

Concept:

zero all subdiagonal entries of each active column with one direct rotation.

Ⓐ

Input an $m \times n$ matrix A_1 and find an $m \times m$ matrix U_1 such that

- $U_1^T = U_1^{-1}$ with $\det(U_1) = 1$
- $A_2 = U_1 A_1$
- $\text{col1}(A_2) = r_{1,1} \vec{e}_1$ where $r_{1,1}$ is a scalar
- $|r_{1,1}| = \ell_1(\text{length}(\text{col1}(A_1)))$

In other words, U rotates A_1 so that its first column aligns with $\vec{e}_1$.

If $\text{col1}(A_1)$ is already a multiple of $\vec{e}_1$,

$$U_1 = I$$

In this case, the diagonal entry $r_{1,1}$

of the resultant matrix A_2 may be positive, negative or 0.

Otherwise,

①

We define the rotation plane spanned by $\text{col1}(A_1)$ and $\vec{e}_1$.

②

Construct the second unit vector $\hat{u}_2$ inside the plane such that

- $\hat{u}_2 \cdot \vec{e}_1 = 0$
- $|\hat{u}_2| = 1$

$\hat{u}_2$ can be obtained by subtracting the projection of $\text{col1}(A_1)$ onto $\vec{e}_1$ and normalizing the resultant vector

③

Define an orthonormal basis for the plane:

$$B_1 = [\vec{e}_1 \mid \hat{u}_2]$$

④

Compute the angle θ between the direction of $\text{col1}(A_1)$ and $\vec{e}_1$ using their normalized dot product

⑤

Extend the plane rotation to a rotation U_1 of $\mathbb{R}^m$

Denote J_1 as the 2×2 rotation by θ :

$$\begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix}$$

in the coordinate system described by B_1 .

U_1 performs the following geometric actions:

- ① leaves every vector perpendicular to the plane described by B_1 unchanged
- ② rotates every vector inside the plane by θ using $B_1 J_1 B_1^T$

Any vector $\vec{v}$ can be decomposed as

$$\vec{v} = \vec{v}_{\parallel} + \vec{v}_{\perp}.$$

Since the columns of B_1 are orthonormal,

$$B_1^T B_1 = I_2.$$

The two components are

- $\vec{v}_{\parallel}$ is the projection onto the plane described by B_1 :

$$\vec{v}_{\parallel} = B_1 B_1^T \vec{v}$$

- $\vec{v}_{\perp}$ is perpendicular to that plane:

$$\begin{aligned}\vec{v}_{\perp} &= \vec{v} - \vec{v}_{\parallel} \\ &= (I_m - B_1 B_1^T) \vec{v}\end{aligned}$$

U_1 transforms these two components as follows:

- the component inside the plane is rotated:

$$\vec{v}'_{\parallel} = B_1 J_1 B_1^T \vec{v}$$

- the component perpendicular to the plane is unchanged:

$$\vec{v}'_{\perp} = \vec{v}_{\perp} = (I_m - B_1 B_1^T) \vec{v}$$

↓

$$\vec{v}' = \vec{v}'_{\parallel} + \vec{v}'_{\perp} =$$

$$B_1 J_1 B_1^T \vec{v} + (I_m - B_1 B_1^T) \vec{v}$$

↓

$$U_1 = B_1 J_1 B_1^T + (I_m - B_1 B_1^T)$$

⑥

Simplify $U = B J B^T + (I_m - B B^T)$:

$$\begin{aligned} U &= I_m + B J B^T - B B^T = \\ &= I_m + B (J - I_2) B^T \end{aligned}$$

⑦

Apply U_1 to A_1 :

$$A_2 = U_1 A_1.$$

The first column of A_2 is now complete:

$$\text{col1}(A_2) = r_{1,1} \vec{e}_1.$$

↓

$$A_2 = \left[\begin{array}{c|c} r_{1,1} & \star^T \\ \hline \vec{0} & \tilde{A}_2 \end{array} \right]$$

The first column has the required upper-triangular form.
It does not need to be changed again.

Deflate A_2 by setting aside its first row and first column.

The remaining active matrix is the trailing submatrix $\tilde{A}_2$,

which has dimensions $(m - 1) \times (n - 1)$.

Repeat construction ① on $\tilde{A}_2$.

This produces an $(m - 1) \times (m - 1)$ rotation $\tilde{U}_2$
that transforms the first column of $\tilde{A}_2$ into a multiple
of its first coordinate vector

$$\text{Embed } \tilde{U}_2 \text{ into an } m \times m \text{ matrix } U_2 = \left[\begin{array}{c|c} 1 & \vec{0}^T \\ \hline \vec{0} & \tilde{U}_2 \end{array} \right]$$

Thus U_2 leaves the first row of A_2 unchanged
and acts only on the deflated rows.

Set

$$A_3 = U_2 A_2.$$

Using the block forms of U_2 and A_2 :

↓

$$A_3 = \left[\begin{array}{c|c} 1 & \vec{0}^T \\ \hline \vec{0} & \tilde{U}_2 \end{array} \right] \left[\begin{array}{c|c} r_{1,1} & \vec{*}^T \\ \hline \vec{0} & \tilde{A}_2 \end{array} \right] = \left[\begin{array}{c|c} r_{1,1} & \vec{*}^T \\ \hline \vec{0} & \tilde{U}_2 \tilde{A}_2 \end{array} \right]$$

By construction,

the first column of $\tilde{U}_2 \tilde{A}_2$ is a multiple of its first coordinate vector.



Therefore, the first two columns of A_3 have upper-triangular form

Repeat this deflation and rotation process for each remaining active column

- The product of the transposed rotations forms an orthogonal matrix Q
- The final upper-triangular or upper-trapezoidal matrix R satisfies $A_1 = Q R$



Optimization, page 1

On this page, m denotes the number of active rows,
and we omit the iteration subscripts.

For each column, the rotation matrix was derived as

$$U = I_m + B (J - I_2) B^T.$$

The following quantities have already been defined (not yet computed):

- $\ell = \text{length}(\text{col1}(A))$
 - $c = \cos(\theta)$
 - $s = \sin(\theta)$
 - $B = [\vec{e}_1 \mid \hat{u}_2]$

Let

$$d = c - 1$$

and

$$D = J - I_2.$$

↓

$$D = J - I_2 = \left[\begin{array}{c|c} d & s \\ \hline -s & d \end{array} \right]$$

↓

$$U = I_m + B D B^T$$

$$B D = \left[\begin{array}{c|c} \vec{e}_1 & \hat{u}_2 \end{array} \right] \left[\begin{array}{c|c} d & s \\ \hline -s & d \end{array} \right] = \left[\begin{array}{c|c} d \vec{e}_1 - s \hat{u}_2 & s \vec{e}_1 + d \hat{u}_2 \end{array} \right]$$

$$B D B^T = \left[\begin{array}{c|c} d \vec{e}_1 - s \hat{u}_2 & s \vec{e}_1 + d \hat{u}_2 \end{array} \right] \begin{bmatrix} \vec{e}_1^T \\ \hat{u}_2^T \end{bmatrix}$$

$$= d \vec{e}_1 \vec{e}_1^T - s \hat{u}_2 \vec{e}_1^T + s \vec{e}_1 \hat{u}_2^T + d \hat{u}_2 \hat{u}_2^T.$$

↓

$$U = I_m + d \vec{e}_1 \vec{e}_1^T - s \hat{u}_2 \vec{e}_1^T + s \vec{e}_1 \hat{u}_2^T + d \hat{u}_2 \hat{u}_2^T.$$

$$\text{Since the first entry of } \hat{u}_2 \text{ is zero, } \hat{u}_2 = \begin{bmatrix} 0 \\ \bar{u}_2 \end{bmatrix}$$

where $\bar{u}_2$ contains entries 2 through m of $\hat{u}_2$.

Using the block forms of $\vec{e}_1$ and $\hat{u}_2$,

the four outer-product terms affect U as follows:

$$d \vec{e}_1 \vec{e}_1^T = \left[\begin{array}{c|c} d & \vec{0}^T \\ \hline \vec{0} & 0_{m-1} \end{array} \right]$$

$$s \vec{e}_1 \hat{u}_2^T = \left[\begin{array}{c|c} 0 & s \bar{u}_2^T \\ \hline \vec{0} & 0_{m-1} \end{array} \right]$$

$$-s \hat{u}_2 \vec{e}_1^T = \left[\begin{array}{c|c} 0 & \vec{0}^T \\ \hline -s \bar{u}_2 & 0_{m-1} \end{array} \right]$$

$$d \hat{u}_2 \hat{u}_2^T = \left[\begin{array}{c|c} 0 & \vec{0}^T \\ \hline \vec{0} & d \bar{u}_2 \bar{u}_2^T \end{array} \right]$$

$$1 + d = c$$

$$\downarrow$$

$$U = I_m + d \vec{e}_1 \vec{e}_1^T - s \hat{u}_2 \vec{e}_1^T + s \vec{e}_1 \hat{u}_2^T + d \hat{u}_2 \hat{u}_2^T = \left[\begin{array}{c|c} c & s \bar{u}_2^T \\ \hline -s \bar{u}_2 & I_{m-1} + d \bar{u}_2 \bar{u}_2^T \end{array} \right]$$

We can now summarize the quantities that must actually be computed.

Let $a = (\text{col1}(A))_1$, and let $\bar{p}$ contain entries 2 through m

of $\text{col1}(A)$. No projection computation is required.

- $\rho^2 = \bar{\mathbf{p}}^T \bar{\mathbf{p}}$
- $\ell = \sqrt{a^2 + \rho^2}$
- $\ell^{-1} = \frac{1}{\ell}$
- $c = a \ell^{-1}$
- $\bar{\mathbf{g}} = \bar{\mathbf{p}} \ell^{-1}$
- $h = \frac{c - 1}{\rho^2}$

For m active rows, computing these quantities requires

- $2m$ multiplications
- m additions and subtractions
- 2 divisions
- 1 square root.

Therefore, the block form becomes $U = \left[\begin{array}{c|c} c & \bar{\mathbf{g}}^T \\ \hline -\bar{\mathbf{g}} & I_{m-1} + h \bar{\mathbf{p}} \bar{\mathbf{p}}^T \end{array} \right]$

The quantities c , $\bar{\mathbf{g}}$, and h determine the block matrix U .

However, instead of multiplying the quantities to compute U ,
we will derive the algebraic product $U A$



Optimization, page 2

The block form derived on the preceding page contains two quantities that can be simplified before they are computed:

- $s \bar{u}_2$
- $d \bar{u}_2 \bar{u}_2^\top$.

Let $\bar{p}$ contain entries 2 through m of $\text{col1}(A)$.
Since subtracting the projection onto $\vec{e}_1$ only removes the first entry of $\text{col1}(A)$,

$$\vec{p} = \text{col1}(A) - (\text{col1}(A))_1 \vec{e}_1 = \begin{bmatrix} 0 \\ \bar{p} \end{bmatrix}$$

Let
 $\rho = \text{length}(\vec{p})$.

Then

$$\bar{u}_2 = \frac{\bar{p}}{\rho}$$

and

$$s = \frac{\rho}{\ell}.$$

Simplify $s \bar{u}_2$:

$$\begin{aligned} s \bar{u}_2 &= \begin{pmatrix} \rho \\ \ell \end{pmatrix} \begin{pmatrix} \bar{p} \\ \rho \end{pmatrix} \\ &= \frac{\bar{p}}{\ell}. \end{aligned}$$

Therefore, define

$$\bar{g} = \frac{\bar{p}}{\ell}.$$

Then

$$s \bar{u}_2 = \bar{g},$$

so neither s nor ρ needs to be computed
for the off-diagonal blocks of U .

②

Simplify $d \bar{u}_2 \bar{u}_2^T$:

$$\begin{aligned} &d \bar{u}_2 \bar{u}_2^T \\ &= d \begin{pmatrix} \bar{p} \\ \rho \end{pmatrix} \begin{pmatrix} \bar{p}^T \\ \rho \end{pmatrix} \\ &= \begin{pmatrix} d \\ \rho^2 \end{pmatrix} \bar{p} \bar{p}^T. \end{aligned}$$

Only ρ^2 is required in this expression.
The square root $\rho = \text{length}(\vec{p})$ is not required.

Since
 $\rho^2 = \bar{\mathbf{p}}^T \bar{\mathbf{p}}$,
the same sum of squares used in

$$\ell^2 = \left(\text{col1}(\mathbf{A}) \right)_1^2 + \rho^2$$

can be reused directly.

Thus, let

$$h = \frac{d}{\rho^2}.$$

The two simplified quantities are

$$s \bar{\mathbf{u}}_2 = \bar{\mathbf{g}}$$

and

$$d \bar{\mathbf{u}}_2 \bar{\mathbf{u}}_2^T = h \bar{\mathbf{p}} \bar{\mathbf{p}}^T.$$

Therefore, U may be written as

$$\left[\begin{array}{c|c} \mathbf{c} & \bar{\mathbf{g}}^T \\ \hline -\bar{\mathbf{g}} & \mathbf{I}_{m-1} + h \bar{\mathbf{p}} \bar{\mathbf{p}}^T \end{array} \right]$$

where

$$\bar{\mathbf{g}} = \frac{\bar{\mathbf{p}}}{\ell}$$

and

$$h = \frac{c - 1}{\rho^2}.$$

The rotation can therefore be computed using only one square root:

$$\ell = \text{length}(\text{col1}(A)).$$



Optimization, page 3

Algebraic derivation of $U A$

This page concerns the general case $\rho^2 > 0$.
In the zero and parallel cases, $U = I_m$ and $U A = A$.

Separate the first column of A from its remaining columns:

$$A = [\text{col1}(A) \mid C]$$

The first column of $U A$ is already known from the construction of U :

$$U \text{col1}(A) = \ell \vec{e}_1$$



Only the remaining columns C must be transformed.

Partition C into its first row and lower block,
and write the first column of A using a and $\bar{p}$:



$$C = \begin{bmatrix} \vec{b}^T \\ C_b \end{bmatrix} \quad A = \begin{bmatrix} a & \vec{b}^T \\ \bar{p} & C_b \end{bmatrix}$$

↓

$$U C = \left[\begin{array}{c|c} c & \bar{g}^T \\ \hline -\bar{g} & I_{m-1} + h \bar{p} \bar{p}^T \end{array} \right] \begin{bmatrix} \bar{b}^T \\ C_b \end{bmatrix} = \begin{bmatrix} c \bar{b}^T + \bar{g}^T C_b \\ -\bar{g} \bar{b}^T + C_b + h \bar{p} \bar{p}^T C_b \end{bmatrix}$$

The row vector $\bar{p}^T C_b$ is used in both blocks.

Compute it once:

$$\vec{z}^T = \bar{p}^T C_b$$

Since $\bar{g} = \bar{p} \ell^{-1}$,

$$\bar{g}^T C_b = \ell^{-1} \vec{z}^T.$$

↓

$$U C = \begin{bmatrix} c \bar{b}^T + \ell^{-1} \vec{z}^T \\ -\bar{g} \bar{b}^T + C_b + h \bar{p} \vec{z}^T \end{bmatrix} = \begin{bmatrix} c \bar{b}^T + \ell^{-1} \vec{z}^T \\ C_b + (-\bar{g} \bar{b}^T + h \bar{p} \vec{z}^T) \end{bmatrix}$$

Combining the known first column with $U C$ gives

$$U A = \left[\begin{array}{c|c} \ell & c \bar{b}^T + \ell^{-1} \vec{z}^T \\ \hline \vec{0} & C_b - \bar{g} \bar{b}^T + h \bar{p} \vec{z}^T \end{array} \right]$$

Thus U is never constructed.

The active matrix is updated using one row-vector product

and scaled outer-product terms.

Since $\bar{g} = \bar{p} \ell^{-1}$, define
 $\vec{w}^T = -\ell^{-1} \vec{b}^T + h \vec{z}^T$.

Then
 $-\bar{g} \vec{b}^T + h \bar{p} \vec{z}^T = \bar{p} \vec{w}^T$.
 $\downarrow$

$$U C = \begin{bmatrix} c \vec{b}^T + \ell^{-1} \vec{z}^T \\ C_b + \bar{p} \vec{w}^T \end{bmatrix}$$

For an active $m \times n$ matrix, computing $U C$ requires

- $(2m + 2)(n - 1)$ multiplications
- $(2m - 1)(n - 1)$ additions and subtractions

Including the quantities computed on the preceding page,
one complete active step requires

- $2m + (2m + 2)(n - 1)$ multiplications
- $m + (2m - 1)(n - 1)$ additions and subtractions
 - 2 divisions
 - 1 square root.

The first column of $U A$ is now complete.

Deflate $U A$ by setting aside its first row and first column,
and repeat the same construction on the trailing
 $(m - 1) \times (n - 1)$ active matrix.

When no active column has entries below its diagonal position,
return the resulting upper-triangular or upper-trapezoidal matrix R
and the accumulated orthogonal matrix Q .



Computation cost: comparison with Givens and Householder methods (Full explicit Q and R for an $n \times n$ matrix)

Method	Givens rotations	Direct plane rotations	Householder reflections
Multiplications	$\approx \left(\frac{10}{3}\right)n^3$	$\left(\frac{5}{3}\right)n^3 + \left(\frac{9}{2}\right)n^2 - \left(\frac{25}{6}\right)n - 2$	$\approx \left(\frac{4}{3}\right)n^3$
Additions and subtractions	$\approx \left(\frac{5}{3}\right)n^3$	$\left(\frac{5}{3}\right)n^3 - \left(\frac{2}{3}\right)n - 1$	$\approx \left(\frac{4}{3}\right)n^3$
Flops (multiplications + additions and subtractions)	$\approx 5n^3$	$\left(\frac{10}{3}\right)n^3 + \left(\frac{9}{2}\right)n^2 - \left(\frac{29}{6}\right)n - 3$	$\approx \left(\frac{8}{3}\right)n^3$
Divisions	$\approx n(n - 1)$	$2(n - 1)$	Implementation-dependent lower-order terms
Square roots	$\frac{n(n - 1)}{2}$	$n - 1$	$n - 1$
Weighted cost	$\approx 5n^3 + 12n(n - 1)$	$\left(\frac{10}{3}\right)n^3 + \left(\frac{9}{2}\right)n^2 + \left(\frac{115}{6}\right)n - 27$	$\geq \left(\frac{8}{3}\right)n^3 + 8(n - 1)$

Blue: best cost among the three methods in that operation category.
Green: direct plane rotations improve on Givens rotations, but not on Householder.

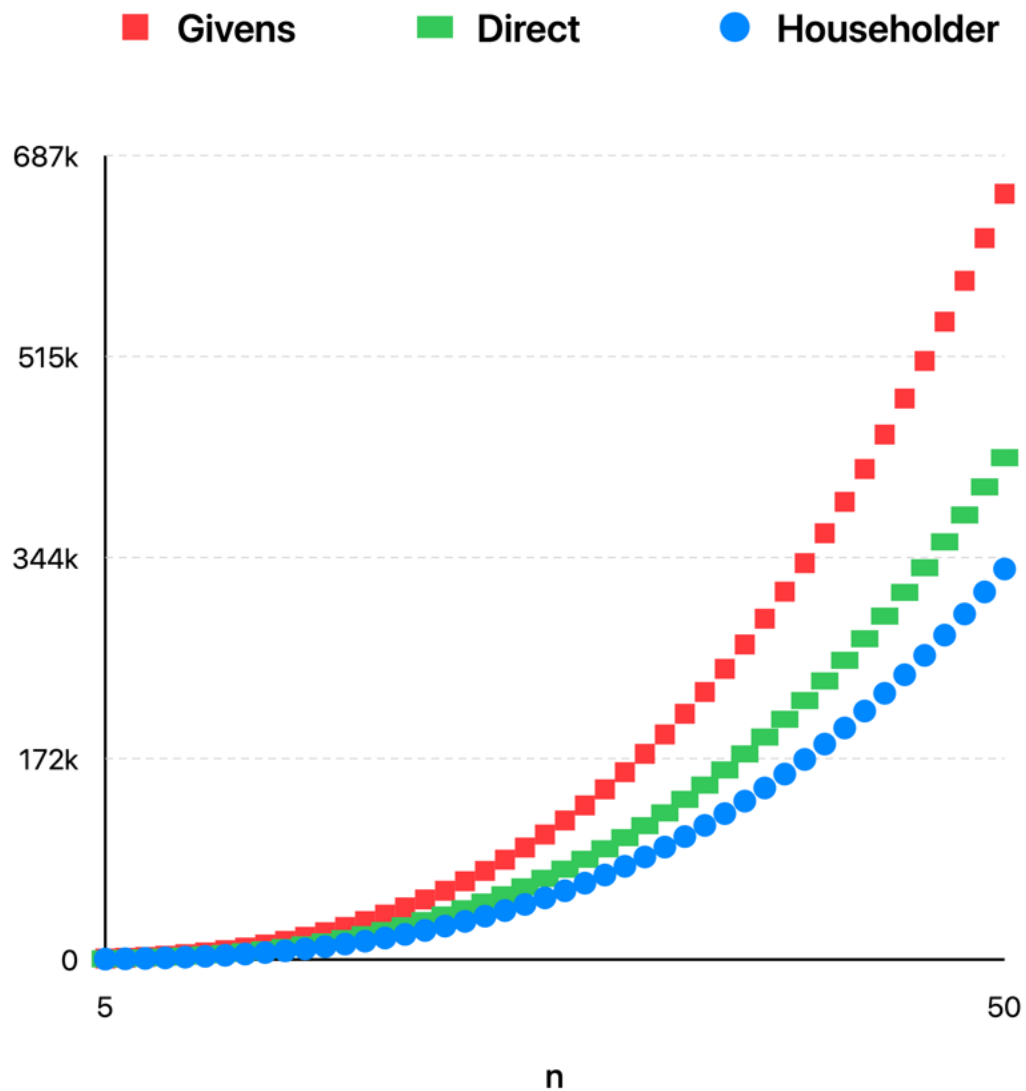
Weighted-cost convention:

- additions, subtractions, and multiplications are assigned weight 1
- divisions and square roots are assigned weight 8

These weights are approximate and hardware-dependent.
The plot should be interpreted as a comparative estimate,
not as measured execution time.

The Householder value is a lower bound because its
implementation-dependent division count is not included.

The plot shows estimated weighted arithmetic cost for dense square matrices
where
 $5 \leq n \leq 50$



There may be significant opportunities to improve efficiency by computing some of the quantities simultaneously.
We have not yet implemented these optimizations.

Open questions:

- ① How does numerical stability compare with Householder when the target column is close to $\vec{e}_1$?

If we are lucky and the comparison is favorable, a hybrid algorithm may be useful.

- ② How does efficiency compare with Givens rotations for sparse matrices?

If we are lucky and the comparison is favorable, a hybrid algorithm may also be useful.

- ③ Is there an advantage to a hybrid algorithm selecting among Householder, Givens, and direct plane rotations on a column-by-column basis?

- ④ Does this simplify computations for rank-1 QR updates?

- ⑤ Are there any other algebraic shortcuts that we have not implemented?

AI assistance:

AI tools were used extensively for information gathering, checking references, and assisting with operation counts.

All mathematical claims, formulas, visualizations, and final editorial decisions were reviewed and accepted by the authors.



