

# DEBY\_Raw\_Envr\_Data

2024-04-03

```
setwd("/Users/nicolemongillo/Desktop/GitHub/MVP_Chesapeake_VIMS_hatchery/src/NM")
```

Load required packages.

```
library("dplyr") #Used for working with data frames
```

```
##  
## 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
```

```
library("lubridate") #Used for time-date conversions
```

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

```
library("readr") #Used to read the CSV file  
library("ggplot2")
```

Note the date of data download and source. All available data should be used for each site regardless of year. Note from the CSV file how often the site was sampled, and if there are replicates in the data. Also describe if the sampling occurred at only low tide, only high tide, or continuously.

```
#Data was downloaded on 04/02/2024  
#Source - http://vecos.vims.edu/StationDetail.aspx?param=YRK005.40&program=CMON  
#The site was sampled continuously every 15 min from 2003
```

```
#Create text strings with metadata information that we want to include in the final data frame.
download_date <- ("04-02-2024")
source_description <- ("Virginia Estuarine and Coastal Observing System, VIMS")
site_name <- ("DEBY")
collection_type <- ("continuous")
```

Use the file path name in your working directory or desktop, see example below. Or, import data set through the “Files” window in R studio. Store the file in a variable with the “raw\_ID\_Site” format. If salinity and temperature data are in separate files, read in both and store them with “\_sal” or “\_temp” in the variable names.

```
#The files we will be working with are from Gloucester Point, VA, which is for the DEBY selection line.

#Environmental data could only be downloaded by year, so first we need to merge the yearly data sets.

#set working directory to files location
setwd("/Users/nicolemongillo/Desktop/GitHub/MVP_Chesapeake_VIMS_hatchery/data/envr_raw_data/DEBY_York")

#merge files into one
raw_DEBY_env <- list.files(path="/Users/nicolemongillo/Desktop/GitHub/MVP_Chesapeake_VIMS_hatchery/data",
  lapply(read.csv) %>%
  bind_rows

#The metadata for these data (located at the bottom of this site: http://vecos.vims.edu/Content.aspx?id)

subset_DEBY_env_error <- subset(raw_DEBY_env, select = c(SAMPLE_DATETIME, WTEMP, WTEMP_A, SALINITY, SALINITY_A))

subset1_DEBY_env <- subset_DEBY_env_error[(subset_DEBY_env_error$WTEMP_A %in% c("", NA)), ] #keep only rows where WTEMP_A is not empty

subset2_DEBY_env <- subset1_DEBY_env[(subset1_DEBY_env$SALINITY_A %in% c("", NA)),] #keep only rows where SALINITY_A is not empty

View(subset2_DEBY_env)
#sorting the data frame by ascending and descending values for WTEMP_A and SALINITY_A shows only blank values

subset_DEBY_env <- subset(subset2_DEBY_env, select = c(SAMPLE_DATETIME, WTEMP, SALINITY)) #remove WTEMP_A and SALINITY_A

# View how the data are stored. Note the variable names and the format and units that the data are stored in
summary(subset_DEBY_env)
```

```
## SAMPLE_DATETIME      WTEMP      SALINITY
## Length:667345      Min.   : 0.020      Min.   : 8.06
## Class :character    1st Qu.: 9.639      1st Qu.:17.69
## Mode  :character    Median :17.300      Median :19.68
##                      Mean   :17.122      Mean   :19.34
##                      3rd Qu.:25.010      3rd Qu.:21.31
##                      Max.   :32.290      Max.   :26.09
```

```
#rename columns. "datetime" = date and time of data collection, "temp" = water temperature in degrees Celsius, "sal" = salinity in practical salinity units
colnames(subset_DEBY_env) <- c("datetime", "temp", "salinity")
```

Start with the date and time of collection. We will use the lubridate package to standardize all values into the date-time format called POSIXct. This format stores the date and time in number of seconds since a past point (1/1/1970). This makes comparisons easy and helps to standardizes values.

```
#Convert to POSIXct format. Tell R what the current date/time format is so it knows how to convert it.
subset_DEBY_env$datetime <- as.POSIXct(subset_DEBY_env$datetime, "%m/%d/%Y %H:%M:%S %p", tz = "")

#Print the new data frame and examine to make sure the new datetime column is in the correct format.
head(subset_DEBY_env)
```

```
##           datetime  temp salinity
## 1 2003-05-28 04:15:00 19.07   15.57
## 2 2003-05-28 04:30:00 19.15   15.45
## 3 2003-05-28 04:45:00 19.38   15.49
## 4 2003-05-28 05:00:00 19.81   15.58
## 5 2003-05-28 05:15:00 19.80   15.60
## 6 2003-05-28 05:30:00 19.89   15.56
```

#Standardize column and variable names. We will use “lat” for latitude in degrees, and “lon” for longitude in degrees.

```
#Store variables that we will include in the final data frame
lat <- 37.247284
lon <- -76.499369
firstyear <- 2003
finalyear <- 2024
```

Filter any of the variables that have data points outside of normal range. We will use 0-40 as the accepted range for salinity (ppt) and temperature (C) values. Note, in the summer, salinity values can sometimes exceed 40. Check to see if there are values above 40. In this case, adjust the range or notify someone that the site has particularly high salinity values.

```
#Filter the data between the values of 0 and 40 for both salinity and temperature.
filtered_DEBY_sal <- subset_DEBY_env %>%
  filter(between(salinity, 0, 40))

summary(subset_DEBY_env)
```

```
##           datetime           temp           salinity
## Min.   :2003-05-28 04:15:00.00 Min.    : 0.020 Min.    : 8.06
## 1st Qu.:2008-12-20 09:30:00.00 1st Qu.: 9.639 1st Qu.:17.69
## Median :2014-01-13 02:30:00.00 Median :17.300 Median :19.68
## Mean   :2013-12-24 11:40:09.11 Mean    :17.122 Mean    :19.34
## 3rd Qu.:2019-01-08 04:00:00.00 3rd Qu.:25.010 3rd Qu.:21.31
## Max.   :2024-01-11 12:45:00.00 Max.     :32.290 Max.     :26.09
## NA's   :152
```

```

filtered_DEBY_env <- filtered_DEBY_sal %>%
  filter(between(temp, 0, 40))

# Sanity check - print the ranges to ensure values are filtered properly. We can see that the ranges for
print(summary(filtered_DEBY_env$salinity))

```

```

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      8.06   17.69   19.68   19.34   21.31   26.09

```

```

print(summary(filtered_DEBY_env$temp))

```

```

##      Min. 1st Qu.  Median    Mean 3rd Qu.    Max.
##      0.020   9.639  17.300  17.122  25.010  32.290

```

```

#Store our data into a variable name with just the site name.
DEBY_env <- filtered_DEBY_env

# we have NAs in the our data frame in the datetime column - need to remove these
count.nas_env <- is.na(DEBY_env$datetime) # store our NAs in a variable
summary(count.nas_env) # we have 152 NAs that are stored as "TRUE" in our count.nas

```

```

##      Mode  FALSE    TRUE
## logical 667193    152

```

```

nrow(DEBY_env) # figure out how many rows we have in the original df: 667345

```

```

## [1] 667345

```

```

which(count.nas_env == TRUE) # find the number of NA rows that we need to remove: 152

```

```

##      [1] 19905 19906 19907 19908 19953 19954 19955 19956 46203 46204
##      [11] 46205 46206 46251 46252 46253 46254 79678 79679 79680 79681
##      [21] 79726 79727 79728 79729 108089 108090 108091 108092 108137 108138
##      [31] 108139 108140 141594 141595 141596 141597 141642 141643 141644 141645
##      [41] 174170 174171 174172 174173 174218 174219 174220 174221 206883 206884
##      [51] 206885 206886 206931 206932 206933 206934 237216 237217 237218 237219
##      [61] 237264 237265 237266 237267 271307 271308 271309 271310 271355 271356
##      [71] 271357 271358 305058 305059 305060 305061 305106 305107 305108 305109
##      [81] 338850 338851 338852 338853 338898 338899 338900 338901 372369 372370
##      [91] 372371 372372 372417 372418 372419 372420 405987 405988 405989 405990
##     [101] 406035 406036 406037 406038 439794 439795 439796 439797 439842 439843
##     [111] 439844 439845 472348 472349 472350 472351 472396 472397 472398 472399
##     [121] 506313 506314 506315 506316 506358 506359 506360 506361 539401 539402
##     [131] 539403 539404 539449 539450 539451 539452 605533 605534 605535 605536
##     [141] 605581 605582 605583 605584 640317 640318 640319 640320 640365 640366
##     [151] 640367 640368

```

```

DEBY_env <- na.omit(DEBY_env) #remove NAs using na.omit
nrow(DEBY_env) #there are 667193 rows in the new data frame

```

```
## [1] 667193
```

```
check_env <- 667345 - 667193 #the value of the check should be 152  
check_env #we removed 152 NA rows!
```

```
## [1] 152
```

```
# check for NAs in our temperature column  
count.nas_temp <- is.na(DEBY_env$temp) # store our NAs in temp in a variable  
summary(count.nas_temp) # we have 0 NAs
```

```
##      Mode      FALSE  
## logical 667193
```

```
#check for NAs in our temperature column  
count.nas_sal <- is.na(DEBY_env$salinity) #store our NAs in salinity in a variable  
summary(count.nas_sal) #we have no NAs
```

```
##      Mode      FALSE  
## logical 667193
```

```
#Data sets for violin plots
```

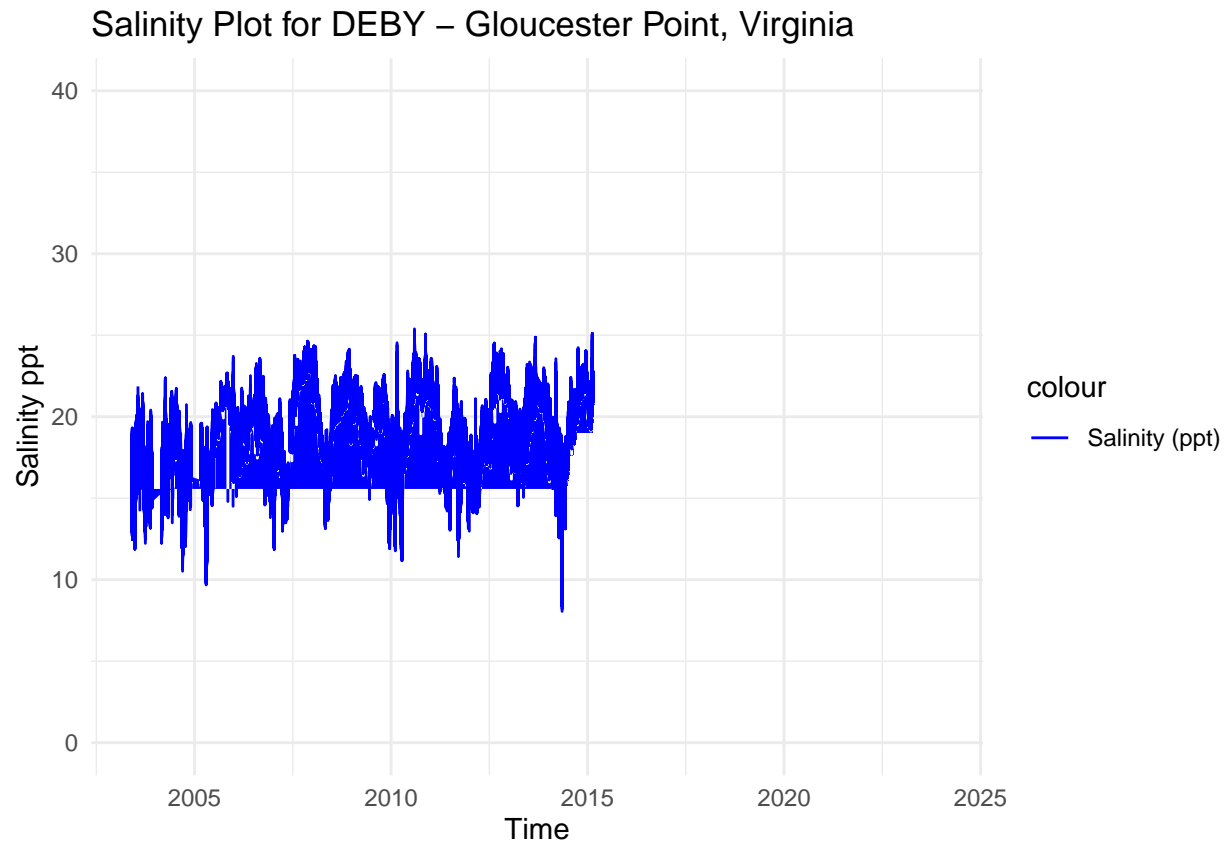
```
#add site name and create new data frame with full envr data set  
DEBY_env_full <- DEBY_env %>%  
  mutate(site_name, site_name = "DEBY")  
  
#reorder columns with site_name first  
DEBY_env_full <- DEBY_env_full[, c(4, 1, 2, 3)]  
  
DEBY_temp_full <- DEBY_env_full[, c(1,2,3)]  
  
DEBY_sal_full <- DEBY_env_full[, c(1, 2, 4)]  
  
#save DEBY_env_full as csv for future analyses  
write.csv(DEBY_temp_full, "/Users/nicolemongillo/Desktop/GitHub/MVP_Chesapeake_VIMS_hatchery/data/envr_r  
write.csv(DEBY_sal_full, "/Users/nicolemongillo/Desktop/GitHub/MVP_Chesapeake_VIMS_hatchery/data/envr_r
```

Visualize the salinity, temperature, and date ranges over time. This can help us see if there are any anomalies or gaps in the data and make sure the filtering was done correctly. Sanity check - do the temperature and salinity ranges look appropriate for the geography of the site (ex. near full ocean salinity for coastal sites, lower salinity for estuaries or near rivers)?

```
salplot <- ggplot(DEBY_env, aes(x = datetime)) +  
  geom_line(aes(y = salinity, color = "Salinity (ppt)")) +  
  ylim(0,40) +  
  labs(x = "Time", y = "Salinity ppt", title = "Salinity Plot for DEBY - Gloucester Point, Virginia") +  
  scale_color_manual(values = c("Salinity (ppt)" = "blue")) +
```

```
theme_minimal()
```

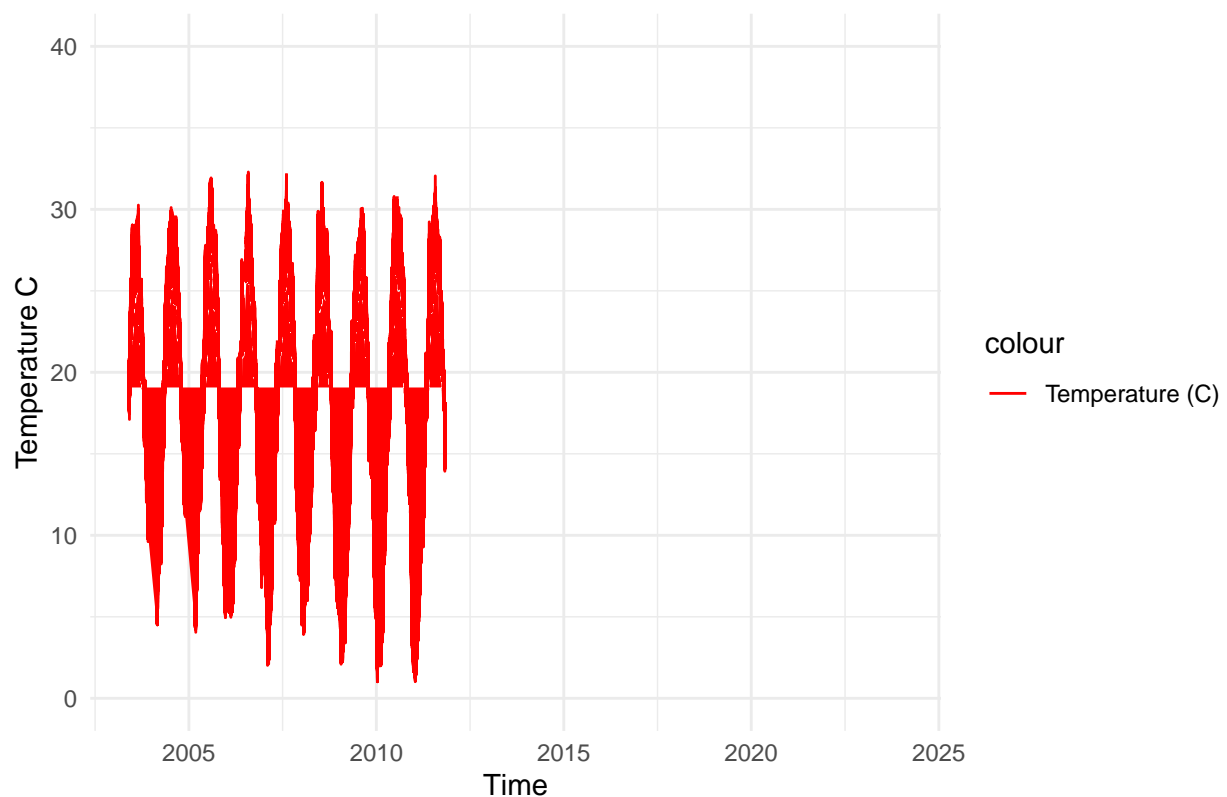
```
salplot
```



```
tempplot <- ggplot(DEBY_env, aes(x = datetime)) +  
  geom_line(aes(y = temp, color = "Temperature (C)")) +  
  ylim(0, 40) +  
  labs(x = "Time", y = "Temperature C", title = "Temperature Plot for DEBY - Gloucester Point, Virginia") +  
  scale_color_manual(values = c( "Temperature (C)" = "red")) +  
  theme_minimal()
```

```
tempplot
```

Temperature Plot for DEBY – Gloucester Point, Virginia



We need to calculate the mean, maximum, and minimum values for salinity and temperature per month and year. First make two data frames to contain each of the annual and monthly averages.

```
#Calculate the mean, maximum, and minimum values for salinity and temperature for each month.
DEBY_envrmonth_sal <- DEBY_env %>%
  mutate(year = year(datetime), month = month(datetime)) %>%
  group_by(year, month) %>%
  summarise(
    min_salinity = min(salinity),
    max_salinity = max(salinity),
    mean_salinity = mean(salinity),
    length_salinity = length(salinity))
```

## 'summarise()' has grouped output by 'year'. You can override using the  
## '.groups' argument.

```
DEBY_envrmonth_temp <- DEBY_env %>%
  mutate(year = year(datetime), month = month(datetime)) %>%
  group_by(year, month) %>%
  summarise(
    min_temp = min(temp),
    max_temp = max(temp),
```

```
mean_temp = mean(temp),
length_temp = length(temp))
```

## 'summarise()' has grouped output by 'year'. You can override using the  
## '.groups' argument.

```
print(DEBY_envrmonth_sal)
```

```
## # A tibble: 246 x 6
## # Groups:   year [22]
##   year month min_salinity max_salinity mean_salinity length_salinity
##   <dbl> <dbl>      <dbl>      <dbl>      <dbl>      <int>
## 1 2003     5      12.7      15.7      14.4        319
## 2 2003     6      11.9      19.3      14.7       2880
## 3 2003     7      14.3      21.9      18.0       2976
## 4 2003     8      14.3      20.6      18.1       2976
## 5 2003     9      13.0      21.4      17.4       2880
## 6 2003    10      12.2      19.7      16.4       2975
## 7 2003    11      13.2      20.4      16.8       2880
## 8 2003    12      14.4      16.2      15.5         33
## 9 2004     2      12.2      15.5      14.1        241
## 10 2004     3      13.0      19.7      16.4       1448
## # i 236 more rows
```

```
print(DEBY_envrmonth_temp)
```

```
## # A tibble: 246 x 6
## # Groups:   year [22]
##   year month min_temp max_temp mean_temp length_temp
##   <dbl> <dbl>      <dbl>      <dbl>      <dbl>      <int>
## 1 2003     5      18.6      20.9      19.6        319
## 2 2003     6      17.1      29.0      22.6       2880
## 3 2003     7      23.7      29.1      26.5       2976
## 4 2003     8      25.0      30.3      27.0       2976
## 5 2003     9      21.7      27.8      24.4       2880
## 6 2003    10      15.4      23.2      19.2       2975
## 7 2003    11       9.61      19.5      14.9       2880
## 8 2003    12      10.4      11.9      11.2         33
## 9 2004     2       4.5       7.64      5.59        241
## 10 2004     3       5.27      12.7      8.63       1448
## # i 236 more rows
```

*#Calculate the mean, maximum, and minimum values for salinity and temperature for each year.*

```
DEBY_envryear_sal <- DEBY_env %>%
  mutate(year = year(datetime)) %>%
  group_by(year) %>%
  summarise(
    min_salinity = min(salinity),
    max_salinity = max(salinity),
    mean_salinity = mean(salinity))
```



```
DEBY_envryear_temp <- DEBY_env %>%
  mutate(year = year(datetime)) %>%
  group_by(year) %>%
  summarise(
    min_temp = min(temp),
    max_temp = max(temp),
    mean_temp = mean(temp))

print(DEBY_envryear_sal)
```

```
## # A tibble: 22 x 4
##   year min_salinity max_salinity mean_salinity
##   <dbl>      <dbl>      <dbl>      <dbl>
## 1 2003         11.9         21.9         16.9
## 2 2004         10.5         22.4         17.2
## 3 2005          9.68         23.7         18.6
## 4 2006         13.8         23.6         19.1
## 5 2007         11.9         24.6         20.1
## 6 2008         13.2         24.4         20.1
## 7 2009         11.9         22.6         19.8
## 8 2010         11.2         25.4         20.1
## 9 2011         11.4         23.6         18.1
## 10 2012         14.1         24.5         19.7
## # i 12 more rows
```

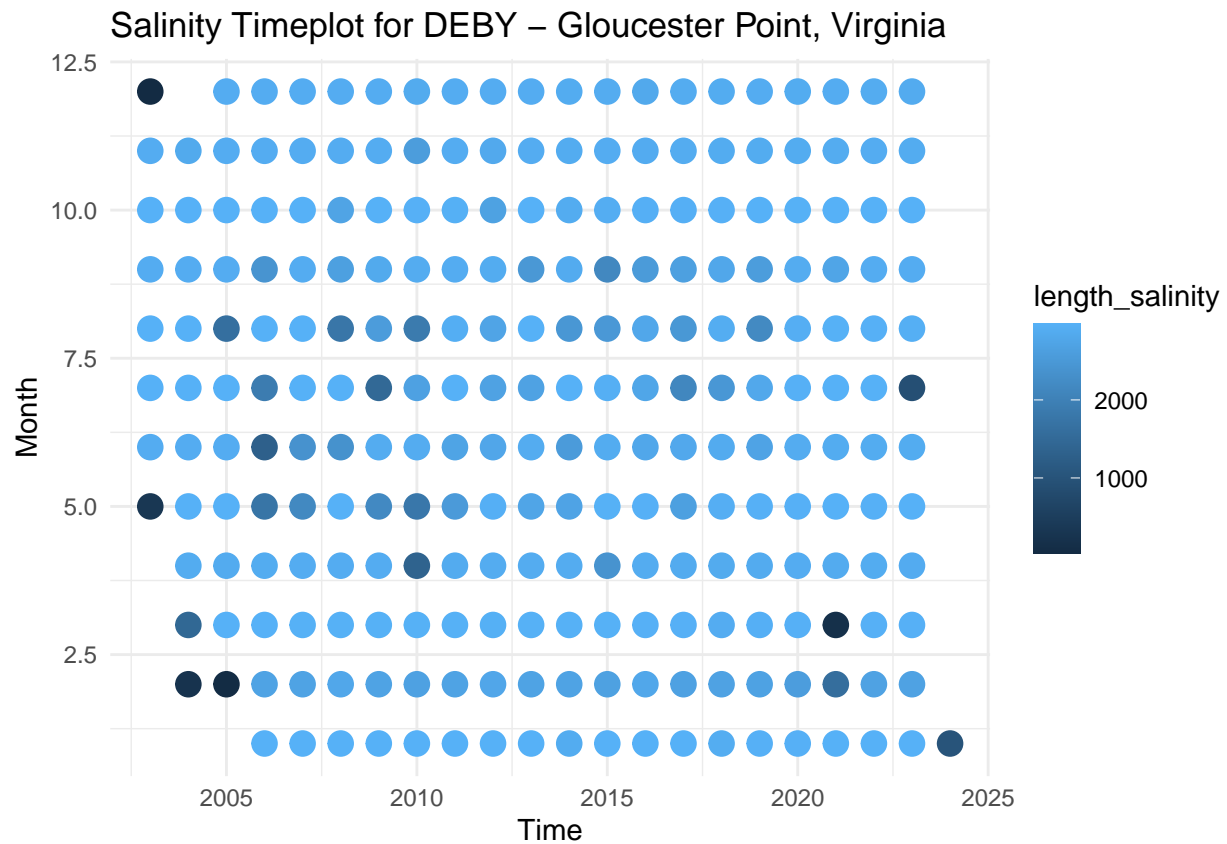
```
print(DEBY_envryear_temp)
```

```
## # A tibble: 22 x 4
##   year min_temp max_temp mean_temp
##   <dbl>   <dbl>   <dbl>   <dbl>
## 1 2003    9.61    30.3    22.4
## 2 2004    4.5     30.1    20.6
## 3 2005    4.05    31.9    18.4
## 4 2006    4.97    32.3    16.0
## 5 2007    2.02    32.2    17.1
## 6 2008    3.92    31.7    16.4
## 7 2009    2.09    30.1    15.9
## 8 2010    1.01    30.8    16.0
## 9 2011    1.02    32.0    16.9
## 10 2012    5.86    30.8    17.5
## # i 12 more rows
```

Plot the months and years of data collection to check if there are any collection gaps in the data.

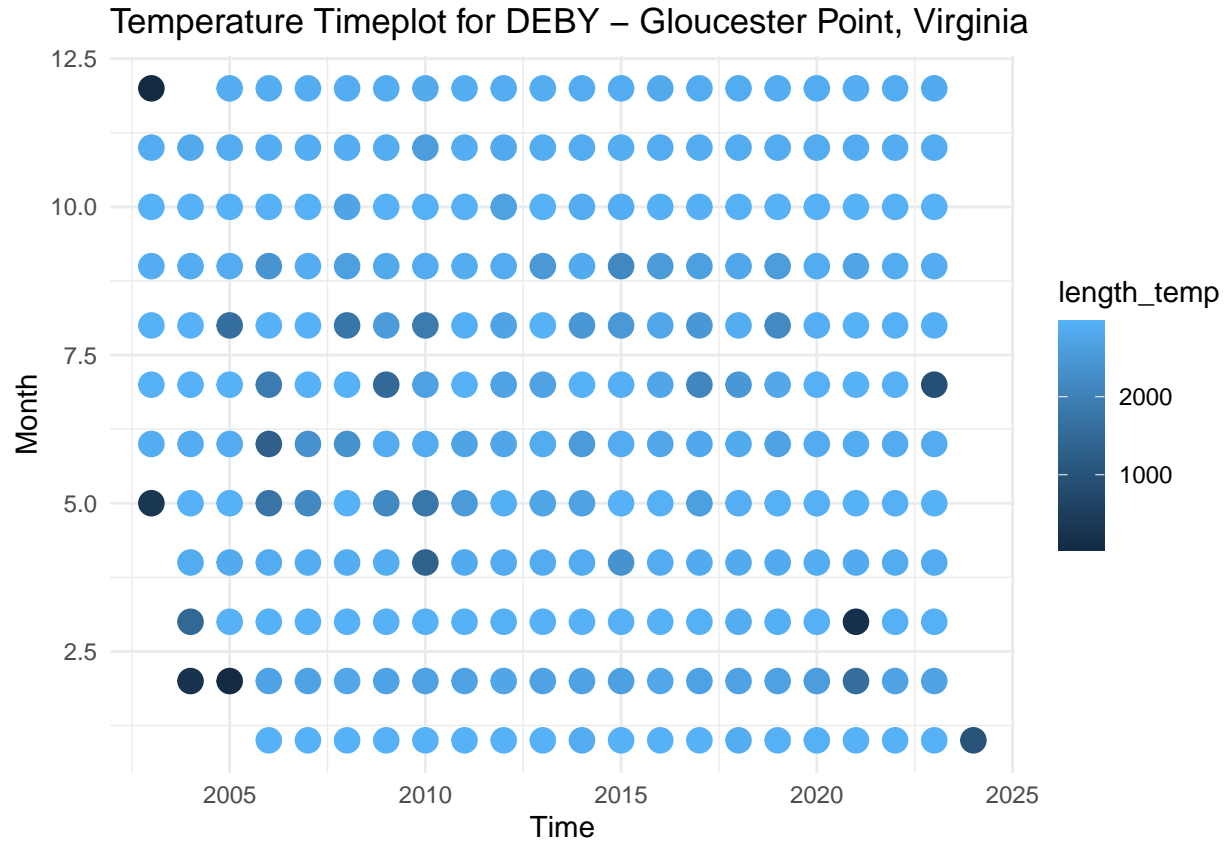
```
timeplot <- ggplot(DEBY_envrmonth_sal, aes(x = year)) +
  geom_point(aes(y = month, color = length_salinity), size = 4) +
  labs(x = "Time", y = "Month", title = "Salinity Timeplot for DEBY - Gloucester Point, Virginia") +
  ylim(1,12) +
  theme_minimal()
```

```
timeplot
```



Plot the months and years of data collection to check if there are any collection gaps in the data.

```
timeplot_temp <- ggplot(DEBY_envrmonth_temp, aes(x = year)) +  
  geom_point(aes(y = month, color = length_temp), size = 4) +  
  labs(x = "Time", y = "Month", title = "Temperature Timeplot for DEBY - Gloucester Point, Virginia")  
  ylim(1,12) +  
  theme_minimal()  
  
timeplot_temp
```



We can now calculate a list of variables that we will have collected for all sites. This will allow us to compare sites easily. We will calculate the number of observations from each site, the mean annual, maximum annual, and minimum annual value for all variables.

Our list of variables includes:

- Mean\_Annual\_Temperature\_C: average of all available data
- Mean\_max\_temperature\_C: average of maximums for each year
- Mean\_min\_temperature\_C: average of minimums for each year
- Temperature\_st\_dev: standard deviation of all available data
- Temperature\_n: total number of data points
- Temperature\_years: number of years in data set
- Mean\_Annual\_Salinity\_ppt: average of all available data
- Mean\_min\_Salinity\_ppt: average of minimums for each year
- Mean\_max\_Salinity\_ppt: average of maximums for each year
- Salinity\_st\_dev: standard deviation of all available data
- Salinity\_n: total number of data points
- Salinity\_years: number of years in data set

```

#Calculate temperature variables.
#Calculate temperature variables.
Mean_Annual_Temperature_C <- mean(DEBY_env$temp)
Mean_max_temperature_C <- mean(DEBY_envryear_temp$max_temp)
Mean_min_temperature_C <- mean(DEBY_envryear_temp$min_temp)
Temperature_st_dev <- sd(DEBY_env$temp)
Temperature_n <- nrow(DEBY_env)
Temperature_years <- nrow(DEBY_envryear_temp)

#Create a data frame to store the temperature results
DEBY_temp <- cbind(site_name, download_date, source_description, lat, lon, firstyear, finalyear, Mean_A
print(DEBY_temp)

```

```

##      site_name download_date
## [1,] "DEBY"      "04-02-2024"
##      source_description      lat
## [1,] "Virginia Estuarine and Coastal Observing System, VIMS" "37.247284"
##      lon      firstyear finalyear Mean_Annual_Temperature_C
## [1,] "-76.499369" "2003"      "2024"      "17.1242424156129"
##      Mean_max_temperature_C Mean_min_temperature_C Temperature_st_dev
## [1,] "30.1437727272727"      "3.54609090909091"      "8.07505645766884"
##      Temperature_n Temperature_years collection_type
## [1,] "667193"      "22"      "continuous"

```

```

# Write to a unique new CSV file
write.csv(DEBY_temp, "/Users/nicolemongillo/Desktop/GitHub/MVP_Chesapeake_VIMS_hatchery/data/envr_raw_d

```

```

#Calculate the salinity variables
Mean_Annual_Salinity_ppt <- mean(DEBY_env$salinity)
Mean_max_Salinity_ppt <- mean(DEBY_envryear_sal$max_salinity)
Mean_min_Salinity_ppt <- mean(DEBY_envryear_sal$min_salinity)
Salinity_st_dev <- sd(DEBY_env$salinity)
Salinity_n <- nrow(DEBY_env)
Salinity_years <- nrow(DEBY_envryear_sal)

#Create a data frame to store the temperature results
DEBY_salinity <- cbind(site_name, download_date, source_description, lat, lon, firstyear, finalyear, Mea
print(DEBY_salinity)

```

```

##      site_name download_date
## [1,] "DEBY"      "04-02-2024"
##      source_description      lat
## [1,] "Virginia Estuarine and Coastal Observing System, VIMS" "37.247284"
##      lon      firstyear finalyear Mean_Annual_Salinity_ppt
## [1,] "-76.499369" "2003"      "2024"      "19.3391307162995"
##      Mean_max_Salinity_ppt Mean_min_Salinity_ppt Salinity_st_dev      Salinity_n
## [1,] "24.1890909090909"      "12.9063636363636"      "2.56014876611224" "667193"
##      Salinity_years collection_type
## [1,] "22"      "continuous"

```

```
# Write to the combined file with all sites  
# Write to a unique new CSV file  
write.csv(DEBY_salinity, "/Users/nicolemongillo/Desktop/GitHub/MVP_Chesapeake_VIMS_hatchery/data/envr_r
```