

Mlfuns Sample Script

Simulating with Parameter Uncertainty

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1 Purpose

This script shows how to conduct a simulation that considers uncertainty in the parameter estimates.

2 Data

Here we load Mifuns and read in the data to be used for simulations.

Listing 1:

```
> library(Mifuns)
```

```
Mifuns 4.1.0
```

Listing 2:

```
> data <- read.csv("../data/derived/phase1.csv")
> head(data)
```

	C	ID	TIME	SEQ	EVID	AMT	DV	SUBJ	HOUR	TAFD	TAD	LDOS	MDV	HEIGHT	WEIGHT	SEX
1	C	1	0.00	0	0	.	0	1	0.00	0.00	.	.	0	174	74.2	0
2	.	1	0.00	1	1	1000	.	1	0.00	0.00	0	1000	1	174	74.2	0
3	.	1	0.25	0	0	. 0.363		1	0.25	0.25	0.25	1000	0	174	74.2	0
4	.	1	0.50	0	0	. 0.914		1	0.50	0.50	0.5	1000	0	174	74.2	0
5	.	1	1.00	0	0	. 1.12		1	1.00	1.00	1	1000	0	174	74.2	0
6	.	1	2.00	0	0	. 2.28		1	2.00	2.00	2	1000	0	174	74.2	0

	AGE	DOSE	FED	SMK	DS	CRCN	predose	zerodv
1	29.1	1000	1	0	0	83.5	1	1
2	29.1	1000	1	0	0	83.5	0	0
3	29.1	1000	1	0	0	83.5	0	0
4	29.1	1000	1	0	0	83.5	0	0
5	29.1	1000	1	0	0	83.5	0	0
6	29.1	1000	1	0	0	83.5	0	0

We use NONMEM output from a simple two compartment model to generate parameters. We use 1005.lst and 1005.cov output from NM7 to populate a call to Mifuns::simpar().

Listing 3:

```
> cov <- read.table("../nonmem/1005/1005.cov", skip=1, header=T)
> head(cov)
```

	NAME	THETA1	THETA2	THETA3	THETA4	THETA5
1	THETA1	0.665158000	0.31249200	1.65973e-04	0.02989100	2.13169000
2	THETA2	0.312492000	4.08110000	6.94328e-03	0.69166700	9.76609000
3	THETA3	0.000165973	0.00694328	3.02940e-05	0.00193292	-0.00658463
4	THETA4	0.029891000	0.69166700	1.93292e-03	0.26105200	1.50038000
5	THETA5	2.131690000	9.76609000	-6.58463e-03	1.50038000	283.10500000
6	THETA6	-0.046470200	-0.02244780	-9.69369e-05	-0.02437590	0.05291770

```

      THETA6      THETA7      SIGMA.1.1.      OMEGA.1.1. OMEGA.2.1.      OMEGA.2.2.
1 -4.64702e-02 -0.146935000  9.41749e-04 -1.56849e-04      0 -9.04248e-04
2 -2.24478e-02  0.186394000 -8.82373e-03  9.40159e-03      0 -2.00309e-02
3 -9.69369e-05  0.000253729 -2.62223e-05 -8.61550e-06      0 -9.88614e-05
4 -2.43759e-02  0.043642100 -1.18030e-03  6.64550e-04      0 -4.82235e-03
5  5.29177e-02 -0.671658000  1.53099e-02  2.17642e-01      0  3.31492e-02
6  1.86049e-02 -0.009628420 -7.39197e-05  2.54053e-03      0 -1.02414e-04
      OMEGA.3.1. OMEGA.3.2.      OMEGA.3.3.
1          0          0 -9.34269e-04
2          0          0 -8.34612e-03
3          0          0 -2.33533e-06
4          0          0  2.75895e-03
5          0          0  1.11018e-02
6          0          0 -5.50259e-04

```

We are interested in theta covariance, so we remove extra columns and rows.

Listing 4:

```
> cov<- cov[1:7,c(2:8)]
```

3 Parameters

Now we generate 10 sets of population parameters based on the 1005.lst results.

Listing 5:

```

> set.seed(10)
> PKparms <- simpar(
+   nsim=10,
+   theta=c(8.58,21.6, 0.0684, 3.78, 107, 0.999, 1.67),
+   covar=cov,
+   omega=list(0.196, 0.129, 0.107),
+   odf=c(40,40,40),
+   sigma=list(0.0671),
+   sdf=c(200)
+ )
> PKparms

```

```

      TH.1 TH.2      TH.3 TH.4 TH.5 TH.6 TH.7 OM.1.1.1 OM.2.1.1 OM.3.1.1
1  8.858 19.33 0.06423 4.091 106.8 0.9002 1.1870 0.1847 0.15400 0.13630
2 10.270 20.15 0.06250 3.433 110.1 0.8190 1.2940 0.2862 0.12000 0.16400
3  9.371 22.89 0.06297 3.585 130.1 1.0860 1.7050 0.1647 0.12770 0.11300
4 10.160 19.98 0.06527 3.399 117.1 1.1520 0.8838 0.1886 0.11460 0.08460
5  9.540 19.84 0.07016 3.908 102.1 0.8257 1.6340 0.1526 0.08448 0.13140
6  8.855 21.08 0.07458 4.227 100.4 0.9416 1.6640 0.2462 0.17640 0.08805
7  9.377 24.16 0.07357 4.054 127.3 0.9219 1.4800 0.2221 0.14440 0.09957
8  9.408 22.03 0.06965 4.473 113.1 0.8532 1.6320 0.2287 0.13820 0.06118
9  8.784 20.74 0.06608 3.686 134.4 0.8937 1.6620 0.1765 0.12310 0.08504

```

```

10  8.719 20.77 0.06393 3.896 111.3 1.0180 1.4060  0.2116  0.11940  0.09954
    SG.1.1.1
1   0.06894
2   0.06099
3   0.06041
4   0.07700
5   0.06269
6   0.07274
7   0.06160
8   0.06692
9   0.06092
10  0.06269
    
```

4 Control Streams

We read in a control stream and clean out extra xml markup.

Listing 6:

```

> ctl <- as.nmcontrol(readLines("../nonmem/ctl/1005.ctl"))
> ctl[] <- lapply(ctl,function(rec)sub("<.*","",rec))
    
```

Now we iterate across the rows of PKparms, writing out a separate ctl for each.

Listing 7:

```

> dir.create("../nonmem/sim")
> set <- lapply(
+   rownames(PKparms),
+   function(row,params,ctl){
+     params <- as.character(PKparms[row,])
+     ctl$prob <- sub(1005,row,ctl$prob)
+     ctl$theta <- params[1:7]
+     ctl$omega <- params[8:10]
+     ctl$sigma <- params[11]
+     names(ctl)[names(ctl)=='estimation'] <- 'simulation'
+     ctl$simulation <- paste(
+       '(',
+       as.numeric(row) + 7995,
+       'NEW) (' ,
+       as.numeric(row) + 8996,
+       'UNIFORM) ONLYSIMULATION'
+     )
+     ctl$cov <- NULL
+     ctl$table <- NULL
+     ctl$stable <- NULL
+     ctl$stable <- 'ID TIME DV WT SEX LDOS NOPRINT NOAPPEND FILE=sim.tab
    
```

```
+         write.nmcontrol(ctl, file=file.path('../nonmem/sim', paste(sep='.',
+ row, 'ctl'))))
+         return(ctl)
+     },
+     params=PKparms,
+     ctl=ctl
+ )
```

5 Simulation

Finally, we run NONMEM simulations using NONR.

Listing 8:

```
> NONR (
+     run=1:10,
+     command="/common/NONMEM/nm7_osxi/test/nm7_osxi.pl",
+     project="../nonmem/sim",
+     diag=FALSE,
+     checkrunno=FALSE,
+     grid=TRUE
+ )
```