# IrisSpatialFeatures - An R package to quantify the tumor microenvironment based on multiplex IF data

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## Short example on how to use the IrisSpatialFeatures package

### Reading the dataset

This is a toy example, based on 2 sample with 2 coordinates each. The 20x images were acquired on the Mantra system of PerkinElmer, analyzed in inForm. There are three different phenotypes present in this example: melanoma cells (SOX10+ cells), CD8 cells indicating cytotoxic T-cells and other or undefined cells which have neither SOX10 or CD8 protein expression, but show up as cells according to the nuclear stain. In addition, PD1 and PD-L1 expression was scored in each sample and a threshold was determined in inForm that let's us distinguish between PD1+ and PD1- cells as well as PD-L1+/-. PD-L1 is only relevant in melanoma cells, whereas PD1 is relevant in the other two cell types. Since the toy example only shows the area of a very small image, the resulting statistics often contain NA value, because cell types are not present, which is especially the case in ROI analyses. Look at the full example datasets for more realistic examples.

```
require(IrisSpatialFeatures)
raw data <- new("ImageSet")</pre>
raw_data<- read_raw(raw_data,</pre>
                     raw_dir_name=system.file("extdata", package = "IrisSpatialFeatures"),
                     format='Mantra')
## [1] "Sample: MEL2"
## [1] "Sample: MEL3"
#apply all the thresholds PD1 for T and other cells, PD-L1 for macrophages and tumor cells
dataset <- threshold_dataset(raw_data,</pre>
                               marker='PD-Ligand-1 (Opal 690)',
                              marker_name='PDL1',
                              base=c('SOX10+'))
dataset <- threshold_dataset(dataset,</pre>
                               marker='PD-1 (Opal 540)',
                              marker_name='PD1',
                               base=c('CD8+','OTHER'))
```

#### Overview plots

Next we plot all the cell coordinates color coordinated in .pdf format

```
plot_dir <- file.path(tempdir(),'plots')
if (!file.exists(plot_dir)){
    dir.create(file.path(plot_dir))
}
p <- overview_plot(dataset,outdir=plot_dir,palette=NULL,type='png')</pre>
```



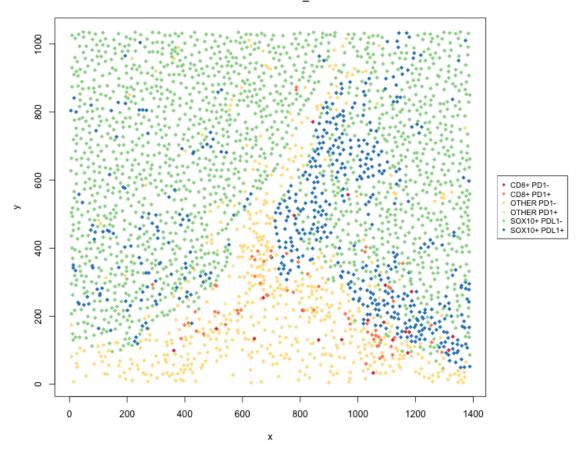


Figure 1: Example of an Overview plot

#### Extract counts and ratios

Here we extract counts per mm2 for each marker, both for each coordinate and collapsed across the multiple images per sample.

```
get_counts_per_mm2(dataset)
##
                MEL2
                                  MEL3
                             9.8" " 95 +/-
## CD8+ PD1-
                " 21 +/-
                                              36.5"
                " 327 +/- 189.5" " 147 +/-
## CD8+ PD1+
                                              29.5"
## OTHER PD1-
                "2235 +/- 373.4" "1475 +/-
                                              71.6"
## OTHER PD1+
                " 157 +/- 44.9" " 14 +/-
                                               8.4"
## SOX10+ PDL1- "3431 +/- 480.1" "3814 +/- 1009.4"
## SOX10+ PDL1+ " 303 +/- 36.5" "2186 +/-
                                             714.6"
get_counts_per_mm2_noncollapsed(dataset)
## $MEL2
##
            CD8+ PD1- CD8+ PD1+ OTHER PD1- OTHER PD1+ SOX10+ PDL1-
## 080416 2
             30.88565 516.6327
                                   2608.434
                                              202.1606
                                                           2950.984
## 080416 7
             11.23115 137.5815
                                   1861.562
                                              112.3115
                                                           3911.247
##
            SOX10+ PDL1+
## 080416 2
                266.7397
## 080416 7
                339.7422
##
## $MEL3
            CD8+ PD1- CD8+ PD1+ OTHER PD1- OTHER PD1+ SOX10+ PDL1-
## 120116_1 58.96352 176.8905
                                   1547.090
                                             22.462292
                                                           4823.777
## 120116_2 131.96596 117.9270
                                   1403.893
                                              5.615573
                                                           2804.979
##
            SOX10+ PDL1+
## 120116_1
                1471.280
## 120116_2
                2900.443
get_count_ratios(dataset, 'SOX10+ PDL1-', 'SOX10+ PDL1+')
##
             MEL2
                            MEL3
                                            <NA>
                                                           <NA>
## "11.06 +/- NA" "11.51 +/- NA" " 3.28 +/- NA" " 0.97 +/- NA"
```

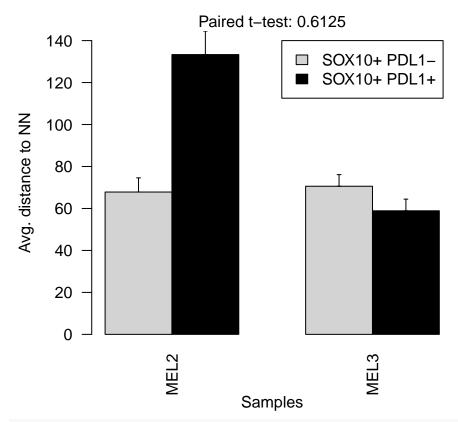
#### Nearest neighbor analysis

##

Next step we calculate the average nearest neighbor distances for each cell-type, plot barplots compare different distances and finally generate ray plots that show a visual representation of these distances for each coordinate.

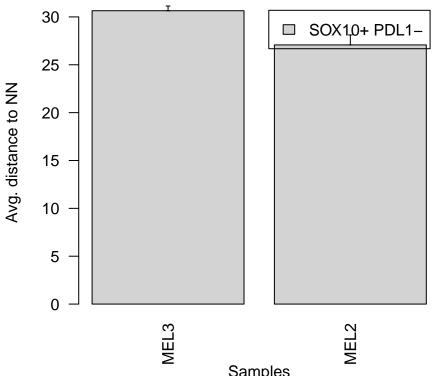
```
dataset <- extract_nearest_neighbor(dataset,min_num_cells=2)</pre>
get_nearest_neighbors(dataset, "SOX10+ PDL1+")
## $mean
##
                      MEL2
                                MEL3
## CD8+ PD1-
                 219.95758 177.05966
## CD8+ PD1+
                  47.26242 140.91845
## OTHER PD1-
                  28.21915 79.21432
## OTHER PD1+
                 108.54489 463.99940
## SOX10+ PDL1-
                 27.07400
                            30.64719
## SOX10+ PDL1+
                  0.00000
                             0.00000
```

## Distance from CD8+ PD1+ to SOX10+ PDL1 +/-



p <- plot\_nearest\_neighbor(dataset,'SOX10+ PDL1+','SOX10+ PDL1-')</pre>

## Distance from SOX10+ PDL1+ to SOX10+ PDL1-



##

##

## \$MEL3\$`120116\_2` ## null device

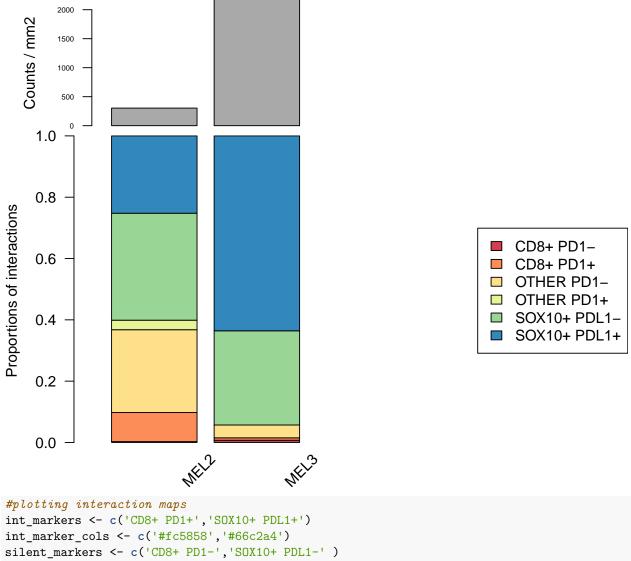
```
Samples
#ray plots for
plot_dir <- file.path(tempdir(),'ray_plots')</pre>
if (!file.exists(plot_dir)){
    dir.create(file.path(plot_dir))
neighbor_ray_plot(dataset,from_type='CD8+ PD1+',to_type='SOX10+ PDL1+',plot_dir=plot_dir,format ='pdf')
## $MEL2
## $MEL2$\`080416 2\`
## null device
##
##
## $MEL2$\`080416_7\`
## null device
##
##
## $MEL3
## $MEL3$`120116_1`
## null device
##
```

## Interaction analysis

Here we extract the interactions between different cell types, generate an interaction profile for SOX10+PDL1+ cells and also generate interaction maps for each coordinate showing the interactions between CD8+PD1+ cells and SOX10+ PD-L1+ cells.

```
dataset <- extract_interactions(dataset)</pre>
## MEL2 ... processing...
## 29 solved of 36 issues
## 11 solved of 13 issues
## MEL3 ... processing...
## 16 solved of 18 issues
## 15 solved of 19 issues
get_interactions(dataset,'SOX10+ PDL1+')
##
                        MEL2
                                    MEL3
## CD8+ PD1-
                0.002687426 0.007307972
## CD8+ PD1+
                0.095003831 0.007819282
## OTHER PD1-
                0.270041866 0.041832681
## OTHER PD1+
                0.031083077 0.000000000
## SOX10+ PDL1- 0.348876723 0.307243647
## SOX10+ PDL1+ 0.252307077 0.635796418
#plotting interaction summaries
p <- plot_interactions(dataset,'SOX10+ PDL1+',xlim_fix=4)</pre>
```

## Interactions with SOX10+ PDL1+



## Working on sample: MEL2
## Working on sample: MEL3

## Running the proximity analysis

Calculating the number of cells within 25 pixels distance for each cell and then showing the the profile for SOX10+ PDL1-

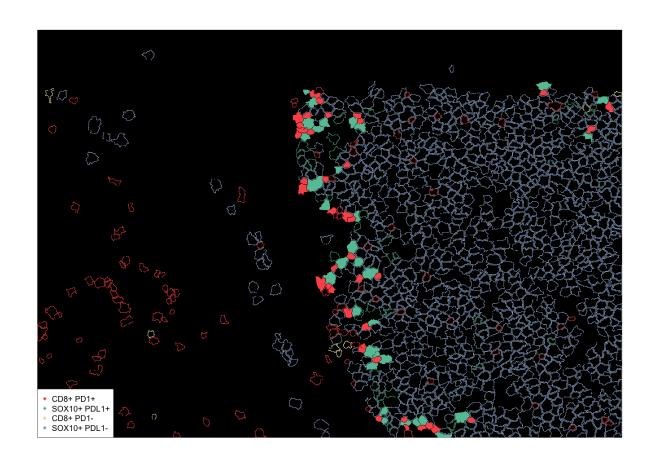


Figure 2: Example of a Interaction map

```
dataset <- extract_proximity(dataset,only_closest=TRUE,radii=25)

## MEL2 ... processing...

## MEL3 ... processing...

p <- plot_proximities(dataset, "SOX10+ PDL1-", xlim_fix=3)</pre>
```

## In proximity to SOX10+ PDL1-

