

PRO 5971 - Statistical Process Monitoring

About the control limits

Linda Lee Ho

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Department of Production Engineering
University of São Paulo

- Let X , be a random variable with PDF $f(X)$ and CDF as $F(X)$ with $E(X) = \mu$ and $Var(X) = \sigma^2$, both $< \infty$
- Let us consider a sample random of size n **independent identical distributed** (iid) random variables: $X_1, X_2, \dots, X_n$.
- By Central Limit Theorem, the **asymptotic distribution** (that is, when $n \rightarrow \infty$) of $\bar{X} = \frac{\sum_{i=1}^n X_i}{n}$ follows a normal distribution $\sim N\left(\mu, \frac{\sigma^2}{n}\right)$
- For the $\bar{X}$ control chart, the control limits are of the form $\mu_0 \pm z_{1-\alpha/2} \frac{\sigma}{\sqrt{n}}$
- These control limits are known as **asymptotic** one
- $z_{1-\alpha/2} = 3$, the control limits are known as **3 sigma control limits**
- They are symmetrical and easier to deal with them

- If $X \sim N(\mu, \sigma^2)$ then the **exact distribution** of $\bar{X}$ is also normal $\sim N\left(\mu, \frac{\sigma^2}{n}\right)$
- So if the quality characteristic X follows a normal distribution, the asymptotic and the exact distribution of $\bar{X}$ are **identical**.
- However when the quality characteristic X does not follow a normal distribution, the **exact distribution of $\bar{X}$ may be far from the asymptotic distribution of $\bar{X}$**
- Control limits determined by using the exact distribution of $\bar{X}$ is known as **probability control limits**.
- If $X \sim N(\mu, \sigma^2)$ then the probability and asymptotic control limits of $\bar{X}$ are **identical**.

- Assumption: The quality characteristic X does not follow a normal distribution
- Let $F_{\bar{X}}(\cdot)$ be the CDF of $\bar{X}$; $q_{\alpha/2}$ and $q_{1-\alpha/2}$ are respectively the $\alpha/2$ and $1 - \alpha/2$ quantiles of $F_{\bar{X}}$ such that:

$$F_{\bar{X}}(q_{\alpha/2}) = P(\bar{X} < q_{\alpha/2}) = \alpha/2$$

and

$$F_{\bar{X}}(q_{1-\alpha/2}) = P(\bar{X} < q_{1-\alpha/2}) = 1 - \alpha/2$$

- Consequently $P(q_{\alpha/2} < \bar{X} < q_{1-\alpha/2}) = 1 - \alpha$
- The quantiles $q_{\alpha/2}$ and $q_{1-\alpha/2}$ are **the probability control limits** considering as error of type I equal to α
- However sometimes the exact distribution of $\bar{X}$ is unknown and the probability control limits are determined by simulation.

- If X is r.v. following a Bernoulli distribution with parameter p , then $\sum_{i=1}^n X_i \sim \text{Binomial}(n; p)$
- If X is r.v. following a Poisson distribution with parameter λ , then $\sum_{i=1}^n X_i \sim \text{Poisson}(n\lambda)$
- If X is r.v. following an exponential distribution with parameter λ , then $\sum_{i=1}^n X_i \sim \text{Gamma}(n; \lambda)$
- etc

- X , some quality characteristic with PDF $f_X(\cdot)$ and CDF $F_X(\cdot)$
- Let $X_1, X_2, \dots, X_n$ a random sample of X ; $\tilde{\Theta} = G(X_1, X_2, \dots, X_n)$ a statistic used to monitor some parameter of interest with $E(\tilde{\Theta})$ and $Var(\tilde{\Theta})$ both known/available.
- The **asymptotic control limits** are: $E(\tilde{\Theta}) \pm z_{1-\alpha/2} \sqrt{Var(\tilde{\Theta})}$
- The quantiles $q_{\alpha/2}$ and $q_{1-\alpha/2}$ are the **probability control limits** with as error of type I equal to α such that

$$F_{\tilde{\Theta}}(q_{\alpha/2}) = P(\tilde{\Theta} < q_{\alpha/2}) = \alpha/2$$

and

$$F_{\tilde{\Theta}}(q_{1-\alpha/2}) = P(\tilde{\Theta} < q_{1-\alpha/2}) = 1 - \alpha/2$$

$F_{\tilde{\Theta}}(\cdot)$ is the CDF of $\tilde{\Theta}$

- Consequently $P(q_{\alpha/2} < \tilde{\Theta} < q_{1-\alpha/2}) = 1 - \alpha$