## Package 'nanostringr'

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Type Package

**Title** Performs Quality Control, Data Normalization, and Batch Effect Correction for 'NanoString nCounter' Data

Version 0.1.4 Date 2019-04-23

Description Provides quality control (QC), normalization, and batch effect correction operations for 'NanoString nCounter' data, Talhouk et al. (2016) <doi:10.1371/journal.pone.0153844>. Various metrics are used to determine which samples passed or failed QC. Gene expression should first be normalized to housekeeping genes, before a reference-based approach is used to adjust for batch effects. Raw NanoString data can be imported in the form of Reporter Code Count (RCC) files.

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URL https://github.com/OVCARE/nanostringr,
 https://ovcare.github.io/nanostringr

BugReports https://github.com/OVCARE/nanostringr/issues

**Depends** R (>= 3.5.0)

Imports assertthat, ccaPP, dplyr, epiR, magrittr, purrr, rlang

Suggests covr, knitr, rmarkdown, testthat

VignetteBuilder knitr

**Encoding** UTF-8

LazyData TRUE

RoxygenNote 6.1.1

NeedsCompilation no

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Concordance Correlation Plot

## Description

Plotting function for reliability measure.

## Usage

```
CCplot(method1, method2, Ptype = "None", metrics = FALSE,
    xlabel = "", ylabel = "", title = "", subtitle = NULL,
    xrange = NULL, yrange = NULL, MArange = c(-3.5, 5.5))
```

## Arguments

method1	measurements obtained in batch 1 or using method 1
method2	measurements obtained in batch 2 or using method 2
Ptype	type of plot to be outputted c("scatter", "MAplot")
metrics	if TRUE, prints Rc, Ca, and R2 to console
xlabel	x-axis label for scatterplot
ylabel	y-axis label for scatterplot
title	title for the main plot
subtitle	subtitle of plot
xrange	range of x axis
yrange	range of y axis
MArange	MA range

## Value

Either a scatterplot or MA plot showing concordance correlation.

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#### Author(s)

Aline Talhouk Aline Talhouk

#### **Examples**

```
# Simulate normally distributed data
set.seed(12)
a1 < - rnorm(20) + 2
a2 <- a1 + rnorm(20, 0, 0.15)
a3 < -a1 + rnorm(20, 0, 0.15) + 1.4
a4 <- 1.5 * a1 + rnorm(20, 0, 0.15)
a5 < -1.3 * a1 + rnorm(20, 0, 0.15) + 1
a6 <- a1 + rnorm(20, 0, 0.8)
# One scatterplot
CCplot(a1, a2, Ptype = "scatter")
m2 <- list(a1, a2, a3, a4, a5, a6)
mains <- c("Perfect Agreement", "Very Good Agreement", "Location Shift",</pre>
           "Scale Shift", "Location and Scale Shift", "Measurement Error")
subs <- letters[1:6]</pre>
par(mfrow = c(3, 2), mar = c(5.1, 4.1, 1.5, 1.5))
# Scatterplots
mapply(function(y, t, s)
  CCplot(method1 = a1, method2 = y, Ptype = "scatter",
         xlabel = "X", ylabel = "Y", title = t, subtitle = s),
  y = m2, t = mains, s = subs)
# MAplots and show metrics
mapply(function(y, t, s)
  CCplot(method1 = a1, method2 = y, Ptype = "MAplot",
         title = t, subtitle = s, metrics = TRUE),
  y = m2, t = mains, s = subs)
```

cohort

NanoString Experiment Cohorts

## **Description**

There were five different cohorts used in NanoString experiments.

#### Usage

hld.r

ovd.r

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ovc.r

hlo.r

ovo.r

#### **Format**

- hld.r Hodgkin Lymphoma Clinical Samples: a data frame with 232 rows and 77 columns
- ovd.r Ovarian Cancer Clinical Samples: a data frame with 133 rows and 261 columns
- ovc.r Ovarian Cancer Cell Lines: a data frame with 133 rows and 29 columns
- hlo.r DNA Oligonucleotides for the HL CodeSet: a data frame with 40 rows and 71 columns
- ovo.r DNA Oligonucleotides for the OC CodeSet: a data frame with 133 rows and 138 columns

#### **Details**

Each data object contains raw expression counts, so no normalization has been applied. The format is a data frame with genes as rows, samples as columns. Note that the first three columns contain gene metadata and are always labelled "Code.Class", "Name", and "Accession", and the rest are sample names. Hence, for the hld.r data, the raw counts are contained in 232 genes for 77 - 3 = 74 samples. The total number of samples is 74 + 258 + 26 + 68 + 135 = 561, which matches the number of rows in expQC, the expression QC data.

#### Source

See Table 1 of https://journals.plos.org/plosone/article?id=10.1371/journal.pone. 0153844 for details.

## See Also

expQC

expQC

Expression QC data

#### **Description**

Quality control metrics for the five cohorts analyzed in NanoString experiments.

#### **Format**

A data frame with 561 rows and 23 columns.

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## **Details**

The total number of samples from the five cohorts is 561.

#### See Also

cohort

НК	norm	Normalization to Housekeeping Genes

## Description

Normalizes the gene expression of NanoString nCounter data to housekeeping genes. This is done by subtracting the average log housekeeping gene expression from the expression level of every gene in each sample.

#### Usage

```
HKnorm(raw, is.logged = FALSE, corr = 1e-04)
```

#### **Arguments**

raw data frame of raw counts obtained from nCounter (rows represent genes, columns

represent samples). The first three columns must be labeled: c("Code.Class", "Name", "Accession")

and contain that information.

is.logged logical; If TRUE, normalization has already been done on log base 2 scale, no

need log the data

corr small correction to avoid error

#### Value

data frame of log normalized data in the same format but without reference genes

#### Author(s)

Aline Talhouk, Derek Chiu

#### **Examples**

```
HKnorm(ovd.r)
HKnorm(ovd.r, is.logged = TRUE)
```

NanoStringQC

NanoStringQC	QC metrics for NanoString Data	

## Description

Computes and returns NanoString quality control metrics and flags.

## Usage

```
NanoStringQC(raw, exp, detect = 80, sn = 150)
```

## Arguments

raw	data frame of raw counts obtained from nCounter (rows represent genes, columns represent samples). The first three columns must be labeled: c("Code.Class", "Name", "Accession") and contain that information.
exp	data frame of annotations with rows in the same order as the columns of raw. Requires a column labeled "File.Name" with entries corresponding to sample names in raw, also needs columns c("fov.counted", "fov.count", "binding.density"). These fields can be extracted from the nanostring RCC files.
detect	threshold of percentage of genes expressed over limit of detection (LOD) that we would like to detect (not decimal), defaults to 80 percent.
sn	signal to noise ratio of the housekeeping genes we are willing to tolerate, defaults to 150.

## Value

data frame of annotations updated with normalization parameters

## Author(s)

Aline Talhouk, Derek Chiu

## **Examples**

```
exp.OVD <- subset(expQC, OVD == "Yes")
expOVD <- NanoStringQC(ovd.r, exp.OVD)</pre>
```

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rcc

Read NanoString RCC files

## **Description**

Read RCC files and extract count and attribute data. Use read\_rcc() for multiple files, and use the parse\_\*() functions for single files.

## Usage

```
read_rcc(path = ".")
parse_counts(file)
parse_attributes(file)
```

#### **Arguments**

path directory path for multiple RCC files

file RCC file name

#### **Details**

RCC files for a sample are direct outputs from NanoString runs. We can extract counts for each gene in a sample. Sample attributes include sample ID, GeneRLF, date, cartridge ID, lane number, Fov count, Fov counted, and binding density. read\_rcc() merges both count and attribute data across samples.

If path points to a zipped RCC file with multiple samples, the zip file is uncompressed and a directory of RCC sample files is created with the same name. Only file extensions ".RCC" or ".rcc" are allowed.

#### Value

read\_rcc() reads in a directory of RCC files and outputs a list with two elements:

- raw: A tibble of parsed counts for multiple RCC files created by calling parse\_counts() on each sample. Columns include "Code.Class", "Name", "Accession", and a column for each sample ID. There is one row per gene.
- exp: A tibble of parsed attributes for multiple RCC files created by calling parse\_attributes() on each sample. Columns include "File.Name" (sample ID), "geneRLF", "nanostring.date", "cartridgeID", "lane.number", fov.count", "fov.counted", "binding.density". There is one row per sample.

parse\_counts() reads a single RCC file and returns a tibble of parsed counts.

parse\_attributes() reads a single RCC file and returns a list of parsed attributes.

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#### Author(s)

Derek Chiu

#### **Examples**

```
rcc_file <- system.file("extdata", "example.RCC", package = "nanostringr")
parse_counts(rcc_file)
parse_attributes(rcc_file)</pre>
```

refMethod

Reference-based approach for batch adjustment

## **Description**

Batch adjustment by considering a measure relative to a reference sample

## Usage

```
refMethod(Y, R1, R2)
```

## Arguments

in the same batch as Y, if calibrating one batch to the other Y represents the data from batch 2 and R1 would be reference run in batch 1 and R2 would be reference from batch 2
R1 reference data run in the first batch
R2 reference data run in the second batch

### Value

The Y data adjusted calibrated to batch 1 (if two batches are presented) or the data with reference sample expression removed if only one data is provided

#### Author(s)

Aline Talhouk

### **Examples**

```
set.seed(12)
A <- matrix(rnorm(120), ncol = 10)
B <- matrix(rnorm(80), ncol = 10)
C <- matrix(rnorm(50), ncol = 10)
refMethod(A, B, C)</pre>
```

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