

Universidade Federal do Rio Grande do Norte
Instituto Metr pole Digital
IMD0601 - Bioestat stica

Organiza  o dos dados em R

Prof. Dr. Tetsu Sakamoto
Instituto Metr pole Digital - UFRN
Sala A224, ramal 182
Email: tetsu@imd.ufrn.br



Baixe a aula (e os arquivos)

- Para aqueles que não clonaram o repositório:

```
> git clone https://github.com/tetsufmbio/IMD0601.git
```

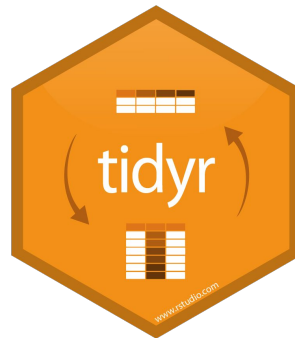
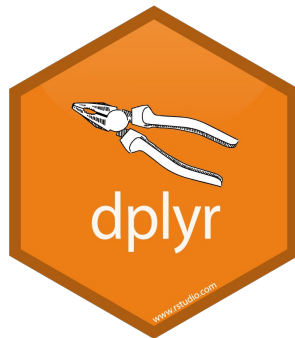
- Para aqueles que já tem o repositório local:

```
> cd /path/to/IMD0601
```

```
> git pull
```

Revisão

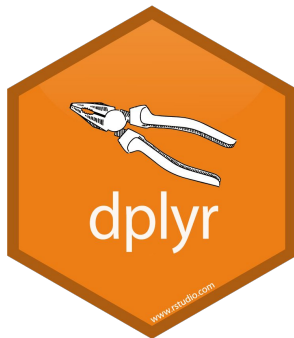
Entender e arrumar os dados



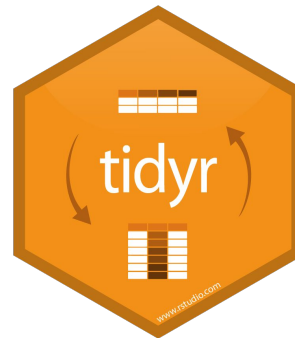
Conceito de uma tabela arrumada:

- Cada variável deve estar em uma coluna;
- Cada observação forma uma linha;
- Cada tipo de unidade observacional forma uma tabela.

Revisão



- `select(<tbl_df>, <var1>, <var2>, ...)`
- `filter(<tbl_df>, <condition1>, <condition2>, ...)`
- `arrange(<tbl_df>, <var1>, desc(<var2>), ...)`
- `mutate(<tbl_df>, <new_var1> = <value>, <new_var2> = <value>, ...)`
- `summarize(<tbl_df>, <column> = <function>)`



- `gather(<tbl_df>, <col1>, <col2>, ...)`
- `separate(<tbl_df>, <col>, into = <vector>)`
- `spread(<tbl_df>, <col>, <col_value>)`
- `unique(<tbl_df>)`
- `bind_rows(<tbl_df1>, <tbl_df2>)`

Revisão

Messy data:

- Valores representando colunas de uma tabela;
- Múltiplas variáveis em uma coluna;
- Variáveis armazenadas tanto em linhas quanto em colunas;
- Múltiplas unidades observacionais em uma única tabela. Dados redundantes na tabela.
- Duas tabelas separadas com a mesma unidade observacional.

<https://makingnoiseandhearingthings.com/2018/04/19/datasets-for-data-cleaning-practice/>

Exercício

Trabalhe com o dados de iris e manipule os dados de forma que ele apresente as seguintes estruturas:

a)

	Species	Part	Measure	Value
1	setosa	Sepal	Length	5.1
2	setosa	Sepal	Length	4.9
3	setosa	Sepal	Length	4.7
4	setosa	Sepal	Length	4.6
5	setosa	Sepal	Length	5.0
6	setosa	Sepal	Length	5.4

b)

	Species	Flower	Part	Length	Width
1	setosa	1	Petal	1.4	0.2
2	setosa	1	Sepal	5.1	3.5
3	setosa	2	Petal	1.4	0.2
4	setosa	2	Sepal	4.9	3.0
5	setosa	3	Petal	1.3	0.2
6	setosa	3	Sepal	4.7	3.2

as.<class>()

```
# logical  
class(TRUE)
```

```
# character  
class("8484.00")
```

```
# numeric  
class(99)
```

```
# character  
class("factor")
```

```
# character  
class("FALSE")
```

```
# Converter para "character"  
class(as.character(TRUE))
```

```
# Converter para "numeric"  
class(as.numeric("8484.00"))
```

```
# Converter para "integer"  
class(as.integer(99))
```

```
# Converter para "factor"  
class(as.factor("factor"))
```

```
# Converter para "logical"  
class(as.logical("FALSE"))
```

lubridate

Pacote que lida com dados temporais em R

```
library(lubridate)
```

```
ymd("2012, Aug 13") # "2012-08-13"
```

```
ymd_hm("2012-08-13 12:43") # "2012-08-13 12:43:00 UTC"
```

```
hms("12:43:54") # "12H 43M 54S"
```



Dates and times with lubridate : : CHEAT SHEET



Date-times



2017-11-28 12:00:00
A **date-time** is a point on the timeline, stored as the number of seconds since 1970-01-01 00:00:00 UTC
`dt <- as_datetime(1511870400)`
"2017-11-28 12:00:00 UTC"

PARSE DATE-TIMES (Convert strings or numbers to date-times)

1. Identify the order of the year (y), month (m), day (d), hour (h), minute (m) and second (s) elements in your data.
2. Use the function below whose name replicates the order. Each accepts a wide variety of input formats.

2017-11-28 14:02:00 `ymd_hms()`, `ymd_hm()`, `ymd_h()`
`ymd_hms("2017-11-28T14:02:00")`

2017-22-12 10:00:00 `ydym_hms()`, `ydym_hm()`, `ydym_h()`
`ydym_hms("2017-22-12 10:00:00")`

11/28/2017 1:02:03 `mdy_hms()`, `mdy_hm()`, `mdy_h()`
`mdy_hms("11/28/2017 1:02:03")`

1 Jan 2017 23:59:59 `dmy_hms()`, `dmy_hm()`, `dmy_h()`
`dmy_hms("1 Jan 2017 23:59:59")`

20170131 `ymd()`, `ymd()`, `ymd(20170131)`

July 4th, 2000 `mdy()`, `myd()`, `mdy("July 4th, 2000")`

4th of July '99 `dmy()`, `dym()`, `dmy("4th of July '99")`

2001: Q3 `yq()` Q for quarter, `yq("2001: Q3")`

2:01 `hms()`, `hms()` Also `lubridate::hms()`, `hm()` and `ms()`, which return periods: `hms(sec = 0, min = 1, hours = 2)`

2017.5 `date_decimal()` (decimal, tz = "UTC")
`date_decimal(2017.5)`

`now(tzone = "")` Current time in tz (defaults to system tz). `now()`

`today(tzone = "")` Current date in a tz (defaults to system tz). `today()`

`fast_strptime()` Faster `strptime`.
`fast_strptime("9/1/01", "%m/%d")`

`parse_date_time()` Easier `strptime`.
`parse_date_time("9/1/01", "ymd")`



2017-11-28
A **date** is a day stored as the number of days since 1970-01-01

`d <- as_date(17498)`
"2017-11-28"

GET AND SET COMPONENTS

Use an accessor function to get a component.
Assign into an accessor function to change a component in place.

2018-01-31 11:59:59

2018-01-31 11:59:59

2018-01-31 11:59:59

2018-01-31 11:59:59

2018-01-31 11:59:59

2018-01-31 11:59:59

2018-01-31 11:59:59



12:00:00
An **hms** is a time stored as the number of seconds since 00:00:00

`t <- hms::as_hms(85)`
"00:01:25"

`d ## "2017-11-28"`
`day(d) ## 28`
`day(d) <- 1`
`d ## "2017-11-01"`

`date(x)` Date component. `date(dt)`

`year(x)` Year. `year(dt)`
`isoyear(x)` The ISO 8601 year.
`epiyear(x)` Epidemiological year.

`month(x, label, abbr)` Month.
`month(dt)`

`day(x)` Day of month. `day(dt)`
`wday(x, label, abbr)` Day of week.
`qday(x)` Day of quarter.

`hour(x)` Hour. `hour(dt)`

`minute(x)` Minutes. `minute(dt)`

`second(x)` Seconds. `second(dt)`

`week(x)` Week of the year. `week(dt)`
`isoweek()` ISO 8601 week.
`epiweek()` Epidemiological week.

`quarter(x, with_year = FALSE)`
Quarter. `quarter(dt)`

`semester(x, with_year = FALSE)`
Semester. `semester(dt)`

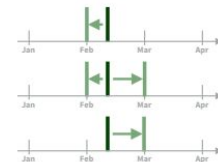
`am(x)` Is it in the am? `am(dt)`
`pm(x)` Is it in the pm? `pm(dt)`

`dst(x)` Is it daylight savings? `dst(d)`

`leap_year(x)` Is it a leap year?
`leap_year(d)`

`update(object, ..., simple = FALSE)`
`update(dt, mday = 2, hour = 1)`

Round Date-times



`floor_date(x, unit = "second")`
Round down to nearest unit.
`floor_date(dt, unit = "month")`

`round_date(x, unit = "second")`
Round to nearest unit.
`round_date(dt, unit = "month")`

`ceiling_date(x, unit = "second")`
change_on_boundary = NULL
Round up to nearest unit.
`ceiling_date(dt, unit = "month")`

`rollback(dates, roll_to_first = FALSE, preserve_hms = TRUE)`
Roll back to last day of previous month. `rollback(dt)`

Stamp Date-times

`stamp()` Derive a template from an example string and return a new function that will apply the template to date strings. Also `stamp_date()` and `stamp_time()`.

1. Derive a template, create a function
`sf <- stamp("Created Sunday, Jan 17, 1999 3:34")`
2. Apply the template to dates
`sf(ymd("2010-04-05"))`
[1] "Created Monday, Apr 05, 2010 00:00"

Tip: use a date with day > 12

Time Zones

R recognizes ~600 time zones. Each encodes the time zone, Daylight Savings Time, and historical calendar variations for an area. R assigns one time zone per vector.

Use the **UTC** time zone to avoid Daylight Savings.

`OlsonNames()` Returns a list of valid time zone names. `OlsonNames()`



`with_tz(time, tzone = "")` Get the same date-time in a new time zone (a new clock time).
`with_tz(dt, "US/Pacific")`

`force_tz(time, tzone = "")` Get the same clock time in a new time zone (a new date-time).
`force_tz(dt, "US/Pacific")`

stringr

```
# Pacote que lida com strings em R
```

```
library(stringr)
```

```
str_detect(students3$dob, "1997")
```

```
str_length("TESTE") # 5
```

```
str_replace("Female", "F", "f") # female
```

```
str_trim("  T  ") # "T"
```

```
str_c("T","E","S","T") # "TEST"
```

```
str_pad(x, width = 6, side = "left", pad = ".")
```



String manipulation with stringr: : CHEAT SHEET



The **stringr** package provides a set of internally consistent tools for working with character strings, i.e. sequences of characters surrounded by quotation marks.

Detect Matches



str_detect(string, pattern) Detect the presence of a pattern match in a string.
`str_detect(fruit, "a")`



str_which(string, pattern) Find the indexes of strings that contain a pattern match.
`str_which(fruit, "a")`



str_count(string, pattern) Count the number of matches in a string.
`str_count(fruit, "a")`



str_locate(string, pattern) Locate the positions of pattern matches in a string. Also **str_locate_all**.
`str_locate(fruit, "a")`

Subset Strings



str_sub(string, start = 1L, end = -1L) Extract substrings from a character vector.
`str_sub(fruit, 1, 3); str_sub(fruit, -2)`



str_subset(string, pattern) Return only the strings that contain a pattern match.
`str_subset(fruit, "b")`



str_extract(string, pattern) Return the first pattern match found in each string, as a vector. Also **str_extract_all** to return every pattern match.
`str_extract(fruit, "[aeiou]")`



str_match(string, pattern) Return the first pattern match found in each string, as a matrix with a column for each () group in pattern. Also **str_match_all**.
`str_match(sentences, "(a|the) ([^]+)")`

Manage Lengths



str_length(string) The width of strings (i.e. number of code points, which generally equals the number of characters). `str_length(fruit)`



str_pad(string, width, side = c("left", "right", "both"), pad = " ") Pad strings to constant width. `str_pad(fruit, 17)`



str_trunc(string, width, side = c("right", "left", "center"), ellipsis = "...") Truncate the width of strings, replacing content with ellipsis. `str_trunc(fruit, 3)`



str_trim(string, side = c("both", "left", "right")) Trim whitespace from the start and/or end of a string. `str_trim(fruit)`

Mutate Strings



str_sub() <- value. Replace substrings by identifying the substrings with **str_sub()** and assigning into the results.
`str_sub(fruit, 1, 3) <- "str"`



str_replace(string, pattern, replacement) Replace the first matched pattern in each string. `str_replace(fruit, "a", "-")`



str_replace_all(string, pattern, replacement) Replace all matched patterns in each string. `str_replace_all(fruit, "a", "-")`



str_to_lower(string, locale = "en")¹ Convert strings to lower case.
`str_to_lower(sentences)`



str_to_upper(string, locale = "en")¹ Convert strings to upper case.
`str_to_upper(sentences)`



str_to_title(string, locale = "en")¹ Convert strings to title case. `str_to_title(sentences)`

Join and Split



str_c(..., sep = "", collapse = NULL) Join multiple strings into a single string.
`str_c(letters, LETTERS)`



str_c(..., sep = "", collapse = NULL) Collapse a vector of strings into a single string.
`str_c(letters, collapse = "")`



str_dup(string, times) Repeat strings times times. `str_dup(fruit, times = 2)`



str_split_fixed(string, pattern, n) Split a vector of strings into a matrix of substrings (splitting at occurrences of a pattern match). Also **str_split** to return a list of substrings.
`str_split_fixed(fruit, " ", n=2)`



str_glue(..., sep = "", .envir = parent.frame()) Create a string from strings and (expressions) to evaluate. `str_glue("Pi is {pi}")`



str_glue_data(x, ..., sep = "", .envir = parent.frame(), na = "NA") Use a data frame, list, or environment to create a string from strings and (expressions) to evaluate.
`str_glue_data(mtcars, "(rownames(mtcars)) has {hp} hp")`

Order Strings



str_order(x, decreasing = FALSE, na_last = TRUE, locale = "en", numeric = FALSE, ...) Return the vector of indexes that sorts a character vector. `x[str_order(x)]`



str_sort(x, decreasing = FALSE, na_last = TRUE, locale = "en", numeric = FALSE, ...) Sort a character vector.
`str_sort(x)`

Helpers



str_conv(string, encoding) Override the encoding of a string. `str_conv(fruit, "ISO-8859-1")`



str_view(string, pattern, match = NA) View HTML rendering of first regex match in each string. `str_view(fruit, "[aeiou]")`

str_view_all(string, pattern, match = NA) View HTML rendering of all regex matches.
`str_view_all(fruit, "[aeiou]")`

str_wrap(string, width = 80, indent = 0, exdent = 0) Wrap strings into nicely formatted paragraphs. `str_wrap(sentences, 20)`

Exercício

Carregue o arquivo data.csv no R(studio). Este arquivo contém dados climatológicos de algumas localidades no Brasil.

Ao explorar os dados, você verificará que os dados estão bagunçados. Utilizando os recursos visto nesta e na aula passada, tente organizar estes dados de forma que cada observação corresponda a cada hora medido.

Descrição sobre as colunas pode ser acessada em:

<https://www.kaggle.com/PROPPG-PPG/hourly-weather-surface-brazil-southeast-region>

Exercício

1. Explore os dados;
2. Divida a tabela de forma que cada tabela tenha dados de uma unidade observacional;
3. Retire colunas redundantes na tabela que contém os dados climatológicos;
4. Formate a coluna que contém a data e a hora utilizando o pacote lubridate;
5. Explore a coluna prcp. O que representa o valor 0 e o NA nesta coluna?
6. Explore a coluna stp. O que representa o valor 0 e o NA nesta coluna?
7. Corrija os valores das colunas prcp, stp, smax e smin em relação aos dados faltantes.

Referência

Esta aula foi baseada no curso “**Cleaning Data in R**” de Nick Carchedi
(<https://www.datacamp.com/courses/cleaning-data-in-r>)