# ENV 790.30 - Time Series Analysis for Energy Data | Spring 2021 Assignment 2 - Due date 02/05/21

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#### Submission Instructions

You should open the .rmd file corresponding to this assignment on RStudio. The file is available on our class repository on Github.

Once you have the file open on your local machine the first thing you will do is change "Student Name" on line 4 with your name. Then you will start working through the assignment by **creating code and output** that answer each question. Be sure to use this assignment document. Your report should contain the answer to each question and any plots/tables you obtained (when applicable).

When you have completed the assignment, **Knit** the text and code into a single PDF file. Rename the pdf file such that it includes your first and last name (e.g., "LuanaLima\_TSA\_A02\_Sp21.Rmd"). Submit this pdf using Sakai.

### R packages

##

filter, lag

R packages needed for this assignment: "forecast", "tseries", and "dplyr". Install these packages, if you haven't done yet. Do not forget to load them before running your script, since they are NOT default packages.\

```
#Load/install required package here
getwd()
## [1] "/Users/ethel/Desktop/ENV 790 Time Series/ENV790_TSA_S2021/Assignments"
library(lubridate)
##
## Attaching package: 'lubridate'
## The following objects are masked from 'package:base':
##
##
       date, intersect, setdiff, union
library(ggplot2)
library(dplyr)
##
## Attaching package: 'dplyr'
## The following objects are masked from 'package:lubridate':
##
##
       intersect, setdiff, union
## The following objects are masked from 'package:stats':
##
```

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

#install.packages("forecast")
library(forecast)

## Registered S3 method overwritten by 'quantmod':
## method from
## as.zoo.data.frame zoo

#install.packages("tseries")
library(tseries)
```

### Data set information

Consider the data provided in the spreadsheet "Table\_10.1\_Renewable\_Energy\_Production\_and\_Consumption\_by\_Source.x on our **Data** folder. The data comes from the US Energy Information and Administration and corresponds to the January 2021 Monthly Energy Review. The spreadsheet is ready to be used. Use the command read.table() to import the data in R or  $panda.read\_excel()$  in Python (note that you will need to import pandas package). }

```
#Importing data set

#install.packages("readxl")
library("readxl")
MonthlyData <- read_excel("../Data/Table_10.1_Renewable_Energy_Production_and_Consumption_by_Source.xls.

#AnnualData <- read_excel("../Data/Table_10.1_Renewable_Energy_Production_and_Consumption_by_Source.xls.
#AnnualData</pre>
```

### Question 1

You will work only with the following columns: Total Biomass Energy Production, Total Renewable Energy Production, Hydroelectric Power Consumption. Create a data frame structure with these three time series only. Use the command head() to verify your data.

```
#specific rows and columns just trim the table
MonthlyData_subset<- MonthlyData[2:575, c(1, 4, 5, 6)]
#Checking data
head(MonthlyData_subset)
## # A tibble: 6 x 4
                          `Total Biomass Ene~ `Total Renewable E~ `Hydroelectric Po~
##
    Month
##
     <dttm>
                         <chr>>
                                              <chr>>
                                                                   <chr>>
## 1 1973-01-01 00:00:00 129.787
                                                                   272.703
                                              403.981
## 2 1973-02-01 00:00:00 117.338
                                              360.9
                                                                  242.199
## 3 1973-03-01 00:00:00 129.938
                                              400.161
                                                                   268.81
## 4 1973-04-01 00:00:00 125.636
                                              380.47
                                                                   253.185
## 5 1973-05-01 00:00:00 129.834
                                              392.141
                                                                  260.77
## 6 1973-06-01 00:00:00 125.611
                                              377.232
                                                                   249.859
str(MonthlyData_subset)
## tibble [574 x 4] (S3: tbl_df/tbl/data.frame)
## $ Month
                                        : POSIXct[1:574], format: "1973-01-01" "1973-02-01" ...
```

```
## $ Total Biomass Energy Production : chr [1:574] "129.787" "117.338" "129.938" "125.636" ... ## $ Total Renewable Energy Production: chr [1:574] "403.981" "360.9" "400.161" "380.47" ... ## $ Hydroelectric Power Consumption : chr [1:574] "272.703" "242.199" "268.81" "253.185" ...
```

### Question 2

Transform your data frame in a time series object and specify the starting point and frequency of the time series using the function ts().

### Question 3

## [1] 236.9515

Compute mean and standard deviation for these three series.

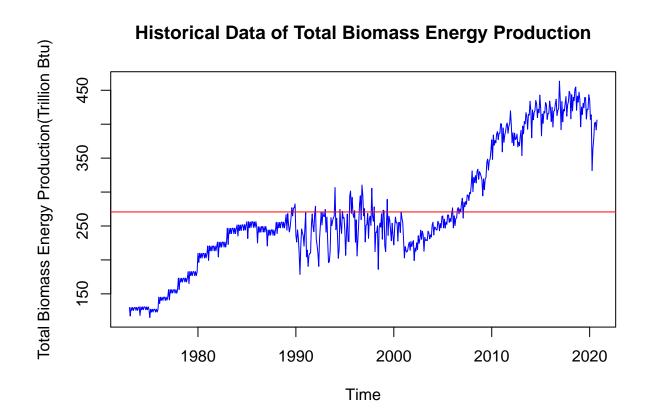
```
Biomass_mean <- mean(MonthlyData_subset_ts[,"Biomass"])</pre>
Biomass_sd <- sd(MonthlyData_subset_ts[,"Biomass"])</pre>
Biomass_mean
## [1] 270.6961
str(Biomass_mean)
## num 271
Biomass_sd
## [1] 87.36311
Renewable_mean <- mean(MonthlyData_subset_ts[,"Renewable"])</pre>
Renewable_sd <- sd(MonthlyData_subset_ts[,"Renewable"])</pre>
Renewable_mean
## [1] 572.7321
Renewable_sd
## [1] 168.4588
Hydroelectric_mean <- mean(MonthlyData_subset_ts[,"Hydroelectric"])</pre>
Hydroelectric_sd <- sd(MonthlyData_subset_ts[,"Hydroelectric"])</pre>
Hydroelectric_mean
```

```
Hydroelectric_sd
```

## [1] 43.90392

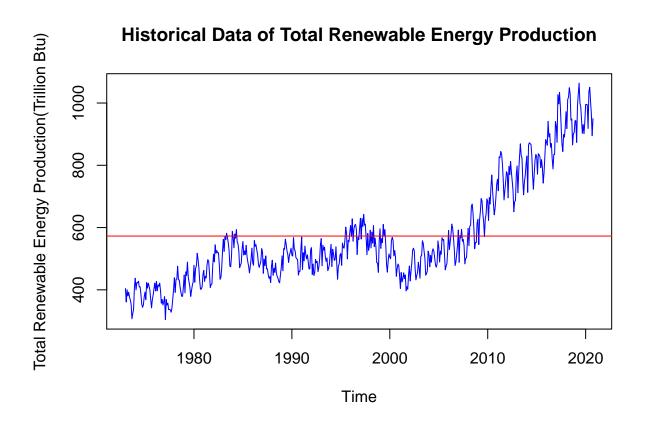
### Question 4

Display and interpret the time series plot for each of these variables. Try to make your plot as informative as possible by writing titles, labels, etc. For each plot add a horizontal line at the mean of each series in a different color.

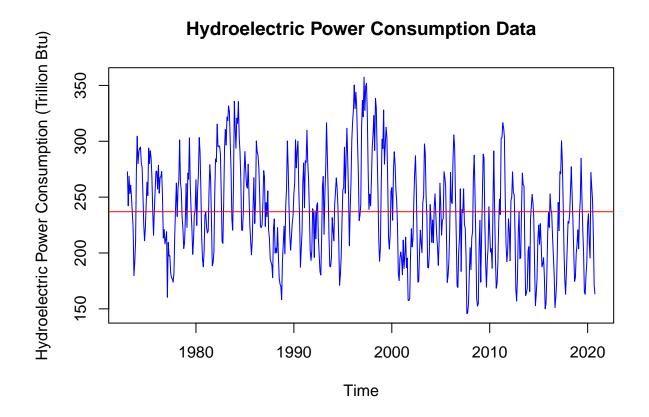


>This time series plot shows a drastic shift in 1975 and 2000 and shows a clear upward trend. Between 1990 and 2000, it shows random variation.

```
plot(MonthlyData_subset_ts[,"Renewable"],col="blue",
        ylab = "Total Renewable Energy Production(Trillion Btu)",
        main = "Historical Data of Total Renewable Energy Production"
    )
#add a line with the mean
abline(h=Renewable_mean,col="red")
```



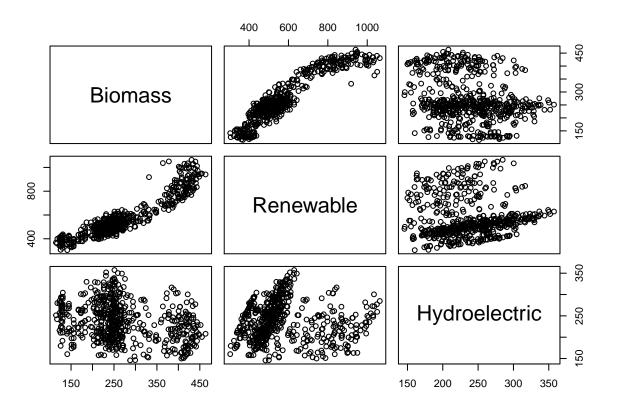
>This plot shows increase in the data values seems to accelerate after 2000. Before the 2000, the cycles do not repeat at regular intervals and do not have the same shape.



>These plot displays the pattern of Hydroelectric Power Consumption increasing and decreasing from 1980 to 2020, which seems changing seasonal.

## Question 5

Compute the correlation between these three series. Are they significantly correlated? Explain your answer.  $pairs(MonthlyData_subset[, c(2,3,4)])$ 



#### cor(MonthlyData\_subset\_ts)

```
## Biomass Renewable Hydroelectric

## Biomass 1.0000000 0.923460855 -0.255567465

## Renewable 0.9234609 1.000000000 -0.002756852

## Hydroelectric -0.2555675 -0.002756852 1.000000000
```

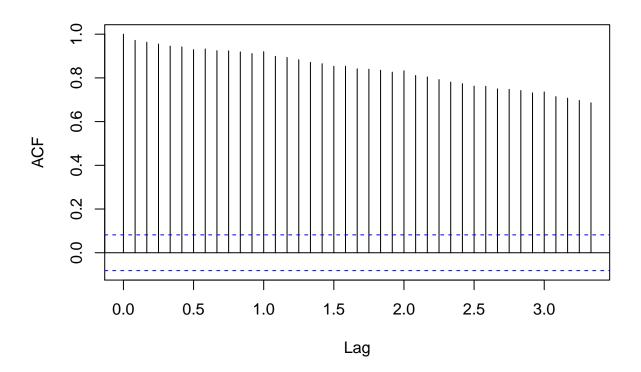
Due to the results of correlation and plots, Total Biomass Energy Production and Total Renewable Energy Production are significantly correlated, whereas the other two are not.

## Question 6

Compute the autocorrelation function from lag 1 up to lag 40 for these three variables. What can you say about these plots? Do the three of them have the same behavior?

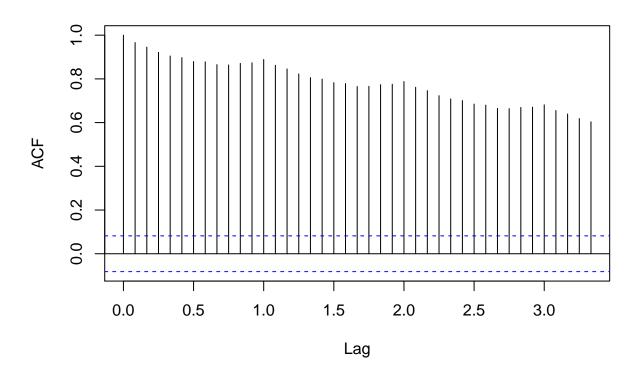
Biomass\_acf = acf(MonthlyData\_subset\_ts[,1], lag.max=40, type = "correlation", plot=TRUE, main="Biomass

# **Biomass ACF**



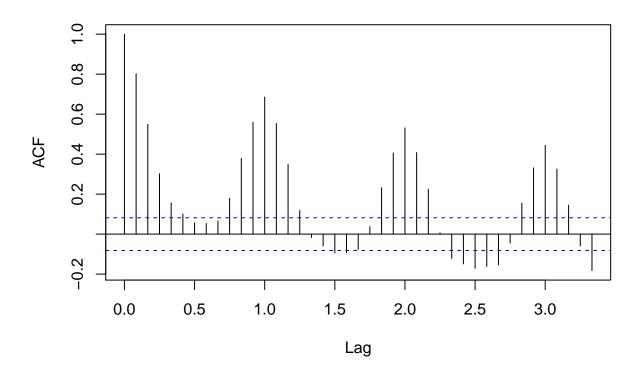
Renewable\_acf = acf(MonthlyData\_subset\_ts[,2], lag.max=40, type = "correlation", plot=TRUE, main="Renewable\_acf"

# Renewable ACF



Hydroelectric\_acf = acf(MonthlyData\_subset\_ts[,3], lag.max=40, type = "correlation", plot=TRUE, main="Hydroelectric\_acf"

# **Hydroelectric ACF**



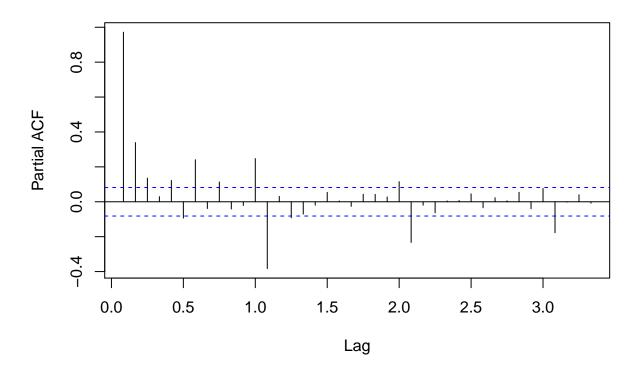
> The Biomass and Renewable plots have very similar trend, but Hydroelectric one is kind of a periodic trend. They do not have same behavior.

## Question 7

Compute the partial autocorrelation function from lag 1 to lag 40 for these three variables. How these plots differ from the ones in Q6?

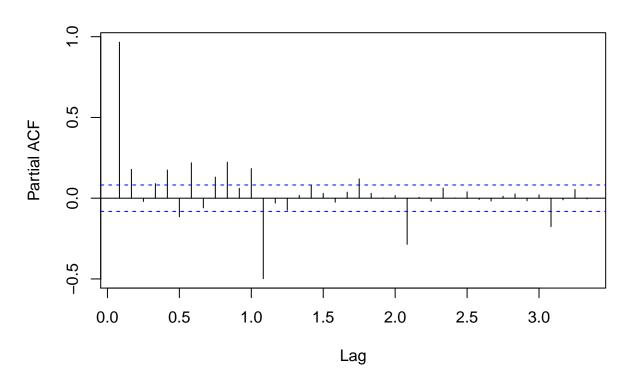
Biomass\_pacf = pacf(MonthlyData\_subset\_ts[,1], lag.max=40,plot=TRUE, main="Biomass PACF")

# **Biomass PACF**



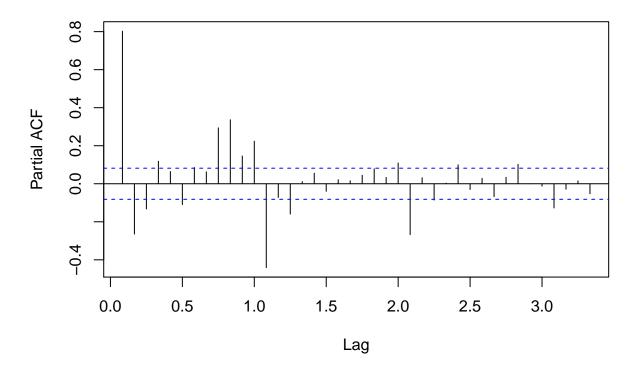
Renewable\_pacf = pacf(MonthlyData\_subset\_ts[,2], lag.max=40, plot=TRUE, main="Renewable PACF")

# Renewable PACF



Hydroelectric\_pacf = pacf(MonthlyData\_subset\_ts[,3], lag.max=40, plot=TRUE, main="Hydroelectric PACF")

# **Hydroelectric PACF**



> The three plots in Q7 is much more similar, but Hydroelectric plot is total different than Biomass and Renewable plots in Q6.