Introduction to package nngeo

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Introduction

Package purpose

This document introduces the nngeo package. The nngeo package includes functions for spatial join of layers based on *k-nearest neighbor* relation between features. The functions work with spatial layer object defined in package sf, namely classes sfc and sf.

Installation

GitHub version:

```
install.packages("devtools")
devtools::install_github("michaeldorman/nngeo")
```

Sample data

The nngeo package comes with three sample datasets:

- cities
- towns
- water

The cities layer is a point layer representing the location of the three largest cities in Israel.

```
cities
#> Simple feature collection with 3 features and 1 field
#> geometry type: POINT
#> dimension: XY
#> bbox: xmin: 34.78177 ymin: 31.76832 xmax: 35.21371 ymax: 32.79405
#> geographic CRS: WGS 84
#> name geometry
#> 1 Jerusalem POINT (35.21371 31.76832)
#> 2 Tel-Aviv POINT (34.78177 32.0853)
#> 3 Haifa POINT (34.98957 32.79405)
```

The towns layer is another **point** layer, with the location of all large towns in Israel, compiled from a different data source:

```
towns
#> Simple feature collection with 193 features and 4 fields
#> geometry type: POINT
#> dimension:
                        XY
#> bbox:
                         xmin: 34.27 ymin: 29.56 xmax: 35.6 ymax: 33.21
#> geographic CRS: WGS 84
#> First 10 features:
#>
                  name country.etc pop capital
                                                                            geometry
#> 12
                 'Afula Israel 39151 0 POINT (35.29 32.62)
#> 17
          'Akko Israel 45606
'Ar'ara Israel 15841
'Arad Israel 22757
'Arrabe Israel 20316
'Atlit Israel 4686
'Eilabun Israel 4296
'Ein Mahel Israel 11014
'Ein Qiniyye Israel 2101
'Ilut Israel 6536
                  'Akko
                                Israel 45606
                                                         0 POINT (35.08 32.94)
                                                   0 POINT (35.1 32.49)
0 POINT (35.22 31.26)
0 POINT (35.33 32.85)
#> 40
#> 41
#> 43
                                                   0 POINT (34.95 C...

0 POINT (35.4 32.83)

0 POINT (35.35 32.72)
#> 52
#> 103
#> 104
#> 105 'Ein Qiniyye
                  'Ilut
                           Israel 6536 0 POINT (35.25 32.72)
#> 112
```

The water layer is an example of a **polygonal** layer. This layer contains four polygons of water bodies in Israel.

```
water
#> Simple feature collection with 4 features and 1 field
#> geometry type: POLYGON
#> dimension:
                  XY
#> bbox:
                  xmin: 34.1388 ymin: 29.45338 xmax: 35.64979 ymax: 33.1164
#> geographic CRS: WGS 84
#>
                 name
                                             geometry
#> 1
             Red Sea POLYGON ((34.96428 29.54775...
#> 2 Mediterranean Sea POLYGON ((35.10533 33.07661...
            Dead Sea POLYGON ((35.54743 31.37881...
#> 4 Sea of Galilee POLYGON ((35.6014 32.89248,...
```

Figure 1 shows the spatial configuration of the cities, towns and water layers.

```
plot(st_geometry(water), col = "lightblue")
plot(st_geometry(towns), col = "grey", pch = 1, add = TRUE)
plot(st_geometry(cities), col = "red", pch = 1, add = TRUE)
```

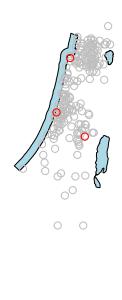


Figure 1: Visualization of the water, towns and cities layers

Usage examples

The st_nn function

The main function in the nngeo package is st_nn.

The st_nn function accepts two layers, x and y, and returns a list with the same number of elements as x features. Each list element i is an integer vector with all indices j for which x[i] and y[j] are nearest neighbors.

For example, the following expression finds which feature in towns[1:5,] is the nearest neighbor to each feature in cities:

```
nn = st_nn(cities, towns[1:5, ], progress = FALSE)
#> lon-lat points
nn
#> [[1]]
#> [1] 4
#>
#> [[2]]
#> [1] 3
#> [1] 2
```

This output tells us that towns[4,] is the nearest among the five features of towns[1:5,] to cities[1,], etc.

The st_connect function

The resulting nearest neighbor matches can be visualized using the $st_connect$ function. This function builds a line layer connecting features from two layers x and y based on the relations defined in a list such the one returned by st_nn :

```
1 = st_connect(cities, towns[1:5, ], ids = nn)
#> Calculating nearest IDs
#> Calculating lines
1
#> Geometry set for 3 features
#> geometry type: LINESTRING
#> dimension: XY
#> bbox: xmin: 34.78177 ymin: 31.26 xmax: 35.22 ymax: 32.94
#> geographic CRS: WGS 84
#> LINESTRING (35.21371 31.76832, 35.22 31.26)
#> LINESTRING (34.78177 32.0853, 35.1 32.49)
#> LINESTRING (34.98957 32.79405, 35.08 32.94)
```

Plotting the line layer 1 gives a visual demonstration of the nearest neighbors match, as shown in Figure 2.

```
plot(st_geometry(towns[1:5, ]), col = "darkgrey")
plot(st_geometry(1), add = TRUE)
plot(st_geometry(cities), col = "red", add = TRUE)
text(st_coordinates(cities)[, 1], st_coordinates(cities)[, 2], 1:3, col = "red", pos = 4)
text(st_coordinates(towns[1:5, ])[, 1], st_coordinates(towns[1:5, ])[, 2], 1:5, pos = 4)
```

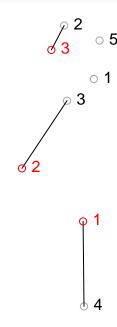


Figure 2: Nearest neighbor match between cities (in red) and towns[1:5,] (in grey)

Dense matrix representation

The st_nn can also return the complete logical matrix indicating whether each feature in x is a neighbor of y. To get the dense matrix, instead of a list, use sparse=FALSE.

```
nn = st_nn(cities, towns[1:5, ], sparse = FALSE, progress = FALSE)
#> lon-lat points
nn
#> [,1] [,2] [,3] [,4] [,5]
#> [1,] FALSE FALSE TRUE FALSE
#> [2,] FALSE FALSE TRUE FALSE FALSE
#> [3,] FALSE TRUE FALSE FALSE
```

k-Nearest neighbors where k>0

It is also possible to return any **k-nearest** neighbors, rather than just one. For example, setting k=2 returns both the 1st and 2nd nearest neighbors:

```
nn = st_nn(cities, towns[1:5, ], k = 2, progress = FALSE)
#> lon-lat points
nn
#> [[1]]
#> [1] 4 3
#>
#> [[2]]
#> [1] 3 1
#>
#> [[3]]
#> [1] 2 5
nn = st_nn(cities, towns[1:5, ], sparse = FALSE, k = 2, progress = FALSE)
#> lon-lat points
nn
         [,1] [,2] [,3] [,4] [,5]
#>
#> [1,] FALSE FALSE TRUE TRUE FALSE
#> [2,] TRUE FALSE TRUE FALSE FALSE
#> [3,] FALSE TRUE FALSE FALSE TRUE
```

Distance matrix

Using returnDist=TRUE the distances list is also returned, in addition the the neighbor matches, with both components now comprising a list:

```
nn = st_nn(cities, towns[1:5, ], k = 2, returnDist = TRUE, progress = FALSE)
#> lon-lat points
nn
#> $nn
#> $nn[[1]]
#> [1] 4 3
#>
#> $nn[[2]]
#> [1] 3 1
#>
#> $nn[[3]]
#> [1] 2 5
#>
#> $dist
#> $dist[[1]]
#> [1] 56364.74 80742.62
#>
#> $dist[[2]]
#> [1] 53968.63 76186.87
#>
#> $dist[[3]]
#> [1] 18265.72 32476.24
```

Search radius

Finally, the search for nearest neighbors can be limited to a **search radius** using maxdist. In the following example, the search radius is set to 50,000 meters (50 kilometers). Note that no neighbors are found within the search radius for cities[2,]:

```
nn = st_nn(cities, towns[1:5, ], k = 2, maxdist = 50000, progress = FALSE)
#> lon-lat points
nn
#> [[1]]
#> integer(0)
#>
#> [[2]]
#> integer(0)
#>
#> [[3]]
#> [1] 2 5
```

Spatial join

The st_nn function can also be used as a **geometry predicate function** when performing spatial join with sf::st_join. For example, the following expression spatially joins the two nearest towns[1:5,] features to each cities features, using a search radius of 50 km:

```
cities1 = st_join(cities, towns[1:5, ], join = st_nn, k = 2, maxdist = 50000)
#> lon-lat points
```

Here is the resulting layer:

```
cities1
#> Simple feature collection with 4 features and 5 fields
#> geometry type: POINT
#> dimension:
#> bbox:
                   xmin: 34.78177 ymin: 31.76832 xmax: 35.21371 ymax: 32.79405
#> geographic CRS: WGS 84
#>
         name.x name.y country.etc
                                       pop capital
                                                                     geometry
#> 1
      Jerusalem
                    <NA>
                                <NA>
                                                NA POINT (35.21371 31.76832)
                                        NA
#> 2
      \mathit{Tel-Aviv}
                    <NA>
                                <NA>
                                        NA
                                                NA POINT (34.78177 32.0853)
#> 3
          Haifa
                   'Akko
                              Israel 45606
                                               0 POINT (34.98957 32.79405)
                              Israel 20316
#> 3.1
          Haifa 'Arrabe
                                                 0 POINT (34.98957 32.79405)
```

Another example

Here is another example, finding the 10-nearest neighbor towns features for each cities feature:

```
x = st_nn(cities, towns, k = 10)
#> lon-lat points
1 = st_connect(cities, towns, ids = x)
```

The result is visualized in Figure 3.

```
plot(st_geometry(1))
plot(st_geometry(cities), col = "red", add = TRUE)
plot(st_geometry(towns), col = "darkgrey", add = TRUE)
```

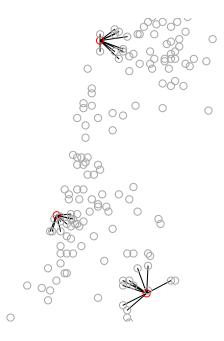


Figure 3: Nearest 10 towns features from each cities feature

Polygons

Nearest neighbor search also works for non-point layers. The following code section finds the 20-nearest towns features for each water body in water[-1,].

```
nn = st_nn(water[-1, ], towns, k = 20, progress = FALSE)
#> lines or polygons
```

Again, we can calculate the respective lines for the above result using st_connect. Since one of the inputs is line/polygon, we need to specify a sampling distance dist, which sets the resolution of connecting points on the shape exterior boundary.

```
1 = st_connect(water[-1, ], towns, ids = nn, dist = 100)
#> Calculating nearest IDs
#> Calculating lines
```

The result is visualized in Figure 4.

```
plot(st_geometry(water[-1, ]), col = "lightblue", border = "grey")
plot(st_geometry(towns), col = "darkgrey", add = TRUE)
plot(st_geometry(l), col = "red", add = TRUE)
```

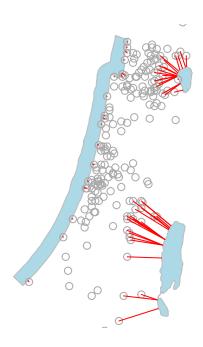


Figure 4: Nearest 20 towns features from each water polygon