

Calibration Interval Adjustment Methods Programmatic Cost Effects

Dr. Howard Castrup, hcastrup@isgmax.com

Greg Cenker, Greg.Cenker@sce.com

Mark Kuster, mkuster@pantex.com

Integrated Sciences Group, Southern California Edison, Pantex Metrology

2012 NCSL International Workshop and Symposium

Learning Objectives

- Understand interval analysis (IA) simulation concepts.
- Judge IA effectiveness by excess test and measurement costs.
- Examine the impacts of realistic calibration scenarios on interval analysis.
- See work that will lead to quantitative method selection guidance in RP-1.

Why Do Interval Analysis?

- Reduce operating costs
- Reduce consequence costs
- Improve product and process quality
- Justify interval changes quantitatively (Z540.3)
- Leverage your software investment
- Gain competitive advantage

Why Simulate?

Complexity

Each method's performance varies with a complex condition set.

● Inventory Parameters

- Group size
- Instrument life
- Reliability behavior
- Correct interval
- Resubmission time
- Interval closure rate

● Programmatic Parameters

- Reliability target
- Initial intervals
- Interval resolution
- Interval restrictions
- False accept rate
- Resubmission windowing

Terminology

Acronyms

IA–Interval Analysis

BOPR–Beginning Of Period Reliability, $1 - p_{FA}$

EOPR–End Of Period Reliability

AOPR–Average Over Period Reliability

Note: All values computed at the instrument level

Method Designators

Ax–Algorithmic, Sx–Statistical, Ex–Experimental

Definition

Relative excess cost metric–The excess cost in the test and measurement budget due to imperfect interval analysis, relative to the optimal budget

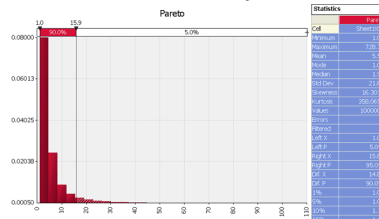
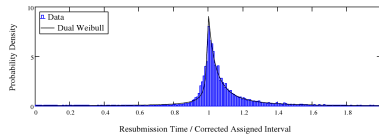
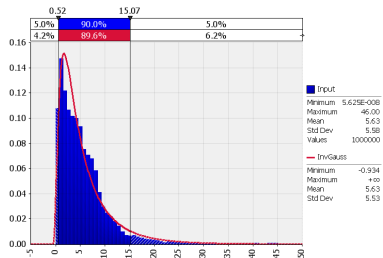
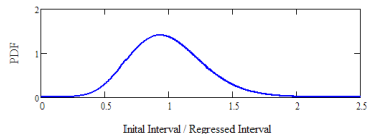
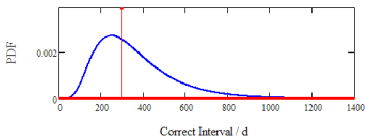
Review: Implemented Methods

- Currently simulating Methods A1, A2, A3, S1, and E1
- Method Parameters
 - A1 $a \in \{0.01, 0.1\}$
 - A2 $v \in \{1, 2\}$
 - A3 $C \in \{0.5, 0.7, 0.9\}$, interpolation & extrapolation choices
 - S1 “out of the box”, always adjust mode only
- Experimental Method E1
 - Greg Cenker constructed it.
 - Fits a two-parameter model to the specified BOPR, observed EOPR, and average resubmission time
 - Linear, exponential, random walk
 - Biases EOPR to speed response with high reliability
 - Configurable action delay
 - Potentially Method A4 in a future RP

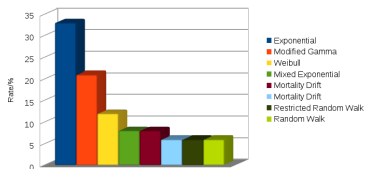
Recent Work: Test Bed Upgrades

- Distributed equipment group size
- Distributed instrument-application life
- Individual instrument lives
- Distributed resubmission time
- Distributed reliability behavior, BOPR and AOPR selection
- Functional failures and unreturned equipment
- Uniform interval rounding and restrictions
- Analyzed almost 13 years' actual calibration interval data
- The paper details the distributions, simulation steps and reliability models.
- Short-lived conservative interval scenario
- The scenario budget contains $\approx 43\%$ excess cost.
- Preliminary data—from 1.3 simulators

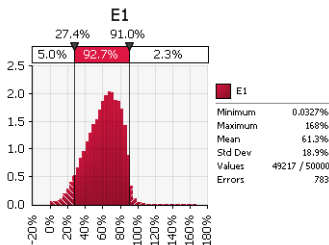
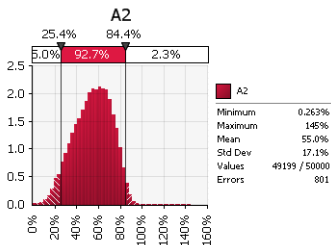
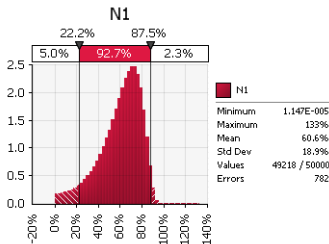
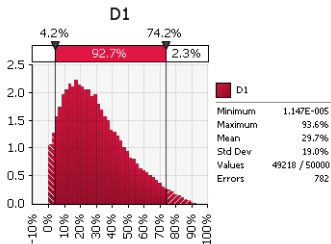
Example Simulation Input Distributions



Reliability Model Occurrence Rates



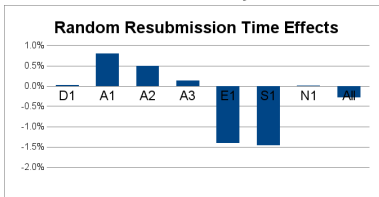
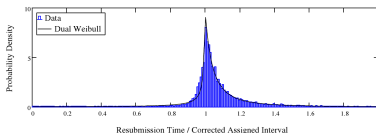
Example Simulation Output Distributions



Programmatic Effects Investigated

- Benchmark: Distributed correct and initial intervals, distributed instrument life, distributed group size
 - Baseline 1: Fixed resubmission time, exponential reliability model, zero inop rate
 - Distributed resubmission time effects
 - Distributed reliability model effects
 - Baseline 2: Distributed resubmission time and model, zero inop rate
 - Default ($\approx 6.5\%$) and 50 % inop rates
 - Baseline 3: Distributed resubmission time and model, default inop rate
 - AOPR and BOPR effects
 - Interval rounding effects
 - Interval restriction effects

Distributed Resubmission Time



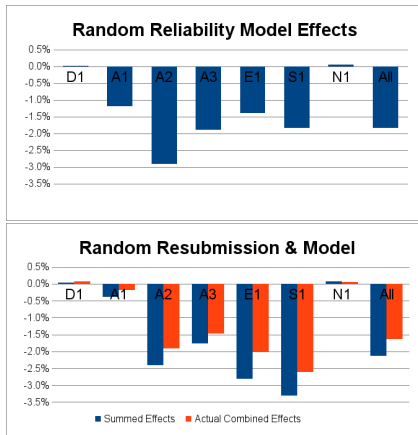
- D1 and N1: only simulation error
- Helps E1 and S1: quicker OOTs?
- Hurts A1 and A2.
- A3 loses less when its aggression increases with reliability target.

- No large overall effect on IA
- Different management strategies for different methods

Distributed Reliability Behavior

Reliability Model	Rate
Exponential	33 %
Modified Gamma	21 %
Weibull	12 %
Mixed Exponential	8 %
Mortality Drift	8 %
Warranty	6 %
Restricted Random Walk	6 %
Random Walk	6 %

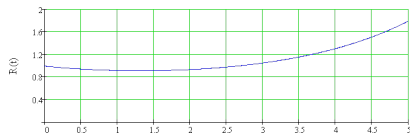
- Dr. Castrup's experience
- Mortality model questionable
- Algorithmic methods more resilient (with more data)



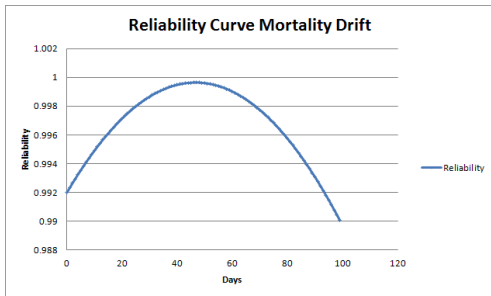
- Non-linear total response

Mortality Drift Model

$$R_{\text{mort}}(t, \theta) = \theta_0 e^{-(\theta_1 t + \theta_2 t^2)}$$

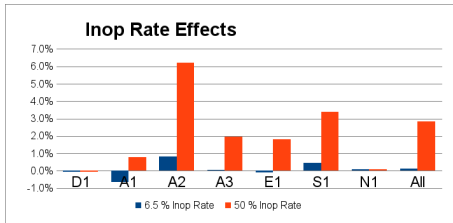


Time/yr since calibration
Mortality Drift Reliability Model



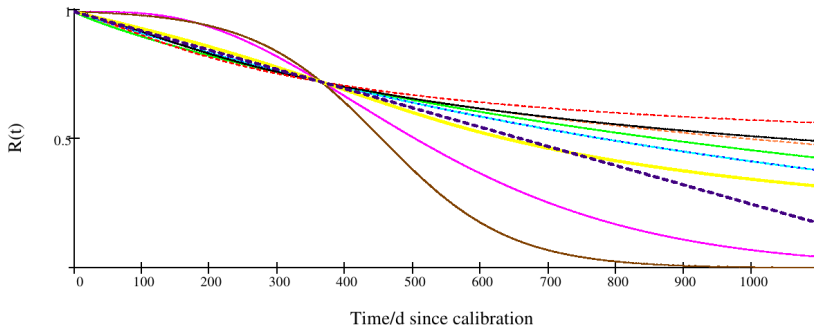
Inop Rate: Wait! No Calibration Data

- $\cong 6.5\%$ of calibrations yielded **no accuracy-related data**
- Inoperative or unreturned equipment: unknown state
- Calibration intervals only control **measurement** reliability



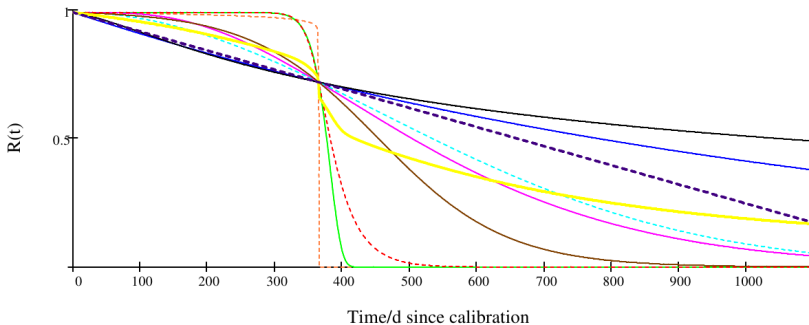
- Hurts both algorithmic and statistical methods
- Small overall effect
- Reduces A1's activity

Mid-Range AOPR Behavior



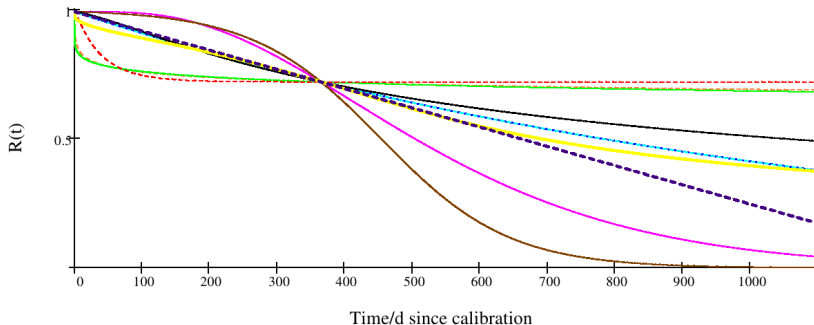
- exponential
- - - mixed exponential
- Weibull
- modified gamma
- - - mortality
- warranty
- random walk
- - - restricted random walk
- weighted average
- - - linear

Extremely High AOPR



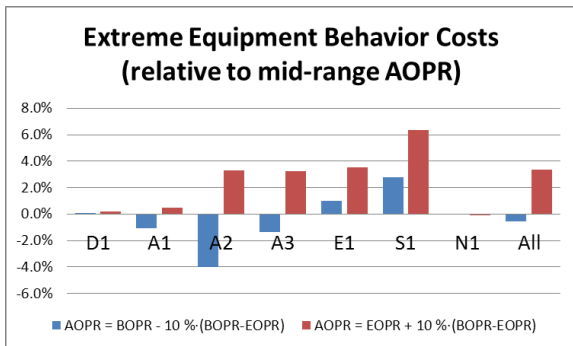
- exponential
- - - mixed exponential
- Weibull
- modified gamma
- · - · mortality
- warranty
- random walk
- · - · restricted random walk
- weighted average
- · - · linear

Extremely Low AOPR



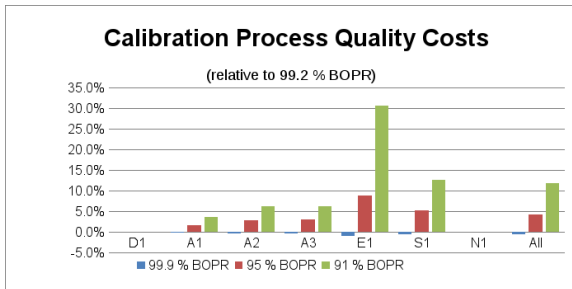
- exponential
- - - mixed exponential
- Weibull
- modified gamma
- - - mortality
- warranty
- random walk
- - - restricted random walk
- weighted average
- - - linear

Extreme Equipment Behavior



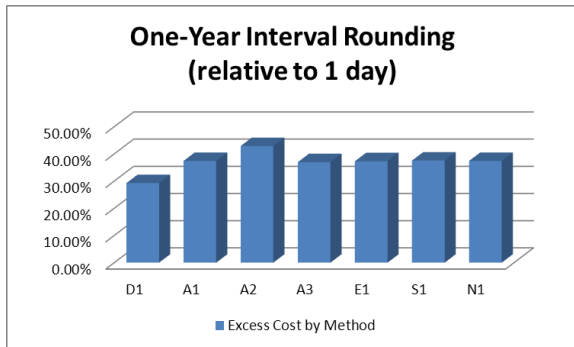
- Extreme behavior counters modeling advantages (S1, E1).
- Low AOPR reduces interval error detection sensitivity.
- A likely opportunity for Method S2

BOPR (False Accept) Effects



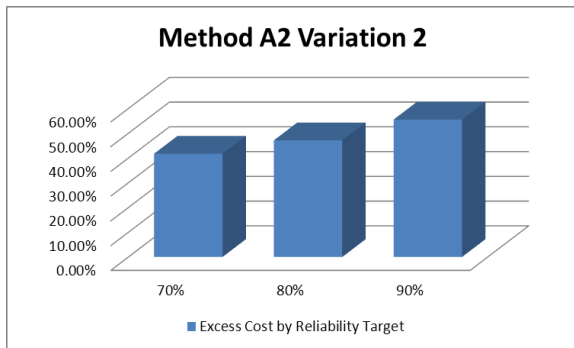
- High sensitivity to BOPR, especially at high reliability targets
- Violates S1's model.
- E1 should compensate but doesn't.
- Even effects algorithmic methods.

Large Interval Units



- One-year interval increments negates all methods.
- Method A2 does worse than nothing.
- Even perfect analysis fares poorly.
- RP-1 does not recommend interval rounding.

Method A2, Variation 2 Detail



- Higher targets allow larger interval adjustments.
- A2 V2 allows interval increases as large as decreases.
- May jump to and converge on the wrong interval increment.
- Other methods do not overcome the large increments.

Interval Restrictions



- Correct intervals ranged from ≈ 40 d to ≈ 3.3 yr.
- The maximum interval restrained Method A2 excesses.
- The minimum interval restrained Method S1 excesses.
- RP-1 does not recommend arbitrary restrictions.

Summary

- Statistical methods handle random resubmission time.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.
- Inoperative equipment has no large impact.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.
- Inoperative equipment has no large impact.
- Extreme equipment behavior counters simple modeling.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.
- Inoperative equipment has no large impact.
- Extreme equipment behavior counters simple modeling.
- Low calibration quality strongly impacts costs, especially for statistical methods.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.
- Inoperative equipment has no large impact.
- Extreme equipment behavior counters simple modeling.
- Low calibration quality strongly impacts costs, especially for statistical methods.
- One-year interval increments negate interval analysis.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.
- Inoperative equipment has no large impact.
- Extreme equipment behavior counters simple modeling.
- Low calibration quality strongly impacts costs, especially for statistical methods.
- One-year interval increments negate interval analysis.
- Interval restrictions **may** help Methods A2 and S1.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.
- Inoperative equipment has no large impact.
- Extreme equipment behavior counters simple modeling.
- Low calibration quality strongly impacts costs, especially for statistical methods.
- One-year interval increments negate interval analysis.
- Interval restrictions **may** help Methods A2 and S1.
- Method E1's linear model performs as well as the exponential and random walk models overall.

Summary

- Statistical methods handle random resubmission time.
- No methods fail under non-exponential behavior.
- Inoperative equipment has no large impact.
- Extreme equipment behavior counters simple modeling. ✓
- Low calibration quality strongly impacts costs, especially for statistical methods. ✓
- One-year interval increments negate interval analysis.
- Interval restrictions **may** help Methods A2 and S1.
- Method E1's linear model performs as well as the exponential and random walk models overall.

Method S2 has an opportunity to regain some ground.

Summary

- Statistical methods handle random resubmission time. ✓
- No methods fail under non-exponential behavior. ✗
- Inoperative equipment has no large impact. ✗
- Extreme equipment behavior counters simple modeling. ✓
- Low calibration quality strongly impacts costs, especially for statistical methods. ✓
- One-year interval increments negate interval analysis. ?
- Interval restrictions **may** help Methods A2 and S1. ?
- Method E1's linear model performs as well as the exponential and random walk models overall. ?

Some results also apply to large data sets. Unexpected gains may evaporate.

Research on the Horizon

- Work remaining
 - Distribute AOPR and BOPR
 - Simulate remaining methods
 - Particularly Method S2
 - S1 (Dwyer, Huang, Jackson) and A3 (Castrup) variants
 - S3? Variables data Methods?
 - Optimize each method over the reliability target space
 - Map optimal methods to equipment management scenarios
- Caveats
 - Probably more subtleties in the data
 - Metric limitations
 - Missing niceties: resubmission windows, equipment aging, etc.

Acknowledgments

Acknowledgments

- B&W Pantex Metrology
- Southern California Edison
- Integrated Sciences Group
- Cherine-Marie Kuster

Thank You for your time! Questions?