

osrm: Interface Between R and the OpenStreetMap-Based Routing Service OSRM

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Software

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Summary

osrm is an interface between R and the Open Source Routing Machine (OSRM) API. OSRM (Luxen & Vetter, 2011) is a routing service based on OpenStreetMap (OSM) data. This package enables the computation of routes, trips, isochrones and travel distances matrices (travel times and kilometric distances) based on the OSM road network. It contains five functions that interface with OSRM services:

- osrmTable() uses the table service to query time/distance matrices,
- osrmRoute() uses the *route* service to query routes,
- osrmTrip() uses the trip service to query trips,
- osrmIsochone() and osrmIsodistance() use multiple osrmTable() calls to create isochrones or isodistance polygons.

An instance of OSRM can be installed on a local or remote server, allowing free and heavy usage of the routing engine.

Statement of need

Distance based computations and models are at the core of many spatial analysis operations in various scientific fields. The simplest distance metric is the Euclidean distance (or distance as the crow flies) which is easy and inexpensive to compute. The use of this simple metric may be well suited to study some phenomenon, such as species distribution or pollution diffusion. But whenever research aims at studying human activities (transport of persons or goods for example) it is common to use more realistic metrics based on road distance or travel time.

To compute these metrics with R one has to use packages that interface routing engines. Most routing engines are commercial, use tokens to limit the number of requests, or heavily restrict the usage of derived datasets. The use of open source software based on open source data enables a high level of transparency useful to research works that aim at reproducibility.

We argue that osrm offers such a level of transparency by relying on the open source software OSRM, which itself uses the open data source OSM.

osrm is already used in various fields such as transport (Danesi & Tengattini, 2020; Fernandes Barroso et al., 2021; Savaria et al., 2021), education (Odell, 2017), health (Chen et al., 2021; Rice et al., 2021; Snyder & Smucker, 2022; Wisch et al., 2022), applied geography (Kandlbinder et al., 2019; Oberst & Voigtländer, 2021; Wieland, 2018, 2021), environmental science (Hickisch et al., 2019; E. Walker et al., 2021), urban planning (Ballantyne et al., 2022; Dey et al., 2022; Xu, 2019) and linguistics (Jeszenszky et al., 2019).

The use of osrm is also suggested by the package for sustainable transport planning stplanr (Lovelace & Ellison, 2018)



State of the field

Several packages exist to compute routes, trips or distance matrices. Most of them rely on commercial and non-free software and use non-free data. See for example hereR (Unterfinger, 2021) that uses here services, gmapsdistance (Azuero Melo & Zarruk, 2022), googleway (Cooley, 2022) and mapsapi (Dorman, 2022) that use Google Maps Platform or mapboxapi (K. Walker, 2022) that relies on Mapbox. Using these packages imposes many restrictions on data extraction, analysis and sharing. Other packages use open source routing engines and open data: graphhopper (Kuethe, 2021) uses GraphHopper, opentripplanner (Morgan et al., 2019) uses OpenTripPlanner, valhallr (Belanger, 2021) uses Valhalla. osrmr (Staempfli & Strauss, 2021) uses OSRM, however it exposes only a small subset of OSRM services and does not handle spatial data formats. Among these packages, osrm has the advantage of using OSRM, which is easy to install and run on a local or remote server, to give access to most OSRM services and to handle spatial data formats.

Features

This is a short overview of the main features of osrm. The dataset used here is shipped with the package and contains a sample of 100 random pharmacies in Berlin (© OpenStreetMap contributors) stored in a geopackage file.

osrmTable() gives access to the table OSRM service. In this example we use this function
to get the median time needed to access any pharmacy from any other pharmacy.

```
library(osrm)
## Data: (c) OpenStreetMap contributors, ODbL 1.0 - http://www.openstreetmap.org/copyrig
## Routing: OSRM - http://project-osrm.org/
library(sf)
## Linking to GEOS 3.9.0, GDAL 3.2.2, PROJ 7.2.1; sf_use_s2() is TRUE
pharmacy <- st_read(system.file("gpkg/apotheke.gpkg", package = "osrm"),</pre>
                     quiet = TRUE)
travel_time <- osrmTable(loc = pharmacy)</pre>
travel_time$durations[1:5,1:5]
             2
                  3
## 1 0.0 21.1 33.4 21.2 12.6
## 2 22.1 0.0 42.3 16.1 20.2
## 3 33.0 43.0 0.0 30.5 27.4
## 4 20.1 15.3 29.7 0.0 12.7
## 5 10.2 20.3 26.8 12.3 0.0
diag(travel_time$durations) <- NA</pre>
median(travel_time$durations, na.rm = TRUE)
## [1] 21.4
```

The median time needed to access any pharmacy from any other pharmacy is 21.4 minutes.

• osrmRoute() is used to compute the shortest route between two points. Here we compute the shortest route between the two first pharmacies.

```
(route <- osrmRoute(src = pharmacy[1, ], dst = pharmacy[2, ]))
## Simple feature collection with 1 feature and 4 fields
## Geometry type: LINESTRING</pre>
```



```
## Dimension: XY
## Bounding box: xmin: -13170.51 ymin: 5837172 xmax: -3875.771 ymax: 5841047
## Projected CRS: WGS 84 / UTM zone 34N
## src dst duration distance geometry
## 1_2 1 2 21.11667 12.348 LINESTRING (-13170.51 58410...
This route is 12.3 kilometers long and it takes 21.1 minutes to drive.
plot(st_geometry(route))
plot(st_geometry(pharmacy[1:2,]), pch = 20, add = T, cex = 1.5)
```

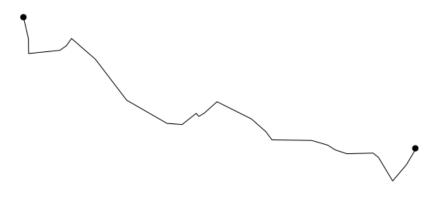


Figure 1: Shortest route between two points.

 osrmTrip() can be used to resolve the travelling salesman problem, it gives the shortest trip between a set of unordered points. In this example we want to obtain the shortest trip between the first five pharmacies.

```
(trips <- osrmTrip(loc = pharmacy[1:5, ], overview = "full"))</pre>
## [[1]]
## [[1]]$trip
## Simple feature collection with 5 features and 4 fields
## Geometry type: LINESTRING
## Dimension:
                  XY
## Bounding box: xmin: -13431.24 ymin: 5837172 xmax: -3875.582 ymax: 5856332
## Projected CRS: WGS 84 / UTM zone 34N
     start end duration distance
##
                                                        geometry
## 1
             2 21.11667 12.3480 LINESTRING (-13170.77 58410...
## 2
            4 16.10833
                         8.4273 LINESTRING (-3875.582 58379...
## 3
            3 29.69000 18.1448 LINESTRING (-7444.513 58427...
            5 27.39833 16.4265 LINESTRING (-8027.41 585621...
## 4
         3
                         4.2289 LINESTRING (-11716.36 58435...
## 5
             1 10.15333
##
```



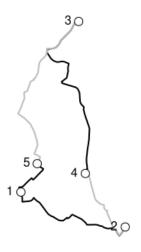


Figure 2: Shortest trip between a set of unordered points.

osrmIsochrone() computes areas that are reachable within a given time span from a
point and returns the reachable regions as polygons. These areas of equal travel time
are called isochrones. Here we compute the isochrones from a specific point defined by
its longitude and latitude.

```
(iso <- osrmIsochrone(loc = c(13.43,52.47), breaks = seq(0,12,2)))
## Simple feature collection with 5 features and 3 fields
## Geometry type: MULTIPOLYGON
## Dimension: XY
## Bounding box: xmin: 13.34397 ymin: 52.41642 xmax: 13.50187 ymax: 52.51548
## Geodetic CRS: WGS 84
## id isomin isomax geometry</pre>
```



```
## 1
                    4 MULTIPOLYGON (((13.43743 52...
                    6 MULTIPOLYGON (((13.42356 52...
## 3 3
             6
                    8 MULTIPOLYGON (((13.40345 52...
## 4 4
             8
                   10 MULTIPOLYGON (((13.4077 52....
## 5 5
                   12 MULTIPOLYGON (((13.42257 52...
bks <- sort(unique(c(iso$isomin, iso$isomax)))</pre>
pals <- hcl.colors(n = length(bks) - 1, palette = "Light Grays", rev = TRUE)</pre>
plot(iso["isomax"], breaks = bks, pal = pals,
     main = "Isochrones (in minutes)", reset = FALSE)
points(x = 13.43, y = 52.47, pch = 4, lwd = 2, cex = 1.5)
```

Isochrones (in minutes)

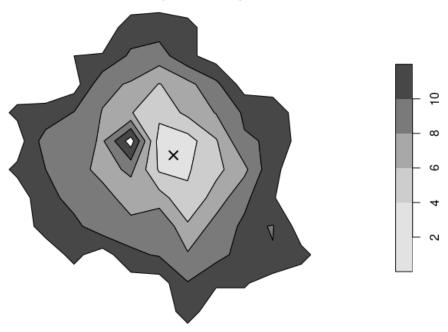


Figure 3: Isochrone map.

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