

Will Firmin

Math 151AH: HW2

Problem 1)

```
In [ ]: import numpy as np

a = np.array([[0.4218,0.6557,0.6787,0.6555],
              [0.9157,0.0357,0.7577,0.1712],
              [0.7922,0.8491,0.7431,0.7060],
              [0.9595,0.9340,0.3922,0.0318]])
c = np.array([0.2769,0.0462,0.0971,0.8235])

def gaussian_elimination(A,b):
    A = A.copy()
    b = b.copy()
    for i in range(A.shape[1]):
        b[i] = b[i] / A[i,i]
        A[i,:] = A[i,:]/A[i,i]
        if i < A.shape[1]-1:
            for j in range(i+1,A.shape[1]):
                b[j] = b[j] - A[j,i] * b[i]
                A[j,:] -= A[j,i] * A[i,:]
    for i in range(A.shape[1]-2,-1,-1):
        for j in range(i+1,A.shape[1]):
            b[i] = b[i] - A[i,j]*b[j]
            A[i,:] = A[i,:] - A[i,j] * A[j,:]
    return b

x=gaussian_elimination(a,c)
print(f"x = {x}")
```

```
x = [-1.26040472  1.40084283  2.03133593 -2.27103223]
```

Problem 2)

Part a)

$$\text{WTS } Av = \lambda v \Rightarrow \rho(A)v = \rho(\lambda)v$$

Suppose $\rho(A) = \sum_{i=0}^m c_i A^i$ for coefficients c_i . Then $\rho(A)v = \sum_{i=0}^m c_i A^i v$

Note that $A^i v = \lambda A^{i-1} v = \dots = \lambda^{i-1} A v = \lambda^i v$

Thus $\rho(A)v = \sum_{i=0}^m c_i \lambda^i v = v \sum_{i=0}^m c_i \lambda^i = \rho(\lambda)v$

Part b)

$$\begin{aligned}
\rho(A) &= \rho(P\Lambda P^{-1}) \\
&= \sum_{i=0}^m c_i (P\Lambda P^{-1})^i \\
&= \sum_{i=0}^m c_i P\Lambda^i P^{-1} \\
&= P \left(\sum_{i=0}^m c_i \Lambda^i \right) P^{-1} \\
&= P\rho(\Lambda)P^{-1}
\end{aligned}$$

Since Λ is diagonal, $(\Lambda^k)_{i,j} = (\Lambda_{i,j})^k$ and its diagonal entries of Λ are the eigenvalues of A , therefore $\rho(\Lambda) = \text{diag}(\rho(\lambda_1), \dots, \rho(\lambda_n))$.

Part c)

$$\begin{aligned}
e^A &= e^{P\Lambda P^{-1}} \\
&= \sum_{k=1}^{\infty} \frac{1}{k!} (P\Lambda P^{-1})^k \\
&= \sum_{k=1}^{\infty} \frac{1}{k!} P\Lambda^k P^{-1} \\
&= P \left(\sum_{k=1}^{\infty} \frac{1}{k!} \Lambda^k \right) P^{-1} \\
&= P e^{\Lambda} P^{-1}
\end{aligned}$$

Since Λ is diagonal, $e^{\Lambda} = \text{diag}(e^{\lambda_1}, \dots, e^{\lambda_n})$ as before in part b.

Problem 3)

Part a)

$$\begin{aligned}
\|x + y\|_1 &= \sum_{i=1}^n |x_i + y_i| \\
&\leq \sum_{i=1}^n (|x_i| + |y_i|) \\
&= \sum_{i=1}^n |x_i| + \sum_{i=1}^n |y_i| \\
&= \|x\|_1 + \|y\|_1
\end{aligned}$$

Part b)

$$\begin{aligned}
\|x + y\|_{\infty} &= \max_i |x_i + y_i| \\
&= |x_I + y_I| && \text{for some } I \leq n \\
&\leq |x_I| + |y_I| \\
&\leq \max_i |x_i| + \max_i |y_i| \\
&= \|x\|_{\infty} + \|y\|_{\infty}
\end{aligned}$$

Problem 4)

Part a)

For $u(t) = \frac{t^p}{p} + \frac{1}{q} - t$,

$$\frac{du}{dt} = t^{p-1} - 1 = 0 \Rightarrow t = 1$$

$$\frac{d^2u}{dt^2} = (p-1)t^{p-2} = p-1 > 0 \text{ at } t = 1$$

$$u(1) = 0, u(0) = \frac{1}{q} > 0$$

$$\Rightarrow 1 = \argmin_t u(t)$$

For $t^* = ab^{-\frac{q}{p}}$,

$$u(1) \leq u(t^*)$$

$$\Rightarrow 0 \leq \frac{a^p b^{-q}}{p} + \frac{1}{q} - ab^{-\frac{q}{p}}$$

$$\Rightarrow 0 \leq \frac{a^p}{p} + \frac{b^q}{q} - ab^{q-\frac{q}{p}}$$

$$= \frac{a^p}{p} + \frac{b^q}{q} - ab^{q(1-\frac{1}{p})}$$

$$= \frac{a^p}{p} + \frac{b^q}{q} - ab^{q\frac{1}{q}}$$

$$= \frac{a^p}{p} + \frac{b^q}{q} - ab$$

$$\Rightarrow ab \leq \frac{a^p}{p} + \frac{b^q}{q}$$

Part b)

$$\begin{aligned}
\frac{1}{\|x\|_p \|y\|_q} \sum_{k=1}^n |x_k y_k| &= \sum_{k=1}^n \frac{|x_k|}{\|x\|_p} \frac{|y_k|}{\|y\|_q} \\
&\leq \sum_{k=1}^n \left(\frac{1}{p} \frac{|x_k|^p}{\|x\|_p^p} + \frac{1}{q} \frac{|y_k|^q}{\|y\|_q^q} \right) \\
&= \frac{1}{p} \|x\|_p^{-p} \sum_{k=1}^n |x_k|^p + \frac{1}{q} \|y\|_q^{-q} \sum_{k=1}^n |y_k|^q \\
&= \frac{1}{p} \|x\|_p^{-p} \|x\|_p^p + \frac{1}{q} \|y\|_q^{-q} \|y\|_q^q \\
&= \frac{1}{p} + \frac{1}{q} \\
&= 1 \\
\Rightarrow \sum_{k=1}^n |x_k y_k| &\leq \|x\|_p \|y\|_q
\end{aligned}$$

Part c)

$$\begin{aligned}
\|x + y\|_p^p &= \sum_{i=1}^n |x_i + y_i|^{p-1} |x_i + y_i| \\
&\leq \sum_{i=1}^n |x_i| |x_i + y_i|^{p-1} + \sum_{i=1}^n |y_i| |x_i + y_i|^{p-1} \\
&\leq \|x\|_p \left(\sum_{i=1}^n (x_i + y_i)^{q(p-1)} \right)^{\frac{1}{q}} + \|y\|_p \left(\sum_{i=1}^n (x_i + y_i)^{q(p-1)} \right)^{\frac{1}{q}} \\
&= (\|x\|_p + \|y\|_p) \left(\sum_{i=1}^n (x_i + y_i)^p \right)^{\frac{1}{q}} \\
&= (\|x\|_p + \|y\|_p) \left(\sum_{i=1}^n (x_i + y_i)^p \right)^{1 - \frac{1}{p}} \\
\Rightarrow \|x + y\|_p &\leq (\|x\|_p + \|y\|_p) \frac{\sum_{i=1}^n (x_i + y_i)^p}{\|x + y\|_p^p} \\
&= (\|x\|_p + \|y\|_p) \left(\sum_{i=1}^n (x_i + y_i)^p \right)^{1-1} \\
&= \|x\|_p + \|y\|_p
\end{aligned}$$

Problem 5)

Part a)

$$v^T w = \sum_{k=1}^n v_k w_k \leq \sum_{k=1}^n |v_k w_k| \leq \|v\|_p \|w\|_q \leq \|v\|_p$$

Part b)

We are looking for w^* s.t. the above inequality holds as an equality. The inequality comes from two sources: the first is $v_k w_k \leq |v_k w_k|$, which holds equal when $\text{sign}(v_k) = \text{sign}(w_k)$. The second is Hölder's inequality, which holds as an equality when Young's inequality in turn holds as an equality. This occurs when $ab^{-\frac{q}{p}} = 1$. Translating to the current problem, this condition becomes

$$\frac{|v_k|}{\|v\|_p} \left(\frac{|w_k|}{\|w\|_q} \right)^{-\frac{q}{p}} = 1$$

Assuming for now that $\|w\|_q = 1$, this gives that

$$|w_k| = \left(\frac{|v_k|}{\|v\|_p} \right)^{\frac{p}{q}}$$

The first condition dictates that $\text{sign}(w_k) = \text{sign}(v_k)$.

Now I confirm that $\|w\|_q = 1$ holds:
$$\begin{aligned} \|w\|_q^q &= \left(\sum_{i=1}^n |w_i|^q \right)^{\frac{1}{q}} = \left(\sum_{i=1}^n \left(\frac{|v_i|}{\|v\|_p} \right)^p \right)^{\frac{1}{q}} = \|v\|_p^{-\frac{pq}{q}} \left(\sum_{i=1}^n |v_i|^p \right)^{\frac{1}{q}} \\ &= \|v\|_p^{-\frac{pq}{q}} (\|v\|_p^p)^{\frac{1}{q}} = 1 \end{aligned}$$

Part c)

Part a shows that for the region that we are maximizing $v^T w$ over, the $v^T w \leq \|v\|_p$. Therefore the maximized value must also be less than or equal to $\|v\|_p$. Part b shows that there exists w such that the value $\|v\|_p$ is attained. Since all greater values have been shown to be unattainable, we can conclude that $\|v\|_p$ is the maximum.

Problem 6)

Let $p(x) = \sum_{i=0}^n c_i x^i \in P_n$ for $c_i \in \mathbb{R}$ be represented by coordinates $c = (c_0, \dots, c_n) \in \mathbb{R}^n$ with respect to the given basis.

Then $\frac{d}{dx} p(x) = \sum_{i=0}^n i c_i x^{i-1} = \sum_{i=0}^{n-1} (i+1) c_{i+1} x^i$

can be represented by $\hat{c} = (\hat{c}_0, \dots, \hat{c}_n)$ where $\hat{c}_i = (i+1) c_{i+1}$ for $i < n$ and $\hat{c}_n = 0$.

The derivative operator is linear: let $p, q \in P_n$. Then for $\alpha, \beta \in \mathbb{R}$,

$$\begin{aligned}
\frac{d}{dx}(\alpha p(x) + \beta q(x)) &= \frac{d}{dx}(\alpha \sum_{i=0}^n p_i x^i + \beta \sum_{i=0}^n q_i x^i) \\
&= \alpha \sum_{i=0}^n i p_i x^{i-1} + \beta \sum_{i=0}^n i q_i x^{i-1} \\
&= \sum_{i=0}^n (\alpha p_i + \beta q_i) i x^{i-1} \\
&= \sum_{i=0}^{n-1} (\alpha p_{i+1} + \beta q_{i+1}) (i+1) x^i
\end{aligned}$$

has coordinates with respect to $\{x^i\}_{0 \leq i \leq n}$ and is thus in P_n . Furthermore, the derivative operator can be represented by matrix $A \in \mathbb{R}^{n \times n}$ where $A_{i,i+1} = i$ for $i < n+1$ and all other entries are 0. A has no eigenvalues since every nonzero polynomial loses a degree under differentiation: the "highest" nonzero coordinate becomes zero.

Problem 7)

Part a)

$$\begin{aligned}
\|D\|_\infty &= \sup_{\|v\|_1=1} \|Dv\|_\infty \\
&= \max_v (\max_i \{D_{i,i} v_i\}) \text{ s.t. } \sum_i |v_i| = 1 \\
|v_i| &\leq 1 \Rightarrow D_{i,i} v_i \leq |D_{i,i} v_i| \leq |D_{i,i}| \\
D_{i,i} v_i &= |D_{i,i}| \text{ for } v_i = \text{sign}(D_{i,i}), v_j = 0 \forall j \neq i \\
&\text{so } D_{i,i} v_i \text{ is maximized with } v \text{ s.t. all weight is on } D_{i,i}. \\
&\text{This makes the maximization problem a choice of } i. \\
\Rightarrow \|D\|_\infty &= \max_v \{ \max_i \{D_{i,i} v_i\} \} \\
&= \max_i |D_{i,i}|
\end{aligned}$$

Part b)

$$\begin{aligned}
\|U\|_2 &= \sup_{v \neq 0} \frac{\|Qv\|_2}{\|v\|_2} \\
&= \sup_{v \neq 0} \left(\frac{\langle Qv, Qv \rangle}{\langle v, v \rangle} \right)^{\frac{1}{2}} \\
&= \sup_{v \neq 0} \left(\frac{\langle v, Q^* Q v \rangle}{\langle v, v \rangle} \right)^{\frac{1}{2}} \\
&= \sup_{v \neq 0} \left(\frac{\langle v, v \rangle}{\langle v, v \rangle} \right)^{\frac{1}{2}} \\
&= 1
\end{aligned}$$

Part c)

By Schur decomposition there exists a unitary matrix U s.t. $A = U^*TU$ for upper triangular T .

$$\begin{aligned}UAU^* &= T \\ \Rightarrow (UAU^*)^* &= T^* \\ \Rightarrow UA^*U^* &= T^* \\ \Rightarrow UAU^* &= T^* \\ \Rightarrow T &= T^*\end{aligned}$$

Since T is upper triangular, this implies that T must be diagonal.

Problem 8)

In []:

```
def diagonal(A):
    n = len(A)
    D = np.zeros([n,n])
    D[np.diag_indices(n)] = A[np.diag_indices(n)]
    return D

def error(x):
    return ((x - np.array([1,2,3,4]))**2).sum()**0.5

def Jacobi(A, b, N=100):
    n = len(a)
    A2 = diagonal(A)
    A2inv = np.zeros([n,n])
    A2inv[np.diag_indices(n)] = np.reciprocal(A2[np.diag_indices(n)])
    A1 = A - A2
    x = [np.zeros(n)]
    count = 0
    while error(x[-1]) > 10**(-8) and count < N:
        x.append(A2inv.dot(b-A1.dot(x[-1])))
        count += 1
    return x

a = np.array([[2.4117,0.6557,0.6787,0.6555],
              [0.9157,1.8804,0.7577,0.1712],
              [0.7922,0.8491,3.0905,0.7060],
              [0.9595,0.9340,0.3922,2.3175]])
c = np.array([8.3813,7.6345,14.5862,13.2743])

X = Jacobi(a,c, N=200)
eJ = [np.log(error(x)) for x in X]
```

In []:

```
def GaussSeidel(A, b, N=100):
    n = len(a)
    x = [np.zeros(n)]
    count = 0
    while error(x[-1]) > 10**(-8) and count < N:
        xk = np.zeros(n)
        xk1 = x[-1]
        for i in range(n):
            s1 = sum([A[i,j]*xk[j] for j in range(i)])
            s2 = sum([A[i,j]*xk1[j] for j in range(i+1,n)])
            xk[i] = 1/A[i,i]*(b[i] - s1 - s2)
        x.append(xk)
```

```

        count += 1
    return x

X = GaussSeidel(a,c,N=200)
eG = [np.log(error(x)) for x in X]

import matplotlib.pyplot as plt
plt.plot(range(len(eJ)), eJ, label="Jacobi's Method")
plt.plot(range(len(eG)), eG, label="GaussSeidel Method")
plt.legend()
plt.show()

```

