Column-wise operations

It's often useful to perform the same operation on multiple columns, but copying and pasting is both tedious and error prone:

```
df %>%
  group_by(g1, g2) %>%
  summarise(a = mean(a), b = mean(b), c = mean(c), d = mean(d))
```

(If you're trying to compute mean(a, b, c, d) for each row, instead see vignette("rowwise"))

This vignette will introduce you to the across() function, which lets you rewrite the previous code more succinctly:

```
df %>%
  group_by(g1, g2) %>%
  summarise(across(a:d, mean))
```

We'll start by discussing the basic usage of across(), particularly as it applies to summarise(), and show how to use it with multiple functions. We'll then show a few uses with other verbs. We'll finish off with a bit of history, showing why we prefer across() to our last approach (the _if(), _at() and _all() functions) and how to translate your old code to the new syntax.

```
library(dplyr, warn.conflicts = FALSE)
```

Basic usage

across() has two primary arguments:

- The first argument, .cols, selects the columns you want to operate on. It uses tidy selection (like select()) so you can pick variables by position, name, and type.
- The second argument, .fns, is a function or list of functions to apply to each column. This can also be a purrr style formula (or list of formulas) like ~ .x / 2. (This argument is optional, and you can omit it if you just want to get the underlying data; you'll see that technique used in vignette("rowwise").)

Here are a couple of examples of across() in conjunction with its favourite verb, summarise(). But you can use across() with any dplyr verb, as you'll see a little later.

```
starwars %>%
 summarise(across(where(is.character), n_distinct))
#> # A tibble: 1 × 8
     name hair_color skin_color eye_color sex gender homeworld species
#>
             <int> <int> <int> <int> <int><</pre>
                                                       <int>
    <int>
                                  15 5
                         31
                                                 3
                                                          49
#> 1
       87
                12
                                                                 38
starwars %>%
 group_by(species) %>%
```

```
filter(n() > 1) %>%
 summarise(across(c(sex, gender, homeworld), n_distinct))
#> # A tibble: 9 × 4
#> species sex gender homeworld
    <chr> <int> <int>
                           <int>
               1
#> 1 Droid
                      2
#> 2 Gungan
              1
                     1
                              1
#> 3 Human
                      2
                              15
                2
#> 4 Kaminoan
#> # i 5 more rows
starwars %>%
 group by(homeworld) %>%
 filter(n() > 1) %>%
 summarise(across(where(is.numeric), ~ mean(.x, na.rm = TRUE)))
#> # A tibble: 10 × 4
  homeworld height mass birth year
    <chr>
            <dbl> <dbl>
                             <dbL>
#> 1 Alderaan 176. 64
                              43
#> 2 Corellia 175 78.5
                              25
#> 3 Coruscant 174. 50
                              91
                          31.5
#> 4 Kamino
           208. 83.1
#> # i 6 more rows
```

Because across() is usually used in combination with summarise() and mutate(), it doesn't select grouping variables in order to avoid accidentally modifying them:

Multiple functions

You can transform each variable with more than one function by supplying a named list of functions or lambda functions in the second argument:

```
min_max <- list(
    min = ~min(.x, na.rm = TRUE),
    max = ~max(.x, na.rm = TRUE)
)
starwars %>% summarise(across(where(is.numeric), min_max))
#> # A tibble: 1 × 6
#> height_min height_max mass_min mass_max birth_year_min birth_year_max
```

```
<dbL>
                                                          <dbL>
#>
           <int>
                      <int>
                                          <dbL>
                                                                          <dbL>
#> 1
              66
                        264
                                   15
                                           1358
                                                              8
                                                                            896
starwars %>% summarise(across(c(height, mass, birth year), min max))
#> # A tibble: 1 × 6
     height_min height_max mass_min mass_max birth_year_min birth_year_max
          <int>
#>
                      <int>
                                <dbL>
                                          <dbL>
                                                          <dbl>
                                                                          <dbL>
#> 1
              66
                        264
                                   15
                                           1358
                                                              8
                                                                            896
```

Control how the names are created with the .names argument which takes a glue spec:

```
starwars %>% summarise(across(where(is.numeric), min max, .names = "{.fn}.{.col}"))
         #> # A tibble: 1 × 6
              min.height max.height min.mass max.mass min.birth_year max.birth_year
         #>
                   <int>
                               <int>
                                         <dbl>
                                                  <dbl>
                                                                  <dbl>
                                                                                 <dbL>
         #> 1
                       66
                                 264
                                           15
                                                   1358
                                                                                   896
                                                                      8
         starwars %>% summarise(across(c(height, mass, birth_year), min_max, .names = "{.fn}.
{.col}"))
         #> # A tibble: 1 × 6
              min.height max.height min.mass max.mass min.birth_year max.birth_year
         #>
                   <int>
                               <int>
                                         <dbL>
                                                  <dbl>
                                                                  <dbL>
                                                                                 <dbL>
         #> 1
                       66
                                                   1358
                                                                      8
                                                                                   896
                                 264
                                           15
```

If you'd prefer all summaries with the same function to be grouped together, you'll have to expand the calls yourself:

```
starwars %>% summarise(
  across(c(height, mass, birth year), ~min(.x, na.rm = TRUE), .names = "min {.col}"),
  across(c(height, mass, birth_year), ~max(.x, na.rm = TRUE), .names = "max_{.col}")
)
#> # A tibble: 1 × 6
     min height min mass min birth year max height max mass max birth year
                                   <dbL>
          <int>
                   <dbL>
                                               <int>
                                                        <dbL>
                                                                        <dbL>
#>
#> 1
             66
                       15
                                       8
                                                 264
                                                         1358
                                                                          896
```

(One day this might become an argument to across() but we're not yet sure how it would work.)

We cannot however use where(is.numeric) in that last case because the second across() would pick up the variables that were newly created ("min_height", "min_mass" and "min_birth_year").

We can work around this by combining both calls to across() into a single expression that returns a tibble:

```
starwars %>% summarise(
  tibble(
    across(where(is.numeric), ~min(.x, na.rm = TRUE), .names = "min_{.col}"),
    across(where(is.numeric), ~max(.x, na.rm = TRUE), .names = "max_{.col}")
  )
)
#> # A tibble: 1 × 6
     min_height min_mass min_birth_year max_height max_mass max_birth_year
          <int>
                                   <dhl>
#>
                    <dbL>
                                               <int>
                                                        <dbL>
                                                                        <dbL>
#> 1
             66
                      15
                                       8
                                                 264
                                                         1358
                                                                          896
```

Alternatively we could reorganize results with relocate():

Current column

If you need to, you can access the name of the "current" column inside by calling <code>cur_column()</code>. This can be useful if you want to perform some sort of context dependent transformation that's already encoded in a vector:

Gotchas

Be careful when combining numeric summaries with where(is.numeric):

Here n becomes NA because n is numeric, so the across() computes its standard deviation, and the standard deviation of 3 (a constant) is NA. You probably want to compute n() last to avoid this problem:

Alternatively, you could explicitly exclude n from the columns to operate on:

Another approach is to combine both the call to n() and across() in a single expression that returns a tibble:

```
df %>%
   summarise(
     tibble(n = n(), across(where(is.numeric), sd))
)
#> n x     y
#> 1 3 1 4.041452
```

Other verbs

So far we've focused on the use of across() with summarise(), but it works with any other dplyr verb that uses data masking:

• Rescale all numeric variables to range 0-1:

For some verbs, like <code>group_by()</code>, <code>count()</code> and <code>distinct()</code>, you don't need to supply a summary function, but it can be useful to use tidy-selection to dynamically select a set of columns. In those cases, we recommend using the complement to <code>across()</code>, <code>pick()</code>, which works like <code>across()</code> but doesn't apply any functions and instead returns a data frame containing the selected columns.

Find all distinct

```
starwars %>% distinct(pick(contains("color")))
#> # A tibble: 67 × 3
#> hair_color skin_color eye_color
#> <chr> <chr> <chr> #> 1 blond fair blue
#> 2 <NA> gold yellow
#> 3 <NA> white, blue red
```

```
#> 4 none white yellow
#> # i 63 more rows
```

Count all combinations of variables with a given pattern:

```
starwars %>% count(pick(contains("color")), sort = TRUE)
#> # A tibble: 67 × 4
     hair_color skin_color eye_color
     <chr>>
                <chr>
                           <chr>>
                                      <int>
#> 1 brown
                Light
                           brown
                                          6
#> 2 brown
                fair
                           blue
                                          4
#> 3 none
                           black
                grey
                                          4
#> 4 black
                dark
                           brown
                                          3
#> # i 63 more rows
```

across() doesn't work with select() or rename() because they already use tidy select syntax; if you want to transform column names with a function, you can use rename_with().

filter()

We cannot directly use <code>across()</code> in <code>filter()</code> because we need an extra step to combine the results. To that end, <code>filter()</code> has two special purpose companion functions:

• if_any() keeps the rows where the predicate is true for at least one selected column:

```
starwars %>%
  filter(if_any(everything(), ~ !is.na(.x)))
#> # A tibble: 87 × 14
    name
              height mass hair_color skin_color eye_color birth_year sex
                                                                              aender
     <chr>>
                <int> <dbl> <chr>
                                       <chr>
                                                   <chr>>
                                                                  <dbl> <chr> <chr>
#> 1 Luke Skv...
                  172
                         77 blond
                                       fair
                                                  blue
                                                                  19
                                                                       male mascu...
#> 2 C-3P0
                  167
                       75 <NA>
                                                  yellow
                                       gold
                                                                 112
                                                                        none mascu...
#> 3 R2-D2
                  96
                        32 <NA>
                                       white, bl... red
                                                                   33 none mascu...
#> 4 Darth Va...
                  202
                       136 none
                                       white
                                                  yellow
                                                                   41.9 male mascu...
#> # i 83 more rows
#> # i 5 more variables: homeworld <chr>, species <chr>, films <list>,
     vehicles <list>, starships <list>
```

• if_all() keeps the rows where the predicate is true for all selected columns:

```
starwars %>%
  filter(if_all(everything(), ~ !is.na(.x)))
#> # A tibble: 29 × 14
    name
               height mass hair color skin color eye color birth year sex
     <chr>>
                <int> <dbl> <chr>
                                       <chr>
                                                   <chr>>
                                                                  <dbl> <chr> <chr>
                       77 blond
#> 1 Luke Sky...
                 172
                                       fair
                                                  blue
                                                                  19 male mascu...
#> 2 Darth Va...
                  202
                      136 none
                                       white
                                                  yellow
                                                                   41.9 male mascu...
#> 3 Leia Org...
                  150
                        49 brown
                                       Light
                                                  brown
                                                                   19 fema... femin...
#> 4 Owen Lars
                 178
                       120 brown, gr... light
                                                  blue
                                                                   52 male mascu...
#> # i 25 more rows
```

```
#> # i 5 more variables: homeworld <chr>, species <chr>, films <list>,
#> # vehicles <list>, starships <list>
```

```
_if, _at, _all
```

Prior versions of dplyr allowed you to apply a function to multiple columns in a different way: using functions with _if, _at, and _all() suffixes. These functions solved a pressing need and are used by many people, but are now superseded. That means that they'll stay around, but won't receive any new features and will only get critical bug fixes.

Why do we like across()?

Why did we decide to move away from these functions in favour of across()?

1. across() makes it possible to express useful summaries that were previously impossible:

```
df %>%
  group_by(g1, g2) %>%
  summarise(
    across(where(is.numeric), mean),
    across(where(is.factor), nlevels),
    n = n(),
)
```

- 2. across() reduces the number of functions that dplyr needs to provide. This makes dplyr easier for you to use (because there are fewer functions to remember) and easier for us to implement new verbs (since we only need to implement one function, not four).
- 3. across() unifies _if and _at semantics so that you can select by position, name, and type, and you can now create compound selections that were previously impossible. For example, you can now transform all numeric columns whose name begins with "x": across(where(is.numeric) & starts_with("x")).
- 4. across() doesn't need to use vars(). The _at() functions are the only place in dplyr where you have to manually quote variable names, which makes them a little weird and hence harder to remember.

Why did it take so long to discover across()?

It's disappointing that we didn't discover across() earlier, and instead worked through several false starts (first not realising that it was a common problem, then with the $_{each()}$ functions, and most recently with the $_{if()/_{at()}/_{at()}}$ functions). But across() couldn't work without three recent discoveries:

- You can have a column of a data frame that is itself a data frame. This is something provided by base R, but it's not very well documented, and it took a while to see that it was useful, not just a theoretical curiosity.
- We can use data frames to allow summary functions to return multiple columns.
- We can use the absence of an outer name as a convention that you want to unpack a data frame column into individual columns.

How do you convert existing code?

Fortunately, it's generally straightforward to translate your existing code to use across():

- Strip the _if(), _at() and _all() suffix off the function.
- Call across(). The first argument will be:
 - For _if(), the old second argument wrapped in where().
 - For _at(), the old second argument, with the call to vars() removed.
 - For _all(), everything().

The subsequent arguments can be copied as is.

For example:

```
df %>% mutate_if(is.numeric, ~mean(.x, na.rm = TRUE))
# ->
df %>% mutate(across(where(is.numeric), ~mean(.x, na.rm = TRUE)))
df %>% mutate_at(vars(c(x, starts_with("y"))), mean)
# ->
df %>% mutate(across(c(x, starts_with("y")), mean))

df %>% mutate_all(mean)
# ->
df %>% mutate(across(everything(), mean))
```

There are a few exceptions to this rule:

- rename_*() and select_*() follow a different pattern. They already have select semantics, so are generally used in a different way that doesn't have a direct equivalent with across(); use the new rename_with() instead.
- Previously, filter_*() were paired with the all_vars() and any_vars() helpers. The new helpers
 if_any() and if_all() can be used inside filter() to keep rows for which the predicate is true for at
 least one, or all selected columns:

```
df <- tibble(x = c("a", "b"), y = c(1, 1), z = c(-1, 1))
# Find all rows where EVERY numeric variable is greater than zero
df %>% filter(if_all(where(is.numeric), ~ .x > 0))
#> # A tibble: 1 × 3
            У
#> <chr> <dbl> <dbl>
#> 1 b
       1 1
# Find all rows where ANY numeric variable is greater than zero
df %>% filter(if_any(where(is.numeric), ~ .x > 0))
#> # A tibble: 2 × 3
   X
           y z
#> <chr> <dbl> <dbl>
#> 1 a
           1 -1
      1 1
#> 2 b
```

• When used in a mutate(), all transformations performed by an across() are applied at once. This is different to the behaviour of mutate_if(), mutate_at(), and mutate_all(), which apply the transformations one at a time. We expect that you'll generally find the new behaviour less surprising: