ABC-XYZ Premise and Dataset Creation

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The purpose of this script is to create a dataset with an imaginary patient population for a future study.

⁻ Lockfile written to "~/repo/jonahlyon/ABC-XYZ/renv.lock".

1 Load PopPK Model Run for ABC-XYZ

Building 1001_mod ... done.

2 Create population

- Study of N = 120
- pancreatic cancer (HV = 0)
- ECOG of 0 (45%) and 1 (55%)
- obese study BMI > 30
- hepatic dysfunciton (mild n = 40, moderate n = 60)
- U.S Study 80% White, 15% Black, 3% Asian, 2% Other 52% Female, 48% Male
- QOD, 4.8 mg (n = 80) and 7.2 mg (n = 40) Dosing Groups. All doses fasted.

3 Define Cohorts and Subjects

```
drug <- "ABC-XYZ"
site <- "001"
cohort1 <- str_c("01", str_pad(1:80, pad = 0, width = 3)) # 4 mg
cohort2 <- str_c("02", str_pad(1:40, pad = 0, width = 3)) # 8 mg

cohort1_usubjid <- str_c(drug, site, cohort1, sep = "-")
cohort2_usubjid <- str_c(drug, site, cohort2, sep = "-")</pre>
```

4 Random Simulation for subject covariates

4.1 Demographics

```
set.seed(20240107)
subject_dm <- data.frame(
   USUBJID = c(cohort1_usubjid, cohort2_usubjid),
   COHORT = c(rep(1, 80), rep(2, 40)),</pre>
```

```
SEXC = sample(c("F", "M"), size = 120,
              replace = TRUE, prob = c(0.52, 0.48)),
BBMI = round(rnorm(n = 120, mean = 34.2, sd = 1.4), 1),
RACEC = sample(c("White", "Black", "Asian", "Other"), size = 120,
              replace = T, prob = c(0.8, 0.15, 0.03, 0.02)),
BAGE = round(rnorm(n = 120, 62, sd = 4.7)),
BECOG = sample(c(0, 1), size = 120, replace = TRUE, prob = c(0.45, 0.55))
mutate(
  SEXF = ifelse(SEXC == "M", 0, 1),
  RACEN = case_when(
    RACEC == "White" ~ 1,
    RACEC == "Black" ~ 2,
    RACEC == "Asian" ~ 3,
    RACEC == "Other" ~ 4
  ),
  BHT = ifelse(SEXF == 1,
               round(rnorm(n = 120, mean = 163, sd = 6.35)/100, 2),
               round(rnorm(n = 120, mean = 175, sd = 7.62)/100, 2)),
  BWT = round(BHT^2*BBMI, 1),
  PTYPEC = "Pancreatic Cancer",
  PTYPE = 4
)
```

4.2 Labs

```
subject_lb <- data.frame(
   USUBJID = c(cohort1_usubjid, cohort2_usubjid),
   BHFCC = sample(c("Mild", "Moderate"), size = 120, replace = TRUE, prob = c(0.4, 0.6)),
   BALB = round(rnorm(n = 120, mean = 3.6, sd = 0.25), 2),
   BCRCL = round(rnorm(n = 120, mean = 85, sd = 12), 1)
) |>
   mutate(
   BHFC = ifelse(BHFCC == "Mild", 1, 2)
)
```

4.3 Combine Demographics and Labs

```
subject_dm_lb <- subject_dm |>
  left_join(subject_lb, by = "USUBJID")
```

- 5 Dosing History
- 6 28-day cycles

7 QOD for 4.8, QOD for 7.2

```
subject_ex <- subject_dm_lb |>
 select(USUBJID, COHORT) |>
 mutate(
   AMT = ifelse(COHORT == 1, 4.8, 7.2)
 ) |>
 crossing(
   ATFD = seq(0, 2016, by = 48) # QOD for 3 28 day cycles, in hours
 ) |>
 mutate(
   NTFD = ATFD,
   ATLD = 0,
   NTLD = 0,
   EVID = 1,
   CMT = 1,
   BLQ = 0,
   MDV = 1,
   FOOD = 0
  )
```

8 Create Base (Demographics + Dose History) Dataset

```
doses_demogs <- subject_ex |>
  left_join(subject_dm_lb, by = c("USUBJID", "COHORT")) |>
  mutate(
    C = NA_integer_,
```

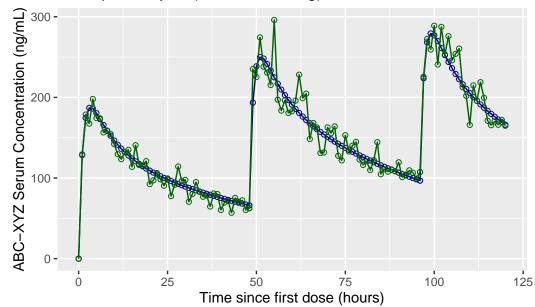
[1] "99a885e785f94ba870ccc7155c76288c"

9 Test Simulation Run (validate dataset for use in mrgsolve and NONMEM)

Note - hourly saampling is meant to demonstrate variability in PK samples, not show an example study design.

```
geom_point(col = "darkblue", shape = 1) +
geom_line(col = "darkblue") +
geom_point(aes(y = Y), col = "darkgreen", shape = 1) +
geom_line(aes(y = Y), col = "darkgreen") +
xlim(0, 120) +
xlab("Time since first dose (hours)") +
ylab("ABC-XYZ Serum Concentration (ng/mL)") +
ggtitle("Example Subject (ID = 47, 4.8 mg) First 3 Doses")
```

Example Subject (ID = 47, 4.8 mg) First 3 Doses



BLUE = individual predicted value; GREEN = simulation of value based on known variability (residual error)