

1. (8 pts) Subspaces of R^n are generally described either as the column space of some given matrix, or the null space of some given matrix.

a) If we define a subspace V as the nullspace of the matrix $A = \begin{bmatrix} -1 & 1 & -2 & 1 \\ 1 & -1 & 1 & -2 \end{bmatrix}$

express V instead as the column space of some matrix B . Explain how/why this works in general.

Find the special solutions of $A\bar{x} = \bar{0}$ and put them in a matrix N :

$$\begin{bmatrix} -1 & 1 & -2 & 1 \\ 1 & -1 & 1 & -2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & -1 & 0 & -3 \\ 0 & 0 & 1 & 1 \end{bmatrix}, N = \begin{bmatrix} 1 & 3 \\ 1 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix}. \text{ Every vector in the null}$$

space is a linear combination of the special solutions, (the special solutions are a basis for the null space). This means that the column space of N is V . So our matrix B is the matrix N of special solutions of $A\bar{x} = \bar{0}$.

b) If we define a subspace V as the column space of the matrix $A = \begin{bmatrix} -1 & 1 \\ 1 & -1 \\ -2 & 1 \\ 1 & -2 \end{bmatrix}$ express V

instead as the null space of some matrix B . Explain how/why this works in general.

The column space of A is the orthogonal complement of the left null space of A , which is the null space of A^T . The null space of A^T comes from the solutions of $A^T\bar{x} = \bar{0}$, which, conveniently enough, were calculated above as the vectors in the column space of

$$N = \begin{bmatrix} 1 & 3 \\ 1 & 0 \\ 0 & -1 \\ 0 & 1 \end{bmatrix}. \text{ Now the orthogonal complement of the column space of } N \text{ can be}$$

obtained as the null space of $N^T = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 3 & 0 & -1 & 1 \end{bmatrix}$. So V is the null space of

$$N^T = \begin{bmatrix} 1 & 1 & 0 & 0 \\ 3 & 0 & -1 & 1 \end{bmatrix}. \text{ In general we can say the the column space of } A \text{ can be}$$

characterized as the null space of the matrix whose rows are composed of a basis of the left null space of A .

2. (8 pts) Find an orthonormal basis for the subspace V spanned by

$$\begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \begin{bmatrix} 2 \\ 0 \\ -1 \\ 0 \end{bmatrix}, \begin{bmatrix} 3 \\ 0 \\ 0 \\ -1 \end{bmatrix}$$

This is just standard Gram-Schmidt.

$$\begin{aligned} \bar{u}_1 &= \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \bar{u}_2 = \bar{v}_2 - \frac{\bar{u}_1 \cdot \bar{v}_2}{\bar{u}_1 \cdot \bar{u}_1} \bar{u}_1 = \begin{bmatrix} 2 \\ 0 \\ -1 \\ 0 \end{bmatrix} - \frac{2}{2} \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix} = \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix} \\ \bar{u}_3 &= \bar{v}_3 - \frac{\bar{u}_1 \cdot \bar{v}_3}{\bar{u}_1 \cdot \bar{u}_1} \bar{u}_1 - \frac{\bar{u}_2 \cdot \bar{v}_3}{\bar{u}_2 \cdot \bar{u}_2} \bar{u}_2 = \begin{bmatrix} 3 \\ 0 \\ 0 \\ -1 \end{bmatrix} - \frac{3}{2} \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix} - \frac{3}{3} \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix} = \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ 1 \\ -1 \end{bmatrix} \end{aligned}$$

(check your calculations by confirming orthogonality - it works)

Finally, normalize each vector:

$$\bar{q}_1 = \frac{1}{\sqrt{2}} \begin{bmatrix} 1 \\ -1 \\ 0 \\ 0 \end{bmatrix}, \bar{q}_2 = \frac{1}{\sqrt{3}} \begin{bmatrix} 1 \\ 1 \\ -1 \\ 0 \end{bmatrix}, \bar{q}_3 = \frac{1}{\sqrt{5/2}} \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \\ 1 \\ -1 \end{bmatrix}$$

3. (5 points) The normal equations $A^T(\bar{b} - A\bar{x}) = \bar{0}$ arise how/why in calculating the projection \bar{p} of a vector \bar{b} onto the column space of A . What property do the normal equations express?

The projection \bar{p} must have the form $A\bar{x}$ because it is in the column space of A . It is characterized by the requirement that the error vector $\bar{e} = \bar{b} - \bar{p}$ is orthogonal to the column space. This is expressed by the equation $A^T(\bar{b} - A\bar{x}) = \bar{0}$ which says that $\bar{b} - A\bar{x}$ is in the left null space of A , $N(A^T)$, which is the orthogonal complement of the column space.

4. (8 pts) Use the (amazing) cofactor matrix to calculate the inverse of $\begin{bmatrix} -1 & 1 & 1 \\ 2 & 2 & 1 \\ 1 & 1 & 1 \end{bmatrix}$. If

$$A = \begin{bmatrix} a_{11} & a_{12} & \cdots & a_{1n} \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix} \text{ is an } n \times n \text{ matrix, and } C = \begin{bmatrix} c_{11} & c_{12} & \cdots & c_{1n} \\ c_{21} & c_{22} & \cdots & c_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ c_{n1} & c_{n2} & \cdots & c_{nn} \end{bmatrix} \text{ is its matrix of}$$

cofactors, then $\det \begin{bmatrix} x_1 & x_2 & \cdots & x_n \\ a_{21} & a_{22} & \cdots & a_{2n} \\ \vdots & \vdots & \vdots & \vdots \\ a_{n1} & a_{n2} & \cdots & a_{nn} \end{bmatrix}$ can be expressed as the dot product of $\underline{\bar{x}}$ with $\underline{\text{row 1 of } C}$.

The cofactor matrix is $C = \begin{bmatrix} 1 & -1 & 0 \\ 0 & -2 & 2 \\ -1 & 3 & -4 \end{bmatrix}$. The determinant of A can be calculated

directly, or by noting that $\det A$ is the dot product of any row/column of C with the corresponding row/column of A . In any case $\det A = -2$ and

$$A^{-1} = \frac{1}{\det A} C^T = \frac{1}{(-2)} \begin{bmatrix} 1 & 0 & -1 \\ -1 & -2 & 3 \\ 0 & 2 & -4 \end{bmatrix} = \begin{bmatrix} -\frac{1}{2} & 0 & \frac{1}{2} \\ \frac{1}{2} & 1 & -\frac{3}{2} \\ 0 & -1 & 2 \end{bmatrix}$$

As for the second part, expanding by the first row of the matrix with the x'_i s, the cofactors appear in the first row of the matrix C so we get the determinant from the dot product of \bar{x} with row 1 of C .

5. (10 pts) Complete the following by filling in the blanks:

If $A\bar{x} = \bar{b}$ then \bar{b} is in the column space of A .

If \bar{y} is in the orthogonal complement of the column space of A , then \bar{y} is in the left null space of A or the null space of A^T .

If \bar{x} is in the row space of A , then \bar{x} is in the orthogonal complement of the null space of A .

If V is a subspace of R^n of dimension k , then V^\perp has dimension $n - k$. A basis for V^\perp can be calculated from a basis of V as the special solutions of $A\bar{x} = \bar{0}$ where A contains the basis of V as its rows and the rank of A is k .

6. (5 points) If $A\bar{x} = \bar{b}$ has a solution, then the solution \bar{x} of shortest length is in the row space of A . Why is that true? Explain.

Any solution \bar{x} , as a vector in R^n , can be expressed as $\bar{x} = \bar{x}_{\text{row}} + \bar{x}_{\text{null}}$. Now $A\bar{x} = A\bar{x}_{\text{row}} + A\bar{x}_{\text{null}} = A\bar{x}_{\text{row}}$ so \bar{x}_{row} is a solution by itself. Any solution must be of the form $\bar{x} = \bar{x}_{\text{row}} + \bar{n}$ where \bar{n} is in $N(A)$. However, note that $\|\bar{x}\|^2 = \|\bar{x}_{\text{row}}\|^2 + \|\bar{n}\|^2$ which is minimum when $\bar{n} = \bar{0}$ so that $\bar{x} = \bar{x}_{\text{row}}$ gives the shortest length.

7. (10 points) Find the values of a and b so that the vector $\bar{p} = a \begin{bmatrix} -1 \\ 1 \\ 1 \\ -1 \end{bmatrix} + b \begin{bmatrix} 1 \\ 1 \\ 1 \\ -1 \end{bmatrix}$ is as

close as possible to the vector $\bar{b} = \begin{bmatrix} 1 \\ 1 \\ 1 \\ 1 \end{bmatrix}$. How would you describe the vector \bar{p} using the idea of projection?

This is a projection problem, $\bar{p} = \begin{bmatrix} -1 & 1 \\ 1 & 1 \\ 1 & 1 \\ -1 & -1 \end{bmatrix} \bar{x}$ where $\bar{x} = \begin{bmatrix} a \\ b \end{bmatrix}$ so \bar{p} is a vector in the

column space of $A = \begin{bmatrix} -1 & 1 \\ 1 & 1 \\ 1 & 1 \\ -1 & -1 \end{bmatrix}$ and the \bar{p} closest to \bar{b} is the projection of \bar{b} onto the

column space of A . We calculate from the normal equations

$A^T A \bar{x} = A^T \bar{b}$, $\begin{bmatrix} 4 & 2 \\ 2 & 4 \end{bmatrix} \bar{x} = \begin{bmatrix} 0 \\ 2 \end{bmatrix}$. We can solve by elimination:

$$\begin{bmatrix} 4 & 2 & 0 \\ 2 & 4 & 2 \end{bmatrix} \rightarrow \begin{bmatrix} 1 & 0 & -\frac{1}{3} \\ 0 & 1 & \frac{2}{3} \end{bmatrix} \text{ so } \bar{x} = \begin{bmatrix} a \\ b \end{bmatrix} = \begin{bmatrix} -\frac{1}{3} \\ \frac{2}{3} \end{bmatrix}$$