Using sirgraph

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1 Graph data

A graph object suitable for sirgraph is an igraph object with some extra attributes on the graph and the nodes. A sample can be created with the makedata function.

```
g = makedata()
g

## IGRAPH U--- 34 78 -- Zachary
## + attr: name (g/c), start (g/n), time (g/n), stepsize (g/n),
## vaccinated (v/x), sex (v/c), age (v/n), state (v/c)
```

The easiest way to get the node attributes is via the get.data.frame function.

```
head(get.data.frame(g, "vert"))
     vaccinated sex age state
## 1
                  M 19
          FALSE
## 2
           TRUE
                  F
                     19
                             S
## 3
          FALSE
                  F
                     19
                             S
          FALSE
                  M
                     19
                             S
## 5
                             S
          FALSE
                  Μ
                     19
## 6
           TRUE
                  F 21
                             S
```

As well as extra node attributes, the graph itself has some extra attributes.

```
g$time
## [1] "2014-01-01"
g$start
## [1] "2014-01-01"
g$stepsize
## [1] 1
```

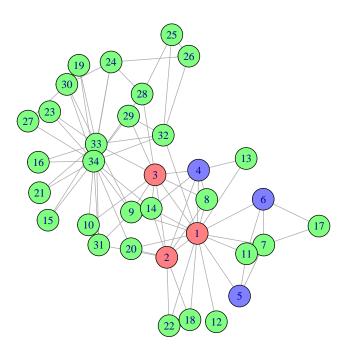
2 Plotting

First we'll set some infected and recovered nodes on the graph.

```
V(g)$state[1:3] = "I"
V(g)$state[4:6] = "R"
```

a simple plotting function can then be used:

```
plotSIR(g)
```



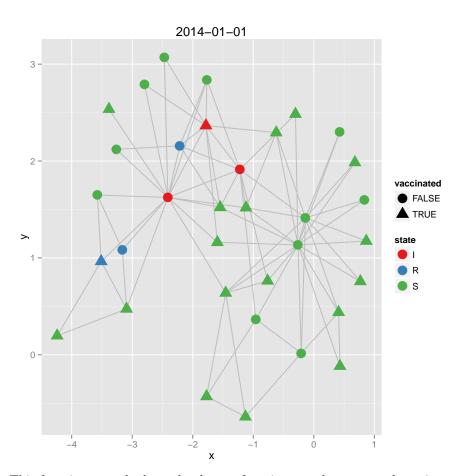
To use the ggplot2 graphics functions we first have to assign coordinates to the graph using the glayout function:

```
g = glayout(g)
```

By default this uses the Kamada-kawai layout, but any of the layout functions from <code>igraph</code> can be used.

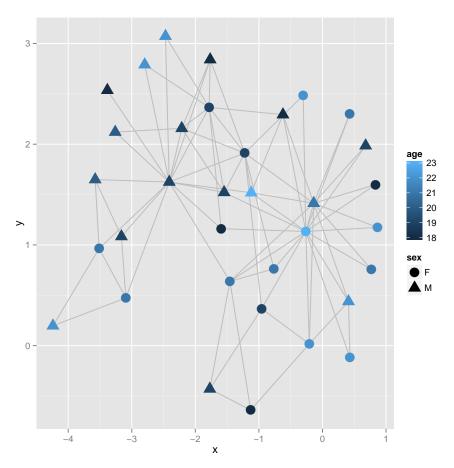
Now we can plot it:

gplotgraph(g)



This function uses the lower-level <code>geom</code> functions to plot state and vaccination status. You can use these lower-level functions to build other plots, for example to plot the network coloured by age and shaped by sex:

```
ggplot() + geom_edge(aes(x = x, y = y), col = "grey", data = g) + geom_node(aes(x = x, y = y, col = age, shape = sex), size = 5, data = g)
```



Full control of the plot is then available using the ggplot2 functions, for example to change the colour scheme or shapes.

3 Discrete Time Infection Modelling

Running an SIR model on a graph network requires the following:

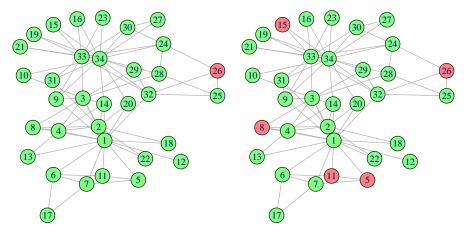
- A graph
- An initial state of each node (S, I, or R).
- A method for the spread of infection (S to I).
- A method for the removal of infection (I to R).
- A way of deciding when to stop the simulation

3.1 Starter Functions

A "starter" function is one which sets the initial infection state of a graph. Possible starter functions could be to set a fixed number of nodes as infectious, or to set infectious status with a fixed probability, or to set infectious status according to attributes of the nodes. A starter function always takes a graph as parameter, and returns the graph with a modified state attribute.

The infectN function is a starter function generator, in that it returns starter functions from a family of functions based on N, the number of initial infectious nodes. So infectN(1) is a starter function that infects one node, infectN(5) is a starter function that infects five nodes.

```
g = makedata()
g = glayout(g)
g = infectN(1)(g)
par(mfrow = c(1, 2))
par(mar = c(0, 0, 0, 0))
plotSIR(g)
g = infectN(5)(g)
plotSIR(g)
```



Exercise: write a starter function generator, infectP, perhaps, that starts the infection with each node having a probability, p of being infected.

3.2 Spreader Functions

A "spreader" function is responsible for stepping the infection along by a single time step. Currently it has the job of infecting new nodes and removing infected nodes (but this may be worth changing at some point).

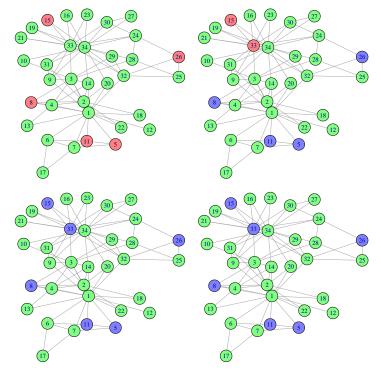
As with a starter function, it takes an SIR graph as input and returns the updated graph. Function generators are again useful for creating spreader functions from parameterised families. We illustrate that now.

The spreadP2 spreader function generator produces spreader functions based on two probabilities. The probability of a susceptible node being infected by an infectious node is pSI in any time step, and the probability of an infected node recovering is pIR in any time step. Hence a spreader function that gives each infectious possibility and recovery possibility an even chance is created thus:

```
sEvens = spreadP2(pSI = 0.5, pIR = 0.5)
```

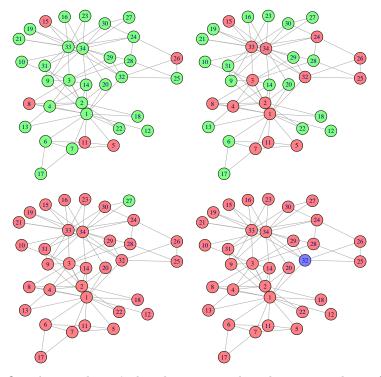
We can now run this spreader a few times and see the change.

```
par(mfrow = c(2, 2))
par(mar = c(0, 0, 0, 0))
plotSIR(g)
g = sEvens(g)
plotSIR(g)
g = sEvens(g)
plotSIR(g)
g = sEvens(g)
plotSIR(g)
g = sEvens(g)
```



Now we repeat, but with a lower recovery probability and a higher infectious probability:

```
sMore = spreadP2(pSI = 0.8, pIR = 0.1)
g = infectN(5)(g)
par(mfrow = c(2, 2))
par(mar = c(0, 0, 0, 0))
plotSIR(g)
g = sMore(g)
plotSIR(g)
g = sMore(g)
plotSIR(g)
g = sMore(g)
plotSIR(g)
g = sMore(g)
plotSIR(g)
```



One thing we haven't done here is to update the time attribute of the graph. This can be done within a loop that repeatedly calls the spreader function, but first we need to look at stopper functions.

3.3 Stopper Functions

A "stopper" function takes an SIR graph and returns TRUE if it decides you should stop running a simulation. Possible stopper functions include stopping after a given time, or stopping if there are no infectious cases. Those two stopper functions are currently implemented.

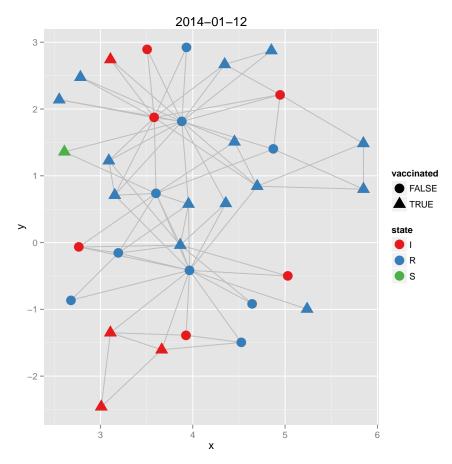
Stopper functions aren't much use on their own, but are designed to be passed into the main looping function stepSim.

4 Running Simulations

The main simulation looping function, stepSim takes a graph, a starter function, a spreader function, and a stopper function, and runs the spreader until the stopper indicates it has finished. The graph is returned with final state and infections and removal times as node attributes.

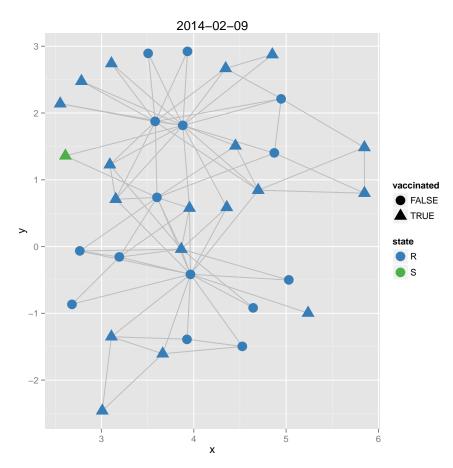
The default start time in makedata is 2014-01-01, so here we'll run for 12 days:

```
g = makedata()
g = stepSim(g, infectN(1), sMore, stopAfter("2014-01-12"))
g$time
## [1] "2014-01-12"
gplotgraph(g)
```



The infection is clearly still going at that point. Now we'll start again and let it run until it has finished and there are no infectious nodes.

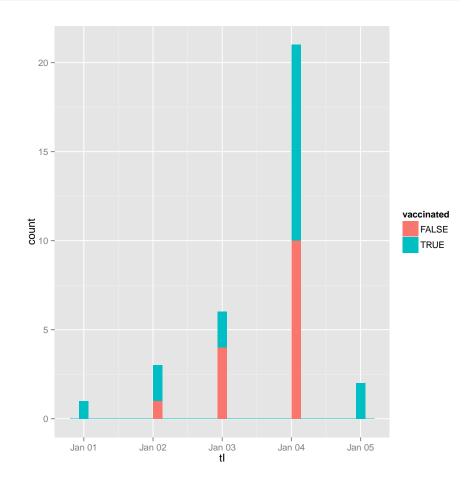
```
g = makedata()
g = stepSim(g, infectN(1), sMore, stopWhenClear)
g$time
## [1] "2014-02-09"
gplotgraph(g)
```



The timing of the state changes can be obtained by getting the vertex data frame. Note that dates aren't kept as dates and so need converting from the number of days from the start of 1970. Then we can, for example, plot a histogram of the infection times, broken down by vaccination status (which in this case is irrelevant):

```
d = get.data.frame(g, "vertices")
d$tI = as.Date(d$tI, "1970-01-01")
d$tR = as.Date(d$tR, "1970-01-01")
head(d)
##
     vaccinated sex age state
                                  X
                                          У
## 1
          FALSE
                  M 19
                            R 3.965 -0.4153 2014-01-03 2014-01-08
                     19
                            R 3.865 -0.0386 2014-01-04 2014-01-12
## 2
           TRUE
                  F
## 3
                  F
                     19
                            R 3.602 0.7354 2014-01-04 2014-01-06
          FALSE
## 4
          FALSE
                  Μ
                     19
                            R 3.195 -0.1572 2014-01-04 2014-01-10
          FALSE
                 M 19
                            R 4.526 -1.4961 2014-01-04 2014-01-12
## 5
```

```
## 6 TRUE F 21 R 3.108 -1.3501 2014-01-04 2014-01-13
ggplot(d, aes(x = tI, fill = vaccinated)) + geom_histogram()
## stat_bin: binwidth defaulted to range/30. Use 'binwidth = x' to
adjust this.
## Warning: position_stack requires constant width: output may be
incorrect
```



5 Design

This code has been written with a fairly strong functional program methodology, hence the sprinkling of function generators here and there. The design of the **stepSim** function has been influenced by the idea of *dependency injection*, so that the function code represents the minimal amount of code needed for any

generic simulation, and specific behaviour is passed in via spreader and stopper functions.

This design also means that spreader and stopper functions can be easily run and hence tested outside any looping code.

6 To-do

Some ideas:

- make the SIR graph objects a new class
- separate the infectious and removal aspects of the spreader function
- write some functions to plot the number of S, I, R cases over time