

# Package ‘widy’

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**Type** Package

**Title** Widen, Process, then Re-Tidy Data

**Version** 0.1.2

**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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**LazyData** TRUE

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**URL** <http://github.com/dgrtwo/widy>

**BugReports** <http://github.com/dgrtwo/widy/issues>

**VignetteBuilder** knitr

**Imports** dplyr, tidyr, reshape2, tidytext, purrr, Matrix, broom

**Suggests** ggraph, igraph, gapminder, testthat, covr, knitr, topicmodels, janeaustenr, rmarkdown, unvotes (>= 0.2.0), countrycode, fuzzyjoin, ggplot2, maps, irlba

**RoxygenNote** 6.1.1

**NeedsCompilation** no

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cor_sparse	<i>Find the Pearson correlation of a sparse matrix efficiently</i>
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### 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: <http://stackoverflow.com/a/9626089/712603>.

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pairwise_cor	<i>Correlations of pairs of items</i>
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### 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-retain 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
library(unvotes)

country_cors <- un_votes %>%
  mutate(vote = as.numeric(vote)) %>%
  pairwise_cor(country, rcid, vote, sort = TRUE)

country_cors
```

---

pairwise_count	<i>Count pairs of items within a group</i>
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---

**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**

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

**See Also**

[squarely](#)

**Examples**

```
library(dplyr)
dat <- data_frame(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 <a href="#">dist</a>
...	Extra arguments passed on to <a href="#">squarely</a> , 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)
```

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pairwise_dist	<i>Distances of pairs of items</i>
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### 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 <a href="#">dist</a>
...	Extra arguments passed on to <a href="#">squarely</a> , 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 <- data_frame(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 <a href="#">squarely</a> , 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")
```



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squarely	<i>A special case of the widely adverb for creating tidy square matrices</i>
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---

### Description

A special case of [widely](#). Used to pre-prepare and post-tidy functions that take an  $m \times n$  ( $m$  items,  $n$  features) matrix and return an  $m \times m$  (item  $\times$  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	<i>Adverb for functions that operate on matrices in "wide" format</i>
--------	---

---

**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\_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

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