

Committee-based Congressional Network Regression Analysis

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```
library(readr)
df <- read_csv("final_analysis.csv")
```

```
## Rows: 481 Columns: 11
## -- Column specification -----
## Delimiter: ","
## chr (6): Legislator name, Connections, Gender Estimates, Predicted Gender, E...
## dbl (5): Legislative Effectiveness Score, bc, ec, cc, dc
##
## i Use 'spec()' to retrieve the full column specification for this data.
## i Specify the column types or set 'show_col_types = FALSE' to quiet this message.
```

```
df$factor <- as.factor(sample(1:2, nrow(df), replace = TRUE))
head(df)
```

```
## # A tibble: 6 x 12
##   'Legislator name' Connections 'Legislative E~'      bc      ec      cc      dc
##   <chr>             <chr>             <dbl> <dbl> <dbl> <dbl> <dbl>
## 1 Jack Reed        ['Kirsten ~        0.346 0.00178 -2.74e-18 0      0.117
## 2 Kirsten Gillibra~ ['Jack Ree~        0.339 0.00178 -3.85e-18 0.00181 0.108
## 3 Tammy Duckworth  ['Jack Ree~        1.53  0.00178  1.94e-18 0.00361 0.116
## 4 Gary Peters      ['Jack Ree~        5.02  0.00225  1.74e-17 0.00722 0.119
## 5 Elizabeth Warren ['Jack Ree~        0.725 0.00180  2.01e-17 0.00903 0.112
## 6 Martin Heinrich  ['Jack Ree~        0.781 0.00239  1.96e-17 0.0108  0.125
## # ... with 5 more variables: 'Gender Estimates' <chr>,
## #   'Predicted Gender' <chr>, 'Ethnicity Estimates' <chr>,
## #   'Predicted Ethnicity' <chr>, factor <fct>
```

```
library(glmnet)
```

```
## Loading required package: Matrix
## Loaded glmnet 4.1-3
```

```
# the original data frame and formula
```

```
set.seed(23)
```

```
f <- as.formula(df$`Legislative Effectiveness Score` ~ (df$bc+ df$ec+ df$cc+ df$dc + df$`Predicted Gender`))
```

```
# transform dataframe to matrices as required by glmnet
```

```
x <- model.matrix(f, df)
```

```
y <- as.matrix(df$`Legislative Effectiveness Score`, ncol=1)
```

```

# fit ridge regression model with a wide range of penalty parameter
lambdas <- 10^seq(3, -2, by = -.1)
cv_fit <- cv.glmnet(x,y, lambda=lambdas, alpha = 0)
fit <- cv_fit$glmnet.fit
summary(fit)

```

```

##           Length Class      Mode
## a0           51    -none-   numeric
## beta        1734 dgCMatrix S4
## df           51    -none-   numeric
## dim           2    -none-   numeric
## lambda        51    -none-   numeric
## dev.ratio     51    -none-   numeric
## nulldev        1    -none-   numeric
## npasses        1    -none-   numeric
## jerr           1    -none-   numeric
## offset         1    -none-   logical
## call           5    -none-   call
## nobs           1    -none-   numeric

```

```

# get BIC value
tLL <- fit$nulldev - deviance(fit)
k <- fit$df
n <- fit$nobs

BIC<-log(n)*k - tLL
BIC

```

```

## [1] 203.7148 203.6922 203.6636 203.6279 203.5832 203.5275 203.4580 203.3717
## [9] 203.2649 203.1332 202.9715 202.7790 202.5427 202.2597 201.9245 201.5326
## [17] 201.0809 200.5688 199.9986 199.3758 198.7085 198.0079 197.2856 196.5534
## [25] 195.8211 195.0955 194.3794 193.6714 192.9683 192.2578 191.5308 190.7818
## [33] 189.9976 189.1778 188.3069 187.3999 186.4561 185.4871 184.5072 183.5323
## [41] 182.5796 181.6579 180.7752 179.9356 179.1394 178.3807 177.6596 176.9605
## [49] 176.2710 175.6056 174.9604

```

```

opt_lambda = lambdas[which(BIC == min(BIC))]
# Choose best model
best_model <- glmnet(x,y, lambda=opt_lambda, alpha = 0)
print(coef(best_model))

```

```

## 35 x 1 sparse Matrix of class "dgCMatrix"
##
## (Intercept) 1.497341e+00
## (Intercept) .
## df$bc -1.130367e+02
## df$ec -2.108612e-01
## df$cc 4.343528e-01
## df$dc -2.443166e+00
## df$'Predicted Gender'male 4.855377e-01
## df$'Predicted Ethnicity'black 1.842198e-01

```

## df\$'Predicted Ethnicity'hispanic	5.753947e-01
## df\$'Predicted Ethnicity'white	-2.049838e-01
## df\$bc:df\$ec	-3.244254e+03
## df\$bc:df\$cc	3.456645e+02
## df\$bc:df\$dc	3.972438e+02
## df\$bc:df\$'Predicted Gender'male	-2.179875e+01
## df\$bc:df\$'Predicted Ethnicity'black	8.332574e+02
## df\$bc:df\$'Predicted Ethnicity'hispanic	-1.106393e+02
## df\$bc:df\$'Predicted Ethnicity'white	3.791436e+01
## df\$ec:df\$cc	-1.353747e-01
## df\$ec:df\$dc	4.584725e+01
## df\$ec:df\$'Predicted Gender'male	2.392189e+00
## df\$ec:df\$'Predicted Ethnicity'black	4.392340e+01
## df\$ec:df\$'Predicted Ethnicity'hispanic	-4.033550e+01
## df\$ec:df\$'Predicted Ethnicity'white	4.196871e-01
## df\$cc:df\$dc	2.115944e+00
## df\$cc:df\$'Predicted Gender'male	-1.231721e+00
## df\$cc:df\$'Predicted Ethnicity'black	-2.429899e+00
## df\$cc:df\$'Predicted Ethnicity'hispanic	-9.938116e-01
## df\$cc:df\$'Predicted Ethnicity'white	-1.672829e+00
## df\$dc:df\$'Predicted Gender'male	-1.118700e+00
## df\$dc:df\$'Predicted Ethnicity'black	-2.158242e+01
## df\$dc:df\$'Predicted Ethnicity'hispanic	-4.427393e-01
## df\$dc:df\$'Predicted Ethnicity'white	1.293783e-02
## df\$'Predicted Gender'male:df\$'Predicted Ethnicity'black	5.450884e-01
## df\$'Predicted Gender'male:df\$'Predicted Ethnicity'hispanic	-1.401436e-01
## df\$'Predicted Gender'male:df\$'Predicted Ethnicity'white	4.131083e-02