Areal Data Model Fitting, Part 2

Continuing with the Tester election dataset.

Now consider some covariates to explain the response

Consider a linear model with county population, using both 1m and S.glm

```
lm_dat <- Tester %>%
 left_join(usmap::countypop %>% rename("county_fips" = fips), by = 'county_fips') %>%
 mutate(scale_pop = scale(pop_2015))
pop_model <- lm(Tester_Prop ~ scale_pop, data = lm_dat)</pre>
display(pop_model)
## lm(formula = Tester_Prop ~ scale_pop, data = lm_dat)
               coef.est coef.se
## (Intercept) 0.40
                        0.02
                        0.02
## scale_pop
               0.06
## ---
## n = 56, k = 2
## residual sd = 0.13, R-Squared = 0.18
S.glm(Tester_Prop ~ scale_pop, data = lm_dat, family = 'gaussian',
      burnin = 100, n.sample = 1000, verbose = F)
##
## ###############
## #### Model fitted
## ################
## Likelihood model - Gaussian (identity link function)
## Random effects model - None
## Regression equation - Tester_Prop ~ scale_pop
## Number of missing observations - 0
##
## ###########
## #### Results
## ###########
## Posterior quantities and DIC
##
##
                        2.5% 97.5% n.effective Geweke.diag
               Median
## (Intercept) 0.3982 0.3650 0.4348
                                            900
                                                        -1.5
                                           1053
## scale_pop
               0.0621 0.0241 0.0967
                                                        1.4
## nu2
               0.0182 0.0125 0.0262
                                            900
                                                        -1.1
##
## DIC = -62.58489
                         p.d = 2.901435
                                                LMPL = 30.64
```

We previously extract the residuals create a choropleth and test for spatial association. Now we will directly run the CAR model using CARBayes. See vignette for more info and options on model fitting, including GLM models.

This model can be expressed as

$$Y_k \sim N(\mu_k, \nu^2)$$
, where $\mu_k = x_k \beta + \psi_k$.

A CAR prior is placed on ψ_k such that the full conditional is expressed as

$$|\psi_k| - \sim N\left(\frac{\rho \sum_i w_{ki} \psi_i}{\rho \sum_i w_{ki} + 1 - \rho}, \frac{\tau^2}{\rho \sum_i w_{ki} + 1 - \rho}\right)$$

If $\rho = 0$ there is no spatial structure present, if $\rho \to 1$ this is the intrinsic CAR model.

```
##
## ################
## #### Model fitted
## ################
## Likelihood model - Gaussian (identity link function)
## Random effects model - Leroux CAR
## Regression equation - Tester_Prop ~ scale_pop
## Number of missing observations - 0
##
## ###########
## #### Results
## ###########
## Posterior quantities and DIC
##
##
               Median
                        2.5% 97.5% n.effective Geweke.diag
## (Intercept) 0.3977 0.3698 0.4258
                                         18000.0
                                                         0.9
               0.0538 0.0208 0.0881
                                                         0.0
## scale_pop
                                         15840.8
               0.0108 0.0034 0.0201
                                         3285.2
                                                         2.1
## nu2
               0.0123 0.0029 0.0373
## tau2
                                         3096.0
                                                        -1.6
               0.5307 0.0646 0.9454
                                         4799.9
## rho
                                                         1.9
##
                          p.d = 17.83747
## DIC = -81.44965
                                                 LMPL = 34.33
```