# FEM4CFD Notes

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## Contents

1	1D	Advection Diffusion Equation	1
	1.1	Governing Equation	1
	1.2	Weak Form	2
	1.3	Galerkin Approximation	2
		1.3.1 Implementation	į
	1.4	Petrov-Galerkin Method	7
		1.4.1 Implementation	Ć
	1.5	Stream-Upwind Petrov-Galerkin Method	10
		1.5.1 Implementation	10
	1.6	Postivity Preserving Variational Method	11

## 1 1D Advection Diffusion Equation

## 1.1 Governing Equation

The governing equation of the advection diffusion reaction is of type (1). The equation models the transport phenomena including all three advection, diffusion and reaction. In the equation (2),  $\mathbf{F}$  and  $\mathbf{G}$  represent the advection and diffusion coefficients, while the term  $\mathbf{Q}$  represents either a reaction or source term. The advection diffusion equation is of the type

$$\frac{\partial \Phi}{\partial t} + \frac{\partial \mathbf{F_i}}{\partial x_i} + \frac{\partial \mathbf{G_i}}{\partial x_i} + \mathbf{Q} = 0 \tag{1}$$

$$F_{i} = F_{i}(\Phi)$$

$$G_{i} = G_{i} \frac{\partial \Phi}{\partial x_{j}}$$

$$Q = Q(x_{i}, \Phi)$$
(2)

where in general,  $\Phi$  is a basic dependent vector variable. A linear reaction between the source term and the scalar variable is referred to as the reaction term. In (2),  $x_i$  and i refer to Cartesian coordinates and the associated quantities and as a whole.

Thus, the equation in scalar terms becomes

$$\Phi \to \phi,$$
  $\mathbf{Q} \to Q(x_i, \phi) = s\phi$   
 $\mathbf{F_i} \to F_i = a\phi,$   $\mathbf{G_i} \to G_i = -k\frac{\partial \phi}{\partial x}$  (3)

$$\frac{\partial \phi}{\partial t} + \frac{\partial (a\phi)}{\partial x_i} - \frac{\partial}{\partial x_i} \left( k \frac{\partial \phi}{\partial x_i} \right) + Q = 0 \tag{4}$$

Here, U is the velocity field and  $\phi$  is the scalar quantity being transported by this velocity. But, diffusion can also occur, and k is the diffusion coefficient.

A linear reaction term can be written associated, where c is a scalar parameter.

$$Q = c \ \phi$$

Here, a is the velocity field and  $\phi$  is the scalar quantity being transported by this velocity. But, diffusion can also occur, and k is the diffusion coefficient. A linear reaction term can be associated, where c is a scalar parameter. The equation represented by (4) is the strong form or the differential form of the advection diffusion governing equation. For the steady state solution, the first term becomes zero, leaving only the advection, diffusion and reaction terms.

### 1.2 Weak Form

The use of 4 requires computation of second derivatives to solve the problem, as such a weakened form can be considered by solving the equation over a domain  $\Omega$  using an integral, like

$$\int_{\Omega} w \left( a \cdot \frac{\partial \phi}{\partial x} \right) d\Omega - \int_{\Omega} w \frac{d}{dx} \left( k \frac{d\phi}{dx} \right) + \int_{\Omega} w Q = 0$$
 (5)

where w is an arbitrary weighting function, chosen such that w = 0 on Dirichlet boundary condition,  $\Gamma_D$ . Also at the same Dirichlet boundary condition, the variable  $\phi = \phi_D$ . Assuming a source term, s is present, the right hand side of the above equation changes resulting in the weak form the governing equation.

$$\int_{\Omega} w \left( a \cdot \nabla \phi \right) d\Omega - \int_{\Omega} w \nabla \cdot (k \nabla \phi) d\Omega + \int_{\Omega} w Q d\Omega = \int_{\Omega} w s d\Omega \tag{6}$$

Noting that, w = 0 on  $\Gamma_D$ , using divergence theore, we get

$$\int_{\Omega} w \left( a \cdot \nabla \phi \right) d\Omega - \int_{\Omega} \nabla w \cdot (k \nabla \phi) d\Omega + \int_{\Omega} w Q d\Omega = \int_{\Omega} w s d\Omega + \int_{\Gamma_N} w h d\Gamma \tag{7}$$

where  $\Gamma_N$  and h represent the Neumann boundary condition and the normal diffusive flux on the Neumann boundary condition.

## 1.3 Galerkin Approximation

The Galerkin approximation is a technique used to approximate numerical solutions of PDEs by replacing the infinite-dimensional spaces into finite dimensional spaces. The finite spaces are constructed using finite elements over a domain. Since spaces are finite-dimensional, the weighting function is a discrete weighting function  $w_h$ .

Using Galerkin approximation,  $\phi$  can be written

$$\phi(x) = N_1(x)\phi_1 + N_2(x)\phi_2 + \dots + N_n(x)\phi_n$$

$$\phi(x) = \sum_{i=1}^{n_{el}} N_i(x)\phi_i$$
(8)

Here,  $N_i$  is the shape function or basis function at that node,  $n_{el}$  is the total number of elements and  $\phi$  is the solution, also known as degrees of freedom (DOFs). For Galerkin approximation, the weighting function is equal to the shape function i.e.,  $w_i = N_i$ . Now, equation 7 becomes:

$$\int_{\Omega} a \left( N_a \frac{\partial N_b}{\partial N_t} \right) d\Omega + \int_{\Omega} k \left( \frac{\partial N_a}{\partial t} \frac{\partial N_b}{\partial t} \right) d\Omega \int_{\Omega} Q \left( N_a \cdot N_b \right) d\Omega = \int_{\Omega} f N_a d\Omega + \int_{\Gamma} h N_a d\Gamma \qquad (9)$$

Here, in (9) f represents the source term and for problems with only Dirichlet boundary conditions, the second term on right hand vanishes, effectively giving us the weak form of the Galerkin approximatin of the ADR equation.

Solving the equation for cell Pe = 0.1, where Pe is defined by  $Pe = \frac{a \cdot he}{2k}$  with 10 linear elements, with the domain length 1 and (0, 1) Dirichlet boundary conditions on the left and the right of the domain respectively. The results of the comparison of the Galerkin solution with the analytical solution 11 without a source or a reaction term can be seen in Figure 1a.

### 1.3.1 Implementation

The following section explores the implementation of the Galerkin approximation for the one dimensional advection diffusion equation in MATLAB.

#### 1. Initialization:

```
%% 1D steady state advection diffusion
 2
   clear;
   clc;
   close all;
   xL = 0;
   xR = 1;
   nelem = 10;
9
10
   L = xR - xL;
11
   he = L / nelem;
12
13
   % boundary conditions
14
   uL = 0;
15
   uR = 1;
16
17
   Pe = 0.1;
   mu = 1;
   c = Pe * (2 * mu) / he;
20 \mid f = 0:
```

```
21
22
   nGP = 2;
23
   [gpts, gwts] = get_Gausspoints_1D(nGP);
24
25
   nnode = nelem + 1;
26
27
   ndof = 1;
28
29
   totaldof = nnode * ndof;
30
   node_coords = linspace(xL, xR, nnode);
32
33
   elem_node_conn = [1:nelem; 2:nnode]';
34
   elem_dof_conn = elem_node_conn;
35
36
   dofs_full = 1:totaldof;
37
   dofs_fixed = [1, totaldof];
   dofs_free = setdiff(dofs_full, dofs_fixed);
38
39
40
  % solution array
   soln_full = zeros(totaldof, 1);
```

The first section of the code is dedicated for the initial boundary values, domain properties and initializing the solution arrays for calculation. he is the elemental length, while L is the length of the domain. xL and xR are the left and the right boundaries of the domain and nelem is the number of elements the domain will be discretized into, totaldof is the total degrees of freedom of the entire system, and  $soln_full$  is the final solution array i.e.,  $\phi(x)$ . This solution utilizes Gaussian points for numerical integration over an element. For the purpose of this solution, two Gausspoints (nGP) are considered (linear element) with xi and wt being  $\pm \frac{1}{\sqrt{(3)}}$  and 1.0, respectively.

#### 2. Processing:

```
1
   %% Processing
2
   for iter = 1:9
3
       Kglobal_g = zeros(totaldof, totaldof);
4
       Fglobal_g = zeros(totaldof, 1);
6
7
       for elnum = 1:nelem
           elem_dofs = elem_dof_conn(elnum, :);
8
           Klocal = zeros(2, 2);
9
           Flocal = zeros(2, 1);
10
11
12
           %% Galerkin Approximation
13
           [Klocal, Flocal] = galerkinApproximation(c, mu, he, s,
              nGP, gpts, gwts, elem_dofs, node_coords, soln_full,
              Klocal, Flocal);
```

```
14
15
            Kglobal_g(elem_dofs, elem_dofs) = Kglobal_g(elem_dofs,
               elem_dofs) + Klocal;
            Fglobal_g(elem_dofs, 1) = Fglobal_g(elem_dofs, 1) +
16
               Flocal;
17
       end
18
19
       Fglobal_g = forceVector(Kglobal_g, Fglobal_g, iter, uL, uR,
          totaldof);
20
21
       rNorm = norm(Fglobal_g);
22
23
       if (rNorm < 1.0e-10)
24
            break;
25
       end
26
27
       Kglobal_g = stiffnessMatrix(Kglobal_g, totaldof);
28
29
       soln_incr = Kglobal_g \ Fglobal_g;
30
       soln_full = soln_full + soln_incr;
31
32
   end
```

This solution uses an iterative solver to solve the advection diffusion equation. The solver iteratives over the number of elements and calls the function *galerkinApproximation* on each iteration to calculate the terms of the global stiffness matrix. The global stiffness matrix is a combination of the advection, diffusion and reaction contribution to the global matrix. These are:

$$K_{ad} = \frac{a}{2} \begin{bmatrix} -1 & 1 \\ -1 & 1 \end{bmatrix}, K_{diff} = \frac{k}{he} \begin{bmatrix} 1 & -1 \\ -1 & 1 \end{bmatrix}, K_{re} = \frac{s \cdot he}{6} \begin{bmatrix} 2 & 1 \\ 1 & 2 \end{bmatrix}$$
 (10)

```
1
   %% Galerkin Approximation function
   function [Klocal, Flocal] = galerkinApproximation(a, mu, h, s,
      nGP, gpts, gwts, elem_dofs, node_coords, soln_full, Klocal,
      Flocal)
3
       Klocal_g = zeros(2, 2);
       Flocal_g = zeror(2, 1);
4
5
6
       n1 = elem_dofs(n1);
7
       n2 = elem_dofs(n2);
8
9
       u1 = soln_full(n1);
       u2 = soln_full(n2);
10
11
       u = [u1 \ u2];
       for gp = 1:nGP
12
13
               xi = gpts(gp);
14
               wt = gwts(gp);
               N = [0.5 * (1 - xi), 0.5 * (1 + xi)];
15
```

```
dNdxi = [-0.5, 0.5];
16
17
               Jac = h / 2;
18
               dNdx = dNdxi / Jac;
19
               du = dNdx * u';
20
               x = N * [x1 x2];
21
22
               % advection
23
               Klocal = Klocal + (a * N' * dNdx) * Jac * wt;
24
               % reaction
25
               Klocal = Klocal + (s * N' * N) * Jac * wt;
               % diffusion
26
27
               Klocal = Klocal + (mu * dNdx' * dNdx) * Jac * wt:
28
29
               % force vector
30
               Flocal = Flocal + N' * f * Jac * wt;
31
       end
32
33
               Klocal_g = Klocal_g + Klocal;
34
               Flocal_g = Flocal_g + Flocal;
35
   end
```

Once the global stiffness matrix and the force vectors are assembled, the force Vector function is called. The force Vector function sets the first and the last elements of the force Vector as the left and the right boundary conditions (for this case i.e., Dirichlet BCs). Similar to the force Vector function, the stiffness Matrix function also sets the boundary conditions for the stiffness matrix. The stiffness Matrix subroutine changes the first element in the first row and the first column of the stiffness matrix to 1 and every other element to 0. This enforces the Dirichlet boundary condition.

```
%% Force vector assembly
2
   function Fglobal = forceVector(Kglobal, Fglobal, iter, uL, uR,
      totaldof)
3
4
          if iter == 1
               Fglobal = Fglobal - Kglobal(:, 1) * uL;
5
6
               Fglobal = Fglobal - Kglobal(:, totaldof) * uR;
7
               Fglobal(1, 1) = uL;
8
               Fglobal(end, 1) = uR;
9
          else
               Fglobal(1, 1) = 0.0;
10
11
               Fglobal(end, 1) = 0.0;
12
          end
13
14
   end
```

```
Kglobal(:, 1) = zeros(totaldof, 1);
Kglobal(1, 1) = 1.0;

Kglobal(end, :) = zeros(totaldof, 1);
Kglobal(:, end) = zeros(totaldof, 1);
Kglobal(:, end) = 1.0;
end
```

It can be seen in Figure 1a that the Galerkin approximation for the one dimensional advection diffusion equation closely follows the analytical solution. However, at higher Peclet numbers, or due to a finer mesh, spurious oscillations are introduced, as seen in Figure 1b. 11 represents the analytical solution for this case of the advection diffusion reaction equation.

$$\phi(x) = \frac{e^{axk} - 1}{e^{Pe} - 1} \tag{11}$$

Spurious oscillations in the Galerkin approximation of the advection diffusion reaction equation

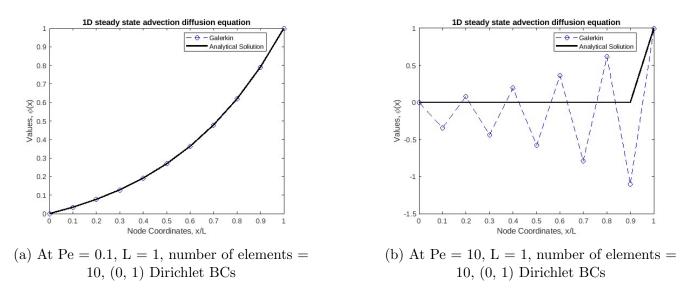


Figure 1: Galerkin Approximation solution

occurs primarily in advection dominated regimes. The standard Galerkin approximation struggles to resolve the layers in advection dominated regions (a >> k), which causes steep gradients and non-physical oscillations. There are several ways to mitigate these oscillations. 3 are presented in the following sections: Petrov-Galerkin (PG), Stream-Upwind Petrov-Galerkin (SUPG), and Positivity Preserving Variational method (PPV). Compared to PG and SUPG methods, PPV is used to preserve the non-linear positivity condition, making sure discontinuity in the solution are captured and corrected.

### 1.4 Petroy-Galerkin Method

The Galerkin method introduces several problems i.e., sharp gradients and spurious oscillations, in advection dominated problems. These problems or defeciencies can be cured in two ways: one, adding a diffusion term to balance the numerical diffusion introduced by the Galerkin approximation, and two, an upwind approximation of the convective term of the Galerkin solution. Early

solutions to the problems in Galerkin method were solved using these two philosophies; however both methodologies are actually equivalent, in that an upwind approximation induces numerical diffusion.

The finite element method approximates the advection term with a second order central difference method. Numerical diffusion, however can be introduced by replacing the central differencing method by a first-order upwind method, for a > 0:

$$u_x(x_j) \approx \frac{u_j - u_{j-1}}{h}$$

$$a\frac{u_j u_{j-1}}{h} - k\frac{u_{j+1} - 2u_j + u_{j-1}}{h^2} = 0$$
(12)

Using Taylor expansion for the advection term, an added diffusion of magnitude  $\frac{ah}{2}$  is introduced by the upwind approximation. However this upwind treatment of the ADR equation leads to a stable but highly diffusive solution. The Petrov-Galerkin method also utilizes a weighting function that is different than the standard Galerkin method in which  $W_a \neq N_a$ . For the Petrov-Galerkin method the weighting function becomes

$$W_a = N_a + \alpha W_a^*$$

$$\alpha \text{ or } \alpha_{opt} = \coth|Pe| - \frac{1}{|Pe|}$$
(13)

This value of  $\alpha$  is chosen to give the exact nodal values for all values of Pe. But, it can be observed from Figure 2 that the solution is over diffusive and deviates from the analytical solution of the governing equation.

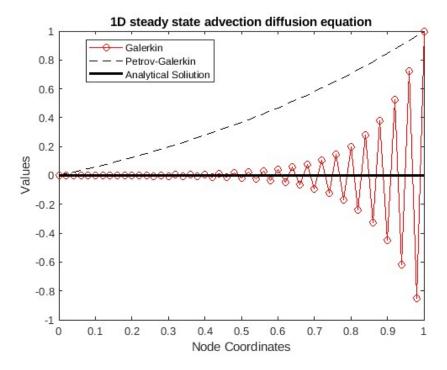


Figure 2: For Pe = 5 with no source

#### 1.4.1 Implementation

```
function [Klocal_pg, Flocal_pg] = petrovGalerkin(a, mu, h, alpha,
      nGP, gpts, gwts, elem_dofs, node_coords, soln_full_pg, Klocal,
      Flocala)
2
           Klocal = zeros(2, 2);
3
           Flocal = zeros(2, 1);
4
5
           n1 = elem_dofs(1);
           n2 = elem_dofs(2);
6
7
8
           u1 = soln_full_pg(n1);
           u2 = soln_full_pg(n2);
9
10
11
           for gp = 1:nGP
12
                    xi = gpts(gp);
13
                    wt = gwts(gp);
14
15
                    N = [0.5 * (1 - xi), 0.5 * (1 + xi)];
                    dNdxi = [-0.5, 0.5];
16
17
                    Jac = h / 2;
18
19
20
                    dNdx = dNdxi / Jac;
21
22
                    % advection
23
                    Klocal = Klocal + (alpha * a * dNdx' * dNdx) * Jac *
                        wt:
24
25
                    % diffusion
26
                    Flocal = Flocal - (alpha * a * dNdx' * du) * Jac *
                       wt;
27
                    end
28
            Klocal_pg = Klocal_pg + Klocal;
29
           Flocal_pg = Flocal_pg + Flocal;
30
31
   end
```

This is the subroutine for the Petrov-Galerkin implementation. To implement this correctly, these lines must be added to the pre-processing and the main loop of the main function, respectively.

Fglocal\_pg = zeros(totaldof, 1);

4 \% inside the element loop

## 1.5 Stream-Upwind Petrov-Galerkin Method

The SUPG method is a stabilization technique used to stabilize the convective term in a consistent manner. This ensures the solution of the differential equation is similar to the solution of the weak form of the governing equation. A standard stabilization technique adds an extra term to the Galerkin weak form, and this term is a function of the residual of the weak form. We know the residual of the ADR equation is

$$\mathcal{R}(u) = a \cdot \nabla \phi - \nabla \cdot (\nu \nabla \phi) + \sigma \phi - f = \mathcal{L}(u) - f \tag{14}$$

Here  $\mathcal{L}$  is the differential operator. Now, the standard form of stabilization techniques is

$$a \cdot \nabla \phi - \nabla \cdot (\nu \nabla \phi) + \sigma \phi + \sum_{e} \int_{\Omega} \mathcal{P}(w) \tau \mathcal{R}(u) d\Omega = f$$
 (15)

where  $\tau$  is the stabilization term and  $\mathcal{P}(w)$  is the operator applied to the test function. For the SUPG method, the terms are:

$$\mathcal{P}(w) = a \cdot \nabla w$$

$$\tau = \frac{he}{2a} \left( \coth(Pe) - \frac{1}{Pe} \right) \tag{16}$$

#### 1.5.1 Implementation

```
function [Klocal_supg, Flocal_supg] = supg(a, mu, h, alpha, tau, s,
     nGP, gpts, gwts, elem_dofs, node_coords, soln_full_supg, Klocal,
     Flocal)
   Klocal_supg = zeros(2, 2);
   Flocal_supg = zeros(2, 1);
4
   n1 = elem_dofs(1);
   n2 = elem_dofs(2);
6
   x1 = node_coords(n1);
9
   x2 = node_coords(n2);
10
11
   u1 = soln_full_supg(n1);
   u2 = soln_full_supg(n2);
   u = [u1 \ u2];
13
14
  for gp = 1:nGP
           xi = gpts(gp);
16
```

```
17
           wt = gwts(gp);
18
           N = [0.5 * (1 - xi), 0.5 * (1 + xi)];
19
20
            dNdxi = [-0.5, 0.5];
21
22
            Jac = h / 2;
23
24
            dNdx = dNdxi / Jac;
25
           du = dNdx * u';
           uh = N * u';
26
           x = N * [x1 x2]';
27
28
29
           mod_test = a * dNdx';
31
           % stabilization
32
            Klocal = Klocal + tau * (mod_test * (a * dNdx + s * N) * Jac
                * wt):
33
            % force vector
34
            res = a * du + s * uh - f; % residual
            % Flocal = Flocal + tau * a * dNdx' * f * Jac * wt;
36
           Flocal = Flocal + tau * (mod_test * f) * Jac * wt;
           Flocal = Flocal - tau * dNdx' * a ^ 2 * du * Jac * wt;
38
   end
39
40
   Klocal_supg = Klocal_supg + Klocal;
41
   Flocal_supg = Flocal_supg + Flocal;
42
43
   end
```

This is the subroutine for the SUPG implementation. Similar to the Petrov-Galerkin method, certain variables in the same manner must be initialized before the iteration and the element loop. As seen in Figure 3, the solution to the ADR equation in production regime (s > 1), and even with Pe > 1, the solution is dissipated from its Galerkin approximation solution. The linear stabilization term introduces diffusion and helps stabilize the solution in higher Peclet number cases, but oscillations and sharp gradients still persist. The case in Figure 3 is with Pe = 10 and Da = 10, where Da is the Damkohler number and is responsible for the production (s > 0) or destruction (s < 0) regime. The cases from this point onward will include the Da number as part of the problem.

## 1.6 Postivity Preserving Variational Method

To reduce the presence of sharp gradients and oscillations in the solution due to discontinuities, a positivity preserving method is introduced. The PPV method adds a non-linear stabilization term to the weak form of the Galerkin approximation, which is added only when the gradient of the solution is steeper than a certain desired value. It makes sure the solution follows the natural condition and remains within the bounds of the actual solution of the differential equation.

A positivity condition is enforced in the element stiffness matrix i.e.,

$$k_{ij} = k_{ji} = max(0, a_{ij}, a_{ji})$$
 (17)

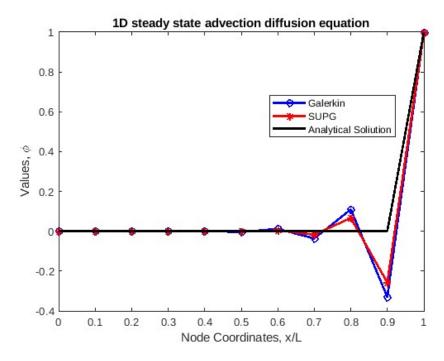


Figure 3: For Pe = 10, Da = 10, (1, 0)

where,  $a_{ij}$  is an element of the positivity preserving matrix A and  $k_{ij}$  is an element of the stiffness matrix. Even though stabilization methods like SUPG and PG reduce the oscillations in the solution, there still exist overshoots and undershoots near regions of high gradient, becauseneither method is monotone or positivity preserving. Natural physical solutions must follow the sign of their neighbour while excluding oscillations.