

# Building a graph from raw data

CASE STUDIES: NETWORK ANALYSIS IN R



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Instructor

# Exploring the data

- Data is several days of all the tweets mentioning #rstats
- Key attributes for building a graph are:
  - screen name
  - raw text of the tweet

# Anatomy of a tweet

1. *ReecheshJC*: "Hey #rstats, how do I do fct\_lump but where I lump based on count values in a column?"
2. *kom\_256*: "RT @elenagbg: Retweeted R-Ladies Madrid (@RLadiesMAD):\n\nEn el #OCSummit17... Fast Talks sobre #rstats organizado por... <https://t.co/CKY5aG...>"

```
library(igraph)
library(stringr)
raw_tweets <- read.csv("datasets/rstatstweets.csv",
  stringsAsFactors = FALSE)
```

## Data sample, single row

```
user_name:      Karen Millidine
screen_name:    KJMillidine
tweet_text:     RT @Rbloggers: RStudio v1.1 Released
https://t.co/kCMHc689nY #rstats #DataScience
favorites:      0
retweets:       96
location:       None
expanded_url:   https://wp.me/pMm6L-ExV
in_reply_to_tweet_id: NA
in_reply_to_user_id: NA
dt:             10/10/17
```

# Building the graph

```
## Get all the screen names  
all_sn <- unique(raw_tweets$screen_name)  
  
## Create graph  
retweet_graph <- graph.empty()  
  
## Add screen names as vertices  
retweet_graph <- retweet_graph + vertices(all_sn)
```

# Building the graph

```
## Extract name and add edges
for(i in 1:dim(raw_tweets)[1]){
  # Extract retweet name
  rt_name <- find_rt(raw_tweets$tweet_text[i])
  # If there is a name add an edge
  if(!is.null(rt_name)){
    # Check to make sure the vertex exists, if not, add it
    if(!rt_name %in% all_sn){
      retweet_graph <- retweet_graph + vertices(rt_name)
    }
    # add the edge
    retweet_graph <- retweet_graph +
      edges(c(raw_tweets$screen_name[i], rt_name))
  }
}
```

# Cleaning the graph

```
## Size the number of degree 0 vertices  
sum(degree(retweet_graph) == 0)  
  
## Trim and simplify  
retweet_graph <- simplify(retweet_graph)  
retweet_graph <- delete.vertices(retweet_graph,  
degree(retweet_graph) == 0)
```

# **Let's practice!**

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# Building a mentions graph

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# Recall tweet anatomy

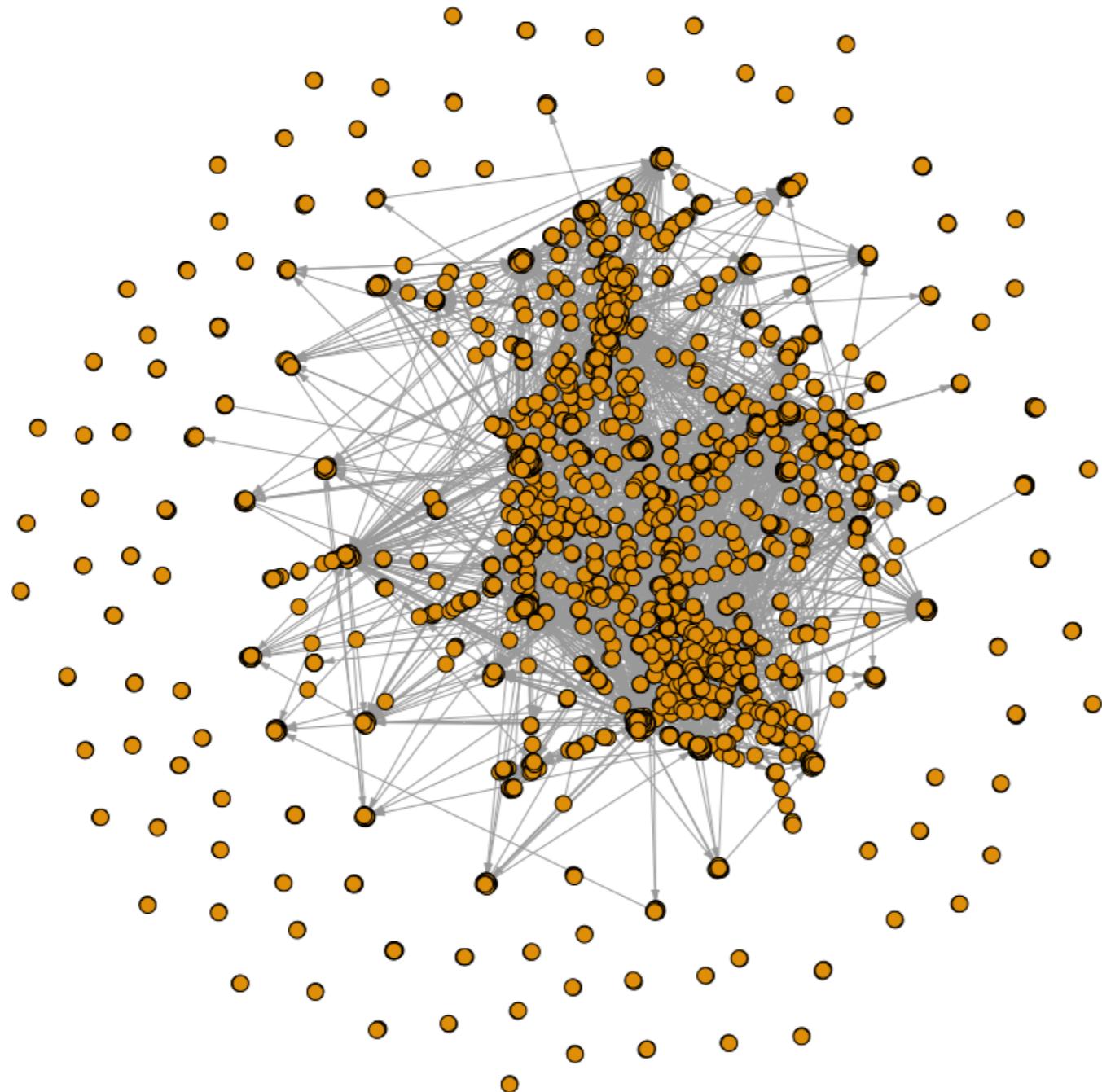
*AlexisAchim*: "@LAStools @Lees\_Sandbox @jhollist  
@LeahAWasser LidR is also available directly on CRAN #rstats"

*timelyportfolio*: "just might have a demo of @emeeks new  
#reactjs/#d3js semiotic in #rstats in the works"

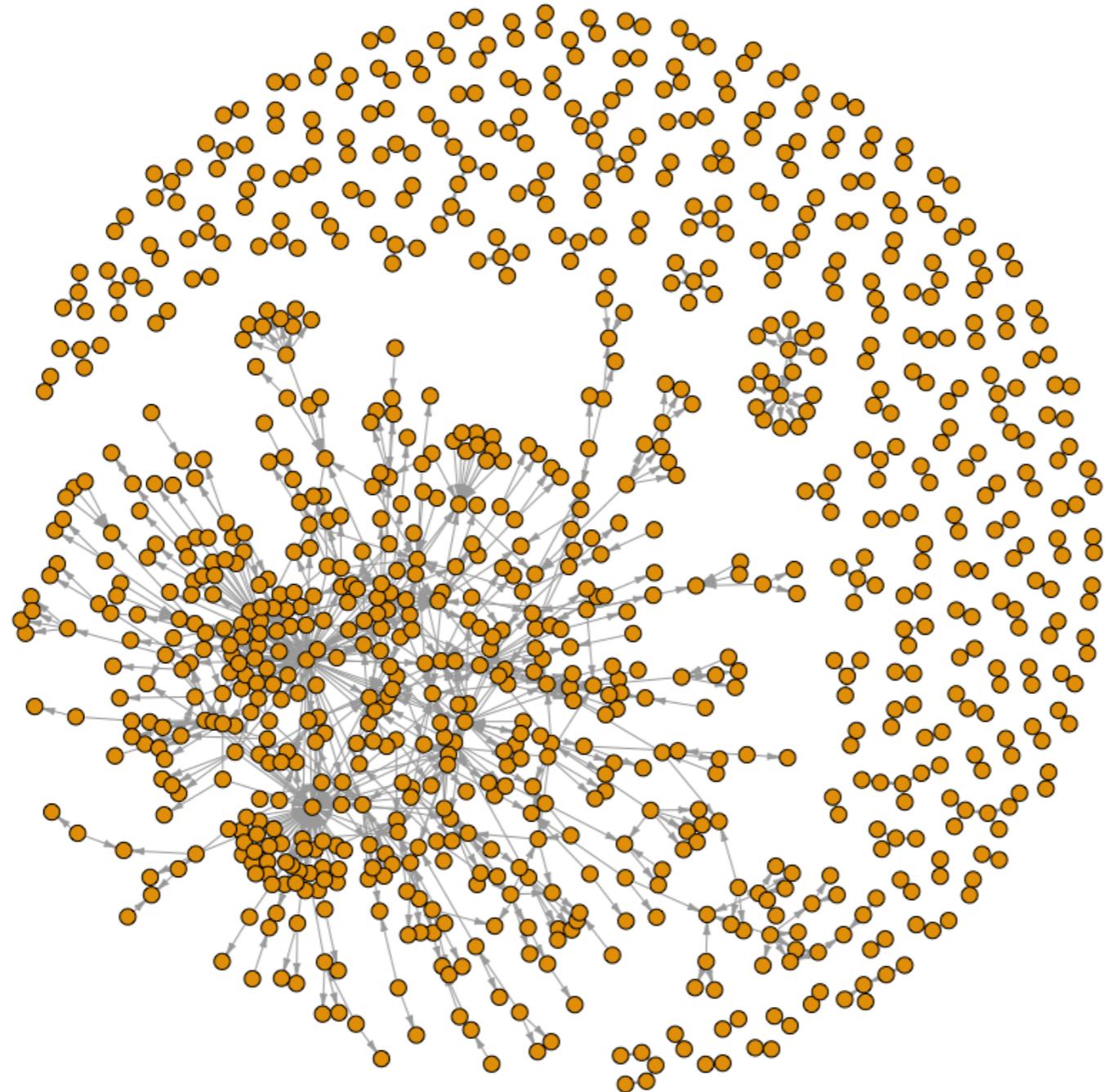
# Build your mentions graph

```
ment_g <- graph.empty()
ment_g <- ment_g + vertices(all_sn)
for(i in 1:dim(raw_tweets)[1]) {
  ment_name <- mention_ext(raw_tweets$ tweet_text[i])
  if(length(ment_name) > 0 ) {
    # Add the edge(s)
    for(j in ment_name) {
      # Check to make sure the vertex exists, if not, add it
      if(!j %in% all_sn) {
        ment_g <- ment_g + vertices(j)  }
      ment_g <- ment_g + edges(c(raw_tweets$screen_name[i], j))
    }
  }
}
ment_g <- simplify(ment_g)
ment_g <- delete.vertices(ment_g, degree(ment_g) == 0)
```

# Retweet Graph



## Mentions Graph



# **Let's practice!**

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# Finding communities

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# Three different communities

```
undirected_ment_g <- as.undirected(ment_g)

ment_edg <- cluster_edge_betweenness(undirected_ment_g)
ment_eigen <- cluster_leading_eigen(undirected_ment_g)
ment_lp <- cluster_label_prop(undirected_ment_g)
```

# Sizing the communities

```
length(ment_edg)  
length(ment_eigen)  
length(ment_lp)
```

```
173  
168  
212
```

```
table(sizes(ment_edg))
```

	2	3	4	5	6	7	8	9	11	12	18	19	20	23	24	26	28
103	21	14	7	3	3	1	2	1	2	2	1	1	1	1	2	1	1
31	33	38	40	41	52	58											
	1	1	1	1	1	1	1										

```
table(sizes(ment_eigen))
```

	2	3	4	5	6	7	9	10	12	18	23	26	29	30	32	34	35	58
103	22	14	7	4	3	1	1	1	1	1	1	1	1	1	1	1	1	1
64	66	101																
	1	1	1															

```
table(sizes(ment_lp))
```

	2	3	4	5	6	7	8	9	10	11	12	13	16	25	26	67	70
103	32	22	19	8	5	4	3	5	1	2	3	1	1	1	1	1	1

# Comparing communities

```
compare(ment_edg, ment_eigen, method = 'vi')
```

```
0.9761792
```

```
compare(ment_eigen, ment_lp, method = 'vi')
```

```
1.192238
```

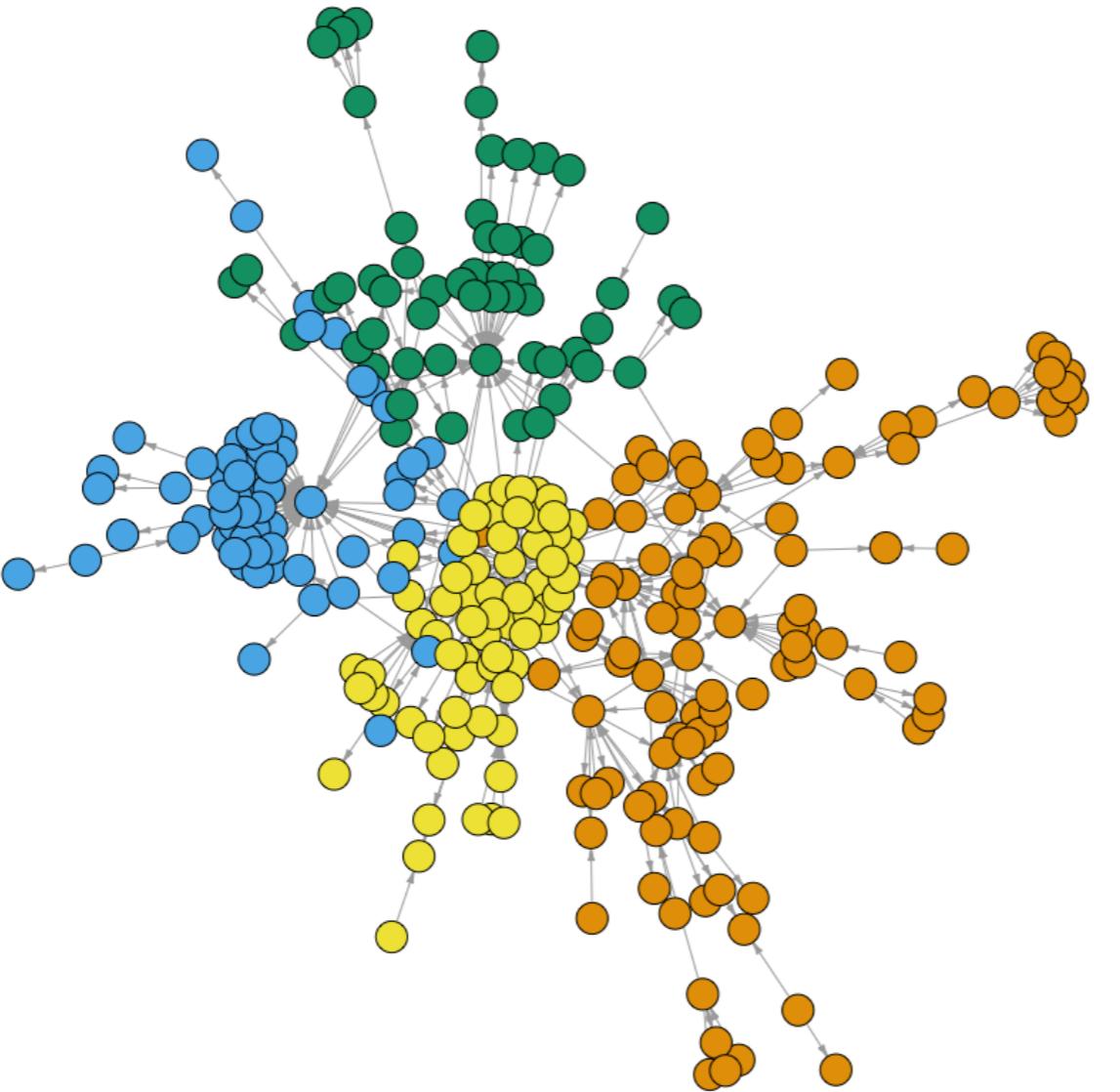
```
compare(ment_lp, ment_edg, method = 'vi')
```

```
0.9631608
```

# Plotting community structure

```
lrg_eigen <- as.numeric(  
  names(ment_eigen[which(sizes(ment_eigen) > 45)])  
)  
  
eigen_sg <- induced.subgraph(ment_g,  
  V(ment_g)[ eigen %in% lrg_eigen])  
plot(eigen_sg, vertex.label = NA, edge.arrow.width = .8,  
  edge.arrow.size = 0.2,  
  coords = layout_with_fr(ment_sg), margin = 0,  
  vertex.size = 6, vertex.color =  
  as.numeric(as.factor(V(eigen_sg)$eigen)))
```

## Mentions subgraph communities



# **Let's practice!**

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