

## 4: Part 1 - Data Wrangling

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### Objectives

1. Describe the usefulness of data wrangling and its place in the data pipeline
2. Wrangle datasets with dplyr functions
3. Apply data wrangling skills to a real-world example dataset

### Set up your session

Today we will work with a dataset from the North Temperate Lakes Long-Term Ecological Research Station. The NTL-LTER is located in the boreal zone in northern Wisconsin, USA. We will use the chemical and physical limnology dataset, running from 1984-2016.

Opening discussion: why might we be interested in long-term observations of temperature, oxygen, and light in lakes?

Add notes here: looking at lakes, chemical and physical data, temp o2 levels ect first load in packages and the data

```
getwd()
```

```
## [1] "/home/guest/module1/EDE_Fall2023"
```

```
#install.packages(tidyverse)
```

```
library(tidyverse)
```

```
#install.packages(lubridate)
```

```
library(lubridate)
```

```
#importing data add "stringsAsFactors = TRUE" for reading csv
```

```
NTL.phys.data <- read.csv("./Data/Raw/NTL-LTER_Lake_ChemistryPhysics_Raw.csv", stringsAsFactors = TRUE)
```

```
#taking a look at the data
```

```
colnames(NTL.phys.data) #some at same date but different depths
```

```
## [1] "lakeid"      "lakename"    "year4"       "daynum"
## [5] "sampledate" "depth"       "temperature_C" "dissolvedOxygen"
## [9] "irradianceWater" "irradianceDeck" "comments"
```

```
head(NTL.phys.data)
```

```
##   lakeid lakename year4 daynum sampledate depth temperature_C dissolvedOxygen
## 1      L Paul Lake 1984   148    5/27/84  0.00             14.5             9.5
## 2      L Paul Lake 1984   148    5/27/84  0.25              NA             NA
## 3      L Paul Lake 1984   148    5/27/84  0.50              NA             NA
## 4      L Paul Lake 1984   148    5/27/84  0.75              NA             NA
## 5      L Paul Lake 1984   148    5/27/84  1.00             14.5             8.8
## 6      L Paul Lake 1984   148    5/27/84  1.50              NA             NA
##   irradianceWater irradianceDeck comments
## 1              1750             1620    <NA>
## 2              1550             1620    <NA>
```

```
## 3      1150      1620      <NA>
## 4      975      1620      <NA>
## 5      870      1620      <NA>
## 6      610      1620      <NA>
```

```
summary(NTL.phys.data)
```

```
##      lakeid      lakename      year4      daynum
## R      :11288 Peter Lake   :11288 Min.   :1984 Min.   : 55.0
## L      :10325 Paul Lake   :10325 1st Qu.:1991 1st Qu.:166.0
## T      : 6107 Tuesday Lake : 6107 Median :1997 Median :194.0
## W      : 4188 West Long Lake: 4188 Mean   :1999 Mean   :194.3
## E      : 3905 East Long Lake: 3905 3rd Qu.:2006 3rd Qu.:222.0
## M      : 1234 Crampton Lake : 1234 Max.   :2016 Max.   :307.0
## (Other): 1567 (Other)      : 1567
##      sampledate      depth      temperature_C      dissolvedOxygen
## 5/17/94: 84 Min.   : 0.00 Min.   : 0.30 Min.   : 0.00
## 9/5/90 : 64 1st Qu.: 1.50 1st Qu.: 5.30 1st Qu.: 0.30
## 10/1/07: 61 Median : 4.00 Median : 9.30 Median : 5.60
## 9/10/90: 61 Mean   : 4.39 Mean   :11.81 Mean   : 4.97
## 5/10/87: 60 3rd Qu.: 6.50 3rd Qu.:18.70 3rd Qu.: 8.40
## 5/9/88 : 60 Max.   :20.00 Max.   :34.10 Max.   :802.00
## (Other):38224 NA's      :3858 NA's      :4039
##      irradianceWater      irradianceDeck      comments
## Min.   : -0.337 Min.   : 1.5 DO Probe bad - Doesn't go to zero: 206
## 1st Qu.: 14.000 1st Qu.: 353.0 DO taken with Jones Lab Meter : 162
## Median : 65.000 Median : 747.0 NA's :38246
## Mean   : 210.242 Mean   : 720.5
## 3rd Qu.: 265.000 3rd Qu.:1042.0
## Max.   :24108.000 Max.   :8532.0
## NA's   :14287 NA's   :15419
```

```
str(NTL.phys.data) # find data type of each column
```

```
## 'data.frame': 38614 obs. of 11 variables:
## $ lakeid : Factor w/ 9 levels "C","E","H","L",...: 4 4 4 4 4 4 4 4 4 ...
## $ lakename : Factor w/ 9 levels "Central Long Lake",...: 5 5 5 5 5 5 5 5 5 ...
## $ year4 : int 1984 1984 1984 1984 1984 1984 1984 1984 1984 1984 ...
## $ daynum : int 148 148 148 148 148 148 148 148 148 148 ...
## $ sampledate : Factor w/ 1712 levels "10/1/07","10/1/93",...: 134 134 134 134 134 134 134 134 134 134 ...
## $ depth : num 0 0.25 0.5 0.75 1 1.5 2 3 4 5 ...
## $ temperature_C : num 14.5 NA NA NA 14.5 NA 14.2 11 7 6.1 ...
## $ dissolvedOxygen: num 9.5 NA NA NA 8.8 NA 8.6 11.5 11.9 2.5 ...
## $ irradianceWater: num 1750 1550 1150 975 870 610 420 220 100 34 ...
## $ irradianceDeck : num 1620 1620 1620 1620 1620 1620 1620 1620 1620 1620 ...
## $ comments : Factor w/ 2 levels "DO Probe bad - Doesn't go to zero",...: NA NA NA NA NA NA NA NA NA NA
```

```
dim(NTL.phys.data)
```

```
## [1] 38614 11
```

```
# fixing dates
```

```
class(NTL.phys.data$sampledate)
```

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

```
# Format sampleddate as date
NTL.phys.data$sampleddate <- as.Date(NTL.phys.data$sampleddate, format = "%m/%d/%y") #formats the dates t
```

## Data Wrangling

Data wrangling extends data exploration: it allows you to process data in ways that are useful for you. An important part of data wrangling is creating *tidy datasets*, with the following rules:

1. Each variable has its own column
2. Each observation has its own row
3. Each value has its own cell

What is the best way to wrangle data? There are multiple ways to arrive at a specific outcome in R, and we will illustrate some of those approaches. Your goal should be to write the simplest code that will get you to your desired outcome. However, there is sometimes a trade-off of the opportunity cost to learn a new formulation of code and the time it takes to write complex code that you already know. Remember that the best code is one that is easy to understand for yourself and your collaborators. Remember to comment your code, use informative names for variables and functions, and use reproducible methods to arrive at your output.

Notes: make it simple, get what you want and reusable

## Dplyr Wrangling Functions

dplyr is a package in R that includes functions for data manipulation (i.e., data wrangling or data munging). dplyr is included in the tidyverse package, so you should already have it installed on your machine. The functions act as verbs for data wrangling processes. For more information, run this line of code:

note: functions have names that fit well to what you would expect it to

```
vignette("dplyr") #shows basics of what the package does, shows which does things for rows and which fo
```

```
## starting httpd help server ... done
```

## Filter

Filtering allows us to choose certain rows (observations) in our dataset.

Here are the relevant commands used in the `filter` function. Add some notes to designate what these commands mean. `== != < <= > >= & |`

note: works on rows , based on logical operations, can also use `&` for and and `|` for or if there are multiple conditions. rows have to be of the same type

```
class(NTL.phys.data$lakeid)
```

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

```
class(NTL.phys.data$depth)
```

```
## [1] "numeric"
```

```
# matrix filtering
```

```
NTL.phys.data.surface1 <- NTL.phys.data[NTL.phys.data$depth == 0,]
```

```
#gets only the surface data
```

```
# dplyr filtering
```

```
NTL.phys.data.surface2 <- filter(NTL.phys.data, depth == 0) # another way to do the same thing
```

```
NTL.phys.data.surface3 <- filter(NTL.phys.data, depth < 0.25) #includes ones that arent quite on the su
```

```
# Did the methods arrive at the same result?
```

```
head(NTL.phys.data.surface1)
```

```
##   lakeid    lakename year4 daynum sampledate depth temperature_C
## 1      L    Paul Lake 1984   148 1984-05-27    0          14.5
## 18     R    Peter Lake 1984   149 1984-05-28    0          14.8
## 40     T Tuesday Lake 1984   150 1984-05-29    0          15.0
## 56     L    Paul Lake 1984   155 1984-06-03    0          18.8
## 72     R    Peter Lake 1984   156 1984-06-04    0          18.8
## 90     T Tuesday Lake 1984   157 1984-06-05    0          21.0
##   dissolvedOxygen irradianceWater irradianceDeck comments
## 1                9.5             1750           1620    <NA>
## 18               9.2             1630           1540    <NA>
## 40               9.5             1850           1960    <NA>
## 56               8.0             1100           1050    <NA>
## 72               9.0              275            275    <NA>
## 90               8.4             1200           1200    <NA>
```

```
dim(NTL.phys.data.surface1)
```

```
## [1] 1902   11
```

```
head(NTL.phys.data.surface2)
```

```
##   lakeid    lakename year4 daynum sampledate depth temperature_C
## 1      L    Paul Lake 1984   148 1984-05-27    0          14.5
## 2      R    Peter Lake 1984   149 1984-05-28    0          14.8
## 3      T Tuesday Lake 1984   150 1984-05-29    0          15.0
## 4      L    Paul Lake 1984   155 1984-06-03    0          18.8
## 5      R    Peter Lake 1984   156 1984-06-04    0          18.8
## 6      T Tuesday Lake 1984   157 1984-06-05    0          21.0
##   dissolvedOxygen irradianceWater irradianceDeck comments
## 1                9.5             1750           1620    <NA>
## 2                9.2             1630           1540    <NA>
## 3                9.5             1850           1960    <NA>
## 4                8.0             1100           1050    <NA>
## 5                9.0              275            275    <NA>
## 6                8.4             1200           1200    <NA>
```

```
dim(NTL.phys.data.surface2)
```

```
## [1] 1902   11
```

```
head(NTL.phys.data.surface3)
```

```
##   lakeid    lakename year4 daynum sampledate depth temperature_C
## 1      L    Paul Lake 1984   148 1984-05-27    0          14.5
## 2      R    Peter Lake 1984   149 1984-05-28    0          14.8
## 3      T Tuesday Lake 1984   150 1984-05-29    0          15.0
## 4      L    Paul Lake 1984   155 1984-06-03    0          18.8
## 5      R    Peter Lake 1984   156 1984-06-04    0          18.8
## 6      T Tuesday Lake 1984   157 1984-06-05    0          21.0
##   dissolvedOxygen irradianceWater irradianceDeck comments
## 1                9.5             1750           1620    <NA>
## 2                9.2             1630           1540    <NA>
## 3                9.5             1850           1960    <NA>
## 4                8.0             1100           1050    <NA>
```

```
## 5          9.0          275          275      <NA>
## 6          8.4          1200         1200      <NA>
```

```
dim(NTL.phys.data.surface3)
```

```
## [1] 1902  11
```

```
# Choose multiple conditions to filter
summary(NTL.phys.data$lakename)
```

```
## Central Long Lake      Crampton Lake      East Long Lake      Hummingbird Lake
##              539              1234              3905              430
##      Paul Lake      Peter Lake      Tuesday Lake      Ward Lake
##      10325              11288              6107              598
##      West Long Lake
##      4188
```

```
#needs quotation marks cause it is a character, either paul or peter
NTL.phys.data.PeterPaul1 <- filter(NTL.phys.data, lakename == "Paul Lake" | lakename == "Peter Lake")
#can also do by saying not any of the others
NTL.phys.data.PeterPaul2 <- filter(NTL.phys.data, lakename != "Central Long Lake" &
                                   lakename != "Crampton Lake" & lakename != "East Long Lake" &
                                   lakename != "Hummingbird Lake" & lakename != "Tuesday Lake" &
                                   lakename != "Ward Lake" & lakename != "West Long Lake")
#can do by checking inclusion in a list
NTL.phys.data.PeterPaul3 <- filter(NTL.phys.data, lakename %in% c("Paul Lake", "Peter Lake"))

# Choose a range of conditions of a numeric or integer variable
summary(NTL.phys.data$daynum)
```

```
##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      55.0  166.0   194.0   194.3   222.0   307.0
```

```
#exclude all other dates
NTL.phys.data.JunethruOctober1 <- filter(NTL.phys.data, daynum > 151 & daynum < 305)
#can use comma instead of &
NTL.phys.data.JunethruOctober2 <- filter(NTL.phys.data, daynum > 151, daynum < 305)
#inclusive but one inside
NTL.phys.data.JunethruOctober3 <- filter(NTL.phys.data, daynum >= 152 & daynum <= 304)
#create a vector with all acceptable day numbers
NTL.phys.data.JunethruOctober4 <- filter(NTL.phys.data, daynum %in% c(152:304))

# Exercise:
# filter NTL.phys.data for the year 1999
# what code do you need to use, based on the class of the variable?
class(NTL.phys.data$year4)
```

```
## [1] "integer"
```

```
# Exercise:
# filter NTL.phys.data for Tuesday Lake from 1990 through 1999.
```

Question: Why don't we filter using row numbers?

Answer:

## Arrange

Arranging allows us to change the order of rows in our dataset. By default, the arrange function will arrange rows in ascending order. basically a sort function

```
NTL.phys.data.depth.ascending <- arrange(NTL.phys.data, depth)
NTL.phys.data.depth.descending <- arrange(NTL.phys.data, desc(depth)) #need to specify to do decending

# Exercise:
# Arrange NTL.phys.data by temperature, in descending order.
# Which dates, lakes, and depths have the highest temperatures?
```

## Select

Selecting allows us to choose certain columns (variables) in our dataset.

```
NTL.phys.data.temps <- select(NTL.phys.data, lakename, sampledate:temperature_C)
#can choose just relevant columns (full dataset, names of columns)
```

## Mutate

Mutating allows us to add new columns that are functions of existing columns. Operations include addition, subtraction, multiplication, division, log, and other functions.

```
NTL.phys.data.temps <- mutate(NTL.phys.data.temps, temperature_F = (temperature_C*9/5) + 32)
# adds a new column so you can just save to the same place
```

## Lubridate

A package that makes coercing date much easier is lubridate. A guide to the package can be found at <https://lubridate.tidyverse.org/>. The cheat sheet within that web page is excellent too. This package can do many things (hint: look into this package if you are having unique date-type issues), but today we will be using two of its functions for our NTL dataset.

```
# add a month column to the dataset
NTL.phys.data.PeterPaul1 <- mutate(NTL.phys.data.PeterPaul1, month = month(sampledate))
#lubidate lets you just pull out month

# reorder columns to put month with the rest of the date variables
NTL.phys.data.PeterPaul1 <- select(NTL.phys.data.PeterPaul1, lakeid:daynum, month, sampledate:comments)
#just resave by selecting all the columns but in a different order

# find out the start and end dates of the dataset
interval(NTL.phys.data.PeterPaul1$sampledate[1], NTL.phys.data.PeterPaul1$sampledate[21613]) #if ordered

## [1] 1984-05-27 UTC--2016-08-16 UTC

interval(first(NTL.phys.data.PeterPaul1$sampledate), last(NTL.phys.data.PeterPaul1$sampledate)) #gets start and end

## [1] 1984-05-27 UTC--2016-08-16 UTC
```

## Pipes

Sometimes we will want to perform multiple functions on a single dataset on our way to creating a processed dataset. We could do this in a series of subsequent functions or create a custom function. However, there is another method to do this that looks cleaner and is easier to read. This method is called a pipe. We designate a pipe with %>%. A good way to think about the function of a pipe is with the word “then.”

Let's say we want to take our raw dataset (NTL.phys.data), *then* filter the data for Peter and Paul lakes, *then* select temperature and observation information, and *then* add a column for temperature in Fahrenheit:

```
NTL.phys.data.processed <-  
  NTL.phys.data %>%  
    filter(lakename == "Paul Lake" | lakename == "Peter Lake") %>%  
    select(lakename, sampledate:temperature_C) %>%  
    mutate(temperature_F = (temperature_C*9/5) + 32)  
#make a list of steps to do %>% basically means then  
#dont need to provide the dataset in all of the functions
```

Notice that we did not place the dataset name inside the wrangling function but rather at the beginning.

### Saving processed datasets

be sure to save things as csv once it is tidy give path and file name

```
write.csv(NTL.phys.data.PeterPaul1, row.names = FALSE, file = "./Data/Processed/NTL-LTER_Lake_Chemistry")
```

### Closing Discussion

When we wrangle a raw dataset into a processed dataset, we create a code file that contains only the wrangling code. We then save the processed dataset as a new spreadsheet and then create a separate code file to analyze and visualize the dataset. Why do we keep the wrangling code separate from the analysis code?