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## Practice Problems

1. Consider the matrix

$$A = \begin{pmatrix} 1 & -1 & -1 \\ -1 & 1 & -1 \\ -1 & -1 & 1 \end{pmatrix}.$$

One eigenvalue of A is -1. The other eigenvalue is a double root of  $\det(A - \lambda I)$ . What is the other eigenvalue? (Hint: use the trace, don't do much calculation.)

Solution. We will use the fact that the trace of A is the sum of its eigenvalues. Say the mystery eigenvalue is  $\lambda$ . Then we have

$$3 = -1 + 2\lambda$$
$$4 = 2\lambda$$
$$\lambda = 2.$$

**2.** Suppose M is a positive Markov matrix (so one eigenvalue equals 1, all other eigenvalues have  $|\lambda| < 1$ ). Why is  $M^{\infty}$  a rank-1 matrix?

Solution. We know that the eigenvalues of  $M^n$  are the nth powers of eigenvalues of M. So for n large,  $M^n$  has one eigenvalue equal to 1 and the remaining eigenvalues are extremely small numbers. This means that  $M^{\infty}$  has one eigenvalue equal to 1 and all other eigenvalues equal to zero. The number of nonzero eigenvalues is the rank of the matrix (the number of zero eigenvalues is the dimension of the nullspace).

**3.**  $x^T A y = \text{tr}(AB)$  where B is what matrix? (Hint: use the cyclic property of the trace and recall that the trace of a 1x1 matrix a is a).

Solution. We would like to get a trace on the left hand side. The left hand side is a scalar, so is equal to its own trace. That is,  $x^TAy = \operatorname{tr}(x^TAy)$ . We also know that  $\operatorname{tr}(BC) = \operatorname{tr}(CB)$  for any B, C. So  $\operatorname{tr}(x^TAy) = \operatorname{tr}(Ayx^T)$ , and  $B = yx^T$ .

**4.** Suppose A is an  $m \times n$  full column-rank matrix with thin SVD  $U\Sigma V^T$  (so that V is square/unitary and U is  $m \times n$ ). By inspection of  $A^TA$  in comparison with the diagonalization formula, the eigenvectors of  $A^TA$  are \_\_\_ and its eigenvalues are \_\_\_ .

Solution. First, we compute  $A^TA$ :

$$A^TA = V\Sigma U^T(U\Sigma V^T) = V\Sigma^2 V^T$$

since U has orthonormal columns so  $U^TU=I$ . The diagonalization formula is  $B=X\Lambda X^{-1}$ , where the columns of X are the eigenvectors of B and  $\Lambda$  is diagonal and the diagonal entries are eigenvalues of B. Comparing, we see that the columns of V are the eigenvectors of  $A^TA$  and the eigenvalues are  $\sigma^2$ , the squares of the singular values.

**5.** Suppose A is  $m \times m$ , full rank, and we compute its QR factorization A = QR, e.g. by Gram-Schmidt. Claim: the matrix B = RQ has the same eigenvalues as A. Why?

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Solution. We start by relating B to A. Since A = QR, the matrix

$$RQ = Q^{-1}(QR)Q = Q^{-1}AQ.$$

This tells us that A and RQ are similar matrices, so they have the same eigenvalues.