

Problem 1 (Shift invariance of the Laplace operator)

The Laplace equation is shift invariant, i.e., invariant under translations

$$x' = x + a \quad , \quad y' = y + b, \quad a, b \in \mathbb{R}.$$

The shift invariance can then be written as

$$u_{xx} + u_{yy} = u_{x'x'} + u_{y'y'}$$

To see this explicitly, we consider x, y as mappings depending on x' and y' , respectively, and we compute

$$y(y') = y' - b, \quad x(x') = x' - a \Rightarrow \frac{\partial y(y')}{\partial y'} = 1, \frac{\partial x(x')}{\partial x'} = 1.$$

It is not wrong to consider $x \equiv x(x', y'), y \equiv y(x', y')$, so that

$$\frac{\partial y}{\partial y'} = 1 \quad , \quad \frac{\partial x}{\partial x'} = 1.$$

It follows

$$\begin{aligned} u_{x'x'} = u(x, y)_{x'x'} &= \frac{\partial}{\partial x'} \left(\frac{\partial}{\partial x'} u(x, y) \right) \\ &= \frac{\partial}{\partial x'} \left(\underbrace{\frac{\partial x}{\partial x'}}_{=1} \cdot \frac{\partial}{\partial x} u(x, y) \right) \\ &= \frac{\partial}{\partial x'} u_x(x, y) \\ &= \frac{\partial}{\partial x} \underbrace{\frac{\partial x}{\partial x'}}_{=1} u_x(x, y) = u_{xx}(x, y). \end{aligned}$$

$u_{y'y'} = u_{yy}$ follows analogously.

Problem 2 (What's the matrix, what's the matrix?)

1. For the ordering

$$(u_1 \ u_2 \ u_3 \ u_4 \ u_5 \ u_6 \ u_7 \ u_8 \ u_9 \ u_{10} \ u_{11} \ u_{12} \ u_{13} \ u_{14} \ u_{15} \ u_{16})^\top \quad (1)$$

and the underlying process

$$-\left[\frac{u_{i+1,j} - 2u_{ij} + u_{i-1,j}}{\Delta x^2} + \frac{u_{i,j+1} - 2u_{ij} + u_{i,j-1}}{\Delta y^2} \right] = f_{ij} \quad (2)$$

we get the following matrix system

$\begin{array}{cccc} 4 & -1 & & \\ -1 & 4 & -1 & \\ & -1 & 4 & -1 \\ & & -1 & 4 \end{array}$	$\begin{array}{ccc} -1 & & \\ & -1 & \\ & & -1 \end{array}$		
$\begin{array}{ccc} -1 & & \\ & -1 & \\ & & -1 \end{array}$	$\begin{array}{ccc} 4 & -1 & \\ -1 & 4 & -1 \\ & -1 & 4 & -1 \\ & & -1 & 4 \end{array}$	$\begin{array}{ccc} -1 & & \\ & -1 & \\ & & -1 \end{array}$	
	$\begin{array}{ccc} -1 & & \\ & -1 & \\ & & -1 \end{array}$	$\begin{array}{ccc} 4 & -1 & \\ -1 & 4 & -1 \\ & -1 & 4 & -1 \\ & & -1 & 4 \end{array}$	$\begin{array}{ccc} -1 & & \\ & -1 & \\ & & -1 \end{array}$
		$\begin{array}{ccc} -1 & & \\ & -1 & \\ & & -1 \end{array}$	$\begin{array}{ccc} 4 & -1 & \\ -1 & 4 & -1 \\ & -1 & 4 & -1 \\ & & -1 & 4 \end{array}$

(3)

2. For the ordering

$$(u_1 \ u_3 \ u_6 \ u_{10} \ u_2 \ u_5 \ u_9 \ u_{13} \ u_4 \ u_8 \ u_{12} \ u_{15} \ u_7 \ u_{11} \ u_{14} \ u_{16})^\top \quad (4)$$

we get the following matrix system

4	-1 -1					
-1 -1	4 4	-1 -1 -1 -1				
	-1 -1 -1 -1	4 4 4	-1 -1 -1 -1 -1 -1			
		-1 -1 -1 -1 -1 -1	4 4 4 4	-1 -1 -1 -1 -1 -1		
			-1 -1 -1 -1 -1 -1	4 4 4	-1 -1 -1 -1	
				-1 -1 -1 -1	4 4	-1 -1
					-1 -1 -1 -1	-1 -1
					-1 -1 -1 -1	4 4
					-1 -1	-1 4 4

(5)

Problem 3 (Crossing derivatives)

1. We have been given the following cross derivative discretisation by use of central difference methods

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

with $h = \Delta x = \Delta y$. For this 4 points we can compute the 2D-

Taylor expansion

$$\begin{aligned}
u(x_0, y_0) &= f(x, y) + \binom{1}{0} \frac{\partial}{\partial x} f(x, y)(x - x_0) + \binom{1}{1} \frac{\partial}{\partial y} f(x, y)(y - y_0) \\
&+ \frac{1}{2} \left(\binom{2}{0} \frac{\partial^2}{\partial x^2} f(x, y)(x - x_0)^2 + \binom{2}{1} \frac{\partial}{\partial x} \frac{\partial}{\partial y} f(x, y)(x - x_0)(y - y_0) \right. \\
&\left. + \binom{2}{2} f(x, y) \frac{\partial^2}{\partial y^2} (y - y_0)^2 \right)
\end{aligned}$$

This gives for our simple points the following approximation:

$$\begin{aligned}
u(x \pm h, y + h) &= u \pm hu_x + u_y + \frac{1}{2}h^2(u_{xx} \pm 2u_{xy} + u_{yy}) \\
&+ \frac{1}{6}h^3(\pm u_{xxx} + 3u_{xxy} \pm u_{xyy} + 3u_{yyy}) \\
&+ \frac{1}{24}h^4(u_{xxxx} \pm 4u_{xxxy} + 6u_{xxyy} \pm 4u_{xyyy} + u_{yyyy}) + \mathcal{O}(h^5) \\
u(x + h, y - h) &= u + hu_x - u_y + \frac{1}{2}h^2(u_{xx} - 2u_{xy} + u_{yy}) \\
&+ \frac{1}{6}h^3(u_{xxx} - 3u_{xxy} + u_{xyy} - 3u_{yyy}) \\
&+ \frac{1}{24}h^4(u_{xxxx} - 4u_{xxxy} + 6u_{xxyy} - 4u_{xyyy} + u_{yyyy}) + \mathcal{O}(h^5) \\
u(x - h, y - h) &= u - hu_x - u_y + \frac{1}{2}h^2(u_{xx} + 2u_{xy} + u_{yy}) \\
&- \frac{1}{6}h^3(u_{xxx} + 3u_{xxy} + u_{xyy} + 3u_{yyy}) \\
&+ \frac{1}{24}h^4(u_{xxxx} + 4u_{xxxy} + 6u_{xxyy} + 4u_{xyyy} + u_{yyyy}) + \mathcal{O}(h^5)
\end{aligned}$$

Now we can input this approximation into our four pixel scheme:

$$\begin{aligned}
u_{xy} &\approx \frac{1}{4h^2}(u_{i+1,j+1} - u_{i-1,j+1} - u_{i+1,j-1} + u_{i-1,j-1}) \\
&= \frac{1}{4h^2}(u + hu_x + u_y + \frac{1}{2}h^2(u_{xx} + 2u_{xy} + u_{yy})) \\
&+ \frac{1}{6}h^3(u_{xxx} + 3u_{xxy} + u_{xyy} + 3u_{yyy}) \\
&+ \frac{1}{24}h^4(u_{xxxx} + 4u_{xxxy} + 6u_{xxyy} + 4u_{xyyy} + u_{yyyy}) + \mathcal{O}(h^5) \\
&- (u - hu_x + u_y + \frac{1}{2}h^2(u_{xx} - 2u_{xy} + u_{yy})) \\
&+ \frac{1}{6}h^3(-u_{xxx} + 3u_{xxy} - u_{xyy} + 3u_{yyy}) \\
&+ \frac{1}{24}h^4(u_{xxxx} - 4u_{xxxy} + 6u_{xxyy} - 4u_{xyyy} + u_{yyyy}) + \mathcal{O}(h^5) \\
&- (u + hu_x - u_y + \frac{1}{2}h^2(u_{xx} - 2u_{xy} + u_{yy})) \\
&+ \frac{1}{6}h^3(u_{xxx} - 3u_{xxy} + u_{xyy} - 3u_{yyy}) \\
&+ \frac{1}{24}h^4(u_{xxxx} - 4u_{xxxy} + 6u_{xxyy} - 4u_{xyyy} + u_{yyyy}) + \mathcal{O}(h^5) \\
&+ u - hu_x - u_y + \frac{1}{2}h^2(u_{xx} + 2u_{xy} + u_{yy}) \\
&- \frac{1}{6}h^3(u_{xxx} + 3u_{xxy} + u_{xyy} + 3u_{yyy}) \\
&+ \frac{1}{24}h^4(u_{xxxx} + 4u_{xxxy} + 6u_{xxyy} + 4u_{xyyy} + u_{yyyy}) + \mathcal{O}(h^5)
\end{aligned}$$

If we combine these terms together, we will get:

$$\begin{aligned}
& \frac{1}{4h^2} \left(u \underbrace{(1-1-1+1)}_{=0} + hu_x \underbrace{(1+1-1-1)}_{=0} + hu_y \underbrace{(1-1+1-1)}_{=0} \right) \\
& + \frac{1}{2} h^2 \underbrace{(u_{xx} + 2u_{xy} + u_{yy} - u_{xx} + 2u_{xy} - u_{yy} - u_{xx} + 2u_{xy} - u_{yy} + u_{xx} + 2u_{xy} + u_{yy})}_{8u_{xy}} \\
& + \frac{1}{6} h^3 (u_{xxx} \underbrace{(1+1-1-1)}_{=0} + 3u_{xxy} \underbrace{(1-1+1-1)}_{=0}) \\
& + 3u_{xyy} \underbrace{(1+1-1-1)}_{=0} + u_{yyy} \underbrace{(1-1+1-1)}_{=0} \\
& + \frac{1}{24} h^4 (u_{xxxx} \underbrace{(1-1-1+1)}_{=0} + 4u_{xxxy} \underbrace{(1+1+1+1)}_{=4} \\
& + 6u_{xyyy} \underbrace{(1-1-1+1)}_{=0} + 4u_{xyyy} \underbrace{(1+1+1+1)}_{=4} + u_{yyyy} \underbrace{(1-1-1+1)}_{=0}) + \mathcal{O}(h^5)
\end{aligned}$$

This sums up to

$$\begin{aligned}
& \frac{1}{4h^2} (4h^2 u_{xy} + \frac{16}{24} (u_{xxxy} + u_{xyyy}) + \mathcal{O}(h^5)) \\
& = u_{xy} + \frac{1}{6} h^2 (u_{xxxy} + u_{xyyy}) + \mathcal{O}(h^3) \\
& = u_{xy} + \frac{1}{6} h^2 \Delta u_{xy} + \mathcal{O}(h^3) \\
& = \left(1 + \frac{1}{6} h^2 \right) u_{xy} + \mathcal{O}(h^3) \\
& \Rightarrow \mathcal{O}(h^2)
\end{aligned}$$

Overall we get a $\mathcal{O}(h^2)$ error term for the cross derivative approximation.

2. This discretisation is isotropic, as the error term incorporates an additional isotropic Laplace operator onto u_{xy} , which we wanted to approximate in the first place.

Problem 4 (Cooking norms)

1. We want to prove the following statement:

$$\frac{1}{\sqrt{n}} \|x\|_2 \stackrel{1}{\leq} \|x\|_\infty \stackrel{2}{\leq} \|x\|_2 \stackrel{3}{\leq} \sqrt{n} \|x\|_\infty$$

We will do this step by step. So at first we prove the first inequality

$$\begin{aligned}
 \frac{1}{\sqrt{n}}\|x\|_2 &= \frac{1}{\sqrt{n}}\sqrt{\sum_{i=1}^n|x_i|^2} \\
 &= \sqrt{\frac{1}{n}\sum_{i=1}^n|x_i|^2} \\
 &\leq \sqrt{\frac{1}{n}\sum_{i=1}^n\max_i\|x_i\|^2} \\
 &= \sqrt{\max_i\|x_i\|^2} \\
 &= \max_i|x_i|.
 \end{aligned}$$

Now we will have a closer look at the second inequality $\|x\|_\infty \leq \|x\|_2$. For this, however we consider now the squared norms, as this does not violate the monotonicity of the norming function:

$$\|x\|_2^2 = \max_{i=1,\dots,n}|x_i|^2 \leq \sum_{i=1}^n|x_i|^2 = \|x\|_\infty^2.$$

Now we only need to prove the last inequality, so we compute

$$\begin{aligned}
 \|x\|_2 &= \left(\sum_{i=1}^n|x_i|^2\right)^{\frac{1}{2}} \\
 &\leq \sqrt{n\max_i|x_i|^2} \\
 &= \sqrt{n}\sqrt{\max_i|x_i|^2} \\
 &= \sqrt{n}\max_i|x_i| \\
 &= \sqrt{n}\|x\|_\infty
 \end{aligned}$$

which concludes the proof.

2. We want to prove the following statement:

$$\frac{1}{\sqrt{n}}\|x\|_1 \leq \|x\|_2 \leq \|x\|_1 \leq \sqrt{n}\|x\|_2$$

At first, we will prove the third inequality by use of the Cauchy-Schwarz inequality

$$\left(\sum_i x_i y_i \right)^2 \leq \left(\sum_i x_i^2 \right) \cdot \left(\sum_i y_i^2 \right),$$

so we can compute

$$\|x\|_1 = \sum_{i=1}^n 1 \cdot |x_i| \leq \left(\sum_{i=1}^n 1^2 \right)^{\frac{1}{2}} \left(\sum_{i=1}^n |x_i|^2 \right)^{\frac{1}{2}} = \sqrt{n} \|x\|_2.$$

From this, the first inequality is easily derivable. The biggest problem is now the second inequality. For this we will now consider the vector

$$y = \frac{x}{\|x\|_1} = \frac{x}{\sum_{k=1}^n |x_k|}$$

We will now show that $\|y\|_2 \leq 1$ which will help us later.

$$\begin{aligned} \|y\|_2^2 &= \sum_{i=1}^n \frac{|x_i|^2}{\left(\sum_{k=1}^n |x_k| \right)^2} \\ &= \frac{1}{\sum_{k=1}^n |x_k|} \cdot \sum_{i=1}^n \frac{|x_i|^2}{\sum_{k=1}^n |x_k|} \\ \sum_{k=1}^n |x_k| \geq |x_i| &\leq \frac{1}{\|x\|_1} \cdot \underbrace{\sum_{i=1}^n \frac{|x_i|^2}{|x_i|}}_{\|x\|_1} = 1. \end{aligned}$$

We use this result now in:

$$\|x\|_2 = \|x\|_1 \|y\|_2 \leq \|x\|_1 \cdot 1 = \|x\|_1,$$

what we wanted to show.

Problem 5 (Proving Banach) Let an arbitrary $x_0 \in D$ be given. As $F : D \rightarrow D$, the sequence $(x_k)_{k \in \mathbb{N}_0}$ is uniquely determined by $x_{k+1} = F(x_k)$ and for $k \in \mathbb{N}$ it holds:

$$|x_{k+1} - x_k| = |F(x_k) - F(x_{k-1})| \leq L|x_k - x_{k-1}| \leq L^2|x_{k-1} - x_{k-2}| \leq \dots \quad (*)$$

Also, by iteration of the first approximation it follows for $k \geq n$:

$$|x_{k+1} - x_k| \leq L^{k+1-n}|x_n - x_{n-1}| \quad (**).$$

and from that also for $m \geq n$:

$$\begin{aligned} |x_m - x_n| &= |(x_m - x_{m-1}) + (x_{m-1} - x_{m-2}) + \dots + (x_{n+1} - x_n)| \\ &\leq \sum_{k=n}^{m-1} |x_{k+1} - x_k| \quad \text{Triangle-Inequality} \\ &\leq \left(\sum_{k=n}^{m-1} L^{k+1-m} \right) |x_n - x_{n-1}| \\ &\leq \left(\sum_{j=1}^{\infty} L^j \right) |x_n - x_{n-1}| \\ &= L \cdot \left(\sum_{j=0}^{\infty} L^j \right) |x_n - x_{n-1}| \\ &= \frac{L}{1-L} |x_n - x_{n-1}| \\ &\leq \frac{L^n}{1-L} |x_1 - x_0| \quad \text{see (*) or (**).} \end{aligned}$$

Therefore it holds for $m \geq n$:

$$|x_m - x_n| \leq \frac{L}{1-L} |x_n - x_{n-1}| \leq \frac{L^n}{1-L} |x_1 - x_0| \quad (***)$$

Due to $L < 1$ it holds that $L^n \rightarrow 0$ ($n \rightarrow \infty$) and therefore $(x_n)_{n \in \mathbb{N}_0}$ is a Cauchy sequence that is convergent with the limit x^* . As D is compact, with $***$, $x^* \in D$. F is continuous and therefore the limit x^* is a fixed point of F . x^* is the only fixed point in D . If $x^{**} \neq x^*$ were another fixed point, it would follow:

$$|x^{**} - x^*| = |F(x^{**}) - F(x^*)| \leq L|x^{**} - x^*| < |x^{**} - x^*|$$

which is a contradiction. The proposed error approximation follows directly from $***$, if $m \rightarrow \infty$.
