

Creating a Pharmacokinetics (PK) Non-compartmental analysis (NCA) ADaM (ADNCA/ADPC) or a Population PK ADaM (ADPPK)

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Programming workflow

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Introduction

This article describes creating a Pharmacokinetics (PK) Non-compartmental analysis (NCA) ADaM (ADNCA/ADPC) or a Population PK ADaM (ADPPK). The first part of the article describes the NCA file creation while the second part describes Population PK. This initial steps for both files are very similar and could be combined in one script if desired.

Programming PK NCA (ADPC/ADNCA) Analysis Data

The Non-compartmental analysis (NCA) ADaM uses the [CDISC Implementation Guide](<https://www.cdisc.org/standards/foundational/adam/adamig-non-compartmental-analysis-input-data-v1-0>). This example presented uses underlying EX and PC domains where the EX and PC domains represent data as collected and the ADPC ADaM is output. However, the example can be applied to situations where an EC domain is used as input instead of EX and/or ADNCA or another ADaM is created.

One of the important aspects of the dataset is the derivation of relative timing variables. These variables consist of nominal and actual times, and refer to the time from first dose or time from most recent reference dose. The reference dose for pre-dose records may be the upcoming dose. The CDISC Implementation Guide makes use of duplicated records for analysis, which allows the same record to be used both with respect to the previous dose and the next upcoming dose. This is illustrated later in this vignette.

Here are the relative time variables we will use. These correspond to the names in the CDISC Implementation Guide.

Variable	Variable Label	CDISC Notes
NFRLT	Nominal Relative Time from Analyte First Dose	This is the planned elapsed time (for sample point or start of sampling interval) from first exposure to treatment associated with PARAM and ANALYTE. For studies with an extended duration infusion, use the define to document if this is from the start or end of the infusion.
AFRLT	Actual Relative Time from Analyte First Dose	This is the actual elapsed time (for sample point or start of sampling interval) from first exposure to treatment associated with PARAM and ANALYTE. Note that this is referring to the first dose available for a particular drug. It is useful in multiple-dosing situations
NRRLT	Nominal Relative Time from Ref. Dose	This is the planned elapsed time (for sample point or start of sampling interval) from reference exposure to study treatment

Variable	Variable Label	CDISC Notes
ARRLT	Actual Relative Time from Ref. Dose	This is the actual elapsed time (for sample point or start of sampling interval) from reference exposure to study treatment.
MRRLT	Modified Relative Time from Ref. Dose	This variable could be used to modify the ARRLT variable based on analysis needs (e.g., setting negative values to zero or having a mix of nominal and actual time based of TMPCTDF).

Note: All examples assume CDISC SDTM and/or ADaM format as input unless otherwise specified.

Read in Data

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

Additional domains such as VS and LB may be used for additional baseline variables if needed. These may come from either the SDTM or ADaM source.

For the purpose of example, the CDISC Pilot SDTM and ADaM 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':
```

```
date, intersect, setdiff, union
```

At this step, it may be useful to join ADSL to your PC and EX domains as well. Only the ADSL variables used for derivations are selected at this step. The rest of the relevant ADSL variables will be added later.

In this case we will keep TRTSDT/TRTSDTM for day derivation and TRT01P/TRT01A for planned and actual treatments.

In this segment we will use `derive_vars_merged()` to join the ADSL variables and the following {admiral} functions to derive analysis dates, times and days: `derive_vars_dtm()`, `derive_vars_dtm_to_dt()`,

derive_vars_dtm_to_tm(), derive_vars_dy(). We will also create NFRLT for PC data based on PCTPTNUM. We will create an event ID (EVID) of 0 for concentration records and 1 for dosing records. This is a traditional variable that will provide a handy tool to identify records but will be dropped from the final dataset in this example.

```
adsl_vars <- exprs(TRTSDT, TRTSDTM, TRT01P, TRT01A)

pc_dates <- pc %>%
  # Join ADSL with PC (need TRTSDT for ADY derivation)
  derive_vars_merged(
    dataset_add = adsl,
    new_vars = adsl_vars,
    by_vars = exprs(STUDYID, USUBJID)
  ) %>%
  # Derive analysis date/time
  # Impute missing time to 00:00:00
  derive_vars_dtm(
    new_vars_prefix = "A",
    dtc = PCDTC,
    time_imputation = "00:00:00"
  ) %>%
  # Derive dates and times from date/times
  derive_vars_dtm_to_dt(exprs(ADTM)) %>%
  derive_vars_dtm_to_tm(exprs(ADTM)) %>%
  derive_vars_dy(reference_date = TRTSDT, source_vars = exprs(ADT)) %>%
  # Derive event ID and nominal relative time from first dose (NFRLT)
  mutate(
    EVID = 0,
    DRUG = PCTEST,
    NFRLT = if_else(PCTPTNUM < 0, 0, PCTPTNUM), .after = USUBJID
  ) # dim(pc_dates) [1] 4572 32

pc_dates %>% select(USUBJID,PCTEST,ADTM,VISIT,PCTPT,NFRLT) %>% head(n=10)
```

A tibble: 10 x 6

	USUBJID	PCTEST	ADTM	VISIT	PCTPT	NFRLT
	<chr>	<chr>	<dtm>	<chr>	<chr>	<dbl>
1	01-701-1015	XANOMELINE	2014-01-01 23:30:00	BASELINE	Pre-dose	0
2	01-701-1015	XANOMELINE	2014-01-02 00:05:00	BASELINE	5 Min Post-dose	0.08
3	01-701-1015	XANOMELINE	2014-01-02 00:30:00	BASELINE	30 Min Post-dose	0.5
4	01-701-1015	XANOMELINE	2014-01-02 01:00:00	BASELINE	1h Post-dose	1
5	01-701-1015	XANOMELINE	2014-01-02 01:30:00	BASELINE	1.5h Post-dose	1.5
6	01-701-1015	XANOMELINE	2014-01-02 02:00:00	BASELINE	2h Post-dose	2
7	01-701-1015	XANOMELINE	2014-01-02 04:00:00	BASELINE	4h Post-dose	4
8	01-701-1015	XANOMELINE	2014-01-02 06:00:00	BASELINE	6h Post-dose	6
9	01-701-1015	XANOMELINE	2014-01-02 08:00:00	BASELINE	8h Post-dose	8
10	01-701-1015	XANOMELINE	2014-01-02 12:00:00	BASELINE	12h Post-dose	12

Next we will also join ADSL data with EX and derive dates/times. This section uses the {admiral} functions `derive_vars_merged()`, `derive_vars_dtm()`, and `derive_vars_dtm_to_dt()`. Time is imputed to 00:00:00 here for reasons specific to the sample data. Other imputation times may be used based on study details. Here we create NFRLT for EX data based on VISITDY using `dplyr::mutate()`.

```
# ---- Get dosing information ----

ex_dates <- ex %>%
  derive_vars_merged(
    dataset_add = adsl,
    new_vars = adsl_vars,
    by_vars = exprs(STUDYID, USUBJID)
  ) %>%
  # Keep records with nonzero dose
  filter(EXDOSE > 0) %>%
  # Add time and set missing end date to start date
  # Impute missing time to 00:00:00
  # Note all times are missing for dosing records in this example data
  # Derive Analysis Start and End Dates
  derive_vars_dtm(
    new_vars_prefix = "AST",
    dtc = EXSTDTC,
    time_imputation = "00:00:00"
  ) %>%
  derive_vars_dtm(
    new_vars_prefix = "AEN",
    dtc = EXENDTC,
    time_imputation = "00:00:00"
  ) %>%
  # Derive event ID and nominal relative time from first dose (NFRLT)
  mutate(
    EVID = 1,
    NFRLT = case_when(
      VISITDY == 1 ~ 0,
      TRUE ~ 24 * VISITDY
    )
  ) %>%
  # Set missing end dates to start date
  mutate(AENDTM = case_when(
    is.na(AENDTM) ~ ASTDTM,
    TRUE ~ AENDTM
  )) %>%
  # Derive dates from date/times
  derive_vars_dtm_to_dt(exprs(ASTDTM)) %>%
  derive_vars_dtm_to_dt(exprs(AENDTM)) # dim(ex_dates) [1] 365 29

ex_dates %>% select(USUBJID, EXTRT, EXDOSFRQ, ASTDTM, AENDTM, VISIT, VISITDY, NFRLT) %>% head(n=10)
```

```
# A tibble: 10 x 8
  USUBJID EXTRT EXDOSFRQ ASTDTM          AENDTM          VISIT VISITDY
  <chr>    <chr> <chr>    <dtm>          <dtm>          <chr>    <dbl>
1 01-701-- XANO~ QD      2013-07-19 00:00:00 2013-08-01 00:00:00 BASE~      1
2 01-701-- XANO~ QD      2013-08-02 00:00:00 2014-01-06 00:00:00 WEEK~     14
3 01-701-- XANO~ QD      2014-01-07 00:00:00 2014-01-14 00:00:00 WEEK~    168
4 01-701-- XANO~ QD      2014-03-18 00:00:00 2014-03-31 00:00:00 BASE~      1
5 01-701-- XANO~ QD      2014-07-01 00:00:00 2014-07-15 00:00:00 BASE~      1
6 01-701-- XANO~ QD      2014-07-16 00:00:00 2014-12-17 00:00:00 WEEK~     14
7 01-701-- XANO~ QD      2014-12-18 00:00:00 2014-12-30 00:00:00 WEEK~    168
8 01-701-- XANO~ QD      2014-01-01 00:00:00 2014-01-15 00:00:00 BASE~      1
9 01-701-- XANO~ QD      2014-01-16 00:00:00 2014-06-18 00:00:00 WEEK~     14
10 01-701-- XANO~ QD      2014-06-19 00:00:00 2014-07-09 00:00:00 WEEK~    168
# i 1 more variable: NFRLT <dbl>
```

Expand Dosing Records

The function `create_single_dose_dataset()` can be used to expand dosing records between the start date and end date. The nominal time will also be expanded based on the values of `EXDOSFRQ`, for example “QD” will result in nominal time being incremented by 24 hours and “BID” will result in nominal time being incremented by 12 hours. This is a new feature of `create_single_dose_dataset()`. QD: quanque die, once a day. BID: bis in die, twice a day.

Dates and times will be derived after expansion using `derive_vars_dtm_to_dt()` and `derive_vars_dtm_to_tm()`.

For this example study we will define analysis visit (AVISIT) based on the nominal day value from `NFRLT` and give it the format, “Day 1”, “Day 2”, “Day 3”, etc. This is important for creating the `BASETYPE` variable later. `DRUG` is created from `EXTRT` here. This will be useful for linking treatment data with concentration data if there are multiple drugs and/or analytes, but this variable will also be dropped from the final dataset in this example.

```
# ---- Expand dosing records between start and end dates ----

ex_exp <- ex_dates %>%
  create_single_dose_dataset(
    dose_freq = EXDOSFRQ,
    start_date = ASTDT,
    start_datetime = ASTDTM,
    end_date = AENDT,
    end_datetime = AENDTM,
    nominal_time = NFRLT,
    lookup_table = dose_freq_lookup,
    lookup_column = CDISC_VALUE,
    keep_source_vars = exprs(
      STUDYID, USUBJID, EVID, EXDOSFRQ, EXDOSFRM,
      NFRLT, EXDOSE, EXDOSU, EXTRT, ASTDT, ASTDTM, AENDT, AENDTM,
      VISIT, VISITNUM, VISITDY, TRT01A, TRT01P, DOMAIN, EXSEQ, !!!adsl_vars
    )
  )
```

```

) %>%
# Derive AVISIT based on nominal relative time
# Derive AVISITN to nominal time in whole days using integer division
# Define AVISIT based on nominal day
mutate(
  AVISITN = NFRLT %/% 24 + 1,
  AVISIT = paste("Day", AVISITN),
  ADTM = ASTDTM,
  DRUG = EXTRT,
) %>%
# Derive dates and times from datetimes
derive_vars_dtm_to_dt(exprs(ADTM)) %>%
derive_vars_dtm_to_tm(exprs(ADTM)) %>%
derive_vars_dtm_to_tm(exprs(ASTDTM)) %>%
derive_vars_dtm_to_tm(exprs(AENDTM)) %>%
derive_vars_dy(reference_date = TRTSDT, source_vars = exprs(ADT)) # dim(ex_dates) [1] 365 29

ex_exp %>% select(USUBJID, DRUG, EXDOSFRQ, ASTDTM, AENDTM, AVISIT, NFRLT) %>% head(n=10)

```

A tibble: 10 x 7

	USUBJID	DRUG	EXDOSFRQ	ASTDTM	AENDTM	AVISIT	NFRLT
	<chr>	<chr>	<chr>	<dtm>	<dtm>	<chr>	<dbl>
1	01-701-1~	XANO~	ONCE	2013-07-19 00:00:00	2013-07-19 00:00:00	Day 1	0
2	01-701-1~	XANO~	ONCE	2013-07-20 00:00:00	2013-07-20 00:00:00	Day 2	24
3	01-701-1~	XANO~	ONCE	2013-07-21 00:00:00	2013-07-21 00:00:00	Day 3	48
4	01-701-1~	XANO~	ONCE	2013-07-22 00:00:00	2013-07-22 00:00:00	Day 4	72
5	01-701-1~	XANO~	ONCE	2013-07-23 00:00:00	2013-07-23 00:00:00	Day 5	96
6	01-701-1~	XANO~	ONCE	2013-07-24 00:00:00	2013-07-24 00:00:00	Day 6	120
7	01-701-1~	XANO~	ONCE	2013-07-25 00:00:00	2013-07-25 00:00:00	Day 7	144
8	01-701-1~	XANO~	ONCE	2013-07-26 00:00:00	2013-07-26 00:00:00	Day 8	168
9	01-701-1~	XANO~	ONCE	2013-07-27 00:00:00	2013-07-27 00:00:00	Day 9	192
10	01-701-1~	XANO~	ONCE	2013-07-28 00:00:00	2013-07-28 00:00:00	Day 10	216

Find First Dose

FANLDTM

- First Datetime of Dose for Analyte
- Date and time of first exposure to treatment associated with PARAM and ANALYTE for a subject in a study where multiple doses have been given. If treatment is given over a duration multiple times, this variable will reflect the start date and time of the first dose.

In this section we will find the first dose for each subject and drug, using `derive_vars_merged()`. We also create an analysis visit (AVISIT) based on NFRLT. The first dose datetime for an analyte FANLDTM is calculated as the minimum ADTM from the dosing records by subject and drug.

```

# ---- Find first dose per treatment per subject ----
# ---- Join with ADPC data and keep only subjects with dosing ----

adpc_first_dose <- pc_dates %>%
  derive_vars_merged(
    dataset_add = ex_exp,
    filter_add = (EXDOSE > 0 & !is.na(ADTM)),
    new_vars = exprs(FANLDTM = ADTM),
    order = exprs(ADTM, EXSEQ),
    mode = "first",
    by_vars = exprs(STUDYID, USUBJID, DRUG)
  ) %>%
  filter(!is.na(FANLDTM)) %>%
  # Derive AVISIT based on nominal relative time
  # Derive AVISITN to nominal time in whole days using integer division
  # Define AVISIT based on nominal day
  mutate(
    AVISITN = NFRLT %/% 24 + 1,
    AVISIT = paste("Day", AVISITN)
  ) # dim(pc_dates) [1] 4572 32 # dim(ex_exp) [1] 16331 31 # dim(adpc_first_dose) [1] 302

adpc_first_dose %>% select(USUBJID,FANLDTM,AVISIT,ADTM,PCTPT) %>% head(n=10)

```

```

# A tibble: 10 x 5
  USUBJID      FANLDTM      AVISIT ADTM      PCTPT
  <chr>      <dtm>      <chr> <dtm>      <chr>
1 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-18 23:30:00 Pre-dose
2 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 00:05:00 5 Min Post-dose
3 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 00:30:00 30 Min Post-dose
4 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 01:00:00 1h Post-dose
5 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 01:30:00 1.5h Post-dose
6 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 02:00:00 2h Post-dose
7 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 04:00:00 4h Post-dose
8 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 06:00:00 6h Post-dose
9 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 08:00:00 8h Post-dose
10 01-701-1028 2013-07-19 00:00:00 Day 1 2013-07-19 12:00:00 12h Post-dose

```

Find Reference Dose Dates Corresponding to PK Records

Use `derive_vars_joined()` to find the previous dose data. This will join the expanded EX data with the ADPC based on the analysis date ADTM. Note the `filter_join` parameter. In addition to the date of the previous dose (`ADTM_prev`), we also keep the actual dose amount `EXDOSE_prev` and the analysis visit of the dose `AVISIT_prev`.


```
# ---- Find previous dose ----

adpc_prev <- adpc_first_dose %>%
  derive_vars_joined(
    dataset_add = ex_exp,
    by_vars = exprs(USUBJID),
    order = exprs(ADTM),
    new_vars = exprs(
      ADTM_prev = ADTM, EXDOSE_prev = EXDOSE, AVISIT_prev = AVISIT,
      AENDTM_prev = AENDTM
    ),
    join_vars = exprs(ADTM),
    join_type = "all",
    filter_add = NULL,
    filter_join = ADTM > ADTM.join,
    mode = "last",
    check_type = "none"
  ) # dim(adpc_prev) [1] 3024 39 # dim(adpc_first_dose) [1] 3024 35

adpc_prev %>% select(USUBJID,VISIT,ADTM,PCTPT,ADTM_prev,EXDOSE_prev,AVISIT_prev) %>% head(n=10)
```

```
# A tibble: 10 x 7
  USUBJID VISIT ADTM PCTPT ADTM_prev EXDOSE_prev
  <chr> <chr> <dtm> <chr> <dtm> <dbl>
1 01-701-1028 BASELI~ 2013-07-18 23:30:00 Pre~ NA NA
2 01-701-1028 BASELI~ 2013-07-19 00:05:00 5 Mi~ 2013-07-19 00:00:00 54
3 01-701-1028 BASELI~ 2013-07-19 00:30:00 30 M~ 2013-07-19 00:00:00 54
4 01-701-1028 BASELI~ 2013-07-19 01:00:00 1h P~ 2013-07-19 00:00:00 54
5 01-701-1028 BASELI~ 2013-07-19 01:30:00 1.5h~ 2013-07-19 00:00:00 54
6 01-701-1028 BASELI~ 2013-07-19 02:00:00 2h P~ 2013-07-19 00:00:00 54
7 01-701-1028 BASELI~ 2013-07-19 04:00:00 4h P~ 2013-07-19 00:00:00 54
8 01-701-1028 BASELI~ 2013-07-19 06:00:00 6h P~ 2013-07-19 00:00:00 54
9 01-701-1028 BASELI~ 2013-07-19 08:00:00 8h P~ 2013-07-19 00:00:00 54
10 01-701-1028 BASELI~ 2013-07-19 12:00:00 12h ~ 2013-07-19 00:00:00 54
# i 1 more variable: AVISIT_prev <chr>
```

Similarly, find next dose information using `derive_vars_joined()` with the `filter_join` parameter as `ADTM <= ADTM.join`. Here we keep the next dose analysis date `ADTM_next`, the next actual dose `EXDOSE_next`, and the next analysis visit `AVISIT_next`.

```
# ---- Find next dose ----

adpc_next <- adpc_prev %>%
  derive_vars_joined(
    dataset_add = ex_exp,
    by_vars = exprs(USUBJID),
```

```

order = exprs(ADTM),
new_vars = exprs(
  ADTM_next = ADTM, EXDOSE_next = EXDOSE, AVISIT_next = AVISIT,
  AENDTM_next = AENDTM
),
join_vars = exprs(ADTM),
join_type = "all",
filter_add = NULL,
filter_join = ADTM <= ADTM.join,
mode = "first",
check_type = "none" ) # dim(adpc_next) 3024 43

adpc_next %>% select(USUBJID,VISIT,ADTM,PCTPT,ADTM_next,EXDOSE_next,AVISIT_next) %>% head(n=10)

```

```

# A tibble: 10 x 7
  USUBJID VISIT ADTM PCTPT ADTM_next EXDOSE_next
  <chr>    <chr> <dtm>    <chr> <dtm>    <dbl>
1 01-701-1028 BASELI~ 2013-07-18 23:30:00 Pre~ 2013-07-19 00:00:00 54
2 01-701-1028 BASELI~ 2013-07-19 00:05:00 5 Mi~ 2013-07-20 00:00:00 54
3 01-701-1028 BASELI~ 2013-07-19 00:30:00 30 M~ 2013-07-20 00:00:00 54
4 01-701-1028 BASELI~ 2013-07-19 01:00:00 1h P~ 2013-07-20 00:00:00 54
5 01-701-1028 BASELI~ 2013-07-19 01:30:00 1.5h~ 2013-07-20 00:00:00 54
6 01-701-1028 BASELI~ 2013-07-19 02:00:00 2h P~ 2013-07-20 00:00:00 54
7 01-701-1028 BASELI~ 2013-07-19 04:00:00 4h P~ 2013-07-20 00:00:00 54
8 01-701-1028 BASELI~ 2013-07-19 06:00:00 6h P~ 2013-07-20 00:00:00 54
9 01-701-1028 BASELI~ 2013-07-19 08:00:00 8h P~ 2013-07-20 00:00:00 54
10 01-701-1028 BASELI~ 2013-07-19 12:00:00 12h ~ 2013-07-20 00:00:00 54
# i 1 more variable: AVISIT_next <chr>

```

Use the same method to find the previous and next nominal times. Note that here the data are sorted by nominal time rather than the actual time. This will tell us when the previous dose and the next dose were supposed to occur. Sometimes this will differ from the actual times in a study. Here we keep the previous nominal dose time NFRLT_prev and the next nominal dose time NFRLT_next. Note that the filter_join parameter uses the nominal relative times, e.g. NFRLT > NFRLT.join.

```

# ---- Find previous nominal time ----

adpc_nom_prev <- adpc_next %>%
  derive_vars_joined(
    dataset_add = ex_exp,
    by_vars = exprs(USUBJID),
    order = exprs(NFRLT),
    new_vars = exprs(NFRLT_prev = NFRLT),
    join_vars = exprs(NFRLT),
    join_type = "all",
    filter_add = NULL,

```

```

    filter_join = NFRLT > NFRLT.join,
    mode = "last",
    check_type = "none") # dim(adpc_nom_prev) [1] 3024    44

# ---- Find next nominal time ----

adpc_nom_next <- adpc_nom_prev %>%
  derive_vars_joined(
    dataset_add = ex_exp,
    by_vars = exprs(USUBJID),
    order = exprs(NFRLT),
    new_vars = exprs(NFRLT_next = NFRLT),
    join_vars = exprs(NFRLT),
    join_type = "all",
    filter_add = NULL,
    filter_join = NFRLT <= NFRLT.join,
    mode = "first",
    check_type = "none") # dim(adpc_nom_next) [1] 3024    45

adpc_nom_next %>% select(USUBJID,NFRLT,PCTPT,NFRLT_prev,NFRLT_next) %>% head(n=10)

# A tibble: 10 x 5
  USUBJID      NFRLT PCTPT      NFRLT_prev NFRLT_next
  <chr>      <dbl> <chr>      <dbl>      <dbl>
1 01-701-1028 0     Pre-dose      NA          0
2 01-701-1028 0.08 5 Min Post-dose 0          24
3 01-701-1028 0.5   30 Min Post-dose 0          24
4 01-701-1028 1     1h Post-dose 0          24
5 01-701-1028 1.5   1.5h Post-dose 0          24
6 01-701-1028 2     2h Post-dose 0          24
7 01-701-1028 4     4h Post-dose 0          24
8 01-701-1028 6     6h Post-dose 0          24
9 01-701-1028 8     8h Post-dose 0          24
10 01-701-1028 12    12h Post-dose 0          24

```

Combine PC and EX Records and Derive Relative Time Variables

Combine PC and EX records and derive the additional relative time variables. Often NCA data will keep both dosing and concentration records. We will keep both here. Sometimes you will see ADPC with only the concentration records. If this is desired, the dosing records can be dropped before saving the final dataset. We will use the {admiral} function `derive_vars_duration()` to calculate the actual relative time from first dose (AFRLT) and the actual relative time from most recent dose (ARRLT). Note that we use the parameter `add_one = FALSE` here. We will also create a variable representing actual time to next dose (AXRLT) which is not kept, but will be used when we create duplicated records for analysis for the pre-dose records. For now, we will update missing values of ARRLT corresponding to the pre-dose records with AXRLT, and dosing records will be set to zero.

We also calculate the reference dates FANLDTM (First Datetime of Dose for Analyte) and PCRFTDTM (Reference Datetime of Dose for Analyte) and their corresponding date and time variables.

We calculate the maximum date for concentration records and only keep the dosing records up to that date.

```
# ---- Combine ADPC and EX data ----
# Derive Relative Time Variables

adpc_arrlt <- bind_rows(adpc_nom_next, ex_exp) %>%
  group_by(USUBJID, DRUG) %>%
  mutate(
    FANLDTM = min(FANLDTM, na.rm = TRUE),
    min_NFRLT = min(NFRLT_prev, na.rm = TRUE),
    maxdate = max(ADT[EVID == 0], na.rm = TRUE), .after = USUBJID
  ) %>%
  arrange(USUBJID, ADTM) %>%
  ungroup() %>%
  filter(ADT <= maxdate) %>%
  # Derive Actual Relative Time from First Dose (AFRLT)
  derive_vars_duration(
    new_var = AFRLT,
    start_date = FANLDTM,
    end_date = ADTM,
    out_unit = "hours",
    floor_in = FALSE,
    add_one = FALSE
  ) %>%
  # Derive Actual Relative Time from Reference Dose (ARRLT)
  derive_vars_duration(
    new_var = ARRLT,
    start_date = ADTM_prev,
    end_date = ADTM,
    out_unit = "hours",
    floor_in = FALSE,
    add_one = FALSE
  ) %>%
  # Derive Actual Relative Time from Next Dose (AXRLT not kept)
  derive_vars_duration(
    new_var = AXRLT,
    start_date = ADTM_next,
    end_date = ADTM,
    out_unit = "hours",
    floor_in = FALSE,
    add_one = FALSE
  ) %>%
  mutate(
    ARRLT = case_when(
      EVID == 1 ~ 0,
```

```

    is.na(ARRLT) ~ AXRLT,
    TRUE ~ ARRLT
  ),

  # Derive Reference Dose Date
  PCRFTDTM = case_when(
    EVID == 1 ~ ADTM,
    is.na(ADTM_prev) ~ ADTM_next,
    TRUE ~ ADTM_prev
  )
) %>%
# Derive dates and times from datetimes
derive_vars_dtm_to_dt(exprs(FANLDTM)) %>%
derive_vars_dtm_to_tm(exprs(FANLDTM)) %>%
derive_vars_dtm_to_dt(exprs(PCRFTDTM)) %>%
derive_vars_dtm_to_tm(exprs(PCRFTDTM)) # dim(adpc_arrlt) 3522 68

adpc_arrlt %>% select(USUBJID,FANLDTM,AVISIT,PCTPT,AFRLT,ARRLT,AXRLT) %>% head(n=10)

```

```

# A tibble: 10 x 7
  USUBJID      FANLDTM      AVISIT PCTPT      AFRLT      ARRLT AXRLT
  <chr>      <dtm>      <chr> <chr>      <dbl>      <dbl> <dbl>
1 01-701-1028 2013-07-19 00:00:00 Day 1 Pre-dose      -0.5      -0.5     -0.5
2 01-701-1028 2013-07-19 00:00:00 Day 1 <NA>          0          0         NA
3 01-701-1028 2013-07-19 00:00:00 Day 1 5 Min Post-dose 0.0833    0.0833   -23.9
4 01-701-1028 2013-07-19 00:00:00 Day 1 30 Min Post-dose 0.5        0.5     -23.5
5 01-701-1028 2013-07-19 00:00:00 Day 1 1h Post-dose    1          1        -23
6 01-701-1028 2013-07-19 00:00:00 Day 1 1.5h Post-dose 1.5        1.5     -22.5
7 01-701-1028 2013-07-19 00:00:00 Day 1 2h Post-dose    2          2        -22
8 01-701-1028 2013-07-19 00:00:00 Day 1 4h Post-dose    4          4        -20
9 01-701-1028 2013-07-19 00:00:00 Day 1 6h Post-dose    6          6        -18
10 01-701-1028 2013-07-19 00:00:00 Day 1 0-6h Post-dose 6          6        -18

```

For nominal relative times we calculate NRRLT generally as $NFRLT - NFRLT_prev$ and NXRLT as $NFRLT - NFRLT_next$.

```

adpc_nrslt <- adpc_arrlt %>%
  # Derive Nominal Relative Time from Reference Dose (NRRLT)
  mutate(
    NRRLT = case_when(
      EVID == 1 ~ 0,
      is.na(NFRLT_prev) ~ NFRLT - min_NFRLT,
      TRUE ~ NFRLT - NFRLT_prev
    ),
    NXRLT = case_when(
      EVID == 1 ~ 0,

```

```

    TRUE ~ NFRLT - NFRLT_next
  )
) # dim(adpc_nrrlt) [1] 3522 70

adpc_nrrlt %>% select(USUBJID, AVISIT, PCTPT, NFRLT, NRRLT, NXRLT) %>% head(n=10)

```

```

# A tibble: 10 x 6
  USUBJID    AVISIT PCTPT      NFRLT NRRLT NXRLT
  <chr>      <chr> <chr>      <dbl> <dbl> <dbl>
1 01-701-1028 Day 1 Pre-dose      0      0      0
2 01-701-1028 Day 1 <NA>          0      0      0
3 01-701-1028 Day 1 5 Min Post-dose 0.08 0.08 -23.9
4 01-701-1028 Day 1 30 Min Post-dose 0.5 0.5 -23.5
5 01-701-1028 Day 1 1h Post-dose 1 1 -23
6 01-701-1028 Day 1 1.5h Post-dose 1.5 1.5 -22.5
7 01-701-1028 Day 1 2h Post-dose 2 2 -22
8 01-701-1028 Day 1 4h Post-dose 4 4 -20
9 01-701-1028 Day 1 6h Post-dose 6 6 -18
10 01-701-1028 Day 1 0-6h Post-dose 3 3 -21

```

Derive Analysis Variables

Using `dplyr::mutate` we derive a number of analysis variables including analysis value (AVAL), analysis time point (ATPT) analysis timepoint reference (ATPTREF) and baseline type (BASETYPE).

We set ATPT to PCTPT for concentration records and to “Dose” for dosing records. The analysis timepoint reference ATPTREF will correspond to the dosing visit. We will use AVISIT_prev and AVISIT_next to derive. The baseline type will be a concatenation of ATPTREF and “Baseline” with values such as “Day 1 Baseline”, “Day 2 Baseline”, etc. The baseline flag ABLFL will be set to “Y” for pre-dose records.

Analysis value AVAL in this example comes from PCSTRESN for concentration records. In addition we are including the dose value EXDOSE for dosing records and setting BLQ (Below Limit of Quantitation) records to 0 before the first dose and to 1/2 of LLOQ (Lower Limit of Quantitation) for records after first dose. (Additional tests such as whether more than 1/3 of records are BLQ may be required and are not done in this example.) We also create a listing-ready variable AVALCAT1 which includes the “BLQ” record indicator and formats the numeric values to three significant digits.

We derive actual dose DOSEA based on EXDOSE_prev and EXDOSE_next and planned dose DOSEP based on the planned treatment TRT01P. In addition we add the units for the dose variables and the relative time variables.

```

# ---- Derive Analysis Variables ----
# Derive ATPTN, ATPT, ATPTREF, ABLFL and BASETYPE
# Derive planned dose DOSEP, actual dose DOSEA and units
# Derive PARAMCD and relative time units
# Derive AVAL, AVALU and AVALCAT1

```

```

adpc_aval <- adpc_nrrlt %>%
  mutate(
    PARCAT1 = PCSPEC,
    ATPTN = case_when(
      EVID == 1 ~ 0,
      TRUE ~ PCTPTNUM
    ),
    ATPT = case_when(
      EVID == 1 ~ "Dose",
      TRUE ~ PCTPT
    ),
    ATPTREF = case_when(
      EVID == 1 ~ AVISIT,
      is.na(AVISIT_prev) ~ AVISIT_next,
      TRUE ~ AVISIT_prev
    ),
    # Derive baseline flag for pre-dose records
    ABLFL = case_when(
      ATPT == "Pre-dose" ~ "Y",
      TRUE ~ NA_character_
    ),
    # Derive BASETYPE
    BASETYPE = paste(ATPTREF, "Baseline"),

    # Derive Actual Dose
    DOSEA = case_when(
      EVID == 1 ~ EXDOSE,
      is.na(EXDOSE_prev) ~ EXDOSE_next,
      TRUE ~ EXDOSE_prev
    ),
    # Derive Planned Dose
    DOSEP = case_when(
      TRT01P == "Xanomeline High Dose" ~ 81,
      TRT01P == "Xanomeline Low Dose" ~ 54
    ),
    DOSEU = "mg",
  ) %>%
  # Derive relative time units
  mutate(
    FRLTU = "h",
    RRLTU = "h",
    # Derive PARAMCD
    PARAMCD = coalesce(PCTESTCD, "DOSE"),
    ALLOQ = PCLLOQ,
    # Derive AVAL
    AVAL = case_when(
      EVID == 1 ~ EXDOSE,

```

```

PCSTRESC == "<BLQ" & NFRLT == 0 ~ 0,
PCSTRESC == "<BLQ" & NFRLT > 0 ~ 0.5 * ALLOQ,
TRUE ~ PCSTRESN
),
AVALU = case_when(
  EVID == 1 ~ EXDOSU,
  TRUE ~ PCSTRESU
),
AVALCAT1 = if_else(PCSTRESC == "<BLQ", PCSTRESC, prettyNum(signif(AVAL, digits = 3))),
) %>%
# Add SRCSEQ
mutate(
  SRCDOM = DOMAIN,
  SRCVAR = "SEQ",
  SRCSEQ = coalesce(PCSEQ, EXSEQ)
) # dim(adpc_aval) [1] 3522 89

adpc_aval %>% select(USUBJID,NFRLT,AVISIT,ATPT,ABLFL,ATPTREF,AVAL,AVALCAT1) %>% head(n=10)

```

```

# A tibble: 10 x 8
  USUBJID      NFRLT AVISIT ATPT      ABLFL ATPTREF      AVAL AVALCAT1
  <chr>      <dbl> <chr> <chr>      <chr> <chr>      <dbl> <chr>
1 01-701-1028 0 Day 1 Pre-dose      Y Day 1 0 <BLQ
2 01-701-1028 0 Day 1 Dose      <NA> Day 1 54 <NA>
3 01-701-1028 0.08 Day 1 5 Min Post-dose <NA> Day 1 0.102 0.102
4 01-701-1028 0.5 Day 1 30 Min Post-dose <NA> Day 1 0.547 0.547
5 01-701-1028 1 Day 1 1h Post-dose <NA> Day 1 0.925 0.925
6 01-701-1028 1.5 Day 1 1.5h Post-dose <NA> Day 1 1.19 1.19
7 01-701-1028 2 Day 1 2h Post-dose <NA> Day 1 1.37 1.37
8 01-701-1028 4 Day 1 4h Post-dose <NA> Day 1 1.68 1.68
9 01-701-1028 6 Day 1 6h Post-dose <NA> Day 1 1.76 1.76
10 01-701-1028 3 Day 1 0-6h Post-dose <NA> Day 1 24.9 24.9

```

Create Duplicated Records for Analysis

DTYPE

- Derivation Type
- Analysis value derivation method. DTYPE is used to denote, and must be populated, when the value of AVAL or AVALC has been imputed or derived differently than the other analysis values within the parameter. DTYPE is required to be populated even if AVAL and AVALC are null on the derived record. \n Three common situations when DTYPE should be populated: \n * A new row is added within a parameter with the analysis value populated based on other rows within the parameter. \n * A new row is added within a parameter with the analysis value populated based on a constant value or data from other subjects. \n * An analysis value (AVAL or AVALC) on an existing record is being replaced

with a value based on a pre-specified algorithm. \n DTYPE is used to denote analysis values that are “special cases” within a parameter. For each value of DTYPE, the precise derivation algorithm must be defined in analysis variable metadata, even for DTYPE values in the CDISC Controlled Terminology. The controlled terminology for DTYPE is extensible. See Section 4, Implementation Issues, Standard Solutions, and Examples for examples of the use of DTYPE. \n Some examples of DTYPE values: \n * LOCF = last observation carried forward \n * WOCF = worst observation carried forward \n * AVERAGE = average of values

As mentioned above, the CDISC ADaM Implementation Guide for Non-compartmental Analysis uses duplicated records for analysis when a record needs to be used in more than one way. In this example the 24 hour post-dose record will also be used as the pre-dose record for the “Day 2” dose. In addition to 24 hour post-dose records, other situations may include pre-dose records for “Cycle 2 Day 1”, etc.

In general, we will select the records of interest and then update the relative time variables for the duplicated records. In this case we will select where the nominal relative time to next dose is zero. (Note that we do not need to duplicate the first dose record since there is no prior dose.)

DTYPE is set to “COPY” for the duplicated records and the original PCSEQ value is retained. In this case we change “24h Post-dose” to “Pre-dose”. ABLFL is set to “Y” since these records will serve as baseline for the “Day 2” dose. DOSEA is set to EXDOSE_next and PCRFTDTM is set to ADTM_next.

```
# ---- Create DTYPE copy records ----

dtype <- adpc_aval %>%
  filter(NFRLT > 0 & NXRLT == 0 & EVID == 0 & !is.na(AVISIT_next)) %>%
  select(-PCRFTDT, -PCRFTTM) %>%
  # Re-derive variables in for DTYPE copy records
  mutate(
    ABLFL = NA_character_,
    ATPTREF = AVISIT_next,
    ARRLT = AXRLT,
    NRRLT = NXRLT,
    PCRFTDTM = ADTM_next,
    DOSEA = EXDOSE_next,
    BASETYPE = paste(AVISIT_next, "Baseline"),
    ATPT = "Pre-dose",
    ATPTN = -0.5,
    ABLFL = "Y",
    DTYPE = "COPY"
  ) %>%
  derive_vars_dtm_to_dt(exprs(PCRFTDTM)) %>%
  derive_vars_dtm_to_tm(exprs(PCRFTDTM)) # dim(dtype) [1] 330 90 # dim(adpc_aval) [1] 3522 8

dtype %>% select(USUBJID, DTYPE, ATPT, NFRLT, NRRLT, AFRLT, ARRLT, BASETYPE) %>% head(n=10)
```

```
# A tibble: 10 x 8
```

USUBJID	DTYPE	ATPT	NFRLT	NRRLT	AFRLT	ARRLT	BASETYPE
<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<chr>

1	01-701-1028	COPY	Pre-dose	24	0	24	0	Day 2	Baseline
2	01-701-1028	COPY	Pre-dose	48	0	48	0	Day 3	Baseline
3	01-701-1033	COPY	Pre-dose	24	0	24	0	Day 2	Baseline
4	01-701-1033	COPY	Pre-dose	48	0	48	0	Day 3	Baseline
5	01-701-1034	COPY	Pre-dose	24	0	24	0	Day 2	Baseline
6	01-701-1034	COPY	Pre-dose	48	0	48	0	Day 3	Baseline
7	01-701-1097	COPY	Pre-dose	24	0	24	0	Day 2	Baseline
8	01-701-1097	COPY	Pre-dose	48	0	48	0	Day 3	Baseline
9	01-701-1111	COPY	Pre-dose	24	0	24	0	Day 2	Baseline
10	01-701-1111	COPY	Pre-dose	48	0	48	0	Day 3	Baseline

Combine ADPC data with Duplicated Records

Now the duplicated records are combined with the original records. We also derive the modified relative time from reference dose MRRLT. In this case, negative values of ARRLT are set to zero.

This is also an opportunity to derive analysis flags e.g. ANLO1FL, ANLO2FL etc. In this example ANLO1FL is set to “Y” for all records and ANLO2FL is set to “Y” for all records except the duplicated records with DTYPE = “COPY”. Additional flags may be used to select full profile records and/or to select records included in the tables and figures, etc.

```
# ---- Combine original records and DTYPE copy records ----
```

```
adpc_dtype <- bind_rows(adpc_aval, dtype) %>%
  arrange(STUDYID, USUBJID, BASETYPE, ADTM, NFRLT) %>%
  mutate(
    # Derive MRRLT, ANLO1FL and ANLO2FL
    MRRLT = if_else(ARRLT < 0, 0, ARRLT),
    ANLO1FL = "Y",
    ANLO2FL = if_else(is.na(DTYPE), "Y", NA_character_),
  ) # dim(adpc_dtype) [1] 3852 93
```

```
adpc_dtype %>% select(STUDYID, USUBJID, BASETYPE, ADTM, ATPT, NFRLT, NRRLT, ARRLT, MRRLT) %>% head(n=10)
```

```
# A tibble: 10 x 9
```

	STUDYID	USUBJID	BASETYPE	ADTM	ATPT	NFRLT	NRRLT	ARRLT	MRRLT
	<chr>	<chr>	<chr>	<dtm>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>
1	CDISCP~	01-701~	Day 1 B~	2013-07-18 23:30:00	Pre~	0	0	-0.5	0
2	CDISCP~	01-701~	Day 1 B~	2013-07-19 00:00:00	Dose	0	0	0	0
3	CDISCP~	01-701~	Day 1 B~	2013-07-19 00:05:00	5 Mi~	0.08	0.08	0.0833	0.0833
4	CDISCP~	01-701~	Day 1 B~	2013-07-19 00:30:00	30 M~	0.5	0.5	0.5	0.5
5	CDISCP~	01-701~	Day 1 B~	2013-07-19 01:00:00	1h P~	1	1	1	1
6	CDISCP~	01-701~	Day 1 B~	2013-07-19 01:30:00	1.5h~	1.5	1.5	1.5	1.5
7	CDISCP~	01-701~	Day 1 B~	2013-07-19 02:00:00	2h P~	2	2	2	2
8	CDISCP~	01-701~	Day 1 B~	2013-07-19 04:00:00	4h P~	4	4	4	4
9	CDISCP~	01-701~	Day 1 B~	2013-07-19 06:00:00	0-6h~	3	3	6	6
10	CDISCP~	01-701~	Day 1 B~	2013-07-19 06:00:00	6h P~	6	6	6	6

Calculate Change from Baseline and Assign ASEQ

The {admiral} function `derive_var_base()` is used to derive BASE and the function `derive_var_chg()` is used to derive change from baseline CHG.

We also now derive ASEQ using `derive_var_obs_number()` and we drop intermediate variables such as those ending with “_prev” and “_next”.

Finally we derive PARAM and PARAMN from a lookup table.

```
# ---- Derive BASE and Calculate Change from Baseline ----

adpc_base <- adpc_dtype %>%
  # Derive BASE
  derive_var_base(
    by_vars = exprs(STUDYID, USUBJID, PARAMCD, PARCAT1, BASETYPE),
    source_var = AVAL,
    new_var = BASE,
    filter = ABLFL == "Y"
  ) # dim(adpc_base) [1] 3852    94

adpc_chg <- derive_var_chg(adpc_base) # dim(adpc_chg) [1] 3852    95

# ---- Add ASEQ ----

adpc_aseq <- adpc_chg %>%
  # Calculate ASEQ
  derive_var_obs_number(
    new_var = ASEQ,
    by_vars = exprs(STUDYID, USUBJID),
    order = exprs(ADTM, BASETYPE, EVID, AVISITN, ATPTN, PARCAT1, DTYPE),
    check_type = "error"
  ) %>%
  # Remove temporary variables
  select(
    -DOMAIN, -PCSEQ, -starts_with("orig"), -starts_with("min"),
    -starts_with("max"), -starts_with("EX"), -ends_with("next"),
    -ends_with("prev"), -DRUG, -EVID, -AXRLT, -NXRLT, -VISITDY
  ) %>%
  # Derive PARAM and PARAMN
  derive_vars_merged(
    dataset_add = select(param_lookup, -PCTESTCD), by_vars = exprs(PARAMCD)
  ) # dim(adpc_aseq) [1] 3852    73

adpc_aseq %>% select(USUBJID, BASETYPE, DTYPE, AVISIT, ATPT, AVAL, NFRLT, NRRLT, AFRLT, ARRLT, BASE, CHG)

# A tibble: 10 x 12
  USUBJID  BASETYPE DTYPE AVISIT ATPT    AVAL NFRLT NRRLT  AFRLT  ARRLT  BASE
  <chr>    <chr>    <chr> <chr> <chr> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>
```

	<chr>	<chr>	<chr>	<chr>	<chr>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	<dbl>	
1	01-701--	Day 1	B~	<NA>	Day 1	Pre--	0	0	0	-0.5	-0.5	0
2	01-701--	Day 1	B~	<NA>	Day 1	Dose	54	0	0	0	0	NA
3	01-701--	Day 1	B~	<NA>	Day 1	5 Mi~	0.102	0.08	0.08	0.0833	0.0833	0
4	01-701--	Day 1	B~	<NA>	Day 1	30 M~	0.547	0.5	0.5	0.5	0.5	0
5	01-701--	Day 1	B~	<NA>	Day 1	1h P~	0.925	1	1	1	1	0
6	01-701--	Day 1	B~	<NA>	Day 1	1.5h~	1.19	1.5	1.5	1.5	1.5	0
7	01-701--	Day 1	B~	<NA>	Day 1	2h P~	1.37	2	2	2	2	0
8	01-701--	Day 1	B~	<NA>	Day 1	4h P~	1.68	4	4	4	4	0
9	01-701--	Day 1	B~	<NA>	Day 1	0-6h~	24.9	3	3	6	6	NA
10	01-701--	Day 1	B~	<NA>	Day 1	6h P~	1.76	6	6	6	6	0

i 1 more variable: CHG <dbl>

Add Additional Baseline Variables

Here we derive additional baseline values from VS for baseline height HTBL and weight WTBL and compute the body mass index (BMI) with `compute_bmi()`. These values could also be obtained from ADVS if available. Baseline lab values could also be derived from LB or ADLB in a similar manner.

```
# Derive additional baselines from VS
adpc_baselines <- adpc_aseq %>%
  derive_vars_merged(
    dataset_add = vs,
    filter_add = VSTESTCD == "HEIGHT",
    by_vars = exprs(STUDYID, USUBJID),
    new_vars = exprs(HTBL = VSSTRESN, HTBLU = VSSTRESU)
  ) %>%
  derive_vars_merged(
    dataset_add = vs,
    filter_add = VSTESTCD == "WEIGHT" & VSBLFL == "Y",
    by_vars = exprs(STUDYID, USUBJID),
    new_vars = exprs(WTBL = VSSTRESN, WTBLU = VSSTRESU)
  ) %>%
  mutate(
    BMIBL = compute_bmi(height = HTBL, weight = WTBL),
    BMIBLU = "kg/m^2" ) # dim(adpc_baselines) [1] 3852 79

adpc_baselines %>% select(USUBJID,HTBL,HTBLU,WTBL,WTBLU,BMIBL,BMIBLU,BASETYPE,ATPT,AVAL) %>% h
```

A tibble: 10 x 10

	USUBJID	HTBL	HTBLU	WTBL	WTBLU	BMIBL	BMIBLU	BASETYPE	ATPT	AVAL
	<chr>	<dbl>	<chr>	<dbl>	<chr>	<dbl>	<chr>	<chr>	<chr>	<dbl>
1	01-701-1028	178.	cm	99.3	kg	31.4	kg/m^2	Day 1 Baseline	Pre-d~	0
2	01-701-1028	178.	cm	99.3	kg	31.4	kg/m^2	Day 1 Baseline	Dose	54
3	01-701-1028	178.	cm	99.3	kg	31.4	kg/m^2	Day 1 Baseline	5 Min~	0.102
4	01-701-1028	178.	cm	99.3	kg	31.4	kg/m^2	Day 1 Baseline	30 Mi~	0.547

5	01-701-1028	178. cm	99.3 kg	31.4 kg/m ²	Day 1 Baseline 1h Po~	0.925
6	01-701-1028	178. cm	99.3 kg	31.4 kg/m ²	Day 1 Baseline 1.5h ~	1.19
7	01-701-1028	178. cm	99.3 kg	31.4 kg/m ²	Day 1 Baseline 2h Po~	1.37
8	01-701-1028	178. cm	99.3 kg	31.4 kg/m ²	Day 1 Baseline 4h Po~	1.68
9	01-701-1028	178. cm	99.3 kg	31.4 kg/m ²	Day 1 Baseline 0-6h ~	24.9
10	01-701-1028	178. cm	99.3 kg	31.4 kg/m ²	Day 1 Baseline 6h Po~	1.76

Add the ADSL variables

If needed, the other ADSL variables can now be added:

```
# Add all ADSL variables
adpc <- adpc_baselines %>%
  derive_vars_merged(
    dataset_add = select(adsl, !!!negate_vars(adsl_vars)),
    by_vars = exprs(STUDYID, USUBJID)
  ) # dim(adpc) [1] 3852 123
```

Adding attributes to the ADPC file will be discussed [below](#). We will now turn to the Population PK example.

Programming Population PK (ADPPK) Analysis Data

The Population PK Analysis Data (ADPPK) follows the CDISC Implementation Guide (<https://www.cdisc.org/standards/foundational/adam/basic-data-structure-adam-poppk-implementation-guide-v1-0>). The programming workflow for Population PK (ADPPK) Analysis Data is similar to the NCA Programming flow with a few key differences. Population PK models generally make use of nonlinear mixed effects models that require numeric variables. The data used in the models will include both dosing and concentration records, relative time variables, and numeric covariate variables. A DV or dependent variable is often expected. This is equivalent to the ADaM AVAL variable and will be included in addition to AVAL for ADPPK. The ADPPK file will not have the duplicated records for analysis found in the NCA.

Here are the relative time variables we will use for the ADPPK data. These correspond to the names in the forthcoming CDISC Implementation Guide.

Variable	Variable Label
NFRLT	Nominal Rel Time from First Dose
AFRLT	Actual Rel Time from First Dose
NPRLT	Nominal Rel Time from Previous Dose
APRLT	Actual Rel Time from Previous Dose

Find First Dose ADPPK

References

[Creating a BDS Time-to-Event ADaM](#)

[Analysis Data Model Implementation Guide for Non-compartmental Analysis Input Data Version 1.0 \(Final\)](#)