HW3

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```
library(sna)
library(network)
library(networkdata)
library(tidyverse)
library(ergm)
data(coldwar)
data(hansell)
```

Question 1

```
coldwarcc <- coldwar$cc[,,1]
coldwarcc[coldwarcc < 0] <- 0
coldwarcc[is.na(coldwarcc)] <- 0

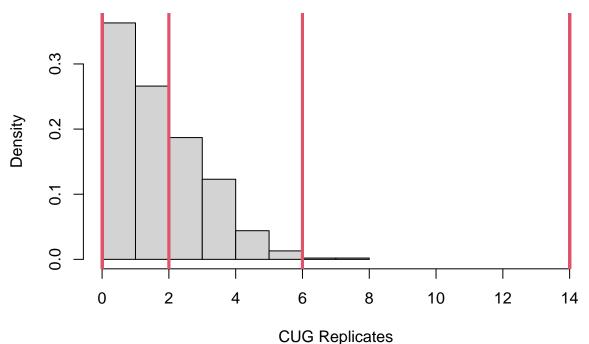
network_cw <- network(coldwarcc, directed = FALSE)
degree_cw <- sna::centralization(dat = network_cw, FUN = sna::degree)
degree_cw</pre>
```

[1] 0.09423077

since the value of degree centralization is 0.0942308, the null hypothesis cannot be rejected, which means it is a simple random graph model.

```
cug_test <- cug.test(dat = network_cw, FUN = degree, cmode = "edges", reps = 1000)
plot(cug_test)</pre>
```

Univariate CUG Test

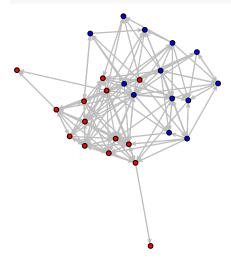


Conditioning: edges Reps: 1000

Question 2

 \mathbf{a}

```
plot(hansell,
    vertex.col = ifelse(get.vertex.attribute(hansell, "sex") == "male", "blue", "red"),
    edge.col = "grey")
```



based on the plot, there is a general preference for same-sex friendship ties, it is easy to see that blue nodes are predominantly connected the other blue nodes, same for red nodes which represents girls.

```
b)
fit.er <- ergm(hansell ~ edges)</pre>
## Starting maximum pseudolikelihood estimation (MPLE):
## Obtaining the responsible dyads.
## Evaluating the predictor and response matrix.
## Maximizing the pseudolikelihood.
## Finished MPLE.
## Evaluating log-likelihood at the estimate.
summary(fit.er)
## Call:
## ergm(formula = hansell ~ edges)
## Maximum Likelihood Results:
##
        Estimate Std. Error MCMC % z value Pr(>|z|)
##
## edges -1.24454 0.09058
                             0 -13.74 <1e-04 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
       Null Deviance: 973.2 on 702 degrees of freedom
## Residual Deviance: 746.2 on 701 degrees of freedom
## AIC: 748.2 BIC: 752.8 (Smaller is better. MC Std. Err. = 0)
log_odds_b <- -1.24454
cat("the log-odds of a tie between two randomly chosen students is", log_odds_b)
## the log-odds of a tie between two randomly chosen students is -1.24454
probability_b <- exp(log_odds_b) / (1 + exp(log_odds_b))</pre>
cat("the MLE of the probability of a tie between two randomly chosen students is",
   probability_b)
## the MLE of the probability of a tie between two randomly chosen students is 0.2236467
hansell
## Network attributes:
##
    vertices = 27
    directed = TRUE
##
    hyper = FALSE
##
##
    loops = FALSE
##
    multiple = FALSE
##
    total edges= 157
##
      missing edges= 0
##
      non-missing edges= 157
##
##
  Vertex attribute names:
##
       sex
##
## No edge attributes
```

```
vertices <- 27
tot_edges <- 157
density_b <- tot_edges / (vertices * (vertices - 1) / 2)
cat("the density of the network is", density_b)</pre>
```

the density of the network is 0.4472934

```
# H_0: The null hypothesis is that all networks among the students are equally likely
# H_1: The alternative hypothesis is that the Erdos-Renyi model fits the data better

Null_Deviance <- 973.2
Residual_Deviance <- 746.2
p_value_b <- 1 - pchisq(Null_Deviance - Residual_Deviance, df = 1)
p_value_b</pre>
```

[1] 0

the MLE probability under the Erdos-Renyi model is lower than the actual observed density of ties in the network (MLE = 0.2236467, density = 0.4472934) and the p-value is 0, suggesting that the Erdos-Renyi model might underestimate the number of ties in this specific network.

```
c)
fit.homo <- ergm(hansell~edges + nodematch("sex"))</pre>
## Starting maximum pseudolikelihood estimation (MPLE):
## Obtaining the responsible dyads.
## Evaluating the predictor and response matrix.
## Maximizing the pseudolikelihood.
## Finished MPLE.
## Evaluating log-likelihood at the estimate.
summary(fit.homo)
## Call:
## ergm(formula = hansell ~ edges + nodematch("sex"))
## Maximum Likelihood Results:
##
##
                Estimate Std. Error MCMC % z value Pr(>|z|)
## edges
                 -1.9841
                          0.1608 0 -12.341
## nodematch.sex 1.2954
                             0.1979
                                           6.547
                                         0
                                                     <1e-04 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
##
       Null Deviance: 973.2 on 702 degrees of freedom
##
   Residual Deviance: 699.1 on 700 degrees of freedom
##
## AIC: 703.1 BIC: 712.2 (Smaller is better. MC Std. Err. = 0)
```

sicne p-value for nodematch.sex is <1e-04, which is highly significant result that indicates strong evidence of homophily based on sex

[1] 0

The p-value is extremely small (p-value = 0), indicating that the model with homophily based on sex fits significantly better than the Erdos-Renyi model.

d) fit.diff.homo <- ergm(hansell~edges + nodematch("sex", diff=TRUE))</pre> ## Starting maximum pseudolikelihood estimation (MPLE): ## Obtaining the responsible dyads. ## Evaluating the predictor and response matrix. ## Maximizing the pseudolikelihood. ## Finished MPLE. ## Evaluating log-likelihood at the estimate. summary(fit.diff.homo) ## Call: ## ergm(formula = hansell ~ edges + nodematch("sex", diff = TRUE)) ## Maximum Likelihood Results: ## ## Estimate Std. Error MCMC % z value Pr(>|z|) ## edges -1.9841 0.1608 0 -12.341 0.2221 0 6.607 ## nodematch.sex.female 1.4674 <1e-04 *** ## nodematch.sex.male 1.0813 0.2389 0 4.526 <1e-04 *** ## ---## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1 ## Null Deviance: 973.2 on 702 degrees of freedom ## ## Residual Deviance: 696.4 on 699 degrees of freedom ## AIC: 702.4 BIC: 716 (Smaller is better. MC Std. Err. = 0) beta_female <- 1.4674

beta_male <- 1.0813 se_female <- 0.2221 se_male <- 0.2389

```
e)
fit.mix <- ergm(hansell~nodemix("sex"))</pre>
## Starting maximum pseudolikelihood estimation (MPLE):
## Obtaining the responsible dyads.
## Evaluating the predictor and response matrix.
## Maximizing the pseudolikelihood.
## Finished MPLE.
## Evaluating log-likelihood at the estimate.
summary(fit.mix)
## Call:
## ergm(formula = hansell ~ nodemix("sex"))
## Maximum Likelihood Results:
##
##
                      Estimate Std. Error MCMC \% z value Pr(>|z|)
## mix.sex.male.female -1.8374 0.2153
                                          0 -8.533 <1e-04 ***
                                   0.2424
                                               0 -8.867 <1e-04 ***
## mix.sex.female.male -2.1493
## mix.sex.male.male
                       -0.9029
                                   0.1767
                                               0 -5.109 <1e-04 ***
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
       Null Deviance: 973.2 on 702 degrees of freedom
##
## Residual Deviance: 707.2 on 699 degrees of freedom
## AIC: 713.2 BIC: 726.9 (Smaller is better. MC Std. Err. = 0)
```

```
log_odds_e <- -1.8374
cat("the log-odds of a tie from a male to a female student is", log_odds_e)

## the log-odds of a tie from a male to a female student is -1.8374

probability_e <- exp(-2.1493) / (1 + exp(-2.1493))
cat("the MLE of the probability of a tie from a female to a male student is",
    probability_e)

## the MLE of the probability of a tie from a female to a male student is 0.1043967

sex_attr <- get.vertex.attribute(hansell, "sex")
edge_list <- as.matrix.network.edgelist(hansell)

female_to_male_count <- 0
for (i in 1:nrow(edge_list)) {
    from <- edge_list[i, 1]
    to <- edge_list[i, 2]

    if (sex_attr[from] == "female" && sex_attr[to] == "male") {
        female_to_male_count <- female_to_male_count + 1</pre>
```

[1] 0.1043956

female_to_male_count / (13 * 14)

they are almost the same, the difference between them is $-1.0491962 \times 10^{-6}$ (empirical frequency - MLE of the probability of a tie from a female to a male student)

```
Null_Deviance <- 973.2
Residual_Deviance <- 707.2
p_value_e <- 1 - pchisq(Null_Deviance - Residual_Deviance, df = 2)
p_value_e</pre>
```

[1] 0

}

the p-value is very small, indicating that the full mixing model fits significantly better than the Erdos-Renyi model.