

Data Assembly

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This script assembles simulated phase 1 data.

Make sure you are in the script directory, where this files resides.

Listing 1:

```
> getwd()
[1] "/data/metrumrg/inst/example/project/script"
```

Load the metrumrg package.

Listing 2:

```
> library(metrumrg)
```

Groom the dose data

Listing 3:

```
> dose <- read.csv('../data/source/dose.csv',na.strings='.',stringsAsFactors=FALSE)
> head(dose)
```

	SUBJ	AMT	HOUR
1	1	1e+03	0
2	2	5e+03	0
3	3	1e+04	0
4	4	5e+04	0
5	5	1e+05	0
6	6	1e+03	0

Listing 4:

```
> dose <- as.keyed(dose, key=c('SUBJ','HOUR'))
> summary(dose)
```

```
SUBJ~HOUR
0 NA keys
0 duplicate keys
```

Looks okay.

Groom the demographic data.

Listing 5:

```
> dem <- read.csv('../data/source/dem.csv',na.strings='.',stringsAsFactors=FALSE)
> head(dem)
```

	SUBJ	HEIGHT	WEIGHT	SEX	AGE	DOSE	FED	SMK	DS	CRCN
1	1	174	74.2	0	29.1	1e+03	1	0	0	83.5
2	2	177	80.3	0	36.8	5e+03	1	0	0	142.0
3	3	180	94.2	0	46.4	1e+04	1	0	0	121.0
4	4	177	85.2	0	30.3	5e+04	1	0	0	127.0
5	5	166	82.8	0	32.5	1e+05	1	0	0	97.2
6	6	164	63.9	0	18.8	1e+03	1	0	0	138.0

Listing 6:

```
> dem <- as.keyed(dem, key='SUBJ')
> summary(dem)
```

```
SUBJ
0 NA keys
0 duplicate keys
```

Looks okay. Note that DOSE is a treatment group, not an actual dose.

Groom the pk data.

Listing 7:

```
> pk   <- read.csv('../data/source/pk.csv',na.strings='.',stringsAsFactors=FALSE)
> head(pk)
```

	SUBJ	HOUR	DV
1	1	0.00	0.000
2	1	0.25	0.363
3	1	0.50	0.914
4	1	1.00	1.120
5	1	2.00	2.280
6	1	3.00	1.630

Listing 8:

```
> pk <- as.keyed(pk, key=c('SUBJ','HOUR'))
> head(pk)
```

	SUBJ	HOUR	DV
1	1	0.00	0.000
2	1	0.25	0.363
3	1	0.50	0.914
4	1	1.00	1.120
5	1	2.00	2.280
6	1	3.00	1.630

Listing 9:

```
> summary(pk)
```

```
SUBJ~HOUR
1 NA keys
2 duplicate keys
unsorted: 2
```

Listing 10:

```
> pk[naKeys(pk),]
```

```
SUBJ  HOUR  DV
561    40    NA 100
```

Listing 11:

```
> pk[dupKeys(pk), ]
```

```
  SUBJ HOUR DV
560   40   72 35.5
562   40   72   NA
```

Listing 12:

```
> bad <- pk[with(pk, is.na(HOUR) | is.na(DV)), ]
> bad
```

```
  SUBJ HOUR DV
561   40   NA 100
562   40   72   NA
```

Listing 13:

```
> pk <- pk - bad
> summary(pk)
```

```
SUBJ~HOUR
0 NA keys
0 duplicate keys
```

Looks okay.

Combine these data sources into an NMTRAN-style data set. The function ‘aug’ adds columns on-the-fly. The function ‘as.nm’ sets up a chain reaction that makes sure the final result has properties of an NMTRAN data set as described in ?nm.

Every source must specify DATETIME or HOUR. All of ours specify HOUR. If HOUR is the same for two records, we want, e.g., pk samples to sort before dose records (assumed predose). SEQ controls the sort order when times and subject identifiers match.

The plus operator means “outer join” or “full merge” when the arguments are “keyed” data.frames. The pipe operator means “left join” (merge, all.x=TRUE) when the arguments are “keyed” data.frames.

Listing 14:

```
> dat <-
+     nm() +
+     aug(dose,SEQ=1,EVID=1) +
+     aug(pk,  SEQ=0,EVID=0) |
+     dem
> summary(dat)
```

	value
rows	600
records	600
comments	0
subjects	40

```

longestCase      72
naKeys          0
dupKeys          0
badDv            0
falseDv          0
zeroDv           25
predoseDv        40
badAmt           0
falseAmt         0
zeroAmt          0
noPk             0
badII            0
    
```

Note predose/zero DV. See ?zeroDv We comment-out these records.

Listing 15:

```

> dat <- hide(dat, where=predoseDv(dat), why='predose')
> summary(dat)
    
```

	value
rows	600
records	560
comments	40
subjects	40
longestCase	72
naKeys	0
dupKeys	0
badDv	0
falseDv	0
zeroDv	10
predoseDv	0
badAmt	0
falseAmt	0
zeroAmt	0
noPk	0
badII	0

We still have some zero DV that are not predose. We comment those as well.

Listing 16:

```

> dat <- hide(dat, where=zeroDv(dat), why='zerodv')
> summary(dat)
    
```

	value
rows	600
records	550
comments	50
subjects	40
longestCase	72
naKeys	0

```

dupKeys      0
badDv        0
falseDv      0
zeroDv       0
predoseDv   0
badAmt       0
falseAmt     0
zeroAmt     0
noPk         0
badII        0
    
```

Listing 17:

```
> head(dat)
```

C	SUBJ	TIME	SEQ	HOUR	EVID	ID	AMT	TAFD	TAD	LDOS	DV	MDV	HEIGHT	WEIGHT	SEX	
1	C	1	0.00	0	0.00	0	1	NA	0.00	NA	NA	0.000	0	174	74.2	0
2	.	1	0.00	1	0.00	1	1	1000	0.00	0.00	1000	NA	1	174	74.2	0
3	.	1	0.25	0	0.25	0	1	NA	0.25	0.25	1000	0.363	0	174	74.2	0
4	.	1	0.50	0	0.50	0	1	NA	0.50	0.50	1000	0.914	0	174	74.2	0
5	.	1	1.00	0	1.00	0	1	NA	1.00	1.00	1000	1.120	0	174	74.2	0
6	.	1	2.00	0	2.00	0	1	NA	2.00	2.00	1000	2.280	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 could rearrange columns for convenience and clarity.

Listing 18:

```
> dat <- shuffle(dat,c('C','ID','TIME','SEQ','EVID','AMT','DV'))
> head(dat)
```

C	ID	TIME	SEQ	EVID	AMT	DV	SUBJ	HOUR	TAFD	TAD	LDOS	MDV	HEIGHT	WEIGHT	SEX	
1	C	1	0.00	0	0	NA	0.000	1	0.00	0.00	NA	NA	0	174	74.2	0
2	.	1	0.00	1	1	1000	NA	1	0.00	0.00	0.00	1000	1	174	74.2	0
3	.	1	0.25	0	0	NA	0.363	1	0.25	0.25	0.25	1000	0	174	74.2	0
4	.	1	0.50	0	0	NA	0.914	1	0.50	0.50	0.50	1000	0	174	74.2	0
5	.	1	1.00	0	0	NA	1.120	1	1.00	1.00	1.00	1000	0	174	74.2	0
6	.	1	2.00	0	0	NA	2.280	1	2.00	2.00	2.00	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 create a file using write.nm to format NAs specially, etc.

Listing 19:

```
> write.nm(dat,file='../../data/derived/phase1.csv')
```

We create a summary of which columns were hidden for which reasons.

Listing 20:

```
> summary(hidden(dat))
```

	predose	zerodv
total	40	10
unique	40	10