
Strings in R

Introduction to Regular expression Regex in {stringr}	
Pattern Matching & Substitution: str_detect, str_sub, str_extract, str_replace	

Introduction to Regular expression Regex in {stringr}

We would now like to delve into a family of functions designed for pattern matching, extraction, and substitution in strings, namely `str_detect`, `str_extract`, and `str_replace`. However, before exploring these functions, it's crucial to understand Regular Expressions (Regex), a powerful tool for pattern matching in strings.

Regular Expressions are sequences of characters forming a search pattern used to identify, match, and manipulate specific parts of strings.

Let's start with `str_sub`, a basic but powerful function in {stringr}, to get a feel for string manipulation, which will serve as a stepping stone to understanding more complex pattern matching with Regex.

Using `str_sub` for String Subsetting

The `str_sub` function in {stringr} is used for extracting or replacing substrings within a string. It's a straightforward yet effective tool for accessing specific parts of strings.

Basic Usage

The basic usage of `str_sub` involves specifying the string, the start position, and the end position of the substring you want to extract:

```
library(stringr)

str_sub("Hello World", 1, 5) # Extracts "Hello"
```

This code snippet extracts the first five characters from "Hello World".

Negative Indices

`str_sub` also allows the use of negative indices. Negative values are counted backward from the end of the string, making it easy to extract substrings from the end:

```
str_sub("Hello World", -5, -1) # Extracts "World"
```

This example extracts the last five characters of the string.

Flexibility without Regex

Even without Regex, `str_sub` can be quite flexible. For instance, you can replace a specific portion of a string:

```
example_string <- "Hello World"
str_sub(example_string, 1, 5) <- "Goodbye"
example_string # Now it's "Goodbye World"
```

This code changes the first five characters of “Hello World” to “Goodbye”.

Using `str_sub` with Regex

While `str_sub` doesn't inherently use Regex for its basic operations, understanding how to extract or replace parts of a string is foundational for more advanced pattern matching with Regex, which will be explored in functions like `str_detect`, `str_extract`, and `str_replace`.

In the next sections, we will explore how these functions use Regex to enable powerful and complex string manipulations, building on the understanding developed through `str_sub`.

Pattern Matching & Substitution: `str_detect`, `str_sub`, `str_extract`, `str_replace`

Next...

But first: Brief Look at Regular Expressions

Regular expressions, commonly referred to as “regex”, are sequences of characters that define search patterns. Think of them as wildcards! When working with strings, these patterns help in recognizing complex sequences, be it specific words, numbers, or a mix of characters in a particular order.

With regex, you can define what you're looking for, making tasks like data extraction and string cleaning more straightforward.

For more information about regex, you can refer to cheatsheets for a quick overview. A handy one for `stringr` with a dedicated section for regex can be found [here](#).

Regular Expressions - Regular expressions, or *regexps*, are a concise language for describing patterns in strings.

MATCH CHARACTERS

string (type this)	regex (to mean this)	matches (which matches this)	example
<code>\\.</code>	<code>.</code>	a (etc.)	<code>see("a")</code> abc ABC 123 .!?\0{
<code>\\!</code>	<code>!</code>	!	<code>see("\\!")</code> abc ABC 123 .!?\0{
<code>\\?</code>	<code>?</code>	?	<code>see("\\?")</code> abc ABC 123 .!?\0{
<code>\\ </code>	<code> </code>		<code>see("\\ ")</code> abc ABC 123 .!?\0{
<code>\\(</code>	<code>(</code>	(<code>see("\\(")</code> abc ABC 123 .!?\0{
<code>\\)</code>	<code>)</code>)	<code>see("\\)")</code> abc ABC 123 .!?\0{
<code>\\{</code>	<code>{</code>	{	<code>see("\\{")</code> abc ABC 123 .!?\0{
<code>\\}</code>	<code>}</code>	}	<code>see("\\}")</code> abc ABC 123 .!?\0{
<code>\\n</code>	<code>\n</code>	new line (return)	<code>see("\\n")</code> abc ABC 123 .!?\0{
<code>\\t</code>	<code>\t</code>	tab	<code>see("\\t")</code> abc ABC 123 .!?\0{
<code>\\s</code>	<code>\s</code>	any whitespace (<i>\S</i> for non-whitespaces)	<code>see("\\s")</code> abc ABC 123 .!?\0{
<code>\\d</code>	<code>\d</code>	any digit (<i>\D</i> for non-digits)	<code>see("\\d")</code> abc ABC 123 .!?\0{
<code>\\w</code>	<code>\w</code>	any word character (<i>\W</i> for non-word chars)	<code>see("\\w")</code> abc ABC 123 .!?\0{
<code>\\b</code>	<code>\b</code>	word boundaries	<code>see("\\b")</code> abc ABC 123 .!?\0{
<code>[[:digit:]]</code>		digits	<code>see("[[:digit:]]")</code> abc ABC 123 .!?\0{
<code>[[:alpha:]]</code>		letters	<code>see("[[:alpha:]]")</code> abc ABC 123 .!?\0{
<code>[[:lower:]]</code>		lowercase letters	<code>see("[[:lower:]]")</code> abc ABC 123 .!?\0{
<code>[[:upper:]]</code>		uppercase letters	<code>see("[[:upper:]]")</code> abc ABC 123 .!?\0{
<code>[[:alnum:]]</code>		letters and numbers	<code>see("[[:alnum:]]")</code> abc ABC 123 .!?\0{
<code>[[:punct:]]</code>		punctuation	<code>see("[[:punct:]]")</code> abc ABC 123 .!?\0{
<code>[[:graph:]]</code>		letters, numbers, and punctuation	<code>see("[[:graph:]]")</code> abc ABC 123 .!?\0{
<code>[[:space:]]</code>		space characters (i.e. \s)	<code>see("[[:space:]]")</code> abc ABC 123 .!?\0{
<code>[[:blank:]]</code>		space and tab (but not new line)	<code>see("[[:blank:]]")</code> abc ABC 123 .!?\0{
<code>.</code>		every character except a new line	<code>see(".")</code> abc ABC 123 .!?\0{

¹ Many base R functions require classes to be wrapped in a second set of [], e.g. `[[:digit:]]`

[[:space:]]

new line

[[:blank:]]

space

tab

[[:graph:]]

[[:punct:]]

[[:symbol:]]

[[:alnum:]]

[[:digit:]]

[[:alpha:]]

[[:lower:]]

[[:upper:]]

Detecting Patterns with `str_detect()`

The `str_detect()` function checks if a string contains a specified pattern. It's best used when you want to filter or categorize data based on the presence of a substring or pattern.

This operation can also be paired with functions like `case_when()` from the `dplyr` package to create new variables or modify existing ones based on whether a pattern is detected.

Suppose that we want to create a new variable based on whether the start date is in a typical format:

```

irs %>%
  mutate(is_typical_format = case_when(
    str_detect(start_date_typical, "\\d{2}/\\d{2}/\\d{4}") ~ TRUE,
    TRUE ~ FALSE))

```

```

## # A tibble: 112 × 10
##   village      target_spray sprayed coverage_p
##   <chr>          <dbl>    <dbl>    <dbl>
## 1 Mess              87      64      73.6
## 2 Nkombedzi        183     169      92.4
## 3 B Compound        16      16     100
## 4 D Compound         3        2     66.7
## 5 Post Office         6        3      50

```

```
## 6 Mangulenje          375    372    99.2
## 7 Mangulenje Senior    7      4    57.1
## 8 Old School          24     23    95.8
## 9 Mwanza              671    636    94.8
## 10 Alumenda           226    226    100
## # i 102 more rows
## # i 6 more variables: start_date_default <date>, ...
```

The syntax is:

- The string input
- The pattern to look for
- If set to TRUE, the negate argument returns non-matches elements.

Creating Subsets of Strings with `str_sub()` and `str_extract`

Sometimes you might want to grab a particular segment of a string, like the year from a date.

`str_sub()` is used to extract or replace substrings from a string! This function allows you to define the start and end positions to slice the string.

Let see a practical example by extracting the first 4 characters from `start_date_default`. In this case, it would be the year:

```
irs$start_year <- str_sub(irs$start_date_default, start = 1, end =
4)
```

Syntax:

- The string input.
- start and end defines the range of characters

SIDE NOTE



Use `str_sub()` when you have a consistent structure across your strings and you know the exact positions you want to extract!

Q: Detecting Patterns

Which rows in the `irs` dataframe have a `start_date_typical` in the standard “DD/MM/YYYY” format?

Q: Extracting Substrings

Extract the day (first two digits) from the `start_date_typical` column.

The `str_extract()` function is versatile in that it extracts matched patterns from a string using regex! It's useful for cases where the string structure isn't consistent.

For example, if we wanted to extract months from the `start_date_long` variable, we can apply the following syntax:

```
# Example: Extracting the month name from the start_date_long
irs$month_name <- str_extract(irs$start_date_long, "[A-Z][a-z]+",
group = F)
```

Syntax:

- The string input
- The pattern to look for
 - Here, the pattern "**[A-Z][a-z]+**" looks for a capital letter followed by one or more lowercase letters, capturing month names effectively.
- The `group` argument specifies whether to return a list of character vector or matrix. - FALSE (default) - returns a list - TRUE - returns a matrix

Replacing Strings with `str_replace()`

The `str_replace()` function replaces the first instance of a matched pattern in a string.

To illustrate, let's transform the hyphens in `start_date_default` to slashes:

```
irs$start_date_default <- str_replace_all(irs$start_date_default, "-",
"/")
```

Syntax:

- The string input
- The pattern to look for
- The replacement value

Q: Diving into Regex with `str_extract()`

From the `start_date_long` column, extract the day of the month, which should be two digits following a space after the month name.

Q: Replacing Patterns with `str_replace()`

For the `start_date_messy` column, suppose some dates use dots (.) instead of slashes (/). Replace all dots with slashes in this column.
