Mi proyecto Subtítulo Subtítulo

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Mi resumen

Keywords: palabra clave 1, palabra clave 2

1 Introducción

Cuál es la distribución espacial de los puntos de calor - Hot Spots para los incendios forestale

Para él o las áreas con mayor índice de influencia.

```
Qué relación existe entre este resultado y las elevaciones del área?
Qué relación existe entre este resultado y las pendientes del área?
Qué relación existe entre este resultado y la cobertura arbórea del área?
Qué relación existe entre este resultado y el uso de suelo del área?
Qué relación existe entre este resultado y la cobertura agrícola del área?
```

Data a usar

Para el desarrollo de este trabajo se estará usando la siguiente información

Incendios forestales "https://firms.modaps.eosdis.nasa.gov/download/create.php" (Información de Variables climaticas "http://www.worldclim.org/" (10m)

Temperatura Minima (C°)
Temperatura Maxima (C°)
Temperatura Promedio (C°)
Precipitacion (mm)
Radiación Solar (Kj m-2 day-1)
Velocidad del viento (m s-1)
Presion de Vapor de agua (KPa)

Es de resaltar que para las siguiente data se realizará una selección basada en los resultados del análisis de hot-spots realizado.

```
Elevaciones "https://www.eorc.jaxa.jp/ALOS/en/aw3d30/index.htm"

Pendientes -> Se realizará un análisis de pendientes basado el modelo DEM (Digital Elevation Mod Cobertura arbórea http://earthenginepartners.appspot.com/science-2013-global-forest

Uso de suelo http://due.esrin.esa.int/page_globcover.php

Cobertura agrícola "Por definir"
```

```
2 Metodología
3 Resultados
4 Discusión o Conclusiones
5 Información de soporte
6 Script reproducible
library(sf)
## Linking to GEOS 3.7.1, GDAL 2.4.2, PROJ 5.2.0
library(sp)
library(raster)
library(tidyverse)
## -- Attaching packages -----
## v ggplot2 3.2.1 v purrr
                                 0.3.3
## v tibble 2.1.3 v dplyr 0.8.3
## v tidyr 1.0.0 v stringr 1.4.0
## v readr 1.3.1 v forcats 0.4.0
## -- Conflicts ------
## x tidyr::extract() masks raster::extract()
## x dplyr::filter() masks stats::filter()
## x dplyr::lag() masks stats::lag()
## x dplyr::select() masks raster::select()
library(tmap)
```

Loading required package: zoo

library(RColorBrewer)

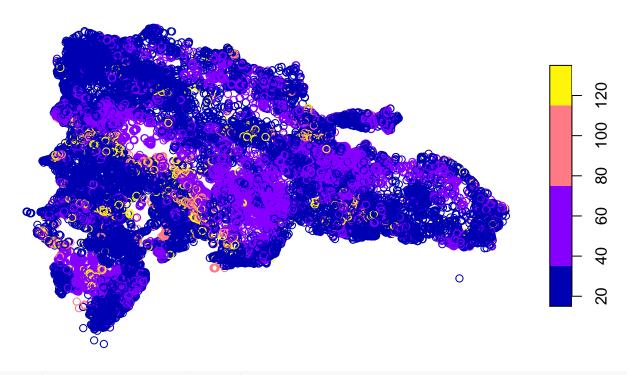
library(lmtest)

```
##
## Attaching package: 'zoo'
## The following objects are masked from 'package:base':
##
##
       as.Date, as.Date.numeric
library(spdep)
## Loading required package: spData
## To access larger datasets in this package, install the spDataLarge
## package with: 'install.packages('spDataLarge',
## repos='https://nowosad.github.io/drat/', type='source')'
#carga de capas y transformación
incendios <- st_read(dsn = "data/Incendios/FiredataBuffer.shp")</pre>
## Reading layer 'FiredataBuffer' from data source '/home/ingdi/unidad-0-asignacion-99-mi-proyec
## Simple feature collection with 66803 features and 17 fields
## geometry type:
                   POINT
## dimension:
                   XY
## bbox:
                   xmin: -71.97234 ymin: 17.66494 xmax: -68.34374 ymax: 19.92763
## epsg (SRID):
                   4326
## proj4string:
                   +proj=longlat +datum=WGS84 +no_defs
mun <- st_read(dsn = 'data/DivisionRD/divisionRD.gpkg', layer = 'MUNCENSO2010')</pre>
## Reading layer 'MUNCENSO2010' from data source '/home/ingdi/unidad-O-asignacion-99-mi-proyecto
## Simple feature collection with 155 features and 5 fields
## geometry type:
                   MULTIPOLYGON
## dimension:
                   XY
## bbox:
                   xmin: 182215.8 ymin: 1933532 xmax: 571365.3 ymax: 2205216
## epsg (SRID):
                   32619
                   +proj=utm +zone=19 +datum=WGS84 +units=m +no_defs
## proj4string:
mun4326 <- st_transform(mun, crs = 4326)</pre>
usoSuelo <- raster('data/UsoSuelo/GLOBCOVER_RD.color.tif')</pre>
#extracción de datos raster
incUsoSuelo <- raster::extract(usoSuelo, incendios,sp=TRUE)</pre>
incForestales <- subset(incUsoSuelo,incUsoSuelo$GLOBCOVER_RD.color>20 & incUsoSuelo$GLOBCOVER_RD
incForestales.df <- data.frame(incForestales)</pre>
summary(incForestales.df[[18]])
##
      Min. 1st Qu. Median
                              Mean 3rd Qu.
                                               Max.
##
     30.00
             30.00 30.00
                             37.91
                                      40.00 120.00
```

colnames(incForestales.df) ## [1] "LONGITUDE" "LATITUDE" "SYMBOL" ## [4] "ACQ_DATE" "ACQ_TIME" "BRIGHTNESS" "BRIGHT_TI5" ## [7] "BRIGHT_T31" "BRIGHT_TI4" ## [10] "CONFIDENCE" "CONFIDEN_1" "INSTRUMENT" "SATELLITE" "SCAN" ## [13] "LAYER" ## [16] "TRACK" "VERSION" "GLOBCOVER_RD.color" ## [19] "coords.x1" "coords.x2" "optional" incForestales.sf <- st_as_sf(incForestales)</pre> incendiosForestales <- st_intersection(incForestales.sf,mun4326)</pre> ## although coordinates are longitude/latitude, st_intersection assumes that they are planar ## Warning: attribute variables are assumed to be spatially constant ## throughout all geometries

GLOBCOVER_RD.color

plot(incendiosForestales['GLOBCOVER_RD.color'])



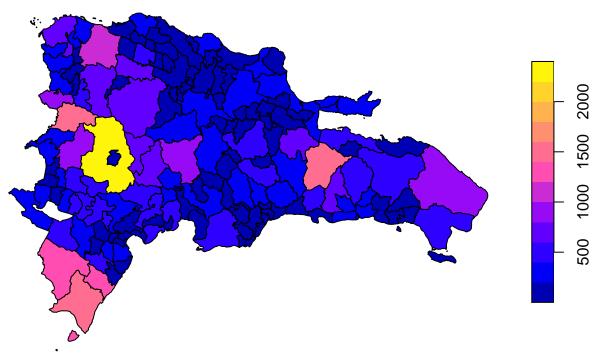
table(incendiosForestales\$ENLACE)

##
010901 010902 010903 010904 011801 011802 011803 011804 011805 011806

```
270
##
       52
               4
                     199
                             31
                                     76
                                            30
                                                     5
                                                           58
                                                                   22
## 011807 011808 011809 012501 012502 012503 012504 012505 012506 012507
##
      230
             129
                     102
                            197
                                     62
                                           187
                                                     4
                                                          783
                                                                   44
                                                                          23
## 012508 012509 021301 021302 021303 021304 022401 022402 022403 022404
##
              26
                     398
                            925
                                    223
                                           115
                                                   495
                                                          180
                                                                   44
                                                                          47
        7
## 022801 022802 022803 030601 030602 030603 030604 030605 030606 030607
##
      360
              31
                     187
                            363
                                    101
                                            15
                                                    49
                                                          170
## 031401 031402 031403 031404 031901 031902 031903 032001 032002 032003
              47
                     107
                            223
                                     60
                                            45
                                                    14
                                                          356
                                                                  338
                                                                         236
## 040501 040502 040503 040504 040505 041501 041502 041503 041504 041505
      224
             371
                     193
                            804
                                     87
                                           634
                                                   284
                                                         1158
                                                                         280
##
                                                                  139
## 041506 042601 042602 042603 042701 042702 042703 050201 050202 050203
      213
             747
                     418
                            130
                                    616
                                           418
                                                   150
                                                          518
                                                                   68
##
                                                                         412
## 050204 050205 050206 050207 050208 050209 050210 051701 051702 052101
             262
                      72
                                    462
                                                   200
                                                          572
##
      621
                              25
                                           369
                                                                         185
## 052102 052103 052104 052105 052106 052107 052108 053101 053102 053103
##
       23
              22
                      42
                            337
                                    176
                                            70
                                                   104
                                                          285
                                                                         104
## 060301 060302 060303 060304 060305 060401 060402 060403 060404 060405
##
      555
             474
                     341
                            475
                                    247
                                           349
                                                   290
                                                         1325
                                                                  126
                                                                         127
## 060406 060407 060408 060409 060410 060411 061001 061002 061003 061004
##
       42
             146
                      84
                             36
                                    551
                                           128
                                                   267
                                                          423
                                                                  213
                                                                         150
## 061005 061006 061601 061602 070701 070702 070703 070704 070705 070706
                    1359
             308
                           1536
                                    216
                                           265
                                                   102
                                                           83
                                                                 1587
## 072201 072202 072203 072204 072205 072206 080801 080802 081101 081102
##
     2221
             784
                     407
                            154
                                    901
                                           441
                                                   579
                                                          148
                                                                  931
                                                                         541
## 081201 081202 081203 092301 092302 092303 092304 092305 092306 092901
                     120
                            265
                                    797
                                                   231
                                                          223
##
       68
              85
                                           178
                                                                  235
                                                                         358
## 092902 092903 092904 092905 093001 093002 093003 100101 103201 103202
                                                    82
                                                            3
                                                                   73
     1512
             653
                     543
                             42
                                    512
                                           484
                                                                           7
##
## 103203 103204 103205 103206 103207
##
      241
             151
                     518
                             10
                                    376
```

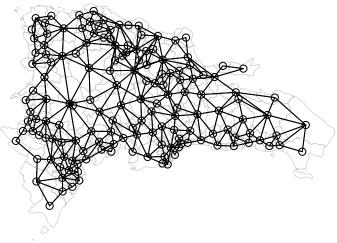
```
#Conversion a Poligono
munInc <-arrange(mun4326, ENLACE)
munInc$NumIncendios <- c(table(incendiosForestales$ENLACE))
plot(munInc['NumIncendios'])</pre>
```

NumIncendios



```
#Vecindad
munInc.sp <- as_Spatial(munInc)</pre>
colnames(munInc.sp@data)
## [1] "PROV"
                      "MUN"
                                      "REG"
                                                     "TOPONIMIA"
## [5] "ENLACE"
                      "NumIncendios"
row.names(munInc.sp) <- as.character(munInc.sp$TOPONIMIA)</pre>
munInc.nb <- poly2nb(munInc.sp, queen = TRUE)</pre>
summary(munInc.nb)
## Neighbour list object:
## Number of regions: 155
## Number of nonzero links: 804
## Percentage nonzero weights: 3.346514
## Average number of links: 5.187097
## Link number distribution:
## 1 2 3 4 5 6 7 8 9 10 11 12 14
## 1 10 20 34 33 22 13 13 4 1 1 2 1
## 1 least connected region:
## JUAN DE HERRERA with 1 link
## 1 most connected region:
## LA VEGA with 14 links
```

```
plot(munInc.sp, border="grey", lwd=0.5)
plot(munInc.nb, coordinates(munInc.sp), add=T)
```



```
#Num Vecinos
coords <- coordinates(munInc.sp)</pre>
ident <- row.names(munInc.sp)</pre>
munInc.nb.k1 <- knn2nb(knearneigh(coords, k = 1), row.names = ident)</pre>
summary(munInc.nb.k1)
## Neighbour list object:
## Number of regions: 155
## Number of nonzero links: 155
## Percentage nonzero weights: 0.6451613
## Average number of links: 1
## Non-symmetric neighbours list
## Link number distribution:
##
##
   1
## 155
## 155 least connected regions:
## MOCA CAYETANO GERMOSÉN GASPAR HERNÁNDEZ JAMAO AL NORTE PUERTO PLATA ALTAMIRA GUANANICO IMBERT
## 155 most connected regions:
## MOCA CAYETANO GERMOSÉN GASPAR HERNÁNDEZ JAMAO AL NORTE PUERTO PLATA ALTAMIRA GUANANICO IMBERT
card(munInc.nb.k1)
##
```

[141] 1 1 1 1 1 1 1 1 1 1 1 1 1 1 1

```
plot(munInc.sp, border="grey", lwd=0.5)
plot(munInc.nb.k1, coordinates(munInc.sp), add=T)

is.symmetric.nb(munInc.nb.k1)

## [1] FALSE

dist <- unlist(nbdists(munInc.nb.k1, coords))
```

Max.

Mean 3rd Qu.

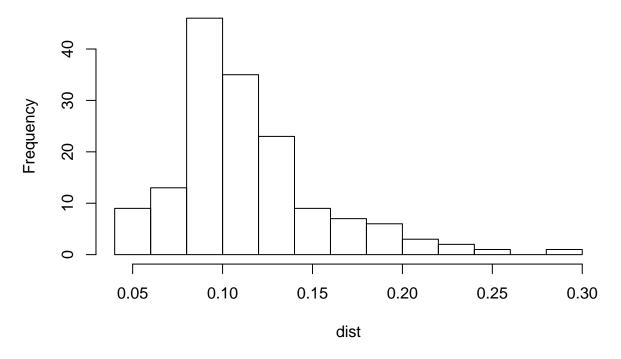
summary(dist)

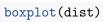
Min. 1st Qu. Median

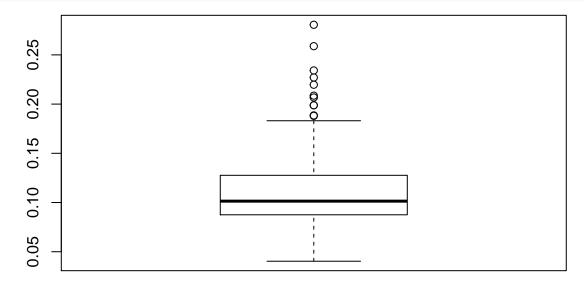
0.04036 0.08754 0.10140 0.11396 0.12765 0.28058

##

Histogram of dist







(distmin <- min(dist))</pre>

[1] 0.04036477

(distmax <- max(dist))</pre>

[1] 0.2805772

```
indicemin <- which(distmin==dist)</pre>
ident[indicemin]
## [1] "SAN JUAN"
                         "JUAN DE HERRERA"
indicemax <- which(distmax==dist)</pre>
ident[indicemax]
## [1] "HIGÜEY"
ident[order(dist)]
     [1] "SAN JUAN"
##
                                      "JUAN DE HERRERA"
##
     [3] "BAJOS DE HAINA"
                                      "SANTO DOMINGO OESTE"
     [5] "LICEY AL MEDIO"
                                      "PUÑAT."
##
##
     [7] "NIZAO"
                                      "SABANA GRANDE DE PALENQUE"
                                      "EL PEÑÓN"
     [9] "LOS ALCARRIZOS"
##
##
    [11] "FUNDACIÓN"
                                      "SANTO DOMINGO DE GUZMÁN"
## [13] "TAMBORIL"
                                      "YAGUATE"
## [15] "SABANA DE LA MAR"
                                      "EL VALLE"
## [17] "NEIBA"
                                      "VILLA JARAGUA"
## [19] "BISONÓ"
                                      "VILLA GONZÁLEZ"
## [21] "LOS RÍOS"
                                      "BARAHONA"
## [23] "POSTRER RÍO"
                                      "SALCEDO"
## [25] "TENARES"
                                      "SANTIAGO"
## [27] "AZUA"
                                      "PUEBLO VIEJO"
## [29] "GUANANICO"
                                      "IMBERT"
## [31] "SAN JOSÉ DE OCOA"
                                      "SABANA LARGA"
## [33] "SAN CRISTÓBAL"
                                      "SAN GREGORIO DE NIGUA"
## [35] "SABANA YEGUA"
                                      "TÁBARA ARRIBA"
## [37] "NAGUA"
                                      "EL FACTOR"
## [39] "YAMASÁ"
                                      "PERALVILLO"
## [41] "QUISQUEYA"
                                      "GUAYACANES"
## [43] "ESPERANZA"
                                      "VICENTE NOBLE"
                                      "LA CIÉNAGA"
## [45] "JAQUIMEYES"
## [47] "PIMENTEL"
                                      "LAS GUÁRANAS"
## [49] "HONDO VALLE"
                                      "JUAN SANTIAGO"
## [51] "FANTINO"
                                      "LA MATA"
                                      "EUGENIO MARÍA DE HOSTOS"
## [53] "CASTILLO"
## [55] "JIMA ABAJO"
                                      "CAYETANO GERMOSÉN"
## [57] "VILLA TAPIA"
                                      "ALTAMIRA"
## [59] "EL LLANO"
                                      "DAJABÓN"
## [61] "PEPILLO SALCEDO"
                                      "PARTIDO"
    [63] "EL PINO"
                                      "SABANA IGLESIA"
## [65] "CRISTÓBAL"
                                      "JAMAO AL NORTE"
## [67] "SOSÚA"
                                      "PARAÍSO"
```

"CABRAL"

[69] "VILLA LOS ALMÁCIGOS"

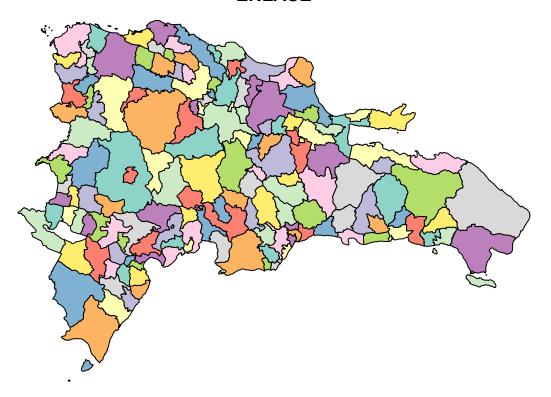
```
[71] "SAN PEDRO DE MACORÍS"
## [73] "PUERTO PLATA"
                                      "VILLA MONTELLANO"
## [75] "CASTAÑUELAS"
                                      "VILLA VÁSQUEZ"
## [77] "MAIMÓN"
                                      "PIEDRA BLANCA"
## [79] "LAS SALINAS"
                                      "MOCA"
## [81] "LA ROMANA"
                                      "GUAYMATE"
## [83] "LOS HIDALGOS"
                                      "GALVÁN"
                                      "POLO"
    [85] "CAMBITA GARABITOS"
## [87] "PEDRO BRAND"
                                      "LA DESCUBIERTA"
## [89] "LAGUNA SALADA"
                                      "EL CERCADO"
## [91] "VILLA HERMOSA"
                                      "MONTE CRISTI"
## [93] "COTUÍ"
                                      "LAS MATAS DE SANTA CRUZ"
## [95] "LA VEGA"
                                      "LOMA DE CABRERA"
## [97] "PERALTA"
                                      "GUAYABAL"
## [99] "LOS LLANOS"
                                      "ENRIQUILLO"
## [101] "VILLA ALTAGRACIA"
                                      "LOS CACAOS"
## [103] "LUPERÓN"
                                      "RAMÓN SANTANA"
## [105] "JÁNICO"
                                      "CABRERA"
## [107] "RÍO SAN JUAN"
                                      "RESTAURACIÓN"
## [109] "RANCHO ARRIBA"
                                      "ESTEBANÍA"
## [111] "ARENOSO"
                                      "VILLA RIVA"
## [113] "SÁNCHEZ"
                                      "MELLA"
## [115] "COMENDADOR"
                                      "BÁNICA"
## [117] "SANTO DOMINGO NORTE"
                                      "OAM"
## [119] "JIMANÍ"
                                      "BOCA CHICA"
## [121] "SAN ANTONIO DE GUERRA"
                                      "CEVICOS"
## [123] "VILLA ISABELA"
                                      "TAMAYO"
## [125] "LAS YAYAS DE VIAJAMA"
                                      "SAN IGNACIO DE SABANETA"
## [127] "GASPAR HERNÁNDEZ"
                                      "SANTO DOMINGO ESTE"
## [129] "LAS TERRENAS"
                                      "SAN FRANCISCO DE MACORÍS"
## [131] "DUVERGÉ"
                                      "MONCIÓN"
                                      "PADRE LAS CASAS"
## [133] "VALLEJUELO"
## [135] "BOHECHÍO"
                                      "HATO MAYOR"
## [137] "JARABACOA"
                                      "LAS MATAS DE FARFÁN"
## [139] "PEDRO SANTANA"
                                      "BONAO"
## [141] "LAS CHARCAS"
                                      "BANÍ"
                                      "CONSTANZA"
## [143] "SABANA GRANDE DE BOYÁ"
## [145] "GUAYUBÍN"
                                      "MONTE PLATA"
## [147] "EL SEIBO"
                                      "MICHES"
## [149] "OVIEDO"
                                      "SAN JOSÉ DE LAS MATAS"
                                      "SAMANÁ"
## [151] "BAYAGUANA"
## [153] "PEDERNALES"
                                      "SAN RAFAEL DEL YUMA"
## [155] "HIGÜEY"
#Ponderadoes espaciales
munInc.w.W <- nb2listw(munInc.nb)</pre>
```

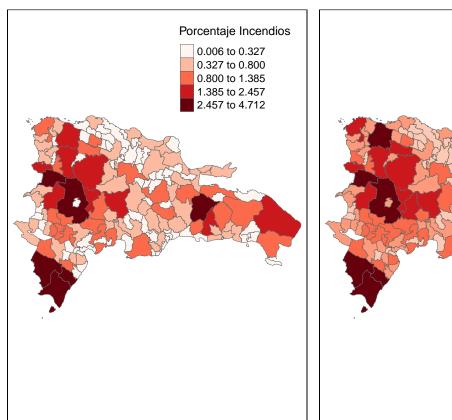
"CONSUELO"

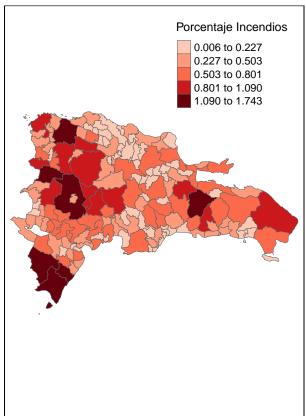
munInc.w.W

```
## Characteristics of weights list object:
## Neighbour list object:
## Number of regions: 155
## Number of nonzero links: 804
## Percentage nonzero weights: 3.346514
## Average number of links: 5.187097
##
## Weights style: W
## Weights constants summary:
      n
           nn S0
                         S1
                                  S2.
## W 155 24025 155 65.94606 650.7687
munInc.w.B <- nb2listw(munInc.nb, style = 'B')</pre>
munInc.w.B
## Characteristics of weights list object:
## Neighbour list object:
## Number of regions: 155
## Number of nonzero links: 804
## Percentage nonzero weights: 3.346514
## Average number of links: 5.187097
##
## Weights style: B
## Weights constants summary:
           nn S0
                     S1
## B 155 24025 804 1608 19520
#Correlacion Incendios
munIncPerc <- munInc %>%st_centroid() %>% mutate(
  'IncPercentage' = munInc$NumIncendios/sum(munInc$NumIncendios)*100,
  'IncPercentage_log' = log1p(munInc$NumIncendios/sum(munInc$NumIncendios)*100),
  'AreaKm2' = as.numeric((st_area(munInc)/1000)),
  'IncXArea' = (munInc$NumIncendios/AreaKm2),
  'IncXArea_log' = log1p(munInc$NumIncendios/AreaKm2),
 x=unlist(map(geom,1)), y=unlist(map(geom,2)))  %>%
 st_drop_geometry()
## 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
munIncPercPol <- munInc %>%
 merge(munIncPerc, all.y=TRUE)
plot(munIncPercPol['ENLACE'])
```

ENLACE

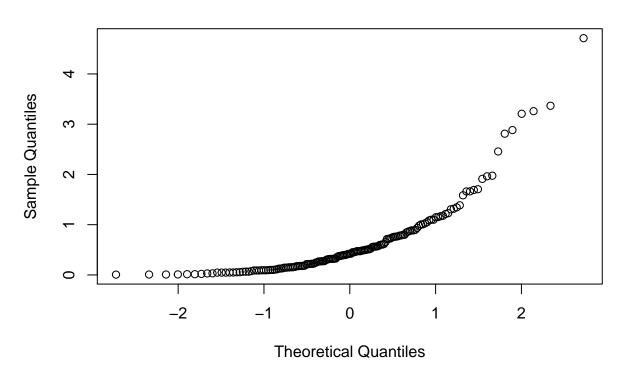






#qq
qqnorm(munIncPerc\$IncPercentage)

Normal Q-Q Plot

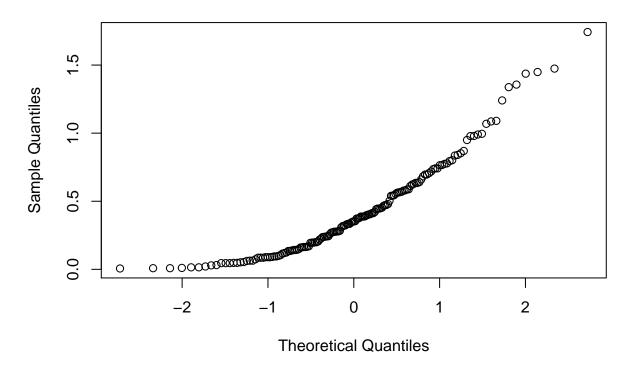


shapiro.test(munIncPerc\$IncPercentage)

```
##
## Shapiro-Wilk normality test
##
## data: munIncPerc$IncPercentage
## W = 0.74127, p-value = 3.202e-15
```

qqnorm(munIncPerc\$IncPercentage_log)

Normal Q-Q Plot



shapiro.test(munIncPerc\$IncPercentage_log)

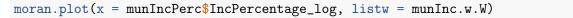
```
##
## Shapiro-Wilk normality test
##
## data: munIncPerc$IncPercentage_log
## W = 0.89833, p-value = 6.953e-09

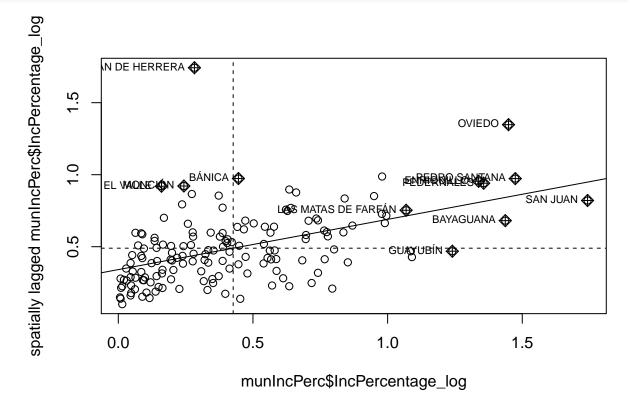
munIncPerc %>% lm(IncPercentage ~ x, .) %>% bptest()

##
## studentized Breusch-Pagan test
##
## data: .
## BP = 1.6534, df = 1, p-value = 0.1985
```

```
munIncPerc %>% lm(IncPercentage ~ y, .) %>% bptest()
##
## studentized Breusch-Pagan test
##
## data:
## BP = 1.1426, df = 1, p-value = 0.2851
munIncPerc %>% lm(IncPercentage_log ~ x, .) %>% bptest()
##
## studentized Breusch-Pagan test
##
## data:
## BP = 0.61098, df = 1, p-value = 0.4344
munIncPerc %>% lm(IncPercentage_log ~ y, .) %>% bptest()
##
## studentized Breusch-Pagan test
##
## data:
## BP = 1.3799, df = 1, p-value = 0.2401
match(attr(munInc.w.W$neighbours, "region.id"), munIncPerc$TOPONIMIA)==1:155
  ## [155] TRUE
(gmoranw <- moran.test(x = munIncPerc$'IncPercentage_log', listw = munInc.w.W ))</pre>
##
## Moran I test under randomisation
## data: munIncPerc$IncPercentage_log
## weights: munInc.w.W
```

```
##
## Moran I statistic standard deviate = 6.9272, p-value = 2.146e-12
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic
                           Expectation
                                                 Variance
##
         0.348674184
                          -0.006493506
                                              0.002628733
(gmoranb <- moran.test(x = munIncPerc$'IncPercentage_log', listw = munInc.w.B))</pre>
##
   Moran I test under randomisation
##
##
## data: munIncPerc$IncPercentage_log
## weights: munInc.w.B
##
## Moran I statistic standard deviate = 7.8512, p-value = 2.06e-15
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic
                           Expectation
                                                 Variance
##
         0.374437198
                          -0.006493506
                                              0.002354052
```



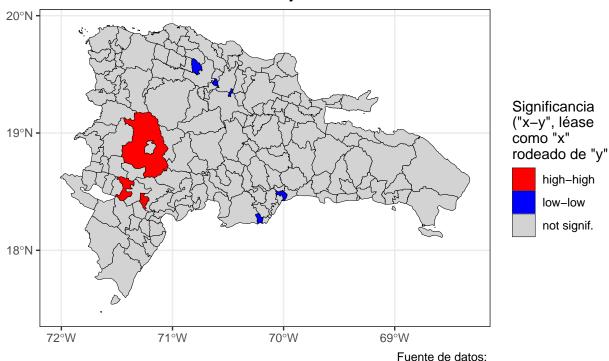


```
source('lisaclusters.R')
lisamap(objesp = munIncPercPol,
```

```
var = 'IncPercentage_log',
pesos = munInc.w.W,
tituloleyenda = 'Significancia\n("x-y", léase\ncomo "x"\nrodeado de "y"',
leyenda = T,
anchuratitulo = 1000,
tamanotitulo = 16,
fuentedatos = '',
titulomapa = pasteO('Clusters LISA de Porcentaje de Incendios Forestales'))
```

\$grafico

Clusters LISA de Porcentaje de Incendios Forestales

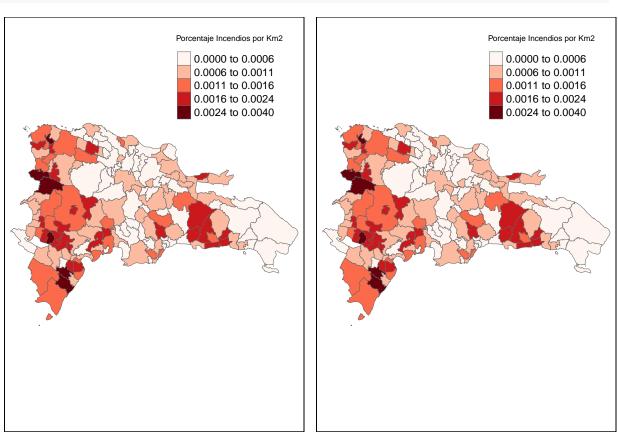


```
##
## $objeto
## Simple feature collection with 155 features and 16 fields
## geometry type:
                  MULTIPOLYGON
## dimension:
                  XΥ
                  xmin: -72.01147 ymin: 17.47033 xmax: -68.32354 ymax: 19.93211
## bbox:
## epsg (SRID):
                  4326
## proj4string:
                  +proj=longlat +datum=WGS84 +no_defs
## First 10 features:
##
      PROV MUN REG
                                 TOPONIMIA ENLACE NumIncendios IncPercentage
       01 01 10 SANTO DOMINGO DE GUZMÁN 100101
## 1
                                                                 0.006364562
                                      AZUA 050201
                                                                 1.098947726
## 2
       02 01
               05
                                                           518
                               LAS CHARCAS 050202
## 3
       02 02
               05
                                                           68
                                                                 0.144263408
                     LAS YAYAS DE VIAJAMA 050203
## 4
       02 03 05
                                                           412
                                                                 0.874066531
```

```
## 5
        02
            04
                05
                           PADRE LAS CASAS 050204
                                                            621
                                                                  1.317464358
## 6
        02
            05
                05
                                   PERALTA 050205
                                                            262
                                                                  0.555838425
## 7
        02
            06
                05
                              SABANA YEGUA 050206
                                                             72
                                                                  0.152749491
        02
            07
                05
                              PUEBLO VIEJO 050207
                                                             25
## 8
                                                                  0.053038018
## 9
        02
            80
                05
                             TÁBARA ARRIBA 050208
                                                            462
                                                                  0.980142566
                                                            369
## 10
        02
            09
                05
                                  GUAYABAL 050209
                                                                  0.782841141
##
      IncPercentage_log
                          AreaKm2
                                       IncXArea IncXArea_log
## 1
            0.006344394 91496.44 3.278816e-05 3.278762e-05 -69.94175
## 2
            0.741436136 416240.47 1.244473e-03 1.243699e-03 -70.80988
## 3
            0.134761118 246535.55 2.758223e-04 2.757843e-04 -70.54611
## 4
            0.628110685 430819.68 9.563166e-04 9.558596e-04 -70.99966
            0.840473639 573560.72 1.082710e-03 1.082124e-03 -70.91135
## 5
            0.442014580 129359.98 2.025356e-03 2.023308e-03 -70.78275
## 6
            0.142149950 113903.67 6.321131e-04 6.319134e-04 -70.88809
## 7
## 8
            0.051679337 48080.24 5.199641e-04 5.198289e-04 -70.78029
## 9
            0.683168845 274864.78 1.680827e-03 1.679415e-03 -70.91510
## 10
            0.578208238 235349.38 1.567882e-03 1.566654e-03 -70.76497
##
                                     geometry puntuacionz lagpuntuacionz
      18.48488 MULTIPOLYGON (((-69.89794 1... -1.19133383
## 1
                                                               0.35324679
     18.42093 MULTIPOLYGON (((-70.71457 1... 0.89259640
                                                              -0.55810348
## 3
     18.39184 MULTIPOLYGON (((-70.50185 1... -0.82728200
                                                               0.35821362
      18.60405 MULTIPOLYGON (((-70.85774 1... 0.57132720
                                                              -0.74175652
     18.80900 MULTIPOLYGON (((-70.77551 1... 1.17336033
                                                               0.21171873
## 6
     18.60198 MULTIPOLYGON (((-70.73131 1... 0.04375864
                                                              -0.10336069
     18.41374 MULTIPOLYGON (((-70.83014 1... -0.80633521
                                                              -0.22051577
      18.39309 MULTIPOLYGON (((-70.79387 1... -1.06281265
                                                               0.31361084
      18.49605 MULTIPOLYGON (((-70.83352 1... 0.72741298
                                                              -0.09605209
## 10 18.71745 MULTIPOLYGON (((-70.68664 1... 0.42985749
                                                               0.45866074
##
         quad_sig
     not signif.
## 1
## 2
     not signif.
## 3
     not signif.
## 4
     not signif.
## 5
    not signif.
## 6
     not signif.
## 7
     not signif.
## 8
     not signif.
## 9
     not signif.
## 10 not signif.
# Mapa Porcentaje por Km2
p3 <- tm_shape(munIncPercPol) +
  tm_fill(col = "IncXArea", style = 'jenks',
          palette = brewer.pal(9, name = 'Reds'), title = 'Porcentaje Incendios por Km2') +
  tm_borders(lwd = 0.5)
p4 <- tm_shape(munIncPercPol) +
  tm_fill(col = "IncXArea_log", style = 'jenks',
```

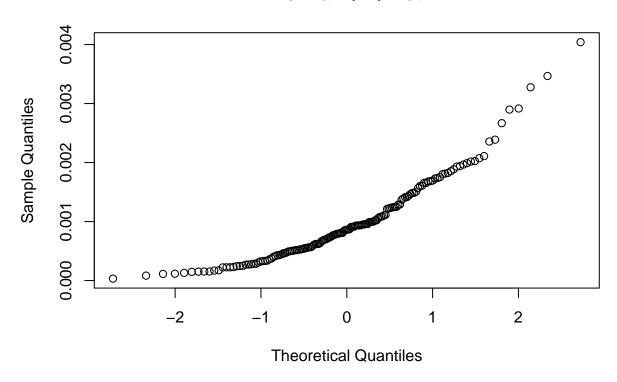
```
palette = brewer.pal(9, name = 'Reds'), title = 'Porcentaje Incendios por Km2') +
tm_borders(lwd = 0.5)

tmap_arrange(p3,p4)
```



#qq
qqnorm(munIncPerc\$IncXArea)

Normal Q-Q Plot

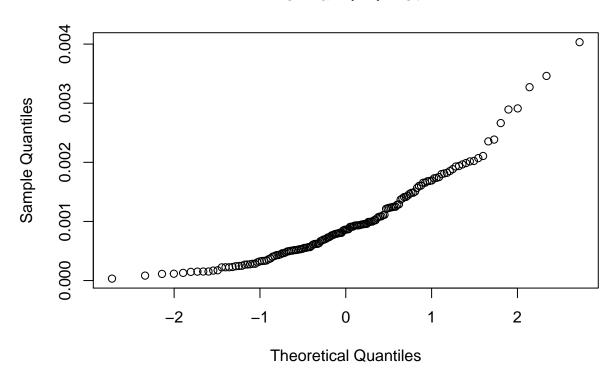


shapiro.test(munIncPerc\$IncXArea)

```
##
## Shapiro-Wilk normality test
##
## data: munIncPerc$IncXArea
## W = 0.89846, p-value = 7.076e-09
```

qqnorm(munIncPerc\$IncXArea_log)

Normal Q-Q Plot



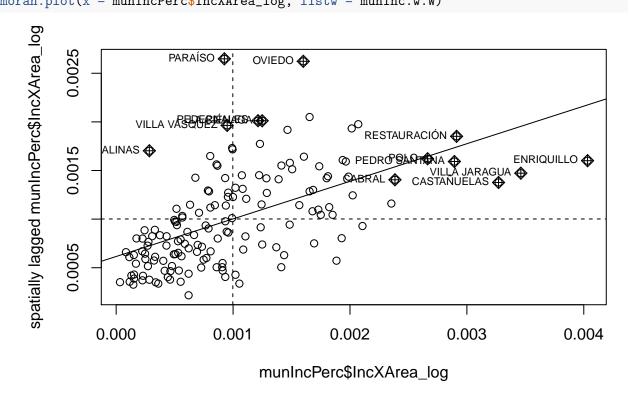
shapiro.test(munIncPerc\$IncXArea_log)

BP = 0.41244, df = 1, p-value = 0.5207

```
##
   Shapiro-Wilk normality test
##
## data: munIncPerc$IncXArea_log
## W = 0.89875, p-value = 7.342e-09
munIncPerc %>% lm(IncXArea ~ x, .) %>% bptest()
##
##
   studentized Breusch-Pagan test
##
## data:
## BP = 3.3994, df = 1, p-value = 0.06522
munIncPerc %>% lm(IncXArea ~ y, .) %>% bptest()
##
   studentized Breusch-Pagan test
##
##
## data:
```

```
munIncPerc %>% lm(IncXArea_log ~ x, .) %>% bptest()
##
##
 studentized Breusch-Pagan test
##
## data:
## BP = 3.3927, df = 1, p-value = 0.06548
munIncPerc %>% lm(IncXArea_log ~ y, .) %>% bptest()
##
## studentized Breusch-Pagan test
##
## data:
## BP = 0.41022, df = 1, p-value = 0.5219
match(attr(munInc.w.W$neighbours, "region.id"), munIncPerc$TOPONIMIA)==1:155
  ##
 ## [155] TRUE
(gmoranw <- moran.test(x = munIncPerc$'IncXArea_log', listw = munInc.w.W))
##
## Moran I test under randomisation
## data: munIncPerc$IncXArea_log
## weights: munInc.w.W
## Moran I statistic standard deviate = 7.6926, p-value = 7.207e-15
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic
            Expectation
                      Variance
##
    0.386575214
            -0.006493506
                     0.002610878
```

```
(gmoranb <- moran.test(x = munIncPerc$'IncXArea_log', listw = munInc.w.B))</pre>
##
##
   Moran I test under randomisation
##
## data: munIncPerc$IncXArea_log
## weights: munInc.w.B
##
## Moran I statistic standard deviate = 7.97, p-value = 7.932e-16
## alternative hypothesis: greater
## sample estimates:
## Moran I statistic
                           Expectation
                                                 Variance
                          -0.006493506
##
         0.378897197
                                              0.002338207
moran.plot(x = munIncPerc$IncXArea_log, listw = munInc.w.W)
```

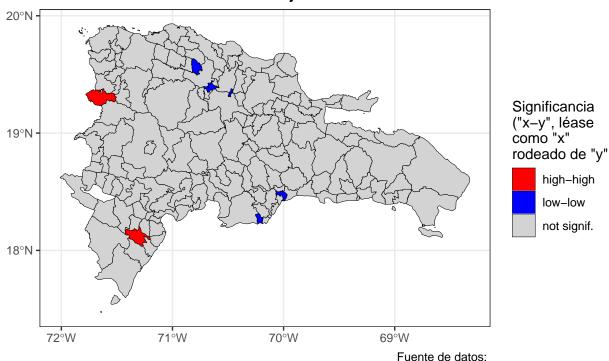


\$grafico

2

3

Clusters LISA de Porcentaje de Incendios Forestales



\$objeto ## Simple feature collection with 155 features and 16 fields ## geometry type: MULTIPOLYGON ## dimension: ## bbox: xmin: -72.01147 ymin: 17.47033 xmax: -68.32354 ymax: 19.93211 ## epsg (SRID): 4326 ## proj4string: +proj=longlat +datum=WGS84 +no_defs ## First 10 features: ## PROV MUN REG TOPONIMIA ENLACE NumIncendios IncPercentage ## 1 10 SANTO DOMINGO DE GUZMÁN 100101 01 01 3 0.006364562 ## 2 02 01 05 AZUA 050201 518 1.098947726 ## 3 02 02 05 LAS CHARCAS 050202 68 0.144263408 05 LAS YAYAS DE VIAJAMA 050203 412 ## 4 02 03 0.874066531 ## 5 02 04 05 PADRE LAS CASAS 050204 621 1.317464358 ## 6 02 05 05 PERALTA 050205 262 0.555838425 ## 7 05 SABANA YEGUA 050206 72 02 06 0.152749491 ## 8 02 05 PUEBLO VIEJO 050207 25 0.053038018 07 ## 9 02 80 05 TÁBARA ARRIBA 050208 462 0.980142566 ## 10 09 05 GUAYABAL 050209 369 02 0.782841141 IncPercentage_log ## AreaKm2 IncXArea IncXArea_log ## 1 0.006344394 91496.44 3.278816e-05 3.278762e-05 -69.94175

0.741436136 416240.47 1.244473e-03 1.243699e-03 -70.80988

0.134761118 246535.55 2.758223e-04 2.757843e-04 -70.54611

```
## 4
            0.628110685 430819.68 9.563166e-04 9.558596e-04 -70.99966
## 5
            0.840473639 573560.72 1.082710e-03 1.082124e-03 -70.91135
## 6
            0.442014580 129359.98 2.025356e-03 2.023308e-03 -70.78275
## 7
            0.142149950 113903.67 6.321131e-04 6.319134e-04 -70.88809
            0.051679337 48080.24 5.199641e-04 5.198289e-04 -70.78029
## 8
## 9
            0.683168845 274864.78 1.680827e-03 1.679415e-03 -70.91510
## 10
            0.578208238 235349.38 1.567882e-03 1.566654e-03 -70.76497
##
                                     geometry puntuacionz lagpuntuacionz
             у
     18.48488 MULTIPOLYGON (((-69.89794 1... -1.34612576
## 1
                                                              0.66116666
     18.42093 MULTIPOLYGON (((-70.71457 1... 0.34298868
                                                              0.29866905
     18.39184 MULTIPOLYGON (((-70.50185 1... -1.00716698
## 3
                                                              0.26239636
## 4
     18.60405 MULTIPOLYGON (((-70.85774 1... -0.05852211
                                                             -0.81213107
     18.80900 MULTIPOLYGON (((-70.77551 1... 0.11760616
## 5
                                                              0.62767676
     18.60198 MULTIPOLYGON (((-70.73131 1... 1.43047345
## 6
                                                              0.26522721
     18.41374 MULTIPOLYGON (((-70.83014 1... -0.51039837
## 7
                                                              0.39803286
     18.39309 MULTIPOLYGON (((-70.79387 1... -0.66674625
                                                              0.55638749
     18.49605 MULTIPOLYGON (((-70.83352 1... 0.95077431
                                                              0.06916901
## 10 18.71745 MULTIPOLYGON (((-70.68664 1... 0.79348187
                                                              0.43751696
##
         quad_sig
## 1 not signif.
## 2 not signif.
## 3 not signif.
## 4 not signif.
## 5 not signif.
## 6 not signif.
## 7 not signif.
## 8 not signif.
## 9
     not signif.
## 10 not signif.
```

7 Referencias