CSR: SMALL: Robust Algorithms for an Open Source Software Reliability Tool UMass Lance Fiondella Dartmouth

Abstract

Systems increasingly depend on software

- Mission and life critical
- Must preserve high reliability and availability

Rush to deploy new technologies may result in

- Inadequate reliability testing
- Severe economic damage and loss of life

Recent National Academies report *Reliability Growth:* Enhancing Defense System Reliability (2015) recommends:

• *"Use of reliability growth models to direct contractor"* design and test activities"

Contributions:

- Free and open source tool for users http://sasdlc.org/lab/projects/srt.html
- **Collaborative environment for researchers**
- **Expectation conditional maximization (ECM)** algorithms to ensure stability of convergence

Goals and Objectives

Software Failure and Reliability Assessment Tool (SFRAT)

- Designed for practitioner and research community
- Programmed in R and provides functionality through Shiny graphical user interface
- Reduces the need for knowledge of the underlying statistical techniques
- Can help user quantitatively assess software as part of their data collection and reporting process

Allows users to answer following questions about a software system during test

- 1. Is software ready to release (Has it achieved a specified reliability goal)?
- 2. How much more time and test effort will be required to achieve a specified reliability goal?
- 3. What will be the consequences to a system's operational reliability if not enough testing resources are available?

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Input Data Format 1. Inter-failure (IF) times data: Time between $(i - 1)^{st}$ and i^{th} failure, $t_i = (T_i - T_{i-1})$ 2. Failure times (FT) data: Vector of failure times, $T = < t_i, t_2, ..., t_n >$ **3. Failure count data:** Length of interval and number of failures observed within it, $< \mathbf{T}, \mathbf{K} > = < (t_i, k_1), (t_2, k_2), \dots, (t_n, k_n) >$ Input File format: Excel or CSV FN 30 113 146 227 81 115 342 Output 1. Trend tests (Assess if data exhibits reliability growth) 2. Model rankings (Select for prediction) 3. Visualization • Cumulative failure, time between failure, failure intensity, and reliability growth plots 4. Predictions • Time to achieve reliability Expected number of faults for next t time units Expected time to next k failures **SFRAT Tab view**

	Software Reliability Assessment in R	Select, Analyze, and Filter Data	Set Up and Apply Models	Query Model Results	Evaluate Mod
Select, Analyze, and Subset Failure Data Specify the input file format	Plot	Data and Troad Test Table			/
 Excel (.xlsx) CSV (.csv) Select a failure data file Choose File No file chosen 	Open, analyze, and	l subset file	_ /	/ /	
Please upload an excel file Choose a view of the failure data.		Apply models, plot result			
Cumulative Failures Draw the plot with data points only, lines only, or both? Both Points Lines	~	Detail	ed model queries		
Plot Data or Trend Test? Data Trend test Does data show reliability growth?			Evaluate model	performance	
Laplace Test	~			-	
Subset the failure data by category or data range Select one or more failure categories to retain					
	5				
1 2 3 4	5				

Tabs enable data input, visualization, model application, model query, and goodness of fit assessment.

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Model fits (smooth lines) can be compared with data (red staircase plot). Predictions (beyond dotted black line).

Tab3: Predictions							
Model	Time to achieve 90% reliability for mission of length 4116	Expected # of failures for next 4116 time units	Nth failure	Expected times to next 1 failures			
layed S-shaped	12401.15	0.246856	1	N/A			
inski-Moranda	59915.29	0.856125	1	4869.807			
el-Okumoto	62829.77	0.903615	1	4591.285			
eibull	259865.77	1.725954	1	2353.053			
ometric	1592716.46	1.877473	1	2170.031			

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Tab 4: Model assessment					
	AIC	PSSE			
S-shaped	2075.146	296.34925			
ric	1937.034	84.32708			
umoto	1953.613	23.07129			
Moranda	1950.534	19.60037			
	1938.161	74.94496			

Models with lower AIC (Akaike information criterion) and PSSE (predictive sum of squares error) preferred.

Conclusions and Future Work

 Free and open source tool to promote collaboration • Application architecture enables integration of

models and measures from research literature

• ECM algorithms to improve stability of model fitting

•Future research

• Expand architecture to encompass additional stages of software development lifecycle

• Explore applications to system level assurance of cyber security

• Multi-stage model fitting algorithms to improve speed and stability of convergence

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