

# Spectral Methods

**Abstract:** First, theoretical introductions to interpolation and spectral expansion are given and a particular emphasis is put on the fast convergence of the spectral approximation. I will present different approaches to solve differential equations, limiting ourselves to the one-dimensional case, with one or several domains.

- Motivation
- Introduction
- Best Polynomial Approximation
- Interpolation
- **Weighted Residual Method**
- Multi-domain
- Non-linear Case
- Conclusion

- Non-locality of the spectral representation.
- The fast convergence of the spectral approximation.

# Introduction

When doing simulations and solving ODEs in  $[-1,1]$ , *finite-differences (FD) methods*:

Set up a grid:

$$\{x_i\}_{i=0,\dots,N} \in [-1, 1] \quad (1)$$

Represent  $f$

$$\{f_i = f(x_i)\}_{i=0,\dots,N} \quad (2)$$

Represent the derivative

$$\forall i < N, f'_i = f'(x_i) \simeq \frac{f_{i+1} - f_i}{x_{i+1} - x_i} \quad (3)$$

Decay as  $\Delta x = 1/N$

# Introduction

*Spectral methods:*

$$f(x) \simeq \sum_{i=0}^N c_i \Phi_i(x) \quad (4)$$

basis functions or trial functions  $\{\Phi_i\}_{i=0,\dots,N}$ .

For a ODE system,

$$\begin{cases} Lu(x) = s(x), x \in U \\ Bu(x) = 0, x \in \partial U \end{cases} \quad (5)$$

$L$  is a **linear** operator;

$B$  is the operator defining the boundary conditions;

$s$  is the source term.

**residual**  $R = L\bar{u} - s$  is small   $\bar{u}$  is a *numerical solution*

# Introduction

Q:How to express the smallness of residual?

A:Weighted Residual

$$\forall i = 0, \dots, N, \quad (\xi_i, R)_\omega = 0 \quad (6)$$

choose a set of *test functions*  $\{\xi_i(x)\}_{i=0\dots N}$ .

# Best Polynomial Approximation

Polynomials are the functions that a computer can exactly evaluate.

Try to approximate any function by a polynomial.

Given a continuous function  $f$ , the best polynomial approximation:  $P_N^*(x)$ ,

$$\|f - P_N^*\|_\infty = \min \{ \|f - P\|_\infty, P \in \mathbb{P}_N \} \quad (7)$$

If and only if there exist  $N + 2$  points  $x_i$ , s.t.

$$f(x_i) - p_N^*(x_i) = (-1)^{i+\delta} \|f - p_N^*\|_\infty, \quad \text{with } \delta = 0 \text{ or } 1 \quad (8)$$

$p_N^*$  should be the **interpolation approximation**

# Interpolation

Continuous function  $f$ , grids  $X$  with  $N + 1$  nodes  $x_i$ , a unique polynomial of degree  $N$ ,  $I_N^X f$ :

$$I_N^X f(x_i) = f(x_i), \quad 0 \leq i \leq N \quad (9)$$

Alternatively,

$$\begin{aligned} I_N^X f &= \sum_{i=0}^N f(x_i) \ell_i^X(x) \\ &= \sum_{i=0}^N \frac{\omega_{N+1}^X(x)}{(x - x_i) \omega_{N+1}'^X(x_i)} f(x_i) \end{aligned} \quad (10)$$

where

$$\ell_i^X(x) = \prod_{j=0, j \neq i}^N \frac{x - x_j}{x_i - x_j}, \quad \omega_{N+1}^X(x) = \prod_{i=0}^N (x - x_i) \quad (11)$$



# Interpolation

Compare to the best approximation:

$$\|f - I_N^X f\|_\infty \leq (1 + \Lambda_N(X)) \|f - p_N^*\|_\infty \quad (12)$$

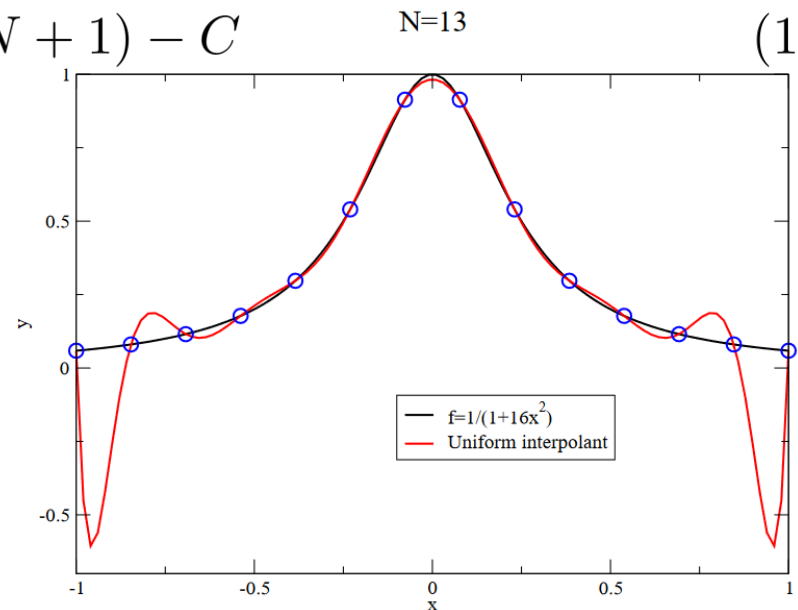
where  $\Lambda_N$  is the *Lebesgue constant* of the grid  $X$ ,

$$\Lambda_N(X) = \max_{x \in [-1, 1]} \sum_{i=0}^N |\ell_i^X(x)|. \quad (13)$$

For any choice of grid  $X$ , there exists a constant  $C > 0$ , such that:

$$\Lambda_N(X) > \frac{2}{\pi} \ln(N + 1) - C \quad (14)$$

Runge phenomenon



# Interpolation

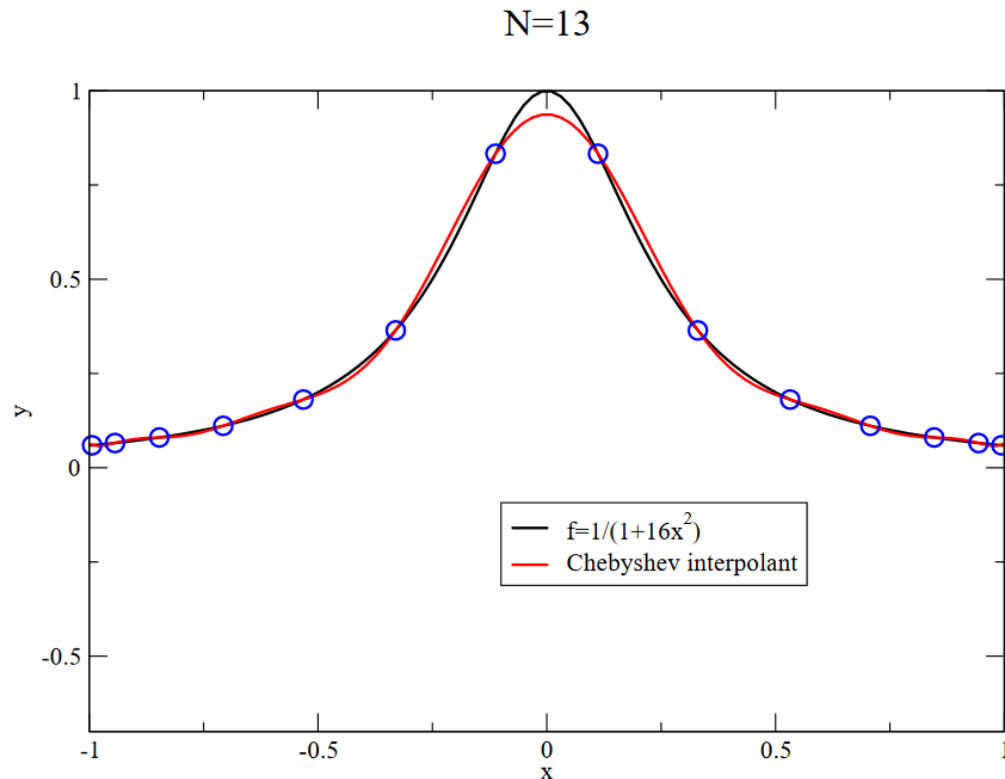
However, one can hope to minimize the interpolation error by choosing a grid such.

$$f(x) - I_N^X f(x) = \frac{f^{N+1}(\epsilon(x))}{(N+1)!} \omega_{N+1}^X(x) \quad (15)$$

For example, we can choose a grid which nodes are **the zeros of the Chebyshev polynomial  $T_{N+1}(x)$** .

$$\|\omega_{N+1}^X\|_\infty = \frac{1}{2^N}$$

**Grid points are important**



# Interpolation

A basis of  $\mathbb{P}_N = \{p_n\}_{n=0,\dots,N}$ . For orthogonal polynomials,  $(p_i, p_j)_\omega = 0, i \neq j$

$$(f, g)_w = \int_{x \in [-1, 1]} f(x)g(x)w(x)dx \quad (16)$$

We can project function  $f$  on this basis,

$$P_N f = \sum_{n=0}^N \hat{f}_n p_n \quad (17)$$

where the coefficients are

$$\hat{f}_n = \frac{(f, p_n)}{(p_n, p_n)} \quad (18)$$

Which requires the evaluation of  $f$  at a great number of points, making the whole numerical scheme impractical.



**Gaussian quadrature**

# Interpolation

Given a measure  $w$ , there exist  $N + 1$  real numbers  $w_n$  and  $N + 1$  real numbers  $x_n$ , s.t.

$$\int_{[-1,1]} f(x)\omega(x)dx = \sum_{j=0}^N f(x_j)\omega_j + E_N[f], \quad (19)$$

When  $f \in \mathbb{P}_{2N+\delta}$ ,  $E_N[f] = 0$ ;  $x_j$  are the collocation points;  $w_j$  can be determined by,

$$w_j = \int_{[-1,1]} \ell_j^X(x)w(x)dx \quad (20)$$

integer  $\delta$  depends on the choice of quadrature.

When  $\{x_n\}$  are the roots of orthogonal polynomial  $P_{N+1}$ , we have  $f = rP_{N+1} + q \in \mathbb{P}_{2N+1}$ ,  $r, q \in \mathbb{P}_N$  and  $\delta = 1$ .

$N + 1$  collocation points enjoy the maximum degree of precision  $2N + 1$ .

There are some example polynomials:

- Legendre polynomials  $P_n(x)$ :  $x_i$  are the nodes of  $P_{N+1}(x)$ ,

$$w_i = \frac{2}{(1 - x_i^2) [P'_{N+1}(x_i)]^2} \quad (21)$$

- Chebyshev polynomials  $T_n(x)$ :  $T_n(x) = 2xT_{n-1}(x) - T_{n-2}(x)$ ,  $T_0 = 1$ ,  $T_1 = x$ ,

$$x_i = \cos \frac{(2i+1)\pi}{2N+2}, \quad w_i = \frac{\pi}{N+1} \quad (22)$$

# Interpolation

By making use of the Gaussian quadrature, define the *interpolant* of a function  $f$  by

$$\text{Spectral interpolation} \quad I_N f = \sum_{n=0}^N \tilde{f}_n p_n(x) \quad (23)$$

where

$$\tilde{f}_n = \frac{1}{\gamma_n} \sum_{i=0}^N f(x_i) p_n(x_i) w_i \quad \text{and} \quad \gamma_n = \sum_{i=0}^N p_n^2(x_i) w_i \quad (24)$$

$\tilde{f}_n$  is that they are computed by estimating  $f$  at the  $N + 1$  collocation points only.

configuration space:

if the function is described by its value at the  $N + 1$  collocation points  $f(x_i)$ .



coefficient space:

if the function is described by  $N + 1$  coefficients  $\tilde{f}_i$

# Interpolation

One of the main advantage of spectral method is the very fast convergence of the interpolant  $I_N f$ . For a  $C^m$  function  $f$

- For Legendre:

$$\|I_N f - f\|_{L^2} \leq \frac{C_1}{N^{m-1/2}} \sum_{k=0}^m \|f^{(k)}\|_{L^2} \quad (25)$$

- For Chebyshev:

$$\|I_N f - f\|_{L_w^2} \leq \frac{C_2}{N^m} \sum_{k=0}^m \|f^{(k)}\|_{L_w^2} \quad (26)$$


When  $m \rightarrow \infty$ , the difference decays faster than any power of  $N$ .

# Weighted Residual Method

More about Weighted residual method:

Let us consider a differential equation,

$$Lu(x) = S(x), \quad x \in [-1, 1] \quad (27)$$

$u$  is an admissible solution  Fulfills the boundary conditions  
& the residual  $R = Lu - S$  is small.

For a set of  $N + 1$  test functions  $\{\xi_n\}_{n=0\dots N}$  on  $[-1, 1]$

$$(R, \xi_k) = \sum_{i=0}^N R(x_i) \xi_k(x_i) w_i = 0, \quad \forall k \leq N \quad (28)$$

Depending on the choice of the test functions and the way the boundary conditions are enforced, there are different solvers.



# Weighted Residual Method

## 1. Tau method

the test functions are the **basis used for the spectral expansion**.

$$\sum_{i=0}^N L_{ni} \tilde{u}_i = \tilde{s}_n, \quad \forall n \leq N \quad (29)$$

Impose the boundary conditions:

In the Tau-method, this is done by relaxing the last two equations in (29)(i.e. for  $n = N - 1$  and  $n = N$ ) and by replacing them by the boundary conditions at  $x = -1$  and  $x = 1$ .

## 2. Collocation method

Test functions are zero at each but one collocation point.

They are indeed **the Lagrange cardinal polynomials**, i.e.  $\ell_i^X(x_j) = \delta_{ij}$

$$\sum_{i=0}^N \sum_{j=0}^N L_{ij} \tilde{u}_j T_i(x_n) = S(x_n), \quad \forall n \leq N \quad (30)$$

Impose the boundary conditions:

In this case, the relaxed conditions are the two associated with the outermost points, i.e.  $n = 0$  and  $n = N$ , which are replaced by appropriate boundary conditions to get an invertible system.

## 3. Galerkin method

Test polynomials  $G_i$  **individually verify the boundary conditions**.  
And solution  $u$  as a sum of polynomials  $G_i$

$$u(x) = \sum_{k=0}^K \tilde{u}_k^G G_k(x), \text{ with } G_i = \sum_{j=0}^N M_{ji} T_j \quad (31)$$

Galerkin system reads:

$$(Lu, G_n) = (S, G_n)$$
$$\sum_{k=0}^{N-2} \tilde{u}_k^G \sum_{i=0}^N \sum_{j=0}^N M_{in} M_{jk} L_{ij}(T_i|T_i) = \sum_{i=0}^N M_{in} \tilde{s}_i(T_i|T_i), \quad \forall n \leq N-2. \quad (32)$$

Spectral methods are very efficient dealing with  $C^\infty$  functions.

If not?

**Multi-domain techniques:** in each domain, do same thing and only deal with  $C^\infty$  functions.

For simplicity the physical space is split into two domains

- First domain:  $x \leq 0 \Rightarrow x_1 = 2x + 1, x_1 \in [-1, 1]$ :

$$I_N u(x) = \sum_{i=0}^N \tilde{u}_i^1 T_i(x_1(x))$$

- Second domain:  $x \geq 0 \Rightarrow x_2 = 2x - 1, x_2 \in [-1, 1]$

➤ Multi-domain tau method: when  $x \leq 0$ ,

$$(T_n, R) = 0 \implies \sum_{i=0}^N L_{ni} \tilde{u}_i^1 = \tilde{s}_n^1 \quad \forall n \leq N \quad (33)$$

➤ Multi-domain collocation method : when  $x \leq 0$ ,

$$\sum_{i=0}^N \sum_{j=0}^N L_{ij} \tilde{u}_j^1 T_i(x_{1n}) = S(x_{1n}) \quad \forall n \leq N \quad (34)$$

- $N - 1$  residual equations in the first domain;
- $N - 1$  residual equations in the second domain;
- 2 boundary conditions;
- 2 matching conditions: solution and its first derivative at  $x = 0$ .

$2N + 2$  equations



$2N + 2$  undetermined coefficients

# Non-linear Case

If  $L$  is a non-linear differential operator, we can separate  $L$  into the linear part  $\hat{L}$  and non-linear part  $N$  as,

$$\begin{aligned}Lu(x) &= s(x) \\ (\tilde{L} + N)u(x) &= s(x) \\ \tilde{L}u(x) &= S(x)\end{aligned}\tag{35}$$

where  $S(x) = s(x) - Nu(x)$ . It is difficult to evaluate the  $\tilde{S}_i$ . And with the collocation method, the residual is

$$\begin{aligned}R &= \tilde{L}u(x_n) - S(x_n) = 0 \\ \sum_{i=0}^N \sum_{j=0}^N \tilde{L}_{ij} \tilde{u}_j T_i(x_n) &= S(x_n), \quad \forall n \leq N\end{aligned}\tag{36}$$

which is equivalent to

$$M\mathbf{u} = \mathbf{S}\tag{37}$$

# Non-linear Case

We can use the Newton method to solve this system.

$$\mathbf{u}^{(k+1)} = \mathbf{u}^{(k)} - \mathbf{J}^{-1} R(\mathbf{u}^{(k)}) \quad (38)$$

where  $\mathbf{J} = R'(\mathbf{u}^{(k)})$  is Jacobian matrix.

relaxation method:  
FD+NM



**Pseudo-spectral method**

# Conclusion

- ✓ Introduce the concepts in one dimension
- ✓ Weighted Residual Method to minimize error in a certain way and spectral methods in one dimension.
- ✓ Spectral method for non-linear system



Thank you for your attention and the  
questions