

CS 3220: PRELIM 1 EXAMPLE QUESTIONS

Instructor: Anil Damle

WHATS IN THIS DOCUMENT

These questions are intended to be somewhat representative of the type of questions that could arise on the first prelim.

QUESTION 1

Here, we consider some properties of $\|A\|_F$. For what follows $A \in \mathbb{R}^{n \times n}$

(a) Prove that

$$\|A\|_F^2 = \sum_{i=1}^n \|A(:, i)\|_2^2.$$

(b) For any two orthogonal matrices $Q_1 \in \mathbb{R}^{n \times n}$ and $Q_2 \in \mathbb{R}^{n \times n}$ prove that

$$\|Q_1 A Q_2\|_F = \|A\|_F$$

(c) Prove that

$$\|A\|_F = \sqrt{\sum_{i=1}^n \sigma_i^2},$$

where $\sigma_1, \dots, \sigma_n$ are the singular values of A .

SOLUTION

(a) We have that $\|A\|_F^2 = \sum_{i,j} A_{i,j}^2$, which is equivalent to

$$\|A\|_F^2 = \sum_{j=1}^n \sum_{i=1}^n A_{i,j}^2.$$

The inner loop is $\|A(:, j)\|_2^2$, thereby completing the proof.

(b) Since the two norm of vectors is invariant under orthogonal transformations, the prior part gives us immediately that

$$\begin{aligned} \|Q_1 A Q_2\|_F^2 &= \sum_{i=1}^n \|Q_1 (A Q_2)(:, i)\|_2^2 \\ &= \sum_{i=1}^n \|(A Q_2)(:, i)\|_2^2 \\ &= \|A Q_2\|_F^2. \end{aligned}$$

Using the fact that for any matrix B , $\|B\|_F^2 = \|B^T\|_F^2$ we can use the same argument to remove Q_2 .

(c) If $A = U\Sigma V^T$ we have that

$$\|A\|_F^2 = \|\Sigma\|_F^2.$$

Explicitly writing out the right hand side and taking the square root yields the result.

QUESTION 2

Here, we ask you to interpret the condition number of a 2×2 matrix geometrically. (Hint: pictures are useful here!)

1. We saw that the SVD of a 2×2 matrix allows us to view any matrix A as mapping a circle to an ellipse. If A becomes increasingly ill-conditioned what is geometrically happening to the ellipse?
2. Geometrically argue why for an ill-conditioned matrix a small relative change in b can result in a big relative change in x .

SOLUTION

- (a) The ratio of the length of the major axis to minor axis of the ellipse is going to infinity, so the ellipse is collapsing to a line segment.
- (b) Because of the elongated shape of the ellipse, changes to b in the direction of v_2 can drastically alter the location along the ellipse.

QUESTION 3

Say you are given a symmetric matrix A and tasked with computing the algebraically smallest eigenvalue. Using only the power method (applied to A or matrices related to A), how might you go about doing this? (Hint: think about how the eigenvalues/vectors of $A - \gamma I$ relate to those of A for any scalar $\gamma \in \mathbb{R}$.)

SOLUTION

We can first use the power method to compute the largest eigenvalue in magnitude of A . If it is negative we are done, if not we need to do a bit more work. Let μ be the eigenvalue of A that we previously computed. Now, we can simply use the power method to compute an eigenvalue of $A - \mu I$, call it λ and then $\lambda + \mu$ is the algebraically smallest eigenvalue of A . This works because the eigenvalues of $A - \mu I$ are simply those of A shifted right by μ . Therefore the algebraically largest eigenvalue of $A - \mu I$ is 0 and the largest magnitude one is necessarily the smallest algebraic.