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Software Quality Factors and Software Quality Metrics to Enhance Software Quality Assurance

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Author's contribution

This whole work was carried out by authors MCL.

Original Research Article

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ABSTRACT

Aims: Software quality assurance is a formal process for evaluating and documenting the quality of the work products during each stage of the software development lifecycle. The practice of applying software metrics to operational factors and to maintain factors is a complex task. Successful software quality assurance is highly dependent on software metrics. It needs linkage the software quality model and software metrics through quality factors in order to offer measure method for software quality assurance. The contributions of this paper build an appropriate method of Software quality metrics application in quality life cycle with software quality assurance.

Design: The purpose approach defines some software metrics in the factors and discussed several software quality assurance model and some quality factors measure method.

Methodology: This paper solves customer value evaluation problem are: Build a framework of combination of software quality criteria. Describes software metrics. Build Software quality metrics application in quality life cycle with software quality assurance.

Results: From the appropriate method of Software quality metrics application in quality life cycle with software quality assurance, each activity in the software life cycle, there is one or more QA quality measure metrics focus on ensuring the quality of the process and the resulting product. Future research is need to extend and improve the methodology to extend metrics that have been validated on one project, using our criteria, valid measures of quality on future software project.

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1. INTRODUCTION

Software quality assurance (SQA) is a technique to help achieve quality. SQA is becoming a critical issue in software development and maintenance [1]. SQA can monitor that the software engineering processes and methods used to ensure quality. Software metric deals with the measurement of software product and software product development process and it guides and evaluates software development [2]. Software metric is quantitative measure of the extent to which a system, component, or process. Software factors are going importance and acceptance in corporate sectors as organizations grow nature and strive to improve enterprise quality. The metrics are the quantitative measures of the degree to which software processes a given attribute that affects its quality. SQA is a formal process for evaluating and documenting the quality of the products produced during each stage of the software development lifecycle.

There have four quality models: McCall's quality model [3], Boehm's quality model [4], FURPS mode; [5], Dromey's quality model [6], and ISO/IEC 25000 standard. Each model contains different quality factors and quality criteria [7]). They are indicators of process and product and are useful in case of software quality assurance [8]. The aim of this paper present what are software quality factors and their criteria and their impact on the SQA function.

The remaining part of this paper is organized as follows section 2 describes literature review, it discuses the content of the relation of quality factors with quality criteria. Section 3 builds a relationship of software quality criteria between metrics. Section 4 describes software metrics, it found in the software engineering. Section 5 builds an appropriate method of Software quality metrics application in quality life cycle with software quality assurance. Fig. 1 is a research framework.

Fig. 1. A Research framework

2. LITERATURE RESEARCH

2.1 Software Quality Assurance Model

In this section, it discusses the contents of the following quality assurance model: McCall quality model, Boehm quality model, FURPS model, and Dromey model.

The McCall quality model [3] has three quality of software product: product transition (adaptability to new environment), product revision (ability to undergo changes), and product operations (its operation characteristics). Product revision includes Maintainability, Flexibility, and Testability. Product Transition includes Portability, Reusability, and Interoperability. This model contains 11 quality factors and 23 quality criteria. The quality factors describe different types of system characteristics and quality criterions are attributes to one or more of the quality factors. Table 1 is denoted as the factors and criteria of McCall quality mode

Table 1. The factors and criteria of McCall quality mode

Boehm quality model attempts to automatics and qualitatively evaluate the quality of software. The high – level characteristics address three classification; general utility into as utility, maintainability, portability. In the intermediate level characteristics, Boehm quality model have 7 quality factors like portability, reliability, efficiency, Usability, Human engineering, understandability, flexibility [4,9]. Table 2 is denoted as the quality factors and quality criteria of Boehm quality mode.

Dromey quality model proposed a framework for evaluate requirement, design and implementation phases. The high-level product properties for the implementation quality model include: correctness, internal, contextual, and descriptive ([6], [10]). Table 3 is denoted as the factors and criteria of Dromey quality mode.

Table 3. The factors and criteria of Dromey quality mode

Furps model originally presented by Grady [5], then it is extended by IBM Rational Software $([11] \sim [12])$ into FURPS+. The "+" indicates such requirements as design constraints, implementation requirements, interface requirements and physical requirements [11]. There are four characteristics in FURPS model. Table 4 is denoted as the factors and criteria of FURPS quality mode.

Table 4. The quality factors and quality criteria of FURPS quality mode

ISO 9000 – it provides guidelines for quality assurance [8]. ISO 9000 is a process oriented approach towards quality management [13]. It processes designing, documenting, implementing, supporting, monitoring, controlling and improving [14]. Recently, the ISO/IEC 9126-1: 2001 software product quality model, which defined six quality characteristics, has replaced by ISO/IEC 205010:2011 system and software product quality model [15]. ISO 25010 is the most commonly used quality standard model. It contains eight quality factors: Functional suitability, reliability, operability, security, performance efficiency, compatibility, maintainability, and portability. The 28 quality factors are arranged in six quality characteristics. Table 5 is denoted as the factors and criteria of ISO/IEC 2510 quality mode $([15) \sim [16]).$

Table 5. The factors and criteria of ISO/IEC 25010 quality mode

The quality models described above contain several factors in common, like Maintainability, Efficiency, and Reliability. However, some of factors like correctness, understandability, modifiability and supportability are not so common and are in one or two models. Table 5 compared of fours quality model and ISO/IEC 25010. Table 6 is denoted as a comparison of the factors of four quality model and ISO/IEC 25010.

Extended Al-Outaish [17]

3. COMBINATION OF SOFTWARE QUALITY CRITERIA AND SOFTWARE METRICS

Under the software quality assume model and ISO/IEC 25000 standard, it found the combination of software quality factors and criteria. In this section, it need describe the relationship of software quality criteria and software quality metrics.

There are four reasons for developing a list of criteria for each factor:

- 1. Criteria offer a more complete, concrete definition of factors.
- 2. Criteria common among factors help to illustrate the interrelation between factors.
- 3. Criteria allow audit and review metrics to be developed with greater ease.
- 4. Criteria allow us to pinpoint that area of quality factors which may not be up to a predefined acceptable standard.

3.1 Software Quality Factors and Quality Criteria

Criteria are the characteristics which define the quality factors. The criteria for the factors are the attributes of the software product or software production process by which the factor can be judged or definition. The relationships between the factors between the criteria can be found in Table 7.

Table 7. The relationships of factors with criteria of software quality

3.2 Quality Criteria and Related Factors

Table 8 is criteria for software quality factors. It provides an illustration of the relationship between these criteria and the factors [3].

Table 8. Criteria of software quality factors

Table 8 Continued…

3.3 Criteria of Software Quality Factors

The following table lists all software metrics. They copied from volume II of the specification of software quality attributes software quality evaluation guidebook [18]. Table 9 is denoted as the relationship of criteria between software quality metrics.

3.4 Quality Assurance in the Software Life Cycle

Product metrics, process metrics, and project metrics are three important types of software metrics. Product metrics measures the efficiency of accomplishing product targets for instance size, complexity, design features, performance, and quality level. Process metrics measures the efficiency of performance the product development process for instance turnover rate. Project metrics measures the efficiency of product development process, for instance schedule performance, cost performance, team performance [19].

In order to be efficient, quality assurance activities should following stage in the software life cycle. For each activity in the software life cycle, there is one or more QA support activities focus on ensuring the quality of the process and the resulting product. A concept framework of QA support software quality life cycle as shown in Fig. 2.

Table 9. The relationship of criteria between software quality metrics

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Fig. 2. A concept framework of QA support software quality life cycle

4. SOFTWARE QUALITY METRICS

This section concentrates on different metrics found in the software engineering literature. A classical classification of the software quality metrics: Halstead's software metrics, McCabe's cyclomatic complexity metric, RADC's methodology, Albrecht's function points metric, Ejiogu's software metrics, and Henry and Kafura's information metric.

4.1 Halstead's Software Metrics

Halstead's measure for calculation of module conciseness is essentially based on the assumption that a well structured program is a function of only its unique operators and operands. The best predictor of time required to develop and run the program successfully was Halstead's metric for program volume.

Halstead [20] defined the following formulas of software characterization for instance.

The measure of vocabulary: $n = n_1 + n_2$ Program length: $N = N_1 + N_2$ Program volume: $V = N \log_2 n$ Program level: $L = \frac{V^*}{V}$ $=$

Where

 n_1 = the number of unique operators

 n_2 = the number of unique operand N_1 = the total number of operators

 $N₂$ = the total number of operands

Christensen et al. [21] have taken the idea further and produced a metrics called difficulty. V^* is the minimal program volume assuming the minimal set of operands and operators for the implementation of given algorithm:

Program effort:
$$
E = \frac{V^*}{L}
$$

\nDifficulty of implementation: $D = \frac{n_1 N_2}{2n_2}$
\nProgramming time in seconds: $T = \frac{E}{S}$
\nDifficulty: $\frac{n_1}{2} + \frac{N_2}{n_2}$

With S as the Stroud number $(5 \le S \le 20)$ which is introduced from the psychological science.

 $n₂[*]$ is the minimal set of operands. E_0 is determined from programmer's previous work. The based on difficulty and volume Halstead proposed an estimator for actual programming effect, namely.

Effort = difficulty $*$ volume

Table 10 is denoted as the formulas of Halstead's software metrics with software quality factors.

4.2 McCabe's Cyclomatic complexity metrics

McCable [22] has proposed a complexity metric on mathematical graph theory. The complexity of a program is defined in terms of its control structure and is represented by the maximum number of "linearly independent" path through the program. Software developer can use this measure to determine which modules of a program are over-complex and need to be re-coded.

Software metrics	Software quality factors	Formulas
Implementation length N	Maintainability	$N = N_1 + N_2$
	Number of Bugs	$= n_1 \log_2 n_1 + n_2 \log_2 n_2$)
	Modularity	
	Performance	
	Reliability	
Volume V	Complexity	$= N \log$ η n
	Maintainability	
	Number of Bugs	
	Reliability	
	Simplicity	
Potential Volume \overrightarrow{l}	Conciseness Efficiency	$V^* = (n_1 + n_2^*) \log_2(n_1 + n_2^*)$
Program Level L	Conciseness Simplicity	
		$L = \frac{L}{L}$
Program Effort	Clarity Complexity	
	Maintainability Modifiability	$E = \frac{V^*}{V}$
	Modularity Number of Bugs	
	Performance Reliability	
	Simplicity Understandability	
Number of Bugs	Maintainability Number of	
	BugsTestability.	$R =$ Eρ

Table 10. The formulas of Halstead's software metrics with software quality factors

The formulas for the cyclomatic complexity proposed by [3] are:

 $V(G) = e - n + 2p$

Where

e = the number of edges in the graph

 $n =$ the number of nodes in the graph

 $P =$ the number of connected components in the graph.

The Cyclomatic complexity metric is based on the number of decision elements (IF-THEN-ELSE, DO WHILE, DO UNTIL, CASE) in the language and the number of AND, OR, and NOT phrases in each decision. The formula of the metric is: Cyclomatic complexity = number of decisions +number of conditions + 1[23].

The Essential complexity metricb is based on the amount of unstructured code in a program. Modules containing unstructured code may be more difficult to understand and maintain. The essential complexity proposed by McCable [22]:

$$
EV(G) = V(G) - m
$$

Where

 V (G) = the cyclomatic complexity m = the number of proper sub graphs

McCabe's Cyclomatic complexity measure has been correlated with several quality factors. These relationships are listed in Table 11.

4.3 RADC's Methodology

RADS expanded Boehm model. The metrics discussed in this section are based on continuing development effort [18]. The requirements present a ratio of actual occurrence to the possible number of occurrence for each situation: these results in a clear correlation between the quality criteria and their associated factors. Table 12 is denoted as the formulas of RADC's methodology.

Table 12. The formulas of RADC's methodology

Source: Bowen et al. [18]

1 Traceability (*Number of itemized requirements traced total number of requirements*)

$$
2\n\text{Complete } ss = (\frac{1}{1})\n\text{Solve for element } i
$$

4.4 Albrecht's Function Points Metrics

Albrecht developed a metric to determine the number of elementary functions, hence the value of source code. This metric was developed to estimates the amount the effort needed to design and develop customer applications software [24].

1. Calculation the function counts (FCs) based on the following formula:

$$
FC = \sum_{i=1}^{5} \sum_{j=1}^{3} w_{ij} \times x_{ij}
$$

Where w_{ij} are the weighting factors of the five components by complexity level (low, average, high) and x_{ii} are the numbers of each component in the application.

It is a weighted of five major components [25] are:

- ・External input: Low complexity, 3; average complexity, 4; high complexity, 6
- ・External output: Low complexity, 4; average complexity, 5; high complexity, 7
- ・Logical internal file: Low complexity, 5; average complexity, 7; high complexity, 10
- ・External interface file: Low complexity, 7; average complexity, 10; high complexity, 15
- ・External inquiry: Low complexity, 3; average complexity, 4; high complexity, 6
- 2. Calculation the value adjustment factor, it involves a scale from 0 to 5 to assess the impact of 14 general system characteristics in terms of their likely on the application. There are 14 characteristics: data communication distributed function, heavily used configuration, transaction rate, online data entry, end user efficiency, online update, complex processing, reusability, installation ease, operational ease, multiple sites, and facilitation of change.

The scores (ranging from 0 to 5) for these characteristics are then summed, based on the following formulas, to arrive at the value adjustment factor (VAF)

$$
VAF = 0.65 + 0.01 \sum_{i=1}^{14} c_i
$$

 c_i : the score of general system characteristics.

3. The number of function points is obtained by multiplying function counts and the value adjustment factor:

$$
FP = FC \times VAF
$$

4.5 Ejiogu's Software Metrics

Ejiogu's software metrics uses language constructs to determine the structural complexity of a program. The syntactical constructs are nodes. These metrics are related to the structural complexity of a program. They are also related to other quality factors, such as usability, readability, and modifiability.

The structural complexity metrics gives a numerical notion of the distribution and connectedness of a system's components [26].

$$
S_{\mathcal{C}} = H \times R_{t} \times M
$$

Where

H: the height of the deepest nested node, R_t : the Twin number of the root, M: the Monadicity [27]

The height for an individual node is the number of levels that a node is nested below the root node. The Twin number is the number of nodes that branch out from a higher level node. Monads are nodes that do not have branches emanating from them. They also are referred to as "leaf nodes".

Software size is the size of a set of nodes of source code. It is calculated using the number of modes in the tree.

 $S = total$ *umber of nodes* -1

Where

1: represents the root node.

4.6 Henry and Kafura's Information Metrics

Information flow complexity (IFC) [28] describes the amount of information which flows into and out of a procedure. This metrics use the flow between procedures to dhow the data flow complexity of a program. The Formula is:

$$
IFC = Length \times (fan - in * fan - out)^2
$$

Where

- Fan-in: The number of local flows into a procedure plus the number of global data structures from which a procedure retrieves information.
- Fan-out: The number of local flows into a procedure plus the number of global data structures from which a procedure updates. Length is the number of lines of source code in the procedure. In implementing this count, embedded comments are also counted, but not

comments preceding the beginning of the executable code.

4.7 Project Metrics

PMI PMBOK (Project management institute's project management body of knowledge) describes Project Management Processes, tools and techniques and provides one set of high level businesses for all industries. The PMBOK includes all nine knowledge areas and all associated with them tools and techniques: Integration management, Scope management, Time management, Cost management, Quality management, Human Resource management, Communication Management, Risk management, and Procurement management (PMBOK [29]).

Some of those processes often are not applicable or even irrelevant to Software Development industry. CMM (Capability Maturity model) speaks about software project planning processes without mentioning specific methodologies for project estimating described in PMBOK (PMBOK). Basic key process areas (KPA) of the SEI CMM is requirement management, project planning, project tracking and oversight, subcontract management, quality assurance, and configuration management. The Table 13 is mapping of some relevant to CMM activities, tools and techniques:

Table 13. Mapping of project management processes to process groups and knowledge areas

4.8 Reliability Metrics

A varies often used measure of reliability and availability in computer-based system is mean time between failures (MTBF) [30]. The sum of mean time to failure (MTTF) and mean time to repair (MTTR) gives the measure, i.e.

$$
MTBF = MTTF + MTTR
$$

The availability measure of software is the percentage that a program is operating according to requirement at a given time and is given by the formula:

$$
Availableility = MTTP / (MTTF + MITE)^* 100\%
$$

The reliability growth models assume in general that all defects during the development and testing phases are correct, and new errors are not introduced during theses phases. All models seem to include some constraints on the distribution of defects or the hazard rate, i.e. defect remaining in the system.

Increase software reliability gives the metrics:

$$
F\text{ailure rate (FR)} = \frac{Number\ of\ failures}{Execution\ time}
$$

4.9 Readability metrics

Walston and Felix [31] defined a ratio of document pages to LOC as:

 $D - 49I^{1.01}$

Where

D= number of pages of document $L =$ number of 1000 lines of code.

4.10 Metrics-Based Estimation Models

4.10.1 COCOMO Model

Most of the models presented in this subsection are estimators of the effort needed to produce a software product. Probably the best known estimation model is Boehm's COCOMO model [32]. The first one is a basic model which is a single-value model that computes software development effort and cost as a function of program size expressed as estimated lines of code (LOC). The second COCOMO model computes software development effort as a function of program size and a set of "coat drives" that include subjective assessment of product, hardware, personal, and project attributes. The basic COCOMO equations are:

$$
E = a_i (KLOC)^{b_i}, \ D = c_i E^{d_i}
$$

Where

E is the effort applied in person-month. D is the development time in chronological months

The coefficients a_i and c_i and the exponents b_i and d_i are given in Table 14.

Table 14. Basic COCOMO

The second COCOMO has some special features, which distinguish it from other ones. The usage of this method is very wide and its results are accurate. The equations are use to estimate effort and schedule see [33].

4.10.2 Putnam estimation model

The Putnam estimation model ($[34] \sim [35]$) assumes a specific distribution of effort over the software development project. The distribution of effort can be described by the Royleigh-Norden curve. The equation is:

$$
L = c_k K^{1/3} t_d 4/3
$$

Where

 c_k is the state of technology constant (the environment indictor),

 k is the effort expended (in person-years) over the whole life cycle.

 t_d is the development time in year.

The c_k valued ranging from 2000 for poor to 11000 for an excellent environment is used [36].

4.10.3 Source line of code

SLOC is an estimation parameter that illustrates the number of all comments and data definition but it does not include instructions such as comments, blanks, and continuation lines. Since SLOC is computed based on language instructions, comparing the size of software which uses different language is too hard. SLOC usually is computed by considering S_L as the lowest, S_H as the highest and S_M as the most probable size [37].

$$
S = \frac{S_L + 4S_M + S_H}{6}
$$

4.10.4 Productive estimation model

Walston and Fellix [31] give a productivity estimator of a similar form at their document metric. The programming productivity is defined as the ratio of the delivered source lines of code to the total effort in person-months required to produce the delivered product.

 $E = 5.2L^{0.91}$ **E**

Where

E is total effort in person-month

L is the number of 1000lines of code.

4.11 Metrics for software maintenance

During the maintenance phase, the following metrics are very important: [37]

- ・ Fix backlog and backlog management index
- ・ Fix response time and fix responsiveness
- ・ Percent delinquent fixes
- ・ Fix quality

Fix backlog is a workload statement for software maintenance. To manage the backlog of open, unresolved, problems is the backlog management index (BMI). If BMI is large then 100, it means the backlog is reduced. If BMI is less than 100, then the backlog increased.

> $\times 100\%$ *Humber of problrms closed during the month Number of problem arrivals during the month*

4.12 Customer Problem Metrics

The customer problems metric can be regarded as an intermediate measurement between defects measure and customer satisfaction. The problems metric is usually expressed in terms of problem per user month (PUM). PUM is usually calculated for each month after the software is released to market, and also for monthly averages by user.

Several metrics with slight variations can be Constructed and used, depending on the purpose of analysis. For example: $(38] \sim [40]$).

- ・ Percent of completely satisfied customers.
- ・ Percent of satisfied customers (satisfied and completely satisfied)
- ・ Percent of dissatisfied customers(dissatisfied and completely dissatisfied)
- ・ Percent of non (neutral, dissatisfied, and completely dissatisfied).
- ・ Customer founded defects (CFD) total

 CFD *total* $=$ $\frac{Number\ of\ customer - founded\ defects}{Assembly - equivalent\ total\ source\ size}$

・ Customer – founded defects (CFD) delta

Assembly equivalent total source size increasemental software development CFD total Number of customer founded defects caused by PUM = Total problems that customers reported (true defects and non-defects- orients problems) for a time period + Total number of License- months of the software during the period.

Where Number of license- months = Number of install license of the software \times Number of months in the calculation period.

4.13 Test Product and Process Metrics

Test process metrics provide information about preparation for testing, test execution and test progress. Some testing metrics ([36,41,42]) as following:

- 1. Number of test cases designed
- 2. Number of test cases executed
- 3. DA = Number of defects rejected / Total number of defects *100%
- 4. Bad Fix Defect =Number of Bad Fix Defect / Total number of valid defects *100%
- 5. Test case defect density = (Number of failed tests / Number of executed test cases) *100
- 6. Total actual execution time/ total estimated execution time
- 7. Average execution time of a test case

Test product metrics provide information about the test state and testing status of a software product. Using these metrics we can measure the products test state and indicative level quality, useful for product release decision [42].

- 1. Test Efficiency = $(D_T/(D_T+D_U)^*100$
- 2. Test Effectiveness = $(D_T/(D_F+D_U)^*100$
- 3. Test improvement TI = number of defects detected by the test team during / source lines of code in thousands
- 4. Test time over development time TD = number of business days used for product testing / number of business days used for product
- 5. Test cost normalized to product size (TCS) = total cost of testing the product in dollars / source lines of code in thousands
- 6. Test cost as a ration of development cost (TCD) = total cost of testing the product in dollars / total cost of developing the product in dollars
- 7. Test improvement in product quality = Number of defects found in the product after release / source lines of code in thousands
- 8. Cost per defect unit = Total cost of a specific test phase in dollars / number of defects found in the product after release
- 9. Test effectiveness for driving out defects in each test phase = $(D_D/(D_D+D_N)^*100$
- 10. Performance test efficiency (PTE) = requirement during perform test / (requirement during performance time + requirement after signoff of performance time) * 100%
- 11. Cost per defect unit = Total cost of a specific test phase in dollars / number of defects found in the product after release
- 12. Estimated time for testing
- 13. Actual testing time
- 14. % of time spent = (actual time spent / Estimating time)*100

Where

DD : Number of defects of this defect type that are detected after the test phase.

- D_T : Number of defects found by the test team during the product cycle
- *DU* : Number of defects of found in the product under test (before official release)
- D_F : Number of defects found in the product after release the test phase
- D_N : Number of defects of this defect type (any particular type) that remain uncovered after the test phase.

4.14 Method of Statistical Analysis

The revisions on the software measurement methods, developed with the purpose of improving their consistency must be empirically evaluated so as to determine to what extent is the pursued goal fulfilled. The most used statistical methods are given in the following table ($[43] \sim [48]$). Some commonly used statistical methodology (include nonparametric tests) are discussed as follow:

- 1. Ordinary least square regression models: Ordinary least square regression (OLS) model is used to subsystem defects or defect densities prediction
- 2. Poisson models: Poisson analysis applied to library unit aggregation defect analysis
- 3. Binomial analysis: Calculation the probability of defect injection
- 4. Ordered response models: Defect proneness
- 5. Proportional hazards models: Failure analysis incorporating software characteristics
- 6. Factor analysis: Evaluation of design languages based on code measurement
- 7. Bayesian networks: Analysis of the relationship between defects detecting during test and residual defects delivered
- 8. Spearman rank correlation coefficient: Spearman's coefficient can be used when both dependent (outcome; response) variable and independent (predictor) variable are ordinal numeric, or when one variable is an ordinal numeric and the other is a continuous variable.
- 9. Pearson or multiple correlations: Pearson correlation is widely used in statistics to measure the degree of the relationship between linear related variables. For the Pearson correlation, both variables should be normally distributed
- 10. Mann Whitney U test: Mann Whitney U test is a non-parametric statistical hypothesis test for assessing whether one of two samples of independent observations tends to have larger values than the other
- 11. Wald-Wolfowitz two-sample Run test: Wald-Wolfowitz two-sample Run test is used to examine whether two samples come from populations having same distribution.
- 12. Median test for two samples: To test whether or not two samples come from same population, median test is used. It is more efficient than run test each sample should be size 10 at least.
- 13. Sign test for match pairs: When one member of the pair is associated with the treatment A and the other with treatment B, sign test has wide applicability.
- 14. Run test for randomness: Run test is used for examining whether or not a set of observations constitutes a random sample from an infinite population. Test of randomness is of major importance because the assumption of randomness underlies statistical inference.
- 15. Wilcoxon signed rank test for matcher pairs: Where there is some kind of pairing between observations in two samples, ordinary two sample tests are not appropriate.
- 16. Kolmogorov-Smirnov test: Where there is unequal number of observations in two samples, Kolmogorov-Smirnov test is appropriate. This test is used to test whether there is any significant difference between two treatments A and B.

5. SOFTWARE QUALITY METRICS IN QUALITY LIFE CYCLE WITH SOFTWARE QUALITY ASSURANCE

Software quality metrics focus on quality aspects of product metrics, process metrics, maintenance metrics, customer metrics and project metrics.

Product metrics are measures of the software product at any stage of its development, from requirements to installed system. Product metrics may measure the complexity of the software design, the size of the final program, or the number of pages of documentation production. Process metrics are measure of the software development process, such as overall development time, the average level of experience of the programming staff, or type of methodology used. The test process metrics provide information about preparation for testing, test execution and test progress. Some test product metrics are number of test cases design, % of test cases execution, or % test cases failed. Test product metrics provide information of about the test state and testing status of a software product and are generated by execution and code fixes or deferment. Some rest product metrics are Estimated time for testing, average time interval between failures, or time remaining to complete the testing.

The software maintenance phases the defect arrivals by time interval and customer problem calls. The following metrics are therefore very important: Fix backlog and backlog management index, fix response time and fix responsiveness, percent delinquent fixes, and fix quality.

Subjective metrics may measure different values for a given metric, since their subjective judgment is involved in arriving at the measured value. An example of a subjective product metric is a classification of the software as "organic", "semi-detached" or "embedded" as required in the COCOMO cost estimation model.

From the customer's perspective, it is bad enough to encounter functional defects when running a business on the software. The problems metric is usually expressed in terms of problem per user month (PUM). PUM is usually calculated for each month after the software is released to market, and also for monthly averages by user.

The customer problems metric can be regarded as an intermediate measurement between defects measure and customer satisfaction. To reduce customer problems, one has to reduce the functional defects in the products, and improve other factors (usability, documentation, problem rediscovery, etc.). Table 15 is denoted as software quality assurance with quality measure metrics in quality life cycle.

Table 15. Software quality assurance with quality measure metrics in quality life cycle

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Source: this study

6. CONCLUSION

Software quality metrics focus on quality aspects of product, process, and project. They group into six categories in accordance with the software life cycle: project metrics, requirements gathering, product metrics, process metrics, ensure Project deployment (customer satisfaction metrics), track support and change management (maintenance metrics). In order to understand the relationship of criteria of software quality factor, we have discussed software quality model and standard, quality factors and quality criteria, quality criteria and quality metric. We detail discussed software quality metrics. It includes Halstead's software metrics, McCabe's Cyclomatic complexity metrics, RADC's methodology, Albrecht's function points metric, Ejiogu's software metrics, Henry and Kafura's information metric, project metric, Reliability metrics, Readability metrics, Metricsbased estimation models, Metrics for software maintenance, In- process quality metrics, Customer problem metrics, Test product and process metrics, and Method of statistical analysis. Under the above 15 software quality metrics, we give table of software quality assurance with quality measure metrics in quality life cycle. It contains software quality factors and software quality measure metric in each software development phase.

In order to continue to improve its software product, processes, and customer services. Future research is need to extend and improve the methodology to extend metrics that have been validated on one project, using our criteria, valid measures of quality on future software project.

COMPETING INTERESTS

Author declares that there are no competing interests.

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