# Computational Physics

Part II: Numerical Solution of PDEs

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## Topics is Part 2

- General tools for solving partial differential equations
  - Spatial discretization
  - Linear algebra
- Methods for different types of PDEs
  - Elliptic, parabolic and hyperbolic
  - Poisson, heat and wave equation

## Today's Topics

- Why PDEs and how they arise in describing physical phenomena
- From infinite to finite dimensions
- Finite difference formulae

## Why Partial Differential Equations?

Omnipresent in physics & engineering:

$$ightharpoonup$$
 Heat  $rac{\partial u}{\partial t}=\Delta u$  Diffusion / Heat flow:  $rac{d}{dt}\int_V u\,d{f r}=-\int_{\partial V} u\,d{f r}$ 

ightharpoonup Schrödinger  $-\frac{1}{2}\Delta\psi+V\psi=E\psi$  Quantum mechanical particle in the potential V

Discuss: What needs to be addressed when finding (numerical) solutions to PDEs?

## Why Partial Differential Equations?

#### Omnipresent in physics & engineering:

- Poisson  $-\Delta u = f$
- Heat  $\frac{\partial u}{\partial t} = \Delta u$
- Wave  $\frac{\partial^2 u}{\partial t^2} = \Delta u$
- Schrödinger  $-\frac{1}{2}\Delta\psi + V\psi = E\psi$

#### Tasks:

- PDEs have infinite dimension
  - → must find a finite-dimensional representations
- Linear operators lead to linear equations
  - → need matrix algebra
- Time dependent problems
  - → need time stepping

Need both spatial and temporal discretization

## Boundary and Initial Conditions

#### Basic boundary conditions:

- Dirichlet u = 0 "zero potential"
- Neumann  $\frac{\partial u}{\partial n} = \vec{n} \cdot \nabla u = 0$

"zero normal component"  $\vec{n} \cdot \vec{E} = 0$ 

• Robin  $\frac{\partial u}{\partial n} = \alpha u + \beta g$ 

"flux depends on the value"  $k \frac{\partial T}{\partial n} = \alpha (T - T_0)$ 

#### Initial conditions:

- Heat equation  $\frac{\partial u}{\partial t} = \Delta u$  set  $u(t=0) = u_0$
- Wave equation  $\frac{\partial^2 u}{\partial t^2} = \Delta u$

set 
$$u(t=0) = u_0$$
 and  $\frac{\partial u}{\partial t}(t=0) = v_0$ 

## Step 1: Spatial discretization

- Take first 1-D:  $-\frac{\partial^2}{\partial x^2}u \rightarrow$
- Simple setting: Interval 0...L, uniform grid  $x_i = i \cdot h$ , h = L/N
- Expand  $u((i-1)h) = u(ih) u'(ih)h + \frac{1}{2}u''(ih)h^2 + \dots$  $u((i+1)h) = u(ih) + u'(ih)h + \frac{1}{2}u''(ih)h^2 + \dots$

Add and reorganize

$$-u''(ih) = \frac{1}{h^2} \left[ 2u(ih) - u((i+1)h) - u((i-1)h) \right] + \dots$$



$$-\frac{\partial^2}{\partial x^2}u\approx\frac{1}{h^2}\left(2u_i-u_{i+1}-u_{i-1}\right) \qquad \text{Famous 3-point stencil}$$

## Step 1: Spatial discretization

- How do you know if this is any good:  $-\frac{\partial^2}{\partial x^2}u \approx \frac{1}{h^2}\left(2u_i u_{i+1} u_{i-1}\right)$
- Spectral testing:

#### Continuous:

$$-u'' = \lambda u, \quad u(0) = u(L) = 0$$

$$u(x) = C_1 \sin\left(\sqrt{\lambda}x\right) + C_2 \cos\left(\sqrt{\lambda}x\right)$$

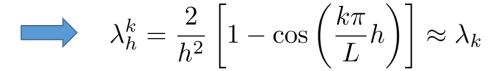
$$\begin{cases} u(0) = 0 \Rightarrow C_2 = 0 \\ u(L) = 0 \Rightarrow \lambda = \lambda_k = \left(\frac{k\pi}{L}\right)^2 \end{cases}$$

#### Discrete:

Try: 
$$u_i^k = \sin\left(\frac{k\pi}{L}ih\right)$$

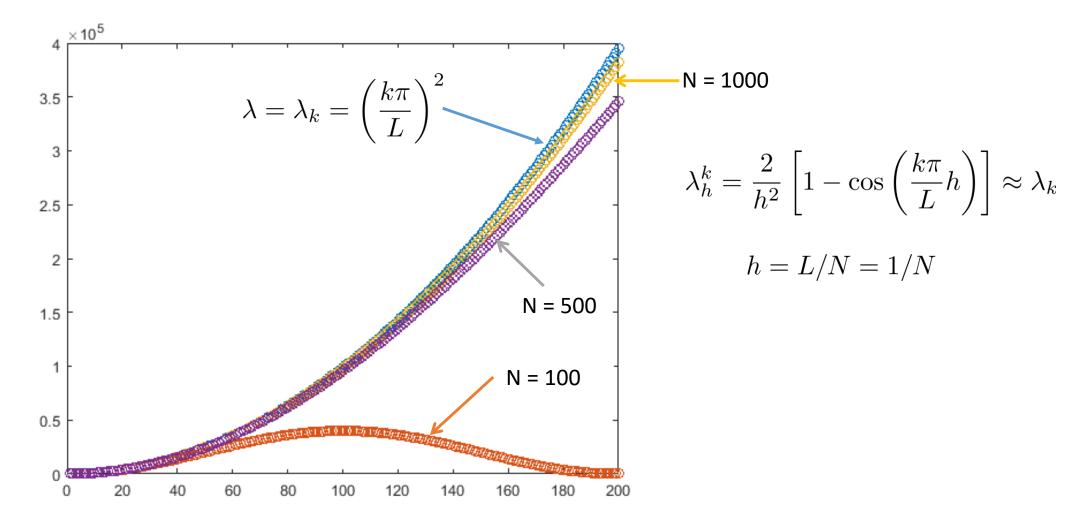


$$\frac{1}{h^2} (2u_i - u_{i+1} - u_{i-1}) = \frac{2}{h^2} \left[ 1 - \cos\left(\frac{k\pi}{L}h\right) \right] u_i$$



Ok, the approximation looks reasonable

## Spatial discretization: Spectral testing



Ok, the approximation looks still reasonable

## Another sanity check: Polynomials

$$-\frac{\partial^2}{\partial x^2}u \approx \frac{1}{h^2} \left(2u_i - u_{i+1} - u_{i-1}\right)$$

• What happens when  $u(x) = ax^2 + bx + c$ ?

$$\frac{1}{h^2} \left( 2u_i - u_{i+1} - u_{i-1} \right) = -2a \quad \text{and even when } \textit{u(x)} = \textit{x}^3 \quad \Longrightarrow \quad \frac{1}{h^2} \left( 2u_i - u_{i+1} - u_{i-1} \right) = -6ih$$

But when 
$$u(x) = x^4$$
  $\frac{1}{h^2} (2u_i - u_{i+1} - u_{i-1}) = (-12i^2 - 2)h^4$ 

Order	OK?
1	Yes
2	Yes
3	Yes
≥4	No

### More dimensions

- Since in 1D:  $-\frac{\partial^2}{\partial x^2}u \approx \frac{1}{h^2}(2u_i u_{i+1} u_{i-1})$
- 2D becomes easy:

$$-\left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2}\right)u \approx \frac{1}{h^2}\left(2u_{i,j} - u_{i+1,j} - u_{i-1,j} + 2u_{i,j} - u_{i,j+1} - u_{i,j-1}\right)$$
$$= \frac{1}{h^2}\left(4u_{i,j} - u_{i+1,j} - u_{i-1,j} - u_{i,j+1} - u_{i,j-1}\right)$$

3D is then crystal clear (7-point stencil):

$$-\left(\frac{\partial^2}{\partial x^2} + \frac{\partial^2}{\partial y^2} + \frac{\partial^2}{\partial z^2}\right)u \approx \frac{1}{h^2}\left(6u_{i,j,k} - u_{i+1,j,k} - u_{i-1,j,k} - u_{i,j+1,k} - u_{i,j-1,k} - u_{i,j,k+1} - u_{i,j,k-1}\right)$$

## Finite difference formulae from polynomials

Fit the data points with a polynomial, take the derivative of the polynomial and evaluate that at desired point.

$$\tilde{u}(x) = \frac{(x - x_1)(x - x_2)}{2h^2}u(x_0) - \frac{(x - x_0)(x - x_2)}{h^2}u(x_1) + \frac{(x - x_0)(x - x_1)}{2h^2}u(x_2) \qquad \qquad \tilde{u}''(x) = \frac{1}{h^2}u(x_0) - \frac{2}{h^2}u(x_1) + \frac{1}{h^2}u(x_2)$$

#### Forward difference formula

$$x_0 = ih, \ x_1 = (i+1)h, \ x_2 = (i+2)h$$
 
$$-u''(ih) = \frac{1}{h^2} \left[ 2u((i+1)h) - u(ih) - u((i+2)h) \right]$$

#### Backward difference formula

$$x_0 = (i-2)h, \ x_1 = (i-1)h, \ x_2 = ih$$
 
$$-u''(ih) = \frac{1}{h^2} \left[ 2u((i-1)h) - u(ih) - u((i-2)h) \right]$$

#### Central difference formula

$$x_0 = (i-1)h, \ x_1 = ih, \ x_2 = (i+1)h$$
 
$$-u''(ih) = \frac{1}{h^2} \left[ 2u(ih) - u((i+1)h) - u((i-1)h) \right]$$

### Finite difference formulae from polynomials

Fit the data points with a polynomial, take the derivative of the polynomial and evaluate that at desired point.

#### Three-point formula for the 2nd derivative:

```
Nn = 1; (*Number of points at both directions*)
Print["Total number of points is ", 2 Nn + 1]
Total number of points is 3
Dd = 2; Print["Derivative of the order ", Dd]
Derivative of the order 2
g[x_] := (*Interpolate a polynomial*)
 Evaluate [InterpolatingPolynomial [
   Table [\{x0+i, y[i]\}, \{i, -Nn, Nn\}], x]
gp[x_] := Evaluate[D[g[x], \{x, Dd\}]]
 (*Derivative of the interpolating polynomial*)
Simplify[
 Expand [Collect [gp[x0+0],
   Table[y[i], {i, -Nn, Nn}]]]]
y[-1] - 2y[0] + y[1]
```

#### Five-point formula for the 2nd derivative:

```
Nn = 2; (*Number of points at both directions*)
Print["Total number of points is ", 2 Nn + 1]
Total number of points is 5
Dd = 2; Print["Derivative of the order ", Dd]
Derivative of the order 2
g[x] := (*Interpolate a polynomial*)
 Evaluate[InterpolatingPolynomial[
   Table [\{x0+i, y[i]\}, \{i, -Nn, Nn\}], x]
gp[x] := Evaluate[D[g[x], \{x, Dd\}]]
 (*Derivative of the interpolating polynomial*)
Simplify[
 Expand [Collect [gp[x0+0],
   Table[y[i], {i, -Nn, Nn}]]]]
\frac{1}{12} (-y[-2] + 16y[-1] - 30y[0] + 16y[1] - y[2])
```

### Convergence of the finite difference formulae

Test the central difference formula for exp(x) at x=0

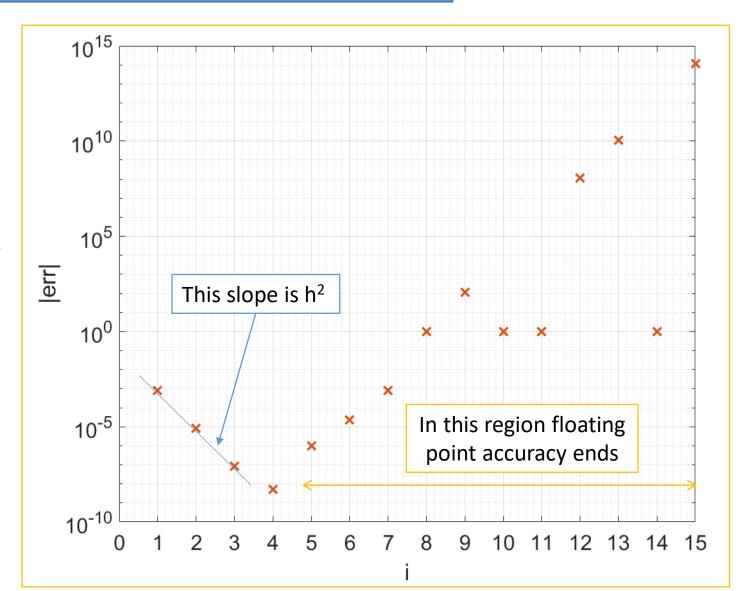
Exact: 
$$\frac{d^2}{dx^2}e^x = e^x$$
  $x = 0 \Rightarrow e^0 = 1$ 

#### Approximation:

$$e^0 \approx \frac{1}{h^2} \left[ e^h - 2 + e^{(-h)} \right], \quad h = 10^{-i}, \ i = 1, \dots, N$$

Measure of error:

$$err = \frac{1}{h^2} \left[ e^h - 2 + e^{(-h)} \right] - 1$$

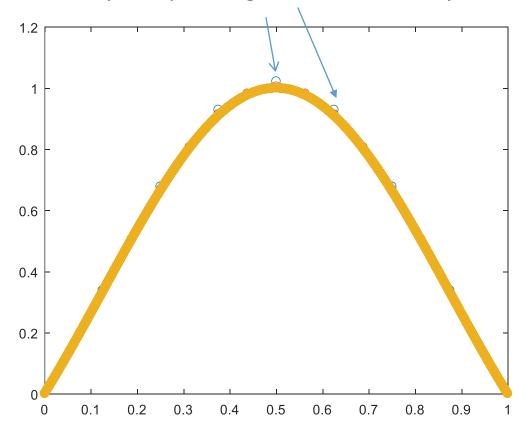


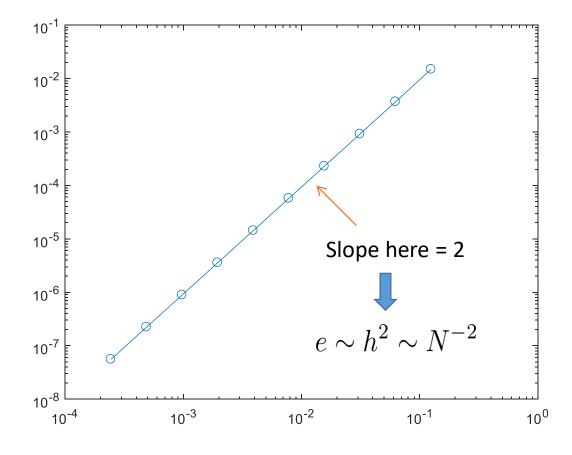
### Solving the Poisson equation: 1D

Easy case: 
$$u(x) = \sin\left(\frac{\pi x}{L}\right) \exp\left(-(x-L/2)^2\right)$$
 
$$N = 2^k, \quad k = 3\dots 12$$

Error as: 
$$e = \sqrt{h \sum_{i} (u(x_i) - u_i)^2}$$

#### Only the sparsest grid differs noticeably

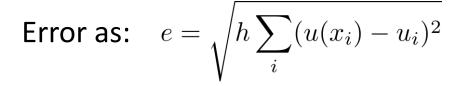


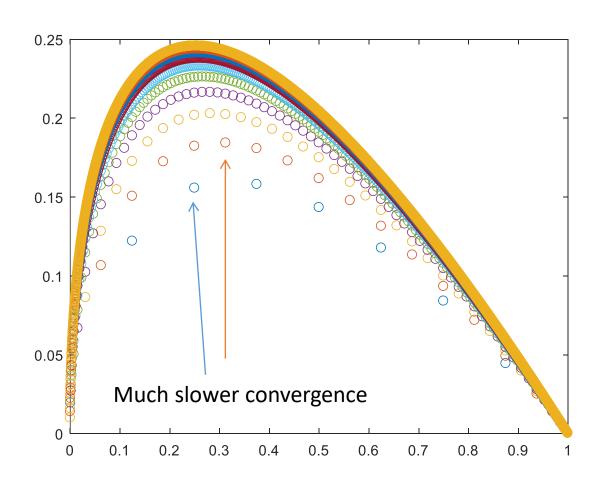


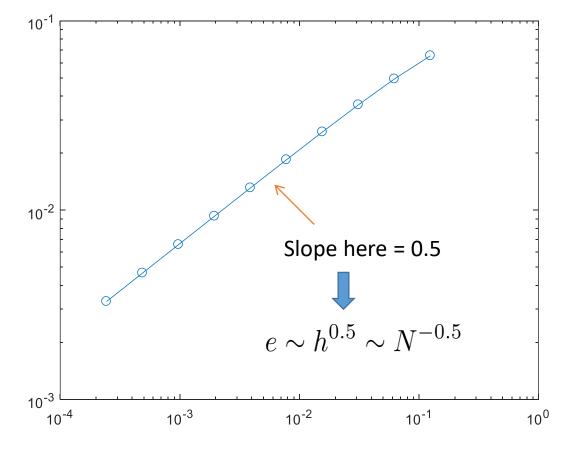
### Solving the Poisson equation: 1D

Tougher case:  $u(x) = \sqrt{x} - x$ 

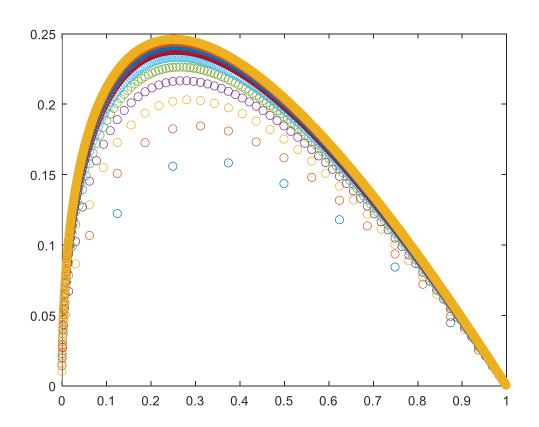
$$N = 2^k, \quad k = 3 \dots 12$$

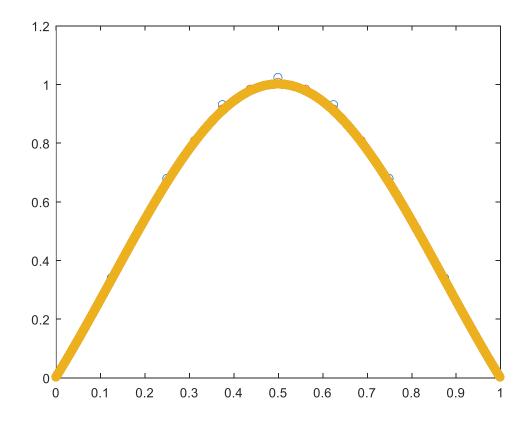






## Finite Difference: Convergence





Discuss: What is the difference between left and right?

## From classical to quantum physics

Classical Hamiltonian of a particle (mass m) in external potential V:

$$H = \frac{p^2}{2m} + V(x)$$

that is the energy of the system, kinetic and potential energy.

We can get to quantum mechanics simply by:

$$p o -\mathrm{i}\hbar \frac{\partial}{\partial x}$$

so that the quantum Hamiltonian is an operator that looks like:

$$\mathcal{H} = -rac{\hbar^2}{2m}rac{\partial^2}{\partial x^2} + V(x)$$

This acts on the wave function, and gives the energy:  $\ensuremath{\mathcal{H}}\psi(x)=E\psi(x)$  .

$$\mathcal{H}\psi(x) = E\psi(x)$$

$$-\frac{1}{2}\frac{\partial^2 \psi(x)}{\partial x^2} + V(x)\psi(x) = E\psi(x).$$

This is known as the 1D Schrödinger equation.

(All constants set to 1)

## Two familiar cases of 1D Schrödinger

a) Particle in a box of size L: V=0, 0 < x < L,  $V=\infty$  outside.

$$\phi_i = \sin\left(irac{\pi x}{L}
ight)\sqrt{rac{2}{L}} \ ,$$

where  $i=1,2,3,\ldots$ , and energy

$$E_i=rac{\pi^2 i^2}{2L^2}\,.$$

b) Particle in a harmonic potential:  $V=rac{1}{2}\omega^2x^2$ .

$$\phi_i = (\omega/\pi)^{1/4} \frac{1}{\sqrt{2^i i!}} H_i(\sqrt{\omega}x) \exp(-\omega x^2/2) ,$$

where  $i=0,1,2,3,\ldots$ , and energy is

$$E_i = \omega \left( i + rac{1}{2} 
ight) \ .$$

## Finite differences for 1D Schrödinger

Take the equation to be solved:

$$-\frac{1}{2}\frac{\partial^2 \psi(x)}{\partial x^2} + V(x)\psi(x) = E\psi(x)$$

and discretize the space using an uniform grid

$$\{x o \{x_i\}_{i=1}^N = -x_m, -x_m+h, -x_m+2h, \dots, x_m-h, x_m\}$$

so that the wave function becomes a vector of length N:

$$\psi(x) \to \{\psi(x_i)\}_{i=1}^N = \{\psi_i\}_{i=1}^N$$

Potential also at the same grid:  $\,V(x_i)\,$ 

with elements: 
$$V_{i,j} = \delta_{ij} V(x_i)$$

## Finite differences for 1D Schrödinger

Kinetic energy by the finite difference:

$$T\psi(x) = -\frac{1}{2}\frac{d^2\psi(x)}{dx^2} \approx \left\{ -\frac{1}{2} \left[ \psi(x+h) + \psi(x-h) \right] + \psi(x) \right\} / h^2$$

or

$$T\psi_i = -rac{1}{2h^2}\psi_{i-1} - rac{1}{2h^2}\psi_{i+1} + rac{1}{h^2}\psi_i$$

Kinetic energy operator couples the wave function values at neighbouring sites.

$$T_{i,i}=rac{1}{h^2}$$
 ,  $T_{i,i+1}=-rac{1}{2h^2}$  ,  $T_{i,i-1}=-rac{1}{2h^2}$  ,

and the of the elements rest are zero. This is a tridiagonal matrix.

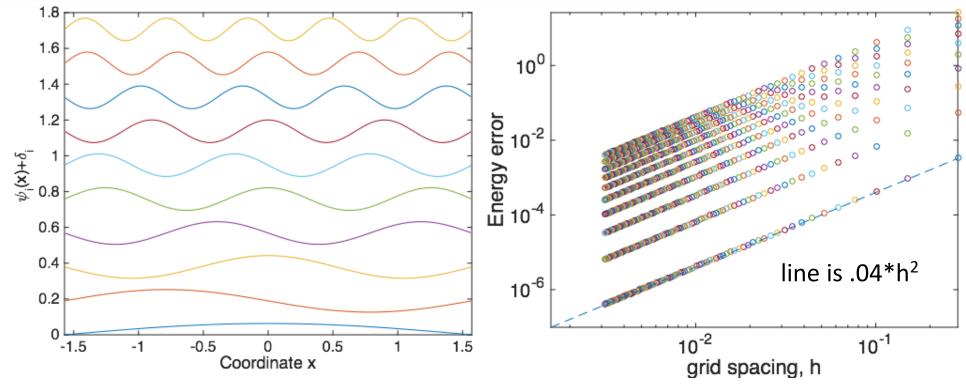
So the Schrödinger equation maps to a tridiagonal matrix eigenvalue equation:

$$(T+V)\psi = E\psi$$

### Test for the particle in a box

```
L=pi;
omega=0;
Nx=500;
h=L/(Nx+1);
x=-.5*L+h*(1:Nx);
V=.5*(omega*x).^2;

T=(diag(ones(Nx,1))-.5*diag(ones(Nx-1,1),1)-.5*diag(ones(Nx-1,1),-1))./h^2;
[wf, enes]=eig(T+diag(V));
```



### Homework 7: FD stencils in 1D and 2D

a) Dimension 1. Implement a solver for the one-dimensional problem

$$\begin{cases} -u''(x) = (x - 0.5)^3 - 2(x - 0.5), & 0 < x < 1 \\ u(0) = u(1) = 0 \end{cases}$$

using the three-point stencil

$$-u''(x_i) \approx \frac{1}{h^2}(-u_{i-1}+2u_i-u_{i+1})$$

Check that when varying the grid spacing h you converge to the exact solution at the rate  $h^2$ .

c) Dimension 2. Implement a solver for the two-dimensional problem

$$\begin{cases} -\Delta u(x,y) = \exp(-\frac{(x-0.5)^2 + (y-0.5)^2}{18}), & (x,y) \in [0,1] \times [0,1] \\ u(0,y) = u(1,y) = u(x,0) = u(x,1) = 0 \end{cases}$$

using the five-point stencil (see, e.g.,

https://en.wikipedia.org/wiki/Discrete\_Poisson\_equation)

$$-\Delta u(x_i, y_j) \approx \frac{1}{h^2} (4u_{i,j} - u_{i-1,j} - u_{i+1,j} - u_{i,j-1} - u_{i,j+1})$$

What can you say about the convergence with respect to the grid specing h? (3 p)

### Finite Differences: Pros & Cons

Pros: Cons:

Discuss: What are the Pros and Cons of Finite Differences