Discretionary Note

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Math 226: HW 8

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- 1. a) We can give $R(UT) = \{U(w) : w \in R(T)\}$ and $R(U) = \{U(w) : w \in W\}$. Because $R(T) \subseteq W$, $R(UT) \subseteq R(U)$. This means that all of the vectors that form a basis of R(UT) are in R(U). Thus, the number of vectors that form a basis of R(UT) is at least the number of vectors that form a basis of $R(U) \Rightarrow dim(R(UT)) \leq dim(R(U))$.
 - b) As given by Theorem 3.5, for any given $D \in M_{m \times n}(\mathbb{F})$, $rank(D) = rank(L_D)$. Thus, given two linear transformations L_A and L_B we get:

$$dim(R(L_AL_B)) \le dim(R(L_A))$$

 $rank(L_{AB}) \le rank(L_A)$
 $rank(AB) \le rank(A)$

The first statement $dim(R(L_AL_B)) \leq dim(R(L_A))$ was proved in part (a).

- c) We can give $R(UT) = \{U(w) : w \in R(T)\}$ and $R(T) = \{T(v) : v \in V\}$. Let us inspect the case in which U is injective and is not injective.
 - (1) U is injective

If U is injective, every pre-image in R(T) maps to a *unique* image in Z through U. Let us define the basis for R(T) as $\beta_{R(T)}$. We consider the two cases where U is bijective and not bijective (i.e. U is not surjective).

(i) U is bijective

As proved in Question 2 of Homework 4, in the case U is bijective, the set $U(\beta_{R(T)}) = \{U(w) : w \in \beta_{R(T)}\}$ acts as a basis for $R(UT) \Rightarrow |U(\beta_{R(T)})| = |\beta_{R(T)}| \Rightarrow dim(R(UT)) = dim(R(T))$.

(ii) U is not surjective

For proof by contradiction, let us assume dim(R(UT)) > dim(R(T)). Let us define dim(R(UT)) = b and dim(R(T)) = a and b > a. This means that there are b linearly independent vectors in R(UT), which are all formed from U applied to b pre-images in R(T). These b pre-images in R(T) are linearly independent a and so this means that $dim(R(T)) = b \neq a$. Thus, by proof by contradiction, we have proven $dim(R(UT)) \leq dim(R(T))$.

Thus, we can see in both cases $dim(R(UT)) \leq dim(R(T))$.

 $(\mathbf{2})$ U is not injective

Let us define a new linear map $U': R(T) \to R(UT)$ that applies function U to all elements in R(T). From Dimension Theorem, we know that dim(R(T)) = nullity(U') + rank(U'). Because U is surjective by definition, R(U') = R(UT) and so rank(U') = dim(R(U')) = dim(R(UT)). Thus, we get that dim(R(T)) = dim(N(U')) + dim(R(UT)). Because $dim(N(U')) \ge 0$ by definition, $dim(R(T)) \ge dim(R(UT)) \Rightarrow dim(R(UT)) \le dim(R(T))$.

Thus, we can see in both cases $dim(R(UT)) \leq dim(R(T))$.

d) We apply Theorem 3.5 in a similar manner as in part (b). Let us suppose we have two linear transformations L_A and L_B .

¹Let us define these b linearly independent vectors in R(UT) as $\{z_1, \ldots, z_b\}$. Because each of these vectors has a pre-image in R(T), we can represent these b linearly independent vectors as $\{U(w_1), \ldots, U(w_b)\}$ where $w_i \in R(T)$. Because these vectors are linearly independent, the solution to $\sum_{i=1}^b a_i z_i = 0$ is for all $a_i = 0$. We can re-express this equation as $U(\sum_{i=1}^b a_i w_i) = 0$. Because $N(U) = \{0\}$ as U is injective, the solution to $\sum_{i=1}^b a_i w_i = 0$ is all $a_i = 0 \Rightarrow$ these b pre-images in R(T) given by $\{w_1, \ldots, w_b\}$ are linearly independent.

$$dim[R(L_AL_B)] \le dim[R(L_B)]$$

 $rank(L_{AB}) \le rank(L_B)$
 $rank(AB) \le rank(B)$

This first statement $dim[R(L_AL_B)] \leq dim[R(L_B)]$ is proven by part (c).

- 2. a) We prove both directions of this if and only if statement below.
 - (1) If $rank(A) \neq rank([A'|b'])$, [A'|b'] contains a row in which the only nonzero entry is in the last column. For proof by contrapositive, let us assume that [A'|b'] does not contain any rows in which the only nonzero entry is in the last column. This means that a valid solution to the equation Ax = b exists $\Rightarrow b \in R(L_A) \Rightarrow b$ is in the span of the columns of $A \Rightarrow rank(A) = rank([A|b])$. Because rank([A|b]) = rank([A'|b']), this means that rank(A) = rank([A'|b']). Thus, by proof by contrapositive, we have proved that if $rank(A) \neq rank([A'|b'])$, [A'|b'] contains a row in which the only nonzero entry is in the last column.
 - (2) If [A'|b'] contains a row in which the only nonzero entry is in the last column, $rank(A) \neq rank([A'|b'])$.

 If there is a row in which the only nonzero entry is in the last column, this means that there is no solution to the equation Ax = b as zero multiplied by each index of x cannot be nonzero. This means that $b \notin R(L_A) \Rightarrow b$ is not in the span of the columns of $A \Rightarrow b$ is linearly independent from the columns of $A \Rightarrow rank(A) \neq rank([A|b])$. Because rank([A|b]) = rank([A'|b']), this means $rank(A) \neq rank([A'|b'])$ if there is a row in which the only nonzero entry is in the last column.
 - b) If the equation Ax = b is consistent, this means that there is a solution to this system. This means that $b \in R(L_A) \Rightarrow b \in$ the span of the columns of $A \Rightarrow rank(A) = rank([A|b]) \Rightarrow rank(A) = rank([A'|b'])$. Thus, this statement is essentially: $rank(A) = rank([A'|b']) \Leftrightarrow [A'|b']$ contains no row in which the only nonzero entry is in the last column. This statement is the negative (on both sides of the if and only if) of the statement proved in part (a). Thus, this statement is true.
- 3. Answer on paper. Please go to the end of this PDF.
- 4. a) The area of a parallelogram is given by the product of the base and the height. The length of the base can be given by $||u_{\theta}||$ and the height can be given as $||v_{\theta}|| sin(\theta)$. Thus, the area A is given as $||u_{\theta}|| ||v_{\theta}|| sin(\theta)$, which we compute below:

$$A = \|u_{\theta}\| \|v_{\theta}\| \sin(\theta) = \|u_{\theta}\| \|v_{\theta}\| \sqrt{1 - (\cos(\theta))^{2}} = \|u_{\theta}\| \|v_{\theta}\| \sqrt{1 - (\frac{u_{\theta} \cdot v_{\theta}}{\|u_{\theta}\| \|v_{\theta}\|})^{2}} =$$

$$\|u_{\theta}\| \|v_{\theta}\| \sqrt{\frac{\|u_{\theta}\|^{2} \|v_{\theta}\|^{2} - (u_{\theta} \cdot v_{\theta})^{2}}{(\|u_{\theta}\| \|v_{\theta}\|)^{2}}} = \sqrt{\|u_{\theta}\|^{2} \|v_{\theta}\|^{2} - (u_{\theta} \cdot v_{\theta})^{2}} =$$

$$\sqrt{(u_{1}^{2} + u_{2}^{2})(v_{1}^{2} + v_{2}^{2}) - (u_{1}v_{1} + u_{2}v_{2})^{2}} = \sqrt{u_{1}^{2}v_{2}^{2} - 2u_{1}v_{1}u_{2}v_{2} + u_{2}^{2}v_{1}^{2}} = \sqrt{(u_{1}v_{2} - u_{2}v_{1})^{2}} =$$

$$u_{1}v_{2} - u_{2}v_{1}$$

Thus, we have computed $A = u_1v_2 - u_2v_1 \ge 0$ due to the square root function. We now show that $A = |det[u_\theta, v_\theta]|$. The determinant of a 2 × 2 matrix is ad - bc and

the matrix $[u_{\theta}, v_{\theta}]$ is given by:

$$\begin{bmatrix} u_1 & v_1 \\ u_2 & v_2 \end{bmatrix}$$

Thus, the determinant of this matrix is given by $u_1v_2 - u_2v_1$. As we saw before, this value is ≥ 0 and so the absolute value of this determinant is $u_1v_2 - u_2v_1 = A$.

b) If $A_{-\theta}(v) = v_{\theta}$, this means that $v_{\theta} = (v_1 cos(\theta) + v_2 sin(\theta), -v_1 sin(\theta) + v_2 cos(\theta))$. Similarly, if $A_{-\theta}(u) = u_{\theta}$, then $u_{\theta} = (u_1 cos(\theta) + u_2 sin(\theta), -u_1 sin(\theta) + u_2 cos(\theta))$. We compute $A_{-\theta}[u, v]$ below:

$$A_{-\theta}[u,v] = \begin{bmatrix} \cos(\theta) & \sin(\theta) \\ -\sin(\theta) & \cos(\theta) \end{bmatrix} \begin{bmatrix} u_1 & v_1 \\ u_2 & v_2 \end{bmatrix} = \begin{bmatrix} u_1 \cos(\theta) + u_2 \sin(\theta) & v_1 \cos(\theta) + v_2 \sin(\theta) \\ -u_1 \sin(\theta) + u_2 \cos(\theta) & -v_1 \sin(\theta) + v_2 \cos(\theta) \end{bmatrix} = \begin{bmatrix} u_\theta & v_\theta \end{bmatrix}$$

where u_{θ} and v_{θ} act as column vectors.

c) By Theorem 4.7, we have $det(A_{-\theta}[u,v]) = det(A_{-\theta})det([u,v])$. As shown in (b), $A_{\theta}[u,v] = [u_{\theta},v_{\theta}]$ and so this simplifies to:

$$det([u_{\theta}, v_{\theta}]) = det(A_{-\theta})det([u, v]).$$

As shown in part (a), $det([u_{\theta}, v_{\theta}]) = A$. We can also compute $det(A_{-\theta}) = cos^2(\theta) + sin^2(\theta) = 1$. Thus, we get:

$$A = 1 * det([u, v])$$
$$A = det([u, v])$$

Furthermore, in part (a) we have shown that $A \ge 0$. Thus, |det([u, v])| = |A| = A.

- 5. We prove both directions of this if and only if statement below.
 - (1) If δ has the form $\delta(A) = \alpha a_{11} a_{22} + B a_{11} a_{21} + C a_{12} a_{22} + D a_{12} a_{21}$ where $\alpha, B, C, D \in \mathbb{F}$, δ is a 2-linear functional.

We prove that δ is a 2-linear functional by proving, without loss of generality, that δ is a 2-linear functional if it's second row is a+cb. Note that this proof can be done identically for the first row. We define $c \in \mathbb{F}$.

$$\delta(\begin{bmatrix} a_{11} & a_{12} \\ a_{21} + cb_{21} & a_{22} + cb_{22} \end{bmatrix}) = \delta(\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}) + c\delta(\begin{bmatrix} a_{11} & a_{12} \\ b_{21} & b_{22} \end{bmatrix})$$

The LHS of this equation can be given as $\alpha a_{11}(a_{22} + cb_{22}) + Ba_{11}(a_{21} + cb_{21}) + Ca_{12}(a_{22} + cb_{22}) + Da_{12}(a_{21} + cb_{21})$. We now compute the RHS:

$$\delta(\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}) + c\delta(\begin{bmatrix} a_{11} & a_{12} \\ b_{21} & b_{22} \end{bmatrix}) = \alpha a_{11} a_{22} + B a_{11} a_{21} + C a_{21} a_{22} + D a_{12} a_{21} a_{21} + c(a_{21} a_{22} + a_{22} a_{21} a_{22} + a_{22} a_{$$

Thus, we have proven that LHS = RHS and so we have proven that if δ has this form, it is a 2-linear functional.

(2) If δ is a 2-linear functional, it has the form $\delta(A) = \alpha a_{11} a_{22} + B a_{11} a_{21} + C a_{12} a_{22} + D a_{12} a_{21}$ where $\alpha, B, C, D \in \mathbb{F}$

To prove this statement, we inspect the value of $\delta(A)$, where $A \in M_{2\times 2}(\mathbb{F})$ and δ is a 2-linear functional:

$$\delta(\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}) = \delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ a_{21}e_1 + a_{22}e_2 \end{bmatrix}) = \delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ a_{21}e_1 \end{bmatrix}) + a_{22}\delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ e_2 \end{bmatrix})$$

We refer to the first and second element in this summation as (i) and (ii), respectively. We compute (i) below²:

$$\delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ a_{21}e_1 \end{bmatrix}) = \delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ 0 \end{bmatrix}) + a_{21}\delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ e_1 \end{bmatrix})$$

$$= \delta(\begin{bmatrix} a_{11}e_1 \\ 0 \end{bmatrix}) + a_{12}\delta(\begin{bmatrix} e_2 \\ 0 \end{bmatrix}) + a_{21}\delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ e_1 \end{bmatrix})$$

$$= \delta(\begin{bmatrix} 0 \\ 0 \end{bmatrix}) + a_{11}\delta(\begin{bmatrix} e_1 \\ 0 \end{bmatrix}) + a_{12}\delta(\begin{bmatrix} e_2 \\ 0 \end{bmatrix}) + a_{21}\delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ e_1 \end{bmatrix})$$

$$= a_{11}\delta(\begin{bmatrix} e_1 \\ 0 \end{bmatrix}) + a_{12}\delta(\begin{bmatrix} e_2 \\ 0 \end{bmatrix}) + a_{21}\delta(\begin{bmatrix} a_{11}e_1 \\ e_1 \end{bmatrix}) + a_{21}a_{12}\delta(\begin{bmatrix} e_2 \\ e_1 \end{bmatrix})$$

$$= a_{11}\delta(\begin{bmatrix} e_1 \\ 0 \end{bmatrix}) + a_{12}\delta(\begin{bmatrix} e_2 \\ 0 \end{bmatrix}) + a_{21}\delta(\begin{bmatrix} e_1 \\ 0 \end{bmatrix}) + a_{21}a_{12}\delta(\begin{bmatrix} e_2 \\ e_1 \end{bmatrix})$$

Note that for any $i \in \mathbb{R}$, $\delta([e_i, 0]^T) = 0$. We show this below (for any vector \vec{k}):

²Note that because δ is a linear functional, $\delta(\mathbf{0}) = 0$. This is proven by the fact that, for any vector \vec{k} , $\delta(\mathbf{0}) = \delta([0, \vec{k} - \vec{k}]^T) = \delta([0, \vec{k}]^T) - \delta([0, \vec{k}]^T) = 0$.

$$\delta([e_i, \vec{k}]^T) = \delta([e_i, 0]^T) + \delta([e_i, \vec{k}]^T)$$
$$\delta([e_i, 0]^T) = 0$$

Similarly, for any $i \in \mathbb{R}$, $\delta([0, e_i]^T) = 0$. We show this for any vector \vec{k} :

$$\delta([\vec{k}, e_i]^T) = \delta([0, e_i]^T) + \delta([\vec{k}, e_i]^T)$$
$$\delta([0, e_i]^T) = 0$$

Because we cannot express $\delta([e_1, e_1]^T)$ and $\delta([e_2, e_1]^T)$ any further, we define constants $B = \delta([e_1, e_1]^T)$ and $D = \delta([e_2, e_1]^T)$. Thus, we can express (\mathbf{i}) as:

$$Ba_{11}a_{21} + Da_{12}a_{21}$$

We now compute (ii) as:

$$\begin{aligned} a_{22}\delta(\begin{bmatrix} a_{11}e_1 + a_{12}e_2 \\ e_2 \end{bmatrix}) &= a_{22}[\delta(\begin{bmatrix} a_{11}e_1 \\ e_2 \end{bmatrix}) + a_{12}\delta(\begin{bmatrix} e_2 \\ e_2 \end{bmatrix})] = a_{22}\delta(\begin{bmatrix} a_{11}e_1 \\ e_2 \end{bmatrix}) + a_{22}a_{12}\delta(\begin{bmatrix} e_2 \\ e_2 \end{bmatrix}) \\ &= a_{22}\delta(\begin{bmatrix} 0 \\ e_2 \end{bmatrix}) + a_{22}a_{11}\delta(\begin{bmatrix} e_1 \\ e_2 \end{bmatrix}) + a_{22}a_{12}\delta(\begin{bmatrix} e_2 \\ e_2 \end{bmatrix}) \\ &= a_{22}a_{11}\delta(\begin{bmatrix} e_1 \\ e_2 \end{bmatrix}) + a_{22}a_{12}\delta(\begin{bmatrix} e_2 \\ e_2 \end{bmatrix}) \end{aligned}$$

Because we cannot express $\delta([e_1, e_2]^T)$ and $\delta([e_2, e_2]^T)$ any further, we define constants $\alpha = \delta([e_1, e_2]^T)$ and $C = \delta([e_2, e_2]^T)$. Thus, we can express (ii) as:

$$\alpha a_{11}a_{22} + Ca_{12}a_{22}$$

Thus, we get that $\delta(A)$, where $A \in M_{2x2}(\mathbb{F})$, can be given by (i) + (ii), or:

$$\delta\left(\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}\right) = \alpha a_{11} a_{22} + C a_{12} a_{22} + B a_{11} a_{21} + D a_{12} a_{21}$$

$$\delta\left(\begin{bmatrix} a_{11} & a_{12} \\ a_{21} & a_{22} \end{bmatrix}\right) = \alpha a_{11} a_{22} + B a_{11} a_{21} + C a_{12} a_{22} + D a_{12} a_{21}$$

Thus we have proved that if δ is a 2-linear functional, it has the form $\delta(A) = \alpha a_{11} a_{22} + B a_{11} a_{21} + C a_{12} a_{22} + D a_{12} a_{21}$ where $\alpha, B, C, D \in \mathbb{F}$.





