# statgraphics®

# Special Purpose Control Charts in Statgraphics Centurion

#### Presented by Dr. Neil W. Polhemus

## **Control Charts**

 $\geq$  A control chart:

- Plots a statistic versus time.
- Contains control limits to detect unusual values.
- Used for 2 primary purposes:
  - To determine whether a process is in a state of statistical control. (Phase 1 or initial study)
  - To monitor that process to detect deviations from pre-established parameters. (Phase 2 or control to standard)



## **Common Control Charts**

- For variable data:
  - X-bar and R, X-bar and S, X and MR
  - Median and MR
  - Moving average and EWMA
  - Cumulative sum charts
- For attribute data:
  - P or NP for binomial proportions and counts
  - C or U for Poisson rates and counts
  - P-prime and U-prime charts for overdispersed data

## **Special Purpose Control Charts**

- 1. ARIMA charts
- 2. Acceptance charts
- 3. Toolwear charts
- 4. Charts for rare events
- 5. Cuscore charts



#### **Usual Model for Variable Data**

Let  $X_t$  be measurement made at time t. Then

 $X_t = \mu + \varepsilon_t$ 

where  $\mu$  is the process mean and the deviations  $\varepsilon_t$  are assumed to be NID(0, $\sigma^2$ ).

Assumptions:

- Stationarity of mean and variance.
- Independence of successive measurements.
- Normal distribution for deviations from mean.



#### **ARIMA Control Charts**

- Used for data that are serially correlated (successive observations are not independent).
- Very common when measurements are taken close together in time from a continuous process.



#### **Example: Chemical concentration**





#### Standard Phase 1 X Chart





#### **ARIMA Charts Model**

- Relax the assumption that successive observations are independent.
- Instead, assume that the process can be described by an autoregressive moving average model.



#### **Autoregressive Models**

• AR(1)

$$X_t = \mu + \phi_1(X_{t-1} - \mu) + \varepsilon_t$$

• AR(2)

### $X_t = \mu + \phi_1(X_{t-1} - \mu) + \phi_2(X_{t-2} - \mu) + \varepsilon_t$



#### **Moving Average Models**

• MA(1)

$$X_t = \mu + \varepsilon_t - \theta_1 \varepsilon_{t-1}$$

• MA(2)

## $X_t = \mu + \varepsilon_t - \theta_1 \varepsilon_{t-1} - \theta_2 \varepsilon_{t-2}$



#### **ARMA Models**

• ARMA(1,1)

 $X_t = \mu + \phi_1 \left( X_{t-1} - \mu \right) + \varepsilon_t - \theta_1 \varepsilon_{t-1}$ 



## Selecting the Best ARIMA Model

	Automatic Forecasting	
Time Concentration	Data:	
	(Time Indices:)	
	or Sampling Interval	
	C       Year(s) (4-digit)       C       Hour(s)       1.0         C       Quarter(s)       C       Minute(s)       1.0         C       Month(s)       C       Second(s)         C       Day(s)       Image: Content of the content of	
	(Seasonality:)	
	(Trading Days Adjustment:)	
Sort column names	(Select:)	68
	Number of Forecasts: Withhold for Validation: 60 0	P.o.
OK Ca	ncel Delete Transform Help	



## Selecting the Best ARIMA Model

Aut	tomatic Forecasting Opt	ions ×
Models to Include		OK Cancel
Random Walk with Drift  Mean  Linear Trend  Quadratic Trend  Exponential Trend  S-Curve Moving Average Simple Exp. Smoothing	<ul> <li>Optimize Parameters</li> </ul>	Help Method Selection Criterion Akaike Information Criterion (AIC) Hannan-Quinn Criterion (HQC) Schwarz Bayesian Inf. Criterion (SBIC) Mean Squared Error (MSE)
<ul> <li>Holt's Linear Exp. Smoothing</li> <li>Quadratic Exp. Smoothing</li> <li>Winters' Exp. Smoothing</li> <li>ARIMA: Optimize Model Order</li> <li>AR Terms (p)</li> <li>Nonseasonal: 2</li> </ul>	<ul> <li>Optimize Parameters</li> <li>Optimize Parameters</li> <li>Optimize Parameters</li> <li>Optimize Parameters</li> <li>Optimize Parameters</li> <li>Differencing (d)</li> <li>Nonseasonal: 0</li> </ul>	C Mean Abs. Percentage Error (MAPE)          Adjustments         Parameters         Estimation
Seasonal: 0 Seasonal: 0	Seasonal: 0	Input series



# Selecting the Best ARIMA Model





Backforecasting: yes

Estimated white noise variance = 0.109914 with 117 degrees of freedom Estimated white noise standard deviation = 0.331533

Number of iterations: 6

imber of iterati

<



## **Selected Model**





## **ARIMA Chart – Data Input**

	ARIMA Individuals Chart	×	
Time Concentration	Observations: Concentration (Date/Time/Labels:) Time (LSL:) (Nominal:) (USL): (Select:) (Select:)		
Sort column names	el Delete Transform Help	]	
	statgraphics.com		

# **ARIMA Chart – Analysis Options**

ARIMA Individuals	s Charts Options
Type of Study	Control to Standard
Initial Study     Control to Strended	Mean:
- ARIMA Control Limite - MR(2) Control Limite -	Sigma:
Upper: Upper:	AB(1): MA(1):
3.0 Sigma 3.0 Sigma	AB(2): MA(2):
✓ Lower:     ✓ Lower:     ✓.3.0 Sigma	AB(4): MA(4):
Model         Estimate sigma           AR         I         MA           O         O         O           O         I         O           O         I         I           O         I         I           O         I         I           I         I         I	from t Chart type Chart type Data with long-term limits Data with one-step limits Residuals Normalized Residuals
O 4 O 4	Transform



# Data with Long-Term Limits





## Data with One-Step Limits





## **Chart of Residuals**





# MR(2) Chart for Residuals





## **Acceptance Control Charts**

- Used to insure that product remains within specification limits.
- Position control limits with respect to the specifications rather than the process mean.
- Useful for high C<sub>pk</sub> processes which can tolerate some wandering of the process mean.



## Sample Data

Plot of strength vs time





## X Chart





#### **Acceptance Charts**

- Begin by specifying *f*, the maximum allowable fraction of nonconforming items.
- Find most extreme values of process mean that would yield no more than f nonconformities:

Largest allowable mean:  $\mu_U = USL - Z_{\delta}\sigma$ Smallest allowable mean:  $\mu_L = LSL + Z_{\delta}\sigma$ 

Position control limits with respect to these extremes.



## **Positioning Limits**

 Sigma multiple method: add "3-sigma" to the extreme values for the mean.

 $UCL = \mu_U + 3\sigma$  $LCL = \mu_L - 3\sigma$ 

Note: Montgomery calls these "modified control limits".

 Beta risk method: specify β, probability of not generating an out of control signal. Position control limits at:

$$UCL = \mu_U - Z_\beta \sigma$$
$$LCL = \mu_L + Z_\beta \sigma$$



## Acceptance Chart – Data Input

	Acceptance Individuals Chart	×	
strength time	Observations:         Image: Strength         (Date/Time/Labels:)         Image: Strength         (LSL:)       (Nominal:)         200       Image: Strength         Strength         (Select:)		
Sort column names			100 mar
	el Delete Iransform Help		100
	statgraphics		

## Acceptance Chart – Analysis Options

Accept	ance Individuals Ch	nart Options	
Type of Study Initial Study Control to Standard	Specify G Sigma Multiple Beta Risk	Control to Standard Mean: 254.653 Std. Dev.: 7.62501	
X Control Limits Fraction Nonconforming: 0.0001 Sigma multiple: 3.0	MR(2) Control Limits Upper: 3.0 Sigma Lower: -3.0 Sigma		
OK Cano	el Exclude	Transform Help	*******
	staturaphics com	enturion	

statgraphics.com

## **Based on Sigma Multiple**

Acceptance Chart for strength LSL = 200.0, Nominal = 250.0, USL = 300.0





## **Based on Beta Risk**

Acceptance Chart for strength LSL = 200.0, Nominal = 250.0, USL = 300.0

300 284.18 280 260 254.65 × 240 220 215.82 200 3:20 6:40 13:20 16:40 0:00 10:00 time



#### **Toolwear Control Charts**

- Used for data that do not have a constant mean, such as measurements affected by wear on a tool.
- Used to determine whether mean is changing at a constant or expected rate.
- Can also add specification limits to help determine when tool should be changed.



## Sample Data (from Duncan)



• Specification: 0.255 to 0.265 inches



#### **Toolwear Chart Model**

Let  $X_{t,j}$  be measurement on  $j^{th}$  sample made at time t. Then

$$X_{t,j} = \alpha + \beta t + \varepsilon_{t,j}$$

where  $\mu = \alpha + \beta t$  is the process mean and the deviations  $\varepsilon_t$  are assumed to be NID(0, $\sigma^2$ ).



## **Toolwear Chart – Data Input**

	Toolwear Chart	×	
 Time X1 X2 X3 X4 X5	Data Deservations X1 X2 X3 X4 X5		
	C Subgroup Statistics Means: Ranges: Sizes:		
	Date/Time/Labels or Size: Time (LSL:) (Nominal:) (USL 0.255	0.265	
Sort column names	(Select:)	Help	



# **Analysis Options**

	Toolwear Chart Opt	ions	×	
Type of Study         Initial Study         Control to Standard         Include Model Estimation         X-bar Control Limits         ✓ Upper:         3.0         Sigma         ✓ Lower:         -3.0         Sigma	Estimate Sigma From R or S Chart Model MSE n Error R or S Control Limits Upper: 3.0 Sigma Lower: -3.0 Sigma	Control to Standard Intercept: Slope: Sigma:		
OK Cano	cel Exclude	Transform	Help	
	statgraphics.com	<b>CS</b> <sup>®</sup> urion		

## **Toolwear Chart**

Toolwear Chart for X1-X5 LSL = 0.255, USL = 0.265





## **Charts for Rare Events**

- Used for monitoring occurrence of rare events
   Hospital infections
  - Factory or mining accidents
  - Unexpected shutdowns
- Two primary charts
  - t chart: for events recorded in continuous time
  - g chart: for events recorded in discrete time



# Examples

	C:\Data\webinar\	infections.sgd		23
	Infection time	Date of infection	Col_3	
1	1/1/12 8:11	1/4/11		
2	1/2/12 5:42	1/5/11		
3	1/3/12 21:55	1/7/11		
4	1/5/12 5:24	1/8/11		
5	1/7/12 10:14	1/10/11		
6	1/9/12 11:11	1/13/11		
7	1/12/12 16:56	1/14/11		
8	1/15/12 13:39	1/14/11		
9	1/17/12 14:26	1/20/11		
10	1/18/12 13:26	1/31/11		
11	1/21/12 10:53	2/2/11		
12	1/22/12 23:58	2/2/11		
13	1/25/12 6:34	2/15/11		
14	1/27/12 6:31	2/19/11		
15	1/28/12 14:18	2/19/11		
16	1/30/12 16:20	2/21/11		
17	2/4/12 9:15	2/23/11		
18	2/8/12 23:19	3/5/11		
19	2/9/12 5:15	3/8/11		
20	2/12/12 13:34	3/9/11		
21	2/15/12 8:46	3/9/11		
22	2/18/12 4:37	3/12/11		-
	infections B C			



#### **Time Between Events**

- When events are rare, standard charts such as a C chart will have many 0's.
- Better approach than plotting counts is to plot the times between consecutive events.
- For a continuous time process, the interevent times are modeled by a Weibull distribution.
- For a discrete time process, the interevent times are modeled by a geometric distribution.



## **Scatterplots**



statgraphics.com

# t Chart Data Input

	t Chart	×	
Infection time Date of infection	Data: Infection time Type of data Time of occurrence Time between occurrences (Labels:) (LSL:) (Nominal:) (USL): (Select:)		
Sort column names	Delete Transform Help	]	· · · · · · · · · · · · · · · · · · ·
			I
	statoraphics		

centurion

statgraphics.com

# t Chart Analysis Options

	t Chart Options	×	
Type of Study  Initial Study  Control to Standard  Recalculate at:	<ul> <li>Normalize</li> <li>✓ Avg. Subgroup Size</li> <li>✓ Use Zone Format</li> </ul>	Control to Standard Specify Parameters Shape: Scale:	
t Chart Control Limits Upper: 3.0 Sigma Lower: -3.0 Sigma		C Specify Control Limits Upper: Centerline: Lower:	
ОК	Cancel Exclu	ide Help	
	statgraphics.com	nics <sup>®</sup>	

## t Chart





# g Chart Data Input

	g Chart	X	
Infection time Date of infection	Data:		
	Type of data Time of occurrence Time between occurrences		
	(Labels:)	]	
	(LSL:) (Nominal:) (USL):		1
Sort column names			65 (A) (A)
OK Cancel	Delete Transform Help		0
	statoraphics®		

centurion

statgraphics.com

# g Chart Analysis Options

	g Chart Options		×
<ul> <li>Type of Study</li> <li>Initial Study</li> <li>C Control to Standard</li> </ul>	<ul> <li>Normalize</li> <li>Avg. Subgroup Size</li> <li>Use Zone Format</li> </ul>	Control to Standard Specify Parameter mean:	
Recalculate at: g Chart Control Limits ✓ Upper: 3.0 Sigma ✓ Lower: -3.0 Sigma		C Specify Control Limits Upper: Centerline: Lower:	536° 0 0 0
OK	Cancel Exclu	ıde Help	
s	statgraph	nturion	

# g Chart







## G Chart – Special Run Test

Extra run test for consecutive zeroes:

ns Tests les nuns above or below centerline of length 8 or greater. runs up or down of length 8 or greater. sets of 5 subgroups with at least 4 beyond 1.0 sigma. sets of 3 subgroups with at least 2 beyond 2.0 sigma. 4 consecutive zeroes (alpha = 0.27%) Plations bservation g Chart H H B e StatAdvisor is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly ting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table ws the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.		G Chart - Date of infection	x
les         nuns above or below centerline of length 8 or greater.         runs up or down of length 8 or greater.         sets of 5 subgroups with at least 4 beyond 1.0 sigma.         sets of 3 subgroups with at least 2 beyond 2.0 sigma.         4 consecutive zeroes (alpha = 0.27%)         Dations         bservation       g Chart         H         H         e StatAdvisor         is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly ting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table ws the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	Runs Tests		
runs above or below centerine of length 8 or greater.         runs up or down of length 8 or greater.         sets of 5 subgroups with at least 4 beyond 1.0 sigma.         sets of 3 subgroups with at least 2 beyond 2.0 sigma.         4 consecutive zeroes (alpha = 0.27%)         Plations         bservation       g Chart         H         H         H         with a subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	Rules		
runs up or down of length 8 or greater. sets of 5 subgroups with at least 4 beyond 1.0 sigma. sets of 3 subgroups with at least 2 beyond 2.0 sigma. 4 consecutive zeroes (alpha = 0.27%) plations bservation g Chart H H e StatAdvisor is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly ting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	(A) runs above o	r below centerline of length 8 or greater.	
sets of 3 subgroups with at least 4 beyond 1.0 sigma. 4 consecutive zeroes (alpha = 0.27%) Plations bservation g Chart H H E StatAdvisor is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly ting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	(B) runs up or do	wn of length 8 or greater.	
sets of 3 subgroups with at least 2 beyond 2.0 sigma.         4 consecutive zeroes (alpha = 0.27%)         plations         bservation       g Chart         H         H         H         is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly tring away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table ows the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	(C) sets of 5 subg	groups with at least 4 beyond 1.0 sigma.	
4 consecutive zeroes (alpha = 0.27%)         plations         bservation       g Chart         H         H         H         is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly thing away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	(D) sets of 3 subs	groups with at least 2 beyond 2.0 sigma.	
baservation       g Chart         bservation       g Chart         H       H         H       H         is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly thing away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table ows the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	(H) 4 consecutive	e zeroes (aipna = 0.27%)	
bservation       g Chart         H       H         e StatAdvisor         is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly tring away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	Violations		
H         H         H         H         H         Is stable searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly ting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table ows the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	Observation	g Chart	
e StatAdvisor is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly ting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table ows the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	59	H	
e StatAdvisor is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly ting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	60	Н	
e StatAdvisor is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly fting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.			
is table searches for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly fting away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	The StatAdvisor		
Tring away from target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table was the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	This table search	es for and identifies any unusual patterns in the data. This is often helpful in detecting processes which are slowly	
ows the subgroup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	drifting away from	n target, even though no points may fall outside the control limits. 2 unusual runs have been detected. The table	
mple, rule H was violated at point 59. At this location, there were 4 consecutive zeroes.	shows the subgro	oup or observation at which the unusual pattern was detected, as well as the particular rule which was violated. For	r
	example, rule H w	as violated at point 59. At this location, there were 4 consecutive zeroes.	
> > > > >	<		>



## **Cuscore Charts**

- Designed to detect specific types of disturbances:
  - Spike
  - Ramp
  - Bump
  - Step change
  - Exponential increase
  - Sine wave
- Automatically selects the best type of control chart to detect that pattern



## Example: sine wave plus noise

I	si	gnal plus noise.sgd		
	t	Noise	Signal	Col_4
1	1	-0.298475889641	0	
2	2	0.158540926563	0	
3	3	2.44921673896	0	
4	4	-2.92901952736	0	
5	5	1.66838656126	0	
6	6	1.25353294413	0	
7	7	0.645732200693	0	
8	8	-1.51840745872	0	
9	9	0.470703738286	0	
10	10	0.162318392636	0	
11	11	-0.0911150761078	0	
12	12	0.156118205884	0	
13	13	0.434819344846	0	
14	14	-0.573646766803	0	
15	15	-0.56808795381	0	
16	16	0.572478593534	0	
17	17	-0.13202972842	0	
18	18	0.719853740624	0	
	I signal plus noise	BC		



# Signal and Noise



# Signal Plus Noise





#### **Q** Score Statistic

 Cuscore chart plots a cumulative "Q score" statistic defined by:

$$Q_t = \sum_{i=1}^{t} \hat{a}_i r_i$$

t

where the  $r_i$  values come from a "detector" series specifically designed to detect the presence of the expected disturbance and the  $a_i$ values are the residuals from an ARIMA model.



## Cuscore Chart – Data Input

	Cuscore Individuals Charts	×
t Signal	Observations: Signal+Noise (Date/Time/Labels:) t (LSL:) (Nominal:) (USL): Select:) (Select:)	
Sort column names		10101000000000
OK Cance	I Delete Transform Help	
	stat <b>graphics</b> ®	

centurion

## **Cuscore Chart – Analysis Options**

	Cuscore Chart Options
Type of Study Estim Initial Study	IR(2) Chart Mean: -0.014877
C Control to Standard	odel MSE         Sigma:         0.982517           Control Limits         AB(1):         MA(1):
3.0 Sigma 3.0	AR(2): MA(2): MA(2): MA(3): MA
I         Lower:         I         Low           -3.0         Sigma         -3.0	Sigma         AB(4):         MA(4):
Noise Model           AR         I         MA           ● 0         ● 0         ● 0           ○ 1         ○ 1         ○ 1           ○ 2         ○ 2         ○ 2           ○ 3         □         Constant         ○ 3	Signal to Detect         Spike       C Ramp       C Bump of Duration       2         Step Change       C Exponential Increase of       0.1         Sine Wave with Period       100       Phase       0
C 4 C 4	Custom         0         0         0         0           0         0         0         0         0           0         0         0         0         0
ОК	Cancel Transform Help



## **Cuscore Chart**



#### References

- Introduction to Statistical Quality Control (7<sup>th</sup> edition) by Douglas Montgomery (2012)
- <u>Quality Control and Industrial Statistics</u> (5<sup>th</sup> edition) by Acheson Duncan (1986)
- <u>Statistical Control by Monitoring and Adjustment</u> (2<sup>nd</sup> edition) by George Box, Alberto Luceño, and Maria del Carmen Paniagua-Quinones (2009)



#### **More Information**

Go to www.statgraphics.com

 Click on "Learn" and then "Instructional Videos".

Go to <u>www.youtube.com</u>

- Search for "Statgraphics control charts".

