

# Assignment 2: Orthogonalization

## Math 327/397 Winter 2019

Due: Friday Feb 22, 2019

### Instructions

Submit a complete paper copy of your solutions in class or to the math department by 4pm on the due date. Questions 1 – 5 are for everyone. Question 6 is for Math 397 only.

### Question 1: Orthogonal Matrices

- (a) Let  $u, v \in \mathbb{R}^n$ ,  $\langle u, v \rangle = u^T v$  denote the standard inner product, and  $\|\cdot\|$  denote the induced (2-)norm. If  $Q \in \mathbb{R}^{n \times n}$  is orthogonal, show that  $Q$  preserves inner products, norms, distances, and angles under matrix-vector multiplication.
- (b) Show that the orthogonal projection of  $u$  onto  $v$ , is the vector in  $\text{span}(v)$  closest to  $u$ . Extend the result to subspaces, that is, if  $W \subseteq \mathbb{R}^n$  is a subspace, show that the orthogonal projection of  $u$  onto  $W$  is the vector inside  $W$  closest to  $u$ . Is the closest vector unique? Justify.

### Question 2: Classical Gram-Schmidt

Consider the matrix

$$A = \begin{bmatrix} 1 & -2 & 1 \\ 0 & 1 & 3 \\ 1 & 1 & 0 \\ 0 & 1 & -1 \end{bmatrix}.$$

In constructing the QR decomposition, ensure that  $R$  has non-negative diagonal elements. For by-hand computations keep at least four significant digits in your intermediate calculations.

- (a) Compute by hand the reduced QR decomposition of  $A$  using the Classical Gram-Schmidt method. Show your work.

- (b) Assuming the input is full rank, write the pseudo-code for the Classical Gram-Schmidt (you may refer to the class notes).
- (c) Implement the Classical Gram-Schmidt into a Matlab m-file.
- (d) Run your algorithm on  $A$  and paste the output.
- (e) How many floating point operations does the algorithm require for an  $m \times n$  input matrix? How much memory storage counted in floating point numbers? Justify, aiming to get at least the order and the first coefficient correct.

### Question 3: Modified Gram-Schmidt

Repeat question 2, (a)–(d) only, with the Modified Gram-Schmidt.

### Question 4: Givens Rotations

Repeat question 2, (a)–(d) only, with the Givens Rotations method. In this case, construct the full QR decomposition instead of the reduced one.

### Question 5: Householder Reflections

Repeat question 2, (a)–(d) only, with the Householder Reflections method. Again, construct the full QR decomposition.

### Question 6 (Math 397): Loss of Orthogonality

Define  $B = \text{rand}(100, 80)$  in Matlab.

- (a) In your m-files for Classical Gram-Schmidt, add lines of code to track the loss of orthogonality using the following metric:  $LCGS_k := \|Q_k^T Q_k - I_k\|_F$  at each step. Here  $\|\cdot\|_F$  refers to the Frobenius Norm. Do the same for Modified Gram-Schmidt defining the loss of orthogonality as  $LMGS_k$ . Run both algorithms on  $B$  and plot  $LCGS_k$  and  $LMGS_k$  on the same chart against the iteration index  $k$ . (You might find it useful to explore different plotting options for better visualization, e.g. `logplot`.) Which method loses orthogonality faster?
- (b) Run your Givens rotations and Householder methods on  $B$ . At completion compute the corresponding loss of orthogonality quantities  $LG_n$ ,  $LH_n$ . Here, only consider the relevant portion of  $Q$ , whose column space is the same as the one of  $B$ . Comparing  $LCGS_n$ ,  $LMGS_n$ ,  $LG_n$ , and  $LH_n$ , which of the four methods gives the most orthogonal  $Q$  according to this metric?