

ENV 797 - Time Series Analysis for Energy and Environment Applications | Spring 2025

Assignment 5 - Due date 02/18/25

Ayoung Kim

Directions

You should open the .rmd file corresponding to this assignment on RStudio. The file is available on our class repository on Github. And to do so you will need to fork our repository and link it to your RStudio.

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_A05_Sp25.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 needed for this assignment: “readxl”, “ggplot2”, “forecast”, “tseries”, and “Kendall”. 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
```

```
library(forecast)
```

```
## Warning: package 'forecast' was built under R version 4.3.3
```

```
## Registered S3 method overwritten by 'quantmod':
```

```
##   method      from
```

```
##   as.zoo.data.frame zoo
```

```
library(tseries)
```

```
## Warning: package 'tseries' was built under R version 4.3.3
```

```
library(ggplot2)
```

```
library(Kendall)
```

```
## Warning: package 'Kendall' was built under R version 4.3.3
```

```
library(lubridate)
```

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

```
library(tidyverse) #load this package so you can clean the data frame using pipes
```

```
## -- Attaching core tidyverse packages ----- tidyverse 2.0.0 --  
## v dplyr   1.1.4      v stringr 1.5.1  
## v forcats 1.0.0      v tibble  3.2.1  
## v purrr   1.0.2      v tidyr   1.3.0  
## v readr   2.1.5
```

```
## -- Conflicts ----- tidyverse_conflicts() --  
## x dplyr::filter() masks stats::filter()  
## x dplyr::lag()     masks stats::lag()  
## i Use the conflicted package (<http://conflicted.r-lib.org/>) to force all conflicts to become errors
```

```
library(zoo)
```

```
##  
## Attaching package: 'zoo'  
##  
## The following objects are masked from 'package:base':  
##  
##     as.Date, as.Date.numeric
```

```
library(dplyr)
```

Decomposing Time Series

Consider the same data you used for A04 from the spreadsheet “Table_10.1_Renewable_Energy_Production_and_Consumption”. The data comes from the US Energy Information and Administration and corresponds to the December 2023 Monthly Energy Review.

```
#Importing data set - using xlsx package  
library(openxlsx)
```

```
## Warning: package 'openxlsx' was built under R version 4.3.3
```

```
library(readxl)
```

```
getwd()
```

```
## [1] "/Users/ayoungkim/TSA_Sp25/Assignments"
```

```
setwd("/Users/ayoungkim/TSA_Sp25/Data")
getwd()
```

```
## [1] "/Users/ayoungkim/TSA_Sp25/Data"
```

```
renewable_data2 <- read.xlsx(xlsxFile="./Table_10.1_Renewable_Energy_Production_and_Consumption_by_Source",
                             sheet = "Monthly Data",
                             startRow = 13,
                             colNames = FALSE)
```

```
read_col_names2 <- read.xlsx(xlsxFile="./Table_10.1_Renewable_Energy_Production_and_Consumption_by_Source",
                              sheet = "Monthly Data",
                              rows = 11,
                              colNames = FALSE)
```

```
colnames(renewable_data2) <- read_col_names2
head(renewable_data2)
```

```
##      Month Wood Energy Production Biofuels Production
## 1 26665          129.630      Not Available
## 2 26696          117.194      Not Available
## 3 26724          129.763      Not Available
## 4 26755          125.462      Not Available
## 5 26785          129.624      Not Available
## 6 26816          125.435      Not Available
##      Total Biomass Energy Production Total Renewable Energy Production
## 1          129.787          219.839
## 2          117.338          197.330
## 3          129.938          218.686
## 4          125.636          209.330
## 5          129.834          215.982
## 6          125.611          208.249
##      Hydroelectric Power Consumption Geothermal Energy Consumption
## 1          89.562          0.490
## 2          79.544          0.448
## 3          88.284          0.464
## 4          83.152          0.542
## 5          85.643          0.505
## 6          82.060          0.579
##      Solar Energy Consumption Wind Energy Consumption Wood Energy Consumption
## 1      Not Available      Not Available          129.630
## 2      Not Available      Not Available          117.194
## 3      Not Available      Not Available          129.763
## 4      Not Available      Not Available          125.462
## 5      Not Available      Not Available          129.624
## 6      Not Available      Not Available          125.435
##      Waste Energy Consumption Biofuels Consumption
## 1          0.157      Not Available
## 2          0.144      Not Available
## 3          0.176      Not Available
## 4          0.174      Not Available
## 5          0.210      Not Available
```

```
## 6          0.176      Not Available
## Total Biomass Energy Consumption Total Renewable Energy Consumption
## 1          129.787          219.839
## 2          117.338          197.330
## 3          129.938          218.686
## 4          125.636          209.330
## 5          129.834          215.982
## 6          125.611          208.249
```

```
nobs=nrow(renewable_data2)
nvar=ncol(renewable_data2)
```

Q1

For this assignment you will work only with the following columns: Solar Energy Consumption and Wind Energy Consumption. Create a data frame structure with these two time series only and the Date column. Drop the rows with *Not Available* and convert the columns to numeric. You can use filtering to eliminate the initial rows or convert to numeric and then use the `drop_na()` function. If you are familiar with pipes for data wrangling, try using it!

```
#Using "select" function, selected only Solar Energy Consumption and Wind Energy Consumption
renewable_data2_filtered <- select(renewable_data2, 'Month', 'Solar Energy Consumption', 'Wind Energy Consumption')

start_date <- "1973-01-01"
renewable_data2_filtered$Month <- as.Date(seq(from = as.Date(start_date),
                                             by = "month",
                                             length.out = nobs))

head(renewable_data2_filtered$Month)
```

```
## [1] "1973-01-01" "1973-02-01" "1973-03-01" "1973-04-01" "1973-05-01"
## [6] "1973-06-01"
```

```
# Converted the format to Year-Month
renewable_data2_filtered$Month <- format(renewable_data2_filtered$Month, "%Y-%m")
head(renewable_data2_filtered)
```

```
##      Month Solar Energy Consumption Wind Energy Consumption
## 1 1973-01          Not Available          Not Available
## 2 1973-02          Not Available          Not Available
## 3 1973-03          Not Available          Not Available
## 4 1973-04          Not Available          Not Available
## 5 1973-05          Not Available          Not Available
## 6 1973-06          Not Available          Not Available
```

```
#Convert to numeric
renewable_data2_filtered$`Solar Energy Consumption` <- as.numeric(renewable_data2_filtered$`Solar Energy Consumption`)
```

```
## Warning: NAs introduced by coercion
```

```
renewable_data2_filtered$`Wind Energy Consumption` <- as.numeric(renewable_data2_filtered$`Wind Energy`)
```

```
## Warning: NAs introduced by coercion
```

```
#Dropped 'Not available'
renewable_data2_filtered <- drop_na(renewable_data2_filtered)
head(renewable_data2_filtered)
```

```
##      Month Solar Energy Consumption Wind Energy Consumption
## 1 1984-01                0.000                0.000
## 2 1984-02                0.000                0.001
## 3 1984-03                0.001                0.001
## 4 1984-04                0.001                0.002
## 5 1984-05                0.002                0.003
## 6 1984-06                0.003                0.002
```

```
#1 ts1 - Solar energy consumption
ts1_renewable_data2_filtered<-ts(renewable_data2_filtered$`Solar Energy Consumption`,
                                start = c(1,1),
                                frequency=12)
ts1_renewable_data2_filtered
```

```
##      Jan      Feb      Mar      Apr      May      Jun      Jul      Aug      Sep
## 1    0.000    0.000    0.001    0.001    0.002    0.003    0.001    0.003    0.003
## 2    0.002    0.000    0.000    0.004    0.005    0.005    0.004    0.007    0.005
## 3    0.002    0.002    0.003    0.005    0.006    0.005    0.006    0.006    0.005
## 4    0.000    0.002    0.003    0.004    0.004    0.006    0.006    0.004    0.003
## 5    0.002    0.003    0.004    0.002    0.004    0.005    0.004    0.004    0.003
## 6    2.821    3.041    4.220    4.660    5.109    5.175    5.476    5.389    4.941
## 7    3.015    3.256    4.489    4.946    5.515    5.534    5.832    5.781    5.228
## 8    3.125    3.369    4.561    5.026    5.739    5.803    6.133    6.035    5.487
## 9    3.196    3.451    4.752    5.284    5.883    5.931    6.274    6.205    5.628
## 10   3.288    3.570    4.954    5.515    6.053    6.058    6.445    6.401    5.818
## 11   3.432    3.680    5.137    5.640    6.216    6.292    6.584    6.471    5.949
## 12   3.415    3.692    5.105    5.735    6.370    6.471    6.750    6.674    6.047
## 13   3.472    3.760    5.293    5.865    6.487    6.562    6.804    6.602    5.956
## 14   3.422    3.762    5.176    5.748    6.292    6.433    6.721    6.612    5.958
## 15   3.375    3.629    5.138    5.664    6.173    6.327    6.679    6.584    5.956
## 16   3.294    3.570    4.984    5.536    6.121    6.184    6.482    6.377    5.775
## 17   3.155    3.440    4.733    5.239    5.832    5.895    6.261    6.144    5.608
## 18   3.008    3.261    4.542    5.020    5.672    5.755    6.029    5.947    5.380
## 19   2.923    3.189    4.435    4.877    5.408    5.585    5.812    5.719    5.162
## 20   2.848    3.083    4.334    4.795    5.298    5.421    5.579    5.522    5.034
## 21   2.809    3.019    4.289    4.722    5.278    5.343    5.575    5.488    4.986
## 22   2.742    2.968    4.158    4.638    5.178    5.247    5.438    5.396    4.894
## 23   2.881    3.124    4.327    4.821    5.368    5.410    5.640    5.657    5.085
## 24   3.016    3.269    4.579    5.051    5.658    5.707    5.981    5.886    5.367
## 25   3.206    3.518    4.933    5.478    6.029    6.181    6.408    6.326    5.758
## 26   3.247    3.578    5.053    5.614    6.200    6.229    6.582    6.501    5.892
## 27   3.485    3.833    5.386    6.033    6.760    6.895    7.158    7.072    6.433
## 28   3.924    4.374    6.043    6.760    7.496    7.667    7.903    7.958    7.178
## 29   4.607    5.077    7.148    8.096    9.316    9.605    9.934    9.685    8.960
```

## 30	5.869	6.663	9.260	10.151	11.264	11.745	12.038	12.336	11.551
## 31	8.157	8.799	12.624	13.934	15.758	16.428	16.395	16.624	15.631
## 32	9.815	11.480	15.989	18.058	19.510	19.804	20.660	20.720	18.026
## 33	11.728	15.428	19.297	21.401	24.459	24.955	27.056	26.741	24.199
## 34	15.555	17.857	27.472	30.175	34.567	36.083	34.635	33.492	30.881
## 35	20.417	23.213	30.918	36.049	40.277	42.476	40.715	39.785	35.355
## 36	22.249	23.942	35.490	40.146	43.146	46.198	47.572	45.914	40.157
## 37	26.741	32.049	38.731	46.045	54.208	54.219	58.159	52.712	44.933
## 38	32.034	35.565	51.477	59.068	66.559	65.882	66.269	64.229	59.026
## 39	41.781	47.414	62.793	71.077	79.465	82.616	82.569	77.172	70.117
## 40	43.772	51.049	67.465	80.376	91.383	92.683	97.656	92.824	81.556
## 41	53.206	65.064	83.696	98.077	111.667	118.688	119.326	117.388	100.303
##	Oct	Nov	Dec						
## 1	0.002	0.001	0.000						
## 2	0.003	0.001	0.001						
## 3	0.004	0.003	0.001						
## 4	0.002	0.001	0.001						
## 5	0.000	0.000	0.000						
## 6	4.501	3.697	3.465						
## 7	4.782	3.884	3.700						
## 8	4.935	4.034	3.792						
## 9	5.115	4.187	3.914						
## 10	5.243	4.297	4.039						
## 11	5.458	4.415	4.136						
## 12	5.471	4.473	4.192						
## 13	5.494	4.495	4.262						
## 14	5.437	4.424	4.187						
## 15	5.395	4.389	4.147						
## 16	5.311	4.307	4.068						
## 17	5.056	4.103	3.882						
## 18	4.773	3.905	3.675						
## 19	4.647	3.821	3.551						
## 20	4.535	3.669	3.452						
## 21	4.470	3.625	3.418						
## 22	4.399	3.547	3.335						
## 23	4.577	3.718	3.487						
## 24	4.848	3.922	3.662						
## 25	5.166	4.167	3.926						
## 26	5.307	4.300	4.023						
## 27	5.693	4.721	4.384						
## 28	6.507	5.259	5.056						
## 29	8.214	6.715	6.439						
## 30	10.946	9.028	8.800						
## 31	14.507	11.805	10.387						
## 32	15.938	13.628	12.547						
## 33	21.438	17.985	16.202						
## 34	28.042	20.501	19.362						
## 35	30.386	23.468	20.576						
## 36	35.724	26.634	22.573						
## 37	40.674	33.068	29.778						
## 38	49.778	42.082	34.895						
## 39	63.195	46.708	39.656						
## 40	74.296	56.874	50.393						
## 41									

```
#2 ts2 - Wind Energy Consumption
ts2_renewable_data2_filtered<-ts(renewable_data2_filtered$'Wind Energy Consumption',
                                start = c(1,1),
                                frequency=12)
ts2_renewable_data2_filtered
```

##	Jan	Feb	Mar	Apr	May	Jun	Jul	Aug	Sep
## 1	0.000	0.001	0.001	0.002	0.003	0.002	0.002	0.001	0.002
## 2	0.002	0.004	0.002	0.002	0.001	0.001	0.000	0.001	0.003
## 3	0.001	0.001	0.002	0.002	0.003	0.002	0.002	0.002	0.000
## 4	0.000	0.000	0.000	0.000	0.001	0.003	0.004	0.001	0.001
## 5	0.002	0.000	0.000	0.000	0.000	0.000	0.000	0.000	0.000
## 6	0.497	0.763	0.873	0.812	0.697	0.534	0.485	0.352	0.956
## 7	1.028	0.789	0.837	1.434	1.195	1.387	0.406	0.239	0.311
## 8	0.212	0.989	0.707	0.813	1.237	0.530	0.530	0.777	0.707
## 9	0.512	1.759	1.120	1.088	0.320	0.256	0.256	0.480	1.088
## 10	0.844	0.169	1.773	1.266	1.266	0.295	0.464	0.253	0.464
## 11	0.000	0.343	1.865	1.408	1.256	1.256	0.647	0.457	1.066
## 12	0.019	0.080	0.016	0.023	1.394	1.701	2.115	1.862	1.955
## 13	0.503	0.382	0.640	0.834	1.336	1.282	1.826	1.416	1.199
## 14	0.411	0.372	0.507	1.106	1.196	1.764	1.738	1.809	0.888
## 15	0.059	0.028	0.021	0.293	0.489	1.348	1.868	1.438	1.623
## 16	1.150	1.054	1.524	1.274	0.940	0.866	1.556	1.304	1.288
## 17	2.202	2.409	1.194	1.126	1.392	1.525	1.296	1.284	1.134
## 18	1.328	1.471	1.816	2.336	2.167	2.285	2.167	1.969	1.673
## 19	2.769	2.436	2.907	3.495	3.677	3.843	3.038	3.332	2.511
## 20	2.158	2.543	3.535	3.729	3.434	3.573	3.253	2.782	3.054
## 21	3.410	3.486	4.406	4.418	5.806	4.767	3.973	3.585	3.718
## 22	3.861	3.298	5.325	5.792	5.959	6.131	4.848	3.883	5.010
## 23	8.130	6.558	8.048	8.434	8.389	7.001	6.671	5.647	6.412
## 24	8.367	8.598	10.397	10.822	10.073	8.941	7.364	9.210	9.781
## 25	14.580	13.142	16.316	17.829	18.221	17.539	13.677	11.138	10.616
## 26	20.304	19.968	24.222	25.446	21.366	19.105	16.906	18.645	15.868
## 27	23.387	18.533	29.306	33.316	29.676	27.463	22.942	22.812	24.244
## 28	29.174	35.661	35.978	42.383	40.167	37.481	25.551	25.500	23.437
## 29	46.508	37.708	47.861	43.363	42.788	40.849	30.104	28.898	29.991
## 30	50.288	48.026	53.758	59.629	55.406	46.909	37.851	32.871	39.832
## 31	61.113	47.798	60.515	63.584	53.232	53.906	41.583	34.702	39.305
## 32	51.733	50.912	52.231	60.963	58.520	45.793	46.661	44.629	47.671
## 33	63.007	68.712	74.857	70.967	64.309	55.627	60.114	46.367	55.969
## 34	70.965	75.375	87.793	86.590	78.707	68.724	55.001	47.355	61.115
## 35	87.343	79.123	90.294	90.182	81.728	84.286	56.116	67.716	63.189
## 36	82.917	77.189	87.936	98.659	87.959	76.586	75.408	68.165	83.640
## 37	95.950	99.325	100.039	101.515	96.824	103.085	78.019	78.576	79.111
## 38	102.566	91.153	133.768	123.370	115.280	91.003	74.093	92.367	98.941
## 39	127.664	128.443	146.820	157.522	143.726	115.215	100.569	84.339	93.254
## 40	130.877	141.340	148.708	145.848	109.959	93.990	95.552	96.881	96.742
## 41	118.833	141.413	155.252	161.170	131.666	129.865	95.229	97.908	98.641
##	Oct	Nov	Dec						
## 1	0.003	0.003	0.004						
## 2	0.002	0.001	0.000						
## 3	0.001	0.000	0.000						
## 4	0.000	0.000	0.001						

```
## 5    0.000    0.000    0.000
## 6    0.599    0.419    0.219
## 7    0.693    0.789    0.406
## 8    1.448    1.448    0.671
## 9    0.832    0.960    1.184
## 10   0.971    1.562    0.929
## 11   1.218    1.675    0.571
## 12   0.876    0.427    0.329
## 13   0.863    0.337    0.417
## 14   0.937    0.248    0.244
## 15   1.019    0.618    1.519
## 16   1.428    1.209    1.720
## 17   1.404    2.808    1.309
## 18   2.071    1.605    2.101
## 19   2.506    2.238    2.577
## 20   3.061    3.280    3.770
## 21   3.511    3.181    3.999
## 22   4.934    5.492    6.237
## 23   8.333    8.668    8.433
## 24  11.521   10.560   11.909
## 25  16.229   17.038   22.573
## 26  23.248   23.458   23.563
## 27  27.104   33.259   30.910
## 28  35.913   42.440   36.358
## 29  43.114   39.745   49.554
## 30  46.523   53.921   47.656
## 31  49.501   64.374   50.195
## 32  55.889   67.154   68.576
## 33  69.384   66.212   78.973
## 34  83.146   77.162   75.749
## 35  72.314   75.118   82.933
## 36  94.255   85.929   90.909
## 37  98.343  113.038  109.220
## 38 109.918  121.983  135.965
## 39 111.725  140.570  131.975
## 40 122.900  124.352  129.787
## 41
```

```
#Converting time series to numeric
solar_energy_numeric <- as.numeric(ts1_renewable_data2_filtered)
wind_energy_numeric <- as.numeric(ts2_renewable_data2_filtered)

#Make two ts and month as a dataframe
df_ts_renewable_data2_filtered <- data.frame(
  Month = renewable_data2_filtered$Month,
  Solar_Energy_Consumption = solar_energy_numeric,
  Wind_Energy_Consumption = wind_energy_numeric)

df_ts_renewable_data2_filtered
```

```
##      Month Solar_Energy_Consumption Wind_Energy_Consumption
## 1  1984-01                0.000                0.000
## 2  1984-02                0.000                0.001
## 3  1984-03                0.001                0.001
```


## 4	1984-04	0.001	0.002
## 5	1984-05	0.002	0.003
## 6	1984-06	0.003	0.002
## 7	1984-07	0.001	0.002
## 8	1984-08	0.003	0.001
## 9	1984-09	0.003	0.002
## 10	1984-10	0.002	0.003
## 11	1984-11	0.001	0.003
## 12	1984-12	0.000	0.004
## 13	1985-01	0.002	0.002
## 14	1985-02	0.000	0.004
## 15	1985-03	0.000	0.002
## 16	1985-04	0.004	0.002
## 17	1985-05	0.005	0.001
## 18	1985-06	0.005	0.001
## 19	1985-07	0.004	0.000
## 20	1985-08	0.007	0.001
## 21	1985-09	0.005	0.003
## 22	1985-10	0.003	0.002
## 23	1985-11	0.001	0.001
## 24	1985-12	0.001	0.000
## 25	1986-01	0.002	0.001
## 26	1986-02	0.002	0.001
## 27	1986-03	0.003	0.002
## 28	1986-04	0.005	0.002
## 29	1986-05	0.006	0.003
## 30	1986-06	0.005	0.002
## 31	1986-07	0.006	0.002
## 32	1986-08	0.006	0.002
## 33	1986-09	0.005	0.000
## 34	1986-10	0.004	0.001
## 35	1986-11	0.003	0.000
## 36	1986-12	0.001	0.000
## 37	1987-01	0.000	0.000
## 38	1987-02	0.002	0.000
## 39	1987-03	0.003	0.000
## 40	1987-04	0.004	0.000
## 41	1987-05	0.004	0.001
## 42	1987-06	0.006	0.003
## 43	1987-07	0.006	0.004
## 44	1987-08	0.004	0.001
## 45	1987-09	0.003	0.001
## 46	1987-10	0.002	0.000
## 47	1987-11	0.001	0.000
## 48	1987-12	0.001	0.001
## 49	1988-01	0.002	0.002
## 50	1988-02	0.003	0.000
## 51	1988-03	0.004	0.000
## 52	1988-04	0.002	0.000
## 53	1988-05	0.004	0.000
## 54	1988-06	0.005	0.000
## 55	1988-07	0.004	0.000
## 56	1988-08	0.004	0.000
## 57	1988-09	0.003	0.000

## 58	1988-10	0.000	0.000
## 59	1988-11	0.000	0.000
## 60	1988-12	0.000	0.000
## 61	1989-01	2.821	0.497
## 62	1989-02	3.041	0.763
## 63	1989-03	4.220	0.873
## 64	1989-04	4.660	0.812
## 65	1989-05	5.109	0.697
## 66	1989-06	5.175	0.534
## 67	1989-07	5.476	0.485
## 68	1989-08	5.389	0.352
## 69	1989-09	4.941	0.956
## 70	1989-10	4.501	0.599
## 71	1989-11	3.697	0.419
## 72	1989-12	3.465	0.219
## 73	1990-01	3.015	1.028
## 74	1990-02	3.256	0.789
## 75	1990-03	4.489	0.837
## 76	1990-04	4.946	1.434
## 77	1990-05	5.515	1.195
## 78	1990-06	5.534	1.387
## 79	1990-07	5.832	0.406
## 80	1990-08	5.781	0.239
## 81	1990-09	5.228	0.311
## 82	1990-10	4.782	0.693
## 83	1990-11	3.884	0.789
## 84	1990-12	3.700	0.406
## 85	1991-01	3.125	0.212
## 86	1991-02	3.369	0.989
## 87	1991-03	4.561	0.707
## 88	1991-04	5.026	0.813
## 89	1991-05	5.739	1.237
## 90	1991-06	5.803	0.530
## 91	1991-07	6.133	0.530
## 92	1991-08	6.035	0.777
## 93	1991-09	5.487	0.707
## 94	1991-10	4.935	1.448
## 95	1991-11	4.034	1.448
## 96	1991-12	3.792	0.671
## 97	1992-01	3.196	0.512
## 98	1992-02	3.451	1.759
## 99	1992-03	4.752	1.120
## 100	1992-04	5.284	1.088
## 101	1992-05	5.883	0.320
## 102	1992-06	5.931	0.256
## 103	1992-07	6.274	0.256
## 104	1992-08	6.205	0.480
## 105	1992-09	5.628	1.088
## 106	1992-10	5.115	0.832
## 107	1992-11	4.187	0.960
## 108	1992-12	3.914	1.184
## 109	1993-01	3.288	0.844
## 110	1993-02	3.570	0.169
## 111	1993-03	4.954	1.773

## 112 1993-04	5.515	1.266
## 113 1993-05	6.053	1.266
## 114 1993-06	6.058	0.295
## 115 1993-07	6.445	0.464
## 116 1993-08	6.401	0.253
## 117 1993-09	5.818	0.464
## 118 1993-10	5.243	0.971
## 119 1993-11	4.297	1.562
## 120 1993-12	4.039	0.929
## 121 1994-01	3.432	0.000
## 122 1994-02	3.680	0.343
## 123 1994-03	5.137	1.865
## 124 1994-04	5.640	1.408
## 125 1994-05	6.216	1.256
## 126 1994-06	6.292	1.256
## 127 1994-07	6.584	0.647
## 128 1994-08	6.471	0.457
## 129 1994-09	5.949	1.066
## 130 1994-10	5.458	1.218
## 131 1994-11	4.415	1.675
## 132 1994-12	4.136	0.571
## 133 1995-01	3.415	0.019
## 134 1995-02	3.692	0.080
## 135 1995-03	5.105	0.016
## 136 1995-04	5.735	0.023
## 137 1995-05	6.370	1.394
## 138 1995-06	6.471	1.701
## 139 1995-07	6.750	2.115
## 140 1995-08	6.674	1.862
## 141 1995-09	6.047	1.955
## 142 1995-10	5.471	0.876
## 143 1995-11	4.473	0.427
## 144 1995-12	4.192	0.329
## 145 1996-01	3.472	0.503
## 146 1996-02	3.760	0.382
## 147 1996-03	5.293	0.640
## 148 1996-04	5.865	0.834
## 149 1996-05	6.487	1.336
## 150 1996-06	6.562	1.282
## 151 1996-07	6.804	1.826
## 152 1996-08	6.602	1.416
## 153 1996-09	5.956	1.199
## 154 1996-10	5.494	0.863
## 155 1996-11	4.495	0.337
## 156 1996-12	4.262	0.417
## 157 1997-01	3.422	0.411
## 158 1997-02	3.762	0.372
## 159 1997-03	5.176	0.507
## 160 1997-04	5.748	1.106
## 161 1997-05	6.292	1.196
## 162 1997-06	6.433	1.764
## 163 1997-07	6.721	1.738
## 164 1997-08	6.612	1.809
## 165 1997-09	5.958	0.888

## 166 1997-10	5.437	0.937
## 167 1997-11	4.424	0.248
## 168 1997-12	4.187	0.244
## 169 1998-01	3.375	0.059
## 170 1998-02	3.629	0.028
## 171 1998-03	5.138	0.021
## 172 1998-04	5.664	0.293
## 173 1998-05	6.173	0.489
## 174 1998-06	6.327	1.348
## 175 1998-07	6.679	1.868
## 176 1998-08	6.584	1.438
## 177 1998-09	5.956	1.623
## 178 1998-10	5.395	1.019
## 179 1998-11	4.389	0.618
## 180 1998-12	4.147	1.519
## 181 1999-01	3.294	1.150
## 182 1999-02	3.570	1.054
## 183 1999-03	4.984	1.524
## 184 1999-04	5.536	1.274
## 185 1999-05	6.121	0.940
## 186 1999-06	6.184	0.866
## 187 1999-07	6.482	1.556
## 188 1999-08	6.377	1.304
## 189 1999-09	5.775	1.288
## 190 1999-10	5.311	1.428
## 191 1999-11	4.307	1.209
## 192 1999-12	4.068	1.720
## 193 2000-01	3.155	2.202
## 194 2000-02	3.440	2.409
## 195 2000-03	4.733	1.194
## 196 2000-04	5.239	1.126
## 197 2000-05	5.832	1.392
## 198 2000-06	5.895	1.525
## 199 2000-07	6.261	1.296
## 200 2000-08	6.144	1.284
## 201 2000-09	5.608	1.134
## 202 2000-10	5.056	1.404
## 203 2000-11	4.103	2.808
## 204 2000-12	3.882	1.309
## 205 2001-01	3.008	1.328
## 206 2001-02	3.261	1.471
## 207 2001-03	4.542	1.816
## 208 2001-04	5.020	2.336
## 209 2001-05	5.672	2.167
## 210 2001-06	5.755	2.285
## 211 2001-07	6.029	2.167
## 212 2001-08	5.947	1.969
## 213 2001-09	5.380	1.673
## 214 2001-10	4.773	2.071
## 215 2001-11	3.905	1.605
## 216 2001-12	3.675	2.101
## 217 2002-01	2.923	2.769
## 218 2002-02	3.189	2.436
## 219 2002-03	4.435	2.907

## 220	2002-04	4.877	3.495
## 221	2002-05	5.408	3.677
## 222	2002-06	5.585	3.843
## 223	2002-07	5.812	3.038
## 224	2002-08	5.719	3.332
## 225	2002-09	5.162	2.511
## 226	2002-10	4.647	2.506
## 227	2002-11	3.821	2.238
## 228	2002-12	3.551	2.577
## 229	2003-01	2.848	2.158
## 230	2003-02	3.083	2.543
## 231	2003-03	4.334	3.535
## 232	2003-04	4.795	3.729
## 233	2003-05	5.298	3.434
## 234	2003-06	5.421	3.573
## 235	2003-07	5.579	3.253
## 236	2003-08	5.522	2.782
## 237	2003-09	5.034	3.054
## 238	2003-10	4.535	3.061
## 239	2003-11	3.669	3.280
## 240	2003-12	3.452	3.770
## 241	2004-01	2.809	3.410
## 242	2004-02	3.019	3.486
## 243	2004-03	4.289	4.406
## 244	2004-04	4.722	4.418
## 245	2004-05	5.278	5.806
## 246	2004-06	5.343	4.767
## 247	2004-07	5.575	3.973
## 248	2004-08	5.488	3.585
## 249	2004-09	4.986	3.718
## 250	2004-10	4.470	3.511
## 251	2004-11	3.625	3.181
## 252	2004-12	3.418	3.999
## 253	2005-01	2.742	3.861
## 254	2005-02	2.968	3.298
## 255	2005-03	4.158	5.325
## 256	2005-04	4.638	5.792
## 257	2005-05	5.178	5.959
## 258	2005-06	5.247	6.131
## 259	2005-07	5.438	4.848
## 260	2005-08	5.396	3.883
## 261	2005-09	4.894	5.010
## 262	2005-10	4.399	4.934
## 263	2005-11	3.547	5.492
## 264	2005-12	3.335	6.237
## 265	2006-01	2.881	8.130
## 266	2006-02	3.124	6.558
## 267	2006-03	4.327	8.048
## 268	2006-04	4.821	8.434
## 269	2006-05	5.368	8.389
## 270	2006-06	5.410	7.001
## 271	2006-07	5.640	6.671
## 272	2006-08	5.657	5.647
## 273	2006-09	5.085	6.412

## 274 2006-10	4.577	8.333
## 275 2006-11	3.718	8.668
## 276 2006-12	3.487	8.433
## 277 2007-01	3.016	8.367
## 278 2007-02	3.269	8.598
## 279 2007-03	4.579	10.397
## 280 2007-04	5.051	10.822
## 281 2007-05	5.658	10.073
## 282 2007-06	5.707	8.941
## 283 2007-07	5.981	7.364
## 284 2007-08	5.886	9.210
## 285 2007-09	5.367	9.781
## 286 2007-10	4.848	11.521
## 287 2007-11	3.922	10.560
## 288 2007-12	3.662	11.909
## 289 2008-01	3.206	14.580
## 290 2008-02	3.518	13.142
## 291 2008-03	4.933	16.316
## 292 2008-04	5.478	17.829
## 293 2008-05	6.029	18.221
## 294 2008-06	6.181	17.539
## 295 2008-07	6.408	13.677
## 296 2008-08	6.326	11.138
## 297 2008-09	5.758	10.616
## 298 2008-10	5.166	16.229
## 299 2008-11	4.167	17.038
## 300 2008-12	3.926	22.573
## 301 2009-01	3.247	20.304
## 302 2009-02	3.578	19.968
## 303 2009-03	5.053	24.222
## 304 2009-04	5.614	25.446
## 305 2009-05	6.200	21.366
## 306 2009-06	6.229	19.105
## 307 2009-07	6.582	16.906
## 308 2009-08	6.501	18.645
## 309 2009-09	5.892	15.868
## 310 2009-10	5.307	23.248
## 311 2009-11	4.300	23.458
## 312 2009-12	4.023	23.563
## 313 2010-01	3.485	23.387
## 314 2010-02	3.833	18.533
## 315 2010-03	5.386	29.306
## 316 2010-04	6.033	33.316
## 317 2010-05	6.760	29.676
## 318 2010-06	6.895	27.463
## 319 2010-07	7.158	22.942
## 320 2010-08	7.072	22.812
## 321 2010-09	6.433	24.244
## 322 2010-10	5.693	27.104
## 323 2010-11	4.721	33.259
## 324 2010-12	4.384	30.910
## 325 2011-01	3.924	29.174
## 326 2011-02	4.374	35.661
## 327 2011-03	6.043	35.978

## 328 2011-04	6.760	42.383
## 329 2011-05	7.496	40.167
## 330 2011-06	7.667	37.481
## 331 2011-07	7.903	25.551
## 332 2011-08	7.958	25.500
## 333 2011-09	7.178	23.437
## 334 2011-10	6.507	35.913
## 335 2011-11	5.259	42.440
## 336 2011-12	5.056	36.358
## 337 2012-01	4.607	46.508
## 338 2012-02	5.077	37.708
## 339 2012-03	7.148	47.861
## 340 2012-04	8.096	43.363
## 341 2012-05	9.316	42.788
## 342 2012-06	9.605	40.849
## 343 2012-07	9.934	30.104
## 344 2012-08	9.685	28.898
## 345 2012-09	8.960	29.991
## 346 2012-10	8.214	43.114
## 347 2012-11	6.715	39.745
## 348 2012-12	6.439	49.554
## 349 2013-01	5.869	50.288
## 350 2013-02	6.663	48.026
## 351 2013-03	9.260	53.758
## 352 2013-04	10.151	59.629
## 353 2013-05	11.264	55.406
## 354 2013-06	11.745	46.909
## 355 2013-07	12.038	37.851
## 356 2013-08	12.336	32.871
## 357 2013-09	11.551	39.832
## 358 2013-10	10.946	46.523
## 359 2013-11	9.028	53.921
## 360 2013-12	8.800	47.656
## 361 2014-01	8.157	61.113
## 362 2014-02	8.799	47.798
## 363 2014-03	12.624	60.515
## 364 2014-04	13.934	63.584
## 365 2014-05	15.758	53.232
## 366 2014-06	16.428	53.906
## 367 2014-07	16.395	41.583
## 368 2014-08	16.624	34.702
## 369 2014-09	15.631	39.305
## 370 2014-10	14.507	49.501
## 371 2014-11	11.805	64.374
## 372 2014-12	10.387	50.195
## 373 2015-01	9.815	51.733
## 374 2015-02	11.480	50.912
## 375 2015-03	15.989	52.231
## 376 2015-04	18.058	60.963
## 377 2015-05	19.510	58.520
## 378 2015-06	19.804	45.793
## 379 2015-07	20.660	46.661
## 380 2015-08	20.720	44.629
## 381 2015-09	18.026	47.671

## 382 2015-10	15.938	55.889
## 383 2015-11	13.628	67.154
## 384 2015-12	12.547	68.576
## 385 2016-01	11.728	63.007
## 386 2016-02	15.428	68.712
## 387 2016-03	19.297	74.857
## 388 2016-04	21.401	70.967
## 389 2016-05	24.459	64.309
## 390 2016-06	24.955	55.627
## 391 2016-07	27.056	60.114
## 392 2016-08	26.741	46.367
## 393 2016-09	24.199	55.969
## 394 2016-10	21.438	69.384
## 395 2016-11	17.985	66.212
## 396 2016-12	16.202	78.973
## 397 2017-01	15.555	70.965
## 398 2017-02	17.857	75.375
## 399 2017-03	27.472	87.793
## 400 2017-04	30.175	86.590
## 401 2017-05	34.567	78.707
## 402 2017-06	36.083	68.724
## 403 2017-07	34.635	55.001
## 404 2017-08	33.492	47.355
## 405 2017-09	30.881	61.115
## 406 2017-10	28.042	83.146
## 407 2017-11	20.501	77.162
## 408 2017-12	19.362	75.749
## 409 2018-01	20.417	87.343
## 410 2018-02	23.213	79.123
## 411 2018-03	30.918	90.294
## 412 2018-04	36.049	90.182
## 413 2018-05	40.277	81.728
## 414 2018-06	42.476	84.286
## 415 2018-07	40.715	56.116
## 416 2018-08	39.785	67.716
## 417 2018-09	35.355	63.189
## 418 2018-10	30.386	72.314
## 419 2018-11	23.468	75.118
## 420 2018-12	20.576	82.933
## 421 2019-01	22.249	82.917
## 422 2019-02	23.942	77.189
## 423 2019-03	35.490	87.936
## 424 2019-04	40.146	98.659
## 425 2019-05	43.146	87.959
## 426 2019-06	46.198	76.586
## 427 2019-07	47.572	75.408
## 428 2019-08	45.914	68.165
## 429 2019-09	40.157	83.640
## 430 2019-10	35.724	94.255
## 431 2019-11	26.634	85.929
## 432 2019-12	22.573	90.909
## 433 2020-01	26.741	95.950
## 434 2020-02	32.049	99.325
## 435 2020-03	38.731	100.039

## 436 2020-04	46.045	101.515
## 437 2020-05	54.208	96.824
## 438 2020-06	54.219	103.085
## 439 2020-07	58.159	78.019
## 440 2020-08	52.712	78.576
## 441 2020-09	44.933	79.111
## 442 2020-10	40.674	98.343
## 443 2020-11	33.068	113.038
## 444 2020-12	29.778	109.220
## 445 2021-01	32.034	102.566
## 446 2021-02	35.565	91.153
## 447 2021-03	51.477	133.768
## 448 2021-04	59.068	123.370
## 449 2021-05	66.559	115.280
## 450 2021-06	65.882	91.003
## 451 2021-07	66.269	74.093
## 452 2021-08	64.229	92.367
## 453 2021-09	59.026	98.941
## 454 2021-10	49.778	109.918
## 455 2021-11	42.082	121.983
## 456 2021-12	34.895	135.965
## 457 2022-01	41.781	127.664
## 458 2022-02	47.414	128.443
## 459 2022-03	62.793	146.820
## 460 2022-04	71.077	157.522
## 461 2022-05	79.465	143.726
## 462 2022-06	82.616	115.215
## 463 2022-07	82.569	100.569
## 464 2022-08	77.172	84.339
## 465 2022-09	70.117	93.254
## 466 2022-10	63.195	111.725
## 467 2022-11	46.708	140.570
## 468 2022-12	39.656	131.975
## 469 2023-01	43.772	130.877
## 470 2023-02	51.049	141.340
## 471 2023-03	67.465	148.708
## 472 2023-04	80.376	145.848
## 473 2023-05	91.383	109.959
## 474 2023-06	92.683	93.990
## 475 2023-07	97.656	95.552
## 476 2023-08	92.824	96.881
## 477 2023-09	81.556	96.742
## 478 2023-10	74.296	122.900
## 479 2023-11	56.874	124.352
## 480 2023-12	50.393	129.787
## 481 2024-01	53.206	118.833
## 482 2024-02	65.064	141.413
## 483 2024-03	83.696	155.252
## 484 2024-04	98.077	161.170
## 485 2024-05	111.667	131.666
## 486 2024-06	118.688	129.865
## 487 2024-07	119.326	95.229
## 488 2024-08	117.388	97.908
## 489 2024-09	100.303	98.641

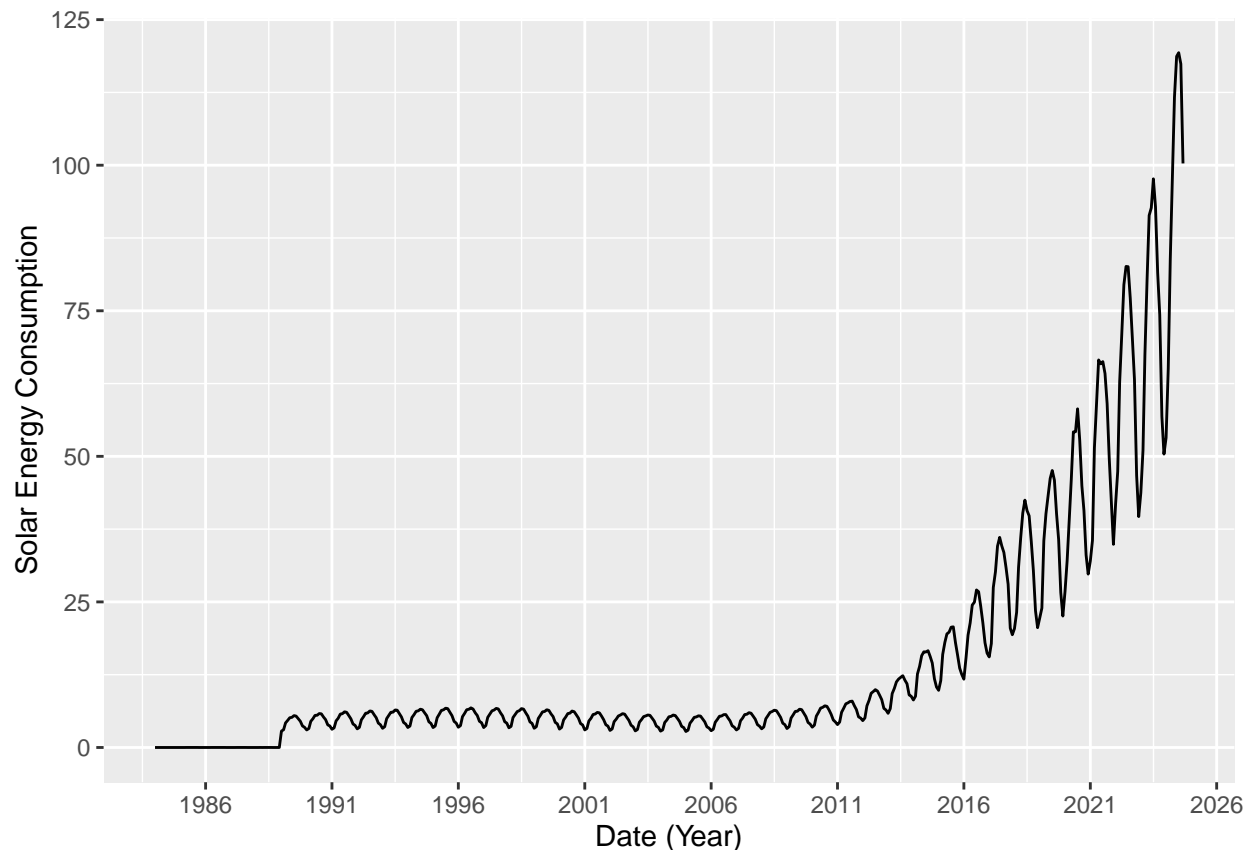
Q2

Plot the Solar and Wind energy consumption over time using ggplot. Plot each series on a separate graph. No need to add legend. Add informative names to the y axis using `ylab()`. Explore the function `scale_x_date()` on ggplot and see if you can change the x axis to improve your plot. Hint: use `scale_x_date(date_breaks = "5 years", date_labels = "%Y")`

```
# Convert the 'Month' column to Date format (Confirmed my code with gpt)
df_ts_renewable_data2_filtered$Month <- as.Date(paste0(df_ts_renewable_data2_filtered$Month, "-01"), for

#1 Plot - Solar Energy Consumption + ylab, scale_x_date
plot_ts1 <- ggplot(df_ts_renewable_data2_filtered, aes(x = Month, y = Solar_Energy_Consumption)) +
  geom_line() +
  ylab("Solar Energy Consumption") +
  xlab("Date (Year)") +
  scale_x_date(date_breaks = "5 years", date_labels = "%Y")

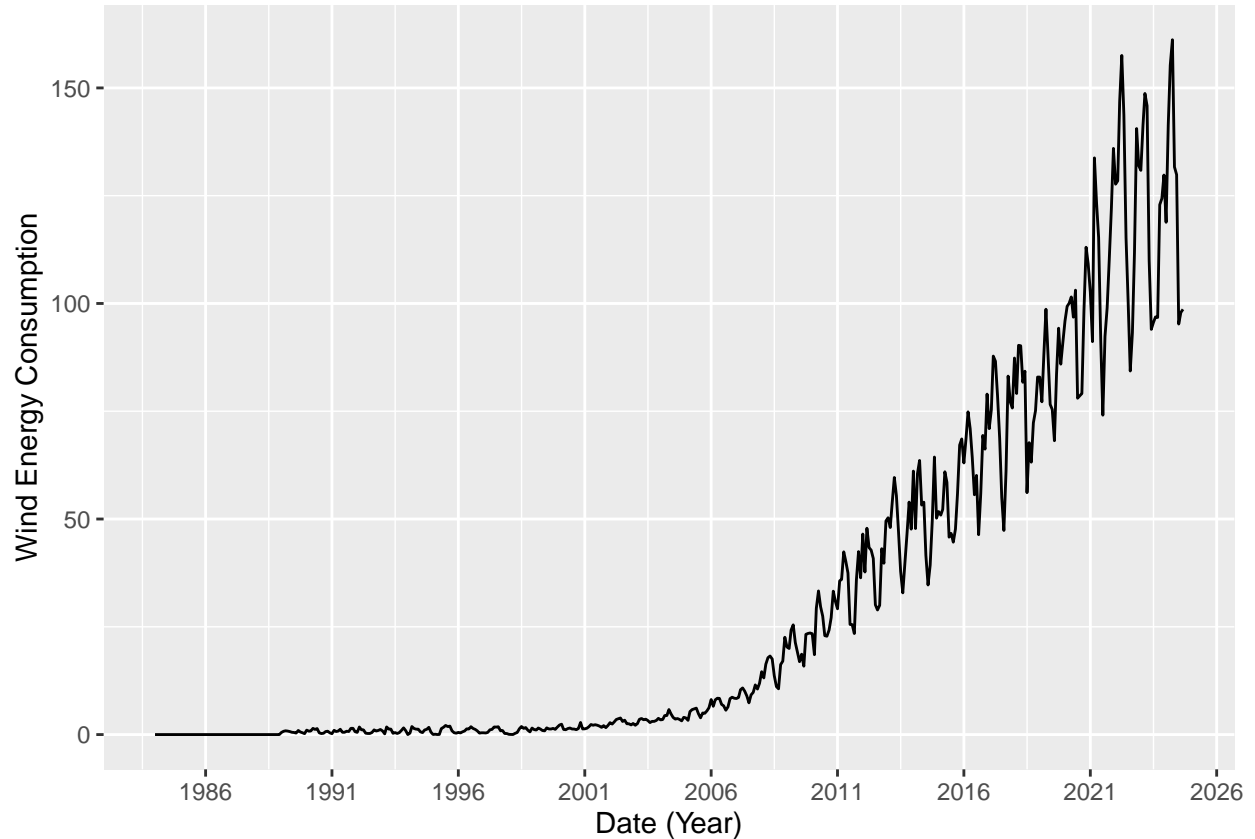
print(plot_ts1)
```



```
#2 Plot - Wind Energy Consumption
plot_ts2 <- ggplot(df_ts_renewable_data2_filtered,
  aes(x = Month,
    y = Wind_Energy_Consumption)) +
  ylab("Wind Energy Consumption") +
  xlab("Date (Year)") +
```

```
geom_line()+
scale_x_date(date_breaks = "5 years", date_labels = "%Y")

print(plot_ts2)
```

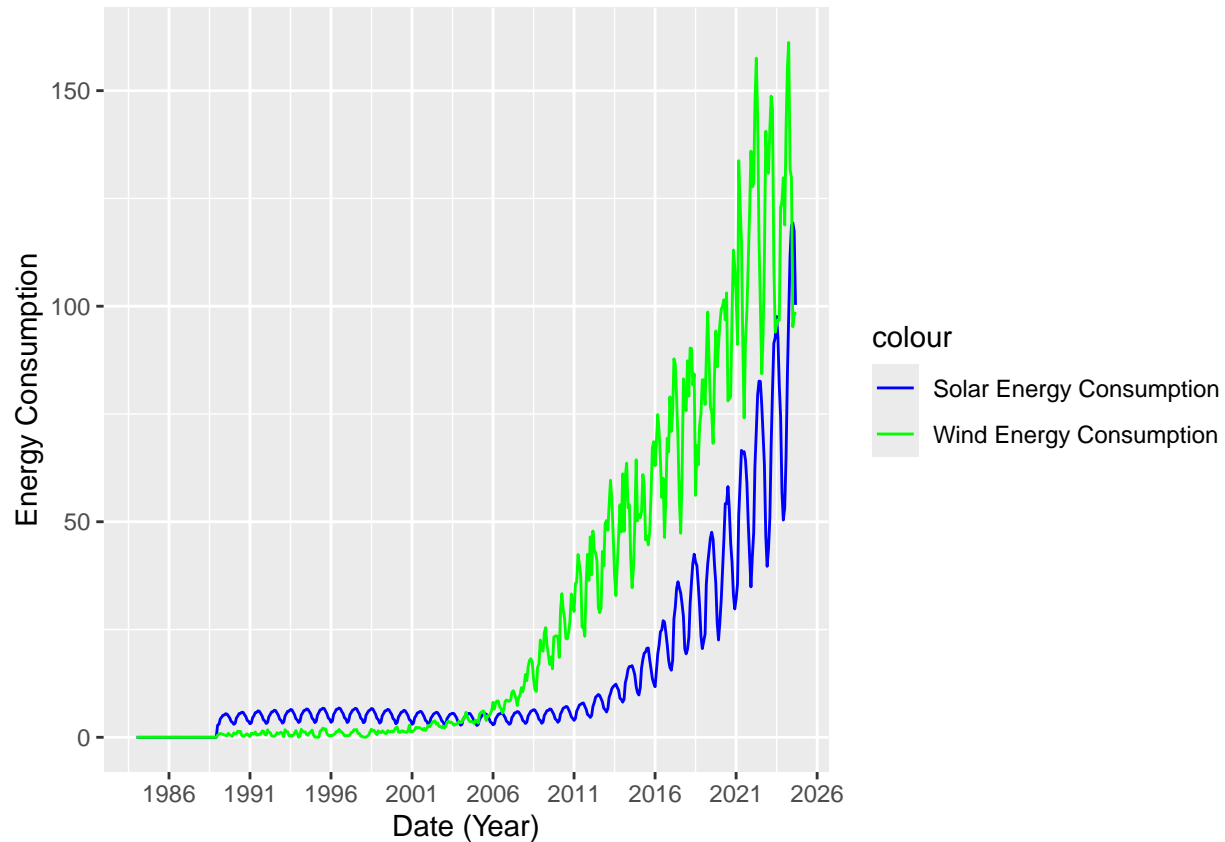


Q3

Now plot both series in the same graph, also using `ggplot()`. Use function `scale_color_manual()` to manually add a legend to `ggplot()`. Make the solar energy consumption red and wind energy consumption blue. Add informative name to the y axis using `ylab("Energy Consumption")`. And use function `scale_x_date()` to set x axis breaks every 5 years.

```
plot_ts_combined <- ggplot() +
  geom_line(data = df_ts_renewable_data2_filtered,
    aes(x = Month,
      y = Solar_Energy_Consumption,
      color = "Solar Energy Consumption")) +
  geom_line(data = df_ts_renewable_data2_filtered,
    aes(x = Month,
      y = Wind_Energy_Consumption,
      color = "Wind Energy Consumption")) +
  scale_color_manual(values = c("Solar Energy Consumption" = "blue", "Wind Energy Consumption" = "green")) +
  xlab("Date (Year)") +
  ylab("Energy Consumption") +
```

```
scale_x_date(date_breaks = "5 years", date_labels = "%Y")
print(plot_ts_combined)
```



Decomposing the time series

The stats package has a function called `decompose()`. This function only take time series object. As the name says the decompose function will decompose your time series into three components: trend, seasonal and random. This is similar to what we did in the previous script, but in a more automated way. The random component is the time series without seasonal and trend component.

Additional info on `decompose()`.

- 1) You have two options: alternative and multiplicative. Multiplicative models exhibit a change in frequency over time.
- 2) The trend is not a straight line because it uses a moving average method to detect trend.
- 3) The seasonal component of the time series is found by subtracting the trend component from the original data then grouping the results by month and averaging them.
- 4) The random component, also referred to as the noise component, is composed of all the leftover signal which is not explained by the combination of the trend and seasonal component.

Q4

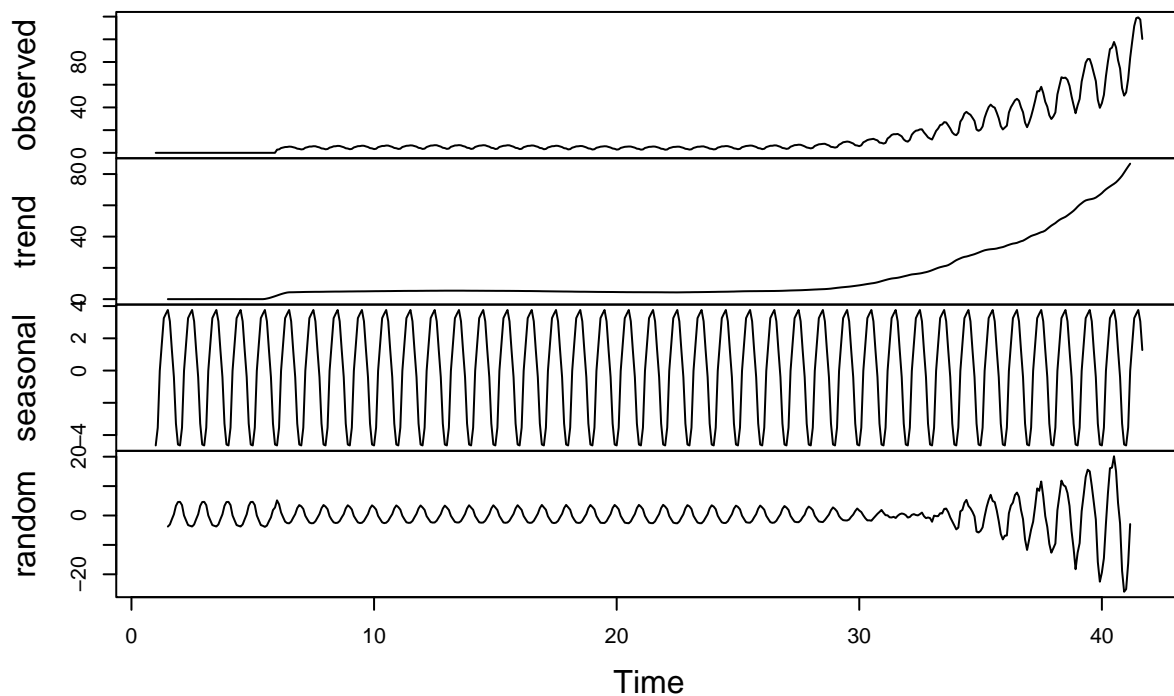
Transform wind and solar series into a time series object and apply the decompose function on them using the additive option, i.e., `decompose(ts_data, type = "additive")`. What can you say about the trend

component? What about the random component? Does the random component look random? Or does it appear to still have some seasonality on it?

Answer: First of all, the trend of solar (ts1) is upwarding. It seems like it still have some seasonality from 0 to 25 lag, while it looks random after that. The trend of wind (ts2) is also upwarding overall, while there is a slight drop at the end. Similar to the former one, it seems like it still have some seasonality from 0 to 27 lag, while it looks random after that.

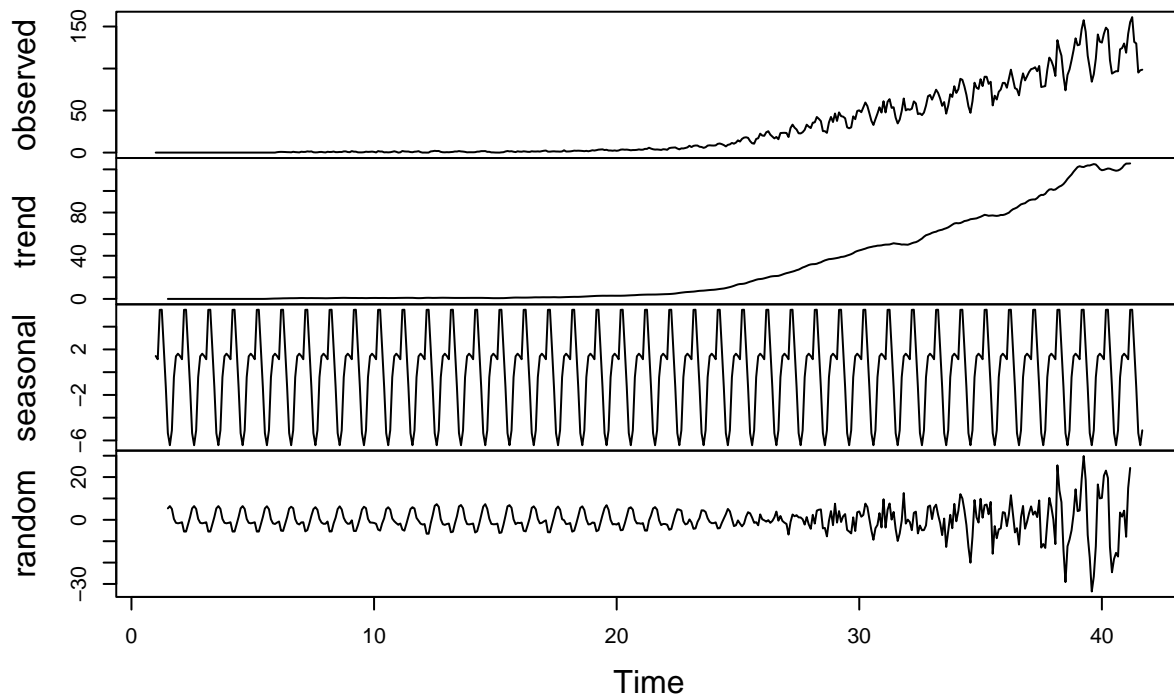
```
#Wind and Solar series - Already transformed into a time series object in Q1  
#Decomposition of Solar - Type: Additive  
ts1_renewable_data2_filtered_decompose<-decompose(ts1_renewable_data2_filtered,type="additive")  
plot(ts1_renewable_data2_filtered_decompose)
```

Decomposition of additive time series



```
#Decomposition of Wind - Type: Additive  
ts2_renewable_data2_filtered_decompose<-decompose(ts2_renewable_data2_filtered,type="additive")  
plot(ts2_renewable_data2_filtered_decompose)
```

Decomposition of additive time series



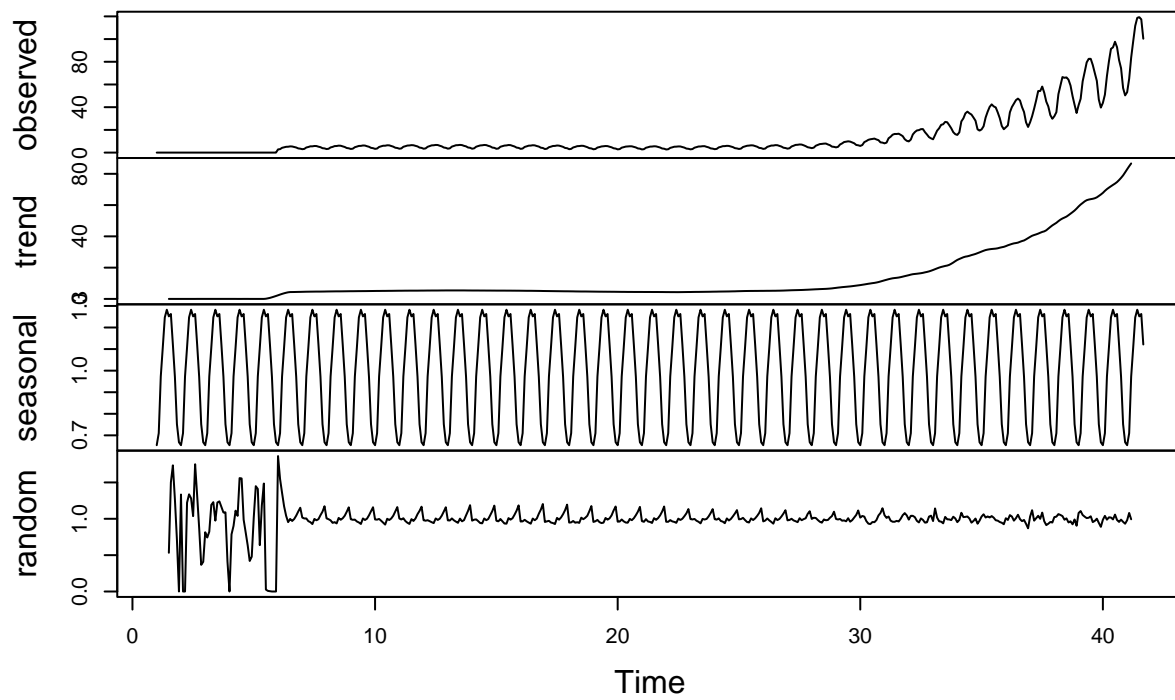
Q5

Use the `decompose` function again but now change the type of the seasonal component from additive to multiplicative. What happened to the random component this time?

Answer: The random component of solar energy consumption looks random at first, but it seems to have some seasonality after that. The wind energy one shows more random pattern through 15 or 16 unit of time. After that, it also seems to have some seasonality.

```
#Decomposition of Solar - Multiplicative  
ts1_renewable_data2_filtered_decompose_M<-decompose(ts1_renewable_data2_filtered,type="multiplicative")  
plot(ts1_renewable_data2_filtered_decompose_M)
```

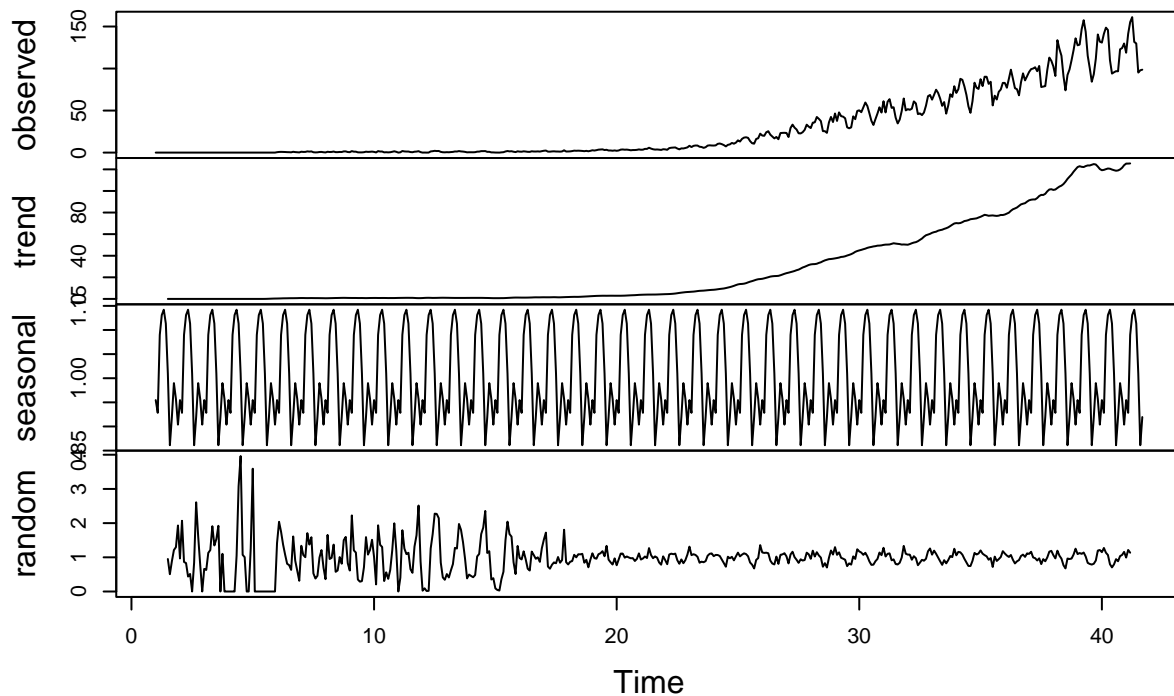
Decomposition of multiplicative time series



```
#Decomposition of Wind - Multiplicative
```

```
ts2_renewable_data2_filtered_decompose_M<-decompose(ts2_renewable_data2_filtered,type="multiplicative")
plot(ts2_renewable_data2_filtered_decompose_M)
```

Decomposition of multiplicative time series



Q6

When fitting a model to this data, do you think you need all the historical data? Think about the data from 90s and early 20s. Are there any information from those years we might need to forecast the next six months of Solar and/or Wind consumption. Explain your response.

Answer: I don't think we need all the historical data, seeing that the trend has significantly changed in more recent years for both Solar and Wind energy consumption.

Q7

Create a new time series object where historical data starts on January 2012. Hint: use `filter()` function so that you don't need to point to row numbers, i.e, `filter(xxxx, year(Date) >= 2012)`. Apply the `decompose` function `type=additive` to this new time series. Comment the results. Does the random component look random? Think about our discussion in class about seasonal components that depends on the level of the series.

```
#Converted to Date (Checked my code with AI)
renewable_data2_filtered$Month <- as.Date(paste0(renewable_data2_filtered$Month, "-01"), format = "%Y-%m-%d")
head(renewable_data2_filtered)
```

```
##           Month Solar Energy Consumption Wind Energy Consumption
## 1 1984-01-01                0.000                0.000
## 2 1984-02-01                0.000                0.001
```



```
## 3 1984-03-01          0.001          0.001
## 4 1984-04-01          0.001          0.002
## 5 1984-05-01          0.002          0.003
## 6 1984-06-01          0.003          0.002
```

```
#Filtered out the historical data starts on Jan 2012
```

```
renewable_data2_filtered_recent <- renewable_data2_filtered %>%
  filter(Month >= as.Date("2012-01-01"))
```

```
#1 Time Series object starts on Jan 2012 - Solar
```

```
ts1_renewable_data2_filtered_recent <- ts(renewable_data2_filtered_recent$`Solar Energy Consumption`,
  start = c(2012, 1),
  frequency = 12)
```

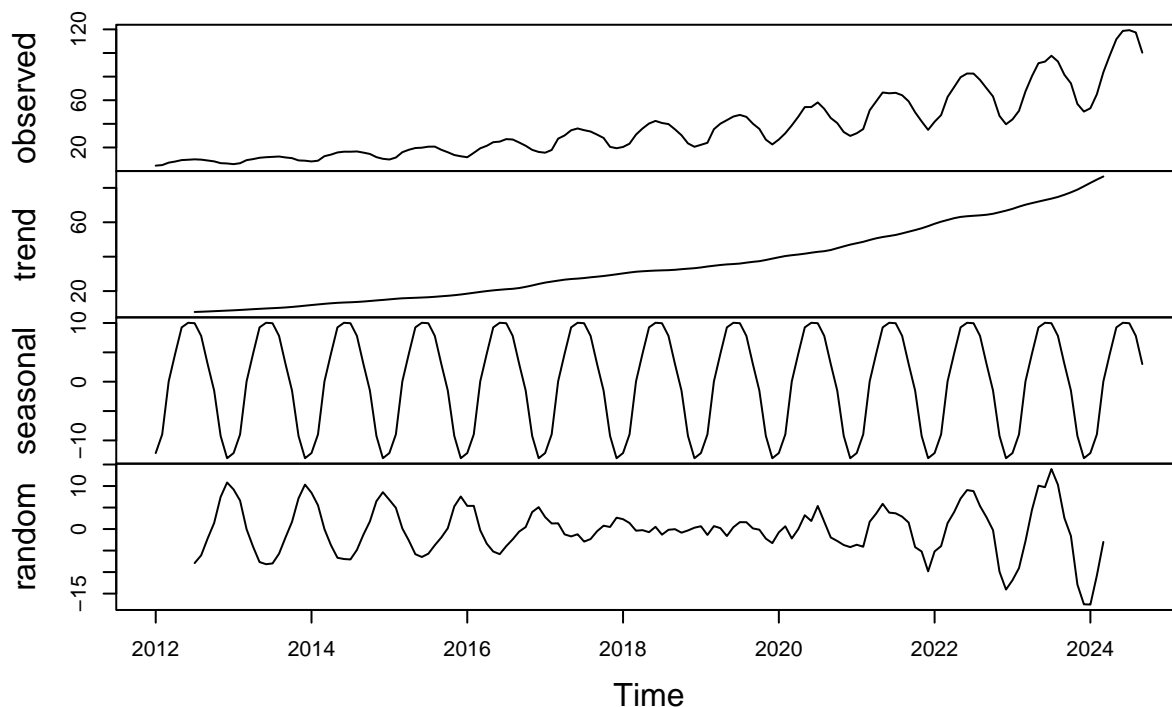
```
#2 Time Series object starts on Jan 2012 - Wind
```

```
ts2_renewable_data2_filtered_recent <- ts(renewable_data2_filtered_recent$`Wind Energy Consumption`,
  start = c(2012, 1),
  frequency = 12)
```

```
#1 Decomposition - Solar
```

```
ts1_renewable_data2_filtered_recent_decompose<-decompose(ts1_renewable_data2_filtered_recent,type="addi
plot(ts1_renewable_data2_filtered_recent_decompose)
```

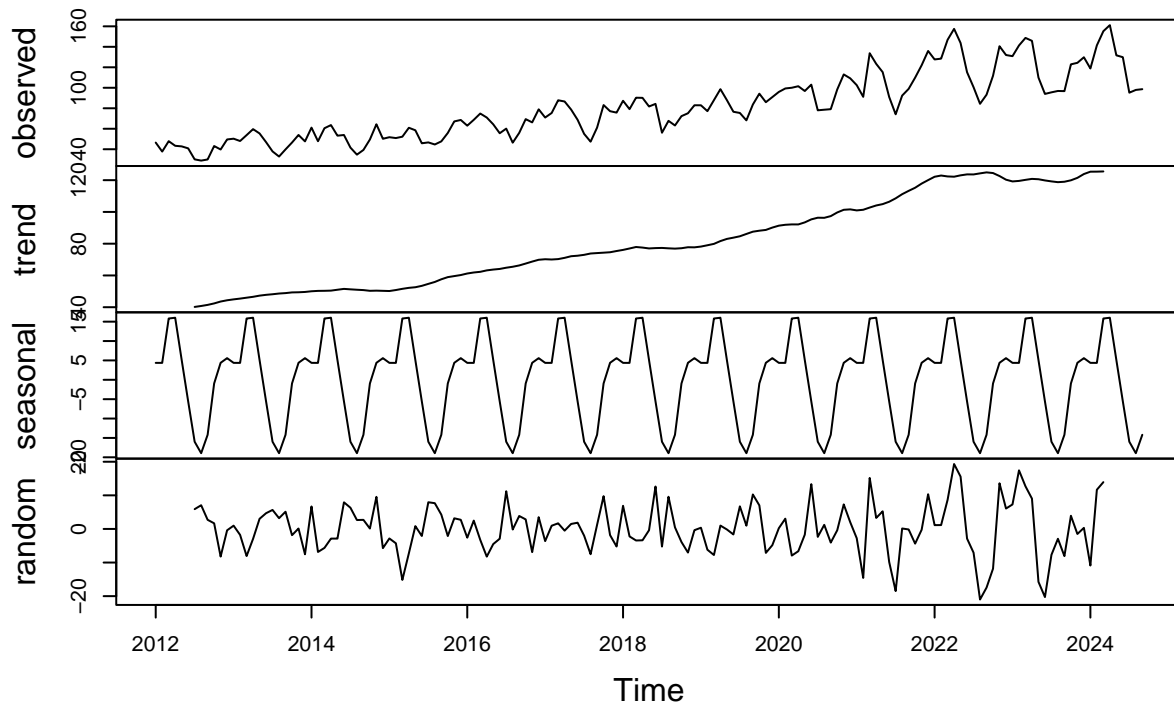
Decomposition of additive time series



```
#2 Decomposition - Wind
```

```
ts2_renewable_data2_filtered_recent_decompose<-decompose(ts2_renewable_data2_filtered_recent,type="addi
plot(ts2_renewable_data2_filtered_recent_decompose)
```

Decomposition of additive time series



Answer: According to the decomposed time series of Solar and Wind, the random component in more recent data appears to be more unpredictable compared to the results in Q4 and Q5. The overall trends continue to rise, which contrasts with the previous trend component that remained stable in the past before shifting upward after a certain period.

Identify and Remove outliers

Q8

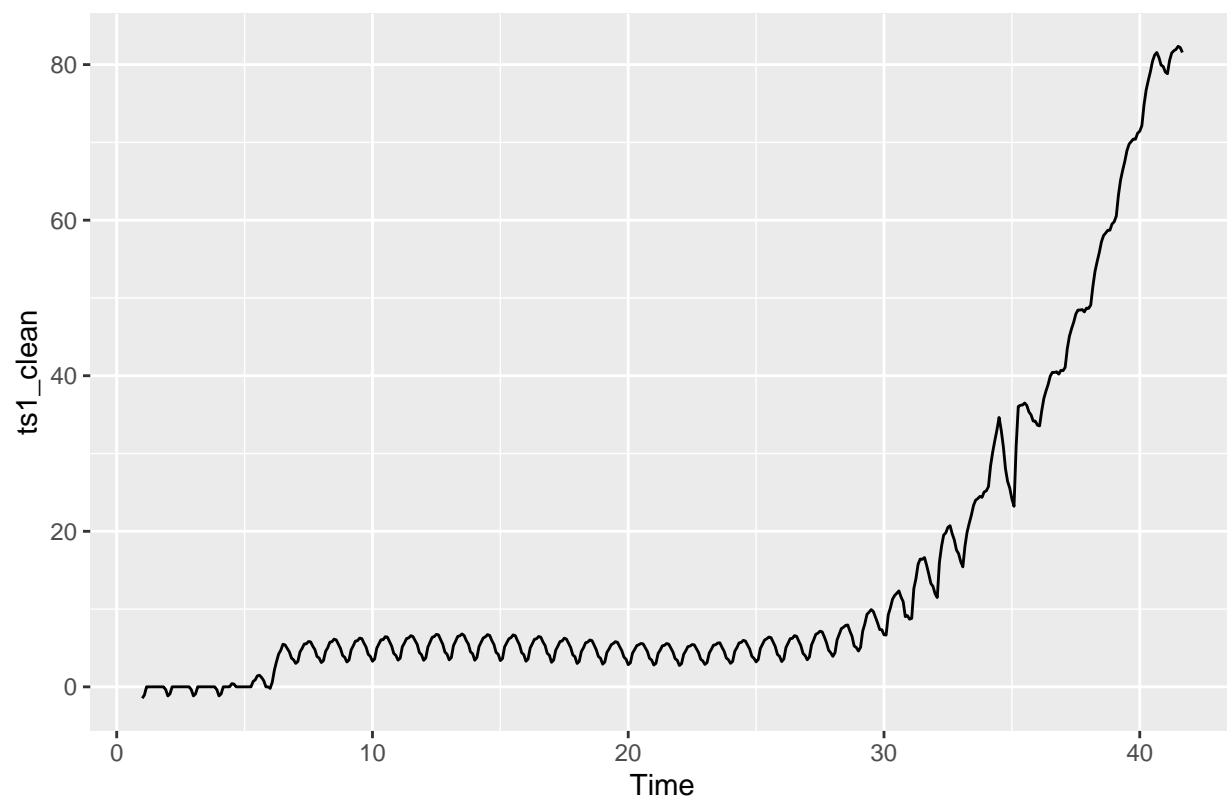
Apply the `tsclean()` to both series from Q7. Did the function removed any outliers from the series? Hint: Use `autoplot()` to check if there is difference between cleaned series and original series.

Answer: Yes, it seems like the outliers are removed from the series. Compared to the original plots, the cleaned ones are less fluctuating with fewer outliers.

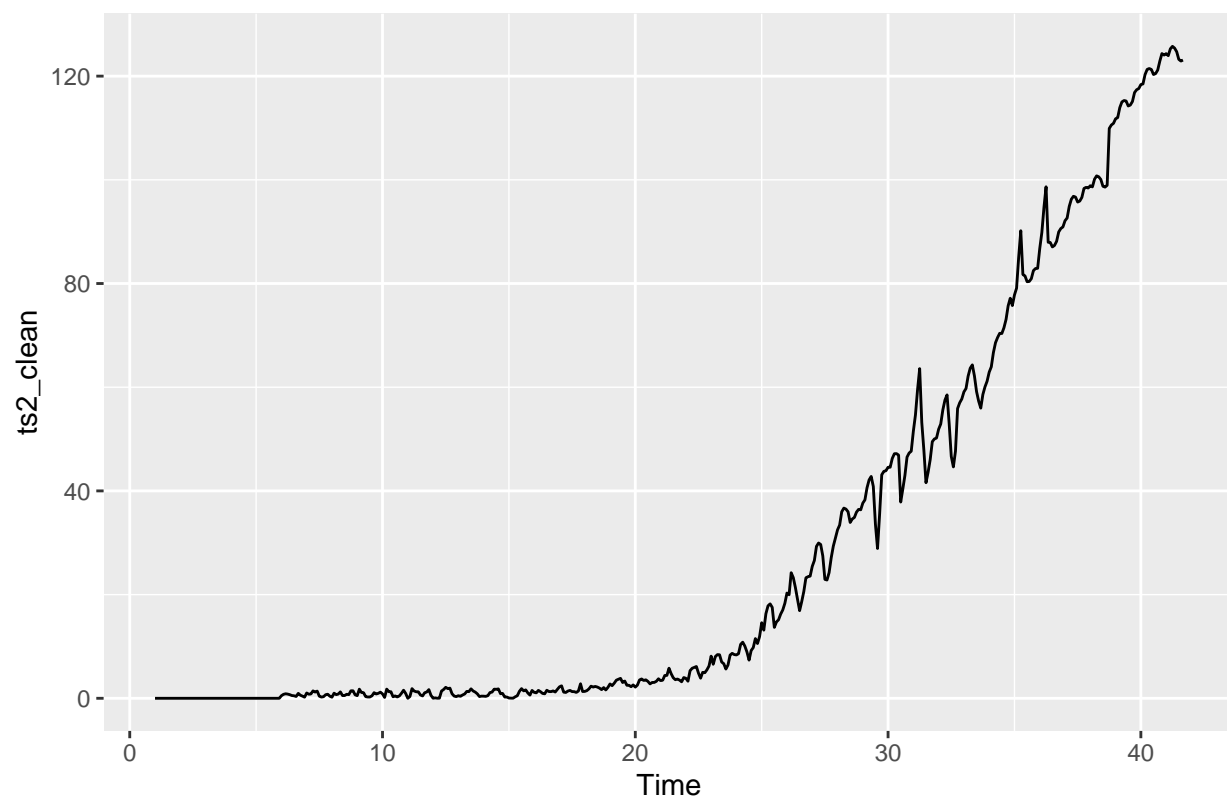
```
#1 Clean - Solar
ts1_clean<-tsclean(ts1_renewable_data2_filtered)

#2 Clean - Wind
ts2_clean <-tsclean(ts2_renewable_data2_filtered)

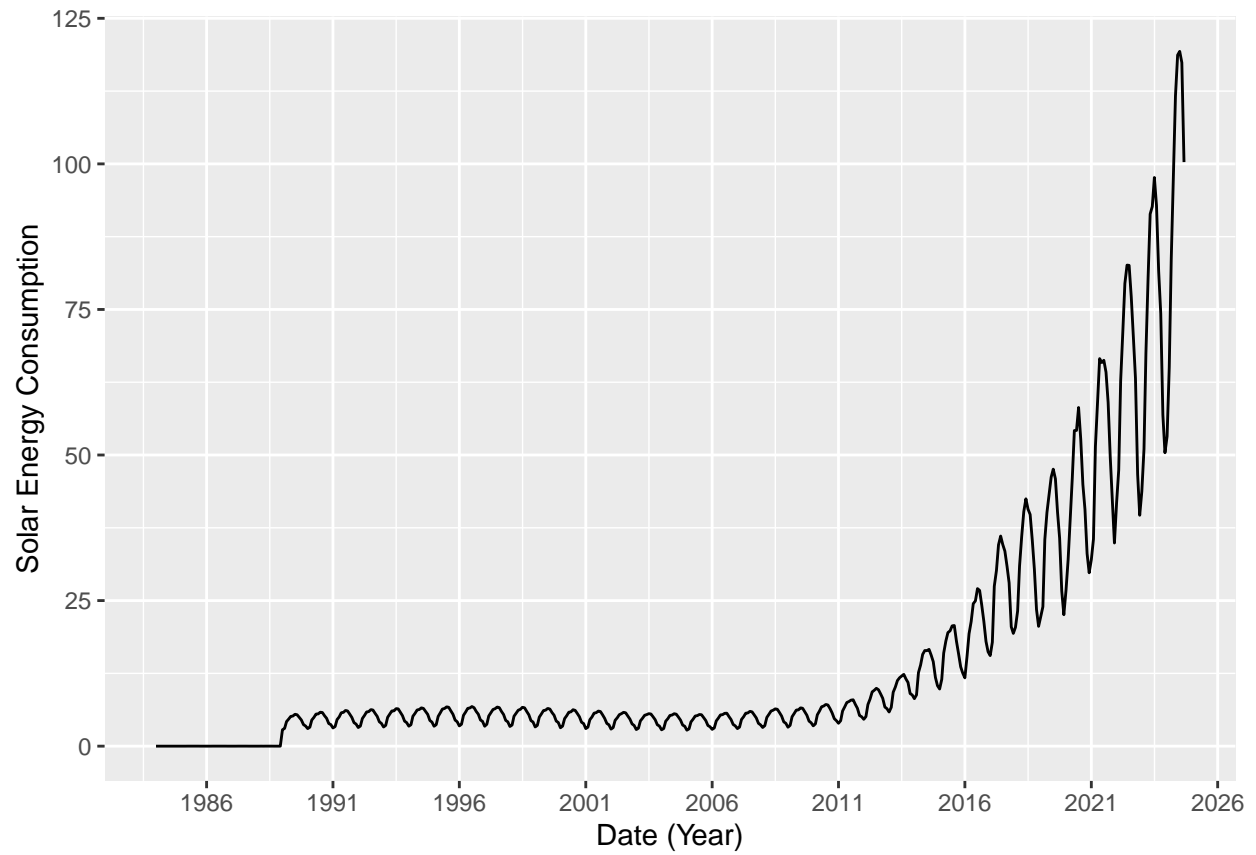
#Cleaned ver. autoplots
autoplot(ts1_clean)
```



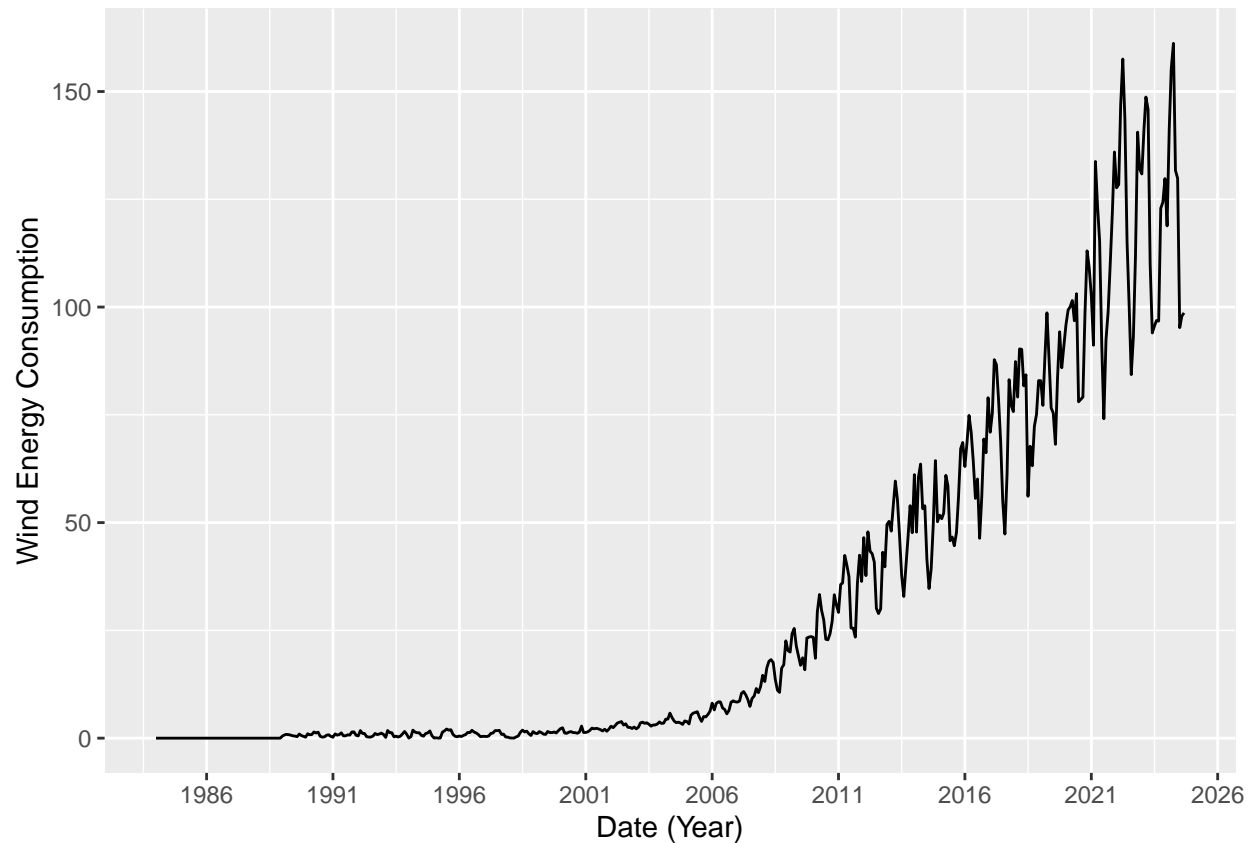
```
autoplot(ts2_clean)
```



```
#Original plot  
print(plot_ts1)
```



```
print(plot_ts2)
```



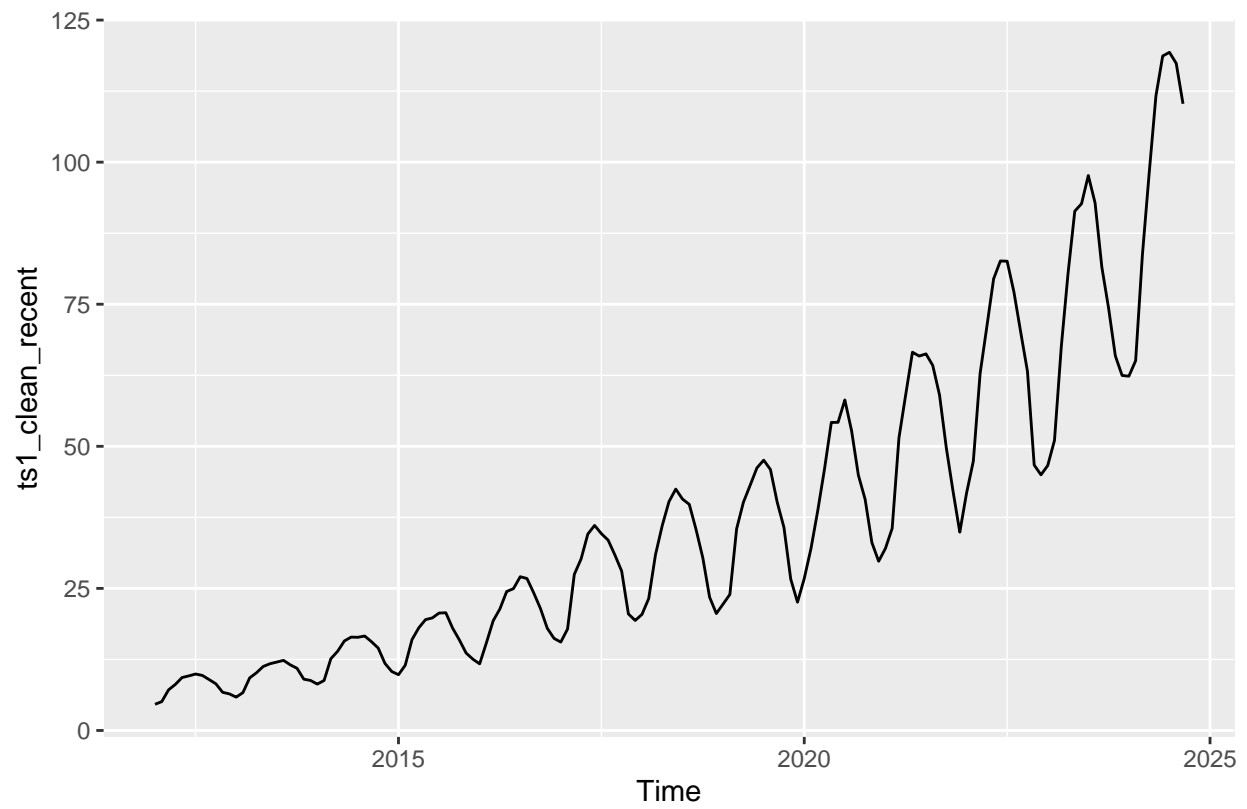
Q9

Redo number Q8 but now with the time series you created on Q7, i.e., the series starting in 2014. Using what `autoplot()` again what happened now? Did the function removed any outliers from the series?

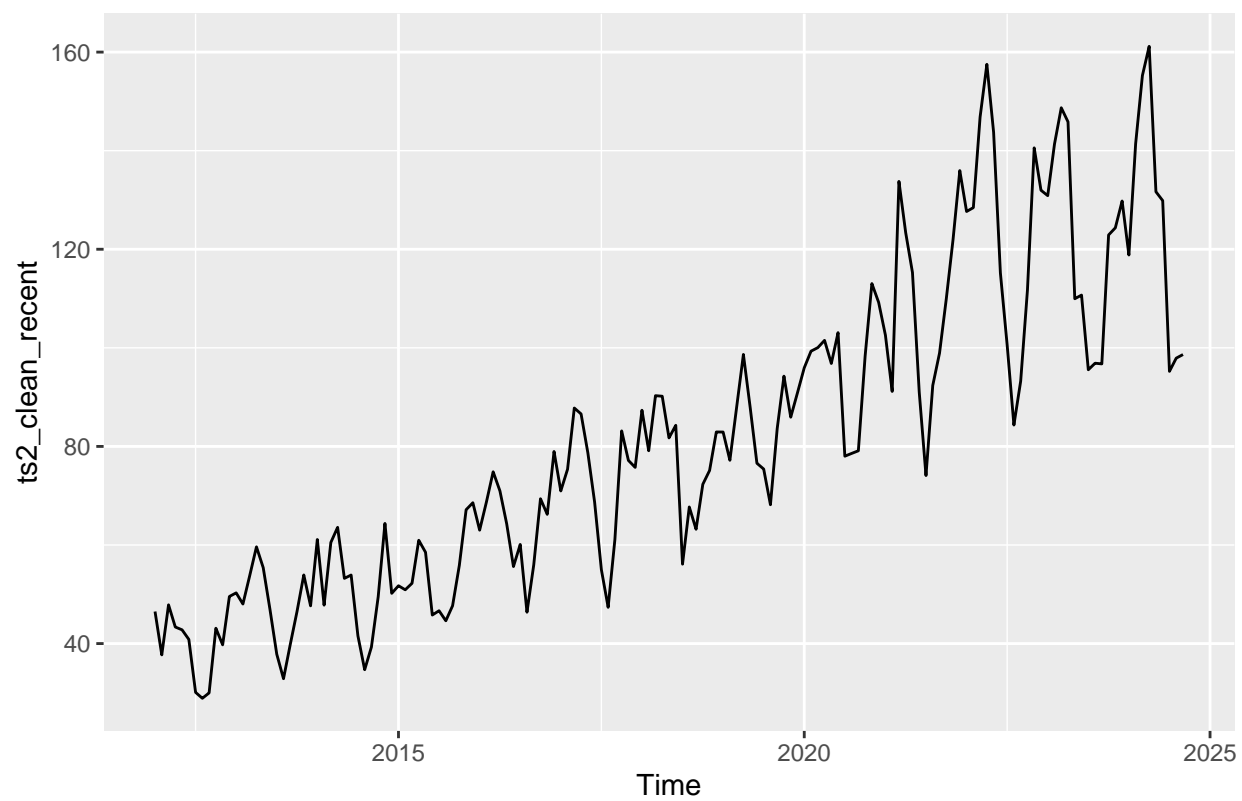
```
#1 Clean - Solar
ts1_clean_recent<-tsclean(ts1_renewable_data2_filtered_recent)

#2 Clean - Wind
ts2_clean_recent <-tsclean(ts2_renewable_data2_filtered_recent)

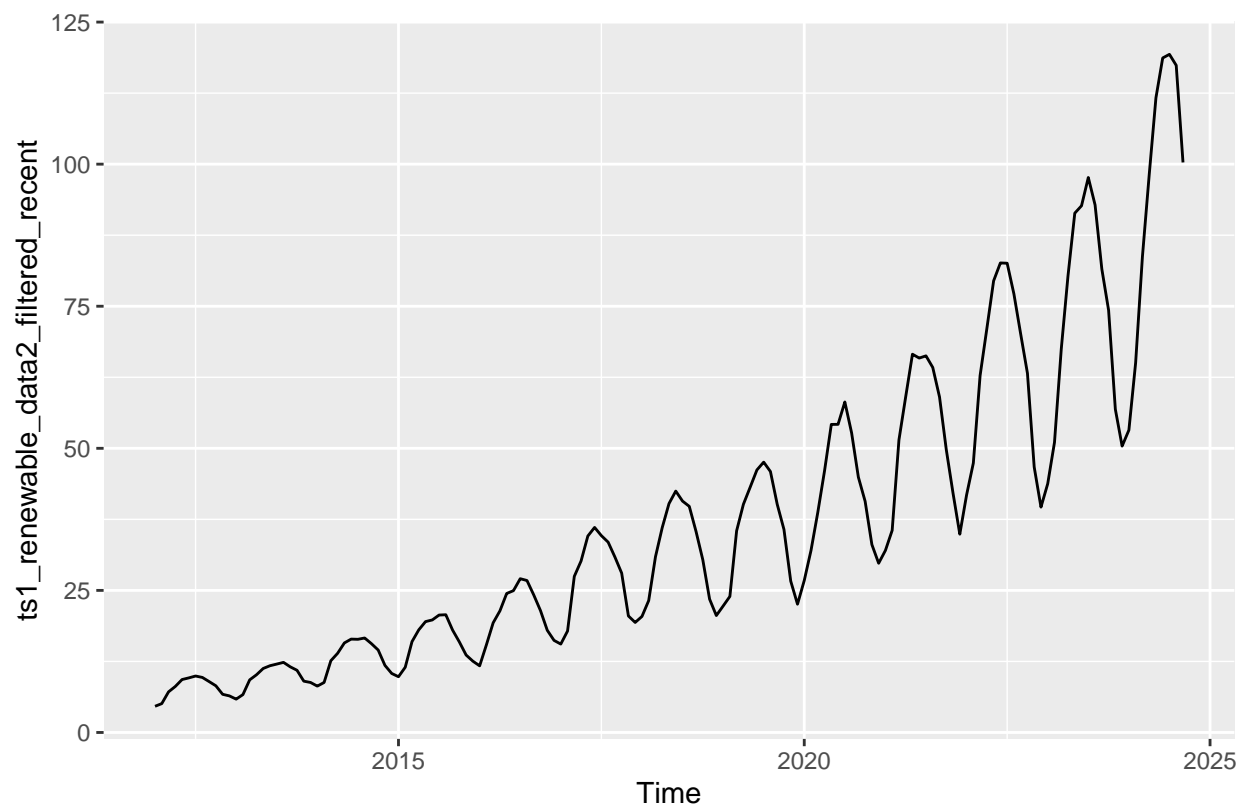
#Cleaned ver. autoplots
autoplot(ts1_clean_recent)
```



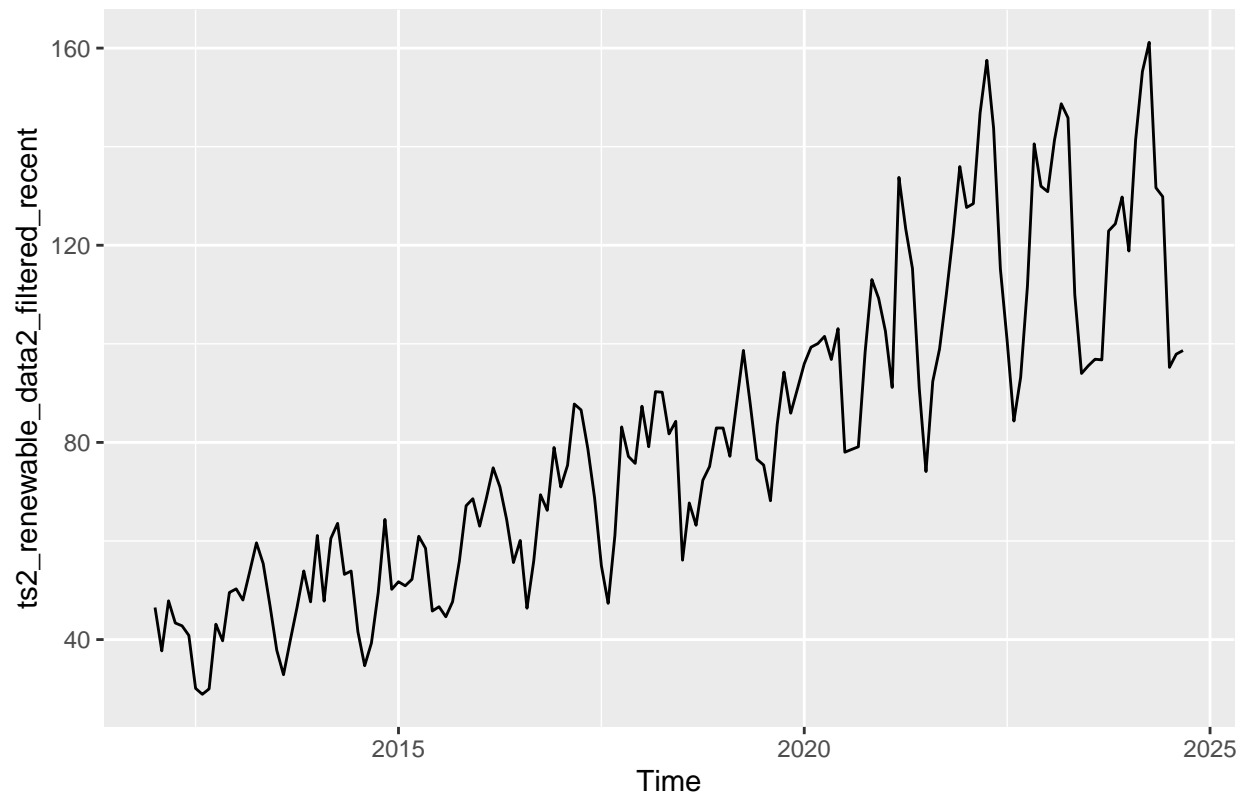
```
autoplot(ts2_clean_recent)
```



```
#Original autoplots  
autoplot(ts1_renewable_data2_filtered_recent)
```

```
autoplot(ts2_renewable_data2_filtered_recent)
```



Answer: Answer: Similar to Q8, the plots with `tsclean()` function tend to have fewer outliers than the original plots. Compared to the original plots, the cleaned plots are less fluctuating with fewer outliers.