

LAMBDA

NOTES

News for the Society of Reliability Engineers

Volume 41, Issue 1, August 2008



SRE President's Message

Presidents Corner

Greetings!

Well, we had another great RAMS in Las Vegas! We had a number of SRE functions and activities at the Symposium including a luncheon that was quite well attended by our membership. Other SRE activities included recognition of the Stan Ofsthun Best Paper award winners at the RAMS Banquet and the presentation of the award itself at our SRE Annual Board meeting. We also awarded two iPod Shuffles at our luncheon and an 80 GB iPod Classic to Robert Bergman of our At-Large Chapter. You had to be present at the luncheon to have a chance at one of the Shuffles, but all SRE members in good standing were entered in the iPod Classic drawing.

The next RAMS is scheduled for Ft. Worth, TX from 26-29 January, 2009. If you are familiar with RAMS, you probably recognize that it tends to alternate between the East and West coasts, so this is different. I have been to the Renaissance Worthington hotel and it is really first-class, and Ft. Worth has a lot to offer in addition to the Symposium itself. As always, you can get the big picture at www.rams.org.

Please take a few minutes and read about the goings-on of your Society. We have put in place initiatives for a chapter coordinator, student support to attend RAMS and the establishment of a Fellow class of membership to supplement our existing activities like the Stan Ofsthun Award, plus continuing developmental items such as our website project.

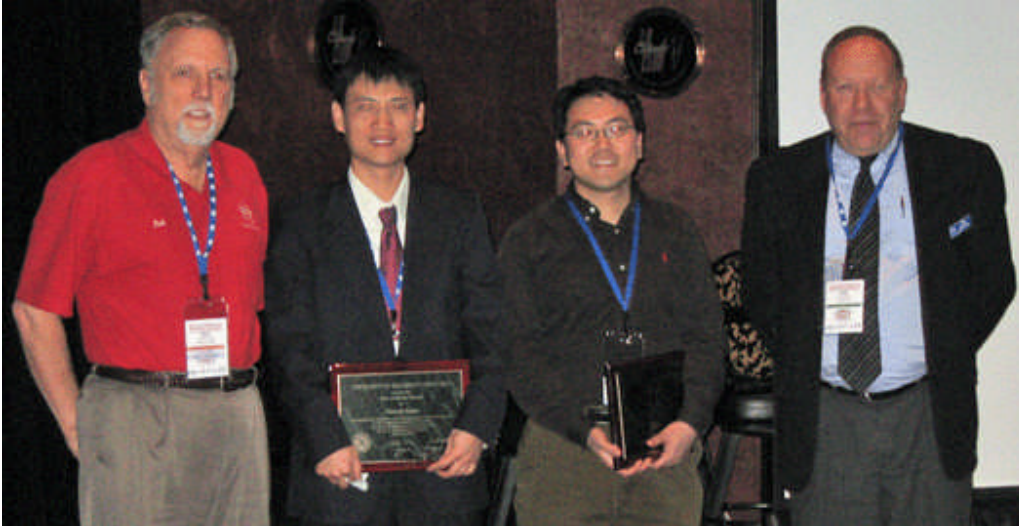
Respectfully,

Bob Loomis

SRE President

Stan Ofsthun Award - RAMS 2008

Harry Guo and Rong Pan received the SRE Stan Ofsthun Award for best paper at the 2008 RAMS for their paper entitled "On Determining Sample Size and Testing Duration of Repairable Systems Test" .



Presentation of the 2008 Stan Ofsthun Award at the 2008 RAMS. Left to right, Bob Loomis, SRE President, Harry Guo, Rong Pan and Joel Nachlas, Award Facilitator

The Stan Ofsthun Award is provided for the best paper at RAMS by an SRE member. Candidate papers receive a blind review in the areas of originality, significance, accuracy, applicability, and clarity. Joel Nachlas is the coordinator of the award. If you are a SRE member and are authoring a RAMS paper, be sure and list your affiliation with SRE in the biographical part of the paper and apply for the competition through your chapter chair. The winning paper for 2008 is the Technical Supplement in this edition of Lambda Notes (see page 6).

SRE Chapters Corner

KSC-Orlando Chapter

The past year has been a good time for the KSC/Orlando Chapter. We continued our partnership with the local Cape Canaveral ASQ Section sponsoring joint meetings throughout the year. The seven joint meetings held during the year included speakers from NASA at the Kennedy Space Center as well as companies such as Nokia, Spherion Staffing, and Brietideas, Inc . The meetings included excellent presentations on Risk Decision Making, Product Liability, Independent Assessment, Project Management, Workplace Happiness, and the future of Human Spaceflight. The Chapter also continued its recruiting drive for new members and was able to increase the membership 25, a new high for the Chapter.

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Your Chapter could be here!

SRE Chapter Chair

Let me begin with an introduction of the chapter chair team. Vice chairs are Colie Warren and Macy Messalti, in support of the Chapter Chair mission. For those new members, our mission is to charter new chapters, support existing chapters, and offer growth support wherever needed.

We are pleased to announce that over the last several months we have established new chapters in the Twin Cities, Los Angeles and the San Francisco Bay Area. We are in the process of developing chapters in Houston, Toronto and Sydney Australia. These chapter formations would not have been possible without dedicated and competent member leaders at the chapter level. We salute those member leaders for their leadership of, and contribution to, the society.

It is also worth noting that the Buffalo New York chapter is under reconstruction. As you may already know, the Buffalo chapter was the founding chapter of the society, and is in the process of rebuilding their membership. We have already established a high degree of collaboration with the many other professional societies in the upstate New York area, and are looking forward to a fruitful program year.

We would also encourage chapters to send their program announcements to other professional societies in their respective geographical areas, as well as to solicit speakers for SRE chapter meetings. As Reliability practitioners, we seek to share the knowledge and usefulness of our discipline as well as to learn new tools and skills from other disciplines. So, in closing, feel free to contact any member of the SRE leadership team if you need program ideas, contacts, or would like to promote the charter of a new chapter.

Respectfully,
David Auda
SRE Chapter Chair

Other New SRE Initiatives

SRE Approves Fellow Designation

The establishment of a designation of SRE Fellow was approved at the SRE Board Meeting at the RAM Symposium in January 2008. The SRE Fellows Program is oriented towards practitioners doing reliability-related work. It is an honorary designation designed to recognize both service and contributions to reliability and to SRE. Requirements to become an SRE Fellow are posted on the SRE website and include practical reliability experience, service to the reliability community and supporting SRE as a member.

SRE RAMS[®] SCHOLARSHIP

In order to encourage awareness of and the pursuit of careers in Reliability Engineering or closely related fields, the SRE has established the SRE RAMS[®] Scholarship to sponsor student attendance at the annual Reliability and Maintainability Symposium (www.rams.org).

The SRE RAMS[®] Scholarship was approved at the annual meeting in January 2008 and will be awarded annually to one college student who meets the criteria and wins the competition. The scholarship is actually a grant for up to \$1200 to cover the student's expenses at the annual RAMS[®] held the end of each January. Eligible expenses include transportation costs, conference registration fees, lodging during the conference, and other incidental expenses. Also included in the scholarship is a paid membership in the SRE for two years.

To be considered students must complete an application process that includes an evaluation letter from at least one instructor who taught the student a reliability related course, unofficial transcripts of completed courses, and a short essay describing academic accomplishments, professional activities,

work history, career goals and the relevance of reliability engineering to each of them.

All completed applications received by the 15 November deadline will be judged by a panel of SRE board members and one winner will be selected. The winner will be contacted by 15 December and the necessary arrangements coordinated for attending the annual RAMS® the following January. The winning student will be expected to attend the annual SRE meeting held in conjunction with the annual RAMS® and submit a report to SRE by 15 February detailing participation at the symposium. Additional information can be found on the application and on the SRE web site.

New SRE Banners Have Arrived

SRE has new banners for chapter use. If a chapter is interested in acquiring a new SRE banner (or needs SRE lapel pins) please contact the SRE webmaster.

The Last Word

We want to publish Lambda Notes quarterly. To make this newsletter more useful to the members of SRE we would like to include information about each of the chapters of SRE. Please send us copies of your chapter newsletters or just information about your chapter. The more chapters contributing to this effort the more useful it will be to the entire society. This is the third or fourth "rebirth" of Lambda Notes since the mid 1960s. None of the earlier efforts lasted for more than three issues. To make this one different we must have input from the chapters, not just the officers and committee chairs. We know many of you are doing good things in your chapters, please share it with all of us.

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Technical Supplement

The SRE Stan Ofsthun Award for best paper at the 2008 RAMS.

"On Determining Sample Size and Testing Duration of Repairable Systems Test"

Written by: Harry Guo and Rong Pan

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On Determining Sample Size and Testing Duration of Repairable System Test

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Rong Pan, PhD, Arizona State University

Key Words: Repairable system; Maximum likelihood estimation; Reliability test plan; Fisher information matrix

SUMMARY & CONCLUSIONS

Reliability demonstration tests for non-repairable systems have been extensively discussed by many researchers. However, very few works have been done for repairable system tests. Demonstration tests for repairable systems can be time consuming and costly. Carefully planning sample size and test duration is very important. This paper develops a theoretical method, based on pivotal quantities and a confidence bound requirement for the reliability metrics of interest, to help test planners to determine the minimal sample sizes and test duration. A case study was given and the developed theoretical results were compared with simulation results. The comparison shows that the proposed method is accurate and efficient.

ACRONYMS

CI	Confidence Interval
NHPP	Non-homogeneous Poisson Process
MLE	Maximum Likelihood Estimation

1 INTRODUCTION

Life test design for non-repairable parts has been extensively discussed in literature. For example, Lawless [1] developed a procedure for determining the required sample size and number of failures based on the required confidence intervals of model parameters, providing a table for Weibull and Extreme Value distributions. McKane, Escobar and Meeker [2] generalized Lawless's method and developed charts for sample size determination for log-location-scale distributions. In order to reduce the test time and the cost, life tests are usually conducted at higher stress levels than the normal usage condition. Many accelerated test plans for non-repairable systems have also been discussed in the recent years (see, e.g., [3-5]). Recent research has made extensions of previous works, such as designing test plans for high usage rates [6] and for multiple objective functions [7].

However, very few papers have discussed test plans for repairable systems. Singh and Swaminathan [8] derived the exact confidence interval of system availability based on the assumption of exponential distributions for time-to-failure and time-to-repair. Usher and Taylor [9] showed the design of availability test plans that meet the stated levels of producer

and consumer risks. The existing papers focus on a single system only and have not discussed test plans for a repairable system with certain failure intensity function. Due to the high cost and lengthy testing time that a typical system test consumes, it is necessary to develop a method for sample size determination for repairable system tests. With a proper sample size, the test results can be used to predict the number of failures in a given time period with greater confidence. Resources, such as budget, crew and spare parts, also can be properly allocated in advance.

In this paper, a method for planning a life test for repairable systems is proposed. A repairable system is assumed to undergo "minimal repair" at the occurrence of each failure so that the rate of system failure is not disturbed by the repair. This assumption is used by many NHPP (non-homogeneous Poisson process) models, such as the log linear failure intensity model by Cox and Lewis [10], Crow-AMSAA model [11], and bounded failure intensity models by Pulcini [12] and Attardi and Pulcini [13]. With the proposed method, a planner can determine the sample size and/or the total testing time so that the failure intensity function or the expected number of failures can be estimated with the required precision. This type of test can be applied at the final system validation stage, where the required system reliability needs to be demonstrated. The test results can also assist in allocating resources to maintain the system. For example, an automobile manufacturer may need to collect vehicle failure data with minimal effort in order to optimize its warranty policy. Therefore, it is important to design a test that will yield sufficient data and accurate results in the most time- and cost-effective way.

2 METHODOLOGY

In this section, based on the asymptotic properties of parameter estimation, a method for determining required sample size and test duration for repairable system tests is proposed. According to the required accuracy of predicted number of failures or model parameters, the sample size for the test can be properly determined. For convenience, the Crow-AMSAA model, which is a power law NHPP model, will be used in this section. The proposed method also can be extended to other NHPP models, such as the log-linear model and the bounded failure intensity model. The analytical

solution for the power law model is provided and the results are compared with the simulation results provided in Section 3. The comparison shows that the proposed method is valid and will have a wide application for repairable system test planning.

2.1 Maximum Likelihood Estimation of a Power Law NHPP model

For a power law NHPP model the expected number of failures and failure intensity function for a single repairable system are derived from the following equation:

$$\lambda(t) = \theta \beta t^{\beta-1} \quad (1)$$

where θ is a scale parameter (sometimes called the intrinsic failure rate) and β is a shape parameter. The expected number of failures at time t of a Poisson process is the same as the cumulative failure intensity value, which is:

$$E[N(t)] = \Lambda(t) = \int_0^t \theta \beta u^{\beta-1} du = \theta t^\beta \quad (2)$$

For a single repairable system, the likelihood function of observing n failures at times t_1, t_2, \dots, t_n with the censoring time at T is:

$$L(\theta, \beta | t_1, t_2, \dots, t_n, T) = \prod_{i=1}^n (\lambda(t_i) e^{-\Lambda(t_i) - \Lambda(t_{i-1})}) \times e^{-[\Lambda(T) - \Lambda(t_n)]} \quad (3a)$$

$$= \prod_{i=1}^n \lambda(t_i) \times e^{-\Lambda(T)} = \theta^n \beta^n e^{-\theta T^\beta} \prod_{i=1}^n t_i^{\beta-1}$$

Taking the logarithm of the likelihood function, we get the log-likelihood function:

$$\ln L(\theta, \beta | t_1, t_2, \dots, t_n, T) = n \ln(\theta \beta) - \theta T^\beta + \sum_{i=1}^n (\beta - 1) \ln t_i \quad (3b)$$

Assume there are m identical repairable systems. These systems are tested until time T and the failure times of these systems are recorded as $\{t_{1,1}, t_{1,2}, \dots, t_{1,n_1}\}, \dots, \{t_{m,1}, t_{m,2}, \dots, t_{m,n_m}\}$, then the likelihood and log-likelihood functions for all the m systems are:

$$L(\theta, \beta | \{t_{j,1}, t_{j,2}, \dots, t_{j,n_j}\}_{j=1, \dots, m}, T) = (\theta \beta)^{\sum_{j=1}^m n_j} e^{-\theta T^\beta m} \prod_{j=1}^m \prod_{i=1}^{n_j} t_{j,i}^{\beta-1} \quad (4a)$$

and:

$$\begin{aligned} \ln L(\theta, \beta | \{t_{j,1}, t_{j,2}, \dots, t_{j,n_j}\}_{j=1, \dots, m}, T) \\ = (\sum_{j=1}^m n_j) \ln(\theta \beta) - \theta T^\beta m + (\beta - 1) \sum_{j=1}^m \sum_{i=1}^{n_j} \ln t_{j,i} \end{aligned} \quad (4b)$$

In order to obtain the MLEs of θ and β , we take the first derivative of the log-likelihood function with respect to θ and β and set them to zero. It can be shown that [11]:

$$\hat{\theta} = \frac{\sum_{j=1}^m n_j}{T^{\hat{\beta}} m} \quad (5)$$

and:

$$\hat{\beta} = \frac{\sum_{j=1}^m n_j}{\sum_{j=1}^m n_j \ln T - \sum_{j=1}^m \sum_{i=1}^{n_j} \ln t_{j,i}} \quad (6)$$

2.2 Confidence Intervals for θ , β and $N(t)$

Normal-approximation confidence regions are usually used for ML estimates of model parameters and functions of model parameters. In this section, we will build the confidence interval for θ , β and $N(t)$. Similar work was done by Lawless [1] for non-repairable systems with Weibull and Extreme Value distributions, Nelson and Schmee [3] for lognormal distribution and McKane, Escobar and Meeker [2] for general log-location-scale distributions. The Fisher information matrix is used to construct the variance-covariance matrix for model parameters. By taking the expected value of the negative of second derivatives of the log-likelihood function, we obtain the expected Fisher information matrix as:

$$\begin{aligned} I = -E \begin{bmatrix} \frac{\sum_{j=1}^m n_j}{\beta^2} - \theta m T^\beta \ln^2 T & -m T^\beta \ln T \\ -m T^\beta \ln T & \frac{\sum_{j=1}^m n_j}{\theta^2} \end{bmatrix} \\ = m T^\beta \begin{bmatrix} \frac{\theta}{\beta^2} + \theta \ln^2 T & \ln T \\ \ln T & \frac{1}{\theta} \end{bmatrix} \end{aligned} \quad (7)$$

The approximated variance-covariance matrix of the β and θ estimators is:

$$\begin{aligned} \Sigma = \begin{bmatrix} \text{var}(\hat{\beta}) & \text{cov}(\hat{\beta}, \hat{\theta}) \\ \text{cov}(\hat{\beta}, \hat{\theta}) & \text{var}(\hat{\theta}) \end{bmatrix} = I^{-1} \\ = \begin{pmatrix} \beta^2 \\ m T^\beta \end{pmatrix} \begin{bmatrix} \frac{1}{\theta} & -\ln T \\ -\ln T & \frac{\theta}{\beta^2} + \theta \ln^2 T \end{bmatrix} \end{aligned} \quad (8)$$

Therefore, the large sample approximations of the standard errors of β and θ estimators are:

$$\text{std}(\hat{\beta}) = \frac{\beta}{\sqrt{m T^\beta \theta}} \quad (9)$$

and:

$$\text{std}(\hat{\theta}) = \sqrt{\frac{\theta}{m T^\beta} (1 + \beta^2 \ln^2 T)} \quad (10)$$

Usually, the model parameter estimators, especially for the parameters that can only have positive values, are assumed lognormally distributed [1, 3, 14]. In the simulation part of Section 3, we will see that the estimates of β indeed can be fitted very well by a lognormal distribution, while the estimates of θ slightly deviate from a lognormal distribution. In this analysis, following the tradition, we will assume that both estimators of β and θ are lognormally distributed. The $(1 - \alpha)\%$ confidence interval of β is:

$$[\hat{\beta} e^{-z_{\alpha/2} \text{std}(\hat{\beta})/\hat{\beta}}, \hat{\beta} e^{z_{\alpha/2} \text{std}(\hat{\beta})/\hat{\beta}}] \quad (11)$$

The confidence interval of θ is:

$$[\hat{\theta} e^{-z_{\alpha/2} \text{std}(\hat{\theta})/\hat{\theta}}, \hat{\theta} e^{z_{\alpha/2} \text{std}(\hat{\theta})/\hat{\theta}}] \quad (12)$$

where $z_{\alpha/2}$ is the upper $\alpha/2$ percentile of a standard normal distribution.

Since the expected number of total failures is a function of β and θ , we may derive the variance of the expected number of failures at a given time t for a single repairable system. The estimated number of failures expected for a single

reparable system is $\hat{\Lambda}(t) = \hat{E}[N(t)] = \hat{\theta}t^\beta$ and its variance is:

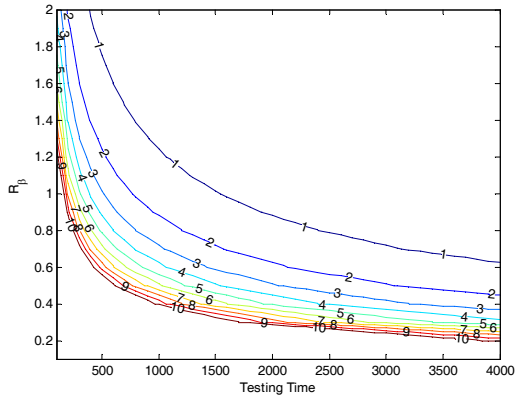
$$\begin{aligned} \text{Var}(\hat{\Lambda}(t)) &\approx (t^\beta)^2 \text{Var}(\hat{\theta}) + (\theta t^\beta \ln(t))^2 \text{Var}(\hat{\beta}) \\ &\quad + 2(t^\beta)(\theta t^\beta \ln(t)) \text{Cov}(\hat{\theta}, \hat{\beta}) \\ &= t^{2\beta} \frac{\theta(1 + \beta^2 \ln^2 T)}{mT^\beta} + \theta^2 t^{2\beta} \ln^2(t) \frac{\beta^2}{mT^\beta \theta} \\ &\quad - 2t^{2\beta} \theta \ln(t) \frac{\beta^2 \ln T}{mT^\beta} \\ &= \frac{t^{2\beta} \theta}{mT^\beta} [1 + \beta^2 (\ln T - \ln t)^2] \end{aligned} \quad (13)$$

The corresponding confidence interval of $\Lambda(t)$ can be easily obtained as $[\hat{\Lambda}(t)e^{-z_{\alpha/2} \text{std}(\hat{\Lambda}(t))/\hat{\Lambda}(t)}, \hat{\Lambda}(t)e^{z_{\alpha/2} \text{std}(\hat{\Lambda}(t))/\hat{\Lambda}(t)}]$.

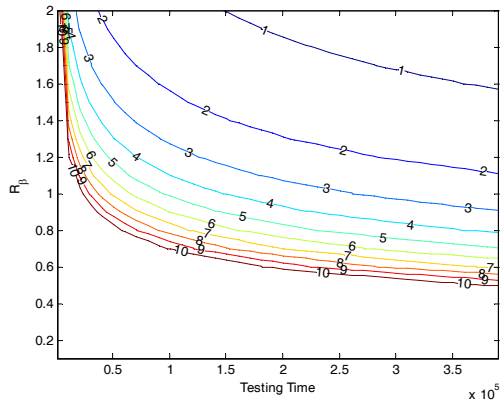
2.3 Sample Size and Test Duration Determination

As discussed at the beginning of this paper, the purpose of a test is to get the accurate estimation of model parameters or the function of model parameters for a reparable system. To achieve the required precision of an estimation, for example, where the expected log ratio of the upper and lower bounds in the confidence interval of β (or equivalently, where the expected difference of the upper and lower bounds of the $\ln \beta$), is less than R_β , we have:

$$2z_{\alpha/2} / \sqrt{mT^\beta \theta} \leq R_\beta \quad (14)$$



(a)



(b)

Similarly, for the precision requirement of θ , we have:

$$2z_{\alpha/2} \sqrt{\frac{1}{mT^\beta \theta}} (1 + \beta^2 \ln^2 T) \leq R_\theta \quad (15)$$

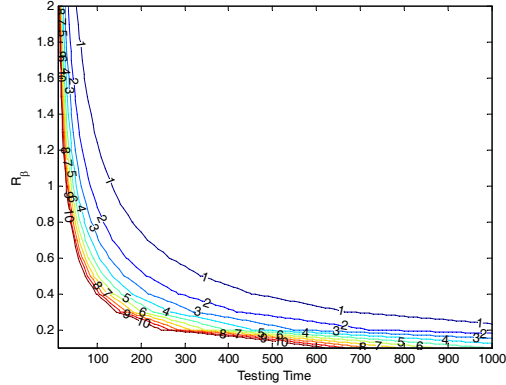
The expected number of failures at time T for m reparable systems is $E[N_m(T)] = E[\sum_{j=1}^m n_j] = m\theta T^\beta$. From (13) and (14), one

can see that the larger the sample size, the higher the precision of the estimation of model parameters β and θ will be.

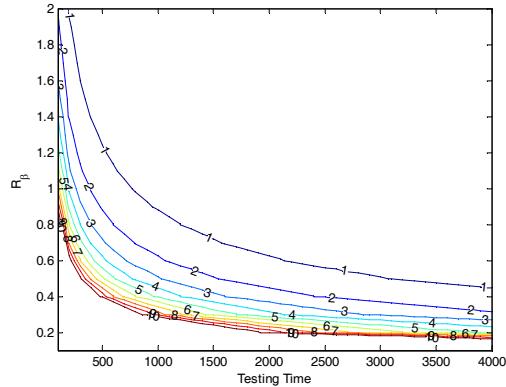
If the test plan criterion is in terms of the precision of the expected number of failures $\Lambda(t)$ at time t , we can find the similar relationship of test sample size, m , and test duration, T , with the required precision, $R_{\Lambda(t)}$. It is:

$$2z_{\alpha/2} \sqrt{\frac{1}{mT^\beta \theta} (1 + \beta^2 (\ln T - \ln t)^2)} \leq R_{\Lambda(t)} \quad (16)$$

From Eqns. 14-16, the contour plots of the minimum number of systems and the allowable test duration for a given



(c)



(d)

Figure 1- Contour plots of minimal sample size versus testing time and precision requirement of model parameter β .

The values of the parameters for the curves in Fig. 1 are:

- (a) Constant failure intensity, $\beta = 1$ $\theta = 0.01$;
- (b) Decreasing failure intensity function, $\beta = 0.5$ $\theta = 0.01$ (note: the scale of time axis is enlarged);
- (c) Increasing failure intensity function, $\beta = 1.5$ $\theta = 0.01$ (note: the scale of time axis is shrunk);
- (d) Constant failure intensity and larger intrinsic failure rate, $\beta = 1$ $\theta = 0.02$.

For all plots, $\alpha = 0.05$ and $z_{\alpha/2} = 1.96$

precision can be drawn. In order to get these plots, the pivotal

values of β and θ should be provided. This contour plot can be used by test planners to design an efficient test for a repairable system. Figure 1 provides the contour plots based on the precision requirement of model parameter β . Similar graphs can be plotted for the precision of θ or $\Lambda(t)$.

If, instead of defining the required precision of each parameter directly, we define the ratios of each parameter to its estimator's standard error as $S\hat{N}_\beta = \beta / \text{std}(\hat{\beta})$ and $S\hat{N}_\theta = \theta / \text{std}(\hat{\theta})$, we obtain:

$$S\hat{N}_\beta = \sqrt{mT^\beta \theta} \quad (17)$$

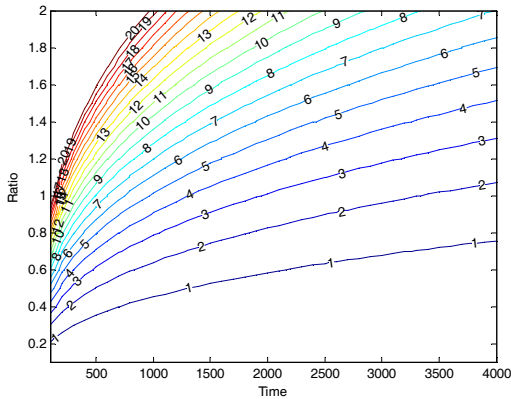
and

$$S\hat{N}_\theta = \frac{\sqrt{mT^\beta \theta}}{\sqrt{1 + \beta^2 \ln^2 T}} \quad (18)$$

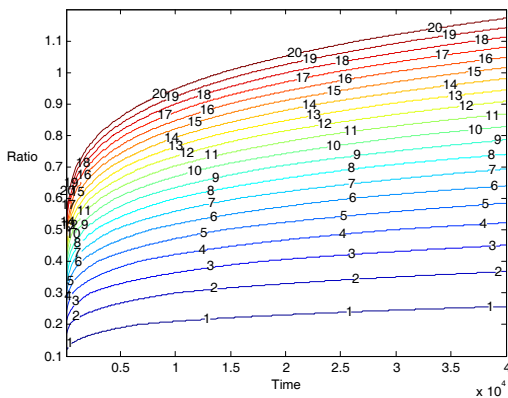
We can define that these ratios must be larger than a specific value, r . Since the denominator of the latter equation is always larger than 1, to have both ratios larger than r is equivalent to having:

$$\frac{mT^\beta}{1 + \beta^2 \ln^2 T} \geq r^2 \quad (19)$$

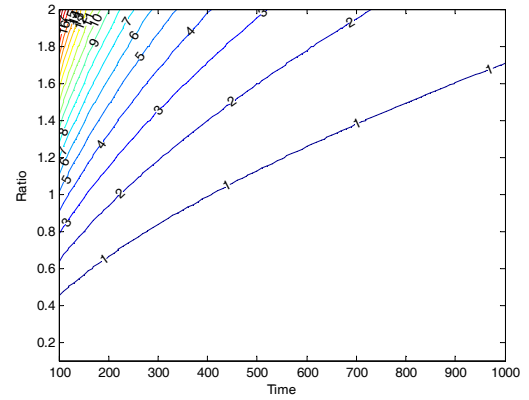
The minimum requirement of sample size and testing time is given in the above formula.



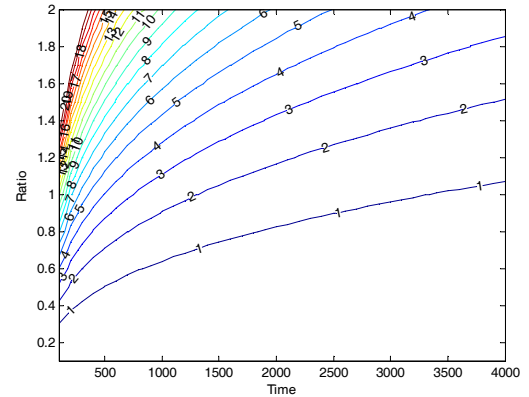
(a)



(b)



(c)



(d)

Figure 2- Contour plots of minimal sample size versus testing time and precision requirement, r .

The values of the parameters for the curves in Fig. 2 are:

- (a) Constant failure intensity, $\beta = 1$ $\theta = 0.01$;
- (b) Decreasing failure intensity function, $\beta = 0.5$ $\theta = 0.01$;
- (c) Increasing failure intensity function, $\beta = 1.5$;
- (d) Constant failure intensity and larger intrinsic failure rate, $\beta = 1$ $\theta = 0.02$.

For all plots, $\alpha = 0.05$ and $z_{\alpha/2} = 1.96$

An engineer is required to design a test to estimate the expected number of failures for a system. From the test results, the engineer needs to predict the number of failures at 3000 operation cycles. It is required that the log ratio of 90% upper confidence bound to the lower confidence bound of the predicted number of failures be less than 1.2. Suppose that the power law NHPP model is an appropriate model to describe the failure process. From the historical data, it is known that if 10 units are tested for 200 cycles each, 5 failures are expected and if 10 units are tested for 400 cycles each, 12 failures are expected. Because of the constraints of the test, the allowable test duration for the ongoing test is 400 cycles, and the minimal sample size for the test must be decided.

3 CASE STUDY AND SIMULATION COMPARISON

3.1 Case Study of Sample Size Determination

From the above information, the expected number of failures can be obtained from θt^β , and

$$\theta \times 200^\beta = 0.5 \text{ and } \theta \times 400^\beta = 1.2$$

Solving for β and θ , we get the pivotal value of $\beta = 1.263$ and $\theta = 0.00062$. Using (16), we get

$$2z_{\alpha/2} \sqrt{\frac{1}{mT^\beta \theta} (1 + \beta^2 (\ln T - \ln t)^2)} \leq R_{N(t)}$$

$$\Rightarrow m \geq 46.5$$

The engineer needs at least 47 units in the test to meet the precision requirement. Therefore, he decides to use 50 units. Using 50 units, the expected standard deviation of β , θ and $N(3000)$ will be:

$$std(\hat{\beta}) = \frac{\beta}{\sqrt{mT^\beta \theta}} = 0.1631$$

$$std(\hat{\theta}) = \sqrt{\frac{\theta}{mT^\beta} (1 + \beta^2 \ln^2 T)} = 0.00061$$

$$std[\hat{N}(t = 3000)] = t^\beta \sqrt{\frac{\theta}{mT^\beta} [1 + \beta^2 (\ln T - \ln t)^2]} = 5.39$$

Using (16), the expected bound ratio for expected number of failures at time 3000 is

$$2z_{\alpha/2} \sqrt{\frac{1}{mT^\beta \theta} (1 + \beta^2 (\ln T - \ln t)^2)} = 1.158$$

which is less than the required ratio value of 1.2.

3.2 Comparison with the Simulation Results

In order to verify the accuracy of the theoretical results, a simulation study is conducted. There are a total of 5000 simulation runs. In each simulation run, 50 systems are tested for 400 cycles. The simulated test results are used to estimate model parameters of β and θ . From the obtained 5000 β s and θ s, the mean and the standard deviation of β and θ can be calculated. Similarly, the mean and standard deviation of $N(3000)$ also can be calculated from the simulation results. For easy comparison, the standard deviations calculated from the simulation results and from the Fisher information matrix are given in the following table.

	Theta	Std (theta)	Beta	Std (beta)	N(t)	Std (N(t))
Theoretical Result	0.00062	0.00061	1.263	0.1631	15.3	5.39
Simulation Result	0.00085	0.00085	1.282	0.1660	16.9	6.75

Table 1- Theoretical and Simulation Results

In the process of getting theoretical results, we assume that β , θ and $N(t)$ follow lognormal distribution. In order to verify this assumption, we plot the lognormal probability plot for β , θ and $N(t)$.

From the probability plot, it can be seen that β is indeed lognormally distributed, while θ and $N(t)$ are slightly off from lognormal distribution.

Using the mean and standard deviation of $N(t)$. That are obtained from the simulation, and assuming $N(t)$ is lognormally distributed, we can get the upper and lower bound

ratio, which is

$$2z_{\alpha/2} \frac{std[\hat{N}(t)]}{\hat{N}(t)} = 1.29$$

- This is close to the required bound ratio 1.2. The difference is caused by the lognormal approximation in (13) and the simulation itself.

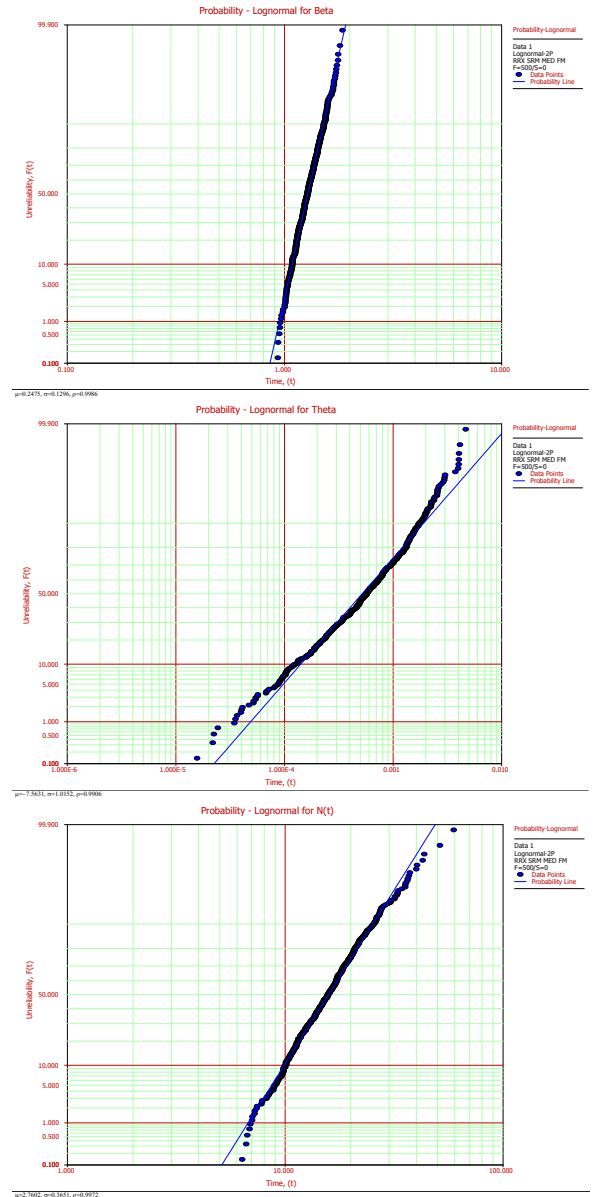


Figure 3 – Probability Plots for the Estimated Parameters

4 CONCLUSIONS

In this paper we discussed the reliability test plan for a repairable system, where the power law NHPP model is used for its failure intensity function. Confidence intervals for model parameters and for the expected number of failures at a given time are derived. They are used to determine the minimal sample size and testing time for a repairable system to meet the precision requirement of estimated system

reliability. In general, the higher the precision required, the larger the number of samples and the longer the testing time that are needed. A case study was given and the developed theoretical results were compared with simulation results. The comparison shows that the proposed method is accurate and efficient.

5 ACKNOWLEDGEMENT

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