



# Linear Algebra Refresher

Basics of Tensor



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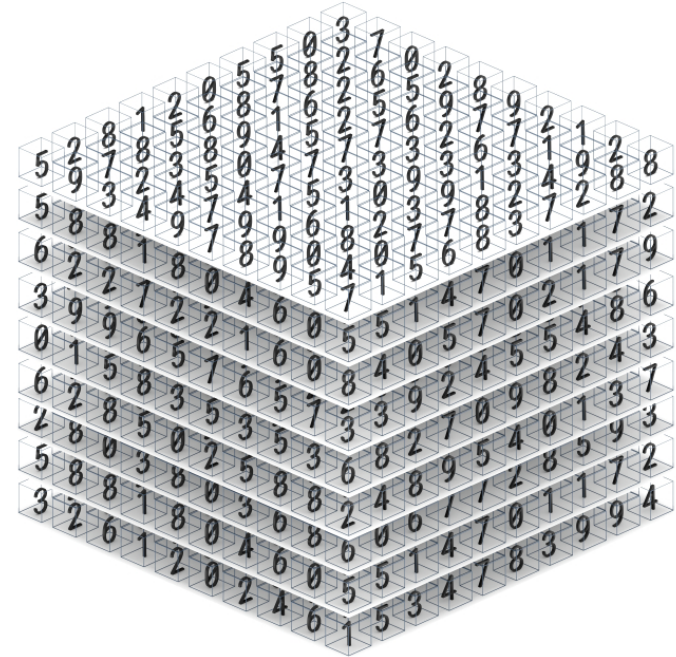
# What is a Tensor?



*In many cases, we need an array with more than two axes.*

*In the general case, an array of numbers arranged on a regular grid with a variable number of axes is known as a tensor.*

Structurally a tensor can be thought of as a generalised matrix. Dynamically, a tensor elements transform uniformly according to certain rules under a change of coordinates.



A tensor in 3 dimensions is a 3D matrix or cube of numbers

# Scalar, Vector, Matrix and Tensor



So how are scalars, vectors, matrices and tensors related?

A vector is a first order tensor, a matrix is a second order **tensor** or two dimensional tensor. A tensor may just have a single number, in which case it is as a **tensor** of order zero, or simply a scalar.

Tensor notation is like matrix notation with capital letter represent a tensor and elements are represented with small letters with subscript.

A three-dimensional tensor can be represented as  $\mathbf{A} = ((a_{i,j,k}))$

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Scalar

$$\begin{bmatrix} 1 & 2 \end{bmatrix}$$

Vector

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

Matrix

$$\begin{bmatrix} \begin{bmatrix} 1 & 2 \\ 3 & 4 \end{bmatrix} & \begin{bmatrix} 3 & 2 \\ 4 & 4 \end{bmatrix} \\ \begin{bmatrix} 5 & 2 \\ 8 & 6 \end{bmatrix} & \begin{bmatrix} 2 & 2 \\ 1 & 1 \end{bmatrix} \end{bmatrix}$$

Tensor

# Tensor – Applications in Deep Learning



Tensors are used as the fundamental data structure in Deep Learning. In computer vision, tensor data structures are used to represent multi-channel colour images and extract colour features. Tensor data structures are also used to represent node inputs, outputs and weights in Artificial Neural Networks.

For example, following is a tensor representation of a 3 channels (RGB) image. Order of the tensor is  $3 \times 5 \times 4$ .

127	194	187	23				
141	170	135	152	84	70		
234	180	75	115	2	252	215	33
228	150	166	160	175	135	69	91
199	34	83	25	197	243	62	56
		208	199	182	133	217	149
				233	84	178	173

# Basic Tensor Operations – Addition and Subtraction



Tensor Addition of two tensors **A** and **B** is simply the element wise addition of two tensors resulting in a matrix **C** of same dimension.

Let  $A = (a_1, a_2)$  and  $B = (b_1, b_2)$

$$\begin{aligned}\text{Then } \mathbf{C} &= \mathbf{A} + \mathbf{B} \\ &= (a_1 + b_1, a_2 + b_2)\end{aligned}$$

Similarly Tensor Subtraction of two tensors **A** and **B** is simply the element wise subtraction of two tensors resulting in a matrix **C** of same dimension. Therefore:

$$\begin{aligned}\mathbf{C} &= \mathbf{A} - \mathbf{B} \\ &= (a_1 - b_1, a_2 - b_2)\end{aligned}$$

# Basic Tensor Operations - Product



Consider a tensor **A** with m dimensions and tensor **B** with n dimensions, the product of these tensors will be a new tensor **C** with q + r dimensions.

Let  $A = (a_1, a_2)$  and  $B = (b_1, b_2)$

Then  $C = A \otimes B$

$$C = (a_1 \otimes [b_1, b_2] \\ a_2 \otimes [b_1, b_2])$$

$$\text{Or } C = (a_1 \otimes b_1 \ a_1 \otimes b_2 \\ a_2 \otimes b_1 \ a_2 \otimes b_2)$$