Exploring our data

CASE STUDIES: NETWORK ANALYSIS IN R



Edmund Hart
Instructor



```
library(igraph)
library(dplyr)
library(lubridate)
bike_dat <- read.csv("datasets/bike2_test3.csv", stringsAsFactors = FALSE)
str(bike_dat)</pre>
```

```
'data.frame':
               52800 obs. of 13 variables:
$ tripduration : int 295 533 1570 2064 2257 296 412...
$ from_station_id : int 49 165 25 300 85 174 75 45 85 99 ...
$ from_station_name: chr "Dearborn St & Monroe St" ...
$ to_station_id : int 174 308 287 296 313 198 56 147 174 99 ...
$ to_station_name : chr "Canal St & Madison St" ...
$ usertype
                  : chr "Subscriber" "Subscriber" "Customer"...
$ gender
                  : chr "Male" "Male" "" ...
$ birthyear
                  : int 1964 1972 NA NA 1963 1973 1989 1965 1983 1983 ...
$ from_latitude
                  : num 41.9 42 41.9 41.9 41.9 ...
$ from_longitude
                  : num -87.6 -87.7 -87.6 -87.6 -87.6 ...
$ to_latitude
                  : num 41.9 41.9 41.9 41.9 ...
$ to_longitude
                  : num -87.6 -87.7 -87.6 -87.6 -87.6 ...
$ geo_distance
                  : num 859 1882 2159 288 3044 ...
```



Creating the bike sharing graph

```
trip_df <- bike_dat %>%
    group_by(from_station_id, to_station_id) %>%
    summarize(weights = n())
head(trip_df)
```

```
# A tibble: 6 x 3
# Groups: from_station_id [1]
 from_station_id to_station_id weights
            <int>
                          <int>
                                  <int>
                              5
                             14
3
                5
                                      1
                             16
                             25
                                      3
                                      3
5
                             29
                             33
```



Creating the bike sharing graph

```
trip_g <- graph_from_data_frame(trip_df[, 1:2])
# add edge weights
E(trip_g)$weight <- trip_df$weights
# Quick exploration of our graph
gsize(trip_g)</pre>
```

19052

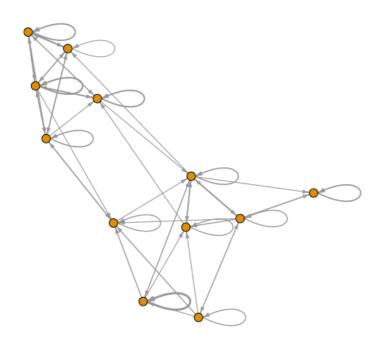
gorder(trip_g)

300



Explore the graph

```
sg <- induced_subgraph(trip_g, 1:12)
plot(sg, vertex.label = NA, edge.arrow.width = 0.8,
    edge.arrow.size = 0.6,
    margin = 0,
    vertex.size = 6,
    edge.width = log(E(sg)$weight + 2))</pre>
```



Let's practice!

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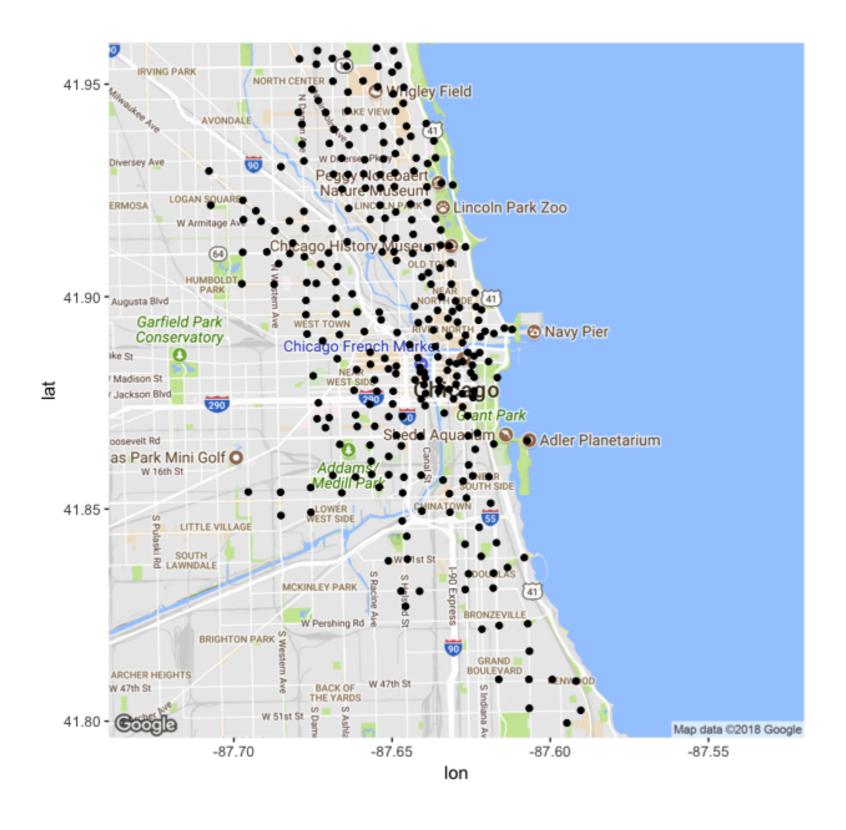
Compare graph distance vs. geographic distance

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Graph distance

```
farthest_vertices(trip_g_simp)
$vertices
+ 2/300 vertices, named, from 20dcfff:
[1] 336 340
$distance
[1] 5
get_diameter(trip_g_simp)
+ 4/300 vertices, named, from 20dcfff:
[1] 336 267 76 340
```







Geographic distance

```
library(geosphere)
# Get the to stations coordinates
st_to <- bike_dat %>%
            filter(from_station_id == 336) %>%
            sample_n(1) %>%
            select(from_longitude, from_latitude)
# Get the from stations coordinates
st_from <- bike_dat %>%
            filter(from_station_id == 340) %>%
            sample_n(1) %>%
            select(from_longitude, from_latitude)
# find the geographic distance
farthest_dist <- distm(st_from, st_to, fun = distHaversine)</pre>
farthest_dist
```

[1,] 13660.66



Geographic distance

```
bike_dist <- function(station_1, station_2, divy_bike_df){</pre>
    st1 <- divy_bike_df %>%
                filter(from_station_id == station_1) %>%
                sample_n(1) %>%
                select(from_longitude, from_latitude)
    st2 <- divy_bike_df %>%
                filter(from_station_id == station_2) %>%
                sample_n(1) %>%
                select(from_longitude, from_latitude)
    farthest_dist <- distm(st1, st2, fun = distHaversine)</pre>
    return(farthest_dist)
```

Let's practice!

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Connectivity

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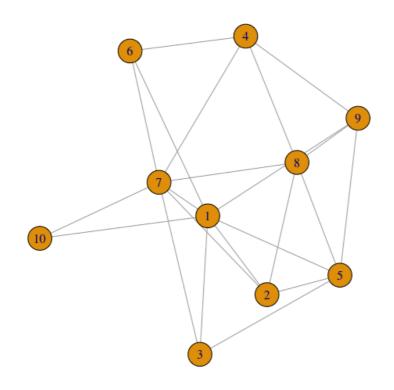


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Measuring connectivity

```
rand_g <- erdos.renyi.game(10, 0.4, "gnp", directed = FALSE)
plot(rand_g)</pre>
```



Measuring connectivity

```
rand_g <- erdos.renyi.game(10, 0.4, "gnp", directed = FALSE)
vertex_connectivity(rand_g)</pre>
```

2

```
edge_connectivity(rand_g)
```

2



Minimum number of cuts

```
min_cut(rand_g, value.only = FALSE)
```

```
$value
[1] 2
$cut
+ 2/18 edges from 17a8fad:
[1] 10--7 10--1
$partition1
+ 1/10 vertex, from 17a8fad:
[1] 10
$partition2
+ 9/10 vertices, from 17a8fad:
[1] 1 2 3 4 5 6 7 8 9
```



Connectivity randomizations

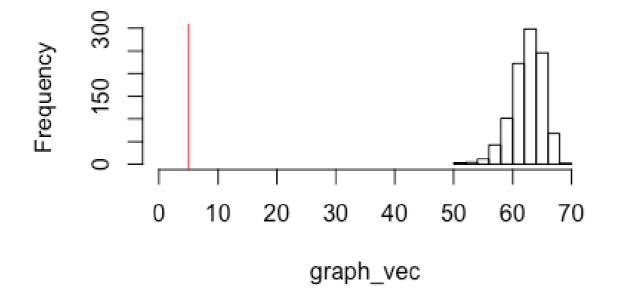
```
# Get parameters to simulate graph
nv <- gorder(trip_g_ud)</pre>
ed <- edge_density(trip_g_ud)</pre>
# Empty vector to store output
graph_vec <- rep(NA, 1000)
# Generate 1000 random graphs and find the edge connectivity
for(i in 1:1000) {
  w1 <- erdos.renyi.game(nv, ed, "gnp", directed = TRUE)</pre>
  graph_vec[i] <- edge_connectivity(w1)</pre>
```



Connectivity randomizations

```
# Find actual connectivity
econn <- edge_connectivity(trip_g_ud)
hist(graph_vec, xlim = c(0, 140))
abline(v = edge_connectivity(trip_g_ud))</pre>
```

Connectivity Randomizations



Let's practice!

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