

Manipulação de Dados

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Atualização: 2022-05-18

Vamos usar a library `palmerpenguins`

```
penguins %>%  
  glimpse
```

Rows: 344

Columns: 8

```
$ species      <fct> Adelie, Adelie, Adelie, Adelie, Adelie, Adelie, Adel...  
$ island       <fct> Torgersen, Torgersen, Torgersen, Torgersen, Torgerse...  
$ bill_length_mm <dbl> 39.1, 39.5, 40.3, NA, 36.7, 39.3, 38.9, 39.2, 34.1, ...  
$ bill_depth_mm <dbl> 18.7, 17.4, 18.0, NA, 19.3, 20.6, 17.8, 19.6, 18.1, ...  
$ flipper_length_mm <int> 181, 186, 195, NA, 193, 190, 181, 195, 193, 190, 186...  
$ body_mass_g  <int> 3750, 3800, 3250, NA, 3450, 3650, 3625, 4675, 3475, ...  
$ sex          <fct> male, female, female, NA, female, male, female, male...  
$ year         <int> 2007, 2007, 2007, 2007, 2007, 2007, 2007, 2007, 2007...
```

Manipulações de dados elementares

- Selecionar linhas: `filter()`
- Selecionar colunas: `select()`
- Ordenar linhas: `arrange()`
- Contar coisas: `count()`
- Criar novas colunas: `mutate()`
- Analisar segmentos: `group_by()` and `summarize()`

Mas, primeiro: o operador pipe %>%

%>% é pronunciado "e então"

O pipe %>% alimenta os dados em funções

```
head(penguins)
```

```
# A tibble: 6 × 8
  species island bill_length_mm bill_depth_mm flipper_length_... body_mass_g sex
  <fct>   <fct>         <dbl>         <dbl>         <int>         <int> <fct>
1 Adelie  Torge...         39.1          18.7          181          3750 male
2 Adelie  Torge...         39.5          17.4          186          3800 fema...
3 Adelie  Torge...         40.3           18          195          3250 fema...
4 Adelie  Torge...          NA           NA           NA           NA <NA>
5 Adelie  Torge...         36.7          19.3          193          3450 fema...
6 Adelie  Torge...         39.3          20.6          190          3650 male
# ... with 1 more variable: year <int>
```

O pipe %>% alimenta os dados em funções

```
# head(penguins)
```

```
penguins %>%  
  head()
```

```
# A tibble: 6 × 8
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_...	body_mass_g	sex
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>	<fct>
1	Adelie	Torge...	39.1	18.7	181	3750	male
2	Adelie	Torge...	39.5	17.4	186	3800	fema...
3	Adelie	Torge...	40.3	18	195	3250	fema...
4	Adelie	Torge...	NA	NA	NA	NA	<NA>
5	Adelie	Torge...	36.7	19.3	193	3450	fema...
6	Adelie	Torge...	39.3	20.6	190	3650	male

```
# ... with 1 more variable: year <int>
```

O pipe `%>%` alimenta os dados em funções

```
ggplot(penguins, aes(bill_length_mm, bill_depth_mm, color = species)) + geom_point()
```


O pipe `%>%` alimenta os dados em funções

```
# ggplot(penguins, aes(bill_length_mm, bill_depth_mm, color = species)) + geom_point()
```

```
penguins %>%  
  ggplot(aes(bill_length_mm, bill_depth_mm, color = species)) + geom_point()
```

Desde R 4.1: O pipe nativo: |>

```
penguins |>  
  ggplot(aes(bill_length_mm, bill_depth_mm, color = species)) + geom_point()
```

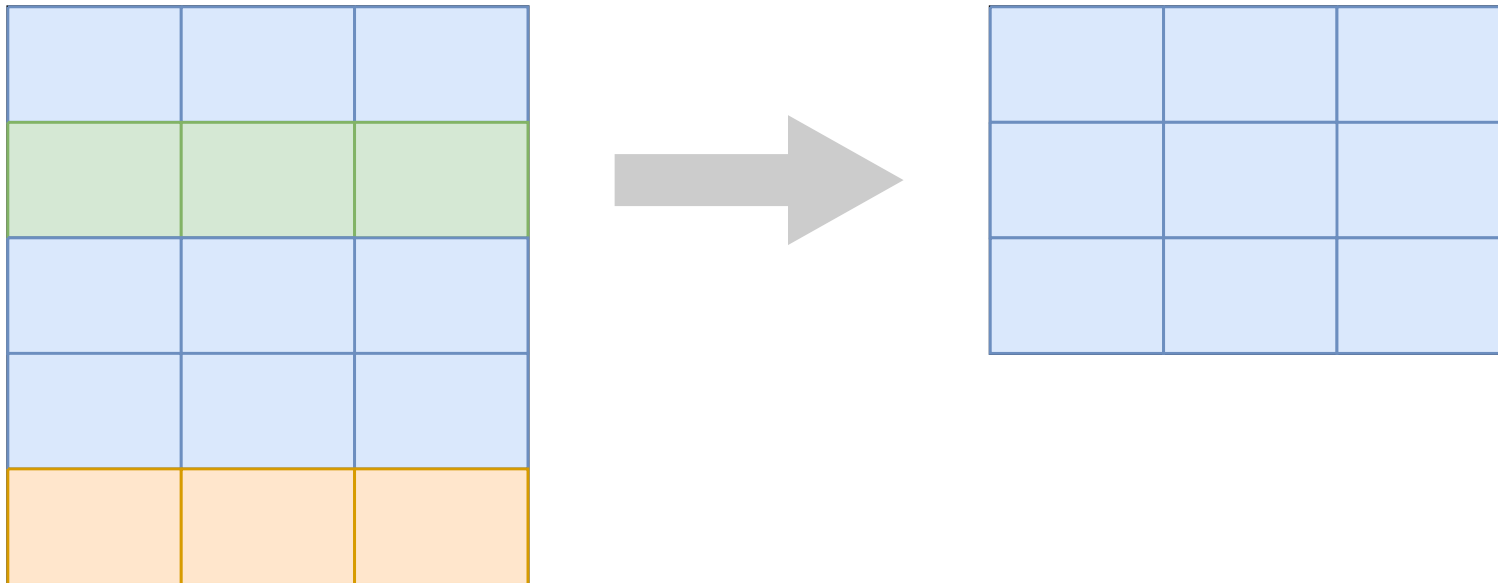
Qual usar? O pipe nativo ou o pipe tradicional?

- `|>` é o futuro. Se puder, use-o.
- `%>%` trabalha em versões mais antigas. É mais seguro por agora

Nós usamos `%>%` porque muitas pessoas ainda usam o R 3.x ou 4.0.

Selecione linhas ou colunas, e ordene

Selecione linhas de uma tabela: `filter()`



Selezione pinguins da espécie Gentoo

```
penguins %>%  
  filter(species == "Gentoo")
```

```
# A tibble: 124 × 8
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>
1	Gentoo	Biscoe	46.1	13.2	211	4500
2	Gentoo	Biscoe	50	16.3	230	5700
3	Gentoo	Biscoe	48.7	14.1	210	4450
4	Gentoo	Biscoe	50	15.2	218	5700
5	Gentoo	Biscoe	47.6	14.5	215	5400
6	Gentoo	Biscoe	46.5	13.5	210	4550
7	Gentoo	Biscoe	45.4	14.6	211	4800
8	Gentoo	Biscoe	46.7	15.3	219	5200
9	Gentoo	Biscoe	43.3	13.4	209	4400
10	Gentoo	Biscoe	46.8	15.4	215	5150

```
# ... with 114 more rows, and 2 more variables: sex <fct>, year <int>
```

Filtre pinguins com comprimento de bico > 50 mm

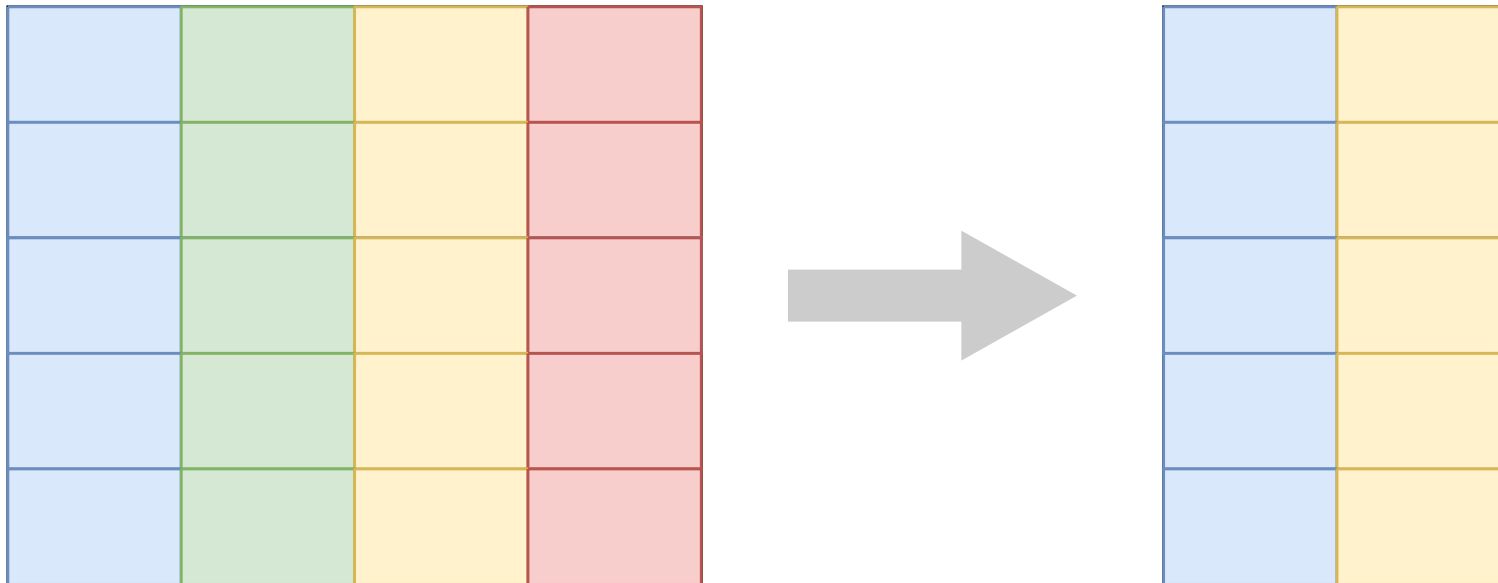
```
penguins %>%  
  filter(bill_length_mm > 50)
```

```
# A tibble: 52 × 8
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>
1	Gentoo	Biscoe	50.2	14.3	218	5700
2	Gentoo	Biscoe	59.6	17	230	6050
3	Gentoo	Biscoe	50.5	15.9	222	5550
4	Gentoo	Biscoe	50.5	15.9	225	5400
5	Gentoo	Biscoe	50.1	15	225	5000
6	Gentoo	Biscoe	50.4	15.3	224	5550
7	Gentoo	Biscoe	54.3	15.7	231	5650
8	Gentoo	Biscoe	50.7	15	223	5550
9	Gentoo	Biscoe	51.1	16.3	220	6000
10	Gentoo	Biscoe	52.5	15.6	221	5450

```
# ... with 42 more rows, and 2 more variables: sex <fct>, year <int>
```

Selecione colunas de uma tabela: `select()`

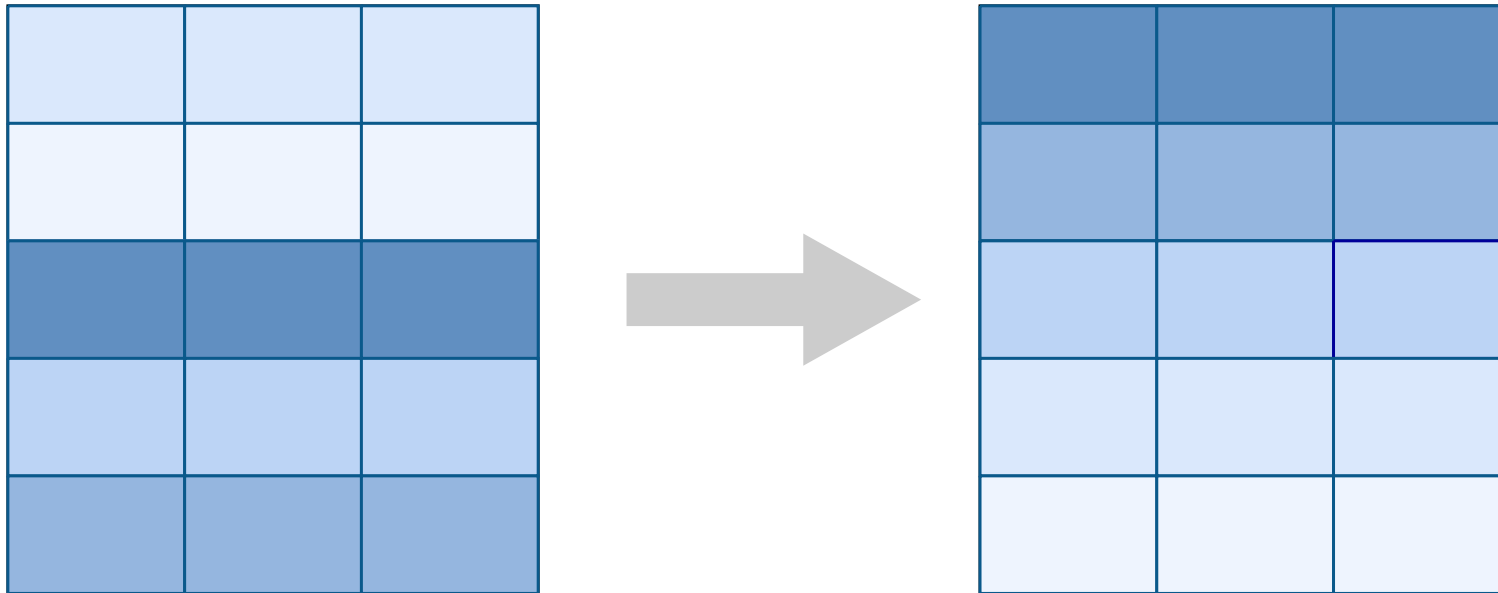


Selecione as colunas `species`, `island`, and `sex`

```
penguins %>%  
  select(species, island, sex)
```

```
# A tibble: 344 × 3  
  species island    sex  
  <fct>   <fct>   <fct>  
1 Adelie  Torgersen male  
2 Adelie  Torgersen female  
3 Adelie  Torgersen female  
4 Adelie  Torgersen <NA>  
5 Adelie  Torgersen female  
6 Adelie  Torgersen male  
7 Adelie  Torgersen female  
8 Adelie  Torgersen male  
9 Adelie  Torgersen <NA>  
10 Adelie Torgersen <NA>  
# ... with 334 more rows
```

Ordene as linhas em uma tabela: `arrange()`



Ordene os pinguins pelo comprimento do bico, crescente

```
penguins %>%  
  arrange(bill_length_mm)
```

```
# A tibble: 344 × 8
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>
1	Adelie	Dream	32.1	15.5	188	3050
2	Adelie	Dream	33.1	16.1	178	2900
3	Adelie	Torgersen	33.5	19	190	3600
4	Adelie	Dream	34	17.1	185	3400
5	Adelie	Torgersen	34.1	18.1	193	3475
6	Adelie	Torgersen	34.4	18.4	184	3325
7	Adelie	Biscoe	34.5	18.1	187	2900
8	Adelie	Torgersen	34.6	21.1	198	4400
9	Adelie	Torgersen	34.6	17.2	189	3200
10	Adelie	Biscoe	35	17.9	190	3450

```
# ... with 334 more rows, and 2 more variables: sex <fct>, year <int>
```

Classifique os pinguins pelo comprimento do bico, decrescente

```
penguins %>%  
  arrange(desc(bill_length_mm))
```

```
# A tibble: 344 × 8
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>
1	Gentoo	Biscoe	59.6	17	230	6050
2	Chinstrap	Dream	58	17.8	181	3700
3	Gentoo	Biscoe	55.9	17	228	5600
4	Chinstrap	Dream	55.8	19.8	207	4000
5	Gentoo	Biscoe	55.1	16	230	5850
6	Gentoo	Biscoe	54.3	15.7	231	5650
7	Chinstrap	Dream	54.2	20.8	201	4300
8	Chinstrap	Dream	53.5	19.9	205	4500
9	Gentoo	Biscoe	53.4	15.8	219	5500
10	Chinstrap	Dream	52.8	20	205	4550

```
# ... with 334 more rows, and 2 more variables: sex <fct>, year <int>
```

Contando coisas

Contando coisas

```
penguins
```

```
# A tibble: 344 × 8
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>
1	Adelie	Torgersen	39.1	18.7	181	3750
2	Adelie	Torgersen	39.5	17.4	186	3800
3	Adelie	Torgersen	40.3	18	195	3250
4	Adelie	Torgersen	NA	NA	NA	NA
5	Adelie	Torgersen	36.7	19.3	193	3450
6	Adelie	Torgersen	39.3	20.6	190	3650
7	Adelie	Torgersen	38.9	17.8	181	3625
8	Adelie	Torgersen	39.2	19.6	195	4675
9	Adelie	Torgersen	34.1	18.1	193	3475
10	Adelie	Torgersen	42	20.2	190	4250

```
# ... with 334 more rows, and 2 more variables: sex <fct>, year <int>
```

Contando coisas

```
penguins %>%  
  count(species)
```

```
# A tibble: 3 × 2  
  species      n  
  <fct>    <int>  
1 Adelie    152  
2 Chinstrap  68  
3 Gentoo   124
```

Contando coisas

```
penguins %>%  
  count(species, island)
```

```
# A tibble: 5 × 3  
  species island      n  
  <fct>   <fct>   <int>  
1 Adelie  Biscoe     44  
2 Adelie  Dream      56  
3 Adelie  Torgersen  52  
4 Chinstrap Dream     68  
5 Gentoo  Biscoe    124
```


Use o pipe para construir pipelines (tubulações) de dados

```
penguins %>%  
  filter(species == "Adelie")
```

```
# A tibble: 152 × 8
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>
1	Adelie	Torgersen	39.1	18.7	181	3750
2	Adelie	Torgersen	39.5	17.4	186	3800
3	Adelie	Torgersen	40.3	18	195	3250
4	Adelie	Torgersen	NA	NA	NA	NA
5	Adelie	Torgersen	36.7	19.3	193	3450
6	Adelie	Torgersen	39.3	20.6	190	3650
7	Adelie	Torgersen	38.9	17.8	181	3625
8	Adelie	Torgersen	39.2	19.6	195	4675
9	Adelie	Torgersen	34.1	18.1	193	3475
10	Adelie	Torgersen	42	20.2	190	4250

```
# ... with 142 more rows, and 2 more variables: sex <fct>, year <int>
```

Use o pipe para construir pipelines (tubulações) de dados

```
penguins %>%  
  filter(species == "Adelie") %>%  
  select(island, sex)
```

```
# A tibble: 152 × 2  
  island    sex  
  <fct>    <fct>  
1 Torgersen male  
2 Torgersen female  
3 Torgersen female  
4 Torgersen <NA>  
5 Torgersen female  
6 Torgersen male  
7 Torgersen female  
8 Torgersen male  
9 Torgersen <NA>  
10 Torgersen <NA>  
# ... with 142 more rows
```

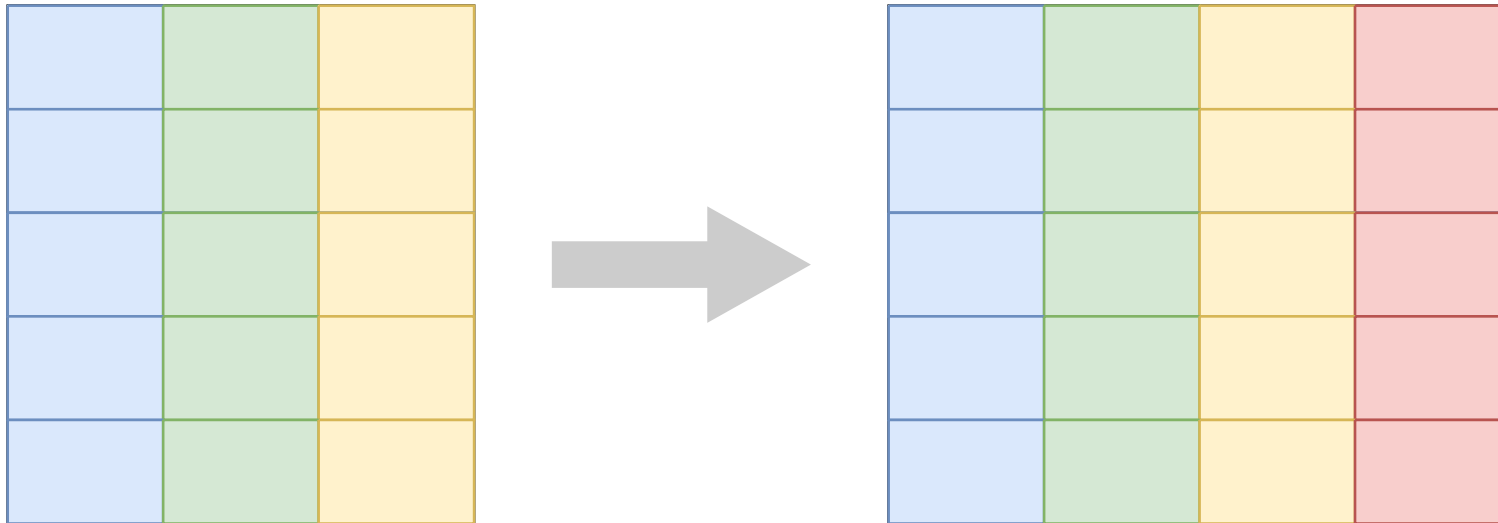
Use o pipe para construir pipelines (tubulações) de dados

```
penguins %>%  
  filter(species == "Adelie") %>%  
  select(island, sex) %>%  
  count(island, sex)
```

```
# A tibble: 8 × 3  
  island    sex      n  
  <fct>    <fct> <int>  
1 Biscoe  female   22  
2 Biscoe  male     22  
3 Dream   female   27  
4 Dream   male     28  
5 Dream   <NA>     1  
6 Torgersen female   24  
7 Torgersen male     23  
8 Torgersen <NA>     5
```

Acréscimo de novas colunas ao banco de dados

Construa uma nova coluna: `mutate()`



Exemplo: comprimento da nadadeira em cm

```
penguins %>%  
  select(species, island, flipper_length_mm)
```

```
# A tibble: 344 × 3  
  species island    flipper_length_mm  
  <fct>    <fct>          <int>  
1 Adelie  Torgersen         181  
2 Adelie  Torgersen         186  
3 Adelie  Torgersen         195  
4 Adelie  Torgersen          NA  
5 Adelie  Torgersen         193  
6 Adelie  Torgersen         190  
7 Adelie  Torgersen         181  
8 Adelie  Torgersen         195  
9 Adelie  Torgersen         193  
10 Adelie Torgersen         190  
# ... with 334 more rows
```

Exemplo: comprimento do bico em cm

```
penguins %>%  
  select(species, island, flipper_length_mm) %>%  
  mutate(flipper_length_cm = flipper_length_mm / 10)
```

```
# A tibble: 344 × 4
```

	species	island	flipper_length_mm	flipper_length_cm
	<fct>	<fct>	<int>	<dbl>
1	Adelie	Torgersen	181	18.1
2	Adelie	Torgersen	186	18.6
3	Adelie	Torgersen	195	19.5
4	Adelie	Torgersen	NA	NA
5	Adelie	Torgersen	193	19.3
6	Adelie	Torgersen	190	19
7	Adelie	Torgersen	181	18.1
8	Adelie	Torgersen	195	19.5
9	Adelie	Torgersen	193	19.3
10	Adelie	Torgersen	190	19

```
# ... with 334 more rows
```

Faça várias colunas de uma só vez

```
penguins %>%  
  select(species, island, flipper_length_mm) %>%  
  mutate(  
    flipper_length_cm = flipper_length_mm / 10, # <- observe a virgula  
    flipper_length_in = flipper_length_mm / 25.4  
  )
```

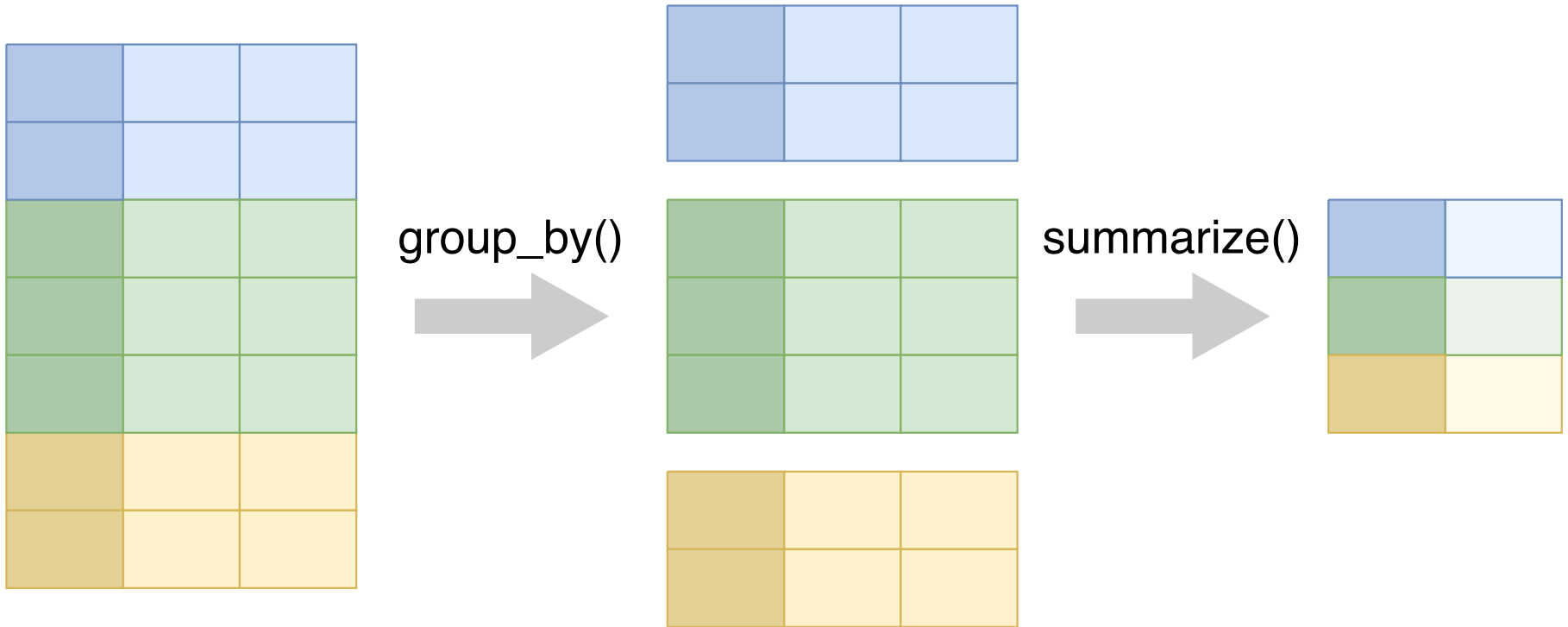
A tibble: 344 × 5

	species	island	flipper_length_mm	flipper_length_cm	flipper_length_in
	<fct>	<fct>	<int>	<dbl>	<dbl>
1	Adelie	Torgersen	181	18.1	7.13
2	Adelie	Torgersen	186	18.6	7.32
3	Adelie	Torgersen	195	19.5	7.68
4	Adelie	Torgersen	NA	NA	NA
5	Adelie	Torgersen	193	19.3	7.60
6	Adelie	Torgersen	190	19	7.48
7	Adelie	Torgersen	181	18.1	7.13
8	Adelie	Torgersen	195	19.5	7.68
9	Adelie	Torgersen	193	19.3	7.60
10	Adelie	Torgersen	190	19	7.48

... with 334 more rows

Analise segmentos: `group_by()` and `summarize()`

Analyse segmentos: `group_by()` and `summarize()`



Exemplo de aplicação de agrupamento: Contagem

Anteriormente, contamos assim:

```
penguins %>%  
  count(species)
```

```
# A tibble: 3 × 2  
  species      n  
  <fct>    <int>  
1 Adelie   152  
2 Chinstrap 68  
3 Gentoo  124
```

Agora vamos fazer da maneira mais difícil

Exemplo de aplicação de agrupamento: Contagem

Vamos voltar aos dados brutos:

```
penguins
```



```
# A tibble: 344 × 8
  species island bill_length_mm bill_depth_mm flipper_length_mm body_mass_g
  <fct>   <fct>         <dbl>         <dbl>           <int>         <int>
1 Adelie  Torgersen         39.1          18.7            181          3750
2 Adelie  Torgersen         39.5          17.4            186          3800
3 Adelie  Torgersen         40.3           18             195          3250
4 Adelie  Torgersen          NA           NA              NA           NA
5 Adelie  Torgersen         36.7          19.3            193          3450
6 Adelie  Torgersen         39.3          20.6            190          3650
7 Adelie  Torgersen         38.9          17.8            181          3625
8 Adelie  Torgersen         39.2          19.6            195          4675
9 Adelie  Torgersen         34.1          18.1            193          3475
10 Adelie Torgersen         42           20.2            190          4250
# ... with 334 more rows, and 2 more variables: sex <fct>, year <int>
```

Exemplo de aplicação de agrupamento: Contagem

vamos agrupá-lo:

```
penguins %>%  
  group_by(species)
```

```
# A tibble: 344 × 8
```

```
# Groups:   species [3]
```

	species	island	bill_length_mm	bill_depth_mm	flipper_length_mm	body_mass_g
	<fct>	<fct>	<dbl>	<dbl>	<int>	<int>
1	Adelie	Torgersen	39.1	18.7	181	3750
2	Adelie	Torgersen	39.5	17.4	186	3800
3	Adelie	Torgersen	40.3	18	195	3250
4	Adelie	Torgersen	NA	NA	NA	NA
5	Adelie	Torgersen	36.7	19.3	193	3450
6	Adelie	Torgersen	39.3	20.6	190	3650
7	Adelie	Torgersen	38.9	17.8	181	3625
8	Adelie	Torgersen	39.2	19.6	195	4675
9	Adelie	Torgersen	34.1	18.1	193	3475
10	Adelie	Torgersen	42	20.2	190	4250

```
# ... with 334 more rows, and 2 more variables: sex <fct>, year <int>
```

Exemplo de aplicação de agrupamento: Contagem

e summarize:

```
penguins %>%  
  group_by(species) %>%  
  summarize(  
    n = n() # n() returns the number of observations per group  
  )
```

```
# A tibble: 3 × 2  
  species      n  
  <fct>    <int>  
1 Adelie    152  
2 Chinstrap  68  
3 Gentoo    124
```

Exemplo de aplicação de agrupamento: Contagem

agora vamos agrupar por multiplas variáveis:

```
penguins %>%  
  group_by(species, island)
```

```
# A tibble: 344 × 8  
# Groups:   species, island [5]  
  species island    bill_length_mm bill_depth_mm flipper_length_mm body_mass_g  
  <fct>    <fct>          <dbl>          <dbl>          <int>          <int>  
1 Adelie  Torgersen         39.1           18.7           181           3750  
2 Adelie  Torgersen         39.5           17.4           186           3800  
3 Adelie  Torgersen         40.3           18            195           3250  
4 Adelie  Torgersen         NA            NA            NA            NA  
5 Adelie  Torgersen         36.7           19.3           193           3450  
6 Adelie  Torgersen         39.3           20.6           190           3650  
7 Adelie  Torgersen         38.9           17.8           181           3625  
8 Adelie  Torgersen         39.2           19.6           195           4675  
9 Adelie  Torgersen         34.1           18.1           193           3475  
10 Adelie Torgersen         42            20.2           190           4250  
# ... with 334 more rows, and 2 more variables: sex <fct>, year <int>
```

Exemplo de aplicação de agrupamento: Contagem

e summarize:

```
penguins %>%  
  group_by(species, island) %>%  
  summarize(  
    n = n() # n() returns the number of observations per group  
  )
```

`summarise()` has grouped output by 'species'. You can override using the
`.groups` argument.

```
# A tibble: 5 × 3  
# Groups:   species [3]  
  species island      n  
  <fct>   <fct>   <int>  
1 Adelie  Biscoe     44  
2 Adelie  Dream      56  
3 Adelie  Torgersen  52  
4 Chinstrap Dream     68  
5 Gentoo  Biscoe    124
```


Exemplo de aplicação de agrupamento: Contagem

`count(...)` is a short-cut for `group_by(...)` `%>% summarize(n = n())`

```
penguins %>%  
  count(species)
```

```
# A tibble: 3 × 2  
  species      n  
  <fct>    <int>  
1 Adelie   152  
2 Chinstrap 68  
3 Gentoo  124
```

```
penguins %>%  
  group_by(species) %>%  
  summarize(  
    n = n()  
  )
```

```
# A tibble: 3 × 2  
  species      n  
  <fct>    <int>  
1 Adelie   152  
2 Chinstrap 68  
3 Gentoo  124
```

o output é exatamente o mesmo

Realizando múltiplos sumários de uma vez

```
penguins %>%  
  group_by(species) %>%  
  summarize(  
    n = n(),                                # number of penguins  
    mean_mass = mean(body_mass_g),          # mean body mass  
    max_flipper_length = max(flipper_length_mm), # max flipper length  
    percent_female = sum(sex == "female")/n() # percent of female penguins  
  )
```

```
# A tibble: 3 × 5  
  species      n mean_mass max_flipper_length percent_female  
  <fct>    <int>    <dbl>          <int>          <dbl>  
1 Adelie   152      NA              NA              NA  
2 Chinstrap  68  3733.             212             0.5  
3 Gentoo   124      NA              NA              NA
```

Cada etapa do `summarize()` cria uma nova coluna

Mas por que os `NAs`?

Realizando múltiplos sumários de uma vez

```
penguins %>%  
  group_by(species) %>%  
  summarize(  
    n = n(),  
    mean_mass = mean(body_mass_g, na.rm = TRUE),  
    max_flipper_length = max(flipper_length_mm, na.rm = TRUE),  
    percent_female = sum(sex == "female", na.rm = TRUE)/sum(!is.na(sex))  
  )
```

```
# A tibble: 3 × 5  
  species      n mean_mass max_flipper_length percent_female  
  <fct>    <int>    <dbl>          <int>          <dbl>  
1 Adelie   152    3701.           210            0.5  
2 Chinstrap  68    3733.           212            0.5  
3 Gentoo  124    5076.           231            0.487
```

Nós precisamos dizer para o R exatamente como os **NAs** devem ser tratados

