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How Good Is the Software: A Review of Defect Prediction Techniques

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Version 1.0

A Review of Defect Prediction Techniques - page 1



Objectives

Awareness of defect prediction and estimation techniques

Awareness of the value to project management and process improvement activities of analyzing defect data

Recommendations for getting started



Why Analyze and Predict Defects?

Project Management

- Assess project progress
- Plan defect detection activities

Work Product Assessment

• Decide work product quality

Process Management

- Assess process performance
- Improve capability



Outline

- F Background
 - Definitions
 - Measuring Defects
 - Tools and techniques
 - Attributes

Technique Review

- Project Management
- Work Product Assessment
- Process Improvement



Definition - Software Defect

Software Defect: any flaw or imperfection in a software work product or software process

- software work product is any artifact created as part of the software process
- software process is a set of activities, methods, practices, and transformations that people use to develop and maintain software work products

A defect is frequently referred to as a fault or bug

Focus on Predicting those Defects that affect Project and Product Performance



Defects as the Focus of Prediction

Distinguish between major and minor defects

- do not use minor or documentation defects in predictions
- minor defects will inflate estimate of latent product defects

Most defect prediction techniques used in planning rely on historical data

Defect prediction techniques vary in the types of data they require

- some require little data, others require more
- some use work product characteristics, others require defect data only

Techniques have strengths and weaknesses depending on the quality of the inputs used for prediction



Defect Attributes Available for Analysis

Problem Status Open Recognized Evaluated Resolved Closed Problem Type Software Defect Requirements Defect Design Defect Design Defect Code Defect Operational Doc. Defect Test Case Defect Other Product Defect

Problem Type (con't) Other Problems Hardware Problem **Operating System Problem** User Mistake **Operations Mistake** New Req't / Enhancement Undetermined Not repeatable Value not identified Uniqueness Original Duplicate **Criticality Level** Urgency



Additional Attributes to Consider

Recognition

- What is the problem?
- When was the problem reported?
- Who reported the problem?

Evaluation

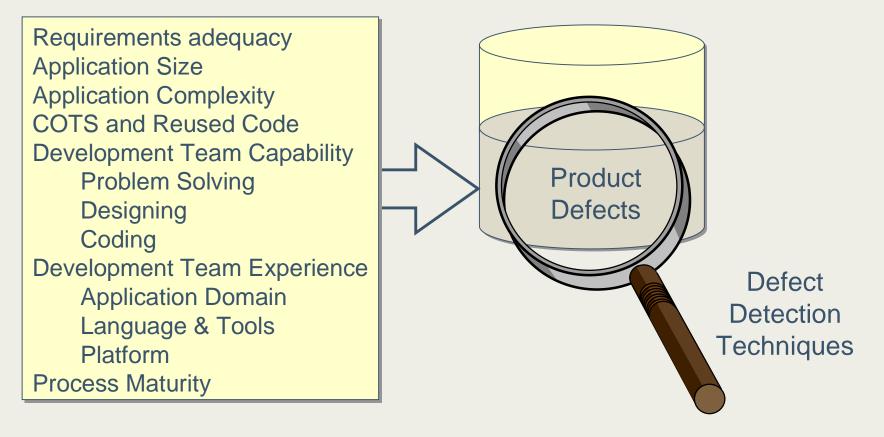
- What work product caused the problem?
- What activity discovered the problem?
- What activity introduced the problem? Resolution
- What work needs to be done?
- What work products will be affected by the change?
- What are the prerequisite changes?

Closure

- When are the changes expected?
- What configuration contains the changes?



Project and Process Factors Correlated with Defect Insertion

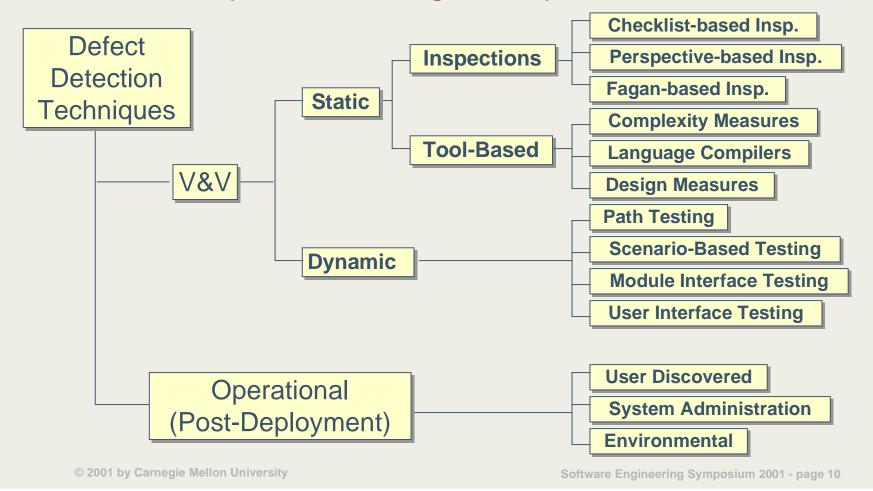


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Defect Discovery Sources

(how are the data generated)





Outline

Background

- Definitions
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F Technique Review

- Project Management
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Defect Prediction Techniques

Project Management

- Empirical Defect Prediction
- Defect Discovery Profile
- COQUALMO
- Orthogonal Defect Classification

Work Product Assessment

- Fault Proneness Evaluation (Size, Complexity, Prior History)
- Capture/Recapture Analysis

Process Improvement

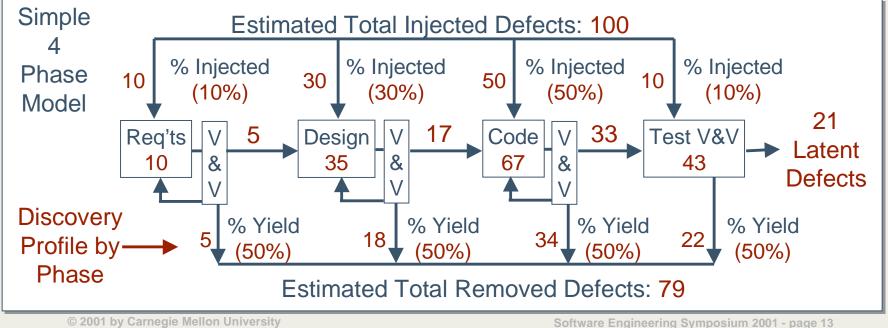
- Defect Prevention Program
- Statistical Process Control



Empirical Defect Prediction Technique Review

Description - number of Defects per Size (Defect Density)

- defect density (Number of Defects / Thousands Lines of Code) based on historical data
- enhanced with historical data on injection distribution and yield





Empirical Defect Prediction - 2

When to use - use for planning (total defects) and in-process monitoring of defect discovery numbers (latent defects)

Required Data - historical defect density data required for planning; in-process data required for monitoring

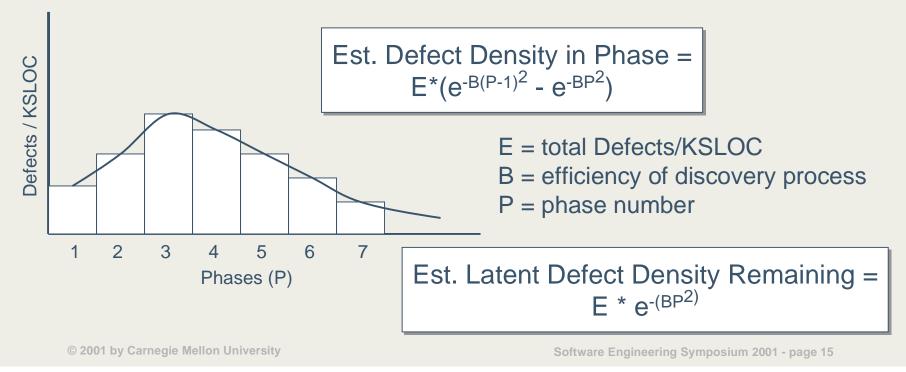
Strengths - easy to use and understand; can be implemented with minimal data

Weaknesses - requires stable processes and standardized life cycle; does not account for changes in the project, personnel, platform, or project



Defect Discovery Profile Technique Review

Description - projection, based on time or phases, of defect density (or number of defects) found "in-process" onto a theoretical discovery curve (Rayleigh). Found in the SWEEP and STEER models





Defect Discovery Profile - 2

When to use - as early in the life cycle as defect data collection permits

Required Data - historical defect density data, estimated and actual size, and consistently tracked defect counts

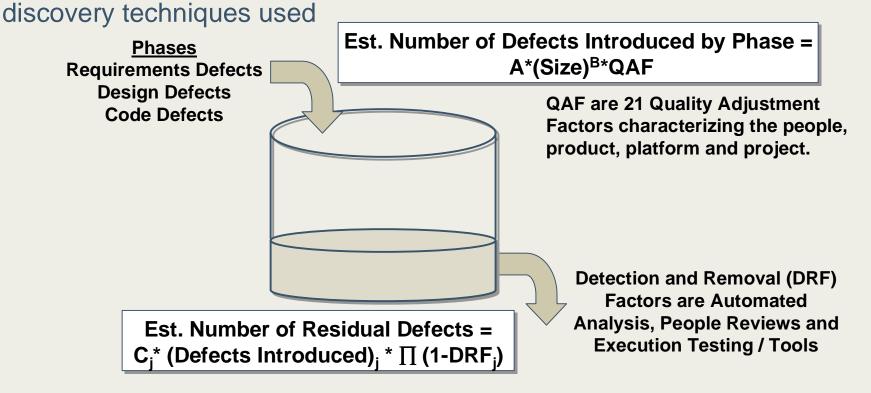
Strengths - predicts defect density by time period enabling the estimation of defects to be found in test

Weaknesses - no insight or adjustment mechanism for B to account for changes in the product, personnel, platform, or project will impact defect predictions.



COQUALMO Technique Review

Description - a defect prediction model for the requirements, design, and coding phases based on sources of introduction and





COQUALMO - 2

When to use – used in the planning phase of a project

Required Data - size of the product and ratings for 21 Quality Adjustment Factors

Strengths - predicts defects for three phases; quantifies the effect of different discovery techniques on the detection and removal of defects. Considers the effects of attributes such as product, personnel, project, and platform

Weaknesses - covers a small number of phases; does not predict test or post-deployment defects

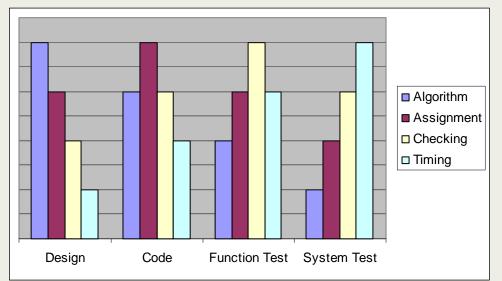


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Orthogonal Defect Classification Technique Review

Description – classification and analysis of defects to identify project status based on comparison of current defects with historical patterns; identify areas for process improvement based on analysis of defect types, "triggers," impact, and source

- Types are what was required for the fix, not the cause of the defect (e.g. function, assignment, interface)
- Triggers are catalysts that cause defects to surface (e.g. testing, inspection, conditions of operational use)





Orthogonal Defect Classification - 2

When to use – ongoing throughout project

Required Data – orthogonal defect classification scheme mapped to development process; historical defect profiles

Strengths – classifications linked to process provide valuable insight; classification takes little time

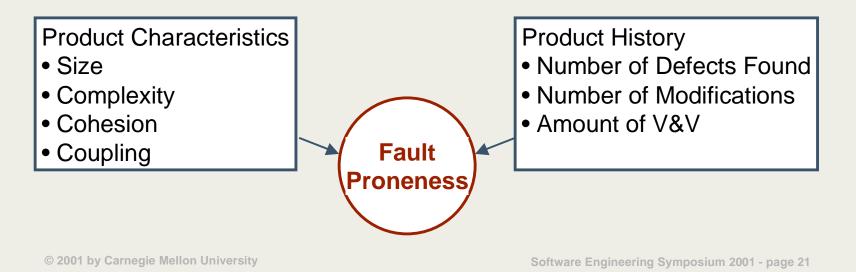
Weaknesses – requires development of classification scheme; reliable classification of defects; ongoing data collection and analysis; does not account for changes in people, process, or product



Fault Proneness Technique Review

Description – analysis of work product attributes to plan for allocation of defect detection resources (inspection and testing)

- A variety of models and heuristics
- comparing cyclomatic complexity against a threshold
- various parametric models (e.g., discriminant analysis)
- reviewing module or component defect histories





Fault Proneness - 2

When to use - test planning, during coding and testing,

Required Data – size, complexity, coupling, historical defect data, etc.

Strengths – efficient and effective focus of defect detection activities

Weaknesses – "in-process" fault density by module or component may not predict operational fault density, effort may be misdirected; models and assumptions not likely to hold from one system to the next



Capture Recapture Technique Review

Description – analysis of pattern of defects detected within an artifact by independent defect detection activities (inspectors or inspection versus test)

Number of remaining defects is estimated from the overlap in defects identified independently by individual inspectors according to the following formula:

N (estimated) in work product =

n(inspector 1) * n(inspector 2)

m (number defects found by both inspectors)

N(estimated) – N (unique discovered) = Remaining defects (est.)

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Capture Recapture - 2

When to use – determining whether a work product should undergo re-inspection

Required Data – detailed defect descriptions from each inspector

Strengths – can be used as soon as data are available

Weaknesses – estimates of number of remaining defects best when stringent assumptions are met. Relaxing assumptions requires more complicated estimation. More robust when simply used to predict whether a criterion for re-inspection has been exceeded

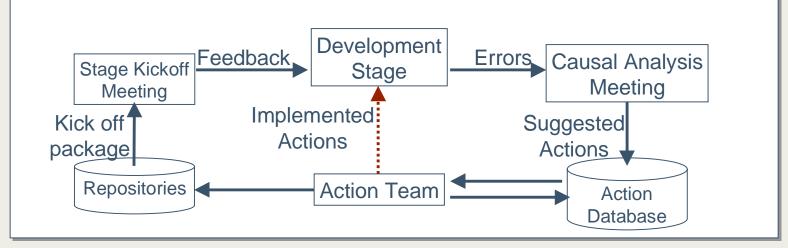


Defect Prevention Technique Review

Description – root cause analysis of most frequently occurring defects

Sample of defect reports selected for in-depth causal analysis

Actions taken to make process changes or improve training to eliminate the root cause of defects and prevent their recurrence



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Defect Prevention-2

When to use – prior to launching new projects or beginning new phases of a project

Required Data – historical defect data

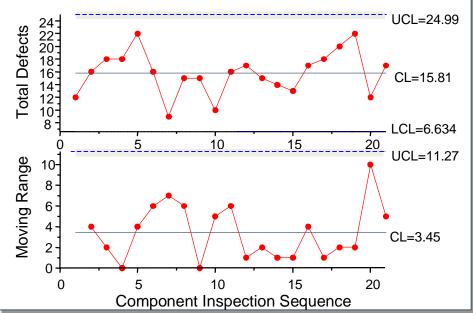
Strengths – allows for comparison of defect trends over time to assess impact and ROI for defect prevention activities

Weaknesses – requires sampling of defects and in-depth analysis and participation by engineers to identify root causes



Statistical Process Control Technique Review

Description – use of control charts to determine whether inspection performance was consistent with prior process performance. Process capability depicts expected range of performance in terms of selected attribute.



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Statistical Process Control - 2

When to use – when inspections are being conducted

Required Data – current measures of inspection process (e.g., defects found, prep time, review rate); historical measures to develop control chart

Strengths – gives indication of inspection and development process performance. Signals provide rapid feedback suggesting re-inspection or process anomaly.

Weaknesses – requires stable process and "real time" data collection and analysis



Observations

For Project Management

- models predict total defects in a product and latent defects from "in-process" measures
- models use estimated and actual software size as a parameter
- models use additional factors to adjust defect estimates

For Product Quality

• predictions can be developed from inspection or product characteristics data

For Process Improvement

• expected process behavior can be used to gauge performance and identify opportunities for improvement



Observations -2

Prediction models are useful for planning and establishing expectations.

Tracking against expectations

- when deviations occur some action is taken such as reallocation of resources towards defect detection, specific modules, or re-inspection.
- most analysis techniques are not very explicit on the threshold that triggers investigation. The exception is control limits in SPC.

Estimates are often inaccurate but suggestive and value-added for decision making and planning



Recommendations for Getting Started

Get started even with simple techniques

- the data available will help determine the technique
- availability of historical data will drive model selection
- analyze for patterns across defects, don't just fix the defects

Measure product defects early in the life cycle

- Post-release defect tracking is the least helpful
- Pre-release defect tracking by phase and by type is most helpful but also more burdensome

Defect definition should meet the intended use of the data

- Track project progress
- Determine product quality
- Assess process performance

Changes in the product, personnel, platform, or project must be measured and accounted for in predictions



Call for Collaboration

If you are interested in using these techniques or studying their effectiveness, please contact:

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I look forward to hearing from you.



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