

## FINITE ELEMENT METHODS

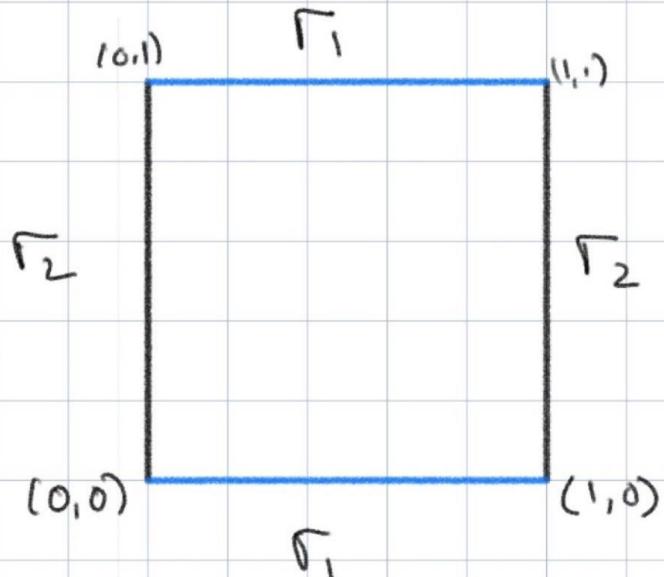
$$-\nabla^2 u = f \quad \text{on} \quad (x, y) \in [0, 1]^2 \quad (+)$$

$$f(x, y) = 2\pi^2 \sin(\pi x) \cos(\pi y)$$

$$u(0, y) = u(1, y) = 0$$

$$\partial_y u(x, y) \Big|_{y=0} = \partial_y u(x, y) \Big|_{y=1} = 0$$

diagram



STEP 1 part a)

⇒ we first introduce a test function,  
which is here denoted  $\varphi$ :

$w(x, y)$

$\Rightarrow$  we then multiply ( $\nabla u$ ) by the test function and integrate by parts over the entire domain ( $\Omega$ )

$$-\int_0^1 \int_0^1 w(x, y) \nabla^2 u \, dx \, dy = \int_0^1 \int_0^1 w(x, y) f(x, y) \, dx \, dy$$

$\Rightarrow$  integrating the L.H.S by parts then gives

$$\int_0^1 \int_0^1 \nabla(w) \cdot \nabla u \, d\Omega - \int w \hat{n} \cdot \nabla u \, d\Gamma = \int_0^1 \int_0^1 w f \, d\Omega$$

If we then look at the boundary conditions provided

$$\Gamma_2: [\hat{n} \cdot \nabla u] = (0) \cdot \left( \frac{\partial u}{\partial y} \right) = \partial_y u$$

\* from B.Cs  $\partial_y u(x, y)|_{y=0} = \partial_y v(x, y)|_{y=1} = 0$  (neumann)

such that  $\partial_y v$  at  $\Gamma_2 = 0$

and then

$$\Gamma_1: \int \hat{n} \cdot \nabla u = \begin{pmatrix} 1 \\ 0 \end{pmatrix} \cdot \begin{pmatrix} \frac{\partial u}{\partial x} \\ \frac{\partial u}{\partial y} \end{pmatrix} = \sum \frac{\partial u}{\partial x} = 0$$

does not work

and

$$u(0,y) = a, \quad w(0,y) = 0 \\ u(1,y) = b, \quad w(1,y) = 0$$

$$\text{so } \Gamma_1 \rightarrow 0$$

$$\int \int \int \nabla(w) \cdot \nabla u \, d\Omega - \cancel{\int \int \hat{n} \cdot \nabla u \, d\Gamma}^0 = \int \int \int f \, d\Omega$$

and the weak formulation becomes:

$$\int \int \int \nabla(w) \cdot \nabla u \, dx \, dy = \int \int \int f \, dx \, dy$$

$$\Rightarrow \boxed{\int \int (\nabla(w) \cdot \nabla u - wf) \, dx \, dy = 0}$$

weak form

$$\therefore \delta \int \int \frac{1}{2} |\nabla u|^2 - fu \, dx \, dy = 0 \quad (++)$$

## Step 1 part b)

⇒ the minimisation problem

$$F[u] = \iint \frac{1}{2} |\nabla u|^2 - uf \, dx \, dy$$

⇒ consider a class of functions

$$u(x) = v(x) + \epsilon s v(x, y)$$

⇒ which is the variation around  $v(x)$

⇒ The condition for the existence  
of a minimum

$$\frac{d F(u)}{d \epsilon} = 0$$

⇒ plugging the class of functions  
into the minimization problem  
and differentiating with respect  
to  $\epsilon$  gives us the weak formulation

$$0 = \frac{d}{d \epsilon} \left[ \iint \frac{1}{2} |\nabla(v + \epsilon s v)|^2 - (v + \epsilon s v) f \, dx \, dy \right]$$

$$0 = \iint_D \left[ \frac{d}{dx} \left( \frac{1}{2} \frac{\partial (U + \epsilon S U)}{\partial x} \right)^2 + \left( \frac{1}{2} \frac{\partial (U + \epsilon S U)}{\partial y} \right)^2 \right] - \frac{\partial (U + \epsilon S U)}{\partial x} f \, dx \, dy$$

which then reduces to

$$0 = \iint_D \frac{\partial (U + \epsilon S U)}{\partial x} \frac{\partial (S u)}{\partial x} + \frac{\partial (U + \epsilon S U)}{\partial y} \frac{\partial (S u)}{\partial y} - f S u \, dx \, dy$$

and that becomes

$$\boxed{\iint_D \nabla U \cdot \nabla S u - f S u \, dx \, dy = 0}$$

This is the same as the weak form derived in step 1a with  
 $w(x, y) = S u(x, y)$

## STEP 2

⇒ The solution is approximated by a finite dimensional subset of the global basis function ( $\phi_j(x,y)$ ) such that

$$u_n = \sum_{j=1}^n \hat{u}_j \phi_j(x,y)$$

where  $\hat{u}_j$ 's are the coefficients and  $\phi_j(x,y)$  are the global basis functions. These basis functions have compact support.

⇒  $U_n$  is further split by considering the boundary conditions such that

$$u_n(x,y) = \sum_{j=1}^{N_n} u_j \phi_j(x,y) = \sum_{j' \neq N\text{dof-Ndirichlet}} u_j \phi_j(x,y) + \sum_{j=N\text{dof-Ndirichlet}} \cancel{u_0 \phi_j(x,y)}$$

⇒ from the B.C's we know that  $u_0=0$  and that term cancels out

we then substitute this approximation into the R-G principle defined in step 1.

$$\delta \iint \frac{1}{2} |\nabla \bar{U}|^2 - f u dxdy = 0$$

$$\delta \iint \frac{1}{2} \sum_{i=1}^N \sum_{j=1}^N u_i \nabla \phi_i \cdot u_j \nabla \phi_j - f \sum u_i \phi_j dxdy = 0$$

$$\Rightarrow \sum_{j=1}^N \sum_{i=1}^N u_i u_j \frac{1}{2} \iint \nabla \phi_i(x,y) \cdot \nabla \phi_j(x,y) dxdy$$

$$- \sum_{j=1}^N u_j \iint f(x,y) \phi_j(x,y) dxdy = 0$$

if we then take

$$A_{ij} = \iint \phi_i(x,y) \cdot \phi_j(x,y) dxdy$$

$$\text{and } b_j = \iint f \phi_j(x,y) dxdy$$

we can write the R.G principle as

$$\sum_{i=1}^n \sum_{j=1}^n u_i u_j \frac{1}{2} A_{ij} - \sum_{j=1}^n u_j b_j = 0$$

variation around  $u_i$  and  $u_j$ :

$$u_j = \bar{u}_j + \epsilon s_{uj}$$
$$u_i = \bar{u}_i + \epsilon s_{ui}$$

d

$$\frac{d}{d\epsilon} \left[ \sum \sum (u_j + \epsilon s_{uj})(u_i + \epsilon s_{ui}) \frac{1}{2} A_{ij} - \sum (u_j s_{uj}) b_j \right] = 0$$

$$\frac{d}{d\epsilon} \left[ \sum \sum [u_j \bar{u}_i + u_j \epsilon s_{ui} + \epsilon s_{uj} \bar{u}_i + \epsilon^2 s_{ui} s_{uj}] \frac{A_{ij}}{2} - \sum (u_j + \epsilon s_{uj}) b_j \right] = 0$$

then we differentiate.  $\Rightarrow$

$$\sum \sum [u_j s_{ui} + u_i s_{uj} + 2\epsilon s_{ui} s_{uj}] \frac{A_{ij}}{2} - \sum s_{uj} b_j \xrightarrow{\epsilon=0} 0$$

since the variations show that the equation reached a minimum when  $\epsilon=0$

$$\sum \sum (u_j s_{ui} + u_i s_{uj}) \frac{A_{ij}}{2} - \sum s_{uj} b_j = 0$$

$$u_i \delta u_i \quad i \leftrightarrow j \quad u_j \delta u_j$$

$$\sum_{i=1}^n u_i \delta u_j A_{ij} = \sum_{j=1}^n \delta u_j b_j$$

$$\therefore \sum_{j=1}^n \delta u_j \left( \sum_{i=1}^n u_i A_{ij} - b_j \right) = 0$$

*arbitrary*

$$\therefore \sum_{i=1}^n A_{ij} u_i - b_j = 0$$

(similar to  $Ku = F$ )

Discrete weak formulation

substituting  $u_h(x,y) = \sum_{i=1}^n u_i \phi_i(x,y)$

into the weak formulation derived in  
step 1:

$$\int_0^1 \int_0^1 \nabla v \cdot \nabla \delta u - f \delta u \, dx dy = 0$$

$$\Rightarrow \int_0^1 \int_0^1 \nabla \left( \sum_{j=1}^n \delta u_j \phi_j \right) \cdot \nabla \left( \sum_{i=1}^n u_i \phi_i \right) \, dx dy$$

$$= \int_0^1 \int_0^1 \sum_{j=1}^N S u_j \phi_j f \, dx dy$$

$$\text{L.H.S.} \Rightarrow \sum_{j=1}^N \sum_{i=1}^N S u_j v_i \int_0^1 \int_0^1 \nabla \phi_j \cdot \nabla \phi_i \, dx dy$$

$$\text{R.H.S.} \Rightarrow \sum_{j=1}^N S u_j \int_0^1 \int_0^1 \phi_j f \, dx dy$$

we can then factor out  $S u_j$

$$\Rightarrow \sum_{j=1}^N S u_j \left( \sum_{i=1}^N v_i \int_0^1 \int_0^1 \nabla \phi_j \cdot \nabla \phi_i \, dx dy - \int_0^1 \int_0^1 \phi_j f \, dx dy \right) = 0$$

\* since  $S u_j$  is arbitrary, the terms in the bracket must vanish for all  $j$

$$\sum_{i=1}^N v_i \int_0^1 \int_0^1 \nabla \phi_j \cdot \nabla \phi_i \, dx dy = \int_0^1 \int_0^1 \phi_j f \, dx dy$$

we then define

$$A_{ij} = \int_0^1 \int_0^1 \nabla \phi_j \cdot \nabla \phi_i \, dx dy$$

symmetric

$$b_j = \int_0^1 \int_0^1 \phi_j f \, dx dy$$

so the equations become

$$\sum_{i=1}^n s v_i A_{ij} - b_j = 0$$

$$\Rightarrow A_{ij} v_i - b_j = 0$$

(similar to  $Ku = F$ )

D

## Step 4

The solution to the system using the provided Firedrake code is depicted in Figure 1 for a value of  $nx = 16$  ( $h = 1/16$ ).

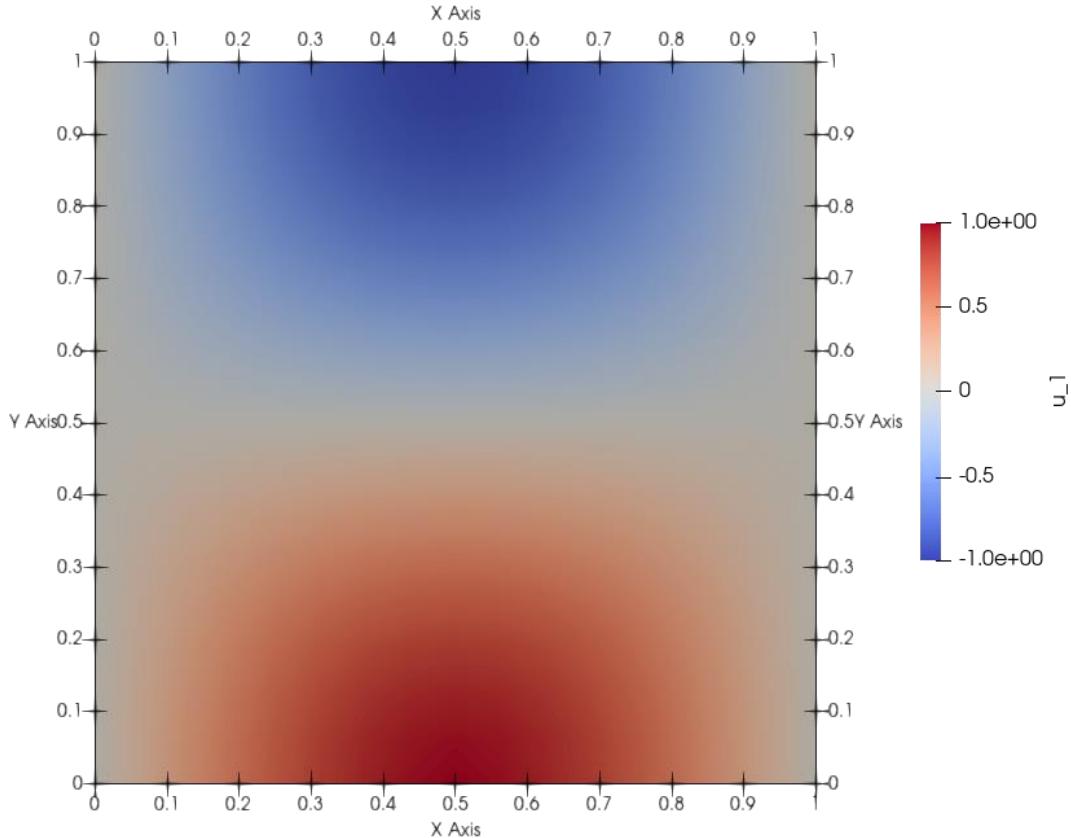


Figure 1: Numerical results for  $u_h(x, y)$  for  $h = 1/16$ .

The difference  $|u_h(x, y) - u_e(x, y)|$  is then plotted for varying  $h$  –refinements. The values of  $nx, ny$  are varied as 16, 32, 64, 128, and 256. This was all done for  $p = 1$ , so the order of accuracy for these is  $O(\Delta x^2)$ .

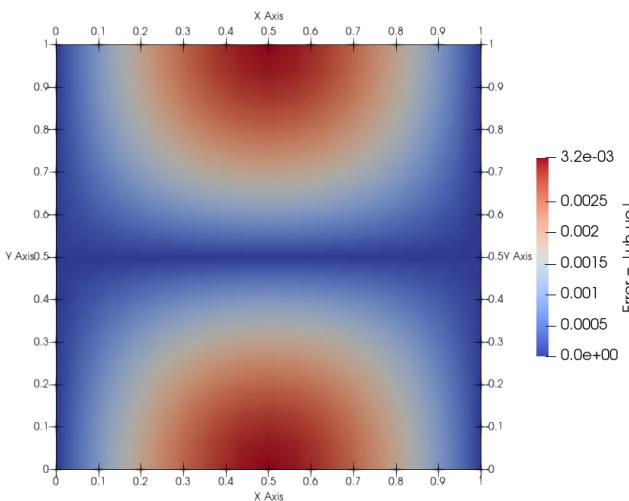


Figure 2:  $p = 1, h = 1/16$

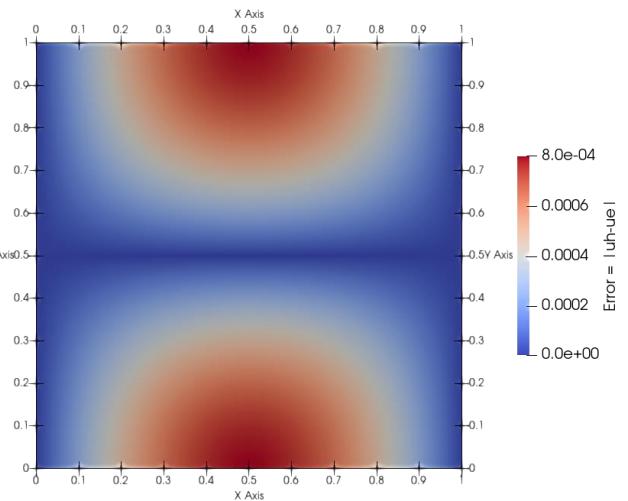


Figure 3:  $p = 1, h = 1/32$

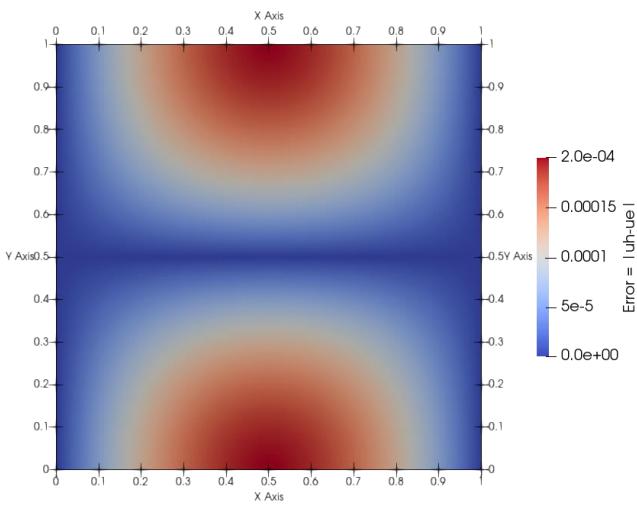


Figure 4:  $p = 1, h = 1/64$

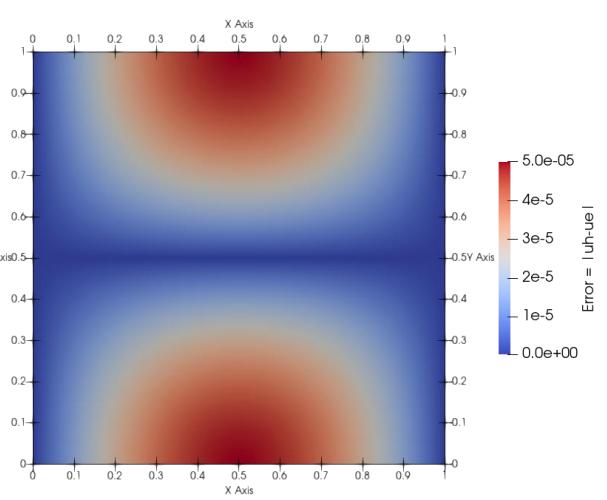


Figure 5:  $p = 1, h = 1/128$

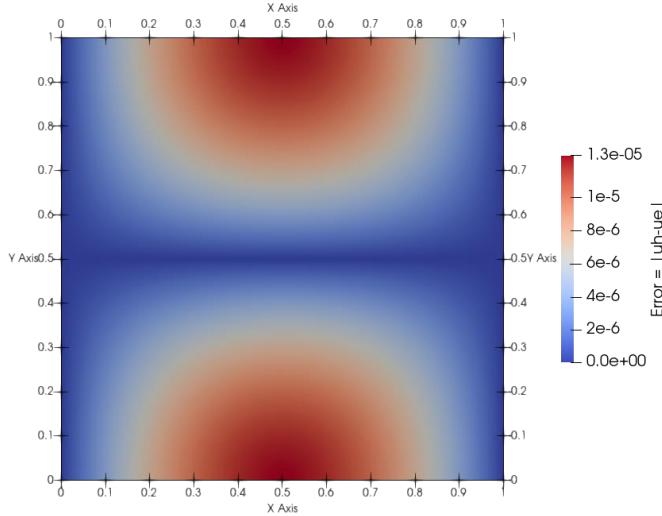


Figure 6:  $p = 1, h = 1/256$

From these figures, it is clear that as the value of  $h$  is halved, the error decreases by approximately a factor of four each time, suggesting the relationship:

$$\text{Error} \sim h^2$$

If we utilize the same colour map, we can confirm convergence occurs after  $h = 1/128$ . This is shown below, indicated by the error tending to zero as we decrease the element size.

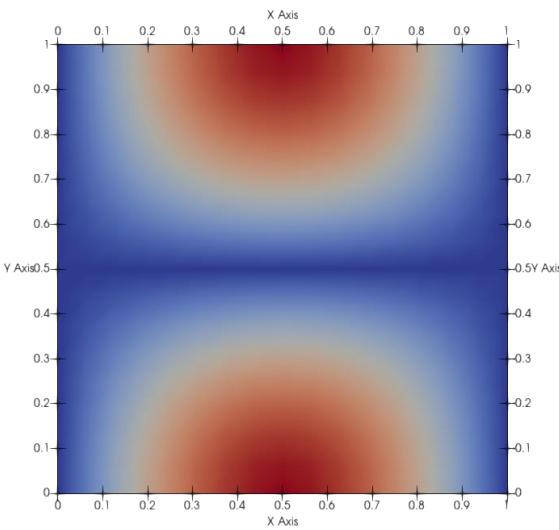


Figure 7:  $p = 1, h = 1/16$

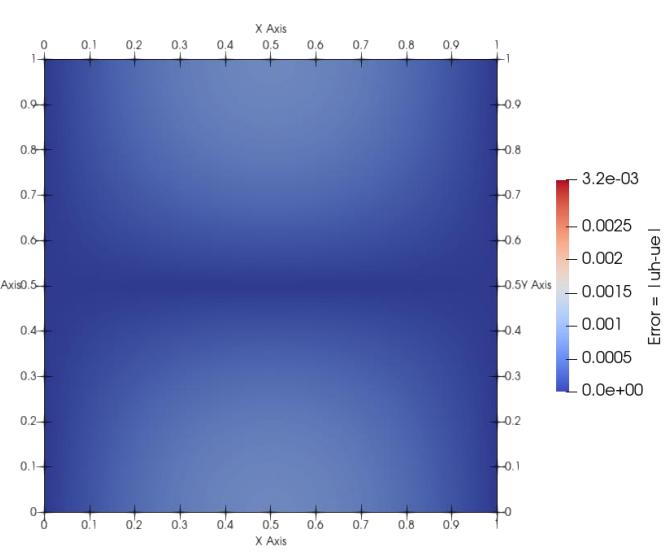


Figure 8:  $p = 1, h = 1/32$

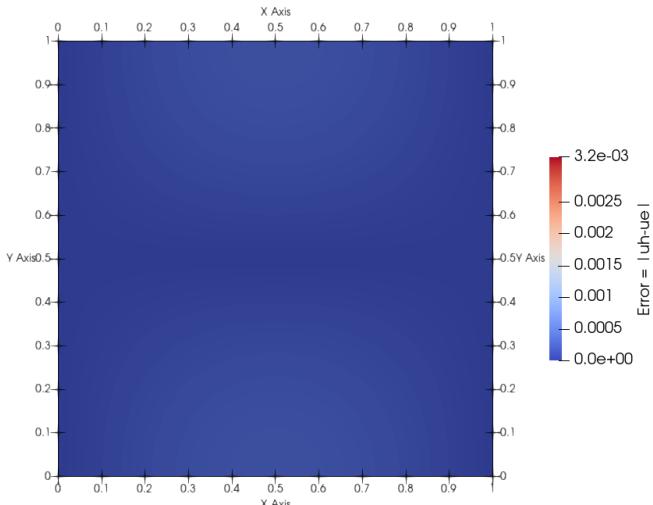


Figure 9:  $p = 1, h = 1/64$

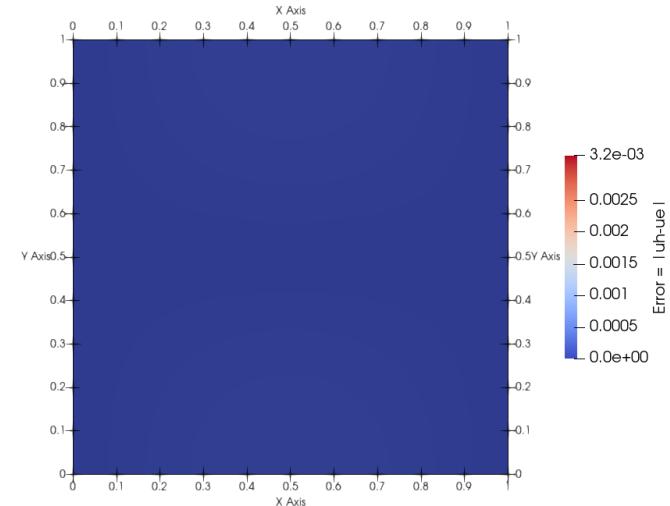


Figure 10:  $p = 1, h = 1/128$

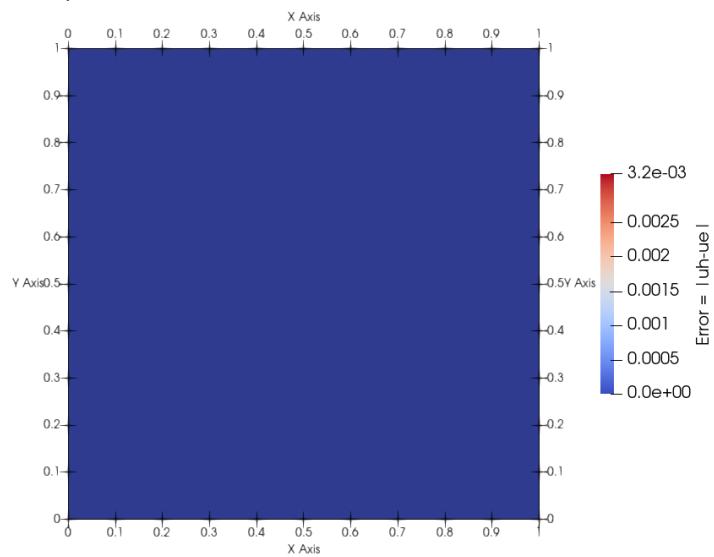


Figure 11:  $p = 1, h = 1/256$

The value of  $p$  can then be altered to change the polynomial function and the changes in error can be observed. Generally, as  $p$  increases the order of accuracy scales as follows:

$$\text{Order of accuracy} \sim O\Delta x^{p+1}$$

The order of accuracy for different  $p$ -refinements is summarized below in Table 1.

Table 1:  $p$  vs. order of accuracy.

$p$	Order of accuracy
<b>1</b>	$O(\Delta x^2)$
<b>2</b>	$O(\Delta x^3)$
<b>3</b>	$O(\Delta x^4)$
<b>4</b>	$O(\Delta x^5)$

To investigate the effect of changing the degree polynomial ( $p$ ), the value of  $p$  is altered whilst the value of  $h$  is kept constant in a p-refinement study. The value of  $h$  chosen is  $h = 1/128$  ( $nx = 128$ ) and  $p$  is varied from 1 till 4. This is shown in the figures below.

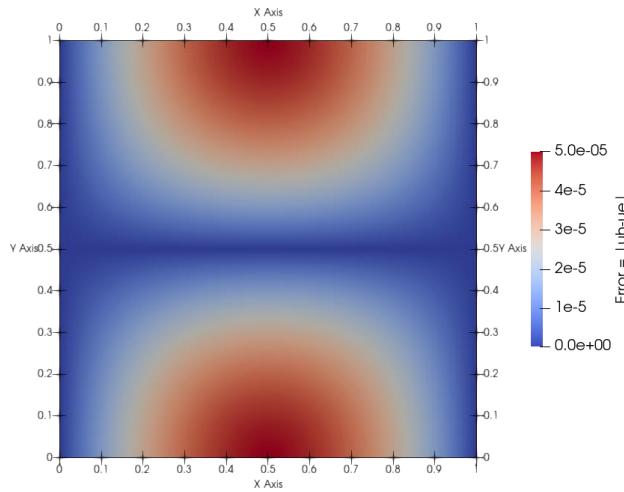


Figure 12:  $p = 1 (O\Delta(x^2))$ ,  $h = 1/128$

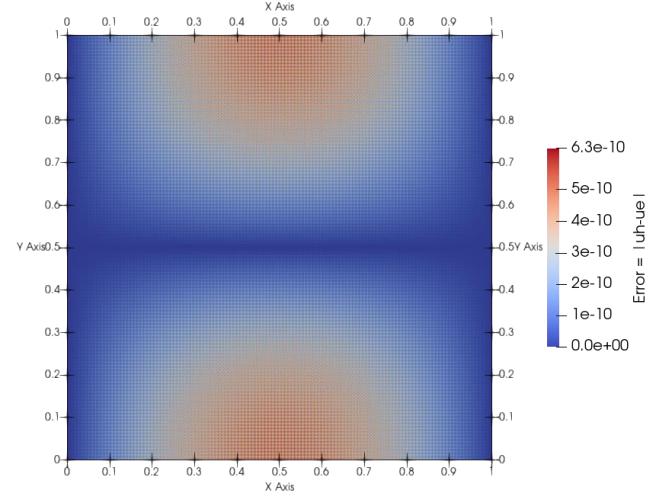


Figure 13:  $p = 2 (O\Delta(x^3))$ ,  $h = 1/128$

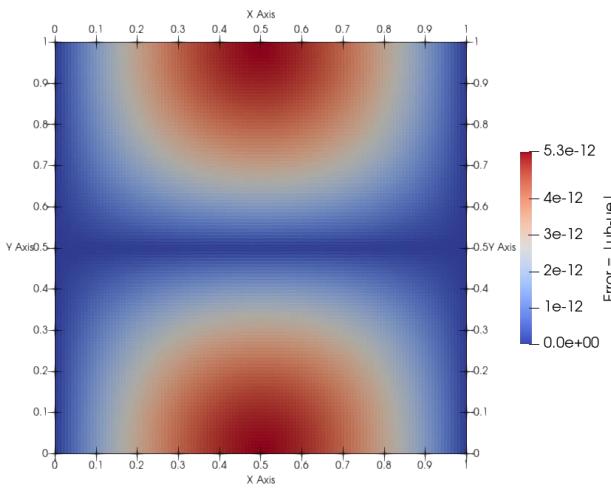


Figure 13:  $p = 3 (O\Delta(x^4))$ ,  $h = 1/128$

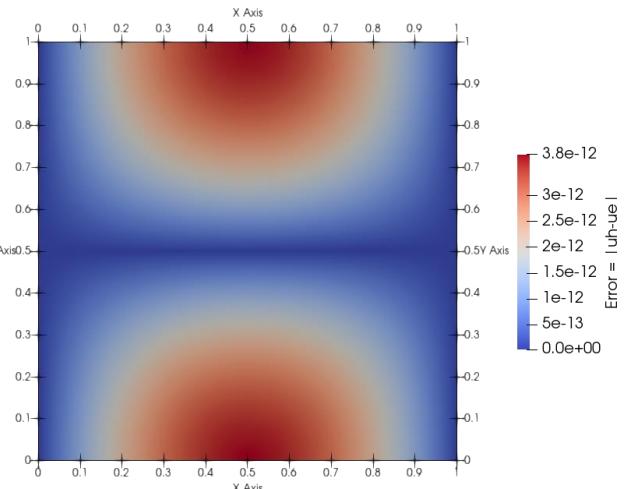


Figure 14:  $p = 4, (O\Delta(x^4)) h = 1/128$

From these graphs, we can see that as  $p$  increases, the error reduces by a much larger magnitude than just refining  $h$ . Generally, the error for these systems scales as:

$$\text{Error} \sim O(h^{p+1})$$

For some  $\{h, p\}$  refinements, the errors are approximately the same. Three observed cases with similar errors are shown below.

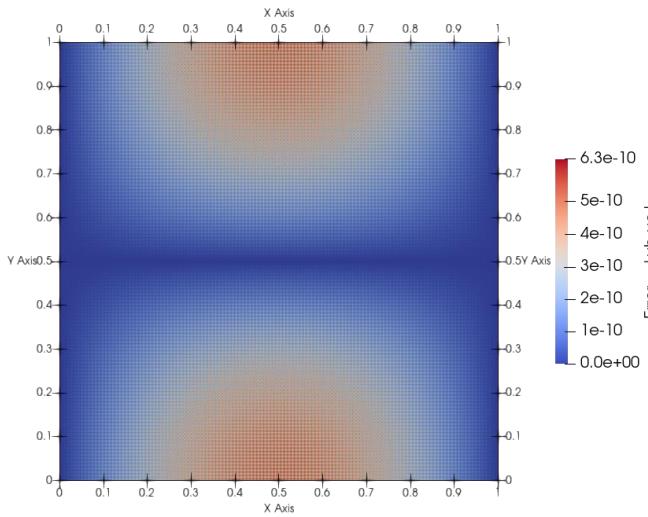


Figure 15:  $p = 2(O\Delta(x^3)), h = 1/128$

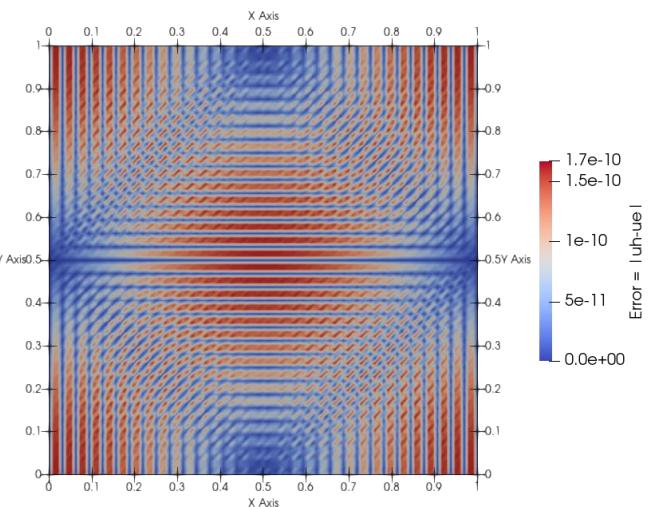


Figure 16:  $p = 3(O\Delta(x^4)), h = 1/32$

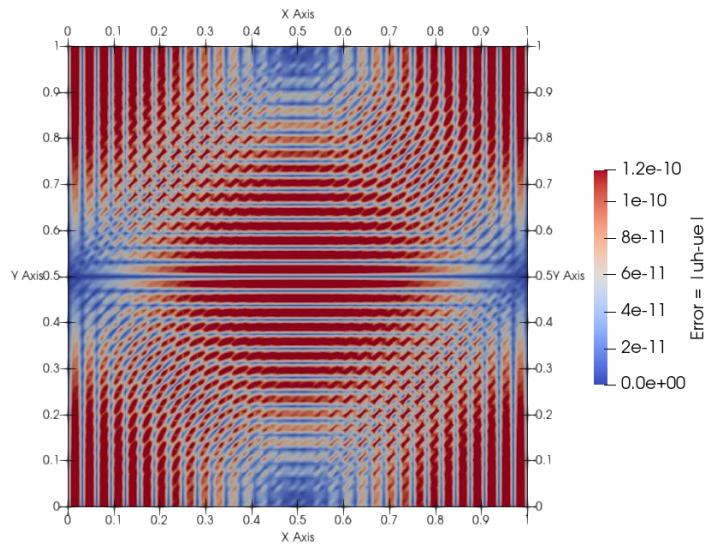


Figure 17:  $p = 4(O\Delta(x^5)), h = 1/16$

The reason these errors are similar is because their order of accuracy is also quite similar:

$$\left\{ p = 2, h = \frac{1}{128} \right\} \rightarrow \text{Error} \sim O\left(\frac{1}{128}\right)^3 \sim 4.99 \times 10^{-7}$$

$$\left\{ p = 3, h = \frac{1}{32} \right\} \rightarrow \text{Error} \sim O\left(\frac{1}{32}\right)^4 \sim 9.53 \times 10^{-7}$$

$$\left\{ p = 4, h = \frac{1}{16} \right\} \rightarrow \text{Error} \sim O\left(\frac{1}{16}\right)^5 \sim 9.54 \times 10^{-7}$$