# Software Reliability and Safety CSE 8317 — Fall 2008

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# **SRE.4: Applications & Frontiers**

- Non-traditional applications.
- Data treatment for better results
- Data clustering and DCRMs
- Research issues and improvement

## **Current SRE Assessment**

- Reliability assessment:
  - ▷ IDRMs and SRGMs (TBRMs too)
- Reliability prediction: SRGMs
  - ▶ OP accuracy?
  - ▶ Data appropriate/meaningful?
- Reliability improvement:
  - → TBRMs and other emerging approaches
- More research needed.
  (Why we are working on related topics.)

## **SRE Applications**

- Traditional applications:
  - "medium-reliable" systems.
  - ▶ Telecommunication software/systems:
    - Musa/AT&T/Lucent, and others.
  - Operation systems and system software:
    - DEC/HP/IBM/etc.
  - Commercial software systems:
    - IBM examples in this class.
  - Dother similar applications.
- New application domains:
  - Adapting existing models/techniques.
  - ▶ Data selection/treatment necessary?
  - New models/techniques needed?

## Web Testing and SRE

- SMU Project background:
  - NSF 0204345 and industrial partners
  - research team: Alaeddine, Nguyen, Shahdad, Kallepalli, Li, Ma, Rudraraju, Tian.
  - ▶ key publications: 2 TSE papers
    - (Kallepalli and Tian, 2001)
    - (Tian/Rudraraju/Li, 2004)
- Key activities and results:
  - - unified Markov models
    - details in Tian/SQE book: Ch.10.
  - ▷ Statistical web testing and SRE:
    - (Kallepalli and Tian, 2001)
  - ▶ Web SRE: time/activity measurement:
    - (Tian/Rudraraju/Li, 2004)
  - ▶ Web ODC: (Ma and Tian, 2007)

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# Web Characteristics and QA

- Web applications and QA:
  - ▶ Large, diverse, general population.
  - ▷ Document/information vs. computation.
  - ▷ Diverse usage patterns/environments.
  - ▶ Reliability one of the key concerns.
- Usage environment:
  - ▶ Traditional hardware/software env.

  - ▶ Layered structure.
  - > Failure analysis necessary.
- SRE and statistical testing appropriate.

# Web Failure, QA, and Reliability

- Web failure: inability to deliver information or document required by a web user.
- Infrastructure failure

  - Network failure: down/congested.
  - ▷ Browser failure: software problem.

  - Existing hardware/software reliability
- Information source failure
  - > Individual page problems
  - ▷ Overall reliability: focus here.
- User errors: beyond our control.

# Web Logs and Usage/Reliability

- Access and error logs: Information source for usage modeling and reliability analysis (Tian/Rudraraju/Li, 2004)
- Access log: hits

  - ▶ but not operations using local cache
  - > specific information recorded at server
- Error log: problems

  - ▷ problem type: Table 3
  - > similar info and format

## Error Logs and Reliability Analysis

## Error logs

- Detailed problem information
- ▷ "failures" for reliability analysis
- ▷ In conjunction with other measurements
- > When absent: use response code.

# Reliability analysis

- ▶ Reliability by Nelson model
- ▶ Mean-time-between-failures

$$MTBF = \frac{1}{f} \sum_{i} t_{i}$$

for usage time  $t_i$ 

hd MTBF when  $t_i$  not available

$$\mathsf{MTBF} = \frac{n}{f}$$

Reliability growth using SRGMs (software reliability growth models)

## **Case Studies**

- Site, information sources, and tools

  - Information sources: access/error logs (Apache web server)
  - Analysis tools:
    - FastStat and Perl programs
  - ▶ Manual analysis also
- Cross-validation:
  - ▷ www.kde.org.
  - Different types of web site.
  - ▶ Heavier traffic.

# Case Study: Error Analysis

- General error analysis result:

  - > Types A through K, but two key types:
    - Type A: permission denied
    - Type E: file does not exist
- Further analysis of errors
  - Type A: may or may not be considered as failures

  - Relating to usage information
  - ▶ Reliability analysis

- Error over time: Fig 1.
  - Ups and downs (calendar time)
  - ▷ Impact of traffic/workload
  - Conclusion: proper workload/usage measurement for reliability analysis
- Possible workload/usage measurements:
  - Hits: already done in (Kallepalli and Tian, 2001)

  - $\triangleright$  Sessions ( $\approx$  to data grouping)
- Measurement results: Figs 3-7.

- Overall reliability:
  - Relating failures to usage
    - errors vs workload measurements
  - - errors vs. bytes
- Reliability evaluation results:
  - > Application of Nelson model.

  - $\triangleright$  MTBF = 26.6
  - $\triangleright$  Reliability = 0.962
  - > Results in other units possible
    - but need to be cautious.

- Reliability growth for statistical testing
- Hypothetical situation:
  - □ usage-based testing

  - > studied over 26 days
  - ▷ calculated from error/access logs
- SRGM results:

  - Other results: Table 6.
  - ▶ Purification level  $\rho$  (from SRE.3, and SQE Ch.22)

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- Cross validation using KDE data.
- Overall workload and reliability:

  - Session profile: Fig 12 − two variations(2 hr vs. 15 min)
    - only 2 hrs used for SMU/SEAS
  - → Hourly traffic: Fig 13.
  - ▷ Overall results: Table 7.
  - Similar results (better reliability)
- Reliability growth:
  - ⊳ GO model for errors vs. bytes: Fig. 14
  - $\rho = 86.7\%$  to 88.9%

## Web SRE Summary

- What has been done?
  - Reliability assessment/prediction by analyzing both access and error logs
  - Case study to demonstrate viability and effectiveness
- SRE specific results:
  - Data and modeling from existing sources.
  - Good operational reliability assessment.
  - ▶ Reliability growth potential assessment.
  - $\triangleright$  2 diverse web sites  $\Rightarrow$  generalization.

## **Data-Model Mismatch**

- Data-model mismatch:
  - > Assumption mismatch.
  - Data appropriate?
    - ⇒ data selection and/or treatment
  - ▶ Model appropriate?
    - ⇒ choose alternative model
    - ⇒ develop new model
  - Research community: new models (but often impractical)
  - ▷ Industry: model/data selection/treatment
- Data treatment:

  - Grouping/clustering techniques

#### **Data Treatment**

- Data censoring techniques:
  - - ⇒ censored data reflects usage
  - ► Technique: K.-Y. Cai, IEEE Trans.
    Reliability 46(1):69-75, 1997.
- Data compression:

  - ▷ Basis: coverage
    - less likely to fail if tested
    - coverage as multiplier
  - ► Technique: M.-H. Chen et al, IEEE Trans.
    Reliability 50(2):165-170, 2001.
- Works with individual data points directly.

## **Data Grouping**

- Need for data grouping:
  - Already grouped from applications:
    - hourly/daily/weekly/monthly data
    - data collection practicality
  - ▶ Local fluctuations
  - Data dependencies
- Basis for data grouping:
  - External clock/time (most of the existing work)
  - Model (result) optimization
    - Schneidewind approachTSE 19(11):1095-1104, 11/1995
  - Data clustering

#### **DCRM**

- General information/strategy:

  - DCRM: DCRM1 + DCRM2 data cluster based reliability models
  - Automatic clustering
  - ▷ DCRM1: direct usage
  - DCRM2: use with existing SRGMs (grouped data as input)
- Basic ideas: How?
  - ▷ Clustering of homogeneous runs.
  - Data driven/sensitive partitions.
  - ▶ Method: Tree-based modeling (TBM).

# The Case for Grouping

- Scenario-based vs. random testing:
  - ▷ Parallelism/interleaving in testing.
  - ▶ Randomized workload.
  - ▷ Similar overall picture.
  - $\triangleright \Rightarrow$  Data grouping.
- Defect fixing and run dependencies:
  - > Strong short term dependency.
  - ▶ Lack of long term dependency.
  - $\triangleright \Rightarrow$  Clustering.
- Develop DCRMs

#### **DCRM Construction**

- Clustering/grouping test runs:
  - By similar failure intensity.
  - Computation: Tree-based modeling (TBM) supported by S-PLUS.
- Generic procedure:
  - > Identify period, runs, and time.
  - ⊳ Failure intensity = failure / time,
  - ▷ Simple algebraic mean for segment:

$$\frac{\sum_{j=1}^{n_i} f_{ij}}{n_i} = \frac{f_i}{n_i} = \lambda_i$$

Weighted average for segment:

$$\frac{\sum_{j,l_i < d_j \le u_i} t_j \lambda_j}{\sum_{j,l_i < d_j \le u_i} t_j} = \frac{\sum_{j,l_i < d_j \le u_i} f_j}{\sum_{j,l_i < d_j \le u_i} t_j} = \frac{F_i}{T_i} = \Lambda_i$$

▷ Other: as special cases of above.

# Model Usage and Performance

- Direct usage: DCRM1
  - ▶ Reliability for each segment.
  - > Overall trend assessment.

  - ▶ Prediction: extrapolation.
  - ▷ Risk/anomaly identification.
- DCRM1 performance:
  - $\triangleright$  Goodness-of-fit:  $R^2$ 
    - 304 vs. 6329 for Goel-Okumoto.
  - ▶ Prediction comparison:
    - use training and testing sets.
    - linear extrapolation.
    - good short term results

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## **DCRMs vs Other Models**

## • DCRM1 vs IDRMs:

- > Partition by failure intensity in runs
  - similar to Nelson model.
- ▶ Partition by general failure intensity
  - similar to Brown-Lipow model.

#### DCRM1 vs SRGMs:

- $\triangleright$  Constant  $\lambda$  for given period
  - similar to Jelinski-Moranda model.
- But variable steps in consecutive steps
  - similar to Littlewood-Verrall model.
- $\triangleright$  Non-function form for progression of  $\lambda$ 's
- PFC-SRGMs: used in DCRM2.

#### DCRM2

- DCRM2: SRGMs with grouped data (each segment as a data point)
- Choosing SRGMs for DCRM2:
  - ▷ Only PFC (FC) models usage
  - ▶ NHPP choices:
    - Goel-Okumoto (GO)
    - Musa-Okumoto (MO, log Poisson)
    - Schneidewind and data req.
    - S-shaped as descriptive model
- More about GO and MO choices:

  - ▶ Past experience at IBM
  - Empirical data elsewhere

## **DCRM** Performance

- Product and comparison points:
  - ▶ Products E (and D) from IBM
  - ▷ E: last 8 weeks
    - 7 point comparison for DCRM2
  - DCRM1,GO, MO, DCRM2.GO, DCRM2.MO
- Applicability:
  - ▷ DCRM1 clearly superior
  - Others about equal
- Goodness-of-fit:
  - ▷ DCRM1 clearly superior
  - Others about equal
  - Caution: use more important

#### **DCRM** Performance

- Reliability assessment:
  - DCRM1 not as stable but available early

  - DCRM2 provide tighter bound (more stable also, see prediction)
- Reliability prediction:
  - DCRM1 only for short term (mixed results)
    - only one available early
  - Prediction accuracy tables
  - Direct comparison graphs
- Model stability: DCRM2 better

# **DCRM Summary**

- Easily satisfiable assumptions:
  - Rough operational profiles.
  - ▶ No long term dependencies
    - but short term dependencies
  - > Failure intensity clusters.
- Implementation and applications:
  - ▶ Model construction: S-PLUS.
  - > Practical applications.

  - > Further studies underway.

# Other Data/Models

## Trend analysis:

- Qualitative/visual inspection:
  - curvature (super-/sub-additive)
- Quantitative analysis.
- Avoid meaningless modeling results.

# Other data/models:

- Reliability simulation: especially for mixed h/w-s/w systems
- ▷ Composite models.
- ▶ General models for correlated data: Goceva-Popstojanova and Trivedi, IEEE Trans. Reliability, 49(1):37-48, 3/2000.
- ⊳ etc., Lyu book.

# **Improvement Strategies**

- Traditional models/techniques:
  - Assessment/prediction focus.
  - ▶ Limited used in improvement.
  - ▶ Testing/QA as semi-separate.
- TBRM and extensions:
  - > TBRMs for risk identification.

  - - analyze other (e.g., inspection) data
    - reliability composition

# Other Research Topics

- Linking SRE with metrics/analysis.
  - Musa prescriptive models refinement.
  - ▶ Metrics-SRE: still separate
    - risk (problem-prone) identification
  - Quantitative linkage?
- Reliability composition:
  - ▷ Small scale: Hamlet/Mason/Woit work.
  - ▷ RE to requirement: Smidts work.
  - Work at SMU: OP-mapping, fault-injection, embedded systems.
- Reliability optimization.
  - Other factors: cost, schedule, etc.
  - ▶ Lyu/Rangarajan/van Moorsel work.