Creating ADaM Subject-level Analysis (ADSL) using R admiral package

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Read in Data

To start, all data frames needed for the creation of ADSL should be read into the environment. This will be a company specific process. Some of the data frames needed may be DM, EX, DS, AE, and LB.

For example purpose, the CDISC Pilot SDTM datasets—which are included in {pharmaversesdtm}—are used.

```
Warning: package 'pharmaversesdtm' was built under R version 4.4.2

Warning: package 'lubridate' was built under R version 4.4.2

Attaching package: 'lubridate'

The following objects are masked from 'package:base':
```

The DM domain is used as the basis for ADSL:

date, intersect, setdiff, union

```
adsl <- dm %>%
  dplyr::select(-DOMAIN) %>%
  dplyr::mutate(TRT01P = ARM, TRT01A = ACTARM) # dim(adsl) 306 26
```

Derive Period, Subperiod, and Phase Variables

See the "Visit and Period Variables" vignette for more information.

If the variables are not derived based on a period reference dataset, they may be derived at a later point of the flow. For example, phases like "Treatment Phase" and "Follow up" could be derived based on treatment start and end date.

Derive treatment variables

TRTxxP

- Planned Treatment for Period xx (e.g. TRT01P for the first planned treatment)
- Subject-level identifier that represents the planned treatment for period xx. In a one-period randomized trial, TRT01P would be the treatment to which the subject was randomized. TRTxxP might be derived from the SDTM DM variable ARM. At least TRT01P is required.
- Derived from DM.ARM

TRTxxA

- Actual Treatment for Period xx
- Subject-level identifier that represents the actual treatment for the subject for period xx. Required when actual treatment does not match planned and there is an analysis of the data as treated.
- · Derived from DM.ACTARM

Derive treatment datetime, duration variables

EXSTDTM

- Numeric datetime of first exposure derived from character EXSTDTC
- Date with missing time can be imputated. e.g., EXSTDTC="2014-01-02" -> EXSTDTM= 2014-01-02 00:00:00

EXSTTMF

EXENDTM

- Numeric datetime of last exposure derived from character EXENDTC
- Date with missing time can be imputed. E.g., EXENDTC="2014-01-16" -> EXENDTM=2014-01-16 23:59:59

EXENTMF

The function derive_vars_merged() can be used to derive the treatment start and end date/times using the ex domain. A pre-processing step for ex is required to convert the variable EXSTDTC and EXSTDTC to datetime variables and impute missing date or time components. Conversion and imputation is done by derive_vars_dtm().

Example calls:

```
# Derive a datetime object --DTM from a date character vector --DTC
ex_ext <- ex %>%
  admiral::derive_vars_dtm(
    dtc = EXSTDTC
    ,new_vars_prefix = "EXST"
    ,time_imputation = "first") %>%
  admiral::derive_vars_dtm(
    dtc = EXENDTC
    ,new_vars_prefix = "EXEN"
    ,time_imputation = "last") # dim(ex_ext) 591 21

ex_ext %>% select(EXSTDTC, EXSTDTM, EXSTTMF, EXENDTC, EXENDTM, EXENTMF) %>% head()
```

```
# A tibble: 6 x 6
 EXSTDTC
          EXSTDTM
                                EXSTTMF EXENDTC
                                                   EXENDTM
                                                                       EXENTMF
            <dttm>
                                                   <dttm>
  <chr>
                                 <chr>
                                        <chr>
                                                                        <chr>>
1 2014-01-02 2014-01-02 00:00:00 H
                                        2014-01-16 2014-01-16 23:59:59 H
2 2014-01-17 2014-01-17 00:00:00 H
                                         2014-06-18 2014-06-18 23:59:59 H
                                         2014-07-02 2014-07-02 23:59:59 H
3 2014-06-19 2014-06-19 00:00:00 H
4 2012-08-05 2012-08-05 00:00:00 H
                                        2012-08-27 2012-08-27 23:59:59 H
5 2012-08-28 2012-08-28 00:00:00 H
                                         2012-09-01 2012-09-01 23:59:59 H
6 2013-07-19 2013-07-19 00:00:00 H
                                        2013-08-01 2013-08-01 23:59:59 H
```

TRTSDTM

- Datetime of First Exposure to Treatment
- Numeric version of datetime derived from DM.RFXSTDTC (but here it is from EXSTDTM)

TRTEDTM

- Datetime of Last Exposure to Treatment
- Numeric version of datetime derived from DM.RFXENDTC (but here it is from EXENDTM)

```
# Left join adsl and ex_ext
# new variables added: "TRTSDTM" "TRTSTMF" "TRTEDTM" "TRTETMF"
adsl <- adsl %>%
  derive_vars_merged(
   dataset_add = ex_ext
   # Observations from dataset_add that meet the conditions will be merged to adsl
    ,filter_add = (EXDOSE > 0 |
                    (EXDOSE == 0 &
                     str_detect(EXTRT, "PLACEBO"))) & !is.na(EXSTDTM)
    ,new_vars = exprs(TRTSDTM = EXSTDTM, TRTSTMF = EXSTTMF)
    ,order = exprs(EXSTDTM, EXSEQ)
    ,mode = "first"
    ,by_vars = exprs(STUDYID, USUBJID)) %>%
 derive_vars_merged(
   dataset_add = ex_ext,
   filter_add = (EXDOSE > 0 |
                    (EXDOSE == 0 &
                       str_detect(EXTRT, "PLACEBO"))) & !is.na(EXENDTM),
   new_vars = exprs(TRTEDTM = EXENDTM, TRTETMF = EXENTMF),
   order = exprs(EXENDTM, EXSEQ),
   mode = "last",
   by_vars = exprs(STUDYID, USUBJID)
 ) # dim(adsl) after merging: 306 30 # dim(adsl) before merging: 306 26
# Old variables
ex_ext %>% select(EXSTDTM, EXSTTMF, EXENDTM, EXENTMF) %>% head()
# A tibble: 6 x 4
 EXSTDTM
                     EXSTTMF EXENDTM
                                                  EXENTMF
  <dt.tm>
                     <chr>
                             <dttm>
                                                  <chr>
1 2014-01-02 00:00:00 H
                              2014-01-16 23:59:59 H
2 2014-01-17 00:00:00 H
                            2014-06-18 23:59:59 H
3 2014-06-19 00:00:00 H
                            2014-07-02 23:59:59 H
                           2012-08-27 23:59:59 H
2012-09-01 23:59:59 H
4 2012-08-05 00:00:00 H
5 2012-08-28 00:00:00 H
6 2013-07-19 00:00:00 H
                            2013-08-01 23:59:59 H
# New variables
adsl %>% select(TRTSDTM, TRTSTMF, TRTEDTM, TRTETMF) %>% head()
# A tibble: 6 x 4
 TRTSDTM
                      TRTSTMF TRTEDTM
                                                  TRTETMF
 <dttm>
                     <chr> <dttm>
                                                  <chr>>
1 2014-01-02 00:00:00 H
                              2014-07-02 23:59:59 H
                           2012-09-01 23:59:59 H
2 2012-08-05 00:00:00 H
3 2013-07-19 00:00:00 H
                            2014-01-14 23:59:59 H
```

```
4 2014-03-18 00:00:00 H 2014-03-31 23:59:59 H
5 2014-07-01 00:00:00 H 2014-12-30 23:59:59 H
6 2013-02-12 00:00:00 H 2013-03-09 23:59:59 H
```

TRTSDT

- Date of First Exposure to Treatment
- Numeric version of date portion of DM.RFXSTDTC formatted as a SAS date (But here it is from TRTSDTM)

TRTEDT

- Date of Last Exposure to Treatment
- Numeric version of date portion of DM.RFXENDTC formatted as a SAS date (But here it is from TRTEDTM)

TRTDURD

- Total Treatment Duration (Days)
- Total treatment duration, as measured in days. More than one of TRTDURD, TRTDURM, and TRTDURY can be populated, but each represents the entire duration in its respective units.
- 'TRTDURD= TRTEDT- TRTSDT+1'

This call returns the original data frame with the column TRTSDTM, TRTSDTM, and TRTETMF added. Exposure observations with incomplete date and zero doses of non placebo treatments are ignored. Missing time parts are imputed as first or last for start and end date respectively.

The datetime variables returned can be converted to dates using the derive_vars_dtm_to_dt() function.

Now, that TRTSDT and TRTEDT are derived, the function derive_var_trtdurd() can be used to calculate the Treatment duration (TRTDURD).

```
# New variables added: "TRTSDT" "TRTEDT" "TRTDURD"
adsl <- adsl %>%

# Derive date variables from datetime variables
admiral::derive_vars_dtm_to_dt(source_vars = exprs(TRTSDTM, TRTEDTM)) %>%

# Derives total treatment duration (days) (TRTDURD). TRTDURD= TRTEDT- TRTSDT+1
admiral::derive_var_trtdurd() # dim(adsl) 306 33

adsl %>% select(TRTSDT, TRTEDT, TRTDURD) %>% head()
```

```
# A tibble: 6 x 3
  TRTSDT
             TRTEDT
                         TRTDURD
             <date>
  <date>
                           <dbl>
1 2014-01-02 2014-07-02
                             182
                              28
2 2012-08-05 2012-09-01
3 2013-07-19 2014-01-14
                             180
4 2014-03-18 2014-03-31
                              14
5 2014-07-01 2014-12-30
                             183
6 2013-02-12 2013-03-09
                              26
```

Derive disposition variables

DSSTDT

• Convert character disposition date DS.DSSTDTC to numeric date DSSTDT using derive_vars_dt()

EOSDT

- End of Study Date
- Date subject ended the study either date of completion or date of discontinuation or data cutoff date for interim analyses.
- Numeric version of DS.DSSTDTC or data cutoff date

The functions derive_vars_dt() and derive_vars_merged() can be used to derive a disposition date. First the character disposition date (DS.DSSTDTC) is converted to a numeric date (DSSTDT) calling derive_vars_dt(). The DS dataset is extended by the DSSTDT variable because the date is required by other derivations, e.g., RANDDT as well. Then the relevant disposition date is selected by adjusting the filter_add argument.

To add the End of Study date (EOSDT) to the input dataset, a call could be:

```
# A tibble: 6 x 2
DSSTDTC DSSTDT
<hr/>
<hr>
<1 2013-08-01 2013-08-01
2 2013-08-08 2013-08-08
3 2012-12-17 2012-12-17
4 2013-02-18 2013-02-18
5 2013-02-18 2013-02-18
```

```
6 2013-06-03 2013-06-03
# Check protocol milestones
ds_ext %>% filter(DSCAT=="PROTOCOL MILESTONE") %>% distinct(DSDECOD)
# A tibble: 1 x 1
  DSDECOD
  <chr>
1 RANDOMIZED
# Check disposition events
ds_ext %>% filter(DSCAT=="DISPOSITION EVENT") %>% distinct(DSDECOD)
# A tibble: 10 \times 1
   DSDECOD
   <chr>
 1 COMPLETED
 2 ADVERSE EVENT
 3 STUDY TERMINATED BY SPONSOR
 4 SCREEN FAILURE
 5 DEATH
 6 WITHDRAWAL BY SUBJECT
 7 PHYSICIAN DECISION
8 PROTOCOL VIOLATION
 9 LOST TO FOLLOW-UP
10 LACK OF EFFICACY
# Left join adsl and ds_ext
# New variable added: EOSDT
adsl <- admiral::derive_vars_merged(</pre>
  dataset=adsl # dim(adsl) 306 33
  ,dataset_add = ds_ext
  ,by_vars = exprs(STUDYID, USUBJID)
  ,new_vars = exprs(EOSDT = DSSTDT)
  ,filter_add = DSCAT == "DISPOSITION EVENT" & DSDECOD != "SCREEN FAILURE") # dim(adsl) 306 34
adsl %>% select(USUBJID, EOSDT) %>% tail()
# A tibble: 6 x 2
  USUBJID
             EOSDT
  <chr>
              <date>
1 01-718-1250 2014-02-08
2 01-718-1254 2014-01-09
3 01-718-1328 2013-05-01
4 01-718-1355 2013-08-29
5 01-718-1371 2013-08-08
```

6 01-718-1427 2013-02-18

The derive_vars_dt() function allows to impute partial dates as well. If imputation is needed and missing days are to be imputed to the first of the month and missing months to the first month of the year, set highest_imputation = "M".

EOSSTT

- End of Study Status
- The subject's status as of the end of study or data cutoff. Examples: COMPLETED, DISCONTINUED, ONGOING.
- Derived based on DS.DSCAT and DS.DSDECOD

```
# Example function format_eosstt():
format_eosstt <- function(x) {
   case_when(
     x %in% c("COMPLETED") ~ "COMPLETED"
     ,x %in% c("SCREEN FAILURE") ~ NA_character_
     ,TRUE ~ "DISCONTINUED")
}</pre>
```

The customized mapping function format_eosstt() can now be passed to the main function. For subjects without a disposition event the end of study status is set to "ONGOING" by specifying the missing_values argument.

```
# New variables added: EOSSTT (End of Study Status)
adsl <- adsl %>%
  derive_vars_merged(
    dataset_add = ds
    ,by_vars = exprs(STUDYID, USUBJID)
    ,filter_add = DSCAT == "DISPOSITION EVENT"
    ,new_vars = exprs(EOSSTT = format_eosstt(DSDECOD))
    ,missing_values = exprs(EOSSTT = "ONGOING")
    ) # dim(adsl) 306 34 before merging # dim(adsl) 306 35 after merging
adsl %>% select(USUBJID, EOSDT,EOSSTT) %>% tail()
```

This call would return the input dataset with the column EOSSTT added.

If the derivation must be changed, the user can create his/her own function to map DSDECOD to a suitable EOSSTT value.

DCSREAS

- · Reason for Discontinuation from Study
- Reason for subject's discontinuation from study. The source would most likely be the SDTM DS dataset. Null for subjects who completed the study.
- If DS.DSDECOD <> "COMPLETED where DSSCAT ="STUDY PARTICIPATION" (i.e. ADSL.EOSSTT is "DISCONTINUED") then ADSL.DCSREAS = DS.DSDECOD; If DS.DSDECOD = "COMPLETED" where DSSCAT = "STUDY PARTICIPATION", then ADSL.DCSREAS is; If there is no DS record where DSSCAT = "STUDY PARTICIPATION" (i.e. EOSSTT is "ONGOING") then ADSL.DCSREAS is null.

DCSREASP

- Reason Specified for Discontinuation from Study
- Additional detail regarding subject's discontinuation from study (e.g., description of "other").
- If DS.DSDECOD <> "COMPLETED" where DS.DSSCAT = "STUDY PARTICIPATION" (i.e. ADSL.EOSSTT is "DISCONTINUED") CO.COVAL / CO.COVAL1 where COREF = "PRIMARY REASON FOR STUDY DISCONTINUATION" (if populated); otherwise ADSL.DCSREASP is null.

Disposition Reasons

The main reason for discontinuation is usually stored in DSDECOD while DSTERM provides additional details regarding subject's discontinuation (e.g., description of "OTHER").

The function derive_vars_merged() can be used to derive a disposition reason (along with the details, if required) at a specific timepoint. The relevant observations are selected by adjusting the filter_add argument.

To derive the End of Study reason(s) (DCSREAS and DCSREASP), the function will map DCSREAS as DSDECOD, and DCSREASP as DSTERM if DSDECOD is not "COMPLETED", "SCREEN FAILURE", or NA, NA otherwise.

This call would return the input dataset with the column DCSREAS and DCSREASP added.

If the derivation must be changed, the user can define that derivation in the filter_add argument of the function to map DSDECOD and DSTERM to a suitable DCSREAS/DCSREASP value.

The call below maps DCSREAS and DCREASP as follows:

- DCSREAS as DSDECOD if DSDECOD is not "COMPLETED" or NA, NA otherwise
- DCSREASP as DSTERM if DSDECOD is equal to OTHER, NA otherwise

```
adsl <- adsl %>%
  derive_vars_merged(
    dataset_add = ds
    ,by_vars = exprs(USUBJID)
    ,new_vars = exprs(DCSREAS = DSDECOD, DCSREASP = DSTERM)
    ,filter_add = DSCAT == "DISPOSITION EVENT" &
        !(DSDECOD %in% c("SCREEN FAILURE", "COMPLETED", NA))
    ) # dim(adsl) 306 35 before merging # dim(adsl) 306 37 after merging
adsl %>% select(USUBJID,EOSDT,EOSSTT,DCSREAS,DCSREASP) %>% head()
```

```
# A tibble: 6 x 5
 USUBJID EOSDT
                                      DCSREAS
                                                                   DCSREASP
                         EOSSTT
  <chr>
             <date>
                         <chr>
                                      <chr>
                                                                   <chr>
1 01-701-1015 2014-07-02 COMPLETED
                                      <NA>
                                                                   <NA>
2 01-701-1023 2012-09-02 DISCONTINUED ADVERSE EVENT
                                                                   ADVERSE EVENT
3 01-701-1028 2014-01-14 COMPLETED
                                      <NA>
                                                                   <NA>
4 01-701-1033 2014-04-14 DISCONTINUED STUDY TERMINATED BY SPONSOR SPONSOR DECIS~
5 01-701-1034 2014-12-30 COMPLETED
                                      <NA>
                                                                   <NA>
6 01-701-1047 2013-03-29 DISCONTINUED ADVERSE EVENT
                                                                   ADVERSE EVENT
```

RANDDT

- Date of Randomization
- Required in randomized trials.
- DS.DSSTDTC is a character (text) variable with date in ISO 8601 format: YYYY-MM-DD (e.g. 1997-07-16). ADSL.RANDDT is the DS.DSSTDTC where DSDECOD = "RANDOMIZED", SAS date format DATE11.; If a subject was not randomized (e.g. Screen Failure) and there is no record in DS for the subject where DSDECOD = "Randomized" then ADSL.RANDDT is null.

The function derive_vars_merged() can be used to derive randomization date variable. To map Randomization Date (RANDDT), the call would be:

```
adsl <- adsl %>%
  derive_vars_merged(
    dataset_add = ds_ext
    ,filter_add = DSDECOD == "RANDOMIZED"
    ,by_vars = exprs(STUDYID, USUBJID)
    ,new_vars = exprs(RANDDT = DSSTDT)
)
adsl %>% select(USUBJID,RANDDT) %>% head()
```

```
# A tibble: 6 x 2
USUBJID RANDDT
<chr> <date>
```

```
1 01-701-1015 2014-01-02
2 01-701-1023 2012-08-05
3 01-701-1028 2013-07-19
4 01-701-1033 2014-03-18
5 01-701-1034 2014-07-01
6 01-701-1047 2013-02-12
```

Derive death variables

DTHDT

- · Date of death
- Date of subject's death. Derived from DM.DTHDTC.

The function derive_vars_dt() can be used to derive DTHDT. This function allows the user to impute the date as well.

Example calls:

```
adsl <- adsl %>%
 derive_vars_dt(
   new_vars_prefix = "DTH"
    ,dtc = DTHDTC
   #,date imputation = "first"
    ) # dim(adsl) 306 39
adsl %>% select(USUBJID,TRTEDT, DTHDTC, DTHDT) %>% filter(!is.na(DTHDT)) %>% head()
# A tibble: 3 x 4
          TRTEDT
 USUBJID
                         DTHDTC
                                    DTHDT
  <chr>
             <date>
                         <chr>
                                    <date>
1 01-701-1211 2013-01-12 2013-01-14 2013-01-14
2 01-704-1445 2014-11-01 2014-11-01 2014-11-01
3 01-710-1083 2013-08-01 2013-08-02 2013-08-02
```

This call would return the input dataset with the columns DTHDT added and, by default, the associated date imputation flag (DTHDTF) populated with the controlled terminology outlined in the ADaM IG for date imputations. If the imputation flag is not required, the user must set the argument flag_imputation to "none".

If imputation is needed and the date is to be imputed to the first day of the month/year the call would be:

```
adsl <- adsl %>%
  derive_vars_dt(
    new_vars_prefix = "DTH",
    dtc = DTHDTC,
    date_imputation = "first"
)
```

Warning: Variable "DTHDT" already exists in the dataset.

See also Date and Time Imputation.

DTHCAUS

- · Cause of death
- if the date of death is collected in the AE form when the AE is Fatal, the cause of death would be set to the preferred term (AEDECOD) of that Fatal AE, while if the date of death is collected in the DS form, the cause of death would be set to the disposition term (DSTERM). To achieve this, the 'event()' objects within 'derive_vars_extreme_event()' must be specified and defined such that they fit the study requirement.

DTHDOM

- · Death Domain
- Store the domain where the date of death is collected

DTHSEQ

- Death Sequence Number
- Store the xxSEQ value of that domain

The cause of death DTHCAUS can be derived using the function derive_vars_extreme_event().

Since the cause of death could be collected/mapped in different domains (e.g. DS, AE, DD), it is important the user specifies the right source(s) to derive the cause of death from.

For example, if the date of death is collected in the AE form when the AE is Fatal, the cause of death would be set to the preferred term (AEDECOD) of that Fatal AE, while if the date of death is collected in the DS form, the cause of death would be set to the disposition term (DSTERM). To achieve this, the event() objects within derive_vars_extreme_event() must be specified and defined such that they fit the study requirement.

An example call to derive_vars_extreme_event() would be:

```
),
      event(
        dataset_name = "ds",
        condition = DSDECOD == "DEATH" & grepl("DEATH DUE TO", DSTERM),
        set values to = exprs(DTHCAUS = DSTERM, DTHDOM = "DS", DTHSEQ = DSSEQ),
      )
    ),
    source_datasets = list(ae = ae, ds = ds),
    tmp_event_nr_var = event_nr,
    order = exprs(event_nr),
   mode = "first",
    new_vars = exprs(DTHCAUS, DTHDOM, DTHSEQ)
  ) # dim(adsl) 306 42
adsl %>% select(USUBJID, DTHDT, DTHCAUS, DTHDOM, DTHSEQ) %>% filter(!is.na(DTHDT)) %>% head()
# A tibble: 3 x 5
 USUBJID
          DTHDT
                         DTHCAUS
                                               DTHDOM DTHSEQ
  <chr>>
              <date>
                         <chr>
                                                <chr>
                                                        <dbl>
1 01-701-1211 2013-01-14 SUDDEN DEATH
                                                ΑE
                                                            9
2 01-704-1445 2014-11-01 COMPLETED SUICIDE
                                                ΑE
                                                            1
3 01-710-1083 2013-08-02 MYOCARDIAL INFARCTION AE
                                                            1
```

Following the derivation of DTHCAUS and related traceability variables, it is then possible to derive grouping variables such as death categories (DTHCGRx) using standard tidyverse code.

```
adsl <- adsl %>%
 mutate(DTHCGR1 = case_when(
   is.na(DTHDOM) ~ NA_character_,
   DTHDOM == "AE" ~ "ADVERSE EVENT",
   str_detect(DTHCAUS, "(PROGRESSIVE DISEASE|DISEASE RELAPSE)") ~ "PROGRESSIVE DISEASE",
   TRUE ~ "OTHER"
 )) # dim(adsl) 306 43
adsl %>% filter(!is.na(DTHDT)) %>% select(USUBJID, DTHCAUS, DTHDOM, DTHCGR1) %>% head()
# A tibble: 3 x 4
 USUBJID
             DTHCAUS
                                    DTHDOM DTHCGR1
  <chr>
              <chr>
                                    <chr> <chr>
1 01-701-1211 SUDDEN DEATH
                                    ΑE
                                           ADVERSE EVENT
2 01-704-1445 COMPLETED SUICIDE
                                    ΑE
                                           ADVERSE EVENT
3 01-710-1083 MYOCARDIAL INFARCTION AE
                                           ADVERSE EVENT
```

Duration relative to death

DTHADY

- Relative Day of Death
- DTHADY=DTHDT-TRTSDT+1

The function derive_vars_duration() can be used to derive duration relative to death like the Relative Day of Death (DTHADY) or the numbers of days from last dose to death (LDDTHELD).

Example calls:

• Relative Day of Death

```
adsl <- adsl %>%
 derive_vars_duration(
   new_var = DTHADY,
   start_date = TRTSDT,
   end_date = DTHDT
 ) # dim(adsl) 306 44
adsl %>% filter(!is.na(DTHDT)) %>% select(USUBJID, TRTSDT, DTHDT, DTHADY) %>% head()
# A tibble: 3 x 4
 USUBJID TRTSDT
                        DTHDT
                                   DTHADY
  <chr>
          <date>
                        <date>
                                     <dbl>
1 01-701-1211 2012-11-15 2013-01-14
                                       61
2 01-704-1445 2014-05-11 2014-11-01
                                      175
3 01-710-1083 2013-07-22 2013-08-02
                                        12
```

LDDTHELD

- Numbers of days from last dose to death
- LDDTHELD=DTHDT-TRTEDT

```
adsl <- adsl %>%
  derive_vars_duration(
    new_var = LDDTHELD,
    start_date = TRTEDT,
    end_date = DTHDT,
    add_one = FALSE
) # dim(adsl) 306 45

adsl %>% filter(!is.na(DTHDT)) %>% select(USUBJID, TRTEDT, DTHDT, LDDTHELD) %>% head()
```

```
# A tibble: 3 x 4
USUBJID TRTEDT DTHDT LDDTHELD
<chr> <chr> <date> <date> <date> <dbl>
1 01-701-1211 2013-01-12 2013-01-14 2
2 01-704-1445 2014-11-01 2014-11-01 0
3 01-710-1083 2013-08-01 2013-08-02 1
```

Derive the Last Date Known Alive

LSTALVDT

- Last Date Known Alive
- Similarly as for the cause of death (DTHCAUS), the last known alive date (LSTALVDT) can be derived from multiples sources using 'derive_vars_extreme_event()`.

Similarly as for the cause of death (DTHCAUS), the last known alive date (LSTALVDT) can be derived from multiples sources using derive vars extreme event().

An example could be (DTC dates are converted to numeric dates imputing missing day and month to the first):

```
adsl <- adsl %>%
  derive_vars_extreme_event(
   by_vars = exprs(STUDYID, USUBJID),
    events = list(
      event(
        dataset name = "ae",
        order = exprs(AESTDTC, AESEQ),
        condition = !is.na(AESTDTC),
        set_values_to = exprs(
          LSTALVDT = convert_dtc_to_dt(AESTDTC, highest_imputation = "M"),
          seq = AESEQ
        ),
      ),
      event(
        dataset_name = "ae",
        order = exprs(AEENDTC, AESEQ),
        condition = !is.na(AEENDTC),
        set_values_to = exprs(
         LSTALVDT = convert_dtc_to_dt(AEENDTC, highest_imputation = "M"),
          seq = AESEQ
       ),
      ),
      event(
       dataset_name = "lb",
        order = exprs(LBDTC, LBSEQ),
```

```
condition = !is.na(LBDTC),
        set_values_to = exprs(
         LSTALVDT = convert dtc to dt(LBDTC, highest imputation = "M"),
          seq = LBSEQ
       ),
      ),
      event(
       dataset_name = "adsl",
       condition = !is.na(TRTEDT),
        set_values_to = exprs(LSTALVDT = TRTEDT, seq = 0),
      )
    ),
    source_datasets = list(ae = ae, lb = lb, adsl = adsl),
    tmp_event_nr_var = event_nr,
    order = exprs(LSTALVDT, seq, event_nr),
   mode = "last",
   new_vars = exprs(LSTALVDT)
  ) # dim(adsl) 306 46
adsl %>% select(USUBJID, TRTEDT, DTHDT, LSTALVDT) %>% head()
```

```
# A tibble: 6 x 4
 USUBJID TRTEDT
                      DTHDT LSTALVDT
       <date> <date> <date>
 <chr>
1 01-701-1015 2014-07-02 NA 2014-07-02
2 01-701-1023 2012-09-01 NA
                             2012-09-02
3 01-701-1028 2014-01-14 NA
                            2014-01-14
4 01-701-1033 2014-03-31 NA
                            2014-04-14
5 01-701-1034 2014-12-30 NA
                            2014-12-30
6 01-701-1047 2013-03-09 NA
                             2013-04-07
```

Traceability variables can be added by specifying the variables in the set_values_to parameter of the event() function.

```
LALVDOM = "AE",
      LALVVAR = "AESTDTC"
    ),
  ),
  event(
    dataset_name = "ae",
    order = exprs(AEENDTC, AESEQ),
    condition = !is.na(AEENDTC),
    set_values_to = exprs(
      LSTALVDT = convert_dtc_to_dt(AEENDTC, highest_imputation = "M"),
      LALVSEQ = AESEQ,
      LALVDOM = "AE",
     LALVVAR = "AEENDTC"
    ),
  ),
  event(
    dataset_name = "lb",
    order = exprs(LBDTC, LBSEQ),
    condition = !is.na(LBDTC),
    set_values_to = exprs(
      LSTALVDT = convert_dtc_to_dt(LBDTC, highest_imputation = "M"),
      LALVSEQ = LBSEQ,
     LALVDOM = "LB",
      LALVVAR = "LBDTC"
    ),
  ),
  event(
    dataset_name = "adsl",
    condition = !is.na(TRTEDT),
    set_values_to = exprs(LSTALVDT = TRTEDT, LALVSEQ = NA_integer_, LALVDOM = "ADSL", LALV
  )
),
source_datasets = list(ae = ae, lb = lb, adsl = adsl),
tmp_event_nr_var = event_nr,
order = exprs(LSTALVDT, LALVSEQ, event_nr),
mode = "last",
new_vars = exprs(LSTALVDT, LALVSEQ, LALVDOM, LALVVAR)
```

Derive grouping, population variables

AGEGRy

• Pooled Age Group y

• Character description of a grouping or pooling of the subject's age for analysis purposes. For example, AGEGR1 might have values of "<18", "18-65", and ">65"; AGEGR2 might have values of "Less than 35 y old" and "At least 35 y old".

REGIONy

- Geographic Region y
- Character description of geographical region. For example, REGION1 might have values of "Asia",
 "Europe", "North America", "Rest of World"; REGION2 might have values of "United States", "Rest
 of World".

Grouping (e.g., AGEGR1 or REGION1)

Numeric and categorical variables (AGE, RACE, COUNTRY, etc.) may need to be grouped to perform the required analysis. {admiral} does not **currently** have functionality to assist with all required groupings. So, the user will often need to create his/her own function to meet his/her study requirement.

For example, if

- AGEGR1 is required to categorize AGE into <18, 18-64 and >64, or
- REGION1 is required to categorize COUNTRY in North America, Rest of the World,

the user defined functions would look like the following:

```
format_agegr1 <- function(var_input) {
   case_when(
    var_input < 18 ~ "<18",
    between(var_input, 18, 64) ~ "18-64",
    var_input > 64 ~ ">64 ~ ">64",
    TRUE ~ "Missing"
   )
}

format_region1 <- function(var_input) {
   case_when(
    var_input %in% c("CAN", "USA") ~ "North America",
    !is.na(var_input) ~ "Rest of the World",
    TRUE ~ "Missing"
   )
}</pre>
```

These functions are then used in a mutate() statement to derive the required grouping variables:

```
adsl <- adsl %>%
  mutate(
    AGEGR1 = format_agegr1(AGE),
    REGION1 = format_region1(COUNTRY)
) # dim(adsl) 306 48
```

```
adsl %>% select(USUBJID, AGE, COUNTRY, AGEGR1, REGION1) %>% head()
```

```
# A tibble: 6 x 5
 USUBJID
                AGE COUNTRY AGEGR1 REGION1
  <chr>
                            <chr>
                                   <chr>
              <dbl> <chr>
1 01-701-1015
                 63 USA
                            18-64 North America
2 01-701-1023
                 64 USA
                            18-64 North America
3 01-701-1028
                 71 USA
                            >64
                                   North America
                 74 USA
4 01-701-1033
                            >64
                                   North America
5 01-701-1034
                 77 USA
                            >64
                                   North America
6 01-701-1047
                 85 USA
                            >64
                                   North America
```

Population flags

SAFFL

- Safety Population Flag
- These flags identify whether or not the subject is included in the specified population. A minimum of one subject-level population flag variable is required in ADSL. Not all of the indicators listed here need to be included in ADSL. As stated in Section 3.1.4, Item 2, only those indicators corresponding to populations defined in the statistical analysis plan or populations used as a basis for analysis need be included in ADSL. This list of flags is not meant to be all-inclusive. Additional population flags may be added. The values of subject-level population flags cannot be blank. If a flag is used, the corresponding numeric version (*FN, where 0=no and 1=yes) of the population flag can also be included. Please also refer to Section 3.1.4.
- Since the populations flags are mainly company/study specific no dedicated functions are provided, but in most cases they can easily be derived using derive_var_merged_exist_flag.

Since the populations flags are mainly company/study specific no dedicated functions are provided, but in most cases they can easily be derived using derive_var_merged_exist_flag.

An example of an implementation could be:

```
adsl <- adsl %>%
  derive_var_merged_exist_flag(
    dataset_add = ex,
    by_vars = exprs(STUDYID, USUBJID),
    new_var = SAFFL,
    condition = (EXDOSE > 0 | (EXDOSE == 0 & str_detect(EXTRT, "PLACEBO")))
) # dim(adsl) 306 49

adsl %>% select(USUBJID, ARM, ACTARM, SAFFL) %>% head()
```

```
# A tibble: 6 x 4
 USUBJID ARM
                                   ACTARM
                                                        SAFFL
             <chr>
                                   <chr>
                                                        <chr>
  <chr>
1 01-701-1015 Placebo
                                   Placebo
                                                        Y
2 01-701-1023 Placebo
                                   Placebo
                                                        Y
3 01-701-1028 Xanomeline High Dose Xanomeline High Dose Y
4 01-701-1033 Xanomeline Low Dose Xanomeline Low Dose Y
5 01-701-1034 Xanomeline High Dose Xanomeline High Dose Y
6 01-701-1047 Placebo
                                   Placebo
```

Derive Other Variables

The users can add specific code to cover their need for the analysis.

The following functions are helpful for many ADSL derivations:

- derive_vars_merged() Merge Variables from a Dataset to the Input Dataset
- derive_var_merged_exist_flag() Merge an Existence Flag
- derive_var_merged_summary() Merge Summary Variables

See also Generic Functions.

Add Labels and Attributes

Adding labels and attributes for SAS transport files is supported by the following packages:

- metacore: establish a common foundation for the use of metadata within an R session.
- metatools: enable the use of metacore objects. Metatools can be used to build datasets or enhance columns in existing datasets as well as checking datasets against the metadata.
- xportr: functionality to associate all metadata information to a local R data frame, perform data set level validation checks and convert into a transport v5 file(xpt).

NOTE: All these packages are in the experimental phase, but the vision is to have them associated with an End to End pipeline under the umbrella of the pharmaverse. An example of applying metadata and perform associated checks can be found at the pharmaverse E2E example.

References

Creating ADSL

ADaM Subject-level Analysis - ADSL Dataset

ADaMIG V1.3