

Numerical Analysis & Scientific Computing II

Lesson 4

Numerical Solution of PDE

4.3 Hyperbolic PDE

- Finite Difference Methods



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Numerical Methods for PDE: Hyperbolic PDE



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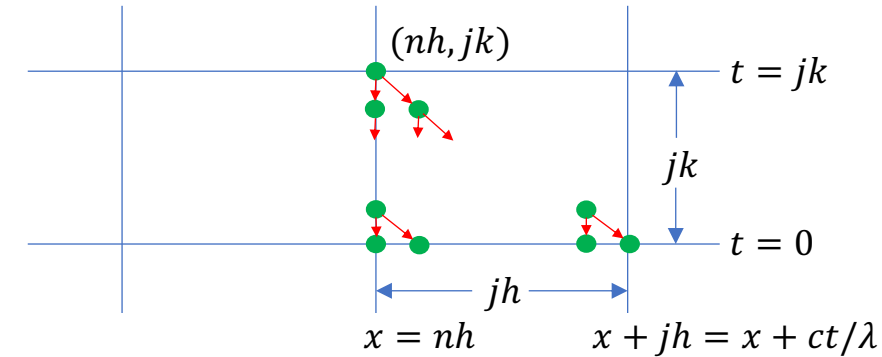
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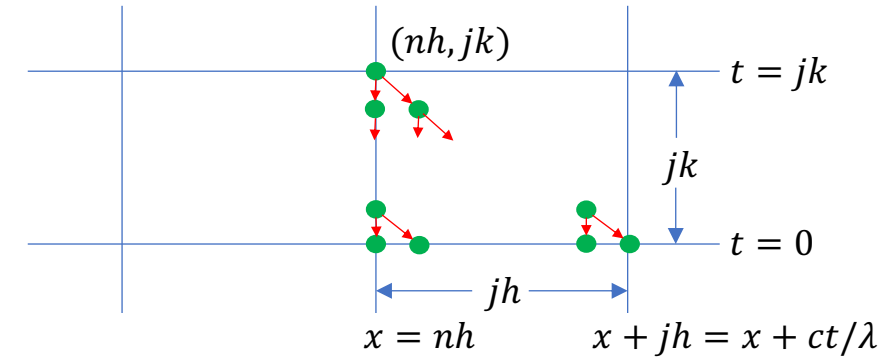
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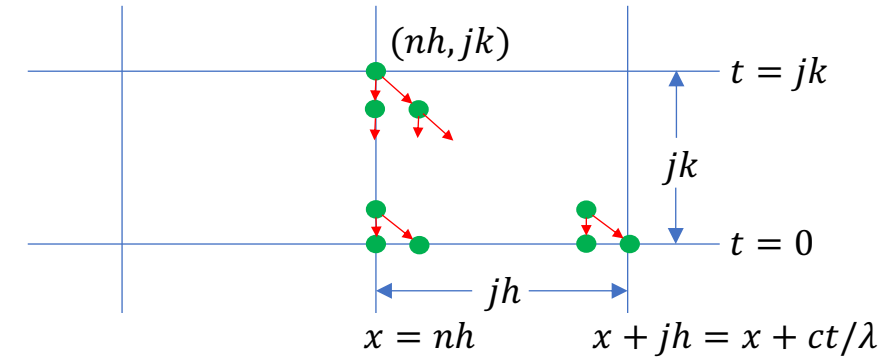
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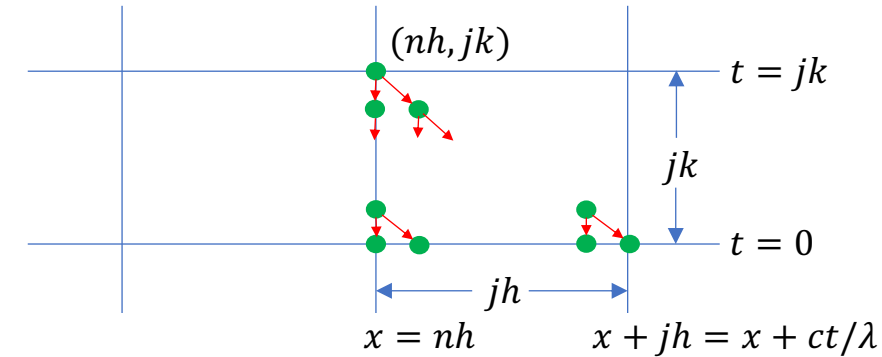
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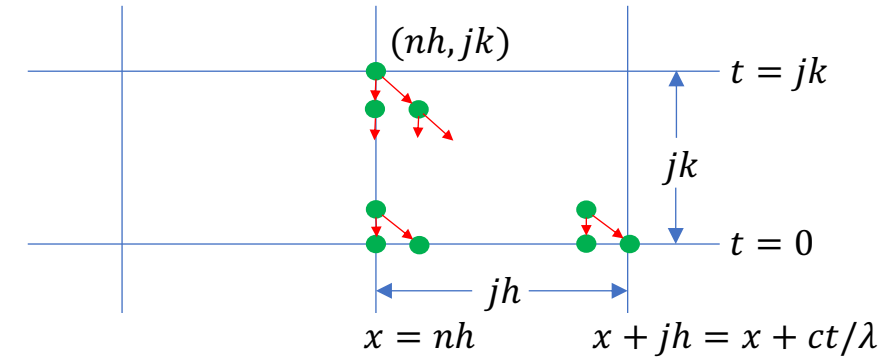
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This necessary condition, which fails for forward-forward difference method, is called the **Courant-Friedrichs-Levy condition**, or **CFL condition**.



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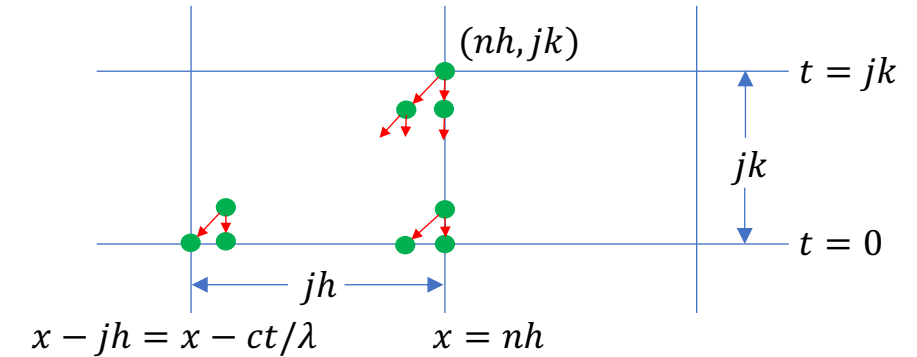
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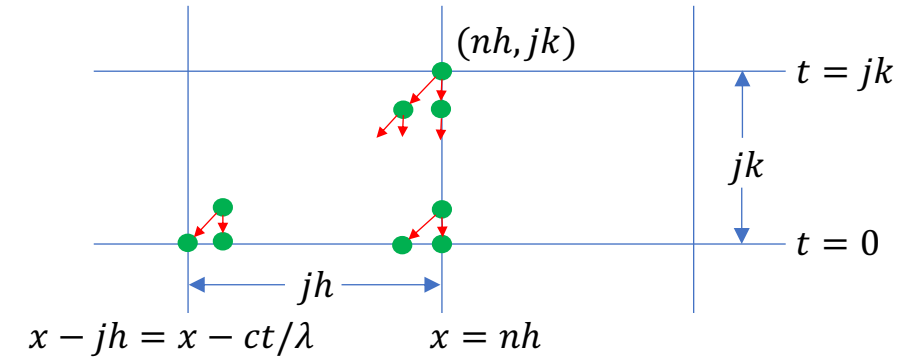
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$$x - ct \in [x - ct/\lambda, x]$$

that is,

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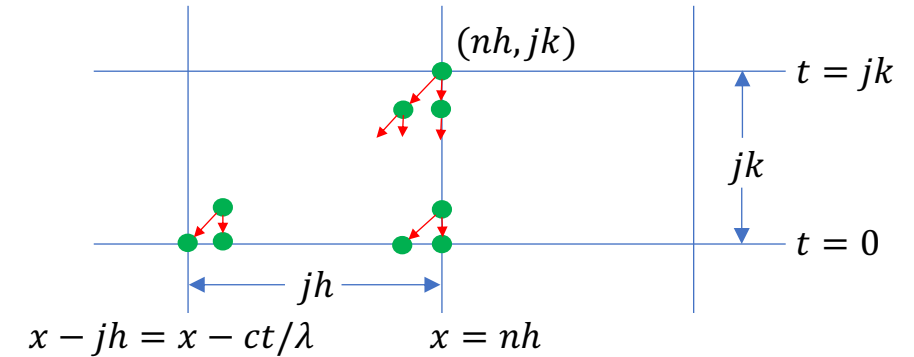
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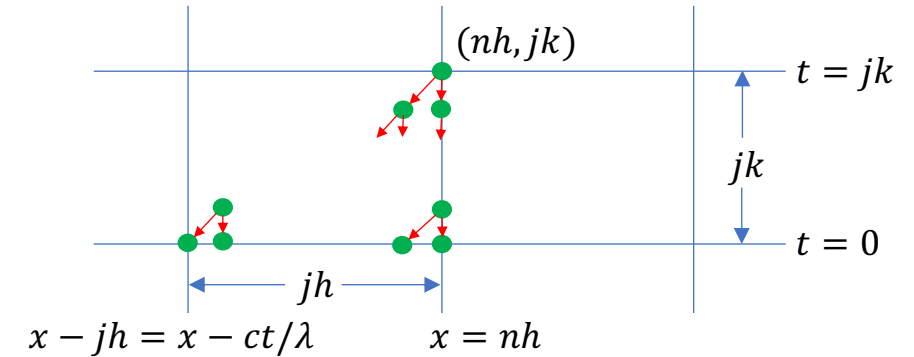
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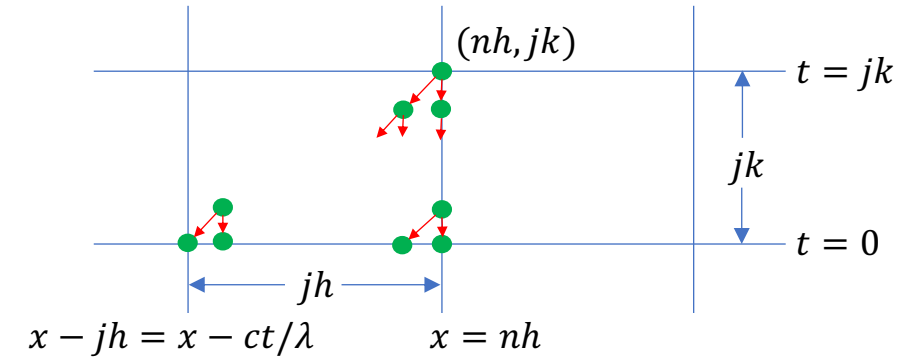
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In general, however, CFL is not sufficient for convergence. It turns out that, for forward-central scheme, while the CFL condition is $|\lambda| \leq 1$, the method is **unconditionally unstable**.



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- *Finite Difference Methods*
- ***Stability Analysis***



Numerical Methods for PDE: Hyperbolic PDE



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For simplicity, let's consider a 1-periodic problem rather than a boundary value problem:

$$\begin{aligned}\frac{\partial u}{\partial t} + c \frac{\partial u}{\partial x} &= 0, & x \in \mathbb{R}, t > 0, \\ u(x, 0) &= u_0(x), & x \in \mathbb{R}, \\ u(x+1, t) &= u(x, t), & x \in \mathbb{R}, t > 0.\end{aligned}$$



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The ψ_m are orthogonal with respect to the inner product (**exercise**)

$$\langle \phi, \psi \rangle_h = h \sum_{n=0}^{N-1} \phi(nh) \overline{\psi(nh)}.$$



Note that ψ_m is an eigenvector for the forward difference operator D_h^+ , the backward difference operator D_h^- and the centered difference operator D_h . For example,

$$D_h^- \psi_m(x) = \frac{\psi_m(x) - \psi_m(x - h)}{h} = \frac{1 - e^{-2\pi i m h}}{h} \psi_m(x).$$

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$$D_h^- \psi_m(x) = \frac{\psi_m(x) - \psi_m(x-h)}{h} = \frac{1 - e^{-2\pi i m h}}{h} \psi_m(x).$$

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As $1 - \lambda + \lambda e^{-2\pi i m h}$ describes a circle centered at $1 - \lambda$ of radius $|\lambda|$, we see that the method is stable if and only if $0 \leq \lambda \leq 1$.

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Numerical Methods for PDE: Hyperbolic PDE



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As all eigenvalues

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As the final example, we look at the Lax-Friedrichs method which can be seen as a (conditionally) stable variant of the forward-centered difference scheme and maintains $O(k + h^2)$ accuracy.



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Remark

Note that the Lax-Friedrichs scheme can be rewritten as

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Numerical Analysis & Scientific Computing II

Lesson 5

Integral Equations



Akash Anand
MATH, IIT KANPUR

Numerical Analysis & Scientific Computing II

Lesson 5

Integral Equations

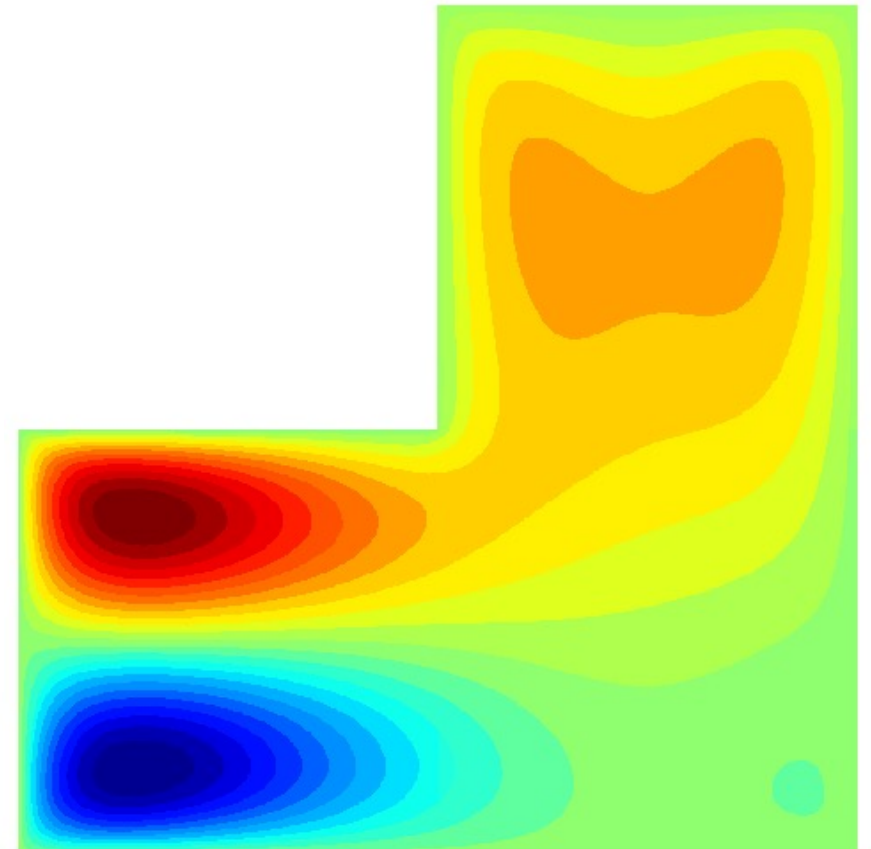
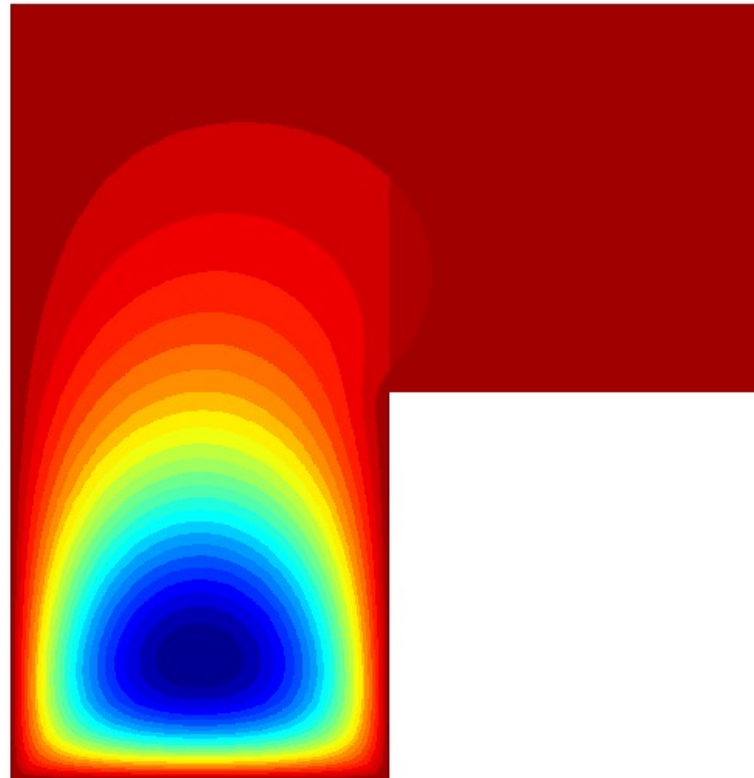
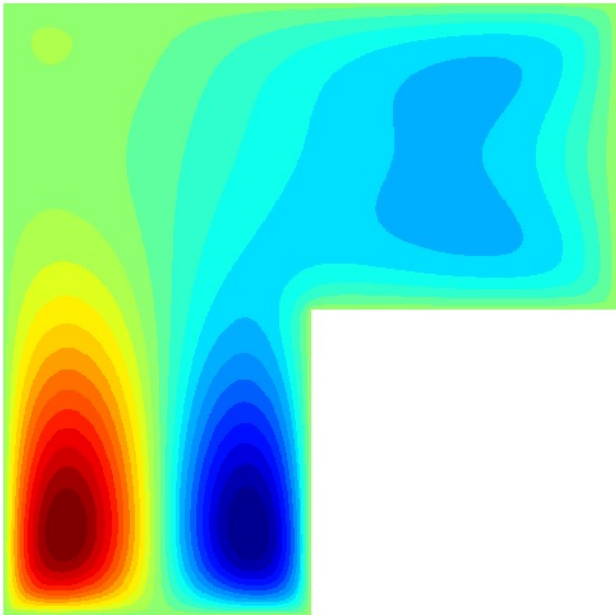
5.1 Some solutions of boundary value problems for PDEs via integral equations



Integral Equations: Some examples



Some solutions of integral Fredholm integral equations

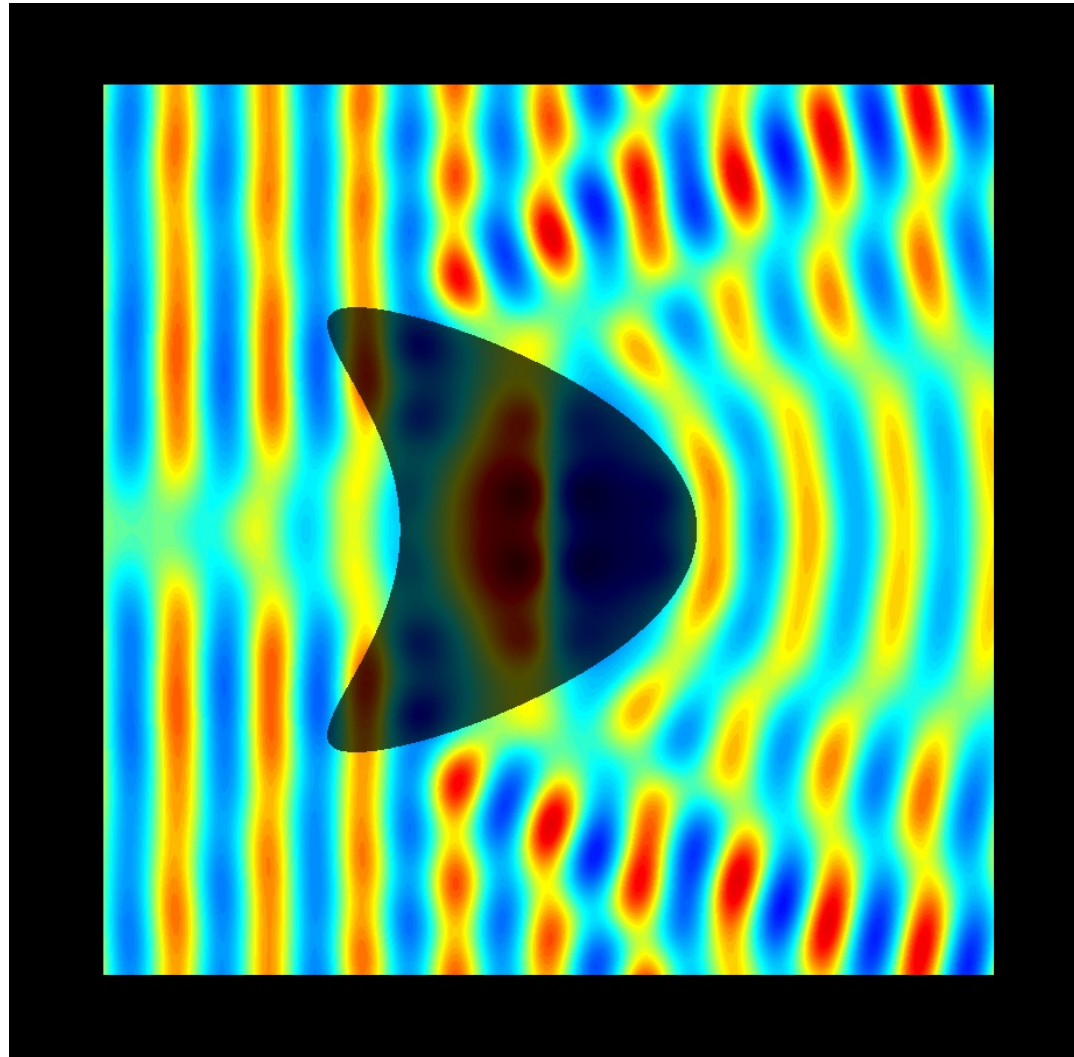


(Akash Anand)

Integral Equations: Some examples



Some solutions of integral Fredholm integral equations in wave scattering

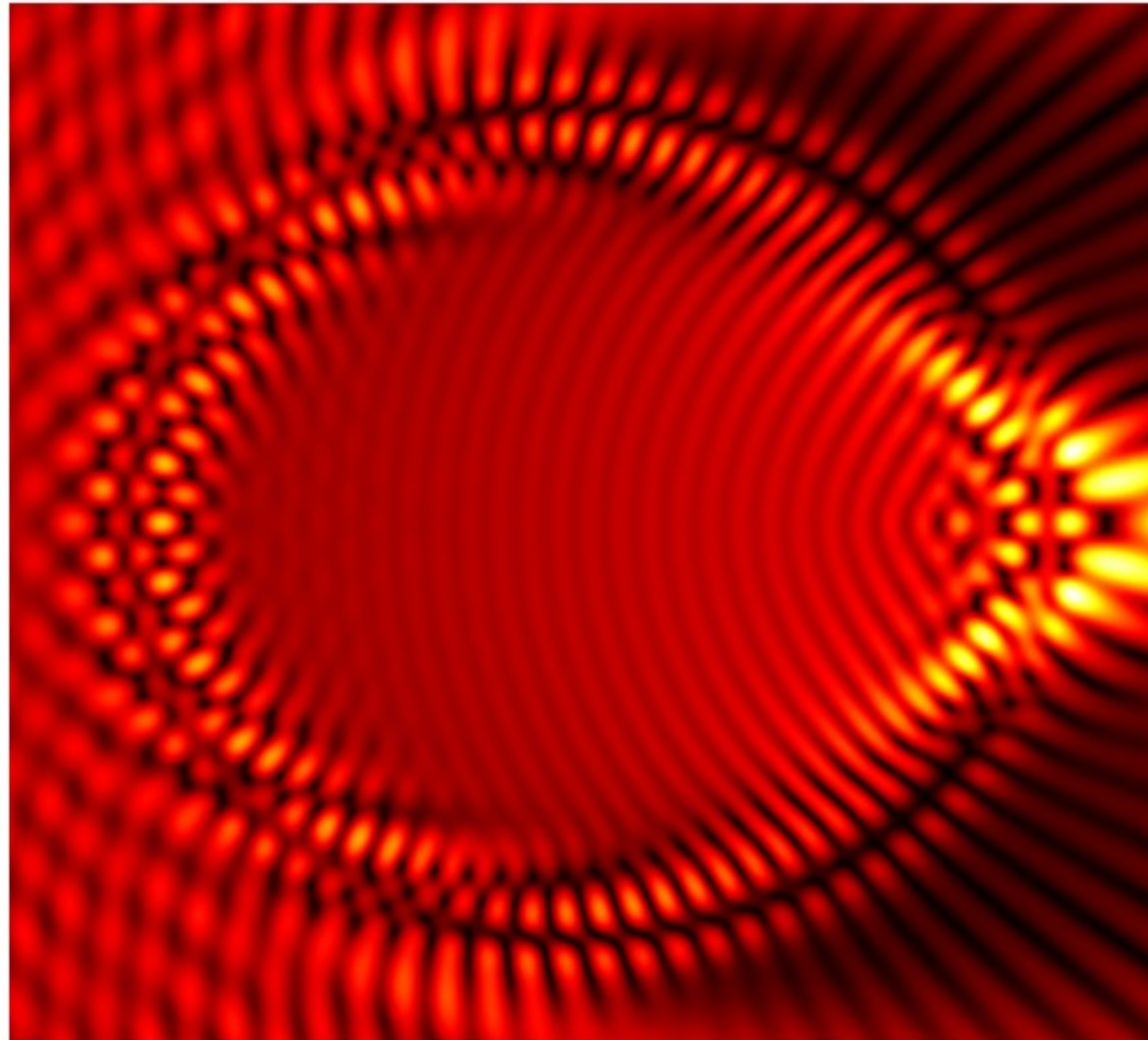


*(Ambuj Pandey,
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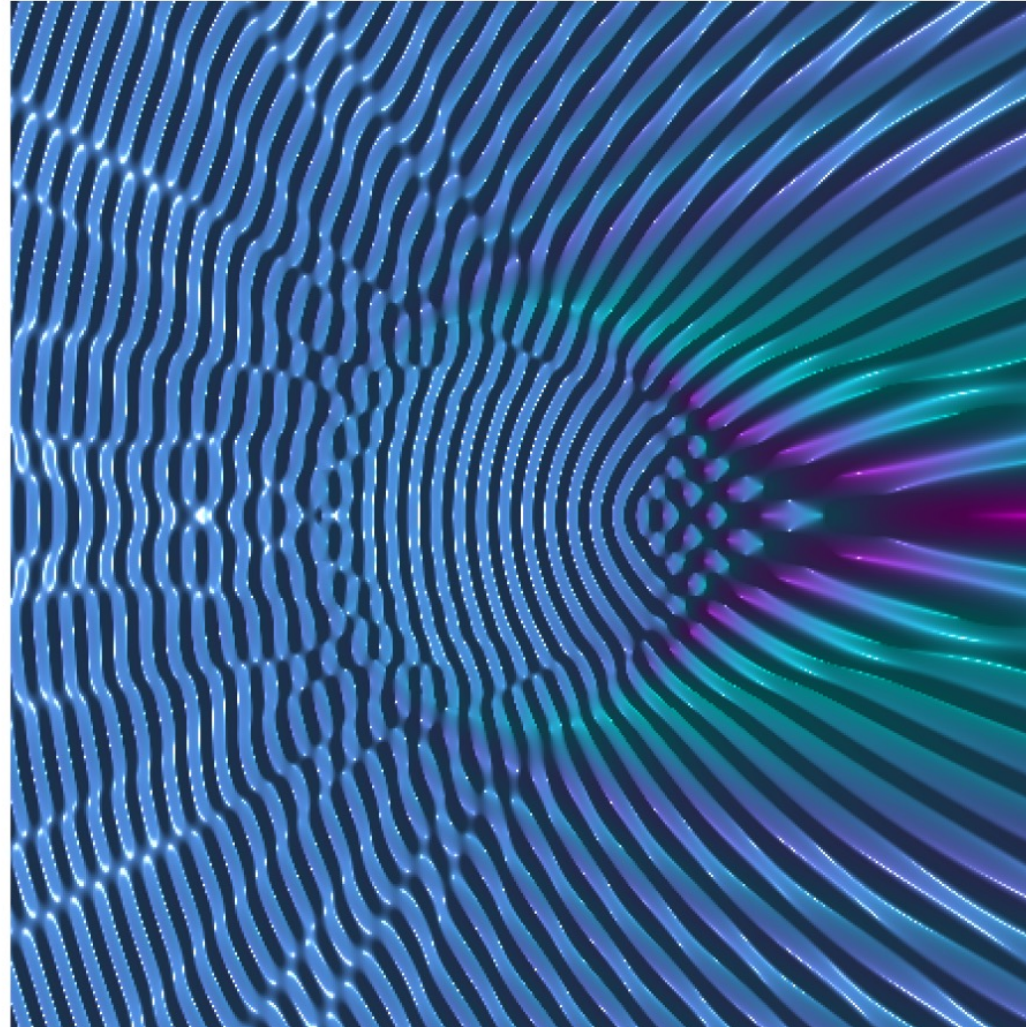


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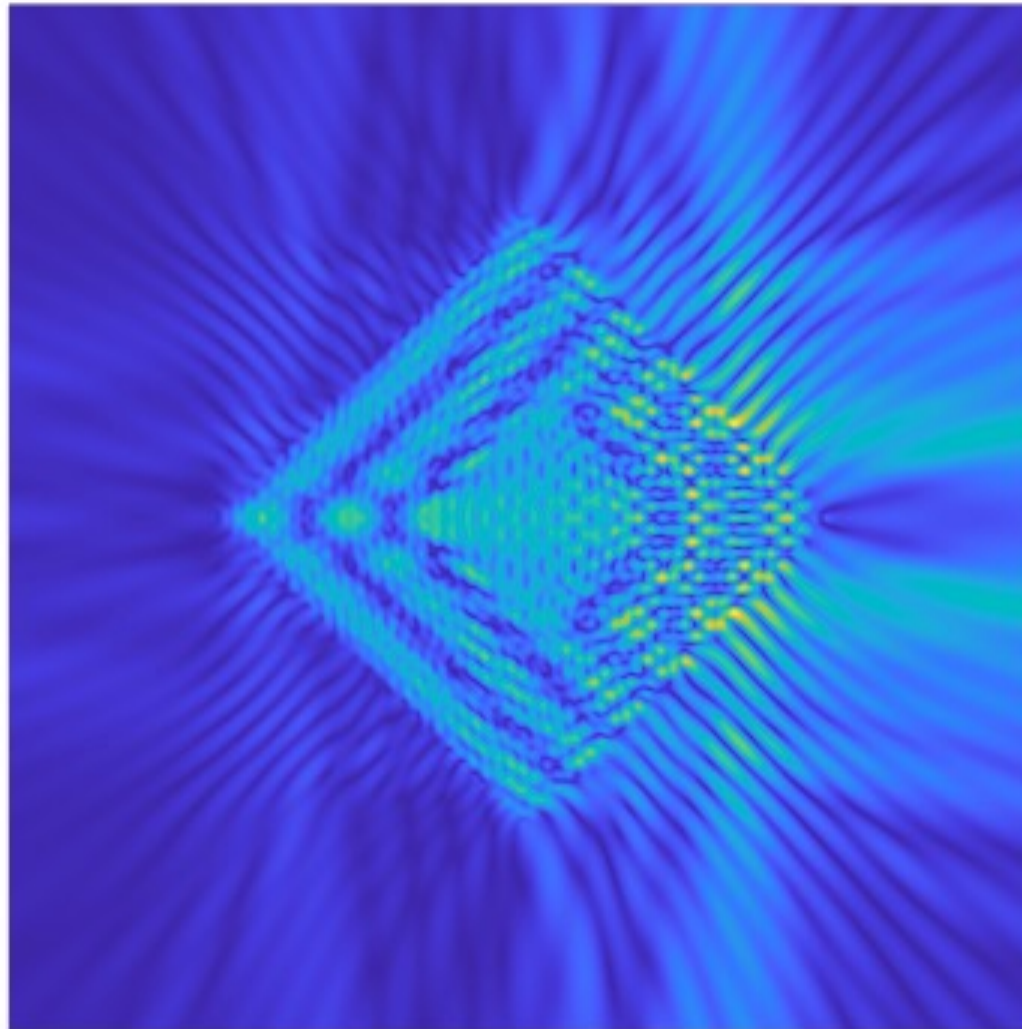


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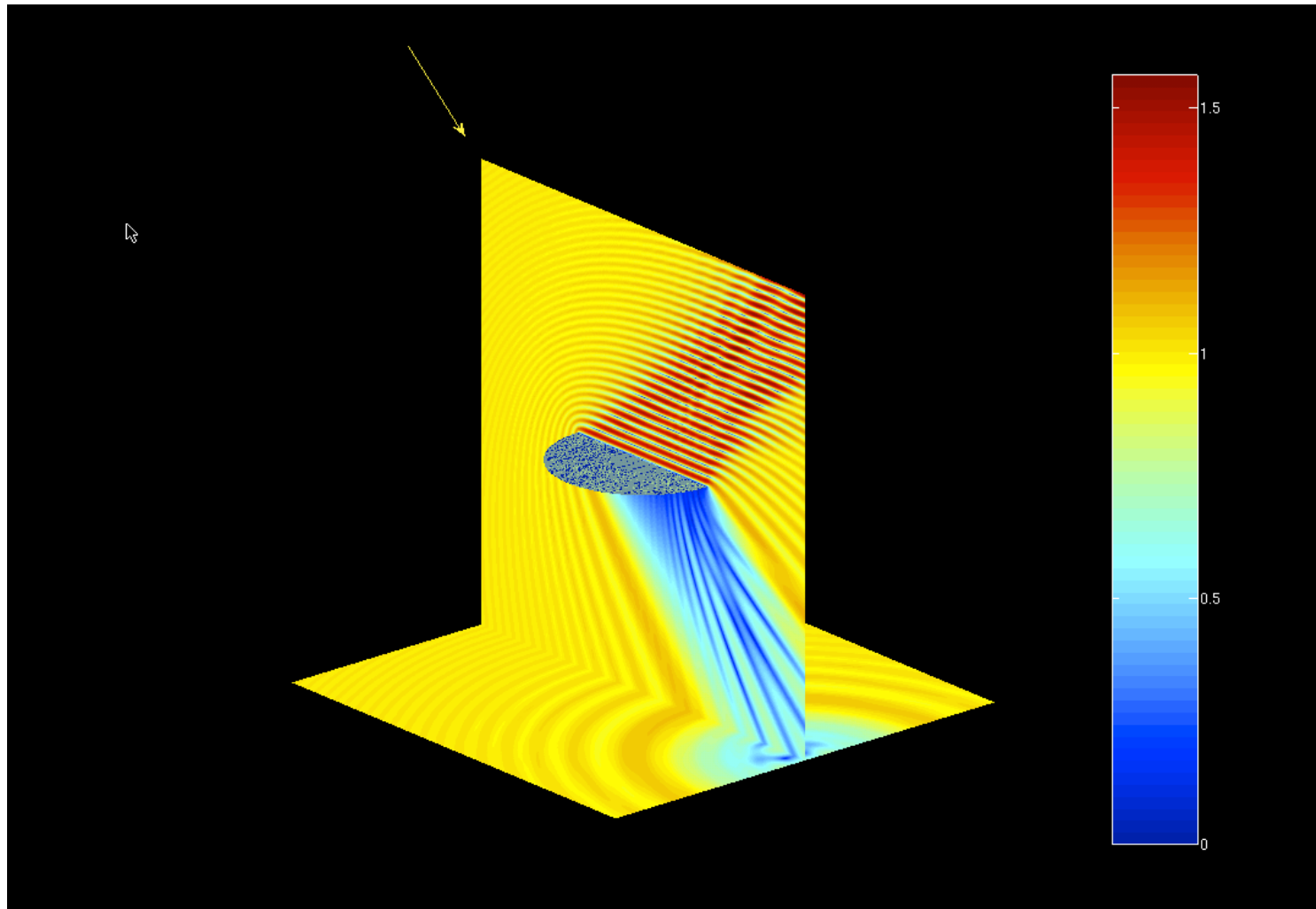


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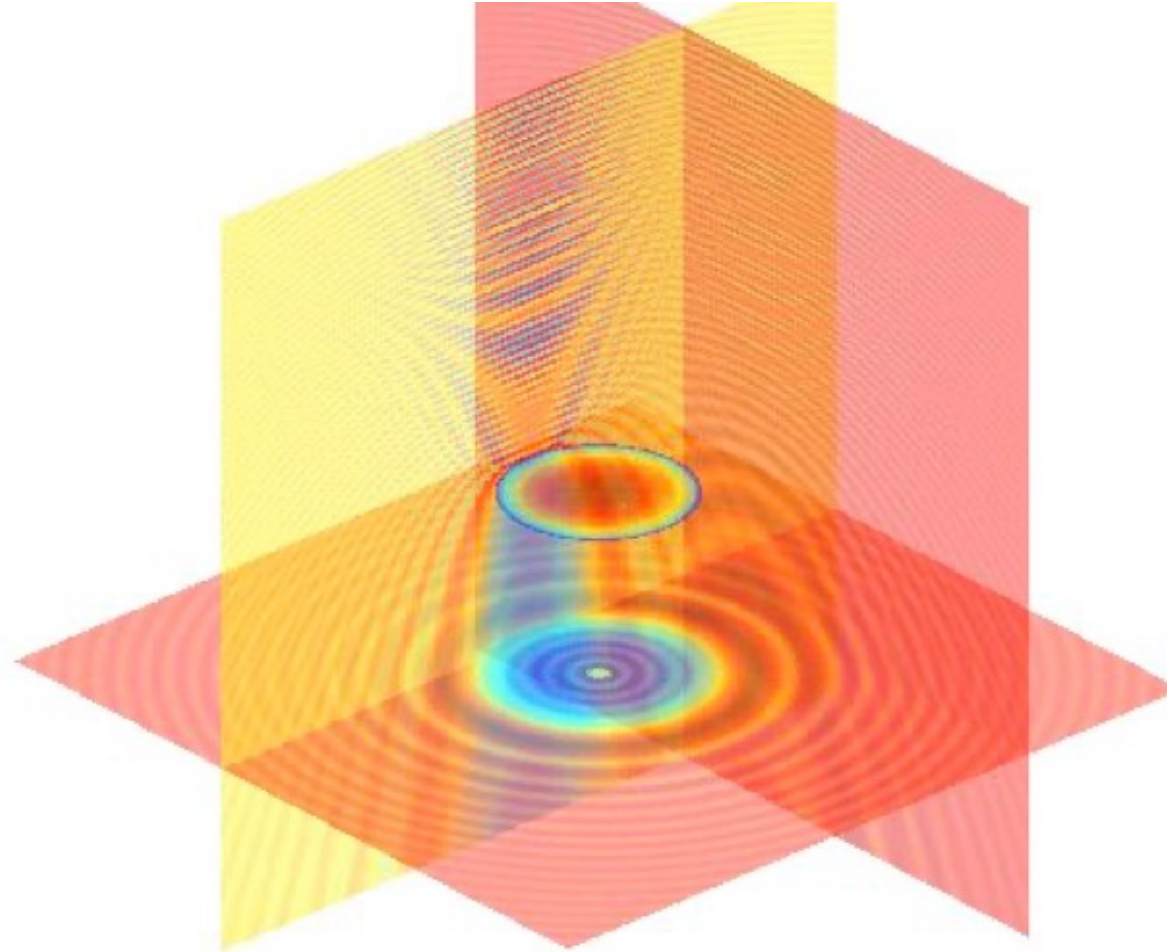


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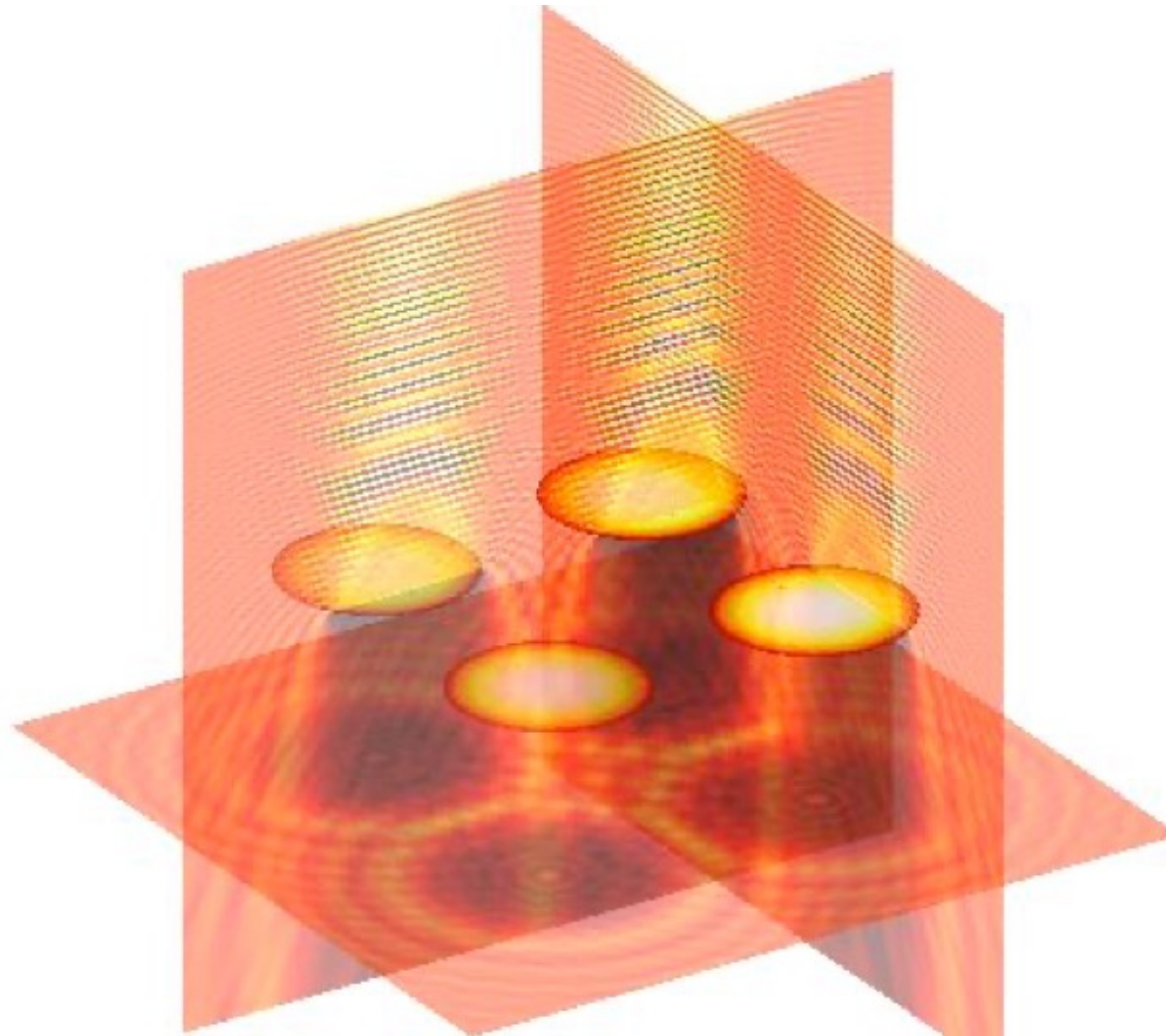


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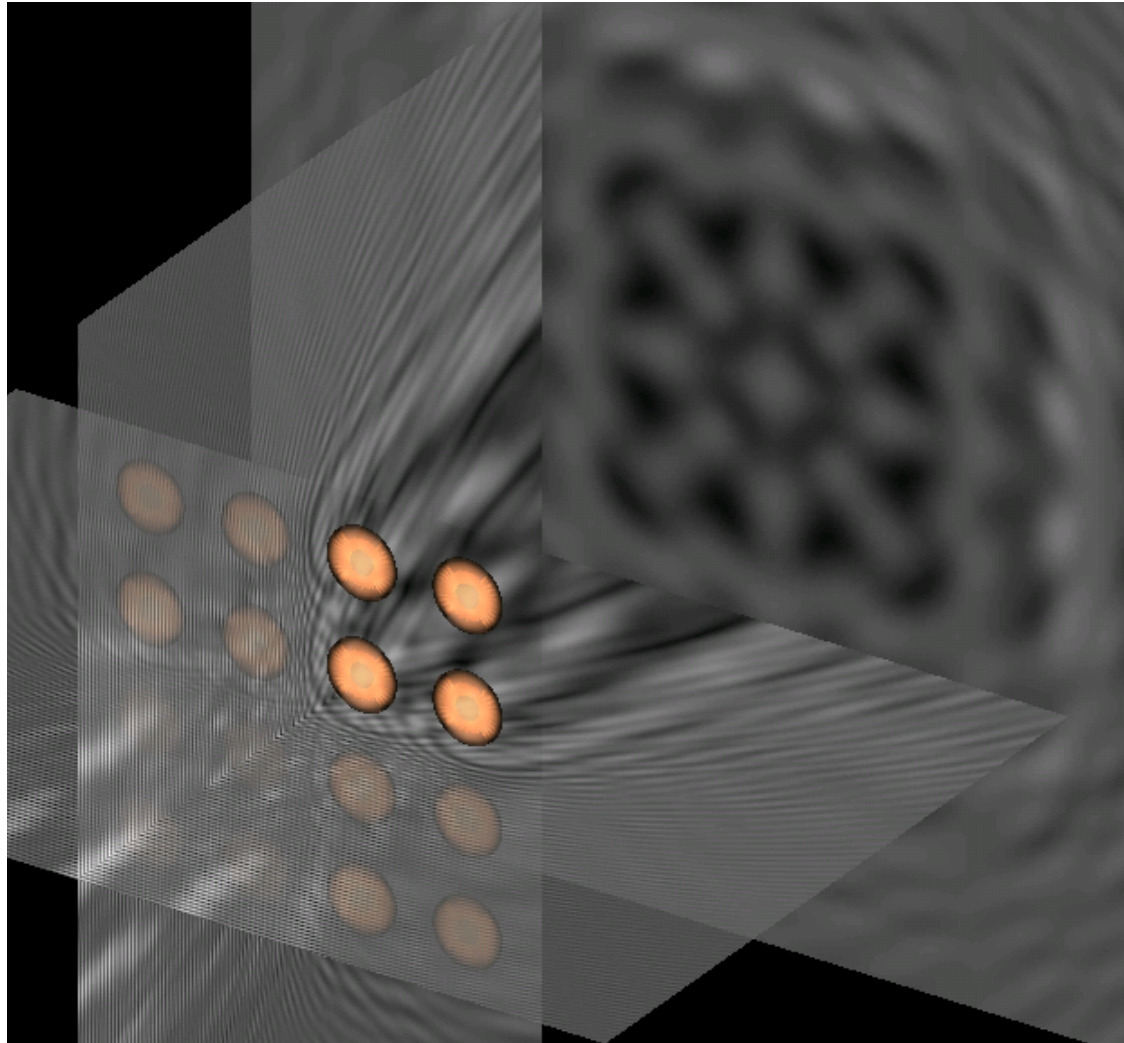


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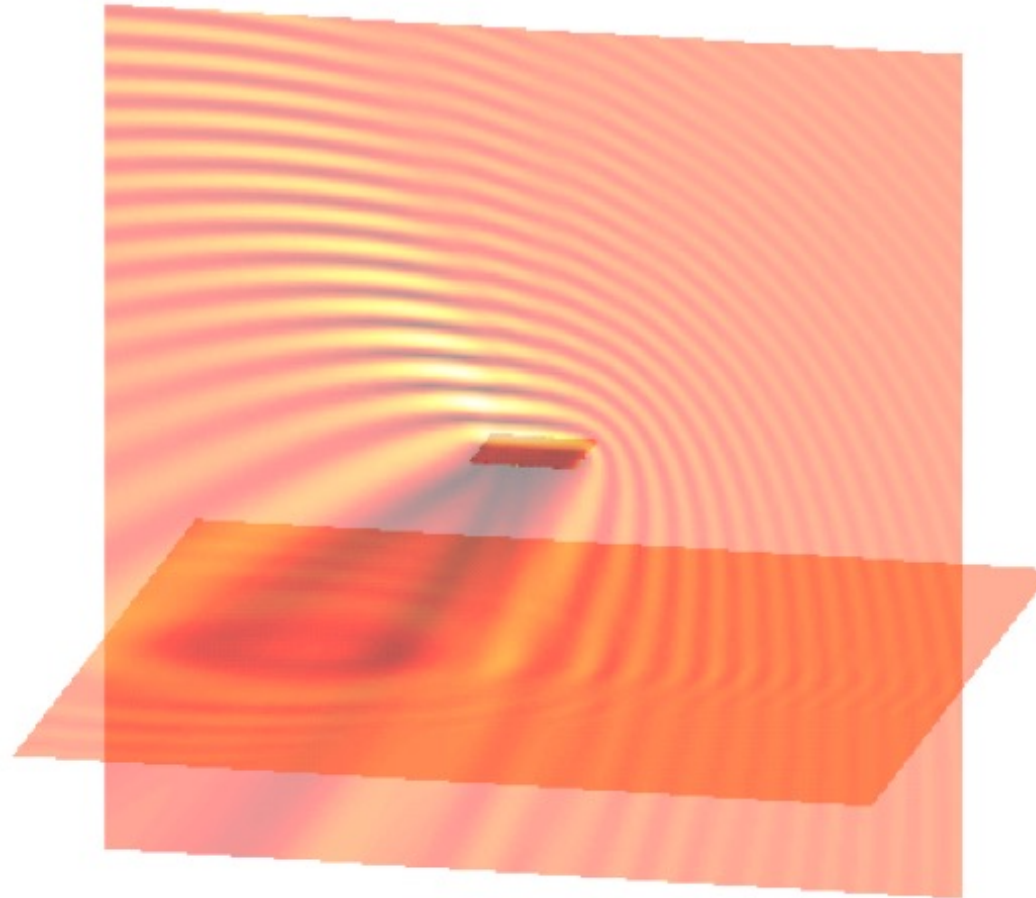


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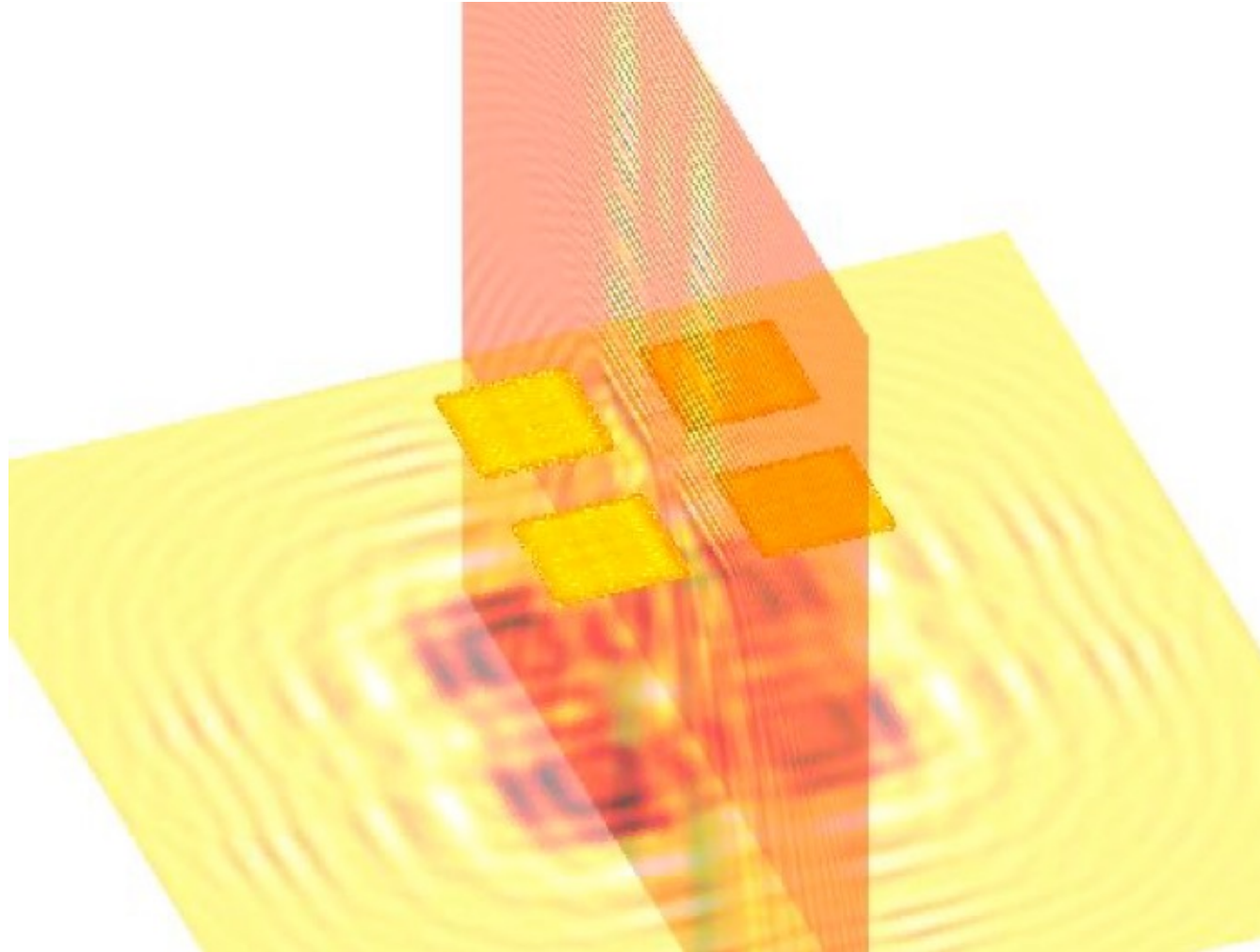


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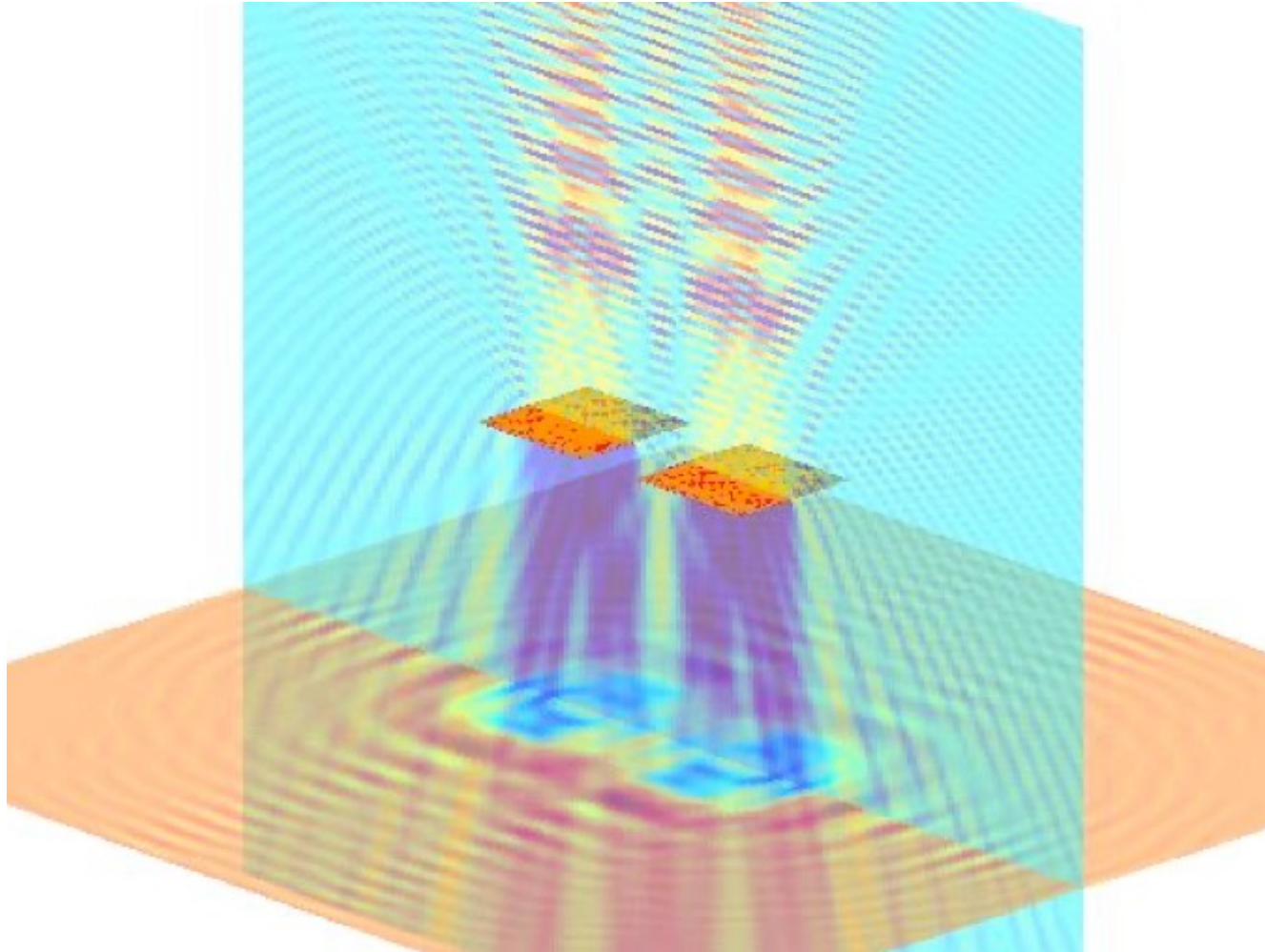


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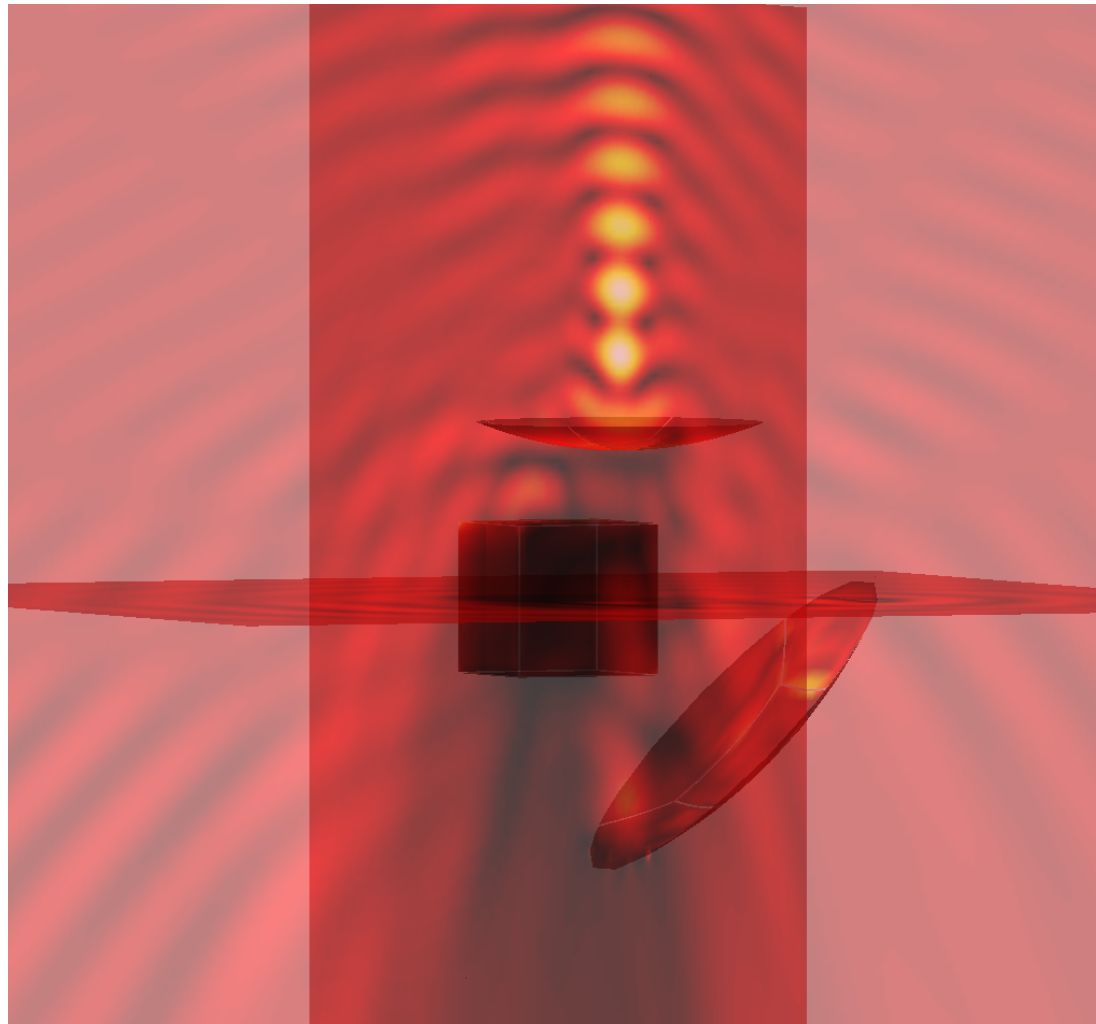


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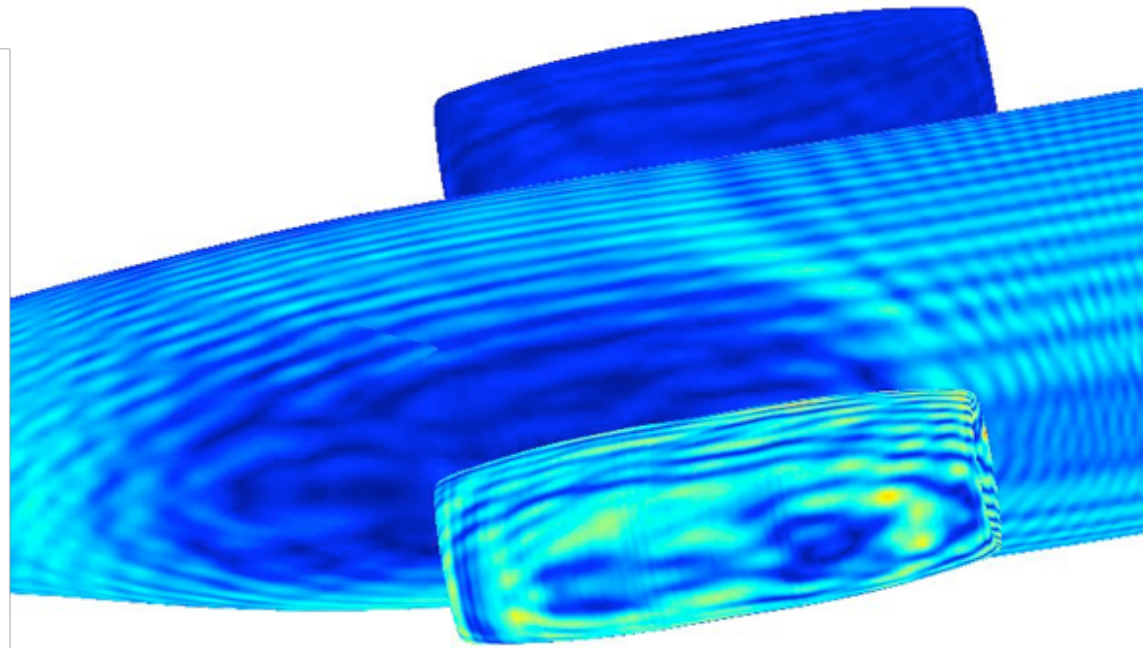
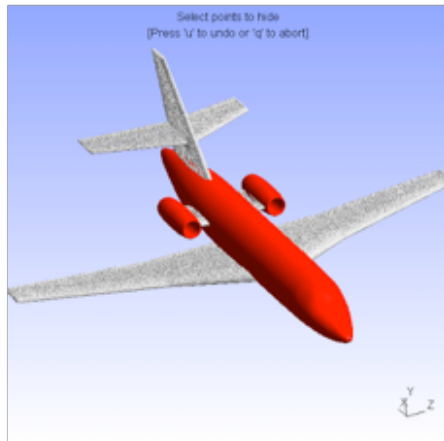
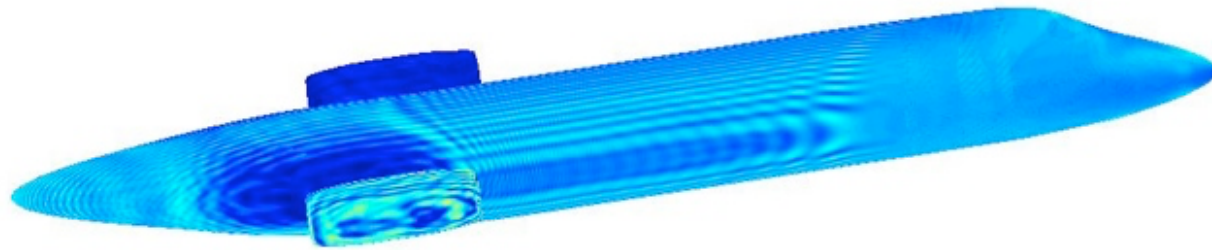


(Akash Anand)

Integral Equations: Some examples



Some solutions of integral Fredholm integral equations in wave scattering



(Akash Anand)

Numerical Analysis & Scientific Computing II

Lesson 5

Integral Equations

5.1 Some solutions of boundary value problems for PDEs via integral equations

5.2 An Introduction



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Integral Equations: An Introduction



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Recall that the initial value problem

$$y' = f(t, y), y(t_0) = y_0,$$

is equivalent to finding y satisfying

$$y(t) = y_0 + \int_{t_0}^t f(s, y(s)) ds.$$

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The difficulty with these equations, linear or non-linear, is that they are ill-conditioned.

The numerical solution of these equations are closely related to the initial value problem. We will, however, focus on a different type of integral equations known as Fredholm integral equations, in particular, of the second kind.

Integral Equations: An Introduction



The general form of such an integral equation is

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Consider solving the problem

$$\begin{aligned} \Delta u(x) &= 0, & x \in \Omega, \\ u(x) &= g(x), & x \in \Gamma, \end{aligned}$$

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$$u(x) = \int_{\Gamma} \frac{1}{|x-y|} \rho(y) dy, \quad x \in \Omega,$$

where $\rho(y)$ is called a *single layer density* function and it is the unknown in the equation.

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a Fredholm integral equation of the first kind. If we see the solution in the form of a *double layer potential*

$$u(x) = \int_{\Gamma} \frac{\partial}{\partial \nu(y)} \left(\frac{1}{|x - y|} \right) \mu(y) dy, \quad x \in \Omega,$$

then $\mu(y)$ satisfies a Fredholm integral equation of the second kind.

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then $\mu(y)$ satisfies a Fredholm integral equation of the second kind. Indeed, the **double layer density** function $\mu(y)$ satisfies

$$\frac{1}{2} \mu(x) - \int_{\Gamma} \frac{\partial}{\partial \nu(y)} \left(\frac{1}{|x - y|} \right) \mu(y) dy = -g(x), \quad x \in \Gamma,$$

(another fact from “Potential Theory” known as jump relation for double layer potential).



Integral Equations: An Introduction

We say that a kernel $K: \Omega \times \Omega \rightarrow \mathbb{C}$ is **weakly singular** if K is defined and continuous for all $x, y \in \Omega \subseteq \mathbb{R}^m$, $x \neq y$, and there exist positive constants M and $\alpha \in (0, m]$ such that

$$|K(x, y)| \leq M|x - y|^{-(m-\alpha)}$$

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Remark

One can show that, the integral operator

$$(Au)(x) = \int_{\Omega} K(x, y)u(y)dy,$$

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Moreover, it is known that, for such integral operators, the **Fredholm alternative** holds, that is,

$$(I - A)u = f$$

has a unique solution for every $f \in C(\Omega)$ if and only if the homogeneous equation $(I - A)v = 0$ has only the trivial solution $v = 0$.

Numerical Analysis & Scientific Computing II

Lesson 5

Integral Equations

5.1 Some solutions of boundary value problems for PDEs via integral equations

5.2 An Introduction

5.3 Numerical Methods



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Integral Equations: An Introduction

There are three main ideas for numerical solution of the second kind Fredholm integral equation

$$(I - A)u = f$$

with the linear integral operator

$$(Au)(x) = \int_{\Omega} K(x, y)u(y)dy .$$

Approximate the integral operator by

- *approximating the kernel $K(x, y)$.*
- *approximating the solution $u(x)$.*
- *approximating the integral $\int_{\Omega} f(y)dy$ by a quadrature.*

Numerical Analysis & Scientific Computing II

Lesson 5

Integral Equations

5.2 An Introduction

5.3 Numerical Methods

- Degenerate Kernel Method



Integral Equations: Numerical Methods



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The solution of the integral equation of the second kind, $u - Au = f$, is then obtained as

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where the coefficients $\gamma_1, \gamma_2, \dots, \gamma_n$ satisfy the linear system

$$\gamma_j - \sum_{k=1}^n \langle a_k, b_j \rangle \gamma_k = \langle f, b_j \rangle, \quad j = 1, 2, \dots, n.$$

Integral Equations: Numerical Methods



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Theorem

Let X and Y be Banach spaces and let $A: X \rightarrow Y$ be a bounded linear operator with a bounded operator $A^{-1}: Y \rightarrow X$. Assume the sequence $A_n: X \rightarrow Y$ of bounded linear operators to be norm convergent, that is, $\|A_n - A\| \rightarrow 0$ as $n \rightarrow \infty$. Then, for sufficiently large n , more precisely, for all n with $\|A^{-1}(A_n - A)\| < 1$, the inverse operators $A_n^{-1}: Y \rightarrow X$ exist and are bounded by

$$\|A_n^{-1}\| \leq \frac{\|A^{-1}\|}{1 - \|A^{-1}(A_n - A)\|}.$$

For all solutions of the equations $A\varphi = f$ and $A_n\varphi_n = f_n$, we have the error estimate

$$\|\varphi_n - \varphi\| \leq \frac{\|A^{-1}\|}{1 - \|A^{-1}(A_n - A)\|} \{\|(A_n - A)\varphi\| + \|f_n - f\|\}.$$

Numerical Analysis & Scientific Computing II

Lesson 5

Integral Equations

5.2 An Introduction

5.3 Numerical Methods

- Degenerate Kernel Method**
- via interpolation**



Integral Equations: Numerical Methods

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Consider the domain to be an interval, that is, $\Omega = (a, b)$ and let K be continuous.

One idea that works when K is continuous is interpolation -- approximate K by interpolating in x with respect to the points x_1, x_2, \dots, x_n in $[a, b]$ for each $y \in [a, b]$, we have

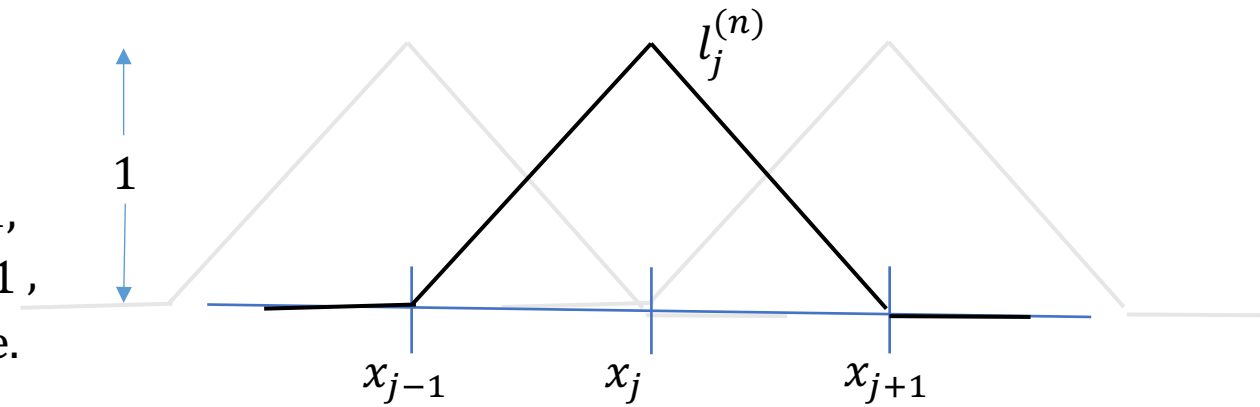
$$K_n(x, y) = \sum_{j=1}^n K(x_j, y) l_j^{(n)}(x).$$

Integral Equations: Numerical Methods

Example

Equidistant piecewise linear interpolation:

$$l_j^{(n)}(x) = \begin{cases} (x - x_{j-1})/h, & x \in [x_{j-1}, x_j], j \geq 1, \\ (x_{j+1} - x)/h, & x \in [x_j, x_{j+1}], j \leq n-1, \\ 0, & \text{otherwise.} \end{cases}$$



Integral Equations: Numerical Methods

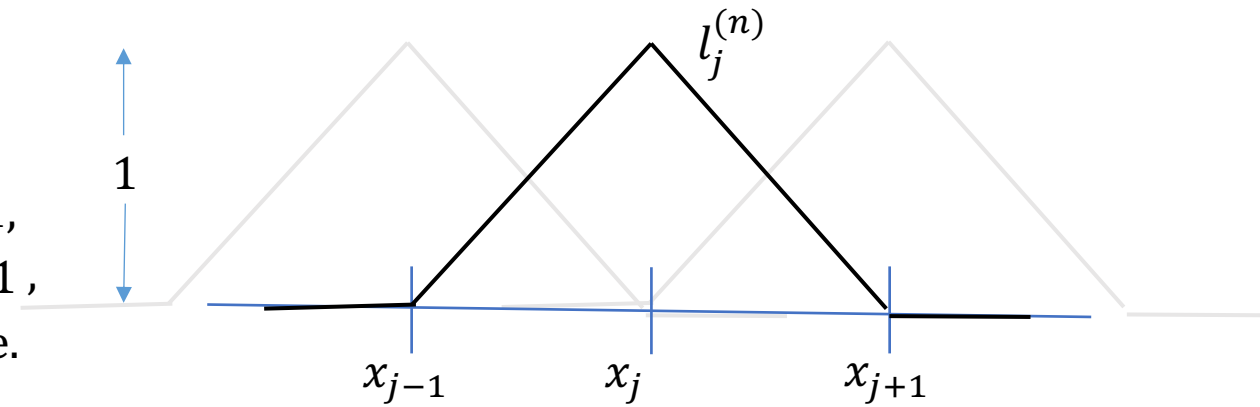
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Integral Equations: Numerical Methods

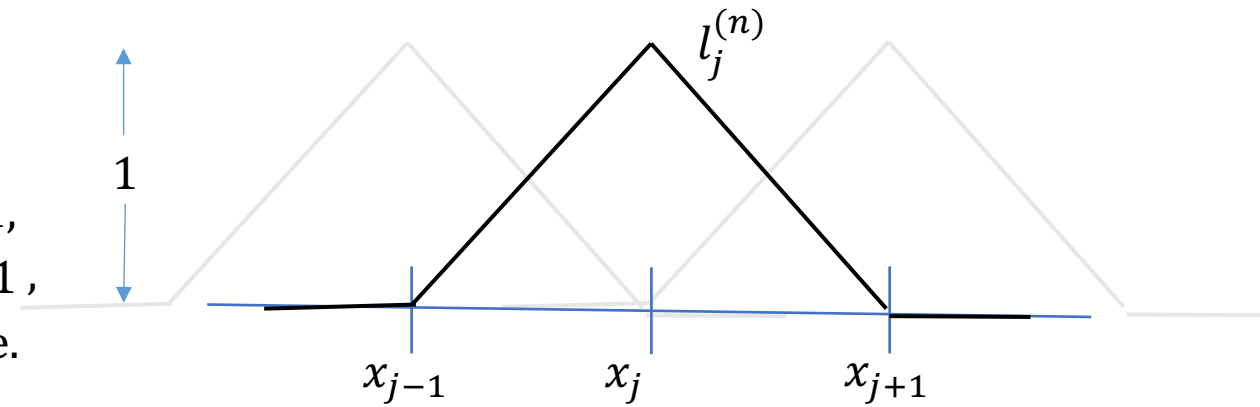
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Integral Equations: Numerical Methods

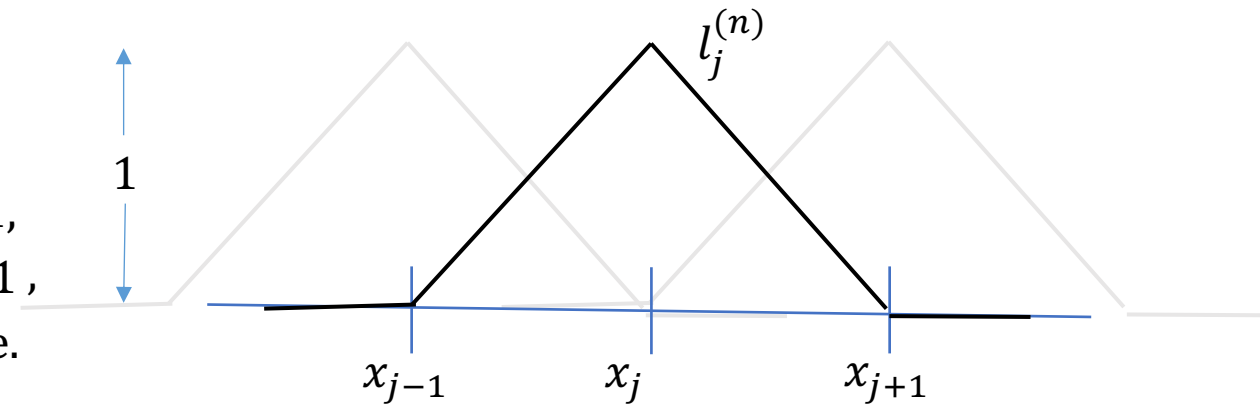
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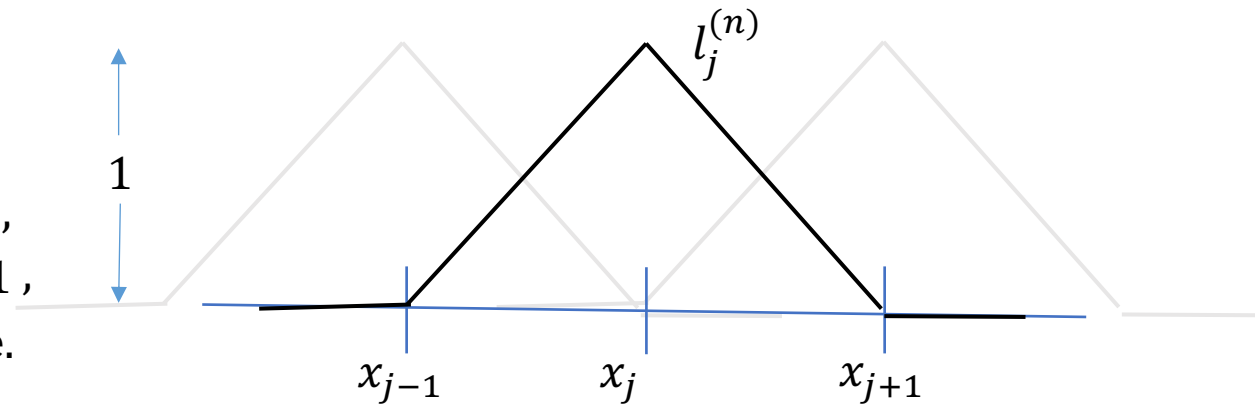


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where

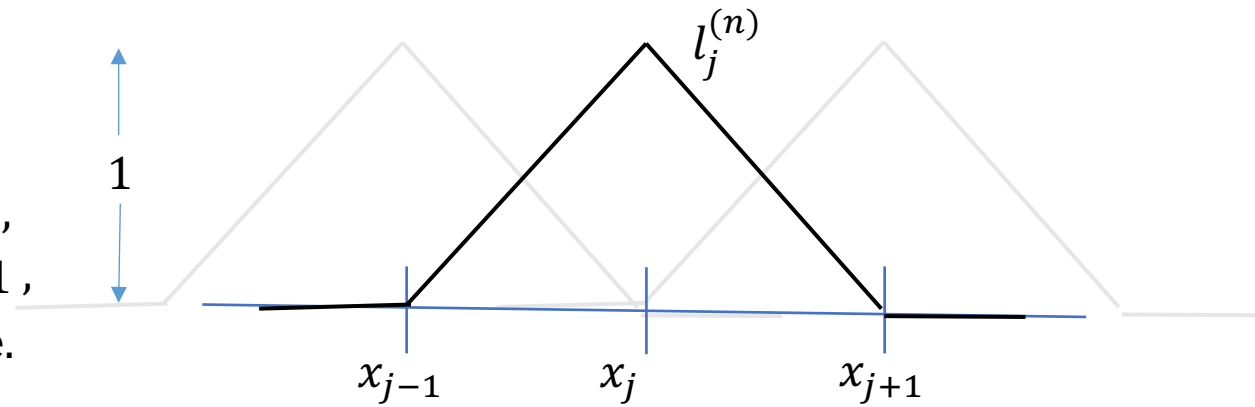
$$W = \frac{h}{6} \begin{bmatrix} 2 & 1 & & & \\ 1 & 4 & 1 & & \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ & & 1 & 4 & 1 \\ & & & 1 & 2 \end{bmatrix}.$$

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$$m_{ij} = \langle a_j, b_i \rangle = \int_a^b a_j(x) b_i(x) dx = \int_a^b l_j^{(n)}(x) K(x_i, x) dx \approx \sum_{k=1}^n K(x_i, x_k) \int_a^b l_j^{(n)}(x) l_k^{(n)}(x) dx = \sum_{k=1}^n K(x_i, x_k) w_{jk},$$

where

$$W = \frac{h}{6} \begin{bmatrix} 2 & 1 & & & \\ 1 & 4 & 1 & & \\ \cdot & \cdot & \cdot & \cdot & \cdot \\ & & 1 & 4 & 1 \\ & & & 1 & 2 \end{bmatrix}.$$

Similarly,

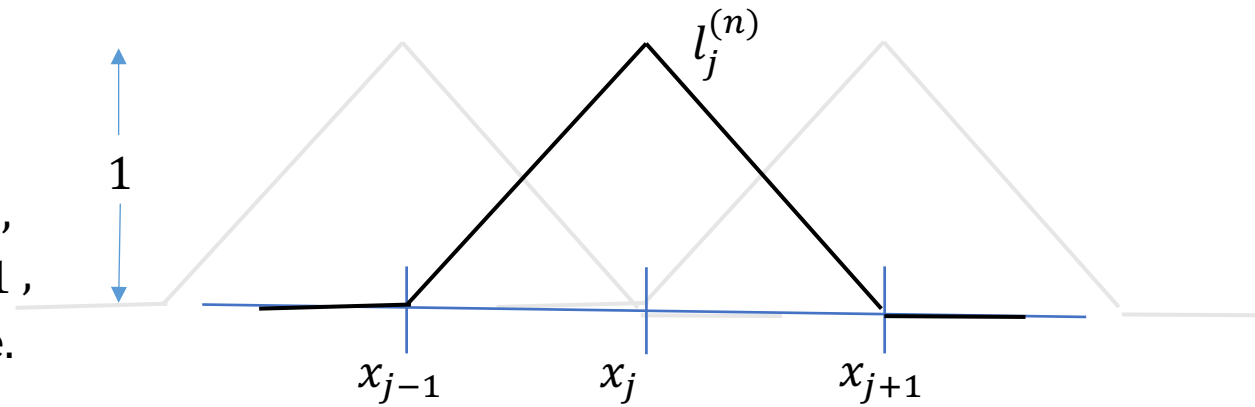
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Integral Equations: Numerical Methods

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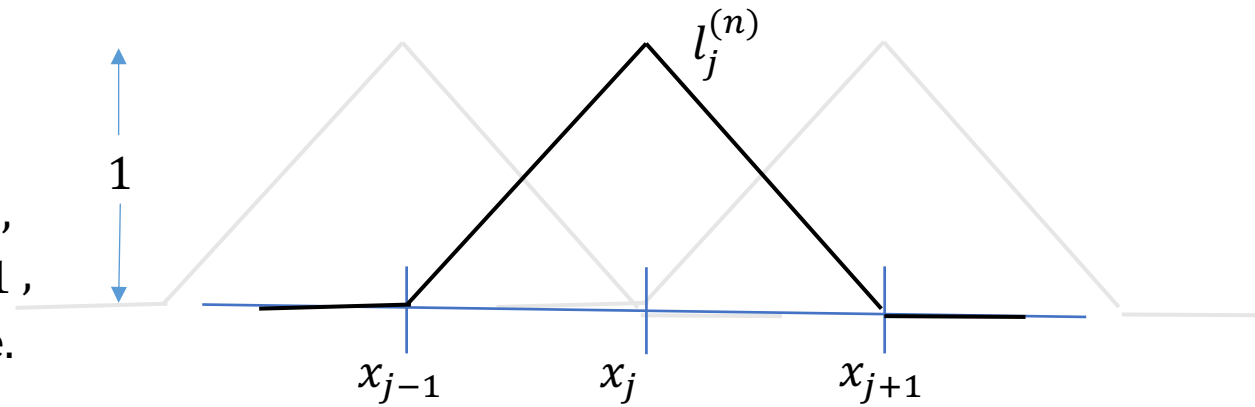
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Integral Equations: Numerical Methods

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Integral Equations: Numerical Methods



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Integral Equations: Numerical Methods



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Consider the following integral equation

$$\varphi(x) - \frac{1}{2} \int_0^1 (x+1)e^{-xy} \varphi(y) dy = e^{-x} - \frac{1}{2} + \frac{1}{2} e^{-(x+1)}, \quad 0 \leq x \leq 1.$$

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Table showing the error between the approximate and the exact solution



n	x = 0	x = 0.25	x = 0.5	x = 0.75	x = 1
4	0.004808	0.005430	0.006178	0.007128	0.008331
8	0.001199	0.001354	0.001541	0.001778	0.002078
16	0.000300	0.000385	0.000385	0.000444	0.000519
32	0.000075	0.000085	0.000096	0.000111	0.000130

Numerical Analysis & Scientific Computing II

Lesson 5

Integral Equations

5.2 An Introduction

5.3 Numerical Methods

- Degenerate Kernel Method**
- via orthogonal expansion**



Integral Equations: Numerical Methods



Another possibility for constructing degenerate kernels is by expansions, in particular, by orthogonal expansions.



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If $\langle \cdot, \cdot \rangle$ denotes an inner product on $C(\Omega)$ and $\{u_1, u_2, \dots\}$ is a complete orthonormal system, then a given continuous kernel K is expanded with respect to x for each y , that is, $K(x, y)$ is approximated by the partial sum

$$K_n(x, y) = \sum_{j=1}^n u_j(x) \langle K(\cdot, y), u_j \rangle$$

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For this degenerate kernel, the linear system reads

$$\gamma_j - \sum_{k=1}^n \gamma_k \int_{\Omega} u_k(y) \langle K(\cdot, y), u_k \rangle dy = \int_{\Omega} f(y) \langle K(\cdot, y), u_k \rangle dy, j = 1, 2, \dots, n.$$

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Note that the setting up the linear system requires a double integration for each coefficient and for each right-hand side.

Integral Equations: Numerical Methods



Example (Expansion with respect to the Chebyshev system)

Let $\Omega = (-1,1)$.

Integral Equations: Numerical Methods



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Easy to check that

$$T_0(z) = 1, \quad T_1(z) = z, \quad T_{n+1}(z) + T_{n-1}(z) = 2zT_n(z), \quad n \geq 1.$$

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Integral Equations: Numerical Methods

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The approximate solution u_n to $(I - A)u = f$ is obtained as

$$u_n(x) = f(x) + \sum_{j=1}^n \gamma_j T_j(x)$$

where the coefficients γ_j solve the linear system

$$\gamma_j - \sum_{k=1}^n \gamma_k \int_{-1}^1 T_k(y) \langle K(\cdot, y), T_k \rangle dy = \int_{-1}^1 f(y) \langle K(\cdot, y), T_k \rangle dy, \quad j = 1, 2, \dots, n.$$

Integral Equations: Numerical Methods

Theorem

Let $g: [-1, 1] \rightarrow \mathbb{R}$ be analytic. Then, there exists an ellipse E with foci at -1 and 1 such that g can be extended to a holomorphic and bounded function $g: D \rightarrow \mathbb{C}$ where D denotes the open interior of E . The orthonormal expansion with respect to the Chebyshev polynomials

$$g = \frac{a_0}{2} T_0 + \sum_{n=1}^{\infty} a_n T_n, \quad a_n = \frac{2}{\pi} \int_{-1}^1 \frac{g(x) T_n(x)}{\sqrt{1-x^2}} dx = \frac{2}{\pi} \langle g, T_n \rangle$$

is uniformly convergent with the estimate

$$\left\| g - \frac{a_0}{2} T_0 - \sum_{k=1}^n a_k T_k \right\|_{\infty} \leq \frac{2M}{R-1} R^{-n}.$$

Here R is given through the semi-axis a and b of E by $R = a + b$ and M is a bound on g in D .