sdcSpatial: Privacy protected density maps

Edwin de Jonge @edwindjonge

Statistics Netherlands Research & Development @edwindjonge

useR! 2019, July 11 2019



sdcSpatial: Privacy protected maps





sdcSpatial: Privacy protected maps

sdcSpatial has methods for:

- Creating a raster map: sdc_raster for pop density, value density and mean density, using the excellent raster package by Hijmans (2019).
- Finding out which locations are sensitive: plot_sensitive, is sensitive.
- Adjusting raster map for protecting data: protect_smooth, protect_quadtree.
- · Removing sensitive locations.



Who am I?

- Statistical consultant, Data Scientist @cbs.nl / Statistics NL
- Expertise:
 - R programming
 - Data Cleaning with R
 - Data visualization
 - Complex networks analysis
 - @edwindjonge / https://github.com/edwindj



What is SN / CBS?

Statistics Netherlands is producer main official statistics in the Netherlands:

- Stats on Demographics, economy (GDP), education, environment, agriculture, Finance etc.
- Part of the European Statistical System, ESS.

Motivation for sdcSpatial

 ESS has European Code of Statistical Practice (predates GDPR, European law on Data Protection):
 no individual information may be revealed.



Sdc in sdcSpatial?

SDC = "Statistical Disclosure Control"

Collection of statistical methods to:

- · Check if data is safe to be published
- Protect data by slightly altering (aggregated) data
 - adding noise
 - shifting mass
- Most SDC methods operate on records.
- sdcSpatial works upon locations.



Let's create a sdc_raster

Creation:

What is it about?

```
print(unemployed)
## logical sdc_raster object:
## resolution: 500 500 , max_risk: 0.95 , min_count: 10
## mean sensitivity score [0,1]: 0.4249471
```

42% of the data on this map is sensitive!



What is sensitivity?

Binary score (logical) per raster cell indicating if it's unsafe to publish.

Calculated:

- a) Per location (x_i, y_i) (raster cell)
- b) Using risk function disclosure_risk $r(x,y) \in [0,1]$. How accurate can an attacker estimate the value of an individual? If $r(x_i, y_i) > \max_{x \in \mathbb{R}} then (x_i, y_i)$ is sensitive.
- c) Using a minimum number of observations. If $count_i < min_count$, then (x_i, y_i) is sensitive.



Type of raster density maps:

Density can be area-based:

- **number of people** per square (unemployed\$value\$count): population density.
- (total) value per square (unemployed\$value\$sum): number of unemployed per square.

Or density can population-based:

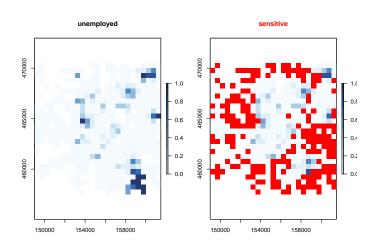
 Mean value per square (unemployed\$value\$mean): probability of being unemployed per square.

Note: All density types are valid, but (total) value per square strongly interacts with population density. (e.g. https://xkcd.com/1138).



Plotting a sdc_raster

plot(unemployed, "mean")





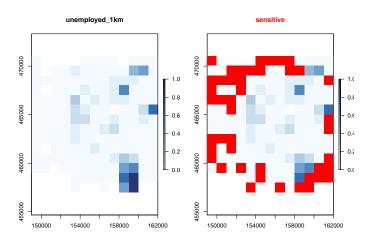
How to reduce sensitivity?

Options:

- a) Use a coarser raster: sdc_raster.
- b) Apply spatial smoothing: protect_smooth method by Wolf and Jonge (2018), Jonge and Wolf (2016).
- c) Aggregate with a Quad tree: protect_quadtree method by Suñé et al. (2017).
- d) Remove sensitive locations: remove_sensitive.



Option: coarser raster





Option: Coarsening

Pros

• Simple and easy explainable

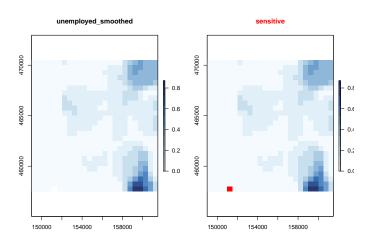
Cons

- · Detailed spatial patterns are removed
- visually unattractive: "Blocky"



Option: KDE-smoothing

unemployed_smoothed <- protect_smooth(unemployed, bw = 1500)
plot(unemployed_smoothed, "mean")</pre>





Options: KDE-smoothing

Pro's

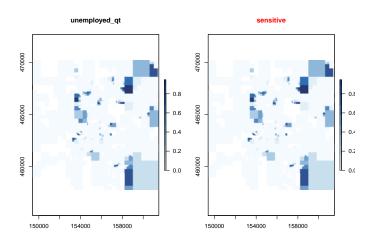
- Often enhances spatial pattern visualization, removing spatial noise.
- Makes it a density map and used as source for e.g. contour map.

Con's

- Does not remove all sensitive values (depends on bandwidth bw)
- A fixed band width is used for all locations: may remove detailed patterns...
 spatial processes often have location dependent band widths. (= future work)



Option: Quad tree





Option: Quad tree

Pro

- · Adapts to data density
- Adjusts until no sensitive data is left.

Cons

• Visually: "Blocky" / "Mondrian-like" result.

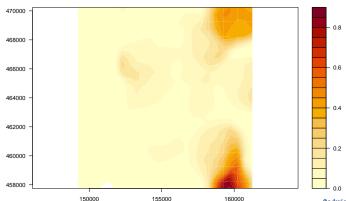


Publish: visual interpolation

So in 5 lines we create a visual attractive map that is safe:

```
unemployed <- sdc_raster(dwellings[c("x","y")], dwellings$unemployed, r=500)
unemployed_smoothed <- protect_smooth(unemployed, bw = 1500)
unemployed_safe <- remove_sensitive(unemployed_smoothed)
mean_unemployed <- mean(unemployed_safe)
raster::filledContour(mean_unemployed, main="Unemployment rate")</pre>
```







The end

Thank you for your attention!

Questions?

Curious?

install.packages("sdcSpatial")

Feedback and suggestions?

https://github.com/edwindj/sdcSpatial/issues



References

Hijmans, Robert J. 2019. *Raster: Geographic Data Analysis and Modeling*. https://CRAN.R-project.org/package=raster.

Jonge, Edwin de, and Peter-Paul de Wolf. 2016. "Spatial Smoothing and Statistical Disclosure Control." In *Privacy in Statistical Databases*, edited by Josep Domingo-Ferrer and Mirjana Pejić-Bach, 107–17. Springer.

Suñé, E., C. Rovira, D. Ibáñez, and M. Farré. 2017. "Statistical Disclosure Control on Visualising Geocoded Population Data Using Quadtrees."

http://nt17.pg2.at/data/x_abstracts/x_abstract_286.docx.

Wolf, Peter-Paul de, and Edwin de Jonge. 2018. "Spatial Smoothing and Statistical Disclosure Control." In *Privacy in Statistical Databases - Psd 2018*, edited by Josep Domingo-Ferrer and Francisco Montes Suay. Springer.

