

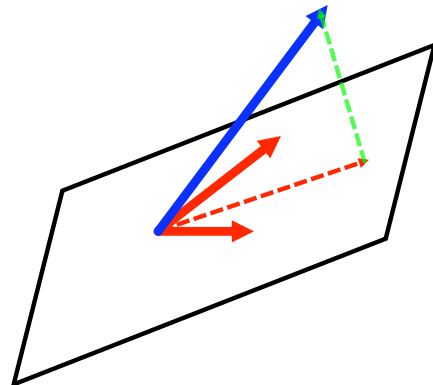
## Lecture 19: Numerical Linear Algebra #3

### Direct methods for solving Linear Least Squares problems

#### Outline

- 1) Quick Review: Linear Least Squares Problems and  $A^T Ax = A^T b$
- 2) Issues with solving the Normal Equations
- 3) Orthogonalization, the QR Factorization and Least Squares
  - A) Introduction
  - B) QR by modified Gram-Schmidt  $A \rightarrow Q$
  - C) QR by Householder Transformation  $A \rightarrow R$  (similarity to LU factorization)

***Review: Linear Least-squares problems and the Normal Equations.***



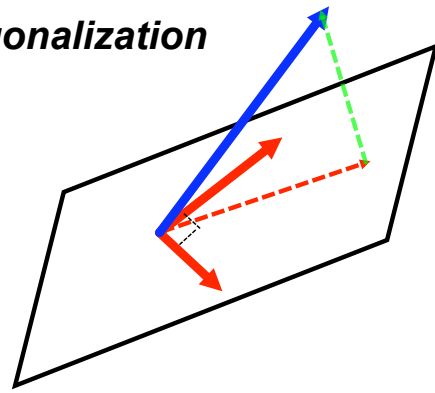
***Normal Equations and Condition number squaring***

The SVD and the  $\text{cond}(A)_2$

***Normal Equations and Condition number squaring***

Example:  $A = \begin{bmatrix} 1 & a & 0 \\ 1 & 0 & a \end{bmatrix}$ ,  $b = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$

***Solution of Least Squares by Orthogonalization  
(QR factorization)***



***Solution of Least Squares by Orthogonalization  
(QR factorization)***

**Theorem:** Every  $m \times n$  full-column rank Matrix  $A$  can be factored as  $A=QR$  where  $Q$  is  $m \times n$  and contains an orthonormal basis for  $C(A)$  and  $R$  is  $n \times n$  upper-triangular and invertible. (Proof by construction/Gram-Schmidt).

***Solution of Least Squares by Orthogonalization  
(QR factorization)***

**Theorem:** if  $A=QR$ , the solution of the normal Equations  $A^T A \hat{x} = A^T b$ ,  
is the solution of  $R\hat{x} = Q^T b$

***QR Factorization Algorithms:  
#1 Modified Gram-Schmidt Orthogonalization***

Idea: Take  $A \rightarrow Q$  by repeated projections, get  $R=Q^T A$

**QR Factorization Algorithms:  
#1 Modified Gram-Schmidt Orthogonalization**

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

**QR Factorization Algorithms:  
#1 Modified Gram-Schmidt Orthogonalization**

Algorithm:

**QR Factorization Algorithms:  
#2 Orthogonalization by Householder transformation**

Idea: Take  $A \rightarrow R$  directly by repeated applications of Q matrices  
(similar algorithm to LU factorization)

**QR Factorization Algorithms:  
#2 Orthogonalization by Householder transformation**

The Householder Reflection Matrix  $H = I - 2vv^T/v^T v$

Properties:

- 1) Symmetric matrix
- 2) Q matrix
- 3) Reflects a vector around the hyperplane orthogonal to  $\underline{v}$

**QR Factorization Algorithms:  
#2 Orthogonalization by Householder transformation**

*Example: Find  $H$  to transform  $a=[1\ 2\ 2]^T$  to  $[c\ 0\ 0]^T$*

**QR Factorization Algorithms:  
#2 Orthogonalization by Householder transformation**

*General Idea:*

**QR Factorization Algorithms:  
#2 Orthogonalization by Householder transformation**

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

**QR Factorization Algorithms:  
#2 Orthogonalization by Householder transformation**

Algorithm: (quick and dirty)

```
[m,n]=size(A); % get size of A

for k=1:n % loop over columns
    i=k:m; j=k:n; % sub-block indices

    % calculate the reflection vector v
    v=A(i,k); % extract subcolumn k
    v(1)=v(1)+sign(v (1))*norm(v); % modify first component by alpha
    v=v/norm(v); % normalize v

    % Multiply by Qk to transform subblock
    A(i,j) = A(i,j)-2*v*(v'*A(i,j)) ;
end
```

***QR Factorization Algorithms:  
#2 Orthogonalization by Householder transformation***

Operation Counts: