Software Reliability and Safety CSE 8317 — Fall 2009

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SRE.3: Reliability Models

- Reliability functions and definitions
- Software Reliability Growth Models
- Combinatorial and Other Models
- Model Assumptions/Limitations/Usage

Reliability Models

- Reliability modeling
 - Reliability-fault relations
 - Exposure assumptions
 - ▶ Lyu book: Chapter 3; Tian/AIC paper
- Time domain SRGMs

 - Stochastic process for failure arrivals
 - Reliability growth due to fault removal
- Combinatorial & other models
 - Reliability-fault relation over input

 - ▶ Input domain (ID) models
 - Cleanroom and coverage-based models

Develop and Use Models

1. Preparation:

- > study failure data and environment
- choose reliability model(s)(reliability expressed as math functions)
- ▷ influence of past experience

2. Modeling (function with parameters):

- > estimate model parameters
- ▷ obtain fitted model

Followup and decision making: (assessment/prediction/control aspects)

Environment and Choice of Models

- Environment and data
 - > Modeling goals under the environment
 - > Environmental constraints:
 - project/process environment
 - data availability/cost
 - Preliminary choice of models
- Model choice: goal driven
 - ▶ Goal: assessment/prediction/control?
 - ▶ Proper definition of reliability
 - time/input/stage/coverage?

 - ▶ Reliability goals as exit criteria
 - Management and improvement

Choice of Models

- Choice based on experience
 - > Previous choices and experience
 - models fitted obs. well?
 - other results: positive/negative?
 - overall feedback from development?
 - ▶ Both local and non-local experience
 - Baseline for comparison
 - > Adaptation and refinement for now
- Other factors
 - ▶ Match model assumptions with reality
 - implications/limitations later
 - > Tools and software support
 - SMERFS, CASRE, etc. (Lyu Book)
 - integration with other tools?

Basic Functions and Definitions

- Some basic functions/definitions:
 - $\triangleright F(t)$: cdf for failure over time
 - $\triangleright f(t)$: pdf, f(t) = F'(t)
 - \triangleright Reliability function R(t) = 1 F(t)

$$R(t) = P(T \ge t) = P(\text{no failure by } t)$$

$$z(t)\Delta t = P\{t < T < t + \Delta t | T > t\}$$

- \triangleright Mean function m(t) in NHPP
- \triangleright Failure rate/intensity, $\lambda(t) = m'(t)$
- > Time domain definition:

$$R = \frac{s}{n} = \frac{n-f}{n} = 1 - \frac{f}{n} = 1 - r$$

- ▷ MTBF, MTTF, etc.
- Details/relations: Tian/SQE book Ch.22.

SRGM Classification

Data used:

- - r.v.: failure interval
- - r.v.: failure count for given interval
- Most widely used (in this class)
- Some models can use both TBF and FC data

Other classifications possible

- > Time measurement:
 - calendar/wall-clock/execution/etc. time
- ▷ Distribution/f-arrival function:
 - Poisson/binomial/etc.
- > Finite vs infinite failures

TBF Models

- Model characteristics
 - > Failure intervals as r.v.
 - T_i : r.v. for the time between (i-1)st and ith failures
 - \triangleright Distribution/density: $F_i(t)$ or $f_i(t)$
 - \triangleright Directly define $z_i(t)$
 - \triangleright Relate $z_i(t)$ to failures/faults
- Defining TBF models
 - \triangleright Sequence of $z_i(t)$ over i
 - ▶ Initial value?
 - ▶ Physical interpretation possible?
 - ▶ Rate (or cumulative) data plotting

TBF1: Jelinski-Moranda

- One of the earliest model using TBF (time-between-failure) measurement
- Failure rate $(z_i \text{ or } \lambda_i)$:
 - > Proportional to defects remaining
 - \triangleright Step function: $z_i = \phi(N (i 1))$
 - $\triangleright z_i$: failure rate for the *i*-th failure
 - > Two model parameters:
 - $-\phi$ constant for failure exposure
 - N constant for total defects
 - \triangleright Plotting z_i 's and reliability growth
- Relation to later models
 - Similar assumptions
 - Do Other failure rate: geometric etc.
 - ▷ Continuous version: Goel-Okumoto etc.

TBF2-3: Schick-Wolverton

- Variations of TBF1 model
- Schick-Wolverton linear model (TBF2):
 - ▶ Proportional to defects remaining & time
 - > Slope function with renewal
 - $\triangleright \lambda_i = \phi(N (i-1))t$
 - ▷ Assumptions/parameters similar to TBF1
- Schick-Wolverton parabolic model (TBF3):
 - ▷ 2nd order (parabolic) time renewal
 - $> \lambda_i = \phi(N (i 1))(at^2 + bt + c)$
 - ▶ Assumptions/parameters similar to TBF2
- ullet Plotting λ_i 's and reliability growth

TBF4: Geometric Models (Moranda)

- Similar to Jelinski-Moranda
- Failure rate
 - > Step function but geometric step sizes
 - $\triangleright \lambda_i = \lambda_0 \phi^{i-1}$
 - $\triangleright \lambda_i$: failure rate for the *i*-th failure
 - > Two model parameters:
 - $-\phi$: step reduction/curvature
 - $-\lambda_0$: initial failure rate
 - Plotting and comparison to JM
- Relation to later models
 - Close relation to Musa-Okumoto model (logarithmic Poisson)
 - Models defect discovery situations
 - Hybrid geometric Poisson

$$\lambda_i = \lambda_0 \phi^{i-1} + c$$

TBF5: Imperfect Debugging

- Goel-Okumoto
- Failure rate

 - > Allow for imperfect debugging
 - $\triangleright \lambda_i = \phi(N p(i-1))$
 - ▷ p: prob(imperfect debugging)
 - Other parameters same (parameter re-interpretation as JM)
- Relation to later models
 - Close relation to Goel-Okumoto NHPP model
 - Models defect removal process

TBF6: Littlewood-Verrall

- Bayesian model
 - $\triangleright t_i$: *i*-th inter-failure interval
 - \triangleright Distribution (pdf) for t_i :

$$f(t_i|\lambda_i) = \lambda_i e^{-\lambda_i t_i}$$

- $\triangleright \lambda_i$: failure rate parameter
- \triangleright Distribution (pdf) for λ_i :

$$f(\lambda_i | \alpha, \psi(i)) = \frac{[\psi(i)]^{\alpha} \lambda_i^{\alpha - 1} e^{-\psi(i)\lambda_i}}{\Gamma(\alpha)}$$

- $\triangleright \psi(i)$: increasing function of i
- $\triangleright \alpha$: constant
- In SMERFS, LV model with $\psi(i)$:

$$\triangleright \psi(i) = \beta_0 + \beta_1 i$$
 , or

$$\triangleright \psi(i) = \beta_0 + \beta_1 i^2$$

FC Models

- Model characteristics
 - \triangleright Failure count N_i as r.v.
 - > Time interval: predefined
 - equal: Schneidewind model
 - different: other models
 - Distribution: failure arrival process
 - Directly define process parameters
 - ▶ NHPP most common
- Defining FC models

 - ▶ Underlying stochastic processes
 - > Physical interpretation
 - ▷ Cumulative (or rate) data plotting

FC1: Goel-Okumoto

- Process assumption: NHPP
 (Non-homogeneous Poisson Process)
- Model definition:
 - \triangleright Probability of n failures in [0, t]:

$$P(N(t) = n) = \frac{m(t)^n}{n!} e^{-m(t)}$$

 $\triangleright m(t)$: mean function

$$m(t) = N(1 - e^{-bt})$$

 $\triangleright \lambda(t) = m'(t)$: failure rate

$$\lambda(t) = Nbe^{-bt}$$

 $\triangleright N$: total estimated failures

 \triangleright b: failure exposure as model curvature

• Data: period failure count (PFC model) (N(t)) is the random variable)

FC Models: Other NHPP

Similar to Goel-Okumoto model

$$P(N(t) = n) = \frac{m(t)^n}{n!} e^{-m(t)}$$

S-shaped SRGM (2 variations)

▷ Allow for slow start

Modified Goel-Okumoto

$$M(t) = N(1 - e^{-bt^c})$$

Similar to modified Jelinski-Moranda

• Logarithmic Poisson (Musa-Okumoto)

$$m(\tau) = \frac{1}{\theta} \log(\lambda_0 \theta \tau + 1)$$

FC Models: Generalized Poisson

- Differences with previous NHPP:
 - > Segmented rather that global NHPP

 - > Sequence follows some function
- Schneidewind & Generalized Poisson:
 - ▶ NHPP overall

$$d_i(t) = \lambda_i(t) = \alpha e^{-\beta i}$$

Generalized Poisson

$$m_i(t) = \phi(N - M_{i-1})g_i(x_1, x_2, \dots, x_i)$$

Can treat many models as special cases of this model

FC Models: Brooks-Motley

- Binomial/Poisson process with
 - $\triangleright n_{ij}$ failures for ith session, jth module
 - \triangleright session length K_{ij} or t_{ij}
 - $\triangleright q_{ij}$, ϕ_{ij} : binomial/Poisson constant
- Binomial: $q_{ij} = 1 (1 q)^{K_{ij}}$

$$P(X = n_{ij}) = \binom{N_{ij}}{n_{ij}} q_{ij}^{n_{ij}} (1 - q_{ij})^{N_{ij} - n_{ij}}$$

• Poisson: $\phi_{ij} = 1 - (1 - \phi)^{t_{ij}}$

$$P(X = n_{ij}) = \frac{\left(N_{ij}\phi_{ij}\right)^{n_{ij}} e^{-N_{ij}\phi_{ij}}}{n_{ij}!}$$

FC Models: Musa

- Variations of Musa models
 - Prescriptive: derived from product/process characteristics
 - Descriptive: fitted, similar to prev. SRGMs
 - Execution time: used in modeling
 - > Calendar time: used in management
 - Conversion between the two times
- Musa models (descriptive):
 - Basic Musa: resembles Jelinski-Moranda
 - (Musa-Okumoto) logarithmic Poisson (a variation of NHPP model)

$$m(\tau) = \frac{1}{\theta} \log(\lambda_0 \theta \tau + 1)$$

> Execution time used in both above

FC Models: Musa

- Practicality of Musa models
 - Software usage: operational profile and execution time
 - Predictions (prescriptive) based on process and product characteristics
 - Practical issues dealt in Musa book
 - ▷ Practicality vs. theoretical focus
- Applications of Musa models
 - ⊳ AT&T projects: 10-20%
 - ▷ Best practice at AT&T (Lyu/HSRE Ch.6)
 - > Adoption in other environments
 - - AT&T's SRE ToolKit
 - training and benchmarking
 - Most publicized success stories

Choice of SRGMs

- Issues discussed before:
 - ⊳ Goal/environment/experience
- Other model choice issues:
 - > Time measurement and model fit.
 - ▷ Single vs. multiple models.
 - ▷ Composite models possible/meaningful?
 - ▷ Existing vs. new models.
 - Assumptions/limitations/applicability.

Choice of SRGMs

- Time measurement and model fit:
 - ▷ experience at AT&T (exec. time!)
 - ▷ IBM experience
 - bad fit ⇒ time appropriate? (compare to: bad fit ⇒ other model)
- Single vs. multiple models:
 - best fitted vs. optimistic (fast rel. growth) vs. pessimistic (slow ..)
 - ▷ band/range instead of single estimate
 - > related: synthesized/composite models
- Existing vs. new models:
 - ▷ simplicity of existing models
 - > validation of new models
 - > caution against ad-hoc new models

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Alternatives to SRGMs

- Reliability: Prob(failure-free operation)
 - ▷ Time: how to measure ⇒ SRGMs
 - ▶ Input: characterize/classify

 - Applicability and limitations
- Alternatives to SRGMs:
 - ▶ Input domain/combinatorial
 - also fault seeding

 - Coverage-based and predictive
 - → TBRMs: tree-based reliability models
 - both time/input info. (SRE.2)

Mills Fault Seeding Model

- Assumptions (BIG!)
 - > Random seeding, same distribution
 - > Same probability for detection
- Seeding/tagging to estimate population
 - $\triangleright n_s$ seeded, x_s captured
 - $\triangleright n_o$ original, x_o captured
 - \triangleright Prob(finding exactly x_s and x_o):

$$P = \frac{\binom{n_o}{x_o} \binom{n_s}{x_s}}{\binom{n_o + n_s}{x_o + x_s}}$$

 \triangleright ML estimate of n_o given by $\hat{n_0}$

$$\widehat{n_0} = \frac{n_s x_o}{x_s}$$

Nelson's Input Domain Model

Nelson Model:

- \triangleright Running for a sample of n inputs.
- \triangleright Randomly selected from set E:

$$E = \{E_i : i = 1, 2, \dots, N\}$$

Sampling probability vector:

$$\{P_i: i=1,2,\ldots,N\}$$

- $\triangleright \{P_i\}$: Operational profile.
- \triangleright Number of failures: f.
- ▷ Estimated reliability = success rate:

$$R = \frac{n-f}{n} = 1 - \frac{f}{n} = 1 - r$$

 $\triangleright r$: failure rate.

Repeated sampling without fixing.

Other Input Domain Models

- Brown-Lipow model:
 - ▷ Explicit input state distribution.
 - hd Known probability for sub-domains E_i
 - $\triangleright f_i$ failures for n_i runs from subdomain E_i

$$R = 1 - \sum_{i=1}^{N} \frac{f_i}{n_i} P(E_i)$$

- Ramamoorthy-Bastani:
 - \triangleright Safety critical systems, $\hat{R}=1$
 - \triangleright Confidence level for \widehat{R}
 - $\triangleright x_i$ specific set of inputs
 - \triangleright P(program correct | correct for x_i 's)

$$P = e^{-\lambda V} \prod_{i=1}^{n-1} \frac{2}{1 + e^{-\lambda x_i}}$$

- $\triangleright \lambda$ source code complexity
- ▶ Recent development by Woit-Parnas

Ho's Input Domain Model

- Step 1: Symbolic execution tree
 - > Execution tree generation
 - \triangleright Path identification T_i
 - \triangleright Path frequency assignment p_i
- Step 2: Path reliability R_i
 - ▷ Estimate vs. bound

 - Ramamoorthy-Bastani model
- Step 3: System reliability for m paths with probability p_i and reliability R_i

$$R = \sum_{i=1}^{m} p_i R_i$$

Cleanroom Reliability Model

- Hybrid model
 - Reliability growth over stages
 - Random sampling within stage
- Factors affecting reliability
 - ▷ Increment testing: reliability change
 - Mixture of untested and tested codes
- Certifying statistical quality

 \triangleright MTTF = MR^c

▶ M: Initial MTTF

▷ R: Effective ratio for change

▷ c: software changes

Coverage and Coverage-Based Models

- Alternative: coverage analysis
 - Defect fixing effect
 - ▷ Infeasibility of exhaustive testing
 - > Pure coverage vs. cov-based models
- Focus on input/internal state coverage:
 - > Function/data/statement coverage.
 - ▶ Path and dependency coverage.
 - Assumption: coverage↑ ⇒ reliability ↑
 (qualitative relation, not quantified)
- Coverage-based modeling:
 - ▷ Analytical: Weyuker etc.
 - ▷ Empirical: Mathur etc.

General Assumptions and Implications

- Times between failures are independent
 - ▶ Implies randomized testing
 - Practical scenarios:
 - defect fixing effect
 - structure/progression in testing
- Immediate defect removal
 - Duplicate defect counting
 - ▶ Related but not duplicate?
 - ▶ Infeasible for in-field defects
- No new fault injected
 - Reliability growth assured
 - ▶ Practical: injection < removal</p>
 - ▶ Related: Decreasing failure rate

Assumptions and Implications

- Relating failure rate to number of faults
 - Variations to the assumption
 - proportionality between the two
 - functional relation between the two
 - time dependent relation
 - Implications of failure detection and detection sequences
- Operational profile
 - Ensures reasonable/meaningful reliability assessments and predictions
 - ▶ Limits applicability
- Time as a basis for failure rate
 - Equivalent time units
 - Requires proper time measurement

Assumptions and Applicability

- General considerations
 - Assumptions for different model types

 - Match them to application environment
 - models necessarily simple
 - impossible perfect match
- Applicability to different processes
 - Waterfall generally assumed
 - Testing phases
 - ▶ UBST (BBT also?): SRGMs and ID

 - ▷ Incremental development: cleanroom
 - Spiral model: iterations
 - Operational phases
 - difference in defect removal
 - data availability

Applicability to Different Phases

- Requirement and specification
 - Reliability goal from customer expectation and feasibility (also affordable?)
 - Operational profile construction
 - Prepare for random testing
- Design and coding

 - Musa's prescriptive model
 - Other existing models not applicable
 - Alternative models may be needed:
 - fault and error based models
 - constructive information (white box)
 - predictive models relating to reliability

Applicability to Different Phases

Unit testing

- White-box deterministic testing
- Applicable: fault seeding, coverage-based, (Musa's prescriptive?)
- Other models not applicable

Integration and system testing

- > FVT, SVT, regression, integration
- > Focus: customer oriented operations
- Less emphasis on coverage
- Main phase for SRGMs
- > FC models more robust
- ▶ Random testing conformance?

Applicability to Different Phases

Acceptance testing

- Gate: accept/release or not (also plan for product support)
- ▶ Basis: snapshot(s) or random sampling
- ▶ Input domain model appropriate

Operational phase:

- Actual operations (post-release)
- ▶ Beta or ECI programs (pre-release)
- Difference in operational environments
- Data availability and treatment
- ▶ Reliability vs. availability
- Defect fix and product refreshing
- Business decisions

Applications and Examples

Overall procedure

- ▷ Generic: preparation/modeling/followup
- Routine procedure once started
- Often periodic activities

Application examples

- Data: telecommunications (Musa)
- Wide applications of Goel-Okumoto, Musa, and other models
- Shuttle: Schneidewind and Keller