

Data Cleaning

Module 9

Andrew Jaffe

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Data

- We will be using multiple data sets in this lecture:
 - Salary, Monument, Circulator, and Restaurant from OpenBaltimore: <https://data.baltimorecity.gov/browse?limitTo=datasets>
 - Gap Minder - very interesting way of viewing longitudinal data
 - * Data is here - <http://www.gapminder.org/data/>
 - http://spreadsheets.google.com/pub?key=rMsQHawTObBb6_U2ESjKXYw&output=xls

Data Cleaning

In general, data cleaning is a process of investigating your data for inaccuracies, or recoding it in a way that makes it more manageable.

MOST IMPORTANT RULE - LOOK AT YOUR DATA!

Again - `table`, `summarize`, `is.na`, `any`, `all` are useful.

Data Cleaning

```
table(c(0, 1, 2, 3, NA, 3, 3, 2,2, 3),  
      useNA="ifany")
```

```
##  
##    0    1    2    3 <NA>  
##    1    1    3    4     1
```

```
table(c(0, 1, 2, 3, 2, 3, 3, 2,2, 3),  
      useNA="always")
```

```
##  
##    0    1    2    3 <NA>  
##    1    1    4    4     0
```

```
tab <- table(c(0, 1, 2, 3, 2, 3, 3, 2,2, 3),  
             c(0, 1, 2, 3, 2, 3, 3, 4, 4, 3),  
             useNA="always")  
margin.table(tab, 2)
```

```
##
##    0    1    2    3    4 <NA>
##    1    1    2    4    2    0
```

```
prop.table(tab)
```

```
##
##           0    1    2    3    4 <NA>
##    0    0.1 0.0 0.0 0.0 0.0 0.0
##    1    0.0 0.1 0.0 0.0 0.0 0.0
##    2    0.0 0.0 0.2 0.0 0.2 0.0
##    3    0.0 0.0 0.0 0.4 0.0 0.0
##   <NA> 0.0 0.0 0.0 0.0 0.0 0.0
```

```
prop.table(tab,1)
```

```
##
##           0    1    2    3    4 <NA>
##    0    1.0 0.0 0.0 0.0 0.0 0.0
##    1    0.0 1.0 0.0 0.0 0.0 0.0
##    2    0.0 0.0 0.5 0.0 0.5 0.0
##    3    0.0 0.0 0.0 1.0 0.0 0.0
##   <NA>
```

Download Salary FY2014 Data

<https://data.baltimorecity.gov/City-Government/Baltimore-City-Employee-Salaries-FY2014/2j28-xzd7>

Download as a CSV and then read it into R as the variable `Sal`

```
Sal = read.csv("../data/Baltimore_City_Employee_Salaries_FY2014.csv",
               as.is=TRUE)
```

Data Cleaning

- `any()` - checks if there are any TRUES
- `all()` - checks if ALL are true

```
Sal[1:4,]
```

```
##           Name                JobTitle AgencyID
## 1  Aaron,Keontae E             AIDE BLUE CHIP  W02200
## 2  Aaron,Patricia G Facilities/Office Services II A03031
## 3   Aaron,Petra L  ASSISTANT STATE'S ATTORNEY  A29005
## 4 Abaineh,Yohannes T          EPIDEMIOLOGIST  A65026
##           Agency  HireDate AnnualSalary  GrossPay
## 1      Youth Summer 06/10/2013   $11310.00   $873.63
## 2      OED-Employment Dev 10/24/1979   $53428.00 $52868.38
## 3 States Attorneys Office 09/25/2006   $68300.00 $67439.19
## 4  HLTH-Health Department 07/23/2009   $62000.00 $58654.74
```

```
any(is.na(Sal$Name))
```

```
## [1] FALSE
```

Example of Cleaning:

For example, let's say gender was coded as Male, M, m, Female, F, f. Using Excel to find all of these would be a matter of filtering and changing all by hand or using if statements.

In R, you can simply do something like:

```
data$gender[data$gender %in%  
  c("Male", "M", "m")] <- "Male"
```

Sometimes though, it's not so simple. That's where functions that find patterns come in very useful.

```
table(gender)
```

```
## gender  
##      F FeMAle FEMALE      Fm      M      Ma      mAle      Male      MaLe      MALE  
##      75      82      74      89      89      79      87      89      88      95  
##      Man      Woman  
##      73      80
```

Find/Replace and Regular Expressions

- R can do much more than find exact matches for a whole string
- Like Perl and other languages, it can use regular expressions.
- What are regular expressions?
- Ways to search for specific strings
- Can be very complicated or simple
- Highly Useful

‘Find’ functions

grep: `grep`, `grepl`, `regexpr` and `gregexpr` search for matches to argument pattern within each element of a character vector: they differ in the format of and amount of detail in the results.

`grep(pattern, x, fixed=FALSE)`, where:

- `pattern` = character string containing a regular expression to be matched in the given character vector.
- `x` = a character vector where matches are sought, or an object which can be coerced by `as.character` to a character vector.
- If `fixed=TRUE`, it will do exact matching for the phrase anywhere in the vector (regular find)

‘Find’ functions

```
grep("Rawlings",Sal$Name)
```

```
## [1] 13832 13833 13834 13835
```

These are the indices/elements where the pattern match occurs

grep() returns something similar to which() on a logical statement

‘Find’ functions

```
grep("Rawlings",Sal$Name)
```

```
## [1] 13832 13833 13834 13835
```

```
grep("Rawlings",Sal$Name,value=TRUE)
```

```
## [1] "Rawlings,Kellye A"          "Rawlings,MarqWell D"
## [3] "Rawlings,Paula M"          "Rawlings-Blake,Stephanie C"
```

```
Sal[grep("Rawlings",Sal$Name),]
```

```
##              Name              JobTitle AgencyID
## 13832      Rawlings,Kellye A EMERGENCY DISPATCHER  A40302
## 13833      Rawlings,MarqWell D      AIDE BLUE CHIP   W02384
## 13834      Rawlings,Paula M      COMMUNITY AIDE    A04015
## 13835 Rawlings-Blake,Stephanie C      MAYOR      A01001
##              Agency  HireDate AnnualSalary  GrossPay
## 13832 M-R Info Technology 01/06/2003   $47980.00 $68426.73
## 13833 Youth Summer      06/15/2012   $11310.00  $507.50
## 13834 R&P-Recreation    12/10/2007   $19802.00  $8195.79
## 13835 Mayors Office     12/07/1995  $163365.00 $161219.24
```

grep() Options

```
head(grep("Tajhgh",Sal$Name, value=TRUE))
```

```
## [1] "Reynold,Tajhgh J"
```

```
grep("Jaffe",Sal$Name)
```

```
## [1] 8603
```

```
length(grep("Jaffe",Sal$Name))
```

```
## [1] 1
```

A bit on Regular Expressions

- <http://www.regular-expressions.info/reference.html>
- They can use to match a large number of strings in one statement
- . matches any single character
- * means repeat as many (even if 0) more times the last character
- ? makes the last thing optional

Using Regular Expressions

- Look for any name that starts with:
 - Payne at the beginning,
 - Leonard and then an S
 - Spence then a capital C

```
grep("Payne.*", x=Sal$Name, value=TRUE)
```

```
## [1] "Payne El,Jackie"      "Payne Johnson,Nickole A"
## [3] "Payne,Chanel"        "Payne,Connie T"
## [5] "Payne,Denise I"      "Payne,Dominic R"
## [7] "Payne,James R"       "Payne,Jasman T"
## [9] "Payne,Joey D"        "Payne,Jordan A"
## [11] "Payne,Karen V"       "Payne,Karen V"
## [13] "Payne,Leonard S"    "Payne,Mary A"
## [15] "Payne,Micah W"       "Payne,Michael C"
## [17] "Payne,Michael N"     "Payne,Morag"
## [19] "Payne,Nora M"        "Payne,Shelley F"
```

```
grep("Leonard.?S", x=Sal$Name, value=TRUE)
```

```
## [1] "Payne,Leonard S"      "Szumlanski,Leonard S"
```

```
grep("Spence.*C.*", x=Sal$Name, value=TRUE)
```

```
## [1] "Greene,Spencer C"     "Spencer,Charles A"   "Spencer,Christian O"
## [4] "Spencer,Clarence W"   "Spencer,Michael C"
```

Replace

Let's say we wanted to sort the data set by Annual Salary:

```
class(Sal$AnnualSalary)
```

```
## [1] "character"
```

```
sort(c("1", "2", "10")) # not sort correctly (order simply ranks the data)
```

```
## [1] "1" "10" "2"
```

```
order(c("1", "2", "10"))
```

```
## [1] 1 3 2
```

Replace

So we must change the annual pay into a numeric:

```
head(as.numeric(Sal$AnnualSalary), 4)
```

```
## [1] NA NA NA NA
```

R didn't like the \$ so it thought turned them all to NA.

sub() and gsub() can do the replacing part.

Replacing and subbing

Now we can replace the \$ with nothing (used fixed=TRUE because \$ means something in regular expressions):

```
Sal$AnnualSalary <- as.numeric(gsub(pattern="$", replacement="",
                                   Sal$AnnualSalary, fixed=TRUE))
Sal <- Sal[order(Sal$AnnualSalary,decreasing=TRUE), ] # use negative to sort descending
Sal[1:5, c("Name", "AnnualSalary", "JobTitle")]
```

```
##           Name AnnualSalary           JobTitle
## 1222 Bernstein,Gregg L      238772    STATE'S ATTORNEY
## 3175 Charles,Ronnie E      200000 EXECUTIVE LEVEL III
## 985  Batts,Anthony W      193800 EXECUTIVE LEVEL III
## 1343 Black,Harry E        190000 EXECUTIVE LEVEL III
## 16352 Swift,Michael      187200 CONTRACT SERV SPEC II
```

Useful String Functions

Useful String functions

- toupper(), tolower() - uppercase or lowercase your data:
- str_trim() (in the stringr package) - will trim whitespace
- nchar - get the number of characters in a string
- substr(x, start, stop) - substrings from position start to position stop
- strsplit(x, split) - splits strings up - returns list!
- paste() - paste strings together - look at ?paste

Paste

Paste can be very useful for joining vectors together:

```
paste("Visit", 1:5, sep="_")
```

```
## [1] "Visit_1" "Visit_2" "Visit_3" "Visit_4" "Visit_5"
```

```
paste("Visit", 1:5, sep="_", collapse=" ")
```

```
## [1] "Visit_1 Visit_2 Visit_3 Visit_4 Visit_5"
```

```
paste("To", "is going be the ", "we go to the store!", sep="day ")
```

```
## [1] "Today is going be the day we go to the store!"
```

```
# and paste0 can be even simpler see ?paste0  
paste0("Visit",1:5)
```

```
## [1] "Visit1" "Visit2" "Visit3" "Visit4" "Visit5"
```

```
paste(1:5, letters[1:5], sep="_")
```

```
## [1] "1_a" "2_b" "3_c" "4_d" "5_e"
```

```
paste(6:10, 11:15, 2000:2005, sep="/")
```

```
## [1] "6/11/2000" "7/12/2001" "8/13/2002" "9/14/2003" "10/15/2004"  
## [6] "6/11/2005"
```

```
paste(paste("x",1:5,sep=""),collapse="+")
```

```
## [1] "x1+x2+x3+x4+x5"
```

Strsplit

```
x <- c("I really", "like writing", "R code")  
y <- strsplit(x, split=" ")  
y[[2]]
```

```
## [1] "like"    "writing"
```

```
sapply(y, "[", 1) # on the fly
```

```
## [1] "I"      "like" "R"
```

```
sapply(y, "[", 2) # on the fly
```

```
## [1] "really" "writing" "code"
```

Data Merging/Append

- Merging - joining data sets together - usually on key variables, usually “id”
- `merge()` is the most common way to do this with data sets
- `rbind/cbind` - row/column bind, respectively
 - `rbind` is the equivalent of “appending” in Stata or “setting” in SAS
 - `cbind` allows you to add columns in addition to the previous ways
- `reshape2` package also has a lot of information about different ways to reshape data (wide to long, etc)
 - but has a different (and sometimes more intuitive syntax)
- `t()` is a function that will transpose the data

Merging

```
base <- data.frame(id=1:10, Age= seq(55,60, length=10))
base[1:2,]
```

```
##   id      Age
## 1  1 55.00000
## 2  2 55.55556
```

```
visits <- data.frame(id=rep(1:8, 3), visit= rep(1:3, 8),
                     Outcome= seq(10,50, length=24))
visits[1:2,]
```

```
##   id visit Outcome
## 1  1     1 10.00000
## 2  2     2 11.73913
```

```
merged.data <- merge(base, visits, by="id")
merged.data[1:5,]
```

```
##   id      Age visit Outcome
## 1  1 55.00000     1 10.00000
## 2  1 55.00000     3 23.91304
## 3  1 55.00000     2 37.82609
## 4  2 55.55556     2 11.73913
## 5  2 55.55556     1 25.65217
```



```
dim(merged.data)
```

```
## [1] 24 4
```

```
all.data <- merge(base, visits, by="id", all=TRUE)
tail(all.data)
```

```
##      id      Age visit Outcome
## 21  7 58.33333      2 48.26087
## 22  8 58.88889      2 22.17391
## 23  8 58.88889      1 36.08696
## 24  8 58.88889      3 50.00000
## 25  9 59.44444     NA      NA
## 26 10 60.00000     NA      NA
```

```
dim(all.data)
```

```
## [1] 26 4
```

Aside: Dates

You can convert date-like strings in the `Date` class (<http://www.statmethods.net/input/dates.html> for more info)

```
circ = read.csv("../data/Charm_City_Circulator_Ridership.csv", as.is=TRUE)
head(sort(circ$date))
```

```
## [1] "01/01/2011" "01/01/2012" "01/01/2013" "01/02/2011" "01/02/2012"
## [6] "01/02/2013"
```

```
circ$date <- as.Date(circ$date, "%m/%d/%Y") # creating a date for sorting
head(circ$date)
```

```
## [1] "2010-01-11" "2010-01-12" "2010-01-13" "2010-01-14" "2010-01-15"
## [6] "2010-01-16"
```

```
head(sort(circ$date))
```

```
## [1] "2010-01-11" "2010-01-12" "2010-01-13" "2010-01-14" "2010-01-15"
## [6] "2010-01-16"
```

Data Reshaping

Disclaimer: the `reshape` command in R is not remarkably intuitive.

- Wide - multiple measurements are variables / columns so that the data gets wider with more measurements
- Long - multiple measurements are rows so data gets longer with more measurements
- One example would be many ids with multiple visits

Example of Long/Wide

```
head(wide)
```

```
##   id visit1 visit2 visit3
## 1  1   Good   Good   Bad
```

```
head(long)
```

```
##   id visit Outcome
## 1  1     1    Good
## 2  1     2    Good
## 3  1     3    Bad
```

Data Reshaping

- Good resource: <http://www.ats.ucla.edu/stat/r/faq/reshape.htm>

```
head(Indometh) # this is long
```

```
## Grouped Data: conc ~ time | Subject
##   Subject time conc
## 1         1 0.25 1.50
## 2         1 0.50 0.94
## 3         1 0.75 0.78
## 4         1 1.00 0.48
## 5         1 1.25 0.37
## 6         1 2.00 0.19
```

Data Reshaping

```
wide <- reshape(Indometh, v.names = "conc", idvar = "Subject",
                 timevar = "time", direction = "wide")
head(wide)
```

```
##   Subject conc.0.25 conc.0.5 conc.0.75 conc.1 conc.1.25 conc.2 conc.3
## 1         1      1.50      0.94      0.78  0.48      0.37  0.19  0.12
## 12        2      2.03      1.63      0.71  0.70      0.64  0.36  0.32
## 23        3      2.72      1.49      1.16  0.80      0.80  0.39  0.22
## 34        4      1.85      1.39      1.02  0.89      0.59  0.40  0.16
## 45        5      2.05      1.04      0.81  0.39      0.30  0.23  0.13
## 56        6      2.31      1.44      1.03  0.84      0.64  0.42  0.24
##   conc.4 conc.5 conc.6 conc.8
## 1   0.11  0.08  0.07  0.05
## 12  0.20  0.25  0.12  0.08
## 23  0.12  0.11  0.08  0.08
## 34  0.11  0.10  0.07  0.07
## 45  0.11  0.08  0.10  0.06
## 56  0.17  0.13  0.10  0.09
```

Data Reshaping

```
dim(Indometh)
```

```
## [1] 66 3
```

```
wide
```

```
##      Subject conc.0.25 conc.0.5 conc.0.75 conc.1 conc.1.25 conc.2 conc.3
## 1         1         1.50      0.94      0.78  0.48      0.37  0.19  0.12
## 12        2         2.03      1.63      0.71  0.70      0.64  0.36  0.32
## 23        3         2.72      1.49      1.16  0.80      0.80  0.39  0.22
## 34        4         1.85      1.39      1.02  0.89      0.59  0.40  0.16
## 45        5         2.05      1.04      0.81  0.39      0.30  0.23  0.13
## 56        6         2.31      1.44      1.03  0.84      0.64  0.42  0.24
##      conc.4 conc.5 conc.6 conc.8
## 1      0.11  0.08  0.07  0.05
## 12     0.20  0.25  0.12  0.08
## 23     0.12  0.11  0.08  0.08
## 34     0.11  0.10  0.07  0.07
## 45     0.11  0.08  0.10  0.06
## 56     0.17  0.13  0.10  0.09
```

Data Reshaping

- If you've reshaped a data set - to get it back, just reshape it again

```
reshape(wide, direction = "long")[1:10,]
```

```
##      Subject time conc
## 1.0.25      1 0.25 1.50
## 2.0.25      2 0.25 2.03
## 3.0.25      3 0.25 2.72
## 4.0.25      4 0.25 1.85
## 5.0.25      5 0.25 2.05
## 6.0.25      6 0.25 2.31
## 1.0.5       1 0.50 0.94
## 2.0.5       2 0.50 1.63
## 3.0.5       3 0.50 1.49
## 4.0.5       4 0.50 1.39
```

Note the row name change

Data Reshaping - A Better Example

```
TB <- read.xlsx(file="../data/indicator_estimatedincidencealltbper100000.xlsx",
               sheetName="Data")
head(TB, 1)
```

```
## TB.incidence..all.forms..per.100.000.population.per.year. X1990 X1991
## 1 Afghanistan 168 168
## X1992 X1993 X1994 X1995 X1996 X1997 X1998 X1999 X2000 X2001 X2002 X2003
## 1 168 168 168 168 168 168 168 168 168 168 168 168
## X2004 X2005 X2006 X2007 NA.
## 1 168 168 168 168 NA
```

```
TB$NA. <- NULL
head(TB, 1)
```

```
## TB.incidence..all.forms..per.100.000.population.per.year. X1990 X1991
## 1 Afghanistan 168 168
## X1992 X1993 X1994 X1995 X1996 X1997 X1998 X1999 X2000 X2001 X2002 X2003
## 1 168 168 168 168 168 168 168 168 168 168 168 168
## X2004 X2005 X2006 X2007
## 1 168 168 168 168
```

Data Reshaping - A Better Example

```
colnames(TB) <- c("Country", paste("Year",
                                   1990:2007, sep="."))
head(TB,1)
```

```
## Country Year.1990 Year.1991 Year.1992 Year.1993 Year.1994 Year.1995
## 1 Afghanistan 168 168 168 168 168 168
## Year.1996 Year.1997 Year.1998 Year.1999 Year.2000 Year.2001 Year.2002
## 1 168 168 168 168 168 168
## Year.2003 Year.2004 Year.2005 Year.2006 Year.2007
## 1 168 168 168 168 168
```

Data Reshaping - More is better!

```
TB.long <- reshape(TB, idvar="Country",
                   v.names="Cases", times=1990:2007,
                   direction="long", timevar="Year",
                   varying = paste("Year", 1990:2007, sep="."))
head(TB.long, 4)
```

```
## Country Year Cases
## Afghanistan.1990 Afghanistan 1990 168
## Albania.1990 Albania 1990 25
## Algeria.1990 Algeria 1990 38
## American Samoa.1990 American Samoa 1990 21
```

```
rownames(TB.long) <- NULL
head(TB.long, 4)
```

##	Country	Year	Cases
## 1	Afghanistan	1990	168
## 2	Albania	1990	25
## 3	Algeria	1990	38
## 4	American Samoa	1990	21