

# ENV 790.30 - Time Series Analysis for Energy Data | Spring 2026

Assignment 2 - Due date 01/27/26

Leo Zhang

## 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 rename the file such that it includes your first and last name (e.g., 'LuanaLima\_TSA\_A02\_Sp24.Rmd'). Then 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. Submit this pdf using Sakai.

## R packages

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.\

```
library(forecast)
library(gt)
library(readxl)
library(tidyverse)
library(tseries)
```

## Data set information

Consider the data provided in the spreadsheet 'Table\_10.1\_Renewable\_Energy\_Production\_and\_Consumption\_by\_Source' on our **Data** folder. The data comes from the US Energy Information and Administration and corresponds to the December 2023 Monthly Energy Review. The spreadsheet is ready to be used. You will also find a .csv version of the data install.packages(forecast) install.packages(tseries). You may use the function read.table() to import the .csv data in R. Or refer to the file 'M2\_ImportingData\_CSV\_XLSX.Rmd' in our Lessons folder for functions that are better suited for importing the .xlsx.

```
outlook <- read_excel('C:/Users/24169/Desktop/e797/797_spring26/Data/Table_10.1_Renewable_Energy_Produc
```

## 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.

```
outlook <- outlook[-c(1), ]
outlook <- outlook %>% select('Total Biomass Energy Production', 'Total Renewable Energy Production', 'Hydroelectric Power Consumption')
outlook <- outlook %>% mutate(across(c('Total Biomass Energy Production', 'Total Renewable Energy Production', 'Hydroelectric Power Consumption'), ~.x))
head(outlook)
```

```
## # A tibble: 6 x 3
##   Total Biomass Energy Production~1 Total Renewable Energy Production~2 Hydroelectric Power Consumption~3
##   <dbl> <dbl> <dbl>
## 1 1549. 1873. 323.
## 2 1562. 1907. 344.
## 3 1535. 1891. 356.
## 4 1474. 1849. 374.
## 5 1419. 1793. 374.
## 6 1394. 1775. 381.
## # i abbreviated names: 1: 'Total Biomass Energy Production',
## # 2: 'Total Renewable Energy Production',
## # 3: 'Hydroelectric Power Consumption'
```

## 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()`.

```
history <- ts(outlook, start = 1949, end = 2023, frequency = 1)
history
```

```
## Time Series:
## Start = 1949
## End = 2023
## Frequency = 1
##   Total Biomass Energy Production Total Renewable Energy Production
## 1949 1549.262 1872.627
## 1950 1562.307 1906.525
## 1951 1534.669 1890.800
## 1952 1474.369 1848.694
## 1953 1418.601 1792.616
## 1954 1394.327 1775.242
## 1955 1424.143 1820.740
## 1956 1415.871 1843.178
## 1957 1333.581 1788.598
## 1958 1323.123 1813.136
## 1959 1352.874 1834.493
## 1960 1319.870 1829.873
## 1961 1294.762 1825.773
## 1962 1300.242 1887.502
## 1963 1323.316 1900.483
```

## 1964	1336.802	1952.686
## 1965	1334.761	2007.517
## 1966	1368.985	2044.989
## 1967	1340.249	2108.853
## 1968	1419.495	2191.661
## 1969	1440.487	2307.418
## 1970	1430.962	2289.021
## 1971	1432.323	2353.833
## 1972	1503.065	2449.491
## 1973	1529.068	2475.544
## 1974	1539.662	2586.001
## 1975	1498.734	2544.167
## 1976	1713.373	2704.698
## 1977	1838.332	2613.474
## 1978	2037.605	3014.948
## 1979	2151.906	3131.031
## 1980	2475.500	3445.378
## 1981	2596.283	3515.922
## 1982	2663.452	3745.796
## 1983	2904.414	4069.164
## 1984	2971.120	4104.121
## 1985	3016.233	4018.174
## 1986	2932.095	3970.473
## 1987	2874.884	3774.443
## 1988	3016.049	3822.682
## 1989	3159.357	4206.136
## 1990	2735.112	3862.708
## 1991	2781.797	3901.431
## 1992	2931.678	3931.748
## 1993	2908.446	4007.312
## 1994	3027.534	4056.461
## 1995	3099.082	4294.583
## 1996	3155.301	4479.994
## 1997	3107.908	4466.063
## 1998	2928.930	4174.139
## 1999	2965.132	4202.601
## 2000	3005.661	4093.033
## 2001	2624.162	3513.751
## 2002	2705.406	3770.941
## 2003	2804.778	3914.186
## 2004	2996.017	4093.471
## 2005	3101.186	4220.160
## 2006	3211.515	4429.647
## 2007	3472.080	4582.116
## 2008	3868.250	5084.563
## 2009	3956.617	5309.146
## 2010	4552.532	5942.580
## 2011	4712.436	6404.003
## 2012	4553.559	6186.970
## 2013	4834.795	6560.783
## 2014	5049.215	6832.704
## 2015	5024.958	6839.536
## 2016	5121.688	7178.348
## 2017	5155.644	7494.525

## 2018	5304.078	7734.125
## 2019	5204.933	7742.749
## 2020	4699.703	7455.133
## 2021	4903.546	7797.231
## 2022	5063.341	8297.453
## 2023	5164.910	8437.463
##	Hydroelectric Power Consumption	
## 1949	323.365	
## 1950	344.218	
## 1951	356.131	
## 1952	374.325	
## 1953	374.015	
## 1954	380.915	
## 1955	396.597	
## 1956	427.307	
## 1957	455.017	
## 1958	490.013	
## 1959	481.619	
## 1960	509.889	
## 1961	530.690	
## 1962	586.917	
## 1963	576.594	
## 1964	615.189	
## 1965	672.111	
## 1966	675.363	
## 1967	767.525	
## 1968	770.679	
## 1969	864.834	
## 1970	856.267	
## 1971	919.641	
## 1972	941.469	
## 1973	939.769	
## 1974	1037.971	
## 1975	1034.357	
## 1976	978.986	
## 1977	762.919	
## 1978	967.183	
## 1979	965.855	
## 1980	952.569	
## 1981	900.238	
## 1982	1065.820	
## 1983	1144.012	
## 1984	1106.550	
## 1985	970.068	
## 1986	1003.146	
## 1987	862.745	
## 1988	771.456	
## 1989	927.985	
## 1990	999.258	
## 1991	986.048	
## 1992	863.536	
## 1993	957.046	
## 1994	887.549	
## 1995	1060.561	

```
## 1996      1184.517
## 1997      1216.219
## 1998      1103.221
## 1999      1090.257
## 2000       940.254
## 2001       740.271
## 2002       901.890
## 2003       941.051
## 2004       915.840
## 2005       922.336
## 2006       986.909
## 2007       844.504
## 2008       869.485
## 2009       932.995
## 2010       887.813
## 2011      1089.639
## 2012       942.532
## 2013       916.345
## 2014       884.959
## 2015       849.861
## 2016       913.775
## 2017      1024.736
## 2018       998.092
## 2019       982.225
## 2020       973.355
## 2021       858.407
## 2022       869.339
## 2023       835.948
```

### Question 3

Compute mean and standard deviation for these three series.

```
statement <- c('Mean', 'Standard Deviation')
biomass <- history[, 'Total Biomass Energy Production']
first_statistic <- c(mean(biomass), sd(biomass))
renewable <- history[, 'Total Renewable Energy Production']
second_statistic <- c(mean(renewable), sd(renewable))
hydro <- history[, 'Hydroelectric Power Consumption']
third_statistic <- c(mean(hydro), sd(hydro))
general_statistic <- data.frame(statement, first_statistic, second_statistic, third_statistic)
p1 <- general_statistic %>% gt() %>% tab_options(table.width = pct(100)) %>% tab_header(title = 'Figure')
p1
```

### Question 4

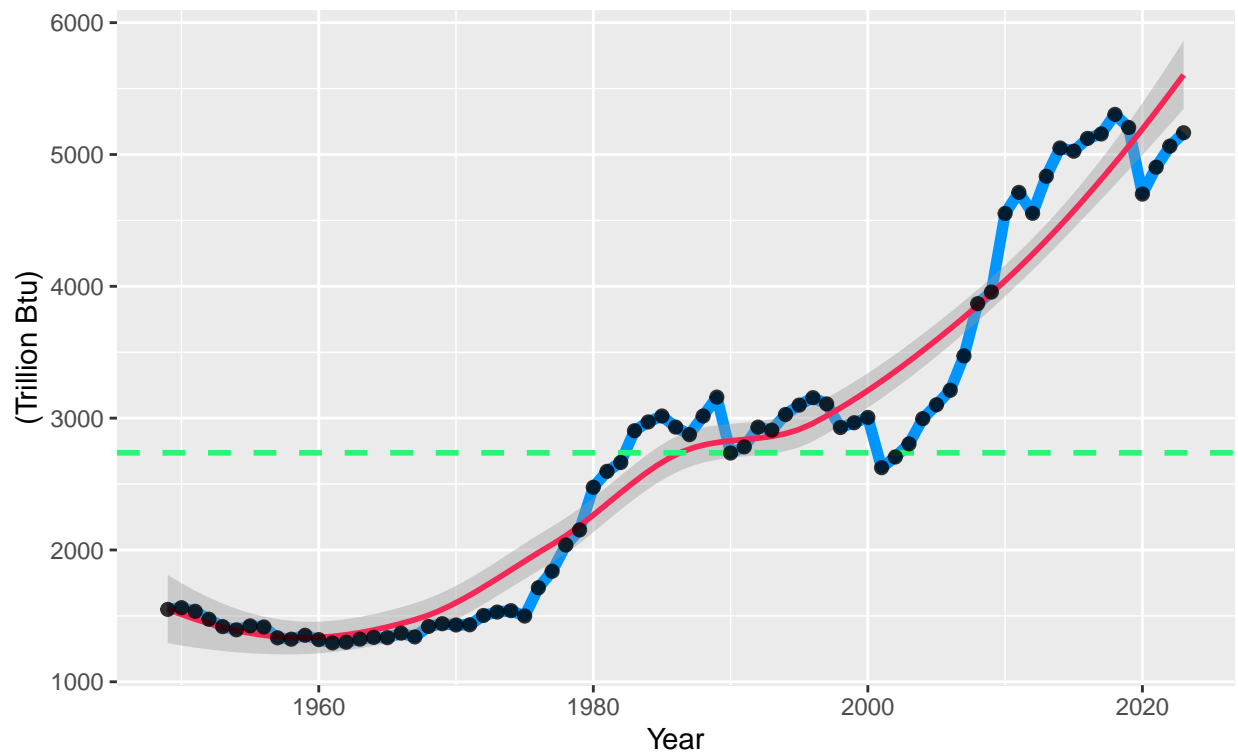
Display and interpret the time series plot for each of these variables. Try to make your plot as informative as possible by writing titles, labels, etc. For each plot add a horizontal line at the mean of each series in a different color.

Figure 1: Summary Statistics  
time series from 1949 to 2023

	Total Biomass Energy Production	Total Renewable Energy Production	Hydroele
Mean	2737.046	3871.594	
Standard Deviation	1298.192	1899.766	

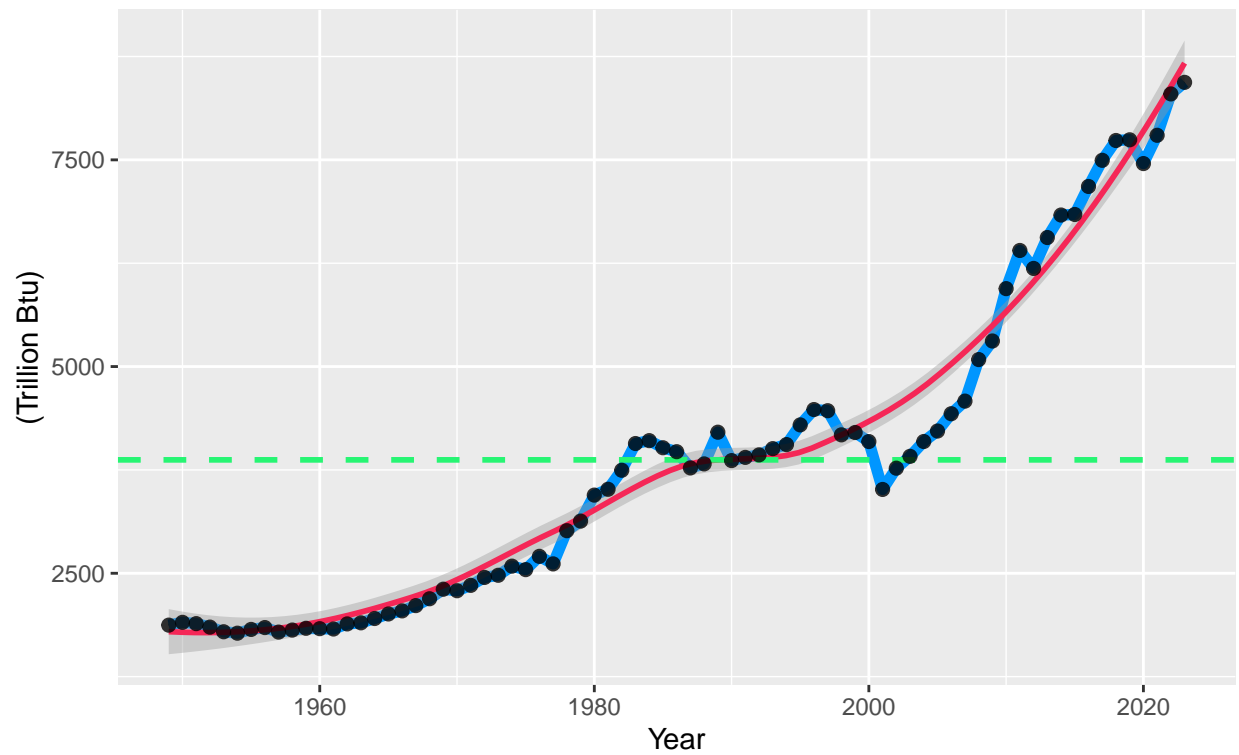
```
p2 <- autoplot(biomass) + geom_line(color = '#0096FF', linewidth = 2) + geom_smooth(color = '#F52757', linewidth = 2)
p2
```

Figure 2: Time Series For Total Biomass Energy Production  
from 1949 to 2023



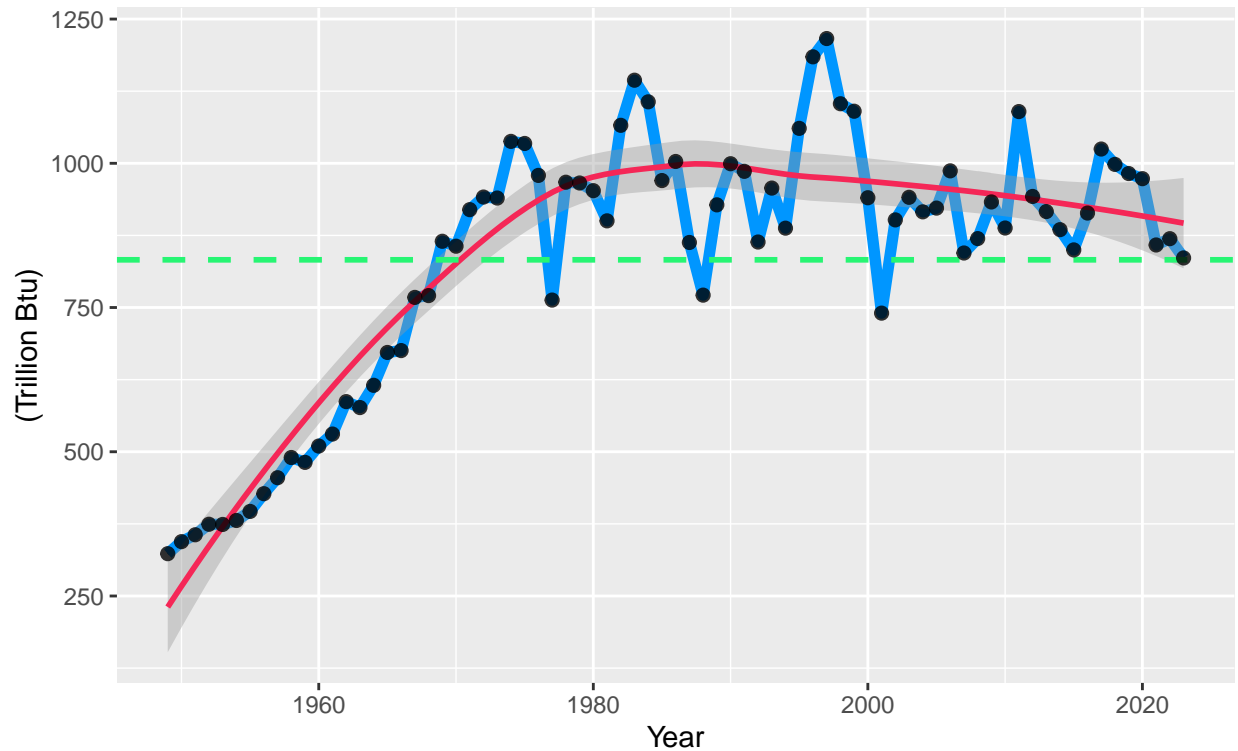
```
p3 <- autoplot(renewable) + geom_line(color = '#0096FF', linewidth = 2) + geom_smooth(color = '#F52757', linewidth = 2)
p3
```

Figure 3: Time Series For Total Renewable Energy Production  
from 1949 to 2023



```
p4 <- autoplot(hydro) + geom_line(color = '#0096FF', linewidth = 2) + geom_smooth(color = '#F52757', lin
p4
```

Figure 4: Time Series For Hydroelectric Power Consumption  
from 1949 to 2023



For 75 years, the total biomass energy production has been increasing. Particularly, between 1970 and 1980, the total biomass energy production increased rapidly. The growth was more considerable after 2000. For 75 years, the total renewable energy production has also been increasing. The growth was also more considerable after 2000. Before 1980, the hydroelectric power consumption seemed to increase steadily. However, after 1980, with apparent fluctuations, the hydroelectric power consumption had a declining tendency.

## Question 5

Compute the correlation between these three series. Are they significantly correlated? Explain your answer.

```
correlate <- cor(history)
correlate
```

```
##                                Total Biomass Energy Production
## Total Biomass Energy Production      1.0000000
## Total Renewable Energy Production    0.9861982
## Hydroelectric Power Consumption      0.5276925
##                                Total Renewable Energy Production
## Total Biomass Energy Production      0.9861982
## Total Renewable Energy Production    1.0000000
## Hydroelectric Power Consumption      0.5457158
##                                Hydroelectric Power Consumption
## Total Biomass Energy Production      0.5276925
```



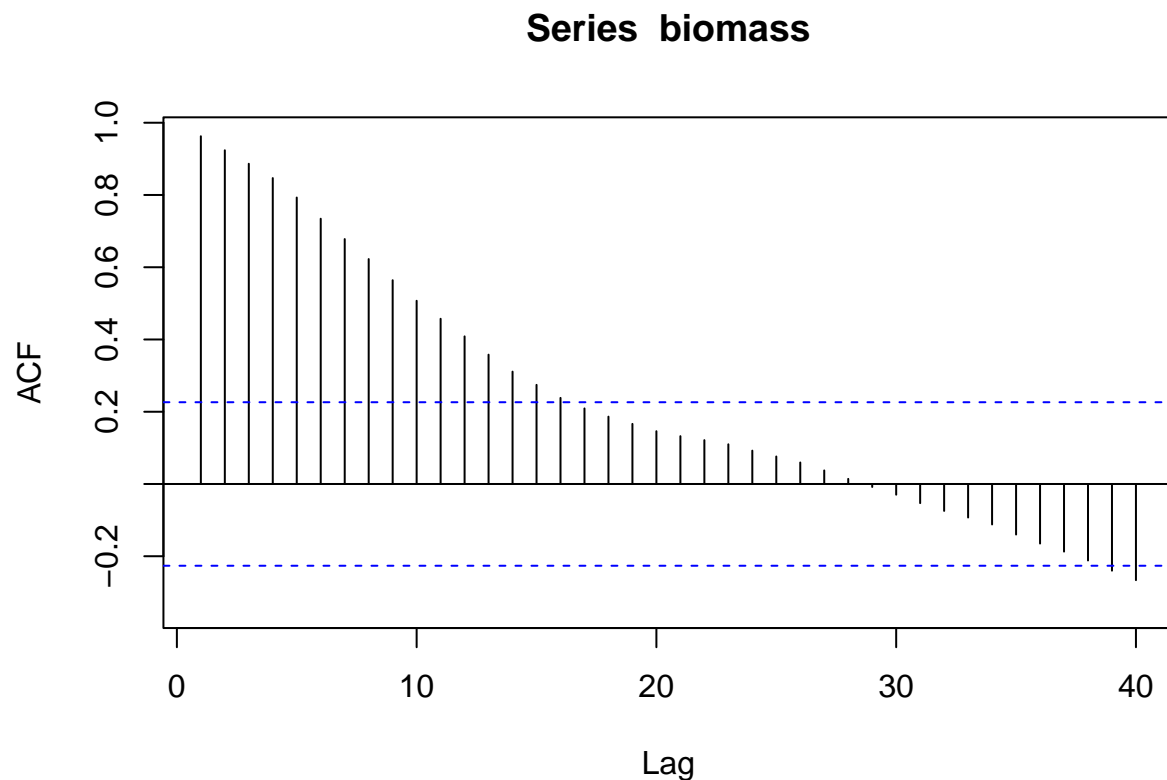
```
## Total Renewable Energy Production      0.5457158
## Hydroelectric Power Consumption         1.0000000
```

The correlation coefficient between total biomass energy production and total renewable energy production is 0.9861982, which may indicate a very strong positive correlation. There is a moderate positive correlation between hydroelectric power consumption and the rest.

## 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?

```
first_acf <- Acf(biomass, lag.max = 40, type = 'correlation', plot = TRUE)
```



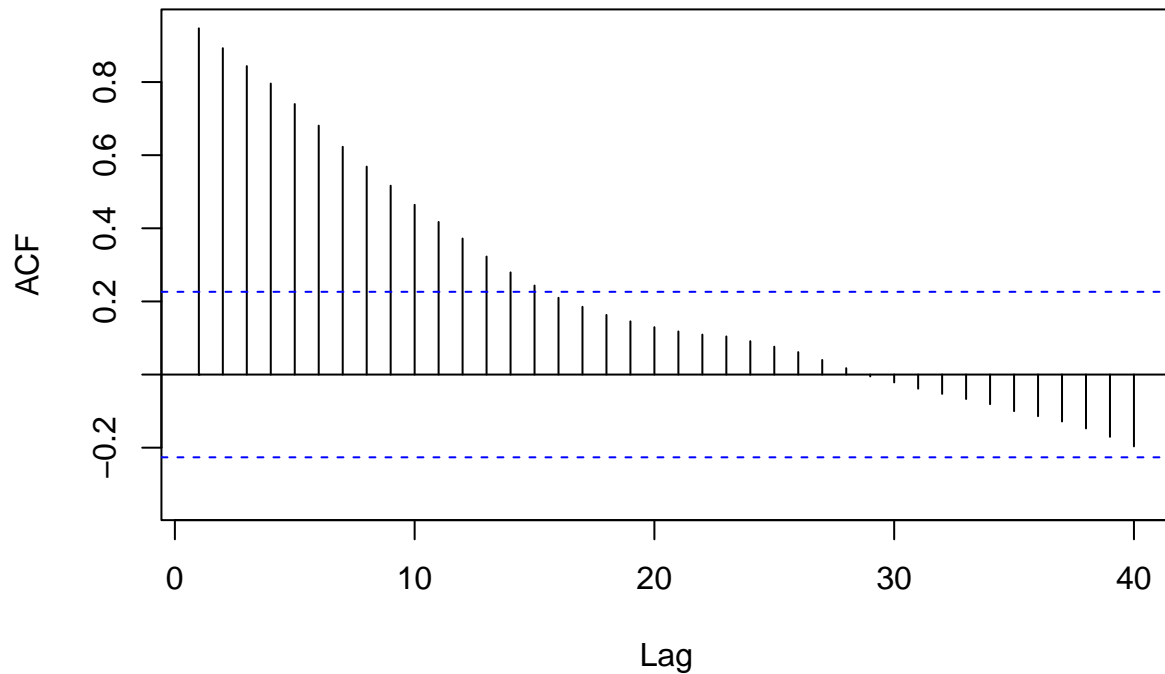
```
first_acf
```

```
##
## Autocorrelations of series 'biomass', by lag
##
##      0      1      2      3      4      5      6      7      8      9     10
## 1.000 0.963 0.924 0.887 0.847 0.793 0.734 0.678 0.623 0.564 0.507
##    11    12    13    14    15    16    17    18    19    20    21
## 0.457 0.409 0.358 0.311 0.274 0.238 0.209 0.186 0.166 0.146 0.132
##    22    23    24    25    26    27    28    29    30    31    32
```

```
## 0.121 0.110 0.092 0.076 0.060 0.038 0.014 -0.008 -0.030 -0.053 -0.075
##      33      34      35      36      37      38      39      40
## -0.093 -0.112 -0.140 -0.165 -0.187 -0.212 -0.240 -0.266
```

```
second_acf <- Acf(renewable, lag.max = 40, type = 'correlation', plot = TRUE)
```

## Series renewable

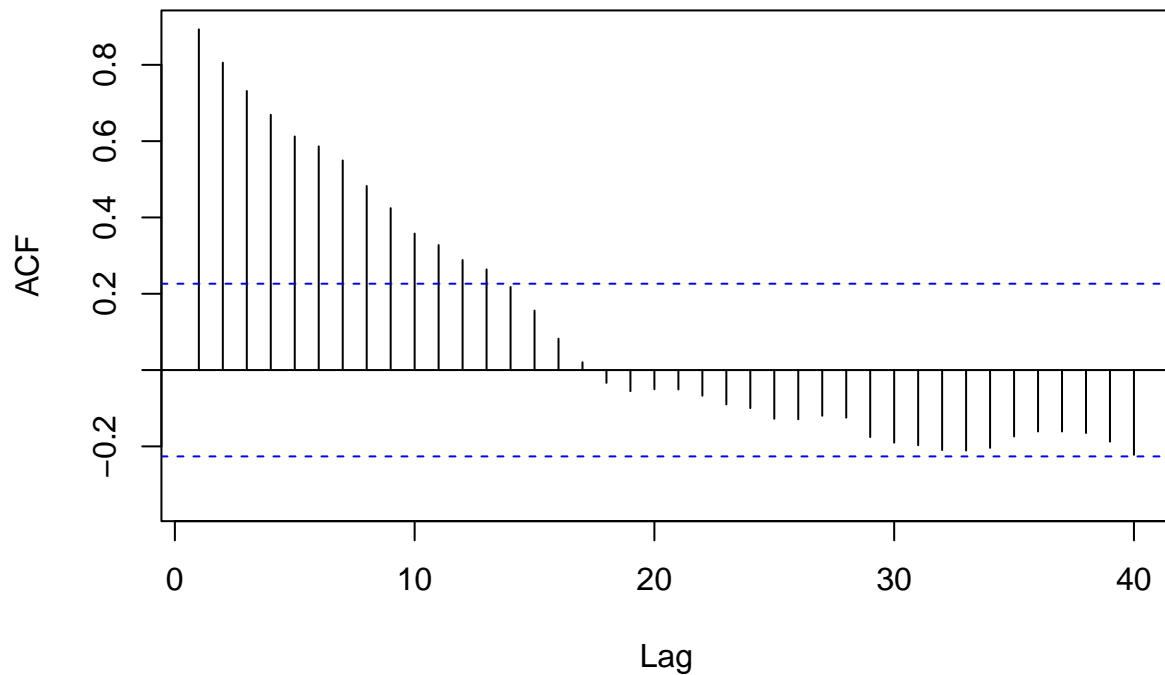


```
second_acf
```

```
##
## Autocorrelations of series 'renewable', by lag
##
##      0      1      2      3      4      5      6      7      8      9     10
## 1.000 0.947 0.893 0.844 0.796 0.740 0.681 0.623 0.569 0.517 0.464
##     11     12     13     14     15     16     17     18     19     20     21
## 0.417 0.372 0.323 0.279 0.243 0.210 0.185 0.163 0.146 0.130 0.118
##     22     23     24     25     26     27     28     29     30     31     32
## 0.109 0.104 0.091 0.076 0.061 0.040 0.017 -0.005 -0.022 -0.039 -0.053
##     33     34     35     36     37     38     39     40
## -0.067 -0.081 -0.100 -0.114 -0.128 -0.147 -0.170 -0.196
```

```
third_acf <- Acf(hydro, lag.max = 40, type = 'correlation', plot = TRUE)
```

## Series hydro



```
third_acf
```

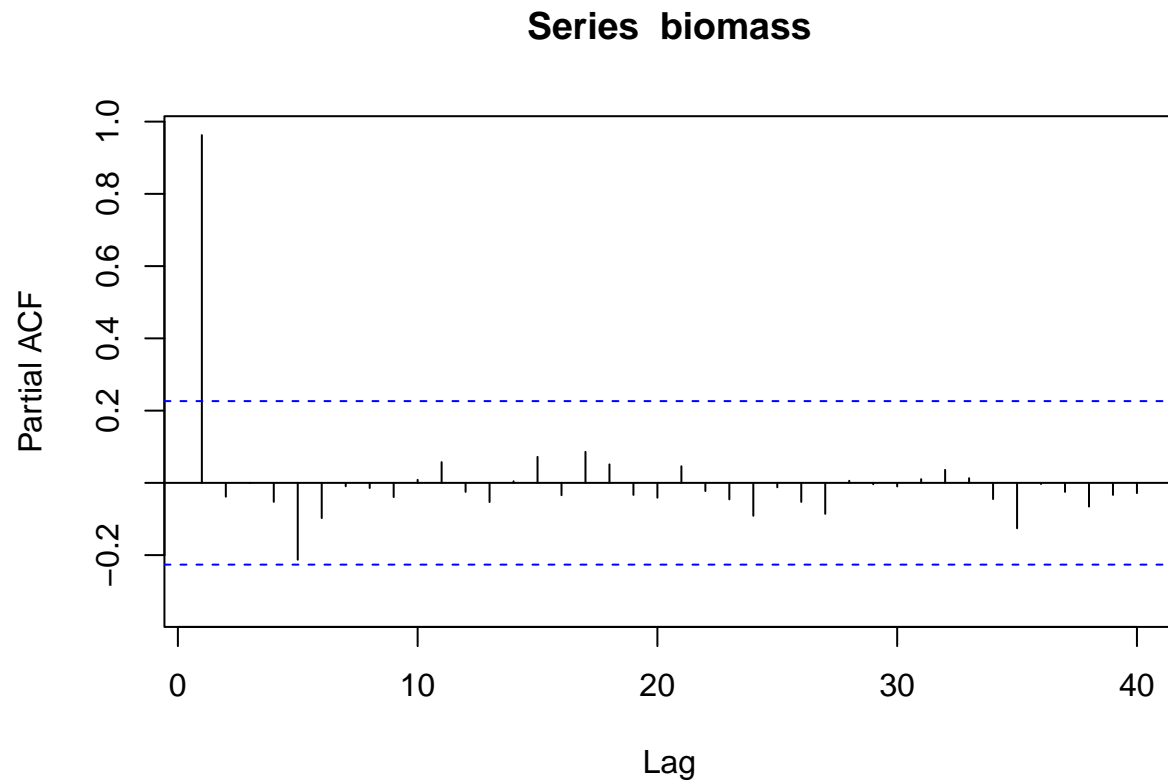
```
##
## Autocorrelations of series 'hydro', by lag
##
##      0      1      2      3      4      5      6      7      8      9     10
## 1.000 0.893 0.806 0.731 0.669 0.612 0.586 0.549 0.483 0.424 0.358
##    11    12    13    14    15    16    17    18    19    20    21
## 0.328 0.288 0.264 0.218 0.156 0.082 0.020 -0.033 -0.055 -0.050 -0.051
##    22    23    24    25    26    27    28    29    30    31    32
## -0.067 -0.090 -0.100 -0.128 -0.129 -0.120 -0.125 -0.176 -0.190 -0.197 -0.210
##    33    34    35    36    37    38    39    40
## -0.211 -0.204 -0.174 -0.161 -0.161 -0.165 -0.187 -0.222
```

The behavior of total biomass energy production is similar to that of total renewable energy production. For both total biomass energy production and total renewable energy production, the coefficient becomes negative with lag 29. Hydroelectric power consumption behave differently because its coefficient becomes negative with lag 18. This faster decline may indicate that the temporal dependence is weaker for hydroelectric power consumption.

## 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?

```
first_pacf <- Pacf(biomass, lag.max = 40, plot = TRUE)
```

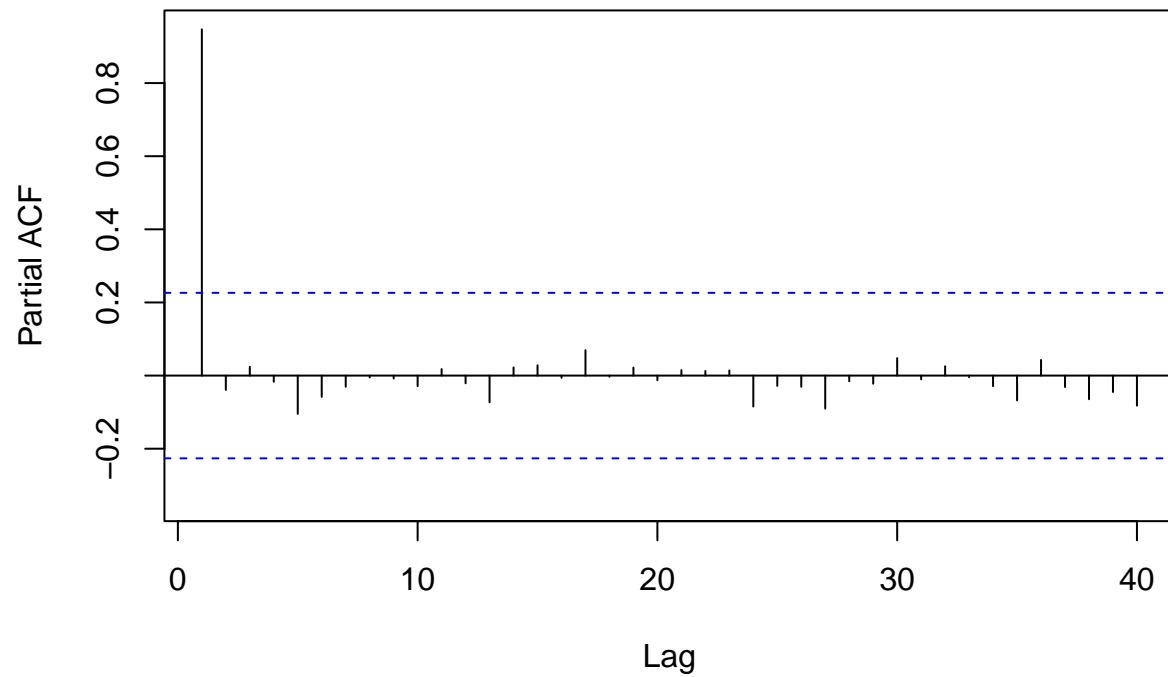


```
first_pacf
```

```
##
## Partial autocorrelations of series 'biomass', by lag
##
##      1      2      3      4      5      6      7      8      9     10     11
## 0.963 -0.038  0.000 -0.053 -0.213 -0.097 -0.009 -0.014 -0.039  0.008  0.057
##     12     13     14     15     16     17     18     19     20     21     22
## -0.025 -0.053  0.004  0.072 -0.034  0.086  0.051 -0.033 -0.041  0.046 -0.023
##     23     24     25     26     27     28     29     30     31     32     33
## -0.046 -0.091 -0.012 -0.053 -0.086  0.006 -0.004 -0.010  0.010  0.036  0.013
##     34     35     36     37     38     39     40
## -0.045 -0.126 -0.003 -0.025 -0.066 -0.033 -0.028
```

```
second_pacf <- Pacf(renewable, lag.max = 40, plot = TRUE)
```

## Series renewable

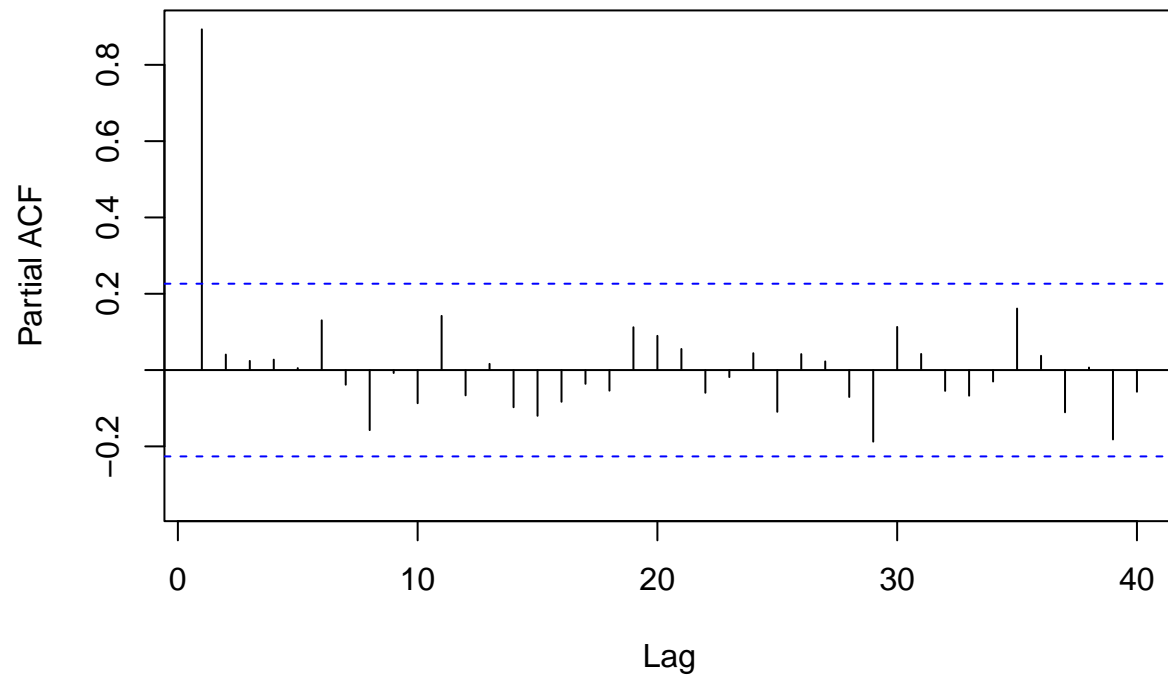


```
second_pacf
```

```
##
## Partial autocorrelations of series 'renewable', by lag
##
##      1      2      3      4      5      6      7      8      9     10     11
## 0.947 -0.039  0.024 -0.017 -0.105 -0.058 -0.031 -0.005 -0.008 -0.029  0.018
##     12     13     14     15     16     17     18     19     20     21     22
## -0.021 -0.073  0.022  0.028 -0.006  0.069 -0.003  0.022 -0.013  0.015  0.013
##     23     24     25     26     27     28     29     30     31     32     33
##  0.014 -0.085 -0.028 -0.031 -0.091 -0.016 -0.022  0.048 -0.010  0.025 -0.004
##     34     35     36     37     38     39     40
## -0.029 -0.068  0.043 -0.031 -0.065 -0.045 -0.082
```

```
third_pacf <- Pacf(hydro, lag.max = 40, plot = TRUE)
```

## Series hydro



```
third_pacf
```

```
##
## Partial autocorrelations of series 'hydro', by lag
##
##      1      2      3      4      5      6      7      8      9     10     11
## 0.893 0.041 0.024 0.027 0.005 0.130 -0.038 -0.157 -0.008 -0.087 0.142
##     12     13     14     15     16     17     18     19     20     21     22
## -0.066 0.016 -0.097 -0.120 -0.083 -0.036 -0.054 0.112 0.090 0.055 -0.059
##     23     24     25     26     27     28     29     30     31     32     33
## -0.018 0.044 -0.109 0.042 0.023 -0.070 -0.188 0.113 0.042 -0.055 -0.067
##     34     35     36     37     38     39     40
## -0.030 0.161 0.037 -0.111 0.007 -0.182 -0.057
```

Coefficients no longer follow a declining trend. PACF eliminates the influence of intermediate variables.