Social Network Analysis with R *

Yanchang Zhao
http://www.RDataMining.com

R and Data Mining Course
Beijing University of Posts and Telecommunications,
Beijing, China

July 2019

^{*}Chapter 11: Social Network Analysis, in R and Data Mining: Examples and Case Studies. http://www.rdatamining.com/docs/RDataMining=bookspdf E + A + A +

Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

Advanced Graph Visualization

R Packages

Wrap Up

Network and Graph

- Nodes, vertices or entities
- Edges, links or relationships
- Network analysis, graph mining
- ► Link prediction, community/group detection, entity resolution, recommender system, information propagation modeling

Graph Databases

- ► Neo4j: https://neo4j.com
- Giraph on Hadoop: http://giraph.apache.org
- GraphX on Spark: http://spark.apache.org/graphx/

Social Network Analysis

- ► Graph construction
- Graph query
- Centrality measures
- Graph visualization
- Clustering and community detection

Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

Advanced Graph Visualization

R Packages

Wrap Up

Graph Construction

- Tom, Ben, Bob and Mary are friends of John.
- ► Alice and Wendy are friends of Mary.
- Wendy is a friend of David.

```
library(igraph)
# nodes
nodes <- data.frame(</pre>
 name = c("Tom", "Ben", "Bob", "John", "Mary", "Alice", "Wendy", "David"),
  gender = c("M", "M", "M", "F", "F", "F", "M"),
  age = c(16, 30, 42, 29, 26, 32, 18, 22)
# relations
edges <- data.frame(
 from = c("Tom", "Ben", "Bob", "Mary", "Alice", "Wendy", "Wendy"),
  to = c("John", "John", "John", "Mary", "Mary", "David")
# build a graph object
g <- graph.data.frame(edges, directed=F, vertices=nodes)</pre>
```

Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

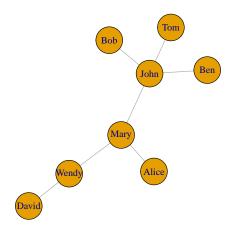
Advanced Graph Visualization

R Packages

Wrap Up

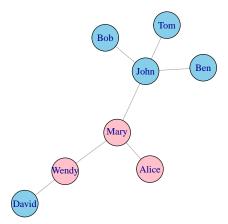
Graph Visualization

```
layout1 <- g %>% layout_nicely() ## save layout for reuse
g %>% plot(vertex.size = 30, layout = layout1)
```



Graph Visualization (cont.)

```
## use blue for male and pink for female
colors <- ifelse(V(g)$gender=="M", "skyblue", "pink")
g %>% plot(vertex.size=30, vertex.color=colors, layout=layout1)
```



Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

Advanced Graph Visualization

R Packages

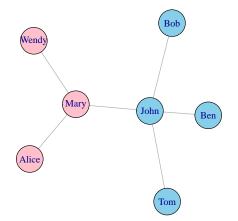
Wrap Up

Graph Query

```
## nodes
V(g)
## + 8/8 vertices, named, from 8dfec3f:
## [1] Tom Ben Bob John Mary Alice Wendy David
## edges
E(g)
## + 7/7 edges from 8dfec3f (vertex names):
## [1] Tom --John Ben --John Bob --John John --Mary
## [5] Mary --Alice Mary --Wendy Wendy--David
## immediate neighbors (friends) of John
friends <- ego(g,order=1,nodes="John",mindist=1)[[1]] %>% print()
## + 4/8 vertices, named, from 8dfec3f:
## [1] Tom Ben Bob Mary
## female friends of John
friends[friends$gender == "F"]
## + 1/8 vertex, named, from 8dfec3f:
## [1] Mary
```

Graph Query (cont.)

```
## 1- and 2-order neighbors (friends) of John
g2 <- make_ego_graph(g, order=2, nodes="John")[[1]]
g2 %>% plot(vertex.size=30, vertex.color=colors)
```



Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

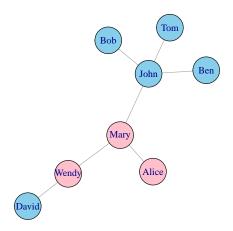
Centrality Measures

Advanced Graph Visualization

R Packages

Wrap Up

Friendship Graph



Centrality Measures

- Degree: the number of adjacent edges; indegree and outdegree for directed graphs
- ► Closeness: the inverse of the average length of the shortest paths to/from all other nodes
- Betweenness: the number of shortest paths going through a node

```
degree <- g %>% degree() %>% print()
    Tom Ben Bob John Mary Alice Wendy David
##
   1 1 1 4 3 1 2 1
##
closeness <- g %>% closeness() %>% round(2) %>% print()
##
    Tom Ben Bob John Mary Alice Wendy David
   0.06 0.06 0.06 0.09 0.09 0.06 0.07 0.05
##
betweenness <- g %>% betweenness() %>% print()
    Tom Ben Bob John Mary Alice Wendy David
##
                    15
                         14 0
##
                0
                                      4 D > 4 A > 4 B > 4 B >
```

Centrality Measures (cont.)

- Eigenvector centrality: the values of the first eigenvector of the graph adjacency matrix
- Transivity, a.k.a clustering coefficient, measures the probability that the adjacent nodes of a node are connected.

```
eigenvector <- evcent(g)$vector %>% round(2) %>% print()
## Tom Ben Bob John Mary Alice Wendy David
## 0.45 0.45 0.45 1.00 0.85 0.38 0.48 0.22

transitivity <- g %>% transitivity(type = "local") %>% print()
## [1] NaN NaN NaN 0 0 NaN
```

Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

Advanced Graph Visualization

R Packages

Wrap Up

Static Network Visualization

- Static network visualization
- ► Fast in rendering big graphs
- For very big graphs, the most efficient way is to save visualization result into a file, instead of directly to screen.
- Save network diagram into files: pdf(), bmp(), jpeg(), png(), tiff()

```
library(igraph)

## plot directly to screen when graph is small
plot(g)

## for big graphs, save visualization to a PDF file
pdf("mygraph.pdf")
plot(g)
graphics.off() ## or dev.off()
```

Interactive Network Visualization

- Coordinates of other nodes are not adjusted when moving a node.
- Can be slow when rendering big graphs
- Save network diagram into files: visSave(), visExport()

```
visIgraph(g, idToLabel=T) %>%
  ## highlight nodes connected to a selected node
  visOptions(highlightNearest=T) %>%
  ## use different icons for different types (groups) of nodes
  visGroups(groupname="person", shape="icon",
            icon=list(code="f007")) %>%
  . . . %>%
  ## use FontAwesome icons
  addFontAwesome() %>%
  ## add legend of nodes
  visLegend() %>%
  ## to save to file
  visSave(file = "network.html")
```

Interactive Network Visualization (cont.)

- Dynamically adjusting coordinates for better visualization
- Very slow when rendering big graphs

Load Graph Data

```
## download graph data
url <- "http://www.rdatamining.com/data/graph.rdata"
download.file(url, destfile = "./data/graph.rdata")</pre>
```

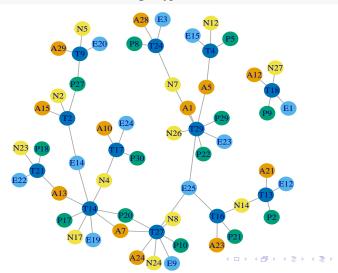
```
library(igraph)
# load graph data into R
# what will be loaded: g, nodes, edges
load("../data/graph.rdata")
```

Build a Graph

```
head(nodes, 3)
##
     name type
## 1
      T9 tid
## 2 T24 tid
## 3 T13 tid
head(edges, 3)
##
     from to
## 1
       T9 P27
## 2 T24 P8
## 3 T13 P2
## buid a graph object
g <- graph.data.frame(edges, directed = F, vertices = nodes)</pre>
g
## IGRAPH 9597c42 UN-B 61 60 --
## + attr: name (v/c), type (v/c)
## + edges from 9597c42 (vertex names):
    [1] T9 --P27 T24--P8 T13--P2 T27--P10 T29--P29 T2 --P27
##
##
    [7] T16--P21 T27--P20 T17--P30 T14--P20 T29--P22 T14--P17
## [13] T21--P18 T18--P9 T4 --P5 T9 --A29 T24--A28 T13--A21
```

Example of Static Network Visualization

```
library(igraph)
plot(g, vertex.size=12, vertex.label.cex=0.7,
    vertex.color=as.factor(V(g)$type), vertex.frame.color=NA)
```



24 / 37

Example of Interactive Network Visualization

```
library(visNetwork)
V(g)$group <- V(g)$type
## visualization
data <- toVisNetworkData(g)
visNetwork(nodes=data$nodes, edges=data$edges) %>%
visGroups(groupname="tid",shape="icon",icon=list(code="f15c")) %>%
visGroups(groupname="person",shape="icon",icon=list(code="f007")) %>%
visGroups(groupname="addr",shape="icon",icon=list(code="f015")) %>%
visGroups(groupname="phone",shape="icon",icon=list(code="f095")) %>%
visGroups(groupname="email",shape="icon",icon=list(code="f095")) %>%
visGroups(groupname="email",shape="icon",icon=list(code="f000")) %>%
```





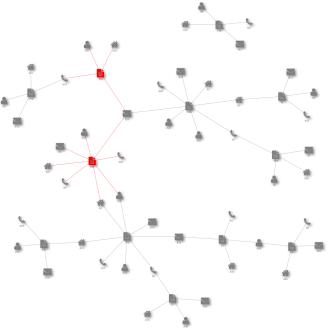
Address











Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

Advanced Graph Visualization

R Packages

Wrap Up

R Packages

- ▶ Network analysis: igraph, sna, statnet
- ► Network visualization: *visNetwork*
- ► Interface with graph databases: RNeo4j

Package igraph †

- ▶ V(g), E(g): nodes and edges of graph g
- degree, betweenness, closeness, transitivity: various centrality scores
- neighborhood: neighborhood of graph vertices
- cliques, largest.cliques, maximal.cliques, clique.number: find cliques, ie. complete subgraphs
- clusters, no.clusters: maximal connected components of a graph and the number of them
- fastgreedy.community, spinglass.community: community detection
- cohesive.blocks: calculate cohesive blocks
- induced.subgraph: create a subgraph of a graph (igraph)
- read.graph, write.graph: read and writ graphs from and to files of various formats

[†]https://cran.r-project.org/package=igraph

Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

Advanced Graph Visualization

R Packages

Wrap Up

Wrap Up

- Package igraph and sna
 - Static visualization
 - Can visualize nodes with shapes, images and icons
 - Visualise very large graph
 - Support network analysis and graph mining
- Package visNetwork
 - Interactive visualization
 - Can visualize nodes with shapes, images and icons
 - Image rendering can be very slow for large graphs
 - Designed for visualization only, and does not support network analysis and graph mining

Graph and Social Network Analysis

Graph Construction

Graph Visualization

Graph Query

Centrality Measures

Advanced Graph Visualization

R Packages

Wrap Up

Further Readings

- Social network analysis (SNA) https://en.wikipedia.org/wiki/Social_network_analysis
- igraph a network analysis package, supporting R, Python and C/C++ http://igraph.org
- sna an R package for social network analysis https://cran.r-project.org/web/packages/sna/index.html
- statnet software tools for the analysis, simulation and visualization of network data; also available as an R package http://www.statnet.org
- visNetwork an R package for network visualization http://datastorm-open.github.io/visNetwork/

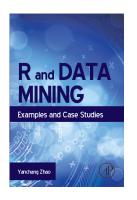
Online Resources

▶ Book titled *R* and *Data Mining: Examples and Case Studies* [Zhao, 2012]

http://www.rdatamining.com/docs/RDataMining-book.pdf

- R Reference Card for Data Mining http://www.rdatamining.com/docs/RDataMining-reference-card.pdf
- ► Free online courses and documents http://www.rdatamining.com/resources/
- ▶ RDataMining Group on LinkedIn (27,000+ members) http://group.rdatamining.com
- Twitter (3,300+ followers)@RDataMining

The End





Thanks!

Email: yanchang(at)RDataMining.com

Twitter: @RDataMining

How to Cite This Work

Citation

Yanchang Zhao. R and Data Mining: Examples and Case Studies. ISBN 978-0-12-396963-7, December 2012. Academic Press, Elsevier. 256 pages. URL: http://www.rdatamining.com/docs/RDataMining-book.pdf.

▶ BibTex

```
@BOOK{Zhao2012R,
    title = {R and Data Mining: Examples and Case Studies},
    publisher = {Academic Press, Elsevier},
    year = {2012},
    author = {Yanchang Zhao},
    pages = {256},
    month = {December},
    isbn = {978-0-123-96963-7},
    keywords = {R, data mining},
    url = {http://www.rdatamining.com/docs/RDataMining-book.pdf}}
```

References I



Zhao, Y. (2012).

R and Data Mining: Examples and Case Studies, ISBN 978-0-12-396963-7. Academic Press, Elsevier.