

# **Course 18.327 and 1.130**

## **Wavelets and Filter Banks**

**Numerical solution of PDEs: Galerkin approximation; wavelet integrals (projection coefficients, moments and connection coefficients); convergence**

# Numerical Solution of Differential Equations

**Main idea:** look for an approximate solution that lies in  $V_j$ .  
Approximate solution should converge to true solution as  $j \rightarrow \infty$ .

**Consider the Poisson equation**

$$\frac{\partial^2 u}{\partial x^2} = f(x) \text{ ----- } \square \quad \left( \begin{array}{l} \text{leave boundary} \\ \text{conditions till later} \end{array} \right)$$

**Approximate solution:**

$$u_{\text{approx}}(x) = \sum_k c[k] 2^{j/2} \underbrace{\phi(2^j x - k)}_{\phi_{j,k}(x)} \text{ ----- } \square$$

**trial functions**

**Method of weighted residuals: Choose a set of test functions,  $g_n(x)$ , and form a system of equations (one for each  $n$ ).**

$$\int \frac{\partial^2 u_{\text{approx}}}{\partial x^2} g_n(x) dx = \int f(x) g_n(x) dx$$

**One possibility: choose test functions to be Dirac delta functions. This is the collocation method.**

$$g_n(x) = \delta(x - n/2^j) \quad n \text{ integer}$$

$\Rightarrow$   $\sum_k c[k] \phi_{j,k}''(n/2^j) = f(n/2^j)$  -----□

**Second possibility: choose test functions to be scaling functions.**

- Galerkin method if synthesis functions are used (test functions = trial functions)
- Petrov-Galerkin method if analysis functions are used

**e.g. Petrov-Galerkin**

$$\mathbf{g}_n(\mathbf{x}) = \tilde{\phi}_{j,n}(\mathbf{x}) \in \tilde{\mathbf{V}}_j$$

$$\Rightarrow \sum_{\mathbf{k}} \mathbf{c}[\mathbf{k}] \int_{-\infty}^{\infty} \frac{\partial^2}{\partial \mathbf{x}^2} \phi_{j,\mathbf{k}}(\mathbf{x}) \cdot \tilde{\phi}_{j,n}(\mathbf{x}) \, d\mathbf{x} = \int_{-\infty}^{\infty} f(\mathbf{x}) \tilde{\phi}_{j,n}(\mathbf{x}) \, d\mathbf{x} \quad \text{-----} \square$$

**Note: Petrov-Galerkin  $\equiv$  Galerkin in orthogonal case**

Two types of integrals are needed:

(a) Connection Coefficients

$$\begin{aligned}\int_{-\infty}^{\infty} \frac{\partial^2}{\partial x^2} \phi_{j,k}(x) \cdot \tilde{\phi}_{j,n}(x) dx &= 2^{2j} \int_{-\infty}^{\infty} 2^{j/2} \phi''(2^j x - k) 2^{j/2} \tilde{\phi}(2^j x - n) dx \\ &= 2^{2j} \int_{-\infty}^{\infty} \phi''(\tau) \tilde{\phi}(\tau + k - n) d\tau \\ &= 2^{2j} h_{\partial^2/\partial x^2} [n - k]\end{aligned}$$

where  $h_{\partial^2/\partial x^2} [n]$  is defined by

$$h_{\partial^2/\partial x^2} [n] = \int_{-\infty}^{\infty} \phi''(t) \tilde{\phi}(t - n) dt \quad \square$$

↑  
connection coefficients

## (b) Expansion coefficients

The integrals  $\int_{-\infty}^{\infty} f(x) \tilde{\phi}_{j,n}(x) dx$  are the coefficients for the expansion of  $f(x)$  in  $V_j$ .

$$f_j(x) = \sum_k r_j[k] \phi_{j,k}(x)$$

with

$$r_j[k] = \int_{-\infty}^{\infty} f(x) \tilde{\phi}_{j,k}(x) dx$$



So we can write the system of Galerkin equations as a convolution:

$$2^{2j} \sum_k c[k] h_{\partial^2/\partial x^2}[n-k] = r_j[n]$$



⇒ Solve a deconvolution problem to find  $c[k]$  and then find  $u_{\text{approx}}$  using equation □.

**Note:** we must allow for the fact that the solution may be non-unique, i.e.  $H_{\partial^2/\partial x^2}(\omega)$  may have zeros.

**Familiar example: 3-point finite difference operator**

$$h_{\partial^2/\partial x^2}[n] = \{1, -2, 1\}$$

$$H_{\partial^2/\partial x^2}(z) = 1 - 2z^{-1} + z^{-2} = (1 - z^{-1})^2$$

⇒  $H_{\partial^2/\partial x^2}(\omega)$  has a 2<sup>nd</sup> order zero at  $\omega = 0$ .

Suppose  $u_0(x)$  is a solution. Then  $u_0(x) + Ax + B$  is also a solution. Need boundary conditions to fix  $u_{\text{approx}}(x)$ .

## Determination of Connection Coefficients

$$h_{\partial^2/\partial x^2}[n] = \int_{-\infty}^{\infty} \phi''(t) \tilde{\phi}(t - n) dt$$

Simple numerical quadrature will not converge if  $\phi''(t)$  behaves badly.

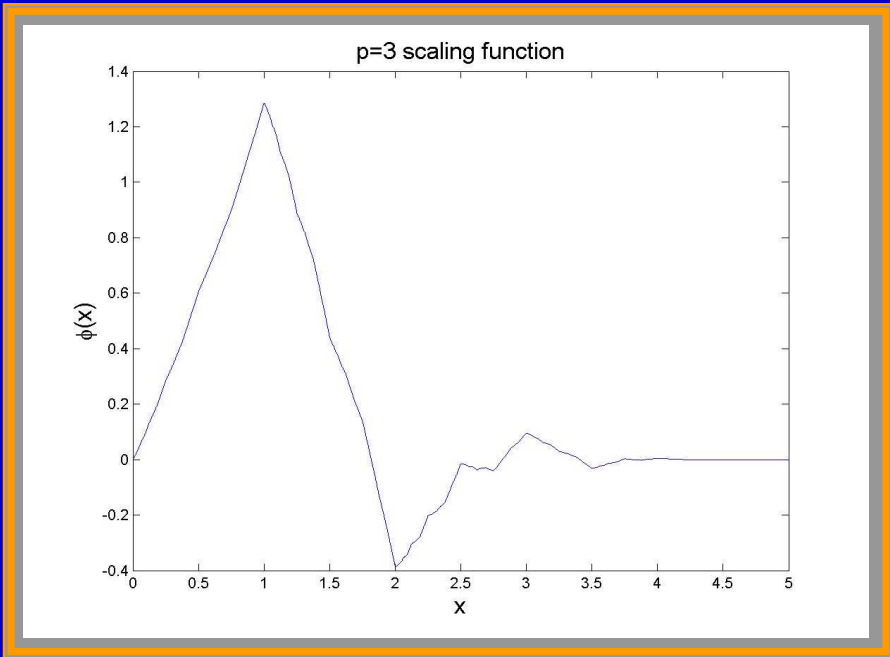
Instead, use the refinement equation to formulate an eigenvalue problem.

$$\left. \begin{aligned} \phi(t) &= 2 \sum_k f_0[k] \phi(2t - k) \\ \phi''(t) &= 8 \sum_k f_0[k] \phi''(2t - k) \\ \tilde{\phi}(t - n) &= 2 \sum_\ell h_0[\ell] \tilde{\phi}(2t - 2n - \ell) \end{aligned} \right\} \begin{array}{l} \text{Multiply and} \\ \text{Integrate} \end{array}$$

So

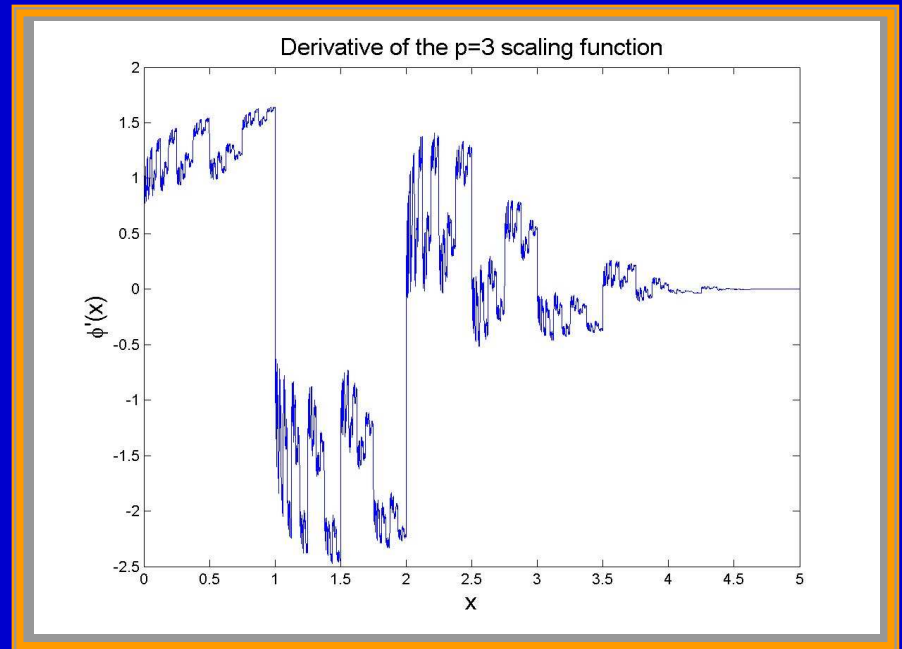
$$h_{\partial^2/\partial x^2}[n] = 8 \sum_k f_0[k] \sum_\ell h_0[\ell] h_{\partial^2/\partial x^2}[2n + \ell - k]$$





## Daubechies 6 scaling function

## First derivative of Daubechies 6 scaling function



Reorganize as

$$h_{\partial^2/\partial x^2}[n] = 8 \sum_m h_0[m - 2n] (\sum_k f_0[m - k] h_{\partial^2/\partial x^2}[k])$$

$$m = 2n + \ell$$

Matrix form

$$h_{\partial^2/\partial x^2} = 8 \mathbf{A} \mathbf{B} h_{\partial^2/\partial x^2} \longrightarrow \text{eigenvalue problem}$$

Need a normalization condition  $\longrightarrow$  use the moments of the scaling function:

If  $h_0[n]$  has at least 3 zeros at  $\pi$ , we can write

$$\sum_k \mu_2[k] \phi(t - k) = t^2 ; \mu_2[k] = \int_{-\infty}^{\infty} t^2 \tilde{\phi}(t - k) dt$$

Differentiate twice, multiply by  $\tilde{\phi}(t)$  and integrate:

$$\sum_k \mu_2[k] h_{\partial^2/\partial x^2}[-k] = 2! \longrightarrow \text{Normalizing condition}$$

## Formula for the moments of the scaling function

$$\mu_k^l = \int_{-\infty}^{\infty} \tau^l \phi(\tau - k) d\tau$$

### Recursive formula

$$\mu_0^0 = \int_{-\infty}^{\infty} \phi(\tau) d\tau = 1$$

$$\mu_0^r = \frac{1}{2^{r-1}} \sum_{i=0}^{r-1} \binom{r}{i} \left( \sum_{k=0}^N h_0[k] k^{r-i} \right) \mu_0^i$$

$$\mu_k^l = \sum_{r=0}^l \binom{l}{r} k^{l-r} \mu_0^r$$

## How to enforce boundary conditions?

One idea – extrapolate a polynomial:

$$u(\mathbf{x}) = \sum_k c[k] \phi_{j,k}(\mathbf{x}) = \sum_{l=0}^{p-1} a[l] x^l$$

Relate  $c[k]$  to  $a[l]$  through moments. Extend  $c[k]$  by extending underlying polynomial.

Extrapolated polynomial should satisfy boundary constraints:

Dirichlet:

$$u(x_0) = \alpha \Rightarrow \sum_{l=0}^{p-1} a[l] x_0^l = \alpha$$

Neumann:

$$u'(x_0) = \beta \Rightarrow \sum_{l=0}^{p-1} a[l] l x_0^{l-1} = \beta$$

Constraint  
on  $a[l]$



## Convergence

**Synthesis scaling function:**

$$\phi(x) = 2 \sum_k f_0[k] \phi(2x - k)$$

**We used the shifted and scaled versions,  $\phi_{j,k}(x)$ , to synthesize the solution. If  $F_0(\omega)$  has  $p$  zeros at  $\pi$ , then we can exactly represent solutions which are degree  $p - 1$  polynomials.**

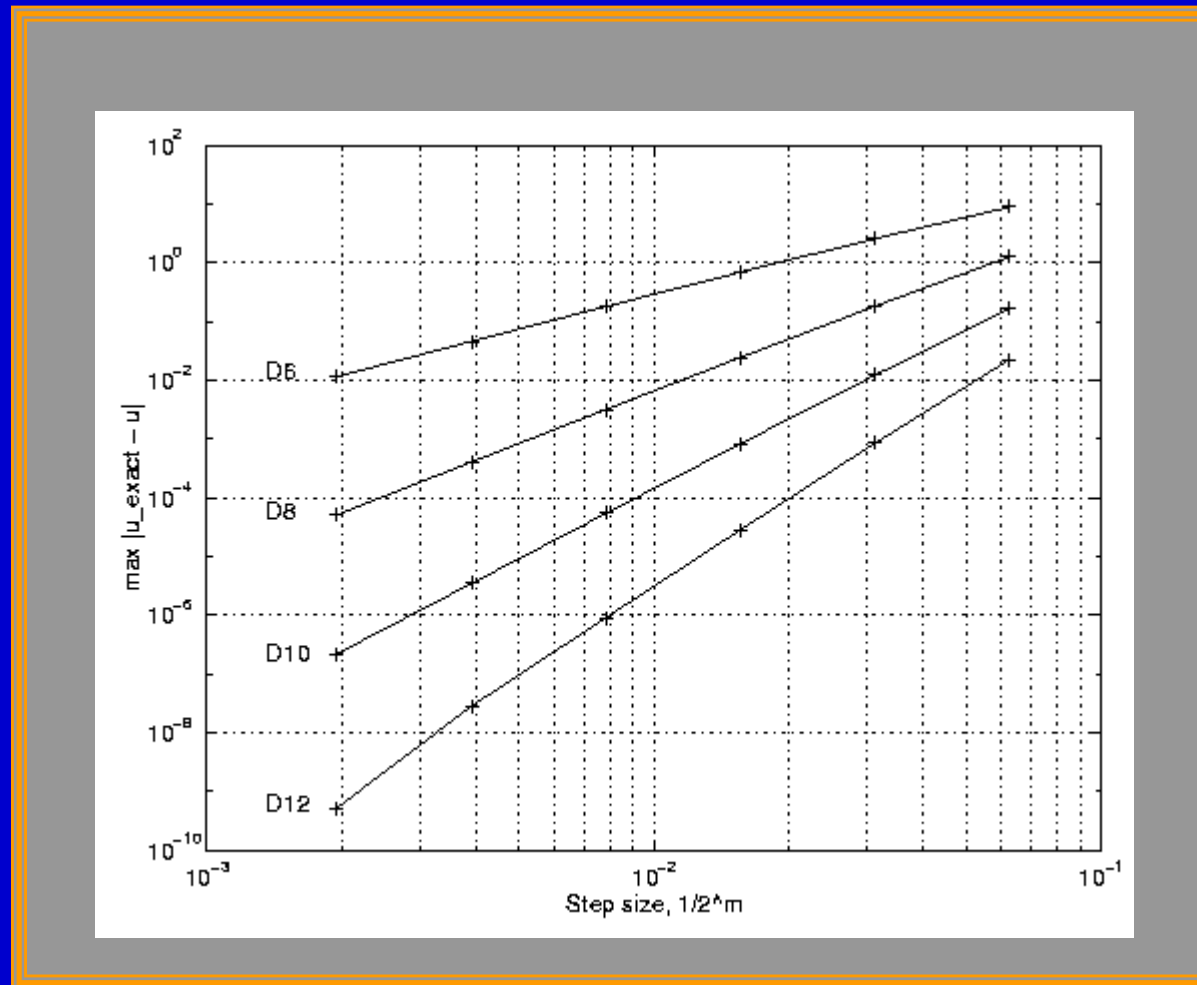
**In general, we hope to achieve an approximate solution that behaves like**

$$u(x) = \sum_k c[k] \phi_{j,k}(x) + O(h^p)$$

**where**

$$h = \frac{1}{2^j} = \text{spacing of scaling functions}$$

## Reduction in error as a function of h



# Multiscale Representation

e.g.  $\partial^2 u / \partial x^2 = f$

Expand as

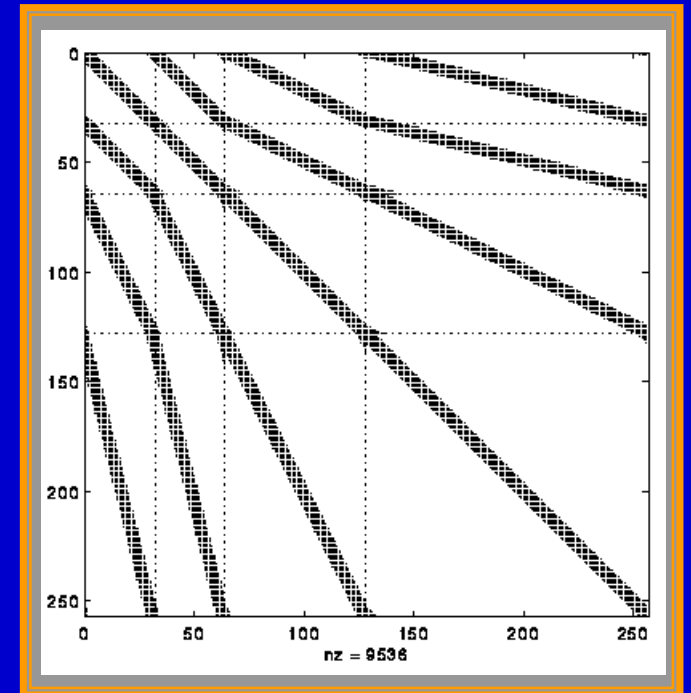
$$u = \sum_k c_k \phi(x - k) + \sum_{j=0}^J \sum_k d_{j,k} w(2^j x - k)$$

Galerkin gives a system

$$Ku = f$$

with typical entries

$$K_{m,n} = 2^{2j} \int_{-\infty}^{\infty} \frac{\partial^2}{\partial x^2} w(x - n) w(x - m) dx$$







# **Matlab Example**

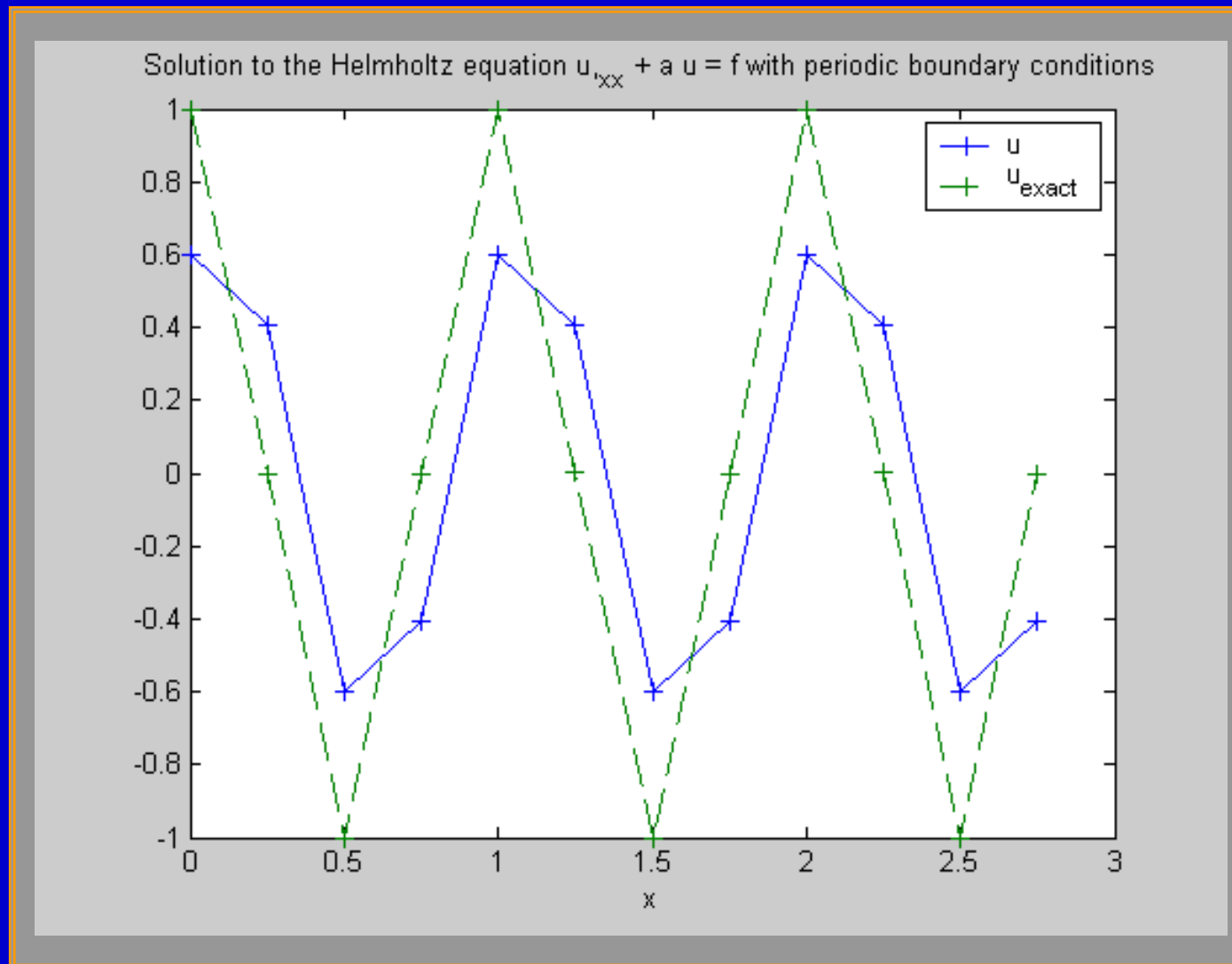
## **Numerical solution of Partial Differential Equations**

# The Problem

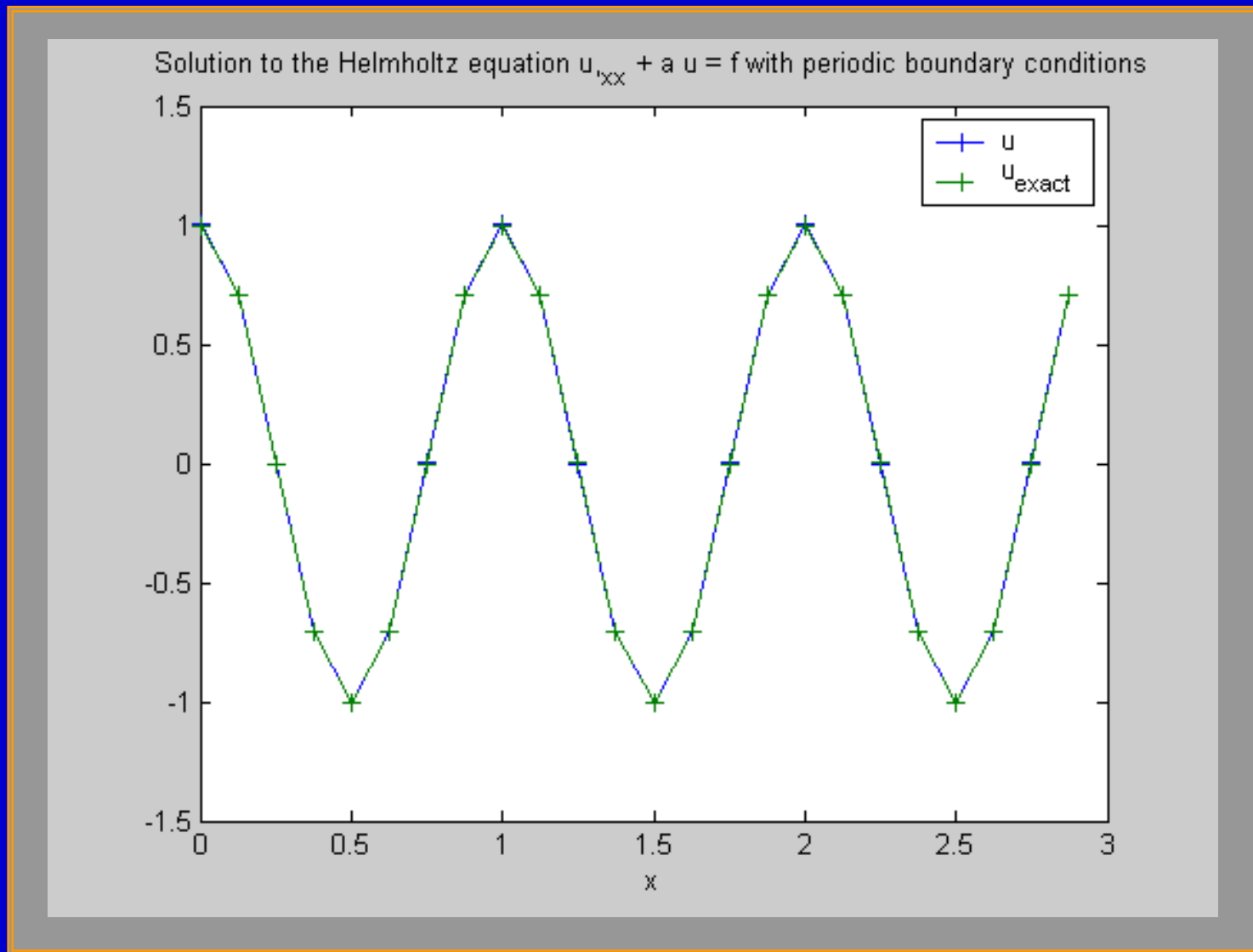
1. Helmholtz equation:  $u_{xx} + a u = f$

- $p=6;$             % Order of wavelet scheme ( $p_{min}=3$ )
- $a = 0$
- $L = 3;$             % Period.
- $nmin = 2;$         % Minimum resolution
- $nmax = 7;$         % Maximum resolution

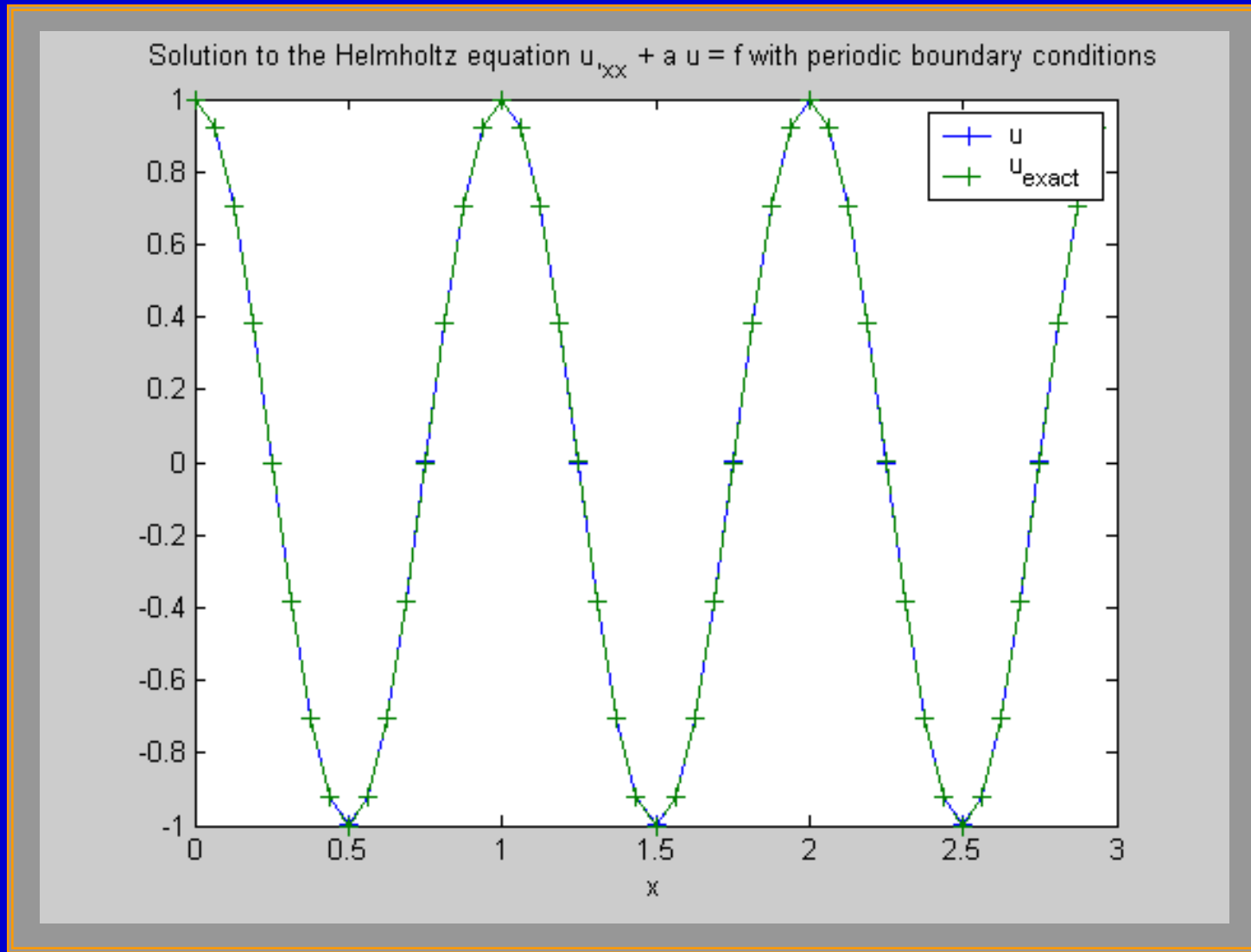
# Solution at Resolution 2



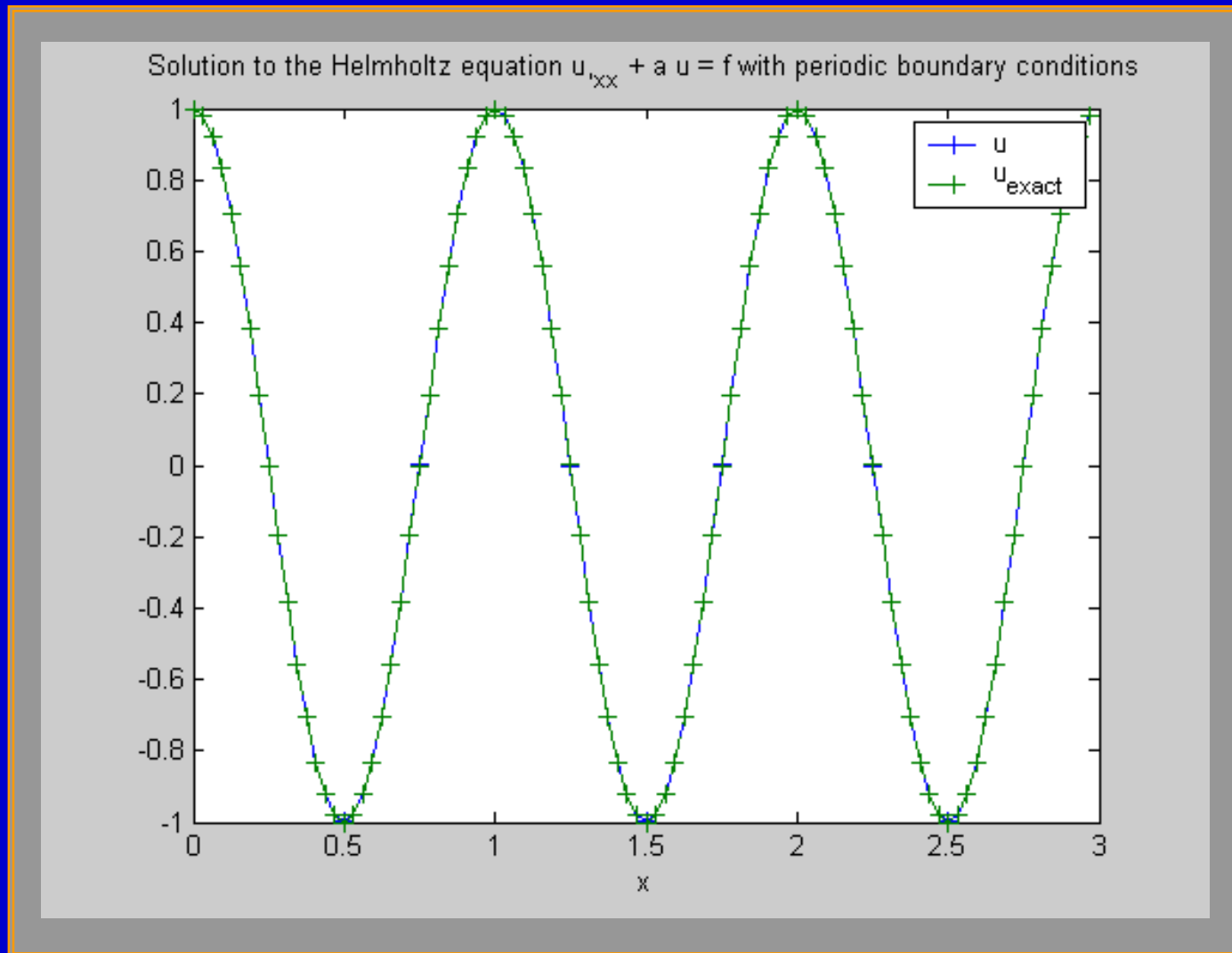
# Solution at Resolution 3



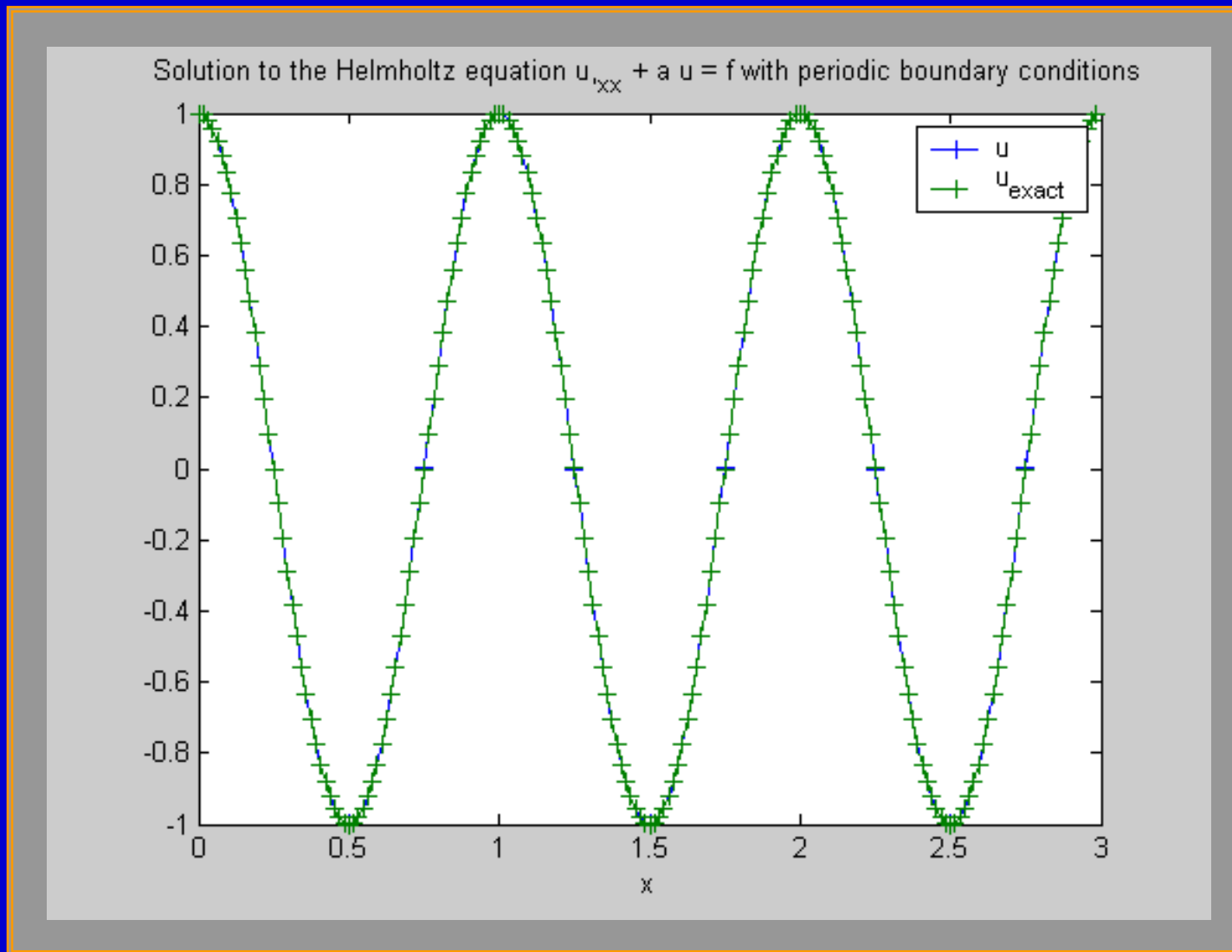
# Solution at Resolution 4



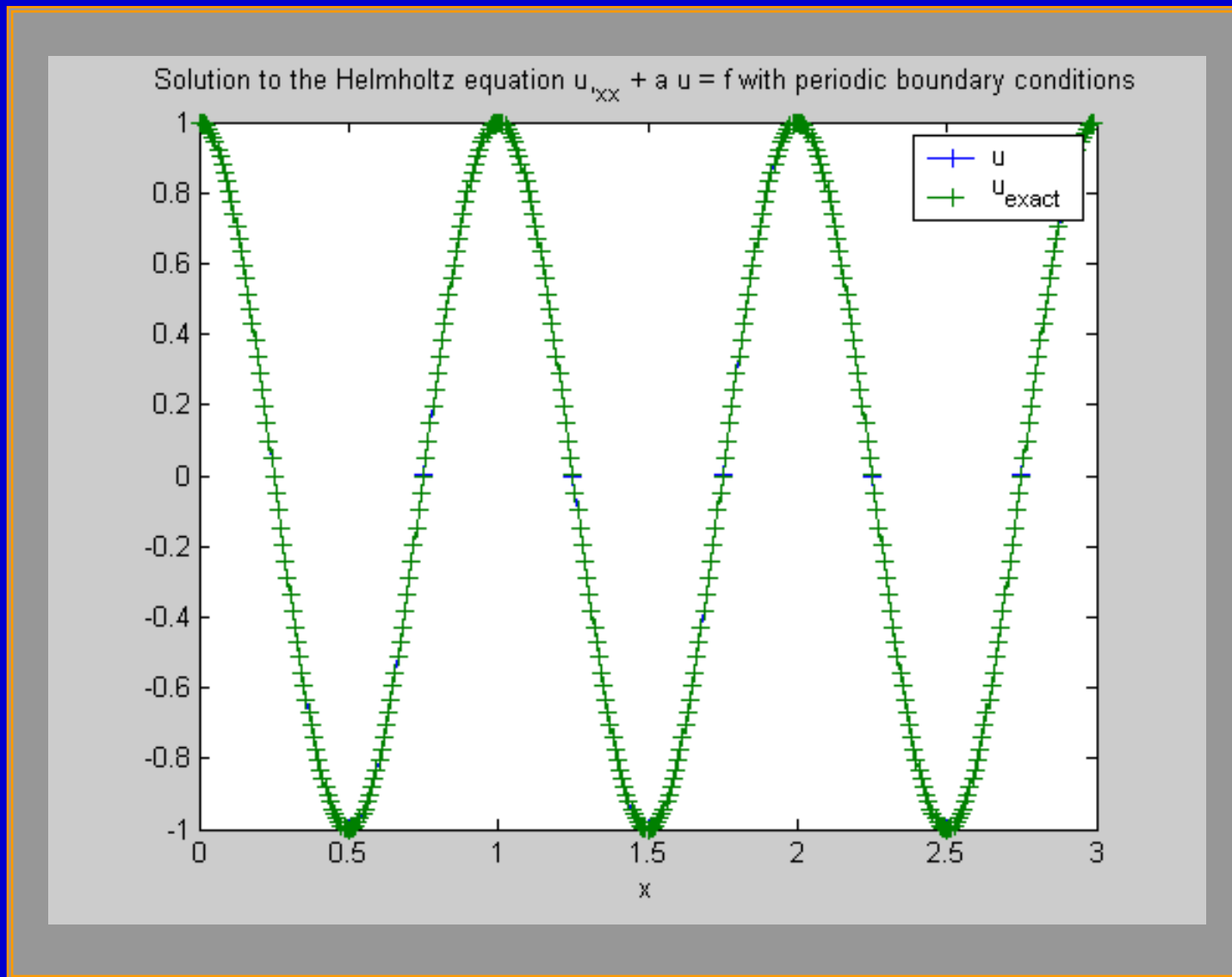
# Solution at Resolution 5



# Solution at Resolution 6

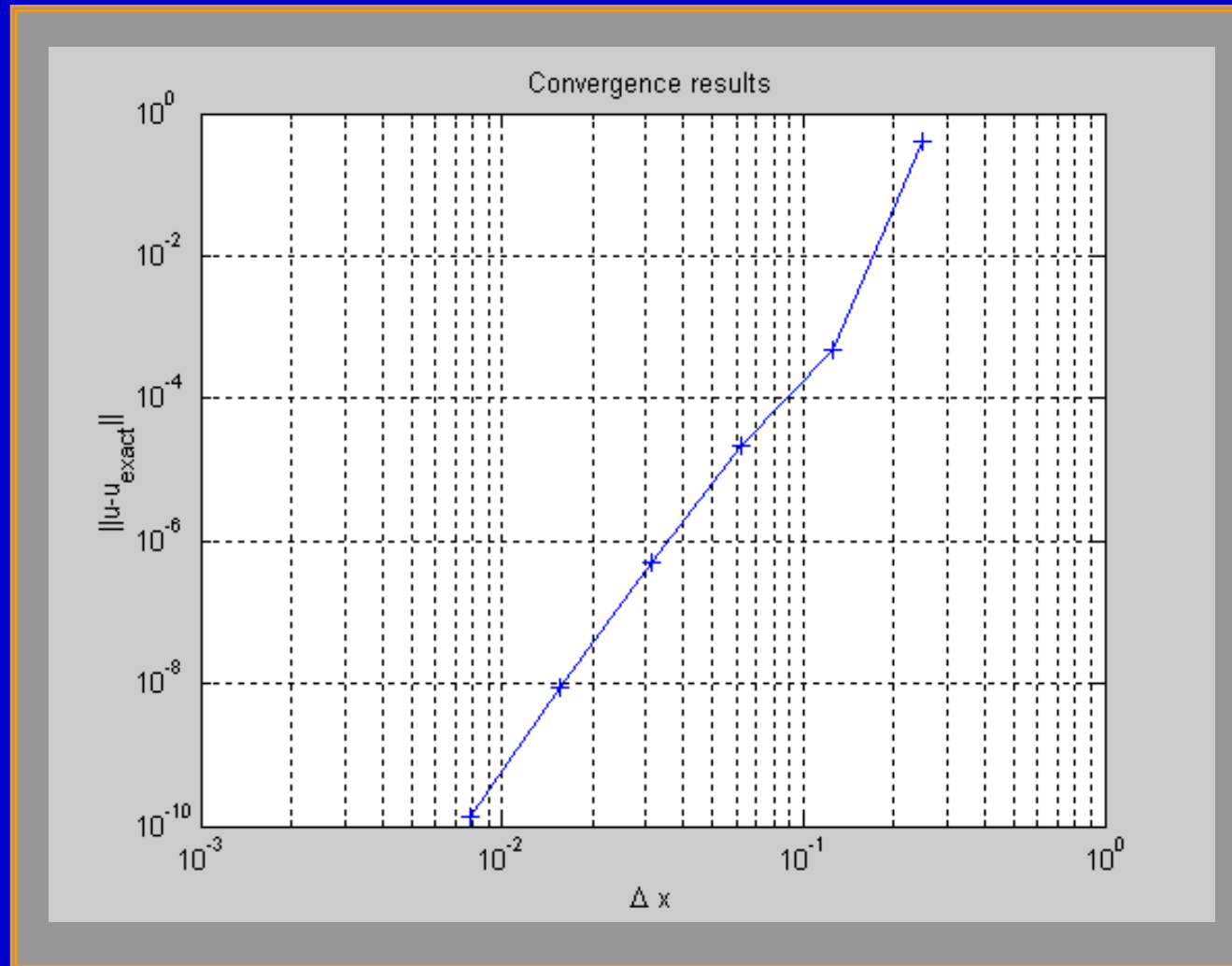


# Solution at Resolution 7





# Convergence Results



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>> helmholtz slope = 5.9936
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