

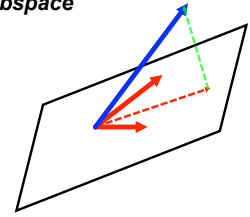
Lecture13

Lecture 13: Orthonormal Bases and Q matrices (towards the QR Factorization)

Outline:

- 1) Motivation: $A^T A x = A^T b$ as a projection problem using columns of A as a basis for $C(A)$
- 2) But some bases are better than others: *Orthonormal Bases*
 - A) Definition and Q matrices ($Q^T Q = I$)
 - B) Q matrices and Least-Squares problems
 - C) Examples of Q matrices
 - D) General Properties of Q Matrices
 - E) Proof by basketball redux
- 3) Turning A to Q: Gram-Schmidt Orthogonalization

Orthogonal Projection onto a Subspace



Orthonormal bases

Definition: a set of vectors $q_1, q_2, \dots, q_n$ in $\mathbb{R}^m$ ($m \geq n$) are said to be *orthonormal* if

- 1) they are all unit vectors $\|q_i\|=1$
- 2) they are all mutually orthogonal: $q_i^T q_j = 0$ if $i \neq j$

ultra compact definition $q_i^T q_j = \delta_{ij} =$

Orthonormal bases

Theorem: any set of orthogonal vectors is linearly independent (and therefore form a basis for $\text{span}(q_i)$)

Proof: just use the fundamental definition of linear independence. the q_i are linearly independent if

$$c_1 q_1 + c_2 q_2 + \dots + c_n q_n = 0 \text{ iff all } c_i = 0$$

Q Matrices (Orthogonal Matrices)

Definition: a matrix whose columns consist of orthonormal vectors we'll call a Q matrix (Orthogonal/Unitary matrix if Q is square)

Comments:

- 1) the columns of Q form an orthonormal basis for $C(Q)$
- 2) the fundamental property of Q matrices is $Q^T Q = \underline{\hspace{1cm}}$
- 3) If Q is square $Q^T = \underline{\hspace{1cm}}$ and $Q Q^T = \underline{\hspace{1cm}}$ (only true if square)
- 4) In general $Q Q^T$ is a projection matrix onto $C(Q)$

Q Matrices: Immediate practicality

Consider the overdetermined Least-Squares problem $A\mathbf{x}=\mathbf{b}$ if $A=Q$.

Point: in general $A \neq Q$, however A can be transformed to Q via a new algorithm, which leads to a new Factorization $A=QR$...this is where we are going.

Q Matrices: Examples

Example #1: Permutation Matrices

Q Matrices: Examples

Example #2: Rotation Matrices

Q Matrices: Examples

Example #3: Householder reflection matrices $H = I - 2\mathbf{u}\mathbf{u}^T$ ($\|\mathbf{u}\|=1$)

Q Matrices: Fundamental Properties

Property 1: Q matrices do not change the length of Vectors

Property 2: Q matrices preserve dot products

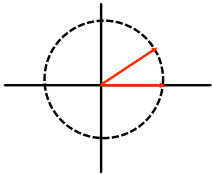
Comment: these properties make Q matrices very useful for numerical linear algebra

Q Matrices: Utility #2: Proof by basketball redux

Theorem: the dot product of any two unit vectors in $\mathbb{R}^n$ is

$$\mathbf{u}_1^T \mathbf{u}_2 = \cos(\theta)$$

Proof: Part 1--dot product of simple vectors in $\mathbb{R}^2 = \cos(\theta)$



Part 2: The dot product is invariant to rotations

Q Matrices: Utility #2: Proof by basketball redux

Theorem: the dot product of any two unit vectors in $\mathbb{R}^n$ is

$$\mathbf{u}_1^T \mathbf{u}_2 = \cos(\theta)$$

Proof: Part 3--Basketball time: The dot product of 2 unit vectors in $\mathbb{R}^n$ is equal to the dot product of 2 unit vectors in $\mathbb{R}^2 = \cos\theta$

Comment: Idea of **Change of basis** is very powerful

Lecture13

Gram-Schmidt Orthogonalization:

A new algorithm for transforming an arbitrary basis into an orthonormal basis (or transforming A to Q)

Problem: given a set of vectors,

$\mathbf{a}_1, \mathbf{a}_2, \dots, \mathbf{a}_n$
that form a basis for some subspace S , find an orthonormal basis

$\mathbf{q}_1, \mathbf{q}_2, \dots, \mathbf{q}_n$

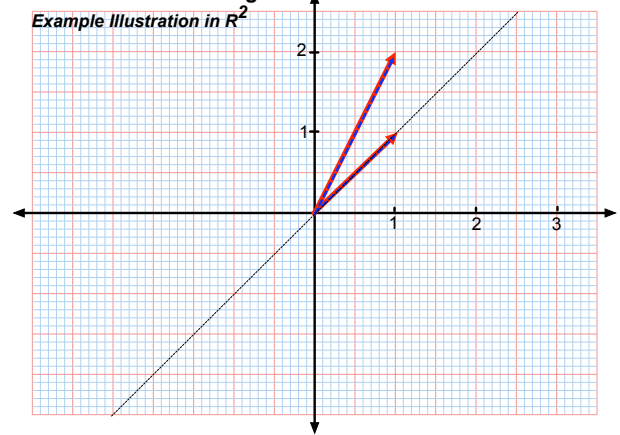
for the same space.

Comments:

- 1) if $\mathbf{a}_i$ form the columns of a matrix A and $\text{span } C(A)$, find an orthogonal matrix Q such that $C(Q)=C(A)$
- 2) An algorithm to do this (Gram-Schmidt Orthogonalization) is just another application of projections.
- 3) GS...is just a fancy name for "straightening out a bunch of vectors"

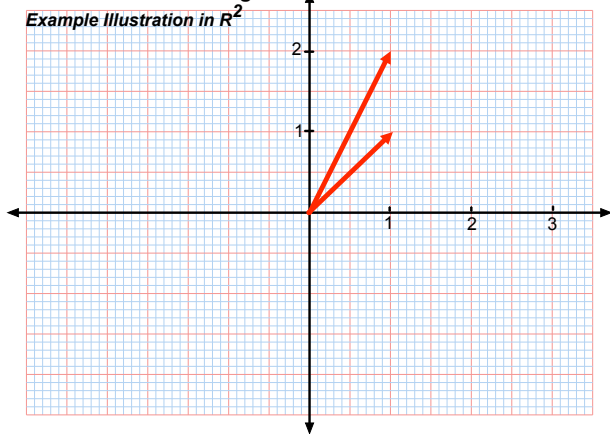
Gram-Schmidt Orthogonalization:

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



Gram-Schmidt Orthogonalization:

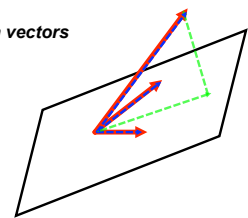
Example Illustration in $\mathbb{R}^2$



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 R^3 : find an orthonormal basis for the $C(A)$ where
 $A = [1 \ 1 \ 1; 1 \ 1 \ 0; 1 \ 0 \ 0]$;

2) Specific example

Gram-Schmidt Orthogonalization:

Example in R^3 : find an orthonormal basis for the $C(A)$ where
 $A = [1 \ 1 \ 1; 1 \ 1 \ 0; 1 \ 0 \ 0]$;

3) Comment Q not unique: suppose $A = [1 \ 1 \ 1; 0 \ 1 \ 1; 0 \ 0 \ 1]$