Package ‘rPref’

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Description Routines to select and visualize the maxima for a given strict partial order. This especially includes the computation of the Pareto frontier, also known as (Top-k) Skyline operator (see Börzsönyi, et al. (2001) <doi:10.1109/ICDE.2001.914855>), and some generalizations known as database preferences (see Kießling (2002) <doi:10.1016/B978-155860869-6/50035-4>).
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base_pref ......................................................... 2
base_pref_macros .............................................. 4
complex_pref ................................................. 6
general_pref ............................................... 8
get_hasse_diag ............................................... 10
plot_btg ......................................................... 11
plot_front ....................................................... 13
pred_succ ....................................................... 14
psel ............................................................ 16
rPref ........................................................... 19
show.pref ...................................................... 20
show.query ..................................................... 21

Description

Base preferences are used to describe the different goals (dimensions, in case of a Skyline query) of a preference query.

Usage

low(expr, df = NULL)
low_(expr, df = NULL)
high(expr, df = NULL)
high_(expr, df = NULL)
true(expr, df = NULL)
true_(expr, df = NULL)
is.base_pref(x)

Arguments

expr A numerical/logical expression which is the term to evaluate for the current preference. The objective is to search for minimal/maximal values of this expression (for low/high) or for logical TRUE values (for true). For low_., high_. and true_., the argument must be an expression, a call or a string.
df

(optional) A data frame, having the same structure (i.e., columns) like that data
frame, where this preference is evaluated later on. Causes a partial evaluation of
the preference and the preference is associated with this data frame. See below
for details.

x

An object to be tested if it is a base preference.

Details

Mathematically, all base preferences are strict weak orders (irreflexive, transitive and negative tran-
sitive).

The three fundamental base preferences are:

low(a), high(a) Search for minimal/maximal values of a, i.e., the induced order is the "smaller
than" or "greater than" order on the values of a. The values of a must be numeric values.

true(a) Search for true values in logical expressions, i.e., TRUE is considered to be better than
FALSE. The values of a must be logical values. For a tuplewise evaluation of a complex logical
expression one has to use the & and | operators for logical AND/OR (and not the && and ||
operators).

The term expr may be just a single attribute or may contain an arbitrary expression, depending on
more than one attribute, e.g., low(a+2*b+f(c)). There a, b and c are columns of the addressed
data set and f has to be a previously defined function.

Functions contained in expr are evaluated over the entire data set, i.e., it is possible to use aggregate
functions (min, mean, etc.). Note that all functions (and also variables which are not columns of
the data set, where expr will be evaluated on) must be defined in the same environment (e.g.,
environment of a function or global environment) as the base preference is defined.

The function is_base_pref returns TRUE if x is a preference object and FALSE otherwise.

Using Expressions in Preferences

The low_, high_ and true_ preferences have the same functionality as low, high and true but ex-
pect an expression, a call or a string as argument. For example, low(a) is equivalent to low_(expression(a))
or low_("a"). Lazy expressions (see the lazyeval package) are also possible.

This is helpful for developing your own base preferences. Assume you want to define a base Prefer-
ence false as the dual of true. A definition like false <- function(x) -true(x) is the wrong
approach, as psel(data.frame(a = c(1,2)), false(a == 1)) will result in the error "object
'a' not found". This is because a is considered as a variable and not as an (abstract) symbol to be
evaluated later. By defining

false <- function(x, ...) -true_(substitute(x), ...)

one gets a preference which behaves like a "built-in" preference. Additional optional parameters
(like df) are bypassed. The object false(a == 1) will output [Preference] -true(a == 1)
on the console and psel(data.frame(a = c(1,2)), false(a==1)) returns correctly the second
tuple with a==2.

There is a special symbol df__ which can be used in preference expression to access the given data
set df, when psel is called on this data set. For example, on a data set where the first column has
the name A the preference low(df__[[1]]) is equivalent to low(A).
Partial Evaluation and Associated Data Frames

If the optional parameter \texttt{df} is given, then the expression is evaluated at the time of definition as far as possible. All variables occurring as columns in \texttt{df} remain untouched. For example, consider

\begin{verbatim}
f <- function(x) 2*x
p <- true(cyl == f(1), mtcars)
\end{verbatim}

Then \texttt{p} is equivalent to the preference \texttt{true(cyl == 2)} as the variable \texttt{cyl} is a column in \texttt{mtcars}. Additionally the data set \texttt{mtcars} is associated with the preference \texttt{p}, implying that the preference selection can be done with \texttt{peval}. See \texttt{assoc.df} for details on associated data sets.

The preference selection, i.e., \texttt{psel(mtcars, p)} can be invoked without the partial evaluation. But this results in an error, if the function \texttt{f} has meanwhile removed from the current environment. Hence it is safer to do an early partial evaluation of all preferences, as far as they contain user defined functions.

The partial evaluation can be done manually by \texttt{partial.eval.pref}.

See Also

See \texttt{complex_pref} how to compose complex preferences to retrieve e.g., the Skyline. See \texttt{general_pref} for functions applying to all kind of preferences. See \texttt{base_pref_macros} for more base preferences.

Examples

\begin{verbatim}
# define a preference with a score value combining mpg and hp
p1 <- high(4 * mpg + hp)
# perform the preference selection
psel(mtcars, p1)

# define a preference with a given function
f <- function(x, y) (abs(x - mean(x))/max(x) + abs(y - mean(y))/max(y))
p2 <- low(f(mpg, hp))
psel(mtcars, p2)

# use partial evaluation for weighted scoring
p3 <- high(mpg/sum(mtcars$mpg) + hp/sum(mtcars$hp), df = mtcars)
p3
# select Pareto optima
peval(p3)
\end{verbatim}
Usage

```r
around(expr, center, df = NULL)

between(expr, left, right, df = NULL)

pos(expr, pos_value, df = NULL)

layered(expr, ..., df = NULL)
```

Arguments

- `expr`: A numerical expression (for `around` and `between`) or an arbitrary expression (for `pos` and `layered`). The objective are those tuples where `expr` evaluates to a value within the preferred interval, layer, etc. Regarding attributes, functions and variables, the same requirements as for `base_pref` apply.
- `center`: Preferred numerical value for `around`.
- `df` (optional): Data frame for partial evaluation and association of preference and data set. See `base_pref` for details.
- `left`: Lower limit (numerical) of the preferred interval for `between`.
- `right`: Upper limit (numerical) of the preferred interval for `between`.
- `pos_value`: A vector containing the preferred values for a `pos` preference. It has to be of the same type (numeric, logical, character, ...) as `expr`.
- `...`: Layers (sets) for a `layered` preference. Each parameter corresponds to a layer and the first one characterizes the most preferred values.

Definition of the Preference Macros

- `between(expr, left, right)` Those tuples are preferred where `expr` evaluates to a value between `left` and `right`. For values not in this interval, the values nearest to the interval are preferred.
- `around(expr, center)` Same as `between(expr, center, center)`.
- `pos(expr, pos_value)` Those tuples are preferred, where `expr` evaluates to a value which is contained in `pos_value`.
- `layered(expr, layer1, layer2, ..., layerN)` For the most preferred tuples `expr` must evaluate to a value in `layer1`. The second-best tuples are those where `expr` evaluates to a value in `layer2` and so forth. Values occurring in none of the layers are considered worse than those in `layerN`. Technically, this is realized by a prioritization chain (lexicographical order) of `true` preferences.

Note that only the argument `expr` may contain columns from the data frame, all other variables must evaluate to explicit values. For example `around(mpg, mean(mpg))` is not allowed. In this case, one can use `around(mpg, mean(mtcars$mpg))` instead. Or alternatively, without using the base preference macros, `low(abs(mpg - mean(mpg)))` does the same. There, the actual mean value of `mpg` is calculated just when the preference selection via `psel` is called.
Examples

```r
c # search for cars where mpg is near to 25
psel(mtcars, around(mpg, 25))

c # cyl = 2 and cyl = 4 are equally good, cyl = 6 is worse
psel(mtcars, layered(cyl, c(2, 4), 6))
```

**Description**

Complex preferences are used to compose different preference orders. For example the Pareto composition (via operator `*`) is the usual operator to compose the preference for a Skyline query. The Skyline is also known as Pareto frontier. All complex preferences are mathematically strict partial orders (irreflexive and transitive).

**Usage**

```r

## S4 method for signature 'preference,preference'
e1 * e2

## S4 method for signature 'preference,preference'
e1 & e2

## S4 method for signature 'preference,preference'
e1 | e2

## S4 method for signature 'preference,preference'
e1 + e2

reverse(p)
is.complex_pref(x)
```

**Arguments**

- `p`, `e1`, `e2` Preference objects (they can be either base preferences, see `base_pref`, or complex preferences)
- `x` An object to be tested if it is a complex preference.

**Skylines**

The most important preference composition operator is the Pareto operator (`p1 * p2`) to formulate a Skyline query. A tuple `t1` is better than `t2` w.r.t. `p1 * p2` if it is strictly better w.r.t. one of the preferences `p1`, `p2` and is better or equal w.r.t. the other preference.

The syntactical correspondence to other query languages supporting Skylines/preferences to rPref is given as follows:
• A query in the syntax from Borzsonyi et. al (2001) like

"... SKYLINE OF a MAX, b MIN, c MAX"

corresponds in rPref to the preference

high(a) * low(b) * high(c).

• A query in the syntax from Kiessling (2002) like

"... PREFERRING a LOWEST AND (b HIGHEST PRIOR TO c LOWEST)"

corresponds in rPref to

low(a) * (high(b) & low(c)).

• A query in the syntax of the "Skyline" feature of the commercial database "EXASOL EXASolution 5" like

"... PREFERRING LOW a PLUS (b = 1 PRIOR TO LOW c))"

corresponds in rPref to

low(a) * (true(b == 1)) & low(c).

Note that preferences in rPref can be translated to some of this query dialects by `show.query`.

**Definition of Additional Preference Operators**

Additionally, rPref supports the following preference composition operators:

- `p1 & p2` Prioritization (lexicographical order): A tuple t1 is better than t2 w.r.t. p1 & p2 if it is strictly better w.r.t. p1 or is equal w.r.t. p1 and is better w.r.t. p2.
- `p1 | p2` Intersection preference: A tuple t1 is better than t2 w.r.t. p1 | p2 if it is strictly better w.r.t. both preferences. This is a stricter variant of the Pareto operator. The evaluation of `psel(df, p1 | p2)` is always a subset of `psel(df, p1 * p2)`.
- `p1 + p2` Union preference: A tuple t1 is better than t2 w.r.t. p1 + p2 if it is strictly better w.r.t. to one of the preferences. Note that this can violate the strict partial order property, if the domains (the tuples on which `p1` and `p2` define better-than-relationships) of the preferences are not disjoint.
- `reverse(p1)` or `-p1` Reverse preference (converse relation): A tuple t1 is better than t2 w.r.t. -p1 if t2 is better than t1 w.r.t. p1. The unary minus operator, i.e. -p1, is a short hand notation for `reverse(p1)`.

The function `is.complex_pref` returns TRUE if x is a complex preference object (i.e., was constructed by one of these binary operators or the unary operator `reverse`) and FALSE otherwise.

**Associated Data Sets**

If one of the preferences for a binary operator are associated with a data set (see `base_pref`), then this association is propagated. For example, the preference

`p <= high(mpg, df = mtcars) * high(hp)`

as well as

`p <= high(mpg) * high(hp, df = mtcars)`

both result in the same complex preference which is associated with `mtcars`. A partial evaluation is also invoked for all preferred which are added. For example, using this p,
p <- p * true(cyl == max(mtcars$cyl))

generates the following console output:

[Preference] high(mpg) * high(hp) * true(cyl == 8)
* associated data source: data.frame "mtcars" [32 x 11]

We see that the association with the data set is propagated and max(mtcars$cyl) is partially evaluated.

References


See Also

See `base_pref` for the construction of base preferences. See `general_pref` for functions applicable to all kind of preferences. See `psel` for the evaluation of preferences.

Examples

# define preference for cars with low consumption (high mpg-value)
# and simultaneously high horsepower
p1 <- high(mpg) * high(hp)

# perform the preference search
psel(mtcars, p1)

# alternative way: create preference with associated data set
p2 <- high(mpg, df = mtcars) * high(hp)
peval(p2)
### Usage

```r
empty()

is.empty_pref(x)
```

```r
## S4 method for signature 'preference'
length(x)

is.preference(x)
```

```r
## S4 method for signature 'preference'
as.expression(x, ...)

## S4 method for signature 'preference'
assoc.df(x)

## S4 replacement method for signature 'preference'
assoc.df(x) <- value
```

### Arguments

- `x`  
  A preference, or, for `is.preference`, an object to be tested if it is an (empty) preference.

- `...` 
  Optional arguments passed to `as.expression`.

- `value`  
  A data frame to associate with a preference object.

### Details

The empty preference `empty()` is a neutral element for the complex preference compositions `{*, &`, `+}`. It holds that `empty() * p` and `empty() & p` is equal to `p` for all preferences `p`.

The function `length(p)` returns the term length of the preference term `p` which is defined as the number of base preferences in a complex preference term. The empty preference `empty()` has length 0, and all base preferences have length 1.

With `as.expression(p)` for a preference `p` the call to the preference is constructed. This means, `eval(as.expression(p))` returns the preference `p`, evaluated in the current environment.

The function `is.empty_pref` returns `TRUE` if `x` is the empty preference object `empty()` and `FALSE` otherwise.

With `assoc.df` the associated data frame of a preference can be retrieved or set. Setting the associated data frame means that a partial evaluation based on this data frame is done. See `show.pref` for details on partial evaluation of preferences. Next, the preference is linked to that data frame, such that `peval(p)` can be used instead of `psel(df, p)`. It returns `NULL` if no data frame is associated. Use `set.assoc.df(NULL)` to delete an associated data frame.

### See Also

See `base_pref` for the construction of base preferences, and `complex_pref` for the construction of complex preferences. See `show.pref` for string output and partial evaluation of preference terms.
Examples

```r
# Same as low(a) * low(b)
p <- low(a) * low(b) * empty()

# returns 2, as empty() does not count
length(p)

# the preference expression (without empty())
as.expression(p)
```

---

**get_hasse.diag**

**Adjacency List of Hasse diagramm**

**Description**

Returns the adjacency list of the Hasse diagram of a preference as an (n x 2) matrix. This is the transitive reduction of the preference relation.

**Usage**

get_hasse.diag(df, pref)

**Arguments**

- `df` A data frame.
- `pref` A preference on the columns of `df`, see `psel` for details.

**Details**

A row (i, j) in the resulting matrix means that `df[i,]` is better than `df[j,]` with regard to the preference `p`. The matrix is the transitive reduction (Hasse diagram) of the induced relations, i.e., if (1,2) and (2,3) occur in the result, then (1,3) will not be contained. The number of rows in the result depends on the number of non-transitive Better-Than-Relationships in `df` w.r.t. `p`.

**See Also**

- `get_btg` to plot the Hasse diagram.

**Examples**

```r
get_hasse.diag(mtcars, low(mpg))
```
**plot_btg**  
*Better-Than-Graphs*

**Description**

Returns/plots a Hasse diagram of a preference order (also called the Better-Than-Graph, short BTG) on a given data set. Ploting within R relies on the igraph package or the Rgraphviz package. Alternatively, a dot file for an external graphviz/dot interpreter can be generated.

**Usage**

```r
plot_btg(df, pref, labels = 1:nrow(df), flip.edges = FALSE,  
levelwise = TRUE, use_dot = "Rgraphviz" %in%  
rownames(installed.packages()))
```

```r
get_btg(df, pref, flip.edges = FALSE, use_dot = "Rgraphviz" %in%  
rownames(installed.packages()))
```

```r
get_btg_dot(df, pref, labels = 1:nrow(df), flip.edges = FALSE,  
levelwise = TRUE, file = NULL)
```

**Arguments**

- `df`  
  A data frame.

- `pref`  
  A preference on the columns of `df`, see `psel` for details.

- `labels`  
  (optional) Labels for the vertices. Default values are the row indices.

- `flip.edges`  
  (optional) Flips the orientation of edges, if TRUE than arrows point from worse nodes to better nodes.

- `levelwise`  
  (optional) Only relevant is the dot layouter is used. If TRUE, all tuples from the same level are placed on one row. If FALSE, the row arrangement is subject to the dot layouter.

- `use_dot`  
  (optional) If TRUE, the dot layouter from Rgraphviz is used. If FALSE, igraph is used. By default this is TRUE if Rgraphviz is available.

- `file`  
  (optional) If specified, then `get_btg_dot` writes the graph specification to given file path. If not specified, the graph specification is returned as a string.

**Details**

The Hasse diagram of a preference visualizes all the better-than-relationships on a given data set. All edges which can be retrieved by transitivity of the order are omitted in the graph.

The functions `get_btg` and `plot_btg` either use the igraph package (if `use_dot` = FALSE) or the dot layouter from the Rgraphviz package (if `use_dot` = TRUE). If Rgraphviz is available it is used by default, otherwise the igraph Package is used. Note that Rgraphviz is only available on BioConductor and not on CRAN.
The dot layouter from Rgraphviz is more appropriate for Better-Than-Graphs than the igraph layouter, as all edges will be directed in the same direction (rank based ordering). Using `levelwise = TRUE` (the default), all tuples of the same level are placed on the same row.

**BTGs with igraph**

If used with `use_dot = FALSE`, the function `get_btg` returns a list `l` with the following list entries:

- `l$graph`: An igraph object, created with the `igraph` package.
- `l$layout`: A typical Hasse diagram layout for plotting the graph, also created with igraph.

To plot the resulting graph returned from `get_btg`, use the `plot` function as follows:

```r
plot(l$graph, layout = l$layout)
```

For more details, see `igraph.plotting`.

**BTGs with Rgraphviz**

If used with `use_dot = FALSE`, the function `get_btg` returns a `graphnel` object from the `graph` package (Rgraphviz is build on top of that package). This object can also be plotted using `plot(...)`.

**Direct Plotting**

In both cases (whether Rgraphviz is used or not), the function `plot_btg` directly plots the Better-Than-Graph. There is an additional parameter `labels`, specifying the node labels. The default are the row numbers (not the `rownames` of the data frame), ranging from `"1"` to `as.character(nrow(df))`.

**Dot (Graphviz) String Output**

The function `get_btg_dot` produces the source code of the Better-Than-Graph in the dot language of the Graphviz software. This is useful for an external dot interpreter. Depending on the `file` parameter the output is either written to a file (if a file path is given) or returned as a string (if `file = NULL`).

**Additional Parameters**

By default, the directed edges in the diagram point from better to worse nodes w.r.t. the preference. This means an arrow can be read as "is better than". If `flip.edges = TRUE` is set, then the arrows point from worse nodes to better nodes ("is worse than"). In any case, the better nodes are plotted at the top and the worse nodes at the bottom of the diagram.

If Rgraphviz is used for `plot_btg` and for `get_btg_dot`, the option `levelwise` controls if all nodes of the same level are placed in one row. If this parameter is `FALSE`, then the vertical arrangement is subject to the dot layouter.

**Examples**

```r
# pick a small data set and create preference and BTG
df <- mtcars[1:10,]
pref <- high(mpg) * low(wt)
```
# directly plot the BTG with row numbers as labels
# uses Rgraphviz if available and igraph otherwise
plot_btg(df, pref)

# plot the graph with labels with relevant values
labels <- paste0(df$mpg, " ", df$wt)
plot_btg(df, pref, labels)

# show lattice structure of 3-dimensional Pareto preference
df <- merge(merge(data.frame(x = 1:3), data.frame(y = 1:3)), data.frame(z = 1:2))
labels <- paste0(df$x, ",", df$y, ",", df$z)
plot_btg(df, low(x) * low(y) * low(z), labels)

# Create a graph with external Graphviz (requires installed Graphviz)
## Not run:
# creates tmpgraph.dot in the current working directory
get_btg_dot(df, pref, labels, file = "tmpgraph.dot")
# convert to diagram tmpgraph.png using Graphviz
shell(paste0("C:\Program Files (x86)\Graphviz2.38\bin\dot.exe",
            " -Tpng tmpgraph.dot -o tmpgraph.png"))
# open resulting image
shell("tmpgraph.png")
## End(Not run)

## plot_front

### Pareto Front Plot

**Description**

Connects the points of a Pareto front (also known as Pareto frontier) and hence visualizes the dominance region of a Skyline.

**Usage**

plot_front(df, pref, ...)

**Arguments**

- **df**
  The data frame for which the Pareto front is plotted. This may be already a maximal set w.r.t. the preference `pref`, but anyway the maximal set is recalculated via `psel(df, pref)`.

- **pref**
  The preference representing the Skyline goals. This must be a Pareto composition (p1 * p2) or intersection composition (p1 | p2) of two `low` or `high` preferences.

- **...**
  Additional graphic parameters which are passed to the `segments` function (internally used to plot the front).
Details

plot_front assumes that there is an existing plot, where the value of the first preference was plotted as x-coordinate and the value of the second preference as y-coordinate.

Note that plot_front is only recommended if you want to use the plotting functionality from base R. If you prefer to use ggplot2, we recommend using geom_step for plotting the Pareto front. See vignette("visualization", package = "rPref") for examples.

Examples

```r
# plots Pareto fronts for the hp/mpg values of mtcars
show_front <- function(pref) {
  plot(mtcars$hp, mtcars$mpg)
  sky <- psel(mtcars, pref)
  plot_front(mtcars, pref, col = rgb(0, 0, 1))
  points(sky$hp, sky$mpg, lwd = 3)
}

# do this for all four combinations of Pareto compositions
show_front(low(hp) * low(mpg))
show_front(low(hp) * high(mpg))
show_front(high(hp) * low(mpg))
show_front(high(hp) * high(mpg))

# compare this to the front of a intersection preference
show_front(high(hp) | high(mpg))
```

pred_succ

Predecessor and Successor Functions

Description

Function for traversing the BTG (Better-Than-Graph or Hasse diagram) of a preference.

Usage

```r
init_pred_succ(p, df = NULL)
hasse_pred(p, v, intersect = FALSE)
hasse_succ(p, v, intersect = FALSE)
all_pred(p, v, intersect = FALSE)
all_succ(p, v, intersect = FALSE)
```
Arguments

**p**
A preference. Worse tuples in the induced order are successors and better tuples are predecessors.

**df**
(optional) A data frame characterizing the set wherein predecessors/successors are searched. If df is NULL then the data frame associated with the preference is used. Causes an error if df == NULL and no data frame is associated.

**v**
A numeric vector of indices in df. This represents the set of tuples for which predecessors/successors are searched.

**intersect**
(optional) Logical value. If it is FALSE (by default) the union of all predecessors/successors of v is returned. For intersect = TRUE the intersection of those values is returned.

Details

These functions return the predecessors and successors in the Better-Than-Graph of a preference. Note that the successors/predecessors can be plotted via \texttt{get_btg}. Before any of the successor/predecessor functions can be used the initialization has to be called as follows:

\begin{verbatim}
init_pred_succ(p, df)
\end{verbatim}

There \( p \) is a preference object and \( df \) a data frame. When this done, the data frame \( df \) is associated with \( p \), i.e., implicitly \texttt{assoc.df} is called. If the preference has already an associated data frame, \( df \) can be omitted. For example

\begin{verbatim}
p <- low(mpg, df = mtcars)
init_pred_succ(p)
\end{verbatim}

does the initialization of the preference \texttt{low(mpg)} on the data set \texttt{mtcars}.

The \texttt{init_pred_succ} function calculates the Better-Than-Relation on \( df \) w.r.t. \( p \). Afterwards the predecessor and successor functions, as subsequently described, can be called. The value of \( v \) is a numeric vector within \( 1:nrow(df) \) and characterizes a subset of tuples in \( df \). The return value of these functions is again a numeric vector referring to the row numbers in \( df \) and it is always ordered ascending, independently of the order of the indices in \( v \).

\begin{verbatim}
all_pred(p, v) Returns all predecessors of \( v \), i.e., indices of better tuples than \( v \).
all_succ(p, v) Returns all successors of \( v \), i.e., indices of worse tuples than \( v \).
hasse_pred(p, v) Returns the direct predecessors of \( v \), i.e., indices of better tuples than \( v \) where the better-than-relation is contained in the transitive reduction.
hasse_succ(p, v) Returns the direct successors of \( v \), i.e., indices of worse tuples than \( v \) where the better-than-relation is contained in the transitive reduction.
\end{verbatim}

If \( v \) has length 1, then the value of intersect does not matter, as there is nothing to intersect or join. For scalar values \( x \) and \( y \) the following identities hold, where \( f \) is one of the predecessor/successor functions:

\begin{verbatim}
f(p, c(x, y), intersect = FALSE) == union(f(p, x), f(p, y))
f(p, c(x, y), intersect = TRUE) == intersect(f(p, x), f(p, y))
\end{verbatim}
Examples

```r
# preference on mtcars for high mpg and low weight
p <- high(mpg) * low(wt)
init_pred_succ(p, mtcars)

# helper to show mpg/_hp values
show_vals <- function(x) mtcars[c('mpg', 'wt')]

# pick some tuple "in the middle"
show_vals(10)

# show (direct) predecessors/successors of tuple 10
show_vals(hasse_pred(p, 10)) # Next better car
show_vals(hasse_succ(p, 10)) # Next worse car
show_vals(all_pred(p, 10)) # All better cars
show_vals(all_succ(p, 10)) # All worse cars
```

desc

Preferences Selection

Evaluates a preference on a given data set, i.e., returns the maximal elements of a data set for a given preference order.

Usage

```r
psel(df, pref, ...)
psel.indices(df, pref, ...)
peval(pref, ...)
```

Arguments

- `df` A data frame or, for a grouped preference selection, a grouped data frame. See below for details.
- `pref` The preference order constructed via `complex_pref` and `base_pref`. All variables occurring in the definition of `pref` must be either columns of the data frame `df` or variables/functions of the environment where `pref` was defined.
- `...` Additional (optional) parameters for top(-level)-k selections:
  - `top` A top value of k means that the k-best tuples of the data set are returned. This may be non-deterministic, see below for details.
  - `at_least` An at_least value of k returns the top-k tuples and additionally all tuples which are not dominated by the worst tuple (i.e. the minima) of the Top-k set. The number of tuples returned is greater or equal than at_least. In contrast to top-k, this is deterministic.
psel

**top_level** A top_level value of k returns all tuples from the k-best levels. See below for the definition of a level.

**and_connected** Logical value, which is only relevant if more than one of the above {top, at_least, top_level} values is given, otherwise it will be ignored. Then and_connected = TRUE (which is the default) means that all top-conditions must hold for the returned tuples: Let cond1 and cond2 be top-conditions like top=2 or top_level=3, then psel([...], cond1, cond2) is equivalent to the intersection of psel([...], cond1) and psel([...], cond2). If we have and_connected = FALSE, these conditions are or-connected. This corresponds to the union of psel([...], cond1) and psel([...], cond2).

**show_level** Logical value. If TRUE, a column .level is added to the returned data frame, containing all level values. If at least one of the {top, at_least, top_level} values are given, then show_level is true by default for the psel function. Otherwise, and for psel.indices in all cases, this option is FALSE by default.

**Details**

The difference between the three variants of the preference selection is:

- The psel function returns a subset of the data set which are the maxima according to the given preference.
- The function psel.indices returns just the row indices of the maxima (except top-k queries with show_level = TRUE, see top-k preference selection). Hence psel(df, pref) is equivalent to df[psel.indices(df, pref),] for non-grouped data frames.
- Finally, peval does the same like psel, but assumes that p has an associated data frame which is used for the preference selection. Consider base_pref to see how base preferences are associated with data sets or use assoc.df to explicitly associate a preference with a data frame.

**Top-k Preference Selection**

For a given top value of k the k best elements and their level values are returned. The level values are determined as follows:

- All the maxima of a data set w.r.t. a preference have level 1.
- The maxima of the remainder, i.e., the data set without the level 1 maxima, have level 2.
- The n-th iteration of "Take the maxima from the remainder" leads to tuples of level n.

By default, psel.indices does not return the level values. By setting show_level = TRUE this function returns a data frame with the columns '.indices' and '.level'. Note that, if none of the top-k values {top, at_least, top_level} is set, then all level values are equal to 1.

By definition, a top-k preference selection is non-deterministic. A top-1 query of two equivalent tuples (equivalence according to pref) can return on both of these tuples. For example, a top=1 preference selection on the tuples (a=1, b=1), (a=1, b=2) w.r.t. low(a) preference can return either the 'b=1' or the 'b=2' tuple.

On the contrary, a preference selection using at_least is deterministic by adding all tuples having the same level as the worst level of the corresponding top-k query. This means, the result is filled
with all tuples being not worse than the top-k result. A preference selection with top-level-k returns all tuples having level k or better.

If the top or at_least value is greater than the number of elements in df (i.e., nrow(df)), or top_level is greater than the highest level in df, then all elements of df will be returned without further warning.

**Grouped Preference Selection**

Using psel it is also possible to perform a preference selection where the maxima are calculated for every group separately. The groups have to be created with `group_by` from the dplyr package. The preference selection preserves the grouping, i.e., the groups are restored after the preference selection.

For example, if the summarize function from dplyr is applied to `psel(group_by(...), pref)`, the summarizing is done for the set of maxima of each group. This can be used to e.g., calculate the number of maxima in each group, see the examples below.

A \{top, at_least, top_level\} preference selection is applied to each group separately. A top=k selection returns the k best tuples for each group. Hence if there are 3 groups in df, each containing at least 2 elements, and we have top = 2, then 6 tuples will be returned.

**Parallel Computation**

On multi-core machines the preference selection can be run in parallel using a divide-and-conquer approach. Depending on the data set, this may be faster than a single-threaded computation. To activate parallel computation within rPref the following option has to be set:

```r
options(rPref.parallel = TRUE)
```

If this option is not set, rPref will use single-threaded computation by default. With the option `rPref.parallel.threads` the maximum number of threads can be specified. The default is the number of cores on your machine. To set the number of threads to the value of 4, use:

```r
options(rPref.parallel.threads = 4)
```

**See Also**

See `complex_pref` on how to construct a Skyline preference.

**Examples**

```r
# Skyline and top-k/at-least skyline
psel(mtcars, low(mpg) * low(hp))
psel(mtcars, low(mpg) * low(hp), top = 5)
psel(mtcars, low(mpg) * low(hp), at_least = 5)

# preference with associated data frame and evaluation
p <- low(mpg, df = mtcars) * (high(cyl) & high(gear))
peval(p)

# visualize the skyline in a plot
sky1 <- psel(mtcars, high(mpg) * high(hp))
plot(mtcars$mpg, mtcars$hp)
```
rPref

Summary of the rPref Package

Description

rPref contains routines to select and visualize the maxima for a given strict partial order. This especially includes the computation of the Pareto frontier, also known as (Top-k) Skyline operator, and some generalizations (database preferences).

Preference Composition/Selection

- Preferences are primarily composed from base preferences (see base_pref) and complex preferences (see complex_pref), where especially the Pareto operator for Skylines is such a complex preference.
- Some utility functions for preferences are collected in general_pref.
- Additionally some base preference macros are provided in base_pref Macros.
- The (top(-level)-k) preference selection psel allows to retrieve the maxima of a preference (or Pareto frontier, Skyline), constructed with the functions above, on a given data set.

Visualization and Analysis of Preferences

- The visualization of the preference order in a Better-Than-Graph (Hasse diagram) is possible via plot_btg.
- The adjacency list of the Hasse diagram can be accessed via get_hasse_diag.
- Predecessors/successors in the Hasse diagram are calculated with the pred_succ functions.
- The Pareto frontier can be plotted using the plot_front function.

String Output of Preferences

- The preference query for some preference-supporting DBMS can be given by show.query.
- A preference is partially evaluated and printed with show.pref.
Vignettes

To learn the basics of rPref, start with the vignettes:

- A general introduction and some examples are given in vignette("introduction", package = "rPref")
- The visualization of preferences is explained in vignette("visualization", package = "rPref")

Further Information

The rPref homepage is http://www.p-roocks.de/rpref. To submit bugs, feature requests or other comments, feel free to write a mail to me.

Author(s)

Patrick Roocks, <mail@p-roocks.de>

---

**show.pref**

Partial Evaluation and String Output of Preferences

---

Description

Functions to substitute variables and functions in preferences which can be calculated before the preference is evaluated on a data frame and character output of preferences.

Usage

```
show.pref(p, df = NULL)
```

```
# S4 method for signature 'preference'
as.character(x, ...)
```

```
pref.str(p, df = NULL)
```

```
partial.eval.pref(p, df = NULL)
```

Arguments

- `p, x` The preference to be shown or partially evaluated.
- `df` (optional) A data frame on which the preference operates. Used for partial evaluation.
- `...` Optional arguments passed to `as.character`. 
Details

The function `pref.str` (or `as.character(p)` for a preference `p`) returns the preference string while `show.pref` outputs it directly to the console, preceded by '[Preference]'. If `df` is specified, then a partial evaluation of the preference is done before converting it to a string.

The function `partial.evalpref` (with given data frame `df`) partially evaluates the internal preference expression and returns again a preference object. All expressions in `p` are evaluated in the environment where `p` was defined, except the the column names in `df` (which are potential attributes in `p`) and except the special variable `df__`, which accesses the entire data set (see `psel`). The content of the data frame `df` does not matter; only `names(df)` is used to get the "free variables" in `p`.

If `p` has already an associated data frame (see `assoc.df`), then a partial evaluation was already done when the data frame was associated. In this case, the `df` parameter should not be used. The association will not be changed if one of these function are called with a given data frame on a preference object having an associated data frame.

Partial Evaluation Before String Output

The functions `show.pref` and `pref.str` have the optional parameter `df`. If this parameter is given, these functions call `partial.evalpref` before they output or return the preference string. The following equalities hold:

- `as.character(partial.evalpref(p, df)) == pref.str(p, df)`
- `show(partial.evalpref(p, df))` produces the same console output as `show.pref(p, df)`

See Also

See `general_pref` for more utility functions for preferences.

Examples

```r
f <- function(x) 2*x
p <- true(cyl == f(1))

# prints 'true(cyl == f(1))'
show(p)

# prints 'true(cyl == 2)'
show.pref(p, mtcars)
partial.evalpref(p, mtcars)
```

Description

For a given preference this shows the `PREFERERING` clause of a database query in different SQL dialects which support preferences.
show.query(p, dialect = "EXASOL", df = NULL)

Arguments

p A preference.
dialect The preference query dialect, which determines the syntax of the returned query. This has to be one of the following (not case sensitive):

'EXASOL': Syntax of the "Skyline" feature of the commercial database EXASOL EXASolution 5.

'Preference SQL' or 'PSQL': Syntax of the Preference SQL system. This is a research prototype developed at the Chair of Databases and Information Systems of the University of Augsburg. See references for details.
df Optional parameter to specify a data frame on which the preference operates causing a partial evaluation. See show.pref for details.

Details

There are few database systems supporting Skyline queries. A Skyline query consists of a usual SQL query followed by a PREFERENCE-clause (in some rarely used dialects also SKYLINE OF). For example consider a database table r(a,b). The preference selection \( \text{psel}(r, \text{low}(a) \times \text{high}(b)) \) can be expressed by (in the Exasol dialect):

SELECT * FROM r PREFERENCE LOW a PLUS HIGH b

The show.query function generates just the PREFERENCE-clause, i.e. show.query(low(a) * high(b)) returns

PREFERENCE LOW a PLUS HIGH b

As usual in SQL queries, all keywords are not case sensitive, i.e., PLUS or plus does not make any difference.

References


Examples

show.query(low(a) * high(b))

show.query(low(a) * high(b), dialect = 'Preference SQL')
*preference,preference-method
  (complex_pref), 6
*preference-method (complex_pref), 6
+preference,preference-method
  (complex_pref), 6
+preference-method (complex_pref), 6
&preference,preference-method
  (complex_pref), 6
&preference-method (complex_pref), 6
all_pred (pred_succ), 14
all_succ (pred_succ), 14
around (base_pref_macros), 4
as.character,basepref-method
  (show.pref), 20
as.character,complexpref-method
  (show.pref), 20
as.character,emptypref-method
  (show.pref), 20
as.character,preference-method
  (show.pref), 20
as.character,reversepref-method
  (show.pref), 20
as.expression,basepref-method
  (general_pref), 8
as.expression,complexpref-method
  (general_pref), 8
as.expression,emptypref-method
  (general_pref), 8
as.expression,preference-method
  (general_pref), 8
as.expression,reversepref-method
  (general_pref), 8
assoc.df, 4, 15, 17, 21
assoc.df (general_pref), 8
assoc.df,preference-method
  (general_pref), 8
assoc.df< (general_pref), 8
assoc.df<,preference-method
  (general pref), 8
base pref, 2, 5–9, 16, 17, 19
base_pref_macros, 4, 4, 19
between (base_pref Macros), 4
complex pref, 4, 6, 9, 16, 18, 19
empty (general pref), 8
general pref, 4, 8, 8, 19, 21
get_btg, 10, 15
get_btg (plot bg), 11
get_btg dot (plot bg), 11
get_hasse diag, 10, 19
group_by, 18
hasse pred (pred succ), 14
hasse_succ (pred succ), 14
high, 13
high (base pref), 2
high (base pref), 2
igraph, 11, 12
igraph.plotting, 12
init_pred_succ (pred succ), 14
is.base pref (base pref), 2
is.complex pref (complex pref), 6
is.empty pref (general pref), 8
is.preference (general pref), 8
layered (base pref macros), 4
length,basepref-method (general pref), 8
length,complexpref-method
  (general pref), 8
length,emptypref-method (general pref), 8
length,preference-method
  (general pref), 8
length,reversepref-method
  (general pref), 8
low, 13
low (base pref), 2
low_ (base_pref), 2

partial_eval pref, 4
partial_eval pref (show pref), 20
peval, 4, 9
peval (psel), 16
plot_btg, 11, 19
plot_front, 13, 19
pos (base_pref Macros), 4
pred_succ, 14, 19
pref_str (show pref), 20
psel, 3, 8, 10, 11, 16, 19, 21

reverse (complex pref), 6
rPref, 19
rPref-package (rPref), 19

segments, 13
show pref, 9, 19, 20, 22
show query, 7, 19, 21

ture, 5
ture_ (base pref), 2
ture_ (base pref), 2