

Preprocessing your data with R

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Missing data.

```
?is.na
example <- c("A",1,6,7,NA,"B")
example
```

```
## [1] "A" "1" "6" "7" NA  "B"
```

```
mean(example)
```

```
## Warning in mean.default(example): argument is not numeric or logical:
## returning NA
```

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

```
is.na(example)
```

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

```
example2 <- c(2,1,6,7,NA,4)
example2
```

```
## [1]  2  1  6  7 NA  4
```

```
is.na(example2)
```

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

```
mean(example2)
```

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

```
length(example2)
```

```
## [1] 6
```

```
mean(example2,na.rm=TRUE)
```

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

```
median(example2,na.rm=TRUE)
```

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

Replacing just the missing values

```
example3 <- example2  
example3
```

```
## [1] 2 1 6 7 NA 4
```

```
example3[is.na(example3)] <- 0  
example3
```

```
## [1] 2 1 6 7 0 4
```

```
length(example3)
```

```
## [1] 6
```

Missing values can behave strangely

| | |
|-----------|--|
| NA == NA | |
| ## [1] NA | |
| NA+8 | |
| ## [1] NA | |
| NA^0 | |
| ## [1] 1 | |
| 1/NA | |
| ## [1] NA | |

Other strange values...

| |
|------------|
| 1/0 |
| ## [1] Inf |
| 1/0-1/0 |
| ## [1] NaN |

Data imputation

```
library("mice")
data(mammalsleep)
?mammalsleep
dim(mammalsleep)
```

```
## [1] 62 11
```

```
nic(mammalsleep)
```

```
## [1] 20
```

```
md.pattern(mammalsleep)
```

| ## | species | bw | brw | pi | sei | odi | ts | mls | gt | ps | sws |
|-------|---------|----|-----|----|-----|-----|----|-----|----|----|-----|
| ## 42 | | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 0 |
| ## 2 | | 1 | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 1 |
| ## 3 | | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 |
| ## 9 | | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 1 | 0 | 2 |
| ## 2 | | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 1 | 2 |
| ## 1 | | 1 | 1 | 1 | 1 | 1 | 1 | 0 | 0 | 1 | 2 |
| ## 2 | | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 1 | 0 | 3 |
| ## 1 | | 1 | 1 | 1 | 1 | 1 | 1 | 0 | 1 | 0 | 3 |
| ## | | 0 | 0 | 0 | 0 | 0 | 4 | 4 | 4 | 12 | 38 |

Missing at random (MCAR) versus systematic patterns (MNAR).

```
?mice
```

Outlier detection

```
summary(mammalsleep)
```

```
##           species      bw      brw
## African elephant      : 1  Min.   :  0.005  Min.   :  0.14
## African giant pouched rat: 1 1st Qu.:  0.600 1st Qu.:  4.25
## Arctic Fox              : 1  Median :  3.342 Median : 17.25
## Arctic ground squirrel  : 1  Mean    : 198.790 Mean    : 283.13
## Asian elephant          : 1 3rd Qu.: 48.203 3rd Qu.: 166.00
## Baboon                  : 1  Max.    :6654.000 Max.    :5712.00
## (Other)                  :56
##      sws      ps      ts      mls
## Min.   : 2.100  Min.   :0.000  Min.   : 2.60  Min.   : 2.000
## 1st Qu.: 6.250 1st Qu.:0.900 1st Qu.: 8.05 1st Qu.: 6.625
## Median : 8.350 Median :1.800 Median :10.45 Median : 15.100
## Mean    : 8.673 Mean    :1.972 Mean    :10.53 Mean    : 19.878
## 3rd Qu.:11.000 3rd Qu.:2.550 3rd Qu.:13.20 3rd Qu.: 27.750
## Max.    :17.900 Max.    :6.600 Max.    :19.90 Max.    :100.000
## NA's    :14    NA's    :12    NA's    :4    NA's    :4
##      gt      pi      sei      odi
## Min.   : 12.00  Min.   :1.000  Min.   :1.000  Min.   :1.000
## 1st Qu.: 35.75 1st Qu.:2.000 1st Qu.:1.000 1st Qu.:1.000
## Median : 79.00 Median :3.000 Median :2.000 Median :2.000
## Mean    :142.35 Mean    :2.871 Mean    :2.419 Mean    :2.613
## 3rd Qu.:207.50 3rd Qu.:4.000 3rd Qu.:4.000 3rd Qu.:4.000
## Max.    :645.00 Max.    :5.000 Max.    :5.000 Max.    :5.000
## NA's    :4
```

```
which.max(mammalsleep$bw)
```

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

```
mammalsleep[which.max(mammalsleep$bw),]
```

```
##           species  bw  brw sws ps  ts  mls  gt pi sei odi
## 1 African elephant 6654 5712  NA  NA 3.3 38.6 645  3  5  3
```

Document them, find the reason they occurred, then remove them.

Make the data easier to look at interactively

```
View(pressure)  
View(iris)
```


Grouping Data

```
load('births.RData')
head(births)
```

```
##   year month date_of_month day_of_week births
## 1 2000     1              1           6   9083
## 2 2000     1              2           7   8006
## 3 2000     1              3           1  11363
## 4 2000     1              4           2  13032
## 5 2000     1              5           3  12558
## 6 2000     1              6           4  12466
```

```
birthn <- births
save(birthn, file="birthn.RData")
```

Different ways of filtering the data

Choosing only the Saturday births.

```
###Subsetting
Sat <- birthn[birthn$day_of_week==6,]
Sat[1:5,]
```

```
##   year month date_of_month day_of_week births
## 1  2000     1             1           6   9083
## 8  2000     1             8           6   8934
## 15 2000     1            15           6   8525
## 22 2000     1            22           6   8855
## 29 2000     1            29           6   8805
```

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
```

```
## The following objects are masked from 'package:stats':
##
##   filter, lag
```

```
## The following objects are masked from 'package:base':
##
##   intersect, setdiff, setequal, union
```

```
Sat1 <- filter(birthn, day_of_week == 6)
Sat1[1:5,]
```

```
##   year month date_of_month day_of_week births
## 1  2000     1             1           6   9083
## 2  2000     1             8           6   8934
## 3  2000     1            15           6   8525
## 4  2000     1            22           6   8855
## 5  2000     1            29           6   8805
```

```
Sat2 <- birthn %>% filter(day_of_week == 6)
Sat2[1:5,]
```

```
##   year month date_of_month day_of_week births
## 1  2000     1             1           6   9083
## 2  2000     1             8           6   8934
## 3  2000     1            15           6   8525
## 4  2000     1            22           6   8855
## 5  2000     1            29           6   8805
```

Another way of looking at data is to make them into what is called a tibble: (tbl: tibble).

tbl s have the advantage of always showing themselves in the console optimally.

tbl_df gives similar information as str we have been using.

```
tbl_df(Sat1)
```

```
## # A tibble: 783 × 5
##   year month date_of_month day_of_week births
##   <int> <int>         <int>         <int> <int>
## 1  2000     1             1             6  9083
## 2  2000     1             8             6  8934
## 3  2000     1            15             6  8525
## 4  2000     1            22             6  8855
## 5  2000     1            29             6  8805
## 6  2000     2             5             6  8624
## 7  2000     2            12             6  8836
## 8  2000     2            19             6  8861
## 9  2000     2            26             6  9026
## 10 2000     3             4             6  9054
## # ... with 773 more rows
```

```
str(Sat1)
```

```
## 'data.frame':    783 obs. of  5 variables:
## $ year          : int  2000 2000 2000 2000 2000 2000 2000 2000 2000 2000 ...
## $ month         : int   1 1 1 1 1 2 2 2 2 3 ...
## $ date_of_month: int   1 8 15 22 29 5 12 19 26 4 ...
## $ day_of_week  : int   6 6 6 6 6 6 6 6 6 6 ...
## $ births       : int  9083 8934 8525 8855 8805 8624 8836 8861 9026 9054 ...
```

Sequences of Transformations

The %>% operator helps when we are doing several nested operations.

Here is an example

```
GroupBirths <- group_by(birthn, day_of_week)
GroupMeans <- summarise(GroupBirths, mean(births))
SortedBirths <- arrange(GroupMeans, `mean(births)`)
SortedBirths
```

```
## # A tibble: 7 × 2
##   day_of_week `mean(births)`
##       <int>         <dbl>
## 1         7      7518.377
## 2         6      8562.573
## 3         1     11897.830
## 4         5     12596.162
## 5         4     12845.826
## 6         3     12910.766
## 7         2     13122.444
```

```
str(SortedBirths)
```

```
## Classes 'tbl_df', 'tbl' and 'data.frame':   7 obs. of  2 variables:
## $ day_of_week : int  7 6 1 5 4 3 2
## $ mean(births): num 7518 8563 11898 12596 12846 ...
```

```
birthn %>%
  group_by(day_of_week) %>%
  summarise(avg = mean(births)) %>%
  arrange(avg)
```

```
## # A tibble: 7 × 2
##   day_of_week    avg
##       <int>    <dbl>
## 1         7 7518.377
## 2         6 8562.573
## 3         1 11897.830
## 4         5 12596.162
## 5         4 12845.826
## 6         3 12910.766
## 7         2 13122.444
```

```
####More succinctly
birthn %>%
  group_by(day_of_week) %>%
  summarise(mean(births)) %>%
  arrange()
```

```
## # A tibble: 7 × 2
##   day_of_week `mean(births)`
##       <int>         <dbl>
## 1         1     11897.830
## 2         2     13122.444
```

```
## 3      3      12910.766
## 4      4      12845.826
## 5      5      12596.162
## 6      6       8562.573
## 7      7       7518.377
```

`x %>% f(y)` is equivalent to just executing `f(x,y)`

If we need to execute a sequence of functions: `h(g(f(x,y),z),m)` can be hard to parse and read.

`x %>% f(y) %>% g(z) %>% h(m)` gives the same answer.

To find out the average of Friday 13th births:

```
birthn %>%
  filter(day_of_week == 5) %>%
  filter(date_of_month == 13) %>%
  summarise(mean(births))
```

```
## mean(births)
## 1      11949.96
```

```
birthn %>%
  filter(day_of_week < 5) %>%
  filter(date_of_month != 13) %>%
  summarise(mean(births))
```

```
## mean(births)
## 1      12700.61
```

Bad Drivers Data

Five Thirty Eight Article

You need an internet connection for this to work:

```
drivers <- read.csv(url("https://raw.githubusercontent.com/fivethirtyeight/data/master/bad-drivers/bad-drivers.csv"))
head(drivers)
```

| | | |
|------|--|---------|
| ## | State | |
| ## 1 | Alabama | |
| ## 2 | Alaska | |
| ## 3 | Arizona | |
| ## 4 | Arkansas | |
| ## 5 | California | |
| ## 6 | Colorado | |
| ## | Number.of.drivers.involved.in.fatal.collisions.per.billion.miles | |
| ## 1 | | 18.8 |
| ## 2 | | 18.1 |
| ## 3 | | 18.6 |
| ## 4 | | 22.4 |
| ## 5 | | 12.0 |
| ## 6 | | 13.6 |
| ## | Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Speeding | |
| ## 1 | | 39 |
| ## 2 | | 41 |
| ## 3 | | 35 |
| ## 4 | | 18 |
| ## 5 | | 35 |
| ## 6 | | 37 |
| ## | Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Alcohol.Impaired | |
| ## 1 | | 30 |
| ## 2 | | 25 |
| ## 3 | | 28 |
| ## 4 | | 26 |
| ## 5 | | 28 |
| ## 6 | | 28 |
| ## | Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Not.Distracted | |
| ## 1 | | 96 |
| ## 2 | | 90 |
| ## 3 | | 84 |
| ## 4 | | 94 |
| ## 5 | | 91 |
| ## 6 | | 79 |
| ## | Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Had.Not.Been.Involved.In.Any.Previous.Accidents | |
| ## 1 | | 80 |
| ## 2 | | 94 |
| ## 3 | | 96 |
| ## 4 | | 95 |
| ## 5 | | 89 |
| ## 6 | | 95 |
| ## | Car.Insurance.Premiums.... | |
| ## 1 | | 784.55 |
| ## 2 | | 1053.48 |
| ## 3 | | 899.47 |
| ## 4 | | 827.34 |
| ## 5 | | 878.41 |
| ## 6 | | 835.50 |
| ## | Losses.incurred.by.insurance.companies.for.collisions.per.insured.driver.... | |
| ## 1 | | 145.08 |
| ## 2 | | 133.93 |
| ## 3 | | 110.35 |
| ## 4 | | 142.39 |

| | |
|------|--------|
| ## 5 | 165.63 |
| ## 6 | 139.91 |

```
tbl_df(drivers)
```

```
## # A tibble: 51 × 8
##       State
##       <fctr>
## 1      Alabama
## 2      Alaska
## 3      Arizona
## 4      Arkansas
## 5      California
## 6      Colorado
## 7      Connecticut
## 8      Delaware
## 9 District of Columbia
## 10     Florida
## # ... with 41 more rows, and 7 more variables:
## #   Number.of.drivers.involved.in.fatal.collisions.per.billion.miles <dbl>,
## #   Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Speeding <int>,
## #   Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Alcohol.Impaired <int>,
## #   Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Not.Distracted <int>,
## #   Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Had.Not.Been.Involved.In.Any.Previous.Accident
## #   Car.Insurance.Premiums.... <dbl>,
## #   Losses.incurred.by.insurance.companies.for.collisions.per.insured.driver.... <dbl>
```

```
glimpse(drivers)
```

```
## Observations: 51
## Variables: 8
## $ State
## $ Number.of.drivers.involved.in.fatal.collisions.per.billion.miles
## $ Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Speeding
## $ Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Alcohol.Impaired
## $ Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Not.Distracted
## $ Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Had.Not.Been.Involved.In.Any.Previous.Accidents
## $ Car.Insurance.Premiums....
## $ Losses.incurred.by.insurance.companies.for.collisions.per.insured.driver....
```

```
summary(drivers)
```

```
##       State
## Alabama   : 1
## Alaska    : 1
## Arizona   : 1
## Arkansas  : 1
## California: 1
## Colorado  : 1
## (Other)   :45
## Number.of.drivers.involved.in.fatal.collisions.per.billion.miles
## Min.      : 5.90
## 1st Qu.:12.75
## Median    :15.60
## Mean      :15.79
## 3rd Qu.:18.50
## Max.      :23.90
##
## Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Speeding
## Min.      :13.00
## 1st Qu.:23.00
## Median    :34.00
## Mean      :31.73
```

```
## 3rd Qu.:38.00
## Max. :54.00
##
## Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Alcohol.Impaired
## Min. :16.00
## 1st Qu.:28.00
## Median :30.00
## Mean :30.69
## 3rd Qu.:33.00
## Max. :44.00
##
## Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Were.Not.Distracted
## Min. : 10.00
## 1st Qu.: 83.00
## Median : 88.00
## Mean : 85.92
## 3rd Qu.: 95.00
## Max. :100.00
##
## Percentage.Of.Drivers.Involved.In.Fatal.Collisions.Who.Had.Not.Been.Involved.In.Any.Previous.Accidents
## Min. : 76.00
## 1st Qu.: 83.50
## Median : 88.00
## Mean : 88.73
## 3rd Qu.: 95.00
## Max. :100.00
##
## Car.Insurance.Premiums....
## Min. : 642.0
## 1st Qu.: 768.4
## Median : 859.0
## Mean : 887.0
## 3rd Qu.:1007.9
## Max. :1301.5
##
## Losses.incurred.by.insurance.companies.for.collisions.per.insured.driver....
## Min. : 82.75
## 1st Qu.:114.64
## Median :136.05
## Mean :134.49
## 3rd Qu.:151.87
## Max. :194.78
##
```

```
colnames(drivers)=
  c("State","NperB","PrcSpeed","PrcAlco","PrcNotDist","PrcNoPrev","Premium","Loss")
sort(drivers[,2])
```

```
## [1] 5.9 8.2 9.6 10.6 10.8 11.1 11.2 11.3 11.6 12.0 12.3 12.5 12.7 12.8
## [15] 12.8 13.6 13.6 13.8 14.1 14.1 14.5 14.7 14.9 15.1 15.3 15.6 15.7 16.1
## [29] 16.2 16.8 17.4 17.5 17.6 17.8 17.9 18.1 18.2 18.4 18.6 18.8 19.4 19.4
## [43] 19.5 19.9 20.5 21.4 21.4 22.4 23.8 23.9 23.9
```

```
drivers[1:10,1:3]
```

```
##           State NperB PrcSpeed
## 1      Alabama  18.8      39
## 2      Alaska   18.1      41
## 3      Arizona  18.6      35
## 4      Arkansas 22.4      18
## 5      California 12.0      35
## 6      Colorado 13.6      37
## 7      Connecticut 10.8      46
## 8      Delaware 16.2      38
```



```
## 9 District of Columbia 5.9 34
## 10 Florida 17.9 21
```

```
drivers[order(drivers[,2]),1:3]
```

```
##          State NperB PrcSpeed
## 9 District of Columbia 5.9 34
## 22 Massachusetts 8.2 23
## 24 Minnesota 9.6 23
## 48 Washington 10.6 42
## 7 Connecticut 10.8 46
## 40 Rhode Island 11.1 34
## 31 New Jersey 11.2 16
## 45 Utah 11.3 43
## 30 New Hampshire 11.6 35
## 5 California 12.0 35
## 33 New York 12.3 32
## 21 Maryland 12.5 34
## 47 Virginia 12.7 19
## 14 Illinois 12.8 36
## 38 Oregon 12.8 33
## 6 Colorado 13.6 37
## 46 Vermont 13.6 30
## 50 Wisconsin 13.8 36
## 23 Michigan 14.1 24
## 36 Ohio 14.1 28
## 15 Indiana 14.5 25
## 29 Nevada 14.7 37
## 28 Nebraska 14.9 13
## 20 Maine 15.1 38
## 13 Idaho 15.3 36
## 11 Georgia 15.6 19
## 16 Iowa 15.7 17
## 26 Missouri 16.1 43
## 8 Delaware 16.2 38
## 34 North Carolina 16.8 39
## 51 Wyoming 17.4 42
## 12 Hawaii 17.5 54
## 25 Mississippi 17.6 15
## 17 Kansas 17.8 27
## 10 Florida 17.9 21
## 2 Alaska 18.1 41
## 39 Pennsylvania 18.2 50
## 32 New Mexico 18.4 19
## 3 Arizona 18.6 35
## 1 Alabama 18.8 39
## 42 South Dakota 19.4 31
## 44 Texas 19.4 40
## 43 Tennessee 19.5 21
## 37 Oklahoma 19.9 32
## 19 Louisiana 20.5 35
## 18 Kentucky 21.4 19
## 27 Montana 21.4 39
## 4 Arkansas 22.4 18
## 49 West Virginia 23.8 34
## 35 North Dakota 23.9 23
## 41 South Carolina 23.9 38
```

```
arrange(drivers,NperB)
```

```
##          State NperB PrcSpeed PrcAlco PrcNotDist PrcNoPrev
## 1 District of Columbia 5.9 34 27 100 100
## 2 Massachusetts 8.2 23 35 87 80
## 3 Minnesota 9.6 23 29 88 88
## 4 Washington 10.6 42 33 82 86
```

| | | | | | | |
|-------|----------------|------|----|----|----|-----|
| ## 5 | Connecticut | 10.8 | 46 | 36 | 87 | 82 |
| ## 6 | Rhode Island | 11.1 | 34 | 38 | 92 | 79 |
| ## 7 | New Jersey | 11.2 | 16 | 28 | 86 | 78 |
| ## 8 | Utah | 11.3 | 43 | 16 | 88 | 96 |
| ## 9 | New Hampshire | 11.6 | 35 | 30 | 87 | 83 |
| ## 10 | California | 12.0 | 35 | 28 | 91 | 89 |
| ## 11 | New York | 12.3 | 32 | 29 | 88 | 80 |
| ## 12 | Maryland | 12.5 | 34 | 32 | 71 | 99 |
| ## 13 | Virginia | 12.7 | 19 | 27 | 87 | 88 |
| ## 14 | Illinois | 12.8 | 36 | 34 | 94 | 96 |
| ## 15 | Oregon | 12.8 | 33 | 26 | 67 | 90 |
| ## 16 | Colorado | 13.6 | 37 | 28 | 79 | 95 |
| ## 17 | Vermont | 13.6 | 30 | 30 | 96 | 95 |
| ## 18 | Wisconsin | 13.8 | 36 | 33 | 39 | 84 |
| ## 19 | Michigan | 14.1 | 24 | 28 | 95 | 77 |
| ## 20 | Ohio | 14.1 | 28 | 34 | 99 | 82 |
| ## 21 | Indiana | 14.5 | 25 | 29 | 95 | 95 |
| ## 22 | Nevada | 14.7 | 37 | 32 | 95 | 99 |
| ## 23 | Nebraska | 14.9 | 13 | 35 | 93 | 90 |
| ## 24 | Maine | 15.1 | 38 | 30 | 87 | 84 |
| ## 25 | Idaho | 15.3 | 36 | 29 | 85 | 98 |
| ## 26 | Georgia | 15.6 | 19 | 25 | 95 | 93 |
| ## 27 | Iowa | 15.7 | 17 | 25 | 97 | 87 |
| ## 28 | Missouri | 16.1 | 43 | 34 | 92 | 84 |
| ## 29 | Delaware | 16.2 | 38 | 30 | 87 | 99 |
| ## 30 | North Carolina | 16.8 | 39 | 31 | 94 | 81 |
| ## 31 | Wyoming | 17.4 | 42 | 32 | 81 | 90 |
| ## 32 | Hawaii | 17.5 | 54 | 41 | 82 | 87 |
| ## 33 | Mississippi | 17.6 | 15 | 31 | 10 | 100 |
| ## 34 | Kansas | 17.8 | 27 | 24 | 77 | 85 |
| ## 35 | Florida | 17.9 | 21 | 29 | 92 | 94 |
| ## 36 | Alaska | 18.1 | 41 | 25 | 90 | 94 |
| ## 37 | Pennsylvania | 18.2 | 50 | 31 | 96 | 88 |
| ## 38 | New Mexico | 18.4 | 19 | 27 | 67 | 98 |
| ## 39 | Arizona | 18.6 | 35 | 28 | 84 | 96 |
| ## 40 | Alabama | 18.8 | 39 | 30 | 96 | 80 |
| ## 41 | South Dakota | 19.4 | 31 | 33 | 98 | 86 |
| ## 42 | Texas | 19.4 | 40 | 38 | 91 | 87 |
| ## 43 | Tennessee | 19.5 | 21 | 29 | 82 | 81 |
| ## 44 | Oklahoma | 19.9 | 32 | 29 | 92 | 94 |
| ## 45 | Louisiana | 20.5 | 35 | 33 | 73 | 98 |
| ## 46 | Kentucky | 21.4 | 19 | 23 | 78 | 76 |
| ## 47 | Montana | 21.4 | 39 | 44 | 84 | 85 |
| ## 48 | Arkansas | 22.4 | 18 | 26 | 94 | 95 |
| ## 49 | West Virginia | 23.8 | 34 | 28 | 97 | 87 |
| ## 50 | North Dakota | 23.9 | 23 | 42 | 99 | 86 |
| ## 51 | South Carolina | 23.9 | 38 | 41 | 96 | 81 |

| | | |
|-------|---------|--------|
| ## | Premium | Loss |
| ## 1 | 1273.89 | 136.05 |
| ## 2 | 1011.14 | 135.63 |
| ## 3 | 777.18 | 133.35 |
| ## 4 | 890.03 | 111.62 |
| ## 5 | 1068.73 | 167.02 |
| ## 6 | 1148.99 | 148.58 |
| ## 7 | 1301.52 | 159.85 |
| ## 8 | 809.38 | 109.48 |
| ## 9 | 746.54 | 120.21 |
| ## 10 | 878.41 | 165.63 |
| ## 11 | 1234.31 | 150.01 |
| ## 12 | 1048.78 | 192.70 |
| ## 13 | 768.95 | 153.72 |
| ## 14 | 803.11 | 139.15 |
| ## 15 | 804.71 | 104.61 |
| ## 16 | 835.50 | 139.91 |
| ## 17 | 716.20 | 109.61 |
| ## 18 | 670.31 | 106.62 |
| ## 19 | 1110.61 | 152.26 |
| ## 20 | 697.73 | 133.52 |
| ## 21 | 710.46 | 108.92 |
| ## 22 | 1029.87 | 138.71 |

```
## 23 732.28 114.82
## 24 661.88 96.57
## 25 641.96 82.75
## 26 913.15 142.80
## 27 649.06 114.47
## 28 790.32 144.45
## 29 1137.87 151.48
## 30 708.24 127.82
## 31 791.14 122.04
## 32 861.18 120.92
## 33 896.07 155.77
## 34 780.45 133.80
## 35 1160.13 144.18
## 36 1053.48 133.93
## 37 905.99 153.86
## 38 869.85 120.75
## 39 899.47 110.35
## 40 784.55 145.08
## 41 669.31 96.87
## 42 1004.75 156.83
## 43 767.91 155.57
## 44 881.51 178.86
## 45 1281.55 194.78
## 46 872.51 137.13
## 47 816.21 85.15
## 48 827.34 142.39
## 49 992.61 152.56
## 50 688.75 109.72
## 51 858.97 116.29
```

```
arrange(drivers, desc(PrcSpeed))
```

| ## | State | NperB | PrcSpeed | PrcAlco | PrcNotDist | PrcNoPrev |
|-------|----------------------|-------|----------|---------|------------|-----------|
| ## 1 | Hawaii | 17.5 | 54 | 41 | 82 | 87 |
| ## 2 | Pennsylvania | 18.2 | 50 | 31 | 96 | 88 |
| ## 3 | Connecticut | 10.8 | 46 | 36 | 87 | 82 |
| ## 4 | Missouri | 16.1 | 43 | 34 | 92 | 84 |
| ## 5 | Utah | 11.3 | 43 | 16 | 88 | 96 |
| ## 6 | Washington | 10.6 | 42 | 33 | 82 | 86 |
| ## 7 | Wyoming | 17.4 | 42 | 32 | 81 | 90 |
| ## 8 | Alaska | 18.1 | 41 | 25 | 90 | 94 |
| ## 9 | Texas | 19.4 | 40 | 38 | 91 | 87 |
| ## 10 | Alabama | 18.8 | 39 | 30 | 96 | 80 |
| ## 11 | Montana | 21.4 | 39 | 44 | 84 | 85 |
| ## 12 | North Carolina | 16.8 | 39 | 31 | 94 | 81 |
| ## 13 | Delaware | 16.2 | 38 | 30 | 87 | 99 |
| ## 14 | Maine | 15.1 | 38 | 30 | 87 | 84 |
| ## 15 | South Carolina | 23.9 | 38 | 41 | 96 | 81 |
| ## 16 | Colorado | 13.6 | 37 | 28 | 79 | 95 |
| ## 17 | Nevada | 14.7 | 37 | 32 | 95 | 99 |
| ## 18 | Idaho | 15.3 | 36 | 29 | 85 | 98 |
| ## 19 | Illinois | 12.8 | 36 | 34 | 94 | 96 |
| ## 20 | Wisconsin | 13.8 | 36 | 33 | 39 | 84 |
| ## 21 | Arizona | 18.6 | 35 | 28 | 84 | 96 |
| ## 22 | California | 12.0 | 35 | 28 | 91 | 89 |
| ## 23 | Louisiana | 20.5 | 35 | 33 | 73 | 98 |
| ## 24 | New Hampshire | 11.6 | 35 | 30 | 87 | 83 |
| ## 25 | District of Columbia | 5.9 | 34 | 27 | 100 | 100 |
| ## 26 | Maryland | 12.5 | 34 | 32 | 71 | 99 |
| ## 27 | Rhode Island | 11.1 | 34 | 38 | 92 | 79 |
| ## 28 | West Virginia | 23.8 | 34 | 28 | 97 | 87 |
| ## 29 | Oregon | 12.8 | 33 | 26 | 67 | 90 |
| ## 30 | New York | 12.3 | 32 | 29 | 88 | 80 |
| ## 31 | Oklahoma | 19.9 | 32 | 29 | 92 | 94 |
| ## 32 | South Dakota | 19.4 | 31 | 33 | 98 | 86 |
| ## 33 | Vermont | 13.6 | 30 | 30 | 96 | 95 |
| ## 34 | Ohio | 14.1 | 28 | 34 | 99 | 82 |

| | | | | | | |
|-------|---------------|------|----|----|----|-----|
| ## 35 | Kansas | 17.8 | 27 | 24 | 77 | 85 |
| ## 36 | Indiana | 14.5 | 25 | 29 | 95 | 95 |
| ## 37 | Michigan | 14.1 | 24 | 28 | 95 | 77 |
| ## 38 | Massachusetts | 8.2 | 23 | 35 | 87 | 80 |
| ## 39 | Minnesota | 9.6 | 23 | 29 | 88 | 88 |
| ## 40 | North Dakota | 23.9 | 23 | 42 | 99 | 86 |
| ## 41 | Florida | 17.9 | 21 | 29 | 92 | 94 |
| ## 42 | Tennessee | 19.5 | 21 | 29 | 82 | 81 |
| ## 43 | Georgia | 15.6 | 19 | 25 | 95 | 93 |
| ## 44 | Kentucky | 21.4 | 19 | 23 | 78 | 76 |
| ## 45 | New Mexico | 18.4 | 19 | 27 | 67 | 98 |
| ## 46 | Virginia | 12.7 | 19 | 27 | 87 | 88 |
| ## 47 | Arkansas | 22.4 | 18 | 26 | 94 | 95 |
| ## 48 | Iowa | 15.7 | 17 | 25 | 97 | 87 |
| ## 49 | New Jersey | 11.2 | 16 | 28 | 86 | 78 |
| ## 50 | Mississippi | 17.6 | 15 | 31 | 10 | 100 |
| ## 51 | Nebraska | 14.9 | 13 | 35 | 93 | 90 |

| | | |
|-------|---------|--------|
| ## | Premium | Loss |
| ## 1 | 861.18 | 120.92 |
| ## 2 | 905.99 | 153.86 |
| ## 3 | 1068.73 | 167.02 |
| ## 4 | 790.32 | 144.45 |
| ## 5 | 809.38 | 109.48 |
| ## 6 | 890.03 | 111.62 |
| ## 7 | 791.14 | 122.04 |
| ## 8 | 1053.48 | 133.93 |
| ## 9 | 1004.75 | 156.83 |
| ## 10 | 784.55 | 145.08 |
| ## 11 | 816.21 | 85.15 |
| ## 12 | 708.24 | 127.82 |
| ## 13 | 1137.87 | 151.48 |
| ## 14 | 661.88 | 96.57 |
| ## 15 | 858.97 | 116.29 |
| ## 16 | 835.50 | 139.91 |
| ## 17 | 1029.87 | 138.71 |
| ## 18 | 641.96 | 82.75 |
| ## 19 | 803.11 | 139.15 |
| ## 20 | 670.31 | 106.62 |
| ## 21 | 899.47 | 110.35 |
| ## 22 | 878.41 | 165.63 |
| ## 23 | 1281.55 | 194.78 |
| ## 24 | 746.54 | 120.21 |
| ## 25 | 1273.89 | 136.05 |
| ## 26 | 1048.78 | 192.70 |
| ## 27 | 1148.99 | 148.58 |
| ## 28 | 992.61 | 152.56 |
| ## 29 | 804.71 | 104.61 |
| ## 30 | 1234.31 | 150.01 |
| ## 31 | 881.51 | 178.86 |
| ## 32 | 669.31 | 96.87 |
| ## 33 | 716.20 | 109.61 |
| ## 34 | 697.73 | 133.52 |
| ## 35 | 780.45 | 133.80 |
| ## 36 | 710.46 | 108.92 |
| ## 37 | 1110.61 | 152.26 |
| ## 38 | 1011.14 | 135.63 |
| ## 39 | 777.18 | 133.35 |
| ## 40 | 688.75 | 109.72 |
| ## 41 | 1160.13 | 144.18 |
| ## 42 | 767.91 | 155.57 |
| ## 43 | 913.15 | 142.80 |
| ## 44 | 872.51 | 137.13 |
| ## 45 | 869.85 | 120.75 |
| ## 46 | 768.95 | 153.72 |
| ## 47 | 827.34 | 142.39 |
| ## 48 | 649.06 | 114.47 |
| ## 49 | 1301.52 | 159.85 |
| ## 50 | 896.07 | 155.77 |
| ## 51 | 732.28 | 114.82 |

Make new variables

```
driversp=mutate(drivers,prem_c=Loss/Premium)
select(arrange(driversp,prem_c),State,prem_c)
```

```
##           State    prem_c
## 1      Montana 0.1043236
## 2 District of Columbia 0.1067989
## 3      New York 0.1215335
## 4      Arizona 0.1226834
## 5    New Jersey 0.1228179
## 6      Florida 0.1242792
## 7    Washington 0.1254115
## 8      Alaska 0.1271310
## 9      Idaho 0.1289021
## 10    Rhode Island 0.1293136
## 11     Oregon 0.1299971
## 12     Delaware 0.1331259
## 13    Massachusetts 0.1341357
## 14      Nevada 0.1346869
## 15      Utah 0.1352640
## 16    South Carolina 0.1353831
## 17     Michigan 0.1370958
## 18    New Mexico 0.1388170
## 19     Hawaii 0.1404120
## 20    South Dakota 0.1447311
## 21      Maine 0.1459026
## 22    Louisiana 0.1519878
## 23     Vermont 0.1530438
## 24     Indiana 0.1533091
## 25    West Virginia 0.1536958
## 26     Wyoming 0.1542584
## 27     Texas 0.1560886
## 28    Connecticut 0.1562789
## 29     Georgia 0.1563818
## 30     Nebraska 0.1567979
## 31     Kentucky 0.1571673
## 32     Wisconsin 0.1590607
## 33    North Dakota 0.1593031
## 34    New Hampshire 0.1610229
## 35     Colorado 0.1674566
## 36    Pennsylvania 0.1698253
## 37      Kansas 0.1714396
## 38     Minnesota 0.1715819
## 39     Arkansas 0.1721058
## 40     Illinois 0.1732639
## 41    Mississippi 0.1738369
## 42      Iowa 0.1763627
## 43    North Carolina 0.1804755
## 44     Missouri 0.1827741
## 45     Maryland 0.1837373
## 46     Alabama 0.1849213
## 47     California 0.1885566
## 48      Ohio 0.1913634
## 49     Virginia 0.1999090
## 50     Tennessee 0.2025888
## 51     Oklahoma 0.2029018
```

Document all the changes you make using a script.

The best way to make a report is to put everything into an .Rmd document and then knit into an html file using the knitr package.

Summary of this Session:

- Careful data preprocessing is necessary at the beginning of any data exploration exercise.
- Missing data and outliers need to be identified.
- Missing data may be imputed if there are only a few in a column or row and if their occurrence patterns are random.
- We saw how to use the package `dplyr` that allows us to easily do a sequence of actions on data using the `%>%` operator.
- Some of the possible actions are:
 - `filter()`
 - `arrange()`
 - `select()`
 - `mutate()`
 - `summarise()`
 - `sample_n()`
- We saw that preprocessing your data should be documented with scripts you save. A good way to do this is to use RStudio's Rmd editor and html generator.

Question: Look up the RStudio data wrangling cheatsheet: [R Data Wrangling Cheatsheet](#)

Activity: Re-analyze the drivers data and make your own Rmd and html reports.