

# TUTORIAL

## Cpk Plots – An Application of Individuals Control Charts

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In a previous issue of the Statistics Division Newsletter, I introduced readers to a variety of Process Capability indices used today in industry. In this article I offer a case study on how an Individuals control chart is used to track Cpk as a means of measuring ongoing process improvement activities. Recall that Cpk is defined as:

$$Cpk = \text{minimum of } \left\{ \frac{\bar{X} - LSL}{3\sigma}, \frac{USL - \bar{X}}{3\sigma} \right\} \quad \text{where } \sigma = \text{process standard deviation}$$

Practitioners often forget that process capability indices are merely point estimates obtained by taking samples from a population. If one must report Cpk (many customers now require Cpk for key quality characteristics on suppliers' Certificates of Analysis), then one must understand the associated errors of sample size and other related limitations and constraints in estimating Cpk. First, Cpk should only be estimated for a stable process. A process lacking statistical control is not predicable. In such cases Ppk is the more correct estimate of process performance.

$$Ppk = \text{minimum of } \left\{ \frac{\bar{X} - LSL}{3s}, \frac{USL - \bar{X}}{3s} \right\} \quad \text{where } s = \text{sample standard deviation}$$

Second, Cpk (and Ppk) estimates of population capability assume the data come from a normal (or bell-shaped) distribution, with symmetrical specification limits about the target. Because process capability indices are determined from estimates of standard deviation, they are affected by sample size (degrees of freedom). As expected, the standard error of the estimate of Cpk decreases as the sample size increases.

It can be shown by use of the Chi-square distribution and Monte Carlo simulation techniques that a sample size (n) as small as 10 cannot be relied upon to give results of much practical value. Even when n is as large as 40 there is still substantial uncertainty in the estimator of Cpk. Table 1 provides an estimate of the 95% lower bound on the sample Cpk, for various known population (or lot) capabilities (assuming the data follow a normal distribution). Table 2 shows confidence intervals for Ppk.

**Table 1.** Approximate 95% lower CI for Cpk. (1)

		Sample Size Used to Estimate Cpk		
		30	50	75
True Cpk	1.00	0.72	0.79	0.83
	1.40	0.80	0.87	0.91
	1.50	1.12	1.21	1.26
	1.667	1.25	1.35	1.40

**Table 2.** Approximate 95% Confidence Interval for Ppk.\*

		True Ppk		
		1.00	1.33	1.67
Sample Size	30	.76-1.31	1.02-1.76	1.29-2.19
	60	.83-1.21	1.11-1.61	1.49-2.01
	120	.88-1.14	1.17-1.52	1.47-1.90

As Table 1 shows, a stable process possessing a "true" process capability of 1.00 could exhibit a sample Cpk value as small as 0.72, even when the sample size is as large as 30. Although not usually of interest, a similar approach would illustrate that the sample Cpk could be much larger than the population Cpk. These points are important to note with regards to shipment Certificates of Analysis (CoA) required by your customer. Small sample sizes could result in a "capable" shipment appearing to be not capable of meeting customer specifications, but only because the sampling variability allows the sample estimate of Cpk to be much smaller than the population or lot characteristic (referred to as 'producer's risk'). But also, an incapable shipment could 'appear' capable because the sampling variability allows the sample Cpk to be larger than the population or lot characteristic (referred to as 'consumer's risk'). Because the sampling distribution of the Cpk statistic is so variable it should not be used unless relatively large sample sizes (100~200) are obtained. If Lot Cpk is requested on a shipment CoA the supplier should pull in historical data to get the sample size (for Cpk estimates only) greater than 100. Similar variability in Ppk is illustrated by Table 2.

As with any process data, Cpk estimates are better understood and interpreted when they are plotted over time. To avoid the pitfalls of making decisions based on 2-point comparisons and of single sample variation the concepts of Statistical Thinking should be employed and Cpk values plotted on control charts. A graphical plot of Cpk assists in the visual summarization of process improvement (or deterioration). The following example illustrates both common cause variation in Cpk and special cause variation in Cpk.

### Case Study: Slit Roll Tab Control

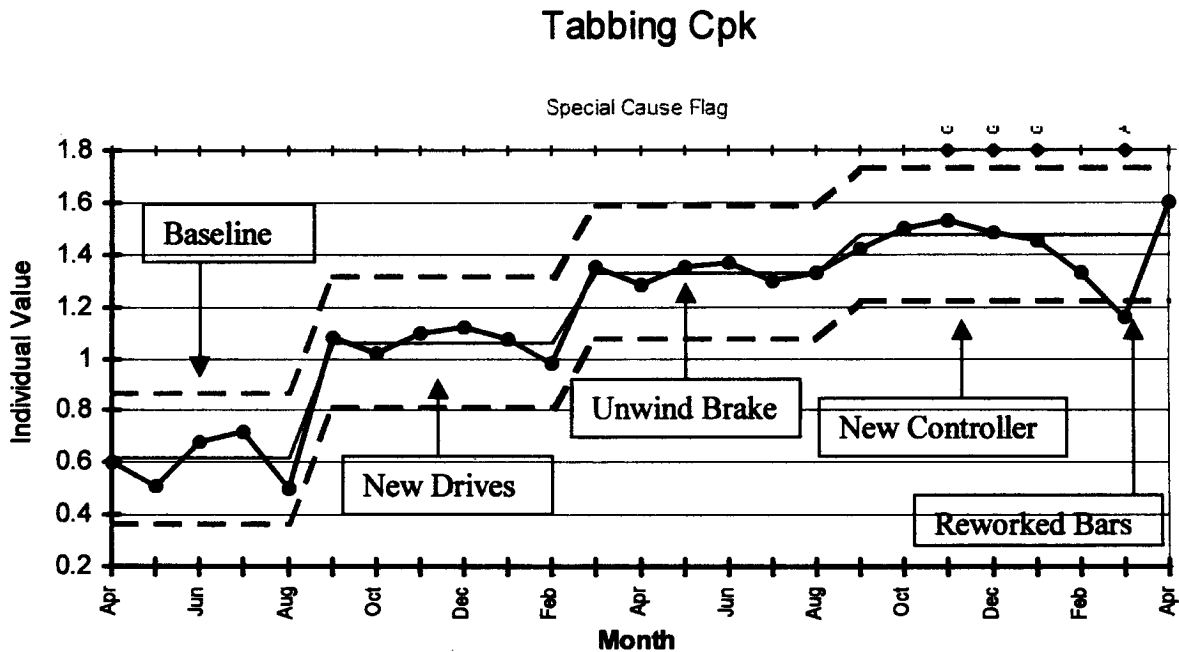
A key quality characteristic for the starter tab on a roll of office tape must be controlled to target, with minimal variation. Undertabbing results in a "flagged" tab which causes roll unraveling and strand entanglement in the automated conveying equip-

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ment; whereas, overtabbing causes end-user frustration due to difficulty starting the roll. In essence we have identified an Internal customer need for equipment efficiency, and an External customer need for ease of use. The parties agreed that this need might be 'summarized' by requiring a Cpk of 1.33 or greater. An Individuals control chart of Tabbng Cpk, (how well the starter tab is placed to the correct length on the roll of office tape), given below, shows the affect of slitting and winding process modifications over time. The "Individual Value" reported on the y-axis is the Cpk statistic calculated for the month. The monthly sample size was set at n=100 in order to minimize the effect of sampling variation in the estimate of sigma for the Cpk statistic.



The 3-sigma control limits on the Individuals chart are determined from the historical Moving Range of the monthly averages. The limits were then "fixed" about the process mean. The centerline was adjusted only after we had determined that the process shift was indeed attributed to our process modifications, and only after we were comfortable that the change was permanent. (Remember we are collecting a minimum 100 Tab Length measurements per month). During the first five months of charting this key quality characteristic of the tabbing process was in-control (predictable) but not capable of meeting tab length specifications, i.e. Cpk was less than 1.0. A cross-functional team of operator, maintenance, and process engineers was formed to improve the tabbing capability. Possible causes for the tabbing variation were brainstormed and prioritized. Process Capability was plotted over time as process modifications were implemented. The monthly chart of Cpk clearly illustrates that the installation of new drives raised the process capability from the 0.6 vicinity to the 1.0 vicinity, and we knew that this special cause change (Cpk values outside the former limits) was 'significant'. Similarly, the next change (unwind brake) was illustrated to have raised the process capability from the 1.0 vicinity to the 1.3 vicinity; and the new controller raised it further to the 1.4 vicinity. The most recent change (reworked bars) is still being evaluated. Note that the decisions as to whether Cpk has improved and whether it meets the 1.3 standard are not based on a single value but rather on a collection of values, and the graphical representation helps visualize these judgements.

Through the use of this "process behavior chart" on Cpk we have been able to quickly and easily display the effect of various process changes on process capability, and to illustrate whether these process changes resulted in significant changes in Cpk (special cause) or merely random changes in Cpk (common cause). However, although the Individuals control chart of Cpk values helps illustrate process improvements over time and differentiate Special cause from Common cause variation month to month, it does not show the short-term underlying stability of the roll tabbing process. In other words, though a Cpk value can be calculated, is the underlying process stable and in a state of statistical control (a requirement of Cpk)? Tab Length X-bar and R charts were maintained by production operators in order to verify that the new process was stable, following each of the above process changes. The estimates of Cpk were then summarized monthly by the process engineer and used in the above chart.

## SUMMARY

Cpk values are estimates (statistics) calculated from a sample and used to estimate a population characteristic. These estimates are subject to variation (sampling variability). The larger the sample used to estimate Cpk, the smaller the variation in that estimate. A simple control chart can be used to illustrate this 'random variability in Cpk', and to help us identify when process changes have resulted in significant improvements in process capability.

### Reference:

(1) Franklin, L. A. and Wasserman, G. S., "Bootstrap Lower Confidence Limits for Capability Studies", JQT, Vol. 24 No. 4. and Quality Assurance for the Chemical and Process Industries, 2nd edition, (ASQ Quality Press).