

Numerical Methods in Holonomic Gradient Method (HGM)

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- TYZ[10] N.Takayama, T.Yaguchi, Y.Zhang, Comparison of Numerical Solvers for Differential Equations for Holonomic Gradient Method in Statistics,

<https://arxiv.org/abs/2111.10947>

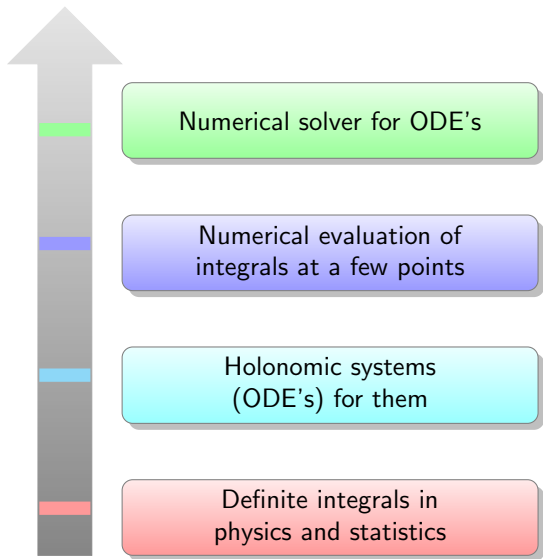
- OpenXM-hgm[7]

<http://www.math.kobe-u.ac.jp/OpenXM/Math/hgm/ref-hgm.html>

- chebfun[2] <https://chebfun.org>

- <http://www.math.kobe-u.ac.jp/OpenXM/Math/defusing/ref.html>

Sample codes.



What is a difficulty in numerical solver in HGM?

The ODE may contain solutions $f(t)$ such that

$$f(t) \gg Z(t) (\text{normalizing constant, ...})$$

Example 1

$$\frac{dY}{dt} = \begin{pmatrix} \lambda_1 & 0 \\ 0 & -\lambda_2 \end{pmatrix} Y$$

$\lambda_1 > 0 > -\lambda_2$. We assume $Z(t) = Y_1(t) + Y_2(t) \sim \exp(-\lambda_2 t)$.
A small numerical error ε in the initial condition

$$Y(0) = (\varepsilon, 1)^T$$

gives the solution $Y(t) = (\varepsilon \exp(\lambda_1 t), \exp(-\lambda_2 t))^T$ and then

$$Y_1(t) + Y_2(t) = \varepsilon \exp(\lambda_1 t) + \exp(-\lambda_2 t)$$

Example 2

(Airy function, running example 1)

$$\frac{d^2y}{dt^2} - ty = 0 \quad (1)$$

$$\text{Ai}(t) \sim \frac{1}{2\sqrt{\pi}t^{1/4}} \exp\left(-\frac{2}{3}t^{3/2}\right) O(1)$$

$$\text{Bi}(t) \sim \frac{1}{\sqrt{\pi}t^{1/4}} \exp\left(\frac{2}{3}t^{3/2}\right) O(1)$$

https://en.wikipedia.org/wiki/Airy_function

The initial value problem to obtain $\text{Ai}(t)$ will have the difficulty.

Example 3

($H_n^k(x, y)$, running example 2) Let n and k be positive integers.
([OpenXM/Math/defusing/Hkn/19-a19-n-pf.rr](https://openxm.math.defusing/hkn/19-a19-n-pf.rr))

$$H_n^k(x, y) = \int_0^x t^k \exp(-t) {}_0F_1(; n; yt) dt \quad (2)$$

$$= \frac{\Gamma(n)}{\sqrt{\pi}\Gamma(n-1/2)} \int_{D(x)} t^k (1-s^2)^{n-3/2} \exp(-t-2s\sqrt{yt}) dt ds$$

where $D(x) = \{(t, s) \in [0, x] \times [-1, 1]\}$ (3)

This function appears in studies of the outage probability of MIMO WiFi systems KA[6]. The function $H_n^k(x, y)$ is annihilated by the following ordinary differential operator w.r.t y .

$$\begin{aligned} & y^2 \partial_y^4 + (-y + 2n + 2)y \partial_y^3 \\ & + (-yx + (-k - n - 3)y + n(n + 1)) \partial_y^2 \\ & + ((y - n)x - n(k + 2)) \partial_y + (k + 1)x \end{aligned} \quad (4)$$

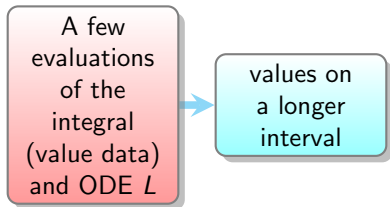
Initial value problem.

1. Runge-Kutta methods work in a short range. Implicit Runge-Kutta methods work in a longer range, but are not enough.
2. Geometric integrators like symplectic methods cannot be applied in most cases.

Boundary value problem.

1. A naive approaches do not work well.

Sparse interpolation/extrapolation methods



1. Chebyshev function method Trefethen[12], `chebfun`[2].
2. Minimizing $\int_D |Lf|^2 dt$ with constraints by value data¹.

¹Perhaps it is well-known and used in numerical analysis, but it seems not to be well-known in HGM community.

Chebyshev function method, chebfun[2]

The chebfun project was initiated in 2002 by Lloyd N. Trefethen and his student Zachary Battles.

<https://en.wikipedia.org/wiki/Chebfun>.

The n -th Chebyshev function (polynomial) is

$$T_n(x) = \cos(n\theta), \quad x = \cos \theta \quad (5)$$

The extreme points of the curve $y = T_n(x)$ in $[-1, 1]$, which we mean points that take the value $y = 1$ or $y = -1$, are called Chebyshev points (of the second kind) of T_n . For example, $T_2(x) = 2x^2 - 1$, the Chebyshev points are $\{-1, 0, 1\}$.

$T_3(x) = 4x^3 - 3x$, $\{-1, -0.5, 0.5, 1\}$.

Chebyshev interpolant

Let $f(x)$ be a function. Fix the set of Chebyshev points for $T_n(x)$. Let the value of f at Chebyshev point x_j be f_j . The Chebyshev interpolant is

$$p(x) = \sum_{j=0}^{n'} \frac{(-1)^j f_j}{x - x_j} / \sum_{j=0}^{n'} \frac{(-1)^j}{x - x_j} \quad (6)$$

The primes on the summation signs signify that the terms $j = 0$ and $j = n$ are multiplied by $1/2$.

$p(x_j) = f_j$. Degree n polynomial.

Convergence rate

$a_k = \frac{2}{\pi} \int_{-1}^1 \frac{f(x)T_k(x)}{\sqrt{1-x^2}} dx$. $f = \sum_{k=0}^{\infty} a_k T_k(x)$ when f is Lipschitz continuous.

Theorem 4

(Bernstein 1911, 1912. See, e.g., Th 8.2, Th 8.3 in Trefethen[12])
If f is analytic on $[-1, 1]$, its Chebyshev coefficients a_k decrease geometrically. If f is analytic and $|f| \leq M$ in the Bernstein ρ -ellipse² about $[-1, 1]$, then $|a_k| < 2M\rho^{-k}$. The degree n Chebyshev interpolant has accuracy $O(M\rho^{-n})$ by the sup norm.

²The radius ρ circle in the z -plane. Map it by $x = (z + z^{-1})/2$ and then we obtain the Bernstein ρ -ellipse:

chebmat $M(n - m, n; s)$ chebfun[2]

Let X be the set of the n chebyshev points (of the second kind) for the Chebyshev function T_{n-1} .

$\ell_j(X; t)$: the j -th polynomial of the Lagrange interpolation for X .
Let Y be the set of the $(n - m)$ Chebyshev points where $m \geq 0$ ³.

Definition 5

chebfun[2] $M(n - m, n; s)$: $(n - m) \times n$ matrix with (i, j) entries

$$\sum_{k=0}^{n-m-1} \ell_k(Y; Y_i) \ell_j^{(s)}(X; Y_k) \quad (7)$$

When $f(t)$ is the Chebyshev interpolant w.r.t. X ,

$$f^{(s)}(Y_i) = (i\text{-th row of } M(n - m, n; s)) \cdot (f_0, \dots, f_{n-1})^T$$

³We approximate $f(t)$ by the values at Y , which is called “down-sampling” in DH2016[3].

From ODE to a (dense) matrix equation

Example 6

The Airy equation

$$f'' - tf = 0$$

Symbolically, we solve

$$(M(n-2, n; 2) - \text{diag}(Y)M(n-2, n; 0))F = 0 \quad (8)$$

where $F = [f_0, f_1, \dots, f_{n-1}]^T$ with given, e.g., values of f_0 and f_{n-1} (boundary values) or values of f_0 and the first entry of $M(n-2, n; 1)(f_0, \dots, f_n)^T$ (initial values f and f').

See <https://www.chebfun.org/examples/ode-linear/SpectralDisc.html>.

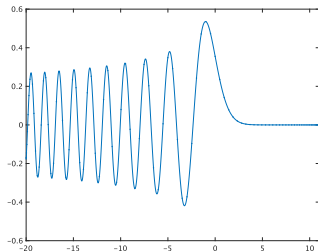
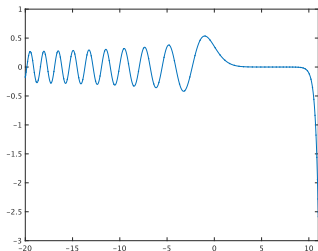


Figure: Solving the Airy differential equation by chebfun

Initial value problem for Airy $Ai(t)$. ([OpenXM/Math/defusing/intro/y2023_07_16_airy_initial_value.m](#))

$Ai(-20) = -0.176406127077984689590192292219$, $Ai'(-20) = 0.892862856736471238398409934114$

Chebfun gives reasonable values⁴ upto $t = 9$, but divergent values appear when t is larger than 9. The left graph of Figure 1.

Boundary value problem for Airy $Ai(t)$. ([OpenXM/Math/defusing/intro/y2023_07_16_airy_boundary_value.m](#))

$Ai(-20) = -0.176406127077984689590192292219$, $Ai(-11) = 4.22627586496035959129883545080 \times 10^{-12}$.

Divergent values do not appear. See the right graph of Figure 1.

⁴Values are evaluated by Mathematica.

$$H_n^k(x, y)$$

Example 7

Boundary value problem for $H_n^k(x, y)$ for $x = 1$ and $y \in [10^8, 10^8 + 2 \times 10^5]$.

We give the boundary values of $H_1^{10}(1, y)$ and $\frac{\partial H_1^{10}}{\partial y}(1, y)$ at $y = 10^8$ and $y = 10^8 + 2 \times 10^5$. We apply the chebfun package for this boundary value problem.

([OpenXM/Math/defusing/Hkn/y2023-07-25_hkn_valid10power8.m](#))

To check the accuracy, we compare the values by the chebfun package and by the numerical integration by Mathematica at $y = 10^8 + 200$. The chebfun package keeps 4 digits accuracy at the point and the ODE is solved in $1.66s^5$. On the otherhand, the numerical integration by Mathematica (2022)

([OpenXM/Math/defusing/Hkn/2023-07-09-hkn-int.m](#)) took $23.58s^6$.

⁵Apple M1, 2020, Matlab 2022b

⁶AMD EPYC 7552 48-Core Processor, 1499.534MHz

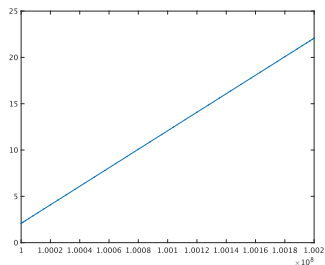
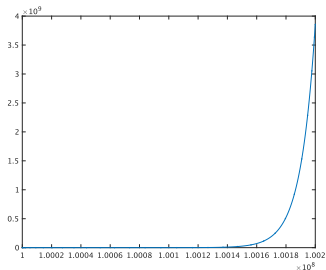


Figure: Left: $H_1^{10}(1, y)$. Right: $\log H_1^{10}(1, y)$. Values should be magnified by 10^{8678} .

Sparse interpolation/extrapolation TYZ[10]

Known: $Lf = 0$ (ODE), $f(p_i) = q_i$ for some points p_i 's. $\{e_j\}$: a set of basis functions. Put $f(t) = \sum_{k=0}^M f_k e_k(t)$ (unknown constants f_j 's). Minimize

$$\int_a^b |Lf(t)|^2 dt, \quad f(p_i) = q_i, \quad i = 1, 2, \dots \quad (9)$$

A numerical integration for a function g :

$$I_N(g) = \sum_{j=0}^N T_j g(t_j) \quad (10)$$

where $t_0 = a < t_1 < \dots < t_{N-1} < t_N = b$ and $T_j \in \mathbf{R}_{\geq 0}$. Fix it. Then, the loss function is

$$\begin{aligned} \ell(\{f_k\}) &:= \sum_{j=0}^N |(Lf)(t_j)|^2 T_j \\ &= \sum_{j=0}^N \left| \sqrt{T_j} \sum_{k=0}^M f_k (Le_k)(t_j) \right|^2 \end{aligned} \quad (11)$$

We minimize it under $f(p_i) = q_i$ (least square for the data $(Le_k)(t_j) \sqrt{T_j}$).

Chebyshev function method as a sparse interpolation

The Chebyshev function method can be regarded as a special case of this method. The numerical integration scheme of the Chebyshev quadrature:

$$\int_{-1}^1 \sqrt{1-t^2} g(t) dt \sim \sum_{i=1}^{n-2} w_i g(Y_i) \quad (12)$$

where Y is the set of the Chebyshev points for T_{n-1} and the weight w_i is

$$w_i = \frac{\pi}{n-1} \sin^2 \left(\frac{i}{n-1} \pi \right)$$

Put $g(t) = |Lf|^2$. Since the left hand side of (8) are values at the set of Chebyshev points Y , assuming it is equal to the zero vector is equivalent to that the integral by the Chebyshev quadrature over Y is equal to zero.

Todo

A different solver with validation and Chebyshev functions is proposed in BBJ2018[1].

The advantage of the method is that matrices in the solver are banded and validation is given. We will test this method for the HGM as a next try.

Example 8

$E[\chi(M_x)]$, TJKZ2020[11] (Expectation of Euler characteristic of random manifolds).

Extrapolation of some values near $t = 4.8$ by the sparse interpolation/extrapolation method; The degree 29 polynomial and the rectangle integration is used for a rank 11 ODE (26KB).

<https://colab.research.google.com/drive/1XhysmF1DMZfAhTt10tc9A7tFYRBeI6tI?usp=sharing> See Figure 3.

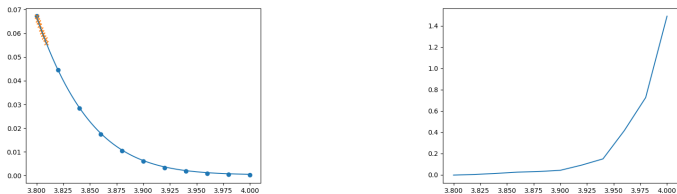


Figure: The graph of $F_{29}(t)$ and simulation values in the left and relative errors in the right. The data points are marked with 'x'.

The initial value problem of the ODE

$$\frac{dF}{dt} = P(t)F \quad (13)$$

$$F(t_0) = F_0^{\text{true}} \in \mathbf{R}^r \quad (14)$$

where $P(t)$ is an $r \times r$ matrix, $F(t)$ is a column vector function of size r , and F_0^{true} is the initial value of F at $t = t_0$.

Situation 1

1. The initial value has at most 3 digits of accuracy. We denote this initial value F_0 .
2. The property $|F| \rightarrow 0$ when $t \rightarrow +\infty$ is known, e.g., from a background of the statistics.
3. There exists a solution \tilde{F} of (13) such that $|\tilde{F}| \rightarrow +\infty$ or non-zero finite value when $t \rightarrow +\infty$.

Defusing method

Numerical schemes such as the Runge-Kutta method obtain a numerical solution by the recurrence

$$F_{k+1} = Q(k, h)F_k \quad (15)$$

from F_0 where $Q(k, h)$ is an $r \times r$ matrix determined by a numerical scheme and h is a small number. The vector F_k is an approximate value of $F(t)$ at $t = t_k = t_0 + hk$. Let N be a suitable natural number and put

$$Q = Q(N-1, h)Q(N-2, h) \cdots Q(1, h)Q(0, h) \quad (16)$$

We call Q the *matrix factorial* of $Q(k, h)$. The matrix Q approximates the fundamental solution matrix of the ODE.

Project F_0 to eigenspaces of negative eigenvalues.

Algorithm 1

1. Obtain eigenvalues $\lambda_1 > \lambda_2 > \dots > \lambda_r > 0$ of Q and the corresponding eigenvectors v_1, \dots, v_r .
2. Let λ_m be the first negative eigenvalue.
3. Express the initial value vector F_0 containing errors in terms of v_i 's as

$$F_0 = f_1 v_1 + \dots + f_r v_r, \quad f_i \in \mathbf{R} \quad (17)$$

4. Choose a constant c such that $F'_0 := c(f_m v_m + \dots + f_r v_r)$ approximates F_0 .
5. Determine F_N by $F_N = QF'_0$ with the new initial value vector F'_0 .

Example 9

Solving Airy differential equation by the defusing method.

([OpenXM/Math/defusing/intro/2023-07-21-airy.rr](#)) Give initial values at $t = -20$ as

$F_0 = [-0.17640612707798468959, 0.89286285673647123840]$

($\text{Ai}[-20]$ and $\text{Ai}'[-20]$).

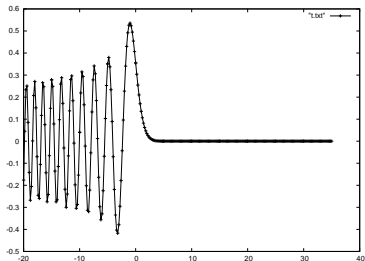


Figure: Solving initial value problem, $t \in [-20, 30]$

Example 10

We implement the defusing method in `tk_ode_by_mpfr.rr`⁷ for the Risa/Asir [9]. It generates C codes utilizing the MPFR [8] for bigfloat and the GSL [4] for eigenvalues and eigenvectors.

We apply the defusing method for initial value problem to $H_1^{10}(1, y)$ which is a solution of the ODE (4). We apply the defusing method for a transformed ODE with a gauge function $\exp(y)y^{1-n+k}$ to make the target solution decrease to 0 when $y \rightarrow \infty$. We use the step size $h = 10^{-3}$ and the bigfloat of 30 digits of accuracy.

([OpenXM/Math/defusing/asir-tmp/tk-ode-assert.rr](#) (code generation), `tk-ode-assert.hkn1()`, `tk-ode-assert.hkn2()`) The Figure 5 shows that the adaptive Runge-Kutta method of GSL [4] fails before y becomes 30. The Figure 6 presents the relative error of values by the defusing method and exact values. It shows that the defusing method works even when $y = 10^3$.

⁷http://www.math.kobe-u.ac.jp/OpenXM/Current/doc/asir-contrib/ja/tk_ode_by_mpfr-html/tk_ode_by_mpfr-ja.html. **Todo, English manual.**

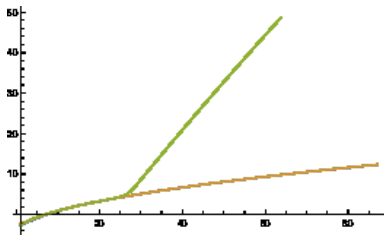


Figure: $\log H_1^{10}(1, y)$. Exact value (by numerical integration) and the value by our defusing method agree. The adaptive Runge-Kutta method with the initial relative error 10^{-20} (upper curve) does not agree with the exact value when y is larger than about 25.

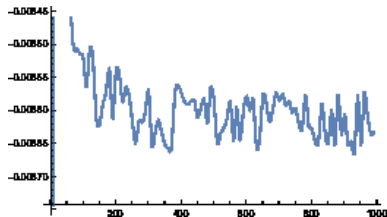












Figure: The relative error of $H_1^{10}(1, y)$ of our defusing method. The relative error is defined as $(H_d - H)/H$ where H_d is the value by the defusing method and H is the exact value.

Summary


1. The use of implicit Runge-Kutta method will be a good choice for solving ODE in a short range.
2. In order to solve unstable HGM initial value problem, the defusing method (filter method) will be a good choice.
3. In order to solve unstable HGM boundary value or sparse interpolation problem, the Chebyshev function method and other sparse interpolation method will be a good choice.
4. See also todo.

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