

RELIABILITY OF SOFTWARE RELIABILITY GROWTH MODELS: A REVIEW

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Abstract- Today we live in the world which is revolutionised by technology. Technology has become part of our everyday life: from accomplishing a simple task like turning on a light bulb to launching satellites, technology has become an inseparable part of everyone's life. In order to be able to use this technology reliably, the software which we interact with technology needs to be reliable. Software reliability refers to probability that a software execution will be as expected under specified terms and conditions in a given amount of time without any fail. Software reliability is measured using software reliability growth models. Over time hundreds of models have been developed so far. This paper aims to review and examine several different non-homogenous Poisson process software reliability growth models.

Keywords— *Software reliability, non-homogenous Poisson process, Software Reliability Growth Models.*

I. Introduction

Software in today's technological world is an integral part of our day to day life. We interact with technology at our homes, in school, in our offices even in our cars; from searching on Alexa to teaching students on smart boards and finding routes on GPS or doing complex financial transactions we rely on technology. We interact with technology by means of software's. In order to be able to use this software for carrying out our chores efficiently & reliably, the software needs to possess certain characteristics or qualities [1] which include

- Functionality
- Usability
- Efficiency
- Maintainability
- Probability
- Reliability

Out of all the characteristics mentioned above Reliability is widely considered as key quality parameter. Software Reliability is defined as the "Degree to which a system, product or component performs specified functions under specified conditions for a specified period of time. This characteristic is composed of the following sub-characteristics: Maturity, Availability, Fault tolerance and Recoverability"[2] as per ISO standards. Software Reliability being an intangible in nature is a bit tedious to measure. Since 1960s several scientists have established Models called the software reliability growth models (SRGM) that calculate reliability of the software in terms of time for which software is expected to run reliably without any flaw. SRGMs also help software practitioner's in determining the right time to stop testing as to when the reliability is achieved and hence releasing the software[3].

The main objective of this paper is to go through a systematic approach for readers to gain knowledge about the software reliability, NHPP SRGMs and also put forward a critical review of what a particular model lacks and how consequent modeling technique overcame that/those drawback(s). This review paper also aims to present to readers, young scholars, researchers an insight into the growth, development of SRGMs in general. Thus it will also aim to persuade researcher to study existing model from a critical point of view motivating him/her to develop better models for Software Reliability.

II. Classification of SRGMs

Based on the parameters taken into consideration for measurement of reliability, this paper discusses SRGMs belonging to 3 categories as listed below:

1. PERFECT DEBUGGING MODELS

Models under this category assume that a failure when encountered is removed in entirety in no time in addition to assumptions listed below:

- The failures are governed by non-homogenous Poisson process (NHPP).
- A failure once detected is removed immediately.
- Removal of failure is completely perfect.
- All faults in software are mutually independent.

This category is further divided into two sub categories based on the fact whether time is calculated in terms of execution time of software or calendar time.

Perfect debugging models are described by a general formula

$$y(t) = a(1-X)$$

Where

$y(t)$ is no. of expected failures at time t

a is the estimated total no. of failures at an infinite time i.e

$$y(\infty) = a$$

X is parameter defined differently by each models discussed below and summarized in Table 1

Execution time model

J D Musa [4] devised the basic execution model in which time t is expressed in execution units. The mean value function for failure occurrence for software under this model is given by formula

Where

$$y(t) = a(1-X)$$

$$X = e^{-\frac{\lambda_0 t}{a}}$$

λ_0 is initial failure intensity .

Calendar time model

Goel Okumoto (GO Model) [5] also known as exponential model, is defined by Non-

homogenous Poisson Process (NHPP) in which failures are given by

$$y(t) = a(1 - X)$$

$$X = e^{-bt}$$

$$y(t) = a(1 - e^{-bt}) \tag{1.1}$$

where b is the fault removal rate, t is time expressed in calendar units.

Delayed S-Shaped Model

Then came the model given by Yamada & Osaki called the delayed S-shaped Model [6] which described testing as a 2- step process

Step I: fault detection Step II: fault removal

For step I $y(t)$ is same as that for GO i.e

$$y(t) = a(1 - e^{-bt})$$

While for step II X is given as

$$X = (1 + bt)e^{-bt}$$

Table 1

	Value for X	Formula
Basic time execution model [4]	$\lambda_0 t$ $-()$ e^{-a}	$y(t) = a(1 - e^{-a})$
GO model [5]	e^{-bt}	$y(t) = a(1 - e^{-bt})$
Delayed S-shaped model Step I [6]	e^{-bt}	$y(t) = a(1 - e^{-bt})$
Delayed S-shaped model Step II [6]	$(1 + bt)e^{-bt}$	$y(t) = a(1 - (1 + bt)e^{-bt})$

They [6] further enhanced their model with better estimation techniques [7] Following GO, Ohba [8] proposed hyper exponential model that assumed software to contain independent clusters of independent units/modules. Each unit having a different initial number of errors and a different failure rates, such as new and existing/reused units, simple and composite units, and interactive units and units which do not interact.

Further Yamada & Osaki [9] assumed software to contain two types of faults; ones easy to detect and correct and another difficult to detect which resulted in a model called Fault Categorization Model. On the similar lines Erlang Model [10] assumed software to contain 3 types of faults ; simple, hard & complex. Kapur & Garg [11] Another remarkable Model was

added by Bittanti et al.[12] that replaced constant error rate b by initial and final Fault detection rates b_i & b_f . With mean value for occurrence of failures given by

$$y(t) = a \left(\frac{b_i (e^{b_f t} - 1)}{b_f + b_i (e^{b_f t} - 1)} \right)$$

Recently Hanagal and Bhalerao[13],[14],[15]obtained SRGMs based on general inverse exponential distribution & extended inverse exponential distributions.

All of the models discussed above lack onething i.e the presumption that the errors as and when found are debugged immediately which is unrealistic.

This is the critique point for most of the conventional SRGMs including the ones discussed above.The models discussed in the next segment try to overcome this by taking into account the imperfection debugging as a parameter for SRGMs.

2. IMPERFECT DEBUGGING MODELS

As the name implies this section discusses models where debugging process is not perfect as assumed in models we discussed above.Models that work onthe principle

- a) that a debugging process can induce new faults into the software and /or
- b) detected faults are not removed completely

are known as Imperfect debugging models. The models described here under this category are basedon the general assumptions of NHPP SRGMs listedin the first section except that the debugging process is not at all perfect which is quite realistic. In addition to listed assumption, each model discussedbelow have some additional conditions.

Pure error generation Model

Ohba & Chou [16] formulated a model under imperfect debugging that induced new faults during the debugging process. The inductions of new faults were assumed to occur at a constant rate ' α '. The overall mean value fault function for model is given by

$$y(t) = a(1 - \alpha)(1 - e^{-b(1-\alpha)t})$$

Pure Imperfect Debugging Model

Kapur & Garg[17] incorporated imperfect debugging by assuming that there exists a probability p that an error is perfectly detected & removed and hence the mean value function formodel is given separately for fault detection and fault removal and are defined by

$$y_d(t) = a(1 - e^{-bpt})$$

$$y_r(t) = (a/p)(1 - e^{-bpt})$$

Where $y_d(t)$ is the function for error detection

& $y_r(t)$ is the function for error removal at time t . This model was further improvised by Yamada etal.[18] by assuming fault introduction as exponentialin pure error generation model.

Testing Efficiency Model

Zhang et al.[19] developed a model that incorporated features from both the models discussed above i.e a)introduction of new faults from Ohba & Chou[16] & b)probabilistic removal of

$$y(t) = \frac{a}{p-\alpha} \left(1 - \left(\frac{(1+\beta)e^{-bt}}{1+\beta e^{-bt}} \right)^{(c/b)(p-\alpha)} \right)$$

faults from Kapur & Garg[11]. The mean value fault function for model is given by

Another remarkable Model called PNZ[20] Model incorporated fault introduction probability into imperfect debugging. Kumar et al.[21] incorporated imperfect debugging into Yamada[6] Model called the delayed S-shaped model with pure error generation. This way the researchers introduced imperfect debugging into SRGMs. In reality error detection and correction is a tedious and time consuming process. It takes expertise and considerable amount of time to fix an error which to some extent is accounted in the next section.

3. TESTING EFFORT BASED SRGMs

Testing is an integral part of Software Development Life Cycle and heart & soul of SRGMs. The idea behind development of testing effort based SRGMs was to incorporate testing resources (manpower, hardware and software) into regular SRGMs. Putnam[22] proposed use of Rayleigh model to describe the test-effort expenses time-dependent behavioral which is given by cumulative distribution of test efforts as

$$z(t) = (1 - e^{-at^2}) \tag{2.1}$$

Yamada et al. [23] presumed the fault detection rate is dependent to the amount of test effort used during the testing phase and is proportionate to the present failure content. Thus the amount of test effort used throughout the software testing phase was incorporated into a software-reliability growth model. The model is devised using a non-homogeneous Poisson process. Yamada[23] incorporated $z(t)$ from “(2.1)” in GO model yielding

$$y(t) = a(1 - e^{-bz(t)})$$

Where b is the fault detection/removal rate per remaining error.

The technique for data analysis for measuring software dependability is built using the model. This model is used to anticipate the cost of further test efforts needed to meet the objective number of software testing faults found, as well as the best time to end software testing before a release. Yamada [24] soon developed model calculating test effort using Weibull distribution and incorporating it in SRGM as above. On the similar lines, Bokhari & Ahmad[25] used formulated Log Logistic Test Effort model followed by Kapur et al.[26] and Khatri[27], [28]. They didn't stop here Khatri[29] began studying Ohba & Chou Imperfect Debugging Model. They came up with [30] Model beautifully quantifying Software Reliability under imperfect debugging and incorporating testing effort ' $z(t)$ ' in Ohba & Chou[16] as

$$y(t) = \frac{a}{1-\alpha} (1 - e^{-pb(1-\alpha)z(t)})$$

The next step in SRGM development was to incorporate Testing effort in imperfect debugging Models and initiated by Kapur et al.[31] This was followed by incorporating test effort and imperfect debugging in Yamada’s Delayed S-shaped SRGM [6], [32] Later Khatri[33] Kapur et al.[34] proposed a generalized framework utilizing Test effort for modeling multi release of a software introducing the effect of fault reduction factor.

The question still remains that even with better test efforts; testing process is never going to be simple, it is still going to be complex & time consuming. In SRGMs discussed above certain factors are assumed to remain constant over time. However it may not be the case , which is why the next category of Models came into existence.

4. **CHANGE POINT MODELS**

In the applications of SRGM it is assumed that the debugging /testing environment stays same. In reality this environment changes over time e.g test teams/tools/resources can change affecting the error detection rate. Change Point Models capture such transitions and incorporate them into SRGMs.

Exponential Change Point Model

This model was given by Chang[35] suggesting single change point ‘ ρ ’ in basic GO [5] such that the rate of failure detection b before and after single change point is b_1 and b_2 respectively.

Using in 1.1, we have,

$$y(t) = \begin{cases} a(1 - e^{-b_1 t}), & 0 \leq t \leq \rho, \\ a(1 - e^{-(b_1 \rho + b_2(t - \rho))}), & \rho < t \end{cases}$$

S-shaped Change Point model

Inoue & Yamada [36] devised a model by incorporating single change point in delayed S-shaped Model[6] ‘ ρ ’. The mean value fault function for model is given by

$$Y(t) = \begin{cases} a(1 - \frac{(1 + b_1 t)e^{-b_1 t}}{(1 + b_1 \rho)}), & 0 \leq t \leq \rho, \\ a(1 - \frac{(1 + b_1 \rho)(1 + b_2 t)e^{-b_1 \rho - b_2(t - \rho)}}{(1 + b_2 \rho)}), & \rho < t \end{cases}$$

This was followed by Kapur[37] who devised model with single change points where failures followed different probability distributions before and after change point. In the similar manner Exponential Change point Model [38] under imperfect debugging with different fault induction rates before and after Change point followed by [39]. Recent advances [40] uses test effort with fault reduction factor, under imperfect debugging in change point. A pioneering breakthrough was led by Huang[41] introducing Multiple Change Points. Later Kapur & Khatri[42],[43],[44],[45][46] introduced multiple change point by means of categorisation of faults as easy and difficult faults.

III. **Conclusion:**

Some of major NHPP models that have appeared in literature are discussed in this paper. Reliability models are a powerful tool for estimating, administering and examining software reliability. They are especially useful to describe reliability growth and fault deterioration,

making it simple to analyse software reliability & predict software release. In this paper, we first studied the basic execution model followed by GO model to provide a quantification for software reliability. In order to capture both execution and calendar time, we looked at a few widely used variations of NHPP models based on time-dependent transition probabilities. Then, we studied the modelling approach utilising test efforts. We further discussed model incorporating much more realistic imperfect debugging in which the earlier models can be tailored using probabilities of fault correction. Finally, we discussed change point models where in, the change(s) in software environment is included in estimation of failures. By increasing the testing effort intensity and properly allocating and managing the testing resources and by adding more realistic parameters to these SRGMs can aid in the removal of flaws, assisting software practitioners in determining when a software system is prepared for release and if its reliability has reached a predetermined threshold

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