

# MATH714 HW2

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## 1 Problem A

### 1.1 (a)

Assume  $v = \sum_{i=1}^n a_i w_i$ , so  $\langle v, w_i \rangle = a_i \|w_i\|^2$ , which means  $\frac{\langle v, w_i \rangle}{\|w_i\|^2} w_i = a_i w_i$ .

From the equation above, we can get  $v = \sum_{i=1}^n \frac{\langle v, w_i \rangle}{\|w_i\|^2} w_i$

### 1.2 (b)

First for  $n = 1$ ,  $\langle p_1, p_0 \rangle = \langle p_1, r_0 \rangle = \langle r_1, r_0 \rangle - \langle \frac{\langle r_0, r_0 \rangle}{\|r_0\|^2} r_0, r_0 \rangle$ .

$$\begin{aligned} r_1 &= f - Au_1, r_0 = f - Au_0 \\ r_1^T A r_0 &= (f - Au_1)^T A (f - Au_0) \end{aligned}$$

So  $\langle p_1, p_0 \rangle = 0$ .

Now when  $\langle p_n, p_j \rangle = 0$  for any  $0 \leq j < n$ ,

for  $j < n$ ,  $\langle p_{n+1}, p_j \rangle = \langle r_{n+1} - \sum_{i=1}^n \frac{\langle r_{n+1}, p_i \rangle}{\|p_i\|^2} p_i, p_j \rangle = \langle r_{n+1}, p_j \rangle = \langle f - Au_n, p_j \rangle = 0$ .

And  $\langle p_{n+1}, p_n \rangle = \langle r_{n+1} - \frac{\langle r_{n+1}, p_n \rangle}{\|p_n\|^2} p_n, p_n \rangle$ . Like  $\langle p_1, p_0 \rangle$ , it equals to 0.

### 1.3 (c)

#### 1.3.1 i

From (a) we can get  $v = \sum_{j=1}^N \frac{\langle v, \phi_j \rangle}{\|\phi_j\|^2} \phi_j$  and  $w = \sum_{j=1}^N \frac{\langle w, \phi_j \rangle}{\|\phi_j\|^2} \phi_j$ .

So  $\langle Av, w \rangle = \langle \sum_{j=1}^N \frac{\langle v, \phi_j \rangle}{\|\phi_j\|^2} A\phi_j, \sum_{j=1}^N \frac{\langle w, \phi_j \rangle}{\|\phi_j\|^2} \phi_j \rangle = \sum_{j=1}^N \lambda_j \langle v, \phi_j \rangle \langle \phi_j, w \rangle$

#### 1.3.2 ii

Since  $A$  is positive definite,  $0 < \phi_n^T A \phi_n = \lambda_n \phi_n^T \phi_n$ . Obviously, we can get  $\lambda_n > 0$  from  $\phi_n^T \phi_n \geq 0$

### 1.3.3 iii

From i we can know  $\langle Av, v \rangle = \sum_{n=1}^N \lambda_n \langle v, \phi_n \rangle^2 = \sum_{n=1}^N \lambda_n \|v\|^2 \leq \lambda_N \|v\|^2$  where  $v^* = \frac{v}{\|v\|}$

Since  $\{\phi_n\}$  are orthogonal,  $\lambda_1 \leq \sum_{n=1}^N \langle v^*, \phi_n \rangle^2 \leq \lambda_N$

### 1.3.4 iv

$\|Av\| = \sum_{i=1}^N \langle v, \phi_i \rangle \lambda_i \phi_i \leq \lambda_N \|v\|$  as iii.

## 1.4 (d)

$$p_{n+1} = r_{n+1} + \beta_n p_n = r_n - \alpha_n w_n + \beta_n p_n = r_n - \alpha_n A p_n + \beta_n p_n = p_n - \beta_{n-1} p_{n-1} - \alpha_n A p_n + \beta_n p_n = (1 + \beta_n) p_n - \alpha_n A p_n - \beta_{n-1} p_{n-1}$$

## 1.5 (e)

Now the character polynomial of  $A$  is  $p(\lambda) = \lambda^n + \sum_{i=0}^{n-1} a_i \lambda^i$ .

From Cayley-Hamilton theorem,  $p(A) = 0$ , which means  $A^n = -\sum_{i=0}^{n-1} a_i A^i$ , where  $A^0 = I_n$ .

## 1.6 (f)

### 1.6.1 i

$$e_{n+1} = u_{n+1} - u = u_n - u - \alpha A(u_n - u) + \alpha f - \alpha A u = e_n - \alpha A e_n = (I - \alpha A) e_n$$

### 1.6.2 ii

$$\|e_{n+1}\| = \|(I - \alpha A) e_n\| \leq \|I - \alpha A\| \|e_n\| \leq \max\{\lambda(I - \alpha A)\} \|e_n\| = \max\{1 - \alpha \lambda_j\} \|e_n\|$$

### 1.6.3 iii

Now  $\rho = |1 - \alpha \lambda_1|$  or  $|1 - \alpha \lambda_N|$ .

for each case,  $\alpha = \frac{1}{\lambda}$  minimize  $\rho$ . But to minimize both cases, it's to minimize  $(1 - \alpha \lambda_j)^2$ . After calculating its Jacobi matrix, we can get  $\alpha = \frac{2}{\lambda_1 + \lambda_N}$

### 1.6.4 iv

$$\text{Now } \max |1 - \alpha \lambda_j| = \max \left| \frac{C+c-2\lambda_j}{C+c} \right| \leq \max \left| \frac{C-c}{C+c} \right| = \frac{C-c}{C+c}$$

## 1.7 g

### 1.7.1 i

$$r_1 = f - A u_1 = f - A(u_0 + \alpha_0(f - A u_0)) = f - A u_0 - \alpha_0 A r_0 = r_0 - \alpha_0 A r_0$$

### 1.7.2 ii

Just like i.

### 1.7.3 iii

$$Aq_0 = A \frac{r_0}{\|r_0\|} = \frac{r_0 - r_0 + \alpha_0 A r_0}{\alpha_0 \|r_0\|} = \frac{r_0 - r_1}{\alpha_0 \|r_0\|} = \frac{r_0}{\alpha_0 \|r_0\|} - \frac{r_1}{\alpha_0 \|r_0\|} = \frac{1}{\alpha_0} q_0 - \frac{\sqrt{\beta_0}}{\alpha_0} q_1$$

$$Aq_n = A \frac{r_n}{\|r_n\|} = \frac{r_{n-1} - \alpha_{n-1} A r_{n-1} + \frac{\alpha_{n-1} \beta_{n-2}}{\alpha_{n-2}} (r_{n-1} - r_{n-2})}{\|r_n\|} = -\delta_{n-1} q_{n-1} + \gamma_n q_n - \delta_n q_{n+1}$$

### 1.7.4 iv

From (iii),  $Aq_0 = (Q_n T_n)_0 + 0$ ,  $Aq_n = (Q_n T_n)_n - \delta_{n-1} q_n e_n^T$

### 1.7.5 v

$$Q_n^T A Q_n = Q_n^T Q_n T_n - Q_n^T \delta_{n-1} q_n e_n^T = T_n$$

## 2 Problem B

The result of the problem is  $N = 100$ , and the link to the code is:

<https://github.com/Skystaryou/Study/tree/main/MA>

TH714\_MethodOfComputatnlMathI/linearInterpolant\_HW2

Just run "main.m".

## 3 Problem C

### 3.1 (a)

Now the scheme is  $\frac{u_{t+1}^{x,y} - 2u_t^{x,y} + u_{t-1}^{x,y}}{\Delta t^2} = \frac{u_t^{x+1,y} - 2u_t^{x,y} + u_t^{x-1,y}}{\Delta x^2} + \frac{u_t^{x,y+1} - 2u_t^{x,y} + u_t^{x,y-1}}{\Delta x^2}$

Dirichlet boundary condition: we can simply let  $u(0, y, t) = u(1, y, t) = u(x, 0, t) = u(x, 1, t) = 0$ .

Initial condition:  $u(x, y, 0) = 0$  and  $\frac{u(x, y, \Delta t) - u(x, y, 0)}{\Delta t} = f(x)f(y)$

The link to the code is:

[https://github.com/Skystaryou/Study/tree/main/MATH714\\_MethodOfComputatnlMathI/2DWave\\_HW2](https://github.com/Skystaryou/Study/tree/main/MATH714_MethodOfComputatnlMathI/2DWave_HW2)

thodOfComputatnlMathI/2DWave\_HW2

Just run main.m and the plot will show up.

### 3.2 (b)

Now  $f'(u_{t+1}) \frac{u_{t+2} - u_t}{2\Delta t}$ ,  $f'(u_{t-1}) \frac{u_t - u_{t-2}}{2\Delta t}$ .

So the scheme:  $\frac{u_{t+2} - 2u_t + u_{t-2}}{4\Delta t^2} = \lambda u(t)$