

28.1) Firstly, the QR algorithm without shifts is given by,

$$\begin{cases} \bullet A_0 = A \\ \bullet \text{for } k=1, 2, \dots \\ \bullet Q_k R_k = A_{k-1} \quad (\text{QR factorization}) \\ \bullet A_k = R_k Q_k \end{cases}$$

If A is orthogonal, the QR factorization

$$A = A_0 = Q_1 R_1 = A \cdot I \quad \left(\text{If } A \text{ is already orthogonal } QR(A) = [A, I] \right)$$

$$\text{So } A_1 = R_1 Q_1 = I \cdot A = A.$$

It follows that after every iteration, we will have $Q_k = A$, $R_k = I$, and $A_k = A$, so the algorithm really doesn't do much at all.

Comparing with Theorem 28.4, A_k no longer (always)* converges to $\text{diag}(\vec{\lambda})$ as $k \rightarrow \infty$, $A_k = A$ $\forall k \geq 0$, but Q_k still converges to $Q = A$.

In general, an orthogonal matrix need not have eigenvalues of distinct modulus, and need not be real and symmetric. The matrix

* If $A = \text{diag}(\vec{\lambda})$ with $\vec{\lambda}$ a vector of distinct ^{real} eigenvalues, it is technically orthogonal, real, and symmetric, so Theorem 28.4 trivially applies, but this is a special case, and is already diagonalized.

30.1) we are looking for a 2×2 rotation matrix

$$J = \begin{bmatrix} \cos\theta & \sin\theta \\ -\sin\theta & \cos\theta \end{bmatrix} = \begin{bmatrix} c & s \\ -s & c \end{bmatrix} \text{ such that } (\lambda_i \neq 0)$$

$$J^T A J = \begin{bmatrix} c & -s \\ s & c \end{bmatrix} \begin{bmatrix} a & d \\ d & b \end{bmatrix} \begin{bmatrix} c & s \\ -s & c \end{bmatrix} = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix} = D$$

we have, by multiplying everything out, the following system of equations

$$\begin{bmatrix} c(ac-ds) - s(cd-bs) & c(cd-bs) + s(ac-ds) \\ c(as+cd) - s(bc+ds) & c(bc+ds) + s(as+cd) \end{bmatrix} = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

Focusing on equation ① and ②,

$$\text{①} \rightarrow c(as+cd) = s(bc+ds)$$

$$\text{②} \rightarrow c(cd-bs) = s(ds-ac)$$

Dividing ① with ②,

$$\frac{(as+cd)}{(cd-bs)} = \frac{(bc+ds)}{(ds-ac)}$$

$$\rightarrow (as+cd)(ds-ac) = (cd-bs)(bc+ds)$$

$$\underline{ads^2} - a^2 sc + \underline{scd^2} - \underline{adc^2} = \underline{bdc^2} + \underline{d^2sc} - \underline{b^2sc} - \underline{bds^2}$$

$$\rightarrow bd(c^2-s^2) + ad(c^2-s^2) = sc(b^2-a^2)$$

$$\rightarrow (b/a)d \cos(2\theta) = \frac{1}{2} \sin(2\theta) (b-a)(b/a)$$

$$(\text{since } \cos^2\theta - \sin^2\theta = \cos(2\theta) \text{ and } 2\sin\theta\cos\theta = \sin(2\theta))$$

Dividing both sides by $\cos(2\theta)$ we see

$$d = \frac{1}{2} \tan(2\theta) [b-a], \text{ or}$$

$$\tan(2\theta) = \frac{2d}{b-a}, \text{ as desired.}$$

We have chosen θ such that D is diagonal.

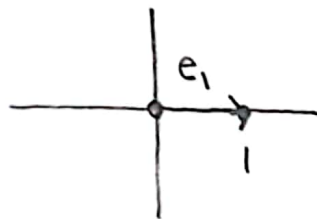
Geometrically speaking, we have

$$J^T A J = D = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

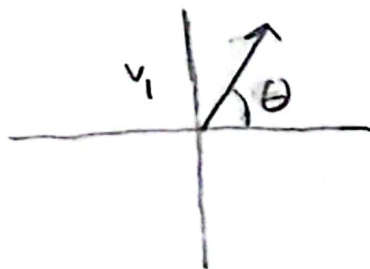
or

$$J^T A J = J^T A J I = J^T A J [\vec{e}_1 | \vec{e}_2] = \begin{bmatrix} \lambda_1 & 0 \\ 0 & \lambda_2 \end{bmatrix}$$

Consider the vector e_1



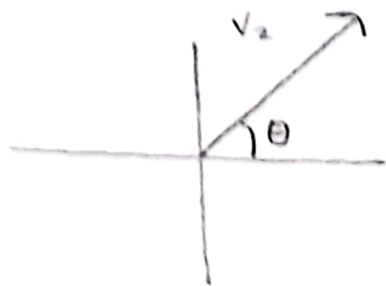
$$J e_1 = v_1$$



(still unit length)

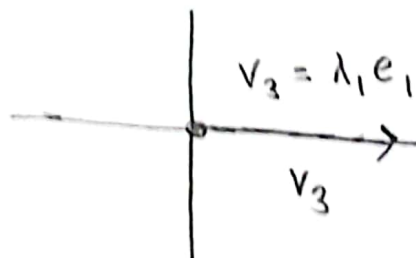
J rotates e_1 by the θ we found

$$\downarrow v_2 = A J e_1$$



Now A stretches v_1 so that its signed length is λ_1

$$\downarrow v_3 = J^T A J e_1$$



Now J^T rotates v_2 back to be along the e_1 axis but now has signed length λ_1

The same happens for e_2 , but it will get mapped to $\lambda_2 e_2$ instead. The significance of choosing θ such that $\tan(2\theta) = \frac{2d}{b-a}$, is that if we chose any other rotation angle, we wouldn't have this nice property that the e_i 's get mapped to $\lambda_i e_i$, meaning if this angle was not chosen, at the end of our rotating/stretching, we wouldn't land back onto the principle axes.

30.3 Put $S_0 = \sum_{i \neq j} (a_{ij})^2$, the sum of the off diagonal entries, then after 1 step we have

$$S_1 = S_0 - 2a^2 \quad (\text{where } a = \max_{i \neq j} (a_{ij})^2)$$

So,

$$\frac{S_1}{S_0} = 1 - \frac{2a^2}{S_0} \quad \left[\begin{array}{l} \text{since } A \text{ is symmetric we will} \\ \text{out 2 terms} \end{array} \right]$$

However, we can say

$$\sum_{i \neq j} a_{ij}^2 \leq \overbrace{(m^2 - m)}^{\# \text{ of off-diags}} a^2 \quad \swarrow \text{max off-diag square}$$

$$\rightarrow \frac{\sum_{i \neq j} a_{ij}^2}{m^2 - m} \leq a^2$$

$$\rightarrow -a^2 \leq -\frac{S_0}{m^2 - m}$$

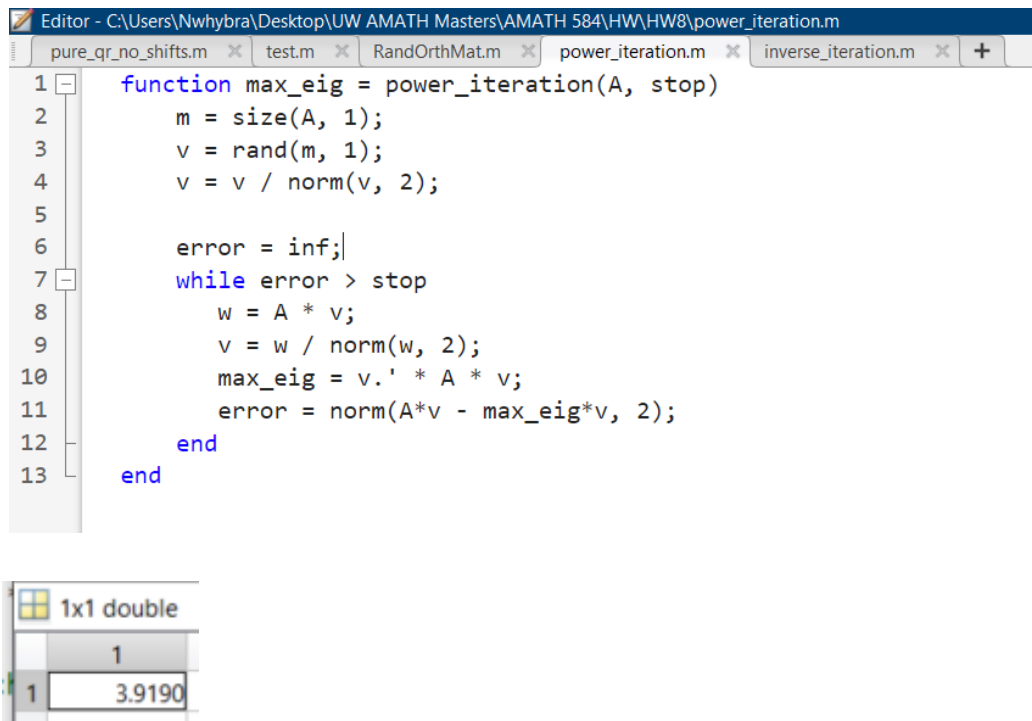
So

$$\textcircled{A} = 1 - \frac{2a^2}{S_0} \leq 1 - \frac{2S_0'}{S_0(m^2 - m)} = 1 - \frac{2}{(m^2 - m)} \quad \textcircled{A}$$

So the ratio of the sum decreases by a factor of at least \textcircled{A} in going from the current step to the next, so we are done.

Problem A1:

- a) Gerschgorin's here tells us that eigenvalues of A must either satisfy $|\lambda - 2| \leq 1$ (the sum of the absolute-values of the off-diagonals in the first/last row is 1 and the diagonal is 2), $|\lambda - 2| \leq 2$ (the sum of the absolute-values of the off-diagonals in the rest of the rows is 2). So the eigenvalues are either in the interval $[1, 3]$ or $[0, 4]$, as our matrix is real and symmetric it must have only real eigenvalues.
- b)



The image shows a MATLAB Editor window with the following tabs: pure_qr_no_shifts.m, test.m, RandOrthMat.m, power_iteration.m, and inverse_iteration.m. The active tab is power_iteration.m, which contains the following code:

```
1 function max_eig = power_iteration(A, stop)
2     m = size(A, 1);
3     v = rand(m, 1);
4     v = v / norm(v, 2);
5
6     error = inf;
7     while error > stop
8         w = A * v;
9         v = w / norm(w, 2);
10        max_eig = v.' * A * v;
11        error = norm(A*v - max_eig*v, 2);
12    end
13 end
```

Below the editor, the Command Window shows the output of the function. It displays a 1x1 double array with the value 3.9190.

1x1 double	
1	3.9190

c)

```

Editor - C:\Users\Nwhybra\Desktop\UW AMATH Masters\AMATH 584\HW\HW8\pure_qr_no_shifts.m
pure_qr_no_shifts.m  test.m  RandOrthMat.m  power_iteration.m  inverse_iteration.m  +
1  function [B, Q] = pure_qr_no_shifts(A, stop)
2      B = A;
3      m = size(A, 1);
4      % A matrix with 1's on the off-diagonals and 0's on the diagonals.
5      % Used to conveniently grab the matrix entries I want to compute the
6      % error.
7      off_diagonal_mask = logical(ones(m, m) - eye(m));
8
9      error = inf;
10     while error > stop
11         [Q, R] = qr(B);
12         B = R * Q;
13         % Here the error is the maximum of the off-diagonal entries.
14         error = sum(max(abs(B(off_diagonal_mask))));
15     end
16 end

```

10x10 double

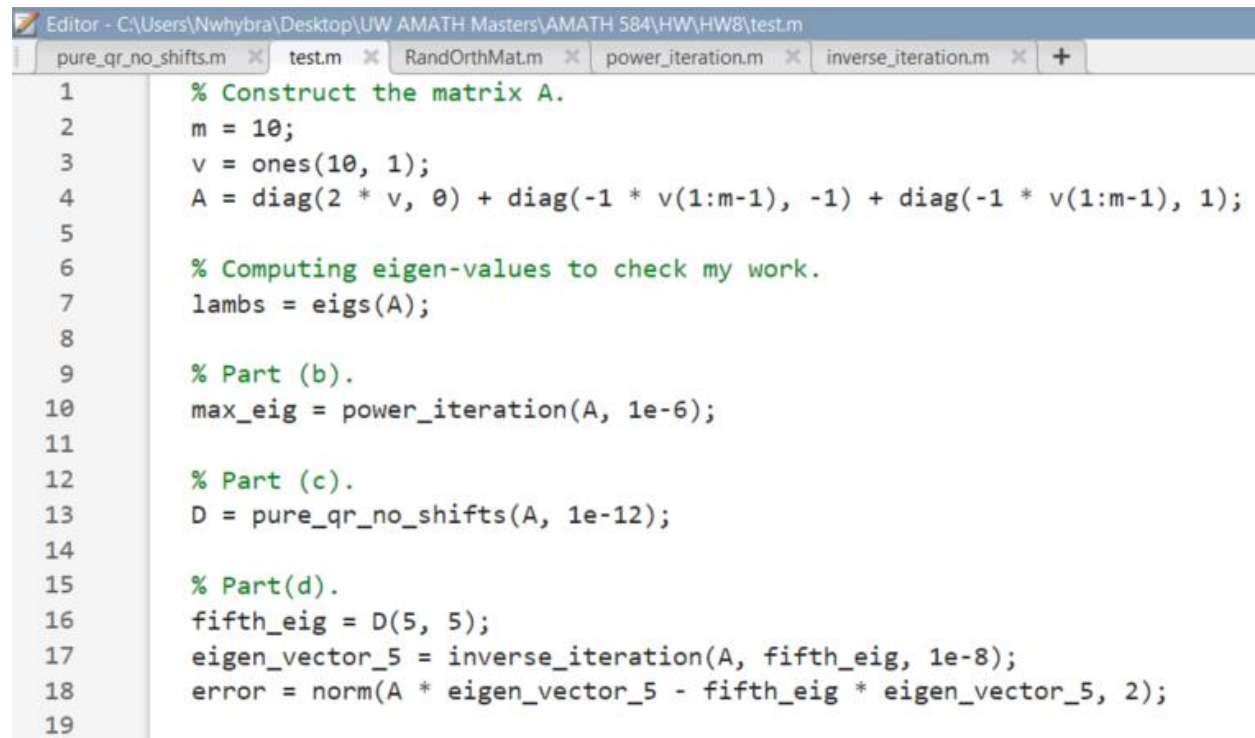
	1	2	3	4	5	6	7	8	9	10
1	3.9190	-9.5441e-13	2.5982e-16	-1.8796e-16	-2.8892e-16	-1.3452e-16	-2.5568e-16	-2.5253e-17	5.7313e-17	1.8443e-17
2	-9.5467e-13	3.6825	1.2341e-16	1.5202e-16	-1.4776e-16	-2.1867e-16	7.2846e-17	-1.3462e-16	-8.3503e-17	1.9793e-16
3	0	-1.2700e-20	3.3097	1.3249e-16	1.5238e-16	1.5350e-16	5.3961e-18	1.5278e-16	-2.8109e-17	-4.7965e-16
4	0	0	-1.1176e-29	2.8308	2.7220e-16	2.9658e-16	1.3209e-16	-1.7298e-16	-1.7246e-16	1.9614e-16
5	0	0	0	-1.8971e-40	2.2846	5.9651e-17	2.5885e-16	3.2062e-16	3.1835e-16	1.4064e-16
6	0	0	0	0	-5.8826e-54	1.7154	-5.9967e-17	2.4485e-16	1.4407e-16	1.3595e-17
7	0	0	0	0	0	-3.7753e-72	1.1692	1.5039e-17	8.4684e-17	1.7691e-16
8	0	0	0	0	0	0	-3.2122e-99	0.6903	1.4902e-17	-2.3424e-16
9	0	0	0	0	0	0	0	-2.6214e-1...	0.3175	2.0518e-16
10	0	0	0	0	0	0	0	0	-2.5458e-2...	0.0810

d)

```
Editor - C:\Users\Nwhybra\Desktop\UW AMATH Masters\AMATH 584\HW\HW8\inverse_iteration.m
pure_qr_no_shifts.m x test.m x RandOrthMat.m x power_iteration.m x inverse_iteration.m x +
1 function eigen_vector = inverse_iteration(A, lambda, stop)
2     m = size(A, 1);
3     v = rand(m, 1);
4     v = v / norm(v, 2);
5
6     error = inf;
7     I = eye(m);
8     while error > stop
9         w = (A - lambda * I) \ v;
10        v = w / norm(w, 2);
11        error = norm(A * v - lambda * v, 2);
12    end
13
14    eigen_vector = v;
15 end
```

10x1 double	
	1
1	-0.4221
2	0.1201
3	0.3879
4	-0.2305
5	-0.3223
6	0.3223
7	0.2305
8	-0.3879
9	-0.1201
10	0.4221

Full test script:



```
Editor - C:\Users\Nwhybra\Desktop\UW AMATH Masters\AMATH 584\HW\HW8\test.m
pure_qr_no_shifts.m test.m RandOrthMat.m power_iteration.m inverse_iteration.m +
1      % Construct the matrix A.
2      m = 10;
3      v = ones(10, 1);
4      A = diag(2 * v, 0) + diag(-1 * v(1:m-1), -1) + diag(-1 * v(1:m-1), 1);
5
6      % Computing eigen-values to check my work.
7      lambs = eigs(A);
8
9      % Part (b).
10     max_eig = power_iteration(A, 1e-6);
11
12     % Part (c).
13     D = pure_qr_no_shifts(A, 1e-12);
14
15     % Part(d).
16     fifth_eig = D(5, 5);
17     eigen_vector_5 = inverse_iteration(A, fifth_eig, 1e-8);
18     error = norm(A * eigen_vector_5 - fifth_eig * eigen_vector_5, 2);
19
```