

Robust Uncertainty Quantification for Measurement Problems with Limited Information

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Tathagata Basu, Jochen Einbeck and Matthias C. M. Troffaes
Department of Mathematical Sciences, Durham University, UK
Albstar Forbes
Data Science Group, National Physical Laboratory, London, UK


Introduction

- We investigate the use of imprecise probability for metrology.
- Uncertainty quantification of end-gauge calibration process.
- Limited data.
- Lack of knowledge regarding the source of uncertainty.
- Lack of knowledge of dependencies between measurements.

Model and Objective [2]

Input-output model:

- Uncertain input: $y = (y_1, \dots, y_n)$
- Quantity of interest (output): $z = f(y)$
- Known functional relationship $z = f(y)$



Challenges: The measurements are noisy and limited.
Objective: Obtain a robust estimate of $f(y)$ based on an estimate of μ

Delta Method [3]

Let $X = (X_1, \dots, X_n)$ be an estimator of μ . Assume $X \sim \mathcal{N}(\mu, \Sigma)$ (approximately).

Method:

- Taylor expand f , such that
$$f(X) \approx f(\mu) + \nabla f(\mu)^T (X - \mu)$$
- Compute the covariance matrix of $f(X)$,
$$\text{Cov}(f(X)) \approx \nabla f(\mu)^T \Sigma \nabla f(\mu)$$

Obtain a confidence interval around $f(\mu)$ using the relation:
$$f(X) \approx f(\mu) \pm \sqrt{\lambda} \sqrt{\nabla f(\mu)^T \Sigma \nabla f(\mu)}$$

Limitations:

- f must be differentiable with respect to the input variables.
- f has to be approximately linear around μ for the distributional range of X .
- $f(X)$ may not be Gaussian, when f is highly non-linear.
- We may have to use sample standard deviation instead of Σ and we may have to use $\nabla f(X)$ instead of $\nabla f(\mu)$.

P-box [1]

When a p-box is specified by two cumulative distribution functions \underline{F} and \overline{F} , and compute the set of all cumulative distribution functions bounded by \underline{F} and \overline{F} .

$$\{F \in \mathcal{F} : \underline{F}(x) \leq F(x) \leq \overline{F}(x) \forall x\}$$

- Easy propagation through non-linear operators.
- Relax/revise distributional assumption.
- Able to relax assumptions about dependence.

Example: End gauge calibration [2]

Problem: Estimate length $f(\mu)$ of an end gauge (M) by comparing it with length l of a known standard (S) using the relation:

$$f(\mu) = \frac{l(\alpha_{12} + \alpha_{21} + \alpha_{11} + \alpha_{22})}{2}$$

where, α_{12} and β_{12} (α_{21} and β_{21}) are thermal expansion coefficient and temperature deviation of M (S) and d is the difference between $f(\mu)$ and l .

Linearization gives:

$$f(\mu) \approx l + d - \beta_{12}(\mu - \mu_0) - \beta_{21}(\mu - \mu_0)$$

where, $\Delta \mu = \mu - \mu_0$ and $\Delta l = l - l_0$.

Parameter	single mean	single standard deviation	unit
μ^a	0	0.075-06	cm
α_{12}	50	2.5e-06	cm
α_{21}	11.5e-06	1.35e-06	Celsius
α_{11}	11.5e-06	1.2e-06	Celsius
α_{22}	0	0.5e-06	Celsius
β_{12}	20	0.41	Celsius
β_{21}	20	0.413	Celsius
d	0	0.029	Celsius

Estimates: We compare inferences from delta method and p-box method, under various dependence assumptions.

Method	mean	variance
Δ -method	50	1.15e-11

Independent variables:

Assumption	mean	variance
Gaussian	50.000000410908	(2.8235e-06, 4.4617e-10)
Distribution free	[49.99999, 50.00007]	[0, 0.00034]

Uncorrelated μ :

Assumption	mean	variance
Gaussian	[49.99999, 50.00007]	(2.8235e-06, 4.4617e-10)
Distribution free	[49.99999, 50.00007]	[0, 0.00034]

Uncorrelated μ, β :

Assumption	mean	variance
Gaussian	[49.9998, 50.00012]	(7.8375e-07, 8.3596e-10)
Distribution free	[49.97665, 50.02333]	[0, 0.00044]

Uncorrelated μ, β, d :

Assumption	mean	variance
Gaussian	[49.99978, 50.00021]	(7.8375e-07, 8.3596e-10)
Distribution free	[49.97664, 50.02334]	[0, 0.00044]

Discussion

- We investigate the use of p-boxes to propagate uncertainty in measurement problems.
- We illustrate our approach using an end gauge calibration problem.
- We compare our result with the classical delta method.
- Using p-boxes, we can relax distributional assumptions.

References

[1] Scott Ferson et al. 'Constructing probability boxes and Dempster-Shafer structures'. In: *Intelligence Journal* (manuscript, Not yet accepted for publication (May 2015)). arXiv: 0909.4054.

[2] Evaluation of measurement data - Guide to the expression of uncertainty in measurement. Accessed: 2019-05-24 2016.

[3] And W. van der Vaart and Jan A. Wellner. *Robust Convexity and Empirical Processes: With Applications to Statistics*. Springer New York, 1996. doi: 10.1007/978-1-4937-2545-2.

- Use of p-boxes to propagate uncertainty in metrology.
- Illustration using an end gauge calibration problem.
- Comparison with the classical delta method.

We look forward to seeing you at our poster!