

# Tensor Empirical Interpolation Method for multivariate functions

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## Introduction

- We consider a multivariate function  $f(x_1, \dots, x_d)$  defined over  $\prod_{i=1}^d I_i$ . We focus on tensors and/or low-rank tensors for the approximation of  $f$  in a separated form:

$$f(x_1, x_2, \dots, x_d) \approx \sum_{k=1}^K \prod_{i=1}^d \varphi_i^k(x_i) = \sum_{k=1}^K \varphi_1^k(x_1) \dots \varphi_d^k(x_d)$$

where,  $\varphi_i^k(x_i), i \in \{1, \dots, d\}$  are univariate functions.

- We propose an algorithm for the decomposition of multivariate functions using the Empirical Interpolation Method (EIM)<sup>(1)</sup>.

<sup>1</sup> M. Barrault, Y. Maday, N. C. Nguyen, and A. T. Patera, *An empirical interpolation method: application to efficient reduced-basis discretization of partial differential equations*, Comptes Rendus Mathématique, 339(9) : 667-672, 2004.

## Tensor Empirical Interpolation Method (TEIM) for bivariate functions<sup>(2)</sup>

- $f : I \times J \rightarrow \mathbb{R}$  a uniformly continuous function,  $m$  and  $n$  two fixed integers.

- First step: EIM in  $x$  direction: “ $x$  space variable and  $y \in J$  parameter”

- $\tilde{y}_k = \arg \max_{y \in J} \|f(., y) - \mathcal{I}_x^{(k-1)} f(., y)\|_{L_x^\infty(I)}$
- $x_k = \arg \max_{x \in I} |f(x, \tilde{y}_k) - \mathcal{I}_x^{(k-1)} f(x, \tilde{y}_k)|$
- The basis function

$$q_k^{(k)}(x) = \frac{f(x, \tilde{y}_k) - \mathcal{I}_x^{(k-1)} f(x, \tilde{y}_k)}{f(x_k, \tilde{y}_k) - \mathcal{I}_x^{(k-1)} f(x_k, \tilde{y}_k)}.$$

- A slight modification of the original EIM: update all the previous basis functions

$$q_i^{(k)}(x) = q_i^{(k-1)}(x) - q_i^{(k-1)}(x_k) q_k^{(k)}(x) \quad \forall i \in \{1, k-1\}.$$

- The rank- $k$  interpolation operator  $\mathcal{I}_x^{(k)} f$

$$\mathcal{I}_x^{(k)} f(x, y) = \sum_{i=1}^k f(x_i, y) q_i^{(k)}(x), \quad y \in J.$$

- Second step: EIM in  $y$  direction: “ $y$  space variable and  $x \in I$  parameter”

- A set of interpolation points :  $\{y_1, \dots, y_n\}$
- A set of basis functions  $\{s_1^{(n)}, s_2^{(n)}, \dots, s_n^{(n)}\}$
- The rank- $n$  interpolation operator

$$\mathcal{I}_y^{(n)} f(x, y) = \sum_{i=1}^n f(x, y_i) s_i^{(n)}(y), \quad x \in I.$$

- Third step: tensorized interpolation of  $f$ :

$$\mathcal{I}^{m,n} f(x, y) = \mathcal{I}_y^{(n)} \mathcal{I}_x^{(m)} f(x, y) = \sum_{i=1}^m \sum_{j=1}^n f(x_i, y_j) q_i^{(m)}(x) s_j^{(n)}(y).$$

<sup>2</sup> F. De Vuyst, A. Toumi, *A mixed EIM-SVD tensor decomposition for bivariate functions*. ArXiv preprint 2017.

## Extension of the TEIM to the multidimensional case

- $f(x_1, \dots, x_d)$  uniformly continuous function on  $\Omega = \prod_{i=1}^d I_i$ ,  $d \geq 2$ .

- EIM algorithm “direction-by-direction”: for each variable  $x_i$ ,  $1 \leq i \leq d$ , we compute:

Interpolation points:  $\{x_i^j\}_{j=1, m_i}$ , basis functions:  $\{q_{i,j}^{(m_i)}(x_i)\}_{j=1, m_i}$  and  $\mathcal{I}_{x_i}^{(m_i)} f$ .

- The tensorized interpolation of the function  $f$

$$\mathcal{I}^{m_1, \dots, m_d} f(x_1, \dots, x_d) = \sum_{j_1=1}^{m_1} \dots \sum_{j_d=1}^{m_d} f(x_1^{j_1}, \dots, x_d^{j_d}) q_{1,j_1}(x_1) \dots q_{d,j_d}(x_d),$$

## Properties of TEIM

- The basis functions  $q_{i,j}^{(m_i)}$ ,  $1 \leq j \leq m_i$  satisfy the Lagrange property
- Reinforced interpolation property: for all  $1 \leq i \leq d$

$$\mathcal{I}^{m_1, \dots, m_d} f(x_1, \dots, x_{i-1}, x_i^k, x_{i+1}, \dots, x_d) = f(x_1, \dots, x_{i-1}, x_i^k, x_{i+1}, \dots, x_d),$$

- A priori error estimation :  $f^*$  best approximation of  $f$ ,  $L_i^{m_i}$  the Lebesgue constants

$$\|f - \mathcal{I}^{m_1, \dots, m_d} f\|_{L^\infty} \leq \inf_{f^* \in \mathcal{W}^{m_1, \dots, m_d}} \|f - f^*\|_{L^\infty} \left( 1 + \prod_{i=1}^d L_i^{m_i} \right)$$

## Complexity of the TEIM

- Numerical complexity: curse of dimensionality for the multivariate extension
  - First solution: Singular Value Decomposition (SVD)
  - Second Solution: Sparse collocation method

## Numerical results

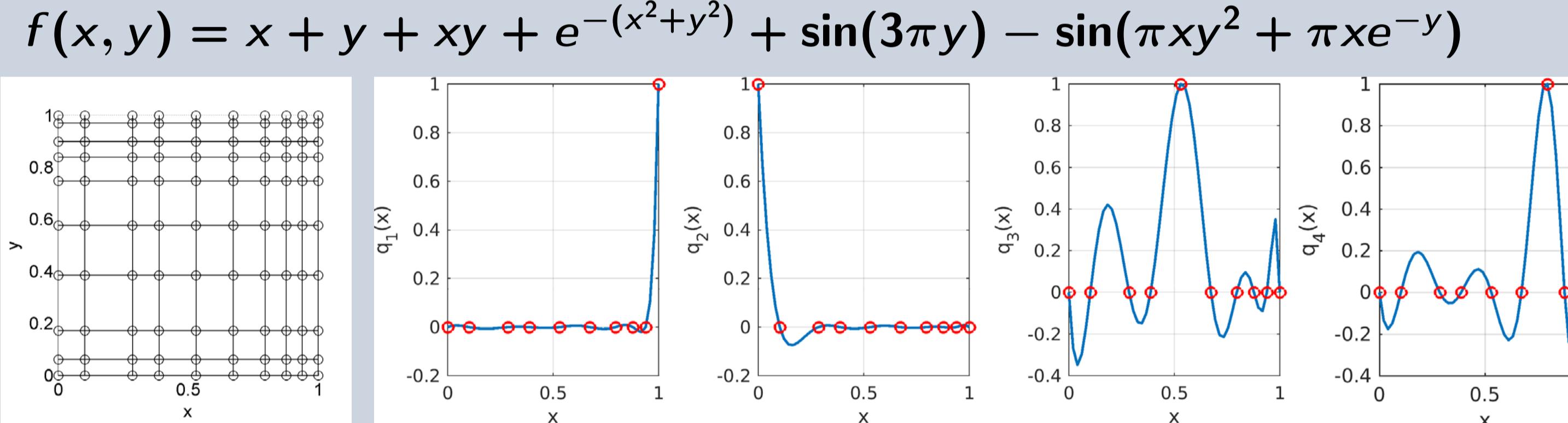


Figure: The first 100 interpolation points and the first four basis functions with the 10 TEIM points  $x_i$ .

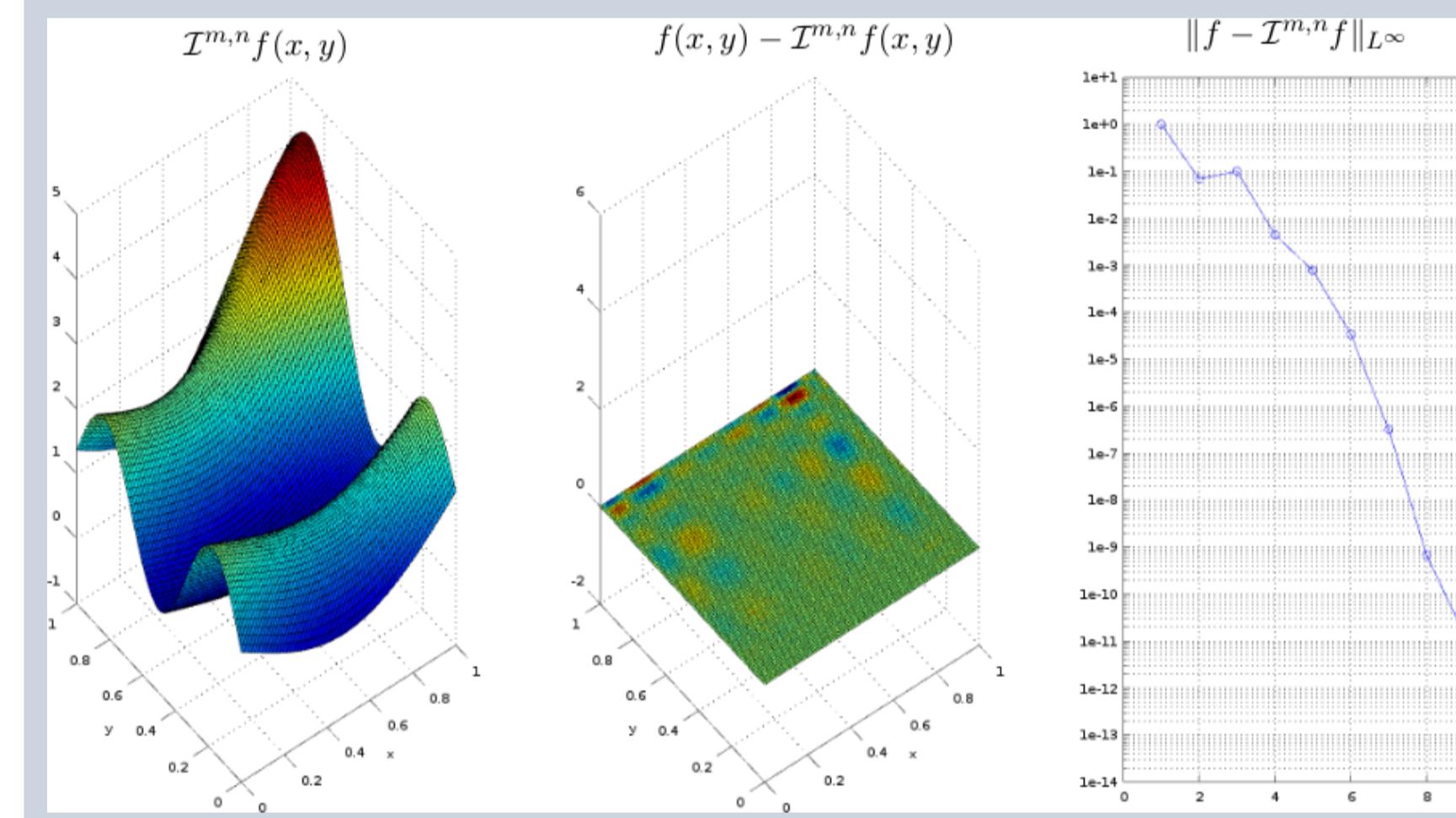


Figure: The TEIM interpolation

- The tensorized interpolation operator  $\mathcal{I}^{10,10}$  (left).
- The difference between  $f$  and its approximation resulting from TEIM (middle).
- Convergence of the TEIM interpolation approach with respect to  $m$  (right)

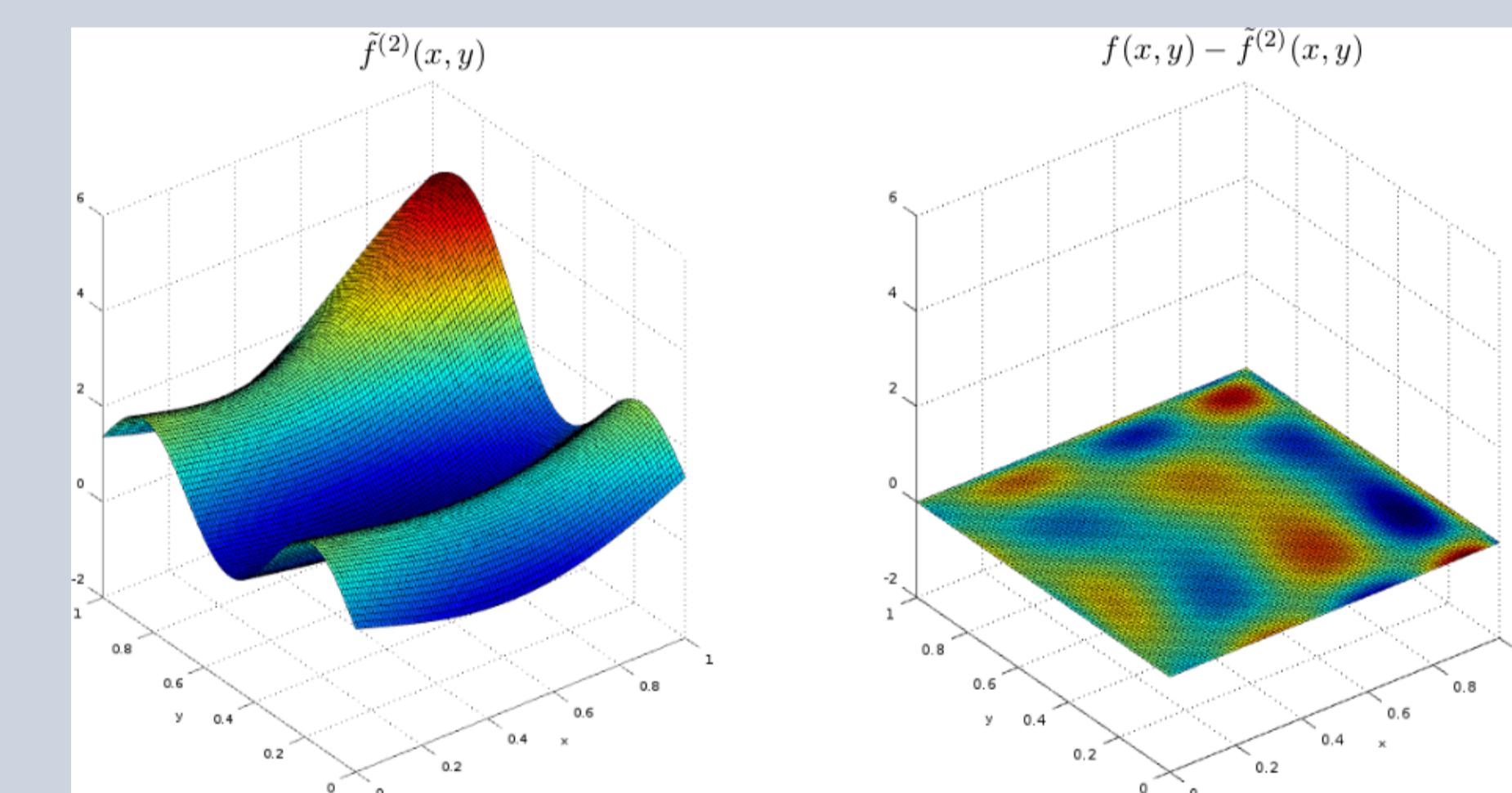


Figure: The SVD decomposition

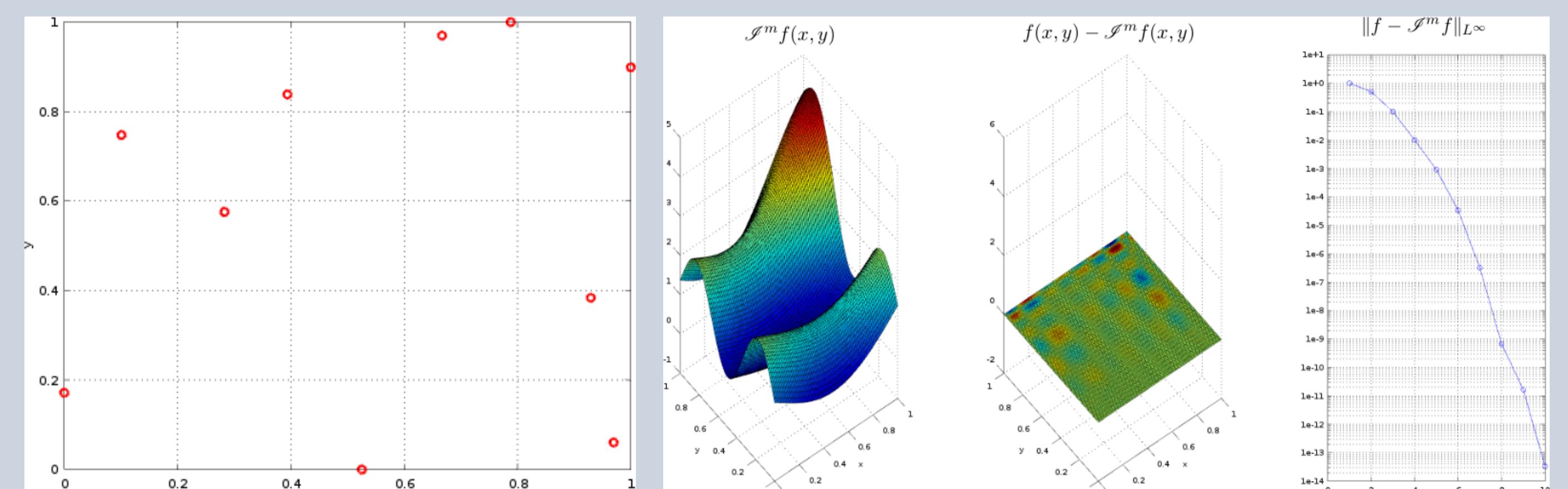
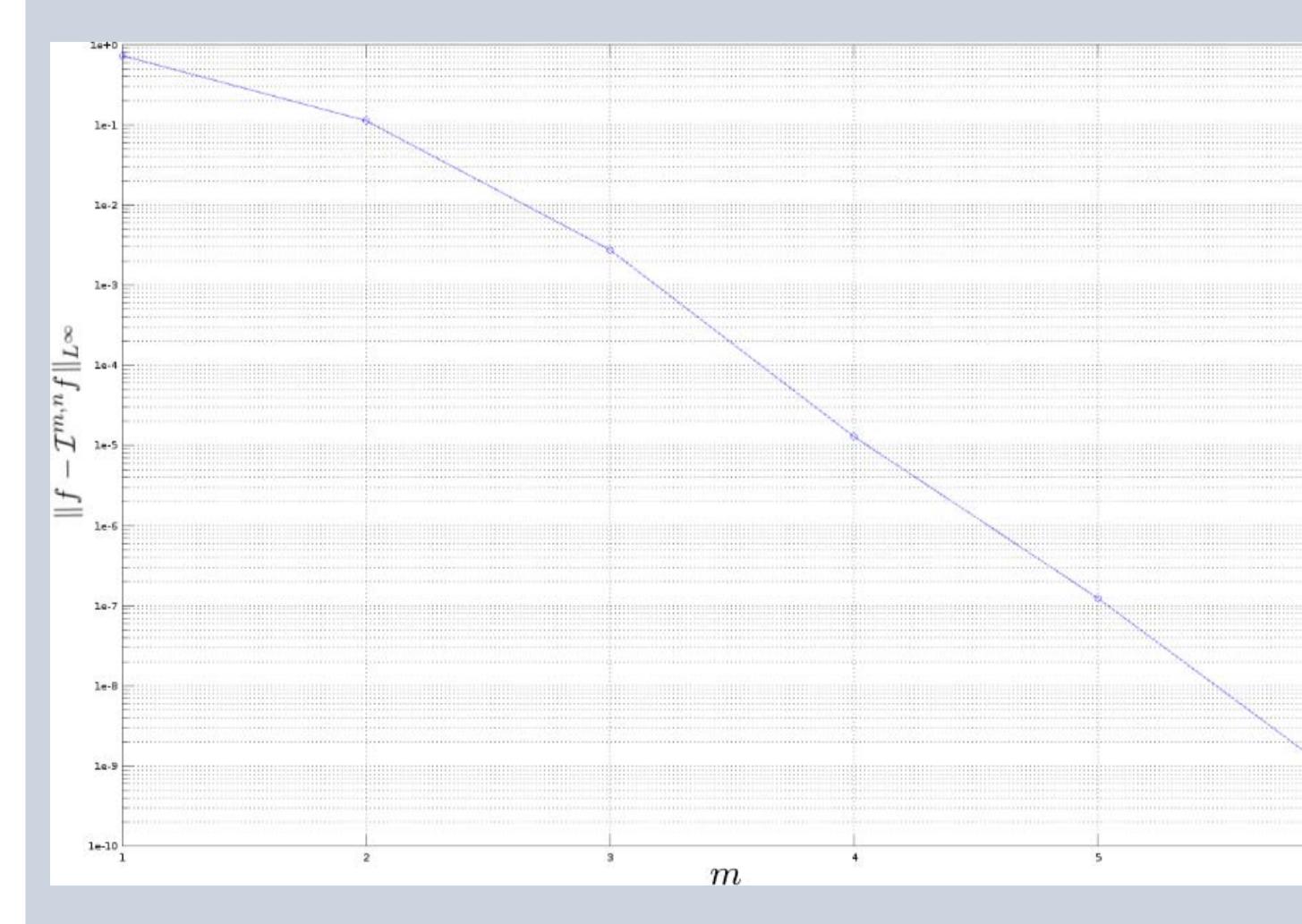


Figure: The sparse collocation points method



- $f = \sin(\pi x_1 x_2 x_3 x_4 x_5)$

- Numerical convergence confirm that even in the five dimensions case, the TEIM still have good properties.

## Acknowledgements

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