### **NERRS/SWMP**

Training Workshop: R Intro & SWMPr

October 25, 2015

## SWMPr overview, retrieve, and organize

Marcus W. Beck<sup>1</sup> Todd D. O'Brien<sup>2</sup>

<sup>1</sup>ORISE, USEPA NHEERL Gulf Ecology Division Email: beck.marcus@epa.gov

> <sup>2</sup>NOAA/NMFS COPEPOD Project Email: todd.obrien@noaa.gov

## Objectives for the session

- Why and what is SWMPr?
- How can data get from CDMO into R using SWMPr?
- What is the basic structure of a swmpr data object?
- What is data organization and how can SWMPr help?

### Interactive portion

We will use the swmpr1. Rproj project for this session, double-click to open in RStudio

- location on flash drive
- location online

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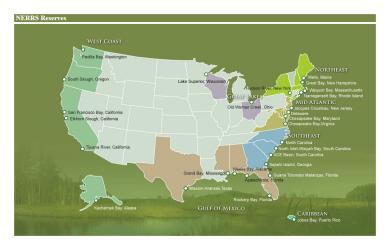
- location on flash drive
- location online

You will run examples whenever you see this guy:



Don't forget to use your stickies: green for done/ok, red for problem

SWMP - System Wide Monitoring Program, initiated in 1995 to provide continuous monitoring data at over 140 stations in 28 US estuaries



CDMO is your one-stop shop for retrieving SWMP data



The raw data will look like this...

4	Α	В	С	D	E	F	G	Н	I	J	K	L
1	StationCo	isSWMP	DateTimeStamp	Historical	Provisiona	CollMetho	REP	F_Record	PO4F	F_PO4F	NH4F	F_NH4F
2	apacpnut	P	1/10/2012 10:20	0	1	1	1		0.003	<-4>[SBL]	0.03	<0>
3	apacpnut	P	2/7/2012 11:41	0	1	1	1		0.005	<0>	0.019	<0>
4	apacpnut	P	3/5/2012 11:51	0	1	1	1		0.003	<-4>[SBL]	0.041	<0>
5	apacpnut	P	4/4/2012 10:30	0	1	1	1		0.003	<-4>[SBL]	0.043	<0>
6	apacpnut	P	5/9/2012 10:12	0	1	1	1		0.003	<0>	0.053	<0>
7	apacpnut	P	5/9/2012 10:15	0	1	1	2		0.003	<-4>[SBL]	0.022	<0>
8	apacpnut	P	5/9/2012 10:20	0	1	1	3		0.003	<0>	0.016	<0>
9	apacpnut	P	6/5/2012 8:30	0	1	1	1		0.003	<-4>[SBL]	0.04	<0>
10	apacpnut	P	7/3/2012 9:58	0	1	1	1	{CSM}	0.004	<0>	0.094	<0>
11	apacpnut	P	7/3/2012 9:59	0	1	1	2	{CSM}	0.004	<0>	0.066	<0>
12	apacpnut	P	7/3/2012 10:01	0	1	1	3	{CSM}	0.005	<0>	0.069	<0>
13	apacpnut	P	8/7/2012 9:53	0	1	1	1	{CSM}	0.003	<-4>[SBL]	0.05	<0>
14	apacpnut	P	9/5/2012 10:56	0	1	1	1		0.003	<-4>[SBL]	0.026	<0>
15	apacpnut	P	10/2/2012 9:22	0	1	1	1		0.003	<-4>[SBL]	0.042	<0>
16	apacpnut	P	10/2/2012 9:27	0	1	1	2		0.003	<-4>[SBL]	0.024	<0>
17	apacpnut	P	10/2/2012 9:32	0	1	1	3		0.003	<0>	0.042	<0>
18	apacpnut	P	11/6/2012 10:30	0	1	1	1		0.003	<-4>[SBL]	0.07	<0>
19	apacpnut	P	11/26/2012 11:39	0	1	1	1		0.003	<-4>[SBL]	0.041	<0>

What are the challenges for evaluating SWMP data?

What are the challenges for evaluating SWMP data?

- Knowing what we want
- Dealing with QAQC columns and removing 'bad' observations
- Data we don't want... extra columns or irrelevant parameters
- Combining data for comparison
- Issues inherent with time series, e.g., missing data
- Others?



**What**: An R package to **augment** existing CDMO services and to provide a **bridge** to analysis



 ${\it What}$ : An R package to  ${\it augment}$  existing CDMO services and to provide a  ${\it bridge}$  to analysis

**Why**: There are many challenges working with SWMP data... a toolkit for addressing these challenges will be useful



 ${\it What}$ : An R package to  ${\it augment}$  existing CDMO services and to provide a  ${\it bridge}$  to analysis

**Why**: There are many challenges working with SWMP data... a toolkit for addressing these challenges will be useful

**How**: Use the SWMPr functions to **retrieve**, **organize**, and **analyze** SWMP data

#### Some housekeeping...

```
# install from CRAN (only do once)
install.packages('SWMPr')

# load for your current session
library(SWMPr)
```

https://cran.r-project.org/web/packages/SWMPr/index.html



Uses an *object-oriented* structure... data are imported into R as a swmpr data object, with functions built to use this object

What are the **retrieve**, **organize**, and **analyze** functions?

Run this code one line at a time... What comes up?

```
help.search('retrieve', package = 'SWMPr')
help.search('organize', package = 'SWMPr')
help.search('analyze', package = 'SWMPr')
```

Uses an *object-oriented* structure... data are imported into R as a swmpr data object, with functions built to use this object

What are the **retrieve**, **organize**, and **analyze** functions?

Run this code one line at a time... What comes up?

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help.search('retrieve', package = 'SWMPr')
help.search('organize', package = 'SWMPr')
help.search('analyze', package = 'SWMPr')
```

What about this?

```
?import local
```

What pieces of information are in the help file?

Let's get some data into R!

The *retrieval* functions do two things:

Import data directly from the CDMO:

all\_params
all\_params\_dtrng
single\_param
site\_codes
site\_codes\_ind

These functions require registering your IP address with CDMO Import data from a local path:

import\_local

Imports data obtained from (and only from) the zip downloads feature

The 'zip\_ex' folder in the project is a sample dataset that looks exactly like a folder you get from CDMO

Let's import some data from that folder, try to import 'apacpwq'...

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Let's import some data from that folder, try to import 'apacpwq'...

```
# get data for apacpwq, all years

# location of data
mypath <- 'zip_ex'

# import and assign to 'dat'
dat <- import_local(mypath, 'apacpwq', trace = T)</pre>
```

The 'zip\_ex' folder in the project is a sample dataset that looks exactly like a folder you get from CDMO

Let's import some data from that folder, try to import 'apacpwq'...

```
# get data for apacpwq, all years
# location of data
mypath <- 'zip_ex'
# import and assign to 'dat'
dat <- import_local(mypath, 'apacpwq', trace = T)</pre>
```

#### What about this?

```
dat2 <- import_local(mypath, 'apacp2012', trace = T)
dat3 <- import_local(mypath, 'apadbnut', trace = F)</pre>
```

# Structure of the swmpr data object



Now we have data in our 'workspace' that we can organize/analyze

Try running the following...

```
head(dat)
tail(dat)
View(dat)
str(dat)
attributes(dat)
```

# Structure of the swmpr data object



Now we have data in our 'workspace' that we can organize/analyze

Try running the following...

```
head(dat)
tail(dat)
View(dat)
str(dat)
attributes(dat)
```

How are the data organized?

What are the column names?

What are the attributes?

### Structure of the swmpr data object

The swmpr object is a data.frame and a list of attributes

```
head(dat, 3)
         datetimestamp temp f_temp spcond f_spcond sal f_sal do_pct f_do_pct
## 1 2011-01-01 00:00:00 11
                          < 0>
                                  44
                                       < 0>
                                            28 <0>
                                                             < 0>
## 2 2011-01-01 00:15:00 11
                         <0> 44
                                    <0>
                                                < 0>
                                                       68
                                                            < 0>
## 3 2011-01-01 00:30:00 11 <0>
                              44
                                    <0>
                                            28 <0>
                                                   68
                                                          <0>
    do_mgl f_do_mgl depth f_depth cdepth f_cdepth level f_level clevel f_clevel
## 1
             < 0>
                    2
                        < 0>
                                 2
                                      <3>
                                            NA <-1>
                                                         NΑ
## 2
             <0>
                    2 <0>
                                      <3>
                                            NA <-1>
                                                         NA
                                                                NA
                2 <0>
## 3
    6 <0>
                                 2 <3> NA <-1>
                                                        NA
                                                                NA
    ph f ph turb f turb chlfluor f chlfluor
    8 <0>
             3 <0>
                         NA
                                <-1>
## 2 8 <0> 3 <0>
                         NA <-1>
## 3 8 <0> 2 <0>
                      NA <-1>
names(attributes(dat))
## [1] "names" "row.names" "class" "station"
                                                    "parameters"
## [6] "qaqc cols" "date rng" "timezone" "stamp class"
attr(dat, 'parameters')
   [1] "temp" "spcond" "sal" "do pct" "do mgl"
                                                    "depth"
##
   [7] "cdepth" "level" "clevel"
                                  "ph"
                                           "turb"
                                                    "chlfluor"
```

First problem is solved... we know how to get SWMP data from CDMO into R:

- Download a dataset from zip downloads
- Find where the data have downloaded
- Import using import\_local
- Have a look at the data (head, View, attributes)
- Lost? Check the help files: ?import\_local

Now we can think about preprocessing or organizing prior to analysis

What are the challenges for evaluating SWMP data?

What are the challenges for evaluating SWMP data?

- Knowing what we want
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Take a few minutes to acquaint yourself with the *organize* functions:

```
help.search('organize', package = 'SWMPr')
```

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```
help.search('organize', package = 'SWMPr')
```

Which function would you use first?

Which would you use to reduce data volume or select certain variables?

Can any be used to combine swmpr data objects?



Perhaps you want to deal with QAQC columns first...

From the zips folder, import all of the weather data for apaelmet (?import\_local)



Perhaps you want to deal with QAQC columns first...

From the zips folder, import all of the weather data for apaelmet (?import\_local)

```
# import data
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apaebmet')</pre>
```



Perhaps you want to deal with QAQC columns first...

From the zips folder, import all of the weather data for apaelmet (?import\_local)

```
# import data
mvpath <- 'zip ex'
dat <- import_local(mypath, 'apaebmet')</pre>
```

View the data, what are the columns?

Try running qaqc (?qaqc) and view again, what happened?



Perhaps you want to deal with QAQC columns first...

From the zips folder, import all of the weather data for apaelmet (?import\_local)

```
# import data
mypath <- 'zip ex'
dat <- import_local(mypath, 'apaebmet')</pre>
```

View the data, what are the columns?

Try running gage (?gage) and view again, what happened?

```
View(dat)
dat2 <- qaqc(dat)</pre>
View(dat2)
```



Try playing with the qaqc\_keep argument (?qaqc)...

How are these different?

```
# different options for gage
dat2 <- gagc(dat)</pre>
dat3 <- qaqc(dat, qaqc_keep = c('0', '-1'))
dat4 <- qaqc(dat, qaqc_keep = NULL)</pre>
dat5 <- gagc(dat, gagc_keep = 'CSM')</pre>
```



Try playing with the qaqc\_keep argument (?qaqc)...

How are these different?

```
# different options for gage
dat2 <- gagc(dat)</pre>
dat3 <- qaqc(dat, qaqc_keep = c('0', '-1'))
dat4 <- gagc(dat, gagc keep = NULL)</pre>
dat5 <- gagc(dat, gagc_keep = 'CSM')</pre>
```

Changes are hard to visualize for lots of data - as a proof of concept, try running qaqcchk on any of the datasets

```
qaqcchk(dat)
gaqcchk(dat2)
gaqcchk(dat3)
qaqcchk(dat4)
gagcchk(dat5)
```



We'll continue with the water quality data for apadb - import again and run the qaqc function



We'll continue with the water quality data for apadb - import again and run the qaqc function

```
# import apawa
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apadbwg')</pre>
dat <- qaqc(dat)</pre>
```



We'll continue with the water quality data for apadb - import again and run the qaqc function

```
# import apawa
mypath <- 'zip ex'
dat <- import_local(mypath, 'apadbwg')</pre>
dat <- qaqc(dat)</pre>
```

What is the next logical step after dealing with QAQC values?

How would we further want to organize the data?



We'll continue with the water quality data for apadb - import again and run the qaqc function

```
# import apawa
mvpath <- 'zip ex'
dat <- import_local(mypath, 'apadbwg')</pre>
dat <- qaqc(dat)</pre>
```

What is the next logical step after dealing with QAQC values?

How would we further want to organize the data?

Maybe we want to subset the data...

```
# view help file
?subset.swmpr
```

The subset function has several arguments (help file ?subset.swmpr)

Not all are necessary for every task

- swmpr\_in: input data (swmpr object)
- subset: dates to keep
- select: parameters to keep
- operator: less than, greater than, etc. if only one date in subset
- rem\_rows: remove empty rows
- rem\_cols: remove empty columns

The select argument of subset is used to select parameters of interest - one to many

```
# select the DO column
tmp <- subset(dat, select = 'do_mgl')</pre>
head(tmp)
           datetimestamp do mgl
     2011-01-01 00:00:00
                              NΑ
   2 2011-01-01 00:15:00
                              NΑ
   3 2011-01-01 00:30:00
                              NA
  4 2011-01-01 00:45:00
                              NΑ
   5 2011-01-01 01:00:00
                              NA
## 6 2011-01-01 01:15:00
                              NA
```

Selecting more than one column...

The select argument of subset is used to select parameters of interest - one to many

```
# select the DO column
tmp <- subset(dat, select = 'do mgl')</pre>
head(tmp)
           datetimestamp do mgl
## 1 2011-01-01 00:00:00
                              NΑ
## 2 2011-01-01 00:15:00
                              NΑ
## 3 2011-01-01 00:30:00
                         NA
## 4 2011-01-01 00:45:00
                             NΑ
## 5 2011-01-01 01:00:00
                              NA
## 6 2011-01-01 01:15:00
                              NA
```

Selecting more than one column...

```
# select DO and salinity
tmp <- subset(dat, select = c('do_mgl', 'sal'))
head(tmp)</pre>
```

The subset argument of subset.swmpr selects a date range

The dates must have a specific format: 'YYYY-mm-dd HH:MM'

```
# select a date range, July 2012
dates <- c('2012-07-01 12:00', '2012-07-31 6:30')
tmp <- subset(dat, subset = dates)</pre>
head(tmp) # view first six rows
##
          datetimestamp temp spcond sal do_pct do_mgl depth cdepth level clevel
    2012-07-01 12:00:00
                                 50
                                     33
                                           104
                                                                NΑ
                          NΑ
                                                                      NΑ
                                                                             NΑ
  2 2012-07-01 12:15:00
                                 50 33
                                          101
                          NΑ
                                                                NΑ
                                                                      NΑ
                                                                             NΑ
  3 2012-07-01 12:30:00
                          NA
                                 50 33 104
                                                                NA
                                                                      NA
                                                                             NA
    2012-07-01 12:45:00
                          NΑ
                                 50 33 104
                                                                NΑ
                                                                      NΑ
                                                                             NΑ
  5 2012-07-01 13:00:00
                          NA
                                 50 33 104
                                                                NA
                                                                      NA
                                                                             NA
## 6 2012-07-01 13:15:00
                          NA
                                 52 34
                                          104
                                                                NA
                                                                      NA
                                                                             NA
    ph turb chlfluor
##
## 1
     8
          3
                  NA
## 2
     8 11
                  NΑ
## 3
                  NΑ
## 4
        10
                  NA
## 5
         15
                  NΑ
         12
                  NA
## 6
```



• Import the weather data at apaeb

• Deal with QAQC columns

• Select two columns of interest



• Import the weather data at apaeb

```
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apaebmet')</pre>
```

• Deal with QAQC columns

• Select two columns of interest



• Import the weather data at apaeb

```
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apaebmet')</pre>
```

• Deal with QAQC columns

```
tmp <- qaqc(dat)</pre>
```

• Select two columns of interest





```
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apaebmet')</pre>
```

• Deal with QAQC columns

```
tmp <- gagc(dat)
```

• Select two columns of interest

```
tmp <- subset(tmp, select = c('temp', 'wind'))</pre>
```



```
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apaebmet')</pre>
```

• Deal with QAQC columns

```
tmp <- gagc(dat)
```

• Select two columns of interest

```
tmp <- subset(tmp, select = c('temp', 'wind'))</pre>
```

```
dates <- c('2012-01-01 0:0', '2012-01-31 0:0')
tmp <- subset(tmp, subset = dates)</pre>
```





```
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apaebmet')</pre>
```

• Deal with QAQC columns

```
tmp <- gagc(dat)
```

Select two columns of interest

```
tmp <- subset(tmp, select = c('temp', 'wind'))</pre>
```

Subset a date range

```
dates <- c('2012-01-01 0:0', '2012-01-31 0:0')
tmp <- subset(tmp, subset = dates)</pre>
```

• Bonus: can you select all observations before or after a date?





```
mypath <- 'zip_ex'</pre>
dat <- import_local(mypath, 'apaebmet')</pre>
```

• Deal with QAQC columns

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tmp <- gagc(dat)
```

• Select two columns of interest

```
tmp <- subset(tmp, select = c('temp', 'wind'))</pre>
```

Subset a date range

```
dates <- c('2012-01-01 0:0', '2012-01-31 0:0')
tmp <- subset(tmp, subset = dates)</pre>
```

• Bonus: can you select all observations before or after a date?

```
# get observations after Jan 1, 2013
dates <- '2013-01-01 00:00'
tmp <- subset(dat, subset = dates, operator = '>=')
```



A final nod to the comb and setstep functions

Run the following, view the results, discuss with your neighbors:

```
mvpath <- 'zip ex'
dat met <- import local(mypath, 'apaebmet')</pre>
dat_met <- qaqc(dat_met)</pre>
dat_wq <- import_local(mypath, 'apadbwq')</pre>
dat_wq <- qaqc(dat_wq)</pre>
# what does this do (hint: use View to see the data)?
tmp1 <- comb(dat_wq, dat_met, timestep = 120)</pre>
```





A final nod to the comb and setstep functions

Run the following, view the results, discuss with your neighbors:

```
mvpath <- 'zip ex'
dat met <- import local(mypath, 'apaebmet')</pre>
dat_met <- qaqc(dat_met)</pre>
dat_wq <- import_local(mypath, 'apadbwq')</pre>
dat_wq <- qaqc(dat_wq)</pre>
# what does this do (hint: use View to see the data)?
tmp1 <- comb(dat_wq, dat_met, timestep = 120)</pre>
```

Now try this...

```
tmp2 <- setstep(dat_wq, timestep = 60)</pre>
```

What happened?

The setstep function is used to standardize the time step of a swmpr object

The comb function is used to combine swmpr objects

setstep is used within comb so you should not have to use it directly

Arguments for comb:

- ...: input swmpr data, separated by comma
- timestep: minutes defining the standardized time step
- differ: maximum difference in minutes for matching observations with original time steps to standardized time steps
- method: how the data are combined using the time stamps union, intersect, or using a station

A final note about combining... what about combining data with different *time ranges*?

Consider combining two datasets

The method argument of allows flexibility under different scenarios time range 'intersect', 'union', or range of one station

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A final note about combining... what about combining data with different *time ranges*?

Consider combining two datasets

• Scenario 1: Time ranges are the same

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- Scenario 1: Time ranges are the same
  - ▶ all methods can be used, same results for each

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A final note about combining... what about combining data with different *time ranges*?

Consider combining two datasets

- Scenario 1: Time ranges are the same
  - ▶ all methods can be used, same results for each
- Scenario 2: Time ranges are not the same, but there is overlap

The method argument of allows flexibility under different scenarios time range 'intersect', 'union', or range of one station

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- Scenario 1: Time ranges are the same
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- Scenario 2: Time ranges are not the same, but there is overlap
  - ▶ all methods can be used, different results for each
- Scenario 3: Time ranges are not the same, there is no overlap

The method argument of allows flexibility under different scenarios time range 'intersect', 'union', or range of one station

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A final note about combining... what about combining data with different *time ranges*?

Consider combining two datasets

- Scenario 1: Time ranges are the same
  - all methods can be used, same results for each
- Scenario 2: Time ranges are not the same, but there is overlap
  - ▶ all methods can be used, different results for each
- Scenario 3: Time ranges are not the same, there is no overlap
  - only 'union' will work

The method argument of allows flexibility under different scenarios time range 'intersect', 'union', or range of one station

#### **Summary**

Now you have an idea of how to organize SWMP data for analysis!

Here's what we did:

- Import SWMP data into R
- ullet Evaluate and  $handle\ QAQC$  flags in the data
- Subset to select variables or time ranges of interest
- Combine data for comparison or data simplification

Consult the SWMP cookbook for an example workflow

Check the help files for usage (reference manual on CRAN)

## A final exercise

*Import* 2012 apadbwq and apaebmet data, deal with *QAQC*, *subset* one month of data and one variable of interest, *combine* at hourly time step and intersect method, and *plot* two variables against each other...

## A final exercise

Import 2012 apadbwq and apaebmet data, deal with QAQC, subset one month of data and one variable of interest, combine at hourly time step and intersect method, and plot two variables against each other...

?import\_local, ?qaqc, ?subset.swmpr, ?comb, ?plot



*Import* 2012 apadbwq and apaebmet data, deal with *QAQC*, *subset* one month of data and one variable of interest, *combine* at hourly time step and intersect method, and *plot* two variables against each other...

?import\_local, ?qaqc, ?subset.swmpr, ?comb, ?plot

```
wqdat <- import_local('zip_ex', 'apadbwq2012')
wqdat <- qaqc(wqdat)
wqdat <- subset(wqdat, select = 'turb',
    subset = c('2012-08-01 0:0', '2012-08-31 0:0'))
metdat <- import_local('zip_ex', 'apaebmet2012')
metdat <- qaqc(metdat)
metdat <- subset(metdat, select = 'wspd',
    subset = c('2012-08-01 0:0', '2012-08-31 0:0'))
dat <- comb(wqdat, metdat, method = 'intersect', timestep = 60)
plot(turb " wspd, data = dat)</pre>
```



*Import* 2012 apadbwq and apaebmet data, deal with *QAQC*, *subset* one month of data and one variable of interest, *combine* at hourly time step and intersect method, and *plot* two variables against each other...

?import\_local, ?qaqc, ?subset.swmpr, ?comb, ?plot

```
wqdat <- import_local('zip_ex', 'apadbwq2012')
wqdat <- qaqc(wqdat)
wqdat <- subset(wqdat, select = 'turb',
    subset = c('2012-08-01 0:0', '2012-08-31 0:0'))
metdat <- import_local('zip_ex', 'apaebmet2012')
metdat <- qaqc(metdat)
metdat <- subset(metdat, select = 'wspd',
    subset = c('2012-08-01 0:0', '2012-08-31 0:0'))
dat <- comb(wqdat, metdat, method = 'intersect', timestep = 60)
plot(turb ~ wspd, data = dat)</pre>
```

Bonus: can you plot a regression line on this plot? Hint: ?lm, ?abline

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After break... what are some ways we can analyze or visualize the data?

#### $Questions \ref{eq:questions}$