Partitioning Mutate, Example 2 John Mount, Win-Vector LLC 2017-11-24

Sparklyr, with its dplyr translations allows R, to perform the heavy lifting that has traditionally been the exclusive domain of proprietary systems such as SAS. In general, dplyr is good at handling intermediate variables in the mutate function so users don't need to think about it. However, some of that breaks down when the processing is done on the Apache Spark side. Win-Vector LLC developed the seplyr package to use with consulting clients to mitigate some of these situations. In this article we will demonstrate we seplyr functions: if_else_device() and partition_mutate_qt().

This is a follow-on example building on our "Partitioning Mutate" article, showing a larger block sequence based on swaps.² For more motivation and context please see the first article.

Please consider the following example data (on a remote Spark cluster).

```
class(d)
## [1] "tbl_spark" "tbl_sql" "tbl_lazy"
## [4] "tbl"
d %.>%
# avoid https://qithub.com/tidyverse/dplyr/issues/3216
```

rowNum a_2 b_2 d_2 a_1 b_1 c_1 c_2 d_1 e_1 e_2 NANA NANANA NA NANA NANA1 2 NA NANA NANA NANANANA NA 3 NA NA NANA NA NA NA NA NA NA 4 NANA NANA NANA NA NANA NA 5 NA NA NA NA NA NA NA NA NA NA

We find in non-trivial projects it is often necessary to simulate block-if(){}else{} structures in dplyr pipelines.

For our example: suppose we wish to assign columns in a complementary to treatment and control design^3

To write such a procedure in pure $\tt dplyr$ we might simulate block with code such as the following 4

```
library("seplyr")
packageVersion("seplyr")
```

dplyr::collect(.) %.>%

knitr::kable(.)

- $^{\rm 1}$ And we distribute the package as open-source to give back to the R community.
- ² The source code for this article can be found here.

³ Abraham Wald designed some sequential analysis procedures in this way as Nina Zumel remarked. Another string example is conditionals where you are trying to vary on a perrow basis which column is assigned to, instead of varying what value is assigned from.

⁴ Only showing work on the a group right now. We are assuming we want to perform this task on all the grouped letter columns.

```
## [1] '0.5.1'
plan <- if_else_device(</pre>
  testexpr =
    "rand()>=0.5".
     thenexprs = c(
       "a_1" := "'treatment'",
       "a 2" := "'control'"),
     elseexprs = c(
       "a 1" := "'control'",
       "a 2" := "'treatment'")) %.>%
  partition_mutate_se(.)
```

We are using the indent notation to indicate the code-blocks we are simulating with row-wise if(){}else{} blocks.⁵ The if else device is also using quoted expressions (or value-oriented standard notation).⁶ In the end we can examine and execute the mutate plan:

```
print(plan)
## $group00001
## ifebtest_w49z9gg2vgoj
           "rand()>=0.5"
##
##
  $group00002
##
##
## "ifelse( ifebtest_w49z9gg2vgoj, 'treatment', a_1)"
##
     "ifelse( ifebtest_w49z9gg2vgoj, 'control', a_2)"
##
##
## $group00003
##
                                                        a_1
##
     "ifelse(!(ifebtest w49z9gg2vgoj), 'control', a 1)"
## "ifelse( !( ifebtest_w49z9gg2vgoj ), 'treatment', a_2)"
d %.>%
  mutate_seb(., plan) %.>%
  select_se(., grepdf('^ifebtest_.*', ., invert=TRUE)) %.>%
  dplyr::collect(.) %.>%
  knitr::kable(.)
```

Our advice is to compose the expressions using your smart R-code editor of choice and then throw on the additional quote marks after you have the statements as you want them.

```
rowNum
        a 1
                   a_2
                              b_1
                                   b_{2}
                                         c_1
                                              c_2
                                                    d_1
                                                         d_2
                                                               e_1
                                                                    e_2
                                   NA
                                         NA
                                              NA
                                                    NA
                                                                    NA
                              NA
                                                         NA
                                                               NA
         treatment
                   control
      2
         control
                   treatment
                             NA
                                   NA
                                         NA
                                              NA
                                                    NA
                                                         NA
                                                               NA
                                                                    NA
```

⁵ For more on this concept, please see: the if_else_device reference.

⁶ One can over-worry about this, but in the end all a non-standard evaluation scheme saves you is a few quote marks (at the cost of transparency, and a lot of downstream headaches).

rowNum	a_1	a_2	b_1	b_2	c_1	c_2	d_1	d_2	e_1	e_2
3	treatment	control	NA							
4	treatment	control	NA							
5	treatment	control	NA							

Our larger goal was to perform this same operation on each of the 5 letter groups.

We do this easily as follows:⁷

```
plan <- lapply(c('a', 'b', 'c', 'd', 'e'),</pre>
               function(gi) {
                 if_else_device(
                   "rand()>=0.5",
                   thenexprs = c(
                     paste0(gi, "_1") := "'treatment'",
                     paste0(gi, "_2") := "'control'"),
                   elseexprs = c(
                     paste0(gi, "_1") := "'control'",
                     paste0(gi, "_2") := "'treatment'"))
               }) %.>%
  unlist(.) %.>%
  partition_mutate_se(.)
d %.>%
  mutate_seb(., plan) %.>%
  select_se(., grepdf('^ifebtest_.*', ., invert=TRUE)) %.>%
  dplyr::collect(.) %.>%
  knitr::kable(.)
```

⁷ A better overall design would be to use cdata::moveValuesToRowsN(), then perform a single bulk operation on rows, and then pivot/transpose back with cdata::moveValuesToColumnsN(). But let's see how we simply work with a problem at hand.

${\rm rowNum}$	a_1	a_2	b_1	b_2	c_1	c_2	d_1	d_2	e_1
1	control	treatment	treatment	control	control	treatment	treatment	control	control
2	treatment	control	control	treatment	control	treatment	treatment	control	treatment
3	treatment	control	control	treatment	control	treatment	treatment	control	treatment
4	treatment	control	treatment	control	treatment	$\operatorname{control}$	control	treatment	control
5	treatment	control	control	treatment	treatment	control	treatment	control	control

Please keep in mind: we are using a very simple and regular sequence only for purposes of illustration. The intent is to show the types of issues one runs into when standing-up non-trivial applications in Sparklyr.

The purpose of seplyr::partition_mutate_qt() is to re-arrange statements and break them into blocks of non-dependent statements (no statement in a block depends on any other in the same block, and all value dependencies are respected by the block order). seplyr::partition_mutate_qt() if further defined to do this in a performant manner.⁸

Without such partition planning the current version of dplyr (0.7.4) the results of dplyr::mutate() do not seem to be well-defined when values are created and re-used in the same dplyr::mutate() block. This is not a currently documented limitation, but it is present:

```
ex <- dplyr::mutate(d,
                      condition_tmp = rand()>=0.5,
                      a_1 = ifelse( condition_tmp,
                                      'treatment',
                                      a<sub>1</sub>),
                      a_2 = ifelse( condition_tmp,
                                      'control',
                                      a<sub>2</sub>),
                      a_1 = ifelse( !( condition_tmp ),
                                      'control',
                                      a_1),
                      a 2 = ifelse(!( condition tmp ),
                                      'treatment',
                                      a 2))
```

knitr::kable(dplyr::collect(dplyr::select(ex, a_1, a_2)))

a_1	a_2
NA	control
control	treatment

Notice above the many NA columns, which are errors.⁹

⁹ Note: no mere re-ordering of the statements would give this result.

```
dplyr::show_query(ex)
## <SQL>
## SELECT `rowNum`, `a_1`, `b_1`, `b_2`, `c_1`, `c_2`, `d_1`, `d_2`, `e_1`, `e_2`, `condition_tmp`, CAS
## FROM (SELECT `rowNum`, `b_1`, `b_2`, `c_1`, `c_2`, `d_1`, `d_2`, `e_1`, `e_2`, `condition_tmp`, CASE
## FROM (SELECT `rowNum`, `a_1`, `a_2`, `b_1`, `b_2`, `c_1`, `c_2`, `d_1`, `d_2`, `e_1`, `e_2`, RAND()
## FROM `d`) `htzbsqefec`) `bredwjgpwq`
```

Looking at the query we see that one of the conditional statements is missing (notice only 3 case statements, not 4):¹⁰

¹⁰ Likely the dplyr SQL generator does not perform a correct live-value analysis and therefor gets fooled into thinking a statement can safely be eliminated (when it can not). seplyr::partition_mutate_qt() performs a correct live value calculation and make sure dplyr::mutate() is only seeing trivial blocks (blocks where no value depends on any calculation in the same block).

⁸ That is to pick a small number of blocks, in our case the plan consisted of 3 blocks. The simple method of introducing a block boundary at each first use of derived value (without statement re-ordering) would create a very much larger set of blocks (which cause problems of their own). In particular the impression code and comments of upcoming dplyr fix appear to indicate an undesirable large number of blocks solution.

Conclusion

seplyr::if_else_device() and seplyr::partition_mutate_qt() type capability is essential for executing non-trivial code at scale in Sparklyr. For more on the if_else_device we suggest reading up on the function reference example, and for a review on the partition_mutate variations we suggest the "Partitioning Mutate" article.

Links

Win-Vector LLC supplies a number of open-source R packages for working effectively with big data. These include:

- wrapr: supplies code re-writing tools that make coding over "non standard evaluation" interfaces (such as dplyr) much easier.
- cdata: supplies pivot/un-pivot functionality at big data scale.
- seplyr: supplies improved interfaces for many data manipulation
- replyr: supplies tools and patches for using dplyr on big data.

Topics such as the above are often discussed on the Win-Vector blog.