Single Cell Response Curve (SCRC) - R package $_{\mathrm{User\ Manual}}$

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Preliminaries

Requirements - Hardware

- A 32 or 64 bit processor (recommended: 64bit)
- 1GHz processor (recommended: multicore for a comprehensive analysis)
- 2GB MB RAM (recommended: 4GB+, depends on the size of experimental data)

Requirements - Software

The main software requirement is the installation of the R environment (version: >= 3.2), which can be downloaded from R project website and is distributed for all common operating systems. We tested the package in R environment installed on Windows 7, 10; Mac OS X 10.11 - 10.13 and Ubuntu 18.04 with no significant differences in the performance. The use of a dedicated Integrated development environment (IDE), e.g. RStudio is recommended.

Apart from a base installation of R, SLEMI requires the following R packages:

- 1. for installation
- devtools
- 2. for estimation
- nne
- doParallel (if parallel computation are needed)
- 3. for visualisation
- ggplot2
- ggthemes
- grDevices
- viridis
- 4. for data handling
- data.table
- reshape2
- dplyr
- foreach

Each of the above packages can be installed by executing

```
install.packages("name_of_a_package")
```

in the R console.

Importantly, during installation availability of the above packages will be verified and missing packages will be automatically installed.

Installation

The package can be directly installed from GitHub. For installation, open RStudio (or base R) and run following commands in the R console

```
install.packages("devtools") # run if 'devtools' is not installed
library(devtools)
install_github("sysbiosig/SCRC")
```

Are required packages not found, they will be installed automatically.

Citing and support

All problems, issues and bugs can be reported here:

https://github.com/sysbiosig/SCRC/issues

or directly via e-mail: karol.nienaltowski a t gmail.com.

Package functionalities

• Define SCRC and confusion matrix

Package structure

The SCRC package provides their functionalities with three main functions:

- 1. SCRC()- estimates the single cell dose-response curve and cell-to-cell heterogeneity structure.
- 2. plotConfusionMatrix() visualises the cell-to-cell heterogeneity structure as fractions of cells stimulated with one dose (rows) with responses typical to either of the doses (columns).
- 3. plotSCRCWaves() visualises single cell dose-response curve, line correspons do the fractions of cells that exhibit different response levels as the dose increases, whereas the bands representing the cell-to-cell heterogeneity structure.

Morevoer, package contains examplary datasets, that were used in the publication:

- 1. data.scrc.cytof
- 2. data.scrc.ps1
- 3. data.scrc.ps3
- 4. data.scrc.nfkb

Input data

The function SCRC() takes data in the form of the object data.frame with a specific structure of rows and columns. Responses y_j^i are assumed to be measured for a finite set of stimuli levels x_1, x_2, \ldots, x_m . The responses y_j^i can be multidimensional. Usually, experimental dataset is represented as a table with rows and columns organized as shown in Figure 1.

Data example

An example of the input data.frame, which contains the multivariate dose-responses to IFN-a2a in monocytes CD14+ CD16- presented in the MP is available within the package under the variable SCRC::data.scrc.cytof. It has the following format

Stim	pSTAT1	pSTAT3	pSTAT4	pSTAT5	pSTAT6
25	9.423005	0.0000000	0.000000	0.2226261	1.0420985
25	14.077361	0.0000000	0.000000	4.4516195	0.0000000
25	6.634479	2.3057582	0.000000	0.0000000	0.8014894
25	5.739593	0.0000000	3.807480	0.3100144	0.6536241
25	24.686699	0.0000000	0.000000	0.1629745	1.2835708
25	21.104772	0.3036974	1.324072	3.2628903	0.6757941

where each row represents measurements of a single-cell, the column named Stim specifies the dose level of IFN-a2a, while pSTAT1, pSTAT3, pSTAT4,pSTAT5,pSTAT6 are the normalized levels of phosporylated STATs in an individual cell. The above table can be shown in R by calling

input	output 1	output 2	output 3	
$n_1 \left\{ \begin{array}{c} x_1 \\ \vdots \\ x_1 \end{array} \right.$	$y_{1,1}^1 \\ \vdots \\ y_{n_1,1}^1$	$y_{1,2}^1 \\ \vdots \\ y_{n_1,2}^1$	$y_{1,m}^1$ \vdots $y_{n_1,m}^1$	
$n_2 \left\{ \begin{array}{c} x_2 \\ \vdots \\ x_2 \end{array} \right.$	$\begin{array}{c} y_{1,1}^2 \\ \vdots \\ y_{n_2,1}^2 \end{array}$	$y_{n_1,2}^1$ $y_{1,2}^2$ \vdots $y_{n_2,2}^2$	$y_{1,m}^2$ \vdots $y_{n_2,m}^2$	
÷	i :	i :	i :	•••
$n_m \left\{ \begin{array}{c} x_m \\ \vdots \\ x_m \end{array} \right.$	$\begin{array}{c c} y_{1,1}^m \\ \vdots \\ y_{n_m,1}^m \end{array}$	$y_{1,2}^m \\ \vdots \\ y_{n_m,2}^m$	$y_{1,m}^m \\ \vdots \\ y_{n_m,m}^m$	

Figure 1: Standard output graph presenting probabilities of correct discrimination between each pair of input values.

```
head(SCRC::data.scrc.cytof)
```

Esitmation of the SCRC

Then, main function is called as:

```
model <- SCRC::SCRC(
  data = data,
  signal = "input",
  response = c("output_1", "output_2", "output_3", ...),
  bootstrap.number = bootstrap.number,
  ...
)</pre>
```

Variables signal and response describes respectively dose level and single-cell responses. These columns should be of type numeric; order and number of outputs should be the same for all cells. Number of observations in data should large, possibly >100, per input value is required.

The variable bootstrap.number represents number of bootstrap samples required for estimation of cell-to-cell heterogeneity. It is crucial to choose this value carefully, as it induce estimator accuracy.

The result of the function is an object of class SCRCModel, that contains results of the estimator. To see the results call :

```
print(model)
```

To get the SCRC (cumulative frequency) and cell-to-cell heterogeneity (confusion matrix) please call model\$corfusion.matrix, respectively.

The result can be visualised using one of our plots, as it was presented in the publication.

SCRC can be plotted using function:

```
plotSCRCWaves(model)
```

The cell-to-cell heterogeneity can be plotted using function:

```
plotConfusingMatrix(model)
```

Example of usage

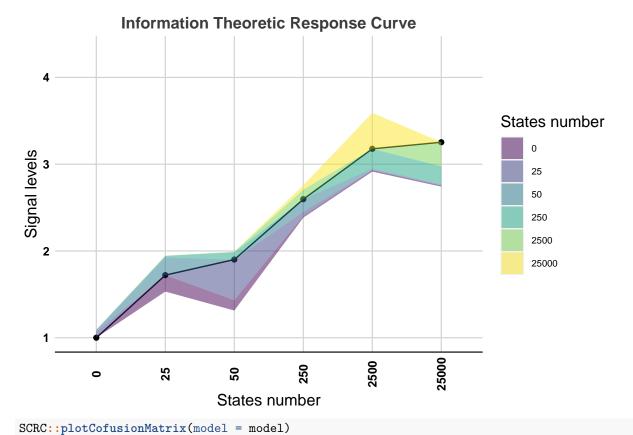
Below, we present an application of SCRC package to the case of the multivariate dose-responses to IFN-a2a in monocytes CD14+ CD16- described above.

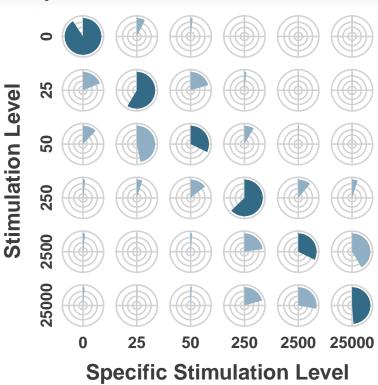
```
library(SCRC)
model <-
SCRC(
   data = SCRC::data.itrc.cytof,
   signal = "Stim",
   response = c("pSTAT1", "pSTAT3", "pSTAT4", "pSTAT5", "pSTAT6"),
   bootstrap.number = 64)</pre>
```

The result is called by:

```
print(model)
```

```
## SCRCModel
## formula : Stim ~ pSTAT1+pSTAT3+pSTAT4+pSTAT5+pSTAT6
## SCRC :
##
           25
                 50
                      250
                           2500 25000
## 1.00 1.72 1.90 2.60 3.18 3.25
## confusion matrix :
##
               25
                    50 250 2500 25000
## 0
        0.91 0.07 0.02 0.00 0.00 0.00
        0.19 0.59 0.20 0.02 0.00 0.00
## 25
        0.12 0.47 0.32 0.08 0.00 0.00
## 50
## 250
        0.02 0.05 0.14 0.63 0.11 0.05
## 2500 0.02 0.00 0.02 0.23 0.32 0.41
## 25000 0.01 0.00 0.01 0.21 0.28 0.49
SCRC::plotSCRCWaves(model = model)
```





Details of SCRC packgae functions

SCRC

Calculation of the SCRC and cell-to-cell heterogeneity with default settings is perfored by the command

```
model <- SCRC(
  data,
  signal = "signal",
  response = "response",
  sample = "sample",
  bootstrap.number = 0,
  bootstrap.sample_size = 1000,
  parallel_cores = 1,
  lr_maxit = 1000,
  MaxNWts = 5000,
  ...
)</pre>
```

The required arguments are:

- data a data.frame or data.table object in a wide format that describe response (might be multidimmensional) of the samples to the signal (now only one dimmensional); data.frame data consists columns of names defined by sample, signal (optional), and response; each row represents a response of one sample to the input signal; column signal define the input signal; columns response define the multidimmensional (optional) response to the input signal; column sample specify identifaction of sample; if sample is not defined then sample is identified by row number;
- signal character, specify name of the column that represents the input signal;
- response vector of characters, that specify names of the columns that represents the output response;
- sample character (optional), specify name of the column that consists identification of sample;
- parallel_cores specify number of cores used for computations, default = 1
- bootstrap.number (default = 1) numeric, bootstrap.number >= 1, specify nymber of bootstrap samples used for estimation SCRC and cell-to-cell heterogeneity. It is crucial to choose this value carefully, as it induce estimator accuracy. The proper value depends on data dimmensions and density distribution. The practice indicates that the higher number of bootstrap samples are required to obtain satisfying level of the accuracy of the cell-to-cell heterogeneity estimator. The bootstrap.number = 1 denotes that one bootstrap sampling is performed to guarantee equipotence between number of cells for each dose, that is assumed in method;
- bootstrap.sample size numeric, size of the bootstrap sample;
- lr_maxit (default = 1000) a maximum number of iterations of fitting step of logistic regression algorithm in nnet function. If a warning regarding lack of convergence of logistic model occurs, should be set to a larger value (possible if data is more complex or of a very high dimension);
- MaxNWts (default = 5000) a maximum number of parameters in logistic regression model. A limit is set to prevent accidental over-loading the memory. It should be set to a larger value in case of exceptionally high dimension of the output data or very high number of input values. In principle, logistic model requires fitting $(m-1) \cdot (d+1)$ parameters, where m is the number of unique input values and d is the dimension of the output.

The function returns a list with the following elements

plotSCRCWaves

```
plotSCRCWaves(
  model,
  title_ = "Information Theoretic Response Curve",
  xlab_ = "States number",
  ylab_ = "Signal levels",
  fill.guide_ = "legend",
  ylimits_ = TRUE,
  alpha_ = 0.5,
  getScaleY = getScaleY.SCRC,
  theme.signal = NULL,
  ...
)
```

The required arguments are:

- model SCRCModel object return by SCRC function
- title_ character, specify title of plot, default "Information Theoretic Response Curve"
- xlab_ character, label of x axes, default "States number"
- ylab_ character, label of y axes and legend title, default "Signal levels"
- fill.guide logical, specify if legend should be displayed
- ylimits logical (TRUE or FALSE) or vector of minimum and maximum of y axes,
- theme.signal optional, object returned by GetRescaledSignalTheme

... Arguments passed on to rescaleSignalsValues

rescale.fun parameter, that defines a function used for rescaling signals in plots. There are three built-in functions, that can be chosen: (1) 'factor' - signals treated as factors (default) with levels defined in list rescale.fun.args, (2) 'numeric', (3) logarithmic - with base defined in rescale.fun.args - default: $e = \exp(1)$. Function must be defined as a lambda construct function(x, ...).

rescale.fun.args list of the arguments to defaults rescale.fun

plotConfusionMatrix