

Assembling my Network

The ecology is theoretical but the fun is real.

Network basics with R and igraph (part I of III)

Posted on [May 29, 2013](#)

In this post I am going to go over some of the basics that I have learned about networks in R by using the igraph package. I originally intended to have this as a reference for myself, but decided to turn it into more of a tutorial for others who, like me, are just getting started with programming in R and interested in networks and network theory.

```

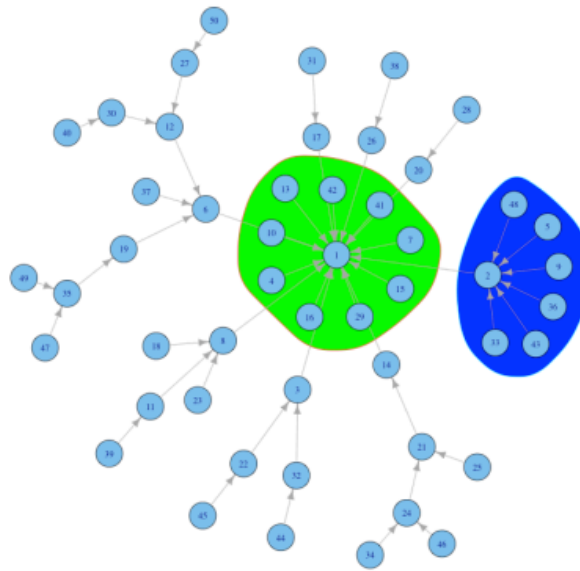
1 #####
2 #####
3 #####
4 #####          PART I: NETWORKS
5 #####
6 #####
7 #####
8 # the igraph library is good for network based analyses
9 library(igraph)
10
11 # First lets build a network using basic formulas:
12
13 graph.onelink<-graph.formula(A-+B)
14
15 # This gives us a two species network (A and B) with one link (represented by A-+B).
16 # We can visualize our simple 2 species network with plot.igraph().
17
18 plot.igraph(graph.onelink)
19
20 # Using graph.formula() we can create any graph we want, as long as we are willing to
21
22 graph.foodchain<-graph.formula(A-+C,B-+C,C-+D)
23
24 # and plot it:
25
26 plot.igraph(graph.foodchain)
27 #A and B are eaten by C while D eats C
28
29 #igraph has a function for generating random networks of varying size and connectance
30
31 graph.random.gnp<-erdos.renyi.game(n=20,p.or.m=.5,type="gnp",directed=T)
32 plot.igraph(graph.random.gnp)
33
34 # We can also change the layout of the graph, here we will plot the nodes in a circle:
35 plot.igraph(graph.random.gnp,layout=layout.circle)
36
37 # Here we have created a random directed graph with 20 species ("n") and a connectance
38 # By setting "type='gnp'" we tell the function to assign links with the probability
39 # Similarly we can set the number of links that we want in the system to a value "m"
40

```

```

41 graph.random.gnm<-erdos.renyi.game(n=20,p.or.m=100,type="gnm",directed=T)
42 plot.igraph(graph.random.gnm)
43
44 # Here the number of links in the network is set to 100, and they are assigned unif
45
46 # Rather than being truly random, many real networks exhibit some type of organizat
47 # To model scale free networks Barabasi and Albert developed the preferential attac
48
49 # In igraph we can use the barabasi.game() function:
50 graph.barabasi.1<-barabasi.game(n=50,power=1)
51
52 # For this graph I will introduce some new plotting tools to specify layout and vert
53
54 plot.igraph(graph.barabasi.1,
55 layout=layout.fruchterman.reingold,
56 vertex.size=10,          # sets size of the vertex, default is 15
57 vertex.label.cex=.5,     # size of the vertex label
58 edge.arrow.size=.5       # sets size of the arrow at the end of the edge
59 )
60
61 # there are a number of different plotting parameters see
62 #?igraph.plotting
63 #for details
64
65 plot.igraph(graph.barabasi.1,
66 layout=layout.fruchterman.reingold,
67
68 vertex.size=10,
69 vertex.label.cex=.5,
70 edge.arrow.size=.5,
71 mark.groups=list(c(1,7,4,13,10,16,15,41,42,29),
72 c(2,48,5,36,43,33,9)), # draws polygon around nodes
73 mark.col=c("green","blue")
74 )

```

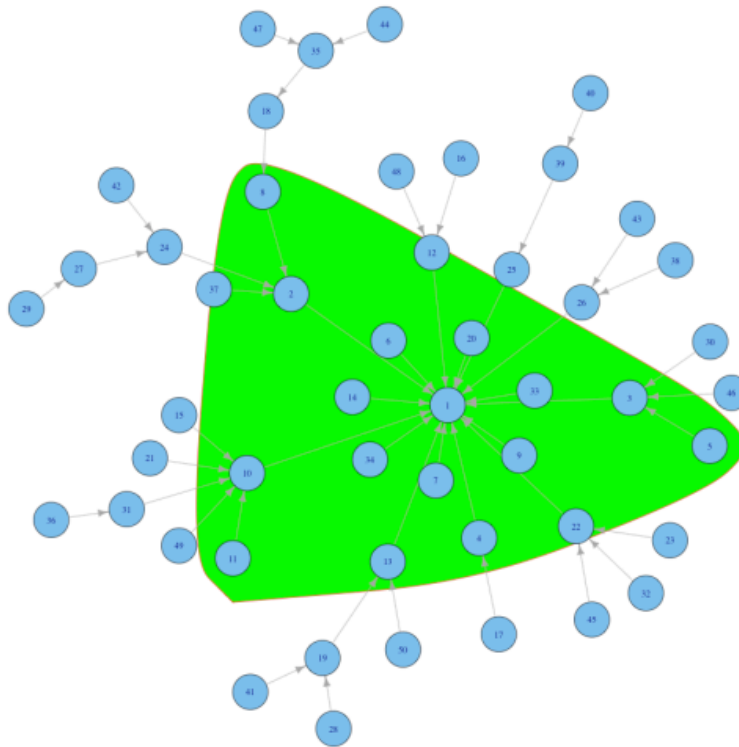


Nodes 1 and 2 are highlighted as hubs in this network

```

1 # In the above plot a green and blue polygon are used to highlight nodes 1 and 2 as
2
3 # We can use a community detection algorithm to determine the most densely connected
4
5 barabasi.community<-walktrap.community(graph.barabasi.1)
6
7 # This algorithm uses random walks to find the most densely connected subgraphs.
8
9 members<-membership(barabasi.community)
10 # The members() function picks out the membership vector (list of nodes in the most
11
12 par(mar=c(.1,.1,.1,.1)) # sets the edges of the plotting area
13 plot.igraph(graph.barabasi.1,
14 layout=layout.fruchterman.reingold,
15 vertex.size=10,
16 vertex.label.cex=.5,
17 edge.arrow.size=.5,
18 mark.groups=list(members),
19 mark.col="green"
20 )

```



A preferential attachment model network with a highlighted walktrap community.

```

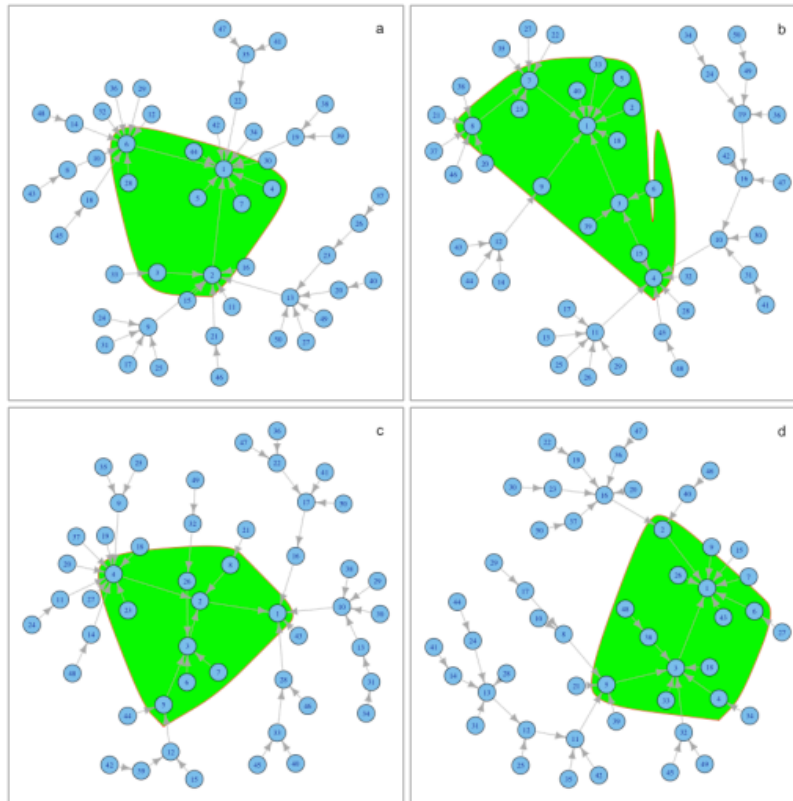
1 # With the above plot the group with the green polygon surrounding it is the nodes :
2
3 # Now we will play around with the "power" argument to see how that impacts the graph
4 # We will generate 4 networks with preferential attachment at varying levels.
5 barabasi.game.2<-barabasi.game(n=50,power=.75)
6 barabasi.game.3<-barabasi.game(n=50,power=.5)
7 barabasi.game.4<-barabasi.game(n=50,power=.25)
8 barabasi.game.5<-barabasi.game(n=50,power=0)
9
10 # These can be organized into a list for convenience.
11 barabasi.graphs<-list(barabasi.game.2,barabasi.game.3,barabasi.game.4,barabasi.game.5)
12
13 # Now lets use community detection, this time with the walktrap algorithm.
14 bg.community.list<-lapply(barabasi.graphs,walktrap.community)
15 bg.membership.list<-lapply(bg.community.list,membership)
16
17 txt<-c("a","b","c","d") # vector for labeling the graphs
18
19 # Plot these four graphs in one window with:
20 par(mfrow=c(2,2),mar=c(.2,.2,.2,.2))
21 # The for loop here plots each graph in the list one by one into the window prepared
22 for(i in 1:4){
23   plot.igraph(barabasi.graphs[[i]],
24     layout=layout.fruchterman.reingold,
25     vertex.size=10,
26     vertex.label.cex=.5,
27     edge.arrow.size=.5,
28     mark.groups=list(bg.membership.list[[i]]),

```

```

29 mark.col="green",
30 frame=T # the frame argument plots a box around the graph
31 )
32 text(1,1,txt[i]) # calls from the vector to label the graph, adds to the
33 }
34
35 # Later we will look at the properties of these graphs to see exactly how they are c

```



Four preferential attachment model networks with walktrap communities highlighted

```

1 # Graphs can also be visualized as a matrix, the adjacency matrix. The adjacency matrix
2
3 # The function get.adjacency() converts a graph into a matrix.
4 barabasi.adjacency<-get.adjacency(graph.barabasi.1)
5
6 # A number of other functions will use the adjacency matrix to calculate different metrics
7
8 # And we can get a graph from the adjacency matrix with graph.adjacency(). Often the
9 graph.adjacency(barabasi.adjacency)
10
11 # Similarly we can arrange the information in a number of other ways
12
13 # Edgelist
14 barabasi.edgelist<-get.edgelist(graph.barabasi.1)
15
16 # Adjacency list
17 barabasi.adjlist<-get.adjlist(graph.barabasi.1,mode="all")
18
19 # Dataframe
20 barabasi.data.frame<-get.data.frame(graph.barabasi.1,what="edges")

```

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4 Responses to *Network basics with R and igraph (part I of III)*

**Tony says:**

June 12, 2013 at 10:07 AM

A very helpful post!

[Reply](#)**Jon Borrelli says:**

June 12, 2013 at 1:30 PM

thanks!

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**Sai Krishna says:**

August 23, 2014 at 10:42 PM

Thanks a lot

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