

QR by Householder Reflectors

Example 1. Using classical Gram-Schmidt, compute the (reduced) QR factorization of

$$\begin{bmatrix} 1 & 1 & 1 \\ \delta & 0 & 0 \\ 0 & \delta & 0 \\ 0 & 0 & \delta \end{bmatrix} \text{ where } \delta = 10^{-10} \text{ in double precision.}$$

First, $\mathbf{y} = \begin{bmatrix} 1 \\ \delta \\ 0 \\ 0 \end{bmatrix}$. Since $\|\mathbf{y}\|_2 = \sqrt{1 + \delta^2} = \sqrt{1} = 1$, $\mathbf{q}_1 = \mathbf{y}$.

$$Q = \begin{bmatrix} 1 \\ \delta \\ 0 \\ 0 \end{bmatrix} \quad R = \begin{bmatrix} 1 & & \\ 0 & & \\ 0 & & \\ 0 & & \end{bmatrix}$$

Then setting $\mathbf{y} = \begin{bmatrix} 1 \\ 0 \\ \delta \\ 0 \end{bmatrix}$, we have $r_{12} = [1 \ \delta \ 0 \ 0] \begin{bmatrix} 1 \\ 0 \\ \delta \\ 0 \end{bmatrix} = 1$ and $\mathbf{y} = \mathbf{y} - (1) \begin{bmatrix} 1 \\ \delta \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ -\delta \\ \delta \\ 0 \end{bmatrix}$.

Then $r_{22} = \|\mathbf{y}\|_2 = \sqrt{2\delta^2} = \sqrt{2}\delta$, so $\mathbf{q}_2 = \begin{bmatrix} 0 \\ -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \\ 0 \end{bmatrix}$.

$$Q = \begin{bmatrix} 1 & 0 \\ \delta & -\frac{\sqrt{2}}{2} \\ 0 & \frac{\sqrt{2}}{2} \\ 0 & 0 \end{bmatrix} \quad R = \begin{bmatrix} 1 & 1 \\ 0 & \sqrt{2}\delta \\ 0 & 0 \end{bmatrix}$$

Now for the third column, set $\mathbf{y} = \begin{bmatrix} 1 \\ 0 \\ 0 \\ \delta \end{bmatrix}$. Then $r_{13} = [1 \ \delta \ 0 \ 0] \begin{bmatrix} 1 \\ 0 \\ 0 \\ \delta \end{bmatrix} = 1$ and $\mathbf{y} =$

$$\begin{bmatrix} 1 \\ 0 \\ 0 \\ \delta \end{bmatrix} - \begin{bmatrix} 1 \\ \delta \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ -\delta \\ 0 \\ \delta \end{bmatrix}. \text{ Then } r_{23} = \begin{bmatrix} 0 & -\frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \end{bmatrix} \begin{bmatrix} 1 \\ 0 \\ 0 \\ \delta \end{bmatrix} = 0 \text{ so } \mathbf{y} \text{ is unchanged.}$$

Then $r_{33} = \|\mathbf{y}\|_2 = \sqrt{2}\delta$ and $\mathbf{q}_3 = \begin{bmatrix} 0 \\ -\frac{\sqrt{2}}{2} \\ 0 \\ \frac{\sqrt{2}}{2} \end{bmatrix}$.

$$Q = \begin{bmatrix} 1 & 0 & 0 \\ \delta & -\frac{\sqrt{2}}{2} & -\frac{\sqrt{2}}{2} \\ 0 & \frac{\sqrt{2}}{2} & 0 \\ 0 & 0 & \frac{\sqrt{2}}{2} \end{bmatrix} \quad R = \begin{bmatrix} 1 & 1 & 1 \\ 0 & \sqrt{2}\delta & 0 \\ 0 & 0 & \sqrt{2}\delta \end{bmatrix}$$

Example 2. The check that $A = QR$ works out fine here, but recall that the goal is for Q to have orthogonal columns so that Q^T is the inverse of Q . Does Q have orthogonal columns?

$$\mathbf{q}_2 \cdot \mathbf{q}_3 = \begin{bmatrix} 0 & -\frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \end{bmatrix} \begin{bmatrix} 0 \\ -\frac{\sqrt{2}}{2} \\ 0 \\ \frac{\sqrt{2}}{2} \end{bmatrix} = \frac{1}{2}, \text{ which is nowhere near } 0.$$

A modified version of the Gram-Schmidt Algorithm does better in this example. The algorithms produce identical answers in exact arithmetic. The only change is at the line where we update \mathbf{y} : use the current version of \mathbf{y} instead of the original column A_j .

Algorithm 1 Modified Gram-Schmidt Algorithm

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for  $j = 1, \dots, n$  do
   $\mathbf{y} = A_j$ 
  for  $i = 1, \dots, j - 1$  do
     $r_{ij} = \mathbf{q}_i^T \mathbf{y}$  ▷ Instead of the original column, use the updated column
     $\mathbf{y} = \mathbf{y} - r_{ij}\mathbf{q}_i$ 
   $r_{jj} = \|\mathbf{y}\|_2$ 
   $\mathbf{q}_j = \frac{1}{r_{jj}}\mathbf{y}$ 

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Example 3. Recalculate the (reduced) QR factorization of $\begin{bmatrix} 1 & 1 & 1 \\ \delta & 0 & 0 \\ 0 & \delta & 0 \\ 0 & 0 & \delta \end{bmatrix}$ where $\delta = 10^{-10}$,

using Modified Gram-Schmidt in double precision.

For the first two columns, there are no changes in the algorithm (if the norm of the first column isn't one, then you do have a new result in the second column). So we pick up where

$$Q = \begin{bmatrix} 1 & 0 \\ \delta & -\frac{\sqrt{2}}{2} \\ 0 & \frac{\sqrt{2}}{2} \\ 0 & 0 \end{bmatrix} \quad R = \begin{bmatrix} 1 & 1 \\ 0 & \sqrt{2}\delta \\ 0 & 0 \end{bmatrix}.$$

We also still have $r_{13} = 1$ and $\mathbf{y} = \begin{bmatrix} 0 \\ -\delta \\ 0 \\ \delta \end{bmatrix}$. Then $r_{23} = \begin{bmatrix} 0 & -\frac{\sqrt{2}}{2} & \frac{\sqrt{2}}{2} & 0 \end{bmatrix} \begin{bmatrix} 0 \\ -\delta \\ 0 \\ \delta \end{bmatrix} = \frac{\sqrt{2}}{2}\delta$ and

$$\mathbf{y} = \begin{bmatrix} 0 \\ -\delta \\ 0 \\ \delta \end{bmatrix} - \frac{\sqrt{2}}{2}\delta \begin{bmatrix} 0 \\ -\frac{\sqrt{2}}{2} \\ \frac{\sqrt{2}}{2} \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ -\frac{1}{2}\delta \\ -\frac{1}{2}\delta \\ \delta \end{bmatrix}. \text{ The norm of the new } \mathbf{y} \text{ is } \left(\sqrt{\frac{1}{4} + \frac{1}{4} + 1}\right) \delta = \frac{\sqrt{6}}{2}\delta.$$

Updating \mathbf{y} for Q gives the factorization

$$Q = \begin{bmatrix} 1 & 0 & 0 \\ \delta & -\frac{\sqrt{2}}{2} & -\frac{\sqrt{6}}{6} \\ 0 & \frac{\sqrt{2}}{2} & -\frac{\sqrt{6}}{6} \\ 0 & 0 & \frac{\sqrt{6}}{3} \end{bmatrix} \quad R = \begin{bmatrix} 1 & 1 & 1 \\ 0 & \sqrt{2}\delta & \frac{\sqrt{2}}{2}\delta \\ 0 & 0 & \frac{\sqrt{6}}{2}\delta \end{bmatrix}.$$

How is orthogonality in Q now?

Well, $\mathbf{q}_2 \cdot \mathbf{q}_3 = \frac{\sqrt{2(6)}}{12} - \frac{\sqrt{2(6)}}{12} = 0$, so that's much better.

In both versions, $\mathbf{q}_1 \cdot \mathbf{q}_2 = -\frac{\sqrt{2}}{2}\delta$, which... leaves room for improvement.

Householder Reflectors

The next (final) process of finding a QR factorization bears a closer resemblance to the process of Gaussian Elimination, in that we are manipulating the matrix A into upper triangular form R . However, we cannot get to R using row operations, as that will not introduce orthogonality.

The theory is to multiply A on the left by orthogonal matrices until it is transformed into R . That is, after multiplying by one orthogonal matrix Q_1 , we should have:

$$Q_1 A = \begin{bmatrix} r_{11} & ? & \dots & ? \\ 0 & ? & \dots & ? \\ \vdots & & & \vdots \\ 0 & ? & \dots & ? \end{bmatrix}$$

The orthogonal matrices used to transform A are a type of orthogonal matrix called a Householder reflector, denoted by H . Householder reflectors are also symmetric, so not only is $H^T = H^{-1}$, $H^T = H$ implies $H = H^{-1}$.

So we'll have a series of Householder reflectors H_1, \dots, H_n where

$$H_n \dots H_2 H_1 A = R,$$

and then because each H is orthogonal and symmetric,

$$A = H_1 H_2 \dots H_n R.$$

Example 4. Verify that the product $H_1 \dots H_n$ is an orthogonal matrix.

Using the theorem/definition that Q is orthogonal if $Q^{-1} = Q^T$, we check the product

$$(H_1 \dots H_n)^T (H_1 \dots H_n) = H_n \dots H_2 H_1 H_1 H_2 \dots H_n = H_n \dots H_2 H_2 \dots H_n = I.$$

Therefore, since $H_1 \dots H_n$ is square, $H_1 \dots H_n$ is orthogonal.

We still need more information to define a Householder reflector. We begin with \mathbf{x} equal to the first column of A ; we desire H that will reflect \mathbf{x} to the vector where the first element is nonzero, and all remaining entries are zero (basically, the x -axis of \mathbb{R}^n). So we require

$$H\mathbf{x} = \mathbf{u} \text{ and } \|\mathbf{x}\|_2 = \|\mathbf{u}\|_2.$$

Question 1. So what is the nonzero element of \mathbf{u} when $\mathbf{x} = A_1$?

It must be that $r_{11} = \|\mathbf{x}\|_2 = \|A_1\|_2$, or else the length would not be preserved.

More generally, the goal at a particular step of the process is to take a vector \mathbf{x} with m entries and reflect it over an $m - 1$ dimensional plane to another vector \mathbf{u} of the same length. (Draw triangle with \mathbf{x} and \mathbf{u})

Question 2. Verify that $\mathbf{u} - \mathbf{x}$ is orthogonal to $\mathbf{u} + \mathbf{x}$.

$$(\mathbf{u} - \mathbf{x}) \cdot (\mathbf{u} + \mathbf{x}) = \mathbf{u} \cdot (\mathbf{u} + \mathbf{x}) - \mathbf{x} \cdot (\mathbf{u} + \mathbf{x}) = \mathbf{u} \cdot \mathbf{u} + \mathbf{x} \cdot \mathbf{u} - \mathbf{x} \cdot \mathbf{u} - \mathbf{x} \cdot \mathbf{x} = \|\mathbf{u}\|_2^2 - \|\mathbf{x}\|_2^2 = 0$$

For convenience, let \mathbf{v} represent $\mathbf{u} - \mathbf{x}$. We want to reflect \mathbf{x} over $\mathbf{u} + \mathbf{x}$. To do so, we remove from \mathbf{x} , twice the projection of \mathbf{x} onto \mathbf{v} .

The projection of \mathbf{x} onto \mathbf{v} is traditionally given in linear algebra as

$$\frac{\mathbf{x} \cdot \mathbf{v}}{\mathbf{v} \cdot \mathbf{v}} \mathbf{v}.$$

We rewrite the projection formula in a matrix form (working towards our orthogonal matrix H). Since $\frac{\mathbf{x} \cdot \mathbf{v}}{\mathbf{v} \cdot \mathbf{v}}$ is a scalar, we can multiply by it second. We can also reverse the order of the vectors in the dot product, by the properties of the dot product.

$$\mathbf{v} \frac{\mathbf{v} \cdot \mathbf{x}}{\mathbf{v} \cdot \mathbf{v}} = \frac{1}{\mathbf{v} \cdot \mathbf{v}} \mathbf{v} (\mathbf{v}^T \mathbf{x}) = \frac{1}{\mathbf{v} \cdot \mathbf{v}} (\mathbf{v} \mathbf{v}^T) \mathbf{x} = \frac{\mathbf{v} \mathbf{v}^T}{\mathbf{v}^T \mathbf{v}} \mathbf{x}$$

Let P represent $\frac{\mathbf{v} \mathbf{v}^T}{\mathbf{v}^T \mathbf{v}}$ (which is now a matrix, not a scalar).

Definition 3. A *projection matrix* is a matrix that satisfies $P^2 = P$.

Question 4. Is P a projection matrix?

Left for homework!

$$\mathbf{u} = \mathbf{x} - 2P\mathbf{x} = (I - 2P)\mathbf{x}$$

So we define $H = I - 2P = I - 2\frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}}$, where \mathbf{v} is the difference between \mathbf{x} and the desired vector \mathbf{u} .

Example 5. Find a Householder reflector that transforms $\mathbf{x} = \begin{bmatrix} 3 \\ 4 \end{bmatrix}$ into an equal length vector on the x -axis.

First, the goal is to transform into the vector $\mathbf{u} = \begin{bmatrix} \|\mathbf{x}\|_2 \\ 0 \end{bmatrix}$, and here, $\|\mathbf{x}\|_2 = 5$. So $\mathbf{u} = \begin{bmatrix} 5 \\ 0 \end{bmatrix}$.

Now we subtract to form $\mathbf{v} = \begin{bmatrix} 2 \\ -4 \end{bmatrix}$.

Next, $P = \frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}} = \left(\frac{1}{20}\right) \begin{bmatrix} 4 & -8 \\ -8 & 16 \end{bmatrix}$. Then

$$H = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \frac{1}{10} \begin{bmatrix} 4 & -8 \\ -8 & 16 \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \frac{1}{5} \begin{bmatrix} 2 & -4 \\ -4 & 8 \end{bmatrix} = \begin{bmatrix} \frac{3}{5} & \frac{4}{5} \\ \frac{4}{5} & -\frac{3}{5} \end{bmatrix}.$$

We can check our answer:

$$H\mathbf{x} = \begin{bmatrix} \frac{3}{5} & \frac{4}{5} \\ \frac{4}{5} & -\frac{3}{5} \end{bmatrix} \begin{bmatrix} 3 \\ 4 \end{bmatrix} = \begin{bmatrix} \frac{9+16}{5} \\ \frac{12-12}{5} \end{bmatrix} = \begin{bmatrix} 5 \\ 0 \end{bmatrix} \text{ as desired.}$$

To complete a process for the QR factorization, we only need to know how to handle the later columns. On the first step, H_1 will be exactly as described above, and we form H_1A which looks like:

$$\begin{bmatrix} r_{11} & r_{12} & \dots & r_{1n} \\ 0 & ? & \dots & ? \\ \vdots & & & \vdots \\ 0 & ? & \dots & ? \end{bmatrix}$$

We repeat this process with just the vector below those that have been completed. So on the second step, we use the $n - 1$ elements from the diagonal down to generate a Householder reflector. However, doing so generates \hat{H}_2 of size $(n - 1) \times (n - 1)$. We want to apply \hat{H}_2 only to the elements below the completed rows, so to leave the previous rows unchanged we set H_2 as the lower part of a matrix with identity on the diagonal. That is, in general, the H_k Householder reflector is the block matrix

$$\begin{bmatrix} I_k & 0 \\ 0 & \hat{H}_k \end{bmatrix}.$$

Example 6. Compute the full QR factorization of $\begin{bmatrix} 2 & 0 \\ 1 & -1 \\ 2 & -7 \end{bmatrix}$.

First, $\mathbf{x} = \begin{bmatrix} 2 \\ 1 \\ 2 \end{bmatrix}$ with norm 3, so $\mathbf{u} = \begin{bmatrix} 3 \\ 0 \\ 0 \end{bmatrix}$ and $\mathbf{v} = \begin{bmatrix} 1 \\ -1 \\ -2 \end{bmatrix}$. Then

$$\begin{aligned} H_1 &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} - \left(\frac{2}{1+1+4} \right) \begin{bmatrix} 1 \\ -1 \\ -2 \end{bmatrix} \begin{bmatrix} 1 & -1 & -2 \end{bmatrix} \\ &= \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix} - \left(\frac{1}{3} \right) \begin{bmatrix} 1 & -1 & -2 \\ -1 & 1 & 2 \\ -2 & 2 & 4 \end{bmatrix} \\ &= \begin{bmatrix} \frac{2}{3} & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{3} & \frac{2}{3} & -\frac{2}{3} \\ \frac{2}{3} & -\frac{2}{3} & -\frac{1}{3} \end{bmatrix} \end{aligned}$$

Then $H_1 A = \begin{bmatrix} \frac{2}{3} & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{3} & \frac{2}{3} & -\frac{2}{3} \\ \frac{2}{3} & -\frac{2}{3} & -\frac{1}{3} \end{bmatrix} \begin{bmatrix} 2 & 0 \\ 1 & -1 \\ 2 & -7 \end{bmatrix} = \begin{bmatrix} 3 & -5 \\ 0 & 4 \\ 0 & 3 \end{bmatrix}$.

Next, $\mathbf{x} = \begin{bmatrix} 4 \\ 3 \end{bmatrix}$ with norm 5, so $\mathbf{u} = \begin{bmatrix} 5 \\ 0 \end{bmatrix}$ and $\mathbf{v} = \begin{bmatrix} 1 \\ -3 \end{bmatrix}$.

Then,

$$\begin{aligned} \hat{H}_2 &= \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix} - \frac{2}{10} \begin{bmatrix} 1 & -3 \\ -3 & 9 \end{bmatrix} \\ &= \begin{bmatrix} \frac{4}{5} & \frac{3}{5} \\ \frac{3}{5} & -\frac{4}{5} \end{bmatrix}. \end{aligned}$$

This makes $H_2 = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{4}{5} & \frac{3}{5} \\ 0 & \frac{3}{5} & -\frac{4}{5} \end{bmatrix}$. Then R equals $H_2 H_1 A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{4}{5} & \frac{3}{5} \\ 0 & \frac{3}{5} & -\frac{4}{5} \end{bmatrix} \begin{bmatrix} 3 & -5 \\ 0 & 4 \\ 0 & 3 \end{bmatrix} = \begin{bmatrix} 3 & -5 \\ 0 & 5 \\ 0 & 0 \end{bmatrix}$.

Then Q of the QR factorization is $H_1 H_2 = \begin{bmatrix} \frac{2}{3} & \frac{1}{3} & \frac{2}{3} \\ \frac{1}{3} & \frac{2}{3} & -\frac{2}{3} \\ \frac{2}{3} & -\frac{2}{3} & -\frac{1}{3} \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & \frac{4}{5} & \frac{3}{5} \\ 0 & \frac{3}{5} & -\frac{4}{5} \end{bmatrix} = \frac{1}{15} \begin{bmatrix} 10 & 10 & -5 \\ 5 & 2 & 14 \\ 10 & -11 & -2 \end{bmatrix}$.

We should be able to check: $\frac{1}{15} \begin{bmatrix} 10 & 10 & -5 \\ 5 & 2 & 14 \\ 10 & -11 & -2 \end{bmatrix} \begin{bmatrix} 3 & -5 \\ 0 & 5 \\ 0 & 0 \end{bmatrix} = \frac{1}{15} \begin{bmatrix} 30 & 0 \\ 15 & -15 \\ 30 & -105 \end{bmatrix} = \begin{bmatrix} 2 & 0 \\ 1 & -1 \\ 2 & -7 \end{bmatrix}$.

A note on full vs. reduced: Householder Reflectors are naturally producing a full factorization. If you wanted to give a reduced factorization, take the answer and erase any below diagonal elements in R and the corresponding columns in Q . For Example 6, we have

$$R = \begin{bmatrix} 3 & -5 \\ 0 & 5 \\ 0 & 0 \end{bmatrix} \text{ which becomes } R = \begin{bmatrix} 3 & -5 \\ 0 & 5 \end{bmatrix}$$

and

$$Q = \frac{1}{15} \begin{bmatrix} 10 & 10 & -5 \\ 5 & 2 & 14 \\ 10 & -11 & -2 \end{bmatrix} \text{ which becomes } Q = \frac{1}{15} \begin{bmatrix} 10 & 10 \\ 5 & 2 \\ 10 & -11 \end{bmatrix}.$$

I leave it to you to check again that $QR = A$.

Stability and Operation Counts

Computing a QR factorization by Householder reflectors takes about $2mn^2 - \frac{2}{3}n^3$ operations, whereas using Gram-Schmidt is about $2mn^2$. It also typically requires less memory.

As for numerical stability, the Householder reflectors method is used in practice and is known to deliver better orthogonality in Q .

Question 5. In practice, routines will use $\pm\|\mathbf{x}\|_2$ where the sign is chosen to be the opposite of the first element of \mathbf{x} . Why might this be better, numerically, than always choosing the positive norm?

It avoids subtracting nearly equal numbers, which we discussed previously may result in loss of significance.