Math 151BH - HW3

Will Firmin

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Problem 1:

 $\langle u_i, u_j \rangle = \langle \frac{Av_i}{\sigma_i}, \frac{Av_j}{\sigma_j} \rangle$ $= (\frac{Av_i}{\sigma_i})^* (\frac{Av_j}{\sigma_j})$ $= \frac{1}{\sigma_i \sigma_j} v_i^* A^* A v_j$ $= \frac{\lambda_j}{\sigma_i \sigma_j} v_i^* v_j$ $= \frac{\lambda_j}{\sigma_i \sigma_j} \langle v_i, v_j \rangle$ $= \begin{cases} 0 \text{ if } i \neq j \text{ by orthogonality} \\ \frac{\lambda_i}{\sigma_i^2} = 1 \text{ if } i = j \text{ by orthonormality} \end{cases}$

(b) Let $\bar{V}, \bar{U}, \bar{\Sigma}$ be the added columns.

$$\begin{split} AV &= A \begin{bmatrix} \hat{V}_r & \bar{V} \end{bmatrix} \\ &= \begin{bmatrix} A\hat{V}_r & A\bar{V} \end{bmatrix} \\ &= \begin{bmatrix} \hat{U}_r\hat{\Sigma}_r & A\bar{V} \end{bmatrix} \\ &\Rightarrow A = AVV^* = \begin{bmatrix} \hat{U}_r\hat{\Sigma}_r & A\bar{V} \end{bmatrix} \begin{bmatrix} \hat{V}_r^* \\ \bar{V}_r^* \end{bmatrix} \\ &= \hat{U}_r\hat{\Sigma}_r\hat{V}_r^* + A\bar{V}\bar{V}^* \\ &= \hat{U}_r\hat{\Sigma}_r\hat{V}_r^* \end{split}$$

 $A\bar{V} = 0$ since $Av_i = 0$ for all i > r.

(c)
$$U\Sigma V^* = \begin{bmatrix} \hat{U}_r & \bar{U} \end{bmatrix} \begin{bmatrix} \hat{\Sigma}_r & 0 \\ 0 & 0 \end{bmatrix} \begin{bmatrix} \hat{V}_r^* \\ \bar{V}^* \end{bmatrix}$$
$$= \begin{bmatrix} \hat{U}_r & \bar{U} \end{bmatrix} \begin{bmatrix} \hat{\Sigma}_r \hat{V}_r^* \\ 0 \end{bmatrix}$$
$$= \hat{U}_r \hat{\Sigma}_r \hat{V}_r^*$$
$$= A$$

Problem 2:

$$A^*Ax = A = A^*b$$

$$\iff \hat{V}_r \hat{\Sigma}_r^* \hat{U}_r^* \hat{U}_r \hat{\Sigma}_r \hat{V}_r^* x = \hat{V}_r \hat{\Sigma}_r^* \hat{U}_r^* b$$

$$\iff \hat{V}_r \hat{\Sigma}_r^* \hat{\Sigma}_r^* \hat{V}_r^* x = \hat{V}_r \hat{\Sigma}_r^* \hat{U}_r^* b$$

$$\iff \hat{V}_r^* x = (\hat{\Sigma}_r^* \hat{\Sigma}_r)^{-1} \hat{\Sigma}_r^* \hat{U}_r^* b$$

$$\iff \hat{V}_r^* x = \hat{\Sigma}_r^{-1} \hat{U}_r^* b$$

$$\iff x = \hat{V}_r \hat{\Sigma}_r^{-1} \hat{U}_r^* b$$

The converse of the last line holds as long as $b \neq 0$ since \hat{V}_r has full rank, so only one value of x can produce the nonzero product on the right. Therefore x is the unique minimizer.

Problem 3:

(a) $f(x_i) = -D_h^+ D_h^- y(x_i) \text{ for all } 0 < i < n$ $= -D_h^+ \frac{y(x_i) - y(x_i - h)}{h}$ $= -\frac{1}{h} \left[\frac{y(x_i + h) - y(x_i + h - h)}{h} - \frac{y(x_i) - y(x_i - h)}{h} \right]$ $= \frac{2y(x_i) - y(x_{i+1}) - y(x_{i+1})}{h^2}$

Then we have n-1 equations $\frac{2}{h^2}y_i - \frac{1}{h^2}y_{i-1} - \frac{1}{h^2}y_{i+1} = f(x_i)$ for 0 < i < n. However this uses n+1 variables $\{y_i\}_{i=0}^n$. The boundary conditions $y_0 = 1$, $y_n = 2$ give

$$\frac{2}{h^2}y_1 - \frac{1}{h^2}y_2 - \frac{1}{h^2} = f(x_1) \Rightarrow \alpha = \frac{1}{h^2}$$
$$\frac{2}{h^2}y_{n-1} - \frac{1}{h^2}y_{n-2} - \frac{2}{h^2} = f(x_{n-1}) \Rightarrow \beta = \frac{2}{h^2}$$

So $(A_h)_{i,i} = \frac{2}{h^2}$ for $1 \le i \le n-1$, $(A_h)_{i,i-1} = -\frac{1}{h^2}$ for $2 \le i \le n-1$, and $(A_h)_{i,i+1} = -\frac{1}{h^2}$ for $1 \le i \le n-2$. A_h is tridiagonal, so all other entries are zero. Then $A_h y_h = f_h$ enforces all the conditions.

(b) A_h is clearly symmetric from the construction above: each entry on the subdiagonal and superdiagonal is $-\frac{1}{h^2}$, and everything else besides the diagonal is zero. Let $v \in \mathbb{R}^{n-1}$, $v \neq 0$.

$$v_{i}(Av)_{i} = \frac{1}{h^{2}} (2v_{i}^{2} - v_{i}v_{i-1} - v_{i}v_{i+1}) \text{ for } 1 < i < n-1$$

$$\Rightarrow v^{\top}Av = \frac{1}{h^{2}} \left[\sum_{i=2}^{n-2} (2v_{i}^{2} - v_{i}v_{i-1} - v_{i}v_{i+1}) + 2v_{1}^{2} - v_{1}v_{2} + 2v_{n-1}^{2} - v_{n-1}v_{n-2} \right]$$

$$= \frac{1}{h^{2}} \left[2 \sum_{i=1}^{n-1} v_{i}^{2} - 2 \sum_{i=1}^{n-2} v_{i}v_{i+1} \right]$$

$$= \frac{1}{h^{2}} \left[\sum_{i=1}^{n-2} (v_{i} - v_{i+1})^{2} + v_{1}^{2} + v_{n-1}^{2} \right]$$

$$> 0$$

The result is strictly greater than 0 because if $(v_i - v_{i+1}) = 0 \ \forall i$ then $v_i = v_j \ \forall i, j$ which must mean that $v_i^2 > 0$ or else v = 0.

Problem 4:

(a) From last quarter's midterm, we know that for A hermitian and positive definite, $||A||_2 = \lambda_{max}$ and $||A^{-1}||_2 = \lambda_{min}^{-1}$. Therefore $\kappa = \frac{\lambda_{max}}{\lambda_{min}} \geq 1$ since SPD implies positive real eigenvalues. Let $a = \frac{\sqrt{\kappa}+1}{\sqrt{\kappa}-1} > 1$. Then $a^k > 1$ and $a^{-k} \in (0,1)$, giving

$$2\left[\left(\frac{\sqrt{\kappa}+1}{\sqrt{\kappa}-1}\right)^{k} + \left(\frac{\sqrt{\kappa}+1}{\sqrt{\kappa}-1}\right)^{-k}\right]^{-1} = 2(a^{k}+a^{-k})^{-1} \le 2(a^{k})^{-1} = 2\left(\frac{\sqrt{\kappa}-1}{\sqrt{\kappa}+1}\right)^{k}$$

(b)
$$u(x) = \gamma - \frac{2x}{\lambda_{max} - \lambda_{min}} = \frac{\lambda_{max} + \lambda_{min} - 2x}{\lambda_{max} - \lambda_{min}}$$

Since u is linear with respect to x, its values between λ_{min} and λ_{max} will be between $u(\lambda_{min})$ and $u(\lambda_{max})$.

$$u(\lambda_{min}) = \frac{\lambda_{max} - \lambda_{min}}{\lambda_{max} - \lambda_{min}} = 1$$

$$u(\lambda_{max}) = \frac{\lambda_{min} - \lambda_{max}}{\lambda_{max} - \lambda_{min}} = -1$$

So $-1 \le u(x) \le 1$ for $x \in [\lambda_{min}, \lambda_{max}]$

(c) $T_k(x) = \cos(k \arccos(x))$. p(x) is a k degree polynomial with p(0) = 1. Therefore $p \in \tilde{\mathcal{P}}^k$.

$$|p(\lambda)| = \left| \frac{T_k(u(x))}{T_k(\gamma)} \right| \le \left| \frac{1}{T_k(\gamma)} \right|$$

If the given equality can be shown, it implies that

$$|p(\lambda)| \le \left| \frac{1}{T_k(\gamma)} \right| = 2 \left[\left(\frac{\sqrt{\kappa} + 1}{\sqrt{\kappa} - 1} \right)^k + \left(\frac{\sqrt{\kappa} + 1}{\sqrt{\kappa} - 1} \right)^{-k} \right]^{-1}$$

Which means the first inequality will hold.

$$w = \frac{1}{2}(z + \frac{1}{z})$$

$$\Rightarrow 2zw = z^2 + 1$$

$$\Rightarrow 0 = z^2 - 2zw + 1$$

$$\Rightarrow z = w \pm \sqrt{w^2 - 1} \text{ by the quadratic formula.}$$
Let $z_1 = w + \sqrt{w^2 - 1}$, $z_2 = w - \sqrt{w^2 - 1}$. Then
$$z_1 z_2 = w^2 - w^2 + 1 = 1 \Rightarrow z_1 = z_2^{-1}$$

$$\Rightarrow J(z_1^k) = \frac{1}{2}(z_1^k + z_1^{-k})$$

 $= \frac{1}{2}(z_2^{-k} + z_2^k)$

 $=J(z_2^k)$

Finally, for $w \in [-1, 1]$,

$$\cos(k \arccos(w)) = \cos(\frac{k}{i} \operatorname{Log}(w + \sqrt{w^2 - 1}))$$

$$= \cos(\frac{k}{i} \operatorname{Log}(z_1))$$

$$= \cos(\frac{1}{i} \operatorname{Log}(z_1^k))$$

$$= \frac{1}{2} [e^{\operatorname{Log}(z_1^k)} + e^{-\operatorname{Log}(z_1^k)}]$$

$$= \frac{1}{2} [z_1^k + z_1^{-k}]$$

$$= J(z_1^k)$$

$$w = \gamma \Rightarrow z_1 = \gamma + \sqrt{\gamma^2 - 1}$$

$$= \frac{\kappa + 1}{\kappa - 1} + \sqrt{\left(\frac{\kappa + 1}{\kappa - 1}\right)^2 - 1}$$

$$= \frac{\kappa + 1}{\kappa - 1} + \sqrt{\frac{4\kappa}{(\kappa - 1)^2}}$$

$$= \frac{\kappa + 1}{\kappa - 1} + \frac{2\sqrt{\kappa}}{\kappa - 1}$$

$$= \frac{(\sqrt{\kappa} + 1)^2}{(\sqrt{\kappa} + 1)(\sqrt{\kappa} - 1)}$$

$$= \frac{\sqrt{\kappa} + 1}{\sqrt{\kappa} - 1}$$

$$\Rightarrow T_k(\gamma) = J\left(\left(\frac{\sqrt{\kappa} + 1}{\sqrt{\kappa} - 1}\right)^k\right)$$

$$= \frac{1}{2}\left[\left(\frac{\sqrt{\kappa} + 1}{\sqrt{\kappa} - 1}\right)^k + \left(\frac{\sqrt{\kappa} + 1}{\sqrt{\kappa} - 1}\right)^{-k}\right]$$

Problem 5:

(a)
$$f(x) = \sum_{k \in \mathbb{Z}} \frac{c_k}{\sqrt{2\pi}} e^{ikx}$$

$$\Rightarrow f'(x) = \sum_{k \in \mathbb{Z}} \frac{ikc_k}{\sqrt{2\pi}} e^{ikx}$$

$$\Rightarrow \mathcal{F}(f') = (ikc_k)_{k \in \mathbb{Z}} \text{Suppose } f^{(p)}(x) = \sum_{k \in \mathbb{Z}} \frac{(ik)^p c_k}{\sqrt{2\pi}} e^{ikx}$$

$$\Rightarrow f^{(p+1)}(x) = \sum_{k \in \mathbb{Z}} \frac{(ik)^{p+1} c_k}{\sqrt{2\pi}} e^{ikx}$$

Thus the general statement is proved by induction.

(b) Let
$$c_k^{(p)} = (ik)^p c_k$$
. By Parseval's relation,

$$\sum_{k} |c_{k}^{(p)}|^{2} < \infty$$

$$\Rightarrow 0 = \lim_{k \to \pm \infty} |c_{k}^{(p)}|^{2}$$

$$\Rightarrow 0 = \lim_{k \to \pm \infty} |c_{k}^{(p)}|$$

$$= \lim_{k \to \pm \infty} |(ik)^{p} c_{k}|$$

$$= \lim_{k \to \pm \infty} |k^{p} c_{k}|$$

$$= \lim_{k \to \pm \infty} \left| \frac{c_{k}}{k^{-p}} \right|$$

$$\Rightarrow |c_{k}| = o(|k|^{-p})$$

Problem 6:

```
(a) Let y_{\text{exact}} = 1 - x + x^2 + x^3 and f(x) = -2 - 6x
(b) import numpy as np
   import math
   import matplotlib.pyplot as plt
   yexact = lambda x: 1 -x + x**2 + x**3
   f = lambda x: -2 -6*x
   ns = [2**i for i in range(3,11)]
   def CG(n,b):
       m = len(b)
       bnorm = np.linalg.norm(b,2)
       Aprod = lambda v: (n)**2 *(((v[:-1]-v[1:])**2).sum() + v[0]**2 + v[-1]**2)
       def Amult(v):
           w = np.zeros(len(v))
           w[1:-1] = (n)**2 *(2*v[1:-1]-v[:-2]-v[2:])
           w[0] = (n)**2 *(2*v[0]-v[1])
           w[-1] = (n)**2 *(2*v[-1]-v[-2])
           return w
       x = np.zeros(m)
       r = b
       p = r
       count = 0
       while(count \leq n and np.linalg.norm(r,2)/bnorm \geq 10**(-10)):
           alpha = np.linalg.norm(r,2)**2/Aprod(p)
           x = x + alpha*p
           r1 = r
           r = r - alpha*Amult(p)
           beta = np.linalg.norm(r,2)**2/np.linalg.norm(r1,2)**2
           p = r + beta*p
           count += 1
       return x
   results = []
   for N in ns:
       X = np.linspace(0,1,N+1)[1:-1]
       fh = np.array([f(x) for x in X])
       fh[0] += N**2
       fh[-1] += 2*N**2
       yexac = np.array([yexact(x) for x in X])
       CG(N,fh)
       results.append(np.linalg.norm(CG(N,fh)-yexac,np.Inf))
   plt.plot(np.log([1/n for n in ns]),np.log(results))
```

