Refit with Saved Parameters

Overview

This vignette walks through using a text file of previously fit model parameters to use in the bdotsRefit function. This is convenient if you have already gone through the refitting process and would like to save/load the refitted parameters in a new session.

To demonstate this process, we start with fitting a set of curves to our data

From this, we can create an appropriate data.table that can be used in a later session

```
parDT <- coefWriteout(refit)</pre>
head(parDT)
#>
      Subject Group LookType
                                         ht
                                               sig1
                                                      siq2
                                                                base1
                                                                         base2
                                 mu
#> 1:
          1
                 50 Cohort 429.76 0.19860 159.89 314.64 0.0097098 0.033761
#> 2:
                     Cohort 634.93 0.26350 303.81 215.38 -0.0206361 0.028924
            1
                 65
#> 3:
            2
                 50
                      Cohort 647.07 0.25438 518.96 255.99 -0.2130875 0.013682
            2
                 65
                      Cohort 723.05 0.25821 392.95 252.94 -0.0548262 0.031973
#> 4:
#> 5:
            3
                 50
                      Cohort 501.48 0.22477 500.85 158.42 -0.3316790 0.025227
#> 6:
            3
                 65
                      Cohort 460.72 0.30677 382.73 166.08 -0.2433086 0.039922
```

It's important that columns are included that match the unique identifying columns in our bdotsObj, and that the parameters match the coefficients used from bdotsFit

```
## Subject, Group, and LookType
head(refit)
#>
      Subject Group LookType
                                    fit
                                             R2
                                                  AR1 fitCode
                    Cohort <qnls[18]> 0.96972
#> 1:
           1
                 50
                                                 TRUE
                                                            0
#> 2:
           1
                 65 Cohort <gnls[18]> 0.98049
                                                 TRUE
#> 3:
           2
                50
                    Cohort <gnls[18]> 0.98117
                                                 TRUE
                                                            0
            2
#> 4:
                 65
                    Cohort <gnls[18]> 0.96975
                                                 TRUE
                                                            0
#> 5:
            3
                 50
                     Cohort <gnls[18]> 0.97619 TRUE
                                                            0
            3
                                                            3
#> 6:
                 65
                      Cohort <qnls[18]> 0.95349 FALSE
## doubleGauss pars
```

```
colnames(coef(refit))
#> [1] "mu" "ht" "sig1" "sig2" "base1" "base2"
```

We can save our parameter data.table for later use, or read in any other appropriately formatted data.frame

```
## Save this for later using data.table::fwrite
fwrite(parDT, file = "mypars.csv")
parDT <- fread("mypars.csv")</pre>
```

Once we have this, we can pass it as an argument to the bdotsRefit function. Doing so will ignore the remaining arguments

```
new_refit <- bdotsRefit(refit, paramDT = parDT)</pre>
```

We end up with a bdotsObj that matches what we had previously. As seeds have not yet been implemented, the resulting parameters may not be exact. It will, however, assist with not having to go through the entire refitting process again manually (although, there is always the option to save the entire object with save(refit, file = "refit.RData))

```
head(new refit)
                                                      R2 AR1 fitCode
#>
      Subject Group
                            LookType
                                             fit
#> 1:
            1
                 50
                               Cohort <gnls[18]> 0.96972 TRUE
                                                                     0
#> 2:
            1
                 50 Unrelated_Cohort <gnls[18]> 0.97900 TRUE
                                                                     0
#> 3:
                               Cohort <gnls[18]> 0.98049 TRUE
                                                                     0
            1
                 65
            1
                 65 Unrelated_Cohort <qnls[18]> 0.87164 TRUE
                                                                     1
#> 4:
                                                                     0
#> 5:
            2
                 50
                               Cohort <gnls[18]> 0.98117 TRUE
#> 6:
                 50 Unrelated_Cohort <gnls[18]> 0.95612 TRUE
                                                                     0
```