# Spatio-temporal overlay and aggregation





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#### Abstract

The so-called "map overlay" is not very well defined and does not have a simple equivalent in space-time. This paper will explain how the over method for combining two spatial features (and/or grids), defined in package sp and extended in package rgeos, is implemented for spatio-temporal objects in package spacetime. It may carry out the numerical spatio-temporal overlay, and can be used for aggregation of spatio-temporal data over space, time, or space-time.

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### 1 Introduction

The so-called *map overlay* is a key GIS operation that does not seem to have a very sharp definition. The over vignette in package sp (Pebesma, 2012) comments on what paper (visual) overlays are, and discusses the over and aggregate methods for spatial data.

In the ESRI ArcGIS tutorial (ESRI, 2012), it can be read that

An overlay operation is much more than a simple merging of linework; all the attributes of the features taking part in the overlay are carried through, as shown in the example below, where parcels (polygons) and flood zones (polygons) are overlayed (using the Union tool) to create a new polygon layer. The parcels are split where they are crossed by the flood zone boundary, and new polygons created. The FID\_flood value indicates whether polygons are outside (-1) or inside the flood zone, and all polygons retain their original land use category values.

It later on mentions raster overlays, such as the addition of two (matching) raster layers (so, potentially the whole of map algebra functions, where two layers are involved).

In the open source arena, with no budgets for English language editing, the Grass 7.0 documentation mentions the following:

v.overlay allows the user to overlay two vector area maps. The resulting output map has a merged attribute-table. The origin columnnames have a prefix (a\_ and b\_) which results from the ainput- and binput-map. [...] Operator defines features written to output vector map Feature is written to output if the result of operation 'ainput operator binput' is true. Input feature is considered to be true, if category of given layer is defined. Options: and, or, not, xor.

## 2 Overlay with method over

We loosely define map overlay as

- an operation involving at least two maps
- asymmetric overlay is different from underlay
- either a visual or a numerical activity.

The method over, as defined in package sp, provides a way to numerically combine two maps. In particular,

```
R> over(x, geometry(y))
```

returns an integer vector of length length(x) with x[i] the index of y, spatially corresponding to x[i], so x[i]=j means that x[i] and y[j] match (have the same location, touch, or overlap/intersect etc.), or x[i]=NA if there is no match. If y has data values (attributes), then

```
R > over(x, y)
```

retrieves a data.frame with length(x) rows, where row i contains the attributes of y at the spatial location of x[i], and NA values if there is no match.

If the relationship is more complex, e.g. a polygon or grid cell x containing more than one point of y, the command

```
R> over(x, y, returnList = TRUE)
```

returns a list of length length(x), with each list element a numeric vector with all indices if y is geometry only, or else a data frame with all attribute table rows of y that spatially matches x[i].

### 3 Spatio-temporal overlay with method over

Package spacetime adds over methods to those defined for spatial data in package sp:

```
R> library(spacetime)
R> showMethods(over)
Function: over (package sp)
x="ST", y="STS"
x="STF", y="STF"
x="STF", y="STFDF"
x="STF", y="STI"
x="STF", y="STIDF"
x="STF", y="STSDF"
x="STI", y="STF"
x="STI", y="STFDF"
x="STI", y="STI"
x="STI", y="STIDF"
x="STI", y="STSDF"
x="STS", y="STF"
x="STS", y="STFDF"
x="STS", y="STI"
x="STS", y="STIDF"
x="STS", y="STSDF"
x="SpatialGrid", y="SpatialPolygons"
x="SpatialGrid", y="SpatialPolygonsDataFrame"
x="SpatialPoints", y="SpatialGrid"
x="SpatialPoints", y="SpatialGridDataFrame"
x="SpatialPoints", y="SpatialPixels"
x="SpatialPoints", y="SpatialPixelsDataFrame"
x="SpatialPoints", y="SpatialPoints"
x="SpatialPoints", y="SpatialPointsDataFrame"
x="SpatialPoints", y="SpatialPolygons"
x="SpatialPoints", y="SpatialPolygonsDataFrame"
x="SpatialPolygons", y="SpatialGrid"
x="SpatialPolygons", y="SpatialGridDataFrame"
x="SpatialPolygons", y="SpatialPoints"
x="SpatialPolygons", y="SpatialPointsDataFrame"
x="xts", y="xts"
```

### 3.1 Time intervals or time instances?

When computing the overlay

```
R> over(x, y)
```

A space-time feature matches another space-time feature when their spatial locations match (coincide, touch, intersect or overlap), and when their temporal properties match. For temporal properties, it is crucial whether time is a time interval, or a time instance. When all endTime values are equal to the time

times, time is considered instance. When one or more endTime values are larger than time, time is considered to reflect intervals.

Suppose we have two time sequences,  $T: t_1, t_2, ..., t_n$  and  $U: u_1, u_2, ..., u_m$ . Both are ordered:  $t_i \leq t_{i+1}$ .

Both T and U can reflect time *instances* or time *intervals*. In case they reflect time *instances*, an observation at  $t_i$  takes place at the time instance  $t_i$ , and has an unregistered (possibly ignorable) duration. In case they reflect time *intervals*, an observation "at"  $t_i$  takes place during, or is representative for, the time interval  $t_i \leq t < t_{i+1}$ . (The last time interval  $t_n$  is obtained by adopting the one-but-last time interval duration:  $t_n \leq t < t_n + (t_n - t_{n-1})$ ).

We define the time (instance or interval) pair  $\{t_i, u_j\}$  to match if

for T instance, U instance:

$$t_i = u_j$$

for T interval, U instance

$$t_i \le u_j < t_{i+1}$$

for T instance, U interval

$$u_j \le t_i < u_{j+1}$$

for T interval, U interval

$$\exists t : t_i \le t < t_{i+1} \land u_j \le t < u_{j+1}$$

which can be rephrased as the negation of  $t_{i+1} \leq u_j \vee t_i \geq u_{j+1}$  (where  $\vee$  denotes 'or'), or alternatively expressed as

$$t_{i+1} > u_j \wedge t_i < u_{j+1}$$

where  $\wedge$  denotes 'and'.

All these conditions fail for intervals having zero width (empty intervals), i.e. the case where T is interval and for some i,  $t_{i+1} - t_i = 0$  or the case where U is interval and for some j,  $u_{j+1} - u_j = 0$ .

## 4 Aggregating spatio-temporal data

The aggregate method for a data.frame is defined as

$$R > aggregate(x, by, FUN, ..., simplify = TRUE)$$

where x is the data.frame to be aggregated, by indicates how groups of x are formed, FUN is applied to each group, and simplify indicates whether the output should be simplified (to vector), or remain a data.frame. The ... are passed to FUN, e.g. passing na.rm=TRUE is useful when FUN is mean and missing values need to be ignored.

For spatio-temporal data, the x argument needs to be of class STFDF, STSDF or STIDF. The by argument needs to specify an aggregation medium: time, space, or space-time.

### 4.1 Example data: PM10

Air quality example data are loaded by

```
R> data(air)
R> class(rural)

[1] "STFDF"
attr(,"package")
[1] "spacetime"

R> class(DE_NUTS1)

[1] "SpatialPolygonsDataFrame"
attr(,"package")
[1] "sp"
```

it provides PM10 daily mean values (taken from AirBase - the European Air quality dataBase), for Germany, 1998-2009, where only stations classified as rural background were selected. The object DE\_NUTS1 contains NUTS-1 level state boundaries for Germany, downloaded from GADM.

### 4.2 Spatial aggregation

To aggregate *completely* over space, we can coerce the data to a matrix and apply a function to the rows:

```
R> x = as(rural[,"2008"], "xts")
R> apply(x, 1, mean, na.rm=TRUE)[1:5]

2007-12-31 2008-01-01 2008-01-02 2008-01-03 2008-01-04
17.34950 16.06945 25.60065 27.24141 24.03417
```

A more refined spatial aggregation of time series can be obtained by grouping them to the state ("Bundesland") level. Here, states are passed as a SpatialPolygons object:

```
Coordinates:
       min
                max
x 5.871619 15.03811
y 47.269858 55.05653
Is projected: FALSE
proj4string:
[+init=epsg:4326 +proj=longlat +ellps=WGS84 +datum=WGS84
+no_defs +towgs84=0,0,0]
Data attributes:
      ID_0
                           NAME_O
              IS0
                                        ID_1
 Min.
        :60
              DEU:13
                      Germany:13
                                   Min. :753.0
 1st Qu.:60
                                    1st Qu.:756.0
 Median:60
                                   Median :761.0
 Mean:60
                                   Mean :760.5
 3rd Qu.:60
                                    3rd Qu.:764.0
 Max.
       :60
                                   Max.
                                          :768.0
   NAME_1
                                        VARNAME_1 NL_NAME_1
 Length:13
                   Bavaria
                                              :1
                                                  NA's:13
 Class :character
                   Hesse
                                              :1
 Mode :character
                   Lower Saxony
                                              :1
                   Mecklenburg-West Pomerania:1
                   North Rhine-Westphalia
                                              :1
                    (Other)
                                              :3
                   NA's
                                              :5
     HASC_1
              CC_1
                       TYPE_1
                                ENGTYPE_1
                                              VALIDFR_1
                                                           VALIDTO_1
 DE.BE :1
            NA's:13
                      Land:13
                                State:13
                                           Unknown:13 Present:13
 DE.BR
       :1
 DE.BW
       :1
 DE.BY :1
 DE.HE :1
 DE.MV :1
 (Other):7
 REMARKS_1
            Shape_Leng
                             Shape_Area
 NA's:13
          Min. : 2.631
                           Min. :0.1172
           1st Qu.:14.529
                           1st Qu.:2.1541
          Median :16.891
                           Median :2.6645
          Mean :18.068
                           Mean :3.3126
           3rd Qu.:24.519
                            3rd Qu.:4.3832
          Max.
                 :32.255
                           Max. :8.6561
[[Temporal:]]
     Index
                           ..1
                           :3653
 Min.
       :2007-12-31
                     Min.
 1st Qu.:2008-03-31
                     1st Qu.:3744
 Median :2008-06-30
                     Median:3836
 Mean
      :2008-06-30
                     Mean
                            :3836
 3rd Qu.:2008-09-29
                     3rd Qu.:3927
 Max. :2008-12-30
                     Max. :4018
[[Data attributes:]]
```

```
PM10
Min. : 2.181
1st Qu.: 9.933
Median :13.755
Mean :15.023
3rd Qu.:18.371
Max. :68.750
NA's :366

R> stplot(x, mode = "tp")
the result of which is shown in
```

the result of which is shown in figure 1, which was created by

```
R> stplot(x, mode = "tp", par.strip.text = list(cex=.5))
```

An aggregation for all stations selected within a single area is obtained by merging all states:

```
R> library(rgeos)
R> DE = gUnionCascaded(DE_NUTS1)
```

and then aggregating the observations within Germany for each moment in time:

```
R> x = aggregate(rural[,"2008"], DE, mean, na.rm=TRUE)
R> class(x)
[1] "xts" "zoo"
R> plot(x)
```

### 4.3 Temporal aggregation

shown in figure 2.

To aggregate *completely* over time, we can coerce the data to a matrix and apply a function to the columns:

```
R> x = as(rural[,"2008"], "xts")
R> apply(x, 2, mean, na.rm=TRUE)[1:5]

DESH001 DENI063 DEUB038 DEBE056 DEBE062
    NaN 18.41594 NaN 20.76446 NaN
```

Aggregating values *temporally* is done by passing a character string or a function to the by argument. For monthly data, we will first select those stations that have measured (non-NA) values in 2008,

```
R> sel = which(!apply(as(rural[,"2008"], "xts"), 2, function(x) all(is.na(x))))
R> x = aggregate(rural[sel,"2008"], "month", mean, na.rm=TRUE)
R> stplot(x, mode = "tp")
```

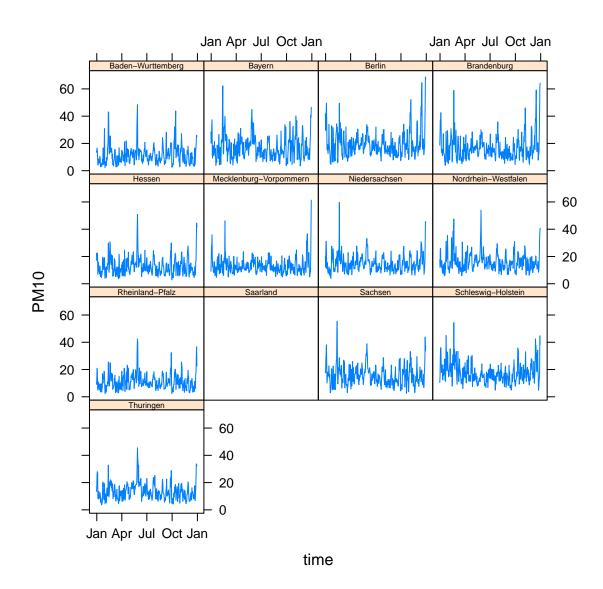


Figure 1: Daily PM10 values, aggregated (averaged) over states  $\,$ 

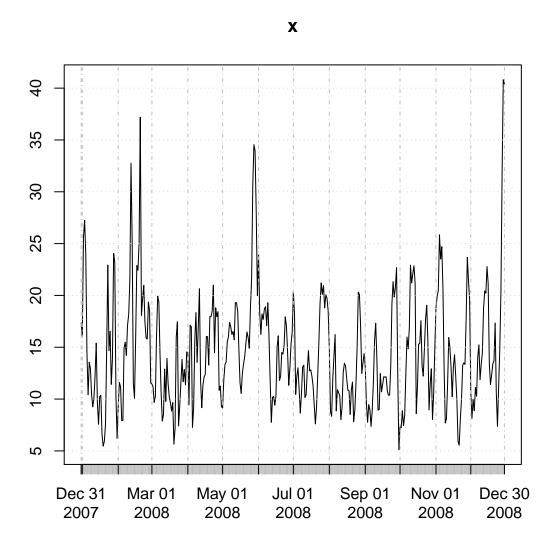


Figure 2: Time series plot of daily rural background PM10, averaged over Germany  $\,$ 

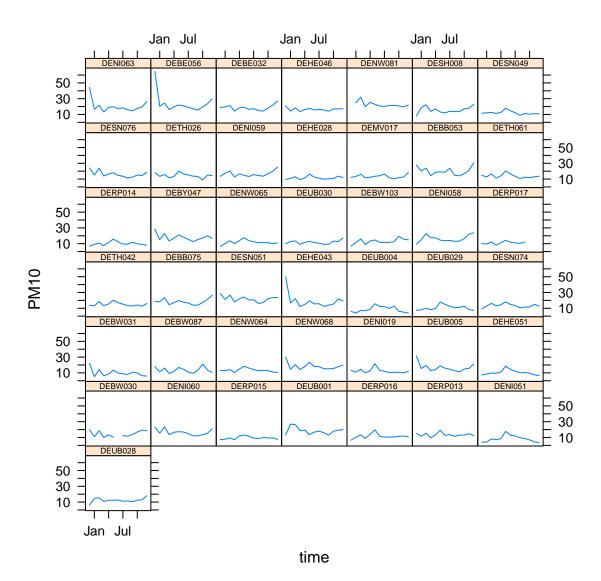


Figure 3: Monthly averaged PM10 values, for those rural background stations in Germany having measured values  $\,$ 

shown in figure 3

The strings that can be passed are e.g. "year", but also "3 days". See ?cut.Date for possible values. Aggregation using this way is only possible if the time index is of class Date or POSIXct.

An alternative is to provide a function for temporal aggregation. The function as.yearqtr from package zoo transforms dates to quarters, and hence allows aggregation to quarterly values, in this example medians:

```
R> x = aggregate(rural[sel,"2005::2011"], as.yearqtr, median, na.rm=TRUE)
R> stplot(x, mode = "tp")
```

shown in figure 4. Aggregating to monthly values is obtained by function as yearmon, aggregating to years by creating the function

```
R> as.year <- function(x) as.numeric(floor(as.yearmon(x)))</pre>
```

Further information can be found in ?aggregate.zoo, which is the function used to do the processing.

### 4.4 Spatio-temporal aggregation

Aggregation over spatio-temporal volumes can be done by passing an object inheriting from ST to the by argument:

```
R> DE.years = STF(DE, as.Date(c("2008-01-01", "2009-01-01")))
R> aggregate(rural[,"2008::2009"], DE.years, mean, na.rm=TRUE)
```

```
PM10 timeIndex
2008-01-01 14.66116 1
2009-01-01 15.12886 2
```

### 4.5 Time intervals

If all data concern time instances (endTime equals time), then time instances are being matched, else overlapping time intervals are being matched. In case intervals are being matched, empty intervals are never matched.

#### References

- ESRI (2012) ESRI ArcGIS Tutorial. http://webhelp.esri.com/arcgisdesktop/ 9.2/index.cfm?TopicName=Overlay\_analysis
- Pebesma, 2012. Map overlay and spatial aggregation in sp. Vignette in package sp, http://cran.r-project.org/web/packages/sp/vignettes/over.pdf

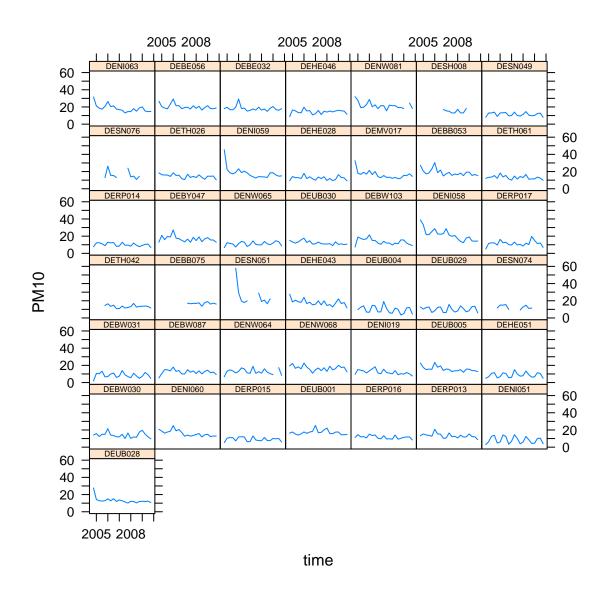


Figure 4: PM10 values, averaged to quarterly medians of daily averages  $\frac{1}{2}$