

ECSE 343 Numerical Methods in Engineering

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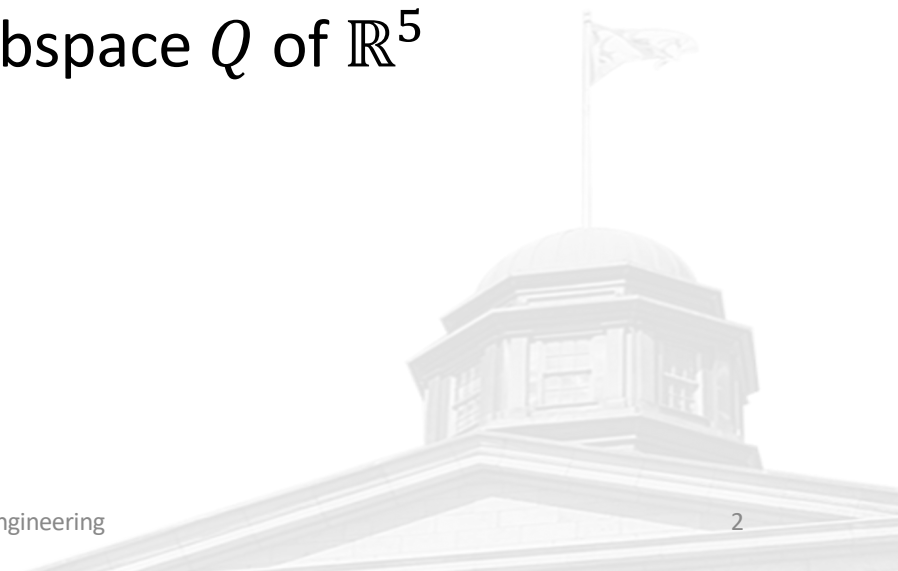
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Overdetermined Systems

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} \\ a_{21} & a_{22} & a_{23} \\ a_{31} & a_{32} & a_{33} \\ a_{41} & a_{42} & a_{43} \\ a_{51} & a_{52} & a_{53} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 \\ b_3 \\ b_4 \\ b_5 \end{bmatrix}$$

- Assume columns of A are linearly independent
- Columns of A form a 3D subspace Q of \mathbb{R}^5





Overdetermined Systems

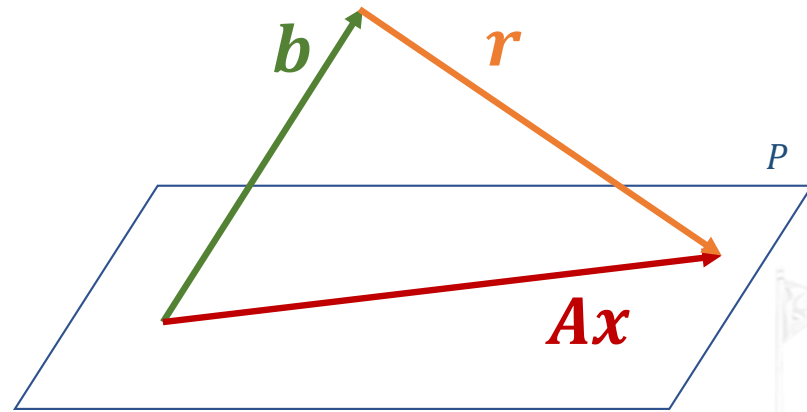
Find \mathbf{x} such that: $\mathbf{Ax} = \mathbf{b}$

Impossible:

$$\mathbf{Ax} \in P \quad \mathbf{b} \notin P$$

Minimize $\mathbf{r} = \mathbf{Ax} - \mathbf{b}$

i.e. minimize $\|\mathbf{r}\| = \|\mathbf{Ax} - \mathbf{b}\|$





Overdetermined Systems

Minimize $\|\mathbf{r}\| = \|\mathbf{Ax} - \mathbf{b}\|$

$$\Rightarrow \mathbf{r} \perp P$$

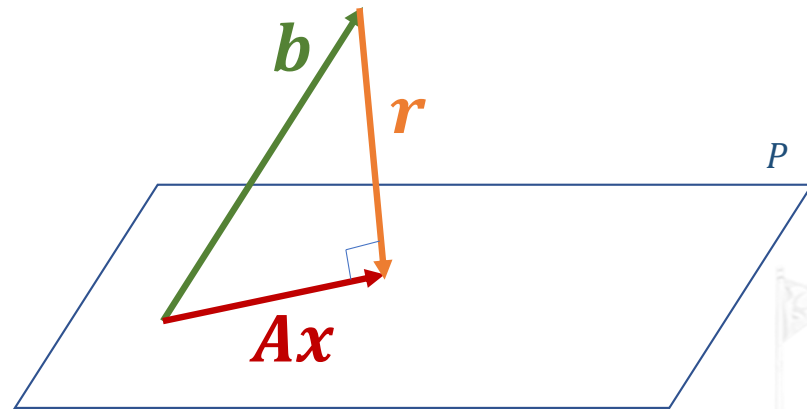
$$\Rightarrow \mathbf{A}^T \mathbf{r} = \mathbf{0}$$

$$\Rightarrow \mathbf{A}^T (\mathbf{Ax} - \mathbf{b}) = \mathbf{0}$$

$$\Rightarrow \mathbf{A}^T \mathbf{Ax} = \mathbf{A}^T \mathbf{b}$$

$$\Rightarrow \mathbf{x} = (\mathbf{A}^T \mathbf{A})^{-1} \mathbf{A}^T \mathbf{b}$$

Avoid $\mathbf{A}^T \mathbf{A}$ due to possible ill-conditioning





Overdetermined Systems

Minimize $\|\mathbf{r}\| = \|\mathbf{Ax} - \mathbf{b}\|$

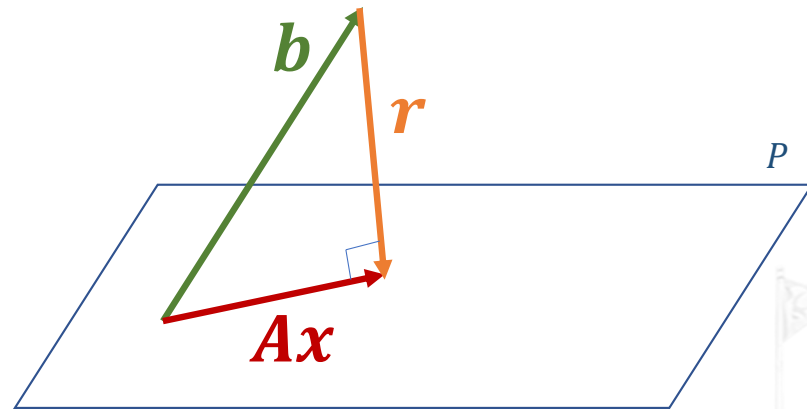
$$\Rightarrow \mathbf{r} \perp P$$

$$\Rightarrow \mathbf{Q}^T \mathbf{r} = \mathbf{0}$$

$$\Rightarrow \mathbf{Q}^T (\mathbf{Ax} - \mathbf{b}) = \mathbf{0}$$

$$\Rightarrow \boxed{\mathbf{Q}^T \mathbf{Ax} = \mathbf{Q}^T \mathbf{b}}$$

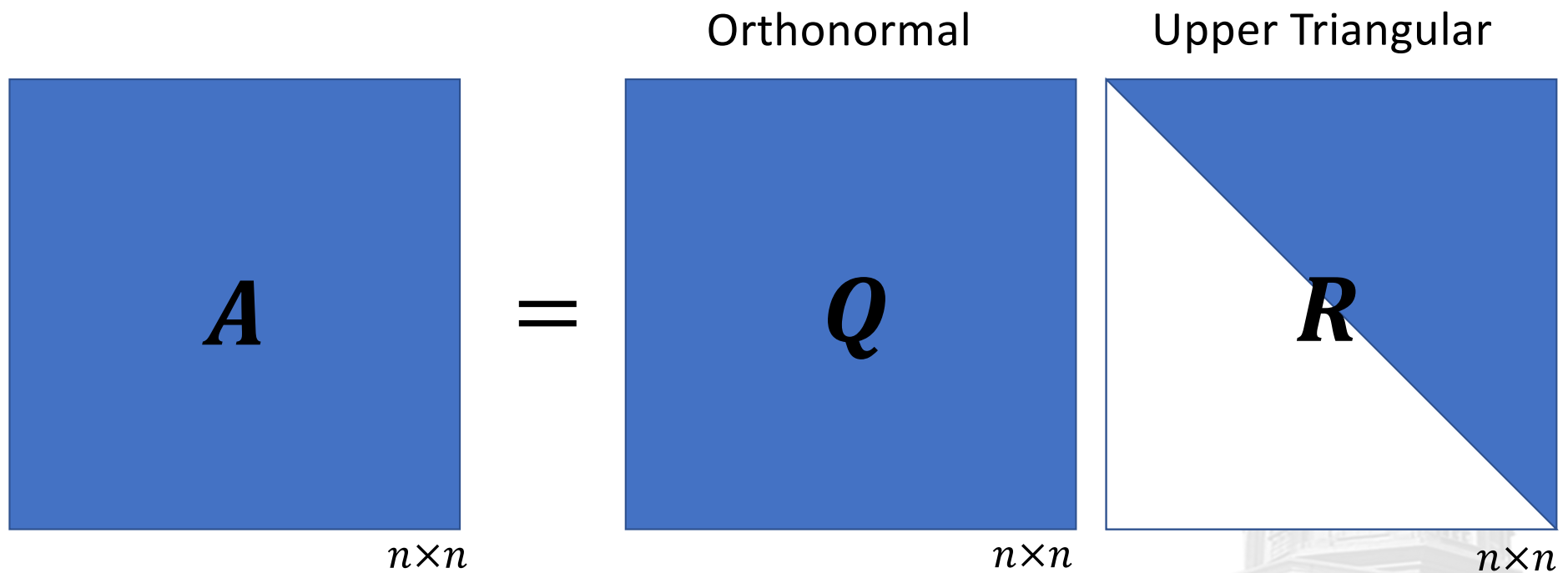
$$\Rightarrow \mathbf{Rx} = \mathbf{Q}^T \mathbf{b}$$



$$\mathbf{A} = \mathbf{QR}$$

$$\text{colsp}\{\mathbf{A}\} = \text{colsp}\{\mathbf{Q}\}$$

QR Decomposition





Modified Gram-Schmidt

$$\underbrace{A R_1 R_2 R_3 \cdots R_m}_{R^{-1}} = Q$$

$$A = QR$$





Another Approach

$$\underbrace{E_m \cdots E_4 E_3 E_2 E_1}_{\text{Elimination Matrices}} \mathbf{A} = \mathbf{U}$$

\mathbf{L}^{-1}

$$\underbrace{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}}_{\text{Elimination Matrices}} \underbrace{\begin{bmatrix} 2 & 1 & -2 \\ 4 & 5 & -3 \\ 6 & 9 & -2 \end{bmatrix}}_{\mathbf{A}} = \underbrace{\begin{bmatrix} 2 & 1 & -2 \\ 0 & 3 & +1 \\ 0 & 0 & +2 \end{bmatrix}}_{\mathbf{U}}$$



Another Approach

$$\underbrace{E_m \cdots E_4 E_3 E_2 E_1}_{\text{Elimination Matrices}} \mathbf{A} = \mathbf{U}$$

\mathbf{L}^{-1}

$$\underbrace{H_m \cdots H_4 H_3 H_2 H_1}_{\text{Orthonormal Matrices}} \mathbf{A} = \mathbf{R}$$

\mathbf{Q}^T





Matrix Vector Multiplication

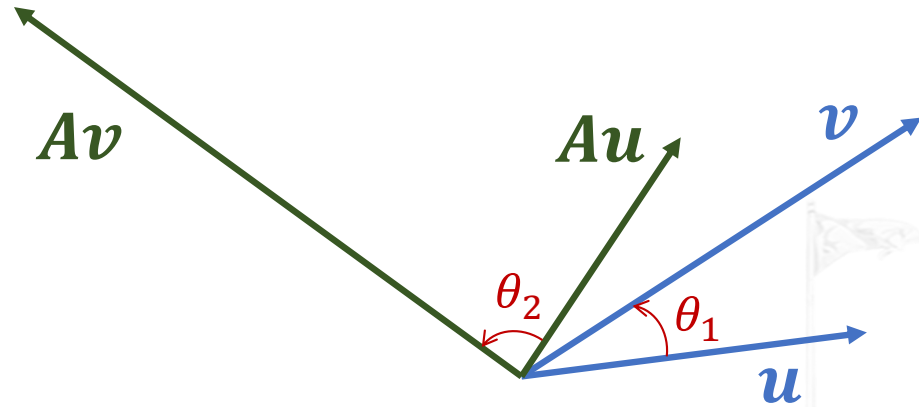
In General:

$$\|Av\| \neq \|v\|$$

$$\|Au\| \neq \|u\|$$


$$\frac{\|Au\|}{\|u\|} \neq \frac{\|Av\|}{\|v\|}$$

$$\theta_1 \neq \theta_2$$






Multiplication by an Orthonormal Matrix

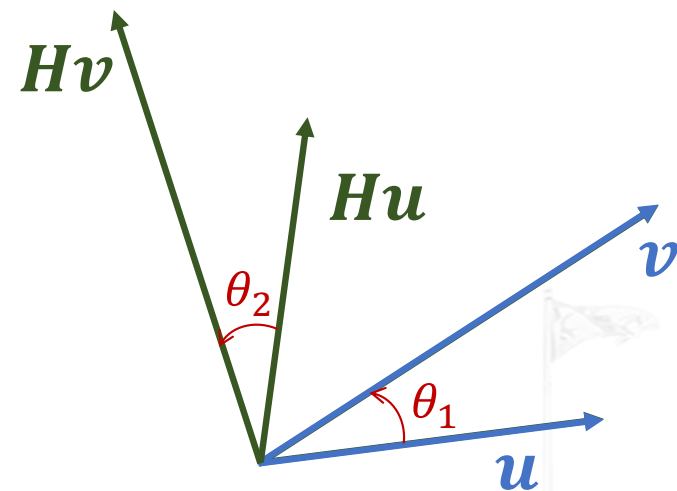

 $\|Hv\| = \|v\|$

$$(Hv)^T(Hv) = v^T \overbrace{H^T H}^U v = v^T v$$

Similarly: $\|Hu\| = \|u\|$


 $\theta_1 = \theta_2$

$$\left\{ \begin{array}{l} \langle v, u \rangle = \|v\| \|u\| \cos(\theta_1) \\ \langle Hv, Hu \rangle = \|Hv\| \|Hu\| \cos(\theta_2) = \|v\| \|u\| \cos(\theta_2) \\ \langle Hv, Hu \rangle = (Hu)^T (Hv) = u^T H^T H v = u^T v = \langle v, u \rangle \end{array} \right.$$



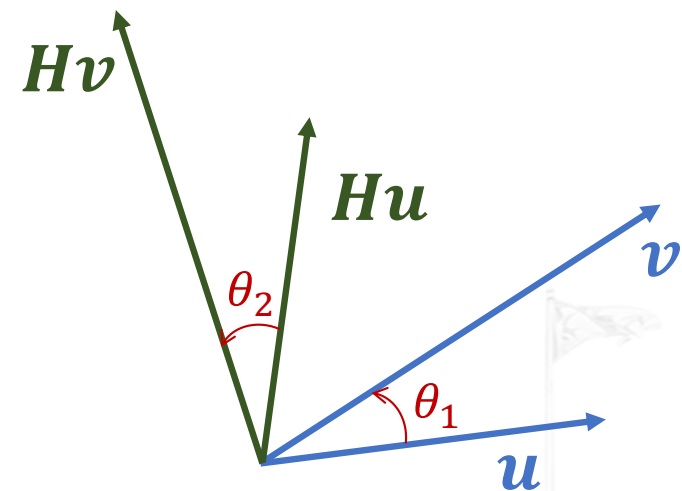


Multiplication by an Orthonormal Matrix

$$\|Hv\| = \|v\| \quad \|Hu\| = \|u\|$$

$$\theta_1 = \theta_2$$

- Rotation



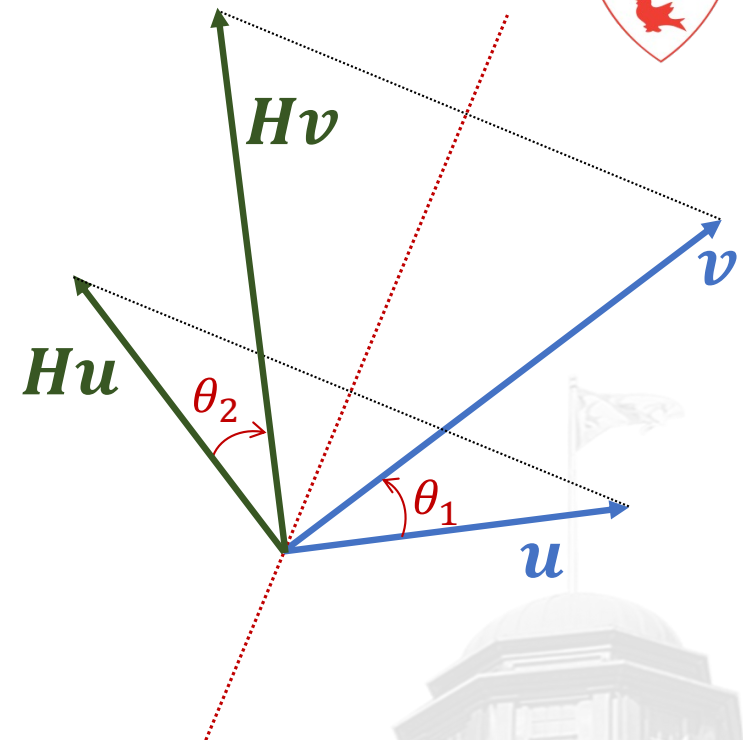


Multiplication by an Orthonormal Matrix

$$\|Hv\| = \|v\| \quad \|Hu\| = \|u\|$$

$$|\theta_1| = |\theta_2| \quad \theta_1 = -\theta_2$$

- Rotation
- Reflection





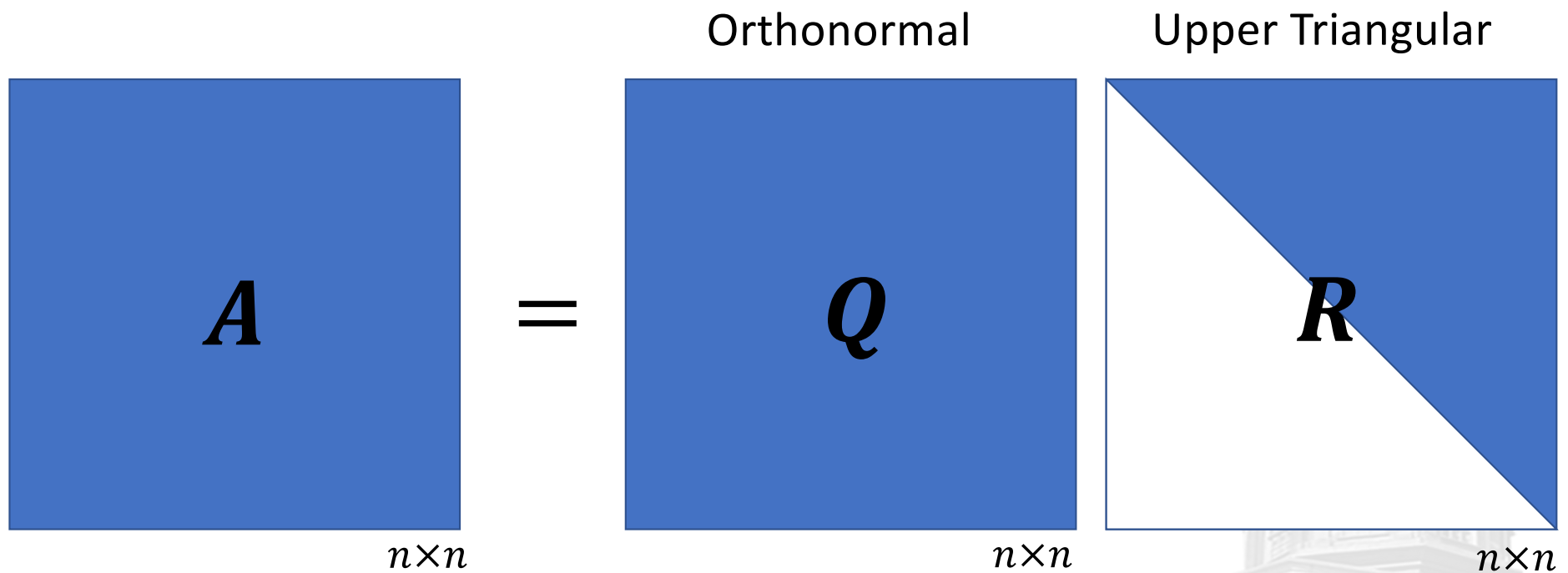
Householder Reflections

$$\underbrace{H_m \cdots H_4 H_3 H_2 H_1}_{\text{Orthonormal Matrices } Q^T} A = R$$

Carefully choose reflection matrices



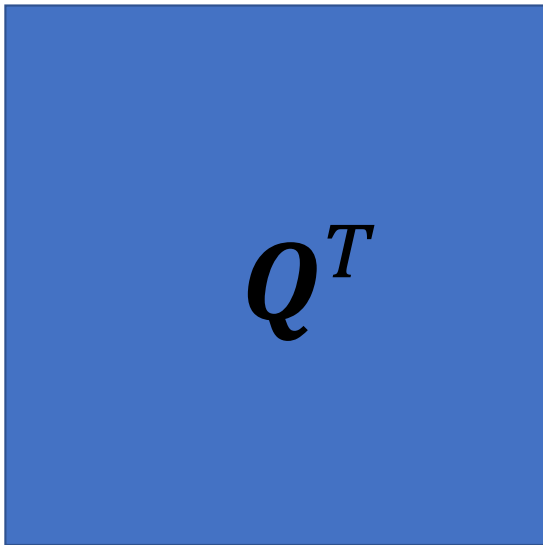
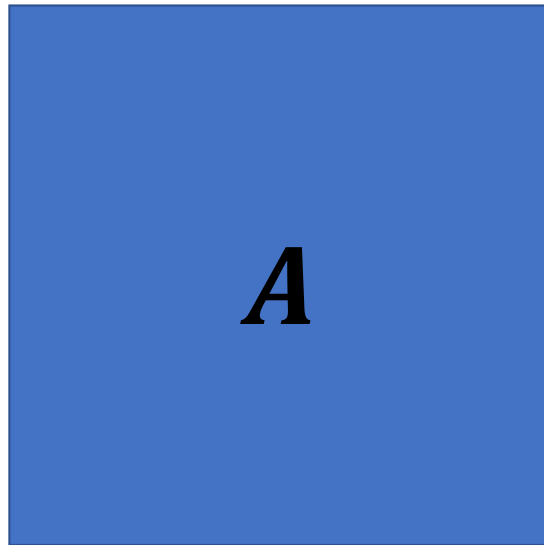
QR Decomposition



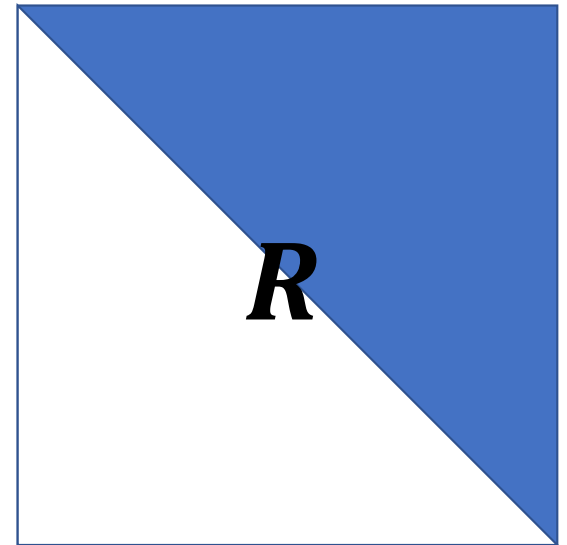
Householder Reflections



Orthonormal

 Q^T  A $=$

Upper Triangular

 R

$$H_m \cdots H_4 H_3 H_2 H_1$$



Elimination Matrices

$$\underbrace{E_m \cdots E_4 E_3 E_2 E_1}_{\text{Elimination Matrices}} \mathbf{A} = \mathbf{U}$$

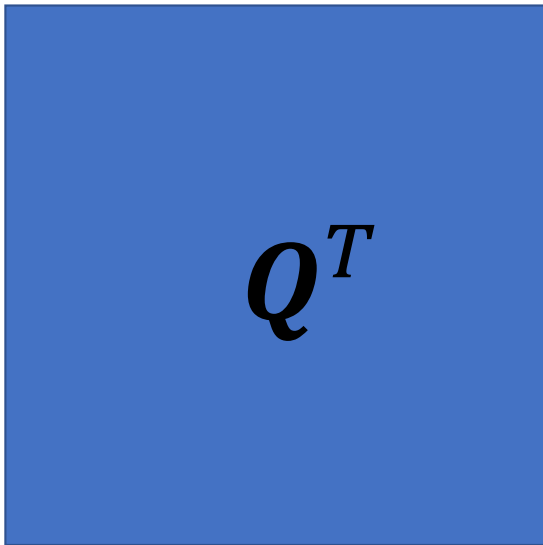
\mathbf{L}^{-1}

$$\underbrace{\begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & -2 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ -3 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 \\ -2 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}}_{\text{Elimination Matrices}} \underbrace{\begin{bmatrix} 2 & 1 & -2 \\ 4 & 5 & -3 \\ 6 & 9 & -2 \end{bmatrix}}_{\mathbf{A}} = \underbrace{\begin{bmatrix} 2 & 1 & -2 \\ 0 & 3 & +1 \\ 0 & 0 & +2 \end{bmatrix}}_{\mathbf{U}}$$

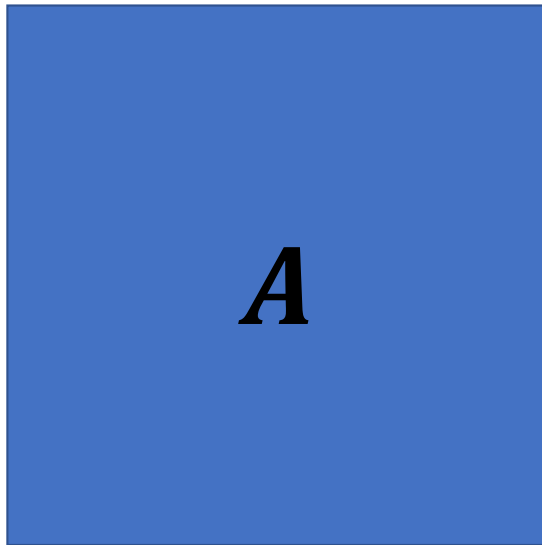
QR Decomposition



Orthonormal



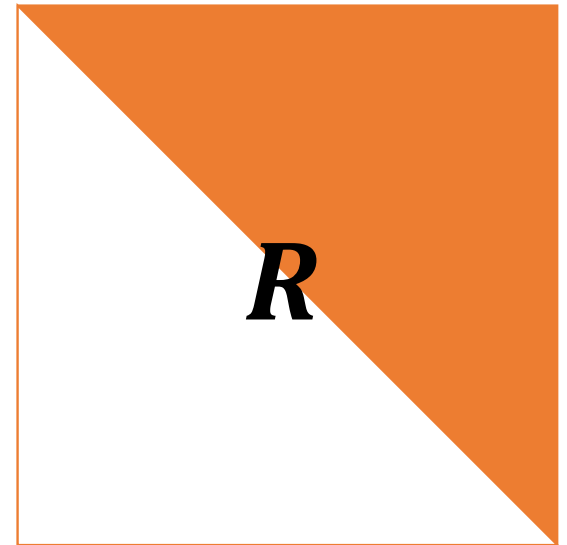
Q^T



A

=

Upper Triangular



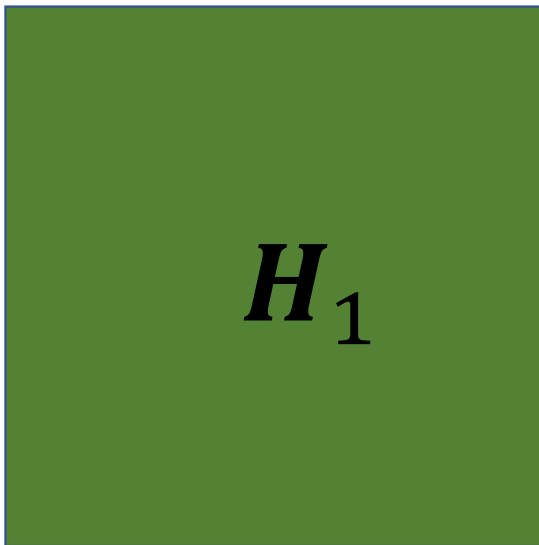
R

$H_m \cdots H_4 H_3 H_2 H_1$

QR Decomposition



Orthonormal

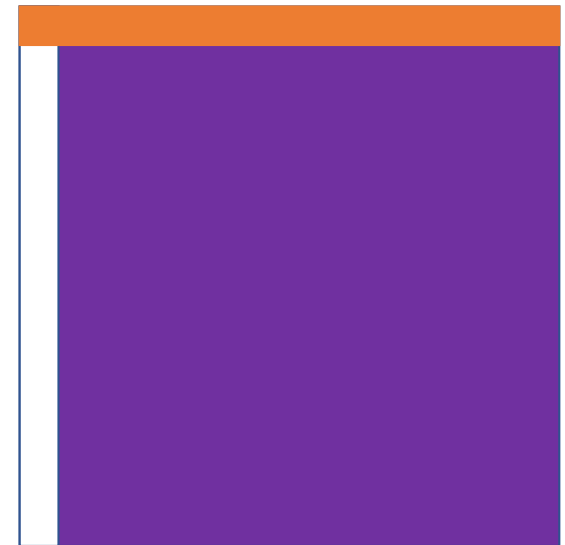


H_1



A

=



Householder Reflections

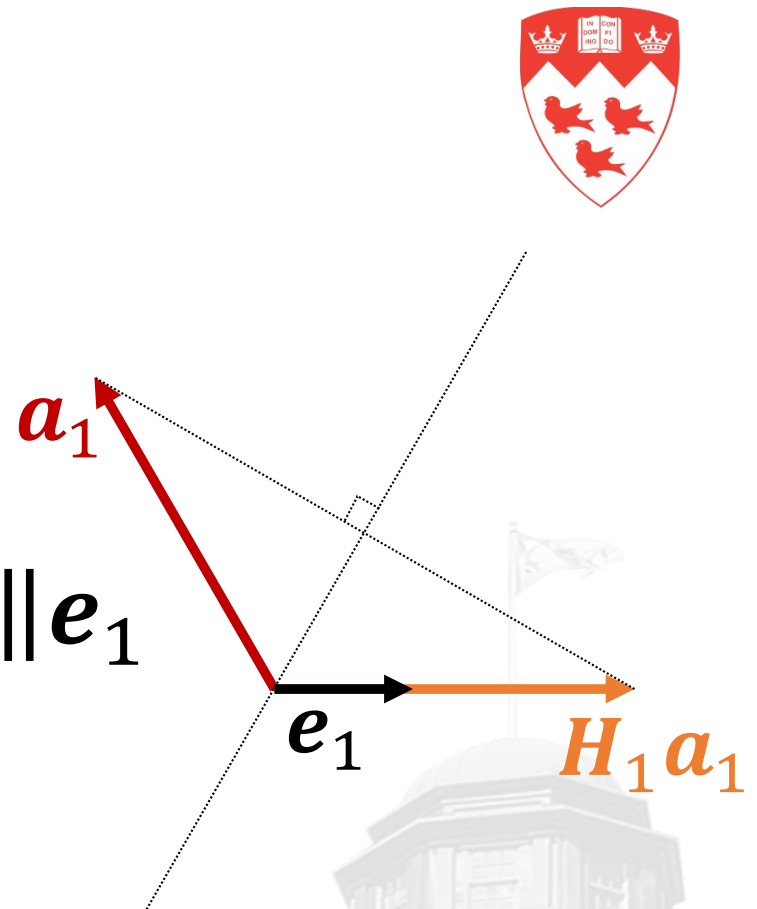


$$\left[\begin{array}{c} H_1 \end{array} \right] \left[\begin{array}{c} \left[\begin{array}{c} a_1 \end{array} \right] \quad \left[\begin{array}{c} a_2 \end{array} \right] \quad \cdots \quad \left[\begin{array}{c} a_m \end{array} \right] \end{array} \right] = \left[\begin{array}{c} \left[\begin{array}{c} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{array} \right] \quad \left[\begin{array}{c} r_{12} \\ \tilde{a}_2 \end{array} \right] \quad \cdots \quad \left[\begin{array}{c} r_{1m} \\ \tilde{a}_m \end{array} \right] \end{array} \right]$$

Householder Reflections

$$H_1 \mathbf{a}_1 = \begin{bmatrix} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} = r_{11} \mathbf{e}_1 = \pm \|\mathbf{a}_1\| \mathbf{e}_1$$

We know what $H_1 \mathbf{a}_1$ is (or must be) even before we learn how to find the reflection matrix H_1



Householder Reflections

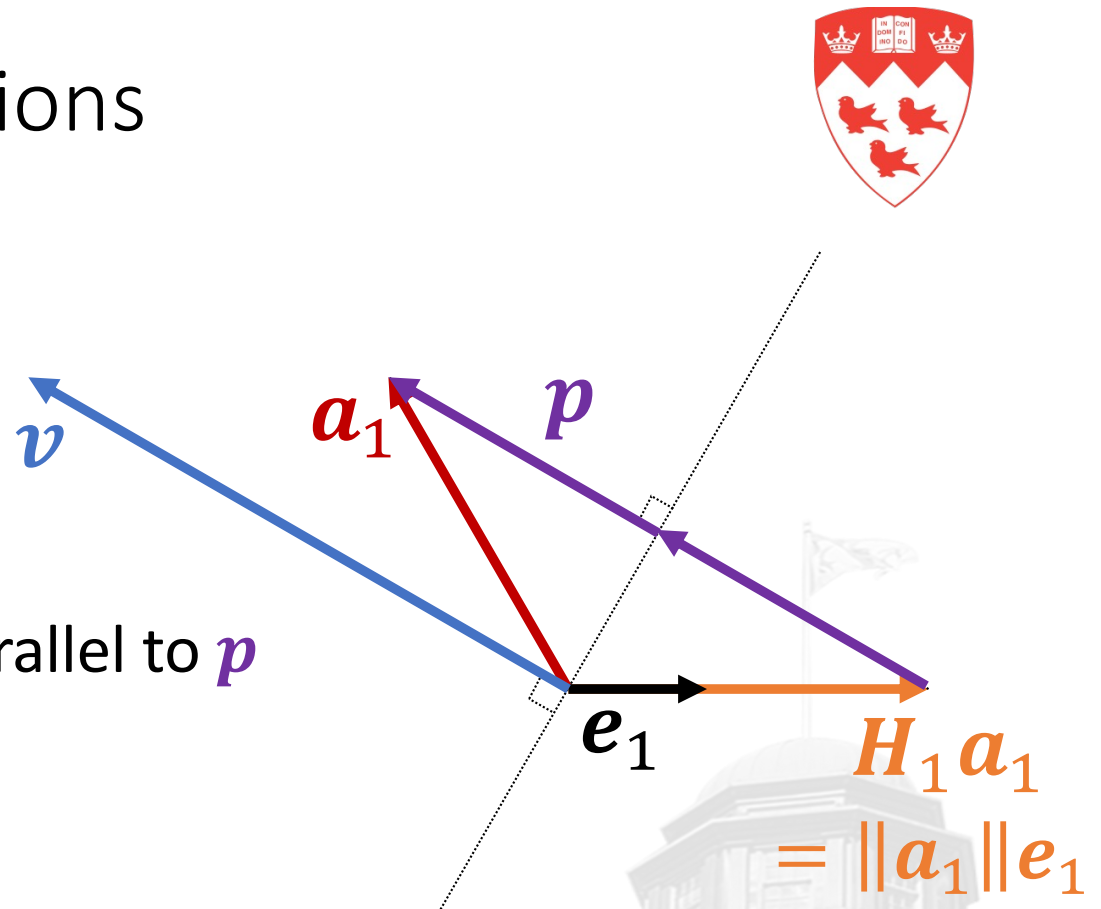
$$H_1 \mathbf{a}_1 = \pm \|\mathbf{a}_1\| \mathbf{e}_1$$

$$H_1 \mathbf{a}_1 = \|\mathbf{a}_1\| \mathbf{e}_1$$

$$H_1 \mathbf{a}_1 = \mathbf{a}_1 - 2\mathbf{p}$$

Assume we have a vector \mathbf{v} parallel to \mathbf{p}

The length of \mathbf{v} is arbitrary





Householder Reflections

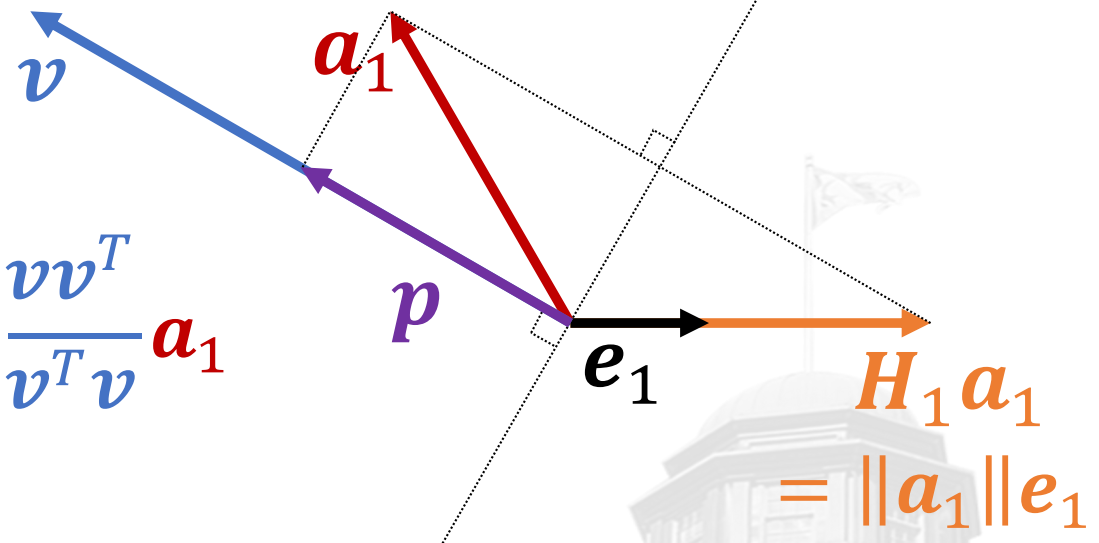
p is the projection of a_1 on v .

$$p = \underbrace{\frac{vv^T}{v^T v}} a_1$$

Projection matrix onto v

$$H_1 a_1 = a_1 - 2p = a_1 - 2 \frac{vv^T}{v^T v} a_1$$

$$H_1 a_1 = \left(U - 2 \frac{vv^T}{v^T v} \right) a_1$$





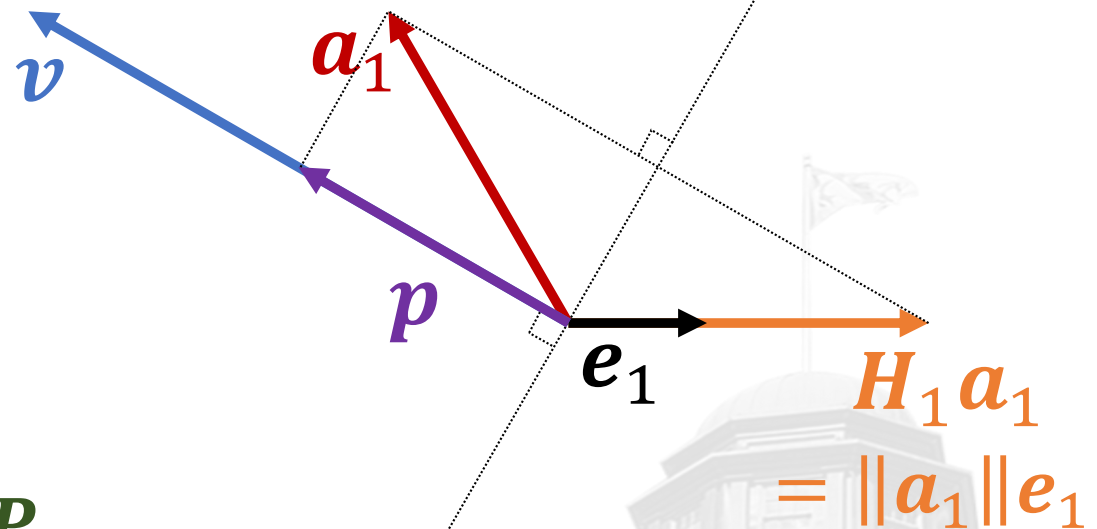
Householder Reflections

$$\mathbf{p} = \underbrace{\frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}}}_{\text{Projection matrix onto } \mathbf{v}} \mathbf{a}_1 = \mathbf{P}\mathbf{a}_1$$

Projection matrix onto \mathbf{v}

$$\mathbf{H}_1 \mathbf{a}_1 = \left(\mathbf{U} - 2 \frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}} \right) \mathbf{a}_1$$

$$\mathbf{H}_1 = \mathbf{U} - 2 \frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}} = \mathbf{U} - 2\mathbf{P}$$



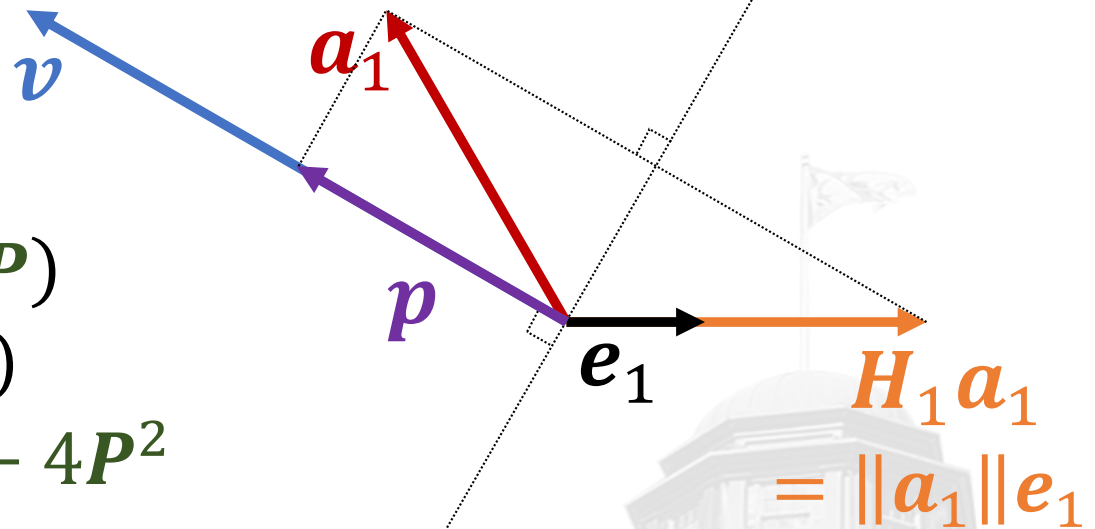


Is \mathbf{H}_1 Orthonormal?

$$\mathbf{H}_1 = \mathbf{U} - 2 \frac{\mathbf{v}\mathbf{v}^T}{\mathbf{v}^T\mathbf{v}} = \mathbf{U} - 2\mathbf{P}$$

$$\mathbf{P} = \mathbf{P}^T \quad \mathbf{P}^2 = \mathbf{P}$$

$$\begin{aligned} \mathbf{H}_1^T \mathbf{H}_1 &= (\mathbf{U} - 2\mathbf{P})^T (\mathbf{U} - 2\mathbf{P}) \\ &= (\mathbf{U} - 2\mathbf{P})(\mathbf{U} - 2\mathbf{P}) \\ &= \mathbf{U} - 2\mathbf{U}\mathbf{P} - 2\mathbf{P}\mathbf{U} + 4\mathbf{P}^2 \\ &= \mathbf{U} - 2\mathbf{P} - 2\mathbf{P} + 4\mathbf{P} = \mathbf{U} \end{aligned}$$





Choice of v

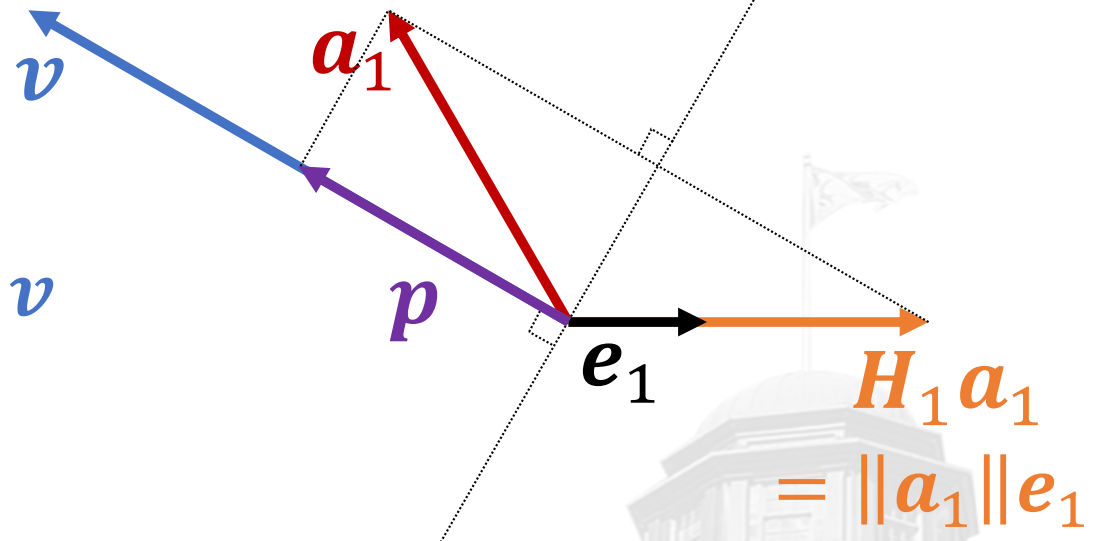
$$\mathbf{H}_1 \mathbf{a}_1 = \mathbf{a}_1 - 2 \frac{\mathbf{v} \mathbf{v}^T}{\mathbf{v}^T \mathbf{v}} \mathbf{a}_1 = \pm \|\mathbf{a}_1\| \mathbf{e}_1$$

$$\mathbf{a}_1 \pm \|\mathbf{a}_1\| \mathbf{e}_1 = 2 \frac{\mathbf{v} \mathbf{v}^T}{\mathbf{v}^T \mathbf{v}} \mathbf{a}_1$$

$(\mathbf{a}_1 \pm \|\mathbf{a}_1\| \mathbf{e}_1)$ is parallel to \mathbf{v}

Note: Length of v is arbitrary

$$\mathbf{v} \equiv \mathbf{a}_1 \pm \|\mathbf{a}_1\| \mathbf{e}_1$$



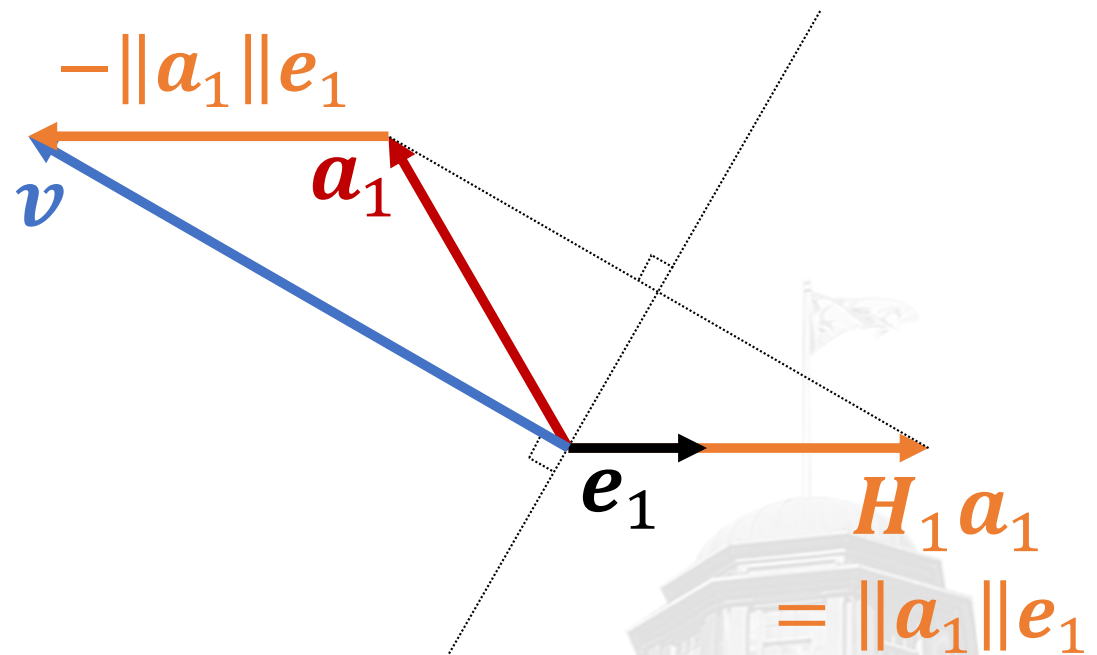


Choice of \mathbf{v} Option #1

$$\mathbf{H}_1 \mathbf{a}_1 = \pm \|\mathbf{a}_1\| \mathbf{e}_1$$

$$\mathbf{H}_1 \mathbf{a}_1 = \|\mathbf{a}_1\| \mathbf{e}_1$$

$$\mathbf{v} \equiv \mathbf{a}_1 - \|\mathbf{a}_1\| \mathbf{e}_1$$



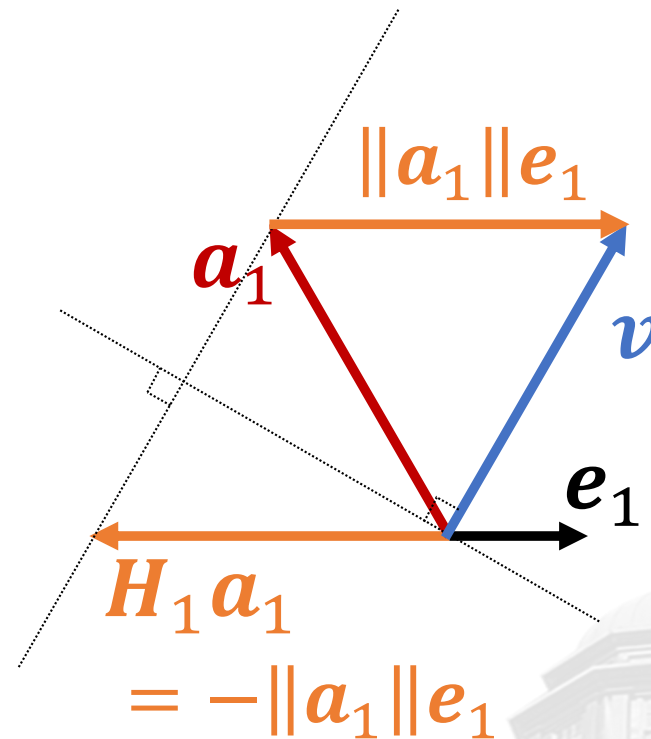


Choice of \mathbf{v} Option #2

$$\mathbf{H}_1 \mathbf{a}_1 = \pm \|\mathbf{a}_1\| \mathbf{e}_1$$

$$\mathbf{H}_1 \mathbf{a}_1 = -\|\mathbf{a}_1\| \mathbf{e}_1$$

$$\mathbf{v} \equiv \mathbf{a}_1 + \|\mathbf{a}_1\| \mathbf{e}_1$$





Householder Reflections

$$\left[\begin{array}{c} H_1 \end{array} \right] \left[\begin{array}{c} a_1 \\ a_2 \\ \vdots \\ a_m \end{array} \right] = \left[\begin{array}{c} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{array} \right] \left[\begin{array}{c} r_{12} \\ \tilde{a}_2 \\ \vdots \\ \tilde{a}_m \end{array} \right] \dots \left[\begin{array}{c} r_{1m} \\ \tilde{a}_m \end{array} \right]$$

$$H_1 = U - 2 \frac{vv^T}{v^T v}$$

$$v \equiv a_1 \pm \|a_1\| e_1$$

Chose sign that makes $v^T v$ larger and therefore avoid dividing by a small value

Next Steps

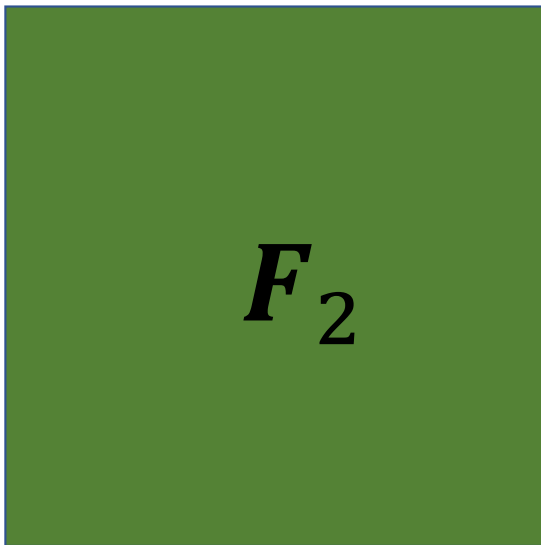


$$\left[\begin{array}{c} H_1 \end{array} \right] \left[\begin{array}{c} a_1 \\ a_2 \\ \vdots \\ a_m \end{array} \right] = \left[\begin{array}{c} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{array} \right] \left[\begin{array}{c} r_{12} \\ \tilde{a}_2 \\ \vdots \\ r_{1m} \\ \tilde{a}_m \end{array} \right]$$

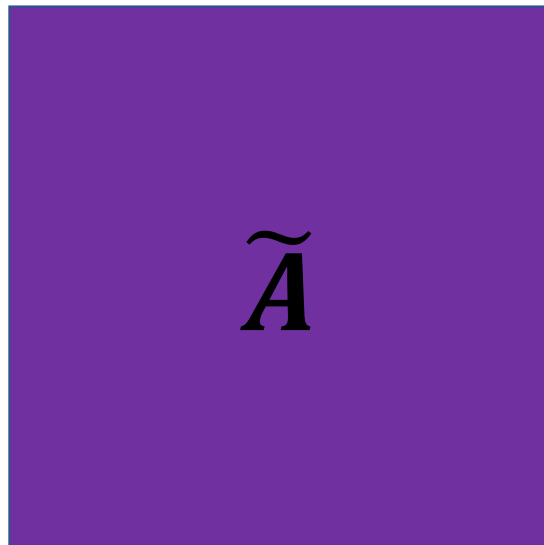
\tilde{A}

Next Steps

Orthonormal

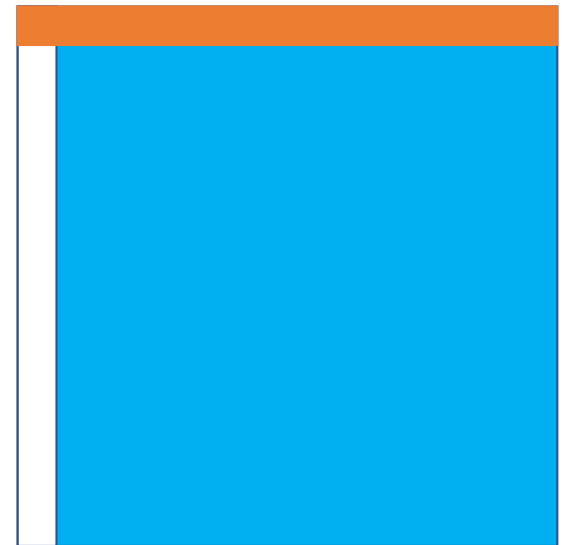


F_2



\tilde{A}

=



Next Steps

\tilde{A}



$$\begin{bmatrix} \mathbf{F}_2 \end{bmatrix} \begin{bmatrix} \tilde{\mathbf{a}}_2 \\ \tilde{\mathbf{a}}_3 \\ \dots \\ \tilde{\mathbf{a}}_m \end{bmatrix} = \begin{bmatrix} r_{22} \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \begin{bmatrix} r_{23} \\ \hat{\mathbf{a}}_3 \end{bmatrix} \dots \begin{bmatrix} r_{2m} \\ \hat{\mathbf{a}}_m \end{bmatrix}$$

We can find \mathbf{F}_2 using the same process we used to find \mathbf{H}_1



Second Householder Reflection

$$\mathbf{H}_2 = \begin{bmatrix} 1 & 0 & 0 & \cdots & 0 \\ 0 & & & & \\ 0 & & \mathbf{F}_2 & & \\ \vdots & & & & \\ 0 & & & & \end{bmatrix}$$



Next Steps



$$\left[\begin{array}{c} H_1 \end{array} \right] \left[\begin{array}{c} a_1 \end{array} \right] \left[\begin{array}{c} a_2 \end{array} \right] \cdots \left[\begin{array}{c} a_m \end{array} \right] = \left[\begin{array}{c} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{array} \right] \left[\begin{array}{c} r_{12} \\ \tilde{a}_2 \end{array} \right] \cdots \left[\begin{array}{c} r_{1m} \\ \tilde{a}_m \end{array} \right]$$

$$\left[\begin{array}{c} H_2 H_1 \end{array} \right] \left[\begin{array}{c} a_1 \end{array} \right] \left[\begin{array}{c} a_2 \end{array} \right] \cdots \left[\begin{array}{c} a_m \end{array} \right] = ?$$



Next Steps

$$\begin{bmatrix} & & & & \\ & \mathbf{H}_2 & & & \\ & & & & \end{bmatrix} \begin{bmatrix} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \begin{bmatrix} r_{12} \\ \tilde{a}_2 \end{bmatrix} \dots \begin{bmatrix} r_{1m} \\ \tilde{a}_m \end{bmatrix} = ?$$

$$\begin{bmatrix} 1 & 0 & 0 & \dots & 0 \\ 0 & & & & \\ 0 & & & & \\ \vdots & & & & \\ 0 & & & & \end{bmatrix} \begin{bmatrix} & & & & \\ & \mathbf{F}_2 & & & \\ & & & & \end{bmatrix} \begin{bmatrix} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} \begin{bmatrix} r_{12} \\ \tilde{a}_2 \end{bmatrix} \dots \begin{bmatrix} r_{1m} \\ \tilde{a}_m \end{bmatrix} = ?$$



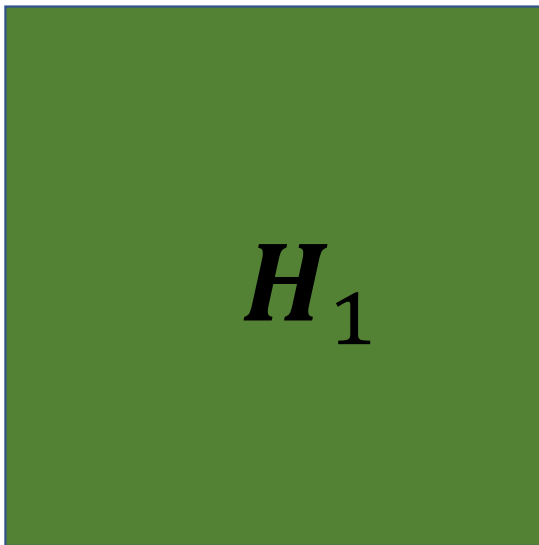
Next Steps

$$\left[\begin{matrix} H_2 H_1 \end{matrix} \right] \left[\begin{matrix} a_1 \\ a_2 \\ \vdots \\ a_m \end{matrix} \right] = \left[\begin{matrix} \begin{bmatrix} r_{11} \\ 0 \\ 0 \\ \vdots \\ 0 \end{bmatrix} & \begin{bmatrix} r_{11} \\ r_{22} \\ 0 \\ \vdots \\ 0 \end{bmatrix} & \begin{bmatrix} r_{11} \\ r_{22} \\ \hat{a}_3 \end{bmatrix} & \dots & \begin{bmatrix} r_{1m} \\ r_{2m} \\ \hat{a}_m \end{bmatrix} \end{matrix} \right]$$

QR Decomposition



Orthonormal



H_1



A

=





Third Householder Reflection

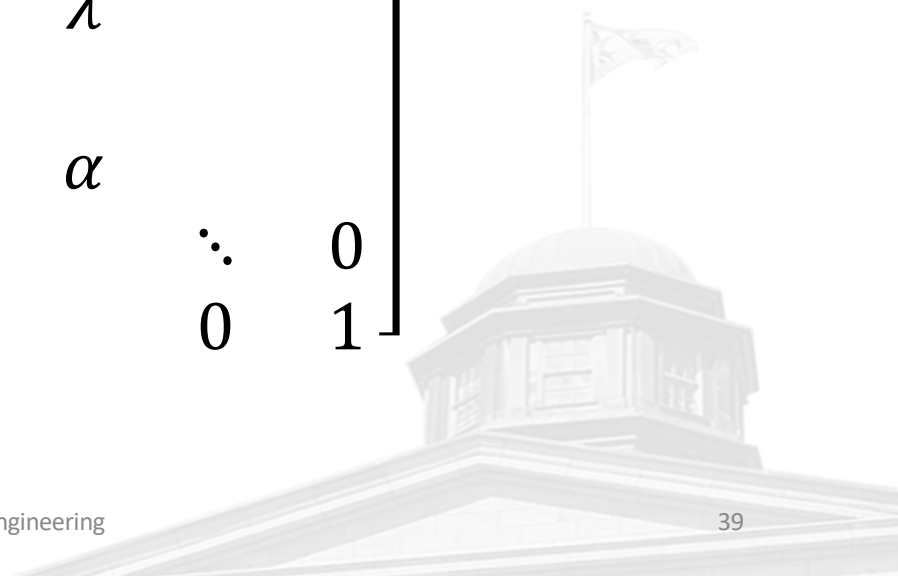
$$\mathbf{H}_3 = \begin{bmatrix} 1 & 0 & 0 & \cdots & 0 \\ 0 & 1 & 0 & \cdots & 0 \\ 0 & 0 & \mathbf{F}_3 & & \\ \vdots & \vdots & & & \\ 0 & 0 & & & \end{bmatrix}$$





QR Using Givens Rotations

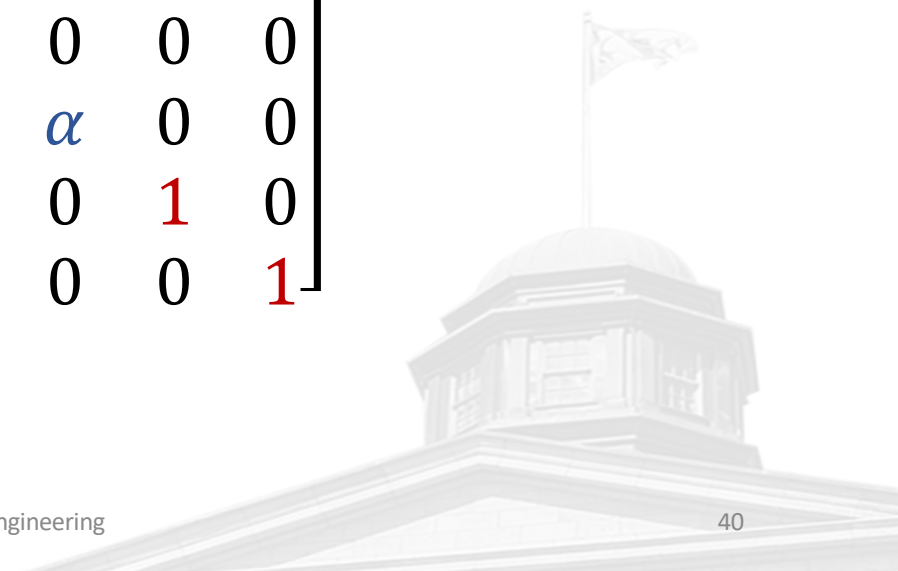
$$H_i = \begin{bmatrix} 1 & 0 & & & \dots & & & 0 \\ 0 & 1 & \dots & & \dots & & & 0 \\ & & \ddots & & & & & \\ & & & \alpha & & \lambda & & \\ \vdots & \vdots & & & \ddots & & & \\ & & & -\lambda & & \alpha & & \\ & & & & & & \ddots & 0 \\ 0 & 0 & & & & & 0 & 1 \end{bmatrix}$$



Example



$$H_i = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 & 0 & 0 & 0 \\ 0 & 0 & \alpha & 0 & 0 & \lambda & 0 & 0 \\ 0 & 0 & 0 & 1 & 0 & 0 & 0 & 0 \\ 0 & 0 & 0 & 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & -\lambda & 0 & 0 & \alpha & 0 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 1 & 0 \\ 0 & 0 & 0 & 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$





Example

$$\mathbf{H} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & \alpha & 0 & \lambda & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & -\lambda & 0 & \alpha & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix}$$

Orthonormal iff $\alpha^2 + \lambda^2 = 1$

$$\mathbf{H}^T \mathbf{H} = \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & \alpha & 0 & -\lambda & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & \lambda & 0 & \alpha & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & \alpha & 0 & \lambda & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & -\lambda & 0 & \alpha & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} = \mathbf{U}$$



QR using Givens Rotations

$$\mathbf{A} = \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \\ a_{41} & a_{42} & a_{43} & a_{44} & a_{45} \\ a_{51} & a_{52} & a_{53} & a_{54} & a_{55} \end{bmatrix}$$





QR Using Givens Rotations

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & 1 & 0 & 0 \\ 0 & 0 & 0 & \alpha & \lambda \\ 0 & 0 & 0 & -\lambda & \alpha \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \\ a_{41} & a_{42} & a_{43} & a_{44} & a_{45} \\ \textcolor{red}{a}_{51} & a_{52} & a_{53} & a_{54} & a_{55} \end{bmatrix} =$$

$$\begin{cases} \alpha^2 + \lambda^2 = 1 \\ -\lambda a_{41} + \alpha a_{51} = 0 \end{cases}$$

$$\alpha = \frac{a_{41}}{\sqrt{a_{41}^2 + a_{51}^2}} \quad \lambda = \frac{a_{51}}{\sqrt{a_{41}^2 + a_{51}^2}}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \\ \tilde{a}_{41} & \tilde{a}_{42} & \tilde{a}_{43} & \tilde{a}_{44} & \tilde{a}_{45} \\ \textcolor{red}{0} & \tilde{a}_{52} & \tilde{a}_{53} & \tilde{a}_{54} & \tilde{a}_{55} \end{bmatrix}$$



Example

$$\begin{bmatrix} 1 & 0 & 0 & 0 & 0 \\ 0 & 1 & 0 & 0 & 0 \\ 0 & 0 & \alpha & \lambda & 0 \\ 0 & 0 & -\lambda & \alpha & 0 \\ 0 & 0 & 0 & 0 & 1 \end{bmatrix} \begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ a_{31} & a_{32} & a_{33} & a_{34} & a_{35} \\ \tilde{a}_{41} & \tilde{a}_{42} & \tilde{a}_{43} & \tilde{a}_{44} & \tilde{a}_{45} \\ 0 & \tilde{a}_{52} & \tilde{a}_{53} & \tilde{a}_{54} & \tilde{a}_{55} \end{bmatrix} =$$

$$\begin{cases} \alpha^2 + \lambda^2 = 1 \\ -\lambda a_{31} + \alpha \tilde{a}_{41} = 0 \end{cases}$$

$$\alpha = \frac{a_{31}}{\sqrt{a_{41}^2 + a_{51}^2}} \quad \lambda = \frac{\tilde{a}_{41}}{\sqrt{a_{41}^2 + a_{51}^2}}$$

$$\begin{bmatrix} a_{11} & a_{12} & a_{13} & a_{14} & a_{15} \\ a_{21} & a_{22} & a_{23} & a_{24} & a_{25} \\ \hat{a}_{31} & \hat{a}_{32} & \hat{a}_{33} & \hat{a}_{34} & \hat{a}_{35} \\ 0 & \hat{a}_{42} & \hat{a}_{43} & \hat{a}_{44} & \hat{a}_{45} \\ 0 & \tilde{a}_{52} & \tilde{a}_{53} & \tilde{a}_{54} & \tilde{a}_{55} \end{bmatrix}$$