Calibration Statistics

- Why We Care About Statistics
- Alternative Statistics to Consider
- Selecting a Best Statistic
- Setting Standards

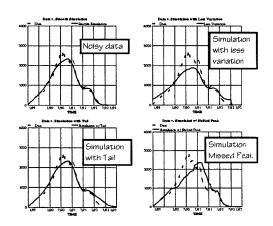
Calibration Statistics Selecting a Statistic and Setting a Standard

1996 System Dynamics Conference

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Why We Care about Statistics

- To enhance the confidence of others
- For comparison against other calibrations
- To assure that minimum standards are met
- For use in "automatic" calibration and sensitivity analysis software



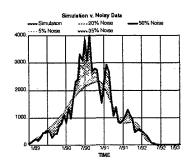
The Ideal Statistic

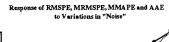
- Provides meaningful results in a variety of circumstances
- · Easily interpreted
- Easy to explain ("intuitive")
- Provides a consistent measure
- Is sensitive to important differences
- Is relatively insensitive to small differences when either the simulation or the data is close to zero

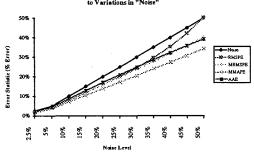
Alternative Statistics

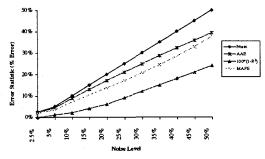
- R²
- Mean Absolute Percent Error (MAPE)
- Root Mean Square Percent Error (RMSPE)
- Modified Mean Absolute Percent Error (MMAPE)
- Modified Root Mean Square Percent Error (MRMSPE)
- Theil Statistics (U)
- Average Absolute Error (AAE)

Handling Noisy Data

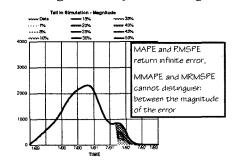




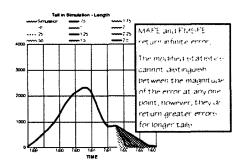




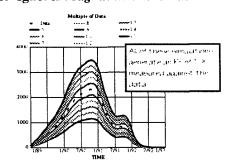
Measuring Sensitivity to Tail Height



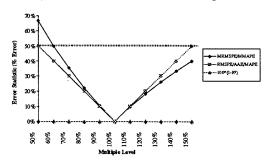
Measuring Sensitivity to Tail Length



R² Ignores Magnitude of the Data



Most Statistics Do Not Produce Symmetrical Results for Multiples



AAE Provides the Best Measure for Life Cycle Models

- Returns symmetric errors for bias
- Gives reasonable errors for increasing magnitude and duration of any "tails"
- Provides predictable response to increasing levels of noise
- Yields similar values to other statistics "the rest of the time"
- Is intuitive and easy to explain

All other measures suffer some fatal flaw

How Good a Fit is Necessary?

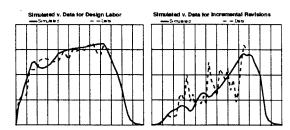
Standard will depend on ...

- The purpose for which the model is constructed
- The amount of noise in the data
- Time and budget available

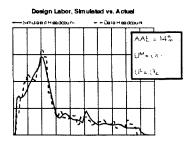
Noise Levels Vary Between Projects



Noise Levels Vary Within a Project

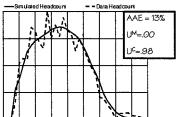


Examples of Good Fit



Examples of Good Fit

Design Labor, Simulated vs. Actual



Smoothed Data May Provide a More Fair Basis for Comparison







Summary

- · Statistics cannot substitute for a careful visual
- . We should continue to compute and report all statistics to allow comparison to other results
- used as the standard
- The Theil components should be computed and monitored to ensure that "bias" is small (<=10%) and most error is due to covariance (>=70%).

inspection and analysis of the fit to historical data

• The "Average Absolute Error" statistic should be

• What fit is "good enough" is inherently subjective.

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