Clasificación de uso de suelo en Zapopan: estadisticas de signaturas

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language: output - spanish, code comments - english

```
verbose = 0
control_parcels_path <- "Control_zones.shp"</pre>
classification_path <- "../2016_classif/random/random_points_50000_buff100_dissok.shp"</pre>
raster_path <- "../2016_corrections/compose_LC80290462016_feb_mar_8PC_incl_b8.TIF"
raster_layers_names <- c("pc1","pc2","pc3","pc4","pc5","pc6","pc7","pc8")
raster_layers_count <- 8
class_codes <- c(100,101,103,105,110,115,120,130,135,150,155,160,161,190,192,195,197,200)
class_legend <- c("Urbano",</pre>
                   "Urbano disperso",
                   "Infraestructura",
                   "SVA",
                   "Agr. riego",
                   "Agr. temporal",
                   "Pastizal",
                   "Matorral",
                   "BEsp",
                   "BTC",
                   "BTSC",
                   "BTemp disp.",
                   "BTemp dens.",
                   "Golf",
                   "AVU",
                   "VAS",
                   "BGal"
                   "Agua")
```

```
# Load zone shape
control_parcels <- readShapePoly(control_parcels_path)
# Load classification
main_class <- readShapePoly(classification_path)
# Load only the first layer of the source raster
raster_l1 <- raster(raster_path)
dim(raster_l1)</pre>
```

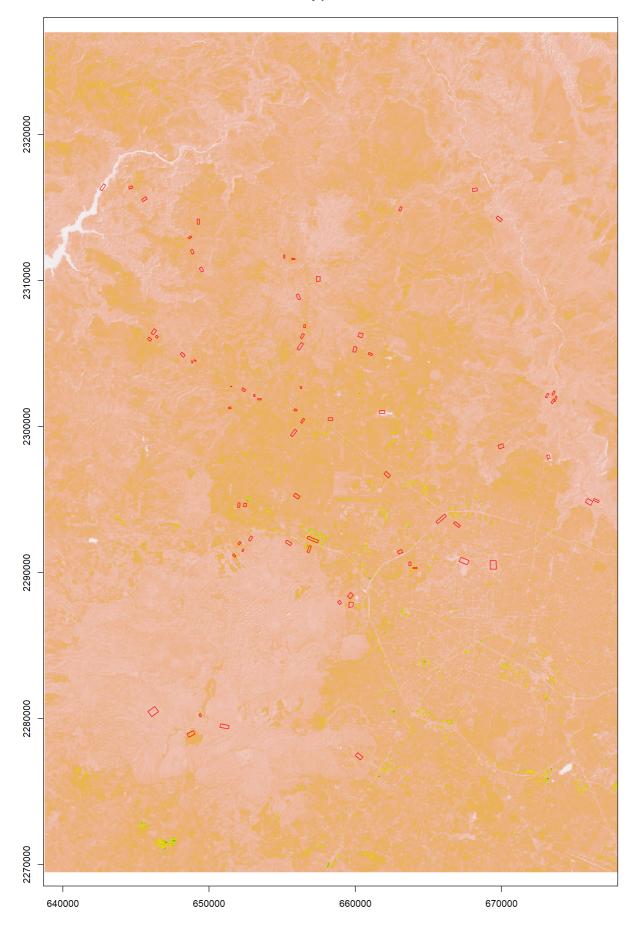
```
## [1] 3840 2618   1
```

```
# load all layers of the source raster
raster_stack<-stack(raster_path)
dim(raster_stack)</pre>
```

```
## [1] 3840 2618 8
```

```
names(raster_stack) <- raster_layers_names</pre>
```

Raster fuente y parcelas de control



```
# reduce number of sampling zones to less then 100 for each class
for (j in 1:length(class_codes)) {
  main_class_subset <- main_class[main_class@data$CLASS == class_codes[j],]</pre>
  subzone count <- 1 + floor(length(main class subset) / 100)</pre>
  if (verbose) {
    cat("Clase: ", class_codes[j], "\n")
               Número de zonas: ", length(main_class_subset), "\n")
    cat("
               Número de subzonas inicial: ", subzone_count, "\n")
  }
  main_class_kfold_index <- kfold(main_class_subset, k = subzone_count)</pre>
  main_class_subset_kfold <- main_class_subset[main_class_kfold_index == 1,]</pre>
  if (verbose) {
    cat("
               Numero de subzonas restante: ", length(main_class_subset_kfold), "\n")
  }
  if (j == 1) {
    main_class_kfold <- main_class_subset_kfold</pre>
    main_class_kfold <- rbind(main_class_kfold, main_class_subset_kfold)</pre>
  }
}
cat("Numero de zonas de muestreo restante para todos clases: ",length(main_class_kfold),
```

```
## Numero de zonas de muestreo restante para todos clases: 1224
```

```
# sampling
main_sampling <- extract(raster_stack, main_class_kfold)</pre>
```

```
# sampling
control_sampling <- extract(raster_stack, control_parcels)</pre>
```

Ejemplo de datos en class (promedio, sd, boxplot)

```
# review example unit of data
head(main_sampling[[1]])
```

```
##
                                        pc4
                                                           pc6
             pc1
                      pc2
                               pc3
                                                 pc5
                                                                    pc7
## [1,] 45805.14 29610.96 18907.39 51923.57 19219.10 24441.85 23007.39
## [2,] 44700.15 31225.41 18423.84 51593.07 18567.01 24885.54 23339.13
## [3,] 44281.46 30665.68 18514.77 51473.67 17598.44 24758.86 23278.13
## [4,] 44636.04 31140.85 18424.63 51673.10 18458.59 24802.95 23297.02
## [5,] 47318.12 31790.99 19541.12 51235.63 20485.86 24921.20 23424.73
## [6,] 46158.75 30253.41 19820.68 50629.48 17738.49 24726.99 23344.65
##
             pc8
## [1,] 25196.95
## [2,] 24911.82
## [3,] 25385.37
## [4,] 25256.88
## [5,] 24903.62
## [6,] 25604.50
```

```
main_class_kfold@data$CLASS[1]
```

```
## [1] 100
```

```
dim(main_sampling[[1]])[1]
```

```
## [1] 89
```

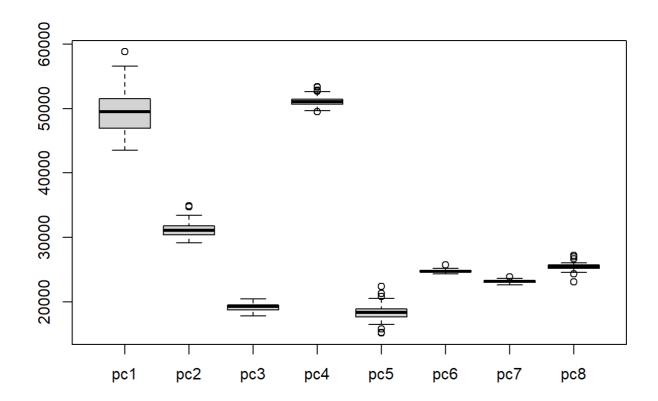
```
apply(main_sampling[[1]],2,mean)
```

```
## pc1 pc2 pc3 pc4 pc5 pc6 pc7 pc8
## 49161.03 31150.13 19136.92 51155.57 18326.49 24749.79 23166.13 25434.48
```

```
apply(main_sampling[[1]],2,sd)
```

```
## pc1 pc2 pc3 pc4 pc5 pc6 pc7
## 3036.6408 1098.1172 547.4662 739.2341 1185.9368 227.9075 258.0665
## pc8
## 514.5812
```

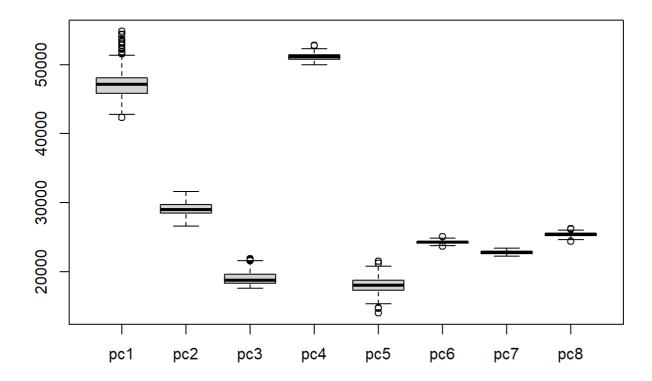
```
boxplot(main_sampling[[1]], col = "lightgray")
```



Ejemplo de datos en parcela (promedio, sd, boxplot)

```
# review example unit of data
head(control_sampling[[1]])
```

```
##
             pc1
                      pc2
                                        pc4
                                                 pc5
                                                          pc6
## [1,] 46846.09 28860.39 19415.59 51007.31 17822.86 24288.69 22561.76
## [2,] 47768.64 29868.83 18943.46 51170.73 18726.11 24581.70 22561.53
## [3,] 47759.44 29846.29 18951.05 51171.02 18675.30 24590.62 22561.20
## [4,] 44951.42 29854.37 20262.03 51523.89 19139.68 24484.58 22914.38
## [5,] 44610.00 29397.61 20340.51 51393.83 18336.76 24402.23 22875.52
## [6,] 44446.09 27259.75 20071.73 50769.47 17499.38 24321.08 22780.80
##
             pc8
## [1,] 25693.04
## [2,] 25126.93
## [3,] 25043.83
## [4,] 24997.36
## [5,] 25293.41
## [6,] 25345.50
control_parcels@data$Code_1[1]
## [1] 101
dim(control_sampling[[1]])[1]
## [1] 246
apply(control_sampling[[1]],2,mean)
##
                 pc2
                          pc3
                                   pc4
                                            pc5
                                                     pc6
                                                              pc7
## 47308.77 29085.91 19163.39 51119.87 18062.19 24326.99 22829.84 25388.83
apply(control_sampling[[1]],2,sd)
##
         pc1
                   pc2
                             pc3
                                       pc4
                                                 pc5
                                                           pc6
## 2307.7048 949.9949 999.3554 453.0548 1237.6978 226.4232 246.5324
##
         pc8
   289,4789
##
boxplot(control_sampling[[1]], col = "lightgray")
```



Tratamiento por clase

Reacomodo de datos por clases de muestreo principal

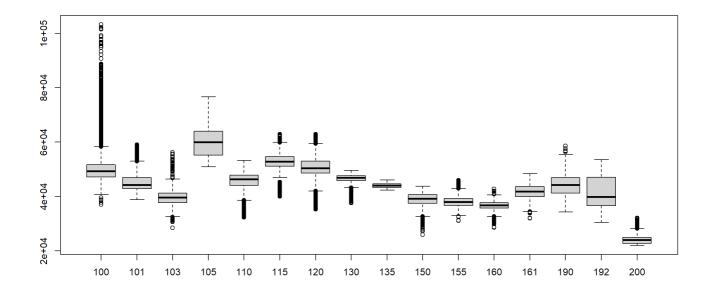
```
# rearrange data to data.frame
# create empty dataframe for main samping by class
main_sampling_df <- data.frame(matrix(ncol = raster_layers_count + 1, nrow = 0))
colnames(main_sampling_df) <- c(raster_layers_names, "CL")

# put sampling result into dataframe, the extra column is for class (CL)
for (i in 1:length(main_sampling)) {
    c_length <- dim(main_sampling[[i]])[1]
    id <- main_class_kfold@data$CLASS[i]
    fl <- cbind(main_sampling[[i]], CL = rep(id,c_length))
    main_sampling_df <- rbind(main_sampling_df,fl)
}

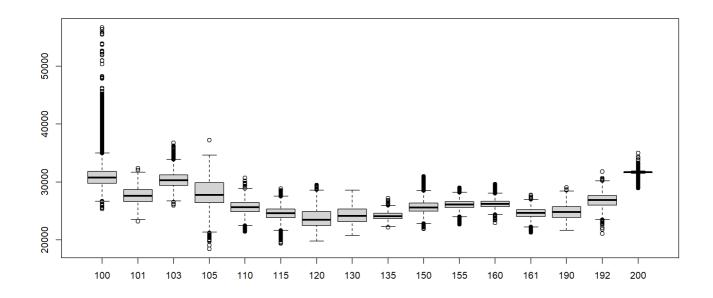
# optional standardize variables (only data columns)
#main_sampling_df[,c(1:raster_layers_count)] <- scale(main_sampling_df[,c(1:raster_layers_count)])</pre>
```

Resumen de datos por clase y calculo de promedios por clase

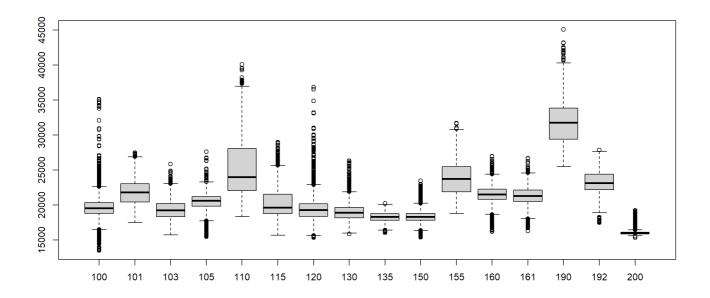
```
# visualize variables by class
boxplot( pc1 ~ CL, data = main_sampling_df, col = "lightgray")
```



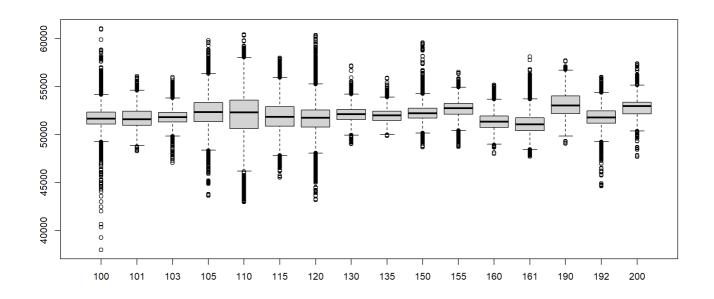
boxplot(pc2 ~ CL, data = main_sampling_df, col = "lightgray")



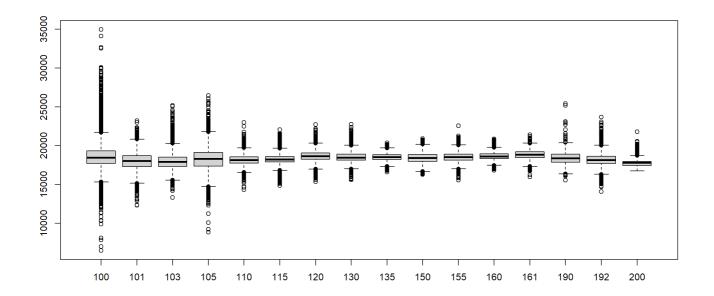
boxplot(pc3 ~ CL, data = main_sampling_df, col = "lightgray")



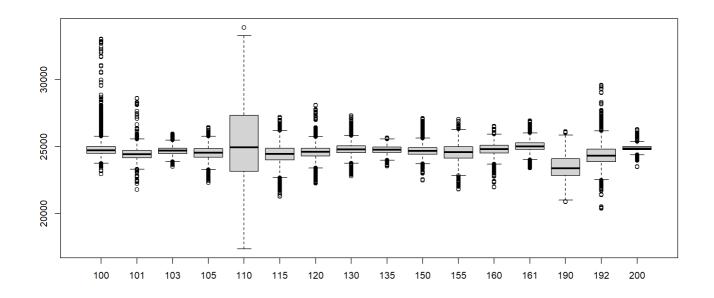
boxplot(pc4 ~ CL, data = main_sampling_df, col = "lightgray")



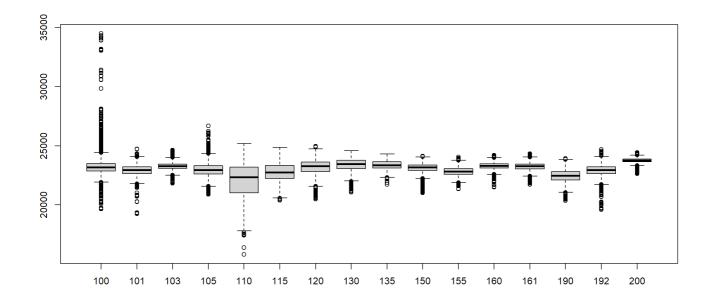
boxplot(pc5 ~ CL, data = main_sampling_df, col = "lightgray")



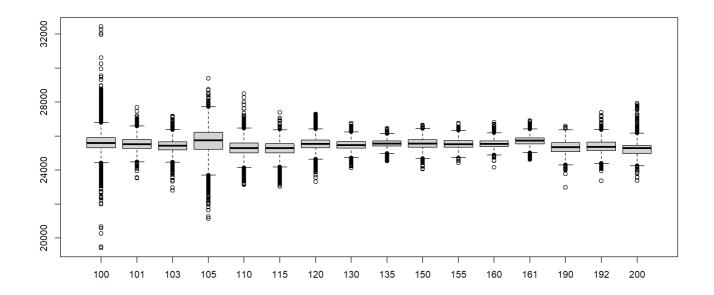
boxplot(pc6 ~ CL, data = main_sampling_df, col = "lightgray")



boxplot(pc7 ~ CL, data = main_sampling_df, col = "lightgray")



boxplot(pc8 ~ CL, data = main_sampling_df, col = "lightgray")



```
# calculate class means
M1 <-aggregate( . ~ CL, data = main_sampling_df, FUN = mean)
row.names(M1) <- M1$CL
raster_layers_count1 <- raster_layers_count + 1
M1 <- M1[,c(2:raster_layers_count1)]
M1</pre>
```

```
##
            pc1
                     pc2
                              pc3
                                       pc4
                                                pc5
                                                          pc6
## 100 50955.04 31417.18 19686.99 51755.79 18673.07 24871.93 23273.72
## 101 45976.93 27681.62 21942.05 51768.15 18008.76 24462.50 22919.29
## 103 39405.37 30357.05 19361.31 51855.39 18000.92 24687.69 23270.62
## 105 60126.03 28040.96 20476.66 52330.88 18277.87 24538.77 22979.45
## 110 45569.22 25641.19 25318.06 52162.26 18166.06 25120.40 22086.78
## 115 52622.71 24556.73 20368.36 51859.68 18244.42 24445.27 22834.40
## 120 50414.76 23783.50 19377.48 51606.33 18648.65 24572.07 23175.60
## 130 46566.27 24274.30 19022.24 52110.84 18536.03 24815.91 23410.79
## 135 44032.84 24177.78 18297.46 51971.44 18509.03 24767.52 23382.10
## 150 38769.17 25813.25 18367.64 52260.74 18425.79 24680.99 23118.98
## 155 38108.93 26129.90 23701.52 52642.15 18542.93 24539.77 22819.87
## 160 36569.86 26283.04 21554.60 51343.15 18656.97 24772.93 23288.91
## 161 41836.76 24625.12 21328.81 51123.45 18814.24 25031.11 23237.72
## 190 44104.79 24801.93 31942.06 53127.35 18391.51 23462.38 22404.65
## 192 41557.33 26823.65 23237.02 51754.29 18197.15 24377.12 22867.19
## 200 24281.28 31606.47 16163.56 52879.62 17803.82 24855.37 23731.07
##
            pc8
## 100 25639.30
## 101 25548.86
## 103 25415.60
## 105 25680.70
## 110 25302.16
## 115 25256.33
## 120 25535.20
## 130 25474.76
## 135 25561.17
## 150 25559.56
## 155 25543.02
## 160 25549.83
## 161 25711.46
## 190 25328.59
## 192 25399.21
## 200 25283.90
```

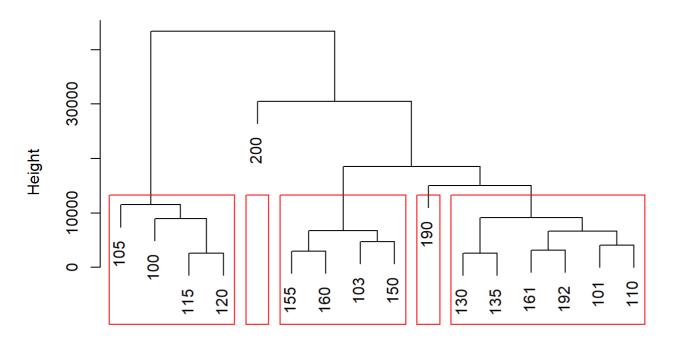
Relación entre clases

```
# Ward Hierarchical Clustering
D <- dist(M1, method = "euclidean")
fit <- hclust(D, method="ward")</pre>
```

```
## The "ward" method has been renamed to "ward.D"; note new "ward.D2"
```

```
plot(fit) # display dendogram
groups <- cutree(fit, k=5) # cut tree into 5 clusters
# draw dendogram with red borders around the 5 clusters
rect.hclust(fit, k=5, border="red")</pre>
```

Cluster Dendrogram



D hclust (*, "ward.D")

Reacomodo de datos por zonas de muestreo principal

```
# rearrange data to data.frame
# create empty dataframe for main samping by zone
main_sampling_df2 <- data.frame(matrix(ncol = raster_layers_count + 1, nrow = 0))
colnames(main_sampling_df2) <- c(raster_layers_names,"ZN")

# put sampling result into dataframe, the extra column is for class (ZN)
for (i in 1:length(main_sampling)) {
    c_length <- dim(main_sampling[[i]])[1]
    id <- main_class_kfold@data$DISS_CODE[i]
    fl <- cbind(main_sampling[[i]], ZN = rep(id,c_length))
    main_sampling_df2 <- rbind(main_sampling_df2,f1)
    if (verbose) {
        print(id)
        print(id)
        print(main_class_kfold@data$CLASS[i])
    }
}</pre>
```

Calculo de promedios por zona

```
# calculate main sampling zone means
M3 <-aggregate( . ~ ZN, data = main_sampling_df2, FUN = mean)
row.names(M3) <- M3$ZN
raster_layers_count1 <- raster_layers_count + 1
M3 <- M3[,c(2:raster_layers_count1)]
#M3</pre>
```

Determinar clase mas probable para cada zona de "main sampling" (K-nearets neighbour)

```
# the order of elements in results of "aggregate" is not the same as in the original class,
# so it is necessary to sort the class labels in same order as training set
# following is to get original class label vector with same order as elements in M3
main_sampling_classes <- c()</pre>
M3_length <- dim(M3)[1]
for (i in 1:M3_length) {
  zone code <- row.names(M3[i,])</pre>
  zone_subset <- main_class_kfold[main_class_kfold@data$DISS_CODE == zone_code,]</pre>
  zone_class <- zone_subset@data$CLASS</pre>
  if (verbose) {
    cat("Zona: ", zone_code,"\n")
    cat("Clase de zona: ", zone_class, "\n")
  main_sampling_classes <- c(main_sampling_classes, zone_class)</pre>
# check the dataset dimensions
#dim(main_sampling_df2[,c(1:raster_layers_count)])
#length(main_sampling_df2[,"ZN"])
# Predict the most probable class for each parcel
prediction_M3 <- knn(main_sampling_df[,c(1:8)], # training set</pre>
                                              # test set
                  main_sampling_df[,"CL"], # training class labels
                  k = 50, prob = TRUE)
M3_predict <- cbind(M3, prediction = prediction_M3, probability = attr(prediction_M3,
"prob"), original = main_sampling_classes)
table_M3_predict <- M3_predict[,c("original","prediction","probability")]</pre>
write.csv(table_M3_predict,file = "tabla_M3_predict.csv")
```

Tabla con valores de predicción para cada zona de "main sampling"

table_M3_predict

##		original	prediction	probability	
##	12192	192	192	0.5200000	
##	17103	103	103	1.0000000	
##	19192	192	192	1.0000000	
##	28110	110	101	0.6800000	
##	33130	130	130	1.0000000	
##	53101	101	101	0.4800000	
##	59120	120	120	0.8600000	
##	60160	160	160	1.0000000	
##	63150	150	150	0.4400000	
##	68135	135	135	1.0000000	
##	75155	155	155	0.9400000	
##	84161	161	161	1.0000000	
##	91101	101	101	1.0000000	
##	93161	161	161	1.0000000	
##	96120	120	120	0.9000000	
##	100130	130	130	0.7600000	
##	119150	150	150	1.0000000	
##	140135	135	135	1.0000000	
##	145115	115	115	1.0000000	
##	147155	155	161	0.5200000	
##	150103	103	103	1.0000000	
##	154192	192	110	0.5200000	
##	158161	161	161	0.9800000	
##	175161	161	161	0.9200000	
##	177160	160	160	1.0000000	
##	179150	150	150	1.0000000	
##	191135	135	135	1.0000000	
##	203135	135	135	1.0000000	
##	205130	130	130	1.0000000	
##	227103	103	103	1.0000000	
##	233135	135	135	1.0000000	
##	235150	150	150	0.9800000	
##	243155	155	155	0.6400000	
##	256103	103	103	1.0000000	
##	268160	160	160	1.0000000	
##	281101	101	101	0.9600000	
##	283160	160	160	1.0000000	
##	292150	150	150	1.0000000	
##	323135	135	135	1.0000000	
##	337103	103	103	1.0000000	
##	340100	100	100	1.0000000	
##	350160	160	160	0.9200000	
##	361135	135	135	1.0000000	
##	362105	105	105	1.0000000	
##	363105	105	105	0.5800000	
##	371161	161	161	1.0000000	
##	376192	192	192	0.7000000	
##	378103	103	103	1.0000000	
##	392160	160	160	1.0000000	
##	395192	192	192	0.4800000	
##	400192	192	110	0.5000000	
##	406130	130	130	1.0000000	
##	421105	105	105	0.9400000	
##	428155	155	161	0.7000000	
##	476115	115	115	1.0000000	
##	480135	135	135	1.0000000	

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## 483100	100	100	1.0000000
## 484120	120	120	0.6600000
## 504130	130	130	0.9400000
## 510101	101	101	0.9400000
## 515155	155	155	1.0000000
## 524200	200	200	1.0000000
## 528155	155	155	0.8600000
## 530120	120	115	0.9400000
## 537150	150	150	1.0000000
## 553103	103	103	1.0000000
## 566110	110	110	1.0000000
## 567100	100	100	1.0000000
## 588161	161	161	1.0000000
## 589105	105	120	0.6800000
## 592155	155	160	0.8000000
## 600130			
	130	130	1.0000000
	135	135	1.0000000
## 607115	115	115	0.9600000
## 608161	161	161	0.9600000
## 610115	115	115	1.0000000
## 627110	110	192	
## 631130	130	130	1.0000000
## 641160	160	160	1.0000000
## 645120	120	120	1.0000000
## 647115	115	115	0.6000000
## 660130	130	130	1.0000000
## 669110	110	110	1.0000000
## 673110	110	110	0.8000000
## 676105	105	105	
## 699105	105	105	1.0000000
## 703100	100	100	1.0000000
## 709100	100	100	1.0000000
## 722103	103	103	1.0000000
## 734190	190	190	0.8000000
## 736115	115	120	0.7800000
## 738160	160	150 161	0.8400000
## 743150	150	161	0.5200000
## 750150	150	161	0.5400000
## 758160	160	160	0.9200000
## 761160	160	160	0.7600000
## 764101	101	101	0.9400000
## 775120	120	115	0.7400000
## 777110	110	110	0.7400000
## 785120	120	120	1.0000000
## 786100	100	100	1.0000000
## 795160	160	160	0.9800000
## 810192	192	110	0.4800000
## 819100	100	100	
## 824100	100	100	1.0000000
## 827161	161 150	161 150	0.8600000
## 833150	150 161	150 161	1.0000000
## 838161	161	161	0.9800000
## 850120	120	120	1.0000000
## 852155	155	155	0.9000000
## 856110	110	110	0.5000000
## 857105	105	105 161	
## 858130	130	161 161	0.6800000
## 862161	161	161	1.0000000

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## 875103	103	103	1.0000000	
## 882105	105	105	0.9200000	
## 889160	160	160	0.9600000	
## 914200	200	200	1.0000000	
## 918103	103	103		
## 920101	101	101		
## 920101	150	150	1.0000000	
			1.0000000	
## 923120	120	120		
## 927103	103	150		
## 930115	115	110		
## 931150	150	150	1.0000000	
## 936120	120	120	1.0000000	
## 938130	130	130		
## 954100	100	100		
## 1007161	161	161	1.0000000	
## 1009110	110	110		
## 1020150	150	150		
## 1022101	101	192		
## 1023150	150	150	1.0000000	
## 1037161	161	161	1.0000000	
## 1046105	105	105		
## 1053135	135	135		
## 1061105	105	105	1.0000000	
## 1066130	130	130	1.0000000	
## 1069103	103	103	1.0000000	
## 1081101	101	100	1.0000000	
## 1088120	120	120	0.9600000	
## 1107130	130	130	1.0000000	
## 1119120	120	120	0.6800000	
## 1138110	110	101	0.6800000	
## 1145103	103	103	1.0000000	
## 1149135	135	135	1.0000000	
## 1175130	130	130	0.9400000	
## 1191155	155	155	0.9200000	
## 1192135	135	135	1.0000000	
## 1204120	120	120	1.0000000	
## 1208160	160	160	0.9200000	
## 1214120	120	120	0.7000000	
## 1219115	115	115	1.0000000	
## 1229161	161	161	1.0000000	
## 1244155	155	155	1.0000000	
## 1255101	101	155	0.4600000	
## 1269100	100	100	0.9800000	
## 1276103	103	103	1.0000000	
## 1281100	100	100	1.0000000	
## 1283155	155	160	0.4600000	
## 1288120	120	120	1.0000000	
## 1294135	135	135	1.0000000	
## 1301100	100	100	1.0000000	
## 1302130	130	130	1.0000000	
## 1304115	115	115	1.0000000	
## 1305190	190	190	0.8000000	
## 1313192	192	101	0.9000000	
## 1330192	192	192	0.6400000	
## 1335190	190	190	0.6400000	
## 1348130	130	130	0.8600000	
## 1361130	130	130	0.5000000	
## 1368135	135	135	1.0000000	

## 1373130	130	130	1.0000000
## 1382155	155	155	0.9200000
## 1384150	150	150	1.0000000
## 1387101	101	101	0.8800000
## 1388161	161	161	1.0000000
	161	161	1.0000000
## 1391200	200	200	1.0000000
## 1393105	105	105	1.0000000
## 1394120	120	120	0.8200000
## 1401135	135	135	1.0000000
## 1411135	135	135	1.0000000
## 1416190	190	190	0.6400000
## 1424101	101	101	0.6400000
## 1433115	115	115	0.8600000
## 1463161	161	161	1.0000000
## 1486130	130	130	0.9800000
## 1493192	192	101	0.5000000
## 1495155	155	155	0.8800000
## 1497130	130	130	0.9800000
## 1514110	110	110	1.0000000
## 1519101	101	101	0.9200000
## 1529101	101	101	0.8200000
## 1536120	120	120	1.0000000
## 1538120	120	120	1.0000000
## 1541110	110	110	0.8200000
## 1562105	105	105	1.0000000
## 1581130	130	130	1.0000000
## 1583161	161	161	1.0000000
## 1587160	160	160	0.900000
## 1589101	101	101	0.8400000
## 1606120	120	120	1.0000000
## 1608160	160	160	1.0000000
## 1611155	155	161	0.5200000
## 1616120	120	120	1.0000000
## 1635135	135	135	1.0000000
## 1640100	100	100	1.0000000
## 1644190	190	190	0.9800000
## 1645160	160	160	0.9200000
## 1648150	150	150	1.0000000
## 1649192	192	160	0.5600000
## 1653100	100	100	1.0000000
## 1659100	100	100	1.0000000
## 1663150	150	150	0.6400000
## 1695120	120	120	1.0000000
## 1697115	115	115	0.8400000
## 1700200	200	200	1.0000000
## 1702120	120	120	0.9600000
## 1702120	120	115	0.5400000
## 1718100	100	100	1.0000000
## 1723101	101	101	1.0000000
## 1730130	130	130	0.7000000
## 1731130	130	130	1.0000000
## 1733135	135	135	1.0000000
## 1735150	150	150	0.8800000
## 1736115	115	120	0.6000000
## 1737110	110	192	0.6000000
## 1757100	100	100	1.0000000
## 1759130	130	130	1.0000000
•			

7-0-10		Olasilica	icion de uso de sucio en Zapopan. Estadist
## 1770150	150	150	1.000000
## 1773135	135	135	1.000000
## 1788103	103	103	1.0000000
## 1810135	135	135	1.0000000
## 1812161	161	161	1.0000000
## 1816135	135	135	1.0000000
## 1823200	200	200	1.0000000
## 1836115	115	115	0.9200000
## 1841103	103	103	1.0000000
## 1846135	135	135	1.0000000
## 1855115	115	115	1.0000000
## 1865115	115	115	0.8600000
## 1877150	150	150	0.8600000
## 1877136	105	105	1.0000000
## 1878163			1.0000000
	160	160	1.0000000
## 1898103	103	103	
## 1899100	100	100	1.0000000
## 1901115	115	115	1.000000
## 1920101	101	101	0.940000
## 1924150	150	150	1.0000000
## 1942160	160	160	1.0000000
## 1947100	100	100	0.6600000
## 1949103	103	103	1.0000000
## 1950101	101	101	0.8600000
## 1952192	192	110	0.7200000
## 1954105	105	105	1.0000000
## 1985100	100	100	1.0000000
## 2011103	103	103	1.0000000
## 2012130	130	130	1.0000000
## 2013160	160	160	0.9600000
## 2019120	120	120	0.9800000
## 2022120	120	120	0.9600000
## 2032130	130	130	1.0000000
## 2043101	101	101	0.9600000
## 2046161	161	161	1.0000000
## 2056135	135	135	1.0000000
## 2057161	161	161	1.0000000
## 2059192	192	160	0.7000000
## 2061110	110	110	1.0000000
## 2067115	115	115	0.5000000
## 2068135	135	135	1.0000000
## 2076161	161	161	1.0000000
## 2079155	155	155	0.7400000
## 2081120	120	120	0.9200000
## 2094101	101	101	0.8400000
## 2108155	155	160	0.5400000
## 2112120	120	120	1.000000
## 2127110	110	110	1.0000000
## 2135150	150	150	1.0000000
## 2146103	103	103	1.0000000
## 2166101	101	101	0.9200000
## 2167135	135	135	1.0000000
## 2170101	101	101	0.9800000
## 2172200	200	200	1.0000000
## 2177110	110	110	0.9000000
## 2196135	135	135	1.000000
## 2197155	155	155	0.8800000
## 2202101	101	101	0.8600000

7-0-10		Olasilica	cion de dos de sucio en Zapopan. Estadist
## 2206100	100	100	1.0000000
## 2207115	115	115	0.6800000
## 2209160	160	160	0.8400000
## 2210100	100	100	1.0000000
## 2211161	161	161	1.0000000
## 2213161	161	161	1.0000000
## 2232161	161	161	1.0000000
## 2245120	120	120	0.9200000
## 2247101	101	101	0.9800000
## 2254120	120	120	0.9600000
## 2256161	161	161	1.0000000
## 2259103	103	103	1.0000000
## 2260101	101	101	0.9019608
## 2262110	110	110	1.0000000
## 2275192	192	192	0.8800000
## 2281190	190	190	0.7200000
## 2301150	150	150	1.0000000
## 2311100	100	100	1.0000000
## 2320103	103	103	1.0000000
## 2339120	120	120	1.0000000
## 2358200	200	200	1.0000000
## 2361150	150	150	1.0000000
## 2369110	110	110	0.7800000
## 2371101	101	101	0.9800000
## 2386160	160	160	0.8000000
## 2387110	110	110	1.0000000
## 2388103	103	103	1.0000000
## 2417103	103	103	1.0000000
## 2434150	150	150	0.6800000
## 2441105	105	101	0.5200000
## 2446192	192	192	0.7200000
## 2454155	155	160	0.5800000
## 2457200	200	200	1.0000000
## 2465155	155	155	0.5200000
## 2475101	101	101	0.6000000
## 2478110	110	110	1.0000000
## 2482150	150	150	1.0000000
## 2510190	190	110	0.6800000
## 2531100	100	100	1.0000000
## 2533100	100	100	1.0000000
## 2535130	130	130	0.9600000
## 2553110	110	110	0.9600000
## 2561150	150	150	1.0000000
## 2572115	115	115	1.0000000
## 2579155	155	155	0.9400000
## 2581115	115	115	0.6800000
## 2582160	160	160	1.0000000
## 2599161	161	161	1.0000000
## 2601135	135	135	1.0000000
## 2602101	101	100	1.0000000
## 2617103	103	103	0.8000000
## 2618105	105	105	1.0000000
## 2624160	160	160	0.8600000
## 2640103	103	103	1.0000000
## 2652115	115	115	0.9200000
## 2658120	120	120	0.9800000
## 2666161	161	161	1.0000000
## 2672160	160	160	1.0000000

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## 2675120	120	120	0.7000000
## 2676135	135	135	1.0000000
## 2677160	160	160	0.9600000
## 2679192	192	192	0.4600000
## 2690192	192	192	0.5800000
## 2714110	110	110	0.7600000
## 2727190	190	190	0.8200000
## 2747200	200	200	1.0000000
## 2753110	110	110	0.8400000
## 2769130	130	130	
## 2770160	160	160	0.9800000
## 2777100	100	100	1.0000000
## 2789115	115	115	1.0000000
## 2791161	161	161	0.9800000
## 2792160	160	192	0.6078431
## 2794120	120	120	1.0000000
## 2796110	110	110	0.8000000
## 2798110	110	110	0.7800000
## 2806110	110	110	0.8000000
## 2826115	115	115	1.0000000
## 2849190	190	190	0.8000000
## 2859103	103	103	1.0000000
## 2868105	105	105	0.4800000
## 2875115	115	115	0.6200000
## 2892135	135	135	1.0000000
## 2901120	120	115	0.6400000
## 2921161	161	161	1.0000000
## 2936150	150	150	1.0000000
## 2939150	150	150	1.0000000
## 2947130	130	130	1.0000000
## 2950160	160	160	0.9600000
## 2959100	100	100	1.0000000
## 2966150	150	150	1.0000000
## 2982120	120	110	0.5882353
## 2985120	120	120	1.0000000
## 2989160	160	160	1.0000000
## 2993150	150	150	1.0000000
## 2997120	120	120	0.8400000
## 3025200	200	200	1.0000000
## 3026150	150	150	1.0000000
## 3036192	192	110	0.5600000
## 3038115	115	115	0.6400000
## 3040110	110	110	0.7200000
## 3045105	105	105	1.0000000
## 3050161	161	161	1.0000000
## 3051120	120	120	1.0000000
## 3064101	101	101	0.7400000
## 3065120	120	120	1.0000000
## 3089192	192	160	0.6800000
## 3151105	105	105	1.0000000
## 3161115	115	115	0.6800000
## 3169161	161	161	0.8000000
## 3205160	160	160	0.8000000
## 3215105	105	115	0.6200000
## 3216130	130	155	0.4200000
## 3222130	130	130	1.0000000
## 3230160	160	160	1.0000000
## 3241103	103	103	1.0000000
I			

7-0-10		Olasilica	icion de aso de sacio e	ii Zapol
## 3244155	155	155	0.7000000	
## 3261115	115	115	0.9400000	
## 3262130	130	130	1.0000000	
## 3263200	200	200	1.0000000	
## 3266200	200	200	1.0000000	
## 3281161	161	161	1.0000000	
## 3283150	150	150	0.9200000	
## 3284120	120	130	1.0000000	
## 3296115	115	115	0.9800000	
## 3310115	115	115	0.7000000	
## 3311161	161	161	0.8600000	
## 3312161	161	161	1.0000000	
## 3345190	190	190	0.6000000	
## 3346110	110	110	0.7400000	
## 3352105	105	105	1.0000000	
## 3359100	100	100	1.0000000	
## 3360101	101	101	0.6000000	
## 3364192	192	192	0.8600000	
## 3367161	161	161	1.0000000	
## 3372103	103	103	1.0000000	
## 3399100	100	100	1.0000000	
## 3403120	120	120	1.0000000	
## 3414135	135	135	1.0000000	
## 3419161	161	161	1.0000000	
## 3431192	192	110	0.5200000	
## 3438200	200	200	1.0000000	
## 3439190	190	190	0.8600000	
## 3449105	105	105	1.0000000	
## 3453155	155	160	0.8400000	
## 3465150	150	150	1.0000000	
## 3475110	110	110	1.0000000	
## 3478161	161	161	0.9600000	
## 3502190	190	190	0.6400000	
## 3504100	100	100	1.0000000	
## 3510160	160	160	0.9800000	
## 3518192	192	110	0.7400000	
## 3525100	100	100	1.0000000	
## 3532135	135	135	1.0000000	
## 3560120	120	120	1.0000000	
## 3564105	105	105	0.9600000	
## 3575192	192	120	0.3000000	
## 3578190	190	190	0.6000000	
## 3580135	135	135	1.0000000	
## 3589160	160	160	1.0000000	
## 3592101	101	101	0.4400000	
## 3602161	161	161	0.8200000	
## 3640101	101	101	0.7200000	
## 3649100	100	100		
## 3653105	105	105		
## 3698192	192	155	0.6000000	
## 3720161	161	161	1.0000000	
## 3723150	150	150		
## 3730120	120	120	1.0000000	
## 3732115	115	115	1.0000000	
## 3742115	115	115	1.0000000	
## 3771103	103	103		
## 3777160	160	160	0.8400000	
## 3794200	200	200	1.0000000	

 7-0-10		Olasilloa	icion de dos de sucio en Zapopan. Estadisticas
## 3814200	200	200	1.0000000
## 3818135	135	135	1.0000000
## 3821115	115	115	0.5400000
## 3826161	161	161	1.0000000
## 3829120	120	120	0.9600000
## 3830150	150	150	1.0000000
## 3831100	100	100	1.0000000
## 3854161	161	161	0.8200000
## 3874103	103	103	1.0000000
## 3883155	155	155	0.8000000
## 3897130	130	130	1.0000000
## 3903130	130	130	1.0000000
## 3907161	161	161	1.0000000
## 3921161	161	161	0.9800000
## 3923115	115	115	1.0000000
## 3928155	155	155	0.7400000
## 3935101	101	101	0.7200000
## 3951150	150	150	1.0000000
## 3952155	155	155	0.5800000
## 3957135	135	135	1.0000000
## 3977200	200	200	1.0000000
## 4003130	130	130	1.0000000
## 4029192	192	192	0.7200000
## 4031192	192	192	1.0000000
## 4033100	100	100	1.0000000
## 4048103	103	103	1.0000000
## 4055100	100	100	1.0000000
## 4060103	103	103	1.0000000
## 4063135	135	135	1.0000000
## 4067103	103	103	1.0000000
## 4071103	103	103	1.0000000
## 4075120	120	120	0.6400000
## 4076110	110	110	1.0000000
## 4095161	161	161	1.0000000
## 4111135	135	135	1.0000000
## 4130105	105	105	1.0000000
## 4135135	135	135	1.0000000
## 4139130	130	130	0.9800000
## 4142161	161	161	0.7600000
## 4146110	110	110	1.0000000
## 4153155	155	155	0.5000000
## 4175105	105	105	1.0000000
## 4179120	120	120	1.0000000
## 4181150	150	150	0.9400000
## 4183161	161	161	0.9000000
## 4191160	160	160	0.9400000
## 4214155	155	155	0.5000000
## 4217100	100	100	0.9800000
## 4237100	100	100	1.0000000
## 4239130	130	130	1.0000000
## 4242105	105	105	1.0000000
## 4260150	150	150	0.9800000
## 4261192	192	192	0.8800000
## 4273100	100	100	1.0000000
## 4274100	100	100	0.9800000
## 4285150	150	150	1.0000000
## 4290120	120	120	0.9600000
## 4204161	1.61	1.61	0 0000000

0.9000000

161

4294161

161

7-0-10		Olasillo	dolon de doo de sacio en Zapopan. es	lac
## 4295155	155	155	0.7600000	
## 4296101	101	101	0.6800000	
## 4315155	155	155	0.8800000	
## 4317103	103	103	1.0000000	
## 4319115	115	115	0.9200000	
## 4331105	105	105	1.0000000	
## 4335150	150	150	0.9400000	
## 4341120	120	120	1.000000	
## 4377100	100	100	0.7400000	
## 4379155	155	155	0.8000000	
## 4397130	130	130	1.000000	
## 4399160	160	160	1.000000	
## 4405120	120	120	1.000000	
## 4419105	105	105	1.000000	
## 4421200	200	200	1.000000	
## 4437120	120	120	1.000000	
## 4439160	160	160	0.5600000	
## 4444100	100	100	1.000000	
## 4445103	103	103	1.000000	
## 4449135	135	135	1.0000000	
## 4451103	103	103	1.000000	
## 4460120	120	120	0.9200000	
## 4463100	100	100	1.000000	
## 4477103	103	103	1.000000	
## 4481115	115	115	0.9800000	
## 4482110	110	110	0.6800000	
## 4485130	130	130	1.000000	
## 4489103	103	103	1.000000	
## 4504130	130	130	1.000000	
## 4518130	130	130	1.000000	
## 4530150	150	150	1.0000000	
## 4545130	130	130	1.000000	
## 4562110	110	190	0.4200000	
## 4567192	192	101	0.7000000	
## 4568100	100	100	1.0000000	
## 4580150	150	150	1.0000000	
## 4584100	100	100	1.0000000	
## 4613160	160	160	1.0000000	
## 4623115	115	115	1.0000000	
## 4628115	115	115	1.0000000	
## 4633103	103	103	1.0000000	
## 4639150	150	150	1.0000000	
## 4641150	150	150	1.0000000	
## 4682135	135	135	1.0000000	
## 4690161	161	161	1.0000000	
## 4691161	161	161	1.0000000	
## 4697100	100	100	1.0000000	
## 4702100	100	100	1.0000000	
## 4717115	115	115	1.0000000	
## 4719161	161	161	1.0000000	
## 4726110	110	110	0.7600000	
## 4750155	155	155	0.4600000	
## 4782101	101	101	0.6400000	
## 4784200	200	200	1.0000000	
## 4790130	130	130	1.0000000	
## 4794130	130	130	0.9400000	
## 4804150	150	150	1.0000000	
## 4811115	115	115	1.0000000	

7-0-10		Olasilica	icion de aso de s	sucio en Zapopan. est
## 4813105	105	105	1.0000000	
## 4816150	150	150	1.0000000	
## 4828115	115	115	0.9400000	
## 4832161	161	161	1.0000000	
## 4843115	115	115	0.8600000	
## 4844115	115	115	1.0000000	
## 4846100	100	100	1.0000000	
## 4847190	190	190	0.8400000	
## 4864130	130	130	1.0000000	
## 4870101	101	101	0.7400000	
## 4875161	161	161	1.0000000	
## 4882160	160	160	0.8800000	
## 4890105	105	105	0.9000000	
## 4896101	101	101	0.7000000	
## 4918100	100	100	1.0000000	
## 4920150	150	161	0.9200000	
## 4922135	135	135	1.0000000	
## 4925130	130	130	1.0000000	
## 4931155	155	155	0.9200000	
## 4949115	115	115	0.7800000	
## 4959190	190	190	0.9400000	
## 4961100	100	100	0.9200000	
## 4962135	135	135	1.0000000	
## 4964161	161	161	0.8000000	
## 4967105	105	105	1.0000000	
## 4969115	115	115	1.0000000	
## 4972101	101	101	0.8800000	
## 4997115	115	115	0.8400000	
## 5001161	161	161	1.0000000	
## 5004103	103	103	1.0000000	
## 5006150	150	150	0.9800000	
## 5013161	161	161	1.0000000	
## 5016130	130	130	1.0000000	
## 5025200	200	200	1.0000000	
## 5026155	155	155	0.5400000	
## 5040115	115	115	1.0000000	
## 5044200	200	200	1.0000000	
## 5053105	105	105	1.0000000	
## 5064200	200	200	1.0000000	
## 5069105	105	105	1.0000000	
## 5074150	150	150	1.0000000	
## 5079120	120	120	0.8800000	
## 5098155	155	155	0.8400000	
## 5099155	155	155	0.7400000	
## 5105115	115	105	0.7600000	
## 5109160	160	160	0.9200000	
## 5122115	115	115	1.0000000	
## 5133130	130	130	1.0000000	
## 5138105	105	105	0.7200000	
## 5143135	135	135	1.0000000	
## 5146160	160	160	0.9200000	
## 5150192	192	192	0.7000000	
## 5152120	120	120	0.9800000	
## 5154130	130	130	0.9200000	
## 5156130	130	130	0.9400000	
## 5158110	110	110	0.7800000	
## 5159105	105	105	1.0000000	
## 5177200	200	200	1.0000000	
	= = =			

1-0-10		Olasilica	cion de do de sacio en Zapopan. Estadistic
## 5181103	103	103	1.0000000
## 5183103	103	103	1.0000000
## 5188130	130	130	0.9400000
## 5192130	130	130	1.0000000
## 5217105	105	101	0.7600000
## 5218135	135	135	1.0000000
## 5219150	150	150	0.9607843
## 5221101	101	101	0.9800000
## 5241120	120	110	0.5800000
## 5244115	115	115	0.8800000
## 5253130	130	130	1.0000000
## 5257150	150	150	1.0000000
## 5275160	160	160	1.0000000
## 5289105	105	105	1.0000000
## 5293160	160	150	0.8600000
## 5294120	120	120	1.0000000
## 5299101	101	101	0.9800000
## 5311115	115	115	0.9200000
## 5318115	115	115	1.0000000
## 5358130	130	130	1.0000000
## 5365101	101	100	0.6800000
## 5370115	115	115	1.0000000
## 5371160	160	160	1.0000000
	160	160	1.0000000
## 5378115	115	115	1.0000000
## 5384103	103	103	1.0000000
## 5396110	110	190	0.5800000
	101	101	0.9600000
## 5401100	100	100	1.0000000
## 5422130	130	130	0.9200000
	101	101	0.6000000
			1.0000000
			1.000000
			0.5600000
			0.7800000
			1.000000
			0.5800000
			0.8200000
			0.7200000
			1.000000
			1.000000
			0.780000
			0.7600000
			1.0000000
			1.0000000
			1.0000000
			0.8600000
			0.5800000
			0.6200000 0.4800000
			0.9400000
			1.000000
			1.0000000
			1.0000000
			1.0000000
			1.0000000
			0.9800000
			0.7800000
иш Э/ЭОТZ	120	120	0.,00000
	## 5183103 ## 5188130 ## 5192130 ## 5217105 ## 5218135 ## 5219150 ## 5221101 ## 5241120 ## 5244115 ## 5253130 ## 5257150 ## 5275160 ## 5293160 ## 5293160 ## 5294120 ## 5311115 ## 5318115 ## 5358130 ## 5365101 ## 5370115 ## 5370115 ## 5374160 ## 5374160 ## 5378115 ## 5378115 ## 5399101 ## 5399101 ## 5399101 ## 5401100	## 5183103	## 5181103

	7-0-10		Olasillo	icion de aso de sacio e
	## 5760192	192	103	0.6600000
	## 5766130	130	130	1.0000000
	## 5770115	115	115	1.0000000
	## 5776120	120	120	0.8200000
	## 5794150	150	150	1.0000000
	## 5796150	150	150	1.0000000
	## 5797135	135	135	1.0000000
	## 5812160	160	160	1.0000000
	## 5816155	155	160	0.5800000
	## 5829135	135	135	1.0000000
	## 5832110	110	110	1.0000000
	## 5837110	110	190	0.5800000
	## 5846103	103	103	1.0000000
	## 5848200	200	200	1.0000000
	## 5849103	103	103	1.0000000
	## 5858135	135	135	1.0000000
	## 5876100	100	100	1.0000000
	## 5879160	160	160	1.0000000
	## 5889103	103	103	1.0000000
	## 5899115	115	115	0.7000000
	## 5909160	160	160	0.9800000
	## 5911101	101	101	1.0000000
	## 5923200	200	200	1.0000000
	## 5925192	192	101	0.8600000
	## 5932130	130	130	1.0000000
	## 5937115	115	115	0.8000000
	## 5938135	135	135	1.0000000
	## 5947130	130	130	1.0000000
	## 5949120	120	120	1.0000000
	## 5952160	160	160	1.0000000
	## 5954155	155	155	0.9600000
	## 5964100	100	100	1.0000000
	## 5998130	130	130	1.0000000
	## 6008115	115	115	0.6000000
	## 6012110	110	130	0.5800000
	## 6014120	120	120	1.0000000
	## 6020115	115	115	0.9600000
	## 6023100	100	100	1.0000000
	## 6031105	105	105	1.0000000
	## 6042103	103	103	1.0000000
	## 6043150	150	150	1.0000000
	## 6045100	100	100	1.0000000
	## 6046135	135	135	1.0000000
	## 6049110	110	110	0.7000000
	## 6057155	155	155	0.9800000
	## 6062115	115	101	0.6666667
	## 6069150	150	150	1.0000000
	## 6077105	105	105	1.0000000
	## 6090161	161	161	0.9200000
	## 6091160	160	160	0.9200000
	## 6092101	101	101	0.880000
	## 6098103	103	103	1.0000000
	## 6105130	130	130	0.5200000
	## 6120115	115	115	0.9800000
	## 6129150	150	150	1.0000000
	## 6131115	115	115	0.8400000
	## 6134120	120	120	0.9200000
	## 6139150	150	150	1.0000000
Į	I			

7-0-10		Olasillo	acion de uso de sucio en Zapopan.
## 6147120	120	120	0.9200000
## 6155155	155	155	0.5200000
## 6162100	100	100	1.0000000
## 6169161	161	161	1.0000000
## 6172103	103	103	1.0000000
## 6186100	100	100	1.0000000
## 6190120	120	120	0.9200000
## 6194150	150	150	1.0000000
## 6210192	192	192	
## 6212155	155	155	
## 6220135	135	135	
## 6221150	150	150	
## 6242100	100	100	
## 6243110	110	110	
## 6253110	110	110	
## 6264110	110	110	
## 6277115	115	115	
## 6279200	200	200	
## 6280155	155	155	
## 6281135	135	135	
## 6298110	110	110	
## 6300190	190	190	
## 6317155	155	160	
## 6325100	100	100	
## 6331155	155	161	
## 6339161	161	161	
## 6345115	115		
## 6348101	101	115 101	
## 6350160	160	160	
## 6352120	120	120	
## 6352120	135	135	
## 6362161	161	161	
## 6382101			0.9000000
## 6392161	161	161	
## 6394150	150	150	
## 6398160	160	160	
## 6400101	100	101	
## 6402200	200	200	1.0000000
## 6432130	130	130	
## 6445110	110	110	
## 6447190	190	110	
## 6462115	115	115	
## 6475160	160	160	
## 6477200	200	200	
## 6481100	100	100	
## 6489110	110	110	0.9000000
## 6501105	105	105	
## 6503130	130	130	
## 6508200	200	200	
## 6510110	110	110	0.9800000
## 6517150	150	150	
## 6539103	103	103	
## 6549115	105	115	
## 6553160	160	160	0.9600000
## 6574130	130	130	
## 6575161	161	161	
## 6590105	105	101	
## 6597115	105	115	
ππ 039/113	11)	113	1.000000

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## 6601160	160	150	0.9400000
## 6603135	135	135	1.0000000
## 6616101	101	192	0.5400000
## 6624110	110	110	0.900000
## 6626120	120	120	1.0000000
## 6643155	155	160	0.7000000
## 6664120	120	120	0.9400000
## 6666100	100	100	1.0000000
## 6674130	130	130	
## 6676161	161	161	0.8200000
## 6679161	161	161	1.0000000
## 6695200	200	200	1.0000000
## 6702161	161	161	
## 6721120	120	120	1.0000000
## 6737120	120	115	0.6400000
## 6768100	100	100	
## 6773161	161	161	
## 6776115	115	115	1.0000000
## 6777130	130	130	0.9800000
## 6784150	150 100	150 100	1.0000000
## 6792100 ## 6800115	100 115	115	0.9200000 0.9600000
## 6802161			1.0000000
## 6802101	161 110	161 110	
## 6828155	155	155	0.7000000
## 6832161	161	161	1.000000
## 6871135	135	135	1.0000000
## 6882110	110	110	0.5400000
## 6885100	100	100	0.9800000
## 6893200	200	200	0.7400000
## 6921161	161	161	1.0000000
## 6922103	103	103	1.000000
## 6937161	161	161	1.000000
## 6960161	161		0.9800000
## 6978103	103	103	1.0000000
## 7019100	100	100	1.0000000
## 7023110	110	110	1.0000000
## 7039161	161	161	1.0000000
## 7045101	101	101	0.6000000
## 7050103	103	103	1.000000
## 7054160	160	160	0.9200000
## 7055192	192	110	0.5200000
## 7057200	200	200	1.0000000
## 7073100	100	100	1.0000000
## 7080150	150	150	1.0000000
## 7081115	115	115	1.0000000
## 7083101	101	101	0.8600000
## 7086161	161	161	1.0000000
## 7093130	130	130	1.0000000
## 7094115	115	115	1.0000000
## 7132130	130	130	1.0000000
## 7140100	100	100	1.0000000
## 7155103	103	103	1.0000000
## 7163135	135	135	1.0000000
			0.8800000
			0.6200000
			0.7000000
## /1//100	TAA	TOO	1.0000000
	## 7164150 ## 7174155 ## 7175110 ## 7177100	## 7174155 155 ## 7175110 110	## 7174155 155 155 ## 7175110 110 110

## 7189160	160	160 1.0000000
## 7193150	150	150 1.0000000
## 7201110	110	110 0.9600000
## 7202105	105	101 0.6000000
## 7204160	160	160 0.8200000
## 7206155	155	155 0.8000000
## 7215150	150	150 1.0000000
## 7219155	155	160 0.9600000
## 7234120	120	120 1.0000000
## 7237110	110	110 1.0000000
## 7241103	103	103 0.4800000
## 7244101	101	101 0.5200000
## 7249135	135	135 1.0000000
## 7251135	135	135 1.0000000
## 7255100	100	100 1.0000000
## 7262115	115	115 1.0000000
## 7275135	135	135 1.0000000
## 7292192	192	101 0.6400000
## 7300155	155	155 0.7400000
## 7310130	130	130 1.0000000
## 7328161	161	161 1.0000000
## 7360150	150	150 0.9215686
## 7382115	115	115 0.8400000
## 7389105	105	105 0.6800000
## 7409150	150	150 1.0000000
## 7411155	155	160 0.6200000
## 7441103	103	103 1.0000000
## 7448130	130	130 1.0000000
## 7449155	155	155 0.5800000
## 7451161	161	161 1.0000000
## 7459100	100	100 1.0000000
## 7468130	130	130 1.0000000
## 7476135	135	135 1.0000000
## 7479150	150	150 1.0000000
## 7484160	160	160 1.0000000
## 7497192	192	192 0.5400000
## 7502135	135	135 1.0000000
## 7522100	100	100 1.0000000
## 7527160	160	160 0.9800000
## 7529130	130	130 1.0000000
## 7532135	135	135 1.0000000
## 7539115	115	115 0.8000000
## 7558103	103	103 1.0000000
## 7561161	161	161 0.9400000
## 7572160	160	150 0.5800000
## 7575115	115	115 1.0000000
## 7576135	135	135 1.0000000
## 7587160	160	160 0.7600000
## 7594110	110	110 0.8200000
## 7604100	100	100 1.0000000
## 7607192	192	150 0.7000000
## 7610130	130	130 1.0000000
## 7635160	160	160 1.0000000
## 7638110	110	110 1.0000000
## 7642100	100	100 1.0000000
## 7645130	130	130 1.0000000
## 7659192	192	101 0.6000000
## 7662161	161	161 0.8400000
1		

•	7-0-10		Olasilloa	icion de aso de sacio (
	## 7673161	161	161	1.0000000
	## 7675110	110	110	1.0000000
	## 7679115	115	155	0.5400000
	## 7706155	155	155	0.3600000
	## 7707100	100	100	1.0000000
	## 7712135	135	135	1.0000000
	## 7737110	110	110	0.7800000
	## 7748120	120	120	0.9800000
	## 7749105	105	105	1.0000000
	## 7754115	115	120	0.7400000
	## 7762161	161	161	0.9400000
	## 7767160	160	160	0.9800000
	## 7775150	150	150	1.0000000
	## 7785120	120	120	1.0000000
	## 7790120	120	120	0.8000000
	## 7797155	155	155	0.9000000
	## 7799160	160	160	0.9400000
	## 7807115	115	115	1.0000000
	## 7809160	160	160	0.7200000
	## 7819135	135	135	0.9400000
	## 7822200	200	200	1.0000000
	## 7823100	100	100	1.0000000
	## 7837110	110	110	0.8200000
	## 7838192	192	192	0.6800000
	## 7853200	200	200	1.0000000
	## 7854103	103	103	1.0000000
	## 7857101	101	101	1.0000000
	## 7892110	110	110	1.0000000
	## 7896130	130	130	1.0000000
	## 7902135	135	135	1.0000000
	## 7909110	110	110	0.6200000
	## 7914161	161	161	1.0000000
	## 7918110	110	110	0.9600000
	## 7924105	105	105	1.0000000
	## 7936130	130	130	1.0000000
	## 7969150	150	150	1.0000000
	## 7984192	192	192	0.4200000
	## 7990160	160	160	0.7000000
	## 8006150	150	150	1.0000000
	## 8013120	120	120	1.0000000
	## 8016115	115	115	0.9800000
	## 8023110	110	130	0.5600000
	## 8025100	100	100	1.0000000
	## 8031100	100	100	1.0000000
	## 8042135	135	135	1.0000000
	## 8053161	161	155	0.7200000
	## 8058192	192	101	0.5800000
	## 8073115	115	115	0.7400000
	## 8085120	120	120	1.0000000
	## 8086200	200	200	1.0000000
	## 8115190	190	190	0.9400000
	## 8131130	130	130	1.0000000
	## 8132103	103	103	1.0000000
	## 8133103	103	103	1.0000000
	## 8136115	115	115	1.0000000
	## 8154192	192	192	0.5800000
	## 8163200	200	200	1.0000000
	## 8174101	101	110	0.3200000
4				

## 8175135	135	135	1.0000000	
## 8178161	161	161	1.0000000	
## 8197160	160	160	1.0000000	
## 8198115	115		0.8600000	
## 8202161			1.0000000	
## 8206100			0.5400000	
## 8210161	161		1.0000000	
## 8210101				
	150		1.0000000	
## 8227120	-		1.0000000	
## 8244160		160		
## 8247160	160		0.8000000	
## 8256100	100		1.0000000	
## 8261155			0.4000000	
## 8262100			1.0000000	
## 8265120	120		1.0000000	
## 8268135	135		1.0000000	
## 8277155			0.5400000	
## 8292160	160		0.9400000	
## 8297150	150	150	1.0000000	
## 8305135	135	135	1.0000000	
## 8306110	110	110	0.7000000	
## 8308130	130		0.9800000	
## 8323110	110	110	0.7800000	
## 8326161	161	161	1.0000000	
## 8340100	100	100	1.0000000	
## 8348150	150	150	1.0000000	
## 8352100	100	100	1.0000000	
## 8356192	192	192	0.6200000	
## 8358160	160	160	0.9400000	
## 8364192	192	155	0.3000000	
## 8382103	103	103	1.0000000	
## 8384120	120	120	0.8400000	
## 8387200	200	200	1.0000000	
## 8399155	155	155	0.6800000	
## 8403135	135	135	1.0000000	
## 8405130	130	130	1.0000000	
## 8421155	155	155	0.5400000	
## 8424110	110	110	0.5200000	
## 8440120	120	120	1.0000000	
## 8442103	103	103	1.0000000	
## 8445150	150	161	0.5000000	
## 8464150	150	150	0.6400000	
## 8468110	110	110	1.0000000	
## 8476155	155	155	1.0000000	
## 8494150	150	150	1.0000000	
## 8509115	115	115	1.0000000	
## 8515100	100	100	1.0000000	
## 8527150	150	150	1.0000000	
## 8531192	192	192	0.7600000	
## 8541192	192	101	0.8600000	
## 8544105	105	105	1.0000000	
## 8554190	190	190	0.7254902	
## 8566161	161	161	1.0000000	
## 8569100	100	100	1.0000000	
## 8571150	150	150	1.0000000	
## 8581135	135	135	1.0000000	
## 8603103	103	103	1.0000000	
## 8606100	100	100	1.0000000	

##	8628120	120	120	0.8200000
##	8637155	155	155	0.8400000
##	8646110	110	110	0.5600000
##	8663120	120	120	1.0000000
##	8670150	150	150	1.0000000
##	8684130	130	130	1.0000000
	8687192	192	192	1.0000000
	8717110	110	110	0.8200000
	8718155	155	160	0.6800000
	8724161	161	161	1.0000000
##	8727161	161	161	1.0000000
	8732155	155	155	0.7000000
	8735100	100	100	1.0000000
	8738160	160	160	1.0000000
	8760161	161	161	1.0000000
	8762100	100	100	1.0000000
	8771105	105	105	1.0000000
	8778150	150	150	1.0000000
	8785135	135	135	1.0000000
	8790115	115	115	0.9400000
	8803103	103	103	1.0000000
	8806135	135	135	1.0000000
	8809130	130	130	1.0000000
	8810160	160	160	1.0000000
	8819101	101	101	1.0000000
	8824105	105	105	1.0000000
	8830160	160	160	1.0000000
	8833200	200	200	1.0000000
	8837105	105	105	1.0000000
	8839103	103	103	1.0000000
	8846110	110	110	1.0000000
	8851110	110	110	0.9800000
	8885101	101	101	1.0000000
	8889130	130	130	1.0000000
	8926161	161	161	0.9800000
	8929135	135	135	1.0000000
	8936160	160	160	1.0000000
##	8940100	100	100	1.0000000
	8947155	155	161	0.6200000
##	8961103	103	103	1.0000000
##	8973130	130	130	1.0000000
##	8979135	135	135	1.0000000
##	8981190	190	190	0.7800000
##	9013115	115	115	0.9600000
##	9033130	130	130	0.8800000
##	9034110	110	110	0.7600000
##	9053135	135	135	1.0000000
##	9062100	100	100	1.0000000
	9077160	160	160	0.9600000
##	9078120	120	101	0.8600000
##	9079161	161	161	1.0000000
	9083160	160	160	0.9200000
	9087150	150	150	1.0000000
	9089130	130	130	0.9400000
##	9117101	101	101	0.7600000
	9125103	103	103	1.0000000
	9135135	135	135	1.0000000
	9144150	150	150	1.0000000
				

7-0-10		Olasillo	acion de aso de sacio en Zapop
## 9145135	135	135	1.0000000
## 9159120	120	120	0.8800000
## 9165120	120	115	0.6000000
## 9175150	150	150	1.0000000
## 9177100	100	100	0.9800000
## 9188130	130	101	0.3000000
## 9189110	110	110	0.9000000
## 9207130	130	130	1.0000000
## 9210130	130	150	0.5000000
## 9213120	120	120	1.0000000
## 9214161	161	161	1.0000000
## 9219161	161	161	1.0000000
## 9220120	120	120	0.9400000
## 9223161	161	161	0.8600000
## 9235135	135	135	1.0000000
## 9254150	150	150	1.0000000
## 9262150	150	150	1.0000000
## 9263103	103	103	1.0000000
## 9264120	120	120	0.8400000
## 9272120	120	120	1.0000000
## 9275120	120	120	0.7600000
## 9276150	150	150	1.0000000
## 9278155	155	155	0.9400000
## 9292100	100	100	1.0000000
## 9316115	115	115	1.0000000
## 9328150	150	150	
## 9331160	160	160	0.9800000
## 9335135	135	135	1.0000000
## 9362150	150	150	1.0000000
## 9406130	130	130	1.0000000
## 9408115	115	115	1.0000000
## 9422101	101	100	0.8600000
## 9427115	115	115	1.0000000
## 9434192	192	192	0.5600000
## 9436120	120	120	1.0000000
## 9475105	105	105	0.7600000
## 9476120	120	110	0.3400000
## 9505120	120	120	0.8800000
## 9508105	105	105	0.6600000
## 9514101	101	101	0.9800000
## 9518115	115	115	1.0000000
## 9540100	100	100	1.0000000
## 9541120	120	120	1.0000000
## 9545150	150	150	1.0000000
## 9566115	115	115	1.0000000
## 9573103	103	103	1.0000000
## 9574130	130	130	1.0000000
## 9582160	160	160	1.0000000
## 9586120	120	120	1.0000000
## 9600103	103	103	1.0000000
## 9603160	160	160	0.9400000
## 9604160	160	160	0.9400000
## 9613103	103	103	1.0000000
## 9614130	130	130	1.0000000
## 9619192	192	192	0.9000000
## 9626110	110	110	1.0000000
## 9662120	120	120	1.0000000
## 9666135	135	135	1.0000000

7-0-10	,		Olasilica	cion de uso de sucio en Zapop
##	9667115	115	115	1.0000000
##	9669160	160	160	0.8800000
##	9670103	103	103	1.0000000
##	9689190	190	190	0.7000000
##	9696161	161	161	0.8000000
##	9702150	150	150	1.0000000
##	9704155	155	155	0.8600000
##	9709155	155	155	0.5200000
##	9715155	155	155	0.7200000
	9718130	130	130	1.0000000
	9729200	200	200	1.0000000
	9746150	150	150	1.0000000
	9748155	155	155	0.6800000
	9749103	103	103	1.0000000
##	9755120	120	120	0.9800000
##	9756101	101	192	0.6000000
##	9757130	130	130	1.0000000
##	9768115	115	115	1.0000000
##	9773161	161	161	1.0000000
##	9775120	120	120	1.0000000
	9781101	101	101	1.0000000
	9784155	155		0.9400000
	9787135	135	135	1.0000000
	9806192	192	192	0.5600000
	9811101	101	101	0.9400000
	9812161	161	161	0.9400000
	9822103	103	103	0.9200000
	9824135	135	135	1.0000000
	9825103	103	100	1.0000000
	9833130	130	130	1.0000000
		115	115	0.6800000
		130		0.9800000
##	9880135	135	135	1.0000000
	9908100	100		1.0000000
##	9918105	105		0.6400000
##	9919115	115	115	1.0000000
##	9921130	130	130	1.0000000
##	9940135	135	135	1.0000000
##	9942120	120	120	0.6800000
##	9960100	100	100	1.0000000
##	9970110	110	110	0.7400000
##	9975160	160	160	0.9400000
##	10003155	155	155	0.9800000
##	10004115	115	115	0.9600000
##	10006135	135	135	1.0000000
##	10011200	200	200	1.0000000
##	10014100	100	100	1.0000000
##	10025101	101	101	0.5800000
##	10042115	115	120	1.0000000
##	10082103	103	103	1.0000000
##	10083200	200	200	1.0000000
##	10096150	150	150	1.0000000
##	10098130	130	130	1.0000000
##	10114130	130	130	1.0000000
##	10116155	155	155	0.5600000
##	10123190	190	190	0.8000000
##	10135155	155	155	0.6800000
##	10152135	135	135	1.0000000

```
## 10154155
                  155
                             155
                                    0.8000000
## 10195155
                  155
                             155
                                    0.8800000
## 10197150
                  150
                             150
                                    0.8800000
## 10205130
                  130
                             130
                                    1.0000000
## 10209192
                  192
                             192
                                    0.6600000
## 10213135
                  135
                             135
                                    1.0000000
## 10226103
                  103
                             103
                                    1.0000000
## 10231150
                             150
                                    1,0000000
                  150
```

Tratamiento por zonas de muestreo (test)

```
# rearrange data to data.frame
# create empty dataframe
control_sampling_df <- data.frame(matrix(ncol = raster_layers_count + 1, nrow = 0))</pre>
colnames(control_sampling_df) <- c(raster_layers_names,"ZN")</pre>
# take every list unit and put it into dataframe, the extra column is for index
for (i in 1:length(control_sampling)) {
  c_length <- dim(control_sampling[[i]])[1]</pre>
  id <- control_parcels@data$Id[i]</pre>
 fl <- cbind(control_sampling[[i]], ZN = rep(id,c_length))</pre>
  control_sampling_df <- rbind(control_sampling_df,fl)</pre>
}
# resumen of data per parcel
for (i in 1:length(control_sampling)) {
  if (verbose) {
    # get parcel id
    parcel_id <- control_parcels@data$Id[i]</pre>
    cat("Parcela de control: ", as.character(parcel_id),"\n")
    cat("Clase original: ",control_parcels@data$Code_2[i],"\n")
    # subset dataframe on parcel id
    df_selection <- control_sampling_df[control_sampling_df$ZN == parcel_id,c(1:8)]</pre>
    # get summary
    cat("Resumen","\n")
    #print(summary(df_selection))
    boxplot(df_selection, col = "lightgray")
    cat("\n\n")
  }
}
# optional standardize variables (only data columns)
# control_sampling_df[,c(1:raster_layers_count)] <- scale(control_sampling_df[,c(1:raster_lay</pre>
ers count)])
M2 <-aggregate( . ~ ZN, data = control_sampling_df, FUN = mean)
row.names(M2) <- M2$ZN
raster_layers_count1 <- raster_layers_count + 1</pre>
M2 <- M2[,c(2:raster_layers_count1)]</pre>
Μ2
```

```
##
           pc1
                    pc2
                                      pc4
                                               pc5
                                                        pc6
     47308.77 29085.91 19163.39 51119.87 18062.19 24326.99 22829.84 25388.83
     49269.58 29485.53 20040.02 51232.92 18116.91 24448.50 22848.52 25466.72
      39042.96 30080.34 19102.90 51708.31 17961.04 24627.67 23210.57 25371.07
      34594.71 27920.68 22835.90 51420.00 17939.53 24293.90 23193.68 25103.56
## 4
     47603.79 23913.97 31609.64 53192.29 18326.76 23218.52 22843.87 25313.64
## 5
      60327.16 34418.49 18984.79 50434.81 18243.29 25363.64 23235.63 25782.14
## 6
      39743.70 25644.29 32129.24 52651.92 18170.61 24166.70 22030.14 25164.85
      55345.96 23159.52 19422.27 51509.74 18442.42 25420.36 22517.60 25416.27
## 7
      55056.58 22043.50 20758.32 51436.58 18503.53 24366.01 23375.51 25369.55
      38275.09 25965.71 22387.23 51602.93 19112.27 24622.73 23133.62 25824.54
## 10 37538.19 25798.38 22203.05 51139.88 18739.45 24891.40 23304.68 25570.79
## 11 39614.83 25779.88 18645.33 52621.30 17919.48 24576.86 23121.87 25342.68
## 12 39013.84 26632.30 15703.51 48111.80 19295.28 23884.87 21221.16 26488.96
## 13 62983.41 30012.55 19624.73 52270.31 18917.73 24267.55 22510.21 26011.76
## 14 40155.24 24754.33 22953.68 51445.08 18659.45 25106.77 23348.01 25781.06
## 15 54531.27 22298.05 21131.53 52682.46 19015.67 24387.55 23281.30 25544.35
## 16 43131.71 24778.42 19017.82 52459.94 18470.09 24499.13 23327.99 25493.62
## 17 65737.94 27701.47 20435.84 53153.42 18837.72 24135.55 22406.54 25886.45
## 18 40033.52 25105.81 22369.52 51397.72 18924.04 25132.30 23050.29 25998.85
## 19 52023.80 24592.03 18578.12 50519.60 18291.60 24129.37 22179.94 25211.69
## 20 49658.89 23320.96 19465.75 51517.91 18430.38 25161.22 24231.00 25291.22
## 21 43794.84 24177.39 21318.82 50613.87 19117.71 24947.40 23324.48 25932.96
## 22 45888.27 25301.97 18826.75 52513.29 18667.49 24615.43 23995.99 25552.47
## 23 46588.10 25912.39 21711.61 51509.00 17968.03 25125.15 22407.68 24980.65
## 24 51233.40 23129.36 21981.47 54700.54 18041.52 24594.94 22952.68 25674.02
## 25 48422.30 23386.46 19226.13 52534.48 18861.94 24331.70 23346.81 25508.55
## 26 43232.58 24401.96 17678.88 51900.14 18453.86 24437.41 23374.56 25409.05
## 27 48477.39 23486.34 18024.25 51572.98 18444.59 24641.36 23598.84 25527.92
## 28 42543.75 24525.98 20718.52 50970.03 19078.58 25440.98 22722.15 25817.63
## 29 47618.96 24731.55 18965.73 52830.84 18266.66 24354.49 23001.33 25491.31
## 30 22539.47 31721.00 16007.33 52393.01 17624.80 24906.09 23847.75 25205.80
## 31 43321.15 23791.11 19329.76 49400.55 19024.64 25004.41 23445.74 25652.66
## 32 40980.17 25566.53 22519.30 51980.22 18677.10 24956.31 23008.12 25859.24
## 33 38961.28 26175.53 18432.95 52489.54 18546.36 24733.65 22989.32 25726.10
## 34 37962.44 26005.99 24738.91 52901.39 18351.44 24190.27 23060.50 25444.11
## 35 44102.80 24922.50 29154.53 54114.96 18632.44 23421.22 22630.82 25415.10
## 36 31052.74 31542.99 17152.48 52975.40 18167.64 25016.03 23808.84 25429.08
## 37 45183.14 23058.43 17663.13 52548.07 18467.95 24727.66 23840.30 25471.20
## 38 29788.80 30610.58 19036.44 52551.82 17893.82 24646.80 23704.06 25222.66
## 39 46567.16 30073.15 18874.14 51329.67 18018.01 24533.19 22977.73 25499.91
## 40 35841.55 27357.77 17941.22 53554.62 18310.84 24611.59 23099.81 25641.15
## 41 36536.76 28209.48 22273.83 52917.03 18378.74 24180.33 23170.58 25675.06
## 42 41510.86 25048.34 18873.87 51991.57 18549.85 24685.17 22903.12 25578.70
## 43 44374.90 23588.01 19904.10 49571.77 19116.29 25144.68 23385.98 25767.94
## 44 52658.27 24325.31 19552.70 54456.89 18997.43 24117.19 22665.76 25885.96
## 45 45025.18 23718.01 18203.37 52173.45 18512.43 24549.29 23302.96 25638.91
## 46 54716.25 29098.87 20354.41 50996.00 18278.01 24429.04 22888.38 25415.66
## 47 49214.87 24462.74 17811.95 53399.98 18096.87 24171.50 22265.15 25718.34
## 48 44392.17 25122.44 18272.33 51518.57 18285.42 24748.95 23441.97 25394.96
## 49 37068.96 32341.65 17967.83 52565.83 17629.20 25093.82 23701.98 25310.35
## 50 47273.67 24254.03 18032.22 51340.71 18263.70 24833.54 23854.34 25334.28
## 51 38955.87 25314.71 23757.08 52286.53 18711.76 25011.09 23185.42 25831.27
## 52 50064.52 32281.21 18733.48 53067.11 18650.39 25044.66 23397.91 25922.02
## 53 50389.85 22908.96 22486.12 52294.72 18784.21 24422.62 23370.85 25449.19
## 54 43628.02 24207.81 21083.13 52244.56 18524.94 25121.08 22821.60 25659.93
## 55 48091.98 24127.51 19623.70 51279.23 18302.60 25007.53 23374.12 25441.69
```

```
## 56 56135.35 23280.95 20096.89 52473.81 18570.67 24539.27 22392.77 25571.97
## 57 40300.34 24345.35 20259.61 51523.77 18424.76 25641.13 22935.79 25527.62
## 58 42732.45 26234.75 22696.22 52384.79 18964.24 24380.35 22810.45 25751.37
## 59 58843.72 25178.51 19168.00 51921.09 17682.60 23856.83 22335.27 24611.64
## 60 51722.79 22364.25 20338.31 51990.40 18818.60 24721.61 23480.09 25521.29
## 61 41908.90 25621.88 21100.09 52368.37 19180.59 24591.59 23069.09 25825.70
## 62 38247.72 25642.75 18559.53 52304.03 17566.10 24904.62 23424.64 25287.52
## 63 45064.19 25443.84 21420.03 53639.46 18082.91 25484.52 23506.00 25493.22
## 64 35301.68 29574.10 18735.65 52988.64 18896.58 24241.98 22971.66 25708.54
## 65 42075.99 25019.24 24803.43 53541.14 18269.29 23878.33 22899.74 25548.48
## 66 52102.98 25433.35 18271.06 49475.70 18421.25 23755.49 21808.40 25264.85
## 67 52048.12 24935.65 18815.44 52113.55 18198.66 24749.77 22143.86 25440.60
## 68 45832.92 29618.25 19102.74 52790.18 17791.19 24577.91 23595.88 25369.02
## 69 51720.93 29508.00 19576.45 51487.03 18021.32 24544.99 22954.31 25460.20
## 70 40997.74 24488.30 35653.79 54237.54 17938.18 24557.66 20769.49 25225.41
```

Determinar clase mas probable para cada zona de muestreo complementario (K-nearets neighbour)

```
# the order of elements in results of "aggregate" is not the same as in the original class,
# so it is necessary to sort the class labels in same order as training set
# following is to get original class label vector with same order as elements in M2
control sampling classes <- c()</pre>
M2\_length <- dim(M2)[1]
for (i in 1:M2_length) {
  zone_code <- row.names(M2[i,])</pre>
  zone_subset <- control_parcels[control_parcels@data$Id == zone_code,]</pre>
  zone_class <- zone_subset@data$Code_2</pre>
  if (verbose) {
    cat("Zona: ", zone_code,"\n")
    cat("Clase de zona: ", zone_class, "\n")
  }
  control_sampling_classes <- c(control_sampling_classes, zone_class)</pre>
}
# check the dataset dimensions
#dim(main_sampling_df[,c(1:raster_layers_count)])
#dim(M2)
#length(main_sampling_df[,"CL"])
# Predict the most probable class for each parcel
prediction_M2 <- knn(main_sampling_df[,c(1:8)], # training set</pre>
                  M2,
                                              # test set
                  main_sampling_df[,"CL"], # training class labels
                  k = 50, prob = TRUE)
M2_predict <- cbind(M2, prediction = prediction_M2, probability = attr(prediction_M2,
"prob"), original = control_sampling_classes)
table_M2_predict <- M2_predict[,c("original","prediction","probability")]</pre>
write.csv(table M2 predict,file = "tabla M2 predict.csv")
```

Tabla con valores de predicción para cada zona de muestreo complementario

```
table_M2_predict
```

##	original	prediction	probability
## 0	100	100	1.000000
## 1	100	100	1.000000
## 2	103	103	
## 3	192	192	
## 4	190	190	0.720000
## 5	100	105	0.620000
## 6	190	190	
## 7	120	115	
## 8	120	120	
## 9	160	161	0.500000
## 10	160	160	
## 11	150	150	1.000000
## 12	120	120	1.000000
## 13	105	105	1.000000
## 14	161	161	1.000000
## 15	115	120	0.620000
## 16	135	135	0.960000
## 17	105	105	1.000000
## 18	161	161	1.000000
## 19	115	115	1.000000
## 20	120	120	0.820000
## 21	161	161	1.000000
## 22	135	130	0.900000
## 23	110	110	0.540000
## 24	115	115	0.660000
## 25	130	130	0.900000
## 26	135	135	1.000000
## 27	120	130	1.000000
## 28	150	161	1.000000
## 29	120	130	1.000000
## 30	200	200	1.000000
## 31	161	161	0.800000
## 32	192	161	0.840000
## 33	150	150	1.000000
## 34	155	155	0.920000
## 35	190	110	0.440000
## 36	200	200	0.745098
## 37	135	135	0.960000
## 38	200	200	1.000000
## 39	100	100	1.000000
## 40	150	150	1.000000
## 41	155	155	0.460000
## 42	150	150	1.000000
## 43	161	161	1.000000
## 44	120	115	0.820000
## 45	135	135	1.000000
## 46	105	105	0.760000
## 47	120	115	0.480000
## 48	130	135	1.000000
## 49	105	103	1.000000
## 50	120	130	1.000000
## 51	160	161	0.920000
## 52	105	100	1.000000
## 53	120	120	0.580000
## 54	135	161	1.000000
## 55	120	120	0.680000
	•	_	

## !	56	120	115	0.940000
## !	57 :	160	150	0.700000
## !	58 :	192	130	0.420000
## !	59 :	120	120	0.500000
## (60	120	120	1.000000
## (61	161	161	0.920000
## (62	150	150	1.000000
## (63	120	130	1.000000
## (64	200	150	0.500000
## (65	110	155	0.580000
## (66	120	115	1.000000
## (67	120	115	1.000000
## (68	103	100	1.000000
## (69	101	100	1.000000
##	70	110	190	0.820000