# Lab 10: Tidyverse I: Pipes and Dplyr

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This week's agenda: learning to master pipes and dplyr.

#### Pipes to base R

For each of the following code blocks, which are written with pipes, write equivalent code in base R (to do the same thing).

```
• 1a.
```

```
# Pipes:
letters <- "abcde"</pre>
letters %>%
 toupper %>%
 paste(collapse="+")
## [1] "ABCDE"
# Base R:
paste(toupper(letters), collapse="+")
## [1] "ABCDE"
  • 1b.
# Pipes:
phrase <- "
               Ceci n'est pas une pipe
     Ceci n'est pas une pipe
                                  " %>%
 gsub("une", "un", .) %>%
 trimws
## [1] "Ceci n'est pas un pipe"
# Base R:
trimws(gsub("une", "un", phrase))
```

```
## [1] "Ceci n'est pas un pipe"
```

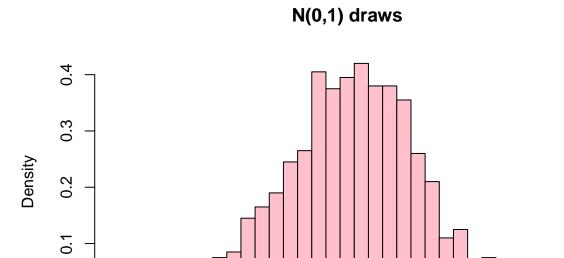
• 1c.

0.0

-3

-2

```
# Pipes:
rnorm(1000) %>%
hist(breaks=30, main="N(0,1) draws", col="pink", prob=TRUE)
```



-1

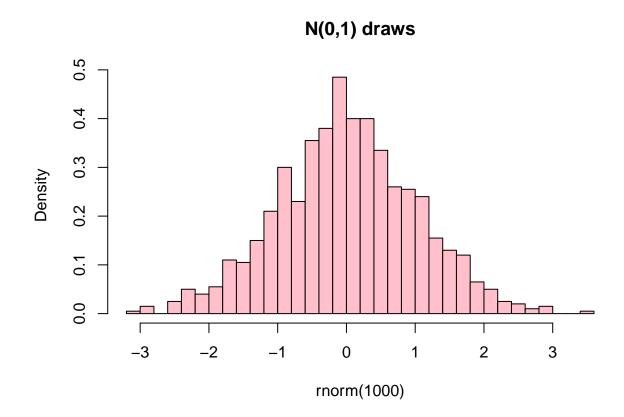
# Base R:
hist(rnorm(1000), breaks=30, main="N(0,1) draws", col="pink", prob=TRUE)

0

1

2

3



• 1d.

```
# Pipes:
rnorm(1000) %>%
  hist(breaks=30, plot=FALSE) %>%
  "[["("density") %>%
  max

## [1] 0.445
# Base R:
max(hist(rnorm(1000), breaks=30, plot=FALSE)$density)

## [1] 0.44
```

## Base R to pipes

For each of the following code blocks, which are written in base R, write equivalent code with pipes (to do the same thing).

• 2a. Hint: you'll have to use the dot ., as seen above in Q1b, or in the lecture notes.

```
# Base R:
paste("Your grade is", sample(c("A","B","C","D","R"), size=1))
```

## [1] "Your grade is A"

```
# Pipes:
c("A","B","C","D","R") %>%
  sample(size=1) %>%
 paste("Your grade is", .)
## [1] "Your grade is C"
  • 2b. Hint: you can use the dot . again, in order to index state.name directly in the last pipe command.
# Base R:
state.name[which.max(state.x77[,"Illiteracy"])]
## [1] "Louisiana"
# Pipes:
state.x77 %>%
  .[,"Illiteracy"] %>%
  which.max %>%
  state.name[.]
## [1] "Louisiana"
  • 2c. Hint: if x is a list of length 1, then x[[1]] is the same as unlist(x).
str.url = "http://www.stat.cmu.edu/~ryantibs/statcomp/data/trump.txt"
# Base R:
lines = readLines(str.url)
text = paste(lines, collapse=" ")
words = strsplit(text, split="[[:space:]]|[[:punct:]]")[[1]]
wordtab = table(words)
wordtab = sort(wordtab, decreasing=TRUE)
head(wordtab, 10)
## words
##
                          to our will
                                                 I have
         the and
                     of
                                          in
   524 189 146
                   127 126
                                90
                                    83
                                          69
                                                67
                                                     58
# Pipes:
readLines(str.url) %>%
  paste(collapse=" ") %>%
  strsplit(split="[[:space:]]|[[:punct:]]") %>%
 unlist %>%
  table %>%
  sort(decreasing = TRUE) %>%
 head(10)
## .
                              our will
##
                     of
                                                 I have
         the and
                          to
## 524 189
             146
                   127
                         126
                                90
                                          69
                                                67
  • 2d. Hint: the only difference between this and the last part is the line words = words [words != ""].
     This is a bit tricky line to do with pipes: use the dot., once more, and manipulate it as if were a
     variable name.
```

```
words = words[words != ""]
wordtab = table(words)
wordtab = sort(wordtab, decreasing=TRUE)
head(wordtab, 10)
## words
## the and
               of
                         our will
                                    in
                                          I have
                                                     а
                    to
## 189 146 127
                   126
                          90
                               83
                                              58
                                                    51
# Pipes:
readLines(str.url) %>%
  paste(collapse=" ") %>%
  strsplit(split="[[:space:]]|[[:punct:]]") %>%
  unlist %>%
  .[.!=""] %>%
  table %>%
  sort(decreasing = TRUE) %>%
 head(10)
## .
## the
                         our will
         and
               of
                    to
                                    in
                                          I have
    189
         146
              127
                   126
                          90
                               83
                                    69
                                         67
                                              58
                                                    51
```

#### Sprints data, revisited

Below we read in a data frame sprint.w.df containing the top women's times in the 100m sprint, as seen in previous labs. We also define a function factor.to.numeric() that was used in Lab 8, to convert the Wind column to numeric values. In what follows, use dplyr and pipes to answer the following questions on sprint.w.df.

• 3a. Convert the Wind column to numeric using factor.to.numeric(). Hint: use mutate\_at(), and reassign sprint.w.df to be the output.

```
sprint.w.df <- sprint.w.df %>%
mutate_at("Wind", factor.to.numeric)
```

```
## Warning in ifelse(length(x) > 1, as.numeric(paste(x, collapse = ".")), ## as.numeric(x)): NAs introduced by coercion
```

• **3b.** Run a linear regression of the Time on Wind columns, but only using data where Wind values that are nonpositive, and report the coefficients. Hint: use filter(), and use the dot. to pipe into the lm() function appropriately.

```
sprint.w.df %>%
filter(Wind<0) %>%
```

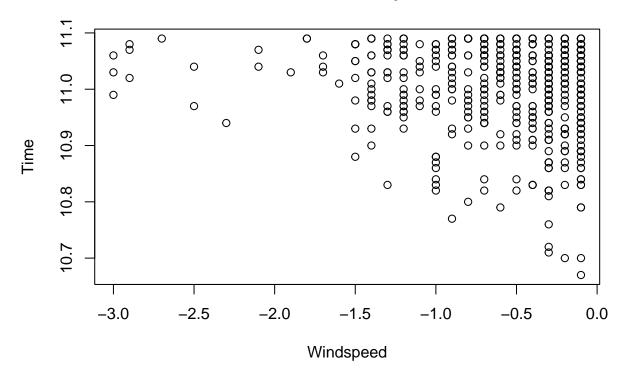
```
lm(Time ~ Wind, data = .) %>%
coef
```

```
## (Intercept) Wind
## 10.98967275 -0.02250043
```

• 3c. Plot the Time versus Wind columns, but only using data where Wind values that are nonpositive, and label the axes appropriately. Hint: recall that for a data frame, with columns colX and colY, you can use plot(colY ~ colX, data=df), to plot df\$colY versus df\$colX.

```
sprint.w.df %>%
  filter(Wind<0) %>%
  plot(Time ~ Wind, data=., main="Time vs. Windspeed", xlab="Windspeed", ylab="Time")
```

### Time vs. Windspeed



• 3d. Reorder the rows in terms of increasing Wind, and then display only the women who ran at most 10.7 seconds. Hint: do this with one single flow of pipe commands; use arrange(), filter().

```
sprint.w.df %>%
  arrange(Wind) %>%
  filter(Time<=10.7)</pre>
```

```
##
      Rank
            Time Wind
                                            Name Country Birthdate
## 1
         2 10.70 -0.2
                           Svetlana Goncharenko
                                                      RUS
                                                           26.05.71
## 2
         6 10.67 -0.1
                                Carmelita Jeter
                                                      USA
                                                           24.11.79
## 3
         7 10.70 -0.1
                                   Marion Jones
                                                      USA
                                                          12.10.75
## 4
                  0.0 Florence Griffith-Joyner
                                                           21.12.59
         1 10.49
                                                      USA
## 5
         1 10.60
                  0.0
                                   Zhanna Block
                                                      UKR
                                                           06.07.72
         2 10.70
## 6
                  0.0
                                   Zhanna Block
                                                     UKR
                                                          06.07.72
```

```
## 7
         7 10.70
                  0.3
                                Elaine Thompson
                                                     JAM
                                                          28.06.92
## 8
                                                     JAM
                                                          27.12.86
         7 10.70
                  0.6
                       Shelly-Ann Fraser-Pryce
## 9
         3 10.62
                  1.0 Florence Griffith-Joyner
                                                     USA
                                                          21.12.59
## 10
         5 10.65
                                   Marion Jones
                                                     USA
                                                          12.10.75
                  1.1
## 11
         2 10.70
                  1.1
                                Juliet Cuthbert
                                                     JAM
                                                          09.04.64
## 12
         2 10.61
                  1.2 Florence Griffith-Joyner
                                                          21.12.59
                                                     USA
                                                          24.11.79
## 13
         4 10.64
                  1.2
                                Carmelita Jeter
                                                     USA
                                                          10.09.88
## 14
         2 10.70
                  1.3
                              Blessing Okagbare
                                                     NGR
##
  15
         7 10.70
                  1.6 Florence Griffith-Joyner
                                                     USA
                                                          21.12.59
##
  16
         7 10.70
                  2.0
                                Carmelita Jeter
                                                     USA
                                                          24.11.79
##
                      City
                                 Date
           Rostov-na-Donu 30.05.1998
##
  1
##
  2
      Thessaloníki 13.09.2009
## 3
                  Sevilla 22.08.1999
## 4
             Indianapolis 16.07.1988
## 5
                      Kiev 12.06.1997
## 6
                     Kiev 12.06.1997
## 7
                 Kingston 01.07.2016
## 8
                 Kingston 29.06.2012
## 9
                    Seoul 24.09.1988
## 10
             Johannesburg 12.09.1998
## 11
                 Kingston 04.07.1992
## 12
             Indianapolis 17.07.1988
## 13
                 Shanghai 20.09.2009
## 14
                  El Paso 10.04.2010
## 15
             Indianapolis 17.07.1988
## 16
                    Eugene 04.06.2011
```

• **3e.** Now reorder the rows in terms of increasing Time, and *then* increasing Wind, and again display only the women who ran at most 10.7 seconds, but only display the Time, Wind, Name, and Date columns. Hint: a single flow of pipe commands will do; note that **arrange()** can take multiple columns that you want to sort by, and the order you pass them specifies the priority.

```
sprint.w.df %>%
  arrange(Time, Wind) %>%
  filter(Time <= 10.7) %>%
  select(Time, Wind, Name, Date)
```

```
##
       Time Wind
                                      Name
                                                 Date
## 1
     10.49
            0.0 Florence Griffith-Joyner 16.07.1988
                              Zhanna Block 12.06.1997
      10.60
             0.0
## 3
      10.61
             1.2 Florence Griffith-Joyner 17.07.1988
## 4
      10.62
             1.0 Florence Griffith-Joyner 24.09.1988
     10.64
             1.2
## 5
                          Carmelita Jeter 20.09.2009
## 6
      10.65 1.1
                             Marion Jones 12.09.1998
                          Carmelita Jeter 13.09.2009
## 7
      10.67 -0.1
## 8
      10.70 -0.2
                     Svetlana Goncharenko 30.05.1998
## 9
     10.70 -0.1
                             Marion Jones 22.08.1999
## 10 10.70
             0.0
                              Zhanna Block 12.06.1997
## 11 10.70
             0.3
                          Elaine Thompson 01.07.2016
## 12 10.70
             0.6
                  Shelly-Ann Fraser-Pryce 29.06.2012
## 13 10.70
             1.1
                          Juliet Cuthbert 04.07.1992
## 14 10.70
             1.3
                        Blessing Okagbare 10.04.2010
  15 10.70
             1.6 Florence Griffith-Joyner 17.07.1988
## 16 10.70
                          Carmelita Jeter 04.06.2011
            2.0
```

#### Prostate cancer data, revisited

Below we read in a data frame pros.df containing measurements on men with prostate cancer, as seen in previous labs. As before, in what follows, use dplyr and pipes to answer the following questions on pros.df.

```
pros.df =
  read.table("http://www.stat.cmu.edu/~ryantibs/statcomp/data/pros.dat")
```

• 4a. Among the men whose lcp value is equal to the minimum value, report the lowest and highest lpsa score.

```
pros.df %>%
  filter(lcp==min(lcp)) %>%
  filter(lpsa==min(lpsa) | lpsa==max(lpsa))
```

```
##
          lcavol lweight age
                                                  1cp gleason pgg45
                                   lbph svi
                                                                           lpsa
     -0.5798185 2.769459
                          50 -1.386294
                                          0 -1.386294
                                                             6
                                                                   0 -0.4307829
## 92
       2.5329028 3.677566 61
                              1.348073
                                          1 -1.386294
                                                             7
                                                                     4.1295508
                                                                  15
```

• **4b.** Order the rows by decreasing age, then decreasing lpsa score, and display the rows from men who are older than 70, but only the age, lpsa, lcavol, and lweight columns.

```
pros.df %>%
  arrange(desc(age), desc(lpsa)) %>%
  filter(age>70) %>%
  select(age, lpsa, lcavol, lweight)
```

```
##
                lpsa
                         lcavol lweight
      age
## 47
       79
           2.5687881
                      2.7278528 3.995445
##
  78
       78
           3.4355988
                      2.5376572 4.354784
## 83
       77
           3.5652984
                      2.6130067 3.888754
  72
       77
           3.0373539
                      1.1600209 3.341093
       76
           3.9936030
                      1.5623463 3.695110
## 90
##
  3
       74 -0.1625189 -0.5108256 2.691243
                      0.4574248 4.524502
## 61
      73
           2.8419982
           2.1575593
## 37
       73
                      1.4231083 3.657131
## 77
       72
           3.3928291
                      2.0108950 4.433789
## 70
       72
           2.9729753
                      1.1939225 4.780383
## 68
       72
           2.9626924
                      2.1983351 4.050915
## 63
       72
           2.8535925
                      2.7757089 3.524889
## 33
       71
           2.0082140
                      1.2753628 3.037354
```

• 4c. Run a linear regression of the lpsa on lcavol and lweight columns, but only using men whose lcp value is strictly larger than the minimum value, and report a summary of the fitted model.

```
pros.df %>%
  filter(lcp>min(lcp)) %>%
  lm(lpsa ~ lcavol + lweight, data = .) %>%
  summary
```

```
##
## Call:
## lm(formula = lpsa ~ lcavol + lweight, data = .)
##
## Residuals:
##
        Min
                  1Q
                        Median
                                      3Q
                                              Max
##
  -1.69447 -0.39524
                      0.06257
                                0.33075
                                         1.87082
##
```

```
## Coefficients:
              Estimate Std. Error t value Pr(>|t|)
##
                0.6699
## (Intercept)
                          1.0512
                                   0.637
## lcavol
                0.6462
                           0.1221
                                    5.291 2.83e-06 ***
## lweight
                0.2846
                           0.2907
                                   0.979
                                            0.332
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
##
## Residual standard error: 0.787 on 49 degrees of freedom
## Multiple R-squared: 0.4013, Adjusted R-squared: 0.3769
## F-statistic: 16.42 on 2 and 49 DF, p-value: 3.48e-06
```

• 4d. Extend your code in the last part, still just using a single flow of pipe commands in total, to extract the p-values associated with each of the coefficients in the fitted model.

```
pros.df %>%
  filter(lcp>min(lcp)) %>%
  lm(lpsa ~ lcavol + lweight, data = .) %>%
  summary %>%
  .$coefficients %>%
  .[,"Pr(>|t|)"]
```

```
## (Intercept) lcavol lweight
## 5.268747e-01 2.832674e-06 3.322792e-01
```