

Preparado para:



REFORM/SC2022/126

DELIVERABLE 4

MÓDULO 2

GESTÃO E TRATAMENTO

DE DADOS EM R

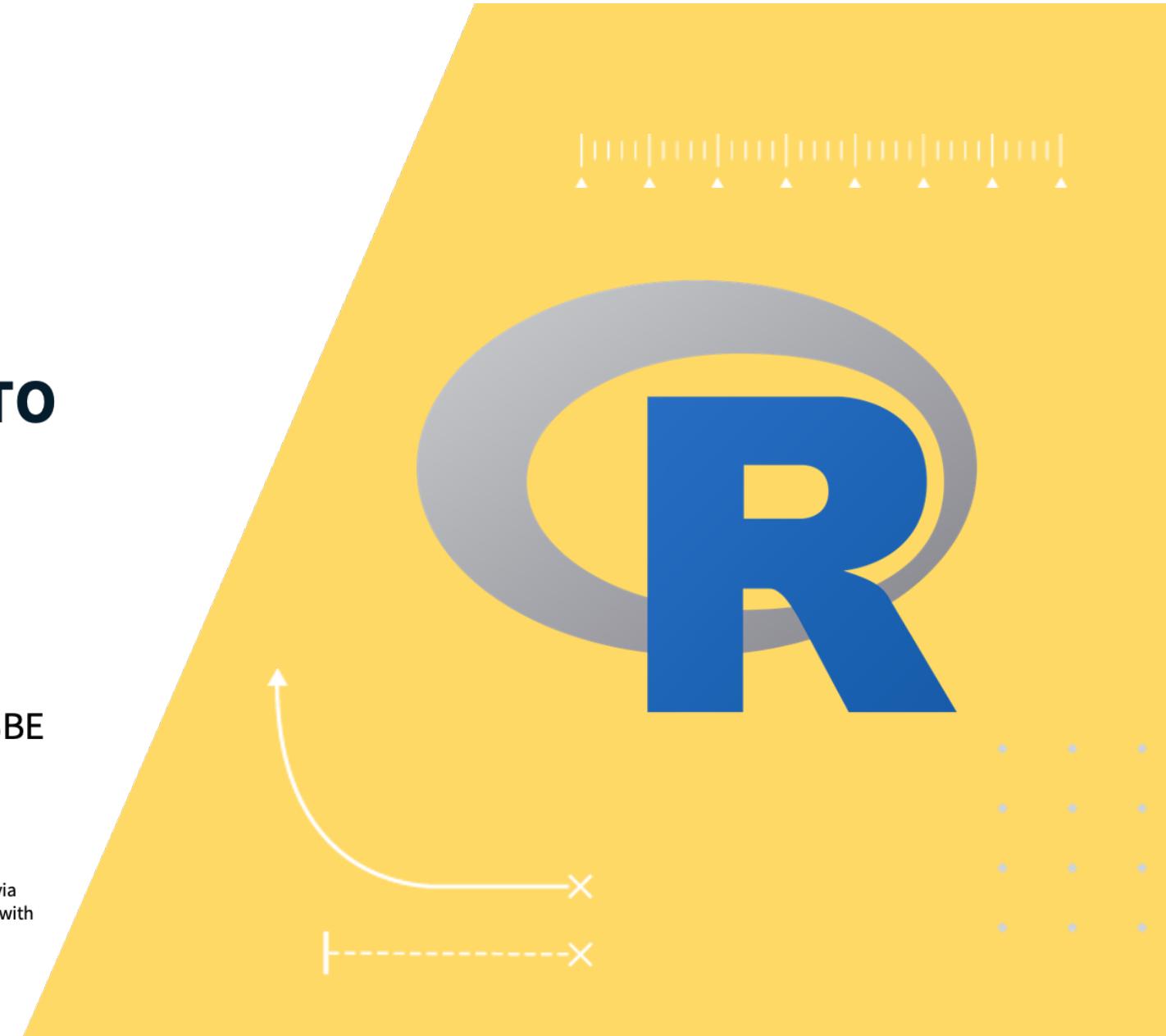
DESIGNING A NEW VALUATION MODEL
FOR RURAL PROPERTIES IN PORTUGAL

Parte II

Formador: Luís Teles Morais | Nova SBE
Lisboa, 25 maio 2023



This project is carried out with funding by the European Union via the Structural Reform Support Programme and in cooperation with the Directorate General for Structural Reform Support of the European Commission



Programa

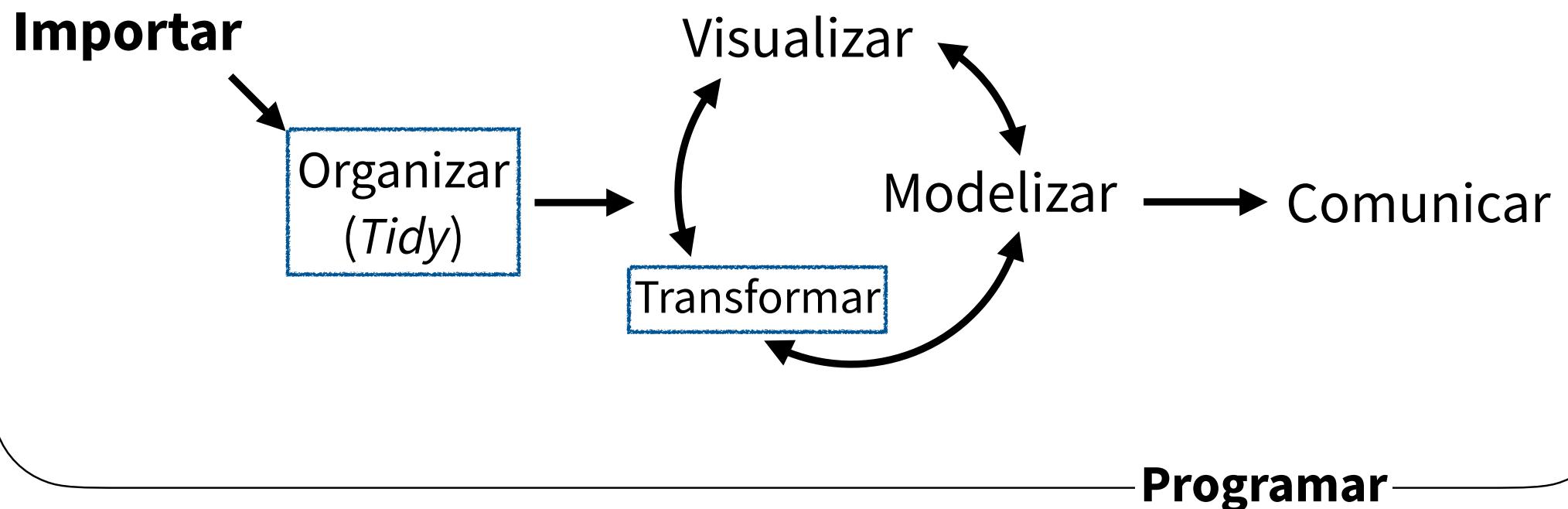
MÓDULOS	DURAÇÃO
Módulo 1 – Introdução ao R: - O que é o R? - Como instalar e configurar o R. - Sintaxe básica e comandos. - Tipos de dados, objetos e classes.	4 Horas
Módulo 2 – Gestão e tratamento de dados em R: - Carregar dados no R. - Perceber as estruturas de dados e <i>subsetting</i> . - Limpeza de dados: <i>missing values</i> , <i>outliers</i> e transformações - Juntar bases de dados	8 Horas
Módulo 3 – Estatística básica em R: - Estatísticas descritivas: medidas de dispersão central e variação. - Distribuições probabilísticas: variáveis discretas e contínuas. - Testes de hipóteses.	8 Horas

MÓDULOS	DURAÇÃO
Módulo 4 – Regressão Linear: - O modelo classico linear. - Estimação de parametros segundo o MMQ. - Testes de hipóteses: significância estatística e ajuste do modelo. - Modelo de regressão múltipla. - Testar as premissas: multicolinearidade, heteroscedasticidade e normalidade dos resíduos. - Critérios de seleção dos modelos.	12 Horas
Módulo 5 – O modelo: - Estrutura do modelo e premissas – Perceber o modelo (4 Hours). - Uso e tratamento dos dados (4 Hours). - Descrição do modelo (4 Hours). - Aplicação do modelo a cada piloto (12 Hours). - Aplicação autónoma do modelo a uma região (8 Hours).	32 Horas

Transformação de dados (I)

Arrumar - *tidy*

Ciência de dados



Vamos a isso

Aceda a este link para começar já

<https://posit.cloud/content/5906356>

Back to babynames

	year	sex	name	n	prop
1	1880	F	Mary	7065	0.0724
2	1880	F	Margar	1578	0.0162
3	1880	F	Martha	1040	0.0107
4	1880	F	Marie	471	0.00483
5	1880	F	Maria	125	0.00128
1	1881	F	Mary	6919	0.0700
2	1881	F	Margar	1658	0.0168
3	1881	F	Martha	1044	0.0106
4	1881	F	Marie	499	0.00505
5	1881	F	Maria	120	0.00121
	1881	M	Gideon	7	0.0001



year	sex	name	n	prop
1880	F	Maria	125	0.00128
1881	F	Maria	120	0.00121
...	...	Maria

filter()

Extrai observações que cumpram todos os critérios fornecidos

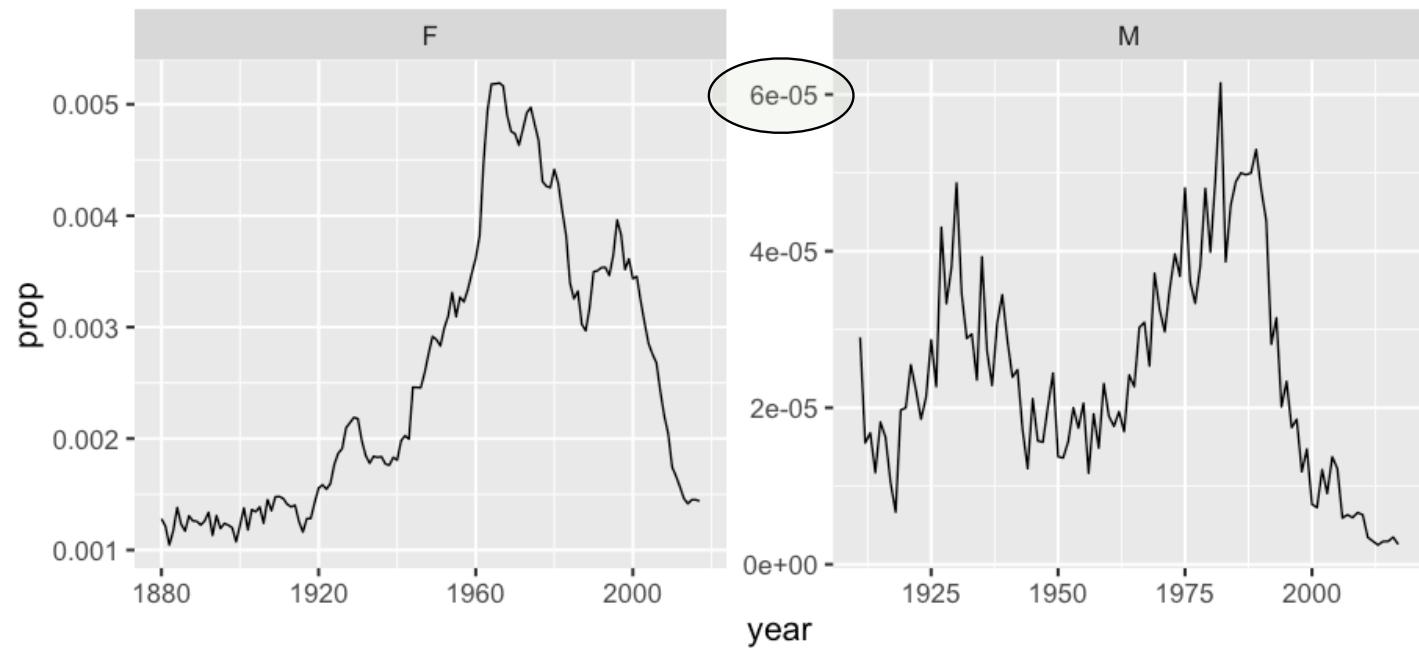
```
filter(babynames, name == "Maria" & year == 1880)
```

babynames				
year	sex	name	n	prop
1880	F	Mary	7065	0.0724
1880	F	Margaret	1578	0.0162
1880	F	Martha	1040	0.0107
1880	F	Marie	471	0.00483
1880	F	Maria	125	0.00128
1881	F	Mary	6919	0.0700

→

year	sex	name	n	prop
1880	F	Maria	125	0.00128

```
babynames %>% filter(name == "Michael") %>%  
  ggplot() +  
  geom_line(mapping = aes(year, prop)) +  
  facet_wrap(~ sex)
```



Quais os nomes
mais populares?

Quiz

Será que conseguimos:

1. calcular o número total de bebés de cada nome?

Transformação de dados (II)

Transformar para derivar nova
informação

Derivar nova informação

summarise() - resumir **variáveis**

group_by() - agrupar por **casos**

mutate() - criar novas **variáveis**

summarise()

e.g. medidas de tendência central (média, mediana) ou dispersão (desvio padrão)
+ na próxima aula!

Calcular tabela de medidas sumárias:

```
babynames %>% summarise(total = sum(n), max = max(n))
```

babynames				
year	sex	name	n	prop
1880	M	John	9655	0.0815
1880	M	William	9532	0.0805
1880	M	James	5927	0.0501
1880	M	Charles	5348	0.0451
1880	M	Garrett	13	0.0001
1881	M	John	8769	0.081

→

total	max
348120517	99686

13



Medidas sumárias em summarise()

Input: vector
Output: valor

Summary Functions

TO USE WITH SUMMARISE ()

summarise() applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.

summary function

COUNTS

dplyr::n() - number of values/rows
dplyr::n_distinct() - # of uniques
sum(is.na()) - # of non-NA's

LOCATION

mean() - mean, also mean(is.na())
median() - median

LOGICALS

mean() - Proportion of TRUE's
sum() - # of TRUE's

POSITION/ORDER

dplyr::first() - first value
dplyr::last() - last value
dplyr::nth() - value in nth location of vector

RANK

quantile() - nth quantile
min() - minimum value
max() - maximum value

SPREAD

IQR() - Inter-Quartile Range
mad() - median absolute deviation
sd() - standard deviation
var() - variance

Vector Functions

TO USE WITH MUTATE ()

mutate() and **transmute()** apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

vectorized function

OFFSETS

dplyr::lag() - Offset elements by 1

dplyr::lead() - Offset elements by -1

CUMULATIVE AGGREGATES

dplyr::cumall() - Cumulative all()

dplyr::cumany() - Cumulative any()

dplyr::cummax() - Cumulative max()

dplyr::cummean() - Cumulative mean()

cummin() - Cumulative min()

dplyr::prod() - Cumulative prod()

dplyr::sum() - Cumulative sum()

RANKINGS

dplyr::cum_dist() - Proportion of all values <=

dplyr::dense_rank() - rank with ties = min, no gaps

dplyr::rank() - rank with ties = min

dplyr::row_number() - row number

dplyr::percent_rank() - rank scaled to [0,1]

dplyr::row_number() - rank with ties = first*

MATH

dplyr::case_when() - multi-case if, else()

dplyr::coalesce() - first non-NA values by element across a set of vectors

dplyr::if_else() - condition-within-if, else()

dplyr::na_if() - replace specific values with NA

dplyr::pmax() - element-wise max()

dplyr::pmin() - element-wise min()

dplyr::recode() - Vectorized switch()

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max() - maximum value

SPREAD

IQR() - Inter-Quartile Range
mad() - median absolute deviation
sd() - standard deviation
var() - variance

Also has_rownames(), remove_rownames()

Combine Tables

COMBINE VARIABLES

Use **bind_cols()** to paste tables beside each other as they are.

Use a "Mutating Join" to join one table to columns from another, matching values with rows that they correspond to. Each join retains a different column of the original tables.

left_join(x, y) - by = NULL, copy = FALSE, suffix = c("x", "y")...)

right_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y")...)

inner_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y")...)

full_join(x, y, by = NULL, copy = FALSE, suffix = c("x", "y")...)

Use **by = c("col1", "col2", ...)** to specify common columns to match on.

Use a named vector, **by = c("col1" = "col2")**, to match on columns that have different names in each table.

Use **suffix** to specify the suffix to give to unmatched columns that have the same name in both tables.

Use **anti_join(x, y, by = c("C" = "D"))** to return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.

COMBINE CASES

bind_rows()

Use **bind_rows(..., id = NULL)** to return rows of the other as a single table. Set id to a column to add a column of the original table names (as pictured).

intersect(x, y, ...)

setdiff(x, y, ...)

union(x, y, ...)

union_all(x, y, ...)

retains duplicates.

EXTRACT ROWS

Use a "Filtering Join" to filter one table against the rows of another.

semi_join(x, y, by = NULL,...)

useful to see what will be joined.

anti_join(x, y, by = NULL,...)

return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.



n()

Número de observações num conjunto de dados

```
babynames %>% summarise(n = n())
```

babynames

year	sex	name	n	prop
1880	M	John	9655	0.0815
1880	M	William	9532	0.0805
1880	M	James	5927	0.0501
1880	M	Charles	5348	0.0451
1880	M	Garrett	13	0.0001
1881	M	John	8769	0.081



n
1924665

15



n_distinct()

Número de casos únicos contidos numa variável (sem repetições)

```
babynames %>% summarise(n = n(), nname = n_distinct(name))
```

babynames				
year	sex	name	n	prop
1880	M	John	9655	0.0815
1880	M	William	9532	0.0805
1880	M	James	5927	0.0501
1880	M	Charles	5348	0.0451
1880	M	Garrett	13	0.0001
1881	M	John	8769	0.081

→

n	nname
1924665	97310

Experimente

Numa única operação, encontre as observações com o nome “**Melo**” e utilize **summarise()**, **sum()** e **min()** para encontrar:

1. O número total de bebés com o nome Melo
2. O primeiro ano (**year**) em que nasceu um bebé com este nome

```
babynames %>%  
  filter(name == "Melo") %>%  
  summarise(total = sum(n), first = min(year))  
# A tibble: 1 × 2  
total first  
<dbl> <dbl>  
1      73    2012
```

Quais são os nomes mais populares?

Nomes mais populares = elevado número de recém-nascidos

1. **Soma** - número de bebés com esse nome, somando as observações de todos os anos disponíveis

```
babynames %>%  
  filter(name == "Melo")
```

A tibble: 6 × 5

	year	sex	name	n	prop
	<dbl>	<chr>	<chr>	<dbl>	<dbl>
1	2012	M	Melo	5	0.00000247
2	2013	M	Melo	11	0.00000545
3	2014	M	Melo	8	0.00000391
4	2015	M	Melo	15	0.00000736
5	2016	M	Melo	16	0.00000793
6	2017	M	Melo	18	0.00000917

Não parece muito popular?

```
babynames %>%  
  filter(name == "Maria", sex == 'F')  
  
# A tibble: 138 × 5  
# ... with 133 more rows  
# ... with 133 more rows  
# i Use `print(n = ...)` to see more rows
```

Como comparar
com outros?

```
babynames %>%
  filter(name == "Melo" & sex == "M") %>%
  summarise(total = sum(n))

total
<dbl>
1     73

babynames %>%
  filter(name == "Maria" & sex == "F") %>%
  summarise(total = sum(n))

total
<dbl>
1 543324
```

group_by()

Agrupa observações em função dos diferentes **casos** de uma ou mais **variável/is**

```
babynames %>%  
  group_by(sex) %>%  
  summarise(total = sum(n))
```

sex	total
F	172371079
M	175749438

group_by()

```
babynames %>%  
  group_by(name) %>%  
  summarise(total = sum(n))
```

Também ordena
pela variável
grupo



sex	total
Aaban	107
Aabha	35
Aabid	10
Aabir	5
Aabullah	22



ungroup()

Remove critérios de group_by

```
babynames %>%  
  group_by(sex) %>%  
  ungroup() %>%  
  summarise(total = sum(n))
```

total
348120517

Experimente

Utilize uma combinação de **group_by()**, **summarise()**, e **arrange()** para ver as 10 combinações mais populares de nome (name) e sexo (**sexo**). Calcule a popularidade como o número *total* de bebés com cada par nome/sexo.

```
babynames %>%  
  group_by(name, sex) %>%  
  summarise(total = sum(n)) %>%  
  arrange(desc(total))
```

A tibble: 107,973 × 3 10+?

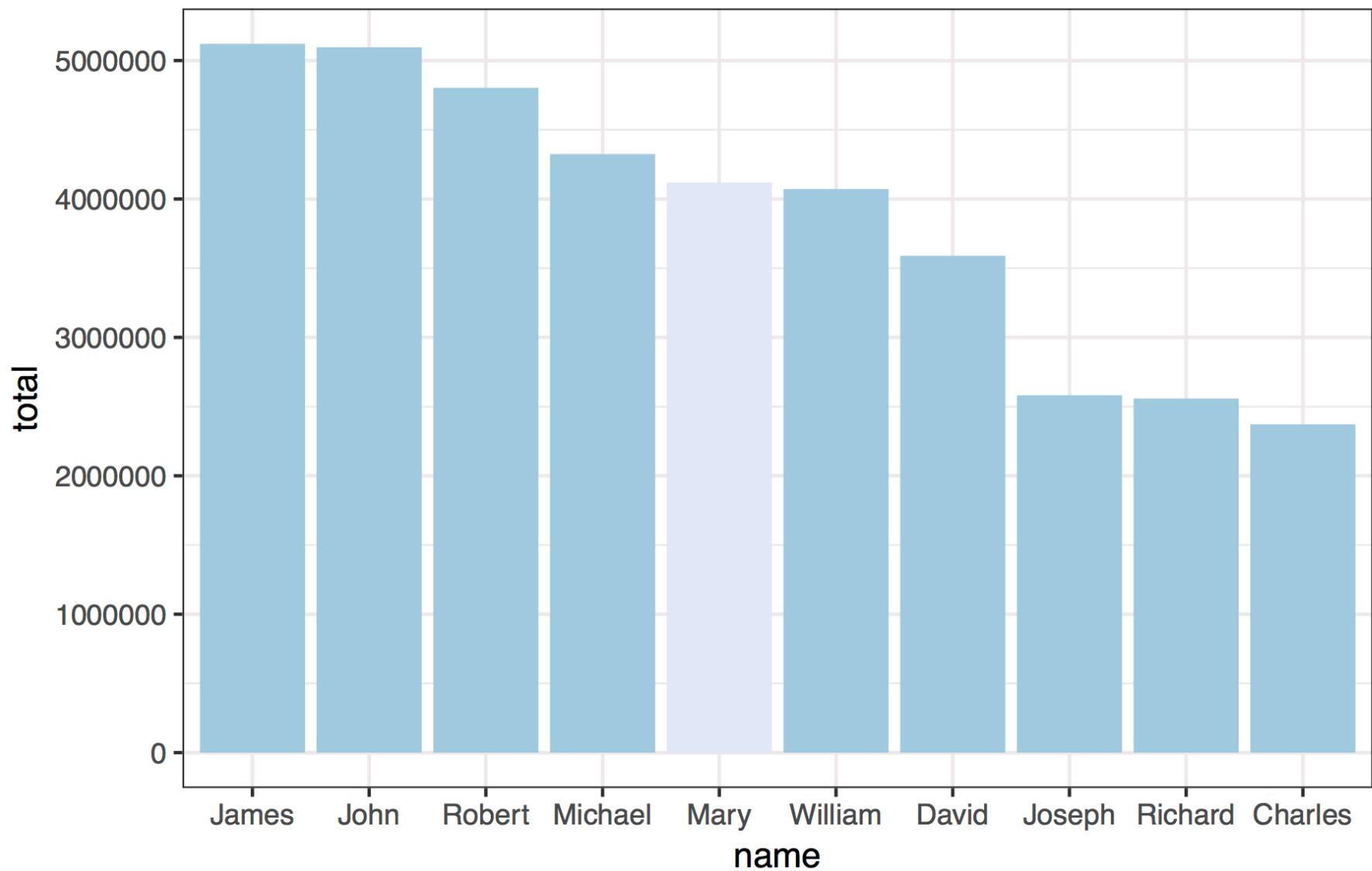
Groups: name [97,310]

	name	sex	total
	<chr>	<chr>	<dbl>
1	James	M	5150472
2	John	M	5115466
3	Robert	M	4814815
4	Michael	M	4350824
5	Mary	F	4123200

... with 107,968 more rows

```
babynames %>%  
  group_by(name, sex) %>%  
  summarise(total = sum(n)) %>%  
  arrange(desc(total)) %>%  
  ungroup %>% slice(1:10)
```

```
# A tibble: 10 × 3  
  name    sex   total  
  <chr>  <chr> <dbl>  
1 James   M     5150472  
2 John    M     5115466  
3 Robert  M     4814815  
4 Michael M     4350824  
5 Mary    F     4123200  
6 William M     4102604  
...  
...
```



```
babynames %>%  
  group_by(name, sex) %>%  
  summarise(total = sum(n)) %>%  
  arrange(desc(total)) %>%  
  ungroup() %>%  
  slice(1:10) %>%  
  ggplot() +  
    geom_col(mapping = aes(x = fct_reorder(name,  
      desc(total)), y = total, fill = sex)) +  
    theme_bw() +  
    scale_fill_brewer() +  
    labs(x = "name")
```

Quais são os nomes mais populares?

Nomes mais populares = elevado número de recém-nascidos

1. **Soma** - número de bebés com esse nome, somando as observações de todos os anos disponíveis

Enviesado? Porquê?

```
babynames %>%  
  filter(year == 2017) %>%  
  summarise(total = sum(n))  
##      total  
## 1 3546301
```

3,546,301

```
babynames %>%  
  filter(year == 1880) %>%  
  summarise(total = sum(n))  
##      total  
## 1 201484
```

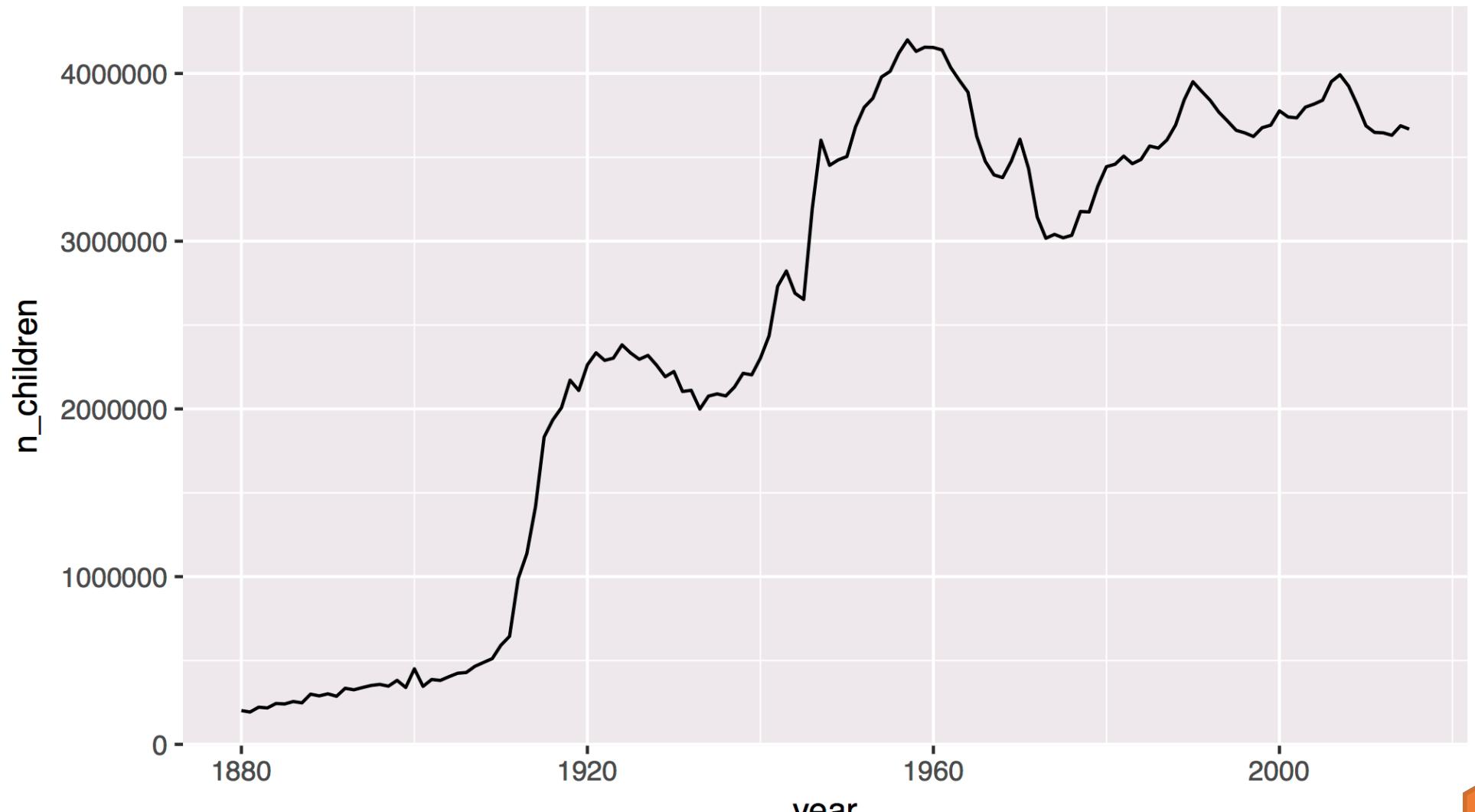
201,484

Experimente

Use group_by() para calcular o número total de bebés registados **em cada ano**.

Mostre os resultados como um gráfico de linhas, uma série temporal

```
babynames %>%  
  group_by(year) %>%  
  summarise(n_children = sum(n)) %>%  
  ggplot() +  
  geom_line(mapping = aes(x = year, y = n_children))
```



35



Qual o nome mais
popular em cada ano?

Quais são os nomes mais populares?

Nomes mais populares = elevado número de recém-nascidos

1. **Soma** - número de bebés com esse nome, somando as observações de todos os anos disponíveis
2. **Ordem** - ranking/posição entre os nomes mais populares *em cada ano*

Será que temos informação suficiente para **ordenar os nomes pela sua popularidade em cada ano?**

mutate()

Comando geral - cria novas variáveis/colunas

```
babynames %>%
```

```
  mutate(percent = round(prop*100, 2))
```

year	sex	name	n	prop
1880	M	John	9655	0.0815
1880	M	William	9532	0.0805
1880	M	James	5927	0.0501
1880	M	Charles	5348	0.0451
1880	M	Garrett	13	0.0001
1881	M	John	8769	0.081



year	sex	name	n	prop	percent
1880	M	John	9655	0.0815	8.15
1880	M	William	9532	0.0805	8.05
1880	M	James	5927	0.0501	5.01
1880	M	Charles	5348	0.0451	4.51
1880	M	Garrett	13	0.0001	0.01
1881	M	John	8769	0.081	8.1



mutate()

```
babynames %>%
```

```
  mutate(percent = round(prop*100, 2), nper = round(percent))
```

babynames

year	sex	name	n	prop
1880	M	John	9655	0.0815
1880	M	William	9532	0.0805
1880	M	James	5927	0.0501
1880	M	Charles	5348	0.0451
1880	M	Garrett	13	0.0001
1881	M	John	8769	0.081



year	sex	name	n	prop	percent	nper
1880	M	John	9655	0.0815	8.15	8
1880	M	William	9532	0.0805	8.05	8
1880	M	James	5927	0.0501	5.01	5
1880	M	Charles	5348	0.0451	4.51	5
1880	M	Garrett	13	0.0001	0.01	0
1881	M	John	8769	0.081	8.1	8

```
babynames %>%  
  mutate(rank = ? )
```

min_rank()

Função dá a posição ordinal de cada valor

```
min_rank(c(50, 100, 100, 1000))  
# [1] 1 2 2 4
```

(em caso de empate, posição seguinte é partilhada)

```
min_rank(desc(c(50, 100, 100, 1000)))  
# [1] 4 2 2 1
```

Experimente

Utilize **mutate()** e **min_rank()** para ordenar as observações pela sua frequência (**n**), da maior para a mais pequena

```
babynames %>%  
  mutate(rank = min_rank(desc(prop)))
```

```
## A tibble: 1,924,665 x 6  
#> #>   year sex   name        n    prop   rank  
#> # 1 1880 F   Mary     7065 0.0724     14  
#> # 2 1880 F   Anna     2604 0.0267     709  
#> # 3 1880 F   Emma     2003 0.0205    1131  
#> # 4 1880 F   Elizabeth 1939 0.0199    1192  
#> # 5 1880 F   Minnie    1746 0.0179    1427  
#> # 6 1880 F   Margaret 1578 0.0162    1683  
## ... with 1,924,659 more rows
```

Experimente

Agrupe os dados por ano e volte a ordená-los por frequência. Filtre os resultados para manter apenas os nomes mais populares em cada ano

```
babynames %>%  
  group_by(year) %>%  
  mutate(rank = min_rank(desc(prop))) %>%  
  filter(rank == 1)  
  
# A tibble: 138 x 6  
# Groups:   year [138]  
  year sex   name     n    prop   rank  
  1 1880 M    John 9655 0.0815     1  
  2 1881 M    John 8769 0.0810     1  
  3 1882 M    John 9557 0.0783     1  
# ... with 135 more rows
```

Funções vetoriais

Input: vector
Output: vector da mesma dimensão (*length*)

Vector Functions

TO USE WITH MUTATE ()

mutate() and **transmute()** apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.

vectorized function ➔

OFFSETS

dplyr::lag() - Offset elements by 1
dplyr::lead() - Offset elements by -1

CUMULATIVE AGGREGATES

dplyr::cumall() - Cumulative all()
dplyr::cumany() - Cumulative any()
 cummax() - Cumulative max()
dplyr::cummean() - Cumulative mean()
 cummin() - Cumulative min()
 cumprod() - Cumulative prod()
 cumsum() - Cumulative sum()

RANKINGS

dplyr::cume_dist() - Proportion of all values <= x
dplyr::dense_rank() - rank with ties = min, no gaps
dplyr::min_rank() - rank with ties = min
dplyr::ntile() - bins into n bins
dplyr::percent_rank() - min_rank scaled to [0,1]
dplyr::row_number() - rank with ties = "first"

MATH

+, -, *, /, ^, %%, %% - arithmetic ops
log(), log2(), log10() - logs
<, <=, >, >=, !=, == - logical comparisons
dplyr::between() - x >= left & x <= right
dplyr::near() - safe == for floating point numbers

MISC

dplyr::case_when() - multi-case if_else()
dplyr::coalesce() - first non-NA values by element across a set of vectors
dplyr::if_else() - element-wise if() + else()
dplyr::na_if() - replace specific values with NA
 pmax() - element-wise max()
 pmin() - element-wise min()
dplyr::recode() - Vectorized switch()
dplyr::recode_factor() - Vectorized switch() for factors

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SPREAD

IQR() - Inter-Quartile Range

mad() - median absolute deviation

sdl() - standard deviation

var() - variance

Row Names

Tidy data does not use rownames, which store a vector outside of the columns. To work with the rownames, first move them into a column.

rownames_to_column()

Move row names into col.

o <- rownames_to_column(iris, var = "C")

column_to_rownames()

Move col in row names.

o <- column_to_rownames(a, var = "C")

Also has_rownames(), remove_rownames()

Summary Functions

TO USE WITH SUMMARISE ()

summarise() applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.

summary function ➔

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 sum(is.na()) - # of non-NAs

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PPOSITION/ORDER

dplyr::first() - first value
dplyr::last() - last value

dplyr::n() - value in nth location of vector

RANK

quantile() - nth quantile

min() - minimum value

max() - maximum value

SPREAD

IQR() - Inter-Quartile Range

mad() - median absolute deviation

sdl() - standard deviation

var() - variance

Combine Tables

COMBINE VARIABLES

bind_cols() - paste tables beside each other as they are.

BE SURE THAT ROWS ALIGN.

bind_rows() - paste tables placed side by side as a single table

Use a "Mutating Join" to join one table to columns from another, matching values with rows that they correspond to. Each join retains a different combination of values from the tables.

left_join(x, y) = NULL,
copy = FALSE, suffix = c("x", "y")...)

Join matching values from x to y.

right_join(x, y) = NULL,
copy = FALSE, suffix = c("x", "y")...)

Join matching values from y to x.

inner_join(x, y) = NULL,
copy = FALSE, suffix = c("x", "y")...)

Join data. Retain only rows with matches.

full_join(x, y, by = NULL,
copy = FALSE, suffix = c("x", "y")...)

Join all data, all rows.

use_by(c("col1", "col2", ...))

To join on specific common columns to match on.

left_join(x, y, by = c("C" = "D"))

use_by to specify the suffix to give to unmatched columns that have the same name in both tables.

left_join(x, y, by = c("C" = "D", "suffix = c("1", "2")))

Return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.

semi_join(x, y = NULL...)

Return rows of x that have a match in y.

anti_join(x, y = NULL...)

Return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.

Use setequal() to test whether two data sets contain the exact same rows (in any order).

EXTRACT ROWS

x + **y** = **z**

Use a "Filtering Join" to filter one table against the rows of another.

semi_join(x, y = NULL...)

Return rows of x that have a match in y.

use_by(c("C" = "D"))

use_by to see what will be joined.

anti_join(x, y = NULL...)

Return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.

Use setdiff() to test whether two data sets contain different rows (in any order).

ROW NAMES

Tidy data does not use rownames, which store a vector outside of the columns. To work with the rownames, first move them into a column.

rownames_to_column()

Move row names into col.

o <- rownames_to_column(iris, var = "C")

column_to_rownames()

Move col in row names.

o <- column_to_rownames(a, var = "C")

Also has_rownames(), remove_rownames()

RStudio

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on back



Resumo: verbos dplyr

(Verbos para uma tabela única)



Selecionar variáveis: **select()**



Extrair casos específicos: **filter()**



Reordenar tabela: **arrange()**



Obter medidas sumárias: **summarise()**



Criar novas variáveis: **mutate()**

Transformação de dados (III)

Reorientar dados (*pivot*)

sales.xlsx

```
clientes <- readxl::read_xlsx("sales_data/sales.xlsx", sheet = 'clientes')
clientes
```

```
## # A tibble: 2 × 4
##   id_cliente item_1    item_2      item_3
##       <dbl> <chr>     <chr>      <chr>
## 1           1 pao       leite      banana
## 2           2 leite     papel higienico <NA>
```

```
precos <- readxl::read_xlsx("sales_data/sales.xlsx", sheet = 'precos')
precos
```

```
## # A tibble: 5 × 2
##   item          price
##   <chr>        <dbl>
## 1 abacate        2
## 2 banana        0.5
## 3 pao            1.5
## 4 leite           1
## 5 papel higienico  3
```

clientes

Temos...

```
## # A tibble: 2 × 4
##   id_cliente item_1 item_2
##       <dbl> <chr>  <chr>
## 1          1 pao    leite
## 2          2 leite  papel higienico
```

Queremos...

```
## # A tibble: 6 × 3
##   id_cliente item_no item
##       <dbl> <chr>  <chr>
## 1          1      1 item_1
## 2          1      1 item_2
## 3          1      1 item_3
## 4          2      2 item_1
## 5          2      2 item_2
## 6          2      2 item_3
```

O objetivo

		wide		
		x	y	z
id	1	a	c	e
	2	b	d	f

Wide vs. long

wide

mais colunas / variáveis

```
## # A tibble: 2 × 4
##   id_cliente item_1 item_2      item_3
##       <dbl> <chr>  <chr>      <chr>
## 1           1 pao    leite      banana
## 2           2 leite  papel higienico <NA>
```

long

mais linhas / observações

```
## # A tibble: 6 × 3
##   id_cliente item_no item
##       <dbl> <chr>  <chr>
## 1           1 item_1  pao
## 2           1 item_2  leite
## 3           1 item_3  banana
## 4           2 item_1  leite
## 5           2 item_2  papel higienico
## 6           2 item_3  <NA>
```

pivot_longer()

- data

```
pivot_longer(  
  data,  
  cols,  
  names_to = "name",  
  values_to = "value"  
)
```

pivot_longer()

- data
- cols: colunas a transpor para formato long

```
pivot_longer(  
  data,  
  cols,  
  names_to = "name",  
  values_to = "value"  
)
```

pivot_longer()

- data
- cols: colunas a transpor para formato long
- names_to: nome da variável que vai receber os nomes das colunas a transpor para long, como valores

```
pivot_longer(  
  data,  
  cols,  
  names_to = "name",  
  values_to = "value"  
)
```

pivot_longer()

- data
- cols: colunas a transpor para formato long
- names_to: nome da variável que vai receber os nomes das colunas a transpor para long, como valores
- values_to: nome da variável que vai receber os valores atualmente dispersos por várias colunas (string)

```
pivot_longer(  
  data,  
  cols,  
  names_to = "name",  
  values_to = "value"  
)
```

clientes → compras

```
compras <- clientes %>%
  pivot_longer(
    cols = item_1:item_3, # variables item_1 to item_3
    names_to = "item_no", # column names -> new column called item_no
    values_to = "item"    # values in columns -> new column called item
  )
```

compras

```
## # A tibble: 6 × 3
##   id_cliente item_no item
##       <dbl> <chr>   <chr>
## 1           1 item_1  pao
## 2           1 item_2  leite
## 3           1 item_3  banana
## 4           2 item_1  leite
## 5           2 item_2  papel higienico
## 6           2 item_3  <NA>
```

Exemplo da importância de dados tidy

Várias operações de transformação requerem-no (e.g. *join* - prox. aula)

precos

```
## # A tibble: 5 × 2
##   item          price
##   <chr>        <dbl>
## 1 abacate       2
## 2 banana      0.5
## 3 pao          1.5
## 4 leite         1
## 5 papel higienico  3
```

compras %>%

```
left_join(precos)
```

```
## # A tibble: 6 × 4
##   id_cliente item_no item          price
##   <dbl>     <chr>    <chr>        <dbl>
## 1 1           1 item_1    pao            1.5
## 2 2           1 item_2    leite           1
## 3 3           1 item_3    banana          0.5
## 4 4           2 item_1    leite           1
## 5 5           2 item_2    papel higienico  3
## 6 6           2 item_3    <NA>            NA
```

compras → clientes

- data
- names_from: variável em formato long a transpor para nomes de novas colunas
- values_from: variável em formato long que contém valores a dispersar por várias colunas no formato wide

```
compras %>%
  pivot_wider(
    names_from = item_no,
    values_from = item
  )

## # A tibble: 2 × 4
##   id_cliente item_1 item_2      item_3
##       <dbl> <chr>  <chr>     <chr>
## 1           1 pao    leite     banana
## 2           2 leite  papel higienico <NA>
```

Transformação de dados (IV)

Juntar bases de dados

band & instrument

```
band <- tribble(  
  ~name,      ~band,  
  "Mick",    "Stones",  
  "John",    "Beatles",  
  "Paul",    "Beatles"  
)
```

band

name	band
Mick	Stones
John	Beatles
Paul	Beatles

```
instrument <- tribble(  
  ~name,    ~plays,  
  "John",   "guitar",  
  "Paul",   "bass",  
  "Keith",  "guitar"  
)
```

instrument

name	plays
John	guitar
Paul	bass
Keith	guitar

join - sintaxe

Cada função **join** tem como input duas tabelas,
e output uma única

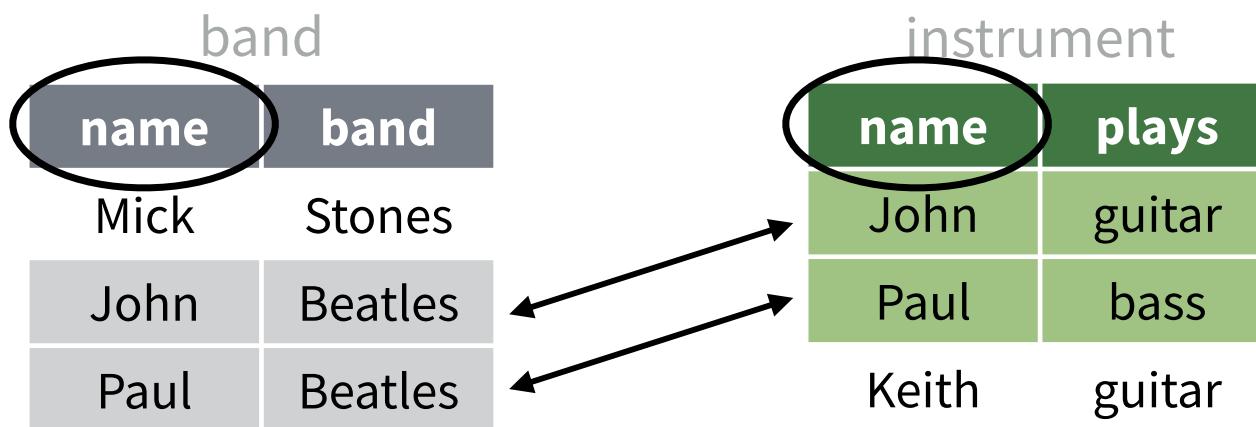
```
left_join(x, y, by = NULL, ... )
```

função

data frames a
juntar (merge)

nomes das variáveis
para mapear as
duas tabelas

Juntar por nome





`x %>% f(y)`
becomes `f(x, y)`

Operador pipe %>%

%>%



instrument %>% `left_join(_____, by = "name")`

Passa resultado da esquerda à função da direita, para o seu primeiro argumento. Experimente:

```
left_join(instrument, by = "name")
instrument %>% left_join(by = "name")
```

left

```
band %>% left_join(instrument, by = "name")
```

band

name	band
Mick	Stones
John	Beatles
Paul	Beatles

+

instrument

name	plays
John	guitar
Paul	bass
Keith	guitar

=

name	band	plays
Mick	Stones	<NA>
John	Beatles	guitar
Paul	Beatles	bass

right

```
band %>% right_join(instrument, by = "name")
```

band

name	band
Mick	Stones
John	Beatles
Paul	Beatles

+

instrument

name	plays
John	guitar
Paul	bass
Keith	guitar

=

name	band	plays
John	Beatles	guitar
Paul	Beatles	bass
Keith	<NA>	guitar

full

```
band %>% full_join(instrument, by = "name")
```

band		instrument		=		
name	band	name	plays	name	band	plays
Mick	Stones	John	guitar	Mick	Stones	<NA>
John	Beatles	Paul	bass	John	Beatles	guitar
Paul	Beatles	Keith	guitar	Paul	Beatles	bass
				Keith	<NA>	guitar

inner

```
band %>% inner_join(instrument, by = "name")
```

band

name	band
Mick	Stones
John	Beatles
Paul	Beatles

+

instrument

name	plays
John	guitar
Paul	bass
Keith	guitar

=

name	band	plays
John	Beatles	guitar
Paul	Beatles	bass

Verbos join



left_join() mantém todos os casos da tabela da **esquerda**



right_join() mantém todos os casos da tabela da **direita**



full_join() mantém todos os casos de **qualquer** tabela



inner_join() mantém apenas os casos em comum de **ambas**



semi_join() extrai os cases em que **há correspondência**



anti_join() extrai os cases em que **não há correspondência**



Mini-teste I

Hip.: os bebés de 1969 com o nome de um destes ícones da música tocarão o respetivo instrumento. (Os outros não tocarão nenhum.)

1. Use um **join** para combinar as tabelas **babynames** e **instruments**.
2. Calcule a percentagem de recém-nascidos nesse ano que tocará algum instrumento.

```
babynames %>% filter(year == 1969) %>%  
  left_join(instrument, by = 'name') %>%  
  group_by(sex, Toca = !is.na(instrument)) %>%  
  summarise(players_pc = sum(prop) * 100)
```

```
# A tibble: 4 × 3  
# Groups:   sex [2]  
  sex   Toca   players_pc  
  <chr> <lgl>     <dbl>  
1 F     FALSE     95.7  
2 F     TRUE      0.0447  
3 M     FALSE     89.2  
4 M     TRUE      8.61
```



```
babynames %>% filter(year == 1969) %>%  
  left_join(instrument, by = 'name') %>%  
  group_by(instrument) %>%  
  summarise(players_no = sum(n)) %>%  
  mutate(players_pc = 100 * players_no / sum(players_no))  
  
# A tibble: 3 × 3  
instrument players_no percent_players  
<chr>          <dbl>            <dbl>  
1 bass           25755            0.741  
2 guitar         132611           3.81  
3 NA             3317849          95.4
```



Mini-teste II

Quanto se gastou em cada categoria de bens?

1. Utilize **read_excel** (do pacote **readxl**) e importe o ficheiro Excel **sales2.xlsx** (ambas as folhas).
2. Reoriente a tabela de **clientes** para formato **long**, de forma a ficar com 4 variáveis: **id_cliente**, **item**, **categoria** e **num_item**.
 - *Dica: vai precisar da opção **names_sep***
3. Junte as tabelas de **preços** e **clientes** numa única.
4. Calcule o montante gasto por categoria. Cada cliente comprou apenas 1 unidade de cada item.

Soluções

join

Data Transformation cheatsheet

on back

Vector Functions

TO USE WITH MUTATE()

- `mutate()` & `transmute()` apply vectorized functions to columns to create new columns. Vectorized functions take vectors as input and return vectors of the same length as output.
- vectorized function** →

OFFSETS

- `dplyr::lag()` - Offset elements by 1
- `dplyr::lead()` - Offset elements by > 1

CUMULATIVE AGGREGATES

- `dplyr::cumall()` - Cumulative all()
- `dplyr::cumany()` - Cumulative any()
- `dplyr::cummax()` - Cumulative max()
- `dplyr::cummean()` - Cumulative mean()
- `dplyr::cumprod()` - Cumulative prod()
- `dplyr::cumsum()` - Cumulative sum()

RANKINGS

- `dplyr::cume_dist()` - Proportion of all values =
- `dplyr::dense_rank()` - rank with ties = min, no gaps
- `dplyr::min_rank()` - rank with ties = min
- `dplyr::ntile()` - bins into n bins
- `dplyr::percent_rank()` - min_rank scaled to [0,1]
- `dplyr::row_number()` - rank with ties = "first"

MATH

- `*, *, /, ^, %%, %%` - arithmetic ops
- `log(), log2(), log10()` - logarithms
- `<, <=, >, >=, !=, ==` - comparisons
- `dplyr::left()` - x <= left & x <= right
- `dplyr::near()` - safe == for floating point numbers

MISC

- `dplyr::case_when()` - multi-case if, else()
- `dplyr::coalesce()` - first non-NA value by element across a set of vectors
- `dplyr::if_else()` - if, else()
- `dplyr::na_if()` - replace specific values with NA
- `dplyr::max()` - element-wise max()
- `dplyr::min()` - element-wise min()
- `dplyr::model()` - Vectorized switch()
- `dplyr::recode()` - Vectorized switch()
- `dplyr::recode_factor()` - Vectorized switch() for factors

Summary Functions

TO USE WITH SUMMARISE()

- `summarise()` applies summary functions to columns to create a new table. Summary functions take vectors as input and return single values as output.
- summary function** →

COUNTS

- `dplyr::n()` - number of values/rows
- `dplyr::n_distinct()` - # of uniqueness
- `sum(is.na())` - # of non-NA's

LOCATION

- `mean()` - mean, also `mean(is.na())`
- `median()` - median
- `sum()` - # of TRUE's

LOGICALS

- `mean()` - Proportion of TRUE's
- `sum()` - # of TRUE's

POSITION/ORDER

- `dplyr::first()` - first value
- `dplyr::last()` - last value
- `dplyr::nth()` - value in nth location of vector

RANK

- `quantile()` - nth quantile
- `min()` - minimum value
- `max()` - maximum value

SPREAD

- `IQR()` - Inter-Quartile Range
- `mad()` - median absolute deviation
- `sdl()` - standard deviation
- `var()` - variance

Row Names

Tidy data does not use rownames, which store a variable outside of the columns. To work with the rownames, first move them into a column:

- `rownames_to_column()` - Move row names into col.
- `as_rownames_(iris, var = "C")` - column_to_rownames(iris, var = "C")

Also has `rownames()`, `remove_rownames()`

Combine Tables

COMBINE VARIABLES

X	y	=
A B C		A B C A B D
a t 1		a t 1 a t 3
b u 2		b u 2 b u 2
c v 3		c v 3 d w 1
d w 1		

Use `bind_cols()` to paste tables beside each other as they are.

bind_cols(...) Returns tables placed side by side as a single table.
BE SURE THAT ROWS ALIGN.

Use a "Mutating Join" to join one table to columns from another, matching values with the rows that they correspond to. Each join retains a different combination of values from the tables.

left_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"), ...)
Join matching values from y to x.

right_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"), ...)
Join matching values from x to y.

inner_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"), ...)
Join data. Retain only rows with matches.

full_join(x, y, by = NULL, copy = FALSE, suffix=c("x","y"), ...)
Join data. Retain all values, all rows.

extract_rows

Use `by = c("col1", "col2", ...)` to specify one or more common columns to match on.

filtering join

Use a named vector, `by = c("col1" = "col2")`, to match on columns that have the same name in both tables. `left_join(x, y, by = c("C" = "D"))`

filtering join

Use a named vector, `by = c("col1" = "col2")`, to match on columns that have different names in each table. `left_join(x, y, by = c("C" = "D"))`

suffix

Use **suffix** to specify the suffix to give to unmatched columns that have the same name in both tables. `left_join(x, y, by = c("C" = "D"), suffix = c("1", "2"))`

COMBINE CASES

X	y	=
A B C		A B C
a t 1		a t 1
b u 2		b u 2
c v 3		c v 3
d w 4		d w 4

Use `bind_rows()` to paste tables below each other as they are.

bind_rows(..., .id = NULL)
Returns tables one on top of the other as a single table. Set `.id` to a column name to add a column of the original table names (as pictured)

intersect(x, y, ...)
Rows that appear in both x and y.

setdiff(x, y, ...)
Rows that appear in x but not y.

union(x, y, ...)
Rows that appear in x or y.
(Duplicates removed). `union_all()` retains duplicates.

Use `setequal()` to test whether two data sets contain the exact same rows (in any order).

EXTRACT ROWS

X	y	=
A B C		A B D
a t 1		a t 3
b u 2		b u 2
c v 3		d w 1

Use a "Filtering Join" to filter one table against the rows of another.

semi_join(x, y, by = NULL, ...)
Return rows of x that have a match in y.

anti_join(x, y, by = NULL, ...)
Return rows of x that do not have a match in y. USEFUL TO SEE WHAT WILL NOT BE JOINED.



Quiz 2

Join at menti.com use code 72 61 65 7



Instructions

Go to

www.menti.com

Enter the code

72 61 65 7



Or use QR code

**Obrigado
e até amanhã!**

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