

COVID-19-Spain-Analysis

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4/11/2020

This notebook reports our initial analysis of COVID-19 incidence in Spain and the climatic correlates of incidence. The data have been organized in a package for ease of access and distribution. The name of the package is `covid19env` and if necessary can be installed from the GitHub repository.

Preliminaries

Load packages:

```
library(covid19env)
library(ggthemes)
library(gridExtra)
library(lubridate)
library(sf)
library(spdep)
library(spsur)
library(tidyverse)
#library(spatialreg)
#library(systemfit)
#library(plm)
#library(splm)
```

Load data from package `covid19env`

```
data("covid19_spain")
```

Summarize the data:

```
covid19_spain %>%
  select(-geometry) %>%
  summary()
```

##	province	CCAA	ID_INE
##	Albacete : 22	Castilla y Leon :198	Min. : 1.0
##	Alicante/Alacant: 22	Andalucia :176	1st Qu.:13.0
##	Almeria : 22	Castilla - La Mancha:110	Median :25.5
##	Araba/alava : 22	Cataluña : 88	Mean :25.5
##	Asturias : 22	Galicia : 88	3rd Qu.:38.0
##	Avila : 22	Aragon : 66	Max. :50.0
##	(Other) :968	(Other) :374	
##	Date	Cases	Incidence
##	Min. :2020-03-14	Min. : 1.0	Min. : 0.4536
##	1st Qu.:2020-03-19	1st Qu.: 126.0	1st Qu.: 20.0996
##	Median :2020-03-24	Median : 378.5	Median : 62.1017
##	Mean :2020-03-24	Mean : 1073.2	Mean :109.3028
##	3rd Qu.:2020-03-30	3rd Qu.: 957.2	3rd Qu.:148.8700
##			Population
##	Min. : 88636		
##	1st Qu.: 331549		
##	Median : 684202		
##	Mean : 974257		
##	3rd Qu.:1149460		

```

## Max. :2020-04-04 Max. :36249.0 Max. :867.5933 Max. :6663394
##
## Older Median_Age Male2Female Area
## Min. :15.16 Min. :40.19 Min. : 91.59 Min. :1.979e+09
## 1st Qu.:18.02 1st Qu.:42.35 1st Qu.: 95.43 1st Qu.:6.637e+09
## Median :19.93 Median :43.70 Median : 98.06 Median :1.001e+10
## Mean :21.03 Mean :44.55 Mean : 97.83 Mean :1.012e+10
## 3rd Qu.:23.07 3rd Qu.:46.01 3rd Qu.:100.08 3rd Qu.:1.377e+10
## Max. :31.36 Max. :50.68 Max. :103.01 Max. :2.179e+10
##
## Altitude Coast Meteo_Station Max_Temp Min_Temp
## Min. : 5.0 Min. :0.00 0016A : 22 Min. : 3.10 Min. : -4.700
## 1st Qu.: 24.0 1st Qu.:0.00 0076 : 22 1st Qu.:13.80 1st Qu.: 3.300
## Median : 215.5 Median :0.00 0367 : 22 Median :16.60 Median : 6.400
## Mean : 369.0 Mean :0.42 1024E : 22 Mean :16.25 Mean : 6.293
## 3rd Qu.: 685.0 3rd Qu.:1.00 1082 : 22 3rd Qu.:19.00 3rd Qu.: 9.100
## Max. :1131.0 Max. :1.00 1111X : 22 Max. :25.50 Max. :18.100
## (Other):968
## Mean_Temp Mean_Temp_lag8 Mean_Temp_lag11 Mean_Temp_lag11w
## Min. : 1.00 Min. : 5.763 Min. : 5.364 Min. : 4.201
## 1st Qu.: 8.90 1st Qu.:10.162 1st Qu.:10.007 1st Qu.: 9.838
## Median :11.40 Median :11.994 Median :12.000 Median :11.764
## Mean :11.27 Mean :12.207 Mean :12.062 Mean :11.951
## 3rd Qu.:13.60 3rd Qu.:13.981 3rd Qu.:13.718 3rd Qu.:14.006
## Max. :21.00 Max. :19.887 Max. :19.636 Max. :19.402
##
## Sunshine_Hours Sunshine_Hours_lag8 Sunshine_Hours_lag11
## Min. : 0.000 Min. : 0.7125 Min. : 1.582
## 1st Qu.: 2.275 1st Qu.: 4.9594 1st Qu.: 5.145
## Median : 6.350 Median : 6.4500 Median : 6.305
## Mean : 5.972 Mean : 6.4370 Mean : 6.317
## 3rd Qu.: 9.500 3rd Qu.: 8.0906 3rd Qu.: 7.623
## Max. :12.400 Max. :10.9375 Max. :10.136
##
## Sunshine_Hours_lag11w Precipitation Precipitation_lag8 Precipitation_lag11
## Min. : 1.115 Min. :0.0000 Min. :0.0000 Min. :0.0000
## 1st Qu.: 4.766 1st Qu.:0.0000 1st Qu.:0.1250 1st Qu.:0.2727
## Median : 6.377 Median :0.0000 Median :0.3750 Median :0.3636
## Mean : 6.368 Mean :0.4491 Mean :0.3516 Mean :0.3682
## 3rd Qu.: 8.037 3rd Qu.:1.0000 3rd Qu.:0.5000 3rd Qu.:0.5455
## Max. :11.041 Max. :1.0000 Max. :1.0000 Max. :1.0000
##
## Precipitation_lag11w Humidity Humidity_lag8 Humidity_lag11
## Min. :0.0000 Min. :20.96 Min. :25.55 Min. :29.44
## 1st Qu.:0.1584 1st Qu.:43.00 1st Qu.:44.94 1st Qu.:45.78
## Median :0.3668 Median :53.57 Median :51.14 Median :50.85
## Mean :0.3593 Mean :53.79 Mean :52.03 Mean :52.39
## 3rd Qu.:0.5489 3rd Qu.:64.49 3rd Qu.:57.74 3rd Qu.:58.18
## Max. :1.0000 Max. :89.51 Max. :76.51 Max. :75.85
##
## Humidity_lag11w geometry
## Min. :25.39 MULTIPOLYGON :1100
## 1st Qu.:44.83 epsg:4326 : 0
## Median :51.24 +proj=long.... : 0

```

```
## Mean      :52.20
## 3rd Qu.   :59.09
## Max.      :78.74
##
```

The dataframe is a simple features object with information at the level of the province. The dataframe includes information about the province, including its Autonomous Community (a superior jurisdiction), an identifier, dates, COVID-19 cases and incidence. The period covered is from March 14, 2020 to April 4, 2020. In addition there are some demographic controls, and various climatic variables. Of interest are the lagged variables. The lagged variables are 8-day moving averages calculated using date-minus-12-days to date-minus-5-days, to account for the latency of the infection. More information about the dataset can be obtained by typing `?covid18_spain`.

There are 50 provinces in Spain:

```
nlevels(covid19_spain$province)
```

```
## [1] 50
```

Shelter in place order in Spain went into effect on March 16, 2020. March 14 is the first day that every province had at least one reported case of COVID-19.

Data visualization

There are 22 days in the dataset. We can summarize the incidence by week (excluding Canarias):

```
week11.plot <- covid19_spain %>%
  filter(CCAA != "Canarias") %>%
  group_by(province, week = isoweek(Date)) %>%
  summarise(mean_weekly_incidence = mean(Incidence)) %>%
  filter(week == 11) %>%
  ggplot() +
  geom_sf(aes(fill = mean_weekly_incidence)) +
  scale_fill_distiller(name = "Mean Weekly Incidence",
    palette = "Reds",
    direction = 1) +

  theme_tufte() +
  theme(axis.text = element_blank(),
    legend.position = "bottom") +
  facet_wrap(~week)
```

```
week12.plot <- covid19_spain %>%
  filter(CCAA != "Canarias") %>%
  group_by(province, week = isoweek(Date)) %>%
  summarise(mean_weekly_incidence = mean(Incidence)) %>%
  filter(week == 12) %>%
  ggplot() +
  geom_sf(aes(fill = mean_weekly_incidence)) +
  scale_fill_distiller(name = "Mean Weekly Incidence",
    palette = "Reds",
    direction = 1) +

  theme_tufte() +
  theme(axis.text = element_blank(),
    legend.position = "bottom") +
  facet_wrap(~week)
```

```
week13.plot <- covid19_spain %>%
```

```

filter(CCAA != "Canarias") %>%
group_by(province, week = isoweek(Date)) %>%
summarise(mean_weekly_incidence = mean(Incidence)) %>%
filter(week == 13) %>%
ggplot() +
geom_sf(aes(fill = mean_weekly_incidence)) +
scale_fill_distiller(name = "Mean Weekly Incidence",
                      palette = "Reds",
                      direction = 1) +

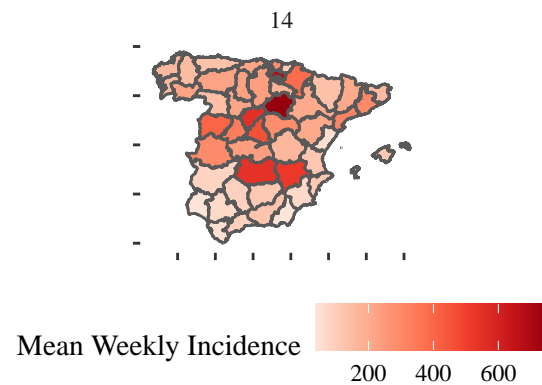
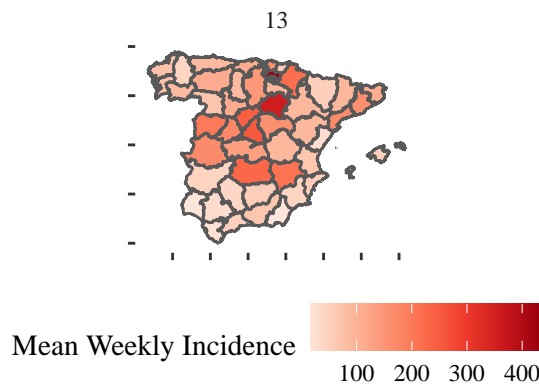
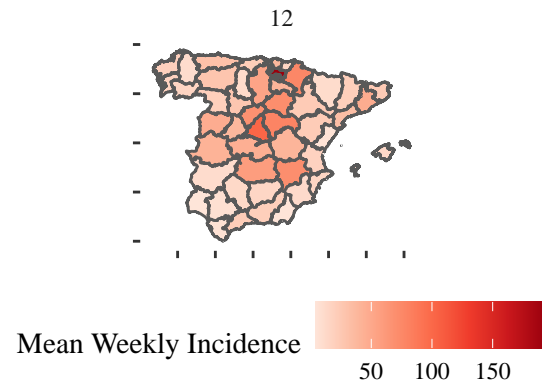
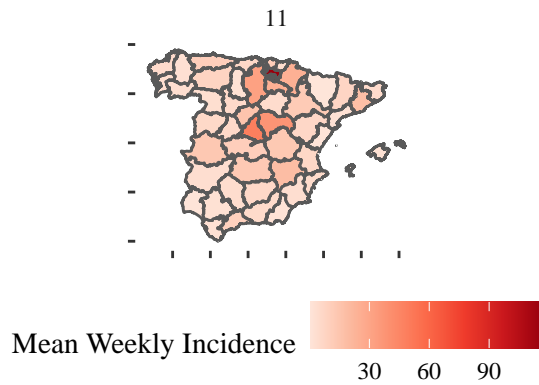
theme_tufte() +
theme(axis.text = element_blank(),
      legend.position = "bottom") +
facet_wrap(~week)

week14.plot <- covid19_spain %>%
  filter(CCAA != "Canarias") %>%
  group_by(province, week = isoweek(Date)) %>%
  summarise(mean_weekly_incidence = mean(Incidence)) %>%
  filter(week == 14) %>%
  ggplot() +
  geom_sf(aes(fill = mean_weekly_incidence)) +
  scale_fill_distiller(name = "Mean Weekly Incidence",
                        palette = "Reds",
                        direction = 1) +

  theme_tufte() +
  theme(axis.text = element_blank(),
        legend.position = "bottom") +
  facet_wrap(~week)

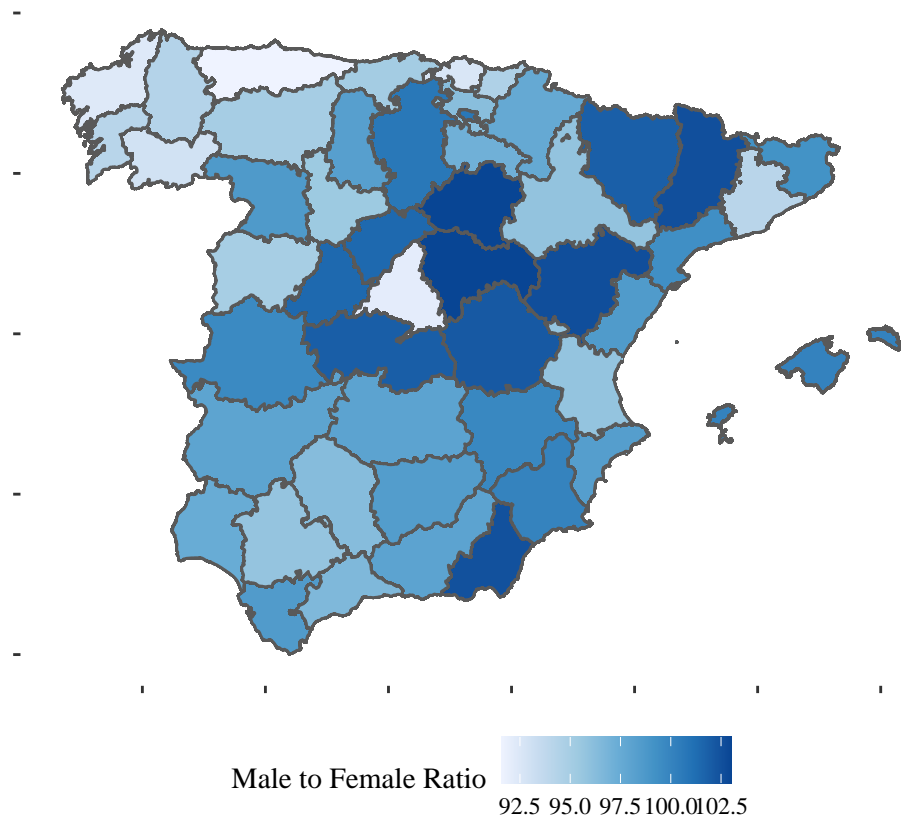
grid.arrange(week11.plot, week12.plot, week13.plot, week14.plot, nrow = 2)

```

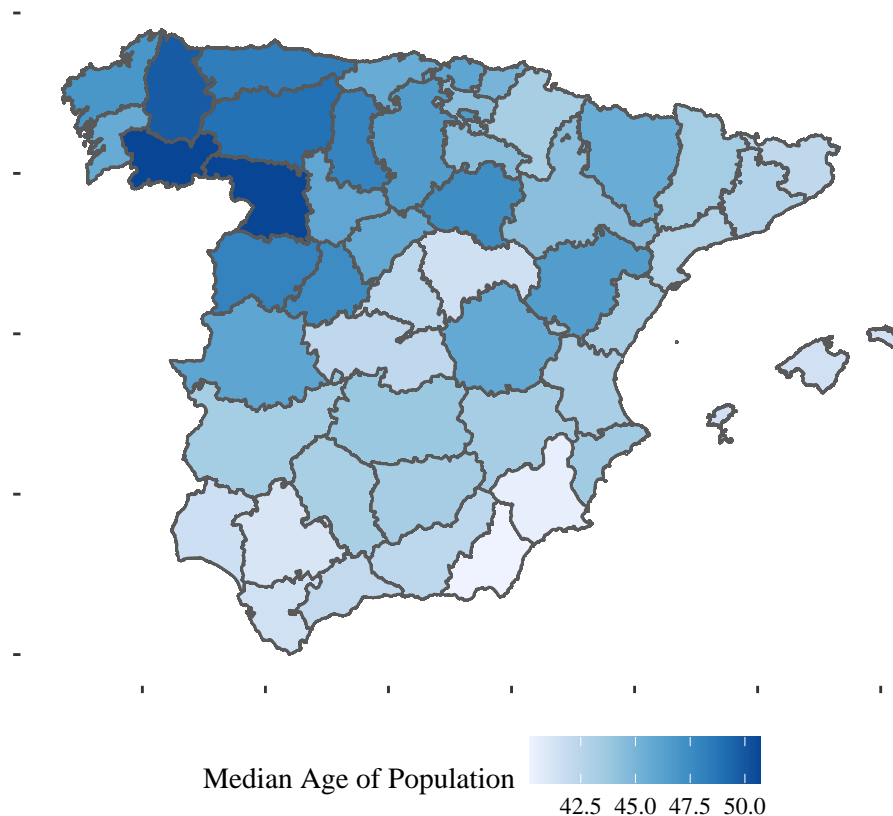


We consider two control variables: ratio of male to female in the province, and median age of the population:

```
covid19_spain %>%
  filter(CCAA != "Canarias") %>%
  ggplot() +
  geom_sf(aes(fill = Male2Female)) +
  scale_fill_distiller(name = "Male to Female Ratio",
                       palette = "Blues",
                       direction = 1) +
  theme_tufte() +
  theme(axis.text = element_blank(),
        legend.position = "bottom")
```



```
covid19_spain %>%
  filter(CCAA != "Canarias") %>%
  ggplot() +
  geom_sf(aes(fill = Median_Age)) +
  scale_fill_distiller(name = "Median Age of Population",
    palette = "Blues",
    direction = 1) +
  theme_tufte() +
  theme(axis.text = element_blank(),
    legend.position = "bottom")
```



We are also interested in the climatic variables. The following plot is the distribution of temperature by CCAA:

```
# Autonomous communities
ccaa.sf <- covid19_spain %>%
  filter(Date == "2020-03-14") %>%
  group_by(CCAA) %>%
  summarize(provinces = n())

# Extract coordinates of autonomous communities
ccaa.coords <- ccaa.sf %>%
  st_centroid() %>%
  st_coordinates() %>%
  as.data.frame()

## Warning in st_centroid.sf(.): st_centroid assumes attributes are constant over
## geometries of x

## Warning in st_centroid.sfc(st_geometry(x), of_largest_polygon =
## of_largest_polygon): st_centroid does not give correct centroids for longitude/
## latitude data

# Join Y coordinate to ccaa.sf
ccaa.sf <- ccaa.sf %>%
  mutate(long = ccaa.coords$Y)

# Sort autonomous communities from north to south
ccaa.levels <- ccaa.sf %>%
```

```

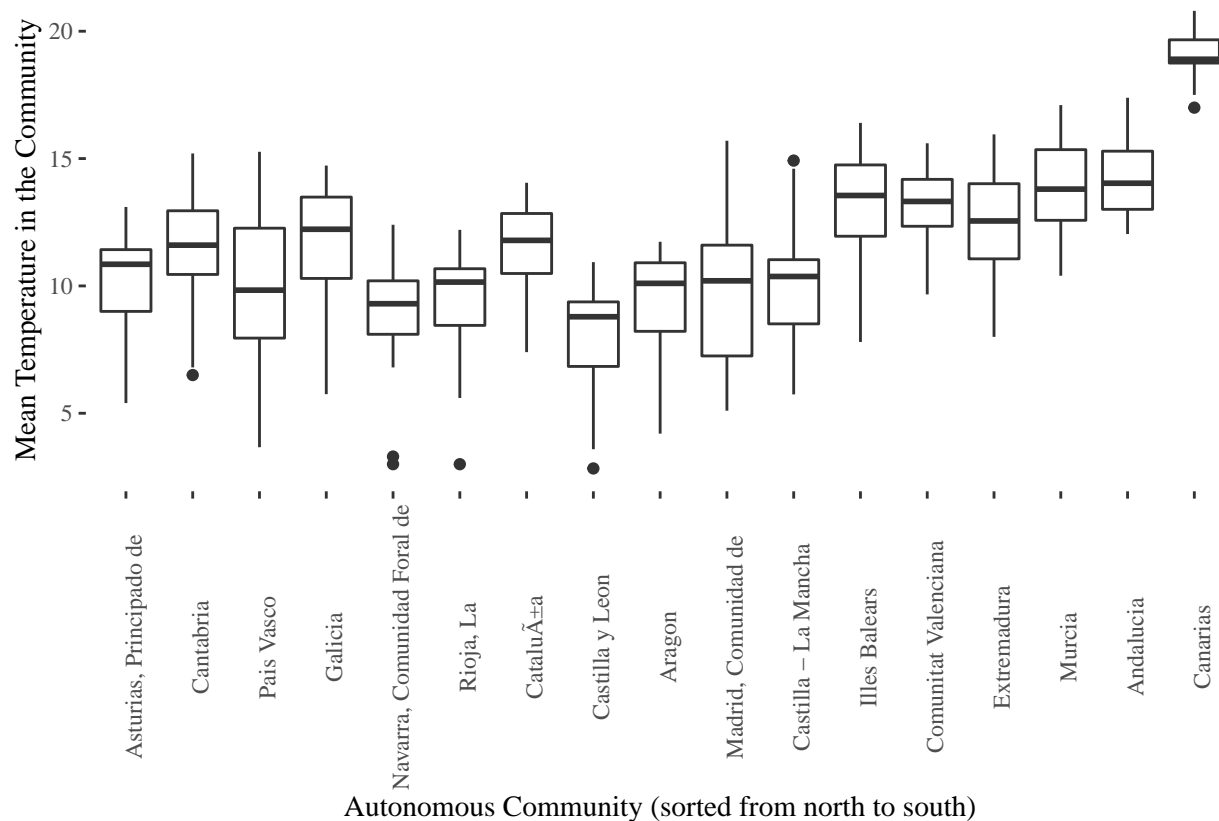
  arrange(desc(long)) %>% select(CCAA)

ccaa.levels <- as.character(ccaa.levels$CCAA)

# Relevel autonomous communities
covid19_spain <- covid19_spain %>%
  mutate(CCAA = factor(CCAA, levels = ccaa.levels, ordered = TRUE))

# Boxplots of temperatures
covid19_spain %>%
  st_drop_geometry() %>%
  group_by(CCAA, Date) %>%
  summarize(Mean_Temp = mean(Mean_Temp)) %>%
  ggplot(aes(x = CCAA, y = Mean_Temp)) +
  geom_boxplot() +
  theme_tufte() +
  theme(axis.text.x = element_text(angle = 90)) +
  xlab("Autonomous Community (sorted from north to south)") +
  ylab("Mean Temperature in the Community")

```



Multivariate analysis: comparison of approaches

Panel

- 1) Panel clásico
- 2) Panel Clásico o Dinámico

- Debe ser un modelo de efectos fijos para recoger la heterogeneidad entre las distintas provincias (efectos)
- Debería incluir estructura dinámica ya que la serie tiene una fuerte estructura temporal
- INCONVENIENTE: considera que la influencia del dato del día anterior es constante (se estima un coeficiente constante)
- INCONVENIENTE: No se pueden incluir variables constantes en T. La heterogeneidad entre provincias queda en el efecto fijo. No podemos por tanto incluir datos sobre estructura de la población.
- INCONVENIENTE: No podemos incorporar efectos espaciales. El paquete **splm** no incluye estimación de paneles dinámicos con efectos espaciales. Tendríamos que hacerlo en matlab con los códigos de P.Elhorst.

Spatial SUR

2) SUR espacial

- Hay un coeficiente para cada variable y cada instante de tiempo. Aunque es posible considerar coeficientes constantes para los periodos temporales que consideremos.
- La heterogeneidad espacial debemos incorporarla mediante variables explicativas. -> Estructura de la población relacionada con COVID-19.
- Permite incluir variables constantes en T.
- la dinámica temporal quedará recogida mediante el término independiente y la estructura de correlaciones en los residuos. EN TODO CASO, ENTIENDO QUE NUESTRO OBJETIVO NO ES EXPLICAR ESA TENDENCIA TEMPORAL (solo modelizarla para no incurrir en errores)

Prepare data for SUR analysis

*El modelo debe considerar efectos del 'individuo' y del 'tiempo' (para incorporar tendencia temporal)**

```
# Definición del panel para plm
GPanel <- plm::pdata.frame(covid19_spain %>%
                           st_drop_geometry() %>%
                           select(province,
                                  Date,
                                  Incidence,
                                  Median_Age,
                                  Male2Female,
                                  Mean_Temp_lag8,
                                  Humidity_lag8,
                                  Sunshine_Hours_lag8,
                                  Mean_Temp_lag11,
                                  Humidity_lag11,
                                  Sunshine_Hours_lag11,
                                  Mean_Temp_lag11w,
                                  Humidity_lag11w,
                                  Sunshine_Hours_lag11w),
                           c("province", "Date"))
```

Modelo SUR espacial

Construcción de W

```
# Definición matriz de contactos
Wmat <- covid19_spain %>%
  filter(Date == "2020-03-14") %>%
  as("Spatial") %>%
```

```

poly2nb(queen = FALSE) %>%
nb2mat(zero.policy = T)

Wmat <- (Wmat > 0) * 1

#W <- poly2nb(as(provincias.sf, "Spatial"), queen = FALSE)
#Wmat <- nb2mat(W,zero.policy = T)
#Wmat <- (Wmat>0)*1
# Conexión de las dos provincias que forman Canarias
Wmat[37, 44] <- 1
Wmat[44, 37] <- 1
# 'Países Catalans'
n = 8
Wmat[9,n] <- 1
Wmat[n,9] <- 1
Wmat[n,47] <- 1
Wmat[47,n] <- 1
Wmat[n,43] <- 1
Wmat[43,n] <- 1
miW <- Wmat/rowSums(Wmat)
listw <- mat2listw(Wmat,style = "W")

```

Se incluyen dos variables de control para recoger la herogeneidad espacial * HM: ratio Hombres/Mujeres (signo esperado +) * EM: Edad media (signo esperado +)

Define formulas with two different lagged variables:

```

formula_lag8 <- log(Incidence) ~ log(Male2Female) +
  log(Median_Age) +
  log(Mean_Temp_lag8) +
  log(Sunshine_Hours_lag8 + 0.1) +
  log(Humidity_lag8)

formula_lag11 <- log(Incidence) ~ log(Male2Female) +
  log(Median_Age) +
  log(Mean_Temp_lag11) +
  log(Sunshine_Hours_lag11 + 0.1) +
  log(Humidity_lag11)

formula_lag11w <- log(Incidence) ~ log(Male2Female) +
  log(Median_Age) +
  log(Mean_Temp_lag11w) +
  log(Sunshine_Hours_lag11w + 0.1) +
  log(Humidity_lag11w)

```

Model with 8-day moving average of climatic variables:

```

sur.slm_lag8 <- spsur::spsurtime(formula = formula_lag8,
                                data=GPanel,
                                time = GPanel$Date,
                                type = "slm",
                                fit_method = "3sls",
                                listw= listw)

```

```
## Time to fit the model: 0.39 seconds
```

```
summary(sur.slm_lag8)
```

```
## Call:
## spsur::spsurtime(formula = formula_lag8, data = GPanel, time = GPanel$Date,
##   listw = listw, type = "slm", fit_method = "3sls")
##
##
## Spatial SUR model type:  slm
##
## Equation  1
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_1      56.590345  25.274531  2.2390 0.0254707 *
## log(Male2Female)_1    -7.257491   4.224569 -1.7179 0.0862570 .
## log(Median_Age)_1     -5.919011   2.603779 -2.2732 0.0233170 *
## log(Mean_Temp_lag8)_1  -1.471130   0.421094 -3.4936 0.0005069 ***
## log(Sunshine_Hours_lag8 + 0.1)_1  0.084006   0.155765  0.5393 0.5898443
## log(Humidity_lag8)_1   0.944923   0.483890  1.9528 0.0512497 .
## rho_1               0.291876   0.139949  2.0856 0.0373813 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.3178
## Equation  2
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_2      47.94777  22.73704  2.1088  0.03532 *
## log(Male2Female)_2    -7.06368   3.88523 -1.8181  0.06948 .
## log(Median_Age)_2     -3.56330   2.28827 -1.5572  0.11988
## log(Mean_Temp_lag8)_2  -1.54689   0.38837 -3.9831 7.522e-05 ***
## log(Sunshine_Hours_lag8 + 0.1)_2  0.16203   0.16072  1.0081  0.31374
## log(Humidity_lag8)_2   0.83170   0.39156  2.1241  0.03402 *
## rho_2               0.13439   0.13355  1.0062  0.31466
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.2951
## Equation  3
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_3      34.21939  20.09783  1.7026  0.08908 .
## log(Male2Female)_3    -4.42740   3.45589 -1.2811  0.20058
## log(Median_Age)_3     -2.89065   2.03158 -1.4229  0.15523
## log(Mean_Temp_lag8)_3  -1.42479   0.34690 -4.1072 4.482e-05 ***
## log(Sunshine_Hours_lag8 + 0.1)_3  0.16152   0.17799  0.9075  0.36445
## log(Humidity_lag8)_3   0.53806   0.30203  1.7815  0.07527 .
## rho_3               0.24524   0.11390  2.1530  0.03166 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.3364
## Equation  4
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_4      23.401417  19.868868  1.1778 0.239283
## log(Male2Female)_4    -3.279917   3.448593 -0.9511 0.341891
## log(Median_Age)_4     -1.454878   2.015831 -0.7217 0.470706
## log(Mean_Temp_lag8)_4  -1.390587   0.333572 -4.1688 3.45e-05 ***
## log(Sunshine_Hours_lag8 + 0.1)_4  0.379034   0.161033  2.3538 0.018863 *
## log(Humidity_lag8)_4   0.436363   0.263870  1.6537 0.098640 .
## rho_4               0.294538   0.095806  3.0743 0.002193 **
```

```

## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.3054
## Equation 5
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_5      2.049188  21.049260  0.0974  0.922475
## log(Male2Female)_5 -0.393102   3.663915 -0.1073  0.914590
## log(Median_Age)_5   0.496715   2.139421  0.2322  0.816473
## log(Mean_Temp_lag8)_5 -0.969965   0.364236 -2.6630  0.007924 **
## log(Sunshine_Hours_lag8 + 0.1)_5 0.314294   0.205808  1.5271  0.127187
## log(Humidity_lag8)_5  0.332435   0.292633  1.1360  0.256345
## rho_5              0.422527   0.094083  4.4910 8.302e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.1987
## Equation 6
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_6      1.374842  19.164408  0.0717  0.9428301
## log(Male2Female)_6   0.991613   3.363995  0.2948  0.7682560
## log(Median_Age)_6  -0.063457   1.947875 -0.0326  0.9740210
## log(Mean_Temp_lag8)_6 -1.147675   0.303964 -3.7757  0.0001732 ***
## log(Sunshine_Hours_lag8 + 0.1)_6 0.105686   0.190358  0.5552  0.5789407
## log(Humidity_lag8)_6  0.053383   0.223245  0.2391  0.8110815
## rho_6              -0.013673   0.084298 -0.1622  0.8711941
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2218
## Equation 7
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_7      -0.4089311  17.8694123 -0.0229  0.98175
## log(Male2Female)_7   0.8338706   3.1436392  0.2653  0.79089
## log(Median_Age)_7    0.6157758   1.8370746  0.3352  0.73758
## log(Mean_Temp_lag8)_7 -1.2788255   0.2786907 -4.5887 5.295e-06 ***
## log(Sunshine_Hours_lag8 + 0.1)_7 0.3806654   0.1810285  2.1028  0.03584 *
## log(Humidity_lag8)_7 -0.0045695   0.2137444 -0.0214  0.98295
## rho_7              0.0397431   0.0801581  0.4958  0.62019
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2344
## Equation 8
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_8      -4.405281  18.086236 -0.2436  0.807635
## log(Male2Female)_8   1.619900   3.156775  0.5132  0.608010
## log(Median_Age)_8    0.694167   1.871799  0.3709  0.710859
## log(Mean_Temp_lag8)_8 -1.452958   0.267622 -5.4291 7.84e-08 ***
## log(Sunshine_Hours_lag8 + 0.1)_8 0.450326   0.171745  2.6221  0.008932 **
## log(Humidity_lag8)_8  0.108661   0.228764  0.4750  0.634942
## rho_8              0.082296   0.079710  1.0324  0.302228
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2703
## Equation 9
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_9      2.744951  17.353754  0.1582  0.874364

```

```

## log(Male2Female)_9          0.580502    3.017012    0.1924    0.847478
## log(Median_Age)_9          0.119317    1.791143    0.0666    0.946907
## log(Mean_Temp_lag8)_9      -1.452979    0.267136   -5.4391    7.431e-08 ***
## log(Sunshine_Hours_lag8 + 0.1)_9 0.407038    0.156577    2.5996    0.009532 **
## log(Humidity_lag8)_9       0.128710    0.223574    0.5757    0.565010
## rho_9                      0.060599    0.082286    0.7364    0.461708
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2733
## Equation 10
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_10      2.834789   15.846039   0.1789   0.85807
## log(Male2Female)_10   0.188904    2.743175   0.0689   0.94512
## log(Median_Age)_10    0.682894    1.644857   0.4152   0.67815
## log(Mean_Temp_lag8)_10 -1.529881    0.271088  -5.6435  2.43e-08 ***
## log(Sunshine_Hours_lag8 + 0.1)_10 0.394323    0.159585   2.4709   0.01372 *
## log(Humidity_lag8)_10  0.216856    0.232729   0.9318   0.35176
## rho_10              -0.025740    0.084262  -0.3055   0.76009
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.3371
## Equation 11
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_11      2.6261320  14.6642367   0.1791   0.85792
## log(Male2Female)_11   0.7588334   2.5334638   0.2995   0.76463
## log(Median_Age)_11    0.2663820   1.5140854   0.1759   0.86040
## log(Mean_Temp_lag8)_11 -1.5429034   0.2230854  -6.9162  1.056e-11 ***
## log(Sunshine_Hours_lag8 + 0.1)_11 0.2222326   0.0897557   2.4760   0.01353 *
## log(Humidity_lag8)_11  0.1273906   0.2029517   0.6277   0.53041
## rho_11              -0.0018179   0.0701847  -0.0259   0.97934
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4012
## Equation 12
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_12      7.118579   14.382602   0.4949   0.6207967
## log(Male2Female)_12  -0.014487    2.476846  -0.0058   0.9953348
## log(Median_Age)_12   -0.150206    1.474119  -0.1019   0.9188693
## log(Mean_Temp_lag8)_12 -1.590710    0.211330  -7.5271  1.618e-13 ***
## log(Sunshine_Hours_lag8 + 0.1)_12 0.346296    0.099599   3.4769   0.0005389 ***
## log(Humidity_lag8)_12  0.412037    0.215851   1.9089   0.0566885 .
## rho_12              -0.104448    0.070929  -1.4726   0.1413219
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4386
## Equation 13
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_13     11.138884   13.452282   0.8280   0.407939
## log(Male2Female)_13  -1.126404    2.324547  -0.4846   0.628135
## log(Median_Age)_13    0.281947    1.360124   0.2073   0.835840
## log(Mean_Temp_lag8)_13 -1.637440    0.192898  -8.4886 < 2.2e-16 ***
## log(Sunshine_Hours_lag8 + 0.1)_13 0.268683    0.078668   3.4154   0.000674 ***
## log(Humidity_lag8)_13  0.338827    0.199626   1.6973   0.090086 .
## rho_13              -0.074246    0.065763  -1.1290   0.259293

```

```

## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4988
## Equation 14
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)_14      14.105360  13.858007  1.0178  0.30910
## log(Male2Female)_14    -1.967121  2.401674 -0.8191  0.41303
## log(Median_Age)_14      0.431891  1.382130  0.3125  0.75477
## log(Mean_Temp_lag8)_14  -1.639816  0.197590 -8.2991 5.479e-16 ***
## log(Sunshine_Hours_lag8 + 0.1)_14  0.287538  0.063693  4.5145 7.458e-06 ***
## log(Humidity_lag8)_14    0.427114  0.219815  1.9431  0.05241 .
## rho_14                -0.061443  0.069297 -0.8867  0.37557
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4804
## Equation 15
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)_15      6.148692  13.335339  0.4611  0.64488
## log(Male2Female)_15   -1.428515  2.318592 -0.6161  0.53802
## log(Median_Age)_15    2.069942  1.319331  1.5689  0.11712
## log(Mean_Temp_lag8)_15 -1.493138  0.184238 -8.1044 2.398e-15 ***
## log(Sunshine_Hours_lag8 + 0.1)_15  0.163444  0.052741  3.0990  0.00202 **
## log(Humidity_lag8)_15    0.207659  0.211255  0.9830  0.32596
## rho_15                -0.025761  0.073195 -0.3520  0.72498
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.51
## Equation 16
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)_16      2.1794037  13.0354064  0.1672  0.867268
## log(Male2Female)_16   -0.9317497  2.2679889 -0.4108  0.681327
## log(Median_Age)_16    2.5930882  1.2846786  2.0185  0.043927 *
## log(Mean_Temp_lag8)_16 -1.5096094  0.1954797 -7.7226 3.998e-14 ***
## log(Sunshine_Hours_lag8 + 0.1)_16  0.1422681  0.0510053  2.7893  0.005428 **
## log(Humidity_lag8)_16    0.1510630  0.1876508  0.8050  0.421083
## rho_16                0.0087468  0.0723014  0.1210  0.903744
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5064
## Equation 17
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)_17      5.523397  12.426436  0.4445  0.6568
## log(Male2Female)_17   -1.200782  2.163160 -0.5551  0.5790
## log(Median_Age)_17    2.076583  1.221978  1.6994  0.0897 .
## log(Mean_Temp_lag8)_17 -1.537280  0.181018 -8.4924 < 2.2e-16 ***
## log(Sunshine_Hours_lag8 + 0.1)_17  0.215539  0.052761  4.0852 4.919e-05 ***
## log(Humidity_lag8)_17    0.148054  0.161515  0.9167  0.3596
## rho_17                -0.013813  0.065373 -0.2113  0.8327
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5276
## Equation 18
##
##           Estimate Std. Error t value Pr(>|t|)
## (Intercept)_18      5.290393  12.656187  0.4180 0.6760704

```

```

## log(Male2Female)_18          -1.556755    2.197782 -0.7083 0.4789786
## log(Median_Age)_18           2.560824    1.244012  2.0585 0.0399135 *
## log(Mean_Temp_lag8)_18       -1.291377    0.172456 -7.4882 2.13e-13 ***
## log(Sunshine_Hours_lag8 + 0.1)_18 0.205285    0.061506  3.3376 0.0008901 ***
## log(Humidity_lag8)_18        0.088840    0.169940  0.5228 0.6012985
## rho_18                       -0.074457    0.072643 -1.0250 0.3057382
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.51
## Equation  19
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_19      0.871139  12.942322  0.0673  0.9464
## log(Male2Female)_19 -0.213160  2.217477 -0.0961  0.9234
## log(Median_Age)_19   2.081718  1.281445  1.6245  0.1047
## log(Mean_Temp_lag8)_19 -1.284355  0.198247 -6.4785 1.756e-10 ***
## log(Sunshine_Hours_lag8 + 0.1)_19 0.128331  0.078135  1.6424  0.1010
## log(Humidity_lag8)_19  0.070632  0.228839  0.3087  0.7577
## rho_19              -0.019762  0.096924 -0.2039  0.8385
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5233
## Equation  20
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_20      1.280947  12.799995  0.1001  0.92031
## log(Male2Female)_20 -0.907790  2.228217 -0.4074  0.68384
## log(Median_Age)_20   2.946753  1.273953  2.3131  0.02101 *
## log(Mean_Temp_lag8)_20 -1.025793  0.148198 -6.9218 1.018e-11 ***
## log(Sunshine_Hours_lag8 + 0.1)_20 0.109231  0.050411  2.1668  0.03059 *
## log(Humidity_lag8)_20 -0.025138  0.141411 -0.1778  0.85896
## rho_20              -0.143481  0.074322 -1.9305  0.05395 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4934
## Equation  21
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_21      2.079365  12.632748  0.1646  0.86931
## log(Male2Female)_21 -0.769726  2.188687 -0.3517  0.72518
## log(Median_Age)_21   2.727955  1.268717  2.1502  0.03189 *
## log(Mean_Temp_lag8)_21 -1.041427  0.149615 -6.9607 7.865e-12 ***
## log(Sunshine_Hours_lag8 + 0.1)_21 0.070565  0.067475  1.0458  0.29602
## log(Humidity_lag8)_21 -0.083360  0.139886 -0.5959  0.55142
## rho_21              -0.178023  0.078059 -2.2806  0.02287 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5048
## Equation  22
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_22      -0.837759  13.119953 -0.0639  0.94911
## log(Male2Female)_22 -0.434774  2.261166 -0.1923  0.84758
## log(Median_Age)_22   3.072665  1.332482  2.3060  0.02141 *
## log(Mean_Temp_lag8)_22 -0.929009  0.142481 -6.5202 1.353e-10 ***
## log(Sunshine_Hours_lag8 + 0.1)_22 -0.017026  0.071616 -0.2377  0.81215
## log(Humidity_lag8)_22 -0.051339  0.159781 -0.3213  0.74807
## rho_22              -0.202667  0.083688 -2.4217  0.01570 *

```

```

## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.4904
##
## Variance-Covariance Matrix of inter-equation residuals:
##  0.7204966 0.6338744 0.5248385 0.5034573 0.4965724 0.4427671 0.3831851
##  0.6338744 0.6298058 0.5309442 0.5189484 0.5005380 0.4385852 0.3878907
##  0.5248385 0.5309442 0.5126003 0.5023227 0.4938828 0.4259262 0.3840332
##  0.5034573 0.5189484 0.5023227 0.5207078 0.5271069 0.4622847 0.4190731
##  0.4965724 0.5005380 0.4938828 0.5271069 0.5892863 0.5117040 0.4657377
##  0.4427671 0.4385852 0.4259262 0.4622847 0.5117040 0.5051340 0.4600011
##  0.3831851 0.3878907 0.3840332 0.4190731 0.4657377 0.4600011 0.4441410
##  0.3758362 0.3747196 0.3759859 0.4141676 0.4589614 0.4589156 0.4426326
##  0.3465666 0.3398036 0.3413130 0.3825363 0.4270146 0.4320255 0.4159743
##  0.3040524 0.3009956 0.2887747 0.3293635 0.3684856 0.3747414 0.3620352
##  0.2592032 0.2556305 0.2589953 0.3009569 0.3311459 0.3414950 0.3284456
##  0.2435988 0.2377375 0.2367796 0.2766476 0.2875362 0.3071727 0.2939594
##  0.2054385 0.1979124 0.1903441 0.2232803 0.2269505 0.2548870 0.2468752
##  0.1913551 0.1854059 0.1739522 0.2057288 0.2041980 0.2355340 0.2307224
##  0.1797603 0.1685549 0.1539840 0.1844245 0.1861417 0.2178577 0.2114469
##  0.1867252 0.1784313 0.1642707 0.1933115 0.1980927 0.2213245 0.2147772
##  0.2072797 0.1968467 0.1799338 0.2036649 0.2067721 0.2267522 0.2190055
##  0.2147320 0.2097544 0.2012665 0.2227491 0.2249899 0.2387266 0.2312747
##  0.1858132 0.1802004 0.1871397 0.2074304 0.2138924 0.2246355 0.2113395
##  0.1934434 0.1925595 0.1947170 0.2201190 0.2268172 0.2435416 0.2344280
##  0.1808506 0.1829938 0.1910915 0.2163209 0.2246524 0.2398431 0.2337634
##  0.1898604 0.1942008 0.2068172 0.2317878 0.2409563 0.2519133 0.2473689
##
##  0.3758362 0.3465666 0.3040524 0.2592032 0.2435988 0.2054385 0.1913551
##  0.3747196 0.3398036 0.3009956 0.2556305 0.2377375 0.1979124 0.1854059
##  0.3759859 0.3413130 0.2887747 0.2589953 0.2367796 0.1903441 0.1739522
##  0.4141676 0.3825363 0.3293635 0.3009569 0.2766476 0.2232803 0.2057288
##  0.4589614 0.4270146 0.3684856 0.3311459 0.2875362 0.2269505 0.2041980
##  0.4589156 0.4320255 0.3747414 0.3414950 0.3071727 0.2548870 0.2355340
##  0.4426326 0.4159743 0.3620352 0.3284456 0.2939594 0.2468752 0.2307224
##  0.4559346 0.4289766 0.3747412 0.3426445 0.3088026 0.2572942 0.2369501
##  0.4289766 0.4173214 0.3657602 0.3352962 0.3064949 0.2573535 0.2409552
##  0.3747412 0.3657602 0.3448464 0.3084914 0.2853803 0.2443765 0.2285279
##  0.3426445 0.3352962 0.3084914 0.2951812 0.2781640 0.2415946 0.2301893
##  0.3088026 0.3064949 0.2853803 0.2781640 0.2804359 0.2504533 0.2414661
##  0.2572942 0.2573535 0.2443765 0.2415946 0.2504533 0.2451296 0.2444020
##  0.2369501 0.2409552 0.2285279 0.2301893 0.2414661 0.2444020 0.2593096
##  0.2166498 0.2209653 0.2103085 0.2096288 0.2216165 0.2281916 0.2446799
##  0.2181580 0.2186134 0.2078068 0.2042845 0.2140455 0.2199090 0.2336014
##  0.2235455 0.2195802 0.2073256 0.2001169 0.2075717 0.2105525 0.2163261
##  0.2366994 0.2298630 0.2117848 0.2027512 0.2062947 0.2065077 0.2096814
##  0.2187901 0.2124077 0.1917928 0.1888356 0.1867801 0.1874331 0.1895215
##  0.2442258 0.2426262 0.2213352 0.2123494 0.2134278 0.2086216 0.2124750
##  0.2469517 0.2441019 0.2220891 0.2115318 0.2089502 0.1989333 0.1988693
##  0.2620160 0.2568315 0.2331108 0.2204748 0.2140491 0.1991887 0.1949911
##
##  0.1797603 0.1867252 0.2072797 0.2147320 0.1858132 0.1934434 0.1808506
##  0.1685549 0.1784313 0.1968467 0.2097544 0.1802004 0.1925595 0.1829938
##  0.1539840 0.1642707 0.1799338 0.2012665 0.1871397 0.1947170 0.1910915

```



```

## 0.1844245 0.1933115 0.2036649 0.2227491 0.2074304 0.2201190 0.2163209
## 0.1861417 0.1980927 0.2067721 0.2249899 0.2138924 0.2268172 0.2246524
## 0.2178577 0.2213245 0.2267522 0.2387266 0.2246355 0.2435416 0.2398431
## 0.2114469 0.2147772 0.2190055 0.2312747 0.2113395 0.2344280 0.2337634
## 0.2166498 0.2181580 0.2235455 0.2366994 0.2187901 0.2442258 0.2469517
## 0.2209653 0.2186134 0.2195802 0.2298630 0.2124077 0.2426262 0.2441019
## 0.2103085 0.2078068 0.2073256 0.2117848 0.1917928 0.2213352 0.2220891
## 0.2096288 0.2042845 0.2001169 0.2027512 0.1888356 0.2123494 0.2115318
## 0.2216165 0.2140455 0.2075717 0.2062947 0.1867801 0.2134278 0.2089502
## 0.2281916 0.2199090 0.2105525 0.2065077 0.1874331 0.2086216 0.1989333
## 0.2446799 0.2336014 0.2163261 0.2096814 0.1895215 0.2124750 0.1988693
## 0.2397552 0.2295874 0.2096668 0.2030287 0.1855671 0.2064033 0.1922656
## 0.2295874 0.2286284 0.2127675 0.2077443 0.1860718 0.2041977 0.1903563
## 0.2096668 0.2127675 0.2076949 0.2050807 0.1817340 0.1980023 0.1869805
## 0.2030287 0.2077443 0.2050807 0.2120664 0.1927143 0.2058265 0.1958122
## 0.1855671 0.1860718 0.1817340 0.1927143 0.2093427 0.1990144 0.1909999
## 0.2064033 0.2041977 0.1980023 0.2058265 0.1990144 0.2204402 0.2142988
## 0.1922656 0.1903563 0.1869805 0.1958122 0.1909999 0.2142988 0.2137079
## 0.1880288 0.1863707 0.1852072 0.1971267 0.1944000 0.2167237 0.2186024
##
## 0.1898604
## 0.1942008
## 0.2068172
## 0.2317878
## 0.2409563
## 0.2519133
## 0.2473689
## 0.2620160
## 0.2568315
## 0.2331108
## 0.2204748
## 0.2140491
## 0.1991887
## 0.1949911
## 0.1880288
## 0.1863707
## 0.1852072
## 0.1971267
## 0.1944000
## 0.2167237
## 0.2186024
## 0.2285957
## Correlation Matrix of inter-equation residuals:
## 1.0000000 0.9445816 0.8938163 0.8661542 0.8373391 0.8101958 0.7708630
## 0.9445816 1.0000000 0.9451757 0.9304032 0.8773063 0.8322519 0.8120188
## 0.8938163 0.9451757 1.0000000 0.9792927 0.9294495 0.8769215 0.8662072
## 0.8661542 0.9304032 0.9792927 1.0000000 0.9639141 0.9168930 0.9050991
## 0.8373391 0.8773063 0.9294495 0.9639141 1.0000000 0.9339933 0.9197071
## 0.8101958 0.8322519 0.8769215 0.9168930 0.9339933 1.0000000 0.9786770
## 0.7708630 0.8120188 0.8662072 0.9050991 0.9197071 0.9786770 1.0000000
## 0.7566578 0.7812539 0.8408729 0.8831944 0.8988455 0.9685694 0.9892148
## 0.7423731 0.7619515 0.8146018 0.8633070 0.8851652 0.9583502 0.9769525
## 0.7207786 0.7492124 0.7745457 0.8290247 0.8617076 0.9339558 0.9549950
## 0.6967877 0.7253362 0.7756754 0.8381620 0.8561085 0.9389615 0.9577203

```

```

## 0.6676240 0.7019273 0.7484964 0.8147587 0.8180856 0.9166077 0.9328076
## 0.6219317 0.6605636 0.7024378 0.7651165 0.7633608 0.8745780 0.8999962
## 0.5990744 0.6479528 0.6872911 0.7473215 0.7278307 0.8366859 0.8675195
## 0.6027280 0.6378640 0.6740610 0.7359286 0.7245940 0.8347616 0.8607348
## 0.6157629 0.6562523 0.6937587 0.7529302 0.7475903 0.8459257 0.8757802
## 0.6434247 0.6771577 0.7104274 0.7631799 0.7589742 0.8609755 0.8927743
## 0.6313686 0.6741060 0.7232996 0.7708472 0.7652871 0.8568941 0.8936217
## 0.5705892 0.5985786 0.6790025 0.7254579 0.7182127 0.8128706 0.8352580
## 0.5926739 0.6349243 0.7014461 0.7533305 0.7503567 0.8454823 0.8789065
## 0.5823713 0.6239287 0.6970513 0.7474359 0.7451136 0.8389184 0.8782199
## 0.5817169 0.6274675 0.7027771 0.7516652 0.7480255 0.8399266 0.8811851
##
## 0.7566578 0.7423731 0.7207786 0.6967877 0.6676240 0.6219317 0.5990744
## 0.7812539 0.7619515 0.7492124 0.7253362 0.7019273 0.6605636 0.6479528
## 0.8408729 0.8146018 0.7745457 0.7756754 0.7484964 0.7024378 0.6872911
## 0.8831944 0.8633070 0.8290247 0.8381620 0.8147587 0.7651165 0.7473215
## 0.8988455 0.8851652 0.8617076 0.8561085 0.8180856 0.7633608 0.7278307
## 0.9685694 0.9583502 0.9339558 0.9389615 0.9166077 0.8745780 0.8366859
## 0.9892148 0.9769525 0.9549950 0.9577203 0.9328076 0.8999962 0.8675195
## 1.0000000 0.9890681 0.9651935 0.9725644 0.9501531 0.9147858 0.8778029
## 0.9890681 1.0000000 0.9747125 0.9778546 0.9608101 0.9226943 0.8890251
## 0.9651935 0.9747125 1.0000000 0.9787840 0.9650226 0.9340517 0.8983736
## 0.9725644 0.9778546 0.9787840 1.0000000 0.9867736 0.9614474 0.9338393
## 0.9501531 0.9608101 0.9650226 0.9867736 1.0000000 0.9808549 0.9570179
## 0.9147858 0.9226943 0.9340517 0.9614474 0.9808549 1.0000000 0.9864019
## 0.8778029 0.8890251 0.8983736 0.9338393 0.9570179 0.9864019 1.0000000
## 0.8730873 0.8839318 0.8962729 0.9256021 0.9506892 0.9808442 0.9890561
## 0.8848736 0.8897946 0.8998215 0.9225337 0.9433285 0.9724035 0.9768180
## 0.9048329 0.9037153 0.9098429 0.9289280 0.9473356 0.9721968 0.9675762
## 0.9056357 0.9023920 0.8985835 0.9196890 0.9328886 0.9554446 0.9509259
## 0.8523283 0.8421470 0.8264951 0.8669846 0.8689958 0.8938894 0.8865751
## 0.8945268 0.8981785 0.8890995 0.9170156 0.9350876 0.9505018 0.9462658
## 0.8979029 0.9005224 0.8872284 0.9150499 0.9305573 0.9396425 0.9306487
## 0.9009673 0.9028524 0.8878954 0.9144090 0.9254097 0.9276195 0.9150496
##
## 0.6027280 0.6157629 0.6434247 0.6313686 0.5705892 0.5926739 0.5823713
## 0.6378640 0.6562523 0.6771577 0.6741060 0.5985786 0.6349243 0.6239287
## 0.6740610 0.6937587 0.7104274 0.7232996 0.6790025 0.7014461 0.6970513
## 0.7359286 0.7529302 0.7631799 0.7708472 0.7254579 0.7533305 0.7474359
## 0.7245940 0.7475903 0.7589742 0.7652871 0.7182127 0.7503567 0.7451136
## 0.8347616 0.8459257 0.8609755 0.8568941 0.8128706 0.8454823 0.8389184
## 0.8607348 0.8757802 0.8927743 0.8936217 0.8352580 0.8789065 0.8782199
## 0.8730873 0.8848736 0.9048329 0.9056357 0.8523283 0.8945268 0.8979029
## 0.8839318 0.8897946 0.9037153 0.9023920 0.8421470 0.8981785 0.9005224
## 0.8962729 0.8998215 0.9098429 0.8985835 0.8264951 0.8890995 0.8872284
## 0.9256021 0.9225337 0.9289280 0.9196890 0.8669846 0.9170156 0.9150499
## 0.9506892 0.9433285 0.9473356 0.9328886 0.8689958 0.9350876 0.9305573
## 0.9808442 0.9724035 0.9721968 0.9554446 0.8938894 0.9505018 0.9396425
## 0.9890561 0.9768180 0.9675762 0.9509259 0.8865751 0.9462658 0.9306487
## 1.0000000 0.9932931 0.9840188 0.9695566 0.9116768 0.9641555 0.9473015
## 0.9932931 1.0000000 0.9934411 0.9848425 0.9233301 0.9730374 0.9577685
## 0.9840188 0.9934411 1.0000000 0.9903833 0.9233172 0.9755836 0.9652907
## 0.9695566 0.9848425 0.9903833 1.0000000 0.9424860 0.9886098 0.9809881
## 0.9116768 0.9233301 0.9233172 0.9424860 1.0000000 0.9461421 0.9380402

```

```
## 0.9641555 0.9730374 0.9755836 0.9886098 0.9461421 1.0000000 0.9959666
## 0.9473015 0.9577685 0.9652907 0.9809881 0.9380402 0.9959666 1.0000000
## 0.9312500 0.9429089 0.9523723 0.9736483 0.9324938 0.9890101 0.9964102
##
## 0.5817169
## 0.6274675
## 0.7027771
## 0.7516652
## 0.7480255
## 0.8399266
## 0.8811851
## 0.9009673
## 0.9028524
## 0.8878954
## 0.9144090
## 0.9254097
## 0.9276195
## 0.9150496
## 0.9312500
## 0.9429089
## 0.9523723
## 0.9736483
## 0.9324938
## 0.9890101
## 0.9964102
## 1.0000000
##
## R-sq. pooled: 0.7738
## Breusch-Pagan: 6954 p-value: ( 0)
```

Model with 11-day moving average of climatic variables:

```
sur.slm_lag11 <- spsur::spsurtime(formula = formula_lag11,
                                data=GPanel,
                                time = GPanel$Date,
                                type = "slm",
                                fit_method = "3sls",
                                listw= listw)
```

```
## Time to fit the model: 0.58 seconds
```

```
summary(sur.slm_lag11)
```

```
## Call:
## spsur::spsurtime(formula = formula_lag11, data = GPanel, time = GPanel$Date,
##   listw = listw, type = "slm", fit_method = "3sls")
##
##
## Spatial SUR model type: slm
##
## Equation 1
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_1  48.41362   26.37951   1.8353 0.066894 .
## log(Male2Female)_1 -5.544849    4.443317 -1.2479 0.212486
## log(Median_Age)_1 -5.107345    2.746269 -1.8597 0.063346 .
## log(Mean_Temp_lag11)_1 -1.639449    0.582809 -2.8130 0.005047 **
```

```

## log(Sunshine_Hours_lag11 + 0.1)_1  0.042746    0.273852    0.1561 0.876006
## log(Humidity_lag11)_1              0.407414    0.550804    0.7397 0.459750
## rho_1                             0.294248    0.134863    2.1818 0.029457 *
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2647
## Equation 2
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_2      38.36960    23.45387   1.6360 0.102302
## log(Male2Female)_2   -5.10121     4.02552  -1.2672 0.205503
## log(Median_Age)_2    -2.81390     2.42078  -1.1624 0.245476
## log(Mean_Temp_lag11)_2 -1.46000     0.50986  -2.8635 0.004316 **
## log(Sunshine_Hours_lag11 + 0.1)_2  0.04241     0.24973   0.1698 0.865196
## log(Humidity_lag11)_2  0.29722     0.45469   0.6537 0.513537
## rho_2              0.10572     0.11578   0.9131 0.361518
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2288
## Equation 3
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_3      30.18608    20.82864   1.4493 0.1477178
## log(Male2Female)_3   -3.55522     3.59894  -0.9879 0.3235708
## log(Median_Age)_3    -2.56329     2.14679  -1.1940 0.2328829
## log(Mean_Temp_lag11)_3 -1.41480     0.41213  -3.4329 0.0006327 ***
## log(Sunshine_Hours_lag11 + 0.1)_3  0.16021     0.24604   0.6512 0.5151574
## log(Humidity_lag11)_3  0.20479     0.36642   0.5589 0.5764135
## rho_3              0.28634     0.10043   2.8511 0.0044862 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2756
## Equation 4
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_4      24.966198    20.480895   1.2190 0.223259
## log(Male2Female)_4   -2.930008     3.551056  -0.8251 0.409594
## log(Median_Age)_4    -1.918448     2.123095  -0.9036 0.366517
## log(Mean_Temp_lag11)_4 -1.486651     0.373174  -3.9838 7.5e-05 ***
## log(Sunshine_Hours_lag11 + 0.1)_4  0.350411     0.244212   1.4349 0.151776
## log(Humidity_lag11)_4  0.171467     0.333066   0.5148 0.606847
## rho_4              0.259670     0.088354   2.9390 0.003402 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2712
## Equation 5
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_5      11.519228    20.928471   0.5504 0.5822159
## log(Male2Female)_5   -0.801973     3.600187  -0.2228 0.8237889
## log(Median_Age)_5    -1.125536     2.185911  -0.5149 0.6067836
## log(Mean_Temp_lag11)_5 -1.456676     0.389410  -3.7407 0.0001987 ***
## log(Sunshine_Hours_lag11 + 0.1)_5  0.371161     0.277678   1.3367 0.1817727
## log(Humidity_lag11)_5  0.337052     0.360012   0.9362 0.3494835
## rho_5              0.293129     0.086841   3.3755 0.0007780 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2577

```

```

## Equation 6
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_6 7.38863 18.62453 0.3967 0.6917
## log(Male2Female)_6 1.12891 3.21418 0.3512 0.7255
## log(Median_Age)_6 -1.47949 1.93751 -0.7636 0.4454
## log(Mean_Temp_lag11)_6 -1.45681 0.34046 -4.2789 2.142e-05 ***
## log(Sunshine_Hours_lag11 + 0.1)_6 -0.12078 0.20592 -0.5865 0.5577
## log(Humidity_lag11)_6 0.12246 0.26976 0.4540 0.6500
## rho_6 -0.12415 0.08074 -1.5377 0.1246
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2884
## Equation 7
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_7 6.429536 16.606563 0.3872 0.6988
## log(Male2Female)_7 1.013027 2.873661 0.3525 0.7246
## log(Median_Age)_7 -0.971000 1.726086 -0.5625 0.5739
## log(Mean_Temp_lag11)_7 -1.712589 0.292389 -5.8572 7.271e-09 ***
## log(Sunshine_Hours_lag11 + 0.1)_7 0.121880 0.157646 0.7731 0.4397
## log(Humidity_lag11)_7 0.048670 0.223012 0.2182 0.8273
## rho_7 -0.029941 0.070892 -0.4223 0.6729
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.3593
## Equation 8
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_8 7.908506 16.617523 0.4759 0.6342859
## log(Male2Female)_8 0.616991 2.862049 0.2156 0.8293812
## log(Median_Age)_8 -0.994125 1.724127 -0.5766 0.5643995
## log(Mean_Temp_lag11)_8 -2.241336 0.291807 -7.6809 5.4e-14 ***
## log(Sunshine_Hours_lag11 + 0.1)_8 0.550106 0.148066 3.7153 0.0002193 ***
## log(Humidity_lag11)_8 0.262748 0.218896 1.2003 0.2304189
## rho_8 0.065160 0.071398 0.9126 0.3617563
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4062
## Equation 9
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_9 17.428126 16.120001 1.0811 0.2800
## log(Male2Female)_9 -1.190685 2.765525 -0.4305 0.6669
## log(Median_Age)_9 -1.361279 1.655441 -0.8223 0.4112
## log(Mean_Temp_lag11)_9 -2.330495 0.291374 -7.9983 5.299e-15 ***
## log(Sunshine_Hours_lag11 + 0.1)_9 0.729471 0.164352 4.4385 1.054e-05 ***
## log(Humidity_lag11)_9 0.354499 0.242343 1.4628 0.1440
## rho_9 0.016548 0.086572 0.1911 0.8485
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4138
## Equation 10
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_10 14.091397 14.900255 0.9457 0.3446235
## log(Male2Female)_10 -1.543865 2.548351 -0.6058 0.5448266
## log(Median_Age)_10 -0.129264 1.525694 -0.0847 0.9325048
## log(Mean_Temp_lag11)_10 -2.118055 0.277509 -7.6324 7.649e-14 ***

```

```

## log(Sunshine_Hours_lag11 + 0.1)_10 0.551808 0.155905 3.5394 0.0004279 ***
## log(Humidity_lag11)_10 0.479776 0.255458 1.8781 0.0607869 .
## rho_10 -0.041520 0.087722 -0.4733 0.6361418
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4493
## Equation 11
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_11 10.765483 13.868288 0.7763 0.43786
## log(Male2Female)_11 -0.545143 2.391240 -0.2280 0.81973
## log(Median_Age)_11 -0.293207 1.405823 -0.2086 0.83485
## log(Mean_Temp_lag11)_11 -1.956502 0.225714 -8.6681 < 2e-16 ***
## log(Sunshine_Hours_lag11 + 0.1)_11 0.218053 0.099786 2.1852 0.02921 *
## log(Humidity_lag11)_11 0.338180 0.229144 1.4758 0.14044
## rho_11 0.033167 0.071482 0.4640 0.64280
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4832
## Equation 12
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_12 11.9847344 14.3838645 0.8332 0.405015
## log(Male2Female)_12 -0.9338541 2.4812392 -0.3764 0.706760
## log(Median_Age)_12 -0.4292512 1.4545800 -0.2951 0.768003
## log(Mean_Temp_lag11)_12 -1.8426582 0.2231855 -8.2562 7.603e-16 ***
## log(Sunshine_Hours_lag11 + 0.1)_12 0.2337495 0.0909874 2.5690 0.010407 *
## log(Humidity_lag11)_12 0.5955628 0.2259649 2.6356 0.008586 **
## rho_12 0.0030812 0.0720207 0.0428 0.965888
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4578
## Equation 13
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_13 16.230181 13.829928 1.1736 0.240976
## log(Male2Female)_13 -1.981472 2.381088 -0.8322 0.405599
## log(Median_Age)_13 -0.176307 1.391695 -0.1267 0.899227
## log(Mean_Temp_lag11)_13 -1.921547 0.223764 -8.5874 < 2.2e-16 ***
## log(Sunshine_Hours_lag11 + 0.1)_13 0.202477 0.074956 2.7013 0.007077 **
## log(Humidity_lag11)_13 0.523438 0.214080 2.4451 0.014731 *
## rho_13 0.065175 0.072450 0.8996 0.368654
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4946
## Equation 14
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_14 18.983310 14.048181 1.3513 0.177040
## log(Male2Female)_14 -2.721227 2.415976 -1.1263 0.260409
## log(Median_Age)_14 -0.037472 1.401664 -0.0267 0.978680
## log(Mean_Temp_lag11)_14 -1.842188 0.231235 -7.9667 6.697e-15 ***
## log(Sunshine_Hours_lag11 + 0.1)_14 0.222262 0.068386 3.2501 0.001209 **
## log(Humidity_lag11)_14 0.519279 0.230076 2.2570 0.024319 *
## rho_14 0.061747 0.075718 0.8155 0.415075
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4881

```

```

## Equation 15
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_15 10.401954 13.092655 0.7945 0.4271835
## log(Male2Female)_15 -1.918158 2.266571 -0.8463 0.3976875
## log(Median_Age)_15 1.333592 1.299989 1.0258 0.3053207
## log(Mean_Temp_lag11)_15 -1.627684 0.186534 -8.7259 < 2.2e-16 ***
## log(Sunshine_Hours_lag11 + 0.1)_15 0.208723 0.055021 3.7935 0.0001615 ***
## log(Humidity_lag11)_15 0.296260 0.188424 1.5723 0.1163361
## rho_15 0.105820 0.069823 1.5155 0.1300925
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5257
## Equation 16
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_16 6.073675 12.689649 0.4786 0.6323513
## log(Male2Female)_16 -1.334684 2.203851 -0.6056 0.5449691
## log(Median_Age)_16 1.807088 1.261448 1.4326 0.1524373
## log(Mean_Temp_lag11)_16 -1.535966 0.179548 -8.5546 < 2.2e-16 ***
## log(Sunshine_Hours_lag11 + 0.1)_16 0.206566 0.056196 3.6758 0.0002554 ***
## log(Humidity_lag11)_16 0.225458 0.167136 1.3489 0.1777945
## rho_16 0.107643 0.068954 1.5611 0.1189629
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5305
## Equation 17
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_17 7.887738 12.620938 0.6250 0.5322
## log(Male2Female)_17 -1.360410 2.194412 -0.6199 0.5355
## log(Median_Age)_17 1.363442 1.255713 1.0858 0.2779
## log(Mean_Temp_lag11)_17 -1.482038 0.169828 -8.7267 < 2.2e-16 ***
## log(Sunshine_Hours_lag11 + 0.1)_17 0.269325 0.057108 4.7160 2.909e-06 ***
## log(Humidity_lag11)_17 0.220004 0.138939 1.5835 0.1138
## rho_17 0.071966 0.065874 1.0925 0.2750
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5222
## Equation 18
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_18 8.400154 12.479203 0.6731 0.5011
## log(Male2Female)_18 -1.733994 2.169552 -0.7992 0.4244
## log(Median_Age)_18 1.771988 1.241646 1.4271 0.1540
## log(Mean_Temp_lag11)_18 -1.376277 0.167192 -8.2317 9.160e-16 ***
## log(Sunshine_Hours_lag11 + 0.1)_18 0.295847 0.067239 4.3999 1.253e-05 ***
## log(Humidity_lag11)_18 0.140308 0.143978 0.9745 0.3301
## rho_18 0.013477 0.071123 0.1895 0.8498
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5307
## Equation 19
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_19 3.854042 13.083411 0.2946 0.76841
## log(Male2Female)_19 -0.240650 2.233310 -0.1078 0.91422
## log(Median_Age)_19 1.145395 1.319022 0.8684 0.38549
## log(Mean_Temp_lag11)_19 -1.363888 0.191515 -7.1216 2.677e-12 ***

```

```

## log(Sunshine_Hours_lag11 + 0.1)_19 0.245130 0.099958 2.4523 0.01444 *
## log(Humidity_lag11)_19 0.140775 0.230444 0.6109 0.54147
## rho_19 0.046550 0.096619 0.4818 0.63011
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5266
## Equation 20
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_20 5.653701 12.833702 0.4405 0.65969
## log(Male2Female)_20 -1.077455 2.222385 -0.4848 0.62796
## log(Median_Age)_20 1.899559 1.298504 1.4629 0.14395
## log(Mean_Temp_lag11)_20 -1.104157 0.151473 -7.2894 8.511e-13 ***
## log(Sunshine_Hours_lag11 + 0.1)_20 0.110393 0.059813 1.8456 0.06537 .
## log(Humidity_lag11)_20 -0.046420 0.170463 -0.2723 0.78546
## rho_20 -0.030479 0.077843 -0.3916 0.69551
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5071
## Equation 21
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_21 7.271966 12.558728 0.5790 0.5628
## log(Male2Female)_21 -1.045307 2.168125 -0.4821 0.6299
## log(Median_Age)_21 1.664692 1.277865 1.3027 0.1931
## log(Mean_Temp_lag11)_21 -1.111652 0.149024 -7.4595 2.605e-13 ***
## log(Sunshine_Hours_lag11 + 0.1)_21 0.029470 0.061270 0.4810 0.6307
## log(Humidity_lag11)_21 -0.167351 0.171868 -0.9737 0.3305
## rho_21 -0.059654 0.078880 -0.7563 0.4498
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5259
## Equation 22
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_22 4.5641e+00 1.3007e+01 0.3509 0.7258
## log(Male2Female)_22 -7.2524e-01 2.2417e+00 -0.3235 0.7464
## log(Median_Age)_22 2.0171e+00 1.3295e+00 1.5172 0.1297
## log(Mean_Temp_lag11)_22 -1.0410e+00 1.4984e-01 -6.9475 8.584e-12
## log(Sunshine_Hours_lag11 + 0.1)_22 -1.1705e-05 6.3680e-02 -0.0002 0.9999
## log(Humidity_lag11)_22 -1.9965e-01 1.8065e-01 -1.1052 0.2695
## rho_22 -6.6000e-02 8.1673e-02 -0.8081 0.4193
##
## (Intercept)_22
## log(Male2Female)_22
## log(Median_Age)_22
## log(Mean_Temp_lag11)_22 ***
## log(Sunshine_Hours_lag11 + 0.1)_22
## log(Humidity_lag11)_22
## rho_22
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5155
##
## Variance-Covariance Matrix of inter-equation residuals:
## 0.8120411 0.7056270 0.5977397 0.5702723 0.5660145 0.4847193 0.4167961
## 0.7056270 0.6937768 0.5918479 0.5791645 0.5609433 0.4728287 0.4147636

```



```

## 0.5977397 0.5918479 0.5663387 0.5525895 0.5380267 0.4503976 0.4004265
## 0.5702723 0.5791645 0.5525895 0.5590887 0.5493546 0.4673474 0.4131887
## 0.5660145 0.5609433 0.5380267 0.5493546 0.5757850 0.4840060 0.4267388
## 0.4847193 0.4728287 0.4503976 0.4673474 0.4840060 0.4662178 0.4086676
## 0.4167961 0.4147636 0.4004265 0.4131887 0.4267388 0.4086676 0.3779338
## 0.3993952 0.3838801 0.3786634 0.3927655 0.4066919 0.3970265 0.3679210
## 0.3753265 0.3559380 0.3485577 0.3633619 0.3780162 0.3681318 0.3431070
## 0.3338609 0.3218883 0.2959867 0.3126347 0.3304431 0.3252263 0.3056945
## 0.2831147 0.2799092 0.2712641 0.2932040 0.3022511 0.3078853 0.2870375
## 0.2627641 0.2630604 0.2548522 0.2790136 0.2804282 0.3018132 0.2797317
## 0.2159196 0.2186631 0.2115736 0.2313112 0.2283665 0.2578441 0.2420319
## 0.2111760 0.2150964 0.2068294 0.2249457 0.2182853 0.2463041 0.2327205
## 0.2017883 0.2001996 0.1922614 0.2087625 0.2061565 0.2328113 0.2197039
## 0.2176515 0.2147667 0.2042642 0.2180205 0.2177597 0.2345462 0.2229551
## 0.2286920 0.2212602 0.2091919 0.2212841 0.2202133 0.2386806 0.2272998
## 0.2229936 0.2217544 0.2146321 0.2260577 0.2238648 0.2367117 0.2257802
## 0.1900137 0.1954066 0.2061940 0.2200444 0.2161950 0.2276962 0.2133018
## 0.2108540 0.2144210 0.2174838 0.2309651 0.2310089 0.2446008 0.2326560
## 0.2082455 0.2117045 0.2164109 0.2284173 0.2282304 0.2377999 0.2291498
## 0.2177786 0.2244016 0.2306529 0.2434490 0.2420955 0.2501162 0.2414299
##
## 0.3993952 0.3753265 0.3338609 0.2831147 0.2627641 0.2159196 0.2111760
## 0.3838801 0.3559380 0.3218883 0.2799092 0.2630604 0.2186631 0.2150964
## 0.3786634 0.3485577 0.2959867 0.2712641 0.2548522 0.2115736 0.2068294
## 0.3927655 0.3633619 0.3126347 0.2932040 0.2790136 0.2313112 0.2249457
## 0.4066919 0.3780162 0.3304431 0.3022511 0.2804282 0.2283665 0.2182853
## 0.3970265 0.3681318 0.3252263 0.3078853 0.3018132 0.2578441 0.2463041
## 0.3679210 0.3431070 0.3056945 0.2870375 0.2797317 0.2420319 0.2327205
## 0.3749958 0.3541071 0.3115622 0.2931748 0.2848373 0.2456083 0.2364708
## 0.3541071 0.3472546 0.3046557 0.2820039 0.2733500 0.2344957 0.2265209
## 0.3115622 0.3046557 0.2932029 0.2635006 0.2580610 0.2258557 0.2171090
## 0.2931748 0.2820039 0.2635006 0.2595524 0.2608781 0.2343400 0.2270012
## 0.2848373 0.2733500 0.2580610 0.2608781 0.2786683 0.2574149 0.2516742
## 0.2456083 0.2344957 0.2258557 0.2343400 0.2574149 0.2536483 0.2520978
## 0.2364708 0.2265209 0.2171090 0.2270012 0.2516742 0.2520978 0.2582067
## 0.2227937 0.2116195 0.2042314 0.2116084 0.2345705 0.2342851 0.2385148
## 0.2262713 0.2148706 0.2061268 0.2072423 0.2252164 0.2226776 0.2259486
## 0.2315683 0.2190383 0.2067165 0.2072975 0.2246326 0.2208924 0.2227347
## 0.2288748 0.2160566 0.2029150 0.2036670 0.2187182 0.2137457 0.2149273
## 0.2106241 0.1901949 0.1766695 0.1895122 0.2025737 0.1963199 0.1951276
## 0.2335005 0.2182119 0.2016051 0.2050060 0.2198689 0.2111505 0.2097947
## 0.2310784 0.2163424 0.1976645 0.1985528 0.2105332 0.1999238 0.1976595
## 0.2429233 0.2266225 0.2049223 0.2061361 0.2168079 0.2035519 0.2015255
##
## 0.2017883 0.2176515 0.2286920 0.2229936 0.1900137 0.2108540 0.2082455
## 0.2001996 0.2147667 0.2212602 0.2217544 0.1954066 0.2144210 0.2117045
## 0.1922614 0.2042642 0.2091919 0.2146321 0.2061940 0.2174838 0.2164109
## 0.2087625 0.2180205 0.2212841 0.2260577 0.2200444 0.2309651 0.2284173
## 0.2061565 0.2177597 0.2202133 0.2238648 0.2161950 0.2310089 0.2282304
## 0.2328113 0.2345462 0.2386806 0.2367117 0.2276962 0.2446008 0.2377999
## 0.2197039 0.2229551 0.2272998 0.2257802 0.2133018 0.2326560 0.2291498
## 0.2227937 0.2262713 0.2315683 0.2288748 0.2106241 0.2335005 0.2310784
## 0.2116195 0.2148706 0.2190383 0.2160566 0.1901949 0.2182119 0.2163424
## 0.2042314 0.2061268 0.2067165 0.2029150 0.1766695 0.2016051 0.1976645

```

```

## 0.2116084 0.2072423 0.2072975 0.2036670 0.1895122 0.2050060 0.1985528
## 0.2345705 0.2252164 0.2246326 0.2187182 0.2025737 0.2198689 0.2105332
## 0.2342851 0.2226776 0.2208924 0.2137457 0.1963199 0.2111505 0.1999238
## 0.2385148 0.2259486 0.2227347 0.2149273 0.1951276 0.2097947 0.1976595
## 0.2270670 0.2175742 0.2146954 0.2082726 0.1912063 0.2055211 0.1949726
## 0.2175742 0.2150386 0.2124267 0.2076010 0.1873559 0.2020256 0.1935406
## 0.2146954 0.2124267 0.2137000 0.2082847 0.1879573 0.2044474 0.1970802
## 0.2082726 0.2076010 0.2082847 0.2079457 0.1917091 0.2060348 0.1991069
## 0.1912063 0.1873559 0.1879573 0.1917091 0.2155187 0.2035494 0.1971637
## 0.2055211 0.2020256 0.2044474 0.2060348 0.2035494 0.2203075 0.2136188
## 0.1949726 0.1935406 0.1970802 0.1991069 0.1971637 0.2136188 0.2102634
## 0.1985999 0.1972417 0.2007025 0.2036024 0.2024764 0.2190456 0.2163732
##
## 0.2177786
## 0.2244016
## 0.2306529
## 0.2434490
## 0.2420955
## 0.2501162
## 0.2414299
## 0.2429233
## 0.2266225
## 0.2049223
## 0.2061361
## 0.2168079
## 0.2035519
## 0.2015255
## 0.1985999
## 0.1972417
## 0.2007025
## 0.2036024
## 0.2024764
## 0.2190456
## 0.2163732
## 0.2250850
## Correlation Matrix of inter-equation residuals:
## 1.0000000 0.9418727 0.8979235 0.8713492 0.8580815 0.8232855 0.7911893
## 0.9418727 1.0000000 0.9484051 0.9400377 0.9062664 0.8560149 0.8408058
## 0.8979235 0.9484051 1.0000000 0.9836517 0.9488438 0.8902893 0.8836218
## 0.8713492 0.9400377 0.9836517 1.0000000 0.9693633 0.9189988 0.9076899
## 0.8580815 0.9062664 0.9488438 0.9693633 1.0000000 0.9336573 0.9201693
## 0.8232855 0.8560149 0.8902893 0.9189988 0.9336573 1.0000000 0.9779079
## 0.7911893 0.8408058 0.8836218 0.9076899 0.9201693 0.9779079 1.0000000
## 0.7730138 0.8001018 0.8534203 0.8783596 0.8931134 0.9625350 0.9858411
## 0.7594705 0.7845102 0.8290109 0.8576277 0.8777197 0.9421682 0.9674957
## 0.7407585 0.7730717 0.7852224 0.8181488 0.8538723 0.9182368 0.9478609
## 0.7122458 0.7468917 0.7879184 0.8310226 0.8466348 0.9297517 0.9560427
## 0.6665176 0.7041866 0.7464551 0.7940148 0.7995132 0.9050094 0.9295493
## 0.6117013 0.6510842 0.6949088 0.7354493 0.7348136 0.8548673 0.8886742
## 0.6032309 0.6496938 0.6935039 0.7310573 0.7189969 0.8308634 0.8697996
## 0.6168515 0.6496255 0.6925008 0.7271069 0.7232367 0.8378556 0.8757448
## 0.6342347 0.6731127 0.7138005 0.7457960 0.7461956 0.8475060 0.8890463
## 0.6426778 0.6743545 0.7128299 0.7423104 0.7424894 0.8509771 0.8935877
## 0.6318924 0.6733929 0.7226432 0.7494996 0.7464736 0.8455958 0.8897114

```

```

## 0.5634466 0.5975547 0.6776267 0.7070040 0.6945162 0.8008515 0.8302588
## 0.5920144 0.6366489 0.7035742 0.7334897 0.7354153 0.8376292 0.8795892
## 0.5888237 0.6333894 0.7031391 0.7316893 0.7329875 0.8318697 0.8794526
## 0.5900192 0.6410802 0.7106373 0.7400674 0.7380289 0.8371331 0.8853015
##
## 0.7730138 0.7594705 0.7407585 0.7122458 0.6665176 0.6117013 0.6032309
## 0.8001018 0.7845102 0.7730717 0.7468917 0.7041866 0.6510842 0.6496938
## 0.8534203 0.8290109 0.7852224 0.7879184 0.7464551 0.6949088 0.6935039
## 0.8783596 0.8576277 0.8181488 0.8310226 0.7940148 0.7354493 0.7310573
## 0.8931134 0.8777197 0.8538723 0.8466348 0.7995132 0.7348136 0.7189969
## 0.9625350 0.9421682 0.9182368 0.9297517 0.9050094 0.8548673 0.8308634
## 0.9858411 0.9674957 0.9478609 0.9560427 0.9295493 0.8886742 0.8697996
## 1.0000000 0.9867944 0.9600631 0.9689279 0.9420993 0.8987102 0.8787952
## 0.9867944 1.0000000 0.9672660 0.9656657 0.9384925 0.8905986 0.8737365
## 0.9600631 0.9672660 1.0000000 0.9700037 0.9464422 0.9069241 0.8838030
## 0.9689279 0.9656657 0.9700037 1.0000000 0.9837024 0.9552088 0.9367945
## 0.9420993 0.9384925 0.9464422 0.9837024 1.0000000 0.9810254 0.9640752
## 0.8987102 0.8905986 0.9069241 0.9552088 0.9810254 1.0000000 0.9897187
## 0.8787952 0.8737365 0.8838030 0.9367945 0.9640752 0.9897187 1.0000000
## 0.8877012 0.8775358 0.8901713 0.9376514 0.9644955 0.9861479 0.9889524
## 0.8998033 0.8874782 0.8986870 0.9357103 0.9574771 0.9755806 0.9760082
## 0.9075057 0.8910418 0.8957359 0.9338541 0.9571775 0.9733272 0.9712325
## 0.9018305 0.8853935 0.8843158 0.9259553 0.9446338 0.9597278 0.9587520
## 0.8349366 0.8010245 0.7940077 0.8666854 0.8789335 0.8952348 0.8926833
## 0.8917158 0.8744508 0.8671223 0.9181560 0.9404809 0.9481809 0.9458024
## 0.8951088 0.8786383 0.8636080 0.9118930 0.9315965 0.9347489 0.9315157
## 0.8989439 0.8822184 0.8636458 0.9123114 0.9286538 0.9270146 0.9246910
##
## 0.6168515 0.6342347 0.6426778 0.6318924 0.5634466 0.5920144 0.5888237
## 0.6496255 0.6731127 0.6743545 0.6733929 0.5975547 0.6366489 0.6333894
## 0.6925008 0.7138005 0.7128299 0.7226432 0.6776267 0.7035742 0.7031391
## 0.7271069 0.7457960 0.7423104 0.7494996 0.7070040 0.7334897 0.7316893
## 0.7232367 0.7461956 0.7424894 0.7464736 0.6945162 0.7354153 0.7329875
## 0.8378556 0.8475060 0.8509771 0.8455958 0.8008515 0.8376292 0.8318697
## 0.8757448 0.8890463 0.8935877 0.8897114 0.8302588 0.8795892 0.8794526
## 0.8877012 0.8998033 0.9075057 0.9018305 0.8349366 0.8917158 0.8951088
## 0.8775358 0.8874782 0.8910418 0.8853935 0.8010245 0.8744508 0.8786383
## 0.8901713 0.8986870 0.8957359 0.8843158 0.7940077 0.8671223 0.8636080
## 0.9376514 0.9357103 0.9338541 0.9259553 0.8666854 0.9181560 0.9118930
## 0.9644955 0.9574771 0.9571775 0.9446338 0.8789335 0.9404809 0.9315965
## 0.9861479 0.9755806 0.9733272 0.9597278 0.8952348 0.9481809 0.9347489
## 0.9889524 0.9760082 0.9712325 0.9587520 0.8926833 0.9458024 0.9315157
## 1.0000000 0.9940076 0.9906452 0.9815198 0.9181101 0.9678634 0.9560617
## 0.9940076 1.0000000 0.9955750 0.9912790 0.9217867 0.9732159 0.9642415
## 0.9906452 0.9955750 1.0000000 0.9923036 0.9200950 0.9774277 0.9717221
## 0.9815198 0.9912790 0.9923036 1.0000000 0.9369400 0.9896569 0.9849052
## 0.9181101 0.9217867 0.9200950 0.9369400 1.0000000 0.9476018 0.9418986
## 0.9678634 0.9732159 0.9774277 0.9896569 0.9476018 1.0000000 0.9968051
## 0.9560617 0.9642415 0.9717221 0.9849052 0.9418986 0.9968051 1.0000000
## 0.9486809 0.9571571 0.9644521 0.9800781 0.9371319 0.9927857 0.9974767
##
## 0.5900192
## 0.6410802
## 0.7106373

```

```
## 0.7400674
## 0.7380289
## 0.8371331
## 0.8853015
## 0.8989439
## 0.8822184
## 0.8636458
## 0.9123114
## 0.9286538
## 0.9270146
## 0.9246910
## 0.9486809
## 0.9571571
## 0.9644521
## 0.9800781
## 0.9371319
## 0.9927857
## 0.9974767
## 1.0000000
##
## R-sq. pooled: 0.7835
## Breusch-Pagan: 7461 p-value: ( 0)
```

Model with 11-day weighted moving average of climatic variables:

```
sur.slm_lag11w <- spsur::spsurtime(formula = formula_lag11w,
                                   data=GPanel,
                                   time = GPanel$Date,
                                   type = "slm",
                                   fit_method = "3sls",
                                   listw= listw)
```

```
## Time to fit the model: 0.36 seconds
```

```
summary(sur.slm_lag11w)
```

```
## Call:
## spsur::spsurtime(formula = formula_lag11w, data = GPanel, time = GPanel$Date,
##   listw = listw, type = "slm", fit_method = "3sls")
##
##
## Spatial SUR model type: slm
##
## Equation 1
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_1    31.218726   26.156528   1.1935 0.2330684
## log(Male2Female)_1 -3.768725    4.483467  -0.8406 0.4008716
## log(Median_Age)_1  -3.150522    2.679908  -1.1756 0.2401551
## log(Mean_Temp_lag11w)_1 -1.070296    0.504505  -2.1215 0.0342359 *
## log(Sunshine_Hours_lag11w + 0.1)_1 0.025972    0.284809   0.0912 0.9273686
## log(Humidity_lag11w)_1  0.427115    0.428354   0.9971 0.3190598
## rho_1             0.403794    0.117744   3.4294 0.0006407 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2052
## Equation 2
```

```

##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_2      24.217204  23.487513  1.0311 0.302869
## log(Male2Female)_2    -3.637960   4.099355 -0.8874 0.375146
## log(Median_Age)_2     -1.264736   2.391334 -0.5289 0.597056
## log(Mean_Temp_lag11w)_2 -1.066531   0.442894 -2.4081 0.016296 *
## log(Sunshine_Hours_lag11w + 0.1)_2 0.140962   0.234510  0.6011 0.547976
## log(Humidity_lag11w)_2   0.308834   0.351107  0.8796 0.379381
## rho_2               0.289792   0.096292  3.0095 0.002712 **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.17
## Equation 3
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_3      12.597745  21.024574  0.5992  0.54924
## log(Male2Female)_3    -1.184885   3.693921 -0.3208  0.74848
## log(Median_Age)_3     -0.975471   2.134354 -0.4570  0.64779
## log(Mean_Temp_lag11w)_3 -0.798061   0.371871 -2.1461  0.03221 *
## log(Sunshine_Hours_lag11w + 0.1)_3 -0.121472   0.216008 -0.5623  0.57406
## log(Humidity_lag11w)_3   0.058640   0.308607  0.1900  0.84935
## rho_3               0.421874   0.080904  5.2145 2.436e-07 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2038
## Equation 4
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_4      11.700176  20.801550  0.5625  0.573980
## log(Male2Female)_4    -1.216784   3.652960 -0.3331  0.739163
## log(Median_Age)_4     -0.452433   2.110770 -0.2143  0.830341
## log(Mean_Temp_lag11w)_4 -0.891664   0.327500 -2.7226  0.006639 **
## log(Sunshine_Hours_lag11w + 0.1)_4 0.012906   0.215104  0.0600  0.952175
## log(Humidity_lag11w)_4  -0.123853   0.300243 -0.4125  0.680094
## rho_4               0.348264   0.072949  4.7741 2.204e-06 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2099
## Equation 5
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_5      -2.300799  21.587759 -0.1066  0.91515
## log(Male2Female)_5    0.701062   3.764219  0.1862  0.85231
## log(Median_Age)_5     0.680827   2.230929  0.3052  0.76032
## log(Mean_Temp_lag11w)_5 -0.745558   0.370682 -2.0113  0.04468 *
## log(Sunshine_Hours_lag11w + 0.1)_5 0.083063   0.308166  0.2695  0.78759
## log(Humidity_lag11w)_5  0.010621   0.357927  0.0297  0.97634
## rho_5               0.358940   0.081402  4.4095 1.201e-05 ***
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.1765
## Equation 6
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_6       1.678543  19.366707  0.0867  0.9309576
## log(Male2Female)_6    1.624680   3.367858  0.4824  0.6296688
## log(Median_Age)_6     -0.607193   2.016196 -0.3012  0.7633847
## log(Mean_Temp_lag11w)_6 -0.970983   0.290733 -3.3398 0.0008834 ***
## log(Sunshine_Hours_lag11w + 0.1)_6 -0.248948   0.233610 -1.0657 0.2869484

```

```

## log(Humidity_lag11w)_6          -0.125553    0.312607 -0.4016 0.6880792
## rho_6                          -0.085298    0.085745 -0.9948 0.3201906
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.2123
## Equation 7
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_7      1.75501183 16.58422845  0.1058   0.9158
## log(Male2Female)_7    1.30665744  2.88687936  0.4526   0.6510
## log(Median_Age)_7    -0.32194853  1.71348958 -0.1879   0.8510
## log(Mean_Temp_lag11w)_7 -1.19473405  0.21745834 -5.4941 5.521e-08
## log(Sunshine_Hours_lag11w + 0.1)_7 0.00338311  0.13008099  0.0260   0.9793
## log(Humidity_lag11w)_7 -0.04982730  0.24890764 -0.2002   0.8414
## rho_7               0.00046549  0.07513578  0.0062   0.9951
##
## (Intercept)_7
## log(Male2Female)_7
## log(Median_Age)_7
## log(Mean_Temp_lag11w)_7      ***
## log(Sunshine_Hours_lag11w + 0.1)_7
## log(Humidity_lag11w)_7
## rho_7
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.3294
## Equation 8
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_8      2.969292 15.573542  0.1907   0.84885
## log(Male2Female)_8    1.041146  2.709672  0.3842   0.70092
## log(Median_Age)_8    -0.333013  1.597970 -0.2084   0.83498
## log(Mean_Temp_lag11w)_8 -1.613850  0.205635 -7.8481 1.604e-14 ***
## log(Sunshine_Hours_lag11w + 0.1)_8 0.160783  0.090990  1.7670   0.07766 .
## log(Humidity_lag11w)_8  0.144623  0.219721  0.6582   0.51062
## rho_8               0.065441  0.072342  0.9046   0.36599
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4331
## Equation 9
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_9     19.827074 14.625032  1.3557 0.1756380
## log(Male2Female)_9  -1.583752  2.517029 -0.6292 0.5294157
## log(Median_Age)_9   -1.452606  1.496548 -0.9706 0.3320672
## log(Mean_Temp_lag11w)_9 -1.938786  0.226833 -8.5472 < 2.2e-16 ***
## log(Sunshine_Hours_lag11w + 0.1)_9 0.320749  0.091384  3.5099 0.0004773 ***
## log(Humidity_lag11w)_9  0.156560  0.223584  0.7002 0.4840180
## rho_9              0.076316  0.081351  0.9381 0.3485155
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.494
## Equation 10
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_10     25.031510 13.469546  1.8584  0.06354 .
## log(Male2Female)_10  -3.449000  2.297409 -1.5013  0.13374
## log(Median_Age)_10   -0.536838  1.359936 -0.3948  0.69315

```

```

## log(Mean_Temp_lag11w)_10      -2.125158    0.233852 -9.0876 < 2.2e-16 ***
## log(Sunshine_Hours_lag11w + 0.1)_10  0.459190    0.092466  4.9660 8.614e-07 ***
## log(Humidity_lag11w)_10        0.338656    0.233909  1.4478 0.14812
## rho_10                        -0.023886    0.084005 -0.2843 0.77623
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.5534
## Equation 11
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_11      23.061968   12.654366   1.8225 0.0688173 .
## log(Male2Female)_11  -2.982100    2.173335  -1.3721 0.1704666
## log(Median_Age)_11   -0.239749    1.250895  -0.1917 0.8480634
## log(Mean_Temp_lag11w)_11 -2.066155    0.209437 -9.8653 < 2.2e-16 ***
## log(Sunshine_Hours_lag11w + 0.1)_11  0.271070    0.073369  3.6946 0.0002376 ***
## log(Humidity_lag11w)_11  0.116594    0.222030  0.5251 0.5996613
## rho_11              -0.027141    0.072928 -0.3722 0.7098834
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.5851
## Equation 12
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_12      20.922015   13.552468   1.5438 0.1231
## log(Male2Female)_12  -2.904554    2.344699  -1.2388 0.2158
## log(Median_Age)_12    0.101964    1.323422  0.0770 0.9386
## log(Mean_Temp_lag11w)_12 -1.931028    0.213062 -9.0632 < 2.2e-16 ***
## log(Sunshine_Hours_lag11w + 0.1)_12  0.312998    0.070812  4.4202 1.145e-05 ***
## log(Humidity_lag11w)_12  0.266170    0.237909  1.1188 0.2636
## rho_12              -0.105780    0.069695  -1.5178 0.1295
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.5328
## Equation 13
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_13      19.678045   13.982628   1.4073 0.159781
## log(Male2Female)_13  -3.010935    2.416270  -1.2461 0.213146
## log(Median_Age)_13    0.730053    1.359940  0.5368 0.591559
## log(Mean_Temp_lag11w)_13 -1.826385    0.228397 -7.9966 5.368e-15 ***
## log(Sunshine_Hours_lag11w + 0.1)_13  0.225996    0.072953  3.0978 0.002028 **
## log(Humidity_lag11w)_13  0.126002    0.263894  0.4775 0.633177
## rho_13              -0.114892    0.070194  -1.6368 0.102131
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.5008
## Equation 14
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_14      21.114069   14.604822   1.4457 0.1487158
## log(Male2Female)_14  -3.170082    2.524453  -1.2557 0.2096298
## log(Median_Age)_14    0.494117    1.425526  0.3466 0.7289811
## log(Mean_Temp_lag11w)_14 -1.760930    0.231257 -7.6146 8.686e-14 ***
## log(Sunshine_Hours_lag11w + 0.1)_14  0.219154    0.063763  3.4370 0.0006233 ***
## log(Humidity_lag11w)_14  0.171270    0.255176  0.6712 0.5023270
## rho_14              -0.121903    0.068550  -1.7783 0.0757915 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```

## R-squared: 0.459
## Equation 15
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_15 10.554179 13.387372 0.7884 0.4308
## log(Male2Female)_15 -1.942054 2.324219 -0.8356 0.4037
## log(Median_Age)_15 1.839151 1.320345 1.3929 0.1641
## log(Mean_Temp_lag11w)_15 -1.552663 0.191953 -8.0888 2.697e-15 ***
## log(Sunshine_Hours_lag11w + 0.1)_15 0.198283 0.049740 3.9864 7.420e-05 ***
## log(Humidity_lag11w)_15 -0.019207 0.194181 -0.0989 0.9212
## rho_15 -0.063899 0.058267 -1.0967 0.2732
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.5107
## Equation 16
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_16 2.822534 13.241142 0.2132 0.831262
## log(Male2Female)_16 -0.817138 2.303213 -0.3548 0.722861
## log(Median_Age)_16 2.312276 1.322258 1.7487 0.080780 .
## log(Mean_Temp_lag11w)_16 -1.225014 0.167326 -7.3211 6.838e-13 ***
## log(Sunshine_Hours_lag11w + 0.1)_16 0.173655 0.053748 3.2309 0.001292 **
## log(Humidity_lag11w)_16 -0.050299 0.167882 -0.2996 0.764567
## rho_16 -0.024994 0.059765 -0.4182 0.675920
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4872
## Equation 17
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_17 2.180781 13.355523 0.1633 0.870340
## log(Male2Female)_17 -0.473969 2.321950 -0.2041 0.838315
## log(Median_Age)_17 1.928080 1.346416 1.4320 0.152592
## log(Mean_Temp_lag11w)_17 -1.012898 0.148957 -6.7999 2.262e-11 ***
## log(Sunshine_Hours_lag11w + 0.1)_17 0.170926 0.056783 3.0102 0.002706 **
## log(Humidity_lag11w)_17 -0.041051 0.153051 -0.2682 0.788611
## rho_17 -0.025699 0.064427 -0.3989 0.690096
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4612
## Equation 18
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_18 2.051648 12.991344 0.1579 0.874562
## log(Male2Female)_18 -0.704834 2.258592 -0.3121 0.755083
## log(Median_Age)_18 2.291176 1.315580 1.7416 0.082027 .
## log(Mean_Temp_lag11w)_18 -0.913034 0.135451 -6.7407 3.321e-11 ***
## log(Sunshine_Hours_lag11w + 0.1)_18 0.189830 0.060415 3.1421 0.001749 **
## log(Humidity_lag11w)_18 -0.050402 0.138358 -0.3643 0.715756
## rho_18 -0.103021 0.071257 -1.4458 0.148692
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared: 0.4789
## Equation 19
##
## Estimate Std. Error t value Pr(>|t|)
## (Intercept)_19 -2.747362 13.364466 -0.2056 0.8372
## log(Male2Female)_19 0.757211 2.284530 0.3315 0.7404
## log(Median_Age)_19 1.835603 1.375436 1.3346 0.1825

```



```

## log(Mean_Temp_lag11w)_19      -0.951805    0.165868 -5.7383 1.429e-08 ***
## log(Sunshine_Hours_lag11w + 0.1)_19  0.176831    0.112369  1.5737  0.1160
## log(Humidity_lag11w)_19      -0.081627    0.195896 -0.4167  0.6770
## rho_19                        -0.078424    0.100091 -0.7835  0.4336
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.4863
## Equation 20
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_20      0.235636  12.988773  0.0181  0.98553
## log(Male2Female)_20 -0.338211  2.244000 -0.1507  0.88024
## log(Median_Age)_20   2.517564  1.326650  1.8977  0.05815 .
## log(Mean_Temp_lag11w)_20 -0.798235  0.130110 -6.1351 1.431e-09 ***
## log(Sunshine_Hours_lag11w + 0.1)_20  0.077695  0.085778  0.9058  0.36537
## log(Humidity_lag11w)_20 -0.149784  0.161505 -0.9274  0.35403
## rho_20               -0.147578  0.082420 -1.7906  0.07380 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.4822
## Equation 21
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_21     -0.342190  12.760412 -0.0268  0.97861
## log(Male2Female)_21 -0.020484  2.195904 -0.0093  0.99256
## log(Median_Age)_21   2.467104  1.312883  1.8791  0.06064 .
## log(Mean_Temp_lag11w)_21 -0.699969  0.121847 -5.7446 1.379e-08 ***
## log(Sunshine_Hours_lag11w + 0.1)_21 -0.069046  0.082047 -0.8415  0.40034
## log(Humidity_lag11w)_21 -0.267353  0.182586 -1.4643  0.14358
## rho_21               -0.167206  0.086094 -1.9421  0.05253 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.4907
## Equation 22
##
##               Estimate Std. Error t value Pr(>|t|)
## (Intercept)_22     -2.538146  13.104478 -0.1937  0.84648
## log(Male2Female)_22  0.310560  2.250044  0.1380  0.89026
## log(Median_Age)_22   2.766201  1.353127  2.0443  0.04130 *
## log(Mean_Temp_lag11w)_22 -0.648306  0.127440 -5.0872 4.685e-07 ***
## log(Sunshine_Hours_lag11w + 0.1)_22 -0.100589  0.079447 -1.2661  0.20590
## log(Humidity_lag11w)_22 -0.404297  0.220082 -1.8370  0.06663 .
## rho_22               -0.151682  0.088374 -1.7164  0.08654 .
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## R-squared:  0.4871
##
## Variance-Covariance Matrix of inter-equation residuals:
##  0.8563332 0.7497242 0.6322792 0.6007832 0.6029492 0.5187877 0.4177832
##  0.7497242 0.7347252 0.6292667 0.6138023 0.6099079 0.5113207 0.4214504
##  0.6322792 0.6292667 0.6008984 0.5867378 0.5806025 0.4860563 0.4052907
##  0.6007832 0.6138023 0.5867378 0.5918263 0.5900993 0.5007987 0.4161214
##  0.6029492 0.6099079 0.5806025 0.5900993 0.6239275 0.5257925 0.4305935
##  0.5187877 0.5113207 0.4860563 0.5007987 0.5257925 0.5161139 0.4311239
##  0.4177832 0.4214504 0.4052907 0.4161214 0.4305935 0.4311239 0.3844768
##  0.3625120 0.3526641 0.3474777 0.3586845 0.3676760 0.3829246 0.3491968
##  0.3041263 0.2886560 0.2819368 0.2931898 0.3042254 0.3219244 0.2958000

```

```

## 0.2550321 0.2408281 0.2136559 0.2219894 0.2376497 0.2570044 0.2395906
## 0.2370496 0.2266858 0.2093509 0.2201508 0.2303050 0.2456785 0.2283934
## 0.2340774 0.2326916 0.2171659 0.2341084 0.2414538 0.2647688 0.2432877
## 0.2080061 0.2193082 0.2086251 0.2278481 0.2301010 0.2577245 0.2380308
## 0.2216757 0.2362124 0.2250390 0.2439842 0.2407950 0.2672806 0.2482261
## 0.2311607 0.2320090 0.2180712 0.2304447 0.2287546 0.2561513 0.2384601
## 0.2545636 0.2525107 0.2363416 0.2449688 0.2452561 0.2676841 0.2514216
## 0.2678014 0.2607718 0.2442968 0.2499241 0.2504577 0.2754850 0.2592658
## 0.2519683 0.2500895 0.2415900 0.2474108 0.2452706 0.2623356 0.2489268
## 0.2077219 0.2100918 0.2247922 0.2327921 0.2279200 0.2444980 0.2326043
## 0.2230292 0.2275485 0.2311797 0.2376763 0.2380552 0.2558500 0.2440500
## 0.2222662 0.2266294 0.2302258 0.2353565 0.2374153 0.2486204 0.2393266
## 0.2265676 0.2351800 0.2385733 0.2460933 0.2480463 0.2586339 0.2482126
##
## 0.3625120 0.3041263 0.2550321 0.2370496 0.2340774 0.2080061 0.2216757
## 0.3526641 0.2886560 0.2408281 0.2266858 0.2326916 0.2193082 0.2362124
## 0.3474777 0.2819368 0.2136559 0.2093509 0.2171659 0.2086251 0.2250390
## 0.3586845 0.2931898 0.2219894 0.2201508 0.2341084 0.2278481 0.2439842
## 0.3676760 0.3042254 0.2376497 0.2303050 0.2414538 0.2301010 0.2407950
## 0.3829246 0.3219244 0.2570044 0.2456785 0.2647688 0.2577245 0.2672806
## 0.3491968 0.2958000 0.2395906 0.2283934 0.2432877 0.2380308 0.2482261
## 0.3391591 0.2985692 0.2421360 0.2305512 0.2444102 0.2363193 0.2446453
## 0.2985692 0.2899408 0.2427850 0.2297525 0.2425349 0.2309601 0.2372488
## 0.2421360 0.2427850 0.2360346 0.2150874 0.2259770 0.2146201 0.2140925
## 0.2305512 0.2297525 0.2150874 0.2083799 0.2177413 0.2093984 0.2088829
## 0.2444102 0.2425349 0.2259770 0.2177413 0.2409546 0.2385367 0.2380995
## 0.2363193 0.2309601 0.2146201 0.2093984 0.2385367 0.2536318 0.2595611
## 0.2446453 0.2372488 0.2140925 0.2088829 0.2380995 0.2595611 0.2779699
## 0.2365750 0.2249668 0.2027239 0.1952304 0.2196417 0.2344326 0.2506091
## 0.2468782 0.2268166 0.1993080 0.1894889 0.2105407 0.2212101 0.2380263
## 0.2547158 0.2286772 0.1944286 0.1827063 0.2023358 0.2106907 0.2275878
## 0.2445054 0.2178066 0.1798397 0.1708389 0.1899406 0.1994579 0.2169882
## 0.2274722 0.1896538 0.1442249 0.1437451 0.1607225 0.1727776 0.1896094
## 0.2374458 0.2049969 0.1633138 0.1556990 0.1747443 0.1838365 0.2003958
## 0.2310646 0.1958998 0.1544069 0.1479350 0.1635375 0.1690506 0.1824259
## 0.2384070 0.2022447 0.1587804 0.1537163 0.1697704 0.1741402 0.1867310
##
## 0.2311607 0.2545636 0.2678014 0.2519683 0.2077219 0.2230292 0.2222662
## 0.2320090 0.2525107 0.2607718 0.2500895 0.2100918 0.2275485 0.2266294
## 0.2180712 0.2363416 0.2442968 0.2415900 0.2247922 0.2311797 0.2302258
## 0.2304447 0.2449688 0.2499241 0.2474108 0.2327921 0.2376763 0.2353565
## 0.2287546 0.2452561 0.2504577 0.2452706 0.2279200 0.2380552 0.2374153
## 0.2561513 0.2676841 0.2754850 0.2623356 0.2444980 0.2558500 0.2486204
## 0.2384601 0.2514216 0.2592658 0.2489268 0.2326043 0.2440500 0.2393266
## 0.2365750 0.2468782 0.2547158 0.2445054 0.2274722 0.2374458 0.2310646
## 0.2249668 0.2268166 0.2286772 0.2178066 0.1896538 0.2049969 0.1958998
## 0.2027239 0.1993080 0.1944286 0.1798397 0.1442249 0.1633138 0.1544069
## 0.1952304 0.1894889 0.1827063 0.1708389 0.1437451 0.1556990 0.1479350
## 0.2196417 0.2105407 0.2023358 0.1899406 0.1607225 0.1747443 0.1635375
## 0.2344326 0.2212101 0.2106907 0.1994579 0.1727776 0.1838365 0.1690506
## 0.2506091 0.2380263 0.2275878 0.2169882 0.1896094 0.2003958 0.1824259
## 0.2384326 0.2326068 0.2263204 0.2158617 0.1906997 0.1999071 0.1855681
## 0.2326068 0.2363078 0.2350985 0.2253884 0.2001729 0.2103706 0.1992798
## 0.2263204 0.2350985 0.2414262 0.2320593 0.2078580 0.2197747 0.2096844

```

```

## 0.2158617 0.2253884 0.2320593 0.2286708 0.2099120 0.2206044 0.2111419
## 0.1906997 0.2001729 0.2078580 0.2099120 0.2301533 0.2149733 0.2081532
## 0.1999071 0.2103706 0.2197747 0.2206044 0.2149733 0.2265371 0.2185486
## 0.1855681 0.1992798 0.2096844 0.2111419 0.2081532 0.2185486 0.2176507
## 0.1894883 0.2027301 0.2120364 0.2128743 0.2094530 0.2184003 0.2205664
##
## 0.2265676
## 0.2351800
## 0.2385733
## 0.2460933
## 0.2480463
## 0.2586339
## 0.2482126
## 0.2384070
## 0.2022447
## 0.1587804
## 0.1537163
## 0.1697704
## 0.1741402
## 0.1867310
## 0.1894883
## 0.2027301
## 0.2120364
## 0.2128743
## 0.2094530
## 0.2184003
## 0.2205664
## 0.2287325
## Correlation Matrix of inter-equation residuals:
## 1.0000000 0.9480206 0.9041731 0.8780157 0.8670069 0.8226090 0.7875695
## 0.9480206 1.0000000 0.9523757 0.9446388 0.9209563 0.8601493 0.8426436
## 0.9041731 0.9523757 1.0000000 0.9848051 0.9542590 0.8904519 0.8831998
## 0.8780157 0.9446388 0.9848051 1.0000000 0.9724348 0.9181212 0.9071031
## 0.8670069 0.9209563 0.9542590 0.9724348 1.0000000 0.9344318 0.9160073
## 0.8226090 0.8601493 0.8904519 0.9181212 0.9344318 1.0000000 0.9796575
## 0.7875695 0.8426436 0.8831998 0.9071031 0.9160073 0.9796575 1.0000000
## 0.7677213 0.8005744 0.8521540 0.8763432 0.8824662 0.9614320 0.9850111
## 0.7392845 0.7663128 0.8056897 0.8334128 0.8476520 0.9308319 0.9563911
## 0.7148796 0.7484447 0.7500434 0.7763393 0.8087051 0.8875587 0.9173747
## 0.7149075 0.7543286 0.7775873 0.8127472 0.8301017 0.9108145 0.9396659
## 0.6681767 0.7173894 0.7431131 0.7874154 0.8000044 0.8984977 0.9255316
## 0.6092464 0.6708908 0.6992569 0.7425747 0.7480056 0.8596438 0.8939176
## 0.5983852 0.6702386 0.6989755 0.7406074 0.7338035 0.8424386 0.8807517
## 0.6422804 0.6990348 0.7276699 0.7632134 0.7614689 0.8711820 0.9051662
## 0.6592964 0.7185728 0.7500242 0.7809307 0.7841886 0.8841586 0.9208291
## 0.6674957 0.7159249 0.7490134 0.7750818 0.7802444 0.8843367 0.9224677
## 0.6544941 0.7094061 0.7533633 0.7791204 0.7788037 0.8712502 0.9121976
## 0.5938477 0.6364650 0.7171035 0.7455681 0.7337886 0.8226166 0.8563631
## 0.6117376 0.6676592 0.7285899 0.7560904 0.7600177 0.8513286 0.8942976
## 0.6144783 0.6683163 0.7300645 0.7548962 0.7601615 0.8461744 0.8929741
## 0.6123803 0.6722243 0.7314936 0.7588995 0.7613426 0.8505173 0.8979832
##
## 0.7677213 0.7392845 0.7148796 0.7149075 0.6681767 0.6092464 0.5983852
## 0.8005744 0.7663128 0.7484447 0.7543286 0.7173894 0.6708908 0.6702386

```

```

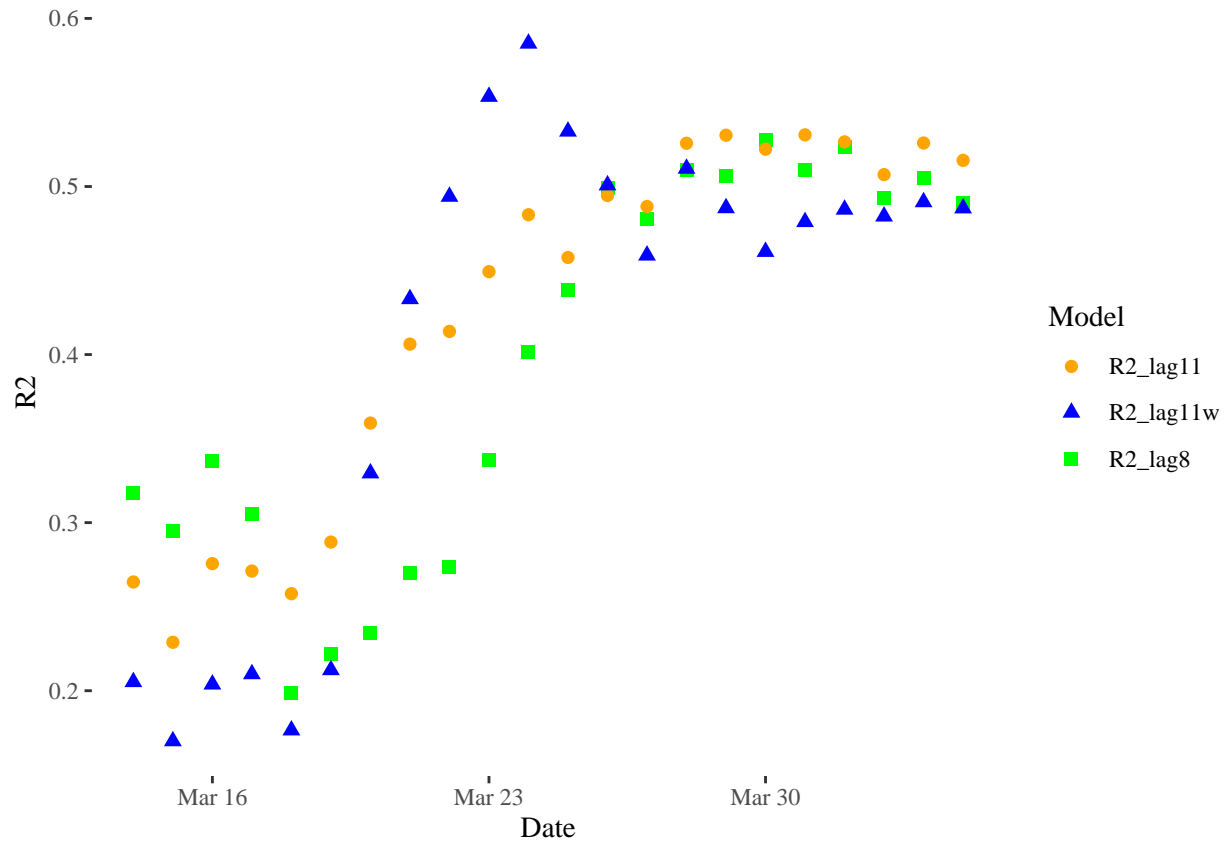
## 0.8521540 0.8056897 0.7500434 0.7775873 0.7431131 0.6992569 0.6989755
## 0.8763432 0.8334128 0.7763393 0.8127472 0.7874154 0.7425747 0.7406074
## 0.8824662 0.8476520 0.8087051 0.8301017 0.8000044 0.7480056 0.7338035
## 0.9614320 0.9308319 0.8875587 0.9108145 0.8984977 0.8596438 0.8424386
## 0.9850111 0.9563911 0.9173747 0.9396659 0.9255316 0.8939176 0.8807517
## 1.0000000 0.9795184 0.9345845 0.9587410 0.9467290 0.9123612 0.8955044
## 0.9795184 1.0000000 0.9568190 0.9729609 0.9630639 0.9238232 0.9067281
## 0.9345845 0.9568190 1.0000000 0.9765729 0.9654988 0.9275270 0.8965200
## 0.9587410 0.9729609 0.9765729 1.0000000 0.9836429 0.9549084 0.9303749
## 0.9467290 0.9630639 0.9654988 0.9836429 1.0000000 0.9797549 0.9540387
## 0.9123612 0.9238232 0.9275270 0.9549084 0.9797549 1.0000000 0.9854090
## 0.8955044 0.9067281 0.8965200 0.9303749 0.9540387 0.9854090 1.0000000
## 0.9223675 0.9205971 0.9124620 0.9396445 0.9607125 0.9807600 0.9859489
## 0.9342251 0.9218003 0.9067353 0.9309860 0.9490218 0.9629303 0.9679850
## 0.9392395 0.9198896 0.8942021 0.9167160 0.9344017 0.9438136 0.9472053
## 0.9283500 0.9069099 0.8685756 0.9002751 0.9173773 0.9295720 0.9384114
## 0.8736185 0.8293017 0.7673602 0.8233512 0.8366235 0.8558438 0.8656963
## 0.9108552 0.8877651 0.8374839 0.8753436 0.8971082 0.9061633 0.9174618
## 0.9092543 0.8817710 0.8272351 0.8641872 0.8819650 0.8865207 0.8956988
## 0.9124530 0.8844654 0.8284442 0.8660872 0.8835756 0.8862380 0.8946651
##
## 0.6422804 0.6592964 0.6674957 0.6544941 0.5938477 0.6117376 0.6144783
## 0.6990348 0.7185728 0.7159249 0.7094061 0.6364650 0.6676592 0.6683163
## 0.7276699 0.7500242 0.7490134 0.7533633 0.7171035 0.7285899 0.7300645
## 0.7632134 0.7809307 0.7750818 0.7791204 0.7455681 0.7560904 0.7548962
## 0.7614689 0.7841886 0.7802444 0.7788037 0.7337886 0.7600177 0.7601615
## 0.8711820 0.8841586 0.8843367 0.8712502 0.8226166 0.8513286 0.8461744
## 0.9051662 0.9208291 0.9224677 0.9121976 0.8563631 0.8942976 0.8929741
## 0.9223675 0.9342251 0.9392395 0.9283500 0.8736185 0.9108552 0.9092543
## 0.9205971 0.9218003 0.9198896 0.9069099 0.8293017 0.8877651 0.8817710
## 0.9124620 0.9067353 0.8942021 0.8685756 0.7673602 0.8374839 0.8272351
## 0.9396445 0.9309860 0.9167160 0.9002751 0.8233512 0.8753436 0.8641872
## 0.9607125 0.9490218 0.9344017 0.9173773 0.8366235 0.8971082 0.8819650
## 0.9807600 0.9629303 0.9438136 0.9295720 0.8558438 0.9061633 0.8865207
## 0.9859489 0.9679850 0.9472053 0.9384114 0.8656963 0.9174618 0.8956988
## 1.0000000 0.9929275 0.9806243 0.9723941 0.9046571 0.9520578 0.9366526
## 0.9929275 1.0000000 0.9931146 0.9879382 0.9209537 0.9679472 0.9586545
## 0.9806243 0.9931146 1.0000000 0.9919387 0.9260344 0.9766508 0.9708744
## 0.9723941 0.9879382 0.9919387 1.0000000 0.9434969 0.9902153 0.9860941
## 0.9046571 0.9209537 0.9260344 0.9434969 1.0000000 0.9514692 0.9488604
## 0.9520578 0.9679472 0.9766508 0.9902153 0.9514692 1.0000000 0.9962962
## 0.9366526 0.9586545 0.9708744 0.9860941 0.9488604 0.9962962 1.0000000
## 0.9357275 0.9573880 0.9677195 0.9836810 0.9458043 0.9915411 0.9971475
##
## 0.6123803
## 0.6722243
## 0.7314936
## 0.7588995
## 0.7613426
## 0.8505173
## 0.8979832
## 0.9124530
## 0.8844654
## 0.8284442

```

```
## 0.8660872
## 0.8835756
## 0.8862380
## 0.8946651
## 0.9357275
## 0.9573880
## 0.9677195
## 0.9836810
## 0.9458043
## 0.9915411
## 0.9971475
## 1.0000000
##
## R-sq. pooled: 0.7764
## Breusch-Pagan: 7014 p-value: ( 0)
```

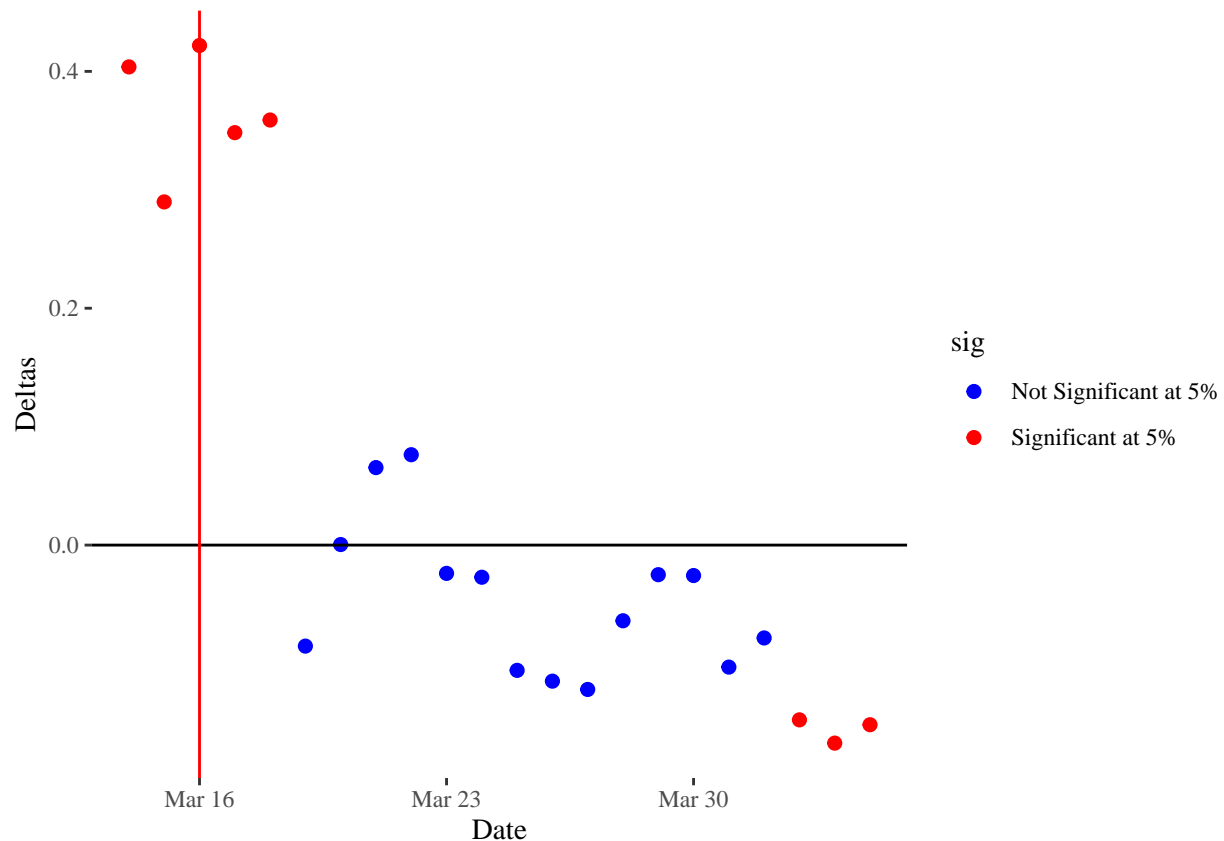
Compare goodness of fit:

```
data.frame(R2_lag8 = sur.slm_lag8$R2,
           R2_lag11 = sur.slm_lag11$R2,
           R2_lag11w = sur.slm_lag11w$R2) %>%
  slice(2:n()) %>%
  rownames_to_column(var = "Equation") %>%
  mutate(Date = seq(ymd("2020-03-14"),
                    ymd("2020-04-04"),
                    by = "days")) %>%
  pivot_longer(cols = starts_with("R"), names_to = "Model", values_to = "R2") %>%
  ggplot(aes(x = Date, y = R2, color = Model, shape = Model)) +
  geom_point(size = 2) +
  scale_color_manual(values = c("R2_lag11w" = "blue", "R2_lag11" = "orange", "R2_lag8" = "green")) +
  theme_tufte()
```



Dinámica temporal de la dependencia espacial (Coef lambda estimado en las T=22 ecuaciones)

```
data.frame(Date = seq(ymd("2020-03-14"),
                      ymd("2020-04-04"),
                      by = "days"),
           Deltas = sur.slm_lag11w$deltas,
           tvalue = sur.slm_lag11w$deltas/sur.slm_lag11$deltas.se) %>%
mutate(sig = ifelse(abs(tvalue) > 1.64, "Significant at 5%", "Not Significant at 5%")) %>%
ggplot(aes(x = Date, y = Deltas, color = sig)) +
geom_point(size = 2) +
scale_color_manual(values = c("Significant at 5%" = "red", "Not Significant at 5%" = "blue")) +
geom_hline(yintercept = 0) +
geom_vline(xintercept = as_date("2020-03-16"), color = "red") +
theme_tufte()
```



- La estructura de dependencia espacial, pasa de positiva (significativa) a no significativa para finalmente ser negativa (significativa)
- Esto puede explicarse:
- en una PRIMERA ETAPA al inicio del brote epidémico había contagio entre provincias colindantes.
- en una SEGUNDA ETAPA Al introducir medidas de confinamiento las provincias quedaron ‘aisladas’ y cada una creció a un ritmo diferente. La estructura de dependencia espacial se ‘disolvió’
- en una TERCERA ETAPA, en muchas provincias se controló parcialmente la epidemia. Salvo en X-provincias que siguieron creciendo fuertemente. Eso dio lugar a patrones de dependencia espacial negativa.

Analysis of autocorrelated residuals

Extract residuals for March 14 and April 2, and compute autocorrelated residuals:

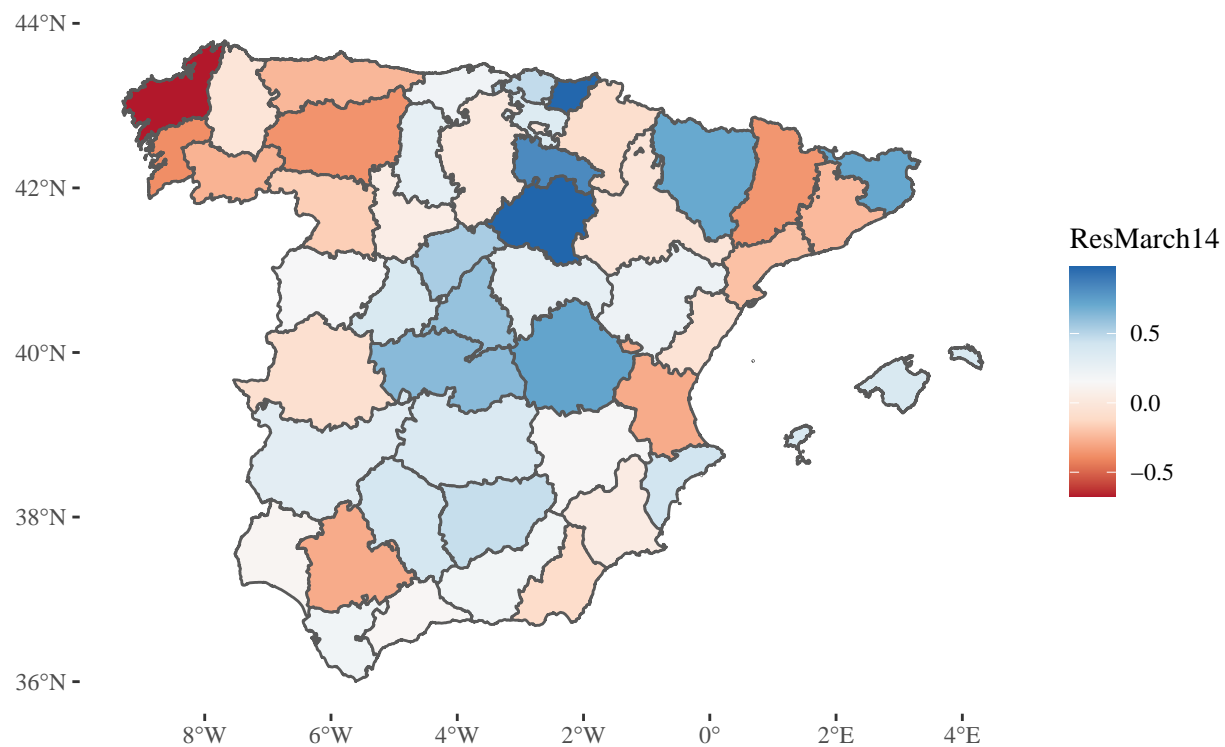
```
residuals_Mar14 <- as.matrix(residuals(sur.slm_lag11w))[[1]]
residuals_Mar14 <- lag.listw(listw, residuals_Mar14)

residuals_Apr04 <- as.matrix(residuals(sur.slm_lag11w))[[22]]
residuals_Apr04 <- lag.listw(listw, residuals_Apr04)
```

Plot residuals on March 14 (positive autocorrelation):

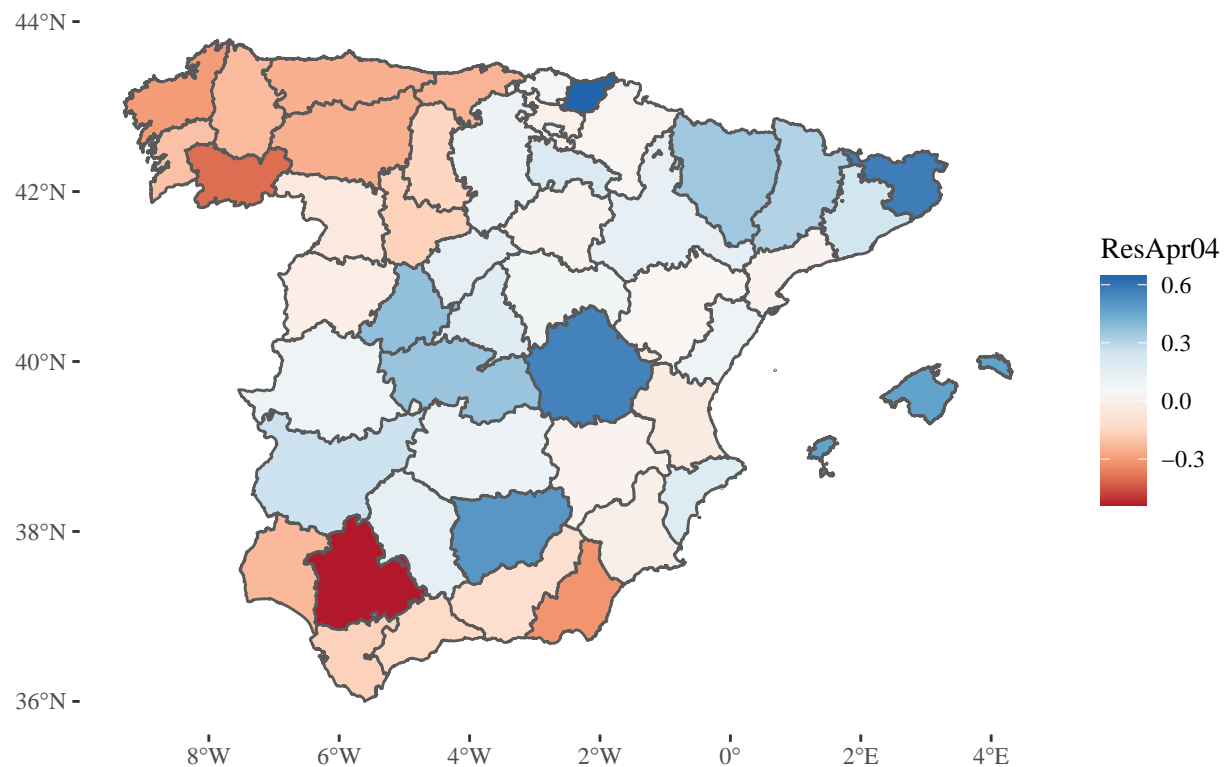
```
covid19_spain %>% filter(Date == "2020-03-14") %>%
  mutate(ResMarch14 = residuals_Mar14) %>%
  filter(CCAA != "Canarias") %>%
  ggplot() +
  geom_sf(aes(fill = ResMarch14)) +
```

```
scale_fill_distiller(palette = "RdBu", direction = 1) +
theme_tufte()
```



Plot residuals on April 4:

```
covid19_spain %>% filter(Date == "2020-03-14") %>%
  mutate(ResApr04 = residuals_Apr04) %>%
  filter(CCAA != "Canarias") %>%
  ggplot() +
  geom_sf(aes(fill = ResApr04)) +
  scale_fill_distiller(palette = "RdBu", direction = 1) +
  theme_tufte()
```

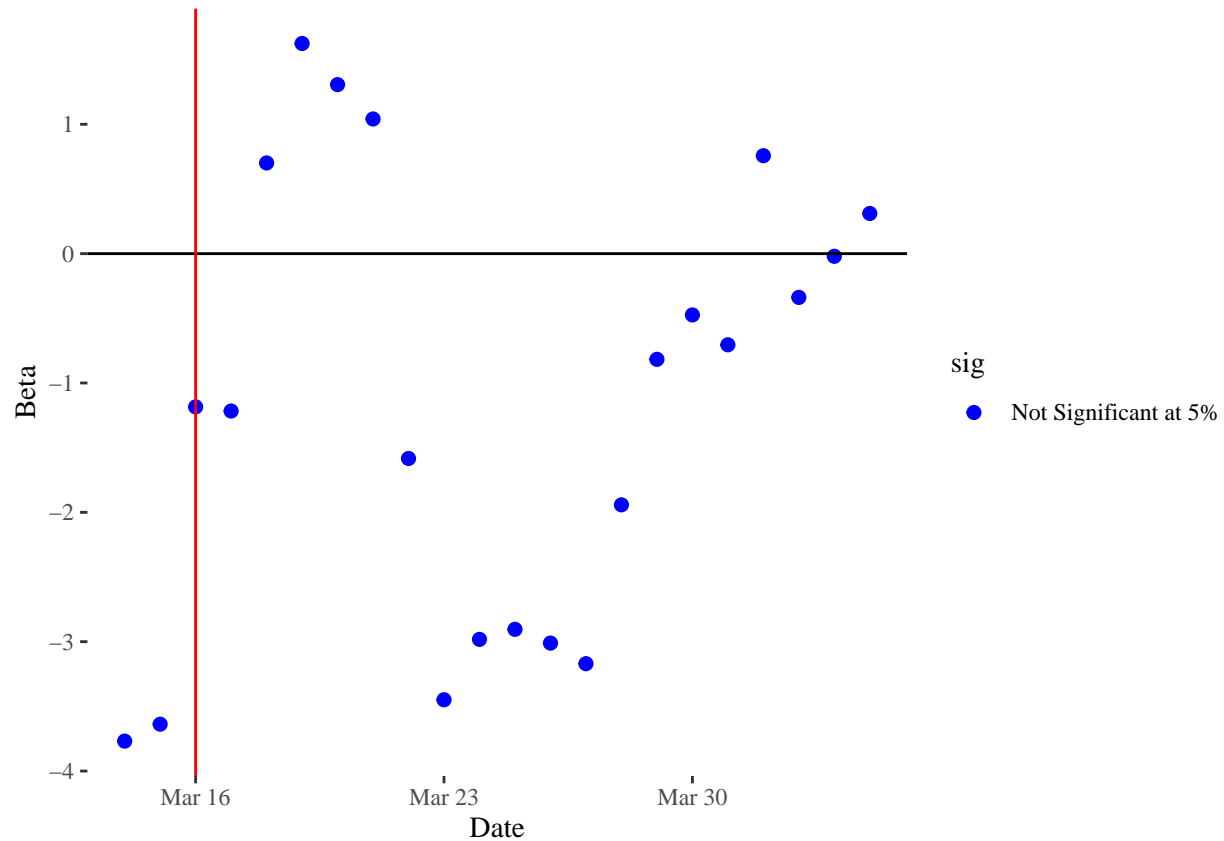



Temporal variation of coefficients of control variables

Male to Female Ratio

- No es significativa

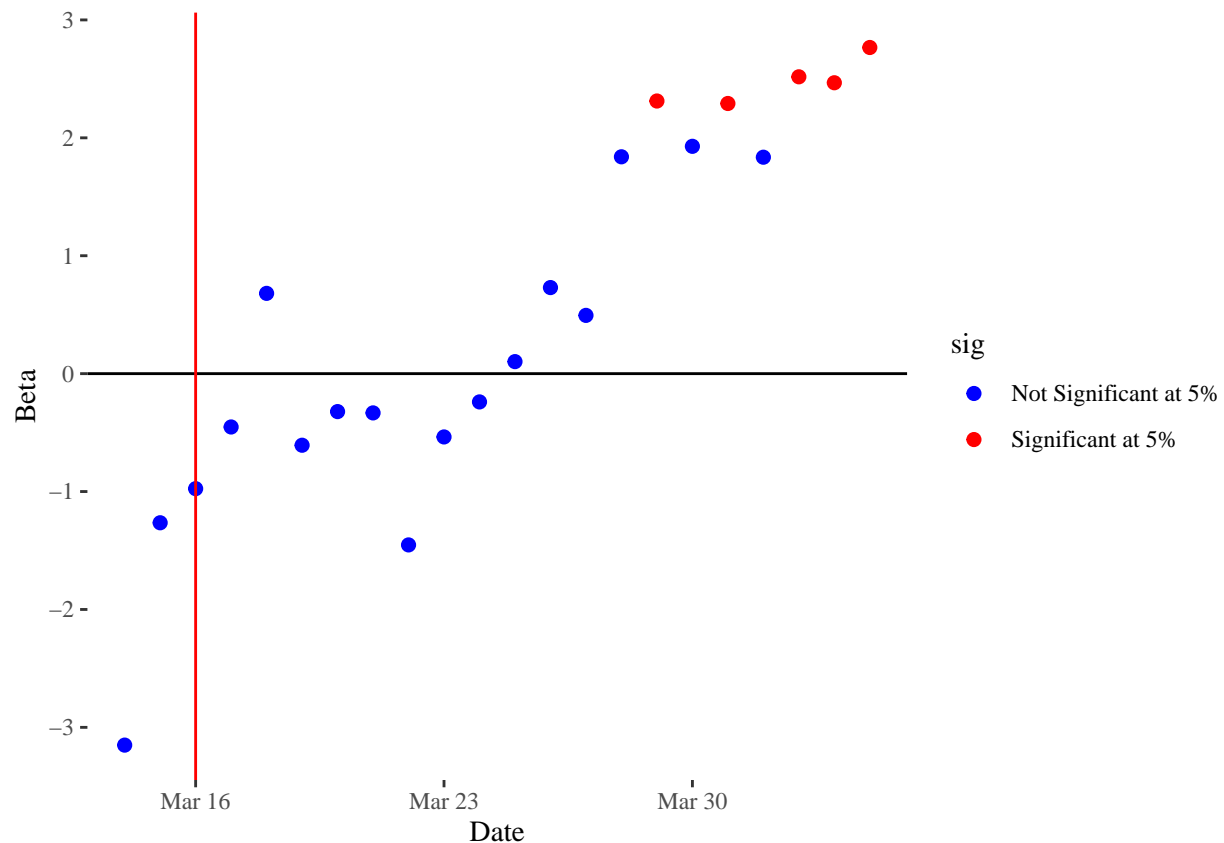
```
n = 2
data.frame(Date = seq(ymd("2020-03-14"),
                      ymd("2020-04-04"),
                      by = "days"),
            Beta = matrix(sur.slm_lag11w$coefficients, ncol = 22)[n,],
            tvalue = matrix(sur.slm_lag11w$coefficients/sur.slm_lag11w$rest.se, ncol = 22)[n,]) %>%
mutate(sig = ifelse(abs(tvalue) > 1.64, "Significant at 5%", "Not Significant at 5%")) %>%
ggplot(aes(x = Date, y = Beta, color = sig)) +
geom_point(size = 2) +
scale_color_manual(values = c("Significant at 5%" = "red", "Not Significant at 5%" = "blue")) +
geom_hline(yintercept = 0) +
geom_vline(xintercept = as_date("2020-03-16"), color = "red") +
theme_tufte()
```



Median Age

- Los coeficientes de las variables de control (EM y HM) tienen el signo esperado
- Los primeros días tienen valores elevados cuando la pandemia no estaba controlada (aún no había confinamiento)

```
n = 3
data.frame(Date = seq(ymd("2020-03-14"),
                      ymd("2020-04-04"),
                      by = "days"),
            Beta = matrix(sur.slm_lag11w$coefficients, ncol = 22)[n,],
            tvalue = matrix(sur.slm_lag11w$coefficients/sur.slm_lag11w$rest.se, ncol = 22)[n,]) %>%
mutate(sig = ifelse(abs(tvalue) > 1.64, "Significant at 5%", "Not Significant at 5%")) %>%
ggplot(aes(x = Date, y = Beta, color = sig)) +
geom_point(size = 2) +
scale_color_manual(values = c("Significant at 5%" = "red", "Not Significant at 5%" = "blue")) +
geom_hline(yintercept = 0) +
geom_vline(xintercept = as_date("2020-03-16"), color = "red") +
theme_tufte()
```

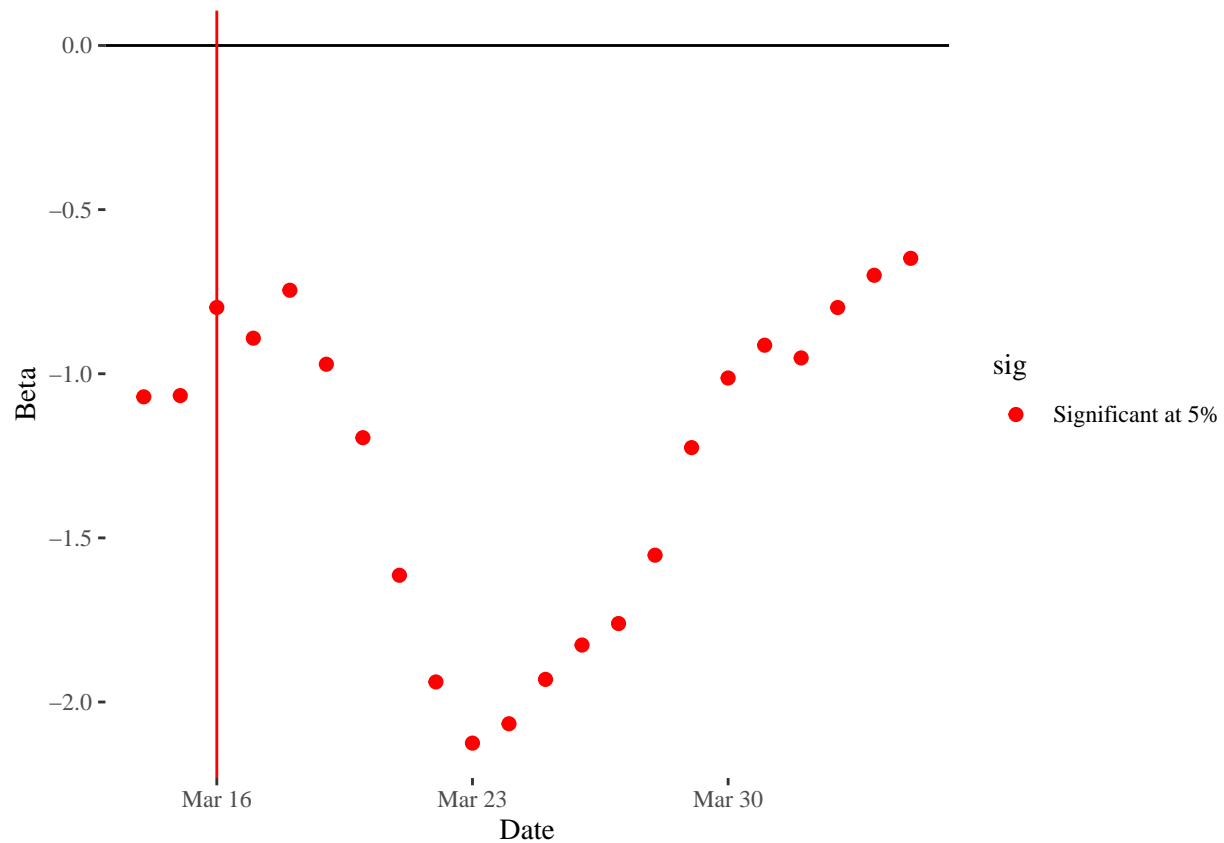


Dinámica temporal de Variables Climáticas

TEMPERATURA

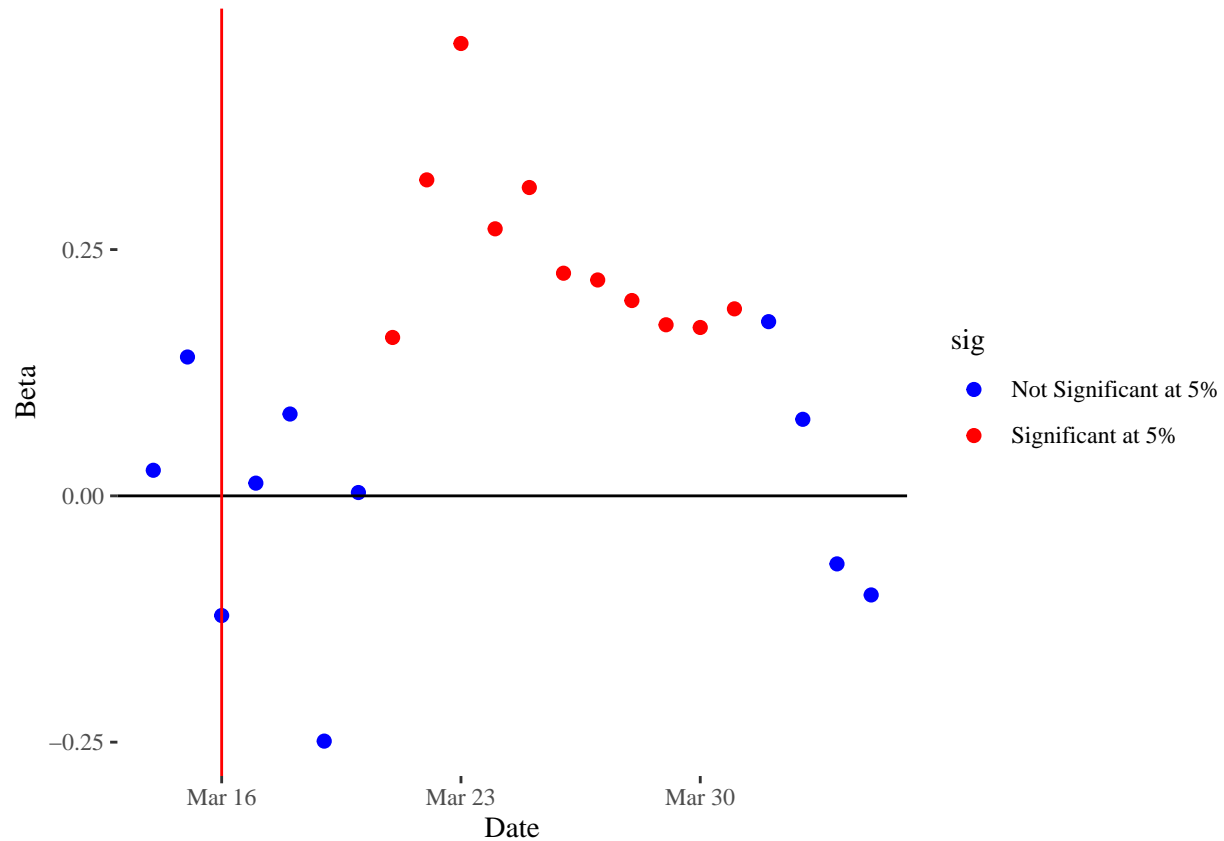
- Los coeficientes de las variables de control (EM y HM) tienen el signo esperado
- Los primeros días tienen valores elevados cuando la pandemia no estaba controlada (aún no había confinamiento)

```
n = 4
data.frame(Date = seq(ymd("2020-03-14"),
                      ymd("2020-04-04"),
                      by = "days"),
            Beta = matrix(sur.slm_lag11w$coefficients, ncol = 22)[n,],
            tvalue = matrix(sur.slm_lag11w$coefficients/sur.slm_lag11w$rest.se, ncol = 22)[n,]) %>%
  mutate(sig = ifelse(abs(tvalue) > 1.64, "Significant at 5%", "Not Significant at 5%")) %>%
  ggplot(aes(x = Date, y = Beta, color = sig)) +
  geom_point(size = 2) +
  scale_color_manual(values = c("Significant at 5%" = "red", "Not Significant at 5%" = "blue")) +
  geom_hline(yintercept = 0) +
  geom_vline(xintercept = as_date("2020-03-16"), color = "red") +
  theme_tufte()
```



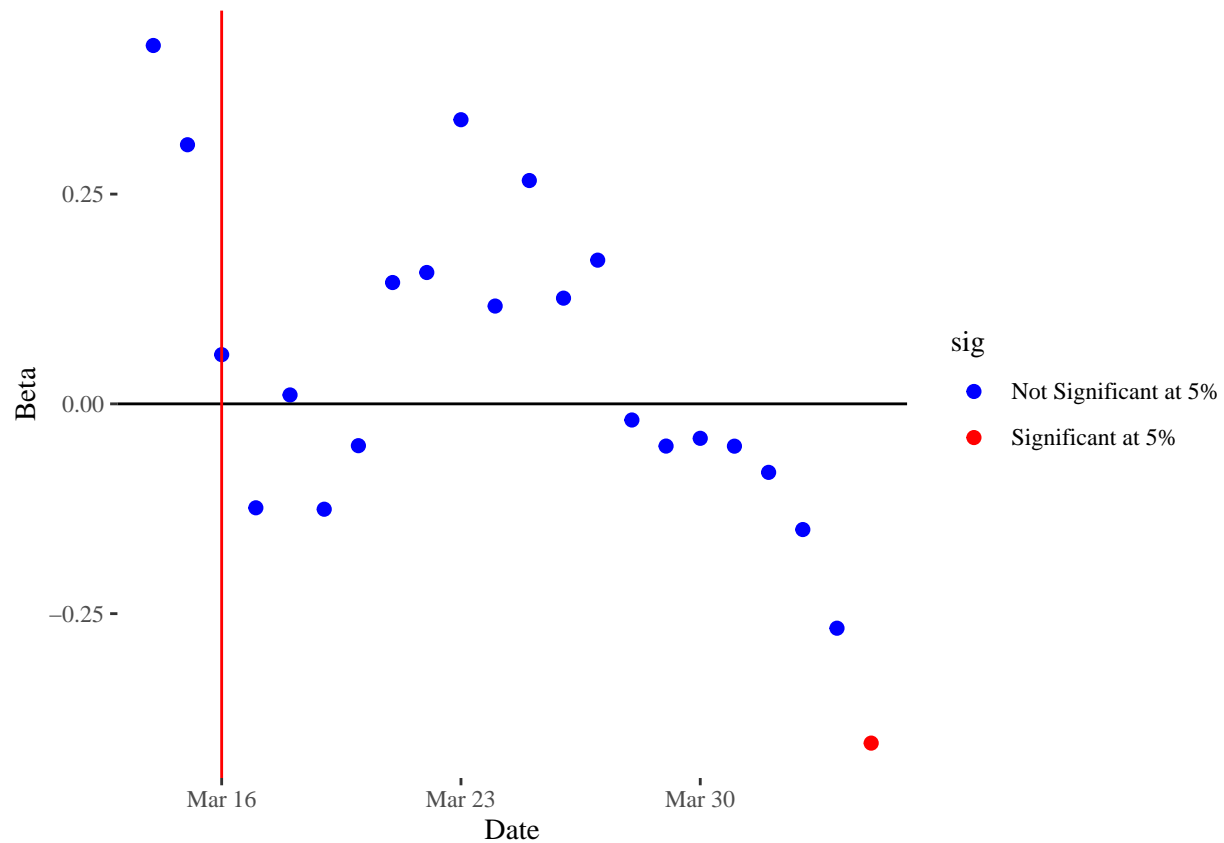
Hours of sunshine

```
n = 5
data.frame(Date = seq(ymd("2020-03-14"),
                      ymd("2020-04-04"),
                      by = "days"),
            Beta = matrix(sur.slm_lag11w$coefficients, ncol = 22)[n,],
            tvalue = matrix(sur.slm_lag11w$coefficients/sur.slm_lag11w$rest.se, ncol = 22)[n,]) %>%
  mutate(sig = ifelse(abs(tvalue) > 1.64, "Significant at 5%", "Not Significant at 5%")) %>%
  ggplot(aes(x = Date, y = Beta, color = sig)) +
  geom_point(size = 2) +
  scale_color_manual(values = c("Significant at 5%" = "red", "Not Significant at 5%" = "blue")) +
  geom_hline(yintercept = 0) +
  geom_vline(xintercept = as_date("2020-03-16"), color = "red") +
  theme_tufte()
```



Humidity

```
n = 6
data.frame(Date = seq(ymd("2020-03-14"),
                      ymd("2020-04-04"),
                      by = "days"),
            Beta = matrix(sur.slm_lag11w$coefficients, ncol = 22)[n,],
            tvalue = matrix(sur.slm_lag11w$coefficients/sur.slm_lag11w$rest.se, ncol = 22)[n,]) %>%
mutate(sig = ifelse(abs(tvalue) > 1.64, "Significant at 5%", "Not Significant at 5%")) %>%
ggplot(aes(x = Date, y = Beta, color = sig)) +
geom_point(size = 2) +
scale_color_manual(values = c("Significant at 5%" = "red", "Not Significant at 5%" = "blue")) +
geom_hline(yintercept = 0) +
geom_vline(xintercept = as_date("2020-03-16"), color = "red") +
theme_tufte()
```



Intercept

```
n = 1
data.frame(Date = seq(ymd("2020-03-14"),
                      ymd("2020-04-04"),
                      by = "days"),
            Beta = matrix(sur.slm_lag11w$coefficients, ncol = 22)[n,],
            tvalue = matrix(sur.slm_lag11w$coefficients/sur.slm_lag11w$rest.se, ncol = 22)[n,]) %>%
  mutate(sig = ifelse(abs(tvalue) > 1.64, "Significant at 5%", "Not Significant at 5%")) %>%
  ggplot(aes(x = Date, y = Beta, color = sig)) +
  geom_point(size = 2) +
  scale_color_manual(values = c("Significant at 5%" = "red", "Not Significant at 5%" = "blue")) +
  geom_hline(yintercept = 0) +
  geom_vline(xintercept = as_date("2020-03-16"), color = "red") +
  theme_tufte()
```

