

ex3

Load the data

```
library(arrow)

## Warning: package 'arrow' was built under R version 4.1.2
##
## Attaching package: 'arrow'
##
## The following object is masked from 'package:utils':
##
##     timestamp

df = read_parquet("~/Desktop/McGill/ORGB/2022-ona-assignments/ex3/app_data_sample.parquet")
```

Predicting examiners' gender based on first name:

The gender package attempts to infer gender (or more precisely, sex assigned at birth) based on first names using historical data, typically data that was gathered by the state.

```
library(gender)
library(dplyr)

##
## Attaching package: 'dplyr'
##
## The following objects are masked from 'package:stats':
##
##     filter, lag
##
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union

first_name = df %>% distinct(examiner_name_first)
gender_probability = gender(first_name$examiner_name_first)
gender_dictionary = gender_probability %>% select(name, gender)
df <- df %>% left_join(gender_dictionary, by = c("examiner_name_first" = "name"))
head(df$gender)

## [1] "female" NA      "female" "female" "male"   "female"
```

The gender package assign gender based on historical data. Some of the name is not in the data set, thus there are some missing gender information. I filled those values by distribution.

```
table(is.na(df$gender))
```

```
##  
##    FALSE    TRUE  
## 1714618  303859
```

```
gender_na = is.na(df$gender)  
gender_fill = sample(df$gender[!gender_na], size = sum(gender_na), replace = TRUE)  
df$gender[is.na(df$gender)] <- gender_fill  
table(is.na(df$gender))
```

```
##  
##    FALSE  
## 2018477
```

All the missing value has been filled.

Predicting examiners' race based on last name:

The “predictrace” package predict the race of a surname using U.S. Census data which says how many people of each race has a certain surname.

```
library(predictrace)  
race = predict_race(df$examiner_name_last, probability = FALSE)  
df$race = race$likely_race  
head(df$race,10)
```

```
## [1] "white" "white" "white" "white" "white" "white" "black" "white" NA  
## [10] "asian"
```

Again, fill the missing values based on distribution.

```
table(is.na(df$race))
```

```
##  
##    FALSE    TRUE  
## 1704131  314346
```

```
race_na = is.na(df$race)  
race_fill = sample(df$race[!race_na], size = sum(race_na), replace = TRUE)  
df$race[is.na(df$race)] <- race_fill  
table(is.na(df$race))
```

```
##  
##    FALSE  
## 2018477
```

Calculate Tenure

To calculate tenure, I need to calculate the time the application stay in the system.

For most applications, the filing date is the date on which PTO received the application.

The `appl_status_date` variable indicates the date that the application entered its most recent status (or status as of the end of 2014).

```
tenure_info <- df %>% select(examiner_id, filing_date, appl_status_date)

library(lubridate)

##
## Attaching package: 'lubridate'

## The following object is masked from 'package:arrow':
##
##     duration

## The following objects are masked from 'package:base':
##
##     date, intersect, setdiff, union

tenure_info = tenure_info %>% mutate(appl_status_date = as_date(dmy_hms(appl_status_date)))
tenure_info$tenure_days = as.numeric(difftime(tenure_info$appl_status_date,tenure_info$filing_date,units="days"))

## detect missing values
table(is.na(tenure_info$tenure_days))

##
##      FALSE      TRUE
## 2013867    4610

## fill missing values
tenure_na = is.na(tenure_info$tenure_days)
tenure_fill = sample(tenure_info$tenure_days[!tenure_na], size = sum(tenure_na), replace = TRUE)
tenure_info$tenure_days[is.na(tenure_info$tenure_days)] <- tenure_fill
table(is.na(tenure_info$tenure_days))

##
##      FALSE
## 2018477

## join with df
df$tenure = tenure_info$tenure_days
```

Pick two workgroup

The two group I pick is 1648 and 1722. 1600 – Biotechnology 1700 – Chemical and Materials Engineering

```

wg = as.numeric(substr(df$examiner_art_unit, 1, 3))
df$wg = wg
group_164 = df %>% filter(df$wg == 164)
group_172 = df %>% filter(df$wg == 172)

```

Examining Group 1648

```

## summary
table(group_164$gender)

```

```

##
## female    male
## 45839    47503

```

```

table(group_164$race)

```

```

##
## american_indian    asian    black    hispanic    white
##                2    24487    3974    1421    63458

```

Examining Group 1722

```

## summary
table(group_172$gender)

```

```

##
## female    male
## 22865    56330

```

```

table(group_172$race)

```

```

##
## american_indian    asian    black    hispanic    white
##                4    18624    1065    2225    57277

```

```

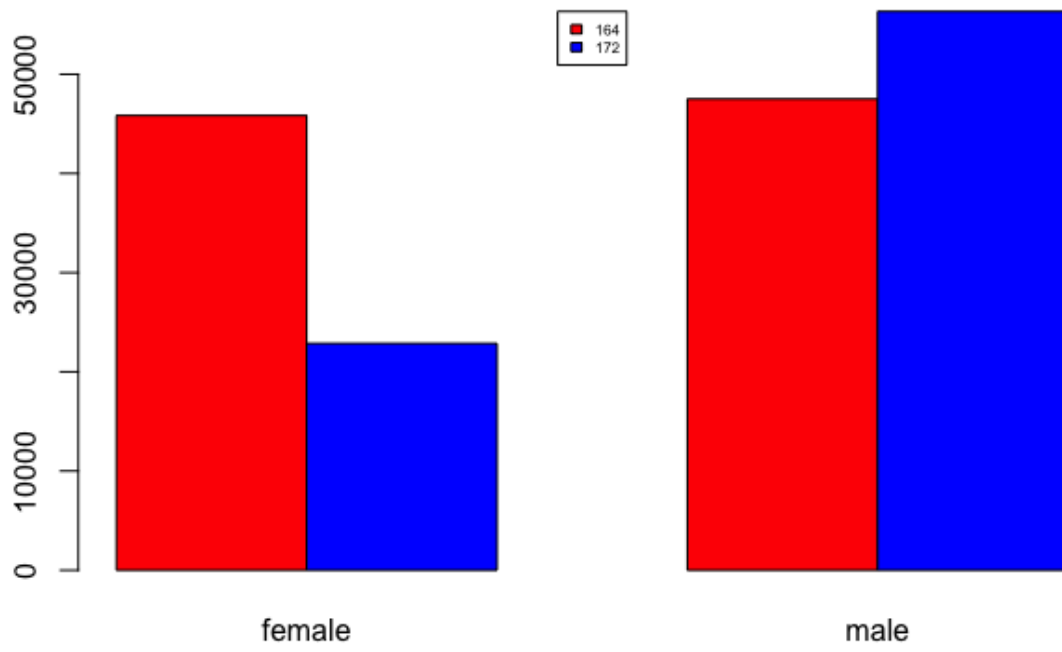
two_group_gender <- t(cbind(table(group_164$gender), table(group_172$gender)))

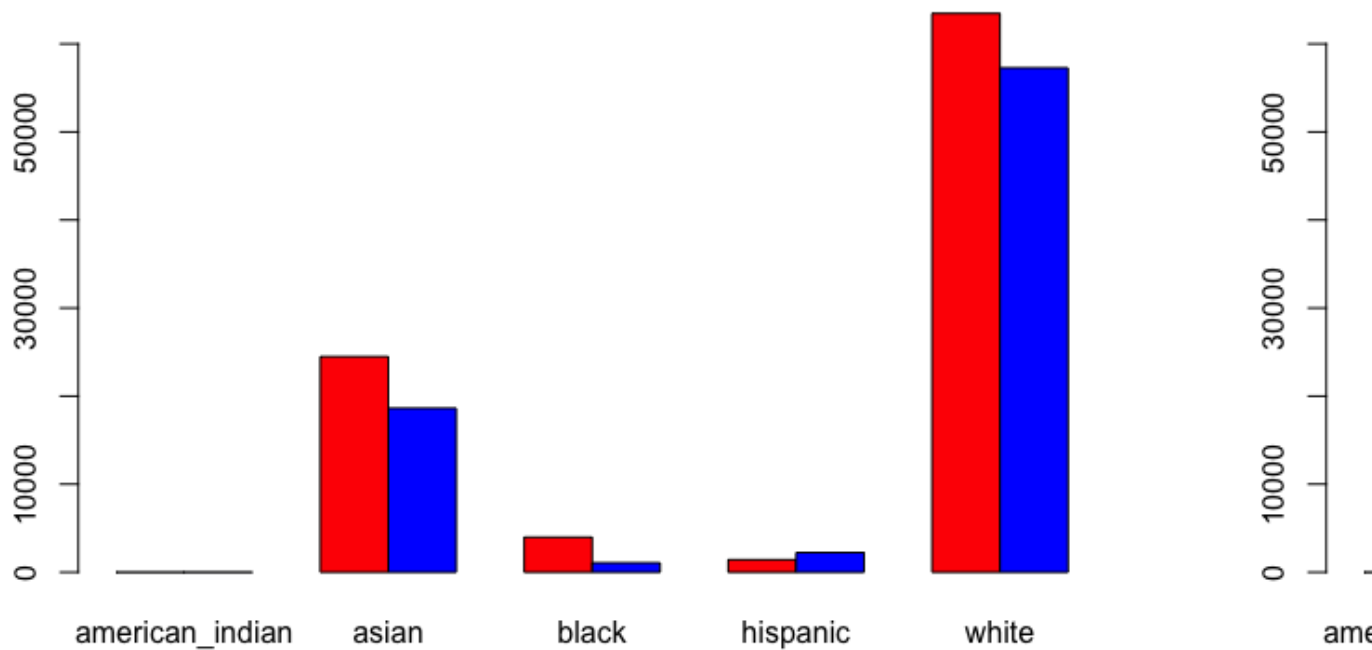
```

```

barplot(two_group_gender, beside=T, col=c("red","blue"))
par(xpd=T)
legend("top", legend = c("164", "172"), fill=c("red", "blue"), cex = 0.5)

```





Create advice networks from edges-sample

```
library(tidyverse)
net = read_csv("~/Desktop/McGill/ORGB/2022-ona-assignments/ex3/edges_sample.csv")

edges_164 = inner_join(df %>% filter(wg == 164), net, by = c("application_number" = "application_number"))

colnames(edges_164) = c("from", "to", "art_unit")
edges_164 = drop_na(edges_164)

edges_172 = inner_join(df %>% filter(wg == 172), net, by = c("application_number" = "application_number"))

colnames(edges_172) = c("from", "to", "art_unit")
edges_172 = drop_na(edges_172)
```

Create Nodes

```
edges = rbind(edges_164, edges_172)
node_ego = edges %>% select(from, art_unit) %>% rename(id=from)
node_alter = edges %>% select(to, art_unit) %>% rename(id=to)
nodes_all <- rbind.data.frame(node_ego, node_alter)

nodes = nodes_all %>% distinct(id)
```

```
nodes = nodes %>% mutate(id = as.character(id))
```

Create Graph

```
library(igraph)
```

```
##  
## Attaching package: 'igraph'  
  
## The following objects are masked from 'package:purrr':  
##  
##   compose, simplify  
  
## The following object is masked from 'package:tidyr':  
##  
##   crossing  
  
## The following object is masked from 'package:tibble':  
##  
##   as_data_frame  
  
## The following objects are masked from 'package:lubridate':  
##  
##   %--%, union  
  
## The following objects are masked from 'package:dplyr':  
##  
##   as_data_frame, groups, union  
  
## The following objects are masked from 'package:stats':  
##  
##   decompose, spectrum  
  
## The following object is masked from 'package:base':  
##  
##   union
```

```
net_164 = graph_from_data_frame(d=edges_164, vertices=nodes, directed=TRUE)  
net_164
```

```
## IGRAPH abe8745 DN-- 382 1320 --  
## + attr: name (v/c), art_unit (e/n)  
## + edges from abe8745 (vertex names):  
## [1] 91688->71059 91688->67669 97910->59738 97910->99004 97910->67669  
## [6] 75775->69583 75775->83794 75775->70306 75775->91151 75775->71534  
## [11] 70204->72882 70204->94911 71120->65790 59338->72882 61757->65024  
## [16] 61757->72882 60067->91747 60067->71087 60067->73722 60067->81365  
## [21] 96963->72882 97910->65790 97910->59738 97910->99004 93839->71946  
## [26] 74224->65024 74224->94911 96963->67657 87897->69583 87897->72882  
## [31] 75775->69583 75775->83794 75775->70306 93839->67669 93839->71946
```

```
## [36] 93839->67669 93839->95981 75775->69583 75775->69583 75775->69583
## + ... omitted several edges
```

```
net_172 = graph_from_data_frame(d=edges_172, vertices=nodes, directed=TRUE)
```

Pick the measure of centrality

1. Degree centrality is defined as the number of links incident upon a node
2. Eigenvector Centrality is an algorithm that measures the transitive influence of nodes. A high eigenvector score means that a node is connected to many nodes who themselves have high scores.
3. Closeness centrality is a measure of the average shortest distance from each vertex to each other vertex
4. Betweenness centrality is a way of detecting the amount of influence a node has over the flow of information in a graph.

```
## Degree Centrality
V(net_164)$dc <- degree(net_164)
V(net_172)$dc <- degree(net_172)

## Eigenvector Centrality
V(net_164)$ec <- evcent(net_164)$vector
V(net_172)$ec <- evcent(net_172)$vector

## Closeness Centrality
V(net_164)$cc <- closeness(net_164)
V(net_172)$cc <- closeness(net_172)

## Betweenness Centrality
V(net_164)$bc <- betweenness(net_164)
V(net_172)$bc <- betweenness(net_172)
```

Plot the network based on centrality

```
library(ggraph)
library(ggplot2)
library(ggpubr)

# Degree Centrality
dc_164 = ggraph(net_164, layout="kk") +
  geom_edge_link() +
  geom_node_point(aes(size=dc), show.legend=T) + ggtitle("Degree Centrality 164")

# Eigenvector Centrality
ec_164<-ggraph(net_164, layout="kk") +
  geom_edge_link() +
  geom_node_point(aes(size=ec), show.legend=T) + ggtitle("Eigenvector Centrality 164")

# Closeness Centrality
cc_164<-ggraph(net_164, layout="kk") +
  geom_edge_link() +
  geom_node_point(aes(size=cc), show.legend=T) + ggtitle("Closeness Centrality 164")
```



```
# Betweenness Centrality
bc_164<-ggraph(net_164, layout="kk") +
  geom_edge_link()+
  geom_node_point(aes(size=bc), show.legend=T) + ggtitle("Betweenness Centrality 164")
```

Centrality Scores

```
centrality_164 <- data.frame(
  id = V(net_164)$name,
  degree      = V(net_164)$dc,
  closeness   = V(net_164)$cc,
  betweenness = V(net_164)$bc,
  eigenvector = V(net_164)$ec)
centrality_164
```

##	id	degree	closeness	betweenness	eigenvector
## 1	91688	2	0.50000000	0.0000000	7.440125e-04
## 2	97910	170	0.01098901	0.0000000	1.000000e+00
## 3	75775	74	0.05882353	0.0000000	5.289362e-02
## 4	70204	50	0.14285714	0.0000000	2.087178e-01
## 5	71120	1	1.00000000	0.0000000	7.229780e-04
## 6	59338	17	0.07142857	0.0000000	2.368914e-02
## 7	61757	15	0.12500000	0.0000000	4.667189e-02
## 8	60067	53	0.06250000	0.0000000	2.659651e-01
## 9	96963	25	0.12500000	26.5000000	3.052454e-02
## 10	93839	33	0.14285714	6.0000000	2.323823e-02
## 11	74224	62	0.12500000	0.0000000	2.960134e-02
## 12	87897	96	0.01694915	0.0000000	1.828619e-01
## 13	97706	22	1.00000000	0.0000000	0.000000e+00
## 14	95604	27	0.04347826	7.3000000	3.620639e-02
## 15	68665	14	0.14285714	0.0000000	2.936425e-02
## 16	59211	43	0.07692308	22.0000000	1.486958e-02
## 17	94046	28	0.03846154	0.0000000	1.094606e-02
## 18	98563	21	0.05555556	0.0000000	1.296573e-02
## 19	84944	2	0.50000000	0.0000000	0.000000e+00
## 20	64659	34	0.16666667	0.0000000	2.892049e-01
## 21	75380	3	0.33333333	0.0000000	8.545698e-03
## 22	71195	1	1.00000000	0.0000000	7.229780e-04
## 23	81117	22	0.14285714	0.0000000	3.273904e-03
## 24	96143	13	0.16666667	0.0000000	5.957523e-02
## 25	97543	13	0.12500000	0.0000000	1.508066e-02
## 26	84896	4	0.25000000	0.0000000	9.202491e-03
## 27	75730	8	0.25000000	0.0000000	2.255907e-03
## 28	62583	21	0.20000000	0.0000000	1.511536e-02
## 29	92784	23	0.33333333	0.0000000	5.269872e-03
## 30	75568	44	0.08333333	0.0000000	2.080392e-01
## 31	85865	22	0.50000000	0.0000000	4.173187e-03
## 32	68436	5	0.25000000	0.0000000	8.786725e-04
## 33	61276	21	0.12500000	2.0000000	9.847097e-02
## 34	85987	24	0.12500000	6.0000000	2.070885e-01
## 35	61048	1	1.00000000	0.0000000	7.438540e-04

## 36	93421	17	0.09090909	0.0000000	5.460819e-02
## 37	72241	7	0.20000000	0.0000000	2.305370e-02
## 38	68922	9	0.14285714	0.0000000	4.583718e-03
## 39	83794	29	0.50000000	16.0000000	3.148191e-02
## 40	63511	13	0.20000000	0.0000000	8.683155e-03
## 41	93869	4	0.16666667	0.0000000	1.535799e-02
## 42	59693	8	0.05263158	0.0000000	2.010957e-03
## 43	79980	2	0.50000000	0.0000000	4.869150e-04
## 44	71259	17	0.14285714	0.0000000	1.169589e-02
## 45	63470	5	0.50000000	0.0000000	1.068662e-03
## 46	64064	4	0.50000000	0.0000000	4.033194e-02
## 47	82047	57	0.04347826	7.6000000	3.348116e-02
## 48	62397	31	0.10000000	0.6666667	1.965911e-01
## 49	90331	11	0.09090909	1.1000000	5.266137e-02
## 50	71414	12	0.20000000	0.0000000	1.801370e-02
## 51	92572	9	0.12500000	5.0000000	2.442997e-02
## 52	61558	2	0.50000000	0.0000000	7.442756e-03
## 53	63394	5	0.04000000	0.0000000	8.039191e-04
## 54	66336	2	0.50000000	0.0000000	4.540176e-05
## 55	96523	5	0.20000000	0.0000000	3.425991e-04
## 56	65654	6	0.14285714	4.0000000	2.454711e-03
## 57	64507	4	1.00000000	2.3333333	2.927285e-02
## 58	59497	6	0.12500000	0.0000000	1.095277e-03
## 59	79495	10	0.14285714	0.0000000	3.255666e-02
## 60	73260	36	0.03030303	0.0000000	3.656860e-02
## 61	62001	1	1.00000000	0.0000000	4.844712e-04
## 62	98076	5	0.33333333	0.0000000	4.055511e-02
## 63	71437	2	1.00000000	0.0000000	6.046637e-04
## 64	64169	4	1.00000000	0.0000000	2.887833e-02
## 65	86115	6	0.25000000	0.0000000	8.494915e-03
## 66	59658	2	0.50000000	0.0000000	0.000000e+00
## 67	96027	10	0.50000000	0.0000000	3.629112e-02
## 68	71858	4	0.25000000	0.0000000	1.207964e-03
## 69	64445	1	1.00000000	0.0000000	1.237095e-03
## 70	73788	3	0.33333333	0.0000000	1.972022e-03
## 71	97772	5	0.33333333	0.0000000	1.763633e-02
## 72	68445	2	0.50000000	0.0000000	7.942560e-03
## 73	80908	3	0.50000000	0.0000000	0.000000e+00
## 74	97072	1	1.00000000	0.0000000	0.000000e+00
## 75	86403	1	1.00000000	0.0000000	0.000000e+00
## 76	84470	4	0.33333333	0.0000000	1.930838e-03
## 77	96576	1	1.00000000	0.0000000	6.705565e-06
## 78	64506	2	0.50000000	0.0000000	8.243366e-03
## 79	96388	3	0.50000000	0.0000000	3.218043e-03
## 80	76749	15	1.00000000	0.0000000	0.000000e+00
## 81	75243	4	0.25000000	0.0000000	0.000000e+00
## 82	90863	11	1.00000000	1.0000000	1.486915e-02
## 83	90219	9	0.33333333	1.5000000	1.114776e-02
## 84	94719	1	1.00000000	0.0000000	1.117014e-04
## 85	88223	1	1.00000000	0.0000000	5.127562e-05
## 86	98470	2	0.50000000	0.0000000	8.456677e-03
## 87	71089	1	1.00000000	0.0000000	7.438540e-04
## 88	74425	4	0.25000000	0.0000000	6.465110e-05
## 89	95634	3	0.08333333	0.0000000	8.122305e-05

## 90	97818	0	NaN	0.0000000	0.000000e+00
## 91	87125	0	NaN	0.0000000	0.000000e+00
## 92	97456	0	NaN	0.0000000	0.000000e+00
## 93	93968	0	NaN	0.0000000	0.000000e+00
## 94	62152	0	NaN	0.0000000	0.000000e+00
## 95	94782	0	NaN	0.0000000	0.000000e+00
## 96	67034	0	NaN	0.0000000	0.000000e+00
## 97	92108	0	NaN	0.0000000	0.000000e+00
## 98	91190	0	NaN	0.0000000	0.000000e+00
## 99	98469	0	NaN	0.0000000	0.000000e+00
## 100	95891	0	NaN	0.0000000	0.000000e+00
## 101	73898	0	NaN	0.0000000	0.000000e+00
## 102	59550	0	NaN	0.0000000	0.000000e+00
## 103	98855	0	NaN	0.0000000	0.000000e+00
## 104	97755	0	NaN	0.0000000	0.000000e+00
## 105	83034	0	NaN	0.0000000	0.000000e+00
## 106	62659	0	NaN	0.0000000	0.000000e+00
## 107	76580	0	NaN	0.0000000	0.000000e+00
## 108	75530	0	NaN	0.0000000	0.000000e+00
## 109	85323	0	NaN	0.0000000	0.000000e+00
## 110	74334	0	NaN	0.0000000	0.000000e+00
## 111	78715	0	NaN	0.0000000	0.000000e+00
## 112	94466	0	NaN	0.0000000	0.000000e+00
## 113	84267	0	NaN	0.0000000	0.000000e+00
## 114	92476	0	NaN	0.0000000	0.000000e+00
## 115	75367	0	NaN	0.0000000	0.000000e+00
## 116	65121	0	NaN	0.0000000	0.000000e+00
## 117	90290	0	NaN	0.0000000	0.000000e+00
## 118	66251	0	NaN	0.0000000	0.000000e+00
## 119	95010	0	NaN	0.0000000	0.000000e+00
## 120	63714	0	NaN	0.0000000	0.000000e+00
## 121	69581	0	NaN	0.0000000	0.000000e+00
## 122	86985	0	NaN	0.0000000	0.000000e+00
## 123	61517	0	NaN	0.0000000	0.000000e+00
## 124	69680	0	NaN	0.0000000	0.000000e+00
## 125	68719	0	NaN	0.0000000	0.000000e+00
## 126	62413	0	NaN	0.0000000	0.000000e+00
## 127	97083	0	NaN	0.0000000	0.000000e+00
## 128	71059	1	NaN	0.0000000	1.085728e-05
## 129	67669	33	NaN	0.0000000	5.097378e-02
## 130	59738	83	NaN	0.0000000	6.982489e-01
## 131	99004	82	NaN	0.0000000	6.836561e-01
## 132	69583	45	NaN	0.0000000	7.015643e-02
## 133	70306	11	NaN	0.0000000	8.490570e-03
## 134	91151	5	NaN	0.0000000	3.859350e-03
## 135	71534	5	NaN	0.0000000	3.859350e-03
## 136	72882	154	NaN	0.0000000	4.947334e-01
## 137	94911	47	NaN	0.0000000	5.922194e-02
## 138	65790	29	NaN	0.0000000	4.954321e-02
## 139	65024	65	NaN	0.0000000	2.618701e-02
## 140	91747	2	NaN	0.0000000	4.369782e-03
## 141	71087	7	NaN	0.0000000	1.123998e-02
## 142	73722	1	NaN	0.0000000	3.881195e-03
## 143	81365	1	NaN	0.0000000	3.881195e-03

## 144 71946	30	NaN	0.0000000	7.298222e-02
## 145 67657	2	NaN	0.0000000	8.908816e-04
## 146 95981	2	NaN	0.0000000	8.727535e-04
## 147 75119	22	NaN	0.0000000	0.000000e+00
## 148 70435	6	NaN	0.0000000	3.170132e-03
## 149 96738	3	NaN	0.0000000	1.529330e-02
## 150 64280	27	NaN	0.0000000	2.826566e-03
## 151 92578	3	NaN	0.0000000	1.227017e-03
## 152 99763	1	NaN	0.0000000	1.892073e-04
## 153 66266	61	NaN	0.0000000	8.477388e-02
## 154 68637	1	NaN	0.0000000	0.000000e+00
## 155 97834	1	NaN	0.0000000	0.000000e+00
## 156 85381	24	NaN	0.0000000	2.008683e-03
## 157 72903	3	NaN	0.0000000	2.608122e-03
## 158 99002	1	NaN	0.0000000	1.342908e-04
## 159 81987	3	NaN	0.0000000	1.327300e-04
## 160 99254	11	NaN	0.0000000	7.423058e-02
## 161 60423	11	NaN	0.0000000	7.423058e-02
## 162 67905	9	NaN	0.0000000	2.464362e-02
## 163 64652	9	NaN	0.0000000	2.464362e-02
## 164 83254	26	NaN	0.0000000	7.179617e-03
## 165 87486	18	NaN	0.0000000	7.654516e-03
## 166 88924	2	NaN	0.0000000	6.044032e-03
## 167 72178	2	NaN	0.0000000	6.044032e-03
## 168 72820	6	NaN	0.0000000	1.915633e-03
## 169 67149	1	NaN	0.0000000	3.035890e-03
## 170 61318	2	NaN	0.0000000	4.881891e-04
## 171 94285	1	NaN	0.0000000	1.459288e-02
## 172 63519	5	NaN	0.0000000	1.620961e-02
## 173 92569	16	NaN	0.0000000	2.071777e-02
## 174 65312	4	NaN	0.0000000	6.237938e-05
## 175 78488	1	NaN	0.0000000	1.267122e-04
## 176 90287	1	NaN	0.0000000	3.022016e-03
## 177 81014	1	NaN	0.0000000	3.022016e-03
## 178 63127	9	NaN	0.0000000	4.730153e-03
## 179 97470	13	NaN	0.0000000	5.868037e-03
## 180 99413	10	NaN	0.0000000	4.402277e-03
## 181 95212	10	NaN	0.0000000	4.402277e-03
## 182 67634	10	NaN	0.0000000	4.402277e-03
## 183 72579	10	NaN	0.0000000	4.402277e-03
## 184 82244	1	NaN	0.0000000	4.885864e-04
## 185 69917	27	NaN	0.0000000	1.619306e-01
## 186 88567	2	NaN	0.0000000	2.918575e-02
## 187 65446	28	NaN	0.0000000	6.449131e-02
## 188 71175	1	NaN	0.0000000	3.456927e-04
## 189 67426	1	NaN	0.0000000	3.456927e-04
## 190 63963	1	NaN	0.0000000	3.456927e-04
## 191 85136	2	NaN	0.0000000	3.024984e-03
## 192 72514	2	NaN	0.0000000	3.024984e-03
## 193 66508	2	NaN	0.0000000	3.024984e-03
## 194 70858	4	NaN	0.0000000	3.073559e-03
## 195 67701	3	NaN	0.0000000	3.766911e-05
## 196 63226	1	NaN	0.0000000	7.968905e-04
## 197 71828	1	NaN	0.0000000	3.364198e-04

## 198 76315	4	NaN	0.0000000	3.513743e-03
## 199 60958	2	NaN	0.0000000	7.878532e-04
## 200 63190	1	NaN	0.0000000	2.668480e-03
## 201 91232	2	NaN	0.0000000	1.536475e-02
## 202 72912	21	NaN	0.0000000	3.319916e-02
## 203 71174	1	NaN	0.0000000	4.285089e-04
## 204 84269	2	NaN	0.0000000	4.339800e-04
## 205 66378	2	NaN	0.0000000	5.345844e-04
## 206 66565	2	NaN	0.0000000	5.345844e-04
## 207 97329	2	NaN	0.0000000	5.345844e-04
## 208 70201	1	NaN	0.0000000	1.239652e-04
## 209 60487	2	NaN	0.0000000	7.762391e-03
## 210 74382	1	NaN	0.0000000	3.881195e-03
## 211 96339	6	NaN	0.0000000	3.790662e-03
## 212 72158	1	NaN	0.0000000	2.169900e-04
## 213 80106	5	NaN	0.0000000	3.856172e-03
## 214 90742	1	NaN	0.0000000	2.169900e-04
## 215 95085	1	NaN	0.0000000	0.000000e+00
## 216 62142	1	NaN	0.0000000	0.000000e+00
## 217 96865	5	NaN	0.0000000	2.647959e-03
## 218 59803	2	NaN	0.0000000	1.607651e-03
## 219 82997	1	NaN	0.0000000	6.688963e-05
## 220 88294	1	NaN	0.0000000	4.319686e-04
## 221 76280	1	NaN	0.0000000	1.239652e-04
## 222 62823	1	NaN	0.0000000	5.336410e-04
## 223 70907	1	NaN	0.0000000	0.000000e+00
## 224 72352	2	NaN	0.0000000	0.000000e+00
## 225 61013	1	NaN	0.0000000	4.285089e-04
## 226 95478	1	NaN	0.0000000	3.292017e-05
## 227 91016	1	NaN	0.0000000	3.456927e-04
## 228 97402	1	NaN	0.0000000	4.777568e-05
## 229 62947	2	NaN	0.0000000	1.296836e-03
## 230 67013	2	NaN	0.0000000	0.000000e+00
## 231 95784	1	NaN	0.0000000	3.035890e-03
## 232 85761	2	NaN	0.0000000	3.196648e-05
## 233 60045	3	NaN	0.0000000	4.377863e-02
## 234 70164	1	NaN	0.0000000	1.459288e-02
## 235 95084	3	NaN	0.0000000	4.377863e-02
## 236 95494	1	NaN	0.0000000	6.688963e-05
## 237 70079	1	NaN	0.0000000	6.688963e-05
## 238 90154	2	NaN	0.0000000	4.401405e-04
## 239 75394	2	NaN	0.0000000	7.762391e-03
## 240 62411	1	NaN	0.0000000	2.817648e-05
## 241 95223	2	NaN	0.0000000	4.595095e-04
## 242 61105	1	NaN	0.0000000	1.459288e-02
## 243 65131	1	NaN	0.0000000	1.459288e-02
## 244 70142	1	NaN	0.0000000	1.459288e-02
## 245 72355	1	NaN	0.0000000	4.319686e-04
## 246 72814	3	NaN	0.0000000	1.564959e-02
## 247 77761	2	NaN	0.0000000	8.908816e-04
## 248 67435	5	NaN	0.0000000	0.000000e+00
## 249 98700	2	NaN	0.0000000	3.413534e-04
## 250 67138	1	NaN	0.0000000	2.668480e-03
## 251 65737	1	NaN	0.0000000	1.762766e-05

##	252	93990	1	NaN	0.0000000	1.762766e-05
##	253	94153	1	NaN	0.0000000	0.000000e+00
##	254	93982	1	NaN	0.0000000	0.000000e+00
##	255	98518	1	NaN	0.0000000	0.000000e+00
##	256	64331	1	NaN	0.0000000	0.000000e+00
##	257	80823	1	NaN	0.0000000	2.668480e-03
##	258	88668	1	NaN	0.0000000	7.718700e-04
##	259	66225	1	NaN	0.0000000	2.668480e-03
##	260	60791	1	NaN	0.0000000	2.668480e-03
##	261	59908	1	NaN	0.0000000	2.668480e-03
##	262	76423	1	NaN	0.0000000	2.668480e-03
##	263	94770	1	NaN	0.0000000	2.668480e-03
##	264	76309	1	NaN	0.0000000	2.668480e-03
##	265	91849	1	NaN	0.0000000	2.668480e-03
##	266	69467	1	NaN	0.0000000	2.668480e-03
##	267	70892	1	NaN	0.0000000	2.668480e-03
##	268	97332	1	NaN	0.0000000	2.668480e-03
##	269	90288	1	NaN	0.0000000	2.668480e-03
##	270	86743	1	NaN	0.0000000	2.668480e-03
##	271	61388	1	NaN	0.0000000	2.668480e-03
##	272	95253	1	NaN	0.0000000	2.668480e-03
##	273	68813	1	NaN	0.0000000	2.668480e-03
##	274	97610	1	NaN	0.0000000	2.668480e-03
##	275	93800	1	NaN	0.0000000	2.668480e-03
##	276	94355	1	NaN	0.0000000	2.668480e-03
##	277	80910	1	NaN	0.0000000	2.668480e-03
##	278	60266	1	NaN	0.0000000	2.668480e-03
##	279	71528	1	NaN	0.0000000	2.668480e-03
##	280	59868	1	NaN	0.0000000	2.668480e-03
##	281	79106	1	NaN	0.0000000	2.668480e-03
##	282	64004	1	NaN	0.0000000	1.459288e-02
##	283	66448	1	NaN	0.0000000	2.668480e-03
##	284	93578	1	NaN	0.0000000	5.336410e-04
##	285	68916	1	NaN	0.0000000	5.336410e-04
##	286	99524	1	NaN	0.0000000	5.336410e-04
##	287	94123	1	NaN	0.0000000	5.336410e-04
##	288	88661	1	NaN	0.0000000	5.336410e-04
##	289	65304	1	NaN	0.0000000	5.336410e-04
##	290	89412	1	NaN	0.0000000	5.336410e-04
##	291	65541	1	NaN	0.0000000	5.336410e-04
##	292	75560	1	NaN	0.0000000	5.336410e-04
##	293	92268	1	NaN	0.0000000	2.668480e-03
##	294	63694	1	NaN	0.0000000	2.668480e-03
##	295	59258	1	NaN	0.0000000	2.668480e-03
##	296	99480	1	NaN	0.0000000	2.668480e-03
##	297	68265	1	NaN	0.0000000	2.668480e-03
##	298	62591	1	NaN	0.0000000	2.668480e-03
##	299	64355	2	NaN	0.0000000	2.295674e-04
##	300	98182	1	NaN	0.0000000	4.285089e-04
##	301	94915	1	NaN	0.0000000	4.750952e-04
##	302	95525	1	NaN	0.0000000	4.750952e-04
##	303	92706	1	NaN	0.0000000	2.668480e-03
##	304	94847	1	NaN	0.0000000	2.668480e-03
##	305	67620	1	NaN	0.0000000	2.668480e-03

## 306 99892	1	NaN	0.0000000	2.668480e-03
## 307 77063	2	NaN	0.0000000	1.056711e-03
## 308 61925	1	NaN	0.0000000	1.459288e-02
## 309 68977	1	NaN	0.0000000	1.459288e-02
## 310 73239	1	NaN	0.0000000	5.283554e-04
## 311 75932	1	NaN	0.0000000	2.628716e-04
## 312 69378	0	NaN	0.0000000	0.000000e+00
## 313 77915	0	NaN	0.0000000	0.000000e+00
## 314 61863	0	NaN	0.0000000	0.000000e+00
## 315 95344	0	NaN	0.0000000	0.000000e+00
## 316 78379	0	NaN	0.0000000	0.000000e+00
## 317 92991	0	NaN	0.0000000	0.000000e+00
## 318 62499	0	NaN	0.0000000	0.000000e+00
## 319 65621	0	NaN	0.0000000	0.000000e+00
## 320 75933	0	NaN	0.0000000	0.000000e+00
## 321 98489	0	NaN	0.0000000	0.000000e+00
## 322 60203	0	NaN	0.0000000	0.000000e+00
## 323 67829	0	NaN	0.0000000	0.000000e+00
## 324 60851	0	NaN	0.0000000	0.000000e+00
## 325 74052	0	NaN	0.0000000	0.000000e+00
## 326 98826	0	NaN	0.0000000	0.000000e+00
## 327 65934	0	NaN	0.0000000	0.000000e+00
## 328 97338	0	NaN	0.0000000	0.000000e+00
## 329 99489	0	NaN	0.0000000	0.000000e+00
## 330 95299	0	NaN	0.0000000	0.000000e+00
## 331 96697	0	NaN	0.0000000	0.000000e+00
## 332 69099	0	NaN	0.0000000	0.000000e+00
## 333 98852	0	NaN	0.0000000	0.000000e+00
## 334 90995	0	NaN	0.0000000	0.000000e+00
## 335 70248	0	NaN	0.0000000	0.000000e+00
## 336 77651	0	NaN	0.0000000	0.000000e+00
## 337 77184	0	NaN	0.0000000	0.000000e+00
## 338 97312	0	NaN	0.0000000	0.000000e+00
## 339 72666	0	NaN	0.0000000	0.000000e+00
## 340 74048	0	NaN	0.0000000	0.000000e+00
## 341 62778	0	NaN	0.0000000	0.000000e+00
## 342 78856	0	NaN	0.0000000	0.000000e+00
## 343 67890	0	NaN	0.0000000	0.000000e+00
## 344 61408	0	NaN	0.0000000	0.000000e+00
## 345 99787	0	NaN	0.0000000	0.000000e+00
## 346 86422	0	NaN	0.0000000	0.000000e+00
## 347 70788	0	NaN	0.0000000	0.000000e+00
## 348 77228	0	NaN	0.0000000	0.000000e+00
## 349 69511	0	NaN	0.0000000	0.000000e+00
## 350 66182	0	NaN	0.0000000	0.000000e+00
## 351 71397	0	NaN	0.0000000	0.000000e+00
## 352 95065	0	NaN	0.0000000	0.000000e+00
## 353 81984	0	NaN	0.0000000	0.000000e+00
## 354 60437	0	NaN	0.0000000	0.000000e+00
## 355 80181	0	NaN	0.0000000	0.000000e+00
## 356 67698	0	NaN	0.0000000	0.000000e+00
## 357 95160	0	NaN	0.0000000	0.000000e+00
## 358 94771	0	NaN	0.0000000	0.000000e+00
## 359 98738	0	NaN	0.0000000	0.000000e+00

```
## 360 63372      0      NaN  0.0000000 0.000000e+00
## 361 95432      0      NaN  0.0000000 0.000000e+00
## 362 59269      0      NaN  0.0000000 0.000000e+00
## 363 63482      0      NaN  0.0000000 0.000000e+00
## 364 94860      0      NaN  0.0000000 0.000000e+00
## 365 89466      0      NaN  0.0000000 0.000000e+00
## 366 95666      0      NaN  0.0000000 0.000000e+00
## 367 89404      0      NaN  0.0000000 0.000000e+00
## 368 73564      0      NaN  0.0000000 0.000000e+00
## 369 64184      0      NaN  0.0000000 0.000000e+00
## 370 67547      0      NaN  0.0000000 0.000000e+00
## 371 67514      0      NaN  0.0000000 0.000000e+00
## 372 97267      0      NaN  0.0000000 0.000000e+00
## 373 64350      0      NaN  0.0000000 0.000000e+00
## 374 99577      0      NaN  0.0000000 0.000000e+00
## 375 91123      0      NaN  0.0000000 0.000000e+00
## 376 88204      0      NaN  0.0000000 0.000000e+00
## 377 97922      0      NaN  0.0000000 0.000000e+00
## 378 94768      0      NaN  0.0000000 0.000000e+00
## 379 81287      0      NaN  0.0000000 0.000000e+00
## 380 66203      0      NaN  0.0000000 0.000000e+00
## 381 82412      0      NaN  0.0000000 0.000000e+00
## 382 62098      0      NaN  0.0000000 0.000000e+00
```

```
centrality_172 <- data.frame(id = V(net_164)$name,
                             degree      = V(net_172)$dc,
                             closeness   = V(net_172)$cc,
                             betweenness = V(net_172)$bc,
                             eigenvector = V(net_172)$ec)
centrality_172
```

```
##      id degree closeness betweenness eigenvector
## 1   91688      0      NaN           0 4.356715e-17
## 2   97910      0      NaN           0 4.356715e-17
## 3   75775      0      NaN           0 4.356715e-17
## 4   70204      0      NaN           0 4.356715e-17
## 5   71120      0      NaN           0 4.356715e-17
## 6   59338      0      NaN           0 4.356715e-17
## 7   61757      0      NaN           0 4.356715e-17
## 8   60067      0      NaN           0 4.356715e-17
## 9   96963      0      NaN           0 4.356715e-17
## 10  93839      0      NaN           0 4.356715e-17
## 11  74224      0      NaN           0 4.356715e-17
## 12  87897      0      NaN           0 4.356715e-17
## 13  97706      0      NaN           0 4.356715e-17
## 14  95604      0      NaN           0 4.356715e-17
## 15  68665      0      NaN           0 4.356715e-17
## 16  59211      0      NaN           0 4.356715e-17
## 17  94046      0      NaN           0 4.356715e-17
## 18  98563      0      NaN           0 4.356715e-17
## 19  84944      0      NaN           0 4.356715e-17
## 20  64659      0      NaN           0 4.356715e-17
## 21  75380      0      NaN           0 4.356715e-17
## 22  71195      0      NaN           0 4.356715e-17
```


## 23	81117	0	NaN	0 4.356715e-17
## 24	96143	0	NaN	0 4.356715e-17
## 25	97543	0	NaN	0 4.356715e-17
## 26	84896	0	NaN	0 4.356715e-17
## 27	75730	0	NaN	0 4.356715e-17
## 28	62583	0	NaN	0 4.356715e-17
## 29	92784	0	NaN	0 4.356715e-17
## 30	75568	0	NaN	0 4.356715e-17
## 31	85865	0	NaN	0 4.356715e-17
## 32	68436	0	NaN	0 4.356715e-17
## 33	61276	0	NaN	0 4.356715e-17
## 34	85987	0	NaN	0 4.356715e-17
## 35	61048	0	NaN	0 4.356715e-17
## 36	93421	0	NaN	0 4.356715e-17
## 37	72241	0	NaN	0 4.356715e-17
## 38	68922	0	NaN	0 4.356715e-17
## 39	83794	0	NaN	0 4.356715e-17
## 40	63511	0	NaN	0 4.356715e-17
## 41	93869	0	NaN	0 4.356715e-17
## 42	59693	0	NaN	0 4.356715e-17
## 43	79980	0	NaN	0 4.356715e-17
## 44	71259	0	NaN	0 4.356715e-17
## 45	63470	0	NaN	0 4.356715e-17
## 46	64064	0	NaN	0 4.356715e-17
## 47	82047	0	NaN	0 4.356715e-17
## 48	62397	0	NaN	0 4.356715e-17
## 49	90331	0	NaN	0 4.356715e-17
## 50	71414	0	NaN	0 4.356715e-17
## 51	92572	0	NaN	0 4.356715e-17
## 52	61558	0	NaN	0 4.356715e-17
## 53	63394	0	NaN	0 4.356715e-17
## 54	66336	0	NaN	0 4.356715e-17
## 55	96523	0	NaN	0 4.356715e-17
## 56	65654	0	NaN	0 4.356715e-17
## 57	64507	0	NaN	0 4.356715e-17
## 58	59497	0	NaN	0 4.356715e-17
## 59	79495	0	NaN	0 4.356715e-17
## 60	73260	0	NaN	0 4.356715e-17
## 61	62001	0	NaN	0 4.356715e-17
## 62	98076	0	NaN	0 4.356715e-17
## 63	71437	0	NaN	0 4.356715e-17
## 64	64169	0	NaN	0 4.356715e-17
## 65	86115	0	NaN	0 4.356715e-17
## 66	59658	0	NaN	0 4.356715e-17
## 67	96027	0	NaN	0 4.356715e-17
## 68	71858	0	NaN	0 4.356715e-17
## 69	64445	0	NaN	0 4.356715e-17
## 70	73788	0	NaN	0 4.356715e-17
## 71	97772	0	NaN	0 4.356715e-17
## 72	68445	0	NaN	0 4.356715e-17
## 73	80908	0	NaN	0 4.356715e-17
## 74	97072	0	NaN	0 4.356715e-17
## 75	86403	0	NaN	0 4.356715e-17
## 76	84470	0	NaN	0 4.356715e-17

## 77	96576	0	NaN	0	4.356715e-17
## 78	64506	0	NaN	0	4.356715e-17
## 79	96388	0	NaN	0	4.356715e-17
## 80	76749	0	NaN	0	4.356715e-17
## 81	75243	0	NaN	0	4.356715e-17
## 82	90863	0	NaN	0	4.356715e-17
## 83	90219	0	NaN	0	4.356715e-17
## 84	94719	0	NaN	0	4.356715e-17
## 85	88223	0	NaN	0	4.356715e-17
## 86	98470	0	NaN	0	4.356715e-17
## 87	71089	0	NaN	0	4.356715e-17
## 88	74425	0	NaN	0	4.356715e-17
## 89	95634	0	NaN	0	4.356715e-17
## 90	97818	7	0.16666667	0	1.603375e-02
## 91	87125	8	0.33333333	12	6.224787e-02
## 92	97456	4	0.50000000	0	1.955749e-06
## 93	93968	11	0.33333333	0	9.513752e-04
## 94	62152	12	0.16666667	0	3.004243e-02
## 95	94782	19	0.33333333	2	1.359346e-03
## 96	67034	31	0.07142857	0	1.000000e+00
## 97	92108	1	1.00000000	0	2.287811e-04
## 98	91190	6	NaN	0	1.753895e-17
## 99	98469	3	0.33333333	0	1.150516e-16
## 100	95891	12	0.25000000	0	7.014774e-05
## 101	73898	4	0.25000000	0	1.515138e-16
## 102	59550	6	0.25000000	0	4.649829e-04
## 103	98855	2	0.16666667	0	1.328568e-04
## 104	97755	1	1.00000000	0	5.318553e-05
## 105	83034	6	0.05555556	0	1.482297e-02
## 106	62659	11	0.11111111	5	7.788714e-03
## 107	76580	1	1.00000000	0	6.496768e-07
## 108	75530	1	1.00000000	0	3.519392e-17
## 109	85323	3	0.50000000	0	5.107478e-04
## 110	74334	1	0.20000000	0	2.407413e-17
## 111	78715	4	0.50000000	0	4.231461e-05
## 112	94466	1	1.00000000	0	2.287811e-04
## 113	84267	3	0.33333333	0	1.105879e-16
## 114	92476	1	1.00000000	0	3.556291e-17
## 115	75367	3	0.50000000	2	1.149087e-16
## 116	65121	1	1.00000000	0	3.722526e-17
## 117	90290	1	1.00000000	0	4.096747e-03
## 118	66251	6	0.16666667	0	2.248197e-16
## 119	95010	5	0.33333333	0	1.392886e-01
## 120	63714	2	0.50000000	0	6.767847e-17
## 121	69581	2	0.50000000	0	4.125087e-03
## 122	86985	2	0.50000000	0	4.125087e-03
## 123	61517	1	1.00000000	0	3.824385e-17
## 124	69680	3	0.11111111	0	1.064280e-02
## 125	68719	15	0.06666667	0	6.341172e-16
## 126	62413	1	1.00000000	0	3.917222e-17
## 127	97083	6	0.16666667	0	2.315594e-16
## 128	71059	0	NaN	0	4.356715e-17
## 129	67669	0	NaN	0	4.356715e-17
## 130	59738	0	NaN	0	4.356715e-17

## 131	99004	0	NaN	0 4.356715e-17
## 132	69583	0	NaN	0 4.356715e-17
## 133	70306	0	NaN	0 4.356715e-17
## 134	91151	0	NaN	0 4.356715e-17
## 135	71534	0	NaN	0 4.356715e-17
## 136	72882	0	NaN	0 4.356715e-17
## 137	94911	0	NaN	0 4.356715e-17
## 138	65790	0	NaN	0 4.356715e-17
## 139	65024	0	NaN	0 4.356715e-17
## 140	91747	0	NaN	0 4.356715e-17
## 141	71087	0	NaN	0 4.356715e-17
## 142	73722	0	NaN	0 4.356715e-17
## 143	81365	0	NaN	0 4.356715e-17
## 144	71946	0	NaN	0 4.356715e-17
## 145	67657	0	NaN	0 4.356715e-17
## 146	95981	0	NaN	0 4.356715e-17
## 147	75119	0	NaN	0 4.356715e-17
## 148	70435	0	NaN	0 4.356715e-17
## 149	96738	0	NaN	0 4.356715e-17
## 150	64280	0	NaN	0 4.356715e-17
## 151	92578	0	NaN	0 4.356715e-17
## 152	99763	0	NaN	0 4.356715e-17
## 153	66266	0	NaN	0 4.356715e-17
## 154	68637	0	NaN	0 4.356715e-17
## 155	97834	0	NaN	0 4.356715e-17
## 156	85381	0	NaN	0 4.356715e-17
## 157	72903	0	NaN	0 4.356715e-17
## 158	99002	0	NaN	0 4.356715e-17
## 159	81987	0	NaN	0 4.356715e-17
## 160	99254	0	NaN	0 4.356715e-17
## 161	60423	0	NaN	0 4.356715e-17
## 162	67905	0	NaN	0 4.356715e-17
## 163	64652	0	NaN	0 4.356715e-17
## 164	83254	0	NaN	0 4.356715e-17
## 165	87486	0	NaN	0 4.356715e-17
## 166	88924	0	NaN	0 4.356715e-17
## 167	72178	0	NaN	0 4.356715e-17
## 168	72820	0	NaN	0 4.356715e-17
## 169	67149	0	NaN	0 4.356715e-17
## 170	61318	0	NaN	0 4.356715e-17
## 171	94285	0	NaN	0 4.356715e-17
## 172	63519	0	NaN	0 4.356715e-17
## 173	92569	8	NaN	0 3.289088e-05
## 174	65312	0	NaN	0 4.356715e-17
## 175	78488	0	NaN	0 4.356715e-17
## 176	90287	0	NaN	0 4.356715e-17
## 177	81014	0	NaN	0 4.356715e-17
## 178	63127	0	NaN	0 4.356715e-17
## 179	97470	0	NaN	0 4.356715e-17
## 180	99413	0	NaN	0 4.356715e-17
## 181	95212	0	NaN	0 4.356715e-17
## 182	67634	0	NaN	0 4.356715e-17
## 183	72579	0	NaN	0 4.356715e-17
## 184	82244	0	NaN	0 4.356715e-17

## 185 69917	0	NaN	0 4.356715e-17
## 186 88567	0	NaN	0 4.356715e-17
## 187 65446	0	NaN	0 4.356715e-17
## 188 71175	0	NaN	0 4.356715e-17
## 189 67426	0	NaN	0 4.356715e-17
## 190 63963	0	NaN	0 4.356715e-17
## 191 85136	0	NaN	0 4.356715e-17
## 192 72514	0	NaN	0 4.356715e-17
## 193 66508	0	NaN	0 4.356715e-17
## 194 70858	0	NaN	0 4.356715e-17
## 195 67701	0	NaN	0 4.356715e-17
## 196 63226	0	NaN	0 4.356715e-17
## 197 71828	0	NaN	0 4.356715e-17
## 198 76315	0	NaN	0 4.356715e-17
## 199 60958	0	NaN	0 4.356715e-17
## 200 63190	0	NaN	0 4.356715e-17
## 201 91232	0	NaN	0 4.356715e-17
## 202 72912	0	NaN	0 4.356715e-17
## 203 71174	0	NaN	0 4.356715e-17
## 204 84269	0	NaN	0 4.356715e-17
## 205 66378	0	NaN	0 4.356715e-17
## 206 66565	0	NaN	0 4.356715e-17
## 207 97329	0	NaN	0 4.356715e-17
## 208 70201	0	NaN	0 4.356715e-17
## 209 60487	0	NaN	0 4.356715e-17
## 210 74382	0	NaN	0 4.356715e-17
## 211 96339	0	NaN	0 4.356715e-17
## 212 72158	0	NaN	0 4.356715e-17
## 213 80106	0	NaN	0 4.356715e-17
## 214 90742	0	NaN	0 4.356715e-17
## 215 95085	0	NaN	0 4.356715e-17
## 216 62142	0	NaN	0 4.356715e-17
## 217 96865	0	NaN	0 4.356715e-17
## 218 59803	0	NaN	0 4.356715e-17
## 219 82997	0	NaN	0 4.356715e-17
## 220 88294	0	NaN	0 4.356715e-17
## 221 76280	0	NaN	0 4.356715e-17
## 222 62823	0	NaN	0 4.356715e-17
## 223 70907	0	NaN	0 4.356715e-17
## 224 72352	0	NaN	0 4.356715e-17
## 225 61013	0	NaN	0 4.356715e-17
## 226 95478	0	NaN	0 4.356715e-17
## 227 91016	0	NaN	0 4.356715e-17
## 228 97402	0	NaN	0 4.356715e-17
## 229 62947	0	NaN	0 4.356715e-17
## 230 67013	0	NaN	0 4.356715e-17
## 231 95784	0	NaN	0 4.356715e-17
## 232 85761	0	NaN	0 4.356715e-17
## 233 60045	0	NaN	0 4.356715e-17
## 234 70164	0	NaN	0 4.356715e-17
## 235 95084	0	NaN	0 4.356715e-17
## 236 95494	0	NaN	0 4.356715e-17
## 237 70079	0	NaN	0 4.356715e-17
## 238 90154	0	NaN	0 4.356715e-17

## 239	75394	0	NaN	0 4.356715e-17
## 240	62411	0	NaN	0 4.356715e-17
## 241	95223	0	NaN	0 4.356715e-17
## 242	61105	0	NaN	0 4.356715e-17
## 243	65131	0	NaN	0 4.356715e-17
## 244	70142	0	NaN	0 4.356715e-17
## 245	72355	0	NaN	0 4.356715e-17
## 246	72814	0	NaN	0 4.356715e-17
## 247	77761	0	NaN	0 4.356715e-17
## 248	67435	0	NaN	0 4.356715e-17
## 249	98700	0	NaN	0 4.356715e-17
## 250	67138	0	NaN	0 4.356715e-17
## 251	65737	0	NaN	0 4.356715e-17
## 252	93990	0	NaN	0 4.356715e-17
## 253	94153	0	NaN	0 4.356715e-17
## 254	93982	0	NaN	0 4.356715e-17
## 255	98518	0	NaN	0 4.356715e-17
## 256	64331	0	NaN	0 4.356715e-17
## 257	80823	0	NaN	0 4.356715e-17
## 258	88668	0	NaN	0 4.356715e-17
## 259	66225	0	NaN	0 4.356715e-17
## 260	60791	0	NaN	0 4.356715e-17
## 261	59908	0	NaN	0 4.356715e-17
## 262	76423	0	NaN	0 4.356715e-17
## 263	94770	0	NaN	0 4.356715e-17
## 264	76309	0	NaN	0 4.356715e-17
## 265	91849	0	NaN	0 4.356715e-17
## 266	69467	0	NaN	0 4.356715e-17
## 267	70892	0	NaN	0 4.356715e-17
## 268	97332	0	NaN	0 4.356715e-17
## 269	90288	0	NaN	0 4.356715e-17
## 270	86743	0	NaN	0 4.356715e-17
## 271	61388	0	NaN	0 4.356715e-17
## 272	95253	0	NaN	0 4.356715e-17
## 273	68813	0	NaN	0 4.356715e-17
## 274	97610	0	NaN	0 4.356715e-17
## 275	93800	0	NaN	0 4.356715e-17
## 276	94355	0	NaN	0 4.356715e-17
## 277	80910	0	NaN	0 4.356715e-17
## 278	60266	0	NaN	0 4.356715e-17
## 279	71528	0	NaN	0 4.356715e-17
## 280	59868	0	NaN	0 4.356715e-17
## 281	79106	0	NaN	0 4.356715e-17
## 282	64004	0	NaN	0 4.356715e-17
## 283	66448	0	NaN	0 4.356715e-17
## 284	93578	0	NaN	0 4.356715e-17
## 285	68916	0	NaN	0 4.356715e-17
## 286	99524	0	NaN	0 4.356715e-17
## 287	94123	0	NaN	0 4.356715e-17
## 288	88661	0	NaN	0 4.356715e-17
## 289	65304	0	NaN	0 4.356715e-17
## 290	89412	0	NaN	0 4.356715e-17
## 291	65541	0	NaN	0 4.356715e-17
## 292	75560	0	NaN	0 4.356715e-17

## 293	92268	0	NaN	0 4.356715e-17
## 294	63694	0	NaN	0 4.356715e-17
## 295	59258	0	NaN	0 4.356715e-17
## 296	99480	0	NaN	0 4.356715e-17
## 297	68265	0	NaN	0 4.356715e-17
## 298	62591	0	NaN	0 4.356715e-17
## 299	64355	0	NaN	0 4.356715e-17
## 300	98182	0	NaN	0 4.356715e-17
## 301	94915	0	NaN	0 4.356715e-17
## 302	95525	0	NaN	0 4.356715e-17
## 303	92706	0	NaN	0 4.356715e-17
## 304	94847	0	NaN	0 4.356715e-17
## 305	67620	0	NaN	0 4.356715e-17
## 306	99892	0	NaN	0 4.356715e-17
## 307	77063	0	NaN	0 4.356715e-17
## 308	61925	0	NaN	0 4.356715e-17
## 309	68977	0	NaN	0 4.356715e-17
## 310	73239	0	NaN	0 4.356715e-17
## 311	75932	0	NaN	0 4.356715e-17
## 312	69378	4	NaN	0 1.190286e-01
## 313	77915	4	NaN	0 6.225029e-02
## 314	61863	4	NaN	0 9.997225e-03
## 315	95344	4	NaN	0 9.997225e-03
## 316	78379	1	NaN	0 3.648349e-03
## 317	92991	1	NaN	0 1.146265e-07
## 318	62499	7	NaN	0 1.108474e-05
## 319	65621	4	NaN	0 2.230405e-04
## 320	75933	13	NaN	0 3.903447e-03
## 321	98489	2	NaN	0 3.521575e-03
## 322	60203	12	NaN	0 1.280462e-03
## 323	67829	18	NaN	0 9.074478e-04
## 324	60851	6	NaN	0 3.516601e-01
## 325	74052	16	NaN	0 7.864214e-01
## 326	98826	10	NaN	0 6.989840e-02
## 327	65934	1	NaN	0 2.684046e-17
## 328	97338	1	NaN	0 2.062738e-17
## 329	99489	1	NaN	0 1.957038e-17
## 330	95299	1	NaN	0 2.194952e-17
## 331	96697	1	NaN	0 2.079526e-17
## 332	69099	1	NaN	0 2.725266e-05
## 333	98852	1	NaN	0 3.601754e-17
## 334	90995	2	NaN	0 1.737549e-03
## 335	70248	1	NaN	0 2.439549e-17
## 336	77651	1	NaN	0 2.621432e-17
## 337	77184	2	NaN	0 9.103444e-03
## 338	97312	1	NaN	0 3.762031e-17
## 339	72666	1	NaN	0 2.611942e-17
## 340	74048	1	NaN	0 2.645324e-17
## 341	62778	1	NaN	0 3.552328e-17
## 342	78856	1	NaN	0 5.152493e-18
## 343	67890	1	NaN	0 7.143847e-18
## 344	61408	1	NaN	0 6.310605e-18
## 345	99787	1	NaN	0 6.960912e-18
## 346	86422	1	NaN	0 3.914374e-18

```

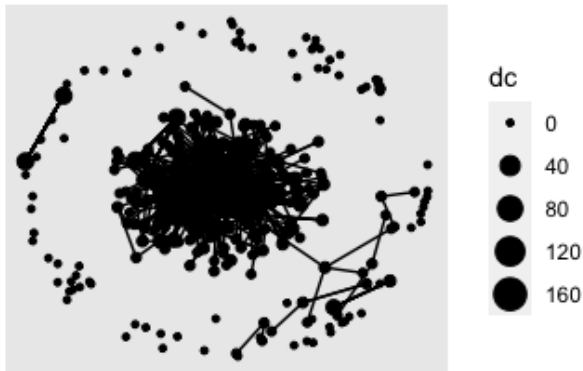
## 347 70788      1      NaN      0 5.802374e-18
## 348 77228      1      NaN      0 9.397387e-04
## 349 69511      1      NaN      0 3.219770e-17
## 350 66182      1      NaN      0 2.820059e-17
## 351 71397      2      NaN      0 4.835429e-04
## 352 95065      1      NaN      0 8.163706e-03
## 353 81984      4      NaN      0 1.187126e-01
## 354 60437      1      NaN      0 2.483594e-17
## 355 80181      1      NaN      0 3.896146e-17
## 356 67698      1      NaN      0 2.724826e-17
## 357 95160      1      NaN      0 4.111361e-06
## 358 94771      1      NaN      0 5.861002e-02
## 359 98738      1      NaN      0 2.178643e-17
## 360 63372      2      NaN      0 6.262549e-04
## 361 95432      1      NaN      0 0.000000e+00
## 362 59269      1      NaN      0 0.000000e+00
## 363 63482      1      NaN      0 0.000000e+00
## 364 94860      1      NaN      0 0.000000e+00
## 365 89466      1      NaN      0 0.000000e+00
## 366 95666      1      NaN      0 0.000000e+00
## 367 89404      1      NaN      0 0.000000e+00
## 368 73564      1      NaN      0 0.000000e+00
## 369 64184      1      NaN      0 0.000000e+00
## 370 67547      1      NaN      0 0.000000e+00
## 371 67514      1      NaN      0 0.000000e+00
## 372 97267      1      NaN      0 0.000000e+00
## 373 64350      1      NaN      0 0.000000e+00
## 374 99577      1      NaN      0 0.000000e+00
## 375 91123      1      NaN      0 0.000000e+00
## 376 88204      1      NaN      0 3.959135e-17
## 377 97922      1      NaN      0 7.108826e-18
## 378 94768      1      NaN      0 8.995829e-18
## 379 81287      1      NaN      0 7.257309e-18
## 380 66203      1      NaN      0 6.032715e-18
## 381 82412      1      NaN      0 6.353455e-18
## 382 62098      1      NaN      0 9.621650e-18

```

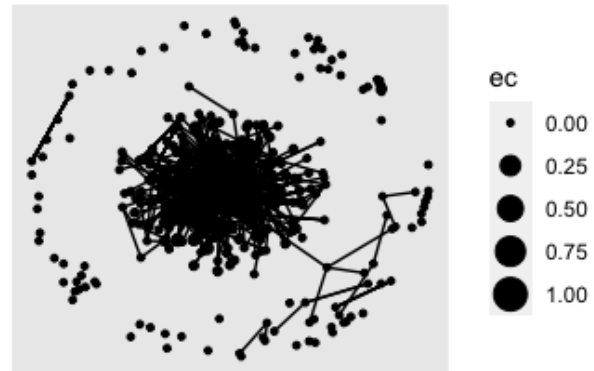
```
ggarrange(dc_164,ec_164,cc_164,bc_164,ncol = 2, nrow = 2)
```

```
## Warning: Removed 293 rows containing missing values (geom_point).
```

Degree Centrality 164



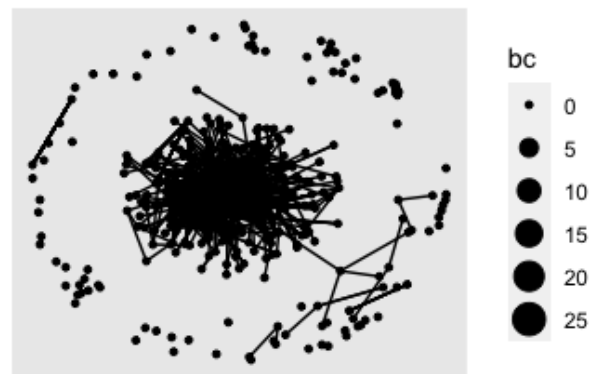
Eigenvector Centrality 164



Closeness Centrality 164



Betweenness Centrality 164



```
dc_172 = ggraph(net_172, layout="kk") +
  geom_edge_link()+
  geom_node_point(aes(size=dc), show.legend=T) + ggtitle("Degree Centrality 172")

# Eigenvector Centrality
ec_172<-ggraph(net_172, layout="kk") +
  geom_edge_link()+
  geom_node_point(aes(size=ec), show.legend=T) + ggtitle("Eigenvector Centrality 172")

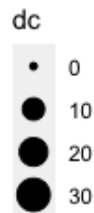
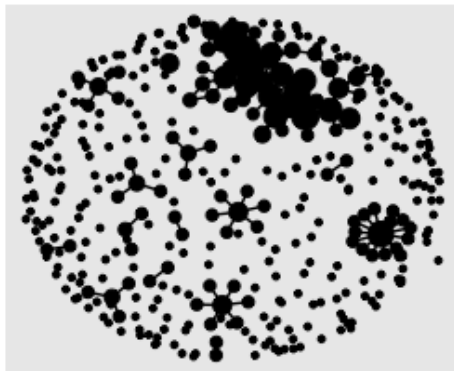
# Closness Centrality
cc_172<-ggraph(net_172, layout="kk") +
  geom_edge_link()+
  geom_node_point(aes(size=cc), show.legend=T) + ggtitle("Closeness Centrality 172")

# Betweenness Centrality
bc_172<-ggraph(net_172, layout="kk") +
  geom_edge_link()+
  geom_node_point(aes(size=bc), show.legend=T) + ggtitle("Betweenness Centrality 172")
```

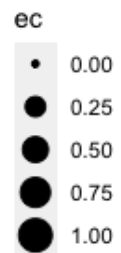
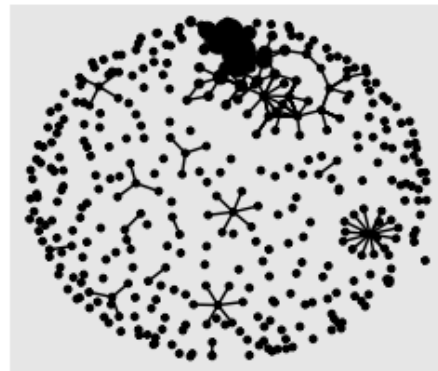
```
ggarrange(dc_172,ec_172,cc_172,bc_172,ncol = 2, nrow = 2)
```

```
## Warning: Removed 345 rows containing missing values (geom_point).
```

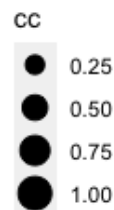
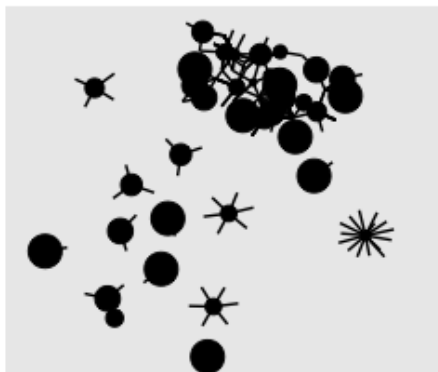

Degree Centrality 172



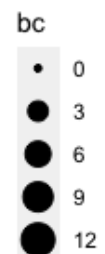
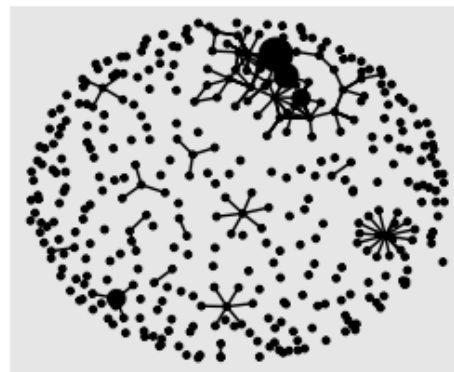
Eigenvector Centrality 172



Closeness Centrality 172



Betweenness Centrality 172



Based on the graph, seems like closeness centrality has clearer cluster center.

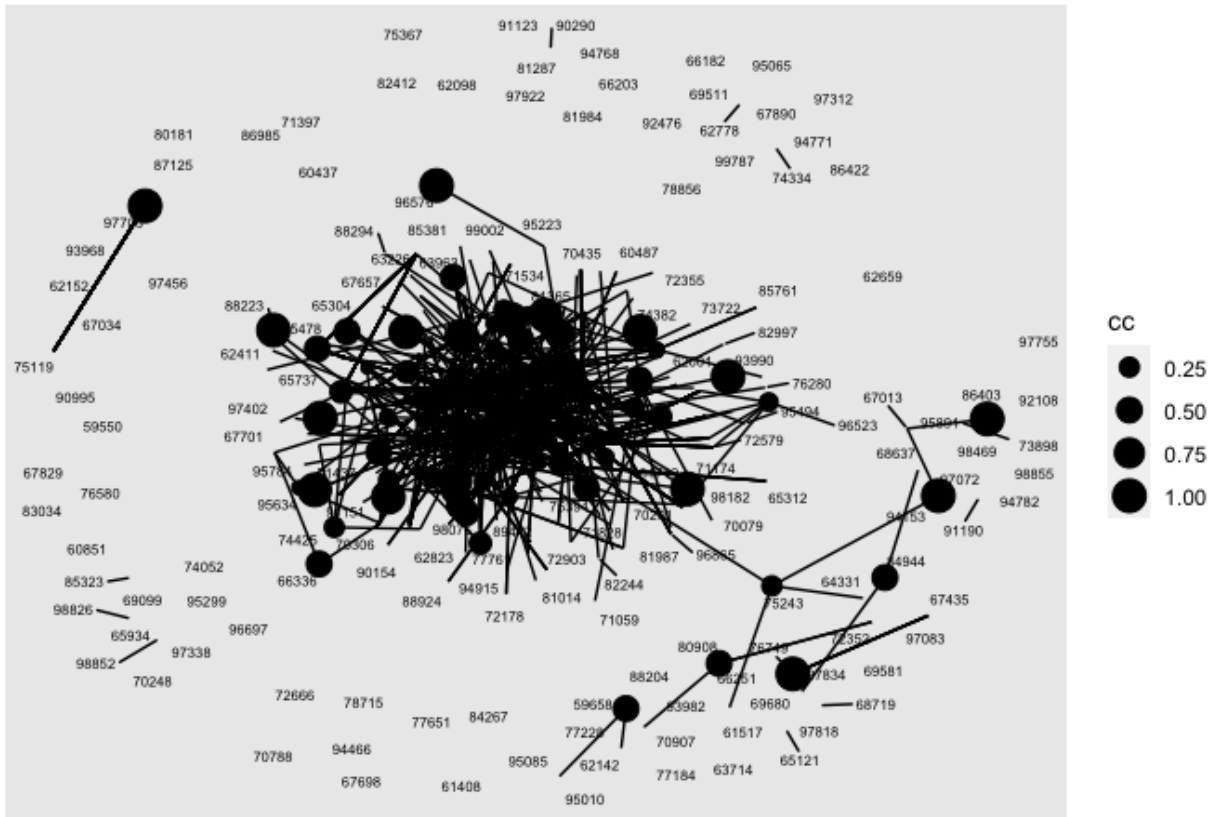
Characterize and discuss the relationship between centrality and other examiners' characteristics

```
ggraph(net_164, layout="kk") +
  geom_edge_link()+
  geom_node_point(aes(size=cc), show.legend=T) +geom_node_text(aes(label = centrality_164$id), repel=TRUE)

## Warning: Removed 293 rows containing missing values (geom_point).

## Warning: ggrepel: 226 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```

Closeness Centrality 164

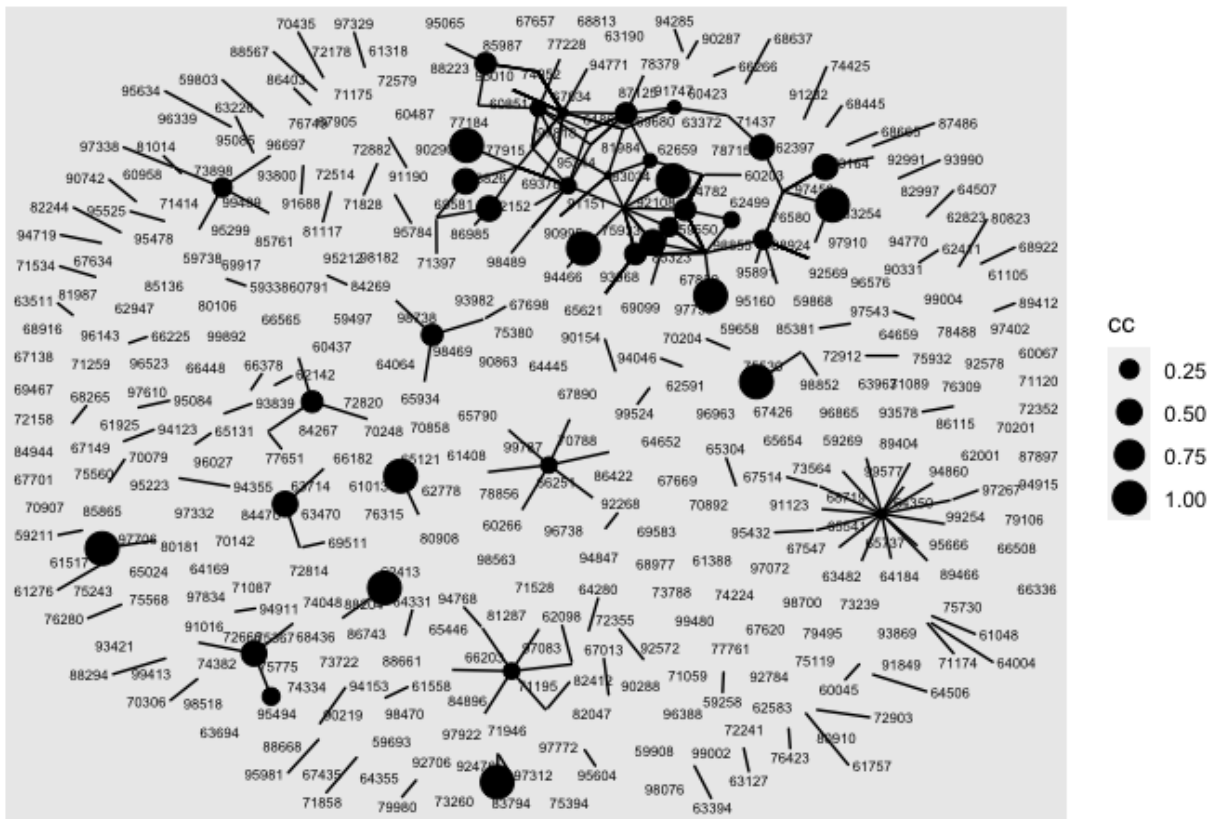


```
ggraph(net_172, layout="kk") +
  geom_edge_link()+
  geom_node_point(aes(size=cc), show.legend=T) +geom_node_text(aes(label = centrality_172$id), repel=TRUE)
```

```
## Warning: Removed 345 rows containing missing values (geom_point).
```

```
## Warning: ggrepel: 7 unlabeled data points (too many overlaps). Consider
## increasing max.overlaps
```

Closeness Centrality 172



Gather all examiner characteristics

```
examiner = df %>% select(examiner_id,examiner_art_unit,gender,race,tenure)
examiner = distinct(examiner)
```

Examiner that are in group 164 and has the highest closeness centrality

```
max_cc_164 = max(centrality_164$closeness[!is.na(centrality_164$closeness)])
max_cc_164_id = centrality_164 %>% filter(centrality_164$closeness ==max_cc_164) %>%select(id)
max_cc_164_id = max_cc_164_id %>% mutate(id = as.numeric(id))
max_cc_164_info = examiner %>%filter(examiner_id == max_cc_164_id$id)

table(max_cc_164_info$gender)
```

```
##
## female    male
##      252    323
```

```
table(max_cc_164_info$race)
```

```
##
##   asian    black hispanic    white
##     163      43         2     367
```

Examiners that has higher closeness centrality in group 164, are more likely to be while male.

Examiner that are in group 172 and has the highest closeness centrality

```
max_cc_172 = max(centrality_172$closeness[!is.na(centrality_172$closeness)])
max_cc_172_id = centrality_172 %>% filter(centrality_172$closeness ==max_cc_172) %>%select(id)
max_cc_172_id = max_cc_172_id %>% mutate(id = as.numeric(id))
max_cc_172_info = examiner %>%filter(examiner_id == max_cc_172_id$id)

table(max_cc_172_info$gender)
```

```
##
## female    male
##      79     366
```

```
table(max_cc_172_info$race)
```

```
##
##   asian hispanic   white
##    63      43     339
```

The examiners that has higher closeness centrality in group 172 are mostly male comparing to group 164. Also, there are more Hispanic examiners that are influential in this group.