

# Empirical Software Engineering

## CSE 8340 — Spring 2014

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### **Module Iib: TBM in Risk Ident.**

- Telecom Case Study
- TBM and Applications
- Results and Discussions

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## Overview

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- Project background:
  - ▷ Reliability and QA:
  - ▷ SMU research under NSF&THECB grants 1998–: Tian, Nguyen, and others.
  - ▷ Industrial support: Nortel Networks Frame, Allen, Appan, and others.
  
- Planned activities:
  - ▷ Defect analysis: TBDMs (Tree-based defect models)
  - ▷ Reliability improvement: TBRMs (Tree-based reliability models)
  - ▷ Other: UMMs, testing, HT, etc.
  
- JSS paper by Tian/Nguyen/Allen/Appan.

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## Overview: Context

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- Objectives of study:  
Understanding problem prone modules for quality management and improvement.
  
- Objects of study:
  - ▷ Large telecommunication software
  - ▷ Nortel Networks products: NT-X
  - ▷ Developed/released/used recently
  - ▷ 5 different releases
  - ▷ Latest finished around 2000/2001
  
- Development environment:
  - ▷ Waterfall-like process
  - ▷ Change/incremental development
  - ▷ Project monitoring through various measurements and tools (e.g., Datrix/EMERALD/COMET)

## Overview: Approach (Design)

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- General approach to the study:
  - ▷ Various measurement data.
  - ▷ Identify problem prone modules
  - ▷ Characterize problem prone modules
  - ▷ Conclusions based on above analyses.
  
- Classification: observational
  
- 5 different releases
  - ▷ Consistency among releases
    - assessment and understanding
  - ▷ Guidance for new release
    - prediction and control

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## Risk Identification: Why?

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- Risk and 80:20 rule
  - ▷ Risk: (high) probability of undesirable situations or consequences
  - ▷ 80:20 rule: 80% of problems traceable to 20% of components
  - ▷ Need risk identification
  
- Problem-prone modules
  - ▷ Likely to contain substantially more internal or development defects.
  - ▷ (Fault-prone: in-field failures)
  - ▷ Identification of the modules
  - ▷ Corrective/remedial actions
  
- Identify problem prone modules
  - ▷ Data: past defect and other metrics.
  - ▷ Technique: risk identification.
  - ▷ Followup: Characterization.

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## Risk Identification: How?

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- Techniques used in Nortel Networks:
  - ▷ EMERALD: mainly multiple regression and logistic analysis, with limited use of neural networks.
  - ▷ COMET: principal component analysis (PCA) and discriminant analysis.
- New techniques:
  - ▷ Tree-based modeling with S-PLUS
  - ▷ CART with SAS
- Primary technique here: TBM.
  - ▷ generic comparison: Tian SQP paper.
  - ▷ specifics: later

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## Product and Defect Metrics

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- Defect metrics:
  - ▷ DF: defect fixes
  - ▷ Applied in response to testing failures.
  - ▷ DF vs. failure/fault counts:
    - DF captures propagation information.
    - DF is identified with specific modules.
  - ▷ Available in project data depository
  - ▷ Data transformed to percentages.

$$DF = \frac{DF_{raw}}{DF_{max}} \times 100\%$$

- Product metrics:
  - ▷ From EMERALD, a Nortel tool/product
  - ▷ Underlying analyzer for procedure-level metrics
  - ▷ Module level metrics  $\sim$  DF

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## Product Metrics: Details

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- 53 raw product metrics:
  - ▷ volume,
  - ▷ testability, decision complexity, dead code, independent path, structuredness,
  - ▷ readability,
  - ▷ section dependability,
  - ▷ software science.
  
- 6 synthetic product metrics:
  - ▷ OurRange: # metrics  $\notin$  acceptable range
    - a rough indicator of module quality
  - ▷ Level (or procedure type)



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## Risk Identification: EMERALD

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- Techniques used:
  - ▷ Mixture of old and new.
  - ▷ Multiple regression & logistic analysis.
  - ▷ Neural network etc.
  
- EMERALD output:
  - ▷ OpRisk: likelihood of field defect.
  - ▷ Values: “green”, “R”, ..., “RRRRRRR”.
  - ▷ Other output also possible.
  - ▷ But not DF for this model
  - ▷ Identifying but not characterizing
  - ▷ Try other models

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## TBDMs: Why?

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- Risk identification:
  - ▷ Assumption (in traditional techniques):
    - linear relation
    - uniformly valid result
  - ▷ Reality of defect distribution:
    - isolated pocket
    - different types of metrics
    - correlation/dependency in metrics
    - qualitative differences
  - ▷ Need new risk id. techniques.
  
- Risk characterization:
  - ▷ Identified, then what?
  - ▷ Result interpretation.
  - ▷ Remedial/corrective actions.
  - ▷ Extrapolation to new product/release.
  - ▷ TBDMs appropriate.

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## TBM & TBDMs: Ideas

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- TBDMs: tree-based defect models using tree-based modeling (TBM) technique
  
- Decision trees:
  - ▷ multiple/multi-stage decisions
  - ▷ may be context-sensitive
  - ▷ natural to the decision process
  - ▷ applications in many problems
    - decision making & problem solving
    - decision analysis/optimization
  
- Tree-based models:
  - ▷ reverse process of decision trees
  - ▷ data  $\Rightarrow$  tree
  - ▷ idea of decision extraction
  - ▷ generalization of "decision"

## TBM: Types and Applications

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- Key "selling" points:
  - ▷ intuitiveness and interpretation
    - compare to PCA, NN
  - ▷ quantitative & qualitative info.
  - ▷ hierarchy/importance/organization
  
- Past applications:
  - ▷ social sciences
  - ▷ Selby&Porter: Amadeus project
  - ▷ Tian et al:
    - NASA/SEL work (area IV)
    - IBM product defects: with Troster
    - IBM TBRM: 8317 coverage
    - SMU: UMM/testing, Nortel work

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## TBM: Technique

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- Technique: tree-based modeling
  - ▷ Tree: nodes=data-set, edges=decision.
  - ▷ Data attributes:
    - 1 response &  $n$  predictor variables.
  - ▷ Construction: recursive partitioning.
  - ▷ Usage: relating response to predictors
    - $Y = Tree(X_1, \dots, X_n)$
    - understanding vs. predicting
    - identification and characterization
  - ▷ Works for mixed-types of data.
  - ▷ Tree growing and pruning.
  
- Algorithm: Fig.1
  - ▷ regression tree and example
  - ▷ classification tree: modify Step 3

## TBDMs: Result for NT-X

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- Overall result: Fig. 2
  - ▷ Similar results for other releases
  - ▷ General understanding: simplicity/pruning
  
- How to read each node?
  - DF and node size summary
  
- Split conditions
  - ▷ Distinguishing characteristics
  - ▷ Root to leaf: order of importance
  - ▷ Metrics selected out of 59 by algorithm:
    - Halstead program length (HalLen)
    - # basic utility routines (Level1)
    - # include files (FilIncNbr)
    - comments volume average (ComStrAvg)
    - Halstead level (HalLv1)

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## TBDMs: Result for NT-X

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- Identifying problem prone modules
  - ▷ Identified leaf nodes (Table 1)
  - ▷ Comparison to other nodes (Fig. 2)
  - ▷ Isolated pockets: llrl, rr (rrl + rrr)
  - ▷ Groups vs. individual identification
  
- Characterizing problem prone modules
  - ▷ Split conditions as characterization
  - ▷ Symptoms of problems
  - ▷ Further analysis  $\Rightarrow$  systematic problems
  - ▷ Corrective/remedial actions
  - ▷ Future process/product improvement

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## Other TBDM Results

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- TBDM performed:
  - ▷ DF  $\sim$  metrics (previous)
  - ▷ DF  $\sim$  *OpRisk*
  - ▷ DF  $\sim$  all
  
- EMERALD result validation
  - ▷ TBDM set1: tot.fix  $\sim$  *OpRisk*
  - ▷ TBDM set2: tot.fix  $\sim$  all
  - ▷ Consistent pattern
  - ▷ Reasonable predictions, but...
    - not much constructive info.



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## TBDMs: NT-X vs. IBM LS and NS

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- IBM products for comparison:
  - ▷ LS and NS: A legacy and a new system
  - ▷ Large s/w systems: 995, 1302 modules
  - ▷ Metrics: DF + (11, 15) other
    - design (6), size (2), complexity (5, 3)
    - change (2 for LS)
  
- Results for IBM LS and NS:
  - ▷ LS: change, size, data complexity
  - ▷ NS: design and control complexity
  - ▷ Problem-prone modules: Table 2
  
- Comparison: NT-X similar to IBM LS
  - ▷ Common traits of legacy systems
  - ▷ Implications: similar initiatives

## Recommendation: Integrated Strategy

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- Main considerations:
  - ▷ Existing tools and infrastructure
  - ▷ Past experience and domain knowledge
  - ▷ Applicability and effectiveness of new risk identification techniques
  - ▷ Tailoring for your environment
  
- Specific for Nortel Networks:
  - ▷ Measurement: existing tools/databases
  - ▷ Identification: EMERALD
  - ▷ Characterization: TBDMs (this paper)
  - ▷ Cross validation: both
  - ▷ Follow-up: causal analysis needed
    - but TBDMs can help/guide

## Recommendation: Lifecycle Integration

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- Main considerations:
  - ▷ Process and data availability
  - ▷ Experience/infrastructure/tools/etc.
  - ▷ Different focus, but similar techniques?
  - ▷ Tailoring for your process/product
  
- Lifecycle integration:
  - ▷ Analysis of inspection/other data
  - ▷ Analysis and feedback loop
  - ▷ Our current/future research projects
  - ▷ QA and improvement focus:
    - defect prevention
    - defect detection and removal
    - defect containment

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## Conclusions and Perspectives

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- Problems addressed:
  - ▷ Large telecommunication systems.
  - ▷ Multiple releases, diverse components
  - ▷ Uneven DF distribution (80:20 rule)
  - ▷ Need risk identification and characterization for corrective/remedial actions
  
- Conclusions: an effective strategy
  - ▷ Existing measurement tools/infrastructure
  - ▷ EMERALD for risk identification
  - ▷ TBDMs for risk characterization
  - ▷ TBDMs guided follow-up actions
  
- Future work:
  - ▷ Lifecycle approach to quality
  - ▷ Progression: qualitative  $\Rightarrow$  quantitative