

Problem 1. Derive the variational form (or weak form) of the BVP

$$-\frac{d}{dx} \left(k(x) \frac{du}{dx} \right) + p(x)u = f(x), \quad 0 < x < 1, \quad (1a)$$

$$u(0) = 0, \quad (1b)$$

$$u(1) = 0, \quad (1c)$$

where p and k satisfy $p(x) > 0$ and $k(x) > 0$ for $x \in [0, 1]$. What is the bilinear form for this BVP? Write a Matlab code using piecewise linear finite elements to solve the above problem. Construct one example. Compare the numerical solution with the exact solution for $n = 20$.

Proof. To simplify the notation we drop the explicit dependence on x and write primes for the derivatives; thus our task is to find the weak form of

$$-(ku')' + pu = f. \quad (2)$$

We will make use of test functions from the space

$$H_0^1 = \{v \in L^2([0, 1]) \mid v' \in L^2([0, 1]), v(0) = 0, v(1) = 0\}.$$

Multiplying Eq. (2) by $v \in H_0^1$ and integrating, we get

$$-\int_0^1 (ku')' v \, dx + \int_0^1 pu v \, dx = \int_0^1 f v \, dx. \quad (3)$$

We notice that an application of the product rule yields

$$ku'v \Big|_0^1 = \int_0^1 (ku'v)' \, dx = \int_0^1 (ku')' v \, dx + \int_0^1 ku'v' \, dx.$$

Substituting back into Eq. (3), we get

$$\int_0^1 ku'v' \, dx - ku'v \Big|_0^1 + \int_0^1 pu v \, dx = \int_0^1 f v \, dx.$$

We then notice that the term

$$ku'v \Big|_0^1 = k(1)u'(1)v(1) - k(0)u'(0)v(0)$$

vanishes because v also vanishes at the endpoints. Thus we conclude that the weak form of the BVP (2) is given by

$$\int_0^1 ku'v' \, dx + \int_0^1 pu v \, dx = \int_0^1 f v \, dx \quad (4)$$

Whence the bilinear form $a(\cdot, \cdot)$ associated with this system is

$$a(u, v) = \int_{[0,1]} (ku'v' + pu v) \, dx.$$

Our job now is to find a suitable solution $u \in H_0^1$ that satisfies (4) for all $v \in H_0^1$. In fact, the space H_0^1 is too large to be of practical use; in the Galerkin approach a suitable finite subspace is used instead. For our purposes, the subspace containing all continuous, piecewise linear functions will suffice. Let $I = [0, 1]$ and let the vector space of linear functions on I be denoted by $P_1(I)$:

$$P_1(I) = \{v \mid v(x) = \alpha_0 + \alpha_1 x; x \in I; \alpha_0, \alpha_1 \in \mathbb{R}\}.$$

We shall make use of $n + 1$ nodes $\{x_i\}_{i=0}^n$ and partition I in the usual way

$$0 = x_0 < x_1 < \dots < x_{n-1} < x_n = 1,$$

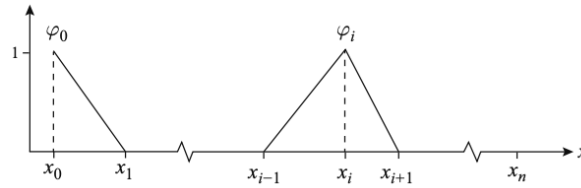
so there are n subintervals $I_i = [x_{i-1}, x_i]$, with $i = 1, \dots, n$, each of length $h_i = x_i - x_{i-1}$.¹ Then the subspace we shall work with is

$$V_n = \{v \mid v \in C^0(I); v|_{I_i} \in P(I_i); v(0) = v(1) = 0\},$$

where $C_0(I)$ denotes as usual the space of all continuous functions on I . Hence, as we alluded to earlier, our work space V_n is the space containing all continuous, piecewise linear functions on the interval I . Moreover, since we need to fulfill the boundary criteria from the original strong-form BVP, we are also imposing the vanishing property at the endpoints in V_n .

Our next order of business is to introduce the basis of hat functions $\{\varphi_j\}_{j=0}^n$ for V_n , which satisfies

$$\varphi_j(x_i) = \begin{cases} 1 & \text{if } i = j; \\ 0 & \text{otherwise.} \end{cases}$$



Referring to the figure, we can easily deduce an explicit expression for the hats:

$$\varphi_i(x) = \begin{cases} \frac{x-x_{i-1}}{h_i} & \text{if } x \in I_i; \\ \frac{x_{i+1}-x}{h_{i+1}} & \text{if } x \in I_{i+1}; \\ 0 & \text{otherwise.} \end{cases} \quad (5)$$

Next we approximate the solution u by a continuous piecewise linear function $^{(n)}u$, so that $^{(n)}u, v \in V_n$. Then, since $\{\varphi_j\}_{j=0}^n$ is a basis for V_n , we can do the following two things: i) replace the v 's with φ 's in Eq. (4), since it suffices to see what happens at the basis only; ii) we consider the ansatz

$$^{(n)}u = \sum_{i=0}^n U_i \varphi_i. \quad (6)$$

We may then rewrite Eq. (4) as

$$\int_I k^{(n)}u' \varphi_i' dx + \int_I p^{(n)}u \varphi_i dx = \int_I f \varphi_i dx \quad i = 0, \dots, n. \quad (7)$$

However, note that, even though this expression is valid for all $i \in \{0, \dots, n\}$, due to the vanishing boundary conditions we only get nonvanishing terms for $i \in \{1, \dots, n-1\}$; whence from now on we shall focus only in i in this range. Plugging in the ansatz (6) on the LHS, we get

$$\begin{aligned} \int_I k \left(\sum_{j=1}^{n-1} U_j \varphi_j' \right) \varphi_i' dx + \int_I p \left(\sum_{j=1}^{n-1} U_j \varphi_j \right) \varphi_i dx &= \sum_{j=1}^{n-1} U_j \int_I k \varphi_j' \varphi_i' dx + \sum_{j=1}^{n-1} U_j \int_I p \varphi_j \varphi_i dx \\ &= \sum_{j=1}^{n-1} U_j \int_I \left(k \varphi_j' \varphi_i' + p \varphi_j \varphi_i \right) dx \quad i = 1, \dots, n-1. \end{aligned}$$

Hence we have a system of the form

$$\boxed{(K + M)U = F} \quad (8)$$

where

$$K_{ij} = \int_I k \varphi_j' \varphi_i' dx \quad (\text{Stiffness Matrix})$$

$$M_{ij} = \int_I p \varphi_j \varphi_i dx \quad (\text{Mass Matrix})$$

$$F_i = \int_I f \varphi_i dx. \quad (\text{Load Vector})$$

¹For future reusability of the code and to ensure flexibility, we will not assume that the partition is uniform; i.e., there will be no h such that $h = h_i \forall i$.

Let us now write out the nonvanishing components of these arrays. Note that, since for $|i - j| > 1$ the hats (and their derivatives) lack common support, both \mathbf{K} and \mathbf{M} will be tridiagonal. In all cases we shall use Simpson's quadrature; we recall that Simpson's method applied to an interval $I = [x_{i-1}, x_i]$ takes the form

$$\int_I f \approx \frac{h_i}{6} [f(x_{i-1}) + 4f(m_i) + f(x_i)],$$

where x_m is the midpoint $m_i = \frac{1}{2}(x_i + x_{i-1})$ and $h_i = x_i - x_{i-1}$. We start with the mass matrix \mathbf{M} ; its diagonal entries are given by

$$\begin{aligned} M_{ii} &= \int_I p \varphi_i^2 dx \\ &= \int_{x_{i-1}}^{x_i} p \varphi_i^2 dx + \int_{x_i}^{x_{i+1}} p \varphi_i^2 dx \\ &= \frac{h_i}{6} \left[p(x_{i-1}) \cdot 0 + 4 \cdot p(m_i) \cdot \left(\frac{1}{2}\right)^2 + p(x_i) \cdot 1 \right] + \frac{h_{i+1}}{6} \left[p(x_i) \cdot 1 + 4 \cdot p(m_{i+1}) \cdot \left(\frac{1}{2}\right)^2 + p(x_{i+1}) \cdot 0 \right] \\ &= \frac{h_i}{6} [p(m_i) + p(x_i)] + \frac{h_{i+1}}{6} [p(x_i) + p(m_{i+1})]. \end{aligned}$$

Similarly, for the subdiagonal entries,

$$\begin{aligned} M_{i+1,i} &= \int_I p \varphi_i \varphi_{i+1} dx \\ &= \int_{x_i}^{x_{i+1}} p \varphi_i \varphi_{i+1} dx \\ &= \frac{h_{i+1}}{6} \left[p(x_i) \cdot 0 + 4 \cdot p(m_{i+1}) \cdot \left(\frac{1}{2}\right)^2 + p(x_{i+1}) \cdot 0 \right] \\ &= \frac{h_{i+1} \cdot p(m_{i+1})}{6}. \end{aligned}$$

By symmetry, the superdiagonal entries are identical to the subdiagonal ones; i.e., $M_{i,i+1} = M_{i+1,i}$. The following Matlab routine will assemble \mathbf{M} :

```

1 function M = MassMatD0(x,p)
2     %input mesh vector x and function p to MassMatD0
3     %output Mass Matrix M
4
5     n = length(x)-1;           %number of subintervals
6     M = zeros(n-1, n-1);       %allocate mass matrix
7     %No need for half-hats due to vanishing BCs; otherwise M would have dim (n+1)x(n+1)
8
9     for i = 1:n-1
10         h_minus = x(i+1) - x(i);           %h_i           (index offset)
11         xmid = (x(i+1) + x(i))/2;           %m_i           (index offset)
12         h_plus = x(i+2) - x(i+1);           %h_{i+1}       (index offset)
13         xmid_plus = (x(i+2) + x(i+1))/2;     %m_{i+1}       (index offset)
14
15         M(i,i) = (h_minus/6) * ( p(xmid) + p(x(i+1)) )
16                 + (h_plus/6) * ( p(xmid_plus) + p(x(i+1)) ) ;
17
18         if i ~= n-1
19             M(i+1,i) = ( h_plus * p(xmid_plus) )/6;
20             M(i,i+1) = M(i+1,i) ;
21         end
22     end
23
24 end

```

Similarly, we now build the stiffness matrix. Before we start, however, we need to know the derivatives of the hat functions. A quick glance at Eq. (5) reveals that

$$\varphi'_i(x) = \begin{cases} \frac{1}{h_i} & \text{if } x \in I_i; \\ -\frac{1}{h_{i+1}} & \text{if } x \in I_{i+1}; \\ 0 & \text{otherwise.} \end{cases} \quad (9)$$

The diagonal entries of K are then given by

$$\begin{aligned}
K_{ii} &= \int_I k (\varphi'_i)^2 dx \\
&= \int_{x_{i-1}}^{x_i} k (\varphi'_i)^2 dx + \int_{x_i}^{x_{i+1}} k (\varphi'_i)^2 dx \\
&= \frac{h_i}{6} \left[k(x_{i-1}) \cdot \left(\frac{1}{h_i}\right)^2 + 4 \cdot k(m_i) \cdot \left(\frac{1}{h_i}\right)^2 + k(x_i) \cdot \left(\frac{1}{h_i}\right)^2 \right] \\
&\quad + \frac{h_{i+1}}{6} \left[k(x_i) \cdot \left(-\frac{1}{h_{i+1}}\right)^2 + 4 \cdot k(m_{i+1}) \cdot \left(-\frac{1}{h_{i+1}}\right)^2 + k(x_{i+1}) \cdot \left(-\frac{1}{h_{i+1}}\right)^2 \right] \\
&= \frac{1}{6h_i} [k(x_{i-1}) + 4k(m_i) + k(x_i)] + \frac{1}{6h_{i+1}} [k(x_i) + 4k(m_{i+1}) + k(x_{i+1})].
\end{aligned}$$

Similarly, for the subdiagonal entries,

$$\begin{aligned}
K_{i+1,i} &= \int_I k \varphi'_i \varphi'_{i+1} dx \\
&= \int_{x_i}^{x_{i+1}} k \varphi'_i \varphi'_{i+1} dx \\
&= \frac{h_{i+1}}{6} \left[k(x_i) \cdot \left(-\frac{1}{h_{i+1}}\right) \left(\frac{1}{h_{i+1}}\right) + 4 \cdot k(m_{i+1}) \cdot \left(-\frac{1}{h_{i+1}}\right) \left(\frac{1}{h_{i+1}}\right) + k(x_{i+1}) \cdot \left(-\frac{1}{h_{i+1}}\right) \left(\frac{1}{h_{i+1}}\right) \right] \\
&= -\frac{1}{6h_{i+1}} [k(x_i) + 4k(m_{i+1}) + k(x_{i+1})].
\end{aligned}$$

Again, by symmetry, $K_{i,i+1} = K_{i+1,i}$. The following Matlab routine assembles K :

```

1 function K = StiffMatD0(x, k)
2     %input mesh vector x and function k to StiffMatD0
3     %output Stiffness Matrix K
4
5     n = length(x)-1;           %number of subintervals
6     K = zeros(n-1, n-1);       %allocate stiffness matrix
7     %No need for half-hats due to vanishing BCs; otherwise M would have dim (n+1)x(n+1)
8
9     for i = 1:n-1
10        h_minus = x(i+1) - x(i);           %h_i           (index offset)
11        xmid = (x(i+1) + x(i))/2;          %m_i           (index offset)
12        h_plus = x(i+2) - x(i+1);          %h_{i+1}       (index offset)
13        xmid_plus = (x(i+2) + x(i+1))/2;    %m_{i+1}       (index offset)
14
15        K(i,i) = (1/(6*h_minus)) * ( k(x(i)) + 4*k(xmid) + k(x(i+1))) )
16               + (1/(6*h_plus)) * ( k(x(i+1)) + 4*k(xmid_plus) + k(x(i+2))) );
17
18        if i ~= n-1
19            K(i+1,i) = - (1/(6*h_plus)) * ( k(x(i+1)) + 4*k(xmid_plus) + k(x(i+2))) );
20            K(i,i+1) = K(i+1,i);
21        end
22    end
23
24 end

```

We are down to the final component that needs to be calculated; the load vector F :

$$\begin{aligned}
F_i &= \int_I f \varphi_i dx \\
&= \int_{x_{i-1}}^{x_i} f \varphi_i dx + \int_{x_i}^{x_{i+1}} f \varphi_i dx \\
&= \frac{h_i}{6} [f(x_{i-1}) \varphi_i(x_{i-1}) + 4f(m_i) \varphi_i(m_i) + f(x_i) \varphi_i(x_i)] \\
&\quad + \frac{h_{i+1}}{6} [f(x_i) \varphi_i(x_i) + 4f(m_{i+1}) \varphi_i(m_{i+1}) + f(x_{i+1}) \varphi_i(x_{i+1})] \\
&= \frac{h_i}{6} \left[f(x_{i-1}) \cdot 0 + 4f(m_i) \cdot \left(\frac{1}{2}\right) + f(x_i) \cdot 1 \right] \\
&\quad + \frac{h_{i+1}}{6} \left[f(x_i) \cdot 1 + 4f(m_{i+1}) \cdot \left(\frac{1}{2}\right) + f(x_{i+1}) \cdot 0 \right]
\end{aligned}$$

$$= \frac{h_i}{6} [2f(m_i) + f(x_i)] + \frac{h_{i+1}}{6} [f(x_i) + 2f(m_{i+1})].$$

The following Matlab routine assembles F :

```

1 function F = LoadVecD0(x, f)
2     %input mesh vector x and function f to LoadVecD0
3     %output Load Vector F
4
5     n = length(x)-1;           %number of subintervals
6     F = zeros(n-1, 1);         %allocate load vector
7     %No need for half-hats due to vanishing BCs; otherwise F would have dim n+1
8
9     for i = 1:n-1
10        h_minus = x(i+1) - x(i);           %h_i           (index offset)
11        xmid = (x(i+1) + x(i))/2;           %m_i           (index offset)
12        h_plus = x(i+2) - x(i+1);           %h_{i+1}       (index offset)
13        xmid_plus = (x(i+2) + x(i+1))/2;     %m_{i+1}       (index offset)
14
15        F(i) = (h_minus/6) * ( f(x(i+1)) + 2*f(xmid) )
16               + (h_plus/6) * ( f(x(i+1)) + 2*f(xmid_plus) ) ;
17
18    end
19
20 end

```

We now construct an example to test our code. Consider the ansatz

$$u(x) = -x^2 + x.$$

The function u certainly satisfies the vanishing Dirichlet boundary conditions. We then choose the following functions p and k :

$$p(x) = 5xe^x;$$

$$k(x) = 1 + x.$$

Then, for the given u , p , and k , the BVP (1) becomes

$$-\frac{d}{dx} \left([1+x] \frac{d}{dx} (x-x^2) \right) + 5xe^x (x-x^2) = 5x^2e^x - 5x^3e^x + 4x + 1.$$

Thus we ended up with

$$f(x) = 5x^2e^x (1-x) + 4x + 1.$$

The following Matlab code solves the system (8) for the proposed example. Even though the code is flexible to handle nonuniform spacing, in this example we use a uniform grid:

```

1 %-----
2 % FEM code to solve the BVP (-ku')' + pu = f w/ vanishing Dirichlet BCs
3 %-----
4
5 %Interval endpoints and number of subintervals
6 a = 0;
7 b = 1;
8 n = 20;
9
10 x = linspace(a,b, n+1); %uniform mesh
11
12 %functions to be called
13 k_funct = @(x) 1+x;
14 p_funct = @(x) 5 .* x .* exp(x);
15 f_funct = @(x) (5 .* (x.^2) .* exp(x)) .* (1-x) + 4 .* x + 1;
16
17 M = MassMatD0(x, p_funct); %call mass matrix
18 K = StiffMatD0(x, k_funct); %call stiffness matrix
19 F = LoadVecD0(x, f_funct); %call load vector
20
21 U = (M+K)\F; %Solve (M+K)U = F
22 U_full = [0; U; 0]; %extend solution to include BCs

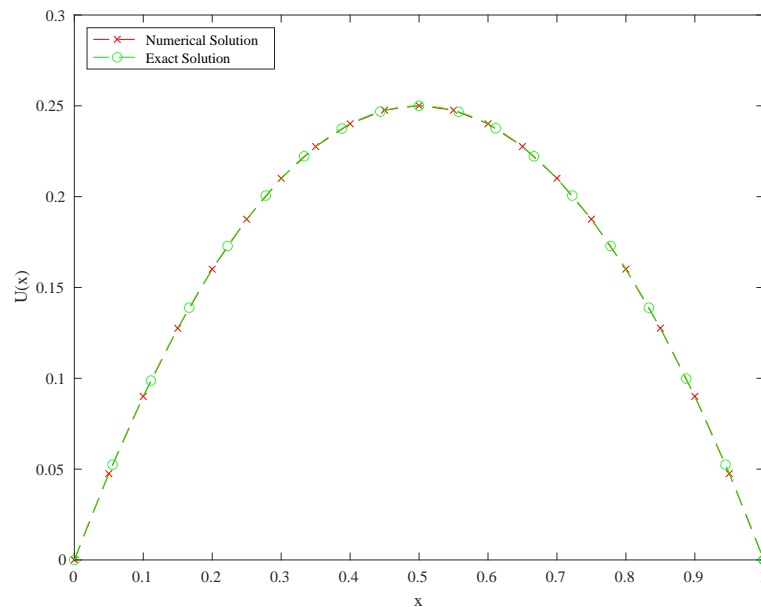
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```

23 %Plot results:
24 plot(x, U_full, "r--x")
25 hold on
26 funct = @(x) x - x.^2; %closed-form solution
27 fplot(funct, [0,1], "g--o")
28 ylabel('U(x)')
29 xlabel('x')
30 legend("Numerical Solution", "Exact Solution", 'Location','northwest')
31 exportgraphics(gcf,'FEM_Prob1.pdf')
32 close

```

The following plot shows an excellent fit between the numerical and the exact solution. All the work was worth it!



□

Problem 2. Write the expression $\nabla \cdot (\kappa \nabla u)$ explicitly in terms of partial derivatives and show that

$$\nabla \cdot (\kappa \nabla u) = \kappa \Delta u + \nabla \kappa \cdot \nabla u.$$

Proof. We work over \mathbb{R}^n and use the notation $\partial_k := \partial/\partial x_k$. Then, expanding the LHS, we have

$$\begin{aligned}
 \nabla \cdot (\kappa \nabla u) &= \begin{bmatrix} \partial_1 \\ \vdots \\ \partial_n \end{bmatrix} \cdot \kappa \begin{bmatrix} \partial_1 u \\ \vdots \\ \partial_n u \end{bmatrix} \\
 &= \partial_1 (\kappa \partial_1 u) + \cdots + \partial_n (\kappa \partial_n u) \\
 &= \kappa \partial_1^2 u + \partial_1 \kappa \partial_1 u + \cdots + \kappa \partial_n^2 u + \partial_n \kappa \partial_n u \\
 &= \kappa (\partial_1^2 + \cdots + \partial_n^2) u + \begin{bmatrix} \partial_1 \kappa \\ \vdots \\ \partial_n \kappa \end{bmatrix} \cdot \begin{bmatrix} \partial_1 u \\ \vdots \\ \partial_n u \end{bmatrix} \\
 &= \kappa \Delta u + \nabla \kappa \cdot \nabla u.
 \end{aligned}$$

□

Problem 3. Show that the following two systems are equivalent when μ and λ are constants:

- System 1:

$$\begin{aligned} -\nabla \cdot \sigma &= f \quad \text{in } \Omega, \\ \sigma &= 2\mu\epsilon + \lambda \operatorname{tr}(\epsilon)I \\ \epsilon &= \frac{1}{2}(\nabla u + \nabla u^\top). \end{aligned}$$

- System 2:

$$\begin{aligned} -(2\mu + \lambda)\frac{\partial^2 u_1}{\partial x^2} - \mu\frac{\partial^2 u_1}{\partial y^2} - (\mu + \lambda)\frac{\partial^2 u_2}{\partial y \partial x} &= f_1, \\ -(\mu + \lambda)\frac{\partial^2 u_1}{\partial y \partial x} - \mu\frac{\partial^2 u_2}{\partial x^2} - (2\mu + \lambda)\frac{\partial^2 u_2}{\partial y^2} &= f_2. \end{aligned}$$

Proof. The divergence operator $\nabla \cdot (\cdot)$ is a rank-lowering operation on tensors. In particular, when applied to the matrix σ (rank-2) we end up with a vector (rank-1). The latter vector, explicitly, has components that are the divergences of the rows of the original matrix σ . Thus,

$$-\nabla \cdot \sigma = - \begin{bmatrix} \partial_x \\ \partial_y \end{bmatrix} \cdot \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix} = - \begin{bmatrix} \partial_x \sigma_{11} + \partial_y \sigma_{12} \\ \partial_x \sigma_{21} + \partial_y \sigma_{22} \end{bmatrix}. \quad (10)$$

In order to expand this expression and show its equivalence to the LHS of System 2, we need to write the matrix σ explicitly; the first order of business then is to write ϵ explicitly in matrix form. To accomplish the latter we first note that, since now $u: \mathbb{R}^2 \rightarrow \mathbb{R}^2$, the gradients are Jacobians:

$$\nabla u = \begin{bmatrix} \partial_x u_1 & \partial_y u_1 \\ \partial_x u_2 & \partial_y u_2 \end{bmatrix}, \quad \nabla u^\top = \begin{bmatrix} \partial_x u_1 & \partial_x u_2 \\ \partial_y u_1 & \partial_y u_2 \end{bmatrix}. \quad (11)$$

Thus,

$$\begin{aligned} \epsilon &= \frac{1}{2} \left(\begin{bmatrix} \partial_x u_1 & \partial_y u_1 \\ \partial_x u_2 & \partial_y u_2 \end{bmatrix} + \begin{bmatrix} \partial_x u_1 & \partial_x u_2 \\ \partial_y u_1 & \partial_y u_2 \end{bmatrix} \right) \\ &= \begin{bmatrix} \partial_x u_1 & \frac{1}{2}(\partial_x u_2 + \partial_y u_1) \\ \frac{1}{2}(\partial_y u_1 + \partial_x u_2) & \partial_y u_2 \end{bmatrix}, \end{aligned}$$

and

$$\operatorname{tr}(\epsilon) = \epsilon_{11} + \epsilon_{22} = \partial_x u_1 + \partial_y u_2.$$

Hence, for constant λ and μ , we get

$$\begin{aligned} \sigma &= 2\mu\epsilon + \lambda \operatorname{tr}(\epsilon)I \\ &= 2\mu \begin{bmatrix} \partial_x u_1 & \frac{1}{2}(\partial_x u_2 + \partial_y u_1) \\ \frac{1}{2}(\partial_y u_1 + \partial_x u_2) & \partial_y u_2 \end{bmatrix} + \begin{bmatrix} \lambda(\partial_x u_1 + \partial_y u_2) & 0 \\ 0 & \lambda(\partial_x u_1 + \partial_y u_2) \end{bmatrix} \\ &= \begin{bmatrix} \partial_x u_1(2\mu + \lambda) + \lambda \partial_y u_2 & \mu(\partial_x u_2 + \partial_y u_1) \\ \mu(\partial_y u_1 + \partial_x u_2) & \partial_y u_2(2\mu + \lambda) + \lambda \partial_x u_1 \end{bmatrix}. \end{aligned}$$

We now substitute into Eq. (10), one row at a time:

$$\begin{aligned} -\partial_x \sigma_{11} - \partial_y \sigma_{12} &= - (2\mu + \lambda) \partial_x^2 u_1 - \lambda \partial_{yx} u_2 - \mu \partial_{xy} u_2 - \mu \partial_y^2 u_1 \\ &= - (2\mu + \lambda) \partial_x^2 u_1 - (\mu + \lambda) \partial_{xy} u_2 - \mu \partial_y^2 u_1 \\ -\partial_x \sigma_{21} - \partial_y \sigma_{22} &= -\mu \partial_{yx} u_1 - \mu \partial_x^2 u_2 - (2\mu + \lambda) \partial_y^2 u_2 - \lambda \partial_{xy} u_1 \\ &= - (\mu + \lambda) \partial_{xy} u_1 - \mu \partial_x^2 u_2 - (2\mu + \lambda) \partial_y^2 u_2. \end{aligned}$$

(In these calculations we used the commutativity of the mixed partials; $\partial_{xy} = \partial_{yx}$.) Hence, since $f = [f_1 \ f_2]^\top$, we conclude that the two systems are identical. \square

Problem 4. Let Ω be the unit square: $\Omega = (0, 1) \times (0, 1)$. Verify that

$$-\int_{\Omega} v \Delta u = \int_{\Omega} \nabla v \cdot \nabla u - \int_{\partial\Omega} v \frac{\partial u}{\partial n}, \quad (12)$$

for

$$u(x, y) = 1 + xy^2, \quad v(x, y) = x + xy.$$

Proof. We start by tackling the LHS; first note that

$$\Delta u = \partial_x^2 u + \partial_y^2 u = 0 + 2x = 2x.$$

Then,

$$\begin{aligned} -\int_{\Omega} v \Delta u &= -\int_0^1 \int_0^1 (x + xy) (2x) \, dx \, dy \\ &= -\int_0^1 \int_0^1 2x^2 (1 + y) \, dx \, dy \\ &= -\int_0^1 \left. \frac{2}{3} x^3 \right|_0^1 (1 + y) \, dy \\ &= -\frac{2}{3} \int_0^1 (1 + y) \, dy \\ &= -\frac{2}{3} \left(y + \frac{1}{2} y^2 \right) \Big|_0^1 \\ &= -1. \end{aligned}$$

Now, on to the first term on the RHS:

$$\begin{aligned} \int_{\Omega} \nabla v \cdot \nabla u &= \int_0^1 \int_0^1 \begin{bmatrix} \partial_x v \\ \partial_y v \end{bmatrix} \cdot \begin{bmatrix} \partial_x u \\ \partial_y u \end{bmatrix} \, dx \, dy \\ &= \int_0^1 \int_0^1 \begin{bmatrix} 1 + y \\ x \end{bmatrix} \cdot \begin{bmatrix} y^2 \\ 2xy \end{bmatrix} \, dx \, dy \\ &= \int_0^1 \int_0^1 (y^3 + y^2 + 2x^2 y) \, dx \, dy \\ &= \int_0^1 \int_0^1 \left(y^3 + y^2 + y \frac{2}{3} x^3 \Big|_0^1 \right) \, dy \\ &= \left(\frac{1}{4} y^4 + \frac{1}{3} y^3 + \frac{2}{3} \frac{1}{2} y^2 \right) \Big|_0^1 \\ &= \frac{11}{12}. \end{aligned}$$

For the last term on the RHS of Eq. (12) we must choose an orientation for the boundary $\partial\Omega$; let us choose the “right-handed” orientation (i.e., counterclockwise). We also use the definition of the normal derivative:

$$\frac{\partial u}{\partial n} := \nabla u \cdot n,$$

where n is the (outward-pointing) unit normal vector. Then,

$$\int_{\partial\Omega} v \frac{\partial u}{\partial n} = \int_0^1 v \nabla u \cdot n \, dx \Big|_{y=0} + \int_0^1 v \nabla u \cdot n \, dy \Big|_{x=1} + \int_1^0 v \nabla u \cdot n \, dx \Big|_{y=1} + \int_1^0 v \nabla u \cdot n \, dy \Big|_{x=0}$$

$$\begin{aligned}
&= \int_0^1 (x+xy) \left[\frac{y^2}{2xy} \right] \cdot \left[\frac{0}{-1} \right] dx \Big|_{y=0} + \int_0^1 (x+xy) \left[\frac{y^2}{2xy} \right] \cdot \left[\frac{1}{0} \right] dy \Big|_{x=1} \\
&+ \int_1^0 (x+xy) \left[\frac{y^2}{2xy} \right] \cdot \left[\frac{0}{1} \right] dx \Big|_{y=1} + \int_1^0 (x+xy) \left[\frac{y^2}{2xy} \right] \cdot \left[\frac{-1}{0} \right] dy \Big|_{x=0} \\
&= \int_0^1 (x+xy) (-2xy) dx \Big|_{y=0}^0 + \int_0^1 (x+xy) y^2 dy \Big|_{x=1} \\
&+ \int_1^0 (x+xy) (2xy) dx \Big|_{y=1} + \int_1^0 (x+xy) (-y^2) dy \Big|_{x=0}^0 \\
&= \int_0^1 (y^3 + y^2) dy + \int_1^0 4x^2 dx \\
&= \frac{1}{4} + \frac{1}{3} + \frac{4}{3} = \frac{23}{12}.
\end{aligned}$$

Hence the RHS of Eq. (12) is

$$\frac{11}{12} - \frac{23}{12} = -1,$$

which proves the validity of Eq. (12). \square

Problem 5. Let $\sigma : \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$ be smooth. Use the ordinary divergence theorem to show that

$$\int_{\Omega} \nabla \cdot \sigma = \int_{\partial\Omega} \sigma n. \quad (13)$$

Proof. We recall from Eq. (10) that

$$\nabla \cdot \sigma = \begin{bmatrix} \partial_x \sigma_{11} + \partial_y \sigma_{12} \\ \partial_x \sigma_{21} + \partial_y \sigma_{22} \end{bmatrix}.$$

On the other hand,

$$\sigma n = \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix} \begin{bmatrix} n_1 \\ n_2 \end{bmatrix} = \begin{bmatrix} \sigma_{11}n_1 + \sigma_{12}n_2 \\ \sigma_{21}n_1 + \sigma_{22}n_2 \end{bmatrix}.$$

Now, since the integral of a vector-valued function is computed by taking the integral of each component of the function, Eq. (13) yields the following system:

$$\int_{\Omega} (\partial_x \sigma_{11} + \partial_y \sigma_{12}) = \int_{\partial\Omega} (\sigma_{11}n_1 + \sigma_{12}n_2) \quad (14a)$$

$$\int_{\Omega} (\partial_x \sigma_{21} + \partial_y \sigma_{22}) = \int_{\partial\Omega} (\sigma_{21}n_1 + \sigma_{22}n_2). \quad (14b)$$

Then, considering the vectors

$$\sigma_{(1)} = \begin{bmatrix} \sigma_{11} \\ \sigma_{12} \end{bmatrix} \quad \sigma_{(2)} = \begin{bmatrix} \sigma_{21} \\ \sigma_{22} \end{bmatrix},$$

the above system becomes

$$\int_{\Omega} \nabla \cdot \sigma_{(1)} = \int_{\partial\Omega} \sigma_{(1)} \cdot n \quad (14c)$$

$$\int_{\Omega} \nabla \cdot \sigma_{(2)} = \int_{\partial\Omega} \sigma_{(2)} \cdot n. \quad (14d)$$

Both of these integral equations hold by the ordinary Divergence Theorem, thereby demonstrating the validity of Eq. (13). \square

Problem 6. Let $\sigma: \mathbb{R}^2 \rightarrow \mathbb{R}^{2 \times 2}$ and $v: \mathbb{R}^2 \rightarrow \mathbb{R}^2$ be smooth. Show that

$$\nabla \cdot (\sigma v) = (\nabla \cdot \sigma^\top) \cdot v + \sigma \cdot \nabla v^\top. \quad (15)$$

Proof. We first expand the LHS:

$$\begin{aligned} \nabla \cdot (\sigma v) &= \begin{bmatrix} \partial_x \\ \partial_y \end{bmatrix} \cdot \left(\begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix} \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \right) \\ &= \begin{bmatrix} \partial_x \\ \partial_y \end{bmatrix} \cdot \begin{bmatrix} \sigma_{11}v_1 + \sigma_{12}v_2 \\ \sigma_{21}v_1 + \sigma_{22}v_2 \end{bmatrix} \\ &= \partial_x (\sigma_{11}v_1 + \sigma_{12}v_2) + \partial_y (\sigma_{21}v_1 + \sigma_{22}v_2) \\ &= v_1 (\partial_x \sigma_{11} + \partial_y \sigma_{21}) + v_2 (\partial_x \sigma_{12} + \partial_y \sigma_{22}) + \sigma_{11} \partial_x v_1 + \sigma_{12} \partial_x v_2 + \sigma_{21} \partial_y v_1 + \sigma_{22} \partial_y v_2. \end{aligned}$$

Now on to the first term on the RHS:

$$\begin{aligned} (\nabla \cdot \sigma^\top) \cdot v &= \left(\begin{bmatrix} \partial_x \\ \partial_y \end{bmatrix} \cdot \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix} \right) \cdot \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \\ &= \begin{bmatrix} \partial_x \sigma_{11} + \partial_y \sigma_{21} \\ \partial_x \sigma_{12} + \partial_y \sigma_{22} \end{bmatrix} \cdot \begin{bmatrix} v_1 \\ v_2 \end{bmatrix} \\ &= v_1 (\partial_x \sigma_{11} + \partial_y \sigma_{21}) + v_2 (\partial_x \sigma_{12} + \partial_y \sigma_{22}). \end{aligned}$$

On the second term of the RHS the dot product we use is the real Frobenius inner product, which is defined by

$$A \cdot B := A \otimes_F B = \sum_{i,j} a_{ij} b_{ij}.$$

Hence,

$$\begin{aligned} \sigma \cdot \nabla v^\top &= \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix} \cdot \begin{bmatrix} \partial_x v_1 & \partial_x v_2 \\ \partial_y v_1 & \partial_y v_2 \end{bmatrix} \\ &= \sigma_{11} \partial_x v_1 + \sigma_{12} \partial_x v_2 + \sigma_{21} \partial_y v_1 + \sigma_{22} \partial_y v_2. \end{aligned}$$

Looking at the color-coded results, we see that the equality (15) does hold. □

Problem 7. Derive the weak form of the following BVP with inhomogeneous boundary conditions:

$$\begin{aligned} -\nabla \cdot \sigma &= f \quad \text{in } \Omega, \\ \sigma &= 2\mu \epsilon + \lambda \text{tr}(\epsilon) I \\ \epsilon &= \frac{1}{2} (\nabla u + \nabla u^\top) \\ u &= g \quad \text{in } \Gamma_1 \\ \sigma n &= h \quad \text{in } \Gamma_2. \end{aligned}$$

Solution. Consider some test function v . We showed in Problem 6 that

$$\nabla \cdot (\sigma v) = (\nabla \cdot \sigma^\top) \cdot v + \sigma \cdot \nabla v^\top.$$

When σ is symmetric (which is indeed true in our case, since ϵ is symmetric), the above expression becomes

$$\nabla \cdot (\sigma v) = (\nabla \cdot \sigma) \cdot v + \sigma \cdot \nabla v. \quad (16)$$

This holds because, when σ is symmetric,

$$\begin{aligned} \sigma \cdot \nabla v &= \begin{bmatrix} \sigma_{11} & \sigma_{12} \\ \sigma_{21} & \sigma_{22} \end{bmatrix} \cdot \begin{bmatrix} \partial_x v_1 & \partial_y v_1 \\ \partial_x v_2 & \partial_y v_2 \end{bmatrix} \\ &= \sigma_{11} \partial_x v_1 + \sigma_{12} \partial_y v_1 + \sigma_{21} \partial_x v_2 + \sigma_{22} \partial_y v_2 \\ &= \sigma_{11} \partial_x v_1 + \sigma_{12} \partial_x v_2 + \sigma_{21} \partial_y v_1 + \sigma_{22} \partial_y v_2 \\ &= \sigma \cdot \nabla v^\top. \end{aligned}$$

But then

$$\sigma \cdot \nabla v = \sigma \cdot \nabla v^\top = \sigma \cdot \epsilon_v,$$

where

$$\epsilon_v := \frac{1}{2} (\nabla v + \nabla v^\top).$$

By the Divergence Theorem, we have

$$\int_{\Omega} \nabla \cdot (\sigma v) = \int_{\partial\Omega} (\sigma v) \cdot n$$

and, moreover, since σ is symmetric,

$$(\sigma v) \cdot n = v \cdot (\sigma n).$$

Thus, combining these results with Eq. (16), we get

$$\int_{\partial\Omega} v \cdot (\sigma n) = \int_{\Omega} (\nabla \cdot \sigma) \cdot v + \int_{\Omega} \sigma \cdot \epsilon_v.$$

Hence, going back to our original BVP, if we multiply through by a test function v and integrate over Ω , we have

$$\begin{aligned} - \int_{\Omega} (\nabla \cdot \sigma) \cdot v &= \int_{\Omega} f \cdot v \\ \int_{\Omega} \sigma \cdot \epsilon_v - \int_{\partial\Omega} v \cdot (\sigma n) &= \int_{\Omega} f \cdot v. \end{aligned}$$

Lastly, since

$$\partial\Omega = \Gamma_1 \cup \Gamma_2,$$

taking into account the imposed boundary conditions we end up with the weak form of the BVP:

$$\int_{\Omega} \sigma \cdot \epsilon_v - \int_{\Gamma_1} v \cdot (\sigma_g n) - \int_{\Gamma_2} v \cdot h = \int_{\Omega} f \cdot v \quad (17)$$

where

$$\begin{aligned} \sigma_g &= 2\mu\epsilon_g + \lambda \operatorname{tr}(\epsilon_g)I; \\ \epsilon_g &= \frac{1}{2} (\nabla g + \nabla g^\top). \end{aligned} \quad \square$$

Problem 8. Solve the following heat equation by using the FEM:

$$\begin{aligned} \frac{\partial u}{\partial t} - \frac{\partial^2 u}{\partial x^2} &= x(1-x) \cos t, \quad 0 < x < 1, \quad t > 0, \\ u(0, x) &= 1, \quad 0 < x < 1, \\ u(t, 0) &= 0, \quad t > 0, \\ u(t, 1) &= 0, \quad t > 0. \end{aligned}$$

Use S_3 as the approximating subspace. Explicitly compute the mass matrix M , the stiffness matrix K , and the load vector $F(t)$. Explicitly set up the system of ODEs and solve it.

Solution. We use the “dot” notation for time-derivatives and “prime” notation for spatial derivatives; moreover we let $I = [0, 1]$. Then, multiplying through by some test function v and integrating, we have

$$\int_I \dot{u}v - \int_I u''v = \int_I f v, \quad (18)$$

with

$$f = f(t, x) := x(1 - x) \cos t.$$

Now, from a straightforward application of the product rule,

$$\left. u'v \right|_0^1 = \int_I (u'v)' = \int_I u''v + \int_I u'v'.$$

Thus, plugging back into Eq. (18), we get

$$\int_I \dot{u}v + \int_I u'v' = \int_I f v \quad (19)$$

This is the weak form of the original Heat Equation. We then recall the ansatz (6); since we are now using S_3 as the approximating subspace, we will only be using three hat-functions. Moreover, the coefficients U_i now depend on time. Thus we have

$$^{(3)}u(t, x) = \sum_{i=1}^3 U_i(t) \varphi_i(x).$$

Plugging this into our weak form (19) and substituting φ_i 's for v 's, we have

$$\begin{aligned} \int_I \left(\sum_{j=1}^3 U_j \varphi_j \right) \varphi_i + \int_I \left(\sum_{j=1}^3 U_j \varphi_j \right)' \varphi_i' &= \int_I f \varphi_i \\ \sum_{j=1}^3 \dot{U}_j \int_I \varphi_j \varphi_i + \sum_{j=1}^3 U_j \int_I \varphi_j' \varphi_i' &= \int_I f \varphi_i, \quad \text{for } i = 1, 2, 3. \end{aligned}$$

This last expression is of the form

$$\mathbf{M}\dot{\mathbf{U}} + \mathbf{K}\mathbf{U} = \mathbf{F}, \quad (20)$$

where \mathbf{M} and \mathbf{K} are, respectively, the mass and stiffness matrices we defined before in Problem 1, except that now $p(x) = k(x) \equiv 1$. We also note that this time the load vector, \mathbf{F} , does depend on time. Using uniform grid-spacing $h \equiv 1/(3+1) = 1/4$ and plugging back into the expressions we derived on Problem 1, the system takes the form

$$\underbrace{\begin{bmatrix} 1/6 & 1/24 & 0 \\ 1/24 & 1/6 & 1/24 \\ 0 & 1/24 & 1/6 \end{bmatrix}}_{\mathbf{M}} \underbrace{\begin{bmatrix} \dot{U}_1 \\ \dot{U}_2 \\ \dot{U}_3 \end{bmatrix}}_{\dot{\mathbf{U}}} + \underbrace{\begin{bmatrix} 8 & -4 & 0 \\ -4 & 8 & -4 \\ 0 & -4 & 8 \end{bmatrix}}_{\mathbf{K}} \underbrace{\begin{bmatrix} U_1 \\ U_2 \\ U_3 \end{bmatrix}}_{\mathbf{U}} = \underbrace{\cos t \begin{bmatrix} 17/384 \\ 23/384 \\ 17/384 \end{bmatrix}}_{\mathbf{F}}.$$

Hence we have reduced a PDE problem to a simple ODE problem, which we can now solve using any of the ODE methods we have previously studied. Since we are only using three hat functions to approximate a solution over the entire interval $[0, 1]$, we cannot realistically expect to get a very smooth solution. Thus, since accuracy is not much of a concern for this exercise, we don't need some highly accurate method like RK4. We shall instead implement Backward Euler, so that Eq. (20) is rewritten as

$$(\mathbf{M} + \Delta t \mathbf{K}) \mathbf{U}^{n+1} = \mathbf{M} \mathbf{U}^n + \Delta t \mathbf{F}^{n+1}, \quad (21)$$

where, per usual notation, the superscripts denote the time step; i.e., $\mathbf{U}^n = \mathbf{U}(t_0 + n\Delta t)$. The following Matlab script implements the Backward Euler method:

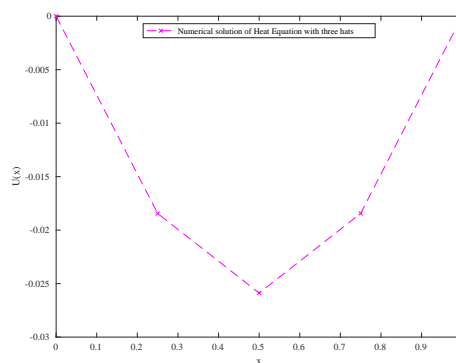
```
1 %Backward Euler solution to the Heat Eq using only three hat functions
2 %(result expected to look rough!)
3
4 m = 3;
5 h = 1/4;
6 dt = h^2/2; %time-stepping (can be larger since we're using an implicit scheme)
7
8 %Generate arrays:
9 M = zeros(m);
10 K = zeros(m);
11 U_0 = ones(m,1); %initial conditions
12 rhs = zeros(m,1); %initialize rhs of Eq
```

```

13 for i = 1:m
14     M(i,i) = 1/6;
15     K(i,i) = 8;
16     if i ~= m
17         M(i,i+1) = 1/24;
18         M(i+1,i) = M(i,i+1);
19         K(i,i+1) = -4;
20         K(i+1,i) = K(i,i+1);
21     end
22 end
23
24 Mat = M + dt*K;
25
26 f_vec = [17/384; 23/384; 17/384];
27 f = @(t) cos(t);
28
29 %-----
30 %       BACKWARD EULER CODE
31 %-----
32 it_max = 500;           %max number of iterations allowed
33 tol = 1e-5;            %tolerance allowed
34 it = 0;
35
36 for n = 1 : it_max
37     it = it + 1;
38     rhs = M * U_0 + dt * f((n+1)*dt) * f_vec;
39     U = Mat\rhs;
40
41     if norm(U - U_0) <= tol
42         disp(['It took ', num2str(it), ' iterations for the solution to converge.'])
43         break
44     elseif it == it_max
45         disp('No convergence; max number of iterations reached.')
46     end
47
48     U_0 = U; %update U_0 value for next iteration
49 end
50 %-----
51 %       END OF BACKWARD EULER CODE
52 %-----
53
54 %extend solution to include boundaries
55 U = [0; U; 0];
56
57 x = linspace(0,1,m+2);
58
59 %Plot results:
60 plot(x,U, "r--x")
61 ylabel('U(x)')
62 xlabel('x')
63 legend("Numerical solution of Heat Equation with three hats", 'Location','north')
64 exportgraphics(gcf, 'BE_Heateq_S_3.pdf')
65 close

```

The code reaches the desired tolerance after 103 iterations and outputs the following plot:



As expected, the solution is not very smooth-looking, since we are only using three φ_i 's over the entire interval $[0, 1]$. However, it does showcase the power of using FEM for the space discretization, since had we used only three interior points for a Finite Differences implementation, the results would look a lot worse! \square