Simple ERGM

Charles Costanzo

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```
# install "intergraph" package (only need to do once, so comment out)
# install.packages("intergraph")
# create a vector containing names of packages we want to load
packages <- c("intergraph", "igraph", "statnet", "ergm")</pre>
# load in packages
lapply(packages, library, character.only = TRUE)
# read in igraph object "g"
path <- "/Users/charlescostanzo/College/Au 2023/Politsc 4998/Data/graphml_zscore_self_1_1_5_85_164/114...
g <- read_graph(path, format = "graphml")</pre>
# print vertex attribute names
names(vertex_attr(g))
## [1] "nokken_poole_dim2"
                                          "nokken_poole_dim1"
## [3] "conditional"
                                          "nominate_number_of_errors"
## [5] "nominate_number_of_votes"
                                          "nominate_geo_mean_probability"
## [7] "nominate_log_likelihood"
                                          "nominate_dim2"
## [9] "nominate_dim1"
                                          "died"
## [11] "born"
                                          "bioguide_id"
## [13] "bioname"
                                          "last_means"
## [15] "occupancy"
                                          "party_code"
## [17] "state_abbrev"
                                          "district_code"
## [19] "state_icpsr"
                                          "icpsr"
## [21] "chamber"
                                          "congress"
## [23] "id"
# set born attribute to numeric type
V(g)$born <- as.numeric(V(g)$born)</pre>
V(g)$died <- ifelse(V(g)$died == "NaN", NA, as.numeric(V(g)$died))
# convert igraph object "g" to network named "net"
net <- asNetwork(g)</pre>
# extract edgelist from "g"
el.g <- get.edgelist(g)</pre>
# extract edgelist from "net"
el.net <- as.matrix(net, "edgelist")</pre>
```

```
# check if edge lists are identical
identical(as.numeric(el.g), as.numeric(el.net))
```

[1] TRUE

```
# set random seed for reproducibility
set.seed(1022)
# run very simple ergm
model <- ergm(net ~ edges +</pre>
                ostar(2:3) +
                nodemix("party_code"),
              control = control.ergm(MCMC.samplesize=5000,
                                     MCMC.burnin=5000,
                                     MCMLE.maxit=10)
## Starting maximum pseudolikelihood estimation (MPLE):
## Obtaining the responsible dyads.
## Evaluating the predictor and response matrix.
## Maximizing the pseudolikelihood.
## Finished MPLE.
## Starting Monte Carlo maximum likelihood estimation (MCMLE):
## Iteration 1 of at most 10:
## Warning: 'glpk' selected as the solver, but package 'Rglpk' is not available;
## falling back to 'lpSolveAPI'. This should be fine unless the sample size and/or
## the number of parameters is very big.
## Optimizing with step length 1.0000.
## The log-likelihood improved by 0.0262.
## Convergence test p-value: 0.1779. Not converged with 99% confidence; increasing sample size.
## Iteration 2 of at most 10:
## Optimizing with step length 1.0000.
## The log-likelihood improved by 0.0963.
## Convergence test p-value: 0.1492. Not converged with 99% confidence; increasing sample size.
## Iteration 3 of at most 10:
## Optimizing with step length 1.0000.
## The log-likelihood improved by 0.0622.
## Convergence test p-value: 0.1028. Not converged with 99% confidence; increasing sample size.
## Iteration 4 of at most 10:
## Optimizing with step length 1.0000.
## The log-likelihood improved by 0.0383.
## Convergence test p-value: 0.0099. Converged with 99% confidence.
## Finished MCMLE.
## Evaluating log-likelihood at the estimate. Fitting the dyad-independent submodel...
## Bridging between the dyad-independent submodel and the full model...
## Setting up bridge sampling...
```

```
## Using 16 bridges: 1 2 3 4 5 6 7 8 9 10 11 12 13 14 15 16 .
## Bridging finished.
## This model was fit using MCMC. To examine model diagnostics and check
## for degeneracy, use the mcmc.diagnostics() function.
# display model summary
summary(model)
## Call:
## ergm(formula = net ~ edges + ostar(2:3) + nodemix("party_code"),
      control = control.ergm(MCMC.samplesize = 5000, MCMC.burnin = 5000,
          MCMLE.maxit = 10))
##
## Monte Carlo Maximum Likelihood Results:
##
##
                           Estimate Std. Error MCMC \% z value Pr(>|z|)
## edges
                         -1.670e+00 6.359e-01
                                                0 -2.626 0.00863 **
## ostar2
                         -2.873e-02 5.099e-02
                                                    0 -0.563 0.57317
                          8.308e-05 2.098e-03
## ostar3
                                                    0 0.040 0.96841
## mix.party_code.200.100 -5.637e-01 3.289e-02
                                                   0 -17.140 < 1e-04 ***
                                                    0 -29.587 < 1e-04 ***
## mix.party_code.100.200 -9.724e-01 3.286e-02
## mix.party_code.200.200 -3.514e-01 3.039e-02
                                                    0 -11.564 < 1e-04 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
##
       Null Deviance: 261719 on 188790 degrees of freedom
## Residual Deviance: 81854 on 188784 degrees of freedom
##
```

AIC: 81866 BIC: 81927 (Smaller is better. MC Std. Err. = 1.138)