

Part I

Statistical Process and Quality Control

Introduction

Quality and Process Control

Definition of Quality

- Quality means fitness for use. and
- Quality is inversely proportional to variability.
- My definition: *Is* and *should* are the same.

Statistical Process Control (SPC)

- Statistical process control is, first and foremost, a way of thinking which happens to have some tools attached.

The Magnificent Seven

1. histogram
2. check sheet
3. Pareto chart
4. defect concentration diagram
5. cause-and-effect diagram
6. control chart
7. scatter diagram

Control Charts

Control Charts versus Hypothesis Testing

$$\bar{X} \approx \mathcal{N}\left(\mu, \frac{\sigma^2}{n}\right) \quad (1)$$

Arithmetic mean

$$\bar{x} = \frac{1}{n} \sum_{i=1}^n x_i \quad (2)$$

Question: is $|\bar{x} - \mu_0|$ significant - μ_0 target value

- \bar{x} Arithmetic mean of the measurements

Given: Level of significance $\alpha = 0.0027$

The Control Chart

No.	sample values						mean	sd	range
1	x ₁₁	x ₁₂	...	x _{1j}	...	x _{1n}	\bar{x}_1	s ₁	R ₁
2	x ₂₁	x ₂₂	...	x _{2j}	...	x _{2n}	\bar{x}_2	s ₂	R ₂
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
i	x _{i1}	x _{i2}	...	x _{ij}	...	x _{in}	\bar{x}_i	s _i	R _i
⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮	⋮
k	x _{k1}	x _{k2}	...	x _{kj}	...	x _{kn}	\bar{x}_k	s _k	R _k

Figure 1: Data Set with Mean, Standard Deviation and Range

Mean values

$$\bar{x}_i = \frac{1}{n} \sum_{j=1}^n x_{ij} \quad (3)$$

Standard deviations

$$s_i = \sqrt{\frac{1}{n-1} \sum_{j=1}^n (x_{ij} - \bar{x}_i)^2} \quad (4)$$

Ranges

$$R_i = \max\{x_{ij} | j \in \{1, \dots, n\}\} - \min\{x_{ij} | j \in \{1, \dots, n\}\} \quad (5)$$

for all $i \in \{1, \dots, k\}$

Control Chart with R and \bar{x}

R Chart

Mean Range

$$\bar{R} = \frac{1}{k} \sum_{i=1}^k R_i \quad (6)$$

Control limits

$$UCL = D_4 \bar{R}; \quad LCL = D_3 \bar{R} \quad (7)$$

\bar{x} Chart

Mean Range

$$\bar{R} = \frac{1}{k} \sum_{i=1}^k R_i \quad (8)$$

Control limits

$$UCL = \mu + 3 \frac{\sigma}{\sqrt{n}}; \quad LCL = \mu - 3 \frac{\sigma}{\sqrt{n}} \quad (9)$$

μ : Process mean

σ : Process standard deviation

Problem: We don't know μ and σ so we have to estimate them:

We know that for an independent sample x_1, \dots, x_n from a normal distribution with parameters μ and σ the mean

$$\bar{x} = \frac{1}{n} \sum_{j=1}^n x_j \quad (10)$$

satisfies

$$E(\bar{x}) = \mu \quad \text{and} \quad Var(\bar{x}) = \frac{\sigma^2}{n} \quad (11)$$

The mean is an unbiased estimator with the standard error

$$SE(\bar{x}) = \frac{\sigma}{\sqrt{n}} \quad (12)$$

Assumption: R chart is under statistical control.

- The value \bar{R} is a reliable estimate for the mean range.

- The value \bar{R} is a reliable estimate for the process standard deviation

$$\hat{\sigma} = \frac{\bar{R}}{d_2} \quad (13)$$

Any samples excluded for construction of the R chart should also be disregarded for construction of the \bar{x} chart. This results in a sample of k^* valid samples, (where k^* denotes the reduced number of samples).

Mean values of $\bar{x}_1, \dots, \bar{x}_{k^*}$ provide an estimate of μ , i.e

$$\hat{\mu} = \bar{\bar{x}} = \frac{1}{k^*} \sum_{i=1}^{k^*} \bar{x}_i \quad (14)$$

Control limits

$$UCL = \bar{\bar{x}} + 3 \frac{\bar{R}}{d_2} \frac{1}{\sqrt{n}} \approx \bar{\bar{x}} + A_2 \bar{R} \quad (15)$$

$$LCL = \bar{\bar{x}} - 3 \frac{\bar{R}}{d_2} \frac{1}{\sqrt{n}} \approx \bar{\bar{x}} - A_2 \bar{R}$$

Control Chart with \bar{x} and s

s Chart

The centreline of the s chart is denoted by \bar{s} and is calculated from the arithmetic mean of the standard deviations

$$\bar{s} = \frac{1}{k} \sum_{i=1}^k s_i \quad (16)$$

Control limits

$$UCL = B_4 \bar{s}; \quad LCL = B_3 \bar{s} \quad (17)$$

\bar{x} Chart

Using an s chart of a process that is under control, the process standard deviation can be estimated by

$$\hat{\sigma} = \frac{\bar{s}}{c_4} \quad (18)$$

Any samples excluded for construction of the s chart should also be disregarded for construction of the \bar{x} chart. This results in a sample of k^* valid samples, (where k^* denotes the reduced number of samples). Mean values of $\bar{x}_1, \dots, \bar{x}_{k^*}$ provide an estimate of μ , i.e

$$\hat{\mu} = \bar{\bar{x}} = \frac{1}{k^*} \sum_{i=1}^{k^*} \bar{x}_i \quad (19)$$

Control limits

$$UCL = \bar{\bar{x}} + 3 \frac{\bar{s}}{c_4} \frac{1}{\sqrt{n}} \approx \bar{\bar{x}} + A_3 \bar{s} \quad (20)$$

$$LCL = \bar{\bar{x}} - 3 \frac{\bar{s}}{c_4} \frac{1}{\sqrt{n}} \approx \bar{\bar{x}} - A_3 \bar{s}$$

Individual Control Charts

Individual control charts have exactly one measurement per sample.
Problem: You cannot estimate variability from a single measurement.
Idea: Use variation of two adjacent measurements. **Moving ranges**

$$MR_i = |x_{i+1} - x_i| \quad (21)$$

for all $i \in \{1, \dots, n-1\}$.

Arithmetic mean of the moving ranges

$$\bar{MR} = \frac{1}{n-1} \sum_{i=1}^{n-1} MR_i \quad (22)$$

Estimated process standard deviation

$$\hat{\sigma} = \frac{\bar{MR}}{d_2} = \frac{\bar{MR}}{1.128} \quad (23)$$

Since two neighboring measurements were used to calculate the moving ranges we have $d_2 = 1.128$.

Centerline The centerline for the individuals control chart is the arithmetic mean of the measured values.

$$\bar{x} = \frac{1}{k} \sum_{i=1}^k x_i \quad (24)$$

Control limits

$$UCL = \bar{x} + 3 \frac{\bar{MR}}{1.128}; \quad LCL = \bar{x} - 3 \frac{\bar{MR}}{1.128} \quad (25)$$

Control Charts for Attributes Data – p Chart

Number of defectives under number tested is a discrete random variable.
Given: Random sample of size n , of which D parts are defective We know: The number of defective D under n examined parts follows a binomial distribution with the unknown probability p of success. **Estimated probability**

$$\hat{p} = \frac{D}{n} \quad (26)$$

Variance

$$Var(\hat{p}) = \frac{p(1-p)}{n} \quad (27)$$

Given: - k random samples with n_1, \dots, n_k values. - Each of these samples contains d_1, \dots, d_k defective products. **k relative frequencies**

$$p_1 = \frac{d_1}{n_1}, \dots, p_k = \frac{d_k}{n_k} \quad (28)$$

Centerline The centreline and the control limits of a p chart are again determined from a stable trial run with k^* valid samples.

Again $k^* \leq k$ is the reduced number of samples.

Distinguish 2 cases: 1. The sample sizes n_1, \dots, n_k are all equal to n . 2. The sample sizes are not all equal.

Case 1
Centerline

$$\bar{p} = \frac{1}{k^*} \sum_{i=1}^{k^*} p_i \quad (29)$$

Control limits

$$UCL = \bar{p} + 3\sqrt{\frac{\bar{p}(1-\bar{p})}{n}}; \quad LSL = \bar{p} - 3\sqrt{\frac{\bar{p}(1-\bar{p})}{n}} \quad (30)$$

Case 2
Centerline

$$\bar{p} = \frac{d_1 + \dots + d_{k^*}}{n_1 + \dots + n_{k^*}} \quad (31)$$

Control limits

$$UCL_i = \bar{p} + 3\sqrt{\frac{\bar{p}(1-\bar{p})}{n_i}}; \quad LSL_i = \bar{p} - 3\sqrt{\frac{\bar{p}(1-\bar{p})}{n_i}} \quad (32)$$

The control limits now depend on the index i .

Part II

Multiple Regression

Part III

Design of Experiment