

# Regular expressions and reshaping using data tables and the nc package

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**Abstract** Regular expressions are powerful tools for extracting tables from non-tabular text data. Capturing regular expressions that describe information to extract from column names can be especially useful when reshaping a data table from wide (one row with many columns) to tall (one column with many rows). We present the R package **nc**, which provides functions for data reshaping, regular expressions, and a uniform interface to three C libraries (PCRE, RE2, ICU). We describe the main features of **nc**, then provide detailed comparisons with related R packages (**stats**, **utils**, **data.table**, **tidyr**, **reshape2**, **cdata**).

## Introduction

Regular expressions are powerful tools for text processing that are available in many programming languages, including R. A regular expression *pattern* defines a set of *matches* in a *subject* string. For example, the pattern `.*[.]*` matches zero or more non-newline characters, followed by a period, followed by zero or more non-newline characters. It would match the subjects `Sepal.Length` and `Petal.Width`, but it would not match in the subject `Species`.

The focus of this article is patterns with capture groups, which are typically defined using parentheses. For example, the pattern `(.*)[.](.*)` results in the same matches as the pattern in the previous paragraph, and it additionally allows the user to capture and extract the substrings by group index (e.g. group 1 matches `Sepal`, group 2 matches `Length`).

Named capture groups allow extracting the a substring by name rather than by index. Using names rather than indices is useful in order to create more readable regular expressions (names document the purpose of each sub-pattern), and to create more readable R code (it is easier to understand the intent of named references than numbered references). For example, the pattern `(?<part>.*)[.](?<dimension>.*)` documents that the flower part appears before the measurement dimension; the part group matches `Sepal` and the dimension group matches `Length`.

Recently, [Hocking \(2019a\)](#) proposed a new syntax for defining named capture groups in R code. Using this new syntax, named capture groups are specified using named arguments in R, which results in code that is easier to read and modify than capture groups defined in string literals. For example, the pattern in the previous paragraph can be written as `part=".*", "[.]", dimension=".*"`. Sub-patterns can be grouped for clarity and/or re-used using lists, and numeric data may be extracted by specifying group-specific type conversion functions.

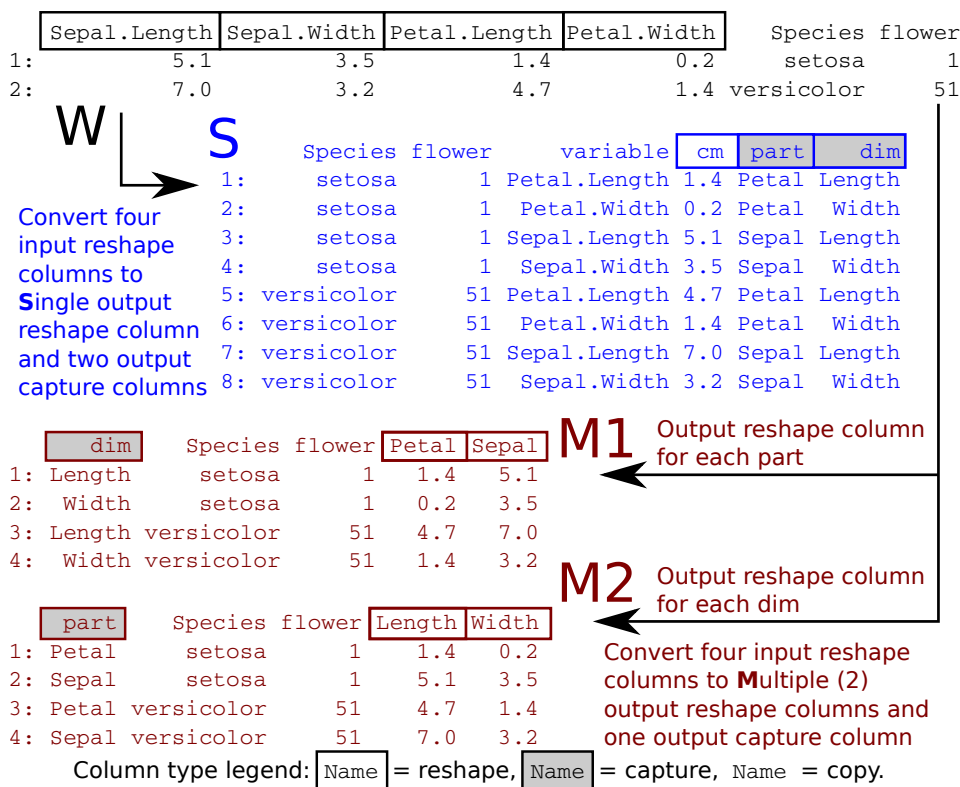
A main thesis of this article is that regular expressions can greatly simplify the code required to specify wide-to-tall data reshaping operations. For one such operation the input is a “wide” table with many columns, and the desired output is a “tall” table with more rows, and some of the input columns converted into a smaller number of output columns (Figure 1). To clarify the discussion we first define three terms that we will use to refer to the different types of columns involved in this conversion:

**Reshape** columns contain the data which is present in the same amount but in different shapes in the input and output.

**Copy** columns contain data in the input which are each copied to multiple rows in the output.

**Capture** columns are only present in the output, and contain data which come from matching a capturing regex pattern to the input reshape column names.

For example the wide iris data (W in Figure 1) have four numeric columns to reshape: `Sepal.Length`, `Sepal.Width`, `Petal.Length`, `Petal.Width`. For some purposes (e.g. displaying a histogram of each reshape input column using facets in [ggplot2](#)) the desired reshaping operation results in a table with a single reshape output column (S in Figure 1), two copied columns, and two columns captured from the names of the reshaped input columns. For other purposes (e.g. scatterplot to compare `Petal` and `Sepal` sizes) the desired reshaping operation results in a table with multiple reshape output columns (M1 with `Sepal` and `Petal` columns in Figure 1), two copied columns, and one column captured from the names of the reshaped input columns. We propose to use the new regular expression syntax of [Hocking \(2019a\)](#), e.g. `part=".*", "[.]", dimension=".*"`, to define both types of wide-to-tall data reshaping operations. In particular, we propose using a single capturing regular expression for defining both (1) the subset of reshape input columns to convert, and (2) the additional capture output columns. We will show that this results in a simple, powerful, non-repetitive syntax for wide-to-tall data reshaping.



**Figure 1:** Two rows of the iris data set (W, black) are considered as the input to a wide-to-tall reshape operation. Four input reshape columns are converted to either a single output reshape column (S, blue) or multiple (2) output reshape columns (M1, M2, red). Other output columns are either copied from the non-reshaped input data, or captured from the names of the reshaped input columns.

In this article our original contribution is the R package `nc` which provides a new implementation of the previously proposed named capture regex syntax of [Hocking \(2019a\)](#), in addition to several new functions that perform wide-to-tall data reshaping using regular expressions. The main new ideas are (1) using un-named capture groups in the regex string literal to provide a uniform interface to three regex C libraries, (2) integration of capture groups and `data.table` functionality ([Dowle and Srinivasan, 2019](#)), and (3) specifying wide-to-tall reshape operations with a concise syntax which results in less repetitive user code than other packages. A secondary contribution of this article is a detailed comparison of current R functions for reshaping data with regular expressions.

The organization of this article is as follows. The rest of this introduction provides an overview of current R packages for regular expressions and data reshaping. The second section describes the proposed functions of the `nc` package. The third section provides detailed comparisons with other R packages, in terms of syntax and computation times. The article concludes with a summary and discussion.

## Related work

There are many R functions which can extract tables from non-tabular text using regular expressions. Recommended R package functions include `base::regexr` and `base::gregexpr` as well as `utils::strcapture`. CRAN packages include `namedCapture` ([Hocking, 2019b](#)), `rematch2` ([Csárdi, 2017](#)), `rex` ([Ushey et al., 2017](#)), `stringr` ([Wickham, 2018](#)), `stringi` ([Gagolewski, 2018](#)), `tidyr` ([Wickham and Henry, 2018](#)), and `re2r` ([Wenfeng, 2017](#)). [Hocking \(2019a\)](#) provides a detailed comparison of these packages in terms of features, syntax, and computation time.

For reshaping data from wide (one row with many columns) to tall (one column with many rows), there are several different R functions that provide similar functionality. Each function supports a different set of features (Table 1); each feature/column is explained in detail below:

- single** refers to support for converting input columns of the same type to a single output column.
- multiple** refers to support for converting input columns of possibly different types to multiple output columns; “sorted” means that conversion works correctly only if the columns are sorted

pkg::function	single	multiple	regex	na.rm	types	list
nc::capture_melt_multiple	no	unsorted	capture	yes	any	yes
nc::capture_melt_single	yes	no	capture	yes	any	yes
tidyr::pivot_longer	yes	unsorted	capture	yes	some	yes
stats::reshape	yes	sorted	capture	no	some	no
data.table::melt, patterns	yes	sorted	match	yes	no	yes
tidyr::gather	yes	no	no	yes	some	yes
reshape2::melt	yes	no	no	yes	no	no
cdata::rowrecs_to_blocks	yes	unsorted	no	no	no	yes
cdata::unpivot_to_blocks	yes	no	no	no	no	yes
utils::stack	yes	no	no	no	no	no

**Table 1:** Reshaping functions in R support various features: “single” for converting input columns into a single output column; “multiple” for converting input columns (either “sorted” in a regular order, or “unsorted” for any order) into multiple output columns of different types; “regex” for regular expressions to “match” input column names or to “capture” and create new output column names; “na.rm” for removal of missing values; “types” for converting input column names to non-character output columns; “list” for output of list columns.

in a regular order, e.g. Sepal.Length, Sepal.Width, Petal.Length, Petal.Width; “unsorted” means that conversion works correctly even if the columns are not sorted, e.g. Sepal.Length, Sepal.Width, Petal.Width, Petal.Length.

**regex** refers to support for regular expressions; “match” means a pattern is used to match the input column names; “capture” means that the specified pattern is used to create new output columns, which is especially useful when the names consist of several distinct pieces of information, e.g. Sepal.Length; “no” means that regular expressions are not directly supported (although `base::grep` can always be used).

**na.rm** refers to support for removing missing values.

**types** refers to support for converting captured text to numeric output columns.

**list** refers to support for output of list columns.

Recommended R package functions include `stats::reshape` and `utils::stack` for reshaping data from wide to tall. Of the features listed in Table 1, `utils::stack` only supports single column output, whereas `stats::reshape` supports the following features. For data with regular input column names (output column, separator, time value), regular expressions can be used to specify the separator (e.g. in `Sepal.Length`, `Sepal` is output column, `dot` is separator, `Length` is time value). Multiple output columns are supported, but incorrect output may be computed if input columns are not sorted in a regular order. The time value is captured by default to an output column named `time`. Automatic type conversion is performed on time values when possible, but custom type conversion functions are not supported. There is neither support for missing value removal nor list column output.

The **tidyr** package provides two functions for reshaping data from wide to tall format: `gather` and `pivot_longer`. The older `gather` function only supports converting input columns to a single output column (not multiple). The input columns to convert may not be directly specified using regular expressions; instead R expressions such as `x:y` can be used to indicate all columns starting from `x` and ending with `y`. It does support limited type conversion; if the `convert=TRUE` argument is specified, the `utils::type.convert` function is used to convert the input column names to numeric, integer, or logical. In contrast the newer `pivot_longer` also supports multiple output columns (even if input columns are unsorted), and regular expressions for capturing new output columns from the input column names (but `grep` must be used to specify the input columns to convert). Limited type conversion is also supported in `pivot_longer`, via the `names_ptypes` argument, which should be a list with names corresponding to output columns and values corresponding to prototypes (zero-length atomic vectors, e.g. `numeric()`). Both functions support list columns and removing missing values, although different arguments are used (`na.rm` for `gather`, `values_drop_na` for `pivot_longer`).

The **reshape2** and **data.table** packages each provide a `melt` function for converting data from wide to tall (Wickham, 2007; Dowle and Srinivasan, 2019). The older **reshape2** version only supports converting input columns to a single output column, whereas the newer **data.table** version also supports multiple output columns. Regular expressions are not supported in **reshape2**, but can be used with `data.table::patterns` to match input column names to convert (although the output can be incorrect if columns are not sorted in a regular order). Neither function supports type conversion, and both functions support removing missing values from the output using the `na.rm` argument. List column output is supported in **data.table** but not **reshape2**.

The `cddata` package provides several functions for data reshaping, including `rowrecs_to_blocks` and `unpivot_to_blocks` which can convert data from wide to tall (Mount and Zume1, 2019). The simpler of the two functions is `unpivot_to_blocks`, which supports a single output column (interface similar to `reshape2::melt/tidyr::gather`). The user of `rowrecs_to_blocks` must provide a control table that describes how the input should be reshaped into the output. It therefore supports multiple output columns, for possibly unsorted input columns. Both functions support list column output, but other features from Table 1 are not supported (regular expressions, missing value removal, type conversion).

## New features in `nc`

The `nc` package provides new regular expression functionality based on the syntax recently proposed by Hocking (2019a). In this section we first discuss how the `nc` package implements this syntax as a front-end to three regex engines, and we then discuss the new features for data reshaping using regular expressions.

### Uniform interface to three regex engines

Several C libraries providing regular expression engines are available in R. The standard R distribution has included and the Perl-Compatible Regular Expressions (PCRE) C library since 2002 (R Core Team, 2002). CRAN package `re2r` provides the RE2 library, and `stringi` provides the ICU library. Each of these regex engines has a unique feature set, and may be preferred for different applications. For example, PCRE is installed by default, RE2 guarantees matching in polynomial time, and ICU provides strong unicode support. For a more detailed comparison of the relative strengths of each regex library, we refer the reader to the recent paper of (Hocking, 2019a).

Each regex engine has also a different R interface, so switching from one engine to another may require non-trivial modifications of user code. In order to make switching between engines easier, Hocking (2019a) introduced the `namedCapture` package, which provides a uniform interface for capturing text using PCRE and RE2. The user may specify the desired engine via e.g. `options(namedCapture.engine="PCRE")`; the `namedCapture` package provides the output in a uniform format. However `namedCapture` requires the engine to support specifying capture group names in regex pattern strings, and to support output of the group names to R (which ICU does not support).

Our proposed `nc` package provides support for the ICU engine in addition to PCRE and RE2. The `nc` package implements this functionality using un-named capture groups, which are supported in all three regex engines. In particular, a regular expression is constructed in R code that uses named arguments to indicate captures, which are translated to un-named groups when passed to the regex engine. For example, consider a user who wants to capture the two parts of the column names of the iris data, e.g. `Sepal.Length`. The user would typically specify the capturing regular expression as a string literal, e.g. `"(.*)[.](.*)"`. Using `nc` the same pattern can be applied to the iris data column names via

```
> nc::capture_first_vec(
+   names(iris), part=".*", "[.]", dim=".*", engine="RE2", nomatch.error=FALSE)

  part    dim
1: Sepal Length
2: Sepal  Width
3: Petal  Length
4: Petal  Width
5:  <NA>   <NA>
```

The first argument specifies the subject and the following three arguments define the pattern. Each named R argument in the pattern generates an un-named capture group by enclosing the specified value in parentheses, e.g. `(.*)`. All of the sub-patterns are pasted together in the sequence they appear in order to create the final pattern that is used with the specified regex engine. The `nomatch.error=FALSE` argument is given because the default is to stop with an error if any subjects do not match the specified pattern (the fifth subject `Species` does not match). In order to see the pattern that is generated and passed to the specified regex engine, we can use

```
> nc::var_args_list(part=".*", "[.]", dim=".*")$pattern

[1] "(.*)[.](.*)"
```

TODO

## Wide-to-tall data reshaping

TODO

## Comparison with other packages

TODO

## Discussion and conclusions

TODO

**Reproducible research statement.** The source code for this article can be freely downloaded from <https://github.com/tdhock/nc-article>

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