

The GRASS GIS Temporal Framework: Object oriented code design with examples

Sören Gebbert
Thünen Institute of Climate-Smart Agriculture, Germany

Edzer Pebesma
Institute for Geoinformatics, University of Münster, Germany

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1 Introduction

This document describes the object oriented design of the GRASS GIS Temporal Framework that underlies the temporally enabled GRASS GIS, called TGRASS. TGRASS is a full featured field-based temporal GIS, see [Gebbert and Pebesma(2014)]. We provide several code examples to show the capabilities of the GRASS GIS Temporal Framework, ranging from spatio-temporal

analysis of space-time dataset relations and their associated time stamped map layers, over topological relations between map layers, single pixel and feature access to the spatio-temporal intersection between trajectory data.

2 Object oriented code design

Objects that represent GRASS map layers and space-time datasets in the temporal framework are stored in the temporal database. The content of these objects can be serialised, since only serialisable Python types are used to represent the map layer and STDS metadata. The Unified Modelling Language (UML) diagram in figure 1 shows the simplified inheritance structure of the spatio-temporal extent and basic metadata classes of map layers and STDS. The metadata content is stored in an object specific Python dictionary. The class *SQLSerializer* takes care of serialising the dictionary content and creates select, insert, update and delete SQL statements that are used by its subclass *SQLDatabaseInterface* to manage metadata in the temporal database. This class also specifies the unique identifier for map layers and space-time datasets. It is a combination of the map layer name or space-time datasets name and the associated mapset. The identifiers for vector layer may contain an additional reference to an attribute table, since a vector layer may have several different attribute tables. The temporal framework allows time stamps for attribute tables. Hence, a single vector layer with multiple attribute tables may have several time stamps and therefore several entries in the temporal database. The class *DatasetBase* inherits from *SQLDatabaseInterface* and specifies basic information about map layer and space-time datasets. Its subclass *STDSBase* specifies in addition the modification time for space-time datasets. The class *TemporalExtent* inherits from *SQLDatabaseInterface* and specifies the temporal extent for map layers and space-time datasets. It provides methods to compute temporal relationships, temporal intersections, unions and disjoint unions between itself and a second object of type *TemporalExtent*. Absolute and relative time temporal extents are specified in subclasses from which space-time dataset specific classes derive that specify the temporal granularity and the temporal type of the registered maps. The spatial extent of map layers and space-time datasets is represented by the class *SpatialExtent*. It inherits from *SQLDatabaseInterface* and supports two-dimensional¹ and three-dimensional spatial extents². Similar to the temporal extent, the class *SpatialExtent* provides methods for the computation of spatial relationships, spatial intersections, unions and disjoint unions between itself and a second object of type *SpatialExtent* in two and three dimensions.

¹north, south, east, west

²north, south, east, west, top, bottom

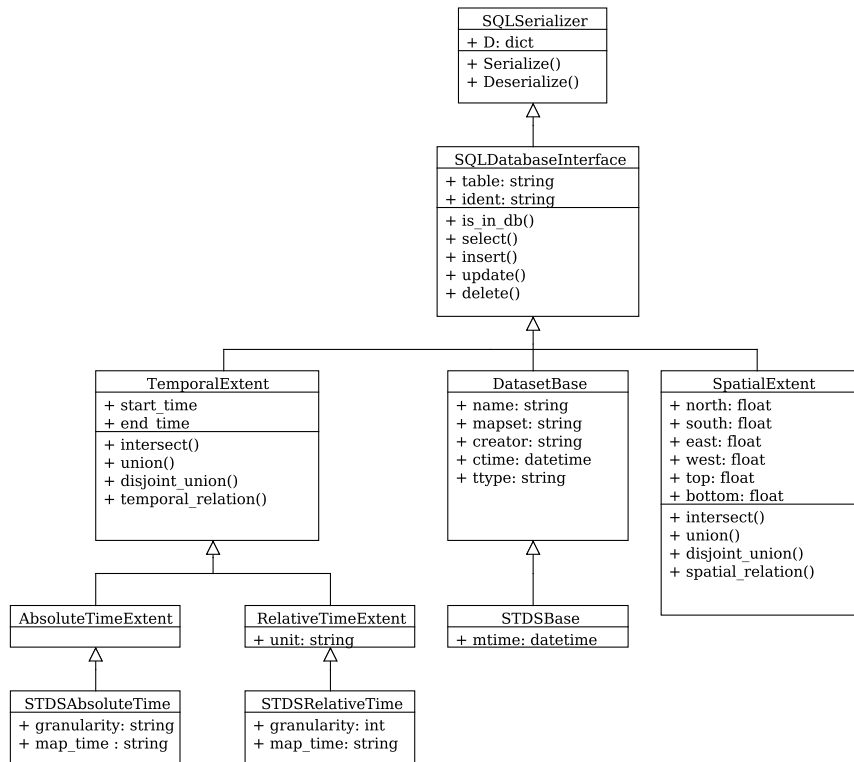


Figure 1: UML diagram showing the simplified functionality and inheritance of the base and spatio-temporal extent classes.

The UML diagram in figure 2 shows the simplified inheritance structure of the map layer type specific metadata classes. The space-time datasets specific classes reflect the fact that raster and 3D raster layers of different spatial resolution can be registered in the same space-time dataset. The following member variables are used to store minimum and maximum resolution values of all registered map layers in a space-time datasets:

- nsres_max specifies the maximum north-south spatial resolution
- nsres_min specifies the minimum north-south spatial resolution
- ewres_max specifies the maximum east-west spatial resolution
- ewres_min specifies the minimum east-west spatial resolution
- tbres_max specifies the maximum top-bottom spatial resolution
- tbres_min specifies the minimum top-bottom spatial resolution

The raster and 3D raster specific metadata classes include the minimum and maximum values of the map layer. The STDS specific metadata classes

aggregate this data into minimum and maximum values of all registered map layer minimums and minimum and maximum values of all registered map layer maximums.

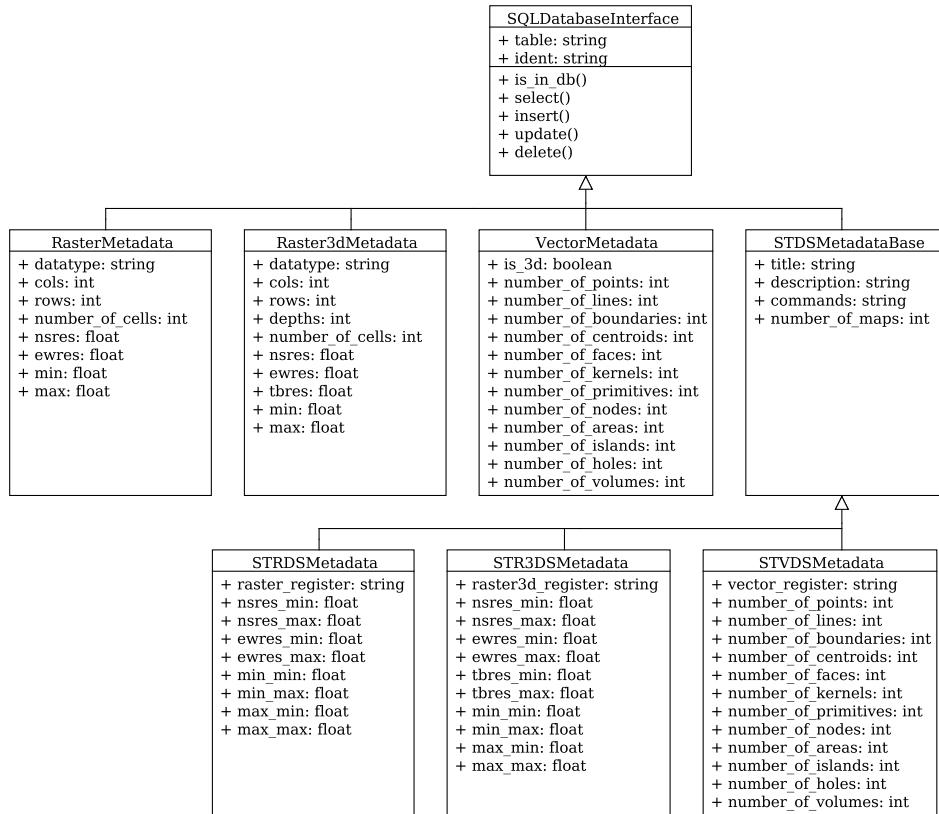


Figure 2: UML diagram showing the simplified functionality and inheritance of map layer and space-time dataset specific metadata classes.

The classes that represent map layers and space-time datasets were implemented based on an abstract view on datasets in the temporal framework, see figure 3.

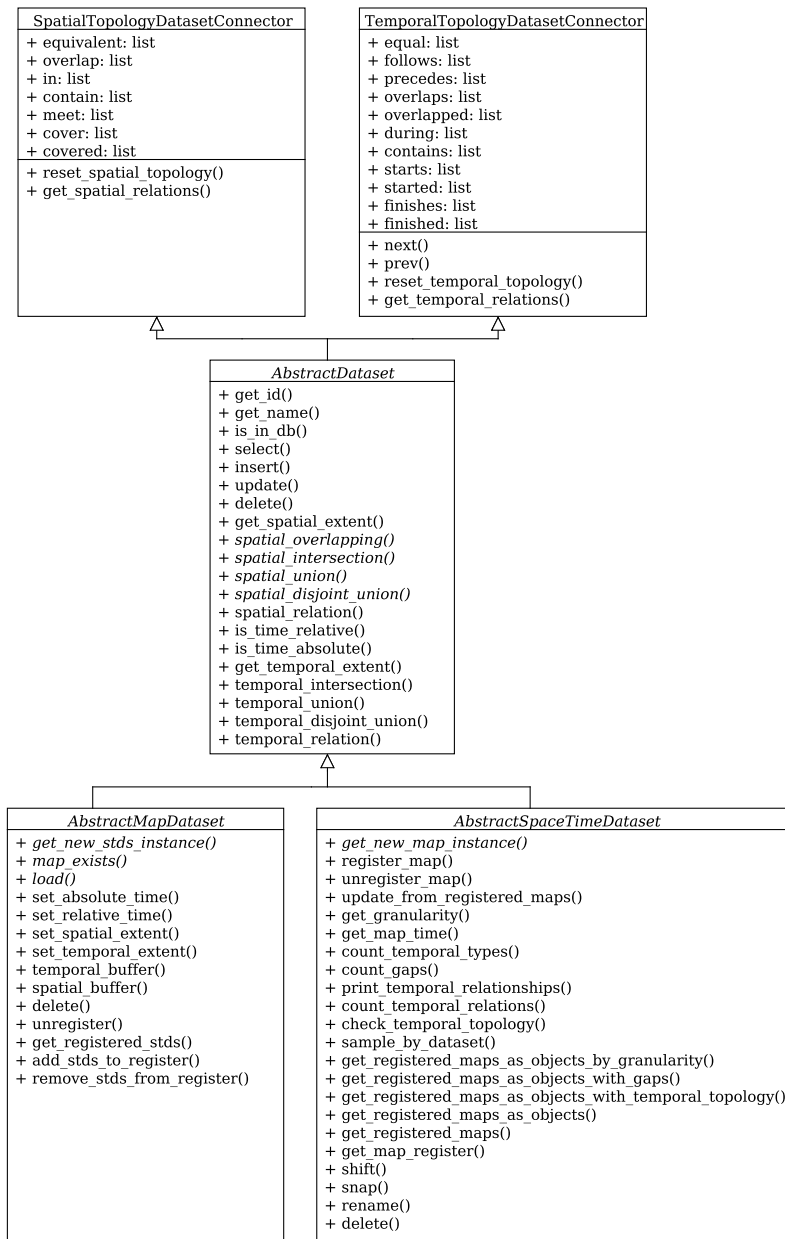


Figure 3: UML diagram showing the simplified functionality and inheritance of abstract map layer and space-time datasets classes.

Map layer and space-time datasets share common properties. They have a spatio-temporal extent and other common metadata. Their object representation can be selected, inserted, updated and deleted from the temporal database. Therefore an abstract base class for map layer and STDS was designed, the *AbstractDataset*. This class implements the common interface

for map layers and space-time datasets. It provides methods for spatial and temporal operations, whereby the spatial methods are specified as abstract methods that must be implemented in subclasses that define a two- or three-dimensional spatial extent. The *AbstractDataset* class is derived from two parent classes, *SpatialTopologyDatasetConnector* and *TemporalTopologyDatasetConnector*. These classes provide the functionality to connect *AbstractDataset* objects to build a spatio-temporal topology structure. To actually build the spatio-temporal topology a dedicated class, *SpatioTemporalTopologyBuilder*, was implemented. This class builds the spatio-temporal topology of an unordered list of *AbstractDataset* objects, or between two unordered lists of *AbstractDataset* objects using their connector functionality. The *SpatioTemporalTopologyBuilder* requires the GRASS GIS vector library R*-Tree that is used for topological vector operations. This library is implemented in C and was improved to allow the fast and memory efficient creation of spatio-temporal topology structures.

Raster, 3D raster and vector layer are represented in the temporal framework by the *AbstractMapDataset* class, a subclass of *AbstractDataset*. It implements methods to set the temporal extent and to buffer the temporal and spatial extents. Map layers can be registered in several different space-time datasets at the same time, hence they must keep track in which space-time datasets they are registered to perform coherent deletion and un-registration operations. This is assured by the methods *get_registered_stds()*, *add_stds_to_register()* and *remove_stds_from_register()*. The class *AbstractMapDataset* defines abstract methods that must be implemented in its subclasses, since they are data type dependent. One of these methods is *load()* that reads the metadata of a specific map from the spatial database support files and stores it into the object dictionary. The metadata can then be accessed and inserted or updated in the temporal database. A specific *delete()* method was implemented for this class to assure that map layer are also deleted from all space-time datasets in which they are registered.

Space-time datasets are represented by the *AbstractSpaceTimeDataset* class, a subclass of *AbstractDataset*, see figure 3. The spatio-temporal metadata of a space-time dataset is defined by its registered map layers. Hence, there are methods to *register()* and *un-register()* map layers. The STDS metadata can be created or updated using the method *update_from_registered_maps()*. This method will use an SQL based update mechanism to update the temporal database structure of the space-time datasets and will load this metadata into the space-time dataset object representation. Based on this information the temporal granularity is computed and can be requested using the *get_granularity()* method. The *AbstractSpaceTimeDataset* class provides functions to analyse the temporal topology. Several methods are implemented to return the registered map layers for different purposes. Three methods are shown in figures 4, 5 and 6 that visualise the resulting object lists of a space-time dataset with three registered map layers A, B and C.

The map layers have different disjoint temporal extents of absolute time.
 The granularity of the space-time dataset is one day.

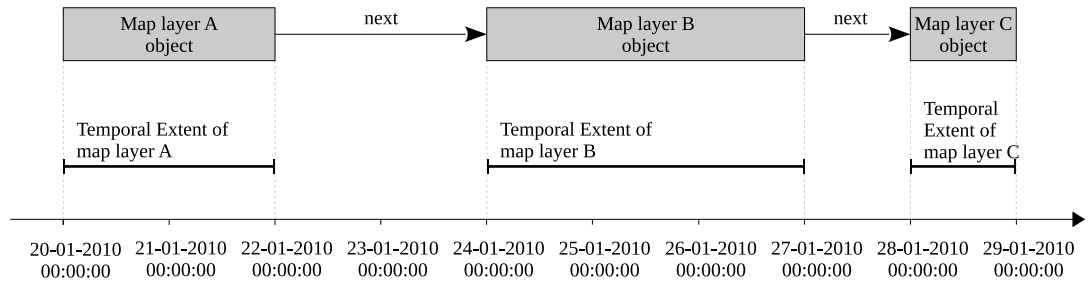


Figure 4: Method *get_registered_maps_as_objects()* returns a temporally ordered list of map layer objects.

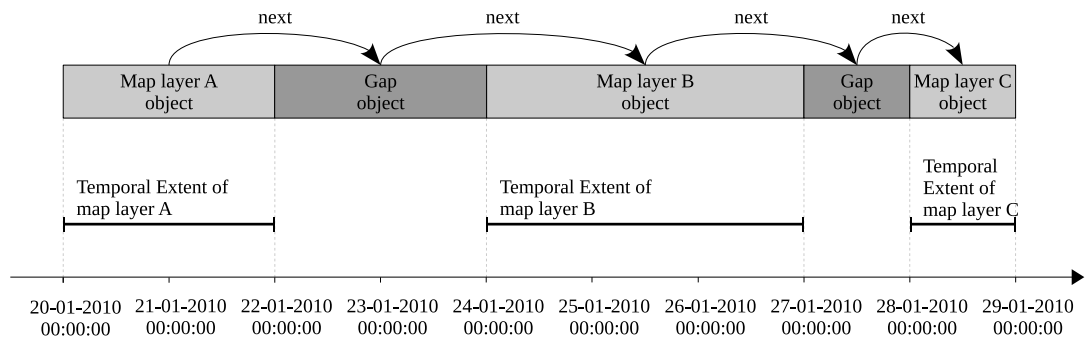


Figure 5: Method *get_registered_maps_as_objects_with_gaps()* returns a temporally ordered list of map layer objects and additional *AbstractMapDataset* objects that represent gaps between the map layers.

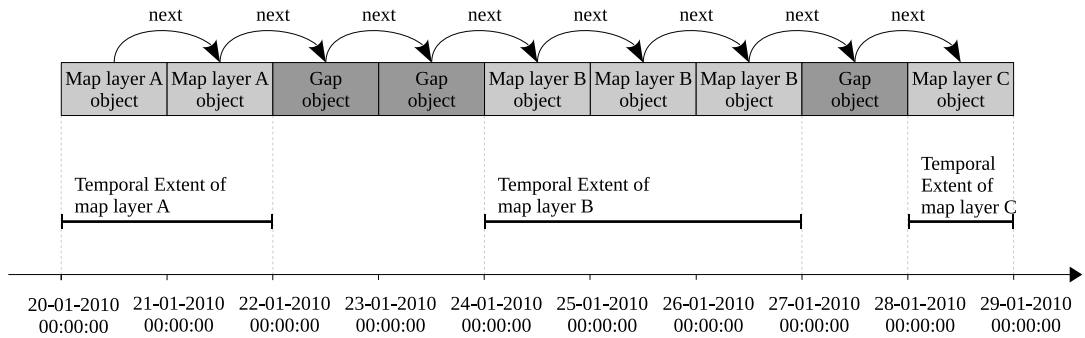


Figure 6: Method `get_registered_maps_as_objects_by_granularity()` was implemented to return a temporally ordered list of map layer objects and *AbstractMapDataset* objects that represent gaps with temporal extents similar to the STDS granularity. This method works only for STDS that have a valid temporal topology.

The temporal operations `shift()` and `snap()` are supported by our framework. Temporal shifting of a space-time dataset means that the time stamps of all registered map layers are moved by a specific granule³ in the future or in the past. To snap a space-time dataset means to create a valid temporal topology by setting the end time of a map layer to the start time of its nearest neighbour in the future. The nearest neighbour will become the temporal topological successor. The end time of the latest map layer will be adjusted⁴ by the temporal granularity of the space-time dataset.

Figures 7 and 8 show the simplified inheritance structure of type specific map layer and space-time dataset classes that implement all abstract methods and specify the type dependent metadata objects for content storage and to interface with the temporal database.

³For example one year or two weeks or 5 minutes

⁴ $\text{end.time} = \text{start.time} + \text{granularity}$

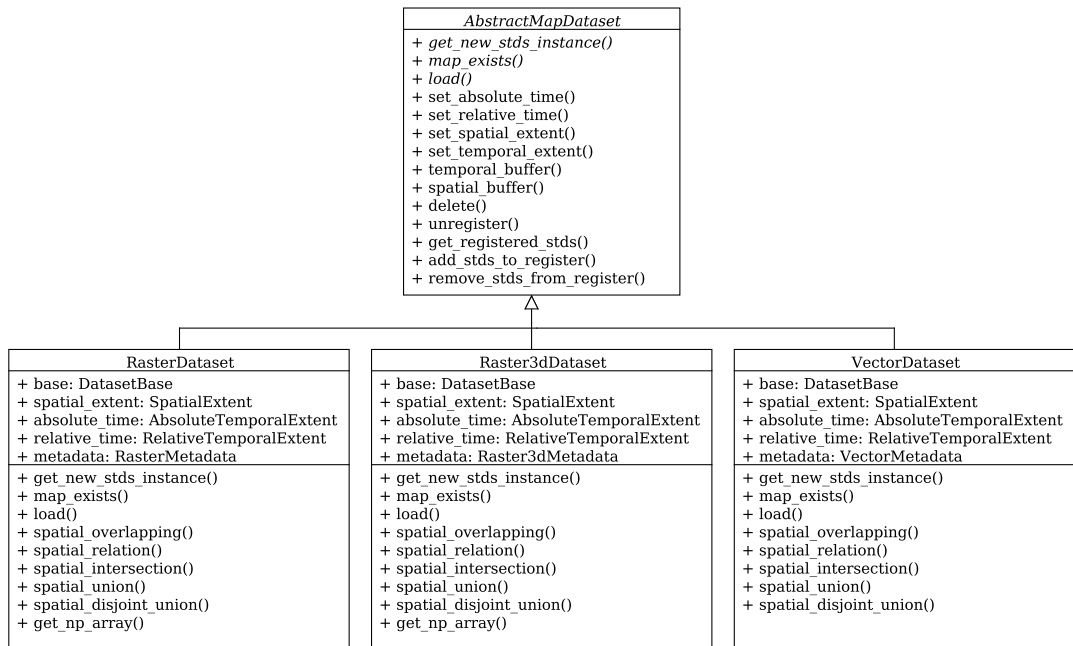


Figure 7: UML diagram showing the simplified functionality and inheritance of type specific map layer classes.

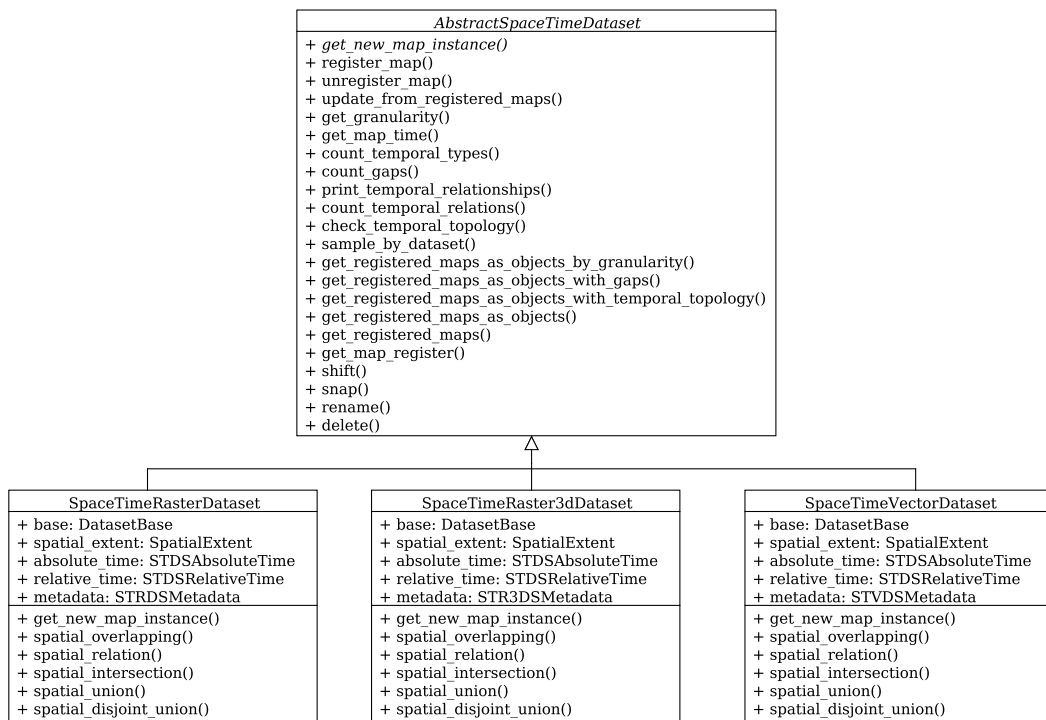


Figure 8: UML diagram showing the simplified functionality and inheritance of type specific space-time dataset classes.

3 Code examples

Several Python code examples were developed to demonstrate the capabilities of the temporal framework.

The first example in listing 1 shows how to import, insert and register 5479 raster layers in the temporal database and in a new STRDS. The raster layers, representing daily mean temperatures of Europe, are provided by the European Climate Assessment and Dataset Project [Haylock et al.(2008)Haylock, Hofstra, Tank, Klok, Jones, and New] as netCDF file.

The imported and registered daily mean temperatures are aggregated to monthly mean temperatures in degree Celsius in the second example in listing 2. The GRASS modules *r.series* and *r.mapcalc* are used for aggregation and temperature unit correction⁵. We use temporal topological relations to decide which map layer are located within a monthly granule.

In the third example, listing 3, the monthly mean temperatures are sampled by three vector points from an existing vector layer that are located in the city centres of Paris, London and Berlin. We demonstrate the direct access of vector features and raster pixel in the temporal framework.

⁵The temperature data must be divided by 10

The last example in listing 4 shows how to use the GRASS GIS Temporal Framework to perform spatio-temporal intersections between trajectory data. The input data of time stamped GPS coordinates from two Zebras in northern Botswana was provided by [Bartlam-Brooks and Harris(2014)]. About 13000 single GPS positions are available in two comma separated value (csv) files, one for each zebra. The spatio-temporal coordinates from the csv files are used directly without the creation of intermediate vector layers. The spatio-temporal locations of each Zebra is buffered in space and snapped in time to allow spatio-temporal intersection computation. The result is filtered to detect only meaningful events in which the two animals may have met.

Listing 5 shows the Python code that was used to perform benchmark runs for typical use cases of the temporal framework.

3.1 Registration of existing raster layers

Listing 1: **Import and registration of European daily mean temperature raster layers from 1950 - 1964** from the gridded (EOB-S) European Climate and Assessment Dataset (ECAD) by [Haylock et al.(2008)Haylock, Hofstra, Tank, Klok, Jones, and New]. We use a dedicated temporal database connection interface to speedup the database access. The GRASS module *r.in.gdal* is used to import the netCDF file with temperature time series raster data. This module will create 5479 new raster layer in the current mapset. The numbering scheme is *temperature_daily.[1-5479]*. It is used to create *RasterDataset* objects that are inserted into the temporal database and registered in a new created STRDS. Finally the metadata of the new STRDS is updated and printed to stdout.

```

1 #!/usr/bin/env python
2 from datetime import datetime
3 from datetime import timedelta
4 from grass.pygrass import modules
5 import grass.temporal as tgis
6
7 tgis.init()
8
9 dbif = tgis.SQLDatabaseInterfaceConnection()
10 dbif.connect()
11
12 modules.Module("r.in.gdal", input="tg_0.25deg_reg_1950-1964_v10.0.nc",
13                 output="temperature_daily", flags="o", overwrite=True)
14
15
16 temperature = tgis.open_new_strds(name="temperature_daily",
17                                  type="strds",
18                                  temporaltype="absolute",
19                                  title="Daily mean temperature Europe",
20                                  descr="Daily mean temperature of Europe 1950 -
21                                         1964",
22                                  semantic="mean",
23                                  overwrite=True, dbif=dbif)
24
25 for day in xrange(5479):
26     temperature_id = "temperature_daily.%i@s"%(day + 1, tgis.get_current_mapset
27     ())
28     temperature_layer = tgis.RasterDataset(temperature_id)
29     temperature_layer.load()

```

```

30 start = datetime(1950, 1, 1) + timedelta(day)
31 end = datetime(1950, 1, 1) + timedelta(day + 1)
32
33 temperature_layer.set_absolute_time(start, end)
34
35 if temperature_layer.is_in_db(dbif=dbif) is False:
36     temperature_layer.insert(dbif=dbif)
37 else:
38     temperature_layer.update(dbif=dbif)
39
40 temperature.register_map(map=temperature_layer, dbif=dbif)
41
42 temperature.update_from_registered_maps(dbif=dbif)
43 temperature.print_info()
44
45 dbif.close()

```

Running the script will result in the following output:

```

1 """
2 |----- Space-Time Raster Dataset -----|
3 |
4 |----- Basic information -----|
5 | Id: ..... temperature_daily@soeren
6 | Name: ..... temperature_daily
7 | Mapset: ..... soeren
8 | Creator: ..... soeren
9 | Temporal type: ..... absolute
10 | Creation time: ..... 2014-06-11 18:55:21.727892
11 | Modification time: ..... 2014-06-11 18:56:41.223508
12 | Semantic type: ..... mean
13 |----- Absolute time -----|
14 | Start time: ..... 1950-01-01 00:00:00
15 | End time: ..... 1965-01-01 00:00:00
16 | Granularity: ..... 1 day
17 | Temporal type of maps: ..... interval
18 |----- Spatial extent -----|
19 | North: ..... 75.5
20 | South: ..... 25.25
21 | East: ..... 75.5
22 | West: ..... -40.5
23 | Top: ..... 0.0
24 | Bottom: ..... 0.0
25 |----- Metadata information -----|
26 | Raster register table: .....
27 | raster_map_register_5c8a728f003a4bf7b676c1b26bb91e6b
28 | North-South resolution min:.. 0.25
29 | North-South resolution max:.. 0.25
30 | East-west resolution min:.. 0.25
31 | East-west resolution max:.. 0.25
32 | Minimum value min: ..... -4961.0
33 | Minimum value max: ..... 517.0
34 | Maximum value min: ..... 969.0
35 | Maximum value max: ..... 3704.0
36 | Aggregation type: ..... None
37 | Number of registered maps:.. 5479
38 |
39 | Title:
40 | Daily mean temperature Europe
41 | Description:
42 | Daily mean temperature of Europe 1950 - 1964
43 | Command history:
44 | # 2014-06-11 18:55:21
45 | import register.py
46 |-----|
47 """

```

3.2 Temporal aggregation of a raster time series

Listing 2: **Aggregation of daily mean temperature to monthly mean temperature.** To guarantee correct aggregation the region of the current mapset is set to the daily mean temperature map layer extent and resolution using the GRASS module *g.region*. The daily STRDS is opened and

a new STRDS that will manage the monthly aggregated data is created. We use a dedicated temporal database connection interface to speed-up the database access. A list of temporary *RasterDataset* objects is created that have monthly time stamps. This list is used together with the daily map layer object list to compute the temporal topology between them. All daily mean temperature layer that are temporally located during a temporary monthly map layer object are aggregated using the GRASS module *r.series*. The module *r.mapcalc* is then used to adjust the unit to degree Celsius. The monthly temporal extent of the temporary map layer object is copied to the resulting map layer. This map layer is then inserted into the temporal database and registered in the monthly STRDS. Finally the metadata of the new monthly STRDS is updated and printed to stdout.

```

1 #!/usr/bin/env python
2 from datetime import datetime
3 from grass.pygrass import modules
4 import grass.temporal as tgis
5
6 tgis.init()
7
8 modules.Module("g.region", rast="temperature-daily.1")
9
10 temperature_daily = tgis.open_old_stds("temperature-daily", "strds")
11
12 temperature_monthly = tgis.open_new_stds(name="temperature-monthly",
13                                         type="strds",
14                                         temporaltype="absolute",
15                                         title="Monthly mean temperature Europe",
16                                         descr="Monthly mean temperature of Europe "
17                                                "1950 - 1964 in degree Celsius",
18                                         semantic="mean",
19                                         overwrite=True)
20
21 dbif = tgis.SQLDatabaseInterfaceConnection()
22 dbif.connect()
23
24 granularity_list = []
25 start = datetime(1950, 1, 1)
26
27 for i in xrange(15 * 12):
28     end = tgis.increment_datetime_by_string(start, "1 month")
29     granule = tgis.RasterDataset(None)
30     granule.set_absolute_time(start, end)
31     granularity_list.append(granule)
32     start = end
33
34 daily_list = temperature_daily.get_registered_maps_as_objects(dbif=dbif)
35
36 topo_builder = tgis.SpatioTemporalTopologyBuilder()
37 topo_builder.build(mapsA=granularity_list, mapsB=daily_list, spatial=None)
38
39 count = 0
40 for granule in granularity_list:
41     count += 1
42     if granule.contains:
43
44         output_name = "%s-%i"%(temperature_monthly, count)
45
46         aggregation_list = []
47         for map_layer in granule.contains:
48             aggregation_list.append(map_layer.get_name())
49
50         modules.Module("r.series", input=aggregation_list,
51                         output="result", method="average",
52                         overwrite=True, quiet=True)
53
54         modules.Module("r.mapcalc",
55                         expression="%s = %s/%f"%(output_name, "result", 100.0),
56                         overwrite=True, quiet=True)
57
58         new_map_layer = tgis.RasterDataset("%s@%s"%(output_name,
59                                                     tgis.get_current_mapset()))
60         new_map_layer.load()
61         new_map_layer.set_temporal_extent(granule.get_temporal_extent())

```

```

62         if new_map_layer.is_in_db(dbif=dbif) is False:
63             new_map_layer.insert(dbif=dbif)
64         else:
65             new_map_layer.update(dbif=dbif)
66
67         temperature_monthly.register_map(new_map_layer)
68
69 temperature_monthly.update_from_registered_maps(dbif=dbif)
70 temperature_monthly.print_info()
71
72 dbif.close()
73

```

Running the script will result in the following output:

```

1 """
2 |----- Space-Time Raster Dataset -----|
3 |
4 |----- Basic information -----|
5 | Id: ..... temperature_monthly@soeren
6 | Name: ..... temperature_monthly
7 | Mapset: ..... soeren
8 | Creator: ..... soeren
9 | Temporal type: ..... absolute
10 | Creation time: ..... 2014-06-17 08:46:31.782002
11 | Modification time: ..... 2014-06-17 08:47:30.440710
12 | Semantic type: ..... mean
13 |----- Absolute time -----|
14 | Start time: ..... 1950-01-01 00:00:00
15 | End time: ..... 1965-01-01 00:00:00
16 | Granularity: ..... 1 month
17 | Temporal type of maps: ..... interval
18 |----- Spatial extent -----|
19 | North: ..... 75.5
20 | South: ..... 25.25
21 | East: ..... 75.5
22 | West: ..... -40.5
23 | Top: ..... 0.0
24 | Bottom: ..... 0.0
25 |----- Metadata information -----|
26 | Raster register table: .....
27 | raster_map_register_2596306ec91749f8843adbdba48222dc
28 | North-South resolution min:.. 0.25
29 | North-South resolution max:.. 0.25
30 | East-west resolution min:.... 0.25
31 | East-west resolution max:.... 0.25
32 | Minimum value min: ..... -32.162258
33 | Minimum value max: ..... 5.229677
34 | Maximum value min: ..... 12.125517
35 | Maximum value max: ..... 33.13
36 | Aggregation type: ..... None
37 | Number of registered maps:.. 180
38 | Title:
39 | Monthly mean temperature Europe
40 | Description:
41 | Monthly mean temperature of Europe 1950 - 1964 in degree Celsius
42 | Command history:
43 | # 2014-06-17 08:46:31
44 | ECAD-aggregation.py
45 |
46 |-----|
47 """

```

3.3 Sampling a raster time series at vector points

Listing 3: **Sampling European monthly mean temperatures with vector points.** The module *g.region* is used to set the correct spatial extent and resolution for sampling. The STRDS *temperature_monthly* is opened and a list of all registered raster layer objects is generated that is ordered by start time. The vector layer *observations* is opened which contains the coordinates of Paris, London and Berlin. The coordinates are read into a

list. A raster layer will be opened for direct pixel access for each map layer object in the list. Each raster layer is sampled at specific positions for each vector point and the sampling result is printed to stdout.

```

1 #!/usr/bin/env python
2 from datetime import datetime
3 from grass.pygrass.modules import Module
4 from grass.pygrass.raster import RasterRow
5 from grass.pygrass.vector import VectorTopo
6 from grass.pygrass.functions import coor2pixel
7 from grass.pygrass.gis.region import Region
8 import grass.temporal as tgis
9
10 tgis.init()
11
12 Module("g.region", raster="temperature_monthly.1")
13
14 temperature_monthly = tgis.open_old_stds("temperature_monthly", "strds")
15
16 monthly_list = temperature_monthly.get_registered_maps_as_objects(order="
    start_time")
17
18 region = Region()
19
20 capitals = ["Paris ", "London", "Berlin"]
21
22 vector_layer = VectorTopo("observations")
23 vector_layer.open("r")
24 vector_iterator = vector_layer.viter("points")
25
26 point_list = []
27
28 for point, capital in zip(vector_iterator, capitals):
29     print capital, point
30     point_list.append(point)
31
32 vector_layer.close()
33
34 for layer in monthly_list:
35     start, end = layer.get_absolute_time()
36     name = layer.get_name()
37
38     raster_layer = RasterRow(name)
39     raster_layer.open("r")
40
41     for point, capital in zip(point_list, capitals):
42         x, y = coor2pixel(point.coords(), region)
43         value = raster_layer[int(x)][int(y)]
44         print capital, start.date(), "-", end.date(), " value: ", value
45
46     raster_layer.close()

```

Running the script will result in the following output:

```

1 ""
2 Paris POINT(2.351667 48.856667)
3 London POINT(-0.118320 51.509390)
4 Berlin POINT(13.408056 52.518611)
5 Paris 1950-01-01 - 1950-02-01 value: 2.80483870968
6 London 1950-01-01 - 1950-02-01 value: 3.59225806452
7 Berlin 1950-01-01 - 1950-02-01 value: -1.71967741935
8 Paris 1950-02-01 - 1950-03-01 value: 7.51857142857
9 London 1950-02-01 - 1950-03-01 value: 6.0625
10 Berlin 1950-02-01 - 1950-03-01 value: 2.82821428571
11 Paris 1950-03-01 - 1950-04-01 value: 8.62870967742
12 London 1950-03-01 - 1950-04-01 value: 7.65387096774
13 Berlin 1950-03-01 - 1950-04-01 value: 5.24225806452
14 Paris 1950-04-01 - 1950-05-01 value: 9.62633333333
15 London 1950-04-01 - 1950-05-01 value: 8.129
16 Berlin 1950-04-01 - 1950-05-01 value: 7.861
17 Paris 1950-05-01 - 1950-06-01 value: 15.2548387097
18 London 1950-05-01 - 1950-06-01 value: 11.6490322581
19 Berlin 1950-05-01 - 1950-06-01 value: 15.4722580645
20 ...
21 Paris 1964-07-01 - 1964-08-01 value: 20.13
22 London 1964-07-01 - 1964-08-01 value: 17.5635483871
23 Berlin 1964-07-01 - 1964-08-01 value: 19.2822580645
24 Paris 1964-08-01 - 1964-09-01 value: 19.2403225806
25 London 1964-08-01 - 1964-09-01 value: 16.8190322581
26 Berlin 1964-08-01 - 1964-09-01 value: 16.6712903226
27 Paris 1964-09-01 - 1964-10-01 value: 17.6486666667
28 London 1964-09-01 - 1964-10-01 value: 15.099

```

```

29 Berlin 1964-09-01 - 1964-10-01 value: 14.168
30 Paris 1964-10-01 - 1964-11-01 value: 10.0380645161
31 London 1964-10-01 - 1964-11-01 value: 8.95903225806
32 Berlin 1964-10-01 - 1964-11-01 value: 7.99741935484
33 Paris 1964-11-01 - 1964-12-01 value: 7.721
34 London 1964-11-01 - 1964-12-01 value: 7.68933333333
35 Berlin 1964-11-01 - 1964-12-01 value: 5.38333333333
36 Paris 1964-12-01 - 1965-01-01 value: 3.32
37 London 1964-12-01 - 1965-01-01 value: 3.30580645161
38 Berlin 1964-12-01 - 1965-01-01 value: 1.46451612903
39 " " "

```

3.4 Spatio-temporal intersection of trajectories

Listing 4: **Spatio-temporal intersection of animal tracking data from two Zebras in northern Botswana.** We need several helper functions to perform the intersection of the GPS tracking paths of two Zebras. The Zebra with the identifier 37743 has about 10900 GPS coordinates measured from 2007-10-28 to 2008-08-28. The Zebra with id 3864 has about 2600 GPS coordinates measured from 2007-10-29 to 2008-01-05. The function `read_csv_file()` will read the prepared CSV file of a single Zebra and creates based on the UTM-34S coordinates and the time stamp new *Vector-Dataset* objects managed in a list. The temporal extent is the event of measurement and the spatial extent is a single GPS point. To perform a spatio-temporal intersection we need to spatially buffer the GPS points. The function `spatial_buffer()` was designed to buffer the spatial extent of a *Vector-Dataset* using the distance to its nearest temporal neighbour in the future. The distance is divided by 8 and used for buffering. The function `temporal_topology_check()` checks for two map layer objects if they are temporally related and calls an intersection function when they are related. The intersection function `intersection()` performs the spatio-temporal intersection of the temporal and spatial extents of two *VectorDataset* objects. It computes the duration of the temporal intersection and the size of the areas resulting from the spatial intersection. This information is then printed to stdout. The code that makes use of these functions is located below the `intersection()` function. First the CSV files for each Zebra is read. Then the time stamps are snapped for each Zebra map layer object list, to create a valid temporal topology without gaps. The function for spatial buffering is called for each list and the spatio-temporal topology is built between the lists. Finally the spatial topology relations are checked and if a relation was found the temporal topology checker is called that performs the spatio-temporal intersection.

```

1 #!/usr/bin/env python
2 from datetime import datetime
3 import csv
4 import math
5 from grass.pygrass import vector
6 from grass.pygrass.vector import geometry
7 from grass.pygrass import modules
8 import grass.temporal as tgis
9
10 def read_csv_file(filename):
11     csv_reader = csv.reader(open(filename, "r"))

```



```

12 layer_list = []
13 for line in csv_reader:
14     east=float(line[0])
15     north=float(line[1])
16     start = datetime.strptime(line[2], "%Y-%m-%d %H:%M:%S")
17
18     tmp_layer = tgis.VectorDataset(None)
19     tmp_layer.set_absolute_time(start, None)
20     tmp_layer.set_spatial_extent_from_values(north=north, south=north,
21                                             east=east, west=east,
22                                             top=0, bottom=0)
23
24     layer_list.append(tmp_layer)
25
26 return layer_list
27 #####
28
29 def spatial_buffering(layer_list):
30     for i in xrange(len(layer_list)):
31         if i < (len(layer_list) - 1):
32             ext_a = layer_list[i].spatial_extent.get_spatial_extent_as_tuple_2d()
33             ext_b = layer_list[i + 1].spatial_extent.get_spatial_extent_as_tuple_2d()
34             dx = ext_a[2] - ext_b[2]
35             dy = ext_a[0] - ext_b[0]
36             distance = math.sqrt(dx*dx + dy*dy)/8.0
37         else:
38             distance = 50
39
40         layer_list[i].spatial_buffer_2d(distance)
41         layer_list[i].buffer_size = distance
42
43 #####
44
45 def temporal_topology_check(layer, sublayer, srelation):
46     if layer.overlaps and sublayer in layer.overlaps:
47         intersection(layer, sublayer, srelation, "overlaps")
48     if layer.overlapped and sublayer in layer.overlapped:
49         intersection(layer, sublayer, srelation, "overlapped")
50     if layer.equal and sublayer in layer.equal:
51         intersection(layer, sublayer, srelation, "equal")
52     if layer.during and sublayer in layer.during:
53         intersection(layer, sublayer, srelation, "during")
54     if layer.contains and sublayer in layer.contains:
55         intersection(layer, sublayer, srelation, "contains")
56
57 #####
58
59 def intersection(layer_a, layer_b, srelation, trelation):
60     spatial_extent = layer_a.spatial_intersection(layer_b)
61     temporal_extent = layer_a.temporal_intersection(layer_b)
62
63     extent = spatial_extent.get_spatial_extent_as_tuple_2d()
64     duration = temporal_extent.get_end_time() - temporal_extent.get_start_time()
65
66     if duration.days == 0 and duration.seconds < 60:
67         return
68     if layer_a.buffer_size > 1000 or layer_b.buffer_size > 1000:
69         return
70
71     area = spatial_extent.get_area()
72     full_area = layer_a.spatial_extent.get_area() + \
73               layer_b.spatial_extent.get_area()
74
75     percent = 100*area/full_area
76
77     print "Zebra 3864 met Zebra 3743 from", temporal_extent.get_start_time(), \
78           "to", temporal_extent.get_end_time()
79     print "    Duration:", duration, "\n    Area:" \
80           int(area), "[m^2] about %i%% of full area,"%(percent)
81     print "    Spatio-temporal relation:", srelation, trelation
82
83 #####
84 tgis.init()
85
86 zebra_3864_list = read_csv_file("Zebra_Id3864.csv")
87 zebra_3743_list = read_csv_file("Zebra_Id3743.csv")
88
89 tgis.AbstractSpaceTimeDataset.snap_map_list(zebra_3864_list)
90 tgis.AbstractSpaceTimeDataset.snap_map_list(zebra_3743_list)
91
92 spatial_buffering(zebra_3864_list)
93 spatial_buffering(zebra_3743_list)
94

```

```

95 topo_builder = tgis.SpatioTemporalTopologyBuilder()
96 topo_builder.build(mapsA=zebra_3864_list, mapsB=zebra_3743_list, spatial="2D")
97
98 for layer in zebra_3864_list:
99     if layer.overlap:
100         for sublayer in layer.overlap:
101             temporal_topology_check(layer, sublayer, "overlap")
102     if layer.cover:
103         for sublayer in layer.cover:
104             temporal_topology_check(layer, sublayer, "cover")
105     if layer.covered:
106         for sublayer in layer.covered:
107             temporal_topology_check(layer, sublayer, "covered")
108     if layer.contain:
109         for sublayer in layer.contain:
110             temporal_topology_check(layer, sublayer, "contain")
111     if layer.in_:
112         for sublayer in layer.in_:
113             temporal_topology_check(layer, sublayer, "in")

```

Running the script will result in the following output:

```

1 """
2 Zebra 3864 met Zebra 3743 from 2007-12-27 23:00:53 to 2007-12-28 00:00:09
3     Duration: 0:59:16
4     Area: 1491 [m^2] about 8% of full area,
5     Spatio-temporal relation: overlap during
6 Zebra 3864 met Zebra 3743 from 2007-12-28 00:00:13 to 2007-12-28 01:00:13
7     Duration: 1:00:00
8     Area: 1351 [m^2] about 15% of full area,
9     Spatio-temporal relation: overlap overlaps
10 Zebra 3864 met Zebra 3743 from 2007-12-28 01:00:24 to 2007-12-28 02:00:43
11     Duration: 1:00:19
12     Area: 11708 [m^2] about 19% of full area,
13     Spatio-temporal relation: contain contains
14 """

```

3.5 Benchmark Python code

Listing 5: This Python code was used to perform the registration, selection and topology building benchmark.

```

1 #!/usr/bin/env python
2 import grass.temporal as tgis
3 import datetime
4 import sys
5
6 def main(filename, strds, number_of_maps):
7     # Initiate the temporal database, global variables and processes
8     tgis.init()
9     # A dedicated database connection object
10    # speeds up the database access significantly
11    dbif = tgis.SQLiteDatabaseInterfaceConnection()
12    dbif.connect()
13
14    # Start map registration in a space time dataset
15    start = datetime.datetime.now()
16    # Create a new space-time raster dataset (STRDS)
17    new_strds = tgis.open_new_strds(name=strds,
18                                   type="strds",
19                                   temporaltype="absolute",
20                                   title="Benchmark STRDS with %i map layers"%(
21                                       number_of_maps),
22                                   descr="Benchmark STRDS with %i map layers"%(
23                                       number_of_maps),
24                                   semantic="mean",
25                                   overwrite=True, dbif=dbif)
26
27    # Register raster layers in the STRDS
28    tgis.register_maps_in_space_time_dataset(type="raster", name=strds,
29                                             file=filename, start="2000-01-01",
30                                             increment="1 second", interval=True,
31                                             dbif=dbif, update_cmd_list=True)
32
33    new_strds.update_from_registered_maps(dbif=dbif)
34
35    end = datetime.datetime.now()
36    print("Time for registration in [s]: ", end - start)

```

```

34
35 # Start SQL based selection operation
36 start = datetime.datetime.now()
37
38 sql_ret = new_strds.get_registered_maps(where="start_time >= '2000-01-01
39         00:00:00'")
40 print("Length of returned dict", len(sql_ret))
41
42 end = datetime.datetime.now()
43 print("Time for map selection in [s]: ", end - start)
44
45 # Start SQL based selection operation and generate a list of map objects
46 start = datetime.datetime.now()
47
48 map_list = new_strds.get_registered_maps_as_objects(where="start_time >=
49         '2000-01-01 00:00:00'")
50 print("Length of map list", len(map_list))
51
52 end = datetime.datetime.now()
53 print("Time for map selection and object generation in [s]: ", end - start)
54
55 # Spatio-temporal topology between map layers
56 start = datetime.datetime.now()
57 # We reuse the map list to count the inner temporal relations
58 relations = new_strds.count_temporal_relations(map_list, dbif=dbif)
59 # We expect only follows and precedes relations
60 print("Follows", relations["follows"])
61 print("Precedes", relations["precedes"])
62
63 end = datetime.datetime.now()
64 print("Time for build st-topology and count relations in [s]: ", end - start)
65
66 dbif.close()
67
68 if __name__ == "__main__":
69     # python benchmark_run.py maps_list.txt A 1000000
70     filename = sys.argv[1]
71     strds = sys.argv[2]
72     number_of_maps = int(sys.argv[3])
73     main(filename, strds, number_of_maps)

```

References

- [Bartlam-Brooks and Harris(2014)] H.L.A. Bartlam-Brooks and S. Harris. In search of greener pastures—using satellite images to predict the effects of environmental change on zebra migration, June 2014. url: <https://www.movebank.org/node/11921>.
- [Gebbert and Pebesma(2014)] Sören Gebbert and Edzer Pebesma. A temporal GIS for field based environmental modeling. *Environmental Modelling & Software*, 53(0):1–12, 2014. ISSN 1364-8152. doi: 10.1016/j.envsoft.2013.11.001. URL <http://www.sciencedirect.com/science/article/pii/S136481521300282X>.
- [Haylock et al.(2008)] Haylock, Hofstra, Tank, Klok, Jones, and New] M. R. Haylock, N. Hofstra, A. M. G. Klein Tank, E. J. Klok, P. D. Jones, and M. New. A European daily high-resolution gridded data set of surface temperature and precipitation for 1950 - 2006. *Journal of Geophysical Research*, 113(D20), October 2008. ISSN 0148-0227. doi: 10.1029/2008JD010201.