Assignment 1

Finite Element Methods

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1. introduction

In this report we will discuss the effect on the error bounds of a finite element method that changing the underlying mesh has, using the Poisson Equation:

$$-u'' = f, -1 < x < 1,$$

$$(1.2) u(-1) = u(1) = 0.$$

Here u is the function of interest, and f the driving force, which is taken to be equal to

$$f(x) = e^{-1000x^2} + 10^{-3}$$

Furthermore we will derive an a-posteriori error-bound on ||u'||.

$2.\ a ext{-}posteriori\ error-estimate$

Using the notation $\|\cdot\|$ to indicate the L^2 norm over the whole domain and $\|\cdot\|_i$ to indicate the L^2 norm over the interval $[x_{i-1}, x_i]$:

It is known that in the 1D case the constant C is unity.¹

¹ The Finite Element Method: Theory, Implementation, and Applications page 7

3. The Finite Element Implementation

We modeled the problem using a finite element method and the standard hat-functions, defined by

(3.1)
$$\phi_i(x_{i-1} \le x \le x_i) = (x - x_{i-1})/h,$$

$$\phi_i(x_i \le x \le x_{i+1}) = (x_{i+1} - x)/h,$$

$$\phi_i(x \setminus [x_{i-1}, x_{i+1}]) = 0.$$

The boundary conditions were enforced by replacing the first and last row of the stiffness matrix A by their respective unit-vectors and the first and last element of the load vector b by the boundary values.

After we have computed our initial solution, we compute the Laplacian by solving

$$(3.2) M\chi = -A\xi$$

for χ . Here M is the mass matrix, defined by

$$M_{ij} = \int_0^1 \phi_j(x)\phi_i(x) \, \mathrm{d}x$$

. We subsequent compute ρ_i using the trapezoidal rule for the interior points, that is:

(3.4)
$$\rho_i = \frac{1}{2}h_i^2 \left((f(x_{i-1}) + \chi_{i-1}) + (f(x_i) + \chi_i) \right),$$

For the boundray intervals, we use a single pair of χ and f instead.

4. Adaptive Mesh Refinement

Having obtained a set of ρ_i (and thus an estimate of the error), we refine our mesh by subdividing each interval i for which

(4.1)
$$\rho_i > \lambda \max_i (\rho_i), \ \lambda \in [0, 1].$$

We have tested $\lambda = 0.05, 0.25, 0.5, 1$, starting from 5 nodes and stopping once we have more than 1000 nodes.

5. Results

Our results are contained in figure 1, 2, and 3.

Looking at figure 1 we note that the result becomes less smooth for increasing λ near its inflection point, as the algorithm puts less points there for higher λ . Visually, all results look similar.

Looking at the total error estimate in figure 2 we see that higher values of λ perform better by up to two orders of magnitude for the same amount of points N. It should be note however, that for high values of λ the mesh refinement algorithm takes significantly longer to reach N = 1000 as it adds fewer points per iteration.

Finally, figure 3 shows how the density of points changes with λ . Surprisingly, larger values of λ give a more uniform distribution while = 0 will always give a perfectly uniform distribution.

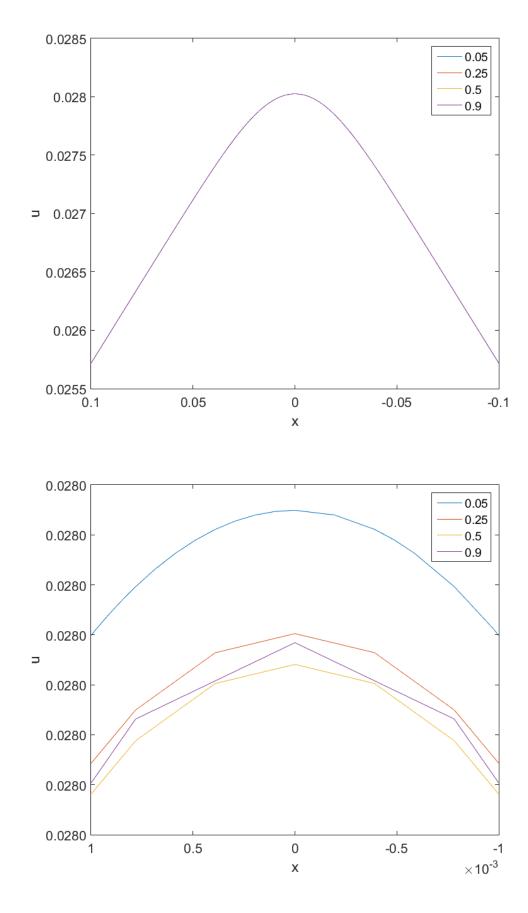


Figure 1: The result of our method for different values of λ .

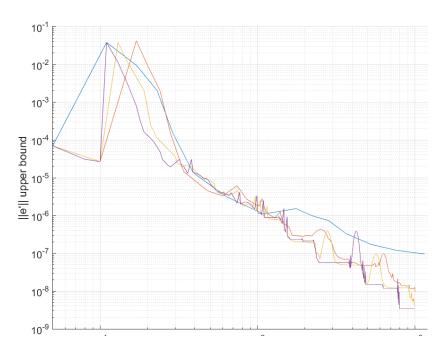


Figure 2: The total sum of ρ_i^2 for different values of λ .

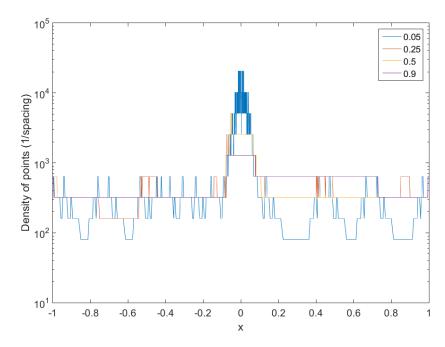


Figure 3: The density of points for different values of λ .

6. Code

6a) Adaptive

```
function adaptive (N, N_max, lambda)
a = -1; % left end point of interval
b = 1; \% right
h = (b-a)/N; \% \text{ mesh size}
x = a:h:b; \% node coords
low = 0;
high = 0.;
figure(4)
hold on;
figure (3)
hold on;
figure (2)
hold on;
figure (1)
hold on;
N_{array} = [];
total_residual_array = [];
while N < N_{max}
    N_{array} = [N_{array} N];
    B=my_load_vector_assembler(x, low, high, @f);
    M = mass_{matrix}(x); %last points kept fixed
    A_fixed =stiffness_matrix_fixed(x,low, high);
    xi_fixed = A_fixed\B; % solve system of equations
    reversed = (-A_fixed*xi_fixed);
    lap fixed = M \setminus (reversed(2:end-1));
    F = \operatorname{arrayfun}(@f, x(2:end-1)).;
    delta = abs(F+lap\_fixed);
    rho = trapezoidal(x, delta);
    total_residual_array = [total_residual_array sqrt(sum(rho.*rho))];
    threshold = max(rho)*lambda;
    for i = 1: length(rho)
         if rho(i) > threshold
             x = [x (x(i+1)+x(i))/2];
         end
    end
    x = sort(x);
    N = size(x,2)
end
%total_residual_array(end)
%rho
N_{array} = [N_{array} N]
B=my_load_vector_assembler(x, low, high, @f);
```

```
M = mass_matrix(x); %last points kept fixed
 A_fixed =stiffness_matrix_fixed(x,low, high);
 xi_fixed = A_fixed\B; % solve system of equations
 reversed = (-A_fixed*xi_fixed);
 lap\_fixed = M \setminus (reversed(2:end-1));
F = \operatorname{arrayfun}(@f, x(2:end-1)).;
 delta = abs(F+lap\_fixed);
 rho = trapezoidal(x, delta);
 total_residual_array = [total_residual_array sqrt(sum(rho.*rho))];
 figure (1)
 plot(x, xi_fixed)
 figure (2)
 semilogy(x(2:end), [1./diff(x)])
 figure (3)
 loglog(N_array, total_residual_array)
 figure (4)
 plot(x(1:end-1), rho)
\%x
 function y=f(x)
\%y = 2;
y=\exp(-1000*x^2)+10^-3;
\%y = pi^2*49*sin(x*pi*7);
\%y = x*(x-1);
       6b) Mass Matrix
 function M = mass_matrix(x)
           N = length(x) - 2;
            diag = zeros(N,1);
            h = zeros(N,1);
            upper = zeros(N-1,1);
            for i = 1:N+1
                       h(i) = x(i+1)-x(i);
            end
           \% \operatorname{diag}(1) = (1/3 * (x(2)^3 - x(1)^3) + x(2)^2 * h(1) - x(2) * (x(2)^2 - x(1)^2)) / h(1)^2;
            for i = 2:N+1
                                                                                      (1/3*(x(i+1)^3-x(i)^3)+x(i+1)^2*h(i)-x(i+1)*(x(i+1)^2-x(i)^3)
                        \operatorname{diag}(i-1) =
                        \mathrm{diag}\,(\,\mathrm{i}\,-1)\,=\,\mathrm{diag}\,(\,\mathrm{i}\,-1)\,+\,(\,1/\,3*(\,x\,(\,\mathrm{i}\,)\,\widehat{}\,3-x\,(\,\mathrm{i}\,-1)\,\widehat{}\,3)+x\,(\,\mathrm{i}\,-1)\,\widehat{}\,2*h\,(\,\mathrm{i}\,-1)-x\,(\,\mathrm{i}\,-1)*(\,x\,(\,\mathrm{i}\,)\,\widehat{}\,2-x\,(\,\mathrm{i}\,-1)\,\widehat{}\,3)
            end
           \% \operatorname{diag}(N+1) = (1/3*(x(N+1)^3-x(N)^3)+x(N)^2*h(N)-x(N)*(x(N+1)^2-x(N)^2))/h(N)^2;
            for i = 1:N-1
                       upper (i) = (-1/3*(x(i+1)^3-x(i)^3)+1/2*(x(i)+x(i+1))*(x(i+1)^2-x(i)^2)-x(i)*x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3)+x(i-1)*(x(i+1)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)^3-x(i)
                       %lower(i) = (-1/3*(x(i+1)^3-x(i)^3)+1/2*(x(i)+x(i+1))*(x(i+1)^2-x(i)^2)-x(i)*x(i+1)^2
            end
           M = gallery('tridiag', upper, diag, upper);
```

end

6c) Trapezoidal

```
\label{eq:normalization} \begin{array}{lll} function & [ & rho & ] = trapezoidal(x, delta) \\ & N = length(delta)+1; \\ & rho = zeros(N,1); \\ & rho(1) = (x(2)-x(1))^2*delta(1); \ \% first interval, project backwards \\ & for & i = 2:N-1 \\ & & h = x(i+1)-x(i); \\ & & rho(i) = 0.5*h^2*(delta(i-1)+delta(i)); \\ & end \\ & rho(N) = (x(N+1)-x(N))^2*delta(N-1); \ \% last interval, project forwards \\ end \end{array}
```