# Assignment 6: Time Series Analysis

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### **OVERVIEW**

This exercise accompanies the lessons in Hydrologic Data Analysis on time series analysis

#### **Directions**

- 1. Change "Student Name" on line 3 (above) with your name.
- 2. Work through the steps, **creating code and output** that fulfill each instruction.
- 3. Be sure to **answer the questions** in this assignment document.
- 4. When you have completed the assignment, **Knit** the text and code into a single pdf file.
- 5. After Knitting, submit the completed exercise (pdf file) to the dropbox in Sakai. Add your last name into the file name (e.g., "A06\_Salk.html") prior to submission.

The completed exercise is due on 11 October 2019 at 9:00 am.

## Setup

- 1. Verify your working directory is set to the R project file,
- 2. Load the tidyverse, lubridate, trend, and dataRetrieval packages.
- 3. Set your ggplot theme (can be theme\_classic or something else)
- 4. Load the ClearCreekDischarge.Monthly.csv file from the processed data folder. Call this data frame ClearCreekDischarge.Monthly.

```
getwd()
```

## [1] "/Users/yixinwen/Box/Duke/2019 Fall/Hydrologic Data Analysis/Hydrologic\_Data\_Analysis/Assignment

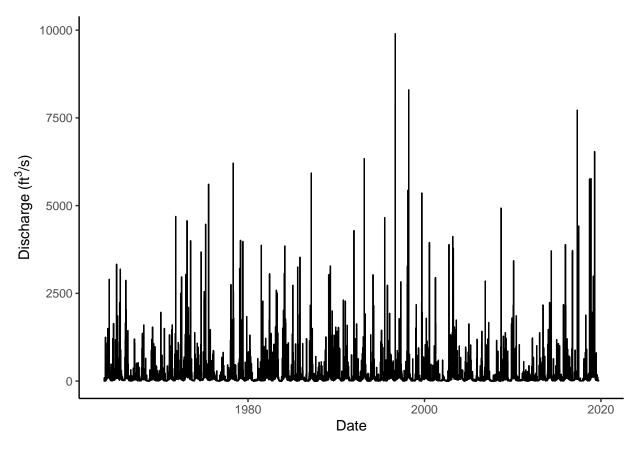
```
library(tidyverse)
## -- Attaching packages -----
## v ggplot2 3.2.1
                   v purrr
                             0.3.2
## v tibble 2.1.3
                    v dplyr
                             0.8.3
          0.8.3
## v tidyr
                    v stringr 1.4.0
## v readr
           1.3.1
                  v forcats 0.4.0
## -- Conflicts ----- tidyverse conflicts() --
## x dplyr::filter() masks stats::filter()
## x dplyr::lag()
                  masks stats::lag()
library(lubridate)
##
## Attaching package: 'lubridate'
## The following object is masked from 'package:base':
##
##
      date
library(trend)
library(dataRetrieval)
library(forecast)
```

```
## Registered S3 method overwritten by 'xts':
##
     method
                from
##
     as.zoo.xts zoo
## Registered S3 method overwritten by 'quantmod':
##
     method
                       from
     as.zoo.data.frame zoo
##
## Registered S3 methods overwritten by 'forecast':
##
     method
                        from
##
     fitted.fracdiff
                        fracdiff
     residuals.fracdiff fracdiff
library(tseries)
theme_set(theme_classic())
library(readr)
ClearCreekDischarge.Monthly <- read_csv("/Users/yixinwen/Box/Duke/2019 Fall/Hydrologic Data Analysis/Hy
## Parsed with column specification:
## cols(
##
     Year = col_double(),
##
    Month = col_double(),
##
    Discharge = col_double()
```

## Time Series Decomposition

## )

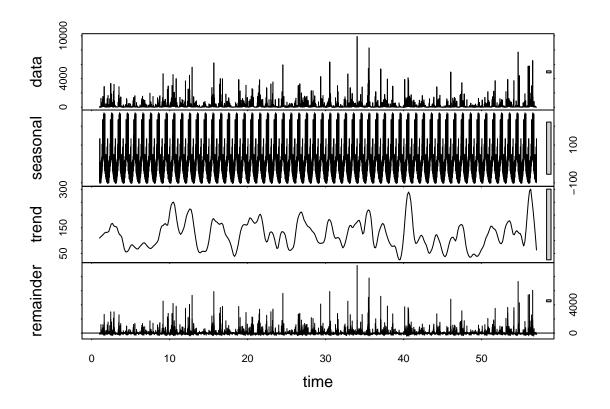
- 5. Create a new data frame that includes daily mean discharge at the Eno River for all available dates (siteNumbers = "02085070"). Rename the columns accordingly.
- 6. Plot discharge over time with geom\_line. Make sure axis labels are formatted appropriately.
- 7. Create a time series of discharge
- 8. Decompose the time series using the stl function.
- 9. Visualize the decomposed time series.



```
#create a time series of discharge
EnoRiver_ts <- ts(EnoRiverDischarge[[4]], frequency = 365)

# decompose the time series
EnoRiver_Decomposed <- stl(EnoRiver_ts, s.window = "periodic")

# visualize the decompozation
plot(EnoRiver_Decomposed)</pre>
```



10. How do the seasonal and trend components of the decomposition compare to the Clear Creek discharge dataset? Are they similar in magnitude?

Seasonal: the range of seasonal change of Eno River discharge is smaller than that of Clear Creek discharge. The Eno River has a more compased seasonal change compared to Clear Creek.

Trend: The range of trend component of Eno River discharge is similar to that of Clear Creek discharge.

## Trend Analysis

Research question: Has there been a monotonic trend in discharge in Clear Creek over the period of study?

- 11. Generate a time series of monthly discharge in Clear Creek from the ClearCreekDischarge.Monthly data frame. This time series should include just one column (discharge).
- 12. Run a Seasonal Mann-Kendall test on the monthly discharge data. Inspect the overall trend and the monthly trends.

```
ClearCreek_ts <- ts(ClearCreekDischarge.Monthly[[3]], frequency = 12)

ClearCreekTrend <- smk.test(ClearCreek_ts)

ClearCreekTrend

##
## Seasonal Mann-Kendall trend test (Hirsch-Slack test)
##
## data: ClearCreek_ts
## z = 1.6586, p-value = 0.09719</pre>
```

```
## alternative hypothesis: true S is not equal to 0
  sample estimates:
            varS
##
        S
      590 126102
##
summary(ClearCreekTrend)
##
    Seasonal Mann-Kendall trend test (Hirsch-Slack test)
##
##
## data: ClearCreek ts
## alternative hypothesis: two.sided
## Statistics for individual seasons
##
## HO
##
                           varS
                                    tau
                                             z Pr(>|z|)
## Season 1:
               S = 0
                       64 11154
                                  0.062
                                         0.597 0.550828
## Season 2:
               S = 0
                       24 10450
                                  0.024
                                         0.225 0.821984
## Season 3:
               S = 0
                                         0.284 0.776650
                        30 10450
                                  0.030
## Season 4:
               S = 0
                       35 10449
                                  0.035
                                         0.333 0.739425
## Season 5:
               S = 0
                        4 10450
                                  0.004
                                         0.029 0.976588
## Season 6:
               S = 0
                      204 10450
                                  0.206
                                         1.986 0.047054 *
               S = 0
                      230 10450
                                         2.240 0.025081
## Season 7:
                                  0.232
## Season 8:
               S = 0
                      148 10450
                                  0.149
                                         1.438 0.150434
## Season 9:
               S = 0
                       94 10450
                                  0.095
                                         0.910 0.362951
## Season 10:
                S = 0 -54 \ 10450 -0.055 -0.518 \ 0.604135
## Season 11:
                S = 0 -99 \ 10449 -0.100 -0.959 \ 0.337703
## Season 12:
                S = 0 -90 \ 10450 -0.091 -0.871 \ 0.383958
## ---
## Signif. codes: 0 '***' 0.001 '**' 0.05 '.' 0.1 ' ' 1
```

- 13. Is there an overall monotonic trend in discharge over time? If so, is it positive or negative? since the p-value = 0.09719, which is larger than 0.05, there's no monotonic trend in discharge over time.
- 14. Are there any monthly monotonic trends in discharge over time? If so, during which months do they occur and are they positive or negative?

June and July, since the z scores are larger than 0, they are positive.

#### Reflection

- 15. What are 2-3 conclusions or summary points about time series you learned through your analysis?
  - 1. when p-value is larger than 0.05, there's no significant trend on change; when p-value is smaller than 0.05, there is monotonic trend.
  - 2. when z score is greater than 0, the trend is positive, and when z score is smaller than 0, the trend is negative.
- 16. What data, visualizations, and/or models supported your conclusions from 12? the Seasonal Mann-Kendall test on ClearCreek can give the results of whether there is monotonic trend on discharge.
- 17. Did hands-on data analysis impact your learning about time series relative to a theory-based lesson? If so, how?

Hands-on data analysis can let me explore the theory on my own, and it can help me understand it better from examples.

18. How did the real-world data compare with your expectations from theory?

The real-world data may have mistakes in the dataset. Before we do analysis on it, we need to remove the abnormal data, otherwise the results may not be correct.