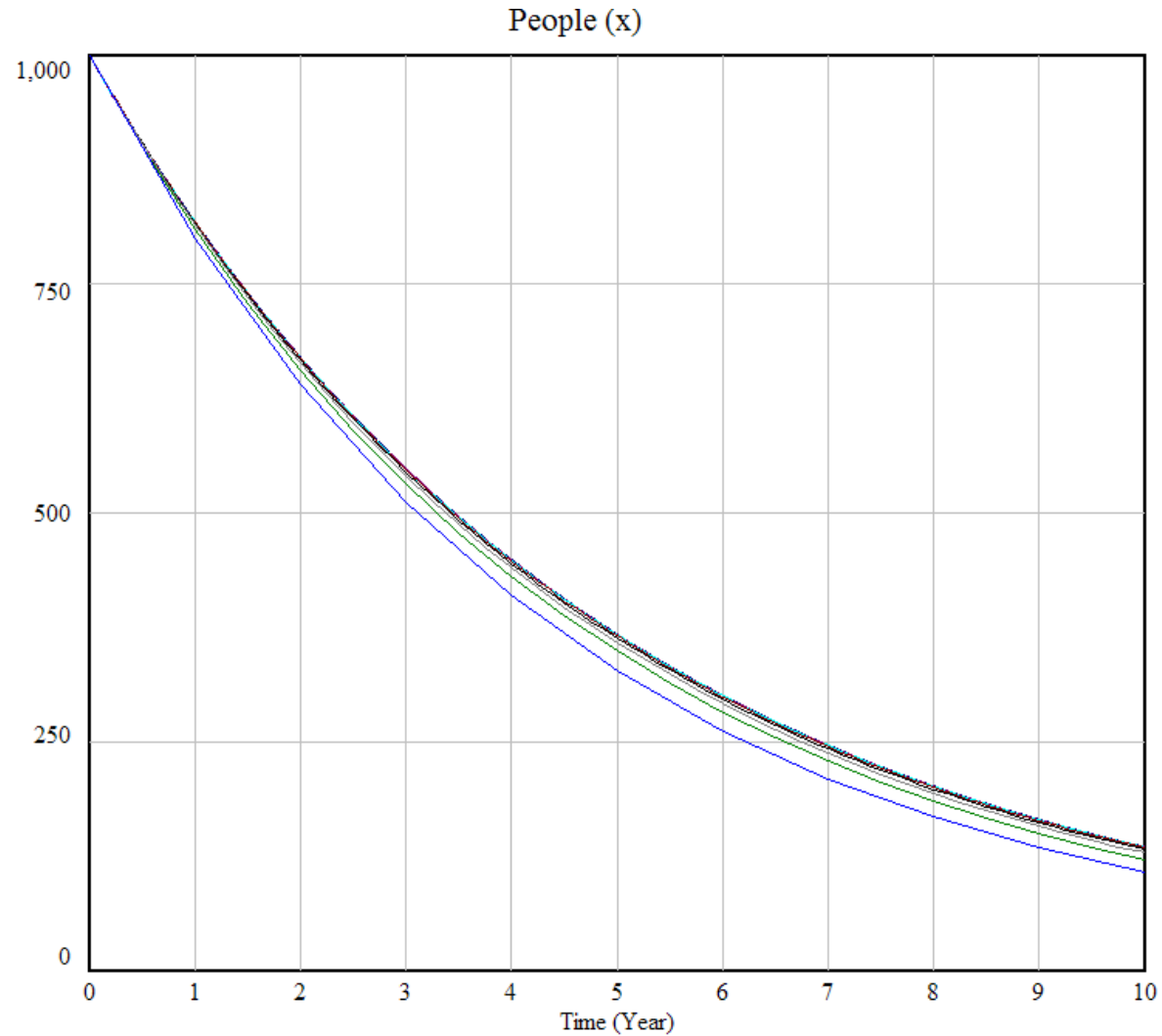
	$\Delta t=1$	$\Delta t=.5$	$\Delta t=.25$
Time (t)	Stock Value (x)	Stock Value (x)	Stock Value (x)
0	1000	1000	1000
1	800	810	814.5
2	640	656.1	663.4
3	512	531.4	540.4
4	409.6	430.5	440.1
5	327.68	348.7	358.5

Vensim has a Step Size!

(Set via Model Menu/Settings Item)

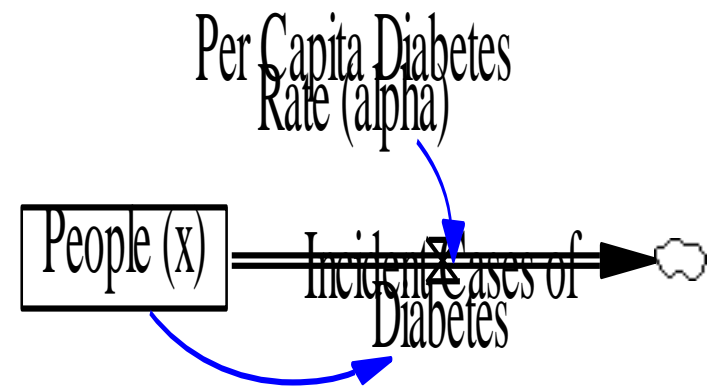


Impact of Step Size on Simulation



"People (x)" : First order decay Timestep = 1
"People (x)" : First order decay Timestep = pt1
"People (x)" : First order decay Timestep = pt5
"People (x)" : First order decay Timestep = pt25
"People (x)" : First order decay Timestep = pt125
"People (x)" : First order decay Timestep = pt0625
"People (x)" : First order decay Timestep = pt03125
"People (x)" : First order decay Timestep = pt015625
"People (x)" : First order decay Timestep = pt007812

Continuous Mathematics (Calculus!)



- Alpha is per-time-unit likelihood of death
 - Chance of death over small dt is αdt
 - If x people are at risk, # dying over dt is $x \cdot (\text{Likelihood of death over } \Delta t) = x(\alpha dt) = x\alpha dt$
 - When people die, they flow out \Rightarrow cause a negative change in x .
 - We denote the change in x over the time dt as Δx
- Thus $dx = -x\alpha dt$
- As x is depleted (becomes smaller), dx becomes smaller as well (for a fixed dt)

Flow Rate Dynamics: Continuous

- The total change in x over the time dt is dx

Thus $dx = -x\alpha dt$

– This might be 10 people over a timeframe of .1 year (~36.5 days)

- The *rate of change* of x over given time dt is dx/dt

This is just the sum of all of the flows!

For system, $dx/dt = (-x\alpha dt)/dt = -\alpha x = -\text{People} * \text{DeathRate}$

Because x (People) changes, this flow rate changes over the course of the time we are observing

- We will sometimes write dx/dt as \dot{x} $\frac{dx}{dt} = \dot{x} = -\alpha x$

The Concept of “Analytic” Solutions

- The model structure describes system behaviour *implicitly*
 - *This indicates how short term changes (flows) depends on the state of the system*
 - *This does not explicitly state how the system evolves*
- Analytic (“closed form”, “exact”) solutions describe system behaviour as an *explicit function of time*
 - *E.g. $a+b*t+c*t^2$, $a +b*t$, $a*\sin(t)$, $e^{\alpha t}$*
- For many systems we will be dealing with (nonlinear systems), an analytic solution *is simply not derivable*
 - Even when an analytic solution is possible, it is often most convenient to deal with simulations for most needs

An Exact Solution to Our Problem

- The state equation formulation of our system

is

$$\frac{dx}{dt} = \dot{x} = -\alpha x$$

This is a linear differential equation with constant coefficients – a type of system that can be solved exactly.

Solution Procedure

$$\frac{dx}{dt} = -\alpha x$$

- Suppose we start x at time 0 with initial value $x(0)$, and we want to find the value of x at time T
- Assuming that x does not start at 0, it will never reach exactly 0, so we can divide the left side by it, and multiply the right side by dt

$$\frac{dx}{x} = -\alpha dt$$

- Integrating both sides

$$\int_{t=0}^{t=T} \frac{dx}{x} = \int_{t=0}^{t=T} -\alpha dt$$

Completion of Derivation

$$\int_{t=0}^{t=T} \frac{dx}{x} = \int_{t=0}^{t=T} -\alpha dt = -\alpha \int_{t=0}^{t=T} dt$$

$$\ln x \Big|_{t=0}^{t=T} = -\alpha t \Big|_{t=0}^{t=T}$$

$$\ln x(T) - \ln x(0) = -\alpha T$$

$$\ln x(T) = \ln x(0) - \alpha T$$

$$x(T) = e^{\ln x(0) - \alpha T} = e^{\ln x(0)} e^{-\alpha T} = x(0) e^{-\alpha T}$$

So the stock x declines as a negative exponential in time T
i.e. # of people remaining in the stock goes down exponentially w/time

Fraction of Original People Still in Stock or Who have Left

- Assuming no inflows, the fraction of people still in the stock at time T is just

(# of people in the stock at time T)/(initial # of people in the stock)=

$$\frac{x(T)}{x(0)} = \frac{x(0)e^{-\alpha T}}{x(0)} = e^{-\alpha T}$$

- Given that people either stay in the stock or leave, the fraction that have left by time T =

$$1 - \frac{x(T)}{x(0)} = 1 - e^{-\alpha T}$$

At Time=1

- At time $t=1$, we have a fraction $e^{-\alpha \cdot 1} = e^{-\alpha}$ in the stock, and a fraction $1 - e^{-\alpha}$ who have left

- Note: By its Taylor Expansion

$$e^{-\alpha t} = \sum_{i=0}^{\infty} \frac{(-\alpha t)^i}{i!} = 1 + (-\alpha t) + \frac{(-\alpha t)^2}{2!} + \frac{(-\alpha t)^3}{3!} + \dots$$

$$= 1 - \alpha t + \frac{(\alpha t)^2}{2} + \dots$$

- For small αt , the higher order terms are very small, and this will be approximately $1 - \alpha t$
- So by time 1 for small α , approx $1 - \alpha$ will remain after, and a fraction of α will have departed

Mean Time to Transition

- People are leaving via the flow
- Suppose we wish to determine the mean (average) time for a given person in the stock to leave
- Recall: A mean for a continuous probability distribution $p(t)$ is given by $\int_{t=-\infty}^{\infty} tp(t)dt$
- Since $p(t)dt$ is the probability that will leave between t and $t+dt$, this is just the continuous version of

$$E[q(a)] = \sum_{a \in \{\text{Possible values of } a\}} aq(a)$$

Mean Time to Leave

- $p(t)dt$ here is the likelihood of a person leaving exactly between time t & $dt+t$
 - We start the simulation at $t=0$, so $p(t)=0$ for $t<0$
 - For $t>0$, $P(\text{leaving exactly between time } t \text{ and } dt+t) = P(\text{leaving exactly between time } t \text{ and } t+dt \mid \text{Still have not left by time } t)P(\text{Still have not left by time } t)$

For $T>0$, $P(\text{Still have not left by time } t) = e^{-\alpha T}$

For $P(\text{leaving exactly between time } t \text{ and } t+dt \mid \text{Still have not left by time } t)$

Recall: For us, probability of leaving in a time dt always $=\alpha dt$

Thus $P(\text{leaving exactly between time } t \text{ and } t+dt \mid \text{Still have not left by time } t) = \alpha dt$

$P(t)dt = P(\text{leaving exact b.t. time } t \text{ & } dt+t) = (e^{-\alpha T})(\alpha dt) = \alpha e^{-\alpha T} dt$

Derivation of Mean

- $P(t)dt = P(\text{leaving exactly between time } t \text{ \& } dt+t) = (e^{-\alpha T})(\alpha dt) = \alpha e^{-\alpha T} dt$
- Now that we have found the function $p(t)$, we must do the integral $\int_{t=-\infty}^{t=\infty} tp(t)dt$ to derive the mean
- Here $E[p(t)] = \int_{t=-\infty}^{t=\infty} tp(t)dt = \int_{t=0}^{t=\infty} tp(t)dt = \int_{t=0}^{t=\infty} t\alpha e^{-\alpha T} dt = \alpha \int_{t=0}^{t=\infty} te^{-\alpha T} dt$

Recall: Integration by Parts

- We have $E[p(t)] = \alpha \int_{t=0}^{t=\infty} te^{-\alpha T} dt = \alpha \left(\int_{t=0}^{t=\infty} te^{-\alpha T} dt \right)$
- To solve the term in brackets, we will use integration by parts
- Integration by parts exploits the following/

$$\frac{d(uv)}{dt} = u \frac{dv}{dt} + v \frac{du}{dt}$$

$$d(uv) = u dv + v du$$

$$\int d(uv) = \int u dv + \int v du$$

$$uv = \int u dv + \int v du$$

and thus

$$\int u dv = uv - \int v du$$

Recall: Integration by Parts

- To solve $\int_{t=0}^{t=\infty} te^{-\alpha T} dt$ we will use integration by parts

$$u = t \Rightarrow du = \frac{du}{dt} dt = 1 dt = dt$$

- Here

$$dv = e^{-\alpha T} dt \Rightarrow v = \int e^{-\alpha T} dt = \frac{-e^{-\alpha T}}{\alpha}$$

- From the previous page, we know

$$\int_{t=0}^{t=\infty} te^{-\alpha T} dt = \int_{t=0}^{t=\infty} u dv = uv - \int_{t=0}^{t=\infty} v du = \left(t \frac{-e^{-\alpha T}}{\alpha} \right) \Big|_{t=0}^{t=\infty} - \int_{t=0}^{t=\infty} \left(\frac{-e^{-\alpha T}}{\alpha} \right) dt$$

$$= \left(\frac{-te^{-\alpha T}}{\alpha} \right) \Big|_{t=0}^{t=\infty} + \frac{1}{\alpha} \int_{t=0}^{t=\infty} e^{-\alpha T} dt = (0 - 0) + \frac{1}{\alpha} \left(\frac{-1}{\alpha} e^{-\alpha T} \right) \Big|_{t=0}^{t=\infty} =$$

$$\frac{1}{\alpha} \left(0 - \frac{-1}{\alpha} \right) = \frac{1}{\alpha^2}$$

Thus

- The mean time (the *delay associated with a first order delay*) is thus given by

$$E[p(t)] = \alpha \int_{t=0}^{t=\infty} t e^{-\alpha T} dt = \alpha \left(\int_{t=0}^{t=\infty} t e^{-\alpha T} dt \right)$$
$$= \alpha \left(\frac{1}{\alpha^2} \right) = \frac{1}{\alpha}$$

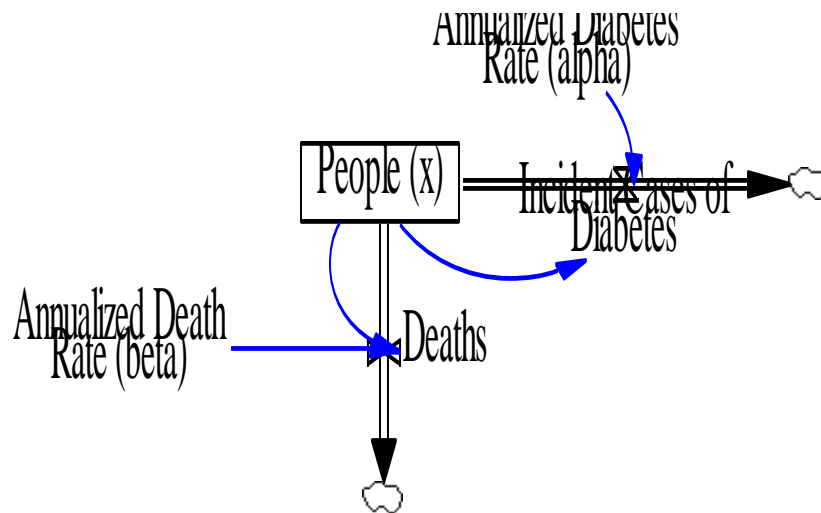
- So e.g. if we have an annualized rate of diabetes incident, the mean time to develop diabetes (independent of other risks) is just the reciprocal of that rate (i.e. 1 over that rate)

Computer Exercise: Simulating a First Order Delay

- Create a first order delay
- Feed in a “step function” that rises suddenly at time 10.
- How does the output from the stock change over time?

Competing Risks

- Suppose we have another outflow from the stock. How does that change our mean time of proceeding specifically down flow 1 (here, developing diabetes)?



Competing Risks Stock Trajectory

Solution Procedure

$$\frac{dx}{dt} = -\alpha x - \beta x = -(\alpha + \beta)x$$

- Suppose we start x at time 0 with initial value $x(0)$, and we want to find the value of x at time T
- This is just like our previous differential equation, except that “ α ” has been replaced by “ $(\alpha + \beta)$ ”
 - The solution must therefore be the same as before, with the appropriate replacement
 - Thus

$$x(T) = x(0)e^{-(\alpha + \beta)T}$$

Mean Time to Leave: Competing Risks

- $p(t)dt$ here is the likelihood of a person leaving *via flow 1* (e.g. developing T2DM) exactly between time t & $t+dt$
 - We start the simulation at $t=0$, so $p(t)=0$ for $t<0$
 - For $t>0$, $P(\text{leaving on flow 1 exactly between time } t \text{ \& } t+dt) = P(\text{leaving on flow 1 exactly between time } t \text{ \& } t+dt \mid \text{Still have not left by time } t) P(\text{Still have not left by time } t)$

For $T>0$, $P(\text{Still have not left by time } T) = e^{-(\alpha+\beta)T}$

For $P(\text{leaving exactly between time } t \text{ and } t+dt \mid \text{Still have not left by time } t)$

Recall: For us, probability of leaving in a time dt always $= \alpha dt$

Thus $P(\text{leaving exactly between time } t \text{ and } t+dt \mid \text{Still have not left by time } t) = \alpha dt$

$$P(t)dt = P(\text{leaving exact b.t. time } t \text{ \& } t+dt) = \alpha e^{-(\alpha+\beta)T} dt$$

Mean Time to Transition via Flow 1: Competing Risks

- By the same procedure as before, we have

$$E[p(t)] = \alpha \int_{t=0}^{t=\infty} t e^{-(\alpha+\beta)T} dt$$

- Using the formula we derived for the integral expression, we have

$$E[p(t)] = \frac{\alpha}{(\alpha + \beta)^2}$$

- Note that this correctly approaches the single-flow case as $\beta \rightarrow 0$

Equilibrium Value of a First-Order Delay

- Suppose we have flow of rate i into a stock with a first-order delay out
 - This could be from just a single flow, or many flows
- The value of the stock will approach an equilibrium where inflow=outflow

Equilibrium Value of 1st Order Delay

- Recall: Outflow rate for 1st order delay= αx
 - Note that this depends on the value of the stock!
- Inflow rate= i
- At equilibrium, the level of the stock must be such that inflow=outflow
 - For our case, we have

$$\alpha x = i$$

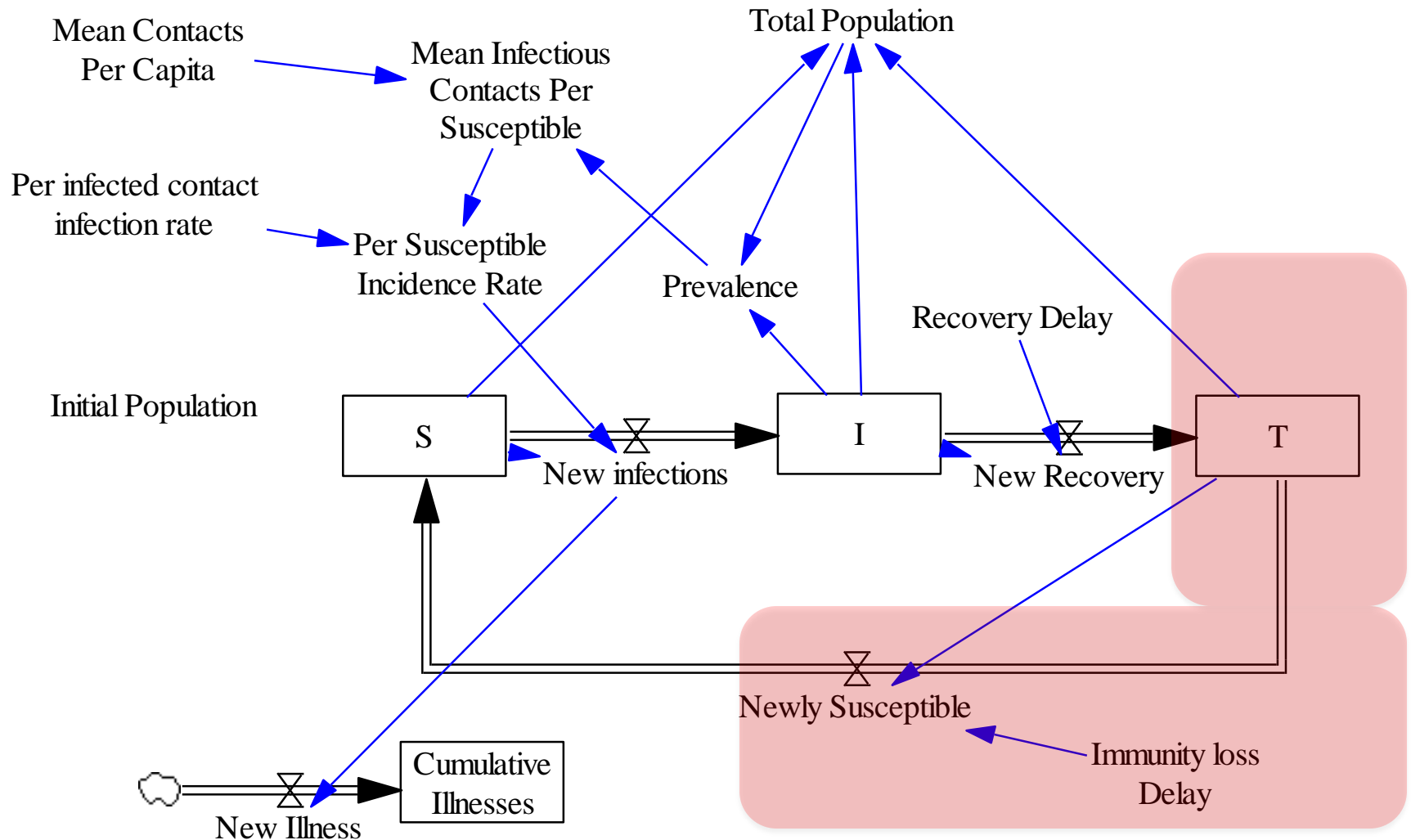
$$\text{Thus } x = i/\alpha$$

The lower the chance of leaving per time unit (or the longer the delay), the larger the equilibrium value of the stock must be to make outflow=inflow

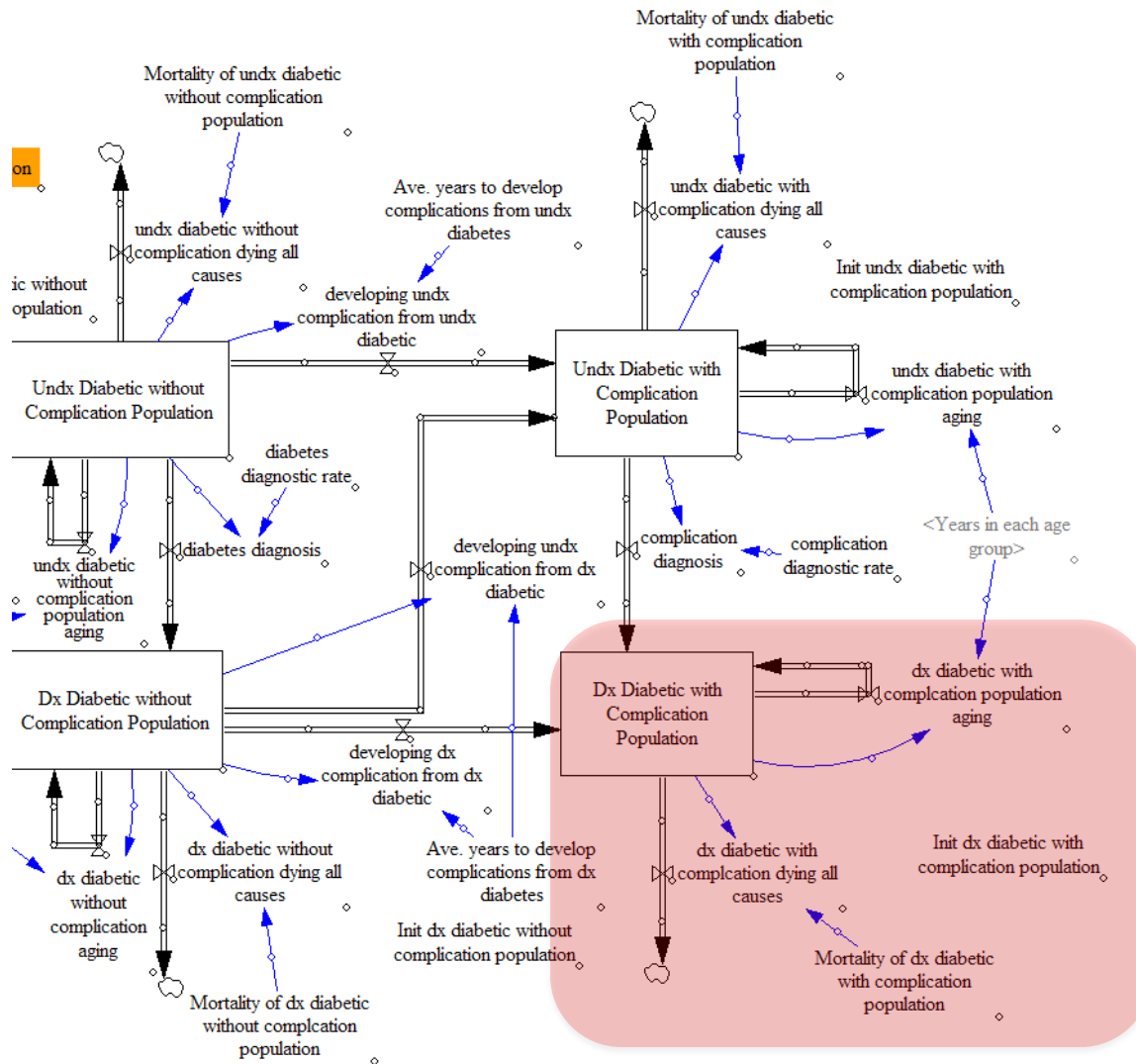
Computer Exercise: Simulating a First Order Delay

- Create a first order delay
- Feed in a “step function” that rises suddenly from 0 to 20 at time 10
 - Use formula if then else($\text{Time} > 10, 20, 0$)
- Questions to ponder
 - How does the output from the stock change over time?
 - How does the equilibrium value of the stock vary with chance of proceeding (alpha)?

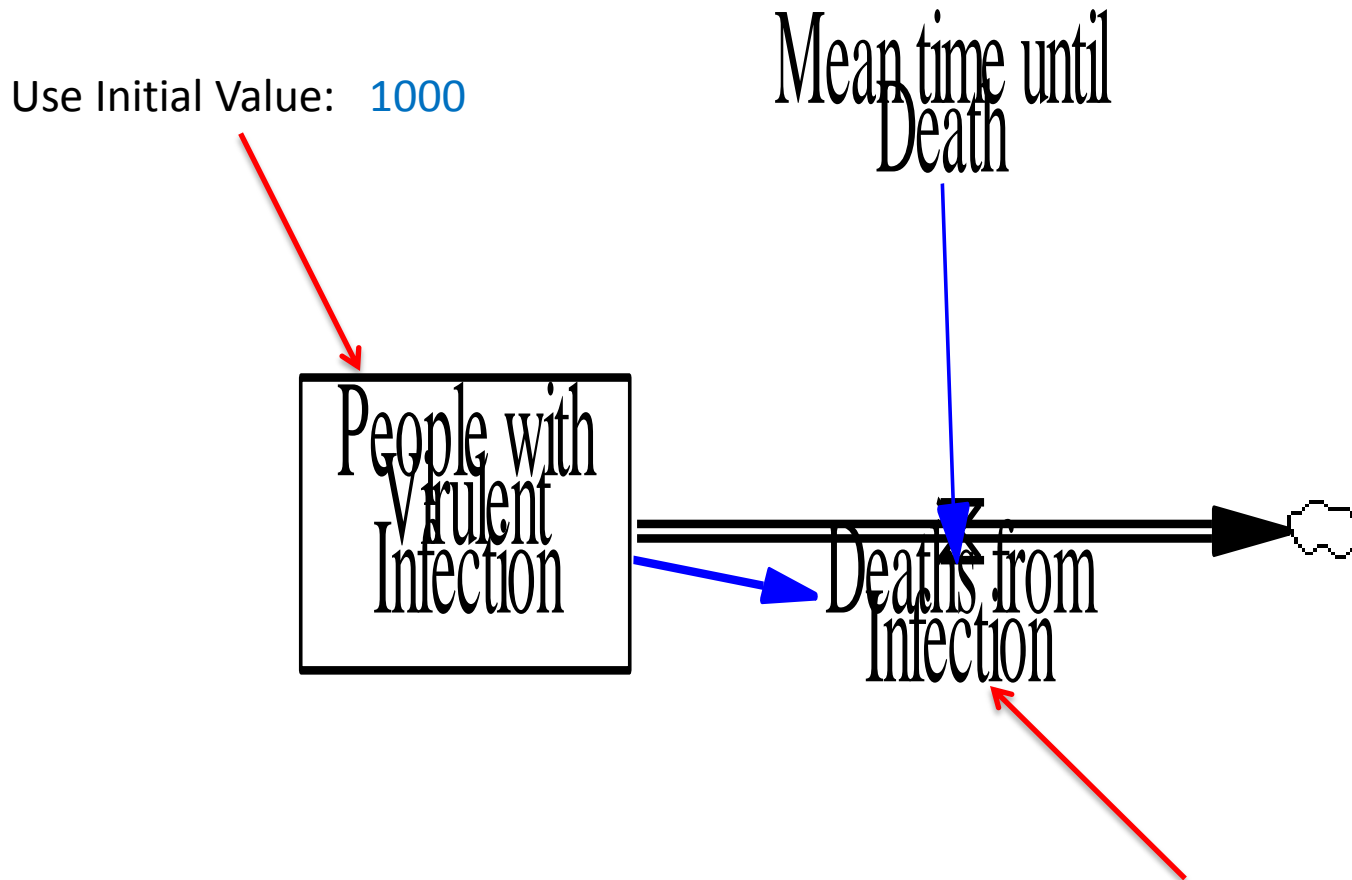
First Order Delays in Action: Simple SIT Model



First Order Delays in Action: Simple SIT Model



Recall: Simple First-Order Decay



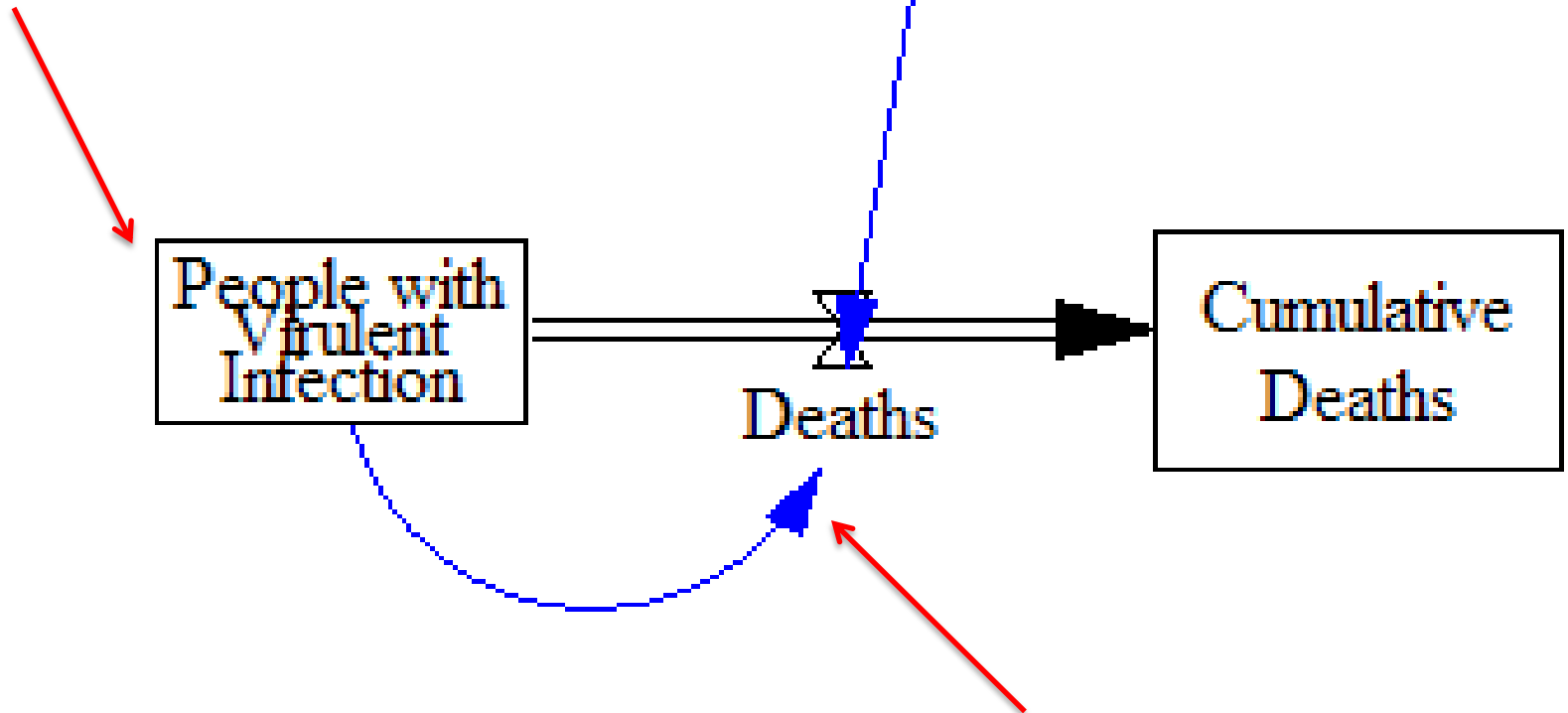
Use Formula: $\text{People with Virulent Infection} / \text{Mean time until Death}$

First-Order Decay (Variant of Last Time)

Recall: How does this relate to the mean time until death?

Use Initial Value: 1000

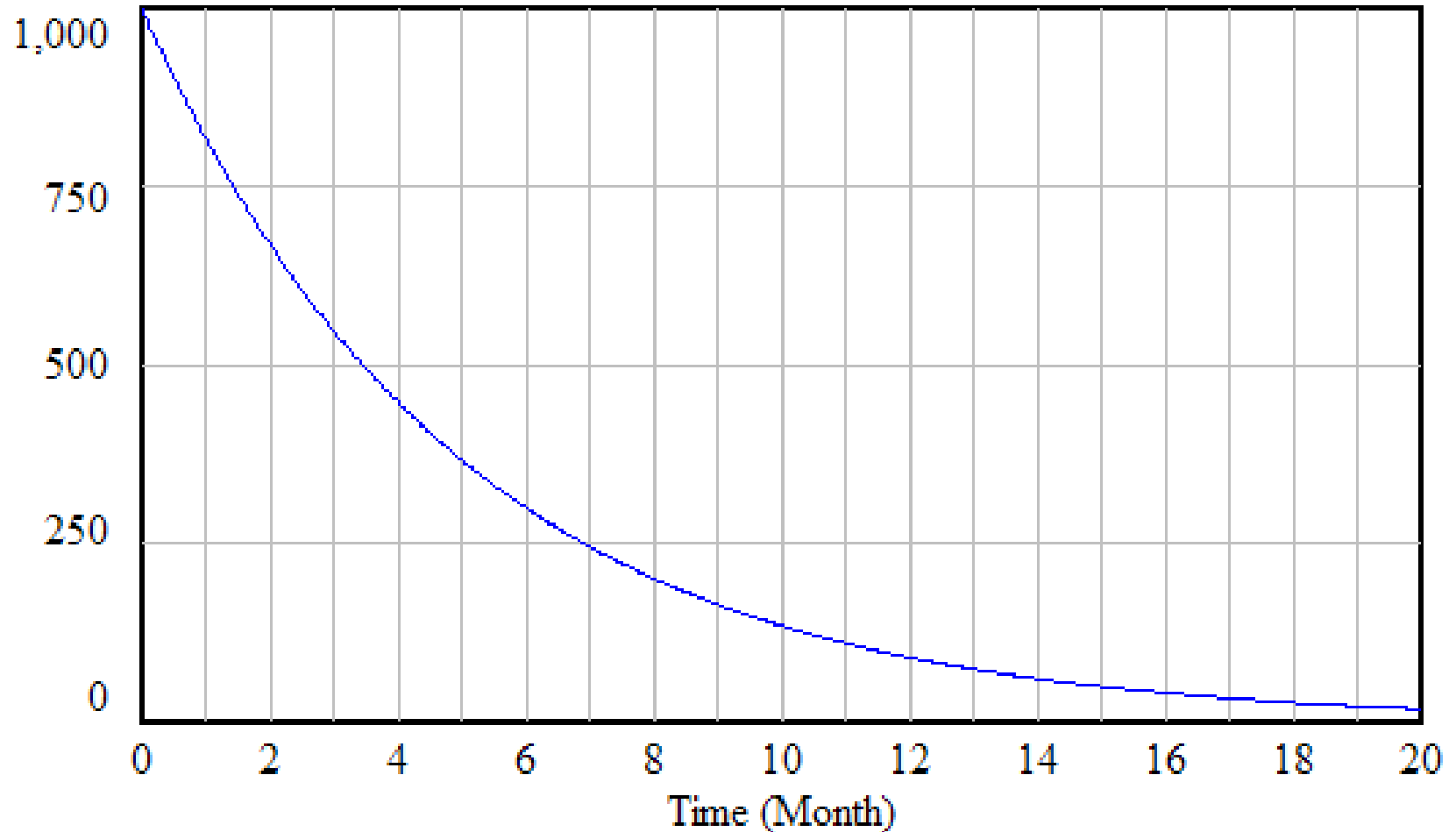
Per Month Use Value: 0.2
Likelihood of Death



Use Formula: $\text{People with Virulent Infection} * \text{Per Month Likelihood of Death}$

People in Stock

People with Virulent Infection

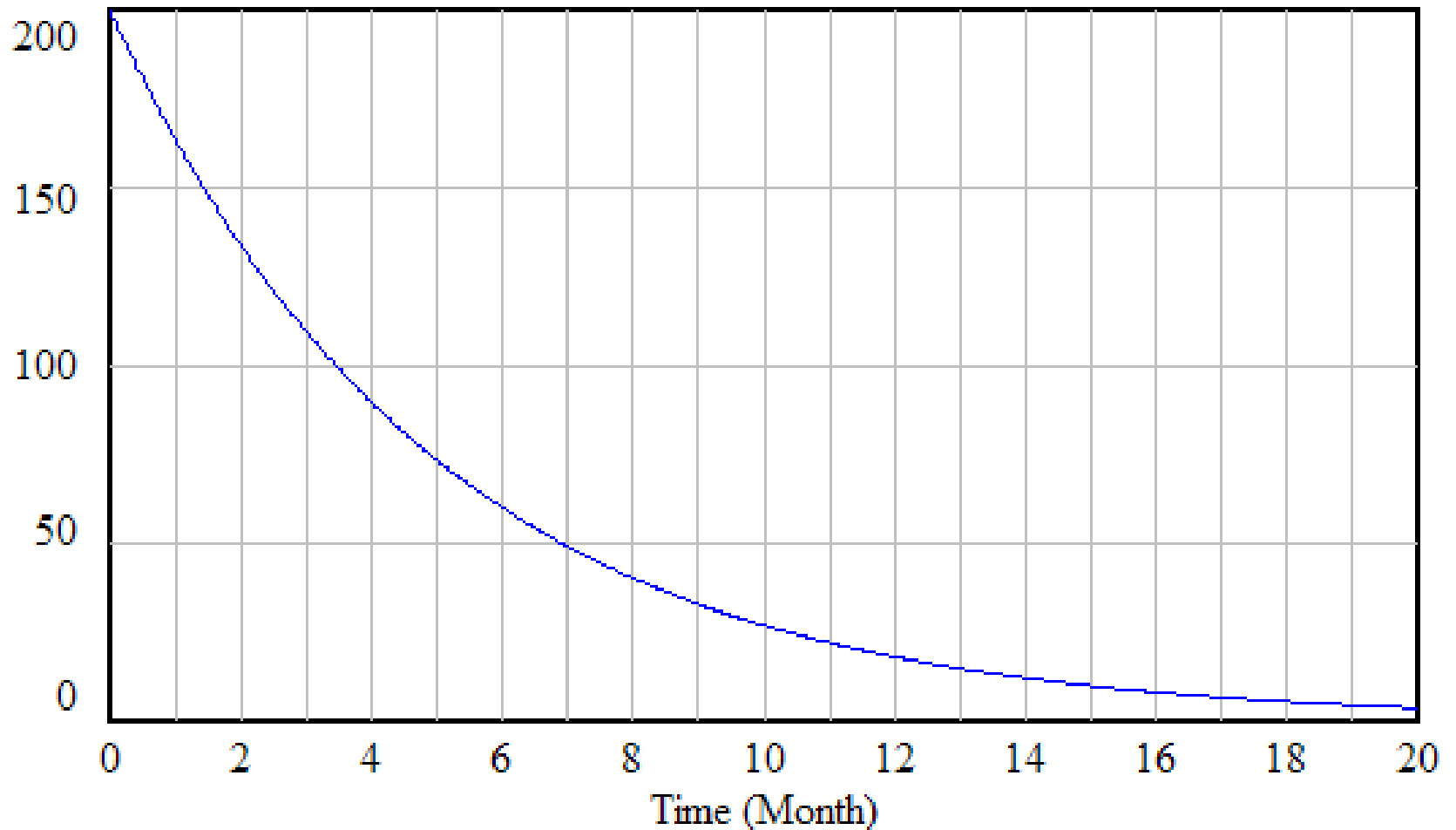


People with Virulent Infection : Baseline



Flow Rate of Deaths

Deaths

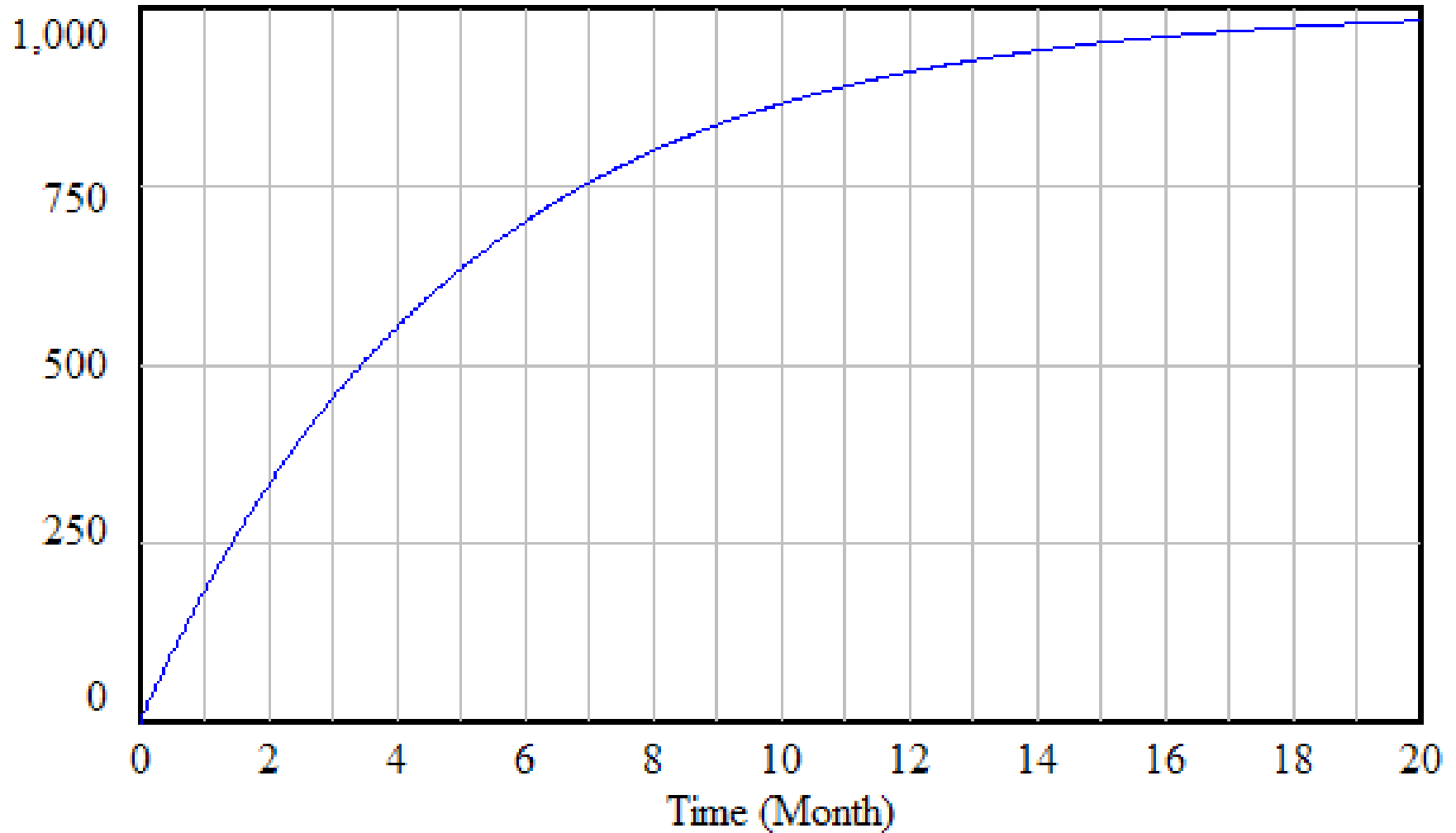


Deaths : Baseline



Cumulative Deaths

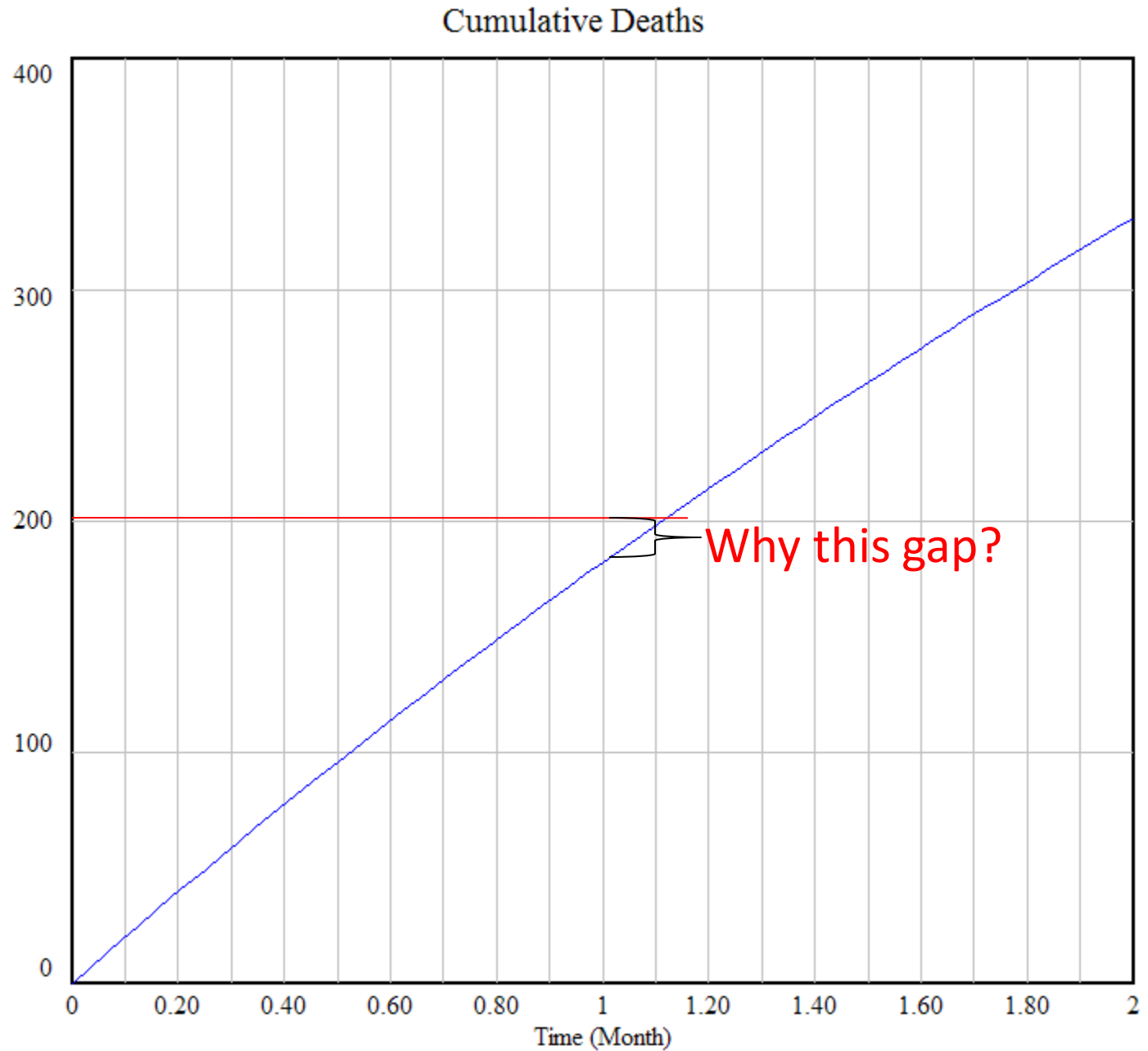
Cumulative Deaths



Cumulative Deaths : Baseline

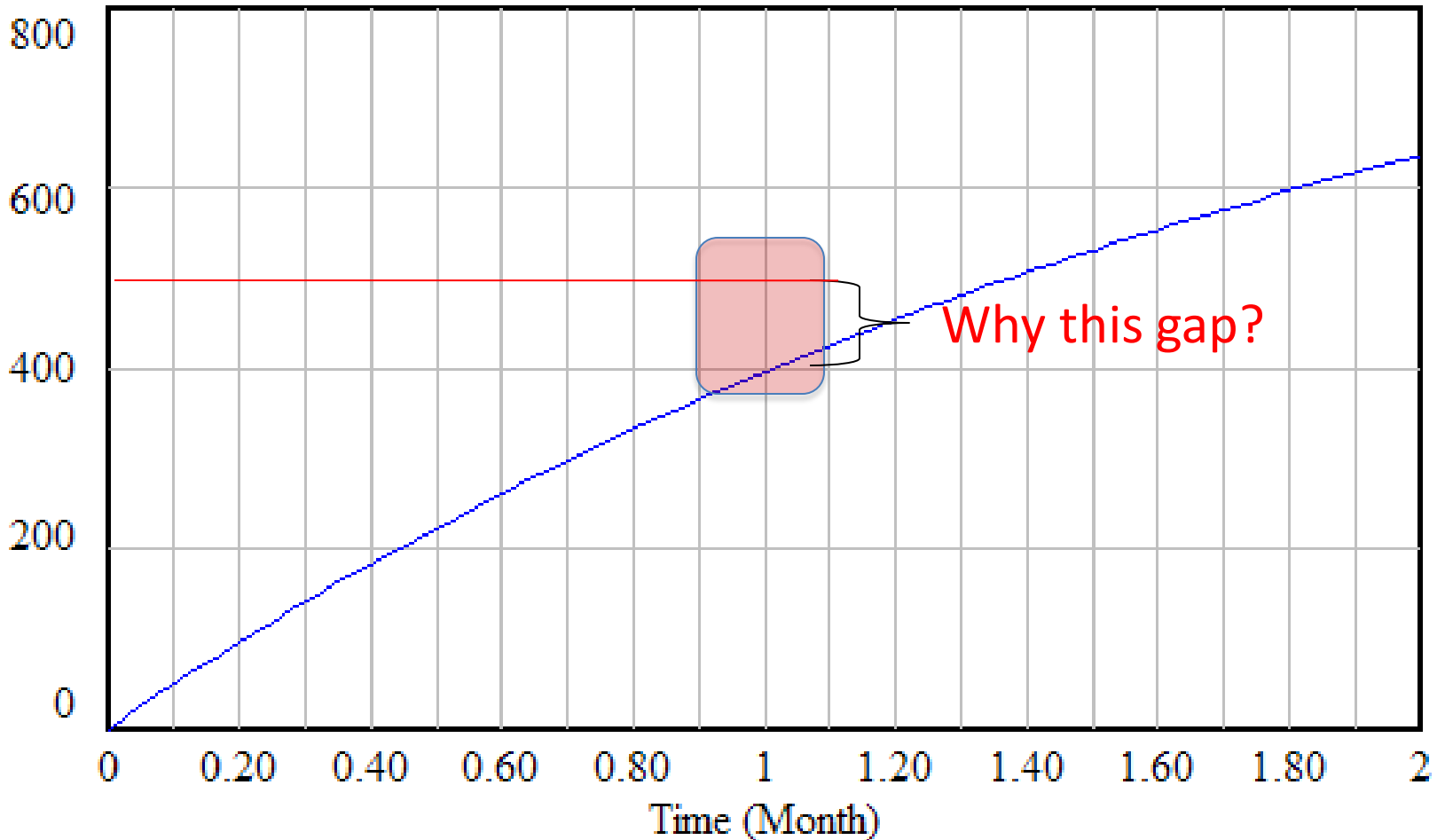


Closeup



50% per Month Risk of Deaths

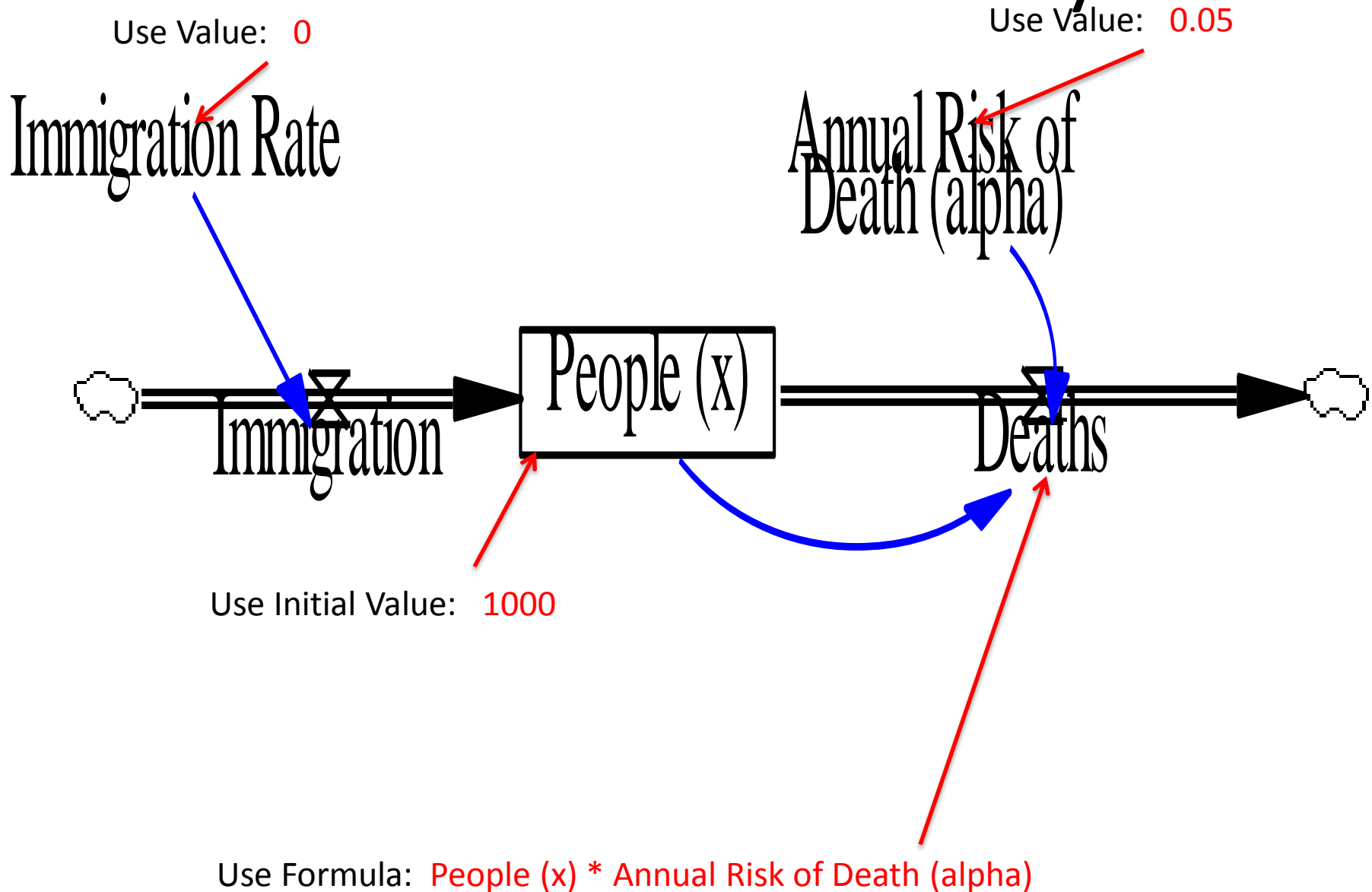
Cumulative Deaths



Answer: The “Gap” is Present Because not all 1000 people are at risk for a month!

- The value of the stock is declining over the first month
- The rate of death indicates that 20% of the population will die per month
- While we may have been expecting 200 people (20% of the 1000) to die, this (erroneously) assumes that all 1000 were at risk for the entire month
 - In fact, because the stock was declining, there were considerably fewer people at risk, meaning that we have fewer deaths
- If we had maintained 1000 people in the stock for the 1st month, 1000 people would have died!

Recall: First Order Delay

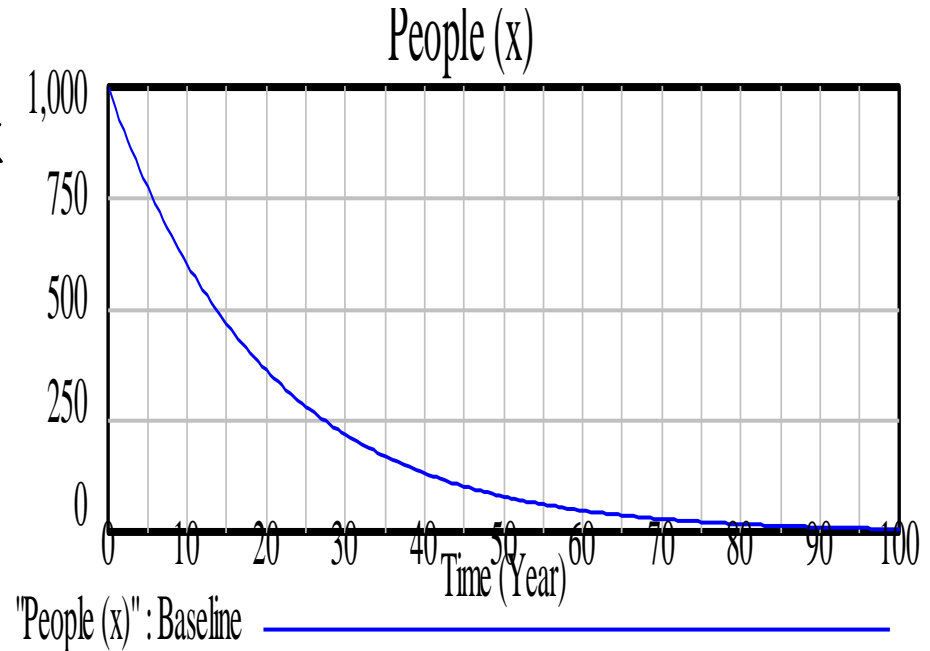


Questions

- What is behaviour of stock x ?
- What is the mean time until people die?
- Suppose we had a constant inflow – what is the behaviour then?

Answers

- Behaviour Of Stock



- Mean Time Until Death

Recall that if coefficient of first order delay is α , then mean time is $1/\alpha$ (Here, $1/0.05 = 20$ years)

Equilibrium Value of a First-Order Delay

- Suppose we have flow of rate i into a stock with a first-order delay out
 - This could be from just a single flow, or many flows
- The value of the stock will approach an equilibrium where inflow=outflow

Equilibrium Value of 1st Order Delay

- Recall: Outflow rate for 1st order delay= αx
 - Note that this depends on the value of the stock!
- Inflow rate= i
- At equilibrium, the level of the stock must be such that inflow=outflow
 - For our case, we have

$$\alpha x = i$$

$$\text{Thus } x = i / \alpha$$

(equivalently, $x = i * \text{Mean time to Transition}$)

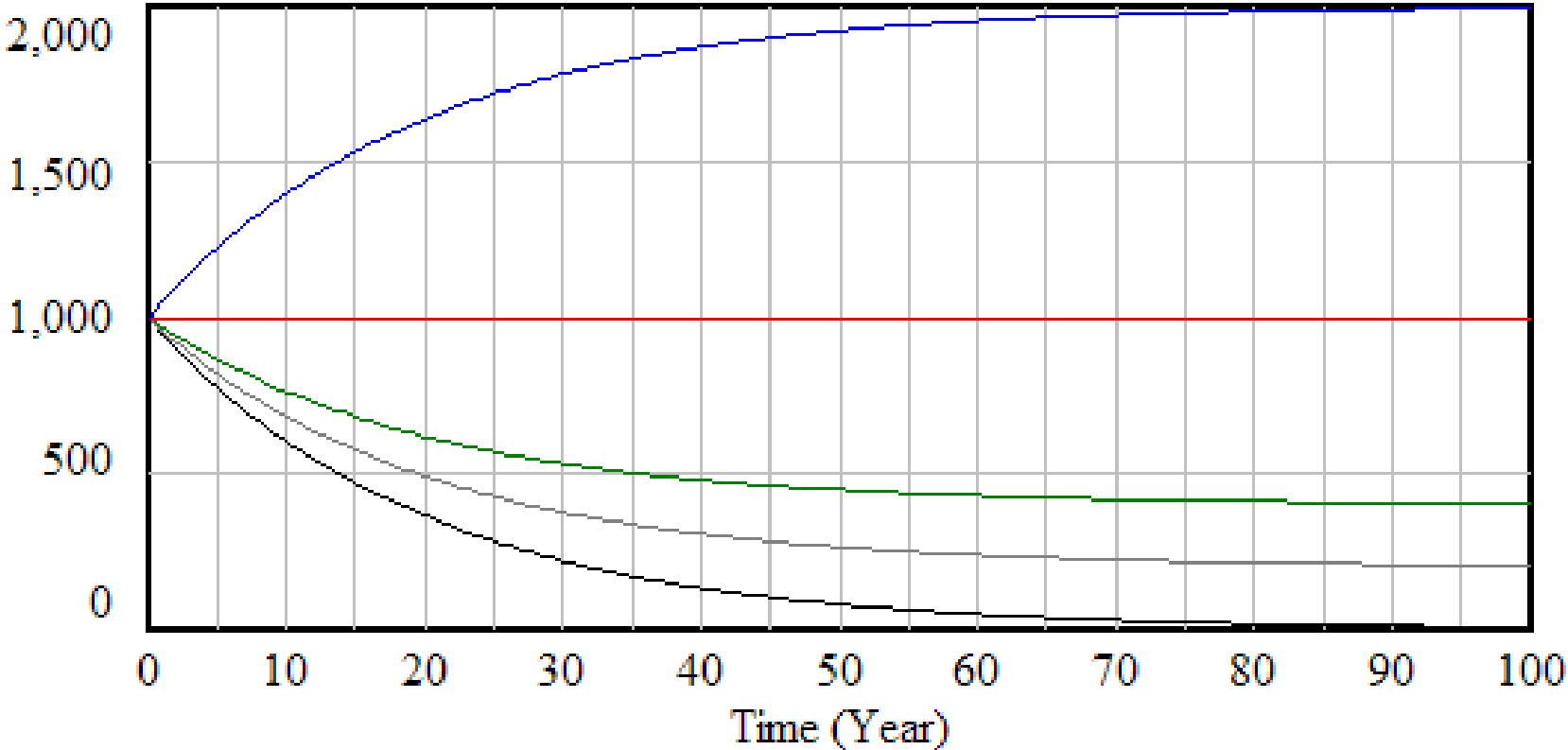
The lower the chance of leaving per time unit (or the longer the delay), the larger the equilibrium value of the stock must be to make outflow=inflow

Scenarios for First Order Delay: Variation in Inflow Rates

- For different immigration (inflows) (what do you expect?)
 - Inflow=10
 - Inflow=20
 - Inflow=50
 - Inflow=100
 - Why do you see this “goal seeking” pattern?
 - What is the “goal” being sought?

Behaviour of Stock for Different Inflows

People (x)

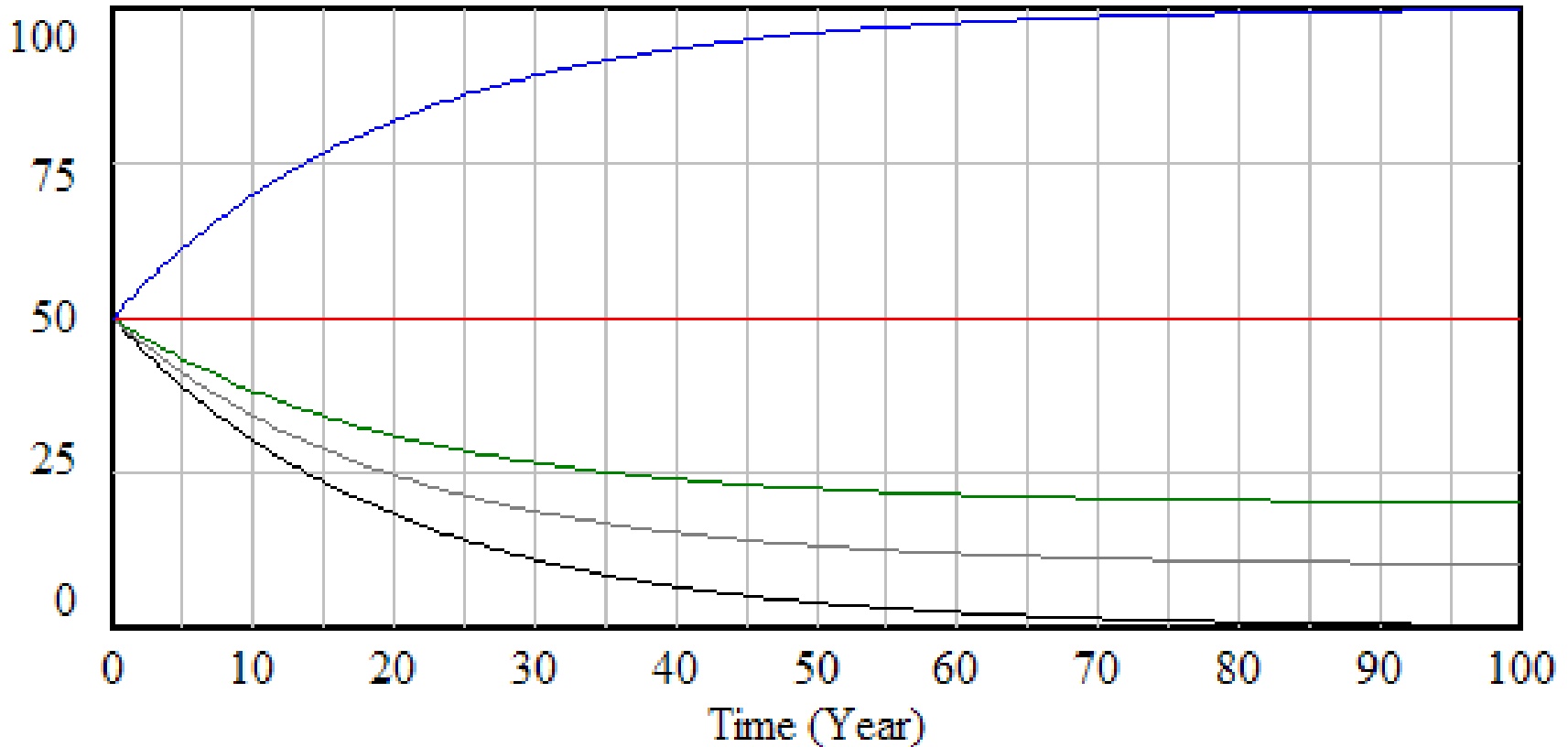


- "People (x)" : Alternative Inflow=100
- "People (x)" : Alternative Inflow=50
- "People (x)" : Alternative Inflow=20
- "People (x)" : Alternative Inflow=10
- "People (x)" : Alternative Inflow=0

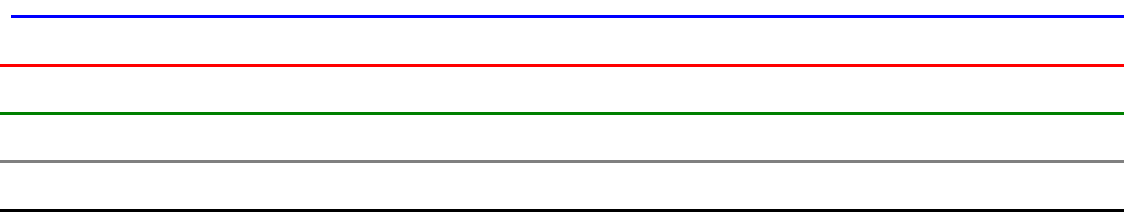
Why do we see this behaviour?

Behaviour of *Outflow* for Different Inflows

Deaths



Deaths : Alternative Inflow=100
Deaths : Alternative Inflow=50
Deaths : Alternative Inflow=20
Deaths : Alternative Inflow=10
Deaths : Alternative Inflow=0

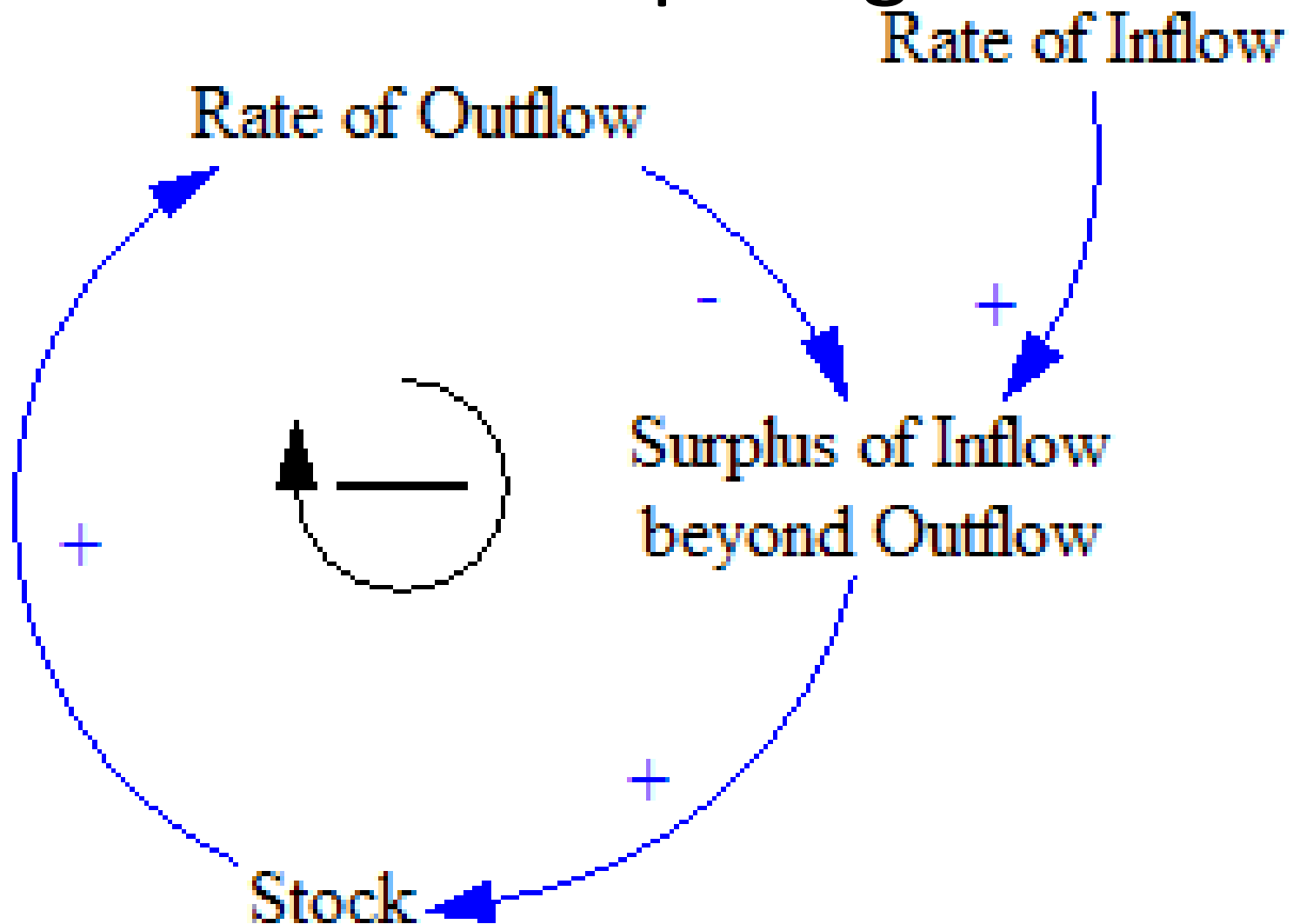


Why do we see this behaviour? Imbalance (gap) causes change to stock (rise or fall) \Rightarrow change to outflow to lower gap **until outflow=inflow**

Goal Seeking Behaviour

- The goal seeking behaviour is associated with a negative feedback loop
 - The larger the population in the stock, the more people die per year
- If we have more people coming in than are going out per year, the stock (and, hence, outflow!) rises until the point where inflow=outflows
- If we have fewer people coming in than are going out per year, the stock declines (& outflow) declines until the point where inflow=outflows

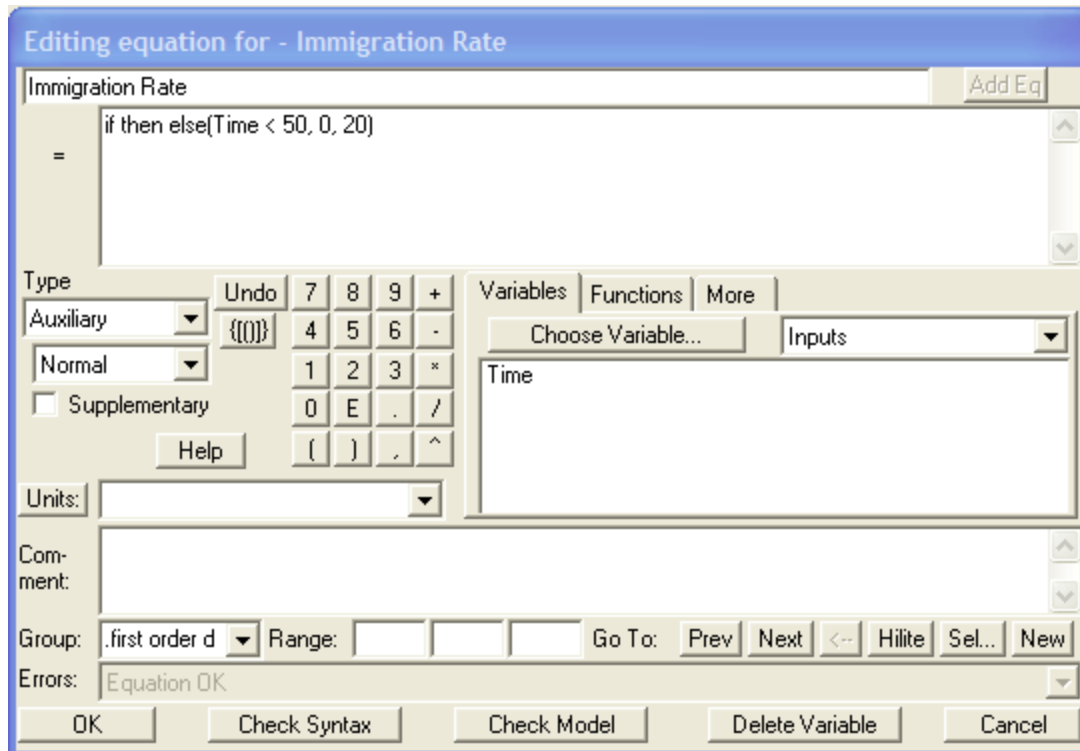
As a Causal Loop Diagram



What does this tell us about how the system would respond to a sudden change in immigration?

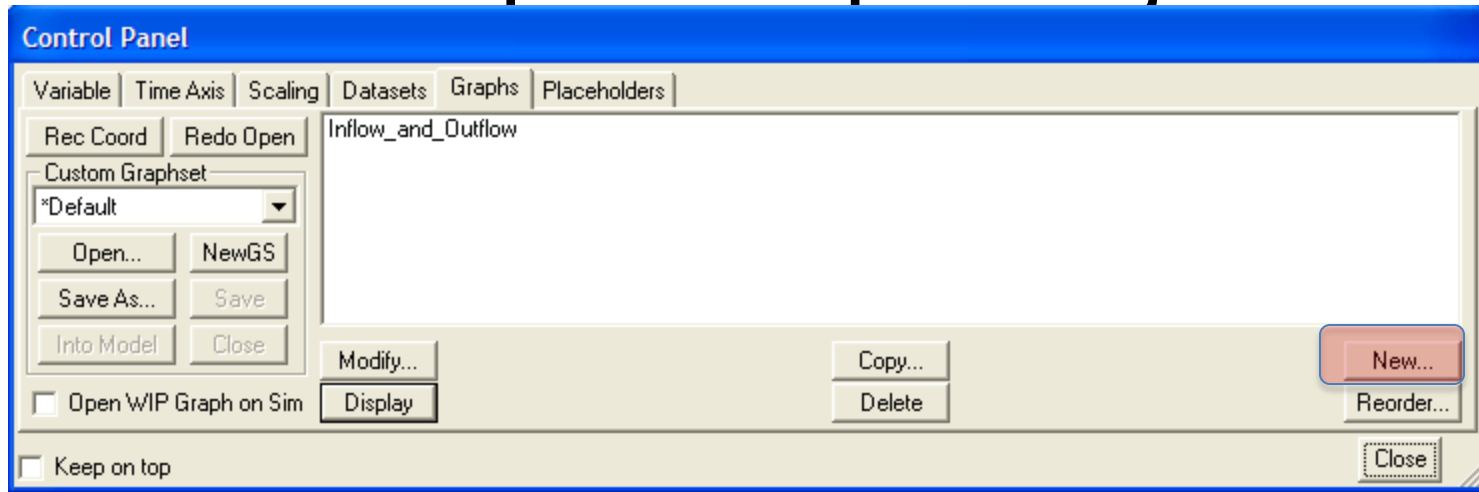
Response to a Change

- Feed in an immigration “step function” that rises suddenly from 0 to 20 at time 50

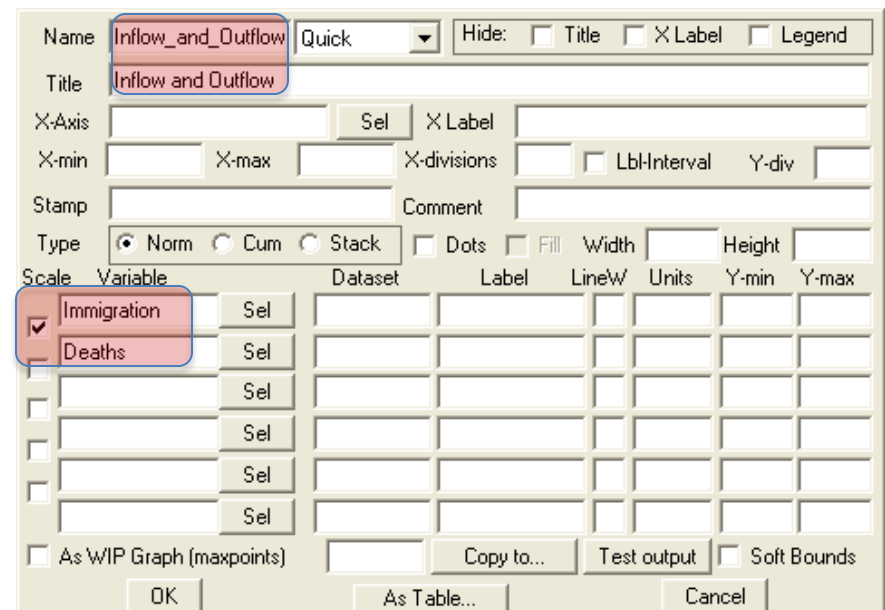


- Set the Initial Value of Stock to 0
- How does the stock change over time?

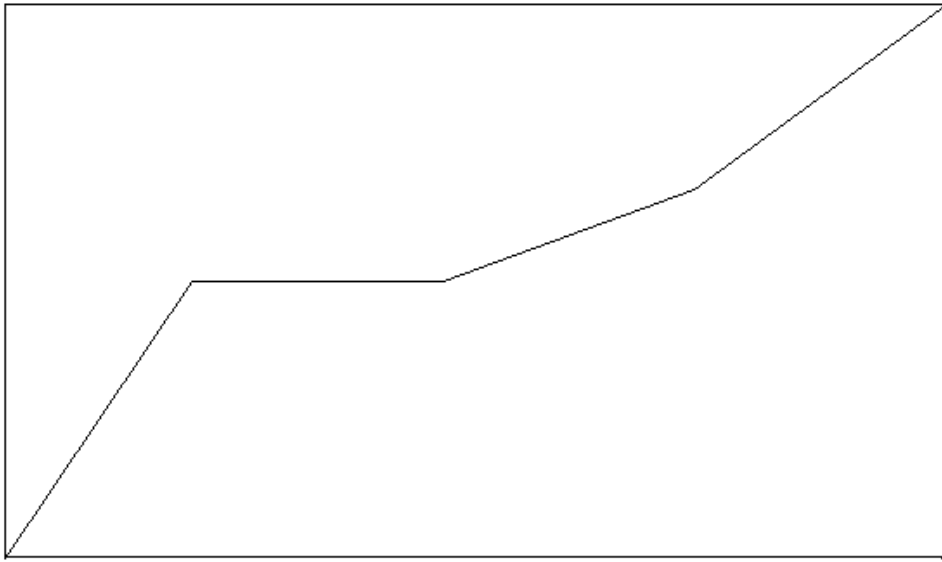
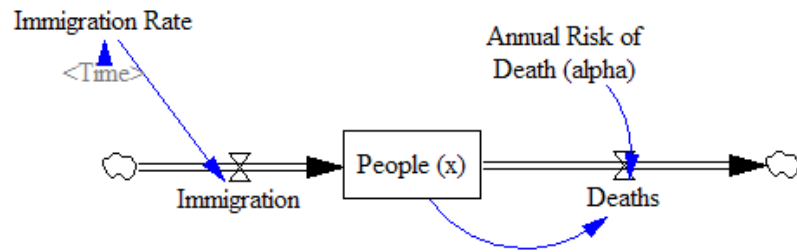
Create a Custom Graph & Display it as an Input-Output Object



- Editing



Create Input-Output Object (for Synthesim)



Input Output Object settings

Object Type
 Input Slider Output Workbench Tool Output Custom Graph

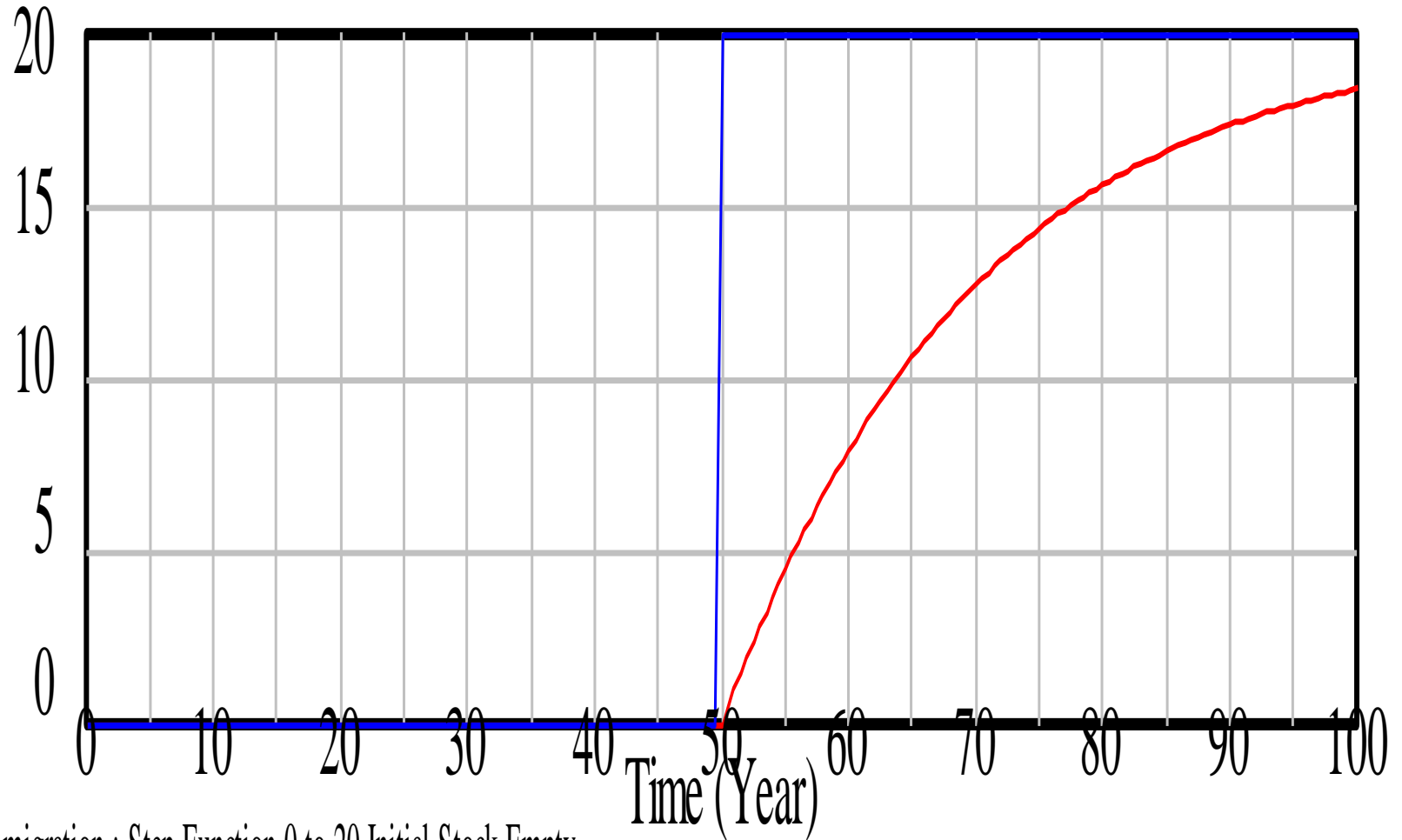
Variable name. Choose:

Slider Settings
Ranging from to with increment
 Label with varname

Custom Graph or Analysis Tool for Output

Stock Starting Empty

Flow Rates Inflow and Outflow



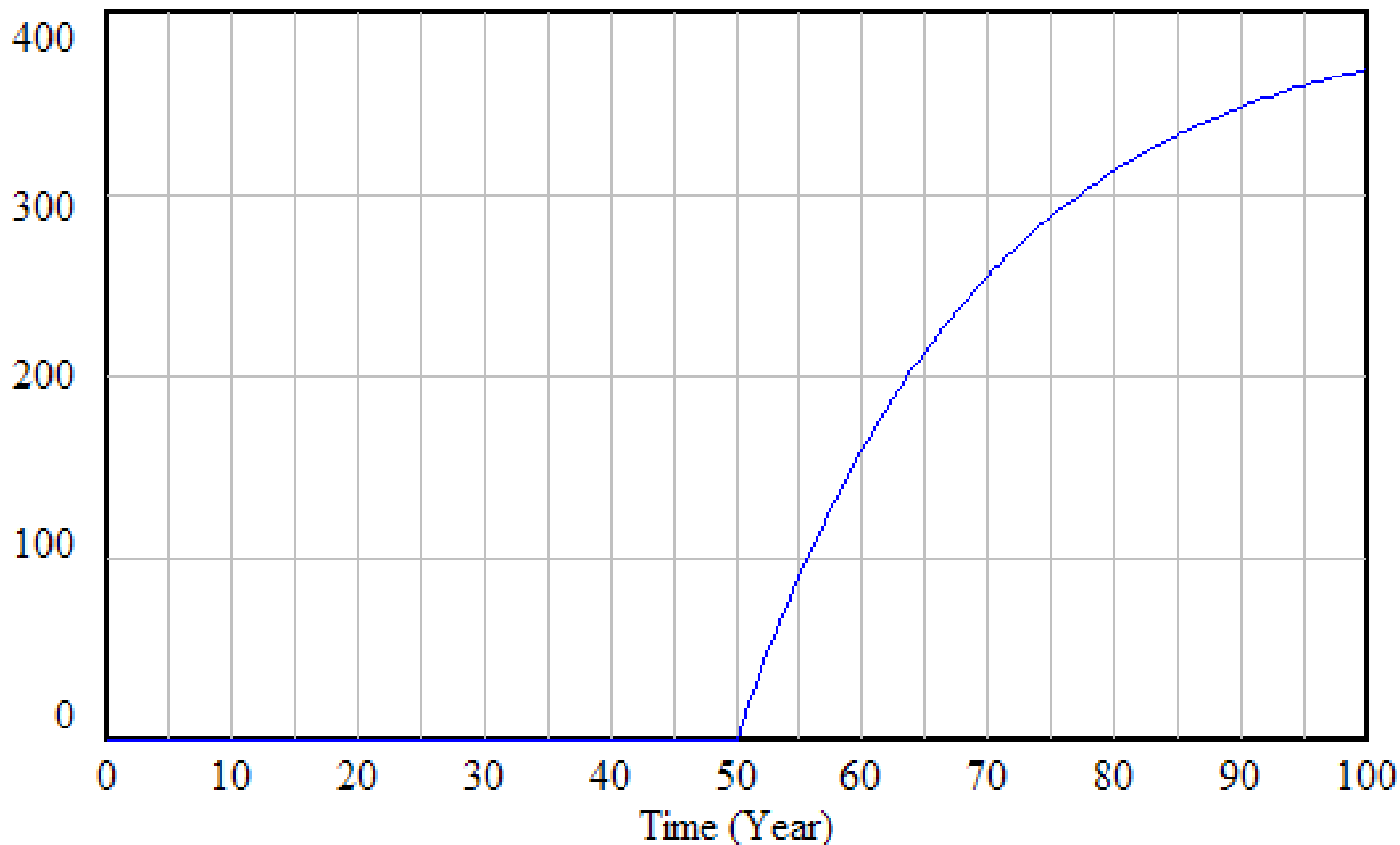
Immigration : Step Function 0 to 20 Initial Stock Empty
Deaths : Step Function 0 to 20 Initial Stock Empty

How would this change with alpha?

Stock Starting Empty?

Value of *Stock* (Alpha=.05)

People (x)

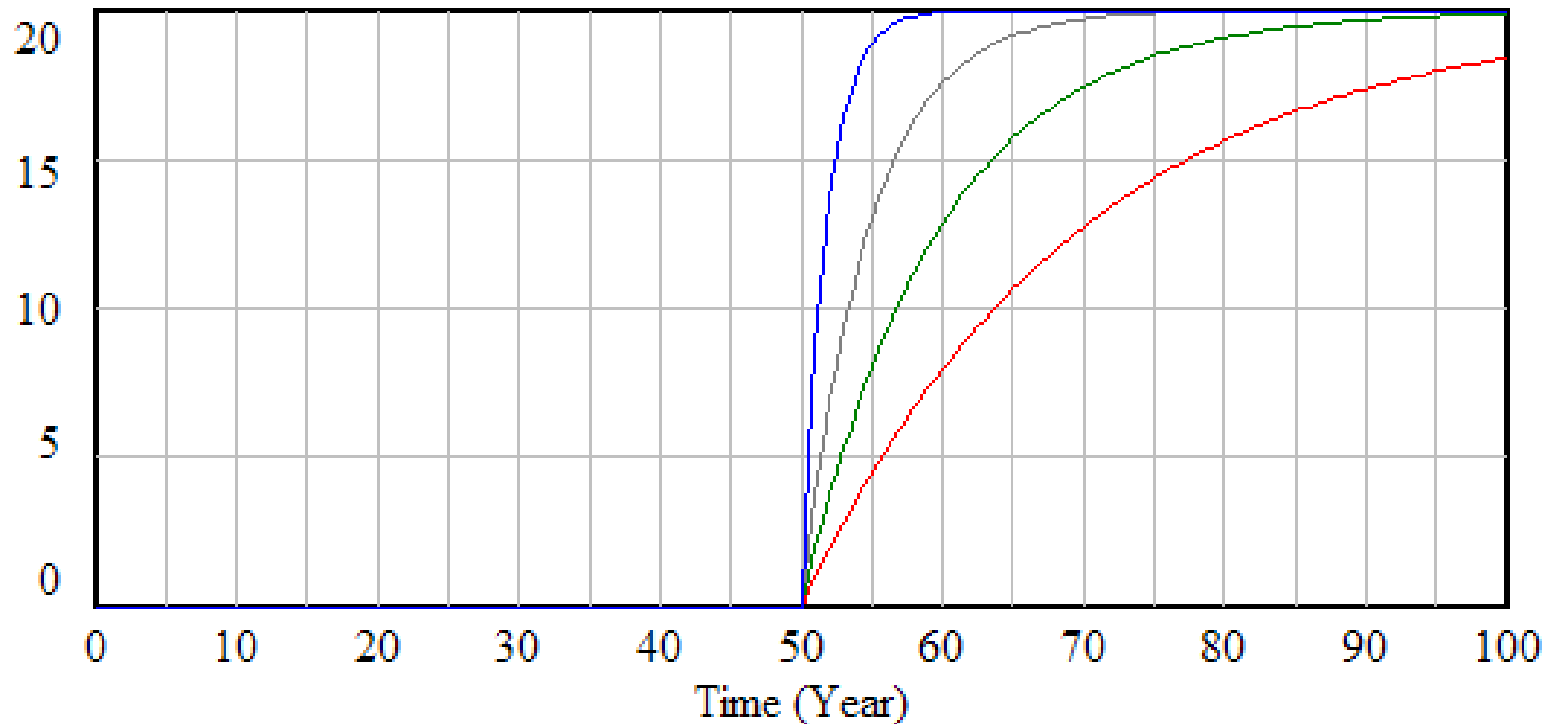


"People (x)" : Step Function 0 to 20 Initial Stock Empty

How would this change with alpha?

For Different Values of (1/) Alpha Flow Rates (Outflow Rises until = Inflow)

Deaths



Deaths : Step Functions 2 yr delay —————

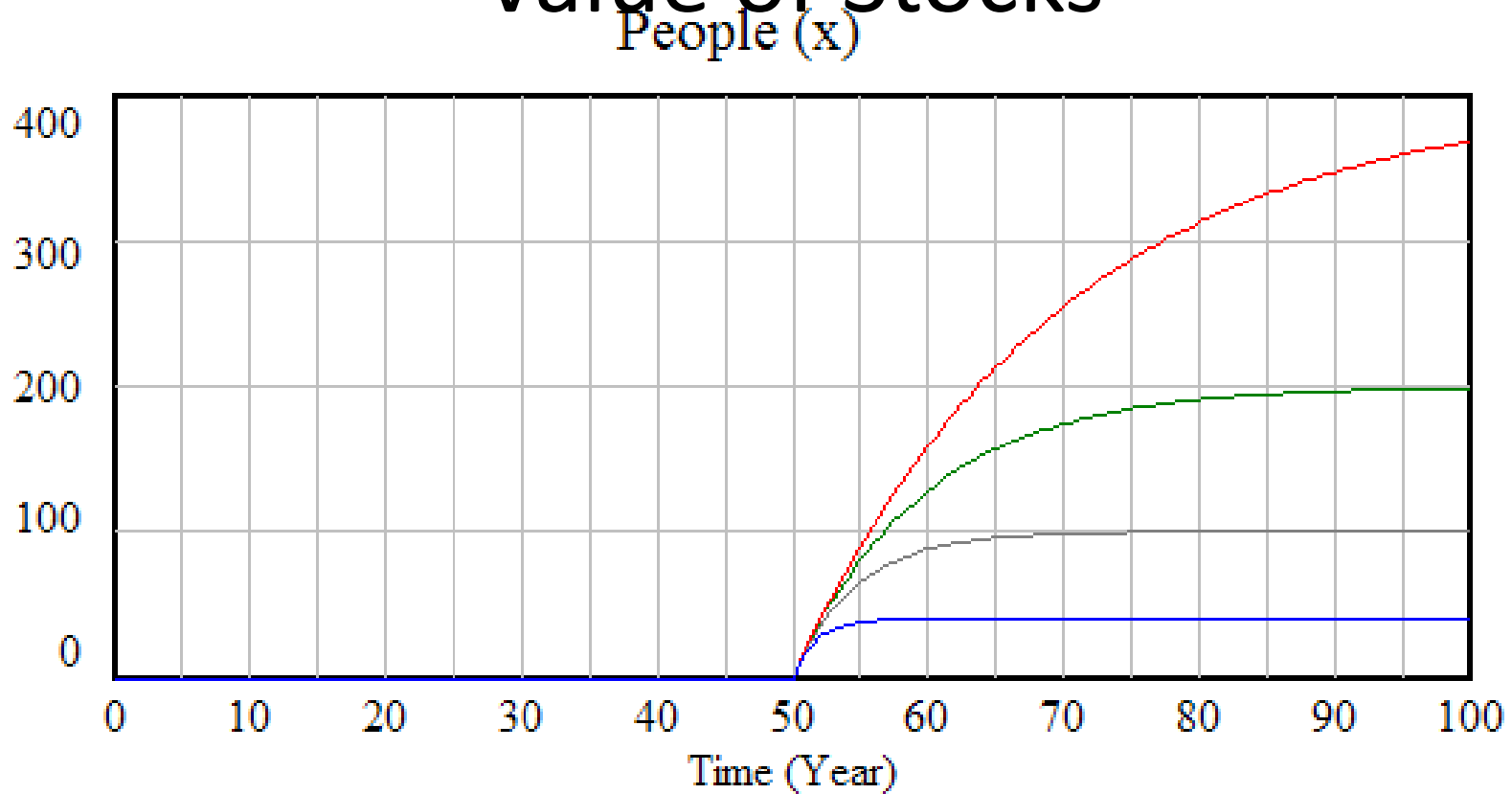
Deaths : Step Functions 20 yr delay —————

Deaths : Step Functions 10 yr delay —————

Deaths : Step Functions 5 yr delay —————

This is for the *flows*. What do stocks do?

For Different Values of (1/) Alpha Value of Stocks



"People (x)" : Step Functions 2 yr delay

"People (x)" : Step Functions 20 yr delay

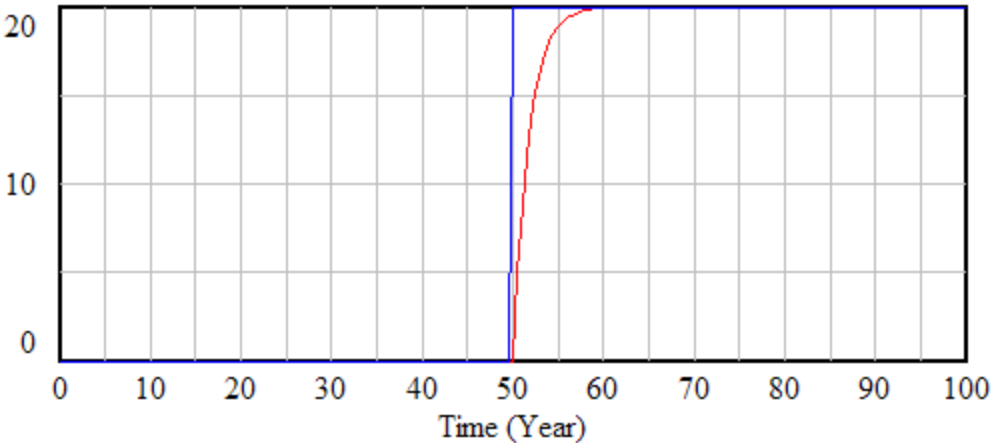
"People (x)" : Step Functions 10 yr delay

"People (x)" : Step Functions 5 yr delay

Why do we see this behaviour? A longer time delay (or smaller chance of leaving per unit time) requires x to be *larger* to make outflow=inflow

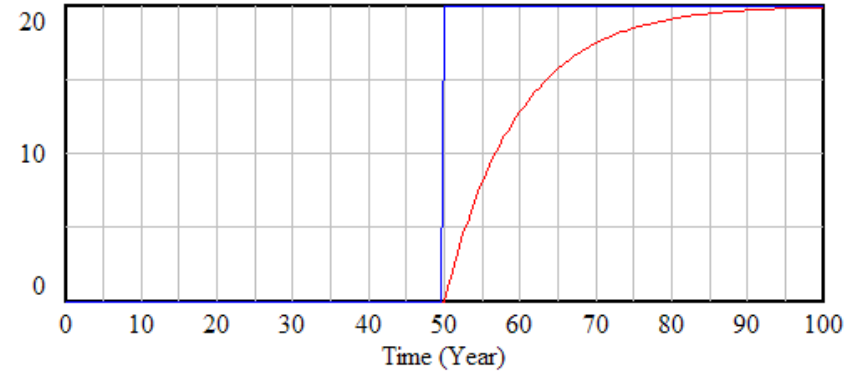
Outflows as Delayed Version of Inputs

Inflow and Outflow



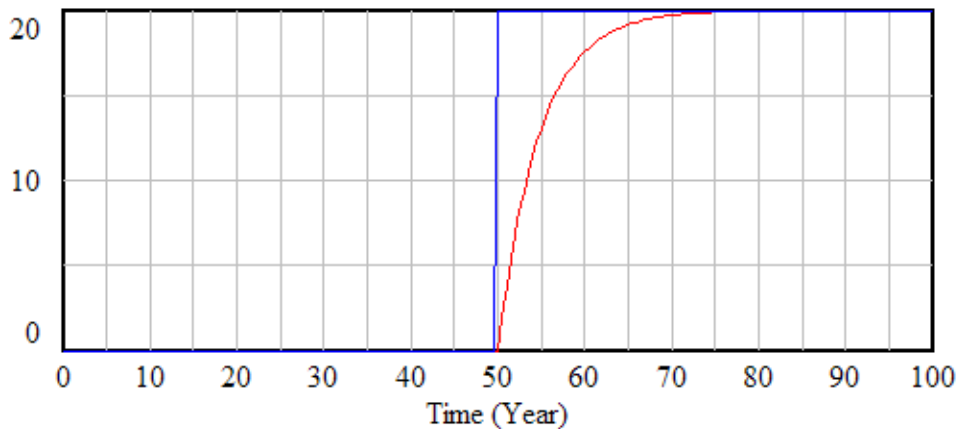
Immigration : Step Functions 2 yr delay —————
 Deaths : Step Functions 2 yr delay —————

Inflow and Outflow



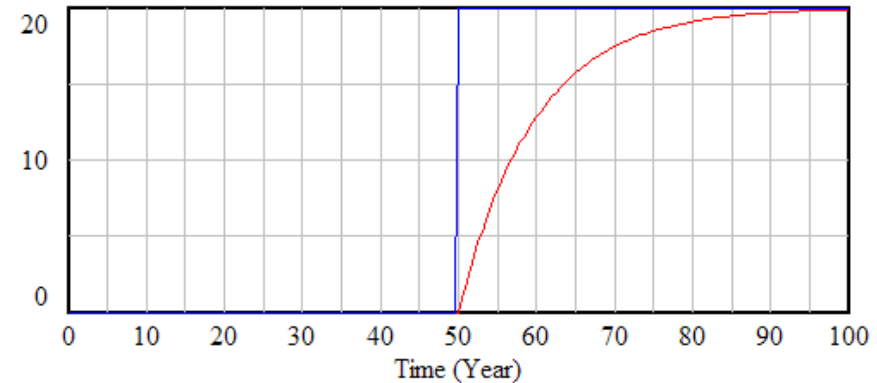
Immigration : Step Functions 10 yr delay —————
 Deaths : Step Functions 10 yr delay —————

Inflow and Outflow



Immigration : Step Functions 5 yr delay —————
 Deaths : Step Functions 5 yr delay —————

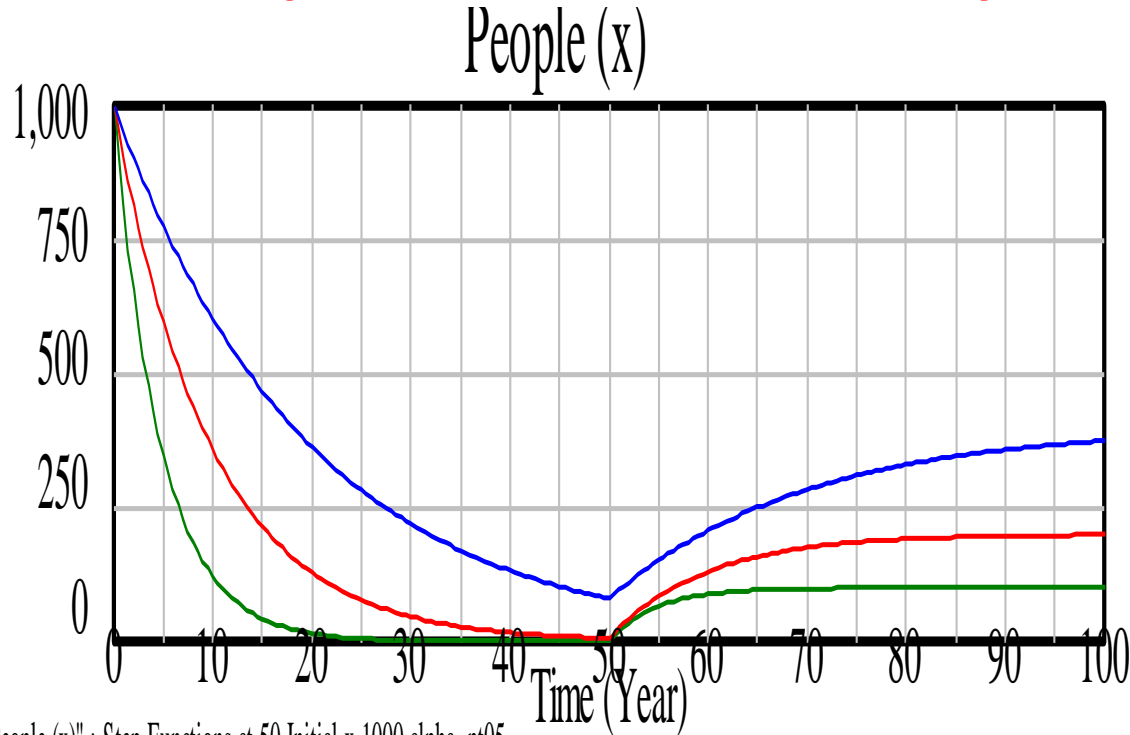
Inflow and Outflow



Immigration : Step Functions 10 yr delay —————
 Deaths : Step Functions 10 yr delay —————

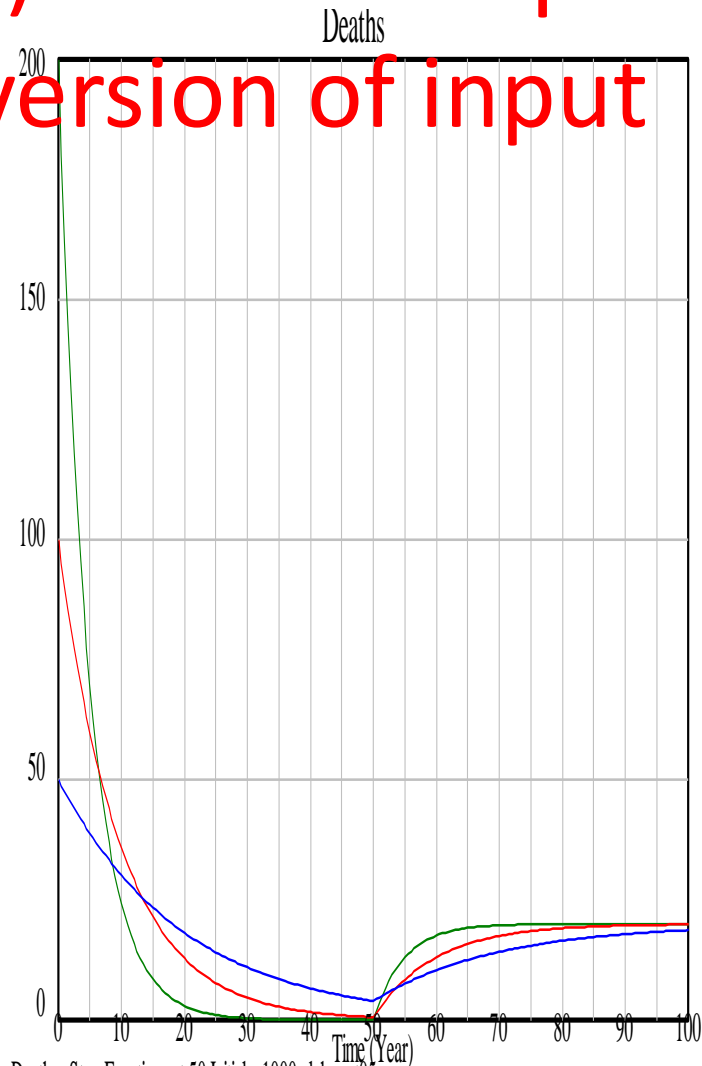
What if stock doesn't start empty?

Decays at first (no inflow) & then output responds with delayed version of input



"People (x)": Step Functions at 50 Initial x 1000 alpha=0.05
"People (x)": Step Functions at 50 Initial x 1000 alpha=0.1
"People (x)": Step Functions at 50 Initial x 1000 alpha=0.2

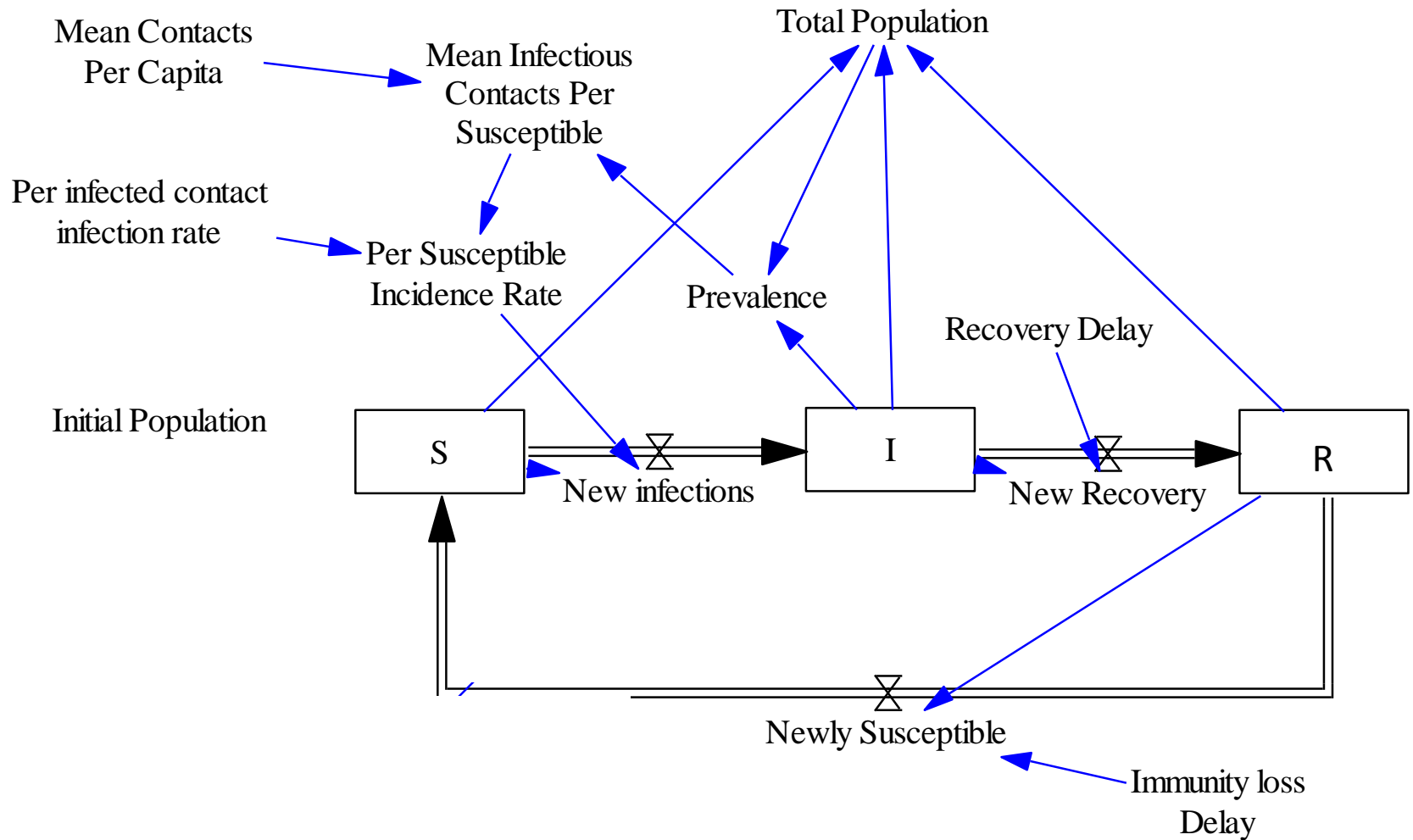
— People (x) alpha=0.05
— People (x) alpha=0.1
— People (x) alpha=0.2



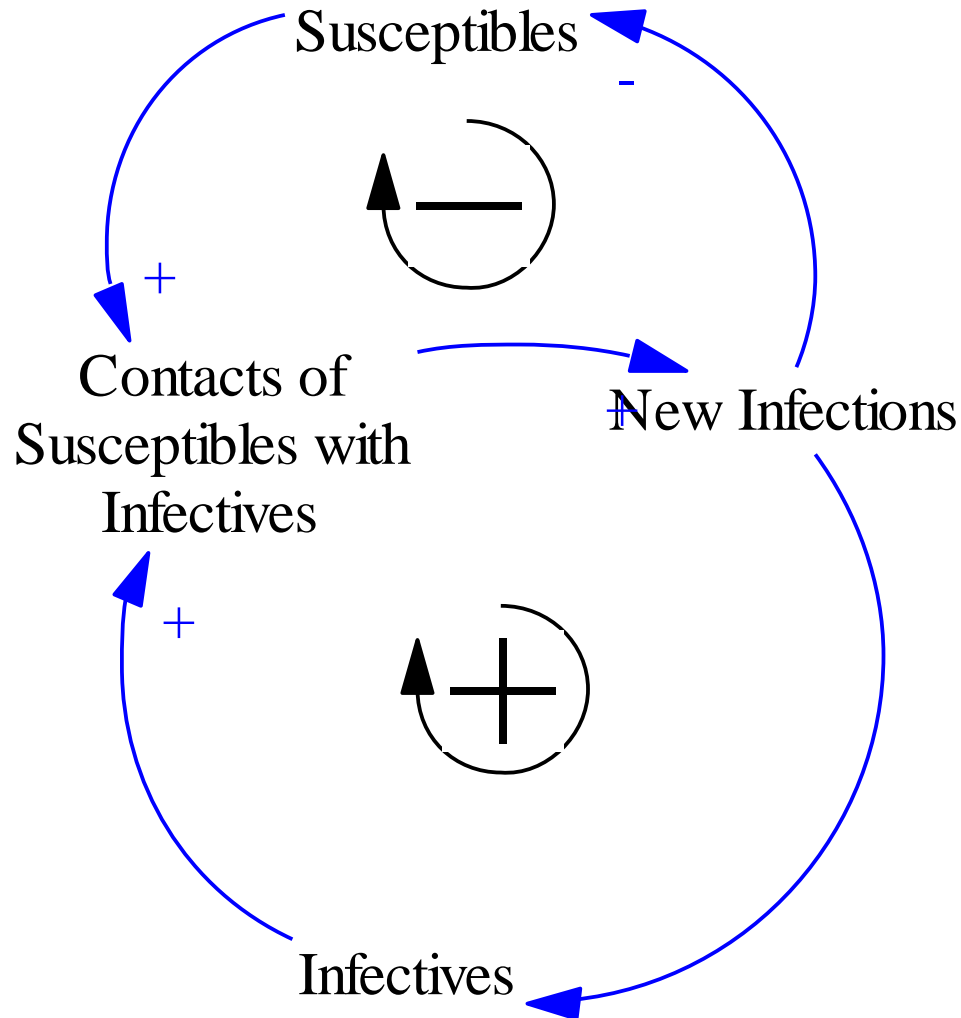
Deaths: Step Functions at 50 Initial x 1000 alpha=0.05
Deaths: Step Functions at 50 Initial x 1000 alpha=0.1
Deaths: Step Functions at 50 Initial x 1000 alpha=0.2

— Deaths alpha=0.05
— Deaths alpha=0.1
— Deaths alpha=0.2

Simple SIT Model

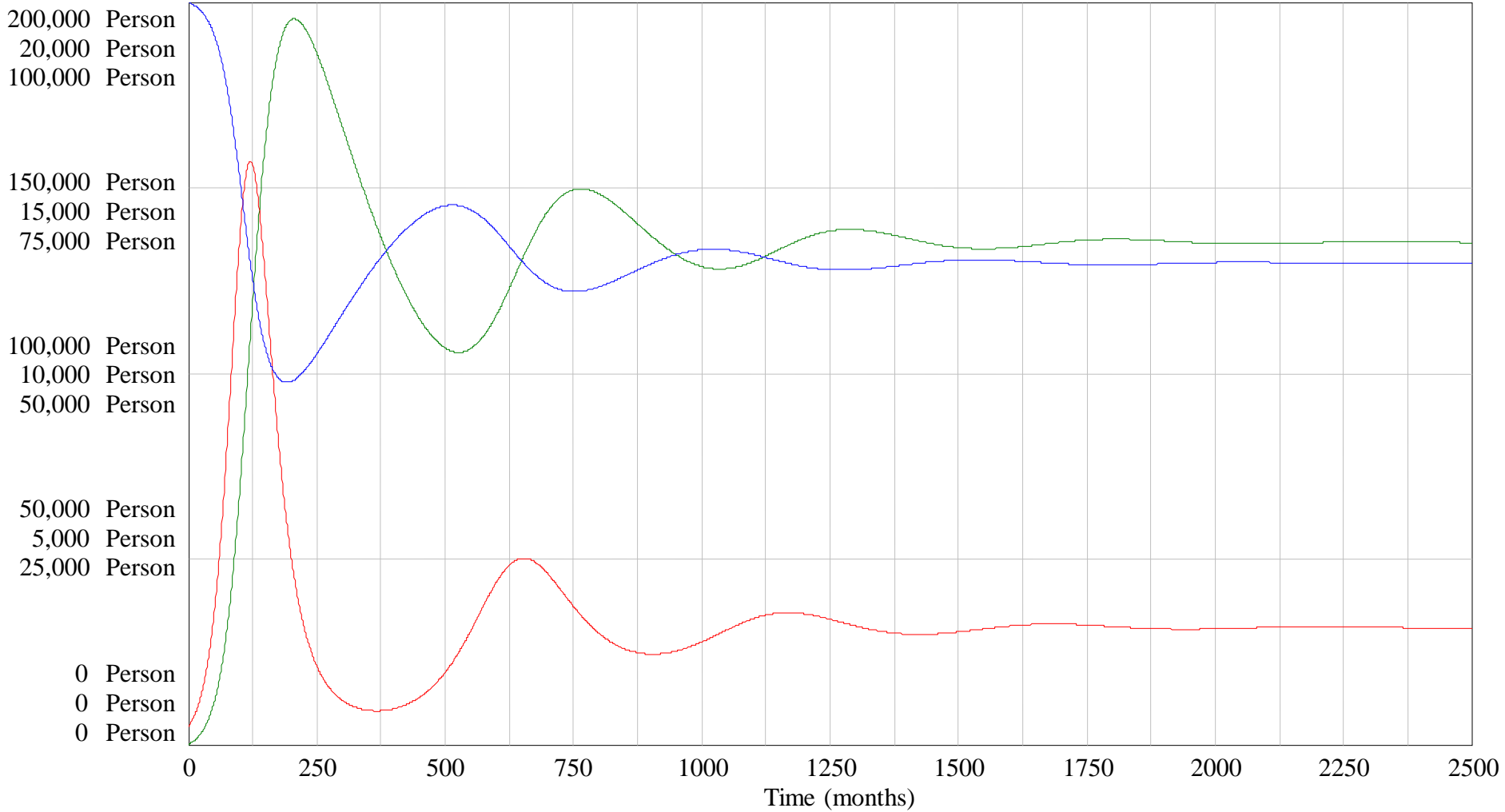


Classic Feedbacks



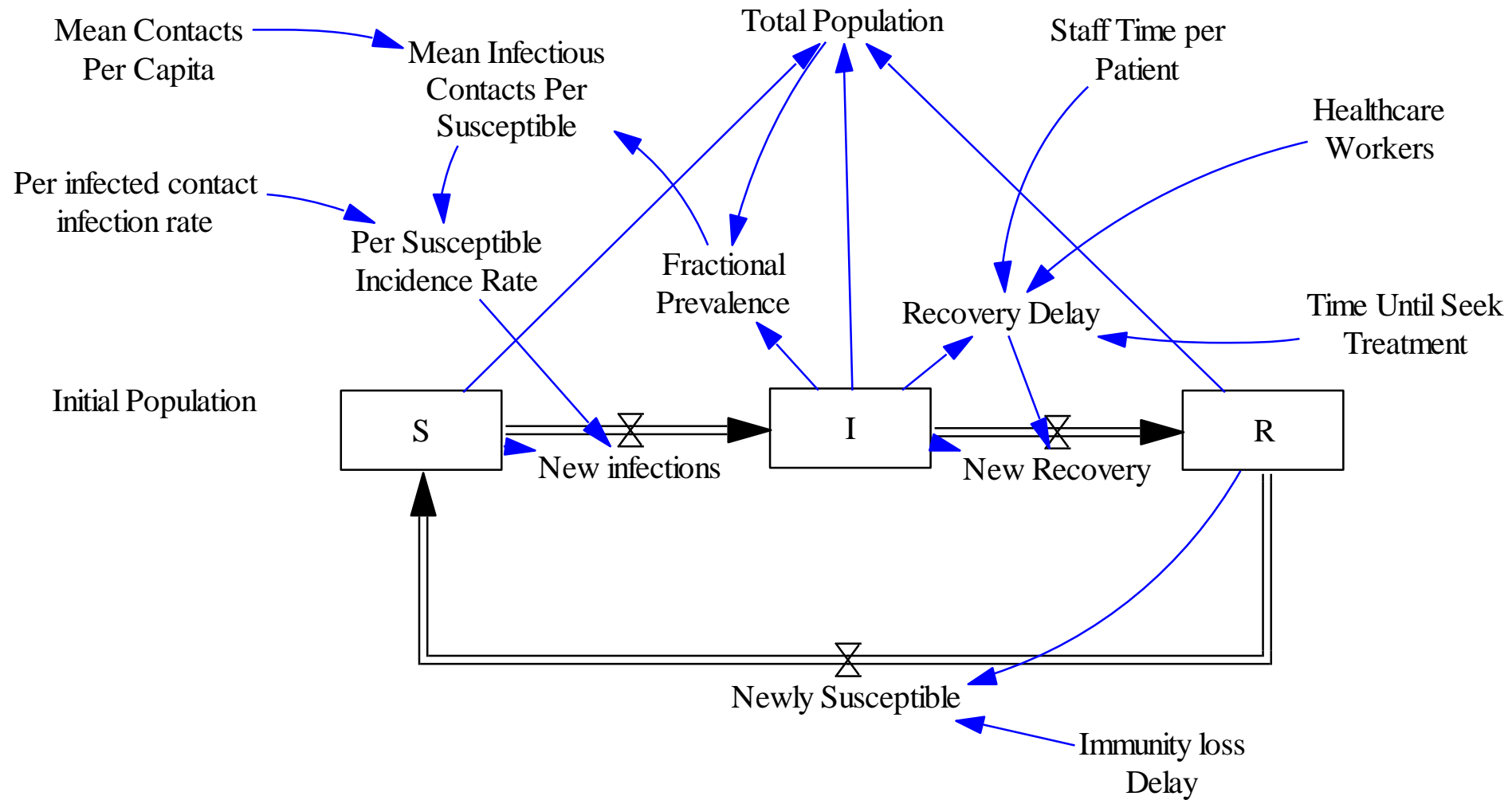
Dynamics

State variables over time

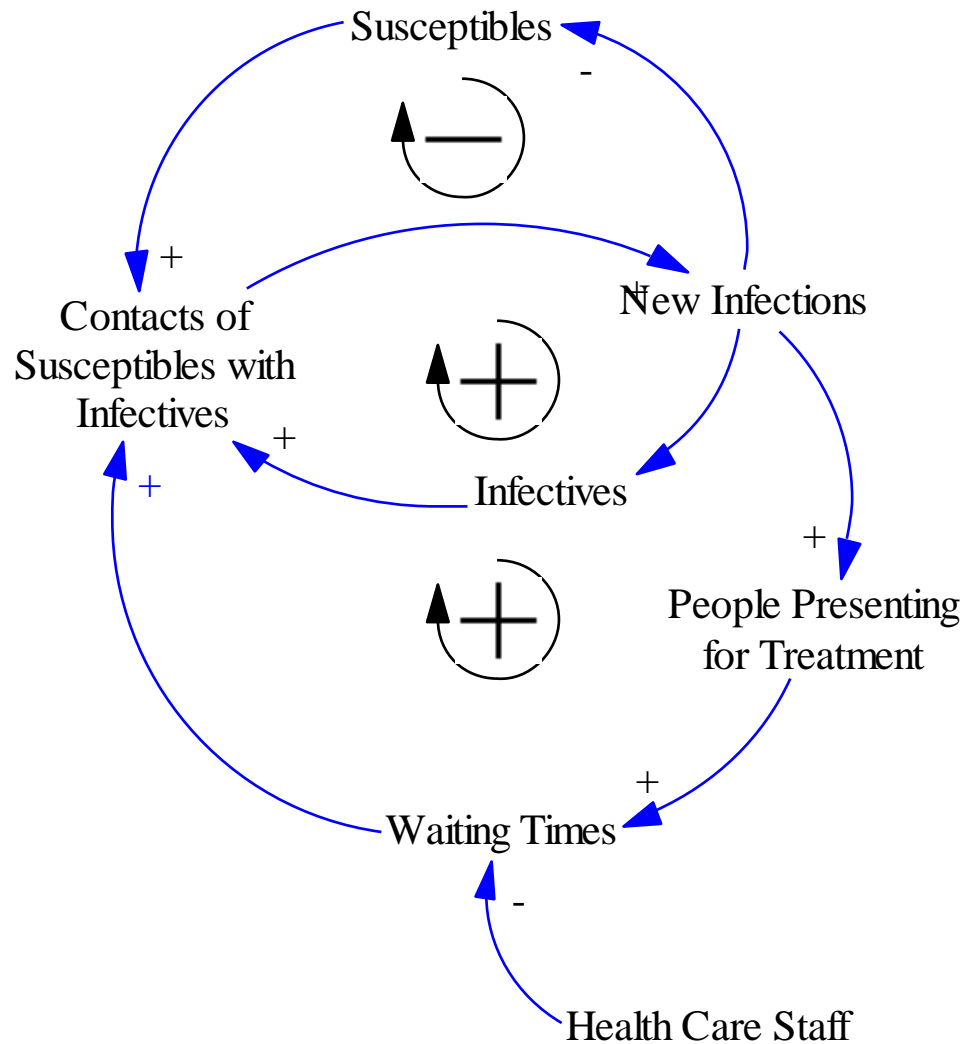


S : Alternative 30 HC Workers Exogenous Recovery Delay Person
I : Alternative 30 HC Workers Exogenous Recovery Delay Person
R : Alternative 30 HC Workers Exogenous Recovery Delay Person

Broadening the Model Boundaries: Endogenous Recovery Delay

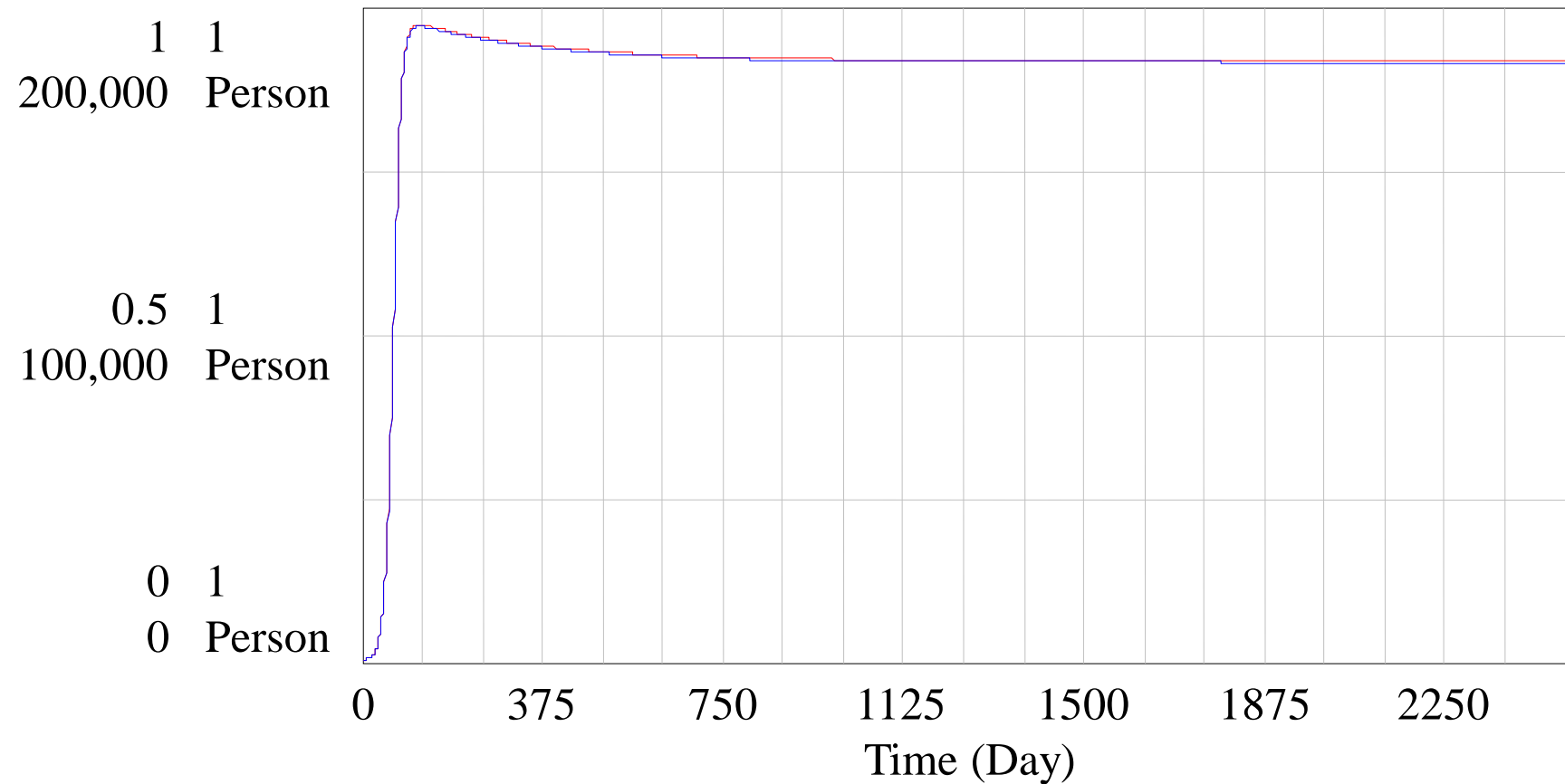


Broadening the Model Boundaries: Endogenous Recovery Delay



A Different Behaviour Mode

Prevalence, Infectious



Prevalence : Baseline 30 HC Workers ————— 1
I : Baseline 30 HC Workers ————— Person

Structure as Shaping Behaviour

- System structure is defined by
 - Stocks
 - Flows
 - Connections between them
- Nonlinearity: The behaviour of the whole is more than the sum of the behaviour of the parts
 - “Emergent” behaviour would not be anticipated from simple behaviour of each piece in turn
- Stock and flow structure (including feedbacks) of a system determines the qualitative behaviour modes that the system can take on