The Pseudospectral Method

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- Any finite-difference operation can be decribed as a convolution operation implying that the specific finite-difference operator has a spectral representation that can be compared with the exact -ik operator
- Using Taylor Series for accuracy improvement (= length of operator)
- Using Fourier concepts for calculating exact derivatives
- ⇒ Fourier method can be interpreted as an infinite order finite-difference scheme



Let us restate the previous result of the partial derivative as an inverse Fourier transform defined as

$$\partial_{x}f(x) = \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} \partial_{x}F(k)e^{-ikx}dk$$
$$= \frac{1}{\sqrt{2\pi}} \int_{-\infty}^{\infty} -ikF(k)e^{-ikx}dk$$

Defining the factors in front of the complex amplitude spectrum F(k) of function f(x) as

$$\partial_X f(X) = \int_{-\infty}^{\infty} D(k) F(k) e^{-ikX} dk, \ D(k) = -ik$$

Two functions d(x) and f(x) with complex spectra D(k) and F(k), are thus linked by

$$D(k) = \mathscr{F}[d]$$

$$F(k) = \mathscr{F}[f]$$

$$d * f = \mathscr{F}^{-1}[D(k)F(k)]$$

where \mathscr{F} represents the Fourier transform, and \ast denotes convolution, defined in the continuous case as

$$(d*f)(x) := \int_{-\infty}^{\infty} d(x')f(x-x')dx'$$

and in the discrete case with vectors d_i , i = 0, 1, ..., m, and f_j , j = 0, 1, ..., n

$$(d*f)_k = \sum_{i=0}^m d_k f_{k-i}, k = 0, 1, ..., m+n$$

D(k) in general is nothing else but a function defined in the spectral domain acting like a filter on the complex spectrum F(k).

The convolution theorem implies that

$$\partial_X f(X) = \int_{-\infty}^{\infty} d(X - X') f(X') dX'$$

where d(x) is a real function, the spatial representation of spectrum D(k), in other words

$$d(x) = \mathscr{F}^{-1}[D(k)].$$

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Limiting the wavenumber domain to the Nyquist wavenumber $k_{max} = \pi/dx$. Thus D(k) becomes

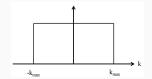
$$D(k) = ik[H(k + k_{max}) - H(k_{kmax})]$$

where H() denotes the Heaviside function, and to obtain d(x) we simply have to inverse transform

$$d(x) = \mathscr{F}^{-1}[ik[H(k+k_{max})-H(k-k_{kmax})]]$$

leading to

$$d(x) = \frac{1}{\pi x^2} [sin(k_{max}x) - k_{max}x cos(k_{max}x)]$$



The r.h.s. of the Fourier integral is a multiplication of two spectra:

- derivative operator ik
- boxcar function and its solution is the *sinc* function of the form sin(x)/x

If space is discretized according to

$$x_n = n dx, \quad n = -N, \ldots, 0, \ldots, N$$

In this case the convolution integral becomes a convolution sum

$$\partial_X f(x) \approx \sum_{n=-N}^{n=N} d_n f(x - n dx)$$

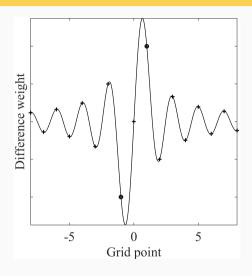
where d_n is the difference operator.

Inserting the discretization into

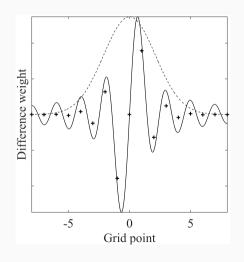
$$d(x) = \frac{1}{\pi x^2} [sin(k_{max}x) - k_{max}x cos(k_{max}x)]$$

we obtain analytically the discrete difference operator

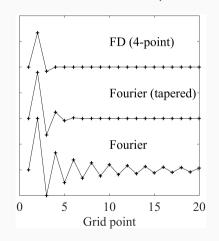
$$d_n = \begin{cases} 0 & \text{for } n = 0 \\ \frac{(-1)^n}{ndx} & \text{for } n \neq 0 \end{cases}$$



Truncated Fourier operator



Convolutional difference operators

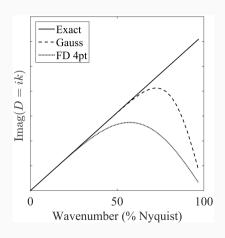


How can we conveniently compare the accuracy of such operators?

 \implies The space representation of the exact difference operator D(k) = -ik in the wavenumber domain

Thus, for a finite-difference operator d_n^{FD} we will obtain

$$D^{FD}(k) = i \, \tilde{k}_{\nu}(k) = \mathscr{F}[d_n^{FD}]$$



Chebyshev Pseudospectral

Method

Let us start with the trigonometric relation

$$\cos \left[(n+1)\phi \right] + \cos \left[(n-1)\phi \right] = 2\cos(\phi)\cos(n\phi) \qquad n \in \mathbb{N}.$$

Inserting n = 0 leads to a trivial statement. However, for $n \ge 1$ we obtain statements like

$$\cos(2\phi) = 2\cos^{2}(\phi) - 1$$

$$\cos(3\phi) = 4\cos^{3}(\phi) - 3\cos(\phi)$$

$$\cos(4\phi) = 8\cos^{4}(\phi) - 8\cos^{2}(\phi) + 1$$

$$\vdots$$

Chebyshev Polynomials

$$cos(n\phi) =: T_n(cos(\phi)) = T_n(x)$$

with

$$X = \cos(\phi)$$
 $X \in [-1,1]$, $n \in \mathbb{N}_0$

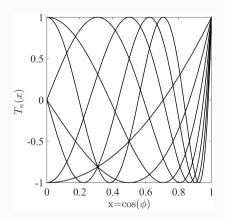
 T_n being the n-th order Chebyshev polynomial. Furthermore

$$|T_n(x)| \leqslant 1 \text{ for } [-1,1], \qquad n \in \mathbb{N}_0$$

Finally, we can write down the first polynomials in $x \in [-1, 1]$

$$T_0(x) = 1$$

 $T_1(x) = x$
 $T_2(x) = 2x^2 - 1$
 $T_3(x) = 4x^3 - 3x$
 $T_4(x) = 8x^4 - 8x^2 - 1$
 \vdots

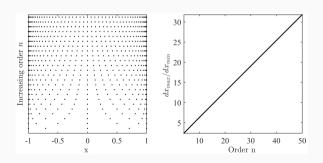


A generating function calculates the Chebyshev polynomials of any order n

$$T_{n+1}(x) = 2xT_n(x) - T_{n-1}(x), n \ge 1$$

The extremal values $x_k^{(e)}$ of these polynomials have a very simple form

$$x_k^{(e)} = \cos(\frac{k\pi}{n})$$
 $k = 0, 1, 2, ..., n$



The Chebyshev polynomials form an orthogonal set with respect to the weighting function $w(x) = 1/\sqrt{1-x^2}$.

⇒ Using them as a basis for function interpolation

$$f(x) \approx g_n(x) = \frac{1}{2}c_0T_0(x) + \sum_{k=1}^n c_kT_k(x)$$

where f(x) is an arbitrary function in the interval [-1,1], $T_n(x)$ are the Chebyshev polynomials, and c_k are real coefficients.

By minimizing the least-squares misfit between f(x) and $g_n(x)$, the coefficients c_k can be found

$$c_k = \frac{2}{\pi} \int_{-1}^{1} f(x) T_k(x) \frac{\mathrm{d}x}{\sqrt{1-x^2}} \qquad k = 0, 1, \dots, n$$

which - after substituting $x = \cos(\phi)$ - can be written as

$$c_k = \frac{1}{\pi} \int_{-\pi}^{\pi} f(\cos(\phi)) \cos(k\phi) d\phi \qquad k = 0, 1, \dots, n$$

These coefficients turn out to be the Fourier coefficients for the even 2π -periodic function $f(\cos(\phi))$ with $x = \cos(\phi)$.

The points we need are the extrema of the Chebyshev polynomials (Chebyshev-) Gauss-Lobatto points defined as

$$x_i = \cos(\frac{\pi}{N}i)$$
 $i = 0, 1, \ldots, n$.

With these unevenly distributed grid points we can define the discrete Chebyshev transform as follows. The approximating function is

$$g_n^*(x) = \frac{1}{2}c_0^*T_0 + \sum_{k=1}^{n-1}c_k^*T_k(x) + \frac{1}{2}c_n^*T_n$$

with the coefficients defined as

$$c_k^* = \frac{2}{m} \left[\frac{1}{2} (f(1) + (-1)^k f(-1)) + \sum_{j=1}^{m-1} f_j \cos(\frac{kj\pi}{m}) \right]$$

$$k = 0, 1, \dots, n, \qquad n = m$$

where f(1) and f(-1) are the function values at the interval boundaries and f_j are the values at the collocation points $f(x = \cos(j\pi/m))$. The fundamental property is

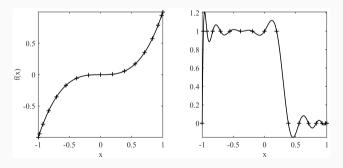
$$g_m^*(x_i) = f(x_i)$$

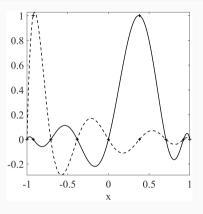
where x_i are the collocation points.

Example

When we have a function $f(x) = x^3$ in the interval [-1, 1] using the Chebyshev transform, the function f(x)

- 1 can be exactly interpolated at the collocation points
- 2 converges very rapidly with just a few polynomials





Two cardinal functions with Chebyshev polynomials for grid points i = n/2 (solid line) and i = n (dashed line) are shown for n = 8

Chebyshev Derivatives,

Differentiation matrices

A convolution operation can be formulated as a matrix-vector product involving Toeplitz matrices. Defining a derivative matrix D_{ij}

$$D_{ij} = \begin{cases} -\frac{2N^2+1}{6} & \text{for } i = j = N \\ -\frac{1}{2} \frac{x_i}{1-x_i^2} & \text{for } i = j = 1,2,...,N-1 \\ \frac{c_i}{c_j} \frac{(-1)^{i+j}}{x_i-x_j} & \text{for } i \neq j = 0,1,...,N \end{cases}$$

where N+1 is the number of Chebyshev collocation points $x_i = \cos(i\pi/N)$, i = 0, ..., N and the c_i are given as

$$c_i = \begin{cases} 2 & \text{for i=0 or N} \\ 1 & \text{otherwise} \end{cases}$$

This differentiation matrix allows us to write the derivative of function $u_i = u(x_i)$ simply as

$$\partial_X u_i = D_{ij} u_j$$

where the right-hand side is a matrix-vector product, and the Einstein summation convention applies.

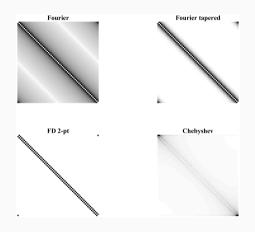
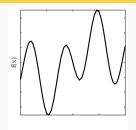
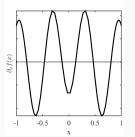


Illustration of differentiation matrices (n=64). **Top left:** Exact Fourier differentiation matrix for regular grid (full). Top right: Exact Chebyshev differentiation matrix for Chebyshev collocation points. Increasing weights at the corners overshadows interior val-Bottom Left: Standard ues. 2-point finite difference operator (banded). Bottom Right: Tapered Fourier operator (12-point). Matrix is banded. For illustration purposes the square root of the absolute values are shown.





By testing the differentiation, we define a function to seismic wavefield calculations as

$$f(x_i) = \sin(2x_i) - \sin(3x_i) + \sin(4x_i) - \sin(10x_i)$$

in the interval $x_i \in [-1,1]$, where the discrete points are the Chebyshev collocation points $x_i = \cos(\pi i/n)$, i = 0, ..., n given for n = 63.

Elastic 1D with Chebyshev Method

Elastic 1D with Chebyshev Method

Elastic 1D wave equation using the standard 3-point operator

$$\rho_{i} \frac{u_{i}^{j+1} - 2u_{i}^{j} + u_{i}^{j-1}}{\mathrm{d}t^{2}} = (\partial_{x} [\mu(x)\partial_{x}u(x,t)])_{i}^{j} + f_{i}^{j}$$

where the lower index *i* corresponds to the spatial discretization and the upper index *j* to the discrete time levels.

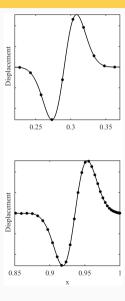
The displacement field as well as the geophysical parameters like density ρ_i and shear modulus μ_i are defined on the irregular Chebyshev collocation points.

Example

- Distance between grid points is 80 times smaller at the boundaries
- The time step for a stable simulation requires cdt/dx ≤ ε
- \Rightarrow Grid distance near boundary is responsible for the global simulation time step

Parameter	Value
nx	200
С	3000 m/s
ho	$2500 \ kg/m^3$
dt	$6 \times 10^{-8} \text{ s}$
dx_{min}	$1.2 \times 10^{-4} \text{ m}$
dx_{max}	0.015 m
f_0	100 kHz
ϵ	1.4

Result



```
Time extrapolation
 Differentiation matrix
D = get cheby matrix(nx)
for it in range (nt):
    # Space derivatives (corrected May 11, 2020)
   du = mu * D @ u
   du = D @ du/rho
    # Time extrapolation
   unew = 2*u - uold + np.transpose(du) * dt**2
    # Source injection
   unew = unew + sq*src[it]*dt**2/rho
    # Remapping
   uold, u = u, unew
   u[0] = 0; u[nx] = 0 # set boundaries pressure free
```

Result

- To obtain a stable solution we need a very small time step that is only needed at the boundaries. Mathematically the time step scales with $O(N^{-2})$.
- In principle we can stretch the spatial grid such that the grid points close to the boundaries are further apart while the grid point distances at the centre remain basically unchanged. If that stretching function is ξ(x) then the derivative of a function f(x) on the stretched grid is defined as

$$\partial_x f(x) = \frac{\partial f}{\partial \xi} \frac{d\xi}{dx}.$$

Summary

- Pseudospectral methods are based on discrete function approximations that allow exact interpolation at so-called collocation points. The most prominent examples are the Fourier method based on trigonometric basis functions and the Chebyshev method based on Chebyshev polynomials.
- The Fourier method can be interpreted as an application of discrete Fourier series on a regular-spaced grid. The space derivatives can be obtained exactly (except for rounding errors). Derivatives can be efficiently calculated with the discrete Fourier transform requiring n log n operations.
- The Fourier method implicitly assumes periodic behavior. Boundary conditions like the free surface or absorbing behaviour are difficult to implement.
- The Chebyshev method is based on the description of spatial fields using Chebyshev polynomials defined
 in the interval [-1,1] (easily generalized to arbitrary domain sizes). Exact interpolation is possible when
 the discrete fields are defined at the Chebyshev collocation points given by x_i = cos(πi/n), i = 0,...,n.
 Therefore, the derivatives can also be evaluated exactly and errors accumulate only due to the
 finite-difference time extrapolation.

Summary

- Because of the grid densification at the boundaries of the Chebyshev collocation points very small time steps are required for stable simulations when *n* is large. This can be avoided by stretching the grids by a coordinate transformation.
- A main advantage of the Chebyshev method is an elegant formulation of boundary conditions (free surface or absorbing) through the definition of so-called characteristic variables.
- Pseudospectral methods have isotropic errors. Therefore they lend themselves to the study of physical anisotropy.
- The derivative operations of pseudospectral methods are of a global nature. That means every point on a
 spatial grid contributes to the gradient. While this is the basis for the high precision, it creates problems
 when implementing pseudospectral algorithms on parallel computers with distributed memory
 architectures. As communication is usually the bottleneck, efficient and scalable parallelization of
 pseudospectral methods is difficult.