# Math 54: Linear Algebra and Differential Equations

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### Applications of Rank-Nullity Theorem

So far, our examples have just been verifying the Rank-Nullity Theorem. Let us actually see why it is useful.

# Problem 1

Let S be the subspace of trace-free matrices in  $M_{n \times n}$  (matrices with trace equal to 0). What is the dimension of S?

Problem 1 demonstrates the following important result. A nonzero linear functional f on a finite dimensional vector space V (linear transformation from V to  $\mathbb{R}$ ) has ker(f) = dim(V)-1. (See Problem Set 6)

# Problem 2

Show that there is no one-to-one linear transformation  $T : \mathbb{R}^3 \to \mathbb{R}^2$ . Show that there is no onto linear transformation  $T : \mathbb{R}^2 \to \mathbb{R}^3$ .

Problem 2 demonstrates the following important result. A linear transformation from a vector space of larger dimension to a vector space of smaller dimension cannot be one-to-one. A linear transformation from a vector space of smaller dimension to a vector space of larger dimension cannot be onto. (The case of when the vector spaces both have the same size is covered on Problem Set 6.)

Finally, let's consider the rank and nullity of a **matrix transformation**. Recall that a **matrix transformation** is a linear transformation between Euclidean spaces that is defined by matrix multiplication. In particular, if A is an m by n matrix, then we can define the linear transformation

 $T: \mathbb{R}^n \to \mathbb{R}^m \quad T(\mathbf{x}) = A\mathbf{x}$ 

For matrix transformations:

• The kernel of T is called the nullspace of A, and it is the solutions to the homogeneous system  $A\mathbf{x} = 0$ . Writing the solution to the homogeneous system in parametric vector form immediately gives a basis for the kernel of T (equivalently the nullspace of A) and hence nullity(T).

• The range of T is called the column space of T, because it is the span of the columns of the matrix A. Putting the columns into the row of a matrix (transpose) and row reducing to reduced row echelon form immediately gives a basis for the column space of T (and hence rank(T)), which will be the nonzero rows of the reduced row echelon form matrix.

#### Problem 3

To illustrate the concepts above, find  $\operatorname{rank}(T) = \operatorname{nullspace}(A)$  and  $\ker(T) = \operatorname{column space}(A)$ for the matrix transformation  $T : \mathbb{R}^5 \to \mathbb{R}^3$  given by

$$T\left(\begin{bmatrix} x_1\\x_2\\x_3\\x_4\\x_5\end{bmatrix}\right) = \begin{bmatrix} 1 & 2 & 1 & 0 & 3\\1 & 1 & 1 & 1 & 4\\4 & 1 & 4 & 1 & 1\end{bmatrix} \begin{bmatrix} x_1\\x_2\\x_3\\x_4\\x_5\end{bmatrix}$$

where we are denoting A to be the 3 by 5 matrix above.

# **Coordinates for Vector Spaces**

Consider an abstract vector space V with dimension n. Fix an ordered basis  $\mathcal{B} = \{v_1, v_2, ..., v_n\}$  for V. Recall that by the spanning and linear independence properties of basis, this means that every vector v in V can be expressed *uniquely* as a linear combination of the basis vectors,

$$v = a_1 v_1 + a_2 v_2 + \dots + a_n v_n$$

Suppose that we have another vector w whose unique linear combination is

$$w = b_1 v_1 + b_2 v_2 + \ldots + b_n v_n$$

Then, note that

$$v + w = (a_1 + b_1)v_1 + (a_2 + b_2)v_2 + \dots + (a_n + b_n)v_n \quad cv = (ca_1)v_1 + (ca_2)v_2 + \dots + (ca_n)v_n$$

Notice that in essence, all vector operations can be done on the coefficients on each of the basis vectors. So if we fix a particular basis,  $\mathcal{B} = \{v_1, v_2, ..., v_n\}$ , we can just deal with the coefficients on the basis vectors in the linear combination and do our computations there. In particular, we can write the **coordinates with respect to**  $\mathcal{B}$  for v and w as

$$v = \begin{bmatrix} a_1 \\ a_2 \\ \vdots \\ a_n \end{bmatrix}_{\mathcal{B}} \quad w = \begin{bmatrix} b_1 \\ b_2 \\ \vdots \\ b_n \end{bmatrix}_{\mathcal{B}}$$

Note that these coordinates only make sense if we know which basis we are writing the coordinates with respect to, since any given abstract vector space has many different bases. In particular, vector spaces do not have preferred bases.

### Problem 4

Define the two bases  $\mathcal{B}_1 = \{1, x, x^2\}$  and  $\mathcal{B}_2 = \{1 - x + 2x^2, 1 + x + 3x^2, 2x + x^2\}$  for  $P_2$ . Calculate the coordinates of  $x^2 + x$  in both bases.

Why are coordinates useful? Well, in some sense, they tell us that every n dimensional vector space is basically like  $\mathbb{R}^n$ , once we fix a basis. Of course, the identification with  $\mathbb{R}^n$  is not **canonical**, meaning that there is no preferred basis in which we should interpret any given vector space. But once we agree on a basis and fix it, we can think of the vector space as just  $\mathbb{R}^n$ .

### The Matrix of a Linear Transformation

Let's motivate the idea of the matrix of a linear transformation with the following example. Suppose that  $T: V \to W$  is a linear transformation, where for concreteness,  $\dim(V) = 3$  and  $\dim(W) = 4$ . Let  $\mathcal{B} = \{v_1, v_2, v_3\}$  be an ordered basis for V and let  $\mathcal{C} = \{w_1, w_2, w_3, w_4\}$  be an ordered basis for W. Suppose that you know that

$$T(v_1) = 2w_1 + 3w_2 + w_3 - w_4$$
$$T(v_2) = -2w_1 + w_2 - w_3 - 2w_4$$
$$T(v_3) = w_1 + w_2 + 2w_3 + 3w_4$$

Then, what is T? Well, since  $\{v_1, v_2, v_3\}$  is a basis, we can write any vector in V in the form

$$v = a_1 v_1 + a_2 v_2 + a_3 v_3$$

So then, applying linearity, we get an explicit formula for T applied to every vector v!

$$T(a_1v_1 + a_2v_2 + a_3v_3) = (2a_1 - 2a_2 + a_3)w_1 + (3a_1 + a_2 + a_3)w_2 + (a_1 - a_2 + 2a_3)w_3 + (-a_1 - 2a_2 + 3a_3)w_4 + (-a_1 - 2a_3)w_4 + (-a_1 -$$

Writing this in coordinates, where we use the basis  $\mathcal{B}$  for V and the basis  $\mathcal{C}$  for W, we get that

$$T\left(\begin{bmatrix}a_{1}\\a_{2}\\a_{3}\end{bmatrix}_{\mathcal{B}}\right) = \begin{bmatrix}2a_{1}-2a_{2}+a_{3}\\3a_{1}+a_{2}+a_{3}\\a_{1}-a_{2}+2a_{3}\\-a_{1}-2a_{2}+3a_{3}\end{bmatrix}_{\mathcal{C}} = \begin{bmatrix}2&-2&1\\3&1&1\\1&-1&2\\-1&-2&3\end{bmatrix} \begin{bmatrix}a_{1}a_{2}a_{3}\end{bmatrix}_{\mathcal{B}}$$

This is amazing! Why? Because even though we have an abstract linear transformation, in some sense, our abstract linear transformation is given by a matrix transformation once we fix bases for V and W! The matrix

$$[T]_{\mathcal{B}\to\mathcal{C}} = \begin{bmatrix} 2 & -2 & 1\\ 3 & 1 & 1\\ 1 & -1 & 2\\ -1 & -2 & 3 \end{bmatrix}$$

is called the matrix of the linear transformation T with respect to the ordered bases  $\mathcal{B}$  and  $\mathcal{C}$ .

In general, given a linear transformation  $T: V \to W$ , once you fix a basis  $\mathcal{B} = \{v_1, v_2, ..., v_m\}$  for V and  $\mathcal{C} = \{w_1, w_2, ..., w_n\}$  for V and W respectively, the matrix of the linear transformation T with respect to the ordered bases  $\mathcal{B}$  and  $\mathcal{C}$  is a matrix where the *j*th column is given by the coordinates of  $T(v_j) \in W$  with respect to the basis  $\mathcal{C}$  for W.

# Problem 5

Find the matrix of the linear transformation  $\frac{d}{dx}: P_3 \to P_2$  with respect to the bases  $\mathcal{B}_1 = \{1, x, x^2, x^3\}$  and  $\mathcal{C}_1 = \{1, x, x^2\}$ . Then find the matrix of the same linear transformation but now with respect to the bases  $\mathcal{B}_1 = \{1 + x, 3 - 2x, x^2 + 1, x^3 - x^2 - x - 1\}$  and  $\mathcal{C}_1 = \{1, x, x^2\}$ .