

有限元方法简介

胡俊

北京大学数学科学学院

June 2, 2021

- 1 两点边值问题
- 2 有限元方法
- 3 先验误差估计
- 4 一维自适应例子
- 5 二维泊松问题及有限元方法
- 6 误差估计

1 两点边值问题

2 有限元方法

3 先验误差估计

4 一维自适应例子

5 二维泊松问题及有限元方法

6 误差估计

两点边值问题

求解如下方程

$$\begin{cases} -u''(x) = f(x), x \in \Omega := (0, 1) \\ u(0) = u(1) = 0 \end{cases} \quad (1)$$

弱导数: 对于函数 $v(x)$, 若存在 $v_g(\int_a^b |v_g|dx < \infty)$, 满足

$$\int_a^b v\varphi'(x)dx = - \int_a^b v_g\varphi(x)dx, \forall \varphi(x) \in C_0^\infty(a, b)$$

则称 $v_g(x)$ 是 $v(x)$ 的弱导数, 也称为广义导数, 记为 $v'(x)$.

例:

$$v(x) = \begin{cases} x, & 0 \leq x \leq 1; \\ -x, & -1 \leq x < 0. \end{cases} \quad \text{弱导数: } v_g(x) = \begin{cases} 1, & 0 \leq x \leq 1; \\ -1, & -1 \leq x < 0. \end{cases}$$

$$H_0^1(\Omega) := \left\{ v : \int_{\Omega} v^2 dx < \infty, \int_{\Omega} v_g^2 dx < \infty, v(0) = v(1) = 0 \right\}.$$

定义

$$\varphi(t) = \begin{cases} 2t, & 0 \leq t \leq \frac{1}{2}, \\ 1, & t > \frac{1}{2}. \end{cases}$$

对 $v(x) \in H_0^1(\Omega)$, 令 $w(x) = v(x)\varphi(x)$, 于是, 有

$$\begin{aligned} w(x) &= w(0) + \int_0^x w'(t)dt \\ &= \int_0^x [v'(t)\varphi(t) + v(t)\varphi'(t)]dt \\ &\leq \int_0^1 |v'(t)|dt + 2 \int_0^1 |v(t)|dt \\ &\leq C\|v\|_{H^{1,1}(\Omega)} \end{aligned}$$

因此, $\forall x \geq \frac{1}{2}$, 有

$$v(x) = w(x) \leq C\|v\|_{H^{1,1}(\Omega)}.$$

同理可证明 $\forall 0 \leq x < \frac{1}{2}$, 有

$$v(x) \leq C \|v\|_{H^{1,1}(\Omega)}$$

对 $\forall v(x) \in H_0^1(\Omega), x, y \in (0, 1)$, 有

$$|v(x) - v(y)| = \left| \int_x^y v'(t) dt \right| \leq |x - y|^{1/2} \left(\int_a^b |v'(t)|^2 dt \right)^{1/2}$$

因此 $v(x)$ Holder 连续.

1 两点边值问题

2 有限元方法

3 先验误差估计

4 一维自适应例子

5 二维泊松问题及有限元方法

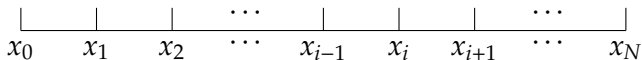
6 误差估计

在方程(1)的两端同乘以 $v(x) \in H_0^1(\Omega)$, 有

$$\begin{aligned} - \int_0^1 u''(x)v(x)dx &= \int_0^1 f(x)v(x)dx \\ \int_0^1 u'(x)v'(x)dx &= \int_0^1 f(x)v(x)dx \end{aligned} \quad (2)$$

问题(2)是问题(1)的变分问题. 将区间 $[0, 1]$ 进行剖分:

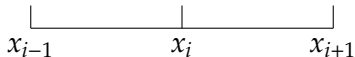
Figure



$x_i, i = 0, 1, 2, \dots, N$ 称为剖分的节点, $[x_i, x_{i+1}]$ 称为单元.
 在内节点 x_i 上, 令 $h_i = x_i - x_{i-1}$, 定义

$$\varphi_i(x) = \begin{cases} \frac{x-x_{i-1}}{h_i}, & x \in [x_{i-1}, x_i] \\ \frac{x_{i+1}-x}{h_{i+1}}, & x \in [x_i, x_{i+1}] \\ 0, & \text{其他} \end{cases}$$

Figure



定义: $V_h = \left\{ v_h : v_h = \sum_{i=1}^{N-1} v_i \varphi_i(x), v_i \in R \right\}$

有限元问题: 求 $u_h = \sum_{j=1}^{N-1} u_j \varphi_j(x)$, 使得

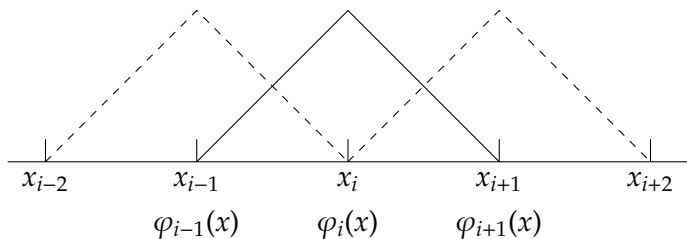
$$\int_0^1 u'_h(x) v'_h(x) dx = \int_0^1 v_h(x) f(x) dx, \forall v_h(x) \in V_h. \quad (3)$$

等价的问题: $\int_0^1 u'_h(x) \varphi'_i(x) dx = \int_0^1 \varphi_i f(x) dx, i = 1, 2, \dots, N-1$, 也即是

$$\sum_{j=1}^{N-1} u_j \int_0^1 \varphi'_i(x) \varphi'_j(x) dx = \int_0^1 f(x) \varphi_i(x) dx$$

只有 $j = i-1, i, i+1$ 时, φ_j 与 φ_i 有共同的支集.

Figure



故有

$$\begin{aligned}
 & u_{i-1} \int_0^1 \varphi'_{i-1}(x) \varphi'_i(x) dx + u_i \int_0^1 \varphi'_i(x) \varphi'_i(x) dx + u_{i+1} \int_0^1 \varphi'_{i+1}(x) \varphi'_i(x) dx \\
 & = \int_0^1 f \varphi_i(x) dx, \text{ 令 } f_i = \int_0^1 f \varphi'_i(x) dx
 \end{aligned}$$

直接计算, 得

$$\begin{aligned} & \int_0^1 \varphi'_{i-1}(x) \varphi'_i(x) dx \\ &= \int_{x_{i-1}}^{x_i} \varphi'_{i-1}(x) \varphi'_i(x) dx = \int_{x_{i-1}}^{x_i} \left(-\frac{1}{h_i}\right) \left(\frac{1}{h_i}\right) dx = -\frac{1}{h_i} \\ & \int_0^1 \varphi'_{i+1}(x) \varphi'_i(x) dx = \int_{x_i}^{x_{i+1}} \left(-\frac{1}{h_{i+1}}\right) \left(\frac{1}{h_{i+1}}\right) dx = -\frac{1}{h_{i+1}} \quad . \\ & \int_0^1 \varphi'_i(x) \varphi'_i(x) dx = \int_{x_{i-1}}^{x_i} \varphi'_i(x) \varphi'_i(x) dx + \int_{x_i}^{x_{i+1}} \varphi'_i(x) \varphi'_i(x) dx \\ &= \frac{1}{h_i} + \frac{1}{h_{i+1}} \end{aligned}$$

即

$$-\frac{u_{i-1}}{h_i} + \left(\frac{1}{h_i} + \frac{1}{h_{i+1}}\right)u_i - \frac{u_{i+1}}{h_{i+1}} = f_i, \quad i = 1, 2, \dots, N-1. \quad (4)$$

令 $a_i = \frac{1}{h_i} + \frac{1}{h_{i+1}}, i = 1, 2, \dots, N-2$, 并令

$$A_h = \begin{pmatrix} a_1 & -\frac{1}{h_2} & & & & \\ -\frac{1}{h_2} & a_2 & -\frac{1}{h_3} & & & \\ & -\frac{1}{h_3} & a_3 & -\frac{1}{h_4} & & \\ & & \ddots & \ddots & \ddots & \\ & & & -\frac{1}{h_{N-2}} & a_{N-2} & -\frac{1}{h_{N-1}} \\ & & & & -\frac{1}{h_{N-1}} & a_{N-1} \end{pmatrix}$$

与 $U_h = (u_1, u_2, \dots, u_{N-1})^T$, $F_h = (f_1, f_2, \dots, f_{N-1})^T$. (4) 的矩阵形式为:

$$A_h U_h = F_h.$$

$$f_i = \int_{x_{i-1}}^{x_{i+1}} \phi_i(x) f(x) dx \approx \frac{h_i f(x_{i-1})}{6} + \frac{h_i f(x_i)}{3} + \frac{h_{i+1} f(x_i)}{3} + \frac{h_{i+1} f(x_{i+1})}{6}.$$

1 两点边值问题

2 有限元方法

3 先验误差估计

4 一维自适应例子

5 二维泊松问题及有限元方法

6 误差估计

有如下关系式:

$$\begin{aligned}\int_0^1 u' v' dx &= \int_0^1 v f dx \\ \int_0^1 u' v'_h(x) dx &= \int_0^1 v_h f dx \quad \forall v_h \in V_h \\ \int_0^1 u'_h(x) v'_h(x) dx &= \int_0^1 v_h(x) f dx \quad \forall v_h \in V_h\end{aligned}$$

可得正交性:

$$\int_0^1 (u' - u'_h) v'_h dx = 0$$

故 $\forall w_h \in V_h$, 有

$$\begin{aligned} & \int_0^1 (u' - u'_h)^2 dx \\ &= \int_0^1 (u' - u'_h)(u' - w'_h) dx + \int_0^1 (u' - u'_h)(w'_h - u'_h) dx \\ & \quad (= 0, \text{ 令 } v_h = w_h - u_h \text{ 即可得}) \\ &= \int_0^1 (u' - u'_h)(u' - w'_h) dx \\ &\leq \left(\int_0^1 (u' - u'_h)^2 dx \right)^{1/2} \left(\int_0^1 (u' - w'_h)^2 dx \right)^{1/2} \end{aligned}$$

即有: $\int_0^1 (u' - u'_h)^2 dx \leq \int_0^1 (u' - w'_h)^2 dx$, 任对 $w_h \in V_h$ 成立, 也即是

$$\int_0^1 (u' - u'_h)^2 dx \leq \inf_{w_h \in V_h} \int_0^1 (u' - w'_h)^2 dx$$

定义插值函数 $I_h u \in V_h$ 使得

$$(I_h u)(x_i) = u(x_i), i = 1, 2, \dots, N$$

变点展开技术: 在单元 $[x_i, x_{i+1}]$ 上, 有:

$$u(x) - I_h u(x) = u(x) - u(x_i)\varphi_i(x) - u(x_{i+1})\varphi_{i+1}(x)$$

展开

$$\begin{aligned} u(x_{i+1}) &= u(x) + u'(x)(x_{i+1} - x) + \int_x^{x_{i+1}} u''(t)(x_{i+1} - t)dt \\ u(x_i) &= u(x) + u'(x)(x_i - x) + \int_x^{x_i} u''(t)(x_i - t)dt \end{aligned}$$

这样

$$\begin{aligned} u(x) - I_h u(x) &= u(x) - (u(x) + u'(x)(x_i - x)) \varphi_i(x) \\ &\quad - (u(x) + u'(x)(x_{i+1} - x)) \varphi_{i+1}(x) \\ &\quad - \int_x^{x_i} u''(t)(x_i - t) dt \varphi_i(x) \\ &\quad - \int_x^{x_{i+1}} u''(t)(x_{i+1} - t) dt \varphi_{i+1}(x) \\ &= -u'(x) ((x_i - x) \varphi_i(x) + (x_{i+1} - x) \varphi_{i+1}(x)) \\ &\quad - \int_x^{x_i} u''(t)(x_i - t) dt \varphi_i(x) \\ &\quad - \int_x^{x_{i+1}} u''(t)(x_{i+1} - t) dt \varphi_{i+1}(x) \\ &= - \int_x^{x_i} u''(t)(x_i - t) dt \varphi_i(x) \\ &\quad - \int_x^{x_{i+1}} u''(t)(x_{i+1} - t) dt \varphi_{i+1}(x) \end{aligned}$$

于是:

$$\begin{aligned}|u(x) - I_h u(x)| &\leq \left(\int_{x_i}^{x_{i+1}} (u''(t))^2 dt \right)^{1/2} \left(\int_{x_i}^{x_{i+1}} (x_i - t)^2 dt \right)^{1/2} \varphi_i(x) \\ &\quad + \left(\int_{x_i}^{x_{i+1}} (u''(t))^2 dt \right)^{1/2} \left(\int_{x_i}^{x_{i+1}} (x_{i+1} - t)^2 dt \right)^{1/2} \varphi_{i+1}(x) \\ &= \frac{1}{\sqrt{3}} \left(\int_{x_i}^{x_{i+1}} (u''(t))^2 dt \right)^{1/2} h_{i+1}^{3/2}\end{aligned}$$

于是

$$\begin{aligned}\sum_{i=0}^{N-1} \int_{x_i}^{x_{i+1}} (u(x) - I_h u(x))^2 dx &\leq \frac{1}{3} \sum_{i=0}^{N-1} \int_{x_i}^{x_{i+1}} (u''(x))^2 dx h_{i+1}^4 \\ &= \frac{1}{3} \sum_{i=0}^{N-1} h_{i+1}^4 \int_{x_i}^{x_{i+1}} (u''(x))^2 dx\end{aligned}$$

另一方面,

$$\begin{aligned}(u(x) - I_h u(x))' &= u'(x) + \frac{u(x_i)}{h_{i+1}} - \frac{u(x_{i+1})}{h_{i+1}} \\&= u'(x) + \frac{u(x) + u'(x)(x_i - x)}{h_{i+1}} - \frac{u(x) + u'(x)(x_{i+1} - x)}{h_{i+1}} \\&\quad + \frac{1}{h_{i+1}} \int_x^{x_i} u''(t)(x_i - t) dt - \frac{1}{h_{i+1}} \int_x^{x_{i+1}} u''(t)(x_{i+1} - t) dt \\&= \frac{1}{h_{i+1}} \left(\int_x^{x_i} u''(t)(x_i - t) dt - \int_x^{x_{i+1}} u''(t)(x_{i+1} - t) dt \right)\end{aligned}$$

于是

$$|(u(x) - I_h u(x))'| \leq \frac{2}{\sqrt{3}h_{i+1}} \left(\int_{x_i}^{x_{i+1}} (u''(x))^2 dx \right)^{1/2} h_{i+1}^{3/2}$$

这样:

$$\int_0^1 ((u(x) - I_h u(x))')^2 dx \leq \frac{4}{3} \sum_{i=0}^{N-1} h_{i+1}^2 \int_{x_i}^{x_{i+1}} (u''(x))^2 dx$$

有如下后验误差估计结果：问题(3)的有限元解 u_h 满足如下估计

$$\left\| \frac{d(u - u_h)}{dx} \right\|_{L^2(\Omega)}^2 \leq C \sum_{i=1}^n \eta_i^2(u_h) \quad (5)$$

其中

$$\eta_i(u_h) = h_i \left\| f + \frac{d^2 u_h}{dx^2} \right\|_{L^2(\Omega_i)}$$

证明:

令 $e = u - u_h$ 为误差, 定义插值算子 $\Pi: V_0 \rightarrow V_{h,0}$ 为连续分片线性插值, 有

$$\begin{aligned}\left\|\frac{de}{dx}\right\|_{L^2(\Omega)}^2 &= \int_{\Omega} \left(\frac{de}{dx}\right)^2 dx \\&= \int_{\Omega} \frac{de}{dx} \frac{d(e - \Pi e)}{dx} dx \\&= \sum_{i=1}^n \int_{x_{i-1}}^{x_i} \frac{de}{dx} \frac{d(e - \Pi e)}{dx} dx \\&= \sum_{i=1}^n \int_{x_{i-1}}^{x_i} \left(-\frac{d^2 e}{dx^2}\right)(e - \Pi e) dx + \left[\frac{de}{dx}(e - \Pi e)\right]_{x_{i-1}}^{x_i} \\&= \sum_{i=1}^n \int_{x_{i-1}}^{x_i} \left(-\frac{d^2 e}{dx^2}\right)(e - \Pi e) dx\end{aligned}$$

因为 $-\frac{d^2 e}{dx^2} = -\frac{d^2(u - u_h)}{dx^2} = -\frac{d^2 u}{dx^2} + \frac{d^2 u_h}{dx^2} = f + \frac{d^2 u_h}{dx^2}$

又由Cauchy-Schwarz不等式, 以及标准误差估计, 有

$$\begin{aligned}\left\|\frac{de}{dx}\right\|_{L^2(\Omega)}^2 &= \sum_{i=1}^n \int_{x_{i-1}}^{x_i} \left(f + \frac{d^2 u_h}{dx^2}\right)(e - \Pi e) dx \\&\leq \sum_{i=1}^n \left\|f + \frac{d^2 u}{dx^2}\right\|_{L^2(\Omega_i)} \|e - \Pi e\|_{L^2(\Omega_i)} \\&\leq \sum_{i=1}^n \left\|f + \frac{d^2 u}{dx^2}\right\|_{L^2(\Omega_i)} Ch_i \left\|\frac{de}{dx}\right\|_{L^2(\Omega_i)} \\&= C \sum_{i=1}^n h_i \left\|f + \frac{d^2 u_h}{dx^2}\right\|_{L^2(\Omega_i)} \left\|\frac{de}{dx}\right\|_{L^2(\Omega_i)} \\&\leq C \left(\sum_{i=1}^n h_i^2 \left\|f + \frac{d^2 u_h}{dx^2}\right\|_{L^2(\Omega_i)}^2\right)^{1/2} \left(\sum_{i=1}^n \left\|\frac{de}{dx}\right\|_{L^2(\Omega_i)}^2\right)^{1/2} \\&= C \left(\sum_{i=1}^n h_i^2 \left\|f + \frac{d^2 u_h}{dx^2}\right\|_{L^2(\Omega_i)}^2\right)^{1/2} \left\|\frac{de}{dx}\right\|_{L^2(\Omega)} \quad \text{证毕}\end{aligned}$$

1 两点边值问题

2 有限元方法

3 先验误差估计

4 一维自适应例子

5 二维泊松问题及有限元方法

6 误差估计

$$\begin{cases} -u'' = f & \Omega = (0, L) \\ u'(0) = \kappa_0(u(0) - g_0) \\ u'(L) = -\kappa_L(u(L) - g_L) \end{cases} \quad (6)$$

其中 $L = 1$.

在上述问题中取 $\kappa_0 = 10^6, \kappa_1 = 0, g_0 = 0$, 取 $f(x) = e^{-100(x-0.5)^2}$, 网格情况如下:

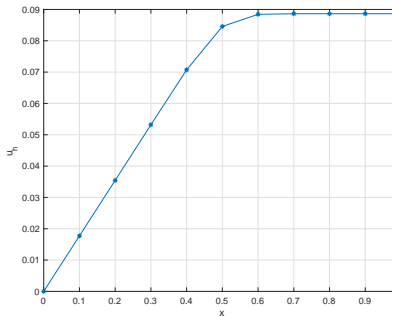


Figure: 均匀分割10

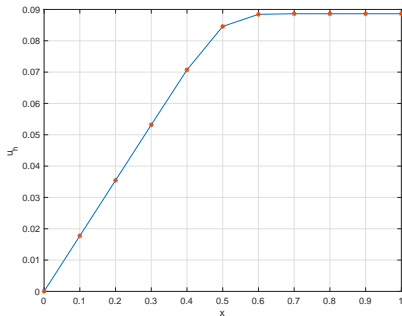


Figure: 自适应初始网格10

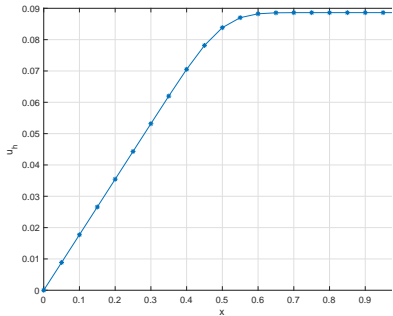


Figure: 均匀剖分20

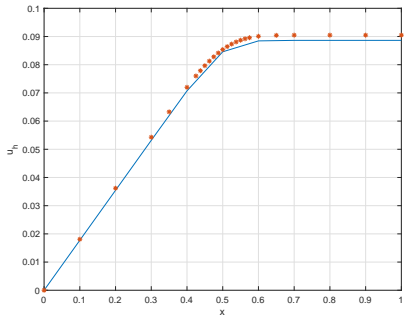


Figure: 自适应网格20

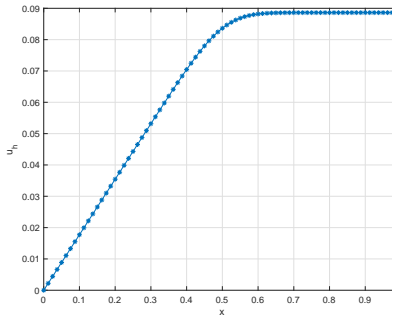


Figure: 均匀剖分80

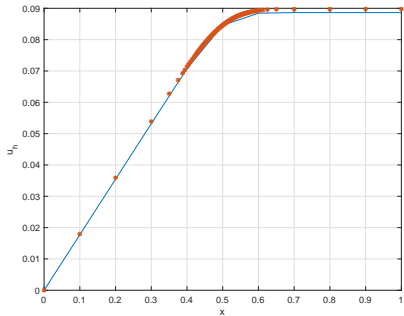


Figure: 自适应网格80

在上述问题中取 $\kappa_0 = 10^6, \kappa_1 = 10^5, g_0 = 0, g_L = 0$,
取 $f(x) = e^{-100(x-0.5)^2}$, 网格情况如下:

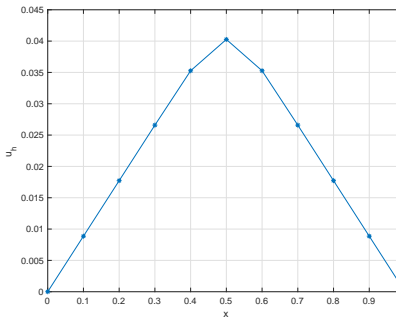


Figure: 均匀剖分10

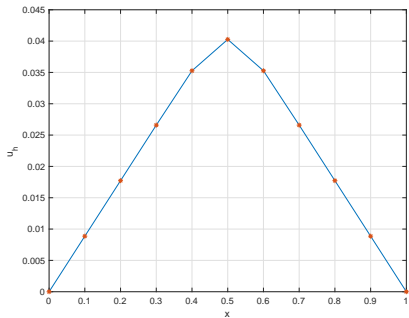


Figure: 自适应初始网格10

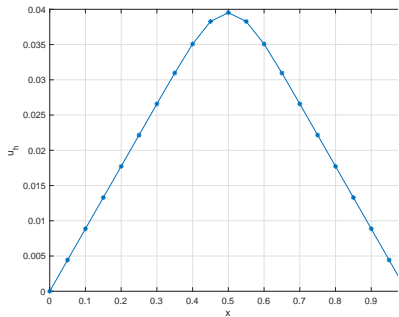


Figure: 均匀剖分20

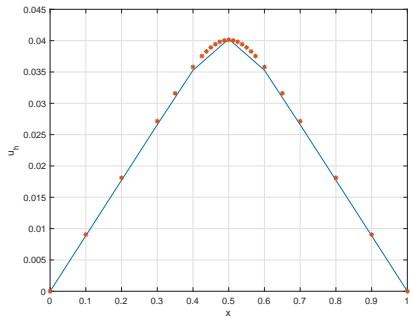


Figure: 自适应网格20

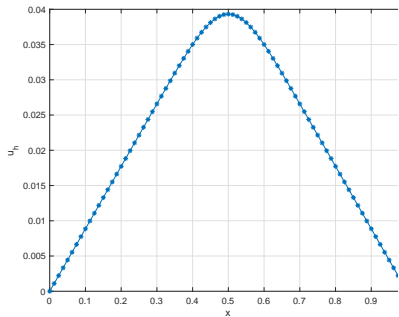


Figure: 均匀剖分80

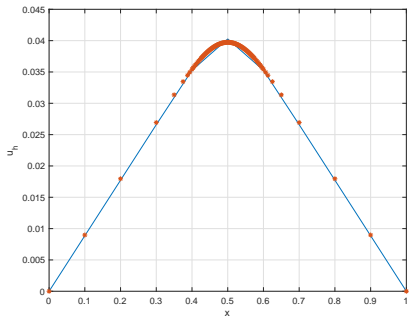


Figure: 自适应网格80

在上述问题中取 $\kappa_0 = 10^6$, $\kappa_1 = 0$, $g_0 = -1$, 取 $f(x) = 0.03(x - 6)^4$, 网格情况如下

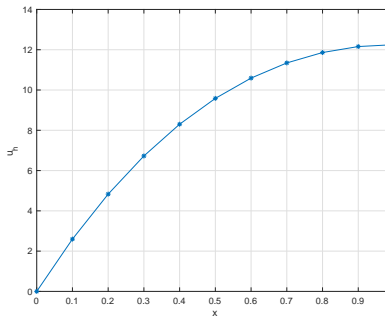


Figure: 均匀剖分10

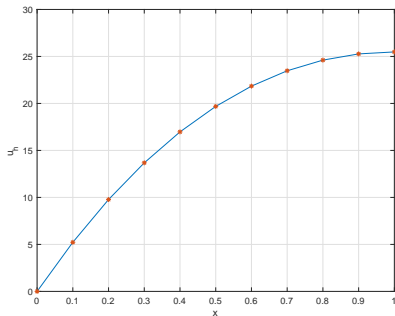


Figure: 自适应初始网格10

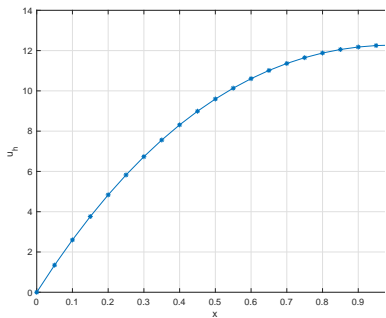


Figure: 均匀剖分20

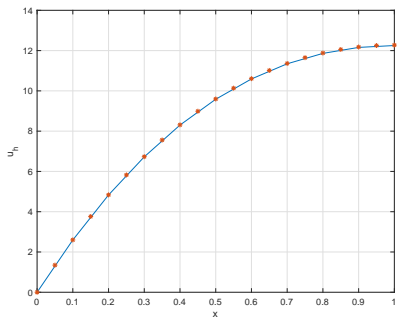


Figure: 自适应网格20

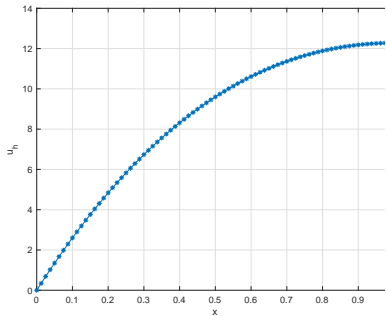


Figure: 均匀剖分80

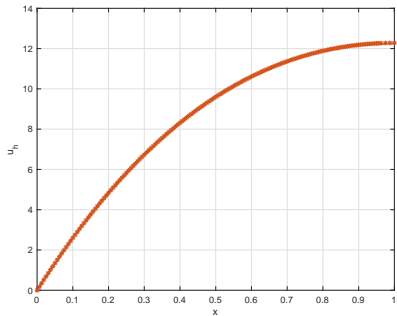


Figure: 自适应网格80

- 1 两点边值问题
- 2 有限元方法
- 3 先验误差估计
- 4 一维自适应例子
- 5 二维泊松问题及有限元方法
- 6 误差估计

二维泊松问题

$$\begin{cases} -\Delta u = f(x, y), & x \in \Omega \\ u|_{\partial\Omega} = 0 \end{cases} \quad (7)$$

定义:

$$H_0^1(\Omega) = \left\{ v : \int_{\Omega} v^2 dx dy + \int_{\Omega} |\nabla v|^2 dx dy < \infty \right\}$$

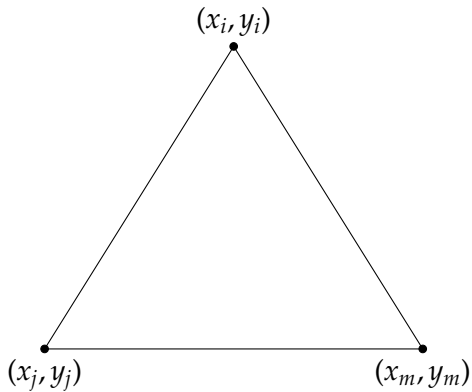
在(7)两边同时乘以 $v(x, y)$, 分步积分有:

$$\int_{\Omega} -\Delta u v dx dy = \int_{\Omega} \nabla u \nabla v dx dy$$

变分问题: 求 $u(x, y) \in H_0^1(\Omega)$, 使得

$$\int_{\Omega} \nabla u \nabla v dx dy = \int_{\Omega} f(x, y) v(x, y) dx dy, \forall v \in H_0^1(\Omega)$$

Figure



求线性函数 $u = ax + by + c$, 使得

$$\begin{cases} ax_i + by_i + c = u_i \\ ax_j + by_j + c = u_j \\ ax_m + by_m + c = u_m \end{cases}$$

这样有:

$$a = \frac{\begin{vmatrix} u_i & y_i & 1 \\ u_j & y_j & 1 \\ u_m & y_m & 1 \end{vmatrix}}{\begin{vmatrix} x_i & y_i & 1 \\ x_j & y_j & 1 \\ x_m & y_m & 1 \end{vmatrix}}, \quad b = \frac{\begin{vmatrix} x_i & u_i & 1 \\ x_j & u_j & 1 \\ x_m & u_m & 1 \end{vmatrix}}{\begin{vmatrix} x_i & y_i & 1 \\ x_j & y_j & 1 \\ x_m & y_m & 1 \end{vmatrix}}, \quad c = \frac{\begin{vmatrix} x_i & y_i & u_i \\ x_j & y_j & u_j \\ x_m & y_m & u_m \end{vmatrix}}{\begin{vmatrix} x_i & y_i & 1 \\ x_j & y_j & 1 \\ x_m & y_m & 1 \end{vmatrix}}$$

也即是

$$a = \frac{1}{2\Delta_e} \left[\begin{vmatrix} y_j & 1 \\ y_m & 1 \end{vmatrix} u_i + \begin{vmatrix} y_m & 1 \\ y_i & 1 \end{vmatrix} u_j + \begin{vmatrix} y_i & 1 \\ y_j & 1 \end{vmatrix} u_m \right]$$

$$b = \frac{1}{2\Delta_e} \left[- \begin{vmatrix} x_j & 1 \\ x_m & 1 \end{vmatrix} u_i - \begin{vmatrix} x_m & 1 \\ x_i & 1 \end{vmatrix} u_j - \begin{vmatrix} x_i & 1 \\ x_j & 1 \end{vmatrix} u_m \right]$$

$$c = \frac{1}{2\Delta_e} \left[\begin{vmatrix} x_j & y_j \\ x_m & y_m \end{vmatrix} u_i + \begin{vmatrix} x_m & y_m \\ x_i & y_i \end{vmatrix} u_j + \begin{vmatrix} x_i & y_i \\ x_j & y_j \end{vmatrix} u_m \right]$$

$$\text{其中: } 2\Delta_e = \begin{vmatrix} x_i & y_i & 1 \\ x_j & y_j & 1 \\ x_m & y_m & 1 \end{vmatrix}.$$

定义:

$$\lambda_i = \frac{1}{2\Delta_e} \left[\begin{vmatrix} y_j & 1 \\ y_m & 1 \end{vmatrix} x - \begin{vmatrix} x_j & 1 \\ x_m & 1 \end{vmatrix} y + \begin{vmatrix} x_j & y_j \\ x_m & y_m \end{vmatrix} \right]$$

$$\lambda_j = \frac{1}{2\Delta_e} \left[\begin{vmatrix} y_m & 1 \\ y_i & 1 \end{vmatrix} x - \begin{vmatrix} x_m & 1 \\ x_i & 1 \end{vmatrix} y + \begin{vmatrix} x_m & y_m \\ x_i & y_i \end{vmatrix} \right]$$

$$\lambda_m = \frac{1}{2\Delta_e} \left[\begin{vmatrix} y_i & 1 \\ y_j & 1 \end{vmatrix} x - \begin{vmatrix} \Delta x_i & 1 \\ x_j & 1 \end{vmatrix} y + \begin{vmatrix} x_i & y_i \\ x_j & y_j \end{vmatrix} \right]$$

则有: $u = u_i\lambda_i + u_j\lambda_j + u_m\lambda_m$. 同时也即是

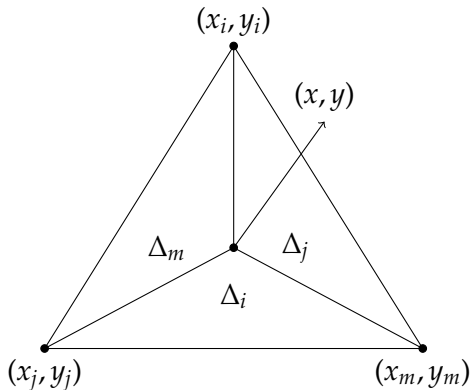
$$\lambda_i = \frac{2\Delta_i}{2\Delta_e}, \lambda_j = \frac{2\Delta_j}{2\Delta_e}, \lambda_m = \frac{2\Delta_m}{2\Delta_e}$$

其中

$$2\Delta_i = \begin{vmatrix} x & y & 1 \\ x_j & y_j & 1 \\ x_m & y_m & 1 \end{vmatrix}, \quad 2\Delta_j = \begin{vmatrix} x & y & 1 \\ x_m & y_m & 1 \\ x_i & y_i & 1 \end{vmatrix}, \quad 2\Delta_m = \begin{vmatrix} x & y & 1 \\ x_i & y_i & 1 \\ x_j & y_j & 1 \end{vmatrix}$$

为下图所示区域面积

Figure



$$\text{满足 } \lambda_i((x_k, y_k)) = \delta_{ik} = \begin{cases} 1, i = k \\ 0, i \neq k \end{cases}.$$

同时还有

$$\begin{aligned}\frac{\partial \lambda_i}{\partial x} &= \frac{1}{2\Delta_e} (y_j - y_m), & \frac{\partial \lambda_i}{\partial y} &= \frac{x_m - x_j}{2\Delta_e} \\ \frac{\partial \lambda_j}{\partial x} &= \frac{1}{2\Delta_e} (y_m - y_i), & \frac{\partial \lambda_j}{\partial y} &= \frac{x_i - x_m}{2\Delta_e} \\ \frac{\partial \lambda_m}{\partial x} &= \frac{y_i - y_j}{2\Delta_e}, & \frac{\partial \lambda_m}{\partial y} &= \frac{x_j - x_i}{2\Delta_e}\end{aligned}$$

与

$$\begin{aligned}1 &= \lambda_i + \lambda_j + \lambda_m \\ x &= x_i \lambda_i + x_j \lambda_j + x_m \lambda_m \\ y &= y_i \lambda_i + y_j \lambda_j + y_m \lambda_m\end{aligned}$$

这样就有

$$\begin{aligned}x &= (x_i - x_m) \lambda_i + (x_j - x_m) \lambda_j + x_m \\ y &= (y_i - x_m) \lambda_i + (y_j - y_m) \lambda_j + y_m\end{aligned}$$

而且

$$\left| \frac{\partial(x, y)}{\partial(\lambda_i, \lambda_j)} \right| = \begin{vmatrix} x_i - x_m & x_j - x_m \\ y_i - y_m & y_j - y_m \end{vmatrix} = 2\Delta_e$$

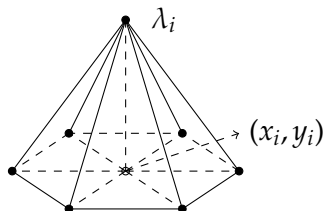
令

$$V_h := \{v \in C^0(\Omega) : v|_T \in P_1(T), \forall T, v|_{\partial\Omega} = 0\}$$

有限元问题: 求 $u_h \in V_h$, 使得

$$\int_{\Omega} \nabla u_h \cdot \nabla v_h dx dy = \int_{\Omega} f v_h dx dy, \forall v_h \in V_h$$

Figure



$u_h = \sum_{j \in \mathcal{N}_0} u_j \lambda_j$, 其中 \mathcal{N}_0 表示内节点的集合.

$$\int_{\Omega} \nabla u_h \cdot \nabla v_h dx dy = \sum_e \int_e \nabla u_h \cdot \nabla v_h dx dy$$

$$u_h|_e = u_i \lambda_i + u_j \lambda_j + u_m \lambda_m = \begin{bmatrix} \lambda_i & \lambda_j & \lambda_m \end{bmatrix} \begin{bmatrix} u_i \\ u_j \\ u_m \end{bmatrix}$$

$$v_h|_e = v_i \lambda_i + v_j \lambda_j + v_m \lambda_m = \begin{bmatrix} v_i & v_j & v_m \end{bmatrix} \begin{bmatrix} \lambda_i \\ \lambda_j \\ \lambda_m \end{bmatrix}$$

于是

$$\begin{aligned}\int_e \nabla u_h \nabla v_h dx dy &= [v_i \ v_j \ v_m] \int_e \begin{bmatrix} \nabla \lambda_i \\ \nabla \lambda_j \\ \nabla \lambda_m \end{bmatrix} \cdot [\nabla \lambda_i \ \nabla \lambda_j \ \nabla \lambda_m] dx dy \begin{bmatrix} u_i \\ u_j \\ u_m \end{bmatrix} \\ &= [v_i \ v_j \ v_m] \int_e \begin{pmatrix} \nabla \lambda_i \cdot \nabla \lambda_i & \nabla \lambda_i \cdot \nabla \lambda_j & \nabla \lambda_i \cdot \nabla \lambda_m \\ \nabla \lambda_j \cdot \nabla \lambda_i & \nabla \lambda_j \cdot \nabla \lambda_j & \nabla \lambda_j \cdot \nabla \lambda_m \\ \nabla \lambda_m \cdot \nabla \lambda_i & \nabla \lambda_m \cdot \nabla \lambda_j & \nabla \lambda_m \cdot \nabla \lambda_m \end{pmatrix} dx dy \begin{bmatrix} u_i \\ u_j \\ u_m \end{bmatrix}\end{aligned}$$

令

$$K_e := \int_e \begin{pmatrix} \nabla \lambda_i \cdot \nabla \lambda_i & \nabla \lambda_i \cdot \nabla \lambda_j & \nabla \lambda_i \cdot \nabla \lambda_m \\ \nabla \lambda_j \cdot \nabla \lambda_i & \nabla \lambda_j \cdot \nabla \lambda_j & \nabla \lambda_j \cdot \nabla \lambda_m \\ \nabla \lambda_m \cdot \nabla \lambda_i & \nabla \lambda_m \cdot \nabla \lambda_j & \nabla \lambda_m \cdot \nabla \lambda_m \end{pmatrix} dx dy$$

这样有

$$\int_{\Omega} \nabla u_h \cdot \nabla v_h dx dy = \sum_e [v_i \ v_j \ v_m] K_e \begin{bmatrix} u_i \\ u_j \\ u_m \end{bmatrix}$$

其中 K_e 称为单元刚度矩阵.(注: $u_i = v_i = 0$, 若 i 为边界节点).

怎样组装总刚度矩阵？

$$\mathbf{K}_e \Rightarrow \begin{pmatrix} k_{ii} & \cdots & k_{ij} & \cdots & k_{im} \\ \vdots & & \vdots & & \vdots \\ k_{ji} & \cdots & k_{jj} & \cdots & k_{jm} \\ \vdots & & \vdots & & \vdots \\ k_{mi} & \cdots & k_{mj} & \cdots & k_{mm} \end{pmatrix}, \quad \mathbf{K}_{N \times N}$$

右端项

$$\int_e f v_h dx dy = [v_i \ v_j \ v_m] \int_e \begin{pmatrix} f \lambda_i \\ f \lambda_j \\ f \lambda_m \end{pmatrix} dx dy$$

令

$$F_e = \int_e \begin{pmatrix} f \lambda_i \\ f \lambda_j \\ f \lambda_m \end{pmatrix} dx dy$$

$$F_e \Rightarrow \begin{pmatrix} f_i \\ \vdots \\ f_j \\ \vdots \\ f_m \end{pmatrix} F_N$$

$$A_{N \times N} U_N = F_N$$

- 1 两点边值问题
- 2 有限元方法
- 3 先验误差估计
- 4 一维自适应例子
- 5 二维泊松问题及有限元方法
- 6 误差估计**

类似一维问题, 有

$$\int_{\Omega} (\nabla u - \nabla u_h) \cdot (\nabla u - \nabla u_h) dx dy = \inf_{v_h \in V_h} \int_{\Omega} |\nabla u - \nabla u_h|^2 dx dy$$

定义插值函数 $I_h u$, 使得

$$I_h u = \sum_{i \in \mathcal{N}} u(x_i, y_i) \lambda_i$$

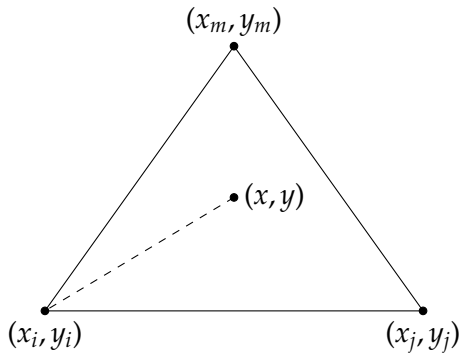
在 Δ_e 上(顶点为 $(x_i, y_i), (x_j, y_j), (x_m, y_m)$), 有

$$\begin{aligned} u(x, y) - I_h u(x, y) &= u(x, y) - u(x_i, y_i) \lambda_i - u(x_j, y_j) \lambda_j \\ &\quad - u(x_m, y_m) \lambda_m \end{aligned}$$

令 $A_i = (x_i, y_i)$, $A = (x, y)$, 作变点展开

$$\begin{aligned} u(x_i, y_i) &= u(x, y) + \nabla u(x, y) \cdot (x_i - x, y_i - y)^T \\ &\quad + \frac{(x_i - x)^2}{2} \int_0^1 s \partial_{xx} u(A_i + s(A - A_i)) ds \\ &\quad + \frac{(y_i - y)^2}{2} \int_0^1 s \partial_{yy} u(A_i + s(A - A_i)) ds \\ &\quad + (x_i - x)(y_i - y) \int_0^1 s \partial_{xy} u(A_i + s(A - A_i)) ds \end{aligned}$$

Figure



作极坐标变化, 得

$$\begin{aligned} u(x_i, y_i) &= u(x, y) + \nabla u(x, y) \cdot (x_i - x, y_i - y)^T \\ &+ \frac{(\cos \theta_i)^2}{2} \int_0^{l_{i(x,y)}} \partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) t dt \\ &+ \frac{(\sin \theta_i)^2}{2} \int_0^{l_{i(x,y)}} \partial_{yy} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) t dt \\ &+ \cos \theta_i \sin \theta_i \int_0^{l_{i(x,y)}} \partial_{xy} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) t dt. \end{aligned}$$

这样

$$\begin{aligned}\nabla u - \nabla I_h u &= \nabla u - \sum_{i=1}^3 u(x_i, y_i) \nabla \lambda_i \\&= - \sum_{i=1}^3 \nabla \lambda_i \left(\frac{(\cos \theta_i)^2}{2} \int_0^{l_{i(x,y)}} \partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) dt \right. \\&\quad \left. + \frac{(\sin \theta_i)^2}{2} \int_0^{l_{i(x,y)}} \partial_{yy} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) dt \right. \\&\quad \left. + \cos \theta_i \sin \theta_i \int_0^{l_{i(x,y)}} \partial_{xy} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) dt \right).\end{aligned}$$

这里用到

$$\sum_{i=1}^3 u(x, y) \nabla \lambda_i = 0, \quad \sum_{i=1}^3 \nabla u \cdot (x, y)^T \nabla \lambda_i = 0,$$

$$\begin{aligned} \sum_{i=1}^3 \nabla u \cdot (x_i, y_i)^T \nabla \lambda_i &= \frac{\partial u}{\partial x} \sum x_i \nabla \lambda_i + \frac{\partial u}{\partial y} \sum y_i \nabla \lambda_i \\ &= \frac{\partial u}{\partial x} \nabla \left(\sum x_i \lambda_i \right) + \frac{\partial u}{\partial y} \nabla \left(\sum y_i \lambda_i \right) \\ &= \frac{\partial u}{\partial x} (1, 0)^T + \frac{\partial u}{\partial y} (0, 1)^T = \nabla u. \end{aligned}$$

于是

$$\begin{aligned} & \int_e |\nabla u - \nabla I_h u|^2 dx dy \\ &= 3 \sum_{i=1}^3 |\nabla \lambda_i|^2 \\ & \quad \times \int_{\alpha_i}^{\beta_i} \int_0^{l_{\theta_i}} \left(\frac{(\cos \theta_i)^2}{2} \int_0^{l_{i(x,y)}} \partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) dt \right. \\ & \quad + \frac{(\sin \theta_i)^2}{2} \int_0^{l_{i(x,y)}} \partial_{yy} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) dt \\ & \quad \left. + \cos \theta_i \sin \theta_i \int_0^{l_{i(x,y)}} \partial_{xy} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) dt \right)^2 l_i dl_i d\theta_i \\ & \quad (l_i = l_{i(x,y)}). \end{aligned}$$

下面估计上式中的每一项

$$\begin{aligned} & \int_0^{l_{i(x,y)}} \partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) t dt \\ & \leq \left(\int_0^{l_{i(x,y)}} (\partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)))^2 t dt \right)^{1/2} \left(\int_0^{l_{i(x,y)}} t dt \right)^{1/2} \\ & \leq \left(\int_0^{l_{\theta_i}} (\partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)))^2 t dt \right)^{1/2} \left(\int_0^{l_{i(x,y)}} t dt \right)^{1/2}. \end{aligned}$$

因此

$$\begin{aligned}& \int_{\alpha_i}^{\beta_i} \int_0^{l_{\theta_i}} \left(\frac{(\cos \theta_i)^2}{2} \int_0^{l_{i(x,y)}} \partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)) t dt \right)^2 l_i dl_i d\theta_i \\& \leq \int_{\alpha_i}^{\beta_i} \frac{(\cos \theta_i)^4}{4} \int_0^{l_{\theta_i}} (\partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)))^2 t dt \\& \quad \times \left(\int_0^{l_{\theta_i}} \int_0^{l_{i(x,y)}} t dt l_i dl_i \right) d\theta_i \\& = \frac{1}{32} \int_{\alpha_i}^{\beta_i} l_{\theta_i}^4 (\cos \theta_i)^4 \int_0^{l_{\theta_i}} (\partial_{xx} u((x_i, y_i) - t(\cos \theta_i, \sin \theta_i)))^2 t dt d\theta_i \\& \leq \max_{\alpha_i \leq \theta_i \leq \beta_i} \frac{l_{\theta_i}^4 (\cos \theta_i)^4}{32} \|\partial_{xx} u\|_{0,e}^2.\end{aligned}$$

其它项可以类似处理, 这样

$$\int_e |\nabla u - \nabla I_h u|^2 dx dy \leq C \|\nabla^2 u\|_{0,e}^2 \sum_{i=1}^3 |\nabla \lambda_i|^2 h^4.$$

谢谢！