Linear Substitutions and Matrix Multiplication

This note interprets matrix multiplication and related concepts in terms of the composition of linear substitutions.

Composition and multiplication We start from the *linear substitution* (cf. Example 3 on page 5 of Anton–Rorres)

$$\begin{cases} u = x + y + 2z \\ v = 2x + 4y - 3z \\ w = 3x + 6y - 5z \end{cases} \text{ with matrix } A = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 4 & -3 \\ 3 & 6 & -5 \end{bmatrix}$$
(1)

(Linear substitutions are usually not allowed to have constant terms.) Then we substitute in turn linear expressions for each of x, y, and z:

$$\begin{cases} x = s + 2t \\ y = 3s -t \\ z = -s + t \end{cases} \text{ with matrix } B = \begin{bmatrix} 1 & 2 \\ 3 & -1 \\ -1 & 1 \end{bmatrix}$$
(2)

We see that u, v, and w then become linear expressions in s and t,

$$\begin{cases} u = x + y + 2z = s + 2t + 3s - t - 2s + 2t = 2s + 3t \\ v = 2x + 4y - 3z = 2s + 4t + 12s - 4t + 3s - 3t = 17s - 3t \\ w = 3x + 6y - 5z = 3s + 6t + 18s - 6t + 5s - 5t = 26s - 5t \end{cases}$$

This is the *composite* linear substitution of the linear substitutions represented by equations (1) and (2). We observe that its matrix D is just the *product* matrix AB,

$$D = AB = \begin{bmatrix} 1 & 1 & 2 \\ 2 & 4 & -3 \\ 3 & 6 & -5 \end{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & -1 \\ -1 & 1 \end{bmatrix} = \begin{bmatrix} 2 & 3 \\ 17 & -3 \\ 26 & -5 \end{bmatrix}$$

In other words, in terms of the vectors

$$\mathbf{s} = \begin{bmatrix} s \\ t \end{bmatrix}, \quad \mathbf{x} = \begin{bmatrix} x \\ y \\ z \end{bmatrix}, \quad \mathbf{u} = \begin{bmatrix} u \\ v \\ w \end{bmatrix}$$

we have

$$\mathbf{u} = A\mathbf{x} = A(B\mathbf{s}) = (AB)\mathbf{s} = D\mathbf{s}$$
(3)

Associativity Suppose we introduce another linear substitution $\mathbf{s} = C\mathbf{p}$, where we find that there will be no need to make \mathbf{p} explicit. Then we can express \mathbf{x} in terms of \mathbf{p} by the linear substitution

$$\mathbf{x} = B\mathbf{s} = B(C\mathbf{p}) = E\mathbf{p},\tag{4}$$

where E = BC denotes another product matrix.

We now have *two* ways to express \mathbf{u} in terms of \mathbf{p} . We can use equation (3) to go by way of \mathbf{s} ,

$$\mathbf{u} = D\mathbf{s} = D(C\mathbf{p}) = (DC)\mathbf{p}$$

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Or we can use equations (1) and (4) to go by way of \mathbf{x} ,

$$\mathbf{u} = A\mathbf{x} = A(E\mathbf{p}) = (AE)\mathbf{p}$$

We therefore conclude, without doing any real work, that AE = DC, that is, A(BC) = (AB)C, which is the *associative law*. Alternatively, we could express this diagrammatically as

$$\mathbf{p} \xrightarrow{C} \mathbf{s} \xrightarrow{B} \mathbf{x} \xrightarrow{A} \mathbf{u}$$

which indicates that (AB)C and A(BC) are really the same. We may simply write $\mathbf{u} = ABC\mathbf{p}$, with no parentheses.

The identity The *identity* linear substitution

$$\begin{cases} x = x \\ y = y \\ z = z \end{cases} \text{ whose matrix is } I = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 1 & 0 \\ 0 & 0 & 1 \end{bmatrix}$$

clearly preserves everything. Its matrix I (in this case I_3) is therefore called the *identity* matrix. Generally, there is an $n \times n$ identity matrix I_n for each n. (But it has to be square.) From this point of view, it is obvious that $I_m F = F = FI_n$ for any $m \times n$ matrix F.

The zero In contrast, the *zero* linear substitution

$$\begin{cases} y_1 &= 0\\ y_2 &= 0\\ y_3 &= 0\\ y_4 &= 0 \end{cases}$$

destroys everything in sight by setting all variables zero. Its matrix is the *zero matrix*, with all entries 0. The zero matrix comes in all sizes $m \times n$. Applying other linear substitutions before or after the zero linear substitution will still leave all variables set to 0. So in matrix language, G0 = 0 and 0G = 0 for any matrix G, whenever the sizes of these zero matrices allow the products to be defined.