Load the dplyr package

This step is important!

```
library(dplyr)
```

```
##
## Attaching package: 'dplyr'
##
## The following object is masked from 'package:stats':
##
##     filter
##
## The following objects are masked from 'package:base':
##
##     intersect, setdiff, setequal, union
```

select

```
chicago <- readRDS("chicago.rds")
dim(chicago)
```

```
## [1] 6940    8
```

```
head(select(chicago, 1:5))
```

```
##    city tmpd   dptp       date pm25tmean2
## 1 chic 31.5 31.500 1987-01-01          NA
## 2 chic 33.0 29.875 1987-01-02          NA
## 3 chic 33.0 27.375 1987-01-03          NA
## 4 chic 29.0 28.625 1987-01-04          NA
## 5 chic 32.0 28.875 1987-01-05          NA
## 6 chic 40.0 35.125 1987-01-06          NA
```

select

```
names(chicago)[1:3]
```

```
## [1] "city" "tmpd" "dptp"
```

```
head(select(chicago, city:dptp))
```

```
##   city tmpd   dptp
## 1 chic 31.5 31.500
## 2 chic 33.0 29.875
## 3 chic 33.0 27.375
## 4 chic 29.0 28.625
## 5 chic 32.0 28.875
## 6 chic 40.0 35.125
```

select

In dplyr you can do

```
head(select(chicago, -(city:dptp)))
```

Equivalent base R

```
i <- match("city", names(chicago))
j <- match("dptp", names(chicago))
head(chicago[, -(i:j)])
```

filter

```
chic.f <- filter(chicago, pm25tmean2 > 30)
head(select(chic.f, 1:3, pm25tmean2), 10)
```

```
##     city tmpd dptp pm25tmean2
## 1   chic   23 21.9      38.10
## 2   chic   28 25.8      33.95
## 3   chic   55 51.3      39.40
## 4   chic   59 53.7      35.40
## 5   chic   57 52.0      33.30
## 6   chic   57 56.0      32.10
## 7   chic   75 65.8      56.50
## 8   chic   61 59.0      33.80
## 9   chic   73 60.3      30.30
## 10 chic   78 67.1      41.40
```

filter

```
chic.f <- filter(chicago, pm25tmean2 > 30 & tmpd > 80)
head(select(chic.f, 1:3, pm25tmean2, tmpd), 10)
```

```
##     city tmpd dptp pm25tmean2
## 1   chic   81 71.2    39.6000
## 2   chic   81 70.4    31.5000
## 3   chic   82 72.2    32.3000
## 4   chic   84 72.9    43.7000
## 5   chic   85 72.6    38.8375
## 6   chic   84 72.6    38.2000
## 7   chic   82 67.4    33.0000
## 8   chic   82 63.5    42.5000
## 9   chic   81 70.4    33.1000
## 10 chic   82 66.2    38.8500
```

arrange

Reordering rows of a data frame (while preserving corresponding order of other columns) is normally a pain to do in R.

```
chicago <- arrange(chicago, date)
head(select(chicago, date, pm25tmean2), 3)
```

```
##         date pm25tmean2
## 1 1987-01-01          NA
## 2 1987-01-02          NA
## 3 1987-01-03          NA
```

```
tail(select(chicago, date, pm25tmean2), 3)
```

```
##            date pm25tmean2
## 6938 2005-12-29    7.45000
## 6939 2005-12-30   15.05714
## 6940 2005-12-31   15.00000
```

arrange

Columns can be arranged in descending order too.

```
chicago <- arrange(chicago, desc(date))
head(select(chicago, date, pm25tmean2), 3)
```

```
##         date pm25tmean2
## 1 2005-12-31   15.00000
## 2 2005-12-30   15.05714
## 3 2005-12-29    7.45000
```

```
tail(select(chicago, date, pm25tmean2), 3)
```

```
##             date pm25tmean2
## 6938 1987-01-03         NA
## 6939 1987-01-02         NA
## 6940 1987-01-01         NA
```

rename

Renaming a variable in a data frame in R is surprising hard to do!

```
head(chicago[, 1:5], 3)
```

```
##    city tmpd dptp       date pm25tmean2
## 1 chic   35 30.1 2005-12-31   15.00000
## 2 chic   36 31.0 2005-12-30   15.05714
## 3 chic   35 29.4 2005-12-29    7.45000
```

```
chicago <- rename(chicago, dewpoint = dptp,
                  pm25 = pm25tmean2)
head(chicago[, 1:5], 3)
```

```
##    city tmpd dewpoint       date     pm25
## 1 chic   35     30.1 2005-12-31 15.00000
## 2 chic   36     31.0 2005-12-30 15.05714
## 3 chic   35     29.4 2005-12-29  7.45000
```

mutate

```
chicago <- mutate(chicago,
                  pm25detrend=pm25-mean(pm25, na.rm=TRUE))
head(select(chicago, pm25, pm25detrend))
```

```
##       pm25 pm25detrend
## 1 15.00000   -1.230958
## 2 15.05714   -1.173815
## 3  7.45000   -8.780958
## 4 17.75000    1.519042
## 5 23.56000    7.329042
## 6  8.40000   -7.830958
```

group_by

Generating summary statistics by stratum

```
chicago <- mutate(chicago,
                  tempcat = factor(1 * (tmpd > 80),
                                   labels = c("cold", "hot"
hotcold <- group_by(chicago, tempcat)
summarize(hotcold, pm25 = mean(pm25, na.rm = TRUE),
          o3 = max(o3tmean2),
          no2 = median(no2tmean2))
```

```
## Source: local data frame [3 x 4]
##
##   tempcat     pm25        o3      no2
## 1    cold 15.97807 66.587500 24.54924
## 2     hot 26.48118 62.969656 24.93870
## 3      NA 47.73750  9.416667 37.44444
```

group_by

Generating summary statistics by stratum

```
chicago <- mutate(chicago,
                  year = as.POSIXlt(date)$year + 1900)
years <- group_by(chicago, year)
summarize(years, pm25 = mean(pm25, na.rm = TRUE),
          o3 = max(o3tmean2, na.rm = TRUE),
          no2 = median(no2tmean2, na.rm = TRUE))
```

```
## Source: local data frame [19 x 4]
##
##    year     pm25       o3      no2
## 1  1987      NaN 62.96966 23.49369
## 2  1988      NaN 61.67708 24.52296
## 3  1989      NaN 59.72727 26.14062
## 4  1990      NaN 52.22917 22.59583
## 5  1991      NaN 63.10417 21.38194
## 6  1992      NaN 50.82870 24.78921
## 7  1993      NaN 44.30093 25.76993
```

%>%

```
chicago %>% mutate(month = as.POSIXlt(date)$mon + 1)
    %>% group_by(month)
    %>% summarize(pm25 = mean(pm25, na.rm = TRUE),
          o3 = max(o3tmean2, na.rm = TRUE),
          no2 = median(no2tmean2, na.rm = TRUE))

## Source: local data frame [12 x 4]
##
##    month     pm25       o3      no2
## 1      1 17.76996 28.22222 25.35417
## 2      2 20.37513 37.37500 26.78034
## 3      3 17.40818 39.05000 26.76984
## 4      4 13.85879 47.94907 25.03125
## 5      5 14.07420 52.75000 24.22222
## 6      6 15.86461 66.58750 25.01140
## 7      7 16.57087 59.54167 22.38442
## 8      8 16.93380 53.96701 22.98333
## 9      9 15.91279 57.48864 24.47917
```

dplyr

Once you learn the dplyr "grammar" there are a few additional benefits

- dplyr can work with other data frame "backends"
- `data.table` for large fast tables
- SQL interface for relational databases via the DBI package

Managing Data Frames with `dplyr`

December 30, 2014

dplyr

The data frame is a key data structure in statistics and in R.

- There is one observation per row
- Each column represents a variable or measure or characteristic
- Primary implementation that you will use is the default R implementation
- Other implementations, particularly relational databases systems

dplyr

- Developed by Hadley Wickham of RStudio
- An optimized and distilled version of `plyr` package (also by Hadley)
- Does not provide any "new" functionality per se, but **greatly** simplifies existing functionality in R
- Provides a "grammar" (in particular, verbs) for data manipulation
- Is **very** fast, as many key operations are coded in C++

dplyr Verbs

- `select`: return a subset of the columns of a data frame
- `filter`: extract a subset of rows from a data frame based on logical conditions
- `arrange`: reorder rows of a data frame
- `rename`: rename variables in a data frame
- `mutate`: add new variables/columns or transform existing variables
- `summarise` / `summarize`: generate summary statistics of different variables in the data frame, possibly within strata

There is also a handy `print` method that prevents you from printing a lot of data to the console.

dplyr Properties

- The first argument is a data frame.
- The subsequent arguments describe what to do with it, and you can refer to columns in the data frame directly without using the $ operator (just use the names).
- The result is a new data frame
- Data frames must be properly formatted and annotated for this to all be useful