UPPSALA UNIVERSITY



ADVANCED NUMERICAL METHODS 1TD050

Assignment 1

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Introduction

In this report we are investigating the numerical solution of the following given PDE problem:

Let Ω be a fixed (open) domain in \mathbb{R}^2 , with boundary $\partial\Omega$ over a time interval [0,T] with initial time zero and the final time T. We are interested in solving the following time-dependent scalar conservation laws:

$$\partial_t u + \nabla \cdot \boldsymbol{f}(u) = 0, \qquad (\boldsymbol{x}, t) \in \Omega \times (0, T],$$

 $u(\boldsymbol{x}, 0) = u_0(\boldsymbol{x}), \qquad \boldsymbol{x} \in \Omega,$

with appropriate boundary conditions. Here u represents the unknown variable, $f \in C^1(\mathbb{R}, \mathbb{R}^2)$ is the flux term and $u_0 \in L^{\infty}(\mathbb{R}^2)$ is given initial data. Let $0 = t_0 < t_1 < \ldots < t_N = T$ be a sequence of discrete time steps with associated time intervals $I_n = (t_{n-1}, t_n]$ of length $k_n = t_n - t_{n-1}, n = 1, 2, \dots, N$. Let

$$\mathcal{X}_h := \left\{ v : v \in H^1(\Omega), v(\boldsymbol{x}) - \text{ cont. pw. linear in } \Omega \right\},$$

be a finite element space consisting of continuous piecewise linear polynomials on a mesh $\mathcal{T}_h = \{K\}$ of mesh-size $h(\boldsymbol{x})$ and let $V_{h,0}$ be the space of all functions in \mathcal{X}_h vanishing on $\partial\Omega$. Next, let us denote by $U_0 = \hat{\pi}_h u_0$ the interpolation of the initial data into the finite element space \mathcal{X}_h . We are now ready to formulate the following finite element approximation: for n = $1, 2, \dots, N$ find $U_n \in \mathcal{X}_h$ such that

$$\frac{1}{k_n} \left(U_n - U_{n-1}, v \right) + \frac{1}{2} \left(\nabla \cdot \left(\boldsymbol{f} \left(U_{n-1} \right) + \boldsymbol{f} \left(U_n \right) \right), v \right) = 0, \quad \forall v \in \mathcal{X}_h, \quad (1)$$

where $U_n = U_h(t_n)$ is the solution at the discrete time steps t_n . The implicit Crank-Nicholson time-stepping is used for the time discretization.

Note: In the tasks we will use $\Omega = \{\mathbf{x} : x_1^2 + x_2^2 \le 1\}$, and Dirichlet boundary condition.

Part I

Problem 1.1

In this task we should implement a Matlab code to solve the finite element approximation given in (1). The implementation should be done with the following values: $u_0(\mathbf{x},0) = \frac{1}{2} \left(1 - \tanh\left(\frac{(x_1 - x_1^0)^2 + (x_2 - x_2^0)^2}{r_0^2} - 1\right)\right)$ is the initial data with $r_0 = 0.25$, $(x_1^0, x_2^0) = (0.3, 0)$. Additionally we should use T = 1, CFL = 0.5 and create a solution using $h_{max} = \frac{1}{8}$ and $h_{max} = \frac{1}{16}$ mesh size.

First let's rewrite the flux term of equation (1) in the given way. So use $\nabla \cdot \mathbf{f}(u) := \mathbf{f'}(u) \cdot \nabla u$, assuming $\mathbf{f'}(u) = 2\pi(-x_2, x_1)$. Then knowing that ∇ is a linear operator, we get this:

$$\frac{1}{k_n}\left(U_n - U_{n-1}, v\right) + \frac{1}{2}\left(\mathbf{f'}(U_{n-1}) \cdot \nabla U_{n-1} + \mathbf{f'}(U_n) \cdot \nabla U_n, v\right) = 0, \quad \forall v \in \mathcal{X}_h,$$

After this step, we should create the system of linear equations from this FEM formulation using the appropriate basis functions. Then we can solve the equation system for U_{n+1} in case U_n is given using the given u_0 value.

We can write U_n in the form $U_n = \sum \xi_i \varphi_i$, where the sum is taken from the φ_i basis function of \mathcal{X}_h space. And in the form of the basis function the statement that $\forall v \in \mathcal{X}_h$, is equivalent with the statement that the equation holds for all basis function. From this we get the following for of the equation:

$$\frac{1}{k_n} \sum_{N_i \in \mathcal{N}_b} (\xi_{i,n} - \xi_{i,n-1}) \cdot \int_{\Omega} \varphi_j \varphi_i d\boldsymbol{x} + \frac{1}{2} \sum_{N_i \in \mathcal{N}_b} (\xi_{i,n} + \xi_{i,n-1}) \cdot \int_{\Omega} \boldsymbol{f'}(u) \nabla \varphi_j \varphi_i d\boldsymbol{x} = 0 \quad \forall i$$

Now we can establish the M mass matrix from the values of $\int_{\Omega} \varphi_j \varphi_i$ and C convection matrix with values from $\int_{\Omega} \mathbf{f'}(u) \nabla \varphi_j \varphi_i$. Using these matrices we can write the equation of FEM with implicit Crank-Nicholson time-discretization in the following form, which solution can be implemented easily:

$$\frac{1}{k_n} M(\xi_n - \xi_{n-1}) + \frac{1}{2} C(\xi_n + \xi_{n-1}) = 0$$

Notes on the implementation: For the calculation of mass and convection matrix I follow the instructions given in the book *The Finite Element Method:* Theory, Implementation, and Applications

The plots of the final solutions at T=1 with $hmax \in \{\frac{1}{8}, \frac{1}{16}\}$ can be found in Figure 1. As it can be seen the problem is a conservation problem, so the solution if the initial data in the given case. It can be seen that solution contains some noise for both meshes over time for both mesh. This means that even if we start from the exact solution, over time some noise occurs in the numerical solution of the equation even with finer meshes. Now let's move to the next part, where we investigate the size of the generated error.

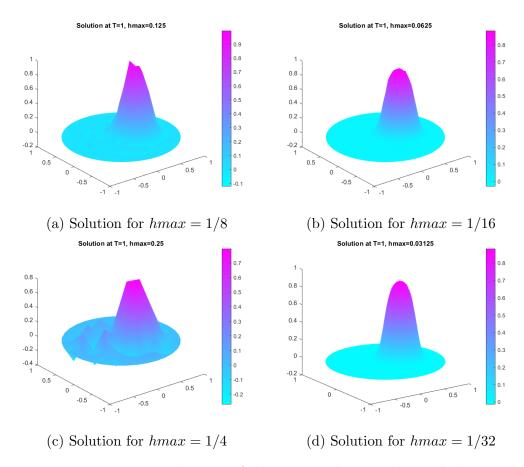


Figure 1: FEM solutions of the PDE with smooth initial data

Problem 1.2

For calculating the error, we need to know the exact solution. As I mentioned in this case that is the initial data. So the error can be calculated $e = u - U_n = u_0 - U_n \approx U_0 - U_n =: e_n$. As it is given in the assignment the norm of the error can be calculated in the following way in the *n*th time step: $||e||_{L^2(\Omega)} = \left(\int_{\Omega} e^2 dx\right)^{\frac{1}{2}} \approx \sqrt{e_n^T M e_n}$, where M is the mass matrix.

Now we should add this to the Matlab implementation and run the simulation for $hmax \in \{\frac{1}{4}, \frac{1}{8}, \frac{1}{16}, \frac{1}{32}, \frac{1}{64}\}$ and calculate the α convergence rate. The errors of the solution can be seen in Figure 2 for the smallest and largest mesh size. As it can be seen the error is getting less with using smaller mesh size.

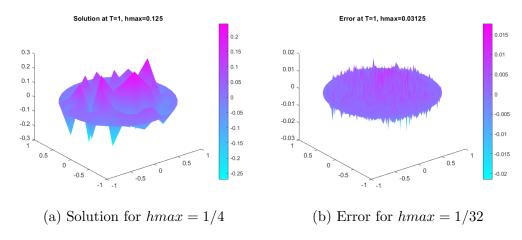


Figure 2: FEM, error of the PDE at T=1, smooth initial data

The $\alpha=1.18$ convergence rate parameter is calculated based on the error norms and a linear interpolation of the log values. The log-log plot of the errors can be seen in Figure 3 together with the h_{max} versus h_{max}^{α} plot.

Actually with sufficient parameter it should be possible to reach second order convergence for the FEM algorithm with the smooth initial data. To investigate this I modified the time stepping parameter CFL and changed it to 0.1 from 0.5. After this I got a much better convergence rate with $\alpha = 1.74$, which can be seen in Figure 4. This is much closer to the expected second order convergence.

(Note: with the same change there wasn't any significant improvement on the

 α convergence rate with the non-smooth initial data).

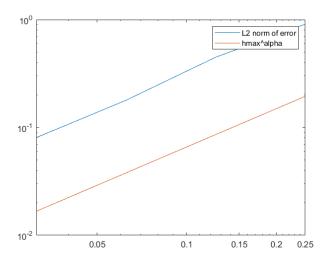


Figure 3: FEM $\alpha = 1.18$ convergence rate with smooth initial data

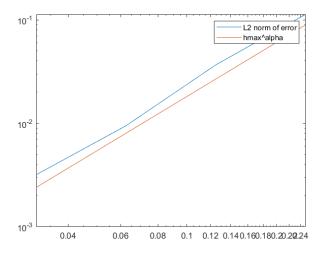


Figure 4: FEM $\alpha=1.74$ convergence rate, smooth initial data, CFL=0.1

Problem 1.3

In this part the task was to solve the same equation as in the previous two parts with different initial data.

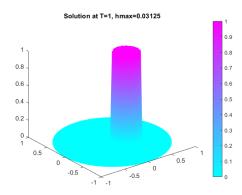


Figure 5: The non-smooth initial data

The theoretical formulation of the problems work the same way with different u_0 data. So the only thing to complete this task was to change the initial data in the MATLAB code and create similar figures as before.

Now the initial data is set to:

$$u_0(\boldsymbol{x}, 0) = \begin{cases} 1 & \text{if } (x_1 - x_1^0)^2 + (x_2 - x_2^0)^2 \le r_0^2 \\ 0 & \text{otherwise} \end{cases}$$

So now the initial data is non-smooth and therefore the simulation is expected to result in worse results. As for a reference the initial data is plotted in Figure 5.

The plots for the solutions at the final time (T=1) with different mesh sizes can be seen in Figure 6. As it could be expected these results are more noisy then the ones in the case of the smooth initial data. But as the mesh is getting finer, the solution is getting better and better. To measure this phenomenon now let's look at the error plots and the convergence rate.

The errors for the smallest and largest mesh is plotted in Figure 7. The shape of the error is similar to the previous case, but the scale is much larger as it

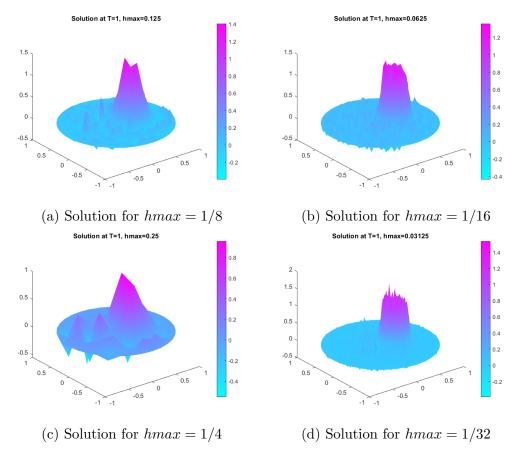


Figure 6: FEM solutions of the PDE, non-smooth initial data

can be expected. Now let's look at the convergence rate plotted in Figure 8. The calculated convergence rate is $\alpha=0.28$, which is much less, than in the previous case. So as it was expected from the numerical solutions we also get that the convergence rate is much worse in this case as in the smooth version of the problem.

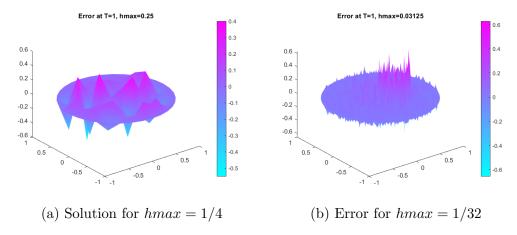


Figure 7: FEM, error of the PDE at T=1, non-smooth initial data

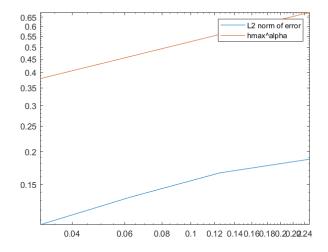


Figure 8: FEM $\alpha = 0.28$ convergence rate for the non-smooth initial data

Part II

In part 2, we will investigate methods for stabilizing the solution for the same problem as the one in part 1. To this end the SUPG and RV method for this problem will be introdued and implemented for both smooth and non smooth initial data.

Problem 2.1

First we have to create the SUPG formulation of the original problem:

$$\partial_t u + \nabla \cdot \boldsymbol{f}(u) = 0, \qquad (\boldsymbol{x}, t) \in \Omega \times (0, T],$$

 $u(\boldsymbol{x}, 0) = u_0(\boldsymbol{x}), \qquad \boldsymbol{x} \in \Omega,$

Reformulate the problem with the given information $\nabla f(u) = f'(u) \cdot \nabla u$.

$$\partial_t u + \mathbf{f'}(u) \cdot \nabla u = 0, \qquad (\mathbf{x}, t) \in \Omega \times (0, T],$$

 $u(\mathbf{x}, 0) = u_0(\mathbf{x}), \qquad \mathbf{x} \in \Omega,$

Now we can get the SUPG formulation of the problem by testing with $v + \delta f'(v) \cdot \nabla v$. Find $u_h \in X_h$ s.t.:

$$(\partial_t u_h + \boldsymbol{f'}(u_h) \cdot \nabla u_h, v) + (\partial_t u_h + \boldsymbol{f'}(u_h) \cdot \nabla u_h, \delta \boldsymbol{f'}(v) \cdot \nabla v) = 0, \quad \forall v \in X_h \quad (2)$$

Note: Here $f'(u_h)$ only depends on the position x, therefore it is a linear PDE, even though in this form $f'(u_h)$ seems to be dependent on the solution. So using the implicit Cranck-Nicholson time-stepping, we get the following formulation:

$$\frac{1}{k}(U_n - U_{n-1}, v + \delta \mathbf{f'}(v) \cdot \nabla u) + \frac{1}{2}(\mathbf{f'}(U_n) \cdot \nabla U_n + \mathbf{f'}(U_{n-1}) \cdot \nabla U_{n-1}, v + \delta \mathbf{f'}(v) \cdot \nabla v) = 0$$

To prove stability we will use the form written in (2) and use v = u to make stability estimates.

$$\frac{1}{2}\partial_t ||u_h||^2 + ||\sqrt{\delta} \boldsymbol{f'}(u_h)\nabla u_h||^2 + (\boldsymbol{f'}(u_h)\nabla u_h, u_h) + (\partial_t u_h, \delta \boldsymbol{f'}(u_h)\nabla u_h) = 0$$

With the notations of the lectures we get that since $\overline{\beta} = \mathbf{f'}$ is divergence free in this case as $\frac{x_2}{\partial_{x_1}} = \frac{x_1}{\partial_{x_2}} = 0$. So $\nabla \cdot \mathbf{f'} = 0$. Therefore the same prove works

as in the lecture to show that $(\mathbf{f'} \cdot \nabla u_h, u_h) = 0$ as it was given in the lecture for $(\overline{\beta} \cdot \nabla u_h, u_h)$ with the homogeneous Dirichlet boundary condition.

For the other term we can show the same thing using the fully discrete form. So in the fully discrete form we get $(\frac{U_n-U_{n-1}}{k_n}, \delta f'(x)\nabla \frac{U_n+U_{n-1}}{2}) = 0$.

Therefore in the fully discrete form we showed stability by gaining $\frac{1}{2}\partial_t ||U_h||^2 + ||\sqrt{\delta} \mathbf{f'}(x)\delta U_h|| = 0$ in the fully discrete equivalent.

Therefore we showed stability.

Problem 2.2 - 2.3

The SUPG (Streamline-Upwind-Petrov-Galerkin) method

First let's discuss and implement the SUPG method. From the previous discussion we obtain the following discretized system to be implemented:

$$\xi_{n+1} \left(\frac{M}{k_n} + \frac{C}{2} + \frac{SM}{k_n} + \frac{SC}{2} \right) = \xi_n \left(\frac{M}{k_n} - \frac{C}{2} + \frac{SM}{k_n} - \frac{SC}{2} \right)$$

Here M and C are the mass and convection matrices from part 1. $SM_{ij} = (\varphi_j, \delta \mathbf{f'}(N_i) \nabla \varphi_i)$, and $SC_{ij} = (\mathbf{f'}(N_j) \nabla \varphi_j, \delta \mathbf{f'}(N_i) \nabla \varphi_i)$. Here $\mathbf{f'}(N_i)$ is used to show that the value of $\mathbf{f'}$ is only space dependent in this case, and N_i denotes the node of the triangulation corresponding to the φ_i basis functions location. From here the implementation is straightforward using discrete integration to calculating the matrices.

The plots for the smooth initial data can be found in Figures 9, 10 and 11. While the plots for the non-smooth initial data can be found in Figures 12,13, 14. As from the plots and error it can be seen the constant 1 initial data parts stays 0 and there is basically no noise to it only close to the "interesting" part of the function. Also the non-smooth data act much better in this case than with the original FEM. This is not surprising as this method is stable. Also we get higher convergence rates with $\alpha = 1.82$ and $\alpha = 0.29$ for the two cases. This means that the SUPG method is also superior to basic FEM in the rate of convergence.

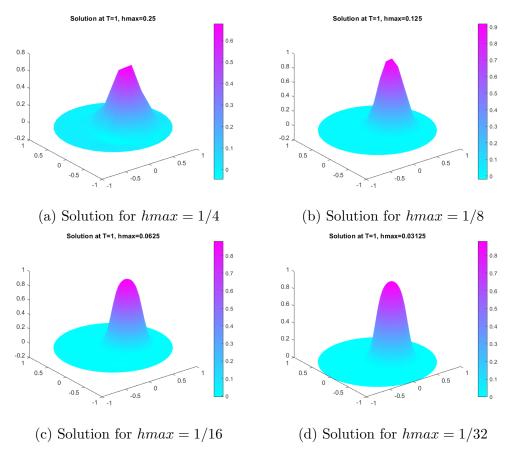


Figure 9: SUPG solutions of the PDE, smooth initial data

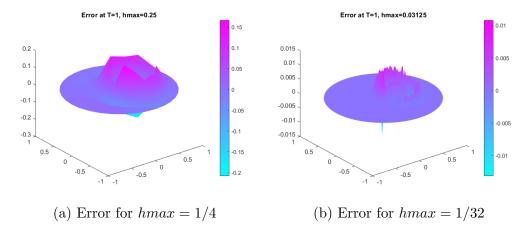


Figure 10: SUPG error of the PDE, smooth initial data

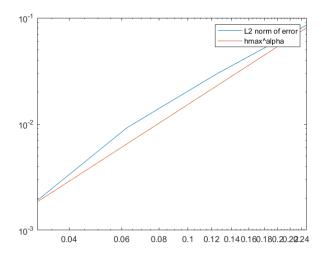


Figure 11: $\alpha=1.82$ convergence rate with SUPG method, smooth initial data

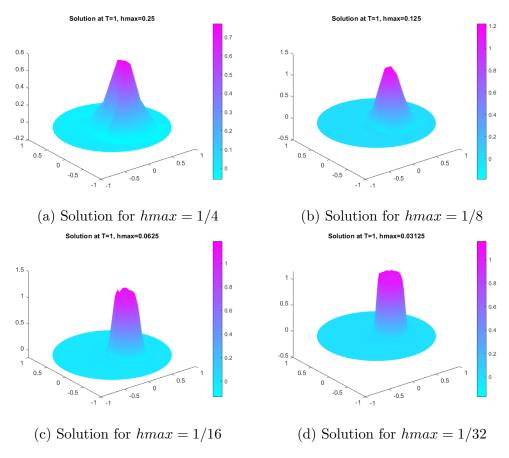


Figure 12: SUPG solutions of the PDE, non-smooth initial data

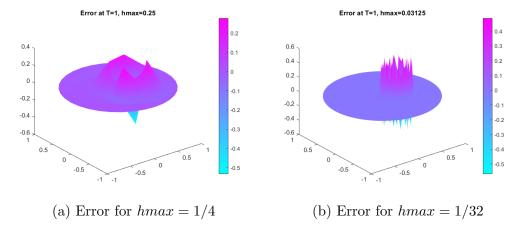


Figure 13: SUPG error of the PDE, non-smooth initial data

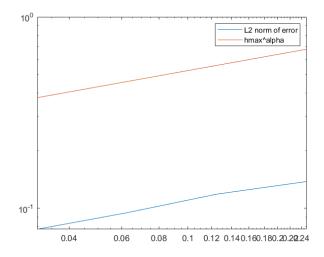


Figure 14: $\alpha=0.29$ convergence rate with SUPG method, non-smooth initial data

The RV (Residual based artificial viscosity) method

The RV method was given in the task description. In this case we add an additional term to the equation.

$$\frac{1}{k_n} (U_n - U_{n-1}, v) + \frac{1}{2} (\nabla \cdot (\boldsymbol{f}(U_{n-1}) + \boldsymbol{f}(U_n)), v)
+ \frac{1}{2} (\varepsilon_n (U_{n-1}) \nabla (U_n + U_{n-1}), \nabla v) = 0, \quad \forall v \in \mathcal{X}_h,$$

Here the artificial viscosity $\varepsilon_n := \varepsilon_h(t_n)$ is computed as

$$\varepsilon_{n,K} = \min \left(C_{\text{vel}} h_K \beta_K, C_{\text{RV}} h_K^2 \frac{\|R_{\text{RV}}\|_{\infty,K}}{\|U_n - \overline{U_n}\|_{\infty,\Omega}} \right),$$

for each cell $K \subset \mathcal{T}_h$. Here $R(U_n) = \frac{1}{k_n} (U_n - U_{n-1}) + \nabla \cdot \boldsymbol{f}(U_n)$, and $\|U_n - \overline{U_n}\|_{\infty,\Omega}$ is a normalization term, with $\overline{U_n}$ being the space average of the solution over Ω , h_K is the mesh-size of the element K, β_K denotes the local element wave speed that is computed as

$$\beta_K \equiv \|\mathbf{f}'(U_n)\|_{L^{\infty}(K)} = \max_{N_i \in K, i=0,1,2} \left(\left[(f_1'(U_n))^2 + (f_2'(U_n))^2 \right]^{\frac{1}{2}} (N_i) \right)$$

, where N_i , i=0,1,2 is the nodes of the element K, $C_{\rm vel}=0.25$ and $C_{\rm RV}=1$ are stabilization parameters. The standard Galerkin solution can be obtained by setting $C_{\rm vel}=0$.

From these given information it is straightforward to do the implementation. All we need is one function to calculate $\epsilon_{n,K}$ and one to calculate the matrix for the new term.

Note: From (R_u, v) the value of R_u can be calculated as $R(U_n) = M^{-1}(M(\frac{U_n - U_{n-1}}{k_n}) + CU_n)$ based on the given form.

The plots for the smooth initial data can be found in Figures 15, 16 and 18. While the plots for the non-smooth initial data can be found in Figures 19,20, 22. In this case for both the smooth and non-smooth data the plots for small mesh sizes are worse than for FEM and SUPG, but as the mesh get finer, the results start to act nicely. Also the convergence rates are here

the bests with $\alpha = 1.97$ and $\alpha = 0.53$. The $\epsilon(u)$ values at the final time is also plotted in Figure 17 and 21. Though this plots are not accurate as ϵ is $\tau_h = \{K\} \mapsto \mathcal{R}$ and the plots are done at the nodal points with calculating an average from the neighbouring elements ϵ value.

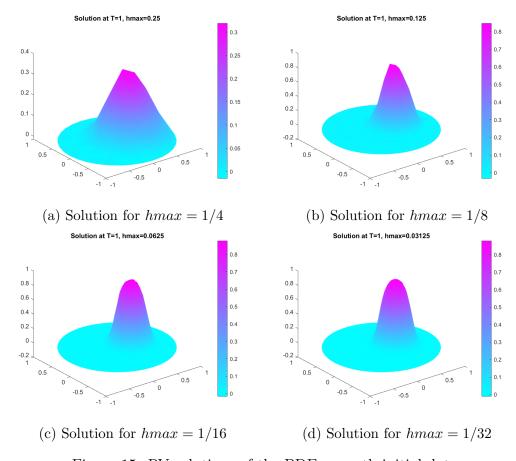


Figure 15: RV solutions of the PDE, smooth initial data

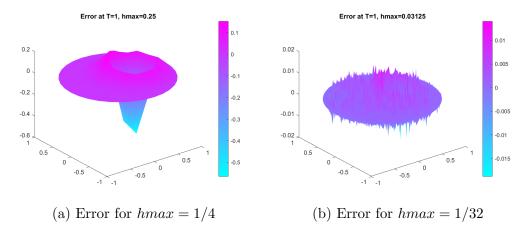


Figure 16: RV, error of the PDE, smooth initial data

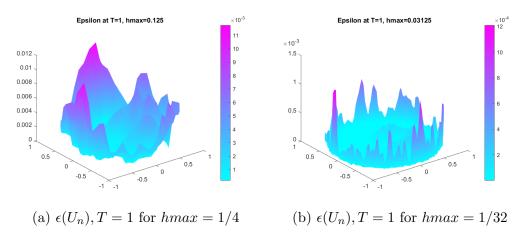


Figure 17: Value of $\epsilon(U)$ at the end of the RV simulation, smooth initial data

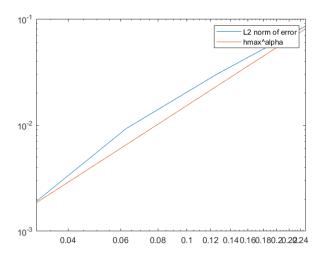


Figure 18: $\alpha = 1.97$ convergence rate with RV method, smooth initial data

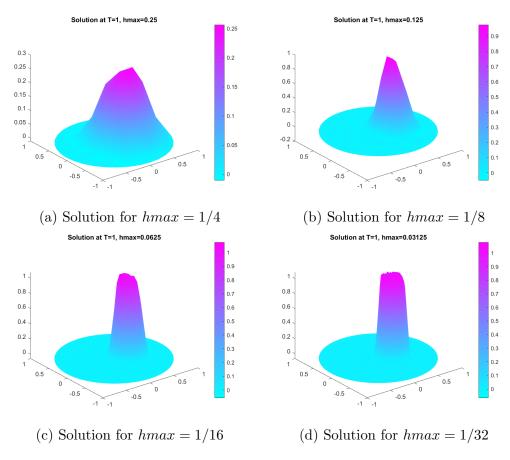


Figure 19: RV solutions of the PDE, non-smooth initial data

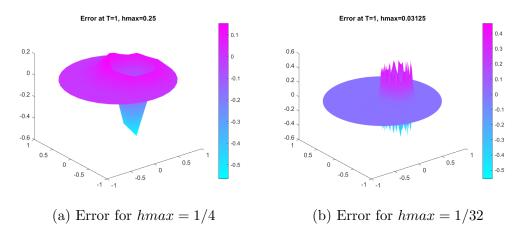


Figure 20: RV, error of the PDE, non-smooth initial data

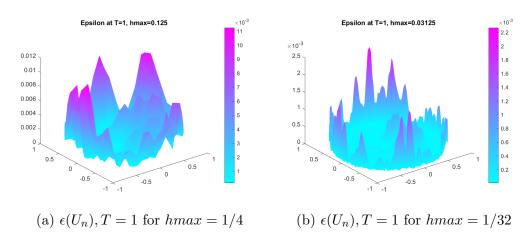


Figure 21: Value of $\epsilon(U)$ at the end of the RV simulation, non-smooth initial data

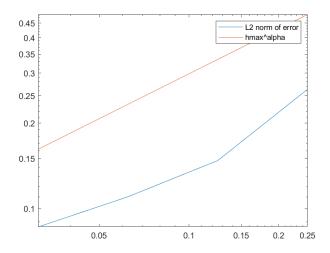


Figure 22: $\alpha=0.53$ convergence rate with RV method, non-smooth initial data

Summary

With this we concluded the analysis and the numerical modelling of the given tasks. We investigated and implemented the basic FEm, the SUPG and RV method to solve the problem. What we obtained is that the FEM works better with smooth initial data, than with the noncontinuous, resulting better measured convergence rate and much less error. These solutions are numerically not stable, therefore we wanted to use methods, where stability can be proven. We implemented the SUPG and RV methods, which are stable and results in better convergence rate and better solutions at least for the non-smooth data the results are much better with these methods, while for the smooth data the FEM worked just fine.

Appendix

Matlab code for parts 1.1 - 1.3: fem.m

```
1 % Given parameters
_{2} T=1;
_{3} CFL = 0.5;
4 \text{ r0} = 0.25;
  x1_{-}0 = 0.3;
  x2_0 = 0;
  %Used geometry
  g = @circleg;
  % Problem 1 - numerical fem solutions and error +
     convergence rate calculation
  %hmax values to run the simulation for
  hmax_list = [1/4, 1/8, 1/16, 1/32];
  initial_data = [1, 2];
  %Iteration on possible hmax values
19
  for init = initial_data
  i = 1;
  for hmax = hmax_list
      %mesh data
      [p, e, t] = initmesh (g, 'hmax', hmax); \%
         function call to create mesh
      x1=p(1,:); % vector of x1 coordinates
      x2=p(2,:)'; % vector of x2 coordinates
26
      % Set up the matrices
      df = ddf(x1, x2)';
29
      M = MassAssembler2D(p, t);
      C = ConvectionAssembler2D(p,t,df(1,:),df(2,:));
31
```

```
%Initial setup
33
34
       if init == 1
35
       U_{-}0 = 1 / 2 * (1 - \tanh((((x_1 - x_1_{-}0).^2 + (x_2 - x_1_{-}0))))))
           x2_{-}0).^{2}) / (r0^{2}) - 1);
        elseif init==2
37
       U_{-}00 = ((x_1 - x_{1-}0).^2 + (x_2 - x_{2-}0).^2) < r_0^2;
38
       U_0 = double(U_00);
39
       end
40
41
       U=U_0;
42
43
       %Time integration Crank-Nicholson
44
       time = 0;
45
       \inf_{n} = \max(abs(df(1,:)) + abs(df(2,:)));
46
       kn = CFL * hmax/inf_norm*1.0; \% timstep - not
47
           depends on U<sub>n</sub> due to f'
48
        while time<T
            %Set up equation matrix / vector
50
            A = ((M/kn) + (C/2));
            b = ((M/kn) * U - (C/2) * U);
52
            %Boundary conditions
54
            I = eye(length(p));
            A(e(1,:),:) = I(e(1,:),:);
56
            b(e(1,:)) = 0;
57
58
            %Solve equation
59
            U = A \setminus b;
60
61
            time = time + kn;
62
       end
63
64
65
       %Plot solution
       figure()
67
       pdeplot (p, e, t, "XYData", U, 'Zdata', U);
```

```
title ("Solution at T="+T+", hmax="+hmax)
69
70
       %Error
71
       error=U-U_0;
72
       L2E(i) = sqrt(error'*M*error);
73
       figure()
74
       pdeplot (p ,[],t,'XYData',error,'ZData',error,'
75
          ColorBar', 'on')
       title ("Error at T="+T+", hmax="+hmax)
76
77
       i = i + 1;
78
   end
79
  % Plot Conv
  p = polyfit (log(hmax_list), log(L2E), 1);
  alpha = p(1)
  figure ()
  %xlabel ('hmax')
86 %ylabel('L2E')
  loglog ( hmax_list , L2E , 'DisplayName' , 'L2 norm of
      error')
  hold on
  loglog( hmax_list , hmax_list.^alpha , 'DisplayName' ,
      'hmax\^alpha')
  hold off
   legend show
92
   end
94
95
96
98
  % Assembler functions based on book (The Finite
      Element Method: Theory, Implementation, and
      101
```

```
% derivative f
   function df = ddf(x1, x2)
       df = 2 * pi * [-x2, x1];
104
   end
105
106
   % Convection Matrix Assembler
107
   function C = ConvectionAssembler2D(p,t,bx,by)
       np = size(p, 2);
109
       nt = size(t,2);
110
       C=sparse(np,np);
111
       for i=1:nt
112
            loc2glb=t(1:3,i);
113
            x=p(1,loc2glb);
114
            y=p(2,loc2glb);
115
            [area, b, c] = HatGradients(x, y);
116
            bxmid=mean(bx(loc2glb));
117
            bymid=mean(by(loc2glb));
            CK=ones(3,1)*(bxmid*b+bymid*c)'*area/3;
119
            C(loc2glb, loc2glb) = C(loc2glb, loc2glb) + CK;
       end
121
   end
122
123
   %Mass Matrix Assembler
   function M = MassAssembler2D(p, t)
125
       np = size(p,2); % number of nodes
126
       nt = size(t,2); \% number of elements
127
       M = \text{sparse}(np, np); \% \text{ allocate mass matrix}
128
       for K = 1:nt % loop over elements
129
            loc2glb = t(1:3,K); \% local-to-global map
130
            x = p(1, loc2glb); % node x-coordinates
131
            y = p(2, loc2glb); \% y
132
            area = polyarea(x,y); \% triangle area
133
            MK = [2, 1, 1; 1, 2, 1; 1, 1, 2]/12*area; \%
134
               element mass matrix
            M(loc2glb, loc2glb) = M(loc2glb, loc2glb) + MK; \%
135
               add element masses to M
       end
136
137 end
```

```
\begin{array}{lll} ^{138} \\ ^{139} \ \% Gradients \\ ^{140} \ function \ [area\,,b\,,c\,] = HatGradients(x\,,y) \\ ^{141} \ area = polyarea\,(x\,,y)\,; \\ ^{142} \ b = [y\,(2\,) - y\,(3\,)\,; \ y\,(3\,) - y\,(1\,)\,; \ y\,(1\,) - y\,(2\,)\,]/2/\,area\,; \\ ^{143} \ c = [x\,(3\,) - x\,(2\,)\,; \ x\,(1\,) - x\,(3\,)\,; \ x\,(2\,) - x\,(1\,)\,]/2/\,area\,; \\ ^{144} \ end \end{array}
```

Matlab code for SUPG method: SUPG.m

```
1 % Given parameters
 T=1;
_{3} CFL = 0.5;
r0 = 0.25;
  x1_0 = 0.3;
  x2_0 = 0;
  %Used geometry
  g = @circleg;
  % Problem 1 - numerical fem solutions and error +
     convergence rate calculation
  % max values to run the simulation for
  hmax_list = [1/4, 1/8, 1/16, 1/32];
  initial_data = [1, 2];
  %Iteration on possible hmax values
  for init = initial_data
  i = 1:
  for hmax = hmax_list
      %mesh data
      delta = 0.05*hmax;
      [p, e, t] = initmesh (g, 'hmax', hmax); %
26
         function call to create mesh
      x1=p(1,:)'; % vector of x1 coordinates
      x2=p(2,:)'; % vector of x2 coordinates
29
      % Set up the matrices
      df = ddf(x1, x2)';
31
      M = MassAssembler2D(p, t);
32
      C = ConvectionAssembler2D(p,t,df(1,:),df(2,:));
      SC = SCAssembler2D(p,t,df(1,:),df(2,:),delta);
34
```

```
SM = delta*C';
35
36
        %Initial setup
37
        if init==1
39
        U_{-}0 = 1 / 2 * (1 - \tanh((((x_1 - x_1_{-}0).^2 + (x_2 - x_1_{-}0))))))
40
            x2_{-}0).^{2} / (r0^{2}) - 1);
        elseif init==2
41
        U_{-}00 \ = \ (\,(\,x1 \ - \ x1_{-}0\,)\,.\,\,\hat{}\ 2 \ + \ (\,x2 \ - \ x2_{-}0\,)\,.\,\,\hat{}\ 2\,) \ <\! r0\,\,\,\hat{}\ 2\,;
42
        U_0 = double(U_00);
43
        end
44
45
        U=U_0;
46
47
        %Time integration Crank-Nicholson
48
        time = 0;
49
        \inf_{n} = \max(abs(df(1,:)) + abs(df(2,:)));
50
        kn = CFL * hmax/inf_norm*1.0; \% timstep - not
51
            depends on U<sub>n</sub> due to f'
52
        while time<T
             %Set up equatddion matrix / vector
54
             A = ((M/kn) + (C/2) + (SM/kn) + (SC/2));
             b = ((M/kn) - (C/2) + (SM/kn) - (SC/2)) * U;
56
57
             %Boundary conditions
58
             I = eye(length(p));
59
             A(e(1,:),:) = I(e(1,:),:);
60
             b(e(1,:)) = 0;
61
62
             %Solve equation
63
             U = A \setminus b;
64
65
             time = time + kn;
66
        end
67
69
        %Plot solution
70
```

```
figure()
71
       pdeplot (p, e, t, "XYData", U, 'Zdata', U);
72
       title ("Solution at T="+T+", hmax="+hmax)
73
       %Error
75
       error=U-U_0;
76
       L2E(i)=sqrt(error'*M*error);
       figure()
78
       pdeplot (p ,[],t,'XYData',error,'ZData',error,'
79
          ColorBar', 'on')
       title ("Error at T="+T+", hmax="+hmax)
80
       i=i+1;
   end
83
  % Plot Conv
   p = polyfit(log(hmax_list), log(L2E), 1);
   alpha = p(1)
  figure ()
  %xlabel ('hmax')
90 %ylabel('L2E')
  loglog ( hmax_list , L2E , 'DisplayName' , 'L2 norm of
      error')
   hold on
   loglog( hmax_list , hmax_list.^alpha , 'DisplayName' ,
      'hmax\^alpha')
   hold off
   legend show
   end
98
99
100
101
102
  % Assembler functions based on book (The Finite
      Element Method: Theory, Implementation, and
```



```
105
   % derivative f
106
   function df = ddf(x1, x2)
       df = 2 * pi * [-x2, x1];
108
   end
109
   % Convection Matrix Assembler
111
   function C = ConvectionAssembler2D(p,t,bx,by)
       np = size(p,2);
113
       nt = size(t,2);
114
       C=sparse(np,np);
115
       for i=1:nt
116
            loc2glb=t(1:3,i);
117
            x=p(1,loc2glb);
118
            y=p(2,loc2glb);
119
            [area, b, c] = HatGradients(x, y);
            bxmid=mean(bx(loc2glb));
121
            by mid = mean(by(loc2glb));
122
            CK=ones(3,1)*(bxmid*b+bymid*c)'*area/3;
123
            C(loc2glb, loc2glb) = C(loc2glb, loc2glb) + CK;
       end
125
   end
126
127
   Mass Matrix Assembler
   function M = MassAssembler2D(p,t)
       np = size(p,2); \% number of nodes
130
       nt = size(t,2); \% number of elements
131
       M = sparse(np, np); \% allocate mass matrix
132
       for K = 1:nt % loop over elements
133
            loc2glb = t(1:3,K); \% local-to-global map
134
            x = p(1, loc2glb); \% node x-coordinates
135
            y = p(2, loc2glb); \% y
136
            area = polyarea(x,y); \% triangle area
137
            MK = [2, 1, 1; 1, 2, 1; 1, 1, 2]/12*area; \%
138
               element mass matrix
            M(loc2glb, loc2glb) = M(loc2glb, loc2glb) + MK; \%
139
               add element masses to M
```

```
end
140
  end
141
  % SC matrix assembly
  function A = SCAssembler2D(p,t,bx, by, delta)
  np = size(p,2);
  nt = size(t,2);
  A = sparse(np, np); % allocate stiffness matrix
   for K = 1:nt
  loc2glb = t(1:3,K); \% local-to-global map
  x = p(1, loc2glb); \% node x-coordinates
  y = p(2, loc2glb); \% node y-
  [area, b, c] = HatGradients(x, y);
  bxmid=mean(bx(loc2glb));
  bymid=mean(by(loc2glb));
  AK = delta*(bxmid^2*b*b'+bymid^2*c*c'+bxmid*bymid*(b*c')
      '+c*b'))*area; % element stiffness matrix
  A(loc2glb, loc2glb) = A(loc2glb, loc2glb) + AK; % add
      element stiffnesses to A
   end
157
   end
158
  %Gradients
   function [area, b, c] = HatGradients(x, y)
  area = polyarea(x, y);
  b=[y(2)-y(3); y(3)-y(1); y(1)-y(2)]/2/area;
  c = [x(3)-x(2); x(1)-x(3); x(2)-x(1)]/2/area;
  end
```

Matlab code for RV method: RV.m

```
1 % Given parameters
_{2} T=1;
_{3} CFL = 0.5;
r0 = 0.25;
5 x1_0 = 0.3;
6 x2_0 = 0;
_{7} C_vel=0.25;
 C_RV=1;
9
 %Used geometry
  g = @circleg;
13
14
  % Problem 1 - numerical fem solutions and error +
     convergence rate calculation
  %hmax values to run the simulation for
  hmax_list = [1/4, 1/8, 1/16, 1/32];
  initial_data = [1,2];
  %Iteration on possible hmax values
  for init = initial_data
  i = 1; L2E = [];
  for hmax = hmax_list
      disp ("hmax: "+hmax);
      %mesh data
27
      [p ,e , t ] = initmesh (g , 'hmax' , hmax); %
28
          function call to create mesh
      x1=p(1,:)'; % vector of x1 coordinates
30
      x2=p(2,:)'; % vector of x2 coordinates
31
32
      % Set up the matrices
      df = ddf(x1, x2)';
34
```

```
M = MassAssembler2D(p, t);
35
       C = ConvectionAssembler2D(p,t,df(1,:),df(2,:));
36
37
       nt = size(t,2); hk = []; bk = [];
       for K = 1:nt \% for each element
39
            loc2glb = t(1:3,K); \% local-to-global map
40
            x = p(1, loc2glb); % node x-coordinates
41
            y = p(2, loc2glb); \% node y-coordinates
42
            hk(K) = min([norm([x(1)-x(2),y(1)-y(2)]),norm([x
43
                (2)-x(3),y(2)-y(3)), norm ([x(3)-x(1),y(3)-y
                (1)), 2*norm([x(1)-sum(x)/3,y(1)-sum(y)/3])
                ); % diameter
            dx=df(1,loc2glb);
44
            dy=df(2,loc2glb);
45
            bk(K) = max([norm([dx(1), dy(1)]), norm([dx(2), dy(1)]))
46
                (2)), norm ([dx(3), dy(3)]);
       end
48
       %Initial setup
50
       if init == 1
       U_{-}0 = 1 / 2 * (1 - \tanh((((x_1 - x_1_{-}0).^2 + (x_2 - x_1_{-}0))))))
52
           x2_{-}0).^{2} / (r0^{2}) - 1);
        elseif init==2
53
       U_{-}00 = ((x_1 - x_{1-}0).^2 + (x_2 - x_{2-}0).^2) < r_0^2;
54
       U_0 = double(U_00);
55
       end
56
57
       U=U_0;
58
       U_old=U_0;
59
60
       %Time integration Crank-Nicholson
61
       time = 0;
62
       \inf_{\text{norm}} = \max(\text{abs}(\text{df}(1,:)) + \text{abs}(\text{df}(2,:)));
63
       kn = CFL * hmax/inf_norm*1.0; \% timstep - not
64
           depends on U<sub>n</sub> due to f'
65
       while time<T
66
```

```
%disp(time);
67
            R = ResidualAssembler(p,t, bk, hk, U, U_old, M,
                C, kn, C_{vel}, C_{RV};
            %Set up equatddion matrix / vector
            A=((M/kn) + (C/2) + (R/2));
70
            b = ((M/kn) - (C/2) - (R/2)) * U;
71
72
            %Boundary conditions
73
            I = eye(length(p));
74
            A(e(1,:),:) = I(e(1,:),:);
75
            b(e(1,:)) = 0;
76
77
            %Solve equation
78
            U_{-}old = U;
79
            U = A \setminus b;
80
81
            time = time + kn;
       end
83
85
       %Plot solution
86
       plt=figure();
87
       pdeplot (p, e, t, "XYData", U, 'Zdata', U);
        title ("Solution at T="+T+", hmax="+hmax)
89
       saveas (plt, "RV_plot_"+init+"_"+hmax+".png")
90
91
       %Error
92
       error=U-U_0;
93
       L2E(i)=sqrt(error'*M*error);
94
       err=figure();
       pdeplot (p ,[],t,'XYData',error,'ZData',error,'
96
           ColorBar', 'on')
        title ("Error at T="+T+", hmax="+hmax)
97
       saveas(err,"RV_err_"+init+"_"+hmax+".png")
98
99
       %Plot epsilon U_n
101
       np = size(p,2);
102
```

```
nt = size(t,2);
103
       eps = zeros(np); % allocate matrix
       occurances = zeros(np);
105
       RU=M\setminus (1/kn*M*(U-U_old)+C*U);
       for K = 1:nt \% for each element
107
       loc2glb = t(1:3,K); \% local-to-global map
108
       x = p(1, loc2glb); \% node x-coordinates
109
       y = p(2, loc2glb); \% node y-coordinates
110
       [area,b,c] = HatGradients(x,y);
111
       h_k = hk(K);
112
       b_k = bk(K);
113
       U_{norm}=\max(abs(U_{mean}(U)));
114
       Rk=max(abs(RU(loc2glb)));
115
       ep=min([C_vel*h_k*b_k,C_RV*h_k^2*Rk/U_norm]);
116
       eps(loc2glb) = eps(loc2glb) + 1/3*ep*[1;1;1]; \% add
117
           element stiffnesses to A
       occurances (loc2glb(1)) = occurances (loc2glb(1)) + 1;
       occurances (loc2glb(2)) = occurances (loc2glb(2)) + 1;
119
       occurances (loc2glb(3)) = occurances (loc2glb(3)) + 1;
       end
121
       eps=eps./occurances;
122
       pdeplot (p ,[],t,'XYData',eps,'ZData',eps,'ColorBar
123
           ', 'on')
        title ("Epsilon at T="+T+", hmax="+hmax)
124
       saveas (err, "RV_eps_"+init+"_"+hmax+".png")
125
126
       i = i + 1;
127
   end
128
129
   % Plot Conv
   p = polyfit(log(hmax_list), log(L2E), 1);
   alpha = p(1)
  alp=figure ();
  %xlabel ('hmax')
  %ylabel('L2E')
   loglog ( hmax_list , L2E , 'DisplayName' , 'L2 norm of
      error')
  hold on
137
```

```
loglog( hmax_list , hmax_list.^alpha , 'DisplayName' ,
      'hmax\^alpha')
   hold off
   legend show
140
       saveas(alp , "RV_conv_"+init +".png")
141
142
   end
143
144
145
146
147
148
  M Assembler functions based on book (The Finite
      Element Method: Theory, Implementation, and
      151
152
  % Convection Matrix Assembler
153
   function C = ConvectionAssembler2D(p, t, bx, by)
154
       np = size(p,2);
155
       nt = size(t,2);
156
       C=sparse(np,np);
157
       for i=1:nt
158
            loc2glb=t(1:3,i);
159
            x=p(1,loc2glb);
160
            y=p(2,loc2glb);
161
            [area, b, c] = HatGradients(x, y);
162
            bxmid=mean(bx(loc2glb));
163
            bymid=mean(by(loc2glb));
164
           CK=ones(3,1)*(bxmid*b+bymid*c)'*area/3;
165
           C(loc2glb, loc2glb) = C(loc2glb, loc2glb) + CK;
166
       end
167
   end
168
169
  %Mass Matrix Assembler
   function M = MassAssembler2D(p,t)
       np = size(p,2); \% number of nodes
172
```

```
nt = size(t, 2); \% number of elements
173
       M = sparse(np, np); \% allocate mass matrix
174
       for K = 1:nt \% loop over elements
175
            loc2glb = t(1:3,K); \% local-to-global map
            x = p(1, loc2glb); \% node x-coordinates
177
            y = p(2, loc2glb); \% y
            area = polyarea(x,y); \% triangle area
179
           MK = [2, 1, 1; 1, 2, 1; 1, 1, 2]/12*area; \%
180
               element mass matrix
           M(loc2glb, loc2glb) = M(loc2glb, loc2glb) + MK; \%
181
               add element masses to M
       end
182
   end
184
  % SC matrix assembly
   function R = ResidualAssembler(p,t, bk,hk, U, U_old, M,
       C, kn, C_{vel}, C_{RV}
   np = size(p,2);
   nt = size(t,2);
  R = sparse(np, np); \% allocate matrix
  RU=M\setminus (1/kn*M*(U-U_old)+C*U);
   for K = 1:nt % for each element
   loc2glb = t(1:3,K); \% local-to-global map
   x = p(1, loc2glb); \% node x-coordinates
   y = p(2, loc2glb); \% node y-coordinates
   [area,b,c] = HatGradients(x,y);
196
   h_k = hk(K);
   b_k=bk(K);
   U_{norm}=\max(abs(U_{mean}(U)));
   Rk=max(abs(RU(loc2glb)));
200
   e=min([C\_vel*h\_k*b\_k,C\_RV*h\_k^2*Rk/U\_norm]);
  AK = e*(b*b'+c*c')*area; \% element stiffness matrix
  R(loc2glb, loc2glb) = R(loc2glb, loc2glb) + AK; % add
      element stiffnesses to A
   end
205
  end
206
```

```
% Gradients % Gradients (x,y) function [area,b,c] = HatGradients(x,y) area=polyarea(x,y); b=[y(2)-y(3); y(3)-y(1); y(1)-y(2)]/2/area; c=[x(3)-x(2); x(1)-x(3); x(2)-x(1)]/2/area; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end df = 2 * pi * end condition df = ddf(x1, x2) end condition df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi * [-x2, x1]; end condition df = ddf(x1, x2) df = 2 * pi
```