

# Problem Set 8

Name: Jianxiang Fan

Email: jianxiang.fan@colorado.edu

1. (a)

$$\begin{aligned} w^T P &= [1 \ 1 \ \cdots \ 1] \begin{bmatrix} p_{11} & p_{12} & \cdots & p_{1n} \\ p_{21} & p_{22} & \cdots & p_{2n} \\ \cdots & \cdots & \cdots & \cdots \\ p_{n1} & p_{n2} & \cdots & p_{nn} \end{bmatrix} \\ &= \left[ \sum_{i=1}^n p_{i1} \quad \sum_{i=1}^n p_{i2} \quad \cdots \quad \sum_{i=1}^n p_{in} \right] \\ &= [1 \ 1 \ \cdots \ 1] \\ &= w^T \end{aligned}$$

(b)

$$\begin{aligned} q_{k+1,1} &= q_{k,1} p_{11} + q_{k,2} p_{12} + q_{k,3} p_{13} \\ q_{k+1,2} &= q_{k,1} p_{21} + q_{k,2} p_{22} + q_{k,3} p_{23} \\ q_{k+1,3} &= q_{k,1} p_{31} + q_{k,2} p_{32} + q_{k,3} p_{33} \end{aligned}$$

That is

$$\mathbf{q}_{k+1} = P \mathbf{q}_k$$

(c) Since  $\|\mathbf{q}_0\|_1 = 1$ , and  $\forall k \geq 1$

$$\begin{aligned} \|\mathbf{q}_{k+1}\|_1 &= |q_{k+1,1}| + |q_{k+1,2}| + |q_{k+1,3}| \\ &= q_{k,1}(p_{11} + p_{21} + p_{31}) + q_{k,2}(p_{12} + p_{22} + p_{32}) + q_{k,3}(p_{13} + p_{23} + p_{33}) \\ &= q_{k,1} + q_{k,2} + q_{k,3} \\ &= \|\mathbf{q}_k\|_1 \end{aligned}$$

So  $\forall k \geq 1$ ,  $\|\mathbf{q}_k\| = 1$ , there is no need to normalize the new vector.

(d) .

---

```
1 P = [0.63,0.18,0.14;0.26,0.65,0.31;0.11,0.17,0.55];
2 q0 = [1/3;1/3;1/3];
3 while 1
4     q = P*q0;
5     if norm(q-q0) < 1e-9
6         break
7     end
8     q0 = q;
9 end
```

---

$\mathbf{q}$  converges to  $[0.3095097 \ 0.4462493 \ 0.2442410]^T$

2. (a) The algorithm may be terminated when we get a reasonable approximation for the eigenvector, which means

$$\|\mathbf{ev}_{new} - \mathbf{ev}_{old}\|_2 < \epsilon$$

(b) .

---

```
1 function [ew, ev2] = sipm(A, mu)
2 [n,~] = size(A);
3 ev1 = randn(n,1);
4 ev1 = ev1 ./ norm(ev1);
5 while 1
6     ev2 = (A - mu * eye(n))\ev1;
7     ev2 = ev2 ./ norm(ev2);
8     ew = ev2'*A*ev2;
9     if ev2(1) < 0
10         ev2 = -ev2;
11     end
12     if norm(ev2 - ev1) < 1e-9
13         break
14     end
15     ev1 = ev2;
16 end
17 end
```

---

(c) Eigenvalue: 0.475207973

Eigenvector:  $\begin{bmatrix} 0.807422688 \\ -0.298585457 \\ -0.508837231 \end{bmatrix}$

Iterations: 23

(d) convergence rate: 0.3831365

- (e) Yes. Using eig function in Matlab, we'll get the eigenvalues of  $P$  are  $\lambda_1 = 1, \lambda_2 = 0.47520797, \lambda_3 = 0.35479203$ . Based on the theory stated in class, the convergence rate is

$$\frac{|\mu - \lambda_2|}{|\mu - \lambda_3|} = \frac{|0.55 - 0.47520797|}{|0.55 - 0.35479203|} = 0.383140$$

We can see that the method stated in (d) gives a reasonable estimation.

3. When doing LU factorization on a  $n \times n$  upper Hessenberg matrix, when iterating for column  $k$ , only row  $k+1$  with nonzero  $A_{k+1,k}$  need to be processed. Then the algorithm can be modified as below.

---

```

1 function [ A, p ] = hessLU( A )
2 [n,~] = size(A);
3 p = 1:n;
4 for k=1:n-1
5     if A(p(k+1),k) > A(p(k), k)
6         t = p(k+1);
7         p(k+1) = p(k);
8         p(k) = t;
9     end
10    A(p(k+1),k) = A(p(k+1),k)/A(p(k),k);
11    for j=k+1:n
12        A(p(k+1),j) = A(p(k+1),j) - A(p(k+1),k) * A(p(k),j);
13    end
14 end
15 end

```

---

FLOPs:

$$\begin{aligned}
 & \sum_{k=1}^{n-1} \sum_{j=k+1}^n 2 \\
 &= 2 \sum_{k=1}^{n-1} (n - k) \\
 &= n(n - 1) \\
 &= n^2 - n
 \end{aligned}$$

Without pivoting,  $L$  is lower bidiagonal with all 1's on the diagonal.

$$L = \begin{bmatrix} 1 & 0 & 0 & \cdots & 0 & 0 \\ * & 1 & 0 & \cdots & 0 & 0 \\ 0 & * & 1 & \cdots & 0 & 0 \\ \cdots & \cdots & \cdots & \cdots & \cdots & \cdots \\ 0 & 0 & 0 & \cdots & 1 & 0 \\ 0 & 0 & 0 & \cdots & * & 1 \end{bmatrix}$$

With pivoting,  $L$  is unit lower triangular with at most one nonzero element in each column besides the diagonal element.