#### 1. Finite Element Discretization

### 1.1. Discretizing the Domain

Discretize the domain [0, L] into E finite elements with nodes  $x_i$ , where i = 1, 2, ..., n and n is the total number of nodes.

### 1.2. Approximating the Solution and Test Functions

Approximate the solution u(x,t) and the test function v(x) using finite element basis functions  $N_i(x)$ :

$$u(x,t) \approx \sum_{j=1}^{n} U_j(t) N_j(x), \quad v(x) = N_i(x)$$

$$\tag{1}$$

Here,  $U_j(t)$  are the time-dependent coefficients (nodal values) of the solution, and  $N_j(x)$  are the shape functions associated with each node.

### 1.3. Substituting into the Weak Form

Substitute these approximations into the weak form and simplify:

$$\sum_{j=1}^{n} \frac{dU_{j}(t)}{dt} \int_{0}^{L} N_{j}(x) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} U(x,t) \frac{\partial N_{j}(x)}{\partial x} N_{i}(x) dx + \nu \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} \frac{\partial N_{i}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(x) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx = \int_{0}^{L} f(x,t) N_{i}(t) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx dx = \int_{0}^{L} f(x,t) N_{i}(t) dx + \sum_{j=1}^{n} U_{j}(t) \int_{0}^{L} \frac{\partial N_{j}(x)}{\partial x} dx dx = \int_{0$$

### 1.4. Defining the Matrices

This can be rewritten in matrix form:

• Mass Matrix M:

$$\mathbf{M}_{ij} = \int_0^L N_i(x) N_j(x) dx \tag{3}$$

• Convection Term C(U):

$$\mathbf{C}_{ij}(\mathbf{U}) = \int_0^L U(x,t) \frac{\partial N_j(x)}{\partial x} N_i(x) dx \tag{4}$$

• Diffusion Matrix K:

$$\mathbf{K}_{ij} = \nu \int_{0}^{L} \frac{\partial N_{i}(x)}{\partial x} \frac{\partial N_{j}(x)}{\partial x} dx \tag{5}$$

• Load Vector F:

$$\mathbf{F}_{i}(t) = \int_{0}^{L} f(x,t)N_{i}(x)dx \tag{6}$$

### 1.5. Discretized System of Equations

The overall system of equations is:

$$\mathbf{M}\frac{d\mathbf{U}(t)}{dt} + \mathbf{C}(\mathbf{U})\mathbf{U} + \mathbf{K}\mathbf{U} = \mathbf{F}(t)$$
(7)

where:

•  $\mathbf{U}(t)$  is the vector of unknowns at the nodes.

# 1.6. Time Discretization

To solve this system over time, you apply a time discretization method such as the implicit Euler scheme:

$$\mathbf{M}\frac{\mathbf{U}^{n+1} - \mathbf{U}^n}{\Delta t} + \mathbf{C}(\mathbf{U}^{n+1})\mathbf{U}^{n+1} + \mathbf{K}\mathbf{U}^{n+1} = \mathbf{F}^{n+1}$$
(8)

This forms a nonlinear system at each time step, which can be solved iteratively.

### 1.7. Solving the Nonlinear System

Once the time discretization is applied, we obtain a nonlinear system of equations at each time step, which can be written as:

$$\mathbf{A}(\mathbf{U}^{n+1})\mathbf{U}^{n+1} = \mathbf{b}^{n+1}$$

where:

$$\mathbf{A}(\mathbf{U}^{n+1}) = \mathbf{M} + \Delta t \mathbf{C}(\mathbf{U}^{n+1}) + \Delta t \mathbf{K}$$

$$\mathbf{b}^{n+1} = \mathbf{M}\mathbf{U}^n + \Delta t \mathbf{F}^{n+1}$$

## 1.7.1. Picard Iteration (Fixed-Point Iteration)

To solve this nonlinear system, we can use the Picard iteration method, which is a fixed-point iterative scheme. The steps are as follows:

- 1. **Initial Guess**: Start with an initial guess  $\mathbf{U}_0^{n+1}$ , which is usually taken as the solution from the previous time step,  $\mathbf{U}^n$ .
- 2. **Iterative Update**: Update the solution using the current approximation:

$$\mathbf{A}(\mathbf{U}_k^{n+1})\mathbf{U}_{k+1}^{n+1} = \mathbf{b}^{n+1}$$

3. Convergence Check: Compute the error between successive iterations:

$$error_{U} = \frac{\|\mathbf{U}_{k+1}^{n+1} - \mathbf{U}_{k}^{n+1}\|}{\|\mathbf{U}_{k+1}^{n+1}\|}$$

4. **Stopping Criterion**: The iteration continues until the error falls below a specified tolerance, or a maximum number of iterations is reached.

This method iteratively solves the linearized system until convergence, gradually refining the approximation for  $\mathbf{U}^{n+1}$ .

**Note:** Newton-Raphson Method (Alternative Approach)

An alternative approach to solving the nonlinear system is the Newton-Raphson method, which involves computing the Jacobian matrix and performing a linearization around the current guess at each iteration. This method typically converges faster but requires the computation and inversion of the Jacobian, which can be computationally expensive.

### 1.7.2. Boundary Conditions

Boundary conditions are crucial in ensuring that the numerical solution adheres to the physical constraints of the problem.

Dirichlet Boundary Condition at x = 0. For the 1D Burgers' equation, the Dirichlet boundary condition is:

$$u(0,t) = \mu_1, \quad \forall t \in [0,T]$$

To enforce this in the FEM system:

• Modify the System Matrix A:

$$\mathbf{A}[0,:] = 0$$
 and  $\mathbf{A}_{00} = 1$ 

This sets the first row of **A** to ensure the solution at x=0 is fixed.

• Adjust the Right-Hand Side Vector b:

$$\mathbf{b}[0] = \mu_1$$

This forces the solution at the boundary node to be  $\mu_1$ .

Initial Condition. The initial condition is:

$$u(x,0) = 1, \quad \forall x \in [0,L]$$

This initializes the solution vector  $\mathbf{U}(0)$  at time t = 0.

## Summary.

- Matrix Modification: Set A[0,:] to 0 and  $A_{00} = 1$  to enforce Dirichlet conditions.
- Vector Adjustment: Set  $\mathbf{b}[0] = \mu_1$  to match the boundary condition.
- Initial Condition: Set U(0) based on the initial state of the system.

This ensures that the boundary and initial conditions are properly incorporated into the numerical solution.