

w1.1- Understanding Statistical Process Control (5:29)

Foundations

- Forms data into patterns
- Promotes understanding
- Grounded in statistical principles
- Uncovers special causes of variation
- Inferences based on samples
- Detects product or process performance shifts

SPC helps us to frame data
into understandable patterns.



Subtitle scale: 0.5

Control Chart Objectives

- Monitor stable processes
- Measure process capability
- Assess expected outcomes
 - Average level
 - Variability
 - Consistency
- Predictive
- Limit exposure
- Continuous improvement



monitor, and control processes.

Benefits

- Lower manufacturing costs
- Stability
- Realistic specifications
- Less inspection
- Better customer perception
- Faster resolution of problems
- Decreased cycle times
- Improved quality

we can lower our
manufacturing costs.

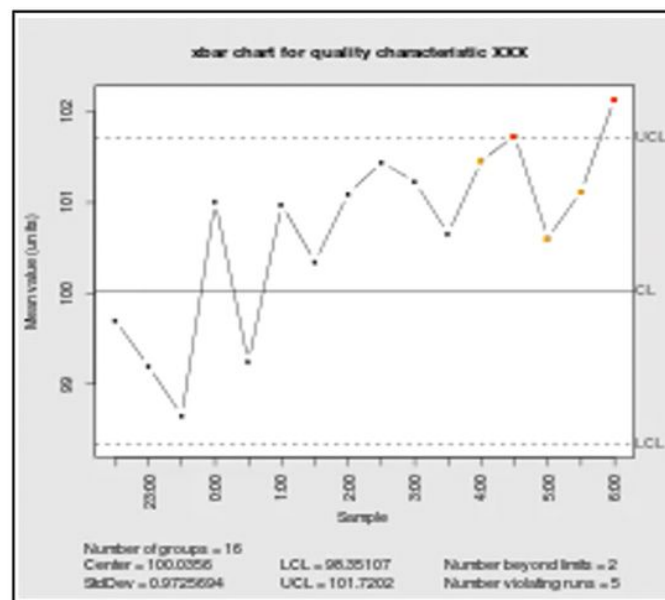


SPC Costs

- Determining what will monitored
- Data collection
- Control charts
- Training
- Oversight
- Corrective action



Structure of a Control Chart










https://en.wikipedia.org/wiki/Control_chart















Statistical Process Control (SPC) uses statistical tools to visualize, analyze, and control processes consistently, favoring a proactive approach to maintain output quality.

Core Objectives and Implementation







- **Upstream Control:** In Six Sigma, the focus shifts from monitoring final outputs to controlling "vital Xs" (input variables) to maintain consistent results.   
- **Targeted Deployment:** SPC is an involved process that should only be deployed where it is critical to business aims.  
- **Implementation Needs** Successful execution requires selecting appropriate variables, setting up data collection methods, providing training, and dedicating resources for investigation and corrective actions.  

The Role of Control Charts


Control charts, also known as Shewhart charts, are the primary feature of SPC.  

- **Components:** These charts display measured statistics (mean, range, or proportion) against a center line (CL) and calculated upper and lower control limits (UCL and LCL).
 
- **Stability and Variation:** They help distinguish between random variation and "special cause" variation, allowing operators to detect performance shifts.     
- **Capability Assessment:** Black Belts use these charts to determine if a process is operating as expected by comparing its functional range to specified tolerances.  
- **Problem Localization:** Control charts signal issues instantly, allowing examinations to be localized to only the material produced since the last check, thereby reducing scrap. 

Key Benefits

- **Cost and Quality:** Proactive SPC lowers manufacturing and inspection costs while improving stability, quality, and customer relationships.   
- **Operational Efficiency:** Implementation often leads to decreased cycle times and faster problem resolution. 
- **Continuous Improvement:** These tools provide a reliable method to measure parameters both before and after process changes to monitor improvement efforts.  

w1.2- Rational Subgrouping (5:46)

In Six Sigma, **rational subgrouping** is the practice of organizing data into groups that allow you to distinguish between two types of variation: variation occurring within a process at a specific point in time, and variation occurring over time. 

The goal is to collect data in a way that makes it easy to see if your process is "in control" or if there are outside factors causing "special cause" shifts.

The Two Dimensions of Variation

Rational subgrouping works by isolating variation into two distinct buckets:

Type of Variation	Definition	Goal in Subgrouping
Within-Subgroup	Variation between items produced under the same conditions (same operator, same batch, same machine).	This should represent the "natural" or common cause noise of the process.
Between-Subgroup	Variation that occurs over time or across different shifts, batches, or locations.	This should reveal special causes or trends that need fixing.

Selection of Variables

- Safety
- Environmental
- High defect rate
- Critical process variables that contribute cost, process or customer complaints
- Regulatory
- Requested items

the benefit is lost
as costs mount.



Rational Subgrouping

- Selection
 - Homogenous
 - Maximum variation
 - Process knowledge
- Types
 - Time based
 - Representative state



Principles for Subgrouping

- Standard sampling
- Never subgroup unlike things
- Minimize variation within subgroup
- Maximize variation between subgroups
- Average across noise



Control Limits and Subgroup Size



This transcript details the strategic implementation of Control Charts and the critical role of rational subgrouping. The core message is that monitoring must be purposeful; over-monitoring creates a burden that can make a process "clouded and ineffective."

Here is a summary of the key concepts discussed:

1. Selecting What to Monitor

You shouldn't monitor everything. Focus on the "vital few" variables that impact the entire process. Selection criteria include:

- **Safety and Regulatory requirements.**
- **High defect rates** or customer complaints.
- **Critical Process Variables:** Identifying the one variable that others depend upon.
- **Tools for selection:** Use Pareto charts, ANOVA, or Design of Experiments (DOE) to find these priorities.

2. The Mechanics of Rational Subgrouping

Subgrouping is how we organize data to see the difference between "noise" (common cause) and "signals" (assignable cause).

- **The Golden Rule:** Minimize variation *within* a subgroup (keep it homogeneous) and maximize the potential for variation *between* subgroups.
- **Time-Based Subgrouping:** Items are collected in the order of production. This is the best way to catch shifts in the process average.
- **Representative Subgrouping:** Used for product acceptance; a random sample is taken from all products made since the last check.
- **Size Matters:** Smaller, frequent subgroups (e.g., 5 groups of 5) are much more informative than one large group (e.g., 1 group of 25). Larger groups are more likely to hide a process shift within the average.

3. Error Types and Control Limits

The transcript explains the balancing act required when setting control limits, typically at $\pm 3\sigma$.

Error Type	Definition	Result of Wide Limits	Result of Narrow Limits
Type 1	Treating random noise as a problem (False Alarm).	Decreases (fewer false alarms).	Increases (too many false alarms).
Type 2	Missing a real process shift (Failed Detection).	Increases (limits are too "loose").	Decreases (more sensitive to change).

Key Recommendation

For most processes, aim for a **subgroup size of 2 to 6**. This range provides enough sensitivity to detect shifts (reducing Type 2 errors) without becoming overly sensitive to natural variation (Type 1 errors).

w1.3- Sources of Variation (3:37)

Special Cause Variation

- Unexpected
- Unpredictable
- Infrequent
- Examples
 - Machine malfunction
 - Operator issues
 - Faulty adjustment
 - Defective material
 - Training

eliminate special
cause variation.

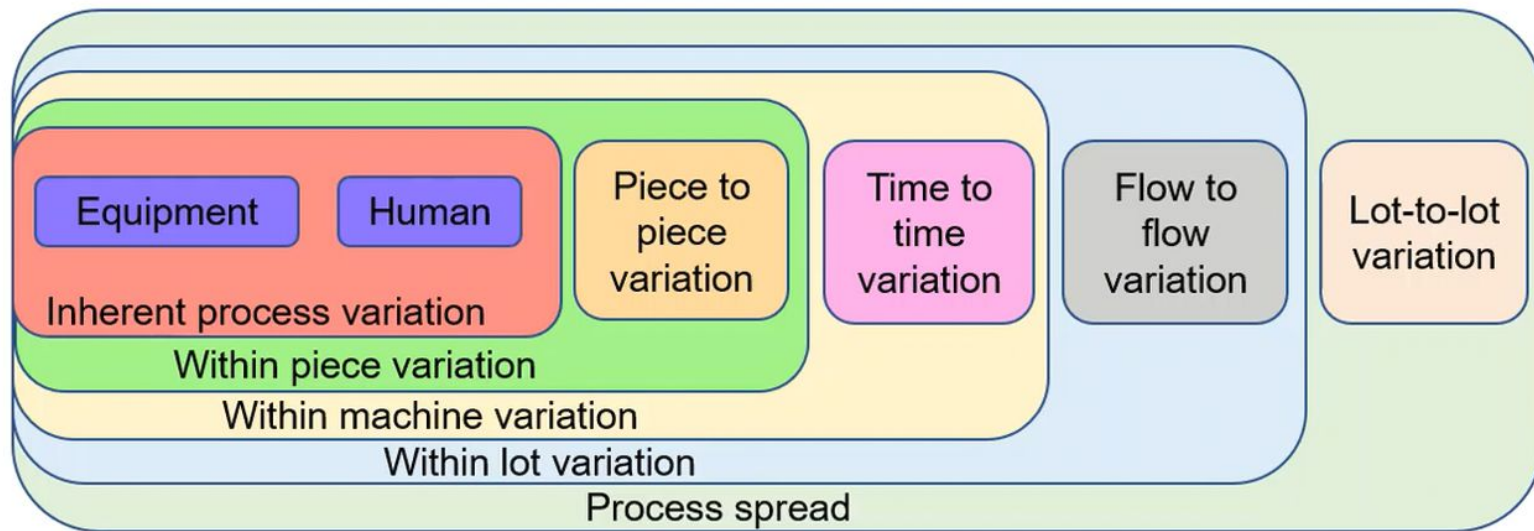


Common Cause Variation

- Due to random chance
- Noise
- Healthy
- Examples
 - Inadequate procedures
 - Poor design
 - Poor maintenance
 - Poor working conditions
 - Inadequate material
 - Measurement error
 - Normal Wear



Sources of Variability



Process capability hinges on

Understanding the different types and sources of variation is critical for effectively managing a manufacturing process. In SPC, the primary goal is to distinguish between variation that can be fixed and variation that is inherent to the system.

1. Special Cause vs. Common Cause Variation

The first step in analyzing a process is identifying the nature of the "noise" or changes observed.

- **Special Cause (Assignable Cause) Variation:** This is variation that can be traced to unexpected and unpredictable behaviors.
 - **Goal:** Find and eliminate these specific issues.
 - **Examples:** Machine malfunctions incorrect equipment settings, defective materials, or a lack of operator training.

- **Common Cause (Chance Cause) Variation:** This is considered "healthy" variation or inherent noise within the system.
 - **Warning:** Mistaking common cause for special cause can lead to increased variation due to unnecessary adjustments.
 - **Examples:** Inadequate procedures, normal equipment wear, and standard measurement error.

2. Peeling Back the Sources of Variability

You can think of long-term process variation like an onion; you must "peel away" specific layers to reveal the inherent process capability.

Variation Layer	Description
Lot-to-Lot	Changes in the process average as you move between different batches.
Within-Lot	The dispersion of measurements inside a single batch or lot.
Machine-to-Machine	Variation based on the flow of products across different pieces of equipment.
Time-to-Time	Isolating changes that occur purely due to the passage of time.
Piece-to-Piece	Differences found when inspecting individual products sequentially.

3. Measurement Error and Inherent Capability

Even after accounting for the layers above, **human and mechanical factors** still contribute to the total variation.

- **Measurement Error:** This results from how a part is handled, the choice of measurement location, and the instrument used. It is often a mix of human factors and equipment condition.
- **Interactions:** Significant variation can come from the interaction between the human operator and the instrument or equipment.
- **Inherent Process Capability:** This is what remains **after all other sources are isolated**. It represents the purest state of reproducibility under ideal operating conditions.