

Tidy data

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August 2011



Wednesday, August 10, 2011

1. What is tidy data?
2. Five common causes of messiness
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4. stringr review

What is tidy data?

- A step along the road to clean data
- Data that is easy to model, visualise and aggregate (i.e. works well with `lm`, `ggplot`, and `ddply`)
- Variables in columns, observations in rows, one type per dataset

	Pregnant	Not pregnant
Male	0	5
Female	1	4

There are three variables in this data set.
What are they?

pregnant	sex	n
no	female	4
no	male	5
yes	female	1
yes	male	0

Storage	Meaning
Table / File	Data set
Rows	Observations
Columns	Variables

Common causes of messiness

- column headers are values, not variable names
- multiple variables are stored in one column
- variables are stored in both rows and columns
- multiple types of experimental unit stored in the same table
- one type of experimental unit stored in multiple tables

```
# Tools
```

```
library(reshape2)
```

```
?melt
```

```
?dcast
```

```
?col_split
```

```
library(stringr)
```

```
?str_replace
```

```
?str_sub
```

```
?str_split_fixed
```

```
library(plyr)
```

```
?arrange
```


**Column headers
values, not
variable names**

```
raw <- read.delim("pew.txt", check.names = F,  
stringsAsFactors = F)
```

```
head(raw)
```

```
# What are the variables in this dataset?
```

```
# Discuss with your neighbour for 1 minute
```

```
# Fixing this problem is easy. We use melt, from  
# reshape2, with two arguments, the input data, and  
# the columns which are already variables:
```

```
library(reshape2)  
tidy <- melt(raw, "religion")
```

```
head(tidy)
```

```
# We can now tweak the variable names  
names(tidy) <- c("religion", "income", "n")
```

Multiple variables in one column

```
raw <- read.csv("tb.csv", stringsAsFactors = FALSE)
raw$new_sp <- NULL

names(raw) <- str_replace(names(raw), "new_sp_", "")
```

```
# What are the variables in this dataset?
# Discuss with your neighbour for 1 minute
# Hint: f = female, u = unknown, 1524 = 15-25
```

Your turn

Use melt in the same way as for the religion-income data to get all variables in columns.

Think about how you might separate the "variable" variable into age and sex.

```
# na.rm = TRUE is useful if the missings don't have
# any meaning
clean <- melt(raw, id = c("iso2", "year"),
  na.rm = TRUE)
names(clean)[4] <- "cases"

# Often a good idea to ensure the rows are ordered
# by the variables
clean <- arrange(clean, iso2, variable, year)
```

```
str_sub(clean$variable, 1, 1)
str_sub(clean$variable, 2)
```

```
ages <- c("04" = "0-4", "514" = "5-14", "014" =
"0-14", "1524" = "15-24", "2534" = "25-34", "3544" =
"35-44", "4554" = "45-54", "5564" = "55-64", "65" =
"65+", "u" = NA)
```

```
ages[str_sub(clean$variable, 2)]
```

```
clean$sex <- str_sub(clean$variable, 1, 1)
clean$age <- factor(ages[str_sub(clean$variable, 2)],
  levels = ages)
clean$variable <- NULL
```

```
tidy <- tidy[c("iso2", "year", "sex", "age", "cases")]
```


Variables in rows and columns

```
raw <- read.delim("weather.txt",  
  stringsAsFactors = FALSE)
```

```
# What are the variables in this dataset?  
# Discuss with your neighbour for 1 minute  
# Hint: TMIN = minimum temperature,  
#       id = weather station identifier
```

Your turn

Melt the data, clean variables, and reorder rows and columns.

What do you need to do next?

```
raw1 <- melt(raw, id = 1:4, na.rm = T)
raw1$day <- as.integer(
  str_replace(raw1$variable, "d", ""))
raw1$variable <- NULL
raw1$element <- tolower(raw1$element)

raw1 <- raw1[c("id", "year", "month", "day",
  "element", "value")]
raw1 <- arrange(raw1, year, month, day, element)
```

```
# dcast shifts variables from rows to columns
tidy <- dcast(raw1, ... ~ element)

# casting syntax:
#   row_var1 + row_var2 ~ col_var1 + col_var2
#   ... = all variables not otherwise mentioned
```

**Multiple
types in the
same table**

Your turn

Practice everything you've learned so far to clean up billboard.csv.

(You might want to peek in billboard-encoding.r)

```
raw <- read.csv("billboard.csv",
  stringsAsFactors = F)
raw$date.peaked <- NULL
raw$artist.inverted <- iconv(raw$artist.inverted,
  "MAC", "UTF-8")
raw$track <- str_replace(raw$track,
  " \\(.*?\\)", "")
names(raw)[-1:6] <- str_c(1:76)

tidy <- melt(raw, 1:6, na.rm = T)
tidy$week <- as.integer(tidy$variable)
tidy$variable <- NULL
```



```
# Fix dates (bonus)
library(lubridate)
tidy$date.entered <- ymd(tidy$date.entered)
tidy$date <- tidy$date.entered +
  weeks(tidy$week - 1)
tidy$date.entered <- NULL

# Tidy column names, order and row order
tidy <- rename(tidy, c("value" = "rank",
  "artist.inverted" = "artist"))
tidy <- tidy[c("year", "artist", "track", "time",
  "genre", "week", "date", "rank")]
tidy <- arrange(tidy, year, artist, track, week)
```

Normalisation

Each fact about a song is repeated many many times. Sign that multiple types of experimental unit stored in the same table.

Need to separate out into song and rank tables.

```
song <- unrowname(unique(tidy[c("artist", "track",  
"genre", "time"])))  
song$song_id <- 1:nrow(song)  
  
rank <- join(tidy, song, match = "first")  
rank <- rank[c("song_id", "date", "rank")]
```

One type in multiple tables

```
# Not shown, but easy with ldply
files <- dir("path", pattern = ".csv", full = T)
names(files) <- basename(files)

all <- ldply(files, read.csv)
```

stringr **review**

<http://bit.ly/stringr>

Function	Parameters	Result
<code>str_detect</code>	string, pattern	logical vector
<code>str_locate</code>	string, pattern	numeric matrix
<code>str_extract</code>	string, pattern	character vector
<code>str_replace</code>	string, pattern, replacement	character vector
<code>str_split_fixed</code>	string, pattern	character matrix

Single	Multiple (output usually a list)
<code>str_detect</code>	
<code>str_locate</code>	<code>str_locate_all</code>
<code>str_extract</code>	<code>str_extract_all</code>
<code>str_replace</code>	<code>str_replace_all</code>
<code>str_split_fixed</code>	<code>str_split</code>

Regular expressions

If you work with text data, I highly recommend learning at least a little about regular expressions.

They are complex, but very powerful.



Ceci n'est pas une pipe.

Magritte

```
# Special characters
```

```
a <- "\\\"
```

```
b <- "\"\""
```

```
c <- "a\nb\nc"
```

```
a
```

```
cat(a, "\n")
```

```
b
```

```
cat(b, "\n")
```

```
c
```

```
cat(c, "\n")
```

Special characters

- Use `\` to “escape” special characters
 - `\” = ”`
 - `\n = new line`
 - `\\ = \`
 - `\t = tab`
- ?Quotes for more

Useful tools

- <http://gskinner.com/RegExr/>
- <http://regexp.resource.googlepages.com/analyzer.html>
- <http://www.txt2re.com/>
- <http://www.regular-expressions.info/reference.html>

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