

Math 4242 Sec 40

Quiz 7 Solutions

Name: _____

1. (3 points) For $A = \begin{pmatrix} 1 & 1 & 1 & 3 \\ 2 & 0 & -2 & 6 \\ -2 & 2 & 2 & -10 \\ 0 & -2 & 0 & 4 \end{pmatrix}$, is the set of columns of A linearly independent? Show that your answer is correct.

Answer: No: We reduce A to row echelon form to obtain the matrix

$$\begin{pmatrix} 1 & 1 & 1 & 3 \\ 0 & -2 & -4 & 0 \\ 0 & 0 & -4 & -4 \\ 0 & 0 & 0 & 0 \end{pmatrix}.$$

This matrix has a non-basic column, so $N(A) \neq \{\vec{0}\}$. Hence, the set of columns of A is linearly dependent.

2. (3 points) For A as in problem 1, give two different bases B_1, B_2 for $\mathcal{R}(A)$, the range of A . Explain why these are in fact bases.

Hint: you may find it helpful to use fact that performing row operations on a matrix preserves any linear dependencies amongst the columns; this was shown in Meyer Chapter 3 (top of page 136), and discussed in class a while ago.

Answer: Usually, a vector space has many bases, and $\mathcal{R}(A)$ is no exception. So there are many correct answers to this problem. Here's one answer:

As explained in Meyer, the basic columns of A form a basis for $\mathcal{R}(A)$. Hence, by problem 1), $S = \{A_{*1}, A_{*2}, A_{*3}\}$ is a basis for $\mathcal{R}(A)$.

Another basis for $\mathcal{R}(A)$ is given, for example, by $S' = \{2A_{*1}, A_{*2}, A_{*3}\}$.

$$spn(S') = \text{Span}(S) = \mathcal{R}(A)$$

because $S' \subset \text{Span}(S)$ and $S \subset \text{Span}(S')$. Furthermore S' is linearly independent because if

$$r_1(2A_{*1}) + r_2A_{*2} + r_3A_{*3} = (2r_1)A_{*1} + r_2A_{*2} + r_3A_{*3} = \vec{0},$$

then by the linear independence of S , $r_1 = r_2 = r_3 = 0$.

For a slightly more interesting answer, yet another basis is given by $S'' = \{A_{*1}, A_{*2}, A_{*4}\}$. By considering the reduced echelon form of A , and appealing to the hint, we can see that S'' is linearly independent and that $\text{Span}(S'') = \text{Span}(S')$, so S'' is in fact a basis.

IMPORTANT NOTE: It is usually not true that for B a row echelon form of A , the basic columns of B themselves form a basis for $\mathcal{R}(A)$: Usually $\mathcal{R}(\mathcal{A})$ and $\mathcal{R}(\mathcal{B})$ will be different. (Can you find a simple example which demonstrates this?)

3. (2 points) For $S = \left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 0 \\ 1 \end{pmatrix}, \begin{pmatrix} 3 \\ 0 \end{pmatrix}, \begin{pmatrix} -1 \\ 1 \end{pmatrix} \right\}$, write down all subsets of S which are bases for \mathbb{R}^2 .

Answer:

$\dim \mathbb{R}^2 = 2$, so by Proposition 3.1 from (the revised version of) my notes, any linearly independent subset $Q \subset S$ with $|Q| = 2$ is a basis for \mathbb{R}^2 .

Hence every 2-element subset of S except $\left\{ \begin{pmatrix} 1 \\ 0 \end{pmatrix}, \begin{pmatrix} 3 \\ 0 \end{pmatrix} \right\}$ is a basis for \mathbb{R}^2 .

4. (2 points) Let $\mathcal{B} = \{b_1, b_2, \dots, b_n\}$ be a basis for a vector space V . Prove that each $v \in V$ can be expressed as a linear combination of the b_i 's:

$$v = \alpha_1 b_1 + \alpha_2 b_2 + \cdots + \alpha_n b_n$$

in only one way—i.e., the coordinates α_i are unique.

Answer: If

$$v = \alpha_1 b_1 + \alpha_2 b_2 + \cdots + \alpha_n b_n = \beta_1 b_1 + \beta_2 b_2 + \cdots + \beta_n b_n$$

then

$$(\alpha_1 - \beta_1)b_1 + (\alpha_2 - \beta_2)b_2 + \cdots + (\alpha_n - \beta_n)b_n = \vec{0}.$$

By the linear independence of B , $\alpha_i - \beta_i = 0$ for each i , so $\alpha_i = \beta_i$ for each i . This gives the result.

5. **Bonus** (2 points) : Note that in view of the result you proved in 4, for any finite basis $\mathcal{B} = \{b_1, b_2, \dots, b_n\}$ of a vector space V , we obtain a well defined function $L_{\mathcal{B}} : V \rightarrow \mathbb{R}^n$, defined by

$$L_{\mathcal{B}}(\alpha_1 b_1 + \alpha_2 b_2 + \cdots + \alpha_n b_n) = (\alpha_1, \alpha_2, \dots, \alpha_n)^T.$$

New terminology: We call an invertible linear map an **isomorphism**.

Prove that L_B is an isomorphism.

Answer: To show that L_B is linear for any $v, v' \in V$, write

$$\begin{aligned} v &= \alpha_1 b_1 + \cdots + \alpha_n b_n, \\ v' &= \alpha'_1 b_1 + \cdots + \alpha'_n b_n. \end{aligned}$$

$$\begin{aligned} L_B(v + v') &= L_B((\alpha_1 + \alpha'_1)b_1 + \cdots + (\alpha_n + \alpha'_n)b_n) \\ &= ((\alpha_1 + \alpha'_1), \dots, (\alpha_n + \alpha'_n)) \\ &= (\alpha_1, \dots, \alpha'_n) + (\alpha'_1, \dots, \alpha'_n) \\ &= L_B(v) + L_B(v'). \end{aligned}$$

A similar argument shows that $L_B(cv) = cL_B(v)$ for all scalars c . Hence L_B is linear.

Further, L_B is invertible: $L_B^{-1} : \mathbb{R}^n \rightarrow V$ is given by

$$L_B^{-1}(\alpha_1, \dots, \alpha_n) = \alpha_1 b_1 + \alpha_2 b_2 + \cdots + \alpha_n b_n.$$

It's clear that $L_B \circ L_B^{-1}(v) = \text{Id}_{\mathbb{R}^n}$ and $L_B^{-1} \circ L_B = \text{Id}_V$, so that these functions are in fact inverses.