

Package ‘imcdatasets’

December 1, 2022

Version 1.6.0

Date 2022-10-21

Title Collection of publicly available imaging mass cytometry (IMC) datasets

Description The imcdatasets package provides access to publicly available IMC datasets. IMC is a technology that enables measurement of > 40 proteins from tissue sections. The generated images can be segmented to extract single cell data. Datasets typically consist of three elements: a SingleCellExperiment object containing single cell data, a CytoImageList object containing multichannel images and a CytoImageList object containing the cell masks that were used to extract the single cell data from the images.

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NeedsCompilation no

Depends R (>= 4.2), SingleCellExperiment, SpatialExperiment, cytomapper,

Imports methods, utils, ExperimentHub, S4Vectors, DelayedArray, HDF5Array

Suggests BiocStyle, knitr, rmarkdown, markdown, testthat

biocViews ExperimentData, ExperimentHub, SingleCellData, SpatialData, Homo_sapiens_Data, ImmunoOncologyData, TechnologyData, PackageTypeData, ReproducibleResearch, Tissue

VignetteBuilder knitr

URL <https://github.com/BodenmillerGroup/imcdatasets>

BugReports <https://github.com/BodenmillerGroup/imcdatasets/issues>

BiocType ExperimentData

RoxygenNote 7.2.1

Encoding UTF-8

git_url <https://git.bioconductor.org/packages/imcdatasets>

git_branch RELEASE_3_16

git_last_commit 525cfdc

git_last_commit_date 2022-11-01

Date/Publication 2022-12-01

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DamondPancreas2019-defunct

Defunct function - 'DamondPancreas2019' dataset

Description

These functions are defunct and no longer available.

Details

Defunct functions are: DamondPancreas2019_sce, DamondPancreas2019_masks, and DamondPancreas2019_images.
 Please use the Damond_2019_Pancreas2019 function instead.

`DamondPancreas2019Data`*Obtain the damond-pancreas-2019 dataset*

Description

This function is provided for compatibility with older versions but is deprecated. As a replacement, please use `Damond_2019_Pancreas`. Obtain the damond-pancreas-2019 dataset, which consists of three data objects: single cell data, multichannel images and cell segmentation masks. The data was obtained by imaging mass cytometry of human pancreas sections from donors with type 1 diabetes.

Usage

```
DamondPancreas2019Data(  
  data_type = c("sce", "images", "masks"),  
  metadata = FALSE,  
  on_disk = FALSE,  
  h5FilePath = NULL,  
  force = FALSE  
)
```

Arguments

<code>data_type</code>	type of object to load, should be 'sce' for single cell data, 'images' for multi-channel images or 'masks' for cell segmentation masks.
<code>metadata</code>	if FALSE (default), the data object selected in <code>data_type</code> is returned. If TRUE, only the metadata associated to this object is returned.
<code>on_disk</code>	logical indicating if images in form of HDF5Array objects (as .h5 files) should be stored on disk rather than in memory. This setting is valid when downloading images and masks.
<code>h5FilePath</code>	path to where the .h5 files for on disk representation are stored. This path needs to be defined when <code>on_disk = TRUE</code> . When files should only temporarily be stored on disk, please set <code>h5FilePath = getHDF5DumpDir()</code>
<code>force</code>	logical indicating if images should be overwritten when files with the same name already exist on disk.

Details

This function is provided for compatibility with older versions but is deprecated. As a replacement, please use `Damond_2019_Pancreas`. This is an Imaging Mass Cytometry (IMC) dataset from Damond et al. (2019), consisting of three data objects:

- `images` contains a hundred 38-channel images in the form of a [CytoImageList](#) class object.
- `masks` contains the cell segmentation masks associated with the images, in the form of a [CytoImageList](#) class object.

- `sce` contains the single cell data extracted from the multichannel images using the cell segmentation masks, as well as the associated metadata, in the form of a [SingleCellExperiment](#). This represents a total of 252,059 cells x 38 channels.

All data are downloaded from ExperimentHub and cached for local re-use.

Mapping between the three data objects is performed via variables located in their metadata columns: `mcols()` for the [CytoImageList](#) objects and `colData()` for the [SingleCellExperiment](#) object. Mapping at the image level can be performed with the `ImageName` or `ImageNumber` variables. Mapping between cell segmentation masks and single cell data is performed with the `CellNumber` variable, the values of which correspond to the intensity values of the `DamondPancreas2019_masks` object. For practical examples, please refer to the "Accessing IMC datasets" vignette.

This dataset is a subset of the complete Damond et al. (2019) dataset comprising the data from three pancreas donors at different stages of type 1 diabetes (T1D). The three donors present clearly diverging characteristics in terms of cell type composition and cell-cell interactions, which makes this dataset ideal for benchmarking spatial and neighborhood analysis algorithms.

The assay slot of the [SingleCellExperiment](#) object contains two assays:

- `counts` contains mean ion counts per cell.
- `exprs` contains arsinh-transformed counts, with cofactor 1.

The marker-associated metadata, including antibody information and metal tags are stored in the `rowData` of the [SingleCellExperiment](#) object.

The cell-associated metadata are stored in the `colData` of the [SingleCellExperiment](#) object. These metadata include cell types (in `colData(sce)$CellType`) and broader cell categories, such as "immune" or "islet" cells (in `colData(sce)$CellCat`). In addition, for cells located inside pancreatic islets, the islet they belong to is indicated in `colData(sce)$ParentIslet`. For cells not located in islets, the "ParentIslet" value is set to 0 but the spatially closest islet can be identified with `colData(sce)$ClosestIslet`.

The donor-associated metadata are also stored in the `colData` of the [SingleCellExperiment](#) object. For instance, the donors' IDs can be retrieved with `colData(sce)$case` and the donors' disease stage can be obtained with `colData(sce)$stage`.

The three donors present the following characteristics:

- 6126 is a non-diabetic donor, with large islets containing many beta cells, severe infiltration of the exocrine pancreas with myeloid cells but limited infiltration of islets.
- 6414 is a donor with recent T1D onset (shortly after diagnosis) showing partial beta cell destruction and mild infiltration of islets with T cells.
- 6180 is a donor with long-duration T1D (11 years after diagnosis), showing near-total beta cell destruction and limited immune cell infiltration in both the islets and the pancreas.

File sizes:

- ``images``: size in memory = 7.4 Gb, size on disk = 1.7 Gb.
- ``masks``: size in memory = 200 Mb, size on disk = 8.2 Mb.
- ``sce``: size in memory = 352 Mb, size on disk = 212 Mb.

When storing images on disk, these need to be first fully read into memory before writing them to disk. This means the process of downloading the data is slower than directly keeping them in memory. However, downstream analysis will lose its memory overhead when storing images on disk.

Original source: Damond et al. (2019): <https://doi.org/10.1016/j.cmet.2018.11.014>

Original link to raw data, also containing the entire dataset: <https://data.mendeley.com/datasets/cydmwfsztj/2>

Value

A [SingleCellExperiment](#) object with single cell data, a [CytoImageList](#) object containing multichannel images, or a [CytoImageList](#) object containing cell masks.

Author(s)

Nicolas Damond

References

Damond N et al. (2019). A Map of Human Type 1 Diabetes Progression by Imaging Mass Cytometry. *Cell Metab* 29(3), 755-768.

Examples

```
# Load single cell data
sce <- DamondPancreas2019Data(data_type = "sce")
print(sce)

# Display metadata
DamondPancreas2019Data(data_type = "sce", metadata = TRUE)

# Load masks on disk
library(HDF5Array)
masks <- DamondPancreas2019Data(data_type = "masks", on_disk = TRUE,
h5FilePath = getHDF5DumpDir())
print(head(masks))
```

Damond_2019_Pancreas *Obtain the Damond_2019_Pancreas dataset*

Description

Obtain the Damond_2019_Pancreas dataset, which consists of three data objects: single cell data, multichannel images and cell segmentation masks. The data was obtained by imaging mass cytometry (IMC) of human pancreas sections from donors with type 1 diabetes.

Usage

```
Damond_2019_Pancreas(
  data_type = c("sce", "spe", "images", "masks"),
  metadata = FALSE,
  on_disk = FALSE,
  h5FilePath = NULL,
  version = "latest",
  force = FALSE
)
```

Arguments

<code>data_type</code>	type of object to load, 'images' for multichannel images or 'masks' for cell segmentation masks. Single cell data are retrieved using either 'sce' for the <code>SingleCellExperiment</code> format or 'spe' for the <code>SpatialExperiment</code> format.
<code>metadata</code>	if FALSE (default), the data object selected in <code>data_type</code> is returned. If TRUE, only the metadata associated to this object is returned.
<code>on_disk</code>	logical indicating if images in form of <code>HDF5Array</code> objects (as .h5 files) should be stored on disk rather than in memory. This setting is valid when downloading images and masks.
<code>h5FilePath</code>	path to where the .h5 files for on disk representation are stored. This path needs to be defined when <code>on_disk = TRUE</code> . When files should only temporarily be stored on disk, please set <code>h5FilePath = getHDF5DumpDir()</code> .
<code>version</code>	dataset version. By default, the latest version is returned.
<code>force</code>	logical indicating if images should be overwritten when files with the same name already exist on disk.

Details

This is an Imaging Mass Cytometry (IMC) dataset from Damond et al. (2019), consisting of three data objects:

- `images` contains a hundred 38-channel images in the form of a `CytoImageList` class object.
- `masks` contains the cell segmentation masks associated with the images, in the form of a `CytoImageList` class object.
- `sce` contains the single cell data extracted from the multichannel images using the cell segmentation masks, as well as the associated metadata, in the form of a `SingleCellExperiment`. This represents a total of 252,059 cells x 38 channels.
- `spe` same single cell data as for `sce`, but in the `SpatialExperiment` format.

All data are downloaded from ExperimentHub and cached for local re-use.

Mapping between the three data objects is performed via variables located in their metadata columns: `mcols()` for the `CytoImageList` objects and `ColData()` for the `SingleCellExperiment` and `SpatialExperiment` objects. Mapping at the image level can be performed with the `image_name` or `image_number` variables. Mapping between cell segmentation masks and single cell data is performed with the `cell_number` variable, the values of which correspond to the intensity values of the masks object. For practical examples, please refer to the "Accessing IMC datasets" vignette.

This dataset is a subset of the complete Damond et al. (2019) dataset comprising the data from three pancreas donors at different stages of type 1 diabetes (T1D). The three donors present clearly diverging characteristics in terms of cell type composition and cell-cell interactions, which makes this dataset ideal for benchmarking spatial and neighborhood analysis algorithms.

The assay slots of the [SingleCellExperiment](#) and [SpatialExperiment](#) objects contain three assays:

- `counts` contains raw mean ion counts per cell.
- `exprs` contains arsinh-transformed counts, with cofactor 1.
- `quant_norm` contains counts censored at the 99th percentile and scaled 0-1.

The marker-associated metadata, including antibody information and metal tags are stored in the `rowData` of the [SingleCellExperiment](#) / [SpatialExperiment](#) objects.

The cell-associated metadata are stored in the `colData` of the [SingleCellExperiment](#) and [SpatialExperiment](#) objects. These metadata include cell types (in `colData(sce)$cell_type`) and broader cell categories, such as "immune" or "islet" cells (in `colData(sce)$cell_category`). In addition, for cells located inside pancreatic islets, the islet they belong to is indicated in `colData(sce)$islet_parent`. For cells not located in islets, the "islet_parent" value is set to 0 but the spatially closest islet can be identified with `colData(sce)$islet_closest`.

The donor-associated metadata are also stored in the `colData` of the [SingleCellExperiment](#) and [SpatialExperiment](#) objects. For instance, the donors' IDs can be retrieved with `colData(sce)$patient_id` and the donors' disease stage can be obtained with `colData(sce)$patient_stage`.

Neighborhood information, defined here as cells that are localized next to each other, is stored as a `SelfHits` object in the `colPairs` slot of the [SingleCellExperiment](#) and [SpatialExperiment](#) objects.

The three donors present the following characteristics:

- 6126 is a non-diabetic donor, with large islets containing many beta cells, severe infiltration of the exocrine pancreas with myeloid cells but limited infiltration of islets.
- 6414 is a donor with recent T1D onset (shortly after diagnosis) showing partial beta cell destruction and mild infiltration of islets with T cells.
- 6180 is a donor with long-duration T1D (11 years after diagnosis), showing near-total beta cell destruction and limited immune cell infiltration in both the islets and the pancreas.

Dataset versions: a `version` argument can be passed to the function to specify which dataset version should be retrieved.

- ``v0``: original version (Bioconductor <= 3.15).
- ``v1``: consistent object formatting across datasets.

File sizes:

- ``images``: size in memory = 7.4 Gb, size on disk = 1.7 Gb.
- ``masks``: size in memory = 200 Mb, size on disk = 8.2 Mb.
- ``sce``: size in memory = 352 Mb, size on disk = 212 Mb.
- ``spe``: size in memory = 371 Mb, size on disk = 212 Mb.

When storing images on disk, these need to be first fully read into memory before writing them to disk. This means the process of downloading the data is slower than directly keeping them in memory. However, downstream analysis will lose its memory overhead when storing images on disk.

Original source: Damond et al. (2019): <https://doi.org/10.1016/j.cmet.2018.11.014>

Original link to raw data, also containing the entire dataset: <https://data.mendeley.com/datasets/cydmwfsztj/2>

Value

A `SingleCellExperiment` object with single cell data, a `SpatialExperiment` object with single cell data, a `CytoImageList` object containing multichannel images, or a `CytoImageList` object containing cell segmentation masks.

Author(s)

Nicolas Damond

References

Damond N et al. (2019). A Map of Human Type 1 Diabetes Progression by Imaging Mass Cytometry. *Cell Metab* 29(3), 755-768.

Examples

```
# Load single cell data
sce <- Damond_2019_Pancreas(data_type = "sce")
print(sce)

# Display metadata
Damond_2019_Pancreas(data_type = "sce", metadata = TRUE)

# Load masks on disk
library(HDF5Array)
masks <- Damond_2019_Pancreas(data_type = "masks", on_disk = TRUE,
h5FilePath = getHDF5DumpDir())
print(head(masks))
```

JacksonFischer2020-defunct

Defunct function - 'JacksonFischer2020' dataset

Description

These functions are defunct and no longer available.

Details

Defunct functions are: `JacksonFischer2020_sce`, `JacksonFischer2020_masks`, and `JacksonFischer2020_images`. Please use the `JacksonFischer_2020_BreastCancer` function instead.

JacksonFischer2020Data

Obtain the jackson-fischer-2020 dataset

Description

This function is provided for compatibility with older versions but is deprecated. As a replacement, please use `JacksonFischer_2020_BreastCancer`. Obtain the `jackson-fischer-2020` dataset, which consists of three data objects: single cell data, multichannel images and cell segmentation masks. The data was obtained by imaging mass cytometry of tumour tissue from patients with breast cancer.

Usage

```
JacksonFischer2020Data(
  data_type = c("sce", "images", "masks"),
  metadata = FALSE,
  on_disk = FALSE,
  h5FilePath = NULL,
  force = FALSE
)
```

Arguments

<code>data_type</code>	type of object to load, should be ‘sce’ for single cell data, ‘images’ for multi-channel images or ‘masks’ for cell segmentation masks.
<code>metadata</code>	if FALSE (default), the data object selected in <code>data_type</code> is returned. If TRUE, only the metadata associated to this object is returned.
<code>on_disk</code>	logical indicating if images in form of HDF5Array objects (as .h5 files) should be stored on disk rather than in memory. This setting is valid when downloading images and masks.
<code>h5FilePath</code>	path to where the .h5 files for on disk representation are stored. This path needs to be defined when <code>on_disk = TRUE</code> . When files should only temporarily be stored on disk, please set <code>h5FilePath = getHDF5DumpDir()</code>
<code>force</code>	logical indicating if images should be overwritten when files with the same name already exist on disk.

Details

This function is provided for compatibility with older versions but is deprecated. As a replacement, please use `JacksonFischer_2020_BreastCancer`. This is an Imaging Mass Cytometry (IMC) dataset from Jackson, Fischer et al. (2020), consisting of three data objects:

- `images` contains a hundred 42-channel images in the form of a `CytoImageList` class object.
- `masks` contains the cell segmentation masks associated with the images, in the form of a `CytoImageList` class object.
- `sce` contains the single cell data extracted from the multichannel images using the cell segmentation masks, as well as the associated metadata, in the form of a `SingleCellExperiment`. This represents a total of 285,851 cells x 42 channels.

All data are downloaded from ExperimentHub and cached for local re-use.

Mapping between the three data objects is performed via variables located in their metadata columns: `mcols()` for the `CytoImageList` objects and `colData()` for the `SingleCellExperiment` object. Mapping at the image level can be performed with the `ImageNb` variable. Mapping between cell segmentation masks and single cell data is performed with the `CellNb` variable, the values of which correspond to the intensity values of the `JacksonFischer2020_masks` object. For practical examples, please refer to the "Accessing IMC datasets" vignette.

This dataset is a subset of the complete Jackson, Fischer et al. (2020) dataset comprising the data from tumour tissue from 100 patients with breast cancer (one image per patient).

The assay slot of the `SingleCellExperiment` object contains three assays:

- `counts` contains mean ion counts per cell.
- `exprs` contains arsinh-transformed counts, with cofactor 1.
- `quant_norm` contains quantile-normalized counts (0 to 1, 99th percentile).

The marker-associated metadata, including antibody information and metal tags are stored in the `rowData` of the `SingleCellExperiment` object.

The cell-associated metadata are stored in the `colData` of the `SingleCellExperiment` object. These metadata include clusters (in `colData(sce)$PhenoGraphBasel`) and metaclusters (in `colData(sce)$metacluster`), as well as spatial information (e.g., cell areas are stored in `colData(sce)$Area`).

The patient-associated clinical data are also stored in the `colData` of the `SingleCellExperiment` object. For instance, the tumor grades can be retrieved with `colData(sce)$grade`.

File sizes:

- `images``: size in memory = 17.8 Gb, size on disk = 2.0 Gb.
- `masks``: size in memory = 433 Mb, size on disk = 10 Mb.
- `sce``: size in memory = 477 Mb, size on disk = 266 Mb.

When storing images on disk, these need to be first fully read into memory before writing them to disk. This means the process of downloading the data is slower than directly keeping them in memory. However, downstream analysis will lose its memory overhead when storing images on disk.

Original source: Jackson, Fischer et al. (2020): <https://doi.org/10.1038/s41586-019-1876-x>

Original link to raw data, containing the entire dataset: <https://doi.org/10.5281/zenodo.3518284>

Value

A `SingleCellExperiment` object with single cell data, a `CytoImageList` object containing multichannel images, or a `CytoImageList` object containing cell masks.

Author(s)

Jana Fischer

References

Jackson, Fischer et al. (2020). The single-cell pathology landscape of breast cancer. *Nature* 578(7796), 615-620.

Examples

```
# Load single cell data
sce <- JacksonFischer2020Data(data_type = "sce")
print(sce)

# Display metadata
JacksonFischer2020Data(data_type = "sce", metadata = TRUE)

# Load masks on disk
library(HDF5Array)
masks <- JacksonFischer2020Data(data_type = "masks", on_disk = TRUE,
h5FilePath = getHDF5DumpDir())
print(head(masks))
```

JacksonFischer_2020_BreastCancer

Obtain the JacksonFischer_2020_BreastCancer dataset

Description

Obtain the JacksonFischer_2020_BreastCancer dataset, which consists of three data objects: single cell data, multichannel images and cell segmentation masks. The data was obtained by imaging mass cytometry (IMC) of tumour tissue from patients with breast cancer.

Usage

```
JacksonFischer_2020_BreastCancer(
  data_type = c("sce", "spe", "images", "masks"),
  metadata = FALSE,
  on_disk = FALSE,
  h5FilePath = NULL,
  version = "latest",
  force = FALSE
)
```

Arguments

data_type	type of object to load, 'images' for multichannel images or 'masks' for cell segmentation masks. Single cell data are retrieved using either 'sce' for the SingleCellExperiment format or 'spe' for the SpatialExperiment format.
metadata	if FALSE (default), the data object selected in data_type is returned. If TRUE, only the metadata associated to this object is returned.
on_disk	logical indicating if images in form of HDF5Array objects (as .h5 files) should be stored on disk rather than in memory. This setting is valid when downloading images and masks.
h5FilePath	path to where the .h5 files for on disk representation are stored. This path needs to be defined when on_disk = TRUE. When files should only temporarily be stored on disk, please set h5FilePath = getHDF5DumpDir().
version	dataset version. By default, the latest version is returned.
force	logical indicating if images should be overwritten when files with the same name already exist on disk.

Details

This is an Imaging Mass Cytometry (IMC) dataset from Jackson, Fischer et al. (2020), consisting of three data objects:

- images contains a hundred 42-channel images in the form of a [CytoImageList](#) class object.
- masks contains the cell segmentation masks associated with the images, in the form of a [CytoImageList](#) class object.
- sce contains the single cell data extracted from the multichannel images using the cell segmentation masks, as well as the associated metadata, in the form of a [SingleCellExperiment](#). This represents a total of 285,851 cells x 42 channels.
- spe same single cell data as for sce, but in the [SpatialExperiment](#) format.

All data are downloaded from ExperimentHub and cached for local re-use.

Mapping between the three data objects is performed via variables located in their metadata columns: `mcols()` for the [CytoImageList](#) objects and `colData()` for the [SingleCellExperiment](#) and [SpatialExperiment](#) objects. Mapping at the image level can be performed with the `image_name` variable. Mapping between cell segmentation masks and single cell data is performed with the `cell_number` variable, the values of which correspond to the intensity values of the masks object. For practical examples, please refer to the "Accessing IMC datasets" vignette.

This dataset is a subset of the complete Jackson, Fischer et al. (2020) dataset comprising the data from tumour tissue from 100 patients with breast cancer (one image per patient).

The assay slot of the [SingleCellExperiment](#) object contains three assays:

- counts contains mean ion counts per cell.
- exprs contains arsinh-transformed counts, with cofactor 1.
- quant_norm contains quantile-normalized counts (0 to 1, 99th percentile).

The marker-associated metadata, including antibody information and metal tags are stored in the `rowData` of the `SingleCellExperiment` and `SpatialExperiment` objects.

The cell-associated metadata are stored in the `colData` of the `SingleCellExperiment` and `SpatialExperiment` objects. These metadata include clusters (in `colData(sce)$cell_cluster_phenograph`) and metaclusters (in `colData(sce)$cell_metacluster`), as well as spatial information (e.g., cell areas are stored in `colData(sce)$cell_area`).

The clinical data are also stored in the `colData` of the `SingleCellExperiment` and `SpatialExperiment` objects. For instance, the tumor grades can be retrieved with `colData(sce)$tumor_grade`.

Dataset versions: a `version` argument can be passed to the function to specify which dataset version should be retrieved.

- `v0`: original version (Bioconductor <= 3.15).
- `v1`: consistent object formatting across datasets.

File sizes:

- `images`: size in memory = 17.8 Gb, size on disk = 2.0 Gb.
- `masks`: size in memory = 433 Mb, size on disk = 10 Mb.
- `sce`: size in memory = 477 Mb, size on disk = 266 Mb.
- `spe`: size in memory = 496 Mb, size on disk = 267 Mb.

When storing images on disk, these need to be first fully read into memory before writing them to disk. This means the process of downloading the data is slower than directly keeping them in memory. However, downstream analysis will lose its memory overhead when storing images on disk.

Original source: Jackson, Fischer et al. (2020): <https://doi.org/10.1038/s41586-019-1876-x>

Original link to raw data, containing the entire dataset: <https://doi.org/10.5281/zenodo.3518284>

Value

A `SingleCellExperiment` object with single cell data, a `SpatialExperiment` object with single cell data, a `CytoImageList` object containing multichannel images, or a `CytoImageList` object containing cell segmentation masks.

Author(s)

Jana Fischer

References

Jackson, Fischer et al. (2020). The single-cell pathology landscape of breast cancer. *Nature* 578(7796), 615-620.

Examples

```
# Load single cell data
sce <- JacksonFischer_2020_BreastCancer(data_type = "sce")
print(sce)

# Display metadata
JacksonFischer_2020_BreastCancer(data_type = "sce", metadata = TRUE)

# Load masks on disk
library(HDF5Array)
masks <- JacksonFischer_2020_BreastCancer(data_type = "masks", on_disk =
TRUE, h5FilesPath = getHDF5DumpDir())
print(head(masks))
```

listDatasets	<i>List all available datasets</i>
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Description

Summary information for all available datasets in the **imcdatasets** package.

Usage

```
listDatasets()
```

Details

Each dataset contains single-cell data, multichannel images and cell segmentation masks.

Value

A [DataFrame](#) where each row corresponds to a dataset, containing the fields:

- `FunctionCall`, the R function call required to construct the dataset.
- `Species`, species of origin.
- `Tissue`, the tissue that was imaged.
- `NumberOfCells`, the total number of cells in the dataset.
- `NumberOfImages`, the total number of images in the dataset.
- `NumberOfChannels`, the number of channels per image.
- `Reference`, a Markdown-formatted citation to `scripts/ref.bib` in the **imcdatasets** installation directory.

Examples

```
listDatasets()
```

ZanotelliSpheroids2020-defunct

Defunct function - 'ZanotelliSpheroids2020' dataset

Description

These functions are defunct and no longer available.

Details

Defunct functions are: ZanotelliSpheroids2020_sce, ZanotelliSpheroids2020_masks, and ZanotelliSpheroids2020_images. Please use the Zanotelli_2020_Spheroids function instead.

ZanotelliSpheroids2020Data

Obtain the zanotelli-spheroids-2020 dataset

Description

This function is provided for compatibility with older versions but is deprecated. As a replacement, please use Zanotelli_2020_Spheroids. Obtain the zanotelli-spheroids-2020 dataset, which consists of three data objects: single cell data, multichannel images and cell segmentation masks. The data were obtained by imaging mass cytometry of sections of 3D spheroids generated from different cell lines.

Usage

```
ZanotelliSpheroids2020Data(  
  data_type = c("sce", "images", "masks"),  
  metadata = FALSE,  
  on_disk = FALSE,  
  h5FilePath = NULL,  
  force = FALSE  
)
```

Arguments

data_type	type of data to load, should be sce for single cell data, images for multichannel images or masks for cell segmentation masks.
metadata	if FALSE (default), the data object selected in data_type is returned. If TRUE, only the metadata associated to this object is returned.
on_disk	logical indicating if images in form of HDF5Array objects (as .h5 files) should be stored on disk rather than in memory. This setting is valid when downloading images and masks.

h5FilePath	path to where the .h5 files for on disk representation are stored. This path needs to be defined when <code>on_disk = TRUE</code> . When files should only temporarily be stored on disk, please set <code>h5FilePath = getHDF5DumpDir()</code>
force	logical indicating if images should be overwritten when files with the same name already exist on disk.

Details

This function is provided for compatibility with older versions but is deprecated. As a replacement, please use `Zanotelli_2020_Spheroids`. This is an Imaging Mass Cytometry (IMC) dataset from Zanotelli et al. (2020), consisting of three data objects:

- `images` contains 517 multichannel images, each containing 51 channels, in the form of a `CytoImageList` class object.
- `masks` contains the cell segmentation masks associated with the images, in the form of a `CytoImageList` class object.
- `sce` contains the single cell data extracted from the multichannel images using the cell segmentation masks, as well as the associated metadata, in the form of a `SingleCellExperiment`. This represents a total of 229,047 cells x 51 channels.

All data are downloaded from ExperimentHub and cached for local re-use.

Mapping between the three data objects is performed via variables located in their metadata columns: `mcols()` for the `CytoImageList` objects and `colData()` for the `SingleCellExperiment` object. Mapping at the image level can be performed with the `ImageName` or `ImageNumber` variables. Mapping between cell segmentation masks and single cell data is performed with the `CellNumber` variable, the values of which correspond to the intensity values of the `ZanotelliSpheroids2020_masks` object. For practical examples, please refer to the "Accessing IMC datasets" vignette.

This dataset was obtained as following (the names of the experimental variables, located in the `colData` of the `SingleCellExperiment` object, are indicated in parentheses): *i*) Cells from four different cell lines (`cellline`) were seeded at three different densities (concentration, relative densities) and grown for either 72 or 96 hours (`time_point`, duration in hours). In the appropriate experimental conditions (see the paper for details), the cells aggregate into 3D spheroids. *ii*) Cells were harvested and pooled into 60-well barcoding plates. *iii*) A pellet of each spheroid pool was generated and cut into several 6 μm -thick sections. *iv*) A subset of these sections (`site_id`) were stained with an IMC panel and acquired as one or more acquisitions (`acquisition_id`) containing multiple spheres each. *v*) Spheres in these acquisitions were identified by computer vision and cropped into individual images (`ImageNumber`).

Other relevant cell metadata include:

- `condition_name`: experimental conditions in the format: "Cell line name"_c"seeding density"_tp"time point".
- `Center_X/Y`: object centroid position in image.
- `Area`: area of the cell (μm^2).
- `dist.rim`: estimated distance to spheroid border.
- `dist.sphere`: distance to spheroid section border.
- `dist.other`: distance to the closest of the other spheroid sections in the same image (if there is any).

- `dist.bg`: distance to background pixels.
- `counts_neighb`: contains arsinh-transformed counts (cofactor = 1).
- `exprs_neighb`: contains arsinh-transformed counts (cofactor 1).

For a full description of the other experimental variables, please refer to the publication (<https://doi.org/10.15252/msb.202097>) and to the original dataset repository (<https://doi.org/10.5281/zenodo.4271910>).

The marker-associated metadata, including antibody information and metal tags are stored in the `rowData` of the `SingleCellExperiment` object. The channels with names starting with "BC_" are the channels used for barcoding. Post-transcriptional modification of the protein targets are indicated in brackets.

The assay slot of the `SingleCellExperiment` object contains four assays:

- `counts`: mean ion counts per cell.
- `exprs`: arsinh-transformed counts per cell, with cofactor 1.
- `counts_neighb`: mean ion counts of the neighboring cells.
- `exprs_neighb`: arsinh-transformed counts (cofactor 1) of the neighboring cells.

The metadata slot of the `SingleCellExperiment` object contains a graph of cell neighbors, generated with the `igraph::graph_from_data_frame` function.

File sizes:

- ``images``: size in memory = 21.2 Gb, size on disk = 860 Mb.
- ``masks``: size in memory = 426 Mb, size on disk = 12 Mb.
- ``sce``: size in memory = 564 Mb, size on disk = 319 Mb.

When storing images on disk, these need to be first fully read into memory before writing them to disk. This means the process of downloading the data is slower than directly keeping them in memory. However, downstream analysis will lose its memory overhead when storing images on disk.

Original source: Zanotelli et al. (2020): <https://doi.org/10.15252/msb.20209798>

Original link to raw data, also containing the entire dataset: <https://doi.org/10.5281/zenodo.4271910>

Value

A `SingleCellExperiment` object with single cell data, a `CytoImageList` object containing multichannel images, or a `CytoImageList` object containing cell segmentation masks.

Author(s)

Nicolas Damond

References

Zanotelli VRT et al. (2020). A quantitative analysis of the interplay of environment, neighborhood, and cell state in 3D spheroids *Mol Syst Biol* 16(12), e9798.

Examples

```
# Load single cell data
sce <- ZanotelliSpheroids2020Data(data_type = "sce")
print(sce)

# Display metadata
ZanotelliSpheroids2020Data(data_type = "sce", metadata = TRUE)

# Load masks on disk
library(HDF5Array)
masks <- ZanotelliSpheroids2020Data(data_type = "masks", on_disk = TRUE,
h5FilePath = getHDF5DumpDir())
print(head(masks))
```

Zanotelli_2020_Spheroids

Obtain the Zanotelli_2020_Spheroids dataset

Description

Obtain the Zanotelli_2020_Spheroids dataset, which consists of three data objects: single cell data, multichannel images and cell segmentation masks. The data were obtained by imaging mass cytometry (IMC) of sections of 3D spheroids generated from different cell lines.

Usage

```
Zanotelli_2020_Spheroids(
  data_type = c("sce", "spe", "images", "masks"),
  metadata = FALSE,
  on_disk = FALSE,
  h5FilePath = NULL,
  version = "latest",
  force = FALSE
)
```

Arguments

data_type	type of object to load, ‘images’ for multichannel images or ‘masks’ for cell segmentation masks. Single cell data are retrieved using either ‘sce’ for the SingleCellExperiment format or ‘spe’ for the SpatialExperiment format.
metadata	if FALSE (default), the data object selected in data_type is returned. If TRUE, only the metadata associated to this object is returned.
on_disk	logical indicating if images in form of HDF5Array objects (as .h5 files) should be stored on disk rather than in memory. This setting is valid when downloading images and masks.

h5FilePath	path to where the .h5 files for on disk representation are stored. This path needs to be defined when <code>on_disk = TRUE</code> . When files should only temporarily be stored on disk, please set <code>h5FilePath = getHDF5DumpDir()</code> .
version	dataset version. By default, the latest version is returned.
force	logical indicating if images should be overwritten when files with the same name already exist on disk.

Details

This is an Imaging Mass Cytometry (IMC) dataset from Zanotelli et al. (2020), consisting of three data objects:

- `images` contains 517 multichannel images, each containing 51 channels, in the form of a [CytolmagineList](#) class object.
- `masks` contains the cell segmentation masks associated with the images, in the form of a [CytolmagineList](#) class object.
- `sce` contains the single cell data extracted from the multichannel images using the cell segmentation masks, as well as the associated metadata, in the form of a [SingleCellExperiment](#). This represents a total of 229,047 cells x 51 channels.
- `spe` same single cell data as for `sce`, but in the [SpatialExperiment](#) format.

All data are downloaded from ExperimentHub and cached for local re-use.

Mapping between the three data objects is performed via variables located in their metadata columns: `mcols()` for the [CytolmagineList](#) objects and `colData()` for the [SingleCellExperiment](#) and [SpatialExperiment](#) objects. Mapping at the image level can be performed with the `image_name` or `image_number` variables. Mapping between cell segmentation masks and single cell data is performed with the `cell_number` variable, the values of which correspond to the intensity values of the masks object. For practical examples, please refer to the "Accessing IMC datasets" vignette.

This dataset was obtained as following (the names of the experimental variables, located in the `colData` of the [SingleCellExperiment](#) and [SpatialExperiment](#) objects, are indicated in parentheses): *i*) Cells from four different cell lines (`cell_line`) were seeded at three different densities (`treatment_concentration`, relative densities) and grown for either 72 or 96 hours (`treatment_time_point`, duration in hours). In the appropriate experimental conditions (see the paper for details), the cells aggregate into 3D spheroids. *ii*) Cells were harvested and pooled into 60-well barcoding plates. *iii*) A pellet of each spheroid pool was generated and cut into several 6 μm -thick sections. *iv*) A subset of these sections (`site_id`) were stained with an IMC panel and acquired as one or more acquisitions (`acquisition_id`) containing multiple spheres each. *v*) Spheres in these acquisitions were identified by computer vision and cropped into individual images (`image_number`).

Other relevant cell metadata include:

- `treatment_name`: experimental conditions in the format: "Cell line name"_c"seeding density"_tp"time point".
- `cell_x/cell_y`: cell centroid position in the image.
- `cell_area`: area of the cell (μm^2).
- `distance_rim`: estimated distance to spheroid border.
- `distance_sphere`: distance to spheroid section border.

- `distance_other_sphere`: distance to the closest of the other spheroid sections in the same image (if there is any).
- `distance_background`: distance to background pixels.

For a full description of the other experimental variables, please refer to the publication (<https://doi.org/10.15252/msb.202097>) and to the original dataset repository (<https://doi.org/10.5281/zenodo.4271910>).

The marker-associated metadata, including antibody information and metal tags are stored in the `rowData` of the `SingleCellExperiment` and `SpatialExperiment` objects. The channels with names starting with "BC_" are the channels used for barcoding. Post-transcriptional modification of the protein targets are indicated in brackets.

The assay slots of the `SingleCellExperiment` and `SpatialExperiment` objects contain three assays:

- `counts` contains raw mean ion counts per cell.
- `exprs` contains arsinh-transformed counts, with cofactor 1.
- `quant_norm` contains counts censored at the 99th percentile and scaled 0-1.

In addition, the `altExp` slot of the `SingleCellExperiment` object contains another `SingleCellExperiment` object where the counts matrix represents raw mean ion counts for cells neighboring the current cell.

Neighborhood information, defined here as cells that are localized next to each other, is stored as a `SelfHits` object in the `colPairs` slot of the `SingleCellExperiment` and `SpatialExperiment` objects. Cells in the `SelfHits` object are represented by unique integers that map to the `cell_number_absolute` column of `colData(sce)`.

Dataset versions: a `version` argument can be passed to the function to specify which dataset version should be retrieved.

- ``v0``: original version (Bioconductor <= 3.15).
- ``v1``: consistent object formatting across datasets.

File sizes:

- ``images``: size in memory = 21.2 Gb, size on disk = 860 Mb.
- ``masks``: size in memory = 426 Mb, size on disk = 12 Mb.
- ``sce``: size in memory = 564 Mb, size on disk = 319 Mb.
- ``spe``: size in memory = 596 Mb, size on disk = 320 Mb.

When storing images on disk, these need to be first fully read into memory before writing them to disk. This means the process of downloading the data is slower than directly keeping them in memory. However, downstream analysis will lose its memory overhead when storing images on disk.

Original source: Zanotelli et al. (2020): <https://doi.org/10.15252/msb.20209798>

Original link to raw data, also containing the entire dataset: <https://doi.org/10.5281/zenodo.4271910>

Value

A `SingleCellExperiment` object with single cell data, a `SpatialExperiment` object with single cell data, a `CytoImageList` object containing multichannel images, or a `CytoImageList` object containing cell segmentation masks.

Author(s)

Nicolas Damond

References

Zanotelli VRT et al. (2020). A quantitative analysis of the interplay of environment, neighborhood, and cell state in 3D spheroids *Mol Syst Biol* 16(12), e9798.

Examples

```
# Load single cell data
sce <- Zanotelli_2020_Spheroids(data_type = "sce")
print(sce)

# Display metadata
Zanotelli_2020_Spheroids(data_type = "sce", metadata = TRUE)

# Load masks on disk
library(HDF5Array)
masks <- Zanotelli_2020_Spheroids(data_type = "masks", on_disk = TRUE,
h5FilesPath = getHDF5DumpDir())
print(head(masks))
```

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