Software Reliability Measurement

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Agenda

- What is Software Reliability and Why Do We Care?
- Measuring and Estimating Software Reliability
  - "Classical" Software Reliability Modeling
  - Fault Modeling Prior To Test
- Where Do We Go From Here?

What is Software Reliability?

Software reliability: "The probability that software will not cause the failure of a system for a specified time under specified conditions. The probability is a function of the inputs to and use of the system as well as a function of the existence of faults in the software. The inputs to the system determine whether existing faults, if any, are encountered". [IEEE89]

Why Do We Care?

- Determine whether software can be released
- Predict resources required to bring software to required reliability
- Determine impact of insufficient resources on operation reliability
- Prioritize testing/inspection of modules having highest estimated fault content
- Develop fault-avoidance techniques:
  - Minimize number of faults inserted
  - Prevent insertion of specific types of faults

Measuring and Estimating Software Reliability

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- "Classical" software reliability modeling
  - Statistical models applied during software test can estimate/forecast reliability
    - Inputs:
      - Time between successive failures, or
      - Number of failures per test interval of a given length
    - Outputs:
      - pdf of time to next failure, or
      - pdf of number of failures in next interval
      - Can estimate/forecast reliability, failure intensity, number of failures observed in next **n** intervals, …
  - Over 100 models published since 1971. Better known models include:
    - Musa Basic
    - Musa-Okumoto
    - Schneidewind (used on Space Shuttle software)
    - Littlewood-Verrall
    - Goal-Okumoto (NHPP for interval failure counts)
  - Following slide shows sample input/output for time between successive failures model.
    - Modeling done using CASRE version 3.0. CASRE is available free of charge from the Open Channel Foundation, "http://www.openchannelfoundation.org/projects/CASRE_3.0"

Measuring and Estimating Software Reliability

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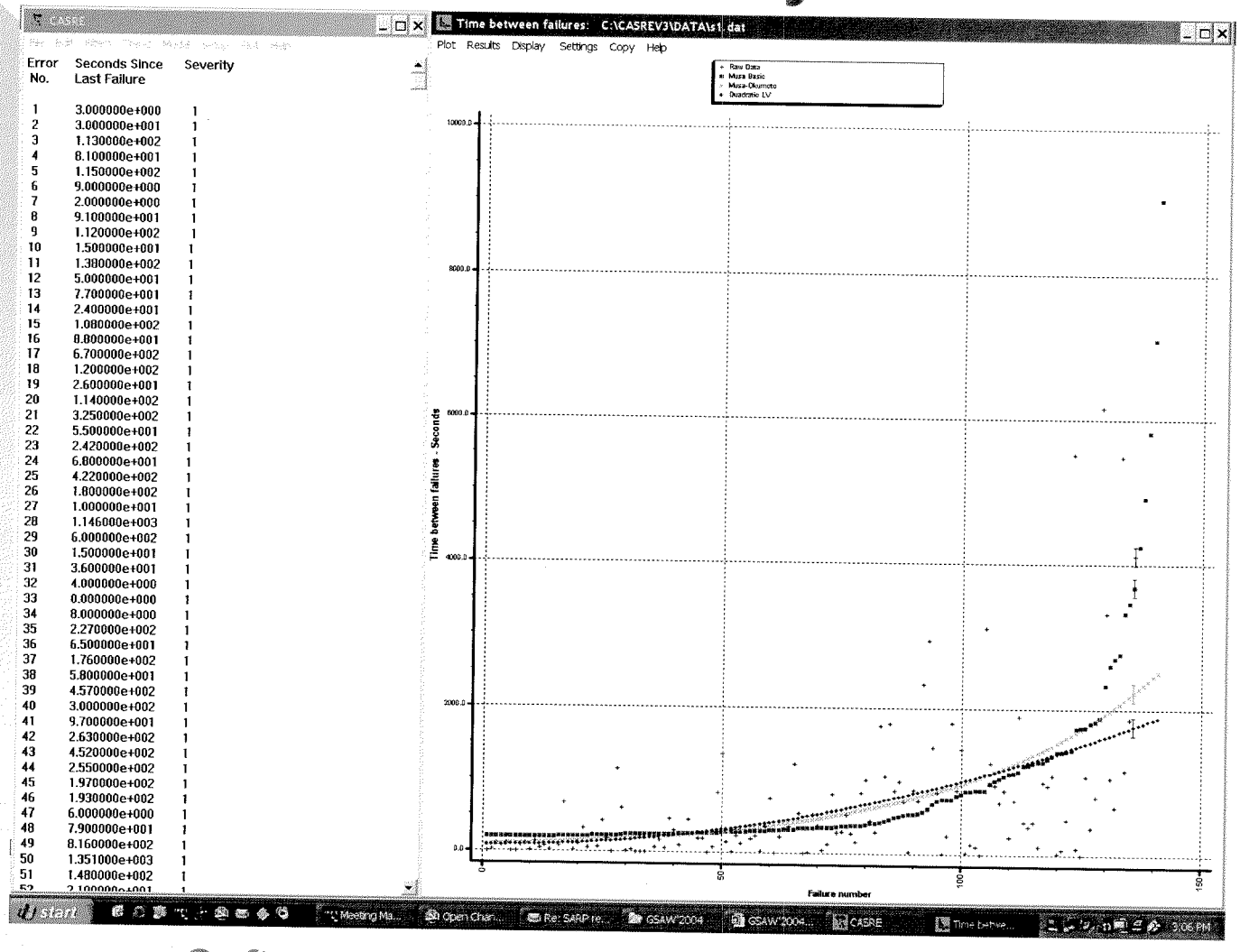

Error No.	Seconds Since Last Failure	Severity
1	3.000000e+000	1
2	3.000000e+001	1
3	1.130000e+002	1
4	8.100000e+001	1
5	1.150000e+002	1
6	9.000000e+000	1
7	2.000000e+000	1
8	9.100000e+001	1
9	1.120000e+002	1
10	1.500000e+001	1
11	1.380000e+002	1
12	5.000000e+001	1
13	7.700000e+001	1
14	2.400000e+001	1
15	1.080000e+002	1
16	8.800000e+001	1
17	6.700000e+002	1
18	1.200000e+002	1
19	2.600000e+001	1
20	1.140000e+002	1
21	3.250000e+002	1
22	5.500000e+001	1
23	2.420000e+002	1
24	6.800000e+001	1
25	4.220000e+002	1
26	1.800000e+002	1
27	1.000000e+001	1
28	1.146000e+003	1
29	6.000000e+002	1
30	1.500000e+001	1
31	3.600000e+001	1
32	4.000000e+000	1
33	0.000000e+000	1
34	8.000000e+000	1
35	2.270000e+002	1
36	6.500000e+001	1
37	1.760000e+002	1
38	5.800000e+001	1
39	4.570000e+002	1
40	3.000000e+002	1
41	9.700000e+001	1
42	2.630000e+002	1
43	4.520000e+002	1
44	2.550000e+002	1
45	1.970000e+002	1
46	1.930000e+002	1
47	6.000000e+000	1
48	7.900000e+001	1
49	8.160000e+002	1
50	1.351000e+003	1
51	1.480000e+002	1

Software reliability models input and output

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Measuring and Estimating Software Reliability Prior to Test

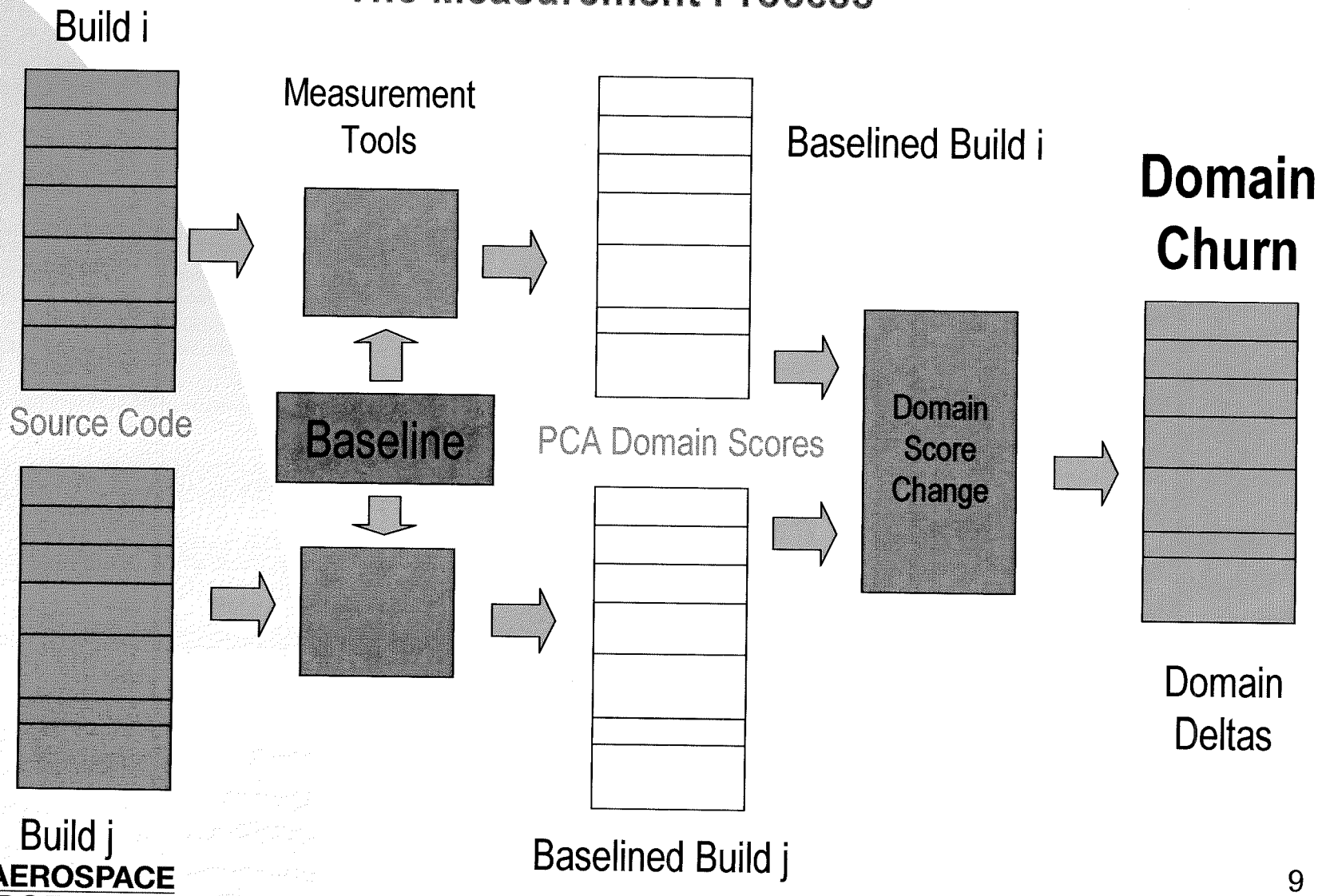
- Identifying Fault-Prone Modules
  - Boolean Discriminant Functions [Schn97]
  - Classification Trees
    - Khoshgoftaar and Allen [Khos01a]
    - Ghokale and Lyu [Gokh97]
  - Logistic Regression [Schn01]

Measuring and Estimating Software Reliability Prior to Test (cont'd)

- Estimating software fault content – representative efforts include:
  - Module-order modeling
  - Neural nets
  - Zero-inflated Poisson regression [Khos01]
  - Measurement of structural evolution [Niko03], [Niko98]
  - ...

Measurement of Structural Evolution

The Measurement Process



Measurement of Structural Evolution

View of Structural Evolution at the System Level

Graph of Code Churn and Code Delta in th...

Darwin
Software Systems International

Darwin Portal

Navigation

Darwin Main

Manager Information

Tester Information

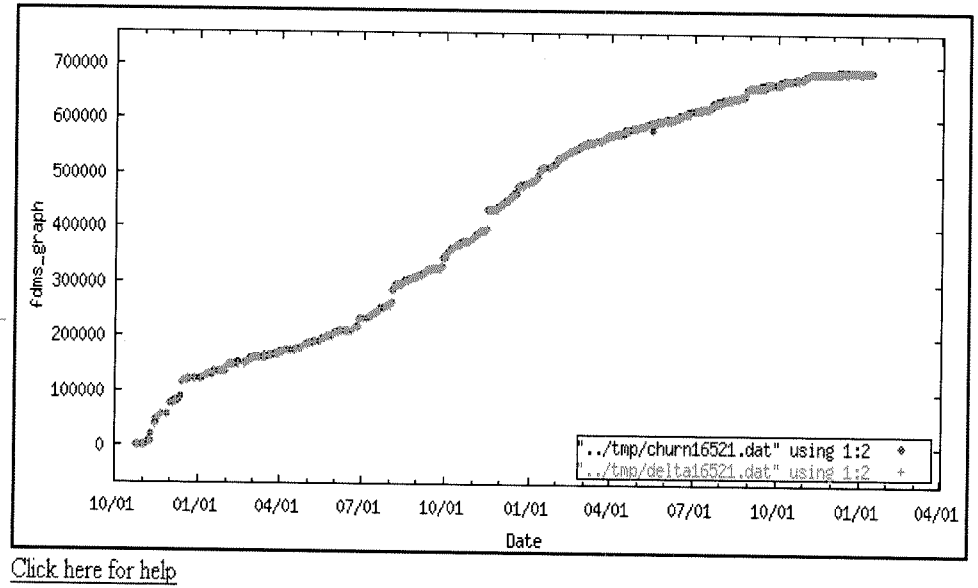
Education

Project Manager

View Available Databases

Feedback

Graph of Code Churn and Code Delta for the project fdms_project_1b.



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documents/Metrics2003cfp.pdf

Measurement of Structural Evolution

View of Structural Evolution at the Module Level

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(Non-zero) Modules for build 2001-11-02 o...

(Non-zero) Modules for build 2001-11-02 of project project_1b, sorted by Churn since baseline.

Modulename	Churn From Baseline
doContent(XML_Parser parser, int startTagLevel, const ENCODING *enc, const char *s, const char *end, const char **nextPtr)	329.523165
doProlog(XML_Parser parser, const ENCODING *enc, const char *s, const char *end, int tok, const char *next, const char **nextPtr)	311.585820
examples()	309.099387
processMeasurementAndPredict(const Mds::Fw::Time::Tmgt::RTEpoch& current, const Mds::Fw::Time::Tmgt::RTEpoch& stop)	289.353008
test2s()	279.141671
TestDiscrete::TestDiscrete()	260.394345
TestIntervallic::TestIntervallic()	256.865234
storeAtts(XML_Parser parser, const ENCODING *enc, const char *attStr, TAG_NAME *tagNamePtr, BINDING **bindingsPtr)	240.951958
GreaseFilterTest(Dispatch& r, const std::string& key, const CGIArgs& args)	240.699311
doTest()	237.752957
PositionEstimateFunctionTest(Dispatch& r, const std::string& key, const CGIArgs& args)	223.900440
DirectedGraph::close()	214.416659
SimpleNormalPositionEstimatorTraits::Thread::updateStateVariables(const Mds::Fw::Time::Tmgt::RTEpoch& start, const Mds::Fw::Time::Tmgt::RTEpoch& stop, const Mds::Fw::Filter::Grease::GreaseBasis& state, const Mds::Fw::Filter::Grease::GreaseBasis& covariance, const Mds::Rd::Mars::Common::SimpleAirDragModel::AirDragModelParameterType& air_drag_model_para, double spacecraft_mass, double avg_engine_thrust)	204.313726
AirDragModelParameterEstimatorTraits::Thread::predictState()	203.781925
ParachuteEstimatorTraits::Thread::predictState()	195.118089
SimpleNormalPositionEstimatorTraits::Thread::changed(const Mds::Fw::Cmp::RefCountComponentInstance monitored_sv, Mds::Fw::Dm::Vhis::ConstItemVectorRef changedItems)	182.312589
PREFIX(prologTok)	175.659645
LengthTest(Dispatch& r, const std::string& key, const CGIArgs& args)	165.230353
vxmain(int ctor, const char* argList)	164.506422
GoalNetTestHarness::createXGoalNet(Dispatch& r, const std::string& key, const CGIArgs& args)	163.479793
vxmain_internal(const char* argList)	162.467209

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Measuring and Estimating Software Reliability Prior to Test (cont'd)

- Architecture-based reliability estimates
  - Architecture-based Reliability Risk Analysis [Ammar2002]
  - Comparison of Architecture-based Software Reliability Models [Gos2001]
  - Cost vs. Reliability Architectural Tradeoffs [Gokh99]

Predicting Software Reliability Prior to Test (cont'd)

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- Architecture-based reliability prediction [Li97]
  - Model software system as a communicating extended finite state machine
  - Map the software architecture into a high-abstraction level design language such as Petri-nets or the Specification and Description Language (SDL)
  - Simulate elements in the system such as work-flow activation, failure occurrence, throughput, and of course the behavior and flows of the architecture from the architectural model
  - Use simulator to analyze the reliability of architectural designs
  - Experiments performed on a Bellcore product
  - Experimental results demonstrated the ability to predict the reliability of a software architecture
  - Verification of the predicted reliability not performed

Where Do We Go From Here?

- Current techniques allow:
  - Estimation/forecasting of software reliability during test and operations
  - Identification of faulty modules during implementation, prior to test
  - Fault content estimates during implementation, prior to test
  - Evaluation of architectures wrt reliability
- Future work
  - Effect of design/requirements changes to reliability
  - Effects of software characteristics, development process characteristics on quality
    - Constructive Quality Model COQUALMO under development at USC Center for Software Engineering – see "http://sunset.usc.edu/research/coqualmo/index.html"
  - Verification of predicted architectural reliability from simulations
    - Aerospace IR&D, "Space Systems Mission Assurance via Software Reliability Monitoring"
  - ...

References and Further Reading

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