

Creating an R data set from STAR

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Abstract

A substantial portion of the data from Tennessee's Student Teacher Achievement Ratio (STAR) project, a large-scale, four-year study of reduced class size, has been made available to the public at <http://www.heros-inc.org/data.htm>. We describe the creation of an R (<http://www.r-project.org>) data set from these data.

1 Introduction

The data from the STAR project are available in several different forms from the web site <http://www.heros-inc.org/data.htm>. The most convenient form for creation of an R data set is the tab-delimited text file. Download and unzip the archive file <http://www.heros-inc.org/ascii-star.zip> producing two files: `readme.txt`, a description of the data, and `webstar.txt`, the data themselves.

2 Reading the data

From the data description file we can see that there are 53 columns in the data set and most of these columns are coded values. Such columns should be represented as factors in R but many of these columns will need to be combined before we can work with them. We will convert the first 5 columns to factors and leave the remaining 48 columns as integers.

```
orig <- read.table("webstar.txt", header = TRUE,  
                  colCl = rep(c("factor", "integer"), c(5, 48)))
```

In the call to `read.table` we used an explicit file name for the data file. In practice it is often more convenient to use the `file.choose` function which brings up a file chooser panel.

We can check the form of the original data with

```
> str(orig)
```

```

`data.frame':      11598 obs. of  53 variables:
 $ NEWID   : Factor w/ 11598 levels "100017","100028",...: 839 943 986 1104 1263 1346 1485 15
 $ SSEX    : Factor w/ 3 levels "1","2","9": 2 2 2 1 1 1 1 2 1 1 ...
 $ SRACE   : Factor w/ 7 levels "1","2","3","4",...: 2 1 2 1 2 1 2 1 1 1 ...
 $ SBIRTHQ : Factor w/ 5 levels "1","2","3","4",...: 3 1 4 4 1 3 1 4 2 3 ...
 $ SBIRTHY : Factor w/ 7 levels "1977","1978",...: 3 4 3 3 4 3 3 3 3 3 ...
 $ STARK   : int    2 1 1 2 1 2 2 2 2 2 ...
 $ STAR1   : int    2 1 1 2 2 2 2 1 1 1 ...
 $ STAR2   : int    2 1 1 2 2 1 2 1 1 1 ...
 $ STAR3   : int    1 1 1 1 2 1 1 1 1 1 ...
 $ CLYPEK  : int    9 1 1 9 3 9 9 9 9 9 ...
 $ CLYPE1  : int    9 1 1 9 9 9 9 3 2 2 ...
 $ CLYPE2  : int    9 1 3 9 9 2 9 3 2 2 ...
 $ CLYPE3  : int    2 1 3 1 9 2 3 3 2 2 ...
 $ SCHYPEK : int    9 3 2 9 1 9 9 9 9 9 ...
 $ HDEGK   : int    9 2 2 9 2 9 9 9 9 9 ...
 $ CLADK   : int    9 1 1 9 6 9 9 9 9 9 ...
 $ TOTEXPK : int    99 7 21 99 0 99 99 99 99 99 ...
 $ TRACEK  : int    9 1 1 9 1 9 9 9 9 9 ...
 $ TREADSSK: int    999 447 450 999 439 999 999 999 999 ...
 $ TMATHSSK: int    999 473 536 999 463 999 999 999 999 ...
 $ SESK    : int    9 2 2 9 1 9 9 9 9 9 ...
 $ SCHYPE1 : int    9 3 2 9 9 9 9 3 3 3 ...
 $ TRACE1  : int    9 1 2 9 9 9 9 1 1 1 ...
 $ HDEG1   : int    9 1 2 9 9 9 9 2 2 1 ...
 $ CLAD1   : int    9 4 3 9 9 9 9 2 4 4 ...
 $ TOTEXP1 : int    99 7 32 99 99 99 99 8 13 7 ...
 $ TREADSS1: int    999 507 579 999 999 999 999 475 999 651 ...
 $ TMATHSS1: int    999 538 592 999 999 999 999 512 999 532 ...
 $ SES1    : int    9 1 9 9 9 9 9 2 2 2 ...
 $ SCHYPE2 : int    9 3 2 9 9 3 9 3 3 3 ...
 $ TRACE2  : int    9 1 2 9 9 1 9 2 1 1 ...
 $ HDEG2   : int    9 1 1 9 9 1 9 2 1 1 ...
 $ CLAD2   : int    9 2 4 9 9 1 9 4 4 4 ...
 $ TOTEXP2 : int    99 3 4 99 99 13 99 13 6 8 ...
 $ TREADSS2: int    999 568 588 999 999 999 999 573 999 596 ...
 $ TMATHSS2: int    999 579 579 999 999 999 999 550 999 590 ...
 $ SES2    : int    9 2 2 9 9 2 9 2 2 2 ...
 $ SCHYPE3 : int    2 3 2 3 9 3 1 3 3 3 ...
 $ TREADSS3: int    580 587 644 686 999 644 999 599 999 626 ...
 $ TMATHSS3: int    564 593 639 667 999 648 999 583 999 618 ...
 $ SES3    : int    1 1 2 2 9 2 1 2 2 2 ...
 $ TRACE3  : int    1 1 1 1 9 1 2 1 1 1 ...
 $ HDEG3   : int    1 1 1 1 9 1 1 1 1 2 ...
 $ CLAD3   : int    4 2 4 4 9 4 1 6 4 4 ...
 $ TOTEXP3 : int    30 1 4 10 99 15 17 23 8 8 ...

```

```

$ SYSIDKN : int  999 30 11 999 11 999 999 999 999 999 ...
$ SYSID1N : int  999 30 11 999 999 999 999 4 40 21 ...
$ SYSID2N : int  999 30 11 999 999 6 999 4 40 21 ...
$ SYSID3N : int  22 30 11 6 999 6 11 4 40 21 ...
$ SCHIDKN : int  999 63 20 999 19 999 999 999 999 999 ...
$ SCHID1N : int  999 63 20 999 999 999 999 5 77 50 ...
$ SCHID2N : int  999 63 20 999 999 8 999 5 77 50 ...
$ SCHID3N : int  54 63 20 8 999 8 31 5 77 50 ...

```

2.1 Missing value codes

All the columns except the first column have missing values present. Typically the missing value code is "9" but "99", "999" and "9999" are also used. We convert these to R's missing value code NA column by column.

```

> mv <- rep("9", 53)
> mv[c(4, 17, 26, 34, 45)] <- "99"
> mv[c(19, 20, 27, 28, 35, 36, 39, 40, 46:53)] <- "999"
> mv[5] <- "9999"
> mv[1] <- "999999"
> for (i in seq(a = orig)) orig[[i]][orig[[i]] == mv[i]] <- NA
> summary(orig[1:5])

```

	NEWID	SSEX	SRACE	SBIRTHQ
100017 :	1	1 :6122	1 :7193	1 :2836
100028 :	1	2 :5456	2 :4173	2 :2851
100045 :	1	9 : 0	3 : 32	3 :3422
100064 :	1	NA's: 20	4 : 21	4 :2423
100070 :	1		6 : 20	99 : 0
100096 :	1		(Other): 14	NA's: 66
(Other):	11592		NA's : 145	
SBIRTHY				
1980 :	6886			
1979 :	3915			
1978 :	645			
1977 :	58			
1981 :	24			
(Other):	1			
NA's :	69			

Notice that level "9" is still present for the SSEX variable even after all the observations at that level have been replaced by the missing value code. To remove these unused levels from this and all the other columns, we loop over the columns selecting all the values but using the optional argument `drop = TRUE`.

```

> for (i in seq(a = orig)) orig[[i]] <- orig[[i]][drop = TRUE]
> summary(orig[1:5])

```

NEWID		SSEX		SRACE		SBIRTHQ		SBIRTHY
100017	:	1	1	:6122	1	:7193	1	:2836
1977	:							58
100028	:	1	2	:5456	2	:4173	2	:2851
1978	:							645
100045	:	1	NA's:	20	3	:32	3	:3422
1979	:							3915
100064	:	1			4	:21	4	:2423
1980	:							6886
100070	:	1			5	:14	NA's:	66
1981	:							24
100096	:	1			6	:20		
1982	:							1
(Other)	:				NA's:			145
								69

For convenience we convert the names of the columns to lower case.

```
> names(orig) <- tolower(names(orig))
```

3 Setting factor levels

In R the levels of a factor can be given meaningful labels instead of numeric codes and in most cases this eliminates the need for a separate codebook. For example storing the labels of `sex` as "M" and "F" makes the coding self-explanatory. When used in a model a factor is automatically converted to a set of “contrasts” (there is a technical definition of the term “contrast” in linear models that is not always fulfilled by these derived variables) and the corresponding coefficients are given meaningful names.

When there is a natural ordering of the levels of a factor it can be created as an ordered factor that will preserve this ordering.

The labels can be set after the factor is created or as part of the creation of the factor. Below we will create a “long form” of the data where each row corresponds to a combination of student and grade. In doing this we will need to concatenate related columns of the original data frame. For example, the columns `cltypek`, `cltype1`, `cltype2` and `cltype3` will be concatenated to form a single column `cltype`. If the coding is consistent across the grades then it is easiest to concatenate the integer codes and set the labels on the “long” version of the variable.

However there are two groups of variables, `hdeg` and `clad`, that are not coded consistently. In each case the codes used for kindergarten teachers are different from those used for teachers of grades 1 to 3 classes. The codes for kindergarten teachers are a superset of those for the other teachers but the numbering is not consistent; a bachelor's degree is coded as 2 for kindergarten but 1 for the others. Thus we cannot combine the numeric values - we must create the labels for each column and then concatenate the labels and convert to a factor.

```
> orig$hdegk <- ordered(orig$hdegk, levels = 1:6, labels = c("ASSOC",
+ "BS/BA", "MS/MA/MEd", "MA+", "Ed.S", "Ed.D/Ph.D"))
> orig$hdeg1 <- ordered(orig$hdeg1, levels = 1:4, labels = c("BS/BA",
+ "MS/MA/MEd", "Ed.S", "Ed.D/Ph.D"))
> orig$hdeg2 <- ordered(orig$hdeg2, levels = 1:4, labels = c("BS/BA",
+ "MS/MA/MEd", "Ed.S", "Ed.D/Ph.D"))
```

```

> orig$hdeg3 <- ordered(orig$hdeg3, levels = 1:4, labels = c("BS/BA",
+ "MS/MA/MEd", "Ed.S", "Ed.D/Ph.D"))
> orig$cladk <- factor(orig$cladk, levels = c(1:3, 5:8),
+ labels = c("1", "2", "3", "APPR", "PROB", "NOT",
+ "PEND"))
> orig$clad1 <- factor(orig$clad1, levels = 1:6, labels = c("NOT",
+ "APPR", "PROB", "1", "2", "3"))
> orig$clad2 <- factor(orig$clad2, levels = 1:6, labels = c("NOT",
+ "APPR", "PROB", "1", "2", "3"))
> orig$clad3 <- factor(orig$clad3, levels = 1:6, labels = c("NOT",
+ "APPR", "PROB", "1", "2", "3"))

```

4 Creating separate data frames

These data are represented in a “wide” format where each row corresponds to a student. Some of the columns, such as `ssex`, are indeed a property of the student; some, such as `hdegk` are properties of teachers; some, such as `sctypek` are properties of schools or classes in schools; and some are unique to a student/grade combination. We will create separate frames for each of these types.

The first 5 columns are student-level data

```

> student <- orig[1:5]
> names(student) <- c("id", "sx", "eth", "birthq", "birthy")
> levels(student$sx) <- c("M", "F")
> levels(student$eth) <- c("W", "B", "A", "H", "I", "O")
> student$birthy <- ordered(student$birthy)
> student$birthq <- ordered(paste(student$birthy, student$birthq,
+ sep = ":"))
> summary(student)

```

	id	sx	eth	birthq	birthy
100017 :	1	M :6122	W :7193	1980:3 :2304	1977: 58
100028 :	1	F :5456	B :4173	1980:1 :2221	1978: 645
100045 :	1	NA's: 20	A : 32	1980:2 :2190	1979:3915
100064 :	1		H : 21	1979:4 :1879	1980:6886
100070 :	1		I : 14	1979:3 : 923	1981: 24
100096 :	1		O : 20	1979:2 : 586	1982: 1
(Other):11592			NA's: 145	(Other):1495	NA's: 69

The other columns refer to a combination of the student and grade. We first create an expanded or “long” version of the table with a row for each student/grade combination.

To create the long version of the table we repeat the student ids four times and add a column for the grade level. Related groups of columns, such as `cltypek`, `cltype1`, `cltype2` and `cltype3`, are concatenated then converted to a factor. However, there are two groups, `hdeg` and `clad`, for which this approach will not work because these groups are not encoded consistently.

```

> long <- data.frame(id = rep(orig$newid, 4), gr = ordered(rep(c("K",
+   1:3), each = nrow(orig)), levels = c("K", 1:3)),
+   star = factor(unlist(orig[6:9])), cltype = factor(unlist(orig[10:13])),
+   schtype = factor(unlist(orig[c(14, 22, 30, 38)])),
+   hdeg = ordered(unlist(lapply(orig[c(15, 24, 32,
+   43)], as.character)), levels = c("ASSOC", "BS/BA",
+   "MS/MA/MEd", "MA+", "Ed.S", "Ed.D/Ph.D")), clad = factor(unlist(lapply(orig[c(16,
+   25, 33, 44)], as.character)), levels = c("NOT",
+   "APPR", "PROB", "PEND", "1", "2", "3")), exp = unlist(orig[c(17,
+   26, 34, 45)]), trace = factor(unlist(orig[c(18,
+   23, 31, 42)]), levels = 1:6, labels = c("W",
+   "B", "A", "H", "I", "O")), read = unlist(orig[c(19,
+   27, 35, 39)]), math = unlist(orig[c(20, 28,
+   36, 40)]), ses = factor(unlist(orig[c(21, 29,
+   37, 41)]), labels = c("F", "N")), sch = factor(unlist(orig[50:53])))

```

We can now eliminate the combinations that are completely missing. Checking

```
> summary(long)
```

id	gr	star	cltype	schtype
100017 :	4	K:11598	1 : 8015	1 : 5624
100028 :	4	1:11598	2 : 9192	2 : 6428
100045 :	4	2:11598	3 : 9589	3 :12561
100064 :	4	3:11598	NA's:19596	4 : 2183
100070 :	4			NA's:19596
100096 :	4			
(Other):	46368			

hdeg	clad	exp	trace
ASSOC : 0	1 :18303	Min. : 0.00	W :21550
BS/BA :16586	APPR : 2030	1st Qu.: 5.00	B : 5005
MS/MA/MEd: 9587	PROB : 1961	Median : 11.00	A : 14
MA+ : 161	NOT : 1757	Mean : 12.04	H : 0
Ed.S : 237	3 : 1059	3rd Qu.: 17.00	I : 0
Ed.D/Ph.D: 58	(Other): 877	Max. : 42.00	O : 0
NA's :19763	NA's :20405	NA's :19789.00	NA's:19823

read	math	ses	sch
Min. : 315	Min. : 288.0	F :13111	51 : 826
1st Qu.: 467	1st Qu.: 505.0	N :12858	27 : 562
Median : 552	Median : 557.0	NA's:20423	9 : 543
Mean : 540	Mean : 553.7		22 : 534
3rd Qu.: 604	3rd Qu.: 603.0		63 : 534
Max. : 775	Max. : 774.0		(Other):23797
NA's :22130	NA's :21779.0		NA's :19596

indicates that fewest missing values are in the `sch`, `cltype`, and `schtype` columns. They are also consistent

```
> with(long, all.equal(is.na(schtype), is.na(sch)))
```

```
[1] TRUE
```

```
> with(long, all.equal(is.na(cltype), is.na(sch)))
```

```
[1] TRUE
```

hence we use these to subset the data frame

```
> long <- long[!is.na(long$sch), ]
```

It turns out that we could have used the `star` column as this simply indicates if the student was in the study that year.

```
> summary(long[1:5])
```

	id	gr	star	cltype	schtype
100173 :	4	K:6325	1:26796	1:8015	1: 5624
100201 :	4	1:6829	2: 0	2:9192	2: 6428
10023 :	4	2:6840		3:9589	3:12561
100236 :	4	3:6802			4: 2183
100302 :	4				
100361 :	4				
(Other):					26772

Because it now contains no information we will drop it.

```
> long$star <- NULL
```

For convenience we set the row names of this data frame to be a combination of the student id and the grade.

```
> rownames(long) <- paste(long$id, long$gr, sep = "/")
```

We can extract the school-level data from this table.

```
> school <- unique(long[, c("sch", "schtype")])
```

```
> length(levels(school$sch)) == nrow(school)
```

```
[1] TRUE
```

```
> row.names(school) <- school$sch
```

```
> school <- school[order(as.integer(as.character(school$sch))),  
+ ]
```

```
> long$schtype <- NULL
```

```
> levels(school$schtype) <- c("inner", "suburb", "rural",  
+ "urban")
```

```
> levels(long$cltype) <- c("small", "reg", "reg+A")
```

We can create a merged data set with

```
> star <- merge(merge(long, school, by = "sch"), student,  
+ by = "id")
```

```
> star$time <- as.integer(star$gr) - 1
```

5 Assigning teacher ids

There are no teacher id numbers available but we can obtain a reasonably accurate surrogate by determining the unique combinations of all the variables associated with the teacher.

```
> teacher <- unique(star[, c("cltype", "trace", "exp",
+   "clad", "gr", "hdeg", "sch")])
> teacher <- teacher[with(teacher, order(sch, gr, cltype,
+   exp, hdeg, clad, trace)), ]
```

To generate the correspondence between the observations and the teacher we create labels that incorporate the levels of each of the variables that defined the unique combinations.

```
> row.names(teacher) <- tch <- teacher$tch <- seq(nrow(teacher))
> names(tch) <- do.call("paste", c(teacher[, 1:7], list(sep = ":")))
> star$tch <- tch[do.call("paste", c(star[c("cltype",
+   "trace", "exp", "clad", "gr", "hdeg", "sch")], list(sep = ":")))]
```

We can check if this is successful by generating tables of class sizes.

```
> table(table(star$tch))

 1  2  3 11 12 13 14 15 16 17 18 19 20 21 22 23 24
45  4  1  2 17 68 81 116 111 106 28 41 49 103 152 137 138
25 26 27 28 29 30 32 44 46
83 47 29 13 11  2  1  1  1

> table(table(subset(star, cltype == "small")$tch))

 1  2 11 12 13 14 15 16 17 18 19 20 32
12  1  2 17 68 81 114 108 100 22  9  1  1

> table(table(subset(star, cltype == "reg")$tch))

 1  2 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30
21  1  1  2  5  4 16 31 47 84 64 70 38 20 11  4  5  1

> table(table(subset(star, cltype == "reg+A")$tch))

 1  2  3 15 16 17 18 19 20 21 22 23 24 25 26 27 28 29 30 44 46
12  2  1  1  1  1  2 16 17 56 68 73 68 45 27 18  9  6  1  1  1
```

We see that there are three classes with sizes greater than 30 and that one of these is labelled as a “small” class. It is likely that each of these represents two or more classes but we do not have enough information to distinguish them.

6 Initial model fits

Some initial model fits are

```
> library(lme4)
> (mm1 <- lmer(math ~ gr + sx + eth + cltype + (1 | id) +
+ (1 | sch), star))
```

Linear mixed-effects model fit by REML
Formula: math ~ gr + sx + eth + cltype + (1 | id) + (1 | sch)
Data: star

	AIC	BIC	logLik	deviance	REMLdeviance
	245170	245291.6	-122570	245172.7	245140

Random effects:

Groups	Name	Variance	Std.Dev.
id	(Intercept)	982.92	31.352
sch	(Intercept)	122.60	11.073
Residual		686.62	26.203

of obs: 24578, groups: id, 10732; sch, 80

Fixed effects:

	Estimate	Std. Error	DF	t value	Pr(> t)
(Intercept)	560.49749	1.47524	24566	379.9371	< 2.2e-16
gr.L	96.41234	0.39614	24566	243.3818	< 2.2e-16
gr.Q	-4.57970	0.36384	24566	-12.5873	< 2.2e-16
gr.C	-3.46016	0.34944	24566	-9.9020	< 2.2e-16
sxF	2.95264	0.71504	24566	4.1293	3.650e-05
ethB	-22.89116	1.27486	24566	-17.9558	< 2.2e-16
ethA	2.12834	7.02898	24566	0.3028	0.76205
ethH	1.17345	10.17194	24566	0.1154	0.90816
ethI	-34.78325	14.50732	24566	-2.3976	0.01651
ethO	2.41930	8.72130	24566	0.2774	0.78147
cltypereg	-7.11133	0.72798	24566	-9.7685	< 2.2e-16
cltypereg+A	-5.91037	0.73984	24566	-7.9888	1.423e-15

```
> (rm1 <- lmer(read ~ gr + sx + eth + cltype + (1 | id) +
+ (1 | sch), star))
```

Linear mixed-effects model fit by REML
Formula: read ~ gr + sx + eth + cltype + (1 | id) + (1 | sch)
Data: star

	AIC	BIC	logLik	deviance	REMLdeviance
	241495.0	241616.4	-120732.5	241497.5	241465.0

Random effects:

Groups	Name	Variance	Std.Dev.
id	(Intercept)	944.01	30.725
sch	(Intercept)	109.48	10.463

Residual 692.05 26.307
of obs: 24226, groups: id, 10621; sch, 80

Fixed effects:

	Estimate	Std. Error	DF	t value	Pr(> t)
(Intercept)	541.70124	1.41601	24214	382.5550	< 2.2e-16
gr.L	131.40304	0.40070	24214	327.9338	< 2.2e-16
gr.Q	-28.20738	0.36784	24214	-76.6840	< 2.2e-16
gr.C	-1.62996	0.35313	24214	-4.6158	3.936e-06
sxF	9.06081	0.70950	24214	12.7707	< 2.2e-16
ethB	-18.80739	1.25458	24214	-14.9910	< 2.2e-16
ethA	8.50278	6.93566	24214	1.2260	0.22023
ethH	1.94184	10.04311	24214	0.1934	0.84669
ethI	-32.09847	14.36068	24214	-2.2352	0.02542
ethO	7.63496	8.61206	24214	0.8865	0.37533
cltypereg	-7.82802	0.72858	24214	-10.7442	< 2.2e-16
cltypereg+A	-4.74488	0.74020	24214	-6.4103	1.479e-10