

## 18.330 Problem Set 6

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**Problem 1** (Source code for this problem attached as `problem1.m`, `problem1.py`, and `mesh.py`)

The construction of the mesh is as follows. At each mesh point  $(i, j)$ , the differential equation can be approximated by

$$u_{i+1,j} + u_{i,j+1} + u_{i-1,j} + u_{i,j-1} - 4u_{i,j} = \left(\frac{1}{1+N}\right)^2,$$

and we can construct an  $N^2 \times N^2$  matrix  $M_N$  where  $M_{(a-1)N+b, (c-1)N+d}$  represents the coefficient of  $u_{c,d}$  in the equation centered at  $u_{a,b}$ . Let  $b_N$  be an  $N^2$ -tall column vector with each entry  $\frac{1}{(N+1)^2}$ . For example,

$$M_2 = \begin{pmatrix} -4 & 1 & 1 & 0 \\ 1 & -4 & 0 & 1 \\ 1 & 0 & -4 & 1 \\ 0 & 1 & 1 & -4 \end{pmatrix} \begin{pmatrix} u_{1,1} \\ u_{1,2} \\ u_{2,1} \\ u_{2,2} \end{pmatrix} = \begin{pmatrix} \frac{1}{9} \\ \frac{1}{9} \\ \frac{1}{9} \\ \frac{1}{9} \end{pmatrix},$$

We solve  $M_N x = b_N$  using software for  $N = 3, 7, 15, 31, 63$  and extrapolate  $u_{(\frac{N+1}{2}, \frac{N+1}{2})}$  four times to get, to 10 digits, **-0.07367135327**. (Both MATLAB and SciPy give the same digits.)

### Problem 2

We first define a few terms. Let

$$M_n = \begin{pmatrix} x_{n-1} - \frac{c_{n-1}}{c_n} & -\frac{c_{n-1}}{c_n} & \cdots & \cdots & -\frac{c_0}{c_n} \\ 1 & x_{n-2} & 0 & \cdots & 0 \\ 0 & 1 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & 0 \\ 0 & \cdots & 0 & 1 & x_0 \end{pmatrix},$$

and  $f_n(x) = \det(xI_n - M_n)$ . We will prove that  $f_n(x) = \frac{1}{c_n} P_n(x)$ , where

$$P_n(x) = c_0 + c_1(x - x_0) + c_2(x - x_0)(x - x_1) + \cdots + c_n(x - x_0) \cdots (x - x_{n-1}),$$

and that therefore the eigenvalues of  $M_n$  are the roots of  $P_n(x)$  (requirement (a)). Note also that  $M_n$  is in upper Hessenberg form (requirement (b)) and that each entry of the matrix can be computed by simple arithmetic on the  $\{c_i\}$  and  $\{x_i\}$  (requirement (c)).

**Conjecture.**  $M_n$  is the companion matrix to  $P_n$ . That is,

$$f_n(x) = \frac{1}{c_n} P_n(x).$$

*Proof.* By induction. Our anchor is at  $n = 2$ .

$$\begin{aligned} f_2(x) &= \det \begin{pmatrix} x - x_1 + \frac{c_1}{c_2} & \frac{c_0}{c_2} \\ -1 & x - x_0 \end{pmatrix} = (x - x_1 + \frac{c_1}{c_2})(x - x_0) - (\frac{c_0}{c_2})(-1) \\ &= (x - x_1)(x - x_0) + \frac{c_1}{c_2}(x - x_0) + \frac{c_0}{c_2} = \frac{1}{c_2}P_2(x) \end{aligned}$$

Now we must show that  $f_{n-1}(x) = \frac{1}{c_{n-1}}P_{n-1}(x) \Rightarrow f_n(x) = \frac{1}{c_n}P_n(x)$ . Consider  $f_n$ :

$$f_n = \det \begin{pmatrix} x - x_{n-1} + \frac{c_{n-1}}{c_n} & \frac{c_{n-1}}{c_n} & \dots & \dots & \frac{c_0}{c_n} \\ -1 & x - x_{n-2} & 0 & \dots & 0 \\ 0 & -1 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & -1 & x - x_0 \end{pmatrix}$$

We expand the determinant around the terms at  $(1, 1)$  and  $(2, 1)$ .

$$\begin{aligned} f_n &= \left( (x - x_{n-1}) + \frac{c_{n-1}}{c_n} \right) \det \begin{pmatrix} x - x_{n-2} & 0 & \dots & \dots & 0 \\ -1 & \ddots & \ddots & & \vdots \\ 0 & \ddots & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & -1 & x - x_0 \end{pmatrix} \\ &\quad - (-1) \det \begin{pmatrix} 0 + \frac{c_{n-2}}{c_n} & \dots & \dots & \dots & \frac{c_0}{c_n} \\ -1 & x - x_{n-3} & 0 & \dots & 0 \\ 0 & -1 & \ddots & \ddots & \vdots \\ \vdots & \ddots & \ddots & \ddots & 0 \\ 0 & \dots & 0 & -1 & x - x_0 \end{pmatrix} \end{aligned}$$

The determinant of the matrix on the left is simply  $\prod_{i=0}^{n-2} (x - x_i)$ . The matrix on the right is, by assumption, simply  $f_{n-1}(x)$  except with the  $(x - x_{n-1})$  term equal to 0, and an  $c_n$  element where the  $c_{n-1}$  element usually is. Thus,

$$\begin{aligned} f_n &= \left( (x - x_{n-1}) + \frac{c_{n-1}}{c_n} \right) \prod_{i=0}^{n-2} (x - x_i) + \left( \frac{c_0}{c_n} + \frac{c_1}{c_n}(x - x_0) + \dots + \frac{c_{n-2}}{c_n}(x - x_0) \dots (x - x_{n-3}) \right) \\ &= (x - x_{n-1}) \dots (x - x_0) + \frac{c_{n-1}}{c_n}(x - x_{n-2}) \dots (x - x_0) + \dots + \frac{c_1}{c_n}(x - x_0) + \frac{c_0}{c_n} = \frac{1}{c_n}P_n(x), \end{aligned}$$

which is what we wanted to show.  $\square$

Thus,  $M_n$  as defined above is the companion matrix for the Newton form polynomial  $P_n$ . As an extra check on our solution, we note that when  $x_0 = \dots = x_{n-1} = 0$ , then  $M_n$  reduces to the companion matrix for  $c_0 + c_1x + \dots + c_nx^n$  given in the problem set statement.

**Problem 3** (Source code for this problem attached as problem3.py, problem3.m, and matlab-eigenvalues.txt)

We define some terms. Let

$$M_n = \begin{pmatrix} a_1 & b_1 & 0 & \dots & 0 \\ b_1 & a_2 & b_2 & \ddots & \vdots \\ 0 & b_2 & \ddots & \ddots & 0 \\ \vdots & \ddots & \ddots & a_{n-1} & b_{n-1} \\ 0 & \dots & 0 & b_{n-1} & a_n \end{pmatrix},$$

where  $a_i = 0$ ,  $b_i^2 = \frac{i^2}{(2i-1)(2i+1)}$  and  $f_n(x) = \det(xI_n - M_n)$ . We will prove that  $f_n(x) = \alpha_n P_n(x)$ , where

$$\alpha_n = \prod_{i=1}^n \frac{n}{2n-1},$$

and  $P_n(x)$  is the  $n^{\text{th}}$  Legendre polynomial, which can be defined recursively by

$$P_n(x) = x \frac{2n-1}{n} P_{n-1}(x) - \frac{n-1}{n} P_{n-2}(x),$$

with  $P_0 = 1$  and  $P_1 = x$ .

**Conjecture.**

$$f_n(x) = \alpha_n P_n(x).$$

*Proof.* By induction. Our anchors are at  $n = 1$  and  $n = 2$ .

$$f_1(x) = x = \alpha_1 P_1(x)$$

$$f_2(x) = \det \begin{pmatrix} x & -\frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{3}} & x \end{pmatrix} = x^2 - \frac{1}{3} = \frac{2}{3} P_2(x) = \alpha_2 P_2(x)$$

Now we must show that  $f_{n-2}(x) = \alpha_{n-2} P_{n-2}(x)$  and  $f_{n-1}(x) = \alpha_{n-1} P_{n-1}(x) \Rightarrow f_n(x) = \alpha_n P_n(x)$ . Consider the relation between the determinants of a symmetric tridiagonal matrix given in the statement of the problem. We know that

$$\det(xI_n - M_n) = (x - a_n) \det(xI_{n-1} - M_{n-1}) - (-b_{n-1})^2 \det(xI_{n-2} - M_{n-2}).$$

If we set  $a_n = 0$ , and let  $b_{n-1}^2 = \frac{(n-1)^2}{(2n-1)(2n-3)}$ , we have

$$f_n(x) = x f_{n-1}(x) - \frac{(n-2)^2}{(2n-1)(2n-3)} f_{n-2}(x)$$

$$f_n(x) = x \alpha_{n-1} P_{n-1}(x) - \frac{(n-1)^2}{(2n-1)(2n-3)} \alpha_{n-2} P_{n-2}(x)$$

We extract an  $\alpha_n = \frac{n}{2n-1} \alpha_{n-1} = \frac{n}{2n-1} \frac{n-1}{2n-3} \alpha_{n-2}$  term:

$$f_n(x) = \alpha_n \left( x \frac{2n-1}{n} P_{n-1}(x) - \frac{(n-1)^2}{(2n-1)(2n-3)} \frac{(2n-1)}{n} \frac{(2n-3)}{(n-1)} P_{n-2}(x) \right)$$

$$f_n(x) = \alpha_n \left( x \frac{2n-1}{n} P_{n-1}(x) - \frac{n-1}{n} P_{n-2}(x) \right),$$

and by the recursive definition of the Legendre polynomial,

$$f_n(x) = \alpha_n P_n(x),$$

which is what we wanted to show.  $\square$

Thus, the eigenvalues of  $M_{32}$  (with  $a_i = 0$  and  $b_i^2 = \frac{i^2}{(2i-1)(2i+1)}$ ) are the roots of  $P_{32}$ . After numerically calculating and sorting the eigenvalues, we get the following list:

-0.99726386	-0.98561151	-0.96476226	-0.93490608
-0.89632116	-0.84936761	-0.7944838	-0.73218212
-0.66304427	-0.58771576	-0.50689991	-0.42135128
-0.3318686	-0.23928736	-0.14447196	-0.04830767
0.04830767	0.14447196	0.23928736	0.3318686
0.42135128	0.50689991	0.58771576	0.66304427
0.73218212	0.7944838	0.84936761	0.89632116
0.93490608	0.96476226	0.98561151	0.99726386

These eigenvalues are calculated using SciPy's `linalg.eig()` function. As given,  $\sim .239287$  is in fact one of the zeros. (For comparison's sake, I've attached a printout of MATLAB's output of the eigenvalues.)

**Problem 4** (*Source code attached as problem4.py*)

Let  $p_N(x) = c_0 + \dots + c_N x^N$  be the degree  $N$  minimax polynomial of  $\sin(x)$  on  $[0, \frac{\pi}{2}]$ , and let

$$\epsilon_N = \max_{0 \leq x \leq \frac{\pi}{2}} | \sin(x) - p_N(x) |.$$

We want to find  $p_7(x)$ . We first note that by choosing seven points  $\{x_i\}$  (with  $x_0 = 0$  and  $x_8 = \frac{\pi}{2}$ ) such that

$$\sin(x_i) - p_7(x_i) = (-1)^i h, \quad 0 \leq i \leq 8,$$

we can write nine equations in  $(c_0, \dots, c_8, h)$ :

$$\begin{aligned} c_0 + c_1 x_0 + \dots + c_7 x_0^7 + h &= \sin(x_0) \\ c_0 + c_1 x_1 + \dots + c_7 x_1^7 - h &= \sin(x_1) \\ &\vdots \\ c_0 + c_1 x_8 + \dots + c_7 x_8^7 + h &= \sin(x_8) \end{aligned}$$

These correspond to the matrix equation

$$\begin{pmatrix} 1 & \dots & x_0^7 & 1 \\ 1 & \dots & x_1^7 & -1 \\ \vdots & & \vdots & \vdots \\ 1 & \dots & x_8^7 & 1 \end{pmatrix} \begin{pmatrix} c_0 \\ \vdots \\ c_7 \\ h \end{pmatrix} = \begin{pmatrix} \sin(x_0) \\ \sin(x_1) \\ \vdots \\ \sin(x_8) \end{pmatrix}$$

We can solve this using software. Thus, given a set of seven points, we can obtain a seventh-degree polynomial that misses  $\sin(x)$  by  $h$  at each of the points. We intend to find the first 8 digits of the  $\{c_i\}$ . We make an initial guess  $\{x_i\}$ , and then refine our guess (a better guess will result in a lower  $\epsilon_7$ ) until we have the required number of digits. The method of refining out initial guess is as follows:

Given a set  $\{x_i\}$  (and an  $h$  and  $\{c_i\}$  derived from it), calculate  $\epsilon_N$ , and also the point  $x'$  at which  $\sin(x') - (c_0 + \dots + c_7x'^7) = \epsilon_N$ . Since the calculated value of  $h$  is typically smaller than  $\epsilon_N$ , if we move one of the  $x_i$  to  $x'$ , we will eliminate this source of error. We decide to move the  $x_i$  closest to  $x'$ .

With this method in mind, and the initial guess  $\{x_i\} = .2, .4, .6, .8, 1, 1.2, 1.4$ , we follow these steps:

Step 1. Given  $\{x_i\}$ , calculate  $\{c_i\}$  to construct a polynomial  $p_7(x)$ .

Step 2. Calculate  $\epsilon_7$ , the maximum error between  $\sin(x)$  and  $p_7(x)$ . Note where the error is happening as  $x'$ .

Step 3. Move the closest  $x_i$  to  $x'$ .

Step 4. Go back to Step 1, and repeat until the  $\{c_i\}$  converge.

In less than twenty iterations, the process settles on (to as many digits as given by SciPy):

$$\begin{pmatrix} x_0 \\ x_1 \\ x_2 \\ x_3 \\ x_4 \\ x_5 \\ x_6 \\ x_7 \\ x_8 \end{pmatrix} = \begin{pmatrix} 0 \\ 0.061204508077236347 \\ 0.23474608626153651 \\ 0.49251533428490585 \\ 0.79392758745194458 \\ 1.092861124015301 \\ 1.344620505292353 \\ 1.5121082344331855 \\ \frac{\pi}{2} \end{pmatrix}, \begin{pmatrix} c_0 \\ c_1 \\ c_2 \\ c_3 \\ c_4 \\ c_5 \\ c_6 \\ c_7 \\ h \end{pmatrix} = \begin{pmatrix} -.0000000195367731583 \\ 1.00000155323 \\ -.0000202292837008 \\ -.166566787675 \\ -.000239703970351 \\ .00863920572651 \\ -.000205700049894 \\ -.000137323212602 \\ .0000000195367731583 \end{pmatrix}$$

Our calculated  $\epsilon_7$  is  $.000000019536773532280449$ , which agrees with  $h$  to eight significant digits. This is reassuring, because we know that by the definition of the minimax function,  $h = \epsilon_7$ . To eight digits, our minimax function of degree 7 of  $\sin(x)$  on  $[0, \frac{\pi}{2}]$  is

$$\begin{aligned} p_7(x) = & (-1.9536773 \cdot 10^{-8}) + (1.0000015)x + (-2.0229283 \cdot 10^{-5})x^2 \\ & + (-.16656678)x^3 + (-2.3970397 \cdot 10^{-4})x^4 + (8.6392057 \cdot 10^{-3})x^5 \\ & + (-2.0570005 \cdot 10^{-4})x^6 + (-1.3732321 \cdot 10^{-4})x^7. \end{aligned}$$

(As an afterthought, this is somewhat close to the seventh Taylor polynomial of  $\sin(x)$ , which is, to eight digits,  $x - .16666666x^3 + .0083333333x^5 - 0.00019841269x^7$ . Well, more in the first terms than in the latter ones.)