

# 有限元方法

## Finite Element Methods

### Chapter 7: Galerkin Approach for Time-Dependent Problems

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# The Standard Galerkin Method for the Heat Equation



We consider the model heat equation:

## Heat equation

$$(P) \quad \begin{cases} u_t - \Delta u = f, & \text{in } \Omega, t > 0, \\ u = 0, & \text{on } \partial\Omega, \\ u(\cdot, 0) = v. \end{cases}$$

### Goals of the chapter:

- Introduce the standard Galerkin finite element method.
- Study spatially semidiscrete approximations.
- Construct fully discrete schemes in time.
- Derive optimal error estimates.

**Strategy:** PDE  $\implies$  weak formulation  $\implies$  finite-dimensional problem

### Main time discretizations:

- Backward Euler method; Crank–Nicolson method;
- Second-order backward difference method (BDF2).



# Sobolev Spaces and Norms

For  $\Omega \subset \mathbb{R}^d$ ,

$$L_2(\Omega) = \{v : \Omega \rightarrow \mathbb{R} \mid \|v\| < \infty\},$$

with norm

$$\|v\| = \left( \int_{\Omega} v^2 dx \right)^{1/2}.$$

For integer  $r \geq 1$ ,

$$H^r(\Omega) = W_2^r(\Omega),$$

with norm

$$\|v\|_r = \left( \sum_{|\alpha| \leq r} \|D^\alpha v\|^2 \right)^{1/2}.$$

Here

$$D^\alpha = \left( \frac{\partial}{\partial x_1} \right)^{\alpha_1} \cdots \left( \frac{\partial}{\partial x_d} \right)^{\alpha_d}, \quad |\alpha| = \sum_{j=1}^d \alpha_j.$$

For functions in  $H_0^1(\Omega)$ ,

$$c\|v\|_1 \leq \|\nabla v\| \leq \|v\|_1.$$

This follows from Friedrichs' inequality.  $\|\nabla v\|$  and  $\|v\|_1$  are equivalent norms.



# Weak Formulation of Poisson's Equation

Consider

$$-\Delta u = f \quad \text{in } \Omega, \quad u = 0 \quad \text{on } \partial\Omega.$$

Multiply by a test function  $\varphi \in H_0^1$  and integrate over  $\Omega$ :

$$-\int_{\Omega} \Delta u \varphi \, dx = \int_{\Omega} f \varphi \, dx.$$

Using Green's formula:

$$(\nabla u, \nabla \varphi) = (f, \varphi), \quad \forall \varphi \in H_0^1.$$

Inner products:

$$(v, w) = \int_{\Omega} vw \, dx, \quad (\nabla v, \nabla w) = \int_{\Omega} \sum_{j=1}^d \frac{\partial v}{\partial x_j} \frac{\partial w}{\partial x_j} \, dx.$$

A function  $u \in H_0^1$  satisfying  $(\nabla u, \nabla \varphi) = (f, \varphi)$  is called the variational solution.



# Elliptic Regularity

It is an easy consequence of the Riesz representation theorem that a unique such solution exists if  $f \in H^{-1}$ , the dual space of  $H_0^1$ .

If  $\partial\Omega$  is smooth, then the solution gains two derivatives:

$$\|u\|_{m+2} \leq C\|\Delta u\|_m = C\|f\|_m, \quad m \geq -1.$$

In particular:

$$f \in H^m \implies u \in H^{m+2}.$$

If  $f \in C^\infty$ , then  $u \in C^\infty$ .

Important consequence:

- Smooth right-hand side  $\implies$  smooth solution.
- Regularity plays a key role in error estimates.

For convex polygonal domains:  $\|u\|_2 \leq C\|f\|$ .

For nonconvex polygonal domains, this may fail.





# Triangulation and Finite Element Space

Let thus  $\Omega$  be a convex domain in the plane with smooth boundary  $\partial\Omega$ , and let  $\mathcal{T}_h$  be a triangulation of  $\Omega$ . Define

$$h = \max_{\tau \in \mathcal{T}_h} \text{diam}(\tau).$$

Assumptions:

- Angles are bounded below.
- Sometimes quasiuniformity is assumed:  $|\tau| \geq ch^2$ .

Finite element space:

$$S_h = \{ \chi \in C(\bar{\Omega}) : \chi|_{\tau} \text{ linear}, \chi = 0 \text{ on } \partial\Omega \}.$$

Let  $P_1, \dots, P_{N_h}$  be interior vertices. Define basis functions  $\Phi_j$ :  $\Phi_j(P_i) = \delta_{ij}$ . Then

$$\chi(x) = \sum_{j=1}^{N_h} \alpha_j \Phi_j(x).$$

The coefficients satisfy  $\alpha_j = \chi(P_j)$ .



# Interpolation Operator

For a smooth function  $v$ , define the interpolant

$$I_h v(x) = \sum_{j=1}^{N_h} v(P_j) \Phi_j(x).$$

Then  $I_h v(P_j) = v(P_j)$ .

Interpolation error estimates:

$$\|I_h v - v\| \leq Ch^2 \|v\|_2, \quad \|\nabla(I_h v - v)\| \leq Ch \|v\|_2.$$

More generally, when  $v \in H^s(\Omega) \cap H_0^1$ ,

$$\inf_{\chi \in S_h} \{\|v - \chi\| + h \|\nabla(v - \chi)\|\} \leq Ch^s \|v\|_s,$$

for  $1 \leq s \leq r$ .

The number  $r$  is referred to as the order of accuracy of the family  $\{S_h\}$ . An interpolation operator  $I_h : H^r \cap H_0^1 \rightarrow S_h$  such that

$$\|I_h v - v\| + h \|\nabla(I_h v - v)\| \leq Ch^s \|v\|_s, \quad \text{for } 1 \leq s \leq r.$$



# Inverse Inequality

For quasiuniform triangulations, finite element functions satisfy the inverse estimate:

$$\|\nabla \chi\| \leq Ch^{-1} \|\chi\|, \quad \forall \chi \in S_h.$$

Key idea of proof:

- Transform each element to a reference triangle.
- Use equivalence of norms on finite-dimensional spaces.
- Sum over all elements.

Interpretation:

Higher derivatives  $\sim h^{-1} \times$  lower derivatives.

This estimate is crucial in:

- stability analysis,
- fully discrete schemes,
- maximum norm estimates.



# Galerkin Approximation for Poisson's Equation

Find  $u_h \in S_h$  such that

$$(\nabla u_h, \nabla \chi) = (f, \chi), \quad \forall \chi \in S_h.$$

Galerkin orthogonality:

$$(\nabla(u_h - u), \nabla \chi) = 0, \quad \forall \chi \in S_h.$$

Using basis functions:

$$u_h(x) = \sum_{j=1}^{N_h} \alpha_j \Phi_j(x).$$

Then

$$\sum_{j=1}^{N_h} \alpha_j (\nabla \Phi_j, \nabla \Phi_k) = (f, \Phi_k), \quad k = 1, \dots, N_h.$$

Matrix form:

$$A\alpha = \tilde{f},$$

where  $a_{jk} = (\nabla \Phi_j, \nabla \Phi_k)$ .

$A$  is symmetric positive definite.



# Error Estimates for the Elliptic Problem

## Theorem (1.1)

Assume that

$$\inf_{\chi \in S_h} \{ \|u - \chi\| + h \|\nabla(u - \chi)\| \} \leq Ch^s \|u\|_s.$$

holds. Then, for  $1 \leq s \leq r$ ,

$$\|u_h - u\| \leq Ch^s \|u\|_s \quad \text{and} \quad \|\nabla u_h - \nabla u\| \leq Ch^{s-1} \|u\|_s.$$

- Energy norm estimate:

$$\|\nabla(u_h - u)\| \leq \inf_{\chi \in S_h} \|\nabla(\chi - u)\|.$$

- $L_2$ -estimate uses the duality argument: Let  $-\Delta\psi = \varphi$ . Then  
 $(u_h - u, \varphi) = (\nabla(u_h - u), \nabla(\psi - \psi_h))$ .

Apply:

- elliptic regularity,
- interpolation estimate,
- Galerkin orthogonality.



# Weak Solution of the Heat Equation

We consider the initial-boundary value problem

$$u_t - \Delta u = f \quad \text{in } \Omega, \quad t > 0, \quad u = 0 \quad \text{on } \partial\Omega, \quad t > 0, \quad u(\cdot, 0) = v.$$

We say that a function  $u = u(x, t)$  is a weak solution on  $[0, \bar{t}]$  if

$$(u_t, \varphi) + (\nabla u, \nabla \varphi) = (f, \varphi), \quad \forall \varphi \in H_0^1,$$

holds for  $t > 0$ , with

$$u \in L_2(0, \bar{t}; H_0^1), \quad u_t \in L_2(0, \bar{t}; H^{-1}),$$

and with initial value

$$u(\cdot, 0) = v.$$

Here

$$(v, w) = \int_{\Omega} vw \, dx, \quad (\nabla v, \nabla w) = \int_{\Omega} \sum_{j=1}^d \frac{\partial v}{\partial x_j} \frac{\partial w}{\partial x_j} \, dx.$$



# Parabolic Regularity Estimate

Since the boundary  $\partial\Omega$  is smooth, the weak solution is smooth provided the data are smooth.

This smoothness is expressed by the parabolic regularity estimate

$$\sum_{j=0}^{m+1} \int_0^{\bar{t}} \|u^{(j)}\|_{2(m-j)+2}^2 dt \leq C \left( \|v\|_{2m+1}^2 + \sum_{j=0}^m \int_0^{\bar{t}} \|f^{(j)}\|_{2(m-j)}^2 dt \right), \quad m \geq 0,$$

where

$$u^{(j)} = \left( \frac{\partial}{\partial t} \right)^j u, \quad C = C_{m, \bar{t}}.$$

Thus, sufficiently smooth initial data  $v$  and source term  $f$  yield higher regularity of both time derivatives and spatial derivatives of  $u$ .

In particular, the solution gains spatial smoothness in a way consistent with the elliptic regularity of the heat operator.



## Compatibility Conditions at $t = 0$

The regularity estimate above is valid only when the data satisfy suitable compatibility conditions at  $t = 0$ .

These conditions express that the boundary condition and the initial condition are consistent with each other.

**First compatibility condition ( $m = 0$ ):** since

$$u(t) = 0 \quad \text{on } \partial\Omega, \quad t > 0,$$

we must also have  $u(0) = v = 0$  on  $\partial\Omega$ .

**Second compatibility condition ( $m = 1$ ):** Since the boundary condition holds for all  $t > 0$ , differentiation in time implies  $u_t = 0$ , on  $\partial\Omega$ . Evaluating the PDE at  $t = 0$ ,

$$u_t(0) = \Delta v + f(0),$$

hence smoothness requires

$$\Delta v + f(0) = 0 \quad \text{on } \partial\Omega.$$

Similarly, for higher time derivatives  $u^{(m)}(0)$ , further compatibility conditions are needed for  $m \geq 2$ .



The above conditions are standard in the regularity theory for parabolic equations.

They ensure that:

- the prescribed initial value agrees with the boundary condition at  $t = 0$ ,
- higher time derivatives are also consistent with the boundary data,
- the solution enjoys the smoothness needed in later finite element error analysis.

Hence, under smoothness of  $\partial\Omega$ ,  $v$ , and  $f$ , together with the required compatibility conditions, the weak solution becomes sufficiently regular for the estimates used in the semidiscrete and fully discrete analyses.

# Semidiscrete Galerkin Method

## Spatially semidiscrete problem

Find  $u_h(t) \in S_h$  such that

$$(\mathcal{W}_h) \quad (u_{h,t}, \chi) + (\nabla u_h, \nabla \chi) = (f, \chi),$$

for all  $\chi \in S_h$ ,  $t > 0$ , with  $u_h(0) = v_h$ .

Expand:

$$u_h(x, t) = \sum_{j=1}^{N_h} \alpha_j(t) \Phi_j(x).$$

Then

$$\sum_{j=1}^{N_h} \alpha_j'(t) (\Phi_j, \Phi_k) + \sum_{j=1}^{N_h} \alpha_j(t) (\nabla \Phi_j, \nabla \Phi_k) = (f, \Phi_k).$$

Matrix form:

$$\mathcal{B} \alpha'(t) + \mathcal{A} \alpha(t) = \tilde{f}(t).$$

Mass and stiffness matrix:

$$b_{jk} = (\Phi_j, \Phi_k), \quad a_{jk} = (\nabla \Phi_j, \nabla \Phi_k).$$

Define the Ritz projection  $R_h : H_0^1 \rightarrow S_h$  by

$$(\nabla R_h v, \nabla \chi) = (\nabla v, \nabla \chi), \quad \forall \chi \in S_h.$$

Interpretation:

- $R_h v$  is the finite element approximation of  $v$ .
- Orthogonal projection in the energy inner product.

Error splitting:

$$u_h - u = \theta + \rho,$$

where  $\theta = u_h - R_h u$ ,  $\rho = R_h u - u$ .

Properties:

$$\|\nabla R_h v\| \leq \|\nabla v\|,$$

and

$$\|R_h v - v\| + h\|\nabla(R_h v - v)\| \leq Ch^s \|v\|_s.$$

## Theorem (1.2)

Let  $u_h$  and  $u$  be the solutions of  $(\mathcal{W}_h)$  and  $(\mathcal{P})$ , and assume  $v = 0$  on  $\partial\Omega$ . Then

$$\|u_h(t) - u(t)\| \leq \|v_h - v\| + Ch^r \left( \|v\|_r + \int_0^t \|u_t\|_r ds \right), \quad \text{for } t \geq 0.$$

Main idea:

$$u_h - u = (u_h - R_h u) + (R_h u - u) =: \theta + \rho.$$

The elliptic part:

$$\|\rho(t)\| \leq Ch^r \|u(t)\|_r.$$

For  $\theta$ :

$$(\theta_t, \chi) + (\nabla\theta, \nabla\chi) = -(\rho_t, \chi).$$

Choose  $\chi = \theta$ . Then

$$\frac{1}{2} \frac{d}{dt} \|\theta\|^2 + \|\nabla\theta\|^2 = -(\rho_t, \theta).$$

After integration:

$$\|\theta(t)\| \leq \|\theta(0)\| + \int_0^t \|\rho_t\| ds.$$



# Gradient Error Estimate

## Theorem (1.3)

Let  $u_h$  and  $u$  be the solutions of  $(\mathcal{W}_h)$  and  $(\mathcal{P})$ , and assume  $v = 0$  on  $\partial\Omega$ . Then

$$\begin{aligned} & \|\nabla u_h(t) - \nabla u(t)\| \\ & \leq C\|\nabla(v_h - v)\| + Ch^{r-1} \left( \|v\|_r + \|u(t)\|_r + \left( \int_0^t \|u_s\|_{r-1}^2 ds \right)^{1/2} \right). \end{aligned}$$

Key step:

Choose  $\chi = \theta_t$  in

$$(\theta_t, \chi) + (\nabla\theta, \nabla\chi) = -(\rho_t, \chi).$$

Then

$$\|\theta_t\|^2 + \frac{1}{2} \frac{d}{dt} \|\nabla\theta\|^2 = -(\rho_t, \theta_t).$$

Hence

$$\frac{d}{dt} \|\nabla\theta\|^2 \leq \|\rho_t\|^2.$$

Integrating in time yields the estimate.



# Superconvergence

Assume  $v_h = R_h v$ . Then  $\theta(0) = 0$ .

Using the previous estimate:

$$\|\nabla\theta(t)\| \leq C \left( \int_0^t \|\rho_t\|^2 ds \right)^{1/2}.$$

Since

$$\|\rho_t\| = \|R_h u_t - u_t\| \leq Ch^r \|u_t\|_r,$$

we obtain

$$\|\nabla\theta(t)\| \leq Ch^r \left( \int_0^t \|u_t\|_r^2 ds \right)^{1/2}.$$

Observation:  $\nabla u_h$  approximates  $\nabla R_h u$  with order  $O(h^r)$ , while  $\nabla u_h$  approximates  $\nabla u$  only with order  $O(h^{r-1})$ .

This phenomenon is called:

Superconvergence



# Maximum Norm Error Estimates

For piecewise linear elements:

$$\|I_h v - v\|_{L_\infty} \leq Ch^2 \|v\|_{W_\infty^2}.$$

## Theorem (1.4)

Let  $u_h$  and  $u$  be the solutions of  $(\mathcal{W}_h)$  and  $(\mathcal{P})$ , and assume  $v = 0$  on  $\partial\Omega$ . Then

$$\|u_h - u\|_{L_\infty} \leq Ch^2 \ell_h \|u\|_{W_\infty^2},$$

where  $\ell_h = \max(1, \log(1/h))$ .

Ritz projection stability:

$$\|R_h v\|_{L_\infty} \leq C \ell_h \|v\|_{L_\infty}.$$

Parabolic estimate:

$$\|u_h(t) - u(t)\|_{L_\infty} \leq C(t, u) h^2 \ell_h.$$

Important fact:

- The logarithmic factor  $\ell_h$  cannot be removed.
- Nevertheless, it grows very slowly.



# Backward Euler Galerkin Method

We introduce a time step  $k$  and the time levels  $t_n = nk$  ( $n \geq 0$ ), and denote by  $U^n = U_h^n \in S_h$  the approximation of  $u(t_n)$ .

Backward difference:

$$\bar{\partial}U^n = \frac{U^n - U^{n-1}}{k}.$$

## Fully Discrete Backward Euler

Find  $U^n \in S_h$  such that

$$(\mathcal{B}\mathcal{E}) \quad (\bar{\partial}U^n, \chi) + (\nabla U^n, \nabla \chi) = (f(t_n), \chi), \quad U^0 = v_h,$$

for all  $\chi \in S_h$ ,  $n \geq 1$ .

Equivalent formulation:

$$(U^n, \chi) + k(\nabla U^n, \nabla \chi) = (U^{n-1} + kf(t_n), \chi).$$

Matrix form:

$$(\mathcal{B} + k\mathcal{A})\alpha^n = \mathcal{B}\alpha^{n-1} + k\tilde{f}(t_n).$$

The matrix  $\mathcal{B} + k\mathcal{A}$  is symmetric positive definite.

# Error Estimate for Backward Euler

## Theorem (1.5)

With  $U^n$  and  $u$  the solutions of  $(\mathcal{BE})$  and  $(\mathcal{P})$ , respectively, we have, if

$\|v_h - v\| \leq Ch^r \|v\|_r$  and  $v = 0$  on  $\partial\Omega$ ,

$$\|U^n - u(t_n)\| \leq Ch^r \left( \|v\|_r + \int_0^{t_n} \|u_t\|_r ds \right) + k \int_0^{t_n} \|u_{tt}\| ds, \quad \text{for } n \geq 0.$$

Error decomposition:  $U^n - u(t_n) = \theta^n + \rho^n.$

Discrete error equation:

$$(\bar{\partial}\theta^n, \chi) + (\nabla\theta^n, \nabla\chi) = -(\omega^n, \chi).$$

Consistency error:

$$\omega^n = (R_h - I)\bar{\partial}u(t_n) + (\bar{\partial}u(t_n) - u_t(t_n)).$$

Main result:

$$\|U^n - u(t_n)\| = O(h^r + k)$$

Backward Euler: • first order in time, • unconditionally stable.



# Crank–Nicolson Galerkin Method

Define  $\widehat{U}^n = \frac{1}{2} (U^n + U^{n-1})$ .

## Fully Discrete Crank–Nicolson

Find  $U^n \in S_h$  such that

$$(\mathcal{CN}) \quad (\bar{\partial}U^n, \chi) + (\nabla\widehat{U}^n, \nabla\chi) = (f(t_{n-\frac{1}{2}}), \chi), \quad U^0 = v_h,$$

for all  $\chi \in S_h$ ,  $n \geq 1$ .

Matrix form:  $(\mathcal{B} + \frac{1}{2}k\mathcal{A})\alpha^n = (\mathcal{B} - \frac{1}{2}k\mathcal{A})\alpha^{n-1} + k\tilde{f}(t_{n-\frac{1}{2}})$ .

## Theorem (1.6)

Let  $U^n$  and  $u$  be the solutions of (CN) and (P), respectively, and let

$\|v_h - v\| \leq Ch^r \|v\|_r$  and  $v = 0$  on  $\partial\Omega$ . Then, for  $n \geq 0$ ,

$$\|U^n - u(t_n)\| \leq Ch^r \left( \|v\|_r + \int_0^{t_n} \|u_t\|_r ds \right) + Ck^2 \int_0^{t_n} (\|u_{ttt}\| + \|\Delta u_{tt}\|) ds.$$

Main result:

$$\|U^n - u(t_n)\| = O(h^r + k^2).$$



# Second Order Backward Difference Method

We define the second order backward difference operator:

$$\bar{D}U^n = \frac{\frac{3}{2}U^n - 2U^{n-1} + \frac{1}{2}U^{n-2}}{k} u_t(t_n) + O(k^2).$$

## Fully Discrete BDF2 Galerkin Method

Find  $U^n \in S_h$  such that

$$(\mathcal{BDF2}) \quad (\bar{D}U^n, \chi) + (\nabla U^n, \nabla \chi) = (f(t_n), \chi), \quad n \geq 2,$$

for all  $\chi \in S_h$ , with  $U^0 = v_h$ .

The first step  $U^1$  is computed by one backward Euler step:

$$(\bar{\partial}U^1, \chi) + (\nabla U^1, \nabla \chi) = (f(t_1), \chi).$$

Matrix form:

$$\left(\frac{3}{2}\mathcal{B} + k\mathcal{A}\right) \alpha^n = 2\mathcal{B}\alpha^{n-1} - \frac{1}{2}\mathcal{B}\alpha^{n-2} + k\tilde{f}(t_n), \quad n \geq 2.$$

## Theorem (1.7)

Let  $U^n$  and  $u$  be the solutions of (BDF2) and (P), with  $U^0 = v_h$  and  $U^1$  defined as above. If  $\|v_h - v\| \leq Ch^r \|v\|_r$  and  $v = 0$  on  $\partial\Omega$ , then

$$\begin{aligned} \|U^n - u(t_n)\| \leq & Ch^r \left( \|v\|_r + \int_0^{t_n} \|u_t\|_r ds \right) \\ & + Ck \int_0^k \|u_{tt}\| ds + Ck^2 \int_0^{t_n} \|u_{ttt}\| ds, \quad n \geq 0. \end{aligned}$$

Main result:

$$\|U^n - u(t_n)\| = O(h^r + k^2).$$



# Variable Time Step Backward Euler Method

Let  $0 = t_0 < t_1 < \dots < t_n < \dots$ ,  $k_n = t_n - t_{n-1}$ . Define  $\bar{\partial}_n U^n = \frac{U^n - U^{n-1}}{k_n}$ .

## Fully Discrete Variable-Step Backward Euler

Find  $U^n \in S_h$  such that

$$(\mathcal{VBE}) \quad (\bar{\partial}_n U^n, \chi) + (\nabla U^n, \nabla \chi) = (f(t_n), \chi), \quad n \geq 1,$$

for all  $\chi \in S_h$ , with  $U^0 = v_h$ .

## Theorem (1.8)

Let  $U^n$  and  $u$  be the solutions of  $(\mathcal{VBE})$  and  $(\mathcal{P})$ , with  $U^0 = v_h$  such that

$\|v_h - v\| \leq Ch^r \|v\|_r$  and  $v = 0$  on  $\partial\Omega$ . Then, for  $n \geq 0$ ,

$$\|U^n - u(t_n)\| \leq Ch^r \left( \|v\|_r + \int_0^{t_n} \|u_t\|_r ds \right) + \sum_{j=1}^n k_j \int_{t_{j-1}}^{t_j} \|u_{tt}\| ds.$$

Advantages:

- adaptive time stepping, efficient for multiscale problems,
- larger steps when the solution varies slowly.