

# Quality Corner

SPC: Measurement Systems Analysis, Process Capability,  
and Control Charts

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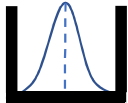
# Statistical Process Control

- Methods and tools used to monitor and improve product quality by monitoring and reducing process variation
- First step to reduce variation is to understand the source:
  1. Common Cause Variation
    - Is stable over time?
    - Is a result of the system or process design?
    - Processes will have some natural level of variation over time
  2. Special Cause Variation
    - Is process unstable over time?
    - Is assignable to a specific event that is outside of the process?
- Process is said to be operating in statistical process control when the only sources of variation are from common causes

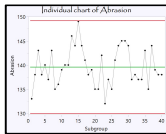
# SPC Common Theme



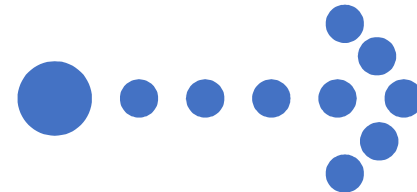
Measurement Systems  
Analysis



Process Capability



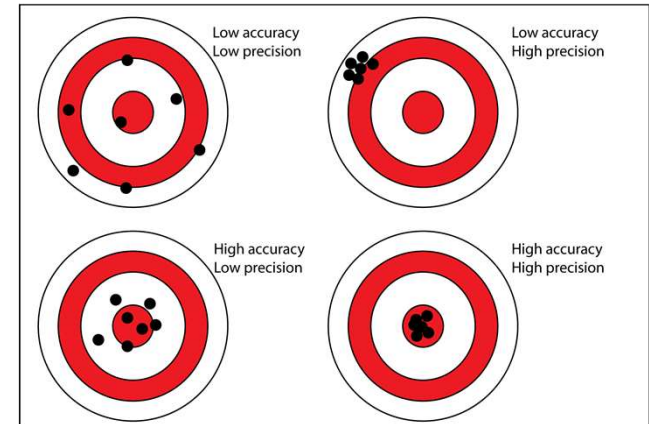
Control Charts



# Measurement Systems Analysis

# MSA

- Assess the measurement system
- What is the true value? (bias/accuracy)
- What is the precision? (variation)
- Range of techniques
  - Linearity
  - Resolution
  - Stability
  - Go/no go
- Helps determine how much variation is due to the measurement system rather than from product or process variation



# Sources of Variation:

Part to Part

Operator

Device (equipment, gage)

$$SS_{Total} = SS_{Part} + SS_{Measurement\ System}$$

$$SS_{Total} = \sum(Y_i - \hat{Y})^2$$

SS: Measure the distance from the average, square it and sum them all together.

SS is a measure of variation.


# Precision Errors

- **Repeatability** – the variation due to the measuring device. It is the variation observed when the same operator measures the same part repeatedly with the same device.
- **Reproducibility** – the variation due to the measurement system. It is the variation observed when different operators measure the same parts using the same device.

# MSA – Gage R&R Studies

$$SS_{Total} = SS_{part} + SS_{Operator} + SS_{Gage}$$

**Reproducibility**      **Repeatability**


$$SS_{Total} = SS_{part} + SS_{R\&R}$$

For a good Measurement System

$$\begin{array}{l} SS_{part} / SS_{Total} \gg SS_{R\&R} / SS_{Total} \\ > 90\% \qquad \qquad < 10\% \end{array}$$

**\* Variance SS Percent (%) of Total \***

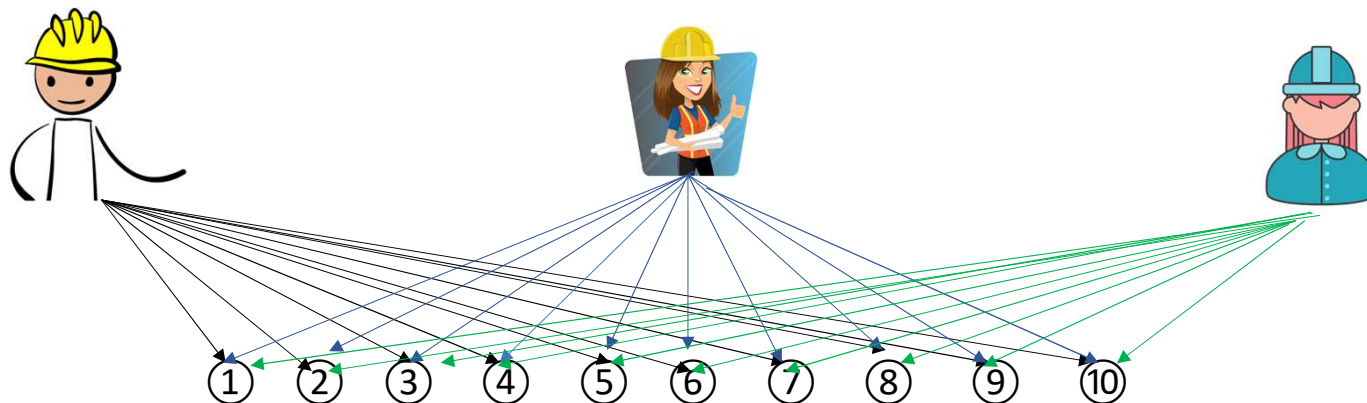


# Crossed Versus Nested Designs

- **Crossed** designs we assume that each operator can measure the same parts that other operators measure. We also plan on having each operator measure the same part more than once, (e.g. board that can be measured again)
- **Nested** designs we assume that the part is destroyed (e.g. breaking strength test, flammability test) to obtain a measurement, therefore, for repeatability the operator must have parts that are assumed to be the “same” to obtain repeatability variation.

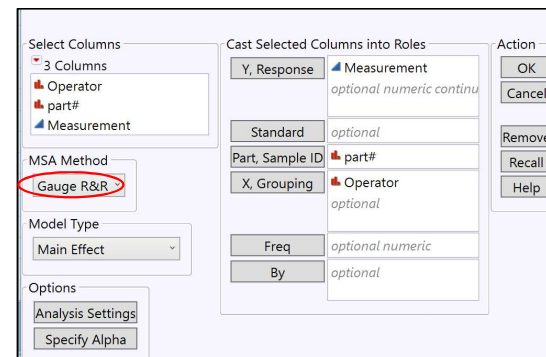
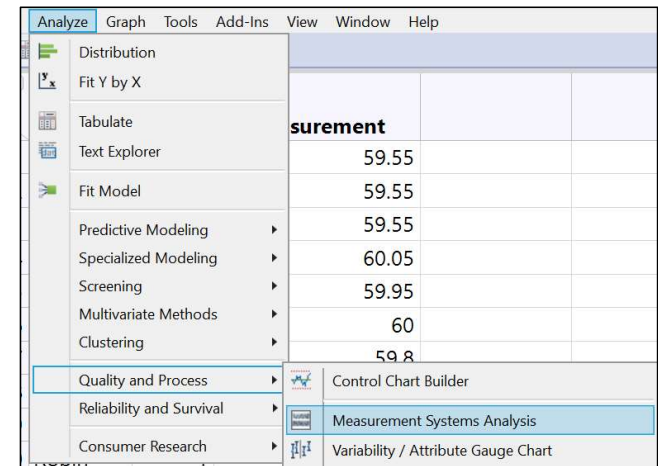
# Example: Measuring Board Length

- Ten boards were selected that represent the expected range of the process variation (usually we want to assure at least 80% coverage).
- Three operators measured the ten parts, three times per part.

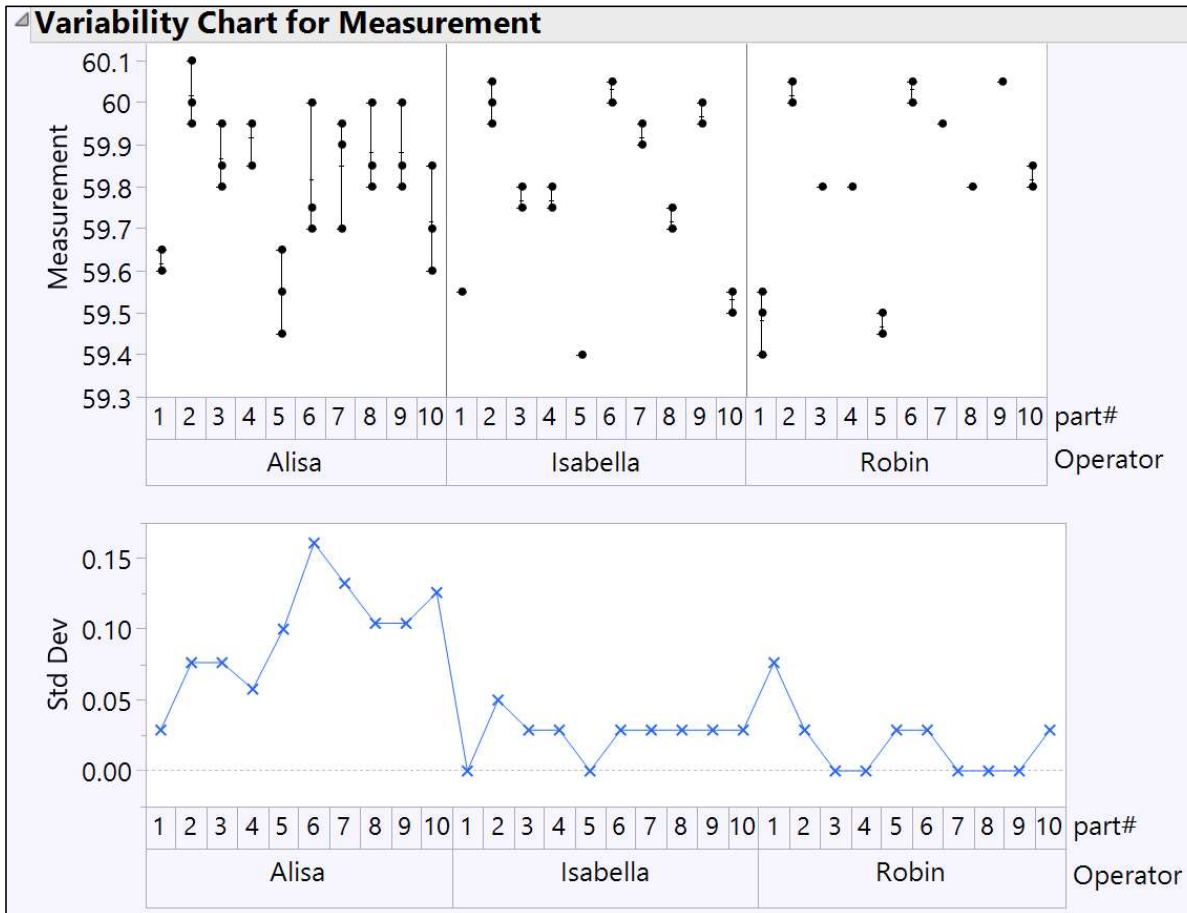


# Instructions

1. Select **Analyze**→ **Quality and Process**  
→**Measurement Systems Analysis**
2. Select **Measurement** for **Y, Response**
3. Select **part#** for **Part, Sample ID**
4. Select **Operator** for **X, Grouping**
5. Make sure **MSA Method** is changed to **Gauge R&R**
6. Select **OK**.



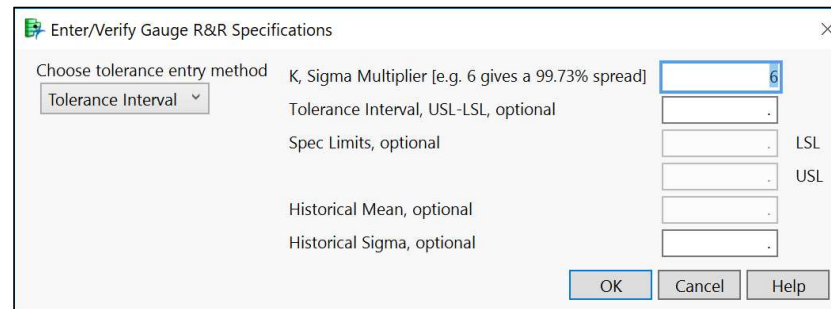
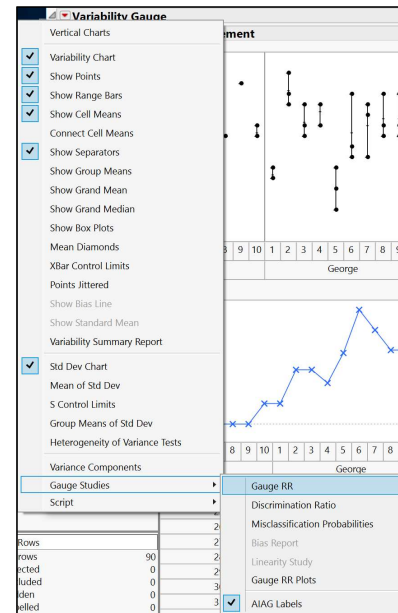
# Output



Does it look like there is variation within operator or between operators?

# Instructions

1. Click on the red triangle next to **Variability Gauge**.
2. Select **Gauge Studies**  
→ **Gauge RR**
3. Select **Crossed**
4. Select **OK**.



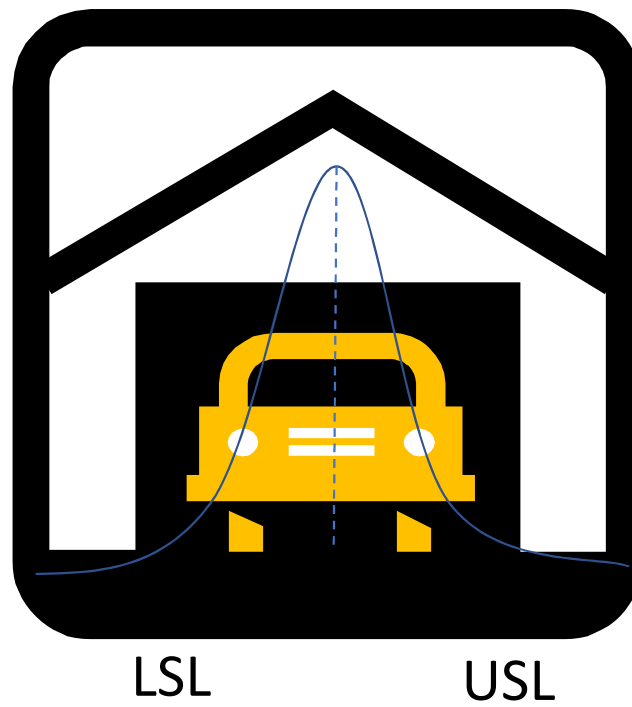
# Output

Variance Components for Gauge R&R			
Component	Var Component	% of Total	20406080
Gauge R&R	0.00864815	21.47	
Repeatability	0.00800000	19.86	
Reproducibility	0.00064815	1.61	
Part-to-Part	0.03162449	78.53	

Look to see what % contribution of Gage R&R is.  
Is this an acceptable Gage R&R?

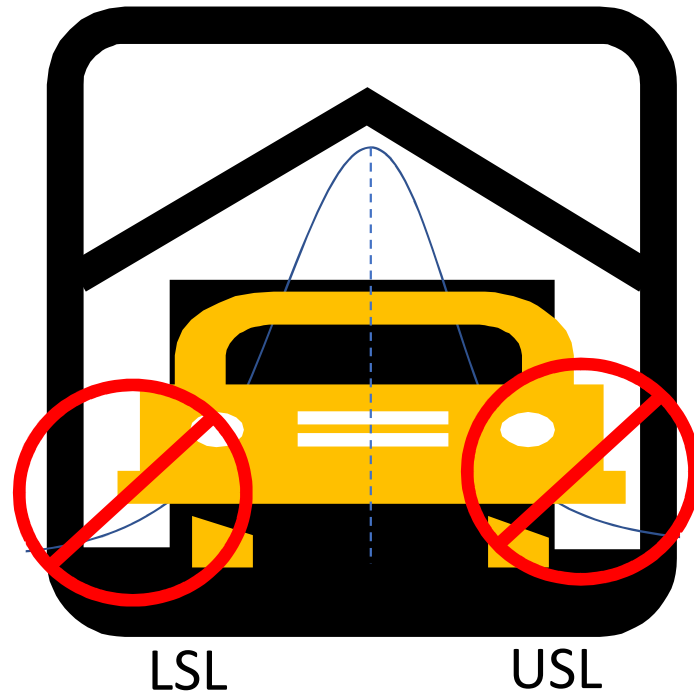
# Process Capability

# Driving into a Garage

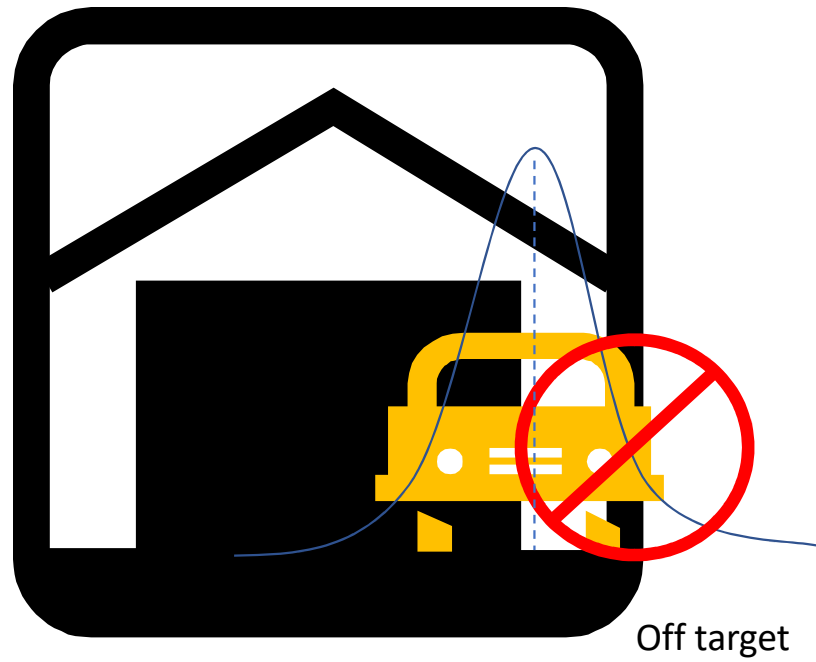




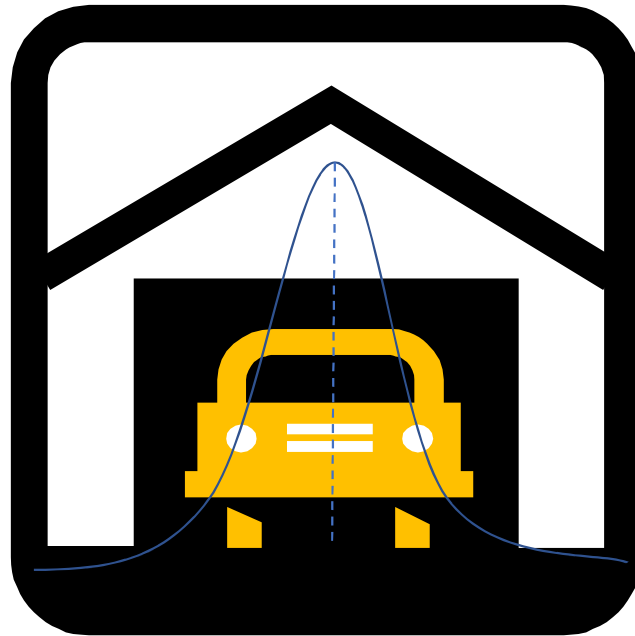
# Process Spread: Too Much Variation



## Process is Not Centered: Creating Defects



Ideal: Centered and On Target



LSL=59.75"

T=60"

USL=60.25"

# Capability Analysis

- Is my process capable of meeting specifications?
- Compare **Voice of the Customer (Specifications) [USL-LSL]** to **Voice of the Process (Process Distribution) [6 $\sigma$ ]**
- Specifications:

Internal, External; one-sided, two-sided

$$\hat{\sigma}_{\text{short-term}} = \frac{\bar{R}}{d_2}$$

$$\hat{\sigma}_{\text{long-term}} = \sqrt{\frac{\sum_{i=0}^N (X_i - \bar{X})^2}{N-1}}$$

$$C_p = \frac{(USL - LSL)}{6 * \hat{\sigma}_{\text{short-term}}}$$

$$P_p = \frac{(USL - LSL)}{6 * \hat{\sigma}_{\text{long-term}}}$$

$$Cp_k = \frac{\text{Min}(USL - \bar{x}, \bar{x} - LSL)}{3 * \hat{\sigma}_{\text{short-term}}}$$

$$Pp_k = \frac{\text{Min}(USL - \bar{x}, \bar{x} - LSL)}{3 * \hat{\sigma}_{\text{long-term}}}$$

# What are Good Process Capability Indices?

- It varies by industry, but generally anything higher than 1 is acceptable

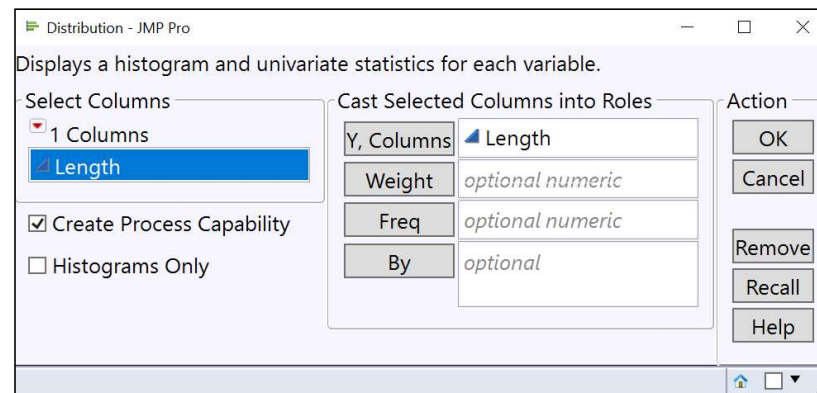
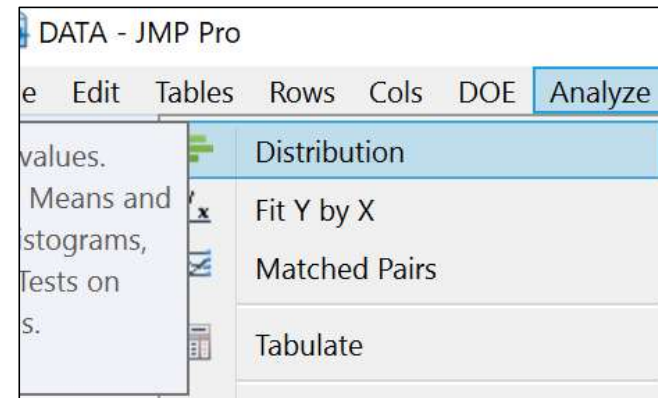
Cp	Parts Per Million Defects
1.00	66,813
1.33	6,210
1.50	1,350
1.67	233
1.83	32
2.00	3.4

# Assumptions

- Stable(No special cause variation)
- Process is normally distributed
  - Non-Normal distributions (Weibull, Gamma, Exponential, etc...) can be fit

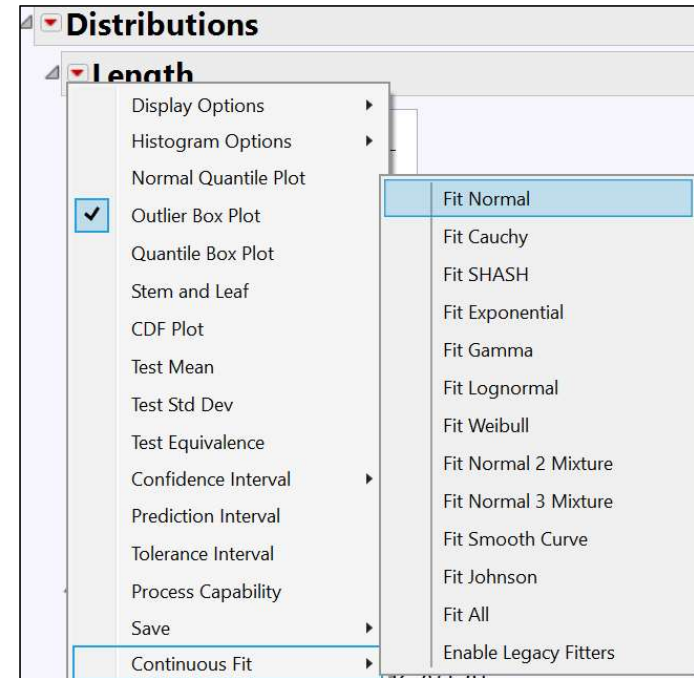
# Process Capability Example

1. Open the data table JMP **Length.jmp**
2. Select **Analyze**→  
**Distribution**
3. Select **Length** for **Y, Columns**
4. Select **OK**.



# Process Capability Example

1. Click on the red triangle beside **Length**
2. Select **Continuous Fit**  
→ **Fit Normal**
3. Click on the red triangle beside **Fitted Normal Distribution** → Select **Goodness of Fit**



The screenshot shows the 'Fitted Normal Distribution' table with the 'Goodness of Fit' option selected. The table displays the following data:

	Std Error	Lower 95
0.0060436	1.337104	
0.0175152	0.034244	

Tests the goodness of fit for this distribution against the data.

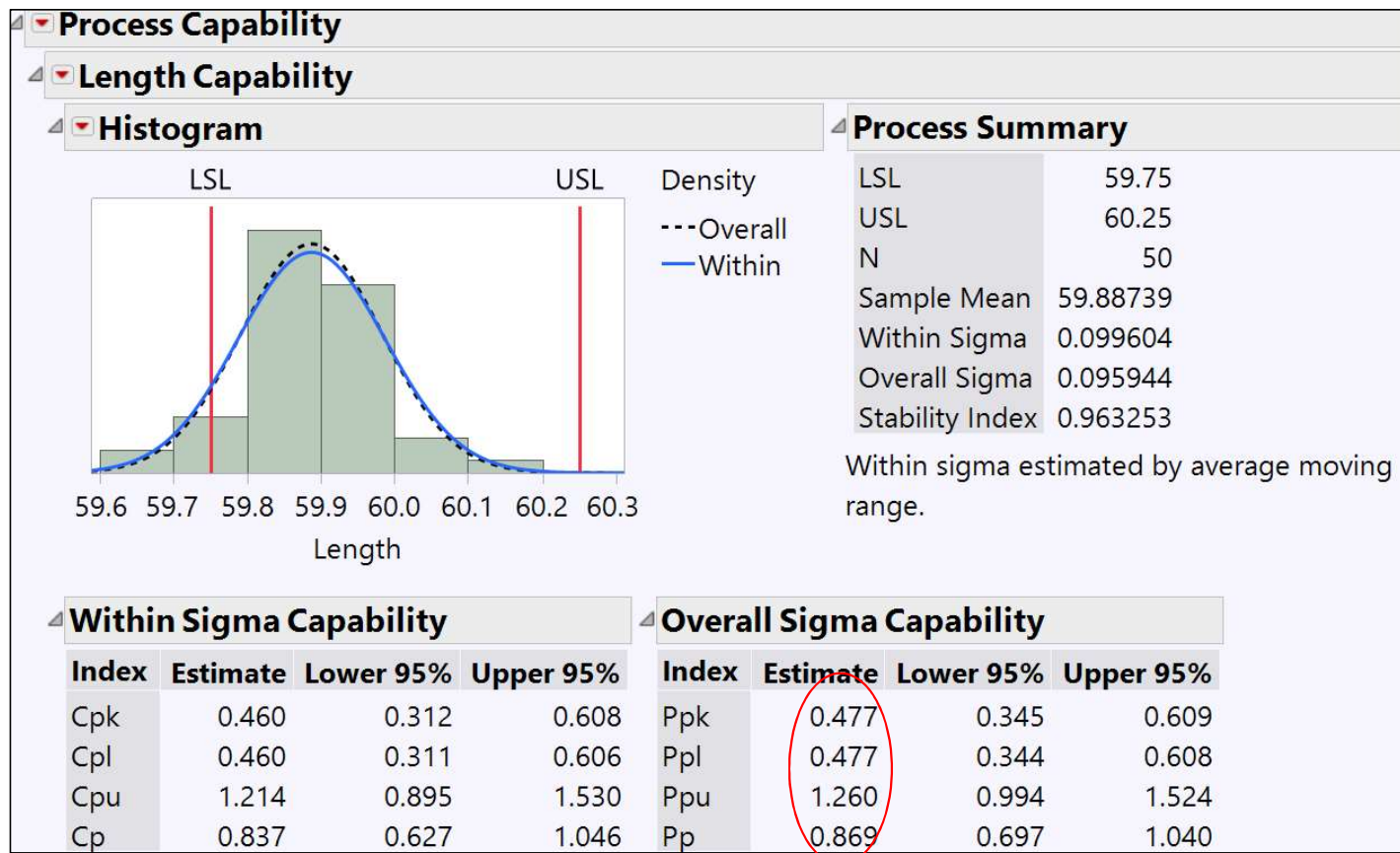


# Example: Length

Fitted Normal Distribution					
Parameter		Estimate	Std Error	Lower 95%	Upper 95%
Location	$\mu$	59.887387	0.0135685	59.860793	59.913981
Dispersion	$\sigma$	0.095944	0.0314502	0.0504657	0.0997013
Measures					
-2*LogLikelihood		-93.50524			
AICc		-89.24992			
BIC		-85.68119			
Goodness-of-Fit Test					
		W	Prob < W		
Shapiro-Wilk	0.9818584		0.6325		
		A2	Prob > A2		
Anderson-Darling	0.3828168		0.3910		

Is the data normal?  
What's next?

# Process Capability Example



Is this process capable?

# Control Charts

# Control Charts

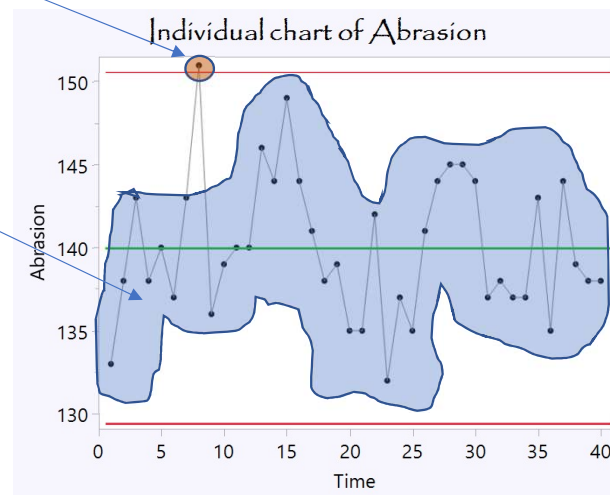
- Control Charts are used to monitor, control, and improve process performance over time
- Sophisticated time run chart
- What do they detect?
  - Changes in process average
  - Changes in process variation
  - Different sources of variation
- Benefits of Control Charts:
  1. Control Limits can distinguish 'signals' (special cause) from 'noise' (common cause)
  2. Allow you to evaluate the stability around the mean
  3. Allows for better decisions regarding process improvements – balance between acting when not necessary (overcontrol) vs. failing to act when necessary
  4. Control Limits are based on what the process is, not what it should be



# Control Limits

- Control Charts use limits to identify signs of trouble, or special cause variation that is not due to chance.
- These limits are expressed as lines plotted above (upper control limit) and below (lower control limit) the central tendency of the process.
- Plot points that fall outside of the limits indicate potential trouble

Special Cause (Assignable)



Upper Control Limit

Common Cause (Chance)

Lower Control Limit

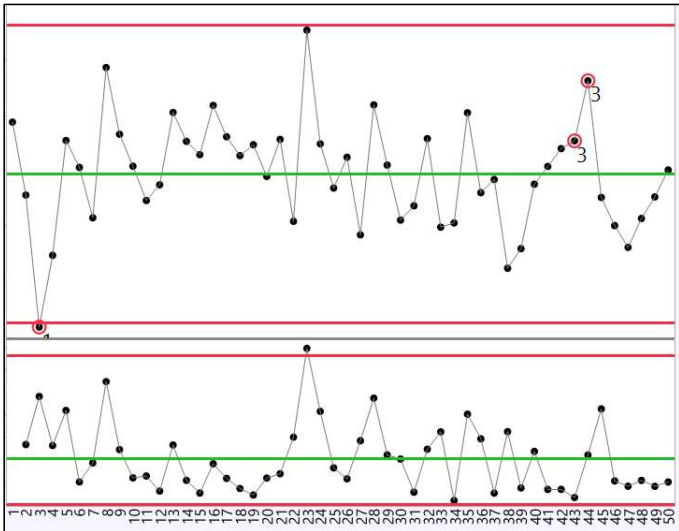
# Western Electric Rules: How they detect changes

<b>Test 1</b>	One point beyond Zone A (upper or lower)	Detects a shift in the mean, an increase in the standard deviation, or a single aberration in the process. For interpreting Test 1, any dispersion chart (R, S, or MR) can be used to rule out increases in variation.	<b>Test 5</b>	Two out of three points in a row in or beyond Zone A and the point itself is in or beyond Zone A; the two points must be on the same side (upper or lower)	Detects a shift in the process average or increase in the standard deviation. Any two out of three points provide a positive test.
<b>Test 2</b>	Nine points in a row in a single (upper or lower) side of Zone C or beyond	Detects a shift in the process mean.	<b>Test 6</b>	Four out of five points in a row in or beyond Zone B and the point itself is in or beyond Zone B; the four points must be on the same side (upper or lower)	Detects a shift in the process mean. Any four out of five points provide a positive test.
<b>Test 3</b>	Six points in a row steadily increasing or decreasing (anywhere on the chart)	Detects a trend or drift in the process mean.	<b>Test 7</b>	Fifteen points in a row in Zone C, above and below the center line	Detects stratification of subgroups when the observations in a single subgroup come from various sources with different means. Also detects a reduction in variation.
<b>Test 4</b>	Fourteen points in a row alternating up and down (anywhere on the chart)	Detects systematic effects such as two alternately used machines, vendors, or operators.	<b>Test 8</b>	Eight points in a row on both sides of the center line with none in Zones C	Detects stratification of subgroups when the observations in one subgroup come from a single source, but subgroups come from different sources with different means.

# What Type of Chart?

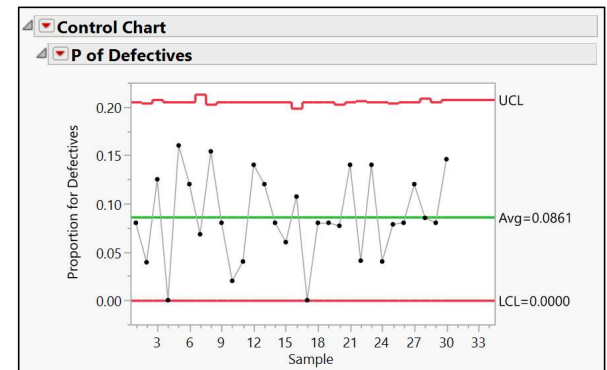
## Continuous

- IR:  $n=1$
- $\bar{X}$ -R:  $2 < n < 9$
- $\bar{X}$ -S:  $n \geq 10$



## Discrete

- np: sample size constant/counting defectives
- p: sample size not constant/counting defectives
- c: sample size constant/counting defects
- u: sample size not constant/counting defects





Observations in Sample, $n$	Chart for Averages			Chart for Standard Deviations						Chart for Ranges						
	Factors for Control Limits			Factors for Center Line		Factors for Control Limits				Factors for Center Line		Factors for Control Limits				
	$A$	$A_2$	$A_3$	$c_4$	$1/c_4$	$B_3$	$B_4$	$B_5$	$B_6$	$d_2$	$1/d_2$	$d_3$	$D_1$	$D_2$	$D_3$	$D_4$
2	2.121	1.880	2.659	0.7979	1.2533	0	3.267	0	2.606	1.128	0.8865	0.853	0	3.686	0	3.267
3	1.732	1.023	1.954	0.8862	1.1284	0	2.568	0	2.276	1.693	0.5907	0.888	0	4.358	0	2.575
4	1.500	0.729	1.628	0.9213	1.0854	0	2.266	0	2.088	2.059	0.4857	0.880	0	4.698	0	2.282
5	1.342	0.577	1.427	0.9400	1.0638	0	2.089	0	1.964	2.326	0.4299	0.864	0	4.918	0	2.115
6	1.225	0.483	1.287	0.9515	1.0510	0.030	1.970	0.029	1.874	2.534	0.3946	0.848	0	5.078	0	2.004
7	1.134	0.419	1.182	0.9594	1.0423	0.118	1.882	0.113	1.806	2.704	0.3698	0.833	0.204	5.204	0.076	1.924
8	1.061	0.373	1.099	0.9650	1.0363	0.185	1.815	0.179	1.751	2.847	0.3512	0.820	0.388	5.306	0.136	1.864
9	1.000	0.337	1.032	0.9693	1.0317	0.239	1.761	0.232	1.707	2.970	0.3367	0.808	0.547	5.393	0.184	1.816
10	0.949	0.308	0.975	0.9727	1.0281	0.284	1.716	0.276	1.669	3.078	0.3249	0.797	0.687	5.469	0.223	1.777
11	0.905	0.285	0.927	0.9754	1.0252	0.321	1.679	0.313	1.637	3.173	0.3152	0.787	0.811	5.535	0.256	1.744
12	0.866	0.266	0.886	0.9776	1.0229	0.354	1.646	0.346	1.610	3.258	0.3069	0.778	0.922	5.594	0.283	1.717
13	0.832	0.249	0.850	0.9794	1.0210	0.382	1.618	0.374	1.585	3.336	0.2998	0.770	1.025	5.647	0.307	1.693
14	0.802	0.235	0.817	0.9810	1.0194	0.406	1.594	0.399	1.563	3.407	0.2935	0.763	1.118	5.696	0.328	1.672
15	0.775	0.223	0.789	0.9823	1.0180	0.428	1.572	0.421	1.544	3.472	0.2880	0.756	1.203	5.741	0.347	1.653
16	0.750	0.212	0.763	0.9835	1.0168	0.448	1.552	0.440	1.526	3.532	0.2831	0.750	1.282	5.782	0.363	1.637
17	0.728	0.203	0.739	0.9845	1.0157	0.466	1.534	0.458	1.511	3.588	0.2787	0.744	1.356	5.820	0.378	1.622
18	0.707	0.194	0.718	0.9854	1.0148	0.482	1.518	0.475	1.496	3.640	0.2747	0.739	1.424	5.856	0.391	1.608
19	0.688	0.187	0.698	0.9862	1.0140	0.497	1.503	0.490	1.483	3.689	0.2711	0.734	1.487	5.891	0.403	1.597
20	0.671	0.180	0.680	0.9869	1.0133	0.510	1.490	0.504	1.470	3.735	0.2677	0.729	1.549	5.921	0.415	1.585
21	0.655	0.173	0.663	0.9876	1.0126	0.523	1.477	0.516	1.459	3.778	0.2647	0.724	1.605	5.951	0.425	1.575
22	0.640	0.167	0.647	0.9882	1.0119	0.534	1.466	0.528	1.448	3.819	0.2618	0.720	1.659	5.979	0.434	1.566
23	0.626	0.162	0.633	0.9887	1.0114	0.545	1.455	0.539	1.438	3.858	0.2592	0.716	1.710	6.006	0.443	1.557
24	0.612	0.157	0.619	0.9892	1.0109	0.555	1.445	0.549	1.429	3.895	0.2567	0.712	1.759	6.031	0.451	1.548
25	0.600	0.153	0.606	0.9896	1.0105	0.565	1.435	0.559	1.420	3.931	0.2544	0.708	1.806	6.056	0.459	1.541

For  $n > 25$ .

*Introduction to Statistical Quality Control, 4<sup>th</sup> Ed. Montgomery, pg. 761*

$$A = \frac{3}{\sqrt{n}} \quad A_3 = \frac{3}{c_4 \sqrt{n}} \quad c_4 \approx \frac{4(n-1)}{4n-3}$$

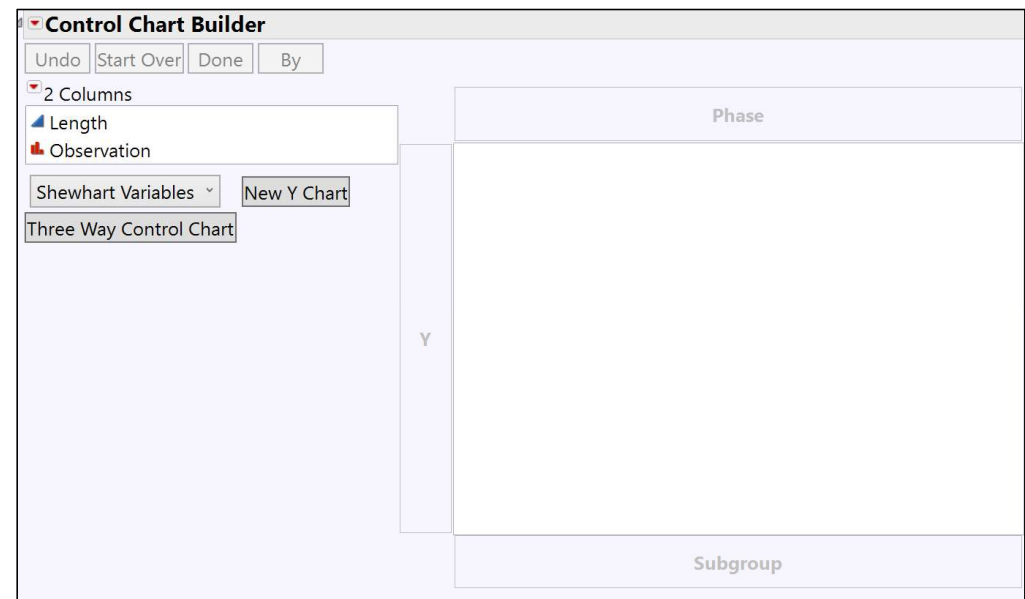
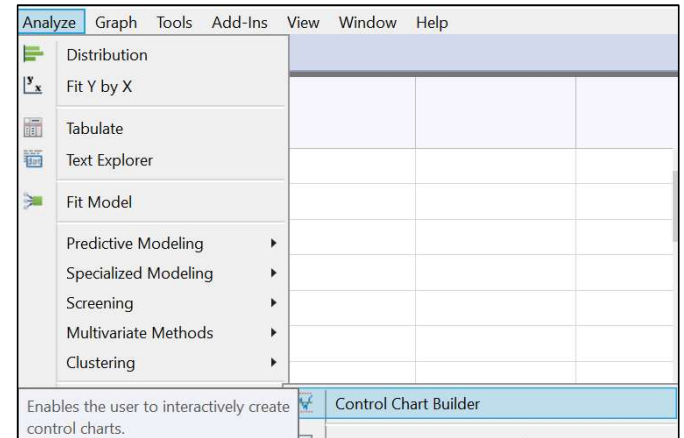
$$B_3 = 1 - \frac{3}{c_4 \sqrt{2(n-1)}} \quad B_4 = 1 + \frac{3}{c_4 \sqrt{2(n-1)}}$$

$$B_5 = c_4 - \frac{3}{\sqrt{2(n-1)}} \quad B_6 = c_4 + \frac{3}{\sqrt{2(n-1)}}$$



## Example-Cont'd

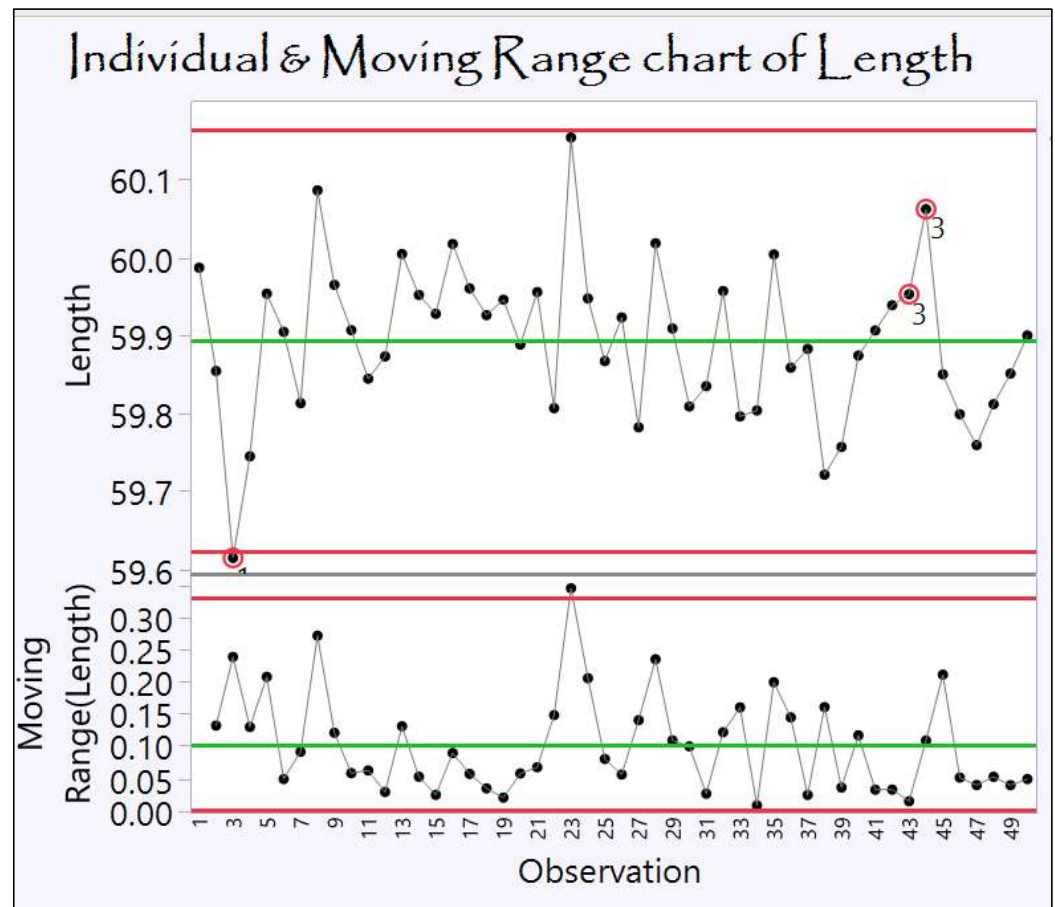
1. Open **Length.jmp**.
2. Select **Analyze** → **Quality and Process** → **Control Chart Builder**.
3. Drag **Length** for **Y** and **Observation** to **Subgroup**.
4. Select **OK**.



# Example

1. Right click on top chart and select **Warnings**→**Tests**→**All Tests**.
2. Click **Done**.

Is the process in control?



# Example

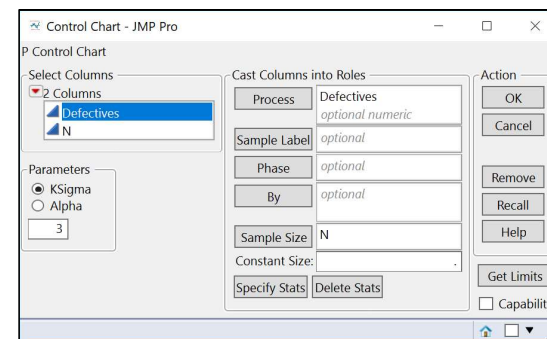
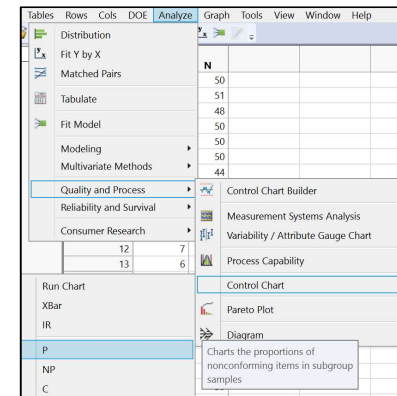
You decide to count the number of product returns over the last month from varying production lot sizes.

Which control chart would you use? Is it in control?

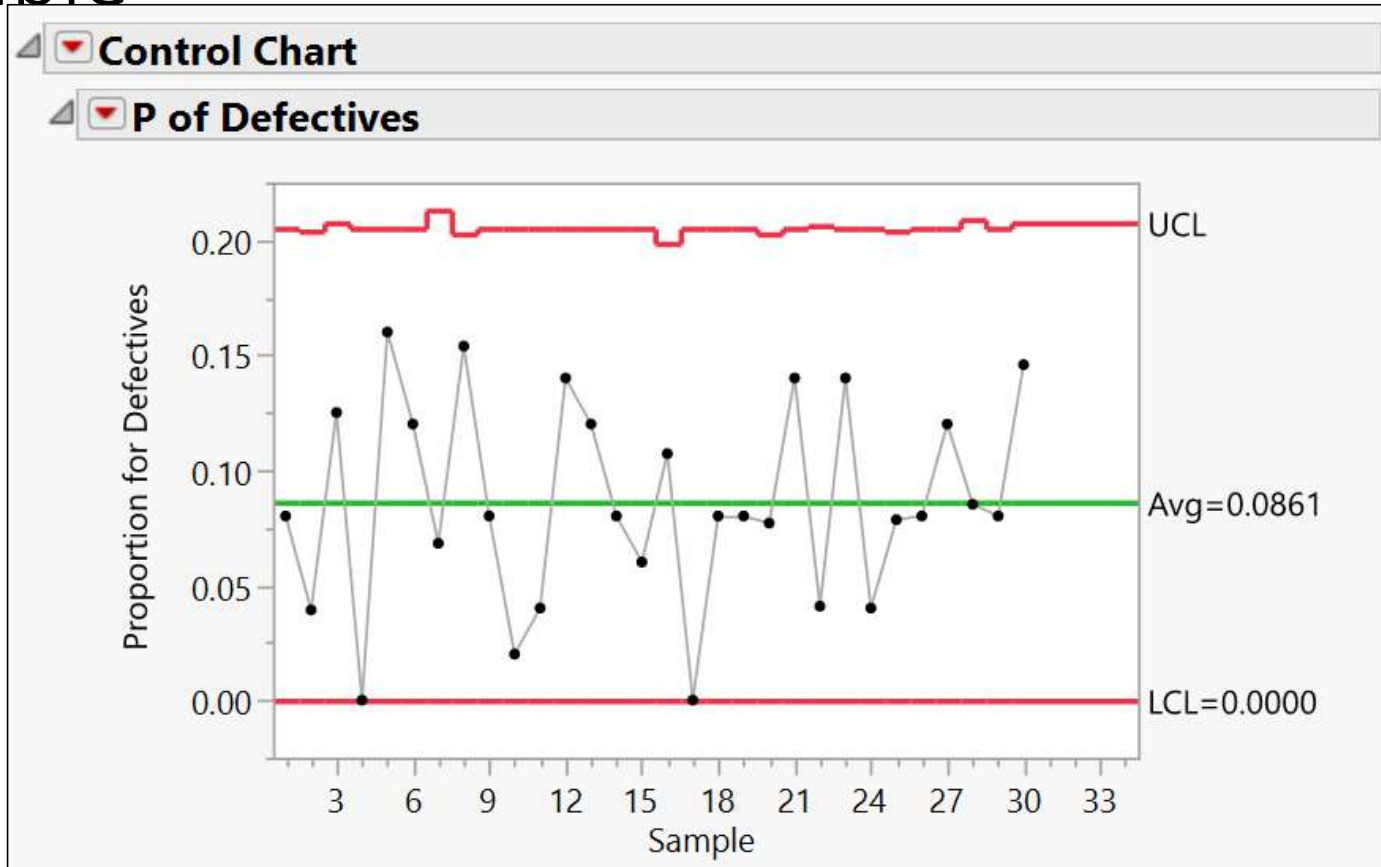
Open **Returns.jmp**

## Example-Cont'd

1. Open **Returns.xlsx**
2. Select **Analyze**→**Quality and Process**→**Control Chart** →**P**.
3. Select **Defectives** for **Process**.
4. Select **N** for **Sample Size**.
5. Select **OK**.



# Example



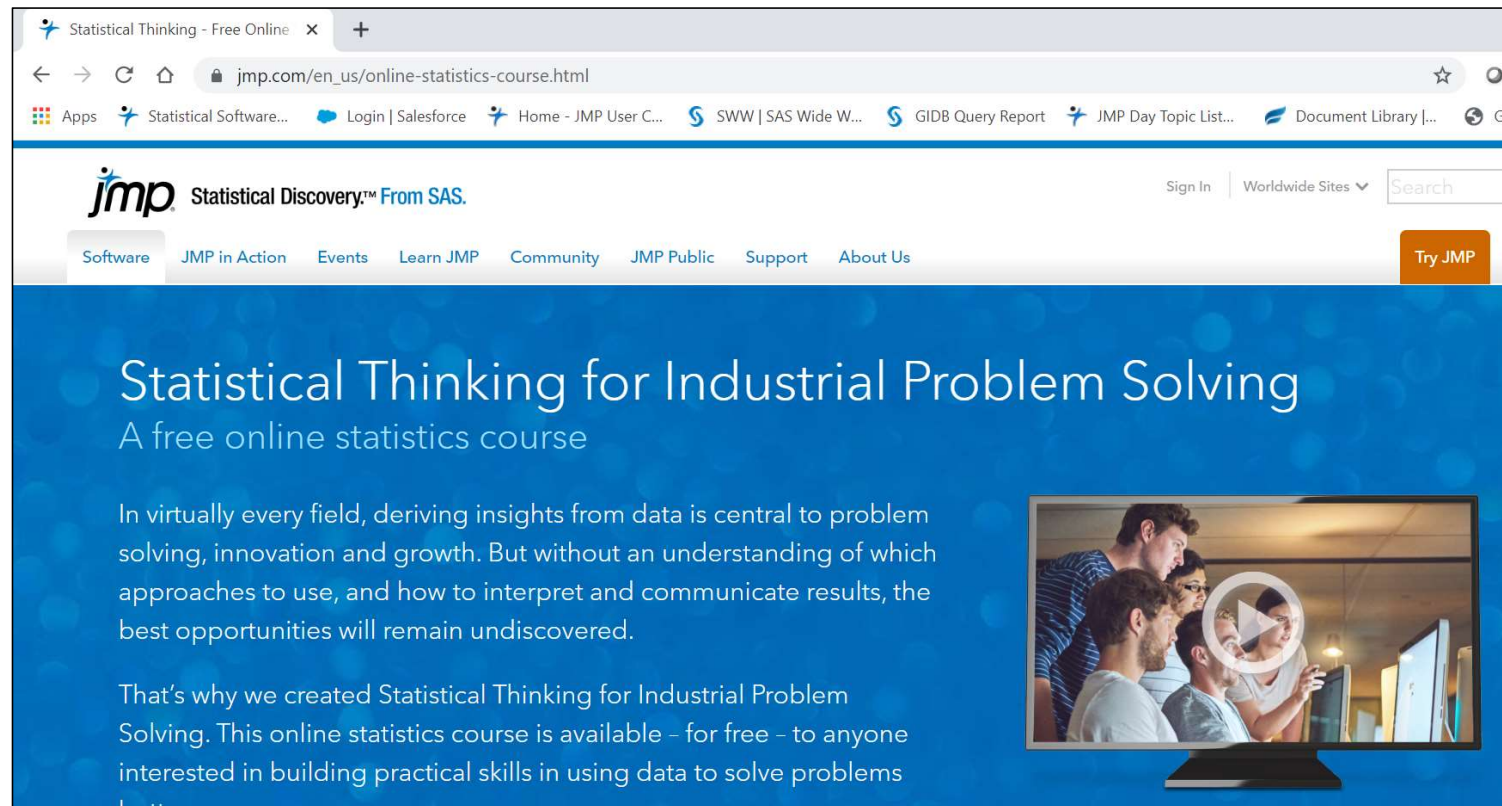
# Summary

- SPC looks at variation
- MSA assesses the measurement system
- Process capability assesses whether meeting customer specifications
- Control charts identify whether process is stable and in control

Just in case you were wondering....



For more information, please check out the free SAS Training course, [Statistical Thinking for Industrial Problem Solving](https://www.jmp.com/en_us/online-statistics-course.html)



The screenshot shows a web browser window with the URL [jmp.com/en\\_us/online-statistics-course.html](https://www.jmp.com/en_us/online-statistics-course.html). The page features the JMP logo and the tagline "Statistical Discovery™ From SAS." in the top left. A navigation menu includes links for Software, JMP in Action, Events, Learn JMP, Community, JMP Public, Support, and About Us. A "Try JMP" button is located in the top right. The main content area has a blue background with the title "Statistical Thinking for Industrial Problem Solving" and the subtitle "A free online statistics course". Below this, there is a paragraph of text: "In virtually every field, deriving insights from data is central to problem solving, innovation and growth. But without an understanding of which approaches to use, and how to interpret and communicate results, the best opportunities will remain undiscovered." This is followed by another paragraph: "That's why we created Statistical Thinking for Industrial Problem Solving. This online statistics course is available - for free - to anyone interested in building practical skills in using data to solve problems better." On the right side of the page, there is a video player showing a group of people looking at a screen, with a play button icon overlaid on the video.