Jump process

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Introduction

This implementation defines the model as a combination of two jump processes, infection and recovery, simulated using the Doob-Gillespie algorithm.

Libraries

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

Transitions

For each process, we define the rate at which it occurs, and how the state variables change at each jump. Note that these are total rates, not *per capita*, and that the change in state variables occurs in-place.

```
function infection_rate(u,p,t)
    (S,I,R) = u
    (\beta, c, \gamma) = p
    N = S+I+R
    \beta*c*I/N*S
function infection! (integrator)
  integrator.u[1] -= 1
  integrator.u[2] += 1
infection_jump = ConstantRateJump(infection_rate,infection!);
function recovery_rate(u,p,t)
    (S,I,R) = u
    (\beta, c, \gamma) = p
    \gamma*I
function recovery!(integrator)
  integrator.u[2] -= 1
  integrator.u[3] += 1
recovery_jump = ConstantRateJump(recovery_rate,recovery!);
```

Time domain

```
tmax = 40.0
tspan = (0.0,tmax);
```

For plotting, we can also define a separate time series.

```
\delta t = 0.1

t = 0:\delta t:tmax;
```

Initial conditions

```
u0 = [990, 10, 0]; # S, I, R
```

Parameter values

```
p = [0.05, 10.0, 0.25]; \# \beta, c, \gamma
```

Random number seed

We set a random number seed for reproducibility.

```
Random.seed! (1234);
```

Running the model

Running this model involves:

- Setting up the problem as a DiscreteProblem;
- Adding the jumps and setting the algorithm using JumpProblem; and
- Running the model, specifying SSAStepper()

Post-processing

In order to get output comparable across implementations, we output the model at a fixed set of times.

```
out_jump = sol_jump(t);
```

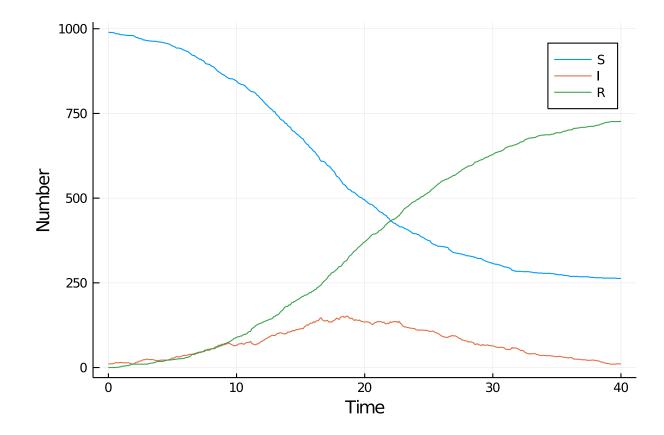
We can convert to a dataframe for convenience.

```
df_jump = DataFrame(out_jump')
df_jump[!,:t] = out_jump.t;
```

Plotting

We can now plot the results.

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



Benchmarking

```
@benchmark solve(prob_jump,FunctionMap())
BenchmarkTools.Trial:
  memory estimate: 14.44 KiB
  allocs estimate: 125
```

samples: 5478
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` [80f14c24-f653-4e6a-9b94-39d6b0f70001] AbstractMCMC 1.0.1 [46ada45e-f475-11e8-01d0-f70cc89e6671] Agents 3.1.0 [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 [a134a8b2-14d6-55f6-9291-3336d3ab0209] BlackBoxOptim 0.5.0 [336ed68f-0bac-5ca0-87d4-7b16caf5d00b] CSV 0.6.2 [be33ccc6-a3ff-5ff2-a52e-74243cff1e17] CUDAnative 3.1.0 [3a865a2d-5b23-5a0f-bc46-62713ec82fae] CuArrays 2.2.0 [717857b8-e6f2-59f4-9121-6e50c889abd2] DSP 0.6.6 [2445eb08-9709-466a-b3fc-47e12bd697a2] DataDrivenDiffEq 0.3.1 [a93c6f00-e57d-5684-b7b6-d8193f3e46c0] DataFrames 0.21.0 [1313f7d8-7da2-5740-9ea0-a2ca25f37964] DataFramesMeta 0.5.1 [ebbdde9d-f333-5424-9be2-dbf1e9acfb5e] DiffEqBayes 2.1.1 [eb300fae-53e8-50a0-950c-e21f52c2b7e0] DiffEqBiological 4.3.0 [459566f4-90b8-5000-8ac3-15dfb0a30def] DiffEqCallbacks 2.13.1 [aae7a2af-3d4f-5e19-a356-7da93b79d9d0] DiffEqFlux 1.10.2 [c894b116-72e5-5b58-be3c-e6d8d4ac2b12] DiffEqJump 6.7.5 [1130ab10-4a5a-5621-a13d-e4788d82bd4c] DiffEqParamEstim 1.14.1 [41bf760c-e81c-5289-8e54-58b1f1f8abe2] DiffEqSensitivity 6.14.1 [6d1b261a-3be8-11e9-3f2f-0b112a9a8436] DiffEqTutorials 0.1.0

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