## **Review Topics**

## for In-Class Midterm Exam on Friday 22 October, 2021

The Midterm Quiz will cover Lectures 1, 2, 3, 4, 5, 6, 7, 8, 10, 11 in Trefethen & Bau. (Lecture 9 will not be covered, but feel free to state algorithms as runnable Matlab!) The problems will be of these types: state definitions, state theorems, state algorithms (as pseudocode or Matlab), describe or illustrate basic geometrical ideas, make basic applications of theorems, make basic calculations, prove simple theorems/corollaries.

## **Definitions**. Know how to define:

- matrix-vector product; matrix-matrix product
- adjoint; hermitian; transpose
- inner product; outer product
- unitary
- $\|\cdot\|_p$  of a vector in  $\mathbb{C}^m$  for any  $1 \leq p \leq \infty$
- induced matrix norm  $\|\cdot\|$
- Frobenius matrix norm  $\|\cdot\|_F$
- projector; orthogonal projector
- eigenvalue; eigenvector
- singular value

Matrix Factorizations and Constructions. Know the properties of the factors in each factorization below. (For example,  $\hat{U}$  has orthonormal columns and is of same size as A in the  $m \geq n$  case of the reduced SVD factorization  $A = \hat{U}\hat{\Sigma}V^*$ .) Assume  $A \in \mathbb{C}^{m \times n}$  unless otherwise stated. Be able to use the factorization in simple calculations. Be able to compute the factorization by hand in sufficiently small and simple cases.

- full SVD:  $A = U\Sigma V^*$
- reduced SVD,  $m \ge n$ :  $A = \hat{U}\hat{\Sigma}V^*$
- full QR,  $m \ge n$ : A = QR
- reduced QR, m > n:  $A = \hat{Q}\hat{R}$
- eigenvalue, m = n:  $A = X\Lambda X^{-1}$
- orthogonal projector:  $P = \hat{Q}\hat{Q}^* = A(A^*A)^{-1}A^*$
- Householder reflector:  $F = I 2vv^*/(v^*v)$

**Algorithms.** Be able to state these algorithms, including the amount of work to leading order.

- matrix-vector and matrix-matrix products
- $\bullet$  Alg. 7.1: classical Gram-Schmidt for QR
- Alg. 8.1: modified Gram-Schmidt for QR
- Alg. 10.1: Householder triangularization for QR
- Alg. 10.2: compute  $Q^*b$  after Householder triangularization
- Alg. 11.2: QR for least squares on overdetermined "Ax = b"
- Alg. 11.1: normal equations for least squares on overdetermined "Ax = b"

Facts and Formulas. Know as facts. Be able to prove unless otherwise stated.

- Cauchy-Schwarz:  $|x^*y| \le ||x|| ||y||$  [proof not required]
- invariance of  $\|\cdot\|_2$  and  $\|\cdot\|_F$  matrix norms under unitaries
- $\bullet \|A\|_F = \sqrt{\sigma_1^2 + \dots + \sigma_r^2}$
- $||A||_2 = \sigma_1$
- rank(A) is number of nonzero singular values (in exact arithmetic)
- for  $A \in \mathbb{C}^{m \times n}$  with  $m \geq n$ , A has full rank if and only if  $A^*A$  is nonsingular
- the singular values of A are the square roots of the eigenvalues of  $A^*A$
- if P is a projector then so is I P
- if P is an orthogonal projector then so is I-P, and furthermore I-2P is unitary

**Ideas**. Please be comfortable with the following ideas! Some ideas correspond to theorems, but otherwise it is just a perspective or paradigm.

- L1 and L2: how to think about Ax,  $A^{-1}b$ , Qx,  $Q^*b$
- L4: the image of the unit sphere under any  $m \times n$  matrix is a hyperellipsoid
- L5: sums like this are optimal (in what sense?) approximations of A:

$$A_{\nu} = \sum_{j=1}^{\nu} \sigma_j u_j v_j^*$$

- L7: construction of orthogonal functions (e.g. orthogonal polynomials) is an application of the Gram-Schmidt process, and/or of A = QR, when the columns of A are infinitely long
- L8: the leading-order number of operations in some algorithm can be counted by geometric arguments
- L10: Gram-Schmidt is "triangular orthogonalization" while Householder is "orthogonal triangularization" <sup>1</sup>

<sup>&</sup>lt;sup>1</sup>Gauss elimination is "triangular triangularization" ... but it will not be on the Midterm Exam!