

## Lab Session 8

1. The purpose of this experiment is to illustrate that QR factorization by Householder reflector (MATLAB command `[Q, R] = qr(A)`) is better than modified Gram-Schmidt scheme (MGS) and classical Gram-Schmidt scheme (CGS).

Consider the  $n$ -by- $n$  Hilbert matrix  $H$  (use MATLAB command `H = hilb(n)` to generate  $H$ ). Your task is to use different methods listed below to orthonormalize the columns of  $H$  for  $n = 7$  and  $n = 12$ .

- (a) Write a MATLAB function implementing classical Gram-Schmidt method (CGS).

```
function [Q, R] = cgsqr(A)
% [Q, R] = cgsqr(A) employs classical Gram-Schmidt scheme to compute
% an isometry Q, an upper triangular matrix R such that A=QR.

[m, n] = size(A); % Assume that m >= n
Q = A; R = zeros(n);
for k = 1:n
    R(1:k-1,k) = Q(:,1:k-1)' * A(:,k);
    Q(:,k) = A(:,k) - Q(:,1:k-1) * R(1:k-1,k);
    R(k,k) = norm(Q(:,k));
    Q(:,k) = Q(:,k)/R(k,k);
end
```

- (b) Write a MATLAB function implementing modified Gram-Schmidt method (MGS).

```
function [Q, R] = mgsqr(A)
% [Q, R] = mgsqr(A) employs modified Gram-Schmidt scheme to compute
% an isometry Q, an upper triangular matrix R such that A=QR.

[m,n] = size(A); % Assume that m >=n
Q = A; R = zeros(n);
for k = 1:n
    R(k,k) = norm(Q(:,k));
    Q(:,k) = Q(:,k)/R(k,k);
    R(k,k+1:n) = Q(:,k)' * Q(:,k+1:n);
    Q(:,k+1:n) = Q(:,k+1:n) - Q(:,k) * R(k,k+1:n);
end
```

- (c) QR decomposition with reflectors. Use MATLAB command `[Q,R] = qr(H, 0)`, which produces an 'economy size' QR decomposition of  $H$  with  $Q$  being an isometry.

Examine the deviation from orthonormality by computing  $\|Q' * Q - \text{eye}(n)\|_2$  in each case (MATLAB command `norm(eye(n)-Q'*Q)`). Setting  $E := QR - H$ , we have  $H + E = QR$ . Check the residual norm  $\|E\|_2 = \text{norm}(H - Q * R)$ . The small residual error of order  $\mathcal{O}(u)$  as well as small deviation from orthonormality of order  $\mathcal{O}(u)$  imply that the algorithm is backward stable. In other words, the algorithm computes QR factorization of a slightly perturbed matrix which is indistinguishable from  $A$ . Test the backward stability of CGS, MGS and Householder QR factorization.

Find the condition number of  $H$  and check whether or not the matrix  $Q$  obtained from the MGS program satisfies  $\|Q' * Q - \text{eye}(n)\|_2 \approx u * \text{cond}(H)$ .

Did you get what you would expect in light of the values of unit roundoff  $u$  and  $\text{cond}(H)$ ? Which among all the above methods produces the smallest deviation from orthonormality?

2. This experiment is about detecting nearly rank deficiency of a matrix  $A$  via QR factorization  $A = QR$  by monitoring the size of diagonal entries of  $R$ . Generate the test matrix  $A$  as follows.

```
[U, X] = qr(randn(80));
[V, X] = qr(randn(80));
S = diag( 2.^ (-1:-1:-80));
A = U*S*V; % Note A has small singular values and is nearly rank deficient.

Now compute QR factorization of A using cgsqr, mgsqr and the matlab function qr:
[QC, RC] = cgsqr(A);
[QM, RM] = mgsqr(A);
[Q, R] = qr(A);
```

To test how close these matrices are to being unitary, compute  $\text{norm}(QC'*QC - \text{eye}(80))$ ,  $\text{norm}(QM'*QM - \text{eye}(80))$ ,  $\text{norm}(Q'*Q - \text{eye}(80))$ . Which method is worse? Which method gives better result?

To explain your results, plot the absolute values of the diagonal entries of  $RC$ ,  $RM$ ,  $R$ . Use commands

```
x= (1:80)';
hold off
semilogy(x, abs(diag( RC ) ), 'bo')
hold on
semilogy(x, abs(diag( RM ) ), 'rx')
semilogy(x, abs(diag( R ) ), 'k+')
title('abs(diag(R)) for cgs, mgs and qr')
gtext('cgs=o, mgs = x, qr=+')
```

Do the diagonal entries of  $R$  show near rank deficiency of  $A$ . Which method is better?

3. **Assignment.** Your task is to find polynomials  $p_5(t)$  and  $p_{18}(t)$  of degree 5 and 18, respectively, that best fit the function  $f(t) = \sin(\pi t/5) + \frac{t}{5}$  for  $t_1 = -5, t_2 = -4.5, \dots, t_{23} = 6$ , that is,  $\mathbf{t} = (-5:.5:6)'$ . For  $k = 5, 18$ , determine the polynomial  $p_k$  whose coefficients are given by  $x$  (that is,  $p_k(t) := \sum_{j=1}^{k+1} x_j t^{j-1}$ ) by solving LSP  $Ax = b$  in two different ways: solve LSP  $Ax = b$  using QR factorization of  $A$  and QR factorization of the augmented matrix  $[A \ b]$ . Here are the details.

Let  $[Q, R] = \text{cgsqr}(A)$  and  $[Q, R] = \text{mgsqr}(A)$  be MATLAB functions implementing classical Gram-Schmidt and modified Gram-Schmidt methods. For  $k = 5$  and  $k = 18$ , perform the following computations.

- Compute  $[Q, R] = \text{cgsqr}([A \ b])$  and use  $R$  to solve the LSP  $Ax = b$ . Compute the residual  $\text{res1} := \|Ax - b\|_2$  from the matrix  $R$ . Call the polynomial  $p_1(t)$ .  
Next compute  $[QC, RC] = \text{cgsqr}(A)$  and use  $QC$  and  $RC$  to solve the LSP  $Ax = b$ . Compute the residual  $\text{res2} := \|Ax - b\|_2$ . Call the polynomial  $p_2(t)$ . Which method gives a better fit (small residual error)? Plot  $p_1(t), p_2(t)$  and  $f(t)$  in a single plot and comment on the result.
- Compute  $[Q, R] = \text{mgsqr}([A \ b])$  and use  $R$  to solve the LSP  $Ax = b$ . Compute the residual  $\text{res3} := \|Ax - b\|_2$  from the matrix  $R$ . Call the polynomial  $p_3(t)$ .  
Next compute  $[QM, RM] = \text{mgsqr}(A)$  and use  $QM$  and  $RM$  to solve the LSP  $Ax = b$ . Compute the residual  $\text{res4} := \|Ax - b\|_2$ . Call the polynomial  $p_4(t)$ . Which method gives a better fit (small residual error)? Plot  $p_3(t), p_4(t)$  and  $f(t)$  in a single plot and comment on the result.
- Compute  $[Q, R] = \text{qr}([A \ b])$  and use  $R$  to solve the LSP  $Ax = b$ . Compute the residual  $\text{res5} := \|Ax - b\|_2$  from the matrix  $R$ . Call the polynomial  $p_5(t)$ .  
Next solve the LSP  $Ax = b$  using MATLAB command  $\mathbf{x} = A \backslash \mathbf{b}$  and compute the residual  $\text{res6} := \|Ax - b\|_2$ . Call the polynomial  $p_6(t)$ . Which method gives a better fit (small residual error)? Plot  $p_5(t), p_6(t)$  and  $f(t)$  in a single plot and comment on the result.

- (d) Plot the data points  $(t_i, f_i)$ ,  $f(t)$ , and the best fit polynomials of degree 5 and 18 obtained from (c) in a single plot. Use the command `axis([-10 10 -3 3])` to set the axes. Which polynomial gives a better fit?

Among the six methods, which method provides the best fit? What is the impact of the condition number of  $A$  on these methods?

**12 marks**

\*\*\* End \*\*\*