

# IAM 961 HW3: QR decomp

Due Monday Nov 2, 2015.

## Problem 1.

Write the following Julia functions for computing the QR decomposition of a matrix

- `qrcgs(A)` via Classical Gram-Schmidt orthogonalization,
- `qrmgs(A)` via Modified Gram-Schmidt orthogonalization, and
- `qrhouse(A)` via Householder triangularization.

```
In [110]: function qrcgs(A)
           # fill in
         end

         function qrmgs(A)
           # fill in
         end

         function qrhouse(A)
           # fill in
         end
```

```
Out[110]: qrhouse (generic function with 1 method)
```

## Problem 2.

Test that your QR algorithms work correctly on a fairly small and well-conditioned matrix (e.g. a 5 x 5 matrix with normally distributed elements,  $A = \text{randn}(5, 5)$ ). You should test that  $Q$  is unitary and that  $QR \approx A$ . Verify to your own satisfaction that  $R$  is upper-triangular. Make these tests as compact and readable as you can!

```
In [ ]:
```

## Problem 3.

Write a `backsub(R, b)` function that computes a solution of  $Rx = b$  by backsubstitution. You can assume that  $R$  is square and nonsingular. Test your backsubstitution function by solving an  $Ax = b$  problem with your 5 x 5  $A$  matrix, one of your QR algorithms, and a known solution  $x$ .

```
In [111]: function backsub(R, b)
           # fill in
           end
```

Out[111]: backsub (generic function with 1 method)

## Problem 4.

Write a function `A = randmatrix(m, n, kappa)` function that returns an  $m \times n$  random matrix with condition number  $\kappa$  and exponentially graded singular values (i.e.  $\sigma_1/\sigma_m = \kappa$  and  $\sigma_{j+1}/\sigma_j = \text{const}$ ). You can use the Matlab code at the top of pg 65 in Trefethen and Bau as a starting point. Test that it works by constructing a  $4 \times 4$  matrix with  $\kappa=10^8$  and then computing its condition number.

```
In [45]: function randmatrix(m,n, kappa)
           # fill in
           end
```

Out[45]: randmatrix (generic function with 1 method)

## Problem 5.

Solve a large number of random  $Ax = b$  problems using your QR decompositions and `randmatrix` and `backsubstitution` functions, and produce a scatter plot of the normalized solution error  $\|\hat{x} - x\|/\|x\|$  versus  $\kappa$ . Plot data points from CGS in blue, MGS in red, and Householder in green.

Specifically: Construct a random  $A$  matrix with  $\kappa = 10^n$  where  $n$  is a random real-valued number uniformly distributed between 0 and 16. Select a random  $x$  vector with  $x = \text{randn}(m, 1)$ , and then set  $b = Ax$ . For each of the CGS, MGS, and Householder QR algorithms, compute the numerical solution  $\hat{x}$  of  $Ax = b$  via QR and then plot  $\|\hat{x} - x\|/\|x\|$  versus  $\kappa$  using log-log axes and the color scheme specified above. Do this for one hundred random  $Ax = b$  problems and for a fairly small value of  $m$  (perhaps 10 or 20).

In [ ]:

## Problem 6

Comment on your results. What can you explain about the scatter plots based on the algorithms and their implementation in finite-precision arithmetic? Or, contrariwise, what can you say about the algorithms based on the scatter plots?

## Problem 7

If you are curious, repeat problem 5 for a different value of  $m$  (perhaps  $m = 100$ ). Does the dimensionality of the matrix (the value of  $m$ ) make any difference?

In [ ]: