

# Function map

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## Introduction

The function map approach taken here is:

- Deterministic
- Discrete in time
- Continuous in state

## Libraries

```
using DifferentialEquations
using SimpleDiffEq
using DataFrames
using StatsPlots
using BenchmarkTools
```

## Utility functions

To assist in comparison with the continuous time models, we define a function that takes a constant rate,  $r$ , over a timespan,  $t$ , and converts it to a proportion.

```
@inline function rate_to_proportion(r::Float64,t::Float64)
    1-exp(-r*t)
end;
```

## Transitions

We define a function that takes the 'old' state variables,  $u$ , and writes the 'new' state variables into  $du$ . Note that the timestep,  $\delta t$ , is passed as an explicit parameter.

```
function sir_map!(du,u,p,t)
    (S,I,R) = u
    ( $\beta$ ,c, $\gamma$ , $\delta t$ ) = p
    N = S+I+R
    infection = rate_to_proportion( $\beta$ *c*I/N, $\delta t$ )*S
    recovery = rate_to_proportion( $\gamma$ , $\delta t$ )*I
    @inbounds begin
        du[1] = S-infection
        du[2] = I+infection-recovery
        du[3] = R+recovery
    end
    nothing
end;
```

## Time domain

Note that even though I'm using fixed time steps, `DifferentialEquations.jl` complains if I pass integer timespans, so I set the timespan to be `Float64`.

```
δt = 0.1
nsteps = 400
tmax = nsteps*δt
tspan = (0.0,nsteps)
t = 0.0:δt:tmax;
```

## Initial conditions

Note that we define the state variables as floating point.

```
u0 = [990.0,10.0,0.0];
```

## Parameter values

```
p = [0.05,10.0,0.25,δt]; #  $\beta, c, \gamma, \delta t$ 
```

## Running the model

```
prob_map = DiscreteProblem(sir_map!,u0,tspan,p)
```

```
DiscreteProblem with uType Array{Float64,1} and tType Float64. In-place: true
timespan: (0.0, 400.0)
u0: [990.0, 10.0, 0.0]
```

```
sol_map = solve(prob_map,solver=FunctionMap);
```

## Post-processing

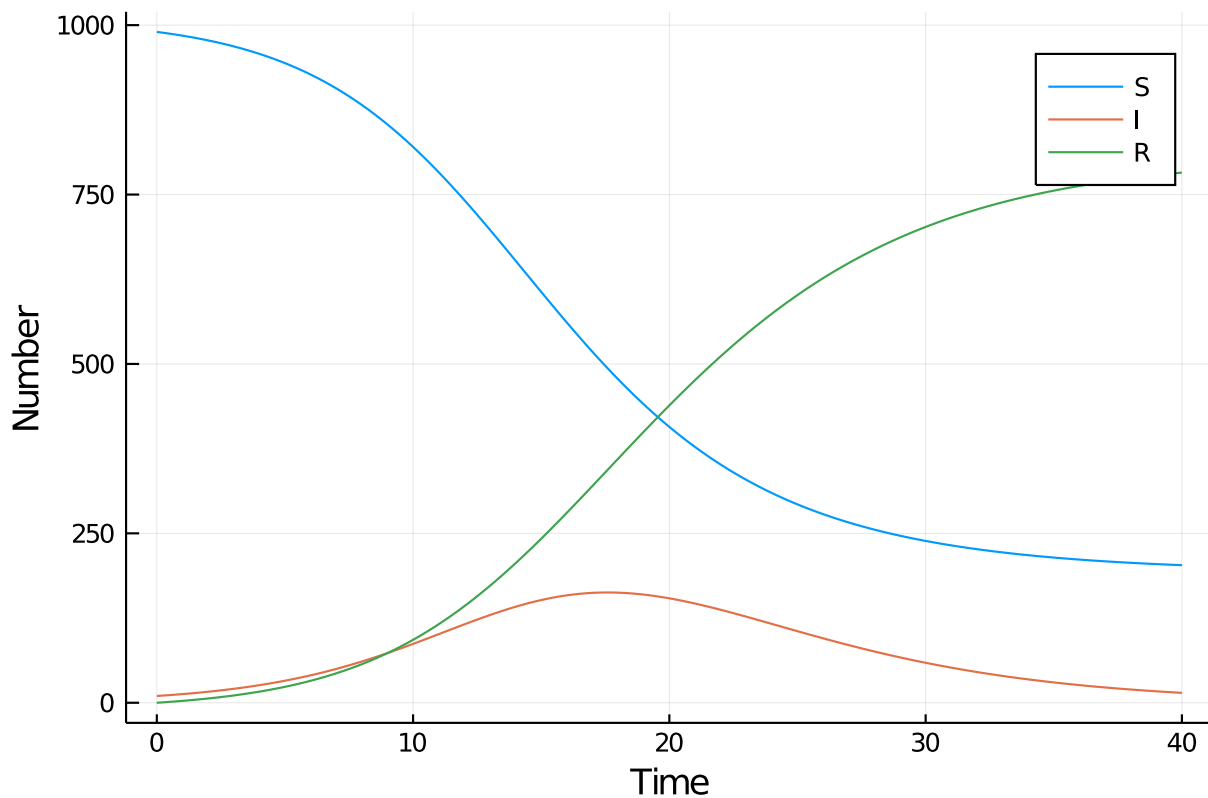
We can convert the output to a dataframe for convenience.

```
df_map = DataFrame(sol_map')
df_map[!,:t] = t;
```

## Plotting

We can now plot the results.

```
@df df_map plot(:t,
    [:x1 :x2 :x3],
    label=["S" "I" "R"],
    xlabel="Time",
    ylabel="Number")
```



## Benchmarking

```
@benchmark solve(prob_map,solver=FunctionMap)
```

BenchmarkTools.Trial:

```
memory estimate: 59.11 KiB
allocs estimate: 479
```

-----

```
minimum time:    101.700 μs (0.00% GC)
median time:     327.550 μs (0.00% GC)
mean time:       360.732 μs (3.05% GC)
maximum time:    31.739 ms (98.47% GC)
```

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```
samples:         10000
evals/sample:    1
```

## Appendix

### Computer Information

Julia Version 1.4.0

Commit b8e9a9ecc6 (2020-03-21 16:36 UTC)

Platform Info:

OS: Windows (x86\_64-w64-mingw32)

CPU: Intel(R) Core(TM) i7-8550U CPU @ 1.80GHz

WORD\_SIZE: 64

LIBM: libopenlibm

```
LLVM: libLLVM-8.0.1 (ORCJIT, skylake)
Environment:
  JULIA_NUM_THREADS = 4
```

## Package Information

```
Status `~\.julia\environments\v1.4\Project.toml`
[46ada45e-f475-11e8-01d0-f70cc89e6671] Agents 3.0.0
[b19378d9-d87a-599a-927f-45f220a2c452] ArrayFire 1.0.6
[c52e3926-4ff0-5f6e-af25-54175e0327b1] Atom 0.12.10
[6e4b80f9-dd63-53aa-95a3-0cdb28fa8baf] BenchmarkTools 0.5.0
[be33ccc6-a3ff-5ff2-a52e-74243cff1e17] CUDAnative 3.0.4
[3a865a2d-5b23-5a0f-bc46-62713ec82fae] CuArrays 2.0.1
[717857b8-e6f2-59f4-9121-6e50c889abd2] DSP 0.6.6
[2445eb08-9709-466a-b3fc-47e12bd697a2] DataDrivenDiffEq 0.2.0
[a93c6f00-e57d-5684-b7b6-d8193f3e46c0] DataFrames 0.20.2
[ebbdde9d-f333-5424-9be2-dbf1e9acfb5e] DiffEqBayes 2.12.1
[eb300fae-53e8-50a0-950c-e21f52c2b7e0] DiffEqBiological 4.3.0
[aae7a2af-3d4f-5e19-a356-7da93b79d9d0] DiffEqFlux 1.8.1
[c894b116-72e5-5b58-be3c-e6d8d4ac2b12] DiffEqJump 6.6.3
[41bf760c-e81c-5289-8e54-58b1f1f8abe2] DiffEqSensitivity 6.13.0
[6d1b261a-3be8-11e9-3f2f-0b112a9a8436] DiffEqTutorials 0.1.0
[0c46a032-eb83-5123-abaf-570d42b7fbaa] DifferentialEquations 6.13.0
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[634d3b9d-ee7a-5ddf-bec9-22491ea816e1] DrWatson 1.10.2
[587475ba-b771-5e3f-ad9e-33799f191a9c] Flux 0.10.4
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[28b8d3ca-fb5f-59d9-8090-bfdbd6d07a71] GR 0.48.0
[523d8e89-b243-5607-941c-87d699ea6713] Gillespie 0.1.0
[7073ff75-c697-5162-941a-fcdaad2a7d2a] IJulia 1.21.2
[e5e0dc1b-0480-54bc-9374-aad01c23163d] Juno 0.8.1
[23fbe1c1-3f47-55db-b15f-69d7ec21a316] Latexify 0.12.0
[d8e11817-5142-5d16-987a-aa16d5891078] MLStyle 0.4.0
[961ee093-0014-501f-94e3-6117800e7a78] ModelingToolkit 3.0.2
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[1dea7af3-3e70-54e6-95c3-0bf5283fa5ed] OrdinaryDiffEq 5.34.1
[91a5bcd-d55d7-5caf-9e0b-520d859cae80] Plots 1.0.12
[e6cf234a-135c-5ec9-84dd-332b85af5143] RandomNumbers 1.4.0
[c5292f4c-5179-55e1-98c5-05642aab7184] ResumableFunctions 0.5.1
[428bdadb-6287-5aa5-874b-9969638295fd] SimJulia 0.8.0
[05bca326-078c-5bf0-a5bf-ce7c7982d7fd] SimpleDiffEq 1.1.0
[2913bbd2-ae8a-5f71-8c99-4fb6c76f3a91] StatsBase 0.33.0
[f3b207a7-027a-5e70-b257-86293d7955fd] StatsPlots 0.14.5
[789caeaf-c7a9-5a7d-9973-96adeb23e2a0] StochasticDiffEq 6.19.2
[a759f4b9-e2f1-59dc-863e-4aeb61b1ea8f] TimerOutputs 0.5.5
[fce5fe82-541a-59a6-adf8-730c64b5f9a0] Turing 0.11.0
[44d3d7a6-8a23-5bf8-98c5-b353f8df5ec9] Weave 0.9.4
[37e2e46d-f89d-539d-b4ee-838fcccc9c8e] LinearAlgebra
```

[cf7118a7-6976-5b1a-9a39-7adc72f591a4] UUIDs