Comparison of the precision of predicted dispensation length when using single or multiple index dates and saturated sampling

SG

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```
library(wtdr)
##
## Caricamento pacchetto: 'wtdr'
## I seguenti oggetti sono mascherati da 'package:stats':
##
##
      dexp, dlnorm
library(bbmle)
## Caricamento del pacchetto richiesto: stats4
library(haven)
library(tidyverse)
## -- Attaching core tidyverse packages ---
                                             ----- tidyverse 2.0.0 --
## v dplyr 1.1.4 v readr
                                  2.1.5
## v forcats 1.0.0 v stringr 1.5.1
## v ggplot2 3.5.1 v tibble 3.2.1
## v lubridate 1.9.4
                    v tidyr
                                  1.3.1
## v purrr
            1.0.2
## -- Conflicts -----
                                          ## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                   masks stats::lag()
## x dplyr::slice() masks bbmle::slice()
## i Use the conflicted package (<a href="http://conflicted.r-lib.org/">http://conflicted.r-lib.org/</a>) to force all conflicts to become error
library(data.table)
##
## Caricamento pacchetto: 'data.table'
## I seguenti oggetti sono mascherati da 'package:lubridate':
```

##

```
##
       hour, isoweek, mday, minute, month, quarter, second, wday, week,
##
       yday, year
##
## I seguenti oggetti sono mascherati da 'package:dplyr':
##
##
       between, first, last
##
## Il seguente oggetto è mascherato da 'package:purrr':
##
##
       transpose
library(rlang)
##
## Caricamento pacchetto: 'rlang'
## Il seguente oggetto è mascherato da 'package:data.table':
##
##
       :=
##
## I seguenti oggetti sono mascherati da 'package:purrr':
##
##
       %0%, flatten, flatten_chr, flatten_dbl, flatten_int, flatten_lgl,
##
       flatten_raw, invoke, splice
library(boot)
library(Deriv)
library(numDeriv)
set.seed(345)
df <- read.csv(file.path("..", "extdata", "df_sat_ex.csv"))</pre>
df <- as.data.table(df)</pre>
df <- df[, X:=NULL]</pre>
df <- df[, rxdate:=as.Date(rxdate)]</pre>
\# id\_sel \leftarrow sample(1:15000, 50) \# filter df to 50 individuals
\# df\_sel \leftarrow df[df$pid %in% id\_sel ,]
# 1) single index date - forward
forw_1 <- wtdttt(data = df,</pre>
                  rxdate ~ dlnorm(logitp, mu, lnsigma),
                  id = "pid",
                  start = as.Date('2014-01-01'),
                  end = as.Date('2014-12-31'),
                  reverse = F
)
```

Warning in wtdttt(data = df, rxdate ~ dlnorm(logitp, mu, lnsigma), id = "pid",

: Some dates are out of the window defined by start and end. Keeping only rows ## within the window.

```
summary(forw_1)
## Maximum likelihood estimation
##
## Call:
## mle2(minuslog1 = form, start = init, fixed = list(delta = delta),
      data = cpy, parameters = parameters_r)
## Coefficients:
          Estimate Std. Error z value
                                        Pr(z)
         ## mu
          4.08699 0.21002 19.4599 < 2.2e-16 ***
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## -2 log L: 336.6988
##
             Estimate Std. Error z value Pr(z) Lower.95 Upper.95
## prevalence 0.7731675 0.0926652 8.343666
                                           0 0.5475307 0.9056694
predict(forw_1, distrx = "rxdate", quantile = 0.8)
                                  p_value Lower.95 Upper.95
##
    Estimate
                  SE
## 1 78.22788 12.03485 6.500115 8.025837e-11 54.64002 101.8157
# 2) single index date - reverse
rev_1 <- wtdttt(data = df,
              rxdate ~ dlnorm(logitp, mu, lnsigma),
               id = "pid",
               start = as.Date('2014-01-01'),
               end = as.Date('2014-12-31'),
               reverse = T
)
## Warning in wtdttt(data = df, rxdate ~ dlnorm(logitp, mu, lnsigma), id = "pid",
## : Some dates are out of the window defined by start and end. Keeping only rows
## within the window.
summary(rev 1)
## Maximum likelihood estimation
##
## Call:
## mle2(minuslog1 = form, start = init, fixed = list(delta = delta),
##
      data = cpy, parameters = parameters_r)
##
## Coefficients:
```

```
Estimate Std. Error z value
## logitp 1.21660 0.60208 2.0207 0.04331 *
         ## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' 1
## -2 log L: 343.3
##
##
            Estimate Std. Error z value Pr(z) Lower.95 Upper.95
## prevalence 0.7714641 0.1061507 7.267631
                                     0 0.5091354 0.9165713
predict(rev_1, distrx = "rxdate", quantile = 0.8)
   Estimate
               SE
                              p_value Lower.95 Upper.95
                        z
## 1 80.80868 22.7642 3.549814 0.0003855036 36.19166 125.4257
# 3) multiple random index date (m =5)
rev_ran_5 <- ranwtdttt(data = df,
                   rxdate ~ dlnorm(logitp, mu, lnsigma),
                   id = "pid",
                   start = as.Date('2014-01-01'),
                   end = as.Date('2014-12-31'),
                   reverse = T,
                   nsamp = 5,
                   robust = T
)
summary(rev_ran_5)
## Maximum likelihood estimation
##
## Call:
## mle2(minuslog1 = form, start = init, fixed = list(delta = delta),
     data = cpy, parameters = parameters_r)
##
## Coefficients:
##
         Estimate Std. Error z value
                                   Pr(z)
        ## logitp
         ## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## -2 log L: 1753.642
##
##
            Estimate Std. Error z value Pr(z) Lower.95 Upper.95
## prevalence 0.7389229 0.0753149 9.811112 0 0.5683695 0.8588237
predict(rev_ran_5, distrx = "rxdate", quantile = 0.8)
                              p_value Lower.95 Upper.95
   Estimate
                        z
## 1 88.59163 6.00726 14.74743 3.196245e-49 76.81761 100.3656
```

```
# 4) multiple random index date (m =50)
rev_ran_50 <- ranwtdttt(data = df,
                      rxdate ~ dlnorm(logitp, mu, lnsigma),
                      id = "pid",
                      start = as.Date('2014-01-01'),
                      end = as.Date('2014-12-31'),
                      reverse = T,
                      nsamp = 50,
                      robust = T
)
summary(rev_ran_50)
## Maximum likelihood estimation
##
## Call:
## mle2(minuslog1 = form, start = init, fixed = list(delta = delta),
      data = cpy, parameters = parameters_r)
##
## Coefficients:
          Estimate Std. Error z value
                                         Pr(z)
         ## logitp
          4.222950 0.030758 137.2961 < 2.2e-16 ***
## mu
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## -2 log L: 17614.93
##
             Estimate Std. Error z value Pr(z) Lower.95 Upper.95
##
## prevalence 0.6935241 0.0796516 8.70697 0 0.5205282 0.8250775
predict(rev_ran_50, distrx = "rxdate", quantile = 0.8)
   Estimate
                 SE
                         z
                                p_value Lower.95 Upper.95
## 1 80.45048 4.00058 20.1097 6.068548e-90 72.60949 88.29147
# 5) saturated sampling
rev_sat_e <- satwtdttt(data = df,
                     rxdate ~ dlnorm(logitp, mu, lnsigma),
                     id = "pid",
                     start = as.Date('2014-01-01'),
                     end = as.Date('2014-12-31'),
                     robust = T,
                     reverse = F
)
summary(rev_sat_e)
## Maximum likelihood estimation
##
```

```
## Call:
## mle2(minuslog1 = form, start = init, fixed = list(delta = delta),
     data = cpy, parameters = parameters_r)
##
## Coefficients:
##
     Estimate Std. Error z value Pr(z)
## logitp 0.93472 0.36930 2.5311 0.01137 *
         4.18727 0.02151 194.6680 < 2e-16 ***
## mu
## ---
## Signif. codes: 0 '*** 0.001 '** 0.01 '* 0.05 '.' 0.1 ' ' 1
## -2 log L: 126619.1
##
##
            Estimate Std. Error z value Pr(z) Lower.95 Upper.95
## prevalence 0.7180311 0.0747688 9.603352 0 0.5525321 0.8400399
predict(rev_sat_e, distrx = "rxdate", quantile = 0.8)
   Estimate
               SE
                               p_value Lower.95 Upper.95
                        Z
```

1 80.16418 2.704524 29.64077 4.459893e-193 74.86341 85.46494