

# Classes and methods for spatio-temporal data in **R**: the spacetime package



**ifgi**  
Institute for Geoinformatics  
University of Münster

Edzer Pebesma

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## Abstract

This document describes a set of classes and methods for spatio-temporal data in R. It builds upon the classes and methods for spatial data taken from package **sp**, and the temporal classes in package **xts**. The goal is to cover a number of useful representations for spatio-temporal sensor data, or results from predicting (spatial and/or temporal interpolation or smoothing), aggregating, or subsetting them.

The goals of this package are to explore how spatio-temporal data can be sensibly represented in classes, and to find out which analysis and visualisation methods are useful and feasible for the classes implemented. It reuses existing classes, methods, and functions present in packages for spatial data (**sp**) and time series data (**zoo** and **xts**). Coercion to the appropriate reduced spatial and temporal classes is provided, as well as to **data.frame** objects in the obvious long or wide format.

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## 1 Introduction

Spatio-temporal data are abundant, and easily obtained. Examples are satellite images of parts of the earth, temperature readings for a number of nearby stations, election results for voting districts and a number of consecutive elections, GPS tracks for people or animals possibly with additional sensor readings, disease outbreaks or volcano eruptions.

Schabenberger and Gotway (2004) argue that analysis of spatio-temporal data often happens *conditionally*, meaning that either first the spatial aspect is analysed, after which the temporal aspects are analysed, or reversed, but not in a joint, integral modelling approach, where space and time are not separated. As a possible reason they mention the lack of good software, data classes and methods to handle, import, export, display and analyse such data. This R package is a start to fill this gap.

Spatio-temporall data are often relatively abundant in either space, or time, but not in both. Satellite imagery is typically very abundant in space, giving lots of detail in high spatial resolution for large areas, but relatively sparse in time. Analysis of repeated images over time may further be hindered by difference in light conditions, errors in georeferencing resulting in spatial mismatch, and changes in obscured areas due to changed cloud coverage. On the other side, data from fixed sensors give often very detailed signals over time, allowing for elaborate modelling, but relatively little detail in space because a very limited number of sensors is available. The cost of an in situ sensor network typically depends primarily on its spatial density; the choice of the temporal resolution with which the sensors register signals may have little effect on total cost.

Although for example Botts et al. (2007) describe a number of open standards that allow the interaction with sensor data (describing sensor characteristics, requesting observed values, planning sensors, and processing raw sensed data to predefined events), the available statistical or GIS software for this is in an early stage, and scattered. This paper describes an attempt to combine available infrastructure in the R statistical environment to a set of useful classes and methods for manipulating, plotting and analysing spatio-temporal data. A number of case studies from different application areas will illustrate its use.

The current version of the package is experimental, class definitions and methods are subject to change.

We use `xts` for time because it has nice tools for aggregation over time and a very flexible syntax to select time periods that adheres ISO 8601<sup>1</sup>. We do not use the `xts` objects to store attribute information, as it is restricted to `matrix` objects, and hence can only store a single type, and not combine numeric and factor. Instead, as in the classes of `sp`, we use `data.frame` to store measured values.

## 2 Space-time data in wide and long formats

Spatio-temporal data for which each location has data for each time can be provided in two so-called **wide formats**. An example where a single column refers to a single moment in time is found in the North Carolina Sudden Infant Death Syndrome (sids) data set, which is in the **time wide format**:

```
> library(foreign)
> read.dbf(system.file("shapes/sids.dbf", package = "maptools"))[1:5,
+      c(5, 9:14)]
```

	NAME	BIR74	SID74	NWBIR74	BIR79	SID79	NWBIR79
1	Ashe	1091	1	10	1364	0	19
2	Alleghany	487	0	10	542	3	12
3	Surry	3188	5	208	3616	6	260
4	Currituck	508	1	123	830	2	145
5	Northampton	1421	9	1066	1606	3	1197

where **columns** refer to a particular **time**: `SID74` contains to the infant death syndrome cases for each county at a particular time period (1974-1978).

The Irish wind data, for which the first six records are

```
> data(wind, package = "gstat")
> wind[1:6, ]
```

	year	month	day	RPT	VAL	ROS	KIL	SHA	BIR	DUB	CLA	MUL	CLO
1	61	1	1	15.04	14.96	13.17	9.29	13.96	9.87	13.67	10.25	10.83	12.58
2	61	1	2	14.71	16.88	10.83	6.50	12.62	7.67	11.50	10.04	9.79	9.67
3	61	1	3	18.50	16.88	12.33	10.13	11.17	6.17	11.25	8.04	8.50	7.67
4	61	1	4	10.58	6.63	11.75	4.58	4.54	2.88	8.63	1.79	5.83	5.88
5	61	1	5	13.33	13.25	11.42	6.17	10.71	8.21	11.92	6.54	10.92	10.34
6	61	1	6	13.21	8.12	9.96	6.67	5.37	4.50	10.67	4.42	7.17	7.50

<sup>1</sup>see [http://en.wikipedia.org/wiki/ISO\\_8601](http://en.wikipedia.org/wiki/ISO_8601)

	BEL	MAL
1	18.50	15.04
2	17.54	13.83
3	12.75	12.71
4	5.46	10.88
5	12.92	11.83
6	8.12	13.17

are in **space wide format**: each *column* refers to another wind measurement **location**, and the rows reflect a single time period; wind was reported as daily average wind speed in knots (1 knot = 0.5418 m/s).

Finally, panel data are shown in **long form**, where the full spatio-temporal information is held in a single column, and other columns denote location and time. In the `Produc` data set (Baltagi, 2001), a panel of 48 observations from 1970 to 1986, the first five records are

```
> data("Produc", package = "plm")
> Produc[1:5, ]
```

	state	year	pcap	hwy	water	util	pc	gsp	emp	unemp
1	ALABAMA	1970	15032.67	7325.80	1655.68	6051.20	35793.80	28418	1010.5	4.7
2	ALABAMA	1971	15501.94	7525.94	1721.02	6254.98	37299.91	29375	1021.9	5.2
3	ALABAMA	1972	15972.41	7765.42	1764.75	6442.23	38670.30	31303	1072.3	4.7
4	ALABAMA	1973	16406.26	7907.66	1742.41	6756.19	40084.01	33430	1135.5	3.9
5	ALABAMA	1974	16762.67	8025.52	1734.85	7002.29	42057.31	33749	1169.8	5.5

where the first two columns denote space and time (a default assumption in package `plm`), and e.g. `pcap` reflects private capital stock.

None of these examples documents has strongly *referenced* spatial or temporal information: it is from the data alone not clear whether the number 1970 refers to a year, or ALABAMA to a state, and where this is. Section ?? shows for each of these three cases how the data can be converted into classes with strongly referenced space and time information.

## 3 Space-time layouts

In the following we will use spatial location to denote a particular point, (set of) line(s), (set of) polygon(s), or pixel, for which one or more measurements are registered at particular moments in time.

Three layouts of space-time data have been implemented, along with convenience methods and coercion methods to get from one to the other. These will be introduced next.

### 3.1 Full space-time grid

A full space-time grid<sup>2</sup> of observations for spatial location (points, lines, polygons, grid cells)  $s_i, i = 1, \dots, n$  and observation time  $t_j, j = 1, \dots, m$  is obtained when the full set of  $n \times m$  set of observations  $z_k$  is stored, with  $k = 1, \dots, nm$ . We choose to cycle spatial locations first, so observation  $k$  corresponds to location

<sup>2</sup>note that neither locations nor time points need to be laid out in a regular sequence

$s_i$ ,  $i = ((k - 1) \% n) + 1$  and with time moment  $t_j$ ,  $j = ((k - 1)/n) + 1$ , with / integer division and % integer division remainder (modulo). The  $t_j$  need to be in time order, as `xts` objects are used to store them.

In this data class (figure 1), for each location, the same temporal sequence of data is sampled. Alternatively one could say that for each moment in time, the same set of spatial entities is sampled. Unsampling combinations of (space, time) are stored in this class, but are assigned a missing value `NA`.

### 3.2 Sparse space-time grid

A sparse grid has the same general layout, with measurements laid out on a space time grid (figure 2), but instead of storing the full grid, only non-missing valued observations  $z_k$  are stored. For each  $k$ , an index  $[i, j]$  is stored that refers which spatial location  $i$  and time point  $j$  the value belongs to.

### 3.3 Irregular space-time data.frame

Space-time irregular `data.frames` (STIDF, figure 3) are those where time and space points of measured values can have no organization: for each measured value the spatial location and time point is stored, as in the long format. This is equivalent to the most sparse grid where the index for observation  $k$  is  $[k, k]$ , and hence can be dropped. For these objects,  $n = m$  equals the number of records. Locations and time points need not be unique, but will be replicated in case they are not.

## 4 Spatio-temporal full grid data.frames (STFDF)

For objects of class `STFDF`, time representation can be regular or irregular, as it is of class `xts` in package `xts`. Spatial locations need to be of a class deriving from `Spatial` in package `sp`.

### 4.1 Class definition

```
> library(spacetime)
> showClass("ST")
```

```
Class "ST" [package "spacetime"]
```

```
Slots:
```

```
Name:      sp      time
Class: Spatial      xts
```

```
Known Subclasses:
```

```
Class "STS", directly
Class "STI", directly
Class "STF", directly
Class "STSDF", by class "STS", distance 2
Class "STIDF", by class "STI", distance 2
```

### STFDF (Space-time full data.frame) layout

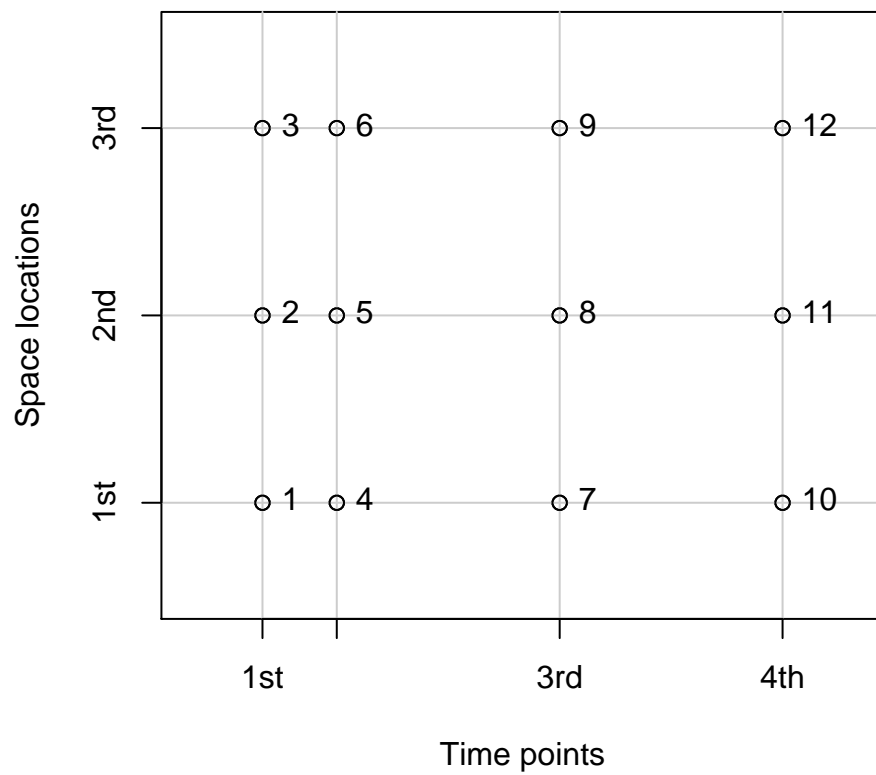


Figure 1: space-time layout of STFDF (STF: ST-Full) objects: all space-time combinations are stored; numbers refer to the ordering of rows in the `data.frame` with measured values: time is kept ordered, space cycles first

## STSDF (Space-time sparse data.frame) layout

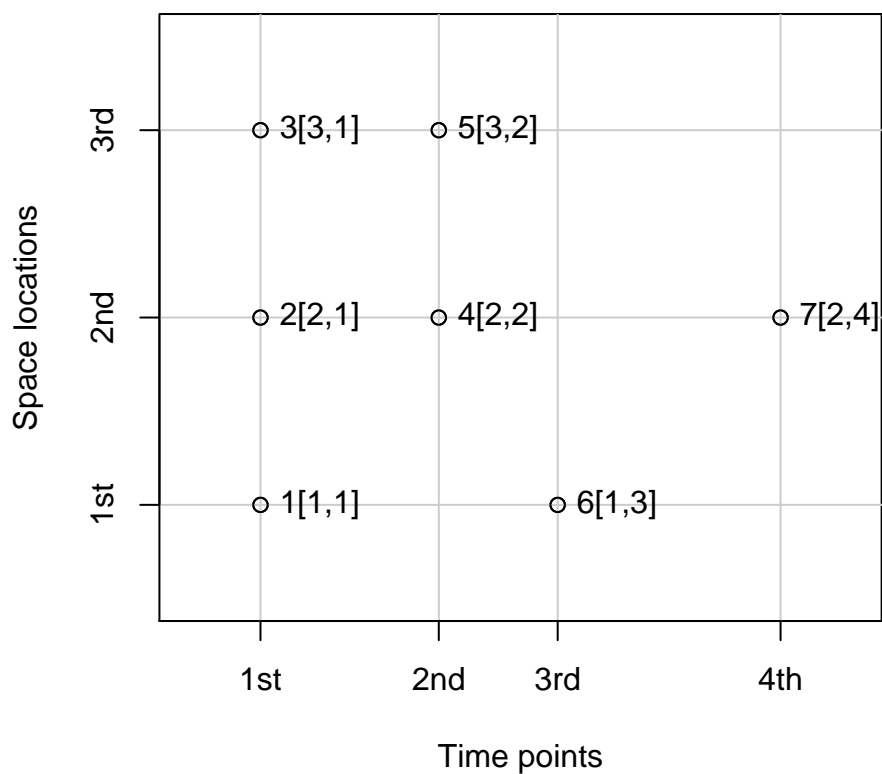


Figure 2: space-time layout of STSDF (STS: ST-Sparse) objects: only the non-missing part of the space-time combinations on a lattice are stored; numbers refer to the ordering of rows in the `data.frame`; an index is kept where `[3,4]` refers to the third item in the list of spatial locations and fourth item in the list of temporal points.

### STIDF (Space–time irregular data.frame) layout

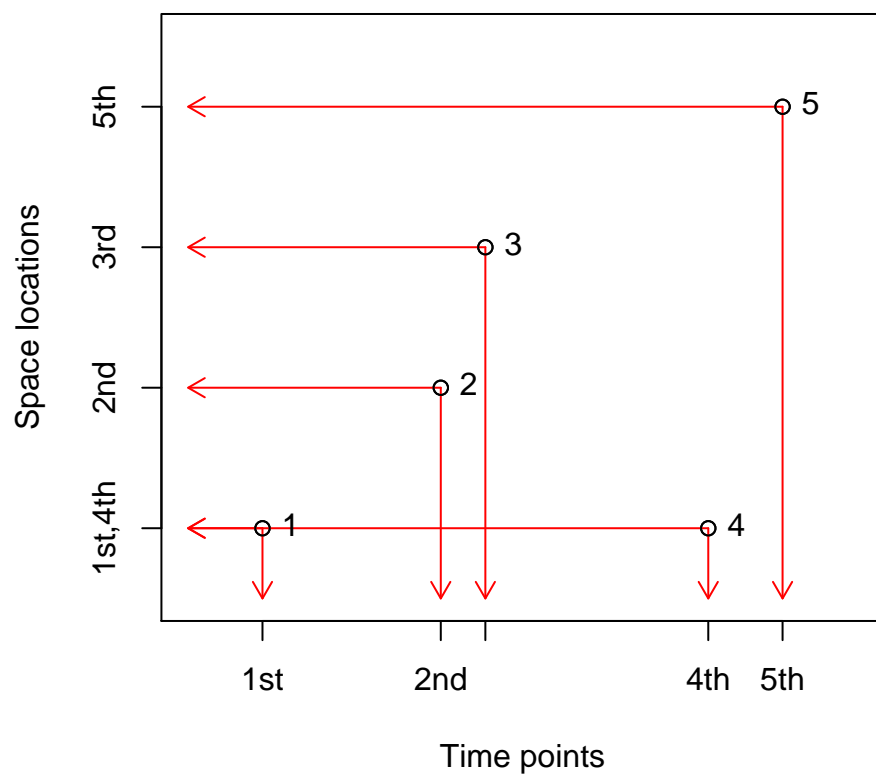


Figure 3: space-time layout of STIDF (STI: ST-Irregular) objects: each observation has its spatial location and time stamp stored; in this example, spatial location 1 is stored twice—observations 1 and 4 having the same location is not registered.



```

Class "STFDF", by class "STF", distance 2
Class "STIDFtraj", by class "STIDF", distance 3

> showClass("STFDF")

Class "STFDF" [package "spacetime"]

Slots:

Name:      data      sp      time
Class: data.frame   Spatial    xts

Extends:
Class "STF", directly
Class "ST", by class "STF", distance 2

> sp = cbind(x = c(0,0,1), y = c(0,1,1))
> row.names(sp) = paste("point", 1:nrow(sp), sep="")
> sp = SpatialPoints(sp)
> time = xts(1:4, as.POSIXct("2010-08-05", tz = "GMT")+3600*(10:13))
> m = c(10,20,30) # means for each of the 3 point locations
> mydata = rnorm(length(sp)*length(time),mean=rep(m, 4))
> IDs = paste("ID",1:length(mydata), sep = "_")
> mydata = data.frame(values = signif(mydata,3), ID=IDs)
> stfdf = STFDF(sp, time, mydata)
> str(stfdf)

Formal class 'STFDF' [package "spacetime"] with 3 slots
 ..@ data:'data.frame':      12 obs. of  2 variables:
 .. ..$ values: num [1:12] 9.73 18.8 31.3 10.4 19 30.5 10 21.3 30.1 9.93 ...
 .. ..$ ID      : Factor w/ 12 levels "ID_1","ID_10",...: 1 5 6 7 8 9 10 11 12 2 ...
 ..@ sp :Formal class 'SpatialPoints' [package "sp"] with 3 slots
 .. ..@ coords      : num [1:3, 1:2] 0 0 1 0 1 1
 .. .. ..- attr(*, "dimnames")=List of 2
 .. .. .. ..$ : chr [1:3] "point1" "point2" "point3"
 .. .. .. ..$ : chr [1:2] "x" "y"
 .. .. ..@ bbox      : num [1:2, 1:2] 0 0 1 1
 .. .. ..- attr(*, "dimnames")=List of 2
 .. .. .. ..$ : chr [1:2] "x" "y"
 .. .. .. ..$ : chr [1:2] "min" "max"
 .. .. ..@ proj4string:Formal class 'CRS' [package "sp"] with 1 slots
 .. .. .. ..@ projargs: chr NA
 ..@ time:An 'xts' object from 2010-08-05 10:00:00 to 2010-08-05 13:00:00 containing:
 Data: int [1:4, 1] 1 2 3 4
 Indexed by objects of class: [POSIXct,POSIXt] TZ: GMT
 xts Attributes:
 NULL

```

## 4.2 Coercion to data.frame

The following coercion function creates a `data.frame` using both the S3 (to set `row.names`) and S4 “`as()`” method. It gives data in the long format, meaning

that time and space are replicated appropriately:

```
> as.data.frame(stfdf, row.names = IDs)
```

	X1	X2	sp.ID	time	values	ID
ID_1	0	0	point1	2010-08-05 10:00:00	9.73	ID_1
ID_2	0	1	point2	2010-08-05 10:00:00	18.80	ID_2
ID_3	1	1	point3	2010-08-05 10:00:00	31.30	ID_3
ID_4	0	0	point1	2010-08-05 11:00:00	10.40	ID_4
ID_5	0	1	point2	2010-08-05 11:00:00	19.00	ID_5
ID_6	1	1	point3	2010-08-05 11:00:00	30.50	ID_6
ID_7	0	0	point1	2010-08-05 12:00:00	10.00	ID_7
ID_8	0	1	point2	2010-08-05 12:00:00	21.30	ID_8
ID_9	1	1	point3	2010-08-05 12:00:00	30.10	ID_9
ID_10	0	0	point1	2010-08-05 13:00:00	9.93	ID_10
ID_11	0	1	point2	2010-08-05 13:00:00	18.40	ID_11
ID_12	1	1	point3	2010-08-05 13:00:00	29.40	ID_12

```
> as(stfdf, "data.frame")[1:4, ]
```

	X1	X2	sp.ID	time	values	ID
1	0	0	point1	2010-08-05 10:00:00	9.73	ID_1
2	0	1	point2	2010-08-05 10:00:00	18.80	ID_2
3	1	1	point3	2010-08-05 10:00:00	31.30	ID_3
4	0	0	point1	2010-08-05 11:00:00	10.40	ID_4

Note that `sp.ID` denotes the ID of the spatial location; coordinates are shown for point, pixel or grid cell centre locations; in case locations refer to lines or polygons, the line's start coordinate and coordinate centre of weight are given, respectively, as the coordinate values in this representation.

For a single attribute, we can obtain a `data.frame` object if we properly unstack the column, giving the data in both its wide formats when in addition we apply transpose `t()`:

```
> unstack(stfdf)
```

	point1	point2	point3
2010-08-05 10:00:00	9.73	18.8	31.3
2010-08-05 11:00:00	10.40	19.0	30.5
2010-08-05 12:00:00	10.00	21.3	30.1
2010-08-05 13:00:00	9.93	18.4	29.4

```
> t(unstack(stfdf))
```

	2010-08-05 10:00:00	2010-08-05 11:00:00	2010-08-05 12:00:00
point1	9.73	10.4	10.0
point2	18.80	19.0	21.3
point3	31.30	30.5	30.1
	2010-08-05 13:00:00		
point1	9.93		
point2	18.40		
point3	29.40		

```
> unstack(stfdf, which = 2)
```

```

              point1 point2 point3
2010-08-05 10:00:00 ID_1   ID_2   ID_3
2010-08-05 11:00:00 ID_4   ID_5   ID_6
2010-08-05 12:00:00 ID_7   ID_8   ID_9
2010-08-05 13:00:00 ID_10  ID_11  ID_12

```

### 4.3 Coercion to xts

We can coerce an object of class STFDF to an xts if we select a single numeric attribute:

```
> as(stfdf[, , "values"], "xts")
```

```

              point1 point2 point3
2010-08-05 10:00:00  9.73  18.8  31.3
2010-08-05 11:00:00 10.40  19.0  30.5
2010-08-05 12:00:00 10.00  21.3  30.1
2010-08-05 13:00:00  9.93  18.4  29.4

```

### 4.4 Attribute retrieval and replacement: [[ and \$

We can define the [[ and \$ retrieval and replacement methods for all classes deriving from ST at once. Here are some examples:

```
> stfdf[[1]]
```

```
[1]  9.73 18.80 31.30 10.40 19.00 30.50 10.00 21.30 30.10  9.93 18.40 29.40
```

```
> stfdf[["values"]]
```

```
[1]  9.73 18.80 31.30 10.40 19.00 30.50 10.00 21.30 30.10  9.93 18.40 29.40
```

```
> stfdf[["newVal"]] = rnorm(12)
```

```
> stfdf$ID
```

```

[1] ID_1 ID_2 ID_3 ID_4 ID_5 ID_6 ID_7 ID_8 ID_9 ID_10 ID_11 ID_12
Levels: ID_1 ID_10 ID_11 ID_12 ID_2 ID_3 ID_4 ID_5 ID_6 ID_7 ID_8 ID_9

```

```
> stfdf$ID = paste("OldIDs", 1:12, sep = "")
```

```
> stfdf$NewID = paste("NewIDs", 12:1, sep = "")
```

```
> stfdf
```

An object of class "STFDF"

Slot "data":

	values	ID	newVal	NewID
1	9.73	OldIDs1	1.23913034	NewIDs12
2	18.80	OldIDs2	-0.60679542	NewIDs11
3	31.30	OldIDs3	0.77112795	NewIDs10
4	10.40	OldIDs4	0.49482942	NewIDs9
5	19.00	OldIDs5	1.10425132	NewIDs8
6	30.50	OldIDs6	1.46351603	NewIDs7

```

7  10.00  OldIDs7  1.48885888  NewIDs6
8  21.30  OldIDs8 -0.02497358  NewIDs5
9  30.10  OldIDs9  1.60574444  NewIDs4
10 9.93  OldIDs10  0.17734963  NewIDs3
11 18.40  OldIDs11 1.96270907  NewIDs2
12 29.40  OldIDs12 -0.48562344  NewIDs1

Slot "sp":
SpatialPoints:
      x y
point1 0 0
point2 0 1
point3 1 1
Coordinate Reference System (CRS) arguments: NA

Slot "time":
      [,1]
2010-08-05 10:00:00    1
2010-08-05 11:00:00    2
2010-08-05 12:00:00    3
2010-08-05 13:00:00    4

```

## 4.5 Space and time selection with [

The idea behind the `[` method for classes in `sp` was that objects would behave as much as possible similar to a matrix or `data.frame` – this is one of the stronger intuitive areas of R syntax. A construct like `a[i,j]` selects row(s) `i` and column(s) `j`. In `sp`, rows were taken as the spatial entities (points, lines, polygons, pixels) and rows as the attributes. This convention was broken for objects of class `SpatialGridDataFrame`, where `a[i,j,k]` would select the  $k$ -th attribute of the spatial grid selection with spatial grid row(s) `i` and column(s) `j`.

For spatio-temporal data, `a[i,j,k]` selects spatial entity/entities `i`, temporal entity/entities `j`, and attribute(s) `k`:

example:

```

> stfdf[,1] # SpatialPointsDataFrame:

  coordinates values      ID      newVal      NewID
1      (0, 0)   9.73 OldIDs1  1.2391303 NewIDs12
2      (0, 1)  18.80 OldIDs2 -0.6067954 NewIDs11
3      (1, 1)  31.30 OldIDs3  0.7711279 NewIDs10

> stfdf[, ,1]

```

An object of class "STFDF"

Slot "data":

```

  values
1    9.73
2   18.80
3   31.30

```

```

4    10.40
5    19.00
6    30.50
7    10.00
8    21.30
9    30.10
10   9.93
11   18.40
12   29.40

Slot "sp":
SpatialPoints:
      x y
point1 0 0
point2 0 1
point3 1 1
Coordinate Reference System (CRS) arguments: NA

Slot "time":
      [,1]
2010-08-05 10:00:00    1
2010-08-05 11:00:00    2
2010-08-05 12:00:00    3
2010-08-05 13:00:00    4

> stfdf[1,,1] # xts

      values
2010-08-05 10:00:00    9.73
2010-08-05 11:00:00   10.40
2010-08-05 12:00:00   10.00
2010-08-05 13:00:00    9.93

> stfdf[,,"ID"]

An object of class "STFDF"
Slot "data":
      ID
1    OldIDs1
2    OldIDs2
3    OldIDs3
4    OldIDs4
5    OldIDs5
6    OldIDs6
7    OldIDs7
8    OldIDs8
9    OldIDs9
10   OldIDs10
11   OldIDs11
12   OldIDs12

```

```

Slot "sp":
SpatialPoints:
      x y
point1 0 0
point2 0 1
point3 1 1
Coordinate Reference System (CRS) arguments: NA

Slot "time":
      [,1]
2010-08-05 10:00:00    1
2010-08-05 11:00:00    2
2010-08-05 12:00:00    3
2010-08-05 13:00:00    4

> stfdf[1,,"values", drop=FALSE] # stays STFDF

An object of class "STFDF"
Slot "data":
      values
1      9.73
4     10.40
7     10.00
10     9.93

Slot "sp":
SpatialPoints:
      x y
point1 0 0
Coordinate Reference System (CRS) arguments: NA

Slot "time":
      [,1]
2010-08-05 10:00:00    1
2010-08-05 11:00:00    2
2010-08-05 12:00:00    3
2010-08-05 13:00:00    4

> stfdf[,1, drop=FALSE] #stays STFDF

An object of class "STFDF"
Slot "data":
      values      ID      newVal      NewID
1      9.73 OldIDs1  1.2391303 NewIDs12
2     18.80 OldIDs2 -0.6067954 NewIDs11
3     31.30 OldIDs3  0.7711279 NewIDs10

Slot "sp":
SpatialPoints:
      x y
point1 0 0

```

```

point2 0 1
point3 1 1
Coordinate Reference System (CRS) arguments: NA

Slot "time":
              [,1]
2010-08-05 10:00:00      1

```

Clearly, unless `drop=FALSE`, selecting a single time or single location object results in an object that is no longer spatio-temporal; see also section 7.

## 5 Space-time sparse data.frames (STSDf)

Space-time sparse `data.frames` have a layout over a grid, meaning that particular times and locations are typically present more than once, but only the data for the time/location combinations are stored. An index keeps the link between the measured values in the data entries (rows), and the locations and times.

### 5.1 Class definition

```

> showClass("STSDf")

Class "STSDf" [package "spacetime"]

Slots:

Name:      data      index      sp      time
Class: data.frame  matrix  Spatial    xts

Extends:
Class "STS", directly
Class "ST", by class "STS", distance 2

```

In this class, `index` is an  $n \times 2$  matrix. If in this index row  $i$  has entry  $[j, k]$ , it means that the  $i$ -th row in the `data` slot corresponds to location  $j$  and time  $k$ .

## 6 Spatio-temporal irregular data.frames (STIDf)

Space-time irregular `data.frames` store for each data record the location and time. No index is kept. Location and time need not be organized. Data are stored such that time is ordered (as it is an `xts` object).

### 6.1 Class definition

```

> showClass("STIDf")

Class "STIDf" [package "spacetime"]

Slots:

```

```

Name:      data      sp      time
Class: data.frame    Spatial    xts

```

```

Extends:
Class "STI", directly
Class "ST", by class "STI", distance 2

```

```

Known Subclasses: "STIDFtraj"

```

```

> sp = expand.grid(x = 1:3, y = 1:3)
> row.names(sp) = paste("point", 1:nrow(sp), sep="")
> sp = SpatialPoints(sp)
> time = xts(1:9, as.POSIXct("2010-08-05", tz = "GMT")+3600*(11:19))
> m = 1:9 * 10 # means for each of the 9 point locations
> mydata = rnorm(length(sp), mean=m)
> IDs = paste("ID",1:length(mydata))
> mydata = data.frame(values = signif(mydata,3),ID=IDs)
> stidf = STIDF(sp, time, mydata)
> stidf

```

```

An object of class "STIDF"

```

```

Slot "data":

```

	values	ID
1	8.88	ID 1
2	19.30	ID 2
3	30.00	ID 3
4	39.70	ID 4
5	49.40	ID 5
6	59.90	ID 6
7	69.60	ID 7
8	78.30	ID 8
9	89.40	ID 9

```

Slot "sp":

```

```

SpatialPoints:

```

	x	y
[1,]	1	1
[2,]	2	1
[3,]	3	1
[4,]	1	2
[5,]	2	2
[6,]	3	2
[7,]	1	3
[8,]	2	3
[9,]	3	3

```

Coordinate Reference System (CRS) arguments: NA

```

```

Slot "time":

```

	[,1]
2010-08-05 11:00:00	1



```

2010-08-05 12:00:00    2
2010-08-05 13:00:00    3
2010-08-05 14:00:00    4
2010-08-05 15:00:00    5
2010-08-05 16:00:00    6
2010-08-05 17:00:00    7
2010-08-05 18:00:00    8
2010-08-05 19:00:00    9

```

## 6.2 Methods

Selection takes place with the `[]` method:

```
> stidf[1:2, ]
```

An object of class "STIDF"

Slot "data":

	values	ID
1	8.88	ID 1
2	19.30	ID 2

Slot "sp":

SpatialPoints:

	x	y
[1,]	1	1
[2,]	2	1

Coordinate Reference System (CRS) arguments: NA

Slot "time":

		[,1]
2010-08-05 11:00:00		1
2010-08-05 12:00:00		2

## 7 Further methods: snapshot, history, coercion

### 7.1 *Snap* and *Hist*

A time snapshot (Galton, 2004) to a particular moment in time can be obtained through selecting a particular time moment:

```
> stfdf[, time[3]]
```

	coordinates	values	ID	newVal	NewID
7	(0, 0)	10.0	OldIDs7	1.48885888	NewIDs6
8	(0, 1)	21.3	OldIDs8	-0.02497358	NewIDs5
9	(1, 1)	30.1	OldIDs9	1.60574444	NewIDs4

by default, a simplified object of the underlying `Spatial` class for this particular time is obtained (`drop=TRUE`); if we specify `drop = FALSE`, the class will not be changed:

```
> class(stfdf[, time[3]])
```

```

[1] "SpatialPointsDataFrame"
attr(,"package")
[1] "sp"

> class(stfdf[, time[3], drop = FALSE])

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

```

A time series (or *history*, according to Galton, 2004) for a single particular location is obtained by selecting this location, e.g.

```

> stfdf[1, , "values"]

              values
2010-08-05 10:00:00  9.73
2010-08-05 11:00:00 10.40
2010-08-05 12:00:00 10.00
2010-08-05 13:00:00  9.93

```

Again, the class is not reduced to the simpler when `drop = FALSE` is specified:

```

> class(stfdf[1, ])

[1] "xts" "zoo"

> class(stfdf[1, drop = FALSE])

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

```

For objects of class `STIDF`, `drop = TRUE` results in a `Spatial` object when a single time value is selected.

## 7.2 Coercion between `STxxx` classes

Coercion from full to sparse and/or irregular space-time `data.frames`, we can use as:

```

> class(stfdf)

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

> class(as(stfdf, "STSDF"))

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

> class(as(as(stfdf, "STSDF"), "STIDF"))

```

```

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

> class(as(stfdf, "STIDF"))

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

```

On our way back, the reverse coercion takes place:

```

> x = as(stfdf, "STIDF")
> class(as(x, "STSDF"))

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

> class(as(as(x, "STSDF"), "STFDF"))

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

> class(as(x, "STFDF"))

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

> xx = as(x, "STFDF")
> identical(stfdf, xx)

[1] TRUE

```

## 8 Spatial footprint or support, time intervals

### 8.1 Time periods

Time series typically store for each record a time stamp, not a time interval. The implicit assumption of time seems to be (i) the time stamp is a moment, (ii) this indicates either the real moment of measurement / registration, or the start of the interval over which something is aggregated (summed, averaged, maximized). For financial "Open, high, low, close" data, the "Open" and "Close" refer to the values at the moments the stock exchange opens and closes, where "high" and "low" aggregated (minimum, maximum taken over the time interval between opening and closing times).

According to [ISO 8601:2004](#), a time stamp like "2010-05" refers to *the full* month of May, 2010, and so reflects a time period rather than a moment. As a selection criterion, `xts` will include everything inside the following interval:

```

> .parseISO8601("2010-05")

```

```
$first.time
[1] "2010-05-01 CEST"
```

```
$last.time
[1] "2010-05-31 23:59:59 CEST"
```

and it seems that this syntax lets one define, unambiguously, yearly, monthly, daily, hourly or minute intervals, but not 10- or 30-minute intervals; for ten minutes, the full specification is needed:

```
> .parseISO8601("2010-05-01T13:30/2010-05-01T13:39")
```

```
$first.time
[1] "2010-05-01 13:30:00 CEST"
```

```
$last.time
[1] "2010-05-01 13:39:59 CEST"
```

## 8.2 Spatial support

All examples above work with spatial points, i.e. data having a point support. The assumption of data having points support is implicit. For polygons, the assumption will be that values reflect aggregates over the polygon. For gridded data, it is ambiguous whether the value at the grid cell centre is meant (e.g. for DEM data) or an aggregate over the grid cell (typical for remote sensing imagery).

## 9 Worked examples

This section shows how existing data in various formats can be converted into ST classes, and how they can be analysed and/or visualised.

### 9.1 North Carolina SIDS

As an example, the North Carolina Sudden Infant Death Syndrome (sids) data in package `maptools` will be used; they are sparse in time (aggregated to 2 periods of unequal length, according to the documentation in package `spdep`), but have polygons in space. Figure 4 shows the plot generated.

```
> library(maptools)
```

```
Note: polygon geometry computations in maptools
      depend on the package gpclib, which has a
      restricted licence. It is disabled by default;
      to enable gpclib, type gpclibPermit()
```

```
Checking rgeos availability as gpclib substitute:
FALSE
```

```
> fname = system.file("shapes/sids.shp", package = "maptools")[1]
> nc = readShapePoly(fname, proj4string = CRS("+proj=longlat +datum=NAD27"))
```

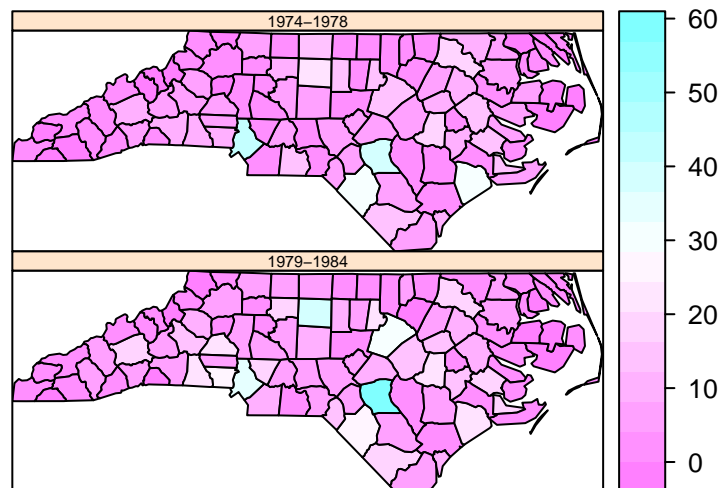


Figure 4: North Carolina sudden infant death syndrome (sids) data

```
> data = data.frame(BIR = c(nc$BIR74, nc$BIR79), NWBIR = c(nc$NWBIR74,
+ nc$NWBIR79), SID = c(nc$SID74, nc$SID79))
> time = xts(1:2, as.POSIXct(strptime(c("1974-01-01", "1979-01-01"),
+ "%Y-%m-%d"), tz = "GMT"))
> nct = STFDF(sp = as(nc, "SpatialPolygons"), time = time, data = data)
> stplot(nct[, , "SID"], c("1974-1978", "1979-1984"))
```

## 9.2 Panel data

The panel data discussed in section 2 are imported as a full ST data.frame (STFDF), and linked to the proper state polygons of maps. Both `Produc` and the states in package `maps` order states alphabetically; the only thing to watch out for is that the former does not include District of Columbia, but the latter does (record 8):

```
> library(maps)
> states.m = map('state', plot=FALSE, fill=TRUE)
> IDs <- sapply(strsplit(states.m$names, ":"), function(x) x[1])
> library(maptools)
> states = map2SpatialPolygons(states.m, IDs=IDs)
> library(plm)
> data(Produc)
> yrs = 1970:1986
> time = xts(1:17, as.POSIXct(paste(yrs, "-01-01", sep=""), tz = "GMT"))
> # deselect District of Columbia, polygon 8, which is not present in Produc:
> Produc.st = STFDF(states[-8], time, Produc[(order(Produc[2], Produc[1])),])
> stplot(Produc.st[, , "unemp"], yrs)
```

The plot itself is left out for reasons of disk space. Time and state were not removed from the data table on construction; printing these data as a `data.frame` confirms that time and state were matched correctly. The `plm` routines can be used on the data, back transformed to a `data.frame`, when `index` is specified (the first two columns from the back-transformed data no longer contain state and year):

```
> zz <- plm(log(gsp) ~ log(pcap) + log(pc) + log(emp) + unemp,
+ data = as.data.frame(Produc.st), index = c("state", "year"))
> summary(zz)
```

Oneway (individual) effect Within Model

Call:

```
plm(formula = log(gsp) ~ log(pcap) + log(pc) + log(emp) + unemp,
     data = as.data.frame(Produc.st), index = c("state", "year"))
```

Balanced Panel: n=48, T=17, N=816

Residuals :

Min.	1st Qu.	Median	3rd Qu.	Max.
-0.12000	-0.02370	-0.00204	0.01810	0.17500

Coefficients :

	Estimate	Std. Error	t-value	Pr(> t )
log(pcap)	-0.02614965	0.02900158	-0.9017	0.3675
log(pc)	0.29200693	0.02511967	11.6246	< 2.2e-16 ***
log(emp)	0.76815947	0.03009174	25.5273	< 2.2e-16 ***
unemp	-0.00529774	0.00098873	-5.3582	1.114e-07 ***

---

Signif. codes: 0 '\*\*\*' 0.001 '\*\*' 0.01 '\*' 0.05 '.' 0.1 ' ' 1

Total Sum of Squares: 18.941

Residual Sum of Squares: 1.1112

R-Squared : 0.94134

Adj. R-Squared : 0.88135

F-statistic: 3064.81 on 4 and 764 DF, p-value: < 2.22e-16

### 9.3 Interpolating Irish wind

This worked example is a modified version of the analysis presented in `demo(wind)` of package `gstat`. This demo is rather lengthy and reproduces much of the original analysis in Haslett and Raftery (1989). Here, we will reduce the intermediate plots and focus on the use of spatio-temporal classes.

First, we will load the wind data from package `gstat`. It has two tables, station locations in a `data.frame`, called `wind.loc`, and daily wind speed in `data.frame` `wind`. We now convert character representation (such as 51d56'N) to proper numerical coordinates, and convert the station locations to a `SpatialPointsDataFrame` object. A plot of these data is shown in figure 6.

```
> library(gstat)
> data(wind)
```

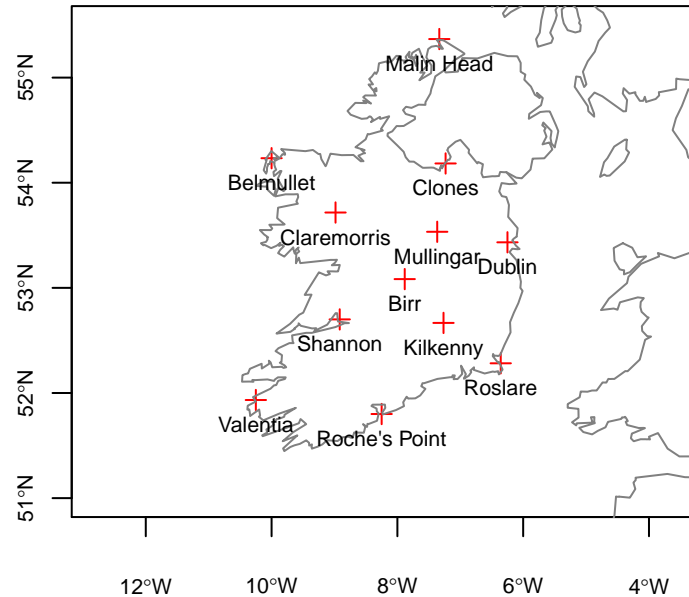


Figure 5: Station locations for Irish wind data

```
> wind.loc$y = as.numeric(char2dms(as.character(wind.loc[["Latitude"]]))))
> wind.loc$x = as.numeric(char2dms(as.character(wind.loc[["Longitude"]]))))
> coordinates(wind.loc) = ~x + y
> proj4string(wind.loc) = "+proj=longlat +datum=WGS84"
```

The first thing to do with the wind speed values is to reshape these data. Unlike the North Carolina SIDS data of section 9.1, for this data space is sparse and time is rich, and so the data in `data.frame wind` come in space wide form with stations time series in columns:

```
> wind[1:3, ]
```

	year	month	day	RPT	VAL	ROS	KIL	SHA	BIR	DUB	CLA	MUL	CLO
1	61	1	1	15.04	14.96	13.17	9.29	13.96	9.87	13.67	10.25	10.83	12.58
2	61	1	2	14.71	16.88	10.83	6.50	12.62	7.67	11.50	10.04	9.79	9.67
3	61	1	3	18.50	16.88	12.33	10.13	11.17	6.17	11.25	8.04	8.50	7.67
				BEL	MAL								
1				18.50	15.04								
2				17.54	13.83								
3				12.75	12.71								

We will recode the time columns to an appropriate time data structure, and subtract a smooth time trend of daily means:

```
> wind$time = ISOdate(wind$year + 1900, wind$month, wind$day)
> wind$jday = as.numeric(format(wind$time, "%j"))
> stations = 4:15
> windsqrt = sqrt(0.5148 * wind[stations])
> Jday = 1:366
> daymeans = apply(sapply(split(windsqrt - mean(windsqrt), wind$jday),
+   mean), 2, mean)
> meanwind = lowess(daymeans ~ Jday, f = 0.1)$y[wind$jday]
> velocities = apply(windsqrt, 2, function(x) {
+   x - meanwind
+ })
```

Next, we will match the wind data to its location, and project the longitude/latitude coordinates and country boundary to the appropriate UTM zone:

```
> # order locations to order of columns in wind;
> # connect station names to location coordinates
> wind.loc = wind.loc[match(names(wind[4:15]), wind.loc$Code),]
> pts = coordinates(wind.loc[match(names(wind[4:15]), wind.loc$Code),])
> rownames(pts) = wind.loc$Station
> pts = SpatialPoints(pts)
> # convert to utm zone 29, to be able to do interpolation in
> # proper Euclidian (projected) space:
> proj4string(pts) = "+proj=longlat +datum=WGS84"
> library(rgdal)
> utm29 = CRS("+proj=utm +zone=29 +datum=WGS84")
> t = xts(1:nrow(wind), wind$time)
> pts = spTransform(pts, utm29)
> # note the t() in:
> w = STFDF(pts, t, data.frame(values = as.vector(t(velocities))))
> library(maptools)
> m = map2SpatialLines(
+   map("worldHires", xlim = c(-11,-5.4), ylim = c(51,55.5), plot=F))
> proj4string(m) = "+proj=longlat +datum=WGS84"
> m = spTransform(m, utm29)
> # setup grid
> grd = SpatialPixels(SpatialPoints(makegrid(m, n = 300)),
+   proj4string = proj4string(m))
> # select april 1961:
> w = w[, "1961-04"]
> # 10 prediction time points, evenly spread over this month:
> n = 10
> tgrd = xts(1:n, seq(min(index(w)), max(index(w)), length=n))
> # separable covariance model, exponential with ranges 750 km and 1.5 day:
> v = list(space = vgm(0.6, "Exp", 750000), time = vgm(1, "Exp", 1.5 * 3600 * 24))
> pred = krigeST(sqrt(values)^1, w, STF(grd, tgrd), v)
> wind.ST = STFDF(grd, tgrd, data.frame(sqrt_speed = pred))
```

the results of which are shown in figure 6, created with `stplot`.



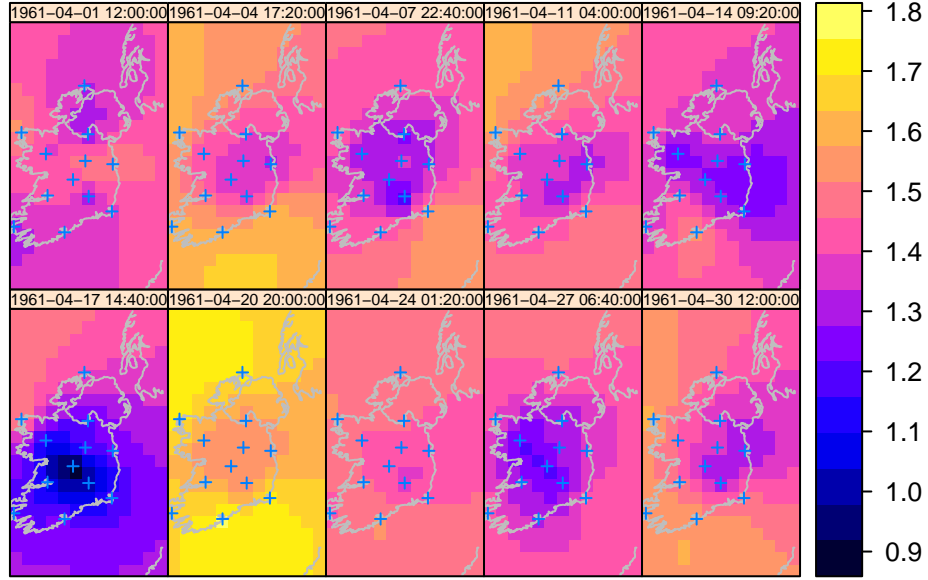


Figure 6: Space-time interpolations of wind (square root transformed, detrended) over Ireland using a separable product covariance model, for 10 time points regularly distributed over the month for which daily data was considered (April, 1961)

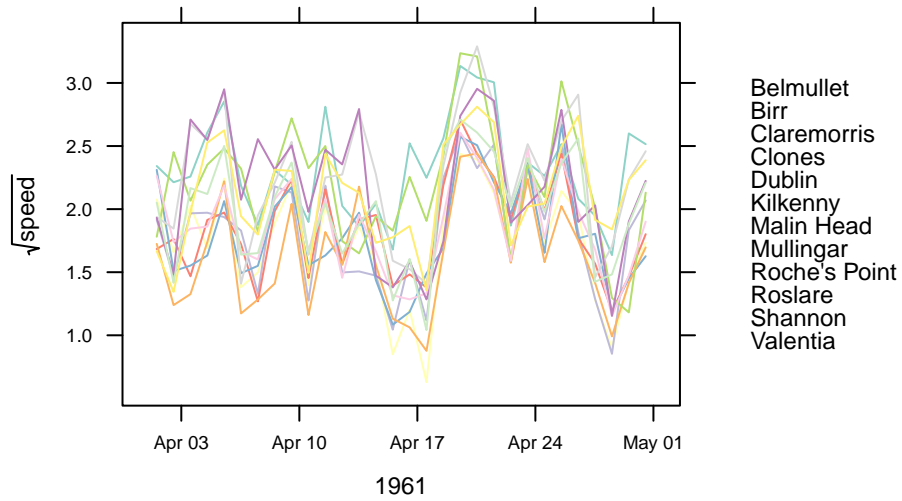


Figure 7: Time series plot of daily wind speed at 12 stations, used for interpolation in figure 6

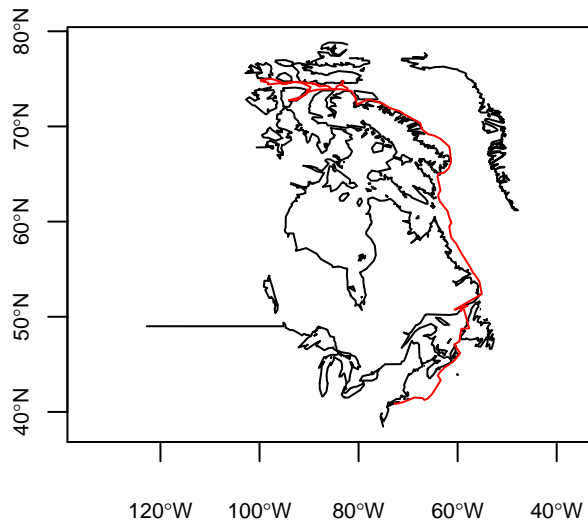
## 9.4 Conversion from and to trip

Objects of class `trip` (Sumner, 2010) extend objects of class `SpatialPointsDataFrame` by indicating in which attribute columns time and trip ID are, in slot `TOR.columns`. To not lose this information (in particular, which column contains the IDs), we will extend class `STIDF` to retain this info.

Currently it does assume that time in a trip object is in order, as xts will order it anyhow:

```
> library(diveMove)
> library(trip)
> locs = readLocs(system.file(file.path("data", "sealLocs.csv"),
+   package = "diveMove"), idCol = 1, dateCol = 2, dtformat = "%Y-%m-%d %H:%M:%S",
+   classCol = 3, lonCol = 4, latCol = 5, sep = ";")
> ringy = subset(locs, id == "ringy" & !is.na(lon) & !is.na(lat))
> coordinates(ringy) = ringy[c("lon", "lat")]
> tr = trip(ringy, c("time", "id"))
> setAs("trip", "STIDFtraj", function(from) {
+   from$burst = from[[from@TOR.columns[2]]]
+   time = from[[from@TOR.columns[1]]]
+   new("STIDFtraj", STIDF(as(from, "SpatialPoints"), time, from@data))
+ })
> x = as(tr, "STIDFtraj")
> m = map2SpatialLines(map("world", xlim = c(-100, -50), ylim = c(40,
+   77), plot = F))
> proj4string(m) = "+proj=longlat +datum=WGS84"
> plot(m, axes = TRUE, cex.axis = 0.7)
> plot(x, add = TRUE, col = "red")
> setAs("STIDFtraj", "trip", function(from) {
+   from$time = index(from@time)
+   trip(SpatialPointsDataFrame(from@sp, from@data), c("time",
+   "burst"))
+ })
> y = as(x, "trip")
> y$burst = NULL
> all.equal(y, tr, check.attributes = FALSE)

[1] TRUE
```



## 9.5 Trajectory data: `ltraj` in `adehabitatLT`

Trajectory objects of class `ltraj` are lists of bursts, sets of sequentially, connected space-time points at which an object is registered. When converting a list to a single `STIDF` object, the ordering is according to time, and the subsequent objects become unconnected. In the coercion back to `ltraj`, based on ID and burst the appropriate bursts are restored. A simple plot is obtained by:

```
> library(adehabitatLT)
> # from: adehabitat/demo/managltraj.r
> # demo(managltraj)
> data(puechabonsp)
> # locations:
> locs = puechabonsp$relocs
> xy = coordinates(locs)
> ### Conversion of the date to the format POSIX
> da = as.character(locs$Date)
> da = as.POSIXct(strptime(as.character(locs$Date), "%y%m%d"), tz = "GMT")
> ## object of class "ltraj"
> ltr = as.ltraj(xy, da, id = locs$Name)
> foo = function(dt) dt > 100*3600*24
> ## The function foo returns TRUE if dt is longer than 100 days
> ## We use it to cut ltr:
> l2 = cutltraj(ltr, "foo(dt)", nextr = TRUE)
> stidfTrj = as(l2, "STIDFtraj")
```

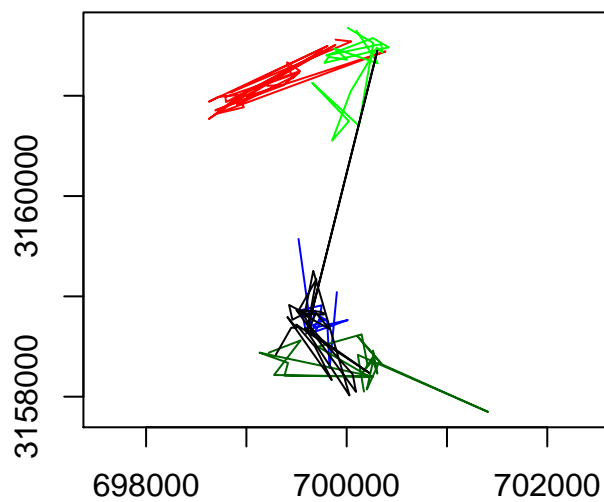
```

> ltr0 = as(stdfTrj, "ltraj")
> all.equal(l2, ltr0, check.attributes = FALSE)

[1] TRUE

> plot(stdfTrj, col = c("red", "green", "blue", "darkgreen", "black"),
+      axes=TRUE)

```



A more complicated plot is shown in figure 8, obtained by the command

```
> stplot(stdfTrj, by = "time*id")
```

the output of which is shown in figure 8.

## 9.6 Country shapes in cshapes

The `cshapes` package contains a GIS dataset of country boundaries (1946-2008), and includes functions for data extraction and the computation of weights matrices. The data set consist of a `SpatialPolygonsDataFrame`, with the following attributes:

```

> library(cshapes)
> cs = cshp()
> names(cs)

```

[1]	"CNTRY_NAME"	"AREA"	"CAPNAME"	"CAPLONG"	"CAPLAT"
[6]	"FEATUREID"	"COWCODE"	"COWSYEAR"	"COWSMONTH"	"COWSDAY"
[11]	"COWEYEAR"	"COWEMONTH"	"COWEDAY"	"GWCODE"	"GWSYEAR"









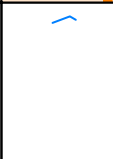















Brock	Brock	Brock	Brock	Brock	Brock
time	time	time	time	time	time
					
Calou	Calou	Calou	Calou	Calou	Calou
time	time	time	time	time	time
					
Chou	Chou	Chou	Chou	Chou	Chou
time	time	time	time	time	time
					
Jean	Jean	Jean	Jean	Jean	Jean
time	time	time	time	time	time
					

Figure 8: trajectories, by id (rows) and time (columns)

```
[16] "GWSMONTH" "GWSDAY" "GWEYEAR" "GWEMONTH" "GWEDAY"
[21] "ISONAME" "ISO1NUM" "ISO1AL2" "ISO1AL3"
```

where two data bases are used, "COW" (correlates of war project, 2008), and "GW" Gleditsch and Ward (1999). The attributes COWSMONTH and COWEMONTH denote the start month and end month, respectively, according to the COW data base.

To select the country boundaries corresponding to a particular date and system, one can use

```
> cshp.2002 <- cshp(date = as.Date("2002-6-30"), useGW = TRUE)
```

In the following fragment, an unordered list of times `t` is passed on to `STIDF`, and this will cause the geometries and attributes to be reordered (in the order of `t`):

```
> t = as.POSIXct(strptime(paste(cs$COWSYEAR, cs$COWSMONTH, cs$COWSDAY,
+   sep = "-"), "%Y-%m-%d"), tz = "GMT")
> st = STIDF(geometry(cs), t, as.data.frame(cs))
> pt = SpatialPoints(cbind(7, 52), CRS(proj4string(cs)))
> as.data.frame(st[pt, , 1:5])
```

	X1	X2	sp.ID	time	CNTRY_NAME	AREA	CAPNAME
1	9.41437	50.57623	188	1955-05-05	Germany Federal Republic	247366.4	Bonn
2	10.38084	51.09070	187	1990-10-03	Germany	356451.5	Berlin
	CAPLONG	CAPLAT					
1	7.1	50.73333					
2	13.4	52.51667					

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