Electronic Component Failure Rate Prediction Analysis

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Introduction:

Many electronic component manufacturers predict failure rates (λ) of their devices by use of an in-house testing process. For example the test may involve placing a certain number of components in an oven, and allowing them to "bake" at a predetermined temperature for a predetermined amount of time. At the end of the test, the number of failed devices are counted to determine the values of r and "DeviceHours", where r = number of failures, and DeviceHours = number of surviving components times the number of hours they remained in the oven. The

quotient $\frac{r}{\text{DeviceHours}}$ is considered an estimate of λ as opposed to the true λ . One reason being

is that the group of devices under test is just a small sample of a much larger population. Since the manufacturer may have difficulty accumulating enough data to calculate true λ , he does the next best thing. He employs a Chi-Square statistical tool that relates the controlled test sample to the entire population. This tool allows him to calculate a "confidence interval" which provides a "statistical" upper bound for the true λ . The equation manufacturers are using is

 $\lambda = \frac{\chi^2(1-\text{CL}, 2r+2)}{2 \text{ x DeviceHours}} \quad [\text{eq. 1}] \text{ where } \chi^2(1-\text{CL}, 2r+2) \text{ is a statistical factor taken from the Chi-Square Table, with } 2r+2 = \text{degrees of freedom, and } \text{CL} = \text{confidence level.}$

Objective:

This paper tries to explain the why and how Equation 1 works, and also where it comes from.

Background Theory:

Calculations for confidence level are based on the binomial distribution function described in many statistics textbooks. The binomial distribution function is written as

$$P_n(k) = \binom{n}{k} p^{n-k} q^k \quad \text{where} \binom{n}{k} \text{ is defined as } \frac{n!}{k!(n-k)!} \quad [\text{eq. 2}] \text{ where } p+q=1$$

When we are interested in the probability that r or fewer events occur in n trials (or, conversely, that greater than r events occur), then the cumulative binomial distribution function (CDF) of Equation 2 is used as shown in Equations 3 or 4:

$$P(\varepsilon \le r) = \sum_{k=0}^{r} P_n(k) = \sum_{k=0}^{r} \left(\frac{n!}{k!(n-k)!} \right) p^{n-k} q^k \quad [eq. 3] \quad \text{or}$$

$$P(\varepsilon > r) = 1 - P(\varepsilon \le r) = \sum_{k=r+1}^{n} \left(\frac{n!}{k!(n-k)!}\right) p^{n-k} q^{k} \quad [eq. 4]$$

In terms of the cumulative binomial distribution function, the confidence level is defined as

$$CL = P(\varepsilon > r) = 1 - \sum_{k=0}^{r} \left(\frac{n!}{k!(n-k)!}\right) (1-q)^{n-k} q^{k} \quad [eq. 5] \quad \text{or}$$

$$1 - CL = \sum_{k=0}^{r} \left(\frac{n!}{k!(n-k)!}\right) (1-q)^{n-k} q^{k} \quad [eq. 6]$$

where CL is the confidence level in percent.

Now if n is very large, and q is very small, then the Poisson expression $\frac{(nq)^k}{k!}e^{-nq}$ provides a conservative estimate of the binomial distribution term for term as shown by Equation 7.

$$\left(\frac{n!}{k!(n-k)!}\right)(1-q)^{n-k}q^k \approx \frac{(nq)^k}{k!}e^{-nq} \quad [eq. 7] \qquad (\text{See attached for a proof.})$$

Joining Equations 6 and 7 together we get

$$1 - CL = \sum_{k=0}^{r} \frac{(nq)^{k}}{k!} e^{-nq} = e^{-nq} \left[1 + nq + \dots + \frac{(nq)^{r-1}}{(r-1)!} + \frac{(nq)^{r}}{r!} \right] \quad [eq. 8]$$

where nq is the expected or mean value.

In terms of Reliability, if λ = the constant failure rate of a component, and t = time in operation, then n can be replaced by t, and q replaced by λ , and λ t will be the expected (mean) number of failures. This implies that P(r) = $\frac{e^{-\lambda t} \cdot (\lambda t)^r}{r!}$ = probability of exactly r failures in time interval 0 to t. Note that when writing each term out

$$P(0) = e^{-\lambda t}$$
, $P(1) = e^{-\lambda t} \cdot \lambda t$, $P(2) = e^{-\lambda t} \frac{(\lambda t)^2}{2!}$, $P(3) = e^{-\lambda t} \frac{(\lambda t)^3}{3!}$, etc.

it becomes apparent that $P(r) = e^{-\lambda t} \left[1 + \lambda t + \dots + \frac{(\lambda t)^{r-1}}{(r-1)!} + \frac{(\lambda t)^r}{r!} \right]$. [eq. 9] In words, Equation 9

reads P(r) = probability (0 failures) or probability (1 failure) or ... or probability (r failures). Another way of saying this same thing is the probability of r or less failures in a time interval t. Replacing λt for nq in Equation 8 yields

$$1 - CL = \sum_{k=0}^{r} \frac{(\lambda t)^{k}}{k!} e^{-\lambda t} = e^{-\lambda t} \left[1 + \lambda t + \dots + \frac{(\lambda t)^{r-1}}{(r-1)!} + \frac{(\lambda t)^{r}}{r!} \right] \text{ [eq. 10]}$$

It turns out that for a given CL, the term $\chi^2(1-CL,2r+2)/2$ is the exact λt solution to Equation 10. So for example choosing CL = 60% or 1 - CL = 0.4, the following table shows exactly what the entries of the Chi-Square Table represent.

Entries from Chi-Square Table (Probability, Degrees of Freedom)		
Failures (r)	$(1-CL, 2r+2)/2 = \lambda t$	exact solution to
0	(0.4, 2) / 2 = 0.916290731	$0.4 = e^{-\lambda t}$
1	(0.4, 4) / 2 = 2.022313245	$0.4 = e^{-\lambda t} \cdot [1 + \lambda t]$
2	(0.4, 6) / 2 = 3.105378597	$0.4 = e^{-\lambda t} \cdot [1 + \lambda t + (\lambda t)^2 / 2!]$
3	(0.4, 8) / 2 = 4.175262733	$0.4 = e^{-\lambda t} \cdot [1 + \lambda t + (\lambda t)^2 / 2! + (\lambda t)^3 / 3!]$

Chi-Square Table:

In other words $\frac{\chi^2(1-CL, 2r+2)}{2} = \lambda t$ in column 2 yields the exact solutions to the Poisson distributions in column 3 of the above table. In other words the Chi Square Table solves for λt .

(Note that the Chi Square Table does not solve for λ or t alone, but λ t.) Now simply divide both

sides of the above equation by t (device hours) to get $\lambda = \frac{\chi^2(1 - CL, 2r + 2)}{2t}$ which is exactly

Equation 1. //

Note:

The above method is reserved for devices that exhibit constant failure rates i.e. electronic devices. For devices that exhibit non-constant failure rates i.e. mechanical devices, other statistical methods should be employed which is a subject for another paper.

Notes on Chi-Square Table

A user friendly Chi-Square Interactive Table can be accessed on line at http://math.uc.edu/~brycw/classes/148/tables.htm

Both Microsoft Excel and MathCad have built in Chi-Square Table generators. Any entry of the Chi-Square Table $\chi^2(1-CL,2r+2)$ can be determined using either one of those two programs. For **Excel** enter Chiinv(1–CL, 2r+2). For **MathCad** enter qchisq(CL, 2r + 2)

More Interesting facts on the Poisson Distribution

The collection of terms e^{-x} , $e^{-x}x$, $e^{-x}\frac{x^2}{2!}$, $e^{-x}\frac{x^3}{3!}$, \cdots is called the Poisson Distribution.

Note that the sum of all the terms is equal to one.

$$\sum_{k=0}^{\infty} e^{-x} \frac{x^{k}}{k!} = e^{-x} \left[1 + x + \frac{x^{2}}{2!} + \frac{x^{3}}{3!} + \cdots \right] = e^{-x} \cdot e^{x} = 1$$

Therefore this Poisson Distribution is a discrete probability density function or PDF.

The cumulative distribution function CDF is
$$\sum_{k=0}^{r} e^{-x} \frac{x^{k}}{k!} = e^{-x} \left[1 + x + \dots + \frac{x^{r-1}}{(r-1)!} + \frac{x^{r}}{r!} \right]$$

For r = 0, 1, 2, 3, etc. is also discrete.

If n is large and q is small the Poisson Approximates the Binomial

Theorem (Equation 6):

If n is large and q is small, then $\frac{n!}{k!(n-k)!}p^{n-k}q^k \approx \frac{(nq)^k}{k!}e^{-nq}$ Proof : $\frac{n!}{k!(n-k)!}p^{n-k}q^k = \frac{n(n-1)(n-2)\cdots(n-k+1)}{k!}(1-q)^{n-k}q^k$ since p=1-q $\approx \frac{n^k}{k!}(1-q)^{n-k}q^k$ since n is large $= \frac{n^k}{k!} \cdot \frac{(1-q)^n}{(1-q)^k} \cdot q^k \approx \frac{n^k}{k!} \cdot \frac{(1-q)^n}{1} \cdot q^k = \frac{(nq)^k(1-q)^n}{k!}$ since q is small (1) now compare $(1-q)^n$ with e^{-nq} by expanding each of them. $(1-q)^n = 1-nq + \frac{n(n-1)}{2!}q^2 - \frac{n(n-1)(n-2)}{3!}q^3 + \cdots$ $\approx 1-nq + \frac{n^2}{2!}q^2 - \frac{n^3}{3!}q^3 + \cdots = 1-nq + \frac{(nq)^2}{2!} - \frac{(nq)^3}{3!} + \cdots$ since n is large $\therefore (1-q)^n \approx 1-nq + \frac{(nq)^2}{2!} - \frac{(nq)^3}{3!} + \cdots$ (2) $e^{-nq} = 1-nq + \frac{(nq)^2}{2!} - \frac{(nq)^3}{3!} + \cdots$ (3) comparing (2) and (3) $\Rightarrow (1-q)^n \approx e^{-nq}$ (4) Replacing e^{-nq} for $(1-q)^n$ in (1) we get $\frac{n!}{k!(n-k)!}p^{n-k}q^k \approx \frac{(nq)^k}{k!}e^{-nq} //$

Proving Equation 6 using L'Hospital's Rule. Theorem 1:

 $\lim_{x\to 0} 1 - e^{-x} = x$ or $\lim_{x\to 0} 1 - x = e^{-x}$

Proof

By using L'Hospital's Rule we see the relationship between x and $1 - e^{-x}$ as x gets very small.

$$\lim_{x \to 0} \frac{f(x)}{g(x)} = \frac{1 - e^{-x}}{x} = \frac{0}{0} \text{ and } \lim_{x \to 0} \frac{f'(x)}{g'(x)} = \frac{e^{-x}}{1} = 1 \implies \lim_{x \to 0} \frac{1 - e^{-x}}{x} = 1 \implies \lim_{x \to 0} \frac{1 - e^{-x}}{x} = 1 \implies \lim_{x \to 0} 1 - e^{-x} = x \text{ or } \lim_{x \to 0} 1 - x = e^{-x}$$

Theorem 2:

If n is very large, and q is very small, then

$$P_{n}(k) = \left(\frac{n!}{k!(n-k)!}\right)(1-q)^{n-k}q^{k} \xrightarrow[n \to \infty]{} \frac{(nq)^{k}}{k!}e^{-nq}$$
Proof
$$P_{n}(0) = (1-q)^{n}$$

$$P_{n}(1) = n(1-q)^{n-1}q$$

$$P_{n}(2) = \frac{n(n-1)(1-q)^{n-2}q^{2}}{2!}$$

$$P_{n}(3) = \frac{n(n-1)(n-2)(1-q)^{n-3}q^{3}}{3!}, \text{ etc. by definition.}$$

Since n is very large we can replace n-k with n and get

$$P_n(0) = (1-q)^n$$

$$P_n(1) = n(1-q)^n q$$

$$P_n(2) \simeq \frac{n^2(1-q)^n q^2}{n^2}$$

$$P_n(2) = 2!$$

 $P_n(3) \cong \frac{n^3(1-q)^n q^3}{3!}$, etc.

Since q is very small we can replace (1–q) with e^{-q} by L'Hospital's Rule. Therefore $P_n(0) \cong (e^{-q})^n = e^{-nq}$

$$P_{n}(1) \cong n(e^{-q})^{n}q = nqe^{-nq}$$

$$P_{n}(2) \cong \frac{n^{2}(e^{-q})^{n}q^{2}}{2!} = \frac{(nq)^{2}e^{-nq}}{2!}$$

$$P_{n}(3) \cong \frac{n^{3}(e^{-q})^{n}q^{3}}{3!} = \frac{(nq)^{3}e^{-nq}}{3!}, \text{ etc. } //$$