

Ecu_RK_3b.R

Antares

2019-04-10

```
#####  
# PROCEDIMIENTO PARA ESTIMAR EL CARBONO ORGANICO EN LOS SUELOS=====  
# ===== DE ECUADOR (KG/M2) Y (TN/HA) =====  
#####  
  
# MODELO EMPLEADO REGRESION - KRIGING.  
# CANTIDAD DE PERFILES DE SUELOS PARA CALIBRACION:12924.  
# CANTIDAD DE PERFILES DE SUELOS DEJADOS PARA VALIDACIO: 1000.  
  
# Establecemos el directorio de trabajo.  
  
setwd("C:/Marsev/Ecuador/")  
  
#load("C:/Marsev/Ecuador/Ecuador_mg_vs_resobaja.Rdata")  
  
# Cargamos las librerias o paquetes requeridos.  
  
library(raster)  
  
## Loading required package: sp  
library(car)  
  
## Loading required package: carData  
library(rgdal)  
  
## rgdal: version: 1.4-3, (SVN revision 828)  
## Geospatial Data Abstraction Library extensions to R successfully loaded  
## Loaded GDAL runtime: GDAL 2.2.3, released 2017/11/20  
## Path to GDAL shared files: C:/Users/Antares/Documents/R/win-library/3.5/rgdal/gdal  
## GDAL binary built with GEOS: TRUE  
## Loaded PROJ.4 runtime: Rel. 4.9.3, 15 August 2016, [PJ_VERSION: 493]  
## Path to PROJ.4 shared files: C:/Users/Antares/Documents/R/win-library/3.5/rgdal/proj  
## Linking to sp version: 1.3-1  
  
library(gstat)  
library(caret)  
  
## Loading required package: lattice  
## Loading required package: ggplot2  
library(reshape)  
library(sp)  
library(lattice)  
library(ggplot2)  
library(automap)  
library(Metrics)  
  
##
```

```

## Attaching package: 'Metrics'

## The following objects are masked from 'package:caret':
##
##      precision, recall

# Cargamos las funciones requeridas.

load("DSM_supportfunctions.RData")

dummyRaster <- function(rast){
  rast <- as.factor(rast)
  result <- list()
  for(i in 1:length(levels(rast)[[1]][[1]])){
    result[[i]] <- rast == levels(rast)[[1]][[1]][i]
    names(result[[i]]) <- paste0(names(rast),
                                levels(rast)[[1]][[1]][i])
  }
  return(stack(result))
}

# Cargamos los datos del splines.

dat <- read.csv("Ecu_cali7.csv")

# Observamos los nombres de los campos o columnas.

# names(dat)

# Transformamos a factor las covariables categoricas.

dat$Bioclivs <- as.factor(dat$Bioclivs)
dat$Climavs <- as.factor(dat$Climavs)
dat$Cobervs <- as.factor(dat$Cobervs)
dat$Pisosvs <- as.factor(dat$Pisosvs)
dat$Suelosvs <- as.factor(dat$Suelosvs)

# Vemos estructura de los datos.

# str(dat)

# Convertimos las columnas de covariables categoricas a dummy,
# el resultado es una matriz:

dat_Bioclivs_du <- model.matrix(~Bioclivs -1, data = dat)
dat_Climavs_du <- model.matrix(~Climavs -1, data = dat)
dat_Cobervs_du <- model.matrix(~Cobervs -1, data = dat)
dat_Pisosvs_du <- model.matrix(~Pisosvs -1, data = dat)
dat_Suelosvs_du <- model.matrix(~Suelosvs -1, data = dat)

dat_Bioclivs_du <- as.data.frame(dat_Bioclivs_du)
dat_Climavs_du <- as.data.frame(dat_Climavs_du)
dat_Cobervs_du <- as.data.frame(dat_Cobervs_du)
dat_Pisosvs_du <- as.data.frame(dat_Pisosvs_du)
dat_Suelosvs_du <- as.data.frame(dat_Suelosvs_du)

```

```
dat <- cbind(dat, dat_Bioclivs_du, dat_Climavs_du, dat_Cobervs_du, dat_Pisosvs_du, dat_Suelosvs_du)
```

```
# Observamos los nombres de los campos o columnas.
```

```
names(dat)
```

```
## [1] "ID"
## [3] "LATITUDE"
## [5] "OCSKGM30"
## [7] "Analytical"
## [9] "Aspect"
## [11] "Longitudin"
## [13] "Closeddepr"
## [15] "Topographi"
## [17] "Channelnet"
## [19] "ValleyDepth"
## [21] "DEMSRE3a"
## [23] "evmmod3a"
## [25] "g01igb3a"
## [27] "g02igb3a"
## [29] "g04esa3a"
## [31] "g05esa3a"
## [33] "g10igb3a"
## [35] "g11igb3a"
## [37] "g13esa3a"
## [39] "g18esa3a"
## [41] "gachws3a"
## [43] "ganhws3a"
## [45] "gcmhws3a"
## [47] "gflhws3a"
## [49] "glcesa3a"
## [51] "glphws3a"
## [53] "glwwwf3a"
## [55] "gplhws3a"
## [57] "gumhws3a"
## [59] "inmsre3a"
## [61] "l02igb3a"
## [63] "l05igb3a"
## [65] "l07igb3a"
## [67] "l09igb3a"
## [69] "l11igb3a"
## [71] "l13igb3a"
## [73] "l3pobi3b"
## [75] "lasmod3a"
## [77] "px1wcl3a"
## [79] "px3wcl3a"
## [81] "slpsrt3a"
## [83] "tdlmod3a"
## [85] "tdsmod3a"
## [87] "tnlmod3a"
## [89] "tnsmod3a"
## [91] "tx1mod3a"
## [93] "tx3mod3a"

"ID"
"LONGITUDE"
"DEM"
"Slope"
"Crosssecti"
"Convergence"
"Flowaccumu"
"LSFactor"
"VerticalDistanceToChannelNetwork"
"RelativeSlopePosition"
"etmnts3a"
"evsmod3a"
"g02esa3a"
"g03esa3a"
"g04igb3a"
"g06esa3a"
"g11esa3a"
"g12igb3a"
"g14esa3a"
"gacgem3a"
"galhws3a"
"garhws3a"
"geaisg3a"
"gglhws3a"
"glcjrc3a"
"glvhws3a"
"gphhws3a"
"grghws3a"
"gvrhws3a"
"inssre3a"
"l04igb3a"
"l06igb3a"
"l08igb3a"
"l10igb3a"
"l12igb3a"
"l14igb3a"
"lammod3a"
"opisre3a"
"px2wcl3a"
"px4wcl3a"
"tdhmod3a"
"tdmmod3a"
"tnhmod3a"
"tnmmod3a"
"twisre3a"
"tx2mod3a"
"tx4mod3a"
```

```
## [95] "tx5mod3a"          "tx6mod3a"
## [97] "Bioclivs"          "Climavs"
## [99] "Cobervs"           "Ecosivs"
## [101] "Geolovs"           "Geomovs"
## [103] "Pisosvs"           "Suelosvs"
## [105] "Bioclivs1"         "Bioclivs2"
## [107] "Bioclivs3"         "Bioclivs4"
## [109] "Climavs1"          "Climavs2"
## [111] "Climavs3"          "Climavs4"
## [113] "Climavs5"          "Climavs6"
## [115] "Climavs7"          "Climavs8"
## [117] "Climavs9"          "Cobervs2"
## [119] "Cobervs3"          "Cobervs4"
## [121] "Cobervs5"          "Cobervs6"
## [123] "Cobervs7"          "Pisosvs1"
## [125] "Pisosvs2"          "Pisosvs3"
## [127] "Pisosvs4"          "Pisosvs7"
## [129] "Pisosvs9"          "Suelosvs1"
## [131] "Suelosvs2"         "Suelosvs3"
## [133] "Suelosvs4"         "Suelosvs5"
## [135] "Suelosvs6"         "Suelosvs7"
## [137] "Suelosvs8"         "Suelosvs9"
## [139] "Suelosvs10"        "Suelosvs11"
```

```
# Vemos un resumen de los datos de carbono organico de los perfiles de suelos en kg/m2.
```

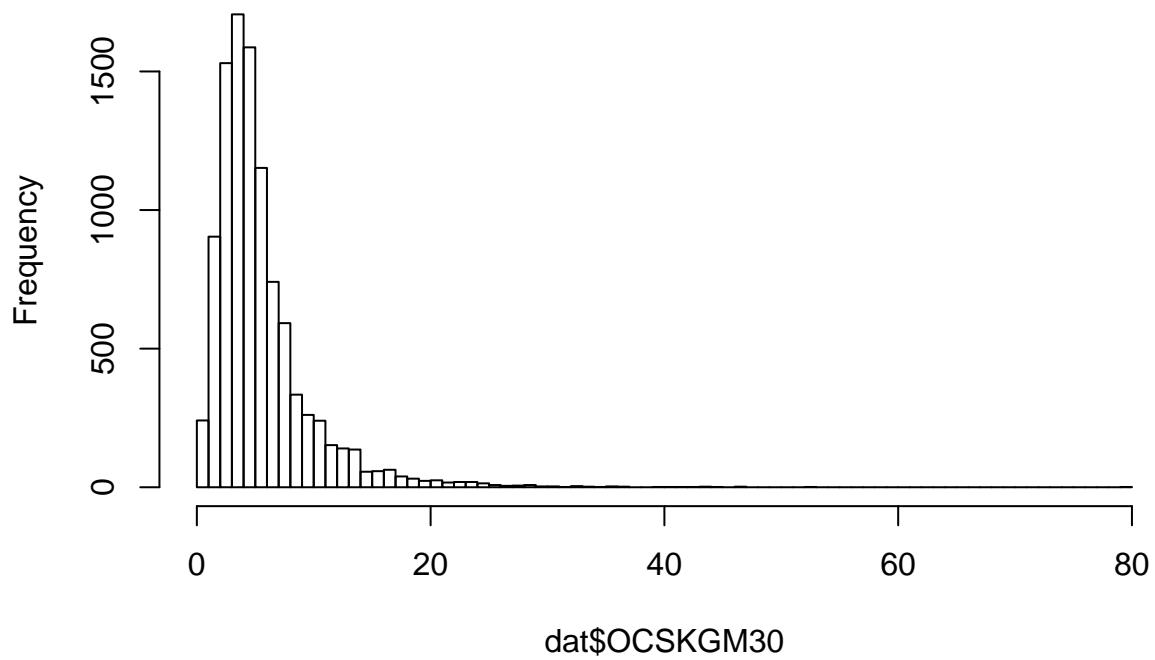
```
summary(dat$OCSKGM30)
```

```
##      Min.   1st Qu.   Median     Mean  3rd Qu.     Max.
## 0.02302  2.89453  4.39886  5.48640  6.63229 79.58376
```

```
# Diseñamos un histogramas de los datos de carbono organico de los perfiles de suelos.
```

```
hist(dat$OCSKGM30, breaks = 100)
```

Histogram of dat\$OCSKGM30



```
# Modificamos valores atipicos.
```

```
dat$OCSKGM30[dat$OCSKGM30 > 45] <- 45
```

```
# Vemos un resumen de los datos de carbono organico de los perfiles de suelos en kg/m2.
```

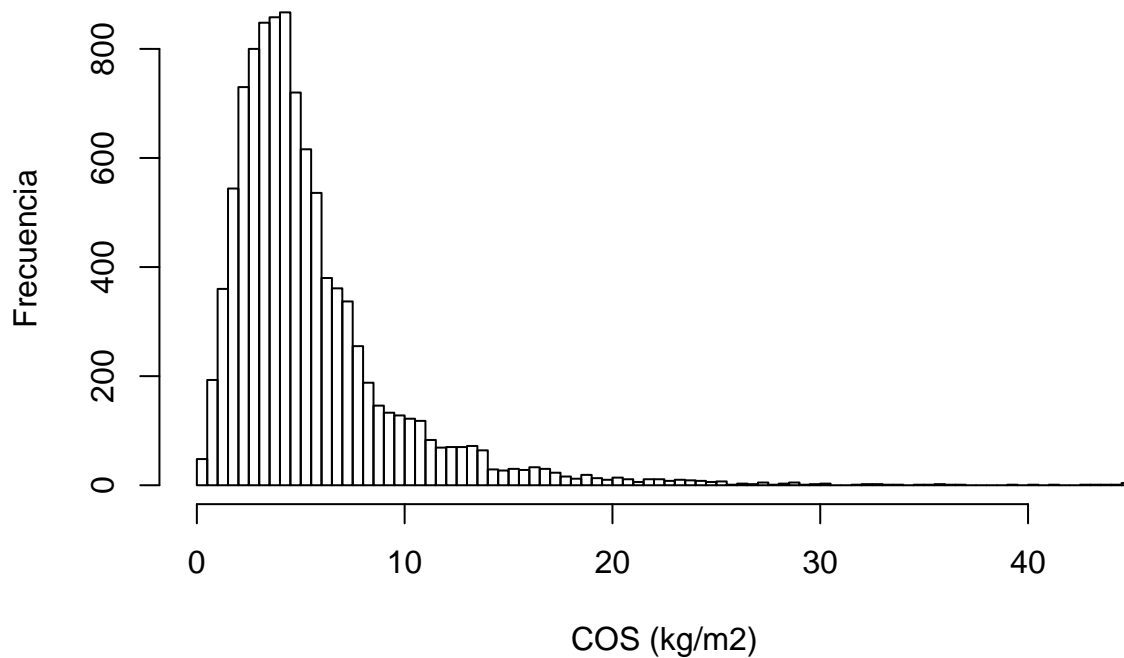
```
summary(dat$OCSKGM30)
```

```
##      Min.   1st Qu.   Median     Mean  3rd Qu.    Max.
##  0.02302  2.89453   4.39886   5.48203  6.63229  45.00000
```

```
# Disenamos un histogramas de los datos de carbono organico de los perfiles de suelos.
```

```
hist(dat$OCSKGM30, breaks = 100, main = "Histograma de Valores de COS (kg/m2)", ylab = 'Frecuencia', xlab = 'Valores de COS (kg/m2)',  
      sub='Histograma sobre datos de COS de perfiles de suelos.' )
```

Histograma de Valores de COS (kg/m2)



```
# Removemos valores atipicos, segun Bonferroni.
```

```
dat <- dat[-c(4496, 2510, 2220, 6999, 9201, 8716, 2214, 2031, 3002, 3869),]
```

```
# Vemos un resumen de los datos de carbono organico de los perfiles de suelos en kg/m2.
```

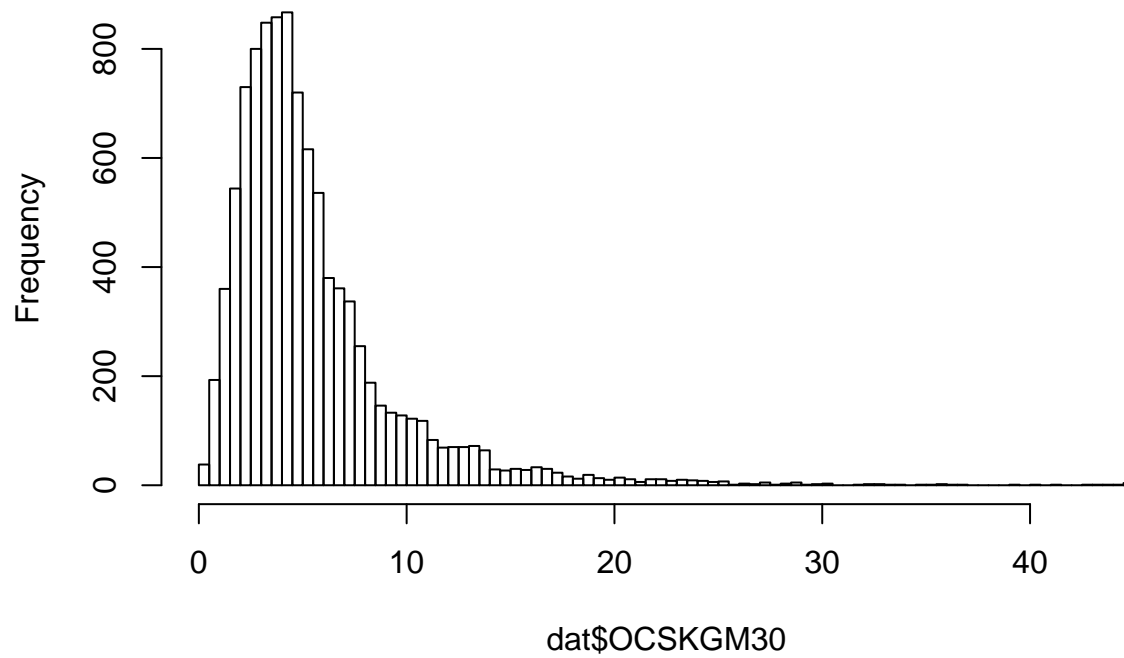
```
summary(dat$OCSKGM30)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##  0.1655  2.9013   4.3989   5.4873  6.6323  45.0000
```

```
# Disenamos un histogramas de los datos de carbono organico de los perfiles de suelos.
```

```
hist(dat$OCSKGM30, breaks = 100)
```

Histogram of dat\$OCSKGM30



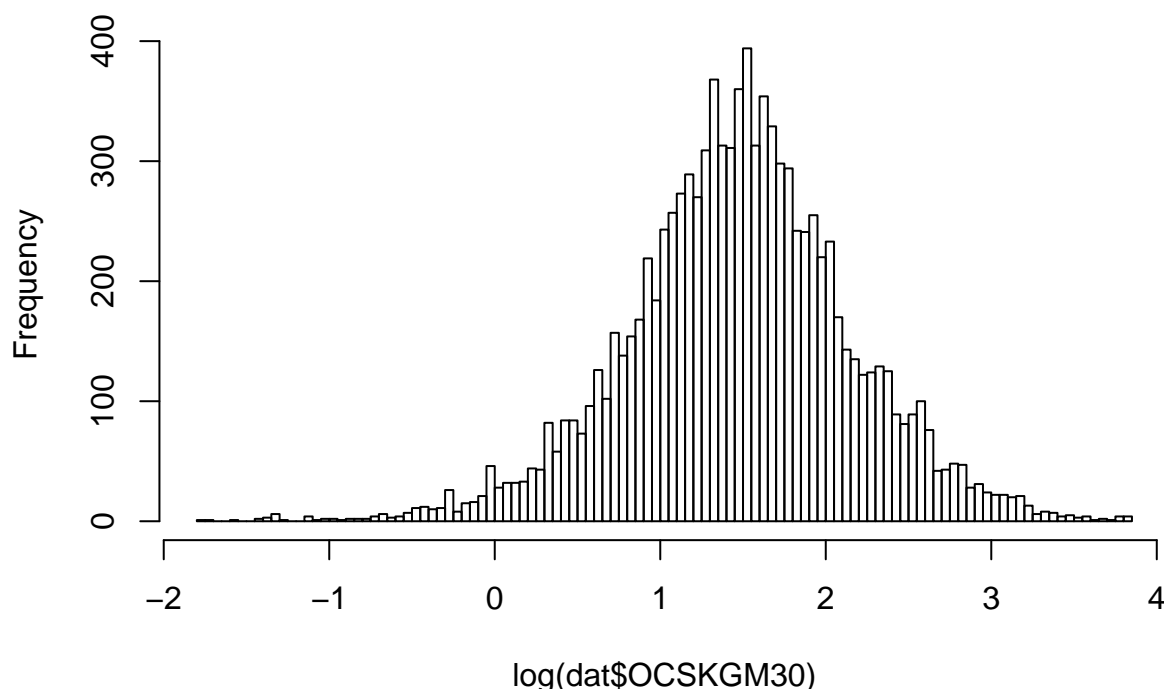
```
# Vemos la estructura de los datos.
```

```
# str(dat)
```

```
# Transformamos a log y diseñamos un histogramas de los datos de COS de los perfiles de suelos.
```

```
hist(log(dat$OCSKGM30), breaks=100)
```

Histogram of log(dat\$OCSKGM30)



```
## Recreamos el objeto con la ubicacion de los puntos
```

```
dat_sp <- dat
coordinates(dat_sp) <- ~ LONGITUDE + LATITUDE
```

```
### Analisis de correlacion
```

```
# names(dat_sp@data)
```

```
COR <- cor(as.matrix(dat_sp@data[,3]), as.matrix(dat_sp@data[, -c(1:3, 95, 96, 97, 101, 102)]))
COR
```

```
## [1,] 0.208929 -0.01369506 0.05634872 0.02661526 0.0944178 0.147155
## Coverage Closeddepr Flowaccumu Topographi LSFactor Channelnet
## [1,] 0.09215387 -0.06614235 -0.03510654 -0.1537406 -0.0678515 0.08534772
## VerticalDistanceToChannelNetwork ValleyDepth RelativeSlopePosition
## [1,] 0.3741556 -0.1803803 0.355123
## DEMSRE3a etmnts3a evmmod3a evsmod3a g01igb3a g02esa3a
## [1,] 0.2105132 0.03984657 0.1475343 -0.05230708 -0.07836442 0.03834311
## g02igb3a g03esa3a g04esa3a g04igb3a g05esa3a g06esa3a
## [1,] -0.07736351 0.02763012 -0.03925473 -0.1048521 0.08657251 0.05915097
## g10igb3a g11esa3a g11igb3a g12igb3a g13esa3a g14esa3a
## [1,] -0.103391 -0.06309548 -0.1066508 -0.1067288 -0.004986846 -0.1429939
## g18esa3a gacgem3a gachws3a galhws3a ganhws3a garhws3a
## [1,] -0.002570222 0.1095029 -0.09087424 0.05191342 0.2407074 -0.0926822
## gcmhws3a geaisg3a gflhws3a gglhws3a glcesa3a
## [1,] -0.03091354 -0.1154044 -0.008979239 -0.09756214 -0.1080372
```



```
##      glcjrc3a  glphws3a  glvhws3a  glwwwf3a  gphhws3a
## [1,] 0.001386801 -0.0885956 -0.05393139 -0.02290974 -0.06567279
##      gplhws3a  grghws3a  gumhws3a  gvrhws3a  inmsre3a  inssre3a
## [1,] -0.07246678 -0.1632326 0.1569082 -0.07845569 0.1837037 0.1925178
##      l02igb3a  l04igb3a  l05igb3a  l06igb3a  l07igb3a  l08igb3a
## [1,] 0.08800435 -0.01624081 0.07691153 -0.0447614 -0.09222192 0.03296024
##      l09igb3a  l10igb3a  l11igb3a  l12igb3a  l13igb3a  l14igb3a
## [1,] -0.07807656 0.1140745 -0.03794976 -0.06704101 -0.07830182 -0.08818114
##      l3pobi3b  lammod3a  lasmod3a  opisre3a  px1wcl3a  px2wcl3a
## [1,] -0.1309477 0.04271475 0.1048257 -0.01459334 0.1067635 0.07436161
##      px3wcl3a  px4wcl3a  slpsrt3a  tdhmod3a  tdlmod3a  tdmmod3a
## [1,] 0.07609933 0.07338385 0.1487879 -0.3683897 -0.2501556 -0.3766725
##      tdsmod3a  tnhmod3a  tnmod3a  tnmod3a  tnsmod3a  twisre3a
## [1,] -0.01636793 -0.2272885 -0.1775566 -0.2068962 0.06277568 -0.2425156
##      tx1mod3a  tx2mod3a  tx3mod3a  tx4mod3a  tx5mod3a  tx6mod3a
## [1,] -0.3867905 -0.4121598 -0.362301 -0.3476275 -0.3676105 -0.3586784
##      Ecosivs  Geolovs  Geomovs  Bioclivs1  Bioclivs2  Bioclivs3
## [1,] -0.003836575 -0.04858467 -0.02815205 0.1822291 -0.02969367 -0.1848079
##      Bioclivs4  Climavs1  Climavs2  Climavs3  Climavs4  Climavs5
## [1,] -0.02582213 0.1267681 -0.08554551 -0.08056989 -0.09176811 -0.1248557
##      Climavs6  Climavs7  Climavs8  Climavs9  Cobervs2  Cobervs3
## [1,] 0.161981 0.06076828 -0.08873329 0.05683279 0.0343383 -0.03353496
##      Cobervs4  Cobervs5  Cobervs6  Cobervs7  Pisosvs1  Pisosvs2
## [1,] 0.03273343 -0.003925262 -0.03322118 -0.04611703 -0.01916704 0.1556606
##      Pisosvs3  Pisosvs4  Pisosvs7  Pisosvs9  Suelosvs1  Suelosvs2
## [1,] 0.178582 0.04207305 0.01095232 -0.1793011 -0.04141558 0.4049269
##      Suelosvs3  Suelosvs4  Suelosvs5  Suelosvs6  Suelosvs7  Suelosvs8
## [1,] -0.0759617 -0.1284575 0.01885043 -0.1214687 -0.06770708 -0.02046923
##      Suelosvs9  Suelosvs10  Suelosvs11
## [1,] 0.002592527 -0.02810099 -0.07658166
```

```
x <- subset(melt(COR), value != 1 | value != NA)
```

```
x <- x[with(x, order(-abs(x$value))),]
```

```
#as.character(x$X2[1:10])
```

```
# Vemos las primeras 10 covariables de mayor correlacion con el COS.
```

```
x[1:10,]
```

```
##      X1      X2      value
## 87  1      tx2mod3a -0.4121598
## 121 1      Suelosvs2 0.4049269
## 86  1      tx1mod3a -0.3867905
## 79  1      tdmmod3a -0.3766725
## 13  1 VerticalDistanceToChannelNetwork 0.3741556
## 77  1      tdhmod3a -0.3683897
## 90  1      tx5mod3a -0.3676105
## 88  1      tx3mod3a -0.3623010
## 91  1      tx6mod3a -0.3586784
## 15  1      RelativeSlopePosition 0.3551230
```

```
idx <- as.character(x$X2[1:25])
```

```
idx
```

```
## [1] "tx2mod3a"      "Suelosvs2"
## [3] "tx1mod3a"      "tdmmod3a"
```

```
## [5] "VerticalDistanceToChannelNetwork" "tdhmod3a"
## [7] "tx5mod3a" "tx3mod3a"
## [9] "tx6mod3a" "RelativeSlopePosition"
## [11] "tx4mod3a" "tdlmod3a"
## [13] "twisre3a" "ganhws3a"
## [15] "tnhmod3a" "DEMSRE3a"
## [17] "DEM" "tnmmod3a"
## [19] "inssre3a" "Bioclivs3"
## [21] "inmsre3a" "Bioclivs1"
## [23] "ValleyDepth" "Pisosvs9"
## [25] "Pisosvs3"

# Creamos el archivo de datos para emplear en la regresion lineal multiple.

dat2 <- dat[c('OCSKGM30', idx, 'LATITUDE', 'LONGITUDE')]

# Observamos los nombres de los campos o columnas.

#names(dat2)

dat2[dat$OCSKGM30 == 0, 1] <- NA

## Modelo de Regresion lineal multiple.

modelo.MLR <- lm(log(OCSKGM30) ~ . -LATITUDE-LONGITUDE, data = dat2)

# Vemos un resumen de los resultados del modelo de regresion lineal multiple.

summary(modelo.MLR)

##
## Call:
## lm(formula = log(OCSKGM30) ~ . - LATITUDE - LONGITUDE, data = dat2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.86687 -0.31757  0.04076  0.35801  2.38386
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)      3.760e+00  3.543e-01  10.611 < 2e-16
## tx2mod3a        -3.250e-02  4.469e-03  -7.271 3.82e-13
## Suelosvs2       3.553e-01  1.721e-02  20.643 < 2e-16
## tx1mod3a        -2.515e-04  3.466e-03  -0.073  0.94216
## tdmmod3a        -2.309e-02  7.292e-03  -3.167  0.00155
## VerticalDistanceToChannelNetwork 2.277e-04  3.787e-05   6.012 1.89e-09
## tdhmod3a        -1.944e-02  3.099e-03  -6.272 3.71e-10
## tx5mod3a        -1.138e-03  3.787e-03  -0.301  0.76377
## tx3mod3a        -9.507e-03  4.717e-03  -2.016  0.04386
## tx6mod3a        2.161e-02  3.365e-03   6.423 1.40e-10
## RelativeSlopePosition 3.072e-01  7.500e-02   4.096 4.24e-05
## tx4mod3a        -7.014e-03  4.293e-03  -1.634  0.10229
## tdlmod3a        1.114e-02  2.486e-03   4.482 7.46e-06
## twisre3a        -7.165e-04  5.298e-04  -1.352  0.17629
## ganhws3a        1.279e-03  1.931e-04   6.624 3.68e-11
```

```

## tnhmod3a          -5.563e-02  6.609e-03  -8.417  < 2e-16
## DEMSRE3a          5.567e-04  2.150e-04   2.589  0.00964
## DEM              -1.047e-03  2.177e-04  -4.809  1.54e-06
## tnmmod3a          8.216e-03  5.067e-03   1.622  0.10492
## inssre3a          1.914e-02  1.252e-02   1.529  0.12631
## Bioclivs3        -9.057e-03  1.852e-02  -0.489  0.62474
## inmsre3a          5.214e-03  3.352e-03   1.556  0.11985
## Bioclivs1         4.538e-02  1.697e-02   2.673  0.00752
## ValleyDepth       1.651e-05  1.609e-05   1.026  0.30490
## Pisosvs9         -1.452e-01  2.235e-02  -6.495  8.70e-11
## Pisosvs3          1.230e-01  4.035e-02   3.048  0.00231
##
## (Intercept)      ***
## tx2mod3a          ***
## Suelosvs2         ***
## tx1mod3a
## tdmmod3a          **
## VerticalDistanceToChannelNetwork ***
## tdhmod3a          ***
## tx5mod3a
## tx3mod3a          *
## tx6mod3a          ***
## RelativeSlopePosition ***
## tx4mod3a
## tdlmod3a          ***
## twisre3a
## ganhws3a          ***
## tnhmod3a          ***
## DEMSRE3a          **
## DEM              ***
## tnmmod3a
## inssre3a
## Bioclivs3
## inmsre3a
## Bioclivs1         **
## ValleyDepth
## Pisosvs9          ***
## Pisosvs3          **
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5705 on 10101 degrees of freedom
## Multiple R-squared:  0.298, Adjusted R-squared:  0.2963
## F-statistic: 171.5 on 25 and 10101 DF, p-value: < 2.2e-16

# Analisis de varianza.

anova(modelo.MLR)

## Analysis of Variance Table
##
## Response: log(OCSKGM30)
##
##              Df Sum Sq Mean Sq  F value    Pr(>F)
## tx2mod3a      1  745.9   745.86 2291.3286 < 2.2e-16
## Suelosvs2     1  291.8   291.77  896.3471 < 2.2e-16

```

```

## tx1mod3a          1    1.6    1.57    4.8342 0.0279235
## tdmmod3a          1    4.7    4.68   14.3817 0.0001501
## VerticalDistanceToChannelNetwork 1   23.5   23.49   72.1620 < 2.2e-16
## tdhmod3a          1  104.0  104.02  319.5491 < 2.2e-16
## tx5mod3a          1   17.4   17.39   53.4108 2.911e-13
## tx3mod3a          1    0.1    0.15    0.4536 0.5006617
## tx6mod3a          1    7.3    7.26   22.3150 2.345e-06
## RelativeSlopePosition 1   22.5   22.50   69.1320 < 2.2e-16
## tx4mod3a          1    0.1    0.13    0.4077 0.5231504
## tdlmod3a          1   41.8   41.77  128.3207 < 2.2e-16
## twisre3a          1    0.7    0.70    2.1359 0.1439195
## ganhws3a          1    8.7    8.70   26.7201 2.397e-07
## tnhmod3a          1   39.7   39.72  122.0136 < 2.2e-16
## DEMSRE3a          1   51.3   51.35  157.7358 < 2.2e-16
## DEM               1    6.5    6.52   20.0209 7.744e-06
## tnmmod3a          1    4.5    4.50   13.8329 0.0002009
## inssre3a          1    0.1    0.08    0.2444 0.6210258
## Bioclivs3         1    0.5    0.51    1.5600 0.2116923
## inmsre3a          1    0.8    0.85    2.6050 0.1065574
## Bioclivs1         1    3.5    3.52   10.8117 0.0010120
## ValleyDepth       1    2.8    2.85    8.7511 0.0031014
## Pisosvs9          1   12.9   12.91   39.6628 3.144e-10
## Pisosvs3          1    3.0    3.02    9.2925 0.0023069
## Residuals        10101 3288.0    0.33
##
## tx2mod3a          ***
## Suelosvs2         ***
## tx1mod3a          *
## tdmmod3a          ***
## VerticalDistanceToChannelNetwork ***
## tdhmod3a          ***
## tx5mod3a          ***
## tx3mod3a          ***
## tx6mod3a          ***
## RelativeSlopePosition ***
## tx4mod3a          ***
## tdlmod3a          ***
## twisre3a          ***
## ganhws3a          ***
## tnhmod3a          ***
## DEMSRE3a          ***
## DEM               ***
## tnmmod3a          ***
## inssre3a          ***
## Bioclivs3         ***
## inmsre3a          ***
## Bioclivs1         **
## ValleyDepth       **
## Pisosvs9          ***
## Pisosvs3          **
## Residuals
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

```

```
## Hacemos seleccion de variables por stepwise
```

```
modelo.MLR.step <- step(modelo.MLR, direction="both")
```

```
## Start: AIC=-11340.09
```

```
## log(OCSKGM30) ~ (tx2mod3a + Suelosvs2 + tx1mod3a + tdmmod3a +  
## VerticalDistanceToChannelNetwork + tdhmod3a + tx5mod3a +  
## tx3mod3a + tx6mod3a + RelativeSlopePosition + tx4mod3a +  
## tdlmod3a + twisre3a + ganhws3a + tnhmod3a + DEMSRE3a + DEM +  
## tnmmmod3a + inssre3a + Bioclivs3 + inmsre3a + Bioclivs1 +  
## ValleyDepth + Pisosvs9 + Pisosvs3 + LATITUDE + LONGITUDE) -  
## LATITUDE - LONGITUDE
```

	Df	Sum of Sq	RSS	AIC
## - tx1mod3a	1	0.002	3288.0	-11342
## - tx5mod3a	1	0.029	3288.0	-11342
## - Bioclivs3	1	0.078	3288.1	-11342
## - ValleyDepth	1	0.343	3288.4	-11341
## - twisre3a	1	0.595	3288.6	-11340
## <none>			3288.0	-11340
## - inssre3a	1	0.761	3288.8	-11340
## - inmsre3a	1	0.788	3288.8	-11340
## - tnmmmod3a	1	0.856	3288.9	-11340
## - tx4mod3a	1	0.869	3288.9	-11339
## - tx3mod3a	1	1.323	3289.3	-11338
## - DEMSRE3a	1	2.182	3290.2	-11335
## - Bioclivs1	1	2.327	3290.3	-11335
## - Pisosvs3	1	3.025	3291.0	-11333
## - tdmmod3a	1	3.264	3291.3	-11332
## - RelativeSlopePosition	1	5.461	3293.5	-11325
## - tdlmod3a	1	6.540	3294.6	-11322
## - DEM	1	7.529	3295.5	-11319
## - VerticalDistanceToChannelNetwork	1	11.767	3299.8	-11306
## - tdhmod3a	1	12.805	3300.8	-11303
## - tx6mod3a	1	13.429	3301.4	-11301
## - Pisosvs9	1	13.731	3301.7	-11300
## - ganhws3a	1	14.282	3302.3	-11298
## - tx2mod3a	1	17.211	3305.2	-11289
## - tnhmod3a	1	23.060	3311.1	-11271
## - Suelosvs2	1	138.715	3426.7	-10924

```
## Step: AIC=-11342.09
```

```
## log(OCSKGM30) ~ tx2mod3a + Suelosvs2 + tdmmod3a + VerticalDistanceToChannelNetwork +  
## tdhmod3a + tx5mod3a + tx3mod3a + tx6mod3a + RelativeSlopePosition +  
## tx4mod3a + tdlmod3a + twisre3a + ganhws3a + tnhmod3a + DEMSRE3a +  
## DEM + tnmmmod3a + inssre3a + Bioclivs3 + inmsre3a + Bioclivs1 +  
## ValleyDepth + Pisosvs9 + Pisosvs3
```

	Df	Sum of Sq	RSS	AIC
## - tx5mod3a	1	0.031	3288.0	-11344
## - Bioclivs3	1	0.080	3288.1	-11344
## - ValleyDepth	1	0.341	3288.4	-11343
## - twisre3a	1	0.599	3288.6	-11342
## <none>			3288.0	-11342

```

## - inssre3a          1      0.759 3288.8 -11342
## - inmsre3a          1      0.788 3288.8 -11342
## - tnmmmod3a         1      0.856 3288.9 -11342
## - tx4mod3a          1      0.869 3288.9 -11341
## + tx1mod3a          1      0.002 3288.0 -11340
## - tx3mod3a          1      1.325 3289.3 -11340
## - DEMSRE3a          1      2.181 3290.2 -11337
## - Bioclivs1         1      2.362 3290.4 -11337
## - Pisosvs3          1      3.023 3291.0 -11335
## - tdmmod3a          1      3.498 3291.5 -11333
## - RelativeSlopePosition 1      5.462 3293.5 -11327
## - tdlmod3a          1      6.557 3294.6 -11324
## - DEM               1      7.528 3295.5 -11321
## - VerticalDistanceToChannelNetwork 1 11.806 3299.8 -11308
## - tdhmod3a          1     12.805 3300.8 -11305
## - tx6mod3a          1     13.457 3301.5 -11303
## - Pisosvs9          1     13.803 3301.8 -11302
## - ganhws3a          1     14.300 3302.3 -11300
## - tx2mod3a          1     19.361 3307.4 -11285
## - tnhmod3a          1     23.060 3311.1 -11273
## - Suelosvs2         1    138.828 3426.8 -10925
##
## Step: AIC=-11343.99
## log(OCSKGM30) ~ tx2mod3a + Suelosvs2 + tdmmod3a + VerticalDistanceToChannelNetwork +
##      tdhmod3a + tx3mod3a + tx6mod3a + RelativeSlopePosition +
##      tx4mod3a + tdlmod3a + twisre3a + ganhws3a + tnhmod3a + DEMSRE3a +
##      DEM + tnmmmod3a + inssre3a + Bioclivs3 + inmsre3a + Bioclivs1 +
##      ValleyDepth + Pisosvs9 + Pisosvs3
##
##
##              Df Sum of Sq    RSS    AIC
## - Bioclivs3          1      0.071 3288.1 -11346
## - ValleyDepth        1      0.331 3288.4 -11345
## - twisre3a           1      0.618 3288.7 -11344
## <none>                                3288.0 -11344
## - inssre3a          1      0.767 3288.8 -11344
## - inmsre3a          1      0.809 3288.9 -11344
## - tnmmmod3a         1      0.833 3288.9 -11343
## - tx4mod3a          1      1.159 3289.2 -11342
## + tx5mod3a          1      0.031 3288.0 -11342
## + tx1mod3a          1      0.003 3288.0 -11342
## - tx3mod3a          1      1.387 3289.4 -11342
## - DEMSRE3a          1      2.170 3290.2 -11339
## - Bioclivs1         1      2.576 3290.6 -11338
## - Pisosvs3          1      3.058 3291.1 -11337
## - tdmmod3a          1      3.680 3291.7 -11335
## - RelativeSlopePosition 1      5.523 3293.6 -11329
## - tdlmod3a          1      6.636 3294.7 -11326
## - DEM               1      7.543 3295.6 -11323
## - VerticalDistanceToChannelNetwork 1 11.779 3299.8 -11310
## - tdhmod3a          1     12.799 3300.8 -11307
## - Pisosvs9          1     13.773 3301.8 -11304
## - ganhws3a          1     14.300 3302.3 -11302
## - tx6mod3a          1     14.535 3302.6 -11301
## - tx2mod3a          1     19.332 3307.4 -11287

```

```

## - tnhmod3a          1    23.311 3311.4 -11274
## - Suelosvs2         1   138.864 3426.9 -10927
##
## Step: AIC=-11345.77
## log(OCSKGM30) ~ tx2mod3a + Suelosvs2 + tdmmod3a + VerticalDistanceToChannelNetwork +
##      tdhmod3a + tx3mod3a + tx6mod3a + RelativeSlopePosition +
##      tx4mod3a + tdlmod3a + twisre3a + ganhws3a + tnhmod3a + DEMSRE3a +
##      DEM + tnmmod3a + inssre3a + inmsre3a + Bioclivs1 + ValleyDepth +
##      Pisosvs9 + Pisosvs3
##
##
##      Df Sum of Sq    RSS    AIC
## - ValleyDepth      1      0.314 3288.4 -11347
## - twisre3a          1      0.591 3288.7 -11346
## <none>                                3288.1 -11346
## - inssre3a          1      0.766 3288.9 -11345
## - inmsre3a          1      0.798 3288.9 -11345
## - tnmmod3a          1      0.862 3289.0 -11345
## - tx4mod3a          1      1.166 3289.3 -11344
## + Bioclivs3          1      0.071 3288.0 -11344
## + tx5mod3a          1      0.022 3288.1 -11344
## + tx1mod3a          1      0.005 3288.1 -11344
## - tx3mod3a          1      1.445 3289.6 -11343
## - DEMSRE3a          1      2.176 3290.3 -11341
## - Bioclivs1         1      2.736 3290.8 -11339
## - Pisosvs3          1      3.021 3291.1 -11338
## - tdmmod3a          1      4.067 3292.2 -11335
## - RelativeSlopePosition 1      5.473 3293.6 -11331
## - tdlmod3a          1      6.712 3294.8 -11327
## - DEM               1      7.524 3295.6 -11325
## - VerticalDistanceToChannelNetwork 1     11.776 3299.9 -11312
## - tdhmod3a          1     13.095 3301.2 -11308
## - Pisosvs9          1     14.035 3302.1 -11305
## - ganhws3a          1     14.681 3302.8 -11303
## - tx6mod3a          1     15.398 3303.5 -11300
## - tx2mod3a          1     19.275 3307.4 -11289
## - tnhmod3a          1     23.433 3311.5 -11276
## - Suelosvs2         1    138.963 3427.1 -10929
##
## Step: AIC=-11346.81
## log(OCSKGM30) ~ tx2mod3a + Suelosvs2 + tdmmod3a + VerticalDistanceToChannelNetwork +
##      tdhmod3a + tx3mod3a + tx6mod3a + RelativeSlopePosition +
##      tx4mod3a + tdlmod3a + twisre3a + ganhws3a + tnhmod3a + DEMSRE3a +
##      DEM + tnmmod3a + inssre3a + inmsre3a + Bioclivs1 + Pisosvs9 +
##      Pisosvs3
##
##
##      Df Sum of Sq    RSS    AIC
## <none>                                3288.4 -11347
## - twisre3a          1      0.701 3289.1 -11347
## - inssre3a          1      0.705 3289.1 -11347
## - inmsre3a          1      0.779 3289.2 -11346
## - tnmmod3a          1      0.871 3289.3 -11346
## + ValleyDepth      1      0.314 3288.1 -11346
## - tx4mod3a          1      1.128 3289.6 -11345
## + Bioclivs3          1      0.054 3288.4 -11345

```

```
## + tx5mod3a          1      0.015 3288.4 -11345
## + tx1mod3a          1      0.002 3288.4 -11345
## - tx3mod3a          1      1.358 3289.8 -11345
## - DEMSRE3a          1      2.147 3290.6 -11342
## - Bioclivs1         1      2.685 3291.1 -11340
## - Pisosvs3          1      3.047 3291.5 -11339
## - tdmmod3a          1      3.962 3292.4 -11337
## - RelativeSlopePosition 1      6.259 3294.7 -11330
## - tdlmod3a          1      6.601 3295.0 -11328
## - DEM               1      7.396 3295.8 -11326
## - tdhmod3a          1     13.201 3301.6 -11308
## - VerticalDistanceToChannelNetwork 1     14.570 3303.0 -11304
## - ganhws3a          1     14.867 3303.3 -11303
## - tx6mod3a          1     15.104 3303.5 -11302
## - Pisosvs9          1     16.627 3305.1 -11298
## - tx2mod3a          1     19.004 3307.4 -11290
## - tnhmod3a          1     23.120 3311.5 -11278
## - Suelosvs2         1    138.882 3427.3 -10930
```

```
summary(modelo.MLR.step)
```

```
##
## Call:
## lm(formula = log(OCSKGM30) ~ tx2mod3a + Suelosvs2 + tdmmod3a +
##     VerticalDistanceToChannelNetwork + tdhmod3a + tx3mod3a +
##     tx6mod3a + RelativeSlopePosition + tx4mod3a + tdlmod3a +
##     twisre3a + ganhws3a + tnhmod3a + DEMSRE3a + DEM + tnmod3a +
##     inssre3a + inmsre3a + Bioclivs1 + Pisosvs9 + Pisosvs3, data = dat2)
##
## Residuals:
##      Min       1Q   Median       3Q      Max
## -2.86564 -0.31856  0.04048  0.35757  2.38921
##
## Coefficients:
##              Estimate Std. Error t value Pr(>|t|)
## (Intercept)    3.788e+00  3.528e-01  10.738 < 2e-16
## tx2mod3a       -3.216e-02  4.209e-03  -7.642 2.34e-14
## Suelosvs2      3.538e-01  1.713e-02  20.658 < 2e-16
## tdmmod3a       -2.387e-02  6.842e-03  -3.489 0.000487
## VerticalDistanceToChannelNetwork 2.390e-04  3.571e-05   6.691 2.33e-11
## tdhmod3a       -1.964e-02  3.083e-03  -6.369 1.98e-10
## tx3mod3a       -9.518e-03  4.660e-03  -2.043 0.041117
## tx6mod3a       2.099e-02  3.080e-03   6.813 1.01e-11
## RelativeSlopePosition 2.625e-01  5.985e-02   4.386 1.17e-05
## tx4mod3a       -7.375e-03  3.961e-03  -1.862 0.062662
## tdlmod3a       1.115e-02  2.475e-03   4.504 6.75e-06
## twisre3a       -7.684e-04  5.236e-04  -1.468 0.142222
## ganhws3a       1.296e-03  1.918e-04   6.759 1.47e-11
## tnhmod3a       -5.454e-02  6.471e-03  -8.429 < 2e-16
## DEMSRE3a       5.519e-04  2.149e-04   2.568 0.010229
## DEM           -1.036e-03  2.173e-04  -4.767 1.89e-06
## tnmod3a        8.242e-03  5.038e-03   1.636 0.101908
## inssre3a       1.836e-02  1.247e-02   1.472 0.141031
## inmsre3a       5.146e-03  3.326e-03   1.547 0.121895
## Bioclivs1      4.709e-02  1.639e-02   2.872 0.004084
```



```
## Pisosvs9 -1.519e-01 2.125e-02 -7.148 9.42e-13
## Pisosvs3 1.231e-01 4.024e-02 3.060 0.002221
##
## (Intercept) ***
## tx2mod3a ***
## Suelosvs2 ***
## tdmmod3a ***
## VerticalDistanceToChannelNetwork ***
## tdhmod3a ***
## tx3mod3a *
## tx6mod3a ***
## RelativeSlopePosition ***
## tx4mod3a .
## tdlmod3a ***
## twisre3a
## ganhws3a ***
## tnhmod3a ***
## DEMSRE3a *
## DEM ***
## tnmmod3a
## inssre3a
## inmsre3a
## Bioclivs1 **
## Pisosvs9 ***
## Pisosvs3 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.5705 on 10105 degrees of freedom
## Multiple R-squared: 0.2979, Adjusted R-squared: 0.2965
## F-statistic: 204.2 on 21 and 10105 DF, p-value: < 2.2e-16
```

```
# Analisis de varianza.
```

```
anova(modelo.MLR.step)
```

```
## Analysis of Variance Table
```

```
##
```

```
## Response: log(OCSKGM30)
```

```
##
```

	Df	Sum Sq	Mean Sq	F value	Pr(>F)
## tx2mod3a	1	745.9	745.86	2291.9449	< 2.2e-16
## Suelosvs2	1	291.8	291.77	896.5882	< 2.2e-16
## tdmmod3a	1	6.2	6.21	19.0806	1.266e-05
## VerticalDistanceToChannelNetwork	1	23.5	23.45	72.0672	< 2.2e-16
## tdhmod3a	1	103.9	103.91	319.3139	< 2.2e-16
## tx3mod3a	1	0.0	0.05	0.1495	0.6989824
## tx6mod3a	1	0.5	0.46	1.4265	0.2323601
## RelativeSlopePosition	1	29.0	28.96	88.9999	< 2.2e-16
## tx4mod3a	1	2.5	2.47	7.5808	0.0059099
## tdlmod3a	1	49.1	49.11	150.8953	< 2.2e-16
## twisre3a	1	1.2	1.19	3.6494	0.0561175
## ganhws3a	1	7.9	7.93	24.3595	8.121e-07
## tnhmod3a	1	43.3	43.31	133.0753	< 2.2e-16
## DEMSRE3a	1	56.1	56.13	172.4924	< 2.2e-16
## DEM	1	6.6	6.58	20.2212	6.975e-06

```

## tnmod3a          1    4.4    4.44    13.6571 0.0002206
## inssre3a         1    0.1    0.07     0.2198 0.6391878
## inmsre3a         1    0.9    0.92     2.8394 0.0920116
## Bioclivs1        1    3.8    3.83    11.7732 0.0006033
## Pisosvs9         1   15.7   15.69   48.2030 4.079e-12
## Pisosvs3         1    3.0    3.05     9.3622 0.0022209
## Residuals       10105 3288.4    0.33
##
## tx2mod3a          ***
## Suelosvs2         ***
## tdmmod3a          ***
## VerticalDistanceToChannelNetwork ***
## tdhmod3a          ***
## tx3mod3a
## tx6mod3a
## RelativeSlopePosition ***
## tx4mod3a          **
## tdlmod3a          ***
## twisre3a          .
## ganhws3a          ***
## tnhmod3a          ***
## DEMSRE3a          ***
## DEM               ***
## tnmod3a           ***
## inssre3a
## inmsre3a          .
## Bioclivs1         ***
## Pisosvs9          ***
## Pisosvs3          **
## Residuals
## ---
## Signif. codes:  0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1

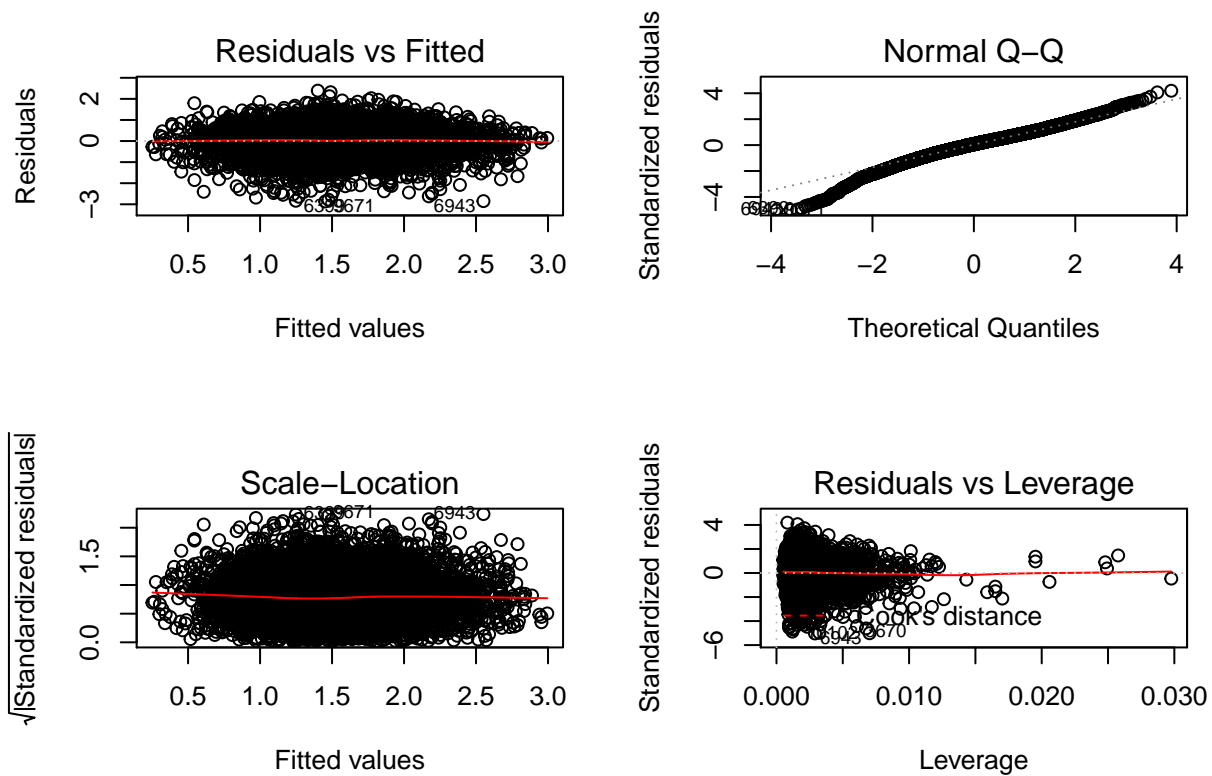
```

```
# Dividimos el area de graficos en 2 filas y 2 columnas.
```

```
par(mfrow=c(2,2))
```

```
# Hacemos los graficos del modelo de regresion lineal multiple.
```

```
plot(modelo.MLR.step)
```



Dividimos el area de graficos en 1 filas y 1 columnas.

```
par(mfrow=c(1,1))
```

*#Falta de multicolinealidad en las variables x: podemos comprobar esto mediante
#el calculo de los Factores de Inflacion de la Varianza (FIVs)*

```
vif(modelo.MLR.step)
```

##	tx2mod3a	Suelosvs2
##	7.707706	1.420587
##	tdmmod3a	VerticalDistanceToChannelNetwork
##	24.949207	7.629238
##	tdhmod3a	tx3mod3a
##	6.238645	10.418959
##	tx6mod3a	RelativeSlopePosition
##	7.007404	6.775654
##	tx4mod3a	tdlmod3a
##	7.884026	7.285933
##	twisre3a	ganhws3a
##	3.795768	1.461294
##	tnhmod3a	DEMSRE3a
##	45.119320	1970.941696
##	DEM	tnmmod3a
##	2014.949317	34.898986
##	inssre3a	inmsre3a

```
##          13.676326          5.760323
##          Bioclivs1          Pisosvs9
##          1.973250          2.826387
##          Pisosvs3
##          1.251000
```

#Variables problematicas tienen sqrt(FIV) > 2

```
sqrt(vif(modelo.MLR.step))
```

```
##          tx2mod3a          Suelosvs2
##          2.776276          1.191884
##          tdmmod3a VerticalDistanceToChannelNetwork
##          4.994918          2.762108
##          tdhmod3a          tx3mod3a
##          2.497728          3.227841
##          tx6mod3a          RelativeSlopePosition
##          2.647150          2.603009
##          tx4mod3a          tdlmod3a
##          2.807851          2.699247
##          twisre3a          ganhws3a
##          1.948273          1.208840
##          tnhmod3a          DEMSRE3a
##          6.717092          44.395289
##          DEM          tnmmod3a
##          44.888187          5.907536
##          inssre3a          inmsre3a
##          3.698152          2.400067
##          Bioclivs1          Pisosvs9
##          1.404724          1.681186
##          Pisosvs3
##          1.118481
```

Eliminamos del MRL multiple las covariables con Multicolinealidad.

```
modelo.MLR.step <- update(modelo.MLR.step, . ~ . - DEM -tnhmod3a -tdmmod3a -inssre3a -tx3mod3a -tnmmod3a)
```

Revisamos de nuevo la multicolinealidad.

```
sqrt(vif(modelo.MLR.step))
```

```
##          tx2mod3a          Suelosvs2
##          2.319083          1.187438
##          VerticalDistanceToChannelNetwork          tdmmod3a
##          2.707139          2.240357
##          tx6mod3a          RelativeSlopePosition
##          2.577834          2.452224
##          tx4mod3a          tdlmod3a
##          2.404800          2.368837
##          twisre3a          ganhws3a
##          1.772470          1.197710
##          DEMSRE3a          inmsre3a
##          2.822557          1.775531
##          Bioclivs1          Pisosvs9
##          1.334821          1.632872
```

```

##                               Pisosvs3
##                               1.099501

#Vamos usar la prueba de Bonferroni para valores atipicos:

outlierTest(modelo.MLR.step)

##          rstudent unadjusted p-value Bonferonni p
## 6399   -5.099740          3.4628e-07   0.0035068
## 6943   -4.994096          6.0100e-07   0.0060863
## 9585   -4.910874          9.2091e-07   0.0093260
## 9671   -4.887380          1.0376e-06   0.0105070
## 10102  -4.746641          2.0966e-06   0.0212320
## 2219   -4.682257          2.8743e-06   0.0291080
## 7367   -4.577421          4.7636e-06   0.0482410

# Incorporamos las covariables requeridas, segun MRL Multiple.

topo <- stack('ECUtopo.tif')
namesTopo <- readRDS('namesTOP0.rds')
names(topo)

## [1] "ECUtopo.1" "ECUtopo.2" "ECUtopo.3" "ECUtopo.4" "ECUtopo.5"
## [6] "ECUtopo.6" "ECUtopo.7" "ECUtopo.8" "ECUtopo.9" "ECUtopo.10"
## [11] "ECUtopo.11" "ECUtopo.12" "ECUtopo.13" "ECUtopo.14" "ECUtopo.15"

names(topo) <- namesTopo
names(topo)

## [1] "DEM" "AnalyticalHillshading"
## [3] "Slope" "Aspect"
## [5] "CrossSectionalCurvature" "LongitudinalCurvature"
## [7] "CovergenceIndex" "ClosedDepressions"
## [9] "FlowAccumulation" "TopographicWetnessIndex"
## [11] "LSFactor" "ChannelNetworkBaseLevel"
## [13] "VerticalDistanceToChannelNetwork" "ValleyDepth"
## [15] "RelativeSlopePosition"

cov <- stack('ECU_worldgridsCOVS.tif')
namesCov <- readRDS('worldgridsCOVS_names.rds')
#names(cov)
names(cov) <- namesCov
#names(cov)

# Incorporamos las covariables categoricas y las adecuamos al resto.

Suelosvs <- raster('Covariables/Suelosvs.tif')
Suelosvs <- resample(Suelosvs,topo, method = 'ngb')
Bioclivs <- raster('Covariables/Bioclivs.tif')
Bioclivs <- resample(Bioclivs,topo, method = 'ngb')
Climavs <- raster('Covariables/Climavs.tif')
Climavs <- resample(Climavs,topo, method = 'ngb')
Cobervs <- raster('Covariables/Cobervs.tif')
Cobervs <- resample(Cobervs,topo, method = 'ngb')
Pisosvs <- raster('Covariables/Pisosvs.tif')
Pisosvs <- resample(Pisosvs,topo, method = 'ngb')

```

```
# Convertimos las covariables categoricas a dummy
```

```
Suelosvs_du <- dummyRaster(Suelosvs)
Bioclivs_du <- dummyRaster(Bioclivs)
Climavs_du <- dummyRaster(Climavs)
Cobervs_du <- dummyRaster(Cobervs)
Pisosvs_du <- dummyRaster(Pisosvs)
```

```
# Apilamos todas las covariables.
```

```
COV <- stack(topo, cov, Suelosvs_du, Bioclivs_du, Climavs_du, Cobervs_du, Pisosvs_du)
```

```
# Observamos los nombres de los campos o columnas.
```

```
names(COV)
```

```
## [1] "AnalyticalHillshading"
## [3] "Slope" "Aspect"
## [5] "CrossSectionalCurvature" "LongitudinalCurvature"
## [7] "CovergenceIndex" "ClosedDepressions"
## [9] "FlowAccumulation" "TopographicWetnessIndex"
## [11] "LSFactor" "ChannelNetworkBaseLevel"
## [13] "VerticalDistanceToChannelNetwork" "ValleyDepth"
## [15] "RelativeSlopePosition" "cntgad3a"
## [17] "DEMSRE3a" "etmnts3a"
## [19] "evmmod3a" "evsmmod3a"
## [21] "g01esa3a" "g01igb3a"
## [23] "g02esa3a" "g02igb3a"
## [25] "g03esa3a" "g04esa3a"
## [27] "g04igb3a" "g05esa3a"
## [29] "g06esa3a" "g07esa3a"
## [31] "g08esa3a" "g09esa3a"
## [33] "g10esa3a" "g10igb3a"
## [35] "g11esa3a" "g11igb3a"
## [37] "g12esa3a" "g12igb3a"
## [39] "g13esa3a" "g14esa3a"
## [41] "g15esa3a" "g16esa3a"
## [43] "g17esa3a" "g18esa3a"
## [45] "g19esa3a" "g20esa3a"
## [47] "g21esa3a" "g22esa3a"
## [49] "gabhws3a" "gacgem3a"
## [51] "gachws3a" "galhws3a"
## [53] "ganhws3a" "garhws3a"
## [55] "gathws3a" "gchhws3a"
## [57] "gclhws3a" "gcmhws3a"
## [59] "gcrhws3a" "geaisg3a"
## [61] "gflhws3a" "gfrhws3a"
## [63] "gglhws3a" "ggyhws3a"
## [65] "ghshws3a" "gkshws3a"
## [67] "glcesa3a" "glcjrc3a"
## [69] "glphws3a" "glvhws3a"
## [71] "glwwf3a" "glxhws3a"
## [73] "gnthws3a" "gphhws3a"
## [75] "gplhws3a" "gpthws3a"
```

```
## [77] "gpzhws3a"      "grghws3a"
## [79] "gschws3a"      "gsnhws3a"
## [81] "gsthws3a"      "gumhws3a"
## [83] "gvrhws3a"      "iflgre3a"
## [85] "inmsre3a"      "inssre3a"
## [87] "l01igb3a"      "l02igb3a"
## [89] "l03igb3a"      "l04igb3a"
## [91] "l05igb3a"      "l06igb3a"
## [93] "l07igb3a"      "l08igb3a"
## [95] "l09igb3a"      "l10igb3a"
## [97] "l11igb3a"      "l12igb3a"
## [99] "l13igb3a"      "l14igb3a"
## [101] "l15igb3a"      "l16igb3a"
## [103] "l3pobi3b"      "lammod3a"
## [105] "lasmod3a"      "lmbgsh3a"
## [107] "lmtgsh3a"      "ln1dms3a"
## [109] "ln2dms3a"      "lnmdms3a"
## [111] "opisre3a"      "px1wcl3a"
## [113] "px2wcl3a"      "px3wcl3a"
## [115] "px4wcl3a"      "SLPSRT3a"
## [117] "smkisir3a"     "tdhmod3a"
## [119] "tdlmod3a"      "tdmmod3a"
## [121] "tdsmod3a"      "tnhmod3a"
## [123] "tnlmod3a"      "tnmmod3a"
## [125] "tnsmod3a"      "twisre3a"
## [127] "tx1mod3a"      "tx2mod3a"
## [129] "tx3mod3a"      "tx4mod3a"
## [131] "tx5mod3a"      "tx6mod3a"
## [133] "wmkmod3a"      "Suelosvs1"
## [135] "Suelosvs2"     "Suelosvs3"
## [137] "Suelosvs4"     "Suelosvs5"
## [139] "Suelosvs6"     "Suelosvs7"
## [141] "Suelosvs8"     "Suelosvs9"
## [143] "Suelosvs10"    "Suelosvs11"
## [145] "Bioclivs1"     "Bioclivs2"
## [147] "Bioclivs3"     "Bioclivs4"
## [149] "Climavs1"      "Climavs2"
## [151] "Climavs3"      "Climavs4"
## [153] "Climavs5"      "Climavs6"
## [155] "Climavs7"      "Climavs8"
## [157] "Climavs9"      "Cobervs1"
## [159] "Cobervs2"      "Cobervs3"
## [161] "Cobervs4"      "Cobervs5"
## [163] "Cobervs6"      "Cobervs7"
## [165] "Pisosvs1"      "Pisosvs2"
## [167] "Pisosvs3"      "Pisosvs4"
## [169] "Pisosvs5"      "Pisosvs7"
## [171] "Pisosvs8"      "Pisosvs9"
```

```
# Seleccionamos solo las primeras 25 covariables de mayor correlacion.
```

```
COV <- COV[[idx]]
```

```
# Observamos los nombres de los campos o columnas.
```

```
names(COV)
```

```
## [1] "tx2mod3a"          "Suelosvs2"
## [3] "tx1mod3a"          "tdmmod3a"
## [5] "VerticalDistanceToChannelNetwork" "tdhmod3a"
## [7] "tx5mod3a"          "tx3mod3a"
## [9] "tx6mod3a"          "RelativeSlopePosition"
## [11] "tx4mod3a"          "tdlmod3a"
## [13] "twisre3a"          "ganhws3a"
## [15] "tnhmod3a"          "DEMSRE3a"
## [17] "DEM"               "tnmmod3a"
## [19] "inssre3a"          "Bioclivs3"
## [21] "inmsre3a"          "Bioclivs1"
## [23] "ValleyDepth"       "Pisosvs9"
## [25] "Pisosvs3"
```

```
# Cambiamos resolucion espacial de las covariables solo para verlos.
# En la corrida final se debe dejar en la resolucion original de 1 km.
```

```
COV <- aggregate(COV, 10)
```

```
# Adecuamos proyeccion cartograficas.
# Projectamos puntos de datos.
```

```
dat_sp@proj4string <- COV@crs
dat_sp <- spTransform(dat_sp, CRS("+init=epsg:32717"))
COV <- projectRaster(COV, crs = CRS("+init=epsg:32717"), method='ngb')
```

```
# Convertimos las covariables a tabla de datos espaciales.
```

```
COV.sp <- as(COV, "SpatialGridDataFrame")
```

```
## Eliminamos Datos duplicados.
```

```
zerodist(dat_sp)
```

```
##      [,1] [,2]
## [1,]    3    4
## [2,]    3    5
## [3,]    4    5
## [4,]    2    7
## [5,]   11   12
## [6,]   11   13
## [7,]   12   13
## [8,]   11   14
## [9,]   12   14
## [10,]  13   14
## [11,]  11   15
## [12,]  12   15
## [13,]  13   15
## [14,]  14   15
## [15,]  11   16
## [16,]  12   16
## [17,]  13   16
```



```
## [2124,] 9832 9833
## [2125,] 9834 9835
## [2126,] 9857 9858
## [2127,] 9841 9861
## [2128,] 9862 9863
## [2129,] 9865 9866
## [2130,] 9867 9868
## [2131,] 9854 9869
## [2132,] 9844 9870
## [2133,] 9875 9876
## [2134,] 9850 9878
## [2135,] 9885 9886
## [2136,] 9896 9897
## [2137,] 9902 9903
## [2138,] 9935 9936
## [2139,] 9929 9938
## [2140,] 9929 9939
## [2141,] 9938 9939
## [2142,] 9920 9944
## [2143,] 9948 9949
## [2144,] 9927 9961
## [2145,] 9925 9965
## [2146,] 9966 9967
## [2147,] 9960 9975
## [2148,] 9976 9977
## [2149,] 9978 9986
## [2150,] 9997 9998
## [2151,] 9925 10007
## [2152,] 9965 10007
## [2153,] 10008 10009
## [2154,] 10011 10012
## [2155,] 10014 10020
## [2156,] 10028 10029
## [2157,] 10049 10050
## [2158,] 10055 10056
## [2159,] 10055 10057
## [2160,] 10056 10057
## [2161,] 10062 10063
## [2162,] 10064 10065
## [2163,] 10062 10068
## [2164,] 10063 10068
## [2165,] 10085 10090
## [2166,] 10104 10105
## [2167,] 10107 10108
## [2168,] 10098 10109
## [2169,] 10114 10127
```

```
dat_sp <- dat_sp[dat_sp$OCSKGM30 != 0,]
```

```
# Ejecutamos estimacion del COS segun ecuacion de RLM Multiple y el kriging
# de los residuos para la parte continental del Ecuador.
```

```
start <- Sys.time()
OCS.krige <- autoKrige(formula = as.formula(modelo.MLR.step$call$formula),
```

```

input_data = dat_sp,
new_data = COV.sp,
verbose = TRUE,
block = c(1000, 1000),
model = c("Sph", "Exp")

```

```

## Warning in autoKrige(formula = as.formula(modelo.MLR.step$call$formula), :
## Removed 2169 duplicate observation(s) in input_data:

```

##	coordinates	ID1	ID
## 3	(753711.4, 10141590)	837	CG4-P158_1.28_-78.72
## 3.1	(753711.4, 10141590)	837	CG4-P158_1.28_-78.72
## 4	(753711.4, 10141590)	10089	PN2-P267_1.28_-78.72
## 2	(753723.5, 10127210)	39	CG1-P021_1.15_-78.72
## 11	(725898.1, 10116130)	1552	CL6-P127_1.05_-78.97
## 11.1	(725898.1, 10116130)	1552	CL6-P127_1.05_-78.97
## 12	(725898.1, 10116130)	1568	CL6-P143_1.05_-78.97
## 11.2	(725898.1, 10116130)	1552	CL6-P127_1.05_-78.97
## 12.1	(725898.1, 10116130)	1568	CL6-P143_1.05_-78.97
## 13	(725898.1, 10116130)	1840	C02-P016_1.05_-78.97
## 11.3	(725898.1, 10116130)	1552	CL6-P127_1.05_-78.97
## 12.2	(725898.1, 10116130)	1568	CL6-P143_1.05_-78.97
## 13.1	(725898.1, 10116130)	1840	C02-P016_1.05_-78.97
## 14	(725898.1, 10116130)	2001	C09-P038_1.05_-78.97
## 11.4	(725898.1, 10116130)	1552	CL6-P127_1.05_-78.97
## 12.3	(725898.1, 10116130)	1568	CL6-P143_1.05_-78.97
## 13.2	(725898.1, 10116130)	1840	C02-P016_1.05_-78.97
## 14.1	(725898.1, 10116130)	2001	C09-P038_1.05_-78.97
## 15	(725898.1, 10116130)	8657	PM1-P076_1.05_-78.97
## 17	(714765.7, 10116120)	9827	PN1-P241_1.05_-79.07
## 11.5	(725898.1, 10116130)	1552	CL6-P127_1.05_-78.97
## 12.4	(725898.1, 10116130)	1568	CL6-P143_1.05_-78.97
## 13.3	(725898.1, 10116130)	1840	C02-P016_1.05_-78.97
## 14.2	(725898.1, 10116130)	2001	C09-P038_1.05_-78.97
## 15.1	(725898.1, 10116130)	8657	PM1-P076_1.05_-78.97
## 16	(725898.1, 10116130)	8659	PM1-P077_1.05_-78.97
## 17.1	(714765.7, 10116120)	9827	PN1-P241_1.05_-79.07
## 18	(714765.7, 10116120)	9829	PN1-P271_1.05_-79.07
## 17.2	(714765.7, 10116120)	9827	PN1-P241_1.05_-79.07
## 18.1	(714765.7, 10116120)	9829	PN1-P271_1.05_-79.07
## 21	(714765.7, 10116120)	9937	PN2-P130_1.05_-79.07
## 17.3	(714765.7, 10116120)	9827	PN1-P241_1.05_-79.07
## 18.2	(714765.7, 10116120)	9829	PN1-P271_1.05_-79.07
## 21.1	(714765.7, 10116120)	9937	PN2-P130_1.05_-79.07
## 22	(714765.7, 10116120)	9962	PN2-P155_1.05_-79.07
## 17.4	(714765.7, 10116120)	9827	PN1-P241_1.05_-79.07
## 18.3	(714765.7, 10116120)	9829	PN1-P271_1.05_-79.07
## 21.2	(714765.7, 10116120)	9937	PN2-P130_1.05_-79.07
## 22.1	(714765.7, 10116120)	9962	PN2-P155_1.05_-79.07
## 23	(714765.7, 10116120)	9964	PN2-P157_1.05_-79.07
## 17.5	(714765.7, 10116120)	9827	PN1-P241_1.05_-79.07
## 18.4	(714765.7, 10116120)	9829	PN1-P271_1.05_-79.07
## 21.3	(714765.7, 10116120)	9937	PN2-P130_1.05_-79.07
## 22.2	(714765.7, 10116120)	9962	PN2-P155_1.05_-79.07

```

## 2442      0      0      0
## 2450      0      0      0
## 2463      0      0      0
## 2480      0      0      0
## 2493      0      0      0
## 2504      0      0      0
## 2508      0      0      0
## 2512      0      0      0
## 2525      0      0      0
## 2533      0      0      0
## 2541      0      0      0
## 2548      0      0      0
## 2556      0      0      0
## 2568      0      0      0
## 2574      0      0      0
## 2573      0      0      0
## 2574.1    0      0      0
## 2575      0      0      0
## 2585      0      0      0
## 2574.2    0      0      0
## 2575.1    0      0      0
## 2579      0      0      0
## 2574.3    0      0      0
## 2575.2    0      0      0
## 2579.1    0      0      0
## 2591      0      0      0
## 2574.4    0      0      0
## [ reached 'max' / getOption("max.print") -- omitted 3619 rows ]
## Checking if any bins have less than 5 points, merging bins when necessary...
##
## Selected:
##   model      psill      range
## 1   Nug 0.24464055      0.00
## 2   Exp 0.08663018 10391.67
##
## Tested models, best first:
##   Tested.models kappa      SSError
## 2           Exp      0 4.429738e-08
## 1           Sph      0 1.033833e-07
## [using universal kriging]
print(Sys.time() - start)

## Time difference of 5.77796 mins
# Devolvemos los valores de CDS a su condcion original.

RKprediction <- exp(raster(OCS.krige$krige_output[1]))
RKpredsd <- exp(raster(OCS.krige$krige_output[3]))

# Vemos el resumen estadistico de los resultados en kg/m2.

summary(RKprediction)

##           layer

```

```
## Min.      1.296340
## 1st Qu.   3.796010
## Median    4.196352
## 3rd Qu.   6.353187
## Max.      22.313380
## NA's      16533.000000
```

```
summary(RKpredsd)
```

```
##           layer
## Min.      1.146581
## 1st Qu.   1.224752
## Median    1.271055
## 3rd Qu.   1.326338
## Max.      1.338849
## NA's      16533.000000
```

```
# Si existen valores atipico se pueden eliminar aqui.
```

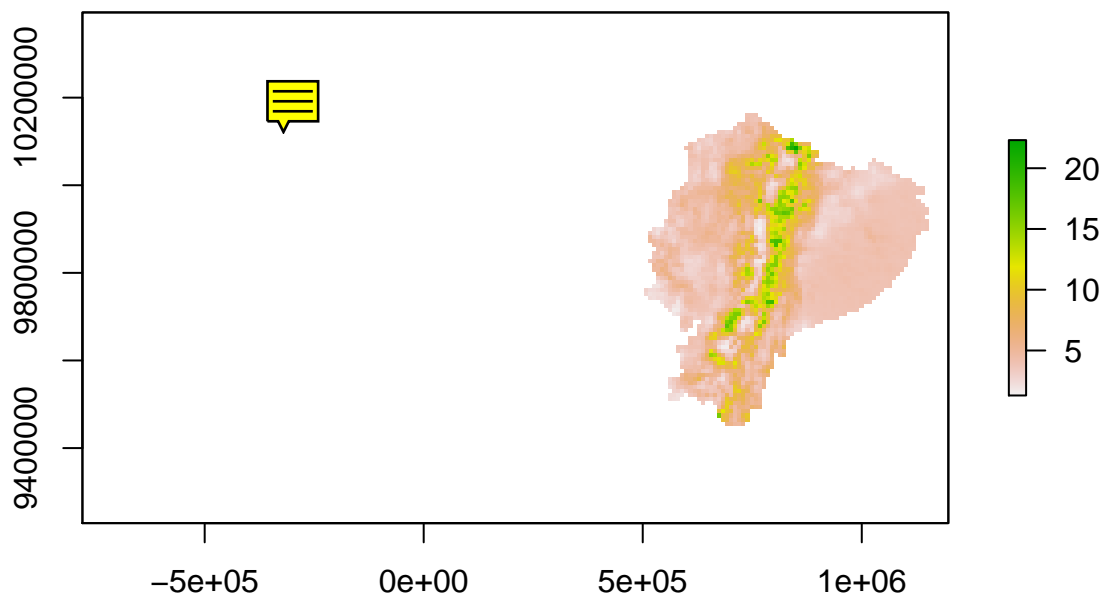
```
#values(RKprediction )[values(RKprediction ) < 0]  <- NA
#values(RKprediction )[values(RKprediction ) > 100] <- NA
#values(RKpredsd)[values(RKpredsd ) > 10]  <- NA
```

```
# Vemos el resumen estadistico de los resultados en kg/m2.
```

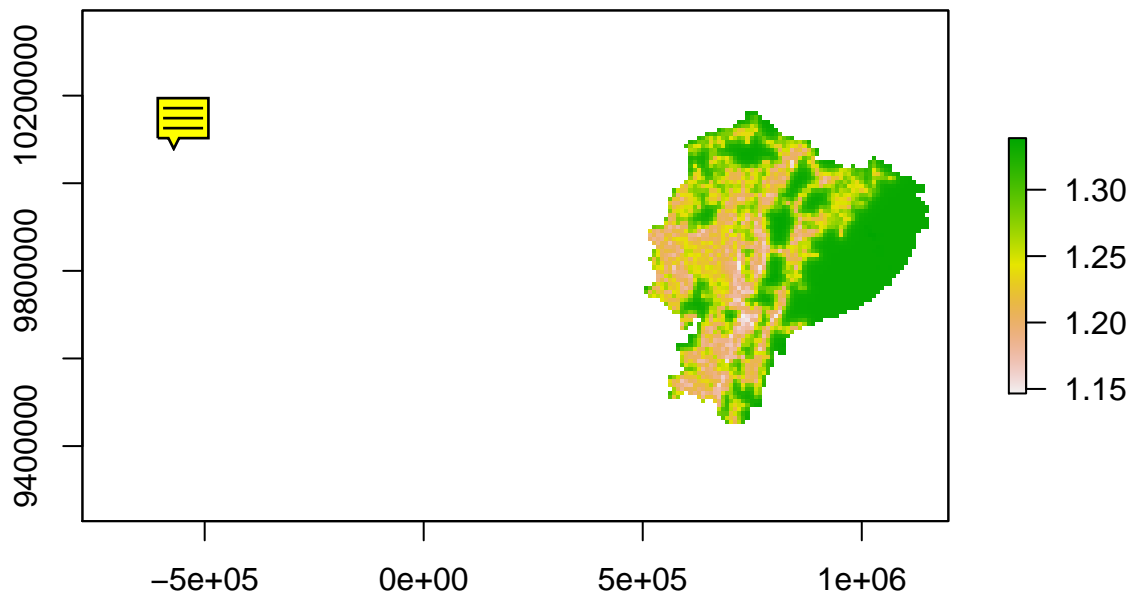
```
#summary(RKprediction)
#summary(RKpredsd)
```

```
# Graficamos los resultados.
```

```
plot(RKprediction)
```



```
plot(RKpredsd)
```



```
# Reproyectamos la prediccion a geografica WGS84.
```

```
RKprediction_geo <- projectRaster(RKprediction, crs = CRS("+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84"))
```

```
# Guardamos los resultados en archivos tiff.
```

```
#writeRaster(RKprediction, filename = "ECU_OCS_RK_kgm2.tif")
```

```
#writeRaster(RKprediction, filename = "ECU_OCS_RK_kgm2a.asc")
```

```
#writeRaster(RKprediction_geo, filename = "ECU_OCS_RK_kgm2_geo.asc")
```

```
#writeRaster(RKprediction_geo, filename = "ECU_OCS_RK_kgm2_geot.tif")
```

```
#writeRaster(RKpredsd, filename = "ECU_OCS_RKpredsd_kgm2.tif")
```

```
# Convertimos los resultados de kg/m2 a Tn/ha.
```

```
# Importamos el raster resultados
```

```
r1 <- raster ('ECU_OCS_RK_kgm2.tif')
```

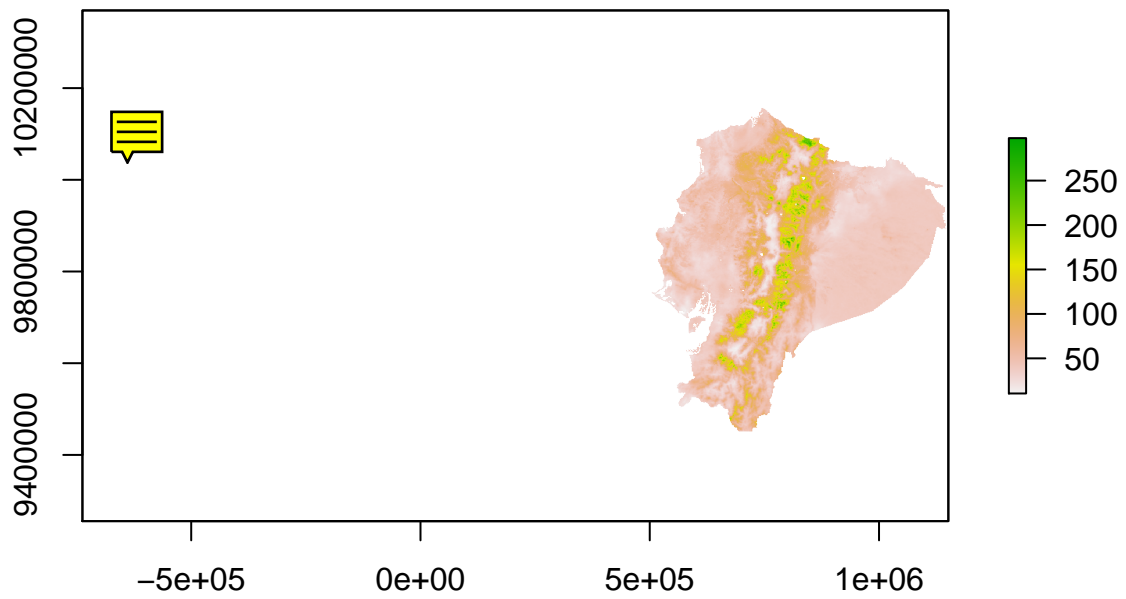
```
r2 <- r1 *10
```

```
r3 <- raster ('ECU_OCS_RKpredsd_kgm2.tif')
```

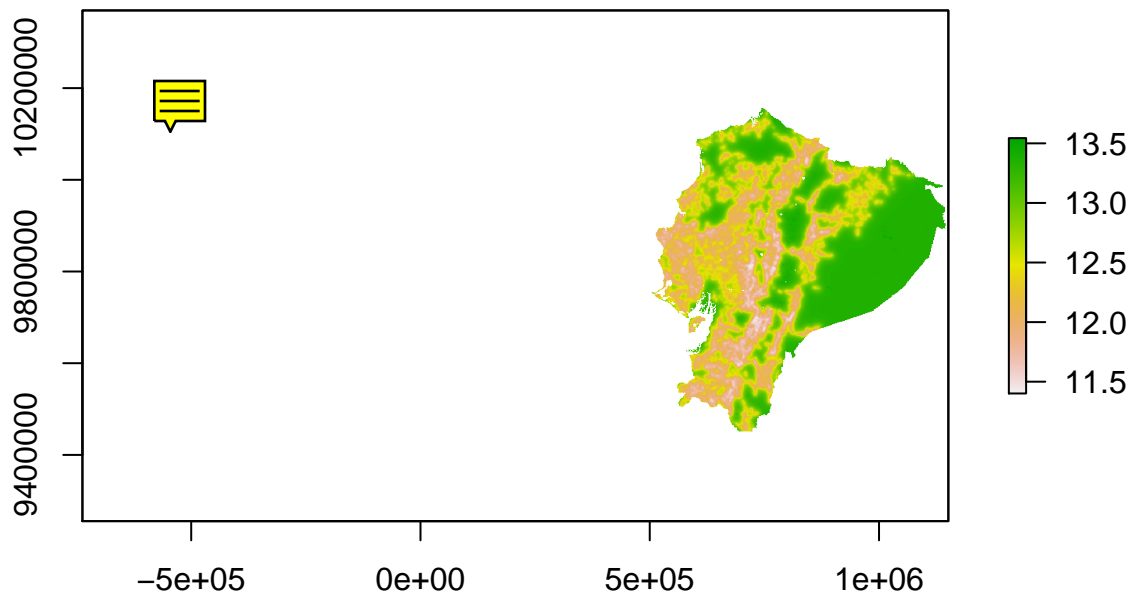
```
r4 <- r3 *10
```

```
# Graficamos los resultados en Tn/ha.
```

```
plot(r2)
```



```
plot(r4)
```



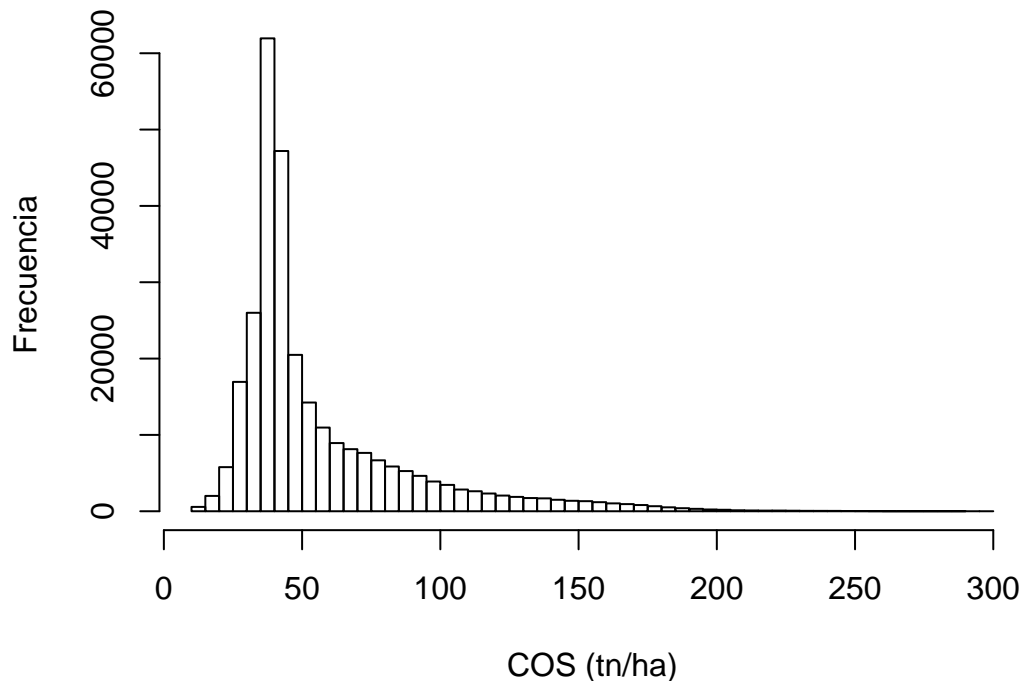
Resumen del mapa de COS en tn/ha.

```
summary(r2)
```

```
##          ECU_OCS_RK_kgm2
## Min.      1.044161e+01
## 1st Qu.   3.759635e+01
## Median    4.226499e+01
## 3rd Qu.   6.507849e+01
## Max.      2.978068e+02
## NA's      1.398030e+06
```

```
hist(r2, breaks = 100, main = "Histograma de frecuencia de COS en mapa de RK (tn/ha)", xlab= 'COS (tn/ha)')
```


Histograma de frecuencia de COS en mapa de RK (tn/ha)



Histograma sobre datos de COS producto de Regresion-Kriging

```
# Reproyectamos la prediccion a geografica WGS84.
```

```
r2_geo <- projectRaster(r2, crs = CRS("+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0"))
r4_geo <- projectRaster(r4, crs = CRS("+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0"))
```

```
# Se guarda en formato tif.
```

```
#writeRaster(r2, 'ECU_Mapa_COS_tnha.tif')
#writeRaster(r4, 'ECU_Mapa_COS_Res_tnha.tif')
#writeRaster(r4_geo, 'ECU_Mapa_COS_Res_tnha_geo.tif')
#writeRaster(r2_geo, 'ECU_Mapa_COS_tnha_geo.tif')
```

```
# Ejecutamos estimacion del COS segun ecuacion de RLM Multiple y el kriging
# de los residuos para las Islas Galapagos.
```

```
start <- Sys.time()
```

```
OCS.krige.g <- autoKrige(formula = log(OCSKGM30) ~ tx2mod3a + tdhmod3a + tx4mod3a + ganhws3a + tx6mod3a +
                        DEMSRE3a + VerticalDistanceToChannelNetwork + RelativeSlopePosition + twisre3a,
                        input_data = dat_sp,
                        new_data = COV.sp,
                        verbose = TRUE,
                        block = c(1000, 1000),
                        model = c("Sph", "Exp"))
```

```
## Warning in autoKrige(formula = log(OCSKGM30) ~ tx2mod3a + tdhmod3a +
## tx4mod3a + : Removed 2169 duplicate observation(s) in input_data:
```

```
## coordinates ID1 ID
```

```

## 2541      0      0      0
## 2548      0      0      0
## 2556      0      0      0
## 2568      0      0      0
## 2574      0      0      0
## 2573      0      0      0
## 2574.1    0      0      0
## 2575      0      0      0
## 2585      0      0      0
## 2574.2    0      0      0
## 2575.1    0      0      0
## 2579      0      0      0
## 2574.3    0      0      0
## 2575.2    0      0      0
## 2579.1    0      0      0
## 2591      0      0      0
## 2574.4    0      0      0
## [ reached 'max' / getOption("max.print") -- omitted 3619 rows ]
## Checking if any bins have less than 5 points, merging bins when necessary...
##
## Selected:
##   model    psill    range
## 1   Nug 0.2431701    0.00
## 2   Exp 0.1039212 10433.87
##
## Tested models, best first:
##   Tested.models kappa    SSError
## 2           Exp      0 7.114677e-08
## 1           Sph      0 1.906273e-07
## [using universal kriging]
print(Sys.time() - start)

## Time difference of 6.066448 mins
# Devolvemos los valores de COS a su condicon original.

RKprediction.g <- exp(raster(OCS.krige.g$krige_output[1]))
RKpredsd.g <- exp(raster(OCS.krige.g$krige_output[3]))

# Vemos el resumen estadistico de los resultados en kg/m2.

summary(RKprediction.g)

##           layer
## Min.      1.271960
## 1st Qu.   3.728430
## Median   4.214753
## 3rd Qu.   6.109628
## Max.     21.034119
## NA's     16350.000000

summary(RKpredsd.g)

##           layer
## Min.      1.154968

```

```
## 1st Qu.    1.244228
## Median    1.301377
## 3rd Qu.   1.367798
## Max.      1.384404
## NA's      16350.000000
```

```
# Si existen valores atipico se pueden eliminar aqui.
```

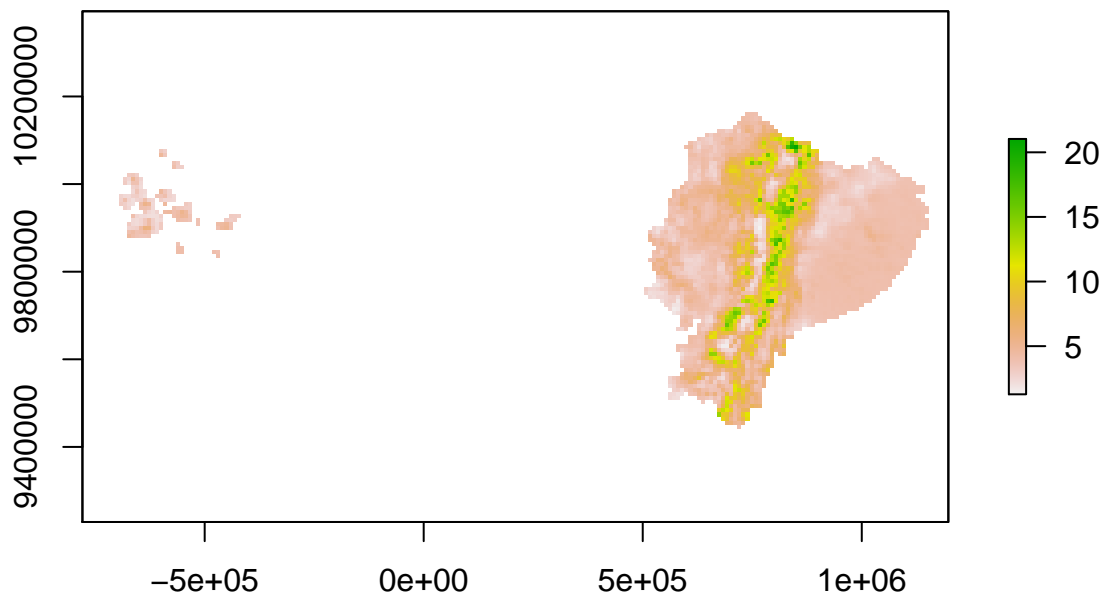
```
#values(RKprediction.g)[values(RKprediction.g) < 0] <- NA
#values(RKprediction.g)[values(RKprediction.g) > 100] <- NA
#values(RKpredsd.g)[values(RKpredsd.g) > 10] <- NA
```

```
# Vemos el resumen estadistico de los resultados en kg/m2.
```

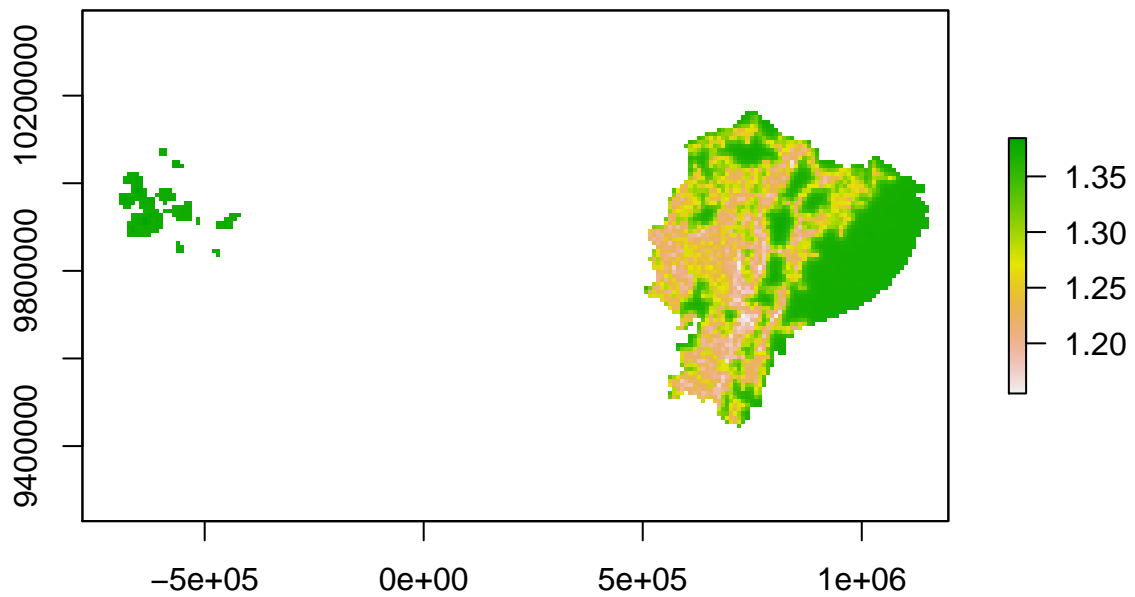
```
#summary(RKprediction.g)
#summary(RKpredsd.g)
```

```
# Graficamos los resultados.
```

```
plot(RKprediction.g)
```



```
plot(RKpredsd.g)
```



```
# Guardamos los resultados en archivos tiff.

# writeRaster(RKprediction.g, filename = "ECU_OCS_RK_G_kgm2.tif")
# writeRaster(RKpredsd.g, filename = "ECU_OCS_RKpredsd_G_kgm2.tif")

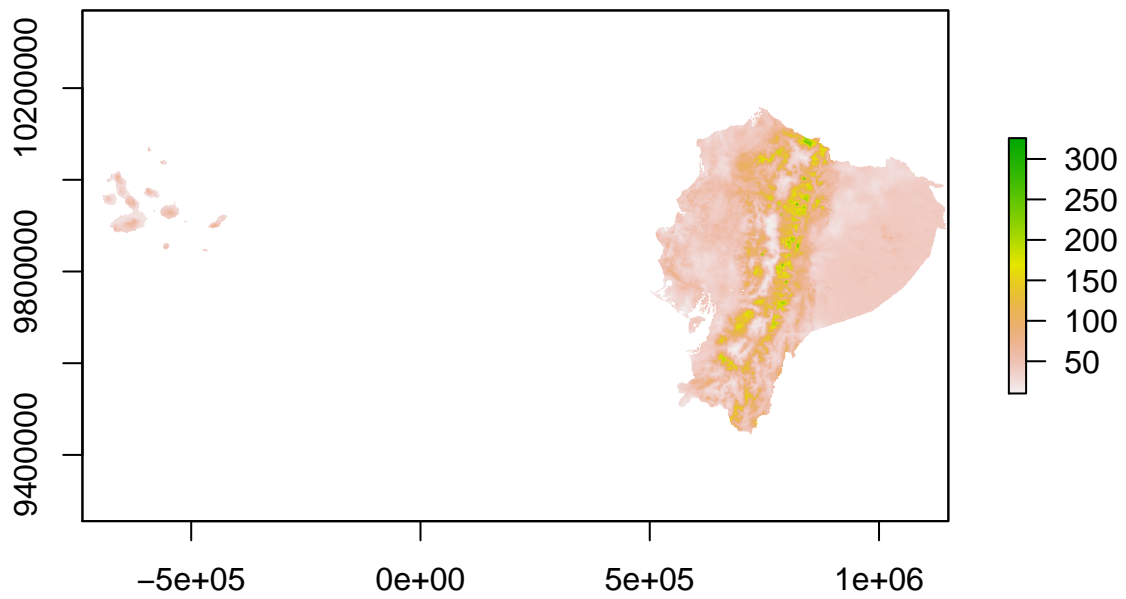
# Convertimos los resultados de kg/m2 a Tn/ha.
# Importamos el raster resultados

r1.g <- raster ('ECU_OCS_RK_G_kgm2.tif')
r2.g <- r1.g *10

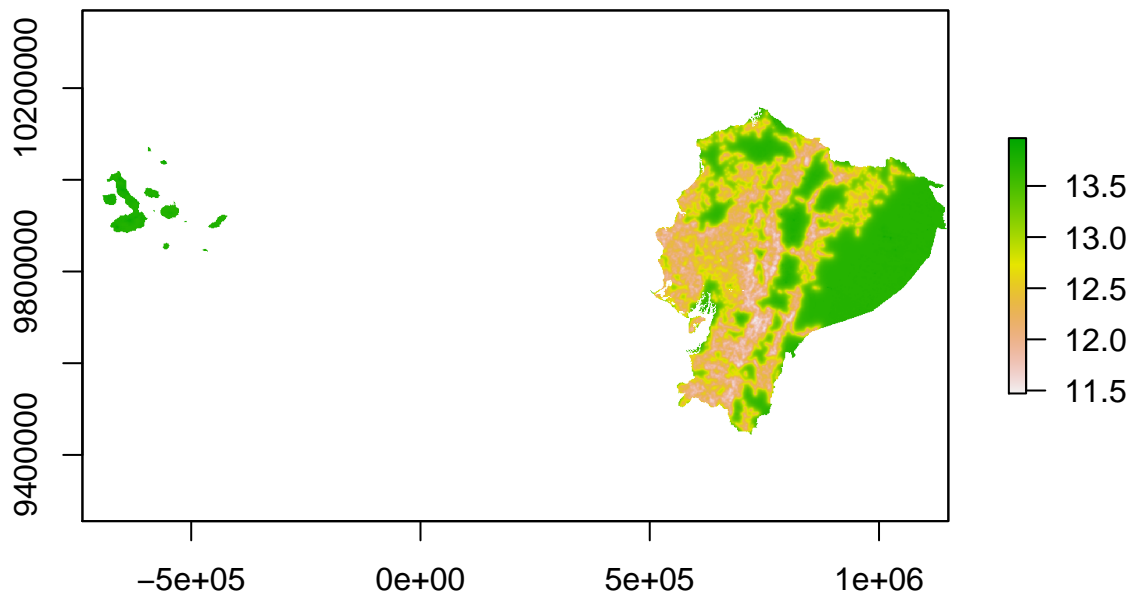
r3.g <- raster ('ECU_OCS_RKpredsd_G_kgm2.tif')
r4.g <- r3.g *10

# Graficamos los resultados en Tn/ha.

plot(r2.g)
```



```
plot(r4.g)
```



```
r2.g_geo <- projectRaster(r2.g, crs = CRS("+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0"))
r4.g_geo <- projectRaster(r4.g, crs = CRS("+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0"))
```

```
# Se guarda en formato tif.
writeRaster(r2.g, 'ECU_Mapas_COS_G_tinha.tif')
writeRaster(r4.g, 'ECU_Mapas_COS_Res_G_tinha.tif')
writeRaster(r2.g_geo, 'ECU_Mapas_COS_G_tinha_geot.tif')
writeRaster(r4.g_geo, 'ECU_Mapas_COS_Res_G_tinha_geot.tif')
```

```
# Estimacion de la incertidumbre segun validacion cruzada.
# Eliminamos datos duplicados.
```

```
dat_sp = dat_sp[which(!duplicated(dat_sp@coords)), ]
```

```
# Corremos la validacion cruzada.
```

```
OCS.krige.cv <- autoKriging(formula = as.formula(modelo.MLR.step$call$formula),
                           input_data = dat_sp, nfold = 5)
```

```
##
```

```
|
|
|
|=====| 0%
|
|=====| 25%
```

```
|=====| 50%
|
|=====| 75%
|
|=====| 100%
```

```
# Vemos un resumen estadístico de la validación cruzada.
```

```
summary(OCS.krige.cv)
```

```
##           [,1]
## mean_error -2.281e-06
## me_mean    -1.544e-06
## MAE         0.394
## MSE         0.2763
## MSNE        0.969
## cor_ospred  0.6335
## cor_predres 0.005655
## RMSE        0.5257
## RMSE_sd     0.7737
## URMSE       0.5257
## iqr         0.6012
```

```
#=====
# ===== SCRIPT PARA VALIDACION EXTERNA =====
# ===== ECUADOR =====
#=====
```

```
# Para esta validación se emplearon los 1000 puntos dejados fuera de la calibración
# del modelo de Regresión - Kriging.
```

```
# Cargamos los datos de los perfiles de validación.
```

```
datv <- read.csv("ecu_vali8.csv", header = TRUE, sep = ",")
```

```
# Observamos los nombres de los campos o columnas.
```

```
names(datv)
```

```
## [1] "Id1"      "Id"      "Latitude" "Longitude" "Ocskgm30"
```

```
# Vemos un resumen de los datos de COS (Kg/m2) de los perfiles de validación.
```

```
summary(datv$Ocskgm30)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
## 0.1376  3.0513  4.4866  5.3964  6.2934 37.7296
```

```
## Recreamos el objeto con la ubicación de los puntos.
```

```
coordinates(datv) <- ~ Longitude + Latitude
```

```
# Adecuamos proyección cartográficas.
```

```
# Project point data
```

```
datv@proj4string <- CRS(projargs = "+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0")
OCSKGM_RK <- raster("ECU_OCS_RK_kgm2_geot.tif")
```

```

# Extraemos los datos de COS en kg/m2 de la capa estimada para los puntos de validacion.
datv <- extract(x = OCSKGM_RK, y = datv, sp = TRUE)

# Calculamos la diferencia entre los valores de COS medidos y los COS estimados.
datv$PE_RK <- datv$ECU_OCS_RK_kgm2_geot - datv$Ocskgm30

# Guardamos los resultados de esta validacion.

# write.csv(datv, "Ecu_validacion8.csv", row.names = F)

# Exponemos un resumen de los errores de prediccion.
summary(res_rk <- abs(datv$ECU_OCS_RK_kgm2_geot - datv$Ocskgm30))

##      Min.   1st Qu.   Median     Mean  3rd Qu.    Max.
## 0.00027  0.56144  1.30736  1.89534  2.33010 34.73290

# Estimacion de las medidas de calidad del mapa.

# Calculamos el cuartil 75%.
s <- quantile(res_rk, .75, na.rm=TRUE)

# Calculamos e imprimimos el error medio cuadrado entre el valor predicho
# y el valor medido.
a <- (rmse(datv$ECU_OCS_RK_kgm2_geot, datv$Ocskgm30))

# Calculamos el R2 entre los valores estimados o predichos y los medidos u observados.
g <- (cor(datv$ECU_OCS_RK_kgm2_geot, datv$Ocskgm30)^2)

# Calculamos el Error medio de todos los puntos de validacion.
ME_RK <- mean(datv$PE_RK, na.rm=TRUE)

# Calculamos el error promedio absoluto (MAE).
MAE_RK <- mean(abs(datv$PE_RK), na.rm=TRUE)

# Calculamos el cuadrado del error promedio (MSE).
MSE_RK <- mean(datv$PE_RK^2, na.rm=TRUE)

# Calculamos la raiz cuadrada del error promedio cuadrado (RMSE).
RMSE_RK <- sqrt(sum(datv$PE_RK^2, na.rm=TRUE) / length(datv$PE_RK))

# Estimamos la varianza explicada (Amount of Variance Explained (AVE)).
AVE_RK <- 1 - sum(datv$PE_RK^2, na.rm=TRUE) /
  sum( (datv$Ocskgm30 - mean(datv$Ocskgm30, na.rm = TRUE))^2,

```



```

na.rm = TRUE)

# Impresion de los errores.

metodo <- factor("Regresion-Kriging")
metodo <- data.frame(metodo)

resultados <- cbind(metodo, ME_RK, MAE_RK, MSE_RK, RMSE_RK, AVE_RK, s, g)
etiquetas <- c("Metodos", "ME", "MAE", "MSE", "RMSE", "AVE", "Err Q75", "R2")
names(resultados) <- etiquetas

print(resultados)

##              Metodos      ME      MAE      MSE      RMSE      AVE
## 75% Regresion-Kriging -0.517241 1.895339 8.988367 2.998061 0.3930775
##      Err Q75      R2
## 75% 2.330099 0.4120027

# Graficamos las medidas de calidad del mapa.

# Graficamos el Scatter.

par(mfrow=c(1,1))
plot(datv$ECU_OCS_RK_kgm2_geot, datv$Ocskgm30, main="Comparacion entre valores COS predichos por Regres",
      ylab='Valor COS observado', text(15,0.5, "La linea roja representa una pendiente de realcion 1: 1.
      valores observados y predichos."))

# Dibujamos una linea con relacion 1:1 color negro.

abline(0,1, lty=2, col='red')

# Establecemos una linea de regresion entre los valores estimados y los medidos color azul.

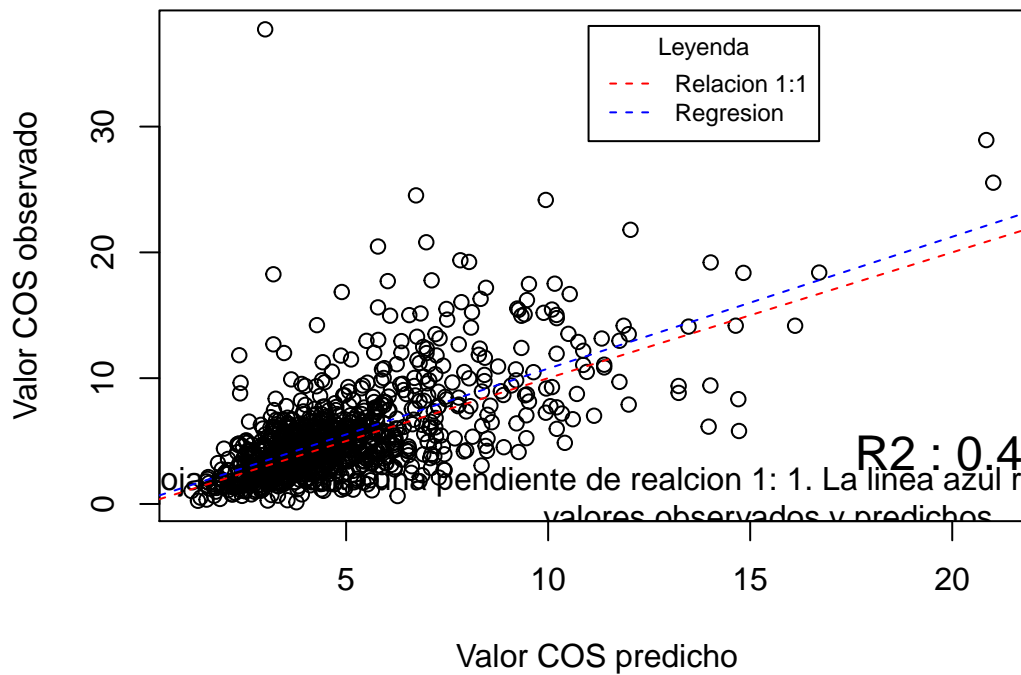
abline(lm(datv$Ocskgm30 ~ datv$ECU_OCS_RK_kgm2_geot), col = 'blue', lty=2)

legend(x = 11, y = 38, legend = c("Relacion 1:1", "Regresion"), col = c("Red", "Blue"),
      title = "Leyenda", lty = 2, cex = 0.75)

text(20,4,"R2 : 0.41", cex = 1.5)

```

e valores COS predichos por Regresion-Kriging y valores reales obser



```
# Graficamos las borbuja de prediccion espacial (Spatial bubbles for prediction errors).
```

```
Ecuador <- shapefile("ecu2.shp")
```

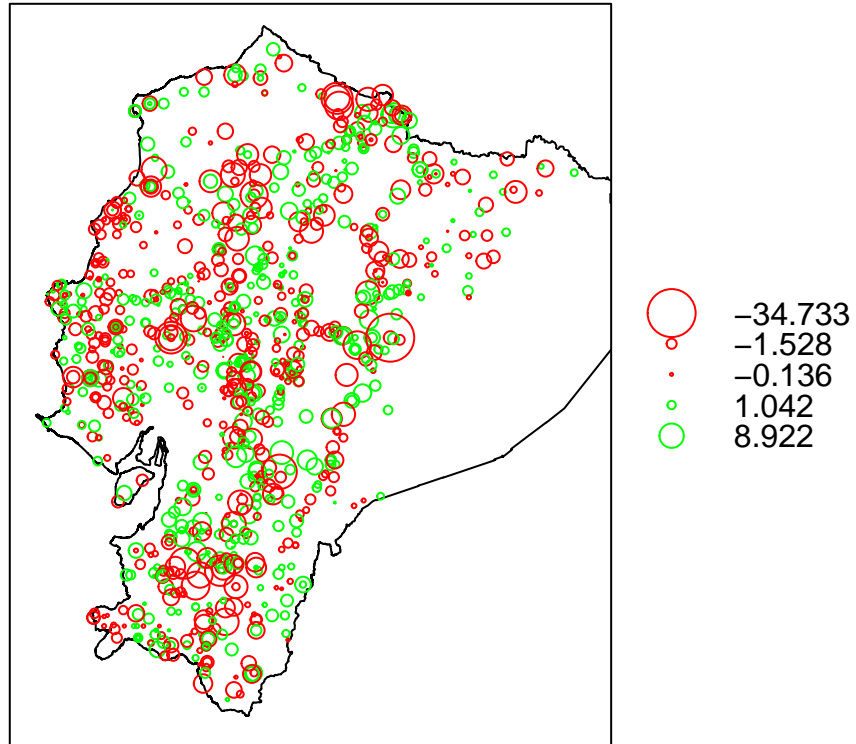
```
## Warning in .local(x, ...): .prj file is missing
```

```
Ecuador@proj4string <- CRS(projargs = "+proj=longlat +datum=WGS84 +no_defs +ellps=WGS84 +towgs84=0,0,0")
```

```
Ecuador.sp <- as(Ecuador, "SpatialPolygonsDataFrame", filled.contour())
```

```
bubble(datv[!is.na(datv$PE_RK),], "PE_RK", pch = 21,
       col=c('red', 'green'), main = "Errores espaciales de predicción", sp.layout=list("sp.polygons", 1
```

Errores espaciales de predicción



```
# Grabamos el espacio de trabajo.
```

```
save.image("C:/Marsev/Ecuador/Ecuador_mg_vs_resobaja.Rdata")
```