
THE MATHEMATICAL BASIS FOR REALISTIC SAMPLE SIZES IN SUCCESS RUN TESTING
(DERIVATION OF THE COMPRESSED SUCCESS RUN THEOREM)

INTRODUCTION

Test sample sizes for success runs are of primary interest in demonstrating a product's reliability to a specified life target. However, the sample sizes required according to the classical assumption used about the population fraction defective in order to demonstrate reliability with some desired confidence level are so ridiculously large that no industrial manager involved in the testing of expensive specimens or assemblies will ever willing to test that many items, for he knows by his own good common sense that such large sample sizes are a bunch of mathematical garbage. Why is this? That's a good question which needs to be answered to everyone's satisfaction before any success run test programs are designed. It is our intention in this bulletin to derive the compressed success run theorem, which should be used in designing success run tests to specific life targets. By using the compressed success run theorem we arrive at very reasonable sample sizes which are only a small fraction of the ridiculously large sample sizes required by the classical approach.

GEOMETRICAL REPRESENTATION OF THE PROBLEM

The problem we are discussing involves the compressed linear interval containing possible fractions failed at some target life for the product to be manufactured and sold.

Let F_0 = Worst Possible Fraction of Items Failed in a Given
Target Service Period

Then, the range of possible values for the fraction defective to the target ranges from 0 to F_0 . This is pictured geometrically in Figure 1 below:

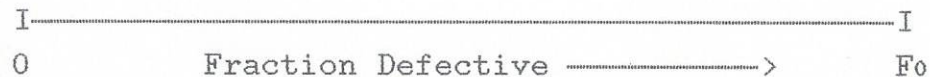


FIGURE 1

Let X_0 = Testing Target (i.e., the test period to which we get
a success run of N consecutive successes.)

QUESTION: What is the population defective at X_0 with confidence C of
not being exceeded if we obtain N consecutive successes to X_0 ?

We pictured this situation in Figure 2, where we start out with a prior Rectangular Distribution between 0 and F_0 . (Worst of $N + 1$ at Z)

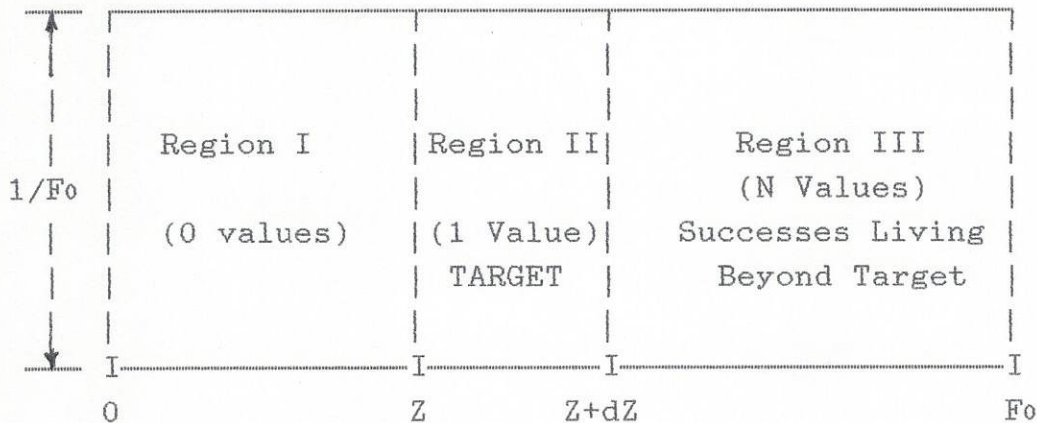


FIGURE 2

Let Z = Fraction Defective at Target X_0 .

(Z lies somewhere between 0 and F_0 .)

By the Multinomial Theory of Probability the probability of having zero items in Region I, exactly 1 item in Region II, and exactly N items in Region III is

$$\frac{(N + 1)!}{0! 1! N!} \begin{bmatrix} \\ Z \\ \\ F_0 \end{bmatrix}^0 \begin{bmatrix} \\ dZ \\ \\ F_0 \end{bmatrix}^1 \begin{bmatrix} \\ F_0 - Z \\ \\ F_0 \end{bmatrix}^N = g(Z)dZ, \text{ where}$$

$$g(Z) = \frac{N + 1}{F_0} \begin{bmatrix} \\ Z \\ \\ 1 - \frac{Z}{F_0} \end{bmatrix}^N \text{ is the PROBABILITY DENSITY FUNCTION of } Z.$$

Hence, the Cumulative Distribution Function of the fraction defective Z is

$$G(Z) = \int_0^Z \frac{N + 1}{F_0} \left[1 - \frac{Z}{F_0} \right]^N dZ = 1 - \left[1 - \frac{Z}{F_0} \right]^{N+1}$$

For confidence C, we must have $1 - (1 - Z/F_0)^{N+1} = C$.

Solving this for Z we obtain $Z = F_0 \left[1 - (1 - C)^{1/N+1} \right]$

Now, $F_0 = 1 - A$; $A = \text{Minimum (worst) Reliability}$

$Z = 1 - R$; $R = \text{Reliability of Population (to target)}$

So , $1 - \frac{Z}{F_0} = 1 - \frac{1 - R}{1 - A} = \frac{R - A}{1 - A}$

Therefore, $\frac{R - A}{1 - A} = (1 - C)^{1/N+1}$

From this $R = A + (1 - A) (1 - C)^{1/N+1}$

The last formula from previous page is the COMPRESSED SUCCESS RUN THEOREM, which tells us what the RELIABILITY is with CONFIDENCE C when we have obtained N consecutive successes to a LIFE TARGET. So, strictly speaking, we should put a subscript C on the RELIABILITY R, and write the formula as follows:

$$R_c = A + (1 - A) (1 - C)^{1/N+1}$$

NOTE: In case $A = 0$ this becomes the

$$\text{CLASSICAL FORMULA} \quad R_c = (1 - C)^{1/N+1}$$

COMPARISON OF RESULTS IN VARIOUS SITUATIONS

From the formula $R_c = A + (1 - A) (1 - C)^{1/N + 1}$, where

A = Worst possible reliability to test target

N = Success Run

C = Confidence

R_c = Reliability with Confidence C , we obtain the following table of required **SUCCESS RUN SAMPLE SIZES** for various values of the Worst Reliability A (i.e., A = 0 , A = .5 , and A = .8)

Let us take C = .90 (90 % Confidence)

DESIRED R. 90	N (FOR A = 0)	N (FOR A = .50)	N (FOR A = .80)
.95	44	21	8
.99	229	114	44
.999	2301	1150	459
.9999	23,024	11,511	4604

NOTE: Solving $R_c = A + (1 - A) (1 - C)^{1/N+1}$ for the **SUCCESS RUN SAMPLE SIZE N** , we obtain

$$N = -1 + \frac{\ln (1 - C)}{\ln \left[\frac{R_c - A}{1 - A} \right]}$$

CONCLUSION

The success run sample size N for a Minimum Reliability of A is approximated by taking the CLASSICAL SUCCESS RUN SAMPLE SIZE for ZERO Minimum Reliability and multiplying it by the factor $(1 - A)$. Thus , in the table on the previous page we see that for $A = .5$ the sample size is about HALF of that for $A = 0$, and that for $A = .8$ the sample size is only about $1/5$ (20%) of that for $A = 0$.

This illustrates the importance of knowing what the worst possible reliability is in a given situation , if we want to avoid excessive testing.