# Nonparametric Analysis of US Dairy Production and Consumption Spatial Nonparametric

Teo Bucci\* Filippo Cipriani† Gabriele Corbo‡ Andrea Puricelli\$2023-02-17

### Contents

1	Load libraries and data	1
2	Study of the Dairy prodction in 2007	1
3	Trend estimates for log_value	2
4	Bandwidth selection	3
5	Variogram estimation	4
6	Automatic modelling	5
7	Kriging	6

### 1 Load libraries and data

```
library("readx1")
library(sf)
library(maps)
library(ggspatial)
library(tidyr)
library(raster)
library(sp)
library(npsp)
library(ggplot2)
source("05-Prepare-data-spatial.R",echo=FALSE)
```

## 2 Study of the Dairy prodction in 2007

Using log values of the population and the dairy sales to avoid scalability problems and setting up the coordinates for spatial analysis

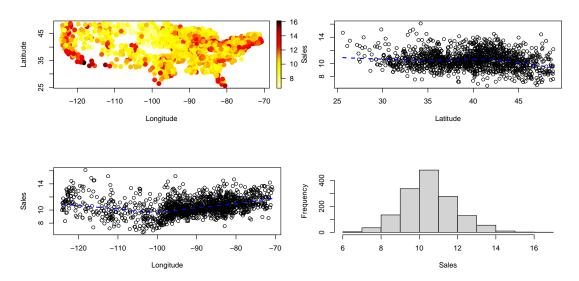
<sup>\*</sup>teo.bucci@mail.polimi.it

 $<sup>^\</sup>dagger filippo.cipriani@mail.polimi.it$ 

 $<sup>^{\</sup>ddagger} gabriele.corbo@mail.polimi.it$ 

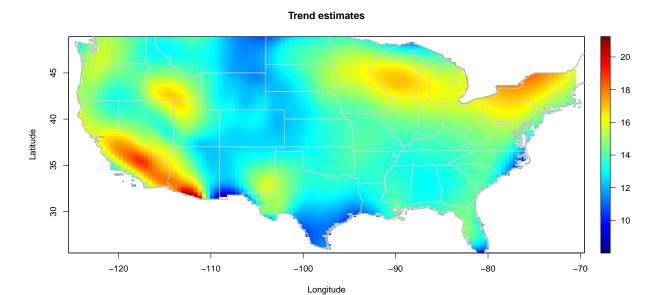
 $<sup>\</sup>S$  and rea 3. puricelli@mail.polimi.it

#### Milk and cheese sales



## 3 Trend estimates for log\_value

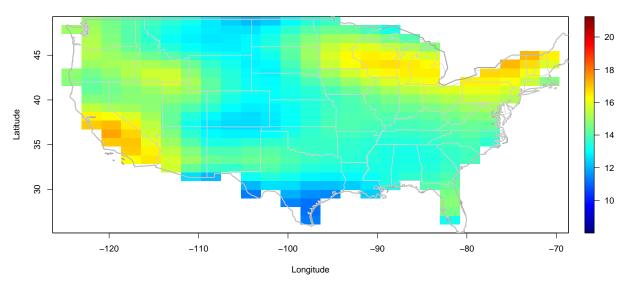
Plotting the estimates of the model using a local linear estimator



### 4 Bandwidth selection

Tuning of the bandwidth using  ${\rm CV}$ 

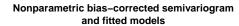
#### Trend estimates

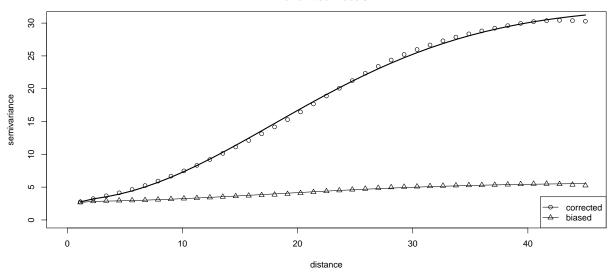


## 5 Variogram estimation

Plotting the estimates of the bias-corrected semivariogram

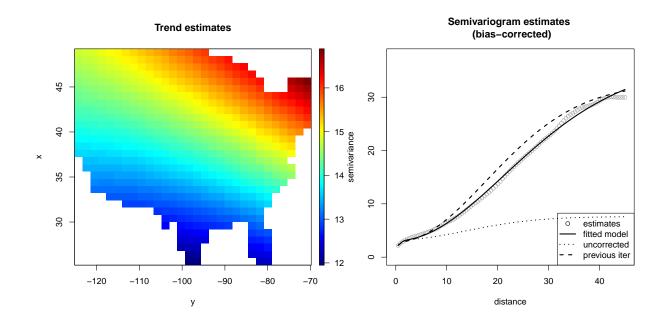
```
lp0 <- locpol(bin, h = lp0.h, hat.bin = TRUE)</pre>
svar.bin <- svariso(x, residuals(lp0), nlags = 40, maxlag = 45)</pre>
svar.h <- h.cv(svar.bin)$h</pre>
svar.np <- np.svar(svar.bin, h = svar.h)</pre>
svar.np2 <- np.svariso.corr(lp0, nlags = 40, maxlag = 45,</pre>
                             h = svar.h, plot = FALSE)
svm0 <- fitsvar.sb.iso(svar.np, dk = 0)</pre>
svm1 <- fitsvar.sb.iso(svar.np2, dk = 0)</pre>
plot(svm1, main = "Nonparametric bias-corrected semivariogram\nand fitted models",
     legend = FALSE, xlim = c(0,max(coords(svar.np2))),
     ylim = c(0,max(svar.np2$biny, na.rm = TRUE)))
plot(svm0, lwd = c(1, 1), add = TRUE)
plot(svar.np, type = "p", pch = 2, add = TRUE)
# abline(h = c(svm1$nuqqet, svm1$sill), lty = 3)
# abline(v = 0, lty = 3)
legend("bottomright", legend = c("corrected", 'biased'),
       lty = c(1, 1), pch = c(1, 2), lwd = c(1, 1))
```





# 6 Automatic modelling

```
np.fitgeo() repeats the same procedure all in one step
```



# 7 Kriging

Using the model estimates and the semivariogram we can use a kriging prediction method

