# Agent-based model using DifferentialEquations

Simon Frost (@sdwfrost), 2020-05-03

#### Introduction

The agent-based model approach is:

- Stochastic
- Discrete in time
- Discrete in state

There are multiple ways in which the model state can be updated. In this implementation, there is the initial state, u, and the next state, u, and updates occur by looping through all the agents (in this case, just a vector of states), and determining whether a transition occurs each state. This approach is relatively simple as there is a chain of states that an individual passes through (i.e. only one transition type per state). After all states have been updated in du, they are then assigned to the current state, u.

#### Libraries

```
using DifferentialEquations
using DiffEqCallbacks
using Distributions
using StatsBase

Error: ArgumentError: Package StatsBase not found in current path:
- Run `import Pkg; Pkg.add("StatsBase")` to install the StatsBase package.

using Random
using DataFrames
using StatsPlots
using BenchmarkTools
```

## Utility functions

#### Transitions

As this is a simple model, the global state of the system is a vector of infection states, defined using an @enum.

```
@enum InfectionStatus Susceptible Infected Recovered
```

```
function sir_abm!(du,u,p,t)
    (\beta,c,\gamma,\delta t) = p
    N = length(u)
    # Initialize du to u
    for i in 1:N
        du[i] = u[i]
    end
    for i in 1:N # loop through agents
        # If recovered
        if u[i] == Recovered
            continue
        # If susceptible
        elseif u[i] == Susceptible
             ncontacts = rand(Poisson(c*\delta t))
             while ncontacts > 0
                 j = sample(1:N)
                 if j==i
                     continue
                 end
                 a = u[j]
                 if a==Infected && rand() < \beta
                     du[i] = Infected
                     break
                 end
                 ncontacts -= 1
             end
        # If infected
        else u[i] == Infected
             if rand() < \gamma
                 du[i] = Recovered
             end
        end
    end
    nothing
end;
sir_abm! (generic function with 1 method)
```

# Time domain

```
\delta t = 0.1

tf = 40.0

tspan = (0.0, tf);

(0.0, 40.0)
```

#### Parameter values

```
\beta = 0.05

c = 10.0

\gamma = rate_to_proportion(0.25,\deltat)

p = [\beta,c,\gamma,\deltat]

4-element Array{Float64,1}:

0.05

10.0

0.024690087971667385

0.1
```

#### Initial conditions

```
N = 1000
I0 = 10
u0 = Array{InfectionStatus}(undef,N)
for i in 1:N
    if i <= I0
        s = Infected
    else
        s = Susceptible
    end
    u0[i] = s</pre>
```

#### Random number seed

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

MersenneTwister(UInt32[0x000004d2], Random.DSFMT.DSFMT state(Int32[-1393240 018, 1073611148, 45497681, 1072875908, 436273599, 1073674613, -2043716458, 1073445557, -254908435, 1072827086 ... -599655111, 1073144102, 367655457, 1 072985259, -1278750689, 1018350124, -597141475, 249849711, 382, 0]), [0.0, 0, 0.0, 0.0, 0.0, 0.0], UInt128[0x00000000000000000000000000000, 0x00000 000], 1002, 0)

### Running the model

We need some reporting functions.

DiscreteCallback{DiffEqCallbacks.var"#30#31",DiffEqCallbacks.SavingAffect{M ain.##WeaveSandBox#420.var"#7#8",Float64,Tuple{Int64,Int64,Int64},DataStructures.BinaryHeap{Float64,DataStructures.LessThan},Array{Float64,1}},typeof(DiffEqCallbacks.saving\_initialize)}(DiffEqCallbacks.var"#30#31"(), DiffEqCallbacks.SavingAffect{Main.##WeaveSandBox#420.var"#7#8",Float64,Tuple{Int64,Int64,Int64},DataStructures.BinaryHeap{Float64,DataStructures.LessThan},Array{Float64,1}}(Main.##WeaveSandBox#420.var"#7#8"(), SavedValues{tType=Float64, savevalType=Tuple{Int64,Int64}}

t:

```
Float64[] saveval:
```

Tuple{Int64,Int64,Int64}[], DataStructures.BinaryHeap{Float64,DataStructure s.LessThan}(DataStructures.LessThan(), [0.0, 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9 ... 39.1, 39.2, 39.3, 39.4, 39.5, 39.6, 39.7, 39.8, 39.9, 40.0]), [0.0, 0.1, 0.2, 0.3, 0.4, 0.5, 0.6, 0.7, 0.8, 0.9 ... 39.1, 39.2, 39.3, 39.4, 39.5, 39.6, 39.7, 39.8, 39.9, 40.0], false, true, 0), DiffEqCallbacks.saving\_initialize, Bool[0, 0])

```
prob_abm = DiscreteProblem(sir_abm!,u0,tspan,p)
```

DiscreteProblem with uType Array{Main.##WeaveSandBox#420.InfectionStatus,1} and tType Float64. In-place: true timespan: (0.0, 40.0)

u0: Main.##WeaveSandBox#420.InfectionStatus[Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Susceptible, Main.##WeaveSandBox#420.Susceptible]

u: 2-element Array{Array{Main.##WeaveSandBox#420.InfectionStatus,1},1}:

[Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.
##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox
#420.Infected, Main.##WeaveSandBox#420.Infected, Main.##WeaveSandBox#420.In
fected, Main.##WeaveSandBox#420.Infected ... Main.##WeaveSandBox#420.Susceptible, Main.##WeaveSandBox#420.Susceptible

[Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered, Main.##WeaveSandBox#420.Recovered ... Main.##WeaveSandBox#420.Susceptible, Main.##WeaveSandBox#420.Recovered, Main.##Wea

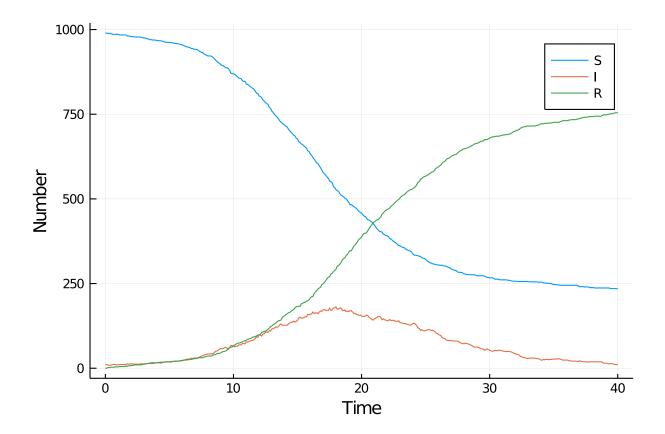
## Post-processing

We can convert the output to a dataframe for convenience.

```
df_abm = DataFrame(saved_values.saveval)
rename!(df_abm,[:S,:I,:R])
df_abm[!,:t] = saved_values.t;
401-element Array{Float64,1}:
 0.0
 0.1
 0.2
 0.3
 0.4
 0.5
 0.6
 0.7
 0.8
 0.9
39.2
39.3
39.4
39.5
39.6
39.7
39.8
39.9
40.0
```

# Plotting

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



# Benchmarking

```
@benchmark solve(prob_abm,
    solver=FunctionMap,
    dt = \delta t,
    callback=cb,
    dense=false,
    save_on=false)
{\tt BenchmarkTools.Trial:}
  memory estimate: 43.41 KiB
  allocs estimate: 79
  -----
  minimum time:
                   46.057 ms (0.00% GC)
  median time: 59.772 ms (0.00% GC) mean time: 60.869 ms (0.00% GC)
                   83.683 ms (0.00% GC)
  maximum time:
                     83
  samples:
  evals/sample:
```

# Appendix

#### **Computer Information**

```
Julia Version 1.4.1
Commit 381693d3df* (2020-04-14 17:20 UTC)
Platform Info:
    OS: Linux (x86_64-pc-linux-gnu)
    CPU: Intel(R) Core(TM) i7-1065G7 CPU @ 1.30GHz
```

WORD\_SIZE: 64 LIBM: libopenlibm

LLVM: libLLVM-8.0.1 (ORCJIT, icelake-client)

Environment:

JULIA NUM THREADS = 4

#### Package Information

```
Status `~/.julia/environments/v1.4/Project.toml`
[46ada45e-f475-11e8-01d0-f70cc89e6671] Agents 3.1.0
[c52e3926-4ff0-5f6e-af25-54175e0327b1] Atom 0.12.11
[6e4b80f9-dd63-53aa-95a3-0cdb28fa8baf] BenchmarkTools 0.5.0
[a134a8b2-14d6-55f6-9291-3336d3ab0209] BlackBoxOptim 0.5.0
[2445eb08-9709-466a-b3fc-47e12bd697a2] DataDrivenDiffEq 0.2.0
[a93c6f00-e57d-5684-b7b6-d8193f3e46c0] DataFrames 0.21.0
[ebbdde9d-f333-5424-9be2-dbf1e9acfb5e] DiffEqBayes 2.14.0
[459566f4-90b8-5000-8ac3-15dfb0a30def] DiffEqCallbacks 2.13.2
[c894b116-72e5-5b58-be3c-e6d8d4ac2b12] DiffEqJump 6.7.5
[1130ab10-4a5a-5621-a13d-e4788d82bd4c] DiffEqParamEstim 1.14.1
[0c46a032-eb83-5123-abaf-570d42b7fbaa] DifferentialEquations 6.14.0
[31c24e10-a181-5473-b8eb-7969acd0382f] Distributions 0.23.2
[634d3b9d-ee7a-5ddf-bec9-22491ea816e1] DrWatson 1.11.0
[587475ba-b771-5e3f-ad9e-33799f191a9c] Flux 0.8.3
[28b8d3ca-fb5f-59d9-8090-bfdbd6d07a71] GR 0.49.1
[523d8e89-b243-5607-941c-87d699ea6713] Gillespie 0.1.0
[7073ff75-c697-5162-941a-fcdaad2a7d2a] IJulia 1.21.2
[4076af6c-e467-56ae-b986-b466b2749572] JuMP 0.21.2
[e5e0dc1b-0480-54bc-9374-aad01c23163d] Juno 0.8.2
[093fc24a-ae57-5d10-9952-331d41423f4d] LightGraphs 1.3.3
[1914dd2f-81c6-5fcd-8719-6d5c9610ff09] MacroTools 0.5.5
[ee78f7c6-11fb-53f2-987a-cfe4a2b5a57a] Makie 0.9.5
[961ee093-0014-501f-94e3-6117800e7a78] ModelingToolkit 3.6.0
[76087f3c-5699-56af-9a33-bf431cd00edd] NLopt 0.6.0
[429524aa-4258-5aef-a3af-852621145aeb] Optim 0.21.0
[1dea7af3-3e70-54e6-95c3-0bf5283fa5ed] OrdinaryDiffEq 5.38.1
[91a5bcdd-55d7-5caf-9e0b-520d859cae80] Plots 1.3.1
[428bdadb-6287-5aa5-874b-9969638295fd] SimJulia 0.8.0
[05bca326-078c-5bf0-a5bf-ce7c7982d7fd] SimpleDiffEq 1.1.0
[f3b207a7-027a-5e70-b257-86293d7955fd] StatsPlots 0.14.6
[789caeaf-c7a9-5a7d-9973-96adeb23e2a0] StochasticDiffEq 6.23.0
[fce5fe82-541a-59a6-adf8-730c64b5f9a0] Turing 0.12.0
[44d3d7a6-8a23-5bf8-98c5-b353f8df5ec9] Weave 0.10.0
```