

# Introduction to package `ngeo`

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## Introduction

### Package purpose

This document introduces the `ngeo` package. The `ngeo` 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/ngeo")
```

### Sample data

The `ngeo` 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
#> epsg (SRID): 4326
#> proj4string: +proj=longlat +datum=WGS84 +no_defs
#>   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
#> epsg (SRID): 4326
#> proj4string: +proj=longlat +datum=WGS84 +no_defs
#> First 10 features:
#>   name country.etc pop capital geometry
#> 12 'Afula Israel 39151 0 POINT (35.29 32.62)
#> 17 'Akko Israel 45606 0 POINT (35.08 32.94)
#> 40 'Ar'ara Israel 15841 0 POINT (35.1 32.49)
#> 41 'Arad Israel 22757 0 POINT (35.22 31.26)
#> 43 'Arrabe Israel 20316 0 POINT (35.33 32.85)
#> 52 'Atlit Israel 4686 0 POINT (34.93 32.68)
#> 103 'Eilabun Israel 4296 0 POINT (35.4 32.83)
#> 104 'Ein Mahel Israel 11014 0 POINT (35.35 32.72)
#> 105 'Ein Qiniyye Israel 2101 0 POINT (35.15 31.93)
#> 112 'Ilut Israel 6536 0 POINT (35.25 32.72)

```

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
#> epsg (SRID): 4326
#> proj4string: +proj=longlat +datum=WGS84 +no_defs
#>   name geometry
#> 1 Red Sea POLYGON ((34.96428 29.54775...
#> 2 Mediterranean Sea POLYGON ((35.10533 33.07661...
#> 3 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)

```

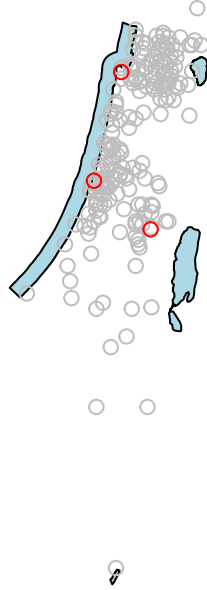


Figure 1: Visualization of the `water`, `towns` and `cities` layers

## Usage examples

### The `st_nn` function

The main function in the `ngeo` 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, ])
#> lon-lat points
#>
#> | 0%
#> |=====| 33%
#> |=====| 67%
#> |=====| 100%
nn
#> [[1]]
#> [1] 4
#>
#> [[2]]
#> [1] 3
#>
#> [[3]]
#> [1] 2
```

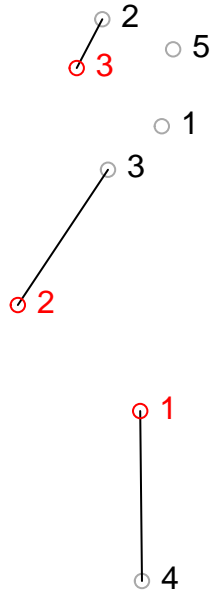


Figure 2: Nearest neighbor match between `cities` (in red) and `towns[1:5, ]` (in grey)

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`:

```
l = st_connect(cities, towns[1:5, ], ids = nn)
#> Calculating nearest IDs
#> Calculating lines
l
#> Geometry set for 3 features
#> geometry type: LINESTRING
#> dimension: XY
#> bbox: xmin: 34.78177 ymin: 31.26 xmax: 35.22 ymax: 32.94
#> epsg (SRID): 4326
#> proj4string: +proj=longlat +datum=WGS84 +no_defs
#> 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 `l` 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(l), 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)
```

## 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`.

```
mn = st_nn(cities, towns[1:5, ], sparse = FALSE)
#> lon-lat points
#>
|
|
|
|=====| 33%
|=====| 67%
|=====| 100%
mn
#>      [,1] [,2] [,3] [,4] [,5]
#> [1,] FALSE FALSE FALSE  TRUE FALSE
#> [2,] FALSE FALSE  TRUE  FALSE FALSE
#> [3,] FALSE  TRUE  FALSE  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 1<sup>st</sup> and 2<sup>nd</sup> nearest neighbors:

```
mn = st_nn(cities, towns[1:5, ], k = 2)
#> lon-lat points
#>
|
|
|
|=====| 33%
|=====| 67%
|=====| 100%
mn
#> [[1]]
#> [1] 4 3
#>
#> [[2]]
#> [1] 3 1
#>
#> [[3]]
#> [1] 2 5
```

```
mn = st_nn(cities, towns[1:5, ], sparse = FALSE, k = 2)
#> lon-lat points
#>
|
|
|
|=====| 0%
```

```

===== | 33%
|
===== | 67%
|
===== | 100%
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 componenets now comprising a list:

```

nn = st_nn(cities, towns[1:5, ], k = 2, returnDist = TRUE)
#> lon-lat points
#>
|
|
===== | 33%
|
===== | 67%
|
===== | 100%
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)
#> lon-lat points
#>
#>
#> | 0%
#> |=====| 33%
#> |=====| 67%
#> |=====| 100%
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: XY
#> bbox: xmin: 34.78177 ymin: 31.76832 xmax: 35.21371 ymax: 32.79405
#> epsg (SRID): 4326
#> proj4string: +proj=longlat +datum=WGS84 +no_defs
#>   name.x name.y country.etc pop capital geometry
#> 1 Jerusalem <NA> <NA> NA NA POINT (35.21371 31.76832)
#> 2 Tel-Aviv <NA> <NA> NA NA POINT (34.78177 32.0853)
#> 3 Haifa 'Akko Israel 45606 0 POINT (34.98957 32.79405)
#> 3.1 Haifa 'Arrabe Israel 20316 0 POINT (34.98957 32.79405)
```

## Another example

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

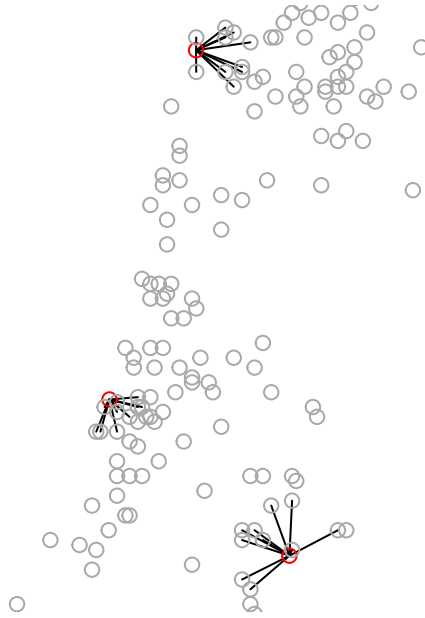


Figure 3: Nearest 10 towns features from each cities feature

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

The result is visualized in Figure 3.

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

## 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, ]`.

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

			0%
	=====		33%
	=====		67%
	=====		100%

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.



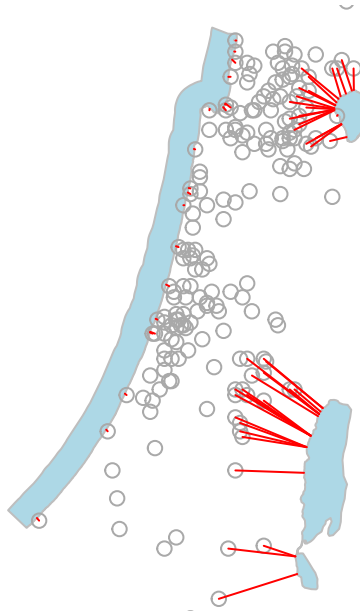


Figure 4: Nearest 20 towns features from each water polygon

```
l = 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)
```