

Simulating with Parameter Uncertainty

September 7, 2011

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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 metrumrg and read in the data to be used for simulations.

Listing 1:

```
> library(metrumrg)  
  
metrumrg 5.0
```

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   0  
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.csv output from NM7 to populate a call to metrumrg::simpar().

Listing 3:

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

      NAME     THETA1     THETA2     THETA3     THETA4     THETA5
1 THETA1  0.669038000  0.3187200  1.58905e-04  0.03757190  2.59715000
2 THETA2  0.318720000  4.0841800  6.94170e-03  0.69266000  9.96862000
3 THETA3  0.000158905  0.0069417  3.02696e-05  0.00193254 -0.00604366
4 THETA4  0.037571900  0.6926600  1.93254e-03  0.26139800  1.58175000
5 THETA5  2.597150000  9.9686200 -6.04366e-03  1.58175000 283.39800000
6 THETA6 -0.055585600 -0.0248295 -1.00494e-04 -0.02667240 -0.03980440
          THETA6     THETA7 SIGMA.1.1. OMEGA.1.1. OMEGA.2.1. OMEGA.2.2.
1 -0.055585600 -0.133741000  1.02030e-03 -7.07190e-04      0 -6.46117e-04
2 -0.024829500  0.187881000 -8.79108e-03  9.36297e-03      0 -1.98732e-02
3 -0.000100494  0.000259341 -2.61526e-05 -8.69484e-06      0 -9.83597e-05
4 -0.026672400  0.044585600 -1.16815e-03  6.89103e-04      0 -4.78282e-03
5 -0.039804400 -0.677987000  1.53154e-02  2.13660e-01      0  3.21359e-02
6  0.021986700 -0.011466100 -9.43146e-05  2.71730e-03      0 -1.45631e-04
      OMEGA.3.1. OMEGA.3.2. OMEGA.3.3.
1      0      0 -7.29033e-04
2      0      0 -8.34369e-03
3      0      0 -2.35296e-06
4      0      0  2.75930e-03
5      0      0  1.20400e-02
6      0      0 -6.06465e-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	OM1.1	OM2.2	OM3.3
1	8.869	19.32	0.06426	4.117	106.8	0.8772	1.2390	0.1847	0.15400	0.13630
2	10.280	20.16	0.06251	3.439	110.1	0.7905	1.3400	0.2862	0.12000	0.16400
3	9.403	22.91	0.06295	3.583	130.1	1.0810	1.6990	0.1647	0.12770	0.11300
4	10.180	19.99	0.06534	3.444	117.1	1.1330	0.9176	0.1886	0.11460	0.08460
5	9.529	19.84	0.07000	3.896	102.1	0.7982	1.7000	0.1526	0.08448	0.13140
6	8.845	21.08	0.07446	4.225	100.4	0.9269	1.7120	0.2462	0.17640	0.08805
7	9.405	24.17	0.07370	4.071	127.3	0.9100	1.4820	0.2221	0.14440	0.09957
8	9.414	22.03	0.06953	4.473	113.1	0.8243	1.6990	0.2287	0.13820	0.06118
9	8.829	20.76	0.06609	3.679	134.5	0.8774	1.6720	0.1765	0.12310	0.08504
10	8.733	20.77	0.06396	3.913	111.4	1.0090	1.4240	0.2116	0.11940	0.09954
	SG1.1									
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.nmctl(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$table <- NULL
+       ctl$table <- 'ID TIME DV WT SEX LDOS NOPRINT NOAPPEND FILE=sim.tab'
+       write.nmctl(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:

```
> NONR72 (
+   run=1:10,
+   command="/common/NONMEM/nm7_osxi/test/nm7_osxi.pl",
+   project="..../nonmem/sim",
+   diag=FALSE,
+   checkrunno=FALSE,
+   grid=TRUE
+ )
```