

# Time-Varying Systems and Computations

On Reflections and Generalized Rotations
Unit 7.5, WS 2024
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## **Exemplary Case**



QR decomposition as a standard computational tool

$$A = Q \cdot R$$
  $A \in \mathcal{R}^{m \times n}$   $Q^T Q = 1$   $R : upper triangular$ 

Readily available in numerical packages such as Matlab, NumPy, etc. (why not QL, or RQ?)

## **Standard QR Decomposition**



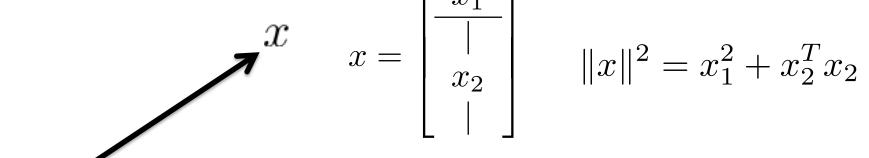
#### Column-wise elimination process

$$Q = H_1 \cdot H_2 \cdot H_3 \cdot H_4$$

Note: Changed Matlab interface

## **Elementary Step**





$$||x||^2 = x_1^2 + x_2^T x_2$$

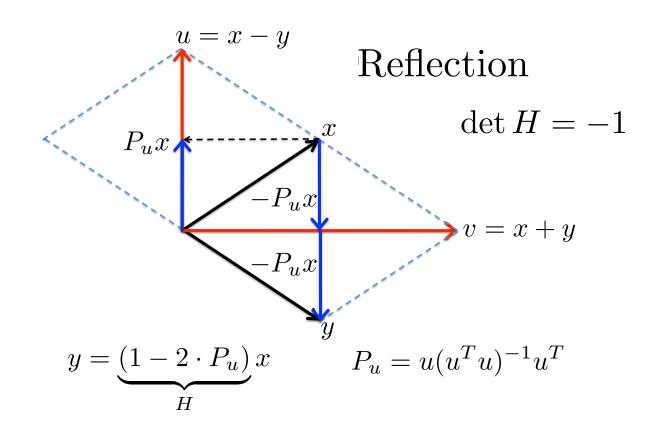
$$y = H \cdot x$$

$$y = \begin{bmatrix} \frac{\|x\|}{0} \\ \vdots \\ 0 \end{bmatrix}$$

$$||x||^2 = ||y||^2 = 1$$

### **Standard Householder Construction**





### **Quest for Rotations**



Computer Vision, Robotics, Machine Intelligence

#### We need Q to be a rotation!

Clifford Algebra

Exponential map of skew-symmetric matrices

2 reflections give a rotation

Quick post-fixes possible (e.g. for SVD)

Use rotations for eliminiation → Givens rotations

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But: Householder is twice as efficient as Givens!

## **QR Decomposition Methods**



What do the textbooks say?

**Gram-Schmidt** 

not recommended (Golub,vLoan)

Householder QR

Reflections (Matlab, NumPy)

Givens QR

Rotations

### Question:

Can we construct a Householder-like rotation?



$$x = \begin{bmatrix} x_1 \\ \\ \\ x_2 \\ \\ \end{bmatrix} \qquad \text{Find } T \qquad \begin{cases} T \in \mathcal{R}^{n \times n} \\ T^T T = 1 \\ \det T = 1 \end{cases}$$

$$y = \begin{bmatrix} \frac{1}{|} \\ 0 \\ \\ \end{bmatrix}$$



Start with a partially specified matrix

$$T = \left[ \begin{array}{cc} x_1 & x_2^T \\ \star & K \end{array} \right]$$

Orthogonality condition

$$T^{T}T = \begin{bmatrix} x_{1}^{2} + \star^{T} \star & x_{1}x_{2}^{T} + \star^{T}K \\ x_{1}x_{2} + K^{T} \star & K^{T}K + x_{2}x_{2}^{T} \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$



From orthogonality condition ...

$$T^TT = \begin{bmatrix} x_1^2 + \star^T \star \\ x_1x_2 + K^T \star \\ x_1x_2 + K^T \star \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ 0 & 1 \end{bmatrix}$$
 ... we can read off 
$$\star = \pm x_2$$
 
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Rank-one Perturbation of Identity with Parameter k

$$K = 1 - kx_2x_2^T$$

Plugged into Orthogonality (22-Block of  $T^TT$ )

$$(1 - kx_2x_2^T) \cdot (1 - kx_2x_2^T) = 1 - x_2x_2^T$$

Quadratic Equation for Parameter *k* 

$$k^{2}(1-x_{1}^{2})-2k+1=0 \implies k_{1}=\frac{1}{1-x_{1}} k_{2}=\frac{1}{1+x_{1}}$$



Resolve sign ambiguity

$$\begin{bmatrix} x_1 & x_2^T \\ \pm x_2 & K \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} x_1^2 + x_2^T x_2 \\ \pm x_2 x_1 + K x_2 \end{bmatrix} = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$k_1 = \frac{1}{1 - x_1} \qquad \Longrightarrow \qquad T_1 = \begin{bmatrix} x_1 & x_2^T \\ \hline x_2 & K \end{bmatrix}$$

$$k_2 = \frac{1}{1+x_1} \quad \Longrightarrow \quad T_2 = \begin{bmatrix} x_1 & x_2^T \\ -x_2 & K \end{bmatrix}$$



Two candidate matrices (check determinant)

$$T_1 = \begin{bmatrix} x_1 & x_2^T \\ x_2 & 1 - \frac{x_2 x_2^T}{1 - x_1} \end{bmatrix} \xrightarrow{\text{Reflection}} \text{Reflection}$$

$$T_2 = \begin{bmatrix} x_1 & x_2^T \\ -x_2 & 1 - \frac{x_2 x_2^T}{1+x_1} \end{bmatrix} \xrightarrow{\text{Rotation}} \det T_2 = 1$$

#### **Householder Rotation**



Rotation is constructed from elementary concepts of Linear Algebra (linear map, orthogonality, ...)

It is very similar to reflections and leads to the same computational complexity than reflection-based QRD

Rotation-based QRD is not covered in textbooks and/or manuals