



**Universität
Zürich** ^{UZH}

Department of Business Administration

Services & Operations Management

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Module Overview

1. Operations strategy
2. Process analytics
- 3. Quality management: SPC**
4. Platform management
5. Sport management



Learning Goals (1/2)

After this lecture you should

- understand the importance of reliable quality
- know how reliable quality can be produced
- be familiar with the methods of Statistical Process Control (SPC)
- be able to calculate upper and lower control limits
- know whether a process is under control or not
- be able to bring an out-of-control process under control



Learning Goals (2/2)

- be able to apply SPC in real time
- understand the cost effects of quality management decisions
- understand the differences between control, performance and specification limits
- be able to calculate process capability indices for symmetric and asymmetric processes



Quality: Consumer vs. Producer

Consumer:

- Fulfilling/exceeding expectations
- Usability
- Serviceability
- Fulfilling/exceeding product-/service standards

Producer:

- Adherence to product-/service specifications

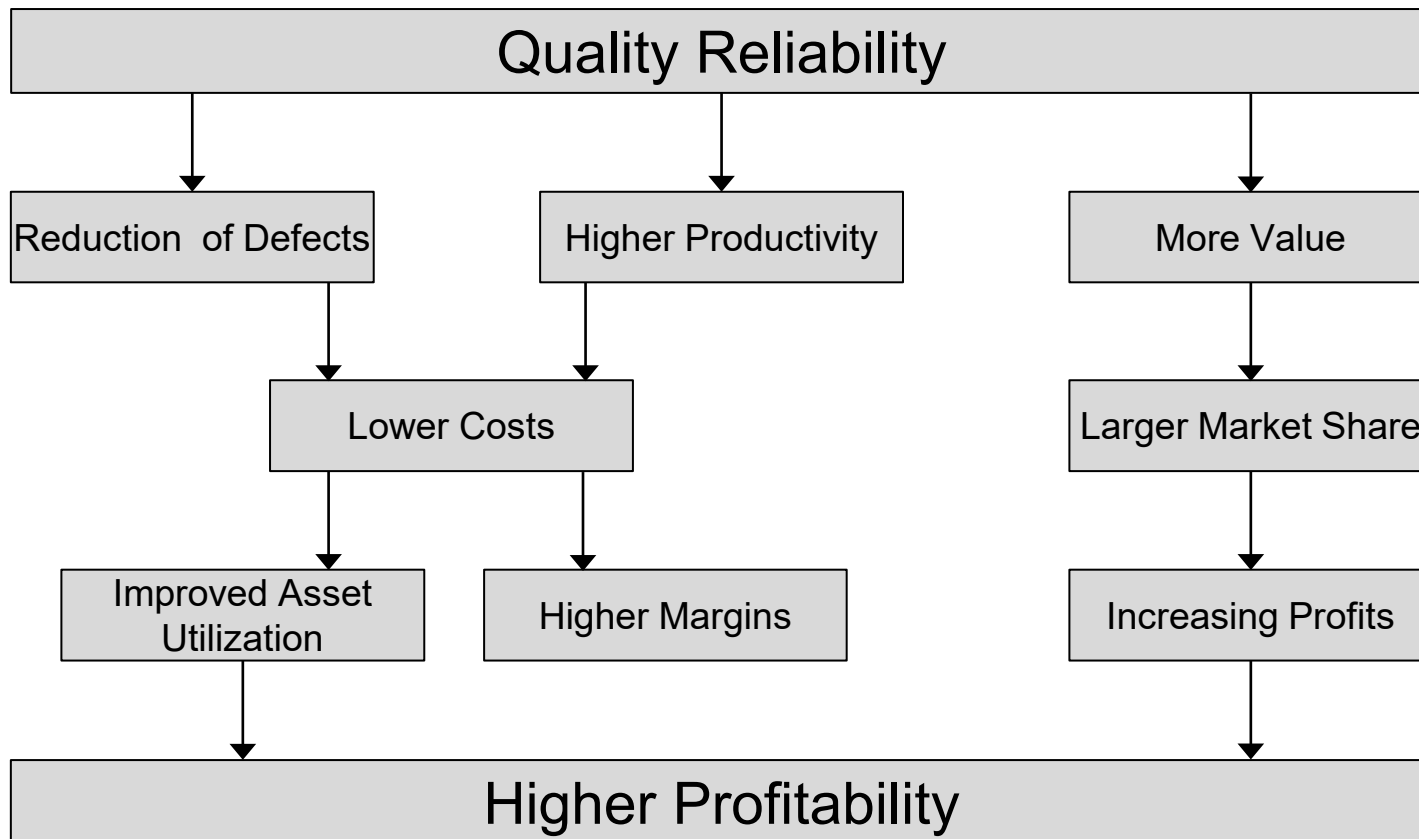


How is Quality Realized?

- Translate customer needs into product/service characteristics
 - Example (Product): gas mileage demanded by customers
 - Example (Service): medical results demanded by patients
- Translate product/service characteristics into product/service specifications
 - Example (Product): weight, wind resistance, etc.
 - Example (Service): 1-, 3-, 10-year survival rates, side effects, etc.
- Design a production system that realizes these product/service specifications



Quality and Profitability





Why is Quality Reliability so Important?

Internal Costs

- Correction of defects
- Inventory cost
- Capacity consumption
- Disruption of production

External Costs

- Loss of reputation and brand-name
- Liability costs (manufacturer liability, legal costs, sanctions, penalties)
- Warrantee costs
- Price reductions

Prevention and Detection Costs

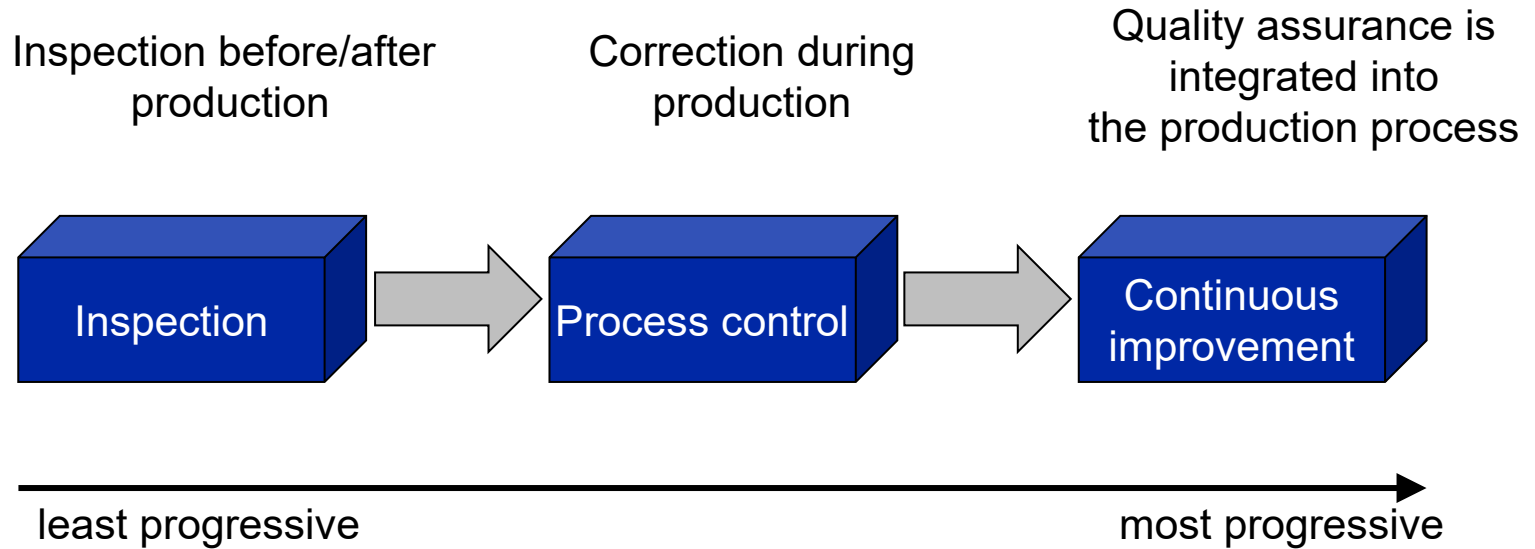
- Monitoring costs
- Inspection costs
- Error diagnosis/fault tracking

Competitive advantages

- Cost reduction for customers
- Profiting from customers' risk aversion (e.g. Disney)
- Brand loyalty/franchising



Quality Assurance





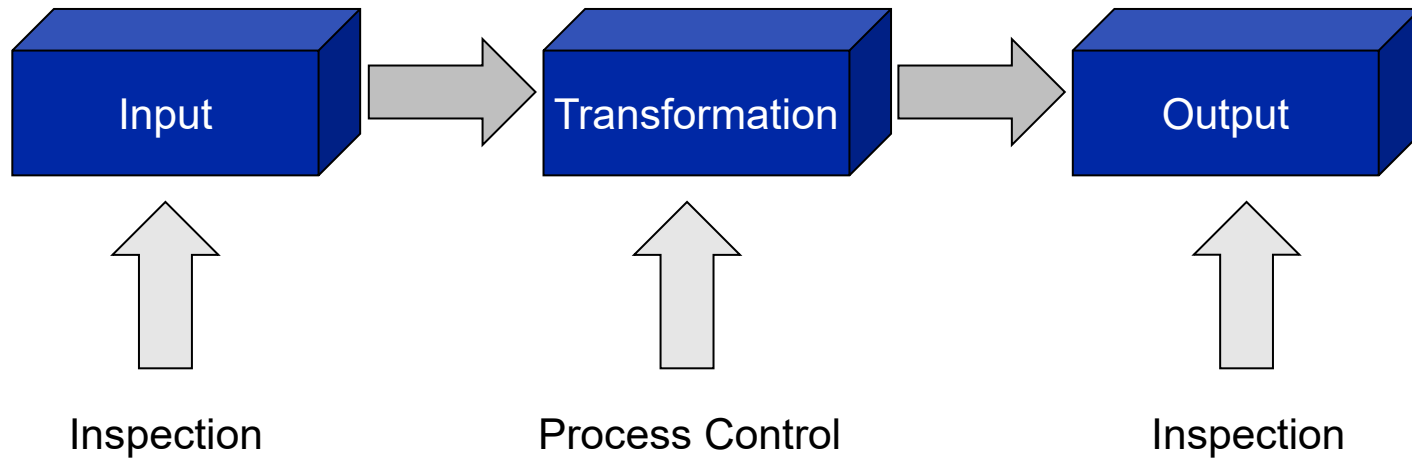
Quality Management Objectives

- Production of reliable quality
- Identification and solution of quality problems
- Minimizing costs

How do we achieve these objectives?



Inspection vs. Process Control





Option 1: Inspection

Idea: Eliminate wrong quality *before* it reaches the customer

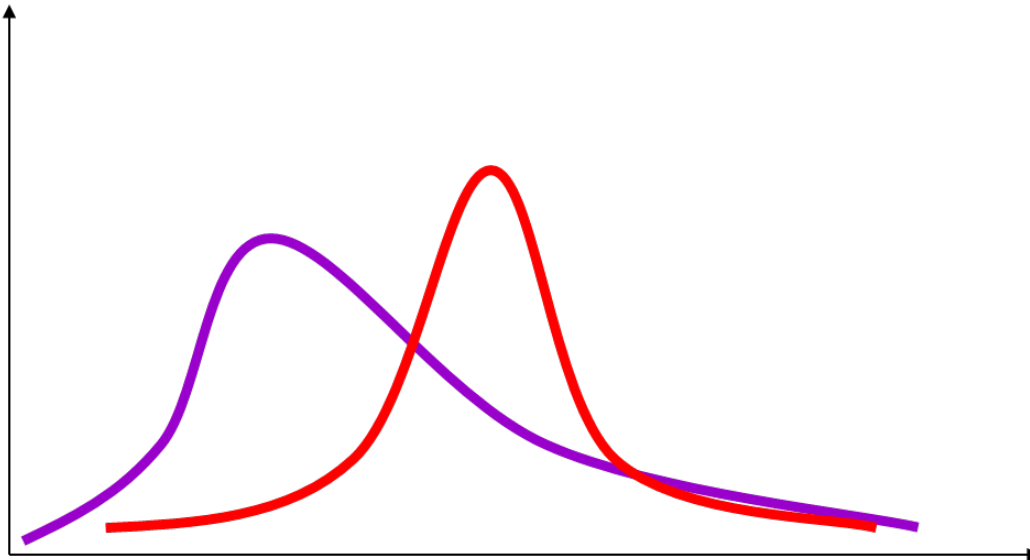
Inspection can be very expansive

- Inspection costs
 - Direct: Inspection personnel, equipment
 - Indirect: Scrap, lost capacity
- Not suitable for industries with
 - Low margins
 - integral product architecture
 - high opportunity costs of production capacity



Central Limit Theorem

The Central Limit Theorem states that the sampling distribution of the sampling means approaches a normal distribution as the sample size gets larger — no matter what the shape of the population distribution.





Samples Inspection: Why?

- In most cases full (100%) inspection is too expensive (large production volumes)
- Often full (100%) inspection is impossible, e.g., if inspection consumes or destroys the product or service (for example: taster in restaurants, bomb testing)
- Often, inspection by the producer is cheaper than by the customer due to economies of scale (fixed costs of inspection)



Option 2: Process Control

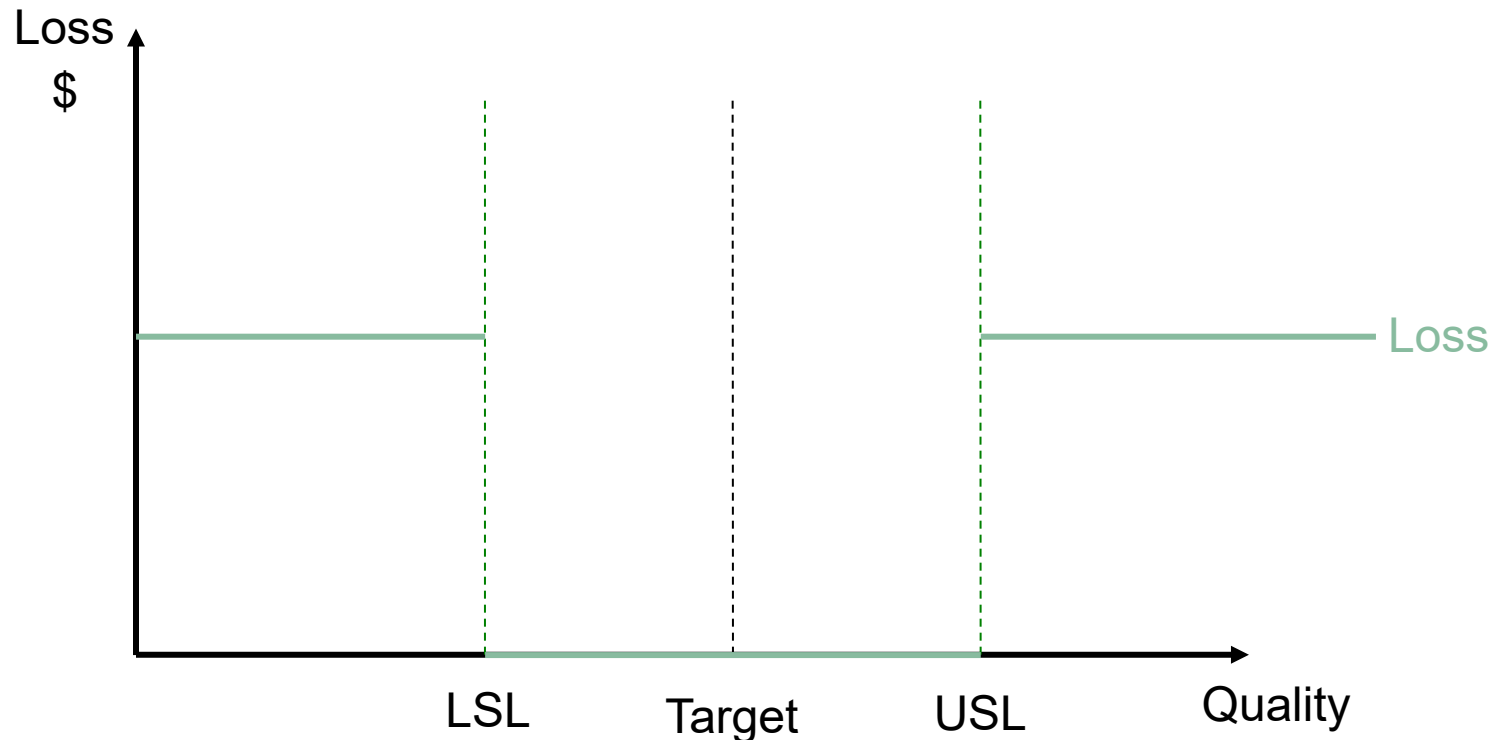
Basic Idea: Control the process which produces quality

- SPC (Statistical Process Control)
- Control of quality dimensions (not only „good“ vs. „bad“ output)
 - How do values change over time?
 - If a product or service is of bad quality how far are the values from the desired quality level?



“Goal Post” Philosophy

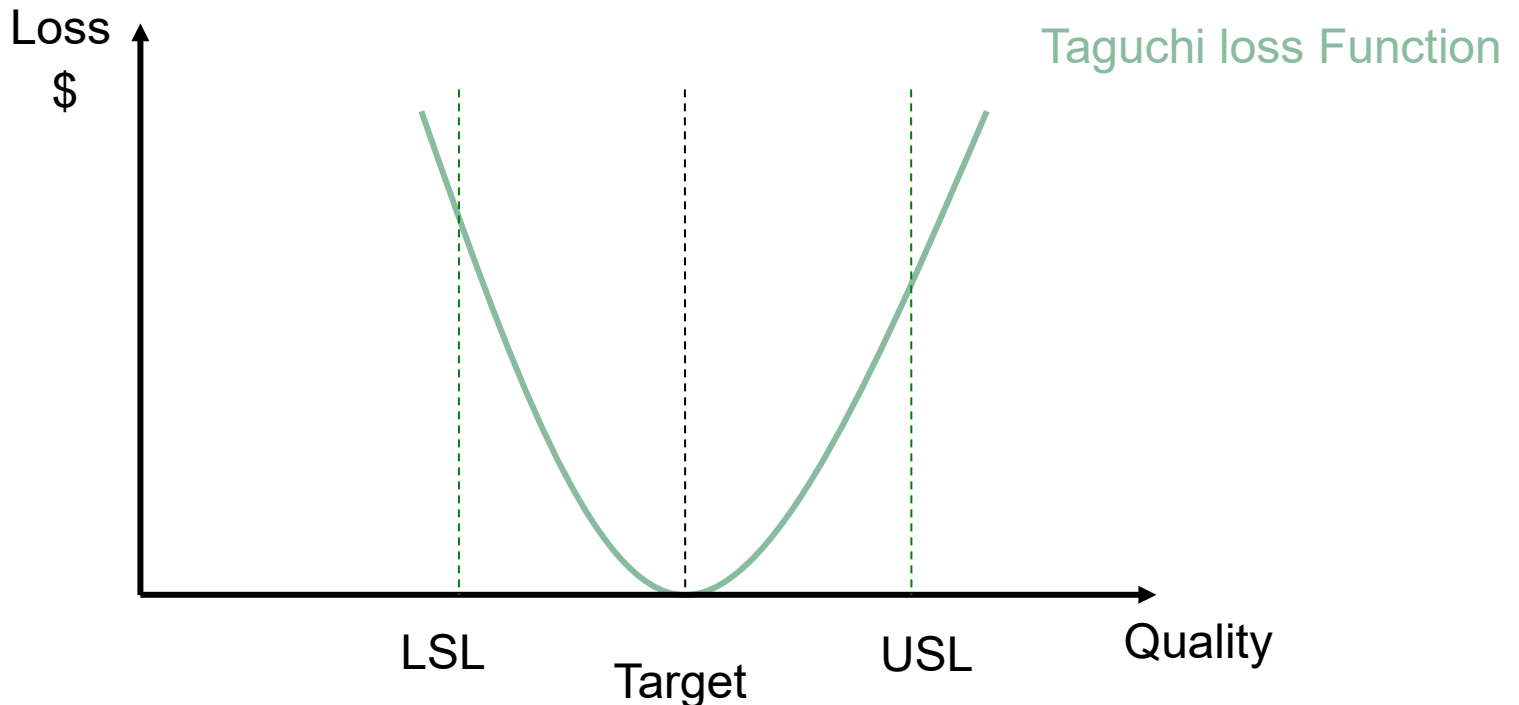
There is no quality problem as long as the quality stays within the lower (LSL) and upper specification limit (USL). If quality is outside the specification limits the product/service is defect/scrap





Taguchi Loss Function

The customer experiences a loss of quality the moment quality deviates from its target value. This loss L is depicted by a quality loss function and it follows a parabolic curve mathematically given by $L = k(y-m)^2$, where m is the theoretical target value, y is the actual quality and k is a parameter



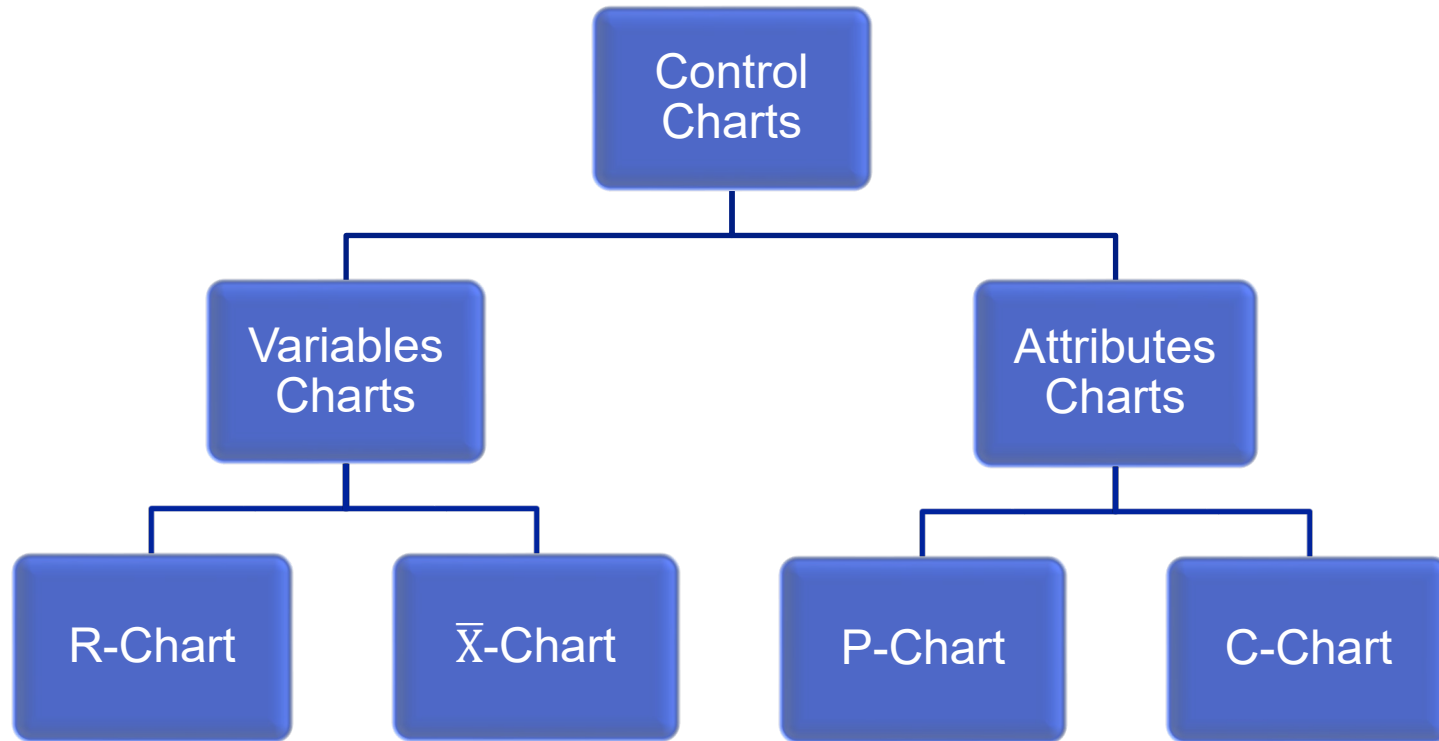


Statistical Process Control (SPC)

- SPC identifies the causes for process variations
 - General causes (lead to random variations)
 - affect the entire output
 - are process-immanent
 - avoidance requires new process design
 - Special causes (lead to systematic variations)
 - affect only part of the output
 - are based on avoidable errors (e.g., human failure)
 - avoidance does not require a new process design
- SPC determines process capability
 - Which quality level can be reliably achieved by the process?

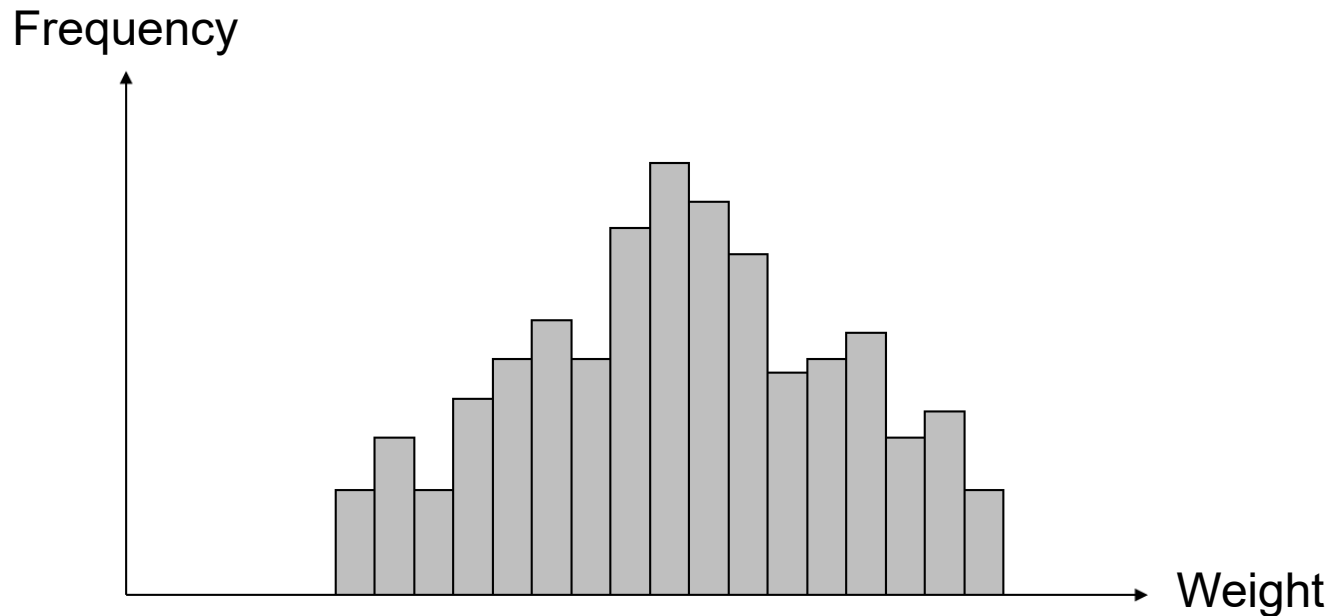


Control Charts





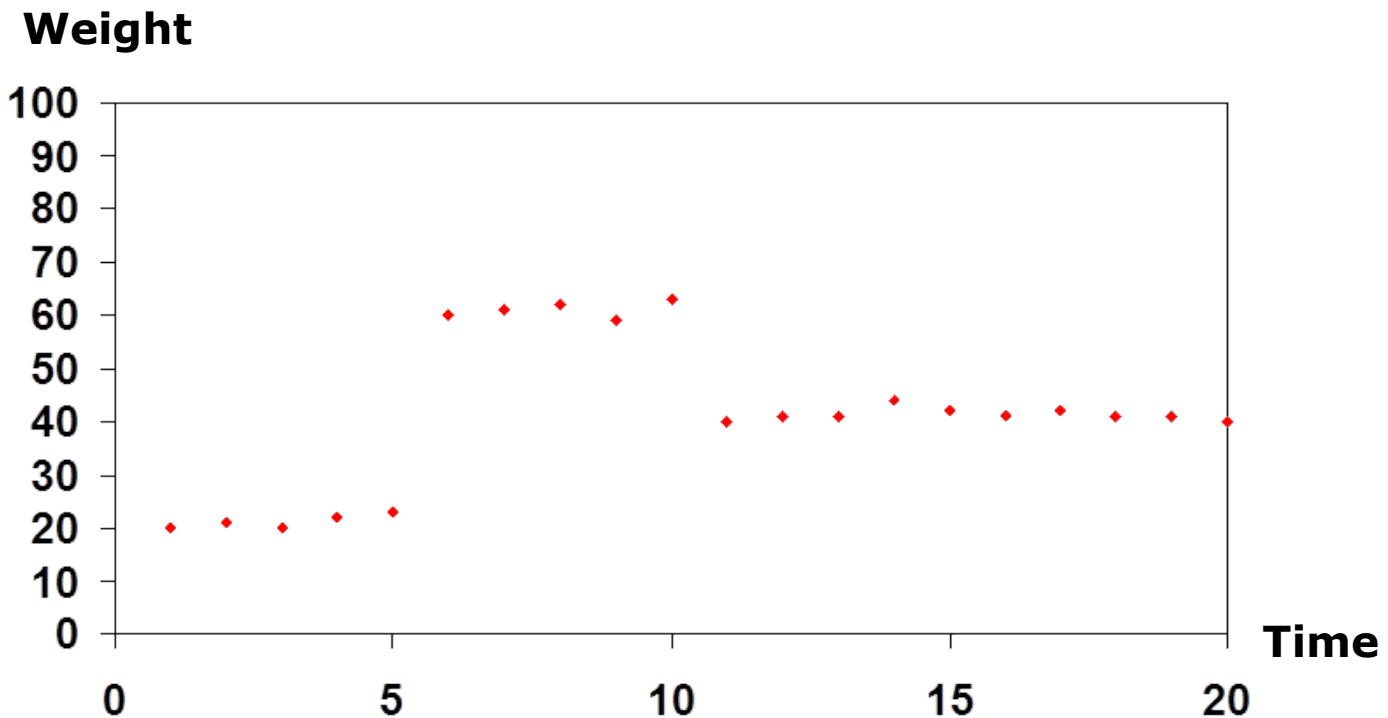
Why Histograms are Problematic



Problem: Histograms do not report variations over time



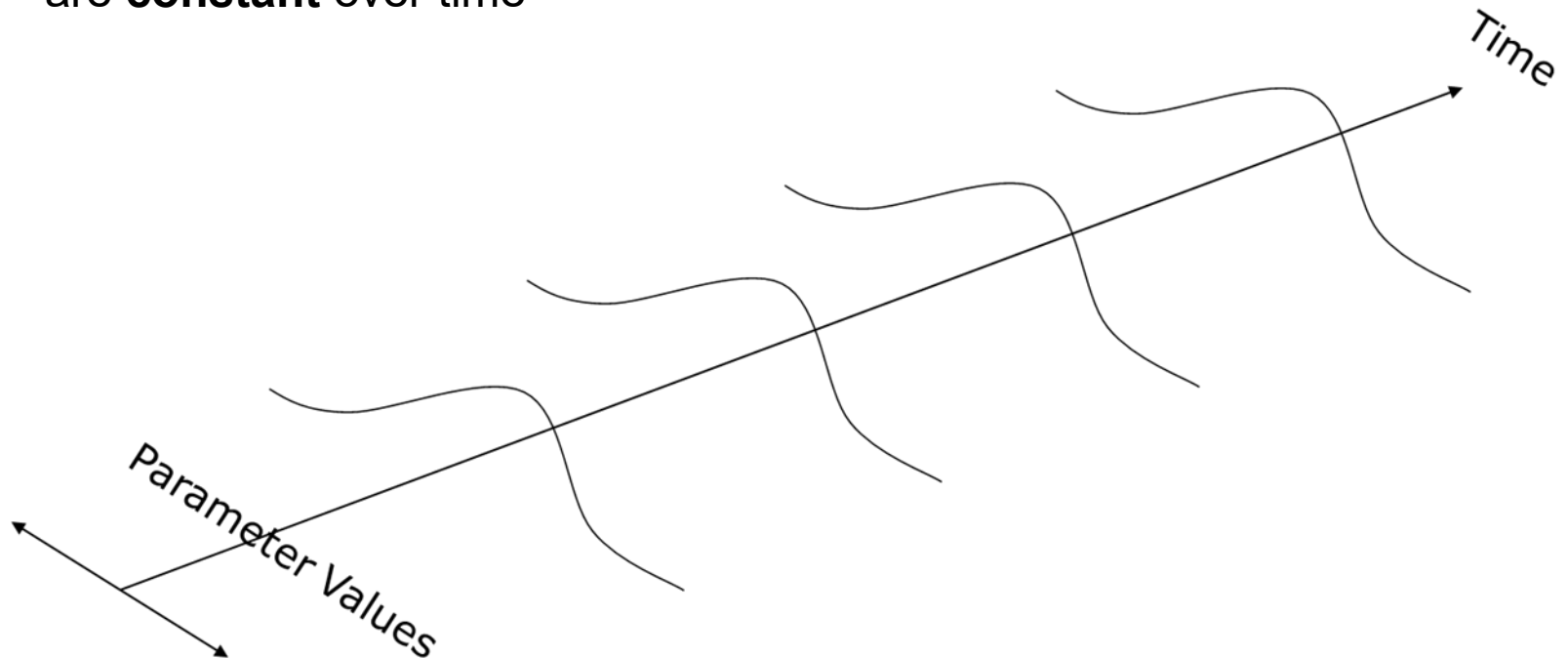
SPC Analyzes Quality Variations over Time





The Concept of SPC (1/2)

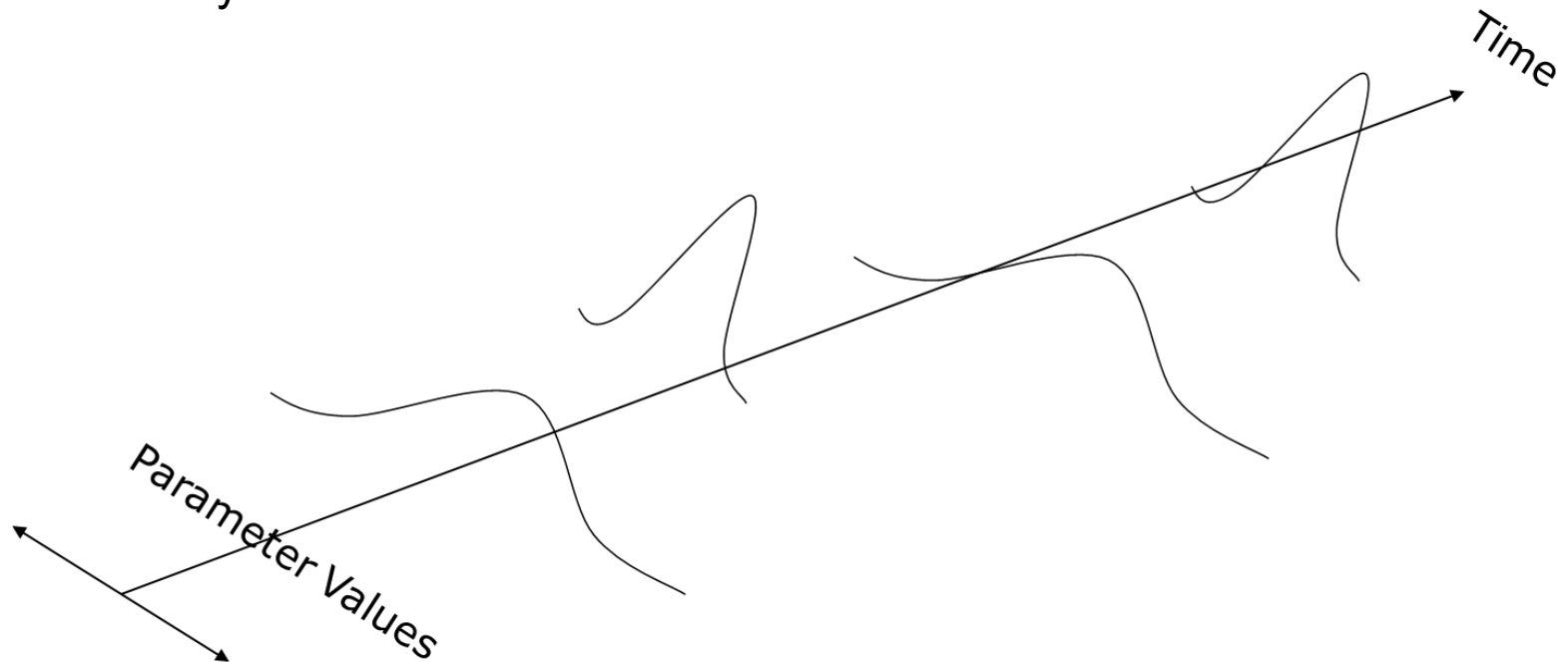
This process is statistically under control because its parameter values are **constant** over time





The Concept of SPC (2/2)

This process is statistically **not** under control because its parameter values vary over time





Control Charts: Tasks

Control Charts determine whether a process is statistically under control

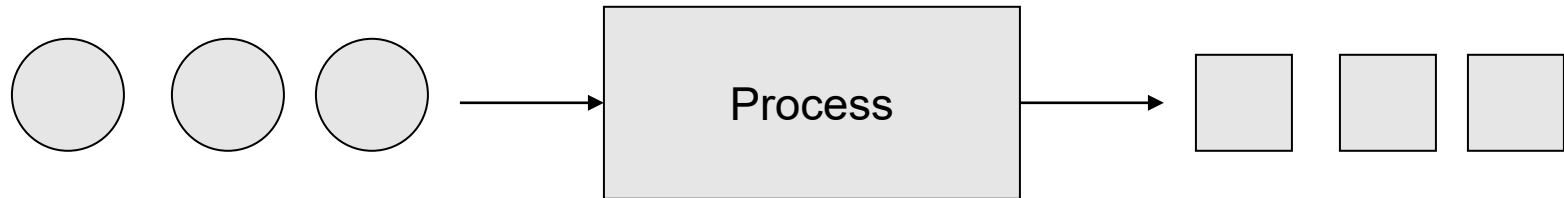
and

identify the causes of quality variations

and

monitor the production process

Data Collection for Control Charts



How to construct samples

- **Goal:**
 - Minimize quality variation within each sample
 - Maximize quality variation across samples
- **Criteria:**
 - Constant environmental conditions within a sample
 - Constant materials within a sample
 - Constant personnel (e.g. one shift) within a sample

Idea: If quality variations have special causes, each sample is affected differently



Control Charts: Symbols

n = Sample size

μ = Mean

σ = Standard deviation

\bar{X} = Sample mean

$\bar{\bar{X}}$ = Average mean (mean of sample means)

R = Sample range

\bar{R} = Average range (mean of sample ranges)



Control Charts: \bar{X} – Chart and R – Chart

\bar{X} – Chart

Shows whether a process is under control with respect to its means

- Control limits if parameters are known: $\bar{\bar{X}} \pm 3 \frac{\sigma}{\sqrt{n}}$
- Control limits if parameters are unknown: $\bar{\bar{X}} \pm A_2 \bar{R}$

R – Chart

Shows whether a process is under control with respect to its variations

- Upper control limit (UCL): $D_4 \bar{R}$
- Lower control limit (LCL): $D_3 \bar{R}$



n	A₂	D₃	D₄
2	1.88	0	3.27
3	1.02	0	2.57
4	0.73	0	2.28
5	0.58	0	2.11
6	0.48	0	2.00
7	0.42	0.08	1.92
8	0.37	0.14	1.86
9	0.34	0.18	1.82
10	0.31	0.22	1.78

Source: Grant E.L. (1988): Statistical Quality Control, 6. Aufl.



n	A₂	D₃	D₄
11	0.29	0.26	1.74
12	0.27	0.28	1.72
13	0.25	0.31	1.69
14	0.24	0.33	1.67
15	0.22	0.35	1.65
16	0.21	0.36	1.64
17	0.20	0.38	1.62
18	0.19	0.39	1.61
19	0.19	0.40	1.60
20	0.18	0.41	1.59



Example 1: Diameter

- Diameter, standard deviation = 0.09 nm
- Table shows results of 5 samples (sample size = 4)
- Is process under control?

Sample	Observations				Mean	Range
	1	2	3	4		
1	0.51	0.63	0.39	0.35	0.47	0.28
2	0.50	0.56	0.42	0.64	0.53	0.22
3	0.68	0.49	0.53	0.62	0.58	0.19
4	0.45	0.33	0.47	0.55	0.45	0.22
5	0.70	0.58	0.64	0.68	0.65	0.12

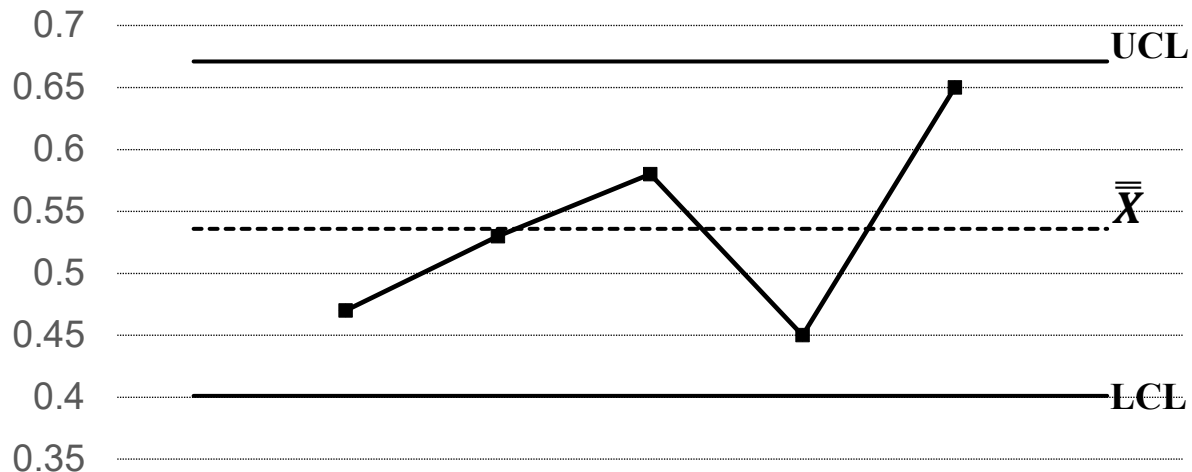


Example 1: \bar{X} – Chart

\bar{X} – Chart

- $\bar{\bar{X}} = \frac{0.47+0.53+0.58+0.45+0.65}{5} = 0.536$
- $UCL = 0.536 + 3 * \left(\frac{0.09}{\sqrt{4}}\right) = 0.536 + 0.135 = 0.671$
- $LCL = 0.536 - 0.135 = 0.401$

→ Process is under control with respect to its means



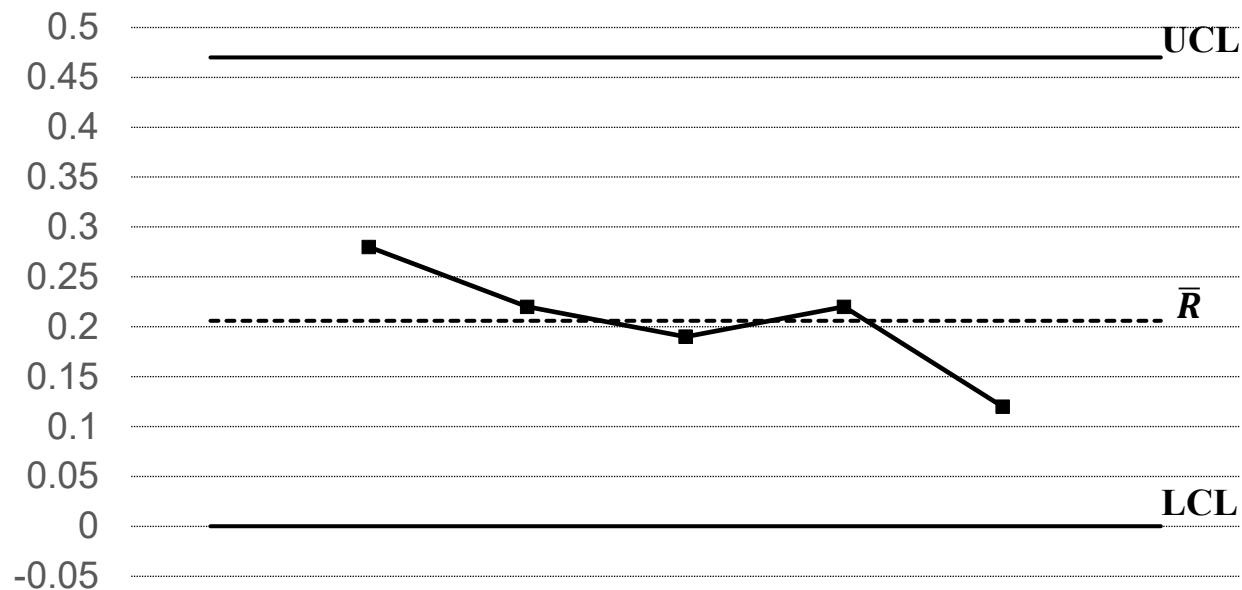


Example 1: R-Chart

R – Chart

- $\bar{R} = \frac{0.28+0.22+0.19+0.22+0.12}{5} = 0.206$
- $UCL = 2.28 * 0.206 = 0.47$
- $LCL = 0 * 0.206 = 0$

→ Process is under control with respect to its range





Example 2: Abrasion

- Tire abrasion in nm, standard deviation unknown
- 20 samples à 10 tires (see Table)
- Is the process under control?

Sample	Average	Range	Sample	Average	Range
1	95.72	1.0	11	95.80	0.6
2	95.24	0.9	12	95.22	0.2
3	95.18	0.8	13	95.56	1.3
4	95.44	0.4	14	95.22	0.5
5	95.46	0.5	15	95.04	0.8
6	95.32	1.1	16	95.72	1.1
7	95.40	0.9	17	94.82	0.6
8	95.44	0.3	18	95.46	0.5
9	95.08	0.2	19	95.60	0.4
10	95.50	0.6	20	95.74	0.6



Example 2: Control Limits

- $\bar{\bar{X}} = 95.398$
- $\bar{R} = 0.665$
- $UCL (\bar{X} - \text{Chart}) = 95.398 + 0.31 * 0.665 = 95.60$
- $LCL (\bar{X} - \text{Chart}) = 95.398 - 0.31 * 0.665 = 95.19$
- $UCL (R - \text{Chart}) = 1.78 * 0.665 = 1.18$
- $LCL (R - \text{Chart}) = 0.22 * 0.665 = 0.15$



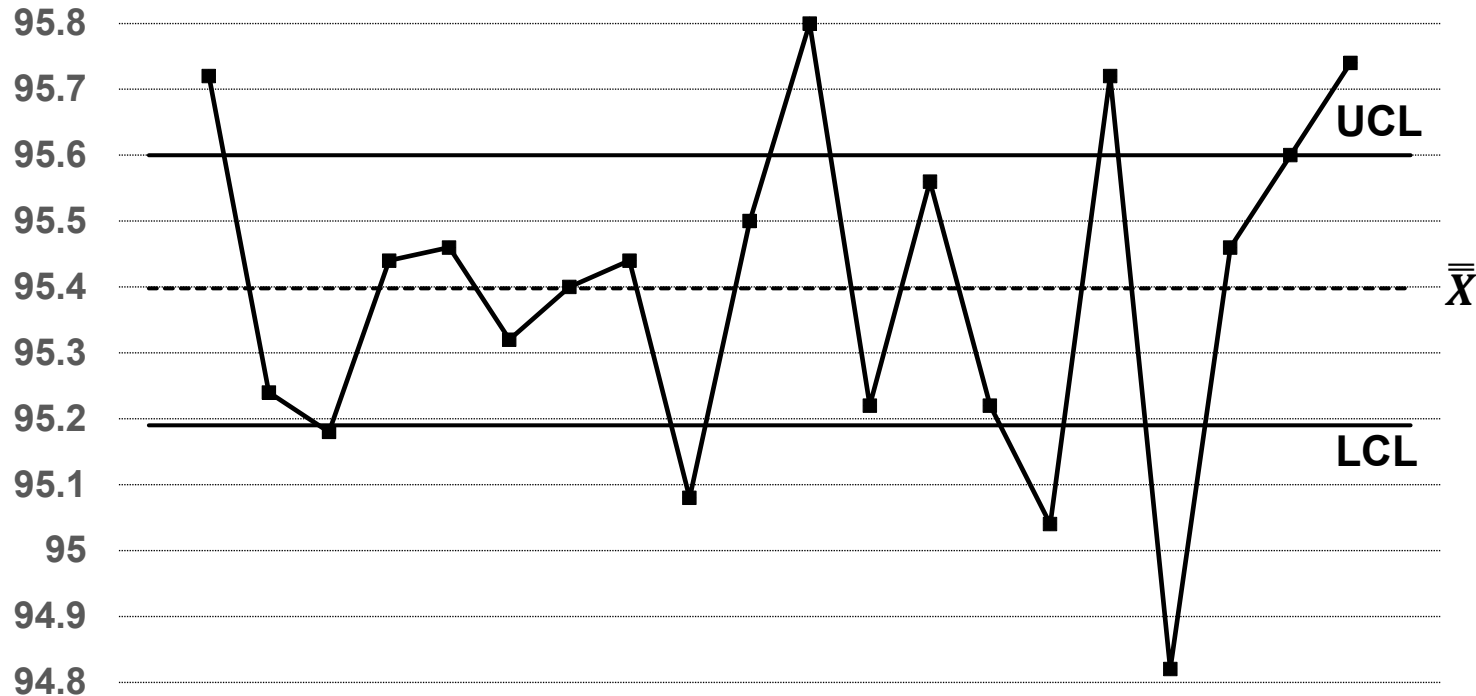
Example 2: Sample Mean

Sample	Average	Range	Sample	Average	Range
1	95.72	1.0	11	95.80	0.6
2	95.24	0.9	12	95.22	0.2
3	95.18	0.8	13	95.56	1.3
4	95.44	0.4	14	95.22	0.5
5	95.46	0.5	15	95.04	0.8
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8	95.44	0.3	18	95.46	0.5
9	95.08	0.2	19	95.60	0.4
10	95.50	0.6	20	95.74	0.6

→ Process is not under control with respect to its mean



Example 2: \bar{X} -Chart



→ Process is not under control with respect to its mean



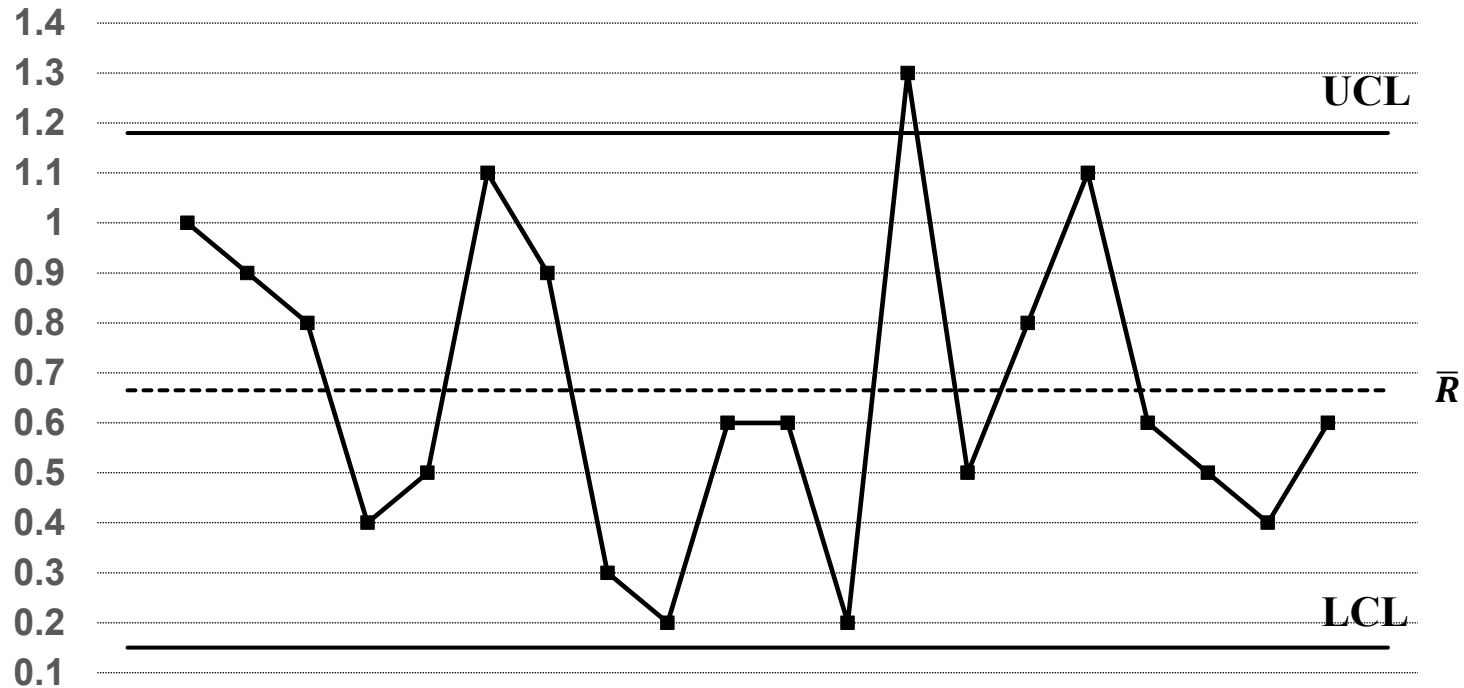
Example 2: Sample Range

Sample	Average	Range	Sample	Average	Range
1	95.72	1.0	11	95.80	0.6
2	95.24	0.9	12	95.22	0.2
3	95.18	0.8	13	95.56	1.3
4	95.44	0.4	14	95.22	0.5
5	95.46	0.5	15	95.04	0.8
6	95.32	1.1	16	95.72	1.1
7	95.40	0.9	17	94.82	0.6
8	95.44	0.3	18	95.46	0.5
9	95.08	0.2	19	95.60	0.4
10	95.50	0.6	20	95.74	0.6

→ Process is not under control with respect to its range



Example 2: R-Chart



→ Process is not under control with respect to its range



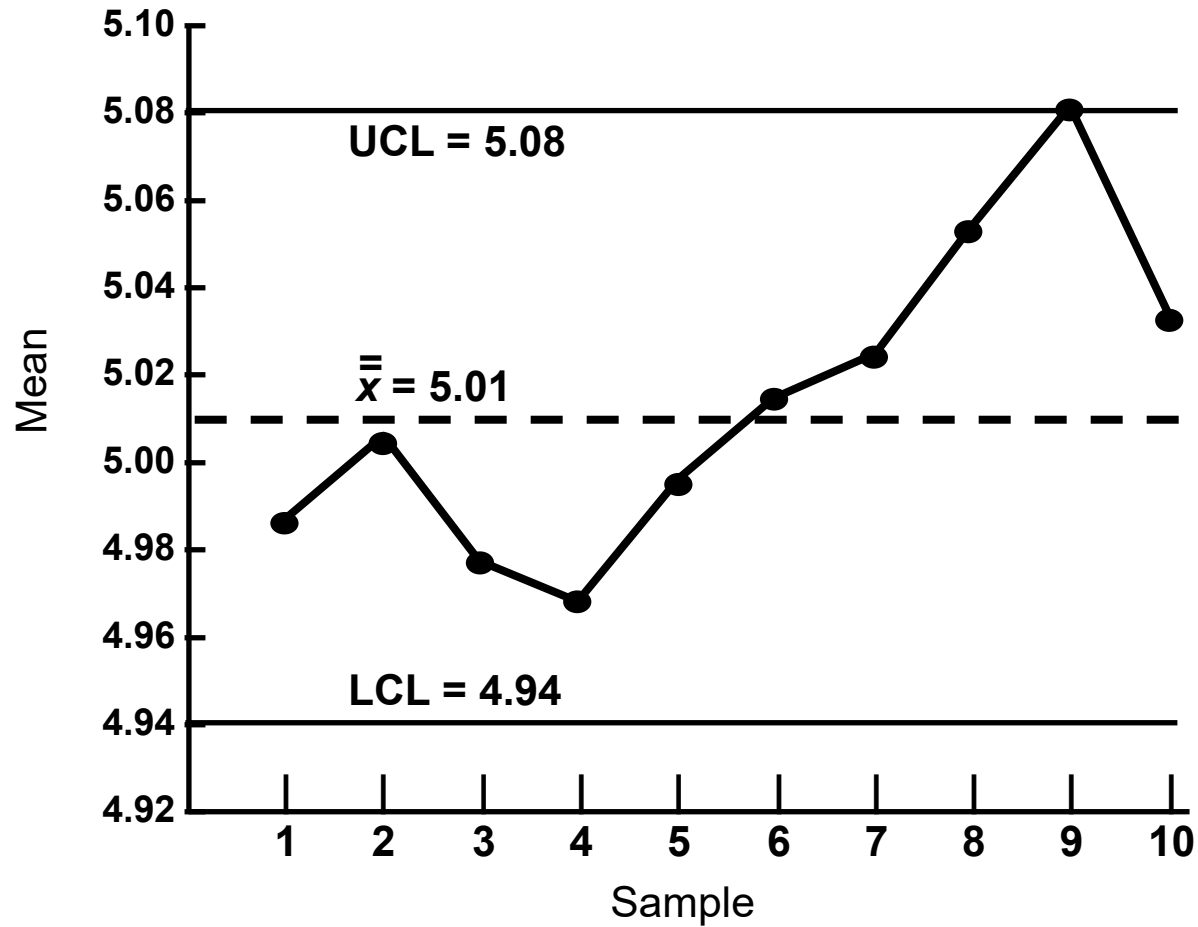
Example 3: Pod Weight Samples

Observations (weight in ng)

	1	2	3	4	5	\bar{X}	R
1	5.02	5.01	4.94	4.99	4.96	4.98	0.08
2	5.01	5.03	5.07	4.95	4.96	5.00	0.12
3	4.99	5.00	4.93	4.92	4.99	4.97	0.08
4	5.03	4.91	5.01	4.98	4.89	4.96	0.14
5	4.95	4.92	5.03	5.05	5.01	4.99	0.13
6	4.97	5.06	5.06	4.96	5.03	5.01	0.10
7	5.05	5.01	5.10	4.96	4.99	5.02	0.14
8	5.09	5.10	5.00	4.99	5.08	5.05	0.11
9	5.14	5.10	4.99	5.08	5.09	5.08	0.15
10	5.01	4.98	5.08	5.07	4.99	5.03	0.10
						<u>50.09</u>	<u>1.15</u>

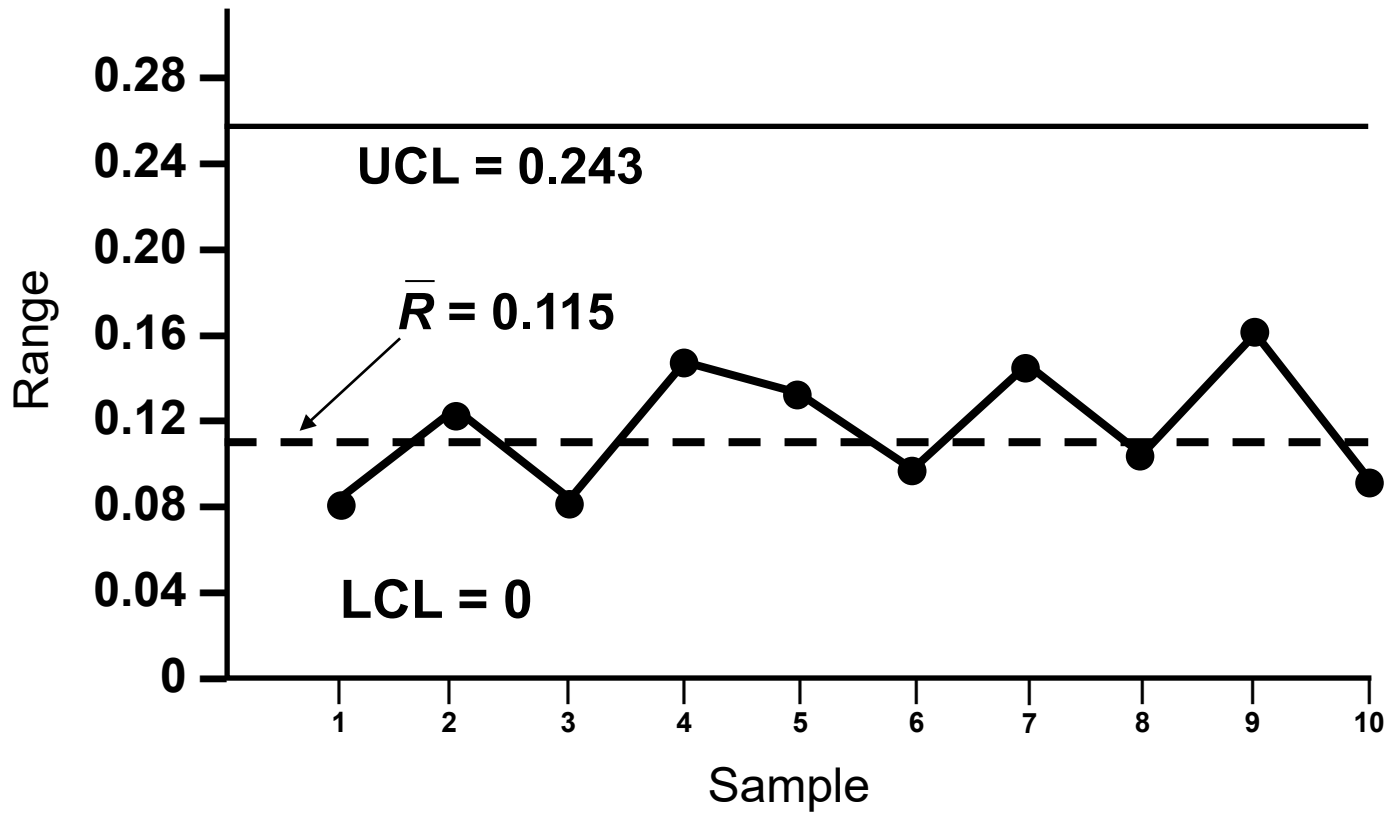


Example 3: \bar{X} -Chart



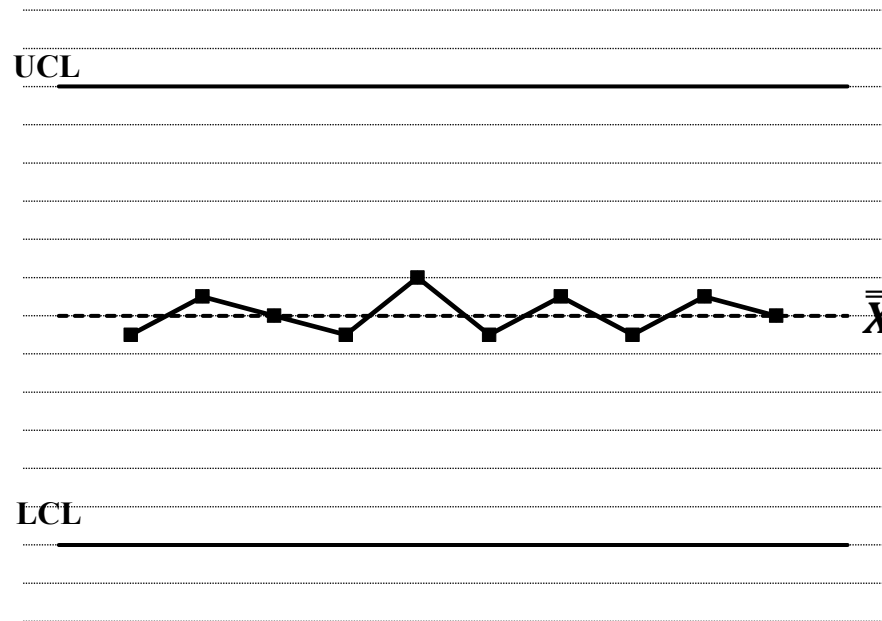


Example: R-Chart





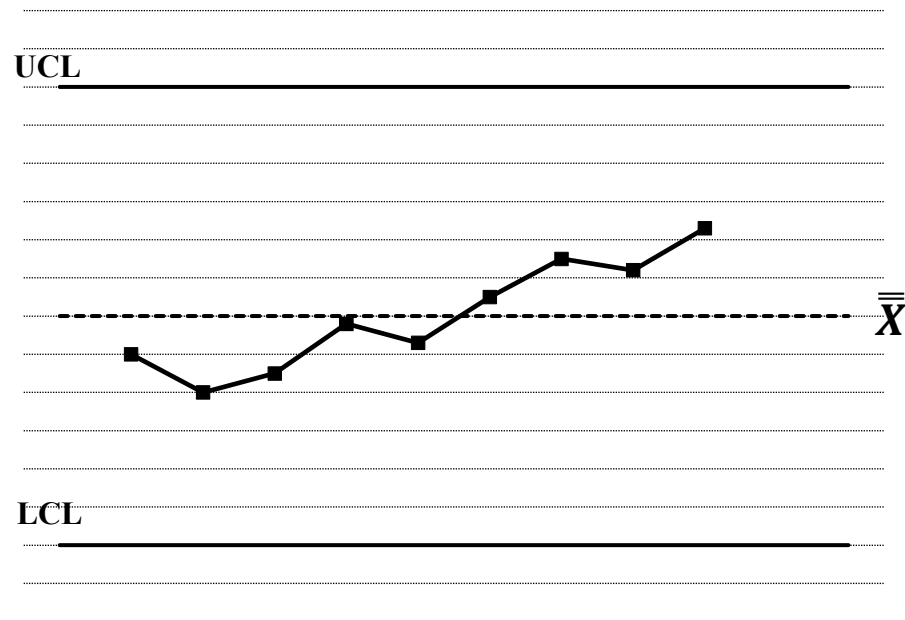
Analyzing Control Charts: Process 1



Process 1 is an ideal process



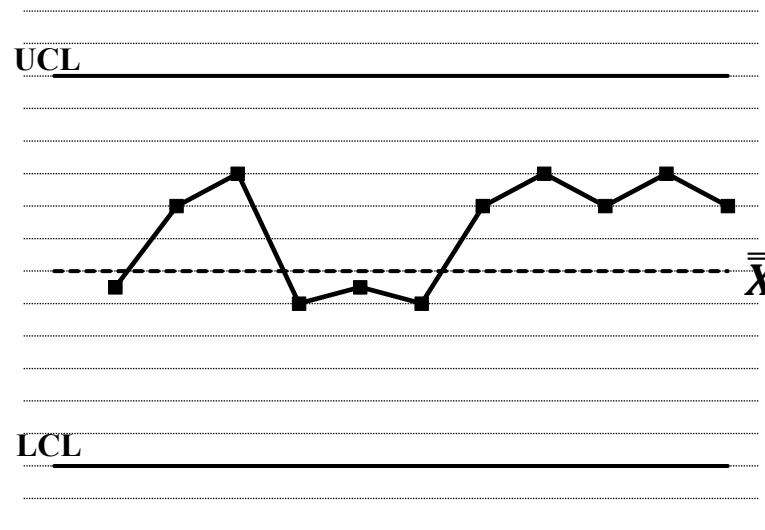
Analyzing Control Charts: Process 2



Process 2 is under control, but problematic because of an (upward) trend



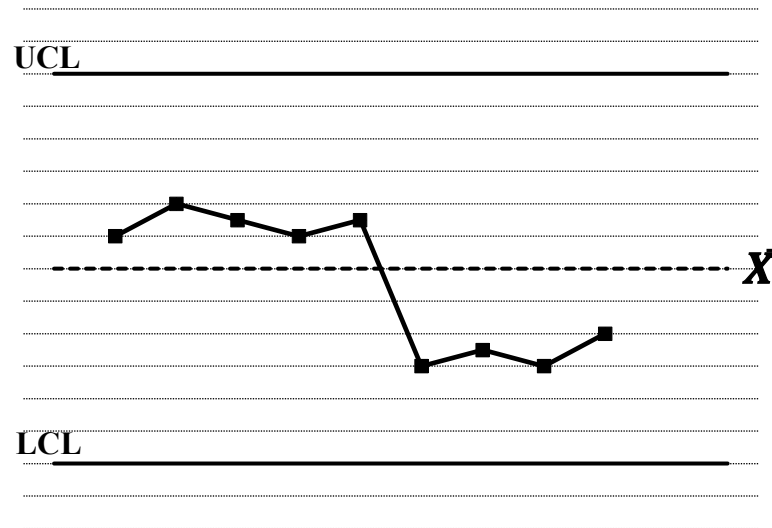
Analyzing Control Charts: Process 3



Process 3 is under control, but problematic because of 5 subsequent observations above \bar{X}



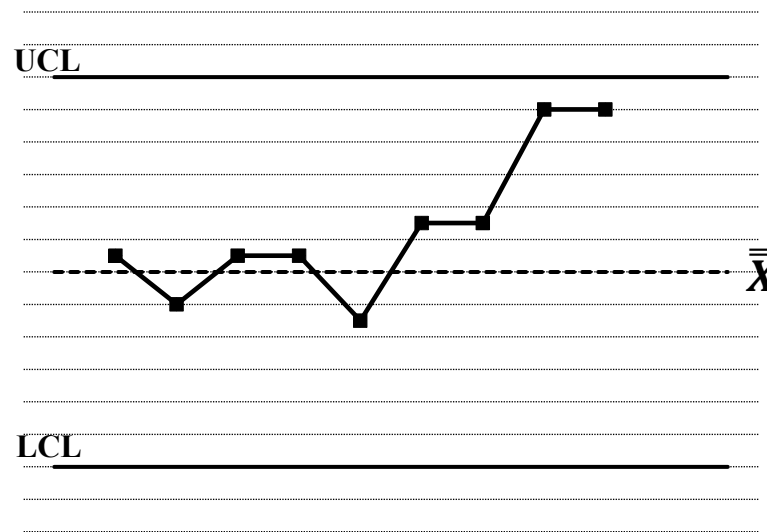
Analyzing Control Charts: Process 4



Process 4 is under control, but problematic because of a sudden shift



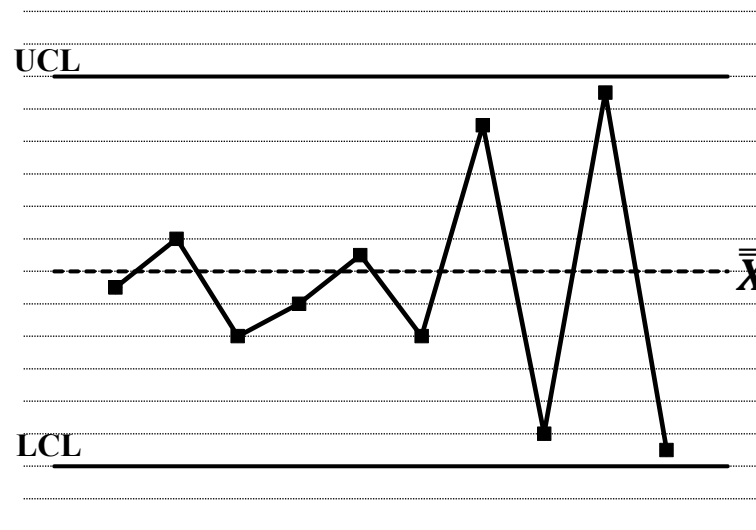
Analyzing Control Charts: Process 5



Process 5 is under control, but problematic because the values are approaching the UCL



Analyzing Control Charts: Process 6



Process 6 is under control, but problematic because of increasing process variance



Control Charts for Attributes

- Used to monitor the proportion of nonconforming units (defects, faults) in a sample
- Example: number of guests with complaints in a five-star hotel
- p-Chart:

$$\bar{p} = \frac{\text{Number of Guests with Complaints}}{\text{Total Number of Guests}}$$

n_j = size of sample j

$$UCL_j = \bar{p} + 3 \sqrt{\frac{\bar{p}(1 - \bar{p})}{n_j}}$$

$$LCL_j = \bar{p} - 3 \sqrt{\frac{\bar{p}(1 - \bar{p})}{n_j}}$$



Example 4: Five-Star Hotel

j	Sample	n_j	# Guests with Complaints	Ratio
1	January	100	8	0.08
2	February	50	4	0.08
3	March	100	10	0.1
4	April	100	8	0.08
5	May	75	6	0.08
6	June	100	10	0.1
7	July	150	15	0.1
8	August	100	12	0.12
9	September	50	8	0.16
10	October	100	10	0.1



Example 4: Calculations

Calculation of \bar{p} :
$$\frac{\sum_j \text{Number of Guests with Complaints}}{\sum_j n_j} = \frac{91}{925} = 0.1$$

Calculation of upper and lower control limits:

$$n_j=150: \text{UCL} = 0.1 + 3\sqrt{\frac{(0.1)(1-0.1)}{150}} = 0.17$$

$$\text{LCL} = 0.1 - 3\sqrt{\frac{(0.1)(1-0.1)}{150}} = 0.03$$

$$n_j=100: \text{UCL} = 0.1 + 3\sqrt{\frac{(0.1)(1-0.1)}{100}} = 0.19$$

$$\text{LCL} = 0.1 - 3\sqrt{\frac{(0.1)(1-0.1)}{100}} = 0.01$$

$$n_j=75: \text{UCL} = 0.1 + 3\sqrt{\frac{(0.1)(1-0.1)}{75}} = 0.20$$

$$\text{LCL} = 0.1 - 3\sqrt{\frac{(0.1)(1-0.1)}{75}} = 0.00$$

$$n_j=50: \text{UCL} = 0.1 + 3\sqrt{\frac{(0.1)(1-0.1)}{50}} = 0.23$$

$$\text{LCL} = 0.1 - 3\sqrt{\frac{(0.1)(1-0.1)}{50}} = -0.03$$

=> LCL = 0!



Example 4: Control Limits

Sample	n_j	# Guests with Complaints	Ratio	LCL	UCL
1	100	8	0.08	0.01	0.19
2	50	4	0.08	0.00	0.23
3	100	10	0.10	0.01	0.19
4	100	8	0.08	0.01	0.19
5	75	6	0.08	0.00	0.20
6	100	10	0.10	0.01	0.19
7	150	15	0.10	0.03	0.17
8	100	12	0.12	0.01	0.19
9	50	8	0.16	0.00	0.23
10	100	10	0.10	0.01	0.19
Sum	925	91	$\bar{p} = 91/925 = 0.1$		



Example 4: Improvement?

- In October, the hotel personnel received special training
- In November and December samples of 150 observations each were analyzed
- The ratio of guests with complaints was 0.02 in November and 0.01 in December
- Were the measures successful?



Performance Limits

Control limits

- are based on actual output data
- help to distinguish special from general (process immanent) causes of quality variations

Performance limits

- predict future process performance
- are calculated for processes which are under control
- make no sense for processes which are not under control



Process Capability

Specification limits

- describe desirable tolerance limits
- represent customer expectations

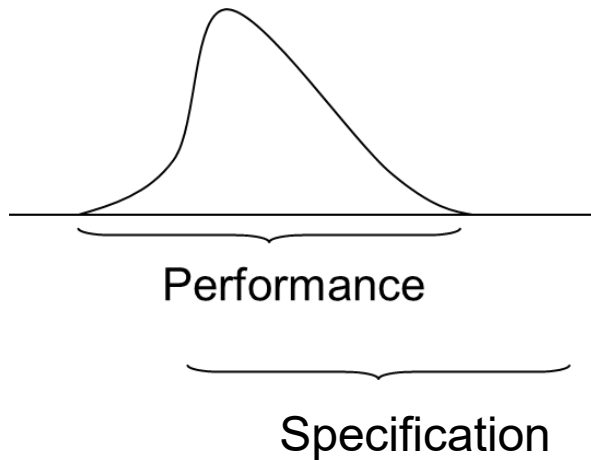
Process capability

- can only be determined for processes which are under control
- If a process is under control, it is capable of producing within its performance limits
- **But:** even processes which are under control may produce undesirable products/services (i.e., quality which is outside the specification limits)

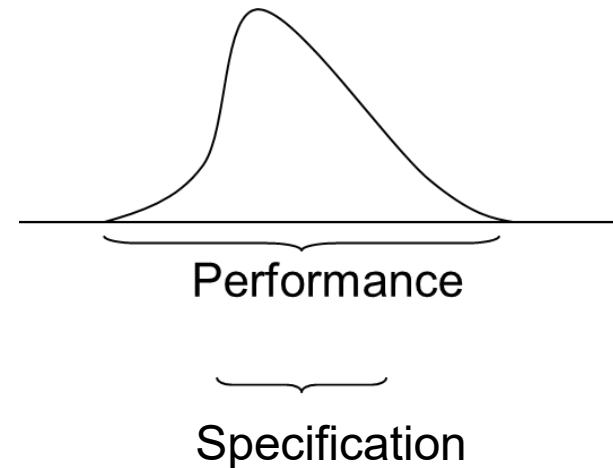


Performance Limits versus Specification Limits (1/2)

Undesirable situation:



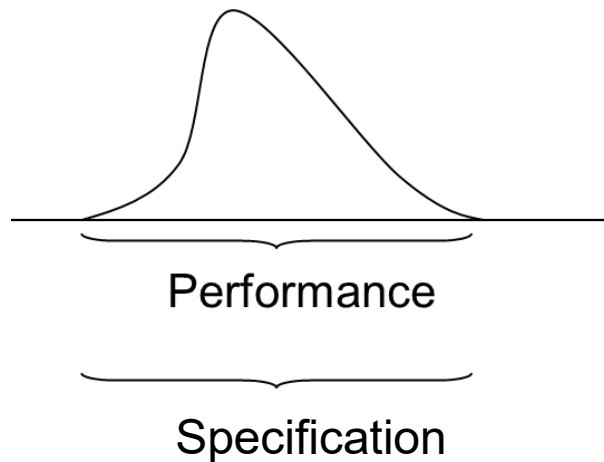
Extremely undesirable situation:



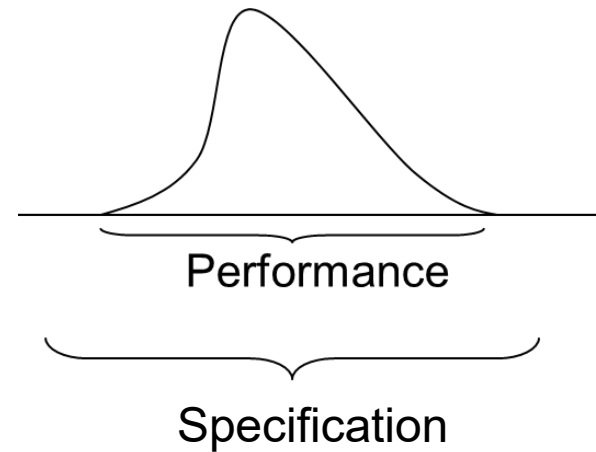


Performance Limits versus Specification Limits (2/2)

Vulnerable situation:



Highly desirable situation:





Capability Index (for symmetric processes)

Assumption: Process mean is centered between specification limits

$$C_p = \frac{\text{Upper Specification Limit (USL)} - \text{Lower Specification Limit (LSL)}}{6 * \sigma}$$

Process is capable if $C_p \geq 1$

Recommended minimum $C_p = 1.33$

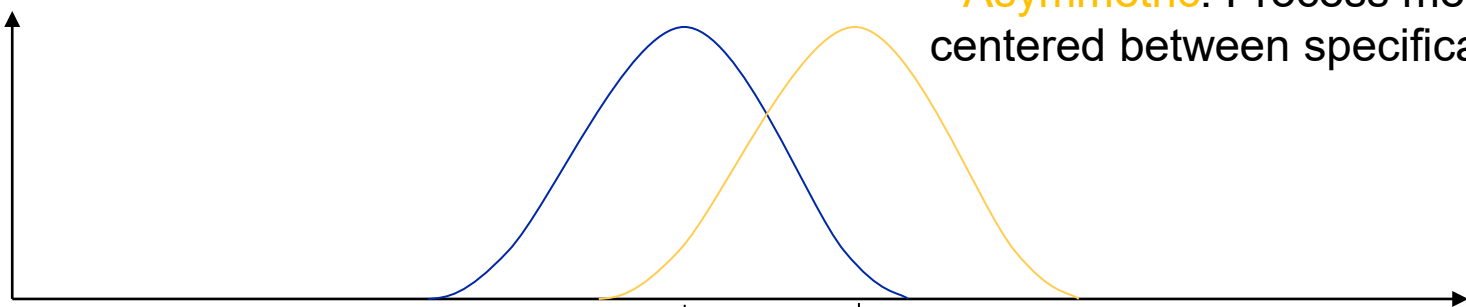
Six Sigma Quality process: $C_p = 2$



Symmetric vs. Asymmetric Processes

Symmetric: Process mean is centered between specification limits

Frequency



Asymmetric: Process mean is not centered between specification limits

Process
Mean

Process
Mean

Per-
formance

Per-
formance

Specification



Special Case: Capability Index for Asymmetric Processes

If process mean is not centered between specification limits:

$$C_{pk} = \min \left[\frac{USL - \mu}{3\sigma}; \frac{\mu - LSL}{3\sigma} \right]$$

C_{pk} Capability Index for asymmetric Processes

USL upper specification limit

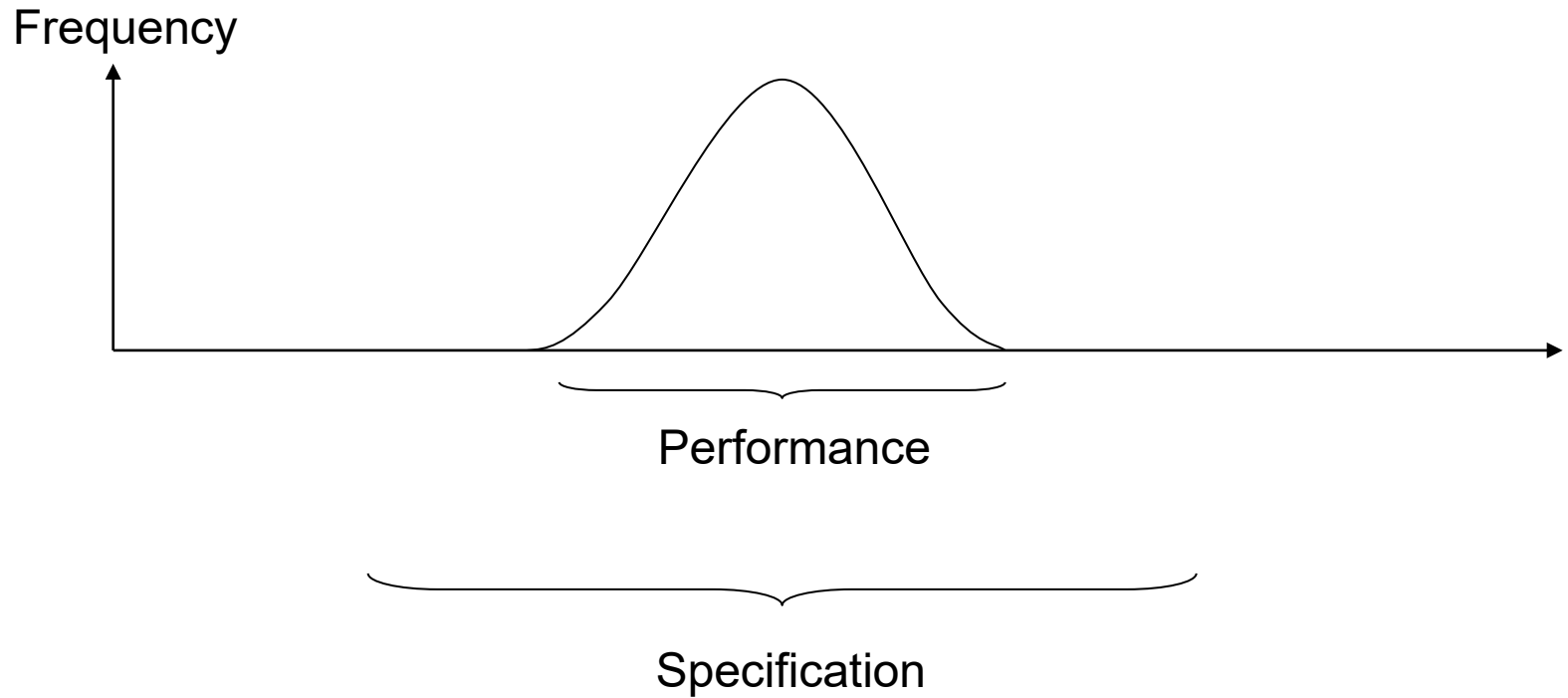
LSL lower specification limit

μ mean of the process (center between UCL and LCL)

σ standard deviation of the process

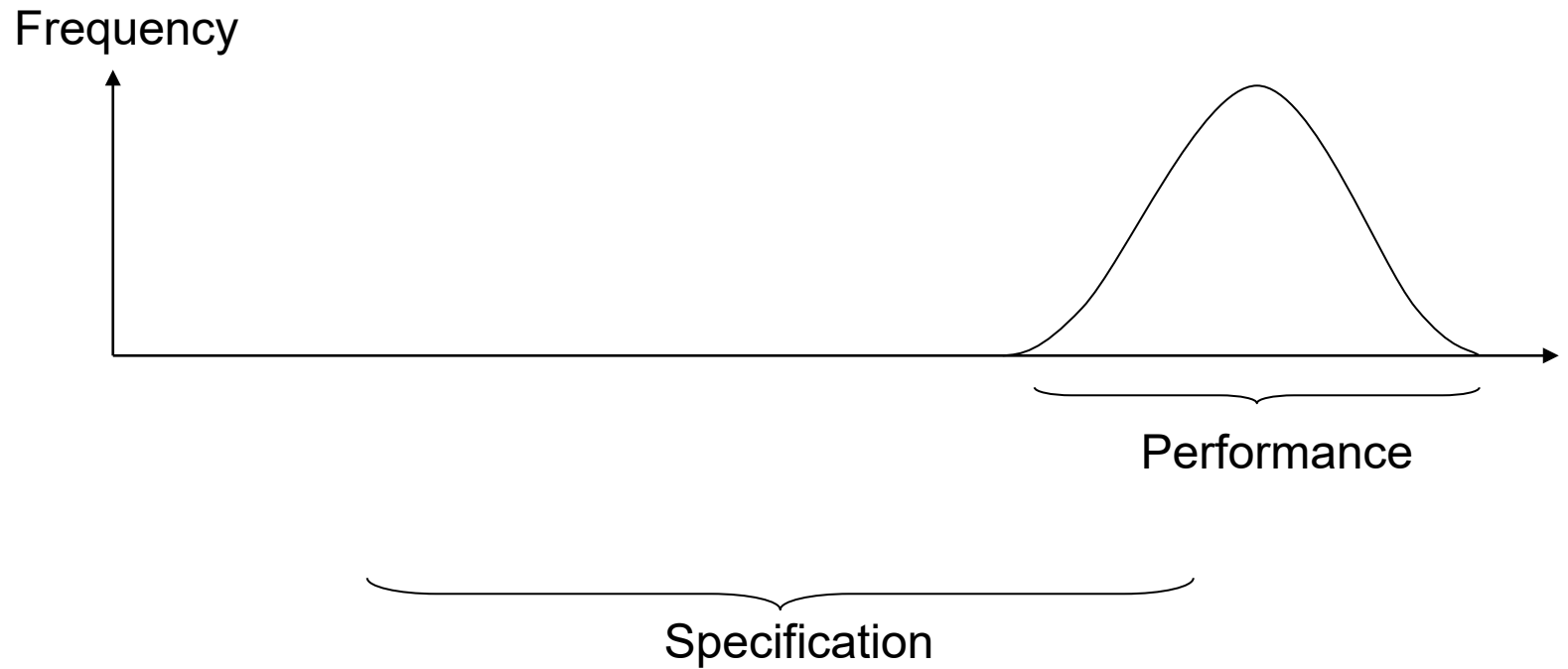


Ideal Situation: $CP > 1$





Bad, But Solvable: $CP > 1$





Not Solvable: $CP \ll 1$

