# Method of Increase of Check Efficiency Based on Small Size Samples 

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#### Abstract

In presented article is offered one more decisive rule of selective check - the equivalent operational characteristic (EOC) which is based on the Weibull distribution. Settlement formulas are given and the table of an assessment of accuracy of calculation of predicted reject by the offered technique is given. The table of the comparative analysis of accuracy of all offered before methods and the existing standard is also submitted. It is proved that use of EOC based on the Weibull distribution gives more exact (by 1,05-5,30 times) results, than use of EOC based on other distribution laws, and $\mathbf{2 , 5 - 8 , 8}$ times more precisely, than operating methods of boundary check on samples of small size sample.


Key words - sample check, the Weibull distribution, equivalent operational characteristic.

## I. Introduction

By production of crystals of integrated microcircuit (IMC) there was a practice when judge quality of 400-5000 working crystals by results of control measurements in 5 (or 10) the test cells (TC). Such ratio of objects of the control and certified samples isn't provided by any operational characteristics therefore on production were compelled to resort to an artificial method of borders that led to considerable false reject with big economic losses [1].

In the previous works [2-6] we found methods of essential increase of accuracy of forecasting of reject due to use of the method of pointed distributions (MPD) which is based on knowledge of the law of distribution of checked parameter, and the equivalent operational characteristic (EOC). As a result it is proved that the subjective component of false reject depending generally on an error of calculation of a mean square deviation (SD) decreased by 2,5-5,3 times that at introduction in production gives considerable economic effect. In article on the basis of development of MTR ways of further increase in accuracy of an assessment of possible reject in relation to each plate of party on the example of the most widespread law of distribution by production of crystals of IMC - Weibull's law which special cases an exponential and normal laws are considered.

## II. Modifications of the Method of Pointed DISTRIBUTIONS

Since the end of the 90th years and to the present by us it is offered and the method of pointed distributions (MPD) [2] which principle is based on rather wide circulation: each measurement should be considered as distribution center with the known law. It allows to pass from the volume of initial small sample of $\mathbf{n}$ to equivalent (virtual) sample which volume of $\mathbf{n}_{\mathbf{e}}$ can be found through Kolmogorov's $D$-statistics and $D_{n^{-}}$ statistics similar to it considering that the amount of information in both cases is identical to the same confidential probability

$$
\begin{equation*}
n_{e}=n\left(\frac{D}{D_{n}}\right)^{2} \tag{1}
\end{equation*}
$$

An assessment of a relative error of one of the most sensitive statistical characteristics $-S$ average quadratic deviation (AQD) - we will conduct on the formula offered by Y.B. Shor [8]

$$
\begin{equation*}
\varepsilon=\frac{\sigma(S)}{\sigma} \cdot 100 \%=\sqrt{\frac{1}{2 n-1,4}} \cdot 100 \% \tag{2}
\end{equation*}
$$

Results of calculations for some volumes of initial sample and two types of laws of distribution at confidential probability $P_{\text {conf }}=0,95$ are presented in tables 1 and 2 . From this tables it is visible that application of MPD allows to increase the accuracy (to reduce a mistake) of calculation by 1,7-2,0 times in case of bilateral distribution and by 2,3-2,5 times in case of unilateral distribution. Improvement of a method of pointed distributions led to creation of the robust method of pointed distributions (RMPD) [4] and at the same time combined method of pointed distributions (KMPD) [5].

The Robust method of pointed distributions differs from initial MPD in use of a steady (robust) assessment of HodgesLehman of an arithmetic average of sample on Walsh's averag-
es. Thus the accuracy of calculations of selective AQD increases in average by 1,2-1,5 times in comparison with classical methods (see tab. 1 and 2).

Unfortunately, RMPD at calculation of the central moments of the second and the highest orders (and, so and AQD) had a systematic mistake because of a little reduced scope of equivalent sample. For elimination of this shortcoming, it is
offered to add the volume of initial sample of $\boldsymbol{n}$ to the intermediate volume of equivalent robust sample of $\boldsymbol{n}_{e}$ of uniforms. Results of calculations are presented in tables 1 and from which analysis it is clear that the accuracy of calculations of selective AQD for KMPD increases in average by 1,5-2,0 times in comparison with MPD or by 2,9-3,4 times in comparison with classical methods of calculation.

TABLE 1. Real n and Equivalent ne Volumes of Samples for Different Modifications of Mpd

| Initial sample |  | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  | MPD | 10 | 14 | 17 | 20 | 23 | 25 | 27 | 29 |
|  | PMPD | 10 | 20 | 29 | 43 | 59 | 77 | 82 | 92 |
|  | KMPD | 20 | 29 | 42 | 59 | 77 | 82 | 92 | 108 |
|  | OMPD | 36 | 51 | 62 | 73 | 85 | 94 | 102 | 112 |
|  | MPD | 13 | 23 | 33 | 34 | 37 | 40 | 41 | 42 |
|  | PMPD | 13 | 34 | 42 | 55 | 66 | 79 | 83 | 96 |
|  | KMPD | 33 | 42 | 55 | 66 | 79 | 83 | 96 | 108 |
|  | OMPD | 49 | 80 | 97 | 108 | 119 | 135 | 145 | 154 |

TABLE 2. Relative Errors of Selective AQD for Different Modifications of MPD, \%

| Sizes of initial samples, $n$ |  | 3 | 4 | 5 | 6 | 7 | 8 | 9 | 10 |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
| $\begin{aligned} & \text { Bilateral distri- } \\ & \text { butions } \end{aligned}$ | Classic. | 46,6 | 38,9 | 34,1 | 30,7 | 28,2 | 26,2 | 24,5 | 23,2 |
|  | MPD | 23,2 | 19,4 | 17,5 | 16,1 | 15,0 | 14,3 | 13,8 | 13,3 |
|  | PMPD | 23,2 | 16,1 | 13,3 | 10,9 | 9,3 | 8,1 | 7,8 | 7,4 |
|  | KMPD | 16,1 | 13,3 | 11,0 | 9,3 | 8,1 | 7,8 | 7,4 | 6,8 |
|  | OMPD | 11,8 | 9,9 | 9,0 | 8,3 | 7,7 | 7,3 | 7,0 | 6,7 |
|  | MPD | 20,0 | 15,1 | 12,4 | 12,3 | 11,7 | 11,3 | 11,1 | 11,0 |
|  | PMPD | 20,0 | 12,3 | 11,0 | 9,6 | 8,8 | 8,0 | 7,8 | 7,2 |
|  | KMPD | 12,3 | 11,0 | 9,6 | 8,8 | 8,0 | 7,7 | 7,2 | 6,7 |
|  | OMPD | 10,1 | 7,9 | 7,2 | 6,8 | 6,5 | 6,1 | 5,9 | 5,7 |

For understanding of further steps on increase in accuracy of calculations it was carried out graphic-analytical comparison of a classical method, MPD, KMPD on a concrete production example.

Let at the next operation of selective control in production of integrated chips crystals on one of plates sample with
five test cells 9,1 was received; 10,$2 ; 11,5 ; 12,8 ; 16,9$ Ohms/sq, distributed in normal law. Then the interval estimates of AQD $S$ received by various methods can be presented in the drawing form.


Fig. 1. Graphic interpretation of interval estimates of $A Q D$ by various methods.

As in all cases initial sample is the same, and the difference consists only in methods of calculation of its parameters thus the fundamental principle is observed: general parameter with the established confidential probability are contained between the lower and top borders of an interval assessment, further increase of accuracy of estimates is reached by use as a support of the left border of an interval assessment of MPD, and the right border - an interval assessment of KMPD for the same parameter arises. Then all estimates of a new method the integrated method of pointed distributions (IMPD) - can be calculated as if back proceeding from borders of an interval assessment of AQD - left for MPD and right for KMPD. Results of calculations are presented in last lines of table 1 and 2.

The analysis of these results showed that OMPD allows to increase the accuracy of calculation of selective AQD (reduce a calculation error) in average by twice in comparison with MPD, by 3,5-4,0 times in comparison with classical methods for unimodal symmetrical distribution, by 1,3-2,0
times in comparison with MPD and by 4,0-4,6 times in comparison with classical methods for unilateral laws of distribution.

## III. Increase of Accuracy of Estimates of Control SAMPLE PARAMETERS

Function of Weibull distribution can be presented in the form [5]

$$
\begin{equation*}
F(x)=1-\exp \left[-\left(\frac{x-\theta}{b}\right)^{\eta}\right] \tag{3}
\end{equation*}
$$

and its density (probability density) is

$$
\begin{equation*}
f(x)=\frac{\eta}{b}\left(\frac{x-\theta}{b}\right)^{\eta-1} \exp \left[-\left(\frac{x-\theta}{b}\right)^{\eta}\right], \tag{4}
\end{equation*}
$$

where $b$ - scale parameter (sometimes $b=\frac{1}{\lambda}$ ); $\eta$ - form parameter; $\theta$ - shift parameter (fig. 2).


Fig. 2. Density of probability (1) and function of Weibull distribution (2) at $b=1, \theta=0$.

Calculation of parameters of distribution of Weibull represents a complex challenge [5], however at $\eta \geq 1$ it can be considerably simplified by means of the following approximations [6, c.151]:

$$
\begin{gather*}
\hat{\eta}=4,8\left(r_{3}+1,23\right)^{-1,4} ; \hat{b}=\frac{\delta}{G} ; \theta=\bar{X}-\delta ; \\
\delta \approx\left(0,5+0,784 \hat{\eta}-\frac{0,35}{\hat{\eta}}\right) \cdot S ; \\
G=\Gamma\left(1+\frac{1}{\eta}\right) \approx 1-0,427(\eta-1) \eta^{-1,9}, \tag{5}
\end{gather*}
$$

where $\bar{X}-$ a sample average; $S-\mathrm{AQD} ; r_{3}$ - the third main issue.

If shift is absent $(\theta=0)$, estimates of parameters have an appearance

$$
\begin{equation*}
\hat{\eta} \approx \frac{n-1}{n}\left(0,465 \frac{s}{\bar{X}}+1,282 \frac{\bar{x}}{s}-0,7\right) ; \hat{b}=\frac{\bar{X}}{G} . \tag{6}
\end{equation*}
$$

Separate samples by the volume of $n=5$ or $n=10$ (quantity of the test cells (TC) on each plate) can be checked for compli-
ance to Weibull's distribution by means of Smirnov-Kramervon Mises criterion

$$
\begin{equation*}
n \omega^{2}=\frac{1}{12 n}+\sum_{i=1}^{n}\left\{F\left(X_{i}-\frac{2 i-1}{2 n}\right)\right\}^{2} \leq n \omega^{2}\left(P_{\text {conf }}\right), \tag{7}
\end{equation*}
$$

where $F\left(X_{i}\right)$ - theoretical function of distribution.
It is necessary to remember that theoretical function of distribution (in our case Weybull distribution [7]) has to be known to within parameters. By researches [8] it is established that use of $F(X)$ as a function of distribution with the parameters estimated on sample (a widespread mistake!) leads to increase in quantity of errors of the second sort.

We looked for the solution of an objective by the way we have already found: equivalent (virtual) increase in volume of sample by means of the method of pointed distributions (MPD) [2] and application of the equivalent operational characteristic (EOC) [3].

To use a formula (3) or (4) for determination of parameters of Weibull distribution, it is necessary to find an arithmetic
average of both the second central and the third main moments of each concrete sample (i.e. each concrete plate). From the point of view of MPD it means to find [9]

$$
\begin{gather*}
m_{X}^{*}=\frac{\sum_{j=1}^{k} \sum_{i=1}^{n} p_{i j} \cdot X_{j}^{\prime} \cdot \exp \left[-4,5\left(\frac{X_{j}^{\prime}-X_{i}}{\rho}\right)^{2}\right]}{\sum_{j=1}^{k} \sum_{i=1}^{n} p_{i j} \cdot \exp \left[-4,5\left(\frac{X_{j}^{\prime}-X_{i}}{\rho}\right)^{2}\right]}  \tag{8}\\
\mu_{2}^{*}=\frac{\sum_{j=1}^{k} \sum_{i=1}^{n} p_{i j} \cdot\left(X_{j}^{\prime}\right)^{2} \cdot \exp \left[-4,5\left(\frac{X_{j}^{\prime}-X_{i}}{\rho}\right)^{2}\right]}{\sum_{j=1}^{k} \sum_{i=1}^{n} p_{i j} \cdot \exp \left[-4,5\left(\frac{X_{j}^{\prime}-X_{i}}{\rho}\right)^{2}\right]}-\left(m_{X}^{*}\right)^{2}  \tag{9}\\
\mu_{3}^{*}=\frac{\sum_{j=1}^{k} \sum_{i=1}^{n} p_{i j} \cdot\left(X_{j}^{\prime}\right)^{2} \cdot \exp \left[-4,5\left(\frac{X_{j}^{\prime}-X_{i}}{\rho}\right)^{2}\right]}{\sum_{j=1}^{k} \sum_{i=1}^{n} p_{i j} \cdot \exp \left[-4,5\left(\frac{X_{j}^{\prime}-X_{i}}{\rho}\right)^{2}\right]}-3 \mu_{2}^{*} m_{X}^{*}-\left(m_{X}^{*}\right)^{3}  \tag{10}\\
r_{3}^{*}=\frac{\mu_{3}^{*}}{\left(\sqrt{\mu_{2}^{*}}\right)^{3}} . \tag{11}
\end{gather*}
$$

Equivalent size of samples for Weibull distribution can be found from expression (10)

$$
\begin{equation*}
n_{e}=\sqrt{-713,7+301,2 n-6,907 n}, \tag{12}
\end{equation*}
$$

that means for $n=5 \rightarrow n_{e}=25$; for $n=10 \rightarrow n_{e}=40$.
Then equivalent operational characteristic is

$$
P(q)=F_{0}\left(\left(U_{1-q}-\frac{k_{s}}{k_{n}}\right) / \sqrt{\frac{1}{n}+\frac{k_{s}^{2}}{2 n-1,4}}\right),
$$

where $k_{n}=\sqrt{\frac{n-1}{2}} \cdot \Gamma\left(\frac{\mathrm{n}-1}{2}\right) / \Gamma\left(\frac{\mathrm{n}}{2}\right)-$ correction index, and $F_{0}(\cdot)$

- Gauss's integral, at substitution of volume $\mathbf{n}$ of initial sample by volume $\mathbf{n}_{\mathbf{e}}$ of equivalent samples. Taking into account concrete values of admissible errors of the first sort (for example, $\alpha=0,10$ ), and also threshold of $100 \%$ acceptance $q_{0}$ (for example, $q_{0}=0,10$ ), we will receive
for $n=5$ and $n_{e}=25$ value $k_{n}=k_{25}=0,10105 ; k_{s}=1,036$; for $n=10$ and $n_{9}=40$ value $k_{n}=k_{40}=0,10064 ; k_{s}=1,107$. Then equivalent operational characteristics will be: for $n=5$

$$
\begin{equation*}
P(q)=F_{\circ}\left(\frac{U_{1-q}-1,025}{0,2492}\right)=\frac{\left|T-m_{x}^{*}\right|}{\sqrt{\mu_{2}^{*}}} ; \tag{13}
\end{equation*}
$$

for $n=10$

$$
\begin{equation*}
P(q)=F_{0}\left(\frac{U_{1-q}-1,100}{0,2015}\right)=\frac{\left|T-m_{x}^{*}\right|}{\sqrt{\mu_{2}^{*}}}, \tag{14}
\end{equation*}
$$

where $T$ - the upper or lower bound of norm, or, taking into account the law of distribution, will look as

$$
\begin{equation*}
P(q)=\exp \left[-\left(\frac{x-\theta}{b}\right)^{\eta}\right] \tag{15}
\end{equation*}
$$

We will remind that range of the predicted reject has dual origin from the objective and subjective reasons. The objective reasons are the following: the volume of sample n , coefficient of overlapping of norm $v=\left(X_{\max }-X_{\min }\right) /\left(T_{B}-T_{H}\right)$ and arrangement of concrete values of controlled parameter on a numerical axis (see fig. 3), to the subjective reasons refers a big mistake of calculation of a mean square deviation (table 3).


Fig. 3. Options of an arrangement of measurements at $m=4$ and $v=1$.
a) minimum reject; b) the most probable reject; c) maximum reject

TABLE 3. Sizes of the Predicted Reject (\%) on a Plate With Five Test Cells (NE=25)

| The predicted reject | $\sqrt{\mu_{2}^{*}}$ | Number of the measurements which aren't going beyond norm, $m$ |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\mathrm{m}=5$ |  |  | $\mathrm{m}=4$ |  |  |  | m=3 |  |  |  |
|  |  | $v=1,0$ | $\boldsymbol{v}=\mathbf{0 , 7}$ | $v=0,5$ | $v=1,3$ | $v=1,0$ | $v=0,7$ | $\boldsymbol{v}=\mathbf{0 , 5}$ | $\boldsymbol{v}=\mathbf{2 , 0}$ | $v=1,0$ | $v=0,7$ | $v=0,5$ |
| Minimum | Min. | - | 0,6 | 0,1 | 4,7 | 2,2 | 2,8 | 3,3 | 13,3 | 5,9 | 4,9 | 4,9 |
|  | Aver. | - | 3,6 | 0,6 | 12,3 | 8,2 | 14,9 | 15,9 | 25,6 | 19,9 | 19,8 | 19,5 |
|  | Max. | - | 6,5 | 2,6 | 19,4 | 13,3 | 14,9 | 16,1 | 32,5 | 21,6 | 19,9 | 19,6 |
| The most probable | Min. | - | 1,0 | 0,7 | 4,9 | 3,5 | 3,4 | 4,0 | 14,3 | 7,9 | 6,0 | 5,9 |
|  | Aver. | - | 7,1 | 6,9 | 15,2 | 14,3 | 16,4 | 17,8 | 26,9 | 23,1 | 21,4 | 21,4 |
|  | Max | - | 8,8 | 7,3 | 19,8 | 16,6 | 16,4 | 17,8 | 33,8 | 25,1 | 21,8 | 21,6 |
| Maximum | Min. | 2,2 | 1,9 | 2,3 | 6,6 | 6,1 | 3,5 | 4,7 | 14,7 | 9,4 | 8,0 | 6,4 |
|  | Aver. | 8,0 | 11,9 | 13,5 | 19,7 | 20,6 | 16,6 | 19,3 | 27,8 | 25,3 | 24,7 | 22,2 |
|  | Max. | 13,1 | 12,2 | 13,6 | 22,9 | 21,9 | 16,6 | 19,3 | 34,2 | 27,3 | 25,3 | 22,5 |

The similar table can be provided for a plate case with ten test cells.

For evaluation of reject forecasting accuracy (or decrease in a share of subjective factors in forecasting), it is necessary to compare results of calculations for different

| The predicted reject | AQD | Number of the measurements which aren't going beyond norm, $m$ |  |  |  |  |  |  |  |  |  |  |
| :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: | :---: |
|  |  | $\mathrm{m}=5$ |  |  | $\mathrm{m}=4$ |  |  |  | $\mathrm{m}=3$ |  |  |  |
|  |  | $v=1,0$ | $\boldsymbol{v}=\mathbf{0 , 7}$ | $v=0,5$ | $v=1,3$ | $v=1,0$ | $\boldsymbol{v}=0,7$ | $v=0,5$ | $v=2,0$ | $v=1,0$ | $\boldsymbol{v}=0,7$ | $v=0,5$ |
| According to the existing standards | Min. | - | 8,83 | 16,58 | 3,96 | 5,22 | 3,89 | 2,94 | 2,56 | 3,22 | 3,01 | 2,58 |
|  | Aver. | - | 6,56 | 5,09 | 3,88 | 4,20 | 3,59 | 2,75 | 2,50 | 2,81 | 2,82 | 2,46 |
|  | Max. | 5,30 | 4,75 | 3,42 | 3,35 | 3,20 | 3,60 | 2,62 | 2,50 | 2,62 | 2,49 | 2,48 |
| On EOC <br> of the normal law | Min. | - | 5,28 | 12,81 | 1,84 | 2,65 | 2,30 | 2,07 | 1,11 | 1,63 | 1,73 | 1,70 |
|  | Aver. | - | 3,90 | 4,56 | 1,80 | 2,13 | 2,09 | 1,88 | 1,07 | 1,41 | 1,58 | 1,55 |
|  | Max.. | 2,68 | 2,81 | 2,46 | 1,55 | 1,63 | 2,07 | 1,73 | 1,05 | 1,30 | 1,36 | 1,49 |
| On EOC of the exponential law | Min. | - | 1,26 | 1,27 | 1,25 | 1,26 | 1,27 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 |
|  | Aver. | - | 1,25 | 1,25 | 1,25 | 1,25 | 1,26 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 |
|  | Max. | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 | 1,25 |

## IV. CONCLUSION

The analysis of table 4 showed that application of the equivalent operational characteristic on base of Weibull distribution law yields the results by $2,5-8,8$ times more exact, than the existing standards, by $1,05-5,30$ times more exact, than EOC with normal distribution law, by 1,25 times more exact, than EOC with exponential distribution (sharply different results: 16,8 in the first case and 12,81 - in the second, didn't take into account as casual deviations). Thus, all offered methods of assessment of results of control on small size samples based on equivalent operational characteristics taking into account laws of distribution (especially Weibull distribution), can be recommended for implementation in the production.

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methods. For this purpose in table 4 are presented the relation of the predicted reject by three techniques to the predicted reject on EOC taking into account Weibull distribution.

> TABLE 4. Relative Reduction of the Maximum Reject (Times)
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