

## The Gram-Schmidt Process

Goal:

Given  $x_1, x_2, \dots, x_p$  linearly independent vectors in  $\mathbb{R}^m$   
 want to produce  $v_1, v_2, \dots, v_p$  orthogonal (orthonormal) vectors in  $\mathbb{R}^m$

such that

$$\text{Span}\{v_i\} = \text{Span}\{x_i\}$$

$$\text{Span}\{v_1, v_2\} = \text{Span}\{x_1, x_2\}$$

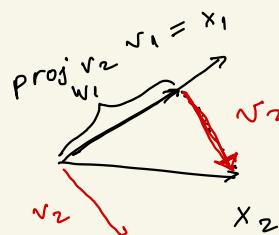
⋮

$$\text{Span}\{v_1, v_2, \dots, v_p\} = \text{Span}\{x_1, x_2, \dots, x_p\}$$

There is an algorithm for that called Gram-Schmidt



$$v_1 = x_1$$



$$w_1 = \text{Span}\{v_1\}$$

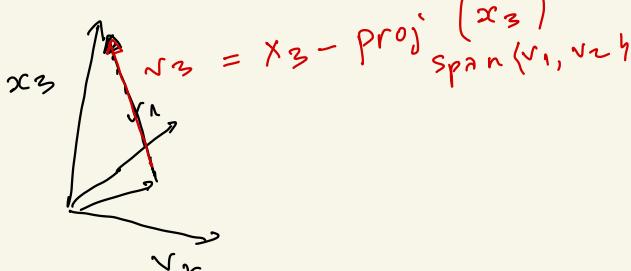
$$v_2 = x_2 - \underbrace{\frac{x_2 \cdot v_1}{v_1 \cdot v_1} v_1}_{\text{proj}_{\text{Span}\{v_1\}} x_2}$$

$$v_1 = x_1$$

$$v_2 =$$

$v_2$  must be  $\perp v_1$

$$\text{Span}\{v_1, v_2\} = \text{Span}\{x_1, x_2\}$$



## Algorithm Gram-Schmidt:

Given  $x_1, \dots, x_p$  linearly independent in  $\mathbb{R}^m$   
 We construct  $v_1, \dots, v_p$  orthogonal (orthonormal)

$$v_1 = x_1$$

$$v_2 = x_2 - \frac{x_2 \cdot v_1}{v_1 \cdot v_1} v_1 = x_2 - P_{W_1}(x_2)$$

$$v_3 = x_3 - \left( \frac{x_3 \cdot v_1}{v_1 \cdot v_1} v_1 + \frac{x_3 \cdot v_2}{v_2 \cdot v_2} v_2 \right) = x_3 - P_{W_2}(x_3)$$

⋮

$$v_p = x_p - \left( \frac{x_p \cdot v_1}{v_1 \cdot v_1} v_1 + \dots + \frac{x_p \cdot v_{p-1}}{v_{p-1} \cdot v_{p-1}} v_{p-1} \right) = x_p - P_{W_{p-1}}(x_p)$$

Notation:  $W_1 = \text{Span}\{v_1\}$   
 $W_2 = \text{Span}\{v_1, v_2\}$

$$\vdots$$

$$W_p = \text{Span}\{v_1, \dots, v_p\}$$

To obtain orthonormal vectors

"scale"

$$\frac{v_1}{\|v_1\|}, \dots, \frac{v_p}{\|v_p\|}$$

Ex 1

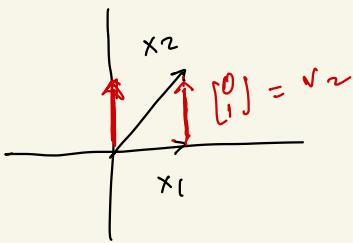
$$x_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \quad x_2 = \begin{bmatrix} 1 \\ 1 \end{bmatrix}$$

$$v_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix} \\ v_2 = \begin{bmatrix} 0 \\ 1 \end{bmatrix}$$

$$v_1 = x_1 = \begin{bmatrix} 1 \\ 0 \end{bmatrix}$$

$$v_2 = x_2 - \frac{x_2 \cdot v_1}{v_1 \cdot v_1} v_1$$

$$= \begin{bmatrix} 1 \\ 1 \end{bmatrix} - \frac{1}{1} \begin{bmatrix} 1 \\ 0 \end{bmatrix} = \begin{bmatrix} 0 \\ 1 \end{bmatrix} = v_2$$



(2)

(3)

Ex 2 Apply G-S to

$$\begin{bmatrix} 1 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix}, \begin{bmatrix} 1 \\ 1 \\ 2 \\ 0 \end{bmatrix}$$

$x_1$        $x_2$        $x_3$

$$v_1 = \begin{bmatrix} 1 \\ -1 \\ 1 \\ 0 \end{bmatrix}$$

$$v_2 = x_2 - \frac{x_2 \cdot v_1}{v_1 \cdot v_1} v_1 = \begin{bmatrix} 1 \\ 0 \\ 1 \\ 0 \end{bmatrix} - \frac{2}{3} \begin{bmatrix} 1 \\ -1 \\ 1 \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{1}{3} \\ \frac{2}{3} \\ \frac{1}{3} \\ 0 \end{bmatrix}$$

$$v_3 = x_3 - \frac{x_3 \cdot v_1}{v_1 \cdot v_1} v_1 - \frac{x_3 \cdot v_2}{v_2 \cdot v_2} v_2$$

$$= \begin{bmatrix} 1 \\ 1 \\ 2 \\ 0 \end{bmatrix} - \frac{2}{3} \begin{bmatrix} 1 \\ -1 \\ 1 \\ 0 \end{bmatrix} - \frac{\frac{5}{2}}{2} \begin{bmatrix} \frac{1}{3} \\ \frac{2}{3} \\ \frac{1}{3} \\ 0 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} \\ 0 \\ \frac{1}{2} \\ 0 \end{bmatrix}$$

—

$$v_1 = \begin{bmatrix} 1 \\ -1 \\ 1 \\ 0 \end{bmatrix}, \quad v_2 = \begin{bmatrix} \frac{1}{3} \\ \frac{2}{3} \\ \frac{1}{3} \\ 0 \end{bmatrix}, \quad v_3 = \begin{bmatrix} \frac{1}{2} \\ 0 \\ \frac{1}{2} \\ 0 \end{bmatrix} \quad \leftarrow \text{orthogonal}$$

$$u_1 = \begin{bmatrix} \frac{1}{\sqrt{3}} \\ -\frac{1}{\sqrt{3}} \\ \frac{1}{\sqrt{3}} \\ 0 \end{bmatrix}, \quad u_2 = \begin{bmatrix} \frac{1}{\sqrt{6}} \\ \frac{2}{\sqrt{3}} \\ \frac{1}{\sqrt{6}} \\ 0 \end{bmatrix}, \quad u_3 = \begin{bmatrix} -\frac{1}{\sqrt{2}} \\ 0 \\ \frac{1}{\sqrt{2}} \\ 0 \end{bmatrix} \quad \leftarrow \text{orthonormal}$$

The QR Factorization of a matrix  $A$   $m \times n$  (4)  
 with linearly independent columns

$$A = Q R \quad (\Rightarrow Q^{-1} A = R \Rightarrow Q^T A = R)$$

$Q$  is orthogonal and  $R$  is invertible upper triangular  
 $m \times m$  with positive diagonal entries

its columns form an orthonormal set

One can use Gram-Schmidt for obtaining QR-factorization

More precisely apply G-S to the columns of  $A$  to obtain  $n$  orthonormal vectors  $u_1, \dots, u_m$

$$\text{form } Q = [u_1, \dots, u_n]$$

$m \times n$  matrix with columns  $u_1, \dots, u_n$

since  $Q$  orthonormal

$$Q^T Q = I_n$$

$$R = Q^T A$$

$\left( \begin{array}{l} \text{if } r_{kk} < 0 \text{ replace} \\ u_k \text{ by } -u_k \text{ in } Q \\ r_{kk} \text{ by } -r_{kk} \end{array} \right)$

**Ex 3**

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

QR-factorization?

by **Ex 2**

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

$$R = Q^T A = \left[ \begin{array}{cccc} \frac{1}{\sqrt{2}} & -\frac{1}{\sqrt{3}} & \frac{1}{\sqrt{3}} & 0 \\ \frac{1}{\sqrt{6}} & \frac{2}{\sqrt{6}} & \frac{1}{\sqrt{6}} & 0 \\ -\frac{1}{\sqrt{2}} & 0 & \frac{1}{\sqrt{2}} & 0 \end{array} \right] \left[ \begin{array}{cccc} 1 & 1 & 1 & 1 \\ -1 & 0 & 1 & 2 \\ 1 & 1 & 0 & 0 \\ 0 & 0 & 0 & 0 \end{array} \right] \quad (5)$$

$$= \left[ \begin{array}{cccc} \frac{1}{\sqrt{3}} & \frac{2}{\sqrt{3}} & \frac{4}{\sqrt{3}} & 0 \\ 0 & \frac{\sqrt{6}}{3} & \frac{5}{\sqrt{6}} & 0 \\ 0 & 0 & \frac{1}{\sqrt{2}} & 0 \end{array} \right]$$

$R$

HW # 23

Hint:

$$A = QR$$

$A$   $m \times n$   
 $Q$  orthonormal  $m \times n$   
 $R$   $n \times n$

$A$  lin. indep. columns.

Argue / prove that  $R$  is invertible.

$R$  invertible  $\Leftrightarrow Rx = 0$  has only trivial sol'n.

Say  $Ry = 0$  for  $y \neq 0$

$$\Rightarrow \underline{QRy = 0} \quad \underline{Ay = 0}$$

$\Rightarrow y = 0$ .  
 (If lin. ind columns)

---

or  $Rx = 0 \Rightarrow QRx = 0 \Rightarrow Ax = 0$

$$\Rightarrow \underline{x = 0} \text{ since } \leftarrow$$

-END-

