

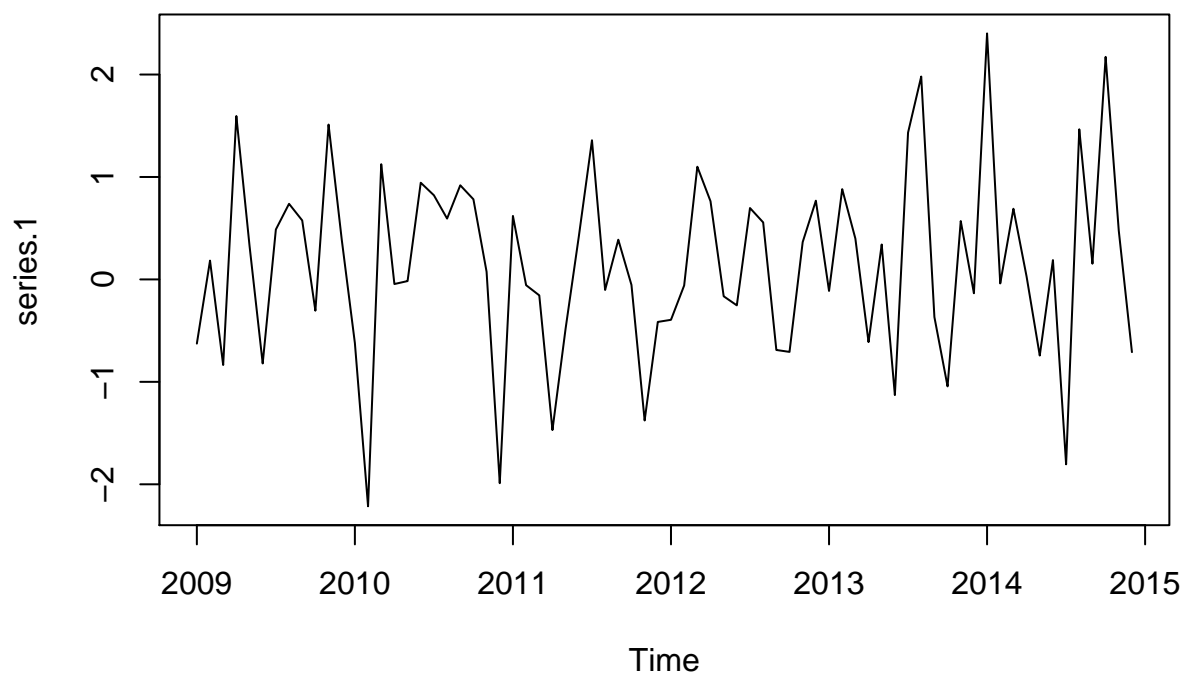
Simulations of Time Series Data

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Generating Time Series Data in R

The function `ts()` can convert a numeric vector into a time series object in R. The syntax is `ts(vectorname, start=, end=, frequency=)`, where start/end are the first/last time points, and frequency is the number of observations per unit in time. `frequency=1` means yearly, `frequency=12` means monthly, etc. Let's generate a time series data set here:

```
set.seed(1)
#randomly generate vector of length 72
vector=rnorm(72,0,1)
#generate time series spanning 6 years
#6 years --> 72 months, frequency=12 puts time series in terms of months
series.1=ts(vector,start=c(2009,1),end=c(2014,12),frequency=12)
plot.ts(series.1)
```

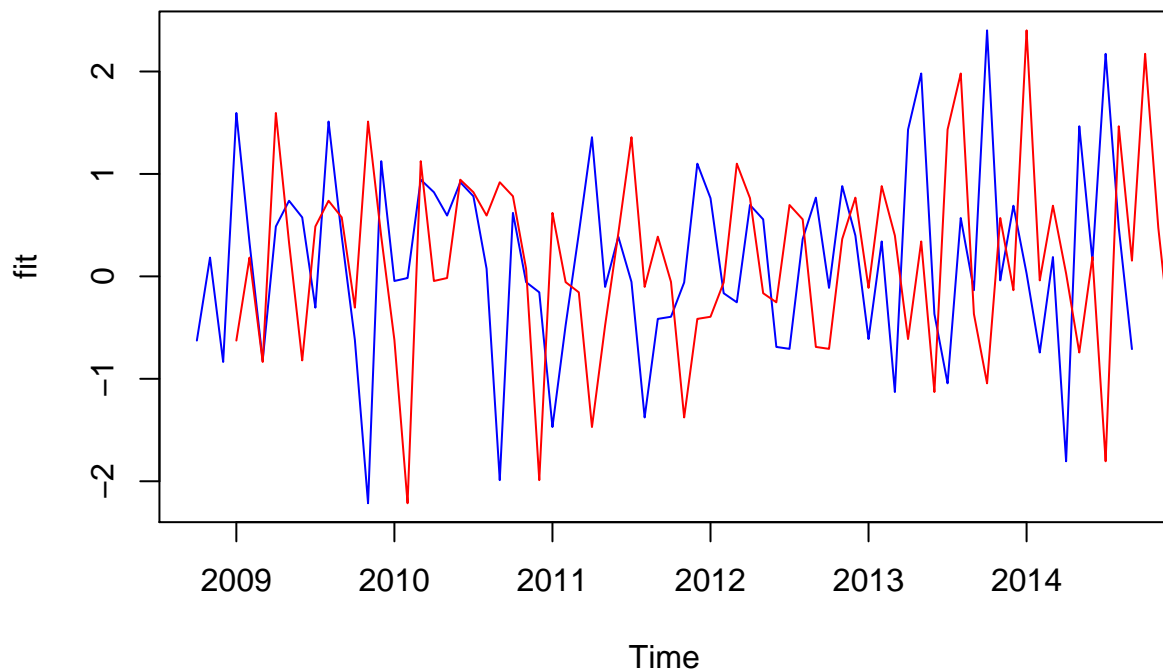


We see that the data revolves around the mean 0, and there is no random walk.

Lag

Lag causes a delay so that you can study how similar a time series is to itself. Lag is an important component of autocorrelation studies...yes, foreshadowing! I shift the time series 3 units and superimpose the two series on a plot. The red lines represent the delayed time series.

```
fit=lag(series.1,3)
plot.ts(fit,col="blue")
lines(series.1,col="red")
library(forecast)
```



Autocorrelation

Autocorrelation functions (**acf**) are useful for measuring the linear predictability of the series at time t (x_t) using only the variable (x_s). The function is given by:

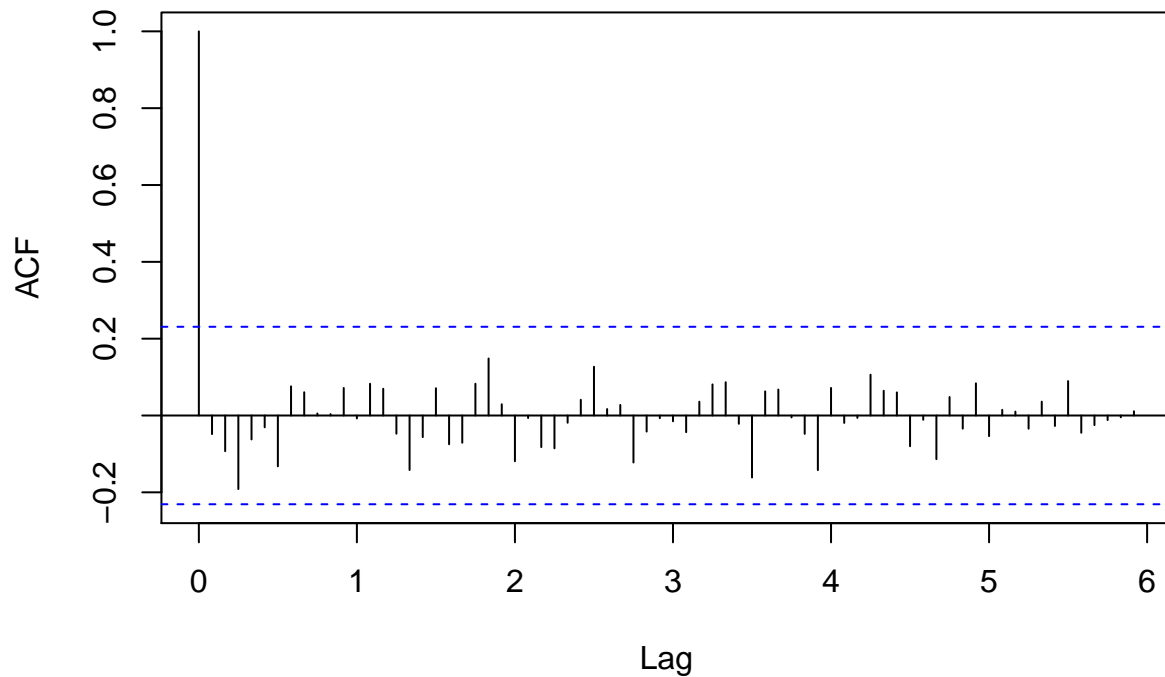
$$p(s, t) = \frac{\gamma(s, t)}{\sqrt{\gamma(s, s)\gamma(t, t)}}, \text{ where } \gamma(s, t) = \text{cov}(x_s, x_t) \text{ and } \gamma(t, t) = \text{cov}(x_t, x_t) = \text{var}(x_t). \text{ Thank you Math Stat.}$$

Another formulation is given measurements Y_1, Y_2, \dots, Y_N at time X_1, X_2, \dots, X_N , the lag k autocorrelation function is defined as:

$$r_k = \frac{\sum_{i=1}^{N-k} (Y_i - \bar{Y})(Y_{i+k} - \bar{Y})}{\sum_{i=1}^N (Y_i - \bar{Y})^2}$$

```
library(forecast)
#generate correlogram
acf=acf(series.1,500)
```

Series series.1



All correlograms start with an autocorrelation of 1; this is because when $t=0$, we are comparing the time series with itself. Periodicity is a good indicator of frequency in the time series data. For example, if each peak in a correlogram occurs when t is a multiple of 7, it is likely that the data is in terms of weeks and it's not just a coincidence.

When using the entirety of time series data, interpreting correlograms might not be easy. Partial autocorrelation functions (**pacf**) controls the values of the time series at shorter lags. This process removes the interference and resonance from multiple cycles and gives a more clear periodicity.

Great reads!

- <https://www.alanzucconi.com/2016/06/06/autocorrelation-function/#part2>
- <https://www.itl.nist.gov/div898/handbook/eda/section3/eda35c.htm>