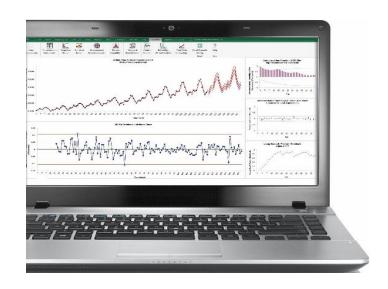


# Lean Six Sigma Statistical Tools, Templates & Monte Carlo Simulation in Excel

### What's New in SigmaXL® Version 9

#### Part 3 of 3: Control Charts for Autocorrelated Data



John Noguera CTO & Co-founder SigmaXL, Inc.

www.SigmaXL.com

Webinar December 10, 2020

### **Agenda**

- Introduction
- Autocorrelation
- Example 1: Chemical Process Concentration
- Simple Exponential Smoothing (EWMA)
- Example 2: Ln(Monthly Airline Passengers-Modified)

### **Agenda**

- Error, Trend, Seasonal (ETS) Exponential Smoothing models
- Autoregressive Integrated Moving Average (ARIMA) models
- ARIMA with Predictors
  - Analyze control chart outlier versus shift
  - Example 3: Electricity Demand with Temperature and Work Day Predictors
- Questions/References

#### Introduction

- Statistical process control for autocorrelated processes typically use the EWMA (Exponentially Weighted Moving Average) one-step-ahead forecast model.
- The time series model forecasts the motion in the mean and an Individuals control chart is plotted of the residuals to detect assignable causes.

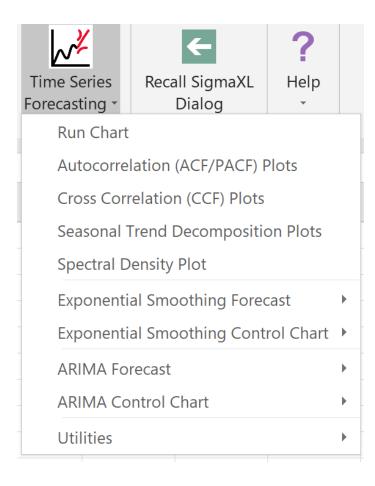
#### Introduction

- Failure to account for the autocorrelation will produce limits that are too narrow resulting in excessive false alarms, or limits that are too wide resulting in misses.
- The challenge with this approach is that if there is seasonality or negative autocorrelation in the data, the user needs an advanced level of knowledge in forecasting methods to pick the correct model, e.g., Seasonal Exponential Smoothing models or Seasonal Autoregressive Integrated Moving Average (ARIMA) models are required.

#### Introduction

- We will review simple exponential smoothing/EWMA, then introduce recent developments in time series forecasting that use automatic model selection to accurately pick the time series model that produces a minimum forecast error.
- An accurate forecast for your time series means the residuals will most often have the right properties to correctly apply a control chart, thus leading to an improved control chart with reduced false alarms and misses.

## SigmaXL Version 9 Time Series Forecasting Menu



#### **Autocorrelation**

- Just as correlation measures the extent of a linear relationship between two variables, autocorrelation (AC) measures the linear relationship between lagged values of data.
- A plot of the data vs. the same data at lag k will show a positive or negative trend. If the slope is positive, the AC is positive; if there is a negative slope, the AC is negative.
- The Autocorrelation Function (ACF) formula is:

$$r_k = \frac{\sum_{t=k+1}^{T} (y_t - \overline{y})(y_{t-k} - \overline{y})}{\sum_{t=1}^{T} (y_t - \overline{y})^2}$$

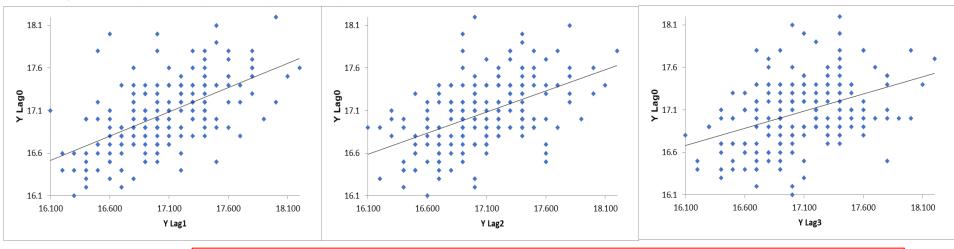
where T is length of the time series [4].

#### **Autocorrelation**

Y Lag0	Y Lag1	Y Lag2	Y Lag3
17			
16.6	17		
16.3	16.6	17	
16.1	16.3	16.6	17
17.1	16.1	16.3	16.6
16.9	17.1	16.1	16.3
16.8	16.9	17.1	16.1
17.4	16.8	16.9	17.1
17.1	17.4	16.8	16.9

Pearson Correlations	Y Lag1	Y Lag2	Y Lag3
Y Lag0	0.571	0.498	0.407

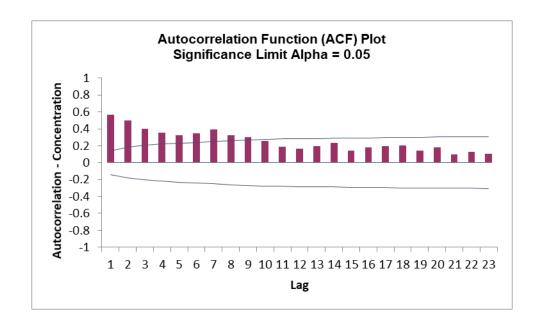
Pearson correlations are used here for demonstration purposes. They are approximately equal to the ACF correlation values.



Any statistically significant correlation  $(r_k > 2/\sqrt{N})$  will adversely affect the performance of a Shewhart control chart.

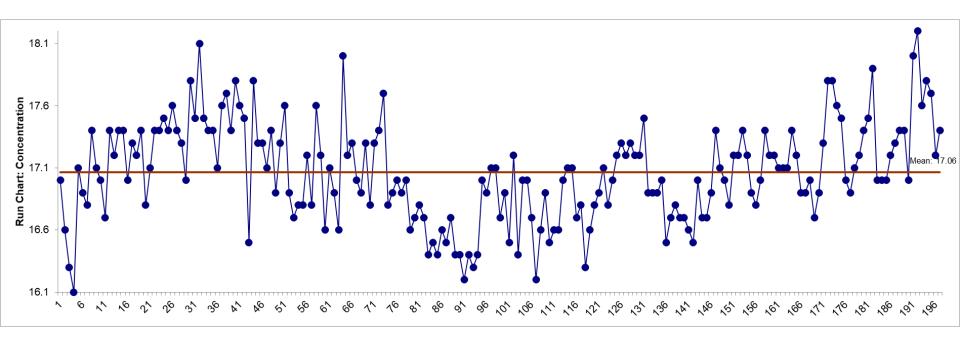
The Ljung-Box test is used to determine if a group of autocorrelations are significant (see formula in Appendix).

### Example 1: Box-Jenkins Series A - Chemical Process Concentration - Autocorrelation Function (ACF) Plot



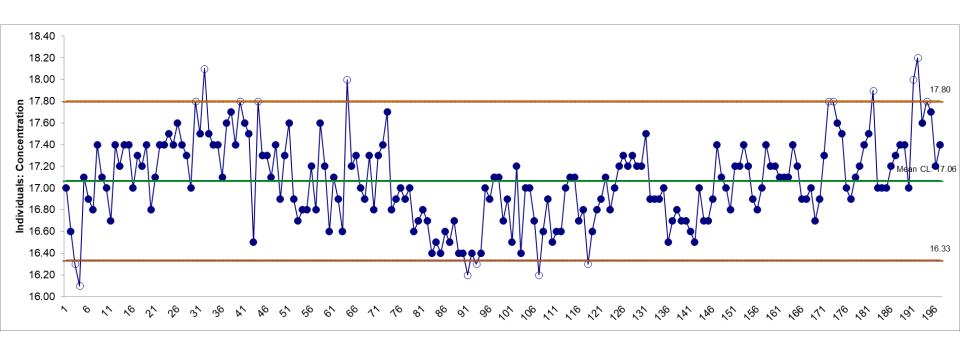
SigmaXL > Time Series Forecasting > Autocorrelation (ACF/PACF) Plots Example 1: Chemical Process Concentration - Series A.xlsx - Concentration

### Example 1: Box-Jenkins Series A - Chemical Process Concentration - Run Chart



SigmaXL > Time Series Forecasting > Run Chart

### **Example 1: Box-Jenkins Series A - Chemical Process Concentration - Individuals Control Chart**



17 out-of-control data points

SigmaXL > Control Charts > Individuals

#### **Autocorrelation**

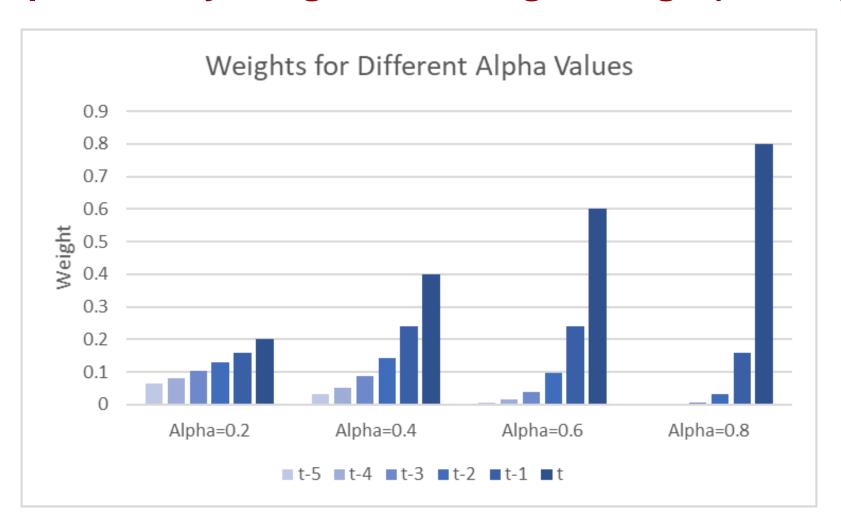
#### Guidelines from Woodall & Faltin [10]:

- If possible, one should first attempt to remove the source of the autocorrelation.
- If the source of autocorrelation cannot be removed directly, then it may be possible to model the autocorrelation and use a feedback control scheme to reduce variability about a specified target value.
- If the source of the autocorrelation cannot be removed directly, and feedback control is not a viable option, then it is important to monitor the process with control charts which do not repeatedly give signals due to presence of the autocorrelation.

Forecasts are calculated using weighted averages, where the weights decrease exponentially as observations come from further in the past with the smallest weights associated with the oldest observations:

$$\hat{y}_{t+1} = \alpha \ y_t + \alpha (1 - \alpha) \ y_{t-1} + \alpha (1 - \alpha)^2 \ y_{t-2} + \cdots$$

where  $0 \le \alpha \le 1$  is the level smoothing parameter [4].



 An equivalent formulation for simple exponential smoothing is:

$$\hat{y}_{t+1} = \alpha y_t + (1 - \alpha)\hat{y}_t$$

with the starting forecast value (initial level)  $y_1$  typically estimated as  $y_1$ .

• The formula used for EWMA is the same, but the smoothing parameter  $\lambda$  is typically used instead of  $\alpha$  and  $X_t$  instead of  $y_t$ :

$$EWMA_{t+1} = \lambda X_t + (1 - \lambda)EWMA_t$$

with the starting forecast value EWMA<sub>1</sub> estimated as the data mean or target value.

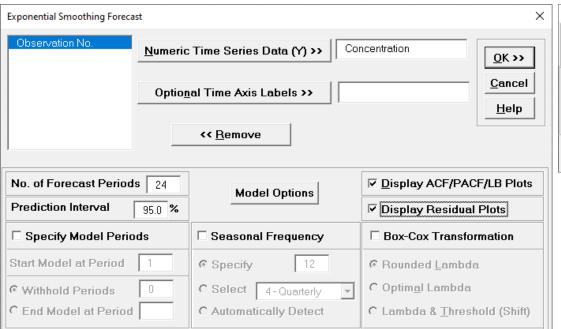
- In the case of an EWMA control chart, the smoothing parameter λ is determined by desired average run length characteristics and is typically 0.2.
- For forecasting or SPC for autocorrelated data, the smoothing parameter and initial level are determined by minimizing the sum-of-square forecast errors (residuals):

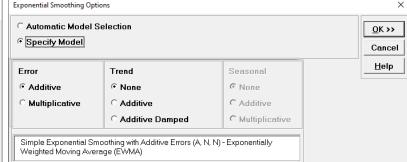
SSE = 
$$\sum_{t=1}^{T} (y_t - \hat{y}_t)^2 = \sum_{t=1}^{T} e_t^2$$
.

 This involves non-linear minimization methods like Newton-Raphson or Nelder-Mead Simplex.

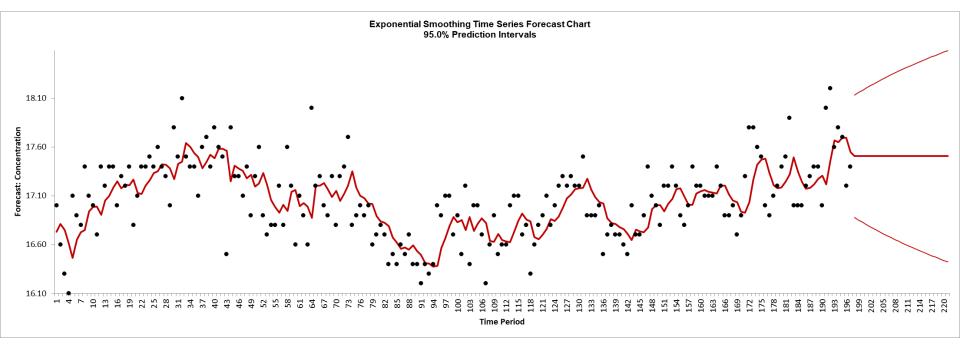
- As usual for any statistical model, the residuals should be normal, independent and identically distributed.
- If this is achieved, this also means that the assumptions for a Shewhart control chart are satisfied.

# Example 1: Box-Jenkins Series A - Chemical Process Concentration - Simple Exponential Smoothing (EWMA) Time Series Forecast





### Example 1: Box-Jenkins Series A - Chemical Process Concentration - Simple Exponential Smoothing (EWMA)



**Exponential Smoothing Model: Concentration** 

Model Type: Simple Exponential Smoothing with Additive Errors (A, N, N) - Exponentially Weighted Moving Average (EWMA)

Model Periods: All observations are used in the Exponential Smoothing model estimation. No withhold periods available for out-of-sample forecast accuracy evaluation.

<b>Exponential Smoothing Model Information</b>		
Seasonal Frequency	1	
Model selection criterion	AICc	
Box-Cox Transformation	N/A	
Lambda		
Threshold		

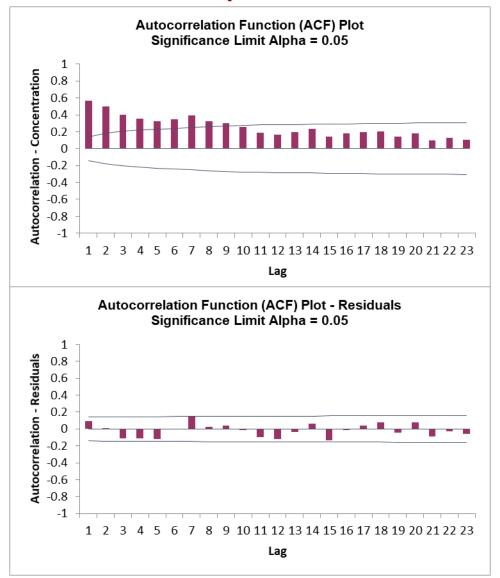
Parameter Estimates		
Term	Coefficient	
alpha (level smoothing)	0.294785988	
l (level initial state)	16.73121246	

		Exponential Smoothing Model Statistics		
		No. Observations	197	
DF 194		194		
StDev 0.319007644		0.319007644		
		Variance	0.101765877	
	Log-Likelihood		-293.8036067	
		AICc	593.7315658	
AIC		AIC	593.6072135	
	BIC		603.4568246	

Forecast Accuracy				
Metric	In-Sample (Estimation) One-Step-Ahead Forecast	Out-of-Sample (Withhold) One-Step-Ahead Forecast	Out-of-Sample (Withhold) Full Period Forecast	
N	197			
RMSE	0.316569334			
MAE	0.247329038			
MAPE	1.446520183			
MASE	0.897712804			

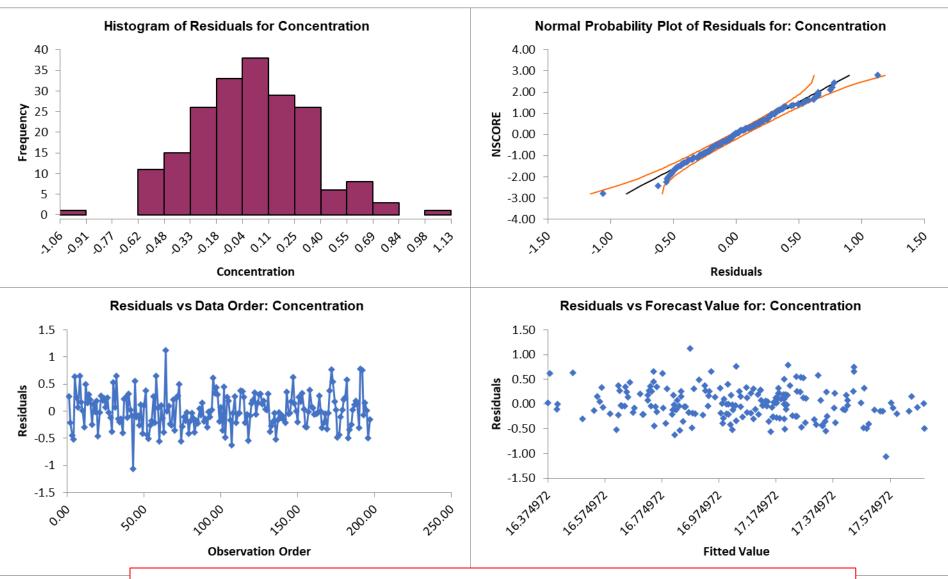
Simple Exponential Smoothing (EWMA) specified. 95% Prediction Intervals for forecast.

### Example 1: Box-Jenkins Series A - Chemical Process Concentration - ACF Plots (Raw Data versus Residuals)



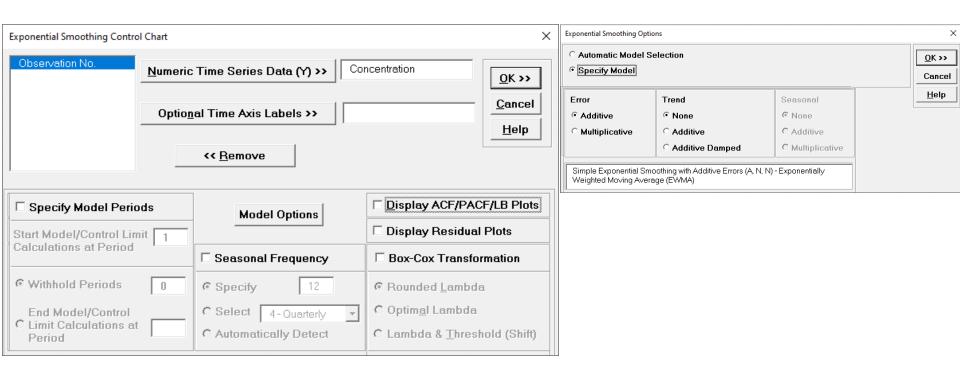
### Example 1: Box-Jenkins Series A - Chemical Process

#### **Concentration - Residuals**



Residuals look good – approximately normal with equal variance.

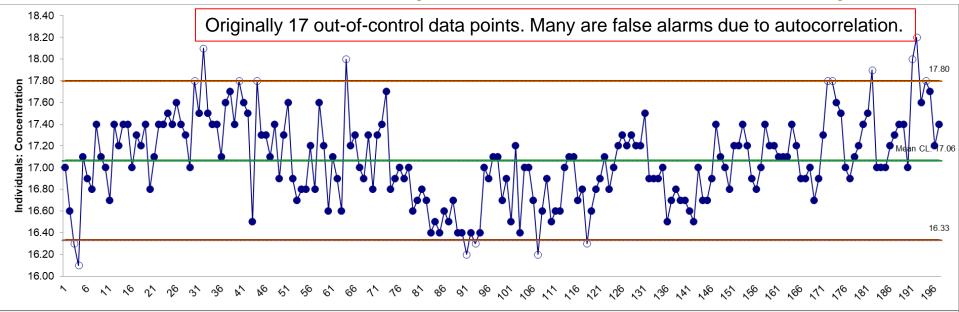
# Example 1: Box-Jenkins Series A - Chemical Process Concentration - Simple Exponential Smoothing (EWMA) Control Chart

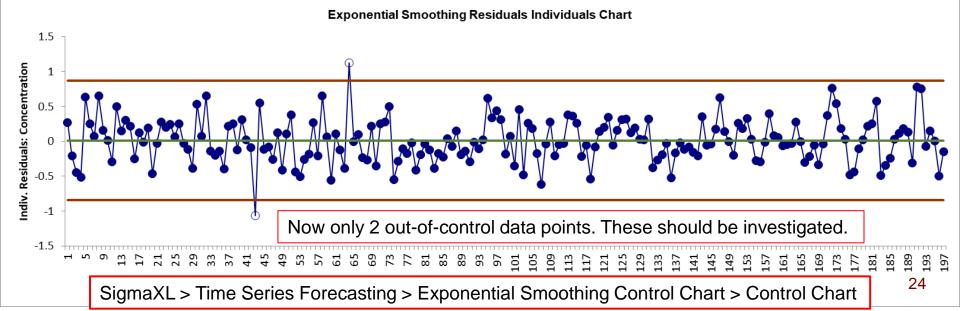


SigmaXL > Time Series Forecasting > Exponential Smoothing Control Chart > Control Chart

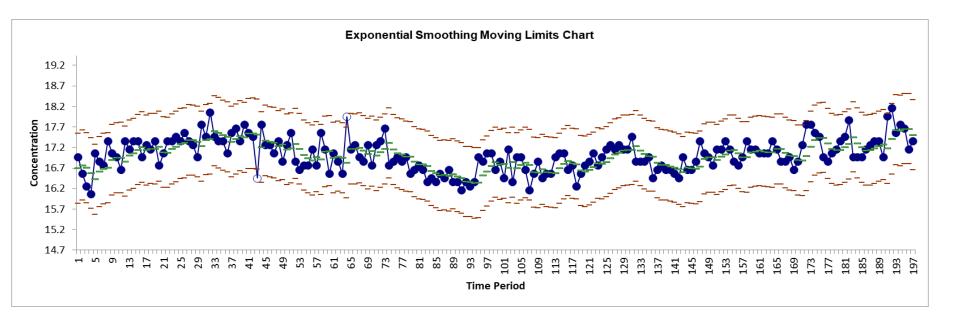
### Example 1: Box-Jenkins Series A - Chemical Process Conc. -

#### Individuals Control Chart (Raw Data versus Residuals)





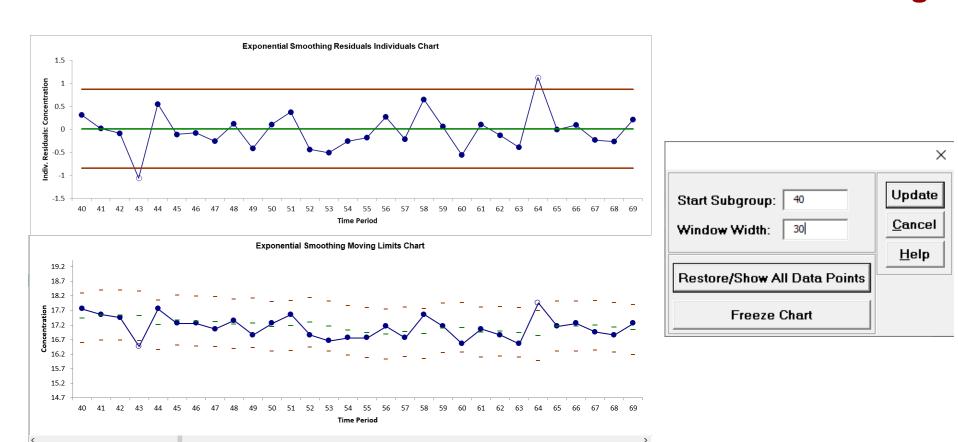
### Example 1: Box-Jenkins Series A - Chemical Process Concentration – Moving Limits Control Chart



The Moving Limits chart uses the one step prediction as the center line, so the control limits will move with the center line.

SigmaXL > Time Series Forecasting > Exponential Smoothing Control Chart > Control Chart

### Example 1: Box-Jenkins Series A - Chemical Process Concentration - Individuals Control Chart: Enable Scrolling



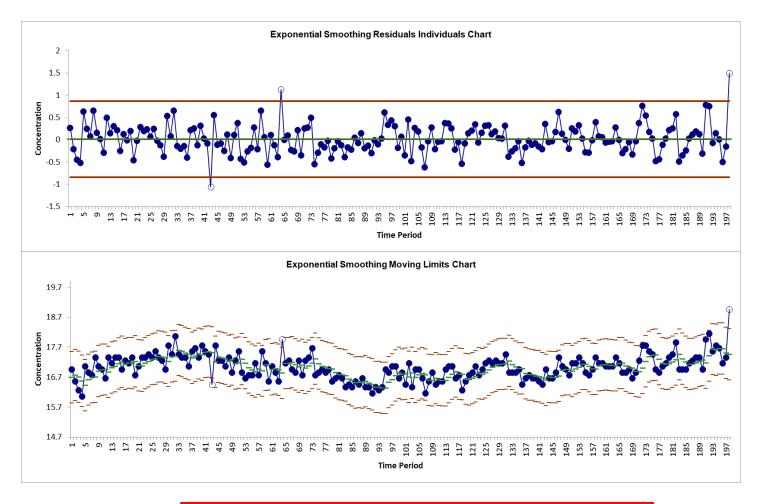
SigmaXL Chart Tools > Enable Scrolling

### Example 1: Box-Jenkins Series A - Chemical Process Concentration - Individuals Control Chart: Add Data

Now we will add a new data point to the Series A Concentration Data. The residuals will be computed using the same model as above without re-estimation of the model parameters or recalculation of the control limits. This is also known as the "Phase II" application of a Control Chart, where an out-of-control signal should lead to an investigation into the assignable cause and corrective action or process adjustment applied.

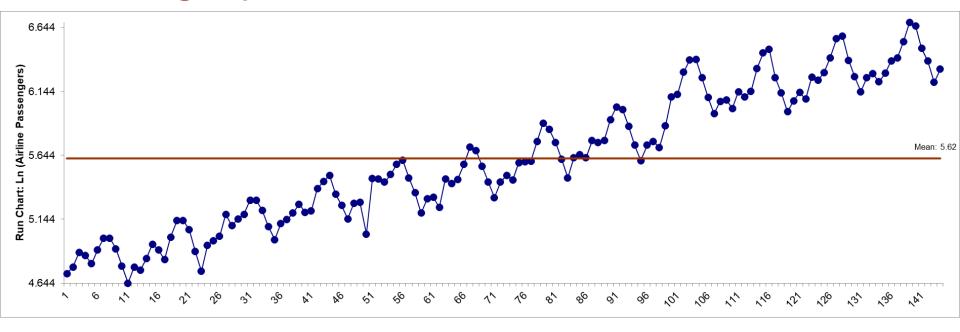
196	195	17.7
197	196	17.2
198	197	17.4
199	198	19

### Example 1: Box-Jenkins Series A - Chemical Process Concentration - Individuals Control Chart: Add Data



SigmaXL Chart Tools > Add Data to this Control Chart

### Example 2a: Box-Jenkins Series G – Ln(Monthly Airline Passengers) - Run Chart



Data modified with negative outlier at 50 (-.25) and level shift (+.25) starting at 100.

Data shows strong positive trend and strong seasonality (monthly data).

SigmaXL > Time Series Forecasting > Run Chart

Example 2: Airline Passengers Modified.xlsx – Ln(Airline Passengers)

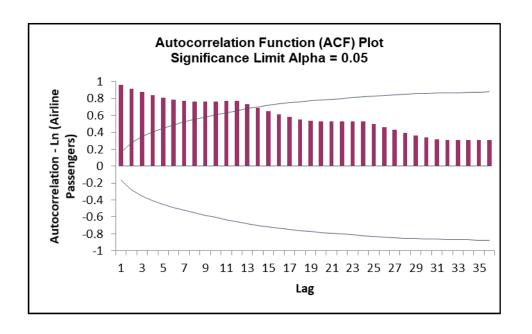
### Example 2a: Box-Jenkins Series G – Ln(Monthly Airline Passengers) - Individuals Control Chart



The control chart signals here are meaningless.

SigmaXL > Control Charts > Individuals

### Example 2a: Box-Jenkins Series G – Ln(Monthly Airline Passengers) - Autocorrelation (ACF) Plot



SigmaXL > Time Series Forecasting > Autocorrelation (ACF/PACF) Plots

## Error, Trend, Seasonal (ETS) Exponential Smoothing Models

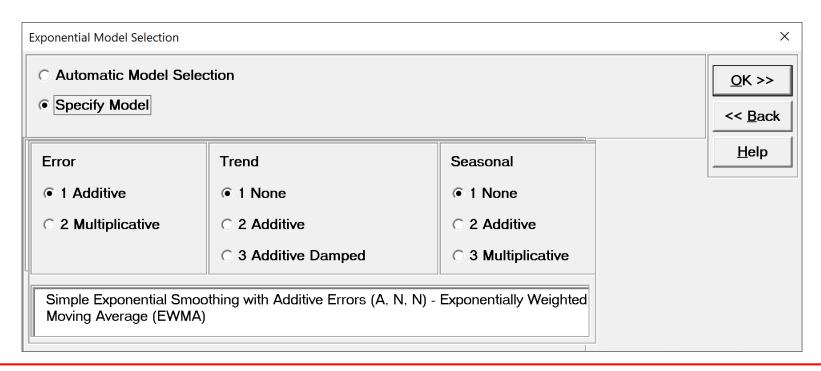
- Error, Trend, Seasonal (ETS) models expand on simple exponential smoothing to accommodate trend and seasonal components as well as additive or multiplicative errors.
- Simple Exponential Smoothing is an Error Model.
- Error, Trend model is Holt's Linear, also known as double exponential smoothing.

## Error, Trend, Seasonal (ETS) Exponential Smoothing Models

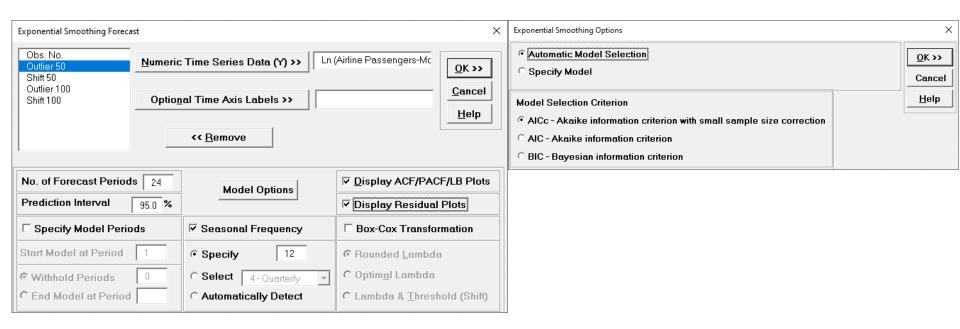
- Error, Trend, Seasonal model is Holt-Winters, also known as triple exponential smoothing.
  - Seasonal frequency must be specified:
    - Quarterly data = 4 (observations per year)
    - Monthly data = 12 (observations per year)
    - Daily data = 7 (observations per week)
    - Hourly data = 24 (observations per day)
  - Frequency is the number of observations per "cycle". This is the opposite of the definition of frequency in physics, or in engineering Fourier analysis, where "period" is the length of the cycle, and "frequency" is the inverse of period.

Reference: https://robjhyndman.com/hyndsight/seasonal-periods/

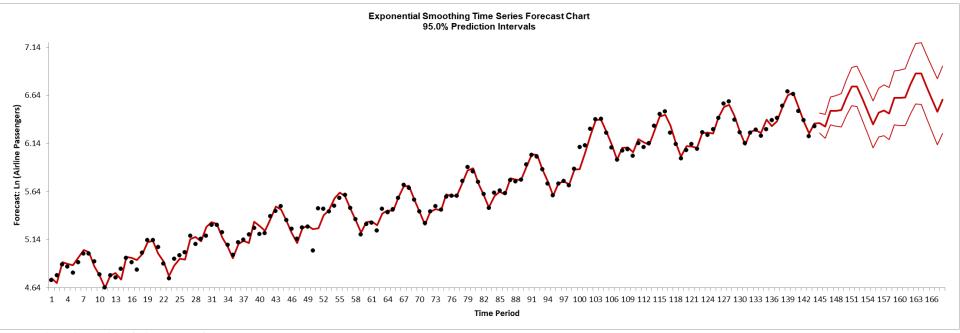
## Error, Trend, Seasonal (ETS) models Hyndman's Taxonomy



# Example 2b: Box-Jenkins Series G – Ln(Monthly Airline Passengers) – Exponential Smoothing Forecast with Automatic Model Selection



### Example 2b: Box-Jenkins Series G – Ln(Monthly Airline Passengers) Seasonal Exponential Smoothing with Trend



Exponential Smoothing Model: Ln (Airline Passengers)

Model Type: Additive Trend, Additive Seasonal Method with Additive Errors (Holt-Winters) (A, A, A)

Model Periods: All observations are used in the Exponential Smoothing model estimation. No withhold periods available for out-of-sample forecast accuracy evaluation.

<b>Exponential Smoothing Model Information</b>		
Seasonal Frequency	12	
Model selection criterion	AICc	
Box-Cox Transformation	N/A	
Lambda		
Throshold		

Parameter Estimates	
Term	Coefficient
alpha (level smoothing)	0.674949361
beta (trend smoothing)	0.0001
gamma (seasonal smoothing)	0.0001
I (level initial state)	4.821176207
b (trend initial state)	0.01120374
s1 (seasonal initial state)	-0.106060076
s2 (seasonal initial state)	-0.218770169
s3 (seasonal initial state)	-0.073902798
s4 (seasonal initial state)	0.065376172
s5 (seasonal initial state)	0.211146551
s6 (seasonal initial state)	0.221577218
s7 (seasonal initial state)	0.118793494
s8 (seasonal initial state)	-0.00223835
s9 (seasonal initial state)	0.001331568

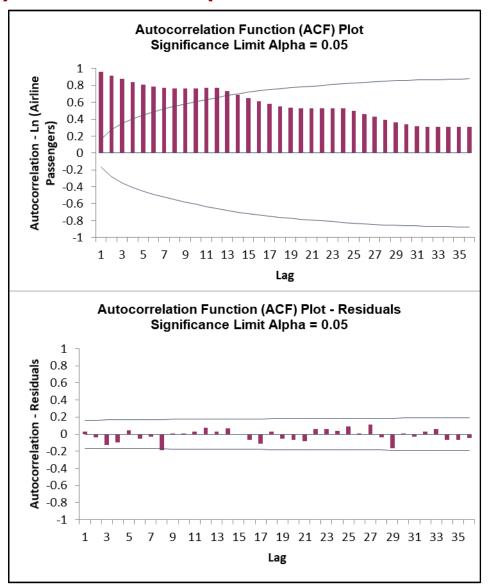
Exponential Smoothing Model Statistics		
No. Observations	144	
DF	127	
StDev	0.052717869	
Variance	0.002779174	
Log-Likelihood	74.98184762	
AICc	-111.1065524	
AIC	-115.9636952	
BIC	65 47696014	

Forecast Accuracy				
Metric	In-Sample (Estimation) One-Step-Ahead Forecast	Out-of-Sample (Withhold) One-Step-Ahead Forecast	Out-of-Sample (Withhold) Full Period Forecast	
N	144			
RMSE	0.049508351			
MAE	0.033506668			
MAPE	0.609538124			
MASE	0.230072112			

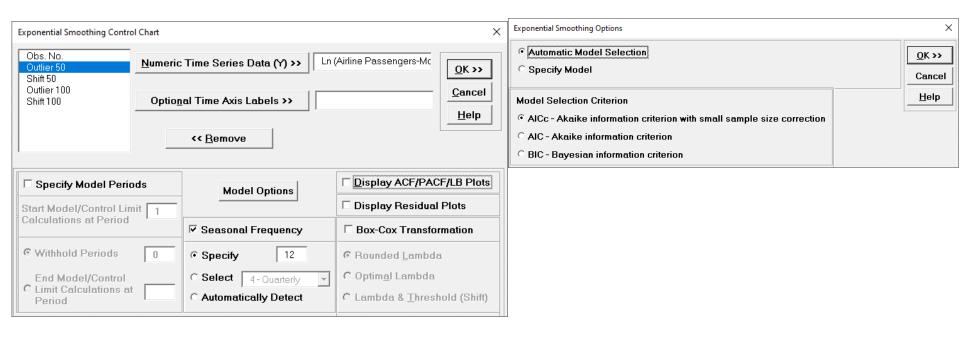
ETS Additive Trend, Additive Seasonal Method with Additive Errors (Holt-Winters) (A, A, A) **automatically selected**. Seasonal Frequency = 12 (Monthly data).

Chart Area

### Example 2b: Box-Jenkins Series G – Ln(Monthly Airline Passengers) - ACF Plots (Raw Data versus Residuals)

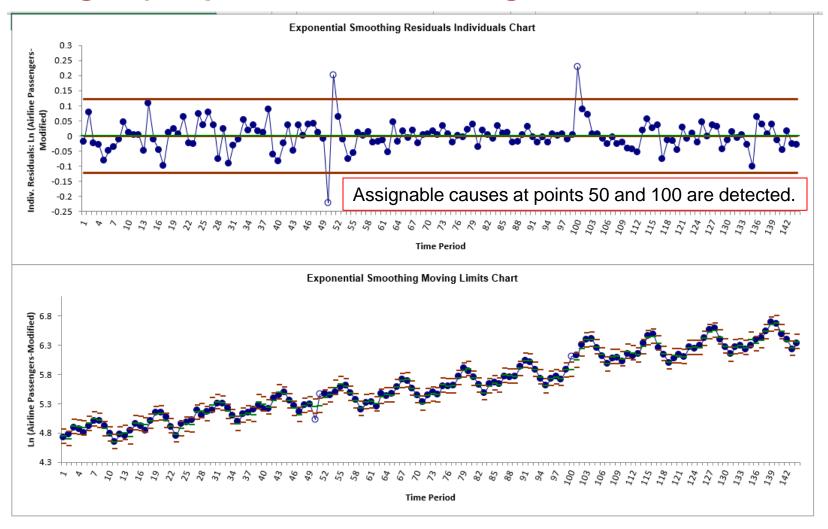


## Example 2b: Box-Jenkins Series G – Ln(Monthly Airline Passengers) – Exponential Smoothing Control Chart with Automatic Model Selection

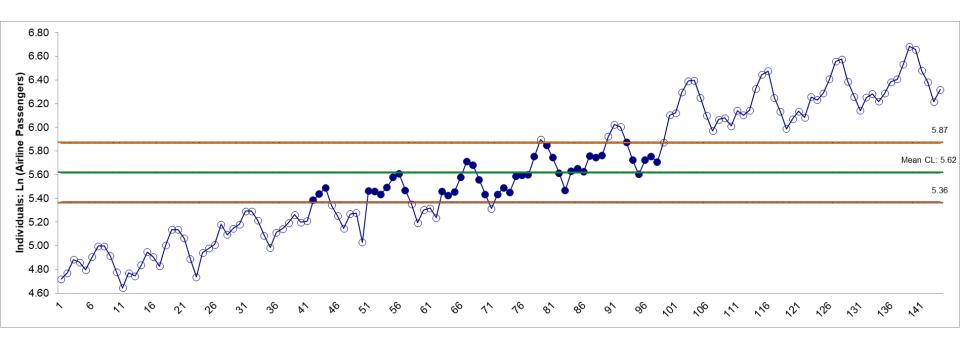


SigmaXL > Time Series Forecasting > Exponential Smoothing Control Chart > Control Chart

### Example 2b: Box-Jenkins Series G – Ln(Monthly Airline Passengers) Exponential Smoothing Control Charts



### Example 2b: Box-Jenkins Series G – Ln(Monthly Airline Passengers) - Individuals Control Chart for Raw Data



The control chart signals here are meaningless.

SigmaXL > Control Charts > Individuals

## **Box-Jenkins AutoRegressive Integrated Moving Average (ARIMA) Models**

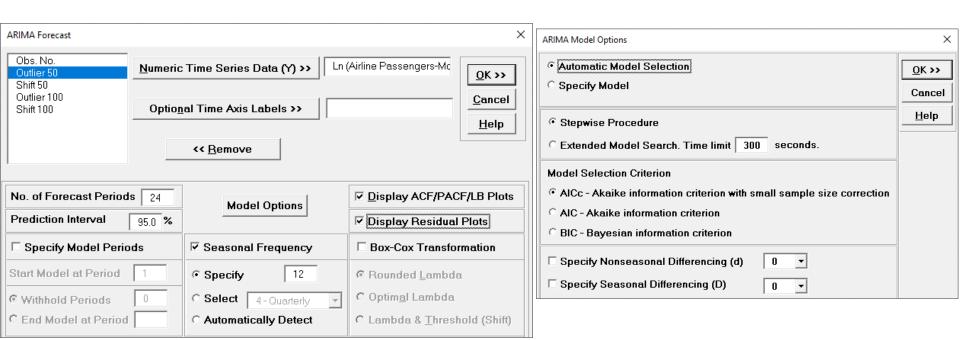
- An ARIMA model includes an Autoregressive (AR) component of order p, an Integrated/Differencing component of order d and a Moving Average component of order q and an optional constant.
- An ARIMA Seasonal model includes a Seasonal Autoregressive (SAR) component of order P, a Seasonal Integrated/Differencing component of order D and a Seasonal Moving Average component of order Q.

## **Box-Jenkins AutoRegressive Integrated Moving Average (ARIMA) Models**

ARIMA Model Selection		×
<ul><li>Automatic Model Selection</li><li>Specify Model</li></ul>		<u>O</u> K >>
Nonseasonal Order	Seasonal Order	<u>H</u> elp
AR - Autoregressive (p)	SAR – Seasonal Autoregressive (P)	
I – Integrated/Differencing (d)	SI - Seasonal Integrated/Differencing (D)	
MA - Moving Average (q)	SMA - Seasonal Moving Average (Q)	
☐ Include Constant (Mean if d & D = 0; Trend/Drift if d or D = 1)		

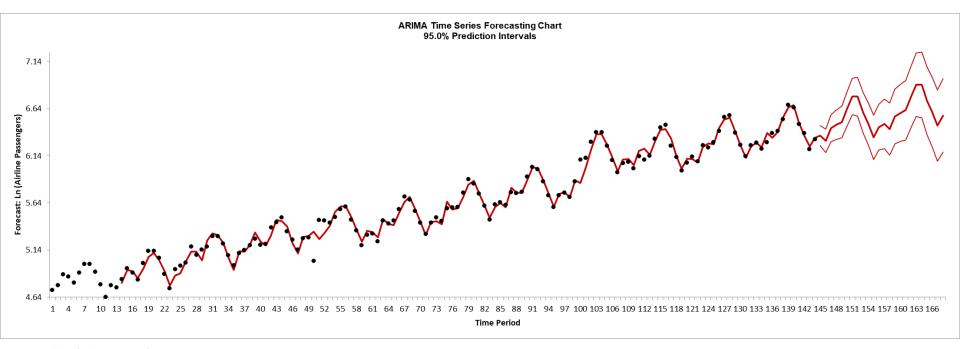
SigmaXL > Time Series Forecasting > ARIMA Forecast > Forecast

# Example 2c: Box-Jenkins Series G – Ln(Monthly Airline Passengers) – ARIMA Forecast with Automatic Model Selection



SigmaXL > Time Series Forecasting > ARIMA Forecast > Forecast

### Example 2c: Box-Jenkins Series G – Ln(Monthly Airline Passengers)



ARIMA Model: Ln (Airline Passengers)

**Model Periods:** 

All observations are used in the ARIMA model estimation. No withhold periods available for out-of-sample forecast accuracy evaluation

ARIMA Model Summary				
AR Order (p)	0			
l Order (d)	1			
MA Order (q)	1			
SAR Order (P)	0			
SI Order (D)	1			
SMA Order (Q)	1			
Seasonal Frequency	12			
Include Constant	0			
No. of Predictors	0			
Model selection criterion	AICc			
Box-Cox Transformation	N/A			
Lambda				

Parameter Estimates				
Term	Coefficient	SE Coefficient	Т	Р
MA_1	0.387990136	0.091662663	4.2328046	0.0000
SMA_1	0.686391161	0.074905258	9.1634577	0.0000

144	
129	
0.054403004	
0.002959687	
191.5831814	
-376.9773864	
-377.1663627	
-368.5407708	

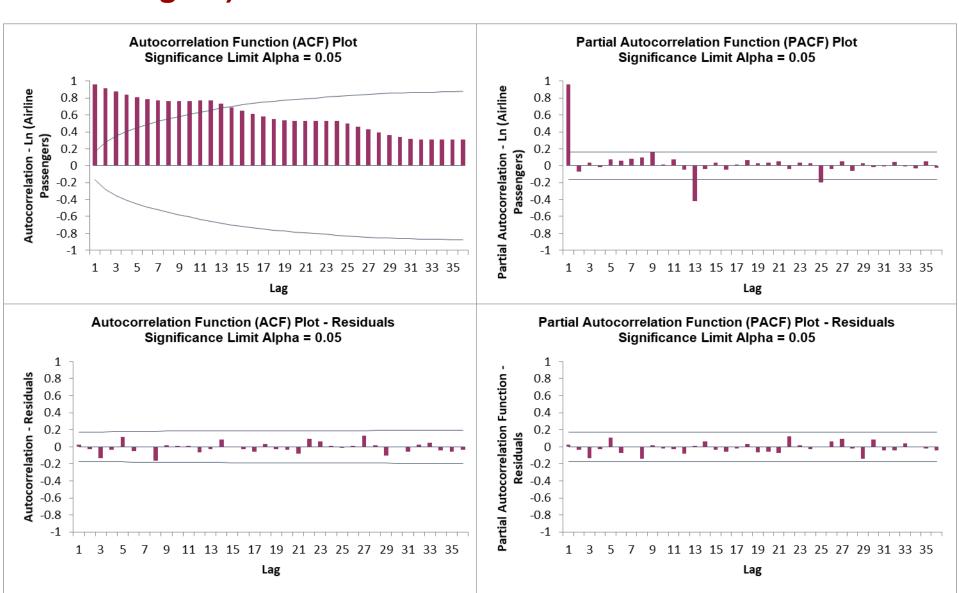
	Metric	One-Step-Ahead Forecast	One-Step-Ahead Forecast
	N	144	
	RMSE	0.055615558	
	MAE	0.036285476	
	MAPE	0.648152199	
	MASE	0.249152676	
1 '			

**Forecast Accuracy** 

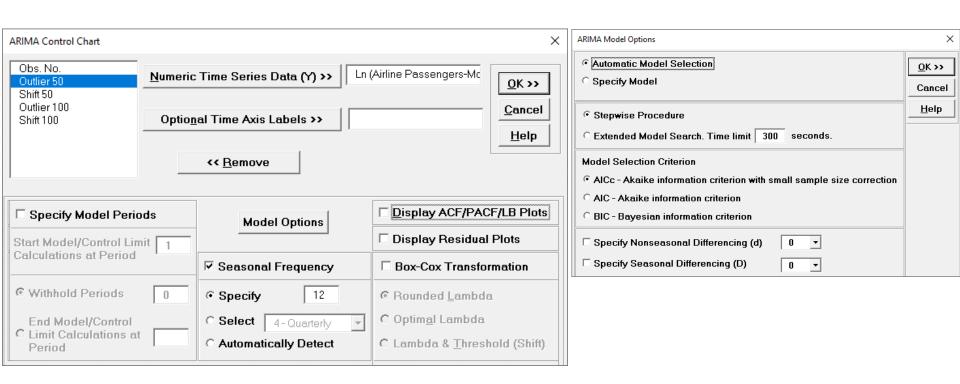
Residuals Randomness Runs Test
P-Value 0.2925

ARIMA (0,1,1) (0,1,1) automatically selected. Seasonal Frequency = 12 (Monthly data).

#### Example 2c: Box-Jenkins Series G – Ln(Monthly Airline Passengers)

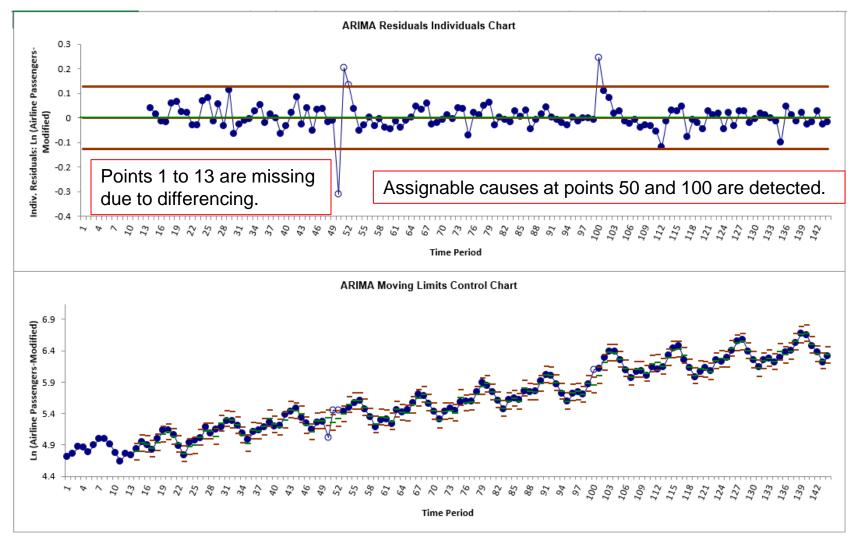


# Example 2c: Box-Jenkins Series G – Ln(Monthly Airline Passengers) – ARIMA Control Chart with Automatic Model Selection



SigmaXL > Time Series Forecasting > ARIMA Control Chart > Control Chart

### Example 2c: Box-Jenkins Series G – Ln(Monthly Airline Passengers) ARIMA Control Charts



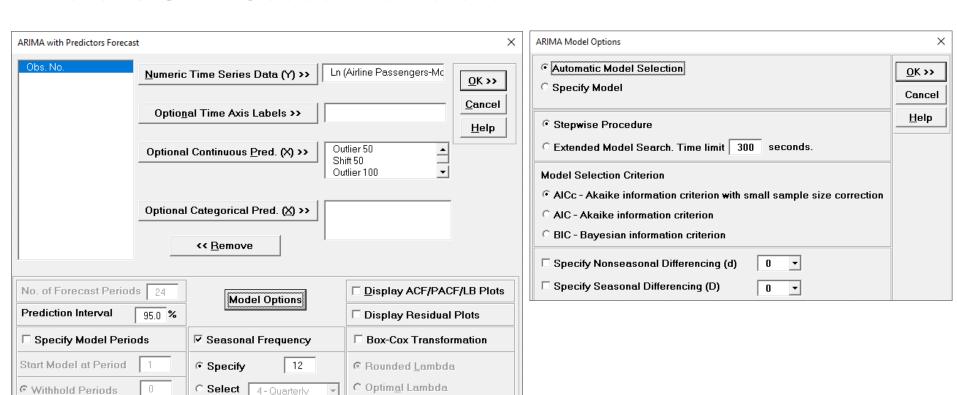
#### **ARIMA** with Predictors

- The ARIMA model supports continuous or categorical predictors, similar to multiple regression.
- In order to provide a forecast, additional predictor (X) values must be added to the dataset prior to running the analysis. The number of forecast periods will be equal to the number of additional predictor rows. Alternatively, the predictor values from a withhold sample may be used.
- As with multiple linear regression, predictors should not be strongly correlated.

#### Example 2d: Box-Jenkins Series G – Ln(Monthly Airline Passengers) – Outlier versus Shift Coded Predictors

Obs. No.	Ln (Airline Passengers- Modified)	Outlier 50	Shift 50	Outlier 100	Shift 100			
49	5.278114659	0	0	0	0			
50	5.028114659	1	1	0	0			
51	5.463831805	0	1	0	0			
52	5.459585514	0	1	0	0			
53	5.433722004	0	1	0	0			
54	5.493061443	0	1	0	0			
55	5.575949103	0	1	0	0			
99	5.874930731	0	1	0	0			
100	6.10220248	0	1	1	1			
101	6.122117789	0	1	0	1			
102	6.295005314	0	1	0	1			
103	6.392037406	0	1	0	1			
104	6.396329258	0	1	0	1			
105	6.251414878	0	1	0	1			

## Example 2d: Box-Jenkins Series G – Ln(Monthly Airline Passengers) – ARIMA Forecast with Predictors: Outlier versus Shift Coded Predictors



SigmaXL > Time Series Forecasting > ARIMA Forecast > Forecast with Predictors

C Lambda & Threshold (Shift)

C End Model at Period

Automatically Detect

#### Example 2d: Box-Jenkins Series G – Ln(Monthly Airline Passengers) – Outlier versus Shift Coded Predictors

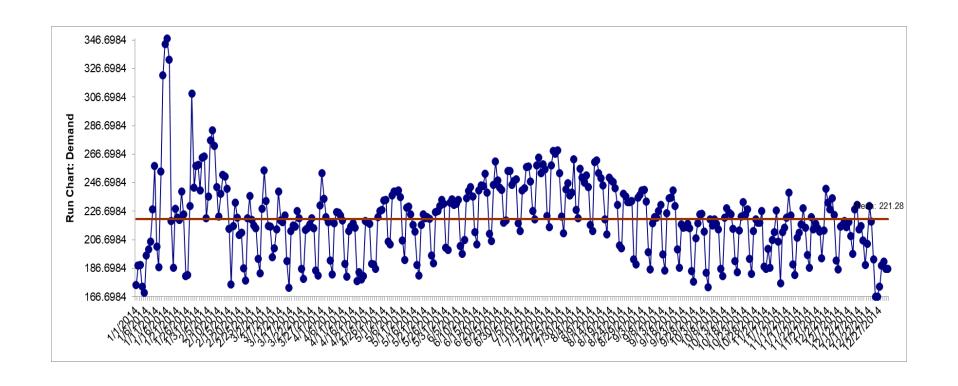
Using ARIMA Forecast with Predictors, we can see that *Outlier50* and *Shift100* are significant denoting Obs. No. 50 as an outlier and 100 as a shift. This is, of course, what we expected since that's how the Ln Airline Passenger data was modified.

This method to identify outlier versus shift is intended as a complement to process knowledge and the search for assignable causes used in classical SPC.

Parameter Estimates					
Term Coefficient SE Coefficient 1		T	Р		
MA_1	0.395854258	0.086277876	4.588132	0.0000	
SMA_1	0.556926317	0.073521579	7.575005	0.0000	
Outlier 50	-0.296267312	0.032292885	9.174384	0.0000	
Shift 50	0.044130277	0.035505653	1.242908	0.2162	
Outlier 100	-0.000597974	0.031822469	0.018791	0.9850	
Shift 100	0.249993132	0.035384072	7.065132	0.0000	

SigmaXL > Time Series Forecasting > ARIMA Forecast > Forecast with Predictors

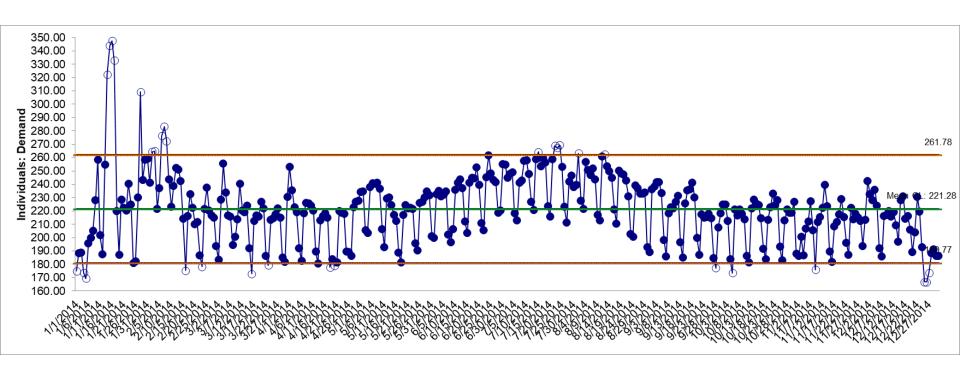
### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – Run Chart



SigmaXL > Time Series Forecasting > Run Chart

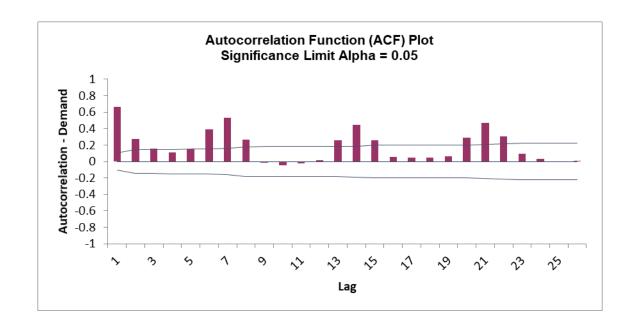
Example 3: Daily Electricity Demand with Predictors – ElecDaily.xlsx Victoria, Australia, 2014.

#### **Example 3: Daily Electricity Demand with Temperature** and Work Day Predictors - Individuals Control Chart



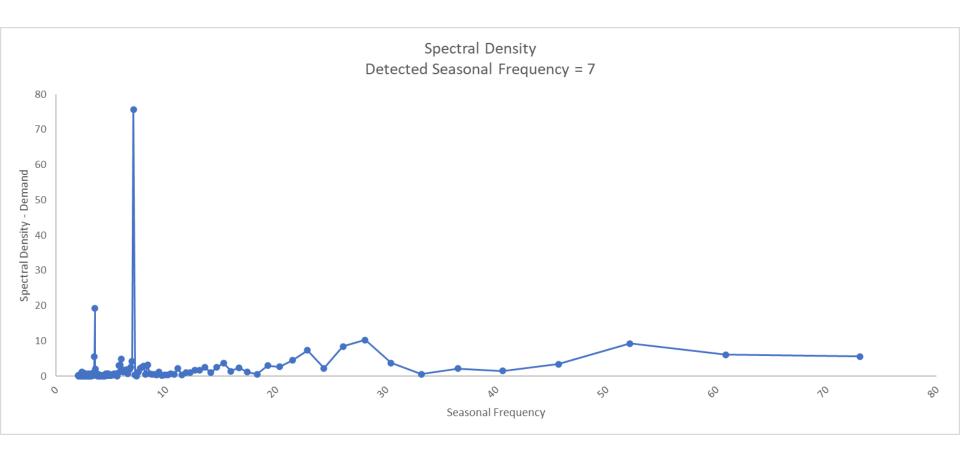
SigmaXL > Control Charts > Individuals

#### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – Autocorrelation (ACF) Plot



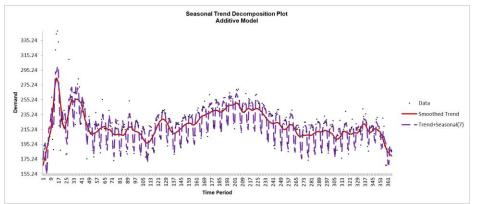
SigmaXL > Time Series Forecasting > Autocorrelation (ACF/PACF) Plots

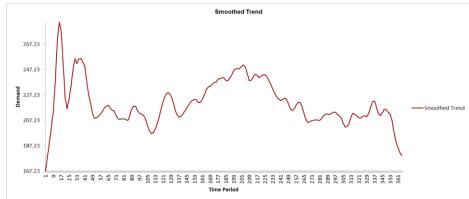
### **Example 3: Daily Electricity Demand with Temperature** and Work Day Predictors - Spectral Density Plot

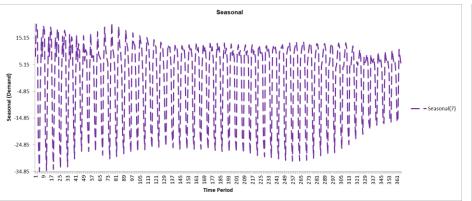


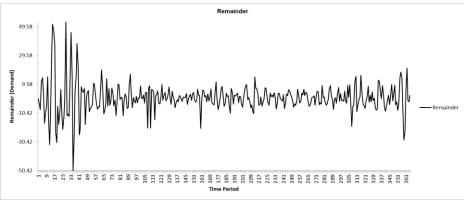
SigmaXL > Time Series Forecasting > Spectral Density Plot

#### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors - Seasonal Trend Decomposition



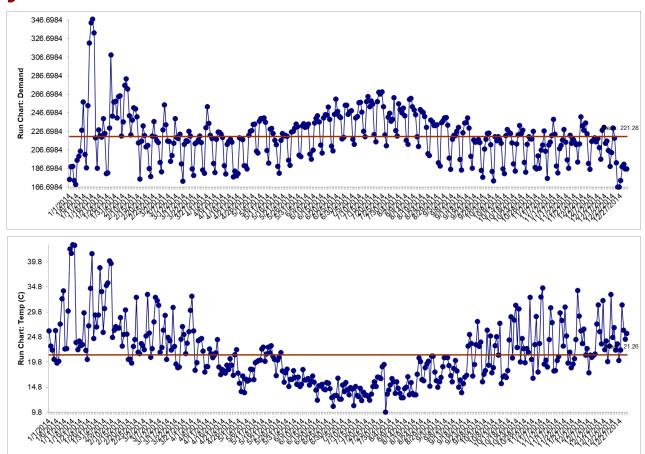






SigmaXL > Time Series Forecasting > Seasonal Trend Decomposition Plots

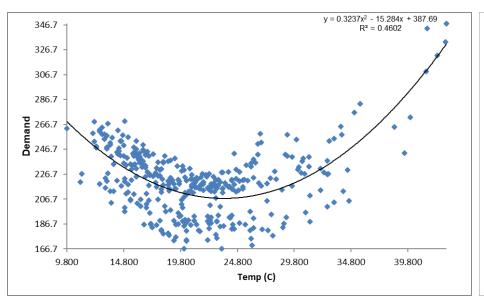
#### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – Run Charts

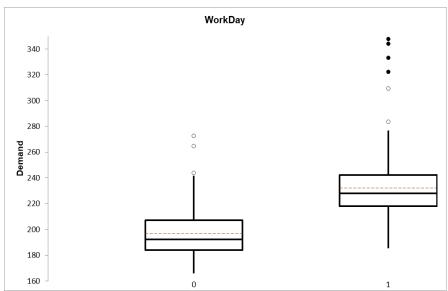


SigmaXL > Time Series Forecasting > Run Chart

Example 3: Daily Electricity Demand with Predictors – ElecDaily.xlsx Victoria, Australia, 2014.

#### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – Scatterplot and Box Plot



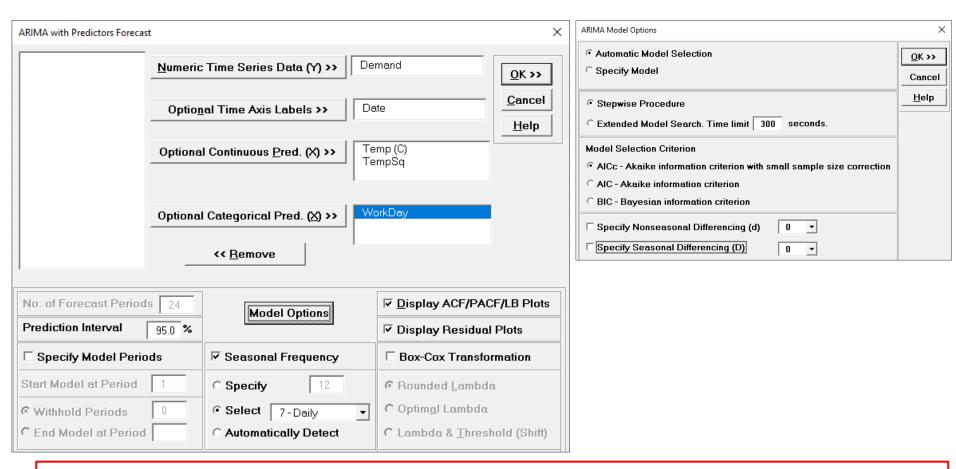


SigmaXL > Graphical Tools > Scatterplots

SigmaXL > Graphical Tools > Boxplots

Example 3: Daily Electricity Demand with Predictors – ElecDaily.xlsx Victoria, Australia, 2014.

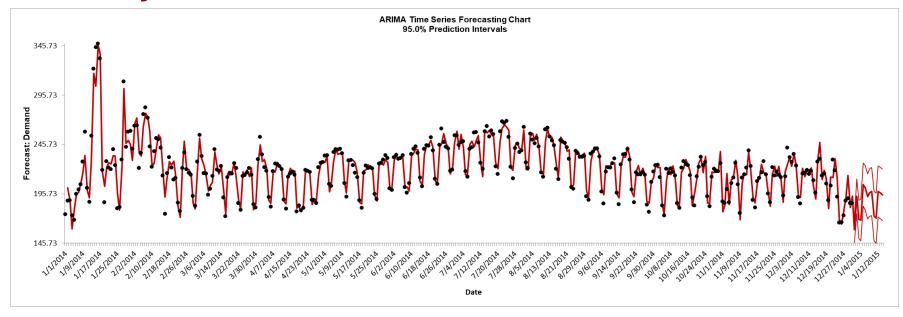
### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – ARIMA Forecast with Predictors



SigmaXL > Time Series Forecasting > ARIMA Forecast > Forecast with Predictors

Example 3: Daily Electricity Demand with Predictors – ElecDaily.xlsx, Sheet "Forecast 2 Weeks".

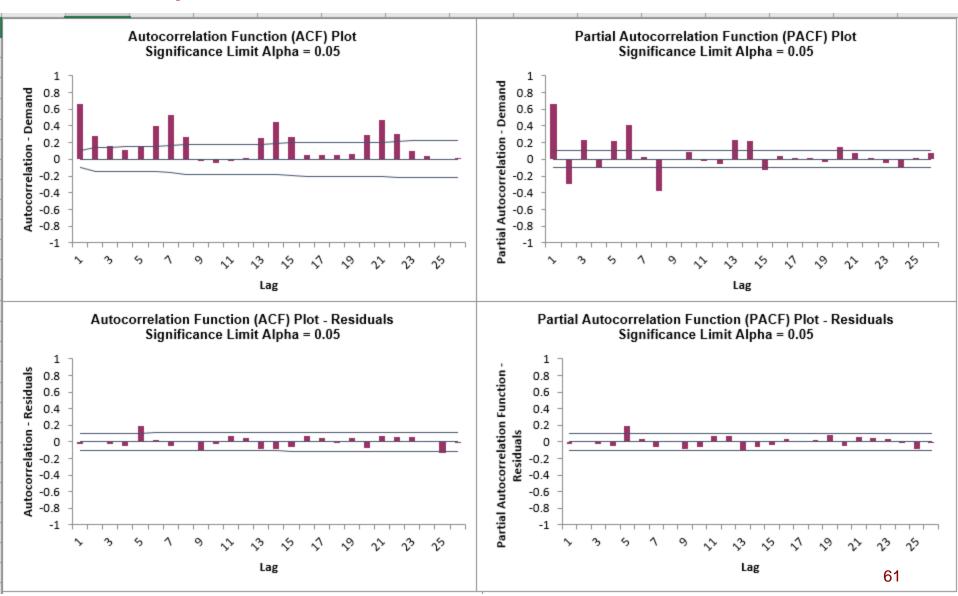
### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – ARIMA Forecast with Predictors



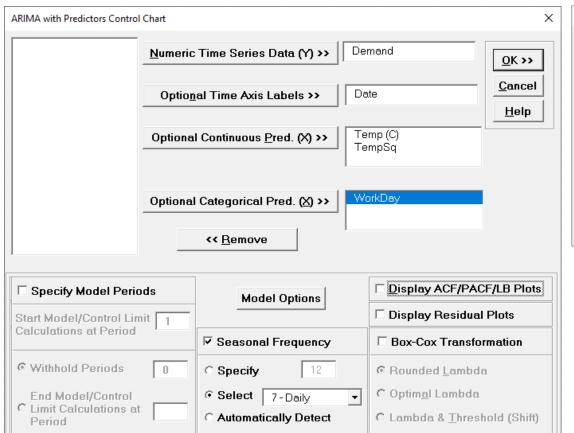
ARIMA Model Summary			
AR Order (p)	2		
l Order (d)	1		
MA Order (q)	2		
SAR Order (P)	2		
SI Order (D)	0		
SMA Order (Q)	0		
Seasonal Frequency	7		
Include Constant	0		
No. of Predictors	3		
Model Selection Criterion	AlCc		
Box-Cox Transformation	N/A		
Lambda			
Threshold			

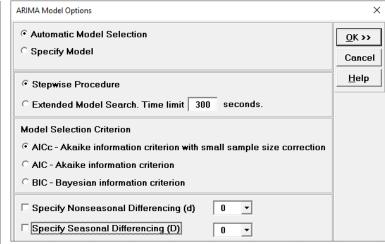
Parameter Estimates				
Term Coefficient SE Coef		SE Coefficient	T	Р
AR_1	-0.063223451	0.075658448	0.83564	0.4039
AR_2	0.673128346	0.067270503	10.0063	0.0000
MA_1	0.022660844	0.043288704	0.52348	0.6010
MA_2	0.929862871	0.039474102	23.5563	0.0000
SAR_1	0.200902989	0.053912363	3.72647	0.0002
SAR_2	0.402632085	0.05676416	7.09307	0.0000
Temp (C)	-7.501559029	0.446098708	16.8159	0.0000
TempSq	0.17890261	0.008530253	20.9727	0.0000
WorkDay_1	30.56943168	1.295720007	23.5926	0.0000

### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – ARIMA Forecast with Predictors



### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – ARIMA Control Chart with Predictors

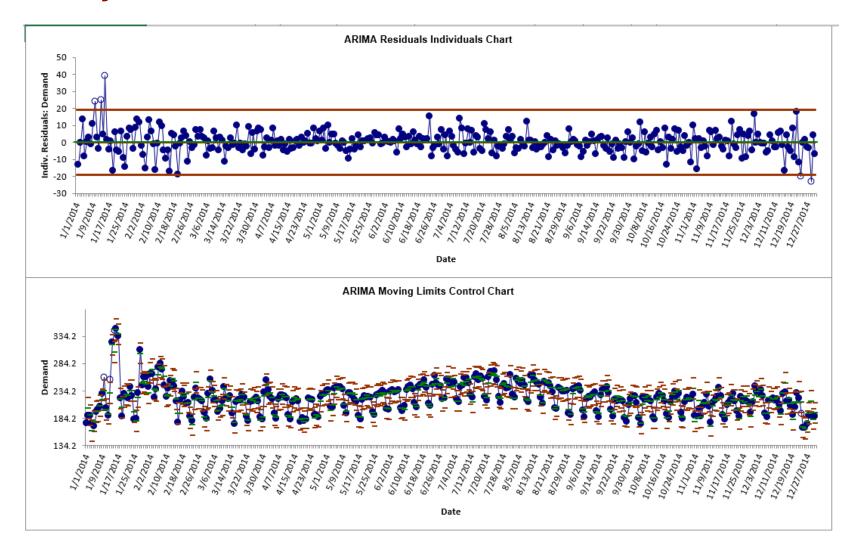




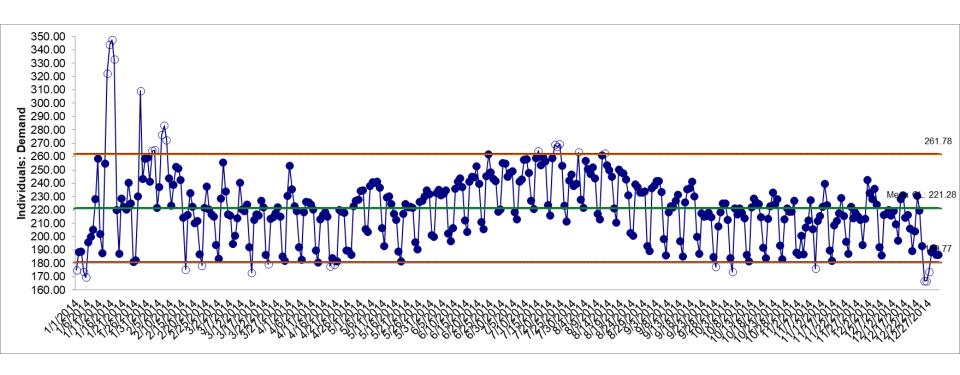
SigmaXL > Time Series Forecasting > ARIMA Control Chart > Control Chart with Predictors

Example 3: Daily Electricity Demand with Predictors – ElecDaily.xlsx, Sheet 1.

#### Example 3: Daily Electricity Demand with Temperature and Work Day Predictors – ARIMA Control Chart with Predictors



## Example 3: Daily Electricity Demand with Temperature and Work Day Predictors - Individuals Control Chart for Raw Data



Note the additional out-of-control signals.

SigmaXL > Control Charts > Individuals





#### What's New in SigmaXL® Version 9

Part 3 of 3: Control Charts for Autocorrelated Data

**Questions?** 



#### References

- 1. Alwan, L.C., and Roberts, H.V. (1988), "Time Series Modeling for Statistical Process Control," **Journal of Business and Economic Statistics**, 6, 87-95.
- 2. Box, G. E. P., Jenkins, G. M., Reinsel, G. C. and Ljung, G.M. (2016). *Time Series Analysis, Forecasting and Control*, 5<sup>th</sup> edition, Wiley.
- 3. Hunter, J.S. (1986), "The Exponentially Weighted Moving Average," **Journal of Quality Technology**, 18, 203-210.
- 4. Hyndman, R.J., & Athanasopoulos, G. (2018). Forecasting: principles and practice, 2<sup>nd</sup> edition, OTexts: Melbourne, Australia. OTexts.com/fpp2.
- 5. Hyndman, R. J. and Y. Khandakar, (2008). "Automatic Time Series Forecasting: The forecast Package for R."

  Journal of Statistical Software, 27(3), 1-22.

#### References

- Montgomery, D. C., and Mastrengelo, C.M. (1991), "Some Statistical Process Control Methods for Autocorrelated Data," Journal of Quality Technology, 23, 179-204.
- 7. Montgomery, D. C. (2013). *Introduction to Statistical Quality Control*, 7<sup>th</sup> edition, Wiley.
- 8. Montgomery, D. C., Jennings, C.L., and Kulahci, M. (2015). *Introduction to Time Series Analysis and Forecasting*, 2<sup>nd</sup> edition, Wiley.
- 9. NIST/SEMATECH e-Handbook of Statistical Methods, <a href="https://www.itl.nist.gov/div898/handbook">https://www.itl.nist.gov/div898/handbook</a>.
- 10. Woodall, W.H. and Faltin, F.W. "Autocorrelated Data and SPC" **ASQC Statistics Division Newsletter**, 13(4).