More on Matrices

D. H. Luecking

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We have the following properties for any invertible matrix A:

- 1. For any integers k and m (positive negative or zero) $A^k A^m = A^{k+m}$.
- 2. For any integers k and m (positive negative or zero) $(A^k)^m = A^{km}$. In particular, if m = -1 we get $(A^k)^{-1} = A^{-k}$.

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 $0.7x_0 + 0.2y_0 = x_1$ $0.3x_0 + 0.8y_0 = y_1$

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Written as a matrix equation, this is

$$\left(\begin{array}{cc} 0.7 & 0.2 \\ 0.3 & 0.8 \end{array}\right) \left(\begin{array}{c} x_0 \\ y_0 \end{array}\right) = \left(\begin{array}{c} x_1 \\ y_1 \end{array}\right)$$

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Let
$$A = \begin{pmatrix} 0.7 & 0.2 \\ 0.3 & 0.8 \end{pmatrix}$$
 and suppose $\begin{pmatrix} x_0 \\ y_0 \end{pmatrix} = \begin{pmatrix} 8000 \\ 2000 \end{pmatrix}$

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$$A^{2} \left(\begin{array}{c} 8000\\ 2000 \end{array}\right) = \left(\begin{array}{c} 5000\\ 5000 \end{array}\right) \text{ and } A^{3} \left(\begin{array}{c} 8000\\ 2000 \end{array}\right) = \left(\begin{array}{c} 4500\\ 5500 \end{array}\right)$$

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We will see much later that

$$A^n \left(\begin{array}{c} 8000\\ 2000 \end{array} \right)$$
 approaches $\left(\begin{array}{c} 4000\\ 6000 \end{array} \right)$

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$$\begin{pmatrix} 1 & -2 & 1 & 3\\ 2 & 1 & 0 & 4\\ \hline 4 & 6 & 1 & -3 \end{pmatrix} = \begin{pmatrix} C_{11} & C_{12}\\ C_{21} & C_{22} \end{pmatrix}$$

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Where C_{ij} are matrices. Note that matrices in a given row must be the same height and matrices in a given column must be the same width.

We've already seen something like this where we think of a matrix as made up of its columns:

$$B = \left(\begin{array}{c|c} -1 & 2 & 4\\ 3 & 1 & 0 \end{array}\right) = \left(\begin{array}{c|c} \mathbf{b}_1 & \mathbf{b}_2 & \mathbf{b}_3 \end{array}\right)$$

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As part of the definition of matrix multiplication we saw that

$$A\left(\begin{array}{ccc} \mathbf{b}_1 & \mathbf{b}_2 & \mathbf{b}_3\end{array}\right) = \left(\begin{array}{ccc} A\mathbf{b}_1 & A\mathbf{b}_2 & A\mathbf{b}_3\end{array}\right)$$

This is true even if the parts contain more that one column: if B_j are matrices with all the same height then

$$A\left(\begin{array}{cccc}B_1 & B_2 & \cdots & B_m\end{array}\right) = \left(\begin{array}{cccc}AB_1 & AB_2 & \cdots & AB_m\end{array}\right)$$

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We have already mentioned this when we talked about finding inverses:

$$M\left(\begin{array}{c|c}A & I\end{array}\right) = \left(\begin{array}{c|c}MA & MI\end{array}\right)$$

There is a similar process that works with partitions into rows:

$$\left(\begin{array}{c} \vec{\mathbf{a}}_1\\ \vec{\mathbf{a}}_2\\ \vdots\\ \vec{\mathbf{a}}_n \end{array}\right) B = \left(\begin{array}{c} \vec{\mathbf{a}}_1 B\\ \vec{\mathbf{a}}_2 B\\ \vdots\\ \vec{\mathbf{a}}_n B \end{array}\right)$$

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And we also don't need the parts to be rows: if ${\cal A}_j$ are matrices with same width then

$$\left(\begin{array}{c}A_1\\A_2\\\vdots\\A_n\end{array}\right)B = \left(\begin{array}{c}A_1B\\A_2B\\\vdots\\A_nB\end{array}\right)$$

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$$AB = \begin{pmatrix} \mathbf{a}_1 & \mathbf{a}_2 & \cdots & \mathbf{a}_k \end{pmatrix} \begin{pmatrix} \vec{\mathbf{b}}_1 \\ \vec{\mathbf{b}}_2 \\ \vdots \\ \vec{\mathbf{b}}_k \end{pmatrix} = \sum_{j=1}^k \mathbf{a}_j \vec{\mathbf{b}}_j$$

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This is called the *outer product expansion* of AB. All the terms in the sum are outer products.

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Again, we don't need rows and columns. If A_j and B_j are matrices where all A_j have the same height, all B_j have the same width and the width of each A_j is the same as the height of the corresponding B_j , then

$$\begin{pmatrix} A_1 & A_2 & \cdots & A_k \end{pmatrix} \begin{pmatrix} B_1 \\ B_2 \\ \vdots \\ B_k \end{pmatrix} = \sum_{j=1}^k A_j B_j$$

Finally, to give an inkling how far we can take this

 $\begin{pmatrix} A_{11} & A_{12} \\ A_{21} & A_{22} \end{pmatrix} \begin{pmatrix} B_{11} & B_{12} \\ B_{21} & B_{22} \end{pmatrix} = \begin{pmatrix} A_{11}B_{11} + A_{12}B_{21} & A_{11}B_{12} + A_{12}B_{22} \\ A_{21}B_{11} + A_{22}B_{21} & A_{21}B_{12} + A_{22}B_{22} \end{pmatrix}$ provided the sizes are such that all the products can be performed and all the rows and columns line up.