- **1.** (15 points)
- (a) If A is a 3 by 4 matrix, what does this tell us about its nullspace? Solution:  $\dim N(A) \ge 1$ , since  $\operatorname{rank}(A) \le 3$ .
- (b) If we also know that

$$Ax = \begin{bmatrix} 1 \\ 1 \\ 1 \end{bmatrix}$$

has no solution, what do we know about the rank of A? Solution: C(A) does not span the entire  $\mathbf{R}^3$ , so  $rank(A) \leq 2$ .

(c) If Ax = b and  $A^Ty = 0$ , find  $y^Tb$  by using those equations. This says that the \_\_\_\_\_ are \_\_\_\_.

Solution:  $y^Tb = y^T(Ax) = (A^Ty)x = 0$ . This says that the **column** space of A and the **null space** of  $A^T$  are **orthogonal**.

**2.** (15 points) Suppose Ax = b reduces by the usual row operations to Ux = c:

$$Ux = \begin{bmatrix} 2 & 6 & 4 & 8 \\ 0 & 0 & 4 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \\ x_3 \\ x_4 \end{bmatrix} = \begin{bmatrix} b_1 \\ b_2 - b_1 \\ b_3 - 2b_2 + b_1 \end{bmatrix} = c.$$

(a) Give a basis for the nullspace of A (that matrix is not shown) and a basis for the row space of A.

Solution: N(A) = N(U), R(A) = R(U). Therefore we can read the bases directly from U:

$$N(A) = span \left\{ \begin{bmatrix} -3\\1\\0\\0 \end{bmatrix}, \begin{bmatrix} -2\\0\\-1\\1 \end{bmatrix} \right\}$$

$$R(A) = span \left\{ \begin{bmatrix} 2\\6\\4\\8 \end{bmatrix}, \begin{bmatrix} 0\\0\\4\\4 \end{bmatrix} \right\}$$

(b) When does Ax = b have a solution? Give a basis for the column space of A. Solution:  $b \in C(A)$ , equivalent to  $c \in C(U)$ .

By looking at c as a function of b we can reconstruct A. Let  $E = \begin{bmatrix} 1 & 0 & 0 \\ -1 & 1 & 0 \\ 1 & -2 & 1 \end{bmatrix}$ . We

have 
$$U = EA$$
,  $c = Eb$ . Hence  $A = E^{-1}U = \begin{bmatrix} 1 & 0 & 0 \\ 1 & 1 & 0 \\ 1 & 2 & 1 \end{bmatrix} \begin{bmatrix} 2 & 6 & 4 & 8 \\ 0 & 0 & 4 & 4 \\ 0 & 0 & 0 & 0 \end{bmatrix} = \begin{bmatrix} 2 & 6 & 4 & 8 \\ 2 & 6 & 8 & 12 \\ 2 & 6 & 12 & 16 \end{bmatrix}$ 

From here we see that

$$C(A) = span \left\{ \begin{bmatrix} 1\\1\\1 \end{bmatrix}, \begin{bmatrix} 1\\2\\3 \end{bmatrix} \right\}.$$

(c) Give a basis for the nullspace of  $A^T$ .

Solution:  $N(A^T) \perp C(A)$ . So  $N(A^T) = span \left\{ \begin{bmatrix} 1 \\ -2 \\ 1 \end{bmatrix} \right\}$  (we only need to find a vector orthogonal to both basis vectors we gave for C(A)).

- **3.** (10 points)
- (a) Suppose  $q_1, q_2$  are orthonormal in  $\mathbb{R}^4$ , and v is NOT a combination of  $q_1$  and  $q_2$ . Find a vector  $q_3$  by Gram-Schmidt, so that  $q_1, q_2, q_3$  is an orthonormal basis for the space spanned by  $q_1, q_2, v$ .

Solution:

$$q_3 = \frac{v - (v^T q_1)q_1 - (v^T q_2)q_2}{\|v - (v^T q_1)q_1 - (v^T q_2)q_2\|}$$

(b) If p is the projection of b onto the subspace spanned by  $q_1$  and  $q_2$  and v, find p as a combination of  $q_1, q_2, q_3$ . (You are solving the least squares problem Ax = b with  $A = [q_1, q_2, q_3]$ .)

Solution:

$$p = A(A^{T}A)^{-1}A^{T}b = AA^{T}b = q_1(q_1^{T}b) + q_2(q_2^{T}b) + q_3(q_3^{T}b)$$

where we used the fact that the columns of A are orthonormal for  $A^TA = I$ . (It is easy to see the final result just by thinking that we are actually projecting onto an orthonormal basis).

- **4.** (10 points)
- (a) To solve a square system Ax = b when det  $A \neq 0$ , Cramer's Rule says that the first component of x is

$$x_1 = \frac{\det B}{\det A}$$
 with  $B = [b a_2 \dots a_n]$ .

So b goes into the first column of A, replacing  $a_1$ . If  $b = a_1$ , this formula gives the right answer  $x_1 = \frac{\det A}{\det A} = 1$ .

- 1. If b = a different column  $a_j$ , show that this formula gives the right answer,  $x_1 = a$ .
- 2. If b is any combination  $x_1a_1 + \cdots + x_na_n$ , why does this formula give the right answer  $x_1$ ?

Solution:

- 1.  $x_1 = 0$ , since det(B) = 0. Let us check that it gives the correct answer: the solution to  $Ax = a_j$  is  $x = e_j$  (it is unique, since  $det(A) \neq 0$ , so A's columns are linearly independent). Hence  $x_1 = 0$ .
- 2. For the same reason as above the first component of x is indeed  $x_1$ , the coefficient of  $a_1$ .

Now let us see what Cramer's Rule gives. In this case  $\det(B) = \det([b \, a_2 \, \dots \, a_n]) = \det(x_1 \cdot a_1 \, a_2 \, \dots \, a_n) = x_1 \det(A)$ . So, indeed,  $x_1 = \det(B)/\det(A)$ .

(b) Find the determinant of

$$A = \left[ \begin{array}{rrrr} 1 & 2 & 0 & 0 \\ 3 & 4 & 0 & 0 \\ 5 & 0 & 6 & 7 \\ 0 & 0 & 8 & 9 \end{array} \right].$$

Cofactor expansion by the first row:

$$C_{11} = \det \begin{bmatrix} 4 & 0 & 0 \\ 0 & 6 & 7 \\ 0 & 8 & 9 \end{bmatrix} = 4 \cdot 6 \cdot 9 - 4 \cdot 8 \cdot 7 = -8$$

$$C_{12} = -\det \begin{bmatrix} 3 & 0 & 0 \\ 5 & 6 & 7 \\ 0 & 8 & 9 \end{bmatrix} = -(3 \cdot 6 \cdot 9 - 3 \cdot 7 \cdot 8) = 6$$

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So  $det(A) = 1 \cdot (-8) + 2 \cdot 6 = 4$ 

- **5.** (15 points)
- (a) Suppose an n by n matrix A has n independent eigenvectors  $x_1, \ldots, x_n$  with eigenvalues  $\lambda_1, \ldots, \lambda_n$ . What matrix equation would you solve for  $c_1, \ldots, c_n$  to write the vector  $u_0$  as a combination  $u_0 = c_1 x_1 + \cdots + c_n x_n$ ?

Solution:

$$\begin{bmatrix} x_1 x_2 \dots x_n \end{bmatrix} \begin{bmatrix} c_1 \\ \vdots \\ c_n \end{bmatrix} = u_0$$

(b) Suppose a sequence of vectors  $u_0, u_1, u_2, \ldots$  starts from  $u_0$  and satisfies  $u_{k+1} = Au_k$ . Find the vector  $u_k$  as a combination of  $x_1, \ldots, x_n$ .

Solution: Let  $u_0 = c_1 x_1 + \dots + c_n x_n$ . Then  $u_k = A^k u_0 = \sum_{i=1}^n (\lambda_i^k c_i) x_i$ .

(c) State the exact requirement on the eigenvalues  $\lambda$  so that  $A^k u_0 \to 0$  as  $k \to \infty$  for every vector  $u_0$ . Prove that your condition **must hold**.

Solution:  $|\lambda_i| < 1$ , for all i. Clearly, if this holds, all the coefficients of  $x_i$  in  $A^k u_0$  go to 0 as  $k \to \infty$ .

For the converse, we require that the coefficient  $\lambda_i^k c_i$  of  $x_i$  in  $A^k u_0$  to go to 0, for any choice of  $u_0$ . Equivalently, we need  $\lambda_i^k c_i \to 0$  for any  $c_i$ . Hence  $|\lambda_i| < 1$ .

- **6.** (10 points)
- (a) Find the eigenvalues of this matrix A (the numbers in each column add to zero).

$$A = \left[ \begin{array}{rrr} -1 & \frac{1}{2} & 0 \\ 1 & -1 & 1 \\ 0 & \frac{1}{2} & -1 \end{array} \right].$$

Solution: The number in each column add to zero, hence  $\mathbf{1} \in N(A^T)$ , so dim  $N(A^T) > 0$ , and thus  $rank(A) = rank(A^T) < 3$ . So  $\lambda_1 = 0$ .

We can easily spot  $\lambda_2 = -1$  as another eigenvalue, since subtracting A + I has two equal columns, and hence  $\det(A + I) = 0$ .

Looking at the trace we get that  $\lambda_3 = tr(A) - \lambda_1 - \lambda_2 = -2$ .

- (b) If you solve  $\frac{du}{dt} = Au$ , is (1) or (2) or (3) true as  $t \to \infty$ ?
  - (1) u(t) goes to zero?
  - (2) u(t) approaches a multiple of (what vector?)
  - (3) u(t) blows up?

Solution: Approaches a multiple of  $\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$ .

Observe that  $\begin{bmatrix} 1\\2\\1 \end{bmatrix}$  is an eigenvector for the 0 eigenvalue, and that the only nonzero eigenvalue of  $e^{At}$  as  $t\to\infty$  is  $e^{0\cdot t}=1$ , with the same eigenvector.

Also,  $u(t) = e^{At}u(0)$ , and that the only nonzero eigenvalue of  $e^{At}$  (as  $t \to \infty$ ) is 1, with the same eigenvector. So  $\lim_{t\to\infty} u(t)$  is a projection of u(0) on the line spanned by

 $\begin{bmatrix} 1 \\ 2 \\ 1 \end{bmatrix}$ , hence a multiple of it.

- 7. (10 points) Every invertible matrix A equals an orthogonal matrix Q times a positive definite matrix S. This famous fact comes directly from the SVD for the square matrix  $A = U\Sigma V^T$ , by choosing  $Q = UV^T$ .
- (a) How can you prove that  $Q = UV^T$  is orthogonal?

Solution: Q is orthogonal if and only if  $Q^TQ = I$ . Notice that since A is invertible, U and V are both square matrices of full rank.

$$Q^{T}Q = (UV^{T})^{T}(UV^{T}) = VU^{T}UV^{T} = V(U^{T}U)V^{T} = VV^{T} = I$$

We used the fact that U is orthogonal, hence  $U^TU = I$ , and that  $V^T$  is orthogonal because V's columns are eigenvectors of a symmetric matrix  $A^TA$ , so  $V^T = V^{-1}$ .

(b) Substitute  $Q^{-1}$  and A to write  $S = Q^{-1}A$  in terms of U, V and  $\Sigma$ . How can you tell that this matrix S is symmetric positive definite?

Solution:

$$S = Q^{-1}A = (UV^T)^{-1}A = (V^T)^{-1}U^{-1}U\Sigma V^T = V\Sigma V^T = (\Sigma^{1/2}V^T)^T(\Sigma^{1/2}V^T)$$

8. (15 points) A 4-node graph has all six possible edges. Its incidence matrix A and its Laplacian matrix  $A^TA$  are

$$A = \begin{bmatrix} -1 & 1 & 0 & 0 \\ -1 & 0 & 1 & 0 \\ -1 & 0 & 0 & 1 \\ 0 & -1 & 1 & 0 \\ 0 & -1 & 0 & 1 \\ 0 & 0 & -1 & 1 \end{bmatrix} \qquad A^{T}A = \begin{bmatrix} 3 & -1 & -1 & -1 \\ -1 & 3 & -1 & -1 \\ -1 & -1 & 3 & -1 \\ -1 & -1 & -1 & 3 \end{bmatrix}$$

(a) Describe the null space of A.

Solution:  $N(A) = span \left\{ \begin{bmatrix} 1\\1\\1\\1 \end{bmatrix} \right\}$  (Recall that applying A to a vector of potentials gives

the potential drops along edges, so in order for a vector of potentials to be in the null space, all the potentials within one connected component must be the same.)

(b) The all-ones matrix B = ones(4) has what eigenvalues? Then what are the eigenvalues of  $A^T A = 4I - B$ ?

Solution:  $B = \vec{1} \cdot \vec{1}^T = 4(\vec{1}/2)(\vec{1}/2)^T$ , where  $\vec{1}$  is the all-ones vector in  $\mathbb{R}^4$ . So B has eigenvalues 4,0,0,0.

I and B diagonalize in the same eigenbasis, so  $\lambda_i(4I - B) = \lambda_i(4I) - \lambda_i(B) = 4\lambda_i(I) - \lambda_i(B)$  for all i. So the eigenvalues of  $A^TA$  are 0, 4, 4, 4.

(c) For the Singular Value Decomposition  $A = U\Sigma V^T$ , can you find the nonzero entries in the diagonal matrix  $\Sigma$  and one column of the orthogonal matrix V?

Solution:  $\sigma_i = \sqrt{\lambda_i(A^T A)}$ , so the nonzero singular values are 2, 2, 2. We only need to find one eigenvector of  $A^T A$ . An obvious one is  $\vec{1}/2$ , since all the rows sum up to 0.