Package ‘ricu’

November 3, 2020

Title  Intensive Care Unit Data with R
Description  Focused on (but not exclusive to) data sets hosted on PhysioNet
(https://physionet.org), ‘ricu’ provides utilities for download, setup
and access of intensive care unit (ICU) data sets. In addition to
functions for running arbitrary queries against available data sets, a
system for defining clinical concepts and encoding their representations
in tabular ICU data is presented.
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### Description

Making a dataset available to `ricu` consists of 3 steps: downloading (`download_src()`), importing (`import_src()`), and attaching (`attach_src()`). While downloading and importing are one-time procedures, attaching of the dataset is repeated every time the package is loaded. Briefly, downloading loads the raw dataset from the internet (most likely in `.csv` format), importing consists of some preprocessing to make the data available more efficiently and attaching sets up the data for use by the package.

### Usage

```
attach_src(x, ...)  
## S3 method for class 'src_cfg'
attach_src(x, assign_env = NULL, data_dir = src_data_dir(x), ...)
```
attach_src

## S3 method for class 'character'
attach_src(x, assign_env = NULL, data_dir = src_data_dir(x), ...)

setup_src_env(x, env, ...)

## S3 method for class 'src_cfg'
setup_src_env(x, env, data_dir = src_data_dir(x), ...)

new_src_tbl(files, col_cfg, tbl_cfg, prefix, src_env)

as_src_tbl(x, ...)

new_src_env(x, env = new.env(parent = data_env()))

as_src_env(x)

**Arguments**

- **x**  
  Data source to attach

- **...**  
  Forwarded to further calls to attach_src()

- **assign_env**  
  Environment in which the data source will become available

- **data_dir**  
  Directory used to look for fst::fst() files; NULL calls data_dir() using the source name as subdir argument

- **env**  
  Environment where data proxy objects are created

- **files**  
  File names of fst files that will be used to create a prt object (see also prt::new_prt())

- **col_cfg**  
  Coerced to col_cfg by calling as_col_cfg()

- **tbl_cfg**  
  Coerced to tbl_cfg by calling as_tbl_cfg()

- **prefix**  
  Character vector valued data source name(s) (used as class prefix)

- **src_env**  
  The data source environment (as src_env object)

**Details**

Attaching a dataset sets up two types of S3 classes: a single src_env object, containing as many src_tbl objects as tables are associated with the dataset. A src_env is an environment with an id_cfg attribute, as well as sub-classes as specified by the data source class_prefix configuration setting (see load_src_cfg()). All src_env objects created by calling attach_src() represent environments that are direct descendants of the data environment and are bound to the respective dataset name within that environment. While attach_src() does not immediately instantiate a src_env object, it rather creates a promise using base::delayedAssign() which evaluates to a src_env upon first access. This allows for data sources to be set up where the data is missing in a way that prompts the user to download and import the data when first accessed.

Additionally, attach_src() creates an active binding using base::makeActiveBinding(), binding a function to the dataset name within the environment passed as assign_env, which retrieves the respective src_env from the data environment. This shortcut is set up for convenience, such that for example the MIMIC-III demo dataset not only is available as ricu::data::mimic_demo, but...
also as ricu::mimic_demo (or if the package namespace is attached, simply as mimic_demo). The ricu namespace contains objects mimic, mimic_demo, eicu, etc. which are used as such links when loading the package. However, new data sets can be set up an accessed in the same way.

If set up correctly, it is not necessary for the user to directly call attach_src(). When the package is loaded, the default data sources are attached automatically. This default can be controlled by setting as environment variable RICU_SRC_LOAD a comma separated list of data source names before loading the library. Setting this environment variable as

```r
Sys.setenv(RICU_SRC_LOAD = "mimic_demo,eciu_demo")
```

will change the default of loading both MIMIC-III and eICU, alongside the respective demo datasets, and HiRID, to just the two demo datasets. For setting an environment variable upon startup of the R session, refer to `base::.First.sys()`.

The src_env promise for each data source is created using the S3 generic function setup_src_env(). This function checks if all required files are available from data_dir. If files are missing the user is prompted for download in interactive sessions and an error is thrown otherwise. As soon as all required data is available, a src_tbl object is created per table and assigned to the src_env.

The S3 class src_tbl inherits from prt, which represents a partitioned fst file. In addition to the prt object, meta data in the form of col_cfg and tbl_cfg is associated with a src_tbl object (see `load_src_cfg()`). Furthermore, as with src_env, sub-classes are added as specified by the source configuration class_prefix entry. This allows certain functionality, for example data loading, to be adapted to data source-specific requirements.

**Value**

The constructors `new_src_env()`/`new_src_tbl()` as well as coercion functions `as_src_env()`/`as_src_tbl()` return src_env and src_tbl objects respectively. The function `attach_src()` is called for side effects and returns NULL invisibly, while `setup_src_env()` instantiates and returns a src_env object.

**Examples**

```r
## Not run:
Sys.setenv(RICU_SRC_LOAD = "")
library(ricu)

ls(envir = data)
exists("mimic_demo")
attach_src("mimic_demo")

ls(envir = data)
exists("mimic_demo")
mimic_demo

## End(Not run)
```
Description

ICU datasets such as MIMIC-III or eICU typically represent patients by multiple ID systems such as patient IDs, hospital stay IDs and ICU admission IDs. Even if the raw data is available in only one such ID system, given a mapping of IDs alongside start and end times, it is possible to convert data from one ID system to another. The function `change_id()` provides such a conversion utility, internally either calling `upgrade_id()` when moving to an ID system with higher cardinality and `downgrade_id()` when the target ID system is of lower cardinality.

Usage

```r
change_id(x, target_id, src, ..., keep_old_id = TRUE)
upgrade_id(x, target_id, src, cols = time_vars(x), ...)
downgrade_id(x, target_id, src, cols = time_vars(x), ...)
```

Details

In order to provide ID system conversion for a data source, the (internal) function `id_map()` must be able to construct an ID mapping for that data source. Constructing such a mapping can be expensive w.r.t. the frequency it might be re-used and therefore, `id_map()` provides caching infrastructure.
The mapping itself is constructed by the (internal) function `id_map_helper()`, which is expected to provide source and destination ID columns as well as start and end columns corresponding to the destination ID, relative to the source ID system. In the following example, we request for `mimic_demo`, with ICU stay IDs as source and hospital admissions as destination IDs.

```r
id_map_helper(mimic_demo, "icustay_id", "hadm_id")
```

### # An 'id_tbl': 136 x 4
### # Id var: icustay_id
### icustay_id hadm_id hadm_id_start hadm_id_end
### <int> <int> <drttn> <drttn>
### # 1 201006 198503 -3290 mins 9114 mins
### # 2 201204 114648 -2 mins 6949 mins
### # 3 203766 126949 -1336 mins 8818 mins
### # 4 204132 157609 -1 mins 10103 mins
### # 5 204201 177678 -368 mins 9445 mins
### # ...
### # 132 295043 170883 -10413 mins 31258 mins
### # 133 295741 176805 -1 mins 3153 mins
### # 134 296804 110244 -1294 mins 4599 mins
### # 135 297782 167612 -1 mins 207 mins
### # 136 298685 151323 -1 mins 19082 mins
### # ...

Both start and end columns encode the hospital admission windows relative to each corresponding ICU stay start time. It therefore comes as no surprise that most start times are negative (hospital admission typically occurs before ICU stay start time), while end times are often days in the future (as hospital discharge typically occurs several days after ICU admission).

In order to use the ID conversion infrastructure offered by `ricu` for a new dataset, it typically suffices to provide an `id_cfg` entry in the source configuration (see `load_src_cfg()`), outlining the available ID systems alongside an ordering, as well as potentially a class specific implementation of `id_map_helper()` for the given source class, specifying the corresponding time windows in 1 minute resolution (for every possible pair of IDs).

While both up- and downgrades for `id_tbl` objects, as well as downgrades for `ts_tbl` objects are simple merge operations based on the ID mapping provided by `id_map()`, ID upgrades for `ts_tbl` objects are slightly more involved. As an example, consider the following setting: we have data associated with `hadm_id` IDs and times relative to hospital admission:

```
data
----|------------------|------------------|------------------|------------------|------------------|
3h 10h 18h 27h 35h 43h 52h 59h

hadm_id
|------------------|------------------|
0h 7h 26h 37h 53h 62h

icustay_id
|------------------|------------------|
0h 19h 0h 16h

ICU_1 ICU_2
```
The mapping of data points from hadm_id to icustay_id is created as follows: ICU stay end times mark boundaries and all data that is recorded after the last ICU stay ended is assigned to the last ICU stay. Therefore data points 1-3 are assigned to ICU_1, while 4-8 are assigned to ICU_2. Times have to be shifted as well, as timestamps are expected to be relative to the current ID system. Data points 1-3 therefore are assigned to time stamps -4h, 3h and 11h, while data points 4-8 are assigned to -10h, -2h, 6h, 15h and 22h. Implementation-wise, the mapping is computed using an efficient data.table rolling join.

Value
An object of the same type as x with modified IDs.

Examples
if (require(mimic.demo)) {
  tbl <- mimic_demo.labevents
  dat <- load_diff_time(tbl, itemid == 50809, c("charttime", "valuenum"))
  dat

  change_id(dat, "icustay_id", tbl, keep_old_id = FALSE)
}

ICU datasets

Description
The Laboratory for Computational Physiology (LCP) at MIT hosts several large-scale databases of hospital intensive care units (ICUs), two of which can be either downloaded in full (MIMIC-III and eICU) or as demo subsets (MIMIC-III demo and eICU demo), while a third data set is available only in full (HiRID). While demo data sets are freely available, full download requires credentialled access which can be gained by applying for an account with PhysioNet. Even though registration is required, the described datasets are all publicly available.

Usage
data

Format
The exported data environment contains all datasets that have been made available to ricu. For datasets that are attached during package loading (see attach_src()), shortcuts to the datasets are set up in the package namespace, allowing the object ricu::data::mimic_demo to be accessed as ricu::mimic_demo (or in case the package namespace has been attached, simply as mimic_demo). Datasets that are made available after the package namespace has been sealed will have their proxy object by default located in .GlobalEnv. Datasets are represented by src_env objects, while individual tables are src_tbl and do not represent in-memory data, but rather data stored on disk, subsets of which can be loaded into memory.
Details

Setting up a dataset for use with ricu requires a configuration object. For the included datasets, configuration can be loaded from

```r
system.file("extdata", "config", "data-sources.json", package = "ricu")
```

by calling `load_src_cfg()` and for dataset that are external to ricu, additional configuration can be made available by setting the environment variable RICU_CONFIG_PATH (for more information, refer to `load_src_cfg()`). Using the dataset configuration object, data can be downloaded (`download_src()`), imported (`import_src()`) and attached (`attach_src()`). While downloading and importing are one-time procedures, attaching of the dataset is repeated every time the package is loaded. Briefly, downloading loads the raw dataset from the internet (most likely in .csv format), importing consists of some preprocessing to make the data available more efficiently (by converting it to .fst format) and attaching sets up the data for use by the package. For more information on the individual steps, refer to the respective documentation pages.

A dataset that has been successfully made available can interactively be explored by typing its name into the console and individual tables can be inspected using the `$` function. For example for the MIMIC-III demo dataset and the icustays table, this gives

```r
mimic_demo

## <mimic_demo Env[25]>
## icustays procedures_icd d_cpt d_items
## [136 x 12] [506 x 5] [134 x 9] [12,487 x 10]
## diagnoses_icd datetimeevents chartevents admissions
## [1,761 x 5] [15,551 x 14] [758,355 x 15] [129 x 19]
## inputevents_cv d_labitems microbiologyevents outputevents
## [34,799 x 22] [753 x 6] [2,003 x 16] [11,320 x 13]
## procedureevents_mv callout d_icd_diagnoses cptevents
## [753 x 25] [77 x 24] [14,567 x 4] [1,579 x 12]
## prescriptions caregivers drgcodes transfers
## [10,398 x 19] [7,567 x 4] [297 x 8] [524 x 13]
## patients labevents d_icd_procedures services
## [100 x 8] [76,074 x 9] [3,882 x 4] [163 x 6]
## inputevents_mv
## [13,224 x 31]
```

```
mimic_demo$icustays

## # <mimic_tbl>: [136 x 12]
## # ID options: subject_id (patient) < hadm_id (hadm) < icustay_id (icustay)
## # Defaults: intime (index), last_careunit (value)
## # Time vars: intime, outtime
## row_id subject_id hadm_id icustay_id dbsource first_careunit last_careunit
## <int> <int> <int> <int> <chr> <chr> <chr>
## 1 12742 10006 142345 206504 carevue MICU MICU
## 2 12747 10011 105331 232110 carevue MICU MICU
## 3 12749 10013 165520 264446 carevue MICU MICU
```

```r
mimic_demo$icustays

   row_id subject_id hadm_id icustay_id dbsource first_careunit last_careunit
   1   12742      10006     142345       carevue       MICU        MICU
   2   12747      10011     105331       carevue       MICU        MICU
   3   12749      10013     165520       carevue       MICU        MICU
```
Table subsets can be loaded into memory for example using the base::subset() function, which uses non-standard evaluation (NSE) to determine a row-subsetting. This design choice stems from the fact that some tables can have on the order of 10^8 rows, which makes loading full tables into memory an expensive operation. Table subsets loaded into memory are represented as data.table objects. Extending the above example, if only ICU stays corresponding to the patient with subject_id == 10124 are of interest, the respective data can be loaded as

```r
subset(mimic_demo$icustays, subject_id == 10124)
```

Much care has been taken to make ricu extensible to new datasets. For example the publicly available ICU database AmsterdamUMCdb provided by the Amsterdam University Medical Center, currently is not part of the core datasets of ricu, but code for integrating this dataset is available on github.

MIMIC-III

The Medical Information Mart for Intensive Care (MIMIC) database holds detailed clinical data from roughly 60,000 patient stays in Beth Israel Deaconess Medical Center (BIDMC) intensive care units between 2001 and 2012. The database includes information such as demographics, vital sign measurements made at the bedside (~1 data point per hour), laboratory test results, procedures, medications, caregiver notes, imaging reports, and mortality (both in and out of hospital). For further information, please refer to the MIMIC-III documentation.

The corresponding demo dataset contains the full data of a randomly selected subset of 100 patients from the patient cohort with conformed in-hospital mortality. The only notable data omission is the noteevents table, which contains unstructured text reports on patients.
eICU

More recently, Philips Healthcare and LCP began assembling the eICU Collaborative Research Database as a multi-center resource for ICU data. Combining data of several critical care units throughout the continental United States from the years 2014 and 2015, this database contains de-identified health data associated with over 200,000 admissions, including vital sign measurements, care plan documentation, severity of illness measures, diagnosis information, and treatment information. For further information, please refer to the eICU documentation.

For the demo subset, data associated with ICU stays for over 2,500 unit stays selected from 20 of the larger hospitals is included. An important caveat that applied to the eICU-based datasets is considerable variability among the large number of hospitals in terms of data availability.

HiRID

Moving to higher time-resolution, HiRID is a freely accessible critical care dataset containing data relating to almost 34,000 patient admissions to the Department of Intensive Care Medicine of the Bern University Hospital, Switzerland. The dataset contains de-identified demographic information and a total of 681 routinely collected physiological variables, diagnostic test results and treatment parameters, collected during the period from January 2008 to June 2016. Dependent on the type of measurement, time resolution can be on the order of 2 minutes.

References


data_dir

File system utilities

Description

Determine the location where to place data meant to persist between individual sessions.
Usage

data_dir(subdir = NULL, create = TRUE)

src_data_dir(srcs)

auto_load_src_names()

src_data_avail(src = auto_load_src_names())

config_paths()

get_config(name, cfg_dirs = config_paths(), combine_fun = c, ...)

set_config(x, name, dir = file.path("inst", "extdata", "config"), ...)

Arguments

subdir A string specifying a directory that will be made sure to exist below the data
directory.
create Logical flag indicating whether to create the specified directory
src, srcs Character vector of data source names
name File name of the configuration file (.json will be appended)
cfg_dirs Character vector of directories searched for config files
combine_fun If multiple files are found, a function for combining returned lists
... Passed to jsonlite::read_json() or jsonlite::write_json()
x Object to be written
dir Directory to write the file to (created if non-existent)

Details

For data, the default location depends on the operating system as

<table>
<thead>
<tr>
<th>Platform</th>
<th>Location</th>
</tr>
</thead>
<tbody>
<tr>
<td>Linux</td>
<td>~/.local/share/ricu</td>
</tr>
<tr>
<td>macOS</td>
<td>~/Library/Application Support/ricu</td>
</tr>
<tr>
<td>Windows</td>
<td>%LOCALAPPDATA%/ricu</td>
</tr>
</tbody>
</table>

If the default storage directory does not exists, it will only be created upon user consent (requiring
an interactive session).

The environment variable RICU_DATA_PATH can be used to overwrite the default location. If desired,
this variable can be set in an R startup file to make it apply to all R sessions. For example, it could
be set within:

- A project-local .Renviron;
- The user-level .Renviron;
- A file at $(R RHOME)/etc/Renviron.site.

Any directory specified as environment variable will recursively be created.

Data source directories typically are sub-directories to data_dir() named the same as the respective dataset. For demo datasets corresponding to mimic and eicu, file location however deviates from this scheme. The function src_data_dir() is used to determine the expected data location of a given dataset.

Configuration files used both for data source configuration, as well as for dictionary definitions potentially involve multiple files that are read and merged. For that reason, get_config() will iterate over directories passed as cfg_dirs and look for the specified file (with suffix .json appended and might be missing in some of the queried directories). All found files are read by jsonlite::read_json() and the resulting lists are combined by reduction with the binary function passed as combine_fun.

With default arguments, get_config() will simply concatenate lists corresponding to files found in the default config locations as returned by config_paths(): first the directory specified by the environment variable RICU_CONFIG_PATH (if set), followed by the directory at

```r
system.file("extdata", "config", package = "ricu")
```

Further arguments are passed to jsonlite::read_json(), which is called with slightly modified defaults: simplifyVector = TRUE, simplifyDataFrame = FALSE and simplifyMatrix = FALSE.

The utility function set_config() writes the list passed as x to file name.json, using jsonlite::write_json() also with slightly modified defaults (which can be overridden by passing arguments as ...): null = "null", auto_unbox = TRUE and pretty = TRUE.

Whenever the package namespace is attached, a summary of dataset availability is printed using the utility functions auto_load_src_names() and src_data_avail(). While the former simply returns a character vector of data sources that are configures for automatically being set up on package loading, the latter returns a summary of the number of available tables per dataset.

**Value**

Functions data_dir(), src_data_dir() and config_paths() return file paths as character vectors, auto_load_src_names() returns a character vector of data source names and src_data_avail() a data.frame describing availability of data sources. Configuration utilities get_config() and set_config() read and write list objects to/from JSON format.

**Examples**

```r
Sys.setenv(RICU_DATA_PATH = tempdir())
identical(data_dir(), tempdir())

dir.exists(file.path(tempdir(), "some_subdir"))
some_subdir <- data_dir("some_subdir")
dir.exists(some_subdir)

cfg <- get_config("concept-dict")
```
download_src

identical(
  cfg,
  get_config("concept-dict",
    system.file("extdata", "config", package = "ricu"))
)

---

download_src  Data download utilities

Description
Making a dataset available to ricu consists of 3 steps: downloading (download_src()), importing (import_src()) and attaching (attach_src()). While downloading and importing are one-time procedures, attaching of the dataset is repeated every time the package is loaded. Briefly, downloading loads the raw dataset from the internet (most likely in .csv format), importing consists of some preprocessing to make the data available more efficiently (by converting it to .fst format) and attaching sets up the data for use by the package.

Usage

```r
download_src(x, ...) 
``` 

## S3 method for class 'src_cfg'
download_src(  
  x,  
  data_dir = src_data_dir(x),  
  tables = NULL,  
  force = FALSE,  
  user = NULL,  
  pass = NULL,  
  ...  
)

Arguments

x  Object specifying the source configuration
...
Generic consistency
data_dir  Destination directory where the downloaded data is written to.
tables  Character vector specifying the tables to download. If NULL, all available tables are downloaded.
force  Logical flag; if TRUE, existing data will be re-downloaded
user, pass  PhysioNet credentials; if NULL and environment variables RICU_PHYSIONET_USER/RICU_PHYSIONET_PASS are not set, user input is required
Details

Downloads by ricu are focused data hosted by PhysioNet and tools are currently available for
downloading the datasets MIMIC-III, eICU and HiRID (see data). While credentials are required
for downloading any of the three datasets, demo dataset for both MIMIC-III and eICU are available
without having to log in. Even though access to full dataset is credentialized, the datasets are in fact
publicly available. For setting up an account, please refer to the registration form.

PhysioNet credentials can either be entered in an interactive session, passed as function arguments
user/pass or as environment variables RICU_PHYSIONET_USER/RICU_PHYSIONET_PASS. For setting
environment variables on session startup, refer to base::First.sys() and for setting environment
variables in general, refer to base::Sys.setenv() If the openssl package is available, SHA256
hashes of downloaded files are verified using openssl::sha256().

Demo datasets MIMIC-III demo and eiCU demo can either be installed as R packages directly by running

```r
install.packages(
  c("mimic.demo", "eicu.demo"),
  repos = "https://septic-tank.github.io/physionet-demo"
)
```

or downloaded and imported using download_src() and import_src(). Furthermore, ricu specifies mimic.demo and eicu.demo as suggests dependencies therefore, passing dependencies = TURE when calling install.packages() for installing ricu, this will automatically install the
demo datasets as well.

While the included data downlosers are intended for data hosted by PhysioNet, download_src() is an S3 generic function that can be extended to new classes. Method dispatch is intended to occur
on objects that inherit from or can be coerced to src_cfg. For more information on data source
configuration, refer to load_src_cfg().

Value

Called for side effects and returns NULL invisibly.

Examples

```r
## Not run:
dir <- tempdir()
list.files(dir)
download_datasource("mimic_demo", data_dir = dir)
list.files(dir)
unlink(dir, recursive = TRUE)

## End(Not run)
```
**Time series utility functions**

ICU data as handled by ricu is mostly comprised of time series data and as such, several utility functions are available for working with time series data in addition to a class dedicated to representing time series data (see `ts_tbl()`). Some terminology to begin with: a time series is considered to have gaps if, per (combination of) ID variable value(s), some time steps are missing. Expanding and collapsing mean to change between representations where time steps are explicit or encoded as interval with start and end times. For sliding window-type operations, `slide()` means to iterate over time-windows, `slide_index()` means to iterate over certain time-windows, selected relative to the index and `hop()` means to iterate over time-windows selected in absolute terms.

**Usage**

```r
expand(
  x, 
  start_var = "start", 
  end_var = "end", 
  step_size = NULL, 
  new_index = NULL, 
  keep_vars = id_vars(x)
)
```

```r
collapse(
  x, 
  id_vars = NULL, 
  index_var = NULL, 
  start_var = "start", 
  end_var = "end", 
  env = NULL,
  ...
)
```

```r
has_no_gaps(x)
```

```r
has_gaps(...)
```

```r
is_regular(x)
```

```r
fill_gaps(x, limits = collapse(x), start_var = "start", end_var = "end")
```

```r
remove_gaps(x)
```

```r
slide(x, expr, before, after = hours(0L), ...)
```
slide_index(x, expr, index, before, after = hours(0L), ...)

hop(
  x,
  expr,
  full_window = FALSE,
  lwr_col = "min_time",
  upr_col = "max_time",
  left_closed = TRUE,
  right_closed = TRUE,
  eval_env = NULL,
  ...
)

Arguments

x  ts_tbl object to use

start_var, end_var  Name of the columns that represent lower and upper windows bounds

step_size  Controls the step size used to interpolate between start_var and end_var

new_index  Name of the new index column

keep_vars  Names of the columns to hold onto

id_vars, index_var  ID and index variables

eval_env  Environment used as parent to the environment used to evaluate expressions passes as ...

...  Passed to hop quo() and ultimately to data.table::[]()

limits  A table with columns for lower and upper window bounds

eexpr  Expression (quoted for * quo and unquoted otherwise) to be evaluated over each window

before, after  Time span to look back/forward

index  A vector of times around which windows are spanned (relative to the index)

windows  An icu_tbl defining the windows to span

full_window  Logical flag controlling how the situation is handled where the sliding window extends beyond available data

lwr_col, upr_col  Names of columns (in windows) of lower/upper window bounds

left_closed, right_closed  Logical flag indicating whether intervals are closed (default) or open.

eval_env  Environment in which expr is substituted; NULL resolves to the environment in which expr was created
A gap in a ts_tbl object is a missing time step, i.e. a missing entry in the sequence \( \text{seq}(\min(index), \max(index), \text{by} = \text{interval}) \) in at least one group (as defined by \( \text{id_vars}() \)), where the extrema are calculated per group. In this case, \( \text{has_gaps()} \) will return \( \text{TRUE} \). The function \( \text{is_regular()} \) checks whether the time series has no gaps, in addition to the object being sorted and unique (see \( \text{is_sorted()} \) and \( \text{is_unique()} \)). In order to transform a time series containing gaps into a regular time series, \( \text{fill_gaps()} \) will fill missing time steps with \( \text{NA} \) values in all \( \text{data_vars()} \) columns, while \( \text{remove_gaps()} \) provides the inverse operation of removing time steps that consist of \( \text{NA} \) values in \( \text{data_vars()} \) columns.

An \( \text{expand()} \) operation performed on an object inheriting from \( \text{data.table} \) yields a ts_tbl where time-steps encoded by columns \( \text{start_var} \) and \( \text{end_var} \) are made explicit with values in \( \text{keep_vars} \) being appropriately repeated. The inverse operation is available as \( \text{collapse()} \), which groups by \( \text{id_vars} \), represents \( \text{index_var} \) as group-wise extrema in two new columns \( \text{start_var} \) and \( \text{end_var} \) and allows for further data summary using . . .

Sliding-window type operations are available as \( \text{slide()} \), \( \text{slide_index()} \) and \( \text{hop()} \) (function naming is inspired by the CRAN package slider). The most flexible of the three, \( \text{hop} \) takes as input a ts_tbl object \( x \) containing the data, an id_tbl object \( \text{windows} \), containing for each ID the desired windows represented by two columns \( \text{lwr_col} \) and \( \text{upr_col} \), as well as an expression \( \text{expr} \) to be evaluated per window. At the other end of the spectrum, \( \text{slide()} \) spans windows for every ID and available time-step using the arguments \( \text{before} \) and \( \text{after} \), while \( \text{slide_index()} \) can be seen as a compromise between the two, where windows are spanned for certain time-points, specified by \( \text{index} \).

Most functions return ts_tbl objects with the exception of \( \text{has_gaps()}/\text{has_no_gaps()}/\text{is_regular()} \), which return logical flags.

### Examples

```r
tbl <- ts_tbl(x = 1:5, y = hours(1:5), z = hours(2:6), val = rnorm(5),
               index_var = "y")
exp <- expand(tbl, "y", "z", step_size = 1L, new_index = "y",
             keep_vars = c("x", "val"))
col <- collapse(exp, start_var = "y", end_var = "z", val = unique(val))
all.equal(tbl, col, check.attributes = FALSE)

tbl <- ts_tbl(x = rep(1:5, 1:5), y = hours(sequence(1:5)), z = 1:15)
win <- id_tbl(x = c(3, 4), a = hours(c(2, 1)), b = hours(c(3, 4)))
hop(tbl, list(z = sum(z)), win, lwr_col = "a", upr_col = "b")
slide_index(tbl, list(z = sum(z)), hours(c(4, 5)), before = hours(2))
slide(tbl, list(z = sum(z)), before = hours(2))

tbl <- ts_tbl(x = rep(3:4, 3:4), y = hours(sequence(3:4)), z = 1:7)
has_no_gaps(tbl)
is_regular(tbl)

tbl[1, 2] <- hours(2)
```
Tabular ICU data classes

Description

In order to simplify handling or tabular ICU data, ricu provides two S3 classes, id_tbl and ts_tbl. The two classes essentially consist of a data.table object, alongside some meta data and S3 dispatch is used to enable more natural behavior for some data manipulation tasks. For example, when merging two tables, a default for the by argument can be chosen more sensibly if columns representing patient ID and timestamp information can be identified.

Usage

id_tbl(..., id_vars = 1L)

is_id_tbl(x)

as_id_tbl(x, id_vars = NULL, by_ref = FALSE)

ts_tbl(..., id_vars = 1L, index_var = NULL, interval = NULL)

is_ts_tbl(x)

as_ts_tbl(x, id_vars = NULL, index_var = NULL, interval = NULL, by_ref = FALSE)

validate_tbl(x)

Arguments

... forwarded to data.table::data.table() or generic consistency

id_vars Column name(s) to be used as id column(s)

x Object to query/operate on

by_ref Logical flag indicating whether to perform the operation by reference

index_var Column name of the index column

interval Time series interval length specified as scalar-valued difftime object
Details

The two classes are designed for two often encountered data scenarios:

- *id_tbl* objects can be used to represent static (with respect to relevant time scales) patient data such as patient age and such an object is simply a *data.table* combined with a non-zero length character vector valued attribute marking the columns tracking patient ID information (*id_vars*). All further columns are considered as *data_vars*.

- *ts_tbl* objects are used for grouped time series data. A *data.table* object again is augmented by attributes, including a non-zero length character vector identifying patient ID columns (*id_vars*), a string, tracking the column holding time-stamps (*index_var*) and a scalar *difftime* object determining the time-series step size *interval*. Again, all further columns are treated as *data_vars*.

Owing to the nested structure of required meta data, *ts_tbl* inherits from *id_tbl*. Furthermore, both classes inherit from *data.table*. As such, *data.table* reference semantics are available for some operations, indicated by presence of a *by_ref* argument. At default, value, *by_ref* is set to *FALSE* as this is in line with base R behavior at the cost of potentially incurring unnecessary data copies. Some care has to be taken when passing *by_ref = TRUE* and enabling by reference operations as this can have side effects (see examples).

For instantiating *ts_tbl* objects, both *index_var* and *interval* can be automatically determined if not specified. For the index column, the only requirement is that a single *difftime* column is present, while for the time step, the minimal difference between two consecutive observations is chosen (and all differences are therefore required to be multiples of the minimum difference).

Upon instantiation, the data might be rearranged: columns are reordered such that ID columns are moved to the front, followed by the index column and a *data.table::key()* is set on meta columns, causing rows to be sorted accordingly. Moving meta columns to the front is done for reasons of convenience for printing, while setting a key on meta columns is done to improve efficiency of subsequent transformations such as merging or grouped operations. Furthermore, *NA* values in either ID or index columns are not allowed and therefore corresponding rows are silently removed.

Coercion between *id_tbl* and *ts_tbl* by default keeps intersecting attributes fixed and new attributes are by default inferred as for class instantiation. Each class comes with a class-specific implementation of the S3 generic function *validate_tbl()* which returns *TRUE* if the object is considered valid or a string outlining the type of validation failure that was encountered. Validity requires

1. inheriting from *data.table* and unique column names
2. for *id_tbl* that all columns specified by the non-zero length character vector holding onto the *id_vars* specification are available
3. for *ts_tbl* that the string-valued *index_var* column is available and does not intersect with *id_vars* and that the index column obeys the specified interval.

Finally, inheritance can be checked by calling *is_id_tbl()* and *is_ts_tbl()*). Note that due to *ts_tbl* inheriting from *id_tbl*, *is_id_tbl()* returns *TRUE* for both *id_tbl* and *ts_tbl* objects, while *is_ts_tbl()* only returns *TRUE* for *ts_tbl* objects.
Value

Constructors \texttt{id_tbl()}/\texttt{ts_tbl()}, as well as coercion functions \texttt{as_id_tbl()}/\texttt{as_ts_tbl()} return \texttt{id_tbl}/\texttt{ts_tbl} objects respectively, while inheritance testers \texttt{is_id_tbl()}/\texttt{is_ts_tbl()} return logical flags and \texttt{validate_tbl()} returns either \texttt{TRUE} or a string describing the validation failure.

Relationship to \texttt{data.table}

Both \texttt{id_tbl} and \texttt{ts_tbl} inherit from \texttt{data.table} and as such, functions intended for use with \texttt{data.table} objects can be applied to \texttt{id_tbl} and \texttt{ts_tbl} as well. But there are some caveats: Many functions introduced by \texttt{data.table} are not S3 generic and therefore they would have to be masked in order to retain control over how they operate on objects inheriting form \texttt{data.table}. Take for example the function \texttt{data.table::setnames()}, which changes column names by reference. Using this function, the name of an index column of an \texttt{id_tbl} object can me changed without updating the attribute marking the column as such and thusly leaving the object in an inconsistent state. Instead of masking the function \texttt{setnames()}, an alternative is provided as \texttt{rename_cols()}. In places where it is possible to seamlessly insert the appropriate function (such as \texttt{base::names<-()} or \texttt{base::colnames<-()}) and the responsibility for not using \texttt{data.table::setnames()} in a way that breaks the \texttt{id_tbl} object is left to the user.

Owing to \texttt{data.table} heritage, one of the functions that is often called on \texttt{id_tbl} and \texttt{ts_tbl} objects is base S3 generic \texttt{[base::[()]}. As this function is capable of modifying the object in a way that makes it incompatible with attached meta data, an attempt is made at preserving as much as possible and if all fails, a \texttt{data.table} object is returned instead of an object inheriting form \texttt{id_tbl}. If for example the index column is removed (or modified in a way that makes it incompatible with the interval specification) from a \texttt{ts_tbl}, an \texttt{id_tbl} is returned. If however the ID column is removed the only sensible thing to return is a \texttt{data.table} (see examples).

Examples

```r
tbl <- id_tbl(a = 1:10, b = rnorm(10))
is_id_tbl(tbl)
is_ts_tbl(tbl)

dat <- data.frame(a = 1:10, b = hours(1:10), c = rnorm(10))
tbl <- as_ts_tbl(dat, "a")
is_id_tbl(tbl)
is_ts_tbl(tbl)

tmp <- as_id_tbl(tbl)
is_ts_tbl(tbl)
is_ts_tbl(tmp)

tmp <- as_id_tbl(tbl, by_ref = TRUE)
is_ts_tbl(tbl)
is_ts_tbl(tmp)

tbl <- id_tbl(a = 1:10, b = rnorm(10))
names(tbl) <- c("c", "b")
tbl

tbl <- id_tbl(a = 1:10, b = rnorm(10))
```

id_vars

validate_tbl(data.table::setnames(tbl, c("c", "b")))

tbl <- id_tbl(a = 1:10, b = rnorm(10))
validate_tbl(rename_cols(tbl, c("c", "b")))

tbl <- ts_tbl(a = rep(1:2, each = 5), b = hours(rep(1:5, 2)), c = rnorm(10))
tbl[, c("a", "c"), with = FALSE]
tbl[, c("b", "c"), with = FALSE]
tbl[, list(a, b = as.double(b), c)]

id_vars  ICU class meta data utilities

Description

The two data classes id_tbl and ts_tbl, used by ricu to represent ICU patient data, consist of a data.table alongside some meta data. This includes marking columns that have special meaning and for data representing measurements ordered in time, the step size. The following utility functions can be used to extract columns and column names with special meaning, as well as query a ts_tbl object regarding its time series related meta data.

Usage

id_vars(x)
id_var(x)
id_col(x)
index_var(x)
index_col(x)
meta_vars(x)
data_vars(x)
data_var(x)
data_col(x)
interval(x)
time_unit(x)
time_step(x)
time_vars(x)
Arguments

x

Object to query

Details

The following functions can be used to query an object for columns or column names that represent a distinct aspect of the data:

• **id_vars()**: ID variables are one or more column names with the interaction of corresponding columns identifying a grouping of the data. Most commonly this is some sort of patient identifier.

• **id_var()**: This function either fails or returns a string and can therefore be used in case only a single column provides grouping information.

• **id_col()**: Again, in case only a single column provides grouping information, this column can be extracted using this function.

• **index_var()**: Suitable for use as index variable is a column that encodes a temporal ordering of observations as *difftime* vector. Only a single column can be marked as index variable and this function queries a ts_tbl object for its name.

• **index_col()**: Similarly to **id_col()**, this function extracts the column with the given designation. As a ts_tbl object is required to have exactly one column marked as index, this function always returns for ts_tbl objects (and fails for id_tbl objects).

• **meta_vars()**: For ts_tbl objects, meta variables represent the union of ID and index variables, while for id_tbl objects meta variables consist pf ID variables.

• **data_vars()**: Data variables on the other hand are all columns that are not meta variables.

• **data_var()**: Similarly to **id_var()**, this function either returns the name of a single data variable or fails.

• **data_col()**: Building on **data_var()**, in situations where only a single data variable is present, it is returned or if multiple data column exists, an error is thrown.

• **time_vars()**: Time variables are all columns in an object inheriting from *data.frame* that are of type *difftime*. Therefore in a ts_tbl object the index column is one of (potentially) several time variables.

• **interval()**: The time series interval length is represented a scalar valued *difftime* object.

• **time_unit()**: The time unit of the time series interval, represented by a string such as "hours" or "mins" (see *difftime*).

• **time_step()**: The time series step size represented by a numeric value in the unit as returned by **time_unit()**.

Value

Mostly column names as character vectors, in case of **id_var()**, **index_var()**, **data_var()** and **time_unit()** of length 1, else of variable length. Functions **id_col()**, **index_col()** and **data_col()** return table columns as vectors, while **interval()** returns a scalar valued *difftime* object and **time_step()** a number.
Examples

```r
tbl <- id_tbl(a = rep(1:2, each = 5), b = rep(1:5, 2), c = rnorm(10),
               id_vars = c("a", "b"))

id_vars(tbl)
tryCatch(id_col(tbl), error = function(...) "no luck")
data_vars(tbl)
data_col(tbl)

tmp <- as_id_tbl(tbl, id_vars = "a")
id_vars(tmp)
id_col(tmp)

tbl <- ts_tbl(a = rep(1:2, each = 5), b = hours(rep(1:5, 2)), c = rnorm(10))
index_var(tbl)
index_col(tbl)

identical(index_var(tbl), time_vars(tbl))

interval(tbl)
time_unit(tbl)
time_step(tbl)
```

---

**import_src**  
**Data import utilities**

**Description**

Making a dataset available to ricu consists of 3 steps: downloading (`download_src()`), importing (`import_src()`) and attaching (`attach_src()`). While downloading and importing are one-time procedures, attaching of the dataset is repeated every time the package is loaded. Briefly, downloading loads the raw dataset from the internet (most likely in `.csv` format), importing consists of some preprocessing to make the data available more efficiently and attaching sets up the data for use by the package.

**Usage**

```r
import_src(x, ...)
```

## S3 method for class 'src_cfg'

```r
import_src(x, data_dir = src_data_dir(x), force = FALSE, ...)
```

```r
import_tbl(x, ...)
```

## S3 method for class 'tbl_cfg'

```r
import_tbl(x, data_dir = src_data_dir(x), progress = NULL, ...)
```
import_src

Arguments

- **x**
  - Object specifying the source configuration
  - Passed to downstream methods (finally to `readr::read_csv/readr::read_csv_chunked`/generic consistency

- **data_dir**
  - The directory where the data was downloaded to (see `download_src()`).

- **force**
  - Logical flag indicating whether to overwrite already imported tables

- **progress**
  - Either NULL or a progress bar as created by `progress::progress_bar()`

Details

In order to speed up data access operations, ricu does not directly use the PhysioNet provided CSV files, but converts all data to `fst::fst()` format, which allows for random row and column access. Large tables are split into chunks in order to keep memory requirements reasonably low.

The one-time step per dataset of data import is fairly resource intensive: depending on CPU and available storage system, it will take on the order of an hour to run to completion and depending on the dataset, somewhere between 50 GB and 75 GB of temporary disk space are required as tables are uncompressed, in case of partitioned data, rows are reordered and the data again is saved to a storage efficient format.

The S3 generic function `import_src()` performs import of an entire data source, internally calling the S3 generic function `import_tbl()` in order to perform import of individual tables. Method dispatch is intended to occur on objects inheriting from `src_cfg` and `tbl_cfg` respectively. Such objects can be generated from JSON based configuration files which contain information such as table names, column types or row numbers, in order to provide safety in parsing of `.csv` files. For more information on data source configuration, refer to `load_src_cfg()`.

Current import capabilities include re-saving a `.csv` file to `.fst` at once (used for smaller sized tables), reading a large `.csv` file using the `readr::read_csv_chunked()` API, while partitioning chunks and reassembling sub-partitions (used for splitting a large file into partitions), as well as re-partitioning an already partitioned table according to a new partitioning scheme. Care has been taken to keep the maximal memory requirements for this reasonably low, such that data import is feasible on laptop class hardware.

Value

Called for side effects and returns NULL invisibly.

Examples

```r
## Not run:

dir <- tempdir()
list.files(dir)
download_src("mimic_demo", dir)
list.files(dir)
import_src("mimic_demo", dir)
list.files(dir)
```
load_concepts

unlink(dir, recursive = TRUE)

## End(Not run)

load_concepts  Load concept data

Description

Concept objects are used in ricu as a way to specify how a clinical concept, such as heart rate can be loaded from a data source. Building on this abstraction, load_concepts() powers concise loading of data with data source specific pre-processing hidden away from the user, thereby providing a data source agnostic interface to data loading. At default value of the argument merge_data, a tabular data structure (either a ts_tbl or an id_tbl, depending on what kind of concepts are requested), inheriting from data.table, is returned, representing the data in wide format (i.e. returning concepts as columns).

Usage

load_concepts(x, ..., cache = FALSE)

## S3 method for class 'character'
load_concepts(
  x,
  src = NULL,
  concepts = NULL,
  ..., 
  dict_name = "concept-dict",
  dict_dirs = NULL
)

## S3 method for class 'concept'
load_concepts(
  x,
  src = NULL,
  aggregate = NULL,
  merge_data = TRUE,
  verbose = TRUE,
  ..., 
  cache = FALSE
)

## S3 method for class 'cncpt'
load_concepts(x, aggregate = NULL, ..., progress = NULL)
## S3 method for class 'num_cncpt'
load_concepts(x, aggregate = NULL, ..., progress = NULL)

## S3 method for class 'fct_cncpt'
load_concepts(x, aggregate = NULL, ..., progress = NULL)

## S3 method for class 'lgl_cncpt'
load_concepts(x, aggregate = NULL, ..., progress = NULL)

## S3 method for class 'rec_cncpt'
load_concepts(x, aggregate = NULL, patient_ids = NULL, id_type = "icustay", interval = hours(1L), ..., progress = NULL, cache = FALSE)

## S3 method for class 'item'
load_concepts(x, patient_ids = NULL, id_type = "icustay", interval = hours(1L), progress = NULL, ...)

## S3 method for class 'itm'
load_concepts(x, patient_ids = NULL, id_type = "icustay", interval = hours(1L), ...

### Arguments

- **x**: Object specifying the data to be loaded
- **...**: Passed to downstream methods
- **cache**: Logical flag indicating whether to cache concepts that are required multiple times (experimental)
- **src**: A character vector, used to subset the concepts; NULL means no subsetting
The concepts to be used or NULL in which case \texttt{load_dictionary()} is called. In case not concepts are passed as concepts, these are forwarded to \texttt{load_dictionary()} as name and file arguments. Controls how data within concepts is aggregated. Logical flag, specifying whether to merge concepts into wide format or return a list, each entry corresponding to a concept. Logical flag for muting informational output. Either NULL, or a progress bar object as created by \texttt{progress::progress\_bar}. Optional vector of patient ids to subset the fetched data with. String specifying the patient id type to return. The time interval used to discretize time stamps with, specified as \texttt{base::difftime()} object.

Details

In order to allow for a large degree of flexibility (and extensibility), which is much needed owing to considerable heterogeneity presented by different data sources, several nested S3 classes are involved in representing a concept and \texttt{load\_concepts()} follows this hierarchy of classes recursively when resolving a concept. An outline of this hierarchy can be described as

- concept: contains many \texttt{cncpt} objects (of potentially differing sub-types), each comprising of some meta-data and an \texttt{item} object.
- item: contains many \texttt{itm} objects (of potentially differing sub-types), each encoding how to retrieve a data item.

The design choice for wrapping a vector of \texttt{cncpt} objects with a container class \texttt{concept} is motivated by the requirement of having several different sub-types of \texttt{cncpt} objects (all inheriting from the parent type \texttt{cncpt}), while retaining control over how this homogeneous w.r.t. parent type, but heterogeneous w.r.t. sub-type vector of objects behaves in terms of S3 generic functions.

Value

An \texttt{id\_tbl/ts\_tbl} or a list thereof, depending on loaded concepts and the value passed as \texttt{merge\_data}.

Concept

Top-level entry points are either a character vector, which is used to subset a concept object or an entire \texttt{concept dictionary}, or a concept object. When passing a character vector as first argument, the most important further arguments at that level control from where the dictionary is taken (\texttt{dict\_name} or \texttt{dict\_dirs}). At concept level, the most important additional arguments control the result structure: data merging can be disabled using \texttt{merge\_data} and data aggregation is governed by the aggregate argument.

Data aggregation is important for merging several concepts into a wide-format table, as this requires data to be unique per observation (i.e. by either id or combination of id and index). Several value types are acceptable as aggregate argument, the most important being \texttt{FALSE}, which disables aggregation, \texttt{NULL}, which auto-determines a suitable aggregation function or a string which
is ultimately passed to `dt_gforce()` where it identifies a function such as `sum()`, `mean()`, `min()` or `max()`. More information on aggregation is available as `aggregate()`. If the object passed as `aggregate` is scalar, it is applied to all requested concepts in the same way. In order to customize aggregation per concept, a named object (with names corresponding to concepts) of the same length as the number of requested concepts may be passed.

Under the hood, a concept object comprises of several `cncept` objects with varying sub-types (for example `num_cncept`, representing continuous numeric data or `fct_cncept` representing categorical data). This implementation detail is of no further importance for understanding concept loading and for more information, please refer to the `concept` documentation. The only argument that is introduced at `cncept` level is `progress`, which controls progress reporting. If called directly, the default value of `NULL` yields messages, sent to the terminal. Internally, if called from `load_concepts()` at concept level (with `verbose` set to `TRUE`), a `progress::progress_bar` is set up in a way that allows nested messages to be captured and not interrupt progress reporting (see `msg_progress()`).

**Item**

A single `cncept` object contains an `item` object, which in turn is composed of several `itm` objects with varying sub-types, the relationship `item` to `itm` being that of `concept` to `cncept` and the rationale for this implementation choice is the same as previously: a container class used representing a vector of objects of varying sub-types, all inheriting form a common super-type. For more information on the `item` class, please refer to the relevant documentation. Arguments introduced at `item` level include `patient_ids`, `id_type` and `interval`. Acceptable values for `interval` are scalar-valued `base::difftime()` objects (see also helper functions such as `hours()`) and this argument essentially controls the time-resolution of the returned time-series. Of course, the limiting factor raw time resolution which is on the order of hours for data sets like MIMIC-III or eICU but can be much higher for a data set like HiRID. The argument `id_type` is used to specify what kind of id system should be used to identify different time series in the returned data. A data set like MIMIC-III, for example, makes possible the resolution of data to 3 nested ID systems:

- `patient(subject_id)`: identifies a person
- `hadm(hadm_id)`: identifies a hospital admission (several of which are possible for a given person)
- `icustay(icustay_id)`: identifies an admission to an ICU and again has a one-to-many relationship to hadm.

Acceptable argument values are strings that match ID systems as specified by the data source configuration. Finally, `patient_ids` is used to define a patient cohort for which data can be requested. Values may either be a vector of IDs (which are assumed to be of the same type as specified by the `id_type` argument) or a tabular object inheriting from `data.frame`, which must contain a column named after the data set-specific ID system identifier (for MIMIC-III and an `id_type` argument of `hadm`, for example, that would be `hadm_id`).

**Extensions**

The presented hierarchy of S3 classes is designed with extensibility in mind: while the current range of functionality covers settings encountered when dealing with the included concepts and datasets, further data sets and/or clinical concepts might necessitate different behavior for data loading. For this reason, various parts in the cascade of calls to `load_concepts()` can be adapted for new requirements by defining new sub-classes to `cncept` or `itm` and providing methods for the
generic function `load_concepts()` specific to these new classes. At `cncept` level, method dispatch defaults to `load_concepts.cncept()` if no method specific to the new class is provided, while at `itm` level, no default function is available.

Roughly speaking, the semantics for the two functions are as follows:

- **cncept**: Called with arguments `x` (the current `cncept` object), `aggregate` (controlling how aggregation per time-point and ID is handled), ... (further arguments passed to downstream methods) and `progress` (controlling progress reporting), this function should be able to load and aggregate data for the given concept. Usually this involves extracting the `itm` object and calling `load_concepts()` again, dispatching on the `itm` class with arguments `x` (the given `itm`), arguments passed as ..., as well as `progress`.

- **itm**: Called with arguments `x` (the current object inheriting from `itm`, `patient_ids` (NULL or a patient ID selection), `id_type` (a string specifying what ID system to retrieve), and `interval` (the time series interval), this function actually carries out the loading of individual data items, using the specified ID system, rounding times to the correct interval and subsetting on patient IDs. As return value, on object of class as specified by the `target` entry is expected and all `data_vars()` should be named consistently, as data corresponding to multiple `itm` objects concatenated in row-wise fashion as in `base::rbind()`.

**Examples**

```r
if (require(mimic.demo)) {
  dat <- load_concepts("glu", "mimic_demo")
  gluc <- concept("gluc",
      item("mimic_demo", "labevents", "itemid", list(c(50809L, 50931L)))
  )
  identical(load_concepts(gluc), dat)
  class(dat)
  class(load_concepts(c("sex", "age"), "mimic_demo"))
}
```

---

**load_dictionary**

**Load concept dictionaries**

**Description**

Data concepts can be specified in JSON format as a concept dictionary which can be read and parsed into `concept/item` objects. Dictionary loading can either be performed on the default included dictionary or on a user-specified custom dictionary. Furthermore, a mechanism is provided for adding concepts and/or data sources to the existing dictionary (see the Details section).
Usage
load_dictionary(
  src = NULL,
  concepts = NULL,
  name = "concept-dict",
  cfg_dirs = NULL
)

Arguments
src NULL or the name of one or several data sources
concepts A character vector used to subset the concept dictionary or NULL indicating no
          subsetting
name Name of the dictionary to be read
cfg_dirs File name of the dictionary

Details
A default dictionary is provided at

system.file(
  file.path("extdata", "config", "concept-dict.json"),
  package = "ricu"
)

and can be loaded into an R session by calling get_config("concept-dict"). The default dictionary can be extended by adding a file concept-dict.json to the path specified by the environment variable RICU_CONFIG_PATH. New concepts can be added to this file and existing concepts can be extended (by adding new data sources). Alternatively, load_dictionary() can be called on non-default dictionaries using the file argument.

In order to specify a concept as JSON object, for example the numeric concept for glucose, is given by

{  
  "glu": {
      "unit": "mg/dL",
      "min": 0,
      "max": 1000,
      "description": "glucose",
      "category": "chemistry",
      "sources": {
        "mimic_demo": [
          {
            "ids": [50809, 50931],
            "table": "labevents",
            "sub_var": "itemid"
          }
        ]
      }
  }
}
Using such a specification, constructors for `cncpt` and `itm` objects are called either using default arguments or as specified by the JSON object, with the above corresponding to a call like

```r
concept(
    name = "glu",
    items = item(
        src = "mimic_demo", table = "labevents", sub_var = "itemid",
        ids = list(c(50809L, 50931L))
    ),
    description = "glucose", category = "chemistry",
    unit = "mg/dL", min = 0, max = 1000
)
```

The arguments `src` and `concepts` can be used to only load a subset of a dictionary by specifying a character vector of data sources and/or concept names.

**Value**

A concept object containing several data concepts as `cncpt` objects.

**Examples**

```r
if (require(mimic.demo)) {
    head(load_dictionary("mimic_demo"))
    load_dictionary("mimic_demo", c("glu", "lact"))
}
```

---

**load_id**

*Load data as id_tbl or ts_tbl objects*

**Description**

Building on functionality provided by `load_src()` and `load_difftime()`, `load_id()` and `load_ts()` load data from disk as id_tbl and ts_tbl objects respectively. Over `load_difftime()` both `load_id()` and `load_ts()` provide a way to specify `meta_vars()` (as `id_var` and `index_var` arguments), as well as an interval size (as interval argument) for time series data.
Usage

load_id(x, ...)

## S3 method for class 'src_tbl'
load_id(
  x,
  rows,
  cols = colnames(x),
  id_var = id_vars(x),
  interval = hours(1L),
  time_vars = ricu::time_vars(x),
  ...
)

## S3 method for class 'character'
load_id(x, src, ...)

## S3 method for class 'itm'
load_id(x, cols = colnames(x), id_var = id_vars(x), ...)

## S3 method for class 'fun_itm'
load_id(x, ...)

## Default S3 method:
load_id(x, ...)

load_ts(x, ...)

## S3 method for class 'src_tbl'
load_ts(
  x,
  rows,
  cols = colnames(x),
  id_var = id_vars(x),
  index_var = ricu::index_var(x),
  interval = hours(1L),
  time_vars = ricu::time_vars(x),
  ...
)

## S3 method for class 'character'
load_ts(x, src, ...)

## S3 method for class 'itm'
load_ts(
  x,
  cols = colnames(x),
  id_var = id_vars(x),
  id_var = id_vars(x),
index_var = ricu::index_var(x),
...

## S3 method for class 'fun_itm'
load_ts(x, ...)

## Default S3 method:
load_ts(x, ...)

**Arguments**

- **x**: Object for which to load data
- **...**: Generic consistency
- **rows**: Expression used for row subsetting (NSE)
- **cols**: Character vector of column names
- **id_var**: The column defining the id of ts_tbl and id_tbl objects
- **interval**: The time interval used to discretize time stamps with, specified as `base::difftime()` object
- **time_vars**: Character vector enumerating the columns to be treated as timestamps and thus returned as `base::difftime()` vectors
- **src**: Passed to `as_src_tbl()` in order to determine the data source
- **index_var**: The column defining the index of ts_tbl objects

**Details**

While for `load_difftime()` the ID variable can be suggested, the function only returns a best effort at fulfilling this request. In some cases, where the data does not allow for the desired ID type, data is returned using the ID system (among all available ones for the given table) with highest cardinality. Both `load_id()` and `load_ts()` are guaranteed to return an object with `id-vars()` set as requested by the `id_var` argument. Internally, the change of ID system is performed by `change_id()`.

Additionally, while times returned by `load_difftime()` are in 1 minute resolution, the time series step size can be specified by the `interval` argument when calling `load_id()` or `load_ts()`. This rounding and potential change of time unit is performed by `change_interval()` on all columns specified by the `time_vars` argument. All time stamps are relative to the origin provided by the ID system. This means that for an `id_var` corresponding to hospital IDs, times are relative to hospital admission.

When `load_id()` (or `load_ts()`) is called on `itm` objects instead of `src_tbl` (or objects that can be coerced to `src_tbl`), The row-subsetting is performed according the the specification as provided by the `itm` object. Furthermore, at default settings, columns are returned as required by the `itm` object and `id_var` (as well as `index_var`) are set accordingly if specified by the `itm` or set to default values for the given `src_tbl` object if not explicitly specified.

**Value**

An `id_tbl` or a `ts_tbl` object.
Examples

```r
if (require(mimic.demo)) {
  load_id("admissions", "mimic_demo", cols = "admission_type")

  dat <- load_ts(mimic_demo$labevents, itemid %in% c(50809L, 50931L),
                  cols = c("itemid", "valuenum"))

  glu <- new_itm(src = "mimic_demo", table = "labevents",
                  sub_var = "itemid", ids = c(50809L, 50931L))

  identical(load_ts(glu), dat)
}
```

---

**load_src**

*Low level functions for loading data*

**Description**

Data loading involves a cascade of S3 generic functions, which can individually be adapted to the specifics of individual data sources. At the lowest level, `load_scr` is called, followed by `load_difftime()`. Functions up the chain, are described in `load_id()`.

**Usage**

```r
load_src(x, ...)
```

## S3 method for class 'src_tbl'

```r
load_src(x, rows, cols = colnames(x), ...)
```

## S3 method for class 'character'

```r
load_src(x, src, ...)
```

```r
load_difftime(x, ...)
```

## S3 method for class 'mimic_tbl'

```r
load_difftime(
  x,
  rows,
  cols = colnames(x),
  id_hint = id_vars(x),
  time_vars = ricu::time_vars(x),
  ...
)
```

## S3 method for class 'eicu_tbl'

```r
load_difftime(
```
load_src

x,
rows,
cols = colnames(x),
id_hint = id_vars(x),
time_vars = ricu::time_vars(x),
...
)

## S3 method for class 'hirid_tbl'
load_difftime(
  x,
  rows,
cols = colnames(x),
id_hint = id_vars(x),
time_vars = ricu::time_vars(x),
...
)

## S3 method for class 'character'
load_difftime(x, src, ...)

Arguments

x Object for which to load data
...
Generic consistency
rows Expression used for row subsetting (NSE)
cols Character vector of column names
src Passed to as_src_tbl() in order to determine the data source
id_hint String valued id column selection (not necessarily honored)
time_vars Character vector enumerating the columns to be treated as timestamps and thus returned as base::difftime() vectors

Details

A function extending the S3 generic load_src() is expected to load a subset of rows/columns from a tabular data source. While the column specification is provided as character vector of column names, the row subsetting involves non-standard evaluation (NSE). Data-sets that are included with ricu are represented by prt objects, which use rlang::eval_tidy() to evaluate NSE expressions. Furthermore, prt objects potentially represent tabular data split into partitions and row-subsetting expressions are evaluated per partition (see the part_safe flag in prt::subset.prt()). The return value of load_src() is expected to be of type data.table.

Timestamps are represented differently among the included data sources: while MIMIC-III and HiRID use absolute date/times, eICU provides temporal information as minutes relative to ICU admission. Other data sources, such as the ICU dataset provided by Amsterdam UMC, opt for relative times as well, but not in minutes since admission, but in milliseconds. In order to smoothen out such discrepancies, the next function in the data loading hierarchy is load_difftime(). This function is expected to call load_src() in order to load a subset of rows/columns from a table.
stored on disk and convert all columns that represent timestamps (as specified by the argument `time_vars`) into `base::difftime()` vectors using `mins` as time unit.

The returned object should be of type `id_tbl`, with the ID vars identifying the ID system the times are relative to. If for example all times are relative to ICU admission, the ICU stay ID should be returned as ID column. The argument `id_hint` may suggest an ID type, but if in the raw data, this ID is not available, `load_difftime()` may return data using a different ID system. In MIMIC-III, for example, data in the `labevents` table is available for `subject_id` (patient ID) or `hadm_id` (hospital admission ID). If data is requested for `icustay_id` (ICU stay ID), this request cannot be fulfilled and data is returned using the ID system with the highest cardinality (among the available ones). Utilities such as `change_id()` can the later be used to resolve data to `icustay_id`.

**Value**

A `data.table` object.

**Examples**

```r
if (require(mimic.demo)) {
  tbl <- mimic_demo$labevents
  col <- c("charttime", "value")

  load_src(tbl, itemid == 50809)
  colnames(
    load_src("labevents", "mimic_demo", itemid == 50809, cols = col)
  )

  load_difftime(tbl, itemid == 50809)
  colnames(
    load_difftime(tbl, itemid == 50809, col)
  )

  id_vars(
    load_difftime(tbl, itemid == 50809, id_hint = "icustay_id")
  )

  id_vars(
    load_difftime(tbl, itemid == 50809, id_hint = "subject_id")
  )
}
```
Description

For a data source to be accessible by ricu, a configuration object inheriting from the S3 class src_cfg is required. Such objects can be generated from JSON based configuration files, using load_src_cfg(). Information encoded by this configuration object includes available ID systems (mainly for use in change_id()), default column names per table for columns with special meaning (such as index column, value columns, unit columns, etc.), as well as a specification used for initial setup of the dataset which includes file names and column names alongside their data types.

Usage

load_src_cfg(src = NULL, name = "data-sources", cfg_dirs = NULL)

Arguments

src (Optional) name(s) of data sources used for subsetting
name String valued name of a config file which will be looked up in the default config directors
cfg_dirs Additional directory/ies to look for configuration files

Details

Configuration files are looked for as files name with added suffix .json starting with the directory (or directories) supplied as cfg_dirs argument, followed by the directory specified by the environment variable RICU_CONFIG_PATH, and finally in extdata/config of the package install directory. If files with matching names are found in multiple places they are concatenated such that in cases of name clashes, the earlier hits take precedent over the later ones. The following JSON code blocks show excerpts of the config file available at

system.file("extdata", "config", "data-sources.json", package = "ricu")

A data source configuration entry in a config file starts with a name, followed by optional entries class_prefix and further (variable) key-value pairs, such as an URL. For more information on class_prefix, please refer to the end of this section. Further entries include id_cfg and tables which are explained in more detail below. As outline, this gives for the data source mimic_demo, the following JSON object:

{
  "name": "mimic_demo",
  "class_prefix": ["mimic_demo", "mimic"],
  "url": "https://physionet.org/files/mimiciii-demo/1.4",
  "id_cfg": {
    ...
  },
  "tables": {
    ...
  }
}
The id_cfg entry is used to specify the available ID systems for a data source and how they relate to each other. An ID system within the context of r icu is a patient identifier of which typically several are present in a data set. In MIMIC-III, for example, three ID systems are available: patient IDs (subject_id), hospital admission IDs (hadm_id) and ICU stay IDs (icustay_id). Furthermore there is a one-to-many relationship between subject_id and hadm_id, as well as between hadm_id and icustay_id. Required for defining an ID system are a name, a position entry which orders the ID systems by their cardinality, a table entry, alongside column specifications id, start and end, which define how the IDs themselves, combined with start and end times can be loaded from a table. This gives the following specification for the ICU stay ID system in MIMIC-III:

```json
{
    "icustay": {
        "id": "icustay_id",
        "position": 3,
        "start": "intime",
        "end": "outtime",
        "table": "icustays"
    }
}
```

Tables are defined by a name and entries files, defaults, and cols, as well as optional entries num_rows and partitioning. As files entry, a character vector of file names is expected. For all of MIMIC-III a single .csv file corresponds to a table, but for example for HiRID, some tables are distributed in partitions. The defaults entry consists of key-value pairs, identifying columns in a table with special meaning, such as the default index column or the set of all columns that represent timestamps. This gives as an example for a table entry for the chartevents table in MIMIC-III a JSON object like:

```json
{
    "chartevents": {
        "files": "CHARTEVENTS.csv.gz",
        "defaults": {
            "index_var": "charttime",
            "val_var": "valuenum",
            "unit_var": "valueuom",
            "time_vars": ["charttime", "storetime"]
        },
        "num_rows": 330712483,
        "cols": {
            ...
        },
        "partitioning": {
            "col": "itemid",
            "breaks": [127, 210, 425, 549, 643, 741, 1483, 3458, 3695, 8440,
                       8553, 220274, 223921, 224085, 224859, 227629]
        }
    }
}
```
The optional `num_rows` entry is used when importing data (see `import_src()`) as a sanity check, which is not performed if this entry is missing from the data source configuration. The remaining table entry, `partitioning`, is optional in the sense that if it is missing, the table is not partitioned and if it is present, the table will be partitioned accordingly when being imported (see `import_src()`). In order to specify a partitioning, two entries are required, `col` and `breaks`, where the former denotes a column and the latter a numeric vector which is used to construct intervals according to which `col` is binned. As such, currently `col` is required to be of numeric type. A partitioning entry as in the example above will assign rows corresponding to `idemid` 1 through 126 to partition 1, 127 through 209 to partition 2 and so on up to partition 17.

Column specifications consist of a name and a `spec` entry alongside a name which determines the column name that will be used by `ricu`. The `spec` entry is expected to be the name of a column specification function of the `readr` package (see `readr::cols()`) and all further entries in a `cols` object are used as arguments to the `readr` column specification. For the admissions table of MIMIC-III the columns `hadm_id` and `admittime` are represented by:

```json
{
  ...
  "hadm_id": {
    "name": "HADM_ID",
    "spec": "col_integer"
  },
  "admittime": {
    "name": "ADMITTIME",
    "spec": "col_datetime",
    "format": "%Y-%m-%d %H:%M:%S"
  },
  ...
}
```

Internally, a `src_cfg` object consist of further S3 classes, which are instantiated when loading a JSON source configuration file. Functions for creating and manipulating `src_cfg` and related objects are marked `internal` but a brief overview is given here nevertheless:

- `src_cfg`: wraps objects `id_cfg`, `col_cfg` and optionally `tbl_cfg`
- `id_cfg`: contains information in ID systems and is created from `id_cfg` entries in config files
- `col_cfg`: contains column default settings represented by `defaults` entries in table configuration blocks
- `tbl_cfg`: used when importing data and therefore encompasses information in `files`, `num_rows` and `cols` entries of table configuration blocks

A `src_cfg` can be instantiated without corresponding `tbl_cfg` but consequently cannot be used for data import (see `import_src()`). In that sense, table config entries `files` and `cols` are optional as well with the restriction that the data source has to be already available in `.fst` format

An example for such a slimmed down config file is available at

```r
system.file("extdata", "config", "demo-sources.json", package = "ricu")
```
The class_prefix entry in a data source configuration is used to create subclasses to src_cfg, id_cfg, col_cfg and tbl_cfg classes and passed on to constructors of src_env (new_src_env()) and src_tbl new_src_tbl() objects. As an example, for the above class_prefix value of c("mimic_demo", "mimic"), the corresponding src_cfg will be assigned classes c("mimic_demo_cfg", "mimic_cfg", "src_cfg") and consequently the src_tbl objects will inherit from "mimic_demo_tbl", "mimic_tbl" and "src_tbl". This can be used to adapt the behavior of involved S3 generic function to specifics of the different data sources. An example for this is how load_difftime() uses these sub-classes to smoothen out different time-stamp representations. Furthermore, such a design was chosen with extensibility in mind. Currently, download_src() is designed around data sources hosted on PhysioNet, but in order to include a dataset external to PhysioNet, the download_src() generic can simply be extended for the new class.

Value

A list of data source configurations as src_cfg objects.

Examples

cfg <- load_src_cfg("mimic_demo")
str(cfg, max.level = 1L)
cfg <- cfg["mimic_demo"]
str(cfg, max.level = 1L)

cols <- as_col_cfg(cfg)
index_var(cols["chartevents"])
time_vars(cols["chartevents"])

as_id_cfg(cfg)

min_or_na

Utility functions

Description

Several utility functions exported for convenience.

Usage

min_or_na(x)

max_or_na(x)

is_val(x, val)

not_val(x, val)

is_true(x)
is_false(x)

last_elem(x)

first_elem(x)

replace_na(x, val, ...)

Arguments

x
Object to use

val
Value to compare against

... Forwarded to other methods

Details

The two functions `min_or_na()` and `max_or_na()` overcome a design choice of `base::min()` (or `base::max()`) that can yield undesirable results. If called on a vector of all missing values with `na.rm = TRUE`, Inf (and -Inf respectively) are returned. This is changed to returning a missing value of the same type as `x`.

The functions `is_val()` and `not_val()` (as well as analogously `is_true()` and `is_false()`) return logical vectors of the same length as the value passed as `x`, with non-base R semanticists of comparing against `NA`: instead of returning `c(NA, TRUE)` for `c(NA, 5) == 5`, `is_val()` will return `c(FALSE, TRUE)`. Passing `NA` as `val` might lead to unintended results but no warning is thrown.

Finally, `first_elem()` and `last_elem()` has the same semantics as `utils::head()` and `utils::tail()` with `n = 1L` and `replace_na()` will replace all occurrences of `NA` in `x` with `val` and can be called on both objects inheriting from `data.table` in which case internally `data.table::setnafill()` is called or other objects.

Value

- `min_or_na()`/`max_or_na()`: scalar-valued extrema of a vector
- `is_val()`/`not_val()`/`is_true()`/`is_false()`: Logical vector of the same length as the object passed as `x`
- `first_elem()`/`last_elem()`: single element of the object passed as `x`
- `replace_na()`/`last_elem()`: modified version of the object passed as `x`

Examples

```r
some_na <- c(NA, sample(1:10, 5), NA)
identical(min(some_na, na.rm = TRUE), min_or_na(some_na))

all_na <- rep(NA, 5)
min(all_na, na.rm = TRUE)
min_or_na(all_na)

is_val(some_na, 5)
some_na == 5
```
is_val(some_na, NA)

identical(first_elem(letters), head(letters, n = 1L))
identical(last_elem(letters), tail(letters, n = 1L))

replace_na(some_na, 11)
replace_na(all_na, 11)
replace_na(1:5, 11)

tbl <- ts_tbl(a = 1:10, b = hours(1:10), c = c(NA, 1:5, NA, 8:9, NA))
res <- replace_na(tbl, 0)
identical(tbl, res)

---

**msg_progress**

*Message signaling nested with progress reporting*

### Description

In order to not interrupt progress reporting by a `progress::progress_bar`, messages are wrapped with class `msg_progress` which causes them to be captured printed after progress bar completion. This function is intended to be used when signaling messages in callback functions.

### Usage

```r
msg_progress(..., envir = parent.frame())
```

### Arguments

- `...` Passed to `base::makeMessage()`
- `envir` Passed to `glue::glue()`.

### Value

Called for side effects and returns `NULL` invisibly.

### Examples

```r
msg_progress("Foo", "bar")

capt_fun <- function(x) {
  message("captured: ", conditionMessage(x))
}

tryCatch(msg_progress("Foo", "bar"), msg_progress = capt_fun)
```
**new_cncpt**

---

**Data Concepts**

**Description**

Concept objects are used in ricu as a way to specify how a clinical concept, such as heart rate can be loaded from a data source and are mainly consumed by `load_concepts()`. Several functions are available for constructing concept (and related auxiliary) objects either from code or by parsing a JSON formatted concept dictionary using `load_dictionary()`.

**Usage**

```r
new_cncpt(
  name,
  items,
  description = NA_character_,
  category = NA_character_,
  aggregate = NULL,
  ..., 
  target = "ts_tbl",
  class = "num_cncpt"
)
```

```r
is_cncpt(x)
```

```r
init_cncpt(x, ...)
```

```r
## S3 method for class 'num_cncpt'
init_cncpt(x, unit = NULL, min = NULL, max = NULL, ...)
```

```r
## S3 method for class 'fct_cncpt'
init_cncpt(x, levels, ...)
```

```r
## S3 method for class 'cncpt'
init_cncpt(x, ...)
```

```r
## S3 method for class 'rec_cncpt'
init_cncpt(x, callback = "identity_callback", interval = NULL, ...)
```

```r
new_concept(x)
```

```r
concept(...)
```

```r
is_concept(x)
```

```r
as_concept(x)
```
Arguments

name  The name of the concept
items  Zero or more itm objects
description  String-valued concept description
category  String-valued category
aggregate  NULL or a string denoting a function used to aggregate per id and if applicable per time step
...  Further specification of the cncpt object (passed to init_cncpt())
target  The target object yielded by loading
class  NULL or a string-valued sub-class name used for customizing concept behavior
x  Object to query/dispatch on
unit  A string, specifying the measurement unit of the concept (can be NULL)
min, max  Scalar valued; defines a range of plausible values for a numeric concept
levels  A vector of possible values a categorical concept may take on
callback  Name of a function to be called on the returned data used for data cleanup operations
interval  Time interval used for data loading; if NULL, the respective interval passed as argument to load_concepts() is taken

Details

In order to allow for a large degree of flexibility (and extensibility), which is much needed owing to considerable heterogeneity presented by different data sources, several nested S3 classes are involved in representing a concept. An outline of this hierarchy can be described as

• concept: contains many cncpt objects (of potentially differing sub-types), each comprising of some meta-data and an item object
• item: contains many itm objects (of potentially differing sub-types), each encoding how to retrieve a data item.

The design choice for wrapping a vector of cncpt objects with a container class concept is motivated by the requirement of having several different sub-types of cncpt objects (all inheriting from the parent type cncpt), while retaining control over how this homogeneous w.r.t. parent type, but heterogeneous w.r.t. sub-type vector of objects behaves in terms of S3 generic functions.

Each individual cncpt object contains the following information: a string-valued name, an item vector containing itm objects, a string-valued description (can be missing), a string-valued category designation (can be missing), a character vector-valued specification for an aggregation function and a target class specification (e.g. id_tbl or ts_tbl). Additionally, a sub-class to cncpt has to be specified, each representing a different data-scenario and holding further class-specific information. The following sub-classes to cncpt are available:

• num_cncpt: The most widely used concept type is indented for concepts representing numerical measurements. Additional information that can be specified includes a string-valued unit specification, alongside a plausible range which can be used during data loading.
• **fct_cncpt**: In case of categorical concepts, such as `sex`, a set of factor levels can be specified, against which the loaded data is checked.

• **lgl_cncpt**: A special case of `fct_cncpt`, this allows only for logical values (TRUE, FALSE and NA).

• **rec_cncpt**: More involved concepts, such as a SOFA score can pull in other concepts. Recursive concepts can build on other recursive concepts up to arbitrary recursion depth. Owing to the more complicated nature of such concepts, a callback function can be specified which is used in data loading for concept-specific post-processing steps.

Class instantiation is organized in the same fashion as for `item` objects: `concept()` maps vector-valued arguments to `new_cncpt()`, which internally calls the S3 generic function `init_cncpt()`, while `new_concept()` instantiates a concept object from a list of `cncpt` objects (created by calls to `new_cncpt()`). Coercion is only possible from list and `cncpt`, by calling `as_concept()` and inheritance can be checked using `is_concept()` or `is_cncpt()`.

**Value**

Constructors and coercion functions return `cncpt` and `concept` objects, while inheritance tester functions return logical flags.

**Examples**

```r
if (require(mimic.demo)) {
  gluc <- concept("glu",
    item("mimic_demo", "labevents", "itemid", list(c(50809L, 50931L))),
    description = "glucose", category = "chemistry",
    unit = "mg/dL", min = 0, max = 1000
  )

  is_concept(gluc)
  identical(gluc, load_dictionary("mimic_demo", "glu"))

  gl1 <- new_cncpt("glu",
    item("mimic_demo", "labevents", "itemid", list(c(50809L, 50931L))),
    description = "glucose"
  )

  is_cncpt(gl1)
  is_concept(gl1)

  conc <- concept(c("glu", "lact"),
    list(
      item("mimic_demo", "labevents", "itemid", list(c(50809L, 50931L))),
      item("mimic_demo", "labevents", "itemid", 50813L)
    ),
    description = c("glucose", "lactate")
  )

  conc
}
```
identical(as_concept(gl1), conc[l])

new_itm               Data items

Description
Item objects are used in ricu as a way to specify how individual data items corresponding to clinical concepts (see also concept()), such as heart rate can be loaded from a data source. Several functions are available for constructing item (and related auxiliary) objects either from code or by parsing a JSON formatted concept dictionary using load_dictionary().

Usage
new_itm(src, ..., target = NA_character_, class = "sel_itm")

is_itm(x)

init_itm(x, ...)

## S3 method for class 'sel_itm'
init_itm(x, table, sub_var, ids, callback = "identity_callback", ...)

## S3 method for class 'hrd_itm'
init_itm(x, table, sub_var, ids, callback = "identity_callback", ...)

## S3 method for class 'col_itm'
init_itm(x, table, unit_val = NULL, callback = "identity_callback", ...)

## S3 method for class 'rgx_itm'
init_itm(x, table, sub_var, regex, callback = "identity_callback", ...)

## S3 method for class 'fun_itm'
init_itm(x, callback, ...)

## S3 method for class 'itm'
init_itm(x, ...)

new_item(x)

item(...)

as_item(x)

is_item(x)
new_itm

Arguments

**src**  
The data source name

...  
Further specification of the itm object (passed to init_itm())

**target**  
Item target class (e.g. "id_tbl"), NA indicates no specific class requirement

**class**  
Sub class for customizing itm behavior

**x**  
Object to query/dispatch on

**table**  
Name of the table containing the data

**sub_var**  
Column name used for subsetting

**ids**  
Vector of ids used to subset table rows. If NULL, all rows are considered corresponding to the data item

**callback**  
Name of a function to be called on the returned data used for data cleanup operations (or a string that evaluates to a function)

**unit_val**  
String valued unit to be used in case no unit_var is available for the given table

**regex**  
String-valued regular expression which will be evaluated by base::grepl() with ignore.case = TRUE

Details

In order to allow for a large degree of flexibility (and extensibility), which is much needed owing to considerable heterogeneity presented by different data sources, several nested S3 classes are involved in representing a concept. An outline of this hierarchy can be described as

- **concept**: contains many cncpt objects (of potentially differing sub-types), each comprised of some meta-data and an item object
- **item**: contains many itm objects (of potentially differing sub-types), each encoding how to retrieve a data item.

The design choice for wrapping a vector of itm objects with a container class item is motivated by the requirement of having several different sub-types of itm objects (all inheriting from the parent type itm), while retaining control over how this homogeneous w.r.t. parent type, but heterogeneous w.r.t. sub-type vector of objects behaves in terms of S3 generic functions.

The following sub-classes to itm are available, each representing a different data-scenario:

- **sel_itm**: The most widely used item class is intended for the situation where rows of interest can be identified by looking for occurrences of a set of IDs (ids) in a column (sub_var). An example for this is heart rate hr on mimic, where the IDs 211 and 220045 are looked up in their emid column of chartevents'.
- **col_itm**: This item class can be used if no row-subsetting is required. An example for this is heart rate (hr) on eicu, where the table vitalperiodic contains an entire column dedicated to heart rate measurements.
- **rgx_itm**: As an alternative to the value-matching approach of sel_itm objects, this class identifies rows using regular expressions. Used for example for insulin in eicu, where the regular expression ^insulin (250.+)?((ml|units)/hr)?\$/ is matched against the drugname column of infusiondrug. The regular expression is evaluated by base::grepl() with ignore.case = TRUE.
fun_itm: Intended for the scenario where data of interest is not directly available from a table, this itm class offers most flexibility. A function can be specified as callback and this function will be called with arguments x (the object itself), patient_ids, id_type and interval (see load_concepts()) and is expected to return an object as specified by the target entry.

hrd_itm: A special case of sel_itm for HiRID data where measurement units are not available as separate column, but as separate table with units fixed per concept.

All itm objects have to specify a data source (src) as well as a sub-class. Further arguments then are specific to the respective sub-class and encode information that define data loading, such as the table to query, the column name and values to use for identifying relevant rows, etc. The S3 generic function init_itm() is responsible for input validation of class-specific arguments as well as class initialization. A list of itm objects, created by calls to new_itm() can be passed to new_item in order to instantiate an item object. An alternative constructor for item objects is given by item() which calls new_itm() on the passed arguments (see examples). Finally as_item() can be used for coercion of related objects such as list, concept, and the like. Several additional S3 generic functions exist for manipulation of item-like objects but are marked internal (see item/concept utilities).

Value

Constructors and coercion functions return itm and item objects, while inheritance tester functions return logical flags.

Examples

```r
if (require(mimic.demo)) {
  gluc <- item("mimic_demo", "labevents", "itemid", list(c(50809L, 50931L)), unit_var = TRUE, target = "ts_tbl")

  is_item(gluc)

  all.equal(gluc, as_item(load_dictionary("mimic_demo", "glu")))

  hr1 <- new_itm(src = "mimic_demo", table = "chartevents", sub_var = "itemid", ids = c(211L, 220045L))

  hr2 <- item(src = c("mimic_demo", "eicu_demo"),
              table = c("chartevents", "vitalperiodic"),
              sub_var = list("itemid", NULL),
              val_var = list(NULL, "heartrate"),
              ids = list(c(211L, 220045L), NULL),
              class = c("sel_itm", "col_itm"))

  identical(as_item(hr1), hr2[1])
  identical(new_item(list(hr1)), hr2[1])
}
```
Description

Owing to increased complexity and more diverse applications, recursive concepts (class `rec_concept`) may specify callback functions to be called on corresponding data objects and perform post-processing steps.

Usage

```r
pafi(..., match_win = hours(2L),
     mode = c("match_vals", "extreme_vals", "fill_gaps"),
     fix_na_fio2 = TRUE,
     interval = NULL)

vent(..., match_win = hours(6L), min_length = mins(10L), interval = NULL)

sed(..., interval = NULL)

gcs(...,
     valid_win = hours(6L),
     set_sed_max = TRUE,
     set_na_max = TRUE,
     interval = NULL)

urine24(...,
       min_win = hours(12L),
       limits = NULL,
       start_var = "start",
       end_var = "end",
       interval = NULL)

vaso60(..., max_gap = mins(5L), interval = NULL)
```

Arguments

- `...` Data input used for concept calculation
- `match_win` Time-span during which matching of values is allowed
mode Method for matching \( \text{PaO}_2 \) and \( \text{FiO}_2 \) values

**fix_na_fio2** Logical flag indicating whether to impute missing \( \text{FiO}_2 \) values with 21

**interval** Expected time series step size (determined from data if `NULL`)

**min_length** Minimal time span between a ventilation start and end time

**valid_win** Maximal time window for which a GCS value is valid if no newer measurement is available

**set_sed_max** Logical flag for considering sedation

**set_na_max** Logical flag controlling imputation of missing GCS values with the respective maximum values

**min_win** Minimal time span required for calculation of urine/24h limits

**limits** Passed to `fill_gaps()` in order to expand the time series beyond first and last measurements

**start_var, end_var** Passed to `fill_gaps()`

**max_gap** Maximum time gap between administration windows that are merged (can be negative).

**Details**

Several concept callback functions are exported, mainly for documenting their arguments, as default values oftentimes represent somewhat arbitrary choices and passing non-default values might be of interest for investigating stability with respect to such choices. Furthermore, default values might not be ideal for some datasets and/or analysis tasks.

**pafi:**

In order to calculate the \( \text{PaO}_2/\text{FiO}_2 \) (or Horowitz index), for a given time point, both a \( \text{PaO}_2 \) and a \( \text{FiO}_2 \) measurement is required. As the two are often not measured at the same time, some form of imputation or matching procedure is required. Several options are available:

- **match_vals** allows for a time difference of maximally `match_win` between two measurements for calculating their ratio
- **extreme_vals** uses the worst \( \text{PaO}_2 \) and a \( \text{FiO}_2 \) values within the time window spanned by `match_win`
- **fill_gaps** represents a variation of `extreme_vals`, where ratios are evaluated at every time-point as specified by `interval` as opposed to only the time points where either a \( \text{PaO}_2 \) or a \( \text{FiO}_2 \) measurement is available

Finally, **fix_na_fio2** imputes all remaining missing \( \text{FiO}_2 \) with 21, the percentage (by volume) of oxygen in (tropospheric) air.

**vent:**

Building on the atomic concepts `vent_start` and `vent_end`, an binary indicator for ventilation status is constructed by combining start and end events that are separated by at most `match_win` and at least `min_length`. Time-points (as determined by `interval`) that fall into such ventilation windows are set to `TRUE`, while missingness (NA) or `FALSE` indicate no mechanical ventilation. Currently, no clear distinction between invasive an non-invasive ventilation is made.
rename_cols

ICU class data utilities

Description

Several utility functions for working with id_tbl and ts_tbl objects are available, including functions for changing column names, removing columns, as well as aggregating or removing rows. An important thing to note is that as id_tbl (and consequently ts_tbl) inherits from data.table, there are several functions provided by the data.table package that are capable of modifying id_tbl in a way that results in an object with inconsistent state. An example for this is data.table::setnames(): if an ID column or the index column name is modified without updating the attribute marking the column as such, this leads to an invalid object. As data.table::setnames() is not an S3 generic function, the only way to control its behavior with respect to id_tbl objects is masking the function. As such an approach has its own down-sides, a separate function, rename_cols() is provided, which is able to handle column renaming correctly.
rename_cols

Usage

rename_cols(x, new, old = colnames(x), skip_absent = FALSE, by_ref = FALSE)

rm_cols(x, cols, skip_absent = FALSE, by_ref = FALSE)

change_interval(x, new_interval, cols = time_vars(x), by_ref = FALSE)

rm_na(x, cols = data_vars(x), mode = c("all", "any"))

## S3 method for class 'id_tbl'
sort(
  x,
  decreasing = FALSE,
  by = meta_vars(x),
  reorder_cols = TRUE,
  by_ref = FALSE,
  ...
)

is_sorted(x)

## S3 method for class 'id_tbl'
duplicated(x, incomparables = FALSE, by = meta_vars(x), ...)

## S3 method for class 'id_tbl'
anyDuplicated(x, incomparables = FALSE, by = meta_vars(x), ...)

## S3 method for class 'id_tbl'
unique(x, incomparables = FALSE, by = meta_vars(x), ...)

is_unique(x, ...)

## S3 method for class 'id_tbl'
aggregate(
  x,
  expr = NULL,
  by = meta_vars(x),
  vars = data_vars(x),
  env = NULL,
  ...
)

dt_gforce(
  x,
  fun = c("mean", "median", "min", "max", "sum", "prod", "var", "sd", "first", "last"),
  by = meta_vars(x),
  vars = data_vars(x),
  na_rm = !fun %in% c("first", "last")
rename_cols

Arguments

<table>
<thead>
<tr>
<th>Argument</th>
<th>Description</th>
</tr>
</thead>
<tbody>
<tr>
<td>x</td>
<td>Object to query</td>
</tr>
<tr>
<td>new, old</td>
<td>Replacement names and existing column names for renaming columns</td>
</tr>
<tr>
<td>skip_absent</td>
<td>Logical flag for ignoring non-existent column names</td>
</tr>
<tr>
<td>by_ref</td>
<td>Logical flag indicating whether to perform the operation by reference</td>
</tr>
<tr>
<td>cols</td>
<td>Column names of columns to consider</td>
</tr>
<tr>
<td>new_interval</td>
<td>Replacement interval length specified as scalar-valued difftime object</td>
</tr>
<tr>
<td>mode</td>
<td>Switch between all where all entries of a row have to be missing (for the selected columns) or any, where a single missing entry suffices</td>
</tr>
<tr>
<td>decreasing</td>
<td>Logical flag indicating the sort order</td>
</tr>
<tr>
<td>by</td>
<td>Character vector indicating which combinations of columns from x to use for uniqueness checks</td>
</tr>
<tr>
<td>reorder_cols</td>
<td>Logical flag indicating whether to move the by columns to the front.</td>
</tr>
<tr>
<td>...</td>
<td>Ignored</td>
</tr>
<tr>
<td>incomparables</td>
<td>Not used. Here for S3 method consistency</td>
</tr>
<tr>
<td>expr</td>
<td>Expression to apply over groups</td>
</tr>
<tr>
<td>vars</td>
<td>Column names to apply the function to</td>
</tr>
<tr>
<td>env</td>
<td>Environment to look up names in expr</td>
</tr>
<tr>
<td>fun</td>
<td>Function name (as string) to apply over groups</td>
</tr>
<tr>
<td>na_rm</td>
<td>Logical flag indicating how to treat NA values</td>
</tr>
</tbody>
</table>

Details

Apart from a function for renaming columns while respecting attributes marking columns a index or ID columns, several other utility functions are provided to make handling of id_tbl and ts_tbl objects more convenient.

Sorting:

An id_tbl or ts_tbl object is considered sorted when rows are in ascending order according to columns as specified by meta_vars(). This means that for an id_tbl object rows have to be ordered by id_vars() and for a ts_tbl object rows have to be ordered first by id_vars(), followed by the index_var(). Calling the S3 generic function base::sort() on an object that inherits id_tbl using default arguments yields an object that is considered sorted. For convenience (mostly in printing), the column by which the table was sorted are moved to the front (this can be disabled by passing FALSE as reorder_cols argument). Internally, sorting is handled by either setting a data.table::key() in case decreasing = FALSE or be calling data.table::setorder() in case decreasing = TRUE.

Uniqueness:

On object inheriting form id_tbl is considered unique if it is unique in terms of the columns as specified by meta_vars(). This means that for an id_tbl object, either zero or a single row is
allowed per combination of values in columns `id_vars()` and consequently for `ts_tbl` objects a maximum of one row is allowed per combination of time step and ID. In order to create a unique `id_tbl` object from a non-unique `id_tbl` object, `aggregate()` will combine observations that represent repeated measurements within a group.

**Aggregating:**

In order to turn a non-unique `id_tbl` or `ts_tbl` object into an object considered unique, the S3 generic function `stats::aggregate()` is available. This applied the expression (or function specification) passed as `expr` to each combination of grouping variables. The columns to be aggregated can be controlled using the `vars` argument and the grouping variables can be changed using the `by` argument. The argument `expr` is fairly flexible: it can take an expression that will be evaluated in the context of the `data.table` in a clean environment inheriting from `env`, it can be a function, or it can be a string in which case `dt_gforce()` is called. The default value `NULL` chooses a string dependent on data types, where numeric resolves to median, logical to sum and character to first.

As aggregation is used in concept loading (see `load_concepts()`), performance is important. For this reason, `dt_gforce()` allows for any of the available functions to be applied using the `GForce` optimization of `data.table` (see `data.table::datatable.optimize`).

**Value**

Most of the utility functions return either an `id_tbl` or a `ts_tbl`, potentially modified by reference, depending on the type of the object passed as `x`. The functions `is_sorted()`, `anyDuplicated()` and `is_unique()` return logical flags, while `duplicated()` returns a logical vector of the length `nrow(x)`.

**Examples**

```r
tbl <- id_tbl(a = rep(1:5, 4), b = rep(1:2, each = 10), c = rnorm(20),
              id_vars = c("a", "b"))
is_unique(tbl)
is_sorted(tbl)
is_sorted(tbl[order(c)])
identical(aggregate(tbl, list(c = sum(c))), aggregate(tbl, "sum"))

tbl <- aggregate(tbl, "sum")
is_unique(tbl)
is_sorted(tbl)
```

---

**Difftime utilities**

**Description**

As `base::difftime()` vectors are used throughout `ricu`, a set of wrapper functions are exported for convenience of instantiation `base::difftime()` vectors with given time units.
sofa_score

Usage

secs(x)
mins(x)
hours(x)
days(x)
weeks(x)

Arguments

x Numeric vector to coerce to base::difftime()

Value

Vector valued time differences as difftime object.

Examples

hours(1L)
mins(NA_real_)
secs(1:10)
hours(numeric(0L))

sofa_score SOFA score label

Description

The SOFA (Sequential Organ Failure Assessment) score is a commonly used assessment tool for tracking a patient’s status during a stay at an ICU. Organ function is quantified by aggregating 6 individual scores, representing respiratory, cardiovascular, hepatic, coagulation, renal and neurological systems. The function sofa_score() is used as callback function to the sofa concept but is exported as there are a few arguments that can used to modify some aspects of the presented SOFA implementation. Internally, sofa_score() calls first sofa_window(), followed by sofa_compute() and arguments passed as ... will be forwarded to the respective internally called function.

Usage

sofa_score(
  ..., interval = NULL,
  win_fun = max_or_na,
explicit_wins = FALSE,
win_length = hours(24L)
)

sofa_resp(..., interval = NULL)
sofa_coag(..., interval = NULL)
sofa_liver(..., interval = NULL)
sofa_cardio(..., interval = NULL)
sofa_cns(..., interval = NULL)
sofa_renal(..., interval = NULL)

Arguments

... Argument Concept data, either passed as list or individual argument
interval Argument Time series interval (only used for checking consistency of input data, NULL will use the interval of the first data object)
win_fun Argument functions used to calculate worst values over windows
explicit_wins Argument The default FALSE iterates over all time steps, TRUE uses only the last time step per patient and a vector of times will iterate over these explicit time points
win_length Argument Time-frame to look back and apply the win_fun

Details

The function sofa_score() calculates, for each component, the worst value over a moving window as specified by win_length, using the function passed as win_fun. The default functions max_or_na() return NA instead of -Inf/Inf in the case where no measurement is available over an entire window. When calculating the overall score by summing up components per time-step, a NA value is treated as 0.

Building on separate concepts, measurements for each component are converted to a component score using the definition by Vincent et. al.:

<table>
<thead>
<tr>
<th>SOFA score</th>
<th>1</th>
<th>2</th>
<th>3</th>
<th>4</th>
</tr>
</thead>
<tbody>
<tr>
<td>Respiration</td>
<td>PaO2/FiO2 [mmHg]</td>
<td>&lt; 400</td>
<td>&lt; 300</td>
<td>&lt; 200</td>
</tr>
<tr>
<td></td>
<td>and mechanical ventilation</td>
<td>yes</td>
<td>yes</td>
<td></td>
</tr>
<tr>
<td>Coagulation</td>
<td>Platelets [×10^3/mm^3]</td>
<td>&lt; 150</td>
<td>&lt; 100</td>
<td>&lt; 50</td>
</tr>
<tr>
<td>Liver</td>
<td>Bilirubin [mg/dl]</td>
<td>1.2-1.9</td>
<td>2.0-5.9</td>
<td>6.0-11.9</td>
</tr>
<tr>
<td>Cardiovascular</td>
<td>MAP</td>
<td>&lt; 70 mmHg</td>
<td>&lt; 5</td>
<td>&gt; 5</td>
</tr>
<tr>
<td></td>
<td>or dopamine</td>
<td>≤ 5</td>
<td>&gt; 5</td>
<td>&gt; 15</td>
</tr>
<tr>
<td></td>
<td>or dobutamine</td>
<td>any dose</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>
or epinephrine \( \leq 0.1 \) \( > 0.1 \)
or norepinephrine \( \leq 0.1 \) \( > 0.1 \)

**Central nervous system**

<table>
<thead>
<tr>
<th>Glasgow Coma Score</th>
<th>13-14</th>
<th>10-12</th>
<th>6-9</th>
<th>&lt; 6</th>
</tr>
</thead>
</table>

**Renal**

<table>
<thead>
<tr>
<th>Creatinine [mg/dl]</th>
<th>1.2-1.9</th>
<th>2.0-3.4</th>
<th>3.5-4.9</th>
<th>&gt; 5.0</th>
</tr>
</thead>
<tbody>
<tr>
<td>or urine output [ml/day]</td>
<td>&lt; 500</td>
<td>&lt; 200</td>
<td></td>
<td></td>
</tr>
</tbody>
</table>

\(^a\)Adrenergic agents administered for at least 1h (doses given are in \( [\mu g/\text{kg} \cdot \text{min}] \))

At default, for each patient, a score is calculated for every time step, from the first available measurement to the last. In instead of a regularly evaluated score, only certain time points are of interest, this can be specified using the `explicit_wins` argument: passing for example `hours(24, 48)` will yield for every patient a score at hours 24 and 48 relative to the origin of the current ID system (for example ICU stay).

**Value**

A `ts_tbl` object.

**References**


---

**Description**

Building on functionality offered by the (internal) function `id_map()`, stay windows as well as (in case of differing values being passed as `id_type` and `win_type`) an ID mapping is computed.

**Usage**

```r
stay_windows(
  x,
  id_type = "icustay",
  win_type = id_type,
  in_time = "start",
  out_time = "end",
  interval = hours(1L)
)
```
Arguments

- **x**: Passed to `as_id_cfg()` and `as_src_env()`.
- **id_type**: Type of ID all returned times are relative to.
- **win_type**: Type of ID for which the in/out times is returned.
- **in_time, out_time**: Column names of the returned in/out times.
- **interval**: The time interval used to discretize time stamps with, specified as `base::difftime()` object.

Value

An `id_tbl` containing the selected IDs and depending on values passed as `in_time` and `out_time`, start and end times of the ID passed as `win_var`.

See Also

- `change_id`

---

**transform_fun**

*Item callback utilities*

**Description**

For concept loading, item callback functions are used in order to handle item-specific post-processing steps, such as converting measurement units, mapping a set of values to another or for more involved data transformations, like turning absolute drug administration rates into rates that are relative to body weight. Item callback functions are called by `load_concepts()` with arguments `x` (the data), a variable number of name/ string pairs specifying roles of columns for the given item, followed by `env`, the data source environment as `src_env` object. Item callback functions can be specified by their name or using function factories such as `transform_fun()`, `apply_map()` or `convert_unit()`.

**Usage**

- `transform_fun(fun, ...)`
- `binary_op(op, y)`
- `comp_na(op, y)`
- `apply_map(map)`
- `convert_unit(rgx, fun, new, ignore_case = TRUE, ...)`
transform_fun

Arguments

- **fun**: Function(s) used for transforming matching values
- ... Further arguments passed to downstream function
- **op**: Function taking two arguments, such as +
- **y**: Value passed as second argument to function op
- **map**: Named atomic vector used for mapping a set of values (the names of map) to a different set (the values of map)
- **rgx**: Regular expression(s) used for identifying observations based on their current unit of measurement
- **new**: Name(s) of transformed units
- **ignore_case**: Forwarded to base::grep()

Details

The most forward setting is where a function is simply referred to by its name. For example in eICU, age is available as character vector due to ages 90 and above being represented by the string "> 89". A function such as the following turns this into a numeric vector, replacing occurrences of "> 89" by the number 90.

```r
eicu_age <- function(x, val_var, ...) {
  data.table::set(
    data.table::set(x, which(x[[val_var]] == "> 89"), j = val_var,
    value = 90),
    j = val_var,
    value = as.numeric(x[[val_var]])
  )
}
```

This function then is specified as item callback function for items corresponding to eICU data sources of the age concept as

```r
item(src = "eicu_demo", table = "patient", val_var = "age",
     callback = "eicu_age", class = "col_itm")
```

The string passed as callback argument is evaluated, meaning that an expression can be passed which evaluates to a function that in turn can be used as callback. Several function factories are provided which return functions suitable for use as item callbacks: `transform_fun()` creates a function that transforms the val_var column using the function supplied as fun argument, `apply_map()` can be used to map one set of values to another (again using the val_var column) and `convert_unit()` is intended for converting a subset of rows (identified by matching rgx against the unit_var column) by applying fun to the val_var column and setting new as the transformed unit name (arguments are not limited to scalar values). As transformations require unary functions, two utility function, `binary_op()` and `comp_na()` are provided which can be used to fix the second argument of binary functions such as * or ==. Taking all this together, an item callback function for dividing the val_var column by 2 could be specified as "transform_fun(binary_op(/, 2))". The supplied function factories create functions that operate on the data using by-reference semantics. Furthermore, during concept loading, progress is reported by a `progress::progress_bar`. In order to signal a message without disrupting the current loading status, see `msg_progress()`.
Value

Callback function factories such as transform_fun(), apply_map() or convert_unit() return functions suitable as item callback functions, while transform function generators such as binary_op(), comp_na() return functions that apply a transformation to a vector.

Examples

dat <- ts_tbl(x = rep(1:2, each = 5), y = hours(rep(1:5, 2)), z = 1:10)

subtract_3 <- transform_fun(binary_op(`-`, 3))
subtract_3(data.table::copy(dat), val_var = "z")

gte_4 <- transform_fun(comp_na(`>=`, 4))
gte_4(data.table::copy(dat), val_var = "z")

map_letters <- apply_map(setNames(letters[1:9], 1:9))
res <- map_letters(data.table::copy(dat), val_var = "z")
res

not_b <- transform_fun(comp_na(`!=`", "b"))
not_b(res, val_var = "z")

write_psv

Read and write utilities

Description

Support for reading from and writing to pipe separated values (.psv) files as used for the PhysioNet Sepsis Challenge.

Usage

write_psv(x, dir, na_rows = NULL)

read_psv(dir, col_spec = NULL, id_var = "stay_id")

Arguments

x Object to write to files
dir Directory to write the (many) files to or read from
na_rows If TRUE missing time steps are filled with NaN values, if FALSE, rows where all data columns entries are missing are removed and if NULL, data is written as-is
col_spec A column specification as created by readr::cols()
id_var Name of the id column (IDs are generated from file names)
Details

Data for the PhysioNet Sepsis Challenge is distributed as pipe separated values (.psv) files, split into separate files per patient ID, containing time stamped rows with measured variables as columns. Files are named with patient IDs and do not contain any patient identifiers as data. Functions `read_psv()` and `write_psv()` can be used to read from and write to such a data format.

Value

While `write_psv()` is called for side effects and returns NULL invisibly, `read_psv()` returns an object inheriting from `id_tbl`.

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