

ggPMX with nlmixr

```
library(ggPMX)
library(nlmixr)
```

It is simple to create a ggPMX controller for a nlmixr object.

Using the theophylline example with a nlmixr model we have:

```
one.compartment <- function() {
  ini({
    tka <- 0.45 # Log Ka
    tcl <- 1 # Log Cl
    tv <- 3.45 # Log V
    eta.ka ~ 0.6
    eta.cl ~ 0.3
    eta.v ~ 0.1
    add.sd <- 0.7
  })
  model({
    ka <- exp(tka + eta.ka)
    cl <- exp(tcl + eta.cl)
    v <- exp(tv + eta.v)
    d/dt(depot) = -ka * depot
    d/dt(center) = ka * depot - cl / v * center
    cp = center / v
    cp ~ add(add.sd)
  })
}

nlmixr(one.compartment)
#> __ RxODE-based ODE model __
#> -- Initialization: -----
#> Fixed Effects ($theta):
#>   tka   tcl   tv
#> 0.45 1.00 3.45
#>
#> Omega ($omega):
#>      eta.ka eta.cl eta.v
#> eta.ka    0.6    0.0    0.0
#> eta.cl    0.0    0.3    0.0
#> eta.v     0.0    0.0    0.1
#> -- mu-referencing ($muRefTable): -----
#>   theta   eta
#> 1   tka eta.ka
#> 2   tcl eta.cl
#> 3   tv  eta.v
#>
#> -- Model: -----
#>      ka <- exp(tka + eta.ka)
#>      cl <- exp(tcl + eta.cl)
```

```
#>      v <- exp(tv + eta.v)
#>      d/dt(depot) = -ka * depot
#>      d/dt(center) = ka * depot - cl / v * center
#>      cp = center / v
#>      cp ~ add(add.sd)
#> -----
```

At the time of this writing, nlmixr requires python; Since cran does not have python installed, we run this model locally by:

```
fit <- nlmixr(one.compartment, theo_sd, est="saem", control=list(print=0))
saveRDS(fit, "fit.rds")
```

Then we can read this into the system

```
fit <- readRDS("fit.rds")
capture.output(print(fit)) %>%
sapply(crayon::strip_style) %>%
unnname %>%
print
#> [1] "-- nlmixr SAEM(ODE); OBJF not calculated fit ----- "
#> [2] " Gaussian/Laplacian Likelihoods: AIC() or $objf etc. "
#> [3] " FOCEi CWRES & Likelihoods: addCwres() "
#> [4] ""
#> [5] "-- Time (sec; $time): ----- "
#> [6] "          saem      setup table covariance      other"
#> [7] "elapsed 20.803 3.921085 0.009          0.007 0.150915"
#> [8] ""
#> [9] "-- Population Parameters ($parFixed or $parFixedDf): ----- "
#> [10] "      Parameter Est.      SE %RSE Back-transformed(95%CI) BSV(CV%) Shrink(SD)%"
#> [11] "tka      Log Ka 0.451  0.196 43.5          1.57 (1.07, 2.31)      71.9      0.411% "
#> [12] "tcl      Log Cl  1.02 0.0836 8.22          2.77 (2.35, 3.26)      27.0      3.36% "
#> [13] "tv       Log V  3.45 0.0469 1.36          31.5 (28.7, 34.5)      14.0      10.0% "
#> [14] "add.sd           0.692                      0.692                      "
#> [15] " "
#> [16] ""
#> [17] " Covariance Type ($covMethod): linFim"
#> [18] " No correlations in between subject variability (BSV) matrix"
#> [19] " Full BSV covariance ($omega) or correlation ($omegaR; diagonals=SDs) "
#> [20] " Distribution stats (mean/skewness/kurtosis/p-value) available in $shrink "
#> [21] ""
#> [22] "-- Fit Data (object is a modified tibble): ----- "
#> [23] "# A tibble: 132 x 18"
#> [24] " ID      TIME      DV EVID  PRED      RES IPRED      IRES  IWRES eta.ka eta.cl  eta.v"
#> [25] " <fct> <dbl> <dbl> <int> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl> <dbl>"
#> [26] "1 1      0      0.74      0 0      0.74 0      0.74 1.07 0.105 -0.487 -0.0800"
#> [27] "2 1      0.25 2.84      0 3.26 -0.423 3.86 -1.02 -1.48 0.105 -0.487 -0.0800"
#> [28] "3 1      0.570 6.57      0 5.84 0.725 6.81 -0.235 -0.340 0.105 -0.487 -0.0800"
#> [29] "# ... with 129 more rows, and 6 more variables: ka <dbl>, cl <dbl>, v <dbl>,"
#> [30] "#      cp <dbl>, depot <dbl>, center <dbl>"
```

The fit object is a nlmixr fit; You can read it into the nlmixr controller by:

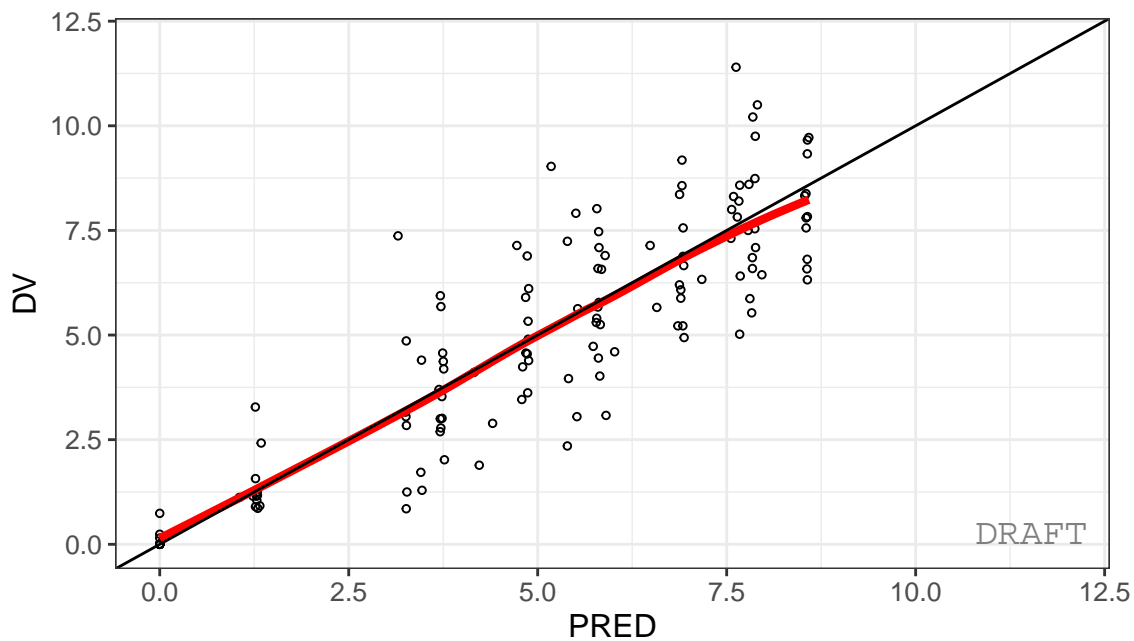
```
fit %>%
  pmx_nlmixr(vpc = FALSE) -> ## VPC is turned on by default, can turn off.
  ctr ## Assigned to controller
```

```
#> Compiling NPDE model...done
```

Once the controller is created, you can of course have the same types of diagnostic plots as the standard ggPMX package; Using the same examples as the user manual:

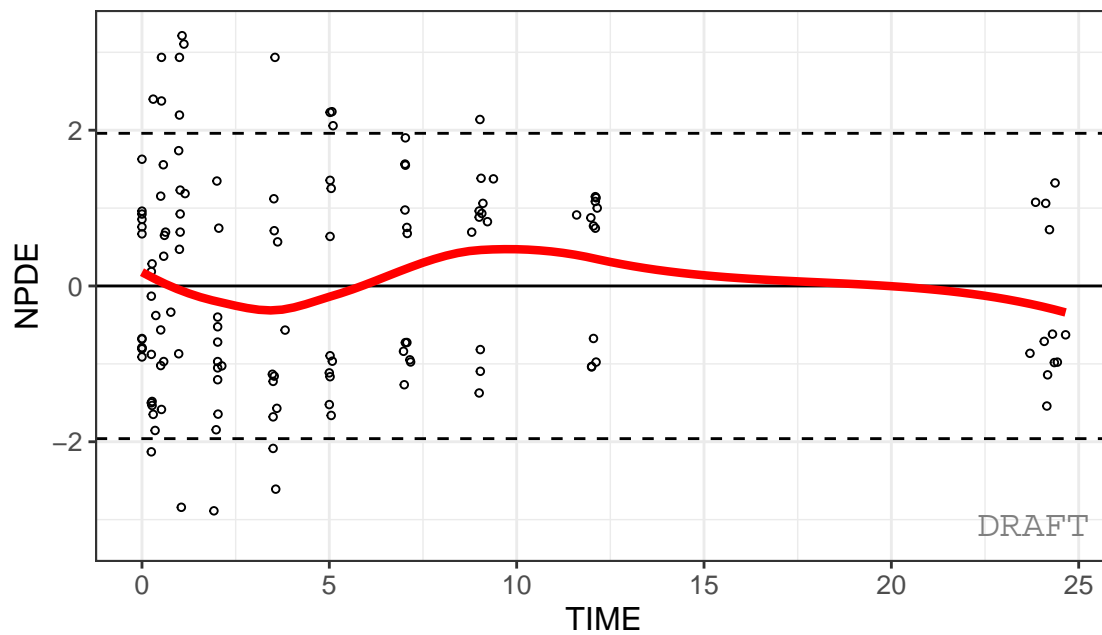
```
ctr %>% pmx_plot_dv_pred  
#> `geom_smooth()` using formula 'y ~ x'
```

DV vs PRED



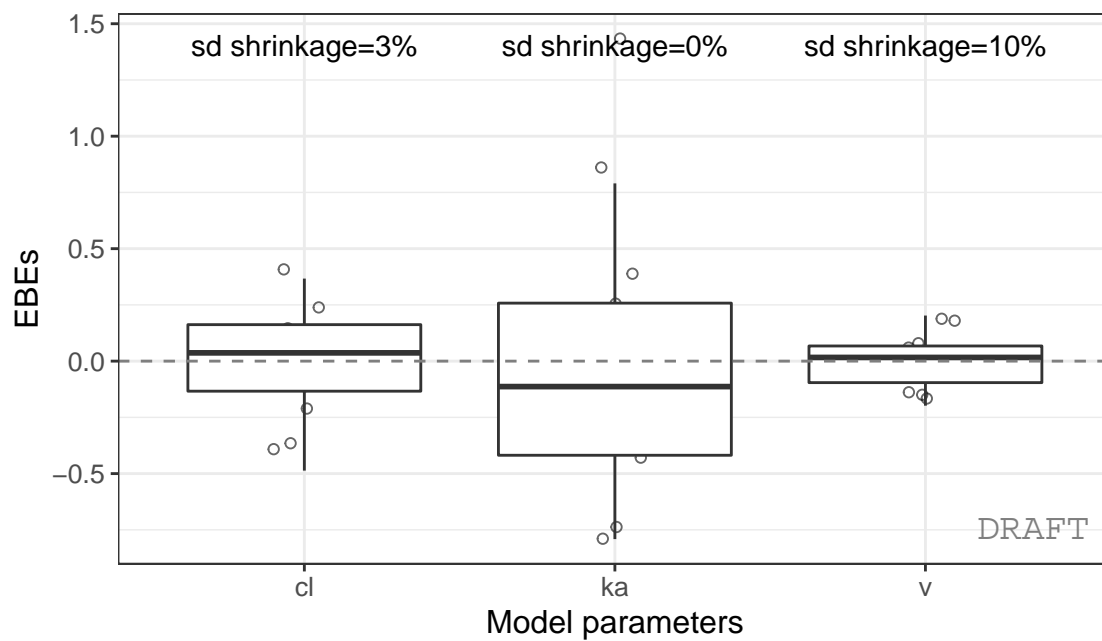
```
ctr %>% pmx_plot_npde_time  
#> `geom_smooth()` using formula 'y ~ x'
```

NPDE vs TIME



```
ctr %>% pmx_plot_vpc
#> NULL
ctr %>% pmx_plot_eta_box
```

EBE distribution

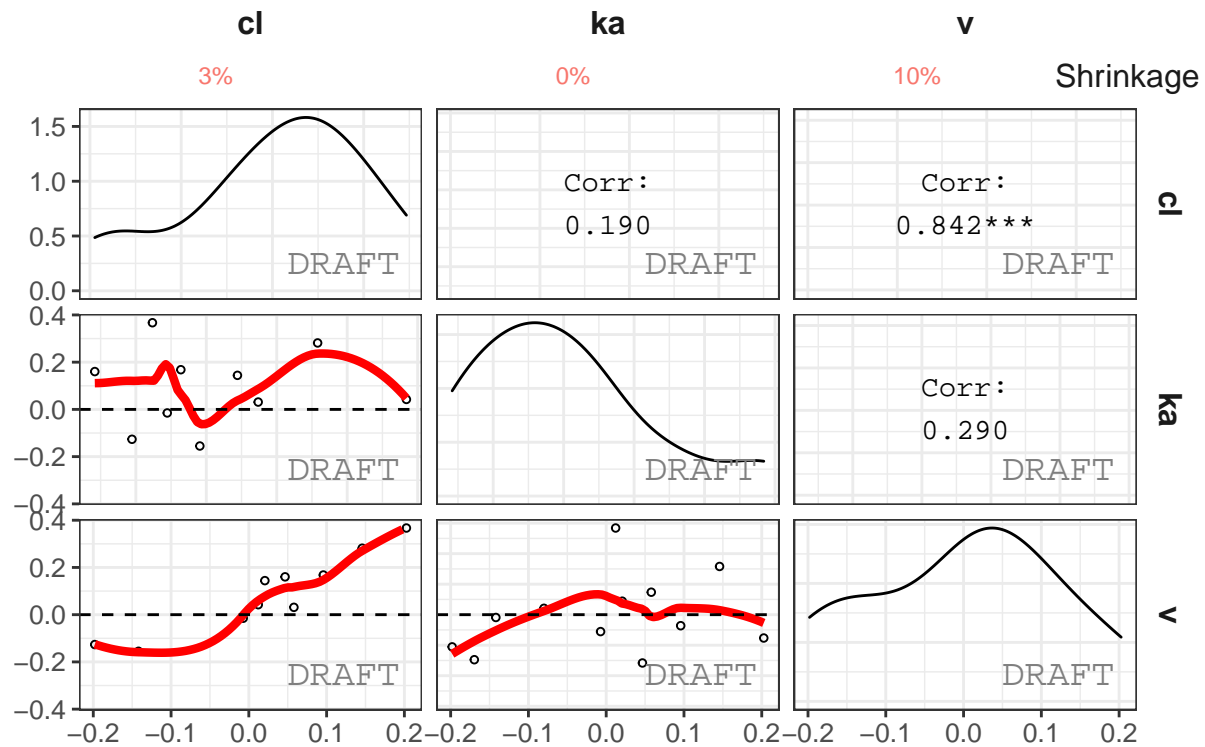


```
ctr %>% pmx_plot_eta_matrix(
  shrink=list(size=3,hjust=1.5))
#> `geom_smooth()` using formula 'y ~ x'
#> Warning: Removed 2 rows containing non-finite values (stat_smooth).
```

```
#> Warning: Removed 2 rows containing missing values (geom_point).
#> `geom_smooth()` using formula 'y ~ x'
#> Warning: Removed 2 rows containing non-finite values (stat_smooth).

#> Warning: Removed 2 rows containing missing values (geom_point).
#> Warning: Removed 1 rows containing missing values (geom_smooth).
#> `geom_smooth()` using formula 'y ~ x'
```

Correlations of random effects



This shows the standard plots for theophylline