

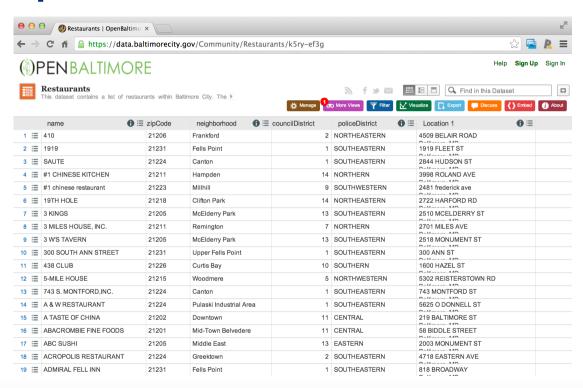
Creating new variables

Jeffrey Leek Johns Hopkins Bloomberg School of Public Health

Why create new variables?

- · Often the raw data won't have a value you are looking for
- · You will need to transform the data to get the values you would like
- · Usually you will add those values to the data frames you are working with
- Common variables to create
 - Missingness indicators
 - "Cutting up" quantitative variables
 - Applying transforms

Example data set



https://data.baltimorecity.gov/Community/Restaurants/k5ry-ef3g

Getting the data from the web

```
if(!file.exists("./data")){dir.create("./data")}
fileUrl <- "https://data.baltimorecity.gov/api/views/k5ry-ef3g/rows.csv?accessType=DOWNLOAD"
download.file(fileUrl,destfile="./data/restaurants.csv",method="curl")
restData <- read.csv("./data/restaurants.csv")</pre>
```

Creating sequences

Sometimes you need an index for your data set

```
s1 <- seq(1,10,by=2) ; s1
```

```
[1] 1 3 5 7 9
```

```
s2 <- seq(1,10,length=3); s2
```

```
[1] 1.0 5.5 10.0
```

```
x <- c(1,3,8,25,100); seq(along = x)
```

```
[1] 1 2 3 4 5
```

Subsetting variables

```
restData$nearMe = restData$neighborhood %in% c("Roland Park", "Homeland")
table(restData$nearMe)
```

```
FALSE TRUE
1314 13
```

Creating binary variables

```
restData$zipWrong = ifelse(restData$zipCode < 0, TRUE, FALSE)
table(restData$zipWrong,restData$zipCode < 0)
```

```
FALSE TRUE
FALSE 1326 0
TRUE 0 1
```

Creating categorical variables

```
restData$zipGroups = cut(restData$zipCode,breaks=quantile(restData$zipCode))
table(restData$zipGroups)
```

```
(-2.123e+04,2.12e+04] (2.12e+04,2.122e+04] (2.122e+04,2.123e+04] (2.123e+04,2.129e+04] 337 375 282 332
```

table(restData\$zipGroups,restData\$zipCode)

21214 21215 21216 21217 21218 21220 21222 21223 21224 21225 21226 2122/5

Easier cutting \square

```
library(Hmisc)
restData$zipGroups = cut2(restData$zipCode,g=4)
table(restData$zipGroups)
```

```
[-21226,21205) [ 21205,21220) [ 21220,21227) [ 21227,21287]
338 375 300 314
```

Creating factor variables

```
restData$zcf <- factor(restData$zipCode)
restData$zcf[1:10]</pre>
```

```
[1] 21206 21231 21224 21211 21223 21218 21205 21211 21205 21231

32 Levels: -21226 21201 21202 21205 21206 21207 21208 21209 21210 21211 21212 21213 21214 ... 21287
```

```
class(restData$zcf)
```

```
[1] "factor"
```

Levels of factor variables

```
yesno <- sample(c("yes", "no"), size=10, replace=TRUE)
yesnofac = factor(yesno, levels=c("yes", "no"))
relevel(yesnofac, ref="yes")</pre>
```

```
[1] yes yes yes no yes yes no no
Levels: yes no
```

```
as.numeric(yesnofac)
```

```
[1] 1 1 1 1 2 1 1 1 2 2
```

Cutting produces factor variables

```
library(Hmisc)
restData$zipGroups = cut2(restData$zipCode,g=4)
table(restData$zipGroups)
```

```
[-21226,21205) [ 21205,21220) [ 21220,21227) [ 21227,21287]
338 375 300 314
```

Using the mutate function

```
library(Hmisc); library(plyr)
restData2 = mutate(restData,zipGroups=cut2(zipCode,g=4))
table(restData2$zipGroups)
```

```
[-21226,21205) [ 21205,21220) [ 21220,21227) [ 21227,21287]
338 375 300 314
```

Common transforms

- · abs(x) absolute value
- sqrt(x) square root
- ceiling(x) ceiling(3.475) is 4
- · floor(x) floor(3.475) is 3
- round(x,digits=n) roun(3.475,digits=2) is 3.48
- signif(x,digits=n) signif(3.475,digits=2) is 3.5
- \cdot cos(x), sin(x) etc.
- · log(x) natural logarithm
- · log2(x), log10(x) other common logs
- exp(x) exponentiating x

http://www.biostat.jhsph.edu/~ajaffe/lec_winterR/Lecture%202.pdf http://statmethods.net/management/functions.html

Notes and further reading

- · A tutorial from the developer of plyr http://plyr.had.co.nz/09-user/
- · Andrew Jaffe's R notes http://www.biostat.jhsph.edu/~ajaffe/lec_winterR/Lecture%202.pdf