

#### Document 4: Data Encoding

Briefly, let's review matrix multiplication. Define the following matrices  $A$  and  $B$ .

$$A = \begin{bmatrix} 1 & 2 & 3 \\ 4 & 5 & 6 \end{bmatrix}, B = \begin{bmatrix} -1 & -4 \\ -2 & 5 \\ -3 & 6 \end{bmatrix}$$

Multiplying components, we can expand the equation.

$$AB = \begin{bmatrix} a & b \\ c & d \end{bmatrix}, a = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} -1 \\ -2 \\ -3 \end{bmatrix}, b = \begin{bmatrix} 1 & 2 & 3 \end{bmatrix} \begin{bmatrix} -4 \\ 5 \\ 6 \end{bmatrix}, c = \begin{bmatrix} 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} -1 \\ -2 \\ -3 \end{bmatrix},$$
$$d = \begin{bmatrix} 4 & 5 & 6 \end{bmatrix} \begin{bmatrix} -4 \\ 5 \\ 6 \end{bmatrix}$$

Going left to right in the first matrix, and top to bottom in the right matrix, we multiply elements of the matrix, and sum the products as such.

$$a = 1(-1) + 2(-2) + 3(-3), b = 1(-4) + 2(5) + 3(6), c = 4(-1) + 5(-2) + 6(-3), d = 4(-4) + 5(5) + 6(6)$$

Hence:

$$a = -14, b = 24, c = -32, d = 45 \Rightarrow AB = \begin{bmatrix} -14 & 24 \\ -32 & 45 \end{bmatrix}$$

In 1D, given an equation  $ax = b$ , we have  $x = \frac{b}{a}$  or  $x = a^{-1}b$ . In 2D, with matrices, given an equation  $A_{2 \times 2}x_{2 \times 1} = b_{2 \times 1}$ , we cannot have  $x = \frac{b}{A}$  because there is no division of matrices. However, we can take the inverse of a matrix and have  $x = a^{-1}b$ . To get an inverse of a matrix, we multiply the matrix by the reciprocal of its determinant.

Given the following matrix, we have determinant  $D = 1 * 4 - 2 * 3 = -2$ .

$$A = \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix}$$

Matrices can be used as complicated functions with multiple variables. We can input systems of equations, multiply by the transformation matrix, and get a new matrix as output. The input is transformed from the decoded space to the encoded space. If you have a system of equations in  $x$ , then

define it in  $y$ , we can call the matrix corresponding with  $x$   $A$ , and the matrix corresponding with  $y$   $A^{-1}$ .

Given the system  $\begin{cases} x_1 + 3x_2 = y_1 \\ 2x_1 + 5x_2 = y_2 \end{cases}$ , we have the corresponding matrix. To solve the system, we place it into reduced row echelon form.

$$A = \left[ \begin{array}{cc|c} 1 & 3 & y_1 \\ 2 & 5 & y_2 \end{array} \right] = \left[ \begin{array}{cc|c} 1 & 0 & 3y_2 - 5y_1 \\ 0 & 1 & -y_2 + 2y_1 \end{array} \right]$$

The reduced row echelon form of the above equation yields the following system, which can be represented using matrices involving the input set containing  $x_1, x_2$  and the solution set involving  $y_1, y_2$ .

$$\text{system} = \begin{cases} x_1 = 3y_2 - 5y_1 \\ x_2 = -y_2 + 2y_1 \end{cases}, \Rightarrow \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} = \begin{bmatrix} -5 & 3 \\ 2 & -1 \end{bmatrix} \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

The second matrix performs a transformation from  $y$  to  $x$ , where the matrix  $A$  performed an transformation from  $x$  to  $y$ . These transformations are exactly opposite, and hence we call this matrix  $A^{-1}$ .

We can apply linear transformations in matrices. Matrices are not commutative, so the order of multiplication is important, just as it is for composition of functions. Given matrices to multiply  $A, B$ , and  $X$ , we define  $A * B * X$  as equal to  $A * (B * X)$ . Imagining the matrices are functions,  $X$  goes into  $B$  goes into  $A$  to produce the output  $A * B * X$ .

Definition: A function  $T : \mathbb{R}^n \rightarrow \mathbb{R}^m$  is a linear transformation if there is an  $m \times n$  matrix  $A$  such that

$$T(\vec{x}) = A\vec{x}, \forall \vec{x} \in \mathbb{R}^n$$

We say  $T$  is acting on vector  $\vec{x}$ .

There are many other transformations that can be represented by matrices. Given a plane defined by some values  $x_1, x_2$ , we have a matrix representation that we can scale and shift to yield a new transformed plane in our solution space. For example, lets define a plane, linearly scale it down by  $\frac{1}{2}$ , and shift it left by  $\frac{1}{2}$ .

$$\begin{bmatrix} \frac{1}{2} & 0 \\ 0 & \frac{1}{2} \end{bmatrix} \begin{bmatrix} x_1 \\ x_2 \end{bmatrix} + \begin{bmatrix} \frac{1}{2} \\ \frac{1}{2} \end{bmatrix} = \begin{bmatrix} y_1 \\ y_2 \end{bmatrix}$$

Beyond scaling and rotating, we can also perform rotations. Let  $T$  be a transformation matrix for a rotation by  $90^\circ$  of the  $xy$ -plane.  $T \equiv R_{90^\circ}$

$$T = \begin{bmatrix} 0 & -1 \\ 1 & 0 \end{bmatrix} = \begin{bmatrix} \cos(\theta) & -\sin(\theta) \\ \sin(\theta) & \cos(\theta) \end{bmatrix}$$