

Q1

a)

Since we know that Q is a orthogonal matrix, we have: $Q^T Q = I$

1. inner products. needs to show: $(Qu)^T \cdot (Qv) = u^T v$

$$(Qu)^T \cdot (Qv) = u^T Q^T Qv = u^T I v = u^T v$$

2. norms. needs to show: $\|Qu\|_2 = \|u\|_2$

$$\begin{aligned}\|Qu\|_2 &= | \langle Qu, Qu \rangle | \\ &= | (Qu)^T \cdot (Qu) | \\ &= | u^T Q^T Qu | = | u^T u | \\ &= | \langle u, u \rangle | = \|u\|_2\end{aligned}$$

then we have: $\|Qu\|_2 = \|u\|_2$

3. distances. needs to show $\|Qu - Qv\|_2 = \|u - v\|_2$

$$\begin{aligned}\|Qu - Qv\|_2 &= | \langle Qu - Qv, Qu - Qv \rangle | \\ &= | \langle Qu, Qu - Qv \rangle - \langle Qv, Qu - Qv \rangle | \\ &= | (\langle Qu, Qu \rangle - \langle Qv, Qu \rangle) - (\langle Qu, Qv \rangle - \langle Qv, Qv \rangle) | \\ &= | (u^T u - v^T u) - (u^T v - v^T v) | \\ &= | (\langle u, u \rangle - \langle v, u \rangle) - (\langle u, v \rangle - \langle v, v \rangle) | \\ &= | \langle u - v, u \rangle - \langle u - v, v \rangle | \\ &= | \langle u, u - v \rangle - \langle v, u - v \rangle | \\ &= | \langle u - v, u - v \rangle | \\ &= \|u - v\|_2\end{aligned}$$

then we have: $\|Qu - Qv\|_2 = \|u - v\|_2$

4. angles. needs to show: $\angle(Qu, Qv) = \angle(u, v)$

$$\begin{aligned}\theta &= \arccos\left(\frac{\langle Qu, Qv \rangle}{\|Qu\| \|Qv\|}\right) \\ &= \arccos\left(\frac{u^T Q^T Q v}{\|Qu\|_2 \|Qv\|_2}\right) = \arccos\left(\frac{u^T v}{\|u\|_2 \|v\|_2}\right)\end{aligned}$$

which implies that: $\angle(Qu, Qv) = \angle(u, v)$

b)

1) show that the orthogonal projection of u onto v , is the vector in $\text{span}(v)$ closest to u

Let $av \in \text{span}(v)$, $a \in \mathbb{R}$, then $d(u, av) = \|u - av\|$

The orthogonal projection of u onto v is denoted by $\text{Proj}_v(u)$, which is colinear with av . By definition, residual $r = u - \text{Proj}_v(u)$ and $u - \text{Proj}_v(u) \perp v$.

Since av and $\text{Proj}_v(u)$ are colinear, we have $\text{Proj}_v(u) - av$ being colinear with av as well.

Therefore, we have $(u - \text{Proj}_v(u)) \perp (\text{Proj}_v(u) - av)$, by Pythagoras theorem, we have:

$$\begin{aligned}& \|u - \text{Proj}_v(u) + \text{Proj}_v(u) - av\|^2 \\ &= \|u - \text{Proj}_v(u)\|^2 + \|\text{Proj}_v(u) - av\|^2 \\ &= \|u - av\|^2, \text{ which is the squared distance between } u \text{ and any vector from } \text{span}(v)\end{aligned}$$

such value is minized when $av = \text{Proj}_v(u)$:

$$\begin{aligned}
& ||u - Proj_v(u)||^2 + ||Proj_v(u) - av||^2 \\
&= ||u - Proj_v(u)||^2 + ||Proj_v(u) - Proj_v(u)||^2 \\
&= ||u - Proj_v(u)||^2 = ||u - av||^2
\end{aligned}$$

Therefore, the orthogonal projection of u onto v is the vector in $\text{span}(v)$ closest to u .

2) show that the orthogonal projection of u onto W is the vector inside W closest to u

Let $x \in W$, Since W is a subspace and $Proj_W(u) \in W$, we can find

$$y = Proj_W(u) + x \in W$$

$$||u - y||^2 = ||u - (Proj_W(u) + x)||^2 = ||(u - Proj_W(u)) - x||^2$$

By definition of residual, $u - Proj_W(u) \perp x$, then by Pythagoras theorem, we have:

$$||(u - Proj_W(u)) - x||^2 = ||u - Proj_W(u)||^2 + ||x||^2$$

where $||(u - Proj_W(u)) - x||^2$ is the squared distance from u to subspace W . The distance is minimized if we have $||x||^2 = 0$, which implies that

$$||u - y||^2 = ||u - Proj_W(u)||^2 \implies ||u - y|| = ||u - Proj_W(u)||$$

Therefore, the orthogonal projection of u onto W is the vector inside W closest to u

The closet vector is not unique because

3) uniqueness of closest vector

Assume there are $x, y \in W$ that are both orthogonal projection of u onto W , and $\|u - x\| \leq \|u - y\|$

Since x, y are both orthogonal projections of u onto W , we have:

$$\begin{aligned}\|u - x\| &= \|u - \text{Proj}_W(u)\| \\ \|u - y\| &= \|u - \text{Proj}_W(u)\|\end{aligned}$$

Plus $\|u - y\|$ on both sides of $\|u - x\| \leq \|u - y\|$, we obtain:

$$\begin{aligned}\|u - x\| + \|u - y\| &\leq \|u - y\| + \|u - y\| \\ 2\|u - \text{Proj}_W(u)\| &\leq 2\|u - \text{Proj}_W(u)\| \implies \|u - \text{Proj}_W(u)\| \leq \|u - \text{Proj}_W(u)\|\end{aligned}$$

Since $\|u - \text{Proj}_W(u)\|$ is always non-negative, the above inequality implies that $\|u - \text{Proj}_W(u)\| = 0$, which means that the closest vector is indeed unique.

Q2

a)

$$A = \begin{bmatrix} 1 & 3 & 1 \\ 0 & 1 & 2 \\ 1 & -1 & 3 \\ 0 & 0 & 4 \end{bmatrix}$$

$$v_1 = a_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, q_1 = \frac{v_1}{||v_1||} = \begin{bmatrix} 0.7071 \\ 0 \\ 0.7071 \\ 0 \end{bmatrix}, ||v_1|| = 1.414$$

$$v_2 = a_2 - (q_1^T a_2)q_1 = \begin{bmatrix} 2 \\ 1 \\ -2 \\ 0 \end{bmatrix}, q_2 = \frac{v_2}{||v_2||} = \begin{bmatrix} 0.6667 \\ 0.3333 \\ -0.6667 \\ 0 \end{bmatrix}, q_1^T a_2 = 1.414, ||v_2|| = 3$$

$$v_3 = a_3 - (q_1^T a_3)q_1 - (q_2^T a_3)q_2 = \begin{bmatrix} -0.1197 \\ 0.4786 \\ 0.1197 \\ 0.8615 \end{bmatrix}, q_3 = \frac{v_3}{||v_3||} = \begin{bmatrix} -0.1197 \\ 0.4786 \\ 0.1197 \\ 0.8615 \end{bmatrix}$$

$$q_1^T a_3 = 2.8284, q_2^T a_3 = -0.6667, ||v_3|| = 4.6428$$

$$Q = \begin{bmatrix} 0.7071 & 0.6667 & -0.1197 \\ 0 & 0.3333 & 0.4786 \\ 0.7071 & -0.6667 & 0.1197 \\ 0 & 0 & 0.8615 \end{bmatrix}$$

$$R = \begin{bmatrix} 1.414 & 1.414 & 2.8284 \\ 0 & 3 & -0.6667 \\ 0 & 0 & 4.6428 \end{bmatrix}$$

b)

• **Input:** $A = [a_1, \dots, a_n] \in \mathbb{R}^{m \times n}, m \geq n$, full column rank

• **Algorithm:**

for $j = 1, \dots, n$

$$v_j = a_j$$

for $i = 1, \dots, j-1$

$$R_{ij} = q_i^T a_j$$

$$v_j = v_j - R_{ij}q_i$$

end

$$r_{jj} = \|v_j\|_2$$

$$q_j = v_j / r_{jj}$$

end

- **Output:** R (result matrix) and $Q = [q_1, \dots, q_n]$

c)

```
1 function [Q, R] = myGS(A)
2 [m,n] = size(A);
3 R = zeros(n,n);
4 Q = zeros(m,n);
5 for j = 1:n
6     v = A(1:m, j);
7     for i = 1:j-1
8         R(i,j) = Q(1:m,i)' * A(1:m,j);
9         v = v - R(i,j) * Q(1:m,i);
10    end
11    R(j,j) = sqrt(v' * v);
12    Q(1:m, j) = v/R(j,j);
13 end
```

d)

% Q2

A = [1 3 1; 0 1 2; 1 -1 3; 0 0 4]

A = 4x3

1	3	1
0	1	2
1	-1	3
0	0	4

[Q,R] = myGS(A)

Q = 4x3

0.7071	0.6667	-0.1197
0	0.3333	0.4786
0.7071	-0.6667	0.1197
0	0	0.8615

R = 3x3

1.4142	1.4142	2.8284
0	3.0000	-0.6667
0	0	4.6428

Q3

a)

If v_3 is orthogonal to both q_1 and q_2 , then we should have:

$$q_1^T v_3 = 0, q_2^T v_3 = 0$$

Take example of $q_1^T v_3$, given that $\|q_1\| = \|q_2\| = 1$, we can express it as:

$$\begin{aligned} q_1^T v_3 &= q_1^T (a_3 - (q_1^T a_3)q_1 - (q_2^T a_3)q_2) \\ &= q_1^T a_3 - q_1^T q_1 (q_1^T a_3) - q_1^T q_2 (q_2^T a_3) \\ &= q_1^T a_3 - q_1^T a_3 - q_1^T q_2 (q_2^T a_3) \\ &= -q_1^T q_2 (q_2^T a_3) \end{aligned}$$

which is a non-zero term. That is contradictory to the fact that $q_1^T v_3 = 0$.

Therefore, if q_1, q_2 are not orthogonal, the $v_3 = a_3 - (q_1^T a_3)q_1 - (q_2^T a_3)q_2$ constructed can not be orthogonal to both q_1 and q_2 .

If q_1, q_2 are indeed orthogonal, then we have $q_1^T q_2 = 0$, which resolves the contradiction.

b)

- **Input:** $A = [a_1, \dots, a_n] \in \mathbb{R}^{m \times n}, m \geq n$, full column rank

- **Algorithm:**

for $j = 1, \dots, n$

$$v_j = a_j$$

for $i = 1, \dots, j-1$

$$R_{ij} = q_i^T v_j$$

$$v_j = v_j - R_{ij}q_i$$

end

$$r_{jj} = \|v_j\|_2$$

$$q_j = v_j / r_{jj}$$

end

- **Output:** R (result matrix) and $Q = [q_1, \dots, q_n]$

c)

```

1  function [Q, R] = myMGS(A)
2  [m,n] = size(A);
3  R = zeros(n,n);
4  Q = zeros(m,n);
5  for j = 1:n
6      v = A(1:m, j);
7      for i = 1:j-1
8          R(i,j) = Q(1:m,i)' * v;
9          v = v - R(i,j) * Q(1:m,i);
10     end
11     R(j,j) = norm(v,2);
12     Q(1:m, j) = v/R(j,j);
13 end

```

d)

% Q3

A = [1 3 1; 0 1 2; 1 -1 3; 0 0 4]

A = 4x3

1	3	1
0	1	2
1	-1	3
0	0	4

[Q,R] = myMGS(A)

Q = 4x3

0.7071	0.6667	-0.1197
0	0.3333	0.4786
0.7071	-0.6667	0.1197
0	0	0.8615

R = 3x3

1.4142	1.4142	2.8284
0	3.0000	-0.6667
0	0	4.6428

Q4

a)

$$A = \begin{bmatrix} 1 & 3 & 1 \\ 0 & 1 & 2 \\ 1 & -1 & 3 \\ 0 & 0 & 4 \end{bmatrix}$$

Step 1:

We modify on first and third row to start off.

- $r^2 = 1^2 + 1^2 \implies r = 1.414$
- $c = \frac{1}{1.414} = 0.707$
- $s = -\frac{1}{1.414} = -0.707$

$$G_1^T = \begin{bmatrix} c & 0 & -s & 0 \\ 0 & 1 & 0 & 0 \\ s & 0 & c & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} = \begin{bmatrix} 0.707 & 0 & 0.707 & 0 \\ 0 & 1 & 0 & 0 \\ -0.707 & 0 & 0.707 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$A_1 = G_1^T A = \begin{bmatrix} 0.707 & 0 & 0.707 & 0 \\ 0 & 1 & 0 & 0 \\ -0.707 & 0 & 0.707 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 3 & 1 \\ 0 & 1 & 2 \\ 1 & -1 & 3 \\ 0 & 0 & 4 \end{bmatrix} = \begin{bmatrix} 1.414 & 1.414 & 2.828 \\ 0 & 1 & 2 \\ 0 & -2.828 & 1.414 \\ 0 & 0 & 4 \end{bmatrix}$$

Step 2:

We modify on second and third row

- $r^2 = 1 + (-2.828)^2 \implies r = 3$
- $c = \frac{1}{3} = 0.333$
- $s = -\frac{-2.828}{3} = 0.943$

$$G_2^T = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0.333 & -0.943 & 0 \\ 0 & 0.943 & 0.333 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$A_2 = G_2^T G_1^T A = \begin{bmatrix} 1.414 & 1.414 & 2.828 \\ 0 & 3 & -0.6674 \\ 0 & 0 & 2.3569 \\ 0 & 0 & 4 \end{bmatrix}$$

Step 3:

We modify on third and fourth row

- $r^2 = 2.3569^2 + 4^2 \implies r = 4.643$
- $c = 2.3569/4.643 = 0.508$
- $s = -4/4.643 = -0.862$

$$G_3^T = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0.508 & 0.862 \\ 0 & 0 & -0.862 & 0.508 \end{bmatrix}$$

$$\tilde{R} = A_3 = G_3^T G_2^T G_1^T A = \begin{bmatrix} 1.414 & 1.414 & 2.828 \\ 0 & 3 & -0.667 \\ 0 & 0 & 4.645 \\ 0 & 0 & 0 \end{bmatrix}$$

$$\tilde{Q} = G_1 G_2 G_3 = \begin{bmatrix} 0.7070 & 0.6667 & -0.1196 & 0.2029 \\ 0 & 0.3330 & 0.4790 & -0.8129 \\ 0.7070 & -0.6667 & 0.1196 & -0.2029 \\ 0 & 0 & 0.8620 & 0.5080 \end{bmatrix}$$

b)

- **Input:** $A = [a_1, \dots, a_n] \in \mathbb{R}^{m \times n}, m \geq n$, full column rank

- **Algorithm:**

$$R = A$$

$$Q = I_m$$

for $j = 1 \dots n$

for $i = 0 \dots m - 1 - j$

$$GT = I_m$$

$$x = R_{m-i-1,j}$$

$$y = R_{m-i,j}$$

$$r = \sqrt{x^2 + y^2}$$

$$c = x/r$$

$$s = -y/r$$

$$GT_{m-i,m-i} = c$$

$$GT_{m-i-1,m-i-1} = c$$

$$GT_{m-i-1,m-i} = s$$

$$GT_{m-i,m-i-1} = -s$$

$$R = GT \cdot R$$

$$Q = Q \cdot GT^T$$

end

end

- **Output:** \tilde{R} and $\tilde{Q} = [q_1, \dots, q_n, q_{n+1}, \dots, q_m]$

c)

```
1 function [Q,R] = myGivens(A)
2 [m,n] = size(A);
3 R = A;
4 Q = eye(m);
5 for j = 1:n
6     for i = 0: m-1-j
7         GT = eye(m);
8         x = R(m-i-1,j);
9         y = R(m-i,j);
10        r = sqrt(x^2 + y^2);
11        c = x/r;
12        s = -y/r;
13        GT(m-i, m-i) = c;
14        GT(m-i-1, m-i-1) = c;
15        GT(m-i-1, m-i) = -s;
16        GT(m-i, m-i-1) = s;
17        R = GT * R;
18        Q = Q * GT';
19    end
20 end
```

d)

% Q4

A = [1 3 1; 0 1 2; 1 -1 3; 0 0 4]

A = 4x3

1	3	1
0	1	2
1	-1	3
0	0	4

[Q,R] = myGivens(A)

Q = 4x4

0.7071	0.6667	-0.1197	0.2031
0	0.3333	0.4786	-0.8123
0.7071	-0.6667	0.1197	-0.2031
0	0	0.8615	0.5077

R = 4x3

1.4142	1.4142	2.8284
0	3.0000	-0.6667
0	0.0000	4.6428
0	-0.0000	0

Q5

a)

$$A = \begin{bmatrix} 1 & 3 & 1 \\ 0 & 1 & 2 \\ 1 & -1 & 3 \\ 0 & 0 & 4 \end{bmatrix}$$

Step 1:

$$u = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, v_1 = \bar{v}_1 = u - \|u\|e_1 = \begin{bmatrix} -0.4142 \\ 0 \\ 1.0000 \\ 0 \end{bmatrix}$$

$$H_{v_1} = I_4 - \frac{2v_1v_1^T}{v_1^Tv_1} = \begin{bmatrix} 0.7071 & 1 & 0.7071 & 1 \\ 0 & 0 & 0 & 0 \\ 0.7071 & 0 & -0.7071 & 0 \\ 0 & 0 & 0 & 0 \end{bmatrix}$$

$$A_1 = H_{v_1}A = \begin{bmatrix} 1.4142 & 1.4142 & 2.8284 \\ 0 & 1 & 2 \\ 0 & 2.8284 & -1.4142 \\ 0 & 0 & 4 \end{bmatrix}$$

Step 2:

$$u = \begin{bmatrix} 1 \\ 2.8284 \\ 0 \end{bmatrix}, \bar{v}_2 = u - \|u\|e_2 = \begin{bmatrix} -2 \\ 2.8284 \\ 0 \end{bmatrix}, v_2 = \begin{bmatrix} 0 \\ -2 \\ 2.8284 \\ 0 \end{bmatrix}$$

$$H_{v_2} = I_4 - \frac{2v_2v_2^T}{v_2^Tv_2} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 0.3333 & 0.9428 & 0 \\ 0 & 0.9428 & -0.3333 & 0 \\ 0 & 0 & 0 & 1 \end{bmatrix}$$

$$A_2 = H_{v_2}H_{v_1}A = \begin{bmatrix} 1.4142 & 1.4142 & 2.8284 \\ 0 & 3 & -0.6667 \\ -0 & -0 & 2.3570 \\ 0 & 0 & 4 \end{bmatrix}$$

Step 3:

$$u = \begin{bmatrix} 2.357 \\ 4 \end{bmatrix}, \bar{v}_3 = u - \|u\|e_3 = \begin{bmatrix} -2.2858 \\ 4 \end{bmatrix}, v_3 = \begin{bmatrix} 0 \\ 0 \\ -2.2858 \\ 4 \end{bmatrix}$$

$$H_{v_3} = I_4 - \frac{2v_3v_3^T}{v_3^Tv_3} = \begin{bmatrix} 1 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 \\ 0 & 0 & 0.5077 & 0.8615 \\ 0 & 0 & 0.8615 & -0.5077 \end{bmatrix}$$

$$\tilde{R} = A_3 = H_{v_3}H_{v_2}H_{v_1}A = \begin{bmatrix} 1.4142 & 1.4142 & 2.8284 \\ 0 & 3 & -0.6667 \\ -0 & -0 & 4.6428 \\ -0 & -0 & 0 \end{bmatrix}$$

$$\tilde{Q} = H_{v_1}H_{v_2}H_{v_3} = \begin{bmatrix} 0.7071 & 0.6667 & -0.1197 & -0.2031 \\ 0 & 0.3333 & 0.4786 & 0.8123 \\ 0.7071 & -0.6667 & 0.1197 & 0.2031 \\ 0 & 0 & 0.8615 & -0.5077 \end{bmatrix}$$

b)

- **Input:** $A = [a_1, \dots, a_n] \in \mathbb{R}^{m \times n}$, $m \geq n$, full column rank

- **Algorithm:**

$$R = A$$

$$Q = I_m$$

for $i = 1 \dots n$

$$u = R_{i:m,i}$$

$$\bar{v} = u - \|u\|e_i$$

$$v = \begin{pmatrix} 0_1 \\ \vdots \\ 0_m \end{pmatrix}$$

$$v_{i:m,1} = \bar{v}$$

$$H_v = I_m - 2 \frac{vv^T}{v^T v}$$

$$R = H_v R$$

$$Q = Q H_v$$

end

- **Output:** \tilde{R} and $\tilde{Q} = [q_1, \dots, q_n, q_{n+1}, \dots, q_m]$

c)

```

1  function [Q,R] = myHouseHolder(A)
2  [m,n] = size(A);
3  Q = eye(m);
4  R = A;
5  for i = 1:n
6      u = R(i:m, i);
7      vb = u - (norm(u,2) * eye(m-i+1, 1));
8      v = zeros(m,1);
9      v(i:m , 1) = vb;
10     Hv = eye(m) - 2 * (v * v') / (v' * v);
11     R = Hv * R;
12     Q = Q * Hv;
13 end

```

d)

% Q5

A = [1 3 1; 0 1 2; 1 -1 3; 0 0 4]

A = 4x3

1	3	1
0	1	2
1	-1	3
0	0	4

[Q,R] = myHouseHolder(A)

Q = 4x4

0.7071	0.6667	-0.1197	-0.2031
0	0.3333	0.4786	0.8123
0.7071	-0.6667	0.1197	0.2031
0	0	0.8615	-0.5077

R = 4x3

1.4142	1.4142	2.8284
0.0000	3.0000	-0.6667
-0.0000	-0.0000	4.6428
-0.0000	-0.0000	0.0000