# Geospatial Point Density

Anna Urbala

1 maja 2020

## Geospatial Point Density

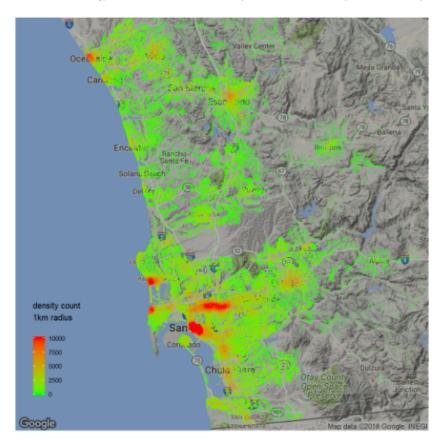
### by Paul F. Evangelista and David Beskow

```
Link
library(ggmap)
## Loading required package: ggplot2
## Google's Terms of Service: https://cloud.google.com/maps-platform/terms/.
## Please cite ggmap if you use it! See citation("ggmap") for details.
library(KernSmooth)
## KernSmooth 2.23 loaded
## Copyright M. P. Wand 1997-2009
library(pointdensityP)
# BKDE2D script (figure 1)
# edit line below to read data from file location
SD<-read.table("extra/GeospatialPointDensity/incidents-5y.csv", sep = ",", header = TRUE)
x<-cbind(SD$lon,SD$lat)
est < -bkde2D(x,bandwidth = c(.01,.01),gridsize = c(750,800),range.x = list(c(-117.45,-116.66),c(32.52,33.26)))
BKD_df <- data.frame(lat=rep(est$x2, each = 750),lon=rep(est$x1, 800),count=c(est$fhat))
map_base <- qmap(location="32.9,-117.1", zoom = 10, darken=0.3)</pre>
## Error: Google now requires an API key.
          See ?register_google for details.
png("SD_bkde2D_test.png", width = 1000, height = 1000, units = "px")
map_base+stat_contour(bins=150,geom="polygon",aes(x=lon, y=lat, z=count, fill = ..level..), data = BKD_
## Error in eval(expr, envir, enclos): nie znaleziono obiektu 'map_base'
dev.off()
## pdf
# point density script (figure 2)
SD_density <- pointdensity(df = SD, lat_col = "lat", lon_col = "lon", date_col = "date", grid_size = 0.
SD_density$count[SD_density$count>10000] <- 10000 ## creates discriminating scale
png("SD_pointdensity_test.png", width = 1000, height = 1000, units = "px")
```

map\_base + geom\_point(aes(x = lon, y = lat, colour = count), shape = 16, size = 0.5, data = SD\_density)

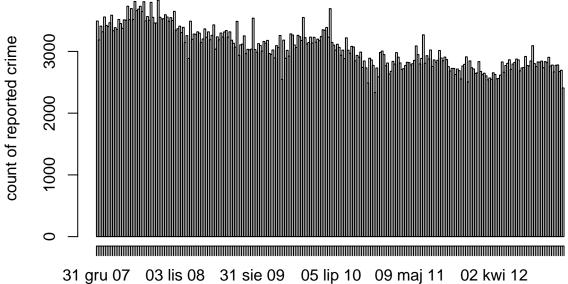
```
## Error in eval(expr, envir, enclos): nie znaleziono obiektu 'map_base'
dev.off()
```

Przez brak dostępu do API nie można wykonać obrazków przedstawionych w artykule:



```
# temporal tendency script (figure 3)
SD_temp_tend <- SD_density[SD_density$dateavg > 0]
#trim upper and lower tails for discriminating visualization
SD_temp_tend$dateavg[SD_temp_tend$dateavg<14711] <- 14711
SD_temp_tend$dateavg[SD_temp_tend$dateavg>14794] <- 14794
png("SD_temp_tend_test.png", width = 1000, height = 1000, units = "px")
map_base + geom_point(aes(x = lon, y = lat, colour = dateavg), shape = 16, size = 0.5, data = SD_temp_t
## Error in eval(expr, envir, enclos): nie znaleziono obiektu 'map_base'
dev.off()
## pdf
## 2
Znowu mamy problem z uzyskaniem grafiki przez brak obiektu pobranego z API Google'a.
#histogram plots in figure 3 and simple linear regression model to measure trends
#San Diego Crime Set
x <- as.Date(SD$date)
hist(x,"weeks",format = "%d %b %y", freq = TRUE, xlab = "week", ylab = "count of reported crime", main simple linear regression.</pre>
```

# weekly count of crime

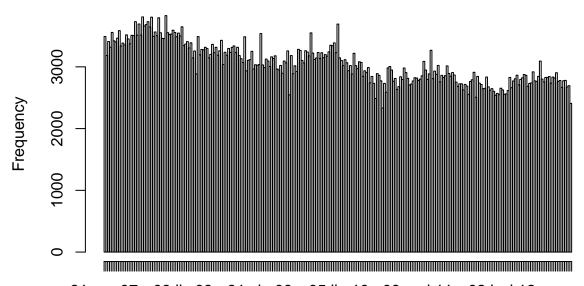


02 kwi 12

week

res <- hist(x,"weeks",format = "%d %b %y", freq = TRUE)</pre>

# Histogram of x



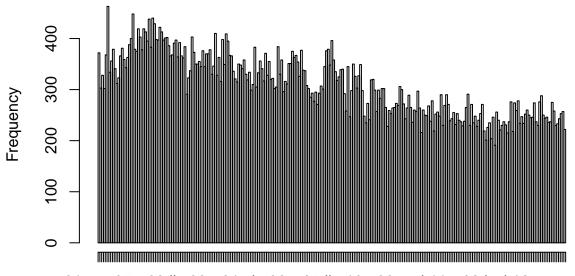
31 gru 07 03 lis 08 31 sie 09 05 lip 10 09 maj 11 02 kwi 12

Χ

SanDiegoTotal <- res\$breaks[1:(length(res\$breaks)-1)]</pre> model <- lm(res\$counts ~ SanDiegoTotal)</pre> summary(model)

```
##
## Call:
## lm(formula = res$counts ~ SanDiegoTotal)
##
## Residuals:
##
       Min
                1Q Median
                                 3Q
                                        Max
   -635.53 -105.51
                     -1.76 107.33
                                    635.61
##
## Coefficients:
##
                   Estimate Std. Error t value Pr(>|t|)
## (Intercept)
                 10565.6442
                               301.8892
                                          35.00
                                                  <2e-16 ***
## SanDiegoTotal
                    -0.5077
                                 0.0204
                                         -24.89
                                                  <2e-16 ***
##
## Signif. codes:
                   0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 173.8 on 259 degrees of freedom
## Multiple R-squared: 0.7051, Adjusted R-squared: 0.704
## F-statistic: 619.3 on 1 and 259 DF, p-value: < 2.2e-16
#Mid-city
mid_city <- subset(SD, lat > 32.69 & lon > -117.14 & lat < 32.79 & lon < -117.08)
x <- as.Date(mid_city$date)</pre>
hist(x,"weeks",format = "%d %b %y", freq = TRUE)
res <- hist(x, "weeks", format = "%d %b %y", freq = TRUE)
```

## Histogram of x



31 gru 07 03 lis 08 31 sie 09 05 lip 10 09 maj 11

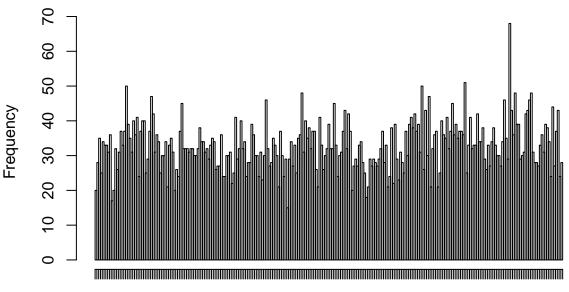
```
Χ
```

```
xres <- res$breaks[1:(length(res$breaks)-1)]</pre>
model <- lm(res$counts ~ xres)</pre>
summary(model)
```

## ## Call:

```
## lm(formula = res$counts ~ xres)
##
## Residuals:
##
       Min
                1Q Median
                                3Q
                                       Max
##
   -92.696 -22.322
                    -2.277
                           22.256
                                    86.782
##
## Coefficients:
##
                 Estimate Std. Error t value Pr(>|t|)
## (Intercept) 1.725e+03 5.491e+01
                                       31.42
                                               <2e-16 ***
               -9.577e-02 3.711e-03 -25.81
## xres
                                               <2e-16 ***
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 31.62 on 259 degrees of freedom
## Multiple R-squared: 0.72, Adjusted R-squared: 0.7189
## F-statistic: 666.1 on 1 and 259 DF, p-value: < 2.2e-16
#Encinitas
encinitas <- subset(SD, lat > 33 & lon > -117.32 & lat < 33.09 & lon < -117.27)
x <- as.Date(encinitas$date)</pre>
hist(x,"weeks",format = "%d %b %y", freq = TRUE)
res <- hist(x,"weeks",format = "%d %b %y", freq = TRUE)</pre>
```

## Histogram of x



31 gru 07 03 lis 08 31 sie 09 05 lip 10 09 maj 11 02 kwi 12

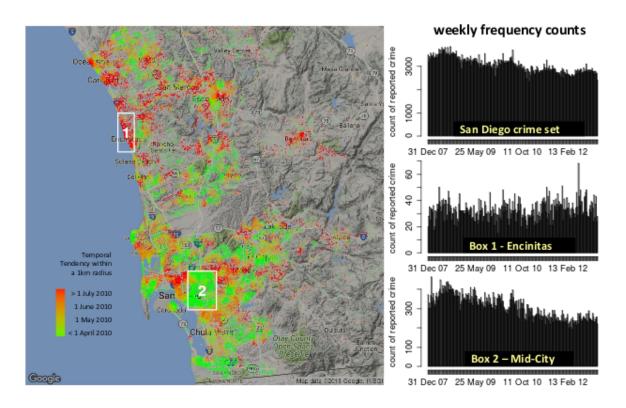
```
xres <- res$breaks[1:(length(res$breaks)-1)]
model <- lm(res$counts ~ xres)
summary(model)
##
## Call:</pre>
```

## lm(formula = res\$counts ~ xres)

##

```
## Residuals:
##
      Min
               1Q Median
                               30
                                      Max
  -17.380 -4.761 -0.201
                            4.115 33.599
##
## Coefficients:
                Estimate Std. Error t value Pr(>|t|)
##
## (Intercept) -1.6726492 11.9752552 -0.140 0.88902
               0.0023281 0.0008093
                                      2.877 0.00435 **
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.01 '*' 0.05 '.' 0.1 ' ' 1
## Residual standard error: 6.895 on 259 degrees of freedom
## Multiple R-squared: 0.03096,
                                 Adjusted R-squared: 0.02722
## F-statistic: 8.275 on 1 and 259 DF, p-value: 0.004353
```

Powyższe histogramy są takie jak w artykule:



Również statystyki są takie same:

	Estimate	Std. Error	t statistic	Pr(> t )
San Diego Total	-0.5077	0.0204	-24.89	<2e-16
Encinitas	0.0023	0.0008	2.88	0.00435
Mid-City	-0.0958	0.0037	-25.81	<2e-16

## Remaining code in the file supports the time comparison shown in table 1
#pointdensity\_original() below is necessary to run the timed comparisons between the hash-based and mat

```
pointdensity_original <- function(df, lat_col, lon_col, date_col = NULL, grid_size, radius){
  grid_size <- round(grid_size/111.2, digits = 3)</pre>
  rad km <- radius
                             ## initial radius measurement in km
  rad_dg <- rad_km/111.2
                                  ## radius as a latitudinal distance
  rad_steps <- round(rad_dg/grid_size) ## number of steps within grid
  rad_km <- rad_steps * grid_size * 111.2 ## radius rounded to nearest grid step</pre>
  cat("\nThe radius was adjusted to ",rad_km,"km in order to accommodate the grid size\n\n")
  cat("algorithm grid step radius is ",rad_steps,"\n\n")
  radius <- rad_steps
                                  ## assign to original variable
  h<-new.env(hash=TRUE)
                                  ## hash that will store the density count
  avg_date<-new.env(hash=TRUE)</pre>
                                      ## hash that will store the average date
  bh <- new.env(hash=TRUE)
                                  ## hash that will store the binned density count for a point
                                      ## hash that will store the binned date cont for a point
  b date<-new.env(hash=TRUE)</pre>
  #round all latitude data to nearest grid
  lat_data <- df[,lat_col]</pre>
  lat<-lat_data*(1/grid_size)</pre>
  lat<-round(lat,0)</pre>
  lat<-lat*(grid_size)</pre>
  #round all longitude data to nearest grid
  lon_data <- df[,lon_col]</pre>
  lon<-lon_data*(1/grid_size)</pre>
  lon<-round(lon,0)</pre>
  lon<-lon*(grid size)</pre>
  if(is.null(date_col)){
    date <- rep(0,length(lon))</pre>
  }
  if(!is.null(date_col)){
    date <- as.Date(df[,date_col])</pre>
    date <- as.numeric(date)</pre>
  key.vec<-paste(lat,lon,sep="-")</pre>
  data_length <- length(lat)</pre>
  ulat <- c()
  ulon <- c()
  cat("binning data...\n\n")
  pb <- txtProgressBar(title="point density calculation progress", label="0% done", min=0, max=100, ini
  for(i in 1:data_length){
    key<-paste(lat[i], lon[i], sep="-")</pre>
    if(is.null(h[[key]])){
      bh[[key]]=1
      h[[key]]=1
      b_date[[key]] = date[i]
      avg_date[[key]] = b_date[[key]]
      ulat <- c(ulat,lat[i])</pre>
      ulon <- c(ulon,lon[i])</pre>
    }
    else{
      bh[[key]]<-bh[[key]]+1
      h[[key]]<-bh[[key]]
      b_date[[key]] = b_date[[key]] + date[i]
```

```
avg_date[[key]] = b_date[[key]]
  }
  \#cat("\n",i,lat[i],lon[i],h[[key]],avg\_date[[key]],"\n")
  setTxtProgressBar(pb, i/(data_length)*100, label=info)
cat("\n", "Data length is ", data_length, "; reduced to ", length(ulat), "bins. Density calculation s
lat <- ulat
lon <- ulon
pb <- txtProgressBar(title="point density calculation progress", label="0% done", min=0, max=100, ini
counter<-0
data_length <- length(lat)</pre>
pb2 <- txtProgressBar(title="point density calculation progress", label="0% done", min=0, max=100, in
for(i in 1:data_length){
  counter <- counter + 1
  if(counter > 99){
    flush.console()
    counter <- 0
  }
  ukey<-paste(lat[i], lon[i], sep="-")</pre>
  lat.vec<-seq(lat[i]-radius*grid_size,lat[i]+radius*grid_size,grid_size)</pre>
  for(lat.temp in lat.vec){
    t<-sqrt(round(((radius*grid_size)^2-(lat.temp-lat[i])^2),8))
    t<-t/cos(lat.temp*2*pi/360)
    t<-t/grid_size
    t<-round(t,0)
    t<-t*grid_size
    lon.vec<-seq(lon[i]-t,lon[i]+t,grid_size)</pre>
    for(lon.temp in lon.vec){
      key<-paste(lat.temp, lon.temp, sep="-")</pre>
  if(is.null(h[[key]])){
        h[[key]]=bh[[ukey]]
        avg_date[[key]]=b_date[[ukey]]
      }
      else{
    if(key != ukey){
      h[[key]]<-h[[key]]+bh[[ukey]]
          avg_date[[key]] = avg_date[[key]] + b_date[[ukey]]
    }
  }
  \#cat(lat.temp, lon.temp, h[[key]], avg\_date[[key]], "\n")
  }
 \#cat("\n here again ",ukey, lat[i],lon[i],h[[ukey]],"avg_date", avg_date[[ukey]],"\n")
 info <- sprintf("%d%% done", round((i/data_length)*100))</pre>
 #setWinProgressBar(pb, i/(data_length)*100, label=info)
 setTxtProgressBar(pb2, i/(data_length)*100, label=info)
}
close(pb)
count_val <- rep(0,length(key.vec))</pre>
avg_date_val <- rep(0,length(key.vec))</pre>
```

```
for(i in 1:length(key.vec)){
    count_val[i] <- h[[key.vec[i]]]</pre>
    avg_date_val[i] <- avg_date[[key.vec[i]]]/count_val[i]</pre>
    count_val[i] <- count_val[i]/(pi*rad_km^2)</pre>
  }
  final <-data.frame(lat=lat_data,lon=lon_data,count=count_val,dateavg = avg_date_val)
  final<-final[order(final$count),]</pre>
 return(final)
  cat("done...\n\n")
}
matrix_time <- rep(0,6)</pre>
hash_time <- rep(0,6)
data_size \leftarrow rep(0,6)
for(i in 1:6){
  number_rows = 10^i
  SD_sample <- SD[sample(nrow(SD), number_rows, replace = TRUE),]</pre>
  data_size[i] = number_rows
  ptm <- proc.time()</pre>
  SD_density_original <- pointdensity_original(df = SD_sample, lat_col = "lat", lon_col = "lon", date_c
  proc_time_original <- proc.time() - ptm</pre>
  hash_time[i] = proc_time_original[[3]]
 ptm <- proc.time()</pre>
  SD_density_n <- pointdensity(df = SD_sample, lat_col = "lat", lon_col = "lon", date_col = "date", gri-
  proc_time_n <- proc.time() - ptm</pre>
  matrix_time[i] = proc_time_n[[3]]
time_compare_table <- data.frame(data_size,hash_time,matrix_time)</pre>
#time_compare_table produces results for comparison to table 1
time_compare_table
##
    data_size hash_time matrix_time
## 1
        1e+01
                 0.105
                               0.033
## 2
        1e+02
                  0.331
                               0.140
                  3.127
                               1.251
## 3
         1e+03
## 4
        1e+04
               24.222
                              10.246
## 5
        1e+05 105.334
                              33.606
## 6
         1e+06 245.455
                              63.317
Wyniki czasowe są inne, ale wynika to z użycia innego sprzętu.
                                                                 Wnioski pozostają te same.
                                 10 	 10^2 	 10^3
                                                           10^{4}
                                                                     10^{5}
                                                                                 10^{6}
   data records
                                 0.06 0.14 1.19 10.60 265.65
  hash-based time (s)
                                                                                 857.15
   matrix-based time (s)
                                 0.03 0.15 1.19 7.97
                                                                     33.35
                                                                                 74.88
```

#### **Problemy**

- trzeba dodatkowo skonfigurować dostęp do API Google'a
- API Google'a jest płatne

#### Jak naprawić

• trzeba zapłacić Google'owi za dostęp do API

#### Podsumowanie

Kategoria	Ocena
Dostęp do zewnętrznych zasobów	++×××
Kompatybilność z nowszymi wersjami	+++++
Kwestie graficzne/estetyczne	+++++
Brak problemów przy dodatkowej konfiguracji	++++×
Odporność na wpływ losowości	
Dostępność kodów źródłowych	++++

#### Session info

```
## R version 3.6.3 (2020-02-29)
## Platform: x86_64-pc-linux-gnu (64-bit)
## Running under: Debian GNU/Linux 9 (stretch)
## Matrix products: default
          /usr/lib/openblas-base/libblas.so.3
## BLAS:
## LAPACK: /usr/lib/libopenblasp-r0.2.19.so
##
## locale:
##
   [1] LC_CTYPE=pl_PL.UTF-8
                                   LC_NUMERIC=C
  [3] LC_TIME=pl_PL.UTF-8
                                   LC_COLLATE=pl_PL.UTF-8
   [5] LC_MONETARY=pl_PL.UTF-8
                                   LC_MESSAGES=pl_PL.UTF-8
   [7] LC_PAPER=pl_PL.UTF-8
                                   LC_NAME=C
                                   LC_TELEPHONE=C
##
  [9] LC_ADDRESS=C
## [11] LC_MEASUREMENT=pl_PL.UTF-8 LC_IDENTIFICATION=C
##
## attached base packages:
## [1] stats
                 graphics grDevices utils
                                               datasets methods
                                                                    base
## other attached packages:
## [1] png_0.1-7
                           pointdensityP_0.3.4 KernSmooth_2.23-16
## [4] ggmap_3.0.0
                           ggplot2_3.2.1
##
## loaded via a namespace (and not attached):
## [1] Rcpp_1.0.3
                            plyr_1.8.4
                                                pillar_1.4.2
## [4] compiler_3.6.3
                            bitops_1.0-6
                                                tools_3.6.3
                                                lifecycle_0.1.0
## [7] zeallot_0.1.0
                            digest_0.6.22
## [10] evaluate_0.14
                            tibble_2.1.3
                                                gtable_0.3.0
## [13] lattice_0.20-40
                            pkgconfig_2.0.3
                                                rlang_0.4.1
## [16] rstudioapi_0.10
                            yam1_2.2.0
                                                xfun_0.10
                            withr_2.1.2
## [19] httr_1.4.1
                                                dplyr_0.8.3
## [22] stringr_1.4.0
                            knitr_1.25
                                                vctrs_0.2.0
## [25] RgoogleMaps_1.4.5.3 grid_3.6.3
                                                tidyselect_0.2.5
## [28] data.table_1.12.6
                                                R6 2.4.0
                            glue_1.3.1
## [31] jpeg_0.1-8.1
                            rmarkdown_1.16
                                                sp_1.4-1
```

## [34] tidyr_1	.0.0 purr	r_0.3.3	magrittr_1.5
## [37] backpor	ts_1.1.5 scal	es_1.0.0	htmltools_0.4.0
## [40] assertt	hat_0.2.1 colo	rspace_1.4-1	stringi_1.4.3
## [43] lazyeva	1_0.2.2 muns	ell_0.5.0	rjson_0.2.20
## [46] crayon_	1.3.4		