Data Cleaning

Module 8

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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")
##
##
            1
                      3 <NA>
##
            1
                 3
                      4
                            1
table(c(0, 1, 2, 3, 2, 3, 3, 2,2, 3),
        useNA="always")
##
##
                      3 <NA>
##
            1
                       4
tab \leftarrow 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)
```

```
##
##
                           4 <NA>
      0
           1
                2
                      3
##
prop.table(tab)
##
##
                     2
                         3
                             4 <NA>
##
          0.1 0.0 0.0 0.0 0.0
                                0.0
          0.0 0.1 0.0 0.0 0.0
##
     1
##
     2
          0.0 0.0 0.2 0.0 0.2
##
          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)
##
##
                     2
                         3
                             4 <NA>
                               0.0
##
     0
          1.0 0.0 0.0 0.0 0.0
##
          0.0 1.0 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-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2ity-Employee-FY2014/2j28-xzd7-2i

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

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
                           ASSISTANT STATE'S ATTORNEY
## 3
          Aaron, Petra L
                                                         A29005
## 4 Abaineh, Yohannes T
                                       EPIDEMIOLOGIST
                                                         A65026
##
                              HireDate AnnualSalary
                                                        GrossPay
                       Agency
## 1
               Youth Summer
                              06/10/2013
                                             $11310.00
                                                         $873.63
          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"</pre>
```

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
                                         М
                                                      mAle
                                                              Male
                                                                      MaLe
                                                                              MALE
                                                Ma
##
        75
                82
                        74
                                89
                                        89
                                                79
                                                        87
                                                                        88
                                                                                95
                                                                89
##
      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
                Rawlings, MarqWell D
## 13833
                                           AIDE BLUE CHIP
                                                            W02384
                                           COMMUNITY AIDE
## 13834
                   Rawlings, Paula M
                                                            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"
                                     "Payne, Morag"
## [17] "Payne, Michael N"
## [19] "Payne, Nora M"
                                     "Payne, Shelley F"
```

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")]
```

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

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

[1] "like"

"writing"

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=" ")</pre>
y[[2]]
```

```
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),</pre>
                    Outcome= seq(10,50, length=24))
visits[1:2,]
     id visit Outcome
          1 10.00000
## 1 1
## 2 2
            2 11.73913
merged.data <- merge(base, visits, by="id")</pre>
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)</pre>
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"

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

## [6] "2010-01-16"</pre>
```

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
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
```

Data Reshaping

```
##
      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
            2
                                      0.71
                                              0.70
## 12
                   2.03
                            1.63
                                                        0.64
                                                               0.36
                                                                      0.32
## 23
            3
                                              0.80
                                                        0.80
                   2.72
                            1.49
                                      1.16
                                                               0.39
                                                                      0.22
## 34
            4
                   1.85
                            1.39
                                      1.02
                                              0.89
                                                        0.59
                                                                      0.16
                                                               0.40
                   2.05
## 45
            5
                            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
                      0.07
                             0.07
        0.11
               0.10
## 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
                                                                        0.22
                                                                 0.39
## 34
            4
                    1.85
                             1.39
                                       1.02
                                               0.89
                                                         0.59
                                                                 0.40
                                                                        0.16
            5
                    2.05
                                       0.81
## 45
                             1.04
                                               0.39
                                                         0.30
                                                                 0.23
                                                                        0.13
                   2.31
                                       1.03
                                                         0.64
                                                                        0.24
## 56
            6
                             1.44
                                               0.84
                                                                 0.42
##
      conc.4 conc.5 conc.6 conc.8
## 1
        0.11
               0.08
                      0.07
                              0.05
        0.20
                      0.12
                              0.08
## 12
               0.25
                      0.08
## 23
        0.12
               0.11
                              0.08
## 34
               0.10
                      0.07
                              0.07
        0.11
## 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 2.03
## 2.0.25
## 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.incidence..all.forms..per.100.000.population.per.year. X1990 X1991
## 1
                                                    Afghanistan
                                                                   168
    X1992 X1993 X1994 X1995 X1996 X1997 X1998 X1999 X2000 X2001 X2002 X2003
##
       168
             168
                   168
                         168
                               168
                                      168
                                            168
                                                  168
                                                        168
                                                              168
                                                                    168
                                                                           168
    X2004 X2005 X2006 X2007 NA.
       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
    X1992 X1993 X1994 X1995 X1996 X1997 X1998 X1999 X2000 X2001 X2002 X2003
##
## 1
                               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",</pre>
                           1990:2007, sep="."))
head(TB,1)
         Country Year.1990 Year.1991 Year.1992 Year.1993 Year.1994 Year.1995
                                  168
                                             168
                                                       168
## 1 Afghanistan
                        168
                                                                  168
                                                                            168
     Year.1996 Year.1997 Year.1998 Year.1999 Year.2000 Year.2001 Year.2002
                      168
                                168
                                           168
                                                     168
                                                                168
     Year.2003 Year.2004 Year.2005 Year.2006 Year.2007
##
## 1
           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
## American Samoa.1990 American Samoa 1990
                                                21
rownames(TB.long) <- NULL</pre>
head(TB.long, 4)
```

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