

STAT 436 / 536 - Lecture 5

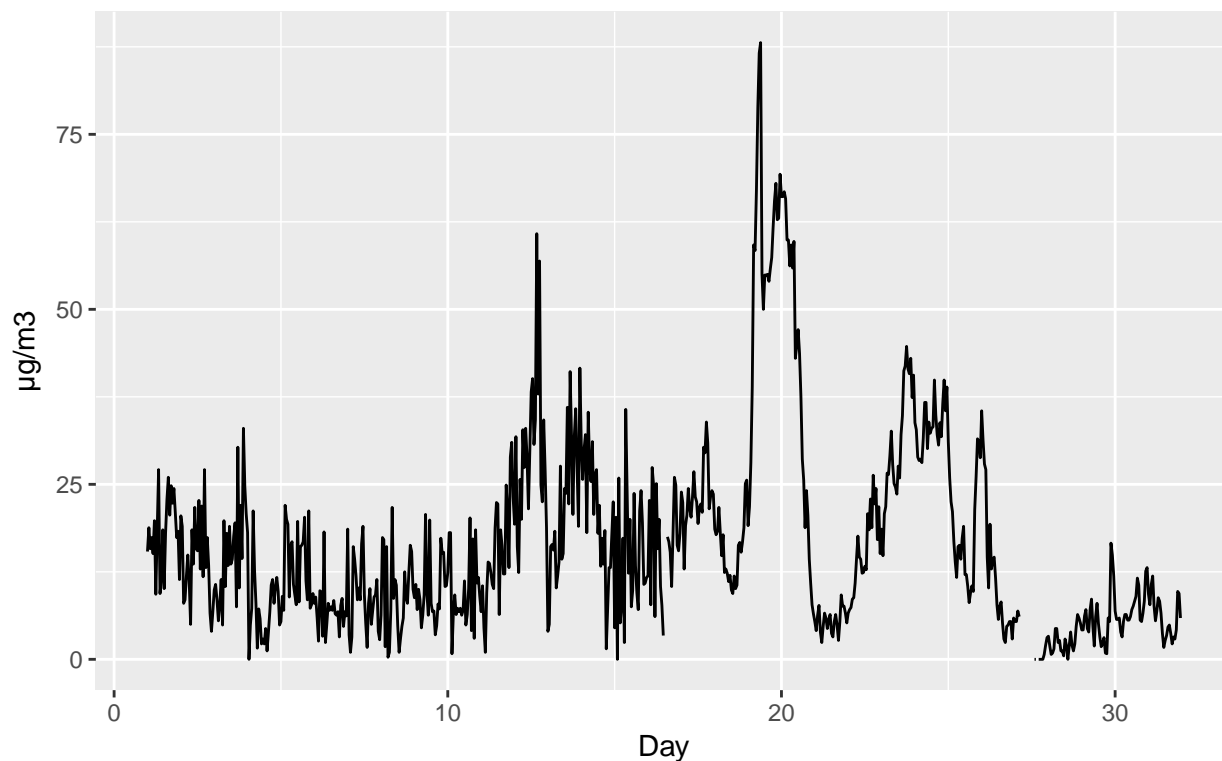
September 12, 2018

Forecasting Strategies

Consider an hourly time series of particulate matter less than 2.5 micrometers from Bozeman, MT in August of 2018.

PM2.5 Measurements in Bozeman, MT for August 2018

Source: MT DEQ



- What information would be useful to predict the PM2.5 measurements on September 1? How about the PM2.5 measurements for tomorrow?

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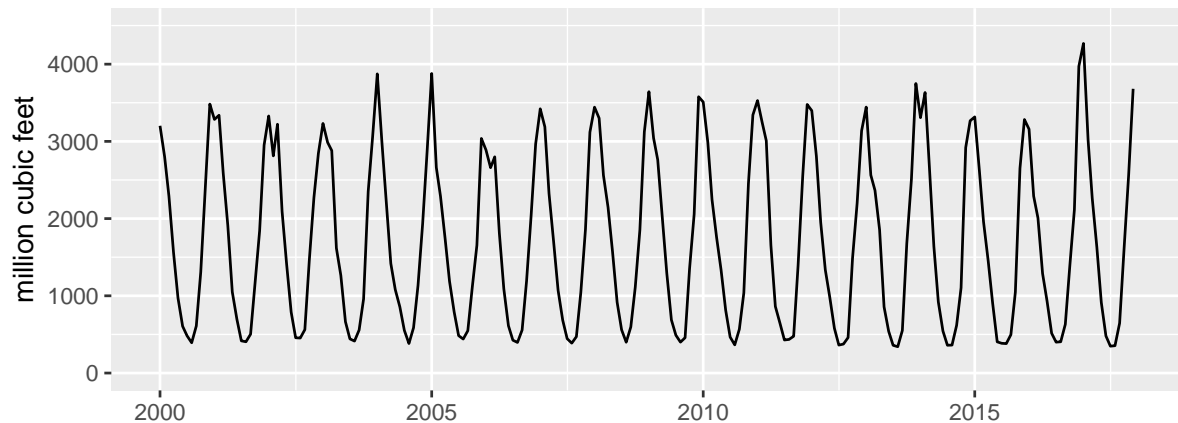
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Leading Variables and Associated Variables

- In some situations another variable may be useful informing our forecasts. A common example of this is in the housing industry where building approvals would be a *leading indicator* of building activity.
- Consider natural gas consumption and heating degree days (HDD).
- What do you assume is the relationship between the heating degree days during a given month and the reported natural gas consumption?

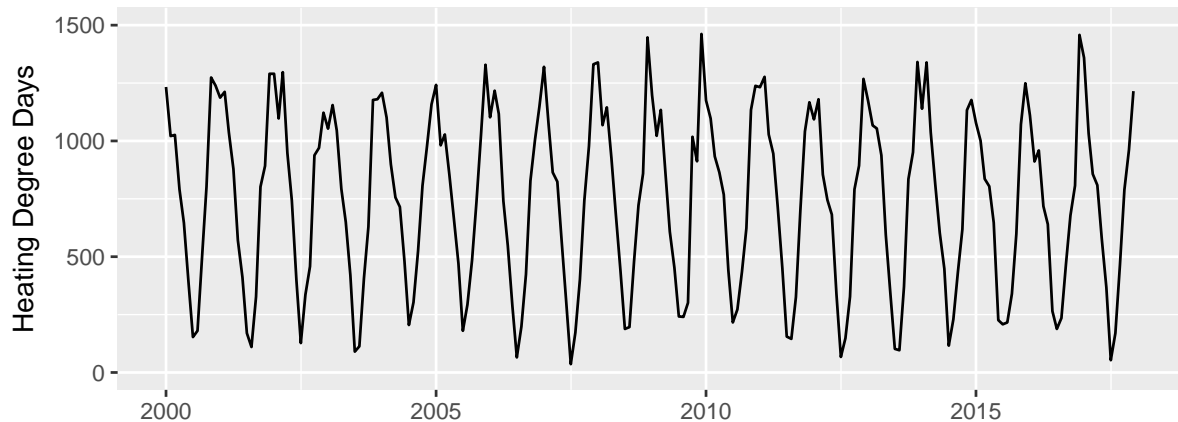
Monthly Natural Gas Consumption in Montana

Source: US DOE



Monthly Heating Degree Days at Sacajawea School, Bozeman, MT

Source: NOAA



Cross-Correlation

- Suppose two times series models, say x and y , are stationary with a constant mean and variance. The variables, x and y , may be serially correlated, perhaps with different time lags.
- The cross covariance function can be written as:

- Q: Is the cross-correlation function symmetric? In other words, does $\gamma_k(x, y) = \gamma_k(y, x)$?
- The sample cross-correlation function (ccf) can be written as
- Similarly to single variable autocorrelation calculations, sample based computations can be made:
 - Sample cross covariance:
 - Sample cross correlation:
- Note the cross correlation and cross covariance functions are defined on the de-seasonalized and de-trended time series, so it will not capture similar patterns in those structures. See HW2 Q1 for a demonstration.

```
acf(ts.union(ts.weather, ts.consumption))
```

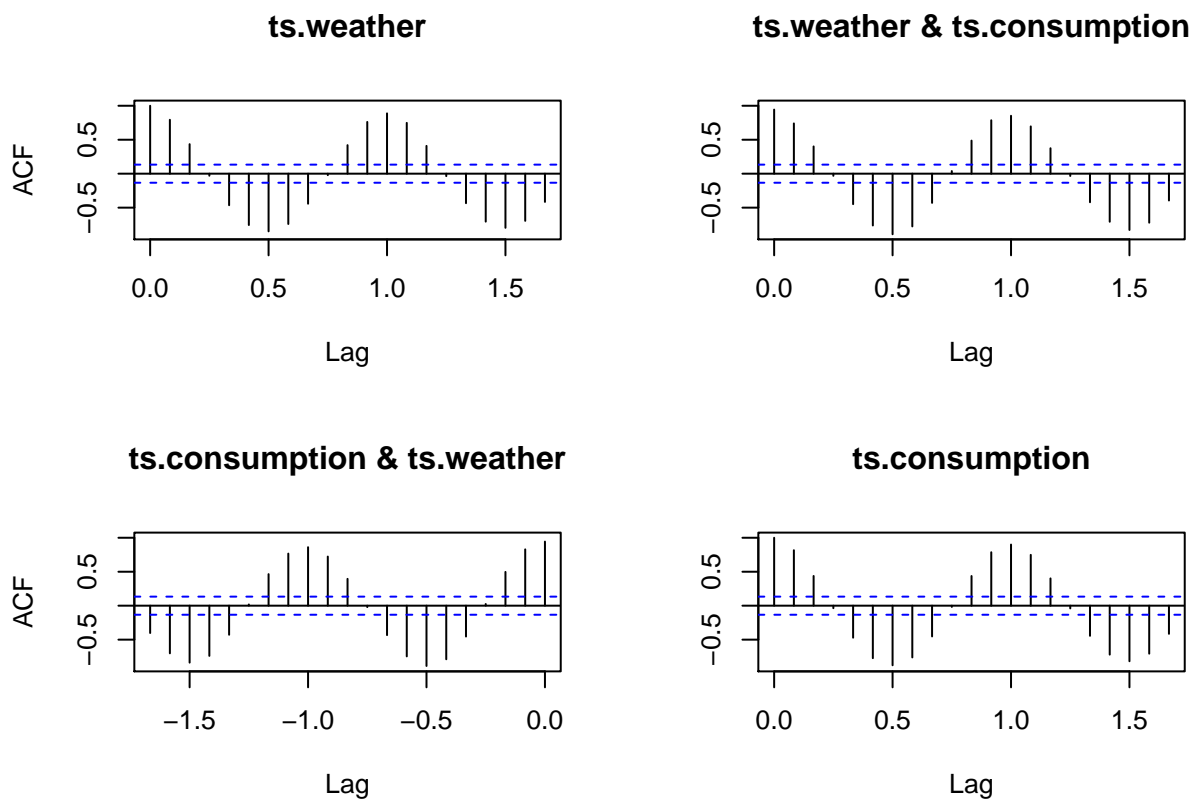


Figure 1: No Decomposition of the Time Series

```
hdd.random <- decompose(ts.weather)$random %>% window(start=c(2001,1), end = c(2016,12))
gas.random <- decompose(ts.consumption)$random %>% window(start=c(2001,1), end = c(2016,12))
acf(ts.union(hdd.random, gas.random))
```

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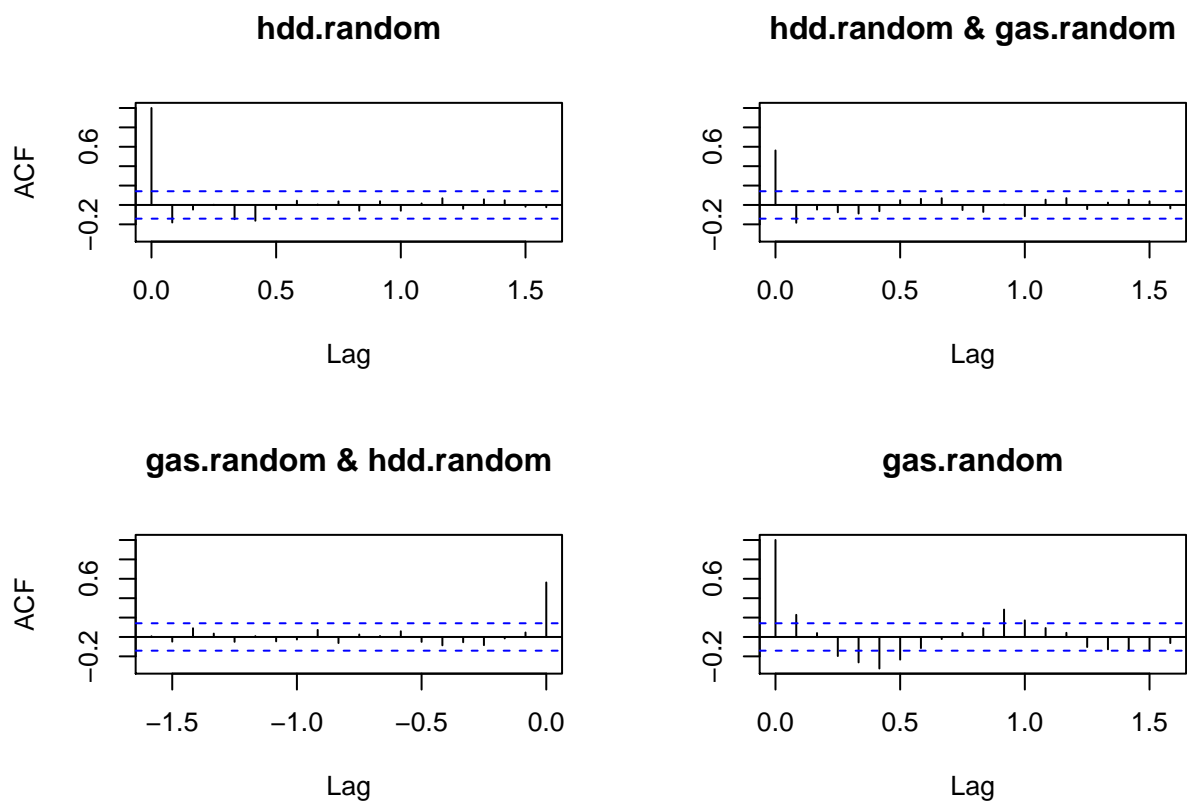


Figure 2: Using the decomposed random component