

## *Lesson 4*

# *Numerical Solution of PDE*

### *4.3 Hyperbolic PDE*

#### *- Finite Difference Methods*



# Numerical Methods for PDE: Hyperbolic PDE



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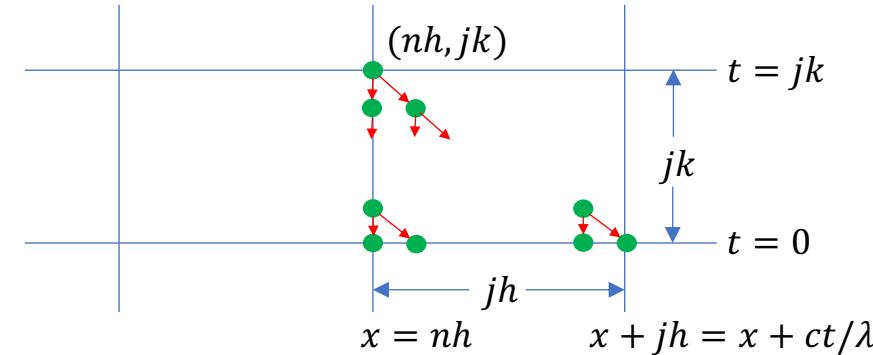
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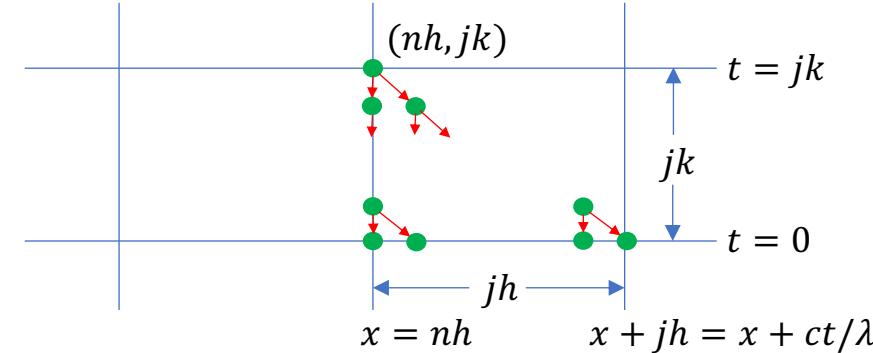
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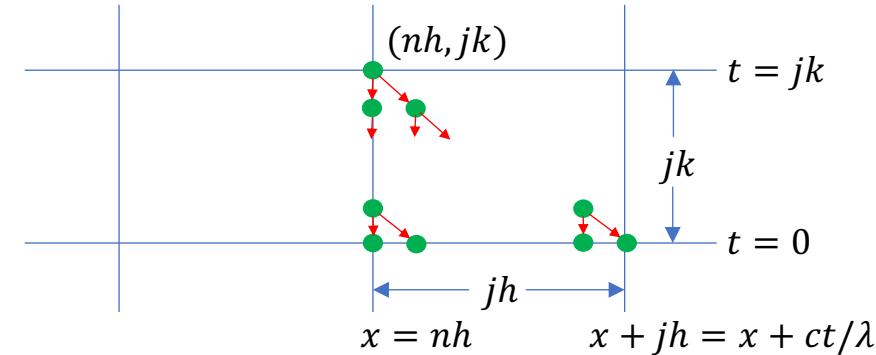
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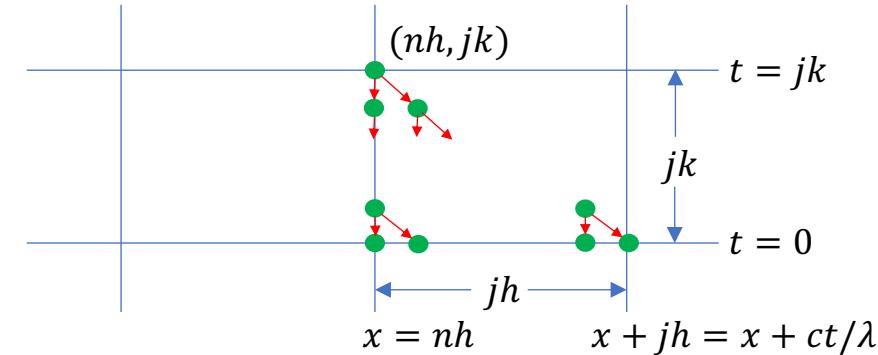
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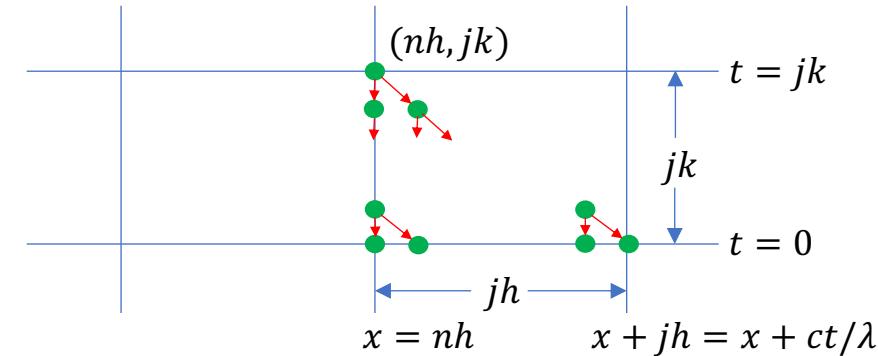
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This necessary condition, which fails for forward-forward difference method, is called the **Courant-Friedrichs-Levy condition**, or **CFL condition**.



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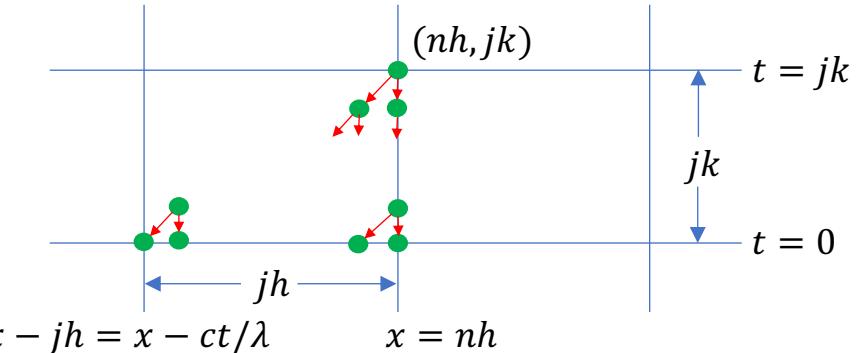
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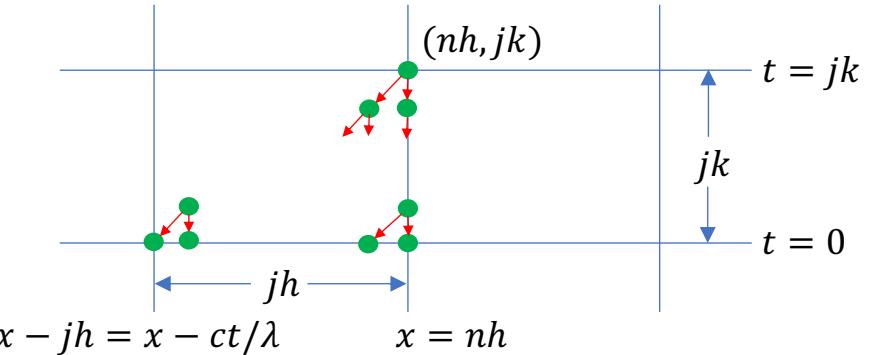
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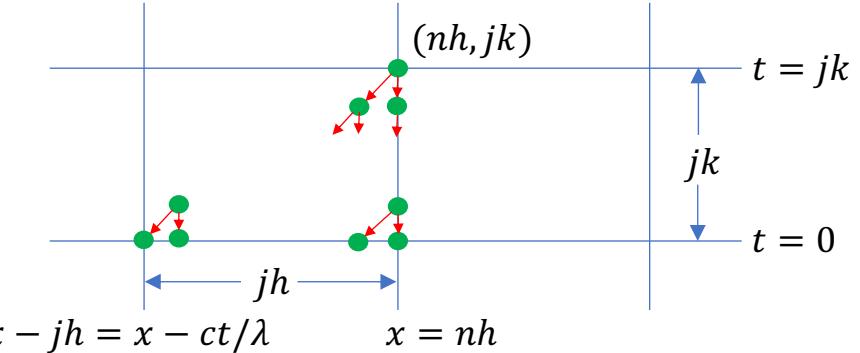
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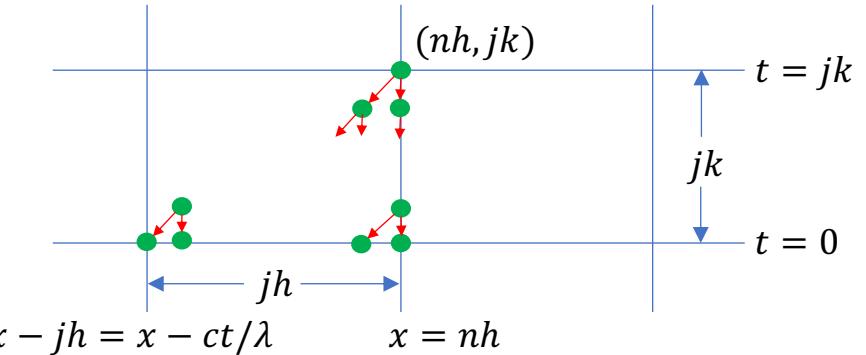
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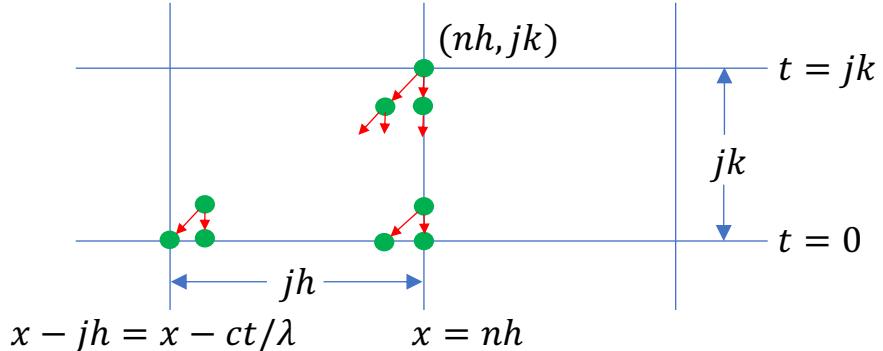
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In general, however, CFL is not sufficient for convergence. It turns out that, for forward-central scheme, while the CFL condition is  $|\lambda| \leq 1$ , the method is **unconditionally unstable**.

## *Lesson 4*

# *Numerical Solution of PDE*

### ***4.3 Hyperbolic PDE***

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



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

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

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$$\psi_m = \exp(2\pi i mx), \quad x \in h\mathbb{Z}, \quad m = 0, 1, \dots, N - 1.$$

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Let  $L_h^{per}$  denote the space of complex valued 1-periodic functions on  $h\mathbb{Z}$ , that is,

$$L_h^{per} = \{v : h\mathbb{Z} \rightarrow \mathbb{C} \mid v((n + N)h) = v(nh + 1) = v(nh)\}.$$

As a basis for this space, we choose

$$\psi_m = \exp(2\pi imx), \quad x \in h\mathbb{Z}, \quad m = 0, 1, \dots, N - 1.$$

The  $\psi_m$  are orthogonal with respect to the inner product (*exercise*)

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

# Numerical Methods for PDE: Hyperbolic PDE



Note that  $\psi_m$  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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## *Lesson 5*

# *Integral Equations*



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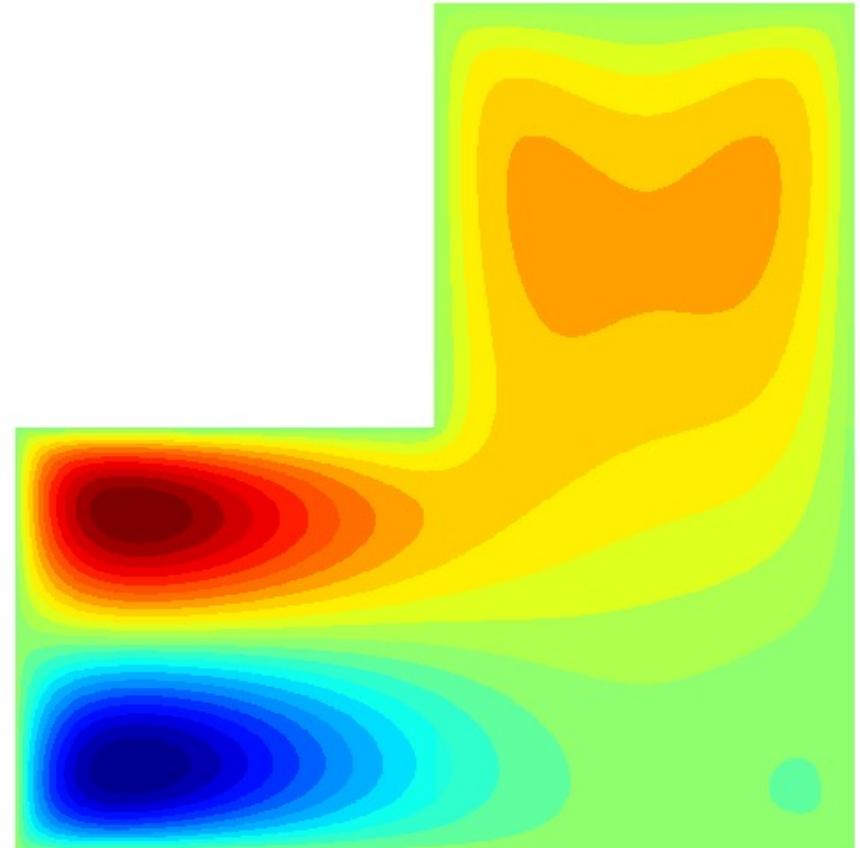
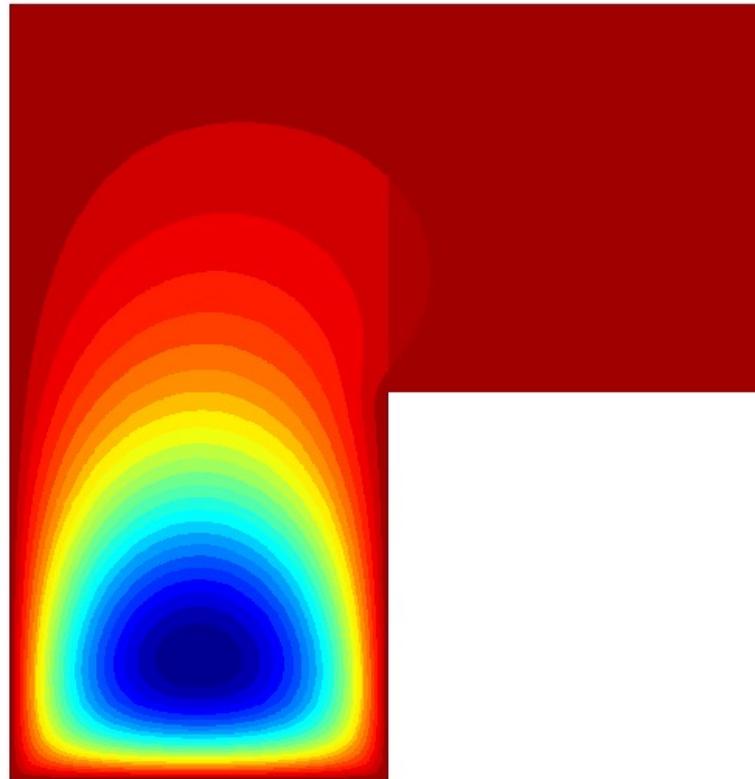
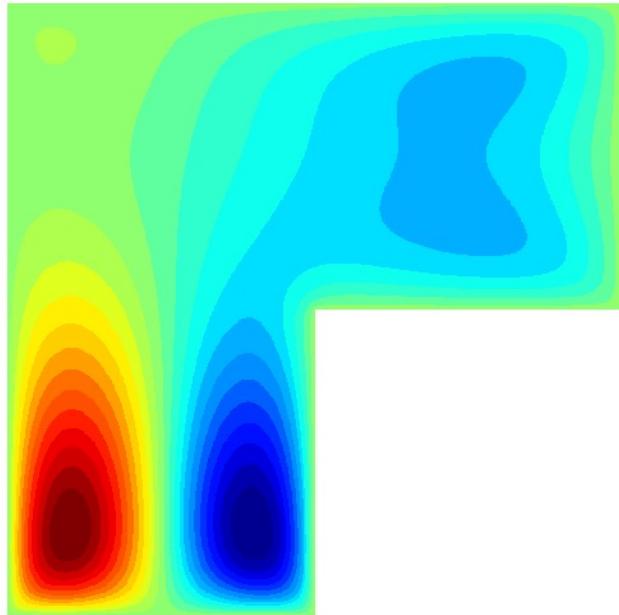
*5.1 Some solutions of boundary value problems for PDEs via integral equations*



# Integral Equations: Some examples



*Some solutions of integral Fredholm integral equations*

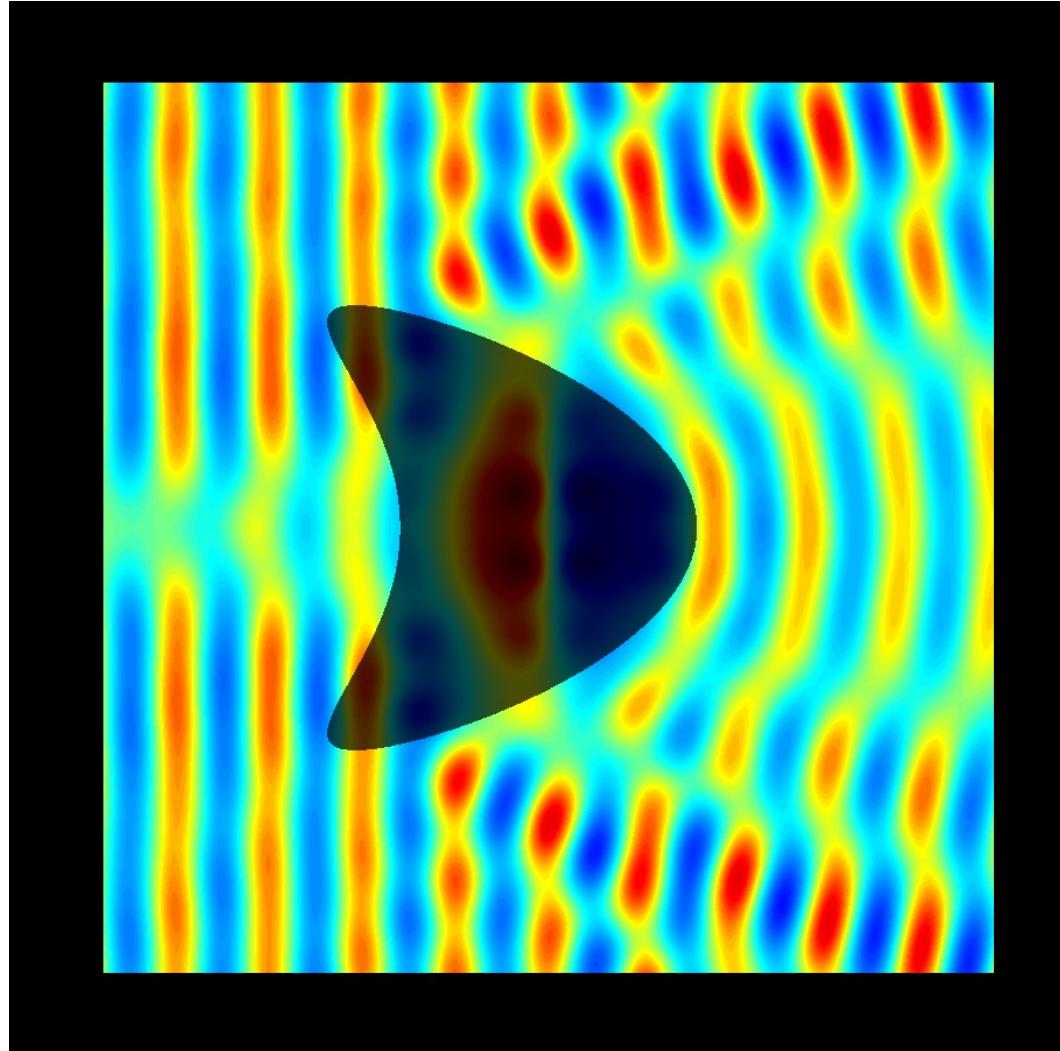


(Akash Anand)

# Integral Equations: Some examples



*Some solutions of integral Fredholm integral equations in wave scattering*

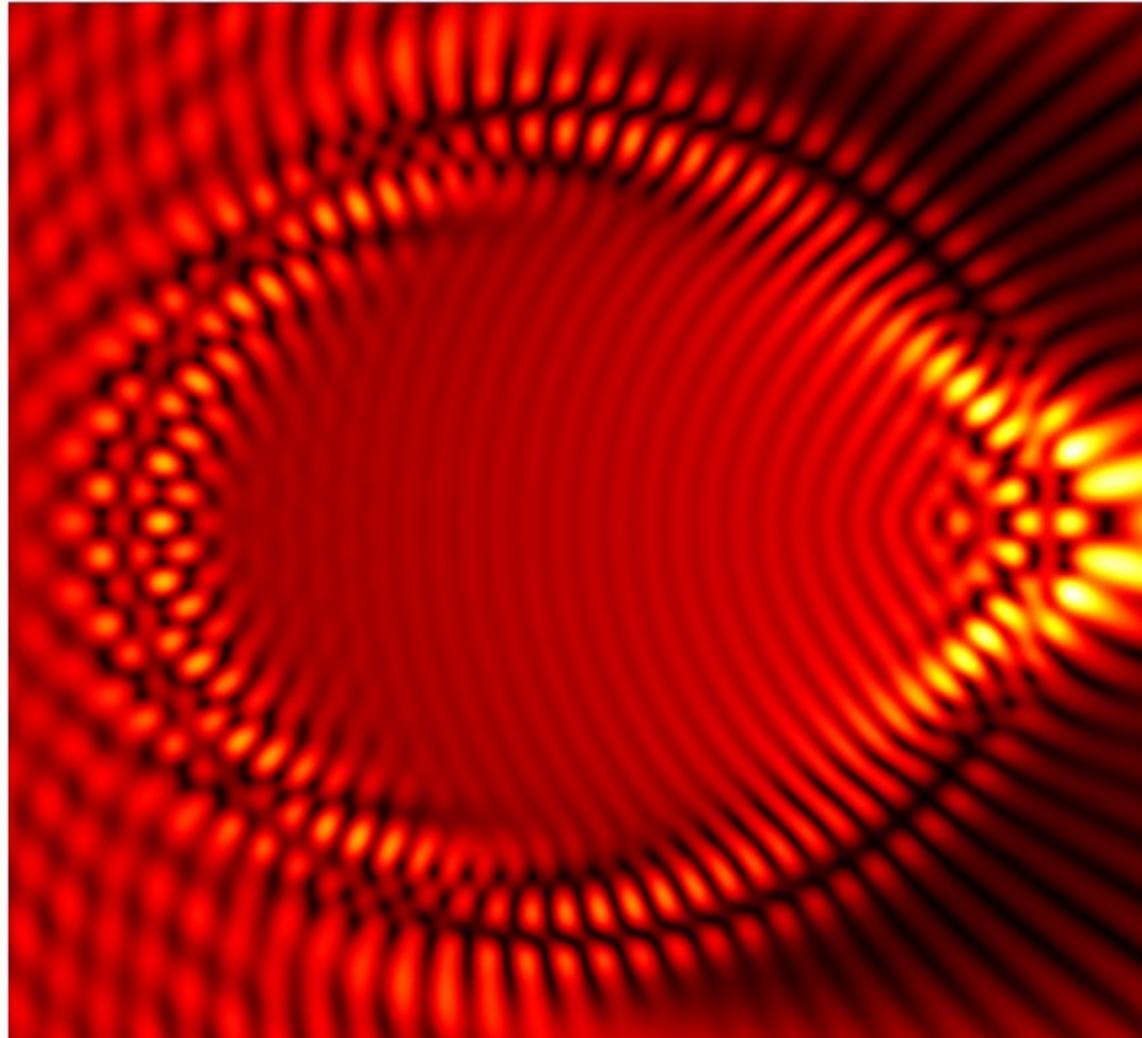


(Ambuj Pandey,  
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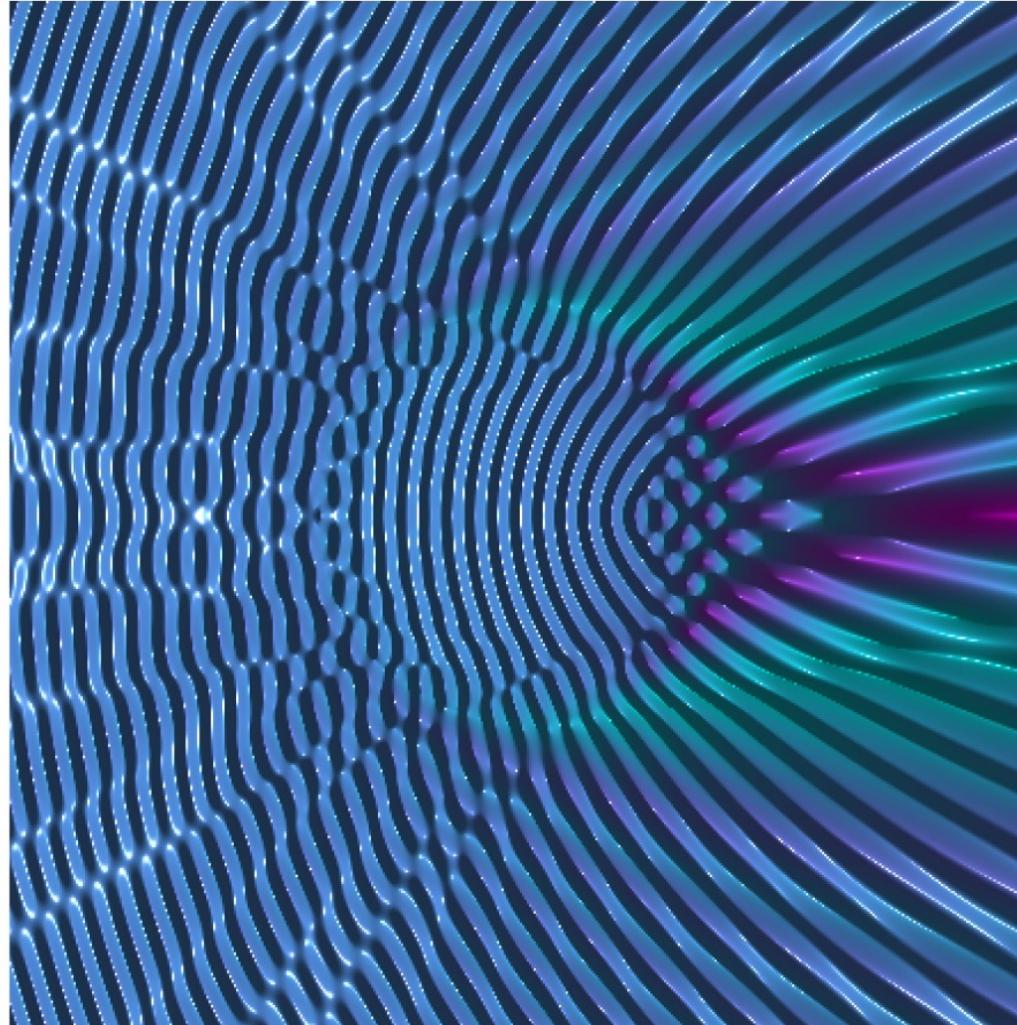


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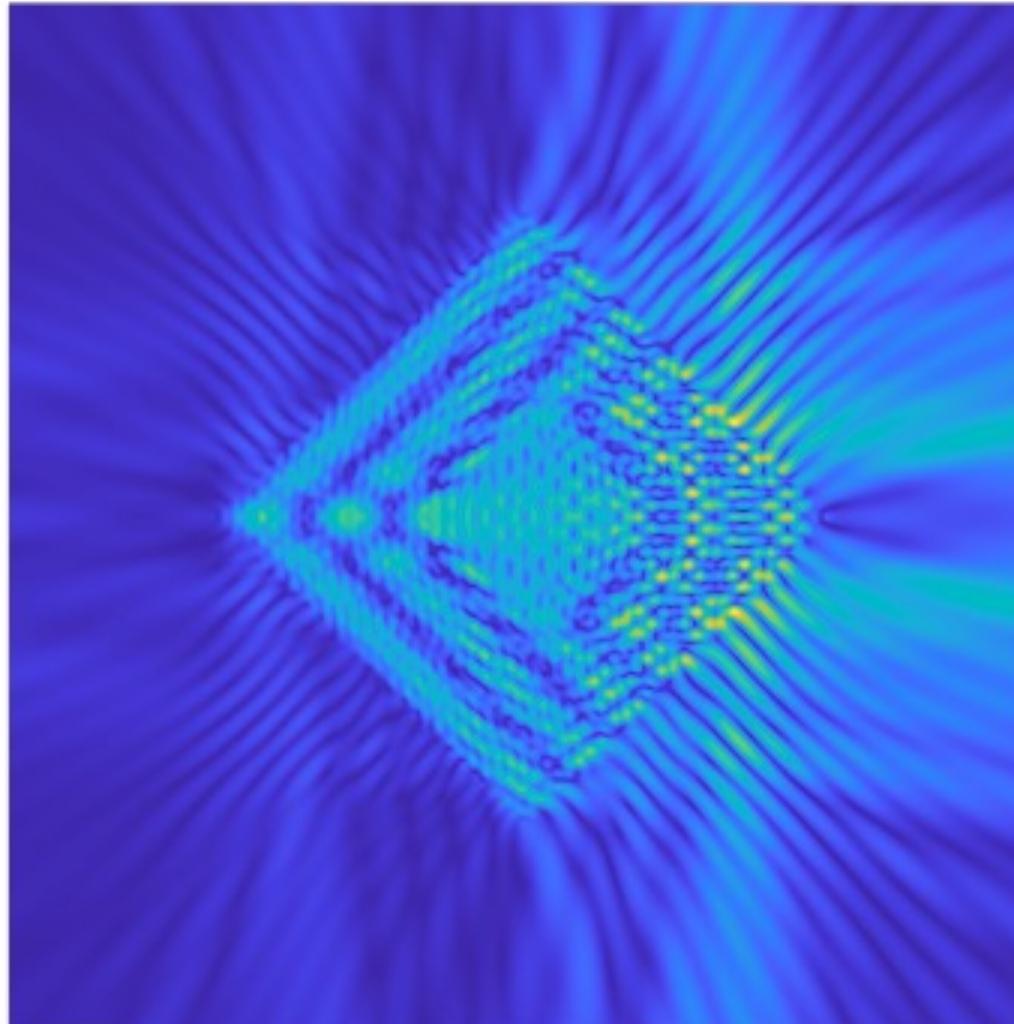


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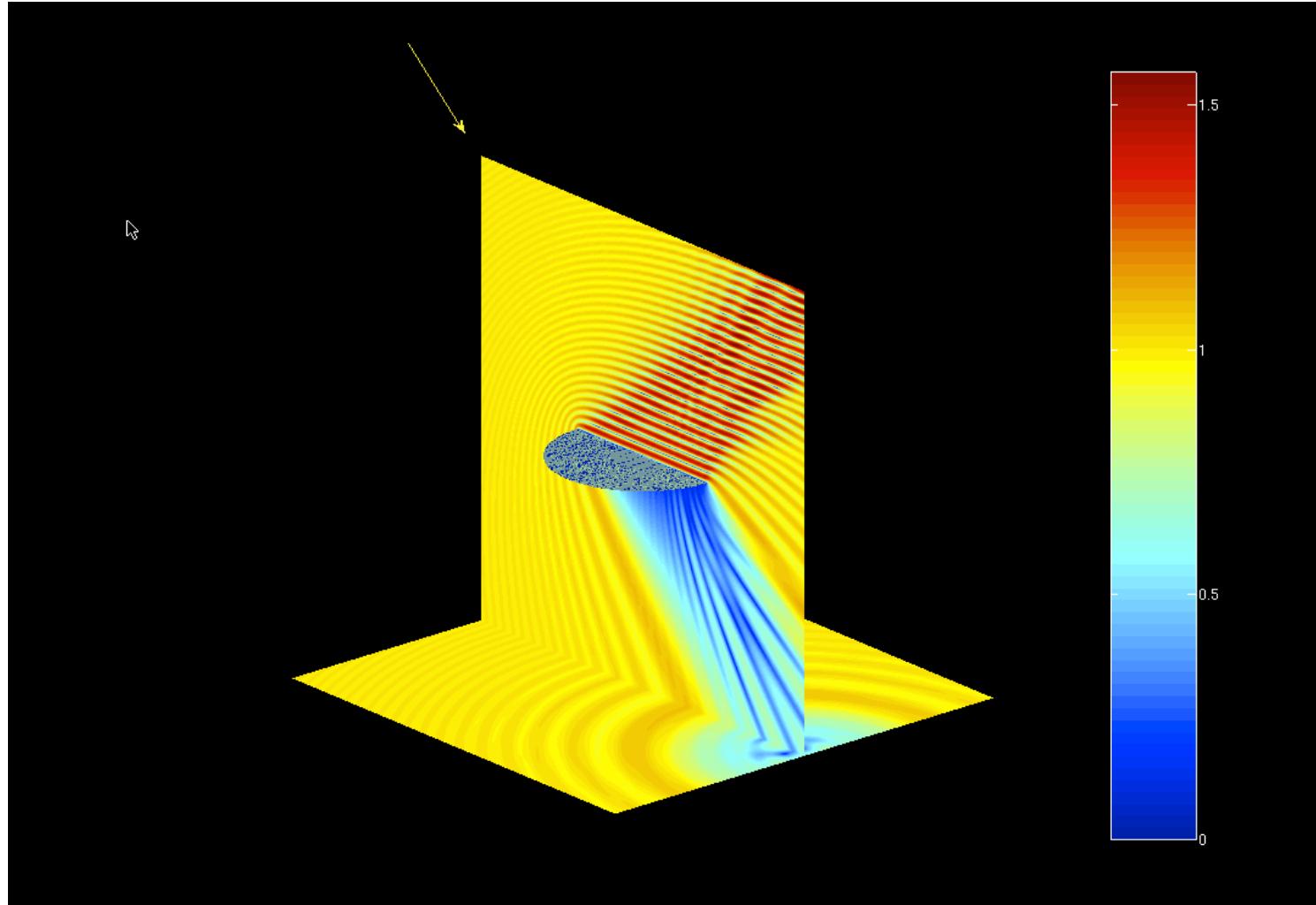


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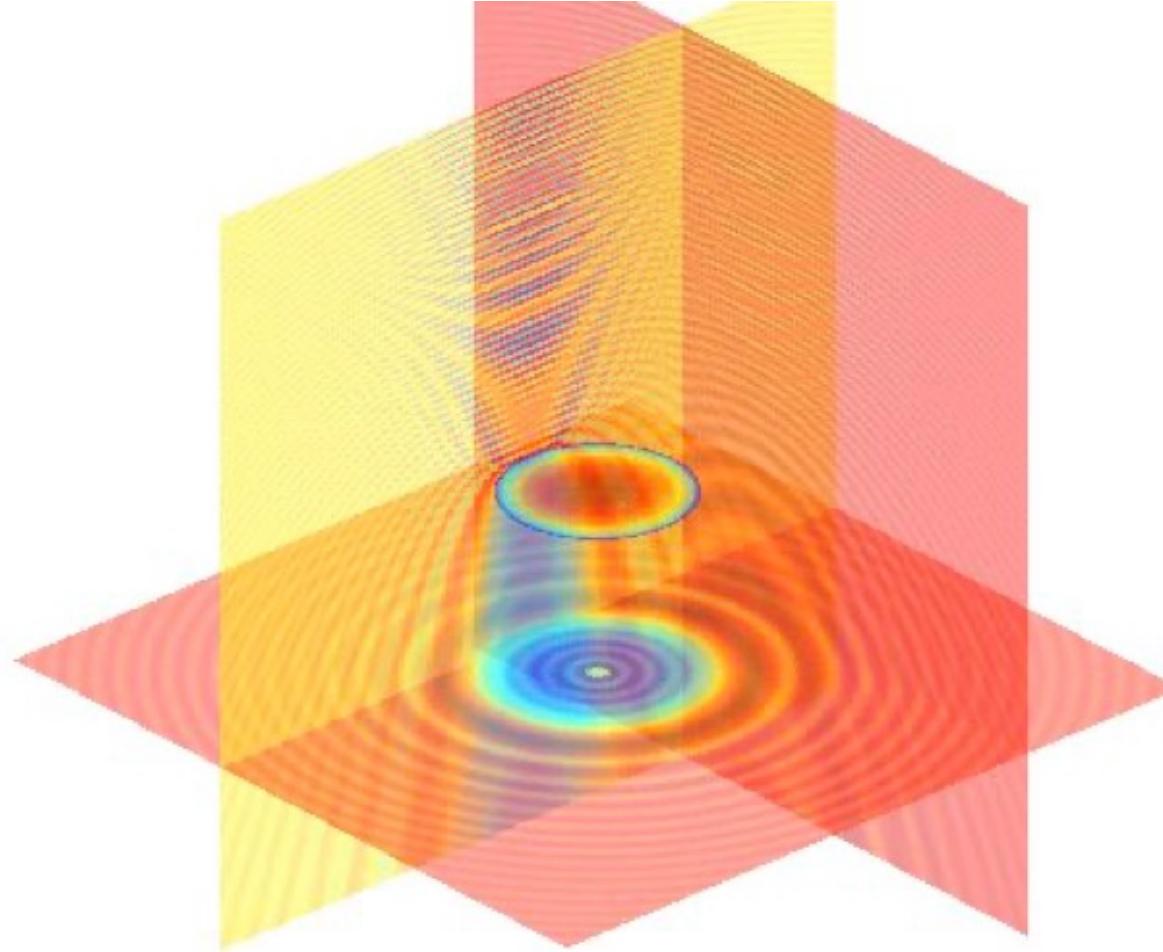


(Stephen Lintner,  
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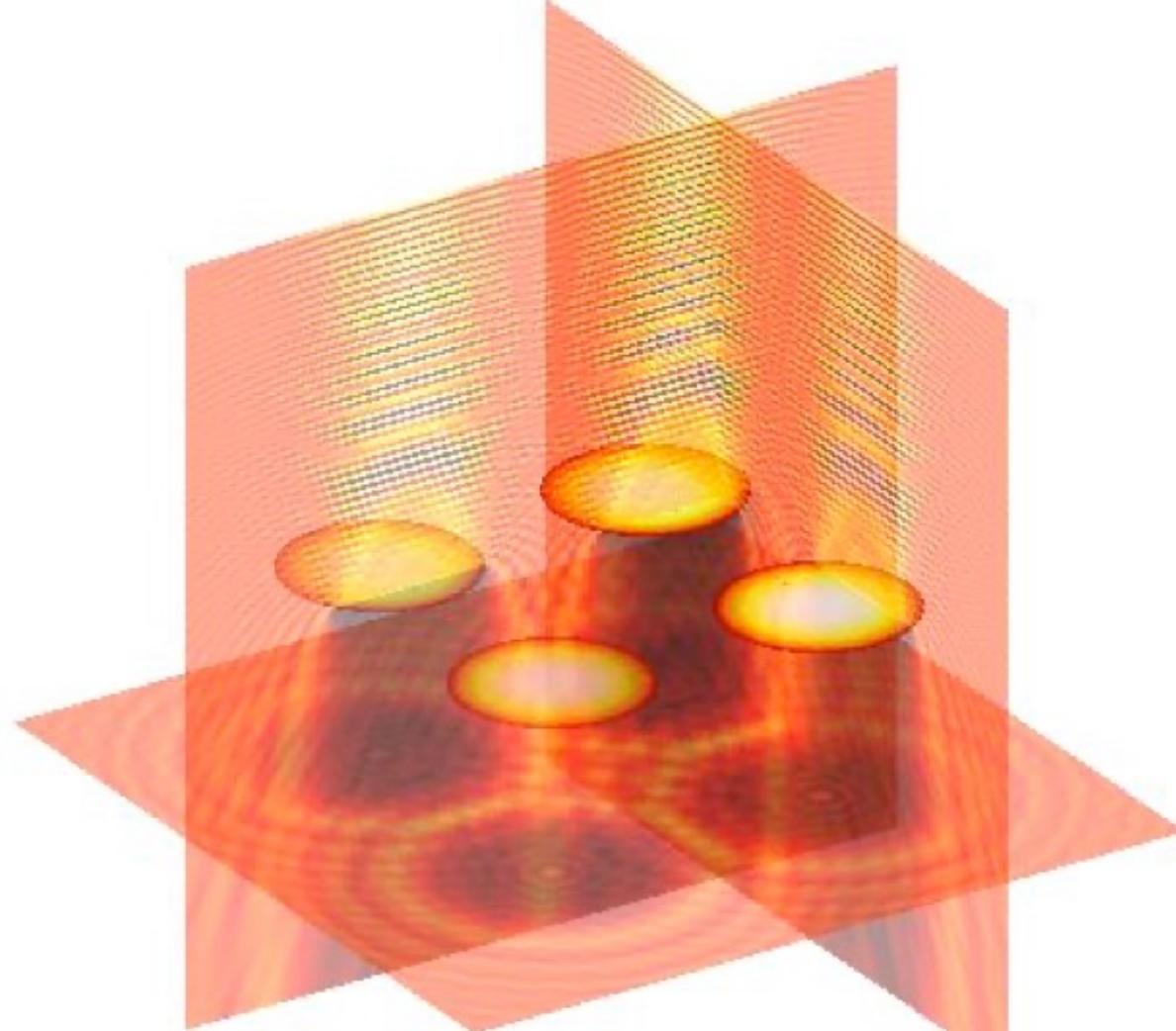


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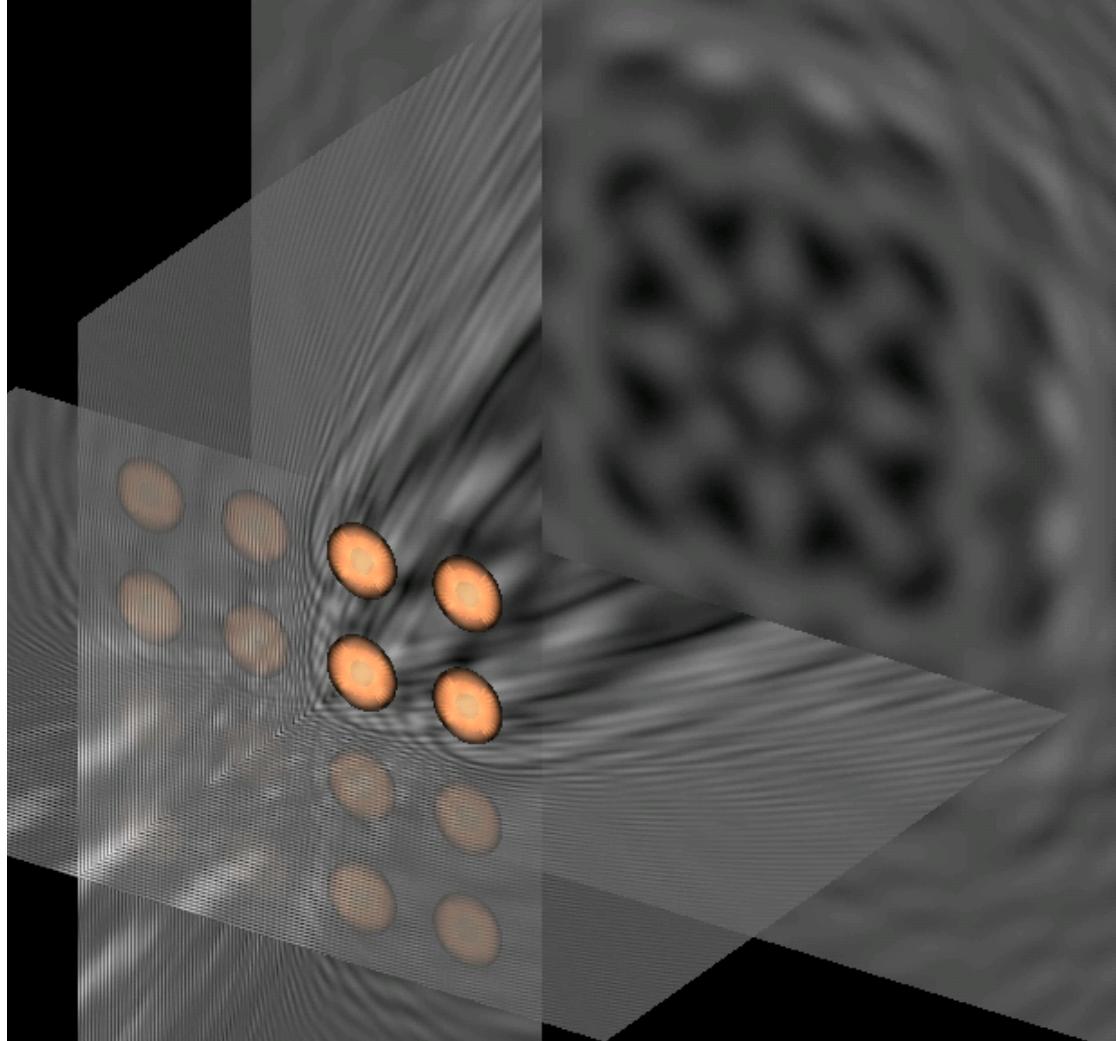


(Akash Anand)

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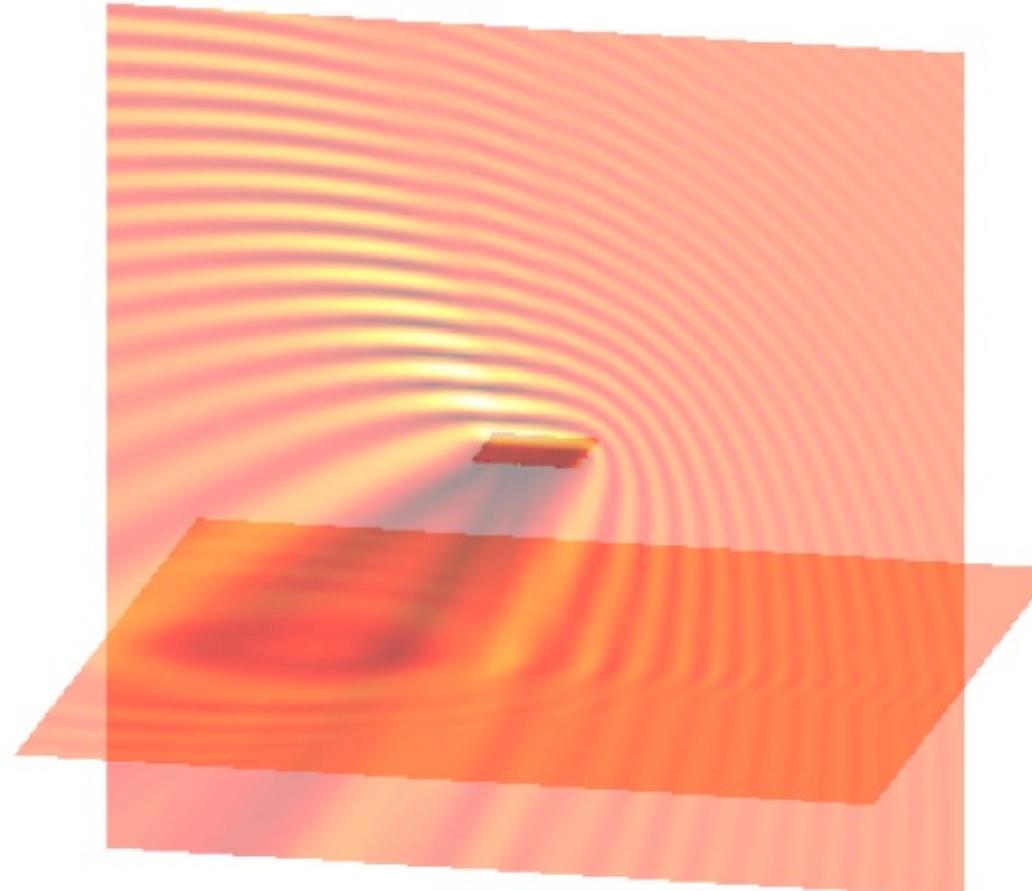


(Stephen Lintner,  
Oscar Bruno)

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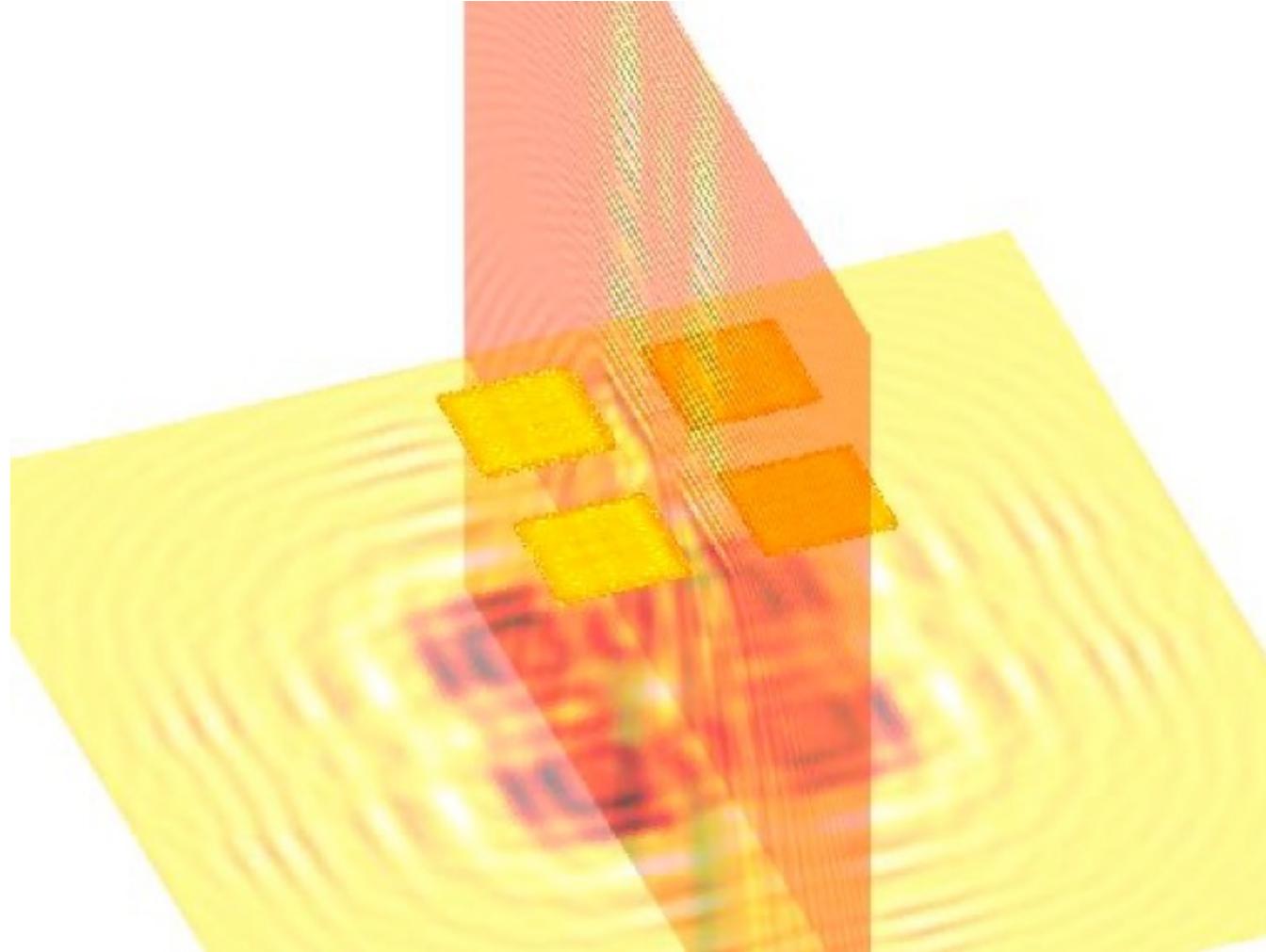


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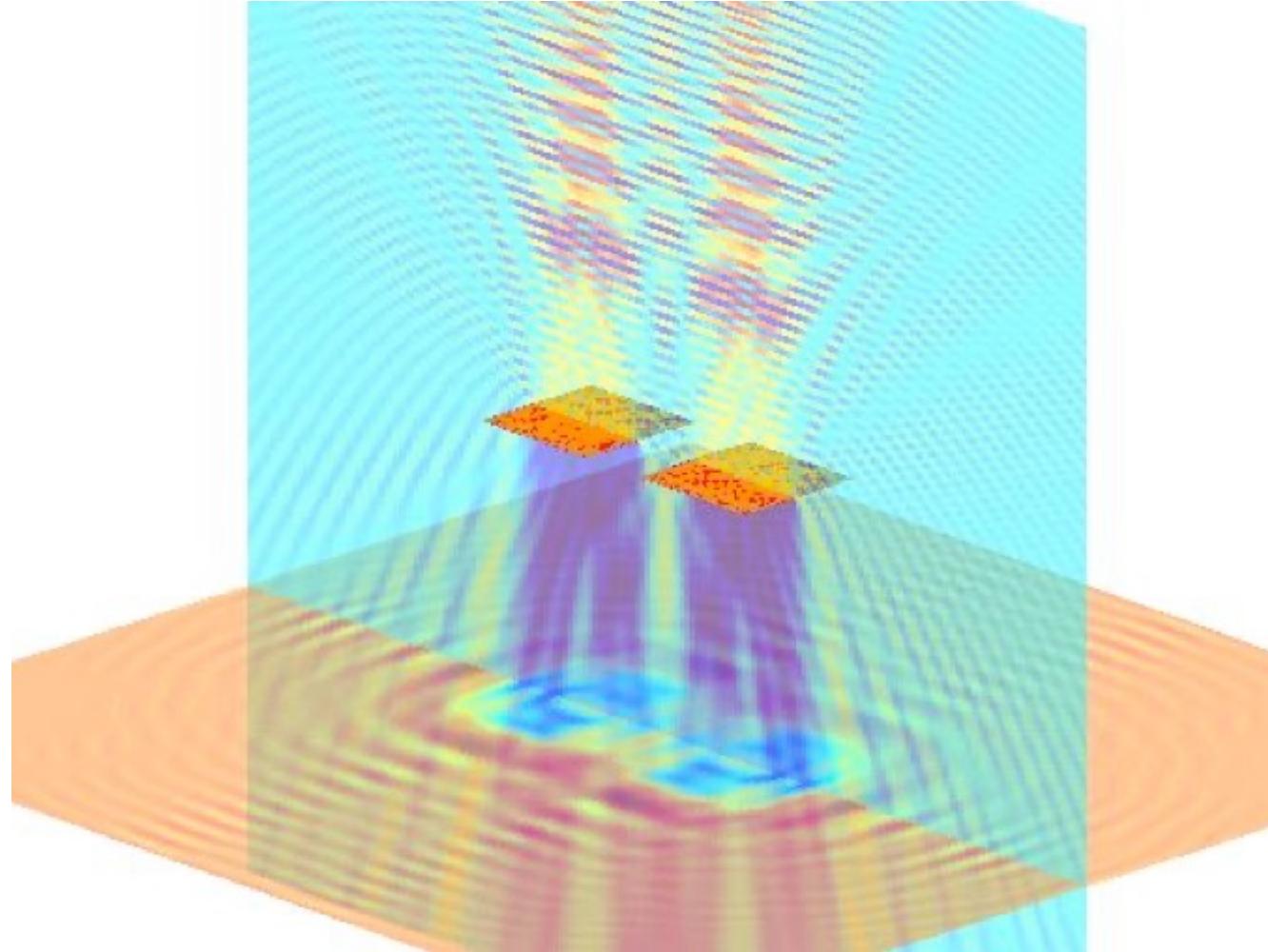


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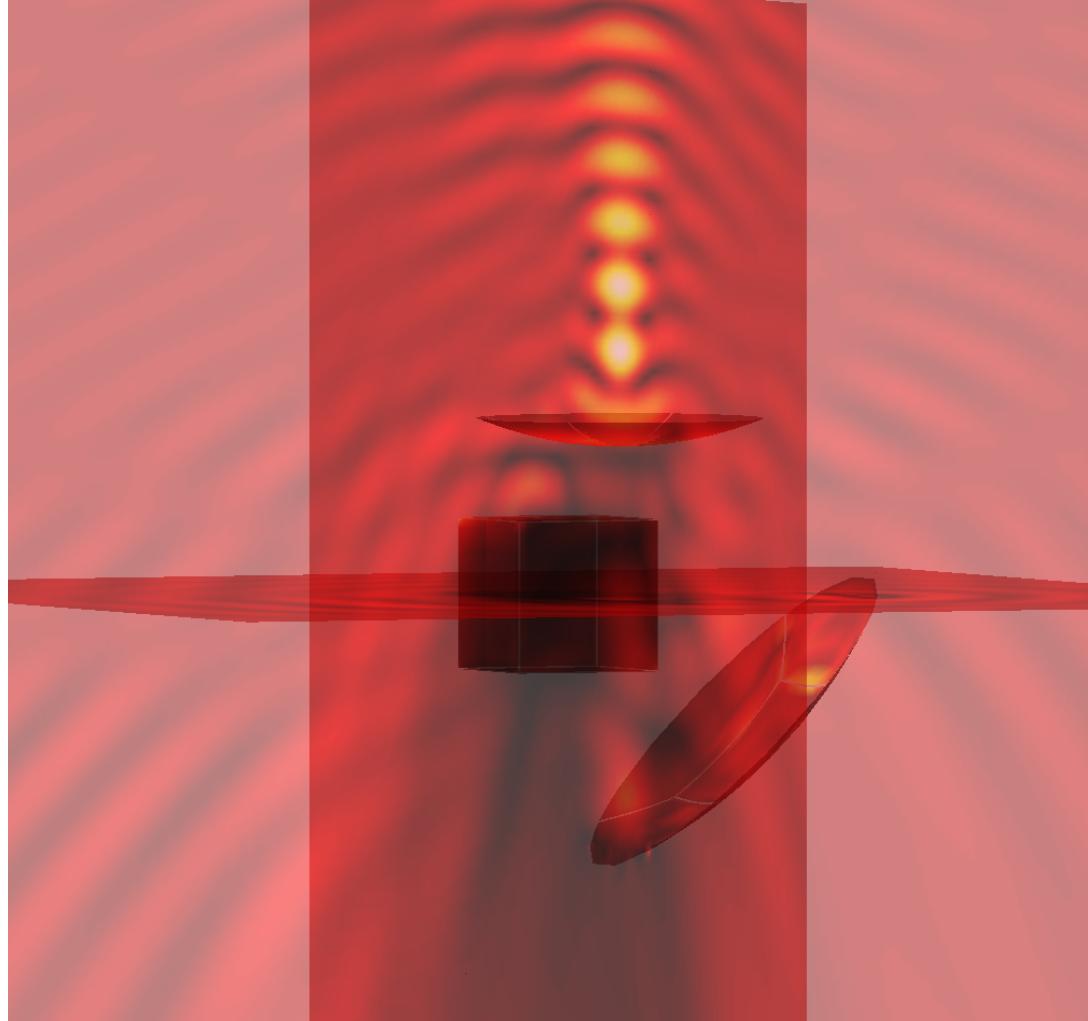


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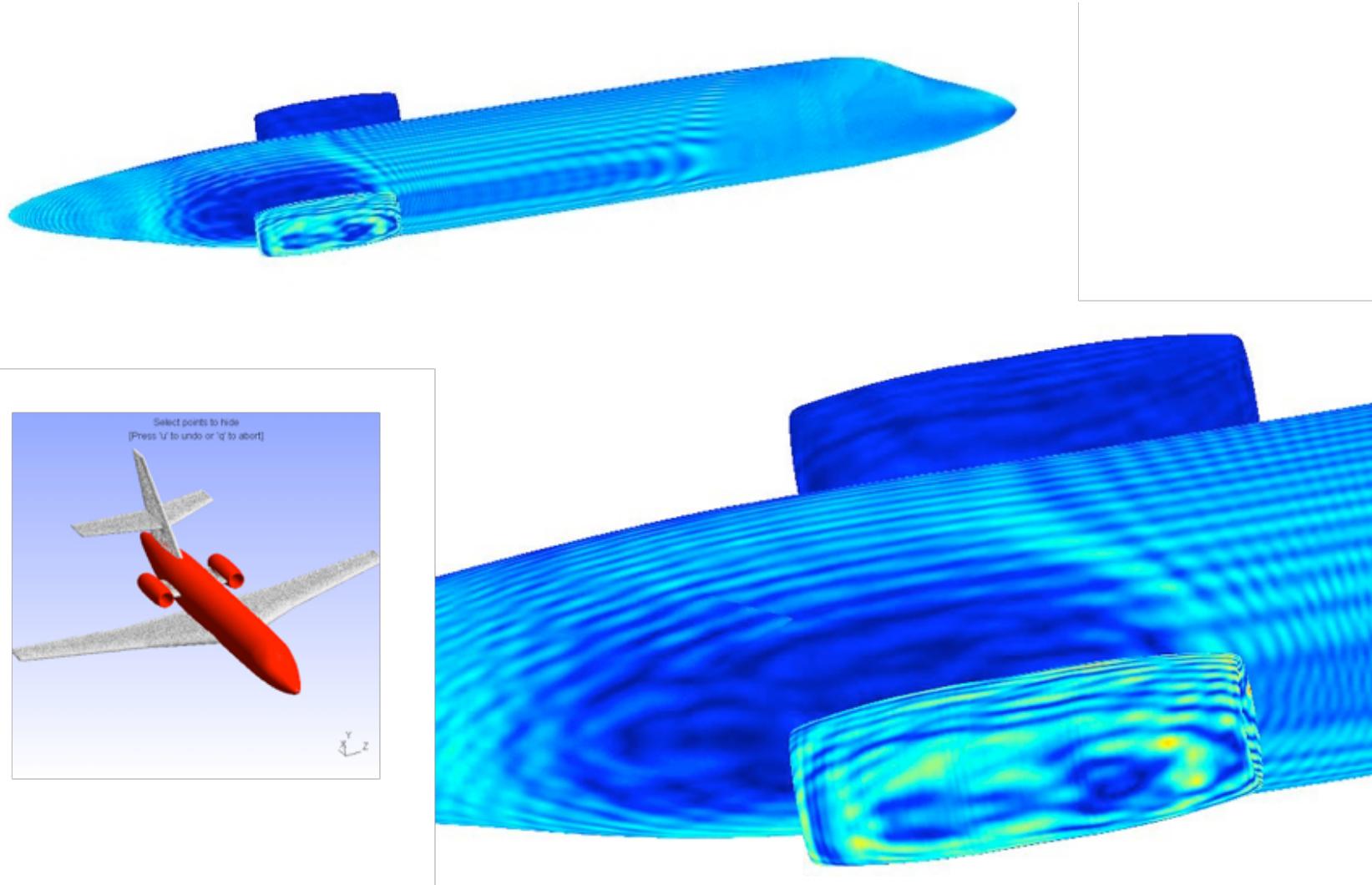


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# Integral Equations: Some examples



*Some solutions of integral Fredholm integral equations in wave scattering*



(Akash Anand)

## *Lesson 5*

# *Integral Equations*

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

***5.2 An Introduction***



# Integral Equations: An Introduction



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# Integral Equations: An Introduction



We have already encountered integral equations in this course.

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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This is a particular case of the more general **Volterra integral equations**

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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

$$u(t) - \int_{\Omega} K(t, s)u(s)ds = f(t), \quad t \in \Omega.$$

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

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

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

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

with weakly singular kernel  $K$  maps continuous functions to continuous function, that is, the operator  $A: C(\Omega) \rightarrow C(\Omega)$  is well-defined.

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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  $v = 0$ .

## *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**



# 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.*

## *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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Thus, the approximate solution has the form

$$u_n = f + \sum_{k=1}^n \gamma_k a_k,$$

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



*How do we construct such an approximation for the kernel? How do we know that this approximation is accurate?*

# Integral Equations: Numerical Methods



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The idea is to construct  $K_n$  such that  $\|A_n - A\| \rightarrow 0$  as  $n \rightarrow \infty$  where

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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

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

## *Lesson 5*

# *Integral Equations*

*5.2 An Introduction*

**5.3 Numerical Methods**

**- Degenerate Kernel Method**

**- via interpolation**



**Akash Anand**  
**MATH, IIT KANPUR**

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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

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

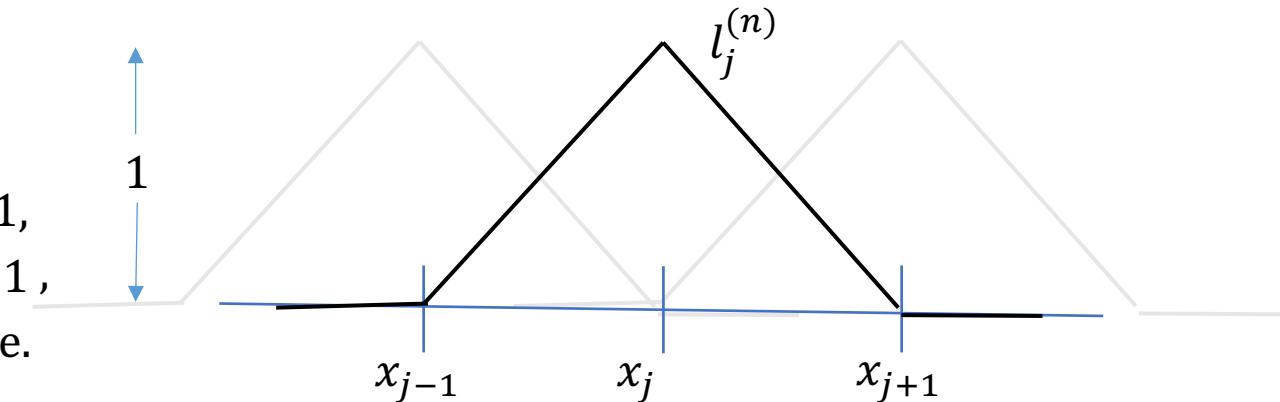
# Integral Equations: Numerical Methods



## 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



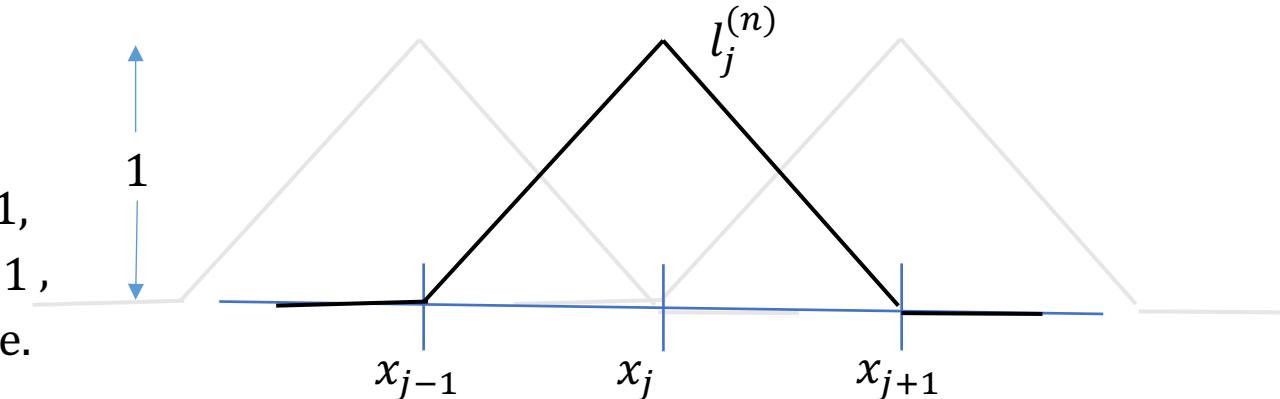
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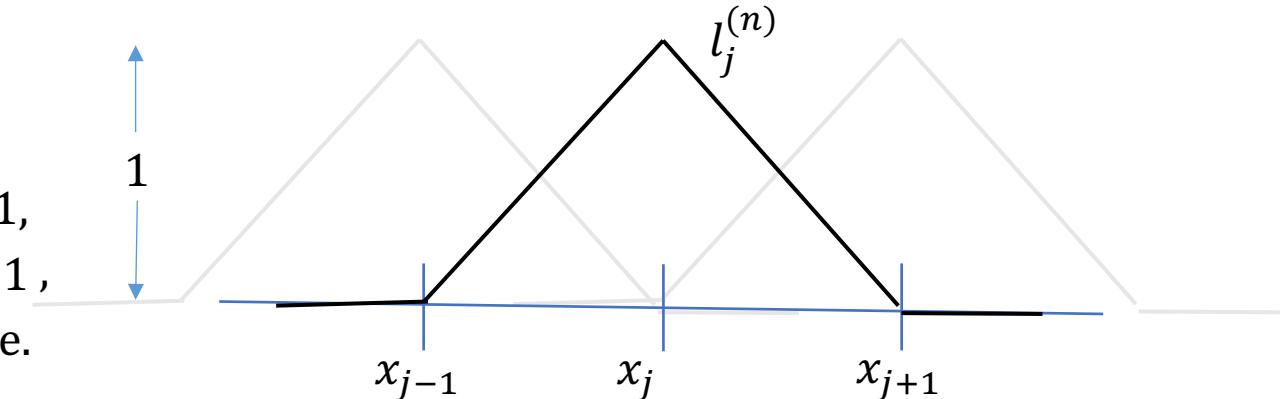
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# Integral Equations: Numerical Methods



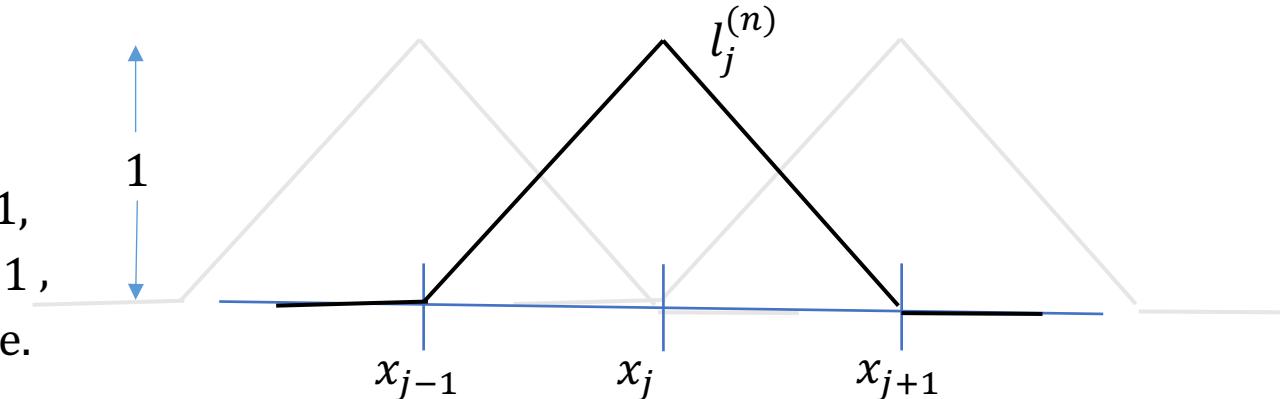
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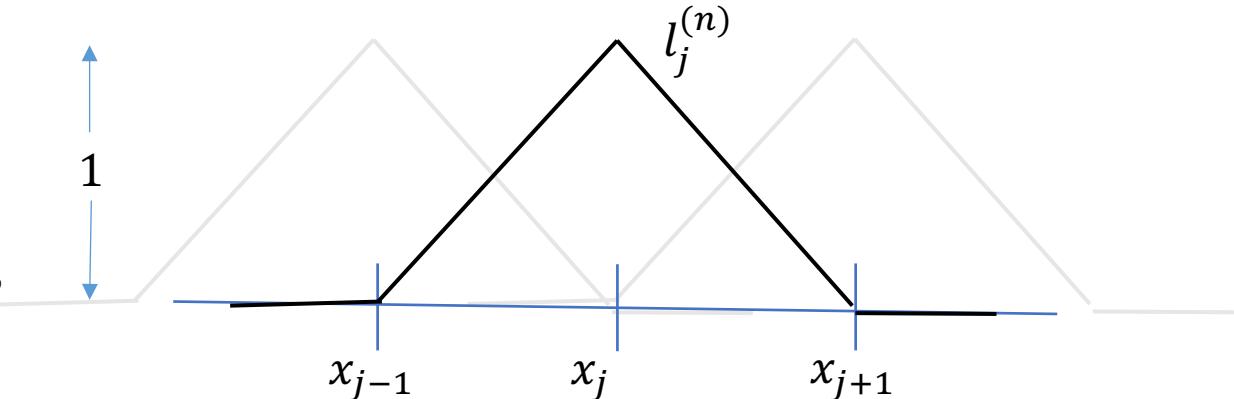
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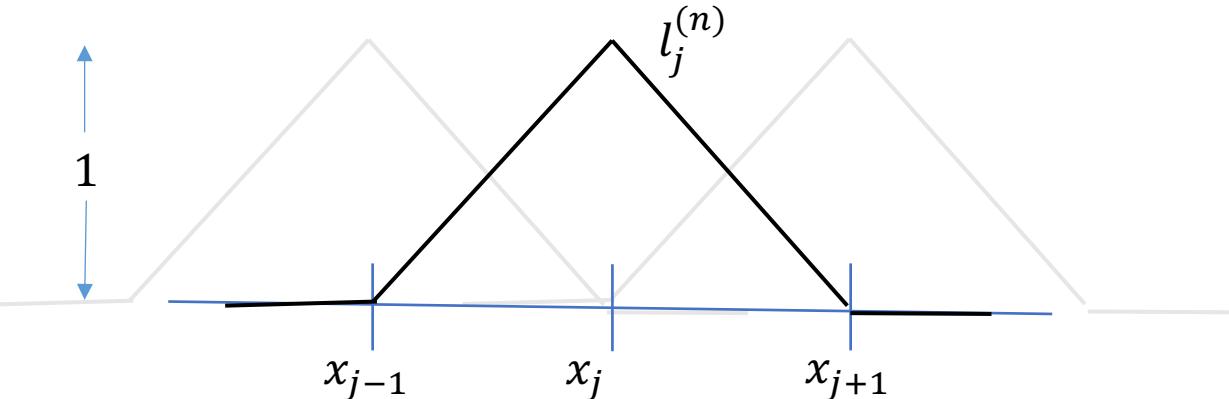
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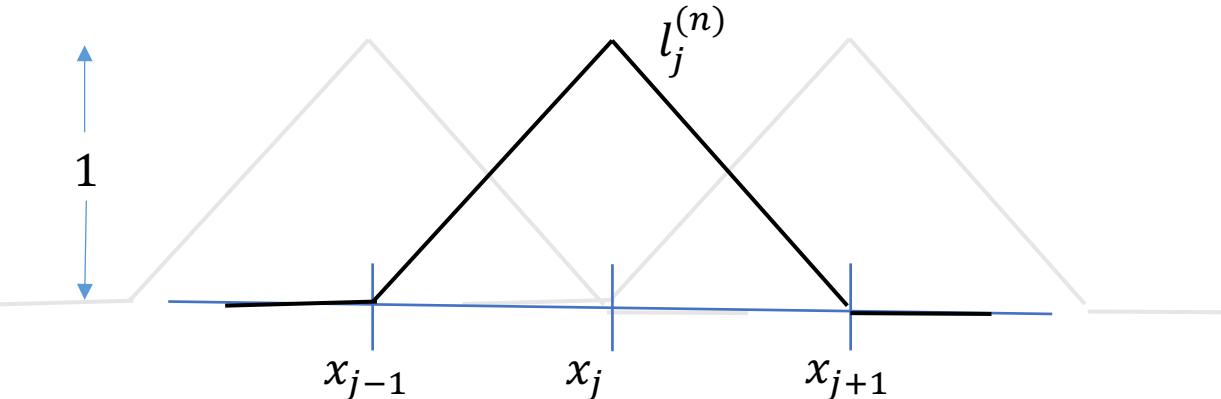
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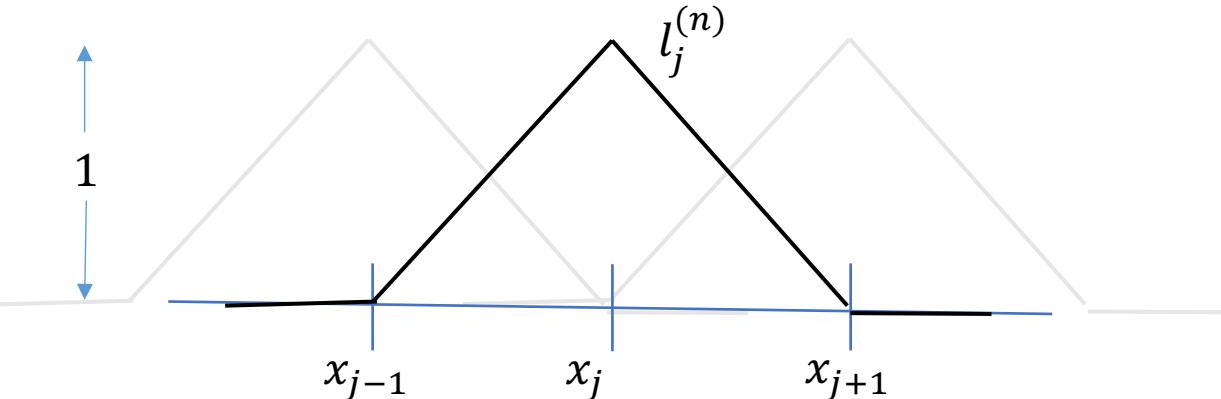
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Consider the following integral equation

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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 |

## *Lesson 5*

# *Integral Equations*

*5.2 An Introduction*

**5.3 Numerical Methods**

*- Degenerate Kernel Method*

*- via orthogonal expansion*



# Integral Equations: Numerical Methods



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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

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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



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



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**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  $D$ .