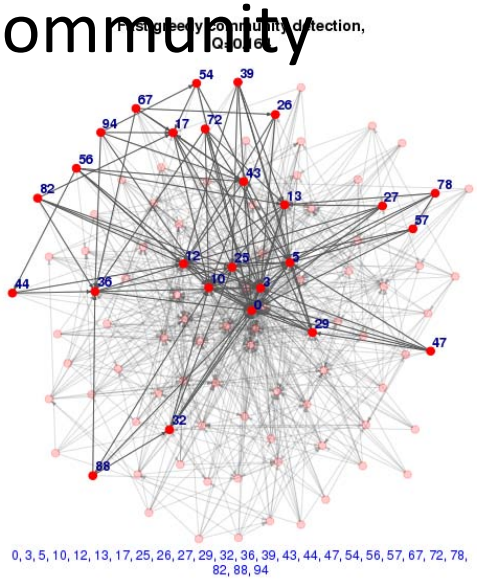
Complex Networks
tools for analyzing networks (R+igraph)

2011.10.17

igraph http://cneurocvs.rmki.kfki.hu/igraph/

- igraph is a free software package for creating and manipulating undirected and directed graphs. It includes implementations for classic graph theory problems like minimum spanning trees and network flow, and also implements algorithms for some recent network analysis methods, like community structure search.



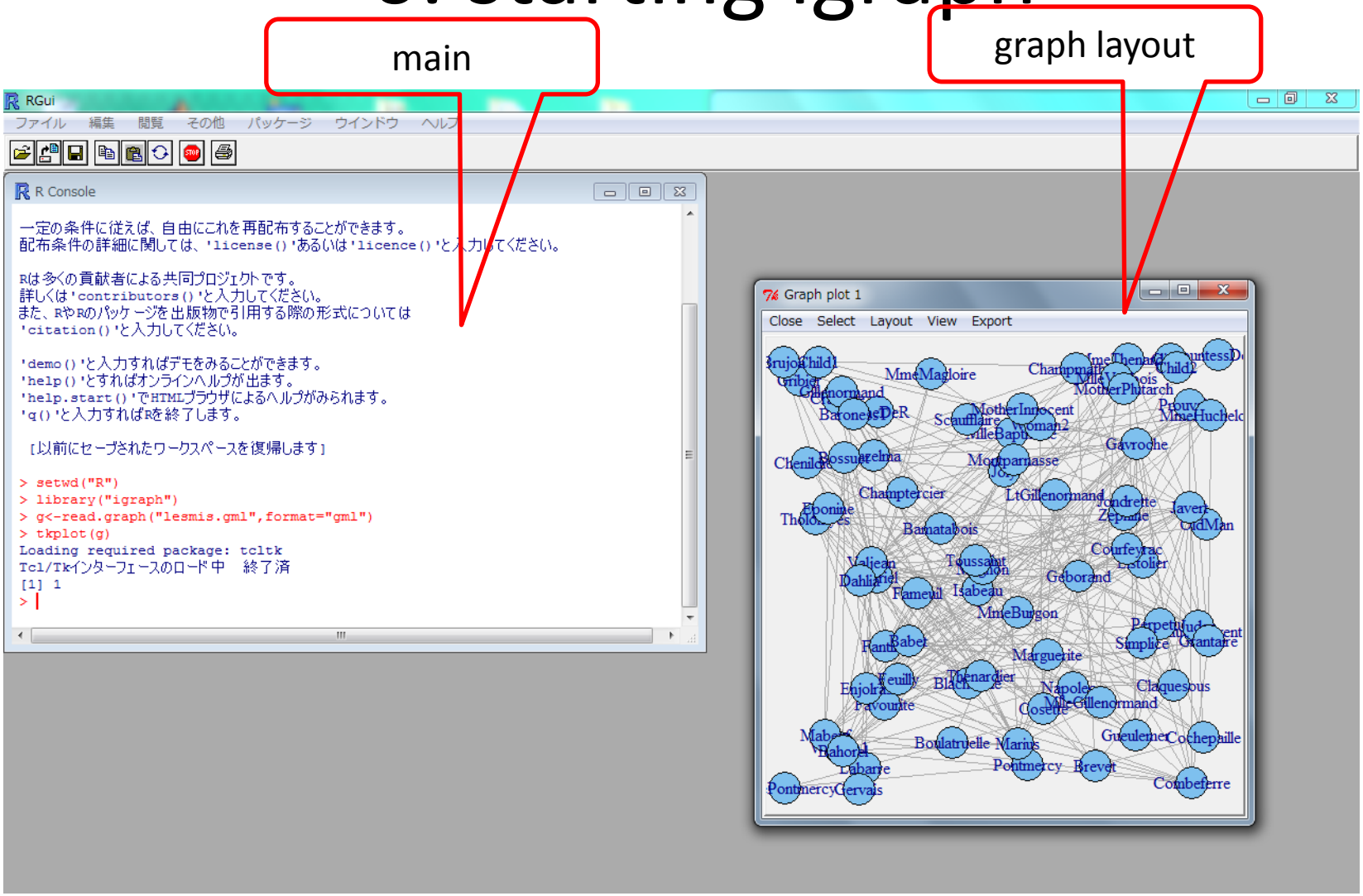
tutorials

- tutorials of R
  - http://cran.r-project.org/other-docs.html (many tutorials in English and other languages)
- tutorial of igraph
  - http://igraph.sourceforge.net/igraphbook/ (English, under development)

procedure for analyzing network

1. create graph object
2. layout the network
3. ranking
4. metrics
5. community detection
6. export

0. starting igraph



1. create graph object

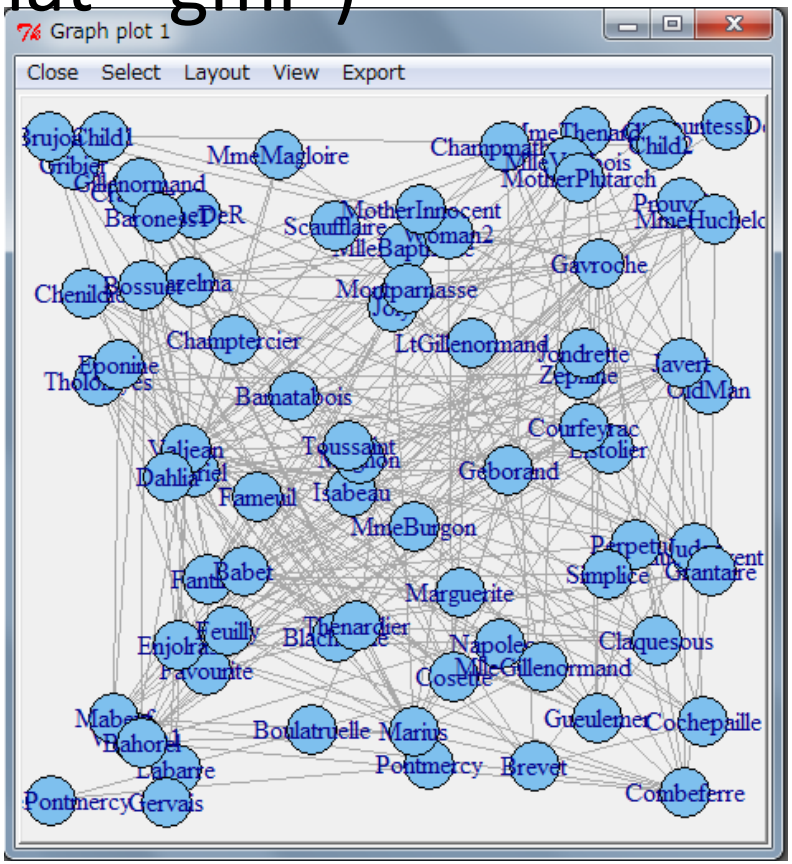
- library("igraph")
- setwd("R")
- g<-read("lesmis.gml", format="gml")
- tkplot(g)

start igraph

set directory

create graph

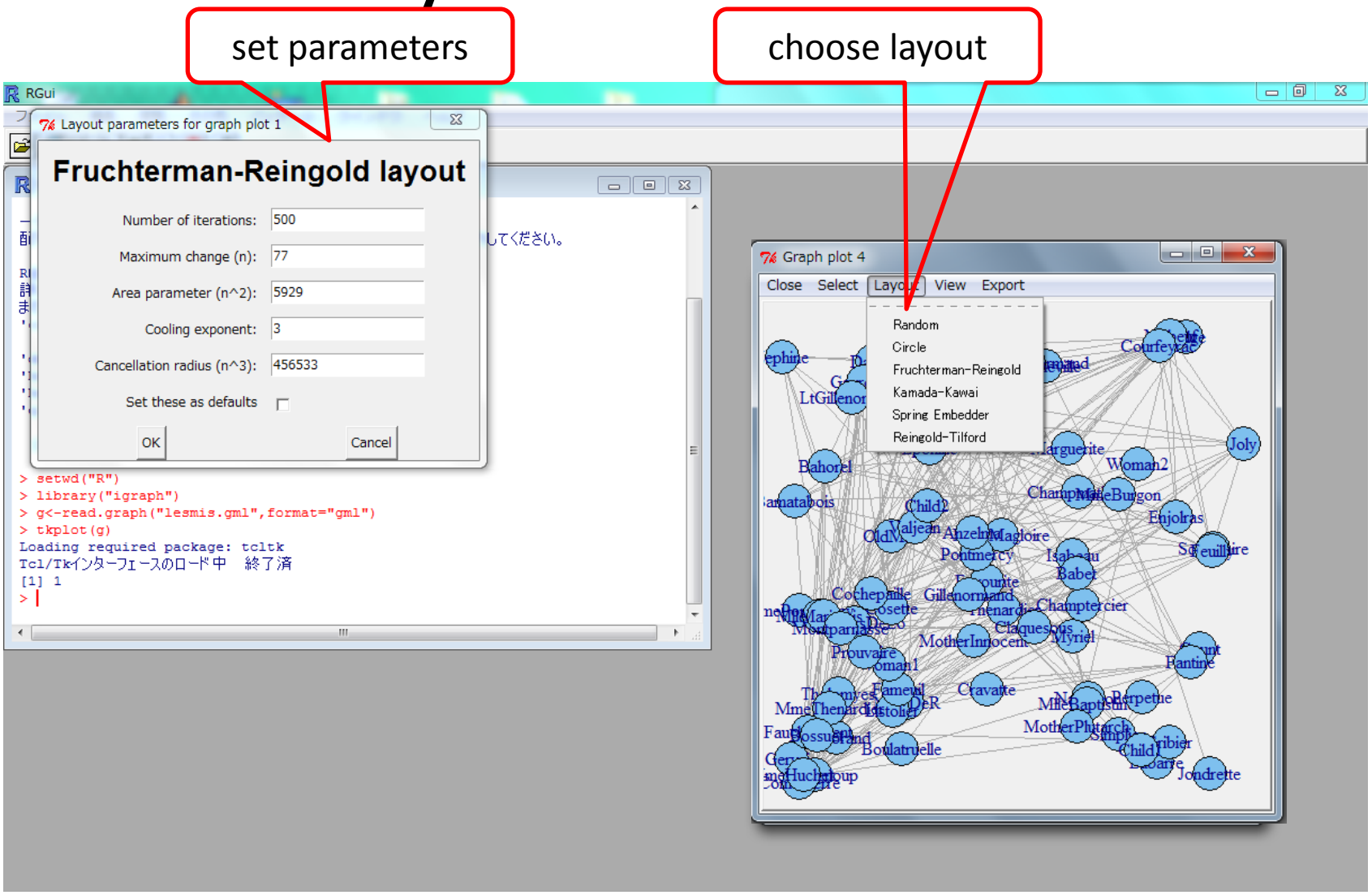
open graph window



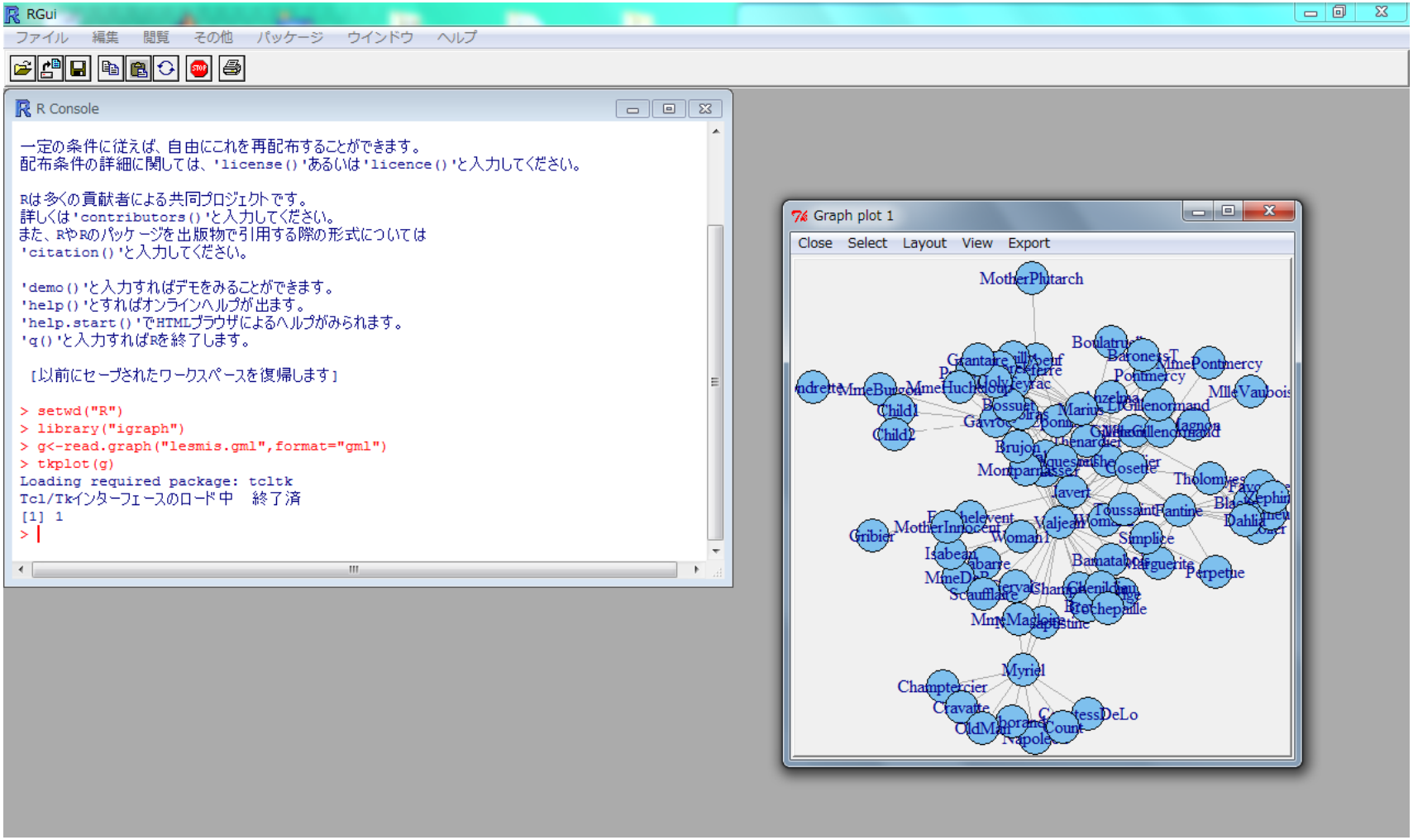
lesmis.gml is available at

Mark Newman's Website

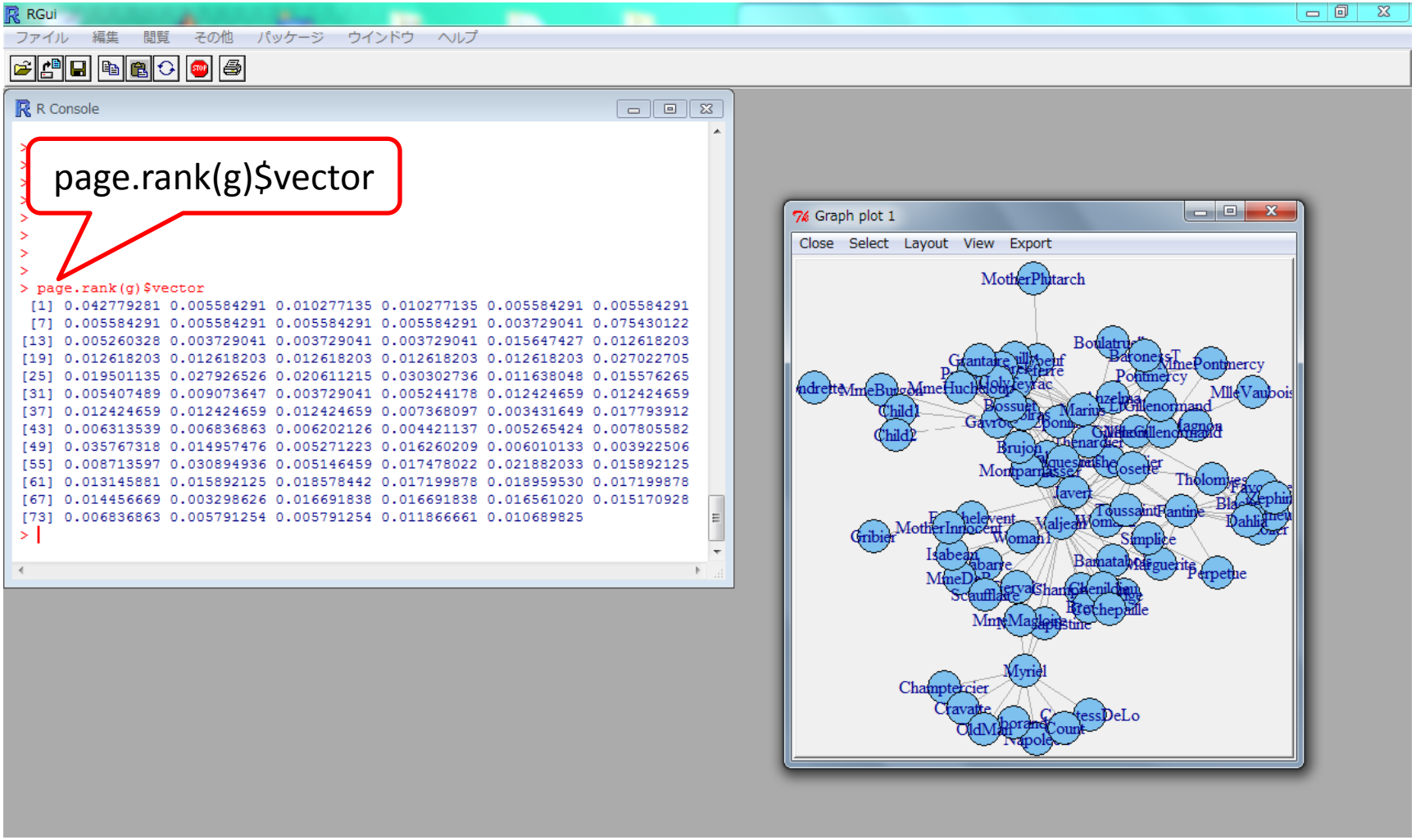
2. layout the network



2. layout the network

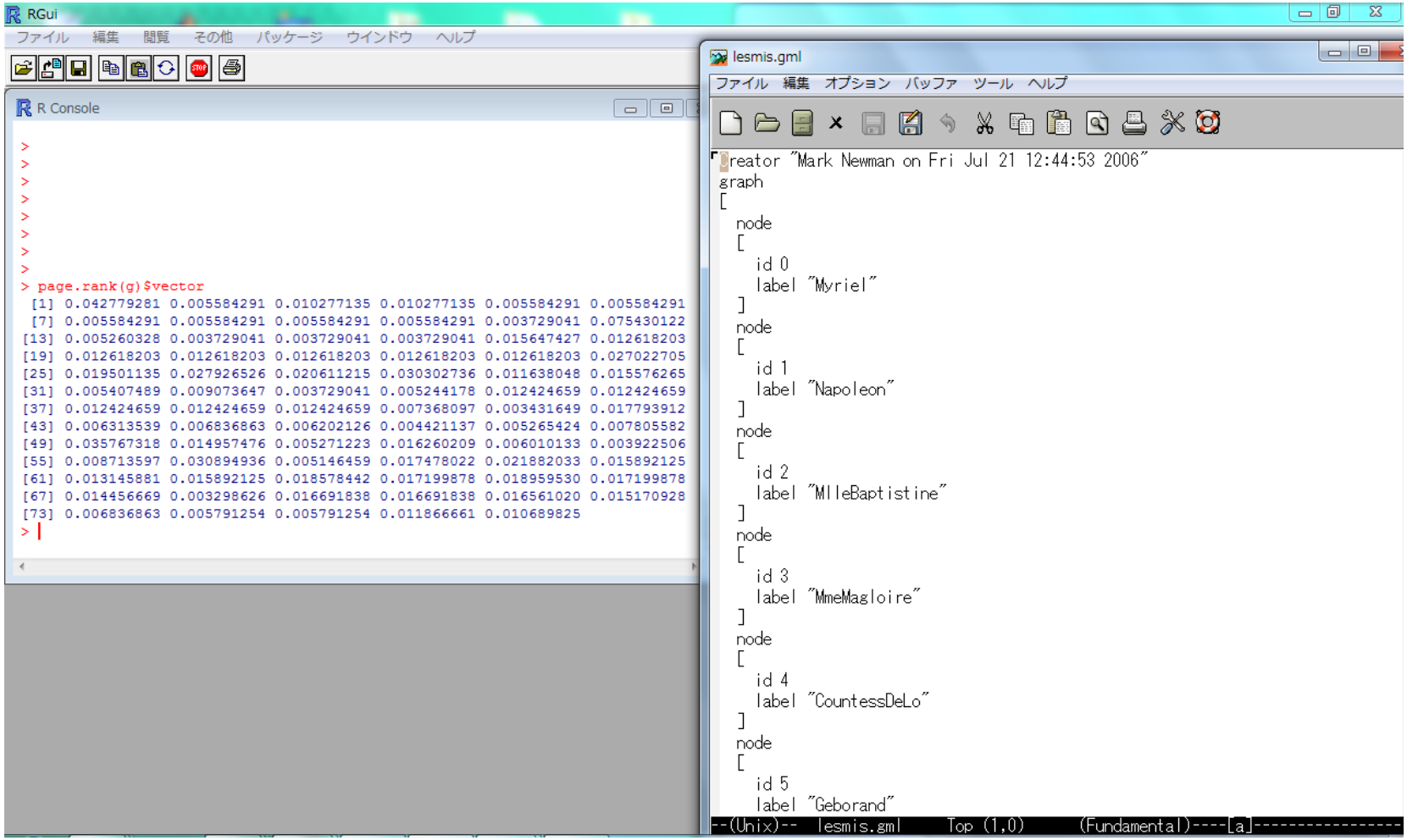


3. ranking


RGui
ファイル 編集 閲覧 その他 パッケージ ウインドウ ヘルプ
R Console
page.rank(g)$vector
> page.rank(g)$vector
[1] 0.042779281 0.005584291 0.010277135 0.010277135 0.005584291 0.005584291
[7] 0.005584291 0.005584291 0.005584291 0.005584291 0.003729041 0.075430122
[13] 0.005260328 0.003729041 0.003729041 0.003729041 0.015647427 0.012618203
[19] 0.012618203 0.012618203 0.012618203 0.012618203 0.012618203 0.027022705
[25] 0.019501135 0.027926526 0.020611215 0.030302736 0.011638048 0.015576265
[31] 0.005407489 0.009073647 0.003729041 0.005244178 0.012424659 0.012424659
[37] 0.012424659 0.012424659 0.012424659 0.007368097 0.003431649 0.017793912
[43] 0.006313539 0.006836863 0.006202126 0.004421137 0.005265424 0.007805582
[49] 0.035767318 0.014957476 0.005271223 0.016260209 0.006010133 0.003922506
[55] 0.008713597 0.030894936 0.005146459 0.017478022 0.021882033 0.015892125
[61] 0.013145881 0.015892125 0.018578442 0.017199878 0.018959530 0.017199878
[67] 0.014456669 0.003298626 0.016691838 0.016691838 0.016561020 0.015170928
[73] 0.006836863 0.005791254 0.005791254 0.011866661 0.010689825
Graph plot 1
Close Select Layout View Export
MotherPlutarch
Myriel
Champtercier
Cravatte
Napoleon
Count
OldMan
Child1
Child2
Gribier
Simplice
Perpetue
Cosette
Javert
Valjean
Woman1
Woman2
Isabeau
Scaufflaire
Brujon
Gavroche
Bossuet
Marius
Pontmercy
MlleVaubois
Tholomyes
ToussaintFantine
Dahlia

3.ranking

- original gml file contain labels



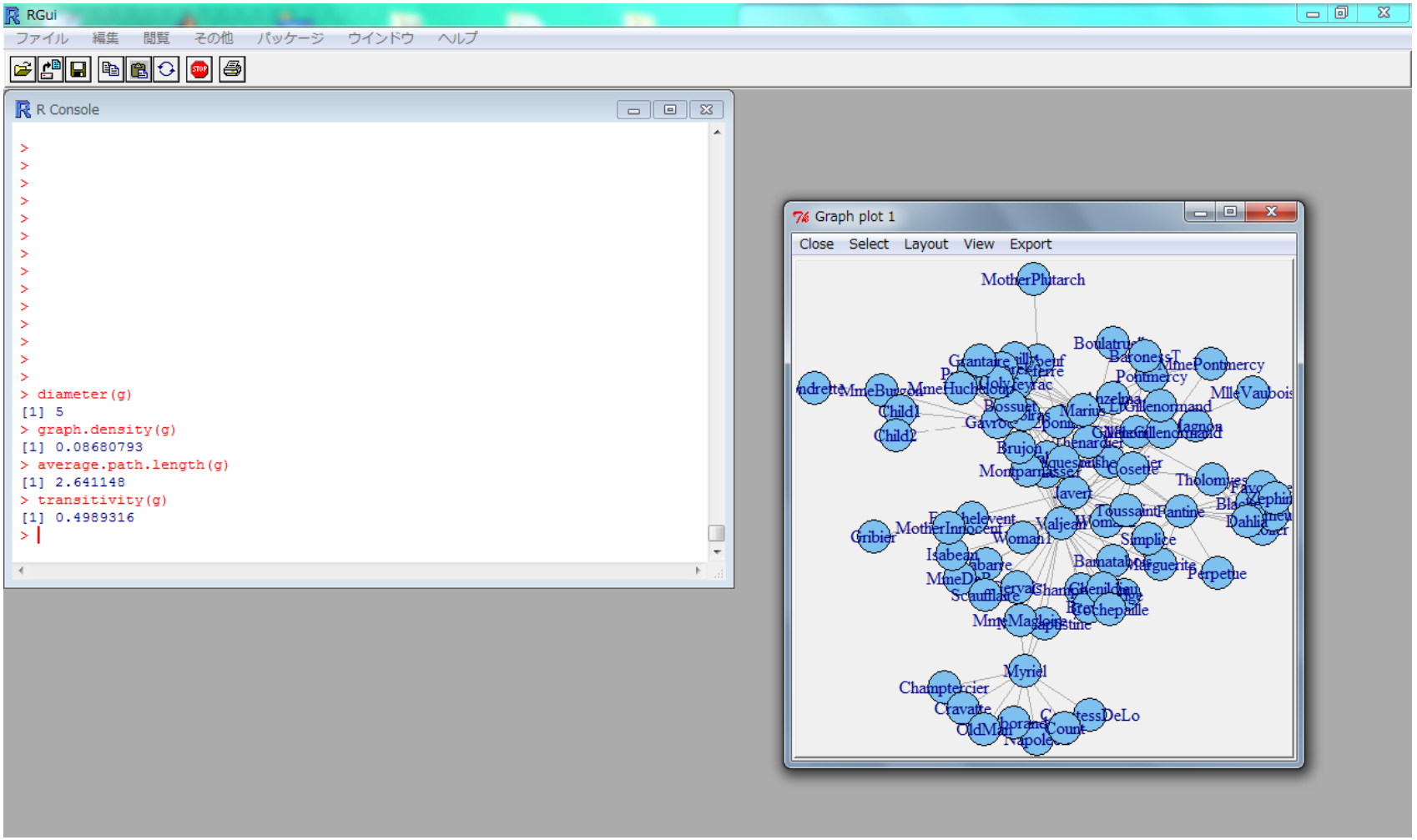
4. metrics

- diameter(g)
- graph.density(g)
- average.path.length(g)
- transitivity(g)

clustering coefficient

- help
  - ??rank
  - help(“page.rank”)

4. metrics


RGui
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R Console
> diameter(g)
[1] 5
> graph.density(g)
[1] 0.08680793
> average.path.length(g)
[1] 2.641148
> transitivity(g)
[1] 0.4989316
Graph plot 1
Close Select Layout View Export
MotherPlutarch
Myriel
Champtercier
Cravatte
Napoleon
Count
Gribier
MotherInnocent
Valjean
Woman1
Simplice
Perpetue
Gillenormand
Marius
Cosette
Javert
Fantine
Toussaint
Child1
Child2
Brujon
Bossuet
Pontmercy
MlleVaubois

5. community detection

modularity optimization

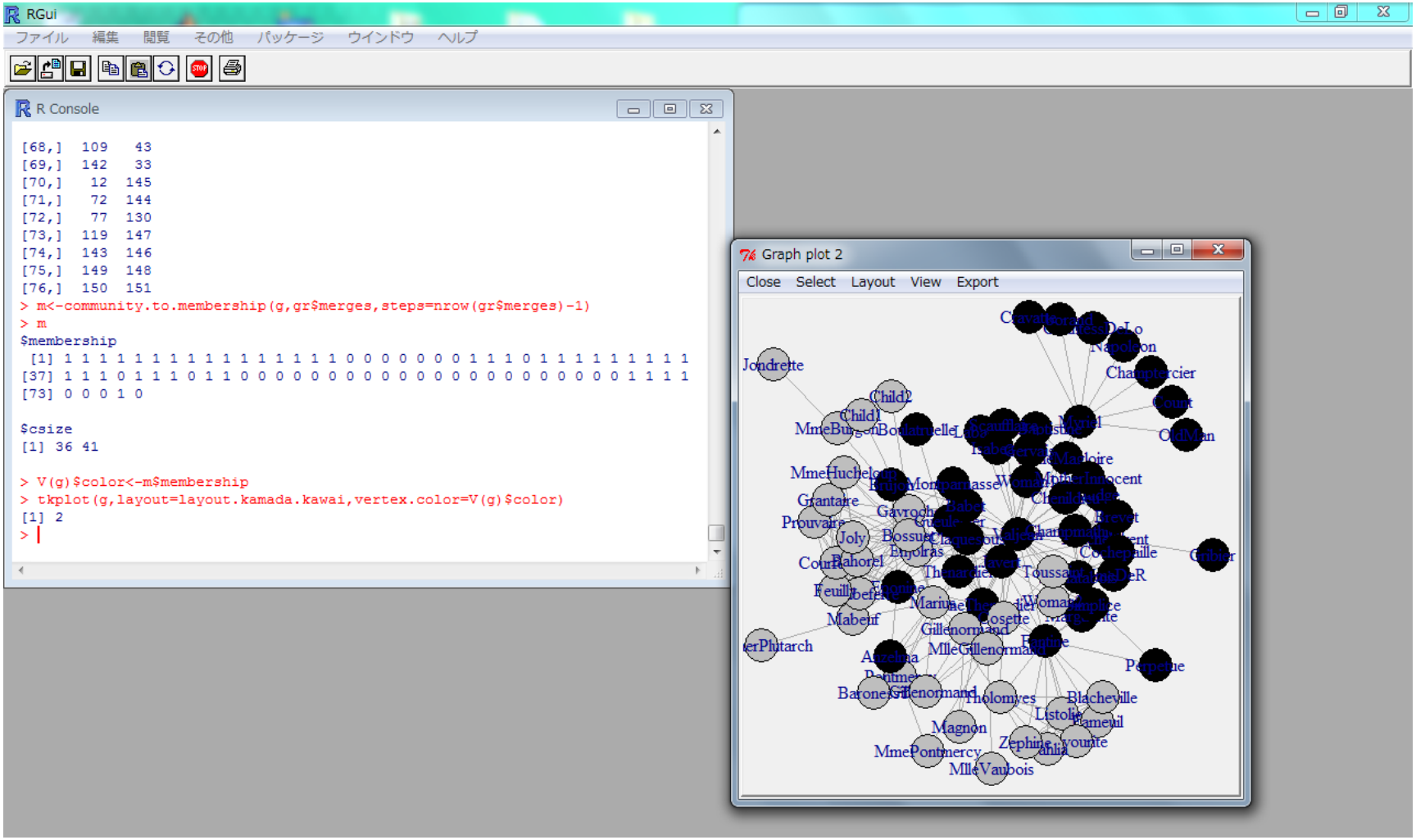
store membership and size

```
> gr<-fastgreedy.community(g)
> m<-community.to.membership(g,gr$merges,steps=nrow(gr$merges)-1)
> m
$membership
 [1] 0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 1 0 0 1 0 1 0 1 1 1 1 1 1 1 1 1 1 1
$csize
[1] 17 17
> V(g)$color<-m$membership
> V(g)$color
 [1] 0 0 0 0 0 0 0 0 1 0 0 0 0 0 1 1 0 0 1 0 1 0 1 1 1 1 1 1 1 1 1 1 1
> tkplot(g,layout=layout.kamada.kawai,vertex.color=V(g)$color)
[1] 2
>
```

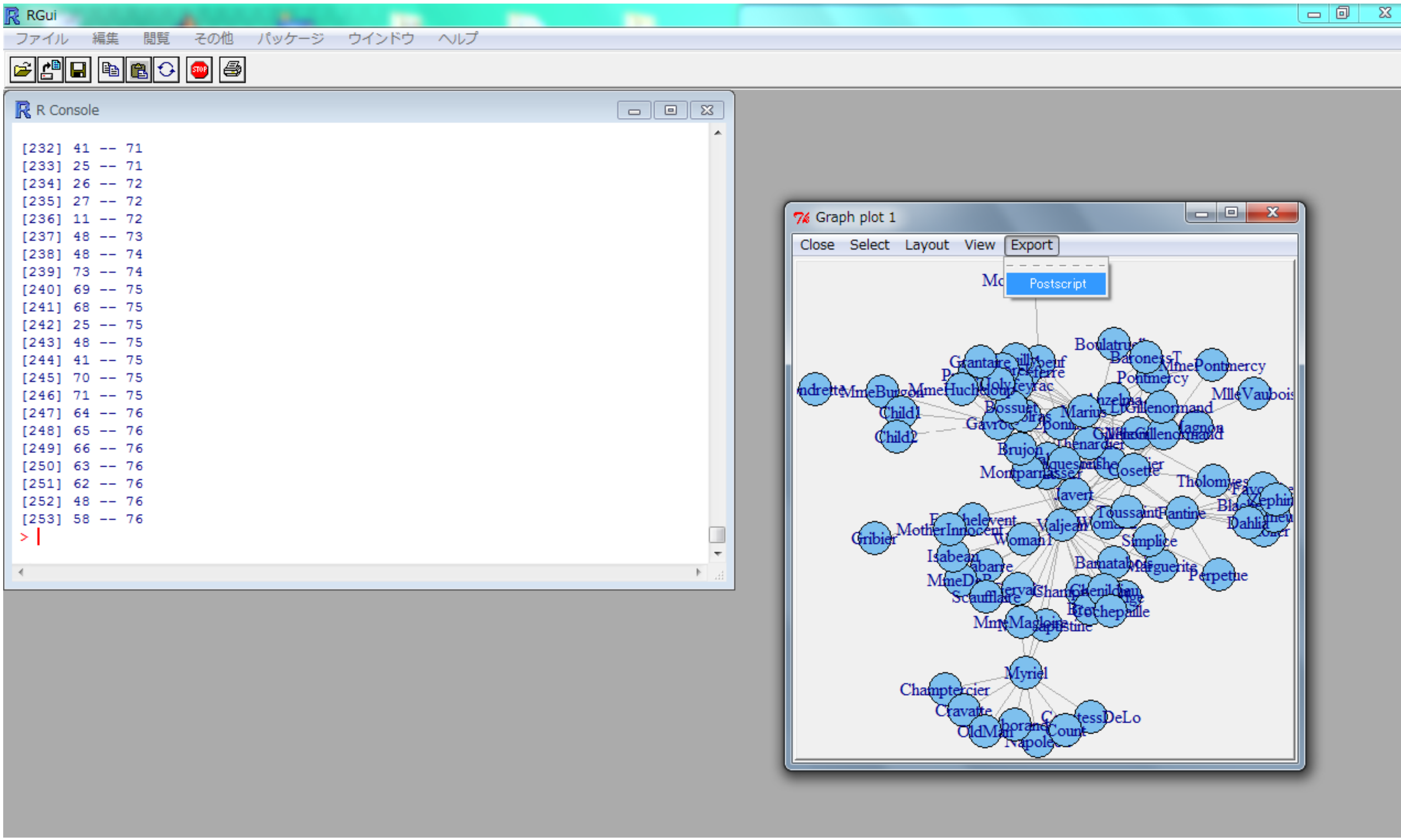
store membership

visualize network

5. community detection



6. export

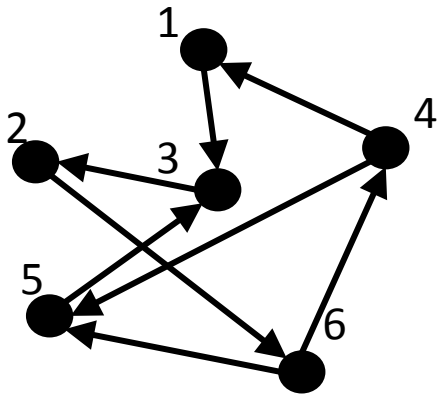

RGui
ファイル 編集 閲覧 その他 パッケージ ウインドウ ヘルプ
R Console
[232] 41 -- 71
[233] 25 -- 71
[234] 26 -- 72
[235] 27 -- 72
[236] 11 -- 72
[237] 48 -- 73
[238] 48 -- 74
[239] 73 -- 74
[240] 69 -- 75
[241] 68 -- 75
[242] 25 -- 75
[243] 48 -- 75
[244] 41 -- 75
[245] 70 -- 75
[246] 71 -- 75
[247] 64 -- 76
[248] 65 -- 76
[249] 66 -- 76
[250] 63 -- 76
[251] 62 -- 76
[252] 48 -- 76
[253] 58 -- 76
>
Graph plot 1
Close Select Layout View Export
Postscript
Myriel
Champtercier
Cravatte
Gribier
Perpetue
Simplice
Child1
Child2
Brujon
Marius
Cosette
Javert
Tholomyes
Dahlia
MlleVaubois
Isabeau
Scaufflaire
Woman1
MotherInnocent

create from adjacency matrix(1)

- from adjacency matrix

```
> a <- matrix(c(0,0,0,1,0,0,
                0,0,1,0,0,0,
                1,0,0,0,1,0,
                0,0,0,0,0,1,
                0,0,0,1,0,1,
                0,1,0,0,0,0),nrow=6,byrow=TRUE)
```

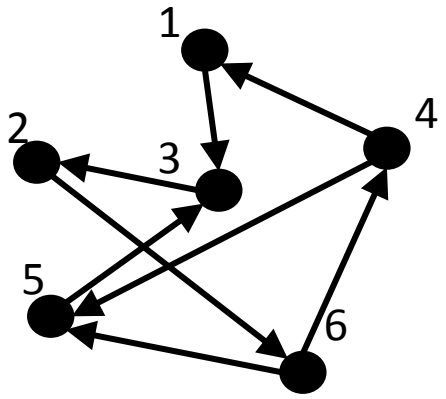
$$A = \begin{array}{c} \\ i \\ \\ \\ \\ \\ \\ \end{array} \begin{array}{c} \begin{array}{cccccc} & & & j & & \end{array} \\ \begin{pmatrix} 0 & 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 & 0 \\ 1 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 1 \\ 0 & 0 & 0 & 1 & 0 & 1 \\ 0 & 1 & 0 & 0 & 0 & 0 \end{pmatrix} \end{array}$$



create from adjacency matrix(2)

```
> ga <- graph.adjacency(t(a))
> ga
Vertices: 6
Edges: 8
Directed: TRUE
Edges:
[0] 0 -> 2
[1] 1 -> 5
[2] 2 -> 1
[3] 3 -> 0
[4] 3 -> 4
[5] 4 -> 2
[6] 5 -> 3
[7] 5 -> 4
```

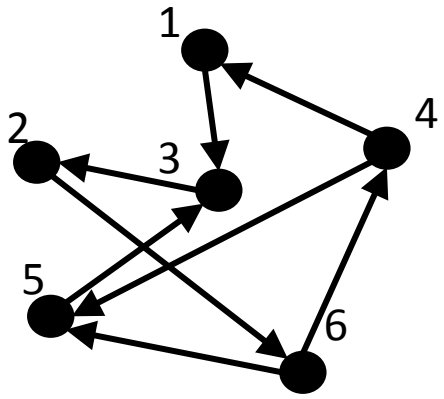
transposition



In igraph,
ID starts from 0 &
direction is from i to j

create from edge list

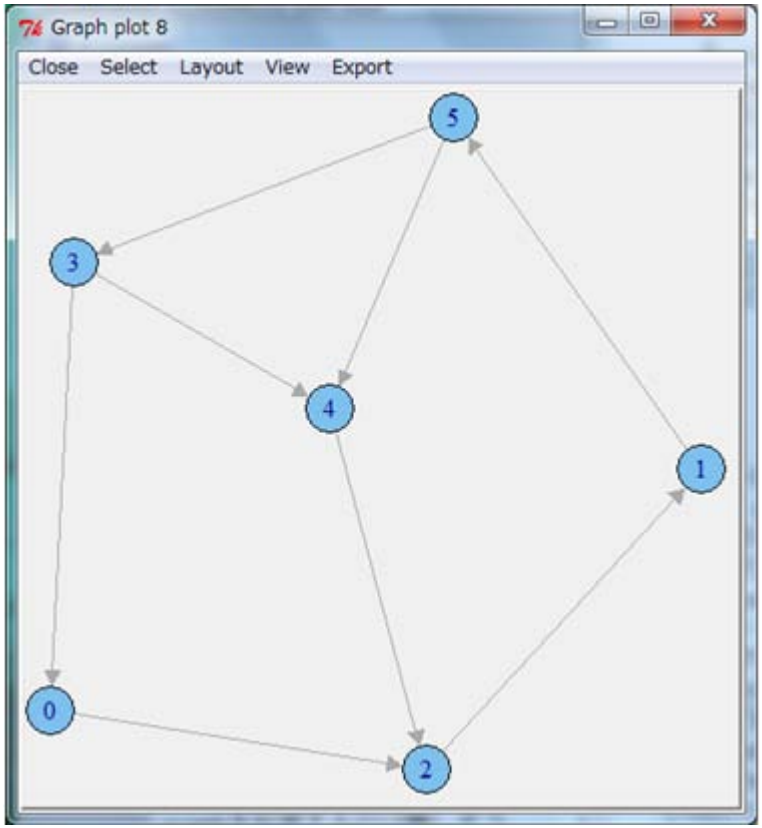
```
> el <-
    matrix(c(0,2,1,5,2,1,3,0,3,4,4,2,5,
    3,5,4),nc=2,byrow=TRUE)
> el
     [,1] [,2]
[1,]    0    2
[2,]    1    5
[3,]    2    1
[4,]    3    0
[5,]    4    2
[6,]    5    3
[7,]    5    4
```




```
> gb<-graph.edgelist(el)
> gb
Vertices: 6
Edges: 8
Directed: TRUE
Edges:
[0] 0 -> 2
[1] 1 -> 5
[2] 2 -> 1
[3] 3 -> 0
[4] 3 -> 4
[5] 4 -> 2
[6] 5 -> 3
[7] 5 -> 4
```

layout the network

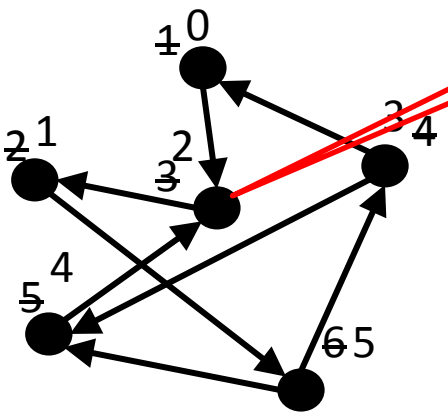
- > tkplot(ga,layout=layout.kamada.kawai)
  - choose layout (random, circle, Fruchterman-Reingold, Kamada-Kawai)
  - deform graph
  - export (Postscript)



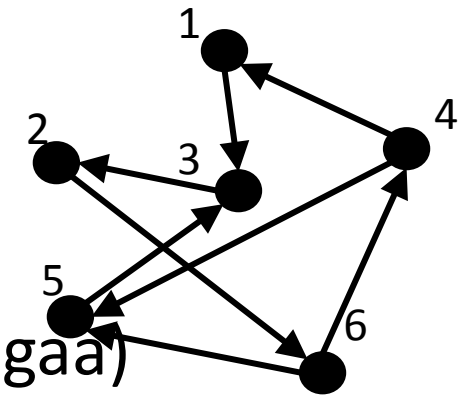
ranking

- PageRank: importance of vertices
  - the probability that a random walker will visit

```
> page.rank(ga)$vector
[1] 0.07337065 0.21643820 0.22522142
    0.11381330 0.16218395 0.20897247
```



metrics (1)



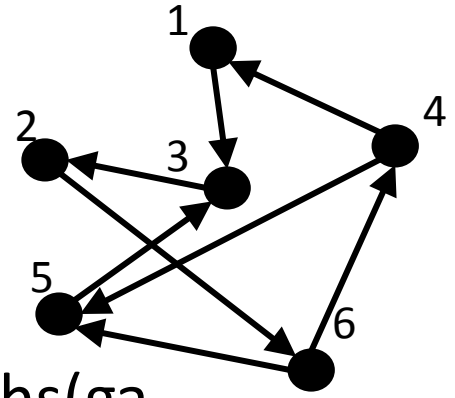
> cocitation(gaa)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]
[1,]	0	0	0	0	1	0
[2,]	0	0	0	0	0	0
[3,]	0	0	0	0	0	0
[4,]	0	0	0	0	1	0
[5,]	1	0	0	1	0	0
[6,]	0	0	0	0	0	0

> bibcoupling(gaa)

	[,1]	[,2]	[,3]	[,4]	[,5]	[,6]
[1,]	0	0	0	0	1	0
[2,]	0	0	0	0	0	0
[3,]	0	0	0	0	0	0
[4,]	0	0	0	0	0	1
[5,]	1	0	0	0	0	0
[6,]	0	0	0	1	0	0

metrics (2)



- undirected

```
> shortest.paths(ga)
     [,1] [,2] [,3] [,4] [,5] [,6]
[1,]    0    2    1    1    2    2
[2,]    2    0    1    2    2    1
[3,]    1    1    0    2    1    2
[4,]    1    2    2    0    1    1
[5,]    2    2    1    1    0    1
[6,]    2    1    2    1    1    0
```

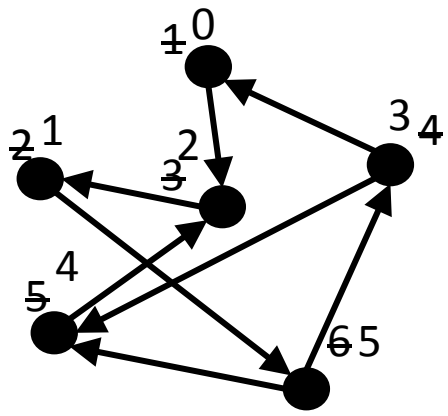
- directed

```
> shortest.paths(ga,
    mode="out")
                 to
     [,1] [,2] [,3] [,4] [,5] [,6]
[1,]    0    2    1    4    4    3
from[2,]    3    0    3    2    2    1
[3,]    4    1    0    3    3    2
[4,]    1    3    2    0    1    4
[5,]    5    2    1    4    0    3
[6,]    2    3    2    1    1    0
```

metrics (3)

```
> average.path.length(ga)
[1] 2.433333
> average.path.length(ga,directed=FALSE)
[1] 1.466667
```

directed




```
> get.all.shortest.paths(ga,0)
[[1]]
[1] 0
[[2]]
[1] 0 2 1
[[3]]
[1] 0 2
[[4]]
[1] 0 3
[[5]]
[1] 0 3 4
[[6]]
[1] 0 2 4
[[7]]
[1] 0 3 5
```

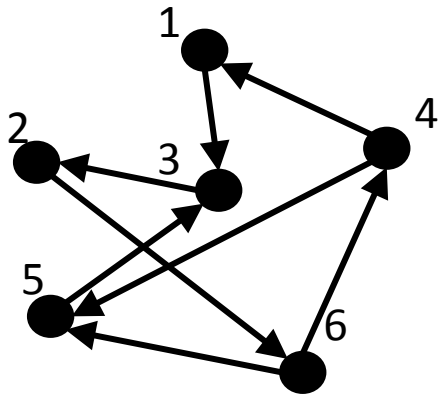
shortest paths from 0

metrics (4)

```
> is.connected(ga)
[1] TRUE

> no.clusters(ga)
[1] 1

> clusters(ga)
$membership
[1] 0 0 0 0 0 0
$csize
[1] 6
$no
[1] 1
```



metrics (5)

> graph.density(ga)

[1] 0.2666667

$$\rho = \frac{m}{n(n-1)} = \frac{8}{6 \cdot 5}$$