A Demonstration of an

Additive Time Series Decomposition

Based on the SAS program Decomposition.sas

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In traditional time series analysis it is often assumed that a time series y_t can be additively decomposed into four components, namely, trend, season, cycle, and irregular components as in

$$y_t = T_t + S_t + C_t + I_t \tag{1}$$

where T_t represents the trend in y_t at time t, S_t the seasonal effect at time t, C_t the cyclical effect at time t and I_t the irregular effect at time t.

To demonstrate this decomposition, consider the following characterizations of trend, cycle, seasonal, and irregular components that have been encoded in a SAS program entitled Decomposition.sas that is available on the class website.

$$T_t = 100 + 4.0 * t \qquad \text{(deterministic trend: intercept} = 100, \text{ slope} = 4)$$

$$C_t = 50 * \cos(3.1416 * t/10) \quad \text{(deterministic cycle: amplitude} = 50,$$

$$\text{Period} = 20 \text{ months, phase} = 0)$$

$$S_t = \{ \text{fixed seasonal effects: -50, -25, 25, -25, -50, 50, 75, 50, 5, -25, -50, } 20 \}$$

$$\text{(i.e. Jan. effect} = -50, \text{Feb. effect} = -25, \dots, \text{Dec. effect} = 20 \}$$

$$I_t \rightarrow NIID(0,100)$$

Therefore, the actual population model can be written as

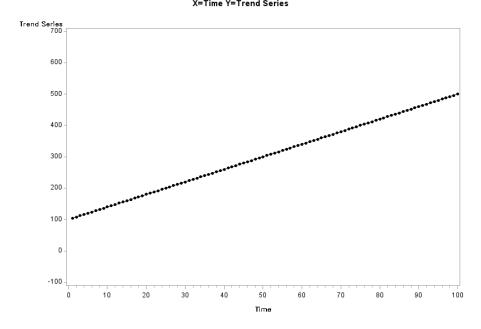
$$\begin{split} y_t &= T_t + C_t + S_t + I_t \\ &= 100 + 4.0t + 50\cos(0.31416t) - 50* \ gamma1 - 25* \ gamma2 \\ &+ 25* \ gamma3 - 25* \ gamma4 - 50* \ gamma5 + 50* \ gamma6 \\ &+ 75* \ gamma7 + 50* \ gamma8 + 5* \ gamma9 - 25* \ gamma10 \\ &- 50* \ gamma11 + 20* \ gamma12 + \varepsilon_t. \end{split}$$

The above seasonal dummies are defined gammai = 1 if the observation is in the i-th month and zero otherwise. Obviously the intercept of the month varies by month. For example, the intercept for all of the January months is (100 - 50 = 50), the intercept for the February months is (100 - 25 = 75), etc. The irregular component is represented by the unobserved error ε_t which is normally and independently distributed with mean zero and variance of 100. The time index t ranges from 1 for the first observation and t = T for the last observation.

Each of these components is plotted in order in the following figures below:

Figure 1

Deterministic Trend Data
X=Time Y=Trend Series



 $Figure\ 2$ Cycle with a=50, w = 2pi/20 (period = 20 months), theta = 0 $_{\text{X=Time Y=Cyclical Series}}$

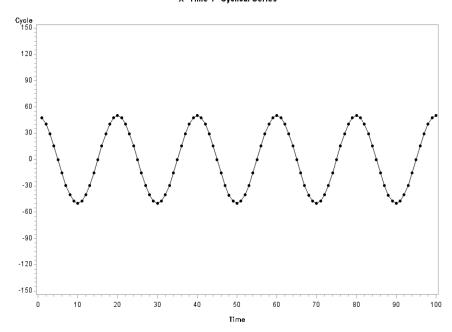


Figure 3

Seasonal Effects by Month X=Time Y=Seasonal Effects

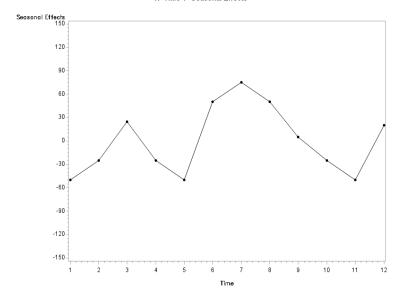
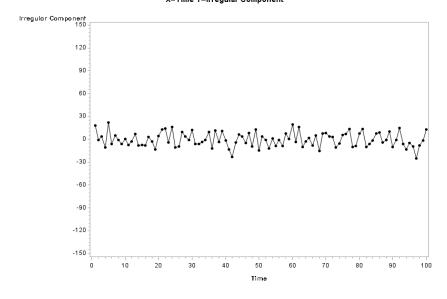


Figure 4

Irregular Component X=Time Y=Irregular Component



In the following graphs we sum these components up as is intended in the additive decomposition (1).

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Figure 5

Deterministic Trend + Cycle X=Time Y=Trend + Cycle

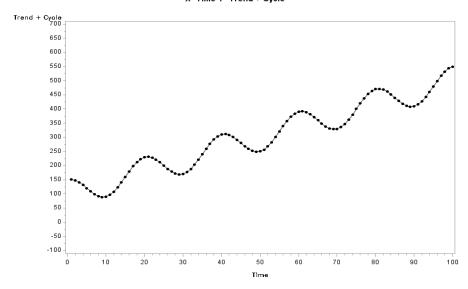


Figure 6

Deterministic Trend + Cycle + Seasonal X=Time Y=Trend + Cycle + Seasonal

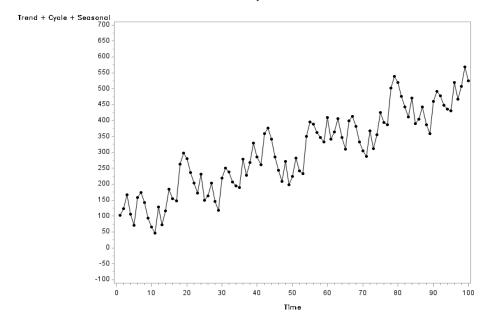
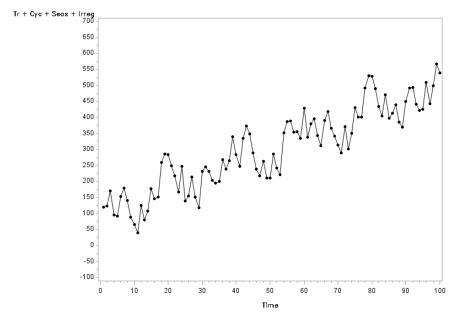


Figure 7

Trend + Cycle + Seasonal + Irregular

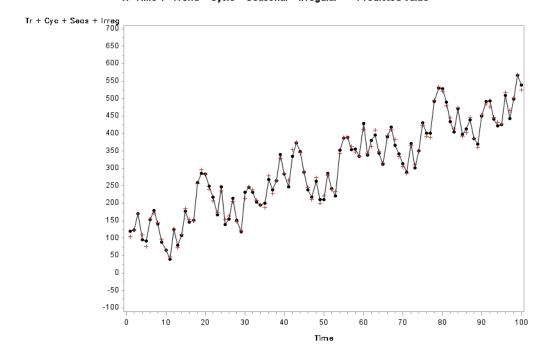
X=Time Y=Trend + Cycle + Seasonal + Irregular



And in Figure 8 below we have the representation of the fitted model obtained by Proc Nlin (a Nonlinear Least Squares procedure).

Figure 8

Fitted Values of Classical Decomposition Model
X=Time Y=Trend + Cycle + Seasonal + Irregular + = Predicted Value



The fitted model is

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\begin{aligned} y_t &= 51.74 + 3.98t + 50.76\cos(0.3140t - 0.0314) + 21.36* \ gamma2 + 75.21* \ gamma3 \\ &+ 23.99* \ gamma4 + 2.21* \ gamma5 + 93.26* \ gamma6 + 120.3* \ gamma7 \\ &+ 100.6* \ gamma8 + 56.16* \ gamma9 + 24.69* \ gamma10 - 0.58* \ gamma11 \\ &+ 69.97* \ gamma12 + \hat{\varepsilon}_t \end{aligned}
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which closely matches the population parameters. Obviously, the January intercept is estimated as 51.74, the February intercept is 51.74 + 21.36 = 73.1, etc. The slope of the trend is 3.98 while the amplitude of the cycle is 50.76, the phase is -0.0314, and the period is $p = 2\pi/w = 2*3.1416/0.3140 = 20.01$ months.