ORIGINAL RESEARCH

A DMAIC approach for process capability improvement an engine crankshaft manufacturing process

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Received: 16 February 2014/Accepted: 15 May 2014/Published online: 27 May 2014 © The Author(s) 2014. This article is published with open access at Springerlink.com

Abstract The define-measure-analyze-improve-control (DMAIC) approach is a five-strata approach, namely DMAIC. This approach is the scientific approach for reducing the deviations and improving the capability levels of the manufacturing processes. The present work elaborates on DMAIC approach applied in reducing the process variations of the stub-end-hole boring operation of the manufacture of crankshaft. This statistical process control study starts with selection of the critical-to-quality (CTQ) characteristic in the define stratum. The next stratum constitutes the collection of dimensional measurement data of the CTO characteristic identified. This is followed by the analysis and improvement strata where the various quality control tools like Ishikawa diagram, physical mechanism analysis, failure modes effects analysis and analysis of variance are applied. Finally, the process monitoring charts are deployed at the workplace for regular monitoring and control of the concerned CTQ characteristic. By adopting DMAIC approach, standard deviation is reduced from 0.003 to 0.002. The process potential capability index ($C_{\rm P}$) values improved from 1.29 to 2.02 and the process performance capability index (C_{PK}) values improved from 0.32 to 1.45, respectively.

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Keywords Critical to quality (CTQ) characteristic · Cause and effect diagram · Statistical process control (SPC) · Process monitoring charts (PMC) · Failure modes and effects analysis (FMEA) · Analysis of variance (ANOVA) · Physical mechanism (PM) analysis

Introduction

An engine crankshaft forms the most dynamic and crucial part of an engine. The manufacture of the crank shaft poses challenges as it involves as many as 23 critical to quality (CTQ) characteristics. Table 1 enlists all the 23 CTQ characteristics associated with the crankshaft manufacturing operations. One such CTQ characteristic is the stubend-hole diameter formed by boring operation. This machining operation is important because, after this operation the finishing line operations of crankshaft journals' and pins' grinding and lapping succeed. So the stub-endhole boring operation is the last roughing line operation of the crank shaft and is also a bottleneck operation. This is the inspiration behind to select the stub-end-hole boring operation for process capability improvement study.

Statistical process control (SPC) widely employs various process monitoring charts for determining whether the process under consideration is performing within the specified limits are not. Process monitoring charts gives a graphical description of the process performance and it instantly helps the process personnel to differentiate chance causes from assignable causes. Process capability indices are the measure of efficiency of the process to produce the product within the specified dimensional tolerance limits. To be specific, C_P is the process potential capability index and C_{PK} is the process performance capability index. C_P gives a measure of the variation and deviation in the



Table 1 The project charter

Objectives

To recognize crankshaft stub end hole operation as a process capable operation

To relieve the stub end hole operation from being a bottle-neck and with a smooth work-in-flow without any staggered inventory

Deliverables and success metrics

To achieve the process potential capability index and process performance capability index, i.e., C_P and C_{PK} values for the stub end hole boring operation of the crankshaft to be >1.33, i.e., more than 4 sigma levels

The C_P and C_{PK} values to be achieved consistently >1.33 for over a persistent period of 3 months

Business impact

Raise the process capability levels and awareness of the importance of process monitoring charts in daily production. Reduce the component rejection and rework by 99 % in the first 6 months after sustenance

S. no.	Components of the project area	Value
1	Total no. of rejects and rework per day, i.e., three production shifts of 8 h each	=15 crankshafts
2	Time taken for segregation and rework of components	=1 h per day
3	Production loss due to rejection and rework per month	$=30 \times 1 = 30 \text{ h}$
4	Monetary loss of 1 h delay in the crankshaft cell	=\$7,000
5	Total monetary loss per month with 25 working days per month	$=25 \times \$7,000 = \$175,000$
6	By avoiding 99 % of rejections & rework, the amount that can be saved per month is	=0.99 × \$175,000 = \$174,240

Scope of this DMAIC project is limited to the Engine Plant of M/s Hindustan Motors Ltd, Pithampur Power Unit Plant, India

process. The higher the $C_{\rm P}$ value, the less the variation and deviation in the process. $C_{\rm PK}$, on the other hand, is obtained from $C_{\rm PKU}$ and $C_{\rm PKL}$. As a precaution of safety, the lower value among the $C_{\rm PKU}$ and $C_{\rm PKL}$ is considered to be the value of $C_{\rm PK}$. Elaborating on this, it can be said that centering of the process within the specification limits is done by $C_{\rm PK}$. It provides an indication whether the process is operating at the center of the specification limits.

Literature review

Schilling (1994), has thrown light on the superiority of process control over the traditional sampling techniques. Locke (1994), stressed on the importance of process charts, cause and effect relationship and control charts. Lin (2004), had emphasized on process capability indices for normal distribution. Tong et al. (2004), focused on define-measure-analyze-improve-control (DMAIC) approach and its



application for printed circuit board quality improvement. Li et al. (2008) adopted DMAIC approach to improve the capability of surface mount technology in solder printing process. Hwang (2006) employed the DMAIC procedure in context of application to manufacturing execution system. Gentili et al. (2006) applied the DMAIC process for a mechanical manufacturing process line, which manufactures both professional and simple kitchen knives. Sahay et al. (2011) used the DMAIC approach for analyzing the manufacturing lines of a brake lever at a Connecticut automotive component manufacturing company. Singh (2011) improved the process capability of polyjet printing for plastic components and charted the procedure for attaining the C_{pk} value attainment >1.33, i.e., >4 sigma level, which is considered as industrial benchmark. Lin et al. (2013) elaborated on the accurate yield assessment of the processes of multiple characteristics of the turbine blade manufacturing process. Kumaravadivel and Natarajan (2013) applied the cause and effect matrix and failure modes and effects analysis (FMEA) for solving problems associated with flywheel casting process. Mariajayaprakash et al. (2013) identified the CTQ characteristics of shock absorber manufacturing process and improved the process by minimizing the defects using Taguchi approach. Genetic algorithm was applied to optimize the parameters using Taguchi approach. Chen et al. (2013) applied Taguchi's orthogonal array and analysis of variance (ANOVA) to find the optimal values for the nine process parameters namely, injection time, injection pressure, packing time, packing pressure, cooling time, cooling temperature, mold open time, melt temperature and mold temperature, in a plastic injection molding process. Lal et al. (2013) in their paper discussed about the performance of piston manufacturing plant through stochastic models. They concluded that the time-dependent availability of the piston manufacturing plant is affected by fixture seat machining and circlip grooving.

The literature cited here reveals that the DMAIC approach is widely used for process capability improvements across the manufacturing sector. Hence, without any iota of doubt, this paper straightaway adopts the DMAIC approach for process capability improvement of the stubend hole boring operation of an engine crankshaft manufacturing process.

Definition stratum

The definition stratum starts with defining the project charter and then the mapping of the machining sequence flow of the crankshaft. It is followed by identifying the CTQ characteristic of interest, for the scope of improvement study.

Project charter

The project charter for the process capability improvement of the stub-end hole boring operation of the crankshaft machining process is depicted in Table 1. The project charter outlines the objectives, deliverables and success metrics of this improvement project. The business impact in terms of monetary benefits is also reflected in the project charter.

Process mapping

The process flow chart for machining line of the crankshaft manufacturing cell consists of the following machining operations sequence, as shown in the Fig. 1. The manufacturing sequence of the crankshaft consists a total of 15 operations. Of all the manufacturing operations, the operation number 90 is of concern. The operations from 10 to 90 constitute the roughing line operations involving mainly facing and centering, turning, whirling, drilling, tapping and chamfering. The operations from 100 to 150 constitute the finishing line operations involving mainly grinding, sursulfing and lapping.

Table 2 depicts the description of the machining operations of crankshaft manufacturing cell along with the associated CTQ characteristics.

Identifying CTQ characteristic

Timing belt pulley is located onto the stub-end-hole of the crank shaft. Hence if the stub-end-hole is undersize or oversize, it leads to incorrect fitment of the timing belt pulley and subsequently the timing belt. Ultimately, this leads to incorrect timing of fuel combustion within the cylinders and detonation and knocking. Therefore, from functional view-point, the Stub-end-hole diameter forms a CTQ characteristic. Second, from manufacturing view-point, the stub-end hole is the last operation performed on the roughing line and it forms the locating reference for the subsequently operations in the finishing line. Any small variations in the stub-end-hole boring operations are carried to the finishing operations of journal and pin grinding.

Hence, unanimously, the stub-end-hole boring operation forms the prime choice for identifying it as a CTQ characteristic. Figure 2 shows the pictorial representation of the CTQ characteristic.

Measurement stratum

The measurement stratum involves data collection of the critical to quality characteristic and is performed for 32 consecutive machined components. Data collection is performed in 3 iterations spanning for a period of 4 weeks each, i.e., about 900 consecutive components. The data are tabulated in the tabular form in Table 3 and the graphical plot of the three sets of measurement iterations is captured in Fig. 3, in the form of process monitoring charts.

Analysis stratum

The analysis stratum comprises of performing the calculations for the C_P and C_{PK} values across each iteration. In this stratum, the root cause analysis is performed with the help of various quality control (QC) tools like cause and effect diagram and physical mechanism analysis. Prioritization of the corrective actions is extracted from the output of process FMEA. This is followed by one-way ANOVA method of investigation to test the differences between the three iterations of the data sets.

Calculations of $C_{\rm P}$ and $C_{\rm PK}$

The calculations of $C_{\rm P}$ and $C_{\rm PK}$ are charted in Table 4. Based on the process monitoring charts and calculations of $C_{\rm P}$ and $C_{\rm PK}$ the following analysis is performed:

Iteration no. 1 primarily indicates the primitive status of the problem on hand before carrying out any improvement work. Continuous sets of measurements of the CTQ characteristic are taken and it is seen that the $C_{\rm P}$ and $C_{\rm PK}$ values here are below the target value of 1.33, with $C_{\rm P}$ value equal to 1.29 and $C_{\rm PK}$ equal to 0.32.



Fig. 1 Process flow chart



Table 2 Machining operations of connecting rod manufacturing cell

Machining operation no.	Description of machining operation	CTQ characteristic pertaining to the machining operation
10	Facing and centering	Center to center distance
		Locating hole diameter and chamfer angle
20	Web milling	Web flatness
30	CNC rough turning of journals	Journal diameter
40	Pin whirling	Phase-out of pins
		Throw of pins
		Pin diameter after whirling
		Journal diameter after whirling
50	Finish turning of journals	Journal diameter after finish turning
60	Oil hole gun drilling	Oil hole diameter
70	Flange end finish turning	Flange diameter
		Flange circularity
80	Bolt hole PCD drilling and tapping	Bolt hole pitch circle diameter
		Thread pitch of bolt holes
90	Stub end boring and	Stub-end-hole diameter
	chamfering using a combination boring-cum- chamfering inserts tool bar	Chamfer angle
100	Journal grinding	Journal finish diameter
		Journal circularity
110	Pin grinding	Pin finish diameter
		Pin circularity
120	Magnetic crack detection	Check for internal cracks
130	Sursulfing heat treatment	Hardness of journals and pins after sursulfing operation
140	Lapping of pins and journals	Surface roughness of journals and pins after lapping
150	Final inspection quality check	

Iteration no. 2 corresponds to the intermediary phase readings after performing moderate improvements like setting up a standard procedure for tool-insert setting on boring bar, cleaning the filter–regulator–lubrication (FRL) unit of pneumatic gauges and replacing the worn out cutting tool insert edge. In Iteration 2, a slight cyclic pattern is observed. This is because of the reason that there is a constant progressive wear out of the boring bar insert on continued boring operation. After every ten components being machined, the boring bar insert must be compensated



for the wear by elevating the insert by about five micrometers, over the diametric dimension. This exercise is performed with the help of a test mandrel and a boring bar tool insert setting Vee-block. Because of worn out Vee-surfaces of tool insert setting Vee-block, though the insert setting is done it is not accurate and this is reflected in the cyclic pattern observed in the second optimization step of Iteration 2. As a measure of corrective action, a new insert setting Vee-block is replaced with the old worn-out piece and the Vee surfaces of the Vee-Block are case hardened to achieve hardness up to 55 HRc with a case depth of 0.5–0.8 mm. In this iteration we see a marginal increase in $C_{\rm P}$ to 1.32 and $C_{\rm PK}$ to 0.90.

Finally, Iteration no. 3 corresponds to the final phase readings which are taken after introducing the corresponding corrective actions identified in the analysis stratum and bringing in a noticeable improvement in the process performance with $C_{\rm P} = 2.02$ and $C_{\rm PK} = 1.45$.

Cause and effect diagram

The cause and effect diagram, also known as Ishikawa diagram or fish-bone analysis, is a directional approach where the common as well as special causes are classified under the heading of 4 M's, i.e., man, material, method and machine. All the causes are directed towards the common effect namely, $C_{\rm P}$ and $C_{\rm PK}$ of the CTQ being <1.33. Figure 4 below charts the detailed cause and effect diagram pertaining to this research.

Through the cause and effect diagram, and the PM analysis, the various causes for the poor performance of the machining operation can be identified and corrective actions are taken upon based on the prioritization by the risk priority number (RPN) generated in the FMEA.

Physical mechanism analysis

The PM analysis or in other words physical mechanism analysis is a QC tool originated during the quality improvements under the Hinshitsu Hozen pillar of total productive maintenance. The same concept is applied to this research.

Physical mechanism analysis starts with the identification of the "physics" related to the machining operation under study. The direction of components of the forces acting at the "junction point" of the cutting tool and the workpiece are identified. Conceptually, this Junction-point of the contact of tool and workpiece is the area which involves the cutting forces acting on tool as well as on workpiece, heat transfer, frictional forces opposing the cutting force. It is at this point where the cutting action starts for single as well as multi-point cutting tool. The "junction point" involves three aspects, namely:



Fig. 2 Stub-end-hole diameter where, X = hole of $\emptyset 30.000^{(+0.020/0.000)}$. a Photograph of the crank shaft, b stub-end-hole modeled in CATIA V5R14

 Table 3 Dimensional readings of Stub-end-hole spanned over three iterations

S. no.	Iteration 1	Iteration 2	Iteration 3
01	30.003	30.007	30.010
02	30.000	30.004	30.010
03	30.004	30.012	30.008
04	30.005	30.010	30.007
05	29.998	30.012	30.008
06	30.005	30.006	30.006
07	30.004	30.007	30.006
08	30.001	30.008	30.007
09	30.006	30.005	30.005
10	29.999	30.007	30.005
11	30.004	30.006	30.010
12	29.999	30.003	30.009
13	30.000	30.011	30.009
14	30.004	30.008	30.008
15	30.002	30.005	30.008
16	30.000	30.007	30.007
17	30.005	30.010	30.006
18	30.005	30.005	30.006
19	30.000	30.007	30.005
20	30.006	30.006	30.005
21	30.002	30.002	30.010
22	30.006	30.010	30.008
23	29.999	30.005	30.008
24	30.006	30.007	30.009
25	30.001	30.005	30.007
26	30.003	30.004	30.007
27	29.999	30.007	30.007
28	30.005	30.003	30.006
29	30.000	30.008	30.006
30	30.003	30.005	30.005
31	30.000	30.008	30.005
32	30.005	30.007	30.006

Cutting tool aspect

Cutting tool insert indexing, insert changing for single point cutting tool like boring bar; cutting tool dressing for a multi point cutting tool like grinding wheel; cutting tool holder location and clamping aspects; calibration of cutting tool setting in the tool-pre-setting area.

Workpiece aspect

Work location, work holding, work clamping, work supporting, power required for driving the work and related machinery, coolant circulation continuity for flushing out the chips generated in the cutting process, heat dissipation by the coolant,

Measurement aspect

Measurement system comprising of the go and no-go gauges on the shop floor, calibration of gauges, the pneumatic pressure fluctuation in the pressure lines corresponding to the pneumatic gauges, the FRL unit maintenance, in-process sensing instrument sensitivity and repeatability.

Process failure modes and effects analysis (FMEA)

Process FMEA is an approach for prioritizing the sequence of corrective actions. Based on the severity, occurrence and detection ratings the RPN is calculated. The RPN >100 are of concern and liable for corrective action. The detailed FMEA sheet for stub-end-hole boring operation is tabulated in Table 5.

Among the various enlisted causes, the cause which most affects and responsible for the poor performance of



Fig. 3 Process monitoring charts

Table 4 Calculations of $C_{\rm P}$ and $C_{\rm PK}$

Formula	Iteration 1	Iteration 2	Iteration 3
USL	30.020	30.020	30.020
LSL	30.000	30.000	30.000
σ	0.003	0.003	0.002
$C_{\rm P} = \frac{({\rm USL-LSL})}{6\sigma}$	1.29	1.32	2.02
$C_{\rm PKU} = \frac{(\rm USL-MEAN)}{3\sigma}$	2.26	1.75	2.60
$C_{\rm PKL} = \frac{(\rm MEAN-LSL)}{3\sigma}$	0.32	0.90	1.45
$C_{\rm PK} = \min \left(C_{\rm PKU,} \ C_{\rm PKL} \right)$	0.32	0.90	1.45

the CTQ characteristic, is found by using the ANOVA technique.

Analysis of variance (ANOVA)

ANOVA starts with the formulation of the hypothesis to be tested, followed by, tests for the assumption about the normality of the data and the homogeneity of variance among the sets of the data.

A post hoc analysis is required if $F_{\text{STATISTIC}}$ is found to be $>F_{\text{CRITICAL}}$.

Formulating the hypothesis and testing the assumptions for normality of data and homogeneity of variance

The null hypothesis (H_0) and the alternate hypothesis (H_1) can be formulated in the present context as

 $H_0: \mu_i = \mu$ all i = 1, 2, 3, $H_1: \mu_i = \mu$ for some i = 1, 2, 3,



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where μ_i is the population mean for level *i*, and μ is the overall grand mean of all levels.

With respect to the data in Table 3, it is seen that there are 3 levels (i.e., 3 iterations) with each level consisting of 32 measurement readings of stub-end-hole diameter of connecting rod. The data plot for normality is captured in Fig. 5, and the histogram bar chart is captured in Fig. 6. The pre-requisite for performing one-way ANOVA test is to find departure from normality and to check the homogeneity of variance among the sets of the data. The normal probability plot is seen linear with equispaced values and this is supported by the histogram with near bell-shaped curvature. The *P* value is <0.005, which is less than the α value of 0.05, thereby indicating that a linear relationship exists with normality retained. The results are further strengthened by the fact that there are no unusual data points. The sample size of 32 is sufficient to detect differences among means and because all the sample sizes are >15, normality is not an issue. Also based on the data observations and alpha level of 0.05, there is at least a 90 % chance of detecting a difference of standard deviation of 0.0023297 and at most a 60 % chance of detecting a difference of standard deviation of 0.0013632.

Finding the F_{STATISTIC}

The ANOVA Table obtained from Minitab software is captured in Table 6. Here it is seen that

$$F_{\text{STATISTIC}} = 41.24. \tag{1}$$

An α value of 0.05 is typically used, corresponding to 95 % confidence levels. If α is defined to be equal



Fig. 4 Cause and effect diagram

to 0.05, then the critical value for rejection region is $F_{\text{CRITICAL}(\alpha, K-1, N-K)}$ and is obtained to be 3.094. Thus, $F_{\text{CRITICAL}} = 3.094$. (2)

Hence, it is seen that

$$F_{\text{STATISTIC}} > F_{\text{CRITICAL}}.$$
(3)

Therefore, the decision will be to reject the null hypothesis. If the decision from the ANOVA is to reject the null hypothesis, then it indicates that at least one of the means (μ_i) is different from the remaining other means. In order to figure out where this difference lies, a post hoc ANOVA test is required.

Post-hoc ANOVA test

Since here the sample sizes are same, we go for the Tukey's test for conducting the Post-hoc ANOVA test. In Tukey's test, the honestly significant difference (HSD) is calculated as

$$\text{HSD} = q \sqrt{\frac{\text{MST}_{\text{E}}}{n}} = 3.38 \sqrt{\frac{0.53 \times 10^{-5}}{32}} = 0.00137, \quad (4)$$

where q is the studentized range statistic which is equal to a value of 3.38, for a df of 93 and k = 3, i.e., number of levels as 3.

If "C1" denotes "Iteration 1", "C2" denotes "Iteration 2" and "C3" denotes for "Iteration 3", then from Minitab

software, the following grouping information using Tukey method is shown in Table 7:

From Table 7, it is seen that means that do not share a letter are significantly different, i.e., Iteration 1 is significantly different from Iterations 2 and 3.

The pairwise comparison using Minitab is depicted in Table 8.

In Table 8, it is seen that the pairwise comparison between C2 and C3 is 0.000375 which is less than HSD is Eq. (3), whereas the pairwise comparison between C1 and C2 is 0.004312 and that between C1 and C3 is 0.004687, which are greater than that of the HSD in Eq. (4), with the difference 0.004687 being the largest. So, it is deduced that the differences are statistically significant. Hence, it is concluded that among all the different causes enumerated in the cause and effect diagram, the most influencing causes are the worn out cutting tool insert, insert setting v-block wear out and non-calibration of the vernier calipers and tool setting mandrel.

Control stratum

In the control stratum, the X-bar and R control charts are implemented at the workplace for monitoring the process and preventing it from deviating. These charts ensure that the process remains capable and is prevented from



Table 5 FMEA sheet

Process name	Potential failure	Potential effect	Severity	Potential cause	Occurrence	Current controls	Detection	RPN
Stub-end- hole boring operation number 90	Stub-end-hole oversize	Component rejection	7	Inaccurate insert pre-setting on the boring bar by operator	7	Tool-presetting check- list	8	392
		Loose fit of timing belt pulley in assembly	9	Tool insert setting v-block wear out	6	Calibration checklist included in measurement system analysis (MSA)	3	162
				Calibration of tool insert setting mandrel	5	Calibration checklist included in measurement system analysis (MSA)	3	135
				Calibration of vernier caliper	5	Calibration checklist included in measurement system analysis (MSA)	3	135
	Stub-end-hole undersize	Component rework	7	Overused and worn out cutting tool inserts	7	Tool-indexing check- list	4	192
				Calibration of vernier caliper	5	Calibration checklist included in measurement system analysis (MSA)	2	70
				Insert pre-setting on the boring bar	7	Tool-presetting check- list	3	147
	Geometrical dimensional variations of stub-end hole	Variations in axiality, cylindricity and position of hole	6	Loose Insert and insert loosening in the middle of operation	2	Insert torque check- list	2	24
				Worn out locating pads on the locating Vee of fixture	3	Fixture calibration check-list	2	36

Fig. 5 The normal plot of residuals for stub-end hole dimensional observations







Fig. 6 Residual histogram for stub-end hole dimensional observations

Table 6 The ANOVA table

One-way ANOVA: Iteration 1, Iteration 2, Iteration 3							
Source	DF	SS	MS	F	Р		
Factor	2	0.0004342	0.0002171	41.24	0.000		
Error	93	0.0004897	0.0000053				
Total	95	0.0009239					

Table 7 Grouping information using Tukey method

Iteration column 'C'	Ν	Mean	Grouping
C3	32	30.007156	А
C2	32	30.006781	А
C1	32	30.002469	В

Table 8 Pairwise comparisons of iterations

S. no.	Pairwise comparison	Value
1	C1 subtracted from: C2	0.004312
2	C1 subtracted from: C3	0.004687
3	C2 subtracted from: C3	0.000375

deviations. The X-Bar and R chart are shown in Fig. 7. It can be deduced that the all the mean and the range values are within the upper and lower control limits and that there are no outliers and no out-of-control subgroups. The data points are also equally distributed across the mean line. Thus, the process mean as well as the process variation are stable and the process is controlled. The causal factor matrix is summarized in Table 9.

Results

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It is seen that replacement of the worn-out insert tip was the major contributor followed by wear out of tool setting mandrel, insert setting v-block and calibration of air gage, air pressure regulation and air filter maintenance. As a part of standardizing the process, the following activities are carried out:

- After every 400 components being bored, the indexing 1. of the cutting edge of tool insert is done.
- 2. Regular machine maintenance schedule has been established and regular checks are included in the checklist.
- 3. After every 10 components bored, the pneumatic gauge is calibrated with the standard ring gauge corresponding to the pneumatic gauging system.
- 4. The gauge calibration is done periodically as a part of measurement system analysis and properly calibrated pneumatic gauge is used at the workplace.
- After every 2 months i.e., after about (1,600)–(1,800) 5. components, the FRL unit maintenance is incorporated in the preventive maintenance checklist.
- 6. Coolant recirculation pressure is set at a value of around 3.0 kgf/sq.cm.
- 7. Insert presetting on the boring-bar is carried out with the help of a portable Vee-block insert setting gauge on the horizontal boring-bar.

The improvement in the process performance is fostered by the fact that the sigma levels (standard deviation) are reduced from 0.003 to 0.002 as captured in Table 3.

Conclusion

This paper traces the DMAIC approach for improving the process capability levels of the stub-end-hole boring operation of the crankshaft manufacturing process. The QC tools predominantly used for tracing out the causes for poor process performance are the Ishikawa diagram, physical mechanism analysis and the failure modes and effects analysis. The process monitoring charts are employed at the workplace for monitoring the process performance and preventing it from deviations. In order to trace out the extent of influence of the causes identified, the ANOVA procedure is adopted. The predominant causes identified are worn out cutting tool inserts, worn out insert setting v-block and non-calibration of the vernier calipers and tool setting mandrel. Finally, on eliminating the causes one-byone, the process potential capability index $(C_{\rm P})$ showed an improvement from 1.29 to 2.02 and the process performance capability index ($C_{\rm PK}$) improved from 0.32 to 1.45.





Fig. 7 X-Bar and R control chart

Table 9 Causal factor matrix

S. no.	Causal factor	Pre-level	Post-level	$C_{\rm P}/C_{\rm PK}$	Iteration level
1	Overused cutting tool inserts	Before indexing the inserts	After indexing the inserts	1.29/0.32	Iteration level 1
2	Calibration of vernier caliper	Before calibration	After calibration		
3	Worn out cutting tool insert edge	Before replacing the worn-out cutting tool insert edge	After replacing the worn-out cutting tool insert edge	1.32/0.90	Iteration level 2
4	Gauge air-pressure variation	Before pressure regulation at 10 kg	After pressure regulation at 13 kg		
5	Cleaning of air-filter of the FRL unit	Before cleaning	After cleaning		
6	Pneumatic gauge calibration	Before calibration	After calibration		
7	Insert pre-setting on the boring bar	Before presetting	After presetting at a value of 15.005 mm		
8	Coolant recirculation pressure	Coolant recirculation pressure at 7 kg	Coolant recirculation pressure at 12 kg		
9	Tool insert setting v-block wear out	Before replacing the insert setting v-Block	After replacing the insert setting v-Block	2.02/1.45	Iteration level 3
10	Calibration of tool insert setting mandrel	Before calibration	After calibration		
11	Untrained operator	Before on-job Training	After on-job Training		
12	Worn out locating pads	Before replacing the worn-out pads	After replacing the worn-out pads		



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