Package 'widyr'

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Type Package

Title Widen, Process, then Re-Tidy Data

Version 0.1.4

Description Encapsulates the pattern of untidying data into a wide matrix,
performing some processing, then turning it back into a tidy form. This
is useful for several operations such as co-occurrence counts,
correlations, or clustering that are mathematically convenient on wide matrices.

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Encoding UTF-8

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URL https://github.com/dgrtwo/widyr

BugReports https://github.com/dgrtwo/widyr/issues

VignetteBuilder knitr

Imports rlang, dplyr, tidyr, reshape2, tidytext, purrr, Matrix, broom,
tibble, gapminder

Suggests ggraph, igraph, testthat, covr, knitr, janeaustenr,
rmarkdown, unvotes (>= 0.3.0), countrycode, fuzzyjoin, ggplot2,
maps, irlba

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R topics documented:

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cor_sparse *Find the Pearson correlation of a sparse matrix efficiently*

Description

Find the Pearson correlation of a sparse matrix. For large sparse matrix this is more efficient in time and memory than cor(as.matrix(x)). Note that it does not currently work on simple_triplet_matrix objects.

Usage

```
cor_sparse(x)
```

Arguments

x   A matrix, potentially a sparse matrix such as a "dgTMatrix" object

Source

This code comes from mike on this Stack Overflow answer: https://stackoverflow.com/a/9626089/712603.

---

pairwise_cor *Correlations of pairs of items*

Description

Find correlations of pairs of items in a column, based on a "feature" column that links them together. This is an example of the spread-operate-retidy pattern.

Usage

```
pairwise_cor(
  tbl,
  item,
  feature,
  value,
  method = c("pearson", "kendall", "spearman"),
  use = "everything",
  ...
)

pairwise_cor_(
  tbl,
  item,
  feature,
  value,
  method = c("pearson", "kendall", "spearman"),
  use = "everything",
  ...
)
```

Arguments

`tbl`	Table
`item`	Item to compare; will end up in `item1` and `item2` columns
`feature`	Column describing the feature that links one item to others
`value`	Value column. If not given, defaults to all values being 1 (thus a binary correlation)
`method`	Correlation method
`use`	Character string specifying the behavior of correlations with missing values; passed on to `cor`
`...`	Extra arguments passed on to `squarely`, such as `diag` and `upper`

Examples

```
library(dplyr)
library(gapminder)

gapminder %>%
  pairwise_cor(country, year, lifeExp)

gapminder %>%
  pairwise_cor(country, year, lifeExp, sort = TRUE)

# United Nations voting data
if (require("unvotes", quietly = TRUE)) {
  country_cors <- un_votes %>%
```

```
    mutate(vote = as.numeric(vote)) %>%
    pairwise_cor(country, rcid, vote, sort = TRUE)
}
```

---

pairwise_count   *Count pairs of items within a group*

Description

Count the number of times each pair of items appear together within a group defined by "feature."
For example, this could count the number of times two words appear within documents).

Usage

```
pairwise_count(tbl, item, feature, wt = NULL, ...)

pairwise_count_(tbl, item, feature, wt = NULL, ...)
```

Arguments

tbl	Table
item	Item to count pairs of; will end up in item1 and item2 columns
feature	Column within which to count pairs item2 columns
wt	Optionally a weight column, which should have a consistent weight for each feature
...	Extra arguments passed on to squarely, such as diag, upper, and sort

See Also

[squarely](#)

Examples

```
library(dplyr)
dat <- tibble(group = rep(1:5, each = 2),
              letter = c("a", "b",
                         "a", "c",
                         "a", "c",
                         "b", "e",
                         "b", "f"))

# count the number of times two letters appear together
pairwise_count(dat, letter, group)
pairwise_count(dat, letter, group, sort = TRUE)
pairwise_count(dat, letter, group, sort = TRUE, diag = TRUE)
```

---

pairwise_delta   *Delta measure of pairs of documents*

Description

Compute the delta distances (from its two variants) of all pairs of documents in a tidy table.

Usage

```
pairwise_delta(tbl, item, feature, value, method = "burrows", ...)

pairwise_delta_(tbl, item, feature, value, method = "burrows", ...)
```

Arguments

`tbl`	Table
`item`	Item to compare; will end up in `item1` and `item2` columns
`feature`	Column describing the feature that links one item to others
`value`	Value
`method`	Distance measure to be used; see [dist](#)
`...`	Extra arguments passed on to [squarely](#), such as `diag` and `upper`

See Also

[squarely](#)

Examples

```
library(janeaustenr)
library(dplyr)
library(tidytext)

# closest documents in terms of 1000 most frequent words
closest <- austen_books() %>%
  unnest_tokens(word, text) %>%
  count(book, word) %>%
  top_n(1000, n) %>%
  pairwise_delta(book, word, n, method = "burrows") %>%
  arrange(delta)

closest

closest %>%
  filter(item1 == "Pride & Prejudice")

# to remove duplicates, use upper = FALSE
```

```
closest <- austen_books() %>%
  unnest_tokens(word, text) %>%
  count(book, word) %>%
  top_n(1000, n) %>%
  pairwise_delta(book, word, n, method = "burrows", upper = FALSE) %>%
  arrange(delta)

# Can also use Argamon's Linear Delta
closest <- austen_books() %>%
  unnest_tokens(word, text) %>%
  count(book, word) %>%
  top_n(1000, n) %>%
  pairwise_delta(book, word, n, method = "argamon", upper = FALSE) %>%
  arrange(delta)
```

---

pairwise_dist   *Distances of pairs of items*

Description

Compute distances of all pairs of items in a tidy table.

Usage

```
pairwise_dist(tbl, item, feature, value, method = "euclidean", ...)

pairwise_dist_(tbl, item, feature, value, method = "euclidean", ...)
```

Arguments

tbl	Table
item	Item to compare; will end up in item1 and item2 columns
feature	Column describing the feature that links one item to others
value	Value
method	Distance measure to be used; see [dist](#)
...	Extra arguments passed on to [squarely](#), such as diag and upper

See Also

[squarely](#)

Examples

```
library(gapminder)
library(dplyr)

# closest countries in terms of life expectancy over time
closest <- gapminder %>%
  pairwise_dist(country, year, lifeExp) %>%
  arrange(distance)

closest

closest %>%
  filter(item1 == "United States")

# to remove duplicates, use upper = FALSE
gapminder %>%
  pairwise_dist(country, year, lifeExp, upper = FALSE) %>%
  arrange(distance)

# Can also use Manhattan distance
gapminder %>%
  pairwise_dist(country, year, lifeExp, method = "manhattan", upper = FALSE) %>%
  arrange(distance)
```

---

pairwise_pmi	*Pointwise mutual information of pairs of items*

Description

Find pointwise mutual information of pairs of items in a column, based on a "feature" column that links them together. This is an example of the spread-operate-retidy pattern.

Usage

```
pairwise_pmi(tbl, item, feature, sort = FALSE, ...)

pairwise_pmi_(tbl, item, feature, sort = FALSE, ...)
```

Arguments

tbl	Table
item	Item to compare; will end up in item1 and item2 columns
feature	Column describing the feature that links one item to others
sort	Whether to sort in descending order of the pointwise mutual information
...	Extra arguments passed on to squarely, such as diag and upper

Value

A tbl_df with three columns, item1, item2, and pmi.

Examples

```
library(dplyr)

dat <- tibble(group = rep(1:5, each = 2),
              letter = c("a", "b",
                         "a", "c",
                         "a", "c",
                         "b", "e",
                         "b", "f"))

# how informative is each letter about each other letter
pairwise_pmi(dat, letter, group)
pairwise_pmi(dat, letter, group, sort = TRUE)
```

---

pairwise_similarity   *Cosine similarity of pairs of items*

Description

Compute cosine similarity of all pairs of items in a tidy table.

Usage

```
pairwise_similarity(tbl, item, feature, value, ...)

pairwise_similarity_(tbl, item, feature, value, ...)
```

Arguments

tbl	Table
item	Item to compare; will end up in item1 and item2 columns
feature	Column describing the feature that links one item to others
value	Value
...	Extra arguments passed on to [squarely](#), such as diag and upper

See Also

[squarely](#)

Examples

```
library(janeaustenr)
library(dplyr)
library(tidytext)

# Comparing Jane Austen novels
austen_words <- austen_books() %>%
  unnest_tokens(word, text) %>%
  anti_join(stop_words, by = "word") %>%
  count(book, word) %>%
  ungroup()

# closest books to each other
closest <- austen_words %>%
  pairwise_similarity(book, word, n) %>%
  arrange(desc(similarity))

closest

closest %>%
  filter(item1 == "Emma")
```

---

squarely   *A special case of the widely adverb for creating tidy square matrices*

Description

A special case of `widely`. Used to pre-prepare and post-tidy functions that take an m x n (m items, n features) matrix and return an m x m (item x item) matrix, such as a distance or correlation matrix.

Usage

```
squarely(.f, diag = FALSE, upper = TRUE, ...)

squarely_(.f, diag = FALSE, upper = TRUE, ...)
```

Arguments

.f	Function to wrap
diag	Whether to include diagonal (i = j) in output
upper	Whether to include upper triangle, which may be duplicated
...	Extra arguments passed on to `widely`

Value

Returns a function that takes at least four arguments:

tbl	A table
item	Name of column to use as rows in wide matrix
feature	Name of column to use as columns in wide matrix
feature	Name of column to use as values in wide matrix
...	Arguments passed on to inner function

See Also

widely, pairwise_count, pairwise_cor, pairwise_dist

Examples

```
library(dplyr)
library(gapminder)

closest_continent <- gapminder %>%
  group_by(continent) %>%
  squarely(dist)(country, year, lifeExp)
```

---

widely   *Adverb for functions that operate on matrices in "wide" format*

Description

Modify a function in order to pre-cast the input into a wide matrix format, perform the function, and then re-tidy (e.g. melt) the output into a tidy table.

Usage

```
widely(.f, sort = FALSE, sparse = FALSE, maximum_size = 1e+07)

widely_(.f, sort = FALSE, sparse = FALSE, maximum_size = 1e+07)
```

Arguments

.f	Function being wrapped
sort	Whether to sort in descending order of value
sparse	Whether to cast to a sparse matrix
maximum_size	To prevent crashing, a maximum size of a non-sparse matrix to be created. Set to NULL to allow any size matrix.

Value

Returns a function that takes at least four arguments:

`tbl`	A table
`row`	Name of column to use as rows in wide matrix
`column`	Name of column to use as columns in wide matrix
`value`	Name of column to use as values in wide matrix
`...`	Arguments passed on to inner function

`widely` creates a function that takes those columns as bare names, `widely_` a function that takes them as strings.

Examples

```
library(dplyr)
library(gapminder)

gapminder

gapminder %>%
  widely(dist)(country, year, lifeExp)

# can perform within groups
closest_continent <- gapminder %>%
  group_by(continent) %>%
  widely(dist)(country, year, lifeExp)
closest_continent

# for example, find the closest pair in each
closest_continent %>%
  top_n(1, -value)
```

---

widely_hclust	*Cluster pairs of items into groups using hierarchical clustering*

Description

Reshape a table that represents pairwise distances into hierarchical clusters, returning a table with `item` and `cluster` columns.

Usage

```
widely_hclust(tbl, item1, item2, distance, k = NULL, h = NULL)
```

Arguments

`tbl`	Table
`item1`	First item
`item2`	Second item
`distance`	Distance column
`k`	The desired number of groups
`h`	Height at which to cut the hierarchically clustered tree

See Also

cutree

Examples

```
library(gapminder)
library(dplyr)

# Construct Euclidean distances between countries based on life
# expectancy over time
country_distances <- gapminder %>%
  pairwise_dist(country, year, lifeExp)

country_distances

# Turn this into 5 hierarchical clusters
clusters <- country_distances %>%
  widely_hclust(item1, item2, distance, k = 8)

# Examine a few such clusters
clusters %>% filter(cluster == 1)
clusters %>% filter(cluster == 2)
```

---

widely_kmeans   *Cluster items based on k-means across features*

Description

Given a tidy table of features describing each item, perform k-means clustering using kmeans and retidy the data into one-row-per-cluster.

Usage

```
widely_kmeans(tbl, item, feature, value, k, fill = 0, ...)
```

Arguments

`tbl`	Table
`item`	Item to cluster (as a bare column name)
`feature`	Feature column (dimension in clustering)
`value`	Value column
`k`	Number of clusters
`fill`	What to fill in for missing values
`...`	Other arguments passed on to [kmeans](#)

See Also

[widely_hclust](#)

Examples

```
library(gapminder)
library(dplyr)

clusters <- gapminder %>%
  widely_kmeans(country, year, lifeExp, k = 5)

clusters

clusters %>%
  count(cluster)

# Examine a few clusters
clusters %>% filter(cluster == 1)
clusters %>% filter(cluster == 2)
```

---

widely_svd	*Turn into a wide matrix, perform SVD, return to tidy form*

Description

This is useful for dimensionality reduction of items, especially when setting a lower nv.

Usage

```
widely_svd(tbl, item, feature, value, nv = NULL, weight_d = FALSE, ...)

widely_svd_(tbl, item, feature, value, nv = NULL, weight_d = FALSE, ...)
```

Arguments

`tbl`	Table
`item`	Item to perform dimensionality reduction on; will end up in `item` column
`feature`	Column describing the feature that links one item to others.
`value`	Value
`nv`	Optional; the number of principal components to estimate. Recommended for matrices with many features.
`weight_d`	Whether to multiply each value by the d principal component.
`...`	Extra arguments passed to `svd` (if nv is NULL) or `irlba` (if nv is given)

Value

A tbl_df with three columns. The first is retained from the `item` input, then `dimension` and `value`. Each row represents one principal component value.

Examples

```
library(dplyr)
library(gapminder)

# principal components driving change
gapminder_svd <- gapminder %>%
  widely_svd(country, year, lifeExp)

gapminder_svd

# compare SVDs, join with other data
library(ggplot2)
library(tidyr)

gapminder_svd %>%
  spread(dimension, value) %>%
  inner_join(distinct(gapminder, country, continent), by = "country") %>%
  ggplot(aes(`1`, `2`, label = country)) +
  geom_point(aes(color = continent)) +
  geom_text(vjust = 1, hjust = 1)
```

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