Interval Analysis and Optimization Applied to Parameter Estimation under Uncertainty

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1 Justification

- 2 The Interval Number System
- 3 Operations of Interval Arithmetic
- 4 Limits and Continuity
- 5 Order Relations for Intervals
- 6 Differentiability
- 7 Optimization Problem Formulation KKT Conditions
- 8 Applications
- 9 References

Justification



Why should we use intervals?

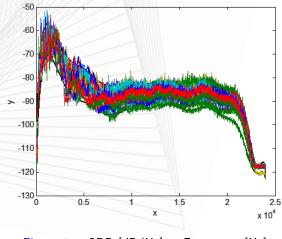


Figure 1 : SPD [dB/Hz] vs Frequency [Hz]

Justification



Why should we use intervals?

Which measurement is the most reliable one?



- ► Which measurement is the most reliable one?
 - If you do not know the value, at least a bounding can be established



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How to estimate the error?



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- How to estimate the error?
 - Uncertainty Dispersion



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- Which measurement is the most reliable one?
 - If you do not know the value, at least a bounding can be established
- How to estimate the error?
 - Uncertainty Dispersion
- Modelling complex dynamics with low information available.
 - Robustness

Definition - Notation

Consider the closed interval denoted by [a, b] which represents the set of real numbers given by

$$[a,b] = \{x \in (R) : a \le x \le b\}$$

Define $I(\mathbb{R}) := \{ [a, b] : a \le b, a, b \in \mathbb{R} \}$ be the set of all closed intervals of \mathbb{R} . We say a interval [a, b] is degenerate if a = b.

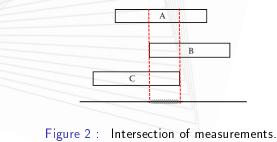
We adopt the *infimum-supremum* notation for intervals:

$$X = [X^{L}, X^{U}] \text{ with } X^{L}, X^{U} \in \mathbb{R}$$
$$X = Y \text{ if } X^{L} = Y^{L} \wedge X^{U} = Y^{U}$$

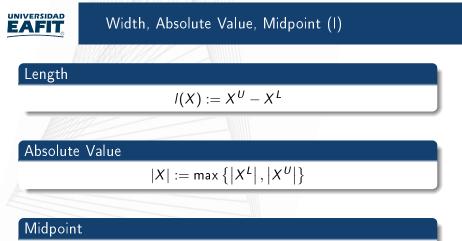


Relevance of Intersection

Intersection plays a key role in interval analysis. If we have two intervals containing a result of interest — regardless of how they were obtained — then the intersection, which may be narrower, also contains the result.



The Interval Number System



$$m(X) := \frac{1}{2}(X^L + X^U)$$

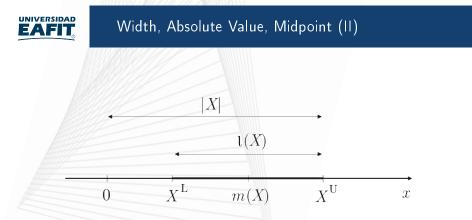


Figure 3 : Width, absolute value, and midpoint of an interval.

Operations of Interval Arithmetic



Definition of Arithmetic Operations

Let $\odot \in \{+, -, \cdot, /\}$ be a binary operation in the real numbers, e.g., addition, subtraction, multiplication and division.

$$X \odot Y := \{x \odot y : x \in X, y \in Y\}$$

In order to simplify notation, the interval [x, x] will be referred as the real number x itself, whenever the context is clear.

Operations of Interval Arithmetic



Endpoint Formulas for the Arithmetic Operations

Let X, $Y \in I(\mathbb{R})$. It can be shown that: 1. $X + Y = [X^{L} + Y^{L}, X^{U} + Y^{U}]$ [Example] 2. $-Y = \begin{bmatrix} -Y^U, -Y^L \end{bmatrix}$ Example 3. $X - Y = X + (-Y) = [X^{L} - Y^{U}, X^{U} - Y^{L}]$ Example 4. $kX = [kX^L, kX^U]$ Example 5. $XY = [\min S, \max S]$, where $S = \{X^{L}Y^{L}, X^{L}Y^{U}, X^{U}Y^{L}, X^{U}Y^{U}\}$ Example 6. $1/Y = [1/Y^U, 1/Y^L]$ Example 7. $X/Y = X \cdot (1/Y)$ Example

Embedding $I(\mathbb{R})$ in a Vector Space

Structure of $I(\mathbb{R})$

Because of this lack of inverse elements under addition, $I(\mathbb{R})$ can not constitute a vector space by itself. However, the work from Radstroem develops the theory of an extension set via equivalence relations in which a commutative semigroup in which the law of cancellation holds, as is indeed true in $I(\mathbb{R})$, can be embedded in a vector space N where the product λA for $\lambda \geq 0$ coincides with the one given on $I(\mathbb{R})$.

Operations of Interval Arithmetic

Hukuhara Difference

Difference

Let
$$X = [X^L, X^U]$$
 and $Y = [Y^L, Y^U]$ be two closed intervals in \mathbb{R} .
If $X^L - Y^L \leq X^U - Y^U$, then the Hukuhara difference $Z = X \ominus Y$
exists and $Z = [Z^L, Z^U] = [X^L - Y^L, X^U - Y^U]$. Example

Note

The usual subtraction and the Hukuhara difference between two intervals need not be the same:

$$[X^L - Y^U, X^U - Y^L] = X - Y \neq X \ominus Y = [X^L - Y^L, X^U - Y^U]$$

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Hausdorff Metric

Let X, $Y \subseteq \mathbb{R}^n$. Then the Hausdorff metric between X and Y is defined by

$$d_{H}(X,Y) = \max\left\{\sup_{x\in X}\inf_{y\in Y}\|x-y\|, \sup_{y\in Y}\inf_{x\in X}\|x-y\|\right\}$$

where $\|\cdot\|$ is a norm in \mathbb{R}^n .

If $X = [X^L, X^U]$ and $Y = [Y^L, Y^U]$ are two closed intervals in \mathbb{R} , it is not hard to see that

$$d_H(X,Y) = \max\left\{|X^L - Y^L|, |X^U - Y^U|\right\}$$

Convergence

Convergence in $I(\mathbb{R})$

Let $\{X_n\}$ and $X \in I(\mathbb{R})$. We say that the sequence of intervals $\{X_n\}$ converges to X, denoted by $\lim_{n\to\infty} X_n = X$, if, for every $\epsilon > 0$, there exists $N \in \mathbb{N}$, such that, for $n \ge N$, we have $d_H(X_n, X) < \epsilon$.

Lemma

$$\lim_{n o \infty} X_n = X$$
 if and only if $X_n^L o X^L \wedge X_n^U o X^U$

Functions in $I(\mathbb{R})$ (I)

Interval-valued Function

The function $f : \mathbb{R}^n \to I(\mathbb{R})$ defined on an Euclidean space \mathbb{R}^n is called an interval-valued function. This function can also be written as $f(\mathbf{x}) = [f^L(\mathbf{x}), f^U(\mathbf{x})]$, where f^L and f^U are real-valued functions defined on \mathbb{R}^n and satisfy $f^L(\mathbf{x}) \leq f^U(\mathbf{x})$ for every $\mathbf{x} \in \mathbb{R}^n$.

Limit of a Function

For $c \in \mathbb{R}^n$ we write $\lim_{x\to c} f(x) = X$ if, for every $\epsilon > 0$, there exists $\delta > 0$ such that, for $||x - c|| < \delta$, we have $d_H(f(x), X) < \epsilon$.

Lemma

Let f be an interval-valued function defined on \mathbb{R}^n and $X = [X^L, X^U]$ be an interval in \mathbb{R} . Then $\lim_{x\to c} f(x) = X$ if and only if $\lim_{x\to c} f^L(x) = X^L$ and $\lim_{x\to c} f^U(x) = X^U$.

Functions in $I(\mathbb{R})$ (II)

Continuity

Let f be an interval-valued function defined on \mathbb{R}^n . We say that f is continuous at $\mathbb{c} \in \mathbb{R}^n$ if

$$\lim_{\mathsf{x}\to \mathsf{c}} f(\mathsf{x}) = f(\mathsf{c})$$

Limits and Continuity

Example

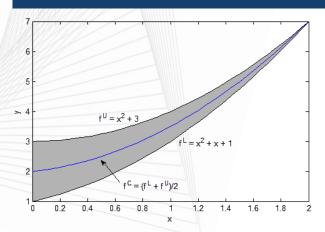


Figure 4 : Graphic representation $f(x) = [x^2 + x + 1, x^2 + 3]$.



Order Relations

Let $X = [X^L, X^U]$ and $Y = [Y^L, Y^U] \in I(\mathbb{R})$. It is possible to express X as a function of its center and width, as $X = \langle m(X), w(X) \rangle$.

Order Relations

$$X \preceq_{LU} Y$$
 if and only if $X^L \leq Y^L$ and $X^U \leq Y^U$

 $X \preceq_{CW} Y$ if and only if $m(X) \leq m(Y)$ and $w(X) \leq w(Y)$

 $X \preceq_{UC} Y$ if and only if $X^U \leq Y^U$ and $m(X) \leq m(Y)$

Order Relations for Intervals

Order Relations

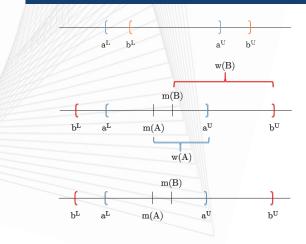


Figure 5 : a) \leq_{LU} b) \leq_{CW} c) \leq_{UC}



Weak Differentiability

Weak Differentiability

Let X be an open set in \mathbb{R} . An interval-valued function $f : X \to I(\mathbb{R})$ with $f(x) = [f^L(x), f^U(x)]$ is called *weakly differentiable* at x_0 if the real valued functions f^L and f^U are differentiable at x_0 (in the usual sense).

H-Differentiability (I)

Derivative

Let X be an open set in \mathbb{R} . We say $f : X \to I(\mathbb{R})$ is *H*-differentiable (strongly differentiable) at x_0 if there exists $A(x_0) \in R(\mathbb{R})$ such that

$$\lim_{h\to 0^+} \frac{f(x_0+h)\oplus f(x_0)}{h} \text{ and } \lim_{h\to 0^+} \frac{f(x_0)\oplus f(x_0-h)}{h}$$

both exist and are equal at $A(x_0)$. Then $A(x_0)$ is the *H*-derivative of f at x_0 .

H-Differentiability (II)

Theorem

Let X be an open set in \mathbb{R} and $f : X \to I(\mathbb{R})$ an interval-valued function defined on X. Suppose that f is weakly differentiable at x_0 with derivatives $(f^L)'(x_0) = \hat{A}^L(x_0)$ and $(f^U)'(x_0) = \hat{A}^U(x_0)$. 1. If $f^L(x_0 + h) - f^L(x_0) \leq f^U(x_0 + h) - f^U(x_0)$ and $f^L(x_0) - f^L(x_0 - h) \leq f^U(x_0) - f^U(x_0 - h)$ for every h > 0, then f is H-differentiable at x_0 with H-derivative $A(x_0) = [\hat{A}^L(x_0), \hat{A}^U(x_0)]$.

2. If $\hat{A^{U}}(x_0) > \hat{A^{L}}(x_0)$, then f is H-nondifferentiable at x_0 .

Optimization Problem Formulation - KKT Conditions UNIVERSIDAD **Optimization Problems** AFIT Problem (RVOP) min $f(\mathbf{x}) = f(x_1, ..., x_n)$ subject to $g_i(x) \leq 0$

Problem (IVOP)

min
$$f(\mathbf{x}) = [f^L(x_1, ..., x_n), f^U(x_1, ..., x_n)] = [f^L(\mathbf{x}), f^U(\mathbf{x})]$$

subject to $g_i(\mathbf{x}) \le 0$

Optimization Problem Formulation - KKT Conditions

Solution Types

Type-I

Let \mathbb{x}^* be a feasible solution, i.e., $\mathbb{x}^* \in X$. We say that \mathbb{x}^* is a *type-I solution* of problem (IVOP) if there exists no $\overline{\mathbb{x}} \in X$ such that $f(\overline{\mathbb{x}}) \prec_{LU} f(\mathbb{x}^*)$.

Type-II

Let \mathbb{x}^* be a feasible solution, i.e., $\mathbb{x}^* \in X$. We say that \mathbb{x}^* is a *type-II solution* of problem (IVOP) if there exists no $\overline{\mathbb{x}} \in X$ such that $f(\overline{\mathbb{x}}) \prec_{LU} f(\mathbb{x}^*)$ or $f(\overline{\mathbb{x}}) \prec_{CW} f(\mathbb{x}^*)$.

KKT Conditions

Theorem (Wu [2])

Assume that the constraint functions $g_i : \mathbb{R}^n \to \mathbb{R}$ are convex on \mathbb{R}^n for i = 1, ..., m. Let $X = \{ \mathbb{x} \in \mathbb{R}^n : g_i(\mathbb{x}) \leq 0, i = 1, ..., m \}$ be a feasible set and a point $\mathbb{x}^* \in X$. Suppose that the **interval-valued** objective function $f : \mathbb{R}^n \to I(\mathbb{R})$ is LU-convex and **weakly** continuously differentiable at $\mathbb{x}^* \in \mathbb{R}^n$. If there exist (Lagrange) multipliers $0 < \lambda^L, \lambda^U \in R$ and $0 \leq \mu_i \in \mathbb{R}, i = 1, ..., m$, such that

1.
$$\lambda^L \nabla f^L(\mathbf{x}^*) + \lambda^U \nabla f^U(\mathbf{x}^*) + \sum_{i=1}^m \mu_i \nabla g_i(\mathbf{x}^*) = 0$$

2.
$$\mu_i g_i(x^*) = 0$$
 for all $i = 1, ..., m_i$

then x^* is a **type-I and type-II**, i.e. optimal under the selected order relation, solution of problem (IVOP).

Interval-Valued Polynomial

Interval-Valued Polynomial

Let $c_i = [c_i^L, c_i^U] \in I(\mathbb{R})$ for $i \in \mathbb{N}$. We say p(x) is an interval-valued polynomial if it can be expressed in the form

$$p(x) = \sum_{i=0}^{n} c_i \cdot x^i = \sum_{i=0}^{n} [c_i^L, c_i^U] \cdot x^i$$

Applications

Matrix Representation

Vandermonde Matrix

Let
$$c_i = [c_i^L, c_i^U] \in I(\mathbb{R})$$
 for $i \in \{1, ..., n\}$.

$$\begin{bmatrix} y_1 \\ \vdots \\ y_m \end{bmatrix} = \begin{bmatrix} 1 & x_1 & x_1^2 & \cdots & x_1^n \\ 1 & x_2 & x_2^2 & \cdots & x_2^n \\ \vdots & \vdots & \vdots & \ddots & \vdots \\ 1 & x_m & x_m^2 & \cdots & x_m^n \end{bmatrix} \begin{bmatrix} c_0 \\ \vdots \\ c_n \end{bmatrix} + \begin{bmatrix} \varepsilon_0 \\ \vdots \\ \varepsilon_n \end{bmatrix}$$
$$\mathbb{Y} = \mathbb{V}\mathbb{C} + \mathbb{E}$$

In this case, $\mathbb V$ is called a Vandermonde matrix.

Applications

Example

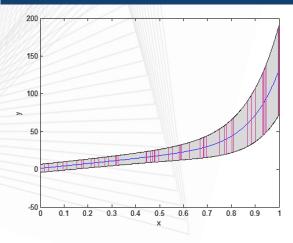


Figure 6 : Interval-valued polynomial graphic.

What do we look for?

In a nutshell

Find a parameter configuration that reduces at most as possible the discrepancies between the observed data and the information provided by the model proposed.

 $\min \sum_{i=1}^{m} [m(y_i) - m(\widehat{y_i})]^2 \qquad \min \sum_{i=1}^{m} d_H(y_i, \widehat{y_i})$ $\ell_2 \text{ Norm - Least Squares} \qquad \ell_1 \text{ Norm - Least Absolute Values}$

Applications



ℓ_2 Norm - Least Squares Estimation

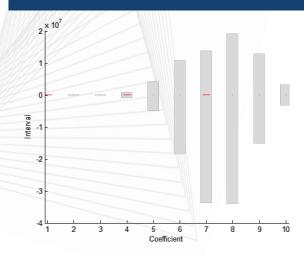


Figure 7 : Parameter estimation result using OLS.

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Applications

ℓ_1 Norm - Heuristic

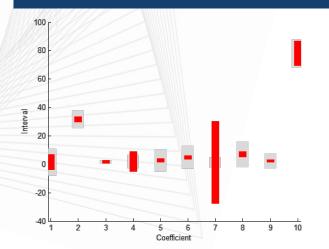


Figure 8 : Parameter estimation result using Differential Evolution.



ℓ_1 Norm - CVX

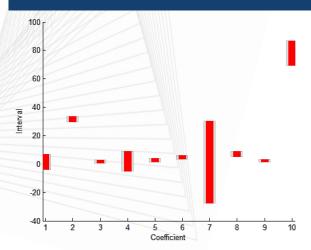


Figure 9 : Parameter estimation result using CVX .



Chaotic Behaviour

Weierstrass Function

In order to evaluate the feasibility of an estimations of the parameters of a model using real data, the used techniques were tested using data sampled from a Weierstrass function, which is an example of a pathological real-valued function on the real line, given by

$$f(x) = \sum_{n=0}^{\infty} a^n \cos\left(b^n \pi x\right)$$

Chaotic Behaviour

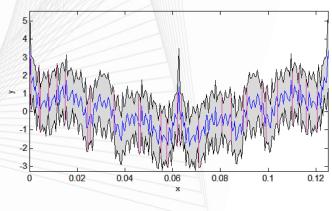


Figure 10 : Weierstrass Function Model.



Experimental Data

Spectral Power Density Measurements

Using hydrophones, measurements of the spectral power density of the sound signals generated by vessels were performed in order to develop a characterization of such crafts. In total 36 measurements were performed, however 12 of those were discarded due to factors that generated changes in behaviour of the spectrum, for example, changes in the speed of the boat and its engines.

Experiment Design

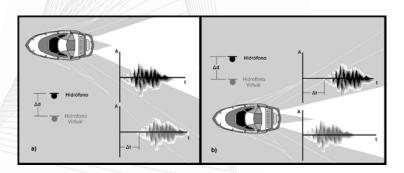


Figure 11 : Experiment designed for the sampling process.

Experimental Data

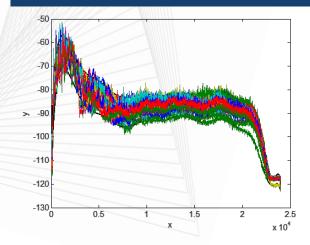


Figure 12 : Measurements - SPD [dB/Hz] vs Frequency [Hz].



Mathematical Model

Fourier Series

In order to describe this behaviour a Fourier series model was proposed. A Fourier series is a way to represent a wave-like function as the sum of simple sine waves, decomposing the signal into the sum of a (possibly infinite) set of simple oscillating functions, namely sines and cosines, as follows:

$$f(x) = a_0 + \sum_{i=1}^n a_i \cos(iwx) + b_i \sin(iwx)$$

Upper - Lower Bounding

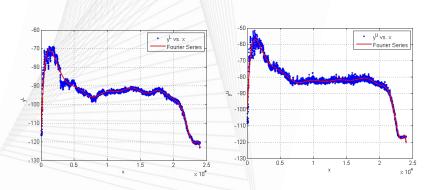


Figure 13 : Fitted lower bound. Figure 14 : Fitted upper bound.

Upper - Lower Bounding

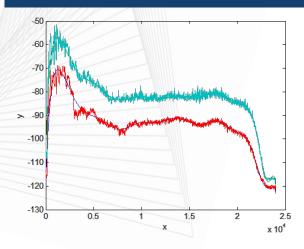


Figure 15 : Fitted model bounds.



Mathematical Model

Fourier Series

Using these estimations an interval-valued function was proposed to enclose the volatility of the measurements using Fourier series to describe the lower and upper functions, i.e. $f : \mathbb{R} \to I(\mathbb{R})$, given by $f(x) = [f^{L}(x), f^{U}(x)]$, where the bounding functions can be expressed by:

$$f^{L}(x) = a_{0}^{L} + \sum_{i=1}^{n} a_{i}^{L} \cos\left(iw^{L}x\right) + b_{i}^{L} \sin\left(iw^{L}x\right)$$

$$f^{U}(x) = a_0^{U} + \sum_{i=1}^{n} a_i^{U} \cos\left(iw^{U}x\right) + b_i^{U} \sin\left(iw^{U}x\right)$$

Estimated Model

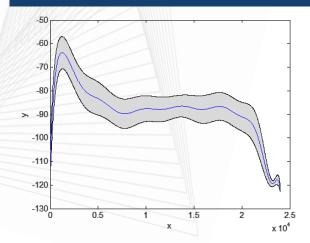


Figure 16 : Interval-valued plot of the estimated Fourier series model.

Modeled Behaviour

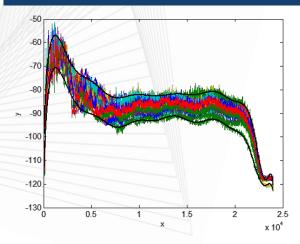


Figure 17 : Real data vs Model output.



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- 3 A. Tarantola, *Inverse problem theory and methods for model parameter estimation*. Philadelphia, PA: Society for Industrial and Applied Mathematics, 2005.



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- 5 Michael Grant and Stephen Boyd. CVX: Matlab software for disciplined convex programming, version 2.0 beta. http://cvxr.com/cvx, September 2013.

Addition - Example

Addition

$$X = [1, 2] \qquad Y = [-4, 5]$$
$$X + Y = [1, 2] + [-4, 5] = [1 + (-4), 2 + 5] = [-3, 7]$$

Negative - Example

Negative

$$X = [-5, 2]$$

 $-X = [-2, -(-5)] = [-2, 5]$

Substraction - Example

Substraction

$$X = [-5, 2]$$
 $Y = [-1, 9]$
 $X + (-Y) = [-5, 2] + [-9, 1] = [-14, 3]$



Scalar Multiplication - Example

Scalar Multiplication (I)

$$X = [-5, 2] \qquad k = 3$$
$$3X = [-5 \cdot 3, 2 \cdot 3] = [-15, 6]$$

Back to Operations

Scalar Multiplication (II)

$$X = [-5, 2] \qquad k = -8$$
$$-8X = [2 \cdot -8, -5 \cdot -8] = [-16, 40]$$



Product - Example

Product

$$X = [-5,2] \qquad Y = [-1,9]$$

$$S = \{(-5)(-1), (-5)(9), (2)(-1), (2)(9)\}$$

$$XY = [\min S, \max S] = [-45, 18]$$





Multiplicative Inverse - Example

Multiplicative Inverse (I)

$$X = [2, 8] \rightarrow \frac{1}{X} = \left[\frac{1}{8}, \frac{1}{2}\right]$$

Back to Operations

Multiplicative Inverse (II)

$$X = [-1, 5]$$
$$\frac{1}{X} = \left\{\frac{1}{x} : x \in X\right\} = (-\infty, \infty)$$

Division - Example

Division

$$X = [-5, 2] \qquad Y = [3, 7]$$
$$\frac{1}{Y} = \left[\frac{1}{7}, \frac{1}{3}\right]$$
$$S = \left\{ (-5) \left(\frac{1}{7}\right), (-5) \left(\frac{1}{3}\right), (2) \left(\frac{1}{7}\right), (2) \left(\frac{1}{3}\right) \right\}$$
$$\frac{X}{Y} = [\min S, \max S] = \left[-\frac{5}{3}, \frac{2}{3}\right]$$



Hukuhara Difference - Example

Hukuhara Difference

$$X = [-5, 2] \qquad Y = [-1, 3]$$
$$X \ominus Y = [-5, 2] \ominus [-1, 3] = [-5 - (-1), 2 - 3]$$
$$X \ominus Y = [-4, -1]$$

Back to H-Difference

Additive Inverses - Example

Additive Inverses

$$X = [-5, 2]$$

Usual Difference

$$X - X = [-5, 2] - [-5, 2] = [-5, 2] + [-2, 5]$$

 $X - X = [-7, 7] = 7[-1, 1] \ni [0, 0]$

Hukuhara Difference

$$X \ominus X = [-5, 2] \ominus [-5, 2] = [-5 - (-5), 2 - 2] = [0, 0]$$

Back to main

Multiplicative Inverses - Example

Multiplicative Inverses

$$X = [3,7]$$

$$\frac{1}{X} = \begin{bmatrix} \frac{1}{7}, \frac{1}{3} \end{bmatrix}$$

$$S = \left\{ (3) \left(\frac{1}{7} \right), (3) \left(\frac{1}{3} \right), (7) \left(\frac{1}{7} \right), (7) \left(\frac{1}{3} \right) \right\}$$

$$\frac{X}{X} = [\min S, \max S] = \begin{bmatrix} \frac{3}{7}, \frac{7}{3} \end{bmatrix} \ni [1,1]$$

Back to main