Numerical Linear Algebra

Sheet 1 - MT20

Norms and SVD (due 9am, 2 working days before class)

This sheet is due 9am two weekdays before the class (Thursday if class is on Monday). For PartC/OMMS students, submit by emailing your work to meier@maths.ox.ac.uk with title 'NLA sheet 1 (your name)'.

Submission via the virtual pigeonhole is recommended once it is set up.

- 1. Show that $||x||_{\infty} = \max_i |x_i|$ satisfies the axioms for a vector norm.
- 2. Show that if ||x|| is a vector norm then $\sup_x \frac{||Ax||}{||x||}$ satisfies the axioms for a matrix norm. Further show that

$$||AB|| \le ||A|| \, ||B||.$$

3. From the definition of the vector 1–norm show that

$$||A||_1 = \max_j \sum_i |a_{ij}|.$$

4. By considering the individual columns a_j of A and b_j of B = QA, show that

$$\|QA\|_{\mathbf{F}} = \|A\|_{\mathbf{F}}$$

if Q is an orthogonal matrix.

- 5. Full SVD. Prove the existence of $A = U \begin{bmatrix} \Sigma \\ 0_{(m-n) \times n} \end{bmatrix} V^*$, where $U \in \mathbb{C}^{m \times m}$ and $V \in \mathbb{C}^{n \times n}$ are unitary matrices i.e., $U^*U = I_m$ and $V^*V = I_n$, and $\Sigma \in \mathbb{R}^{n \times n}$ is diagonal.
- 6. What is the SVD of a normal matrix A, with respect to the eigenvalues and eigenvectors? What if A is symmetric? And unitary?
- 7. If $A \in \mathbb{R}^{n \times n}$ is nonsingular, what is the SVD of A^{-1} in terms of that of A?
- 8. Let B be a square $n \times n$ matrix. Bound the *i*th singular values of AB using $\sigma_i(A)$ and $\sigma_i(B)$: Specifically, prove that for each *i*,

$$\sigma_i(A)\sigma_n(B) \le \sigma_i(AB) \le \sigma_i(A)\sigma_1(B).$$

9. (optional; harder) Let $A \in \mathbb{R}^{m \times n}$, $m \ge n$ and $\sigma_i(A) \ge \sigma_i(A) \ge \cdots \ge \sigma_n(A) \ge 0$ be its singular values. Prove that for $k = 1, 2, \dots, n$,

$$\sum_{i=1}^{k} \sigma_i(A) = \max_{Q^T Q = I_k, W^T W = I_k} \operatorname{trace}(Q^T A W).$$

 $(Q \in \mathbb{R}^{m \times k}, W \in \mathbb{R}^{n \times k} \text{ are orthonormal. Recall for an } k \times k \text{ matrix } B, \text{ trace}(B) = \sum_{i=1}^{k} B_{ii}; \text{ a useful property is trace}(CD) = \text{trace}(DC) \text{ as long as } CD \text{ is square.})$