# Data Cleaning in R

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### Common Tasks

- Data class
- Creating dataframes
- Updating column names
- Combining columns
- ► Transposing, merging, long/wide
- ► Rownames/column names
- Concatenating
- Creating lists
- Subsetting (indices, logic, which)
- Dealing with NA values w/ logic
- Strings
- Functions
- lapply

# Basic background

- dataframe\$column
- read.csv("data.csv")
- write.csv(object, "object.csv")
- ► head()
- str()
- colnames()
- summary()
- unique()
- class()
- ► levels()
- nrow(), length()
- dataframe[row, column]

# **Packages**

```
library(lubridate)
library(tidyr)
library(tibble)
```

#### Structure

[1] "Date"

```
as.Date, as.String, as.Numeric, as.Factor use class() to
find type
class(sales$event dt)
[1] "factor"
sales$event_dt <- as.Date(sales$event_dt)</pre>
class(sales$event_dt)
```

# Splitting columns

```
sales$event_dt[1:3]
[1] "2015-09-12" "2009-09-04" "2006-04-21"
strsplit(as.character(sales$event_dt[1:3]), "-")
\lceil \lceil 1 \rceil \rceil
[1] "2015" "09"
                     "12"
[[2]]
[1] "2009" "09"
                     "04"
[[3]]
[1] "2006" "04"
                     "21"
```

# Creating new dataframes

```
new.df <- data.frame(matrix(nrow=5, ncol=3))</pre>
```

### Updating column names

# Renaming singular column

```
colnames(new.df[colnames(new.df) == "column.2"])
<- "whatever"</pre>
```

# Combining columns

```
[1] "1a" "2b" "3c" "4d" "5e"
```

## Long/wide

```
long <- gather(new.df, key, value)
long[1:3,]</pre>
```

```
key value
1 column.1 1
2 column.1 2
3 column.1 3
```

```
long$ID <- rep(seq(1:5),3)
wide <- spread(long, key, value)
wide[1:3,2:4]</pre>
```

# Transposing

t() returns a nested list. A dataframe must be specified if you want a dataframe.

```
new.df.t <- as.data.frame(t(new.df))
new.df.t</pre>
```

```
V1 V2 V3 V4 V5 column.1 1 2 3 4 5 column.2 a b c d e column.3 1a 2b 3c 4d 5e
```

### Rownames/colnames

```
require(tibble)
rownames_to_column(new.df.t)
```

```
rowname V1 V2 V3 V4 V5
1 column.1 1 2 3 4 5
2 column.2 a b c d e
3 column.3 1a 2b 3c 4d 5e
```

### Concatenating dataframes

```
new.df.2 <- new.df
rbind(new.df, new.df.2)</pre>
```

```
column.1 column.2 column.3
                               1a
                      а
                               2b
                               Зс
                      d
                               4d
5
           5
                               5e
                      е
                               1a
                      a
                      b
                               2b
                               Зс
9
                      d
                               4d
10
           5
                               5e
```

# Creating new lists

```
mylist <- rep(NA, nrow(sales)/500)
mylist</pre>
```

[1] NA NA NA NA NA NA NA NA NA

### Subsetting by indices

R makes use of the [row,column] notation

```
sales[5:10, c(3,9)]
```

```
primary_act_id major_cat_name
5 91c03a34b562436efa3c MISC
6 ac4b847b3fde66f2117e MISC
7 a14232befff04be1e2f3 MISC
8 91c03a34b562436efa3c MISC
9 0efaba7ce3f0d7466b42 MISC
10 f6425a3223e73ea6de5a CONCERTS
```

## Subsetting by >=, ==, <=, !=

R accepts logical statements within a [row, column] subsetting argument

	age_yr	<pre>tickets_purchased_qty</pre>	trans_face_val_amt
11	80	1	20
428	82	1	20
1220	80	1	20
1314	76	1	30
1701	72	2	30
3830	72	1	22
4188	90	1	15
4328	94	1	9
4975	78	1	20

# Subsetting by %in%

```
event dt venue state
732 2015-08-08 RHODE ISLAND
1860 2015-09-17 MANITOBA
2046 2016-02-19 MANITOBA
2177 2015-12-16 RHODE ISLAND
2832 2015-09-17 MANITOBA
3304 2015-09-17 MANITOBA
4064 2016-01-27 MANITOBA
4119 2015-09-17 MANITOBA
4308 2015-11-16 MANTTOBA
4935 2016-01-15 MANTTOBA
```

## Subsetting by which

If you have 500 columns, you may not know which column index to subset by in the previous example. Here the age column is extracted using which.

	tickets_purchased_qty	${\sf age\_yr}$
11	1	80
428	1	82
1220	1	80
1314	1	76
1701	2	72
3830	1	72
4188	1	90
4328	1	94
4975	1	78

# Which.max, which.min

which.max and which.min are useful for removing known outliers.

```
sales[-c(which.max(sales$income_amt)),]
```

### Merging

```
merged <- merge(x1, x2, by.x="column.x", by.y="column.y")</pre>
```

# Reclassifying

```
[1] "(-0.52,305]" "(305,609]" "(609,913]
[4] "(913,1.22e+03]" "(1.22e+03,1.52e+03]"
```

levels(cut(sales\$trans\_face\_val\_amt, 5))

# Removing NA values

```
new.df[2,3] <- NA
new.df[1,2] <- NA
new.df[1,3] <- NA
new.df
```

```
column.1 column.2 column.3
                 <NA>
                           < NA >
                    b
                           < NA >
3
          3
                             Зс
4
          4
                    d
                             4d
5
          5
                             5e
                    е
```

### Na.omit

```
na.omit(new.df)
```

### Column NAs

### Row NAs

# Converting date/time

### lubridate package

```
year()
month()
day()
week()
as_date()
as_datetime()
time_length()
```

# Lubridate examples

```
library(lubridate)
dates <- as.Date(sales$event_dt[1:3])</pre>
dates
[1] "2015-09-12" "2009-09-04" "2006-04-21"
year(dates)
[1] 2015 2009 2006
month(dates)
[1] 9 9 4
```

[1] 37 36 16

week(dates)

## More examples

```
floor_date(dates, "month")
[1] "2015-09-01" "2009-09-01" "2006-04-01"
ceiling_date(dates, "season")
[1] "2015-12-01" "2009-12-01" "2006-06-01"
date1 <- "2009-08-03 12:01:59"
as.Date(date1) # uh-oh
[1] "2009-08-03"
as datetime(date1)
[1] "2009-08-03 12:01:59 UTC"
```

### Strings

```
grepl returns a logical TRUE/FALSE
```

```
files <- files[grepl('name', files) == TRUE]</pre>
```

```
gsub("hello", "goodbye", files)
```

# Creating functions

```
myfunction <- function(x) {
  z <- x + 1
  return(z)
}
myfunction(3)</pre>
```

[1] 4

# Function to calculate percentage of NA in columns

```
calc.na <- function(x, data) {
    calc <- sum(is.na(data[[x]])/nrow(data))
    return(unlist(calc))
}
calc.na("onsale_dt", sales)</pre>
```

[1] 0.0202

# What about the amount of every column?

#### Option 1: For loop

Problem 1: results duplicated Problem 2: returns list indices as column names, which cannot be iterated over or used in the future.

```
results <- rep(NA, length(colnames(sales)))

for (i in colnames(sales)) {
  results[i] <- calc.na(i, sales)
}

results[60]</pre>
```

```
tickets_purchased_qty
```

### Option 2: lapply

```
unlist(lapply(colnames(sales), calc.na, sales))
```

```
[11] 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 [21] 0.0008 0.0000 0.0848 0.0000 0.0000 0.0000 0.0000 0.484 [31] 0.0000 0.0000 0.9646 0.9664 0.9692 0.9722 0.9776 0.985 [41] 0.9740 0.9724 0.9842 0.9846 0.9874 0.9354
```

[1] 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.0000 0.68

### Function - example 2

```
toMatch <- c("x", "y", "z", "...")
sentences <- c("sentence 1", "sentence 2", "...")
subset_sentences <- function(Match, sentences){
    sentences[grep(Match,sentences)]
}
subsetted <- lapply(toMatch, subset_sentences, sentences)</pre>
```

#### Example

Example of creating list of dates to loop over

```
dates <- seq(ymd_hms('2018-03-08 00:00:00'),
              ymd hms('2018-03-12 23:00:00'),
              by="1 hour")
dates <- as.character(dates)</pre>
for (i in seq along(dates)) {
  dates[i] <- gsub(" ", "T", dates[i])</pre>
dates[1:5]
```

```
[1] "2018-03-08T00:00:00" "2018-03-08T01:00:00" "2018-03-08
```

[4] "2018-03-08T03:00:00" "2018-03-08T04:00:00"

#### Read in RDS

Data is saved as an RDS because it is a recursively nested list

```
require(lubridate)
require(tidyr)
weather <- readRDS("data/scraped_data.rds")
class(weather)</pre>
```

```
[1] "matrix"
```

#### Data

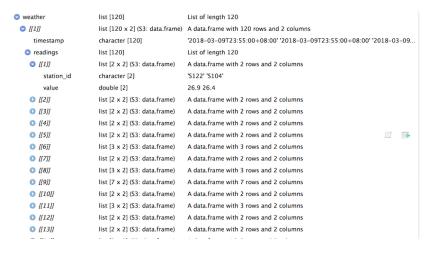


Figure 1:

### Cleaning

```
weather.times <- weather[seq_along(weather) %% 2 > 0]
weather.times <- unlist(weather.times)

weather.readings <- weather[seq_along(weather) %% 2 == 0]
for (i in c(1:length(weather.readings))) {
   weather.readings[[i]][[1]][3] <- i
}

weather.readings <- do.call("rbind", weather.readings)
weather.readings <- do.call("rbind", weather.readings)</pre>
```

#### Continued

# Finished

S116 <sup>‡</sup>	<b>S117</b> <sup>‡</sup>	\$121 <sup>‡</sup>	\$122 <sup>‡</sup>	\$43 <sup>‡</sup>	\$44 <sup>‡</sup>	S50 <sup>‡</sup>	S60 <sup>‡</sup>	date
NA	NA	NA	24.3	NA	NA	NA	NA	2018-03-07T23:55:00+08:00
NA	NA	NA	24.0	NA	NA	NA	NA	2018-03-08T00:55:00+08:00
NA	NA	NA	24.4	NA	24.6	NA	NA	2018-03-08T01:55:00+08:00
NA	NA	NA	24.5	NA	NA	NA	NA	2018-03-08T02:55:00+08:00
NA	NA	NA	24.2	NA	NA	NA	NA	2018-03-08T03:55:00+08:00
NA	NA	NA	24.3	NA	NA	NA	NA	2018-03-08T04:55:00+08:00
NA	NA	NA	23.7	25.7	24.7	24.9	25.6	2018-03-08T05:55:00+08:00
NA	NA	NA	23.8	NA	NA	NA	NA	2018-03-08T06:55:00+08:00
NA	NA	NA	24.8	NA	NA	NA	NA	2018-03-08T07:55:00+08:00
NA	NA	NA	24.8	NA	NA	NA	NA	2018-03-08T07:55:00+08:00
28.0	28.3	27.4	28.7	NA	27.3	28.6	NA	2018-03-08T09:55:00+08:00
NA	NA	NA	30.5	NA	NA	NA	NA	2018-03-08T10:55:00+08:00
NA	NA	NA	30.1	28.9	NA	NA	NA	2018-03-08T11:55:00+08:00
NA	NA	NA	28.8	NA	NA	NA	NA	2018-03-08T12:55:00+08:00
NA	NA	NA	27.7	NA	NA	NA	NA	2018-03-08T13:55:00+08:00
NA	NA	NA	27.5	NA	NA	NA	NA	2018-03-08T14:55:00+08:00
NA	28.6	27.6	27.4	26.7	NA	27.1	27.9	2018-03-08T15:55:00+08:00
NA	NA	NA	28.1	NA	NA	NA	NA	2018-03-08T16:55:00+08:00
NA	NA	28.2	28.2	NA	NA	NA	NA	2018-03-08T17:55:00+08:00

Figure 2:

#### Join metadata

```
weather.metadata <- read.csv("data/weather.metadata.csv")</pre>
weather.melted <- gather(weather.readings, key,</pre>
                            value, -date, -V3)
weather.joined <- merge(weather.melted,</pre>
                          weather.metadata.
                           bv.x="key", by.y="id")
weather.joined$date <- unlist(weather.joined$date)</pre>
weather.joined$day <- as date(weather.joined$date)</pre>
```

### Done!

```
weather.joined[1:5,c(10,4,8,9)]
```

	day	value	${\tt location.latitude}$	location.longitude
1	2018-01-04	29.3	1.4172	103.7485
2	2 2018-01-02	NA	1.4172	103.7485
3	3 2018-01-03	25.7	1.4172	103.7485
4	2018-01-03	25.5	1.4172	103.7485
5	2018-01-04	29.3	1.4172	103.7485

