

DBA5104 igraph Tutorial in R

This markdown is a short tutorial on how to use igraph in R. When using igraph, R is preferred over python as R is more stable and has more comprehensive functions.

You can check out the documentation: <https://igraph.org/r/doc/>

Initialization

```
#install.packages("igraph")
```

```
library(igraph)
```

```
##
```

```
## Attaching package: 'igraph'
```

```
## The following objects are masked from 'package:stats':
```

```
##
```

```
##      decompose, spectrum
```

```
## The following object is masked from 'package:base':
```

```
##
```

```
##      union
```

```
library(igraphdata)
```

```
library(sand)
```

```
## Warning: package 'sand' was built under R version 4.3.2
```

```
##
```

```
## Statistical Analysis of Network Data with R, 2nd Edition
```

```
## Type in C2 (+ENTER) to start with Chapter 2.
```

Creating Network Graphs

```
# For small, toy graphs, the function graph_from_literal can be used,  
# specifying the edges in a symbolically literal manner
```

```
g <- graph_from_literal(1-2, 1-3, 2-3, 2-4, 3-5, 4-5, 4-6, 4-7, 5-6, 6-7)
```

Number of vertices

V(g)

```
## + 7/7 vertices, named, from ede8fcc:  
## [1] 1 2 3 4 5 6 7
```

Number of edges

E(g)

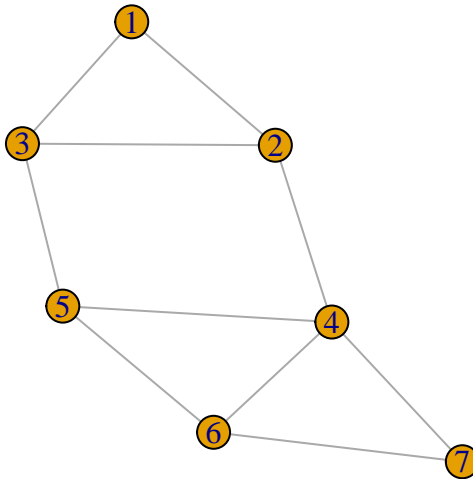
```
## + 10/10 edges from ede8fcc (vertex names):  
## [1] 1--2 1--3 2--3 2--4 3--5 4--5 4--6 4--7 5--6 6--7
```

```
# same information, combined and in a slightly different format, is recovered  
print_all(g)
```

```
## IGRAPH ede8fcc UN-- 7 10 --  
## + attr: name (v/c)  
## + edges (vertex names):  
## 1 -- 2, 3  
## 2 -- 1, 3, 4  
## 3 -- 1, 2, 5  
## 4 -- 2, 5, 6, 7  
## 5 -- 3, 4, 6  
## 6 -- 4, 5, 7  
## 7 -- 4, 6
```

Visual representation

plot(g)



Adjacency matrix

```
as_adjacency_matrix(g)
```

```
## 7 x 7 sparse Matrix of class "dgCMatrix"
##   1 2 3 4 5 6 7
## 1 . 1 1 . . . .
## 2 1 . 1 1 . . .
## 3 1 1 . . 1 . .
## 4 . 1 . . 1 1 1
## 5 . . 1 1 . 1 .
## 6 . . . 1 1 . 1
## 7 . . . 1 . 1 .
```

Graph Concepts

Check for weighted graph

```
is_weighted(g)
```

```
## [1] FALSE
```

Check for simple graph (i.e. no self-loop or multi-edges)

```
is_simple(g)
```

```
## [1] TRUE
```

Check for neighbour of a certain node

```
# Neighbours of node 5 in the graph g
```

```
neighbors(g,5)
```

```
## + 3/7 vertices, named, from ede8fcc:
```

```
## [1] 3 4 6
```

Check for the degrees of the nodes in the graph

```
degree(g)
```

```
## 1 2 3 4 5 6 7
```

```
## 2 3 3 4 3 3 2
```

Check if the graph is connected

```
is_connected(g)
```

```
## [1] TRUE
```

Check for clusters in graph (if the graph is connected - 1 component)

```
clusters(g)
```

```
## $membership
```

```
## 1 2 3 4 5 6 7
```

```
## 1 1 1 1 1 1 1
```

```
##
```

```
## $csize
```

```
## [1] 7
```

```
##
```

```
## $no
```

```
## [1] 1
```

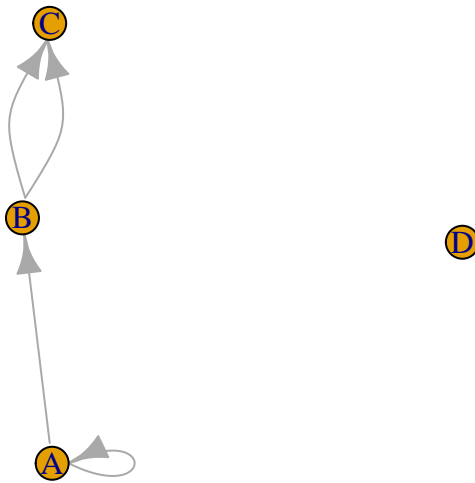
Check for diameter (Longest shortest path)

```
diameter(g, weights=NA)
```

```
## [1] 3
```

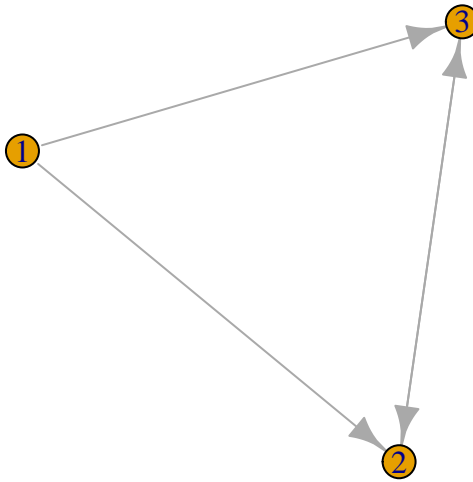
Directed graph in graph with pairs

```
g2 <- graph( c("A", "B", "B", "C", "B", "C", "A", "A"),
isolates=c("D") )
plot(g2)
```



Directed edges in `graph_from_literal` are indicated using a minus/plus convention

```
dg <- graph_from_literal(1-+2, 1-+3, 2++3)
plot(dg)
```



Descriptive Analysis of Network Graph Characteristics

Initialize toy dataset

```
head(elist.lazega)
```

```
##   V1  V2
## 1 V1 V17
## 2 V2  V7
## 3 V2 V16
## 4 V2 V17
## 5 V2 V22
## 6 V2 V26
```

```
head(v.attr.lazega)
```

##	Name	Seniority	Status	Gender	Office	Years	Age	Practice	School
## 1	V1	1	1	1	1	31	64	1	1
## 2	V2	2	1	1	1	32	62	2	1
## 3	V3	3	1	1	2	13	67	1	1
## 4	V4	4	1	1	1	31	59	2	3
## 5	V5	5	1	1	2	31	59	1	2
## 6	V6	6	1	1	2	29	55	1	1

```
g.lazega=graph_from_data_frame(elist.lazega,directed="FALSE",vertices = v.attr.lazega)
```

Betweenness centrality of every node in the graph

```
betweenness(g.lazega)
```

```
##           V1           V2           V3           V4           V5           V6
##  0.0000000  18.4428571  0.6944444  21.3070430  3.5448773  2.0849068
##           V7           V8           V9           V10          V11          V12
##  1.8547619  0.0000000  0.0000000  8.0973957  0.0000000  8.4302572
##           V13          V14          V15          V16          V17          V18
##  0.0000000  13.2791681  22.0448052  70.6494464  101.8843060  21.2662088
##           V19          V20          V21          V22          V23          V24
##  14.0866744  0.3583333  0.0000000  18.4319805  0.0000000  28.6060107
##           V25          V26          V27          V28          V29          V30
##  6.9124542  48.7275072  32.0000000  54.4436758  13.2436480  1.0075758
##           V31          V32          V33          V34          V35          V36
##  62.6624181  52.1727578  3.9833333  2.4651515  8.9770924  0.3409091
```

Eigenvector centrality of every node in the graph

```
eigen_centrality(g.lazega)$vector
```

```
##           V1           V2           V3           V4           V5           V6
##  1.120344e-01  4.780219e-01  1.754210e-01  7.565355e-01  4.163271e-01  3.656392e-01
##           V7           V8           V9           V10          V11          V12
##  1.021647e-01  1.161440e-18  2.538825e-01  3.825620e-01  1.120344e-01  7.376819e-01
##           V13          V14          V15          V16          V17          V18
##  1.141026e-01  4.592049e-01  8.243409e-01  8.772495e-01  1.000000e+00  4.338823e-01
##           V19          V20          V21          V22          V23          V24
##  8.325400e-01  3.604173e-01  2.242129e-02  7.495043e-01  1.161440e-18  6.392531e-01
##           V25          V26          V27          V28          V29          V30
##  3.327115e-01  8.866422e-01  2.001285e-01  7.991835e-01  6.511804e-01  2.643546e-01
##           V31          V32          V33          V34          V35          V36
##  7.355564e-01  8.252849e-01  2.829041e-01  5.020502e-01  5.359166e-01  2.622551e-01
```

Transitivity (i.e. clustering coefficient) of the graph

```
transitivity(g.lazega)
```

```
## [1] 0.3887689
```

Star Wars dataset

Obtained from <https://github.com/pablobarbera/data-science-workshop/blob/master/sna/data/>

Description of dataset

```
edges <- read.csv("C:/Users/Eugene/Downloads/star-wars-network-edges.csv")
nodes <- read.csv("C:/Users/Eugene/Downloads/star-wars-network-nodes.csv")
head(edges)
```

```
##      source target weight
## 1      C-3PO  R2-D2      17
## 2      LUKE   R2-D2      13
## 3     OBI-WAN R2-D2       6
## 4      LEIA   R2-D2       5
## 5       HAN   R2-D2       5
## 6 CHEWBACCA  R2-D2       3
```

```
head(nodes)
```

```
##      name id
## 1      R2-D2 0
## 2 CHEWBACCA 1
## 3      C-3PO 2
## 4      LUKE 3
## 5 DARTH VADER 4
## 6      CAMIE 5
```

Create an igraph object from the above raw data.

```
g <- graph_from_data_frame(d=edges, vertices=nodes, directed=FALSE)
#d describes the edges of the network
g
```

```
## IGRAPH efa65e7 UNW- 22 60 --
## + attr: name (v/c), id (v/n), weight (e/n)
## + edges from efa65e7 (vertex names):
## [1] R2-D2      --C-3PO      R2-D2      --LUKE      R2-D2      --OBI-WAN
## [4] R2-D2      --LEIA       R2-D2      --HAN       R2-D2      --CHEWBACCA
## [7] R2-D2      --DODONNA    CHEWBACCA  --OBI-WAN    CHEWBACCA  --C-3PO
## [10] CHEWBACCA  --LUKE       CHEWBACCA  --HAN       CHEWBACCA  --LEIA
## [13] CHEWBACCA  --DARTH VADER CHEWBACCA  --DODONNA    LUKE       --CAMIE
## [16] CAMIE      --BIGGS      LUKE       --BIGGS      DARTH VADER--LEIA
## [19] LUKE       --BERU       BERU       --OWEN      C-3PO      --BERU
## [22] LUKE       --OWEN      C-3PO      --LUKE      C-3PO      --OWEN
## + ... omitted several edges
```

```
V(g)
```

```
## + 22/22 vertices, named, from efa65e7:
## [1] R2-D2      CHEWBACCA  C-3PO      LUKE      DARTH VADER CAMIE
## [7] BIGGS      LEIA       BERU       OWEN      OBI-WAN     MOTTI
## [13] TARKIN     HAN        GREEDO     JABBA     DODONNA     GOLD LEADER
## [19] WEDGE      RED LEADER RED TEN    GOLD FIVE
```

```
E(g)
```

```
## + 60/60 edges from efa65e7 (vertex names):
## [1] R2-D2      --C-3PO      R2-D2      --LUKE      R2-D2      --OBI-WAN
## [4] R2-D2      --LEIA       R2-D2      --HAN       R2-D2      --CHEWBACCA
## [7] R2-D2      --DODONNA    CHEWBACCA  --OBI-WAN    CHEWBACCA  --C-3PO
```



```
## [10] CHEWBACCA --LUKE      CHEWBACCA --HAN      CHEWBACCA --LEIA
## [13] CHEWBACCA --DARTH VADER CHEWBACCA --DODONNA LUKE      --CAMIE
## [16] CAMIE      --BIGGS      LUKE      --BIGGS      DARTH VADER--LEIA
## [19] LUKE      --BERU      BERU      --OWEN      C-3PO      --BERU
## [22] LUKE      --OWEN      C-3PO      --LUKE      C-3PO      --OWEN
## [25] C-3PO      --LEIA      LUKE      --LEIA      LEIA      --BERU
## [28] LUKE      --OBI-WAN    C-3PO      --OBI-WAN    LEIA      --OBI-WAN
## + ... omitted several edges
```

Character names

```
V(g)$name
```

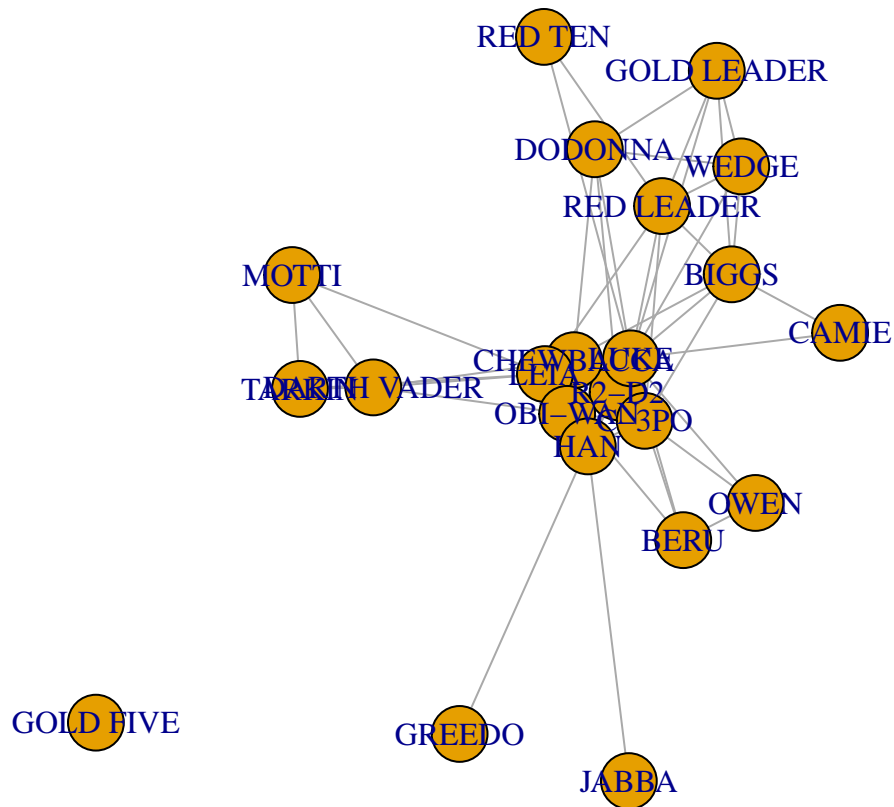
```
## [1] "R2-D2"      "CHEWBACCA"  "C-3PO"      "LUKE"      "DARTH VADER"
## [6] "CAMIE"      "BIGGS"      "LEIA"      "BERU"      "OWEN"
## [11] "OBI-WAN"    "MOTTI"      "TARKIN"     "HAN"      "GREEDO"
## [16] "JABBA"      "DODONNA"    "GOLD LEADER" "WEDGE"     "RED LEADER"
## [21] "RED TEN"    "GOLD FIVE"
```

Edge weights

```
E(g)$weight
```

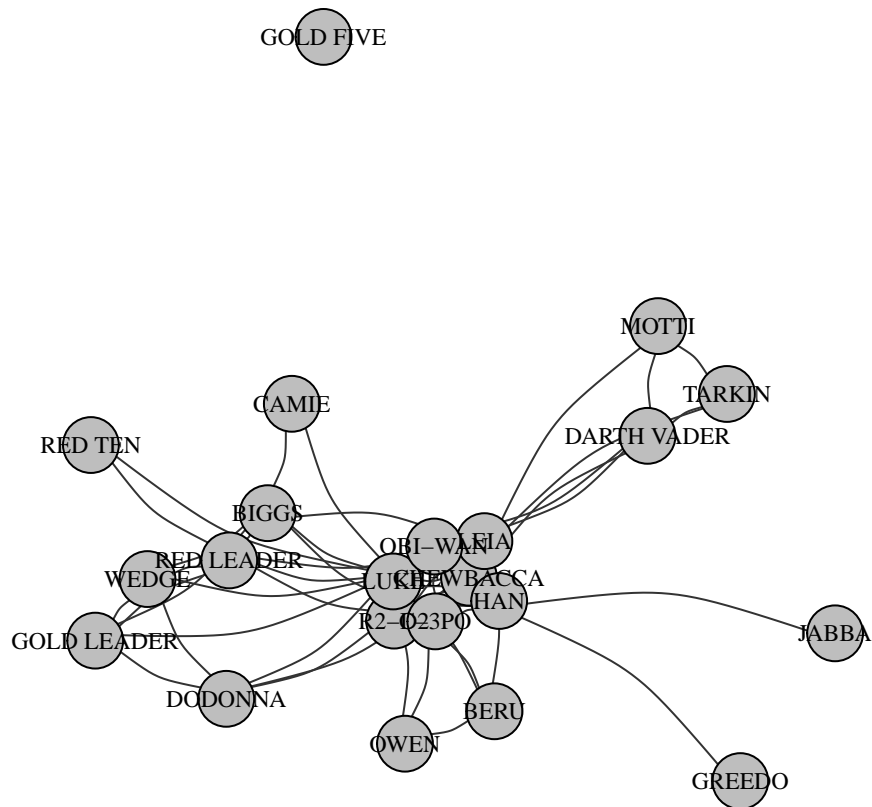
```
## [1] 17 13 6 5 5 3 1 7 5 16 19 11 1 1 2 2 4 1 3 3 2 3 18 2 6
## [26] 17 1 19 6 1 2 1 7 9 26 1 1 6 1 1 13 1 1 1 1 1 1 2 1 1
## [51] 3 3 1 1 3 1 2 1 1 1
```

```
par(mar=c(0,0,0,0))
plot(g)
```



Visualise it better? To see all the available plotting options, you can check `?igraph.plotting`.

```
par(mar=c(0,0,0,0))
plot(g,
vertex.color = "grey", # change color of nodes
vertex.label.color = "black", # change color of labels
vertex.label.cex = .75, # change size of labels to 75% of original size
edge.curved=.25, # add a 25% curve to the edges
edge.color="grey20") # change edge color to grey
```



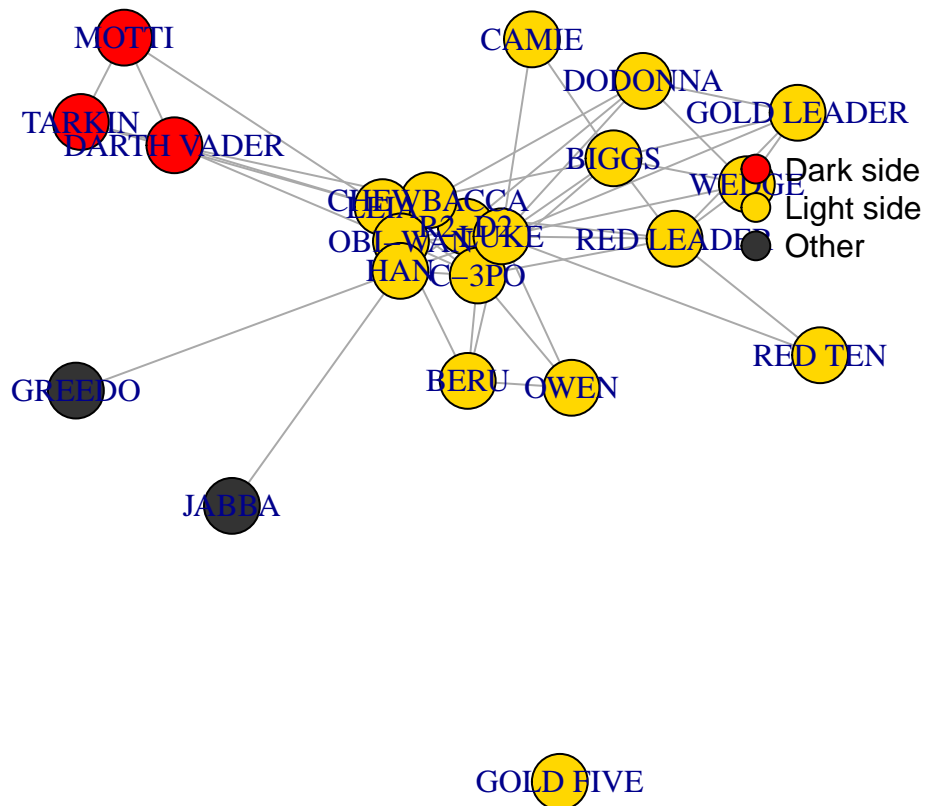
Change based on which side they belong to:

```
# create vectors with characters in each side
dark_side <- c("DARTH VADER", "MOTTI", "TARKIN")
light_side <- c("R2-D2", "CHEWBACCA", "C-3PO", "LUKE", "CAMIE", "BIGGS",
"LEIA", "BERU", "OWEN", "OBI-WAN", "HAN", "DODONNA",
"GOLD LEADER", "WEDGE", "RED LEADER", "RED TEN", "GOLD FIVE")
other <- c("GREEDO", "JABBA")
# node we'll create a new color variable as a node property
V(g)$color <- NA
V(g)$color[V(g)$name %in% dark_side] <- "red"
V(g)$color[V(g)$name %in% light_side] <- "gold"
V(g)$color[V(g)$name %in% other] <- "grey20"
vertex_attr(g)
```

```
## $name
## [1] "R2-D2"      "CHEWBACCA"  "C-3PO"      "LUKE"       "DARTH VADER"
## [6] "CAMIE"      "BIGGS"      "LEIA"       "BERU"       "OWEN"
## [11] "OBI-WAN"    "MOTTI"      "TARKIN"     "HAN"        "GREEDO"
## [16] "JABBA"      "DODONNA"    "GOLD LEADER" "WEDGE"      "RED LEADER"
## [21] "RED TEN"    "GOLD FIVE"
##
## $id
## [1] 0 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 17 18 19 20 21
##
## $color
```

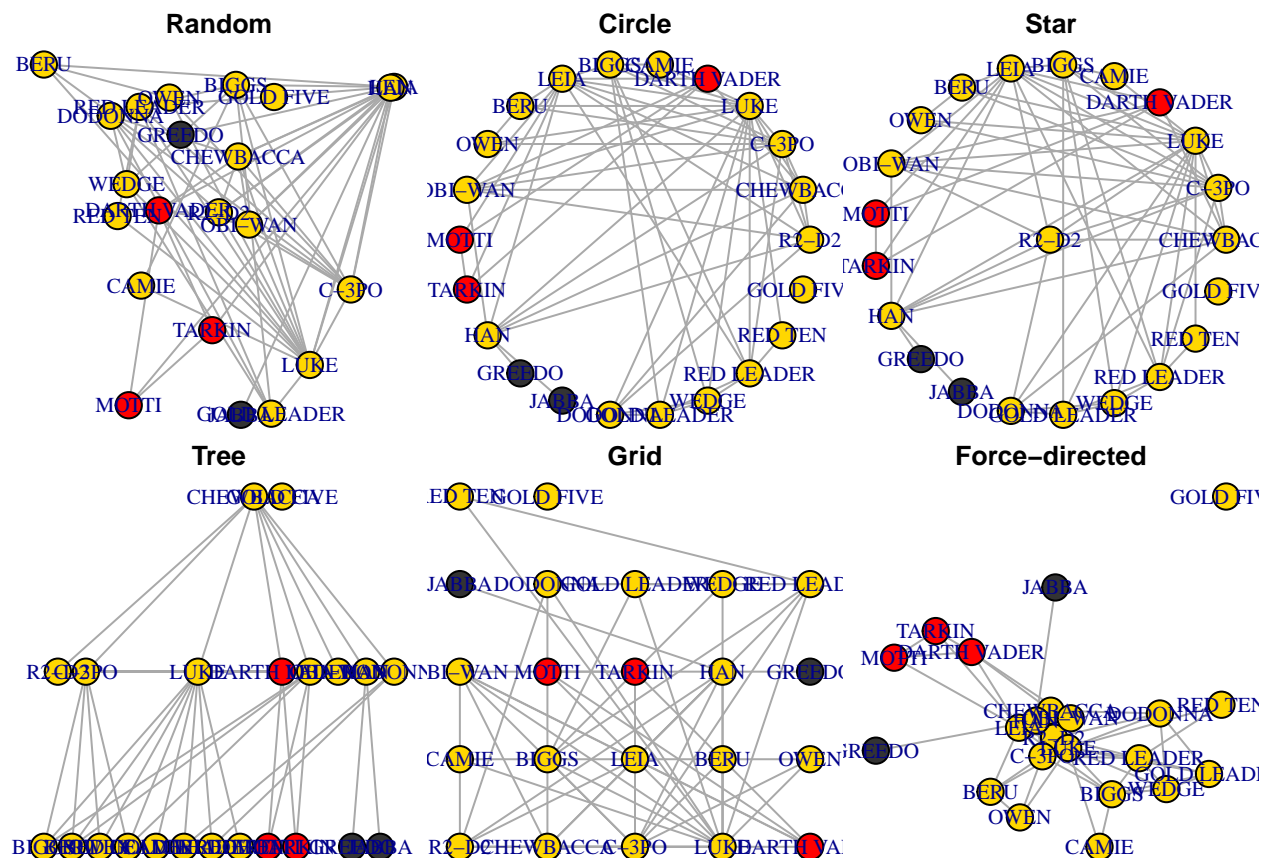
```
## [1] "gold" "gold" "gold" "gold" "red" "gold" "gold" "gold"
## [9] "gold" "gold" "gold" "red" "red" "gold" "grey20" "grey20"
## [17] "gold" "gold" "gold" "gold" "gold" "gold"
```

```
par(mar=c(0,0,0,0)); plot(g)
legend(x=.75, y=.75, legend=c("Dark side", "Light side", "Other"),
pch=21, pt.bg=c("red", "gold", "grey20"), pt.cex=2, bty="n")
```



Up to now, each time we run the plot function, the nodes appear to be in a different location. Why? Because it's running a probabilistic function trying to locate them in the optimal way possible. However, we can also specify the layout for the plot; that is, the (x,y) coordinates where each node will be placed. igraph has a few different layouts built-in, that will use different algorithms to find an optimal distribution of nodes. The following code illustrates some of these:

```
par(mfrow=c(2, 3), mar=c(0,0,1,0))
plot(g, layout=layout_randomly, main="Random")
plot(g, layout=layout_in_circle, main="Circle")
plot(g, layout=layout_as_star, main="Star")
plot(g, layout=layout_as_tree, main="Tree")
plot(g, layout=layout_on_grid, main="Grid")
plot(g, layout=layout_with_fr, main="Force-directed")
```



References

Statistical Analysis of Network Data with R - Eric D. Kolaczyk, Gábor Csárdi

More help:

Use command: `?igraph` or visit <https://cran.r-project.org/web/packages/igraph/igraph.pdf>