

Introduction to Simulation - Lecture 19

Laplace's Equation – FEM Methods

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Thanks to Deepak Ramaswamy, Michal Rewienski,
and Karen Veroy, Jaime Peraire and Tony Patera

Outline for Poisson Equation Section

- Why Study Poisson's equation
 - Heat Flow, Potential Flow, Electrostatics
 - Raises many issues common to solving PDEs.
- Basic Numerical Techniques
 - basis functions (FEM) and finite-differences
 - Integral equation methods
- Fast Methods for 3-D
 - Preconditioners for FEM and Finite-differences
 - Fast multipole techniques for integral equations

Outline for Today

- Why Poisson Equation
 - Reminder about heat conducting bar
- Finite-Difference And Basis function methods
 - Key question of convergence
- Convergence of Finite-Element methods
 - Key idea: solve Poisson by minimization
 - Demonstrate optimality in a carefully chosen norm

Drag Force Analysis of Aircraft

- Potential Flow Equations
 - Poisson Partial Differential Equations.

Engine Thermal Analysis

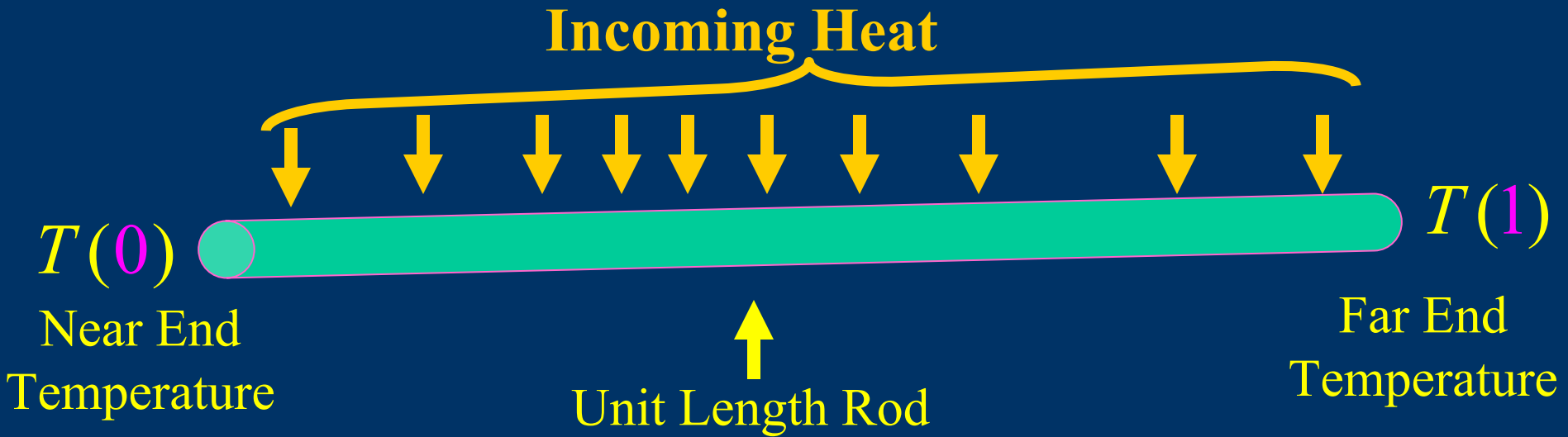
- Thermal Conduction Equations
 - The Poisson Partial Differential Equation.

Capacitance on a microprocessor Signal Line

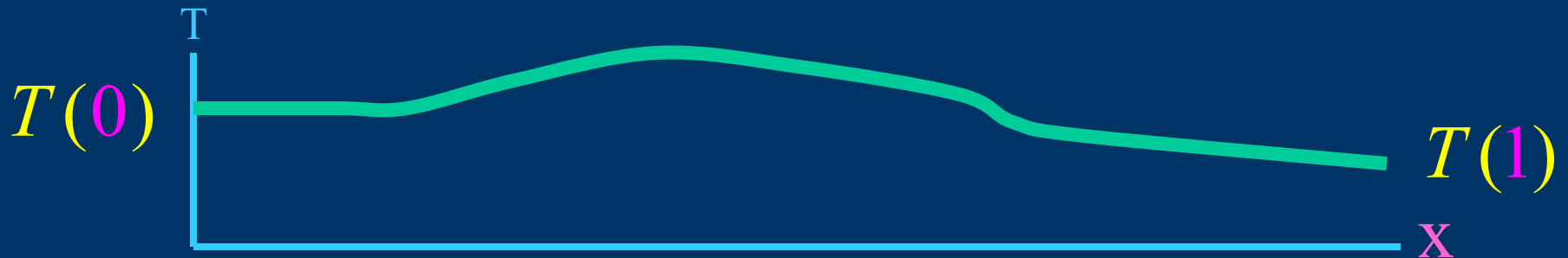
- **Electrostatic Analysis**
 - The Laplace Partial Differential Equation.

Heat Flow

1-D Example



Question: What is the temperature distribution along the bar



- 1) Cut the bar into short sections
- 2) Assign each cut a temperature



Heat Flow through one section

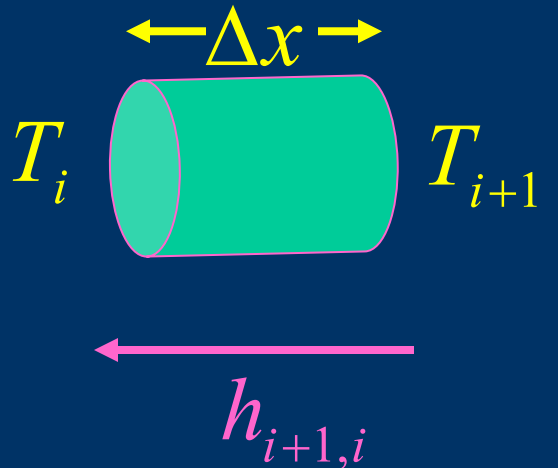
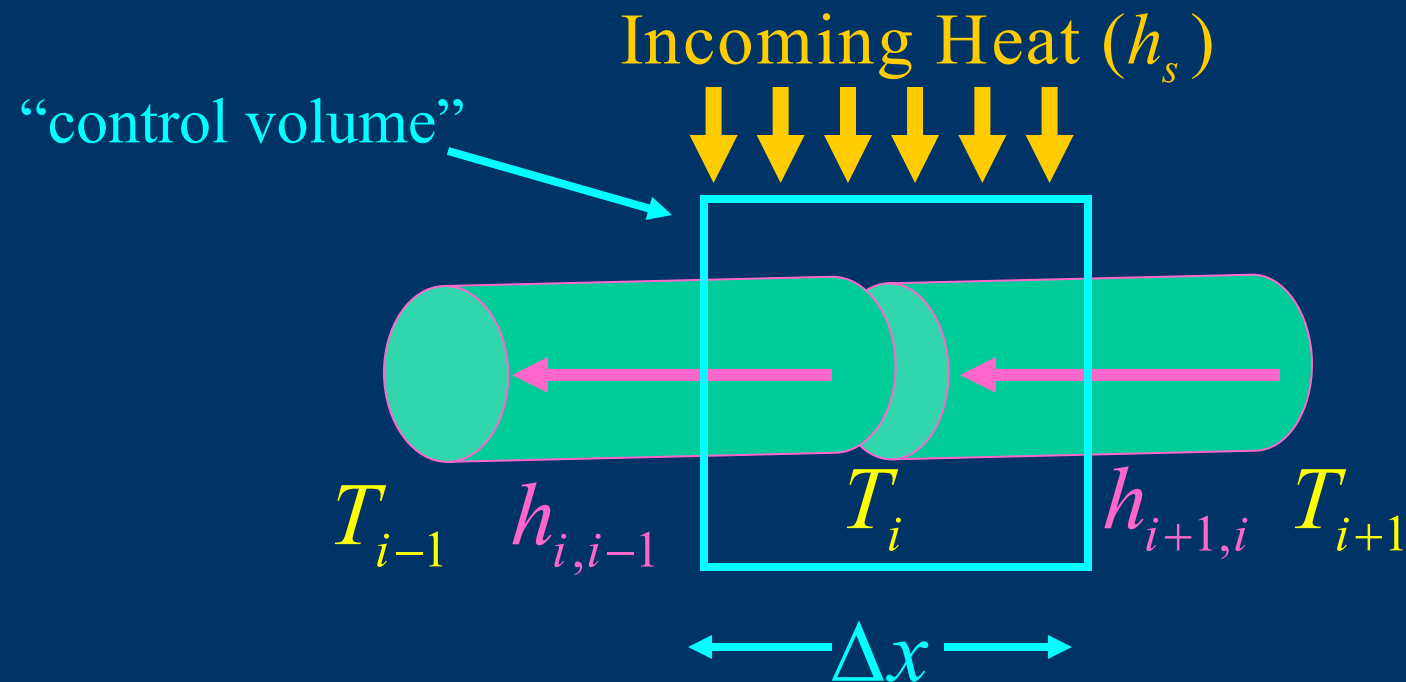


Diagram illustrating heat flow through a section of length Δx . The temperatures at the ends are T_i and T_{i+1} . The heat flow is denoted by $h_{i+1,i}$.

$$h_{i+1,i} = \text{heat flow} = \kappa \frac{T_{i+1} - T_i}{\Delta x}$$

Limit as the sections become vanishingly small

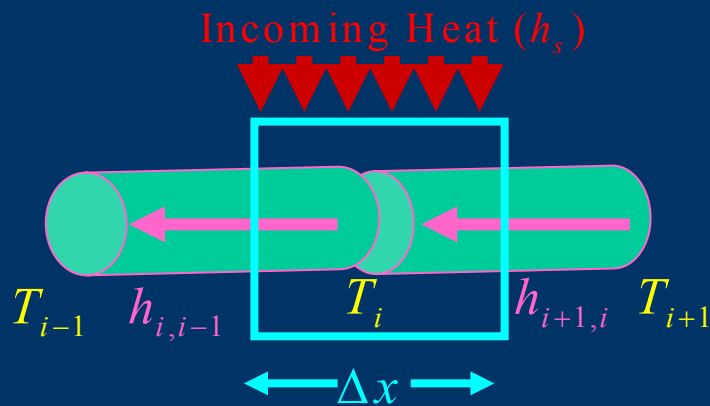
$$\lim_{\Delta x \rightarrow 0} h(x) = \kappa \frac{\partial T(x)}{\partial x}$$

Two Adjacent Sections

Heat Flows into Control Volume Sums to zero

$$h_{i+1,i} - h_{i,i-1} = -h_s \Delta x$$

Heat Flows into Control Volume Sums to zero



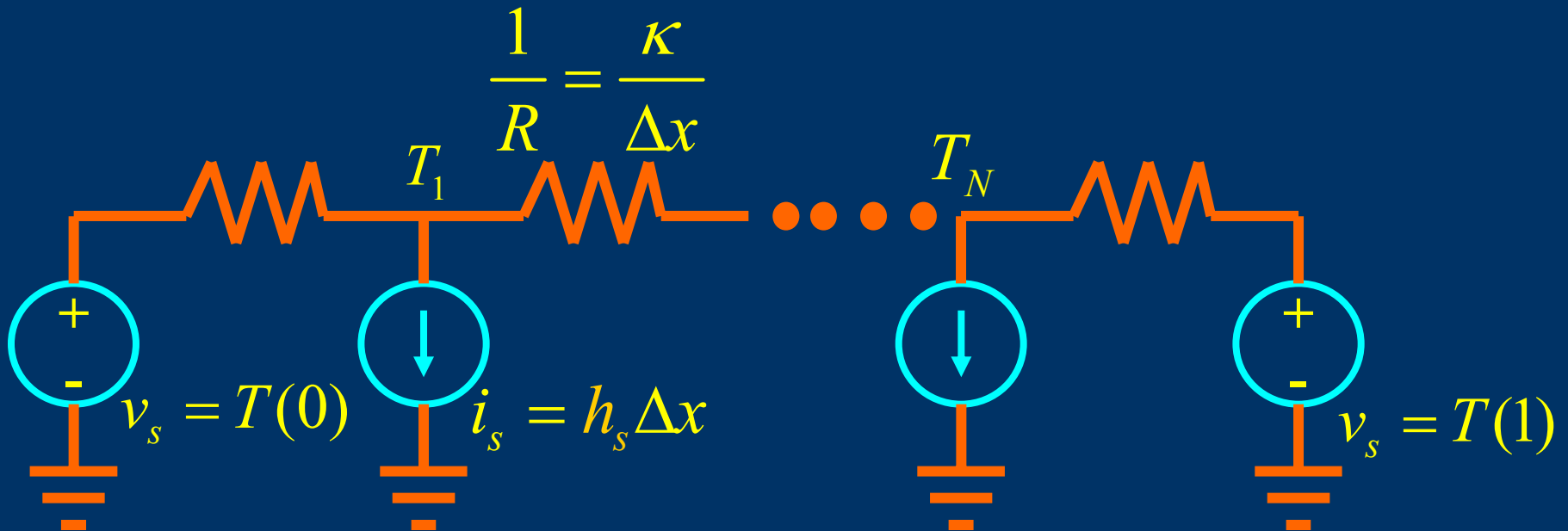
$$h_{i+1,i} - h_{i,i-1} = -h_s \Delta x$$

↑ Heat in from left ↑ Heat out from right ↑ Incoming heat per unit length

Limit as the sections become vanishingly small

$$\lim_{\Delta x \rightarrow 0} h_s(x) = \frac{\partial h(x)}{\partial x} = \frac{\partial}{\partial x} \kappa \frac{\partial T(x)}{\partial x}$$

Temperature analogous to Voltage
Heat Flow analogous to Current



Normalized Poisson Equation

$$\frac{\partial}{\partial x} \kappa \frac{\partial T(x)}{\partial x} = -h_s \implies -\frac{\partial^2 u(x)}{\partial x^2} = f(x)$$

$$-u_{xx}(x) = f(x)$$

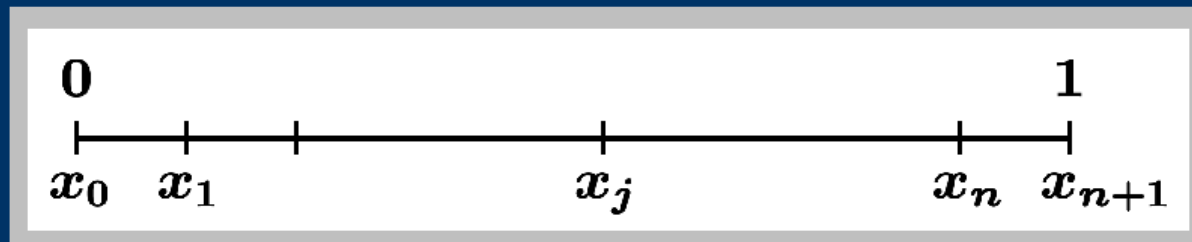
Numerical Solution

Finite Differences

Discretization

Subdivide interval $(0, 1)$ into $n + 1$ equal subintervals

$$\Delta x = \frac{1}{n + 1}$$



$$x_j = j\Delta x, \quad \hat{u}_j \approx u_j \equiv u(x_j)$$

$$\text{for } 0 \leq j \leq n + 1$$

Numerical Solution

Finite Differences

Approximation

For example ...

$$\begin{aligned}v''(x_j) &\approx \frac{1}{\Delta x} (v'(x_{j+1/2}) - v'(x_{j-1/2})) \\ &\approx \frac{1}{\Delta x} \left(\frac{v_{j+1} - v_j}{\Delta x} - \frac{v_j - v_{j-1}}{\Delta x} \right) \\ &= \frac{v_{j+1} - 2v_j + v_{j-1}}{\Delta x^2}\end{aligned}$$

for Δx small

Using Basis Functions

Partial Differential Equation form

$$-\frac{\partial^2 u}{\partial x^2} = f \quad u(0) = 0 \quad u(1) = 0$$

Basis Function Representation

$$u(x) \approx u_h(x) = \sum_{i=1}^n \omega_i \underbrace{\varphi_i(x)}_{\text{Basis Functions}}$$

Plug Basis Function Representation into the Equation

$$R(x) = \sum_{i=1}^n \omega_i \frac{d^2 \varphi_i(x)}{dx^2} + f(x)$$

Using Basis Functions

Example Basis functions

Introduce basis representation $u(x) \approx u_h(x) = \sum_{i=1}^n \omega_i \underbrace{\varphi_i(x)}_{\text{Basis Functions}}$

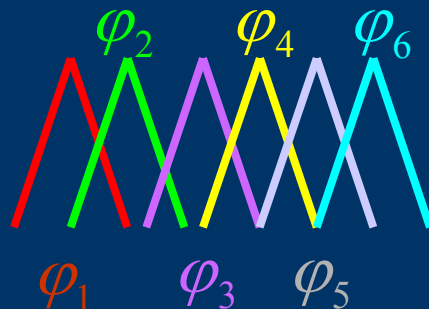
$\Rightarrow u_h(x)$ is a weighted sum of basis functions

The basis functions define a space

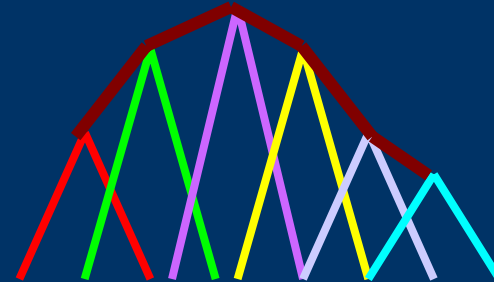
$$X_h = \left\{ v \in X_h \mid v = \sum_{i=1}^n \beta_i \varphi_i \text{ for some } \beta_i \text{'s} \right\}$$

Example

“Hat” basis functions



Piecewise linear Space



Using Basis functions

Basis Weights

Galerkin Scheme

Force the residual to be “orthogonal” to the basis functions

$$\int_0^1 \varphi_l(x) R(x) dx = 0$$

Generates n equations in n unknowns

$$\int_0^1 \varphi_l(x) \left[\sum_{i=1}^n \omega_i \frac{d^2 \varphi_i(x)}{dx^2} + f(x) \right] dx = 0 \quad l \in \{1, \dots, n\}$$

Using Basis Functions

Basis Weights

Galerkin with integration by parts

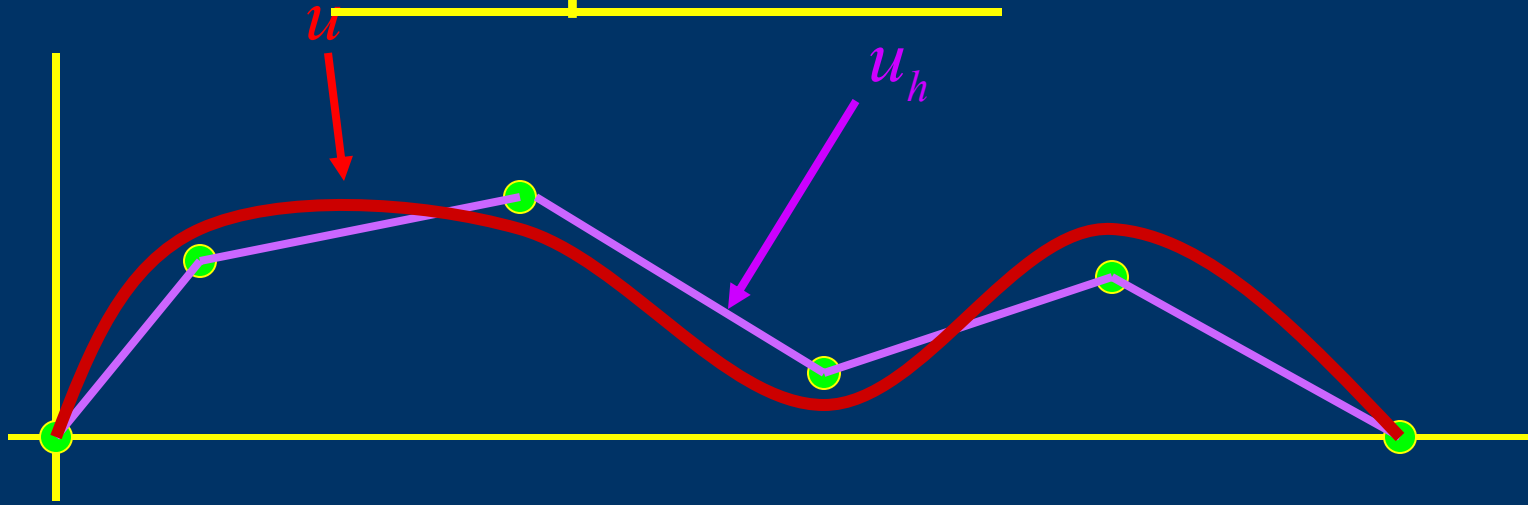
Only first derivatives of basis functions

$$\int_0^1 \frac{d\varphi_l(x)}{dx} \frac{d \sum_{i=1}^n \omega_i \varphi_i(x)}{dx} dx - \int_0^1 \varphi_i(x) f(x) dx = 0$$

$$l \in \{1, \dots, n\}$$

Convergence Analysis

The question is



How does $\underbrace{\|u - u_h\|}_{\text{error}}$ decrease with refinement?

- This time – Finite-element methods
- Next time – Finite-difference methods

Heat Equation

Overview of FEM

Partial Differential Equation form

$$-\frac{\partial^2 u}{\partial x^2} = f \quad u(0) = 0 \quad u(1) = 0$$

“Nearly” Equivalent weak form

$$\underbrace{\int_{\Omega} \frac{\partial u}{\partial x} \frac{\partial v}{\partial x} dx}_{a(u,v)} = \underbrace{\int_{\Omega} f v dx}_{l(v)} \quad \text{for all } v$$

Introduced an abstract notation for the equation u must satisfy

$$a(u, v) = l(v) \quad \text{for all } v$$

Introduce basis representation $u(x) \approx u_h(x) = \sum_{i=1}^n \omega_i \underbrace{\varphi_i(x)}_{\text{Basis Functions}}$

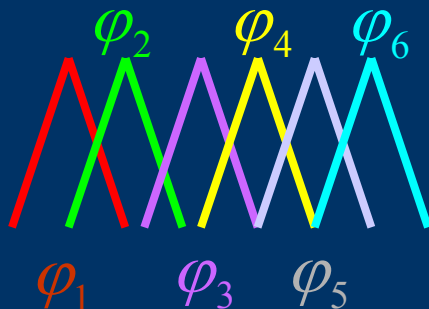
$\Rightarrow u_h(x)$ is a weighted sum of basis functions

The basis functions define a space

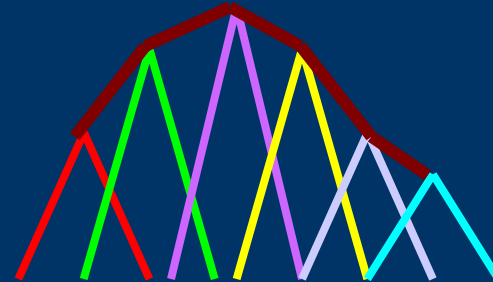
$$X_h = \left\{ v \in X_h \mid v = \sum_{i=1}^n \beta_i \varphi_i \text{ for some } \beta_i \text{'s} \right\}$$

Example

“Hat” basis functions



Piecewise linear Space



Key Idea

$a(u, u)$ defines a norm $a(u, u) \equiv \|u\|$

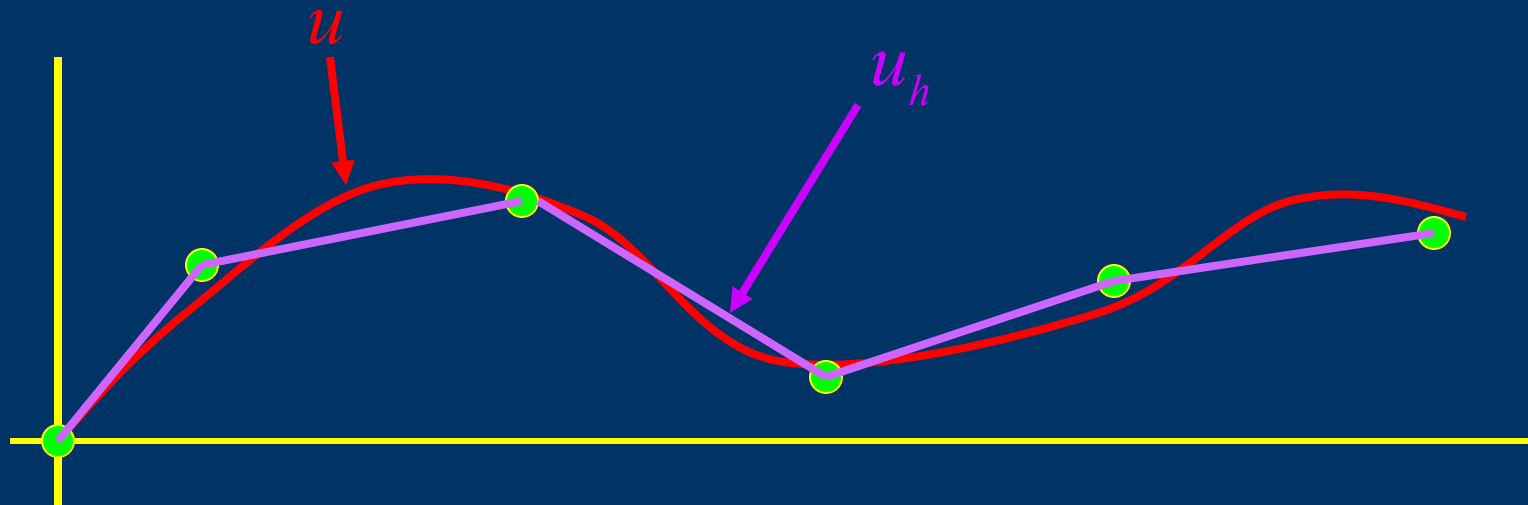
U is restricted to be 0 at 0 and 1!!

Using the norm properties, it is possible to show

If $a(u_h, \varphi_i) = l(\varphi_i)$ for all $\varphi_i \in \{\varphi_1, \varphi_2, \dots, \varphi_n\}$

Then $\underbrace{\|u - u_h\|}_{\text{Solution Error}} = \min_{w_h \in X_h} \underbrace{\|u - w_h\|}_{\text{Projection Error}}$

The question is only



How well can you fit u with a member of X_h

But you must measure the error in the $\| \cdot \|$ norm

For piecewise linear:

$$\underbrace{\|u - u_h\|}_{\text{error}} = O\left(\frac{1}{n}\right)$$

Problem of interest

Helmholtz Equation in 1D

Boundary Value Problem (BVP) - Strong Form

$$-u''(x) + \alpha u(x) = f(x) \quad \alpha \geq 0$$

$$x \in (0, 1), \quad u(0) = u(1) = 0$$

Describes many physical phenomena (e.g.) :

- Temperature distribution in a bar *
- Deformation of an elastic bar
- Deformation of a string under tension

N1

N2

Problem of interest

Solution Properties

- the solution $u(x)$ always *exists*
- $u(x)$ is always smoother than the data $f(x)$
- given $f(x)$ the solution $u(x)$ is *unique*

Minimization Principle

Statement

Find

$$u = \arg \min_{w \in X} J(w)$$

where

$$X = \{v \text{ sufficiently smooth} \mid v(0) = v(1) = 0\},$$

and

$$J(w) = \frac{1}{2} \int_0^1 (w_x w_x + \alpha w w) dx - \int_0^1 f w dx$$

Minimization Principle

Statement

In words:

Over all functions w in X ,
 u that satisfies

$$\begin{aligned} -u_{xx} + \alpha u &= f && \text{in } \Omega \\ u(0) = u(1) &= 0 \end{aligned}$$

makes $J(w)$ as small as possible.

N4

Minimization Principle

Statement

Proof...

Let $w = u + v$.

Then

$$\begin{aligned} J(\underbrace{u}_{\in X} + \underbrace{v}_{\in X}) &= \frac{1}{2} \int_0^1 (u + v)_x (u + v)_x dx \\ &+ \frac{\alpha}{2} \int_0^1 (u + v)(u + v) dx \\ &- \int_0^1 f(u + v) dx . \end{aligned}$$

Minimization Principle

Statement

...Proof...

$$J(u + v) = \frac{1}{2} \int_0^1 (u_x u_x + \alpha u u) dx - \int_0^1 f u dx \quad J(u)$$

$$+ \int_0^1 (u_x v_x + \alpha u v) dx - \int_0^1 f v dx \quad \begin{array}{l} \delta J_v(u) \\ \text{first variation} \end{array}$$

$$+ \frac{1}{2} \int_0^1 (v_x v_x + \alpha v v) dx \quad > 0 \text{ for } v \neq 0$$

Minimization Principle

Statement

...Proof...

$$\begin{aligned}\delta J_v(u) &= \int_0^1 (u_x v_x + \alpha uv) dx - \int_0^1 f v dx \\ &= \cancel{\delta^0}^0(0) u_x(0) - \cancel{\delta^0}^0(1) u_x(1) - \int_0^1 u_{xx} v dx \\ &\quad + \alpha \int_0^1 uv dx - \int_0^1 f v dx \\ &= \int_0^1 v \underbrace{\{-u_{xx} + \alpha u - f\}}_0 dx = 0, \quad \forall v \in X\end{aligned}$$

Minimization Principle

Statement

...Proof

$$J(\underbrace{u + v}_w) = J(u) + \frac{1}{2} \underbrace{\int_0^1 (v_x v_x + \alpha v v) dx}_{> 0 \text{ unless } v = 0}, \quad \forall v \in X$$

\Rightarrow

$$J(w) > J(u), \quad \forall w \in X, w \neq u$$

\Leftrightarrow

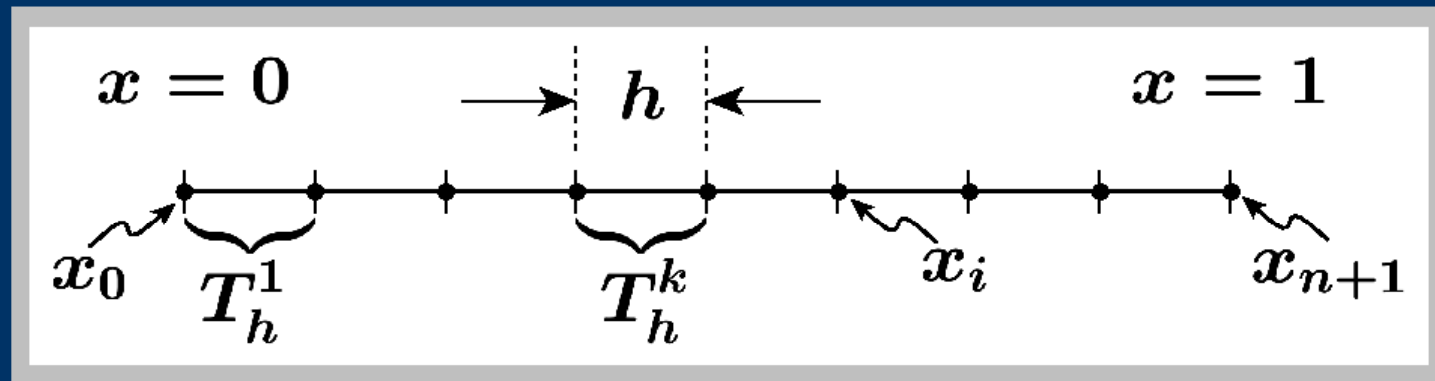
u is *the* minimizer of $J(w)$

E1 N5

Rayleigh-Ritz Approach

Approximation

Mesh



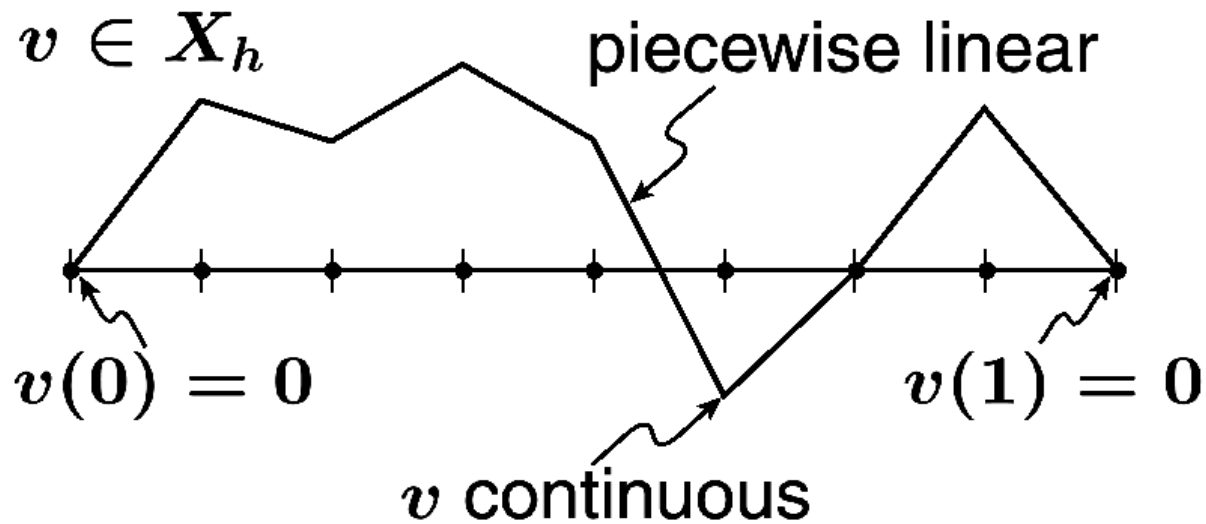
$$\bar{\Omega} = \bigcup_{k=1}^K \bar{T}_h^k \quad T_h^k, k = 1, \dots, K = n + 1: \text{elements}$$
$$x_i, i = 0, \dots, n + 1: \text{nodes}$$

Rayleigh-Ritz Approach

Approximation

Space $X_h \subset X$

$$X_h = \left\{ v \in X \mid v|_{T_h^k} \in \mathbb{P}_1(T_h^k), \quad k = 1, \dots, K \right\}$$



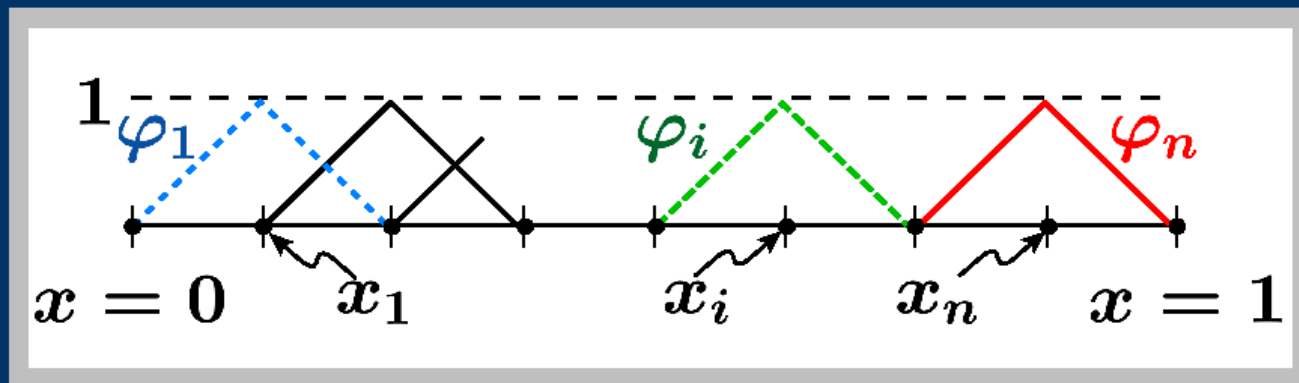
Rayleigh-Ritz Approach

Approximation

Basis

Nodal basis for X_h :

$$\varphi_j, j = 1, \dots, n = \dim(X_h)$$



$$\varphi_i \text{ nonzero only on } \overline{T}_h^i \cup \overline{T}_h^{i+1}$$

N7

N8

Rayleigh-Ritz Approach

“Projection”

Plan...

Let

$$\underbrace{u_h}_{\text{RR/FE Approximation}} (\in X_h) = \sum_{j=1}^n u_{hj} \varphi_j(x);$$

set $u_{hj} = w_j$ that minimize

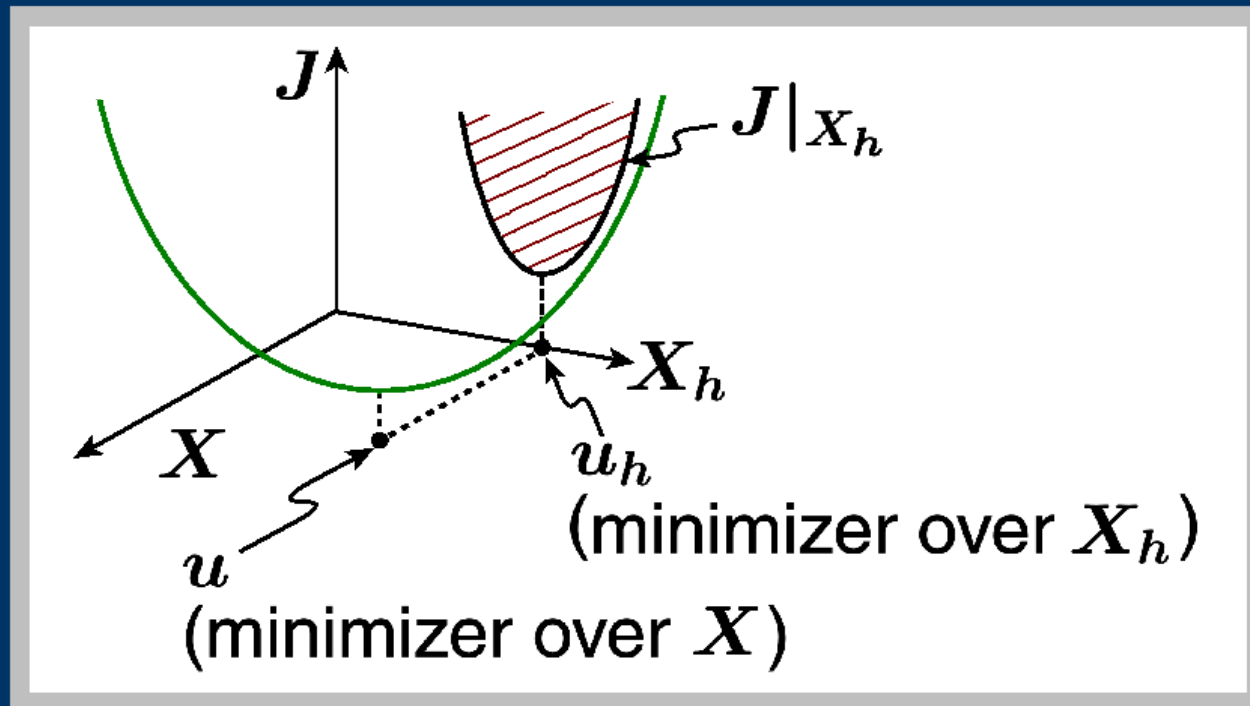
$$J \left(\sum_{j=1}^n w_j \varphi_j \right).$$

Rayleigh-Ritz Approach

“Projection”

...Plan

Geometric Picture:



Rayleigh-Ritz Approach

“Projection”

$J|_{X_h} \dots$

$$\begin{aligned} J \left(\sum_{j=1}^n w_j \varphi_j \right) &= \frac{1}{2} \int_0^1 \frac{d}{dx} \left(\sum_{i=1}^n w_i \varphi_i \right) \frac{d}{dx} \left(\sum_{j=1}^n w_j \varphi_j \right) \\ &\quad + \frac{\alpha}{2} \int_0^1 \sum_{i=1}^n (w_i \varphi_i) \sum_{j=1}^n (w_j \varphi_j) - \int_0^1 f \sum_{j=1}^n w_j \varphi_j \\ &= \frac{1}{2} \sum_{i=1}^n \sum_{j=1}^n w_i w_j \int_0^1 \left(\frac{d\varphi_i}{dx} \frac{d\varphi_j}{dx} + \alpha \varphi_i \varphi_j \right) dx - \sum_{j=1}^n w_j \int_0^1 f \varphi_j dx \end{aligned}$$

by *bilinearity* and *linearity*.

Rayleigh-Ritz Approach

“Projection”

... $J|_{X_h}$

$$\begin{aligned} J^R(\underline{w} \in \mathbb{R}^n) &\equiv J \left(\sum_{j=1}^n w_j \varphi_j \right) \\ &= \frac{1}{2} \underline{w}^T \underline{A}_h \underline{w} - \underline{w}^T \underline{F}_h. \end{aligned}$$

$$\underline{F}_h \in \mathbb{R}^n: F_{hi} = \int_0^1 f \varphi_i dx$$

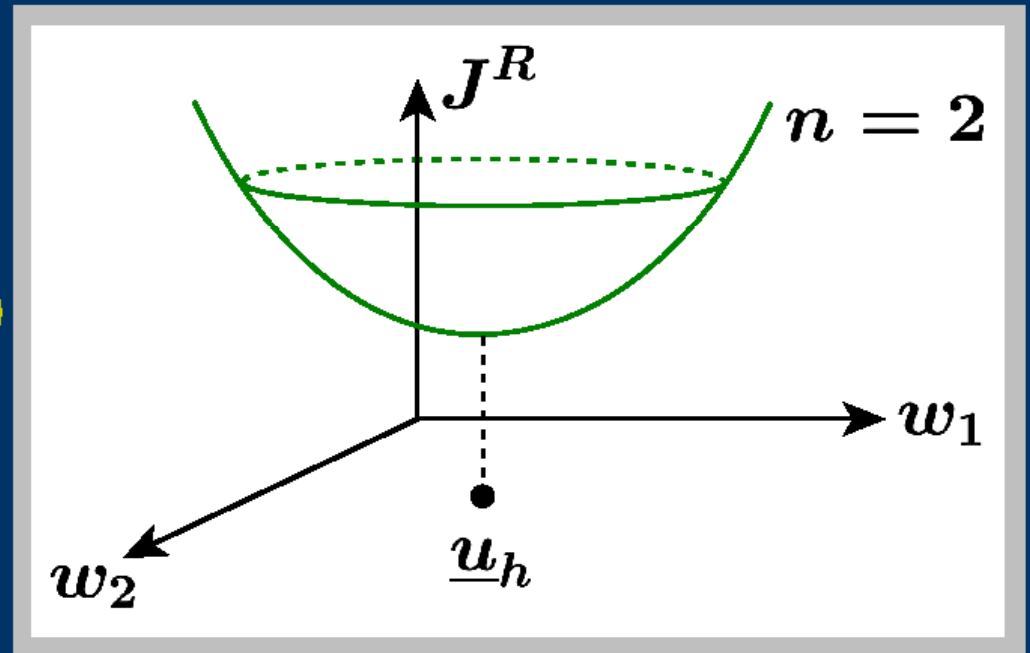
$$\underline{A}_h \in \mathbb{R}^{n \times n}: A_{hij} = \int_0^1 \left(\frac{d\varphi_i}{dx} \frac{d\varphi_j}{dx} + \alpha \varphi_i \varphi_j \right) dx$$

Rayleigh-Ritz Approach

“Projection”

Minimization...

$$\underline{u}_h = \arg \min_{\underline{w} \in \mathbb{R}^n} J^R(\underline{w})$$



Expand $J(\underline{w} = \underline{u}_h + \underline{v})$; require $J(\underline{w}) > J(\underline{u}_h)$ unless $\underline{v} = \mathbf{0}$.

Rayleigh-Ritz Approach

“Projection”

...Minimization...

$$\begin{aligned} J^R(\underline{u}_h + \underline{v}) &= \frac{1}{2} (\underline{u}_h + \underline{v})^T \underline{A}_h (\underline{u}_h + \underline{v}) - (\underline{u}_h + \underline{v})^T \underline{F}_h \\ &= \frac{1}{2} \underline{u}_h^T \underline{A}_h \underline{u}_h - \underline{u}_h^T \underline{F}_h \\ &\quad + \frac{1}{2} \underline{v}^T \underline{A}_h \underline{u}_h + \frac{1}{2} \underline{u}_h^T \underline{A}_h \underline{v} - \underline{v}^T \underline{F}_h \\ &\quad + \frac{1}{2} \underline{v}^T \underline{A}_h \underline{v} \end{aligned}$$

Rayleigh-Ritz Approach

“Projection”

...Minimization...

$$J^R(\underline{u}_h + \underline{v}) = J(\underline{u})$$

$$+ \underbrace{(\underline{A}_h \underline{u}_h - \underline{F}_h)^T}_{\nabla J^R(\underline{u}_h)} \underline{v} \quad \delta J_{\underline{v}}^R(\underline{u}_h) \quad SPD$$

$$+ \frac{1}{2} \underbrace{\underline{v}^T \underline{A}_h \underline{v}}_{>0, \forall \underline{v} \neq 0} \quad SPD$$

Rayleigh-Ritz Approach

“Projection”

...Minimization

If (and only if)

$$\delta J_{\underline{v}}^R(\underline{u}_h) = \mathbf{0}, \quad \forall \underline{v} \in \mathbb{R}^n$$

\Updownarrow

$$\nabla J^R(\underline{u}_h) = \underline{A}_h \underline{u}_h - \underline{F}_h = \underline{\mathbf{0}}$$

then

$$J(\underline{w} = \underline{u}_h + \underline{v}) > J(\underline{u}_h), \quad \forall \underline{v} \neq \mathbf{0}.$$

Rayleigh-Ritz Approach

“Projection”

Final Result

Find $\underline{u}_h \in \mathbb{R}^n$ such that

$$\underline{A}_h \underline{u}_h = \underline{F}_h \quad \Rightarrow \quad u_h(\mathbf{x}) = \sum_{j=1}^n u_{hj} \varphi_j(\mathbf{x}) .$$

SPD \Rightarrow existence and uniqueness.

Energy norm

Error Analysis

Remember

$$J(u + v) = J(u) + \underbrace{\frac{1}{2} \int_0^1 (v_x v_x + \alpha v v) dx}_{\geq 0, SPD}, \quad \forall v \in X$$

Define

$$|||v||| = \left[\int_0^1 (v_x v_x + \alpha v v) dx \right]^{\frac{1}{2}}$$

Energy norm

Error Analysis

Therefore

$$J(u + v) = J(u) + \frac{1}{2}|||v|||^2, \quad \forall v \in X$$

Choose any $w_h \in X_h$, $v \rightarrow (w_h - u) \in X$

$$J(w_h) = J(u) + \frac{1}{2}|||u - w_h|||^2, \quad \forall w_h \in X_h$$

For $w_h = u_h$

$$J(u_h) = J(u) + \frac{1}{2}|||u - u_h|||^2$$

Error Analysis

$$J(u_h) < J(w_h), \quad \forall w_h \in X_h, \quad w_h \neq u_h$$

if $e = u - u_h$

$$\| \underbrace{u - u_h}_e \| < \| u - w_h \|, \quad \forall w_h \in X_h, \quad w_h \neq u_h$$

and

$$\| e \| = \inf_{w_h \in X_h} \| u - w_h \|$$

Error Analysis

In words: even if you *knew* u ,

you could not find a w_h in X_h

more accurate than u_h

in the energy norm.

Error Analysis

A priori error estimates

N9

Energy norm:

$$\|e\| \leq C_1 h$$

L_2 norm:

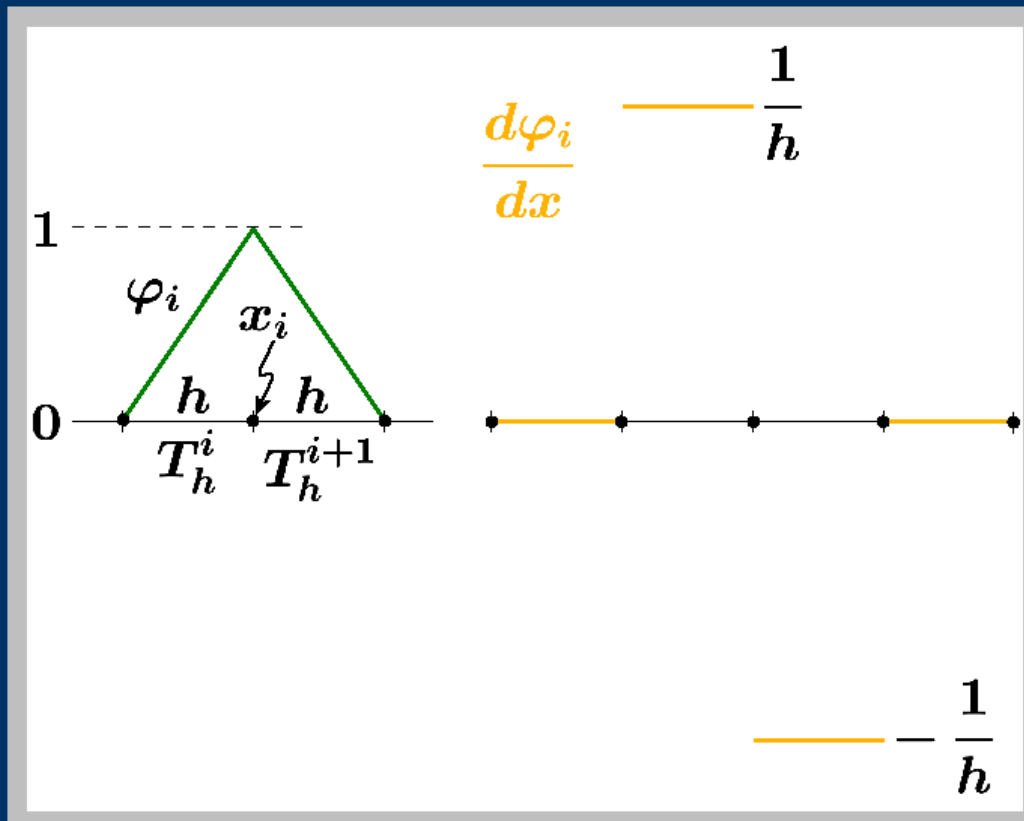
$$\|e\| = \left(\int_0^1 e e \, dx \right)^{1/2} \leq C_2 h^2$$

$C_{1,2} = \mathcal{F}(\Omega, \text{problem parameters, smoothness of } u)$

Discrete Equations

Matrix Elements: \underline{A}_h^1

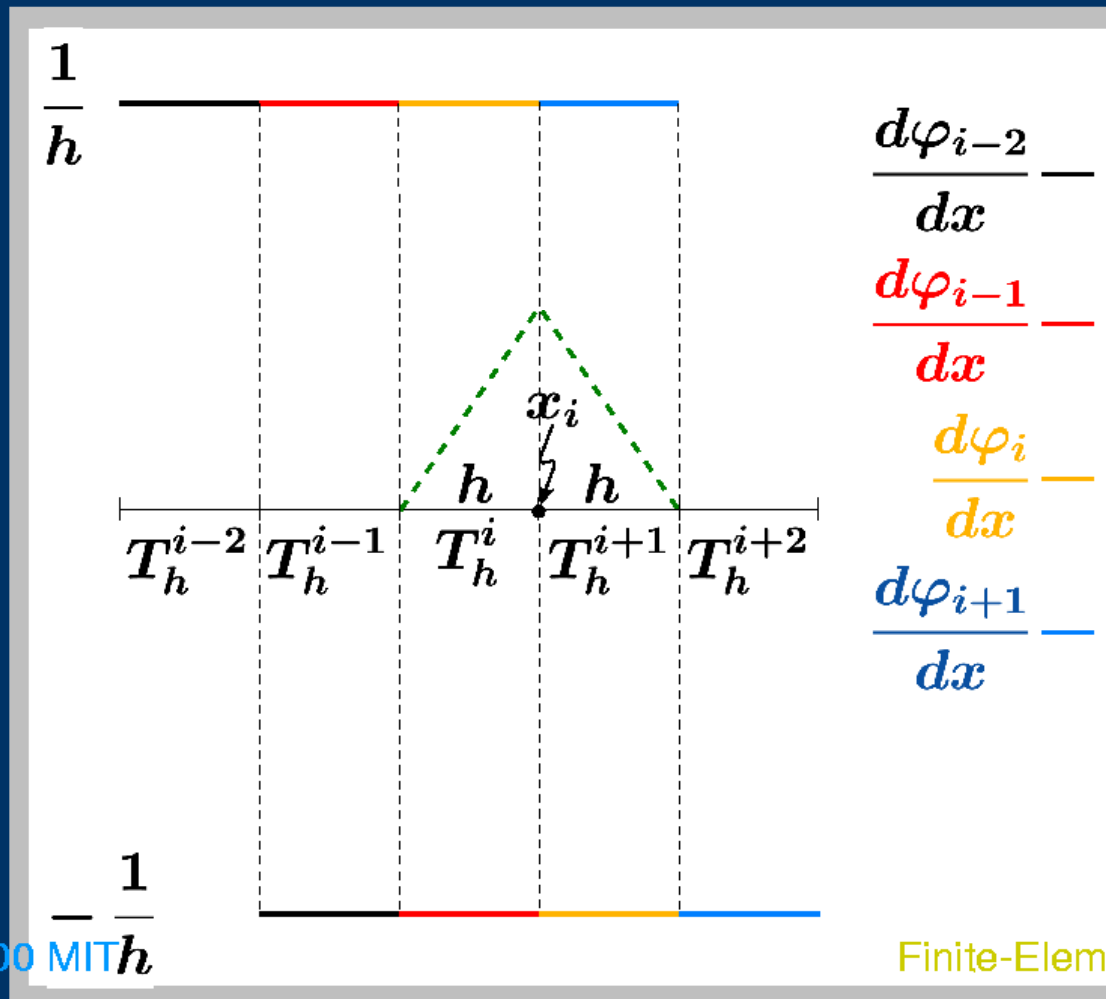
φ_i and $d\varphi_i/dx...$



Discrete Equations

Matrix Elements: \underline{A}_h^1

... φ_i and $d\varphi_i/dx$



Discrete Equations

Matrix Elements: \underline{A}_h^1

Typical Row

$$A_{hij}^1 = \int_{\Omega} \frac{d\varphi_i}{dx} \frac{d\varphi_j}{dx} dx = \int_{T_h^i} \frac{d\varphi_i}{dx} \frac{d\varphi_j}{dx} dx + \int_{T_h^{i+1}} \frac{d\varphi_i}{dx} \frac{d\varphi_j}{dx} dx$$

is nonzero only for $i = j - 1, j, j + 1$

$$A_{hii}^1 = \frac{1}{h^2} (h) + \frac{1}{h^2} (h) = \frac{2}{h}$$

$$A_{hii-1}^1 = \frac{1}{h} \left(-\frac{1}{h}\right) (h) = -\frac{1}{h}$$

$$A_{hii+1}^1 = \left(-\frac{1}{h}\right) \frac{1}{h} (h) = -\frac{1}{h}$$

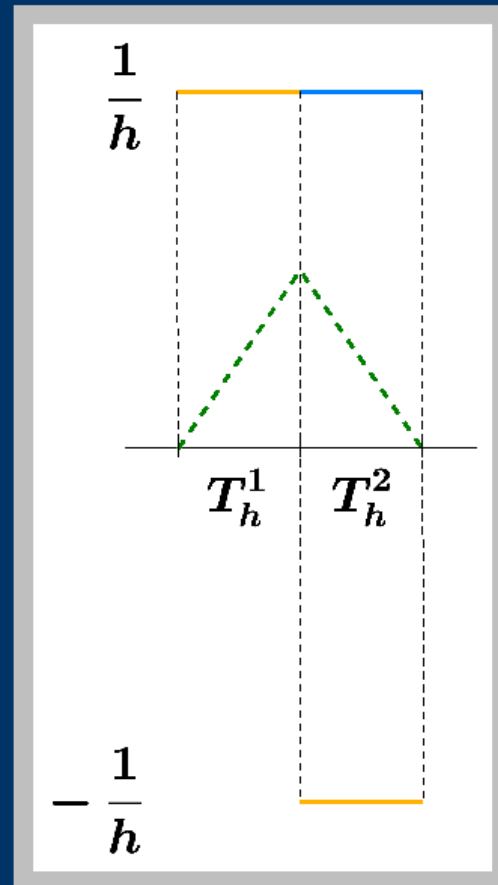
Discrete Equations

Matrix Elements: \underline{A}_h^1

Boundary Rows

$$A_{h11}^1 = \frac{2}{h}, \quad A_{h12}^1 = -\frac{1}{h},$$

$$A_{hnn}^1 = \frac{2}{h}, \quad A_{hnn-1}^1 = -\frac{1}{h}.$$



Discrete
Equations

$\underline{M}_h \in \mathbb{R}^{n \times n}$:

$$M_{hij} = \int_0^1 \varphi_i \varphi_j dx$$

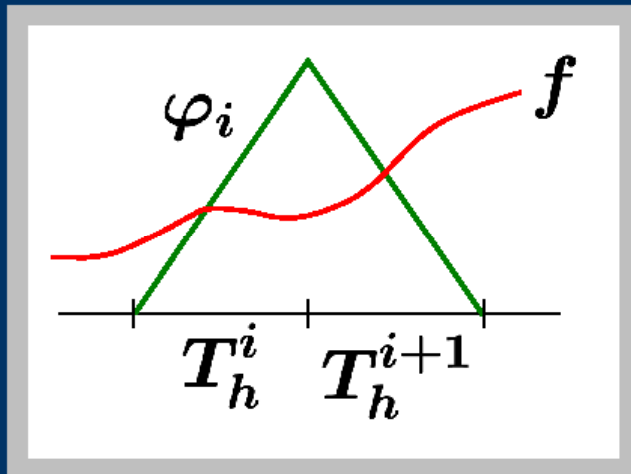
the finite element “identity” (\mathbf{I}) operator

Is nonzero only for $i = j - 1, j, j + 1$

$$M_{hij} = \int_{T_h^i} \varphi_i \varphi_j dx + \int_{T_h^{i+1}} \varphi_i \varphi_j dx$$

Discrete Equations

“Load” Vector Elements: \underline{F}_h



$$F_{hi} = \int_0^1 f \varphi_i dx$$

$$F_{hi} = \int_{T_h^i} f \varphi_i dx + \int_{T_h^{i+1}} f \varphi_i dx, \quad i = 1, \dots, n;$$

Discrete Equations

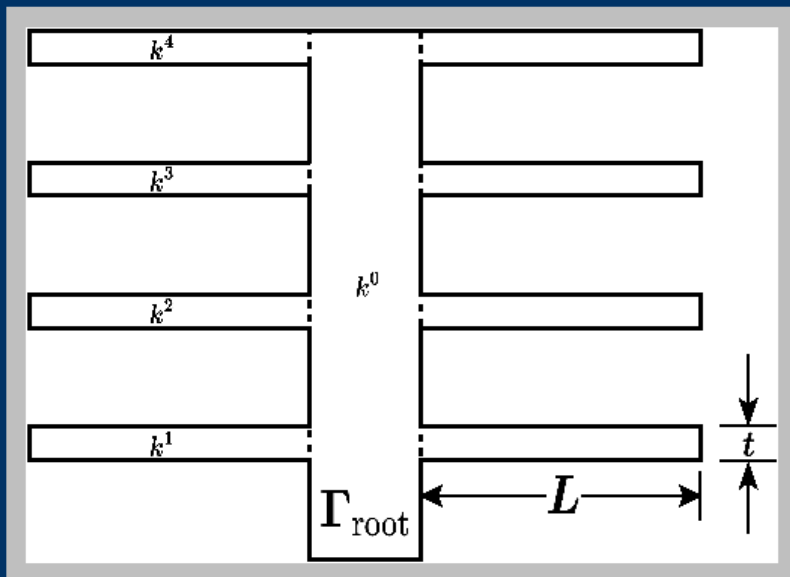
$\underline{u}_h \in \mathbb{R}^n$ satisfies

$$[\underline{A}_h^1 + \alpha \underline{M}_h] \begin{pmatrix} u_{h1} \\ \vdots \\ u_{hn} \end{pmatrix} = \begin{pmatrix} F_{h1} \\ \vdots \\ F_{hn} \end{pmatrix}$$

Heat Transfer Problem

Example

Non-dimensional form



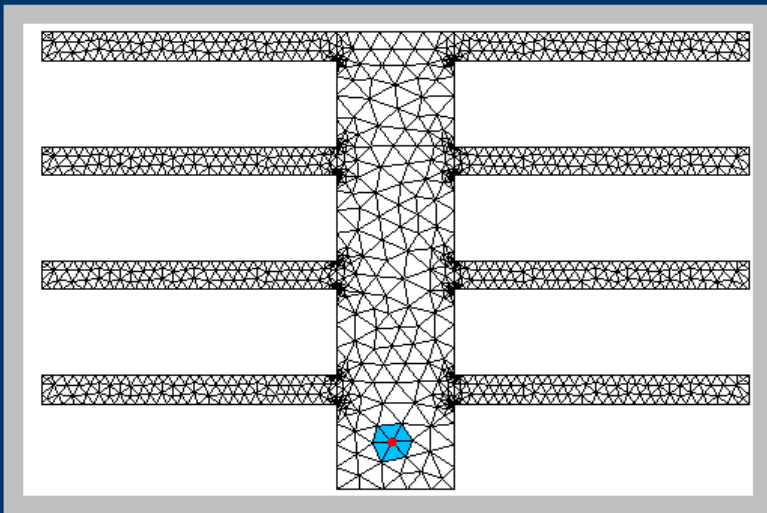
k^i : Thermal conductivity for Ω_i , $i = 0, \dots, 4$

Bi : Heat transfer coefficient

t
 L : Geometric parameters

Finite element method

Example



$$X_h = \text{span}\{\varphi_1, \dots, \varphi_n\}$$

$\varphi_i(x)$:

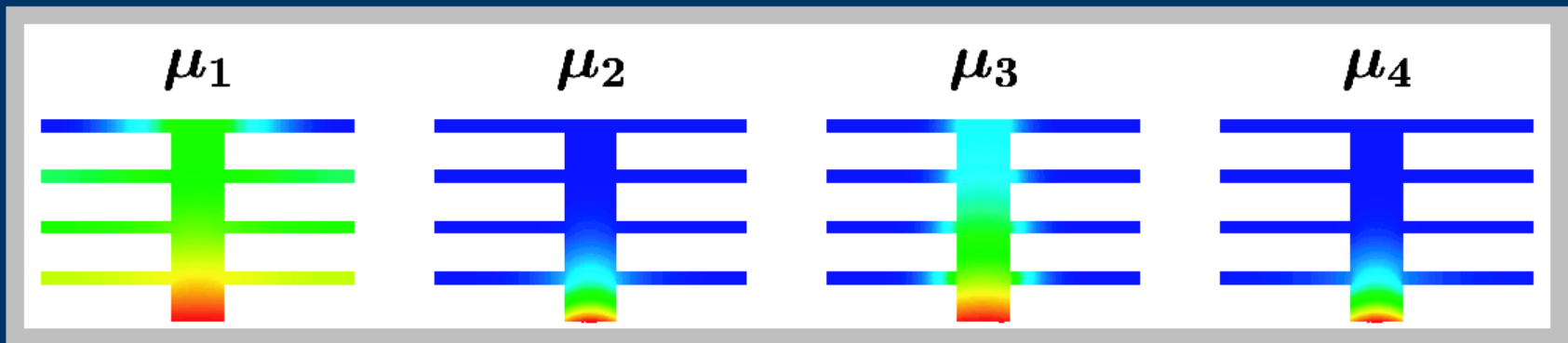
Nodal basis functions

– First order elements

$$\dim(X_h) = n$$

Possible solutions

Example



Example

- Complicated geometries
- General classes of problems
(Good mathematical properties)
- Wider class of operators

Summary

- Why Poisson Equation
 - Reminder about heat conducting bar
- Finite-Difference And Basis function methods
 - Key question of convergence
- Convergence of Finite-Element methods
 - Key idea: solve Poisson by minimization
 - Demonstrate optimality in a carefully chosen norm