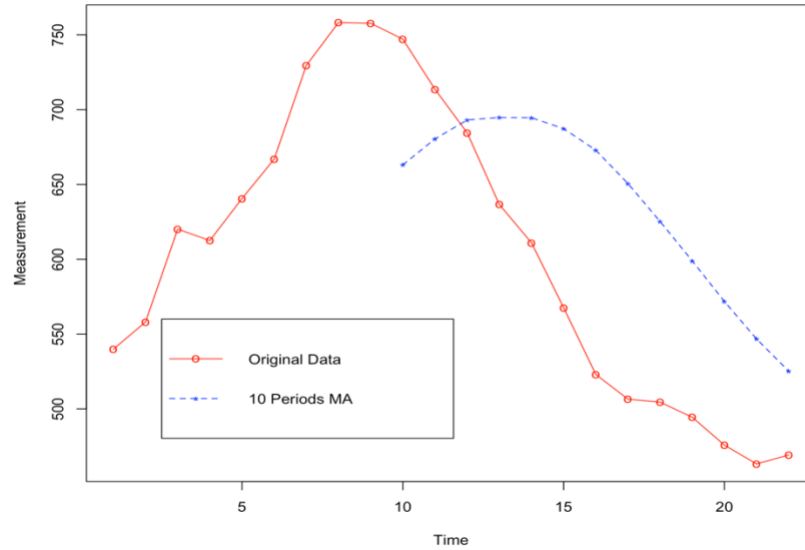


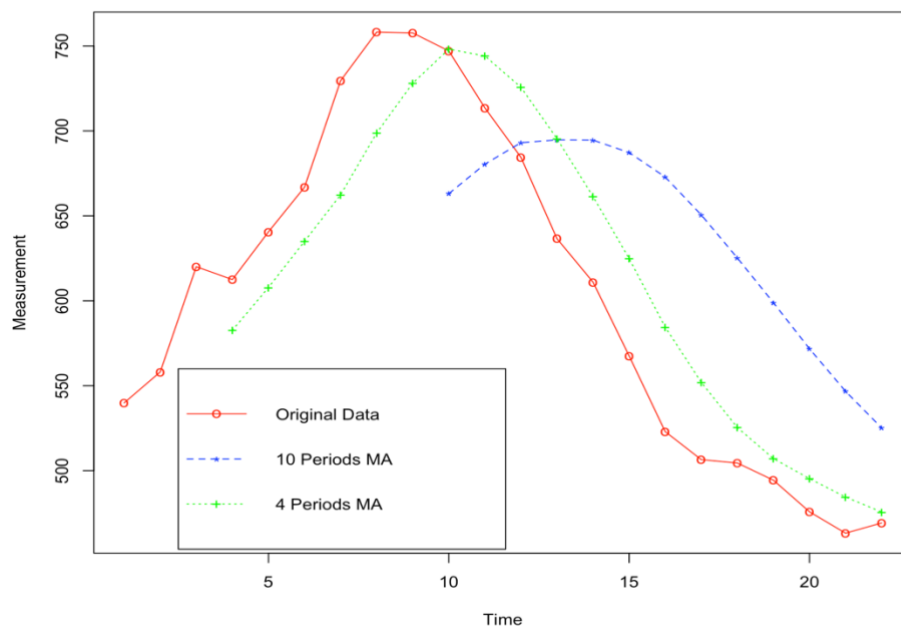
## DSC 275/475: Time Series Analysis and Forecasting (Fall 2019)

### HW #2 Chunlei Zhou

1. Consider the measurement data provided in the file: *Measurement\_Q1.xls*.
  - a. Use a 10-period simple moving average to smooth the data. Plot both the smoothed data and the original data values on the same axes.



- b. Repeat the procedure with a 4-period simple moving average. Overlay this result on the same plot above.



**c. What is the effect of changing the span of the simple moving average?**

Changing the span of the simple moving average from 10 to 4 makes the smoothed data captured the trend of the original data better and obtained more available smoothed data.

**2. FOR GRADUATE STUDENTS ONLY**

- a. Consider the N-span simple moving average applied to data that are uncorrelated with mean  $\mu$  and variance  $\sigma^2$ . Show that the variance of the weighted moving average is  $Var(M_t) = \sigma^2 / N$ .**

The N-span simple moving average  $M_t$  can be denoted as follow:

$$\begin{aligned} M_t &= \frac{y_t + y_{T-1} + \cdots + y_{T-N+1}}{N} \\ &= \frac{1}{N} \sum_{t=T-N+1}^T y_t \end{aligned}$$

Thus, the variance of  $M_t$  can be denoted as:

$$\begin{aligned} Var(M_t) &= Var\left(\frac{1}{N} \sum_{t=T-N+1}^T y_t\right) \\ &= \frac{1}{N^2} \sum_{t=T-N+1}^T Var(y_t) \end{aligned}$$

Since the variance of an individual observation  $y_t$  is  $\sigma^2$ , then the variance of the moving average is:

$$\begin{aligned} Var(M_t) &= \frac{1}{N^2} \times N \times \sigma^2 \\ &= \frac{\sigma^2}{N} \end{aligned}$$

■

- b. Consider an N-span moving average where each observation is weighted by a constant, say,  $a_j > 0$ . Therefore, the weighted moving average at the end of period T is,  $M_T^W = \sum_{t=T-N+1}^{t=T} a_{T+1-t} y_t$ . The variance of the original time series  $y_t$  is  $\sigma^2$ . Show that  $Var(M_t) = \sigma^2 \sum_{j=1}^N a_j^2$ .

The N-span simple moving average  $M_T^W$  can be denoted as follow:

$$M_T^W = \sum_{t=T-N+1}^{t=T} a_{T+1-t} y_t$$

Thus, the variance of  $M_T^W$  can be denoted as:

$$\begin{aligned} Var(M_T^W) &= Var\left(\sum_{t=T-N+1}^{t=T} a_{T+1-t} y_t\right) \\ &= \sum_{t=T-N+1}^{t=T} a_{T+1-t}^2 Var(y_t) \end{aligned}$$

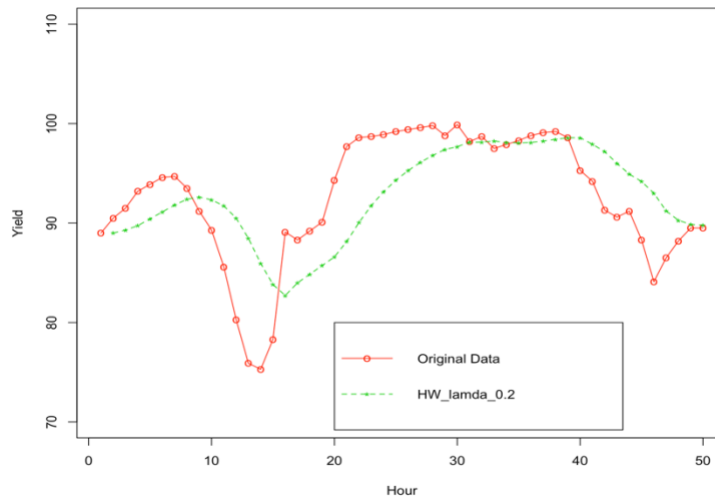
Since the variance of the original time series  $y_t$  is  $\sigma^2$ , we have:

$$\begin{aligned} Var(M_T^W) &= \sum_{j=1}^N a_j^2 Var(y_t) \\ &= \sigma^2 \sum_{j=1}^N a_j^2 \end{aligned}$$

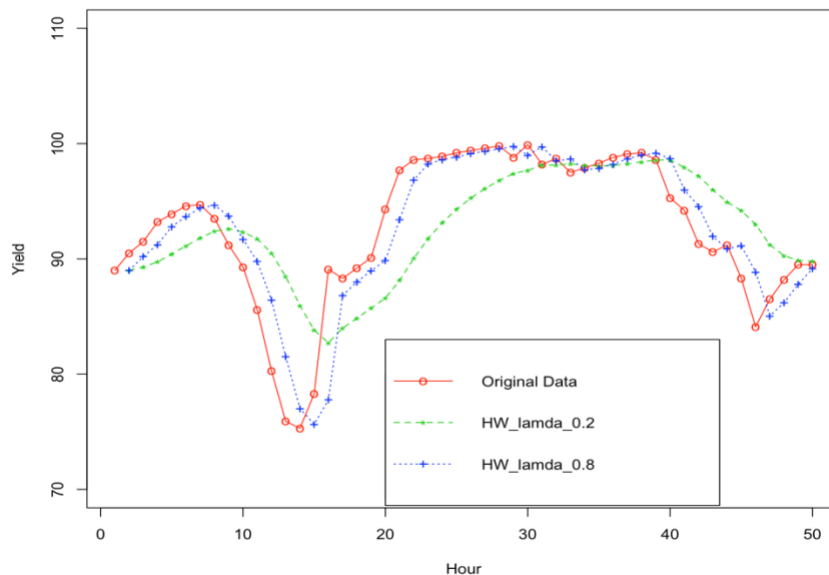
■

3. *Yield\_Data.xls* presents data on the hourly yield from a chemical process.

- a. Use simple (first order) exponential smoothing with  $\lambda = 0.2$  to smooth the data. Overlay the smoothed plot on the original time series.



- b. Change the smoothing constant ( $\lambda$ ) to  $\lambda = 0.8$ . Smooth the data with this new value of  $\lambda$ . Overlay the smoothed plot on the same plot made in (a).



- c. Compute the Mean Square Difference between the original data and the smoothed data for  $\lambda = 0.2$  and  $\lambda = 0.8$ .

For  $\lambda = 0.2$ , the Mean Square Difference is 15.98977.

For  $\lambda = 0.8$ , the Mean Square Difference is 0.3009854.