**Problem 1.** (Operator norm vs. Frobenius norm) For a vector  $x \in \mathbb{R}^n$ , the notation ||x|| will denote the usual Euclidean 2-norm. Recall, that given a matrix  $A \in \mathbb{R}^{m \times n}$ , the operator norm (induced by  $||\cdot||$ ) is defined as

$$||A|| := \sup_{||x||=1} ||Ax||;$$

while the Frobenius norm is

$$||A||_F := \left(\sum_{i,j} |A(i,j)|^2\right)^{1/2} = \left(\operatorname{Tr}\left(A^T A\right)\right)^{1/2}.$$

Recall also that the operator norm is the smallest constant C such that

$$(1)  $||Ax|| \le C||x|| \forall x \in \mathbb{R}^n,$$$

so in particular, the operator norm satisfies the submultiplicative property:

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

In class, we showed that the Frobenius norm also satisfies

$$(2) ||Ax|| \le ||A||_F ||x||,$$

and hence it also has the submultiplicative property.

An alternative way of showing all this would to simply establish the fact that

(3) 
$$||A|| \le ||A||_F$$
,

because then (1) would hold for  $C = ||A||_F$ , which is (2).

There is fairly simple way to compare the operator norm and the Frobenius norm, namely, by writing both in terms of the singular values  $\sigma_1 \ge \cdots \ge \sigma_r > 0$  of the matrix A. Recall that

$$||A|| = \sigma_1 = \sigma_{\max}(A)$$
 and  $||A||_F = \sqrt{\sigma_1^2 + \dots + \sigma_r^2}$ .

In particular,

$$||A||^2 = \sigma_1^2 \le \sigma^2 + \dots + \sigma_r^2 = ||A||_F^2$$

Compute the operator norm and the Frobenius norm of the matrix A below, by first finding its singular values (i.e., solve  $\det(\lambda I - A^T A) = 0$  for  $\lambda$ ):

$$A = \left[ \begin{array}{cc} 2 & 2 \\ -1 & 1 \end{array} \right]$$

**Problem 2.** Consider  $A \in \mathbb{R}^{n \times n}$ . Use the Singular Value Decomposition of A to show that

$$||A^T A||_2 = ||A||_2^2$$

where  $\|\cdot\|_2$  is the operator norm induced by the 2-norm.

Conclude that

$$\kappa_2(A^T A) = \kappa_2(A)^2.$$

Finally, let A be the 2-by-2 matrix from Problem 1, determine its singular value decomposition and deduce its inverse from it.

**Problem 3.** A matrix A is called *power bounded* if there exists a number C>0 so that

$$||A^k|| \le C$$
 for all  $k = 0, 1, 2, ...$ 

(Note that the norm is not specified.)

- (a) Prove that the definition of power boundedness is independent of the norm chosen. Namely, show that power boundedness in one norm implies power boundedness in all norms.
- (b) Let  $\|\cdot\|$  be a matrix norm satisfying  $\|AB\| \le \|A\| \|B\|$ , for instance the operator  $\|\cdot\|_2$ , or the Frobenius norm). Show that if  $\|A\| \le 1$ , then A is power bounded.
- (c) Part (b) shows that  $||A||_2 \le 1$  is a sufficient condition for power boundedness. However, show that it is not a necessary condition, by finding an example of a  $2 \times 2$  power bounded matrix that satisfies  $||A||_2 > 1$ .