

# Solution of Tropical Best Approximation Problems

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# Introduction

- ▶ We consider an approximation problem of an unknown function  $y = f(x)$  given a set of samples  $(x_i, y_i)$  of function input/output
- ▶ Let  $F(x; \theta)$  be an approximating function that depends on the vector  $\theta$  of unknown parameters that are to be determined
- ▶ The purpose is to find a best minimax approximate solution, which is defined in the sense of a distance function  $d$  as follows:

$$\theta_* = \arg \min_{\theta} \max_i d(F(x_i; \theta), y_i)$$

- ▶ We formulate the problem in the framework of a tropical semifield (a semiring with idempotent addition and invertible multiplication)
- ▶ As approximating functions, we use tropical Puiseux polynomials (which allow rational exponents) and tropical rational functions
- ▶ We apply the results to the best Chebyshev approximation of real functions with piecewise linear functions taken as approximants

# Idempotent Algebra: Idempotent Semifield

## Idempotent Semifield

- ▶ **Idempotent semifield**: the algebraic system  $\langle \mathbb{X}, 0, 1, \oplus, \otimes \rangle$
- ▶ The carrier set  $\mathbb{X}$  has neutral elements, **zero**  $0$  and **identity**  $1$
- ▶ The binary operations  $\oplus$  and  $\otimes$  are **associative and commutative**
- ▶ Multiplication  $\otimes$  **distributes** over addition
- ▶ Addition  $\oplus$  is **idempotent**:  $x \oplus x = x$  for all  $x \in \mathbb{X}$
- ▶ Multiplication  $\otimes$  is **invertible**: for each nonzero  $x \in \mathbb{X}$ , there exists an inverse  $x^{-1} \in \mathbb{X}$  such that  $x \otimes x^{-1} = 1$
- ▶ **Linear order**: the order  $x \leq y \iff x \oplus y = y$  is a total order
- ▶ **Algebraic completeness**: the equation  $x^p = a$  is solvable for any  $a \in \mathbb{X}$  and integer  $p$  (there exist powers with rational exponents)
- ▶ **Notational convention**: the multiplication sign  $\otimes$  is omitted

## Semifield $\mathbb{R}_{\max,+}$ (Max-Plus Algebra)

- ▶ **Max-Plus algebra:**  $\mathbb{R}_{\max,+} = \langle \mathbb{R} \cup \{-\infty\}, -\infty, 0, \max, + \rangle$
- ▶ **Carrier set:**  $\mathbb{X} = \mathbb{R} \cup \{-\infty\}$ ; **zero and one:**  $0 = -\infty$ ,  $1 = 0$
- ▶ **Binary operations:**  $\oplus = \max$  and  $\otimes = +$
- ▶ **Idempotent addition:**  $x \oplus x = x$  for each  $x \in \mathbb{X}$  ( $= \max(x, x)$ )
- ▶ **Multiplicative inverse:** there exists  $x^{-1}$  for each  $x \in \mathbb{R}$  ( $= -x$ )
- ▶ **Power notation:**  $x^y$  is well-defined for any  $x, y \in \mathbb{R}$  ( $= x \times y$ )
- ▶ More real idempotent semifields:

$$\mathbb{R}_{\min,+} = \langle \mathbb{R} \cup \{+\infty\}, +\infty, 0, \min, + \rangle,$$

$$\mathbb{R}_{\max} = \langle \mathbb{R}_+ \cup \{0\}, 0, 1, \max, \times \rangle,$$

$$\mathbb{R}_{\min} = \langle \mathbb{R}_+ \cup \{+\infty\}, +\infty, 1, \min, \times \rangle,$$

where  $\mathbb{R}_+ = \{x \in \mathbb{R} \mid x > 0\}$

# Algebra of Matrices and Vectors

- ▶ **Matrix operations** follow the standard entrywise rules, where addition and multiplication are replaced by  $\oplus$  and  $\otimes$
- ▶ A matrix without zero rows and columns is called **regular**
- ▶ A matrix of a single column (row) is a column (row) vector
- ▶ If a vector has no zero elements, it is called **regular**
- ▶ For a column vector  $x = (x_j)$ , its **conjugate** is a row vector  $x^- = (x_j^-)$  with  $x_j^- = x_j^{-1}$  if  $x_j \neq 0$ , and  $x_j^- = 0$  otherwise
- ▶ For regular vectors  $x = (x_j)$  and  $y = (y_j)$ , a **metric** is given by

$$d(x, y) = \bigoplus_j \left( x_j y_j^{-1} \oplus x_j^{-1} y_j \right) = y^- x \oplus x^- y.$$

- ▶ In the context of  $\mathbb{R}_{\max,+}$ , this metric is the **Chebyshev metric**

$$d_\infty(x, y) = \max_j |x_j - y_j| = \max_j \max(x_j - y_j, y_j - x_j)$$

# Polynomials and Rational Functions

## Tropical Puiseux Polynomials

- ▶ A **tropical Puiseux polynomial** of  $n$  monomials is given by

$$P(x) = \bigoplus_{j=1}^n \theta_j x^{p_j}, \quad x \neq \mathbb{0},$$

where  $p_j \in \mathbb{Q}$  are exponents and  $\theta_j \in \mathbb{X}$ ,  $\theta_j \neq \mathbb{0}$ , are coefficients

- ▶ When defined in the context of  $\mathbb{R}_{\max,+}$  (max-plus algebra), a polynomial is represented in terms of the usual operations as

$$P(x) = \max_{1 \leq j \leq n} (p_j x + \theta_j),$$

and therefore specifies a **piecewise-linear convex function** on  $\mathbb{R}$

## Tropical Rational Functions

- ▶ A **tropical rational function** is defined by two polynomials as

$$R(x) = \frac{P(x)}{Q(x)}, \quad P(x) = \bigoplus_{j=1}^n \theta_j x^{p_j}, \quad Q(x) = \bigoplus_{k=1}^l \sigma_k x^{q_k}, \quad x \neq 0$$

- ▶ In the framework of  $\mathbb{R}_{\max,+}$ , the rational function can be written as

$$R(x) = P(x) - Q(x) = \max_{1 \leq j \leq n} (p_j x + \theta_j) - \max_{1 \leq k \leq l} (q_k x + \sigma_k),$$

which presents a **difference of piecewise linear convex functions**

- ▶ We observe that any arbitrary continuous function can be represented as the difference of two convex functions



## Best Approximate Solutions: One-Sided Equation

- ▶ Given an  $(m \times n)$ -matrix  $A$  and  $m$ -vector  $b$ , the problem is to find regular  $n$ -vectors  $x$  that solve the **one-sided equation**

$$Ax = b$$

- ▶ The next result offers a best approximate solution to the equation

### Theorem (K. 2004, 2009, 2012)

Let  $A$  be a regular matrix,  $b$  regular vector and  $\Delta = (A(b^{-}A)^{-})^{-}b$ . Then, the following statements hold:

1. The best approximate error for the equation is equal to  $\sqrt{\Delta}$ ;
2. The best approximate solution of the equation is given by

$$x_* = \sqrt{\Delta}(b^{-}A)^{-};$$

3. If  $\Delta = \mathbb{1}$ , then  $x_* = (b^{-}A)^{-}$  is an exact (the maximum) solution

## Two-Sided Equation

- ▶ Let  $A$  and  $B$  be given matrices of order  $(m \times n)$  and  $(m \times l)$
- ▶ Consider the problem to find regular vectors  $x$  and  $y$  of order  $n$  and  $l$ , which are solutions of the **two-sided equation**

$$Ax = By$$

- ▶ To obtain a best approximate solution of the two-sided equation, we implement an alternating algorithm (K. 2023)
- ▶ The algorithm applies the result of Theorem to solve a series of one-sided equations derived from the two-sided equation
- ▶ The equations are obtained from the two-sided equation where the left and right sides are alternately replaced by constant vectors

## Alternating Algorithm

- ▶ Given a vector  $x_0$ , the algorithm examines the equations

$$Ax_0 = By_1, \quad Ax_2 = By_1, \quad Ax_2 = By_3, \quad Ax_4 = By_3, \quad \dots$$

as one-sided equations in the unknowns  $y_1, x_2, y_3, x_4, \dots$

- ▶ The **alternating algorithm** successively calculate

$$y_1 = \sqrt{\Delta_0}((Ax_0)^- B)^-, \quad \Delta_0 = (B((Ax_0)^- B)^-)^- Ax_0,$$

$$x_2 = \sqrt{\Delta_1}((By_1)^- A)^-, \quad \Delta_1 = (A((By_1)^- A)^-)^- By_1,$$

$$y_3 = \sqrt{\Delta_2}((Ax_2)^- B)^-, \quad \Delta_2 = (B((Ax_2)^- B)^-)^- Ax_2, \quad \dots$$

- ▶ One can verify that the sequence  $\Delta_0, \Delta_1, \dots$  has a limit  $\Delta_* \geq \mathbb{1}$
- ▶ If  $\Delta_* = \mathbb{1}$ , then the two-sided equation has a solution  $(x_*, y_*)$
- ▶ Otherwise the pair  $(x_*, y_*)$  specifies a best approximate solution

# Tropical Approximation: Polynomial Approximation

- ▶ Suppose there are  $m$  samples  $(x_i, y_i)$  where  $x_i$  and  $y_i = f(x_i)$  are input and output of an unknown function  $f : \mathbb{X} \rightarrow \mathbb{X}$
- ▶ Consider the problem of best approximation of these sample data by polynomials that have  $n$  monomials and are given by

$$P(x) = \bigoplus_{j=1}^n \theta_j x^{p_j}$$

with known exponents  $p_j \in \mathbb{Q}$  and unknown coefficients  $\theta_j \in \mathbb{X}$

- ▶ The problem consists in finding the unknown coefficients that make the following equations hold exactly or approximately:

$$P(x_i) = \theta_1 x_i^{p_1} \oplus \cdots \oplus \theta_n x_i^{p_n} = y_i \quad i = 1, \dots, m$$

## Vector Representation

- ▶ Consider the system of scalar equations

$$\theta_1 x_i^{p_1} \oplus \cdots \oplus \theta_n x_i^{p_n} = y_i \quad i = 1, \dots, m$$

- ▶ With the matrix-vector notation

$$\mathbf{X} = \begin{pmatrix} x_1^{p_1} & \cdots & x_1^{p_n} \\ \vdots & & \vdots \\ x_m^{p_1} & \cdots & x_m^{p_n} \end{pmatrix}, \quad \mathbf{y} = \begin{pmatrix} y_1 \\ \vdots \\ y_m \end{pmatrix}, \quad \boldsymbol{\theta} = \begin{pmatrix} \theta_1 \\ \vdots \\ \theta_n \end{pmatrix},$$

we transform the system into the one-sided vector equation

$$\mathbf{X}\boldsymbol{\theta} = \mathbf{y}$$

- ▶ Application of Theorem yields the squared approximation error  $\Delta_*$  and corresponding vector  $\boldsymbol{\theta}_*$  of coefficients, which are given by

$$\Delta_* = (\mathbf{X}(\mathbf{y}^- \mathbf{X})^-)^- \mathbf{y}, \quad \boldsymbol{\theta}_* = \sqrt{\Delta_*} (\mathbf{y}^- \mathbf{X})^-$$

# Rational Approximation

- ▶ Consider a rational function as an approximant, which is given by

$$R(x) = \frac{P(x)}{Q(x)}, \quad P(x) = \bigoplus_{j=1}^n \theta_j x^{p_j}, \quad Q(x) = \bigoplus_{k=1}^l \sigma_k x^{q_k}.$$

- ▶ Given samples  $x_i, y_i \in \mathbb{X}$  for  $i = 1, \dots, m$ , we find the coefficients in polynomials  $P(x)$  and  $Q(x)$  to best approximate the equations

$$R(x_i) = y_i, \quad i = 1, \dots, m.$$

- ▶ We rewrite these equations as

$$\theta_1 x_i^{p_1} \oplus \dots \oplus \theta_n x_i^{p_n} = y_i (\sigma_1 x_i^{q_1} \oplus \dots \oplus \sigma_l x_i^{q_l}), \quad i = 1, \dots, m$$

## Vector Representation

- ▶ To represent the problem in vector form, we introduce the notation

$$\mathbf{X} = \begin{pmatrix} x_1^{p_1} & \dots & x_1^{p_n} \\ \vdots & & \vdots \\ x_m^{p_1} & \dots & x_m^{p_n} \end{pmatrix}, \quad \mathbf{Y} = \begin{pmatrix} y_1 & & 0 \\ & \ddots & \\ 0 & & y_m \end{pmatrix},$$

$$\mathbf{Z} = \begin{pmatrix} x_1^{q_1} & \dots & x_1^{q_l} \\ \vdots & & \vdots \\ x_m^{q_1} & \dots & x_m^{q_l} \end{pmatrix}, \quad \boldsymbol{\theta} = \begin{pmatrix} \theta_1 \\ \vdots \\ \theta_n \end{pmatrix}, \quad \boldsymbol{\sigma} = \begin{pmatrix} \sigma_1 \\ \vdots \\ \sigma_l \end{pmatrix}$$

- ▶ Then, the system takes the form of the two-sided vector equation

$$\mathbf{X}\boldsymbol{\theta} = \mathbf{Y}\mathbf{Z}\boldsymbol{\sigma}$$

- ▶ We use Alternating Algorithm to find the squared approximation error  $\Delta_*$  and obtain the vectors of coefficients  $\boldsymbol{\theta}_*$  and  $\boldsymbol{\sigma}_*$

# Examples: Approximation in Max-Plus Algebra

- ▶ We consider examples in terms of  $\mathbb{R}_{\max,+}$  (max-plus algebra)
- ▶ In this setting, both tropical polynomials and rational functions can be represented as conventional piecewise linear functions
- ▶ We assume the polynomials have a fixed number of monomials, while the exponents of these monomials are not given in advance
- ▶ We combine random search to fix exponents, with the best approximation to find coefficients of monomials in the polynomials
- ▶ To reduce the feasible set of exponents in random search, we consider only polynomials with integer exponents



## Approximation of Convex Function

- ▶ We consider  $m = 21$  input/output points  $(x_i, y_i)$  from the function

$$f(x) = x^2 - 3x^{1/3} + 5/2, \quad 0 \leq x \leq 2,$$

where  $x_i = (i - 1)/10$  and  $y_i = f(x_i)$  for  $i = 1, \dots, 21$

- ▶ We approximate the data by polynomials with  $n = 7$  monomials
- ▶ The exponents for the monomials are produced by random sampling from the discrete uniform distribution over  $[-15, 5]$
- ▶ For each sample set of exponents, coefficients of monomials are found using Theorem to minimize the approximation error
- ▶ After examining 10,000 sample sets, the minimum squared error is  $\Delta_* = 0.0481$  and the polynomial (in the conventional form) is

$$P_*(x) = \max(2.5240 - 15x, 1.4096 - 3x, 0.8736 - x, 0.3503, \\ -0.4760 + x, -1.6720 + 2x, -3.2853 + 3x)$$

## Approximation of Convex Function

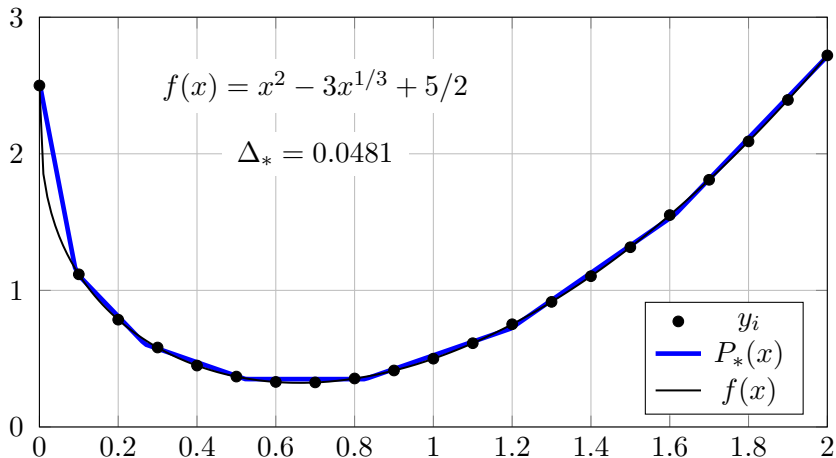


Figure: Approximation of  $f(x)$  by a polynomial  $P_*(x)$  with  $n = 7$  terms

## Approximation of Non-Convex Function

- ▶ Suppose that data points  $(x_i, y_i)$  are obtained from the function

$$g(x) = 3(x - 1)^2 \sin(x) + 1/4, \quad 0 \leq x \leq 2$$

- ▶ We approximate  $g(x)$  by rational functions given by

$$R(x) = P(x)/Q(x),$$

where  $P(x)$  and  $Q(x)$  are polynomials of  $n = 6$  and  $l = 4$  terms

- ▶ Random sampling of 10,000 pairs of sets of exponent in  $[-10, 10]$  and using Alternating Algorithm yield  $\Delta_* = 0.0701$
- ▶ The approximating function is  $R_*(x) = P_*(x) - Q_*(x)$ , where

$$P_*(x) = \max(6.9455 - 3x, 6.0860 - 2x, 4.9978 - x, 3.7461, \\ 0.7639 + 2x, -2.6361 + 4x),$$

$$Q_*(x) = \max(6.6880 - 5x, 6.2962 - 3x, 5.8009 - 2x, 2.4211)$$

## Approximation of Non-Convex Function

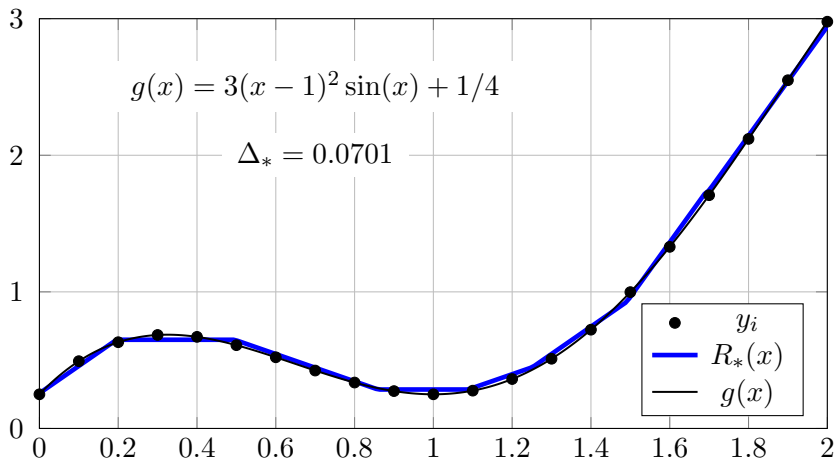


Figure: Approximation of  $g(x)$  by a rational function  $R_*(x)$  with  $n = 6$ ,  $l = 4$

## Conclusion

- ▶ We considered a best approximation problem in which a function is approximated from a set of input/output samples
- ▶ We formulated the problem in terms of a tropical semifield where addition is idempotent and multiplication is invertible
- ▶ We transformed the problem into solving tropical linear vector equations with an unknown vector on one side or on both sides
- ▶ We have derived an exact best approximate solution of the one-sided equation, obtained in direct analytical form
- ▶ To obtain a best approximate solution of the two-sided equation, we have used an iterative alternating algorithm
- ▶ As illustration, we presented results of best discrete Chebyshev approximation of real functions by piecewise linear functions