

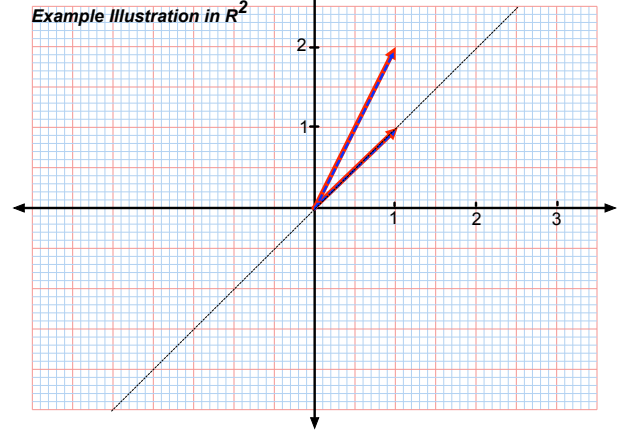
Lecture 14: Gram-Schmidt Orthogonalization and the QR factorization

Outline:

- 1) Review GS for taking A to Q
Examples galore
- 2) Gram-Schmidt orthogonalization leads to a new factorization
 $A=QR$
- 3) Using the QR to solve Linear Least-squares problems
- 4) Matlab
- 5) Computational issues, algorithms (modified Gram-Schmidt), computational costs...
- 6) Quick review of course so far (end of the middle)

Gram-Schmidt Orthogonalization:

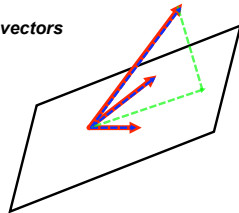
Example Illustration in $\mathbb{R}^2$



Gram-Schmidt Orthogonalization:

Example in $\mathbb{R}^3$: find an orthonormal basis for the $C(A)$ where
 $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$;

- 1) Extension of general algorithm to n vectors



Gram-Schmidt Orthogonalization:

Example in $\mathbb{R}^3$: find an orthonormal basis for the $C(A)$ where
 $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$;

- 2) Specific example

Gram-Schmidt Orthogonalization:

Example in R^3 : find an orthonormal basis for the $C(A)$ where
 $A = \begin{bmatrix} 1 & 1 & 1 \\ 1 & 1 & 0 \\ 1 & 0 & 0 \end{bmatrix}$;

3) Comment Q not unique: suppose $A = \begin{bmatrix} 1 & 1 & 1 \\ 0 & 1 & 1 \\ 0 & 0 & 1 \end{bmatrix}$

Gram-Schmidt Orthogonalization:

Leads to a new Factorization $A=QR$

Analogy with Gaussian Elimination and $PA=LU$

1) Gaussian Elimination

- A) transforms A to U such that $A=LU$
- B) you get L for free
- C) Use the LU to solve square systems by solving $Lc=\underline{b}$, $Ux=\underline{c}$.
- D) Matlab implementation is just $x=A \backslash b$

2) Gram-Schmidt Orthogonalization

- A) transforms A to Q such that $A=QR$
- B) and you get R for free (but I need to explain R)
- C) Use the QR to solve linear least squares problems using $R\hat{x}=Q^T \underline{b}$
- D) Matlab implementation is just $x=A \backslash b$

Gram-Schmidt Orthogonalization:

Leads to a new Factorization $A=QR$

What is the R in $A=QR$?

- 1) Generally: $R = \underline{\hspace{2cm}}$
- 2) R describes how the columns of A can be written (uniquely) in terms of the columns of Q.

Gram-Schmidt Orthogonalization:

Leads to a new Factorization $A=QR$

What is the R in $A=QR$?

- 3) Because of GS algorithm R is **Upper Triangular!** (Nice!)

Gram-Schmidt Orthogonalization:

Leads to a new Factorization $A=QR$

What is the R in $A=QR$?

- 4) Upper Triangular R , makes over-determined least-squares problem $A\hat{x}=\underline{b}$, reduce to the Triangular problem $R\hat{x}=Q^T\underline{b}$. This can be solved quickly by back substitution.

Gram-Schmidt Orthogonalization:

Leads to a new Factorization $A=QR$

What is the R in $A=QR$?

- 5) Comment: only least squares solution $\hat{x}$, needs R

Projection $p=$ _____

Projection Matrix $P=$ _____ (error $e=$ _____)

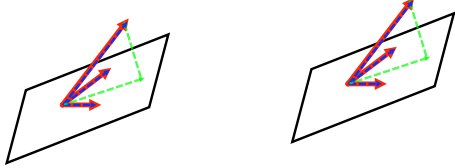
Alternative proof: let $P=A(A^TA)^{-1}A^T$ and $A=QR$

QR Factorization and Applications

Example: Find best fit straight line through $(0,0)$, $(1,2)$, $(2,1)$ by using the Normal Equations and QR...

QR Factorization: Computational issues

Computing the QR: a simple algorithm **Modified Gram-Schmidt**



QR Factorization: Computational issues

Computing the QR: a simple algorithm **Modified Gram-Schmidt**

Computational Costs: Normal Equations vs QR

QR Factorization: Computational issues

Computing the QR: Other Algorithms: A to R by Householder Reflections

Where We are...a mini-course review

Part I)

A) Principal Equation: _____

B) Algorithm: _____

C) Factorization: _____

D) Applications: _____

E) Theory: _____

Part II)

A) Principal Equation: _____

B) Algorithm: _____

C) Factorization: _____

D) Applications: _____

E) Theory: _____