HOMEWORK: 2.4: 4,14,26*,28,40,48

MATRIX PRODUCT. If $B$ is a $m \times n$ matrix and $A$ is a $n \times p$ matrix, then $BA$ is a $m \times p$ matrix with entries 
\[(BA)_{ij} = \sum_{k=1}^{n} B_{ik}A_{kj}.\]

EXAMPLE. If $B$ is a $3 \times 4$ matrix, and $A$ is a $4 \times 2$ matrix then $BA$ is a $3 \times 2$ matrix.

\[
B = \begin{bmatrix} 1 & 3 & 5 & 7 \\ 3 & 1 & 8 & 1 \\ 1 & 0 & 9 & 2 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 3 \\ 3 & 1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix}, \quad BA = \begin{bmatrix} 1 & 3 & 5 & 7 \\ 3 & 1 & 8 & 1 \\ 1 & 0 & 9 & 2 \end{bmatrix} \begin{bmatrix} 1 & 3 \\ 3 & 1 \\ 1 & 0 \\ 0 & 1 \end{bmatrix} = \begin{bmatrix} 15 & 13 \\ 14 & 11 \\ 10 & 5 \end{bmatrix}.
\]

COMPOSING LINEAR TRANSFORMATIONS. If $S : \mathbb{R}^m \to \mathbb{R}^n, x \mapsto Ax$ and $T : \mathbb{R}^n \to \mathbb{R}^p, x \mapsto Bx$ are linear transformations, then their composition $T \circ S$ is a linear transformation from $\mathbb{R}^m$ to $\mathbb{R}^p$. The corresponding matrix is the matrix product $BA$.

EXAMPLE. Find the matrix which is a composition of a rotation around the $x$-axes by $\pi/2$ followed by a rotation around the $y$-axes by $-\pi/2$.

SOLUTION. The first transformation has the property that $e_1 \to e_1, e_2 \to -e_3, e_3 \to e_2$. The second $e_1 \to -e_2, e_2 \to e_1, e_3 \to e_3$. If $A$ is the matrix belonging to the first transformation and $B$ the second, then $BA$ is the matrix to the composition.

\[
B = \begin{bmatrix} 0 & 1 & 0 \\ -1 & 0 & 0 \\ 0 & 0 & 1 \end{bmatrix}, \quad A = \begin{bmatrix} 1 & 0 & 0 \\ 0 & 0 & 1 \\ 0 & -1 & 0 \end{bmatrix}, \quad BA = \begin{bmatrix} 0 & 0 & 1 \\ -1 & 0 & 0 \\ 0 & -1 & 0 \end{bmatrix}.
\]

The composition maps $e_1 \to -e_2 \to e_3 \to e_1$ is a rotation around a long diagonal.

EXAMPLE. A rotation dilation is the composition of a rotation by $\alpha = \arctan(b/a)$ and a scale by $r = \sqrt{a^2 + b^2}$.

WHY? Matrix multiplication is a generalisation of usual multiplication of numbers or the dot product.

PROPERTIES. Note that $AB \neq BA$ in general! Otherwise, the same rules apply as for numbers: $A(BC) = (AB)C, AA^{-1} = A^{-1}A = 1_n, (AB)^{-1} = B^{-1}A^{-1}, A(B + C) = AB + AC, (B + C)A = BA + CA$ etc.

PARTITIONED MATRICES. The entries of matrices can themselves be matrices. If $B$ is a $m \times n$ matrix and $A$ is a $n \times p$ matrix, and assume the entries are $k \times k$ matrices, then $BA$ is a $m \times p$ matrix where each entry $(BA)_{ij} = \sum_{k=1}^{n} B_{ik}A_{kj}$ is a $k \times k$ matrix. Partitioning matrices can improve matrix multiplication (i.e. Strassen algorithm).

EXAMPLE. If $A = \begin{bmatrix} A_{11} & A_{12} \\ A_{00} & A_{22} \end{bmatrix}$, where $A_{ij}$ are $k \times k$ matrices with the property that $A_{11}$ and $A_{22}$ are invertible, then $B = \begin{bmatrix} A_{11}^{-1} & -A_{11}^{-1}A_{12}A_{22}^{-1} \\ 0 & A_{22}^{-1} \end{bmatrix}$ is the inverse of $A$.

NETWORKS. Let us associate to the computer network (shown at the left) a matrix
\[
\begin{bmatrix} 0 & 1 & 1 & 1 \\ 1 & 0 & 1 & 0 \\ 1 & 1 & 0 & 1 \\ 1 & 0 & 1 & 0 \end{bmatrix}
\]
To a worm in the first computer we associate a vector
\[
\begin{bmatrix} 1 \\ 0 \\ 0 \\ 0 \end{bmatrix}
\]
The vector $Ax$ has a 1 at the places, where the worm could be in the next step. The vector $(AA)(x)$ tells, in how many ways the worm can go from the first computer to other hosts in 2 steps. In our case, it can go in three different ways back to the computer itself.

Matrices help to solve combinatorial problems (see movie "Good will hunting"). For example, what does $[A^{1000}]_{22}$ tell about the worm infection of the network? What does it mean if $A^{100}$ has no zero entries?
FRACTALS. Closely related to linear maps are affine maps $x \mapsto Ax + b$. They are compositions of a linear map with a translation. It is not a linear map if $B(0) \neq 0$. Affine maps can be disguised as linear maps in the following way: let $y = \begin{bmatrix} x \\ 1 \end{bmatrix}$ and the $(n+1) \times (n+1)$ and $B = \begin{bmatrix} A & b \\ 0 & 1 \end{bmatrix}$, then $By = \begin{bmatrix} Ax + b \end{bmatrix}$.

Fractals can be constructed by taking for example 3 affine maps $f, g, h$ which contract area. For a given object $Y_0$ define $Y_1 = f(Y_0) \cup g(Y_0) \cup h(Y_0)$ and recursively $Y_k = f(Y_{k-1}) \cup g(Y_{k-1}) \cup h(Y_{k-1})$. Above you see $Y_k$ after some iterations. In the limit, $Y_k$ becomes a fractal, an object with noninteger dimension.

CHAOS. Consider a map in the plane like $T : \begin{bmatrix} x \\ y \end{bmatrix} \mapsto \begin{bmatrix} 2x + 2 \sin(x) - y \\ x \end{bmatrix}$. We apply this map again and again, we look at points $(x_1, y_1) = T(x, y)$, $(x_2, y_2) = T(T(x, y))$, etc. One writes $T^n$ for the n-th iteration of the map and $(x_n, y_n)$ for the image of $(x, y)$ under the map $T^n$. The linear approximation of the map at a point $(x, y)$ is the matrix $DT(x, y) = \begin{bmatrix} 2 + 2 \cos(x) - 1 \\ 1 \end{bmatrix}$. (If $T : \begin{bmatrix} x \\ y \end{bmatrix} = \begin{bmatrix} f(x, y) \\ g(x, y) \end{bmatrix}$, then the row vectors of $DT(x, y)$ are just the gradients of $f$ and $g$). $T$ is called chaotic at $(x, y)$, if the entries of $D(T^n)(x, y)$ grow exponentially fast with $n$. By the chain rule, $D(T^n)$ is the product of matrices $DT(x_i, y_i)$. For example, $T$ is chaotic at $(0, 0)$. If there is a positive probability to hit a chaotic point, then $T$ is called chaotic.

FALSE COLORS. Any color can be represented as a vector $(r, g, b)$, where $r \in [0, 1]$ is the red $g \in [0, 1]$ is the green and $b \in [0, 1]$ is the blue component. Changing colors in a picture means applying a transformation on the cube. Let $T : (r, g, b) \mapsto (g, b, r)$ and $S : (r, g, b) \mapsto (r, g, 0)$. What is the composition of these two linear maps?

OPTICS. Matrices help to calculate the motion of light rays through lenses. A light ray $y(s) = x + ms$ in the plane is described by a vector $(x, m)$. Following the light ray over a distance of length $L$ corresponds to the map $(x, m) \mapsto (x + mL, m)$. In the lense, the ray is bent depending on the height $x$. The transformation in the lense is $(x, m) \mapsto (x, m - kx)$.

$\begin{bmatrix} x \\ m \end{bmatrix} \mapsto A_L \begin{bmatrix} x \\ m \end{bmatrix} = \begin{bmatrix} 1 & L \\ 0 & 1 \end{bmatrix} \begin{bmatrix} x \\ m \end{bmatrix}$, $\begin{bmatrix} x \\ m \end{bmatrix} \mapsto B_k \begin{bmatrix} x \\ m \end{bmatrix} = \begin{bmatrix} 1 & 0 \\ -k & 1 \end{bmatrix} \begin{bmatrix} x \\ m \end{bmatrix}$.

Examples:
1) Eye of length $R$ looking far: $A_R B_k$. 2) Eye of length $R$ looking at distance $L$: $A_R B_k A_L$. 3) Telescope: $B_k A_L B_L$. 