

# Software Evaluation for Developing Software Reliability and Metrics

SRI PAMMI SRINIVAS

Department of Computer Science, DNR College, Bhimavaram (India).

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## ABSTRACT

The Dissertation presents a study and Implementation of different software metrics. We find that there are specific metrics for different stages of the software development cycle. The Metrics in this dissertation as a literature study is Metrics for Software Requirements, Metrics for Design Level. The Software Metrics covered in this Dissertation are Object-Oriented Design Metrics, Metrics for Coding Level, Metrics for Testing Level, Building cost estimation model and Software Reliability Models. Reliability metric was one of the first early prediction models to be used. The late prediction models mostly consist of the Software Reliability growth models. Jelinski and Moranda's model developed one of the earliest reliability models. Musa Basic Execution Time Model postulated that software reliability theory should be based on execution time, which is the actual processor time utilized in executing the program, rather than on calendar time.

**Key words:** Function-oriented metrics, Size-oriented metrics, Network metrics, Stability, Cyclomatic Complexity, Cohesion, Coupling, Object-oriented, Coding, Complexity, Style, Testing, Bugs, Severity, COCOMO, metrics, prediction, halstead, product, process, project, productivity, technical, faults, fault-proneness.

## INTRODUCTION

### Previous History

Size-oriented metrics are derived by normalizing quality and/or productivity measures by considering the "Size" of the software that has been produced. Function-oriented software metrics use a measure of the functionality delivered by the application as a normalization value. Network Metrics for design focus on the structure chart and define some metrics of how "good" the structure or network is in an effort to quantify the complexity of the call graph. Many of the metrics that are traditionally associated with code can be used effectively after detailed design. During detailed design all the metrics covered during the system design are applicable and useful.

### Present System

The Present study makes a number of contributions. This study makes 2 main findings. First, most of the OO design metrics in this study are statistically related to the fault-proneness of classes across fault severity. Second, the predictive abilities of these metrics strongly depend on the severity faults. The main metric of interest during testing is the reliability of the software under testing. Reliability of software depends on the faults in the software. To assess the reliability of software, reliability models are needed. To use a model for given software system, data is needed about the software that can be used in the model to estimate the reliability of the software. Software Failures and bugs are measured by quantity and by relative severity. Severity is usually determined by a local set of criteria.

**Implementation & Methodology**

**Object Oriented Design Metrics**

**Weighted Methods per Class**

The number of methods and their Complexity is a reasonable indicator of the amount of effort required to implement and test a class.

number of methods in the class.

Halstead has proposed metrics for length and volume of a program based on the number of operators and operands. We define the following measurable quantities:

$$WMC = \sum c_i$$

$$N_1 = \sum f_{1,j}$$

$$N_2 = \sum f_{2,j}$$

Where for  $i = 1$  to  $n$ , If the complexity of each method is considered 1, WMC gives the total

Where  $N_1$  is the total occurrences of different operators in the program and  $N_2$  is the total

**Table 2.1: Severity Metrics and Ranking Criteria**

| Severity ranking  | Ranking criteria                                   |
|-------------------|--|
| Severity 1 Errors | Program ceases meaningful operation                |
| Severity 2 Errors | Severe function error but application can continue |
| Severity 3 Errors | Unexpected result or inconsistent operation        |
| Severity 4 Errors | Design or suggestion                               |

**Table 2.2: Relative Seriousness (Composition) of Bugs Found**

| Error Ranking     | Ranking Description                                | Bugs Found | Bugs Reported |
|-------------------|--|------------|---------------|
| Severity 1 Errors | GPF or program ceases meaningful operation         | 18         | 9             |
| Severity 2 Errors | Severe Function error but application can continue | 11         | 11            |
| Severity 3 Errors | Unexpected result or inconsistent operation        | 19         | 19            |
| Severity 4        | Design or suggestion                               | 0          | 0             |
| Totals            |  | 48         | 39            |

**Table 2.3: Multivariate Analysis**

| Metric                    | Precision | Correctness | Completeness | $\Pi$ |
|---------------------------|-----------|-------------|--------------|-------|
| Un-graded Severity Faults |           |             |              |       |
| Model I                   | 68.28%    | 60.34%      | 79.23%       | 0.865 |
| Model I'                  | 71.03%    | 63.79%      | 76.68%       | 0.86  |
| High Severity Faults      |           |             |              |       |
| Model II                  | 84.83%    | 31.25%      | 58.82%       | 0.33  |
| Model II'                 | 84.83%    | 31.25%      | 58.82%       | 0.317 |
| Low Severity Faults       |           |             |              |       |
| Model III                 | 70.34%    | 62.50%      | 79.42%       | 0.87  |
| Model III'                | 69.66%    | 61.40%      | 74.56%       | 0.859 |

occurrences of different operands. The length of the program is defined as:

$$N = N_1 + N_2$$

From the length and vocabulary, the volume  $V$  of the program is defined as

$$V = N \log_2 (n)$$

Where  $\log_2 (n)$  is the number of bits needed to represent every element in the program uniquely, and  $N$  is the total occurrences of the different elements. Volume is used as a size metric for a program.

**Metrics for Testing Level**

**Severity**

Severity is a fundamental measure of a bug or a failure. Many ranking schemes exist for defining severity. Because there is no set standard for establishing bug severity, the magnitude of the severity of a bug is often open to debate. The following table shows the definition of the severity metrics and the ranking criteria used.

Software Failures and bugs are measured by quantity and by relative severity. Severity is usually determined by a local set of criteria.

The above figure shows the graphical representation of the bugs found shown in table 2.2.

**Types of metrics used**

Very few respondents reported using any metrics at all. Lines of Code and percent function

coverage were the two most used metrics cited by survey respondents. Function points, cyclomatic complexity and Halstead's metrics were used only rarely.

**Software Reliability Models**

**MUSA'S Basic Execution Time Model:**

In the basic model, it is assumed that each failure causes the same amount of decrement in the failure intensity.

$$\lambda (\mu) = \lambda_0 [1 - \mu/V_0 ]$$

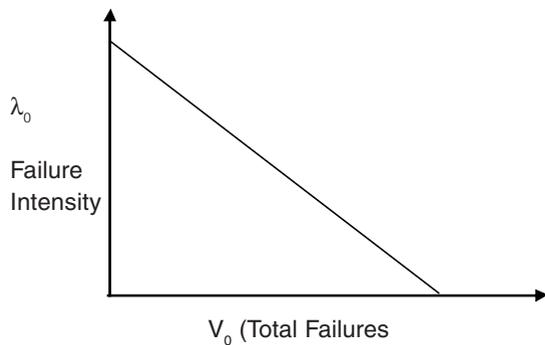
Where,

$\lambda_0$  = Initial failure intensity at the start of execution

$V_0$  = Number of failures experienced, if program is executed for infinite time period

$\mu$  = Average or expected number of failures experienced at a given point in time

The relationship between failure intensity ( $\lambda$ ) and mean failures experienced ( $\mu$ )



**Table 4: Validation Results of the Hypotheses**

| H-M    | Faults             |               |              |
|--------|--------------------|---------------|--------------|
|        | Un-graded Severity | High Severity | Low Severity |
| H-WMC  | √                  | √             | √            |
| H-DIT  | ×                  | ×             | ×            |
| H-RFC  | √                  | √             | √            |
| H-NOC  | ×                  |               | ×            |
| H-CBO  | √                  | √             | √            |
| H-LCOM | √                  | √             | √            |
| H-SLOC | √                  | √             | √            |

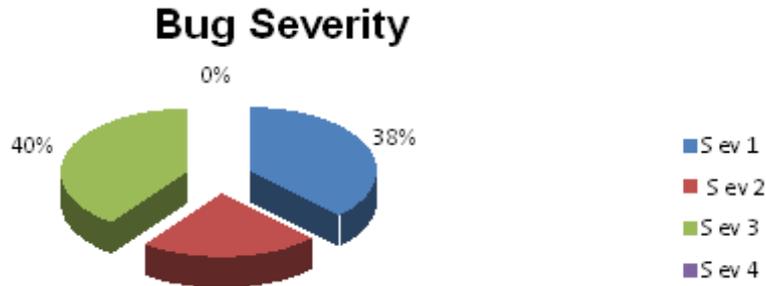


Fig. 2.1: Bug Distribution of Severity

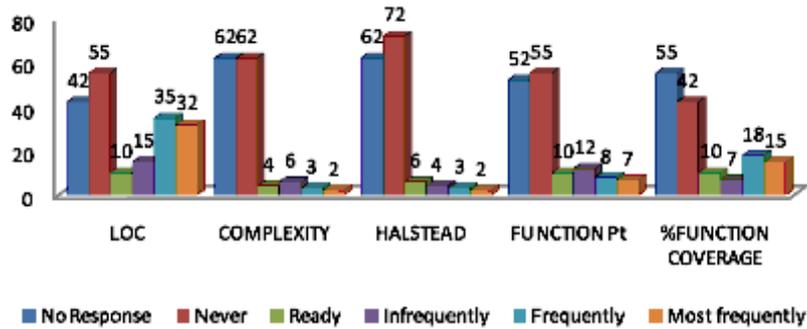


Fig. 2.: Test Metrics used by respondents

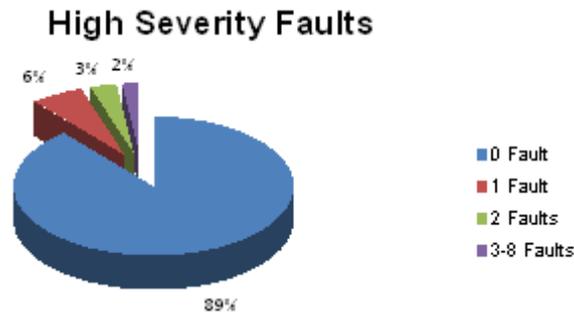


Fig. 2.4(a): High Severity Faults among all the Classes

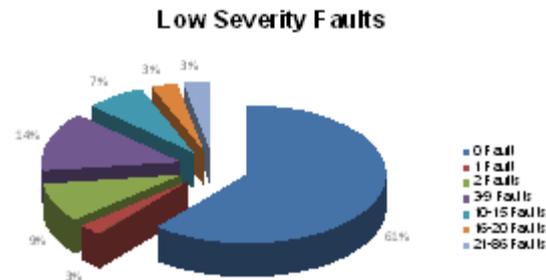


Fig. 2.4(b): Low Severity Faults among all the Classes

### Analysis of object-oriented design metrics for predicting severity faults

This study will test fault severity. Faults are either high severity or low severity. For each Metric M, the hypotheses in this study were Hypotheses Metric (H-M). H-M mean is that a class with a high M value is more likely to have high/low/un-graded severity faults than a class with a low M value. This study makes use of the public domain data set KC1, which provides both method and class-level static metrics. All Faults in KC1 were categorized as of either a high or low severity

The above figures 2.4(a) & 2.4(b) show the distribution of high severity faults and low severity faults among the 145 classes. It can be seen that 11% of all the Classes (i.e., 16 classes) contain high severity faults and may contain between one and eight faults.

### Multivariate Analysis

In this section, I can describe the performance of 3 multivariate fault-proneness prediction models built from design metrics only.

1. For un-graded severity faults
2. For high severity faults
3. For low severity faults.

### Hypothesis

Most of the metrics are aimed at getting empirical laws that relate program size to expected no. of bugs, expected no. of tests required to find bugs, resource requirement, release instant, Reliability & Quality requirement. In this section, I will use findings to validate hypotheses. I will also compare the fault-proneness prediction capability of the metrics for high severity faults and low severity faults. Apart from the analysis results of the

regression analysis, the precision, correctness, and completeness values can also help us in the validation process.

The following table summarizes the validation results of the hypotheses stated above, in which  $\lambda$  means that the hypotheses is supported,  $\chi$  means that the hypotheses is not supported, and a blank entry means that the hypotheses was not examined.

Overall, the results mean that the regression model is more useful to software managers than the simple model in practice. The results suggest that the design metrics investigated perform much better in predicting fault-prone classes in terms of low severity faults than in predicting fault-prone classes in terms of high severity faults.

## CONCLUSION

### Recommendations for further research

Previous studies have raised the need to validate OO design metrics across different faults severities. My study attempts to fill this gap by empirically validating OO design metrics for different fault severities. One limitation of my study is that the fault severity ratings in KC1 data set are subjective and may be inaccurate, which possibly limits the generalizability of my results. The data used in my study is from a single project. As a result, the findings in this paper should be viewed as exploratory and indicative rather than conclusive. In future work, we will replicate this study across projects. There is a need to validate my findings in other OO Programming languages such as Java.

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